

# Diagnostic significance and potential function of miR-338-5p in hepatocellular carcinoma: A bioinformatics study with microarray and RNA sequencing data

LIANG LIANG<sup>1\*</sup>, LI GAO<sup>1\*</sup>, XIAO-PING ZOU<sup>2</sup>, MENG-LAN HUANG<sup>2</sup>,  
GANG CHEN<sup>2</sup>, JIAN-JUN LI<sup>1</sup> and XIAO-YONG CAI<sup>1</sup>

<sup>1</sup>Department of General Surgery, The Second Affiliated Hospital of Guangxi Medical University, Nanning, Guangxi 530007;

<sup>2</sup>Department of Pathology, The First Affiliated Hospital of Guangxi Medical University, Nanning, Guangxi 530021, P.R. China

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**Abstract.** MicroRNA (miR)-338-5p has been studied in hepatocellular carcinoma (HCC); however, the diagnostic value and molecular mechanism underlying its actions remains to be elucidated. The present study aimed to validate the diagnostic ability of miR-338-5p and further explore the underlying molecular mechanism. Data from eligible studies, Gene Expression Omnibus (GEO) chips and The Cancer Genome Atlas (TCGA) datasets were gathered in the data mining and the integrated meta-analysis, to evaluate the significance of miR-338-5p in diagnosing HCC comprehensively. The potential target genes of miR-338-5p were achieved from the intersection of the deregulated targets of miR-338-5p from GEO and TCGA in addition to the predicted target genes from 12 online software. A protein-protein-interaction (PPI) network was drawn to illustrate the interaction between target genes and to define the hub genes. Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analyses were performed to investigate the function of the target genes. From the results, miR-338-5p exhibited favorable value in diagnosing HCC. Types of sample and experiment were defined as the possible sources of heterogeneity in meta-analysis. A total of 423 genes were selected as the potential target genes of miR-338-5p, and five genes were defined as the hub genes from the PPI network. The GO and KEGG analyses indicated that the target genes

were significantly assembled in the pathways of metabolic process and cell cycle. miR-338-5p may function as a novel diagnostic target for HCC through regulating certain target genes and signaling pathways.

## Introduction

Hepatocellular carcinoma (HCC) ranked as the 5th most frequent cancer and was also one of the lethal cancers, particularly in People's Republic of China where liver cancer was the most commonly diagnosed cancer and the most prevalent cause of cancer-related deaths followed by lung, stomach, and esophageal cancers (based on the statistics in 2015) (1,2). Because HCC was often diagnosed at an advanced stage, the prognosis of HCC patients was not optimistic (3). Therefore, a better understanding of the pathogenesis of HCC and a novel target for the early screening of HCC might improve the survival of HCC patients (4).

MicroRNAs (miRNAs) are small non-coding RNAs (18-25 nucleotides in length) that regulate the expression of multiple mRNAs at the post-transcriptional level by suppressing the stability and the translation of mRNAs (5,6). The aberrant expressions of miRNAs was observed in various human cancers, and extensive studies suggested that these deregulated miRNAs had the capacity to distinguish malignant tumors of liver, breast, lung, pancreas and leukemia from adjacent non-tumorous tissue (7-12). miR-338-3p and miR-338-5p originate from an intron of the gene encoding apoptosis-associated tyrosine kinase (AATK). Both miR-338-3p and miR-338-5p are co-expressed because they enjoyed the same promoter together (13). miR-338 was the previous ID of miR-338-3p, which had been reported in variety of diseases (13-16). miR-338\* is one of the members of miR-338 family and usually represented as miR-338-5p (17). As a member of the miRNA family, miR-338-5p was reported to be correlated with the carcinogenesis and progression of several human cancers including gastric cancer (18), colorectal cancer (19), and glioblastoma (20). However, there were limited studies on the clinical significance of miR-338-5p in HCC. Chen *et al* reported the overexpression of miR-338-5p

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*Correspondence to:* Dr Jian-Jun Li or Professor Xiao-Yong Cai, Department of General Surgery, The Second Affiliated Hospital of Guangxi Medical University, 166 Daxue Road East, Nanning, Guangxi 530007, P.R. China  
E-mail: lijianjunmail@163.com  
E-mail: xcy0771@163.com

\*Contributed equally

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in tumor tissues of the liver and preoperative plasma by miRNA array in Asian patients (21). Whether miR-338-5p is indeed a qualified diagnostic biomarker for HCC and the underlying molecular mechanism of miR-338-5p in HCC remained unclarified.

Therefore, this study aimed to investigate the diagnostic significance of miR-338-5p in HCC tissues and the molecular mechanism of miR-338-5p in HCC with a combination of meta-analysis and bioinformatics analysis. Our study confirmed the significance of miR-338-5p for the diagnosis of HCC and might promote the understanding of the molecular mechanism underlying it. The framework of this article was displayed in Fig. 1.

## Materials and methods

*The process of study selection.* In order to obtain the comprehensive data of the diagnostic value of miR-338-5p in HCC, a thorough search for the related studies was conducted in Gene Expression Omnibus (GEO) dataset and other database including PubMed, Embase, Cochrane, Web of Science, Sinomed, Chinese VIP, Wanfang database and China National Knowledge Infrastructure (CNKI) until December 15, 2016 with the searching strategies: (miR-338 or miRNA-338 or microRNA-338 or miR338 or miRNA338 or microRNA338 or 'miR 338' or 'miRNA 338' or 'microRNA 338') and (malignan\* or cancer or tumor or tumour or neoplas\* OR carcinoma) AND (hepatocellular or liver or hepatic or HCC). Studies that meet the following criteria were eligible for the meta-analysis: i) Studies evaluated the expression of miR-338-5p for the diagnosis of HCC; ii) the disease of the patients were validated with golden standard; iii) the number of cases were reported in the study; and iv) the sensitivity and specificity of the diagnostic test were available directly or indirectly from the study. The exclusion criteria of the studies were as follows: i) The content of the studies were irrelevant with HCC; ii) the subjects of experiment were not human beings; iii) there was no sufficient data for researchers to directly acquire or calculate sensitivity or specificity of the diagnostic test; and iv) studies were classified as review, meta-analysis, case study or conference note. Moreover, data of the diagnostic value of miR-338-5p in HCC was also downloaded from the cancer genome atlas (TCGA) (<https://cancergenome.nih.gov/>).

*Data extraction and statistical analysis.* The following information and data were extracted from the included studies: The ID no. of each GSE chip, first author, year of publication, country, experiment type, platform of each GSE chip, sample number for the experiment group and control group, tissue types, true positivity (TP), false positivity (FP), false negativity (FN) and true negativity (TN).

MetaDiSc1.4 and STATA12.0 were applied for all the statistical analysis. To explore expression of miR-338-5p, the continuous outcomes of GEO and TCGA datasets were calculated with standard mean difference (SMD). The sensitivity (SEN), specificity (SPE), positive likelihood ratio (PLR), negative likelihood ratio (NLR) and diagnostic odds ratio (DOR) of the included studies were pooled with the bivariate meta-analysis model (22,23). The summary

receiver operator characteristic (SROC) curve was plotted according to the sensitivity and specificity from each study. The area under the SROC curve (AUC) calculated from the SROC reflected the capacity of miR-338-5p to differentiate HCC patients from non-cancer patients accurately. An AUC value of 0.5 or 1.0 represents a poor or perfect diagnostic value, respectively (24). Additionally, Q test and  $I^2$  statistics were employed to assess the heterogeneity between studies. The random-effects model would be used to pool the results if an  $I^2$  value was more than 50% with a P-value <0.10; otherwise, a fixed-effects model would be applied (25,26). To identify the source of heterogeneity, the subgroup analysis was conducted based on the number and features of the included studies. With regard to the publication bias, the Deeks' funnel plot asymmetry test was carried out to detect the publication bias, and P-value <0.05 was indicative of significance.

*The target genes of miR-338-5p.* The potential target genes of miR-338-5p came from three sources: The differentially expressed genes from GEO, TCGA and the predicted genes from 12 online software (miRWalk, MicroT4, miRanda, mirBridge, miRDB, miRMap, miRMAP, PicTar2, PITA, RNA22, RNAhybrid and TargetScan). We firstly searched Gene Expression Omnibus (GEO) datasets for deregulated target genes of miR-338-5p from the mRNA profiling data of HCC samples on December 15, 2016. All the GSE chips shared the same platform: GPL570 (Affymetrix Human Genome U133 Plus 2.0 Array). After preliminary screening, 54 studies remained for further selection. Among the 54 studies, *Homo sapiens* tissue samples instead of cell lines samples were included for further analysis. Finally, 10 HCC datasets (GSE29721, GSE45436, GSE55092, GSE62232, GSE9843, GSE41804, GSE6764, GSE33006, GSE6222 and GSE19665) comprising 431 HCC samples and 198 control samples were chosen for further analysis. Differentially expressed genes (DEGs) between cancer and normal samples of 10 datasets were acquired via GCBI online tool (<https://www.gcbi.com.cn/gclib/html/index>). Fold-change >1.5, and a P-value <0.05 was set as the threshold for the DEGs. Another database containing high-throughput data: TCGA was also searched. Publicly available miRNA-seq and RNA-seq data of liver HCC was downloaded from the TCGA data portal (December 2016, <https://gdc-portal.nci.nih.gov/>). Since the TCGA data were a community resource project, additional approval by an ethics committee of our hospital was not mandatory. And the present study adhered to the TCGA publication guidelines and data access policies. From the downloaded data of 377 HCC samples and 50 normal liver samples. R language package DESeq was subsequently used for the calculation of DEGs (Padj <0.05 and the absolute log2 fold-change >1). As for the predicted genes, selection was based on the condition that they were recorded in more than 4 of the 12 prediction software. The selected qualified target genes from the online software and the validated target genes from miRWalk were considered as the potential target genes of miR-338-5p.

*The protein-protein-interaction (PPI) network and validation of target genes.* To illustrate the interaction between

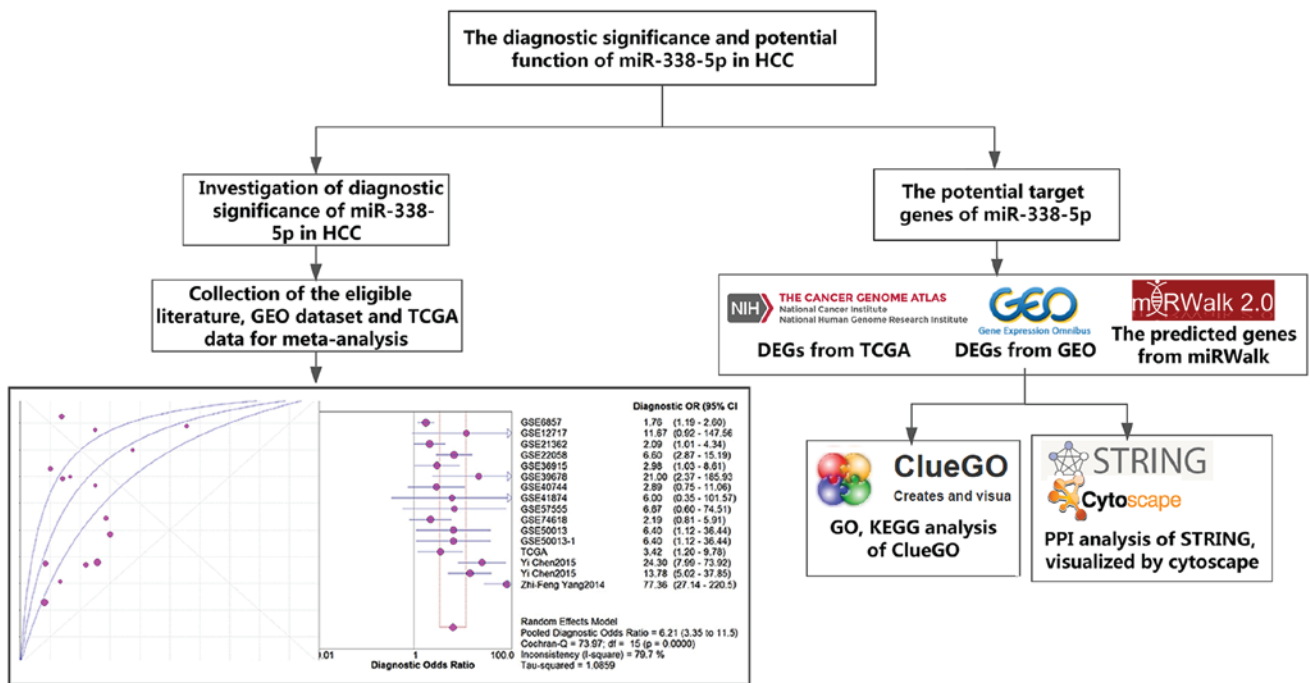


Figure 1. The framework of this study. The framework described the idea of this study.

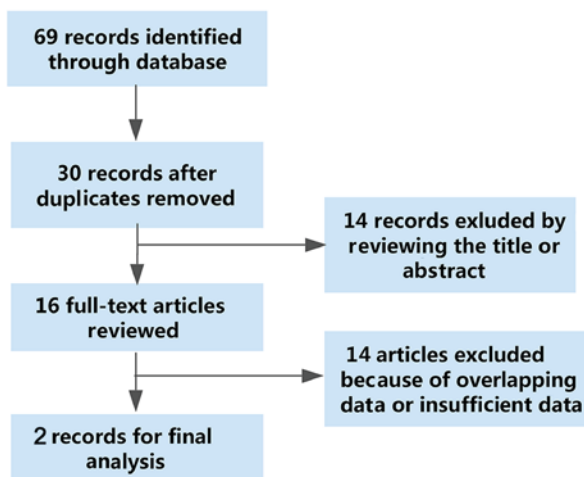


Figure 2. The flowchart: The detailed process of study selection was recorded in the flowchart. After the preliminary screening and full-text scrutinize, three studies were considered eligible studies for the meta-analysis.

the targets of miR-338-5p, a PPI network was drawn by Cytoscape v.3.5.0. The nodes and edges represented target genes and the interactions between target genes, respectively. Hub genes were identified according to the value of degrees of each node. Protein expression of hub gene was validated by The Human Protein Atlas (HPA), an immunohistochemistry (IHC) database (27). Each antibody in the database has been used for IHC staining of both normal and HCC tissues.

*The Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis of the target genes.* The GO and KEGG pathway analysis were performed by the BiNGO and ClueGO plug-in unit

in Cytoscape v.3.5.0 for the functional annotation of the target genes. Three GO terms including biological process (BP), cellular component (CC) and molecular function (MF) were utilized to identify the enrichment of target genes. P-value <0.05 was significant.

## Results

*Eligible studies for the meta-analysis.* As shown in Fig. 2, the flowchart exhibited the selection and retrieval process of the qualified studies. A total of 69 studies were identified as the initial records, and 30 studies remained after the removal of duplicate records. Then, 14 records were excluded in the preliminary screening of the titles and abstracts of the articles. As a consequence, 16 studies were reviewed in the full text. Among the 16 studies, 14 studies were ineligible due to insufficient data of the diagnostic parameters or duplicate data. Eventually, two studies were enrolled for the meta-analysis. Though two included studies were conducted by the same authors, we failed to validate that the two studies shared the same patient cohorts. Thus, we regard the two studies of Chen *et al* as two different studies (21,28).

*Assessment of the diagnostic value and the integrated meta-analysis.* To comprehensively evaluate the diagnostic value of miR-338-5p, we supplemented the literature analysis with GEO data and TCGA data. We searched the GEO dataset with the same searching strategies in literature meta-analysis. Finally, a total of eligible 11 GSE chips were included in our meta-analysis (29-38) (Table I), and the expression level of each study were showed in Fig. 3. With the random-effects model, the forest-plot represented that no significant difference expression was observed between HCC tissue and

Table I. Basic information and clinical data of the included studies.

GEO accession	Author, year	Public year	Country	Experiment type	Platform	HCC	Control	Sample type	TP	FP	TN	(Refs.)
GSE6857	Budhu <i>et al</i> , 2008	2008	USA	Non-coding RNA profiling by array	GPL4700	240	241	Tissue	91	62	149	179 (29)
GSE12717	Su <i>et al</i> , 2009	2008	USA	Non-coding RNA profiling by array	GPL7274	10	6	Tissue	7	1	3	5 (30)
GSE22058	Burchard <i>et al</i> , 2010	2010	USA	Non-coding RNA profiling by array	GPL10457	96	96	Tissue	36	8	60	88 (31)
GSE40744	Diaz <i>et al</i> , 2013	2013	USA	Non-coding RNA profiling by array	GPL14613	26	19	Tissue	8	4	18	26 (32)
GSE50013	Shen <i>et al</i> , 2013	2013	USA	Non-coding RNA profiling by array	GPL15497	18	18	Plasma	16	10	2	8 (33)
GSE41874	Morita (unpublished)	2013	Japan	Non-coding RNA profiling by array	GPL7722	6	4	Tissue	4	1	2	3 -
GSE57555	Murakami <i>et al</i> , 2015	2015	Japan	Non-coding RNA profiling by array	GPL16699	5	16	Tissue	4	6	1	10 (34)
GSE74618	Martinez-Quetglas <i>et al</i> , 2016	2016	Spain	Non-coding RNA profiling by array	GPL14613	223	20	Tissue	131	10	92	10 (35)
GSE36915	Shih <i>et al</i> , 2012	2012	Taiwan	Non-coding RNA profiling by array	GPL8179	68	21	Tissue	37	6	31	15 (36)
GSE39678	Noh <i>et al</i> , 2013	2013	South Korea	Non-coding RNA profiling by array	GPL15852	16	8	Tissue	14	2	2	6 (37)
GSE21362	Sato <i>et al</i> , 2011	2011	Japan	Non-coding RNA profiling by array	GPL10312	73	73	Tissue	27	16	46	57 (38)
TCGA				miRNA-Seq	Illumina	375	50	Tissue	156	8	218	42
Literaure	Chen <i>et al</i> , 2015	2015	China	qRT-PCR		37	31	Plasma	27	0	10	31 (21)
Literaure	Chen <i>et al</i> , 2015	2015	China	qRT-PCR		39	25	Plasma	35	4	4	21 (28)

TP, true positivity; FP, false positivity; FN, false negativity; TN, true negativity.

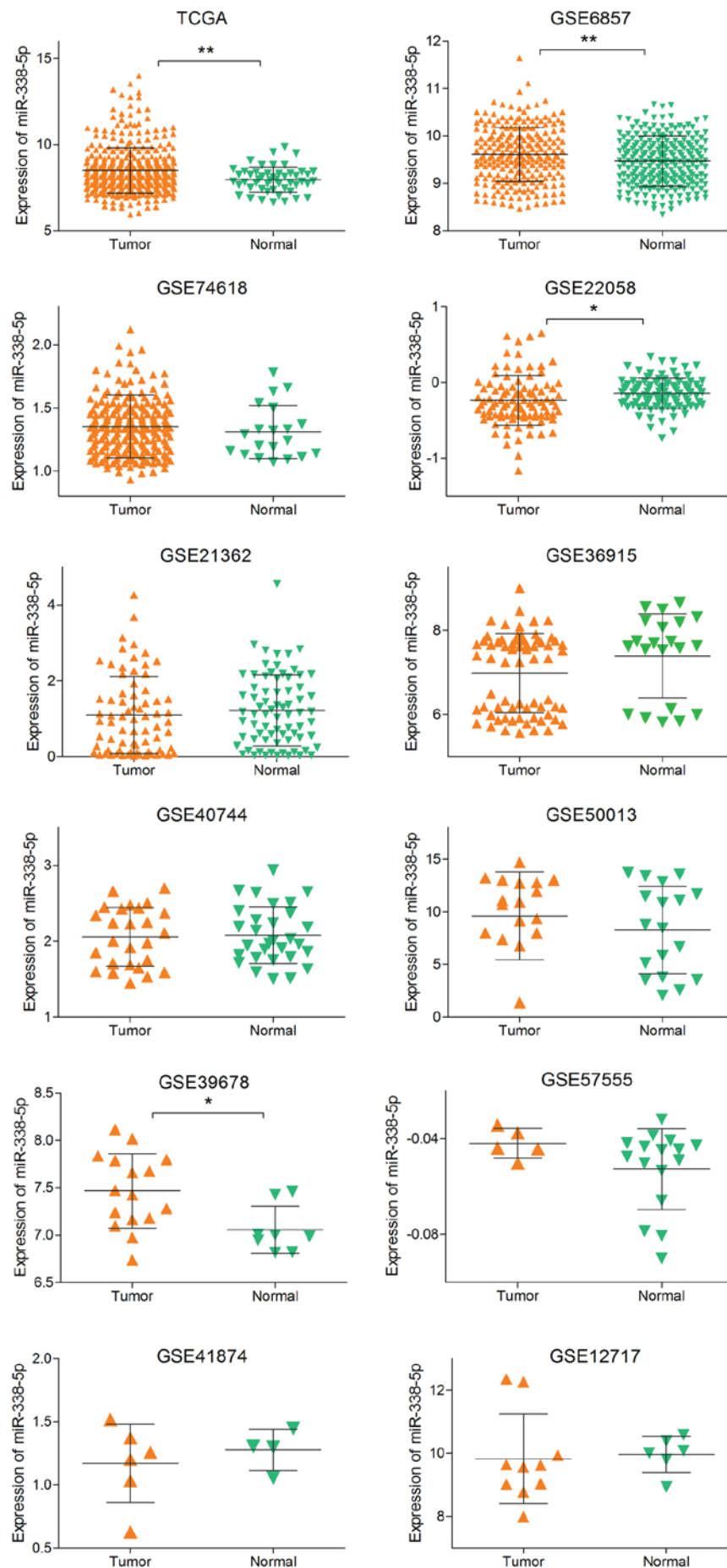


Figure 3. The expression level of each study from TCGA and GEO. Scatter plots were created to illustrate the expression of miR-338-5p in normal tissues and HCC for each study from TCGA and GEO. \* $P < 0.05$ ; \*\* $P < 0.01$ . TCGA, The Cancer Genome Atlas; GEO, Gene Expression Omnibus; HCC, hepatocellular carcinoma.



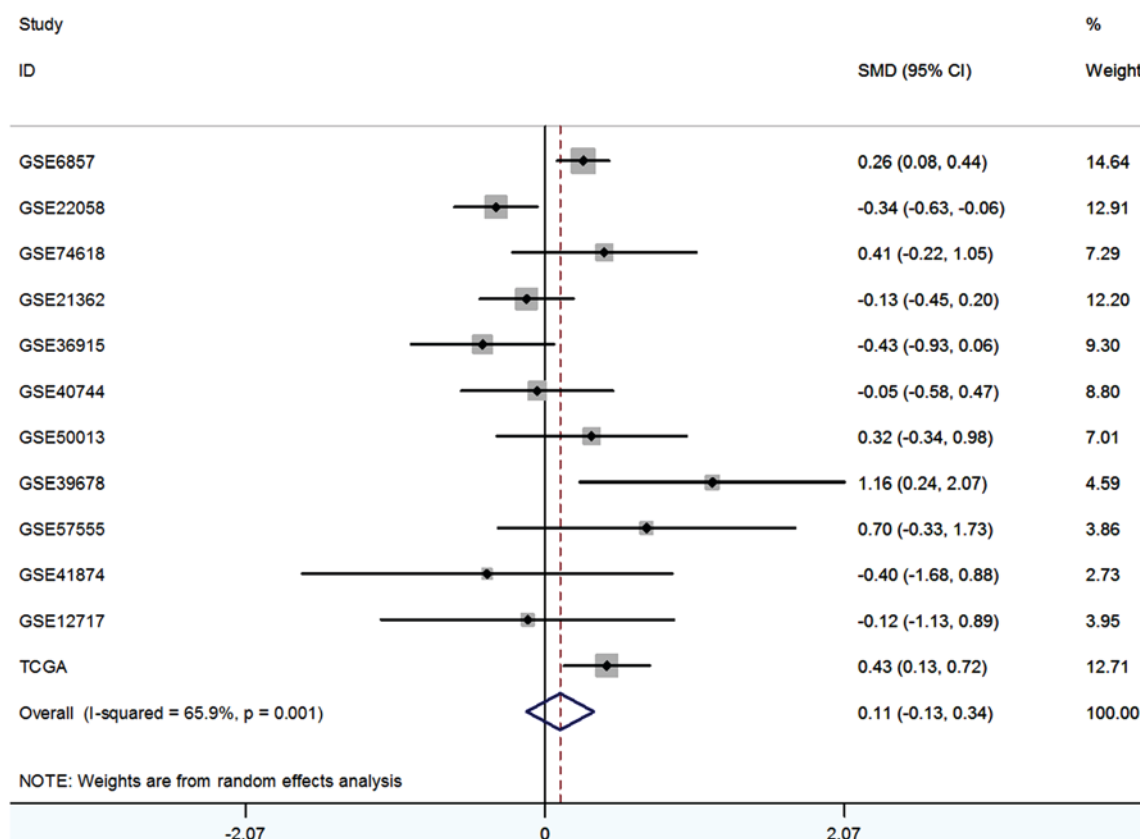


Figure 4. Forest plot of studies calculating SMD of miR-338-5p expression in HCC group compared with control group. SMD, standard mean difference; HCC, hepatocellular carcinoma.

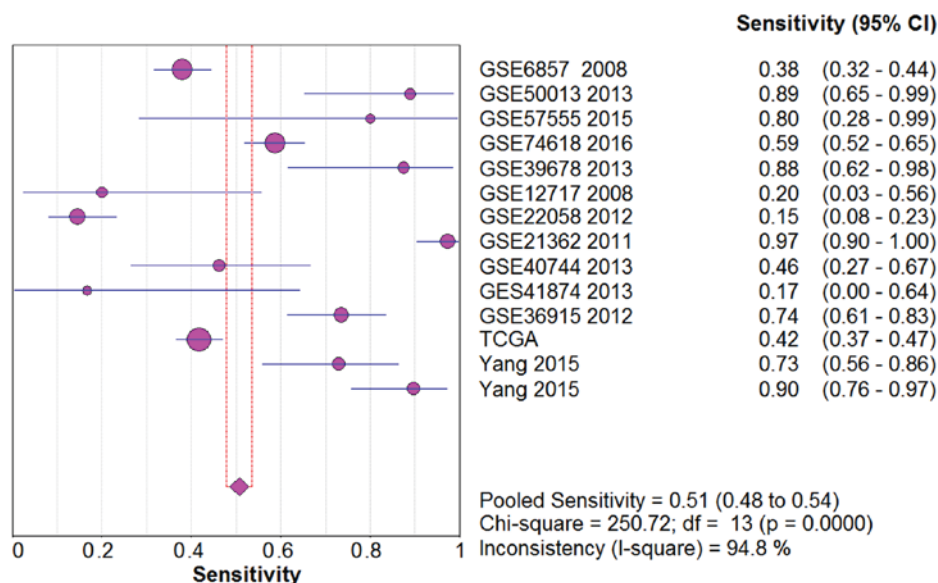


Figure 5. The pooled SE of the integrated meta-analysis. The pooled SE of the integrated meta-analysis was 0.51 (95% CI: 0.48-0.54).

normal tissue. The pooled SMD (0.11, 95% CI: -0.13, 0.34) was showed in Fig. 4.

From the chi-square test and  $I^2$  test, significant heterogeneity existed in all the pooled effects (SE, SP, PLR, NLR and DOR) between studies (All  $I^2 > 50\%$ ;  $P < 0.05$ ). Therefore, random effects model were employed to estimate the overall SE, SP, PLR, NLR and DOR of all the data. As shown

in Figs. 5-9, the SE, SP, PLR, NLR and DOR of all the studies were 0.51 (95% CI: 0.48-0.54), 0.69 (95% CI: 0.65-0.73), 1.76 (95% CI: 1.17-2.66), 0.64 (95% CI: 0.52-0.80) and 3.17 (95% CI: 1.83-5.47). As for the result of SROC, the AUC value of miR-338-5p was 0.691 (Fig. 10). Moreover, the Deeks funnel plot asymmetry test was carried out with Stata 12.0, and no publication bias was detected ( $P > 0.05$ ) (Fig. 11).

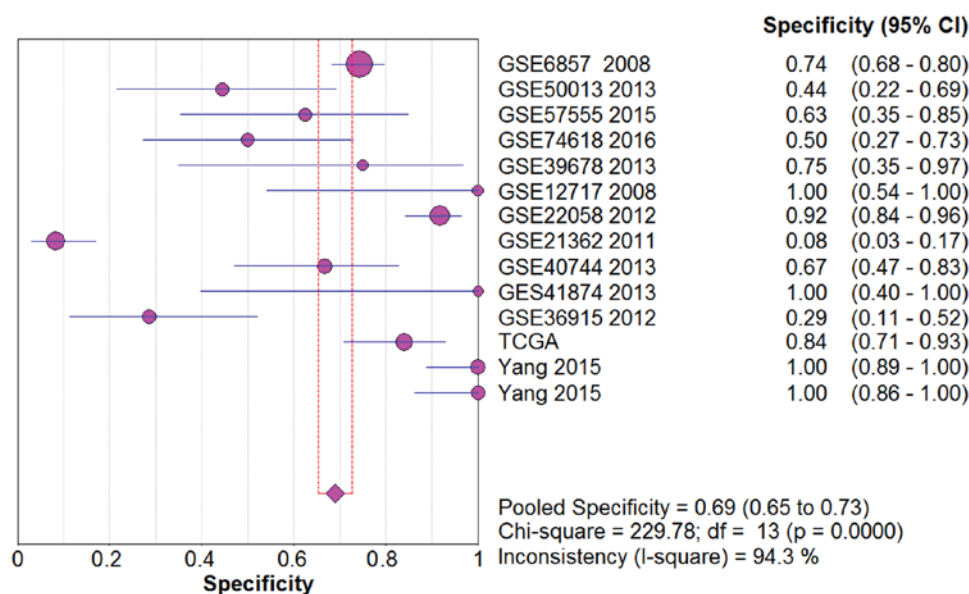


Figure 6. The pooled SP of the integrated meta-analysis. The pooled SP of the integrated meta-analysis was 0.69 (95% CI: 0.65-0.73).

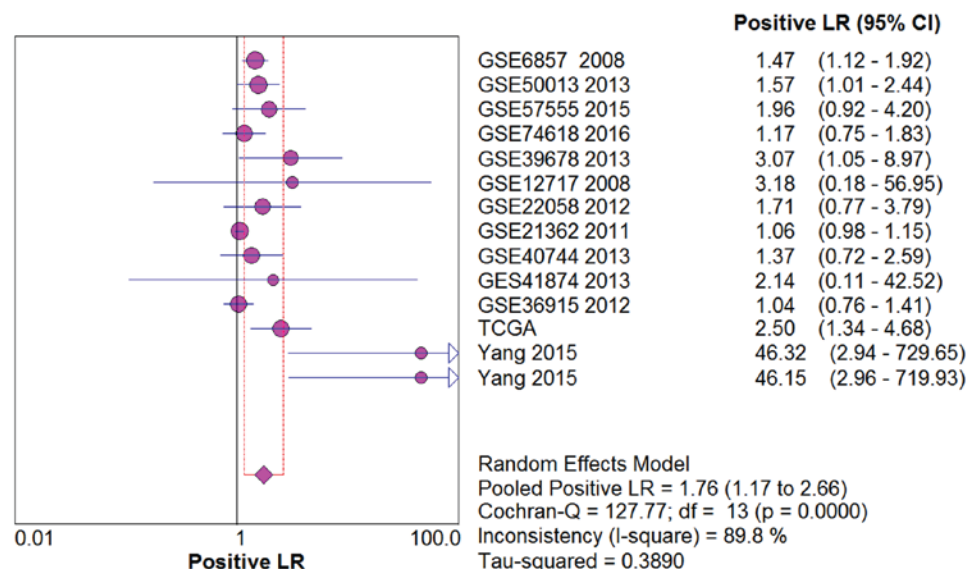


Figure 7. The pooled PLR of the integrated meta-analysis. The pooled PLR of the integrated meta-analysis was 1.76 (95% CI: 1.17-2.66). PLR, positive likelihood ratio.

Now that significant heterogeneity existed between the studies, the subgroup analysis was performed to seek the potential sources of heterogeneity. In the subgroup of sample types, the heterogeneity decreased substantially in the pooling estimates of NLR (49.2%) and DOR (5.6%) in the group of tissue. The value of SE, SP, PLR and DOR were obviously higher in the plasma group (0.83, 0.74-0.90; 0.86, 0.77-0.93; 14.02, 0.08-2,395.96; 61.30, 3.61-1,040.31) than in the tissue group (0.48, 0.45-0.51; 0.67, 0.63-0.71; 1.51, 1.08-2.11; 2.05, 1.51-2.77) and the value of NLR was notably lower in the plasma group (0.21, 0.11-0.38) than in the tissue group (0.81, 0.71-0.92).

With regard to the subgroup of experiment, the heterogeneity decreased substantially in the pooling estimate of DOR (10.3%) in the microarray group, and declined

heterogeneity of SE (0%), PLR (0%) and DOR (0.0%) were also observed in the group of qRT-PCR. The value of SE, SP, PLR and DOR were obviously higher in the qRT-PCR group (0.82, 0.71-0.90; 1.00, 0.94-1.00; 46.23, 6.60-323.68; 254.42, 32.2-2,010.45) than in the microarray group (0.49, 0.46-0.52; 0.66, 0.62-0.70; 1.51, 1.11-2.06; 2.15, 1.56-2.97) and the value of NLR was notably lower in the qRT-PCR group (0.19, 0.08-0.47) than in the microarray group (0.79, 0.69-0.91). This result confirmed that types of sample and experiment were the possible sources of heterogeneity in this study.

**Bioinformatics study of the target genes of miR338-5p.** According to the results, a total of 1,698 and 1,798 genes were identified as DEGs targeted by miR-338-5p from TCGA and

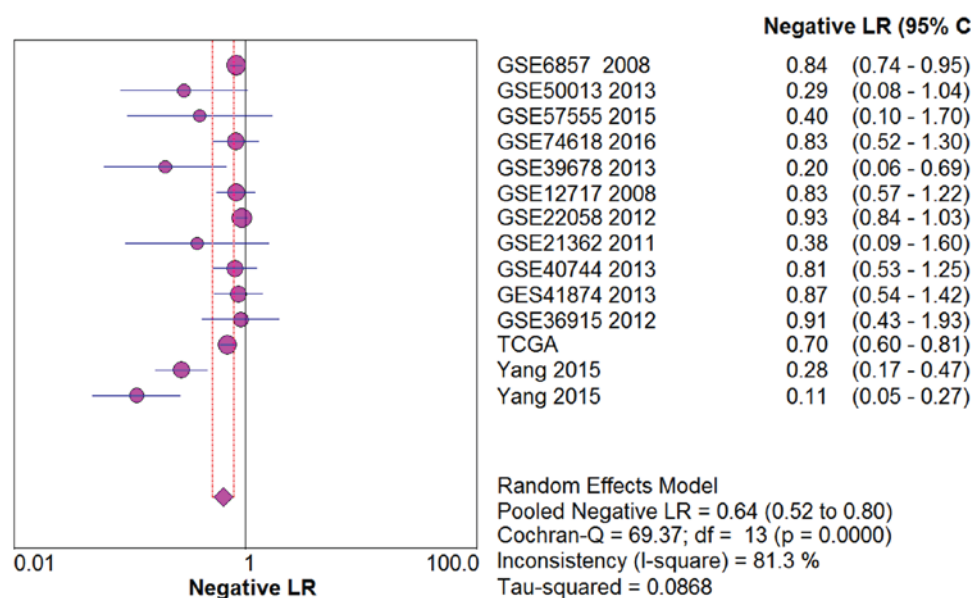


Figure 8. The pooled NLR of the integrated meta-analysis. The pooled NLR of the integrated meta-analysis was 0.64 (95% CI: 0.52-0.80). NLR, negative likelihood ratio.

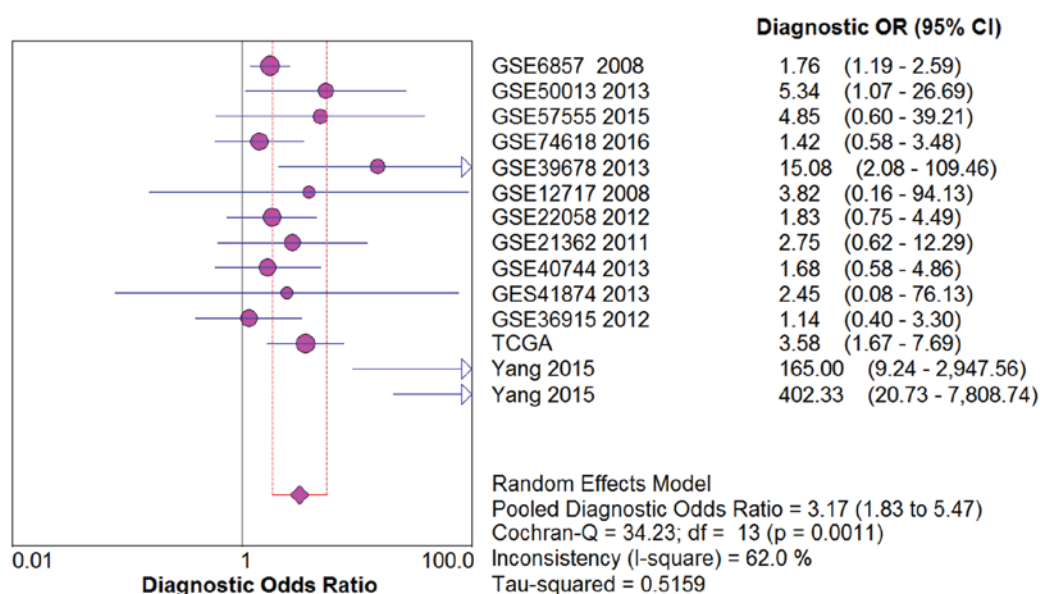


Figure 9. The pooled DOR of the integrated meta-analysis. The pooled DOR of the integrated meta-analysis was 3.17 (95% CI: 1.83-5.47). DOR, diagnostic odds ratio.

GEO, respectively. Additionally, a total of 3,610 predicted target genes that appeared in more than four times of the 12 online software were obtained. Taking the intersection of the DEGs from GEO and TCGA as well as the qualified predicted targets genes, we selected 423 genes for the following bioinformatics analyses (Fig. 12). The PPI network shown in Fig. 13 illustrated the interactions between the target genes of miR-338-5p. There were 147 nodes and 248 edges in the network. Hub genes with a degree values of more than 11 including NCOR1, IGF1, FOXO1, FOS, CDCA8, BUB1B, PCNA, ESR1, BIRC5, MYC and CDK1 were emphasized in red while the remaining were colored in green. To verify that these hub genes are targeted by miR-338-5p, we obtained the immunohistochemical staining of several of the hub genes

including NCOR1 and FOXO1 in HCC tissues and normal tissues. As shown in Fig. 14, NCOR1 and FOXO1 were found to have medium staining and moderate intensity in cytoplasmic/membranous of normal tissues, while a lower staining and weaker intensity of these genes were observed in HCC tissues.

According to the results of GO analysis in cytoscape, the target genes were found to enrich most significantly in the following biological pathways: response to organic substance, response to chemical stimulus and oxoacid metabolic process. As for cellular component and molecular function, target genes mainly assembled in extracellular region part and binding, respectively (Table II; Fig. 15). Moreover, a total of 5 significant pathways were recorded from the KEGG pathway



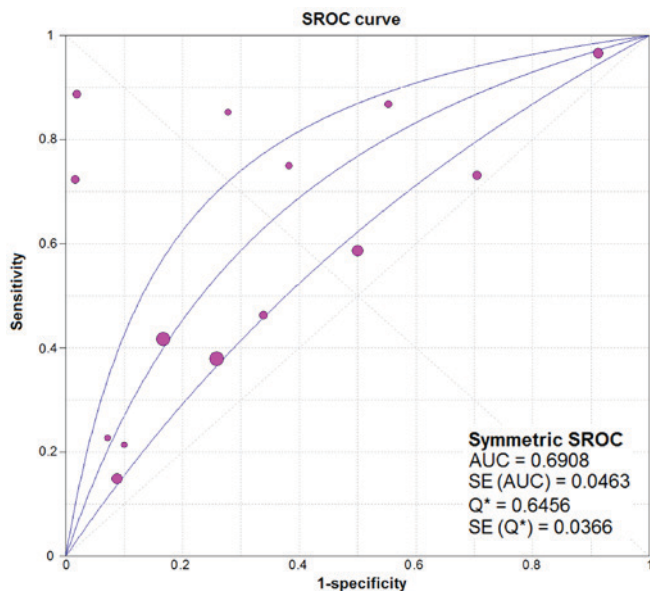


Figure 10. The pooled SROC of the integrated meta-analysis. The pooled SROC of the integrated meta-analysis was 0.691. SROC, summary receiver operator characteristic.

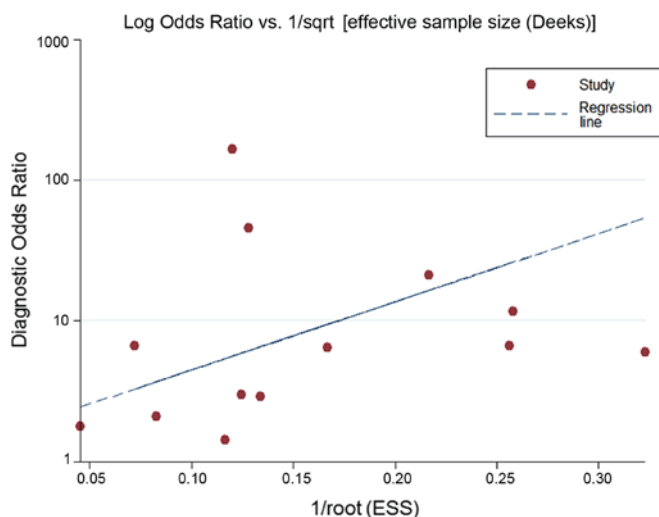


Figure 11. The publication bias. The symmetrical Deeks funnel plot indicated no publication bias.

analysis such as valine, leucine and isoleucine degradation, pathways in cancer and cell cycle (Table III; Fig. 16) were the most significant.

## Discussion

Considerable attention has been attracted to miRNAs as promising diagnostic targets for the early screening of human cancers. Prior to our study, several researches have reported some miRNAs had diagnostic value in HCC. A 3-miRNA panel: miR-92-3p, miR-107, and miR-3126-5p discovered by Zhang *et al* were claimed to distinguish HCC patients in early stage and HCC patients with low-level AFP from their corresponding controls with high accuracy (39). Additionally, some single miRNAs including miR-21 and

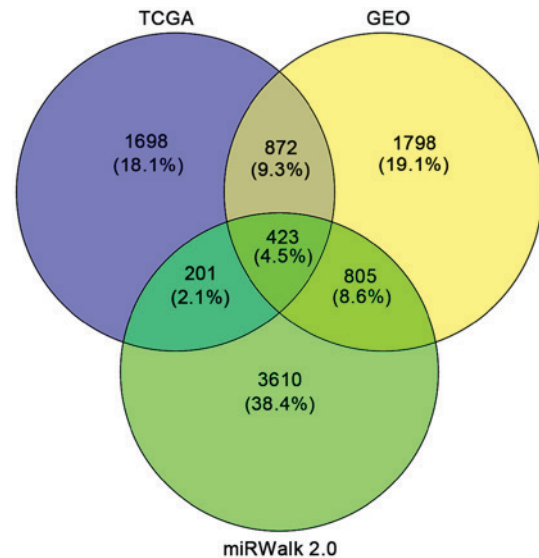
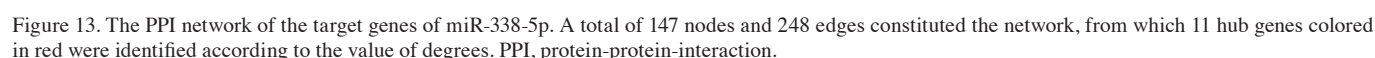


Figure 12. The Venn diagram for the target genes of miR-338-5p. Three circles represented the collected target genes from three different sources (blue for TCGA, yellow for GEO and green for the predicted target genes) There were 1,698, 1,798 and 3,610 potential target genes from TCGA, GEO and online software, respectively. A total of 423 genes were obtained after taking the intersection of the three parts. TCGA, The Cancer Genome Atlas; GEO, Gene Expression Omnibus.

miR-224 also exhibited prominent diagnostic potential for HCC (40,41) and so far, only one study referred to the diagnostic value of miR-338-5p in HCC with the method of miRNA array. Chen *et al* (21) reported a moderate ability of miR-338-5p to differentiate HCC from liver cirrhosis with the AUC of 0.799. Furthermore, an extremely strong diagnostic value of miR-338-5p (AUC=0.909) was observed when diagnosing HCC from healthy controls. Despite some advances has been made in exploring the diagnostic capacity of miRNAs for HCC, the diagnostic significance of miR-338-5p in HCC was indefinite, and the relative molecular mechanism has not been elucidated in these studies; therefore our study was the first one to comprehensively assess the diagnostic value of miR-338-5p in HCC with the data from GEO, TCGA and literature as well as to investigate the underlying molecular mechanism through bioinformatics study.

From the meta-analysis result from our collected literature and the integrated meta-analysis, miR-338-5p may serve as a possible diagnostic target for HCC with fair sensitivity and specificity, which enlightened us that miR-338-5p might play an essential role in the occurrence and progression. Previous studies have pointed out that miR-338-5p exerted a tumor suppressive function in a wide range of cancers. In glioblastoma, miR-338-5p was discovered to inhibit the proliferation, invasion and promote apoptosis by targeting EFEMP1 (42); similarly, miR-338-5p significantly attenuated the malignant potential of gastric cancer cells through regulating BMI1 (13). In contrast, miR-338-5p was increased in both blood and tissue of colorectal cancer (CRC), and represented high area under ROC curve (AUC) of 0.871. The performance of miR-338-5p indicated that it could be a potential biomarker in CRC (19). A similar trend was observed in CRC compared with HCC. Furthermore, according to the



literature meta-analysis, the samples of which were different. Moreover, due to the limited number of literature, we failed to trace the heterogeneity by carrying out subgroup analysis for our literature meta-analysis. Expanding the sample size was necessary for a more reliable assessment of the diagnostic value of miR-338-5p in HCC.

We firstly identified the potential target genes of miR-338-5p and further defined the hub genes from PPI network. The 11 hub genes were assumed to correlate closely with miR-338-5p and play essential roles in the miR-338-5p relevant pathogenesis of HCC. Among the hub genes, CDK1 was important protein for the regulation of cell cycles belonging to the cyclin-dependent kinases family (43). The



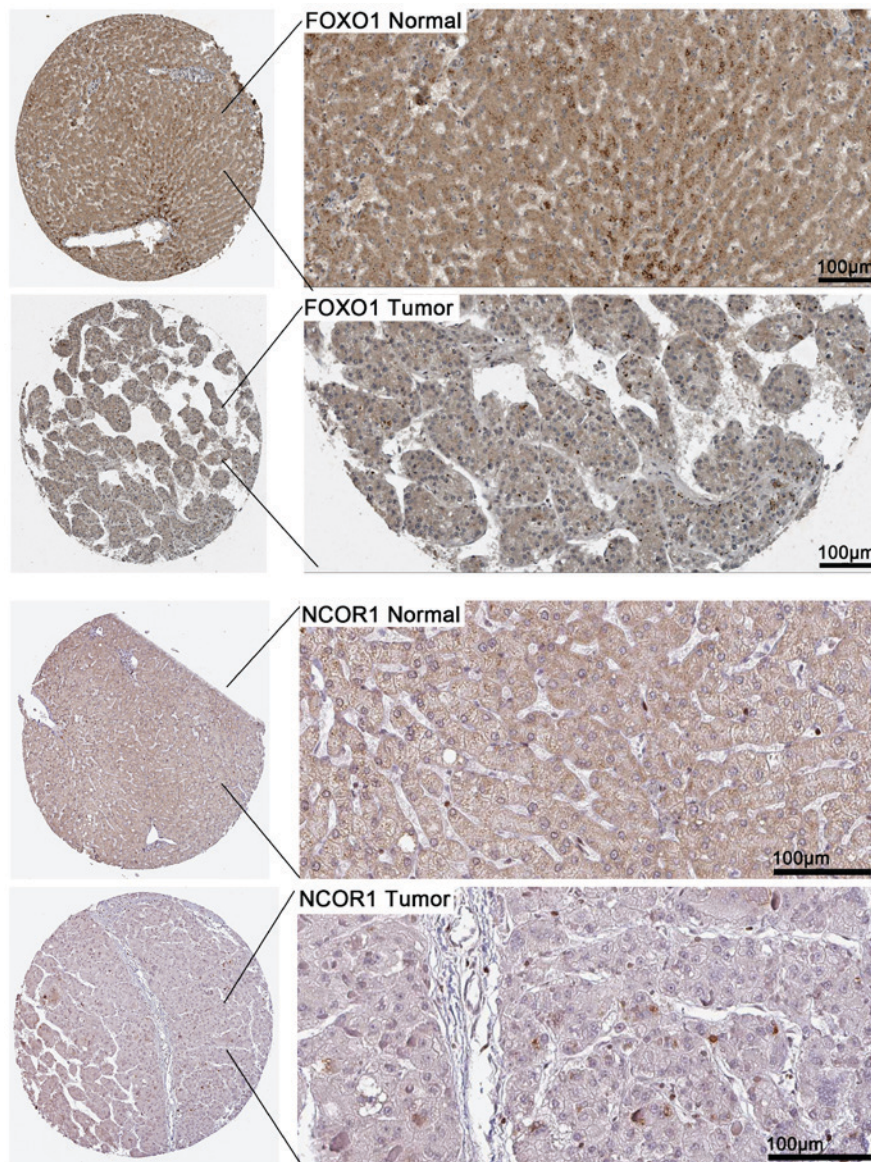


Figure 14. The immunohistochemical staining of NCOR1 and FOXO1 in normal tissues and HCC tissues. According to the immunohistochemical results of these genes in HPA database, NCOR1 and FOXO1 presented medium staining in normal tissues; while a lower staining of these genes were observed in HCC tissues. HCC, hepatocellular carcinoma; HPA, the Human Protein Atlas.

overexpression of CDK1 was detected in various cancers, and a poor prognosis of renal cell carcinoma patients was associated with the high expression of CDK1 and CDK2 (44-48). We hypothesized that CDK1 deregulated by miR-338-5p might promote the deterioration of HCC by affecting the cell cycle of HCC cells. Apart from CDK1, several hub genes such as MYC, BIRC5, IGF1, NCOR1 and FOXO1 participate in the regulation of a wide range of biological processes including cell proliferation, apoptosis and migration (49-57). These genes were reported to be aberrantly expressed in various cancers (58-62) and they were also involved in the malignant progression of HCC (49,50,52,63). It was conceived that miR-338-5p might interact with these molecules through potential signaling pathways to influence the development of HCC. PCNA, a protein that acted as DNA sliding clamp, was found to engage in DNA duplication and repair with its overexpressed in HCC. Further study was necessary to probe into the association between PCNA and miR-338-5p in HCC.

In this study, we analyzed 11 hub genes protein expression by HPA database. The result indicated that the expression of FOXO1 and NCOR1 was downregulated in HCC and most likely regulated by miR-338-5p.

GO enrichment analysis was indicative of the possible functions of the target genes in HCC and the results from three GO terms hinted that the target genes were mainly assembled in response to organic substance. Most of the potential functions of the target genes from the GO analysis were accomplished in signaling pathways. Therefore, it is of great importance to investigate the signaling pathways gathered by the target genes of miR-338-5p. From the results of the KEGG pathway analysis, the most significant ten pathways such as pathways in cancer and cell cycle were closely associated with cancer.

Despite the valuable findings acquired from the meta-analysis and bioinformatics study, there were still some limitations in our study. The sample size of our literature was

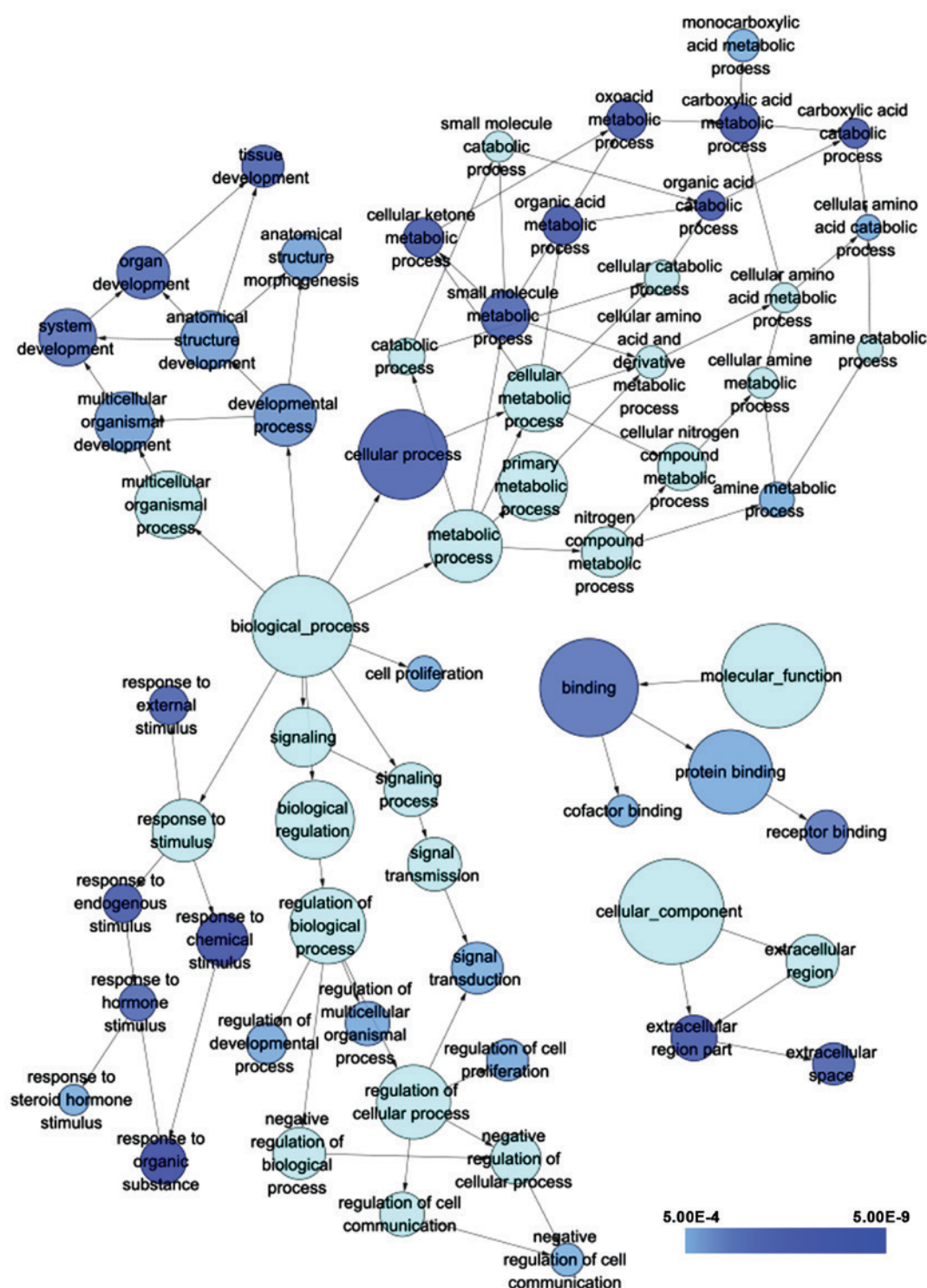


Figure 15. The GO terms of target genes of miR-338-5p. The circles represented different GO terms of BP, CC and MF. The darker the color of the circles presents, the greater the significance the term indicated. The relationship between different GO terms were symbolized by arrows. GO, Gene Ontology; BP, biological process; CC, cellular component; MF, molecular function.

too small for further analysis to identify heterogeneity, which weakened the reliability of our results. Since the samples are from different types, including tissue and plasma, a bias and sensitivity problems might originate from sample types in analysis. Additionally, we only included studies published in Chinese or English, which might cause bias of selection to the meta-analysis. A plausible way to address these issues is to conduct future studies with larger samples and fewer

language restrictions to further verify the diagnostic value of miR-338-5p for HCC.

In conclusion, we anticipated that miR-338-5p may serve as a promising diagnostic marker for HCC and miR-338-5p could affect the development of HCC by targeting certain downstream genes and pathways. The future research will be concentrated on validating the target genes of miR-338-5p and its function in the significant signaling pathways mentioned before.



Table II. GO functional annotation of the target genes of miR-338-5p from Cytoscape.

ID	Category	GO term	P-value	Count
GO:0006082	GO_Biological process	Organic acid metabolic process	3.87x10 <sup>-9</sup>	37
GO:0016054	GO_Biological process	Organic acid catabolic process	6.26x10 <sup>-9</sup>	16
GO:0008152	GO_Biological process	Metabolic process	1.29x10 <sup>-3</sup>	159
GO:0009056	GO_Biological process	Catabolic process	1.15x10 <sup>-1</sup>	28
GO:0007275	GO_Biological process	Multicellular organismal development	1.11x10 <sup>-6</sup>	102
GO:0048731	GO_Biological process	System development	9.71x10 <sup>-8</sup>	91
GO:0048519	GO_Biological process	Negative regulation of biological process	1.82x10 <sup>-5</sup>	72
GO:0048523	GO_Biological process	Negative regulation of cellular process	3.81x10 <sup>-5</sup>	66
GO:0050789	GO_Biological process	Regulation of biological process	3.62x10 <sup>-4</sup>	176
GO:0065007	GO_Biological process	Biological regulation	6.72x10 <sup>-5</sup>	189
GO:0008150	GO_Biological process	Biological_process	1.15x10 <sup>-3</sup>	336
GO:0032501	GO_Biological process	Multicellular organismal process	1.15x10 <sup>-5</sup>	133
GO:0044237	GO_Biological process	Cellular metabolic process	6.22x10 <sup>-3</sup>	132
GO:0034641	GO_Biological process	Cellular nitrogen compound metabolic process	2.19x10 <sup>-2</sup>	59
GO:0044238	GO_Biological process	Primary metabolic process	4.35x10 <sup>-3</sup>	140
GO:0050896	GO_Biological process	Response to stimulus	4.59x10 <sup>-5</sup>	112
GO:0009719	GO_Biological process	Response to endogenous stimulus	3.77x10 <sup>-9</sup>	34
GO:0050794	GO_Biological process	Regulation of cellular process	9.14x10 <sup>-4</sup>	166
GO:0023052	GO_Biological process	Signaling	5.63x10 <sup>-5</sup>	99
GO:0044281	GO_Biological process	Small molecule metabolic process	3.29x10 <sup>-8</sup>	62
GO:0006519	GO_Biological process	Cellular amino acid and derivative metabolic process	5.92x10 <sup>-5</sup>	21
GO:0043436	GO_Biological process	Oxoacid metabolic process	2.77x10 <sup>-9</sup>	37
GO:0019752	GO_Biological process	Carboxylic acid metabolic process	2.77x10 <sup>-9</sup>	37
GO:0009987	GO_Biological process	Cellular process	2.40x10 <sup>-8</sup>	258
GO:0046395	GO_Biological process	Carboxylic acid catabolic process	6.26x10 <sup>-9</sup>	16
GO:0009063	GO_Biological process	Cellular amino acid catabolic process	2.78x10 <sup>-6</sup>	10
GO:0044248	GO_Biological process	Cellular catabolic process	3.54x10 <sup>-2</sup>	25
GO:0009725	GO_Biological process	Response to hormone stimulus	4.39x10 <sup>-8</sup>	30
GO:0048545	GO_Biological process	Response to steroid hormone stimulus	2.46x10 <sup>-6</sup>	18
GO:0006807	GO_Biological process	Nitrogen compound metabolic process	2.95x10 <sup>-3</sup>	67
GO:0006520	GO_Biological process	Cellular amino acid metabolic process	4.99x10 <sup>-5</sup>	16
GO:0009605	GO_Biological process	Response to external stimulus	1.81x10 <sup>-8</sup>	35
GO:0009308	GO_Biological process	Amine metabolic process	4.15x10 <sup>-6</sup>	25
GO:0009310	GO_Biological process	Amine catabolic process	9.50x10 <sup>-6</sup>	10
GO:0032502	GO_Biological process	Developmental process	5.43x10 <sup>-7</sup>	110
GO:0048856	GO_Biological process	Anatomical structure development	8.77x10 <sup>-7</sup>	94
GO:0009653	GO_Biological process	Anatomical structure morphogenesis	1.19x10 <sup>-6</sup>	53
GO:0042221	GO_Biological process	Response to chemical stimulus	3.51x10 <sup>-10</sup>	70
GO:0010033	GO_Biological process	Response to organic substance	4.09x10 <sup>-11</sup>	52
GO:0010646	GO_Biological process	Regulation of cell communication	1.30x10 <sup>-5</sup>	48
GO:0008283	GO_Biological process	Cell proliferation	3.68x10 <sup>-6</sup>	26
GO:0044282	GO_Biological process	Small molecule catabolic process	2.35x10 <sup>-5</sup>	18
GO:0023046	GO_Biological process	Signaling process	8.53x10 <sup>-6</sup>	77
GO:0023060	GO_Biological process	Signal transmission	8.53x10 <sup>-6</sup>	77
GO:0042127	GO_Biological process	Regulation of cell proliferation	1.66x10 <sup>-6</sup>	41
GO:0048513	GO_Biological process	Organ development	5.37x10 <sup>-8</sup>	74
GO:0009888	GO_Biological process	Tissue development	2.37x10 <sup>-8</sup>	42
GO:0044106	GO_Biological process	Cellular amine metabolic process	1.70x10 <sup>-4</sup>	18
GO:0032787	GO_Biological process	Monocarboxylic acid metabolic process	3.81x10 <sup>-6</sup>	21
GO:0007165	GO_Biological process	Signal transduction	2.88x10 <sup>-6</sup>	71
GO:0010648	GO_Biological process	Negative regulation of cell communication	3.44x10 <sup>-6</sup>	22
GO:0051239	GO_Biological process	Regulation of multicellular organismal process	1.60x10 <sup>-6</sup>	48



Table II. Continued.

ID	Category	GO term	P-value	Count
GO:0042180	GO_Biological process	Cellular ketone metabolic process	$5.36 \times 10^{-9}$	37
GO:0050793	GO_Biological process	Regulation of developmental process	$1.88 \times 10^{-6}$	39
GO:0005515	GO_Molecular function	Protein binding	$2.06 \times 10^{-6}$	223
GO:0003674	GO_Molecular function	Molecular_function	$4.03 \times 10^{-4}$	359
GO:0005488	GO_Molecular function	Binding	$8.63 \times 10^{-8}$	316
GO:0048037	GO_Molecular function	Cofactor binding	$3.31 \times 10^{-6}$	19
GO:0005102	GO_Molecular function	Receptor binding	$1.45 \times 10^{-7}$	46
GO:0005575	GO_Cellular component	Cellular_component	$1.14 \times 10^{-2}$	370
GO:0005615	GO_Cellular component	Extracellular space	$1.96 \times 10^{-8}$	42
GO:0005576	GO_Cellular component	Extracellular region	$1.10 \times 10^{-5}$	73
GO:0044421	GO_Cellular component	Extracellular region part	$3.08 \times 10^{-9}$	52

GO, Gene Ontology.

Table III. KEGG pathway analysis of the target genes of miR-338-5p from Cytoscape.

ID	Name	Category	Term P-value	Count
GO:0000280	Valine, leucine and isoleucine degradation	KEGG	$5.1 \times 10^{-6}$	9
GO:0005200	Pathways in cancer	KEGG	$9.5 \times 10^{-5}$	25
GO:0004110	Cell cycle	KEGG	$1.4 \times 10^{-4}$	12
GO:0000071	Fatty acid degradation	KEGG	$1.7 \times 10^{-4}$	7
GO:0000640	Propanoate metabolism	KEGG	$2.0 \times 10^{-4}$	6

KEGG, Kyoto Encyclopedia of Genes and Genomes.

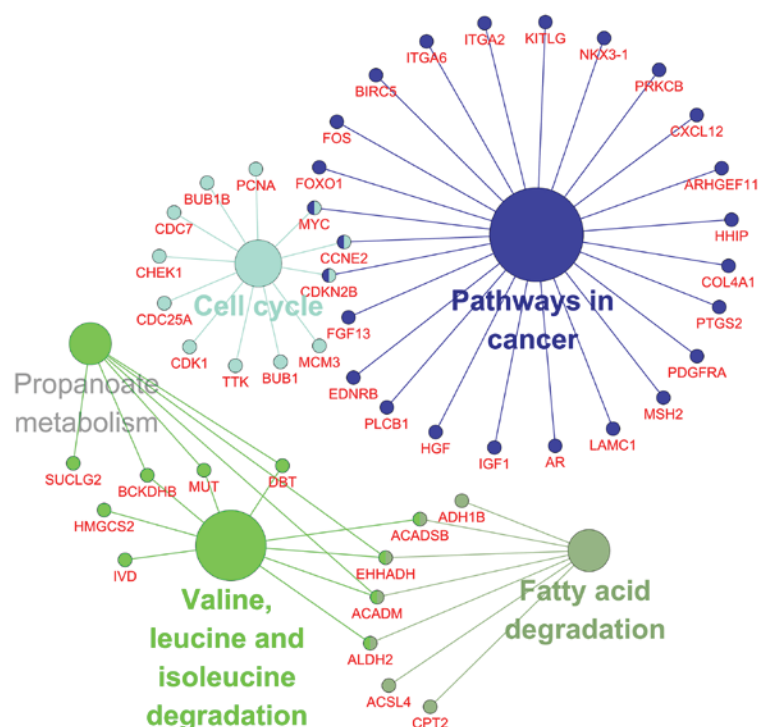


Figure 16. The network of KEGG pathway analysis. Five large circles in the network represented the corresponding significant pathways from KEGG. Small circles linked by the edges between the large circles symbolized the genes contained in these pathways and the common genes shared by multiple pathways were marked by a combination of colors. KEGG, Kyoto Encyclopedia of Genes and Genomes.

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