

Review



# Comparison on Major Gene Mutations Related to Rifampicin and Isoniazid Resistance between Beijing and Non-Beijing Strains of *Mycobacterium tuberculosis*: A Systematic Review and Bayesian Meta-Analysis

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Abstract: Objective: The Beijing strain of Mycobacterium tuberculosis (MTB) is controversially presented as the predominant genotype and is more drug resistant to rifampicin and isoniazid compared to the non-Beijing strain. We aimed to compare the major gene mutations related to rifampicin and isoniazid drug resistance between Beijing and non-Beijing genotypes, and to extract the best evidence using the evidence-based methods for improving the service of TB control programs based on genetics of MTB. Method: Literature was searched in Google Scholar, PubMed and CNKI Database. Data analysis was conducted in R software. The conventional and Bayesian random-effects models were employed for meta-analysis, combining the examinations of publication bias and sensitivity. Results: Of the 8785 strains in the pooled studies, 5225 were identified as Beijing strains and 3560 as non-Beijing strains. The maximum and minimum strain sizes were 876 and 55, respectively. The mutations prevalence of *rpoB*, *katG*, *inhA* and *oxyR-ahpC* in Beijing strains was 52.40% (2738/5225), 57.88% (2781/4805), 12.75% (454/3562) and 6.26% (108/1724), respectively, and that in non-Beijing strains was 26.12% (930/3560), 28.65% (834/2911), 10.67% (157/1472) and 7.21% (33/458), separately. The pooled posterior value of OR for the mutations of rpoB was 2.72 ((95% confidence interval (CI): 1.90, 3.94) times higher in Beijing than in non-Beijing strains. That value for katG was 3.22 (95% CI: 2.12, 4.90) times. The estimate for inhA was 1.41 (95% CI: 0.97, 2.08) times higher in the non-Beijing than in Beijing strains. That for oxyR-ahpC was 1.46 (95% CI: 0.87, 2.48) times. The principal patterns of the variants for the mutations of the four genes were *rpoB* S531L, *katG* S315T, *inhA*-15C > T and oxyR-ahpC intergenic region. Conclusion: The mutations in rpoB and katG genes in Beijing are significantly more common than that in non-Beijing strains of MTB. We do not have sufficient evidence to support that the prevalence of mutations of *inhA* and *oxyR-ahpC* is higher in non-Beijing than in Beijing strains, which provides a reference basis for clinical medication selection.

**Keywords:** *Mycobacterium tuberculosis;* Beijing and non-Beijing strain; mutation of gene; MDR; rifampicin and isoniazid

# 1. Introduction

Tuberculosis (TB) is one of the deadliest transmissible diseases that cause death worldwide. However, only 10% of people infected with *Mycobacterium tuberculosis* (MTB) develop TB disease [1], indicating that either the host or the pathogen's genetic factors may play a critical role in determining the occurrence of TB disease. The Beijing strain of MTB is presented as the predominant strain. It plays a vital role in many countries, such as Bangladesh [2], Upper Myanmar [3] and China [4,5], with the Beijing strain accounting for 26.8%, 71.4% and 81.7%, respectively. The latter country, China, holds the second highest tuberculosis (TB) burden, presenting 8.5% of case notifications worldwide [6].



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**Copyright:** © 2022 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https:// creativecommons.org/licenses/by/ 4.0/). The Beijing strain of MTB is reported to be more virulent, more pathogenic, and fastergrowing, with more histopathological changes and drug resistance, especially multidrugresistance TB (MDR-TB) tendencies, than other strains, leading to a higher mortality rate [7]. The rate of treatment success in MDR-TB remains low, reaching only 47–62.7% [2,8]. MDR TB is not only a severe clinical and epidemiological problem but also entails substantial economic costs of management.

Thus, treating patients with resistance to the main anti-TB agents, such as rifampicin (RIF) and isoniazid (INH), may be many times more expensive compared to treatment costs incurred by the management of TB susceptible to the main medication panel [9]. MDR-TB poses a significant threat not only to the individual faced with diminished chances of cure compared with non-MDR-TB, but also to the community, as outbreaks of MDR-TB have been shown to have devastating consequences [10].

Furthermore, some studies have suggested an association between drug resistance and some MTB genotypes [11–13]. Resistance to anti-TB drugs in MTB mainly arises from genomic mutations in genes encoding either the drug target or enzymes involved in drug activation [14,15]. Even some efflux pump genes, such as *drrA*, *drrB*, *efpA*, *Rv2459*, *Rv1634*, *and Rv1250* [16], were also reported to be related to the resistance of MDR; however, some previous studies suggested the more common candidate genes' mutations to be related to MDR [10,17–19], such as the *rpoB* gene is associated with rifampicin, and *katG*, *inhA*, and *ahpC* genes are related to isoniazid resistance [10]. Other genes mutations related to drug resistance are also reported, such as *rpsL* K43R to streptomycin, *embB* M306V to Ethambutol, *pncA* promoter T (-11) C to pyrazinamide, *gyrA* A90V to fluoroquinolones, *RRS* A1401G to second-line injection drug, and *fabGl\_*promoter C(-15) T to Ethionamide) [20].

It is addressed that 95% of rifampicin resistance (RR) is associated with the mutation in the 81 bp rifampicin resistance determining region (RRDR) [21]. Resistance mutations in RRDR of the *rpoB* gene were found to be associated with phenotypic RIF resistance. [22]. The *rpoB* gene codes the  $\beta$ -subunit of DNA-dependent RNA-polymerase, which acts as a major target for RIF, and up to 95–98% of RIF-resistance strains exhibit mutations in the *rpoB* gene, whereas 90–95% of these mutations are located in RRDR [8,23].

INH resistance is associated with mutations in multiple loci, such as the catalaseperoxidase gene (*katG*), the enoyl-ACP reductase gene (*inhA*) and its promoter, the alkyl hydroperoxide reductase gene (*ahpC*), and the intergenic region between the *oxyR* and *ahpC* (*oxyR-ahpC*) genes, which is distinguished from that of RIF [24–26]. One specific *KatG* variant, S315T, is found in 94% of INH-resistance clinical isolates. Around 15 mutations in *inhA* have been identified in INH-resistance clinical isolates, although two of them were also found in INH-sensitive strains. In this regard, the analysis of gene expression profiling of the Beijing strain of MTB can give us a snapshot of actively expressed genes under various conditions, even though some other researchers hold the opposite issue [27].

Due to the discrepancies between studies possibly resulting from the small sample sizes and variant detection methods of genes in different areas, pooled evidence is needed to provide better evidence that inform policymakers' decisions for controlling TB. Bayesian meta-analysis (BMA) is reported that it harbors more robustness [28] than the conventional meta-analysis (MA) and is not limited to the premise of classical statistical methods, which can be combined with a priori information, sample information and general information, can obtain the posterior distribution easily and is based on its effect quantity variance between the mergers of the values, research, other parameters, and 95% CI, e.g., the shrinkage estimation values with the consideration of the potential publication biases. It is believed that Bayesian statistical methods will be more widely used in evidence-based medicine/meta-analyses [28].

This systematic review focused on combining the results about genes relevant to MDR with the concepts of the classifications of Beijing and non-Beijing using conventional meta-analysis MA and BMA. We aimed to compare the major gene mutations related to RIF and INH resistance between Beijing and non-Beijing genotypes and extract the best

evidence using evidence-based methods for improving the TB control program's service based on the genetics of MTB.

### 2. Methods

## 2.1. Study Design

This systematic review and Bayesian meta-analysis were conducted according to the Preferred Reporting Items for Systematic Reviews and Meta-Analyses Statement (PRISMA, http://links.lww.com/SLA/C529, accessed on 5 September 2022) (Supplementary File S1 PRISMA checklist) [29] and the meta-analyses of observational studies in epidemiology guidelines [30,31]. Bayesian meta-analysis is performed using Bayesian methods, which provide a profitable opportunity for flexible modeling of inter-study heterogeneity by mildly regularizing priors to obtain a stable estimation, which frequency models prove impossible to calculate [32,33].

### 2.2. Literature Search Strategy

To ensure that a piece of relevant contemporary information was obtained [31], limits were applied to years 1960 onward and MTB genetics or clinic research related to MDR, or RIF/INH drug resistance. Eventually, a retrieval of literature relating the genetics from 1 January 1960 to the present was performed.

Search engines: Google Scholar, PubMed, ResearchGate, ResearchGate, Cochrane Library and Chinese National Knowledge Infrastructure (CNKI) Database.

Search terms: MTB AND Beijing AND non-Beijing AND gene mutation AND MDR, or RIF, or INH drug resistance; *rpoB* mutation AND Beijing AND non-Beijing AND MDR, or RIF; *katG* mutation AND Beijing AND non-Beijing AND MDR, or INH; *inhA* mutation AND Beijing AND non-Beijing AND MDR, or INH; *oxyR-ahpC* mutation AND Beijing AND non-Beijing AND MDR, or INH.

### 2.3. Study Selection Criteria

Inclusion criteria: (1) Full article, abstract, letter presenting the major gene mutations related to MDR of MTB classified as Beijing and non-Beijing strains written in English or Chinese; and (2) Gray literature related to the first point above, which is a kind of information produced outside of traditional publishing and distribution channels, and can include reports, policy literature, working papers, newsletters, government documents, speeches, white papers, urban plans, and so on, written in English or Chinese [34].

Exclusion criteria: (1) Studies only related to the genes of MTB produced by the contacts of the studied subjects or produced by the same subject but obtained through follow-up; (2) studies with drug susceptibility test (DST) involving rifampicin and/or isoniazid only related to children analyzed; and (3) studies conducted in unique sites, such as prisons and asylums.

### 2.4. Data Extraction

Screening of studies and all essential data from the included studies meeting the inclusion criteria were extracted by the investigators (S.G. and V.C.). The principal mutations of the four genes, *rpoB*, *katG*, *inhA* and *oxyR-ahpC*, of MTB related to RIF and INH were input into a predesigned Excel sheet. The results were compared electronically according to the two classification variables, Beijing and not-Beijing strains. The place with a supposed gene absence was labeled with "NA" in the Excel sheet.

The study content recorded the data related to the surname of the first author, country of the subjects, date of publication, study design, sample size, and frequency in the relevant sheet (Supplementary File S2).

Any records with discrepancies were resolved by referring to the source articles. Discrepancies between the two reviewers were resolved by consensus involving all the authors. The R package *metagear* [35,36] was performed for the initial screening articles for the literature review.

## 2.5. Data Synthesis and Statistical Analysis

All analyses were conducted using R software (version 3.6.3) with the following packages, epicalc, medorator and Bayesmeta [28]. The Bayesian random-effects model was used for Bayesian meta-analysis [28,37]. Significant heterogeneity between studies would be considered the presence of heterogeneity when the *p* value is less than 0.05 or  $I^2$  is greater than 50%.

The leave-one-out [38] and influence sensitivity analyses were also employed by iteratively removing one study at a time while recalculating the odds ratio (OR) to assess the robustness of the pooled values to explore potential sources of inter-study heterogeneity and to further determine the influence of each study, from which the preprint studies had been excluded.

Subgroup analysis was employed for the three groups according to the regions, East Asia, South/Southeast/West/Central Asia and East/North/Central Europe. Potential publication bias was also assessed by the funnel plot, tests of Egger's liner weighted regression [39] and Begg [40]. Asymmetry of the collected studies' distribution by visual inspection or p value is less than 0.05 was considered as statistically significant [41], indicating the presence of a publication bias evaluated by weight-function. Duval and Tweedie's trim and fill method's assumption was considered to reduce the bias in the pooled estimates [42]. To make it more profitable to interpret, logarithms were converted into corresponding constants where appropriate.

# 3. Results

# 3.1. Literature Search Results

In the initial literature search, 1733 relevant articles were identified. After removing 871 duplicates and 573 articles from primary screening, 198 full-text articles were assessed for eligibility in the meta-analysis. Of these, 91 were excluded due to a paucity of sufficient data. Eventually, a total of 134 articles published between 1 January 1960 and 5 March 2022 were included in the quality review part and 31 in the Bayesian meta-analysis part (Figure 1).



# Article Screening & Identification

Figure 1. Flow diagram of literature screening and identification strategy.

### 3.2. Characteristics of Studies Included

As described in Table 1, a total of 31 studies were included in the final Bayesian meta-analysis. The literature has a relatively wide global range covering Asia (China, Korea, Japan, Thailand, Indonesia, Kyrgyzstan, Bangladesh, India, Iran and Turkey) and Europe (Germany, Latvia, Russia, Ukraine and Sweden). Notably, the proportion of studies conducted in China accounted for the majority.

Sample Size by rpoB-Rif katG-INH inhA-INH oxyR-ahpC-INH Isolate Genotypes Author Country No. Year Sample Non Non-Non-Non-Non-Beijing Beijing Beijing Beijing Beijing Size Beijing <mark>Beijing</mark> Beijing Beijing Beijing Asian NA NA NA Qian NA NA NA Countries Tracevska NA NA NA NA Latvia Toungoussova Russia NA NA NA NA NA NA Park Korea Hillemann Germany outherr NA NA Nikolavevskyv Ukraine Cheunoy Thailand NA NA NA NA NA NA Parwati Indonesia NA NA NA NA H11 China NA NA NA Mäkinen Russia NA NA NA Li China NA NA NA Ma C NA China Y11 China NA NA Mokrousov Kyrgyzstar Zhang China Zhao China NA Vyazovaya NA Russia NA NA NA NA Kisa Turkey Hong China NA NA Wang China NA NA Russia Figueroa NA NA Liu China Bangladesh Uddin NA NA NA NA Wan China Gupta NA NA India NA NA NA NA NA Russia NA Vyazovaya NA NA Gao China China NA NA NA Luo NA Luo China Ghebremichael NA NA Sweden NA NA Khosravi NA NA NA Iran 

Table 1. Characteristics of studies included in the systematic review and Bayesian meta-analysis.

All included studies described critical elements of study design, including study setting, data source, inclusion criteria, participant selection and statistical methods. No studies explained the solution to the missing values, mentioned sample size calculation, or conducted subgroup analysis based on region (Table 1).

# 3.3. Mutations Prevalence for Mutations of Genes

Globally, 8785 pooled MTB isolates were tested to identify MDR-TB, RIF and INH resistance patterns, with 5225 identified as Beijing strains and 3560 as non-Beijing strains. The maximum sample size was 876 strains, and the minimum one was 55 isolates. The prevalence of mutations for *rpoB*, *katG*, *inhA* and *oxyR-ahpC* in Beijing strains was 52.40% (2738/5225), 57.88% (2781/4805), 12.75% (454/3562) and 6.26% (108/1724), respectively; and that in non-Beijing strains was 26.12% (930/3560), 28.65% (834/2911), 10.67% (157/1472) and 7.21% (33/458), separately. The principal variants for the four genes were *rpoB* Ser531Leu, *katG* S315T, *inhA*-15C > T and *oxyR-ahpC* intergenic region, respectively (Table 2).

C	Sample Size		Mutation Isolates		Muta	ation Rates	Principal Mutations	
Genes	Total	Beijing	Non-Beijing	Beijing	Non-Beijing	Beijing	Non-Beijing	Pattern
rpoB	8785	5225	3560	2738	930	52.40	26.12	rpoB Ser531Leu
katG	7716	4805	2911	2781	834	57.88	28.65	katG S315T
inhA	5034	3562	1472	454	157	12.75	10.67	inhA -15 C > T, promoter region of inhA
oxyR-ahpC	2182	1724	458	108	33	6.26	7.21	oxyR-ahpC intergenic region

**Table 2.** Mutations prevalence for mutations of *rpoB*, *katG*, *inhA* and *oxyR-ahpC*.

# 3.4. Publication Bias and Sensitivity Analyses

The symmetrical distributions of the funnel plots were detected when the publication biases were evaluated for all the mutations of *rpoB*, *katG*, *inhA* and *oxyR-ahpC* among Beijing and non-Beijing strains, paralleled with the p > 0.05 of both Egger and Begg tests, indicating the absence of the publication biases. The robustness was detected after sensitivity analysis using leave-one-out and influence tests (Figure 2A–D).



Figure 2. Funnel plots for publication bias tests ((A), rpoB; (B), katG; (C), inhA; (D), oxyR-ahpC).

3.5. Mutations of Major Genes in Beijing and Non-Beijing Strains

Of the 31 studies, 31 studies were evaluated for the mutations of *rpoB*, 27 studies for the mutations of *katG*, 18 studies for the mutations of *inhA* and 9 studies for that of *oxyR*-*ahpC* [5,10,15,17,20–22,24,43–69]. The subgroup analysis was conducted for the mutations of *rpoB*, *katG* and *inhA* instead of the mutations of *oxyR*-*ahpC* because only a few studies were included to be analyzed for the latter. All the ORs were assessed using the Bayesian meta-analysis as well.

The pooled posterior value of OR for the mutations of *rpoB* (100% mutated in locus *rpoB* S531L) was [exp (log1.00)] = 2.72 ((95% confidence interval (CI): 1.90, 3.94) times higher in Beijing than in non-Beijing strains for all 31 studies included that evaluated the mutations of *rpoB*, analogous with the value of the conventional pooled OR (2.76), with a statistical significance being found in east subgroup analysis. Meanwhile, the combined heterogeneity was detected ( $I^2 = 83.5\%$ ), and a prediction interval for the effect as [exp (log1.00)] = 2.72 (95% CI: 0.50, 15.03), meaning there would be an OR of 2.72 for the same indicators for the 32nd ( $\theta_{k+1}$ ) study in the future [28] (Figures 3 and 4).

### quoted estimate shrinkage estimate

study	estimate	95% CI	
Qian (2002)	0.36	[-1.25, 1.96]	
Tracevska (2003)	1.18	[-0.36, 2.73]	
Toungoussova (2004)	0.09	[-1.01, 1.18]	
Park (2005)	0.40	[0.03, 0.77]	===
Hillemann (2005)	0.41	[-3.53, 4.35]	<b>_</b>
Nikolayevskyy (2007)	1.14	[0.57, 1.72]	-
Cheunoy (2009)	-0.02	[-0.96, 0.93]	
Parwati ( 2009 )	0.21	[-0.27, 0.70]	
HU ( 2010 )	1.10	[0.37, 1.83]	
Makinen (2011)	2.14	[1.62, 2.67]	<del></del>
Li ( 2016 )	2.49	[1.47, 3.51]	
Ma (2011)	0.79	[0.05, 1.54]	
Yu(2013)	0.38	[-1.10, 1.85]	
Mokrousov (2013)	1.44	[0.22, 2.67]	
Zhang ( 2015 )	2.57	[1.40, 3.75]	
Zhao(2015)	2.29	[-0.97, 5.55]	
Vyazovaya(2015)	3.03	[1.80, 4.26]	
Kisa ( 2012 )	0.97	[-0.63, 2.57]	
Hong (2020)	-0.09	[-0.61, 0.43]	
Wang ( 2018 )	0.86	[-0.09, 1.82]	
Figueroa (2018)	1.00	[0.34, 1.67]	<b>—</b>
Liu ( 2020 )	-1.54	[-4.39, 1.32]	
Uddin ( 2020 )	-0.36	[-4.29, 3.57]	
Wan (2020)	6.78	[3.91, 9.64]	
Gupta ( 2020 )	1.17	[0.62, 1.72]	-
Vyazovaya (2020)	2.45	[1.46, 3.44]	<b></b>
Gao ( 2020 )	0.07	[-0.31, 0.45]	
Luo ( 2021 )	0.27	[-0.19, 0.72]	
Luo (2019)	-0.40	[-1.48, 0.68]	
Ghebremichael (2010)	0.73	[0.11, 1.34]	-
Khosravi (2014)	2.11	[0.71, 3.52]	
mean	1.00	[0.64, 1.37]	•
prediction	1.00	[-0.69, 2.71]	
Heterogeneity (tau): 0.81 [0	).54, 1,13]		-4 -3 -2 -1 0 1 2 3 4 5 6 7 8 9 10 Log Odds Ratio

**Figure 3.** BMA for the mutations of *rpoB* in Beijing and Non-Beijing Strains (CI: confidence interval; BJ: Beijing strain; Non-BJ: non-Beijing strain).

The converged posterior value of OR for the mutations of *katG* (100% mutated in locus *katG* S531T) was [exp (log1.17)] = 3.22 (95% CI: 2.12, 4.90) times higher in Beijing than in non-Beijing strains for all the 27 studies included. The mutations of *katG*, comparable to the value of the pooled OR (3.26) obtained through the traditional meta-analysis, with a significant difference were found in each subgroup analysis. Simultaneously, the combined heterogeneity was detected ( $I^2 = 90.8\%$ ), and a prediction interval for effect as [exp (log1.17)] = 3.22 (95% CI: 0.42, 24.53) was found for the 28th study in the future (Figures 5 and 6).

	Subli	neage(n)	rpoB	mutation(n)		
Author	BJ	Non-BJ	In BJ	In non-BJ		Odds Ratio [95% CI]
South/Southeast/West/Cen Khosravi (2014) Gupta (2020) Uddin (2020) Kisa (2012) Mokrousov (2013) Parwati (2009) Cheunoy (2009)	tral As 8 76 84 6 62 273 50	<i>ia</i> 152 305 121 89 41 545 26	4 29 84 4 17 29 21	16 49 121 ◀ 36 3 48 11		$\begin{array}{cccccccccccccccccccccccccccccccccccc$
RE Model for Subgroup (Q = 14. <b>East/North/Central Europe</b> Ghebremichael (2010) Vyazovaya (2020) Figueroa (2018) Vyazovaya (2015) Makinen (2011) Nikolayevskyy (2007) Hillemann (2005) Toungoussova (2004) Tracevska (2003)	.60, df = 70 73 130 80 184 89 62 24 63	6, p = 0.02 466 57 49 27 255 136 41 31 46	; I <sup>2</sup> = 60.2 16 40 90 60 83 43 62 16 61	%, $\tau^2 = 0.32$ ) 59 5 22 3 22 31 41 41 20 41		$\begin{array}{cccccccccccccccccccccccccccccccccccc$
RE Model for Subgroup (Q = 31.	.41, df =	8, p < .01;	l <sup>2</sup> = 77.9%	6, τ <sup>2</sup> = 0.61)	-	4.16 [2.23, 7.74]
East Asia						
Luo (2019) Luo (2021) Gao (2020) Wan (2020) Wang (2018) Hong (2020) Zhao (2015) Zhang (2015) Yu (2013) Ma (2011) Li (2016) HU (2010) Park (2005) Qian (2002)	$157 \\ 409 \\ 749 \\ 141 \\ 256 \\ 378 \\ 44 \\ 261 \\ 78 \\ 243 \\ 156 \\ 243 \\ 569 \\ 50 \\ 100 \\ 1$	25 312 42 16 20 69 14 15 7 108 20 108 108 174 16	120 55 437 141 138 132 216 44 258 51 42 132 54 214 45	21 33 72 10 6 41 13 98 4 9 6 9 50 14		$\begin{array}{ccccccc} 0.67 & [0.23, & 1.98] \\ 1.31 & [0.83, & 2.06] \\ 1.07 & [0.73, & 1.56] \\ 875.95 & [50.04, 15334.32] \\ 0.22 & [0.01, & 3.73] \\ 2.37 & [0.91, & 6.18] \\ 0.91 & [0.54, & 1.54] \\ 9.89 & [0.38, & 257.09] \\ 13.12 & [4.07, & 42.32] \\ 1.46 & [0.33, & 6.34] \\ 2.21 & [1.05, & 4.65] \\ 12.06 & [4.35, & 33.48] \\ 3.01 & [1.45, & 6.25] \\ 1.49 & [1.03, & 2.15] \\ 1.43 & [0.29, & 7.12] \end{array}$
RE Model for Subgroup (Q = 65.	.55, df =	= 14, p < .01	; I <sup>2</sup> = 89.3	%, τ <sup>2</sup> = 1.14)		2.39 [1.27, 4.48]
RE Model for All Studies (Q = Test for Subgroup Differences: C	= 147.80 Q <sub>M</sub> = 0.4	0, df = 30,   44, df = 1, p	p < .01; I = 0.505	$r^{2} = 83.5\%, \tau^{2} = 0$	0.75) • • • • • • • • • • • • • • • • • • •	2.76 [1.91, 4.00]

**Figure 4.** MA for the mutations of *rpoB* in Beijing and non-Beijing Strains (CI: confidence interval; BJ: Beijing strain; Non-BJ: non-Beijing strain).

The summarized posterior value of OR for the mutations of *inhA* (100% mutated in *inhA* -15 C > T) was  $[1/\exp(\log - 0.34)] = 1.41$  (95% CI:  $1/\exp(\log 0.03) = 0.97$ ,  $1/\exp(\log - 0.73) = 2.08$ ; the following algorithm is the same) times higher in the non-Beijing than in the Beijing strains for all 18 studies included that evaluated the mutations of *inhA*, with a significant difference found in the East/North/Central Europe group. Although the pooled posterior value of the OR between BMA and MA are close (1/0.71 vs. 1/0.70), the values of the 95% CIs of both diverted with the marginal significance, which was more obvious rather in the BMA compared to that of the MA (OR = 1.43 (95% CI: 0.95, 2.13)). Furthermore, a combined heterogeneity was detected ( $I^2 = 63.3\%$ ), and a prediction interval for effect as 1.41 (95% CI: 0.41, 4.90) was found for the 19th study in the future (Figures 7 and 8).

- 4 4 -	41 4 -	0.54/ 01	
study	estimate	95% CI	_
Tracevska (2003)	0.31	[-3.63, 4.25]	<b>_</b>
Park (2005)	0.72	[0.35, 1.09]	
Hillemann (2005)	1.73	[0.45, 3.02]	
Nikolayevskyy (2007)	0.97	[0.42, 1.52]	
Cheunoy(2009)	-0.05	[-1.02, 0.93]	
HU ( 2010 )	1.36	[0.66, 2.05]	
Makinen (2011)	2.37	[1.84, 2.90]	_ <b>_</b>
Li ( 2016 )	2.23	[1.24, 3.22]	<del></del>
Ma ( 2011 )	0.08	[-0.51, 0.66]	-
Yu ( 2013 )	0.40	[-1.06, 1.87]	
Mokrousov (2013)	0.42	[-0.52, 1.35]	
Zhang ( 2015 )	0.66	[0.21, 1.10]	-
Zhao ( 2015 )	1.12	[-0.09, 2.32]	
Vyazovaya (2015)	3.02	[1.95, 4.10]	
Kisa (2012)	0.97	[-0.63, 2.57]	
Hong (2020)	0.98	[0.44, 1.52]	=
Wang (2018)	0.63	[-0.29, 1.56]	
Figueroa (2018)	1.27	[0.59, 1.96]	<b>_</b> _
Liu (2020)	0.27	[-0.74, 1.28]	
Uddin ( 2020 )	0.79	[0.07, 1.51]	
Wan ( 2020 )	5.58	[4.10, 7.05]	• •
Gupta (2020)	1.41	[0.89, 1.93]	_ <b>_</b>
Gao ( 2020 )	0.49	[0.09, 0.89]	
Luo ( 2021 )	-0.04	[-0.43, 0.36]	
Luo (2019)	-0.61	[-1.51, 0.30]	
Ghebremichael (2010)	2.44	[1.89, 2.99]	_ <del></del>
Khosravi (2014)	3.37	[1.81, 4.94]	<b></b>
mean	1.17	[0.75, 1.59]	•
prediction	1.17	[-0.86, 3.20]	
Heterogeneity (tau): 0.98 [0	0.69, 1.31]		-4 -3 -2 -1 0 1 2 3 4 5 6 7 Log Odds Ratio

	quoted	estim ate	+	shrinkage estimate	
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Figure 5. BMA for the mutations of *katG* in Beijing and non-Beijing strains.

Author	Subl	ineage (n)	katG m	utation (n)			Odde Batic 105% Cll
Autio	BJ	Non-BJ	In BJ	In non-BJ			
South/Southeast/West/Cent	tral As	ia					
Khosravi (2014) Gupta (2020) Uddin (2020) Kisa (2012) Mokrousov (2013) Cheunoy (2009) RE Model for Subgroup (Q = 17.45,	8 76 84 6 62 50 df = 5, p	152 305 121 89 41 26 o < .01; I <sup>2</sup> = 77	6 43 72 4 17 32 7.8%, τ <sup>2</sup> = 0.73	12 73 88 36 8 17		≜ ₽ ₽ ₽	29.22 [ 6.09, 140.35] 4.11 [ 2.44, 6.92] 2.20 [ 1.07, 4.51] 2.64 [ 0.53, 13.10] 1.52 [ 0.60, 3.85] 0.95 [ 0.36, 2.53] 2.86 [1.28, 6.39]
East/North/Central Europe Ghebremichael (2010) Figueroa (2018) Wazovaya (2015) Makinen (2011) Nikolayevskyy (2007) Hillemann (2005) Tracevska (2003) RE Model for Subgroup (Q = 25.94,	70 130 80 184 89 62 63 df = 6, p	466 49 27 255 136 41 46 0 < .01; l <sup>2</sup> = 76	44 99 71 91 52 59 63 6.2%, τ <sup>2</sup> = 0.46	59 23 21 47 31 46		◆ <u>1</u> 1, • <u>1</u> ,	
East Asia Luo (2019) Luo (2021) Gao (2020) Wan (2020) Wang (2018) Hong (2020) Zhao (2015) Zhang (2015) Yu (2015) Ma (2015) Ma (2011) Li (2016) HU (2010) RE Model for Subgroup (Q = 82.10,	157 409 749 157 256 378 44 263 243 243 243 243 243 243 243 243 243 24	25 312 127 42 16 20 69 14 115 7 108 20 108 174 p < .01; l <sup>2</sup> = 9	90 560 139 98 131 301 31 173 46 131 71 250 93.7%, t <sup>2</sup> = 1.4	18 82 9 41 658 3 9 7 41 658 3 9 7 10 48		┈╹┑╹╷┿╋╋┸╄┶┲┶ ┺	$\begin{array}{c} 0.54 & [ \ 0.22 & 1.34] \\ 0.97 & [ \ 0.65 & 1.44] \\ 1.63 & [ \ 1.10 & 2.43] \\ 264. & 12.160.24 & 1157.96] \\ 1.31 & [ \ 0.48 & 3.59] \\ 1.89 & [ \ 0.76 & 4.76] \\ 2.67 & [ \ 1.56 & 4.58] \\ 3.05 & [ \ 0.92 & 1.56] \\ 1.93 & [ \ 1.23 & 3.04] \\ 1.08 & [ \ 0.66 & 1.94] \\ 1.08 & [ \ 0.66 & 1.94] \\ 3.89 & [ \ 1.94 & 7.78] \\ 2.04 & [ \ 1.44] \\ 3.29 & [ \ 1.94 & 7.78] \\ 2.04 & [ \ 1.44] \\ 2.25 & [ \ 1.28 & 4.90] \end{array}$
RE Model for All Studies (Q =	190.1	1, df = 26, p	o < .01; I <sup>2</sup> =	90.8%, τ <sup>2</sup> = 1	.16)	•	3.26 [ 2.09, 5.09]
rescior Subgroup Differences. Q <sub>M</sub> =	0.008,0	ui – i, p = 0.9	30				
				0.05	0.25 1	4	
				Odd	ls Ratio (log s	cale)	

**Figure 6.** MA for the mutations of *katG* in Beijing and non-Beijing Strains (CI: confidence interval; BJ: Beijing strain; Non-BJ: non-Beijing strain).

study	estimate	95% CI	
Park(2005)	-0.37	[-0.86, 0.12]	
Hillemann ( 2005 )	-0.95	[-3.02, 1.11]	<b>_</b>
Nikolayevskyy ( 2007 )	-0.51	[-1.32, 0.30]	
Cheunoy ( 2009 )	-0.43	[-1.91, 1.05]	<b>Ę</b>
Li ( 2016 )	0.88	[-0.49, 2.25]	<b>_</b>
Yu ( 2013 )	-1.24	[-4.52, 2.05]	
Mokrousov (2013)	-1.07	[-2.66, 0.53]	
Zhang(2015)	-0.14	[-0.70, 0.42]	
Zhao ( 2015 )	1.16	[-0.33, 2.64]	<b></b>
Vyazovaya(2015)	-0.20	[-1.30, 0.90]	
Hong ( 2020 )	-0.04	[-1.08, 1.00]	
Wang ( 2018 )	0.27	[-1.10, 1.63]	
Figueroa ( 2018 )	-0.88	[-1.59, -0.17]	<b>_</b>
Liu ( 2020 )	0.04	[-1.20, 1.27]	<b></b>
Wan(2020)	-3.74	[-5.13, -2.35]	
Vyazovaya(2020)	-0.26	[-1.78, 1.27]	
Gao ( 2020 )	0.16	[-0.41, 0.73]	
Luo ( 2019 )	-0.38	[-1.41, 0.66]	
mean	-0.34	[-0.73, 0.03]	•
prediction	-0.34	[-1.59, 0.88]	
	10 004 0 0001		-5 -4 -3 -2 -1 0 1 2 3

quoted estimate shrinkage estimate

Heterogeneity (tau): 0.513 [0.021, 0.920]

-4 -3 **Figure 7.** BMA for the mutations of *inhA* in Beijing and non-Beijing strains.

-2 -1 0 Log Odds Ratio

1

	Sublineage (n)		inhA r	nutation (n)		
Author	BJ	Non-BJ	In BJ	In non-BJ		Odds Ratio [95% Cl
South/Southeast/West	Central Asi	a				
Mokrousov (2013)	62	41	2	4 ⊢		0.34 [0.07, 1.70
Cheunoy ( 2009 )	50	26	4	3	<b>⊢</b> −−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−−	0.65 [0.15, 2.86
RE Model for Subgroup (Q =	0.33, df = 1, p =	0.57; I <sup>2</sup> = 0.0	$0\%, \tau^2 = 0.00$	)		0.48 [0.16, 1.44
East/North/Central Eur	оре					
Vyazovaya (2020)	73	57	3	3	<b>⊢−−−</b> ∎ <u>↓</u> −−−−1	0.77 [0.17, 3.55
Figueroa (2018)	130	49	27	19	<b>⊢</b> − <b>■</b> −−1	0.42 [0.20, 0.84
Vyazovaya (2015)	80	27	13	5	<b>⊢</b>	0.82 [0.27, 2.46
Nikolayevskyy (2007)	89	136	9	22	<b>⊢</b>	0.60 [0.27, 1.35
Hillemann ( 2005 )	62	41	1	2 🖛		0.39 [0.05, 3.03
East Asia	157	25	24	5	<b></b>	0.68 [0.24 1.93
East Asia	157	05		-		0.0010.01 4.00
Gao (2020)	7/9	127	104	15		1 18 [0.66 2.08]
Wan (2020)	1/1	12	2	18 -		0.02 [0.01 0.10
Liu (2020)	157	16	33	3		1 04 [0 30 3 57
Wang (2018)	256	20	38	2		1.30 [0.33, 5.10
Hong (2020)	378	69	23	4		0.96 [0.34, 2.73
Zhao (2015)	44	14	17	2		3 18 [0.72, 14.04
Zhang (2015)	261	115	45	22	<b>F</b>	0.87 [0.50, 1.53
Yu (2013)	78	7	1	0 -		0.29 [0.01, 7.77
Li ( 2016 )	156	20	38	2		2.40 [0.61, 9.46
Park ( 2005 )	569	174	62	26	<b>⊢</b> ∎→	0.69 [0.42, 1.13
RE Model for Subgroup (Q =	33.79, df = 10,	p < .01; l <sup>2</sup> = 8	2.7%, τ <sup>2</sup> = 1.0	00)	-	0.78 [0.39, 1.57
RE Model for All Studies (Q = Test for Subgroup Differences	38.73, df = 17, ∷ Q <sub>M</sub> = 0.001, d	p < .01; l <sup>2</sup> = 6 f = 1, p = 0.97	63.3%, τ <sup>2</sup> = 0	.41)	•	0.70 [0.47, 1.05
				0.05	0.25 1 4	
				Od	ds Ratio (log scale)	

Figure 8. MA for the mutations of *inhA* in Beijing and non-Beijing strains (CI: confidence interval; BJ: Beijing strain; Non-BJ: non-Beijing strain).

The pooled posterior value of OR for the mutations of *oxyR-ahpC* (100% mutated in *oxyR-ahpC* intergenic region) was  $[1/\exp(\log - 0.38)] = 1.46$  (95% CI:  $1/\exp(\log - 0.14) = 0.87$ ,  $1/\exp(\log - 0.91) = 2.48$ ) times higher in the non-Beijing than in the Beijing strains for all nine studies included that evaluated the mutations of *oxyR-ahpC*, without any statistical significances found, neither in BMA nor in MA (OR = 1.45, 95% CI: 0.94, 2.22). A homogeneity ( $I^2 = 0.0\%$ ) and a prediction interval for the effect as 1.46 (95% CI: 0.59, 3.71) were found for the seventh in a future study were identified (Figures 9 and 10).

quoted estimate + shrinkage estimate



Figure 9. BM for the mutations of *oxyR-ahpC* in Beijing and non-Beijing strains.

	Sublineage(n)		oxyR.ah	pC mutation(n)				
Author	BJ	Non-BJ	In BJ	In non-BJ	Odds Ratio		Odds Ratio [95% Cl]	
Park ( 2005 )	569	174	8	6	⊢		0.39 [0.14, 1.11]	
Hillemann (2005)	62	41	0	2 ┥			⊣ 0.13 [0.01, 2.70]	
Yu ( 2013 )	78	7	4	0 🖛		•	→0.91 [0.04, 18.50]	
Zhang ( 2015 )	261	115	32	16	F		0.85 [0.45, 1.62]	
Zhao ( 2015 )	44	14	3	3	-		0.28 [0.05, 1.40]	
Li ( 2016 )	156	20	21	1	<b>—</b>		▶2.06 [0.37, 11.54]	
Wang ( 2018 )	256	20	21	1	<b> </b>	-	→ 1.19 [0.21, 6.62]	
Luo ( 2019 )	157	25	17	4			0.60 [0.19, 1.84]	
Wan(2020)	141	42	2	0			→1.52 [0.07, 32.35]	
RE Model for All Studies (Q = 6.26, df = 8, p = 0.62; $l^2 = 0.0\%$ , $\tau^2 = 0.00$ ) 0.69 [0.45, 1.06]								
				0.05	0.05			
				0.05	0.25	1	4	
				0	das Ratio (la	og scale)		

**Figure 10.** BMA&MA for the mutations of *oxyR-ahpC* in Beijing and non-Beijing strains (CI: confidence interval; BJ: Beijing strain; Non-BJ: non-Beijing strain; OxyR.C: *oxyR-ahpC*).

### 4. Discussion

Heterogeneities were identified by both BMA and MA in most mutations of the genes, and no publication biases were detected. Mutations of *rpoB* and *katG* related to RIF and INH were significantly more common in Beijing than in non-Beijing strains, which were not identified in the mutations of *inhA* and *oxyR-ahpC*. There was not enough evidence to demonstrate that the mutations of *inhA* and *oxyR-ahpC* were higher in the non-Beijing than in the Beijing strains.

RRDR, the so-called "hot" locus of the *rpoB* gene (81-b.p., codon 507–533) harbors around 98% of gene mutations related to RIF drug resistance [8,23]. Compared to the mutations of *katG*, which were more prevalent in European countries, combined with the evidence exhibited in the Beijing and non-Beijing strains in this study, the mutations of *rpoB* were more common in Asian countries. This is equivalent to the finding of Anwaiejiang (isolates collected in China). Despite such miscellaneous mutation locations, most of them are harbored in three *rpoB* codons: 531, 526 and 516 [9]. In this current meta-analysis, 100% of the mutations of *rpoB* were presented in the pattern of S531L. This is slightly different from a survey with isolates collected from Japan, Korea, and China [43], in which although the most prevalent mutations were similar, only *Asp*-516 was found with a higher mutation rate in Beijing than in non-Beijing isolates, different from the study results in the Kyrgyz Republic [9] and Korea [10] that displayed a lower rate in *rpoB* mutation in Beijing vs. in non-Beijing [8,65,66].

However, it might not be comprehensive to use the *rpoB* gene mutation to represent genes with mutation-conferred resistance to RIF to illustrate the drug resistance of the

Beijing strain since some new variants have been found in Beijing strains. According to a previous study, the functional consequence of nonsynonymous *Rv2629* as one of the members of the *dosR* dormancy regulons was found to be upregulated under dormancy conditions in Beijing genotype strains and in a phenotype that might confer a selective advantage under microaerophilic and anaerobic conditions in Beijing strains [70].

The current review demonstrated a prevalence of 100% for the *katG*315 mutation related to INH-resistance, higher than in some previous studies [10,43,71]. The *katG* mutations in Beijing strains of MTB manifested a significantly higher rate than that in non-Beijing isolates (57.88% vs. 28.65%). The rate was higher than that of a study in Southern Xinjiang, China (30.6%; 95% CI, 25.8–35.5%, unclassified by lineages). The prevalence of the *inhA* promoter region mutation in MTB relevant to INH-resistance in Beijing was lower than that of non-Beijing strains in the East/North/Central Europe group, with a significance detected. It might be because the strains of Beijing family strains are not the predominant ones currently [22]. Notably, according to the previous study, some mutations of *inhA* are also found in drug pan-susceptible strains [71]. The drug resistance rate of *oxyR-ahpC* of Beijing strains was lower than that of non-Beijing strains was lower than that of non-Beijing strains (6.26% vs. 7.21%) without significance in MTB relevant to INH-resistance, although both were higher than that in a study of Isakova et al. (1.7%). Similar to the way of the *katG gene* mutation presented as *katG S315T*, almost all the mutations related to *oxyR-ahpC* happened in *the oxyR-ahpC* intergenic region (100%) [9].

This systematic review and Bayesian meta-analysis focused on combining the results of the principal gene mutations of MTB relevant to RIF/INH with the concepts of the classifications of Beijing and non-Beijing strains. It provides a snapshot of the active genes' mutations of the circulating MTB and informs policymakers to make feasible decisions for TB control programs. Furthermore, the pooled data harbors a kind of comprehensive information that the individual study lacks, releasing the clinical practitioners with MTB genetics information for reasonable selections of the anti-TB drugs.

However, entirely relying on genetic methods is not that comprehensive and unreasonable since some potential genes' mutations might have been discovered. Combining considerations based on the merging data of genetics, clinical and epidemiological concerns might be a promising exploration.

### 5. Limitations

There are several limitations in our study. First, due to the local practical conditions, not all the methods used in the included studies followed the World Health Organization criteria, leading to some potential heterogeneities, although corrected technically, which was not as good as it never happening. Second, although the high concern concentrations are focused on the gene mutations of anti-MTB drug resistance, not so many studies are available with the forms meeting the requirements of both the four names of gene mutation related to and the lineages of MTB as well [72], which might lead to some selection biases on the original studies interpretable for the geographical distribution. Third, we included only the major common mutations of MTB genes related to RIF and INH instead of all genes' mutations because of the length limitation of the paper, which might not well interpret the difference in the gene mutations of MTB.

# 6. Conclusions

The mutations in *rpoB* and *katG* genes in Beijing are significantly more common than those in non-Beijing strains of MTB. We do not have sufficient evidence to support that the prevalence of mutations of *inhA* and *oxyR-ahpC* is higher in non-Beijing than in Beijing strains, which provides a reference basis for clinical medication selection.

**Supplementary Materials:** The following supporting information can be downloaded at: https://www.mdpi.com/article/10.3390/genes13101849/s1. File S2 Data of Studies included.

**Author Contributions:** S.G. contributed to the study design, literature review, data extraction and analysis, manuscript draft, and revision. S.L. and V.C. contributed to reach a consensus when there was disagreement on the results of the literature search. All authors have read and agreed to the published version of the manuscript.

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