## Review Article

# An update of the prognostic value of miR-155 in various tumors: a meta-analysis

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Received March 21, 2018; Accepted October 23, 2018; Epub February 15, 2019; Published February 28, 2019

Abstract: Accumulated studies have proved the significant prognostic relevance of Micro-RNAs in many cancers, and miR-155 is one of the most potential Micro-RNAs. However, the association between miR-155 expression and prognostic outcomes of various tumors is still controversial. Therefore this study aims to comprehensively illustrate whether miR-155 has prognostic value for several cancers or not. According to the heterogeneity, pooled hazard ratios (HRs) with 95% confidence intervals (Cls) for overall survival (OS) and disease-free survival (DFS), including relapse-free survival (RFS), event-free survival (EFS), treatment-free survival (TFS), cancer-specific mortality (CSS), time to progression (TTP), and cancer-free survival (CFS) were estimated with the random or fixed effects models. A total of 46 studies containing 5925 cases were selected for analysis. Prognostic analyses on the pooled HR (hazard ratio) of OS and DFS were 1.67 (95% Cl 1.44, 1.93) and 1.99 (95% Cl 1.67, 2.36) for high miR-155 expression and low miR-155 expression, respectively. Further subgroup analyses were performed according to the region, sample sources, tumor types, HR resources, cut off values and independent factors. Most of the subgroup analyses indicated that high-expression of miR-155 cluster was associated with poor OS and DFS. Our study indicated that high expression of miR-155 was significantly related to the poor survival in several cancer patients, and thus could be an important predictor of poor prognosis.

Keywords: miR-155, prognosis, tumour, meta-analysis

#### Introduction

Micro-RNAs (miRNAs) have direct implications for fundamental biology as well as disease etiology and treatment [1], and are small and single-stranded noncoding RNAs consist of approximately 18-22 evolutionarily conserved nucleotides in length. Numerous studies have reported that up-regulation or down-regulation of mi-RNAs contribute to tumor initiation and progression in various malignancies via regulation the target genes [2, 3]. Among these Micro-RNAs, miR-155 family has been reported to develop prognostic role in multiple carcinomas [4-6].

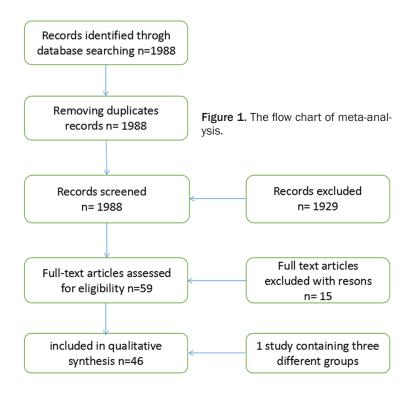
Among all of the miRNAs, miR-155 was well studied in many tumors. Recent studies demonstrated that miR-155 can not only function as a tumor-promoting miRNA by targeting E2F2 in ccRcc [7], but can also play as a tumor-promoting role by regulating the BCL6/cyclin D2 axis in some cancers [8]. These results sugge-

sted that miR-155 has prognostic value in various cancers. In addition, some meta-analyses summarized that miR-155 had prognostic value in cancer patients and regularly measuring miR-155 expression may be useful in clinical practice [9]. But on the other hand, other study demonstrated that high expression of miR-155 in some tumors was not related to the patients' survival [10]. We conduct this meta-analysis to illustrate the relationship between the expression of miR-155 and prognosis of tumor patients.

## Materials and methods

Literature searching strategy

Conducting a systematic literature search through the online databases of Pubmed, Embase and Web of Science, the searching key words included "microRNA-155 OR miR-155 OR MiRNA-155" (all fields). We conducted the search



manually so as to cover all the concerned studies without using too many key words. The search was performed independently by two authors (Li, Xu).

#### Including and excluding criteria

Literatures meeting the following criteria were considered eligible: 1). Various associated human malignancies and survival outcomes with miR-155 expression; 2). Hazard ratios (HRs) and 95% confidence intervals (95% CI) were provided or could be estimated by kaplanmeier curves; 3). MiR-155 expression was detected by qRT-PCR, in situ hybridization or microarray; 4). Study including the patients' prognostic parameters of OS/DFS/PFS/RFS/ MFS/TTP, and miR-155 was measured in tumor tissue, peripheral blood or body liquid. Articles were excluded based on the following items: 1). Case reports, reviews, conference abstracts, letters, and animals trials. 2). The study only conducted on animal models or tumor cell lines. 3). No more than 20 participants in order to make sure the persuasiveness.

#### Data extraction

Based on the including and excluding criteria, we extracted the relevant information from ea-

ch eligible publication. The extracted information included the first author, publication year, study country, journal name, cancer type, sample source, test method, cutoff value, number of participants, survival outcome, independent factor, and the method of analysis. HR and 95% CI and the cut-off value were extracted. Sample source was stratified into tissue, blood, urine, formalin-fixed and paraffinembeded (FFPE). Test methods included in situ hybridization (ISA), reverse transcription-polymerase chain reaction (gRT-PCR) and microarray. Cancer types included solid cancer and more. The results from multivariate analyses or univariate analyses were allowed in the meta-analysis. Additionally, if both multivariate analysis and

univariate analysis were not available, kaplanmeier curves were adopted to extract HR by Engauge Digitizer 4.1.

## Statistical analyses

According to the cut-off value provided in the article, the miR-155 expression was defined as high expression or low expression. Pooled data was calculated by HRs and the corresponding 95% Cls. For the prognostic results, a combined HR > 1 indicated that high expression of miR-155 was associated with the poor outcome. Cochran's Q-test and Higgins-I2 statistics (I2) were used to test the heterogeneity of pooled HRs [11]. If the significant heterogeneity was observed at the percentage of I2 was greater than 50% or the P<0.05, the random-effects model was applied to calculate the pooled HR and 95% CI of survival outcomes. Otherwise, the fixed-effects model was applied. In addition, we performed the sub-analysis to minimize the heterogeneity. Publication bias was assessed by Begg's test [12]. All of the statistical calculations were conducted by Stata version 12.0 (Stata Corporation, College Station. TX, USA). It was considered statistical significant if the P value was less than 0.05.

## Prognostic value of miR-155 in various tumors

Table 1. Subgroup analysis between miR-155 expression and prognosis of patients

	OS (n = 27)				DFS (n = 29)					
Subgroup	No. of studies	Model	Pooled HR (95% CI)	P value	HG I <sup>2</sup> %	No. of studies	Model	Pooled HR (95% CI)	P value	HG I <sup>2</sup> %
	27	random	1.67 (1.44, 1.93)	<0.00001	79	29	random	1.99 (1.67, 2.36)	<0.00001	60
Tumor type										
Colorectal cancer	2	fixed	3.30 (2.04, 5.33)	<0.00001	0	2	fixed	2.72 (1.55, 4.77)	0.0005	0
Blader cancer						3	random	4.13 (1.37, 12.45)	0.0004	53
Renal carcinoma						3	fixed	2.15 (1.35, 3.42)	0.001	0
Oral squamous carcinoma						2	fixed	2.87 (1.15, 7.12)	0.02	26
Lymphoma						3	fixed	2.06 (1.40, 3.03)	0.0003	0
Leukemia	6	random	1.76 (1.31, 2.37)	<0.00001	57	3	fixed	1.41 (1.17, 1.69)	0.0003	5
Lung cancer	5	Random	1.52 (0.77, 2.98)	0.0001	78	7	Random	1.60 (1.16, 2.20)	<0.0001	51
Pancreatic cancer	4	random	1.59 (0.81, 3.13)	0.003	72	1	ND	3.33 (1.20, 9.25)	0.02	ND
Glioma	1	ND	1.74 (0.21, 14.53)	0.61	ND	ND	ND	ND	ND	ND
Gallblader carcinoma	1	ND	1.44 (1.04, 2.00)	0.03	ND	1	ND	1.96 (1.17, 3.27)	0.01	ND
Hepatocellular	2	Random	2.95 (1.02, 8.55)	<0.0001	74	3	Random	2.60 (1.66, 4.06)	<0.00001	56
Breast cancer	3	fixed	2.42 (1.66, 3.52)	<0.00001	0					
Liposarcoma	1	ND	2.90 (1.38, 6.10)	0.005	ND	1	Random	2.11 (1.06, 4.21)	0.03	ND
Chordoma	1	ND	6.35 (1.53, 26.4)	0.001	ND	ND	ND	ND	ND	ND
GBM	1	ND	0.78 (0.63, 0.96)	0.02	ND	ND	ND	ND	ND	ND
Sample type										
Tissues	11	Random	2.03 (1.37, 3.00)	0.0004	83	12	fixed	2.04 (1.67, 2.50)	<0.00001	0
FFPE	8	Random	1.62 (1.14, 2.31)	0.007	77	8	fixed	2.29 (1.80, 2.93)	<0.00001	21
Plasma	1	ND	1.09 (1.02, 1.16)	0.008	ND	2	Random	1.74 (0.64, 4.69)	0.004	75
Urine						2	Random	3.42 (0.79, 14.81)	0.005	56
Blood sample	3	fixed	1.82 (1.40, 2.37)	<0.00001	0	3	fixed	1.41 (1.17, 1.69)	0.0003	5
Serum	1	ND	3.86 (1.58, 9.44)	0.003	ND	2	fixed	2.60 (1.62, 4.18)	<0.0001	0
Marrow	3	Random	1.73 (1.01, 2.99)	0.05	62					
HR resource										
Reported	17	Random	1.55 (1.32, 1.82)	<0.00001	83	21	fixed	1.83 (1.61, 2.07)	<0.00001	43
SC	10	Fixed	1.84 (1.53, 2.23)	<0.00001	7	8	Random	1.74 (1.27, 2.39)	<0.00001	63
Region										
Asian	15	Random	1.61 (1.33, 1.93)	<0.00001	81	17	fixed	2.12 (1.81, 2.48)	<0.00001	18
Europe	4	Random	2.06 (0.83, 5.11)	0.12	85	6	fixed	1.23 (1.11, 1.38)	0.0002	44
America	8	Random	1.86 (1.39, 2.50)	<0.0001	60	5	fixed	1.47 (1.21, 1.79)	<0.0001	50
Egypt						1	ND	2.64 (1.51-4.63)	0.004	ND
Cut-off value										
Median	21	Random	1.83 (1.45, 2.31)	<0.00001	76	21	Random	1.88 (1.55, 2.30)	<0.00001	63
Mean	4	Random	1.13 (1.08, 1.19)	<0.0001	86	4	fixed	2.95 (1.79, 4.88)	<0.0001	0
Others	2	fixed	2.52 (1.63, 3.92)	<0.0001	0	4	fixed	1.94 (1.48, 2.55)	<0.00001	0
Independent risk factor			•					,		
Yes	22	Random	1.72 (1.47, 2.02)	<0.00001	81	22	Random	2.06 (1.70, 2.51)	<0.00001	67
NR	5	Random	1.52 (0.83, 2.79)	0.17	65	7	fixed	1.69 (1.22, 2.35)	0.001	0

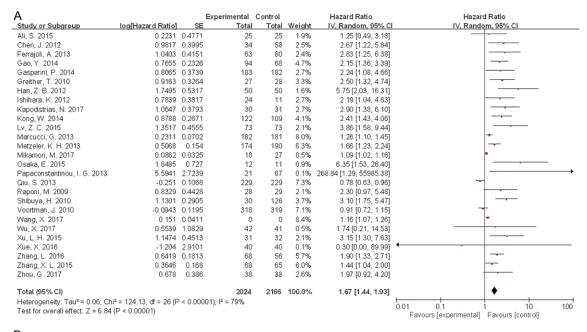
OS-overall survival, DFS-disease free survival, ND-no data, GBM-glioblastoma multiform, FFPE-formalin-fixed paraffin embedded samples, SC-survival curve, NR-not reported, HG-heterogeneity, HR-hazard ratio, Random-random-effect model, fixed-fixed-effect model.

#### Results

#### Characteristics of included studies

A total of 46 studies including 5925 cases were involved in this meta-analysis (**Figure 1**). The basic characteristics were summarized in **Table 1**. Of all the studies, except one from Egypt, 32 were from Asia, 13 were from America, 10 were

from Europe. Tumor types included the colon cancer [13, 14], bladder cancer [15, 16], renal carcinoma [17-19], oral squamous carcinoma [20, 21], lymphoma [22-24], leukemia [25-31], lung cancer [10, 32-39], glioma [40], gallblader carcinoma [41], hepatocellular carcinoma [42, 43], breast cancer [44, 45], liposarcoma [46], pancreatic cancer [47-50], chordoma [51], and glioblastoma multiforme [52]. Meanwhile, there



Baba, O. 2016				Experimental			Hazard Ratio	Hazard Ratio
Bedewy, A. M. 2017		log[Hazard Ratio]	SE	Total				IV, Random, 95% CI
Cui, B. 2014								
Donnem, T. 2011	,,							
Han, Z. B. 2012-2 1.5676 0.3549 50 50 3.5% 4.80 [2.39, 9.61] Huang, Y. H. 2012 0.6941 0.211 107 109 5.4% 2.00 [1.32, 3.03] Idpal, J. 2015 0.3011 1.2818 27 27 0.4% 0.74 [0.06, 9.13] Kapodistrias, N. 2017-2 0.7467 0.3522 30 31 3.5% 2.11 [1.06, 4.21] Kono, H. 2013 0.6729 0.2613 13 13 4.7% 1.96 [1.17, 3.27] U. Y. Z. C. 2015-2 0.9377 0.3536 73 73 3.5% 2.55 [1.28, 5.11] Marcucci, G. 2013-2 0.1823 0.1519 182 181 6.4% 1.20 [0.89, 1.62] Merhautova, J. 2015 0.8805 0.3371 21 42 3.7% 2.41 [1.25, 4.67] Metzeler, K. H. 2013-2 0.392 0.1522 174 190 6.4% 14.8 [1.10, 1.99] Mikamoni, M. 2017-2 1.203 0.5212 18 27 2.1% 3.33 [1.20, 9.25] Saito, M. 2011 0.8629 0.314 44 45 3.9% 2.37 [1.28, 4.39] Saito, M. 2011-2 0.47 0.4043 18 19 3.0% 16.0 [0.72, 3.53] Saito, M. 2011-3 0.2652 0.2803 85 86 4.4% 1.33 [0.77, 2.30] Saito, M. 2011-3 0.2652 0.2803 85 86 4.4% 1.33 [0.77, 2.30] Sanforenzo, C. 2013 0.157 0.0592 26 26 26 7.7% 1.17 [1.04, 1.31] Shil, L. J. 2015 0.585 0.6305 11 19 1.6% 1.75 [0.51, 6.02] Shinmei, S. 2013 0.8372 0.4044 69 68 3.0% 2.31 [1.05, 5.10] Wang, H. 2010-2 1.1207 0.493 30 126 2.3% 3.07 [1.17, 8.06] Shinmei, S. 2013 0.8372 0.4044 69 68 3.0% 2.31 [1.05, 5.10] Wang, H. 2016-2 0.0834 0.7537 40 40 1.2% 0.92 [0.21, 4.03] Yue, X. 2016-2 0.0834 0.7537 40 40 1.2% 0.92 [0.21, 4.03] Yanaihara, N. 2006 1.2296 0.4485 0 0 2.6% 3.42 [1.42, 8.24] Zhang, X. 2016-2 0.8242 0.2402 68 56 5.0% 2.28 [1.42, 8.24] Zhang, X. 2016-2 0.8242 0.2402 68 56 5.0% 2.28 [1.42, 8.24] Zhang, X. 2016-2 2.3171 1.0117 130 32 0.7% 10.15 [1.40, 7.370] Zhong, H. 2012 0.6206 0.2499 39 51 4.8% 186 [1.14, 3.04]								
Huang, Y. H. 2012								
Iqbal, J. 2015       -0.3011       1.2818       27       27       0.4%       0.74 (0.06, 9.13)         Kapodistrias, N. 2017-2       0.7467       0.3522       30       31       3.5%       2.11 [1.06, 4.21]         Kono, H. 2013       0.6729       0.2613       13       13       4.7%       1.96 [1.17, 3.27]         Lv, Z. C. 2015-2       0.9377       0.3536       73       73       3.5%       2.55 [1.28, 5.11]         Marcucci, G. 2013-2       0.1823       0.1519       182       181       6.4%       1.20 [0.89, 162]         Merbautova, J. 2015       0.8805       0.3371       21       42       3.7%       2.41 [1.25, 4.67]         Metzeler, K. H. 2013-2       0.392       0.1522       174       190       6.4%       1.48 [1.10, 1.99]         Mikamori, M. 2017-2       1.203       0.5212       18       27       2.1%       3.33 [1.20, 9.25]         Saito, M. 2011-1       0.8629       0.314       44       45       3.9%       2.37 [1.28, 4.39]         Saito, M. 2011-3       0.2852       0.2803       85       86       4.4%       1.33 [0.77, 2.30]         Saito, M. 2015       0.157       0.0592       26       26       7.7%       1.17 [1.04, 1.31]								
Kapodistrias, N. 2017-2       0.7467 0.3522       30       31 3.5%       2.11 [1.06, 4.21]         Kono, H. 2013       0.6729 0.2613       13 13 4.7%       1.96 [1.17, 3.27]         Lv, Z. C. 2015-2       0.9377 0.3536       73 73 3.5%       2.55 [1.28, 5.11]         Mercucci, G. 2013-2       0.1823 0.1519       182 181 6.4%       1.20 [0.89, 1.62]         Merhautova, J. 2015       0.8805 0.3371       21 42 3.7%       2.41 [1.25, 4.67]         Metzeler, K. H. 2013-2       0.392 0.1522       174 190 6.4%       1.48 [1.10, 1.99]         Mikamori, M. 2017-2       1.203 0.5212       18 27 2.1%       3.33 [1.20, 9.25]         Saito, M. 2011       0.8629 0.314       44 45 3.9%       2.37 [1.28, 4.39]         Saito, M. 2011-3       0.2852 0.2803       85 86 4.4%       1.33 [0.77, 2.30]         Saito, M. 2011-3       0.2852 0.2803       85 86 4.4%       1.33 [0.77, 2.30]         Saito, M. 2015       0.5585 0.6305       11 19 1.6%       1.75 [0.51, 6.02]         Shibuya, H. 2010-2       1.1207 0.493       30 126 2.3%       3.07 [1.17, 8.06]         Shibuya, H. 2010-2       1.1207 0.493       30 126 2.3%       3.07 [1.17, 8.06]         Worag, H. 2015       0.0584 0.7537       40 40 1.2% 0.92 [0.21, 4.03]         Yanaihara, N. 2006       1.2296 0.4485 <td>Huang, Y. H. 2012</td> <td></td> <td></td> <td></td> <td>109</td> <td>5.4%</td> <td>2.00 [1.32, 3.03]</td> <td></td>	Huang, Y. H. 2012				109	5.4%	2.00 [1.32, 3.03]	
Kono, H. 2013	Iqbal, J. 2015			27	27	0.4%	0.74 [0.06, 9.13]	
Lv, Z. C. 2015-2								
Marcucci, G. 2013-2       0.1823       0.1519       182       181       6.4%       1.20 [0.89, 1.62]         Merhautova, J. 2015       0.8805       0.3371       21       42       3.7%       2.41 [1.25, 4.67]         Metzeler, K. H. 2013-2       0.392       0.1522       174       190       6.4%       1.48 [1.10, 1.99]         Mikamori, M. 2017-2       1.203       0.5212       18       27       2.1%       3.33 [1.20, 9.25]         Saito, M. 2011       0.8629       0.314       44       45       3.9%       2.37 [1.28, 4.39]         Saito, M. 2011-3       0.2852       0.2803       85       86       4.4%       1.33 [0.77, 2.30]         Sanforenzo, C. 2013       0.157       0.0592       26       26       7.7%       1.17 [1.04, 1.31]         Shi, L. J. 2015       0.5585       0.6305       11       19       1.6%       1.75 [0.51, 6.02]         Shibuya, H. 2010-2       1.1207       0.493       30       126       2.3%       3.07 [1.17, 8.06]         Shihuya, H. 2015       2.0412       0.7815       52       50       1.1%       7.70 [1.68, 35.62]         Wotschofsky, Z. 2013       0.2776       0.5814       56       55       1.8%       1.32 [0.42, 4.13] <td>Kono, H. 2013</td> <td>0.6729</td> <td>0.2613</td> <td>13</td> <td>13</td> <td>4.7%</td> <td>1.96 [1.17, 3.27]</td> <td> <del></del></td>	Kono, H. 2013	0.6729	0.2613	13	13	4.7%	1.96 [1.17, 3.27]	<del></del>
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Metzeler, K. H. 2013-2       0.392 0.1522       174 190 6.4%       1.48 [1.10, 1.99]         Mikamori, M. 2017-2       1.203 0.5212       18 27 2.1%       3.33 [1.20, 9.25]         Saito, M. 2011       0.8629 0.314 44 45 3.9%       2.37 [1.28, 4.39]         Saito, M. 2011-2       0.47 0.4043 18 19 3.0%       1.60 [0.72, 2.53]         Saito, M. 2011-3       0.2852 0.2803 85 86 4.4%       1.33 [0.77, 2.30]         Sanforenzo, C. 2013 0.157 0.0592 26 26 7.7%       1.17 [1.04, 1.31]         Shi, L. J. 2015 0.5585 0.6305 11 19 1.6% 1.75 [0.51, 6.02]         Shibuya, H. 2010-2 1.1207 0.493 30 126 2.3%       3.07 [1.17, 8.06]         Shimmel, S. 2013 0.8372 0.4044 69 68 3.0% 2.31 [1.05, 5.10]         Wang, H. 2015 2.0412 0.7815 52 50 1.1% 7.70 [1.66, 35.62]         Wotschofsky, Z. 2013 0.2776 0.5814 56 55 1.8% 1.32 [0.42, 4.13]         Xue, X. 2016-2 -0.0834 0.7537 40 40 1.29% 0.92 [0.21, 4.03]         Yanaihara, N. 2006 1.2296 0.4485 0 0 2.8% 3.42 [1.42, 8.65]         Zhang, X. 2016-2 0.8242 0.2402 68 56 5.0% 2.28 [1.42, 3.65]         Zhang, X. 2016-2 2.3171 1.0117 130 32 0.7% 0.15 [1.40, 73.70]         Zhang, H. 2012 0.6206 0.2499 39 51 4.8% 1.86 [1.14, 3.04]         Heterocoreity Tata = 0.40 Chill 7.0 27 cf = 28 (P. 0.0001) 18 - 5006	Marcucci, G. 2013-2	0.1823	0.1519	182	181	6.4%	1.20 [0.89, 1.62]	<del> </del>
Mikamori, M. 2017-2  1.203 0.5212  18 27 2.196  3.33 [1.20, 9.25]  Saito, M. 2011  0.8629  0.314  44 45 3.996  2.37 [1.28, 4.39]  Saito, M. 2011-2  0.47 0.4043  18 19 3.096  1.80 [0.77, 2.30]  Saito, M. 2011-3  0.2852 0.2803  85 86 4.496  1.33 [0.77, 2.30]  Sanforenzo, C. 2013  0.157 0.0592  26 26 7.796  1.17 [1.04, 1.31]  Shi, L. J. 2015  0.5585 0.6305  11 19 1.696  1.75 [0.51, 6.02]  Shibuya, H. 2010-2  1.1207 0.493  30 126 2.396  3.07 [1.17, 8.06]  Shimei, S. 2013  0.8372 0.4044  69 68 3.096  2.31 [1.05, 5.10]  Wong, H. 2015  2.0412 0.7815  52 50 1.196  7.70 [1.66, 35.62]  Wotschofsky, Z. 2013  0.2776 0.5814  55 1.896  1.32 [0.42, 4.13]  Xue, X. 2016-2  -0.0834 0.7537  40 40 1.296  0.92 [0.21, 4.03]  Yanaihara, N. 2006  1.296 0.4485  0 0 2.696  3.42 [1.42, 8.24]  Zhang, X. 2016-2  2.3171 1.0117  130 32 0.796  1.86 [1.14, 3.04]  Total (95% CI)	Merhautova, J. 2015	0.8805	0.3371	21	42	3.7%	2.41 [1.25, 4.67]	
Saito, M. 2011  0.8629  0.314  44  45  3.9%  2.37 [1.28, 4.39]  Saito, M. 2011-2  0.47  0.4043  18  19  3.0%  1.60 [0.72, 3.53]  Santo, M. 2011-3  0.2852  0.2803  85  86  4.4%  1.33 [0.77, 2.30]  Sanforenzo, C. 2013  0.157  0.0592  26  26  27.7%  1.17 [1.04, 1.31]  Shi, L. J. 2015  0.5585  0.6305  11  19  1.6%  1.75 [0.51, 6.02]  Shibuya, H. 2010-2  1.1207  0.493  30  126  2.39%  3.07 [1.17, 8.06]  Shibuya, H. 2010-2  1.1207  0.493  30  126  2.39%  3.07 [1.17, 8.06]  Shimmei, S. 2013  0.8372  0.4044  69  68  3.0%  2.31 [1.05, 5.10]  Wang, H. 2015  2.0412  0.7815  52  50  1.1%  7.70 [1.66, 35.62]  Wotschofsky, Z. 2013  0.2776  0.5814  56  55  1.8%  1.32 [0.42, 4.13]  Xue, X. 2016-2  -0.0834  0.7537  40  40  1.296  0.92 [0.21, 4.03]  Yanaihara, N. 2006  1.296  0.4485  0  0  0.26%  3.89%  2.04 [1.05, 3.85]  Zhang, X. 2016-2  0.8242  0.2402  68  56  5.0%  2.28 [1.42, 3.65]  Zhang, X. 2016-2  2.3171  1.0117  130  32  0.7%  1.86 [1.14, 3.04]  Total (95% CI)  Heteroconciby Tatal = 0.40 Chils = 70.27 eff = 28 (P. 0.0001) is = 60%	Metzeler, K. H. 2013-2	0.392	0.1522	174	190	6.4%	1.48 [1.10, 1.99]	
Saito, M. 2011-2	Mikamori, M. 2017-2	1.203	0.5212	18	27	2.1%	3.33 [1.20, 9.25]	
Saito, M. 2011-3       0.2852       0.2803       85       86       4.4%       1.33 [0.77, 2.30]         Sanforenzo, C. 2013       0.157       0.0592       26       26       7.7%       1.17 [1.04, 1.31]         Shi, L. J. 2015       0.5585       0.6305       11       19       1.6%       1.75 [0.51, 6.02]         Shibuya, H. 2010-2       1.1207       0.493       30       126       2.3%       3.07 [1.17, 8.06]         Shimei, S. 2013       0.8372       0.4044       69       68       3.0%       2.31 [1.05, 5.10]         Wang, H. 2015       2.0412       0.7815       52       50       1.1%       7.70 [1.66, 35.62]         Wotschofsky, Z. 2013       0.2776       0.5814       56       55       1.8%       1.32 [0.42, 4.13]         Xue, X. 2016-2       -0.0834       0.7537       40       40       1.296       0.92 [0.21, 4.03]         Yanaihara, N. 2006       1.296       0.4485       0       0       2.6%       3.42 [1.42, 3.65]         Zhang, X. 2016-2       0.8242       0.2402       68       56       50, 90       2.28 [1.42, 3.65]         Zhang, X. 2016-2       2.3171       1.0117       130       32       3.8%       2.04 [1.08, 3.85]	Saito, M. 2011	0.8629	0.314	44	45	3.9%	2.37 [1.28, 4.39]	
Sanfiorenzo, C. 2013  0.157 0.0592  26 26 7.7%  1.17 [1.04, 1.31]  Shi, L. J. 2015  0.5585 0.6305  11 19 1.6%  1.75 [0.51, 6.02]  Shibuya, H. 2010-2  1.1207 0.493  30 126 2.3%  3.07 [1.17, 8.06]  Shimei, S. 2013  0.8372 0.4044  69 68 3.0%  2.31 [1.05, 5.10]  Wang, H. 2015  2.0412 0.7815  52 50 1.1%  7.70 [1.66, 35.62]  Wotschofsky, Z. 2013  0.2776 0.5814  56 55 1.8%  1.32 [0.42, 4.13]  Xue, X. 2016-2  -0.0834 0.7537  40 40 1.2%  0.92 [0.21, 4.03]  Yanaihara, N. 2006  1.296 0.4485  0 0 2.6%  3.42 [1.42, 8.24]  Zhang, L. 2016-2  0.8242 0.2402  68 56 5.0%  2.28 [1.42, 3.65]  Zhang, X. 2016-2  0.7144 0.323  130 32 3.8%  2.04 [1.08, 3.85]  Zhang, X. 2016-2  2.3171 1.0117  130 32 0.7%  1.86 [1.14, 3.04]  Total (95% CI)  Heterogoapity Tatal = 0.40 Chils = 70.27 ef = 28 (R. c. 0.0001) is = 60%	Saito, M. 2011-2	0.47	0.4043	18	19	3.0%	1.60 [0.72, 3.53]	<del>                                     </del>
Shi, L. J. 2015 0.5585 0.6305 11 19 1.6% 1.75 [0.51, 6.02] Shibuya, H. 2010-2 1.1207 0.493 30 126 2.3% 3.07 [1.17, 8.06] Shinmei, S. 2013 0.8372 0.4044 69 68 3.0% 2.31 [1.05, 5.10] Wang, H. 2015 2.0412 0.7815 52 50 1.1% 7.70 [1.66, 35.62] Wotschofsky, Z. 2013 0.2776 0.5814 56 55 1.8% 1.32 [0.42, 4.13] Xue, X. 2016-2 -0.0834 0.7537 40 40 1.2% 0.92 [0.21, 4.03] Yanaihara, N. 2006 1.2296 0.4485 0 0 2.6% 3.42 [1.42, 8.64] Zhang, L. 2016-2 0.8242 0.2402 68 56 5.0% 2.28 [1.42, 3.65] Zhang, X. 2016 0.7144 0.323 130 32 3.8% 2.04 [1.08, 3.85] Zhang, X. 2016-2 2.3171 1.0117 130 32 0.7% 10.15 [1.40, 73.70] Zhong, H. 2012 0.6206 0.2499 39 51 4.8% 1.86 [1.14, 3.04]	Saito, M. 2011-3	0.2852	0.2803	85	86	4.4%	1.33 [0.77, 2.30]	<del></del>
Shibuya, H. 2010-2       1.1207       0.493       30       126       2.3%       3.07 [1.17, 8.06]         Shimmei, S. 2013       0.8372       0.4044       69       68       3.0%       2.31 [1.05, 5.10]         Wang, H. 2015       2.0412       0.7815       52       50       1.1%       7.70 [1.66, 35.62]         Wotschofsky, Z. 2013       0.2776       0.5814       56       55       1.8%       1.32 [0.42, 4.13]         Xue, X. 2016-2       -0.0834       0.7537       40       40       1.296       0.92 [0.21, 4.03]         Yanaihara, N. 2006       1.2296       0.4485       0       0       2.8%       3.42 [1.42, 8.24]         Zhang, X. 2016-2       0.8242       0.2402       68       56       5.0%       2.28 [1.42, 8.65]         Zhang, X. 2016       0.7144       0.323       130       32       3.8%       2.04 [1.08, 3.85]         Zhang, X. 2016-2       2.3171       1.0117       130       32       0.7%       10.15 [1.40, 73.70]         Zhong, H. 2012       0.6206       0.2499       39       51       4.8%       1.86 [1.14, 3.04]         Hotel percentible Tails = 0.10 Chils = 70.27 (af = 28 (B = 0.0001) is = 5000       1.99 [1.67, 2.36]	Sanfiorenzo, C. 2013	0.157	0.0592	26	26	7.7%	1.17 [1.04, 1.31]	*
Shinmei, S. 2013     0.8372     0.4044     69     68     3.0%     2.31 [1.05, 5.10]       Wang, H. 2015     2.0412     0.7815     52     50     1.1%     7.70 [1.66, 35.62]       Wotschofsky, Z. 2013     0.2776     0.5814     56     55     1.8%     1.32 [0.42, 4.13]       Xue, X. 2016-2     -0.0834     0.7537     40     40     1.2%     0.92 [0.21, 4.03]       Yanaihara, N. 2006     1.2296     0.4485     0     0     2.6%     3.42 [1.42, 3.24]       Zhang, L. 2016-2     0.8242     0.2402     68     56     5.0%     2.28 [1.42, 3.65]       Zhang, X. 2016     0.7144     0.323     130     32     3.8%     2.04 [1.08, 3.85]       Zhang, X. 2016-2     2.3171     1.0117     130     32     0.7%     10.15 [1.40, 73.70]       Zhong, H. 2012     0.6206     0.2499     39     51     4.8%     1.86 [1.14, 3.04]       Total (95% CI)       Heterocorcibit Tatil = 0.10 Chill = 70.27 (d = 28 (R < 0.0001) (R = 500)	Shi, L. J. 2015	0.5585	0.6305	11	19	1.6%	1.75 [0.51, 6.02]	
Wang, H. 2015       2.0412       0.7815       52       50       1.1%       7.70 [1.66, 35.62]         Wotschofsky, Z. 2013       0.2776       0.5814       56       55       1.8%       1.32 [0.42, 4.13]         Xue, X. 2016-2       -0.0834       0.7537       40       40       1.2%       0.92 [0.21, 4.03]         Yanaihara, N. 2006       1.2296       0.4485       0       0       2.6%       3.42 [1.42, 8.24]         Zhang, L. 2016-2       0.8242       0.2402       68       56       5.0%       2.28 [1.42, 3.65]         Zhang, X. 2016       0.7144       0.323       130       32       3.8%       2.04 [1.08, 3.85]         Zhang, X. 2016-2       2.3171       1.0117       130       32       0.7%       10.15 [1.40, 73.70]         Zhong, H. 2012       0.6206       0.2499       39       51       4.8%       1.86 [1.14, 3.04]         Total (95% CI)         Heterococcibit Tails = 0.10 (Chils = 70.27 of = 28 (Pc. 0.0001) is = 60%	Shibuya, H. 2010-2	1.1207	0.493	30	126	2.3%	3.07 [1.17, 8.06]	-
Wotschofsky, Z. 2013       0.2776       0.5814       56       55       1.8%       1.32 [0.42, 4.13]         Xue, X. 2016-2       -0.0834       0.7537       40       40       1.2%       0.92 [0.21, 4.03]         Yanaihara, N. 2006       1.2296       0.4485       0       0       2.6%       3.42 [1.42, 8.24]         Zhang, L. 2016-2       0.8242       0.2402       68       56       5.0%       2.28 [1.42, 3.65]         Zhang, X. 2016       0.7144       0.323       130       32       3.8%       2.04 [1.08, 3.85]         Zhang, X. 2016-2       2.3171       1.0117       130       32       0.7%       10.15 [1.40, 73.70]         Zhong, H. 2012       0.6206       0.2499       39       51       4.8%       1.86 [1.14, 3.04]         Total (95% CI)         Heterogeogible Tails = 0.10 Chils = 70.27 (df = 28 (Pc < 0.0001) is = 50%	Shinmei, S.2013	0.8372	0.4044	69	68	3.0%	2.31 [1.05, 5.10]	-
Xue, X. 2016-2     -0.0834     0.7537     40     40     1.2%     0.92 [0.21, 4.03]       Yanaihara, N. 2006     1.2296     0.4485     0     0     2.8%     3.42 [1.42, 8.24]       Zhang, L. 2016-2     0.8242     0.2402     68     56     5.0%     2.28 [1.42, 3.65]       Zhang, X. 2016     0.7144     0.323     130     32     3.8%     2.04 [1.08, 3.85]       Zhang, X. 2016-2     2.3171     1.0117     130     32     0.7%     10.15 [1.40, 73.70]       Zhong, H. 2012     0.6206     0.2499     39     51     4.8%     1.86 [1.14, 3.04]	Wang, H. 2015	2.0412	0.7815	52	50	1.1%	7.70 [1.66, 35.62]	<del></del>
Yanaihara, N. 2006     1,2296     0,4485     0     0     2,6%     3,42 [1,42, 8,24]       Zhang, L. 2016-2     0,8242     0,2402     68     56     5,0%     2,28 [1,42, 3,85]       Zhang, X. 2016     0,7144     0,323     130     32     3,8%     2,04 [1,08, 3,85]       Zhang, X. 2016-2     2,3171     1,0117     130     32     0,7%     10,15 [1,40, 73,70]       Zhong, H. 2012     0,6206     0,2499     39     51     4,8%     1,86 [1,14, 3,04]       Total (95% CI)       Heterogogoight, Tails = 0,10; Chils = 70,27 (d = 28 /R < 0.0001); R = 60%	Wotschofsky, Z. 2013	0.2776	0.5814	56	55	1.8%	1.32 [0.42, 4.13]	-
Zhang, L. 2016-2     0.8242 0.2402     68     56     5.0%     2.28 [1.42, 3.65]       Zhang, X. 2016     0.7144 0.323     130     32     3.8%     2.04 [1.08, 3.85]       Zhang, X. 2016-2     2.3171 1.0117     130     32     0.7%     10.15 [1.40, 73.70]       Zhong, H. 2012     0.6206 0.2499     39     51     4.8%     1.86 [1.14, 3.04]       Total (95% CI)       Heterogogoight Tails = 0.40 Chils = 70.27 of = 28 (R ≤ 0.0001); R = 60%	Xue, X. 2016-2	-0.0834	0.7537	40	40	1.2%	0.92 [0.21, 4.03]	
Zhang, X. 2016     0.7144     0.323     130     32     3.8%     2.04 [1.08, 3.85]       Zhang, X. 2016-2     2.3171     1.0117     130     32     0.7%     10.15 [1.40, 73.70]       Zhong, H. 2012     0.6206     0.2499     39     51     4.8%     1.86 [1.14, 3.04]       Total (95% CI)       Heterogogoight Tails = 0.10 Chils = 70.27 of = 28 (P. < 0.0001) is = 50%	Yanaihara, N. 2006	1.2296	0.4485	0	0	2.6%	3.42 [1.42, 8.24]	
Zhang, X. 2016-2     2.3171     1.0117     130     32     0.7%     10.15 [1.40, 73.70]       Zhong, H. 2012     0.6206     0.2499     39     51     4.8%     1.86 [1.14, 3.04]       Total (95% CI)       Heterogonoity Tails = 0.10: Chis = 70.27 of = 28 /B ≤ 0.0001 Vis = 50%	Zhang, L. 2016-2	0.8242	0.2402	68	56	5.0%	2.28 [1.42, 3.65]	_ <del>-</del>
Zhong, H. 2012     0.6206 0.2499     39     51     4.8%     1.86 [1.14, 3.04]       Total (95% CI)       Heterogonoity Total = 0.10: Chis = 70.27 of = 28 /P ≤ 0.0001); is = 50%	Zhang, X. 2016	0.7144	0.323	130	32	3.8%	2.04 [1.08, 3.85]	
Total (95% CI) 1698 1675 100.0% 1.99 [1.67, 2.36]	Zhang, X. 2016-2	2.3171	1.0117	130	32	0.7%	10.15 [1.40, 73.70]	<del></del>
Hotorogonoiby: Tay 8 = 0.40; Chi2 = 70.27, df = 28 /P < 0.0004); I2 = 60%	Zhong, H. 2012	0.6206	0.2499	39	51	4.8%	1.86 [1.14, 3.04]	
Hotorogonoity: Tay 8 = 0.10 : Chi8 = 70.27 : df = 28 /D < 0.00013 : i2 = 6004	Total (95% CI)			1698	1675	100.0%	1.99 [1.67, 2.36]	•
		10: Chi <sup>2</sup> = 70.27. df = 2	28 (P < 0	.0001): I <sup>2</sup> = 60%			,	0.01 0.1 1 10

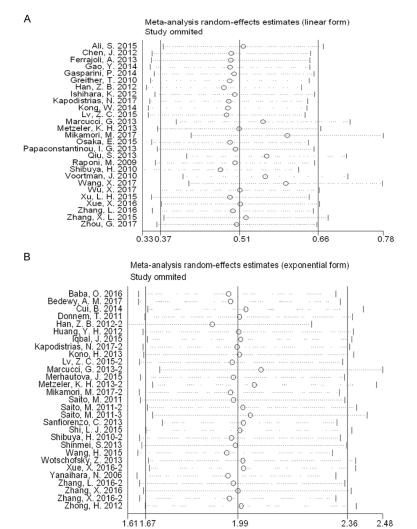
**Figure 2.** A. One way sensitivity analysis of high-expression of miR-155 in various tumors with OS under different types of analysis. B. One way sensitivity analysis of high-expression of miR-155 in various tumors with DFS under different types of analysis.

are such sample types as FFPE, tissue, blood, urine. As for the test methods, most studies applied qRT-PCR, except one study used in situ hybridization (ISH), and another two studies adopted microarray. Most of the cut-off values were median or mean. 44 studies were divided into two datasets: 27 studies for OS analysis

and 29 studies for DFS (including DFS, PFS, RFS, etc) analysis.

#### Meta-analysis of OS

27 studies were involved to evaluate the relation between miR-155 expression and progno-



**Figure 3.** A. One-way sensitivity analysis of high-expression of miR-155 in various tumors. B. One-way sensitivity analysis of high-expression of miR-155 in various tumors.

sis. The results demonstrated that high expression of miR-155 in various tumors was associated with relatively poor OS (HR = 1.67, 95% CI = 1.44, 1.93, P<0.00001; Figure 2A). Because of the heterogeneity among studies (I2 = 79%, P<0.00001), subgroup analysis was performed. Interestingly, high expression of miR-155 could predict worse OS regardless of the region, sample types, tumor types, HR resources, cut off values, independent factors (Table 1). We found the HRs of these tumor types, including lung cancer, pancreatic cancer, were not related to the patients' survival. Similarly, studies from Europe were different from other regions, where there was no relationship between HR and prognosis. Considering the obvious heterogeneity of the results from meta-analysis, a random model was used to calculate the pooled HR and 95% CI from the 29 studies provided the OS of patients.

## Meta-analysis of DFS

Besides OS, all of the following survival data were taken into consideration: PFS, DFS, RFS, EFS, TFS, EFS, CSS, TTP, CFS. We found there were significant association between the survival data and the prognosis (HR = 1.99, 95% CI = 1.67, 2.36, P<0.00001, Figure 2B). Also, since the heterogeneity, we performed sub-analysis and found that except the plasma and urine samples shown no relationship between HR and survival, other sample types, regions, tumor types, HR resources, cut off values, independent factors have certain association with miR-155 expression Table 1. Considering the heterogeneity among the studies, random models were performed to HR and 95% CI.

#### Sensitivity analysis

To eliminate individual studies, sensitivity analysis was per-

formed. In order to explore the potential factors contribute to heterogeneity in OS and DFS, a meta-regression was conducted. The results indicated that there was no stu-dy responsible for *heterogeneity*. The pooled results analyses were shown in **Figure 3A** and **3B**.

## Publication bias

The funnel plots, Begg's analyses were shown in **Figure 3A** and **3B** formed. In order to explore the potential factors contribute to heterogeneity in OS and DFS, a meta-regression was conducted. The result P value shown no evidence of significant publication bias (Pr > |z| = 0.359 > 0.05) (**Figure 4A**). Unpublished studies may existed. Similarly, P-value of Begg's test for DFS

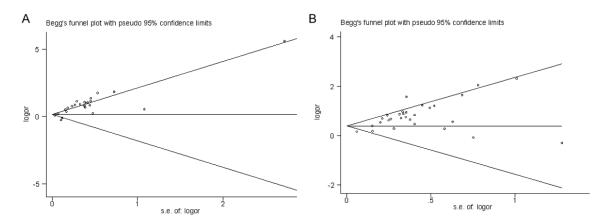


Figure 4. A. Begg's funnel plot of OS for publication bias test (Pr > |z| = 0.359 > 0.05); B. Begg's funnel plot of DFS for publication bias test (Pr > |z| = 0.103 > 0.05).

demonstrated that there was no publication bias (Pr > |z| = 0.103 > 0.05) (Figure 4B).

#### Discussion

Recently, more and more evidences demonstrated that miR-155 was widely expressed in various tumors, and its aberrant expression was reported in many tumors. These results have verified miR-155 as a potential biomarker for prognostic value. A research suggested that associated with lower RAD51 expression, high miR-155 levels indicated better overall survival of patients in breast cancers [44]. Other researches hold the opposite opinion, for example, Jiang S. found that overexpression of miR-155 in breast cancer cells can stimulate breast cancer cells [53]. Our analysis verified that high expression of miR-155 is correlated with poor cancer outcome.

The relation between miR-155 expression and the prognosis of patients with solid tumors was performed by Jing He [9]. However, the analysis is not comprehensive enough to collect all relevant evidence. Compared with former meta-analysis, this study contained more studies and patients, conducted more subgroup analyses and collected new evidences, thus providing powerful new evidence.

Admittedly, there were some inevitable limitations. Firstly, some of the HRs and Cls data were calculated based on the survival curves, despite the method has been previously validated, statistical errors were inevitable due to inaccurate readings. Secondly, tumor type, cutoff value, analysis type, miR-155 detection

method, country, follow-up time and publication year might also contribute to the heterogeneity. Thirdly, studies regarding various tumors without a consistent cut-off value may influence the ultimate results, Meanwhile, we might ignore some undiscovered factors like adjustment for surgery, radiation, chemotherapy, socioeconomic status, tumor characteristics, and so on, which might contribute to the heterogeneity. Finally, although no significant publication bias was detected in this meta-analysis, the funnel plots of the OS analysis were not so symmetric, the results still need to be verified by a large number of studies.

In conclusion, although this study was not the first meta-analysis designed to illustrate the relationship between miR-155 and prognosis, more robust evidence had been revealed through our detailed subgroup analysis, updated data and more studies.

#### Conclusions

Our study indicated that high expression of miR-155 was significantly related to the poor survival in several cancer patients, and thus could be an important predictor of poor prognosis.

## Acknowledgements

This study was supported by grants from National Natural Science Foundation of China (No. 71673193), the Key Technology Research and Development Program of the Sichuan Province (2015SZ0131 and 2017FZ0082).

#### Disclosure of conflict of interest

None.

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#### References

- [1] Carthew RW and Sontheimer EJ. Origins and Mechanisms of miRNAs and siRNAs. Cell 2009; 136: 642-55.
- [2] Yoshimura A, Numakawa T, Odaka H, Adachi N, Tamai Y and Kunugi H. Negative regulation of microRNA-132 in expression of synaptic proteins in neuronal differentiation of embryonic neural stem cells. Neurochem Int 2016; 97: 26-33.
- [3] Guan B, Li Q, Shen L, Rao Q, Wang Y, Zhu Y, Zhou XJ and Li XH. MicroRNA-205 directly targets Kruppel-like factor 12 and is involved in invasion and apoptosis in basal-like breast carcinoma. Int J Oncol 2016; 49: 720-34.
- [4] Jamali Z, Asl Aminabadi N, Attaran R, Pournagiazar F, Ghertasi Oskouei S and Ahmadpour F. MicroRNAs as prognostic molecular signatures in human head and neck squamous cell carcinoma: a systematic review and meta-analysis. Oral Oncol 2015; 51: 321-31.
- [5] Lu L, Mao X, Shi P, He B, Xu K, Zhang S and Wang J. MicroRNAs in the prognosis of triplenegative breast cancer: a systematic review and meta-analysis. Medicine (Baltimore) 2017; 96: e7085.
- [6] Wang F, Zhou J, Zhang Y, Wang Y, Cheng L, Bai Y and Ma H. The value of MicroRNA-155 as a prognostic factor for survival in non-small cell lung cancer: a meta-analysis. PLoS One 2015; 10: e0136889.
- [7] Gao Y, Ma X, Yao Y, Li H, Fan Y, Zhang Y, Zhao C, Wang L, Ma M, Lei Z and Zhang X. miR-155 regulates the proliferation and invasion of clear cell renal cell carcinoma cells by targeting E2F2. Oncotarget 2016; 7: 20324-37.
- [8] Zeng Q, Tao X, Huang F, Wu T, Wang J, Jiang X, Kuang Z and Cheng B. Overexpression of miR-155 promotes the proliferation and invasion of oral squamous carcinoma cells by regulating BCL6/cyclin D2. Int J Mol Med 2016; 37: 1274-80
- [9] He J, Zhang F, Wu Y, Zhang W, Zhu X, He X, Zhao Y, Zhang W and Zhao Y. Prognostic role of microRNA-155 in various carcinomas: results from a meta-analysis. Dis Markers 2013; 34: 379-86.

- [10] Saito M, Schetter AJ, Mollerup S, Kohno T, Skaug V, Bowman ED, Mathe EA, Takenoshita S, Yokota J, Haugen A and Harris CC. The association of microRNA expression with prognosis and progression in early-stage, non-small cell lung adenocarcinoma: a retrospective analysis of three cohorts. Clin Cancer Res 2011; 17: 1875-82.
- [11] Higgins JP and Thompson SG. Quantifying heterogeneity in a meta-analysis. Stat Med 2002; 21: 1539-58.
- [12] Begg CB and Mazumdar M. Operating characteristics of a rank correlation test for publication bias. Biometrics 1994; 50: 1088-101.
- [13] Lv ZC, Fan YS, Chen HB and Zhao DW. Investigation of microRNA-155 as a serum diagnostic and prognostic biomarker for colorectal cancer. Tumour Biol 2015; 36: 1619-25.
- [14] Shibuya H, Iinuma H, Shimada R, Horiuchi A and Watanabe T. Clinicopathological and prognostic value of microRNA-21 and microR-NA-155 in colorectal cancer. Oncology 2010; 79: 313-20.
- [15] Wang H and Men CP. Correlation of increased expression of MicroRNA-155 in bladder cancer and prognosis. Lab Med 2015; 46: 118-22.
- [16] Zhang X, Zhang Y, Liu X, Fang A, Wang J, Yang Y, Wang L, Du L and Wang C. Direct quantitative detection for cell-free miR-155 in urine: a potential role in diagnosis and prognosis for non-muscle invasive bladder cancer. Oncotarget 2016; 7: 3255-66.
- [17] Wotschofsky Z, Busch J, Jung M, Kempkensteffen C, Weikert S, Schaser KD, Melcher I, Kilic E, Miller K, Kristiansen G, Erbersdobler A and Jung K. Diagnostic and prognostic potential of differentially expressed miRNAs between metastatic and non-metastatic renal cell carcinoma at the time of nephrectomy. Clin Chim Acta 2013; 416: 5-10.
- [18] Shinmei S, Sakamoto N, Goto K, Sentani K, Anami K, Hayashi T, Teishima J, Matsubara A, Oue N, Kitadai Y and Yasui W. MicroRNA-155 is a predictive marker for survival in patients with clear cell renal cell carcinoma. Int J Urol 2013; 20: 468-77.
- [19] Merhautova J, Hezova R, Poprach A, Kovarikova A, Radova L, Svoboda M, Vyzula R, Demlova R and Slaby O. miR-155 and miR-484 are associated with time to progression in metastatic renal cell carcinoma treated with sunitinib. Biomed Res Int 2015; 2015: 941980.
- [20] Baba O, Hasegawa S, Nagai H, Uchida F, Yamatoji M, Kanno NI, Yamagata K, Sakai S, Yanagawa T and Bukawa H. MicroRNA-155-5p is associated with oral squamous cell carcinoma metastasis and poor prognosis. J Oral Pathol Med 2016; 45: 248-55.

- [21] Shi LJ, Zhang CY, Zhou ZT, Ma JY, Liu Y, Bao ZX and Jiang WW. MicroRNA-155 in oral squamous cell carcinoma: overexpression, localization, and prognostic potential. Head Neck 2015; 37: 970-6.
- [22] Bedewy AM, Elmaghraby SM, Shehata AA and Kandil NS. Prognostic value of miRNA-155 expression in B-cell Non-Hodgkin's lymphoma. Turk J Haematol 2017; 34: 207-12.
- [23] Iqbal J, Shen Y, Huang X, Liu Y, Wake L, Liu C, Deffenbacher K, Lachel CM, Wang C, Rohr J, Guo S, Smith LM, Wright G, Bhagavathi S, Dybkaer K, Fu K, Greiner TC, Vose JM, Jaffe E, Rimsza L, Rosenwald A, Ott G, Delabie J, Campo E, Braziel RM, Cook JR, Tubbs RR, Armitage JO, Weisenburger DD, Staudt LM, Gascoyne RD, McKeithan TW and Chan WC. Global microRNA expression profiling uncovers molecular markers for classification and prognosis in aggressive B-cell lymphoma. Blood 2015; 125: 1137-45.
- [24] Zhong H, Xu L, Zhong JH, Xiao F, Liu Q, Huang HH and Chen FY. Clinical and prognostic significance of miR-155 and miR-146a expression levels in formalin-fixed/paraffin-embedded tissue of patients with diffuse large B-cell lymphoma. Exp Ther Med 2012; 3: 763-70.
- [25] Ferrajoli A, Shanafelt TD, Ivan C, Shimizu M, Rabe KG, Nouraee N, Ikuo M, Ghosh AK, Lerner S, Rassenti LZ, Xiao L, Hu J, Reuben JM, Calin S, You MJ, Manning JT, Wierda WG, Estrov Z, O'Brien S, Kipps TJ, Keating MJ, Kay NE and Calin GA. Prognostic value of miR-155 in individuals with monoclonal B-cell lymphocytosis and patients with B chronic lymphocytic leukemia. Blood 2013; 122: 1891-9.
- [26] Ishihara K, Sasaki D, Tsuruda K, Inokuchi N, Nagai K, Hasegawa H, Yanagihara K and Kamihira S. Impact of miR-155 and miR-126 as novel biomarkers on the assessment of disease progression and prognosis in adult T-cell leukemia. Cancer Epidemiol 2012; 36: 560-5.
- [27] Marcucci G, Maharry KS, Metzeler KH, Volinia S, Wu YZ, Mrozek K, Nicolet D, Kohlschmidt J, Whitman SP, Mendler JH, Schwind S, Becker H, Eisfeld AK, Carroll AJ, Powell BL, Kolitz JE, Garzon R, Caligiuri MA, Stone RM and Bloomfield CD. Clinical role of microRNAs in cytogenetically normal acute myeloid leukemia: miR-155 upregulation independently identifies high-risk patients. J Clin Oncol 2013; 31: 2086-93.
- [28] Metzeler KH, Maharry K, Kohlschmidt J, Volinia S, Mrozek K, Becker H, Nicolet D, Whitman SP, Mendler JH, Schwind S, Eisfeld AK, Wu YZ, Powell BL, Carter TH, Wetzler M, Kolitz JE, Baer MR, Carroll AJ, Stone RM, Caligiuri MA, Marcucci G and Bloomfield CD. A stem cell-like gene expression signature associates with inferior outcomes and a distinct microRNA expression profile in adults with primary cytoge-

- netically normal acute myeloid leukemia. Leukemia 2013; 27: 2023-31.
- [29] Xu LH, Guo Y, Cen JN, Yan WY, He HL, Niu YN, Lin YX, Chen CS and Hu SY. Overexpressed miR-155 is associated with initial presentation and poor outcome in Chinese pediatric acute myeloid leukemia. Eur Rev Med Pharmacol Sci 2015; 19: 4841-50.
- [30] Zhou G, Cao Y, Dong W, Lin Y, Wang Q, Wu W, Hua X, Ling Y, Xie X, Hu S, Cen J and Gu W. The clinical characteristics and prognostic significance of AID, miR-181b, and miR-155 expression in adult patients with de novo B-cell acute lymphoblastic leukemia. Leuk Lymphoma 2017; 58: 1-9.
- [31] Cui B, Chen L, Zhang S, Mraz M, Fecteau JF, Yu J, Ghia EM, Zhang L, Bao L, Rassenti LZ, Messer K, Calin GA, Croce CM and Kipps TJ. Micro-RNA-155 influences B-cell receptor signaling and associates with aggressive disease in chronic lymphocytic leukemia. Blood 2014; 124: 546-54.
- [32] Gao Y, Fu S, Jiang W, Li B, Tian Y and Fu X. Association of MiR-155 expression with prognosis in resected stage III non-small cell lung cancer. Zhongguo Fei Ai Za Zhi 2014; 17: 417-23.
- [33] Raponi M, Dossey L, Jatkoe T, Wu X, Chen G, Fan H and Beer DG. MicroRNA classifiers for predicting prognosis of squamous cell lung cancer. Cancer Res 2009; 69: 5776-83.
- [34] Voortman J, Goto A, Mendiboure J, Sohn JJ, Schetter AJ, Saito M, Dunant A, Pham TC, Petrini I, Lee A, Khan MA, Hainaut P, Pignon JP, Brambilla E, Popper HH, Filipits M, Harris CC and Giaccone G. MicroRNA expression and clinical outcomes in patients treated with adjuvant chemotherapy after complete resection of non-small cell lung carcinoma. Cancer Res 2010; 70: 8288-98.
- [35] Xue X, Liu Y, Wang Y, Meng M, Wang K, Zang X, Zhao S, Sun X, Cui L, Pan L and Liu S. MiR-21 and MiR-155 promote non-small cell lung cancer progression by downregulating SOCS1, SOCS6, and PTEN. Oncotarget 2016; 7: 84508-19.
- [36] Donnem T, Eklo K, Berg T, Sorbye SW, Lonvik K, Al-Saad S, Al-Shibli K, Andersen S, Stenvold H, Bremnes RM and Busund LT. Prognostic impact of MiR-155 in non-small cell lung cancer evaluated by in situ hybridization. J Transl Med 2011; 9: 6.
- [37] Sanfiorenzo C, Ilie MI, Belaid A, Barlesi F, Mouroux J, Marquette CH, Brest P and Hofman P. Two panels of plasma microRNAs as non-invasive biomarkers for prediction of recurrence in resectable NSCLC. PLoS One 2013; 8: e54596.
- [38] Wang X, Zhang Y and Zhi X. Correlation between microRNA expression, clinicopathological characteristics, and prognosis in patients

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- with non-small cell lung cancer: a retrospective study. Thorac Cancer 2017; 8: 511-6.
- [39] Yanaihara N, Caplen N, Bowman E, Seike M, Kumamoto K, Yi M, Stephens RM, Okamoto A, Yokota J, Tanaka T, Calin GA, Liu CG, Croce CM and Harris CC. Unique microRNA molecular profiles in lung cancer diagnosis and prognosis. Cancer Cell 2006; 9: 189-98.
- [40] Wu X, Wang Y, Yu T, Nie E, Hu Q, Wu W, Zhi T, Jiang K, Wang X, Lu X, Li H, Liu N, Zhang J and You Y. Blocking MIR155HG/miR-155 axis inhibits mesenchymal transition in glioma. Neuro Oncol 2017; 19: 1195-205.
- [41] Kono H, Nakamura M, Ohtsuka T, Nagayoshi Y, Mori Y, Takahata S, Aishima S and Tanaka M. High expression of microRNA-155 is associated with the aggressive malignant behavior of gallbladder carcinoma. Oncol Rep 2013; 30: 17-24.
- [42] Han ZB, Chen HY, Fan JW, Wu JY, Tang HM and Peng ZH. Up-regulation of microRNA-155 promotes cancer cell invasion and predicts poor survival of hepatocellular carcinoma following liver transplantation. J Cancer Res Clin Oncol 2012; 138: 153-61.
- [43] Zhang L, Wang W, Li X, He S, Yao J, Wang X, Zhang D and Sun X. MicroRNA-155 promotes tumor growth of human hepatocellular carcinoma by targeting ARID2. Int J Oncol 2016; 48: 2425-34.
- [44] Gasparini P, Lovat F, Fassan M, Casadei L, Cascione L, Jacob NK, Carasi S, Palmieri D, Costinean S, Shapiro CL, Huebner K and Croce CM. Protective role of miR-155 in breast cancer through RAD51 targeting impairs homologous recombination after irradiation. Proc Natl Acad Sci U S A 2014; 111: 4536-41.
- [45] Kong W, He L, Richards EJ, Challa S, Xu CX, Permuth-Wey J, Lancaster JM, Coppola D, Sellers TA, Djeu JY and Cheng JQ. Upregulation of miRNA-155 promotes tumour angiogenesis by targeting VHL and is associated with poor prognosis and triple-negative breast cancer. Oncogene 2014; 33: 679-89.
- [46] Kapodistrias N, Mavridis K, Batistatou A, Gogou P, Karavasilis V, Sainis I, Briasoulis E and Scorilas A. Assessing the clinical value of microRNAs in formalin-fixed paraffin-embedded liposarcoma tissues: overexpressed miR-155 is an indicator of poor prognosis. Oncotarget 2017; 8: 6896-913.

- [47] Greither T, Grochola LF, Udelnow A, Lautenschlager C, Wurl P and Taubert H. Elevated expression of microRNAs 155, 203, 210 and 222 in pancreatic tumors is associated with poorer survival. Int J Cancer 2010; 126: 73-80.
- [48] Mikamori M, Yamada D, Eguchi H, Hasegawa S, Kishimoto T, Tomimaru Y, Asaoka T, Noda T, Wada H, Kawamoto K, Gotoh K, Takeda Y, Tanemura M, Mori M and Doki Y. MicroR-NA-155 controls exosome synthesis and promotes gemcitabine resistance in pancreatic ductal adenocarcinoma. Sci Rep 2017; 7: 42339.
- [49] Papaconstantinou IG, Manta A, Gazouli M, Lyberopoulou A, Lykoudis PM, Polymeneas G and Voros D. Expression of microRNAs in patients with pancreatic cancer and its prognostic significance. Pancreas 2013; 42: 67-71.
- [50] Ali S, Dubaybo H, Brand RE and Sarkar FH. Differential expression of MicroRNAs in tissues and plasma co-exists as a biomarker for pancreatic cancer. J Cancer Sci Ther 2015; 7: 336-46.
- [51] Osaka E, Kelly AD, Spentzos D, Choy E, Yang X, Shen JK, Yang P, Mankin HJ, Hornicek FJ and Duan Z. MicroRNA-155 expression is independently predictive of outcome in chordoma. Oncotarget 2015; 6: 9125-39.
- [52] Qiu S, Lin S, Hu D, Feng Y, Tan Y and Peng Y. Interactions of miR-323/miR-326/miR-329 and miR-130a/miR-155/miR-210 as prognostic indicators for clinical outcome of glioblastoma patients. J Transl Med 2013; 11: 10.
- [53] Jiang S, Zhang HW, Lu MH, He XH, Li Y, Gu H, Liu MF and Wang ED. MicroRNA-155 functions as an OncomiR in breast cancer by targeting the suppressor of cytokine signaling 1 gene. Cancer Res 2010; 70: 3119-27.