Original Article Clinical value and potential targets of miR-224-5p in hepatocellular carcinoma validated by a TCGA- and GEO- based study

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Abstract: Objective: to explore clinical value and potential targets of MicroRNA-224-5p in the tumorigenesis and progression of hepatocellular carcinoma (HCC). Methods: We evaluated the clinical value of MicroRNA-224-5p from The Cancer Genome Atlas (TCGA) and Gene Expression Omnibus (GEO). Meanwhile, target genes of MicroRNA-224-5p were predicted by bioinformatics method. The target genes of MicroRNA-224-5p were finally analyzed in Gene Ontology (GO) terms, Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway annotation, and Protein-Protein Interaction (PPI) network annotation. Results: MicroRNA-224-5p expression level in HCC was higher than in non-tumor tissues (SMD=1.24; 95% CI, 0.68 to 1.81; P<0.0001) and MicroRNA-224-5p might represent a diagnostic marker (overall AUC=0.92; 95% CI, 0.90 to 0.94). 262 target genes were acquired by overlapping 4927 genes predicted by more than four computational prediction tools with 1,123 down-regulated DEGs in HCC. Furthermore, gene sets enrichment analysis of the 262 overlapping genes was implemented. The mostly significant GO terms within the overlapping target genes of MicroRNA-224-5p were cellular response to chemical stimulus, plasma membrane part and coenzyme binding. KEGG pathway annotation showed the overlapping genes mostly took part in metabolic pathways. In PPI analysis, one hub gene, GNA14, stood out cause for the significant negative correlation with MicroRNA-224. Conclusion: MicroRNA-224-5p is upregulated in HCC and may be a prospective biomarker for diagnosis. Moreover, MicroRNA-224-5p might play an oncogenic role in HCC by targeting GNA14.

Keywords: Hepatocellular carcinoma, gene targeting, bioinformatics, microRNAs

Introduction

Hepatocellular carcinoma (HCC) is one of the most frequent malignant tumors worldwide. Many global studies have focused on the diagnosis and treatment of HCC. Nevertheless, its prognosis is still poor owing to a series of reasons, such as growing incidence, late diagnosis, persistent drug resistance and repeated recurrence. Thus, it is urgent to discover novel and valuable diagnostic and prognostic biomarkers for HCC [1-6].

MicroRNAs (miRNAs) are classified as endogenous ~22 nucleotides RNAs, suppressing mRNA-protein translation or promoting mRNA degradation. The miRNAs have vital characters in various biological processes, such as cell survival, cell apoptosis, cell metastasis, cell proliferation, cell migration and cell invasion [7-13]. MicroRNA-224 (miR-224) has been found to participate in many biological processes of many malignancies, such as ovarian cancer [14], cervical cancer [15], breast cancer [16], gastric cancer [17], osteosarcoma [18], colorectal cancer [19], lung cancer [20], diffuse large B-cell lymphoma [21], esophageal squamous cell carcinoma [22], meningioma [23], glioma [24], prostate cancer [25], and renal cancer [26]. Meanwhile, several studies demonstrated that miR-224 has correlations with carcinogenesis and disease progression of HCC. For example, Scisciani et al identified the mature miR-224 levels significantly upregulated in human HCCs by RT-PCR [27]. Li et al found miR-224 participated in a series of signaling pathways and contributed to the regulation of HCC invasion and migration [28]. However, reli-



Figure 1. General flow chart presenting the design conception in this study.



pathways of miR-224 in HCC is necessary.

We acquired the expression and clinicopathological characteristics data of miR-224-5p from The Cancer Genome Atlas (TCGA) and Gene Expression Omnibus (GEO), analyzing the expression and clinicopathological characteristics of miR-224-5p in HCC, and also performing the Receiver Operating Characteristic (ROC) curves of each dataset. Then, we implemented a comprehensive meta-analysis to assess the expression tendency and the diagnostic value of miR-224-5p. Next, we predicted the target genes of miR-224-5p through 12 miRNA computational prediction tools, and then overlapped with the down-regulated differentially expressed genes (DEGs) in HCC based on TCGA data. Finally, Gene Ontology (GO) terms annotation, Kyoto Encyclopedia of Genes and Genomes (KEpathway annotation GG) and Protein-Protein Interaction (PPI) network annotation were commenced to discover the function of the overlapping genes. Overall, the objective of this study was to evaluate clinical value and potential targets of miR-224-5p in the tumorigenesis and progression of HCC, providing advices for subsequent study of miR-224-5p in HCC (Figure 1).

Materials and methods

TCGA RNA-seq data in HCC patients

ability of previous studies that evaluating miR-224 roles in HCC was limited by small sample sizes to some extent. Therefore, a comprehensive analysis of clinical value, target genes and The expression and clinicopathological characteristics data of miR-224 were acquired from TCGA (http://cancergenome.nih.gov/; 2016), including 364 HCC patients and 50 normal

		miR-224 expression	
Clinicopathological feature	n	(Mean ± SD)	Р
Tissue		· · · ·	
HCC	364	8.88±2.04	
Normal Control	50	6.49±1.40	<0.001
Gender			
Male	247	8.80±2.10	
Female	117	9.06±1.89	0.257
Age (years)			
<60	167	8.79±2.04	
≥60	196	8.96±2.04	0.423
Cirrhosis			
Yes	6	7.57±1.80	
No	340	8.93±2.03	0.105
Grade			
I	55	8.31±2.02	
II	173	9.14±1.95	
III	120	8.79±2.14	
IV	12	9.05±1.72	0.058
Grade			
-	228	8.94±1.99	
III-IV	132	8.81±2.10	0.58
Stage			
	87	8.68±1.91	
II	46	8.46±2.18	
III	36	8.84±2.47	
IV	1	9.88	0.796
Stage			
I-II	133	8.60±2.00	
III-IV	37	8.87±2.44	0.499
Vascular			
Yes	52	8.88±2.31	
No	105	8.61±2.03	0.444
Alcohol			
Yes	84	8.90±2.06	
No	280	8.88±2.03	0.934
Smoking			
Yes	17	8.33±2.60	
No	347	8.91±2.00	0.251
HBV infection			
Yes	81	9.04±1.89	
No	283	8.84±2.08	0.421
HCV infection			
Yes	56	9.24±1.92	
No	308	8.82±2.05	0.155

Table 1. Relationship between the expression of miR-224
and clinicopathological characteristics in HCC

samples. Down-regulated DEGs between the expression profiles of 364 HCC tissues and 50

non-tumor tissues were obtained by performing limma package of R (https:// www.r-project.org/). When log2 fold change (FC) <-1 with the *P* value <0.05 in Student's t-test, gene down-regulated expression difference would be considered as statistically significant.

GEO microarray datasets acquisition

The available data on miR-224-5p expression were obtained from GEO (http://www.ncbi.nlm.nih.gov/geo/). The search strategy that used in GEO datasets was as follows: (hepatocellular OR liver OR hepatic OR HCC) AND (miRNA OR microRNA). The inclusion criteria of the included microarray datasets were: (1) samples within the cancer group were diagnosed as HCC; (2) each microarray included both HCC group and noncancerous group; (3) expression profiling data of miR-224-5p were provided; (4) the species in the microarrays were humans. While the exclusion criteria included: (1) microarrays without available data of miR-224-5p; (2) microarrays without normal controls; (3) the species in the microarrays were animals.

Potential target genes of miR-224-5p

We applied 12 computational prediction tools, including miRWalk, Microt4, miRanda, mirbridge, miRDB, miRMap, miRNAMap, Pictar2, PITA, RNA22, RNAhybrid, and Targetscan, to get predicted target genes of miR-224-5p. Only genes that were simultaneously predicted by four or more than four computational prediction tools were selected. Subsequently, we acquired the extracted genes by overlapping the predicted target genes and the down-regulated DEGs in HCC.

Functional and pathway enrichment analysis

The Database for Annotation, Visualization and Integrated Discovery (DAVID; https://david.ncifcrf.gov/) is a public database providing functional

and pathway enrichment analysis of massive gene lists. We used DAVID to perform GO



Figure 3. Flow chart of study selection for microarray review based on GEO datasets.

enrichment analysis and KEGG pathway enrichment analysis of the extracted target genes of miR-224-5p. Only a *P*-value <0.05 of the terms were selected for analysis. The GO functional network graphs, including molecular function (MF), cellular component (CC), and biological process (BP), was further visualized with Cytoscape 3.2.1 (http://cytoscape.org/). The KEGG enrichment result was visualized by ggplot2 package of R (https://www.r-project. org/).

PPI network construction

Search Tool for the Retrieval of Interacting Genes/Proteins (STRING; http://string-db.org/) is a biological database and web resource of known and predicted PPI. The latest version 10.0 was used to construct PPI network of proteins encoded by the overlapping miR-224-5p target genes. The PPI data were downloaded from STRING database and a confidence score >0.7 were selected as inclusion criteria. At PPI network analysis, the nodes represent proteins, and the lines represent the interactions of proteins. When a node was associated with more than 5 other nodes within the inclusion criteria, it was defined as a hub gene in our study. We downloaded data from STRING and analyzed the association between all nodes in PPI network and got the hub genes.

Statistical analysis

Student's T test which calculated the significant difference between HCC and noncancerous tissues and the scatter plots that exhibited the expression level of miR-224-5p in HCC were both conducted by GraphPad Prism Version 5.0 (GraphPad Software, San Diego CA, USA; https://www.graphpad.com). The Kaplan-Meier survival curve of miR-224 in HCC was also performed by GraphPad Prism Version 5.0.

Standard mean differences (SMDs) with 95% confidence interval (CI) were pooled to estimate the expression level of miR-224-5p in HCC. Heterogeneity in the meta-analysis were assessed by chi-square (x²) test of Q and inconsistency index (I^2), and a P value < 0.05 or I^2 >50% was considered to be significantly heterogeneous and a random-effects model was applied. If a P value >0.05 or l^2 <50%, the fixedeffects model was selected. Forest plots of miR-224-5p expression were conducted using Review Manager Version 5.3 (The Nordic Cochrane Centre, Copenhagen, Denmark; http:// community.cochrane.org/tools/review-production-tools/revman-5). We performed a sensitivity analysis to evaluate the stability of the results through deleting studies one by one and comparing the pooled results of different models. Publication bias was assessed by using a funnel plot with Begg's bias tests via Stata Version 12.0 (StataCorp, College Station, TX, USA; http://www.stata.com). A P value <0.05 suggested the presence of publication bias.

To explore diagnostic value of miR-224-5p in HCC, we performed a series of ROC curves. The standards for assessing the Area Under the Curve (AUC) in ROC were as follows: 0.5-0.7 (poor evidence for diagnosis); 0.7-0.9 (moderate evidence for diagnosis); 0.9-1.0 (high evidence for diagnosis). The ROC curves were drawn by MedCalc Version 9.2.0.1 (MedCalc Software, Ostend, Belgium; https://www.medcalc.org) and Stata Version 12.0 was used for SROC curve analysis.

Results

Expression and clinicopathological characteristics of miR-224 in HCC based on TCGA data

The expression level of miR-224 in HCC specimens (8.88 ± 2.04) was higher than in nontumor tissues (6.49 ± 1.40), and the difference was statistically significant (P<0.001) (**Figure 2A**). While no significant differences were found between miR-224 expression and all the tested clinicopathological characteristics (**Table 1**). To

Dataset	Platform	Country	Sample source	Citation	Samp	le sizes	miR-224-5ı (Mear	o expression n ± SD)
					HCC group	Non-tumor	HCC group	Non-tumor
						group		group
GSE6857	GPL4700	USA	Tissue	Budhu A et al (2008)	238	241	9.83±0.86	10.01±0.76
GSE10694	GPL6542	China	Tissue	Li W et al (2008)	78	88	11.59±0.97	10.68±0.14
GSE12717	GPL7274	USA	Tissue	Su H et al (2009)	10	6	9.55±2.18	4.78±1.40
GSE21362	GPL10312	Japan	Tissue	Sato F et al (2011)	73	73	7.10±1.85	5.87±1.39
GSE22058	GPL6793	USA	Tissue	Burchard J et al (2010)	96	96	0.66±0.67	-0.01±0.36
	GPL9733							
	GPL10457							
GSE31383	GPL10122	USA	Tissue	Wang PR et al (2012)	9	10	5.98±2.40	-0.90±1.73
GSE41874	GPL7722	Japan	Tissue	Missing	3	3	1.26±0.26	1.02±0.28
GSE54751	GPL18262	USA	Tissue	Shen J et al (2015)	10	10	0.03±0.02	0.01±0.01
GSE57555	GPL16699	Japan	Tissue	Murakami Y et al (2015)	5	16	0.05±0.08	-0.04±0.02
	GPL18044							
GSE69580	GPL10850	Taiwan	Tissue	Missing	3	5	2.45±1.75	0.80±1.31

Table 2. Characteristics of the ten included GEO-based microarrays in our study

assess the diagnostic value of miR-224, we conducted the ROC curve analysis based on the expression profile of miR-224 in HCC tissues compared with non-tumor tissues. The AUC of miR-224 was 0.828 (95% Cl: 0.789-0.864), with a sensitivity and specificity of 61.3% and 98.0%, respectively (**Figure 2B**). Kaplan-Meier survival curve was constructed to evaluate prognostic value of miR-224 in HCC. However, no significant prognostic value of miR-224 in HCC was found (HR=1.007, 95% Cl: 0.7121-1.423; P=0.9693) (**Figure 2C**).

Expression of miR-224-5p in HCC based on GEO datasets

In total, ten eligible microarrays, including GSE6857, GSE10694, GSE12717, GSE21362, GSE22058, GSE31383, GSE41874, GSE547-51, GSE57555 and GSE69580 were involved in our study. The flow chart of study selection for microarray review was presented in Figure 3. While the characteristics of the ten included GEO-based microarrays were shown in Table 2. Six microarrays (GSE10694, GSE12717, GSE-21362. GSE22058. GSE31383. and GSE57-555) (Figure 4B-F, 4I) displayed that miR-224-5p expression levels were significantly upregulated in HCC in comparison to noncancerous tissues (P<0.001). Nevertheless, one microarrays (GSE6857) (Figure 4A) demonstrated that the miR-224-5p expression was significantly downregulated in HCC than in noncancerous tissues (P<0.05). There were still three microarrays (GSE41874, GSE54751, and GSE69580) (Figure 4G, 4H, 4J) revealed that miR-224-5p expression levels had no statistically significant difference in HCC and non-tumor tissues (P>0.05). In the diagnostic value of miR-224-5p in HCC analysis, eight microarrays (GSE6857, GSE10694, GSE12717, GSE21362, GSE220-58, GSE31383, GSE57555, GSE69580) (Figure 5A-F, 5I, 5J) showed the significant diagnostic value of miR-224-5p in HCC (P<0.05). While two microarrays (GSE41874, GSE54751) (Figure 5G, 5H) exhibited no significant diagnostic value of miR-224-5p in HCC (P>0.05).

Comprehensive meta-analysis

We performed a comprehensive meta-analysis of miR-224-5p expression based on the data gathered from TCGA and all GEO microarray datasets, including 889 HCC samples and 598 non-tumor liver tissues. According to the heterogeneity test, significant heterogeneity occurred among individual datasets of the expression of miR-224-5p in HCC (P_{heterogeneity} < 0.00001, I²=94%; Figure 6A) and the random-effects model was selected. Significant difference of miR-224-5p expression between HCC group and noncancerous group was observed (SMD=1.24, 95% CI: 0.68-1.81; P<0.0001), which confirmed the expression level of miR-224-5p in HCC tissues was higher than non-tumor tissues. In the sensitivity analysis, after removing the study of GSE31383 and GSE6857, the SMD was statistically changed to 1.17 (95%



CI: 1.02-1.33; P<0.00001) and no significant heterogeneity was observed ($P_{heterogeneity}$ =0.12, l^2 = 37%; **Figure 6B**). The funnel plot was nearly symmetric and the *P* value was greater than

0.05 (Begg's P=0.64; **Figure 7**), suggested that no publication bias existed. To comprehensively evaluate the diagnostic value of miR-224-5p in HCC, we constructed a summary receiver



operating characteristic (SROC) curve and calculated the AUC with its 95% Cl. As shown in **Figure 8**, the overall AUC of miR-224-5p in HCC was 0.92 (95% CI: 0.90-0.94), and the sensitivity and specificity were 0.92 and 0.78, respectively.

MicroRNA-224-5p in hepatocellular carcinoma

A	Expe	rimen	tal	C	ontrol		Std. Mean Difference		Std. Mean Difference
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% Cl	IV, Random, 95% Cl
GSE10694	11.59	0.97	78	10.68	0.14	88	11.5%	1.35 [1.01, 1.69]	
GSE12717	9.55	2.18	10	4.78	1.4	6	7.0%	2.33 [0.95, 3.70]	
GSE21362	7.1	1.85	73	5.87	1.39	73	11.5%	0.75 [0.41, 1.08]	
GSE22058	0.66	0.67	96	-0.01	0.36	96	11.6%	1.24 [0.93, 1.55]	
GSE31383	5.98	2.4	9	-0.9	1.73	10	6.7%	3.17 [1.72, 4.62]	
GSE41874	1.26	0.26	3	1.02	0.28	3	5.6%	0.71 [-1.03, 2.45]	
GSE54751	0.03	0.02	10	0.01	0.01	10	8.9%	1.21 [0.24, 2.18]	
GSE57555	0.05	0.08	5	-0.04	0.02	16	7.7%	2.12 [0.89, 3.35]	
GSE6857	9.83	0.86	238	10.01	0.76	241	11.8%	-0.22 [-0.40, -0.04]	+
GSE69580	2.45	1.75	3	0.8	1.31	5	6.2%	0.98 [-0.61, 2.56]	
TCGA	8.88	2.04	364	6.49	1.4	50	11.6%	1.21 [0.90, 1.52]	-
Total (95% CI)			889			598	100.0%	1.24 [0.68, 1.81]	•
Heterogeneity: Tau ² =	0.70; CI	hi ² = 15	57.15, 0	if = 10 (P < 0.0	00001);	I ² = 94%		
Test for overall effect:	Z = 4.30	(P < 0	.0001)						-4 -2 U 2 4
В	Expe	rimen	tal	C	ontrol			Std. Mean Difference	Std. Mean Difference
B _Study or Subgroup	Expe Mean	rimen SD	tal Total	Co Mean	ontrol SD	Total	Weight	Std. Mean Difference IV, Fixed, 95% Cl	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694	Expe Mean 11.59	rimen SD 0.97	tal <u>Total</u> 78	Co Mean 10.68	ontrol SD 0.14	Total 88	Weight 21.0%	Std. Mean Difference IV, Fixed, 95% Cl 1.35 [1.01, 1.69]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717	Expe <u>Mean</u> 11.59 9.55	rimen <u>SD</u> 0.97 2.18	tal <u>Total</u> 78 10	Co Mean 10.68 4.78	0ntrol SD 0.14 1.4	<u>Total</u> 88 6	Weight 21.0% 1.3%	Std. Mean Difference IV, Fixed, 95% CI 1.35 [1.01, 1.69] 2.33 [0.95, 3.70]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362	Expe Mean 11.59 9.55 7.1	rimen SD 0.97 2.18 1.85	tal <u>Total</u> 78 10 73	Co Mean 10.68 4.78 5.87	0.14 1.4 1.39	Total 88 6 73	Weight 21.0% 1.3% 21.3%	Std. Mean Difference IV, Fixed, 95% Cl 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08]	Std. Mean Difference IV. Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE22058	Expe <u>Mean</u> 11.59 9.55 7.1 0.66	rimen <u>SD</u> 0.97 2.18 1.85 0.67	tal <u>Total</u> 78 10 73 96	Co Mean 10.68 4.78 5.87 -0.01	0.14 1.4 1.39 0.36	Total 88 6 73 96	Weight 21.0% 1.3% 21.3% 25.1%	Std. Mean Difference <u>IV, Fixed, 95% C1</u> 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE22058 GSE41874	Expe Mean 11.59 9.55 7.1 0.66 1.26	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26	tal Total 78 10 73 96 3	Co Mean 10.68 4.78 5.87 -0.01 1.02	0.14 1.4 1.39 0.36 0.28	Total 88 6 73 96 3	Weight 21.0% 1.3% 21.3% 25.1% 0.8%	Std. Mean Difference <u>IV, Fixed, 95% Cl</u> 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> OSE10694 GSE12717 GSE21362 GSE22058 GSE41874 GSE54751	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02	tal 78 10 73 96 3 10	Contraction Contractico Contra	0.14 0.14 1.4 1.39 0.36 0.28 0.01	Total 88 6 73 96 3 10	Weight 21.0% 1.3% 21.3% 25.1% 0.8% 2.5%	Std. Mean Difference <u>IV, Fixed, 95% Cl</u> 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE22058 GSE41874 GSE54751 GSE57555	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03 0.05	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02 0.08	tal <u>Total</u> 78 10 73 96 3 10 5	Contemporation Contemporatica Contemporatica Contemporatica Contemporatica Contem	0.14 1.4 1.39 0.36 0.28 0.01 0.02	Total 88 6 73 96 3 10 16	Weight 21.0% 1.3% 21.3% 25.1% 0.8% 2.5% 1.6%	Std. Mean Difference IV, Fixed, 95% C1 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18] 2.12 [0.89, 3.35]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE21362 GSE41874 GSE54751 GSE57555 GSE69580	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03 0.05 2.45	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02 0.08 1.75	tal <u>Total</u> 78 10 73 96 3 10 5 3	Co 10.68 4.78 5.87 -0.01 1.02 0.01 -0.04 0.8	0.14 1.4 1.39 0.36 0.28 0.01 0.02 1.31	Total 88 6 73 96 3 10 16 5	Weight 21.0% 1.3% 21.3% 25.1% 0.8% 2.5% 1.6% 1.0%	Std. Mean Difference IV, Fixed, 95% C1 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18] 2.12 [0.89, 3.35] 0.98 [-0.61, 2.56]	Std. Mean Difference IV. Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE21362 GSE41874 GSE54751 GSE57555 GSE69580 TCGA	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03 0.05 2.45 8.88	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02 0.08 1.75 2.04	tal 78 10 73 96 3 10 5 3 364	Co Mean 10.68 4.78 5.87 -0.01 1.02 0.01 -0.04 0.8 6.49	0.14 1.4 1.39 0.36 0.28 0.01 0.02 1.31 1.4	Total 88 6 73 96 3 10 16 5 50	Weight 21.0% 1.3% 21.3% 25.1% 0.8% 2.5% 1.6% 1.0% 25.5%	Std. Mean Difference IV, Fixed, 95% C1 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18] 2.12 [0.89, 3.55] 0.98 [-0.61, 2.56] 1.21 [0.90, 1.52]	Std. Mean Difference IV. Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE22058 GSE41874 GSE54751 GSE57555 GSE69580 TCGA Total (95% CI)	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03 0.05 2.45 8.88	stimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02 0.08 1.75 2.04	tal 78 10 73 96 3 10 5 3 364 642	Control Contro	0.14 1.4 1.39 0.36 0.28 0.01 0.02 1.31 1.4	Total 88 6 73 96 3 10 16 5 50 347	Weight 21.0% 1.3% 25.1% 0.8% 2.5% 1.6% 1.0% 25.5% 100.0%	Std. Mean Difference IV, Fixed, 95% Cl 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18] 2.12 [0.89, 3.35] 0.98 [-0.61, 2.56] 1.21 [0.90, 1.52] 1.77 [1.02, 1.33]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE22058 GSE41874 GSE54751 GSE57555 GSE69580 TCGA Total (95% CI) Heterogeneity: Chi ² =	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03 0.05 2.45 8.88 12.73, d	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02 0.08 1.75 2.04 f = 8 (F	tal 78 10 73 96 3 10 5 3 64 642 2 = 0.12	Co Mean 10.68 4.78 5.87 -0.01 1.02 0.01 -0.04 0.8 6.49 2); I ² = 3	0.14 1.4 1.39 0.36 0.28 0.01 0.02 1.31 1.4	Total 88 6 73 96 3 10 16 5 50 347	Weight 21.0% 1.3% 25.1% 0.8% 2.5% 1.6% 1.0% 25.5% 100.0%	Std. Mean Difference IV, Fixed, 95% C1 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18] 2.12 [0.89, 3.35] 0.98 [-0.61, 2.56] 1.21 [0.90, 1.52] 1.17 [1.02, 1.33]	Std. Mean Difference IV, Fixed, 95% Cl
B <u>Study or Subgroup</u> GSE10694 GSE12717 GSE21362 GSE21362 GSE41874 GSE54751 GSE57555 GSE69580 TCGA Total (95% CI) Heterogeneity: Chi ² = Test for overall effect:	Expe Mean 11.59 9.55 7.1 0.66 1.26 0.03 0.05 2.45 8.88 12.73, d Z = 14.8	rimen <u>SD</u> 0.97 2.18 1.85 0.67 0.26 0.02 0.08 1.75 2.04 f = 8 (F 0 (P <	tal 78 10 73 96 3 10 5 3 364 642 2 = 0.12 0.0000	Co Mean 10.68 4.78 5.87 -0.01 1.02 0.01 -0.04 0.8 6.49 2); I ² = 3 1)	0.14 1.4 1.39 0.36 0.28 0.01 0.02 1.31 1.4	Total 88 6 73 96 3 10 16 5 50 347	Weight 21.0% 1.3% 21.3% 25.1% 0.8% 2.5% 1.6% 1.0% 25.5% 100.0%	Std. Mean Difference IV, Fixed, 95% C1 1.35 [1.01, 1.69] 2.33 [0.95, 3.70] 0.75 [0.41, 1.08] 1.24 [0.93, 1.55] 0.71 [-1.03, 2.45] 1.21 [0.24, 2.18] 2.12 [0.89, 3.35] 0.98 [-0.61, 2.56] 1.21 [0.90, 1.52] 1.17 [1.02, 1.33]	Std. Mean Difference IV. Fixed, 95% Cl

Figure 6. Forest plot of studies evaluating SMD of miR-224-5p expression between HCC group and non-tumor group (A) and the Forest plot after sensitivity analysis (B).



Figure 7. Funnel plot for publication bias test on all included studies.

Potential target genes of miR-224-5p in HCC

We used 12 computational prediction tools to predict potential target genes of miR-224-5p, and 14,407 target genes were acquired once in total. The number of genes that were simultaneously predicted by four or more than four computational prediction tools was 4,927. In addition, 1,123 down-regulated DEGs in HCC were acquired from TCGA. We selected the genes that both in the above 4927 predicted genes and 1,123 down-regulated DEGs in HCC, and gained 262 overlapping genes for further bioinformatics analysis, as shown in **Table 3**.

Functional and pathway enrichment analysis

To explore the functions of the 262 overlapping genes signatures, we performed GO annotation analysis by

DAVID. Through GO annotation analysis, the overlapping genes were classified into 3 GO categories, including BP, CC and MF. In BP, the top three processes which overlapping genes significantly participated in were cellular response to chemical stimulus, single-organism metabolic process, and response to organ-



ic substance. Meanwhile, in CC, plasma membrane part, cell periphery, and integral component of plasma membrane were considered the top three. In MF, the top three significant items were coenzyme binding, growth factor binding, and cofactor binding.

KEGG pathway enrichment analysis showed the top three significant enrichments of the overlapping genes were complement and coagulation cascades, metabolic pathways, and tryptophan metabolism. The top ten GO annotations and KEGG pathways that listed according to the *P* values were shown in **Table 4**. Three visualized GO network graphs (BP, CC, and MF) by Cytoscape 3.2.1 were presented in **Figures 9-11**, and the visualized KEGG enrichment result by ggplot2 package of R was rendered in **Figure 12**.

PPI network analysis

To analyze proteins of the 262 overlapping genes whether interrelated on each other, the PPI network was constructed by STRING (Figure 13). In the PPI network analysis, five genes, including G protein subunit alpha 14 (GNA14), insulin like growth factor 1 (IGF1), early growth response 1 (EGR1), aldehyde oxidase 1 (AOX1), and G protein subunit alpha o 1 (GNA01), were identified as hub genes. Subsequently, we performed Spearman correlation between the hub genes expression and miR-224 expression based on TCGA data. A significant negative correlation was found between GNA14 and miR-224 (r=-0.1146, P=0.0277) (Figure 14A), indi-

cating GNA14 have strong potential to be the target gene of miR-224. However, a significant positive correlation was observed between IGF1 and miR-224 (r= 0.1342, P=0.0099) (Figure 14B), and no significant correlation was discovered between EGR1, AOX1, GNA01 and miR-224 (P=0.2298, P=0.0919, and P=0.6387) (Figure 14C-E).

Discussion

In this study, we analyzed the expression and clinicopathological characteristics of

miR-224 in HCC based on TCGA RNA-seq data and found the up-regulation of miR-224 in HCC. Meanwhile, miR-224 might represent a diagnostic marker according to its AUC in ROC curve (AUC=0.828, 95% CI: 0.789-0.864). Then, we verified the conclusions by aggregating the GEO microarray datasets and then analyzed the expression and diagnostic value of miR-224-5p in HCC. We found miR-224-5p expression was higher than non-tumor tissues and miR-224-5p had a good diagnostic value in most of the microarrays. Afterwards, we performed a comprehensive meta-analysis of miR-224-5p expression based on the data gathered from TCGA and all GEO microarray datasets, and finally confirmed the up-regulation of miR-224-5p in HCC. Moreover, the overall AUC of miR-224-5p in HCC by the comprehensive metaanalysis could become a strong evidence to approve the conclusion that miR-224-5p might represent a diagnostic marker.

In recent studies, expression of miR-224-5p has been commonly identified as upregulated in several human cancer types, such as ovarian cancer [14], cervical cancer [15], colorectal cancer [29], lung cancer [20] and glioma [24]. However, there were reports about low expression of miR-224-5p in diffuse large B cell lymphoma [30], prostate cancer [31] and meningioma [32]. In female genital cancer, Hu et al investigated that miR-224 could affect the proliferation abilities of human epithelial ovarian cancer cells by KLLN-cyclin A pathway [14], while Huang et al found that miR-224 pro-

ZNF330	ST3GAL6	SLC39A14	RASD1	OGDHL	MFAP3L	IRF8	GLUD1	EPM2A	CYP1A2	CBFA2T3	ANKRD55
ZG16	SRGN	SLC38A4	RANBP3L	NUDT10	MEGF10	INHBC	GLS2	EPHA2	CXCR2	CAT	ANGPTL3
ZFP36L1	SRD5A1	SLC38A2	PTPN3	NR4A3	MCL1	IL6ST	GCH1	EPB41L4B	CTNNA3	CAMK4	AMOTL2
ZFP1	SPTBN2	SLC35D1	PTGIS	NR4A1	MBNL2	IL1RN	GABARAPL1	ENPEP	CR1	CALN1	AMHR2
YPEL2	SPSB1	SLC31A1	PSAT1	NR3C2	MBL2	IL1RAP	FRMD4B	EIF4E3	CPN2	CA2	ALDOB
XDH	SPRYD4	SIK1	PRRG4	NPY1R	MASP1	IGFBP4	FREM2	EGR2	CPEB3	C8A	ALDH8A1
WNT2	SPG20	SIGLEC1	PROS1	NKX3-1	MAN1A1	IGFBP3	FOXO1	EGR1	COL6A6	C6	ALDH4A1
WEE1	SPATA18	SGMS2	PRKCB	NDST3	MAFF	IGF1	FOSB	EFNB3	COBLL1	C21orf91	AKR1D1
WDR72	SOWAHC	SFRP1	PRKAR2B	NCOR1	LRRC55	HSD17B13	FNIP2	EDNRB	CMTM6	C21orf33	AGBL4
USH2A	SORL1	SERPINE1	PPP1R3B	NAMPT	LRAT	HPGD	FLRT3	DUSP14	CMPK2	C1RL	ADRB2
TSPAN7	SORD	SERPINB9	PLCXD3	NADK2	LPIN2	HOGA1	FHL2	DUSP1	CMBL	BMPER	ADRA1B
TRIB1	SOCS6	SDC4	PLA2G16	NAALADL2	LIFR	HMGCLL1	FGB	DPF3	CLEC12A	BACH2	ADH6
TMPRSS2	SOCS2	SAMD5	PIM1	N4BP2L1	LARP1B	HHIP	FAT4	DNASE1L3	CISH	AVPR1A	ADH1B
TMEM47	SLCO4C1	RUNDC3B	PHLDA1	MY01B	KLRF1	HABP2	FAM65C	DNAJC12	CHST7	ATOH8	ACVR1C
TMEM26	SLC9A9	RSAD2	PELI2	MTTP	KLHL15	GRIN2B	FAM46C	DIRAS3	CHRM2	ATF3	ACSL1
TGFBR3	SLC8A1	RNF152	PDGFRA	MSRA	KLF11	GREM2	FAM180A	DGAT2	CDH19	ASS1	ACAT1
TEK	SLC7A2	RNF125	PDE7B	MS4A6A	KLF10	GRAMD1C	FAM13A	DCUN1D3	CD69	ASPG	ACADSB
TCF21	SLC6A12	RND3	PCDH9	MRO	KDM8	GPR180	FAM110C	DCN	CD302	ARSD	ABLIM3
TAPT1	SLC5A1	RHOB	PANK1	MPPED1	KBTBD11	GPM6A	EXPH5	DBT	CD160	ARL5B	ABCC9
SYT10	SLC4A4	RGAG4	PAIP2B	MPEG1	JDP2	GNE	ESR1	CYP3A4	CCR1	AR	AADAT
SYNP02	SLC46A3	RD3L	OLFML3	MPDZ	IYD	GNA01	ERRFI1	CYP2J2	CCDC71L	APOL6	
STX11	SLC41A2	RBMS3	OIT3	MMAA	IVD	GNA14	ERLIN1	CYP2C18	CCDC25	AOX1	

Table 3. Overlapping target genes of miR-224-5p in HCC

 Table 4. Significant terms of GO annotation (includes biological process, cellular component and molecular function) and KEGG pathway enrichment analysis of miR-224-5p overlapping target genes

ID	Term	Count (%)	P-value	Genes
Biological Process				
G0:0070887	Cellular response to chemical stimulus	78 (29.8)	2.46E-11	CYP3A4, NAMPT, CYP2J2, CYP2C18, CPEB3, IL6ST, FOXO1, CXCR2, CMBL, WNT2, etc.
G0:0044710	Single-organism metabolic process	102 (38.9)	4.62E-11	CYP3A4, ALDH8A1, SGMS2, IL6ST, ADH1B, FOXO1, ENPEP, PRKAR2B, PTGIS, GRIN2B, etc.
G0:0010033	Response to organic substance	80 (30.5)	8.04E-11	NAMPT, CPEB3, IL6ST, F0X01, CXCR2, ACVR1C, WNT2, EDNRB, PRKAR2B, PTGIS, etc.
G0:0044281	Small molecule metabolic process	63 (24.0)	1.35E-10	ALDH8A1, CYP3A4, NAMPT, CYP2J2, CYP2C18, OGDHL, SORL1, ADH1B, FOXO1, CBFA2T3, etc.
G0:0019752	Carboxylic acid metabolic process	38 (14.5)	1.95E-10	ALDH8A1, ACADSB, SORD, CYP2J2, ASS1, CYP2C18, GNE, GLUD1, OGDHL, ALDOB, etc.
G0:0043436	Oxoacid metabolic process	38 (14.5)	2.31E-10	ALDH8A1, ACADSB, SORD, CYP2J2, ASS1, CYP2C18, GNE, GLUD1, OGDHL, ALDOB, etc.
G0:0071310	Cellular response to organic substance	65 (24.8)	1.83E-09	NAMPT, CPEB3, IL6ST, F0X01, CXCR2, WNT2, EDNRB, PRKAR2B, PTGIS, ST3GAL6, etc.
G0:0006082	Organic acid metabolic process	38 (14.5)	2.58E-09	ALDH8A1, ACADSB, SORD, CYP2J2, ASS1, CYP2C18, GNE, GLUD1, OGDHL, ALDOB, etc.
G0:0009725	Response to hormone	36 (13.7)	6.54E-09	NAMPT, SORD, ASS1, NR3C2, FHL2, FOXO1, ACAT1, ACVR1C, ZFP36L1, PRKAR2B, etc.
G0:0048585	Negative regulation of response to stimulus	47 (17.9)	1.10E-08	XDH, MCL1, MASP1, IL6ST, SORL1, FOXO1, DCN, GREM2, TRIB1, TCF21, etc.
Cellular Componen	t			
G0:0044459	Plasma membrane part	62 (23.7)	1.72E-05	GNA14, SGMS2, IL6ST, CPEB3, SLC5A1, SORL1, TSPAN7, CXCR2, ENPEP, SDC4, etc.
G0:0071944	Cell periphery	102 (38.9)	4.39E-05	SLC9A9, GNA14, SGMS2, IL6ST, FAM110C, TSPAN7, ENPEP, AMOTL2, MEGF10, WNT2, etc.
G0:0005887	Integral component of plasma membrane	43 (16.4)	4.83E-05	SLC39A14, SLC38A4, SLC38A2, SGMS2, IL6ST, C6, CCR1, SLC5A1, SORL1, TSPAN7, etc.
G0:0031226	Intrinsic component of plasma membrane	44 (16.8)	5.62E-05	SLC39A14, SLC38A4, SLC38A2, SGMS2, IL6ST, C6, CCR1, SLC5A1, SORL1, TSPAN7, etc.
G0:0005886	Plasma membrane	99 (37.8)	8.77E-05	SLC9A9, GNA14, SGMS2, IL6ST, TSPAN7, ENPEP, AMOTL2, MEGF10, WNT2, PRKAR2B, etc.
G0:0044421	Extracellular region part	79 (30.2)	1.25E-04	ALDH8A1, NAMPT, GNA14, ARSD, CYP2J2, MASP1, IL6ST, SLC5A1, SORL1, ENPEP, etc.
G0:1903561	Extracellular vesicle	60 (22.9)	4.37E-04	ALDH8A1, NAMPT, GNA14, ARSD, CYP2J2, IL6ST, SLC5A1, SORL1, ENPEP, SDC4, etc.
G0:0043230	Extracellular organelle	60 (22.9)	4.41E-04	ALDH8A1, NAMPT, GNA14, ARSD, CYP2J2, IL6ST, SLC5A1, SORL1, ENPEP, SDC4, etc.
G0:0005576	Extracellular region	88 (33.6)	4.53E-04	ALDH8A1, GNA14, NAMPT, ARSD, CYP2J2, MASP1, IL6ST, SLC5A1, SORL1, ENPEP, etc.
G0:0070062	Extracellular exosome	59 (22.5)	6.74E-04	ALDH8A1, NAMPT, GNA14, ARSD, CYP2J2, IL6ST, SLC5A1, SORL1, ENPEP, SDC4, etc.
Molecular Function				
G0:0050662	Coenzyme binding	12 (4.6)	5.37E-05	XDH, ACADSB, SORD, IVD, GLUD1, AOX1, OGDHL, SRD5A1, CAT, ACAT1, etc.
G0:0019838	Growth factor binding	10 (3.8)	7.15E-05	DUSP1, IL6ST, TEK, IL1RN, PDGFRA, LIFR, TGFBR3, IGFBP3, IGFBP4, ACVR1C
G0:0048037	Cofactor binding	14 (5.3)	1.12E-04	AADAT, XDH, ACADSB, SORD, GLUD1, OGDHL, ACAT1, GCH1, IVD, AOX1, etc.
G0:0016491	Oxidoreductase activity	25 (9.5)	1.32E-04	XDH, CYP3A4, ALDH8A1, ACADSB, SORD, CYP2J2, CYP2C18, GLUD1, HSD17B13, OGDHL, etc.
G0:0016725	Oxidoreductase activity, acting on CH or CH2 groups	4 (1.5)	2.90E-04	CYP3A4, XDH, AOX1, CYP1A2
G0:0019199	Transmembrane receptor protein kinase activity	7 (2.7)	9.47E-04	AMHR2, EFNB3, TEK, PDGFRA, TGFBR3, EPHA2, ACVR1C
G0:0038023	Signaling receptor activity	35 (13.4)	1.07E-03	TAPT1, IL6ST, CCR1, SORL1, NR3C2, CXCR2, SDC4, ACVR1C, EDNRB, GRIN2B, etc.
G0:0004860	Protein kinase inhibitor activity	7 (2.7)	1.14E-03	FLRT3, PRKAR2B, SOCS2, SOCS6, DCN, CISH, TRIB1
G0:0003707	Steroid hormone receptor activity	6 (2.3)	1.31E-03	AR, NR3C2, ESR1, NR4A1, NKX3-1, NR4A3
G0:0019210	Kinase inhibitor activity	7 (2.7)	1.37E-03	FLRT3, PRKAR2B, SOCS2, SOCS6, DCN, CISH, TRIB1
KEGG pathway				
hsa04610	Complement and coagulation cascades	8 (3.1)	3.04E-04	C8A, MBL2, CR1, MASP1, FGB, C6, SERPINE1, PROS1
hsa01100	Metabolic pathways	40 (15.3)	4.75E-04	CYP3A4, XDH, NAMPT, ACADSB, NDST3, SORD, CYP2J2, SGMS2, ASS1, CYP2C18, etc.
hsa00380	Tryptophan metabolism	6 (2.3)	8.76E-04	AADAT, AOX1, OGDHL, CAT, CYP1A2, ACAT1

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hsa00830	Retinol metabolism	7 (2.7)	1.37E-03	CYP3A4, LRAT, CYP2C18, AOX1, ADH6, ADH1B, CYP1A2
hsa00280	Valine, leucine and isoleucine degradation	6 (2.3)	1.84E-03	DBT, ACADSB, HMGCLL1, IVD, AOX1, ACAT1
hsa04020	Calcium signaling pathway	10 (3.8)	6.76E-03	EDNRB, GNA14, SLC8A1, ADRB2, CAMK4, CHRM2, PDGFRA, AVPR1A, ADRA1B, PRKCB
hsa04964	Proximal tubule bicarbonate reclamation	4 (1.5)	8.94E-03	GLS2, GLUD1, CA2, SLC4A4
hsa00071	Fatty acid degradation	5 (1.9)	9.32E-03	ACADSB, ACSL1, ADH6, ADH1B, ACAT1
hsa00591	Linoleic acid metabolism	4 (1.5)	1.70E-02	CYP3A4, PLA2G16, CYP2J2, CYP1A2
hsa00250	Alanine, aspartate and glutamate metabolism	4 (1.5)	2.80E-02	GLS2, ASS1, GLUD1, ALDH4A1



moted biological behaviors of cervical cancer cells via targeting RASSF8, such as proliferation, migration and invasion [15]. In digestive system cancer, Liu et al identified miR-224 could suppress the invasion and proliferation of gastric cancer cells through negatively regulating the expression and biological characteristics of RKIP [33], while Ke et al demonstrated miR-224 suppressed the migratory ability of colorectal cancer cells by targeting Cdc42 [34]. Reviewing the above published literature, we could conclude that miR-224 played different roles in different diseases by regulating multiple signaling pathways and target genes.

Up to now, a number of papers have been published about the roles of miR-224 in HCC. For

example, Scisciani et al identified miR-224 upregulation in HCC was association with p65/ NFkB and inflammatory pathways [27], while Wang et al discovered the overexpression of miR-224 in HCC has connection with EP300 [35]. Here, we verified the up-regulation of miR-224-5p from ten eligible GEO microarray datasets and TCGA RNA-seq data. The relatively big sample sizes made the conclusion to be credible; however, the regulation of miR-224 expression in HCC still remains to be further explored. About tumor biology, Zhang et al demonstrated miR-224 bound to multiple target genes such as PAK2, MAPK1, BCL-2, CDC42 and CDH1, influencing HCC cells' anti-apoptosis, migration, invasion, and proliferation [36]. Ma et al and Yu et al both reported that miR-224 upregulation may have connection with HCC tumor



progression by activating AKT signaling pathway [37, 38]. All of the above suggested miR-224 might take part in multiple biological processes and various signaling pathways in HCC tumorigenesis and progression. However, the confirmed target genes of miR-224 were rare due to limited experimental time and cost. Therefore, bioinformatics methods seem to play an important role on predicting the target genes of miR-224.

To explore potential role of miR-224-5p in HCC, bioinformatics methods were applied. We used 12 computational prediction tools to get target genes of miR-224-5p, and only genes that were simultaneously predicted by no less than four computational prediction tools were selected for further study. By overlapping the down-regulated DEGs in HCC, we finally gained 262 extracted target genes of miR-224-5p. Next, we performed GO, KEGG, and PPI network annotation to explore the regulatory role of the 262 overlapping genes. We noticed that these overlapping genes were most enriched in cellular response to chemical stimulus, response to organic substance, response to hormone, carboxylic acid metabolic process, oxoacid metabolic process and small molecule metabolic process, which indicated the overlapping genes mostly participate in the response to external stimulation and metabolic process in BP. As for CC, we found these overlapping genes were significantly involved in plasma membrane part, integral component of plasma membrane, intrinsic component of plasma membrane, and plasma membrane, which showed a possible membrane-associated metabolism. For MF, these overlapping genes were significantly enriched in coenzyme binding, growth factor binding, and cofactor binding, which indicated

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Figure 11. The molecular function (MF) network of miR-224-5p overlapping targeted genes. The circles represented different terms of molecular functions and the arrows represented relationships among terms. The color of nodes indicated the significance of the corresponding term.



Figure 12. Top 10 Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways of miR-224-5p overlapping targeted genes. The color tints indicated the significance of the corresponding pathway. The size of the circle presented the number of enrichment genes in the pathway. Rich factor expressed the percentage of the ratio of miR-224-5p overlapping targeted genes in current study vs total genes in the pathway.

the high binding ability in molecular function they had. Besides, we discovered the overlapping genes were significantly associated with metabolic pathways, tryptophan metabolism, retinol metabolism, linoleic acid metabolism, and alanine, aspartate and glutamate metabolism, indicated the potential miR-224-5p target genes participate in multiple metabolic pathways. Finally, we performed the PPI network annotation of the overlapping genes, and found a complex connection of the proteins by each other. In the hub genes analysis, we found a significant negative correlation between GNA14 and miR-224-5p. GNA14, located in 9g21.2, belongs to the G protein family. Nowadays, there were some studies about GNA14 and

malignant tumors, for example. Lim et al identified GNA14 mutations could cause child vascular tumors, helping to discover the mechanisms of malignant transformation [39], while Oshima et al defined GNA14 as one of the promoters to gastric cancer by TNF- α /TNFR1 signaling pathway [40]. Here, we predicted GNA14 as a hub gene in HCC by bioinformatics analysis and revealed the significant negative correlation between GNA14 and miR-224-5p for the first time.

To sum up, the study investigated the expression and clinicopathological characteristics of miR-224-5p in HCC based on TCGA RNAseq data and GEO microarray datasets, and finally confirmed the up-regulation and good diagnostic value of miR-224-5p in HCC. Furthermore, bioinformatics analysis was performed to predict the potential target genes and pathways of miR-224-5p. In the PPI analysis, five

hub genes were validated and extracted out for further correlation analysis. Surprisingly, GNA14 stood out cause for the significant negative correlation with miR-224. Forthcoming research can focus on the relationship of miR-224-5p with GNA14 which confirmed in vivo and in vitro, as well as the molecular mechanisms within the potential target genes and pathways of miR-224-5p in HCC.

Conclusion

In our study, we identified the up-regulation and diagnostic value of miR-224-5p in HCC. Meanwhile, the results of our study suggested miR-224-5p might play an oncogenic role in HCC by targeting GNA14. We also obtained





other potential target genes and pathways of miR-224-5p via bioinformatics method. Some potential target genes and pathways were already verified by previous studies, while most of others need further study as well as the molecular mechanisms of miR-224-5p in HCC. We hope our research may provide some clues for the subsequent research on miR-224-5p in HCC.

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Disclosure of conflict of interest

None.

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