# **Supplementary Material**

# Global meta-analysis of associations between ambient temperature and pathogenspecific respiratory infections

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# List of abbreviations

OHAT	Office of Health Assessment and Translation
IV	Influenza virus
HPIV	Human parainfluenza virus
RSV	Respiratory syncytial virus
ARI	Acute respiratory infection
GBD	Global Burden of Diseases, Injuries, and Risk Factors Study
DALYs	disability-adjusted life years
RTIs	Respiratory tract infections
HMPV	Human metapneumovirus
HRV	Human rhinovirus
HAdV	Human adenovirus
HBoV	Human bocavirus
MERS-CoV	Middle East Respiratory Syndrome Coronavirus
SARS-CoV	Severe acute respiratory syndrome coronavirus
SARS-CoV-2	Severe acute respiratory syndrome coronavirus 2
HCoV	Human coronavirus
EV	Enterovirus
M. pneumoniae	Mycoplasma pneumoniae
C. pneumoniae	Chlamydophila pneumoniae
S. pneumoniae	Streptococcus pneumoniae
S. pyogenes	Streptococcus pyogenes pharyngiti
H. influenzae	Haemophilus influenzae
L. pneumoniae	Legionella pneumoniae
M. catarrhalis	Moraxella catarrhalis
P. aeruginosa	Pseudomonas aeruginosa
CBM	China Biology Medicine
WHO	World Health Organization

## Table S1. PRISMA Checklist

Section and Topic	Item #	Checklist item	Location where item is reported
TITLE			
Title	1	Identify the report as a systematic review.	1
ABSTRACT			
Abstract	2	See the PRISMA 2020 for Abstracts checklist.	1-2
INTRODUCTION			
Rationale	3	Describe the rationale for the review in the context of existing knowledge.	3-4
Objectives	4	Provide an explicit statement of the objective(s) or question(s) the review addresses.	4
METHODS			
Eligibility criteria	5	Specify the inclusion and exclusion criteria for the review and how studies were grouped for the syntheses.	4-5
Information sources	6	Specify all databases, registers, websites, organisations, reference lists and other sources searched or consulted to identify studies. Specify the date when each source was last searched or consulted.	5
Search strategy	7	Present the full search strategies for all databases, registers and websites, including any filters and limits used.	5
Selection process	8	Specify the methods used to decide whether a study met the inclusion criteria of the review, including how many reviewers screened each record and each report retrieved, whether they worked independently, and if applicable, details of automation tools used in the process.	5
Data collection process	9	Specify the methods used to collect data from reports, including how many reviewers collected data from each report, whether they worked independently, any processes for obtaining or confirming data from study investigators, and if applicable, details of automation tools used in the process.	6
Data items	10a	List and define all outcomes for which data were sought. Specify whether all results that were compatible with each outcome domain in each study were sought (e.g. for all measures, time points, analyses), and if not, the methods used to decide which results to collect.	6
	10b	List and define all other variables for which data were sought (e.g. participant and intervention characteristics, funding sources). Describe any assumptions made about any missing or unclear information.	6
Study risk of bias assessment	11	Specify the methods used to assess risk of bias in the included studies, including details of the tool(s) used, how many reviewers assessed each study and whether they worked independently, and if applicable, details of automation tools used in the process.	7
Effect measures	12	Specify for each outcome the effect measure(s) (e.g. risk ratio, mean difference) used in the synthesis or presentation of results.	8
Synthesis methods	13a	Describe the processes used to decide which studies were eligible for each synthesis (e.g. tabulating the study intervention characteristics and comparing against the planned groups for each synthesis (item #5)).	8
	13b	Describe any methods required to prepare the data for presentation or synthesis, such as handling of missing summary statistics, or data conversions.	8
	13c	Describe any methods used to tabulate or visually display results of individual studies and syntheses.	8

Section and Topic	Item #	Checklist item	Location where item is reported
	13d	Describe any methods used to synthesize results and provide a rationale for the choice(s). If meta-analysis was performed, describe the model(s), method(s) to identify the presence and extent of statistical heterogeneity, and software package(s) used.	8-9
	13e	Describe any methods used to explore possible causes of heterogeneity among study results (e.g. subgroup analysis, meta-regression).	8-9
	13f	Describe any sensitivity analyses conducted to assess robustness of the synthesized results.	8-9
Reporting bias assessment	14	Describe any methods used to assess risk of bias due to missing results in a synthesis (arising from reporting biases).	7-8
Certainty assessment	15	Describe any methods used to assess certainty (or confidence) in the body of evidence for an outcome.	7-8
RESULTS			
Study selection	16a	Describe the results of the search and selection process, from the number of records identified in the search to the number of studies included in the review, ideally using a flow diagram.	9
	16b	Cite studies that might appear to meet the inclusion criteria, but which were excluded, and explain why they were excluded.	9
Study characteristics	17	Cite each included study and present its characteristics.	9
Risk of bias in studies	18	Present assessments of risk of bias for each included study.	10
Results of individual studies	19	For all outcomes, present, for each study: (a) summary statistics for each group (where appropriate) and (b) an effect estimate and its precision (e.g. confidence/credible interval), ideally using structured tables or plots.	9-12
Results of	20a	For each synthesis, briefly summarise the characteristics and risk of bias among contributing studies.	9-12
syntheses	20b	Present results of all statistical syntheses conducted. If meta-analysis was done, present for each the summary estimate and its precision (e.g. confidence/credible interval) and measures of statistical heterogeneity. If comparing groups, describe the direction of the effect.	9-12
	20c	Present results of all investigations of possible causes of heterogeneity among study results.	9-12
	20d	Present results of all sensitivity analyses conducted to assess the robustness of the synthesized results.	9-12
Reporting biases	21	Present assessments of risk of bias due to missing results (arising from reporting biases) for each synthesis assessed.	12
Certainty of evidence	22	Present assessments of certainty (or confidence) in the body of evidence for each outcome assessed.	12-13
DISCUSSION			
Discussion	23a	Provide a general interpretation of the results in the context of other evidence.	13-14
	23b	Discuss any limitations of the evidence included in the review.	14-16

Section and Topic	Item #	Checklist item	Location where item is reported
	23c	Discuss any limitations of the review processes used.	15-16
	23d	Discuss implications of the results for practice, policy, and future research.	16
OTHER INFORMA	TION		
Registration and	24a	Provide registration information for the review, including register name and registration number, or state that the review was not registered.	1
protocol	24b	Indicate where the review protocol can be accessed, or state that a protocol was not prepared.	5
	24c	Describe and explain any amendments to information provided at registration or in the protocol.	5
Support	25	Describe sources of financial or non-financial support for the review, and the role of the funders or sponsors in the review.	Title page
Competing interests	26	Declare any competing interests of review authors.	Title page
Availability of data, code and other materials	27	Report which of the following are publicly available and where they can be found: template data collection forms; data extracted from included studies; data used for all analyses; analytic code; any other materials used in the review.	/

From: Page MJ, McKenzie JE, Bossuyt PM, Boutron I, Hoffmann TC, Mulrow CD, et al. The PRISMA 2020 statement: an updated guideline for reporting systematic reviews.

BMJ 2021;372:n71. doi: 10.1136/bmj.n71

For more information, visit: <a href="http://www.prisma-statement.org/">http://www.prisma-statement.org/</a>

Table S2. Search strategy: terms, databases, limitations and number of articles for review

### PubMed database

#1

- ("temperature" [MeSH Terms] OR "temperature" [All Fields] OR "temperatures" [All Fields] OR "temperatures" [All Fields] OR "temperatures" [All Fields] OR "climate" [MeSH Terms] OR "climate" [All Fields] OR "climates" [All Fields] OR "weatherability" [All Fields] OR "weatherable" [All Fields] OR "weatherables" [All Fields] OR "weatherables" [All Fields] OR "weatherables" [All Fields] OR "meteorologicals" [All Fie
- Fields] OR "global warming"[All Fields]) ("Bacterial Pneumonia" [All Fields] OR "pneumonia bacterial" [All Fields] OR "Bacterial Pneumonias" [All Fields] #2 OR "Bacterial Infections and Mycoses" [All Fields] OR "Bacterial Infections" [All Fields] OR "pneumonia pneumococcal"[All Fields] OR "pneumonia staphylococcal"[All Fields] OR "Streptococcus pneumoniae"[All Fields] OR "Chlamydial Pneumonia" [All Fields] OR "pneumonia mycoplasma" [All Fields] OR "Mycoplasma pneumoniae"[All Fields] OR "Schizoplasma pneumoniae"[All Fields] OR "Chlamydophila pneumoniae"[All Fields] OR "Lower respiratory tract" [All Fields] OR "upper respiratory tract" [All Fields] OR "Human Rhinovirus" [All Fields] OR "Human Enterovirus" [All Fields] OR "Respiratory Tract Infections" [All Fields] OR "Respiratory System Infections" [All Fields] OR "Respiratory System Infection" [All Fields] OR "Upper Respiratory Tract Infections" [All Fields] OR "Upper Respiratory Infections" [All Fields] OR "Upper Respiratory Tract Infection" [All Fields] OR "respiratory infections" [All Fields] OR "respiratory virus" [All Fields] OR "Respiratory syncytial virus"[All Fields] OR "Respiratory Syncytial Viruses"[All Fields] OR "RSV"[All Fields] OR "Chimpanzee Coryza Agent" [All Fields] OR ("adenoviridae" [MeSH Terms] OR "adenoviridae" [All Fields] OR "adenoviruses" [All Fields] OR "adenoviruse" [All Fields]) OR "Human Adenovirus" [All Fields] OR Enterovirus [All Fields] OR "Human Adenoviruses" [All Fields] OR "APC Viruses" [All Fields] OR "APC Virus" [All Fields] OR ("orthomyxoviridae" [MeSH Terms] OR "orthomyxoviridae" [All Fields] OR "orthomyxoviruses" [All Fields]) OR ("orthomyxoviridae" [MeSH Terms] OR "orthomyxoviridae" [All Fields] OR "myxoviruses" [All Fields]) OR "Influenza viruses" [All Fields] OR "Influenza A virus" [All Fields] OR "Influenza A viruses" [All Fields] OR "Orthomyxovirus Type A"[All Fields] OR "Influenza Viruses Type A"[All Fields] OR "Influenza B virus"[All Fields] OR "influenza B viruses" [All Fields] OR "Influenza Viruses Type B" [All Fields] OR "Influenza B virus" [All Fields] OR "Influenzavirus C"[All Fields] OR "Influenza C Virus"[All Fields] OR "Influenza C Viruses"[All Fields] OR "Human parainfluenza viruses" [All Fields] OR "parainfluenza virus 2 human" [All Fields] OR "Parainfluenza Virus Type 2"[All Fields] OR "croup associated viruses"[All Fields] OR "croup associated viruses"[All Fields] OR "Human parainfluenza virus 2"[All Fields] OR "croup associated virus"[All Fields] OR "croup associated virus"[All Fields] OR "HPIV" [All Fields] OR ("metapneumovirus" [MeSH Terms] OR "metapneumovirus" [All Fields] OR "metapneumoviruses"[All Fields]) OR ("metapneumovirus"[MeSH Terms] OR "metapneumovirus"[All Fields] OR "metapneumoviruses"[All Fields]) OR "Human Metapneumoviruses"[All Fields] OR Metapneumovirus"[All Fields] OR "HMPV"[All Fields] OR ("rhinovirus"[MeSH Terms] OR "rhinovirus"[All Fields] OR "rhinoviruses"[All Fields]) OR "Human Rhinovirus"[All Fields] OR "Common Cold Virus"[All Fields] OR "Common Cold Viruses" [All Fields] OR "Coryza Virus" [All Fields] OR ("coronavirus" [MeSH Terms] OR "coronavirus"[All Fields] OR "coronaviruses"[All Fields]) OR "SARS coronavirus"[All Fields] OR "severe acute respiratory syndrome related coronavirus"[All Fields] OR "severe acute respiratory syndrome related coronavirus"[All Fields] OR "HCoV-SARS"[All Fields] OR "SARS Virus"[All Fields] OR ("severe acute respiratory syndrome related coronavirus" [MeSH Terms] OR ("severe" [All Fields] AND "acute" [All Fields] AND "respiratory" [All Fields] AND "syndrome related" [All Fields] AND "coronavirus" [All Fields]) OR "severe acute respiratory syndrome related coronavirus" [All Fields] OR "sarbecovirus" [All Fields]) OR ("severe acute respiratory syndrome related coronavirus" [MeSH Terms] OR ("severe" [All Fields] AND "acute" [All Fields] AND "respiratory"[All Fields] AND "syndrome related"[All Fields] AND "coronavirus"[All Fields]) OR "severe acute respiratory syndrome related coronavirus"[All Fields] OR "sarbecoviruses"[All Fields]) OR "Severe acute respiratory syndrome coronavirus" [All Fields] OR "Severe Acute Respiratory Syndrome Virus" [All Fields] OR "SARS-Cov-2"[All Fields] OR "HcoV"[All Fields] OR "NL63"[All Fields] OR "OC43"[All Fields] OR "229E"[All Fields] OR "HKU1" [All Fields] OR "Human bocaviruses" [All Fields] OR "Coxiella burnetii" [All Fields] OR "Chlamydophila psittaci"[All Fields] OR "Legionella pneumophila"[All Fields] OR "Staphylococcus aureus"[All Fields] OR "Haemophilus influenzae" [All Fields] OR "Pseudomonas aeruginosa" [All Fields] OR "Klebsiella pneumoniae"[All Fields] OR "Escherichia coli"[All Fields] OR "Proteus mirabilis"[All Fields] OR "Providencia stuartii"[All Fields] OR "Moraxella catarrhalis"[All Fields] OR "Streptococcus hemolyticus"[All Fields] OR "Streptococcus anginosus" [All Fields] OR "Streptococcus pneumoniae" [All Fields] OR "Gram-positive cocci" [All Fields] OR "Gram-negative bacteria" [All Fields] OR "gram-negative bacillus" [All Fields] OR "Pseudomonas aeruginosa"[All Fields] OR ("candida"[MeSH Terms] OR "candida"[All Fields] OR "candidae"[All Fields] OR "candidas"[All Fields]) OR "Corynebacterium diphtheriae"[All Fields] OR "Bordetella pertussis"[All Fields] OR "M.tuberculosis" [All Fields] OR "Acute bronchitis" [All Fields] OR "acute angina" [All Fields] OR "acute tonsillitis"[All Fields] OR ("palatine tonsil"[MeSH Terms] OR ("palatine"[All Fields] AND "tonsil"[All Fields]) OR "palatine tonsil"[All Fields] OR "tonsil"[All Fields] OR "tonsils"[All Fields] OR "tonsils"[All Fields] OR "tonsillitis"[MeSH Terms] OR "tonsillitis"[All Fields] OR "tonsillitides"[All Fields] OR "tonsills"[All Fields]) OR "Sinus Infections" [All Fields] OR "Viral Pneumonias" [All Fields] OR "Acute Coryza" [All Fields] OR "Acute viral pharyngitis"[All Fields] OR ("pharyngitis"[MeSH Terms] OR "pharyngitis"[All Fields] OR "pharyngitides"[All

Fields]) OR "Sore Throats" [All Fields] OR "Acute viral laryngitis" [All Fields] OR "acute pharyngitis" [All Fields] OR "acute epiglottitis" [All Fields] OR ("epiglottal" [All Fields] OR "epiglottis" [MeSH Terms] OR "epiglottis" [All Fields] OR "epiglottic" [All Fields] OR "epiglottitis" [MeSH Terms] OR "epiglottitis" [All Fields] OR "epiglottitides"[All Fields]) OR ("bronchitis"[MeSH Terms] OR "bronchitis"[All Fields] OR "bronchitides"[All Fields]) OR ("larynges"[All Fields] OR "laryngitis"[MeSH Terms] OR "laryngitis"[All Fields] OR "laryngitides"[All Fields] OR "larynx"[MeSH Terms] OR "larynx"[All Fields] OR "laryngeal"[All Fields]) OR ("legionellosis" [MeSH Terms] OR "legionellosis" [All Fields] OR "legionelloses" [All Fields]) OR "Lung Abscess"[All Fields] OR ("blastomycosis" [MeSH Terms] OR "blastomycosis" [All Fields] OR "blastomycoses" [All Fields]) OR "pneumonia pneumocystis" [All Fields] OR ("pharyngitis" [MeSH Terms] OR "pharyngitis" [All Fields] OR "pharyngitides" [All Fields]) OR ("nasopharyngitis" [MeSH Terms] OR "nasopharyngitis" [All Fields]) OR "Retropharyngeal Abscess"[All Fields] OR ("pleurisy"[MeSH Terms] OR "pleurisy"[All Fields] OR "pleurisies" [All Fields]) OR ("pleuropneumonia" [MeSH Terms] OR "pleuropneumonia" [All Fields] OR 'pleuropneumonias" [All Fields] OR "pleuropneumoniae" [All Fields]) OR ("pneumonia" [MeSH Terms] OR "pneumonia" [All Fields] OR "pneumonias" [All Fields] OR "pneumoniae" [All Fields] OR "pneumoniae s" [All "bronchopneumonia"[All ("bronchopneumonia"[MeSH Terms] OR Fields] OR "bronchopneumonias" [All Fields] OR "bronchopneumoniae" [All Fields]) OR ("pleuropneumonia" [MeSH Terms] OR "pleuropneumonia" [All Fields] OR "pleuropneumonias" [All Fields] OR "pleuropneumoniae" [All Fields]) OR ("rhinitis"[MeSH Terms] OR "rhinitis"[All Fields] OR "rhinitides"[All Fields]) OR ("rhinoscleroma"[MeSH Terms] OR "rhinoscleroma" [All Fields] OR "rhinoscleromas" [All Fields]) OR "Severe Acute Respiratory Syndrome" [All Fields] OR ("paranasal sinuses"[MeSH Terms] OR ("paranasal"[All Fields] AND "sinuses"[All Fields]) OR "paranasal sinuses"[All Fields] OR "sinuses"[All Fields] OR "sinusal"[All Fields] OR "sinuse"[All Fields] OR "sinusitis" [MeSH Terms] OR "sinusitis" [All Fields] OR "sinusitides" [All Fields]) OR "Allergic Fungal Sinusitis" [All Fields] OR "Ethmoid Sinusitis" [All Fields] OR "Frontal Sinusitis" [All Fields] OR "Maxillary Sinusitis"[All Fields] OR "Sphenoid Sinusitis"[All Fields] OR ("supraglottitis"[MeSH Terms] OR "supraglottitis"[All Fields]) OR ("epiglottal"[All Fields] OR "epiglottis"[MeSH Terms] OR "epiglottis"[All Fields] OR "epiglottic"[All Fields] OR "epiglottitis"[MeSH Terms] OR "epiglottitis"[All Fields] OR "epiglottitides"[All Fields]) OR ("tracheitis"[MeSH Terms] OR "tracheitis"[All Fields]) OR ("silicotuberculosis"[MeSH Terms] OR "silicotuberculosis"[All Fields]) OR "Whooping Cough"[All Fields])

#3	#1	AND	#2

#3	#1 AND #2
Cochra	ne Central Register of Controlled Trials
#1	MeSH descriptor: [Temperature] explode all trees
#2	MeSH descriptor: [Climate Change] explode all trees
#3	MeSH descriptor: [Weather] explode all trees
#4	#1 or #2 or #3
#10	(temperature or "ambient temperature" or "seasurface temperature" or climate or weather or meteorological or "climate change" or Meteorolog* or "extreme weather" or "meteorological factor" or "hot weather" or "cold weather" or "climatic factor" or "global warming"):ti,ab,kw
#11	#9 or #10
#12	MeSH descriptor: [Respiratory Tract Infections] explode all trees
#13	MeSH descriptor: [Pneumonia, Bacterial] explode all trees
#14	MeSH descriptor: [Respiratory Syncytial Viruses] explode all trees
#15	MeSH descriptor: [Mycoplasma pneumoniae] explode all trees
#16	MeSH descriptor: [Adenoviridae] explode all trees
#17	#12 or #13 or #14 or #15 or #16
#18	("Bacterial Pneumonia" or "Pneumonia, Bacterial" or "Bacterial Pneumonias" or "Bacterial Infections and Mycoses" or "Bacterial Infections" or "Pneumonia, Pneumococcal" or "Pneumonia, Staphylococcal" or "Streptococcus pneumoniae" or "Chlamydial Pneumonia" or "Pneumonia, Mycoplasma" or "Mycoplasma pneumoniae" or "Schizoplasma pneumoniae" or "Chlamydophila pneumoniae" or "Lower respiratory tract" or "upper respiratory tract" or "Human Rhinovirus" or "Human Enterovirus" or "Respiratory Tract Infections" or "Respiratory System Infections" or "Upper Respiratory Infections" or "Upper Respiratory Infections" or "Upper Respiratory Syncytial Viruses" or "RSV" or "Chimpanzee Coryza Agent" or Adenoviruses or "Human Adenovirus" or "Human Adenoviruses" or "APC Viruses" or "APC Virus" or "Orthomyxoviruses or Myxoviruses or "Influenza viruses" or "Influenza A virus" or "Influenza A viruses" or "Orthomyxoviruses or "Influenza Viruses Type A" or "Influenza B virus" or "Influenza B virus" or "Influenza C Viruses" or "Human parainfluenza viruses" or "Parainfluenza Viruse 2, Human" or "Parainfluenza Virus Type 2" or "Croup-Associated Viruses" or "Croup Associated Viruses" or "Human parainfluenza virus 2" or "Croup-Associated Virus" or "Croup Associated Virus" or "Human Metapneumoviruses" or "Human Metapneumoviruses" or "Human Metapneumoviruses" or "Group Associated Viruses" or "Group Virus" or "Common Cold Virus" or "Common Cold Viruses" or "Sarbecovirus or "Sarbecoviruses or "Sarbe coronavirus" or "Severe acute respiratory syndrome related coronavirus" or "HCoV-Sars" or "Sarbe Covirus or "Sarbecoviruses or "Sarbecoviruses or "Sarbecoviruses or "Chlamydophila psittaci" or "Legionella pneumophila" or "Staphylococcus aureus" or "Haemophilus influenzae" or "Pseudomonas

aeruginosa" or "Klebsiella pneumoniae" or "Proteus mirabilis" or "Providencia stuartii" or "Moraxella catarrhalis" or "Streptococcus hemolyticus" or "Streptococcus anginosus" or "streptococcus pneumoniae" or "Gram-positive cocci" or "Gram-negative bacteria" or "gram-negative bacillus" or "pseudomonas aeruginosa" or Candida or "Corynebacterium diphtheriae" or "Bordetella pertussis" or "M.tuberculosis" or "Acute bronchitis" or "acute angina" or "acute tonsillitis" or tonsillitis or "Sinus Infections" or "Viral Pneumonias" or "Acute Coryza" or "Acute viral pharyngitis" or Pharyngitides or "Sore Throats" or "Acute viral laryngitis" or "acute pharyngitis" or "acute epiglottitis" or Epiglottitides or Bronchitis or Laryngitis or Legionellosis or "Lung Abscess" or Blastomycosis or "Pneumonia, Pneumocystis" or Pharyngitis or Nasopharyngitis or "Retropharyngeal Abscess" or Pleurisy or Pleuropneumonia or Pneumonia or Bronchopneumonia or Pleuropneumonia or Rhinitis or Rhinoscleroma or "Severe Acute Respiratory Syndrome" or Sinusitis or "Allergic Fungal Sinusitis" or "Ethmoid Sinusitis" or "Frontal Sinusitis" or "Maxillary Sinusitis" or "Sphenoid Sinusitis" or Supraglottitis or Epiglottitis or Tracheitis or Silicotuberculosis or "Whooping Cough"):ti,ab,kw

#19 #17 or #18

#20 #11 and #19

#### Scopus

( TITLE-ABS-KEY ( ( temperature OR "ambient temperature" OR "seasurface temperature" OR climate OR #1 weather OR meteorological OR "climate change" OR meteorolog\* OR "extreme weather" OR "meteorological factor" OR "climatic factor" OR "hot weather" OR "cold weather" OR "global warming" )) AND TITLE-ABS-KEY (( "Bacterial Pneumonia" OR "Pneumonia, Bacterial" OR "Bacterial Pneumonias" OR "Bacterial Infections and Mycoses" OR "Bacterial Infections" OR "Pneumonia, Pneumococcal" OR "Pneumonia, Staphylococcal" OR "Streptococcus pneumoniae" OR "Chlamydial Pneumonia" OR "Pneumonia, Mycoplasma" OR "Mycoplasma pneumoniae" OR "Schizoplasma pneumoniae" OR "Chlamydophila pneumoniae" OR "Lower respiratory tract" OR 'upper respiratory tract" OR "Human Rhinovirus" OR "Human Enterovirus" OR "Respiratory Tract Infections" OR "Respiratory System Infections" OR "Respiratory System Infection" OR "Upper Respiratory Tract Infections" OR "Upper Respiratory Infections" OR "Upper Respiratory Tract Infection" OR "respiratory infections" OR "respiratory virus" OR "Respiratory syncytial virus" OR "Respiratory Syncytial Viruses" OR "RSV" OR "Chimpanzee Coryza Agent" OR adenoviruses OR "Human Adenovirus" OR "Human Adenoviruses" OR "APC Viruses" OR "APC Virus" OR orthomyxoviruses OR myxoviruses OR "Influenza viruses" OR "Influenza A virus" OR "Influenza A viruses" OR "Orthomyxovirus Type A" OR "Influenza Viruses Type A" OR "Influenza B virus" OR "influenza B viruses" OR "Influenza Viruses Type B" OR "Influenza B virus" OR "Influenzavirus C" OR "Influenza C Virus" OR "Influenza C Viruses" OR "Human parainfluenza viruses" OR "Parainfluenza Virus 2, Human" OR "Parainfluenza Virus Type 2" OR "Croup-Associated Viruses" OR "Croup Associated Viruses" OR "Human parainfluenza virus 2" OR "Croup-Associated Virus" OR "Croup Associated Virus" OR "HPIV" OR metapneumovirus OR metapneumoviruses OR "Human Metapneumoviruses" OR "Human Metapneumovirus" OR "HMPV" OR rhinoviruses OR "Human Rhinovirus" OR "Common Cold Virus" OR "Common Cold Viruses" OR "Coryza Virus" OR coronaviruses OR "SARS coronavirus" OR "Severe acute respiratory syndrome-related coronavirus" OR "Severe acute respiratory syndrome related coronavirus" OR "HCoV-SARS" OR "SARS Virus" OR "SARS-Cov-2" OR "HcoV" OR "NL63" OR "OC43" OR "229E" OR "HKU1" OR sarbecovirus OR sarbecoviruses OR "Severe acute respiratory syndrome coronavirus" OR "Severe Acute Respiratory Syndrome Virus" OR "Human bocaviruses" OR "Coxiella burnetii" OR "Chlamydophila psittaci" OR "Legionella pneumophila" OR "Staphylococcus aureus" OR "Haemophilus influenzae" OR enterovirus OR "Pseudomonas aeruginosa" OR "Klebsiella pneumoniae" OR "Proteus mirabilis" OR "Providencia stuartii" OR "Moraxella catarrhalis" OR "Streptococcus hemolyticus" OR "Streptococcus anginosus" OR "streptococcus pneumoniae" OR "Gram-positive cocci" OR "Gram-negative bacteria" OR "gram-negative bacillus" OR "pseudomonas aeruginosa" OR candida OR "Corynebacterium diphtheriae" OR "Bordetella pertussis" OR "M.tuberculosis" OR "Acute bronchitis" OR "acute angina" OR "acute tonsillitis" OR tonsillitis OR "Sinus Infections" OR "Viral Pneumonias" OR "Acute Coryza" OR "Acute viral pharyngitis" OR pharyngitides OR "Sore Throats" OR "Acute viral laryngitis" OR "acute pharyngitis" OR "acute epiglottitis" OR epiglottitides OR bronchitis OR laryngitis OR legionellosis OR "Lung Abscess" OR blastomycosis OR "Pneumonia, Pneumocystis" OR pharyngitis OR nasopharyngitis OR "Retropharyngeal Abscess" OR pleurisy OR pleuropneumonia OR pneumonia OR bronchopneumonia OR pleuropneumonia OR rhinitis OR rhinoscleroma OR "Severe Acute Respiratory Syndrome" OR sinusitis OR Allergic Fungal Sinusitis" OR "Ethmoid Sinusitis" OR "Frontal Sinusitis" OR "Maxillary Sinusitis" OR "Sphenoid Sinusitis" OR supraglottitis OR epiglottitis OR tracheitis OR silicotuberculosis OR " Whooping Cough" ) ) )

## **EMBASE**

- #1 temperature:ab,ti OR 'ambient temperature':ab,ti OR 'seasurface temperature':ab,ti OR rain\*:ab,ti OR climate:ab,ti OR weather:ab,ti OR meteorological:ab,ti OR 'climate change':ab,ti OR 'extreme weather':ab,ti OR 'meteorological factor':ab,ti OR 'hot weather':ab,ti OR 'cold weather':ab,ti OR snow:ab,ti OR 'climatic factor':ab,ti OR 'global warming':ab,ti
- bacterial pneumonia':ab,ti OR 'pneumonia, bacterial':ab,ti OR 'bacterial pneumonias':ab,ti OR 'bacterial infections':ab,ti OR mycoses:ab,ti OR 'bacterial infections':ab,ti OR 'pneumonia, pneumococcal':ab,ti OR 'pneumonia, staphylococcal':ab,ti OR 'chlamydial pneumonia':ab,ti OR 'pneumonia, mycoplasma':ab,ti OR 'mycoplasma pneumoniae':ab,ti OR 'schizoplasma pneumoniae':ab,ti OR 'chlamydophila pneumoniae':ab,ti OR 'lower respiratory tract':ab,ti OR 'upper respiratory tract infections':ab,ti OR 'respiratory system infections':ab,ti OR 'respiratory tract infections':ab,ti OR 'upper respiratory tract infections':ab,ti OR 'upper respiratory tract infections':ab,ti OR 'respiratory virus':ab,ti OR 'respiratory syncytial virus':ab,ti OR 'respiratory virus':ab,ti OR 'respiratory syncytial virus':ab,ti OR 'respiratory

syncytial viruses':ab,ti OR 'rsv':ab,ti OR 'chimpanzee coryza agent':ab,ti OR adenoviruses:ab,ti OR 'human adenovirus':ab,ti OR 'human adenoviruses':ab,ti OR 'apc viruses':ab,ti OR 'apc virus':ab,ti OR orthomyxoviruses:ab,ti OR myxoviruses:ab,ti OR 'influenza viruses':ab,ti OR 'influenza a virus':ab,ti OR 'influenza a viruses':ab,ti OR 'orthomyxovirus type a':ab,ti OR 'influenza viruses type a':ab,ti OR 'influenza b viruses':ab,ti OR 'influenza viruses type b':ab,ti OR 'influenza b virus':ab,ti OR 'influenzavirus c':ab,ti OR 'influenza c virus':ab,ti OR 'influenza c viruses':ab,ti OR 'human parainfluenza viruses':ab,ti OR 'parainfluenza virus 2, human':ab,ti OR 'parainfluenza virus type 2':ab,ti OR 'croup-associated viruses':ab,ti OR 'croup associated viruses':ab,ti OR 'human parainfluenza virus 2':ab,ti OR 'croup-associated virus':ab,ti OR 'croup associated virus':ab,ti OR 'hpiv':ab,ti OR metapneumovirus:ab,ti OR metapneumoviruses:ab,ti OR 'human metapneumoviruses':ab,ti OR 'human metapneumovirus':ab,ti OR 'hmpv':ab,ti OR rhinoviruses:ab,ti OR 'human rhinovirus':ab,ti OR 'common cold virus':ab,ti OR 'common cold viruses':ab,ti OR 'coryza virus':ab,ti OR coronaviruses:ab,ti OR 'sars coronavirus':ab,ti OR 'severe acute respiratory syndrome-related coronavirus':ab,ti OR 'Enterovirus':ab,ti OR 'severe acute respiratory syndrome related coronavirus':ab,ti OR 'hcov-sars':ab,ti OR 'sars virus':ab,ti OR sarbecovirus:ab,ti OR sarbecoviruses:ab,ti OR 'severe acute respiratory syndrome coronavirus':ab,ti OR 'severe acute respiratory syndrome virus':ab,ti OR 'HcoV':ab,ti OR 'NL63':ab,ti OR 'OC43':ab,ti OR '229E':ab,ti OR 'HKU1':ab,ti OR 'human bocaviruses':ab,ti OR 'coxiella burnetii':ab,ti OR 'chlamydophila psittaci':ab,ti OR 'legionella pneumophila':ab,ti OR 'staphylococcus aureus':ab,ti OR 'haemophilus influenzae':ab,ti OR 'klebsiella pneumoniae':ab,ti OR 'proteus mirabilis':ab,ti OR 'providencia stuartii':ab,ti OR 'moraxella catarrhalis':ab,ti OR 'streptococcus hemolyticus':ab,ti OR 'streptococcus anginosus':ab,ti OR 'streptococcus pneumoniae':ab,ti OR 'gram-positive cocci':ab,ti OR 'gramnegative bacteria':ab,ti OR 'gram-negative bacillus':ab,ti OR 'pseudomonas aeruginosa':ab,ti OR candida:ab,ti OR 'corynebacterium diphtheriae':ab,ti OR 'bordetella pertussis':ab,ti OR 'm.tuberculosis':ab,ti OR 'acute bronchitis':ab,ti OR 'acute angina':ab,ti OR 'acute tonsillitis':ab,ti OR tonsillitis:ab,ti OR 'sinus infections':ab,ti OR 'viral pneumonias':ab,ti OR 'acute coryza':ab,ti OR 'acute viral pharyngitis':ab,ti OR pharyngitides:ab,ti OR 'sore throats':ab,ti OR 'acute viral laryngitis':ab,ti OR 'acute pharyngitis':ab,ti OR 'acute epiglottitis':ab,ti OR epiglottitides:ab,ti OR bronchitis:ab,ti OR laryngitis:ab,ti OR legionellosis:ab,ti OR 'lung abscess':ab,ti OR blastomycosis:ab,ti OR 'pneumonia, pneumocystis':ab,ti OR pharyngitis:ab,ti OR nasopharyngitis:ab,ti OR 'retropharyngeal abscess':ab,ti OR pleurisy:ab,ti OR pneumonia:ab,ti OR bronchopneumonia:ab,ti OR pleuropneumonia:ab.ti OR rhinitis:ab.ti OR rhinoscleroma:ab.ti OR 'severe acute respiratory syndrome':ab.ti OR sinusitis:ab,ti OR 'allergic fungal sinusitis':ab,ti OR 'ethmoid sinusitis':ab,ti OR 'frontal sinusitis':ab,ti OR 'maxillary sinusitis':ab,ti OR 'sphenoid sinusitis':ab,ti OR supraglottitis:ab,ti OR epiglottitis:ab,ti OR tracheitis:ab,ti OR silicotuberculosis:ab,ti OR 'whooping cough':ab,ti

#3 #1 AND #2

#### Web of Science

- #1 TS= (temperature or "ambient temperature" or "seasurface temperature" or climate or weather or meteorological or "climate change" or "extreme weather" or "meteorological factor" or "hot weather" or "cold weather" or "climatic factor" or "global warming")
- #2 TS= ("Bacterial Pneumonia" or "Pneumonia, Bacterial" or "Bacterial Pneumonias" or "Bacterial Infections and Mycoses" or "Bacterial Infections" or "Pneumonia, Pneumococcal" or "Pneumonia, Staphylococcal" or "Streptococcus pneumoniae" or "Chlamydial Pneumonia" or "Pneumonia, Mycoplasma" or "Mycoplasma pneumoniae" or "Schizoplasma pneumoniae" or "Chlamydophila pneumoniae" or "Lower respiratory tract" or "upper respiratory tract" or "Human Rhinovirus" or "Human Enterovirus" or "Respiratory Tract Infections" or "Respiratory System Infections" or "Respiratory System Infections" or "Upper Respiratory Tract Infections" or "Upper Respiratory Infections" or "Upper Respiratory Tract Infection" or "respiratory infections" or "respiratory virus" or "Respiratory syncytial virus" or "Respiratory Syncytial Viruses" or "RSV" or "Chimpanzee Coryza Agent" or Adenoviruses or "Human Adenovirus" or "Human Adenoviruses" or "APC Viruses" or "APC Virus" or Orthomyxoviruses or Myxoviruses or "Influenza viruses" or "Influenza A virus" or "Influenza A viruses" or "Orthomyxovirus Type A" or "Influenza Viruses Type A" or "Influenza B virus" or "influenza B viruses" or "Influenza Viruses Type B" or "Influenza B virus" or "Influenzavirus C" or "Influenza C Virus" or "Influenza C Viruses" or "Human parainfluenza viruses" or "Parainfluenza Virus 2, Human" or "Parainfluenza Virus Type 2" or "Croup-Associated Viruses" or "Croup Associated Viruses" or "Human parainfluenza virus 2" or "Croup-Associated Virus" or "Croup Associated Virus" or "HPIV" or Metapneumovirus or Metapneumoviruses or "Human Metapneumoviruses" or "Human Metapneumovirus" or "HMPV" or Rhinoviruses or "Human Rhinovirus" or "Common Cold Virus" or "Common Cold Viruses" or "Coryza Virus" or Coronaviruses or "SARS coronavirus" or "Severe acute respiratory syndrome-related coronavirus" or "Severe acute respiratory syndrome related coronavirus" or "HCoV-SARS" or "SARS Virus" or "SARS-Cov-2" or "HcoV" or "NL63" or "OC43" or "229E" or "HKU1" or Sarbecovirus or Sarbecoviruses or "Severe acute respiratory syndrome coronavirus" or "Severe Acute Respiratory Syndrome Virus" or "Human bocaviruses" or Enterovirus or "Coxiella burnetii" or "Chlamydophila psittaci" or "Legionella pneumophila" or "Staphylococcus aureus" or "Haemophilus influenzae" or "Pseudomonas aeruginosa" or "Klebsiella pneumoniae" or "Proteus mirabilis" or "Providencia stuartii" or "Moraxella catarrhalis" or "Streptococcus hemolyticus" or "Streptococcus anginosus" or "streptococcus pneumoniae" or "Gram-positive cocci" or "Gram-negative bacteria" or "gram-negative bacillus" or "pseudomonas aeruginosa" or Candida or "Corynebacterium diphtheriae" or "Bordetella pertussis" or "M.tuberculosis" or "Acute bronchitis" or "acute angina" or "acute tonsillitis" or tonsillitis or "Sinus Infections" or "Viral Pneumonias" or "Acute Coryza" or "Acute viral pharyngitis" or Pharyngitides or "Sore Throats" or "Acute viral laryngitis" or "acute pharyngitis" or "acute epiglottitis" or Epiglottitides or Bronchitis or Laryngitis or Legionellosis or "Lung Abscess" or Blastomycosis or "Pneumonia, Pneumocystis" or Pharyngitis or Nasopharyngitis or "Retropharyngeal Abscess" or Pleurisy or Pleuropneumonia or Pneumonia or Bronchopneumonia or Pleuropneumonia or Rhinitis or Rhinoscleroma or

	"Severe Acute Respiratory Syndrome" or Sinusitis or "Allergic Fungal Sinusitis" or "Ethmoid Sinusitis" or "Frontal Sinusitis" or "Maxillary Sinusitis" or "Sphenoid Sinusitis" or Supraglottitis or Epiglottitis or Tracheitis or
	Silicotuberculosis or "Whooping Cough")
ш2	#1 AND #2

## **Chinese National Knowledge Infrastructure (CNKI)**

#1 SU=(温度+环境温度+海表温度+气候+天气+气象+气候变化+极端天气+高温+炎热天气+寒冷天气+气象因素+气候因素+全球变暖)\*(呼吸道感染+肺炎支原体+肺炎衣原体+呼吸道病毒+流感病毒+鼻病毒+腺病毒+呼吸道合胞病毒+人类冠状病毒+博卡病毒+肠道病毒+人偏肺病毒+副流感病毒+严重急性呼吸综合征相关冠状病+严重急性呼吸综合征冠状病毒+革兰氏阳性球菌+溶血性链球菌+草绿色链球菌+肺炎链球菌+金黄色葡萄球菌+革兰氏阴性杆菌+流感嗜血杆菌+嗜麦芽窄食单胞菌+肺囊虫+肺炎克雷伯菌+铜绿假单胞菌+革兰氏阴性球菌+化脓性链球菌+卡他莫拉菌+嗜肺军团菌+鲍曼不动杆菌+念珠菌+结核分枝杆菌+结核杆菌+白喉棒状杆菌+百日咳鲍特菌+细菌性感染和真菌病+细菌性感染+肺炎葡萄球菌+下呼吸道感染+上呼吸道感染+急性支气管炎+急性咽峡炎+急性疱疹性咽峡炎+急性扁桃体炎+急性鼻窦炎+细菌性肺炎+病毒性肺炎+非典型肺炎+普通感冒+急性病毒性咽炎+急性病毒性喉炎+急性咽炎+急性会厌炎+咽炎)

## VIP

#1 U= (温度 OR 环境温度 OR 海表温度 OR 气候 OR 天气 OR 气象 OR 气候变化 OR 极端天气 OR 高温 OR 炎热 天气 OR 寒冷天气 OR 气象因素 OR 气候因素 OR 全球变暖) AND (呼吸道感染 OR 肺炎支原体 OR 肺炎衣原体 OR 呼吸道病毒 OR 流感病毒 OR 鼻病毒 OR 腺病毒 OR 呼吸道合胞病毒 OR 人类冠状病毒 OR 肠道病毒 OR 博卡病毒 OR 人偏肺病毒 OR 副流感病毒 OR 严重急性呼吸综合征相关冠状病 OR 严重急性呼吸综合征 冠状病毒 OR 革兰氏阳性球菌 OR 溶血性链球菌 OR 草绿色链球菌 OR 肺炎链球菌 OR 嗜麦芽窄食单胞菌 OR 肺囊虫 OR 金黄色葡萄球菌 OR 革兰氏阴性杆菌 OR 流感嗜血杆菌 OR 肺炎克雷伯菌 OR 铜绿假单胞菌 OR 革兰氏阴性球菌 OR 卡他莫拉菌 OR 嗜肺军团菌 OR 鲍曼不动杆菌 OR 念珠菌 OR 结核分枝杆菌 OR 结核杆菌 OR 白喉棒状杆菌 OR 百日咳鲍特菌 OR 化脓性链球菌 OR 细菌性感染和真菌病 OR 细菌性感染 OR 肺炎葡萄球菌 OR 下呼吸道感染 OR 上呼吸道感染 OR 急性支气管炎 OR 急性咽峡炎 OR 急性疱疹性咽峡炎 OR 急性扁桃体炎 OR 急性鼻窦炎 OR 细菌性肺炎 OR 病毒性肺炎 OR 非典型肺炎 OR 普通感冒 OR 急性病毒性咽炎 OR 急性病毒性喉炎 OR 急性病毒性喉炎 OR 急性病毒性喉炎 OR 急性病毒性疾炎 OR 急性病毒性疾炎 OR 急性病毒性疾炎 OR 急性病毒性

## WanFang

期名或关键词: (温度 OR 环境温度 OR 海表温度 OR 气候 OR 天气 OR 气象 OR 气候变化 OR 极端天气 OR 高温 OR 炎热天气 OR 寒冷天气 OR 气象因素 OR 气候因素 OR 全球变暖) AND (呼吸道感染 OR 肺炎支原体 OR 肺炎衣原体 OR 呼吸道病毒 OR 流感病毒 OR 鼻病毒 OR 腺病毒 OR 呼吸道合胞病毒 OR 人类冠状病毒 OR 肠道病毒 OR 博卡病毒 OR 人偏肺病毒 OR 副流感病毒 OR 严重急性呼吸综合征相关冠状病 OR 严重急性呼吸综合征冠状病毒 OR 革兰氏阳性球菌 OR 溶血性链球菌 OR 草绿色链球菌 OR 肺炎链球菌 OR 嗜麦芽窄食单胞菌 OR 肺囊虫 OR 金黄色葡萄球菌 OR 革兰氏阴性杆菌 OR 流感嗜血杆菌 OR 肺炎克雷伯菌 OR 铜绿假单胞菌 OR 革兰氏阴性球菌 OR 卡他莫拉菌 OR 嗜肺军团菌 OR 鲍曼不动杆菌 OR 念珠菌 OR 结核分枝杆菌 OR 结核杆菌 OR 白喉棒状杆菌 OR 百日咳鲍特菌 OR 化脓性链球菌 OR 细菌性感染和真菌病 OR 细菌性感染 OR 肺炎葡萄球菌 OR 下呼吸道感染 OR 上呼吸道感染 OR 急性支气管炎 OR 急性咽峡炎 OR 急性疱疹性咽峡炎 OR 急性扁桃体炎 OR 急性鼻窦炎 OR 细菌性肺炎 OR 病毒性肺炎 OR 非典型肺炎 OR 普通感冒 OR 急性病毒性咽炎 OR 急性病毒性喉炎 OR 急性病毒性喉炎 OR 急性病毒性喉炎 OR 急性咽炎 OR 急性咽炎 OR 急性病毒性吸炎 OR 急性病毒性喉炎 OR 急性咽炎 OR 急性医炎 OR 网炎)

## **Chinese BioMedical Literature Database (CBM)**

- #1 ("温度"[不加权:扩展] OR "气象学概念"[不加权:扩展]) OR "气候"[不加权:扩展]) OR "天气"[不加权:扩展] OR "极端天气"[不加权:扩展] OR "天气和发病"[不加权:扩展]) OR "气象学概念"[不加权:扩展])
- #2 "温度" [常用字段:智能] OR"环境温度" [常用字段:智能] OR"海表温度" [常用字段:智能] OR "气候" [常用字段:智能] OR "气候" [常用字段:智能] OR "气候变化" [常用字段:智能] OR "极端天气" [常用字段:智能] OR "高温" [常用字段:智能] OR "炎热天气" [常用字段:智能] OR "寒冷天气" [常用字段:智能] OR "气象因素" [常用字段:智能] OR "气候因素" [常用字段:智能] OR "全球变暖" [常用字段:智能]

# #3 #1 OR #2

- #4 ("呼吸道感染"[不加权:扩展] OR "肺炎支原体"[不加权:扩展] OR "肺炎衣原体"[不加权:扩展] OR "呼吸道合胞病毒"[不加权:扩展] OR "冠状病毒 OC43, 人"[不加权:扩展] OR "人博卡病毒"[不加权:扩展] OR "革兰氏阳性球菌"[不加权:扩展] OR "肠道病毒属"[不加权:扩展] OR "肠道病毒感染"[不加权:扩展] OR "嗜麦芽窄食单胞菌"[不加权:扩展] OR "草绿色链球菌"[不加权:扩展] OR "金黄色葡萄球菌"[不加权:扩展] OR "肺炎克雷伯菌"[不加权:扩展] OR "铜绿假单胞菌"[不加权:扩展] OR "白色念珠菌"[不加权:扩展])
- #5 ("呼吸道感染" [常用字段:智能] OR"肺炎支原体" [常用字段:智能] OR"肺炎衣原体" [常用字段:智能] OR"呼吸道病毒" [常用字段:智能] OR"流感病毒" [常用字段:智能] OR"鼻病毒" [常用字段:智能] OR"腺病毒" [常用字段:智能] OR "專玩意事。 [常用字段:智能] OR "專玩意事。 [常用字段:智能] OR "持卡病毒" [常用字段:智能] OR "所述病毒" [常用字段:智能] OR "人类冠状病毒" [常用字段:智能] OR "博卡病毒" [常用字段:智能] OR "肠道病毒" [常用字段:智能] OR "人偏肺病毒" [常用字段:智能] OR "副流感病毒" [常用字段:智能] OR "严重急性呼吸综合征相关冠状病" [常用字段:智能] OR "严重急性呼吸综合征冠状病毒" [常用字段:智能] OR "革兰氏阳性球菌" [常用字段:智能] OR "溶血性链球菌" [常用字段:智能] OR "草绿色链球菌" [常用字段:智能] OR "声段:智能] OR "革兰氏阴性杆菌" [常用字段:智能] OR "肺炎链球菌" [常用字段:智能] OR "金黄色葡萄球菌" [常用字段:智能] OR "革兰氏阴性杆菌" [常

#6 #4 OR #5 #7 #3 AND #6

### Search terms and protocol

PubMed, Scopus, Embase, the Cochrane Library, Web of Science, China National Knowledge Infrastructure, Wanfang Data, Weipu Database, and China Biology Medicine (CBM) for observational studies were comprehensively searched. we identified studies that explored the effect of relationship between temperature and morbidity, hospital admissions, outpatient visits, mortality, or reported cases of respiratory infection surveillance. Zotero and Covidence tools were used for screening. Additionally manual-searched the literature through forward and backward citation tracing using CoCites to obtain remaining studies

#### **Procedure:**

- 1. Search the terms in the table below
- 2. Download references
- 3. Import document management software
- 3. 3. Review all paper titles for eligibility
- 4. Make record of how many research results were found and how many papers were retrieved by each platform
- 5. Remove duplicates; record the number of records before and after
- 6. Screen these abstracts for eligibility- sometimes a screen of the full paper
- 7. Select the list of final papers for review, sending these references and the number of papers reviewed at each stage to other reviewer
- 8. All authors agree on which studies are ultimately included

### **PECO Statement**

## Panel: Population-Exposure-Comparators-Outcomes framework

## **Population**

assessing respiratory disease risk among the general population

### **Exposure**

per  $1^{\circ}C$  increase in temperature (defined as daily/weekly/monthly/mean/min/max temperature)

#### **Comparators**

Comparable populations not exposed to the relevant ambient temperature

## Outcomes

the risk of morbidity due to pathogen-specific respiratory infections

# Table S3. Conversion of effect estimates

If RR values were not directly provided in the study, we performed the following transformations.

Effect estimates	Conversion
Excessrisk (ER)	RR= (ER÷100) +1
Percentage change	Divide by 100 plus 1
Beta coefficients	exp (beta)
Correlation coefficient r	We use the following steps to calculate a confidence interval for the population correlation coefficient based on the sample size n and the sample correlation coefficient r. In order to normalize the distribution of r and to make the variance independent of p, Fisher (1921) proposed the z-transformation $z_r = \ln\left((1+r)/(1-r)\right)/2$ Find the upper and lower limits of the logarithm $L = z_r - (z \ 1-\alpha/2 \ / \ n-3)$ $U = z_r + (z \ 1-\alpha/2 \ / \ n-3)$ Find the confidence interval Confidence interval $Confidence interval = [(e^{2L} - 1)/(e^{2L} + 1), (e^{2U} - 1)/(e^{2U} + 1)]$ Then, correlation coefficient convert to effect sizes (OR), refer to the method of Lenhard 2022. Lenhard, W. & Lenhard, A. (2022). Computation of effect sizes. Retrieved from: $\frac{https://www.psychometrica.de/effect size.html.}{10.13140/RG.2.2.17823.9232}$
OR, SE	Calculate 95% CI: L= OR-1.96 x SE U= OR+1.96 x SE If the value is negative and does not conform to normal distribution, log conversion is performed: L= exp(log(OR) - 1.96 x log(SE)) U= exp(log(OR) + 1.96 x log(SE))

# Table S4. Risk of Bias Assessment in individual studies

Adapted from: Office of Health Assessment and Translation (OHAT) Risk of Bias Rating Tool for Human and Animal Studies

Bias domains	Questions and criteria	Risk level assessment
Exposure	This includes measurement error or	<b>Definitely low</b> risk of bias (++): There is direct
assessment	measurement limitations.	evidence of low risk of bias practices (the true
	Major considerations are:	average population exposure at temperature has
	• Is the exposure data source clearly	high confidence).
	described, not provided, or difficult to	<b>Probably low</b> risk of bias (+): There is indirect
	<ul><li>interpret?</li><li>Were the temporal and spatial</li></ul>	evidence that the method assessment is robust, it is
	resolution of temperature	deemed that deviations from low risk of bias method for these criteria during the study would
	measurements appropriate (e.g. more	not appreciably bias results, including
	than one weather station,	consideration of direction and magnitude of bias.
	meteorological bureau)?	<b>Probably high</b> risk of bias (-): There is indirect
	daily temperature measurements were	evidence of high risk of bias practices, there is
	available?	insufficient information (e.g., not reported or
	• Were missing data present and/or	"NR"), or two out of the three listed
	imputed?	considerations are not applied (e.g. use of single weather station).
	Were selected lag(s) appropriate based on biological plausibility and	<b>Definitely high</b> risk of bias (): There is direct
	other factors?	evidence of high risk of bias practices, or all listed
		considerations are not applied.
Outcome	This includes measurement error or	<b>Definitely low</b> risk of bias (++): Outcome was
assessment	measurement limitations.	Laboratory-confirmed pathogen-specific
dssessificite	Major considerations are:	respiratory infection and the selected timing
	outcome measurements were not	approximates the occurrence of symptoms.
	influenced by knowledge of the	Probably low risk of bias (+): Outcome was
	exposure (data were from different	clinical diagnosis using standard criteria (e.g.
	databases)	WHO definition and International Classification
	• Was Laboratory-confirmed source for	System – ICD code) and the selected timing
	pathogen-specific studies and/or clinical diagnosis appropriate?	approximates the occurrence of symptoms. <b>Probably high</b> risk of bias (-): outcome was not
	Was selected date or timing of	based on standard diagnosis criteria, there is
	outcome appropriate (i.e. close to	evidence that suggests the existence of
	illness onset)	misclassification bias, or there is insufficient
	Was missing data present and/or	information (e.g. not reported) provided.
	imputed?	<b>Definitely high</b> risk of bias (): Outcome was
		based on self-reports (e.g. parents or family
Confounding	Major considerations are:	members) and selected timing was inappropriate. <b>Definitely low</b> risk of bias (++): Study accounted
bias	Did the study appropriately	for all important confounders which were
Dius	adjusted/accounted for all important	measured consistently.
	well studied potential confounders (e.g.	<b>Probably low</b> risk of bias (+): Study accounted
	time, seasonality, other weather	for most of confounding and is not expected to
	parameters, and model specifications,	introduce bias.
	lag effect)?	<b>Probably high</b> risk of bias (-): Study adjusted for
	•Design or analysis accounted for	some but not all of confounding and is expected to introduce bias.
	potential confounding or modifying •Consideration of other exposures that	Definitely high risk of bias (): Study did not
	might bias results	account for potential confounders or were
	3	inappropriately measured.
Selection bias	Major considerations are:	<b>Definitely low</b> risk of bias (++): Similar baseline
	Was selection of participants into the	characteristics, the descriptions of the selection of
	study done in a manner that might	participants were sufficiently detailed to support
	introduce bias in the study?	the assertion that risk of selection effects was minimal.
	Was similar baseline characteristics, application of inclusion/exclusion	Probably low risk of bias (+): Insufficient
	criteria, recruitment strategy?	information about selection of participants to
	Was all data available collected?	permit a judgment of low risk of bias but there is
	Was there any form of randomisation	indirect evidence that suggests low risk of bias.
	of selection of individual cases?	Probably high risk of bias (-): Insufficient
		information about selection of participants to
		permit a judgment of high risk of bias and there is

		indicat avidance that suggests high right of high
		indirect evidence that suggests high risk of bias. <b>Definitely high</b> risk of bias (): Direct indications
		from descriptions of the participant selection of
		high risk of bias.
Incomplete	Major considerations are:	<b>Definitely low</b> risk of bias (++): no missing data
outcome data	• missing data of outcome measures?	present or missing data does not affect the real
	• missing data of exposures?	results
		<b>Probably low</b> risk of bias (+): indirect evidence
		that suggests low risk of bias (e.g., <10% missing
		data, or missing data related to outcome). <b>Probably high</b> risk of bias (-):indirect evidence
		that suggests high risk bias (e.g., $\geq 10\%$ missing
		data without imputed using appropriate method,
		while rational for attrition explained in the study).
		<b>Definitely high</b> risk of bias (): missing outcome
		data are related to true outcome (e.g., substantial
		missing exposure data ( $\geq 10\%$ ), rationale for
		missing data not explained in the study).
Selective	Major considerations are:	<b>Definitely low</b> risk of bias (++): Complete report
reporting	Was selection of participants into the	of the entire study results and analysis.
	study done in a manner that might	<b>Probably low</b> risk of bias (+): There is not enough
	<ul><li>introduce bias in the study?</li><li>Selective reporting of entire studies,</li></ul>	information to judge the risk of low bias, but there is indirect evidence of low risk of bias. (i.e., effect
	outcomes, or analyses.	estimates presented for less than all hypotheses
	Systematic differences between	tested as per aims; but evidence suggests that
	reported and unreported findings.	effect estimates unlikely to be seriously biased).
	Can include potential for bias in	Probably high risk of bias (-): There is not
	reporting through source of funding	enough information to judge the risk of high bias,
		but there is indirect evidence of high risk of bias.
		<b>Definitely high</b> risk of bias (): There is a high
		risk of selectively reporting entire studies, results, or analyses.
Conflict of	Major considerations are:	<b>Definitely low</b> risk of bias (++): the study was not
interest	Potential source of bias in reporting	funded or funded by government or academic
	through source of funding	institutions or non-profit organizations, and the
	Source of funding (government,	authors declare no conflict of interest
	academic institution, commercial	<b>Probably low</b> risk of bias (+): insufficient
	institution, or no funding? • Is conflict of interest declared?	information to judge for low risk, but indirect
	• Is conflict of interest declared?	evidence suggests study was free of financial interest
		Probably high risk of bias (-): insufficient
		information to judge for high risk, but indirect
		evidence suggests study was not free of financial
		interest
		<b>Definitely high</b> risk of bias (): study was
		supported by a commercial institution with a
		financial interest in the findings, and the authors
Other bias	Bias due to other problems not covered	declare a conflict of interest) <b>Definitely low</b> risk of bias (++): No other bias
Other bias	elsewhere	Probably low risk of bias (++): insufficient
	CISC WHOLE	information to judge for low risk, but indirect
		evidence suggests study was free of other
		problems
		Probably high risk of bias (-): insufficient
		information to judge for high risk, but indirect
		evidence suggests study was not free of other
		problems
		<b>Definitely high</b> risk of bias (): at least one important risk of bias (a.g., selective reporting of
		important risk of bias (e.g., selective reporting of subgroups, a potential source of bias related to the
		subgroups, a potential source of bias related to the specific study design used, study has been claimed
		to have been fraudulent)
	<u>l</u>	to have been maddlemy

Table S5. Overall risk of bias rating

Overall Rating	Combinations (three key components of exposure assessment, outcome assessment, and confounding bias)	
Definitely high (DH)	Definitely high + Definitely High + (Definitely high / Probably high / Probably low / Definitely low)	
	Definitely high + Probably high + Probably high	
Probably high (PH)	Definitely high + Probably high + (Probably low / Definitely low)	
	Probably high + Probably high + (Probably high / Probably low/Definitely	
	low)	
Probably low (PL)	Probably high+(Probably low/Definitely low) + (Probably low / Definitely	
	low)	
	Probably low + Probably low + Definitely Low	
Definitely Low (DL)	Probably low + Definitely low + Definitely low	
	Definitely low + Definitely low + Definitely low	

Table S6. Assessing confidence in the Body of Evidence (adapted from Johnson et al. 2014)

<b>Evaluation factors</b>	Criteria
Risk of bias	Study limitations include a significant risk of bias across the body of evidence.
Indirectness	The research evidence was not similar to the primary objective, i.e., participants, exposure, comparison, Outcome (PECO); Indirect evidence.
Inconsistency	High heterogeneity (I <sup>2</sup> ); variability in results; inconsistent magnitude and direction of effect magnitude.
Imprecision	Small sample size and wide confidence intervals.
Publication bias	The funnel plot is obviously asymmetrical; Egger's test or Begg's tests show that publication bias exists. Trim and Fill procedure was used to estimate potentially missing studies.
Large magnitude of effect	The rating was upgraded if modeling suggested that confounding alone was unlikely to explain associations that were judged to be of large magnitude.
<u>Dose response</u>	If dose-response relationships in one or more studies and/or dose-response across studies are consistent, the rating was upgraded.
Confounding minimizes effect	Upgraded if the consideration of all plausible residual confounders or biases would underestimate the effect or suggest a spurious effect when results show no effect.

From:: Johnson, P. I., Sutton, P., Atchley, D. S., Koustas, E., Lam, J., Sen, S., Robinson, K. A., Axelrad, D. A., & Woodruff, T. J. (2014). The Navigation Guide - evidence-based medicine meets environmental health: systematic review of human evidence for PFOA effects on fetal growth. Environmental health perspectives, 122(10), 1028–1039. https://doi.org/10.1289/ehp.1307893

Table S7. Strength of evidence definitions for human evidence according to the Navigation Guide (adapted from Johnson et al., 2014)

Strength rating	Definition
Sufficient evidence	A positive relationship is observed between exposure and outcome, where chance, bias, and confounding can be ruled out with reasonable confidence. The available evidence includes results from one or more well-designed, well conducted studies, and the conclusion is "unlikely to be strongly affected by the results of future studies."
<u>Limited</u> evidence	A positive relationship is observed between exposure and outcome, where chance, bias, and confounding cannot be ruled out with reasonable confidence. Confidence in the relationship is constrained by factors such as "the number, size, or quality of individual studies" or "inconsistency of findings across individual studies." As more information becomes available, the observed effect could change, and this change may be large enough to alter the conclusion.
Inadequate evidence	"The available evidence is insufficient to assess effects" of the exposure. The evidence is insufficient because of "the limited number or size of studies," low quality of individual studies, or "inconsistency of findings across individual studies." More information may allow an assessment of effects.
Lack of evidence	No relationship is observed between exposure and outcome; and chance, bias, and confounding can be ruled out with reasonable confidence. The available evidence includes consistent results from more than one well-designed, well conducted study at the full range of exposure levels that humans are known to encounter, and the conclusion is unlikely to be strongly affected by the results of future studies. The conclusion is limited to the age at exposure and/or other conditions and levels of exposure studied.

From:: Johnson, P. I., Sutton, P., Atchley, D. S., Koustas, E., Lam, J., Sen, S., Robinson, K. A., Axelrad, D. A., & Woodruff, T. J. (2014). The Navigation Guide - evidence-based medicine meets environmental health: systematic review of human evidence for PFOA effects on fetal growth. Environmental health perspectives, 122(10), 1028–1039. https://doi.org/10.1289/ehp.1307893

Table S8. Full texts excluded on basis of eligibility

Reference	Title	Reason
Sohn S 2019	'Pneumonia Weather': Short-term Effects of Meteorological Factors on Emergency Room Visits Due	Target the disease, not the pathogen
Bishop- Williams KE 2017	to Pneumonia in Seoul, Korea  A protocol for a systematic literature review: comparing the impact of seasonal and meteorological parameters on acute respiratory infections in Indigenous and non-Indigenous peoples	Effect estimates are not applicable
Chen J 2021	Ambient Temperature Is an Independent Risk Factor for Acute Tonsillitis Incidence	Target the disease, not the pathogen
Wenfang G	Assessing the effects of meteorological factors on daily children's respiratory disease hospitalizations: A retrospective study	Effect estimates are not applicable
Zhang Lei 2017	Analysis of Mycoplasma pneumoniae infection among children with respiratory tract infection in hospital in Chengdu from 2013 to 2015	Outcome not applicable
Zhang L 2021	Analysis of mycoplasma pneumoniae infection among children with respiratory tract infections in hospital in Chengdu from 2014 to 2020	Outcome not applicable
Fang J 2021	Association between ambient temperature and childhood respiratory hospital visits in Beijing, China: a time-series study (2013-2017)	Target the disease, not the pathogen
Liu Y 2015	Association between Temperature Change and Outpatient Visits for Respiratory Tract Infections among Children in Guangzhou, China	Effect estimates are not applicable
Ruchiraset A 2020	Association of climate factors and air pollutants with pneumonia incidence in Lampang province, Thailand: findings from a 12-year longitudinal study	Target the disease, not the pathogen
Zhang Y 2015	Burden of respiratory syncytial virus infections in China: Systematic review and meta-analysis	Outcome not applicable
Ashmita Gosai 2009	Climate and respiratory disease in Auckland, New Zealand	Target the disease, not the pathogen
Ghia C 2021	Disease Burden Due to Respiratory Syncytial Virus in Indian Pediatric Population: A Literature Review.	Outcome not applicable
Álvaro-Meca A 2022	Environmental factors linked to hospital admissions in young children due to acute viral lower respiratory infections: A bidirectional case-crossover study	Focusing on lower respiratory tract infections
Falagas ME 2008	Effect of meteorological variables on the incidence of respiratory tract infections	Target the upper and lower respiratory tract infections
Jang JY 2021	Effect of diurnal temperature range on emergency room visits for acute upper respiratory tract infections	Improper temperature
Xie MY 2017	Effect of diurnal temperature range on the outpatient visits for acute bronchitis in children: a time-series study in Hefei, China	Target the disease
Álvaro-Meca A 2022	Environmental factors linked to hospital admissions in young children due to acute viral lower respiratory infections: A bidirectional case-crossover study	Target the lower respiratory tract infection
Li Y 2019	Global patterns in monthly activity of influenza virus, respiratory syncytial virus, parainfluenza virus, and metapneumovirus: a systematic analysis	Outcome not applicable
Hossain MZ 2019	Sociodemographic, climatic variability and lower respiratory tract infections: a systematic literature review	Outcome not applicable
Choe YJ 2019	Seasonality of respiratory viruses and bacterial pathogens	Not specific to bacteria
Nastos PT 2006	Weather impacts on respiratory infections in Athens, Greece	Target the disease
Simmering JE 2017	Weather-Dependent Risk for Legionnaires' Disease, United States	Outcome not applicable
TNS 2012	The seasonal variations of respiratory syncytial virus infections in Turkey: a 2-year epidemiological study	Outcome not applicable
Gleason JA 2016	Under the Weather: Legionellosis and Meteorological Factors	Outcome not applicable
Karagiannis I 2009	Warm, wet weather associated with increased Legionnaires' disease incidence in The Netherlands	Target the disease, not the pathogen

Brandsema PS 2014	Summer increase of Legionnaires' disease 2010 in The Netherlands associated with weather conditions and implications for source findin	Target the disease, not the pathogen
Han XY 2019	Solar and Climate Effects Explain the Wide Variation in Legionellosis Incidence Rates in the United States	Outcome not applicable/Temperature not applicable
Conza L 2013	Meteorological factors and risk of community-acquired Legionnaires' disease in Switzerland: an epidemiological study	Temperature not applicable
Fisman DN 2005	It's not the heat, it's the humidity: wet weather increases legionellosis risk in the greater Philadelphia metropolitan area	Outcome not applicable
Ng V 2008	Going with the flow: legionellosis risk in Toronto, Canada is strongly associated with local watershed hydrology	Temperature not applicable
Yuan L 2011	Relationship between Influenza and Meteorological Conditions in Tianjin	Target the disease, not the pathogen
Paynter S 2013	Sunshine, rainfall, humidity and child pneumonia in the tropics: time-series analyses	Temperature not applicable
WU Zhi-qiang 2022	A single-center study on the relationship between meteorological factors and the number of visits for respiratory diseases in children	Outcome not applicable
JIAN Wanlin 2023	A Stratified Comparative Study on the Meteorological Factors Impacting Respiratory Diseases in the Two Counties of Eastern and Western China	Target the disease, not the pathogen
Paynter S 2014	Respiratory syncytial virus seasonality in tropical Australia	effect estimates not applicable
Michelozzi P 2009	High temperature and hospitalizations for cardiovascular and respiratory causes in 12 european cities	effect estimates not applicable
Shao Lin 2009	Extreme high temperatures and hospital admissions for respiratory and cardiovascular diseases	Outcome not applicable
Habeebullah TM 2021	Impact of outdoor and indoor meteorological conditions on the COVID-19 transmission in the western region of Saudi Arabia	did not have convertible estimates
Donzelli G 2022	Role of meteorological factors on SARS-CoV-2 infection incidence in Italy and Spain before the vaccination campaign. A multi-city time series study	effect estimates not applicable
Kaplin A 2021	Evidence and magnitude of the effects of meteorological changes on SARS-CoV-2 transmission	did not have convertible estimates
Mandal CC 2022	Combinatorial influence of environmental temperature, obesity and cholesterol on SARS-CoV-2 infectivity	did not have convertible estimates
Li Ruiying 2018	Study of the Influence of Meteorological Condition on Children Lower Respiratory Tract Infection and the Prediction Model in Qinhuangdao	Outcome not applicable
Dufloo J 2024	Temperature impacts SARS-CoV-2 spike fusogenicity and evolution	did not have convertible estimates
Thi Khanh HN 2023	The impact of ambient temperature and air pollution on SARS-CoV2 infection and Post COVID-19 condition in Belgium (2021-2022)	Under the combined influence of ambient temperature and air pollution/Not a single temperature effect
Kim J 2023	How Does Climate Change Affect the Upper Airway?	No effect estimation
Grover EN 2024	Does behavior mediate the effect of weather on SARS- CoV-2 transmission? Evidence from cell-phone data	No effect estimation
Lee H 2024	Impact of ambient temperature on respiratory disease: a case-crossover study in Seoul	Target the disease, not the pathogen
Maciorowski D 2021	Environmental factors and their role in the transmission of SARS-CoV-2. Biosaf Health.	No effect estimation
Kaplin A 2021	Evidence and magnitude of the effects of meteorological changes on SARS-CoV-2 transmission	effect estimates not applicable
Nichols GL 2021	Coronavirus seasonality, respiratory infections and weather	Temperature not applicable
Tian Y 2023	Ambient temperature variability and hospital admissions for pneumonia: A nationwide study	Target the disease, not the pathogen
Makrufardi F 2024	Extreme temperatures increase the risk of pediatric pneumonia: a systematic review and meta-analysis	Target the disease, not the pathogen

Miyayo SF	Analysis of Pneumonia Occurrence in Relation to	Outcome not emplicable
2021	Climate Change in Tanga, Tanzania	Outcome not applicable

Table S9. Basic characteristics of 137 included studies

Study	Location	Study period	Time sample	Popul ation	Ages	Reporting frequency	Climate	Case source	Exposure (°C)	Pathogen	Income	Overall bias
Raymundo (2023) <sup>1</sup>	Oahu, USA	2016.05- 2019.05	37	7143	<6 y	Monthly	A	Reported cases	T <sub>mean</sub>	EV/HRV	High	PL
Annabel (2023) <sup>2</sup>	Bukit Timah, Singapore	2009-2019	4016	7329	<5 y	Daily	A	Influenza infection	Tmean	IVA, IVB	High	DH
Chee (2023) <sup>3</sup>	Kuala Lumpur, Malaysia	2017-2021	60	2950	<12 y	Monthly	A	Hospitalized for ARI	T <sub>mean</sub>	RSV	Upper- middle	PH
Keita (2023) <sup>4</sup>	47 prefectures in Japan	2014-2019	312	72170 9	all	Weekly	С	Reported cases	Tmean	RSV	High	PL
Ping-Ing (2023) <sup>5</sup>	Taiwan, China	1995-2005, 2000-2005	206	1740	<18 y	Monthly	A/C	Reported bronchiolitis/pneu monia	T <sub>mean</sub>	RSV	High	PL
Meng (2023) <sup>6</sup>	Singapore	2009-2019	4016	15715	<5 y	Daily	A	Reported cases	$T_{mean,}$ $T_{max}$ , $T_{min}$	RSV	High	PL
You (2022)	Gwangju, Korea	2016-2019	48	4195	all	Monthly	D	Hospitalized with ARI	Tmean	IV, HAdV, PIV, RSV, HCoV, HRV, HBoV, HMPV, EV	High	PH
Zhang (2022) <sup>8</sup>	Jiangyin, China	2015-2020	72	12294	28 d-14 y	Monthly	С	Hospitalized for ARTI	Tmean	RSV, IVA, IVB, HPIV 1-3, HAdV	Upper- middle	PH
Yuan (2022) <sup>9</sup>	Chendu, China	2016-2019	48	5127	28 d-14 y	Monthly	С	Hospitalized for ARI	Tmean	M. pneumoniae	Upper- middle	PH
Na (2022) <sup>10</sup>	Nantong, China	2017-2019	156	51665	<14 y	Weekly	С	Reported pneumonia cases	T <sub>mean</sub>	M. pneumoniae	Upper- middle	PL
Pan (2022) <sup>1</sup>	Shenzhen, China	2009-2015	2555	5744	all	Daily	С	Influenza	T <sub>mean,</sub> T <sub>max</sub> , T <sub>min</sub>	IVA, IVB	Upper- middle	DL
Huang (2022) <sup>12</sup>	Suzhou, China	2016-2019	48	7940	1 m-15 y	Monthly	С	Hospitalized for ARI	Tmean	H. influenzae	Upper- middle	PH
Ming (2022) <sup>13</sup>	Wenzhou, China	2008-2017	121	89898	1–3 y	Monthly	С	Hospitalized with ARI	T <sub>mean</sub>	HPIV-3	Upper- middle	PL
You (2022) <sup>14</sup>	13 European countries	2010-2019	521	30965	<5 y	Weekly	Multiple	Hospitalized with ARI	Tmean	RSV	High	DH
Sung (2022) <sup>15</sup>	Changwon, Korea	2020.1.20- 04.29	100	3234	all	Daily	С	Reported cases	Tmean	SARS-CoV-2	High	PL

Lisa (2022) <sup>16</sup>	Helsinki, Finland	2020.08.01- 2021.05.31	62	48013	all	Daily	С	Reported cases	Tmean	SARS-CoV-2	High	PL
Keita (2022) <sup>17</sup>	Six Japanese prefectures	2020.05- 2022.03	699	29822 62	all	Daily	С	Notified cases	T <sub>mean</sub>	SARS-CoV-2	High	PH
Cheng (2022) <sup>18</sup>	Macao, China	2014-2017	1460	4880	1 m-14 y	Daily	С	Hospitalized for ARI	T <sub>mean</sub>	EV/HRV, HAdV, RSV- A, RSV-B, IVA, IVB, HPIV, HMPV, HBoV, HCoV	High	DL
Zhi (2021) <sup>19</sup>	Suzhou, China	2016-2018	36	1157	1-15 y	Monthly	С	Hospitalized with LRTI	Tmean	RSV	Upper- middle	DH
Eun (2021) <sup>20</sup>	Cheonan, Korea	2012-2018	2556	9010	1–9 y	Daily	D	Reported RI	T <sub>mean</sub>	HPIV 1-3	High	PL
Aleix (2021) <sup>21</sup>	52 Spanish provinces	Weeks 10- 16 of 2020	7	24182 50	all	Weekly	С	Reported cases	$T_{mean,}$ $T_{max}$ , $T_{min}$	SARS-CoV-2	High	PL
Liu (2021) <sup>22</sup>	Moscow	2020.03.11- 06.22	103	Not report ed	all	Daily	D	Reported cases	$T_{mean,}$ $T_{max}$ , $T_{min}$	SARS-CoV	High	PH
Na (2021) <sup>23</sup>	Tianjin, China	2015.06- 2019.02	44	63821	0-16 y	Monthly	D	Hospitalized for ARI	Tmean	M. pneumoniae	Upper- middle	DL
Wu (2021) <sup>24</sup>	Haikou, China	2018-2020	36	309	5.05 m	Monthly	A	Hospitalized for RTI	Tmean	RSV, RSV-A, RSV-B	Upper- middle	PL
Chen (2021) <sup>25</sup>	Guangdong, China	2020.01.21- 02.26	36	1347	all	Daily	С	Reported cases	Tmean	SARS-CoV-2	Upper- middle	PL
Stacy (2021) <sup>26</sup>	Singapore	2009-2019	4016	6393	<5 y	Daily	A	Reported cases	T <sub>mean</sub>	HPIV	High	PL
Eun (2021) <sup>27</sup>	Cheonan, Korea	2012-2018	85	9010	<10 y	Monthly	D	Reported RTI	Tmean	HAdV	High	PL
Can (2021) <sup>28</sup>	China	2004-2017	522	19865 36	all	Weekly	C/D	Reported cases	Tmean	IVA/H1N1pdm09, IVA/H3N2, IVB	Upper- middle	PH
Chan (2021) <sup>29</sup>	Hong Kong, China	1993-1997	60	9635	<5 y	Monthly	С	Admission with RI	T <sub>mean</sub>	RSV	High	PL
Dong (2021) <sup>30</sup>	Cheonan, Korea	2012-2018	85	1920	all	Monthly	D	Hospital with RI	Tmean	HRV	High	DH
Therese (2021) <sup>31</sup>	Nairobi, Kenya	2007–2013	85	17261	all	Monthly	A	Reported influenza-like illness	T <sub>mean</sub>	RSV, HPIV 1-3, HAdV	Upper- middle	PL
Lei (2021) <sup>32</sup>	Chengdu, China	2014-2020	85	22882	1 m-17 y	Monthly	С	Hospitalized with RTI	Tmean	M. pneumoniae	Upper- middle	PL
Mohamed (2021) <sup>33</sup>	Safat, Kuwait	2020.02.24- 05.30	96	176	all	Daily	A	Reported cases	Tmean	SARS-CoV-2	High	PH

Liu (2021) <sup>34</sup>	Chongqing, China	2009.06- 2019.06	3681	3107	1 m-18 y	Daily	С	Hospitalized for ARI	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	RSV	Upper- middle	PL
Rosalie (2021) <sup>35</sup>	Amsterdam, The Netherlands	2003-2016	170/730	2161	≤ 24 m	Monthly/ Weekly	С	Admissions for bronchiolitis	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	RSV	High	PL
Jang (2020) <sup>36</sup>	Cheonan, Korea	2012-2018	2556	9010	<10 y	Daily	D	Reported cases	Tmean	HCoV- 229E, HCoV- OC43	High	PL
Ilada (2020) <sup>37</sup>	Bangkok, Thailand	2012-2018	85	8209	≤ 5 y	Monthly	A	Reported ILI cases	Tmean	RSV	Upper- middle	PH
Asmaa (2020) <sup>38</sup>	Riyadh, Saudi Arabia	2012.10- 2018.12	75	712	all	Monthly	A	Reported cases	T <sub>mean</sub>	MERS-CoV	High	DL
Sheikh (2020) <sup>39</sup>	Bukit Timah, Singapore	2005-2015	4016	9905	< 30 m	Daily	A	Reported cases	T <sub>mean,</sub> T <sub>max</sub> , T <sub>min</sub>	RSV	High	PL
Shiv (2020) <sup>40</sup>	Fifty U.S. states and Washington D.C.	2020.1.22- 2020.3.4	42	974	all	Daily	Multiple	Reported cases	T <sub>max</sub>	SARS-CoV-2	High	PL
Zhang (2020) <sup>41</sup>	Suzhou, China	2014-2017	48	7525	2.51±15.93 m	Monthly	С	Hospitalized for CAP	T <sub>mean</sub>	HPIV- 3	Upper- middle	PH
Simone (2020) <sup>42</sup>	Milan; Florence; Trento, Italy	2020.03.08- 2020.06.19	103	Not report ed	all	Daily	C/D	Reported cases	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	SARS-CoV-2	High	PL
Kong (2020) <sup>43</sup>	Wuhan, China	2020.01.20- 02.11	22	Not report ed	all	Daily	С	Reported cases	T <sub>mean</sub>	SARS-CoV-2	Upper- middle	PH
Giselmo (2020) <sup>44</sup>	Maranhao, Brazil	2018.04- 2019.03	12	151	2–12 y	Monthly	A	Hospitalized with RTI	Tmean, Tmax, Tmin	RSV, HRV	Upper- middle	PL
Li (2020) <sup>45</sup>	Shanghai, China	2017.07- 2019.06	104	29135	0-18 y	Weekly	С	Reported cases	$T_{mean}$ , $T_{max}$ , $T_{min}$	IVA, IVB	Upper- middle	PH
Rory (2019) <sup>46</sup>	Edinburgh, Scotland, UK	2009.04- 2015.11	81	52060	all ages	Monthly	С	Hospitalized with RTI	Tmean	HPIV 1-3, RSV, HMPV, IVA, IVB, HRV, HAdV	High	PL
Huang (2019) <sup>47</sup>	Zhongshan, China	2014.09- 2016.08	24	13705	1 m-14 y	Monthly	С	Hospitalized with CAP	$T_{ m mean}$	M. pneumoniae, C. pneumoniae, M. catarrhalis, IVA, IVB, HAdV, HPIV, RSV, H. influenzae, S.pneumoniae	Upper- middle	PH
Pan (2019) <sup>48</sup>	Panzhihua, China	2006-2015	121	6892	all	Monthly	С	Influenza infection	Tmean	IVA, IVB	Upper- middle	DL

He (2019) <sup>49</sup>	Suzhou, China	2012-2015	48	8711	<18 y	Monthly	С	Hospitalized for RTI	Tmean	HBoV	Upper- middle	PL
Wu (2019) <sup>50</sup>	Wu zhishan, China	2015-2016	24	1597	1 m-14 y	Monthly	A	Hospitalized for ARI	T <sub>mean</sub>	M. pneumoniae	Upper- middle	PH
Wei (2019) <sup>51</sup>	Jinan, China	2013-2016	208	9170	all	Weekly	D	Reported cases	Tmean	IVA/H1N1pdm09, IVA/H3N2, IVB	Upper- middle	PL
Yong (2019) <sup>52</sup>	Seoul, Korea	2007-2016	521	23694	<18 y	Weekly	D	Hospitalized for ARI	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	HMPV	High	PL
Katerina (2019) <sup>53</sup>	Liverpool, UK	2011.11- 2017.03	65	374	18-50 y	Monthly	С	Reported cases	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	S. pneumoniae	High	PL
Magali (2018) <sup>54</sup>	Dijon, France	2011-2016	90	4300	all	Weekly	С	Hospitalized for ARI	T <sub>mean</sub>	HMPV, RSV	High	PL
Wen-Kuan (2018) <sup>55</sup>	Guangzhou, China	2009.07- 2016.06	85	11399	≤14 y	Monthly	С	Hospitalized for ARI	Tmean	HBoV-1	Upper- middle	PL
Morley (2018) <sup>56</sup>	Gold Coast, Australia	2007.07- 2016.06	469	15387	<5 y	Weekly	С	Hospitalized with ARI	$T_{mean,} \\ T_{max}, T_{min}$	RSV	High	PL
Hermann (2018) <sup>57</sup>	Garoua, Cameroon	2014-2016	36	1666	0-5 y	Monthly	D	Reported influenza cases	T <sub>mean</sub>	IVA, IVB	Lower- middle	PL
Ines (2018) <sup>58</sup>	Sousse, Northern Africa	2003–2015	158	5131	≤ 5 y	Monthly	В	Hospitalized for bronchiolitis	Tmean	RSV	Lower- middle	PL
Jung (2018) <sup>59</sup>	Seoul, Korea	2005-2012	97	9113	< 3 y	Monthly	D	Hospitalized with bronchiolitis/pneu monia	T <sub>mean</sub>	RSV	High	PL
Adriana (2018) <sup>60</sup>	Toronto, Canada	2010-2015	312	44362	1 m-108 y	Weekly	D	Reported cases	T <sub>mean</sub>	IVA, IVB	High	PL
Benjamin (2018) <sup>61</sup>	Nha Trang, central Vietnam	2007.01.29- 2012.4.26	63	2998	all	Monthly	A	Hospitalized with ARI	T <sub>mean</sub>	IVA, IVB, RSV, HMPV, HPIV 1-4, HRV, HCoV-229E, HCoV-OC43, HAdV, HBoV	Lower- middle	PL
Zhou (2018) <sup>62</sup>	Suzhou, China	2013-2015	36	5994	17 d-15 y	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	HRV	Upper- middle	DH
Jang (2017) <sup>63</sup>	Cheonan, Korea	2006.12- 2014.02	88	6279	<1 y	Monthly	D	Reported cases	T <sub>mean</sub>	RSV	High	PL
Terezinha (2017) <sup>64</sup>	Sao Paulo, Brazil	1996-2010	182	Not report ed	<5 y	Monthly	С	Hospitalized with bronchiolitis/pneu monia	T <sub>mean</sub>	RSV	Upper- middle	PL

Wang (2017) <sup>65</sup>	Suzhou, China	2006-2013	97	15098	1 m-14 y	Monthly	С	Hospitalized with ARI	Tmean	M. pneumoniae	Upper- middle	PL
Gwladys (2017) <sup>66</sup>	Yaounde´, Cameroon	2009.01- 2015.10	360	5216	all	Weekly	A	Reported ILI/ARI cases	Tmean	IVA, IVB	Lower- middle	DH
Yan (2017) <sup>67</sup>	Suzhou, China	2013-2015	36	6194	<14 y	Monthly	С	Hospitalized with LRTI	T <sub>mean</sub>	HRV	Upper- middle	PL
Tian (2017) <sup>68</sup>	Hangzhou, China	2015	365	36500	4.3-52.8 m	Daily	С	Reported pneumonia cases	$T_{mean,}$ $T_{max}$ , $T_{min}$	RSV, M. pneumoniae	Upper- middle	PH
Zhou (2017) <sup>69</sup>	Huangshi, China	2016	12	2326	1 m-14 y	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	HRV	Upper- middle	PL
Raffaella (2017) <sup>70</sup>	Rome, Italy	2004-2014	133	723	< 1 y	Monthly	С	Hospitalized for acute bronchiolitis	T <sub>mean</sub>	RSV	High	PL
Wei (2016) <sup>71</sup>	Shanghai, China	2011.08- 2014.12	41	2819	2 m-12 y	Monthly	С	Hospitalized with ARI	T <sub>mean</sub>	IFV, HPIV 1-4, EV, RSV, HCoV, HAdV, HMPV, HBoV, HRV	Upper- middle	PL
Qing (2016) <sup>72</sup>	Hangzhou, China	2015	365	36500	4.3 m	Daily	С	Hospitalized with ARI	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	RSV	Upper- middle	DH
PSOTER (2016) <sup>73</sup>	48 states of USA	2003-2009	2556	3463	<7 y	Daily	A/B/C/ D	Reported cases	Tmean	P.aeruginosa	High	PH
Nicklas (2016) <sup>74</sup>	Gothenburg, Sweden	2010.10- 2013.07	147	20062	<18 y	Weekly	С	Hospitalized for RTI	T <sub>mean</sub>	IVA, IVB, RSV, HRV, EV, HCoV, HMPV, HAdV, HPIV 1–4, HBoV, C. pneumoniae, M. pneumoniae	High	DH
Shi (2016) <sup>75</sup>	Zhengzhou, China	2012.01- 2015.01	36	2323	18-94 y	Monthly	D	Hospitalized for ARI	Tmean	M. pneumoniae	Upper- middle	PH
Chen (2016) <sup>76</sup>	Yancheng, China	2014-2015	24	3672	1 m-10 y	Monthly	С	Hospitalized for ARI	Tmean	RSV, IVA, IVB, HPIV 1-3, HAdV	Upper- middle	PH
Sobral (2020) <sup>77</sup>	USA	2019.12- 2020.03	118	Not report ed	all	Daily	Multiple	Reported cases	$T_{mean,}$ $T_{max}$ , $T_{min}$	SARS-CoV-2	High	PL
Jaqueline (2016) <sup>78</sup>	Curitiba, Brazil	2012-2013	24	755	all	Monthly	С	Hospitalized for ARI	Tmean	HRV	Upper- middle	PL
Tiina (2016) <sup>79</sup>	Kajaani, Finland	2004.07- 2005.01	214	386	all	Daily	С	Reported cases	T <sub>mean</sub>	HRV	High	DH

Gamba- Sanchez (2016) <sup>80</sup>	Bogota, Colombia	2009-2013	60	13488	<3 y	Monthly	С	Reported cases	T <sub>mean</sub>	RSV	Upper- middle	DL
Rodriguez- Martinez (2015) <sup>81</sup>	Bogota, Colombian	2010.01- 2011.04	16	3931	<3 y	Monthly	С	Hospitalized for ARI	Tmean	RSV	Upper- middle	DL
Geng (2015) <sup>82</sup>	Suzhou, China	2001-2011	133	42664	3 d-15 y	Monthly	С	Hospitalized for ARI	Tmean	RSV	Upper- middle	PL
Daniel (2015) <sup>83</sup>	Mallorca, Spain	2006-2011	312	60659	1– 15 y	Weekly	С	Reported pharyngitis cases	T <sub>mean</sub>	S. pyogenes	High	PL
Lu (2015) <sup>84</sup>	Suzhou, China	2010-2014	60	1803	≤28 d	Monthly	С	Hospitalized with LRTI	T <sub>mean</sub>	RSV	Upper- middle	PL
Cui (2015) <sup>85</sup>	Chaoshan, China	2010.12- 2011.11	364	1074	0–16 у	Daily	С	Hospitalized for ARI	T <sub>mean</sub>	RSV, IVA, IVB, HCoV, HMPV, HPIV 1-4, HRV, EV, HAdV, HBoV	Upper- middle	DH
Lin (2015) <sup>86</sup>	Quanzhou, China	2013	12	6020	0-15 y	Monthly	С	Hospitalized for ARI	$T_{ m mean}$	RSV, HAdV, IVA, IVB, HPIV, M. pneumoniae, C.pneumoniae, L. pneumophila	Upper- middle	PL
Yan (2015) <sup>87</sup>	Shanghai, China	2011-2013	36	2526	0-12 y	Monthly	С	Hospitalized for ARI	Tmean	HPIV 1-4	Upper- middle	PL
Huang (2015) <sup>88</sup>	Suzhou, China	2013	12	1926	1-140 m	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	HRV	Upper- middle	PH
Ni (2014) <sup>89</sup>	Suzhou, China	2006-2011	72	10596	1-13 y	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	M. pneumoniae	Upper- middle	PL
Sun (2014) <sup>90</sup>	Suzhou, China	2006-2010	60	8143	1-180 m	Monthly	С	Hospitalized with ARI	Tmean	M. catarrhalis	Upper- middle	PL
Onozuka (2014) <sup>91</sup>	Fukuoka, Japan	2006-2012	365	30215	all	Weekly	С	Reported cases	T <sub>mean</sub> , T <sub>max</sub> , T <sub>min</sub>	RSV	High	PH
Zheng (2014) <sup>92</sup>	Suzhou, China	2001-2011	133	42104	1 m-14 y	Monthly	С	Hospitalized with ARI	Tmean	RSV, IVA, IVB, HPIV 1-3, HAdV	Upper- middle	PH
Chen (2014) <sup>93</sup>	Suzhou, China	2009-2012	48	7626	<14 y	Monthly	С	Hospitalized with LRTI	T <sub>mean</sub>	HBoV	Upper- middle	PL
Chen (2013) <sup>94</sup>	Suzhou, China	2006	12	1598	1 m-13 y	Monthly	С	Hospitalized for ARI	Tmean	M. pneumoniae, C. pneumoniae	Upper- middle	PL
Tang (2013) <sup>95</sup>	Suzhou, China	2009	12	1883	1 m-10 y	Monthly	С	Hospitalized for RTI	T <sub>mean</sub>	RSV	Upper- middle	PH

Wan (2013) <sup>96</sup>	Suzhou, China	2007-2011	60	28871	3 d-15 y	Monthly	С	Reported cases	Tmean	RSV	Upper- middle	PH
Wang (2013) <sup>97</sup>	Guilin, China	2011-2012	24	1342	30 d-14 y	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	M. pneumoniae	Upper- middle	PL
Zheng (2013) <sup>98</sup>	Luoyang, China	2011	12	256	8-23 m	Monthly	D	Hospitalized for ARI	T <sub>mean</sub>	M. pneumoniae	Upper- middle	PL
Yin (2013) <sup>99</sup>	Suzhou, China	2006-2009	48	8368	1 m-14 y	Monthly	С	Hospitalized for ARI	Tmean	M. pneumoniae	Upper- middle	PL
Chen (2013) <sup>100</sup>	Suzhou, China	2006-2010	60	8157	1 m-14 y	Monthly	С	Hospitalized with RTI	Tmean	M. pneumoniae	Upper- middle	DH
Li (2013) <sup>101</sup>	Zhuhai, China	2010	12	924	1 m-78 y	Monthly	С	Reported ILI cases	T <sub>mean</sub>	IVA/H3N2, IVA/H1N1pdm09, IVB, HPIV 1-3, RSV, HMPV, HAdV	Upper- middle	PL
Wang (2013) <sup>102</sup>	Suzhou, China	2006 - 2009	48	6655	30 d-10 y	Monthly	С	Hospitalized for ARI	Tmean	HMPV	Upper- middle	PL
Chen (2012) <sup>103</sup>	Suzhou, China	2009-2010	24	998	1-24 m	Monthly	С	Hospitalized for bronchiolitis	Tmean	RSV, HBoV, HPIV-3, IVA, HAdV, HMPV, M. pneumoniae	Upper- middle	PH
Chee-Sieng (2012) <sup>104</sup>	Kuala Lumpur, Malaysia	1982-2008	328	10269	≤5 y	Monthly	A	Hospitalized with RTI	T <sub>mean</sub>	RSV	Upper- middle	PL
Chen (2012) <sup>105</sup>	Suzhou, China	2006-2010	60	8197	1-120 m	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	HAdV	Upper- middle	PH
TNS (2012) <sup>106</sup>	Turkey four regions	2008.05- 2010.09	29	3464	<2 y	Monthly	C/D	bronchiolitis/pneu monia	Tmean	RSV	Upper- middle	PL
Daniel Hervás (2012) <sup>107</sup>	Mallorca, Spain	1995-2006	146/625	2384	≤ 2 y	Monthly/ Weekly	С	Hospitalized for bronchiolitis	$T_{mean,}$ $T_{max}$ , $T_{min}$	RSV	High	PL
Susana (2012) <sup>108</sup>	Valencia, Spain	2006-2009	48	243	> 18 y	Monthly	С	Hospitalized for CAP	Tmean	S.pneumoniae, L. pneumophila	High	PH
Ji (2011) <sup>109</sup>	Suzhou, China	2006-2009	48	6655	1 m -10 y	Monthly	С	Hospitalized for ARI	Tmean	RSV, IVA, IVB, HPIV 1-3, HAdV, HMPV, HBoV	Upper- middle	PL
Sun (2011) <sup>110</sup>	Suzhou, China	2006-2009	48	6655	1-120 m	Monthly	С	Hospitalized for RI	T <sub>mean</sub>	HPIV-3	Upper- middle	PH
Wang (2011) <sup>111</sup>	Suzhou, China	2006-2009	48	6599	1-120 m	Monthly	С	Hospitalized for ARI	T <sub>mean</sub>	HMPV	Upper- middle	PH
Loh (2011) <sup>112</sup>	Bukit Timah, Singapore	2003.08– 2008.12	65	44026	0–5 у	Weekly	A	Hospitalized with RTI	T <sub>mean,</sub> T <sub>max</sub> , T <sub>min</sub>	IVA, IVB, RSV, HPIV 1–3, HAdV	High	PL
Cristiana (2010) <sup>113</sup>	Salvador, Brazil	2003.09- 2005.05	20	184	<5 y	Monthly	A	Reported pneumonia cases	T <sub>mean</sub>	IVA, IVB, RSV, HPIV 1-3, HAdV, <i>H</i> .	Upper- middle	PL

	4					1		I	1	: <i>a</i> - c	1	1
										influenzae, S. pneumoniae, M.		
										pneumoniae, M. pneumoniae, C.		
										pneumoniae, M.		
										catarrhalis		
Cao	Beijing, China	2009.08.03-	97	Not	all	Daily	D	Reported ILI	Tmean	IVA/H1N1pdm09	Upper-	PH
$(2010)^{114}$		11.08		report			_	cases	- 11104111	- · · · · · · · · · · · · · · · · · · ·	middle	
(====)				ed								
Jean-	Rheinland-Pfalz.	2001-2006	312	3044	≤16 y	Weekly	С	Hospitalized for	Tmean	IVA, HAdV, HMPV,	High	PH
Baptist	Mainz				3			ARI		HRV, $CoV$ , $M$ .	8	
$(2009)^{115}$										pneumoniae, C.		
,										pneumoniae		
Noyola	San Luis Potosí,	2002.10-	191	1393	<18 y	Weekly	D	Hospitalized with	Tmean	RSV	High	PH
$(2009)^{116}$	Mexico	2006.05			·	_		LRTI				
37.	77 1	2004.06	25	007	2.25	36 11				Day His Hib House	T	DI
Maria	Kathmandu,	2004.06-	37	887	2-35 m	Monthly	A	Reported	T <sub>mean</sub>	RSV, IVA, IVB, HPIV	Lower	PL
$(2009)^{117}$	Nepal	2007.06						pneumonia cases		1-3, HMPV		
Onozuka	Fukuoka,Tokyo,	1999-2007	469	13056	<15 y	Weekly	С	Reported	Tmean	M. pneumoniae	High	PL
$(2009)^{118}$	Japan	1777 2007	407	13030	\13 y	Weekiy		pneumonia cases	1 mean	m. pheumoniae	Ingn	1 L
(200))	Jupun							pricamona cases				
		*****		0.720	1.0	1.5		77 1 11 1 2				
Tang	Hong Kong,	2000.05-	93	8539	<18 y	Monthly	C	Hospitalized for	Tmean	IVA, IVB, RSV	High	PL
$(2009)^{119}$	China	2007.12						RTI				
Onozuka	Fukuoka, Japan	1999-2007	432	13056	all	Weekly	С	Reported cases	Tmean	M. pneumoniae	Upper-	PH
$(2008)^{120}$	T ditaona, tapan	1,,,, 200,		10000		,, coning		reported eases	2 mean	111 procurrence	middle	
Asma	Doha, Qatar	2002.01-	69	3121	≤ 2 y	Monthly	A	Reported cases	Tmean	RSV	High	PL
$(2008)^{121}$	., .,	2007.09			_ 3			1	moun		8	
Yuan	Beijing, China	2003.04.03 -	69	2522	all	Daily	С	Reported cases	Tmean,	SARS-CoV-2	Upper-	PL
$(2006)^{122}$	J 3, 2, 1	06.11						· F	Tmax, Tmin		middle	
Bi (2006) <sup>123</sup>	Beijing and	2003.04.21-	29	Not	all	Daily	C/D	Notified cases	T <sub>mean</sub> .	SARS-CoV	Upper-	PL
21 (2000)	Hong Kong,	05.20		report		Zuny	0,2	Troumed cases	Tmax, Tmin	51115 00 1	middle	12
	China			ed					- 11111			
Wang	Hangzhou,	2001-2003	36	13642	1 m-13 y	Monthly	С	Hospitalized for	T <sub>mean</sub>	RSV	Upper-	PL
$(2005)^{124}$	China				•			pneumonia			middle	
	G 1	2002.01.01	127	501	.11	D. '1	A /C/D	•	T	CADC C V		DII
Feng	Guangzhou,	2003.01.01-	137	581	all	Daily	A/C/D	Reported cases	T <sub>mean</sub> ,	SARS-CoV	Upper-	PH
$(2005)^{125}$	Hong Kong,	05.18							T <sub>max</sub> , T <sub>min</sub>		middle/Hi	
	Toronto,										gh	
	Singapore, Taiwan, Beijing											
Mariana	Buenos Aires,	1998 –2002	60	18561	/5 xx	Monthly	С	Hospitalized with	Tmean	RSV, HPIV, IVA, IVB,	High	PL
$(2004)^{126}$		1998 -2002	00	10301	<5 y	Monthly		Hospitalized with ARI	1 mean	HAdV	High	FL
(2004)***	Argentina							AKI	<u> </u>	паи		

Huang (2004) <sup>127</sup>	Guangdong, China	2003.01-05	129	1491	all	Daily	С	Reported cases	$T_{mean,}$ $T_{max}$ , $T_{min}$	SARS-CoV	Upper- middle	PH
Hailin (2019) <sup>128</sup>	Wenzhou, China	2008-2017	121	89898	<18 y	Monthly	С	Hospitalized with RTI	Tmean	RSV	Upper- middle	PL
Liu (2019) <sup>129</sup>	Guangzhou, China	2009.07- 2016.06	85	11398	≤14 y	Monthly	С	Hospitalized with ARI	Tmean	RSV, HPIV, HMPV	Upper- middle	DL
Natalia (2016) <sup>130</sup>	Athens, Greece	2002-2013	146	7516	0-14 y	Monthly	С	Hospitalized with ARI	T <sub>mean</sub>	RSV	High	PH
Omer (2008) <sup>131</sup>	Lombok, Indonesia	2000-2002	1095	2878	< 2 y	Daily	A	Reported pneumonia cases	$T_{mean,}$ $T_{max}, T_{min}$	RSV	Upper- middle	DH
Patrick (2015) <sup>132</sup>	Kuala Lumpur, Malaysia	1982-1997	194	5691	< 2 y	Monthly	A	Reported bronchiolitis and pneumonia	Tmean	RSV	Upper- middle	DL
Santiago (2004) <sup>133</sup>	Leon, Spain	1995.10- 2000.06	247	221	< 18 y	Weekly	С	Hospitalized with bronchitis/bronchi olitis/pneumonia	T <sub>mean,</sub> T <sub>max</sub> , T <sub>min</sub>	RSV	High	PH
Silvia (2013) <sup>134</sup>	Bologna, Italy	2007-2010	208	327	< 2 y	Weekly	С	Reported cases	$T_{\min}$	RSV	High	PL
Virginia (2015) <sup>135</sup>	Nine states (Arizona, California, Colorado, Iowa, Massachusetts, Maryland, New Jersey, Washington, and Wisconsin), United States	1989-2009	255	Not report ed	all	Monthly	С	Reported acute bronchiolitis	$T_{mean,}$ $T_{max}, T_{min}$	RSV	High	PL
Zhang (2013) <sup>136</sup>	Suzhou, China	2001-2011	133	42664	3 d -15 y	Monthly	С	Hospitalized with ARI	Tmean	RSV	Upper- middle	PH
Tamara (2009) <sup>137</sup>	Amsterdam, Netherlands	1998-2005	417	10672	< 6 m	Weekly	С	Reported cases	$T_{ m min}$	RSV	High	PL

Note: \*d: day(s), wk: week(s), m: month(s), y: year(s); T<sub>mean:</sub> mean temperature; T<sub>max:</sub> maximum temperature; T<sub>min:</sub> minimum temperature; TNS: Turkish Neonatal Society; C: Temperate; A: Tropical; B: Arid; D: Continental; ILI: Influenza-like illness; LRTI: lower respiratory tract infection; ARI: acute respiratory infection; CAP: Community acquired pneumonia; VRI: viral respiratory infection; RSV: Respiratory syncytial virus; IAV: influenza A virus; IBV: influenza B virus; HAdV: human adenovirus; HPIV: human parainfluenza virus; HMPV: human metapneumovirus; HBoV: human bocavirus; HRV: human rhinovirus; EV: enterovirus; HCoV: human coronavirus; EV/HRV: enterovirus/human rhinovirus; MERS-CoV: Middle East Respiratory Syndrome Coronavirus; SARS-CoV: Severe Acute Respiratory Syndrome Coronavirus; M. pneumoniae: Mycoplasma pneumoniae: S. pneumoniae: Streptococcus pneumoniae: C. pneumoniae: Chlamydophila pneumoniae; L. pneumophila: Legionella pneumophila; H. influenzae: Haemophilus influenzae; M. catarrhalis: Moraxella catarrhalis; P. aeruginosa: Pseudomonas aeruginosa; S. pyogenes: Streptococcus pyogenes pharyngitis

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Table S10. Global distribution of common respiratory tract specific pathogen infections studies and sites (applied to the Figure 2)

(applied to the Figure 2)	Number of studies	Number of sites
Country	Number of studies	Number of sites
Leon, Spain	1	1
Valencia, Spain	1	1
Mallorca, Spain	2	1
52 Spanish provinces	1	1
Yaounde, Cameroon	1	1
Garoua, Cameroon	1	1
Riyadh, Saudi Arabia	1	1
Nha Trang, central Vietnam	1	1
Sousse, Northern Africa	1	1
Turkey four regions: Marmara, Mediterranean and Aegean region; Black Sea region; East Anatolian region; Middle Anatolian region	1	4
Kathmandu, Nepal	1	1
Nairobi, Kenya	1	1
Doha, Qatar	1	1
Safat, Kuwait	1	1
Lombok, Indonesia	1	1
Bangkok, Thailand	1	1
Athens, Greece	1	1
Toronto, Canada	1	1
Gold Coast, Australia	1	1
Buenos Aires, Argentina	1	1
Moscow	1	1
Dijon, France	1	1
Gothenburg, Sweden	1	1
Kajaani, Finland	1	1
Helsinki, Finland	1	1
Bukit Timah, Singapore	3	1
Singapore	2	1
Bologna, Italy	1	1
Milan; Florence; Trento, Italy	1	3
Rome, Italy	1	1
Amsterdam, the Netherlands	2	1
Nine states (Arizona, California, Colorado, Iowa, Massachusetts, Maryland, New Jersey, Washington, and Wisconsin), United States	1	9
Rhode Island, USA	1	1
Oahu, USA	1	1
48 states of USA	1	1
Kuala Lumpur, Malaysia	3	1
San Luis Potosí, Mexico	1	1
Rheinland-Pfalz, Mainz	1	1
Curitiba, Brazil	1	1
Sao Paulo, Brazil	1	1
Maranhao, Brazil	1	1
	1 *	1 *

Salvador, Brazil	1	1
Edinburgh, UK	1	1
Liverpool, UK	1	1
Bogota, Colombian	2	1
Cheonan, Korea	5	1
Changwon, Korea	1	1
Gwangju, Korea	1	1
Seoul, Korea	2	1
Fukuoka, Tokyo, Japan	3	1
Six Japanese prefectures (Hokkaido, Tokyo, Aichi, Osaka, Fukuoka, and Okinawa)	1	6
46 Okinawa prefecture, Japan	1	1
47 prefectures in Japan	1	1
13 European countries (Austria, Denmark, Estonia, France, Germany, Greece, Ireland, Netherlands, Poland, Portugal, Slovenia, Spain, United Kingdom)	1	13
Fifty U.S. states and Washington D.C.	1	1
Guangzhou, Hong Kong, Toronto, Singapore, Taiwan, Beijing	1	6
Guangzhou, China	2	1
Guangdong, China	1	1
Zhongshan, China	1	1
Zhuhai, China	1	1
Chaoshan, China	1	1
Shenzhen, China	1	1
Guangzhou, Jiangmen, Foshan, Shenzhen, Zhaoqing, and Zhongshan in Guangdong Province, China	1	6
Hong Kong, China	2	1
Beijing and Hong Kong, China	1	2
Beijing, China	2	1
Shanghai, China	3	1
Wenzhou, China	2	1
Hangzhou, China	3	1
Chengdu, China	2	1
Macao, China	1	1
Chongqing, China	1	1
Panzhihua, China	*	1
- manada, Cama	1	1
Wu zhishan, China	1	1
	1	1
Wu zhishan, China	1 1	1 1
Wu zhishan, China Henan, China	1 1 1	1 1 1
Wu zhishan, China Henan, China Luoyang, China	1 1 1 1	1 1 1 1
Wu zhishan, China Henan, China Luoyang, China Zhengzhou, China Jinan, China	1 1 1 1	1 1 1 1 1
Wu zhishan, China Henan, China Luoyang, China Zhengzhou, China	1 1 1 1 1	1 1 1 1 1
Wu zhishan, China Henan, China Luoyang, China Zhengzhou, China Jinan, China Nantong, China Taiwan, China	1 1 1 1 1 1	1 1 1 1 1 1 1
Wu zhishan, China Henan, China Luoyang, China Zhengzhou, China Jinan, China Nantong, China Taiwan, China Quanzhou, China	1 1 1 1 1 1 1	1 1 1 1 1 1 1
Wu zhishan, China Henan, China Luoyang, China Zhengzhou, China Jinan, China Nantong, China Taiwan, China	1 1 1 1 1 1	1 1 1 1 1 1 1

Haikou, China	1	1
Yancheng, China	1	1
Jiangyin, China	1	1
Wuhan, China	1	1
Suzhou, China	26	1

Table S11. Risk of bias (RoB) Assessment for each study (n=137)

Study	Exposure assessment	Outcome assessment	Confounding bias	Selection bias	Incomplete missing data	Selective outcome reporting	Conflict of interest	Other bias
Santiago (2004)	PL	PL	DL	DL	DL	DL	DL	PH
Zhang (2013)	PL	PL	PH	DL	PL	DL	PL	PL
Liu (2019)	PL	DL	DL	DL	DL	DL	DL	DL
Natalia (2016)	PL	DL	PL	PL	PH	DL	DL	DL
Silvia (2013)	PL	DL	PL	PL	DL	DL	PL	PL
Tamara (2009)	PH	DL	DL	DL	DL	DL	DL	PL
Omer (2008)	PL	DL	PL	PL	DL	DH	PL	PL
Virginia (2015)	PL	PL	PL	DL	PL	PL	DL	DL
Patrick (2015)	PL	DL	DL	DL	DL	DL	DL	DL
Hailin (2019)	PL	PL	DL	PL	PL	PL	PL	DL
Zheng (2014)	PL	PL	DL	DL	DL	PH	DL	DL
Ming (2022)	DL	PL	PL	DL	PL	DL	DL	DL
Eun (2021)	PL	DL	DL	DL	DL	DL	DL	DL
Ilada (2020)	DL	PH	DL	DL	DL	DL	DL	PL
Nicklas (2016)	DL	PL	DH	DL	DH	PH	DL	DL
Jean-Baptist (2009)	PL	PL	PH	PH	DL	DL	DL	DL
Noyola (2009)	DL	PL	PL	PL	PL	PL	PH	DL
Rodriguez-Martinez (2015)	PL	DL	DL	DL	DL	DL	DL	DL
Young (2019)	PH	PL	DL	DL	DL	DL	DL	PL
Zhi (2021)	PL	DH	PL	DH	DL	DL	DL	DL

Huang (2022)	PL	PL	PL	PL	PL	PL	PH	DL
Gamba-Sanchez (2016)	PL	DL						
Cheng (2022)	DL							
Jang (2020)	PL	PL	DL	PL	PL	DL	DL	PL
Chee (2023)	PH	DL						
Wen-Kuan (2018)	PL	PL	DL	PL	DL	PL	DL	DL
Li (2020)	PL	PH	DL	PL	DL	PL	PL	DL
Chen (2014)	PL	PL	DL	PL	DL	PL	DL	DL
Chen (2013)	PL	PL	DL	PL	DL	PL	DL	DL
Liu (2021)	PH	DL						
Wang (2005)	DL	DL	PL	DL	PL	PL	DL	PL
Chen (2014)	PH	DL						
Giselmo (2020)	PL	DL	DL	PL	PL	PL	DL	DL
Chen (2013)	PL	DL	DH	PH	PL	DL	DL	DL
Rory (2019)	PL	DL	DL	PL	DL	PL	DL	DL
Onozuka (2009	PL	DL	DL	PL	PL	DL	DL	DL
Keita (2022)	PL	PH	PH	DL	DL	PL	DL	DL
Mohamed (2021)	PL	PL	PH	DL	PL	PH	DL	DL
Yuan (2006)	PL	DL	PL	PL	PL	PL	DL	DL
Tian (2017)	PL	PH	DL	DL	DL	DL	DL	DL
Tiina (2016)	PL	PL	DL	DL	DH	DL	DL	DL
Cui (2015)	PL	DL	DH	DL	DH	DL	DL	DL
Yan (2017)	PL	DL	PL	DL	PL	DL	DL	DL

Lei (2021)	PL	PL	PL	PL	DL	DL	PL	DL
Jaqueline (2016)	PL	DL	PL	PL	PL	DL	DL	DL
Lu (2015)	PL	PL	PL	DL	PL	DL	DL	DL
Ji (2011)	PL	DL	PL	DL	PL	DL	DL	DL
Geng (2015)	PL	DL	PL	DL	PL	PL	DL	DL
Daniel (2015)	PL	DL	PL	PL	PL	DL	DL	DL
Cristiana (2010)	DL	PL	DL	DL	PL	PL	DL	DL
Susana (2012)	PH	PL	DL	DL	DL	DL	DL	DL
Maria (2009)	DL	DL	PL	PL	PL	PL	PL	DL
Katerina (2019)	PL	PL	DL	DL	DL	PL	PL	DL
Magali (2018)	PL	PL	DL	DL	DL	PL	DL	DL
Loh (2011)	PL	PL	DL	DL	DL	PL	DL	DL
Yong (2019)	PL	PL	DL	DL	PL	PL	DL	DL
Shi (2016)	PH	PL	DL	DL	DL	DL	PL	DL
Therese (2021)	DL	PL	PL	DL	DL	DL	PL	DL
Li (2013)	DL	PL	PL	PL	PL	DL	PL	DL
TNS (2012)	DL	PL	DL	DL	PL	DL	PL	DL
Wang (2013)	PL	PL	PL	DL	PL	PL	PL	DL
Daniel Hervás (2012)	PL	PL	DL	DL	PL	DL	DL	DL
Asma (2008)	PL	PL	DL	PL	PL	DL	PL	PL
Raffaella (2017)	PL	PL	DL	DL	PL	DL	PL	DL
Sheikh (2020)	PL	PL	DL	DL	PL	DL	PL	PL
Ines (2018)	DL	PL	DL	PL	PL	PL	PL	DL

Sun (2014)	DL	PL	DL	DL	PL	PL	PL	DL
You (2022)	PH	PL	DL	DL	PL	PL	PL	DL
Jung (2018)	DL	PL	DL	DL	PL	PL	PL	DL
Adriana (2018)	DL	PL	DL	DL	PL	PL	PL	DL
Chee-Sieng (2012)	DL	PL	DL	DL	PL	PL	PL	DL
Huang (2019)	PH	PL	DL	DL	DL	DL	DL	DL
Wei (2019)	PL	PL	DL	DL	PL	DL	PL	PL
Tang (2009)	PL	PL	DL	DL	PL	DL	PL	DL
Morley (2018)	PL	PL	DL	DL	PL	DL	PL	DL
Hermann (2018)	PL	PL	DL	DL	PL	DL	PL	DL
Rosalie (2021)	PL	PL	DL	DL	DL	DL	PL	DL
Wei (2016)	PL	PL	PL	DL	DL	DL	PL	DL
Keita (2023)	PL	PL	DL	DL	DL	DL	PL	DL
Qing (2016)	PL	PL	DL	DH	PL	DL	PL	DL
Jang (2017)	DL	PL	DL	DL	PL	DL	PL	PL
You (2022)	PL	PL	PL	DH	PL	DL	PL	DL
Terezinha (2017)	PL	PL	DL	DL	PL	DL	PL	DL
Raymundo (2023)	PL	PL	DL	DL	PL	DL	PL	DL
Wang (2017)	PL	DL	DL	DL	PL	DL	PL	DL
Gwladys (2017)	PL	PL	DL	DL	DH	DL	PL	DL
Annabel (2023)	PL	PL	PL	DH	DH	DL	PL	DL
Aleix (2021)	PL	PL	DL	DL	PL	DL	PL	DL

Bi (2006)	PL	DL	DL	DL	PL	DL	PL	DL
Mariana (2004)	PL	PL	DL	DL	PL	DL	PL	DL
Pan (2022)	PH	PL	DL	PL	DL	DL	PL	DL
Asmaa (2020)	DL	PL	DL	DL	DL	DL	PL	DL
Pan (2019)	DL	PL	DL	DL	DL	DL	PL	DL
Can (2021)	DL	PL	DL	DL	DH	DL	PL	DL
Na (2022)	DL	PL	DL	DL	PL	DL	PL	DL
Stacy (2021)	PL	DL	DL	DL	PL	DL	PL	DL
Eun (2021)	PL	PL	DL	DL	PL	DL	PL	DL
Ping-Ing (2023)	PL	PL	DL	DL	PL	DL	PL	DL
Chan (2021)	PL	PL	DL	DL	PL	DL	PL	DL
Dong (2021)	DL	PL	DL	DL	DH	DL	PL	DL
Onozuka (2014)	DL	PH	DL	DL	DL	DL	PL	DL
Benjamin (2018)	PL	PL	DL	DL	PL	DL	PL	DL
Shiv (2020)	DL	PL	DL	PL	DL	DL	DL	DL
PSOTER (2016)	PH	PL	DL	PL	DL	DL	DL	DL
Meng (2023)	DL	PL	DL	PL	DL	DL	DL	DL
Lin (2015)	DL	PL	DL	DL	DL	PL	DL	DL
Wang (2013)	DL	PL	DL	PL	DL	DL	DL	DL
Chen (2012)	PH	PL	DL	PL	DL	DL	DL	DL
Sun (2011)	PH	PL	DL	PL	PL	DL	DL	DL
Onozuka (2008)	PH	PL	DL	PL	DL	DL	DL	DL

Zhou (2018)	PH	DL	ОН	DL	DL	DL	DL	DL
Wu (2019)	PH	DL						
Zhou (2017)	PL	DL	DL	DL	PL	DL	DL	DL
Na (2021)	PL	DL						
Huang (2015)	PH	DL	DL	PH	DL	DL	DL	DL
Wang (2011)	PH	DL	DL	DL	PL	DL	DL	DL
Yuan (2022)	PH	PL	DL	DL	DL	DL	DL	DL
Ni (2014)	PL	PL	DL	DL	DL	DL	DL	DL
Yin (2013)	PL	PL	PL	PL	DL	DL	DL	DL
Zheng (2013)	PL	PL	DL	PL	DL	DL	DL	DL
Sobral (2020)	PL	PL	DL	PL	DL	DL	DL	DL
Wu (2021)	PL	PL	DL	PL	DL	DL	DL	DL
Tang (2013)	PH	DL	DL	PL	DL	DL	DL	DL
Wan (2013)	PH	DL						
Chen (2016)	PH	DL	PL	PL	DL	DL	DL	DL
Zhang (2022)	PH	PH	DL	PH	DL	DL	DL	DL
He (2019)	PL	DL	DL	PL	DL	DL	DL	DL
Yan (2015)	PL	DL	DL	PL	DL	DL	DL	DL
Zhang (2020)	PH	DL	DL	PL	DL	DL	DL	DL
Cao (2010)	PL	PH	DL	PL	DL	DL	DL	DL
Chen (2021)	PL	DL	DL	PL	DL	DL	DL	DL
Feng (2005)	PH	DL	DL	PL	DL	DL	DL	DL

Liu (2021)	PL	DL	DL	PL	DL	DL	DL	DL
Kong (2020)	PL	PH	DL	PL	DL	DL	DL	DL
Huang (2004)	PH	DL	DL	PL	DL	DL	DL	DL
Sung (2022)	PL	DL	DL	PL	DL	DL	DL	DL
Lisa (2022)	PL	PL	DL	PL	DL	DL	DL	DL
Simone (2020)	PL	DL	DL	PL	DL	DL	DL	DL

Note: PL- Probably Low risk; PH- Probably High risk; DH- Definitely High risk; DL- Definitely Low risk

Table S12. Summary of included studies corresponding to each pathogen

Pathogen	Study
Viral infection	
RSV	
Respiratory syncytial virus -A	Zhi (2021) 19 Cheng (2022) 18 Jang (2017) 63 Wu (2021) 24
Respiratory syncytial virus -B	Zhi (2021) 19 Cheng (2022) 18 Jang (2017) 63 Wu (2021) 24
Respiratory syncytial virus - unsubtyped	Santiago (2004) <sup>133</sup> Zhang (2013) <sup>136</sup> Liu (2019) <sup>129</sup> Natalia (2016) <sup>130</sup> Silvia (2013) <sup>134</sup> Omer (2008) <sup>131</sup> Virginia (2015) <sup>135</sup> Patrick (2015) <sup>132</sup> Hailin (2019) <sup>128</sup> Zheng (2014) <sup>92</sup> Ilada (2020) <sup>37</sup> Nicklas (2016) <sup>74</sup> Jean-Baptist (2009) <sup>115</sup> Noyola (2009) <sup>116</sup> Rodriguez-Martinez (2015) <sup>81</sup> Gamba-Sanchez (2016) <sup>80</sup> Chee (2023) <sup>3</sup> Chen (2012) <sup>103</sup> Liu (2021) <sup>34</sup> Giselmo (2020) <sup>44</sup> Rory (2019) <sup>46</sup> Tian (2017) <sup>68</sup> Cui (2015) <sup>85</sup> Lu (2015) <sup>84</sup> Cristiana (2010) <sup>113</sup> Maria (2009) <sup>117</sup> Magali (2018) <sup>54</sup> Therese (2021) <sup>31</sup> Li (2013) <sup>101</sup> Daniel Hervás (2012) <sup>107</sup> Asma (2008) <sup>121</sup> Raffaella (2017) <sup>70</sup> Sheikh (2020) <sup>39</sup> Ines (2018) <sup>58</sup> You (2022) <sup>7</sup> Chee-Sieng (2012) <sup>104</sup> Rosalie (2021) <sup>35</sup> Wei (2016) <sup>71</sup> Keita (2023) <sup>4</sup> Qing (2016) <sup>72</sup> Terezinha (2017) <sup>64</sup> You (2022) <sup>14</sup> Mariana (2004) <sup>126</sup> Ping-Ing (2023) <sup>5</sup> Chan (2021) <sup>29</sup> Onozuka (2014) <sup>91</sup> Benjamin (2018) <sup>61</sup> Meng (2023) <sup>6</sup> TNS (2012) <sup>106</sup> Wang (2005) <sup>124</sup> Geng (2015) <sup>82</sup> Jung (2018) <sup>59</sup> Lin (2015) <sup>86</sup> Tang (2013) <sup>95</sup> Wan (2013) <sup>96</sup> Chen (2016) <sup>76</sup> Zhang (2022) <sup>8</sup> Ji (2011) <sup>109</sup>
Overall respiratory syncytial virus (RSV)	Zhi (2021) <sup>19</sup> Cheng (2022) <sup>18</sup> Jang (2017) <sup>63</sup> Wu (2021) <sup>24</sup> Santiago (2004) <sup>133</sup> Zhang (2013) <sup>136</sup> Liu (2019) <sup>129</sup> Natalia (2016) <sup>130</sup> Silvia (2013) <sup>134</sup> Omer (2008) <sup>131</sup> Virginia (2015) <sup>135</sup> Patrick (2015) <sup>132</sup> Hailin (2019) <sup>128</sup> Zheng (2014) <sup>92</sup> Ilada (2020) <sup>37</sup> Nicklas (2016) <sup>74</sup> Jean-Baptist (2009) <sup>115</sup> Noyola (2009) <sup>116</sup> Rodriguez-Martinez (2015) <sup>81</sup> Gamba-Sanchez (2016) <sup>80</sup> Chee (2023) <sup>3</sup> Chen (2012) <sup>103</sup> Liu (2021) <sup>34</sup> Giselmo (2020) <sup>44</sup> Rory (2019) <sup>46</sup> Tian (2017) <sup>68</sup> Cui (2015) <sup>85</sup> Lu (2015) <sup>84</sup> Cristiana (2010) <sup>113</sup> Maria (2009) <sup>117</sup> Magali (2018) <sup>54</sup> Therese (2021) <sup>31</sup> Li (2013) <sup>101</sup> Daniel Hervás (2012) <sup>107</sup> Asma (2008) <sup>121</sup> Raffaella (2017) <sup>70</sup> Sheikh (2020) <sup>39</sup> Ines (2018) <sup>58</sup> You (2022) <sup>7</sup> Chee-Sieng (2012) <sup>104</sup> Rosalie (2021) <sup>35</sup> Wei (2016) <sup>71</sup> Keita (2023) <sup>4</sup> Qing (2016) <sup>72</sup> Terezinha (2017) <sup>64</sup> You (2022) <sup>14</sup> Mariana (2004) <sup>126</sup> Ping-Ing (2023) <sup>5</sup> Chan (2021) <sup>29</sup> Onozuka (2014) <sup>91</sup> Benjamin (2018) <sup>61</sup> Meng (2023) <sup>6</sup> TNS (2012) <sup>106</sup> Wang (2005) <sup>124</sup> Geng (2015) <sup>82</sup> Jung (2018) <sup>59</sup> Lin (2015) <sup>86</sup> Tang (2013) <sup>95</sup> Wan (2013) <sup>96</sup> Chen (2016) <sup>76</sup> Zhang (2022) <sup>8</sup> Ji (2011) <sup>109</sup>
IV	(2013) Wall (2013) Chen (2010) Enang (2022) 31 (2011)
Influenza virus A	Zheng (2014) <sup>92</sup> Jean-Baptist (2009) <sup>115</sup> Cheng (2022) <sup>18</sup> Li (2020) <sup>45</sup> Chen (2012) <sup>103</sup> Rory (2019) <sup>46</sup> Cui (2015) <sup>85</sup> Maria (2009) <sup>117</sup> Adriana (2018) <sup>60</sup> Gwladys (2017) <sup>66</sup> Annabel (2023) <sup>2</sup> Mariana (2004) <sup>126</sup> Benjamin (2018) <sup>61</sup> Ji (2011) <sup>109</sup> Zhang (2022) <sup>8</sup> Lin (2015) <sup>86</sup> Chen (2016) <sup>76</sup>
Influenza virus A/H1N1pdm09	Li (2013) <sup>101</sup> Wei (2019) <sup>51</sup> Pan (2019) <sup>48</sup> Pan (2022) <sup>11</sup> Can (2021) <sup>28</sup> Cao (2010) <sup>114</sup>
Influenza virus A/H3N2	Li (2013) <sup>101</sup> Adriana (2018) <sup>60</sup> Wei (2019) <sup>51</sup> Pan (2019) <sup>48</sup> Can (2021) <sup>28</sup>
Influenza virus B	Zheng (2014) <sup>92</sup> Nicklas (2016) <sup>74</sup> Jean-Baptist (2009) <sup>115</sup> Cheng (2022) <sup>18</sup> Li (2020) <sup>45</sup> Rory (2019) <sup>46</sup> Maria (2009) <sup>117</sup> Li (2013) <sup>101</sup> Adriana (2018) <sup>60</sup> Gwladys (2017) <sup>66</sup> Mariana (2004) <sup>126</sup> Pan (2022) <sup>11</sup> Ji (2011) <sup>109</sup> Lin (2015) <sup>86</sup> Chen (2016) <sup>76</sup> Zhang (2022) <sup>8</sup> Can (2021) <sup>28</sup> Wei (2019) <sup>51</sup>
Influenza virus B/Victoria	Pan (2019) <sup>48</sup>
Influenza virus B/Yamagata	Pan (2019) <sup>48</sup>
Influenza virus -unsubtyped	You (2022) <sup>7</sup> Adriana (2018) <sup>60</sup> Gwladys (2017) <sup>66</sup> Wei (2016) <sup>71</sup>
Overall influenza virus (IV)	Zheng (2014) <sup>92</sup> Jean-Baptist (2009) <sup>115</sup> Li (2020) <sup>45</sup> Chen (2014) <sup>93</sup> Rory (2019) <sup>46</sup> Cui (2015) <sup>85</sup> Maria (2009) <sup>117</sup> Adriana (2018) <sup>60</sup> Gwladys (2017) <sup>66</sup> Annabel (2023) <sup>2</sup> Mariana (2004) <sup>126</sup> Benjamin (2018) <sup>61</sup> Cao (2010) <sup>114</sup> Ji (2011) <sup>109</sup> Li (2013) <sup>101</sup> Wei (2019) <sup>51</sup> Pan (2022) <sup>11</sup> Can (2021) <sup>28</sup> Nicklas (2016) <sup>74</sup> Cheng (2022) <sup>18</sup> Lin (2015) <sup>86</sup> Chen (2016) <sup>76</sup> Zhang (2022) <sup>8</sup> Pan (2019) <sup>48</sup> You (2022) <sup>7</sup> Wei (2016) <sup>71</sup>
HPIV	g ( ) ( ) ( )
Human parainfluenza virus -1	Eun (2021) <sup>20</sup> Jean-Baptist (2009) <sup>115</sup> Ji (2011) <sup>109</sup> Maria (2009) <sup>117</sup> Chen (2016) <sup>76</sup> Zheng (2014) <sup>92</sup>
Human parainfluenza virus -2	Eun (2021) <sup>20</sup> Ji (2011) <sup>109</sup> Maria (2009) <sup>117</sup> Zheng (2014) <sup>92</sup> Chen (2016) <sup>76</sup>
	Chen (2016) <sup>76</sup> Cui (2015) <sup>85</sup> Eun (2021) <sup>20</sup> Jean-Baptist (2009) <sup>115</sup>

	Ji (2011) <sup>109</sup> Maria (2009) <sup>117</sup> Ming (2022) <sup>13</sup> Zheng (2014) <sup>92</sup> Sun (2011) <sup>110</sup> Chen (2016) <sup>76</sup> Zhang (2022) <sup>8</sup> Zhang (2020) <sup>41</sup>
Human parainfluenza virus -4	Cui (2015) <sup>85</sup>
Human parainfluenza virus - unsubtyped	Huang (2019) <sup>47</sup> Li (2013) <sup>101</sup> Therese (2021) <sup>31</sup> Stacy (2021) <sup>26</sup> Wei (2016) <sup>71</sup> Yan (2015) <sup>87</sup> Cheng (2022) <sup>18</sup> Liu (2019) <sup>129</sup> Mariana (2004) <sup>126</sup> You (2022) <sup>7</sup> Lin (2015) <sup>86</sup>
Overall human parainfluenza virus (HPIV)	Eun (2021) <sup>20</sup> Jean-Baptist (2009) <sup>115</sup> Ji (2011) <sup>109</sup> Maria (2009) <sup>117</sup> Chen (2016) <sup>76</sup> Zheng (2014) <sup>92</sup> Maria (2009) <sup>117</sup> Ming (2022) <sup>13</sup> Sun (2011) <sup>110</sup> Zhang (2022) <sup>8</sup> Zhang (2020) <sup>41</sup> Cui (2015) <sup>85</sup> Huang (2019) <sup>47</sup> Li (2013) <sup>101</sup> Therese (2021) <sup>31</sup> Stacy (2021) <sup>26</sup> Wei (2016) <sup>71</sup> Yan (2015) <sup>87</sup> Cheng (2022) <sup>18</sup> Liu (2019) <sup>129</sup> Mariana (2004) <sup>126</sup> You (2022) <sup>7</sup> Lin (2015) <sup>86</sup>
HMPV	(2022) Liu (2019) Mariana (2004) 1 Ou (2022) Lin (2013)
Human metapneumoviruses (HMPV)	Rory (2019) <sup>46</sup> Nicklas (2016) <sup>74</sup> Liu (2019) <sup>129</sup> Jean-Baptist (2009) <sup>115</sup> Chen (2012) <sup>103</sup> Maria (2009) <sup>117</sup> Magali (2018) <sup>54</sup> Yong (2019) <sup>52</sup> Li (2013) <sup>101</sup> Wang (2013) <sup>102</sup> You (2022) <sup>7</sup> Ji (2011) <sup>109</sup> Cheng (2022) <sup>18</sup> Wang (2011) <sup>111</sup>
HRV	
Human rhinovirus (HRV)	Jean-Baptist (2009) <sup>115</sup> Giselmo (2020) <sup>44</sup> Cui (2015) <sup>85</sup> Tiina (2016) <sup>79</sup> Yan (2017) <sup>67</sup> Jaqueline (2016) <sup>78</sup> You (2022) <sup>7</sup> Dong (2021) <sup>30</sup> Benjamin (2018) <sup>61</sup> Zhou (2018) <sup>62</sup> Zhou (2017) <sup>69</sup> Huang Huang (2015) <sup>88</sup>
HAdV	
Human adenovirus (HAdV)	Zheng (2014) <sup>92</sup> Nicklas (2016) <sup>74</sup> Jean-Baptist (2009) <sup>115</sup> Cheng (2022) <sup>18</sup> Chen (2014) <sup>93</sup> Rory (2019) <sup>46</sup> Therese (2021) <sup>31</sup> Li (2013) <sup>101</sup> You (2022) <sup>7</sup> Mariana (2004) <sup>126</sup> Eun (2021) <sup>27</sup> Ji (2011) <sup>109</sup> Lin (2015) <sup>86</sup> Chen (2012) <sup>105</sup> Chen (2016) <sup>76</sup> Zhang (2022) <sup>8</sup>
HBoV	
Human bocavirus -1	Wen-Kuan (2018) <sup>55</sup>
Human bocavirus -unsubtyped	Cui (2015) <sup>85</sup> Cheng (2022) <sup>18</sup> Chen (2014) <sup>93</sup> You (2022) <sup>7</sup> Nicklas (2016) <sup>74</sup> Chen (2012) <sup>103</sup> Ji (2011) <sup>109</sup> He (2019) <sup>49</sup>
Overall human bocavirus virus (HBoV)	Cui $(2015)^{85}$ Cheng $(2022)^{18}$ Wen-Kuan $(2018)^{55}$ Chen $(2014)^{103}$ You $(2022)^7$ Nicklas $(2016)^{74}$ Chen $(2014)^{103}$ Ji $(2011)^{109}$ He $(2019)^{49}$
EV	
Enterovirus (EV)	You (2022) <sup>7</sup> Cui (2015) <sup>85</sup> Jean-Baptist (2009) <sup>115</sup> Wei (2016) <sup>71</sup>
CoV	
HCoV -229E	Jang (2020) <sup>36</sup>
HCoV -OC43	Jang (2020) <sup>36</sup>
HCoV -unsubtyped	You (2022) <sup>7</sup> Jean-Baptist (2009) <sup>115</sup> Nicklas (2016) <sup>74</sup> Cui (2015) <sup>85</sup> Wei (2016) <sup>71</sup>
MERS-CoV	Asmaa (2020) <sup>38</sup>
SARS-CoV-2	Shiv (2020) <sup>40</sup> Aleix (2021) <sup>21</sup> Keita (2022) <sup>17</sup> Chen (2021) <sup>25</sup> Sung (2022) <sup>15</sup> Lisa (2022) <sup>16</sup> Kong (2020) <sup>43</sup> Simone (2020) <sup>42</sup>
SARS-CoV	Bi (2006) <sup>123</sup> Feng (2005) <sup>125</sup> Liu (2021) <sup>22</sup> Huang (2004) <sup>127</sup>
HRV/EV	
Human rhinovirus/enterovirus (HRV/EV)	Cheng (2022) <sup>18</sup> Raymundo (2023) <sup>1</sup>
Bacterial infection	
Streptococcus pneumoniae	Susana (2012) <sup>108</sup> Cristiana (2010) <sup>113</sup> Katerina (2019) <sup>53</sup> Onozuka (2009) <sup>118</sup>
Streptococcus pyogenes pharyngitis	Daniel (2015) <sup>83</sup>
Moraxella catarrhalis	Huang (2019) <sup>47</sup> Sun (2014) <sup>90</sup>
Pseudomonas aeruginosa	PSOTER (2016) <sup>73</sup>
Legionella pneumophila	Susana (2012) <sup>108</sup> Lin (2015) <sup>86</sup>
Haemophilus influenzae	Huang (2022) <sup>12</sup>
Mycoplasma pneumoniae	Jean-Baptist (2009) <sup>115</sup> Chen (2012) <sup>103</sup> Chen (2013) <sup>94</sup> Huang (2019) <sup>47</sup> Tian (2017) <sup>68</sup> Na (2022) <sup>10</sup> Wang (2017) <sup>65</sup> Lei (2021) <sup>32</sup> Onozuka (2008) <sup>120</sup> Yuan (2022) <sup>9</sup> Lin (2015) <sup>86</sup> Wang (2013) <sup>97</sup> Wu (2019) <sup>50</sup> Na (2021) <sup>23</sup> Ni (2014) <sup>89</sup> Yin (2013) <sup>99</sup> Zheng (2013) <sup>98</sup> Shi (2016) <sup>75</sup> , Chen (2013) <sup>100</sup>

Chlamydia pneumoniae	Huang (2019) <sup>47</sup> Chen (2013)	9 <sup>94</sup> Jean-Baptist (2009) <sup>115</sup> Lin (2015) <sup>86</sup>
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Note: Studies that were included in systematic reviews but not in meta-analyses: Mohamed (2021)<sup>33</sup> Morley (2018)<sup>56</sup> Hermann (2018)<sup>57</sup> Loh (2011)<sup>112</sup> Yuan (2006)<sup>122</sup> Tang (2009)<sup>119</sup> Sobral (2020)<sup>77</sup> Tamara (2009)<sup>137</sup>

Table S13. Meta-regression results

Outcome	Covariates	Coef.	Std. Err.	Z	P> z	Ci.lb	Ci.ub	
	National income leve							
RSV	High	.2724197	0.4951057	0.55	0.583	7095057	1.254345	
	Upper-middle	3211209	0.4932868	-0.65	0.517	-1.299439	0.657197	
	Lower-middle	1973232	1.780681	-0.11	0.912	-3.728884	-3.728884	
	Low	1.765932	2.549974	0.69	0.489	-3.231925	6.763789	
	WHO-region							
	Africa	.8373415	1.45493	0.58	0.565	-2.014268	3.688951	
	Americas	.3209991	1.064084	0.30	0.763	-1.764567	2.406565	
	Eastern	-1.275594	1.838391	-0.69	0.488	-4.878774	2.327586	
	Mediterranean	0010557	5220502	1.07	0.064	2.021097	0571750	
	Europe	9819557	.5239502	-1.87	0.064	-2.021087	.0571759 6.763789	
	South-East Asia	1.765932	2.549974	0.69	0.489	-3.231925	1.701144	
	Western Pacifc	0.7205144	0.4944524	1.46	0.148	2601152	1./01144	
	Study Design				l .			
	cohort	8153311	.4878452	-1.67	0.095	-1.77149	.140828	
	case-control	.8004486	1.203324	0.67	0.506	-1.558023	3.158921	
	cross-sectional	0459937	.5319277	-0.09	0.931	-1.088553	.9965654	
	case-crossover	3570562	1.784618	-0.20	0.841	-3.854844	3.140732	
	time-series	1.268819	.6338912	2.00	0.045	.0264152	2.511223	
	Age	1.188387	0.7958024	1.49	0.138	3898992	2.766673	
V	National income level		0.7730024	1.47	0.130	3070772	2.700073	
	High	5391765	.4383816	-1.23	0.219	-1.398389	.3200357	
	Low	.3938516	1.213169	0.32	0.745	-1.983917	2.77162	
	Upper-middle	.4604176	.426858	1.08	0.743	3762086	1.297044	
	WHO-region		20050	1.00	J.201	.5702000	1.27/077	
	Africa	.8025995	.9249863	0.87	0.386	-1.01034	2.615539	
	Americas	4676699	.5736834	-0.82	0.415	-1.592069	.6567289	
	Europe	0751581	.0119288	-6.30	0.000	0985382	051778	
	South-East Asia	.3938516	1.213169	0.32	0.745	-1.983917	2.77162	
	Western Pacifc	.3576141	.4337681	0.82	0.410	4925558	1.207784	
	Western Pacific   .35/6141   .433/681   0.82   0.410  4925558   1.207/84   Study Design							
	cohort	078567	.0118534	-6.63	0.000	1017992	0553349	
	cross-sectional	4525397	.3953467	-1.14	0.252	-1.227405	.3223255	
	case-crossover	2.617493	1.729004	1.51	0.130	7712925	6.006278	
	time-series	5498703	.389102	-1.41	0.158	-1.312496	.2127556	
	Age	.5640363	.3840765	1.47	0.142	1887397	1.316812	
PIV	National income level		.5040705	1.77	0.142	1007377	1.510012	
	High	-1.278616	.5058212	-2.53	0.011	-2.270007	2872246	
	Low	.9196676	.9207501	1.00	0.318	8849693	2.724305	
	Upper-middle	.8291939			0.092		1.793925	
	Upper-middle         .8291939         .4922186         1.68         0.092        1355368         1.793925           WHO-region							
	Africa	7801076	.9954686	-0.78	0.433	-2.73119	1.170975	
	Americas	-2.247165	1.454936	-1.54	0.122	-5.098788	.6044577	
	Europe	-1.340657	.9798861	-1.37	0.171	-3.261198	.5798848	
	South-East Asia	.9196676	.9207501	1.00	0.318	8849693	2.724305	
	Western Pacifc	.6947843	.565879	1.23	0.220	4143182	1.803887	
	Study Design	.0747043	.505017	1.23	0.220	7173102	1.003007	
	case-control	1.068844	.8694856	1.23	0.219	6353161	2.773005	
	cohort	.7647135	.6160566	1.23	0.219	4427353	1.972162	
	case-crossover	-1.124318	1.514906	-0.74	0.214	-4.093479	1.844843	
	cross-sectional	1.503414	.4132066	3.64	0.438	.6935438	2.313284	
	time-series	.8719242	.5420643	1.61	0.108	1905024	1.934351	
	ecological	9416197	1.381015	-0.68	0.108	-3.648359	1.76512	
	Age	2108411	.8557342	-0.08	0.493	-1.888049	1.466367	
HMPV	National income level		.0331342	0.23	0.003	1.000047	1.100307	
	High	.4654326	.4281547	1.09	0.277	3737352	1.3046	
	Low	.1921004	1.005379	0.19	0.277	-1.778406	2.162607	
	Upper-middle	5276054	.4393033	-1.20	0.230	-1.388624	.3334133	
	WHO-region	52/0034	.TJ/JUJJ	1.20	0.230	-1.300024	دد۱۳ددد.	
	Europe	1917908	.4546588	-0.42	0.673	-1.082906	.699324	
	Western Pacifc	.1413122	.4406817	0.32	0.748	7224082	1.005033	
	South-East Asia	.1921004	1.005379	0.32	0.748	-1.778406	2.162607	
		.1/41004	1.003379	0.17	0.040	-1.//0400	2.102007	
	Study Design	0778105	2250525	2.01	0.004	2105517	1 636060	
	cohort	.9778105	.3358525	2.91	0.004	.3195517	1.636069	
	case-crossover	339844	.9568592	-0.36	0.722	-2.215254	1.535565	
	cross-sectional	6637794 681461	.4392632	-1.51	0.131	-1.524719	.1971607	
	41	6 X I /I 6 I	.537519	-1.27	0.205	-1.734979	.3720568	
	time-series			0.07	0.001			
	Age	.0597434	.0614445	0.97	0.331	0606855	.1801724	
HRV	Age National income			0.97 -1.75	0.331	0606855 -3.242848	.1801724	
HRV	Age	.0597434	.0614445					

_	_						
	(Europe vs Western						
	Pacifc )						
	Study Design					1	1
	case-control	.6886622	1.646415	0.42	0.676	-2.538252	3.915577
	case-crossover	941739	1.122595	-0.84	0.402	-3.141984	1.258506
	cross-sectional	.910875	.9192012	0.99	0.322	8907263	2.712476
	cohort	.5377858	1.383341	0.39	0.697	-2.173513	3.249084
	time-series	-2.493636	1.643218	-1.52	0.129	-5.714284	.7270108
TT A JV/a	Age	-1.67809	.9866331	-1.70	0.089	-3.611856	.2556748
HAdVs	National income level (High vs Upper-middle)	1.844764	.5307367	3.48	0.001	.8045392	2.884989
	WHO-region				ı	- L	
	Africa	4373683	1.014488	-0.43	0.666	-2.425729	1.550992
	Americas	-2.15672	1.392943	-1.55	0.122	-4.886838	.5733973
	Europe	-1.442254	.7782993	-1.85	0.064	-2.967693	.0831848
	Western Pacifc	1.647847	.5619552	2.93	0.003	.5464347	2.749259
	Study Design		•			•	
	cohort	-1.211562	.7724662	-1.57	0.117	-2.725568	.3024437
	case-crossover	9328839	1.473294	-0.63	0.527	-3.820487	1.954719
	cross-sectional	1.465095	.5827668	2.51	0.012	.3228926	2.607296
	time-series	4536482	.7224517	-0.63	0.530	-1.869628	.9623311
	Age	6745164	.7700138	-0.88	0.381	-2.183716	.8346829
HBoV	National income	.5216799	.7176844	0.73	0.467	8849556	1.928315
	level (High vs Upper-middle)						
	WHO-region (Europe vs Western Pacifc)	1.16555	.7904104	1.47	0.140	3836257	2.714726
	Study Design	i	1	1		1	
	cohort	6837422	.1137644	-6.01	0.000	9067163	4607682
	case-crossover	.5538944	1.034743	0.54	0.592	-1.474165	2.581954
	cross-sectional	1.15892	.5094783	2.27	0.023	.160361	2.157479
	time-series	4482961	.8900741	-0.50	0.614	-2.192809	1.296217
	Age	.5538944	1.034743	0.54	0.592	-1.474165	2.581954
EV	National income	.9369969	.2057731	4.55	0.000	.5336889	1.340305
2,	level (High vs Upper-middle)						
	WHO-region (Europe vs Western Pacifc)	1.132873	.2167625	5.23	0.000	.7080263	1.557719
	Study Design						
	cohort	.9369969	.2057731	4.55	0.000	.5336889	1.340305
	case-crossover	1.101775	.9500641	1.16	0.246	7603166	2.963866
	time-series	-1.132873	.2167625	-5.23	0.000	-1.557719	7080263
	Age	1.101775	.9500641	1.16	0.246	7603166	2.963866
SARS-CoV- 2	National income level (High vs Upper-middle)	0344442	.0104926	-3.28	0.001	0550093	0138792
ì	WHO-region	I		1	1	1	
	Americas	.0116355	.0093826	1.24	0.215	0067541	.0300251
	Europe	2472766	.1011741	-2.44	0.015	4455742	048979
	Western Pacifc	0093421	.009325	-1.00	0.316	0276188	.0089345
					1.0.20		
	Study Design	005555	00=====	1	0.00-	1 04:127:	020710
	cohort	.0255756	.0072157	3.54	0.000	.0114331	.039718
	case-crossover	275535	.0114296	-24.11	0.000	2979366	2531335
TTC T	time-series	.9746894	.3969559	2.46	0.014	.1966701	1.752709
HCoV	National income level (High vs Upper-middle)	2.015627	1.031223	1.95	0.051	0055333	4.036786
	WHO-region (Europe vs Western Pacifc)	2.057392	.9107748	2.26	0.024	.2723064	3.842478
	Study Design						
	cohort	2.125852	1.004709	2.12	0.034	.1566589	4.095045
	case-crossover	-2.200029	1.600112	-1.37	0.169	-5.33619	.9361322
	time-series	-1.442606	1.580981	-0.91	0.362	-4.541272	1.656059
	Age	-2.200029	1.600112	-1.37	0.169	-5.33619	.9361322
SARS-CoV	National income level (High vs Upper-middle)	3785387	.7851233	-0.48	0.630	-1.917352	1.160275
	WHO-region (Europe vs Western Pacifc)	-1.766721	.613975	-2.88	0.004	-2.97009	5633517
	Study Design (cross-sectional vs	2922677	.9553481	-0.31	0.760	-2.164716	1.58018
	time-series analysis)					1	

S.	National income	-1.34663	.6411379	-2.10	0.036	-2.603237	0900225
pneumoniae	level (High vs		10.1220.7				
•	Upper-middle)						
	WHO-region		•			•	•
	Americas	7572833	1.182572	-0.64	0.522	-3.075082	1.560515
	Europe	531455	.1121689	-4.74	0.000	751302	311608
	Western Pacifc	.564659	.1103552	5.12	0.000	.3483667	.7809513
	Study Design	.564659	.1103552	5.12	0.000	.3483667	.7809513
	(cohort vs time-						
	series analysis)						
М.	National income	1.390195	2.402839	0.58	0.563	-3.319282	6.099673
pneumoniae	level (High vs						
	Upper-middle )						
	WHO-region	1.770413	3.320054	0.53	0.594	-4.736772	8.277599
	(Europe vs Western						
	Pacifc )						
	Study Design						
	cohort	-1.225525	1.43057	-0.86	0.392	-4.02939	1.57834
	cross-sectional	1.900281	1.456086	1.31	0.192	9535947	4.754157
	time-series	-1.390195	2.402839	-0.58	0.563	-6.099673	3.319282
	Age	-2.350963	2.385202	-0.99	0.324	-7.025874	2.323947
C.	National income	5.287783	.2139942	24.71	0.000	4.868363	5.707204
pneumoniae	level (High vs						
	Upper-middle )						
	WHO-region	5.287783	.2139942	24.71	0.000	4.868363	5.707204
	(Europe vs Western						
	Pacifc )						
	Study Design						
	cohort	5.296347	.2115122	25.04	0.000	4.881791	5.710903
	cross-sectional	-2.004481	3.697434	-0.54	0.588	-9.251319	5.242357
	time-series	-5.287783	.2139942	-24.71	0.000	-5.707204	-4.868363

Crl = credible interval. RSV: Respiratory syncytial virus; IV: influenza virus; HAdV: human adenovirus; HPIV: human parainfluenza virus; HMPV: human metapneumovirus; HBoV: human bocavirus; HRV: human rhinovirus; EV: enterovirus; HCoV: human coronavirus; SARS-CoV: Severe Acute Respiratory Syndrome Coronavirus; S. pneumoniae: Streptococcus pneumoniae; M. pneumoniae: Mycoplasma pneumoniae: C. pneumoniae: Chlamydophila pneumoniae

Table S14: Summary of the assessment of the quality and strength of the evidence on ambient temperature as a risk factor for incidence of viral, mycoplasma and chlamydia respiratory infections

	mbient erature	Respiratory syncytial virus (RSV) ( $n = 62$ )
•	ating	Basis
i. Downgrade		
Risk of bias across studies	-1	Most studies carry a substantial risk of bias.
Indirectness	0	The studies assessed population, exposure, and outcome of interest morbidity was appropriate outcome, studies conducted in the population of interest, mostly direct measures of exposure.
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity.
Imprecision	0	We judged that the incidence CI of meta-analyses was narrow enough.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
iii. Summary of the quali		
Overall quality of evidence (initial rating is "moderate")	e Moderate	Moderate+ $(1)$ + $(-1)$ = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings	n/a	Overall moderate quality of the evidence of the reduction in the incidence of RSV infection associated with ambient temperature exposure (see results from meta-analyses in Figure S2).
Strength considerations		
Quality of body of evidence	n/a	Moderate
Direction of effect estimate	e n/a	Direction largely as expected: increased temperature reduced incidence.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, especially studies with large sample sizes, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant
Other compelling attribute of the data that may influence certainty	s n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an association between temperature exposure and RSV incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded explanations for this association. The available evidence include the results of one or more well-designed, well-conducted studies and we believe that our conclusions are unlikely to strongly influence the outcomes of future studies.
	mbient erature	Influenza viruses (IV) (n = 27)
	ating	Basis
i. Downgrade		

studies

Indirectness	0	The studies assessed population, exposure, and outcome of
Inconsistency	0	interest The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis did not appear to be strongly influenced by the individual studies. measures are consistent in the direction of overall effect estimates.
Imprecision	0	We judged that the incidence CI of meta-analyses was narrow enough.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Moderate	$\label{eq:moderate} Moderate+(1)+(-1) = Moderate.\ Downgrading/upgrading\ resulted$ in moderate rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that most of the reduction in the incidence of IFV infection associated with ambient temperature exposure (see results from meta-analyses in Figure S3).
Strength considerations		
Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	Direction largely as expected: increased temperature reduced incidence.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, especially studies with large sample sizes, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an association between temperature exposure and IFV incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of one or more well-designed, well-conducted studies, and we believe that our conclusions are unlikely to strongly influence the outcomes of future studies.

		influence the outcomes of future studies.
Reference	Ambient	Human parainfluenza virus (HPIV) (n = 23)
	temperature	
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across studies	-1	There is a risk of bias in some studies.
Indirectness	0	The studies assessed population, exposure, and outcome of interest
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of

		HPIV infection do not appear to be strongly influenced by an individual study. measures are consistent in the direction of overall effect estimates.
Imprecision	0	We judged that the incidence CI of meta-analyses was narrow enough.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
quality assessment		
Overall quality of evidence (initial rating is "moderate")	Moderate	Moderate+(1)+(-1) = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that most of the reduction in the incidence of HPIV infection associated with ambient temperature exposure (see results from meta-analyses in Figure S4).
Strength		results from meta-analyses in Figure 34).
considerations		
Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	The direction of effect estimates largely showed an increasing trend in HPIV infection with increasing high temperatures.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, especially studies with large sample sizes, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an positive association between temperature exposure and HPIV incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of one or more well-designed, well-conducted studies, and we believe that our conclusions are unlikely to strongly influence the outcomes of future studies.

		influence the outcomes of future studies.
Reference	Ambient temperature	Human metapneumovirus (hMPV) (n = 14)
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across studies	-1	Some studies carry a certain risk of bias.
Indirectness	0	The studies assessed population, exposure, and outcome of interest
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study.
Imprecision	-1	Except six studies, the 95% CIs of the studies possessed notably narrow confidence intervals.

Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Low	Moderate+ $(-1)$ + $(-1)$ + $(1)$ = Low. Downgrading/upgrading resulted in low rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that most of the reduction in the incidence of hMPV infection associated with ambient temperature exposure (see results from meta-analyses in Figure S5).
Strength		
considerations Quality of body of evidence	n/a	Low
Direction of effect estimate	n/a	The direction of effect estimates largely showed an decreasing trend in hMPV infection with increasing temperatures.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, especially studies with large sample sizes, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an negative association between temperature exposure and hMPV incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of one or more well-designed, well-conducted studies, and we believe that our conclusions are unlikely to strongly influence the outcomes of future studies.

		ruture studies.
Reference	Ambient temperature	Human rhinovirus (HRV) (n = 12)
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across studies	s -1	There is a substantial risk of bias across most studies.
Indirectness	0	The studies assessed population, exposure, and outcome of interest.
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study. The results of all four metaanalyses for change in the respiratory infection. measures are consistent in the direction of overall effect estimates.
Imprecision	-1	We judged that the incidence CI of 3 studies was wide.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		

Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	0	Random effects contain uniformity and are not statistically significant.
Confounding	0	No evidence found to suggest that possible residual confounders
minimizes effect		would reduce effect estimates.
iii. Summary of the		
quality assessment		
Overall quality of	Very low	Moderate+ $(-1)$ + $(-1)$ = Very low. Downgrading/upgrading resulted
evidence (initial		in very low rating for the quality of evidence.
rating is "moderate")		
Summary of findings	n/a	The random effect result contains unity and were statistically
from meta-analysis		insignificant. (see results from meta-analyses in Figure S6).
Summary of	n/a	Studies not included in the meta-analyses presented mixed results,
qualitative findings		mostly insignificant associations.
Strength		
considerations		
Quality of body of	n/a	Very low
evidence		
Direction of effect	n/a	No association was found and the results were inconclusive.
estimate		
Confidence in effect	n/a	Ambient temperature studies directly measure outcomes of
estimate		interest, influence directions are often inconsistent, and most
		studies have a risk of bias, and some have large confidence
		intervals, which makes the results uncertain for the future.
Other compelling	n/a	Differences in study design, exposure measures, statistical
attributes of the data		methods, consideration of lagged effects, and background factors,
that may influence		including completeness of incidence statistics, population
certainty		exposure levels and vulnerability, and differences in physical and
		physiological adaptations of study populations, make
0 11	F	interpretation and comparison difficult.
Overall strength of	Limited	Based on our analysis and interpretation of the evidence, we
evidence		conclude that there is no association between temperature
		exposure and HRV morbidity.

		exposure and HRV morbidity.
Reference	Ambient temperature	Human adenovirus (HAdV) (n = 16)
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across studies	-1	There is a risk of bias in some studies.
Indirectness	0	The studies assessed population, exposure, and outcome of interest.
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study. The results of all four metaanalyses for change in the respiratory infection. measures are consistent in the direction of overall effect estimates.
Imprecision	-1	We judged that the incidence CI of 4 studies was wide.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	0	Random effects contain uniformity and are not statistically significant.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Very low	Moderate+ $(-1)$ + $(-1)$ = Very low. Downgrading/upgrading resulted in very low rating for the quality of evidence.

Summary of findings from meta-analysis Strength	n/a	Random effect contains unity and were statistically insignificant (see results from meta-analyses in Figure S7).
considerations	,	
Quality of body of evidence	n/a	Very low
Direction of effect estimate	n/a	No association was found and the results were inconclusive.
Confidence in effect	n/a	Ambient temperature studies directly measure outcomes of
estimate		interest, influence directions are often inconsistent, and some studies have a risk of bias and some have large confidence intervals.
Other compelling attributes of the data	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors,
that may influence		including completeness of incidence statistics, population
certainty		exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make
		interpretation and comparison difficult.
Overall strength of evidence	Limited	Based on our analysis and interpretation of the evidence, we conclude that there is no association between temperature exposure and incidence, and that future studies may change our
		results.

		results.
Reference	Ambient temperature	Human bocavirus (HBoV) (n = 9)
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across	0	There is no indication that there is substantial risk of bias across
studies		the body of available evidence.
Indirectness	0	The studies assessed population, exposure, and outcome of interest
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study. The results of all four metaanalyses for change in the respiratory infection. measures are consistent in the direction of overall effect estimates.
Imprecision	-1	The number of studies is small, the sample size is insufficient, and the incidence CI of 5 studies was wide.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
quality assessment		
Overall quality of evidence (initial rating is "moderate")	Moderate	Moderate+ $(1)$ + $(-1)$ = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that most of the increase in the incidence of HBoV infection associated with ambient temperature exposure (see results from meta-analyses in Figure S8).
Strength considerations		,
Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	Direction largely as expected: higher risk of incidence at high ambient temperatures.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most 71

Other compelling attributes of the data that may influence certainty	n/a	studies have a low risk of bias, especially studies with large sample sizes, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant. Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Limited	Based on our analysis and interpretation of the evidence, we conclude that there is an positive association between temperature exposure and HBoV incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of one or more well-designed, well-conducted studies, and we believe that our conclusions are unlikely to strongly influence the outcomes of future studies.

		future studies.
Reference	Ambient temperature	Enterovirus (EV) $(n = 4)$
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across	0	There is no indication that there is substantial risk of bias across
studies		the body of available evidence.
Indirectness	0	The studies assessed population, exposure, and outcome of
		interest
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of EVs infection do not appear to be strongly influenced by an individual study. The results of all four meta-analyses for change in the respiratory infection. measures are consistent in the direction of overall effect estimates.
Imprecision	-1	The number of studies is small and the sample size is insufficient.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	0	No dose-response relationship was found.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Low	Moderate+(-1)= Low. Downgrading/upgrading resulted in low rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that most of the reduction in the incidence of respiratory infection associated with ambient temperature exposure (see results from meta-analyses in Figure S9).
Strength		
<b>considerations</b> Quality of body of	n/a	Low
evidence	,	701 111 12 11 1 1 1 1 1 1 1 1 1 1 1 1 1 1
Direction of effect estimate	n/a	The overall low quality evidence showed no association between temperature exposure and EVs morbidity.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure results of interest, influence directions are inconsistent, the number of studies is small, and one study has a large confidence interval.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and

Overall strength of evidence	Sufficient	physiological adaptations of study populations, make interpretation and comparison difficult.  Based on our analysis and interpretation of the evidence, we conclude that there is no association between temperature exposure and EVs incidence, and that the addition of future studies may change the current results.
Reference	Ambient temperature	Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) (n =8)
Quality factor	Rating	Basis
i. Downgrade Risk of bias across studies	0	There is no indication that there is substantial risk of bias across the body of available evidence.
Indirectness	0	The studies assessed population, exposure, and outcome of interest.
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study. measures are consistent in the direction of overall effect estimates.
Imprecision	-1	We judged that the incidence CI of one study was wide and the remaining CI was narrow enough.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
<ul><li>ii. Upgrade</li><li>Large magnitude of effect</li></ul>	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Moderate	Moderate+ $(1)$ + $(-1)$ = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings from meta-analysis Strength	n/a	We found that most of the reduction in the incidence of respiratory infection associated with ambient temperature exposure (see results from meta-analyses in Figure S10).
considerations Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	The direction of effect estimates largely showed an decreasing trend in COVID-19 infection with increasing high temperatures.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are usually consistent, and studies have a low risk of bias, especially studies with large sample sizes, but some studies have large confidence intervals. A new study is unlikely to have an effect estimate that would render the results of the meta-analysis invalid or not significant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an negative association between temperature exposure and COVID-19 incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of one or more well-designed, well-conducted studies, and we believe that our

conclusions are unlikely to strongly influence the outcomes of future studies.

		future studies.
Reference	Ambient	Severe acute respiratory syndrome coronavirus (SARS-CoV)
	temperature	(n=4)
Quality factor	Rating	Basis
i. Downgrade Risk of bias across studies	-1	There is a risk of bias in some studies.
Indirectness	0	The studies assessed population, exposure, and outcome of interest.
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study. The results of all four meta-analyses for change in the respiratory infection. measures are consistent in the direction of overall effect estimates.
Imprecision	-1	The number of studies is small and the sample size is insufficient.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Low	Moderate+ $(1)$ + $(-1)$ + $(-1)$ = Low. Downgrading/upgrading resulted in low rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	Low quality evidence did not show that the incidence of SARS infection was related to ambient temperature exposure (see results from meta-analyses in Figure S11).
Strength		
<b>considerations</b> Quality of body of	n/a	Low
evidence	11/ 41	20 //
Direction of effect estimate	n/a	The direction of effect estimates did not show an association between temperature and SARS infection, and the results were inconclusive.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Limited	Based on our analysis and interpretation of the evidence, we did not find an association between temperature exposure and SARS morbidity, and the available evidence is insufficient.
Reference	Ambient temperature	Human coronavirus (HCoV) (n=5)
Quality factor	Rating	Basis
i. Downgrade Risk of bias across studies	0	There is no indication that there is substantial risk of bias across the body of available evidence.

Indirectness	0	The studies assessed population, exposure, and outcome of
Inconsistency	0	interest.  The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of HCoV infection do not appear to be strongly influenced by an individual study. Measures are consistent in the direction of overall effect estimates.
Imprecision Publication bias	-1 0	The number of studies is small and the sample size is insufficient. We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Moderate	Moderate+(1)+(-1) = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that most of the reduction in the incidence of HCoV infection associated with ambient temperature exposure (see results from meta-analyses in Figure S12).
Strength		•
considerations Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	Direction largely as expected: increased temperature reduced incidence.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an negative association between temperature exposure and HCoV incidence, and we are reasonably confident that opportunity, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of one or more well-designed, well-conducted studies, and we believe that our conclusions are unlikely to strongly influence the outcomes of future studies.  MERS-CoV (n=1)

Reference	Ambient temperature	MERS-CoV (n =1)
0 114 6		
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across studies	0	No risk of bias was found.
Indirectness	0	The studies assessed population, exposure, and outcome of interest.
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods,

		lag structure considered, model) and not be driven by unexpected heterogeneity.
Imprecision Publication bias	-1 0	The number of studies is small and the sample size is insufficient. We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	1	As ambient temperature increases, morbidity increases in doseresponse patterns.
Confounding minimizes effect iii. Summary of the quality assessment	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Moderate	Moderate+(1)+(-1) = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that the increased incidence of respiratory infections was associated with ambient temperature exposure (see results from meta-analyses in Figure S9).
Strength considerations		nom mem analyses in Figure 57).
	/	M. 1
Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	The incidence of respiratory pathogens increased with the increase of temperature.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are usually consistent, studies have low risk of bias, small sample sizes, and large confidence intervals. The emergence of a new study may invalidate or make significant the results of the meta-analysis.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Sufficient	Based on our analysis and interpretation of the evidence, we conclude that there is an association between temperature exposure and incidence, and we have reason to believe that chance, bias, and confounding factors can be excluded as explanations for this association. The available evidence includes the results of only one well-designed study, and our conclusions are unlikely to strongly influence the results of future studies.

		are unlikely to strongly influence the results of future studies.
Reference	Ambient	Human rhinovirus/enterovirus (HRV/EV) ( $n = 2$ )
	temperature	
Quality factor	Rating	Basis
i. Downgrade		
Risk of bias across	0	There is no indication that there is substantial risk of bias across
studies		the body of available evidence.
Indirectness	0	The studies assessed population, exposure, and outcome of interest
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of HRVs/EVs infection do not appear to be strongly influenced by an individual study.
Imprecision	-1	The number of studies is small and the sample size is insufficient.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.

Dose response	1	Most studies report dose-response patterns that are broadly similar, with incidence increasing or decreasing with ambient temperature above a certain threshold.
Confounding minimizes effect iii. Summary of the	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
quality assessment		
Overall quality of evidence (initial rating is "moderate")	Moderate	Moderate+(1)+(-1) = Moderate. Downgrading/upgrading resulted in moderate rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	We found that the increase in the incidence of HRVs/EVs infection associated with ambient temperature rise (see results from meta-analyses in Figure S13).
Strength		
considerations		
Quality of body of evidence	n/a	Moderate
Direction of effect estimate	n/a	Direction largely as expected: higher risk of incidence at high ambient temperatures.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are generally consistent, and most studies have a low risk of bias, especially studies with large sample sizes, but several have large confidence intervals. It is unlikely that a new study would have an effect estimate that would make the results of the meta-analysis null or insignificant.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Limited	Based on our analysis and interpretation of the evidence, we conclude that there is a positive association between temperature exposure and the incidence of HRV/EVs, but this contains only a small amount of available evidence.

n/a, not applicable. "0" quality rating indicates there were no upgrades or downgrades for each factor being evaluated across the body of evidence; -1=downgrade; +1=upgrade.

Table S15: Summary of the assessment of the quality and strength of the evidence on ambient temperature as a risk factor for incidence of bacterial respiratory infections

Reference	Ambient temperature					
	Streptococcus pneumoniae	Streptococcus pyogenes pharyngiti	Moraxella catarrhalis	Pseudomonas aeruginosa	Legionella pneumoniae	Haemophilus influenzae
Initial rating Quality factor i. Downgrade	Moderate	Moderate	Moderate	Moderate	Moderate	Moderate
Risk of bias across studies	-1	-1	0	-1	0	-1
Indirectness	0	0	0	0	0	0
Inconsistency	0	n/a	-1	n/a	-1	n/a
Imprecision	-1	-1	-1	-1	-1	0
Publication bias	0	0	0	0	0	0
ii. Upgrade						
Large magnitude of effect	0					
Dose response	0	+1	0	+1	+1	0
Confounding minimizes effect	0	0	0	0	0	0

iii. Summary of the quality assessment						
Overall quality of evidence (initial rating	Very Low	Low	Very Low	Moderate	Low	Low
is "moderate") Summary of findings from meta-analysis	n/a	n/a	n/a	n/a	n/a	n/a
Summary of qualitative findings	n/a	n/a	n/a	n/a	n/a	n/a
Strength considerations						
Quality of body of evidence	n/a	n/a	n/a	n/a	n/a	n/a
Direction of effect estimate	n/a	n/a	n/a	n/a	n/a	n/a
Confidence in effect estimate	n/a	n/a	n/a	n/a	n/a	n/a
Other compelling attributes of the data that may influence certainty	n/a	n/a	n/a	n/a	n/a	n/a
Overall strength of evidence	Limited	Limited	Limited	Limited	Limited	Limited

n/a, not applicable. "0" quality rating indicates there were no upgrades or downgrades for each factor being evaluated across the body of evidence; -1=downgrade; +1=upgrade.

Table S16: Summary of the assessment of the quality and strength of the evidence on ambient temperature as a risk factor for incidence of mycoplasma and chlamydia respiratory infections

Reference	Ambient temperature	Mycoplasma pneumoniae (n =19)
Quality factor	Rating	Basis
i. Downgrade	ruung	Dubio
Risk of bias across studies	-1	There is a substantial risk of bias across most studies.
Indirectness	0	The studies assessed population, exposure, and outcome of interest
Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study. Measures are consistent in the direction of overall effect estimates.
Imprecision	-1	We judged that the incidence CI of one study was wide.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	0	No significant dose-response relationship was found.
Confounding	0	No evidence found to suggest that possible residual confounders
minimizes effect iii. Summary of the quality assessment		would reduce effect estimates.
Overall quality of evidence (initial rating is "moderate")	Very low	Moderate+(-1)+(-1)=Very low. Downgrading/upgrading resulted in very low rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	The quality of evidence was generally low, and the incidence of Mycoplasma pneumoniae infections was not associated with ambient temperature (see results from meta-analyses in Figure S17).
Summary of qualitative findings <b>Strength considerations</b>	n/a	Studies not included in the meta-analyses presented mixed results, mostly insignificant associations.
Quality of body of evidence	n/a	Low
Direction of effect estimate	n/a	The direction of effect estimates did not show an association between temperature and mycoplasma infection, and the results were inconclusive.
Confidence in effect estimate	n/a	Ambient temperature studies directly measure outcomes of interest, influence directions are inconsistent, and due to these methodological deficiencies, it cannot be ruled out that new studies may show different estimates of effects.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make interpretation and comparison difficult.
Overall strength of evidence	Limited	Based on our analysis and interpretation of the evidence, we conclude that there is no association between temperature exposure and morbidity.
Reference	Ambient temperature	Chlamydophila pneumoniae (n =4)
Quality factor	Rating	Basis
i. Downgrade	-1	There is a risk of bias in some studies.
Risk of bias across	-1	There is a risk of blus in some studies.

Inconsistency	0	The magnitude of effect estimates likely to differ because of differences in study methods (study design, statistical methods, lag structure considered, model) and not be driven by unexpected heterogeneity. The results of the meta-analysis for incidence of respiratory infection do not appear to be strongly influenced by an individual study.
Imprecision	-1	The number of studies is small and the sample size is insufficient.
Publication bias	0	We found no reason to suspect publication bias. The search was comprehensive, and the studies were generally consistent among their findings, regardless of size or funding source.
ii. Upgrade		
Large magnitude of effect	0	We did not consider the estimated effects large.
Dose response	0	Most studies have found no effect of temperature on chlamydophila pneumoniae morbidity.
Confounding minimizes effect iii. Summary of the	0	No evidence found to suggest that possible residual confounders would reduce effect estimates.
quality assessment	** *	W. L. (1) (4) (4) W. L. B. W. (1) W.
Overall quality of evidence (initial rating is "moderate")	Very low	Moderate+(1)+(-1)+(-1) = Very low. Downgrading/upgrading resulted in very low rating for the quality of evidence.
Summary of findings from meta-analysis	n/a	Very low-quality evidence suggests that the incidence of chlamydophila pneumoniae infections is not associated with ambient temperature exposure (see meta-analysis results in Figure S18).
Summary of qualitative findings Strength considerations	n/a	Studies not included in the meta-analyses presented mixed results, mostly insignificant associations.
Quality of body of evidence	n/a	Very low
Direction of effect estimate	n/a	Effect estimates showed no significant difference in temperature and respiratory infections, making the results inconclusive.
Confidence in effect estimate	n/a	No trend was observed in overall random effects, and the number of studies included in the meta-analysis was small, making the results uncertain for the future.
Other compelling attributes of the data that may influence certainty	n/a	Differences in study design, exposure measures, statistical methods, consideration of lagged effects, and background factors, including completeness of incidence statistics, population exposure levels and vulnerability, and differences in physical and physiological adaptations of study populations, make
Overall strength of evidence	Limited	interpretation and comparison difficult.  No association was observed, the quality of evidence was "very low", and a small number of studies led to "limited" evidence to draw conclusions about the association between respiratory infections and temperature.

n/a, not applicable. "0" quality rating indicates there were no upgrades or downgrades for each factor being evaluated across the body of evidence; -1=downgrade; +1=upgrade.

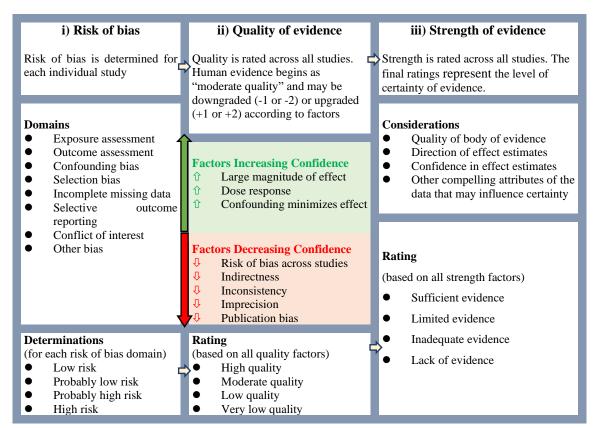


Figure X1 | Overview of Navigation Guide systematic review methodology used for rating the quality and strength of the human evidence.

Picture shows three parts of evidence evaluation, namely risk of bias, quality of evidence, and strength of evidence, respectively showing the subdomains and determinations/ratings of evaluation.

Figure S1. Random-effects meta-analysis of respiratory syncytial virus (RSV) estimates (62 studies)

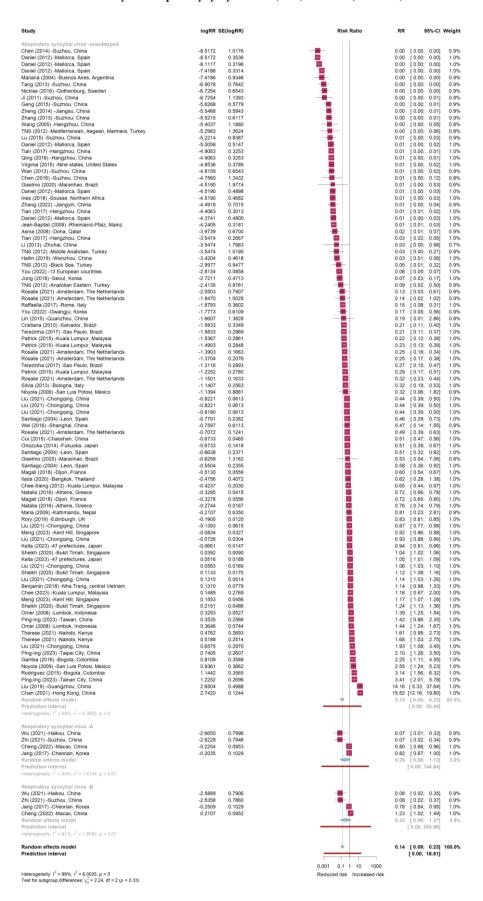


Figure S1-1. Subgroup analysis of respiratory syncytial virus (RSV) meta-analysis by Köppen-Geiger climate

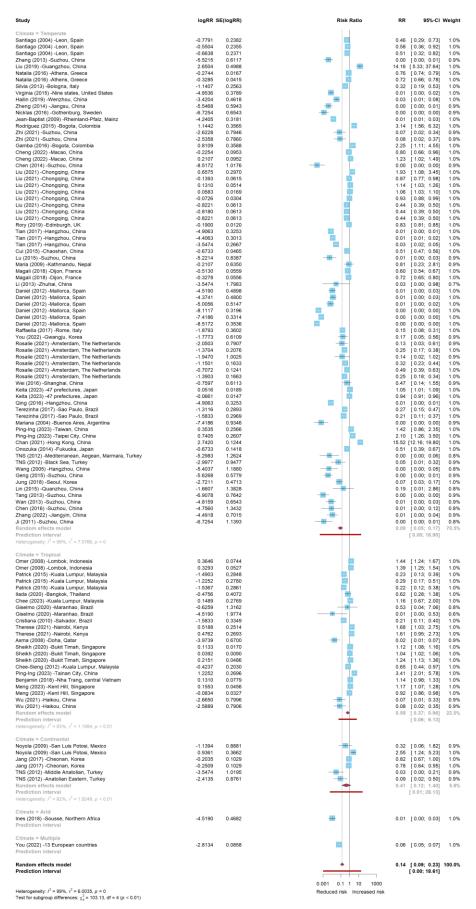


Figure S1-2. Subgroup analysis of respiratory syncytial virus (RSV) meta-analysis by World Bank income category

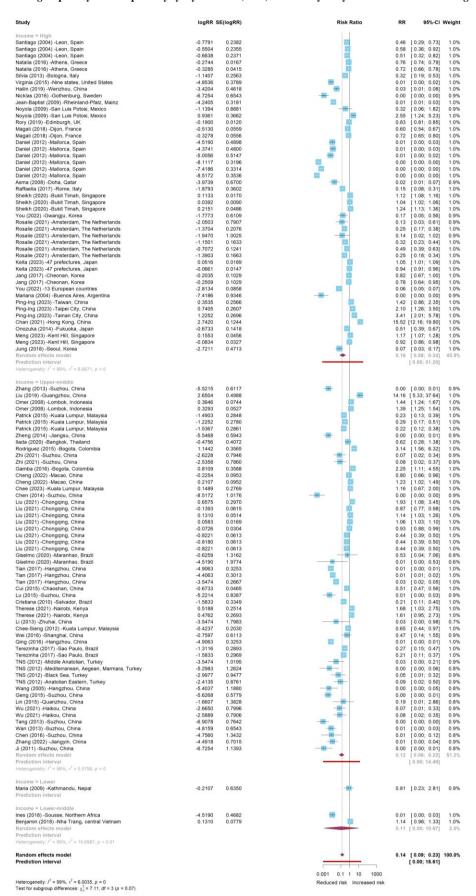
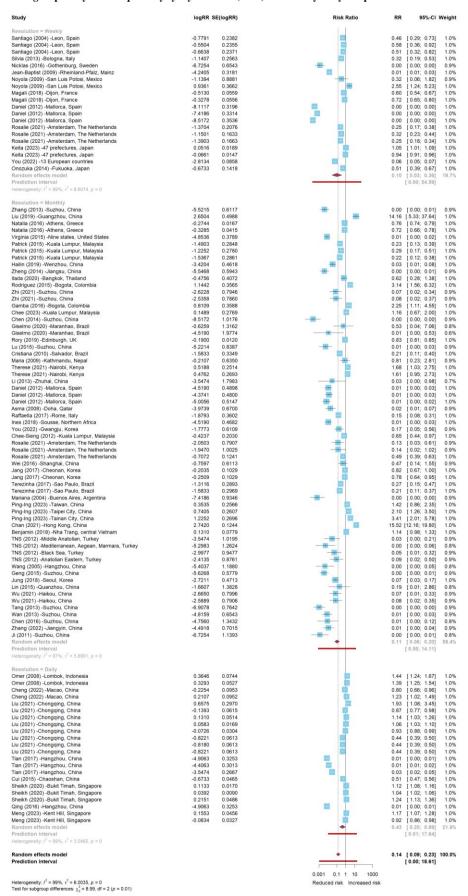


Figure S1-3. Subgroup analysis of respiratory syncytial virus (RSV) meta-analysis by temporal resolution



 $Figure \ S1-4. \ Subgroup \ analysis \ of \ respiratory \ syncytial \ virus \ (RSV) \ meta-analysis \ by \ exposure \ measure$ 

Study	logRR \$	SE(logRR)	Risk Ratio	RR	95%-CI	Weight
Temperature = Mean Santiago (2004) -Leon, Spain	-0.7791	0.2382		0.46	[ 0.29; 0.73]	1.0%
Zhang (2013) -Suzhou, China	-5.5215 2.6504	0.6117	+ 17 <u> </u>	0.00	[ 0.00; 0.01]	0.9%
Liu (2019) -Guangzhou, China Natalia (2016) -Athens, Greece	-0.2744	0.0167	<u> </u>	0.76	[ 0.74; 0.79]	1.0%
Natalia (2016) -Athens, Greece Omer (2008) -Lombok, Indonesia	-0.3285 0.3646	0.0415 0.0744	<b>1</b>	0.72 1.44	[ 0.66; 0.78] [ 1.24; 1.67]	1.0%
Omer (2008) -Lombok, Indonesia Patrick (2015) -Kuala Lumpur, Malaysia	0.3293 -1.4903	0.0527 0.2848	<b>∟</b> †	1.39 0.23	[ 1.25; 1.54] [ 0.13; 0.39]	1.0%
Patrick (2015) -Kuala Lumpur, Malaysia Patrick (2015) -Kuala Lumpur, Malaysia	-1.2252 -1.5367	0.2780		0.29	[0.17; 0.51]	1.0%
Hailin (2019) -Wenzhou, China	-3.4204 -5.5468	0.4618	_=		[ 0.01; 0.08]	1.0%
Zheng (2014) -Jiangsu, China Ilada (2020) -Bangkok, Thailand	-0.4756	0.4072	=	0.62	[ 0.00; 0.01] [ 0.28; 1.38]	1.0%
Nicklas (2016) -Gothenburg, Sweden Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-6.7254 -4.2405	0.6543 0.3181		0.00	[ 0.00; 0.00]	0.9% 1.0%
Noyola (2009) -San Luis Potosi, Mexico Rodriguez (2015) -Bogota, Colombia	-1.1394 1.1442	0.8881 0.3565	-	0.32 3.14	[ 0.06; 1.82]	0.9%
Zhi (2021) -Suzhou, China Zhi (2021) -Suzhou, China	-2.6228 -2.5358	0.7946 0.7860	-	0.07	[0.02; 0.34]	0.9%
Gamba (2016) -Bogota, Colombia	0.8109	0.3588		2.25	[ 1.11; 4.55]	1.0%
Cheng (2022) -Macao, China Cheng (2022) -Macao, China	-0.2254 0.2107	0.0953 0.0952	T		[ 0.66; 0.96] [ 1.02; 1.49]	1.0%
Chee (2023) -Kuala Lumpur, Malaysia Chen (2014) -Suzhou, China	0.1489 -8.5172	0.2769 1.0176	- <b>-</b>   T	1.16	[0.67; 2.00]	1.0%
Liu (2021) -Chongqing, China Rory (2019) -Edinburgh, UK	-0.8221 -0.1900	0.0613		0.44	[ 0.39; 0.50]	1.0%
Tian (2017) -Hangzhou, China Cui (2015) -Chaoshan, China	-4.4063 -0.6733	0.3013 0.0465		0.01	[ 0.01; 0.02]	1.0%
Lu (2015) -Suzhou, China	-5.2214	0.8387	<u> </u>	0.01	[ 0.00; 0.03]	0.9%
Cristiana (2010) -Salvador, Brazil Maria (2009) -Kathmandu, Nepal	-1.5833 -0.2107	0.3349 0.6350		0.21 0.81	[ 0.11; 0.40]	1.0% 0.9%
Magali (2018) -Dijon, France Magali (2018) -Dijon, France	-0.5130 -0.3278	0.0559 0.0556		0.60	[ 0.54; 0.67]	1.0%
Therese (2021) -Nairobi, Kenya Therese (2021) -Nairobi, Kenya	0.5188	0.2514		1.68	[1.03; 2.75]	1.0%
Li (2013) -Zhuhai, China	-3.5474	1.7983		0.03	[ 0.00; 0.98]	0.7%
Daniel (2012) -Mallorca, Spain Daniel (2012) -Mallorca, Spain	-4.5190 -8.1117	0.4898 0.3196	<u> </u>	0.01	[ 0.00; 0.03]	1.0%
Asma (2008) -Doha, Qatar Raffaella (2017) -Rome, Italy	-3.9739 -1.8793	0.6700 0.3602		0.02	[ 0.01; 0.07]	0.9%
Sheikh (2020) -Bukit Timah, Singapore Ines (2018) -Sousse, Northern Africa	0.1133 -4.5190	0.0170 0.4682	<u> </u>	1.12 0.01	[1.08; 1.16]	1.0%
You (2022) -Gwangju, Korea	-1.7773	0.6109	<del>-</del>	0.17	[ 0.05; 0.56]	0.9%
Chee-Sieng (2012) -Kuala Lumpur, Malaysia Rosalie (2021) -Amsterdam, The Netherlands	-0.4237 -2.0503	0.2030 0.7907	-	0.13	[ 0.44; 0.97]	1.0%
Rosalie (2021) -Amsterdam, The Netherlands Wei (2016) -Shanghai, China	-1.3704 -0.7597	0.2076 0.6113	-	0.25	[ 0.17; 0.38]	1.0%
Keita (2023) -47 prefectures, Japan Keita (2023) -47 prefectures, Japan	0.0516	0.0189 0.0147		1.05	[1.01; 1.09]	1.0%
Qing (2016) -Hangzhou, China Jang (2017) -Cheonan, Korea	-4.9063	0.3253	<b>=</b>   <b>I</b>	0.01	[ 0.00; 0.01]	1.0%
Jang (2017) -Cheonan, Korea	-0.2035 -0.2509	0.1029 0.1029	•	0.82 0.78	[ 0.67; 1.00] [ 0.64; 0.95]	1.0%
You (2022) -13 European countries Mariana (2004) -Buenos Aires, Argentina	-2.8134 -7.4186	0.0858 0.9346	- "	0.06	[ 0.05; 0.07]	1.0%
Ping-Ing (2023) -Taiwan, China Ping-Ing (2023) -Taipei City, China	0.3535	0.2566		1.42	[ 0.86; 2.35]	1.0%
Ping-Ing (2023) -Tainan City, China Chan (2021) -Hong Kong, China	1.2252	0.2696		3.41	[ 2.01; 5.78] [12.16; 19.80]	1.0%
Onozuka (2014) -Fukuoka, Japan	-0.6733	0.1418		0.51	[ 0.39; 0.67]	1.0%
Benjamin (2018) -Nha Trang, central Vietnam TNS (2012) -Middle Anatolian, Turkey	0.1310 -3.5474	0.0779 1.0195	<del></del>	0.03	[ 0.98; 1.33]	1.0% 0.9%
TNS (2012) -Mediterranean, Aegean, Marmara, Turkey TNS (2012) -Black Sea, Turkey	-5.2983 -2.9977	1.2624 0.9477	-	0.00	[0.00; 0.06]	0.8%
TNS (2012) -Anatolian Eastern, Turkey Wang (2005) -Hangzhou, China	-2.4135 -5.4037	0.8761 1.1880		0.09	[ 0.02; 0.50]	0.9%
Geng (2015) -Suzhou, China	-5.6268 -2.7211	0.5779	# _	0.00	[ 0.00; 0.01]	0.9%
Jung (2018) -Seoul, Korea Lin (2015) -Quanzhou, China	-1.6607	1.3828	<del></del>	0.19	[ 0.01; 2.86]	0.8%
Wu (2021) -Haikou, China Wu (2021) -Haikou, China	-2.6650 -2.5889	0.7996 0.7906	-	0.07	[ 0.01; 0.33]	0.9%
Tang (2013) -Suzhou, China Wan (2013) -Suzhou, China	-6.9078 -4.8159	0.7642 0.6543	***	0.00	[0.00; 0.00]	0.9%
Chen (2016) -Suzhou, China Zhang (2022) -Jiangyin, China	-4.7560 -4.4918	1.3432 0.7015	-	0.01	[ 0.00; 0.12]	0.8%
Ji (2011) -Suzhou, China Random effects model	-6.7254	1.1393			[ 0.00; 0.01]	0.8% 71.0%
Prediction interval					[ 0.00; 18.97]	71.0%
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 6.1799$ , $p = 0$						
Temperature = Min Santiago (2004) -Leon, Spain	-0.5504	0.2355		0.58	[ 0.36; 0.92]	1.0%
Silvia (2013) -Bologna, Italy Virginia (2015) -Nine states, United States	-1.1407 -4.8536	0.2563 0.3789	<u> </u>	0.32	[ 0.19; 0.53]	1.0%
Liu (2021) -Chongqing, China	0.6575	0.2970			[ 1.08; 3.45]	1.0%
Liu (2021) -Chongqing, China Liu (2021) -Chongqing, China	0.1310	0.0514	<u> </u>	1.14	[ 0.77; 0.98] [ 1.03; 1.26]	1.0%
Liu (2021) -Chongqing, China Liu (2021) -Chongqing, China	0.0583 -0.0726	0.0169 0.0304		1.06 0.93	[ 1.03; 1.10] [ 0.88; 0.99]	1.0%
Liu (2021) -Chongqing, China Giselmo (2020) -Maranhao, Brazil	-0.8180 -0.6259	0.0613 1.3162		0.44	[ 0.39; 0.50]	1.0%
Tian (2017) -Hangzhou, China Daniel (2012) -Mallorca, Spain	-4.9063 -4.3741	0.3253 0.4800	=	0.01	[0.00; 0.01]	1.0%
Daniel (2012) -Mallorca, Spain	-7.4186	0.3314	= -   _	0.00	[ 0.00; 0.00]	1.0%
Sheikh (2020) -Bukit Timah, Singapore Rosalie (2021) -Amsterdam, The Netherlands	0.0392 -1.9470	1.0025	<u>+</u> -T		[ 1.02; 1.06] [ 0.02; 1.02]	0.9%
Rosalie (2021) -Amsterdam, The Netherlands Terezinha (2017) -Sao Paulo, Brazil	-1.1501 -1.3116	0.1633	- T		[ 0.23; 0.44]	1.0%
Random effects model Prediction interval					[ 0.06; 0.57]	16.5%
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 5.2196$ , $\rho < 0.01$						
Temperature = Max Santiago (2004) -Leon, Spain	-0.6638	0.2371		0.61	[ 0.32; 0.82]	1.0%
Noyola (2009) -San Luis Potosi, Mexico	0.9361	0.3662	<del>_</del>	2.55	[ 1.24; 5.23]	1.0%
Liu (2021) -Chongqing, China Giselmo (2020) -Maranhao, Brazil	-0.8221 -4.5190	0.0613 1.9774			[ 0.39; 0.50] [ 0.00; 0.53]	1.0% 0.6%
Tian (2017) -Hangzhou, China Daniel (2012) -Mallorca, Spain	-3.5474 -5.0056	0.2667 0.5147		0.03	[ 0.02; 0.05]	1.0%
Daniel (2012) -Mallorca, Spain Sheikh (2020) -Bukit Timah, Singapore	-8.5172 0.2151	0.3536	= - 1	0.00	[ 0.00; 0.00]	1.0%
Rosalie (2021) -Amsterdam, The Netherlands	-0.7072	0.1241	<b>Ŀ</b> ₹	0.49	[ 0.39; 0.63]	1.0%
Rosalie (2021) -Amsterdam, The Netherlands Terezinha (2017) -Sao Paulo, Brazil	-1.3903 -1.5833	0.1663	<b>□</b>		[0.18; 0.34]	1.0%
Meng (2023) -Kent Hill, Singapore Meng (2023) -Kent Hill, Singapore	0.1553 -0.0834	0.0456 0.0327	#	0.92	[ 1.07; 1.28] [ 0.86; 0.98]	1.0%
Random effects model Prediction Interval				0.15	[ 0.04; 0.66]	12.5%
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 6.9935$ , $p < 0.01$						
Random effects model Prediction interval					[ 0.09; 0.23]	100.0%
, realizabil interval			0.001 0.1 1 10 1000		[ 0.00; 18.61]	
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 6.0035$ , $p = 0$			0.001 0.1 1 10 1000 Reduced risk Increased risk			
Test for subgroup differences: $\chi_2^2 = 0.37$ , df = 2 ( $\rho = 0.83$ )						

Figure S1-5. Subgroup analysis of respiratory syncytial virus (RSV) meta-analysis by modelling approach

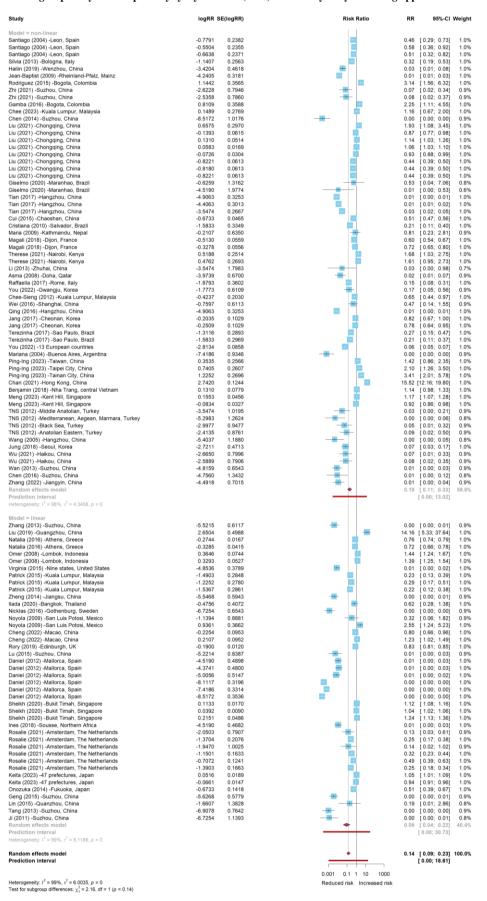


Figure S1-6. Subgroup analysis of respiratory syncytial virus (RSV) meta-analysis by lag type

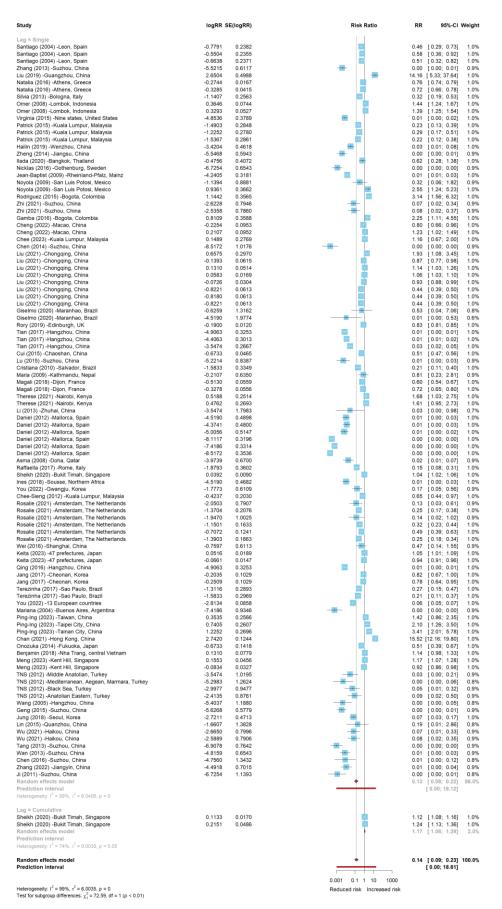


Figure S1-7. Leave-one-out analysis of respiratory syncytial virus (RSV) meta-analysis

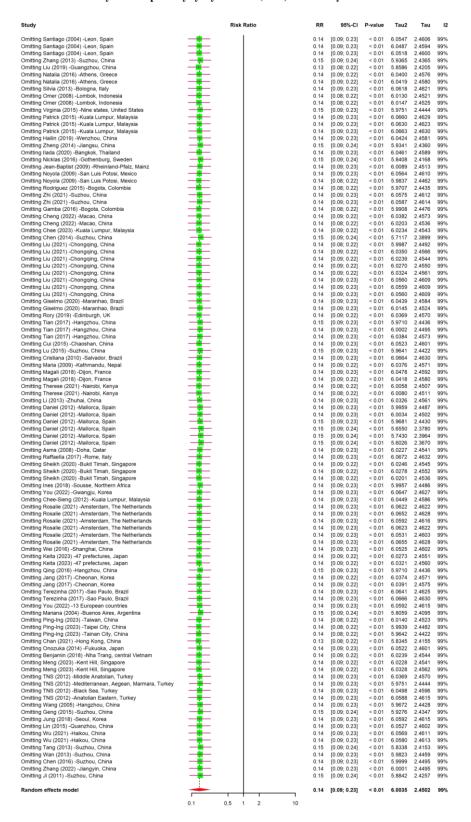
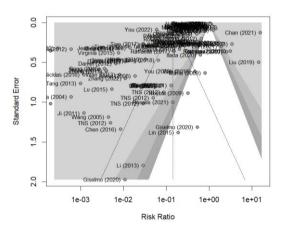


Figure S1-8. Funnel plot of respiratory syncytial virus (RSV) estimates



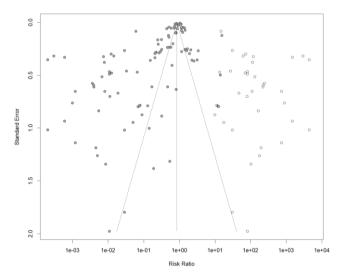
## Linear regression test of funnel plot asymmetry

Egger's test result: t = -6.36, df = 103, p-value < 0.0001

Sample estimates:

bias	se.bias	intercept	se.intercept
-5.1657	0.8258	0.0354	0.0397

Figure S1-9. Trim and fill of respiratory syncytial virus (RSV) meta-analysis  $\,$ 



Number of studies: k = 145 (with 40 added studies)

Adjusted estimates: RR= 0.8329 [0.4510; 1.5382], Z= -0.58, P= 0.5591

 $Quantifying\ heterogeneity:\ tau^2=\ 13.7945\ [11.1059;\ 17.9865];\ tau=3.7141\ [3.3326;\ 4.2411],$ 

 $I^2 = 98.9\%$  [98.8%; 99.0%]; H = 9.50 [9.21; 9.80]

Figure S2. Random-effects meta-analysis of influenza virus (IV) estimates (27 studies)

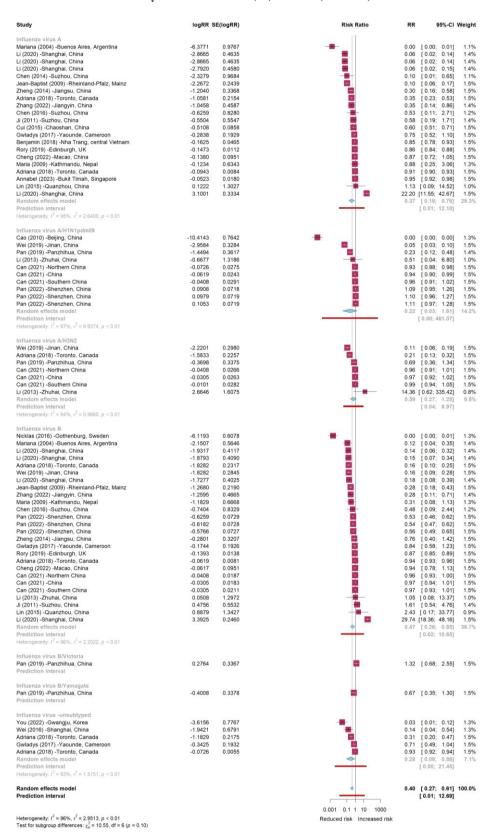


Figure S2-1. Subgroup analysis of influenza virus (IV) meta-analysis by Köppen-Geiger climate

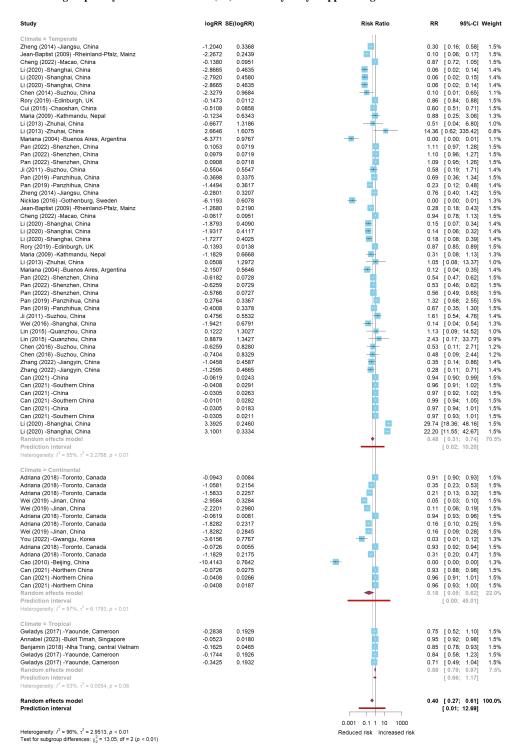


Figure S2-2. Subgroup analysis of influenza virus (IV) meta-analysis by World Bank income category

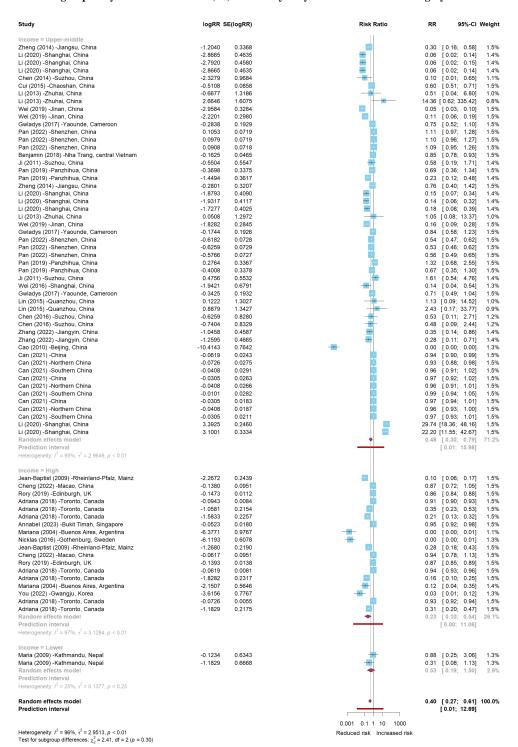


Figure S2-3. Subgroup analysis of influenza virus (IV) meta-analysis by temporal resolution

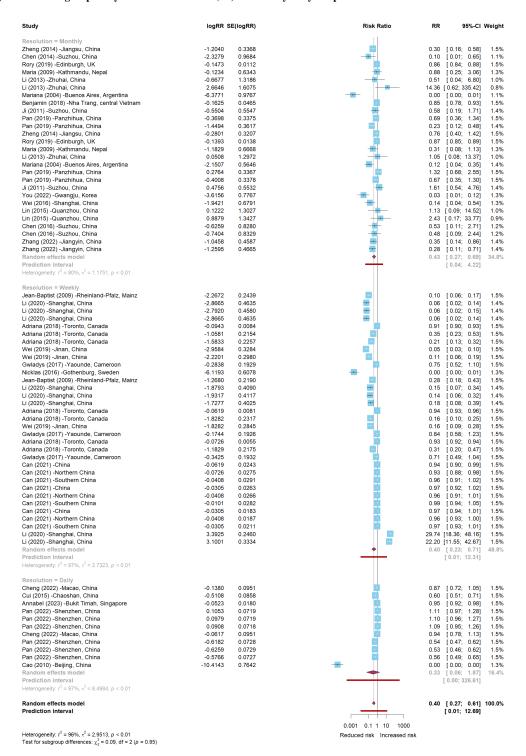
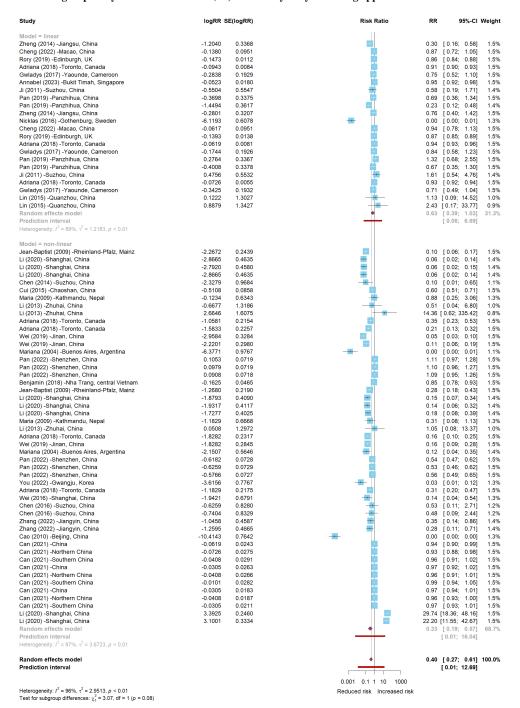


Figure S2-4. Subgroup analysis of influenza virus (IV) meta-analysis exposure measure  $% \left( 1\right) =\left( 1\right) \left( 1\right) \left($ 

Study	logRR	SE(logRR)	Risk Ratio	RR	95%-CI	Weight
Temperature = Mean						
Zheng (2014) -Jiangsu, China	-1.2040	0.3368		0.30	[0.16; 0.58]	1.5%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-2.2672 -0.1380	0.2439 0.0951		0.10	[ 0.06; 0.17]	1.5% 1.5%
Cheng (2022) -Macao, China Li (2020) -Shanghai, China	-2.8665	0.4635	<b>=</b> <sup>™</sup>	0.06	[ 0.72; 1.05] [ 0.02; 0.14]	1.4%
Chen (2014) -Suzhou, China	-2.3279	0.9684		0.10	[ 0.01; 0.65]	1.1%
Rory (2019) -Edinburgh, UK	-0.1473	0.0112		0.86	[0.84; 0.88]	1.5%
Cui (2015) -Chaoshan, China	-0.5108	0.0858		0.60	[0.51; 0.71]	1.5%
Maria (2009) -Kathmandu, Nepal	-0.1234	0.6343	<u>₩</u>	0.88	[ 0.25; 3.06]	1.3%
Li (2013) -Zhuhai, China	-0.6677	1.3186	<del></del> _	0.51	[ 0.04; 6.80]	1.0%
Li (2013) -Zhuhai, China	2.6646 -0.0943	1.6075 0.0084		14.36 0.91	[ 0.62; 335.42]	0.8% 1.5%
Adriana (2018) -Toronto, Canada Adriana (2018) -Toronto, Canada	-1.0581	0.0064		0.35	[0.90, 0.93]	1.5%
Adriana (2018) -Toronto, Canada	-1.5833	0.2154	4	0.33	[ 0.13; 0.32]	1.5%
Wei (2019) -Jinan, China	-2.9584	0.3284		0.05		1.5%
Wei (2019) -Jinan, China	-2.2201	0.2980		0.11	[0.06; 0.19]	1.5%
Gwladys (2017) -Yaounde, Cameroon	-0.2838	0.1929		0.75	[ 0.52; 1.10]	1.5%
Annabel (2023) -Bukit Timah, Singapore	-0.0523	0.0180		0.95	[ 0.92; 0.98]	1.5%
Mariana (2004) -Buenos Aires, Argentina	-6.3771	0.9767	<del></del>	0.00	[ 0.00; 0.01]	1.1%
Pan (2022) -Shenzhen, China	0.1053 -0.1625	0.0719 0.0465		1.11 0.85	[0.97; 1.28]	1.5% 1.5%
Benjamin (2018) -Nha Trang, central Vietnam Ji (2011) -Suzhou, China	-0.5504	0.5547	<b>#</b>	0.58	[ 0.78; 0.93] [ 0.19; 1.71]	1.4%
Pan (2019) -Panzhihua, China	-0.3698	0.3375		0.69	[ 0.36; 1.34]	1.5%
Pan (2019) -Panzhihua, China	-1.4494	0.3617		0.23	[ 0.12; 0.48]	1.4%
Zheng (2014) -Jiangsu, China	-0.2801	0.3207		0.76	[ 0.40; 1.42]	1.5%
Nicklas (2016) -Gothenburg, Sweden	-6.1193	0.6078	-	0.00	[ 0.00; 0.01]	1.3%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-1.2680	0.2190		0.28	[ 0.18; 0.43]	1.5%
Cheng (2022) -Macao, China	-0.0617	0.0951		0.94	[ 0.78; 1.13]	1.5%
Li (2020) -Shanghai, China	-1.8793	0.4090	<b>≖</b>	0.15	[ 0.07; 0.34]	1.4%
Rory (2019) -Edinburgh, UK	-0.1393	0.0138	<u></u>	0.87	[ 0.85; 0.89]	1.5%
Maria (2009) -Kathmandu, Nepal Li (2013) -Zhuhai, China	-1.1829 0.0508	0.6668 1.2972		0.31 1.05	[ 0.08; 1.13]	1.3%
Adriana (2018) -Toronto, Canada	-0.0619	0.0081		0.94	[ 0.00, 10.07]	1.5%
Adriana (2018) -Toronto, Canada	-1.8282	0.2317	<b></b>	0.16	[ 0.10; 0.25]	1.5%
Wei (2019) -Jinan, China	-1.8282	0.2845	<u></u>	0.16	[0.09; 0.28]	1.5%
Gwladys (2017) -Yaounde, Cameroon	-0.1744	0.1926		0.84	[ 0.58; 1.23]	1.5%
Mariana (2004) -Buenos Aires, Argentina	-2.1507	0.5646	<del></del>	0.12	[ 0.04; 0.35]	1.4%
Pan (2022) -Shenzhen, China	-0.6182	0.0728		0.54	[ 0.47; 0.62]	1.5%
Pan (2019) -Panzhihua, China	0.2764	0.3367		1.32	[ 0.68; 2.55]	1.5%
Pan (2019) -Panzhihua, China	-0.4008 0.4756	0.3378		0.67	[ 0.35; 1.30]	1.5%
Ji (2011) -Suzhou, China You (2022) -Gwangju, Korea	-3.6156	0.5532 0.7767		1.61 0.03	[ 0.54; 4.76] [ 0.01; 0.12]	1.4% 1.3%
Adriana (2018) -Toronto, Canada	-0.0726	0.0055		0.03	[0.01, 0.12]	1.5%
Adriana (2018) -Toronto, Canada	-1.1829	0.2175	in the state of t	0.31	[0.20; 0.47]	1.5%
Wei (2016) -Shanghai, China	-1.9421	0.6791	<del></del>	0.14	[0.04; 0.54]	1.3%
Gwladys (2017) -Yaounde, Cameroon	-0.3425	0.1932		0.71	[ 0.49; 1.04]	1.5%
Lin (2015) -Quanzhou, China	0.1222	1.3027	<del>- 1</del>	1.13		1.0%
Lin (2015) -Quanzhou, China	0.8879	1.3427	<u>i</u> -	2.43	[0.17; 33.77]	0.9%
Chen (2016) -Suzhou, China	-0.6259 -0.7404	0.8280 0.8329	<u> </u>	0.53	[0.11; 2.71]	1.2%
Chen (2016) -Suzhou, China Zhang (2022) -Jiangyin, China	-1.0458	0.6329		0.48	[0.09; 2.44]	1.2% 1.4%
Zhang (2022) -Jiangyin, China	-1.2595	0.4665	<u> </u>	0.28	[0.11; 0.71]	1.4%
Cao (2010) -Beijing, China	-10.4143	0.7642	<del></del>	0.00	[0.00; 0.00]	1.3%
Can (2021) -China	-0.0619	0.0243		0.94	[ 0.90; 0.99]	1.5%
Can (2021) -Northern China	-0.0726	0.0275		0.93	[ 0.88; 0.98]	1.5%
Can (2021) -Southern China	-0.0408	0.0291		0.96	[ 0.91; 1.02]	1.5%
Can (2021) -China	-0.0305 -0.0408	0.0263 0.0266	<u></u>	0.97	[ 0.92; 1.02]	1.5%
Can (2021) -Northern China Can (2021) -Southern China	-0.0408	0.0282		0.99	[0.91; 1.01]	1.5% 1.5%
Can (2021) -Southern Crima Can (2021) -China	-0.0305	0.0282		0.97	[ 0.94; 1.01]	1.5%
Can (2021) -Northern China	-0.0408	0.0187		0.96	[ 0.93; 1.00]	1.5%
Can (2021) -Southern China	-0.0305	0.0211		0.97	[0.93; 1.01]	1.5%
Li (2020) -Shanghai, China	3.3925	0.2460	-	29.74	[18.36; 48.16]	1.5%
Li (2020) -Shanghai, China	3.1001	0.3334			[11.55; 42.67]	1.5%
Random effects model			<u> </u>	0.42	[ 0.26; 0.66]	88.3%
Prediction interval					[ 0.01; 15.64]	
Heterogeneity: $I^2 = 96\%$ , $\tau^2 = 3.2193$ , $p < 0.01$						
Temperature = Min						
Li (2020) -Shanghai, China	-2.7920	0.4580	<b>≖</b>	0.06	[ 0.02; 0.15]	1.4%
Pan (2022) -Shenzhen, China	0.0979	0.0719	i i	1.10	[ 0.96; 1.27]	1.5%
Li (2020) -Shanghai, China	-1.9317	0.4117	<b>=</b>	0.14	[ 0.06; 0.32]	1.4%
Pan (2022) -Shenzhen, China	-0.6259	0.0729	2	0.53	[ 0.46; 0.62]	1.5%
Random effects model				0.29	[ 0.08; 1.01]	5.9%
Prediction interval Heterogeneity: $I^2 = 97\%$ , $\tau^2 = 1.5616$ , $p < 0.01$					[ 0.00; 120.84]	
Heterogeneity: 1 = 97%, \(\tau = 1.5616\), \(\rho < 0.01\)						
Temperature = Max						
Li (2020) -Shanghai, China	-2.8665	0.4635	<b>≖</b>	0.06	[ 0.02; 0.14]	1.4%
Pan (2022) -Shenzhen, China	0.0908	0.0718	_ 🙀		[ 0.95; 1.26]	1.5%
Li (2020) -Shanghai, China	-1.7277	0.4025	<b>=</b>	0.18	[ 0.08; 0.39]	1.4%
Pan (2022) -Shenzhen, China	-0.5766	0.0727	31		[ 0.49; 0.65]	1.5%
Random effects model			<b>↑</b>		[ 0.08; 1.05]	5.9%
Prediction interval					[ 0.00; 126.00]	
Heterogeneity: $I^2 = 97\%$ , $\tau^2 = 1.5611$ , $p < 0.01$						
Random effects model				0 40	[ 0.27; 0.61]	100.0%
Prediction interval			_ <del></del>	5.40	[ 0.01; 12.69]	/6
					. ,,	
			0.001 0.1 1 10 1000			
Heterogeneity: $I^2 = 96\%$ , $\tau^2 = 2.9513$ , $p < 0.01$			Reduced risk Increased risk			
Test for subgroup differences: $\chi_2^2 = 0.50$ , df = 2 ( $p = 0.78$ )						

Figure S2-5. Subgroup analysis of influenza virus (IV) meta-analysis by modelling approach



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Figure S2-6. Subgroup analysis of influenza virus (IV) meta-analysis by lag type  $\,$ 

Study	logRR SE	(logRR)	Risk Ratio	RR		95%-CI	Weight
Lag = 0							
Zheng (2014) -Jiangsu, China	-1.2040	0.3368	<u></u>	0.30	[ 0.16;	0.58]	1.5%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-2.2672	0.2439		0.10	[ 0.06;	0.17]	1.5%
Cheng (2022) -Macao, China	-0.1380 -2.8665	0.0951 0.4635		0.87	[ 0.72;	1.05]	1.5%
Li (2020) -Shanghai, China Li (2020) -Shanghai, China	-2.8665 -2.7920	0.4580		0.06	[ 0.02;	0.14] 0.15]	1.4%
Li (2020) -Shanghai, China	-2.8665	0.4635	-	0.06	[ 0.02;	0.13]	1.4%
Chen (2014) -Suzhou, China	-2.3279	0.9684	<del>-</del>	0.10	[ 0.01;	0.65]	1.1%
Rory (2019) -Edinburgh, UK	-0.1473	0.0112		0.86	[ 0.84;	0.88]	1.5%
Cui (2015) -Chaoshan, China	-0.5108	0.0858		0.60	[ 0.51;	0.71]	1.5%
Maria (2009) -Kathmandu, Nepal	-0.1234	0.6343		0.88	[ 0.25;	3.06]	1.3%
Li (2013) -Zhuhai, China	-0.6677	1.3186	<del>+</del>	0.51	[ 0.04;	6.80]	1.0%
Li (2013) -Zhuhai, China	2.6646	1.6075	<u> </u>	14.36	[ 0.62;	335.42]	0.8%
Adriana (2018) -Toronto, Canada	-0.0943	0.0084	<u> </u>	0.91	[ 0.90;	0.93]	1.5%
Adriana (2018) -Toronto, Canada	-1.0581	0.2154	<u></u>	0.35	[ 0.23;	0.53]	1.5%
Adriana (2018) -Toronto, Canada	-1.5833	0.2257		0.21	[ 0.13;	0.32]	1.5%
Wei (2019) -Jinan, China	-2.9584 -2.2201	0.3284 0.2980	**************************************	0.05 0.11	[ 0.03;	0.10]	1.5% 1.5%
Wei (2019) -Jinan, China Gwladys (2017) -Yaounde, Cameroon	-0.2838	0.1929		0.75	[ 0.06;	0.19] 1.10]	1.5%
Mariana (2004) -Buenos Aires, Argentina	-6.3771	0.1929	<b>-</b> T	0.00	[ 0.02;	0.01]	1.1%
Pan (2022) -Shenzhen, China	0.1053	0.0719		1.11	[ 0.97;	1.28]	1.5%
Pan (2022) -Shenzhen, China	0.0979	0.0719		1.10	[ 0.96;	1.27]	1.5%
Pan (2022) -Shenzhen, China	0.0908	0.0718		1.09	[ 0.95;	1.26]	1.5%
Benjamin (2018) -Nha Trang, central Vietnam	-0.1625	0.0465		0.85	[ 0.78;	0.93]	1.5%
Ji (2011) -Suzhou, China	-0.5504	0.5547		0.58	[ 0.19;	1.71]	1.4%
Pan (2019) -Panzhihua, China	-0.3698	0.3375		0.69	[ 0.36;	1.34]	1.5%
Pan (2019) -Panzhihua, China	-1.4494	0.3617		0.23	[ 0.12;	0.48]	1.4%
Zheng (2014) -Jiangsu, China	-0.2801	0.3207		0.76	[ 0.40;	1.42]	1.5%
Nicklas (2016) -Gothenburg, Sweden	-6.1193	0.6078	<u>■</u>	0.00	[ 0.00;	0.01]	1.3%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-1.2680	0.2190	<b>1</b> .	0.28	[ 0.18;	0.43]	1.5%
Cheng (2022) -Macao, China	-0.0617 -1.8793	0.0951 0.4090		0.94 0.15	[ 0.78;	1.13] 0.34]	1.5% 1.4%
Li (2020) -Shanghai, China Li (2020) -Shanghai, China	-1.9317	0.4117		0.15	[ 0.07; [ 0.06;	0.34]	1.4%
Li (2020) -Shanghai, China	-1.7277	0.4025		0.14	[ 0.08;	0.32]	1.4%
Rory (2019) -Edinburgh, UK	-0.1393	0.0138		0.10	[ 0.85;	0.89]	1.5%
Maria (2009) -Kathmandu, Nepal	-1.1829	0.6668		0.31	[ 0.08;	1.13]	1.3%
Li (2013) -Zhuhai, China	0.0508	1.2972		1.05	[ 0.08;	13.37]	1.0%
Adriana (2018) -Toronto, Canada	-0.0619	0.0081		0.94	[ 0.93;	0.96]	1.5%
Adriana (2018) -Toronto, Canada	-1.8282	0.2317	•	0.16	[ 0.10;	0.25]	1.5%
Wei (2019) -Jinan, China	-1.8282	0.2845	•	0.16	[ 0.09;	0.28]	1.5%
Gwladys (2017) -Yaounde, Cameroon	-0.1744	0.1926		0.84	[ 0.58;	1.23]	1.5%
Mariana (2004) -Buenos Aires, Argentina	-2.1507	0.5646	-	0.12	[ 0.04;	0.35]	1.4%
Pan (2022) -Shenzhen, China	-0.6182	0.0728	<u></u>	0.54	[ 0.47;	0.62]	1.5%
Pan (2022) -Shenzhen, China	-0.6259	0.0729		0.53	[ 0.46;	0.62]	1.5%
Pan (2022) -Shenzhen, China	-0.5766 0.2764	0.0727 0.3367		0.56 1.32	[ 0.49; [ 0.68;	0.65] 2.55]	1.5% 1.5%
Pan (2019) -Panzhihua, China Pan (2019) -Panzhihua, China	-0.4008	0.3378		0.67	[ 0.35;	1.30]	1.5%
Ji (2011) -Suzhou, China	0.4756	0.5532		1.61	[ 0.54;	4.76]	1.4%
You (2022) -Gwangju, Korea	-3.6156	0.7767		0.03	[ 0.01;	0.12]	1.3%
Adriana (2018) -Toronto, Canada	-0.0726	0.0055		0.93	[ 0.92;	0.94]	1.5%
Adriana (2018) -Toronto, Canada	-1.1829	0.2175		0.31	[ 0.20;	0.47]	1.5%
Wei (2016) -Shanghai, China	-1.9421	0.6791		0.14	[ 0.04;	0.54]	1.3%
Gwladys (2017) -Yaounde, Cameroon	-0.3425	0.1932		0.71	[ 0.49;	1.04]	1.5%
Lin (2015) -Quanzhou, China	0.1222	1.3027	*	1.13	[ 0.09;	14.52]	1.0%
Lin (2015) -Quanzhou, China	0.8879	1.3427	<u></u>	2.43	[ 0.17;	33.77]	0.9%
Chen (2016) -Suzhou, China	-0.6259	0.8280	The state of the s	0.53	[ 0.11;	2.71]	1.2%
Chen (2016) -Suzhou, China	-0.7404	0.8329		0.48	[ 0.09;	2.44]	1.2%
Zhang (2022) - Jiangyin, China	-1.0458	0.4587 0.4665	<u> </u>	0.35	[ 0.14;	0.86]	1.4% 1.4%
Zhang (2022) -Jiangyin, China Cao (2010) -Beijing, China	-1.2595 -10.4143	0.4665	_ 1	0.28	[ 0.11; [ 0.00:	0.71]	1.4%
Can (2010) -Beijing, China Can (2021) -China	-0.0619	0.0243	-	0.00	[ 0.90;	0.00]	1.5%
Can (2021) -Northern China	-0.0726	0.0275		0.93	[ 0.88;	0.98]	1.5%
Can (2021) -Northern China	-0.0408	0.0291	in the second se	0.96	[ 0.91;	1.02]	1.5%
Can (2021) -China	-0.0305	0.0263	ii ii	0.97	[ 0.92;	1.02]	1.5%
Can (2021) -Northern China	-0.0408	0.0266	i i	0.96	[ 0.91;	1.01]	1.5%
Can (2021) -Southern China	-0.0101	0.0282		0.99	[ 0.94;	1.05]	1.5%
Can (2021) -China	-0.0305	0.0183		0.97	[ 0.94;	1.01]	1.5%
Can (2021) -Northern China	-0.0408	0.0187		0.96	[ 0.93;	1.00]	1.5%
Can (2021) -Southern China	-0.0305	0.0211		0.97	[ 0.93;	1.01]	1.5%
Random effects model			•	0.35	[ 0.24;	0.52]	95.5%
Prediction interval			<u> </u>		[ 0.02;	8.08]	
Heterogeneity: $I^2 = 95\%$ , $\tau^2 = 2.4304$ , $p < 0.01$							
Lag = >1							
Annabel (2023) -Bukit Timah, Singapore	-0.0523	0.0180	<u></u>	0.95	[ 0.92;	0.98]	1.5%
Li (2020) -Shanghai, China	3.3925	0.2460	+	29.74	[18.36;	48.16]	
Li (2020) -Shanghai, China	3.1001	0.3334	<u> </u>	22.20	[11.55;	42.67]	
Random effects model			•	8.42	[ 0.95;	74.40]	4.5%
Prediction interval				- [	0.00; 133271759	86419.18]	
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 3.6514$ , $p < 0.01$							
Random effects model			4	0.40	[ 0.27;		100.0%
Prediction interval			<del></del> _		[ 0.01;	12.69]	
			0.001 1 1000				
Heterogeneity: $I^2 = 96\%$ , $\tau^2 = 2.9513$ , $p < 0.01$							
Heterogeneity: $I^{-} = 96\%$ , $\tau^{-} = 2.9513$ , $p < 0.01$ Test for subgroup differences: $\chi_{1}^{2} = 7.92$ , df = 1 ( $p < 0.01$ )			Reduced risk Increased risk				
anagrap amaranasa (4 - 1.06, 41 - 1.07 - 0.01)							

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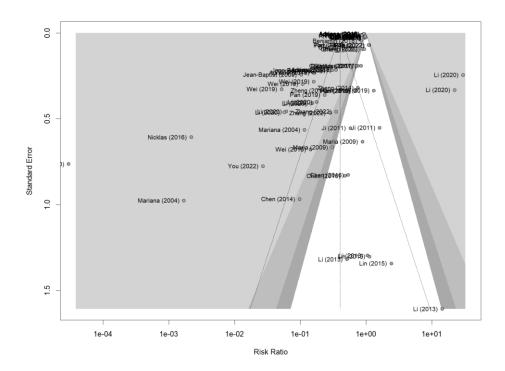
Figure S2-7. Leave-one-out analysis of influenza virus (IV) meta-analysis  $\,$ 

Study	Risk Ratio	RR	95%-CI	P-value	Tau2	Tau	12
Omitting Zheng (2014) -Jiangsu, China		0.40	[0.26; 0.61]	< 0.01	3.0037	1.7331	96%
Omitting Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	_ <del></del>	0.41	[0.27; 0.62]	< 0.01	2.9708	1.7236	95%
Omitting Cheng (2022) -Macao, China	<del>- •</del>	0.40	[0.26; 0.60]	< 0.01	2.9968	1.7311	96%
Omitting Li (2020) -Shanghai, China		0.41	[0.27; 0.63]	< 0.01	2.9374	1.7139	96%
Omitting Li (2020) -Shanghai, China		0.41	[0.27; 0.63]	< 0.01	2.9422	1.7153	96%
Omitting Li (2020) -Shanghai, China Omitting Chen (2014) -Suzhou, China		0.41 0.41	[0.27; 0.63]	< 0.01	2.9374	1.7139	96% 96%
Omitting Crief (2014) -Suzhou, Crima Omitting Rory (2019) -Edinburgh, UK		0.41	[0.27, 0.62]	< 0.01	2.9707	1.7230	96%
Omitting Cui (2015) -Chaoshan, China		0.40	[0.26; 0.61]	< 0.01	3.0048	1.7334	96%
Omitting Maria (2009) -Kathmandu, Nepal	<del></del>	0.40	[0.26; 0.60]	< 0.01	2.9924	1.7299	96%
Omitting Li (2013) -Zhuhai, China	<del></del>	0.40	[0.26; 0.61]	< 0.01	2.9866	1.7282	96%
Omitting Li (2013) -Zhuhai, China	<del></del>	0.39	[0.26; 0.59]	< 0.01	2.9150	1.7073	96%
Omitting Adriana (2018) -Toronto, Canada		0.40 0.40	[0.26; 0.60]	< 0.01	2.9956 3.0062	1.7308	96% 96%
Omitting Adriana (2018) -Toronto, Canada Omitting Adriana (2018) -Toronto, Canada		0.40	[0.26; 0.61]	< 0.01	2.9975	1.7313	96%
Omitting Wei (2019) -Jinan, China		0.41	[0.27; 0.63]	< 0.01	2.9277	1.7313	95%
Omitting Wei (2019) -Jinan, China	<del></del>	0.41	[0.27; 0.62]	< 0.01	2.9733	1.7243	96%
Omitting Gwladys (2017) -Yaounde, Cameroon	<del></del>	0.40	[0.26; 0.61]	< 0.01	3.0002	1.7321	96%
Omitting Annabel (2023) -Bukit Timah, Singapore	<del></del>	0.40	[0.26; 0.60]	< 0.01	2.9943	1.7304	96%
Omitting Mariana (2004) -Buenos Aires, Argentina		0.43	[0.29; 0.64]	< 0.01	2.6690	1.6337	96%
Omitting Pan (2022) -Shenzhen, China Omitting Pan (2022) -Shenzhen, China		0.40	[0.26; 0.60]	< 0.01 < 0.01	2.9887	1.7288	96% 96%
Omitting Pan (2022) -Shenzhen, China		0.40	[0.26; 0.60]	< 0.01	2 9892	1.7289	96%
Omitting Benjamin (2018) -Nha Trang, central Vietnam		0.40	[0.26; 0.60]	< 0.01	2.9975	1.7313	96%
Omitting Ji (2011) -Suzhou, China	<del></del>	0.40	[0.26; 0.61]	< 0.01	3.0007	1.7322	96%
Omitting Pan (2019) -Panzhihua, China	<del>- •</del>	0.40	[0.26; 0.61]	< 0.01	3.0009	1.7323	96%
Omitting Pan (2019) -Panzhihua, China	<del></del>	0.40	[0.27; 0.62]	< 0.01	2.9995	1.7319	96%
Omitting Zheng (2014) -Jiangsu, China		0.40	[0.26; 0.61]	< 0.01 < 0.01	2.9992	1.7318	96% 95%
Omitting Nicklas (2016) -Gothenburg, Sweden Omitting Jean-Baptist (2009) -Rheinland-Pfalz, Mainz		0.43	[0.29; 0.64]	< 0.01	2.5804 3.0039	1.6063	95% 96%
Omitting Cheng (2022) -Macao, China		0.40	[0.26; 0.60]	< 0.01	2.9945	1.7305	96%
Omitting Li (2020) -Shanghai, China		0.41	[0.27; 0.62]	< 0.01	2.9874	1.7284	96%
Omitting Li (2020) -Shanghai, China	<del>- •</del>	0.41	[0.27; 0.62]	< 0.01	2.9855	1.7279	96%
Omitting Li (2020) -Shanghai, China		0.41	[0.27; 0.62]	< 0.01	2.9923	1.7298	96%
Omitting Rory (2019) -Edinburgh, UK	<del>- •</del>	0.40	[0.26; 0.60]	< 0.01	2.9969	1.7312	96%
Omitting Maria (2009) -Kathmandu, Nepal Omitting Li (2013) -Zhuhai, China		0.40	[0.26; 0.61]	< 0.01 < 0.01	2.9990	1.7318	96% 96%
Omitting Er (2013) -Zhuhar, China Omitting Adriana (2018) -Toronto, Canada		0.40	[0.26; 0.60]	< 0.01	2.9946	1.7305	96%
Omitting Adriana (2018) -Toronto, Canada	<del></del> _	0.41	[0.27; 0.62]	< 0.01	2.9899	1.7291	96%
Omitting Wei (2019) -Jinan, China	<del></del>	0.41	[0.27; 0.62]	< 0.01	2.9897	1.7291	96%
Omitting Gwladys (2017) -Yaounde, Cameroon	<del></del>	0.40	[0.26; 0.60]	< 0.01	2.9975	1.7313	96%
Omitting Mariana (2004) -Buenos Aires, Argentina		0.41	[0.27; 0.62]	< 0.01	2.9768	1.7253	96%
Omitting Pan (2022) -Shenzhen, China Omitting Pan (2022) -Shenzhen, China		0.40 0.40	[0.26; 0.61]	< 0.01 < 0.01	3.0062 3.0062	1.7338 1.7338	96% 96%
Omitting Pan (2022) -Shenzhen, China Omitting Pan (2022) -Shenzhen, China		0.40	[0.26; 0.61]	< 0.01	3.0057	1.7337	96%
Omitting Pan (2019) -Panzhihua, China		0.39	[0.26; 0.60]	< 0.01	2.9815	1.7267	96%
Omitting Pan (2019) -Panzhihua, China	<del></del>	0.40	[0.26; 0.61]	< 0.01	3.0014	1.7325	96%
Omitting Ji (2011) -Suzhou, China	<del></del>	0.39	[0.26; 0.60]	< 0.01	2.9733	1.7243	96%
Omitting You (2022) -Gwangju, Korea	<del>- •</del>	0.42	[0.28; 0.63]	< 0.01	2.8984	1.7025	96%
Omitting Adriana (2018) -Toronto, Canada		0.40	[0.26; 0.60]	< 0.01	2.9949	1.7306	96%
Omitting Adriana (2018) -Toronto, Canada Omitting Wei (2016) -Shanghai, China		0.40 0.41	[0.26; 0.61]	< 0.01 < 0.01	3.0050 2.9837	1.7335 1.7274	96% 96%
Omitting Gwladys (2017) -Yaounde, Cameroon		0.41	[0.26; 0.61]	< 0.01	3.0014	1.7325	96%
Omitting Lin (2015) -Quanzhou, China		0.40	[0.26; 0.60]	< 0.01	2.9797	1.7262	96%
Omitting Lin (2015) -Quanzhou, China	<del></del>	0.39	[0.26; 0.60]	< 0.01	2.9637	1.7216	96%
Omitting Chen (2016) -Suzhou, China	<del></del>	0.40	[0.26; 0.61]	< 0.01	2.9963	1.7310	96%
Omitting Chen (2016) -Suzhou, China		0.40	[0.26; 0.61]	< 0.01	2.9968	1.7311	96%
Omitting Zhang (2022) -Jiangyin, China Omitting Zhang (2022) -Jiangyin, China		0.40 0.40	[0.26; 0.61]	< 0.01	3.0035	1.7331	96% 96%
Omitting Cao (2010) -Beijing, China	•	0.46	[0.33; 0.64]	< 0.01	1.8413	1.3570	95%
Omitting Can (2021) -China		0.40	[0.26; 0.60]	< 0.01	2.9946	1.7305	96%
Omitting Can (2021) -Northern China	<del></del>	0.40	[0.26; 0.60]	< 0.01	2.9949	1.7306	96%
Omitting Can (2021) -Southern China	<del>- •</del>	0.40	[0.26; 0.60]	< 0.01	2.9939	1.7303	96%
Omitting Can (2021) -China	<del></del>	0.40	[0.26; 0.60]	< 0.01	2.9936	1.7302	96%
Omitting Can (2021) -Northern China		0.40	[0.26; 0.60]	< 0.01	2.9939	1.7303	96%
Omitting Can (2021) -Southern China Omitting Can (2021) -China		0.40 0.40	[0.26; 0.60]	< 0.01 < 0.01	2.9929 2.9936	1.7300 1.7302	96% 96%
Omitting Can (2021) -Northern China		0.40	[0.26; 0.60]	< 0.01	2.9939	1.7302	96%
Omitting Can (2021) -Notthern China	<del></del>	0.40	[0.26; 0.60]	< 0.01	2.9936	1.7302	96%
Omitting Li (2020) -Shanghai, China	<del></del>	0.38	[0.25; 0.56]	< 0.01	2.6609	1.6312	95%
Omitting Li (2020) -Shanghai, China	<del>- •</del>	0.38	[0.25; 0.57]	< 0.01	2.7176	1.6485	95%
Dondom effects model		0.40	ro 07: 0 641	- 0 01	0.0540	4 7470	00%
Random effects model		0.40	[0.27; 0.61]	< 0.01	2.9513	1.7179	96%
	0.5 1 2						

Figure S2-8. Sensitivity analysis by switching the relative risk for cold effects for Li et al  $2020\,$ 

Study	logRR S	E(logRR)	Risk Ratio	RR	95%-CI	Weight
Influenza virus A Mariana (2004) -Buenos Aires, Argentina Li (2020) -Shanghai, China Li (2020) -Shanghai, China	-6.3771 -2.8665 -2.8665	0.9767 0.4635 0.4635		0.00 0.06 0.06	[0.00; 0.01] [0.02; 0.14] [0.02; 0.14]	1.1% 1.4% 1.4%
Li (2020) -Shanghai, China Li (2020) -Shanghai, China	-2.7920	0.4580	<b>=</b>	0.06	[ 0.02; 0.14]	1.4%
Chen (2014) -Suzhou, China	-2.3279	0.9684	<u>-</u>		[ 0.01; 0.65]	1.1%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz Zheng (2014) -Jiangsu, China	-2.2672 -1.2040	0.2439 0.3368		0.10	[ 0.06; 0.17] [ 0.16; 0.58]	1.5% 1.5%
Adriana (2018) -Toronto, Canada	-1.0581	0.3366		0.35	[ 0.10, 0.58]	1.5%
Zhang (2022) -Jiangyin, China	-1.0458	0.4587	<u>≠</u>	0.35	[ 0.14; 0.86]	1.4%
Chen (2016) -Suzhou, China Ji (2011) -Suzhou, China	-0.6259 -0.5504	0.8280 0.5547	-	0.53	[0.11; 2.71] [0.19; 1.71]	1.2% 1.4%
Cui (2015) -Chaoshan, China	-0.5108	0.0858		0.60	[0.51; 0.71]	1.5%
Gwladys (2017) -Yaounde, Cameroon	-0.2838	0.1929	<u> </u>	0.75	[ 0.52; 1.10]	1.5%
Benjamin (2018) -Nha Trang, central Vietnam Rory (2019) -Edinburgh, UK	-0.1625 -0.1473	0.0465 0.0112		0.85 0.86	[0.78; 0.93]	1.5% 1.5%
Cheng (2022) -Macao, China	-0.1380	0.0951		0.87	[ 0.72; 1.05]	1.5%
Maria (2009) -Kathmandu, Nepal Adriana (2018) -Toronto, Canada	-0.1234 -0.0943	0.6343 0.0084	<u> </u>	0.88	[ 0.25; 3.06] [ 0.90; 0.93]	1.3% 1.5%
Annabel (2023) -Bukit Timah, Singapore	-0.0523	0.0180				1.5%
Lin (2015) -Quanzhou, China	0.1222	1.3027	<del>  </del>		[ 0.09; 14.52]	1.0%
Li (2020) -Shanghai, China Random effects model	2.3561	0.1538	•	0.36	[7.80; 14.26] [0.18; 0.72]	1.5% 29.3%
Prediction interval			<del>-     -</del>		[0.01; 9.68]	
Heterogeneity: $I^2 = 96\%$ , $\tau^2 = 2.3466$ , $p < 0.01$						
Influenza virus A/H1N1pdm09						
Cao (2010) -Beijing, China Wei (2019) -Jinan, China	-10.4143 -2.9584	0.7642 0.3284	<del></del>	0.00	[0.00; 0.00]	1.3% 1.5%
Pan (2019) -Panzhihua, China	-2.9584 -1.4494	0.3284	<b></b>	0.05	[ 0.12; 0.48]	1.5%
Li (2013) -Zhuhai, China	-0.6677	1.3186	<del>-</del>	0.51	[0.04; 6.80]	0.9%
Can (2021) -Northern China Can (2021) -China	-0.0726 -0.0619	0.0275 0.0243		0.93 0.94	[0.88; 0.98]	1.5% 1.5%
Can (2021) -Southern China	-0.0408	0.0291	<u> </u>	0.96	[ 0.91; 1.02]	1.5%
Pan (2022) -Shenzhen, China Pan (2022) -Shenzhen, China	0.0908 0.0979	0.0718 0.0719		1.09	[ 0.95; 1.26] [ 0.96; 1.27]	1.5% 1.5%
Pan (2022) -Shenzhen, China	0.1053	0.0719	į.	1.11	[ 0.97; 1.28]	1.5%
Random effects model Prediction interval			<u></u> _	0.22	[ 0.03; 1.61] [ 0.00; 461.07]	14.2%
Heterogeneity: $I^2 = 97\%$ , $\tau^2 = 9.9374$ , $\rho < 0.01$					[ 0.00, 461.07]	
Influenza virus A/H3N2 Wei (2019) -Jinan, China	-2.2201	0.2980		0.44	[ 0.06; 0.19]	1.5%
Adriana (2018) -Toronto, Canada	-1.5833	0.2960	tent		[0.06, 0.19]	1.5%
Pan (2019) -Panzhihua, China	-0.3698	0.3375		0.69	[ 0.36; 1.34]	1.5%
Can (2021) -Northern China Can (2021) -China	-0.0408 -0.0305	0.0266 0.0263		0.96 0.97	[ 0.91; 1.01] [ 0.92; 1.02]	1.5% 1.5%
Can (2021) -Southern China	-0.0101	0.0282		0.99	[ 0.94; 1.05]	1.5%
Li (2013) -Zhuhai, China Random effects model	2.6646	1.6075		14.36 0.59	[ 0.62; 335.42] [ 0.27; 1.28]	0.8% 9.8%
Prediction interval			<del>-   -</del>		[0.04; 8.97]	
Heterogeneity: $I^2 = 94\%$ , $\tau^2 = 0.9660$ , $p < 0.01$						
Influenza virus B	0.4400		_			
Nicklas (2016) -Gothenburg, Sweden Mariana (2004) -Buenos Aires, Argentina	-6.1193 -2.1507	0.6078 0.5646	-	0.00	[ 0.00; 0.01] [ 0.04; 0.35]	1.3% 1.4%
Li (2020) -Shanghai, China	-1.9317	0.4117	•	0.14	[ 0.06; 0.32]	1.4%
Li (2020) -Shanghai, China Adriana (2018) -Toronto, Canada	-1.8793 -1.8282	0.4090 0.2317			[ 0.07; 0.34] [ 0.10; 0.25]	1.4% 1.5%
Wei (2019) -Jinan, China	-1.8282	0.2845	•			1.5%
Li (2020) -Shanghai, China	-1.7277	0.4025	<u>=</u>	0.18	[ 0.08; 0.39]	1.4%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz Zhang (2022) -Jiangyin, China	-1.2680 -1.2595	0.2190 0.4665	=	0.28	[0.18; 0.43]	1.5% 1.4%
Maria (2009) -Kathmandu, Nepal	-1.1829	0.6668	<del>-</del>	0.31	[ 0.08; 1.13]	1.3%
Chen (2016) -Suzhou, China Pan (2022) -Shenzhen, China	-0.7404 -0.6259	0.8329 0.0729		0.48	[ 0.09; 2.44] [ 0.46; 0.62]	1.2% 1.5%
Pan (2022) -Shenzhen, China	-0.6182	0.0728		0.54	[ 0.47; 0.62]	1.5%
Pan (2022) -Shenzhen, China	-0.5766	0.0727 0.3207		0.56	[ 0.49; 0.65]	1.5%
Zheng (2014) -Jiangsu, China Gwladys (2017) -Yaounde, Cameroon	-0.2801 -0.1744	0.3207		0.76 0.84	[ 0.40; 1.42] [ 0.58; 1.23]	1.5% 1.5%
Rory (2019) -Edinburgh, UK	-0.1393	0.0138		0.87	[ 0.85; 0.89]	1.5%
Adriana (2018) -Toronto, Canada Cheng (2022) -Macao, China	-0.0619 -0.0617	0.0081 0.0951	<u> </u>		[ 0.93; 0.96] [ 0.78; 1.13]	1.5% 1.5%
Can (2021) -Northern China	-0.0408	0.0187			[ 0.93; 1.00]	1.5%
Can (2021) -China Can (2021) -Southern China	-0.0305 -0.0305	0.0183 0.0211	<u> </u>		[ 0.94; 1.01] [ 0.93; 1.01]	1.5% 1.5%
Li (2013) -Zhuhai, China	0.0508	1.2972			[ 0.08; 13.37]	1.0%
Ji (2011) -Suzhou, China	0.4756	0.5532	<u></u>		[ 0.54; 4.76]	1.4%
Lin (2015) -Quanzhou, China Li (2020) -Shanghai, China	0.8879 3.3925	1.3427 0.2460			[ 0.17; 33.77] [18.36; 48.16]	0.9% 1.5%
Random effects model			<u></u>	0.47	[ 0.26; 0.85]	36.7%
Prediction interval Heterogeneity: $I^2 = 96\%$ , $\tau^2 = 2.2022$ , $p < 0.01$					[ 0.02; 10.65]	
Influenza virus B/Victoria Pan (2019) -Panzhihua, China	0.2764	0.3367		1.32	[ 0.68; 2.55]	1.5%
Prediction interval Influenza virus B/Yamagata						
Pan (2019) -Panzhihua, China Prediction interval	-0.4008	0.3378		0.67	[ 0.35; 1.30]	1.5%
Influenza virus -unsubtyped						
You (2022) -Gwangju, Korea	-3.6156	0.7767	-		[ 0.01; 0.12]	1.3%
Wei (2016) -Shanghai, China Adriana (2018) -Toronto, Canada	-1.9421 -1.1829	0.6791 0.2175	-		[ 0.04; 0.54]	1.3% 1.5%
Gwladys (2017) -Yaounde, Cameroon	-0.3425	0.1932		0.71	[ 0.49; 1.04]	1.5%
Adriana (2018) -Toronto, Canada Random effects model	-0.0726	0.0055	<u></u>		[ 0.92; 0.94] [ 0.09; 0.88]	1.5% 7.1%
Prediction interval			<del>-  </del>	0.20	[ 0.00; 21.45]	/0
Heterogeneity: $I^2 = 93\%$ , $\tau^2 = 1.5151$ , $p < 0.01$						
Random effects model Prediction interval			•	0.40	[ 0.26; 0.60] [ 0.01; 11.97]	100.0%
			0.001 0.1 1 10 1000			
Heterogeneity: $I^2 = 96\%$ , $\tau^2 = 2.8670$ , $p < 0.01$ Test for subgroup differences: $\chi_6^2 = 10.89$ , df = 6 ( $p = 0.09$ )			Reduced risk Increased risk			

Figure S2-9. Funnel plot of influenza virus (IV) estimates



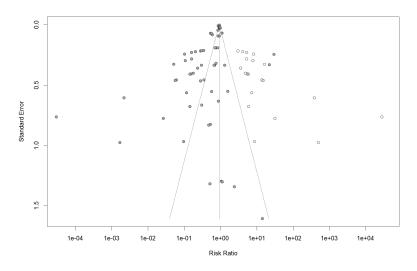
## Linear regression test of funnel plot asymmetry

Egger's test result: t = -4.29, df = 69, p-value < 0.0001

Sample estimates:

bias	se.bias	intercept	se.intercept
-2.4900	0.5804	-0.0516	0.0156

Figure S2-10. Trim and fill of influenza virus (IV) meta-analysis



Number of studies: k = 92 (with 21 added studies)

Random effects model: RR= 0.9162 [0.5580; 1.5041], Z= -0.35, P=0.7292

Quantifying heterogeneity:

 $tau^2 = 5.6364 [4.3703; 8.2222]; tau = 2.3741 [2.0905; 2.8674]$ 

 $I^2 = 96.4\%$  [96.0%; 96.8%]; H = 5.27 [4.98; 5.58]

Figure S3. Random-effects meta-analysis of human parainfluenza virus (HPIV) estimates (23 studies)

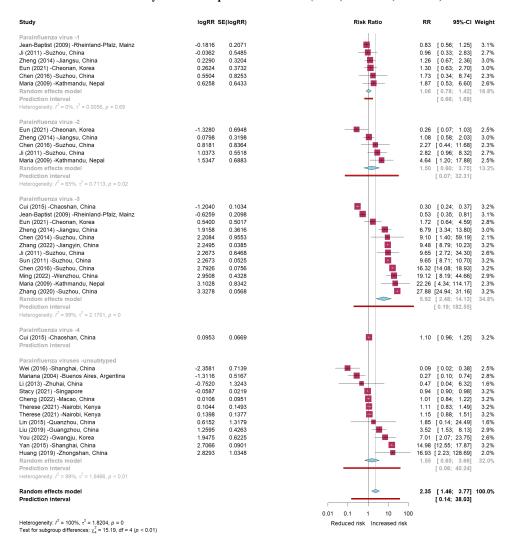


Figure S3-1. Subgroup analysis of human parainfluenza virus (HPIV) meta-analysis by Köppen-Geiger climate

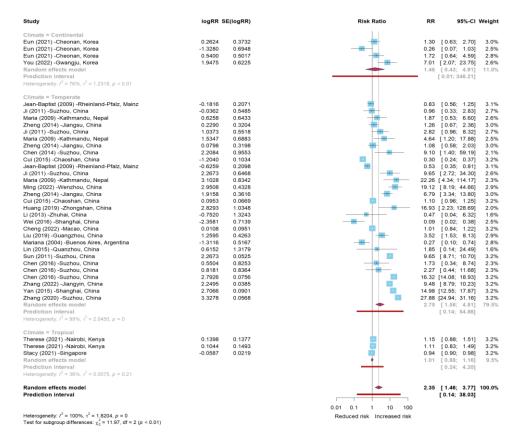
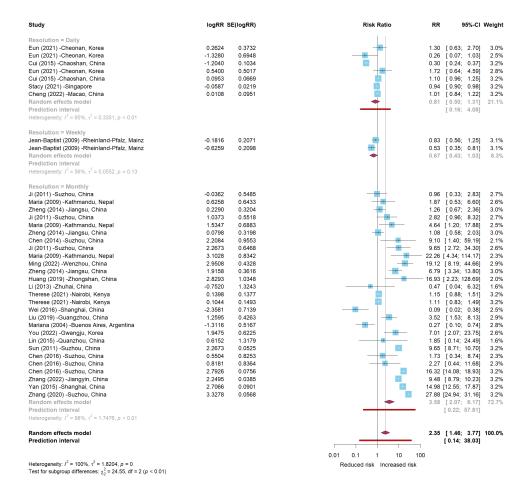


Figure S3-2. Subgroup analysis of human parainfluenza virus (HPIV) meta-analysis by World Bank income category

Study	logRR S	E(logPP)	Risk Ratio	RR		95%-CI	Weight
olday	logitit o	L(logitit)	Hisk Hallo	1414		5070-01	rreigin
Income = High							
Eun (2021) -Cheonan, Korea	0.2624	0.3732	🚍	1.30	[ 0.63;	2.70]	3.0%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-0.1816	0.2071		0.83	[ 0.56;	1.25]	3.1%
Eun (2021) -Cheonan, Korea	-1.3280	0.6948	<u></u>	0.26	[ 0.07;	1.03]	2.5%
Eun (2021) -Cheonan, Korea	0.5400	0.5017		1.72	[ 0.64;	4.59]	2.8%
Jean-Baptist (2009) -Rheinland-Pfalz, Mainz	-0.6259	0.2098		0.53	[ 0.35;	0.81]	3.1%
Stacy (2021) -Singapore	-0.0587	0.0219		0.94	[ 0.90;	0.98]	3.2%
Cheng (2022) -Macao, China	0.0108	0.0951		1.01	[ 0.84;	1.22]	3.2%
Mariana (2004) -Buenos Aires, Argentina	-1.3116 1.9475	0.5167 0.6225	_	0.27 7.01	[ 0.10;	0.74]	2.8%
You (2022) -Gwangju, Korea Random effects model	1.9475	0.6225		0.91	[ 0.56;	1.48]	26.4%
Prediction interval			<u>. I</u>	0.51	[ 0.36,	4.631	20.470
Heterogeneity: $I^2 = 73\%$ , $\tau^2 = 0.4133$ , $p < 0.01$			T		[0.10,	4.03]	
Income = Upper-middle							
Ji (2011) -Suzhou, China	-0.0362	0.5485		0.96	[ 0.33;	2.83]	2.7%
Zheng (2014) -Jiangsu, China	0.2290	0.3204	- 1	1.26	[ 0.67;	2.36]	3.0%
Ji (2011) -Suzhou, China	1.0373	0.5518		2.82	[ 0.96;	8.32]	2.7%
Zheng (2014) -Jiangsu, China	0.0798	0.3198	₩ <u>.</u>	1.08	[ 0.58;	2.03]	3.0%
Chen (2014) -Suzhou, China	2.2084	0.9553		9.10	[ 1.40;	59.19]	2.1%
Cui (2015) -Chaoshan, China	-1.2040	0.1034		0.30	[ 0.24;	0.37]	3.2%
Ji (2011) -Suzhou, China	2.2673	0.6468	_	9.65	[ 2.72;	34.30]	2.6%
Ming (2022) -Wenzhou, China	2.9508	0.4328		19.12	[ 8.19;	44.66]	2.9%
Zheng (2014) -Jiangsu, China	1.9158 0.0953	0.3616 0.0669		6.79 1.10	[ 3.34;	13.80]	3.0% 3.2%
Cui (2015) -Chaoshan, China	2.8293	1.0348	TI	16.93	[ 0.96;	1.25] 128.69]	2.0%
Huang (2019) -Zhongshan, China Li (2013) -Zhuhai, China	-0.7520	1.3243		0.47	[ 2.23;	6.321	1.6%
Therese (2021) -Nairobi, Kenya	0.1398	0.1377		1.15	[ 0.88;	1.51]	3.2%
Therese (2021) -Nairobi, Kenya Therese (2021) -Nairobi, Kenya	0.1398	0.1377		1.15	[ 0.83;	1.49]	3.2%
Wei (2016) -Shanghai, China	-2.3581	0.7139		0.09	[ 0.02;	0.381	2.5%
Liu (2019) -Guangzhou, China	1.2595	0.4263		3.52	[ 1.53;	8.13]	2.9%
Lin (2015) -Guanzhou, China	0.6152	1.3179		1.85	[ 0.14;	24.49]	1.6%
Sun (2011) -Suzhou, China	2.2673	0.0525		9.65	[8.71;	10.70]	3.2%
Chen (2016) -Suzhou, China	0.5504	0.8253		1.73	[ 0.34;	8.74]	2.3%
Chen (2016) -Suzhou, China	0.8181	0.8364	<u>∓</u>	2.27	[ 0.44;	11.68]	2.3%
Chen (2016) -Suzhou, China	2.7926	0.0756		16.32	[14.08;	18.93]	3.2%
Zhang (2022) -Jiangyin, China	2.2495	0.0385		9.48	[8.79;	10.23]	3.2%
Yan (2015) -Shanghai, China	2,7066	0.0901		14.98	[12.55;	17.87]	3.2%
Zhang (2020) -Suzhou, China	3.3278	0.0568		27.88	[24.94]	31.16]	3.2%
Random effects model			<b>→</b>	3.09	[ 1.70;	5.62]	66.2%
Prediction interval			<del>+-</del>		[ 0.16;	59.40]	
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 1.9373$ , $p = 0$							
Income = Lower							
Maria (2009) -Kathmandu, Nepal	0.6258	0.6433	<b>₩</b>	1.87	[ 0.53;	6.60]	2.6%
Maria (2009) -Kathmandu, Nepal	1.5347	0.6883	the state of the s	4.64	[ 1.20;	17.88]	2.5%
Maria (2009) -Kathmandu, Nepal	3.1028	0.8342	-	22.26	[ 4.34;	114.17]	2.3%
Random effects model			•	5.34	[ 1.36;	20.98]	7.5%
Prediction interval				- [	0.00; 2159	7729.55]	
Heterogeneity: $I^2 = 64\%$ , $\tau^2 = 0.9456$ , $p = 0.06$							
Random effects model			•	2.35	[ 1.46;	3,771	100.0%
Prediction interval			<u> </u>		[ 0.14;	38.03]	
					,		
			0.001 0.11 10 1000				
Heterogeneity: $f^2 = 100\%$ , $\tau^2 = 1.8204$ , $p = 0$ Test for subgroup differences: $\chi^2_2 = 12.87$ , df = 2 ( $p < 0.01$ )			Reduced risk Increased risk				
103.101 3abg.oup dilicionces. χ <sub>2</sub> = 12.01, d1 = 2 (β < 0.01)							

Figure S3-3. Subgroup analysis of human parainfluenza virus (HPIV) meta-analysis by temporal resolution



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Figure S3-4. Subgroup analysis of human parainfluenza virus (HPIV) meta-analysis by exposure measure

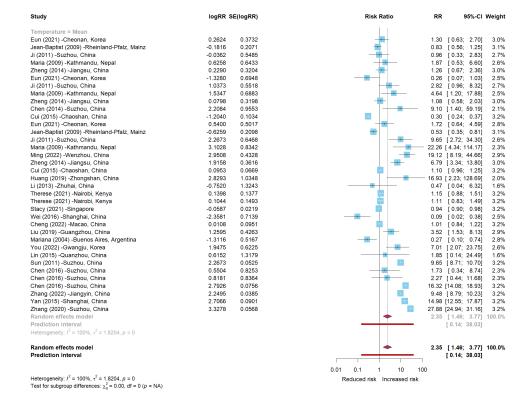


Figure S3-5. Subgroup analysis of human parainfluenza virus (HPIV) meta-analysis by modelling approach

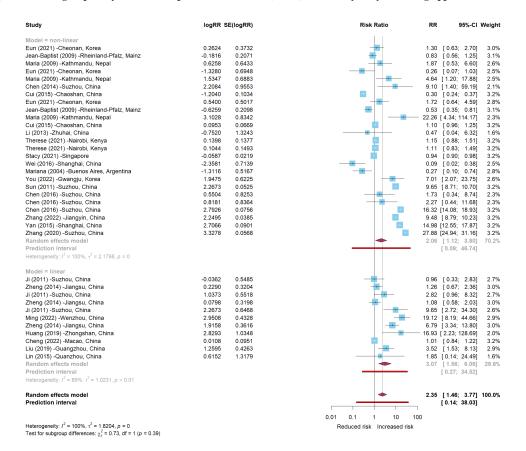


Figure S3-6. Subgroup analysis of human parainfluenza virus (HPIV) meta-analysis by lag type

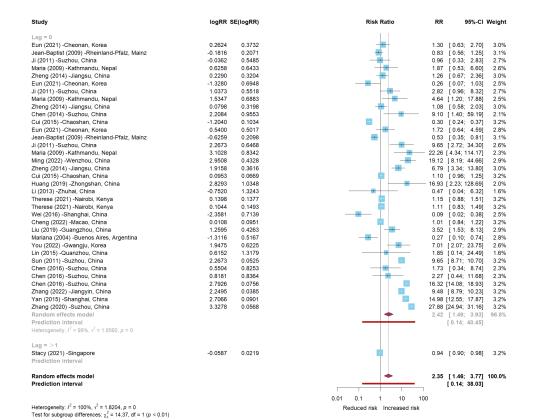


Figure S3-7. Leave-one-out analysis of human parainfluenza virus (HPIV) meta-analysis

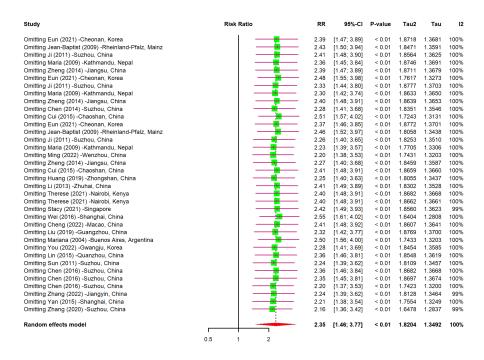
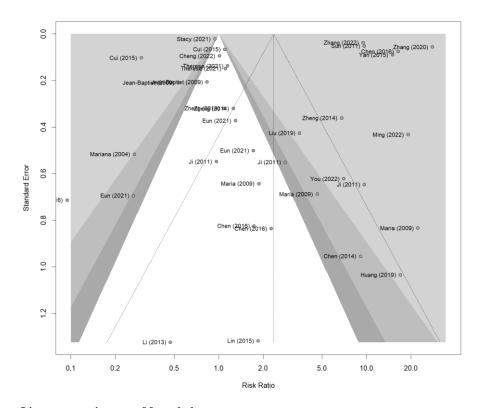


Figure S3-8. Funnel plot of human parainfluenza virus (HPIV) estimates



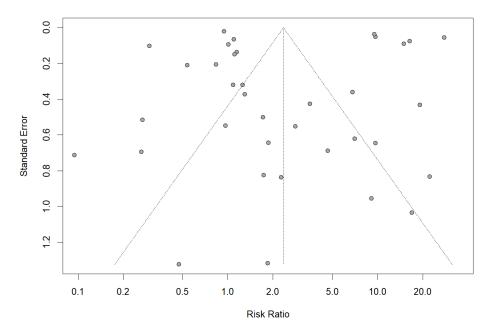
## Linear regression test of funnel plot asymmetry

Egger's test result: t = 0.43, df = 34, p = 0.6721

## Sample estimates:

bias	se.bias	intercept	se.intercept
1.3038	3.0541	0.8533	0.2769

Figure S3-9 Trim and fill of human parainfluenza virus (HPIV) meta-analysis



Number of studies: k = 36 (with 0 added studies)

Random effects model: RR=2.3464 [1.4620-3.7656], z= 3.53, p=0.0004

Quantifying heterogeneity:

 $tau^2 = 1.8204 [1.0997; 3.2263]; tau = 1.3492 [1.0487; 1.7962]$ 

 $I^2 = 99.5\%$  [99.5%; 99.6%]; H = 14.72 [14.06; 15.41]

Figure S4. Random-effects meta-analysis of human metapneumoviruses (HMPV) estimates (14 studies)

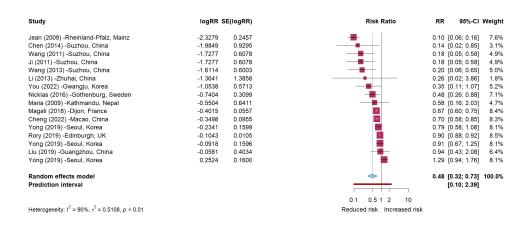


Figure S4-1. Subgroup analysis of human metapneumoviruses (HMPV) meta-analysis by Köppen-Geiger climate

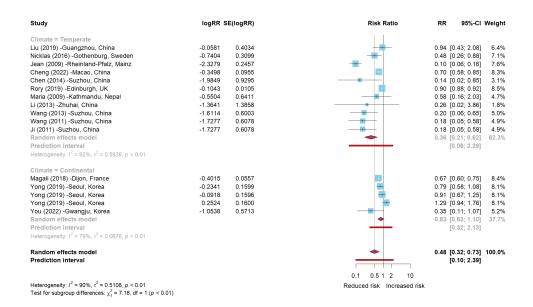


Figure S4-2. Subgroup analysis of human metapneumoviruses (HMPV) meta-analysis by World Bank income category

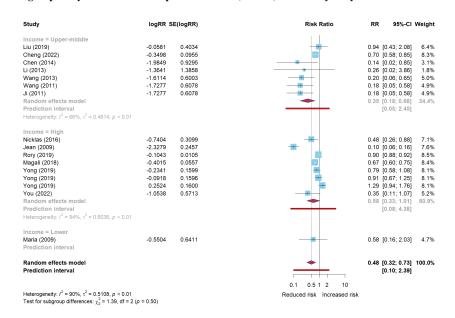


Figure S4-3. Subgroup analysis of human metapneumoviruses (HMPV) meta-analysis by temporal resolution

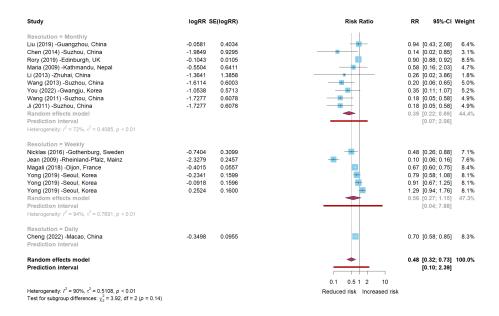


Figure S4-4. Subgroup analysis of human metapneumoviruses (HMPV) meta-analysis by exposure measure

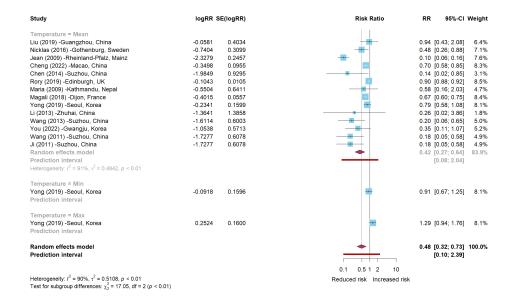


Figure S4-5. Subgroup analysis of human metapneumoviruses (HMPV) meta-analysis by modelling approach

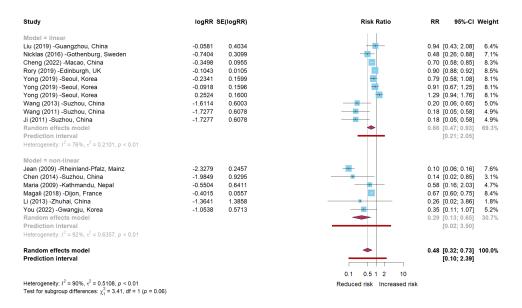


Figure S4-6. Leave-one-out analysis of human metapneumoviruses (HMPV) meta-analysis

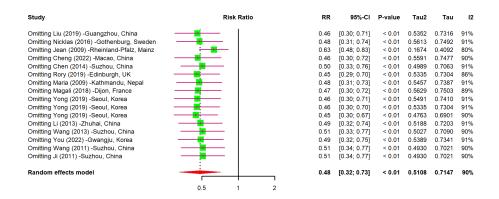
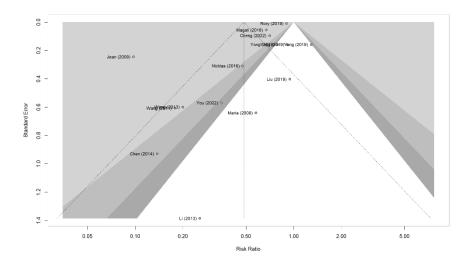


Figure S4-7. Funnel plot of human metapneumoviruses (HMPV) estimates  ${\bf F}$ 

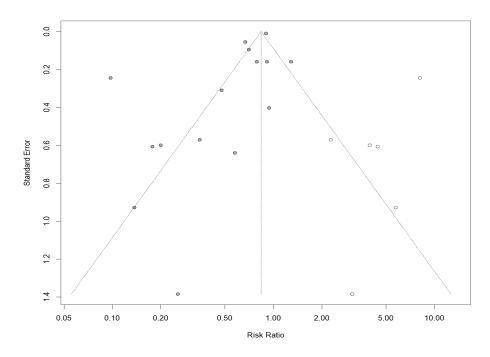


Egger's test result: t = -2.91, df = 14, p-value = 0.0115

Sample estimates:

bias	se.bias	intercept	se.intercept		
-2.0778	0.7151	-0.0876	0.0290		

Figure S4-8. Trim and fill of human metapneumoviruses (HMPV) meta-analysis



Number of studies: k = 23 (with 7 added studies)

Random effects model: RR= 0.8394 [0.5034; 1.3998], Z =-0.67, P= 0.5023

Quantifying heterogeneity:

 $tau^2 = 1.2694$  [0.6562, 2.9522]; tau = 1.1267 [0.8101, 1.7182]

 $I^2 = 91.6\%$  [88.7%, 93.8%]; H = 3.45 [2.97, 4.00]

Figure~S5.~Random-effects~meta-analysis~of~human~rhinovirus~(HRV)~estimates~(12~studies)

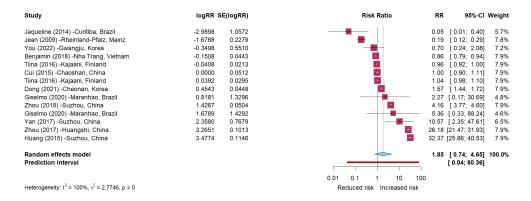


Figure S5-1. Subgroup analysis of human rhinovirus (HRV) meta-analysis by Köppen-Geiger climate

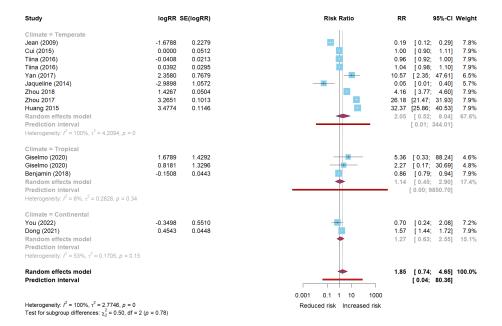


Figure S5-2. Subgroup analysis of human rhinovirus (HRV) meta-analysis by World Bank income category

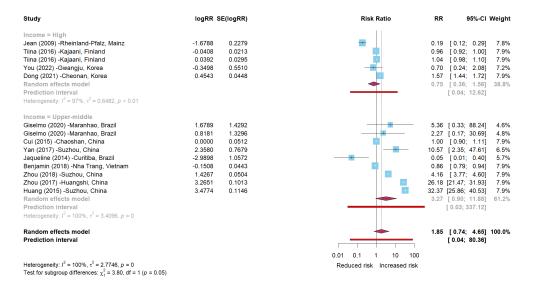


Figure S5-3. Subgroup analysis of human rhinovirus (HRV) meta-analysis by temporal resolution

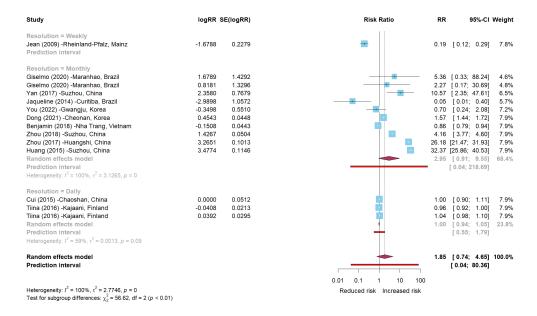


Figure S5-4. Subgroup analysis of human rhinovirus (HRV) meta-analysis by exposure measure

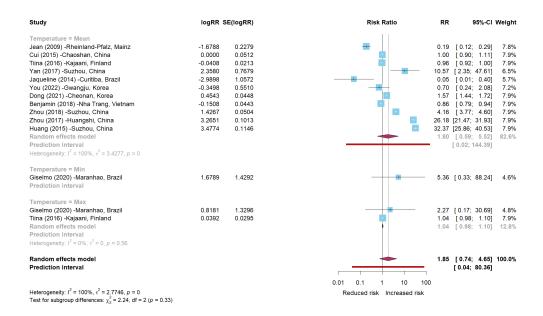
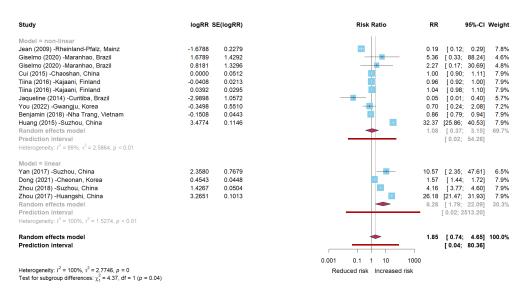


Figure S5-5. Subgroup analysis of human rhinovirus (HRV) meta-analysis by modelling approach



Figure~S5-6.~Leave-one-out~analysis~of~human~rhinovirus~(HRV)~meta-analysis

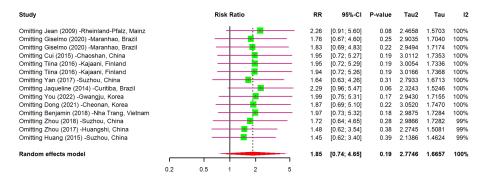
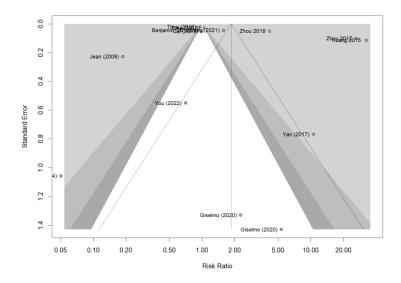
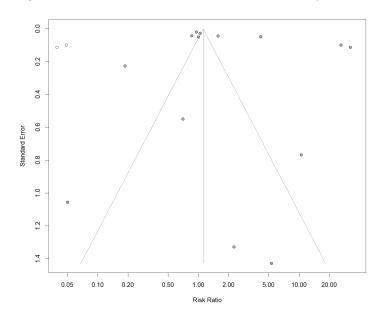


Figure S5-7. Funnel plot of human rhinovirus (HRV) estimates



Egger's test result: t = 1.32, df = 13, p-value = 0.2086

Figure S5-8. Trim and fill of human rhinovirus (HRV) meta-analysis



Number of studies: k = 16 (with 2 added studies)

Random effects model: RR 1.1163 [0.3960; 3.1466], Z=0.21, P= 0.8352

Quantifying heterogeneity:  $tau^2 = 4.1785$  [2.1504; 10.1327]; tau = 2.0441 [1.4664; 3.1832];  $t^2 = 99.7\%$  [99.6%; 99.7%];  $t^2 = 99.7\%$  [1.454]

Figure S6. Random-effects meta-analysis of human adenoviruses (HAdVs) estimates (16 studies)

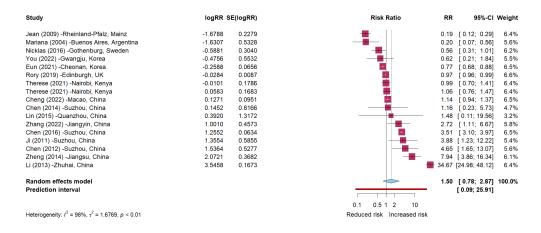


Figure S6-1. Subgroup analysis of human adenoviruses (HAdVs) meta-analysis by Köppen-Geiger climate

Study	logRR S	E(logRR)	Risk Ratio	RR	95%-CI	Weight
Climate = Temperate						
Zheng (2014) -Jiangsu, China	2.0721	0.3682		7.94	[ 3.86; 16.34]	6.1%
Nicklas (2016) -Gothenburg, Sweden	-0.5881	0.3040		0.56	[ 0.31; 1.01]	6.2%
Jean (2009) -Rheinland-Pfalz, Mainz	-1.6788	0.2279		0.19	[ 0.12; 0.29]	6.4%
Cheng (2022) -Macao, China	0.1271	0.0951		1.14	[ 0.94; 1.37]	6.5%
Chen (2014) -Suzhou, China	0.1452	0.8166	<del>-     -   -   -   -   -   -   -   -   -</del>	1.16	[ 0.23; 5.73]	4.7%
Rory (2019) -Edinburgh, UK	-0.0284	0.0087		0.97	[ 0.96; 0.99]	6.6%
Li (2013) -Zhuhai, China	3.5458	0.1673		34.67	[24.98; 48.12]	6.4%
Mariana (2004) -Buenos Aires, Argentina	-1.6307	0.5328		0.20	[ 0.07; 0.56]	5.6%
Ji (2011) -Suzhou, China	1.3554	0.5855	<del>                                    </del>	3.88	[ 1.23; 12.22]	5.4%
Lin (2015) -Quanzhou, China	0.3920	1.3172	<del>-                                      </del>	1.48	[ 0.11; 19.56]	3.2%
Chen (2012) -Suzhou, China	1.5364	0.5277		4.65	[ 1.65; 13.07]	5.6%
Chen (2016) -Suzhou, China	1.2552	0.0634		3.51	[3.10; 3.97]	6.5%
Zhang (2022) -Jiangyin, China	1.0010	0.4573	<del>   </del>	2.72	[ 1.11; 6.67]	5.8%
Random effects model			<del></del>	1.81	[ 0.78; 4.16]	75.1%
Prediction interval				-	[ 0.06; 51.04]	
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 2.1232$ , $p < 0.01$						
Climate = Tropical						
Therese (2021) -Nairobi, Kenya	0.0583	0.1683		1.06	[ 0.76; 1.47]	6.4%
Therese (2021) -Nairobi, Kenya	-0.0101	0.1786	<b>#</b>	0.99	[ 0.70; 1.41]	6.4%
Random effects model			<b>♦</b>	1.03	[ 0.81; 1.31]	12.9%
Prediction interval						
Heterogeneity: $I^2 = 0\%$ , $\tau^2 = 0$ , $p = 0.78$						
Climate = Continental						
You (2022) -Gwangju, Korea	-0.4756	0.5532	<del>- 1</del>	0.62	[ 0.21; 1.84]	5.5%
Eun (2021) -Cheonan, Korea	-0.2588	0.0656		0.77	[ 0.68; 0.88]	6.5%
Random effects model			<b>◆</b>	0.77	[ 0.68; 0.87]	12.1%
Prediction interval						
Heterogeneity: $I^2 = 0\%$ , $\tau^2 = 0$ , $p = 0.70$						
Random effects model				1.50	[ 0.78; 2.87]	100.0%
Prediction interval					[ 0.09; 25.91]	
			0.1 0.5 1 2 10			
Heterogeneity: $I^2 = 98\%$ , $\tau^2 = 1.6769$ , $p < 0.01$ Test for subgroup differences: $\chi^2_2 = 7.68$ , df = 2 ( $p = 0.02$ )			Reduced risk Increased risk			

Figure S6-2. Subgroup analysis of human adenoviruses (HAdVs) meta-analysis by World Bank income category

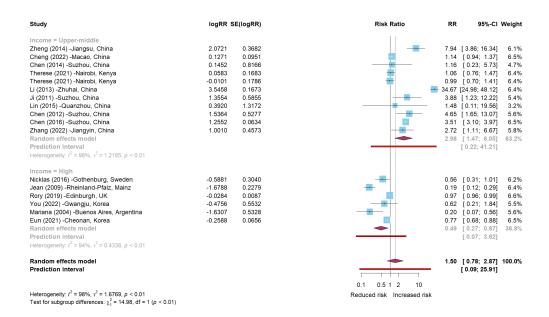
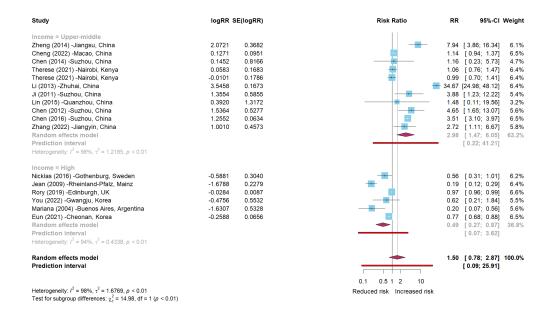
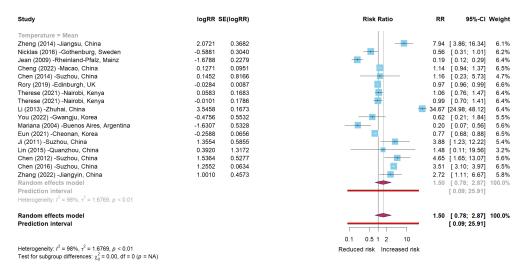


Figure S6-3. Subgroup analysis of human adenoviruses (HAdVs) meta-analysis by temporal resolution



Figure~S6-4.~Subgroup~analysis~of~human~adenoviruses~(HAdVs)~meta-analysis~by~exposure~measure



Figure~S6-5.~Subgroup~analysis~of~human~adenoviruses~(HAdVs)~meta-analysis~by~modelling~approach

Study	logRR S	E(logRR)	Risk Ratio	RR	95%-CI	Weight
Model = linear						
Zheng (2014) -Jiangsu, China	2.0721	0.3682	<del> </del>	7.94	[ 3.86; 16.34]	6.1%
Nicklas (2016) -Gothenburg, Sweden	-0.5881	0.3040		0.56	[ 0.31; 1.01]	6.2%
Cheng (2022) -Macao, China	0.1271	0.0951		1.14	[ 0.94; 1.37]	6.5%
Rory (2019) -Edinburgh, UK	-0.0284	0.0087		0.97	[ 0.96; 0.99]	6.6%
Ji (2011) -Suzhou, China	1.3554	0.5855	<del>                                    </del>	3.88	[ 1.23; 12.22]	5.4%
Lin (2015) -Quanzhou, China	0.3920	1.3172	<del>-   •   • </del>	1.48	[ 0.11; 19.56]	3.2%
Random effects model			<del></del>	1.67	[ 0.71; 3.93]	34.0%
Prediction interval			<del></del>		[ 0.09; 31.29]	
Heterogeneity: $I^2 = 89\%$ , $\tau^2 = 0.9224$ , $p < 0.01$						
Model = non-linear						
Jean (2009) -Rheinland-Pfalz, Mainz	-1.6788	0.2279		0.19	[ 0.12; 0.29]	6.4%
Chen (2014) -Suzhou, China	0.1452	0.8166		1.16	[ 0.23; 5.73]	4.7%
Therese (2021) -Nairobi, Kenya	0.0583	0.1683	# ·	1.06	[ 0.76; 1.47]	6.4%
Therese (2021) -Nairobi, Kenya	-0.0101	0.1786	<del> </del>	0.99	[ 0.70; 1.41]	6.4%
Li (2013) -Zhuhai, China	3.5458	0.1673		34.67	[24.98; 48.12]	6.4%
You (2022) -Gwangju, Korea	-0.4756	0.5532	<del>-    </del>	0.62	[ 0.21; 1.84]	5.5%
Mariana (2004) -Buenos Aires, Argentina	-1.6307	0.5328	<del></del>	0.20	[ 0.07; 0.56]	5.6%
Eun (2021) -Cheonan, Korea	-0.2588	0.0656		0.77	[ 0.68; 0.88]	6.5%
Chen (2012) -Suzhou, China	1.5364	0.5277		4.65	[ 1.65; 13.07]	5.6%
Chen (2016) -Suzhou, China	1.2552	0.0634		3.51	[ 3.10; 3.97]	6.5%
Zhang (2022) -Jiangyin, China	1.0010	0.4573	<del>   </del>	2.72	[ 1.11; 6.67]	5.8%
Random effects model			<del></del>	1.40	[ 0.57; 3.45]	66.0%
Prediction interval				•	[ 0.04; 46.37]	
Heterogeneity: $I^2 = 99\%$ , $\tau^2 = 2.1842$ , $p < 0.01$						
Random effects model				1.50	[ 0.78; 2.87]	100.0%
Prediction interval					[ 0.09; 25.91]	
			0.1 0.5 1 2 10			
Heterogeneity: $I^2 = 98\%$ , $\tau^2 = 1.6769$ , $p < 0.01$ Test for subgroup differences: $\chi^2 = 0.08$ , df = 1 ( $p = 0.78$ )			Reduced risk Increased risk			

Figure~S6-6.~Leave-one-out~analysis~of~human~adenoviruses~(HAdVs)~meta-analysis

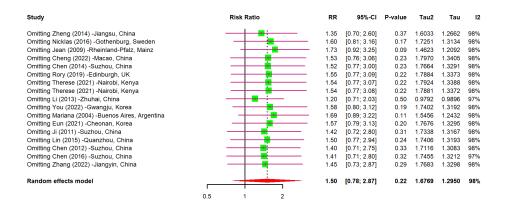
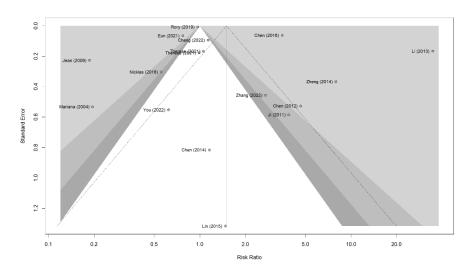


Figure S6-7. Funnel plot of human adenoviruses (HAdVs) estimates

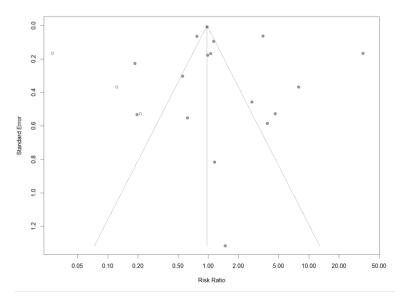


Egger's test result: t = 1.22, df = 15, p-value = 0.2409

Sample estimates:

bias	se.bias	intercept	se.intercept		
2.5030	2.0500	-0.0349	0.0713		

Figure S6-8. Trim and fill of human adenoviruses (HAdVs) meta-analysis



Number of studies: k = 20 (with 3 added studies)

Random effects model: RR=0.9707 [0.4667; 2.0189], Z=-0.08, P=0.9365

Quantifying heterogeneity:

 $tau^2 = 2.5968 [1.4069; 5.4493]; tau = 1.6114 [1.1861; 2.3344]$ 

 $I^2 = 98.7\%$  [98.5%; 98.9%]; H = 8.83 [8.09; 9.65]

Figure S7. Random-effects meta-analysis of human bocavirus (HBoV) estimates (nine studies)

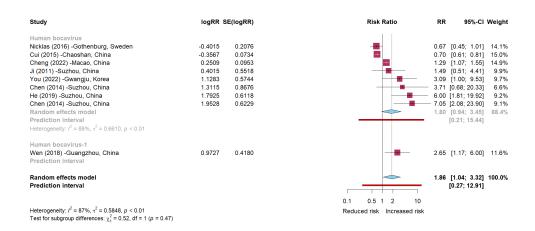


Figure S7-1. Subgroup analysis of human bocavirus (HBoV) meta-analysis by Köppen-Geiger climate

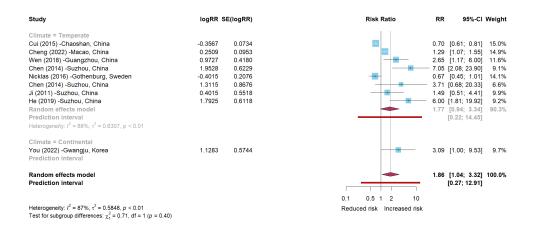


Figure S7-2. Subgroup analysis of human bocavirus (HBoV) meta-analysis by World Bank income category

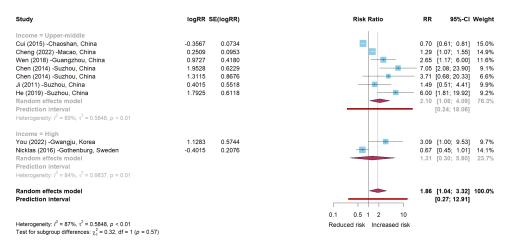


Figure S7-3. Subgroup analysis of human bocavirus (HBoV) meta-analysis by temporal resolution

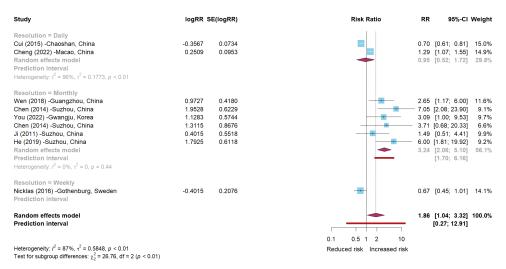


Figure S7-4. Subgroup analysis of human bocavirus (HBoV) meta-analysis by exposure measure

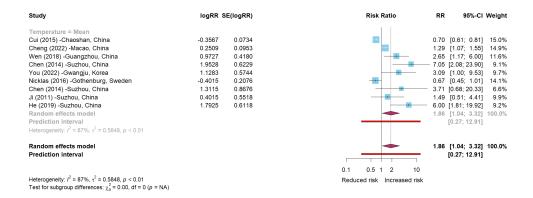


Figure S7-5. Subgroup analysis of human bocavirus (HBoV) meta-analysis by modelling approach

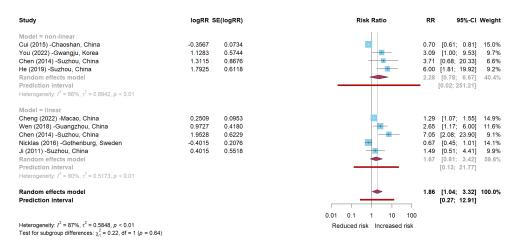


Figure S7-6. Leave-one-out analysis of human bocavirus (HBoV) meta-analysis

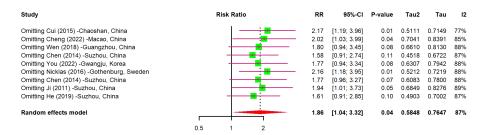


Figure S8. Random-effects meta-analysis of enterovirus(EV) estimates (four studies)

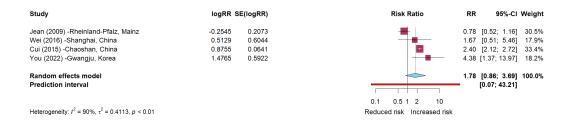


Figure S8-1. Subgroup analysis of enterovirus (EV) meta-analysis by Köppen-Geiger climate

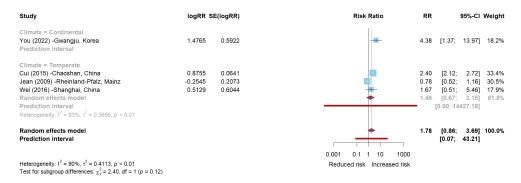


Figure S8-2. Subgroup analysis of enterovirus (EV) meta-analysis by World Bank income category

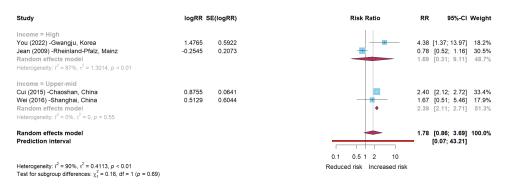


Figure S8-3. Subgroup analysis of enterovirus (EV) meta-analysis by temporal resolution

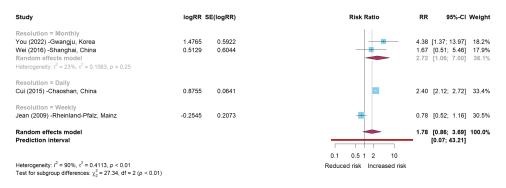


Figure S8-4. Subgroup analysis of enterovirus (EV) meta-analysis by exposure measure

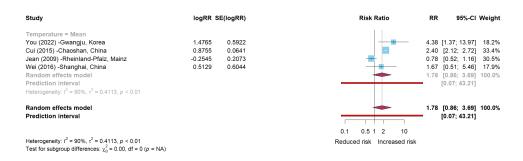


Figure S8-5. Subgroup analysis of enterovirus (EV) meta-analysis by modelling approach

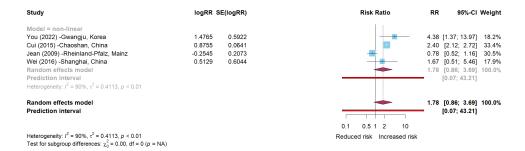


Figure S8-6. Leave-one-out analysis of enterovirus (EV) meta-analysis

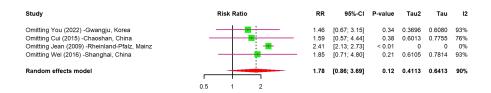


Figure S8-7. Sensitivity analysis by excluding study with high risk of bias for Jean et al 2009

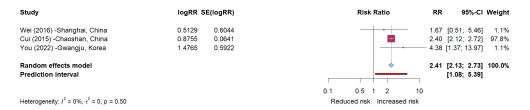


Figure S9. Random-effects meta-analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) estimates (nine studies)

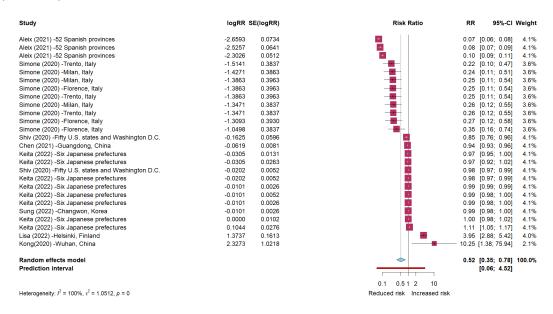


Figure S9-1. Subgroup analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis by Köppen-Geiger climate

Study	logRR SE	(logRR)	Risk Ratio	RR	95%-CI	Weight
Climate = Multiple						
Shiv (2020) -Fifty U.S. states and Washington D.C.	-0.1625	0.0596	<u> </u>	0.85	[0.76; 0.96]	4.1%
Shiv (2020) -Fifty U.S. states and Washington D.C.	-0.0202	0.0052		0.98	[0.97; 0.99]	4.1%
Random effects model			•	0.92	[0.81; 1.06]	8.2%
Prediction interval						
Heterogeneity: $I^2 = 82\%$ , $\tau^2 = 0.0083$ , $p = 0.02$						
Climate = Temperate						
Aleix (2021) -52 Spanish provinces	-2.5257	0.0641	<u> </u>	0.08	[0.07; 0.09]	4.1%
Aleix (2021) -52 Spanish provinces	-2.6593	0.0734	-	0.07	[0.06; 0.08]	4.1%
Aleix (2021) -52 Spanish provinces	-2.3026	0.0512		0.10	[0.09; 0.11]	4.1%
Keita (2022) -Six Japanese prefectures	0.0000	0.0102		1.00	[0.98; 1.02]	4.1%
Keita (2022) -Six Japanese prefectures	-0.0101	0.0026		0.99	[0.99; 0.99]	4.1%
Keita (2022) -Six Japanese prefectures	-0.0101	0.0052		0.99	[0.98; 1.00]	4.1%
Keita (2022) -Six Japanese prefectures	-0.0202	0.0052		0.98	[0.97; 0.99]	4.1%
Keita (2022) -Six Japanese prefectures	0.1044	0.0276		1.11	[1.05; 1.17]	4.1%
Keita (2022) -Six Japanese prefectures	-0.0101	0.0026		0.99	[0.98; 1.00]	4.1%
Keita (2022) -Six Japanese prefectures	-0.0305	0.0131		0.97	[0.95; 1.00]	4.1%
Keita (2022) -Six Japanese prefectures	-0.0305	0.0263		0.97	[0.92; 1.02]	4.1%
Chen (2021) -Guangdong, China	-0.0619	0.0081		0.94	[0.93; 0.96]	4.1%
Sung (2022) -Changwon, Korea	-0.0101	0.0026		0.99	[0.98; 1.00]	4.1%
Lisa (2022) -Helsinki, Finland	1.3737	0.1613	-	3.95	[2.88; 5.42]	4.0%
Kong(2020) -Wuhan, China	2.3273	1.0218	<del>   </del>	- 10.25	[1.38; 75.94]	2.1%
Simone (2020) -Florence, Italy	-1.0498	0.3837	-	0.35	[0.16; 0.74]	3.6%
Simone (2020) -Florence, Italy	-1.3093	0.3930		0.27	[0.12; 0.58]	3.6%
Simone (2020) -Florence, Italy	-1.3863	0.3963			[0.11; 0.54]	3.6%
Simone (2020) -Trento, Italy	-1.5141	0.3837	<del></del>	0.22	[0.10; 0.47]	3.6%
Simone (2020) -Trento, Italy	-1.3863	0.3963		0.25	[0.11; 0.54]	3.6%
Simone (2020) -Trento, Italy	-1.3471	0.3837	-	0.26	[0.12; 0.55]	3.6%
Random effects model			<b>→</b>	0.54	[0.33; 0.89]	81.0%
Prediction interval					[0.05; 5.86]	
Heterogeneity: $I^2 = 100\%$ , $\tau^2 = 1.2263$ , $p = 0$						
Climate = Continental						
Simone (2020) -Milan, Italy	-1.3471	0.3837		0.26	[0.12; 0.55]	3.6%
Simone (2020) -Milan, Italy	-1.3863	0.3963	-	0.25	[0.11; 0.54]	3.6%
Simone (2020) -Milan, Italy	-1.4271	0.3863	-	0.24	[0.11; 0.51]	3.6%
Random effects model			<b>*</b>	0.25	[0.16; 0.39]	10.8%
Prediction interval			<del></del>		[0.01; 4.32]	
Heterogeneity: $I^2 = 0\%$ , $\tau^2 = 0$ , $p = 0.99$						
Random effects model			<b>↓</b>	0.52	[0.35; 0.78]	100.0%
Prediction interval					[0.06; 4.52]	
			0.1 0.5 1 2 10			
Heterogeneity: $l^2 = 100\%$ , $\tau^2 = 1.0512$ , $p = 0$ Test for subgroup differences: $\chi^2_2 = 33.51$ , df = 2 ( $p < 0.01$ )			Reduced risk Increased risk			
***						

Figure S9-2. Subgroup analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis by World Bank income category

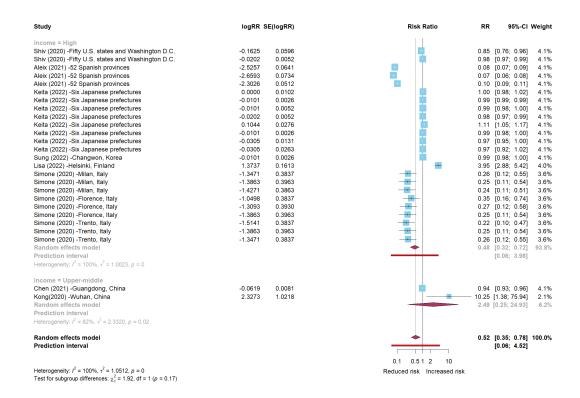


Figure S9-3. Subgroup analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis by temporal resolution

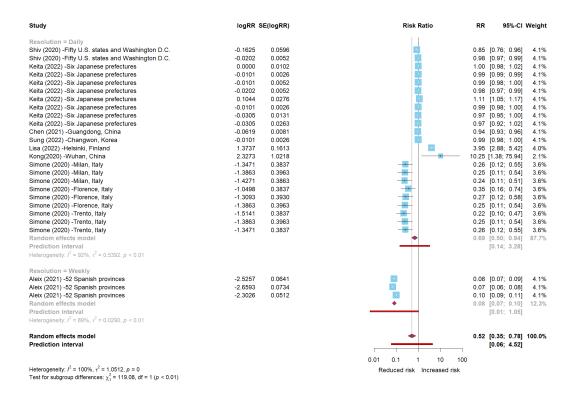


Figure S9-4. Subgroup analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis by exposure measure

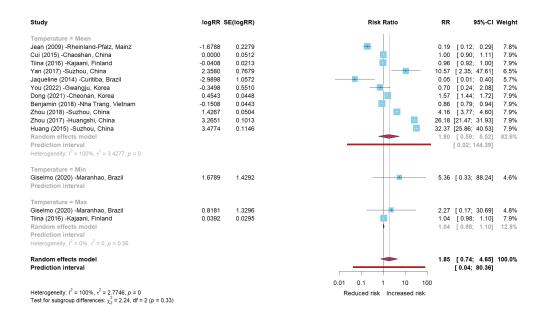


Figure S9-5. Subgroup analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis by modelling approach

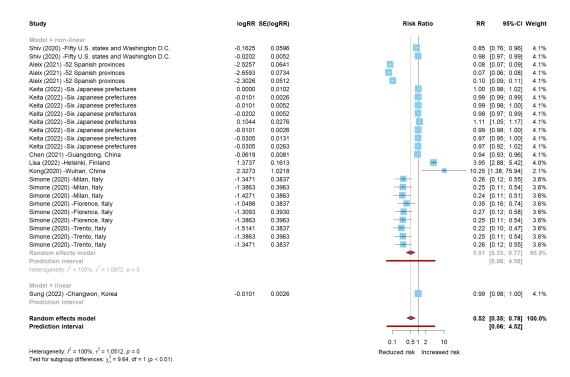


Figure S9-6. Subgroup analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis by lag type

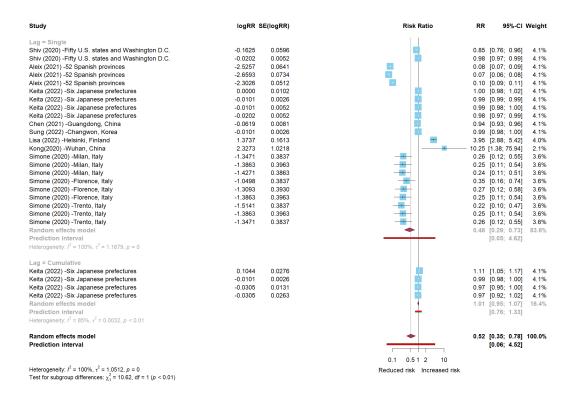


Figure S9-7. Leave-one-out analysis of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) meta-analysis

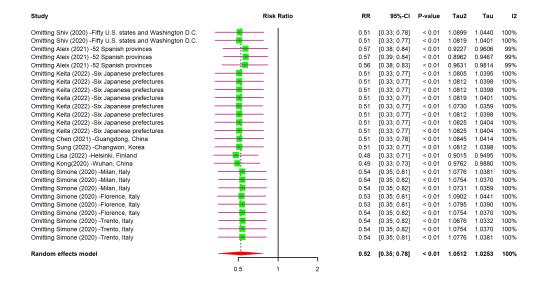


Figure S10. Random-effects meta-analysis of severe acute respiratory syndrome coronavirus (SARS-CoV) estimates (four studies)

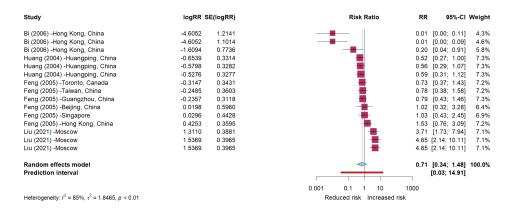


Figure S10-1. Subgroup analysis of severe acute respiratory syndrome (SARS) meta-analysis by Köppen-Geiger climate

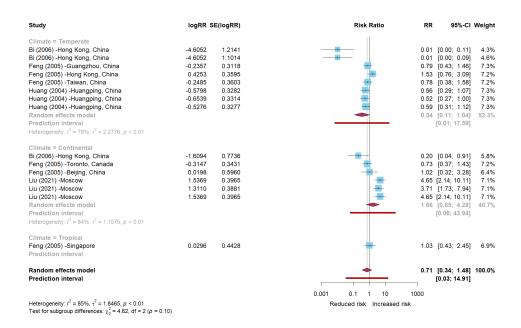


Figure S10-2. Subgroup analysis of severe acute respiratory syndrome coronavirus (SARS-CoV) meta-analysis by World Bank income category

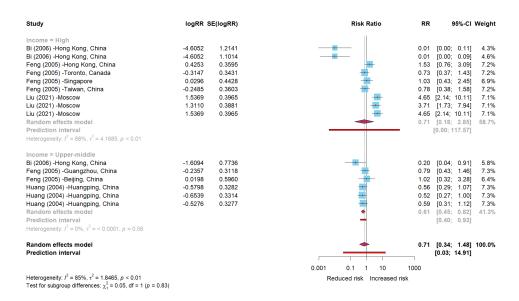


Figure S10-3. Subgroup analysis of severe acute respiratory syndrome coronavirus (SARS-CoV) meta-analysis by temporal resolution

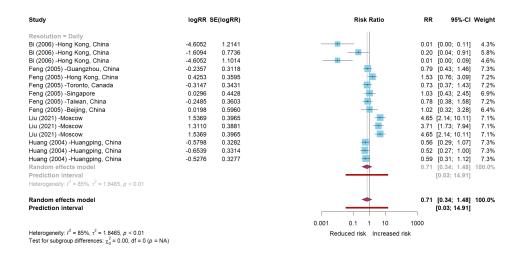


Figure S10-4. Subgroup analysis of severe acute respiratory syndrome coronavirus (SARS-CoV) meta-analysis by exposure measure

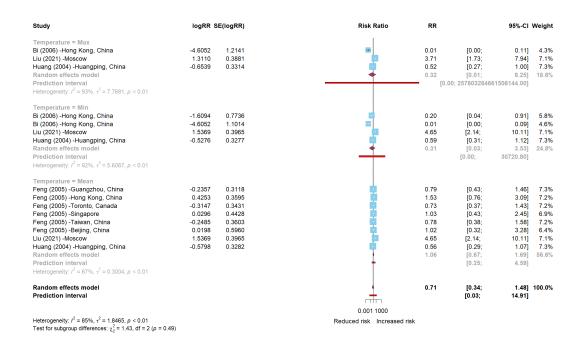


Figure S10-5. Subgroup analysis of severe acute respiratory syndrome coronavirus (SARS-CoV) meta-analysis by modelling approach

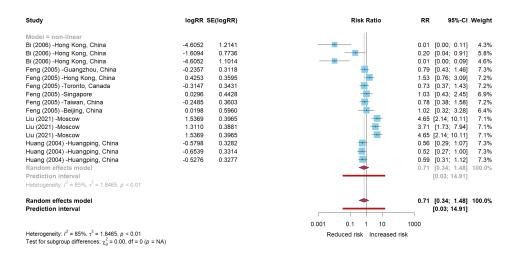


Figure S10-6. Leave-one-out analysis of severe acute respiratory syndrome coronavirus (SARS-CoV) meta-analysis

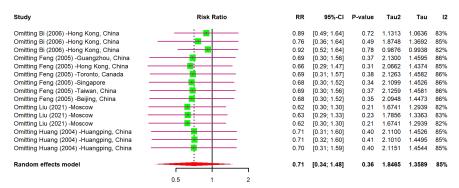


Figure S11. Random-effects meta-analysis of human coronavirus (HCoV) estimates (five studies)

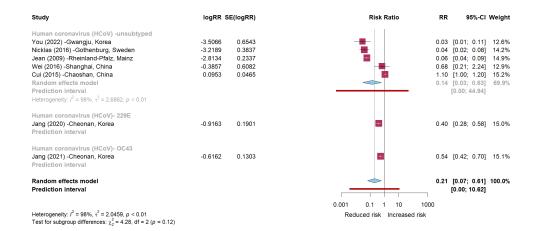


Figure S11-1. Subgroup analysis of human coronavirus (HCoV) meta-analysis by Köppen-Geiger climate

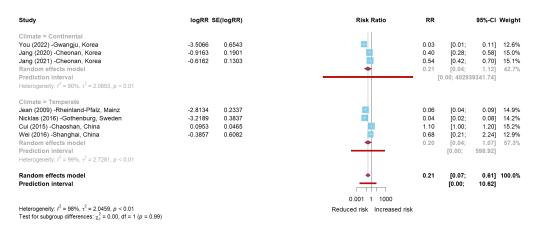
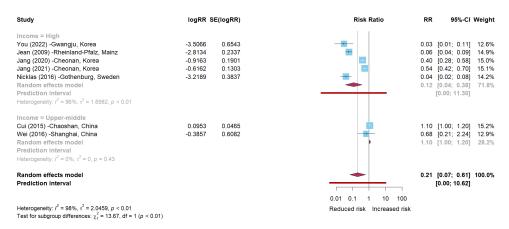


Figure S11-2. Subgroup analysis of human coronavirus (HCoV) meta-analysis by World Bank income category



 $Figure\ S11-3.\ Subgroup\ analysis\ of\ human\ coronavirus\ (HCoV)\ meta-analysis\ by\ temporal\ resolution$ 

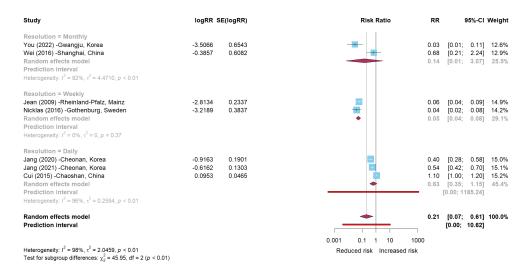


Figure S11-4. Subgroup analysis of human coronavirus (HCoV) meta-analysis by exposure measure

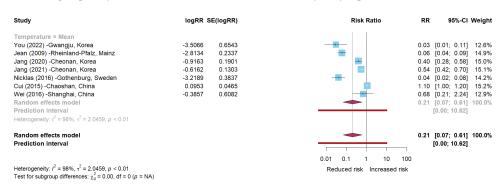


Figure S11-5. Subgroup analysis of human coronavirus (HCoV) meta-analysis by modelling approach

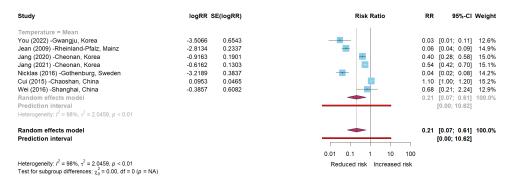


Figure S11-6. Leave-one-out analysis of human coronavirus (HCoV) meta-analysis

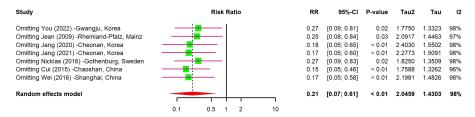


Figure S12. Random-effects meta-analysis of Human rhinoviruses/enteroviruses (HRVs/EVs) estimates (two studies)



Figure S12-1. Subgroup analysis of Human rhinoviruses/enteroviruses (HRVs/EVs) meta-analysis by Köppen-Geiger climate



Figure S12-2. Subgroup analysis of Human rhinoviruses/enteroviruses (HRVs/EVs) meta-analysis by World Bank income category



Figure S12-3. Subgroup analysis of Human rhinoviruses/enteroviruses (HRVs/EVs) meta-analysis by temporal resolution



Figure S12-4. Subgroup analysis of Human rhinoviruses/enteroviruses (HRVs/EVs) meta-analysis by exposure measure

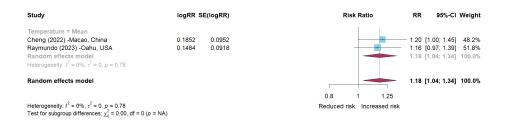


Figure S12-5. Subgroup analysis of Human rhinoviruses/enteroviruses (HRVs/EVs) meta-analysis by modelling approach



Figure S13. Subgroup analysis of bacterial respiratory infections meta-analysis by pathogen

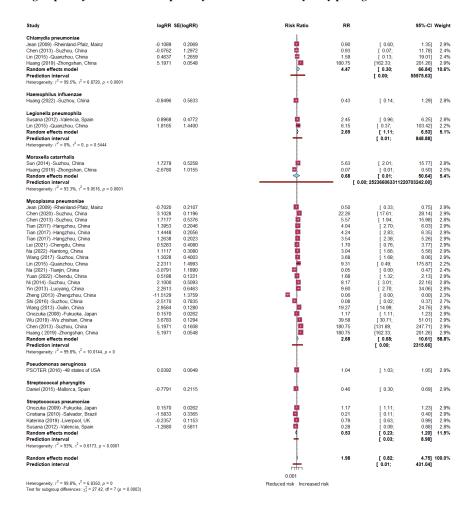


Figure S14. Random-effects meta-analysis of Streptococcus pneumoniae estimates (four studies)

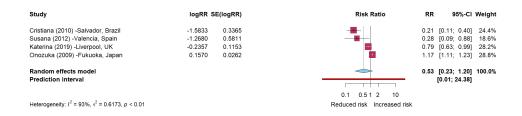


Figure S14-1. Subgroup analysis of Streptococcus pneumoniae meta-analysis by Köppen-Geiger climate

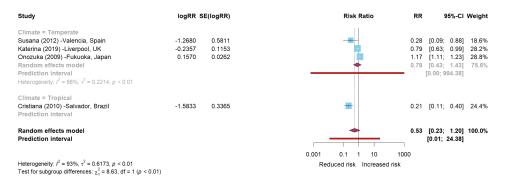


Figure S14-2. Subgroup analysis of Streptococcus pneumoniae meta-analysis by World Bank income category

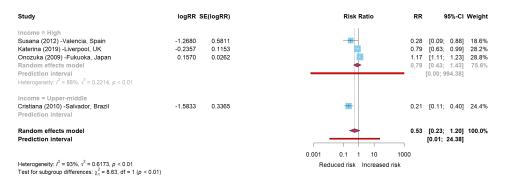


Figure S14-3. Subgroup analysis of Streptococcus pneumoniae meta-analysis by temporal resolution

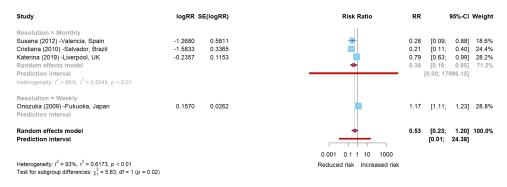


Figure S14-4. Subgroup analysis of Streptococcus pneumoniae meta-analysis by exposure measure

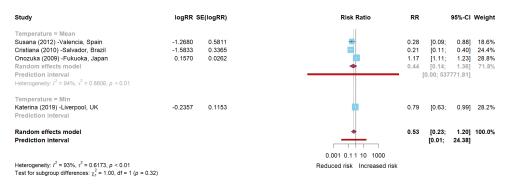


Figure S14-5. Subgroup analysis of Streptococcus pneumoniae meta-analysis by modelling approach

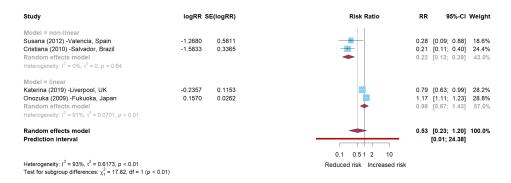


Figure S14-6. Leave-one-out analysis of Streptococcus pneumoniae meta-analysis

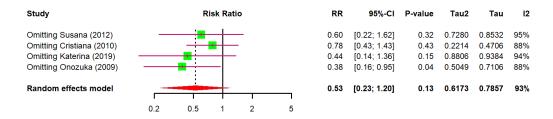


Figure S15. Random-effects meta-analysis of Mycoplasma pneumoniae estimates (19 studies)

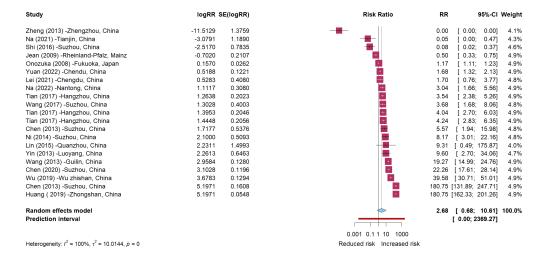
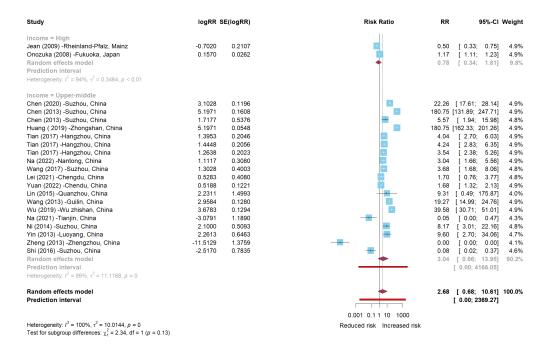


Figure S15-1. Subgroup analysis of Mycoplasma pneumoniae meta-analysis by Köppen-Geiger climate

Study	logRR S	E(logRR)	Risk Ratio	RR		95%-CI	Weight
Climate = Temperate							
Jean (2009) -Rheinland-Pfalz, Mainz	-0.7020	0.2107		0.50	[ 0.33;	0.75]	4.9%
Chen (2020) -Suzhou, China	3.1028	0.1196	F	22.26	[ 17.61;	28.14]	4.9%
Chen (2013) -Suzhou, China	5.1971	0.1608	+	180.75	[131.89;	247.71]	4.9%
Chen (2013) -Suzhou, China	1.7177	0.5376		5.57	[ 1.94;	15.98]	4.8%
Huang ( 2019) -Zhongshan, China	5.1971	0.0548		180.75	[162.33;	201.26]	4.9%
Tian (2017) -Hangzhou, China	1.3953	0.2046		4.04	[ 2.70;	6.03]	4.9%
Tian (2017) -Hangzhou, China	1.4448	0.2056		4.24	[ 2.83;	6.35]	4.9%
Tian (2017) -Hangzhou, China	1.2638	0.2023		3.54	[ 2.38;	5.26]	4.9%
Na (2022) -Nantong, China	1.1117	0.3080		3.04	[ 1.66;	5.56]	4.9%
Wang (2017) -Suzhou, China	1.3028	0.4003	**************************************	3.68	[ 1.68;	8.06]	4.8%
Lei (2021) -Chengdu, China	0.5283	0.4080		1.70	[ 0.76;	3.77]	4.8%
Onozuka (2008) -Fukuoka, Japan	0.1570	0.0262		1.17	[ 1.11;	1.23]	4.9%
Yuan (2022) -Chendu, China	0.5188	0.1221		1.68	[ 1.32;	2.13]	4.9%
Lin (2015) -Quanzhou, China	2.2311	1.4993	<del></del>	9.31	[ 0.49;	175.87]	4.0%
Wang (2013) -Guilin, China	2.9584	0.1280	•	19.27	[ 14.99;	24.76]	4.9%
Ni (2014) -Suzhou, China	2.1000	0.5093	140 2	8.17	[ 3.01;	22.16]	4.8%
Random effects model			•	6.32	[ 2.74;	14.57]	77.3%
Prediction interval			+		[ 0.16;	248.64]	
Heterogeneity: $I^2 = 100\%$ , $\tau^2 = 2.7508$ , $p = 0$							
Climate = Tropical							
Wu (2019) -Wu zhishan, China	3.6783	0.1294	+	39.58	[ 30.71;	51.01]	4.9%
Prediction interval							
Climate = Continental							
Na (2021) -Tianjin, China	-3.0791	1.1890	<del></del>	0.05	[ 0.00;	0.47]	4.3%
Yin (2013) -Luoyang, China	2.2613	0.6463		9.60	[ 2.70;	34.06]	4.7%
Zheng (2013) -Zhengzhou, China	-11.5129	1.3759	<del></del>	0.00	[ 0.00;	0.00]	4.1%
Shi (2016) -Suzhou, China	-2.5170	0.7835		0.08	[ 0.02;	0.37]	4.6%
Random effects model				0.03	[ 0.00;	6.94]	17.8%
Prediction interval				[	0.00; 136365	14406.88]	
Heterogeneity: $I^2 = 97\%$ , $\tau^2 = 31.2086$ , $p < 0.01$							
Random effects model			<b>+</b>	2.68	[ 0.68;	10.61]	100.0%
Prediction interval			<del> </del>		[ 0.00;	2369.27]	
			0.001 1 1000				
Heterogeneity: $I^2 = 100\%$ , $\tau^2 = 10.0144$ , $p = 0$			Reduced risk Increased risk				
Test for subgroup differences: $\chi_2^2 = 23.30$ , df = 2 ( $p < 0.01$ )							

Figure S15-2. Subgroup analysis of Mycoplasma pneumoniae meta-analysis by World Bank income category



 $Figure \ S15-3. \ Subgroup \ analysis \ of \ \textit{Mycoplasma pneumoniae} \ meta-analysis \ by \ temporal \ resolution$ 

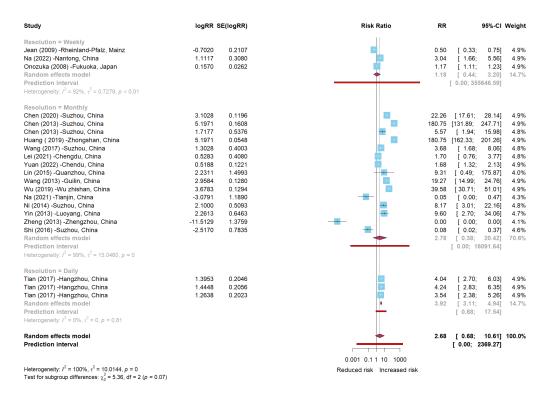


Figure S15-4. Subgroup analysis of Mycoplasma pneumoniae meta-analysis by exposure measure

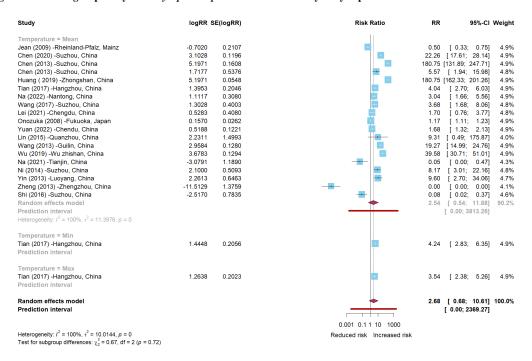


Figure S15-5. Subgroup analysis of Mycoplasma pneumoniae meta-analysis by modelling approach

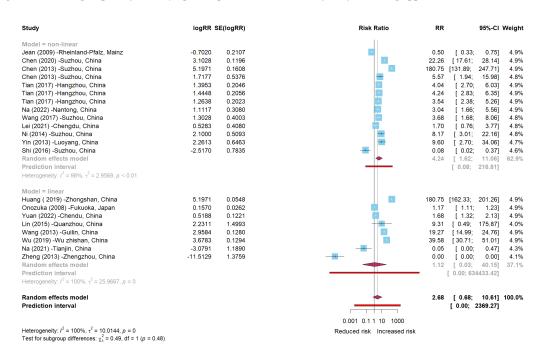


Figure S15-6. Leave-one-out analysis of Mycoplasma pneumoniae meta-analysis

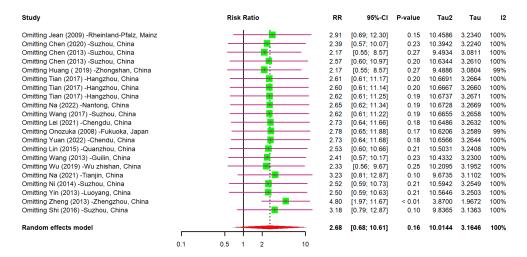
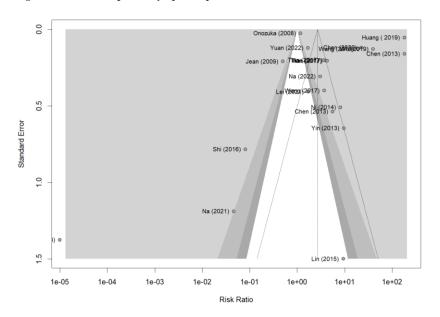


Figure S15-7. Funnel plot of Mycoplasma pneumoniae estimates

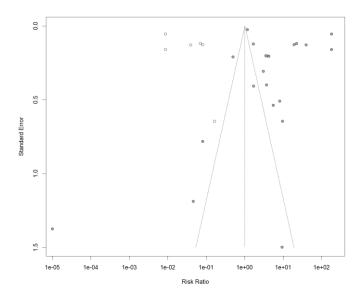


Test result: t = 0.86, df = 19, p-value = 0.4024

Sample estimates:

bias	se.bias	intercept	se.intercept
4.8989	5.7192	1.0285	0.5566

Figure S15-8. Trim and fill of Mycoplasma pneumoniae meta-analysis



Number of studies: k = 27 (with 6 added studies)

Random effects model: RR= 1.0086 [0.2755; 3.6929], Z=0.01, P=0.9896

Quantifying heterogeneity:  $tau^2 = 11.5561$  [7.1899; 23.2580]; tau = 3.3994 [2.6814; 4.8227],  $I^2 = 99.9\%$ ; H = 29.17

Figure S16. Random-effects meta-analysis of Chlamydia pneumoniae estimates (four studies)

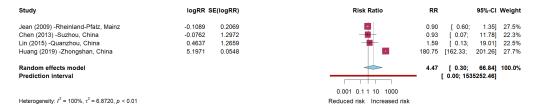


Figure S16-1. Subgroup analysis of Chlamydia pneumoniae meta-analysis by Köppen-Geiger climate

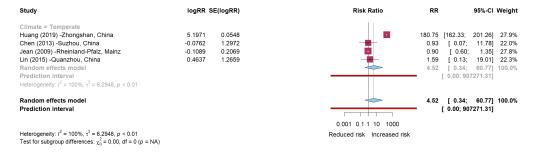


Figure S16-2. Subgroup analysis of Chlamydia pneumonia meta-analysis by World Bank income category

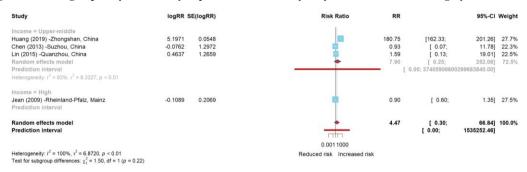


Figure S16-3. Subgroup analysis of Chlamydia pneumoniae meta-analysis by temporal resolution

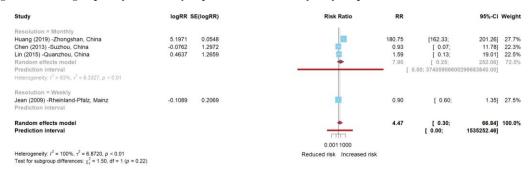


Figure S16-4. Subgroup analysis of Chlamydia pneumoniae meta-analysis by exposure measure

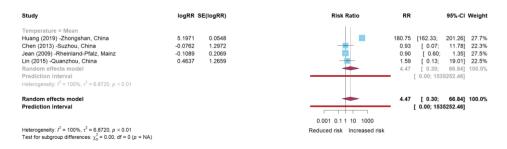


Figure S16-5. Subgroup analysis of Chlamydia pneumoniae meta-analysis by modelling approach

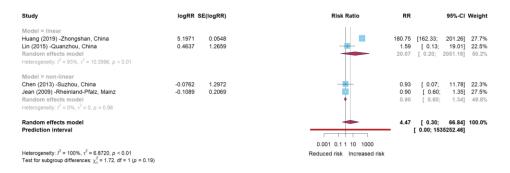


Figure S16-6. Leave-one-out analysis of Chlamydia pneumoniae meta-analysis

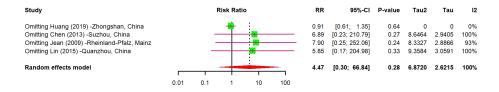


Figure S16-7. Sensitivity analysis by excluding study with high risk of bias for Huang et al 2019

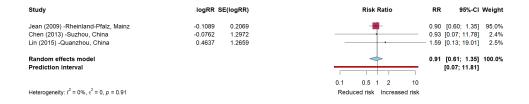


Table X1. Estimates of individual effects from 137 included studies

Study	Location	Cases	Pathogen	RR	lower	upper
Santiago (2004)	Leon, Spain	221	Respiratory syncytial virus -unsubtyped	0.4588	0.2859	0.7274
Santiago (2004)	Leon, Spain	221	Respiratory syncytial virus -unsubtyped	0.5767	0.362	0.9113
Santiago (2004)	Leon, Spain	221	Respiratory syncytial virus -unsubtyped	0.5149	0.3221	0.8159
Zhang (2013)	Suzhou, China	42664	Respiratory syncytial virus -unsubtyped	0.004	0.001	0.011
Liu (2019)	Guangzhou, China	11398	Respiratory syncytial virus -unsubtyped	14.16	5.66	39.99
Natalia (2016)	Athens, Greece	7516	Respiratory syncytial virus -unsubtyped	0.76	0.74	0.79
Natalia (2016)	Athens, Greece	7516	Respiratory syncytial virus -unsubtyped	0.72	0.68	0.8
Silvia (2013)	Bologna, Italy	327	Respiratory syncytial virus -unsubtyped	0.3196	0.1958	0.5348
Omer (2008)	Lombok, Indonesia	2878	Respiratory syncytial virus -unsubtyped	1.44	1.24	1.66
Omer (2008)	Lombok, Indonesia	2878	Respiratory syncytial virus -unsubtyped	1.39	1.22	1.5
Virginia (2015)	Nine states, United States	Not reported	Respiratory syncytial virus -unsubtyped	0.0078	0.0036	0.0159
Patrick (2015)	Kuala Lumpur, Malaysia	5691	Respiratory syncytial virus -unsubtyped	0.2253	0.127	0.3878
Patrick (2015)	Kuala Lumpur, Malaysia	2561	Respiratory syncytial virus -unsubtyped	0.2937	0.1682	0.5002
Patrick (2015)	Kuala Lumpur, Malaysia	2959	Respiratory syncytial virus -unsubtyped	0.2151	0.1209	0.3711
Hailin (2019)	Wenzhou, China	89898	Respiratory syncytial virus -unsubtyped	0.0327	0.0125	0.0764
Zheng (2014)	Jiangsu, China	42104	Respiratory syncytial virus -unsubtyped	0.0039	0.0011	0.0113
Ilada (2020)	Bangkok, Thailand	8209	Respiratory syncytial virus -unsubtyped	0.6215	0.2767	1.3652
Nicklas (2016)	Gothenburg, Sweden	20062	Respiratory syncytial virus -unsubtyped	0.0012	0.0003	0.0039
Jean-Baptist (2009)	Rheinland-Pfalz, Mainz	3044	Respiratory syncytial virus -unsubtyped	0.0144	0.0075	0.0261
Noyola (2009)	San Luis Potosi, Mexico	1393	Respiratory syncytial virus -unsubtyped	0.32	0.04	1.3
Noyola (2009)	San Luis Potosi, Mexico	1393	Respiratory syncytial virus -unsubtyped	2.55	1.34	5.63
Rodriguez (2015)	Bogota, Colombia	3931	Respiratory syncytial virus -unsubtyped	3.14	1.56	6.31
Zhi (2021)	Suzhou, China	1157	Respiratory syncytial virus -A	0.0726	0.0131	0.2951
Zhi (2021)	Suzhou, China	1157	Respiratory syncytial virus -B	0.0792	0.0146	0.318
Gamba (2016)	Bogota, Colombia	3931	Respiratory syncytial virus -unsubtyped	2.25	1.11	4.53
Cheng (2022)	Macao, China	4880	Respiratory syncytial virus -A	0.7982	0.6621	0.9618

Cheng (2022)	Macao, China	4880	Respiratory syncytial virus -B	1.2346	1.0247	1.4884
Chee (2023)	Kuala Lumpur, Malaysia	2950	Respiratory syncytial virus -unsubtyped	1.1605	1.022	3.0255
Chen (2014)	Suzhou, China	998	Respiratory syncytial virus -unsubtyped	0.0002	0.0001	0.0054
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	1.93	1.08	3.46
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	0.87	0.77	0.98
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	1.14	1.03	1.26
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	1.06	1.02	1.09
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	0.93	0.87	0.98
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	0.4395	0.3896	0.4955
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	0.4413	0.3911	0.4974
Liu (2021)	Chongqing, China	3107	Respiratory syncytial virus -unsubtyped	0.4395	0.3896	0.4955
Giselmo (2020)	Maranhao, Brazil	151	Respiratory syncytial virus -unsubtyped	0.5348	0.0353	6.145
Giselmo (2020)	Maranhao, Brazil	151	Respiratory syncytial virus -unsubtyped	0.0109	0.0001	0.2325
Rory (2019)	Edinburgh, UK	52060	Respiratory syncytial virus -unsubtyped	0.827	0.808	0.847
Tian (2017)	Hangzhou, China	36500	Respiratory syncytial virus -unsubtyped	0.0074	0.0038	0.0136
Tian (2017)	Hangzhou, China	36500	Respiratory syncytial virus -unsubtyped	0.0122	0.0066	0.0215
Tian (2017)	Hangzhou, China	36500	Respiratory syncytial virus -unsubtyped	0.0288	0.0168	0.0478
Cui (2015)	Chaoshan, China	1074	Respiratory syncytial virus -unsubtyped	0.51	0.5	0.6
Lu (2015)	Suzhou, China	1803	Respiratory syncytial virus -unsubtyped	0.0054	0.0009	0.0241
Cristiana (2010)	Salvador, Brazil	184	Respiratory syncytial virus -unsubtyped	0.2053	0.1042	0.3873
Maria (2009)	Kathmandu, Nepal	887	Respiratory syncytial virus -unsubtyped	0.81	0.2305	2.7781
Magali (2018)	Dijon, France	4300	Respiratory syncytial virus -unsubtyped	0.5987	0.5365	0.6679
Magali (2018)	Dijon, France	4300	Respiratory syncytial virus -unsubtyped	0.7205	0.646	0.8033
Therese (2021)	Nairobi, Kenya	17261	Respiratory syncytial virus -unsubtyped	1.68	1.03	2.76
Therese (2021)	Nairobi, Kenya	17261	Respiratory syncytial virus -unsubtyped	1.61	0.95	2.73
Li (2013)	Zhuhai, China	924	Respiratory syncytial virus -unsubtyped	0.0288	0.0004	0.4608
Daniel (2012)	Mallorca, Spain	2384	Respiratory syncytial virus -unsubtyped	0.0109	0.0039	0.0266
Daniel (2012)	Mallorca, Spain	2384	Respiratory syncytial virus -unsubtyped	0.0126	0.0046	0.0302

Daniel (2012)	Mallorca, Spain	2384	Respiratory syncytial virus -unsubtyped	0.0067	0.0023	0.0173
Daniel (2012)	Mallorca, Spain	2384	Respiratory syncytial virus -unsubtyped	0.0003	0.0002	0.0007
Daniel (2012)	Mallorca, Spain	2384	Respiratory syncytial virus -unsubtyped	0.0006	0.0003	0.0011
Daniel (2012)	Mallorca, Spain	2384	Respiratory syncytial virus -unsubtyped	0.0002	0.0001	0.0004
Asma (2008)	Doha, Qatar	3121	Respiratory syncytial virus -unsubtyped	0.0188	0.0045	0.0622
Raffaella (2017)	Rome, Italy	723	Respiratory syncytial virus -unsubtyped	0.1527	0.0733	0.3008
Sheikh (2020)	Bukit Timah, Singapore	9905	Respiratory syncytial virus -unsubtyped	1.12	1.16	1.24
Sheikh (2020)	Bukit Timah, Singapore	9905	Respiratory syncytial virus -unsubtyped	1.04	1.12	1.16
Sheikh (2020)	Bukit Timah, Singapore	9905	Respiratory syncytial virus -unsubtyped	1.24	1.19	1.44
Ines (2018)	Sousse, Northern Africa	5131	Respiratory syncytial virus -unsubtyped	0.0109	0.0041	0.0257
You (2022)	Gwangju, Korea	3922	Respiratory syncytial virus -unsubtyped	0.1691	0.0473	0.5186
Chee-Sieng (2012)	Kuala Lumpur, Malaysia	10269	Respiratory syncytial virus -unsubtyped	0.6546	0.4387	0.9721
Rosalie (2021)	Amsterdam, The Netherlands	2161	Respiratory syncytial virus -unsubtyped	0.1287	0.0111	0.2463
Rosalie (2021)	Amsterdam, The Netherlands	2161	Respiratory syncytial virus -unsubtyped	0.254	0.156	0.352
Rosalie (2021)	Amsterdam, The Netherlands	2161	Respiratory syncytial virus -unsubtyped	0.1427	0.0055	0.2799
Rosalie (2021)	Amsterdam, The Netherlands	2161	Respiratory syncytial virus -unsubtyped	0.3166	0.2186	0.4146
Rosalie (2021)	Amsterdam, The Netherlands	2161	Respiratory syncytial virus -unsubtyped	0.493	0.3754	0.6106
Rosalie (2021)	Amsterdam, The Netherlands	2161	Respiratory syncytial virus -unsubtyped	0.249	0.1706	0.3274
Wei (2016)	Shanghai, China	2819	Respiratory syncytial virus -unsubtyped	0.4678	0.1358	1.4916
Keita (2023)	47 prefectures, Japan	721709	Respiratory syncytial virus -unsubtyped	1.053	1.015	1.093
Keita (2023)	47 prefectures, Japan	721709	Respiratory syncytial virus -unsubtyped	0.936	0.91	0.964
Qing (2016)	Hangzhou, China	36500	Respiratory syncytial virus -unsubtyped	0.0074	0.0038	0.0136
Jang (2017)	Cheonan, Korea	6279	Respiratory syncytial virus -A	0.8159	0.6667	0.9978
Jang (2017)	Cheonan, Korea	6279	Respiratory syncytial virus -B	0.7781	0.6357	0.9517
Terezinha (2017)	Sao Paulo, Brazil	Not reported	Respiratory syncytial virus -unsubtyped	0.2694	0.1507	0.4684
Terezinha (2017)	Sao Paulo, Brazil	Not reported	Respiratory syncytial virus -unsubtyped	0.2053	0.1128	0.3612
You (2022)	13 European countries	30965	Respiratory syncytial virus -unsubtyped	0.06	0.05	0.07
Mariana (2004)	Buenos Aires, Argentina	18561	Respiratory syncytial virus -unsubtyped	0.0006	0.0001	0.0039

Ping-Ing (2023)	Taiwan, China	1740	Respiratory syncytial virus -unsubtyped	1.4241	0.8641	2.3629
Ping-Ing (2023)	Taipei City, China	1740	Respiratory syncytial virus -unsubtyped	2.0969	1.2669	3.5199
Ping-Ing (2023)	Tainan City, China	1740	Respiratory syncytial virus -unsubtyped	3.405	2.0309	5.8431
Chan (2021)	Hong Kong, China	9635	Respiratory syncytial virus -unsubtyped	15.5187	12.2083	19.8827
Onozuka (2014)	Fukuoka, Japan	30215	Respiratory syncytial virus -unsubtyped	0.51	0.39	0.68
Benjamin (2018)	Nha Trang, central Vietnam	2998	Respiratory syncytial virus -unsubtyped	1.14	0.98	1.33
Meng (2023)	Kent Hill, Singapore	15715	Respiratory syncytial virus -unsubtyped	1.168	1.068	1.277
Meng (2023)	Kent Hill, Singapore	15715	Respiratory syncytial virus -unsubtyped	0.92	0.863	0.981
TNS (2012)	Middle Anatolian, Turkey	3464	Respiratory syncytial virus -unsubtyped	0.0288	0.003	0.1632
TNS (2012)	Mediterranean, Aegean, Marmara, Turkey	3464	Respiratory syncytial virus -unsubtyped	0.005	0.0003	0.0423
TNS (2012)	Black Sea, Turkey	3464	Respiratory syncytial virus -unsubtyped	0.0499	0.0062	0.2546
TNS (2012)	Anatolian Eastern, Turkey	3464	Respiratory syncytial virus -unsubtyped	0.0895	0.0134	0.4156
Wang (2005)	Hangzhou, China	13642	Respiratory syncytial virus -unsubtyped	0.0045	0.0003	0.0316
Geng (2015)	Suzhou, China	42664	Respiratory syncytial virus -unsubtyped	0.0036	0.0011	0.0106
Jung (2018)	Seoul, Korea	9113	Respiratory syncytial virus -unsubtyped	0.0658	0.0247	0.1567
Lin (2015)	Quanzhou, China	6020	Respiratory syncytial virus -unsubtyped	0.19	0.01	2.26
Wu (2021)	Haikou, China	309	Respiratory syncytial virus -A	0.0696	0.0124	0.2849
Wu (2021)	Haikou, China	309	Respiratory syncytial virus -B	0.0751	0.0137	0.3038
Tang (2013)	Suzhou, China	1883	Respiratory syncytial virus -unsubtyped	0.001	0.0001	0.002
Wan (2013)	Suzhou, China	28871	Respiratory syncytial virus -unsubtyped	0.0081	0.002	0.026
Chen (2016)	Suzhou, China	3672	Respiratory syncytial virus -unsubtyped	0.0086	0.0004	0.0774
Zhang (2022)	Jiangyin, China	12294	Respiratory syncytial virus -unsubtyped	0.0112	0.0025	0.0391
Ji (2011)	Suzhou, China	6655	Respiratory syncytial virus -unsubtyped	0.0012	0.0001	0.0087
Chen (2016)	Suzhou, China	3672	Parainfluenza virus -1	1.7339	0.362	9.1988
Chen (2016)	Suzhou, China	3672	Parainfluenza virus -2	2.2662	0.4746	12.5981
Chen (2016)	Suzhou, China	3672	Parainfluenza virus -3	16.3229	14.0957	18.9573
Zhang (2022)	Jiangyin, China	12294	Parainfluenza virus -3	9.4827	8.7965	10.2296
Yan (2015)	Shanghai, China	2526	Parainfluenza viruses -unsubtyped	14.9778	12.5783	17.9086

Zhang (2020)	Suzhou, China	7525	Parainfluenza virus -3	27.8768	24.9627	31.1838
Liu (2019)	Guangzhou, China	11398	human metapneumovirus (HMPV)	0.9436	0.4276	2.079
Nicklas (2016)	Gothenburg, Sweden	20062	human metapneumovirus (HMPV)	0.4769	0.2572	0.8668
Jean (2009)	Rheinland-Pfalz, Mainz	3044	human metapneumovirus (HMPV)	0.0975	0.0594	0.1556
Cheng (2022)	Macao, China	4880	human metapneumovirus (HMPV)	0.7048	0.5842	0.8495
Chen (2014)	Suzhou, China	998	human metapneumovirus (HMPV)	0.1374	0.0185	0.7074
Rory (2019)	Edinburgh, UK	52060	human metapneumovirus (HMPV)	0.901	0.883	0.92
Maria (2009)	Kathmandu, Nepal	887	human metapneumovirus (HMPV)	0.5767	0.1591	1.9639
Magali (2018)	Dijon, France	4300	human metapneumovirus (HMPV)	0.6693	0.6	0.7464
Yong (2019)	Seoul, Korea	23694	human metapneumovirus (HMPV)	0.7913	0.5779	1.0816
Yong (2019)	Seoul, Korea	23694	human metapneumovirus (HMPV)	0.9123	0.667	1.247
Yong (2019)	Seoul, Korea	23694	human metapneumovirus (HMPV)	1.2871	0.9415	1.7627
Li (2013)	Zhuhai, China	924	human metapneumovirus (HMPV)	0.2556	0.0125	2.8583
Wang (2013)	Suzhou, China	6655	human metapneumovirus (HMPV)	0.1996	0.0574	0.6038
You (2022)	Gwangju, Korea	4195	human metapneumovirus (HMPV)	0.3486	0.1087	1.0205
Wang (2011)	Suzhou, China	6599	human metapneumovirus (HMPV)	0.1777	0.0501	0.5427
Ji (2011)	Suzhou, China	6655	human metapneumovirus (HMPV)	0.1777	0.0501	0.5427
You (2022)	Gwangju, Korea	4195	Human coronavirus (HCoV) -unsubtyped	0.03	0.01	0.13
Jean (2009)	Rheinland-Pfalz, Mainz	3044	Human coronavirus (HCoV) -unsubtyped	0.06	0.04	0.1
Jang (2020)	Cheonan, Korea	9010	Human coronavirus (HCoV)- 229E	0.4	0.28	0.59
Jang (2021)	Cheonan, Korea	9010	Human coronavirus (HCoV)- OC43	0.54	0.42	0.7
Nicklas (2016)	Gothenburg, Sweden	20062	Human coronavirus (HCoV) -unsubtyped	0.04	0.02	0.09
Cui (2015)	Chaoshan, China	1074	Human coronavirus (HCoV) -unsubtyped	1.1	1	1.2
Wei (2016)	Shanghai, China	2819	Human coronavirus (HCoV) -unsubtyped	0.68	0.2	2.17
Jean (2009)	Rheinland-Pfalz, Mainz	3044	Human rhinovirus	0.1866	0.1181	0.2886
Giselmo (2020)	Maranhao, Brazil	151	Human rhinovirus	5.3594	0.4718	127.9191
Giselmo (2020)	Maranhao, Brazil	151	Human rhinovirus	2.2662	0.2005	36.7868
Cui (2015)	Chaoshan, China	1074	Human rhinovirus	1	0.9	1.1

Tiina (2016)	Kajaani, Finland	386	Human rhinovirus	0.96	0.92	1
Tiina (2016)	Kajaani, Finland	386	Human rhinovirus	1.04	0.98	1.1
Yan (2017)	Suzhou, China	6194	Human rhinovirus	10.5699	2.6955	54.6864
Jaqueline (2016)	Curitiba, Brazil	755	Human rhinovirus	0.0503	0.0048	0.3027
You (2022)	Gwangju, Korea	4195	Human rhinovirus	0.7048	0.2358	2.0445
Dong (2021)	Cheonan, Korea	1920	Human rhinovirus	1.575	1.443	1.72
Benjamin (2018)	Nha Trang, Vietnam	2998	Human rhinovirus	0.86	0.79	0.94
Zhou 2018	Suzhou, China	5994	Human rhinovirus	4.1649	3.7751	4.5991
Zhou 2017	Huangshi, China	2326	Human rhinovirus	26.184	21.5272	32.0206
Huang 2015	Suzhou, China	1926	Human rhinovirus	32.3742	25.9496	40.6707
Zheng (2014)	Jiangsu, China	42104	human adenoviruses (HAdVs)	7.9418	3.9794	16.8532
Nicklas (2016)	Gothenburg, Sweden	20062	human adenoviruses (HAdVs)	0.5554	0.3016	0.993
Jean (2009)	Rheinland-Pfalz, Mainz	3044	human adenoviruses (HAdVs)	0.1866	0.1181	0.2886
Cheng (2022)	Macao, China	4880	human adenoviruses (HAdVs)	1.1355	0.9425	1.3684
Chen (2014)	Suzhou, China	998	human adenoviruses (HAdVs)	1.1563	0.2365	5.8082
Rory (2019)	Edinburgh, UK	52060	human adenoviruses (HAdVs)	0.972	0.955	0.988
Therese (2021)	Nairobi, Kenya	17261	human adenoviruses (HAdVs)	1.06	0.76	1.47
Therese (2021)	Nairobi, Kenya	17261	human adenoviruses (HAdVs)	0.99	0.7	1.41
Li (2013)	Zhuhai, China	924	human adenoviruses (HAdVs)	34.668	25.1629	48.4754
You (2022)	Gwangju, Korea	4195	human adenoviruses (HAdVs)	0.6215	0.2059	1.8007
Mariana (2004)	Buenos Aires, Argentina	18561	human adenoviruses (HAdVs)	0.1958	0.0652	0.5264
Eun (2021)	Cheonan, Korea	9010	human adenoviruses (HAdVs)	0.772	0.679	0.878
Ji (2011)	Suzhou, China	6655	human adenoviruses (HAdVs)	3.8783	1.3047	12.9526
Lin (2015)	Quanzhou, China	6020	human adenoviruses (HAdVs)	1.48	0.12	20.97
Chen (2012)	Suzhou, China	8197	human adenoviruses (HAdVs)	4.6479	1.7406	13.7737
Chen (2016)	Suzhou, China	3672	human adenoviruses (HAdVs)	3.5087	3.1008	3.9754
Zhang (2022)	Jiangyin, China	12294	human adenoviruses (HAdVs)	2.721	1.142	6.8566
You (2022)	Gwangju, Korea	4195	Enterovirus (EV)	4.3778	1.4613	14.8906

Cui (2015)	Chaoshan, China	1074	Enterovirus (EV)	2.4	2.1	2.7
Jean (2009)	Rheinland-Pfalz, Mainz	3044	Enterovirus (EV)	0.7753	0.5156	1.162
Wei (2016)	Shanghai, China	2819	Enterovirus (EV)	1.6702	0.5244	5.6051
Shiv (2020)	Fifty U.S. states and Washington D.C.	974	SARS-CoV-2	0.85	0.76	0.96
Shiv (2020)	Fifty U.S. states and Washington D.C.	974	SARS-CoV-2	0.98	0.97	0.99
Aleix (2021)	52 Spanish provinces	2418250	SARS-CoV-2	0.08	0.07	0.09
Aleix (2021)	52 Spanish provinces	2418250	SARS-CoV-2	0.07	0.06	0.08
Aleix (2021)	52 Spanish provinces	2418250	SARS-CoV-2	0.1	0.09	0.11
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	1	0.98	1.02
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	0.99	0.99	1
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	0.99	0.98	1
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	0.98	0.97	0.99
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	1.11	1.05	1.17
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	0.99	0.98	0.99
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	0.97	0.95	1
Keita (2022)	Six Japanese prefectures	2982262	SARS-CoV-2	0.97	0.92	1.02
Chen (2021)	Guangdong, China	1347	SARS-CoV-2	0.94	0.93	0.96
Sung (2022)	Changwon, Korea	3234	SARS-CoV-2	0.99	0.98	0.99
Lisa (2022)	Helsinki, Finland	48013	SARS-CoV-2	3.95	2.88	5.42
Kong(2020)	Wuhan, China	Not reported	SARS-CoV-2	10.25	1.76	96.61
Simone (2020)	Milan, Italy	Not reported	SARS-CoV-2	0.26	0.12	0.54
Simone (2020)	Milan, Italy	Not reported	SARS-CoV-2	0.25	0.11	0.52
Simone (2020)	Milan, Italy	Not reported	SARS-CoV-2	0.24	0.11	0.5
Simone (2020)	Florence, Italy	Not reported	SARS-CoV-2	0.35	0.16	0.72
Simone (2020)	Florence, Italy	Not reported	SARS-CoV-2	0.27	0.12	0.56
Simone (2020)	Florence, Italy	Not reported	SARS-CoV-2	0.25	0.11	0.52
Simone (2020)	Trento, Italy	Not reported	SARS-CoV-2	0.22	0.1	0.45
Simone (2020)	Trento, Italy	Not reported	SARS-CoV-2	0.25	0.11	0.52

Cheng (2022)         Macao, China         4880         Human rhinoviruses/enteroviruses (HRVs/EVs)         1.2035         0.9989         1.4508           Raymundo (2023)         Oahu, USA         7143         Human rhinoviruses/enteroviruses (HRVs/EVs)         1.16         0.97         1.39           Bi (2006)         Hong Kong, China         2142         SARS-CoV         0.01         0.0006         0.07           Bi (2006)         Hong Kong, China         2142         SARS-CoV         0.01         0.0006         0.09           Bi (2006)         Hong Kong, China         2142         SARS-CoV         0.01         0.0012         0.09           Feng (2005)         Guangzhou, China         138         SARS-CoV         0.79         0.43         1.46           Feng (2005)         Hong Kong, China         108         SARS-CoV         1.53         0.76         3.11           Feng (2005)         Toronto, Canada         116         SARS-CoV         0.73         0.37         1.42           Feng (2005)         Singapore         72         SARS-CoV         0.73         0.37         1.42           Feng (2005)         Taiwan, China         106         SARS-CoV         0.78         0.38         1.56           Feng (2005	Simone (2020)	Trento, Italy	Not reported	SARS-CoV-2	0.26	0.12	0.54
Raymundo (2023)         Oahu, USA         7143         Human rhinoviruses/enteroviruses (HRVs/EVs)         1.16         0.97         1.39           Bi (2006)         Hong Kong, China         2142         SARS-CoV         0.01         0.0006         0.07           Bi (2006)         Beijing, China         2142         SARS-CoV         0.2         0.04         0.83           Bi (2006)         Hong Kong, China         2142         SARS-CoV         0.01         0.0012         0.09           Feng (2005)         Guangzhou, China         138         SARS-CoV         0.79         0.43         1.46           Feng (2005)         Hong Kong, China         108         SARS-CoV         1.53         0.76         3.11           Feng (2005)         Toronto, Canada         116         SARS-CoV         0.73         0.37         1.42           Feng (2005)         Tiwan, China         106         SARS-CoV         1.03         0.43         2.44           Feng (2005)         Beijing, China         41         SARS-CoV         1.02         0.32         3.31           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow	Asmaa (2020)	Riyadh, Saudi Arabia	Not reported	MERS-CoV	1.054	1.043	1.065
Bit (2006)   Hong Kong, China   2142   SARS-CoV   0.01   0.0006   0.07     Bit (2006)   Beijing, China   2142   SARS-CoV   0.2   0.04   0.83     Bit (2006)   Hong Kong, China   2142   SARS-CoV   0.01   0.0012   0.09     Feng (2005)   Guangzhou, China   138   SARS-CoV   0.79   0.43   1.46     Feng (2005)   Hong Kong, China   108   SARS-COV   1.53   0.76   3.11     Feng (2005)   Toronto, Canada   116   SARS-COV   0.73   0.37   1.42     Feng (2005)   Singapore   72   SARS-COV   1.03   0.43   2.44     Feng (2005)   Taiwan, China   106   SARS-COV   1.03   0.43   2.44     Feng (2005)   Beijing, China   106   SARS-COV   1.02   0.32   3.31     Liu (2021)   Moscow   Not reported   SARS-COV   1.02   0.32   3.31     Liu (2021)   Moscow   Not reported   SARS-COV   3.71   1.78   8.15     Liu (2021)   Moscow   Not reported   SARS-COV   4.65   2.2   10.41     Huang (2004)   Huangping, China   1491   SARS-COV   0.56   0.29   1.05     Huang (2004)   Huangping, China   1491   SARS-COV   0.59   0.31   1.12     Susana (2012)   Valencia, Spain   243   Streptococcus pneumoniae   0.2814   0.0853   0.391     Katerina (2010)   Salvador, Brazil   184   Streptococcus pneumoniae   0.2053   0.1048   0.3919     Katerina (2015)   Kalosca, Spain   13056   Streptococcus pneumoniae   1.17   1.11   1.23     Daniel (2015)   Mallorca, Spain   60659   Streptococcus pneumoniae   0.458   0.3016   0.6911     Huang (2014)   Suzhou, China   13705   Moravella catarrhalis   5.628   2.0781   16.322     PSOTER (2016)   48 states of USA   3463   Pseudomonas aeruginosa   1.04   1.03   1.05     Bit (2005)   0.20   0.20   0.20   0.20   0.20   0.20     Control (2016)   48 states of USA   3463   Pseudomonas aeruginosa   1.04   1.03   1.05     Control (2016)   48 states of USA   3463   Pseudomonas aeruginosa   1.04   1.03   1.05     Control (2015)   48 states of USA   3463   Pseudomonas aeruginosa   1.04   1.03   1.05     Control (2015)   48 states of USA   3463   Pseudomonas aeruginosa   1.04   1.05     Control (2015)   2.025   2.027   2.030     Control (2016)	Cheng (2022)	Macao, China	4880	Human rhinoviruses/enteroviruses (HRVs/EVs)	1.2035	0.9989	1.4508
Bit (2006)         Beijing, China         2142         SARS-CoV         0.2         0.04         0.83           Bit (2006)         Hong Kong, China         2142         SARS-CoV         0.01         0.0012         0.09           Feng (2005)         Guangzhou, China         138         SARS-COV         0.79         0.43         1.46           Feng (2005)         Hong Kong, China         108         SARS-COV         1.53         0.76         3.11           Feng (2005)         Toronto, Canada         1116         SARS-COV         0.73         0.37         1.42           Feng (2005)         Singapore         72         SARS-COV         1.03         0.43         2.44           Feng (2005)         Taiwan, China         106         SARS-COV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-COV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-COV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-COV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-COV <td>Raymundo (2023)</td> <td>Oahu, USA</td> <td>7143</td> <td>Human rhinoviruses/enteroviruses (HRVs/EVs)</td> <td>1.16</td> <td>0.97</td> <td>1.39</td>	Raymundo (2023)	Oahu, USA	7143	Human rhinoviruses/enteroviruses (HRVs/EVs)	1.16	0.97	1.39
Bit (2006)   Hong Kong, China   2142   SARS-CoV   0.01   0.0012   0.09	Bi (2006)	Hong Kong, China	2142	SARS-CoV	0.01	0.0006	0.07
Feng (2005)         Guangzhou, China         138         SARS-CoV         0.79         0.43         1.46           Feng (2005)         Hong Kong, China         108         SARS-CoV         1.53         0.76         3.11           Feng (2005)         Toronto, Canada         116         SARS-CoV         0.73         0.37         1.42           Feng (2005)         Singapore         72         SARS-CoV         1.03         0.43         2.44           Feng (2005)         Taiwan, China         106         SARS-CoV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-CoV         1.02         0.32         3.31           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Huang (2004)         Huangping, China         1491         SARS-CoV </td <td>Bi (2006)</td> <td>Beijing, China</td> <td>2142</td> <td>SARS-CoV</td> <td>0.2</td> <td>0.04</td> <td>0.83</td>	Bi (2006)	Beijing, China	2142	SARS-CoV	0.2	0.04	0.83
Feng (2005)         Hong Kong, China         108         SARS-CoV         1.53         0.76         3.11           Feng (2005)         Toronto, Canada         116         SARS-CoV         0.73         0.37         1.42           Feng (2005)         Singapore         72         SARS-CoV         1.03         0.43         2.44           Feng (2005)         Taiwan, China         106         SARS-CoV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-CoV         1.02         0.32         3.31           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV <td>Bi (2006)</td> <td>Hong Kong, China</td> <td>2142</td> <td>SARS-CoV</td> <td>0.01</td> <td>0.0012</td> <td>0.09</td>	Bi (2006)	Hong Kong, China	2142	SARS-CoV	0.01	0.0012	0.09
Feng (2005)         Toronto, Canada         116         SARS-CoV         0.73         0.37         1.42           Feng (2005)         Singapore         72         SARS-CoV         1.03         0.43         2.44           Feng (2005)         Taiwan, China         106         SARS-CoV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-CoV         1.02         0.32         3.31           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Strepto	Feng (2005)	Guangzhou, China	138	SARS-CoV	0.79	0.43	1.46
Feng (2005)         Singapore         72         SARS-CoV         1.03         0.43         2.44           Feng (2005)         Taiwan, China         106         SARS-CoV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-CoV         1.02         0.32         3.31           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptoc	Feng (2005)	Hong Kong, China	108	SARS-CoV	1.53	0.76	3.11
Feng (2005)         Taiwan, China         106         SARS-CoV         0.78         0.38         1.56           Feng (2005)         Beijing, China         41         SARS-CoV         1.02         0.32         3.31           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.52         0.27         0.99           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)	Feng (2005)	Toronto, Canada	116	SARS-CoV	0.73	0.37	1.42
Feng (2005)   Beijing, China   41   SARS-CoV   1.02   0.32   3.31	Feng (2005)	Singapore	72	SARS-CoV	1.03	0.43	2.44
Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.52         0.27         0.99           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23	Feng (2005)	Taiwan, China	106	SARS-CoV	0.78	0.38	1.56
Liu (2021)         Moscow         Not reported         SARS-CoV         3.71         1.78         8.15           Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.52         0.27         0.99           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcus pneumoniae         1.04         0.4588         0.301	Feng (2005)	Beijing, China	41	SARS-CoV	1.02	0.32	3.31
Liu (2021)         Moscow         Not reported         SARS-CoV         4.65         2.2         10.41           Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.52         0.27         0.99           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcus pneumoniae         1.04         0.0687         0.0073         0.3909           Sun (2014)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687<	Liu (2021)	Moscow	Not reported	SARS-CoV	4.65	2.2	10.41
Huang (2004)         Huangping, China         1491         SARS-CoV         0.56         0.29         1.05           Huang (2004)         Huangping, China         1491         SARS-CoV         0.52         0.27         0.99           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcus pneumoniae         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628 <td>Liu (2021)</td> <td>Moscow</td> <td>Not reported</td> <td>SARS-CoV</td> <td>3.71</td> <td>1.78</td> <td>8.15</td>	Liu (2021)	Moscow	Not reported	SARS-CoV	3.71	1.78	8.15
Huang (2004)         Huangping, China         1491         SARS-CoV         0.52         0.27         0.99           Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcus pneumoniae         1.17         1.11         1.23           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa	Liu (2021)	Moscow	Not reported	SARS-CoV	4.65	2.2	10.41
Huang (2004)         Huangping, China         1491         SARS-CoV         0.59         0.31         1.12           Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcal pharyngitis         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Huang (2004)	Huangping, China	1491	SARS-CoV	0.56	0.29	1.05
Susana (2012)         Valencia, Spain         243         Streptococcus pneumoniae         0.2814         0.0853         0.8321           Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcal pharyngitis         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Huang (2004)	Huangping, China	1491	SARS-CoV	0.52	0.27	0.99
Cristiana (2010)         Salvador, Brazil         184         Streptococcus pneumoniae         0.2053         0.1048         0.3919           Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcal pharyngitis         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Huang (2004)	Huangping, China	1491	SARS-CoV	0.59	0.31	1.12
Katerina (2019)         Liverpool, UK         374         Streptococcus pneumoniae         0.79         0.63         0.99           Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcal pharyngitis         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Susana (2012)	Valencia, Spain	243	Streptococcus pneumoniae	0.2814	0.0853	0.8321
Onozuka (2009)         Fukuoka, Japan         13056         Streptococcus pneumoniae         1.17         1.11         1.23           Daniel (2015)         Mallorca, Spain         60659         Streptococcal pharyngitis         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Cristiana (2010)	Salvador, Brazil	184	Streptococcus pneumoniae	0.2053	0.1048	0.3919
Daniel (2015)         Mallorca, Spain         60659         Streptococcal pharyngitis         0.4588         0.3016         0.6911           Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Katerina (2019)	Liverpool, UK	374	Streptococcus pneumoniae	0.79	0.63	0.99
Huang (2019)         Zhongshan, China         13705         Moraxella catarrhalis         0.0687         0.0073         0.3909           Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Onozuka (2009)	Fukuoka, Japan	13056	Streptococcus pneumoniae	1.17	1.11	1.23
Sun (2014)         Suzhou, China         8143         Moraxella catarrhalis         5.628         2.0781         16.3229           PSOTER (2016)         48 states of USA         3463         Pseudomonas aeruginosa         1.04         1.03         1.05	Daniel (2015)	Mallorca, Spain	60659	Streptococcal pharyngitis	0.4588	0.3016	0.6911
PSOTER (2016) 48 states of USA 3463 Pseudomonas aeruginosa 1.04 1.03 1.05	Huang (2019)	Zhongshan, China	13705	Moraxella catarrhalis	0.0687	0.0073	0.3909
	Sun (2014)	Suzhou, China	8143	Moraxella catarrhalis	5.628	2.0781	16.3229
Susana (2012)         Valencia, Spain         243         Legionella pneumophila         2.4518         1.1877         7.7123	PSOTER (2016)	48 states of USA	3463	Pseudomonas aeruginosa	1.04	1.03	1.05
	Susana (2012)	Valencia, Spain	243	Legionella pneumophila	2.4518	1.1877	7.7123

Lin (2015)	Quanzhou, China	6020	Legionella pneumophila	6.15	0.53	149.9
Huang (2022)	Suzhou, China	7940	Haemophilus influenzae	0.4276	0.1367	1.2438
Jean (2009)	Rheinland-Pfalz, Mainz	3044	Mycoplasma pneumoniae	0.4956	0.3265	0.7456
Chen (2020)	Suzhou, China	998	Mycoplasma pneumoniae	22.2606	17.6757	28.2451
Chen (2013)	Suzhou, China	1598	Mycoplasma pneumoniae	180.7475	146.4838	275.1224
Chen (2013)	Suzhou, China	8157	Mycoplasma pneumoniae	5.5715	2.0588	16.9388
Huang ( 2019)	Zhongshan, China	13705	Mycoplasma pneumoniae	180.7475	179.2791	222.2747
Tian (2017)	Hangzhou, China	3769	Mycoplasma pneumoniae	4.0363	2.7227	6.0729
Tian (2017)	Hangzhou, China	3769	Mycoplasma pneumoniae	4.2408	2.8561	6.3941
Tian (2017)	Hangzhou, China	3769	Mycoplasma pneumoniae	3.539	2.3968	5.2961
Na (2022)	Nantong, China	51665	Mycoplasma pneumoniae	3.0394	1.6853	5.6372
Wang (2017)	Suzhou, China	15098	Mycoplasma pneumoniae	3.6796	1.7246	8.2816
Lei (2021)	Chengdu, China	22882	Mycoplasma pneumoniae	1.696	0.7718	3.8203
Onozuka (2008)	Fukuoka, Japan	13056	Mycoplasma pneumoniae	1.17	1.11	1.23
Yuan (2022)	Chendu, China	5127	Mycoplasma pneumoniae	1.68	1.32	2.13
Lin (2015)	Quanzhou, China	6020	Mycoplasma pneumoniae	9.31	0.77	274.79
Wang (2013)	Guilin, China	1342	Mycoplasma pneumoniae	19.2665	15.0558	24.8649
Wu (2019)	Wu zhishan, China	1597	Mycoplasma pneumoniae	39.5804	30.8471	51.2382
Na (2021)	Tianjin, China	63821	Mycoplasma pneumoniae	0.046	0.0098	1.0362
Ni (2014)	Suzhou, China	10596	Mycoplasma pneumoniae	8.1661	3.1915	23.4988
Yin (2013)	Luoyang, China	8368	Mycoplasma pneumoniae	9.5957	2.9793	37.5377
Zheng (2013)	Zhengzhou, China	256	Mycoplasma pneumoniae	0.00001	0.00001	0.0022
Shi (2016)	Suzhou, China	2323	Mycoplasma pneumoniae	0.0807	0.015	0.3235
Huang (2019)	Zhongshan, China	13705	Chlamydia pneumoniae	180.7475	179.2791	222.2747
Chen (2013)	Suzhou, China	1598	Chlamydia pneumoniae	0.9266	0.0717	11.5859
Jean (2009)	Rheinland-Pfalz, Mainz	3044	Chlamydia pneumoniae	0.8968	0.5975	1.3444
Lin (2015)	Quanzhou, China	6020	Chlamydia pneumoniae	1.59	0.16	22.87

## The code is as follows:

```
setwd(dir="C:/Users/86183/Desktop/data.csv")
getwd()
bin.metagen <- read.csv("C:/Users/86183/Desktop/data.csv")
head(bin.metagen, 150)
View(bin.metagen)
library("meta")
library(base)
bin.metagen$RR <- log(bin.metagen$RR)
bin.metagen$lower <- log(bin.metagen$lower)</pre>
bin.metagen$upper <- log(bin.metagen$upper)</pre>
bin.metagen$seTE <- (bin.metagen$upper - bin.metagen$lower)/3.92
fit<-metagen(TE=RR,
       seTE=seTE,
       studlab = paste(Study), method.tau="REML",
       sm = "RR", data = bin.metagen,byvar = Suboutcome,print.byvar = FALSE,
       comb.random = TRUE,
       comb.fixed = FALSE, prediction= TRUE, prediction.subgroup = TRUE,
       label.left ="Reduced risk",
       label.right ="Increased risk")
forest(fit,col.diamond = "skyblue",col.square="maroon",col.diamond.lines = "black", sortvar = TE,
    base_family = "Times New Roman", family="sans",fontsize=11,
    col.predict = "#FF0000",colgap.studlab = "4cm", colgap.forest.left = "5.0cm",
    leftlabs = c("Study"))
forest(fit,col.diamond = "maroon",col.square="skyblue",col.diamond.lines = "black",sortvar = TE,
    family="mono",fontsize=11,colgap.studlab = "4cm", colgap.forest.left = "5.0cm",
```

```
base_family = "Times New Roman",col.predict = "#FF0000",
    leftlabs = c("Study"))
Subgroup analysis
fit<-metagen(TE=RR,
       seTE=seTE,
       studlab = paste(Study),
       method.tau="REML",
       sm = "RR",
       data = bin.metagen,byvar = Climate/Income/Resolution/Measurement/Model/Lag,
       comb.random = TRUE,
       comb.fixed = FALSE, prediction.subgroup = TRUE, prediction= TRUE,
       label.left ="Reduced risk",
       label.right ="Increased risk")
forest(fit,col.diamond = "skyblue",col.square="maroon",col.diamond.lines = "black",
    base_family = "Times New Roman",colgap.studlab = "4cm", colgap.forest.left = "5.0cm",
    family="sans",fontsize=11, col.predict = "#FF0000",
    leftlabs = c("Study"))
forest(fit,col.diamond = "maroon",col.square = "skyblue",col.diamond.lines = "black",
    base_family = "Times New Roman",
    family="sans",fontsize=11, col.predict = "#FF0000",colgap.studlab = "4cm",colgap.forest.left = "5.0cm",
    leftlabs = c("Study"))
Sensitivity analysis
fit<-metagen(TE=RR,
```

```
seTE=seTE,
       studlab = paste(Study),
       method.tau="REML",
       sm = "RR",
       data = bin.metagen,
       comb.random = TRUE,
       comb.fixed = FALSE,
       label.left ="Reduced risk",
       label.right ="Increased risk")
forest(metainf(fit,pooled="random"),
    comb.random=TRUE,family="sans",base family = "Arial",refline col = "#004CFFFF",
    fontsize=11,lwd=2,col.diamond.fixed="#4D9221",col.diamond.lines.fixed="#4D9221",
    col.diamond.random="#FF0000",col.diamond.lines.random="maroon",
    col.square="#00FF00",col.study="#C51B7D",lty.fixed=4,
    plotwidth="8cm",colgap.forest.left="1cm",
    colgap.forest.right="1cm",just.forest="right",colgap.left="0.5cm",
    colgap.right="0.5cm")
funnel(fit,
    contour.levels = c (0.9, 0.95, 0.99), studlab = TRUE,
    col.contour = c('darkgray', 'gray', 'lightgray'))
legend(1,0.02, c("0.1 > p > 0.05", "0.05 > p > 0.01", "< 0.01"), fill=c("white", "gray", "darkgray"))
par(mfrow=c(2,2))
dev.off()
```

```
tf.publ<-trimfill(fit)
summary(tf.publ)
funnel (tf.publ)
dev.off()
metabias(fit,method="linreg")
```