

Review Article

The Diagnostic Power of Circulating miR-1246 in Screening Cancer: An Updated Meta-analysis

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Background. MicroRNA-1246 (miR-1246), an oncomiR that regulates the expression of multiple cancer-related genes, has been attracted and studied as a promising indicator of various tumors. However, diverse conclusions on diagnostic accuracy have been shown due to the small sample size and limited studies included. This meta-analysis is aimed at systematically assessing the performance of extracellular circulating miR-1246 in screening common cancers. *Methods*. We searched the PubMed/MEDLINE, Web of Science, Cochrane Library, and Google Scholar databases for relevant studies until November 28, 2022. Then, the summary receiver operating characteristic (SROC) curves were drawn and calculated area under the curve (AUC), diagnostic odds ratio (DOR), sensitivity, and specificity values of circulating miR-1246 in the cancer surveillance. *Results*. After selection and quality assessment, 29 eligible studies with 5914 samples (3232 cases and 2682 controls) enrolled in the final analysis. The pooled AUC, DOR, sensitivity, and specificity of circulating miR-1246 in screening cancers were 0.885 (95% confidence interval (CI): 0.827-0.892), 27.7 (95% CI: 17.1-45.0), 84.2% (95% CI: 79.4-88.1), and 85.3% (95% CI: 80.5-89.2), respectively. Among cancer types, superior performance was noted for breast cancer (AUC = 0.950, DOR = 98.5) compared to colorectal cancer (AUC = 0.905, DOR = 47.6), esophageal squamous cell carcinoma (AUC = 0.757, DOR = 8.0), hepatocellular carcinoma (AUC = 0.872, DOR = 18.6), pancreatic cancer (AUC = 0.767, DOR = 12.3), and others (AUC = 0.887, DOR = 27.5, P = 0.007). No significant publication bias in DOR was observed in the meta-analysis (funnel plot asymmetry test with P = 0.652; skewness value = 0.672, P = 0.071). *Conclusion*. Extracellular circulating miR-1246 may serve as a reliable biomarker with good sensitivity and specificity in screening cancers, especially breast cancer.

1. Introduction

Despite improvements in diagnosis and treatment, cancer is still burdened disease globally with the increased new cases and deaths over the years [1, 2]. Annual screening and earlier detection are crucial strategies that help to reduce cancer incidence and mortality [3–7]. Moreover, early detection of cancers leads to the use of less-aggressive interventions that improve patients' quality of life. Many tools have been used frequently in the surveillance of cancers as low-dose computed tomography, mammography, endoscopy, ultrasound, and serum protein markers such as carbohydrate antigen 125, 15-3, 19-9, CYFRA 21-1, carcinoembryonic antigen, squamous cell carcinoma antigen, alpha-fetoprotein, and prostate-specific antigen. Nevertheless, just a few tests have been well-accepted due to their disadvantages of expensive, invasiveness, discomfort, poor sensitivity, specificity, and a certain false-positive and false-negative rate [3, 7–9].



FIGURE 1: Database searching and study selection.



FIGURE 2: Quality of included studies regarding the risk of bias and applicability.

In recent years, liquid biopsy materials, including micro-RNAs (miR-21, miR-155, miR-486, etc.) in the blood and body fluids, have been attracted and extensively studied as potential biomarkers for cancer diagnosis and prognosis [10]. These are endogenous small noncoding RNAs (19-22 nt) dysregulated in cancer cells. After production, they regulate the translation of target mRNAs or can be released into circulation, then communicate and affect distant cells and tissues, leading to condition changes of tumorigenesis, angiogenesis, invasion, migration, and metastasis [10]. Among microRNAs, miR-1246 plays as an oncogenic molecule that modulates the expression of multiple genes and pathways in various cancers [11]. Previous studies presented an elevated level of miR-1246 in the blood of cancer patients compared to healthy individuals exploring its diagnostic role [12]. However, divergent conclusions on diagnostic accuracy have been shown due to the small sample size and limited cancer types [12, 13]. We aim to systematically assess the performance of extracellular circulating miR-1246 in cancer screening on a larger sample.

2. Materials and Methods

This meta-analysis was conducted according to the guideline of Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) [14].

2.1. Database Searching and Selection of Study. We searched electronic databases of PubMed/MEDLINE, Web of Science, Cochrane Library, and Google Scholar for relevant studies up

Author	Year	Country	Case vs. control	Clinical stage	N, case/control	Sample	Technique	TP	FP 7		R	AUC	Ref.
Takeshita	2013	Japan	ESCC vs. HC	I-IV	101/46	Serum exosome	RT-qPCR	72	12	34	29 (0.754	[19]
Ogata-Kawata	2014	Japan	CRC vs. HC	I-IV	88/11	Serum exosome	Microarray	84	-	10	4	0.948	[20]
Fu	2016	China	BC vs. HC	I-IV	100/40	Serum	RT-qPCR	93	10	30	7	0.904	[21]
Armand-Labit	2016	France	Melanoma vs. HC	VI-III	28/16	Plasma	RT-qPCR	26	_	15	5	0.95^{ℓ}	[22]
Chai	2016	Hong Kong	HCC vs. HC	Na	61/24	Plasma	RT-qPCR	49	0	24	12 0	.982 [£]	[23]
Hannafon	2016	USA	BC vs. HC	0-III	16/16	Plasma exosome	RT-qPCR	6	1	15	~	0.69 [€]	[24]
Machida	2016	Japan	PT vs. HC	I-IV	12/13	Saliva exosome	RT-qPCR	8	0	13	4).814	[25]
Shimomura	2016	Japan	BC vs. HC	0-IV	1206/1397	Serum	Microarray	1065	92 13	305]	[4]	0.91	[26]
Xu	2017	USA	PC vs. HC	I-IIA	15/15	Plasma exosome	RT-qPCR	10	3	12	5	0.73^{\pounds}	[27]
Todeschini	2017	Italy	OC vs. HC	VI-III	168/65	Serum	RT-qPCR	146	15	50	22).893	[28]
Zhai	2018	China	BC vs. HC	Na	46/28	Plasma exosome	Au nanoflare probe	46	7	26	0	0.982	[29]
Bhagirath	2018^{\dagger}	USA	PCa vs. HC	IV	44/8	Serum exosome	RT-qPCR	33	0	8	11	0.926	[30]
Bhagirath	2018^{\ddagger}	USA	PCa vs. HC	IV	43/7	Serum exosome	RT-qPCR	38	0	7	5	0.933	[30]
Moshiri	2018	Italy	HCC vs. cirrhosis	Na	16/27	Plasma	ddPCR	14	4	23	2	0.97	[31]
Wang	2018	China	HCC vs. HC	I-IV	50/50	Serum exosome	RT-qPCR	30	9	44	20 (.825 [£]	[32]
Guo	2018	China	CRC vs. HC	0-IV	107/120	Serum	RT-qPCR	69	38	82	38	0.681	[33]
Shi	2020	China	GC vs. HC	I-IV	85/50	Serum exosome	RT-qPCR	70	, L	43	15	.911	[34]
Wei	2020	China	PC vs. benign+HC	I-IV	120/80	Serum	RT-qPCR	101	29	51	19	0.81	[35]
Ishige	2020	Japan	PC vs. HC	0-IV	41/30	Serum	RT-qPCR	38	~	22	3	0.87	[36]
Salah	2020	Egypt	CRC vs. HC	III-II	37/30	Serum	RT-qPCR	37	9	24	0	0.924	[37]
Huang	2020	China	NSCLC vs. HC	I	33/50	Serum	RT-qPCR	21	, L	43	12 0	.827 [€]	[38]
Hoshino	2020^{\dagger}	Japan	ESCC vs. HC	I-IV	55/39	Serum	RT-qPCR	40	12	27	15).816	[39]
Hoshino	2020^{\ddagger}	Japan	ESCC vs. HC	I-IV	101/34	Serum	RT-qPCR	72	10	24	29 (0.779	[39]
Ueta	2021	Japan	GBC vs. benign+HC	0-IV	50/69	Serum exosome	RT-qPCR	30	23	46	20).646	[40]
Chen	2021	China	BC vs. HC	Na	33/37	Plasma exosome	Molecular beacon	31	-	36	5	0.983	[41]
Zhang	2021	China	BC vs. HC	I-IV	21/9	Plasma exosome	Electrochemical biosensor	17	0	6	4 0	$.931^{\pounds}$	[42]
Chen	2021	China	HCC vs. HC	I-IV	50/50	Serum	RT-qPCR	41	10	40	6).865	[43]
Jang	2021^{\dagger}	Korea	BC vs. HC	0-IV	146/90	Serum	RT-qPCR	136	13	77	10	0.955	[44]
Jang	2021^{\ddagger}	Korea	BC vs. HC	0-IV	80/56	Plasma	RT-qPCR	77	8	48	3).963	[44]
Hoshino	$2021^{\$}$	Japan	ESCC vs. HC	I-IV	72/50	Urine	RT-qPCR	65	19	31	2	0.823	[45]
Hoshino	2021 ⁵	Japan	ESCC vs. HC	I-IV	72/50	Saliva	RT-qPCR	60	17	33	12	0.802	[45]
Rafiee	2022	Iran	CRC vs. HC	III-II	45/45	Serum	RT-qPCR	27	1	44	18	0.84^{\pounds}	[46]
Zhao	2022	China	MM vs. HC	III-II	90/30	Serum	RT-qPCR	78	1	29	12 (0.952	[47]
AUC: area under healthy control; H pancreatobiliary tr FN: false negative	the receive HCC: heps ract cancer ; Ref.: refe	er operating cha atocellular carcir ; ddPCR: drople rence. [†] Test set.	racteristic (ROC) curve; noma; MM: multiple m :t digital polymerase cha *Validation set. *Testir	BC: breast cancer yeloma; Na: not in reaction; RT-qI ig in urine specim	; CRC: colorectal ca available; NSCLC: r PCR: reverse transcr nens. ⁵ Testing in sal	ncer; ESCC: esophag non-small-cell lung c iptase quantitative pc iva specimens. ^E Sens	eal squamous cell carcinoma; (ancer; OC: ovarian cancer; P alymerase chain reaction; TP: t itivity and specificity values cc	3BC: gallb C: pancrea rue positiv rrespondi	ladder ca atic canc e; FP: fal ng to the	ancer; G :er; PCa lse posit e maxim	C: gastr : prosta ive; TN: um You	ic cancer; te cancer; true neg uden's J i	; PT: ative; index
were extracted fro	m the RO	C curve; then, tr	ue positive, false positiv.	e, true negative, ar	nd false negative nun	nbers were calculated							

Study	Events	Total	Sensitivity	Proportion	95%-CI
Takeshita N 2013	72	101	— — — ;	0.713	(0.614; 0.799)
Ogata-Kawata H 2014	84	88		0.955	(0.888; 0.987)
Fu L 2016	93	100	i — 🖬	0.930	(0.861; 0.971)
Armand-Labit V 2016	26	28		0.929	(0.765; 0.991)
Chai S 2016	49	61		0.803	(0.682; 0.894)
Hannafon BN 2016	9	16		0.562	(0.299; 0.802)
Machida T 2016	8	12		0.667	(0.349; 0.901)
Shimomura A 2016	1065	1206		0.883	(0.864; 0.901)
Xu YF 2017	10	15		0.667	(0.384: 0.882)
Todeschini P 2017	146	168		0.869	(0.808; 0.916)
Zhai LY 2018	46	46		1.000	(0.923; 1.000)
Bhagirath D 2018†	33	44		0.750	(0.597; 0.868)
Bhagirath D 2018‡	38	43		0.884	(0.749; 0.961)
Moshiri F 2018	14	16		0.875	(0.617; 0.984)
Wang Y 2018	30	50		0.600	(0.452; 0.736)
Guo S 2018	69	107	—	0.645	(0.546; 0.735)
Shi Y 2020	70	85	— —	0.824	(0.726; 0.898)
Wei J 2020	101	120		0.842	(0.764; 0.902)
Ishige F 2020	38	41		0.927	(0.801; 0.985)
Salah M 2020	37	37		1.000	(0.905; 1.000)
Huang D 2020	21	33		0.636	(0.451; 0.796)
Hoshino I 2020†	40	55		0.727	(0.590; 0.839)
Hoshino I 2020‡	72	101	— <u>—</u> i	0.713	(0.614; 0.799)
Ueta E 2021	30	50		0.600	(0.452; 0.736)
Chen Y 2021	31	33		0.939	(0.798; 0.993)
Zhang Y 2021	17	21	_	0.810	(0.581; 0.946)
Chen S 2021	41	50	— <u>—</u>	0.820	(0.686; 0.914)
Jang JY 2021†	136	146		0.932	(0.878; 0.967)
Jang JY 2021‡	77	80		0.963	(0.894; 0.992)
Hoshino I 2021§	65	72		0.903	(0.810; 0.960)
Hoshino I 2021¶	60	72		0.833	(0.727; 0.911)
Rafiee R 2022	27	45	— — i	0.600	(0.443; 0.743)
Zhao G 2022	78	90	-	0.867	(0.779; 0.929)
Random effects model Heterogeneity: $I^2 = 82.8\%$, τ	^{.2} = 0.7292, p	3232 0 < 0.001	0.3 0.4 0.5 0.6 0.7 0.8 0.9 1	0.842	(0.794; 0.881)

(a)

Study	Events	Total	Specificity	Proportion	95%-CI
Takeshita N 2013	34	46		0.739	(0.589; 0.857)
Ogata-Kawata H 2014	10	11		0.909	(0.587; 0.998)
Fu L 2016	30	40		0.750	(0.558; 0.873)
Armand-Labit V 2016	15	16		0.938	(0.698; 0.998)
Chai S 2016	24	24		1.000	(0.858; 1.000)
Hannafon BN 2016	15	16		0.938	(0.698; 0.998)
Machida T 2016	13	13		1.000	(0.753; 1.000)
Shimomura A 2016	1305	1397	-	0.934	(0.920; 0.947)
Xu YF 2017	12	15		0.800	(0.519: 0.957)
Todeschini P 2017	50	65		0.769	(0.648; 0.865)
Zhai LY 2018	26	28		0.929	(0.765; 0.991)
Bhagirath D 2018†	8	8		1.000	(0.631; 1.000)
Bhagirath D 2018‡	7	7		1.000	(0.590; 1.000)
Moshiri F 2018	23	27	<u>+</u>	0.852	(0.663; 0.958)
Wang Y 2018	44	50		0.880	(0.757; 0.955)
Guo S 2018	82	120	— <u> </u>	0.683	(0.592; 0.765)
Shi Y 2020	43	50		0.860	(0.733; 0.942)
Wei J 2020	51	80		0.637	(0.522; 0.742)
Ishige F 2020	22	30		0.733	(0.541; 0.877)
Salah M 2020	24	30		0.800	(0.614; 0.923)
Huang D 2020	43	50	<mark></mark>	0.860	(0.733; 0.942)
Hoshino I 2020†	27	39		0.692	(0.524; 0.830)
Hoshino I 2020‡	24	34		0.706	(0.525; 0.849)
Ueta E 2021	46	69		0.667	(0.543; 0.776)
Chen Y 2021	36	37		0.973	(0.858; 0.999)
Zhang Y 2021	9	9		1.000	(0.664; 1.000)
Chen S 2021	40	50	—— <u>—</u> —	0.800	(0.663; 0.900)
Jang JY 2021†	77	90		0.856	(0.766; 0.921)
Jang JY 2021‡	48	56		0.857	(0.738; 0.936)
Hoshino I 2021§	31	50 ·	_	0.620	(0.472; 0.753)
Hoshino I 20219	33	50	_	0.660	(0.512; 0.788)
Rafiee R 2022	44	45	i	0.978	(0.882; 0.999)
Zhao G 2022	29	30		0.967	(0.828; 0.999)
Random effects model Heterogeneity: $I^2 = 84.5\%$, a	$r^2 = 0.6303, p$	2682 < 0.001	0.5 0.6 0.7 0.8 0.9 1	0.853	(0.805; 0.892)

(b)

FIGURE 3: Continued.

Ctor Jac	Experir	nental	Con	trol	Diama ati a dila mati a	DOD		147 - 1 - 1 - 4
Study	Events	Total	Events	Total	Diagnostic odds ratio	DOK	95%-CI	weight
Takeshita N 2013	72	84	29	63	= !	7.0	(3.2; 15.4)	4.0%
Ogata-Kawata H 2014	84	85	4	14	<u>+</u> ■	210.0	(21.3; 2068.1)	2.2%
Fu L 2016	93	103	7	37		39.9	(13.9; 113.9)	3.7%
Armand-Labit V 2016	26	27	2	17		195.0	(16.3; 2335.9)	2.1%
Chai S 2016	49	49	12	36		194.0	(11.0; 3415.0)	1.7%
Hannafon BN 2016	9	10	7	22		19.3	(2.0; 183.4)	2.3%
Machida T 2016	8	8	4	17	- = -	51.0	(2.4; 1071.7)	1.6%
Shimomura A 2016	1065	1157	141	1446	•	107.1	(81.4; 141.0)	4.4%
Xu YF 2017	10	13	5	17	- 	8.0	(1.5; 42.0)	2.9%
Todeschini P 2017	146	161	22	72	≠	22.1	(10.7; 45.9)	4.0%
Zhai LY 2018	46	48	0	26		985.8	(45.6; 21312.8)	1.6%
Bhagirath D 2018†	33	33	11	19	— = —	49.5	(2.6; 927.2)	1.7%
Bhagirath D 2018‡	38	38	5	12		105.0	(5.2; 2106.6)	1.6%
Moshiri F 2018	14	18	2	25		40.2	(6.5; 249.1)	2.7%
Wang Y 2018	30	36	20	64		11.0	(4.0; 30.6)	3.7%
Guo S 2018	69	107	38	120		3.9	(2.3; 6.8)	4.2%
Shi Y 2020	70	77	15	58	+	28.7	(10.8; 75.9)	3.8%
Wei J 2020	101	130	19	70		9.3	(4.8; 18.3)	4.1%
Ishige F 2020	38	46	3	25		34.8	(8.4; 145.1)	3.2%
Salah M 2020	37	43	0	24		282.7	(15.2; 5247.5)	1.7%
Huang D 2020	21	28	12	55	i	10.7	(3.7; 31.3)	3.7%
Hoshino I 2020†	40	52	15	42		6.0	(2.4; 14.8)	3.9%
Hoshino I 2020‡	72	82	29	53		6.0	(2.5; 14.0)	3.9%
Ueta E 2021	30	53	20	66		3.0	(1.4; 6.4)	4.0%
Chen Y 2021	31	32	2	38		558.0	(48.2; 6453.5)	2.1%
Zhang Y 2021	17	17	4	13		73.9	(3.6; 1524.1)	1.6%
Chen S 2021	41	51	9	49		18.2	(6.7; 49.6)	3.7%
Jang JY 2021†	136	149	10	87		80.6	(33.7; 192.4)	3.9%
Jang JY 2021‡	17	85	3	51		154.0	(38.9; 609.1)	3.3%
Hoshino I 2021§	65	84	12	38		15.2	(5.8; 39.8)	3.8%
Hoshino I 2021	60	//	12	45	= `_	9./	(4.1; 22.8)	3.9%
Rafiee K 2022	2/	28	18	62		66.0	(8.3; 523.0)	2.5%
Zhao G 2022	/8	/9	12	41		188.5	(23.5; 1515.0)	2.4%
Random effects model		3090		2824	↓ ↓	27.7	(17.1; 45.0)	100.0%
Heterogeneity: $I^2 = 88.4$	$1\%, \tau^2 = 1$	1.3701,	p < 0.00	1				
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FIGURE 3: Continued.



FIGURE 3: Forest plots of sensitivity (a), specificity (b), DOR (c), SROC curves (d, e), and Fagan's nomogram (f) of circulating miR-1246 in screening cancers.







Omitting Shimomura A 2016 Omitting Guo S 2018 Omitting Ueta E 2021 Omitting Hoshino I 2020‡ Omitting Takeshita N 2013 Omitting Hoshino I 2020† Omitting Wei J 2020 Omitting Jang JY 2021‡ Omitting Jang JY 2021† Omitting Hoshino I 2021 Omitting Chen Y 2021 Omitting Zhai LY 2018 Omitting Zhao G 2022 Omitting Wang Y 2018 Omitting Huang D 2020 Omitting Ogata-Kawata H 2014 Omitting Salah M 2020 Omitting Armand-Labit V 2016 Omitting Xu YF 2017 Omitting Chai S 2016 Omitting Hoshino I 2021 § Omitting Bhagirath D 2018‡ Omitting Rafiee R 2022 Omitting Chen S 2021 Omitting Fu L 2016 Omitting Zhang Y 2021 Omitting Todeschini P 2017 Omitting Moshiri F 2018 Omitting Machida T 2016 Omitting Bhagirath D 2018† Omitting Ishige F 2020 Omitting Hannafon BN 2016 Omitting Shi Y 2020



⁽b)

FIGURE 4: Continued.



FIGURE 4: Baujat plot (a) and Leave-One-Out meta-analysis (b) for detecting outliers and important predictors for heterogeneity in DOR (c).

to 28 November 2022. The keywords used in searching were "miR-1246," "miR1246," "miRNA-1246," "miRNA1246," "miRNA1246," "miRNA1246," "miRNA-1246," and "microRNA1246." Also, we reviewed citation reports of potential studies to find additional articles. After searching, all relevant studies were saved as an EndNote list. By removing duplicates (2772 records), 5690 remained for later evaluations (Figure 1). Subsequently, only 41 articles progressed to the detailed assessment step after screening titles and abstracts. Four reduplicated studies, seven with unavailable data, and one included patients on radiotherapy were excluded. Finally, 29 studies were included in this meta-analysis.

2.2. Quality Assessment and Data Extraction. The quality of included studies was assessed by three independent researchers using the QUADAS-2 (Quality Assessment of Diagnostic Accuracy Studies) tool regarding the risk of bias and applicability (Figure 2) [15]. For each signaling question, "yes," "no," or "unclear" are phrased answers corresponding to the "low," "high," or "unclear" risk of bias and applicability concerns. When all signaling questions of a domain are answered "yes," the risk of bias was judged low. If any answer "no" exists, the risk of bias if any "unclear" exist without the "no" answer. In case of no consensus on judgments, three evaluators discussed in detail and determined the final decision.

Data extracted from articles include author names and country, year of publication, cancer, and control type, sample type, sample size, techniques used in experiments, and the AUC value in diagnosis. Besides, the true-positive, false-positive, true-negative, and false-negative numbers were extracted directly from articles or calculated indirectly using sensitivity and specificity corresponding to the maximum Youden's *J* index extracted from the receiving operating characteristic curve.

2.3. Statistical Analysis. We used the random-effects model to estimate pooled DOR, sensitivity, specificity, positive likelihood ratios, and negative likelihood ratios of circulating miR-1246 in cancer screening. Also, we constructed SROC curves and calculated summary AUC values, then compared them between groups using the bootstrap test (B = 2000 resampling iterations). The heterogeneity of diagnostic test accuracy between studies was measured by Higgins and Thompson's I^2 -statistic, which is significant if $I^2 \ge 50\%$. Subsequently, the Leave-One-Out analysis was used to detect outlier studies, while metaregression was performed to explore heterogeneity sources. Moreover, we used the funnel plot asymmetry statistic and the skewness of the standardized deviates to assess publication bias. All data analyses were done with the guidance of Shim et al., Noma et al., and Harrer et al. [16-18], using R statistical software v.4.2.2 (R foundation, 1020 Vienna, Austria) and packages meta, mada, metafor, dmetar, dmetatools, and altmeta. P < 0.05was considered statistically significant.

3. Results

3.1. Study Characteristics. Among 29 included studies [19–47], seven studies demonstrated the diagnostic performance of circulating miR-1246 in breast cancer [21, 24, 26, 29, 41, 42, 44], while four studies showed data for colorectal cancer [20, 33, 37, 46], four others for hepatocellular carcinoma [23, 31, 32, 43], and three for esophageal squamous cell carcinoma or pancreatic cancer [19, 27, 35, 36, 39, 45] (Table 1). Twenty-six out of 29 studies included healthy

	JlN	J 1		Sensitiv	vity			opern	city			DOK		
able	Number of study	Number of case	Estimates, % (95% CI)	$I^{2}, \%$	P value*	value**	Estimates, % (95% CI)	$I^{2}, \%$	P value*	<i>P</i> value**	Estimates, % (95% CI)	$I^{2}, \%$	<i>P</i> value*	P value ^{**}
er type						0.106				<0.001				<0.001
	7	3321	91.8 (83.9-95.9)	70.6	0.001		90.4 (84.9-94.0)	73.5	<0.001		98.5 (72.2-134.2)	29.4	0.194	
C	4	483	89.5 (55.2-98.3)	86.3	<0.001		87.1 (67.4-95.7)	73.3	0.011		47.6 (5.6-401.3)	87.1	<0.001	
CC	Э	620	78.3 (70.2-84.6)	67.1	0.016		68.0 (61.6-73.9)	0.0	0.776		8.0(5.4 - 11.8)	0.0	0.588	
S	4	328	77.1 (65.5-85.6)	67.4	0.027		87.5 (78.1-93.3)	0.0	0.751		18.6 (9.7-35.5)	29.7	0.234	
	ŝ	301	84.7 (78.6-89.3)	61.9	0.072		68.0 (59.3-75.6)	0.2	0.367		12.3 (5.6-26.9)	30.2	0.239	
$hers^{\dagger}$	8	861	80.3 (72.6-86.3)	74.0	<0.001		89.5 (77.9-95.3)	46.4	0.061		27.5 (10.3-73.5)	75.9	<0.001	
rol type						0.292				<0.001				0.068
()	26	5552	84.8 (79.7-88.8)	82.9	<0.001		86.5 (81.6-90.3)	82.9	<0.001		31.5 (19.1-52.1)	87.2	<0.001	
nign	33	362	77.8 (61.5-88.5)	83.4	0.002		68.2 (60.9-74.6)	50.8	0.131		8.6 (2.3-31.7)	77.8	0.011	
ole type						0.479				0.049				0.008
asma	6	544	89.1 (77.9-94.9)	64.4	0.004		92.0 (85.8-95.7)	0.0	0.716		85.8 (30.1-244.0)	50.5	0.040	
um	19	5101	82.5 (76.4-87.3)	87.4	<0.001		82.7 (77.1-87.1)	88.4	<0.001		21.3 (12.2-37.3)	91.8	<0.001	
hers	2	269	85.2 (78.2-90.3)	56.0	0.103		76.8 (43.0-93.6)	0.0	0.917		12.5 (6.7-23.5)	0.0	0.520	
ole size						0.613				0.004				0.048
00	14	1039	85.7 (77.0-91.5)	65.3	<0.001		79.3 (73.0-84.5)	91.9	<0.001		18.8 (10.2-34.5)	93.6	<0.001	
00	15	4875	83.3 (77.4-87.9)	88.4	<0.001		91.9 (85.5-89.2)	23.3	0.185		50.3 (23.4-108.2)	59.2	0.001	
nique						0.009				<0.001				<0.001
'-qPCR	23	2995	81.7 (76.3-86.2)	79.2	<0.001		82.5 (77.1-86.9)	53.0	<0.001		19.7 (12.4-31.3)	75.9	<0.001	
hers	6	2919	92.6 (86.1-96.2)	16.2	0.310		93.4 (92.0-94.5)	0.0	0.598		109.5 (83.9-142.9)	4.3	0.389	
						<0.001				0.016				0675
ction						100.00				010.0				0.000
rect	21	5420	86.9 (82.2-90.5)	82.1	<0.001		82.2 (76.7-86.6)	87.9	<0.001		28.6 (16.1-50.8)	90.8	<0.001	
direct	8	494	71.2 (61.4-79.4)	55.7	0.027		92.6 (85.7-96.3)	0.0	0.642		23.1 (10.2-52.3)	37.6	0.129	

TABLE 2: Subgroup meta-analyses for sensitivity, specificity, and DOR.

Predictor		Coefficient	Standard error	P value
	CRC	-1.121	0.688	0.103
	ESCC	-1.825	0.587	0.002
Cancer type:	HCC	-0.797	0.671	0.235
	PC	-0.632	0.788	0.423
	Others	-0.543	0.601	0.366
Control type:	HC	1.523	0.629	0.015
Technique:	RT-qPCR	-1.528	0.612	0.012

TABLE 3: Meta-regression analysis for the potential sources of heterogeneity in DOR.

CRC: colorectal cancer; DOR: diagnostic odds ratio; ESCC: esophageal squamous cell carcinoma; HC: healthy control; HCC: hepatocellular carcinoma; PC: pancreatic cancer; RT-qPCR: reverse transcriptase quantitative polymerase chain reaction.

individuals as the control group, which did not avoid a casecontrol design and thus might introduce biases according to the QUADAS-2 revised tool (Figure 2). Most studies detected miR-1246 in serum or plasma samples using the reverse transcriptase quantitative polymerase chain reaction (RT-qPCR) method. The total samples included in the meta-analysis were 5914, including 3232 cases and 2682 controls.

3.2. Performance of Circulating miR-1246 in Screening Cancers. The analyzed results indicated that circulating miR-1246 can differentiate cancers with 84.2% sensitivity (95% CI: 79.4-88.1) and 85.3% specificity (95% CI: 80.5-89.2, Figures 3(a) and 3(b)). Besides, the diagnostic odds ratio pooled from 29 studies was 27.7 (95% CI: 17.1-45.0, Figure 3(c)). However, heterogeneity in these analyses was substantial (I^2 were 82.8%, 84.5%, and 88.4%, P < 0.001, respectively). That is why we applied the random-effects model for the analyses.

The SROC curve of included studies shows an AUC of 0.885 (95% CI: 0.827-0.892, Figure 3(d)), suggesting that circulating miR-1246 has high diagnostic power. Remarkably, excellent performance was noted for breast cancer (AUC = 0.950, 95% CI: 0.872-0.958) compared to other types (P = 0.007, Figure 3(e)). With the assumed probability of suffering cancer of 55%, positive result increases the posttest possibility to 88%, while negative result drops that measure to 18% (Figure 3(f)). The positive and negative likelihood ratios were 6.35 and 0.18, respectively.

Because of significant heterogeneity, we performed the influence analysis and detected three outliers that contributed most to overall heterogeneity (Figures 4(a) and 4(b)). However, the heterogeneity remained high after removing these three outliers (DOR = 29.6, I^2 = 67.5%, 95% CI: 52.5-77.7%, P < 0.001). We performed subgroup analyses and observed that cancer type, control type, sample type, sample size, technique used, and data extraction method could contribute to the sensitivity, specificity, and DOR differences between studies (Table 2). The multimodel inference analysis showed that three predictors, including technique, control type, and cancer type, are the most important ones contributing to heterogeneity overall (Akaike's information criterion was the smallest value = 102.4, Figure 4(c)). We fitted these three predictors in a meta-regression and noted that this model could explain $R^2 = 62.8\%$ of the heterogeneity in DOR, and ESCC cancer type (coefficient = -1.825, P = 0.002), healthy control type

(coefficient = 1.523, P = 0.015), and RT-qPCR technique (coefficient = -1.528, P = 0.012) are independent sources (Table 3).

The funnel plot asymmetry test with linear regression indicated a nonsignificant publication bias in the metaanalysis (P = 0.652, Figure 5(a)). That is comparable with the analysis of skewness of the standardized deviates (skewness value = 0.672) (95% CI: -0.213 to 1.254, P = 0.071, Figure 5(b)), suggesting a low potential of publication bias [48].

4. Discussion

miR-1246 has been evidenced as an oncogene that regulates multiple genes (CCNG2, GSK3β, RORa, AXIN2, DYRK1A, *Caspase-9, FOXA2, PDGFR\beta, p53, NFIB, etc.*) and signaling pathways (RAF/MEK/ERK, Wnt/ β -catenin, NF- κ B, STAT3, THBS2/MMP, NOTCH2, etc.) related to the cell proliferation, angiogenesis, antiapoptosis, carcinogenesis, invasion, migration, metastasis, and therapy resistance [11]. Accordingly, recent studies indicated it as a potential biomarker for malignant tumors, but a small sample size resulted in the lack of consistent conclusions [12, 13]. The study of Wei (on 242 cases of colorectal cancer, pancreatic adenocarcinoma, and pancreatobiliary tract cancer from three original reports) exhibited an excellent efficiency of exosome miR-1246 (AUC = 0.969, 92% sensitivity, and 95.8% specificity) [12], whereas in analyses of Xie (conducted on seven individual studies, 975 cases from five cancer types including hepatocellular carcinoma, breast, colorectal, ovarian, and esophageal cancers), authors concluded that miR-1246 is a good indicator with moderate diagnostic accuracy (AUC = 0.83, 80% sensitivity, and 77% specificity) [13].

We conducted a systematic review and performed a metaanalysis on 29 individual studies from 9 countries, including 12 cancer types, over 5900 samples, and confirmed that extracellular circulating miR-1246 has good sensitivity, specificity, and robust performance in screening cancers (Figure 3(d)). Impressively, the diagnostic capacity of miR-1246 is excellent for breast cancer (Figure 3(e), Table 2). These results indicate a superior performance of circulating miR-1246 compared to the combined model of currently used tumor biomarkers [8]. In clinical practice, it is simple to integrate the miR-1246 test into the health examination program without additional



FIGURE 5: The potential of publication bias in DOR: linear regression test for funnel plot asymmetry (a) and skewness value based on the resampling method (b).

blood tubes, thanks to using a small sample volume. Also, it is quantified easily by the RT-qPCR, which is currently the widely used method with a fast turnaround time. Moreover, it is a lower cost and less invasive compared to low-dose computed tomography and endoscopy tests.

This study highlights the diagnostic power of extracellular circulating miR-1246 for cancers. However, most included studies comprise healthy individuals as the control group (Table 1), which is quite different from cancerous, which thus might affect the overall results. Therefore, further clinical trial studies with cancer/benign models and earlystage diseases should be done to confirm the diagnosis role of circulating miR-1246. Another limitation of this study is the existence of significant heterogeneity that requires a cautious use of analyzed results.

5. Conclusion

The results of this study indicated that extracellular circulating miR-1246 has good sensitivity, specificity, and robust performance, which might serve as a reliable biomarker in screening cancers, especially breast cancer.

Abbreviations

AUC:	Area under the receiver operating characteristic
	(ROC) curve
BC:	Breast cancer
CRC:	Colorectal cancer
ESCC:	Esophageal squamous cell carcinoma
GBC:	Gallbladder cancer
GC:	Gastric cancer
HC:	Healthy control
HCC:	Hepatocellular carcinoma
MM:	Multiple myeloma
Na:	Not available
NSCLC:	Non-small-cell lung cancer
OC:	Ovarian cancer
PC:	Pancreatic cancer
PCa:	Prostate cancer
PT:	Pancreatobiliary tract cancer
ddPCR:	Droplet digital polymerase chain reaction
RT-qPCR:	Reverse transcriptase quantitative polymerase
	chain reaction
TP:	True positive
FP:	False positive
TN:	True negative
FN:	False negative
Ref :	Reference.

Data Availability

All data generated or analyzed during this study are included in this published article.

Conflicts of Interest

The authors declared that no conflicts of interest exist.

Authors' Contributions

Study design, protocol writing, and statistical guidance were done by Son Truong Nguyen and Thuc Tri Nguyen. Database searching and reviews were done by Thang Thanh Phan, Toan Trong Ho, Suong Phuoc Pho, Hang Thuy Nguyen, and Binh Thanh Le. Quality assessment and data extraction were done by Thang Thanh Phan, Khanh Quang Huynh, and Anh Tuan Le. Data analysis was done Thang Thanh Phan, Khanh Quang Huynh, and Anh Tuan Le. Manuscript writing was done Thang Thanh Phan, Khanh Quang Huynh, and Anh Tuan Le. Manuscript revision was done by Son Truong Nguyen and Thuc Tri Nguyen. Khanh Quang Huynh and Son Truong Nguyen are responsible for the resources. All authors agreed on the final approval of the manuscript. Khanh Quang Huynh, Anh Tuan Le, and Thang Thanh Phan contributed equally to this work.

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