

SHORT COMMUNICATION

Significance of clinical phenomes of patients with COVID-19 infection: A learning from 3795 patients in 80 reports

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Abstract

A new coronavirus SARS-CoV-2 has caused outbreaks in multiple countries and the number of cases is rapidly increasing through human-to-human transmission. Clinical phenomes of patients with SARS-CoV-2 infection are critical in distinguishing it from other respiratory infections. The extent and characteristics of those phenomes varied depending on the severities of the infection, for example, beginning with fever or a mild cough, progressed with signs of pneumonia, and worsened with severe or even fatal respiratory difficulty in acute respiratory distress syndrome. We summarized clinical phenomes of 3795 patients with COVID-19 based on 80 published reports from the onset of outbreak to March 2020 to emphasize the importance and specificity of those phenomes in diagnosis and treatment of infection, and evaluate the impact on medical services. The data show that the incidence of male patients was higher than that of females and the level of C-reaction protein was increased as well as most patients' imaging included ground-glass opacity. Clinical phenomes of SARS-CoV-2 infection were compared with those of SARS-CoV and MERS-CoV infections. There is an urgent need to develop an artificial intelligence-based machine learning capacity to analyze and integrate radiomics- or imaging-based, patient-based, clinician-based, and molecular measurements-based data to fight the outbreak of COVID-19 and enable more efficient responses to unknown infections in future.

KEYWORDS

acute lung injury, clinical phenome, COVID-19, lung

The coronavirus (CoV) is a single-stranded enveloped virus, belonging to the family Coronaviridae, with a positive-sense

RNA genome about 26–32 kb in size. It is widespread among humans and other mammals, such as camels, bats, masked

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palm civets, mice, dogs, and cats.^{1,2} Two viruses, severe acute respiratory syndrome coronavirus (SARS-CoV)³ and MERS-CoV,⁴ have had a significant impact on humans over the past two decades, although most CoVs cause mild clinical symptoms in humans. In December 2019, pneumonia with an unknown cause broke out in Wuhan, China,⁵ which was different from SARS-CoV and MERS-CoV. This new type of CoV is the seventh family member of CoV to infect humans.⁶ The International Virus Classification Commission has renamed the virus, previously temporarily named 2019-nCoV, as severe acute respiratory syndrome coronavirus-2 (SARS-CoV-2), and WHO has announced a new name for this epidemic: coronavirus disease (COVID-19).⁷ Genome sequencing found it to be a beta CoV with at least 70% genetic-sequence similarity to SARS-CoV. We summarized clinical phenomes of 3795 patients with SARS-CoV-2 infection from 80 international and national reports (Supporting Information Table 1) since the outbreak, mainly including complaints and signs (3795 patients in Table 1), biochemical measurements (3085 patients in Table 2), and imaging information with an autopsy report (2856 patients in Tables 3 and 4). There is only one autopsy report because there is less information in this field, but understanding the pathogenesis of COVID-19 may also help doctors make treatment plans, so we also included them.

The most common clinical phenomes of patients were fever, malaise, dry cough, shortness of breath, and respiratory distress, as reported initially.⁸ One study showed that about 66% of patients had a history of exposure to a wet market in Wuhan, of whom 73% were males and <50% suffered from underlying diseases, including diabetes, high blood pressure, and cardiovascular diseases. More than 90% had fever ranging between 38 and 39°C and bilateral chest abnormalities on radiography, 76% had cough, 63% had reduced lymphocyte number, 55% dyspnea, 44% fatigue, 28% sputum, 25% low white blood cell counts, as well as other clinical features, including headache, hemoptysis, diarrhea, or hypertension.⁹ It would be more beneficial to set up a digital evaluation score system to present the variety of clinical phenomes and integrate with molecular multi-omics data. Table 1 shows that the number of male patients was higher than that of females, with an average age of 46.48 years. The most common phenomes are fever, cough, fatigue, sputum, myalgia, sore throat, and shortness of breath, as summarized from more than 80 published studies (Supporting Information Table 1).

In addition, clinical phenomes related with gastrointestinal phenomes, for example, nausea or vomiting were also noticed.¹⁰ The history of exposure to susceptibility factors is one of the critical factors in the diagnosis, including visit to the seafood market, travel through infected area, or contact with people with infection.¹¹ Radiomics-based phenomes become more important in the early diagnosis and dynamic monitor of disease progression. Imaging phenomes of

SARS-CoV-2-infected pneumonia can be detected about 9 days after the first visit to the clinic, and turbid basal striations in both lower lobes of lungs are shown at about 10 days, during which the rale lung sound could be heard.¹⁰ Multiple peripheral ground-glass opacities appeared in both lobes of lungs in some patients.¹² From the preliminary analysis of about 80 clinical reports (Supporting Information Table 1), about 75% (2856/3795 patients) had chest CTs that showed ground-glass opacity, patchy shadowing, and bilateral lower-lobe inflammation as common radiographic features of patients with COVID-19 infection. The progression of COVID-19-infected lung injury is described by increased density, profusion, confluence, and pleural effusion in chest radiographs⁶ (Table 3). In a study of 159 asymptomatic patients, we collected chest CT imaging of most patients that showed ground-glass opacity shadows, of which 82 were pure ground-glass opacity shadows, and 17 cases presented ground-glass opacity with consolidation. Among patients with asymptomatic infections, a chest CT scan is particularly important to facilitate early detection of suspected cases and guide early isolation. There is an urgent need to develop a mode of artificial intelligence analysis to generate more detailed imaging phenomes and digitalize the image abnormality of density, size, distribution, and geometric figures.

Table 2 demonstrates that the number of lymphocytes, CD3, and CD4 reduced, whereas levels of C-reactive protein, D-dimer, thrombin time, and lactate dehydrogenase increased in patients with SARS-CoV-2 infection. When collecting patient information from multiple studies, because the data were obtained from different hospitals and laboratory platforms have certain differences, different data units were unified and then collated, some indicators have a large difference in normal range, only recorded the abnormal number of people rather than specific values, and others without the specific value also recorded the number of people. The SARS-CoV-2 nucleic acid test kit was being developed rapidly, with a clear improvement in sensitivity, based on different biotechnologies. The rapid development and breakthrough of SARS-CoV-2 kits were carried out by immediately integrating and collaborating multidisciplinary resources and efforts and by swiftly translating new technologies, gene sequencing, and bioinformatics into clinical practice.¹³ Biochemical phenomes become critical criteria to confirm SARS-CoV-2 infection, for example, isolation of SARS-CoV-2, real-time reverse-transcriptase PCR assay, or SARS-CoV-2 full-genome sequencing.¹¹ The latest edition of the National Health Commission, the seventh edition of the new CoV pneumonia diagnosis and treatment program, in addition to nucleic acid detection and sequencing also increased serological testing as the basis for diagnosis. That means detecting new CoV-specific IgM antibodies and following IgG-positive or new CoV-specific IgG antibodies from negative to positive or recovery period from four times

TABLE 1 Patient complaint- and physical examination-based phenomes

| Phenomes | Sum/number | Phenomes | Sum/number | Phenomes | Sum/number | Phenomes | Sum/number | Phenomes | Sum/number | Phenomes | Sum/number | Number/mean \pm SE |
|-------------------------|------------|-----------------------------|------------|-----------------------------|------------|------------------------------------|------------|--|--------------------|----------|------------|----------------------|
| General information | | Nasal congestion | 82 | Myalgia or arthralgia | 365 | Hepatitis B infection | 26 | Arrhythmia | 23 | | | |
| Patient number | 3795 | Chill | 188 | | | COPD | 51 | ARDS | 99 | | | |
| Male | 2081 | Little phlegm | 57 | Abdominal pain and diarrhea | 7 | Chronic lung disease | 26 | AKI | 5 | | | |
| Female | 1705 | Sputum | 586 | Dyspnea | 188 | Cardiac or cardiovascular disease | 205 | Respiratory failure | 11 | | | |
| Age | 46.48 | Blood in sputum | 3 | Chest tightness and dyspnea | 83 | Never smoked | 927 | Hepatic insufficiency | 2 | | | |
| Clinical classification | | Myalgia or fatigue | 103 | Shortness of breath | 323 | Former smoker | 29 | Renal insufficiency | 14 | | | |
| Mild type | 37 | Mild headache and dizziness | 8 | Chest distress | 136 | Current cigarette smoker | 196 | Secondary infection | 4 | | | |
| Moderate type | 487 | Coarse breath sounds | 3 | Chest pain | 41 | Cerebrovascular disease | 32 | DIC | 1 | | | |
| Severe type | 331 | Rhinorrhea | 65 | Rhonchi | 4 | Cardiovascular and cerebrovascular | 53 | Rhabdomyolysis | 2 | | | |
| Critically ill type | 8 | | | Moist rales | 2 | Malignancy | 24 | Co-infection | | | | |
| Signs and symptoms | | Sneeze | 5 | Conjunctivitis | 1 | Chronic kidney disease | 25 | Other viruses | 0 | | | |
| No obvious symptom | 184 | Weakness | 7 | Conjunctival congestion | 9 | HIV infection | 2 | Bacteria | 1 | | | |
| Fever | 1703 | Sore throat | 297 | | | Bacterial co-infections | 2 | Fungus | 4 | | | |
| Higher temperature | 86.8 | Agrypnia | 22 | Hemoptysis | 2 | Digestive system disease | 31 | | | | | |
| Normal | 2 | Gastrointestinal symptoms | | Confusion | 9 | Endocrine system disease | 27 | PaO ₂ (83-108) | 93.76 \pm 4.40 | | | |
| <37.3°C | 742 | Loss of appetite | 208 | Upper airway congestion | 8 | Nervous system disease | 2 | PaO ₂ :FIO ₂ (400-500) | 207.84 \pm 35.92 | | | |
| 37.3-38.0°C | 541 | Nausea and vomit | 216 | Signs of infection | | Immunodeficiency | 6 | SaO ₂ (\geq 95%) | 75.43 \pm 5.79 | | | |
| 38.1-39.0°C | 483 | Nausea | 60 | Throat congestion | 19 | Hyperlipidemia | 22 | Decreased | 9 | | | |
| >39.0°C | 132 | Vomit | 20 | Tonsil swelling | 23 | Cholelithiasis | 6 | PaCO ₂ (35-48) | 36.69 \pm 1.10 | | | |

TABLE 1 Patient complaint- and physical examination-based phenomes

| Phenomes | Sum/ number | Phenomes | Sum/ number | Phenomes | Sum/ number | Phenomes | Sum/ number | Phenomes | Sum/ number | Phenomes | Number/ mean \pm SE |
|---------------------------------------|----------------|---------------------------|----------------|----------------------------|----------------|----------------------|----------------|--------------------|-------------------|----------|--------------------------|
| Fatigue | 943 | Diarrhea | 187 | Enlargement of lymph nodes | 2 | Thyroid diseases | 9 | Heart rate mean | 91.85 \pm 0.92 | | |
| Headache | 290 | Gastrointestinal reaction | 1 | Rash | 2 | Urolithiasis | 3 | Respiratory mean | 21.70 \pm 0.15 | | |
| Headache and mental disorder symptoms | 13 | Abdominal pain | 14 | Comorbidities | | Stroke | 3 | MAP | 90.00 \pm 0.00 | | |
| Dizziness | 39 | Belching | 7 | Diabetes | 214 | Complications | | Systolic pressure | 131.82 \pm 1.36 | | |
| Dizziness or headache | 7 | Constipation | 48 | Hypertension | 415 | Shock | 32 | Diastolic pressure | 79.57 \pm 0.94 | | |
| Cough | 1986 | Hemoptysis | 17 | Chronic liver disease | 40 | Acute cardiac injury | 16 | | | | |

Abbreviations: AKI, acute kidney injury; ARDS, acute respiratory distress syndrome; COPD, chronic obstructive pulmonary disease; DIC, disseminated intravascular coagulation; MAP, mean arterial pressure; PaCO₂, partial pressure of carbon dioxide in artery; PaO₂, partial pressure of oxygen; FiO₂, fraction of inspiration O₂; PaO₂, partial pressure of oxygen; SaO₂, arterial oxygen saturation.

higher than the acute phase can also be diagnosed. The genomes of SARS-CoV-2 are closely related to those of two bat-derived SARS-like CoVs, bat-SL-CoVZC45 and bat-SL-CoVZXC21, but different from SARS-CoV (about 79%) and MERS-CoV (about 50%). However, the homology modeling indicated that SARS-CoV-2 has receptor-binding domain structures, similar to that of SARS-CoV,¹⁴ although there are some differences in the key residues.

In addition to diagnosis, clinical phenomes of patients with SARS-CoV-2 add significant impact to dynamically differentiate it from other diseases, predict the disease severity, and monitor the therapeutic effect, although the specificity should be furthermore defined. Pulmonary edema with hyaline membrane formation, inflammatory cell infiltration (mainly lymphocytes), multinucleated syncytial cells in alveolar space were observed in lung autopsy tissue of patients with SARS-CoV-2 (Table 4), similar to those with SARS-CoV and MERS-CoV infections.^{16,17} Liver biopsy specimens from COVID-19 patients showed moderate microcapsule steatosis and mild lobular and portal activity, which might have been damaged directly by SARS-CoV-2 infection or drug-associated toxicity. Among them, imaging information and genetic detection play important roles. For example, pneumonia of SARS-CoV-2 infection should be differentiated from that of mycoplasma pneumonia and community-acquired pneumonia, which often occur in children and adolescents with a clear seasonal trend. Nucleic acid amplification test can be used in the acute phase of the disease, combined with serological testing, to reduce potential false negative results.¹⁵ Clinical phenomes of mycoplasma pneumonia mainly include alterations in the lower-middle lung area, tracheobronchitis, mucinous or mucous purulent sputum, headache, sore throat, rhinitis, and otitis media. The scoring system of clinical phenomes was developed to identify mycoplasma and guide patient management and adjuvant antimicrobial therapy, with 83% sensitivity.¹⁶ Clinical phenomes of patients with SARS-CoV included dyspnea, recurrent or persistent fever, new pneumonia infiltration on chest imaging, for example, multifocal airspace merger in the lower lung, and ground-glass opacity with consolidation.³ Patients with MERS-CoV in 2012 had clinical phenomes including fever, cough, dyspnea, and pneumonia, like any other lower respiratory diseases, but also had rapidly developing acute respiratory distress syndrome, multiple-organ failure, and death. Compared with SARS-CoV, MERS-CoV develops faster into respiratory failure and causes acute kidney damage.¹⁷

However, there is a lack of clinical phenomes related to the mental health aspect of patients with SARS-CoV-2 or SARS-CoV and MERS-CoV, suspected people with the infections, as well as the normal population, during the early stage, duration, and late stage of infection. The abnormality of mental phenomes can reflect the living environment and influence the quality of people's lives, although the impact

TABLE 2 Biochemical- and molecular measurements-based phenomes

| Phenomes | Number/ mean \pm SE | Phenomes | Number/ mean \pm SE | Phenomes | Number/ mean \pm SE | Phenomes | Number/ mean \pm SE | Phenomes | Number/ mean \pm SE |
|-----------------------|--------------------------|------------------------------|--------------------------|-------------------------|--------------------------|--------------------------|--------------------------|--------------------------|--------------------------|
| Inflammatory cells | | Platelets (125-350) | 180.92 \pm 1.94 | Amyloid A (<10) | 237.73 \pm 55.51 | D-dimer (<0.55) | 16.14 \pm 4.15 | Creatine kinase (<171) | 103.01 \pm 5.05 |
| Patient number | 3085 | Increased | 10 | Increased | 14 | Increased | 262 | \geq 200 U/L | 90 |
| WBC (3.5-9.5) | 5.53 \pm 0.05 | Decreased | 24 | Normal | 4 | Normal | 2 | Creatine kinase-MB (<25) | 13.03 \pm 1.53 |
| Increased | 81 | Normal | 1 | Ferritin (21.0-274.7) | 808.7 \pm 0.00 | FDP (1-5) | 4.00 \pm 0.00 | CTnl (0-0.1) | 0.53 \pm 0.22 |
| Decreased | 230 | Haemoglobin (110-160) | 137.10 \pm 1.04 | CD3 (723-2737) | 363.74 \pm 90.37 | AT (80-120) | 91.00 \pm 0.00 | LDH (125-243) | 277.98 \pm 6.12 |
| Normal | 202 | Decreased | 1 | Decreased | 28 | Blood biochemistry | | Increased | 13 |
| Neutrophils (1.8-6.3) | 15.37 \pm 1.97 | Normal | 10 | CD4 (404-1612) | 244.08 \pm 34.04 | Lactate (0.5-2.0) | 1.20 \pm 0.05 | Myoglobin (0-110) | 58.69 \pm 4.59 |
| Increased | 43 | Eosinophils (0.02-0.52) | 0.05 \pm 0.00 | Decreased | 60 | Albumin (40-55) | 34.89 \pm 0.94 | Glucose (3.9-6.1) | 8.13 \pm 0.49 |
| Decreased | 47 | Increased | 2 | IL-6 (0.1-2.9) | 149.66 \pm 62.27 | Alanine ATase (9-50) | 32.95 \pm 0.97 | BNP (0-23.1) | 43.90 \pm 0 |
| Normal | 54 | Decreased | 14 | CD8 (220-1129) | 247.08 \pm 32.54 | Increased | 160 | HT-1 (<26.2) | 4.90 \pm 0.75 |
| Lymphocytes (1.1-3.2) | 1.72 \pm 0.14 | Normal | 29 | Decreased | 20 | Aspartate ATase (15-40) | 35.48 \pm 0.78 | K (3.5-5.5) | 4.05 \pm 0.10 |
| Increased | 1 | Infection-related biomarkers | | CD19 | 72.77 \pm 15.89 | Increased | 171 | Decreased | 10 |
| Decreased | 347 | Procalcitonin (<0.05) | 0.25 \pm 0.04 | CD 16 (<56) | 64.21 \pm 20.61 | TG (<1.70) | 1.14 \pm 0.00 | Na (136-146) | 137.72 \pm 0.35 |
| Normal | 24 | Increased | 71 | Coagulation function | | Total bilirubin (4.7-24) | 15.67 \pm 1.72 | Cl (96-106) | 98.77 \pm 1.00 |
| RBC (3.5-5.5) | 4.45 \pm 0.14 | Normal | 95 | Thrombin time s (16-18) | 28.90 \pm 0.00 | Increased | 78 | Decreased | 15 |
| Monocytes (0.1-0.6) | 0.46 \pm 0.04 | ESR (20 mm/h) | 35.44 \pm 1.21 | APTT (25.1-36.5) | 32.02 \pm 0.62 | BUN (2.8-7.6) | 4.74 \pm 0.00 | PH (7.35-7.45) | 7.42 \pm 0.00 |
| Increased | 10 | CRP (0-10) | 38.01 \pm 1.49 | PT (9.4-12.5) | 12.58 \pm 0.16 | Serum urea (2.5-7.1) | 3.80 \pm 0.00 | Increased | 21 |
| Decreased | 3 | Increased | 870 | Fibrinogen (2-4) | 5.23 \pm 0.52 | Decreased | 3 | decreased | 4 |
| Normal | 16 | Normal | 66 | Increased | 12 | Creatinine (64-104) | 68.19 \pm 1.45 | Lactate (0.5-1.6) | 1.30 \pm 0.00 |
| | | | | Increased | 15 | Increased | 15 | Increased | 2 |
| | | | | Decreased | 2 | Decreased | 2 | Decreased | 2 |

Abbreviations: APTT, activated partial thromboplastin time; AT, angiotensin; BNP, brain natriuretic peptide; BUN, blood urea nitrogen; CD16, cluster of differentiation 16; CD19, cluster of differentiation 19; CD3, cluster of differentiation 3; CD4, cluster of differentiation 4; CD8, cluster of differentiation 8; Cl, chlorine; CRP, C-reactive protein; CTnl, cardiac troponin I; ESR, erythrocyte sedimentation rate; FDP, fibrin degradation product; HT-1, hypersensitive troponin I; IL-6, interleukin-6; K, potassium; LDH, lactate dehydrogenase; M, mean; Na, sodium; PH, potential of hydrogen; PT, prothrombin time; RBC, red blood cell count; TG, triglyceride; WBC, white blood cell count.

TABLE 3 Imaging-based phenomes, eg radiographs, ultrasounds, pathology

| Phenomes | Sum/ number | Phenomes | Sum/ number | Phenomes | Sum/ number | Phenomes | Sum/ number |
|--|----------------|-------------------------------|----------------|---|----------------|--|----------------|
| Phenomes | SUM | Pleural impairment | 2 | Right | 28 | Crazy paving pattern | 51 |
| Patient number | 2856 | Pleural thicken | 15 | Left upper lobe | 187 | Peripheral distribution | 7 |
| Normal | 75 | Pleural effusion | 21 | Left lower lobe | 285 | Peribronchovascular | 1 |
| Patchy | 491 | Pericardial effusion | 8 | Left | 19 | Scattered opacities | 1 |
| GGO/consolidation | 18 | Lymphadenopathy | 9 | Upper lobes | 43 | Ill-defined | 1 |
| GGO without consolidation | 12 | Lobe affected | | Middle lobe | 30 | Other findings | |
| GGO with consolidation | 113 | 0 | 24 | Lower lobes | 51 | Discrete pulmonary nodules | 14 |
| Consolidation | 119 | 1 | 102 | Predominant distribution | | Lymph node enlargement | 9 |
| Parenchymal abnormalities | 3 | 2 | 59 | Anterior | 9 | Fibrous stripes | 35 |
| Spider web sign | 20 | 3 | 56 | Posterior | 42 | Cystic changes | 8 |
| Ill-defined margins | 124 | 4 | 62 | Inferior | 2 | Bronchiectasis | 11 |
| GGOs with interstitial and/or interlobular septal thickening | 147 | 5 | 136 | Peripheral | 152 | Adjacent pleura thickening | 77 |
| GGO | 999 | >2 lobes | 70 | Subpleural region | 45 | Interlobular septa thickening | 1 |
| Patchy/punctuated GGO | 105 | Unilateral lung | 107 | Central | 5 | Interstitial abnormalities | 143 |
| GGO were bilateral | 182 | Bilateral lung disease | 1119 | Diffuse distribution | 43 | Pleura thickening | 1 |
| GGO involving the posterior lungs | 41 | Bilateral upper lobes | 67 | Peribronchial distribution | 3 | Qualitative change on follow-up chest CT | |
| GGO involving the peripheral | 44 | Bilateral lower lobes | 72 | Both central and peripheral | 1 | No change | 1 |
| Air bronchogram | 165 | Frequency of lobe involvement | | Mixed distribution | 26 | Disease improvement | 1 |
| Bronchial wall thickening | 9 | Right upper lobe | 181 | Opacification distribution and patterns | | Mild disease progression | 7 |
| reticulation | 23 | Right middle lobe | 124 | Rounded morphology | 7 | Moderate disease progression | 2 |
| | | Right lower lobe | 306 | Linear opacities | 58 | Severe disease progression | 1 |

Abbreviation: GGO, ground-glass opacity.

on diagnosis and treatment remain unclear. The COVID-19-associated phobia, depression, and over-responses should be investigated in patients with infection, suspected candidates, relatives, and people in the society. It is important to develop an artificial intelligence evaluation scoring system to dynamically monitor phenomes of mental health as part of the clinical phenomes and enable application to a large number of suspected people before definitive diagnosis and treatment.

Clinical phenomes should be given the priority in the diagnosis and treatment of COVID-19. The development of SARS-CoV-2 infection-associated phenome list required efficient and professional manpower and a big data-based

technology in a short and high-pressure period during a sudden outbreak. The artificial intelligence system can help medical staff to automatically collect, achieve, analyze, and optimize databases of clinical phenomes, and offer an immediate response and indication of the patient's health condition. Such an evaluation system will speed up the diagnosis and save a lot of clinical manpower and resources. The digitalized lung imaging analysis system can effectively quantify dynamic changes of the lung tissue and assist clinical judgment and decision-making. The system should have the capacity for artificial intelligence-based machine learning to analyze radiomics- or imaging-based data (eg, pathology,

TABLE 4 Pathological characteristics of postmortem biopsies

| Phenomes | Sum/ number | Phenomes | Sum/ number |
|---|----------------|---|----------------|
| Patient number | 1 | Viral cytopathic-like changes | 1 |
| Histological examination | | Intranuclear or intracytoplasmic viral inclusions | 0 |
| Bilateral | 1 | Liver biopsy | |
| Alveolar damage | 1 | Microvesicular steatosis | |
| Cellular fibromyxoid exudates | 1 | Mild | |
| Desquamation of pneumocytes | 1 | Moderate | 1 |
| Hyaline membrane formation | 1 | Severe | |
| Pulmonary oedema | 1 | Lobular and portal activity | |
| Interstitial mononuclear inflammatory infiltrates | 1 | Mild | 1 |
| Multinucleated syncytial cells | 1 | Moderate | |
| Atypical enlarged pneumocytes | 1 | Severe | |
| Large nuclei | 1 | Interstitial mononuclear inflammatory infiltrates | 1 |
| Amphophilic granular cytoplasm | 1 | Heart tissue | |
| Nucleoli in the intra-alveolar spaces | 1 | No substantial damage | 1 |

computed tomography, ultrasound, and echocardiography), patient-based information (eg, complaints, feelings, history, and potential factors), clinician-based records (eg, signs, observations, and questions), and molecular measurements-based digitals (eg, biochemistry, liquid biopsies, multi-omics, and RNA and DNA sequencing). The system should also be able to automatically be optimized and updated, to build multi-dimensional models and perform deep learning, and carry out the application of clinical trans-omics for diagnosis and treatment of patients with SARS-CoV-2 infection. In addition to the role discussed above, artificial intelligence also plays an important role in other aspects. Currently, during the initial implementation, recording travel of people during the outbreak and using the internet and artificial intelligence technology to provide online consultation services based on the information obtained from pre-examination,

artificial intelligence can make a preliminary judgment of the information, and give preliminary advice and recommendations for the doctor's reference. There is also an infrared body temperature monitoring system. Artificial intelligence can automatically analyze images captured by thermal imagers and 3D laser cameras, identify and depict human facial features from the images, measure human forehead temperature, and automatically record it. The driverless delivery vehicle would also be responsible for the distribution of some medical and living materials, which is also inseparable from artificial intelligence technology. Artificial intelligence is also playing an important role in forecasting outbreaks and determining the direction of personnel outflow through data orientation and analysis, as well as assisting drug development. Such a system will be of benefit when aiming to control the outbreak of unknown or unexpected infections in the future.

This article is a preliminary retrospective analysis. When collecting information, the patient information in the raw data was insufficient to allow us to further group the patients. In conclusion, we summarized clinical phenomes of patients with SARS-CoV-2 based on published studies to emphasize the importance and specificity of those phenomes in diagnosis and treatment of infection, and evaluate the impact on medical services. Clinical phenomes of SARS-CoV-2 infection were compared with those of SARS-CoV and MERS-CoV infections. There is an urgent need to develop a system with artificial intelligence-based machine learning capacity to analyze and integrate radiomics- or imaging-based, patient-based, clinician-based, and molecular measurements-based data, to fight the outbreak of COVID-19 and enable more efficient responses to unknown infections in the future.

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CONFLICT OF INTERESTS

The authors declare that there is no conflict of interests regarding the publication of this paper.

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REFERENCES

1. Weiss SR, Navas-Martin S. Coronavirus pathogenesis and the emerging pathogen severe acute respiratory syndrome coronavirus. *Microbiol Mol Biol Rev.* 2005;69(4):635-664.

2. Cavanagh D. Coronavirus avian infectious bronchitis virus. *Vet Res.* 2007;38(2):281-297.
3. Peiris JS, Guan Y, Yuen KY. Severe acute respiratory syndrome. *Nat Med.* 2004;10(12 Suppl):S88-S97.
4. Lee J, Chowell G, Jung E. A dynamic compartmental model for the Middle East respiratory syndrome outbreak in the Republic of Korea: a retrospective analysis on control interventions and super-spreading events. *J Theor Biol.* 2016;408:118-126.
5. She J, Jiang J, Ye L, et al. 2019 Novel coronavirus of pneumonia in Wuhan, China: emerging attack and management strategies. *Clin Transl Med.* 2020;9(1):19.
6. Zhu N, Zhang D, Wang W, et al. A novel coronavirus from patients with pneumonia in China, 2019. *N Engl J Med.* 2020;382:727-733.
7. Lai CC, Shih TP, Ko WC, et al. Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) and coronavirus disease-2019 (COVID-19): the epidemic and the challenges. *Int J Antimicrob Agents.* 2020;55(3):105924.
8. Hui DS, E IA, Madani TA, et al. The continuing 2019-nCoV epidemic threat of novel coronaviruses to global health - the latest 2019 novel coronavirus outbreak in Wuhan, China. *Int J Infect Dis.* 2020;91:264-266.
9. Huang C, Wang Y, Li X, et al. Clinical features of patients infected with 2019 novel coronavirus in Wuhan, China. *Lancet.* 2020;395(10223):497-506.
10. Holshue ML, DeBolt C, Lindquist S, et al. First case of 2019 novel coronavirus in the United States. *N Engl J Med.* 2020;382:929-936.
11. Li Q, Guan X, Wu P, et al. Early transmission dynamics in Wuhan, China, of novel coronavirus-infected pneumonia. *N Engl J Med.* 2020. <https://doi.org/10.1056/NEJMoa2001316>.
12. Lei J, Li J, Li X, et al. CT imaging of the 2019 novel coronavirus (2019-nCoV) pneumonia. *Radiology.* 2020. <https://doi.org/10.1148/radiol.2020200236>.
13. Kawakami N, Namkoong H, Ohata T, et al. Clinical features of *Mycoplasma pneumoniae* pneumonia in older adults. *Geriatr Gerontol Int.* 2018;18(5):814-816.
14. Lu R, Zhao X, Li J, et al. Genomic characterisation and epidemiology of 2019 novel coronavirus: implications for virus origins and receptor binding. *Lancet.* 2020;395(10224):565-574.
15. Waites KB, Xiao L, Liu Y, et al. *Mycoplasma pneumoniae* from the respiratory tract and beyond. *Clin Microbiol Rev.* 2017;30(3):747-809.
16. Miyashita N, Kawai Y, Yamaguchi T, et al. Clinical potential of diagnostic methods for the rapid diagnosis of *Mycoplasma pneumoniae* pneumonia in adults. *Eur J Clin Microbiol Infect Dis.* 2011;30(3):439-446.
17. Ajlan AM, Ahyad RA, Jamjoom LG, et al. Middle East respiratory syndrome coronavirus (MERS-CoV) infection: chest CT findings. *AJR Am J Roentgenol.* 2014;203(4):782-787.

SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section at the end of the article.

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