

REVIEW

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Prevalence of antimalaria drug resistance-conferring mutations associated with sulphadoxine-pyrimethamine-resistant *Plasmodium falciparum* in East Africa: a systematic review and meta-analysis

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Abstract

Background The emergence and spread of drug resistance to antimalarial drugs pose a severe threat to effective malaria control and treatment. Although sulfadoxine-pyrimethamine resistance is well-documented, it is still the drug of choice for treating intermittent resistance. Molecular markers play a crucial role in tracking and understanding the prevalence of antimalarial drug resistance. Currently, there is insufficient information on the prevalence of molecular markers associated with sulfadoxine-pyrimethamine resistance in *P. falciparum*.

Objective This systematic review and meta-analysis aimed to determine the pooled prevalence of antimalaria drug resistance-conferring markers associated with sulphadoxine-pyrimethamine in *Plasmodium falciparum* in East Africa.

Methods Systematic search was performed to retrieve articles from PubMed, Scopus, Science Direct databases, and Google Scholar search engine. Sixteen potential studies that provided important data on markers for sulphadoxine-pyrimethamine resistance in *Plasmodium falciparum* were systematically reviewed and analyzed. Nine antimalarial drug resistance markers responsible for sulphadoxine-pyrimethamine resistance in *Plasmodium falciparum* were extracted separately into Microsoft Excel and analyzed using STATA 17.0. The inverse of variance was done to evaluate heterogeneity across studies. A funnel plot was used to determine the presence of publication bias. A trim-and-fill-meta-analysis was carried out to generate a bias-adjusted effect estimate. A random effect model was used to determine the pooled prevalence of markers responsible for sulphadoxine-pyrimethamine resistance. Subgroup analysis was performed based on country and year of publication.

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Results A total of 16 studies were included for this systematic review and meta-analysis. The molecular markers like dhfr (N51I, C59R, S108N, 108N, 59R, and I164L), and dhps (A437G, K540E, & 540E) were selected for meta-analysis. From this meta-analysis, the pooled prevalence of dhfr N51I, dhfr C59R, dhfr S108N, dhfr 108N, dhfr 59R, and dhfr I164L was 88.6%, 85.3%, 89.6%, 92.2%, 71.5%, and 3.9%, respectively. Likewise, the aggregated prevalence of dhps A437G, dhps K540E, and dhps 540E was 90.2%, 80.9%, and 91.5%, respectively. The subgroup analysis based on year of publication showed that the pooled prevalence of dhfr N51I, dhfr C59R, dhfr S108N, dhps A437G, and dhps K540E, in studies conducted 2014–2018 was 97.11%, 90.57%, 96.45%, 90.89%, and 89.45%, respectively, while it was 82.03%, 81.78%, 85.12%, 89.24%, and 73.98%, respectively, in studies conducted 2019–2023. On the other hand, country-based analysis showed that the pooled prevalence of dhfr N51I, dhfr C59R, dhfr S108N, dhps A437G, and dhps K540E, in Kenya was 85.88%, 84.02%, 86.56%, 90.7%, and 77.55%, respectively.

Conclusions This systematic review and meta-analysis reveal a high prevalence of drug resistance markers associated with sulphadoxine-pyrimethamine resistance in *Plasmodium falciparum* across the East African region. This underscores the significant challenges in managing malaria infections caused by *Plasmodium falciparum* in the region. Therefore, regular monitoring, identification, and limiting of drug-resistance markers and drug-resistant *P. falciparum* strains must be sustained to ensure the effectiveness of malaria treatment.

Keywords Prevalence, *Plasmodium falciparum*, Drug resistance, Marker, Sulphadoxine-pyrimethamine, East Africa

Introduction

Malaria is a mosquito-borne infectious disease of humans and animals, which is caused by a protozoan parasite of the genus *Plasmodium* [1]. More than 100 species of *Plasmodium* can infect numerous animals, but only four species of parasite can infect humans, such as *Plasmodium falciparum*, *Plasmodium vivax*, *Plasmodium ovale*, and *Plasmodium malariae* [2]. However, *P. falciparum* is the most common and detrimental malaria parasite, accounting for 99.7% of malaria cases, and it frequently causes severe disease and death, particularly in the World Health Organization African region [3, 4]. In 2020, around 241 million cases and 627 thousand deaths of malaria were estimated worldwide [5]. *P. falciparum* infection is the cause of over 90% of malaria deaths worldwide, making it a persistent danger to public health on a global scale [6]. Over 200 million clinical cases and over 400,000 fatalities in Africa are caused by *P. falciparum* infection each year, accounting for 92% of the malaria burden worldwide [7].

De novo drug resistance to malaria must arise from spontaneous mutations or gene duplications that confer reduced drug susceptibility [8]. These mutations or duplications are selected in individuals when antimalarial drug concentrations are high enough to kill or inhibit susceptible parasites while allowing the resistant clones to thrive [9]. The emergence and spread of drug resistance to commonly used antimalarial drugs pose a severe threat to effective malaria control and treatment [10]. Over the years, the effectiveness of various antimalarial drugs has been compromised by the emergence and spread of drug-resistant malaria parasites [11]. Antimalarial resistance in non-falciparum species has developed more slowly,

possibly due to fewer genetic mutations and lower parasite loads in the human host; for instance, in *P. vivax* and *P. ovale*, this slower development is likely related to their ability to evade blood schizonticides by forming hypnozoites in the liver [12]. The worldwide efforts against malaria are facing significant obstacles due to the increasing prevalence of *P. falciparum* resistance to vital anti-malarial drugs [13]. Drug-resistant *P. falciparum* caused a disastrous rise in sub-Saharan African countries, where case incidence and mortality doubled or tripled [14]. Drug-resistant *P. falciparum* is still found in a variety of locations around the world today, partly due to trends in drug delivery and the intensity of transmission [15].

Despite well-documented resistance to sulfadoxine-pyrimethamine (SP), it remains the drug of choice for treating intermittent malaria and assessing resistance levels to inform national policy decisions [16]. Genetic changes (mutations) in the *P. falciparum* dihydrofolate reductase (*pfdhfr*) and dihydropteroate synthetase (*pfdhps*) genes confer SP resistance [17]. The *pfdhps* A437G codon, when combined with the *pfdhfr* triple mutant/codon change at N51I, C59R, and S108 N allele, has been linked to treatment failure [18]. In East Africa, a quintuple mutant genotype with the *pfdhfr* triple mutant and the *pfdhps* double (A437G + K540E) mutations is a significant predictor of SP treatment failure [19].

Molecular markers play a crucial role in tracking and understanding the prevalence of antimalarial drug resistance, monitoring the efficacy of existing antimalarial treatments, and guiding the development of new therapeutic strategies [20, 21]. These markers are specific genetic variations or mutations within the parasite's genome that confer resistance to antimalarial

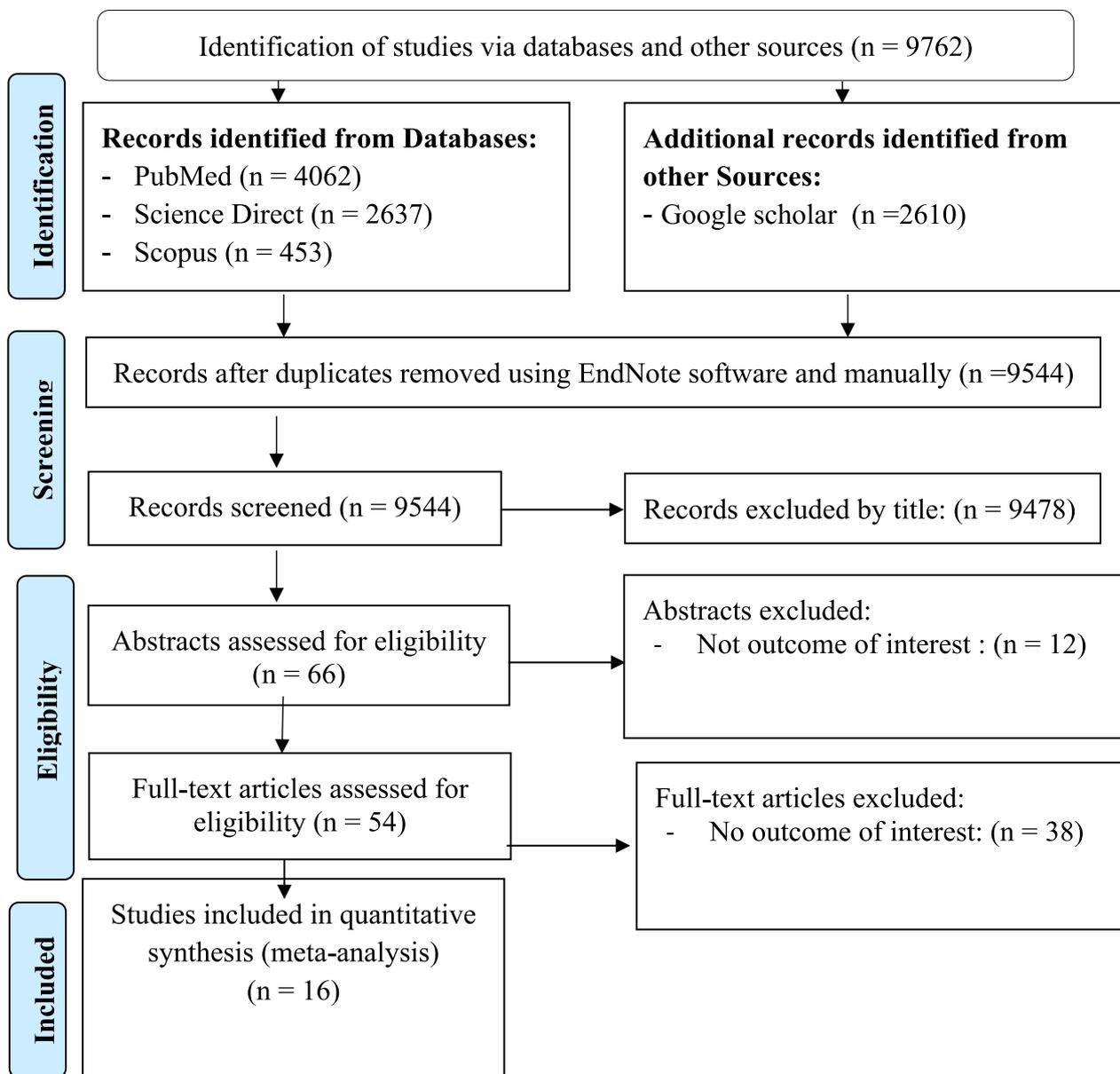


Fig. 1 PRISMA flow diagram indicated the results of the search and reasons for exclusion [32]

drugs [22]. By analyzing these molecular markers, scientists can identify and monitor the spread of drug-resistant malaria parasites in different regions [23]. The prevalence of molecular markers of antimalarial drug resistance varies across different geographical areas and parasite species [24]. For instance, in some regions, there is a high prevalence of molecular markers associated with resistance to artemisinin-based combination therapies, the current frontline treatment for malaria [25]. Besides, the prevalence of molecular markers of anti-malaria drug resistance is a critical issue in the field of malaria

research and public health [26]. This highlights the urgent need for continued surveillance and research to develop new antimalarial drugs and strategies to combat drug resistance [27].

Researchers can determine the degree of drug resistance and adjust treatment regimens by studying the frequency of particular molecular markers, such as *Pfdhfr* and *Pfdhps*, among malaria parasite populations in different regions [28]. This information is crucial for informing national malaria control programs, guiding the selection of appropriate antimalarial drugs, and preventing the

Table 1 Summary of included studies in systematic review and meta-analysis

Author	Year of publication	Country	study design	Gene	Marker	Number of sample sequenced (N)	prevalence of mutation (N/%)	Quality score/9						
Gabriel et al. [33]	2019	Kenya	Cohort study	Pfdhfr	N51I	76	71/93.4	9						
					S108 N	76	70/92							
					Pfdhps	A437G	76		71/93.4					
						K540E	76		69/90.8					
Fredrick et al. [48]	2015	Rwanda	Cross sectional study	Pfdhfr	N51I	376	375/99.7	8						
					C59R	381	344/90.3							
					S108 N	379	379/100							
					I164L	377	19/5							
					Quadruple (N51I-C59R-S108 N-I164L)	399	18/4.5							
					Triple (N51I-C59R-S108 N)	399	339/85							
					Pfdhps	A437G	364		338/93					
						K540E	363		343/94					
						Double (A437G-K540E)	399		334/83.7					
						Triple (A437G-K540E-A581G)	402		339/84.3					
					Paulo et al. [35]	2019	Mozambique		Cross sectional study	Pfdhfr	N51I	100	89/89	9
C59R	100	86/86												
S108 N	100	90/90												
I164L	100	2/2												
Triple [N51I-C59R-S108 N]	100	90/90												
Pfdhps	A437G	100	74/74											
	K540E	100	79/79											
	Double (A437-K540E)	100	72/72											
	Triple [A437G-K540E-A581G]	14	11/79											
Vito et al. [43]	2015	Tanzania	Cross sectional study	Pfdhfr				CIRNI			264	186/70.4	9	
								CICNI			264	28/10.6		
					CNRNI	264	16/6							
					S108 N	264	244/92.4							
					C59R	264	221/83.7							
					Pfdhps	AGEAA	264	96/36.4						
						SGEAA	264	36/13.6						
						SGEGA	264	33/12.5						
					Zhiyong et al. [28]	2022	Kenya	Community based household surveys	Pfdhfr	N51I	323	323/100		9
										S108 N	346	323/93		
C59R	346	323/93												
Pfdhps	A437G	346	323/93											
	K540E	346	319/92											
Ashley et al. [47]	2023	Kenya	NA	Pfdhfr	N51I	53	41/77	7						

Table 1 (continued) drug-resistant malaria parasite strains [27]. Exploring the prevalence of molecular markers of

Author	Year of publication	Country	study design	Gene	Marker	Number of sample sequenced (N)	prevalence of mutation (N/%)	Quality score/9						
Irene et al. [46]	2023	South Sudan	Longitudinal time series	Pfdhps	C59R	52	40/77	8						
					S108 N	52	42/81							
					I164L	52	3/6							
					Quintuple (N51I-C59R- S108 N/ A437G-K540E)	106	44/41.5							
					K540E	48	38/79							
					A437G	42	42/100							
					Pfdhfr	51I	532		517/97					
						108 N	532		516/97					
						59R	532		447/84					
						IRNGE	532		332/62					
						581G	532		34/6.4					
						164L	532		4/0.8					
Monica et al. [41]	2015	Kenya	Cross sectional study	Pfdhps	437G	532	435/82	9						
					540E	532	460/86.5							
					Pfdhfr	Triple (51I/59R/108 N)	253		223/88					
						dhfr/dhps quintuple (dhfr 51I/59R/108 N+ dhps 437G/540E)	253		215/85					
					Pfdhps	Double (437G/540E)	253		243/96					
Melissa et al. [39]	2017	Kenya	Clinical trial study	Pfdhfr	N51I	120	115/95.8	8						
					Pfdhps	C59R	121		117/96.7					
						S108 N	121		117/96.7					
						A437G	50		48/96					
					Pfdhfr	K540E	113		103/91					
Stella et al. [45]	2020	Kenya	Cross sectional study	Pfdhfr	quintuple (51I + 59R + 108 N) + (437G + 540E)	106	91/85.8	9						
					51I	106	94/88.7							
					59R	106	83/78							
					108 N	106	99/93.4							
					Quadruple [(51I + 59R + 108 N) + 437G]	106	11/10.4							
					Quadruple [(59R + 108 N) + (437G + 540E)]	106	11/10.4							
					Triple (51I + 59R + 108 N)	106	4/37.7							
					Pfdhps	51I	106		94/88.7					
						437G	106		100/94					
						540E	106		97/91.5					
					Anthony et al. [40]	2015	Uganda		Molecular epidemiological study	Pfdhfr	108 N	120	120/100	9
											59R	116	112/96.6	
51I	111	111/100												

Table 1 (continued)

Author	Year of publication	Country	study design	Gene	Marker	Number of sample sequenced (N)	prevalence of mutation (N/%)	Quality score/9
					double (51I-108 N)	111	111/100	
					Double (59R-108 N)	116	112/96.6	
					triple (51I-59R-108 N)	110	107/97.3	
					Quadruple (51I- 59R- 108 N) –437G	110	107/97.3	
					Quintuple (51I –59R-108 N) (437G –540E)	110	106/96.4	
					Sextuple (51I- 59R- 108 N) (437G –540E –581G)	101	8/7.9	
				Pfdhps	581G	103	8/7.8	
					437G	120	120/100	
					540E	120	119/99	
					double (437G-540E)	120	119/99	
Mwiche et al. [42]	2015	Zambia	Cross sectional study	Pfdhfr	triple ((Asn-108 + Ile-51 + Arg-59)	72	49/68	9
				Pfdhps	Double (Gly –437 + Glu-540)	72	15/20.8	
Eugenia et al. [38]	2017	Ethiopia	Cross sectional study	Pfdhfr	51I	199	154/77.4	9
					59R	199	99/49.7	
					108 N	199	155/77.9	
				Pfdhps	396 K	199	72/36	
Dennis et al. [37]	2019	Kenya	Unblinded, randomized controlled trial	Pfdhfr	N51I	101	61/60.4	9
					C59R	101	55/54.5	
					S108 N	101	61/60.4	
					I164L	101	4/4	
				Pfdhps	A437G	101	64/63.4	
					K540E	101	66/65	
					A581G	101	1/1	
Reginald et al. [44]	2016	Tanzania		Pfdhps	K540E	928	821/88.5	8
					A581G	854	150/17.6	
Himanshu et al. [34]	2018	Mozambique	Descriptive observational study	Pfdhps	S436 F	345	10/2.8	9
					A437G	345	289/83.8	
					K540E	348	286/82	

NA = Non-applicable

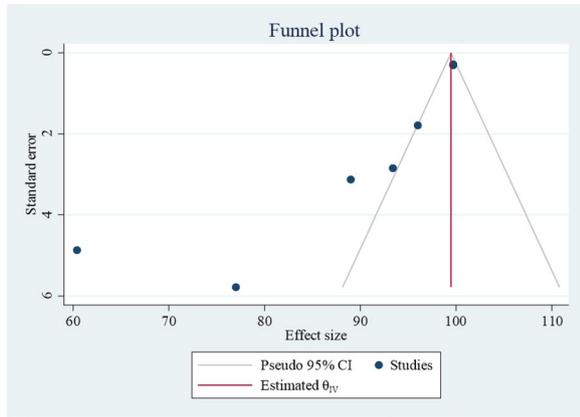


Fig. 2 Funnel plot for prevalence of dhfr N51I mutation

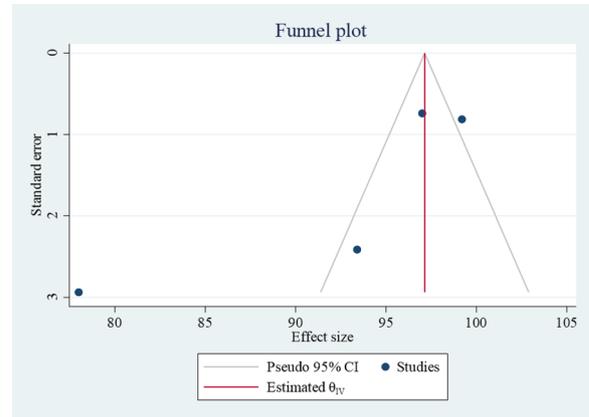


Fig. 5 Funnel plot for prevalence of dhfr 108 N mutation

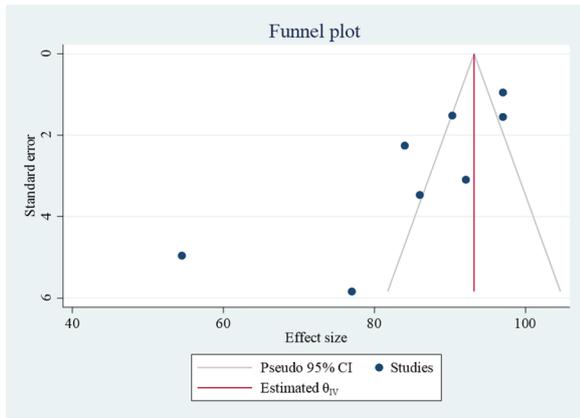


Fig. 3 Funnel plot for prevalence of dhfr C59R mutation

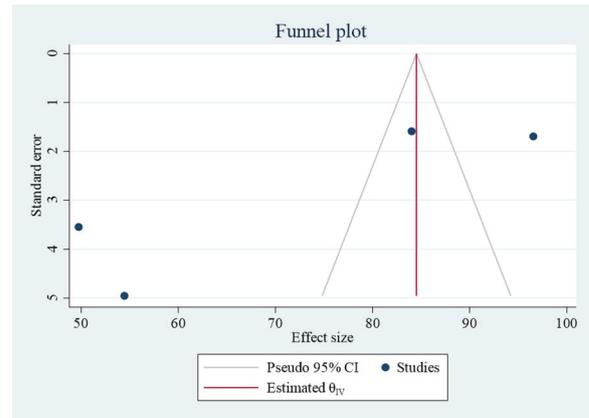


Fig. 6 Funnel plot for prevalence of dhfr 59R mutation

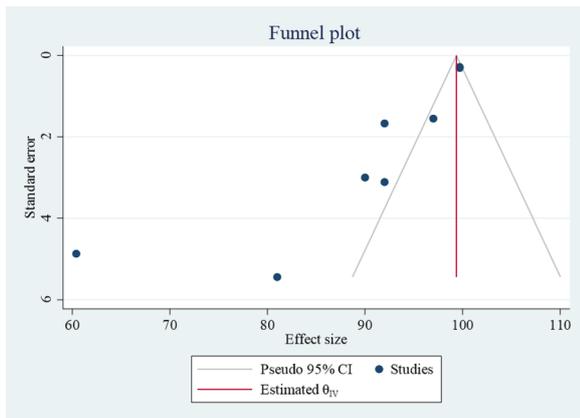


Fig. 4 Funnel plot for prevalence of dhfr S108 N mutation

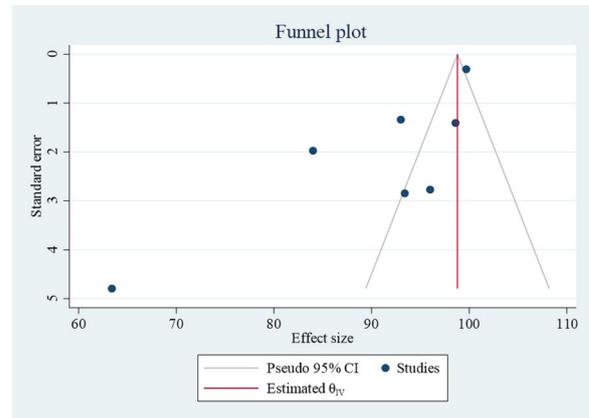


Fig. 7 Funnel plot for prevalence of dhps A437G mutation

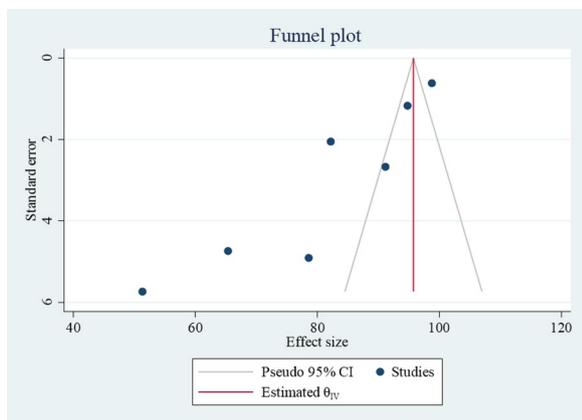


Fig. 8 Funnel plot for prevalence of dhps K540E mutation

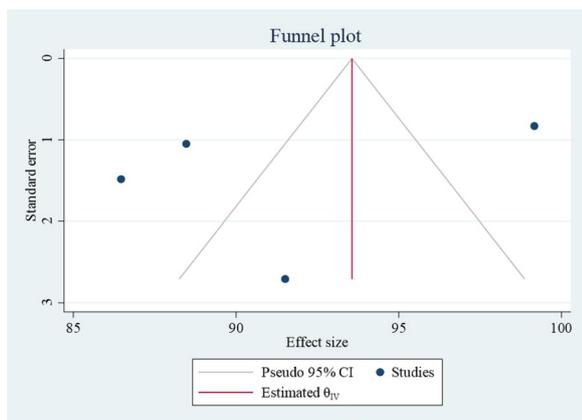


Fig. 9 Funnel plot for prevalence of dhps 540E mutation

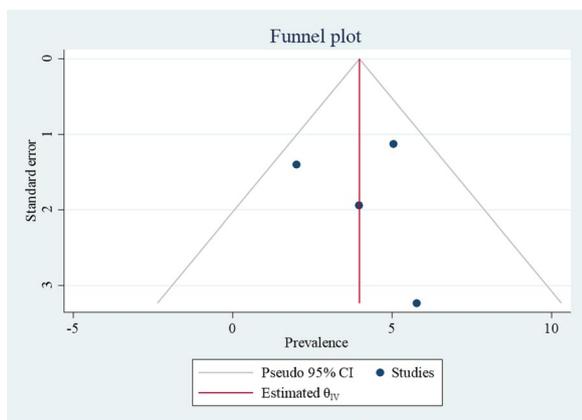


Fig. 10 Funnel plot for prevalence of dhfr I164L mutation

Table 2 Trim-and-fill analysis for the prevalence of dhfrN511 mutation

Studies	Effect size	[95% CI]
Observed (7)	88.55	78.2–98.87
Observed + Imputed (7 + 0)	88.55	78.2–98.87

Table 3 Trim-and-fill analysis for the prevalence of dhfr C59R mutation

Studies	Effect size	[95%CI]
Observed (8)	85.3	76.1–94.59
Observed + Imputed (8 + 2)	81.1	71.98–90.2

Table 4 Trim-and-fill analysis for the prevalence of dhfr S108 N mutation

Studies	Effect size	[95%CI]
Observed (8)	89.6	81.1–98.2
Observed + Imputed (8 + 2)	85.9	77.65–94.2

Table 5 Trim-and-fill analysis for the prevalence of dhfr 108 N mutation

Studies	Effect size	[95%CI]
Observed (4)	92.2	83.1–101.3
Observed + Imputed (4 + 1)	89.7	81.1–98.3

Table 6 Trim-and-fill analysis for the prevalence of dhfr 59R mutation

Studies	Effect size	[95%CI]
Observed (4)	71.5	49.2–93.77
Observed + Imputed (0)	71.5	49.2–93.77

Table 7 Trim-and-fill analysis for the prevalence of dhps A437G mutation

Studies	Effect size	[95%CI]
Observed (7)	90.2	81.46–98.98
Observed + Imputed (7 + 2)	86.39	78.1–94.7

anti-malaria drug resistance provides valuable insights into the dynamics of drug resistance in malaria parasites and underscores the importance of continuous

Table 8 Trim-and-fill analysis for the prevalence of dhps K540E mutation

Studies	Effect size	[95%CI]
Observed (7)	80.9	68.6–93.2
Observed + Imputed (7 + 0)	80.9	68.6–93.2

Table 9 Trim-and-fill analysis for the prevalence of dhps 540E mutation

Studies	Effect size	[95%CI]
Observed (4)	91.5	85.7–97.2
Observed + Imputed (4 + 0)	91.5	85.7–97.2

Table 10 Trim-and-fill analysis for the prevalence of dhfr I164L mutation

Studies	Effect size	[95%CI]
Observed (4)	3.9	2.1–5.77
Observed + Imputed (4 + 1)	3.8	2.1–5.51

surveillance and research efforts to combat this significant public health challenge [29].

Currently, there is insufficient information on the prevalence of molecular markers for SP-resistant *P. falciparum* and their implications for anti-malarial policies. This makes it more difficult to compare resistance patterns throughout the study area and to coordinate efforts to address the problem of drug resistance globally. We present a comprehensive study utilizing systematically extracted data from English-published and unpublished articles conducted over the past decade across East African countries. Therefore, this systematic review and meta-analysis aimed to determine the pooled prevalence of genetic changes responsible for the antimalaria drug, sulfadoxine-pyrimethamine resistance in *P. falciparum* in East African countries from 2014 to 2023.

Methods

Review protocol

We followed the Preferred Reporting Items for Systematic Review and Meta-analyses (PRISMA) guidelines to search articles from online databases, literature screening by title and abstract, and assess the full-text’s appropriateness. The review protocol was developed before literature searching and was registered with the International Prospective Register of Systematic Reviews (PROSPERO) database with registration number CRD42024580210.

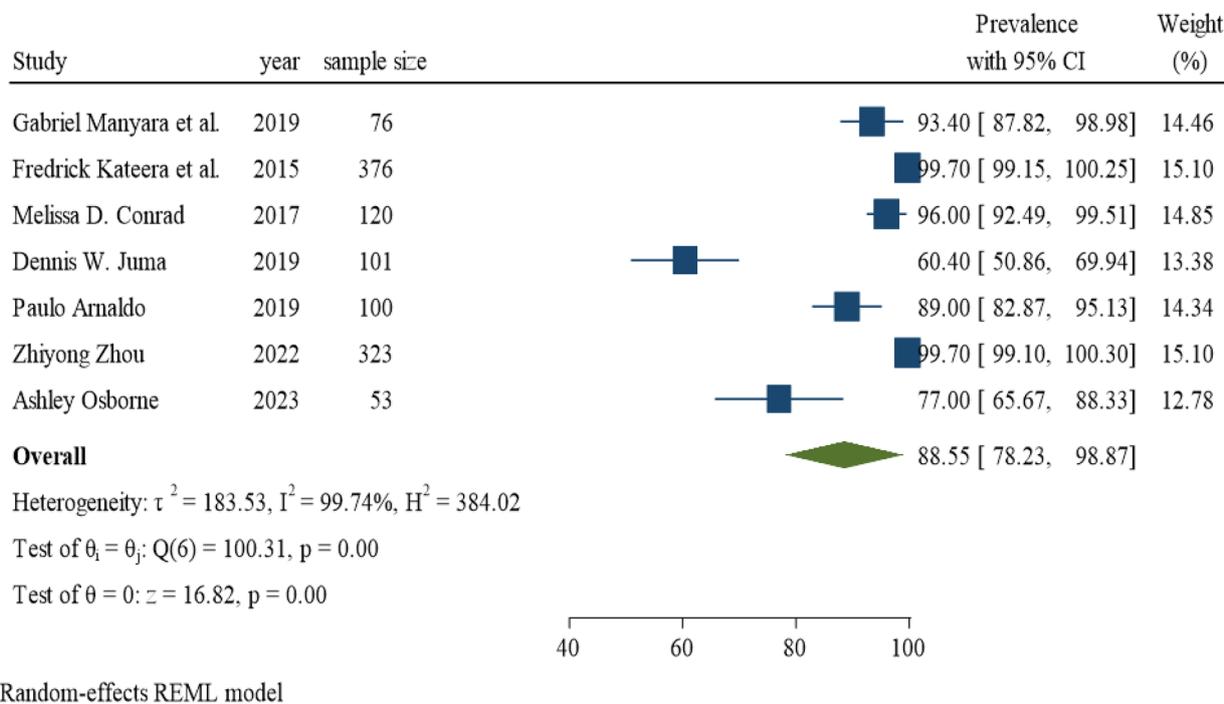
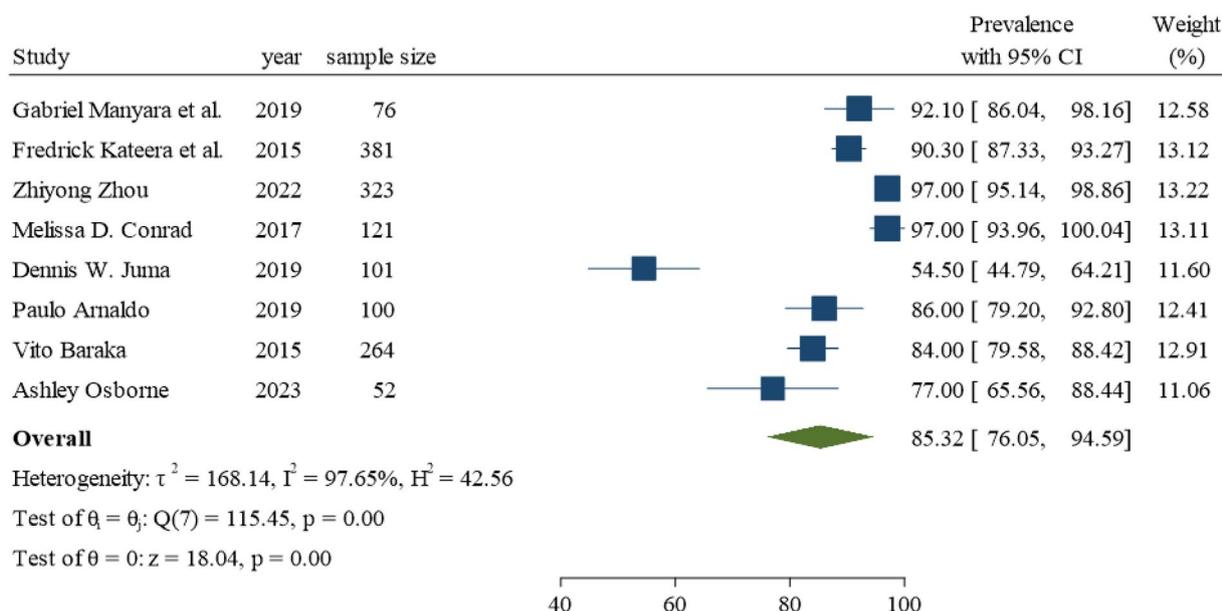
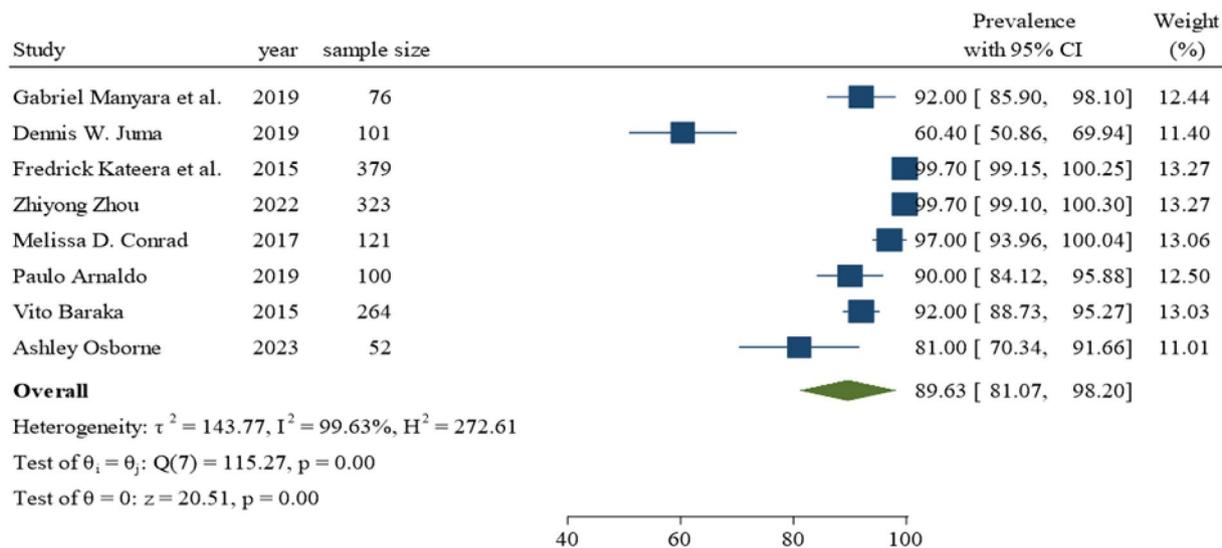


Fig. 11 Forest plot showing the prevalence of dhfr N51I mutation



Random-effects REML model

Fig. 12 Forest plot showing the prevalence of dhfr C59R mutation



Random-effects REML model

Fig. 13 Forest plot showing the prevalence of dhfr S108 N mutation

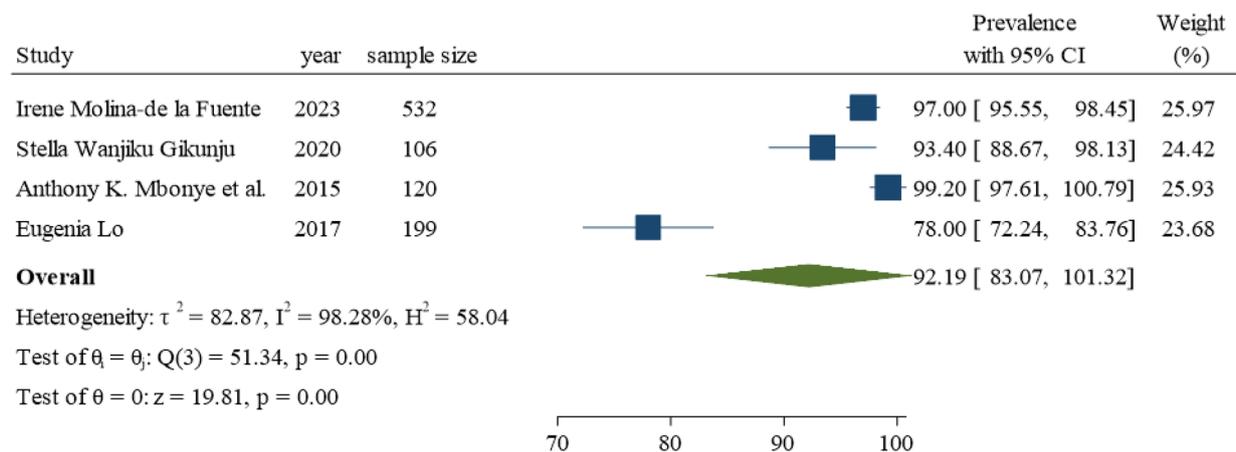
Data management

The data for the review articles was managed using EndNote software version X7 (Thomson Reuters, 2015). The software imported all detected article titles and removed duplicates. Then, using specified criteria, article titles were filtered and classified into several eligibility groups (included or excluded). The Excel 2010 data

extraction form was pretested on five articles and then changed depending on the pilot test results.

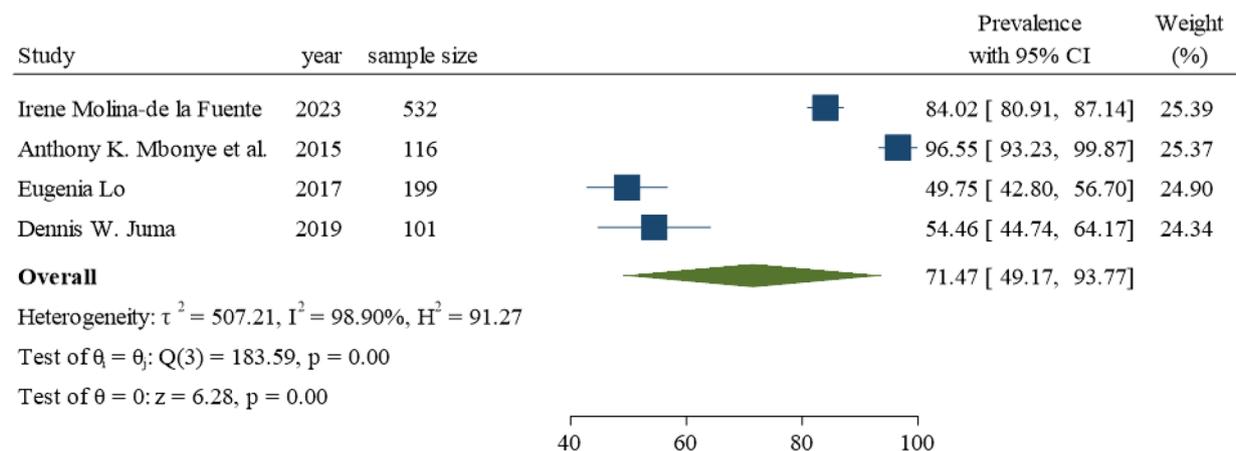
Search strategy

A systematic search strategy, utilizing a combination of keywords, was implemented to search for articles in PubMed, ScienceDirect, Scopus, and the Google Scholar



Random-effects REML model

Fig. 14 Forest plot showing the prevalence of dhfr 108 N mutation



Random-effects REML model

Fig. 15 Forest plot showing the prevalence of dhfr 59R mutation

search engine. Both interventional and observational studies were retrieved for inclusion in the review. The following MeSH search terms were combined using the Boolean operators “OR” and “AND”: “Prevalence”, “Epidemiology”, “Magnitude”, “Biomarkers”, “Molecular markers”, “*plasmodium*”, “*P. falciparum*”, “*P. vivax*”, “*P. ovale*”, “*P. malariae*”, “Drug resistance”, “Antimalaria”, “East Africa”, “2014 to 2023”.

Eligibility criteria

Inclusion criteria

The systematic review and meta-analysis covered the following types of studies: (a) papers published up to December 30, 2023, on human participants of all ages; (b) original articles from studies that explored

asymptomatic, uncomplicated, or severe malaria; (c) studies that included PCR genotyping of *P. falciparum* antimalarial drug resistance markers of SP; (d) research reporting the prevalence of molecular markers; (e) studies written in English; and (f) studies conducted in East African countries.

Exclusion criteria

The following study types were excluded: (a) abstracts; (b) studies on in vitro, ex vivo, and in vivo antimalarial drug resistance without genotyping and reporting marker prevalence; (c) studies on genetic diversity and population structure of *P. falciparum* without drug resistance; (d) studies on diagnostic accuracy of methods for detecting

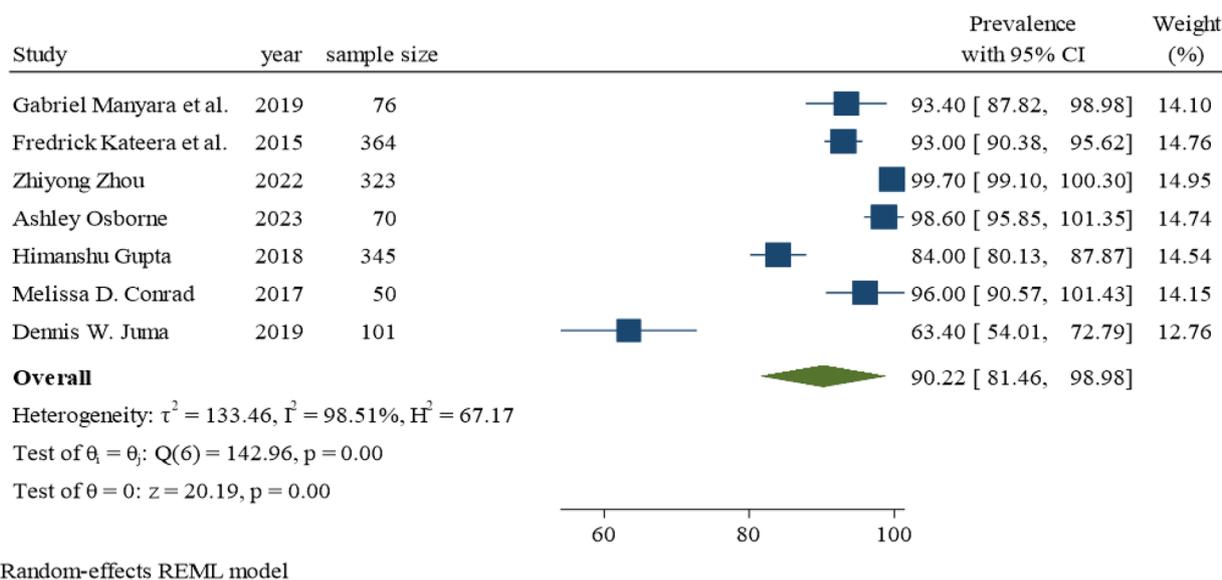


Fig. 16 Forest plot showing the prevalence of dhps A437G mutation

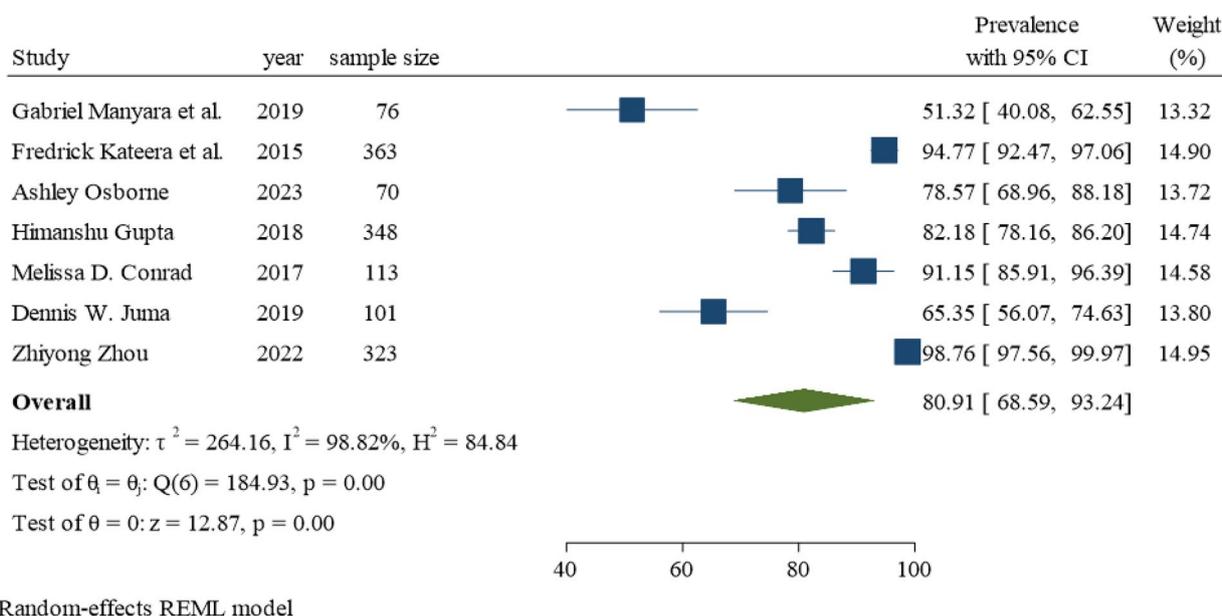


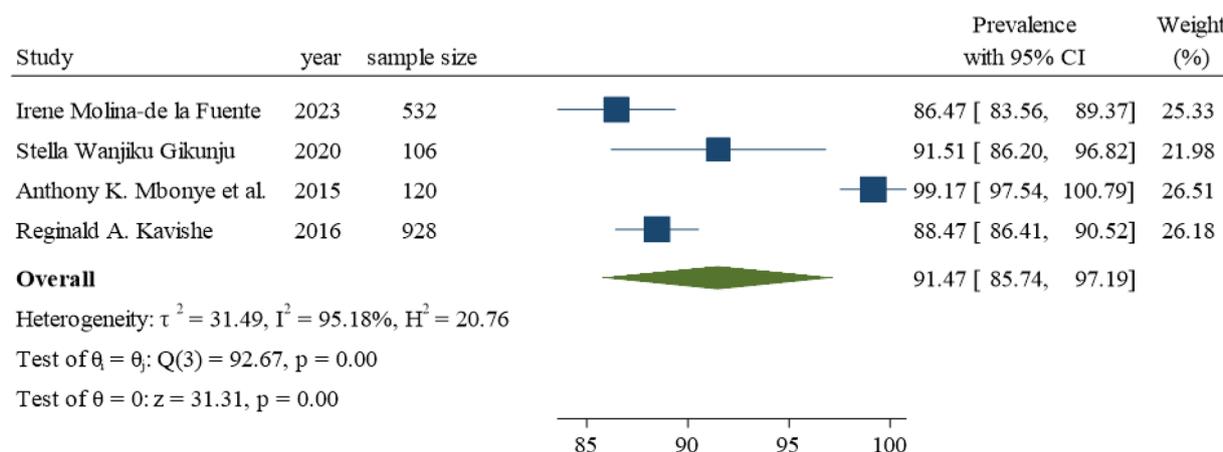
Fig. 17 Forest plot showing the prevalence of dhps K540E mutation

P. falciparum without genotyping for antimalarial drug resistance of marker prevalence.

Review process

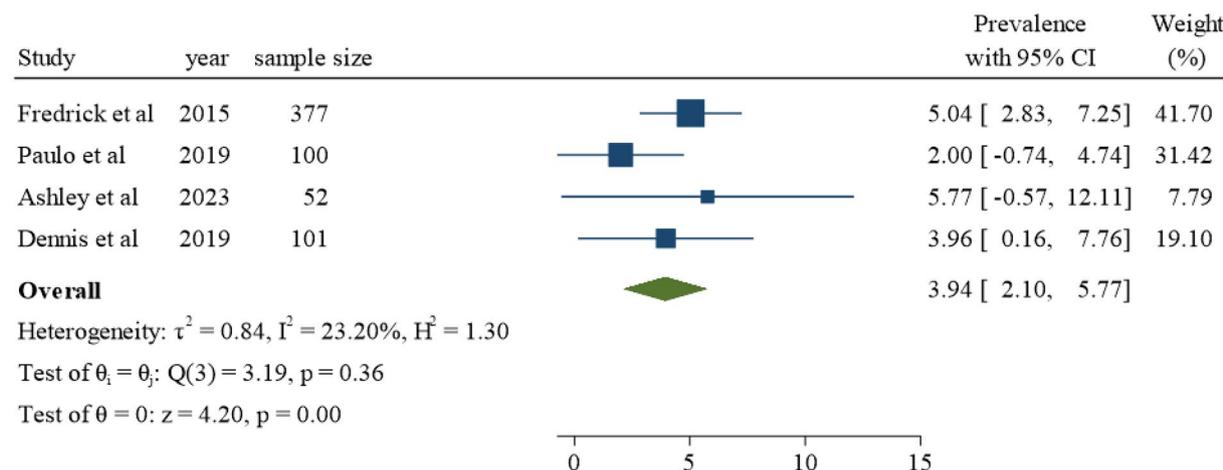
Research articles found through electronic database searches were assessed for eligibility based on their titles, abstracts, and full text. The ineligible articles

and duplicates were eventually removed. Before data extraction began, full-length articles from the selected studies were read to ensure that they met the inclusion criteria. Two independent reviewers (W.A. and Z.A.) inspected the titles and abstracts to identify potentially suitable studies, as well as data derived from full-length articles fulfilling the inclusion criteria.



Random-effects REML model

Fig. 18 Forest plot showing the prevalence of dhps 540E mutation



Random-effects REML model

Fig. 19 Forest plot showing the prevalence of dhfr I164L mutation

Outcome of interest

The major outcome of interest was the prevalence of antimalarial drug resistance-conferring mutations associated with sulphadoxine-pyrimethamine resistant *P. falciparum* in the original paper, expressed as a percentage and the number of cases (n)/total number of participants (N).

Quality assessment

The quality of the articles was assessed using the Joanna Briggs Institute’s (JBI) critical assessment checklist for simple prevalence [30]. Two independent investigators (G.K. and A.A.) assessed the quality of the full-text articles. Disputes were resolved through discussion to

reach an agreement and accept or reject the articles for study. This systematic review and meta-analysis includes studies having a final quality score of at least 50%.

Data extraction procedure

The relevant data extraction was done using Microsoft Excel, an established data extraction tool. This extracting tool contained information about the author(s) names, study site, sample size, study design, sequence genotyping success rate, anti-malarial drug resistance gene (markers), total number of samples genotyped, number of samples genotyped with mutations, and prevalence of molecular markers. Five reviewers (W.A., A.S., M.N., M.A.R., and W.K.) assessed the extracted data for

Table 11 Summary of subgroup analysis of *P. falciparum* molecular markers by year of publication and Country

Variables	Characteristics	Molecular markers	Prevalence with 95% CI	I ² , p-value
Years of publication	2014–2018	dhfr N51I	97.11 [93.48–100.75]	75.96, < 0.001
		dhfr C59R	90.57 [83.29–97.84]	92.68, < 0.001
		dhfr S108 N	96.45 [92.02–100.88]	91.02, < 0.001
		dhpsA437G	90.89 [83.92–97.85]	89.86, < 0.001
		dhpsK540E	89.45 [81.97–96.93]	91.70, < 0.001
	2019–2023	dhfr N51I	82.03 [65.36,98.71]	96.91, < 0.001
		dhfr C59R	81.78 [67.19–96.37]	96.72, < 0.001
		dhfr S108 N	85.12 (72.04–98.21)	96.76, < 0.001
		dhpsA437G	89.24[72.98–105.50]	99.13, < 0.001
		dhpsK540E	73.98 [53.96–93.99]	96.78, < 0.001
Countries	Kenya	dhfr N51I	85.88 [71.68,100.07]	98.39, < 0.001
		dhfr C59R	84.02 [68.27,99.77]	98.53, < 0.001
		dhfr S108 N	86.56 [72.73,100.38]	98.54, < 0.001
		dhps A437G	93.40 [87.82,98.98]	98.72, < 0.001
		dhps K540E	77.55 [60.77,94.34]	97.68, < 0.001
	Mozambique	dhps A437G	84.00 [80.13,87.87]	0, < 0.001
		dhfr N51I	89.00 [82.87,95.13]	0, < 0.001
		dhfr C59R	86.00 [79.20,92.80]	0, < 0.001
		dhfr S108 N	90.00 [84.12,95.88]	0, < 0.001
		dhps K540E	82.12 [78.16,86.20]	0, < 0.001
	Rwanda	dhps A437G	93.00 [90.38,95.62]	0, < 0.001
		dhfr N51I	99.70 [99.15,100.25]	0, < 0.001
		dhfr C59R	90.30 [87.33,93.27]	0, < 0.001
		dhfr S108 N	99.70 [99.15,100.25]	0, < 0.001
		dhps K540E	94.77 [92.47,97.06]	0, < 0.001
Tanzania	dhfr C59R	84 [79.58,88.42]	0, < 0.001	
	dhfr S108 N	92 [88.73,95.27]	0, < 0.001	

correctness and consistency. The sixth reviewer (B.B.A.) was also consulted if needed.

Data analysis

The relevant primary research was retrieved, imported into Microsoft Excel, and exported to STATA 17.0 software (StataCorp, Texas, USA) for final analysis. Forest plots were used to estimate the pooled effect size and effect of each study, along with their confidence interval (CI), and to generate a visual representation of the data. The inverse of variance (I^2) was used to evaluate the degree of heterogeneity among the included studies [31]. The inverse of variance (I^2) values of 25%, 50%, and 75% were thought to indicate low, medium, and high heterogeneity, respectively. The selected studies were assessed for potential publication bias using a funnel plot. Trim and fill meta-analyses were used to assess and adjust for the observed publication bias in the studies, as well as to estimate the number of potentially missing studies. We used a random effect model to analyze the pooled estimate because of the significant heterogeneity seen across

studies. Studies with substantial heterogeneity were subjected to a subgroup analysis based on certain categories.

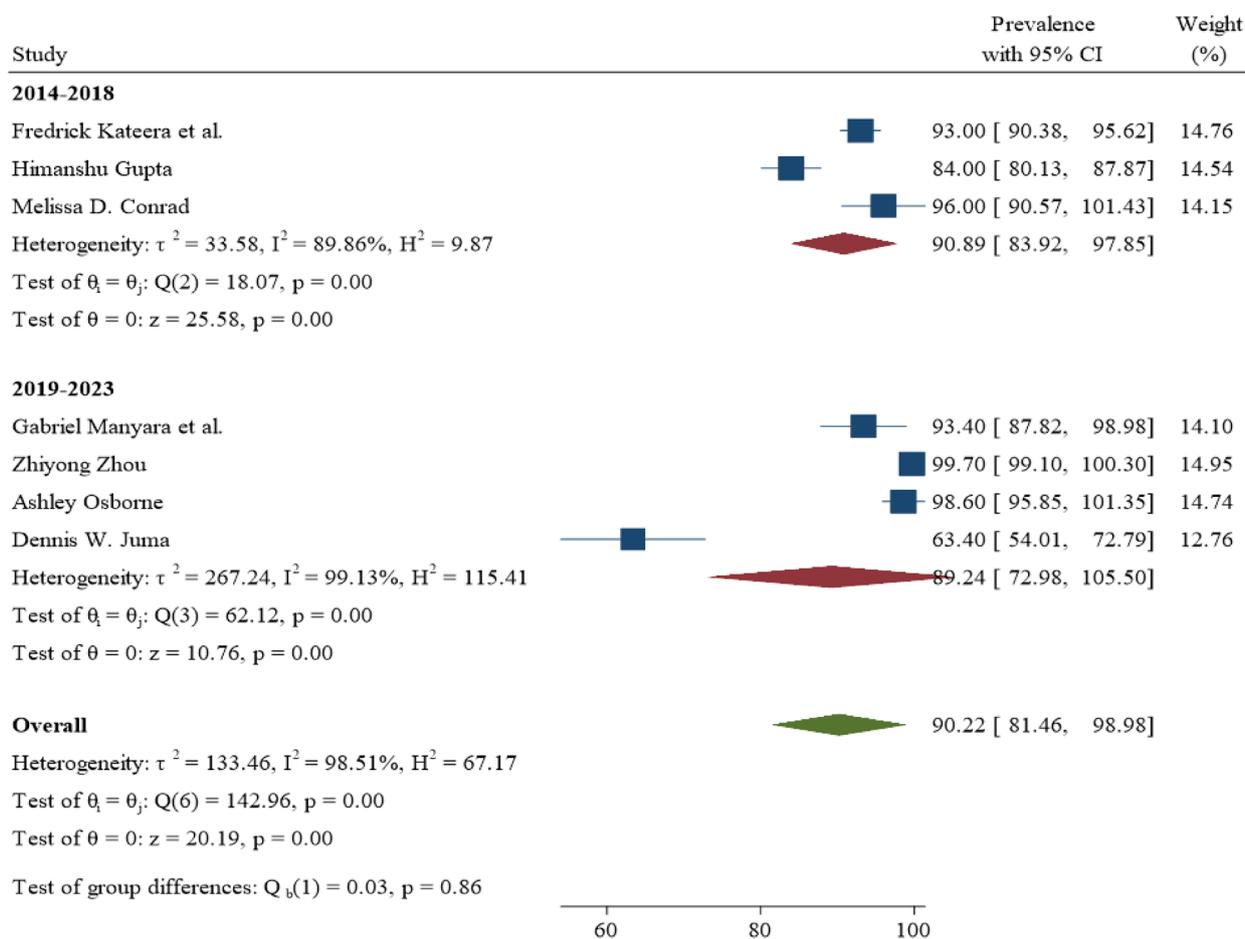
Results

Searching results

The electronic searches yielded a total of 9760 English-published articles and 2 unpublished articles on anti-*P. falciparum* drug resistance markers in East African countries. A total of 9544 studies were identified, after which 218 duplicates were removed. A total of 9544 studies were screened to remove studies by title, abstract, and full-text articles, with 16 studies retained after the screening and eligibility process. Finally, 16 studies were included for both qualitative and quantitative analyses (Fig. 1).

Characteristics of included studies

This study encompasses participants of all ages and genders. A total of 16 studies were included in this systematic review and meta-analysis [28, 33–47]. Out of 16 studies included, 14 studies were obtained from



Random-effects REML model

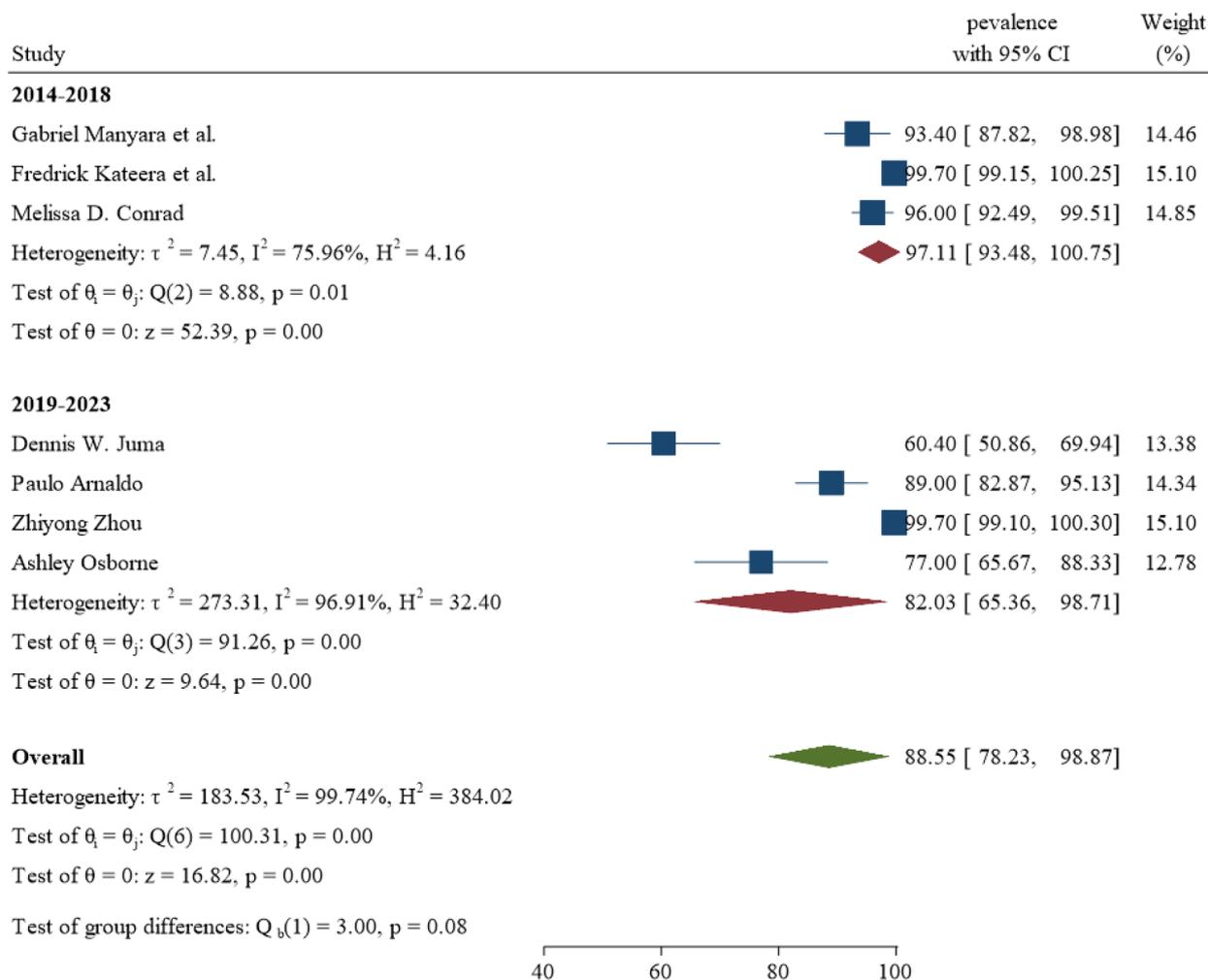
Fig. 20 Subgroup analysis for pooled prevalence of dhps A437G mutations from 2014 to 2023

published articles [28, 33, 34, 37–41, 43–47] and 2 from unpublished data [35, 36]. Among the included studies most studies were from Kenya [28, 33, 37, 41, 45, 47]. Two primary regions determining resistance to SP antimalaria drugs were chosen, and quantitative synthesis was drawn, *Pfdhfr* (n = 14 studies) [28, 33, 35, 37–43, 45–48] and *Pfdhps* (n = 16 studies) [28, 33–35, 37–48]. Among the included studies 87.5% and 100% of them were studied *Pfdhfr* and *Pfdhps*, respectively (Table 1). Five studies reported the presence of double mutation with prevalence of (A437G-K540E) (84%), (59R-108 N) (97%), (437G/540E) (96%), (51I-108 N) (100%), (Gly -437 + Glu-540) (21%), (A437-K540E) (72%), and (437G-540E) (99%) [35, 36, 38, 40, 41]. Similarly, six studies reported the presence of triple mutation with a prevalence of (N51I-C59R-S108 N) (85%), [N51I-C59R-S108 N] (90%), (51I/59R/108 N) (88%), (51I-59R-108 N) (96%), (51I + 59R + 108 N) (4%), (Asn-108 + Ile-51 + Arg-59) (68%), [A437G- K540E-A581G] (79%), and

(A437G-K540E-A581G) (84%) [35, 36, 40–42, 45]. Also, three studies reported the presence of quadruple mutation with prevalence of (N51I-C59R-S108 N-I164L) (4.5%), (51I + 59R + 108 N) + 437G] (10.4%), (59R + 108 N) + (437G + 540E) (10.4%), and (51I-59R- 108 N-437G) (97%) [36, 45]. Likewise, four studies reported the presence of quintuple mutant with prevalence of (N51I-C59R-S108 N-A437G-K540E) (63%), (51I-59R-108 N-437G-540E) (85%), (51I -59R -108N -437G-540E) (86%), and (51I -59R-108 N-437G -540E) (96%) [40, 41, 45, 47]. Moreover, Only one study reported the presence of sextuple mutant mutation with prevalence of (51I-59R-108 N) (437G-540E-581G) (8%) [40].

Heterogeneity and publication bias of included studies

The heterogeneity was assessed for all markers that are incorporated in different studies on *P. falciparum* markers that confer SP resistance. Except for dhfr I164L, there was significant heterogeneity across all markers, with I^2



Random-effects REML model

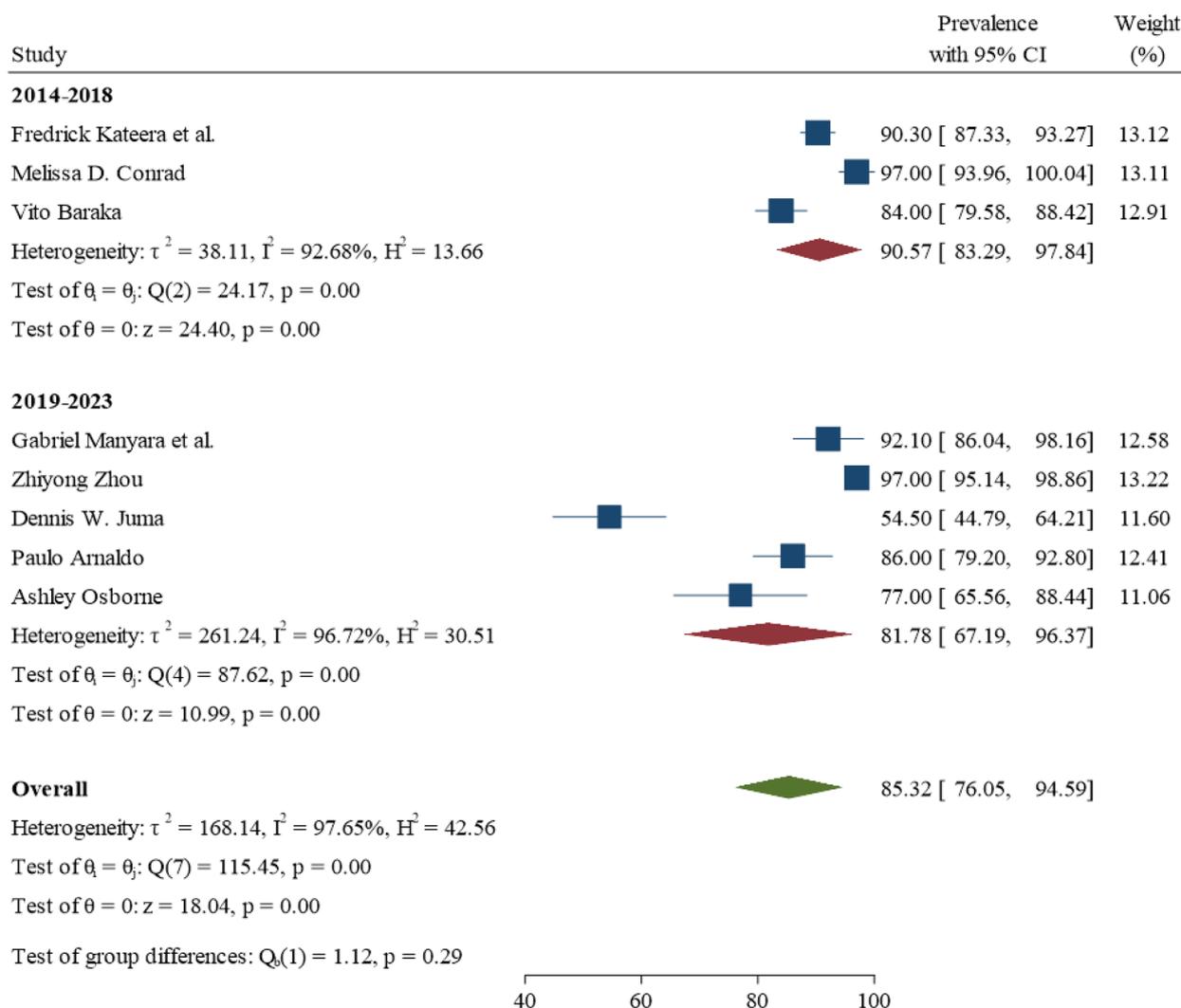
Fig. 21 Subgroup analysis for pooled prevalence of dhps N511 mutations from 2014 to 2023

statistics showing higher than or equal to 95.18% at P value = 0.00. A funnel plot was used for assessing potential publication bias in the included studies. As a result, the funnel plot showed asymmetry, indicating that publication bias existed among studies. To reduce and correct for the observed publication bias in the studies, a trim and fill analysis was done to estimate the number of potentially missing studies. After adjusting for publication bias, the estimated pooled prevalence of dhfr N51I, dhfr C59R, dhfr S108 N, dhfr 108 N, dhfr 59R, dhps A437G, dhps K540E, dhps 540E, and dhfr I164L was 88.55 (95% CI 78.2–98.87), 81.1 (95% CI 71.98–90.2), 85.9 (95% CI 77.65–94.2), 89.7 (95% CI 81.1–98.3), 71.5(95% CI 49.2–93.77), 86.39 (95% CI 78.1–94.7), 80.9 (95% CI 68.6–93.2), 91.5 (95% CI 85.7- 97.2), and 3.8 (95%

CI 2.1–5.51), respectively, based on trim and fill analysis Figs. 2, 3, 4, 5, 6, 7, 8, 9, 10 and Tables 2, 3, 4, 5, 6, 7, 8, 9, 10.

Pooled prevalence of *P. falciparum* anti-malarial drug resistance determining mutations

The analyses of molecular markers revealed that the aggregated prevalence of dhfr N51I, dhfr C59R, dhfr S108 N, dhfr 108 N, dhfr 59R, and dhfr I164L, were 88.6% [95% CI 78.2–98.9], 85.3% [95% CI 76.1–94.6], 89.6% [95% CI = 81.1–98.2], 92.2% [95% CI 83.1–101.3%], 71.5% [95% CI 49.2–93.8], and 3.9% [95% CI 2.1- 5.77], respectively. Likewise, analyses of molecular markers revealed that the aggregated prevalence of dhps A437G, dhps K540E, and dhps 540E were 90.2% [95% CI 81.5–99], 80.9% [95% CI



Random-effects REML model

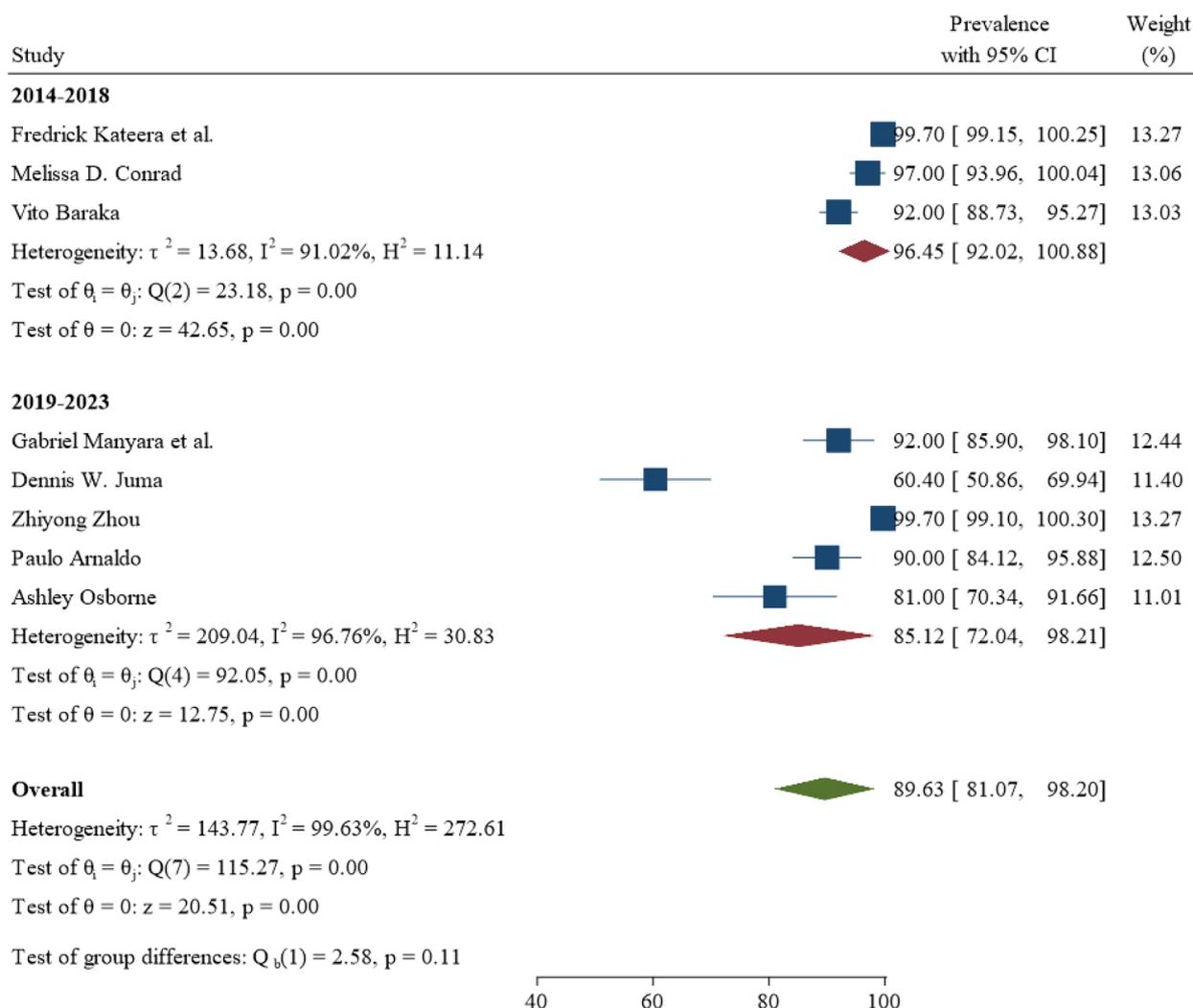
Fig. 22 Subgroup analysis for pooled prevalence of dhfr C59R mutations from 2014 to 2023

68.6–93.2], and 91.5% [95% CI 85.7–97.2%], respectively (Figs. 11, 12, 13, 14, 15, 16, 17, 18, 19).

Subgroup analysis of *P. falciparum* molecular marker by year of publication and country

There was high significant heterogeneity among the included studies. Inverse of variance (I^2) statistics showed greater than or equal to 95.18% heterogeneity among studies for all molecular marker like dhfr (N51I, C59R, S108 N, 108 N, and 59R) and dhps (A437G, K540E, & 540E). To identify the possible source of heterogeneity, subgroup analysis was performed for each molecular markers by year of publication and country. The subgroup analysis by year of publication analysis showing that the pooled prevalence of dhfr N51I, dhfr C59R,

dhfr S108 N, dhps A437G, & dhps K540E, in 2014–2018 [97.11% (95% CI 93.48–100.75%), 90.57% (95% CI 83.29–97.84%), 96.45% (95% CI 92.02–100.88%), 90.89% (95% CI 83.92–97.85%), and 89.45% (95% CI 81.97–96.93%)], and 2019–2023 [82.03% (95% CI 65.36,98.71%), 81.78% (95% CI 67.19–96.37%), 85.12% (95% CI 72.04–98.21%), 89.24% (95% CI 72.98, 105.50%), and 73.98% (95% CI 53.96–93.99%)], respectively. A similar pattern was also observed on country based analysis showed that the pooled prevalence of dhfr N51I, dhfr C59R, dhfr S108 N, dhps A437G, & dhps K540E, in Kenya was 85.88% (95% CI 71.68,100.07%), 84.02% (95% CI 68.27,99.77%), 86.56% (95% CI =72.73,100.38%), 90.7% (95% CI 78.02,103.39%), and 77.55% (95% CI 60.77–94.34%), respectively (Table 11) and (Figs. 20, 21, 22, 23, 24, 25, 26, 27, 28, 29).



Random-effects REML model

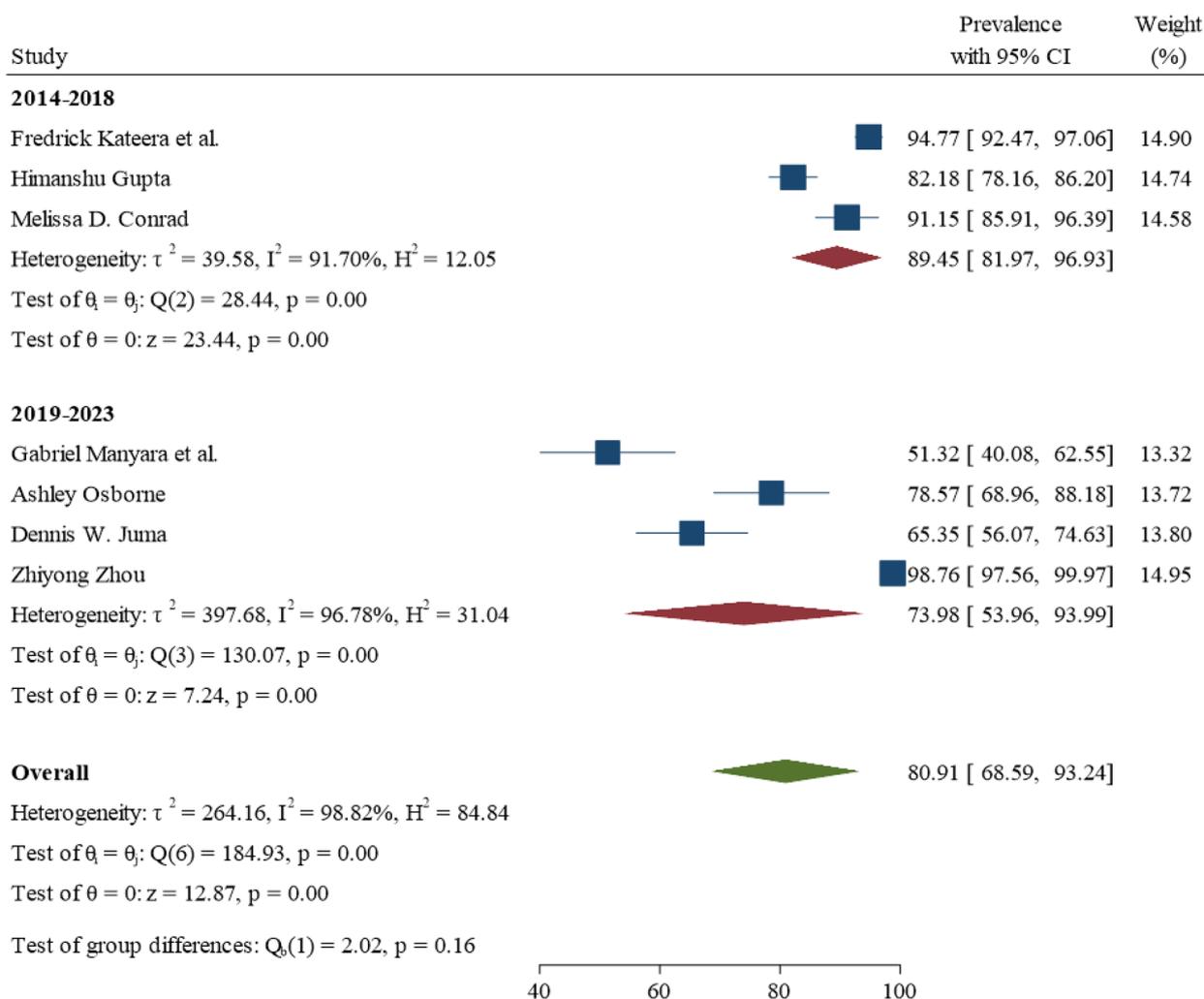
Fig. 23 Subgroup analysis for pooled prevalence of dhfr S108 N mutations from 2014 to 2023

The meta-analysis showed no significant difference in all molecular markers prevalence like dhfr N51I, dhfr C59R, dhfr S108 N, dhps A437G, & dhps K540E among studies on year of publication. However, the meta-analysis showed significant difference in all molecular markers prevalence like dhfr N51I, dhfr S108 N, dhps A437G, & dhps K540E among studies on the country level except dhfr C59R.

Discussion

This systematic review and meta-analysis showed the frequency of *P. falciparum* drug resistance markers of SP over a period of ten years in East Africa. In this systematic review and meta-analysis, the pooled prevalence of dhfr N51I was 88.6%. This finding was higher than that

reported in Nepal [49] and Ghana [50]. This could be due to the widespread use of antifolate drugs, like SP, which can selectively influence the parasite population, causing drug-resistant mutations like dhfr N51I to occur and spread. Additionally, this may suggest that the region in problem is within a stratum with a high risk of malaria transmission and is an urban context with a high degree of variability and intensity in the use of anti-malarial medications and inadequate regulation. Also, this finding was inline with that reported in Senegal [51], Nigeria [52], Central African countries [53], and China [54]. This hypothesizes that the parasites are subjected to similar drug pressure in nations, or that the unrestricted movement of people for work and other purposes among



Random-effects REML model

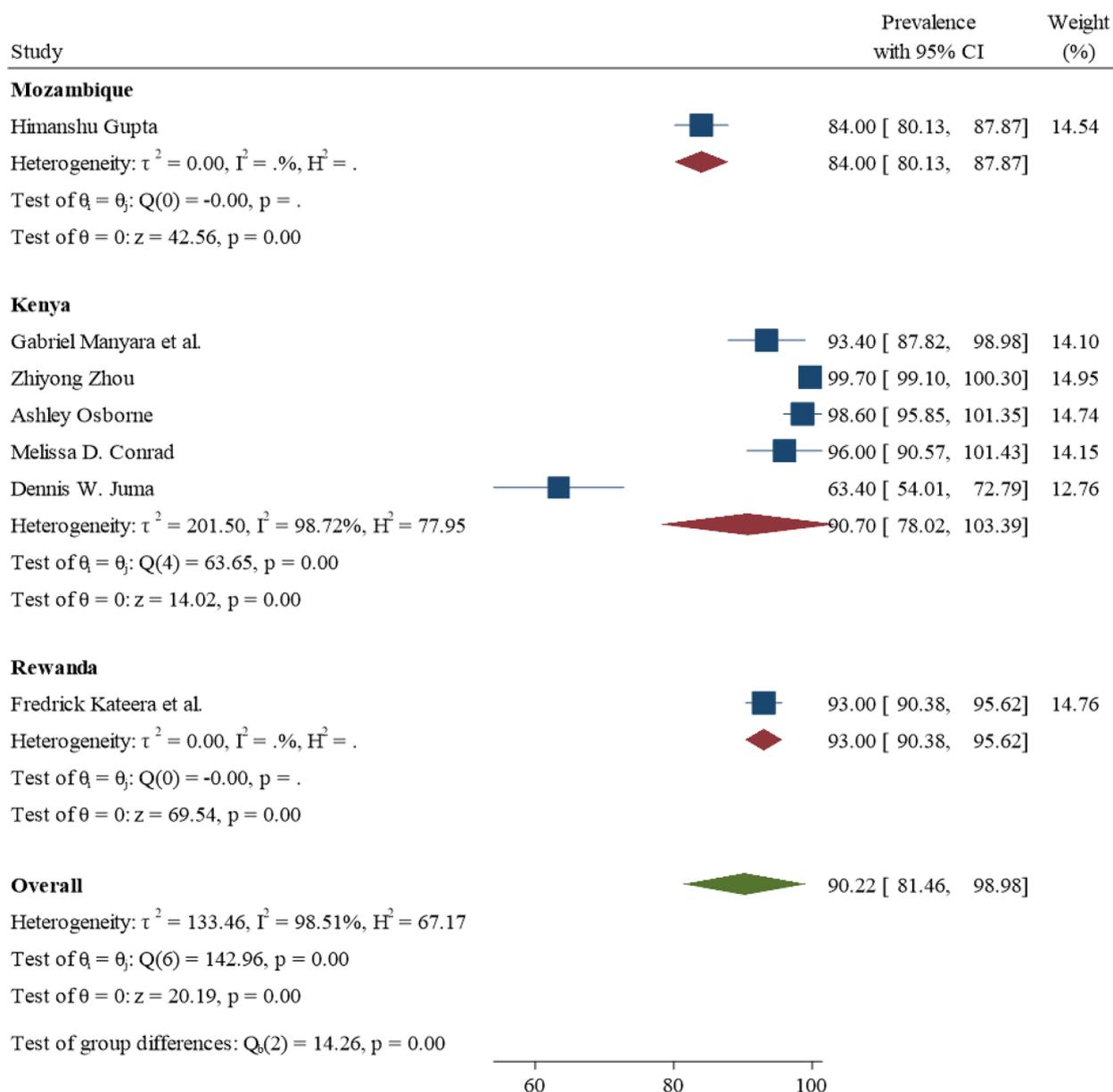
Fig. 24 Subgroup analysis for pooled prevalence of dhps K540E mutations from 2014 to 2023

countries is responsible for the spread of parasites with similar drug resistance profiles.

Similarly, in this systematic review and meta-analysis, the pooled prevalence of dhfr C59R was 85.3%. This finding was inline with that reported in central African countries [53] and China [54]. But this finding was higher than that reported in India [55, 56], Senegal [51], and Ghana [50]. However, this finding was lower than that reported in Nepal [49], Mali [57], and Nigeria [52]. This might be due to the emergence and dissemination of drug resistance mutations like dhfr C59R which are caused by insufficient dosage, unfinished treatment regimens, or the use of substandard antimalarial medications. Furthermore, the high frequency of mutations could be attributed to the use of SP in groups like young children and pregnant women, who serve as reservoirs for infections with resistance alleles as a

direct result of continuous use of SP in seasonal malaria chemotherapy and intermittent preventive treatment of malaria in pregnancy, initiatives that support the alleles' spread among the general population. Since SP is widely accessible at health centers and pharmacies in the study areas, its illegal use for self-medication may be a further major problem [58].

Likewise, in this review, the pooled prevalence of dhfr S108 N was 89.6%. This finding was lower than that reported in Central African Countries [53]. However, this finding was higher than that of reported in India [55, 56], Ghana [50], and Haiti [59]. This might be due to the parasite population in places with high genetic diversity is more likely to contain a range of drug-resistant mutations, such as dhfr S108 N. Furthermore, the large-scale deployment of intermittent preventive treatment for malaria prevention in pregnancy and seasonal



Random-effects REML model

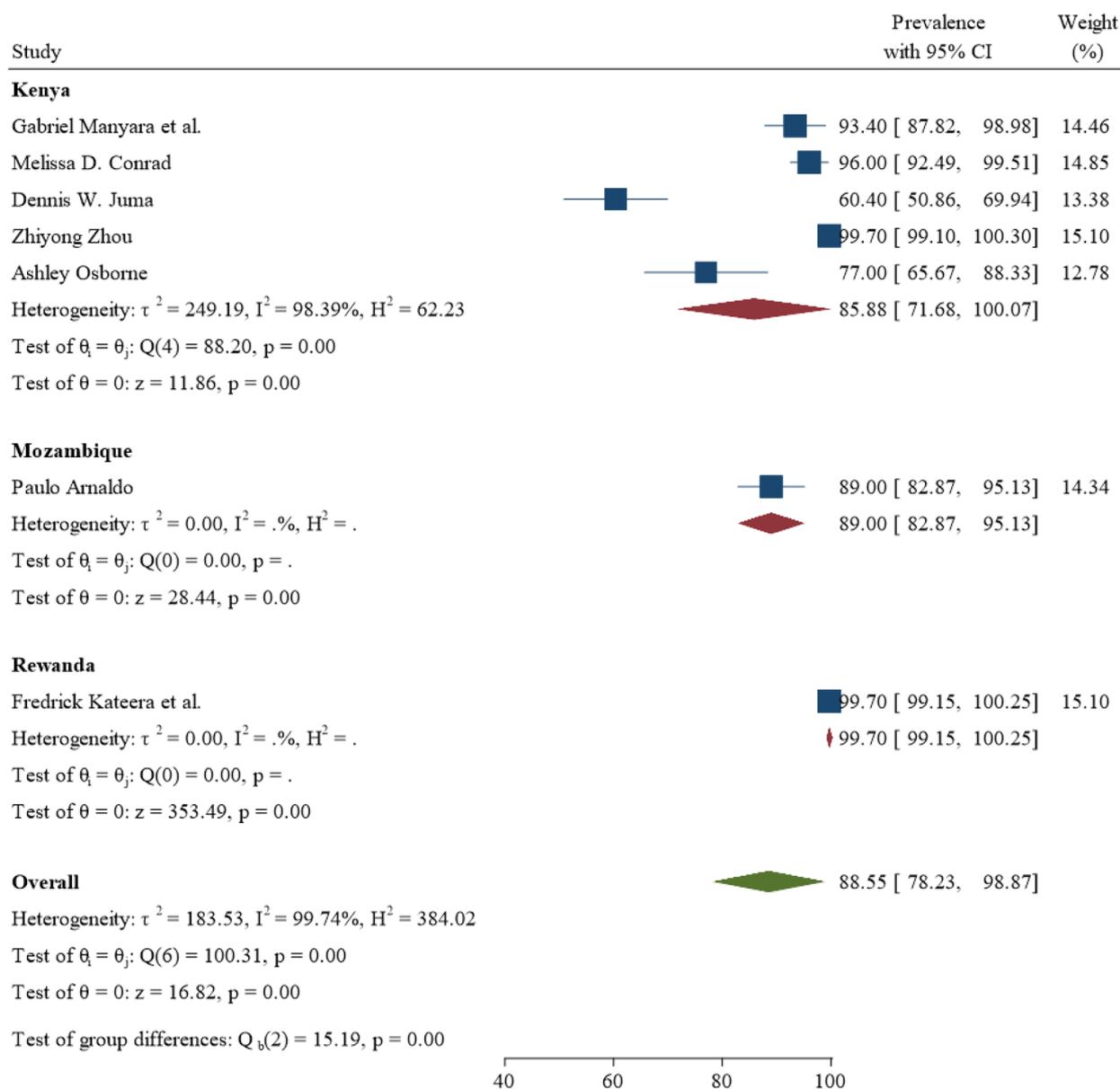
Fig. 25 Subgroup analysis for pooled prevalence of dhps A437G mutations based on country

malaria chemotherapeutic treatments has undoubtedly contributed to the increase in drug pressure, which has promoted the propagation of parasite resistance to SP [60]. This finding was also consistent with that of reported in Nepal [49], Nigeria [52], and China [54].

Furthermore, in this review, the pooled prevalence of dhfr 108 N was 92.2%. This finding was similar to that reported in Cameroon [61], Sudan [62], and Ghana [50]. However, this finding was higher than that reported in Senegal [63]. This might be due to varying topographical

variations and malaria transmission settings. Furthermore, this implies that SP selection is still going on in our study settings.

Moreover, in this review, the pooled prevalence of dhfr 59R was 71.5%. This finding was comparable with that of reported in Senegal [51, 63] and Ghana [50]. This finding was higher than that of reported in Sudan [62]. This might be due to drug resistance markers dispersed as a result of human and vector population movement within the same nation or across other nations. However, this



Random-effects REML model

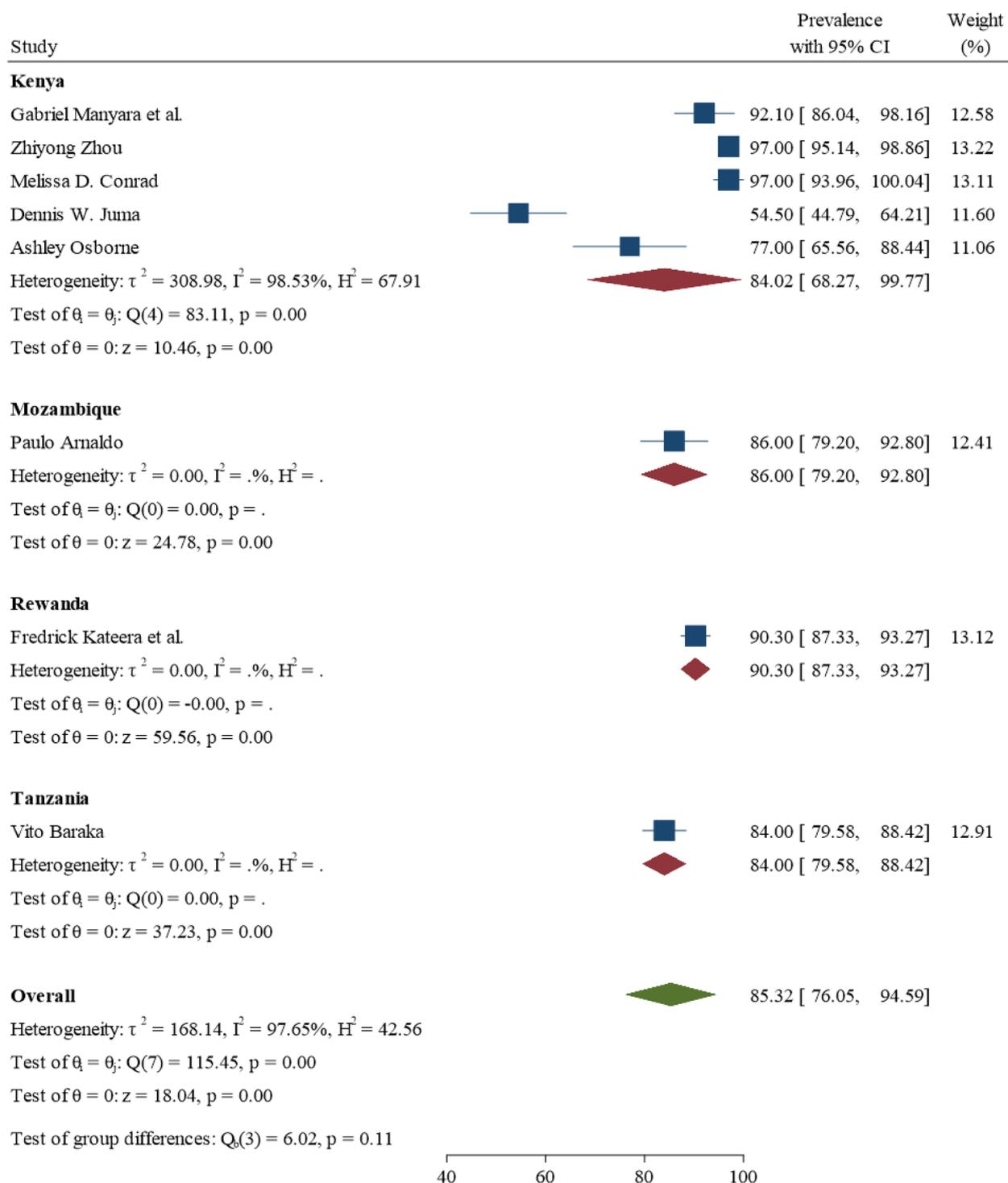
Fig. 26 Subgroup analysis for pooled prevalence of dhfr N511 mutations based on country

finding was lower than that reported in Cameroon [61] and Mali [57]. This discrepancy between studies on the role of the dhfr 59R mutation in SP could be attributed to different study designs like in vitro studies, cross-sectional studies over time at the population level, or clinical trials testing drug levels in patients.

Additionally, in this review, the pooled prevalence of dhfr I164L was 3.9%. This finding was inline with that of reported in India [64]. However, this finding was lower than that reported in Malaysia [65], China [66], and

Thailand [64]. Conversely, this finding was higher than that reported in Senegal [51] and Niger [67]. This might be due to the increased investment in road infrastructure throughout Sub-Saharan Africa, particularly in the Great Lakes region, the risk of the transmission of highly resistant mutations is larger than ever before [68].

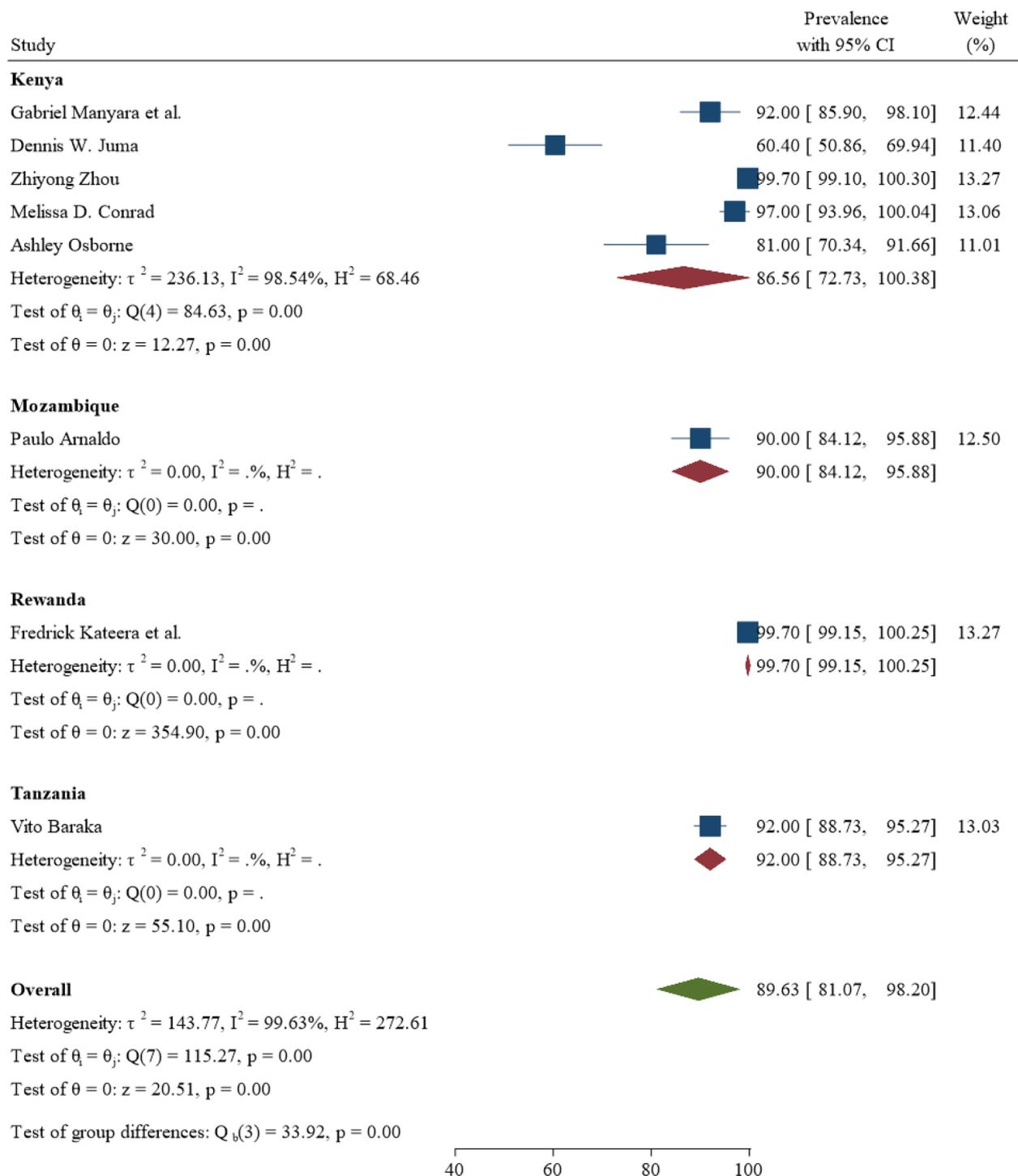
Also, in this systematic review and meta-analysis, the pooled prevalence of dhps A637G was 90.2%, which is higher than that reported in India [55, 56], Nepal [49], Senegal [51], Mali [57], Cameroon [69], Nigeria [52], and



Random-effects REML model
Fig. 27 Subgroup analysis for pooled prevalence of dhfr C59R mutations based on country

Sierra Leone [70]. This could be a result of individuals moving about, which can help drug-resistant parasites spread from one area to another and contribute to the

high occurrence of resistant strains like dhps A637G. It is well known that in Africa, the A437G mutation is highly linked to sulfadoxine resistance and a greater likelihood

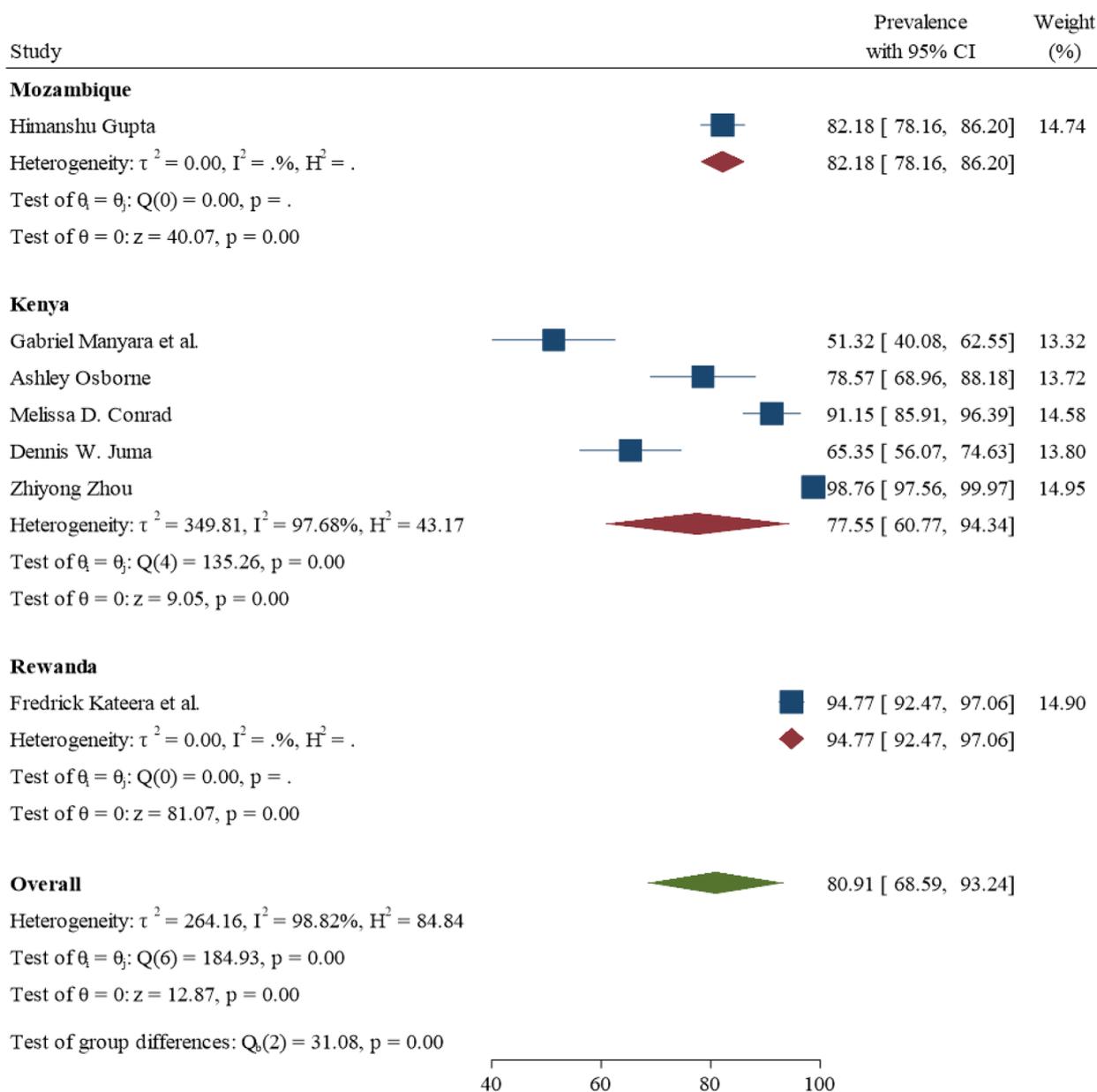


Random-effects REML model

Fig. 28 Subgroup analysis for pooled prevalence of dhfr S108 N mutations based on country

of failing SP treatment [52]. But this finding was inline with that of reported in central African countries [53], Congo [71], and China [54].

Similarly, in this systematic review and meta-analysis, the pooled prevalence of dhps K540E was 80.9%, which is higher than that reported in central African countries [53], Mali [57], Cameroon [69], Nigeria [52], and India



Random-effects REML model

Fig. 29 Subgroup analysis for pooled prevalence of dhps K540E mutations based on country

[55]. This might be because there are not many affordable or readily available alternatives to effective antimalarial drugs, which keeps people depending on antifolate drugs and encourages the selection of parasites that are resistant to treatments. This may also be due to malpractice in drug use, such as the use of the wrong dosage and insufficient information provided to patients about the prescribed treatment, which may lead to an increase in resistance and recurring infection rates. But

this finding was inline with that of reported in in Nepal [49].

Likewise, in this review, the pooled prevalence of dhps 540E was 91.5%. This finding was higher than that reported in Cameroon [61] and Ghana [50]. However, this finding was lower than that reported in Sudan [62]. This could be the result of variations in sample sizes, patient status differences, and geographic differences. Furthermore, this suggests that either the parasites are subject to varying drug pressure among nations,

or the free movement of individuals between various nations for work and other reasons is the cause of the parasites' varied drug resistance profiles. Furthermore, the WHO still advises SP for the intermittent preventive treatment of pregnant women and their unborn children, although this recommendation has been discontinued in populations where 50% or more of the parasites have the dhps540E allele [3, 72]. The Pfdhps 540E has a high prevalence in East Africa [73, 74].

Moreover, in this review, double, triple, quadruple, quintuple, and sextuple mutants were reported in 44%, 50%, 25%, 25%, and 6% of studies, respectively. Correspondingly, these double, triple, quadruple, and quintuple mutants were reported in China [54], Myanmar [75], India [55, 76], and South America [77]. The main factor contributing to the rising frequency of double, triple, quintuple, quadruple, and sextuple mutations in *P. falciparum* is the pressure from drugs. Gene mutations cause resistance mechanisms in parasites when they are frequently exposed to antimalarial drugs. The drug's target site and capacity to enter the parasite or metabolic pathway could all be affected by these mutations. The presence of double, triple, quadruple, quintuple, and sextuple mutations in *P. falciparum* can lead to increased drug resistance. The frequency of these mutations varies geographically; double and triple mutations are more common in some regions, while quintuple and quadruple mutations are more common in others. Mutant genotype combinations are mostly associated with increasing resistance from double to quintuple mutations [73]. The spread of drug-resistant malaria is a serious public health concern because it can result in treatment failure and increased mortality [78, 79]. Furthermore, this systematic review and meta-analysis showed that a significant difference in the prevalence of molecular markers like dhfr N51I, dhfr S108 N, dhps A437G, & dhps K540E among studies on a country level. This shows that the distribution of these markers may vary spatially, with implications for understanding disease risk and creating targeted therapeutics.

Strengths and limitations of the study

The major strength of the present review is that it has presented a picture of the prevalence and distribution of SP resistance markers of *P. falciparum* in East Africa with a total of 16 studies included. However, the data derived from this study did not include the pooled prevalence of dhfr and dhps gene mutations. Since there was no stated prevalence of dhfr and dhps gene mutations in the included studies.

Conclusions

The findings of this systematic review and meta-analysis regarding the markers of SP in East Africa revealed a significant prevalence of *P. falciparum* antimalarial drug resistance markers of SP. This indicates a substantial challenge in managing malaria infection caused by *P. falciparum*. The identified increase in the prevalence of antimalarial drug resistance markers of *P. falciparum* in SP leads to the widespread and quick emergence of drug resistance. This highlights the essential need for ongoing surveillance and research to create new antimalarial drugs and ways to overcome drug resistance. Also, different measures must be taken to prevent drug resistance with the remaining potent compounds as well as any new compounds that may be developed in the future. In addition, regular monitoring, identification, and limiting of drug-resistant *P. falciparum* strains through in vivo efficacy tests, in vitro tests, combination therapy, molecular techniques, and appropriate policies must continue to ensure the effectiveness of malaria treatment.

Abbreviations

CI	Confidence interval
DHFR	Dihydrofolate reductase
DHPS	Dihydropteroate synthetase
PFDHFR	<i>Plasmodium Falciparum</i> Dihydrofolate Reductase
PFDHPS	<i>Plasmodium Falciparum</i> Dihydropteroate Synthetase
SP	Sulfadoxine-pyrimethamine
WHO	World Health Organization

Author contributions

WA led the systematic review and meta-analysis, overseeing the study's conceptualization, article selection, data extraction, statistical analysis, and manuscript preparation. WA, T.M, A.A, T.E, and Z.D played a pivotal role in searching for relevant articles, conducting data extraction, performing statistical analysis, and contributing to manuscript drafting. B.B.A, B.K, and A. AK were involved in statistical analysis consultation of the overall process of this systematic review and meta-analysis. G.K, M.N, A.J, Z.A, Y.G, E.G, M.G, A.S, M.A.R, S.T, S.G, W.K, M.A, and S.A involved in data mining, data extraction, in statistical analysis, manuscript writing, editing, and ensuring accuracy and completeness. Additionally, all authors actively engaged in critically reviewing the study's progress, data analysis, and manuscript preparation, involved in the approval of the final manuscript for submission, thereby affirming their endorsement of its content and findings.

Funding

This systematic review and meta-analysis was not funded by any organization or individual.

Availability of data and materials

No datasets were generated or analysed during the current study.

Declarations

Ethics approval and consent to participate

Not applicable.

Consent for publication

Not applicable.

Competing interests

The authors declare no competing interests.

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Received: 30 August 2024 Accepted: 5 April 2025

Published online: 16 April 2025

References

- Olowe O, et al. Malaria in Africa and the historical perspective: the journey so far. *J Biol Med Sci*. 2015;3:333–41.
- Snow RW, Korenromp EL, Gouws E, Pediatric mortality in Africa: Plasmodium falciparum malaria as a cause or risk? The Intolerable Burden of Malaria II: What's New, What's Needed: Supplement to Volume 71 (2) of the American Journal of Tropical Medicine and Hygiene, 2004.
- WHO, WHO Policy recommendation on Intermittent Preventive Treatment during infancy with sulphadoxine-pyrimethamine (SP-IPTi) for Plasmodium falciparum malaria control in Africa, March 2010. 2010, World Health Organisation.
- Mace KE, Malaria surveillance—United States, 2017. *MMWR. Surveillance Summaries*. 2021. **70**.
- Organization WH. World malaria report 2022. Geneva: World Health Organization; 2022.
- Snow RW. Global malaria eradication and the importance of Plasmodium falciparum epidemiology in Africa. *BMC Med*. 2015;13:1–3.
- Organization WH. World malaria report 2019 (World Health Organization). Geneva: World Health Organization; 2018.
- Cravo P, et al. Mechanisms of drug resistance in malaria: current and new challenges. *Anti-Infect Agents Med Chem*. 2006;5(1):63–73.
- Menard D, Dondorp A. Antimalarial drug resistance: a threat to malaria elimination. *Cold Spring Harb Perspect Med*. 2017;7(7): a025619.
- Sinha S, Medhi B, Sehgal R. Challenges of drug-resistant malaria. *Parasite*. 2014;21:61.
- Eyasu M. Antimalarial drug resistance: in the past, current status and future perspectives. *Br J Pharmacol Toxicol*. 2015;6(1):1–15.
- Maguire JD, et al. Chloroquine-resistant Plasmodium malariae in south Sumatra, Indonesia. *Lancet*. 2002;360(9326):58–60.
- Bhatt S, et al. The effect of malaria control on Plasmodium falciparum in Africa between 2000 and 2015. *Nature*. 2015;526(7572):207–11.
- Trape J-F, The public health impact of chloroquine resistance in Africa. The Intolerable Burden of Malaria: A New Look at the Numbers: Supplement to Volume 64 (1) of the American Journal of Tropical Medicine and Hygiene, 2001.
- van der Pluijm RW, et al. Determinants of dihydroartemisinin-piperazine treatment failure in Plasmodium falciparum malaria in Cambodia, Thailand, and Vietnam: a prospective clinical, pharmacological, and genetic study. *Lancet Infect Dis*. 2019;19(9):952–61.
- Organization WH. World malaria day 2017: malaria prevention works, let's close the gap. Geneva: World Health Organization; 2017.
- Triglia T, et al. Mutations in dihydropteroate synthase are responsible for sulfone and sulfonamide resistance in Plasmodium falciparum. *Proc Natl Acad Sci*. 1997;94(25):13944–9.
- Dunyo S, et al. Randomised trial of chloroquine/sulphadoxine-pyrimethamine in Gambian children with malaria: impact against multidrug-resistant P falciparum. *PLoS Clin Trials*. 2006;1(3):e14.
- Staedke SG, et al. Relationship between age, molecular markers, and response to sulphadoxine-pyrimethamine treatment in Kampala. *Uganda Trop Med Int Health*. 2004;9(5):624–9.
- Mubjer RA, et al. Molecular markers of anti-malarial drug resistance in Lahj Governorate, Yemen: baseline data and implications. *Malar J*. 2011;10:1–9.
- Nsanzabana C, et al. Tools for surveillance of anti-malarial drug resistance: an assessment of the current landscape. *Malar J*. 2018;17:1–16.
- Diakité SA, et al. A comprehensive analysis of drug resistance molecular markers and Plasmodium falciparum genetic diversity in two malaria endemic sites in Mali. *Malar J*. 2019;18:1–9.
- Chenet SM, et al. Molecular profile of malaria drug resistance markers of Plasmodium falciparum in suriname. *Antimicrob Agents Chemother*. 2017. <https://doi.org/10.1128/aac.02655-16>.
- Heuchert A, et al. Molecular markers of anti-malarial drug resistance in southwest Ethiopia over time: regional surveillance from 2006 to 2013. *Malar J*. 2015;14:1–7.
- Arya A, et al. Artemisinin-based combination therapy (ACT) and drug resistance molecular markers: a systematic review of clinical studies from two malaria endemic regions—India and sub-Saharan Africa. *Int J Parasitol Drugs Drug Resist*. 2021;15:43–56.
- Huang F, et al. Molecular epidemiology of drug resistance markers of Plasmodium falciparum in Yunnan province. *China Malaria Journal*. 2012;11:1–7.
- Rasmussen C, Alonso P, Ringwald P. Current and emerging strategies to combat antimalarial resistance. *Expert Rev Anti Infect Ther*. 2022;20(3):353–72.
- Zhou Z, et al. Temporal trends in molecular markers of drug resistance in Plasmodium falciparum in human blood and profiles of corresponding resistant markers in mosquito oocysts in Asembo, western Kenya. *Malar J*. 2022;21(1):265.
- Rocamora F, Winzeler EA. Genomic approaches to drug resistance in malaria. *Annu Rev Microbiol*. 2020;74:761–86.
- Munn Z, et al. Methodological guidance for systematic reviews of observational epidemiological studies reporting prevalence and cumulative incidence data. *JBI Evid Implementation*. 2015;13(3):147–53.
- Higgins JP, Thompson SG. Quantifying heterogeneity in a meta-analysis. *Stat Med*. 2002;21(11):1539–58.
- Page MJ, et al. The PRISMA 2020 statement: an updated guideline for reporting systematic reviews. *BMJ*. 2021;372:14.
- Kishoyian G, et al. Molecular Monitoring of Plasmodium falciparum Resistance to Sulfadoxine-pyrimethamine in Western Kenya, 14 Years after its Withdrawal. *Ann Med Health Sci Res*. 2019;9(4):2.
- Gupta H, et al. Drug-resistant polymorphisms and copy numbers in Plasmodium falciparum, Mozambique, 2015. *Emerg Infect Dis*. 2018;24(1):40.
- Arnaldo P, et al., High-level Plasmodium falciparum sulfadoxine-pyrimethamine resistance with the emergence of sextuple mutant parasites during pregnancy in Mozambique. Evaluation of intermittent preventive treatment during pregnancy (IPTp) in Chókówè district, Southern Mozambique: coverage and effect on pregnancy and parasitological outcomes, 2019: p. 137.
- Kateera F, et al., Molecular surveillance of chloroquine and sulphadoxine Pyrimethamine resistance markers reveals partial recovery of Chloroquine Susceptibility but sustained intense levels of Sulfadoxine-Pyrimethamine mutations at two sites of different malaria transmission intensities in Rwanda. Determinants of malaria control in a rural community in Eastern Rwanda: p. 50.
- Juma DW, et al. The prevalence and antifolate drug resistance profiles of Plasmodium falciparum in study participants randomized to discontinue or continue cotrimoxazole prophylaxis. *PLoS Negl Trop Dis*. 2019;13(3): e0007223.
- Lo E, et al. Transmission dynamics of co-endemic Plasmodium vivax and P falciparum in Ethiopia and prevalence of antimalarial resistant genotypes. *PLoS Negl Trop Dis*. 2017;11(7):e0005806.
- Conrad MD, et al. Impact of intermittent preventive treatment during pregnancy on Plasmodium falciparum drug resistance—mediating polymorphisms in Uganda. *J Infect Dis*. 2017;216(8):1008–17.
- Mbonye AK, et al. Prevalence of Plasmodium falciparum resistance markers to sulfadoxine-pyrimethamine among pregnant women receiving intermittent preventive treatment for malaria in Uganda. *Antimicrob Agents Chemother*. 2015;59(9):5475–82.
- Shah M, et al. Assessment of molecular markers for anti-malarial drug resistance after the introduction and scale-up of malaria control interventions in western Kenya. *Malar J*. 2015;14:1–14.
- Siame MN, et al. High prevalence of dhfr and dhps molecular markers in Plasmodium falciparum in pregnant women of Nchelenge district, Northern Zambia. *Malar J*. 2015;14:1–6.

43. Baraka V, et al. High-level *Plasmodium falciparum* sulfadoxine-pyrimethamine resistance with the concomitant occurrence of septuple haplotype in Tanzania. *Malar J*. 2015;14:1–9.
44. Kavishe RA, et al. Molecular monitoring of *Plasmodium falciparum* super-resistance to sulfadoxine-pyrimethamine in Tanzania. *Malar J*. 2016;15:1–8.
45. Gikunju SW, et al. Prevalence of pfdhfr and pfdhps mutations in *Plasmodium falciparum* associated with drug resistance among pregnant women receiving IPTp-SP at Msambweni county referral hospital, Kwale county, Kenya. *Malar J*. 2020;19:1–7.
46. Molina-De la Fuente I et al., Seasonal malaria chemoprevention in a context of high presumed sulfadoxine-pyrimethamine resistance: malaria morbidity and molecular drug resistance profiles in South Sudan. *Malar J*. 2023; 22(1): 345.
47. Osborne A, et al. Drug resistance profiling of asymptomatic and low-density *Plasmodium falciparum* malaria infections on Ngodhe island, Kenya, using custom dual-indexing next-generation sequencing. *Sci Rep*. 2023;13(1):11416.
48. Kateera F, et al. Malaria case clinical profiles and *Plasmodium falciparum* parasite genetic diversity: a cross sectional survey at two sites of different malaria transmission intensities in Rwanda. *Malar J*. 2016;15:1–10.
49. Ranjitkar S, et al. Prevalence of molecular markers of anti-malarial drug resistance in *Plasmodium vivax* and *Plasmodium falciparum* in two districts of Nepal. *Malar J*. 2011;10:1–8.
50. Abugri J, et al. Prevalence of chloroquine and antifolate drug resistance alleles in *Plasmodium falciparum* clinical isolates from three areas in Ghana. *AAS Open Res*. 2018;1:1.
51. Wurtz N, et al. Prevalence of molecular markers of *Plasmodium falciparum* drug resistance in Dakar, Senegal. *Malar J*. 2012;11:1–10.
52. Quan H, et al. High multiple mutations of *Plasmodium falciparum*-resistant genotypes to sulphadoxine-pyrimethamine in Lagos, Nigeria. *Infect Dis Pov*. 2020;9:1–11.
53. Wang X, et al. Molecular determinants of sulfadoxine-pyrimethamine resistance in *Plasmodium falciparum* isolates from Central Africa between 2016 and 2021: wide geographic spread of highly mutated Pfdhfr and Pfdhps alleles. *Microbiol Spectr*. 2022;10(5):e02005-e2022.
54. Zhao D, et al. Surveillance of antimalarial drug-resistance genes in imported *Plasmodium falciparum* isolates from Nigeria in Henan, China, 2012–2019. *Front Cell Infect Microbiol*. 2021;11: 644576.
55. Sinha A, et al. Meta-analysis on *Plasmodium falciparum* sulfadoxine-pyrimethamine resistance-conferring mutations in India identifies hot spots for genetic surveillance. *Int J Antimicrob Agents*. 2024;63(3): 107071.
56. Rana R, et al. Molecular surveillance of anti-malarial drug resistance genes in *Plasmodium falciparum* isolates in Odisha, India. *Malar J*. 2022;21(1):394.
57. Mahamar A, et al. Effect of three years' seasonal malaria chemoprevention on molecular markers of resistance of *Plasmodium falciparum* to sulfadoxine-pyrimethamine and amodiaquine in Ouelessebouougou, Mali. *Malar J*. 2022;21(1):39.
58. Abuaku BK, Koram KA, Binka FN. Antimalarial drug use among caregivers in Ghana. *Afr Health Sci*. 2004;4(3):171–7.
59. Rogier E, et al. Nationwide monitoring for *Plasmodium falciparum* drug-resistance alleles to chloroquine, sulfadoxine, and pyrimethamine, Haiti, 2016–2017. *Emerg Infect Dis*. 2020;26(5):902.
60. Niba PTN, et al. Drug resistance markers within an evolving efficacy of anti-malarial drugs in Cameroon: a systematic review and meta-analysis (1998–2020). *Malar J*. 2021;20:1–20.
61. Tuedom AGB, et al. Antimalarial drug resistance in the Central and Adamawa regions of Cameroon: Prevalence of mutations in P. falciparum crt, Pfmdr1, Pfdhfr and Pfdhps genes. *PLoS ONE*. 2021;16(8):e0256343.
62. Mohamed NS, et al., Historical literature review and molecular analysis of malaria drug resistance markers of *Plasmodium falciparum* field-isolates from Sudan. 2020.
63. Fall B, et al. *Plasmodium falciparum* susceptibility to anti-malarial drugs in Dakar, Senegal, in 2010: an ex vivo and drug resistance molecular markers study. *Malar J*. 2013;12:1–11.
64. Biswas S, et al. Prevalence of point mutations in the dihydrofolate reductase and dihydropteroate synthetase genes of *Plasmodium falciparum* isolates from India and Thailand: a molecular epidemiologic study. *Tropical Med Int Health*. 2000;5(10):737–43.
65. Norahmad NA, et al. Prevalence of *Plasmodium falciparum* molecular markers of antimalarial drug resistance in a residual malaria focus area in Sabah, Malaysia. *PLoS ONE*. 2016;11(10): e0165515.
66. Wang S, et al. A review of malaria molecular markers for drug resistance in *Plasmodium falciparum* and *Plasmodium vivax* in China. *Front Cell Infect Microbiol*. 2023;13:1167220.
67. Issa I, et al. Prevalence of mutations in the Pfdhfr, Pfdhps, and Pfmdr1 genes of malarial parasites isolated from symptomatic patients in Dogondoutchi, Niger. *Trop Med Infect Dis*. 2022;7(8):155.
68. Signé L. African development, African transformation: how institutions shape development strategy. Cambridge: Cambridge University Press; 2018.
69. L'Episcopia M, et al. Targeted deep amplicon sequencing of antimalarial resistance markers in *Plasmodium falciparum* isolates from Cameroon. *Int J Infect Dis*. 2021;107:234–41.
70. Smith SJ, et al. Efficacy of artemisinin-based combination therapies and prevalence of molecular markers associated with artemisinin, piperaquine and sulfadoxine-pyrimethamine resistance in Sierra Leone. *Acta Trop*. 2018;185:363–70.
71. Ndounga M, et al. Therapeutic efficacy of sulfadoxine-pyrimethamine and the prevalence of molecular markers of resistance in under 5-year olds in Brazzaville. *Congo Trop Med Int Health*. 2007;12(10):1164–71.
72. Naidoo I, Roper C. Drug resistance maps to guide intermittent preventive treatment of malaria in African infants. *Parasitology*. 2011;138(12):1469–79.
73. Naidoo I, Roper C. Mapping 'partially resistant', 'fully resistant', and 'super resistant' malaria. *Trends Parasitol*. 2013;29(10):505–15.
74. Matondo SI, et al. High levels of sulphadoxine-pyrimethamine resistance Pfdhfr-Pfdhps quintuple mutations: a cross sectional survey of six regions in Tanzania. *Malar J*. 2014;13:1–7.
75. Zhao Y, et al. Genetic variations associated with drug resistance markers in asymptomatic *Plasmodium falciparum* infections in Myanmar. *Genes*. 2019;10(9):692.
76. Kojom Foko LP, et al. Nationwide spatiotemporal drug resistance genetic profiling from over three decades in Indian *Plasmodium falciparum* and *Plasmodium vivax* isolates. *Malar J*. 2023;22(1):236.
77. Chaturvedi R, et al. Geographical spread and structural basis of sulfadoxine-pyrimethamine drug-resistant malaria parasites. *Int J Parasitol*. 2021;51(7):505–25.
78. Heinberg A, Kirkman L. The molecular basis of antifolate resistance in *Plasmodium falciparum*: looking beyond point mutations. *Ann N Y Acad Sci*. 2015;1342(1):10–8.
79. Calçada C, et al. Expansion of a specific *Plasmodium falciparum* PfMDR1 haplotype in Southeast Asia with increased substrate transport. *MBio*. 2020. <https://doi.org/10.1128/mbio.02093-20>.

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