

MicroRNA-200 and microRNA-30 family as prognostic molecular signatures in ovarian cancer

A meta-analysis

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Abstract

Background: MicroRNAs (miRs) play a vital role in the occurrence, development, and progression of human cancers, but its role in the prognosis of ovarian cancer is unclear.

Methods: We performed a meta-analysis by searching *PubMed*, *Embase*, and *Web of Science* databases for eligible studies. The pooled hazard ratios (HRs) and 95% confidence intervals (CIs) were used to explore the association between miRs expression and overall survival (OS) and progression-free survival (PFS) on ovarian cancer patients. We also used Kaplan–Meier to analyze the relationship between miRs and OS in OncoLnc dataset.

Results: A total of 15 records were included into the meta-analysis. The expression level of miR-200 family showed significant association with OS (HR=0.78, 95% CI: 0.64–0.94) and insignificant association with PFS (HR=0.72, 95% CI: 0.50–1.03). Subgroup analysis revealed that an increased expression level of miR-200c was associated with better OS (HR=0.59, 95% CI: 0.45–0.74). An increased expression level of miR-200c, and miR-141 was associated with better PFS (miR-200a, HR=0.59, 95% CI: 0.42–0.75; miR-200c, HR=0.50, 95% CI: 0.14–0.87, miR-141, HR=0.38, 95% CI: 0.12–0.63). Similarly, higher expression of miR-30 family was associated with elevated OS/PFS for ovarian cancer (OS, HR=0.43, 95% CI: 0.13–0.74; PFS, HR=0.76, 95% CI: 0.64–0.87). The OncoLnc dataset presented that elevated expression level of miR-30d-5p was associated with better OS (n= 470, P=.0197).

Conclusion: The meta-analysis reveals that miR-200 family and miR-30 family could be promising prognostic biomarkers of ovarian cancer.

Abbreviations: ATF3 = activating transcription factor 3, CI = confidence interval, EMT = epithelial–mesenchymal transition, FFEP = formalin-fixed ethanol-preserved, HR = hazard ratio, ISH = in situ hybridization, miR = microRNA, OS = overall survival, PFS = progression-free survival, PRISMA = preferred reporting items for systematic reviews and meta-analyses reporting, qRT-PCR = quantitative real-time polymerase chain reaction, REMARK = recommendations for tumor marker prognostic studies, ROC = receiver-operating characteristic curve, TCGA = The Cancer Genome Atlas, TET = ten–eleven translocation, TGF- β 1 = transforming growth factor β 1, ZEB1 = zinc finger E-box-binding homeobox 1.

Keywords: biomarker, meta-analysis, microRNA, ovarian cancer, prognosis

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1. Introduction

Ovarian cancer represents one of the most common gynecological neoplasms in the world, with an estimated 238,700 new cases and 151,900 deaths in 2012.^[1] Although there are many new advances in the understanding of the molecular pathogenesis of ovarian cancer, there is still a lack of effective treatment. Therefore, there is an urgent need to offer more reliable prognostic biomarkers for effectively evaluating the outcomes of this disease and improving the treatment effect. Although a number of prognostic biomarkers have been exploited for ovarian cancer, yet reliably prognostic factors are still relative scanty.^[2] As miRs are frequently reported dysregulated in some cancer and often had important roles in the carcinogenesis of various cancers, they seem to be novel and attractive potential indicators for cancer. Some studies have demonstrated that dysregulated expression of miRs can be utilized as a prognostic marker to examine the disease outcome during the treatment of diseases.^[3–7] Therefore, effective prognostic markers for ovarian cancer are urgently needed.

MiRs (19-25 nucleotides) are a kind of short noncoding RNAs, which can regulate posttranscriptional expression of

target genes.^[8] MiR-200, a family of tumor suppressor miRs, consists of miR-141, miR-200a, miR-200b, miR-200c, and miR-429. Study showed that the loss of expression of the miR-200 family members may play an important role in the repression of E-cadherin by zinc finger E-box-binding homeobox 1 (ZEB1) and ZEB2 during epithelial-to-mesenchymal transition (EMT), thereby enhancing migration and invasion during cancer progression.^[9] In addition, some studies suggested that the miR-200 family might be serving as a prognostic marker for the treatment outcome in ovarian cancer.^[3,10,11] Similarly, the miR-30 family is evolutionarily conserved and consists of 5 members, miR-30a, miR-30b, miR-30c, miR-30d, and miR-30e.^[12] Some data had showed that the miR-30 family might be serving as a prognostic marker for the treatment outcome in ovarian cancer.^[13,14] However, their conclusions remain controversial. Therefore, we conducted a meta-analysis of studies that have identified a relationship between miR-200 and miR-30 family expression and survival in ovarian cancer.

In our present study, we performed a meta-analysis to provide a better understanding between the expression of miR-200 and miR-30 family and the prognosis in patients with ovarian cancer.

2. Materials and methods

2.1. Literature search strategy

A computerized literature was performed on *Pubmed*, *Embase*, and *Web of Science* databases for relevant studies that assessed the association between miRs and prognosis in ovarian cancer. All published articles were searched using the following keywords: (microRNA OR miRNA OR miR) AND (ovarian cancer) AND (prognosis OR prognostic OR survival OR outcome OR mortality). The searches were limited to articles published in English. Two investigators (MS and YLM) inspected the titles and abstracts of citations to identify relevant publications and obtained the full texts carefully. We also manually screened the reference lists of retrieved articles in order to identify other relevant studies.

2.2. Inclusion and exclusion criteria

Articles were considered eligible if they met all of the following initial inclusion criteria: focused on patients undergoing treatment for ovarian cancer; measured the expression of miRs in blood/tumor samples; had clearly defined the cutoff values; had clearly described detection methods for miRs; analyzed the correlation between OS/PFS and miRs expression; clearly described the follow-up time; clearly described the sample size; clearly described the study population. Articles were excluded following exclusion criteria: were conference records; had sample size <30 cases; calculated HRs based on the combination of multiple miRs; could not be calculated the HRs and 95% CI; or the survival data originated from TCGA dataset. Thereafter, articles that fulfilled all selection criteria were processed for data extraction. Two individual investigators (MS and YLM) independently assessed the eligibility of the retrieved articles. Disagreements were resolved by consensus and consultation with a third researcher (CZL).

2.3. Quality assessment

The quality of studies were assessed according to the following checklist based on the proposal by preferred reporting items for systematic reviews and meta-analyses (PRISMA)^[15] and

reporting recommendations for tumor marker prognostic studies (REMARK):^[16] had clearly described study population; had clearly described outcome assessment by representing it in OS or PFS; had clearly defined the measurement methods of miRs (quantitative real-time polymerase chain reaction [qRT-PCR] or in situ hybridization [ISH], etc.); had clear definition of cutoff values; measured the miRs expression level in blood/tumor samples; the follow-up time >60 months; and the sample size >30.

2.4. Data extraction

Data were extracted independently by 2 investigators (MS and YLM) who used a standard predefined sheet. The following data were extracted: title; the name of first author; publication year; study design; type of miRs; study population; number of participants; sample types; the measurement methods of miRs expression; cutoff values; follow-up time; HRs together with their 95% CIs and *P* values. If the HRs (95% CIs) and *P* values were not available in the original article, the data were calculated using the Kaplan–Meier curves and the methods illustrated by Parmar et al^[17] and Tierney et al.^[18] An observed HR >1 and *P* value \leq .05 indicated a worse outcome for the group with elevated miR expression. Conversely, an observed HR<1 and *P* value \leq .05 indicated a worse outcome for the group with decreased miR expression.^[19]*P* value >.05 indicated no significance.

2.5. The validation for the results of meta-analysis

We used Kaplan–Meier and log-rank methods to analyze the relationship between miR-200 family and miR-30 family and OS in OncoLnc dataset (http://www.oncolnc.org/).

2.6. Statistical analysis

OS was defined as the time interval between the date of primary surgery and the data of death from any cause.^[20] PFS was defined as the time interval between the start of the treatment and the first sign of appearance of relapse or disease progression.^[3] Metaanalysis was carried out using the Stata 12.0 software (StatCorp, College Station, TX). A test of heterogeneity was conducted using Higgins I^2 statistic and Cochran Q test. P value <.05 for Q test and I^2 value >50% indicated heterogeneity among studies.^[21] The random effect model was applied if the heterogeneity was observed, whereas the fixed-effect model was applied in the absence of between-study heterogeneity. Sensitivity analysis was used by excluding one study by turns and examining the influence of each single study on the combined of HRs. The publication bias was assessed by funnel plots and Egger bias indicator test.

2.7. Ethics statement

The Institutional Review Board of Shandong University (Jinan, China) approved this study.

3. Results

3.1. Selection of studies

A total of 1205 abstracts were found through literature search in *PubMed*, *Embase*, and *Web of Science* databases (Fig. 1). After excluding overlapping abstracts and irrelevant studies, 119 studies were identified as eligible for full-text review. When we checked the full-text, we found that miR-200 family, including miR-200a,^[10,20,22-24] miR-200b,^[3,10,22,25-27] miR-200c,^[3,10,22,25,27-30]



Figure 1. Flow diagram of the study selection procedure. OS=overall survival. PFS=progression-free survival. TCGA=The Cancer Genome Atlas.

miR-141,^[10,22,28,30] and miR-429,^[3,10,22] were the most frequently miRs with prognostic values in ovarian cancer patients. We also found that miR-30 family, including miR-30a,^[3,13,14,31] miR-30b,^[22,24] miR-30c,^[24] miR-30d,^[3,22,24,32] and miR-30e,^[13,14] ranked second. So, the association between miR-200 and miR-30 family and the prognosis were included in this meta-analysis.

3.2. Characteristics of the included studies

All of the included studies were published recently (2009–2016). Those 15 studies included 1417 patients with ovarian cancer, and sample size ranged 30 from 179 patients. MiRs expression was mainly detected in tissue samples; only 3 studies detected in blood. Table 1 shows the main characteristics of studies included into the meta-analysis. Table 2 shows the related data from the included studies. Studies reported a median of 16 (range: 15–19) items of 20 from the REMARK reporting guidelines (Supplementary Table1, http://links.lww.com/MD/C381).

3.3. Meta-analysis for miR-200 family

Eleven studies assessed the association between miR-200 family and survival outcome in ovarian cancer. For miR-200 family, significant interstudy heterogeneity was found (P < .05, $I^2 =$ 83.4%), and so we applied the random-effects model (HR = 0.78, 95% CI: 0.64–0.94) (Supplementary Figure 1, http://links.lww. com/MD/C381). Sensitivity analysis by omitting one study by turns showed there was no obvious influence of individual study on the pooled HRs (Supplementary Figure 2, http://links.lww. com/MD/C381). Stratified analysis by miR-200 family member types revealed that elevated expression level of miR-200c was subsequently significantly associated with better OS (HR = 0.59, 95% CI: 0.45–0.74) (Fig. 2). But there was no significant association with respect to miR-200a, miR-200b, miR-141, or miR-429. The Egger test showed no significant publication bias in these studies for miR-200a (P=.174), for miR-200b (P=.497), for miR-200c (P=.220), for miR-141 (P=.602), and the funnel plot was showed in Supplementary Figure 3A, 3B, 3C, 3D, http://links.lww.com/MD/C381. Only 2 studies showed the association bias was not evaluated.

When considering the sample types, the elevated expression of miR-200c was subsequently significantly associated with better OS in tissue (HR=0.44, 95% CI: 0.27–0.73) (Supplementary Figure 4A, http://links.lww.com/MD/C381). Although the upregulation of miR-200 family indicated a poorer OS in blood (HR=1.97, 95% CI: 1.42–2.73) (Supplementary Figure 4B, http://links.lww.com/MD/C381), subgroup analysis by study kinds revealed that there was no significantly association between the expression level of miR-200 family and OS (Supplementary Figure 5A and 5B, http://links.lww.com/MD/C381). When we divided miR-200 family into 2 clusters based on chromosomal location, Chr1 (miR-200a, miR-200b, and miR-429) and Chr12 (miR-141 and miR-200c), we found that heterogeneity still exists (Supplementary Figure 5C, http://links.lww.com/MD/C381).

Table 1

Main characteristics of the eligible studies.

Study (year)	MiRs	Population	Study design	Number	Assay	Cutoff value	Sample	Survival analysis	Source of HR	Adjusted	Follow-up, mo
Bagnoli (2016)	141	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
	200a	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
	200b	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
	200c	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
	429	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
	30b	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
	30d	Italy	RC	179	qRT-PCR	Median	Tissue	OS	Rep	Yes	120
Calura (2016)	200c	Italy	RC	203	qRT-PCR	ROC	Tissue	OS/PFS	Rep	Yes	204
	141	Italy	RC	203	qRT-PCR	ROC	Tissue	OS/PFS	Rep	Yes	204
Cao (2014)	200a	China	RC	100	qRT-PCR	Median	Tissue	OS	Rep	Yes	60
	200b	China	RC	100	qRT-PCR	Median	Tissue	OS	Rep	Yes	60
	200c	China	RC	100	qRT-PCR	Median	Tissue	OS	Rep	Yes	60
Elgaaen (2014)	200c	Norway	RC	35	qRT-PCR	Tertile	Tissue	OS/PFS	SC	No	120
Gao (2015)	141	China	RC	93	qRT-PCR	Mean	Tissue	OS	Rep	No	110
	200c	China	RC	93	qRT-PCR	Mean	Tissue	OS	Rep	No	110
Hu (2009)	200a	USA	RC	55	qRT-PCR	10.8	Tissue	OS/PFS	SC	No	60
Kapetanakis (2015)	200b	France	RC	51	qRT-PCR	0	Plasma	PFS	Rep	Yes	60
Lee (2012)	30a	Korea	RC	68/58	qRT-PCR	Median	FFPE	OS/PFS	Rep	Yes	122
	30c	Korea	RC	68/58	qRT-PCR	Median	FFPE	OS/PFS	Rep	Yes	122
	30d	Korea	RC	68/58	qRT-PCR	Median	FFPE	OS/PFS	Rep	Yes	122
	30d	Korea	RC	68/58	qRT-PCR	Median	FFPE	OS/PFS	Rep	Yes	122
Leskelä (2011)	200a	Spain	RC	72	ISH	EP	FFPE	OS/PFS	Rep	No	130
	200b	Spain	RC	72	ISH	EP	FFPE	PFS	Rep	No	130
	200c	Spain	RC	72	ISH	EP	FFPE	PFS	Rep	No	130
	141	Spain	RC	72	ISH	EP	FFPE	PFS	Rep	No	130
	429	Spain	RC	72	ISH	EP	FFPE	PFS	Rep	No	130
Marchini (2011)	200b	Italy	RC	89	qRT-PCR	Median	Tissue	OS/PFS	Rep	Yes	144
	30	Italy	RC	89	qRT-PCR	Median	Tissue	OS/PFS	Rep	Yes	144
	30a	Italy	RC	89	qRT-PCR	Median	Tissue	OS/PFS	Rep	Yes	144
	30d	Italy	RC	89	qRT-PCR	Median	Tissue	OS/PFS	Rep	Yes	144
	200c	Italy	RC	89	gRT-PCR	Median	Tissue	OS/PFS	Rep	Yes	144
Meng (2015)	429	Germany	RC	171	gRT-PCR	Median	Serum	OS	Rep	Yes	180
Meng (2016)	200a	Germany	RC	163	gRT-PCR	Median	Exosome	OS/PFS	Rep	Yes	136
	200b	Germany	RC	163	gRT-PCR	Median	Exosome	OS/PFS	Rep	Yes	136
	200c	Germany	RC	163	gRT-PCR	Median	Exosome	OS/PFS	Rep	Yes	136
Sestito (2015)	30a	Italy	RC	39	gRT-PCR	Median	Tissue	PFS	Rep	Yes	130
Wang (2013)	30a	China	RC	69	qRT-PCR	ROC	FFEP	OS	SC	No	70
	30e	China	RC	69	gRT-PCR	ROC	FFEP	OS	SC	No	70
Zhao (2013)	30a	China	RC	30	gRT-PCR	EP	FFEP	OS	SC	No	80
. ,	30e	China	RC	30	qRT-PCR	EP	FFEP	OS	SC	No	80
			-						-		

EP=expression level, FFEP=formalin-fixed ethanol-preserved, HR=hazard ratio, ISH=in situ hybridization, MiRs=microRNAs, Num=number, OS=overall survival, qRT-PCR=quantitative real-time polymerase chain reaction, RC=retrospective cohort, Rep=report, ROC=receiver-operating characteristic curve, SC=survival curve.

Table 2

		05			PFS		
Study (year)	miRs	HR (95% CI)	Р	Poor prognosis	HR (95%CI)	Р	Poor prognosis
Bagnoli (2016)	141	0.82 (0.73-0.92)	.001	Low			
o ()	200a	0.81 (0.71-0.93)	.003	Low			
	200b	0.79 (0.68-0.91)	.003	Low			
	200c	0.79 (0.69-0.91)	.002	Low			
	429	0.84 (0.73-0.96)	.012	Low			
	30b	1.98 (1.24-3.17)	.006	High			
	30d	1.25 (1.03-1.52)	.023	High			
Calura (2016)	200c	0.18 (0.07-0.49)	<.001	Low	0.24 (0.12-0.49)	<.001	Low
	141	0.33 (0.15-0.73)	.006	Low	0.35 (0.18-0.82)	.001	Low
Cao (2014)	200a	0.06 (0.03-0.74)	.010	Low			
5 <i>2</i>	200b	0.06 (0.03-0.88)	.010	Low			
	200c	0.06 (0.03-0.79)	.010	Low			
Elgaaen (2014)	200c	1.33 (1.04-1.70)	.023	High	2.24 (1.08-4.66)	.031	Hiah
Gao (2015)	141	1.81 (1.00-3.33)	.049	High	(/		5
GdU (2013)	200c	0.32 (0.17-0.60)	<.001	Low			
Hu (2009)	200a	0.22 (0.09-0.53)	<.001	low	0.53 (0.38-0.74)	<.001	Low
Kapetanakis (2015)	200b				2.95 (0.94-9.28)	.060	No significance
Lee (2012)	30a	0.48 (0.19-1.18)	.110	No significance	0.46 (0.21-1.00)	.050	Low
	300	0.37 (0.10-1.44)	.150	No significance	0.27 (0.08-0.92)	.040	Low
	30d	0.25 (0.07-0.82)	.020	low	0.33 (0.12-0.90)	.030	Low
	30e	0.48 (0.17-1.37)	170	No significance	0.39 (0.16-0.94)	040	Low
Leskelä (2011)	200c	0.48 (0.24-0.97)	.041	low	0.45 (0.20-1.00)	.050	Low
	200h				0.74 (0.34-1.62)	450	No significance
	200a				0.82(0.39-1.75)	612	No significance
	141				0.42 (0.18-1.02)	054	No significance
	429				0.48 (0.21 - 1.09)	079	No significance
Maschini (2011)	200h	2 05 (0 64-6 57)	226	No significance	2.34 (0.86-6.36)	.010	No significance
maconini (Eo i i)	30	2100 (0101 0101)	.220	no significance	0.70 (0.24-2.02)	502	No significance
	30a	0.50 (0.11-2.23)	364	No significance	0.93 (0.31-2.80)	891	No significance
	30d	0.64 (0.19-2.18)	476	No significance	0.60 (0.20-1.78)	352	No significance
	2000	0.24 (0.08-0.79)	018	Low	0.42 (0.16-1.20)	106	No significance
Mena (2015)	429	1 78 (1 13-2 81)	013	High	0.12 (0.10 1.20)	.100	No olgrinounoo
Meng (2016)	200a	1 70 (0.80-3.50)	220	No significance	1 10 (0 60-1 90)	870	No significance
	200a	2.80 (1.10-6.80)	020	High	1 60 (0.00 1.00)	1/0	No significance
	2005	2.50 (1.10-6.10)	038	High	1 70 (0.80–3.60)	170	No significance
Sestito (2015)	2000	2.00 (1.10 0.10)	.000	riigit	0.84 (0.72-0.98)	030	Low
Wang (2013)	30a	0.18 (0.06-0.53)	002	L OW	0.04 (0.72-0.50)	.050	LOW
wang (2013)	304	0.10 (0.00-0.33)	.002	Low			
7han (2013)	302	0.34 (0.13_0.87)	.004	Low			
21100 (2010)	30e	0.20 (0.05-0.85)	030	Low			
	000	0.20 (0.00 0.00)	.000	LOW			

 $Cl = confidence \ interval, \ HR = hazard \ ratio, \ miRs = microRNAs, \ OS = overall \ survival, \ PFS = progression-free \ survival.$

Study		%
ID	HR (95% CI)	Weight
141		
Calura (2016-141)	0.33 (0.15, 0.73)	5.97
Bagnoli (2016-141)	0.82 (0.73, 0.92)	7.56
Gao (2015-141)	1.81 (1.00, 3.33)	1.31
Subtotal (I-squared = 84.5%, p = 0.002)	0.74 (0.27, 1.22)	14.83
429		
Bagnoli (2016-429)	0.84 (0.73, 0.96)	7 44
Meng (2015-429)	- 178 (1 13 2 81)	2 18
Subtotal (I-squared = 78.8% n = 0.030)	1 21 (0 31 2 12)	9.62
Subiolar (Psquared = 70.0%, p = 0.050)	1.21 (0.01, 2.12)	5.02
200a		
cao (2014-200a)	0.06 (0.03, 0.74)	5.34
Hu (2009-200a) 🖝	0.22 (0.09, 0.53)	6.63
Bagnoli (2016-200a)	0.81 (0.71, 0.93)	7.47
Meng (2016-200a)	1.70 (0.80, 3.50)	1.02
Subtotal (I-squared = 91.6%, p = 0.000)	0.50 (0.02, 0.98)	20.46
2005		
cao (2014-200b)	0.06 (0.03, 0.88)	4 70
Bagnoli (2016-200b)	0.79 (0.68, 0.91)	7 44
Marchini (2011-2006)	2 05 (0.64, 6.57)	0.24
Aeng (2016-200b)	280 (1 10 6 80)	0.25
Subtotal (I-squared = 77.5% p = 0.004)	0.65(-0.03, 1.34)	12 63
	0.00 (0.00, 1.04)	12.00
200c		
cao (2014-200c)	0.06 (0.03, 0.79)	5.10
Calura (2016-200c)	0.18 (0.07, 0.49)	6.72
Marchini (2011-200c)	0.24 (0.08, 0.79)	5.34
Gao (2015-200c)	0.32 (0.17, 0.60)	6.67
eskela" (2011-200c)	0.48 (0.24, 0.97)	5.24
Bagnoli (2016-200c)	0.79 (0.69, 0.91)	7.47
Igaaen (2014-200c) +	1.33 (1.04, 1.70)	5.58
Meng (2016-200c)	2.50 (1.10, 6.10)	0.33
Subtotal (I-squared = 89.5%, p = 0.000)	0.52 (0.22, 0.82)	42.46
Overall (I-squared = 86.8%, p = 0.000)	0.59 (0.45, 0.74)	100.00
NOTE: Weights are from random effects analysis		
	I	
-6.8 0	6.8	

Figure 2. Forest plots of subgroup analysis regarding specific miR-200 family member expression and OS. OS=overall survival.

We also found that elevated expression level of miR-200 family had no significant association with better PFS (HR=0.70, 95% CI: 0.43–1.15) (Supplementary Figure 6, http://links.lww.com/ MD/C381). Stratified analysis by miR-200 family member types revealed that the upregulated expression of miR-200a, miR-200c, and miR-141 was subsequently significantly associated with better PFS (miR-200a, HR=0.59, 95% CI: 0.42–0.75; miR-200c, HR=0.50, 95% CI: 0.14–0.87; miR-141, HR=0.38, 95% CI: 0.12–0.63) (Figs. 3A and B).

3.4. Meta-analysis for miR-30 family

Six studies had showed that miR-30 family could be a predictor of treatment outcome for ovarian cancer. All of the studies used the tissue as sample. For miR-30 family, significant interstudy heterogeneity was found (P < .05, $I^2 = 83.7\%$), and the random-effects model was applied (HR = 0.43, 95% CI: 0.13–0.74) for OS (Fig. 4A). Stratified analysis by miR-30 family member types

revealed that elevated the expression levels of miR-30a and miR-30e were subsequently significantly associated with better OS (Fig. 4A). For PFS, no significant interstudy heterogeneity was found (P=.127, I^2 =41.8%), and the fixed-effect model revealed that miR-30 family expression was inversely related with patient's PFS (HR=0.76, 95% CI: 0.64–0.87) (Fig. 4B). Stratified analysis by miR-30 family member types revealed that elevated the expression level of miR-30a and miR-30d were subsequently significantly associated with better PFS (miR-30a, HR=0.80, 95% CI: 0.69–0.92; miR-30d, HR=0.38, 95% CI: 0.03–0.73) (Fig. 4B).

3.5. The validation for the results of meta-analysis

For miR-200 family and miR-30 family, we only found that the elevated expression of miR-30d-5p was associated with better OS (n=470, P=.0197) (Fig. 5). We used median of miR-30d-5p expression as the cutoff value.



Figure 3. Forest plots of subgroup analysis regarding specific miR-200 family member expression and PFS. (A) miR-141, miR-200a, and PFS. (B) miR-200b, miR-200c, and PFS. PFS = progression-free survival.

4. Discussion

It has been shown that miR-200 family were accumulated in ovarian cancer patients. The pooled analysis of studies demonstrated that, for miR-200 family, the improved OS existed only for enhanced expression of miR-200c, which accords with the results by Shi.^[11] MiR-200 family members have been reported to regulate EMT by targeting ZEB1 and ZEB2, resulting in dysregulation of the cell-cell adhesion

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Marchini (2011-30a) 0.93 (0.31, 2.80) 0.8 Sestito (2015-30a) 0.84 (0.72, 0.98) 78. Subtotal (I-squared = 9.1%, p = 0.348) 0.80 (0.68, 0.92) 89. 30d 0.33 (0.12, 0.90) 8.6 Lee (2012-30a) 0.60 (0.20, 1.78) 2.1 Subtotal (I-squared = 0.0%, p = 0.548) 0.38 (0.03, 0.73) 10. Heterogeneity between groups: p = 0.026 0.76 (0.64, 0.87) 100	larchini (2011-30a)	0.70 (0.24, 2.02)	1.67
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Heterogeneity between groups: p = 0.026 Overall (I-squared = 41.8%, p = 0.127) 0.76 (0.64, 0.87) 100	Subtotal (I-squared = 0.0%, p = 0.548)	0.38 (0.03, 0.73)	10.81
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		0.76 (0.64, 0.87)	100.00

Figure 4. Subgroup analysis regarding specific miR-30 family member expression and OS (A)/PFS (B). OS = overall survival. PFS = progression-free survival.

molecule E-cadherin.^[33,34] E-cadherin downregulation is clearly important in cancer progression, facilitating cell detachment and metastasis. Furthermore, miR-200 family members target EMT regulators, apparently being important in tumor progression.^[29] Studies also found that cells expressing miR-200c played an important role in restoring expression of E-cadherin and altering morphology from mesenchymal to epithelial.^[35] Similarly, the pooled analysis of studies demonstrated that, for miR-200 family, the improved PFS existed for elevated expression of miR-200a and miR-141. EPH receptor A2, one of miR-200a targets,



Figure 5. OS curves of 470 ovarian cancer patients for miR-30d-5p low expression (blue line) and high expression (red line). OS=overall survival.

promotes tumor growth and predicts poor prognosis for ovarian cancer patients. $^{\left[36\right] }$

The pooled analysis of studies demonstrated that higher expression of the miR-30 family significantly improved the OS and PFS in women with ovarian cancer. The subgroup analysis of miR-30 family members revealed that the improved OS existed for enhanced expression of miR-30a and miR-30e. The improved PFS existed for enhanced expression of all miR-30a and miR-30d. Studies showed that activating transcription factor 3 (ATF3), MYC proto-oncogene, and bHLH transcription factor were potential cotargets of miR-30 family, which present as regulators in the different pathways in numerous human cancers.^[14] The OncoLnc dataset showed that only elevated expression of miR-30d-5p associated with better OS, which suggesting the potentiality of miR-30d to be used as prognostic biomarker for ovarian cancer. The ten-eleven translocation (TET) family members are new DNA demethylation-related proteins. The study found that TET3 can block transforming growth factor $\beta 1$ (TGF- β 1) by demethylating the miR-30d precursor gene promoter.^[37] Studies also showed that miR-30d functioned as a suppressor of ovarian cancer progression by decreasing Snail expression and thus blocking TGF-B1-induced EMT process.^[38] The pooled studies can provide a reference for studying the mechanism of ovarian cancer and targeted therapy.

This meta-analysis had several limitations. First, significant heterogeneity existed among the studies. When we analyzed the heterogeneity origin from the study kinds and sample types, the high degree of heterogeneity still exists. Second, the number of studies available was limited. More studies should be conducted to assess these associations in further. Third, circulating markers are more acceptable than tissue markers. More studies may warrant further research to evaluate the prognostic value of miR level in serum.

In conclusion, we found that elevated expression of miR-200 and miR-30 family were indicators of a better treatment outcome in ovarian cancer patients. For further study, investigating the expression of miR-200 and miR-30 family in ovarian cancer may provide a new thinking into cancer prevention and therapeutic strategy.

Author contributions

Conceptualization: Min Shi, Yulan Mu.

Data curation: Yulan Mu.

Formal analysis: Yulan Mu, Xiaoyan Qin.

Funding acquisition: Changzhong Li.

Investigation: Yulan Mu, Hui Zhang, Xiaoyan Qin.

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- Project administration: Hui Zhang, Ming Liu, Jipeng Wan, Xiaoyan Qin.

Resources: Hui Zhang, Ming Liu, Jipeng Wan.

Software: Ming Liu, Jipeng Wan.

Supervision: Changzhong Li, Min Shi, Ming Liu, Jipeng Wan.

Validation: Changzhong Li, Min Shi.

Visualization: Min Shi.

Writing – original draft: Min Shi.

Writing - review and editing: Min Shi.

References

- Torre LA, Bray F, Siegel RL, et al. Global cancer statistics, 2012. CA Cancer J Clin 2015;65:87–108.
- [2] Tognon G, Carnazza M, Ragnoli M, et al. Prognostic factors in earlystage ovarian cancer. Ecancermedicalscience 2013;7:325.
- [3] Marchini S, Cavalieri D, Fruscio R, et al. Association between miR-200c and the survival of patients with stage I epithelial ovarian cancer: a retrospective study of two independent tumour tissue collections. Lancet Oncol 2011;12:273–85.
- [4] Zhang J-X, Song W, Chen Z-H, et al. Prognostic and predictive value of a microRNA signature in stage II colon cancer: a microRNA expression analysis. Lancet Oncol 2013;14:1295–306.
- [5] Ueda T, Volinia S, Okumura H, et al. Relation between microRNA expression and progression and prognosis of gastric cancer: a microRNA expression analysis. Lancet Oncol 2010;11:136–46.
- [6] Dou Z, Lin S, Dai C, et al. Pooling-analysis for diagnostic and prognostic value of MiRNA-100 in various cancers. Oncotarget 2017;8:62703–15.
- [7] Mahdian-Shakib A, Dorostkar R, Tat M, et al. Differential role of microRNAs in prognosis, diagnosis, and therapy of ovarian cancer. Biomed Pharmacother 2016;84:592–600.
- [8] Cheng CJ, Bahal R, Babar IA, et al. MicroRNA silencing for cancer therapy targeted to the tumour microenvironment. Nature 2015;518:107–10.
- [9] Korpal M, Lee ES, Hu G, et al. The miR-200 family inhibits epithelialmesenchymal transition and cancer cell migration by direct targeting of E-cadherin transcriptional repressors ZEB1 and ZEB2. J Biol Chem 2008;283:14910–4.
- [10] Leskela S, Leandro-Garcia LJ, Mendiola M, et al. The miR-200 family controls beta-tubulin III expression and is associated with paclitaxelbased treatment response and progression-free survival in ovarian cancer patients. Endocr Relat Cancer 2011;18:85–95.
- [11] Shi C, Zhang Z. The prognostic value of the miR-200 family in ovarian cancer: a meta-analysis. Acta Obstet Gynecol Scand 2016;95:505–12.

- [12] Wu J, Zheng C, Wang X, et al. MicroRNA-30 family members regulate calcium/calcineurin signaling in podocytes. J Clin Invest 2015;125:4091–106.
- [13] Zhao H, Ding Y, Tie B, et al. miRNA expression pattern associated with prognosis in elderly patients with advanced OPSC and OCC. Int J Oncol 2013;43:839–49.
- [14] Wang Y, Li L, Qu Z, et al. The expression of miR-30a* and miR-30e* is associated with a dualistic model for grading ovarian papillary serious carcinoma. Int J Oncol 2014;44:1904–14.
- [15] Moher D, Liberati A, Tetzlaff J, et al. Preferred reporting items for systematic reviews and meta-analyses: the PRISMA statement. PLoS Med 2009;6:e1000097.
- [16] McShane LM, Altman DG, Sauerbrei W, et al. REporting recommendations for tumour MARKer prognostic studies (REMARK). Br J Cancer 2005;93:387–91.
- [17] Parmar MK, Torri V, Stewart L. Extracting summary statistics to perform meta-analyses of the published literature for survival endpoints. Stat Med 1998;17:2815–34.
- [18] Tierney JF, Stewart LA, Ghersi D, et al. Practical methods for incorporating summary time-to-event data into meta-analysis. Trials 2007;8:16.
- [19] Gasparini P, Cascione L, Fassan M, et al. microRNA expression profiling identifies a four microRNA signature as a novel diagnostic and prognostic biomarker in triple negative breast cancers. Oncotarget 2014;5:1174–84.
- [20] Hu X, Macdonald DM, Huettner PC, et al. A miR-200 microRNA cluster as prognostic marker in advanced ovarian cancer. Gynecol Oncol 2009;114:457–64.
- [21] Tian T, Wang M, Zhu W, et al. MiR-146a and miR-196a-2 polymorphisms are associated with hepatitis virus-related hepatocellular cancer risk: a meta-analysis. Aging 2017;9:381–92.
- [22] Bagnoli M, Canevari S, Califano D, et al. Development and validation of a microRNA-based signature (MiROvaR) to predict early relapse or progression of epithelial ovarian cancer: a cohort study. Lancet Oncol 2016;17:1137–46.
- [23] Cao Q, Lu K, Dai S, et al. Clinicopathological and prognostic implications of the miR-200 family in patients with epithelial ovarian cancer. Int J Clin Exp Pathol 2014;7:2392–401.
- [24] Lee H, Park CS, Deftereos G, et al. MicroRNA expression in ovarian carcinoma and its correlation with clinicopathological features. World J Surg Oncol 2012;10:174.
- [25] Cao J, Cai J, Huang D, et al. miR-335 Represents an Independent Prognostic Marker in Epithelial Ovarian Cancer. Am J Clin Pathol 2014;141:437–42.

- [26] Kapetanakis NI, Uzan C, Jimenez-Pailhes AS, et al. Plasma miR-200b in ovarian carcinoma patients: distinct pattern of pre/post-treatment variation compared to CA-125 and potential for prediction of progression-free survival. Oncotarget 2015;6:36815–24.
- [27] Meng X, Muller V, Milde-Langosch K, et al. Diagnostic and prognostic relevance of circulating exosomal miR-373, miR-200a, miR-200b and miR-200c in patients with epithelial ovarian cancer. Oncotarget 2016;7:16923–35.
- [28] Calura E, Paracchini L, Fruscio R, et al. A prognostic regulatory pathway in stage I epithelial ovarian cancer: new hints for the poor prognosis assessment. Ann Oncol 2016;27:1511–9.
- [29] Elgaaen BV, Olstad OK, Haug KBF, et al. Global miRNA expression analysis of serous and clear cell ovarian carcinomas identifies differentially expressed miRNAs including miR-200c-3p as a prognostic marker. BMC Cancer 2014;14.
- [30] Gao YC, Wu J. MicroRNA-200c and microRNA-141 as potential diagnostic and prognostic biomarkers for ovarian cancer. Tumour Biol 2015;36:4843–50.
- [31] Sestito R, Cianfrocca R, Rosano L, et al. miR-30a inhibits endothelin A receptor and chemoresistance in ovarian carcinoma. Oncotarget 2016;7:4009–23.
- [32] Li N, Kaur S, Greshock J, et al. A combined array-based comparative genomic hybridization and functional library screening approach identifies mir-30d as an oncomir in cancer. Cancer Res 2012;72:154–64.
- [33] Gregory PA, Bert AG, Paterson EL, et al. The miR-200 family and miR-205 regulate epithelial to mesenchymal transition by targeting ZEB1 and SIP1. Nat Cell Biol 2008;10:593–601.
- [34] Vilming Elgaaen B, Olstad OK, Haug KB, et al. Global miRNA expression analysis of serous and clear cell ovarian carcinomas identifies differentially expressed miRNAs including miR-200c-3p as a prognostic marker. BMC Cancer 2014;14:80.
- [35] Hurteau GJ, Carlson JA, Roos E, et al. Stable expression of miR-200c alone is sufficient to regulate TCF8 (ZEB1) and restore E-cadherin expression. Cell Cycle 2009;8:2064–9.
- [36] Lu C, Shahzad MM, Wang H, et al. EphA2 overexpression promotes ovarian cancer growth. Cancer Biol Ther 2008;7:1098–103.
- [37] Ye Z, Li J, Han X, et al. TET3 inhibits TGF-beta1-induced epithelialmesenchymal transition by demethylating miR-30d precursor gene in ovarian cancer cells. J Exp Clin Cancer Res 2016;35:72.
- [38] Ye Z, Zhao L, Li J, et al. miR-30d Blocked transforming growth factor beta1-induced epithelial-mesenchymal transition by targeting snail in ovarian cancer cells. Int J Gynecol Cancer 2015;25:1574–81.