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Clinical values of metagenomic next-generation sequencing in patients with severe pneumonia: a systematic review and meta-analysis

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Background: Clinical values of metagenomic next-generation sequencing (mNGS) in patients with severe pneumonia remain controversial. Therefore, we conduct this meta-analysis to evaluate the diagnostic performance of mNGS for pathogen detection and its role in the prognosis of severe pneumonia.

Methods: We systematically searched the literature published in PubMed, Embase, Cochrane Library, Web of Science, Clinical Trials.gov, CNKI, Wanfang Data, and CBM from the inception to the 28th September 2022. Relevant trials comparing mNGS with conventional methods applied to patients with severe pneumonia were included. The primary outcomes of this study were the pathogen-positive rate, the 28-day mortality, and the 90-day mortality; secondary outcomes included the duration of mechanical ventilation, the length of hospital stay, and the length of stay in the ICU.

Results: Totally, 24 publications with 3220 patients met the inclusion criteria and were enrolled in this study. Compared with conventional methods (45.78%, 705/1540), mNGS (80.48%, 1233/1532) significantly increased the positive rate of pathogen detection [OR = 6.81, 95% CI (4.59, 10.11, P < 0.001]. The pooled 28-day and 90-day mortality in mNGS group were 15.08% (38/252) and 22.36% (36/161), respectively, which were significantly lower than those in conventional methods group 33.05% (117/354) [OR = 0.35, 95% CI (0.23, 0.55), P < 0.001, $I^2 = 0\%$] and 43.43%(109/251) [OR = 0.34, 95% CI (0.21, 0.54), P < 0.001]. Meanwhile, adjusted treatment based on the results of mNGS shortened the length of hospital stay [MD = -2.76, 95% CI (-3.56, -1.96), P < 0.001] and the length of stay in ICU [MD = -4.11, 95% CI (-5.35, -2.87), P < 0.001].

Conclusion: The pathogen detection positive rate of mNGS was much higher than that of conventional methods. Adjusted treatment based on mNGS results can reduce the 28-day and 90-day mortality of patients with severe pneumonia, and shorten the length of hospital and ICU stay. Therefore, mNGS advised to be applied to severe pneumonia patients as early as possible in addition to conventional methods to improve the prognosis and reduce the length of hospital stay.

KEYWORDS

metagenomic next-generation sequencing, severe pneumonia, diagnosis, prognosis, conventional methods

Introduction

Severe pneumonia is a common critical illness with an increasing morbidity and mortality, patients with this disease often require admission to intensive care unit (ICU) (Kalil et al., 2016; Metlay et al., 2019; Zhou et al., 2021). Despite significant advances in its etiological investigation and antimicrobial therapy, severe pneumonia remains the leading cause of death among infectious diseases worldwide (De Pascale et al., 2011). There is no denying that early, rapid, and accurate pathogenic diagnosis is crucial in guiding promt antibiotic treatment, thus improving prognosis and reducing fatality (Cillóniz et al., 2021).

Traditionally, clinicians select antibiotics empirically and then adjust treatment based on the results of conventional methods. However, microbiological culture-based tests do not meet clinical needs due to their time-consuming nature, low sensitivity, lack of diagnostic tests for rare pathogens and vulnerability to external influences (Torres et al., 2021). Inspiringly, metagenomic nextgeneration sequencing (mNGS), a new pathogen detection technique with high efficiency, broad pathogen spectrum and increased sensitivity, has been widely used in clinic gradually (Gu et al., 2019). mNGS theoretically performs unbiased and detailed high-throughput sequencing of the total DNA or RNA content of almost all known pathogens, including bacteria, fungi, viruses, mycobacterium tuberculosis, parasites and atypical pathogens, and then compares the obtained sequence information with databases (Chiu and Miller, 2019). Several case reports and clinical trials have demonstrated the great value of mNGS in the pathogenic diagnosis of severe and complex cases, including rare pathogens, mixed infections (Wang et al., 2019) and infectious diseases presenting with atypical symptoms, such as Chlamydia psittaci (Li et al., 2021), Chlamydia abortus (Zhu et al., 2022), leptospirosis presenting as severe alveolar hemorrhage (Chen M. et al., 2021) and so on. Several studies (Huang et al., 2020; Chen S. et al., 2022; Wu et al., 2022) have revealed that the sensitivity and specificity of mNGS detection are markedly superior than that of conventional methods.

However, the results of studies on the diagnostic and prognositic values of mNGS on severe pneumonia are controversial (Xie et al., 2019; Zhang et al., 2020; Xi et al., 2022). In addition, to our knowledge, there is no relevant systematic review and meta-analysis to provide a higher level of evidence on the clinical values of mNGS results on severe pneumonia. Therefore, we performed this meta-analysis to evaluate and compare mNGS and conventional methods on the diagnostic performance and prognostic impact of severe pneumonia.

Methods

We strictly followed the standards of the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) (http:// www.prisma-statement.org/) in reporting the findings of this review. The protocol for this study was registered in PROSPERO (CRD42022325564).

Search strategy and data sources

PubMed, Embase, Cochrane Library, Web of Science, Clinical Trials.gov, CNKI, Wanfang DATA and CBM were searched from inception to 28th September 2022. Search terms included the following: ("Next generation sequencing" OR "Metagenomic next generation sequencing" OR "NGS" OR "mNGS") AND ("severe pneumonia" OR "serious pneumonia" OR "severe respiratory infection" OR "severe lung infection" OR "severe communityacquired pneumonia" OR "severe hospital-acquired pneumonia"). Researchers manually scanned the references of all retrieved articles and other relevant publications for additional articles. The exhaustive search strategy is reported in Table S1.

Eligibility criteria and study selection

Two authors (Lv M and Zhu C) screened the relevant literature independently and then checked the title and abstract of each retrieved article to decide which required further assessment. The full text of potentially eligible records was retrieved, reviewed and eligible studies were included. When there were disagreements, Lv M, Zhu C and Du X discussed thoroughly to reach an agreement. Severe pneumonia was defined in patients with either one major criterion or at least three minor criteria of the Infectious Diseases Society of America (IDSA)/ American Thoracic Society (ATS) criteria (Metlay et al., 2019). The primary outcomes were pathogen detection positive rate, 28-day mortality, and 90-day mortality. Secondary outcomes included duration of mechanical ventilation, length of hospital stay, and length of stay in the ICU. The eligibility criteria were as follows: (1) patients with severe pneumonia; (2) participants above 18 years old; (3) reports comparing the pathogen detection positive rate or prognosis outcomes of mNGS with conventional methods. Articles were excluded based on the following criteria: (1) reviews, letters, case report, or case series with < 10 patients; (2) enrolled patients did not suffer from severe pneumonia; (3) not provide data on pathogen-detection positive rate, mortality or prognostic factors; (4) studies not compare positive rate, mortality or prognostic factors between mNGS group and conventional methods group; (5) low quality or can not obtain relevant data.

Qualitative assessment and data extraction

The quality of each study was independently evaluated by two authors (Lv M and Zhu C) using the Newcastle-Ottawa scale. Two authors (Lv M and Zhu C) independently extracted data with a customized data extraction form and assessed the risk of bias. The data extraction form included the following detailed information: (1) references and publication date; (2) type of research; (3) sample types; (4) mean age; (5) gender; (6) the initial value of the diagnostic performance indicators; (7) 28-day mortality; (8) 90-day mortality; (9) duration of mechanical ventilation; (10) length of hospital stay; and (11) length of stay in ICU.

Meta-analysis and statistical methods

Meta-analysis was used to synthesize the outcome measure estimates. Data analyses were performed by Review Manager 5.4. A funnel plot was applied to check for publication bias, and I^2 was applied to estimate the total variation attributed to heterogeneity among studies. Sensitivity analysis was performed by Stata 15. For dichotomous variables, odds ratios (*ORs*) were used for statistical calculations, whereas mean differences (*MDs*) were used for continuous variables. For $I^2 \ge 50\%$, the random effect model of the restricted maximum likelihood probability method is used. Otherwise, the fixed effect model of the reverse variance method is used.

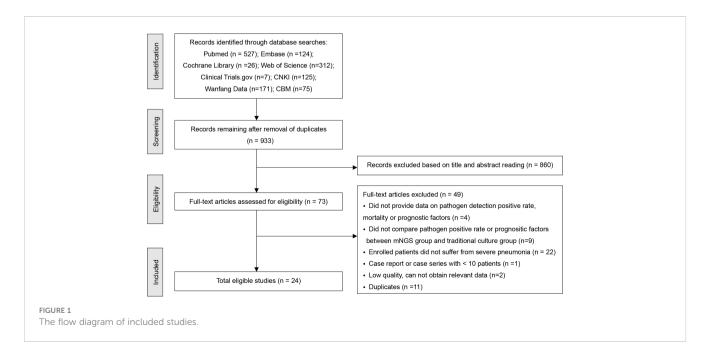
Results

Study selection process and data extraction

The search identified 1367 records, among these 434 were removed as duplicates. After screening titles and abstracts, 860 were found to be ineligible. Overall, 73 full-text articles were assessed for eligibility. Ultimately, 24 studies met the inclusion criteria (Wu, 2019; Xie et al., 2019; Zhuo et al., 2019; Chen, 2020; Fan, 2020; Pan, 2020; Pan and Lv, 2020; Ren et al., 2020; Song et al., 2020; Sun et al., 2020; Wu et al., 2020; Zhang et al., 2020; Chen J. et al., 2021; Huang, 2021; Lu et al., 2021; Ma et al., 2021; Xu, 2021; Zhu and Cao, 2021; Chen Y. et al., 2022; Huang, 2022; Liu et al., 2022; Tan et al., 2022; Zhang et al., 2022; Zhou and Yang, 2022). The detailed literature retrieval and screening process is shown in Figure 1.

Characteristics of the included studies

The characteristics of the eligible studies are presented in Table 1. Totally, 24 studies with 3220 patients were enrolled. Nineteen (Wu, 2019; Xie et al., 2019; Chen, 2020; Fan, 2020; Pan, 2020; Pan and Lv, 2020; Ren et al., 2020; Song et al., 2020; Sun et al., 2020; Zhang et al., 2020; Chen J. et al., 2021; Lu et al., 2021; Ma et al., 2021; Xu, 2021; Chen Y et al., 2022; Huang, 2022;



Liu et al., 2022; Tan et al., 2022; Zhou and Yang, 2022) studies were retrospective and five (Zhuo et al., 2019; Wu et al., 2020; Huang, 2021; Zhu and Cao, 2021; Zhang et al., 2022) were prospective. Eighteen studies included patients with severe pneumonia and six studies included severe pneumonia patients with other complications, such as immunosuppression (Ma et al., 2021), immunodeficiency (Lu et al., 2021), bloodstream infection (Chen J. et al., 2021), acute respiratory distress syndrome (ARDS) (Zhang et al., 2020), after renal transplantation (Zhuo et al., 2019) and autoimmune diseases (Wu, 2019). The sample types collected mainly included blood, sputum, or bronchoalveolar lavage fluid (BALF). Among them, 23 studies (95.8%), 6 studies (25.0%), and 4 studies (16.7%) reported pathogen detection positive rate, 28-day mortality, and 90-day mortality, respectively. Some studies assessed detection positive rate of bacteria, fungi, virus and other pathogens respectively, which are listed in Table S2. Useable data for the duration of mechanical ventilation, length of hospital stay and length of stay in ICU were provided in 5 studies (20.8%), 6 studies (25.0%), and 4 studies (16.7%). The mean study quality score of the studies was 6.38 (SD=0.88) out of 9 on the Newcastle-Ottawa Scale, representing moderate to high methodological quality. A detailed quality assessment is presented in Table S3.

Meta-analysis: mNGS versus conventional methods

Pathogen detection positive rate

A total of 3072 cases in 23 studies were pooled for assessing pathogen detection positive rate. The random-effect model was applied given the heterogeneity test indicated moderate heterogeneity ($I^2 > 50\%$, P < 0.001). We introduced subgroups including BALF and other types of specimens group. The meta-analysis results showed that compared with the conventional methods group (45.78%, 705/1540), mNGS group (80.48%, 1233/1532) significantly increased the pathogen detection positive rate [OR = 6.81, 95% *CI* (4.59, 10.11), P < 0.001 (Figure 2). Sensitivity analysis (Figure 3) was conducted on the included 23 studies, and excluding any one of the studies had no significant impact on the pooled effect value, which confirmed the stability of the final results of this study. The funnel chart showed absence of obvious publication bias (Figure 4).

In addition, twelve studies with a total of 2066 cases assessing detection positive rate of bacteria, fungi, viruses, and other pathogens, respectively. We introduced subgroups including bacteria, fungi, viruses, and other pathogens. In bacteria subgrop, their pooled result indicated that there was no difference in the mNGS group (55.97%, 553/988) and conventional methods group (41.93%, 452/1078) (OR = 1.73, 95% CI (0.73, 4.08), P = 0.21 and $I^2 = 94\%$). The meta-analysis results showed that mNGS (17.75%, 163/973, 22.57%, 223/988 and 23.49%, 128/545) effectively increased the detection positive rate of fungi, virus, and other pathogens compared with the conventional methods group (8.44%, 89/1055) (OR = 2.04, 95% CI (1.13, 3.67), P = 0.02 and $I^2 = 68\%$), (4.64%, 50/1078) (OR = 4.68, 95% CI (1.43, 15.39), P = 0.01 and $I^2 = 82\%$) and (5.79%, 32/553) (OR = 2.59, 95% CI (1.68, 3.98), P=0.001 and $I^2 = 72\%$) (Figure S1). The funnel chart indicated publication bias that are presented in Figure S2.

28-day mortality

A total of six studies investigated the 28-day mortality. 252 cases were included in mNGS group, with a pooled mortality of 15.08% (38/252), significantly lower than that in the conventional methods group (33.05%, 117/354). The synthesis of these results derived from comparison with the conventional methods group indicated that mNGS can significantly promote patient survival, with an OR = 0.35, 95% *CI* (0.23, 0.55), *P* < 0.001, and $I^2 = 0\%$ (Figure 5A). The funnel chart indicated no obvious publication bias (Figure S3).

90-day mortality

Four studies with 161 cases in mNGS group and 251 cases in the conventional methods group reported 90-day mortality. Adjusting treatment based on the pathogen detection results, the pooled 90-day mortality of the mNGS group was significantly lower than that in the conventional methods group (22.36%,36/161 vs 43.43%,109/251) (Figure 5B). There was no obvious publication bias (Figure S4).

Duration of mechanical ventilation

In total, 5 studies including 285 cases in the mNGS group and 296 cases in the conventional methods group investigated the duration of mechanical ventilation. Their pooled result using random model indicated that there was no difference in the mNGS group and the conventional methods group (MD = -1.82, 95% *CI* (-4.39, 0.74), P = 0.16, $I^2 = 57\%$) (Figure 6A). The funnel chart (Figure S5) showed no obvious publication bias.

Length of hospital stay

Six studies including 288 cases in the mNGS group and 296 cases in the conventional methods group reported the length of hospital stay. The mean length of hospital stay of the mNGS group was 16.33, shorter than that of the conventional methods group(19.74). The synthesized results proved that adjusting therapy according to mNGS results could reduce the length of hospital stay when compared with the conventional methods group, MD = -2.76, 95% $CI [-3.56, -1.96], P < 0.00001, I^2 = 47\%$ (Figure 6B). The funnel chart (Figure S6) showed absence of obvious publication bias.

Length of stay in ICU

Four studies including 171 cases in mNGS group and 183 cases in conventional methods group provided data on the length of stay in ICU. The mean length of stay in ICU of mNGS group and conventional methods group were 11.68 and 14.69. Patients who received mNGS detection had shorter length of stay in ICU than those receiving conventional methods detection [MD = -4.11, 95% $CI (-5.35, -2.87), P < 0.001, I^2 = 0\%$] (Figure 6C). The funnel chart (Figure S7) indicated no obvious publication bias.

Antibiotic regiments changes based on NGS results

We carefully examined the studies included. The effects of mNGS results on antibiotic use were noted in four papers. Among them, Zhang et al. (Zhang et al., 2022) reported that early identification of pathogens through mNGS and timely adjustment of treatment regimens significantly reduced the frequency and

TABLE 1 The characteristics of the included preclinical studies.

Author	Year	Type of research	Disease Types	Total, mNGS, control	Sample Types	Detection Positive rate (%)		Mean Age, [mNGS, control (years)]		Proportion of men, [mNGS, control, N(%)]		
				(N)		mNGS	Control	mNGS	Control	mNGS	Control	
Anbing Zhang	2022	Prospective	Severe pneumonia	112,56,56	BALF	92.9	51.8	62.57	60.91	26(46.4)	29(51.8)	
Youlian Chen	2022	Retrospective	Severe pneumonia	116,58,58	BALF	91.4	32.8	62.50	62.50	30(51.7)	30(51.7)	
Xiaolian Zhou	2022	Retrospective	Severe pneumonia	140,70,70	BALF	98.0	69.4	69.20	69.20	51(72.9)	51(72.9)	
Sujun Huang	2022	Retrospective	Severe pneumonia	60,30,30	blood, sputum, or BALF	90.0	56.7	69.72	69.80	18(60.0)	14(46.7)	
Wenwen Tan	2022	Retrospective	Severe pneumonia	320,160,160	BALF	51.9	46.9	75.41	75.41	95(59.4)	95(59.4)	
Hanying Liu	2022	Retrospective	Severe community- acquired pneumonia	346,173,173	sputum or BALF	64.0	28.0	60.00	64.00	126 (72.8)	106(61.3)	
Chunyan Huang	2021	Prospective	Severe pneumonia	60,30,30	blood or BALF	73.3	36.7	56.43	56.83	17(56.7)	16(53.3)	
Fuyao Zhu	2021	Prospective	Severe pneumonia	48,24,24	blood, sputum, or BALF	91.7	66.7	70.75	74.67	13(54.2)	14(58.3)	
Xiaolong Ma	2021	Retrospective	Immunosuppression complicated with severe pneumonia	60,30,30	BALF	83.3	40.0	54.70	55.30	21(70.0)	23(76.7)	
Jiancong Lu	2021	Retrospective	Immunodeficiency with severe pneumonia	152,76,76	BALF	89.5	51.3	38.46	38.46	44(57.9)	44(57.9)	
Jinlian Chen	2021	Retrospective	Severe pneumonia with bloodstream infection	40,20,20	blood or BALF	85.0	50.0	53.05	53.05	12(60.0)	12(60.0)	
Yuhui Xu	2021	Retrospective	Severe pneumonia	220,110,110	blood, sputum, or BALF	74.5	45.5	56.89	56.89	76(69.1)	76(69.1)	
Peng Zhang	2020	Retrospective	ARDS caused by severe pneumonia	95,42,53	blood, sputum, or BALF	91.1	62.2	NR	NR	31(73.8)	38 (71.7)	
Chanyuan Pan	2020	Retrospective	Severe pneumonia	148,115,33	blood, sputum, or BALF	90.4	47.8	61.28	65.64	83(72.2)	26(78.8)	
Di Ren	2020	Retrospective	Severe pulmonary infection	87,43,44	blood or BALF	69.7	36.2	56.05	57.57	29(67.4)	27(61.4)	
C. Song	2020	Retrospective	Severe pneumonia	148,74,74	NR	NR	NR	NR	NR	NR	NR	
Xiaodong Wu	2020	Prospective	Severe community- acquired pneumonia	658,329,329	BALF	90.3	39.5	64.00	64.00	207 (62.9)	207(62.9)	
Ling Chen	2020	Retrospective	Severe pneumonia	56,28,28	BALF	75.6	32.1	54.82	54.82	21(75.0)	21(75.0)	
Xinyuan Fan	2020	Retrospective	Severe pneumonia	40,20,20	BALF	70.0	35.0	58.70	58.70	NR	NR	
Guoxian Sun	2020	Retrospective	Severe pneumonia	16,8,8	BALF	100.0	37.5	62.50	62.50	5(62.5)	5(62.5)	
Chunxi Pan	2020	Retrospective	Severe pneumonia	52,26,26	blood or BALF	92.3	42.3	45.27	45.27	21(80.8)	21(80.8)	

(Continued)

TABLE 1 Continued

Author	Year	Type of research	Disease Types	Total, mNGS, control (N)	Sample Types	Detection Positive rate (%)		Mean Age, [mNGS, control (years)]		Proportion of men, [mNGS, control, N(%)]	
						mNGS	Control	mNGS	Control	mNGS	Control
Huichang Zhuo	2019	Prospective	Severe pneumonia after renal transplantation	38,15,23	BALF	100.0	69.6	40.90	45.30	11(73.3)	18(78.2)
Jing Wu	2019	Retrospective	Autoimmune diseases with severe pneumonia	36,18,18	BALF	83.3	33.3	55.00	55.00	7(41.1)	7(41.1)
Yun Xie	2018	Retrospective	Severe pneumonia	178,48,130	blood, sputum, or BALF	97.9	75.4	NR	NR	NR	NR

BALF, bronchoalveolar lavage fluid; NR, not reported.

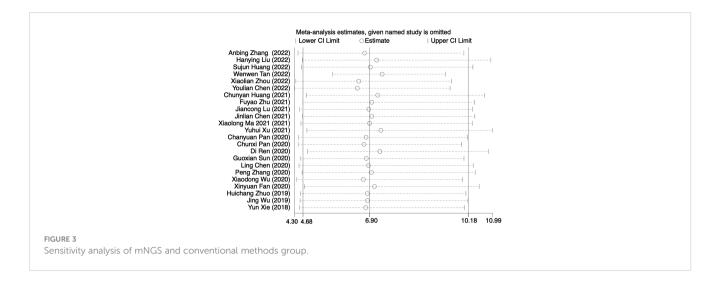
duration of antimicrobial drug adjustment, which provides a new direction for antimicrobial drug management. Xu (Xu, 2021) and Chen (Chen, 2020) reported that adjusting antibiotic regimens based on mNGS detection results improved patient outcomes. Pan (Pan, 2020) reported that adjusted treatment based on mNGS results reduced length of hospital stay and duration of mechanical ventilation, and effectively reduced patient mortality.

Discussion

Severe pneumonia is one of the most critical diseases with leading mortality among infectious diseases in ICU (Nair and Niederman, 2021). In the clinic, conventional methods sometimes fail to find the pathogen in patients with severe pneumonia. A reliable study (Chalmers et al., 2011) shows that the fatality rate is as high as 50% in patients who receive only empiric therapy without reliable evidence of pathogenic microorganisms. Prompt targeted antimicrobial therapy treatment is an effective means to improve the survival rate of patients with severe pneumonia.

The detection of DNA or RNA by mNGS provides rapid, efficient, and accurate access to the entire pathogen genome information within the whole test sample, which has been used widely for pathogen diagnosis in clinic in recent years (Miller et al., 2019). However, to our knowledge, there is no relevant systematic review and meta-analysis to provide more reliable evidence on the clinical values of mNGS on severe pneumonia. This study is performed to evaluate the diagnostic value and prognostic impact

1.1 BALF Anbing Zhang 2022 29 56 52 56 6.6% 0.08 [0.03, 0.26] Werwen Tan 2022 75 160 83 160 6.8% 0.82 [0.53, 1.27] Xiaolian Zhou 2022 44 70 69 70 2.5% 0.02 [0.00, 0.19] Youlian Chen 2022 19 58 53 58 4.8% 0.05 [0.02, 0.13] Jiancong Lu 2021 39 76 68 76 5.5% 0.12 [0.05, 0.29] Xiaolong Ma 2021 12 30 25 30 4.4% 0.13 [0.04, 0.45] Ling Chen 2020 9 28 22 28 4.4% 0.13 [0.04, 0.45] Jing Wu 2019 6 18 15 15 50.38 0.01 [0.06, 0.47] Huichang Zhuo 2019 16 18 15 808 0.38 0.10 [0.06, 0.47] Jing Wu 2019 6 18 15 808 0.38 0.19 [0.12, 0.30]	Study or Subgroup	Events	Total	Events	Total	Weight M-	H, Random, 95% CI	M-H, Random, 95% Cl
$ \begin{array}{c} \text{Were,} mn Tan 2022 & 75 & 160 & 83 & 160 & 6.8\% & 0.82 [0.53, 1.27] \\ \hline \text{Xiaolian Zhou 2022 } 44 & 70 & 69 & 70 & 2.5\% & 0.02 [0.00, 0.13] \\ \hline \text{Jancong Lu 2021 } 39 & 76 & 68 & 76 & 5.5\% & 0.12 [0.05, 0.29] \\ \hline \text{Jancong Lu 2021 } 39 & 76 & 68 & 76 & 5.5\% & 0.12 [0.05, 0.29] \\ \hline \text{Ling Chen 2020 } 3 & 8 & 8 & 8 & 1.3\% & 0.04 [0.00, 0.87] \\ \hline \text{Ling Chen 2020 } 9 & 28 & 22 & 28 & 4.4\% & 0.13 [0.04, 0.45] \\ \hline \text{Ling Chen 2020 } 130 & 329 & 287 & 329 & 7.0\% & 0.10 [0.06, 0.14] \\ \hline \text{Xinoung Mu 2020 } 130 & 329 & 287 & 329 & 7.0\% & 0.10 [0.06, 0.14] \\ \hline \text{Xinoung Fan 2020 } 7 & 20 & 14 & 20 & 4.1\% & 0.23 [0.06, 0.87] \\ \hline \text{Luichang Zhuo 2019 } 6 & 18 & 15 & 1.5\% & 0.07 [0.00, 1.35] \\ \hline \text{Jung Wu 2019 } 6 & 18 & 15 & 1.8\% & 0.07 [0.00, 1.35] \\ \hline \text{Late oversit } 389 & 711 \\ \hline \text{Heterogeneity: Tau" 4 - 1.18; Chi" 2 - 70.27 & df = 11 (P < 0.00001); 1^2 = 84\% \\ \hline \text{Test for overall effect: Z = 5.76 (P < 0.00001)} \\ \hline \text{Lu 2021 } 10 & 20 & 17 & 20 & 3.6\% & 0.19 [0.12, 0.30] \\ \hline \text{Churxy An 2021 } 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \hline \text{Vuhu X 2021 } 50 & 110 & 82 & 110 & 6.5\% & 0.28 [0.16, 0.50] \\ \hline \text{Charyan Pan 2020 } 16 & 33 & 104 & 115 & 5.3\% & 0.07 [0.00, 0.43] \\ \hline \text{Luad 2021 } 11 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \hline \text{Vuhu X 2021 } 50 & 110 & 82 & 110 & 6.5\% & 0.28 [0.16, 0.50] \\ \hline \text{Charyan Pan 2020 } 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ \hline \text{Peng 2Ahug 2020 } 28 & 45 & 41 & 45 & 4.5\% & 0.16 [0.05, 0.53] \\ \hline \text{Lat everts } 316 & 522 \\ \hline \text{Heterogeneity: Tau" = 0.05; \text{Ch}^2 = 79.98, \text{df }= 22 (P < 0.00001); 1^2 = 72\% \\ \hline \text{Total everts } 705 & 1233 \\ \hline \text{Heterogeneity: Tau" = 0.55; \text{Ch}^2 = 79.98, \text{df }= 22 (P < 0.00001); 1^2 = 72\% \\ \hline \text{Total everts } 316 & 522 \\ \hline \text{Heterogeneity: Tau" = 0.55; \text{Ch}^2 = 79.98, \text{df }= 22 (P < 0.00001); 1^2 = 72\% \\ \hline \text{Total everts } 10 & 100 \\ \hline \text{Total everts } 705 & 1233 \\ \hline \text{Heterogeneity: Tau" = 0.55; \text{Ch}^2 = 79.98, \text{df }= 22 (P < 0.00001); 1^2 = 72\% \\ \hline \text{Total everts } 316 & 522 \\ \hline \text{Heterogeneity: Tau" = 0.55; \text{Ch}^$	1.1.1 BALF							
Xiaolian Zhou 2022 44 70 69 70 2.5% 0.02 [0.00, 0.19] Jiancong Lu 2021 39 76 68 76 5.5% 0.12 [0.05, 0.29] Xiaolong Ma 2021 12 30 25 30 4.4% 0.13 [0.04, 0.45] Ling Chen 2020 9 28 22 28 4.4% 0.13 [0.04, 0.43] Xiaolong Wa 2020 30 329 28 22 28 4.4% 0.13 [0.04, 0.43] Michang Zhou 2010 9 28 22 28 4.4% 0.13 [0.04, 0.43] Muchang Zhou 2019 6 18 15 15.3% 0.07 [0.00, 1.34] Huchang Zhou 2019 6 18 15 18 3.4% 0.10 [0.06, 0.24] Total events 389 711 hetrogeneity: Tau ² = 1.18; Chi ² = 70.92; df = 11 (P < 0.00001); l ² = 84% Test for overall effect: 2 = 5.76 (P < 0.00001)	Anbing Zhang 2022	29	56	52	56	4.6%	0.08 [0.03, 0.26]	
Youlian Chen 2022 19 58 53 58 4.8% 0.05 [0.02, 0.13] Jiancong Lu 2021 39 76 68 76 5.5% 0.12 [0.05, 0.29] Xiaolong Ma 2021 12 30 25 30 4.4% 0.13 [0.04, 0.45] Guoxian Sun 2020 3 8 8 1.3% 0.04 [0.00, 0.87] Xiaolong Wu 2020 130 32.9 287 32.9 7.0% 0.10 [0.06, 0.14] Xinvuan Fan 2020 7 20 14 20 4.1% 0.23 [0.06, 0.24] Total events 389 711 Heterogeneity: Tau ² 1.18 Chi 27 30 3.9% 0.12 [0.06, 0.24] Total events 389 711 Heterogeneity: Tau ² 1.16 Chi 27 30 3.9% 0.15 [0.04, 0.59] Churyan Huang 2021 17 30 27 30 3.9% 0.15 [0.0, 0.44] [0.1] Jinitan Chen 2021 10 20 17 20 3.6% 0.18 <td>Wenwen Tan 2022</td> <td>75</td> <td>160</td> <td>83</td> <td>160</td> <td>6.8%</td> <td>0.82 [0.53, 1.27]</td> <td>-+</td>	Wenwen Tan 2022	75	160	83	160	6.8%	0.82 [0.53, 1.27]	-+
Jiancong Li 2021 39 76 68 76 5.5% 0.12 [0.05, 0.29] Xiaolong Ma 2021 12 30 25 30 4.4% 0.13 [0.04, 0.43] Ling Chen 2020 9 28 22 28 4.4% 0.13 [0.04, 0.43] Ling Chen 2020 9 28 22 28 4.4% 0.13 [0.04, 0.43] Xiaodong Wu 2020 13 329 287 329 7.0% 0.10 [0.06, 0.14] Huichang Zhu 2020 17 20 14 20 4.1% 0.23 [0.06, 0.87] Jing Wu 2019 6 18 15 15 15 .15% 0.07 [0.00, 1.35] Jing Wu 2019 6 18 15 18 3.4% 0.10 [0.02, 0.49] Subtotal (95% CI) 876 888 50.3% 0.12 [0.06, 0.24] Heterogeneity: Tau ² 1.18: Ch ² = 7.09 2.7 df = 11 (P < 0.00001); l ² = 84% Test for overall effect: Z = 5.76 (P < 0.00001) Ling Charyuan Pan 2020 16 23 10 27 30 3.9% 0.15 [0.04, 0.59] Jing Wu 2021 10 20 17 20 3.6% 0.19 [0.02, 0.49] Jing U2021 10 20 17 20 3.6% 0.18 [0.04, 0.80] Yuhui Xu 2021 10 20 17 20 3.6% 0.18 [0.04, 0.80] Yuhui Xu 2021 10 22 43 10 6.5% 0.28 [0.16, 0.60] Charyuan Pan 2020 16 33 104 115 5.3% 0.10 [0.04, 0.25] Charyuan Pan 2020 16 43 28 43 5.4% 0.32 [0.13, 0.13] Ji Ren 2020 16 43 28 43 5.4% 0.19 [0.10, 0.31] Ji Ren 2020 16 43 22 4 26 3.3% 0.06 [0.01, 0.31] Subtotal (95% CI) 1540 1522 100.0% 0.15 [0.10, 0.22] Heterogeneity: Tau ² = 0.05; Ch ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 (P < 0.00001); l ² = 72%	Xiaolian Zhou 2022	44	70	69	70	2.5%	0.02 [0.00, 0.19]	←
$\begin{aligned} \begin{array}{c} \hline Xiaolong Ma 2021 & 12 & 30 & 25 & 30 & 4.4\% & 0.13 [0.04, 0.45] \\ \hline Guoxian Sun 2020 & 3 & 8 & 8 & 1.3\% & 0.04 [0.00, 0.87] \\ \hline Ling Chen 2020 & 9 & 28 & 22 & 28 & 4.4\% & 0.13 [0.04, 0.45] \\ \hline Xiaodong Wu 2020 & 130 & 329 & 287 & 329 & 7.0\% & 0.10 [0.06, 0.14] \\ \hline Xinyaua Fan 2020 & 7 & 20 & 14 & 20 & 4.1\% & 0.23 [0.06, 0.87] \\ \hline Huichang Zhuo 2019 & 16 & 23 & 15 & 15 & 5.5\% & 0.07 [0.00, 0.135] \\ \hline Heterogeneity: Tau2 = 1.18; Ch2 = 70.92, df = 11 (P < 0.00001); l2 = 84% \\ \hline Total events & 389 & 711 \\ \hline Heterogeneity: Tau2 = 1.18; Ch2 = 70.92, df = 11 (P < 0.00001); l2 = 84% \\ \hline Total events & 389 & 711 \\ \hline Layouta 10202 & 17 & 30 & 27 & 30 & 3.9\% & 0.15 [0.04, 0.59] \\ \hline Churyan Huang 2021 & 11 & 30 & 20 & 30 & 4.9\% & 0.29 [0.10, 0.84] \\ \hline Tyua 2D21 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \hline Yuhu X 2021 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \hline Churyan Pan 2020 & 16 & 24 & 22 & 42 & 3.2\% & 0.18 [0.04, 0.80] \\ \hline Tuhe 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ \hline Jinitan Chen 2021 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \hline Vuhu X 2021 & 50 & 110 & 652 & 0.28 (0.16, 0.50] \\ \hline Charayuan Pan 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ \hline Di Ren 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ \hline Peng Zhang 2020 & 28 & 45 & 41 & 45 & 4.5\% & 0.16 [0.05, 0.51] \\ \hline Total events & 316 & 522 \\ \hline Heterogeneity: Tau2 = 0.05; Ch2 = 79.98, df = 22 (P < 0.00001); l2 = 72\% \\ \hline Total events & 316 & 522 \\ \hline Heterogeneity: Tau2 = 0.55; Ch2 = 79.98, df = 22 (P < 0.00001); l2 = 72\% \\ \hline Total events & 705 \\ \hline Total events & 705 \\ \hline Total events & 705 \\ \hline Total events & 9.51 (P < 0.00001) \\ \hline Total events & 9.51 (P < 0.00001) \\ \hline \hline Total events & 705 \\ \hline Heterogeneity: Tau2 = 0.55; Ch2 = 79.98, df = 22 (P < 0.00001); l2 = 72\% \\ \hline Heterogeneity: Tau2 = 0.55; Ch2 = 79.98, df = 22 (P < 0.00001); l2 = 72\% \\ \hline Total events & 705 \\ \hline Total events & 705 \\ \hline Total event & 9.51 (P < 0.00001) \\ \hline \hline Total event & 9.51 (P < 0.00001) $	Youlian Chen 2022	19	58	53	58	4.8%	0.05 [0.02, 0.13]	
Guoxian Sun 2020 3 8 8 8 1.3% $0.04 [0.00, 0.87]$ Ling Chen 2020 9 28 22 28 4.4% 0.13 [0.04, 0.03] Xiaodong Wu 2020 130 329 287 329 7.0% 0.10 [0.06, 0.14] Xinyuan Fan 2020 7 20 14 20 4.1% 0.23 [0.06, 0.67] Huichang Zhou 2019 6 18 15 15 15 5.07 (0.00, 1.37] Jing Wu 2019 6 18 15 18 3.4% 0.10 [0.02, 0.49] Subtotal (95% CI) 876 868 50.3% 0.12 [0.06, 0.24] 100 Heterogeneity: Tu² = 1.18: Ch² = 70.92, df = 11 (P < 0.00001); l² = 84%	Jiancong Lu 2021	39	76	68	76	5.5%	0.12 [0.05, 0.29]	
Ling Chen 2020 9 2 28 22 28 4.4% 0.13 [0.04_{0} 0.43] Xiaodong Wu 2020 130 329 287 329 7.0% 0.10 [0.06_{0} 0.43] Xinyuan Fan 2020 7 20 14 20 4.1% 0.23 [0.06_{0} 0.14] Huichang Zhuo 2019 16 23 15 15 1.5% 0.07 [0.00_{1} 1.35] Subtocal (95% CI) 876 888 50.3% 0.12 [0.06_{0} 0.24] Subtocal (95% CI) 876 888 50.3% 0.12 [0.06_{0} 0.24] Heterogeneity: Tau ² = 1.18; Chi ² = 70.92, df = 11 ($P < 0.00001$); $I2 = 84%$ Test for overall effect: Z = 5.76 ($P < 0.00001$) L1.2 Other types of specimens Hanying Liu 2022 43 173 110 173 6.8% 0.19 [$0.12, 0.30$] Subtocal (95% CI) 877 100 27 30 3.9% 0.15 [$0.04, 0.59$] Churyan Huang 2021 11 30 20 30 4.9% 0.29 [$0.10, 0.84$] Churyan Huang 2021 16 24 22 24 3.2% 0.18 [$0.03, 0.97$] Jinilan Chen 2021 10 20 17 20 3.6% 0.18 [$0.04, 0.80$] Yuhui Xu 2021 50 110 62 110 6.5% 0.28 ($0.16, 0.50$] Chanyuan Pan 2020 16 33 104 115 5.3% 0.019 [$0.12, 0.30$] Churyan Pan 2020 16 43 28 43 5.4% 0.32 [$0.13, 0.77$] Peng Zhang 2020 28 45 41 45 4.5% 0.16 [$0.05, 0.53$] Di Ren 2020 16 43 28 43 5.4% 0.32 [$0.13, 0.77$] Peng Zhang 2020 28 45 41 45 4.5% 0.16 [$0.05, 0.53$] Di Ren 2020 16 522 Heterogeneity: Tau ² = 0.05; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Total events 316 522 Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Total events 0.55 ; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Total events 0.55 ; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Total events 0.55 ; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Total events 0.55 ; Chi ² = 79.98, df = 22 ($P < 0.00001$); $I2 = 72\%$ Heterogeneity: Tau ² = 0.55; Chi ² = 0.95; Rf = 2.00001)	Xiaolong Ma 2021	12	30	25	30	4.4%	0.13 [0.04, 0.45]	
$\begin{aligned} \begin{array}{c} \text{Xia} \text{dong } \text{Wu} 2020 & 130 & 329 & 287 & 329 & 7.0\% & 0.10 [0.06, 0.14] \\ \text{Xinyuan Fan } 2020 & 7 & 20 & 14 & 20 & 4.1\% & 0.23 [0.06, 0.08] \\ \text{Yinyuan Fan } 2020 & 7 & 20 & 14 & 20 & 4.1\% & 0.23 [0.06, 0.08] \\ \text{Jing Wu} 2019 & 6 & 18 & 15 & 1.5\% & 0.07 [0.00, 1.35] \\ \text{Jing Wu} 2019 & 6 & 18 & 15 & 1.8 & 3.4\% & 0.10 [0.02, 0.49] \\ \text{Subtotal } (95\% \text{C}) & 876 & 668 & 50.3\% & 0.12 [0.06, 0.24] \\ \text{Test for overall effect: } Z = 5.76 (P < 0.00001); I^2 = 84\% \\ \text{Test for overall effect: } Z = 5.76 (P < 0.00001) \\ \text{L1.2 Other types of specimens} \\ \text{Hanying Liu } 2022 & 43 & 173 & 110 & 173 & 6.8\% & 0.19 [0.12, 0.30] \\ \text{Chunyan Huang } 2021 & 11 & 30 & 20 & 30 & 4.9\% & 0.29 [0.10, 0.84] \\ \text{Fuyao Zhu } 2021 & 16 & 24 & 22 & 43.2\% & 0.18 [0.04, 0.59] \\ \text{Jinilan Chen } 2021 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \text{Yuhui Xu } 2021 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \text{Yuhui Xu } 2021 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \text{Yuhui Xu } 2021 & 10 & 24 & 12 & 3.3\% & 0.10 [0.04, 0.25] \\ \text{Chanyuan Pan } 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ \text{Peng Zhang 2020 & 28 & 45 & 41 & 45 & 4.5\% & 0.16 [0.05, 0.53] \\ \text{Subtotal } (95\% \text{C}) & 1504 & 1532 & 100.0\% & 0.15 [0.10, 0.22] \\ \text{Total events } & 316 & 522 \\ \text{Heterogeneity: Tau^2 = 0.00; Chi^2 = 8.84, df = 10 (P = 0.55); I^2 = 0\% \\ \text{Test for overall effect: Z = 11.78 (P < 0.00001) \\ \text{Total } (95\% \text{C}) & 1504 & 1532 & 100.0\% & 0.15 [0.10, 0.22] \\ \text{Heterogeneity: Tau^2 = 0.55; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ \text{Heterogeneity: Tau^2 = 0.55; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ \text{Heterogeneity: Tau^2 = 0.55; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ \text{Heterogeneity: Tau^2 = 0.55; Chi^2 = 0.95; R^2 < 0.00001) \\ \text{Total events } 10 & 100 \\ \text{Total events } 10 & 100 \\ \text{Total events } 10 & 5.51 (P < 0.00001) \\ \text{Total events } 10 & 100 \\ \text{Total events } 10 & 5.51 (P < 0.00001) \\ \text{Total events } 10 & 5.51 (P < 0.00001) \\ \text{Total events } 10 & 5.51 (P < 0.00001) \\ Tot$	Guoxian Sun 2020	3	8	8	8	1.3%	0.04 [0.00, 0.87]	·
$\begin{aligned} & Xinyuan Tan 2020 & 7 & 20 & 14 & 20 & 4.1\% & 0.23 [0.06, 0.87] \\ & Huichang Zhou 2019 & 16 & 23 & 15 & 15 & 1.5\% & 0.07 [0.00, 1.36] \\ & Jing Wu 2019 & 16 & 23 & 15 & 15 & 1.5\% & 0.07 [0.00, 1.36] \\ & Subtcal (95\% CI) & 876 & 868 & 50.3\% & 0.12 [0.06, 0.24] \\ & Total events & 389 & 711 \\ & Heterogeneity: Tau^2 = 1.18; Chi^2 = 70.92, df = 11 (P < 0.00001); I^2 = 84\% \\ & Test for overall effect: 2 = 5.5 (P < 0.00001) \\ & \mathsf{Tup an Lang 2021 & 11 & 30 & 20 & 30 & 4.9\% & 0.29 [0.10, 0.84] \\ & Fuyao Zhu 2021 & 16 & 24 & 22 & 43 & 2.2\% & 0.18 [0.04, 0.59] \\ & Tup an Lang 2021 & 16 & 24 & 22 & 43 & 2.2\% & 0.18 [0.04, 0.59] \\ & Tup an Lang 2021 & 16 & 24 & 22 & 43 & 2.2\% & 0.18 [0.04, 0.69] \\ & Tup an Lang 2021 & 16 & 24 & 22 & 43 & 2.2\% & 0.18 [0.04, 0.80] \\ & Tup an Lang 2021 & 16 & 24 & 22 & 43 & 3.2\% & 0.18 [0.04, 0.80] \\ & Tup an Lang 2021 & 16 & 24 & 22 & 43 & 3.2\% & 0.18 [0.04, 0.80] \\ & Tup an Lang 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.28 [0.16, 0.50] \\ & Chanyuan Pa 2020 & 11 & 26 & 24 & 26 & 3.3\% & 0.06 [0.01, 0.31] \\ & Di Ren 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ & Peng Zhang 2020 & 28 & 45 & 41 & 45 & 4.5\% & 0.16 [0.05, 0.53] \\ & Subtcal (95\% CI) & 1540 & 1532 100.0\% & \mathsf{0.15 [0.10, 0.22] \\ & Total events & 316 & 522 \\ & \mathsf{Heterogeneity: Tau^2 = 0.05; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 0.95, Gf = 2.20 (P < 0.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 0.95, Gf = 2.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 79.98, df = 22 (P < 0.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 0.95, Gf = 2.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 0.95, Gf = 2.20 (P < 0.00001); I^2 = 72\% \\ & \mathsf{Heterogeneity: Tau^2 = 0.55; Chi^2 = 0.95, Gf = 2.00$	Ling Chen 2020	9	28	22	28	4.4%	0.13 [0.04, 0.43]	
Huichang Zhuo 2019 16 23 15 15 1.5% $0.07 [0.00, 1.35]$ Jing Wu 2019 6 18 15 18 3.4% $0.10 [0.02, 0.49]$ Subtotal (95% CI) 876 868 50.3% $0.12 [0.06, 0.24]$ Total events 389 711 Heterogeneity: Tau ² = 1.18; Ch ² = 70.92, Gf = 11 ($P < 0.00001$); l ² = 84% Test for overall effect: $Z = 5.76 (P < 0.00001$) 1.1.2 Other types of specimens Hanying Liu 2022 43 173 110 173 6.8% 0.19 [0.12, 0.30] Churyan Huang 2021 11 30 20 30 4.9% 0.29 [0.10, 0.84] Fuyao Zhu 2021 16 24 22 24 3.2% 0.18 [0.04, 0.80] Yuhui X 2021 50 110 82 110 6.5% 0.28 [0.16, 0.50] Churyan Pha 2020 16 33 104 115 5.3% 0.10 [0.04, 0.80] Yuhui X 2021 50 110 82 110 6.5% 0.32 [0.13, 0.77] Jinitan Chen 2021 12 62 44 26 3.3% 0.06 [0.01, 0.31] Di Ren 2020 11 26 24 26 3.3% 0.05 [0.01, 0.32] Di Ren 2020 16 43 28 43 5.4% 0.32 [0.13, 0.77] Peng Zhang 2020 28 45 41 45 4.5% 0.16 [0.05, 0.53] Urun Xie 2018 98 130 47 48 2.5% 0.07 [0.01, 0.48] Subtotal (95% CI) 1540 1532 100.0% 0.15 [0.10, 0.22] Heterogeneity: Tau ² = 0.05; Ch ² = 79.98, df = 22 ($P < 0.00001$); l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Ch ² = 79.98, df = 22 ($P < 0.00001$; l ² = 72%	Xiaodong Wu 2020	130	329	287	329	7.0%	0.10 [0.06, 0.14]	
Jing Wu 2019 6 18 15 18 3.4% 0.10 [0.02, 0.49] Subtotal (95% CI) 876 868 50.3% 0.12 [0.06, 0.24] Total events 389 711 Heterogeneity: Tau ² = 1.18; Chi ² = 70.92, df = 11 ($p < 0.00001$); l ² = 84% Test for overall effect: 2 = 5.5; Ch ² = 70.92, df = 11 ($p < 0.00001$); l ² = 84% Tust for overall effect: 2 = 5.5; Chi ² = 70.92, df = 11 ($p < 0.00001$); l ² = 84% Tust for overall effect: 2 = 5.5; Chi ² = 70.92, df = 11 ($p < 0.00001$); l ² = 84% Tust for overall effect: 2 = 0.55; Chi ² = 70.92, df = 11 ($p < 0.0001$); l ² = 84% Tust for overall effect: 2 = 0.55; Chi ² = 70.92, df = 11 ($p < 0.0001$); l ² = 84% Tust for overall effect: 2 = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($p < 0.00001$; l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95; R ⁴ = 0.00001	Xinyuan Fan 2020	7	20	14	20	4.1%	0.23 [0.06, 0.87]	
Subtocle (95% CI) 876 868 50.3% 0.12 [0.06, 0.24] Total events 389 711 Heterogeneity: Tau ² = 1.18; Chi ² = 70.92; df = 11 ($P < 0.00001$); l ² = 84% Test for overall effect: Z = 5.76 ($P < 0.00001$) Hanying Liu 2022 43 173 110 173 6.8% 0.19 [0.12, 0.30] Sujun Huang 2021 11 30 20 30 4.9% 0.29 [0.10, 0.84] Chunyan Huang 2021 10 20 17 20 3.6% 0.18 [0.04, 0.80] Yuhui X 2021 10 20 17 20 3.6% 0.18 [0.04, 0.80] Yuhui X 2021 10 20 17 20 3.6% 0.18 [0.04, 0.80] Yuhui X 2021 10 82 110 6.5% 0.28 [0.16, 0.50]	Huichang Zhuo 2019	16	23	15	15	1.5%	0.07 [0.00, 1.35]	· · · · · · · · · · · · · · · · · · ·
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Hanying Liu 2022 43 173 110 173 6.8% 0.19 [0.12, 0.30] Sujun Huang 2021 17 30 27 30 3.9% 0.15 [0.04, 0.59] Chunyan Huang 2021 16 24 22 24 3.2% 0.18 [0.03, 0.97] Jinlian Chen 2021 16 24 22 24 3.2% 0.18 [0.03, 0.97] Jinlian Chen 2021 10 20 17 20 3.6% 0.18 [0.04, 0.86] Vuhui Xu 2021 50 110 82 110 6.5% 0.28 [0.16, 0.50] Chanyuan Pan 2020 16 33 104 115 5.3% 0.10 [0.04, 0.25] Chunxi Pan 2020 16 43 28 43 5.4% 0.32 [0.13, 0.77] Peng Zhang 2020 28 45 41 45 4.5% 0.16 [0.05, 0.31] Di Ren 2020 16 432 8 43 5.4% 0.32 [0.13, 0.77] Peng Zhang 2020 28 45 41 45 4.5% 0.07 [0.01, 0.49] Subtotal (95% CI) 1540 1532 100.0% 0.15 [0.10, 0.22] Total events 705 1233 Heterogeneity: Tau ² = 0.05; Chi ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.99.8, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.99.8, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001); l ² = 72% Heterogeneity: Tau ² = 0.55; Chi ² = 0.95, df = 22 (P < 0.00001);	Test for overall effect: Z	s = 5.76 (P < 0.0	0001)				
	1.1.2 Other types of sp	pecimens						
$\begin{array}{c cl} Churyna Huang 2021 & 11 & 30 & 20 & 30 & 4.9\% & 0.29 [0.10, 0.84] \\ \hline Fuyao Zhu 2021 & 16 & 24 & 22 & 24 & 3.2\% & 0.18 [0.03, 0.97] \\ \hline Fuyao Zhu 2021 & 10 & 20 & 17 & 20 & 3.6\% & 0.18 [0.04, 0.80] \\ \hline Yuhui Xu 2021 & 50 & 110 & 82 & 110 & 6.5\% & 0.28 [0.16, 0.50] \\ \hline Charyua Pa 2020 & 16 & 33 & 104 & 115 & 5.3\% & 0.10 [0.04, 0.25] \\ \hline Charyua Pa 2020 & 11 & 26 & 24 & 26 & 3.3\% & 0.06 [0.01, 0.31] \\ \hline Di Ren 2020 & 16 & 43 & 28 & 43 & 5.4\% & 0.32 [0.13, 0.77] \\ \hline Peng Zhang 2020 & 28 & 45 & 41 & 45 & 4.5\% & 0.16 [0.05, 0.53] \\ \hline Subtocal (95\% CI) & 664 & 664 & 49.7\% & 0.20 [0.15, 0.26] \\ \hline Total events & 316 & 522 \\ Heterogeneity: Tau2 = 0.05; Chi2 = 79.98, df = 22 (p < 0.00001); l2 = 72\% \\ \hline Total events & 705 & 1233 \\ Heterogeneity: Tau2 = 0.55; Chi2 = 79.98, df = 22 (p < 0.00001); l2 = 72\% \\ \hline Dotal events & 9.51 (p < 0.00001) \\ \hline \end{array}$	Hanying Liu 2022	43	173	110	173	6.8%	0.19 [0.12, 0.30]	
Fuyao Zhu 2021 16 24 22 24 3.2% $0.18 [0.03, 0.97]$ Jinlian Chen 2021 10 20 17 20 3.6% $0.18 [0.04, 0.80]$ Yuhui Xu 2021 10 20 17 20 3.6% $0.18 [0.04, 0.80]$ Chanyuan Pan 2020 16 33 104 115 5.3% $0.06 [0.01, 0.31]$ Di Ren 2020 16 43 28 43 5.4% $0.32 [0.13, 0.77]$ Peng Zhang 202 28 45 44 5.5% $0.07 [0.01, 0.49]$ Peng Zhang 202 28 45 44 2.5% $0.07 [0.01, 0.49]$ Subtotal (95% CI) 664 664 49.7% 0.20 [0.15, 0.26] Total events 316 522 Heterogeneity: Tau ² = 0.05; Chi ² = 79.98, df = 22 (P < 0.00001);		17	30				0.15 [0.04, 0.59]	
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $		11	30	20			0.29 [0.10, 0.84]	
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$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Jinlian Chen 2021							
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	Yuhui Xu 2021	50	110	82	110		0.28 [0.16, 0.50]	
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Peng Zhang 2020 28 45 41 45 4.5% 0.16 $[0.05, 0.53]$ Yun Xie 2018 98 130 47 48 2.5% 0.07 $[0.01, 0.49]$ Subtotal (95% CI) 664 664 49.7% 0.20 $[0.15, 0.26]$ Total events 316 522 Heterogeneity: Tau ² = 0.00; Chi ² = 8.84, df = 10 (P = 0.55); l ² = 0% Test for overall effect: 2 = 11.78 (P < 0.00001) Total (95% CI) 1540 1532 100.0% 0.15 $[0.10, 0.22]$ Total events 705 1233 Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 (P < 0.00001); l ² = 72% Test for overall effect: 2 = 0.51 (P < 0.00001)	Chunxi Pan 2020	11	26	24				
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Heterogeneity: Tau ² = 0.00; Chi ² = 8.84, df = 10 ($P = 0.55$); l ² = 0% Test for overall effect: Z = 11.78 ($P < 0.00001$) Total (95% CI) 1540 1532 100.0% 0.15 [0.10, 0.22] Heterogeneity: Tau ² = 0.55; Chi ² = 79.98, df = 22 ($P < 0.00001$); l ² = 72% Test for overall effect: Z = 9.51 ($P < 0.00001$)			664		664	49.7%	0.20 [0.15, 0.26]	•
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Heterogeneity: $Tau^2 = 0.55$; $Chi^2 = 79.98$, $df = 22$ (P < 0.00001); $l^2 = 72\%$ Test for overall effect: Z = 9.51 (P < 0.00001) Eavours [experimental] Eavours [experimental]			1540		1532	100.0%	0.15 [0.10, 0.22]	◆
Test for overall effect: Z = 9.51 (P < 0.00001) 0.1 1 10 100 Favours (control)								
Test for overall effect: Z = 9.51 (P < 0.00001) Eavours [experimental] Eavours [control]					2 (P <	0.00001); I ²	= 72%	0.01 0.1 1 10 10
Test for subgroup differences: Chi ^{ϵ} = 1.78, dt = 1 (P = 0.18), l ^{ϵ} = 43.9%								
	Test for subgroup differ	rences: Ch	$ni^2 = 1$.78, df =	1 (P =	$0.18), I^2 = 43$	3.9%	

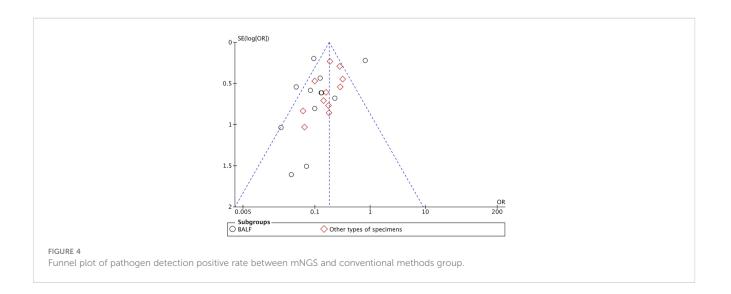


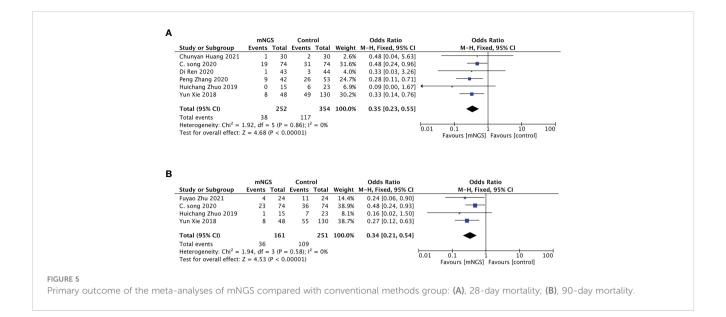
of mNGS compared with conventional methods in patients with severe pneumonia.

In order to accurately assess the pathogenic diagnostic value of mNGS in severe pneumonia and its impact on prognosis, we comprehensively collected the studies published on the application of mNGS to the etiological diagnostic of severe pneumonia compared with conventional methods, and finally included 23 studies for meta-analysis. The pathogen detection positive rate of the mNGS group was 80.48% (1233/1532), much higher than that of the conventional methods group (45.78%, 705/ 1540), which indicated that mNGS has a higher diagnostic value than conventional methods and can be used as an effective method for rapid etiological diagnosis of severe pneumonia. Bacteria can be detected by both mNGS and traditional detection methods. mNGS is superior for detecting fungi, viruses, and rare pathogens, which can be used as an adjunct or complementary tool to conventional microbial testing methods to provide an etiologic basis for the accurate anti-infection treatment of severe pneumonia. It is well known that the traditional microbial detection method is susceptible to the influence of normal bacteria in the body,

especially after using antibiotics. Patients with severe pneumonia are often initially treated with an anti-microbial therapy, which reduces the positive rate of detection by conventional methods. mNGS can improve the pathogen detection positive rate and is less affected by external influences (Gu et al., 2019). Moreover, some reports have confirmed that mNGS achieve a faster diagnosis of pathogens and can detect unknown pathogens and even drugresistance genes (Crofts et al., 2017). Another clinical application of mNGS is to identify microbial colonization or infection by monitoring the immune response of patients, ultimately achieving rational application of antibiotics, suppressing bacterial resistance, and reducing the economic and social burden of infectious diseases (Besser et al., 2018).

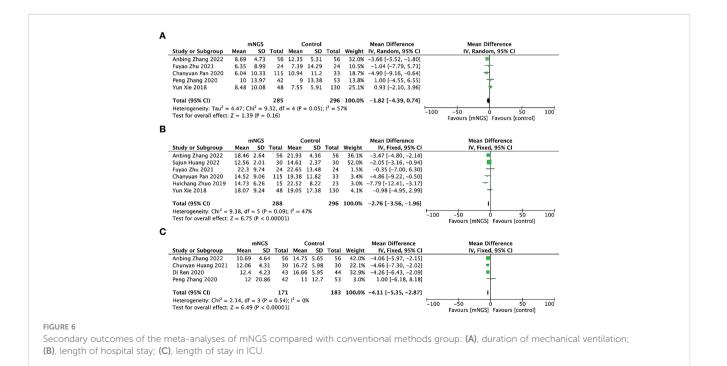
Meanwhile, compared with the conventional methods group, adjusting the treatment based on the mNGS results significantly decreased 28-day and 90-day mortality and shortened the length of hospital and ICU stay of patients with severe pneumonia. Patients with serious conditions have a short window of time for clinicians to save their lives and precise treatment is crucial to their prognosis. mNGS was faster, taking an average of two days, whereas





conventional methods required at least three to five days (Xie et al., 2019). In addition, the higher pathogen-positive rate of mNGS detection gives clinicians the opportunity to select accurate antimicrobial drugs as early as possible, greatly increasing the proportion of target treatments. mNGS lead to more rapid required, accurate diagnosis than conventional methods in severe pneumonia. As a result, mNGS was associated with a better clinical prognosis of severe pneumonia.

mNGS facilitates accurate diagnosis and treatment. Still, there are many challenges, such as the lack of common standards for outcome analysis and common guidelines for report interpretation. The quality and stability of mNGS analysis results may not be stable to some extent, which is closely related to the operator's skill and the detection efficiency of different laboratories may not be compared. A key challenge inherent to mNGS is that microbial nucleic acids from most patients' samples are dominated by human host backgrounds (Hasan et al., 2016). Alternative methods exist for the depletion of human background DNA during the preanalytical phase, one approach is to selectively lyse human white blood cells using saponin or other chemical reagents (Marotz et al., 2018), and a different approach is to target low-molecular-weight cell-free DNA or RNA and remove high-molecular-weight genetic content that is often associated with human genomic material (Bukowska-Ośko et al., 2016). Another potential disadvantage of mNGS is the contamination of the sample. We need strictly adhere to quality control procedures for reagents and workflows to maintain as sterile



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and nucleic acid-free a testing environment as possible. The use of negative controls, reagent evaluation, and regular brushing is needed to ensure that laboratory and sample cross-contamination does not produce false positive results (Schlaberg et al., 2017). In addition, mNGS detection is highly sensitive and requires clinicians with rich experience to comprehensively consider the patient's condition to make a judgment.

The included studies have limitations, such as small sample sizes, single-centered and retrospective analysis (Pan, 2020; Sun et al., 2020; Zhang et al., 2022), and confounding bias that could not be completely excluded. Besides, all 24 studies included were from China, making the inclusion of a homogeneous population and thus limiting the extensibility of the results. This limits the external generalizability of the results to some degree. Finally, we could not conduct a SORC curve analysis to assess the diagnostic efficacy of mNGS and conventional methods for severe pneumonia due to the scarcity of studies reporting sensitivity and specificity.

Conclusion

The pathogen detection positive rate of mNGS was much higher than that of conventional methods, thereby indicating that mNGS has an extremely good diagnostic performance for severe pneumonia. Besides, adjusting treatment based on mNGS results can reduce the 28-day and 90-day mortality of patients with severe pneumonia, and shorten the length of hospital and ICU stay. Therefore, we suggest that patients with severe pneumonia should be tested for mNGS in addition to traditional culture as early as possible to improve the prognosis and reduce the length of hospital stay.

Author contributions

We describe contributions to the paper as follows: conceptualization - ML, XD and GF. methodology - ML and XD. validation - ML and JY. formal analysis - ML and XD. data curation

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Conflict of interest

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Supplementary material

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