

# Identification of potential serum exosomal microRNAs involved in acinar-ductal metaplasia that is a precursor of pancreatic cancer associated with chronic pancreatitis

Li-Ping Sheng, MD, PhD, Chao-Qun Han, MD, PhD, Chi Nie, MD, Tao Xu, MD, Kun Zhang, MD, Xuan-Ji Li, MD, Xin-Ru Xie, MD, Rong Lin, MD, PhD, Zhen Ding, MD, PhD<sup>\*</sup>

# Abstract

**Backgrounds:** Due to difficulty in early diagnosis of chronic pancreatitis (CP), it is urgent to find novel biomarkers to detect CP. Exosomal microRNAs (Exo-miRNAs) located in the serum may be potential diagnostic and therapeutic targets for CP.

**Objective:** To identify differentially expressed Exo-miRNAs (DE-Exo-miRNAs) in the serum of CP patients, we performed a bioinformatics analysis.

**Methods:** The dataset GSE128508 was downloaded from the Gene Expression Omnibus (GEO) database. The analysis was carried out using BRB-ArrayTools and significance analysis of microarrays (SAM). The target genes of DE-S-Exo-miRNAs were predicted by miRWalk databases. Further gene ontology (GO) term and Kyoto Encyclopedia of Genomes (KEGG) pathway analyses were performed with plug-in ClueGO in Cytoscape software 3.7.0. Subsequently, the interaction regulatory network between encoded proteins of target genes was performed with the Search Tool for the Retrieval of Interacting Genes (STRING) database and analyzed using plug-in Molecular Complex Detection (MCODE) and cytoHubba in Cytoscape software 3.7.0.

**Results:** We identified 227 DE-Exo-miRNAs in the serum. Further analysis using the miRWalk database identified 5164 target genes of these miRNAs. The protein–protein interaction (PPI) regulatory network of 1912 potential target genes for hub 10 up-regulated miRNAs with high degrees and one down-regulated miRNAs were constructed using the STRING database and Cytoscape software. The functional analysis using Cytoscape software tool highlighted that target genes involved in pancreatic cancer. Acinar-ductal metaplasia (ADM) in the inflammatory environment of CP is a precursor of pancreatic cancer. Subsequently, we constructed a network of target genes associated with ADM and their miRNAs.

**Conclusions:** Exo-miRNAs in the serum as well as their target genes may be promising targets for the early diagnosis and treatment of CP. In addition, we identified potential Exo-miRNAs involved in ADM that is a precursor of pancreatic cancer associated with CP.

**Abbreviations:** ADM = acinar-ductal metaplasia, AIP = autoimmune pancreatitis, BTRC = beta-transducin repeat containing E3 ubiquitin protein ligase, CCNF = cyclin-F, CP = chronic pancreatitis, DE-Exo-miRNAs = differentially expressed Exo-miRNAs, DM = diabetes mellitus, Exo-miRNAs = Exosomal microRNAs, GEO = Gene Expression Omnibus, GO = gene ontology, HC = healthy control, KEGG = Kyoto Encyclopedia of Genes and Genomes, KRAS = Kirsten rat sarcoma 2 viral oncogene homolog, MAPK = mitogen-activated protein kinase, MCODE = molecular complex detection, miRNAs = microRNAs, MORC4 = MORC family CW-type zinc finger 4, NEDD4L = NEDD4 like E3 ubiquitin protein ligase, PDAC = pancreatic ductal adenocarcinoma, PDK1 = pyruvate dehydrogenase kinase 1, PPI = protein-protein interaction, PSCs = pancreatic stellate cells, RAC1 = Rac family small GTPase 1, RHOA = ras homolog family member A, RNF111 = ring finger protein 111, SAM = significance analysis of microarrays, SMAD3 = SMAD family member 3, STAT3 = signal transducer and activator of transcription 3, STRING = Search Tool for the Retrieval of Interacting Genes, TRPV1 = transient receptor potential cation channel subfamily V member 1, UBE2B = ubiquitin-conjugating

Editor: Jianxun Ding.

The authors have no conflicts of interest to disclose.

Department of Gastroenterology, Union Hospital, Tongji Medical College, Huazhong University of Science and Technology, Wuhan 430022, China.

Copyright © 2021 the Author(s). Published by Wolters Kluwer Health, Inc.

Received: 27 August 2020 / Received in final form: 10 January 2021 / Accepted: 8 April 2021

http://dx.doi.org/10.1097/MD.000000000025753

This study was supported in part by the National Natural Science Foundation of China (81770637).

Data available statement: Data comes from GEO database, which is a public open platform.

The datasets generated and/or analyzed during the current study are available from the corresponding author on reasonable request.

<sup>\*</sup> Correspondence: Zhen Ding, Department of Gastroenterology, Union Hospital, Tongji Medical College, Huazhong University of Science and Technology, 1277 Jiefang Avenue, Wuhan 430022, Hubei Province, China (e-mail: docd720@126.com).

This is an open access article distributed under the terms of the Creative Commons Attribution-Non Commercial License 4.0 (CCBY-NC), where it is permissible to download, share, remix, transform, and buildup the work provided it is properly cited. The work cannot be used commercially without permission from the journal.

How to cite this article: Sheng LP, Han CQ, Nie C, Xu T, Zhang K, Li XJ, Xie XR, Lin R, Ding Z. Identification of potential serum exosomal microRNAs involved in acinar-ductal metaplasia that is a precursor of pancreatic cancer associated with chronic pancreatitis. Medicine 2021;100:18(e25753).

enzyme E2 B, UBE2D3 = ubiquitin conjugating enzyme E2 D3, UBE2K = ubiquitin-conjugating enzyme E2 K, UBE2N = ubiquitinconjugating enzyme E2 N, UBE2S = ubiquitin-conjugating enzyme E2 S, VHL = von Hippel-Lindau tumor suppressor.

Keywords: acinar-ductal metaplasia, bioinformatics analysis, chronic pancreatitis, exosome, miRNA

# 1. Introduction

Chronic pancreatitis (CP) is a form of continuing fibroinflammatory disease associated with pathologically morphological changes of the pancreas, typically causing endocrine and exocrine dysfunction, especially in persons carrying genetic risk factors.<sup>[1]</sup> The diagnosis of CP can be difficulty in early phases of disease. Here, pancreatic functions are preserved and tests of laboratory and imaging may only display minimally abnormal results.<sup>[2]</sup> Thus, a comprehensive understanding about the molecular mechanism underlying CP and the identification of novel targets to diagnose CP in early phase are necessary.

MicroRNAs (miRNAs) are associated with acute or chronic pancreatitis.<sup>[3–7]</sup> miRNAs are single-stranded RNAs ~21–23-nucleotides in length which have important functions in almost every aspect of biology.<sup>[8]</sup> miRNAs play crucial gene regulatory roles by directing post-transcriptional repression of protein-coding genes.<sup>[9]</sup> The roles of miRNAs in many human diseases have made them promising targets for diagnostic and therapeutic strategies.<sup>[10–12]</sup> miRNAs have been discovered in body fluids, such as serum and plasma.<sup>[13,14]</sup> Additionally, researchers have also extracted miRNAs from exosomes.<sup>[15]</sup>

Exosomes are formed by the fusion of multivesicular bodies with the plasma membrane and are then released from the cell into the microenvironment.<sup>[16]</sup> Exosomes regulate many pathophysiological processes, such as inflammation, infection, immune responses, and tumor growth.<sup>[17]</sup> Exosomes contain proteins, lipid, and miRNAs.<sup>[18,19]</sup> As miRNAs in exosomes are prevented from RNase degradation, they are easier to detect compared with freely circulating miRNAs.<sup>[20]</sup> In addition, the miRNA expression profile in exosomes is different compared with the corresponding intracellular miRNAs.<sup>[21]</sup> Acute pancreatitis plasma exosomes contained more pro-inflammatory miRNA than exosomes from plasma control.<sup>[22]</sup> Therefore, exosomal microRNAs (Exo-miRNAs) in the serum could be potential biomarkers for diagnosing CP.

In this study, we analyzed the dataset GSE128508 from the public Gene Expression Omnibus (GEO) database. The samples displayed in the dataset are from the serum exosomes from CP patients, autoimmune pancreatitis (AIP) patients, and healthy control (HC) participants. Previous study identified 165 differentially expressed Exo-miRNAs (DE-Exo-miRNAs) when researchers focused on AIP and HC. However, they only analyzed 10 DE-Exo-miRNAs.<sup>[23]</sup> There, we analyzed more interesting DE-Exo-miRNAs in the serum by only comparing CP patients and HC participants. Subsequently, functional analysis of target genes for DE-Exo-miRNAs from the miRWalk databases was performed using Cytoscape software tool. Then the target genes of 10 up-regulated miRNAs with high degrees and one down-regulated miRNAs selected from miRNAs-mRNA networks were used to construct a protein-protein interaction (PPI) regulatory network. As functionally grouped network of Kyoto Encyclopedia of Genomes (KEGG) pathways show that pancreatic cancer is significant and acinar-ductal metaplasia (ADM) in the inflammatory environment of CP is a precursor of pancreatic cancer,<sup>[24]</sup> the regulatory network of target genes associated with ADM and their DE-Exo-miRNAs in the serum were also constructed.

# 2. Materials and methods

#### 2.1. Data acquisition

In this study, we retrieve datasets from public GEO database. The search keyword was "pancreatitis" and the selection was restricted to series. Finally, the miRNA dataset GSE128508 using the GPL21263 platform, which included results from 10 CP and 10 healthy adults' samples, were downloaded for further analyses. The characteristics of patients and healthy adults were shown in Table 1. Both miRNA profiles data and clinical characteristics of the healthy adults and CP patients are publicly available and open-access. Therefore, approval by a local ethics committee and informed consent were not needed.

### 2.2. Data processing and identifying serous DE-ExomiRNAs

The data from GSE128508 were analyzed by BRB-ArrayTools 4.6.0, R 4.0.1, and Significance Analysis of Microarrays (SAM) 5.0 to identify serous DE-Exo-miRNAs between CP and normal samples. We set Delta = 2.45 as threshold. For more information about statistic method, please visit the url https://brb.nci.nih.gov/BRB-ArrayTools/download.html and http://www-stat.stanford. edu/~tibs/SAM.

#### 2.3. Identification of serous DE-Exo-miRNA target genes

The miRWalk database is a large collection of predicted and experimentally verified binding sites information for miRNAstargets.<sup>[25]</sup> The information for DE-Exo-miRNA target genes was obtained from the miRWalk databases. We selected genes from the intersection between the Targetscan and miRDB databases, as well as experimentally verified miRTarBase database to increase the accuracy of the predictions.

# 2.4. Function analysis of target genes

Potential functions of target genes in biological process were analyzed using gene ontology (GO) term. KEGG is a database for the systematic analysis of gene functions and utilities in biological system from molecular-level information. We obtained informa-

Table 1   The characteristics of patients and healthy participants.							
Variable	Healthy adults (10)	CP patients (10)					
Gender, male: female Age, median (range), y	6:4 61.5 (49–82)	7:3 62 (50–78)					

CP = chronic pancreatitis.

tion from both the GO term and KEGG pathway using the plugin ClueGO (2.5.6) in Cytoscape software 3.7.0.<sup>[26]</sup>

# 2.5. Construction of miRNA-mRNA networks and miRNAs screening

Many gene targets of DE-Exo-miRNA in the serum may be associated with CP. The miRNA-mRNA networks were visualized with Cytoscape software 3.7.0. We identified hub miRNAs in networks according to their intra-network connectivity and correlation with their target genes. The10 up-regulated miRNAs with high degrees and 1 down-regulated miRNAs were identified from the networks.

# 2.6. Construction of PPI networks and significant submodule screening

The functional interactions between encoded proteins of target genes for 11 dysregulated miRNAs were explored using the Search Tool for the Retrieval of Interacting Genes (STRING) database.<sup>[27]</sup> We downloaded the interaction information from the STRING database and use Cytoscape software 3.7.0 to construct the PPI regulatory network. We applied the Molecular Complex Detection (MCODE) tool of Cytoscape software 3.7.0 to identify the significant submodule from the PPI network.<sup>[28]</sup> Furthermore, the hub genes were identified using cytoHubba plugin of Cytoscape software 3.7.0 from the significant submodule.<sup>[29]</sup> We set degree cutoff=2, node score cutoff= 0.2, K-core=2, and maximum depth=100 as the thresholds.

# 2.7. Construction of the regulatory network associated with ADM

KEGG pathways analysis involved in pancreatic cancer and pathways in cancer. Pancreas acinar cells show high plasticity and can undergo ADM after injury. Cells undergoing ADM are precursors for pancreatic intraepithelial neoplasia lesions that can further progress to pancreatic ductal adenocarcinoma (PDAC).<sup>[30]</sup> Among target genes of DE-Exo-miRNAs in the serum, signal transducer and activator of transcription 3 (*STAT3*), pyruvate dehydrogenase kinase 1 (*PDK1*), kirsten rat sarcoma 2 viral oncogene homolog (*KRAS*), Rac family small GTPase 1 (*RAC1*), and ras homolog family member A (*RHOA*) are key factors involving in ADM.<sup>[31–34]</sup> Therefore, we construct the regulatory network of these target genes and their DE-Exo-miRNAs.

## 3. Results

#### 3.1. DE-Exo-miRNAs in CP serum

A total of 227 DE-Exo-miRNAs in the serum were identified from the dataset GSE128508. Among these DE-Exo-miRNAs, 226 were significantly up-regulated and 1 were down-regulated. One down-regulated DE-Exo-miRNAs and the top 50 up-regulated DE-Exo-miRNAs in the serum can be found summarized in Table 2. In supplementary materials, all 227 DE-Exo-miRNAs in the serum are listed.

#### 3.2. Target genes for serous DE-Exo-miRNAs

We utilized the miRWalk database to screen for target genes for the 227 DE-Exo-miRNAs in the serum to explore the potential functions of these DE-Exo-miRNAs. To increase the accuracy of the search, the intersection between the TargetScan and miRDB databases, as well as experimentally verified genes in the miRTarBase database, were selected as target genes. Finally, we obtained 5164 target genes for the 227 serous DE-ExomiRNAs. Among the target genes, many of them are associated with CP, such as SMAD family member 3 (*SMAD3*), transient receptor potential cation channel subfamily V member 1 (*TRPV1*), MORC family CW-type zinc finger 4 (*MORC4*), and so on.<sup>[35,36]</sup> The DE-Exo-miRNAs (hsa-miR-491–5p, hsa-miR-6807–5p, and hsa-miR-6731-5p, etc) may be valuable for the diagnosis and treatment of CP by their target genes.

## 3.3. Function analysis of target genes for serous DE-ExomiRNAs

The function analysis indicated that the enriched GO terms of 5164 target genes primarily included those involved in the cell cycle process, cellular response to signal transduction by p53 class mediator, lymphocyte mediated immunity, regulation of cell death, regulation of cell differentiation, regulation of gene expression, regulation of cell communication, regulation of cellular metabolic process, and regulation of B cell activation, amongst others. The typical 20 GO terms are summarized in Fig. 1.

The KEGG pathways analysis revealed that target genes were involved in numerous different pathways, including pancreatic cancer, pathways in cancer, p53 signaling pathway, mitogenactivated protein kinase (MAPK) signaling pathway, cellular senescence, and signaling pathways regulating pluripotency of stem cells, amongst others. These pathways are shown in Fig. 2.

#### 3.4. The serous DE-Exo-miRNAs with high degrees

We constructed the miRNAs-mRNA networks between the 227 DE-Exo-miRNAs and 5164 target genes using Cytoscape software 3.7.0. The hub 10 up-regulated miRNAs with high degrees identified from the miRNAs-mRNA networks are summarized in Table 3.

# 3.5. The significant submodule from the PPI regulatory network

The 10 up-regulated miRNAs, one down-regulated miRNAs (hsa-miR-5100), and their 1912 target genes were explored further. Using the interaction information of the target genes obtained from the STRING database, the PPI regulatory network of 1912 target genes for 11 key serous DE-Exo-miRNAs were constructed. We selected a significant submodule with the highest scores of 31.226 from the PPI regulatory network (Fig. 3). Furthermore, hub genes were identified from the significant submodule, including von Hippel-Lindau tumor suppressor (VHL), NEDD4 like E3 ubiquitin protein ligase (NEDD4L), ubiquitin-conjugating enzyme E2N (UBE2N), ubiquitin-conjugating enzyme E2 B (UBE2B), beta-transducin repeat containing E3 ubiquitin protein ligase (BTRC), ubiquitin conjugating enzyme E2 D3 (UBE2D3), ubiquitin-conjugating enzyme E2 K (UBE2K), ring finger protein 111 (RNF111), cyclin-F (CCNF), ubiquitin-conjugating enzyme E2S (UBE2S).

# 3.6. Regulatory network of target genes associated with ADM and their miRNAs

Among the target genes of DE-Exo-miRNAs in the serum, STAT3, PKD1, KRAS, RAC1, and RHOA are key factors

# Table 2

### List of partial DE-Exo-miRNAs in the serum<sup>\*</sup>.

	е	Numerator	Denominator	Fold change
Down-regulated microBNAs				
hsa-miB-5100	-4.20	-1.32	0.31	0.82
Up-regulated microBNAs (Top 50)	1.20	1.62	0.01	0.02
hsa-miR-5006-5n	13.60	1.65	0.12	1.50
hsa-miR-6748-5n	13.56	2 11	0.16	1.63
hsa-miR-4485-5n	12.34	1 74	0.14	1.52
hsa-miR-5192	11 75	2 34	0.20	1.02
hsa-miR-0102	11.73	2.01	0.19	1.70
hsa mir +314 hsa-miR-6760-5n	11.75	2.22	0.25	1.07
hsa-miR-6165	10.07	2.00	0.23	1.63
hsa miP 4208	10.05	2.44	0.22	1.00
hsa-min-4230	10.95	2.44	0.22	1.00
haa miD 6997 5n	10.92	2.19	0.20	1.02
115a-1111n-0007-5µ	10.91	3.20	0.30	1.70
$HSa-HIIR-0030-3\mu$	10.00	1.71	0.10	1.01
115a-1111R-0020-0µ	10.62	1.97	0.18	1.09
IISa-IIIIR-370-3p	10.00	2.30	0.23	1.09
nsa-miR-6766-5p	C0.01	2.31	0.22	1.01
nsa-miR-3151-5p	10.64	1.95	0.18	1.57
nsa-miR-6795-5p	10.60	2.20	0.21	1.54
nsa-miR-3622b-5p	10.50	1.66	0.16	1.50
nsa-miR-4419b	10.49	2.51	0.24	1.61
hsa-miR-513a-5p	10.41	2.01	0.19	1.61
hsa-miR-3154	10.41	2.11	0.20	1.55
hsa-miR-6826-5p	10.22	2.01	0.20	1.55
hsa-miR-4728-5p	10.08	2.81	0.28	1.66
hsa-miR-4727-3p	9.93	1.76	0.18	1.51
hsa-miR-3619-3p	9.87	2.74	0.28	1.49
hsa-miR-6849-5p	9.86	2.34	0.24	1.56
hsa-miR-6717-5p	9.82	2.74	0.28	1.60
hsa-miR-5010-5p	9.81	1.68	0.17	1.50
hsa-miR-491-5p	9.77	1.65	0.17	1.50
hsa-miR-6746-5p	9.69	2.29	0.24	1.51
hsa-miR-6890-5p	9.66	1.87	0.19	1.54
hsa-miR-6824-5p	9.55	2.13	0.22	1.53
hsa-miR-6757-5p	9.54	2.51	0.26	1.54
hsa-miR-2110	9.53	1.80	0.19	1.50
hsa-miR-4257	9.51	2.13	0.22	1.48
hsa-miR-6731-5p	9.44	1.86	0.20	1.52
hsa-miR-3714	9.40	1.93	0.21	1.54
hsa-miR-4462	9.29	1.95	0.21	1.50
hsa-miR-920	9.28	2.58	0.28	1.53
hsa-miR-7846-3p	9.10	2.19	0.24	1.55
hsa-miR-7847-3p	9.08	2.43	0.27	1.55
hsa-miR-6812-5p	9.00	2.04	0.23	1.53
hsa-miR-3934-5p	8.84	1.90	0.21	1.49
hsa-miR-583	8.83	1.45	0.16	1.44
hsa-miR-5189-5p	8.83	1.83	0.21	1.51
hsa-miR-6777-5p	8.71	2.28	0.26	1.51
hsa-miR-6511b-5p	8.68	1.79	0.21	1.47
hsa-miR-8059	8.66	2.29	0.26	1.46
hsa-miR-551b-5p	8.61	1.83	0.21	1.51
hsa-miB-6769b-5p	8.61	1 79	0.21	1 43
hsa-miR-6756-5p	8.59	2.03	0.24	1.36

\* All 227 DE-Exo-miRNAs in the serum are listed in supplementary materials. DE-Exo-miRNAs = differentially expressed exosomal miRNAs, Denominator = The denominator of the T-statistic, Numerator = The numerator of the T-statistic, Score = The T-statistic value.

participating in ADM. miRNAs play crucial gene regulatory roles by directing post-transcriptional repression of protein-coding genes. The expression of these target genes may be repressed by the DE-S-Exo-miRNAs in the serum (Fig. 4).

### 4. Discussion

The molecular mechanisms of CP are complex and diagnosis can be difficulty in early phases of the disease.<sup>[2]</sup> Therefore, it is of great importance to understand the molecular mechanism of CP,



Figure 1. GO function annotation of miRNA target genes. Levels of GO terms increase from the upper to the lower. GO=gene ontology, miRNAs=microRNAs.



Figure 2. Functionally grouped network of KEGG pathways of serous DE-ExomiRNA target genes. The node size stands for the pathway significance. The node size increases when the *P*-value of pathways decreases. KEGG=Kyoto Encyclopedia of Genes and Genomes. as well as identify new targets to diagnose CP. Exo-miRNAs in the serum could be effective biomarkers for CP diagnosis, as they are stable and informative. Here, we downloaded the publicly accessible CP patient dataset GSE128508 and performed bioinformatic analysis. Previous study focused on AIP and only identified 10 differentially expressed Exo-miRNAs (DE-ExomiRNAs).<sup>[23]</sup> When only comparing CP and normal samples, we

Table 3

# The top 10 miRNAs with high degrees from miRNAs-mRNA networks.

Rank	miRNA	Degree	Number of target genes	Feature			
1	hsa-miR-24-3p	516	516	Up			
2	hsa-miR-149-3p	449	449	Up			
3	hsa-miR-6785-5p	448	448	Up			
4	hsa-miR-4728-5p	446	446	Up			
5	hsa-miR-6808-5p	402	402	Up			
6	hsa-miR-6779-5p	388	388	Up			
7	hsa-miR-6799-5p	384	384	Up			
8	hsa-miR-6086	343	343	Up			
9	hsa-miR-4722-5p	286	286	Up			
10	hsa-miR-4433a-3p	245	245	Up			

Degree stands for miRNA intra-network connectivity and correlation with their target genes. The miRNA with most high-degree are the most hub node in the network. miRNAs = microRNAs.



Figure 3. The submodule from the PPI network. The hub genes (nodes) are shown with red color. PPI = protein-protein interaction.



Figure 4. The Regulatory network of target genes associated with ADM and their miRNAs. The target genes (nodes) are shown with red color. The expression of these genes may be repressed by DE-S-Exo-miRNAs in the serum. ADM=acinar-ductal metaplasia, miRNAs=microRNAs.

identified 227 interesting DE-Exo-miRNAs in the serum. Moreover, we identified 5164 target genes of the 227 dysregulated Exo-miRNAs in CP serum. GO term and KEGG pathway analysis of these 5164 target genes indicated that they involved in the regulation of numerous biological processes and signaling pathways such as pancreatic cancer, amongst many others. ADM in the inflammatory environment of pancreatitis is a precursor of pancreatic cancer. The expression of the target genes associated with ADM (STAT3, PKD1, KRAS, RAC1, and RHOA) may be repressed by DE-Exo-miRNAs in the serum, which may contribute to blocking ADM initiation.

GO term analysis indicated that these target genes were enriched in the cell cycle process, cellular response to signal transduction by p53 class mediator, lymphocyte mediated immunity, regulation of cell death, regulation of cell differentiation, regulation of gene expression, regulation of cell communication, regulation of cellular metabolic process, and regulation of B cell activation, amongst others. These biological processes are associated with CP. Cyclin D1, an important regulator of the G1 phase of the cell cycle, increases in the pancreas of patients with CP, which may contribute to a hyperproliferative state in CP.<sup>[37]</sup> Inhibition of p53 expression decreased levels of inflammatory factors and suppressed progression of CP in the mouse model.<sup>[38]</sup> The chronic inflammatory process in CP markedly affects immune function in CP patients.<sup>[39]</sup> Pancreatitis also leads cell death, such as apoptosis and necrosis.<sup>[40]</sup> Pancreatic acinar cells undergoing ADM contribute to the regeneration of the pancreas in response to injury.<sup>[41]</sup> The target genes of miRNA in serum exosomes may promote ADM by regulation of cell differentiation and gene expression. Cell communication of pancreatic acinar cells is associated with basal release of amylase.<sup>[42]</sup> The disruption of the metabolism at a molecular level occurs in the pancreas during pancreatitis.<sup>[43]</sup> Finally, the pancreatic B cell can secrete insulin.<sup>[44]</sup> The target genes of miRNA in serum exosomes may remedy or exacerbate diabetes mellitus (DM) secondary to chronic pancreatitis by regulation of B cell activation.

The information regarding KEGG pathway analysis indicated that the target genes involve in pancreatic cancer, pathways in cancer, p53 signaling pathway, MAPK signaling pathway, cellular senescence, and signaling pathways regulating pluripotency of stem cells, amongst others. Chronic pancreatitis increases the risk of pancreatic cancer.<sup>[45]</sup> p53 is one of the most important tumor suppressor genes that is frequently mutated in human cancers.<sup>[46]</sup> p53 mutations can be detected in some cases of CP.<sup>[47]</sup> The target genes associated with pancreatic cancer, pathways in cancer, and p53 signaling pathway may contribute to pancreatic cancer associated with CP. As the major effector cells, activation of pancreatic stellate cells (PSCs) contributes to fibrosis of PC.<sup>[48]</sup> PSCs can be activated by many cellular signals, such as MAPK pathway.<sup>[49]</sup> The number of senescent PSCs are associated with the severity of inflammation and the extension of fibrosis of CP.<sup>[50]</sup> Pancreatic acinar cells show plasticity and dedifferentiate to an embryonic progenitor phenotype in experimental chronic pancreatitis.<sup>[51]</sup> Signaling pathways regulating pluripotency of stem cells may involve in this process. Therefore, a comprehensive understanding of the KEGG pathway analysis could be helpful in exploring complicated mechanisms underlying CP.

We selected the top 10 up-regulated miRNAs with high degrees and 1 up-regulated miRNAs to constructed a PPI regulatory network and selected a significant submodule with the highest scores from the PPI regulatory network. Furthermore, hub genes were identified from the significant submodule. Our current understanding for the functions of hue genes in CP are limited. As tumor suppressor gene, VHL inhibits the proliferation, migration, and invasion of pancreatic cancer cells.<sup>[52]</sup> Moreover, UBE2S regulates pancreatic cancer cell metastasis and invasion by epithelial-mesenchymal transition.<sup>[53]</sup> VHL and UBE2S may involve in pancreatic cancer associated with CP. The roles of other hub genes in CP need to be researched further in future studies.

Functionally grouped network of KEGG pathways show that pancreatic cancer and pathways in cancer are significant. ADM in the inflammatory environment of CP is a precursor of pancreatic cancer. Among the target genes of DE-Exo-miRNAs in the serum, STAT3, PKD1, KRAS, RAC1, and RHOA are key factors participating in ADM. STAT3 and PDK1 are critical effectors of oncogenic KRAS in the pancreas, mediating ADM formation.<sup>[31,32]</sup> Activation of Rac1 is required for transdifferentiation of acinar cells into ADM.<sup>[33]</sup> Moreover, activation of RhoA/ ROCK signaling pathway promotes the formation of ADM.<sup>[34]</sup> We construct the regulatory network of these target genes and their DE-Exo-miRNAs in the serum. Nevertheless, the relationship between DE-Exo-miRNAs and target genes associated with ADM require future experimental verification and could provide interesting therapeutic leads of preventing pancreatic cancer associated with CP.

As a class of non-coding RNAs, one major function of miRNAs is the post-transcriptional repression of their target genes and they may be potential therapeutic targets for CP by interfering with the relative genetic factors.<sup>[54]</sup> miRNA-21 and connective tissue growth factor (CCN2) in PSC-derived exosomes are newly identified participants in PSC fibrogenic signaling during CP.<sup>[55]</sup> The exosomes may easily transport miRNAs to their target cells by encapsulating and protecting them. Therefore, studying the functions of DE-Exo-miRNAs in the serum and target genes is a promising avenue to identifying novel therapeutics.

In our work, we use bioinformatics analysis to identify serous DE-Exo-miRNAs between CP and normal samples. We then obtained their target genes from the miRWalk database. Functional analysis of these target genes as well as interactions between them provided insights into the molecular mechanism of CP. We constructed a network of target genes associated with ADM and their DE-Exo-miRNAs in the serum, as the interactions between these could likely underlie ADM in CP that is a precursor of pancreatic cancer. Nevertheless, further experimental validation is required for these predictions, which was a limitation for our study. Additionally, the fold changes of 227 DE-Exo-miRNAs are lower than 1.78 and detection of the difference is very difficult. That is another limitation.

In summary, Exo-miRNAs in the serum and their target genes may be potential tools for the early diagnosis and treatment of CP. In addition, we identified potential Exo-miRNAs (hsa-miR-4270, hsa-miR-4462, hsa-miR-3622b-5p, hsa-miR-6088, and hsa-miR-3934-5p, etc.) involved in ADM that is a precursor of pancreatic cancer associated with CP and their target genes (*STAT3*, *PKD1*, *RAC1*, *KRAS*, and *RHOA*).

### 5. Authors' contributions

Conceptualization: Li-Ping Sheng. Data curation: Li-Ping Sheng. Methodology: Li-Ping Sheng. Project administration: Zhen Ding. Supervision: Xuan-Ji Li, Xin-Ru Xie, Rong Lin.

Software: Li-Ping Sheng, Chi Nie, Tao Xu, Kun Zhang, Xuan-Ji Li.

Validation: Li-Ping Sheng, Chi Nie, Tao Xu, Kun Zhang, Xuan-Ji Li, Xin-Ru Xie.

Visualization: Li-Ping Sheng, Tao Xu, Kun Zhang, Xin-Ru Xie.

Writing – original draft: Zhen Ding.

Writing - review & editing: Zhen Ding.

#### References

- [1] Whitcomb DC, Frulloni L, Garg P, et al. Chronic pancreatitis: an international draft consensus proposal for a new mechanistic definition. Pancreatology 2016;16:218–24.
- [2] Gupte A, Goede D, Tuite R, et al. Chronic pancreatitis. BMJ 2018;361: k2126.
- [3] Yu P, Liu K, Gao X, et al. Transforming growth factor-β and bone morphogenetic protein 2 regulation of MicroRNA-200 family in chronic pancreatitis. Pancreas 2018;47:252–6.
- [4] Wang Z, Li D, Jin J, et al. Association between microRNA polymorphisms and chronic pancreatitis. Pancreatology 2016;16:244–8.

[5] Xin L, Gao J, Wang D, et al. Novel blood-based microRNA biomarker panel for early diagnosis of chronic pancreatitis. Sci Rep 2017;7:40019.

- [6] Wang D, Xin L, Lin J-H, et al. Identifying miRNA-mRNA regulation network of chronic pancreatitis based on the significant functional expression. Medicine (Baltimore) 2017;96:e6668.
- [7] Fan L, Hui X, Mao Y, et al. Identification of acute pancreatitis-related genes and pathways by integrated bioinformatics analysis. Dig Dis Sci 2020;65:1720–32.
- [8] Li Z, Rana TM. Therapeutic targeting of microRNAs: current status and future challenges. Nat Rev Drug Discov 2014;13:622–38.
- [9] Bartel DP. MicroRNAs: target recognition and regulatory functions. Cell 2009;136:215–33.
- [10] Rupaimoole R, Slack FJ. MicroRNA therapeutics: towards a new era for the management of cancer and other diseases. Nat Rev Drug Discov 2017;16:203–22.
- [11] Párrizas M, Novials A. Circulating microRNAs as biomarkers for metabolic disease. Best Pract Res Clin Endocrinol Metab 2016;30:591– 601.
- [12] Wang J, Liew OW, Richards AM, et al. Overview of MicroRNAs in cardiac hypertrophy, fibrosis, and apoptosis. Int J Mol Sci 2016;17:749.
- [13] Ortiz-Quintero B. Cell-free microRNAs in blood and other body fluids, as cancer biomarkers. Cell Prolif 2016;49:281–303.
- [14] Arner P, Kulyté A. MicroRNA regulatory networks in human adipose tissue and obesity. Nat Rev Endocrinol 2015;11:276–88.
- [15] Valadi H, Ekström K, Bossios A, et al. Exosome-mediated transfer of mRNAs and microRNAs is a novel mechanism of genetic exchange between cells. Nat Cell Biol 2007;9:654–9.
- [16] Mathivanan S, Ji H, Simpson RJ. Exosomes: extracellular organelles important in intercellular communication. J Proteomics 2010;73:1907– 20.
- [17] Barile L, Vassalli G. Exosomes: therapy delivery tools and biomarkers of diseases. Pharmacol Ther 2017;174:63–78.
- [18] Théry C, Zitvogel L, Amigorena S. Exosomes: composition, biogenesis and function. Nat Rev Immunol 2002;2:569–79.
- [19] Ferguson SW, Nguyen J. Exosomes as therapeutics: the implications of molecular composition and exosomal heterogeneity. J Control Release 2016;228:179–90.
- [20] Gallo A, Tandon M, Alevizos I, et al. The majority of microRNAs detectable in serum and saliva is concentrated in exosomes. PLoS One 2012;7:e30679.
- [21] Squadrito ML, Baer C, Burdet F, et al. Endogenous RNAs modulate microRNA sorting to exosomes and transfer to acceptor cells. Cell Rep 2014;8:1432–46.
- [22] Jiménez-Alesanco A, Marcuello M, Pastor-Jiménez M, et al. Acute pancreatitis promotes the generation of two different exosome populations. Sci Rep 2019;9:19887.
- [23] Nakamaru K, Tomiyama T, Kobayashi S, et al. Extracellular vesicles microRNA analysis in type 1 autoimmune pancreatitis: increased expression of microRNA-21. Pancreatology 2020;20:318–24.
- [24] Pinho AV, Chantrill L, Rooman I. Chronic pancreatitis: a path to pancreatic cancer. Cancer Lett 2014;345:203–9.

- [25] Dweep H, Gretz N. miRWalk2.0: a comprehensive atlas of microRNAtarget interactions. Nat Methods 2015;12:697.
- [26] Bindea G, Mlecnik B, Hackl H, et al. ClueGO: a Cytoscape plug-in to decipher functionally grouped gene ontology and pathway annotation networks. Bioinformatics 2009;25:1091–3.
- [27] Szklarczyk D, Gable AL, Lyon D, et al. STRING v11: protein-protein association networks with increased coverage, supporting functional discovery in genome-wide experimental datasets. Nucleic Acids Res 2019;47:D607–13.
- [28] Bader GD, Hogue CWV. An automated method for finding molecular complexes in large protein interaction networks. BMC Bioinformatics 2003;4:2.
- [29] Chin C-H, Chen S-H, Wu H-H, et al. cytoHubba: identifying hub objects and sub-networks from complex interactome. BMC Syst Biol 2014;8 (suppl):S11.
- [30] Storz P. Acinar cell plasticity and development of pancreatic ductal adenocarcinoma. Nat Rev Gastroenterol Hepatol 2017;14:296–304.
- [31] Corcoran RB, Contino G, Deshpande V, et al. STAT3 plays a critical role in KRAS-induced pancreatic tumorigenesis. Cancer Res 2011;71:5020– 9.
- [32] Eser S, Reiff N, Messer M, et al. Selective requirement of PI3K/PDK1 signaling for Kras oncogene-driven pancreatic cell plasticity and cancer. Cancer Cell 2013;23:406–20.
- [33] Saponara E, Visentin M, Baschieri F, et al. Serotonin uptake is required for Rac1 activation in Kras-induced acinar-to-ductal metaplasia in the pancreas. J Pathol 2018;246:352–65.
- [34] Tao X, Chen Q, Li N, et al. Serotonin-RhoA/ROCK axis promotes acinar-to-ductal metaplasia in caerulein-induced chronic pancreatitis. Biomed Pharmacother 2020;125:109999.
- [35] Liu L, Zhu Y, Noë M, et al. Neuronal transforming growth factor beta signaling via SMAD3 contributes to pain in animal models of chronic pancreatitis. Gastroenterology 2018;154:2252.e2–65.e2.
- [36] Zhu Y, Colak T, Shenoy M, et al. Nerve growth factor modulates TRPV1 expression and function and mediates pain in chronic pancreatitis. Gastroenterology 2011;141:370–7.
- [37] Kornmann M, Ishiwata T, Arber N, et al. Increased cyclin D1 expression in chronic pancreatitis. Pancreas 1998;17:158–62.
- [38] Zhou L, Tan J-H, Cao R-C, et al. ATF6 regulates the development of chronic pancreatitis by inducing p53-mediated apoptosis. Cell Death Dis 2019;10:662.
- [39] Gansauge F, Gansauge S, Eh M, et al. Distributional and functional alterations of immunocompetent peripheral blood lymphocytes in patients with chronic pancreatitis. Ann Surg 2001;233:365–70.
- [40] Gukovskaya AS, Pandol SJ. Cell death pathways in pancreatitis and pancreatic cancer. Pancreatology 2004;4:567–86.
- [41] Puri S, Folias AE, Hebrok M. Plasticity and dedifferentiation within the pancreas: development, homeostasis, and disease. Cell Stem Cell 2015;16:18–31.
- [42] Meda P, Bruzzone R, Knodel S, et al. Blockage of cell-to-cell communication within pancreatic acini is associated with increased basal release of amylase. J Cell Biol 1986;103:475–83.
- [43] Ma C, Tian B, Wang J, et al. Metabolic characteristics of acute necrotizing pancreatitis and chronic pancreatitis. Mol Med Rep 2012;6:57–62.
- [44] Petit P, Loubatières-Mariani MM. Potassium channels of the insulinsecreting B cell. Fundam Clin Pharmacol 1992;6:123–34.
- [45] Kirkegård J, Mortensen FV, Cronin-Fenton D. Chronic pancreatitis and pancreatic cancer risk: a systematic review and meta-analysis. Am J Gastroenterol 2017;112:1366–72.
- [46] Hong B, van den Heuvel APJ, Prabhu VV, et al. Targeting tumor suppressor p53 for cancer therapy: strategies, challenges and opportunities. Curr Drug Targets 2014;15:80–9.
- [47] Gansauge S, Schmid RM, Muller J, et al. Genetic alterations in chronic pancreatitis: evidence for early occurrence of p53 but not K-ras mutations. Br J Surg 1998;85:337–40.
- [48] Apte MV, Haber PS, Darby SJ, et al. Pancreatic stellate cells are activated by proinflammatory cytokines: implications for pancreatic fibrogenesis. Gut 1999;44:534–41.
- [49] Jin G, Hong W, Guo Y, et al. Molecular mechanism of pancreatic stellate cells activation in chronic pancreatitis and pancreatic cancer. J Cancer 2020;11:1505–15.
- [50] Fitzner B, Müller S, Walther M, et al. Senescence determines the fate of activated rat pancreatic stellate cells. J Cell Mol Med 2012;16:2620–30.
- [51] Pinho AV, Rooman I, Reichert M, et al. Adult pancreatic acinar cells dedifferentiate to an embryonic progenitor phenotype with concomitant

activation of a senescence programme that is present in chronic pancreatitis. Gut 2011;60:958-66.

- [52] Sun J, Jiang Z, Li Y, et al. Downregulation of miR-21 inhibits the malignant phenotype of pancreatic cancer cells by targeting VHL. Onco Targets Ther 2019;12:7215–26.
- [54] Hu L-H, Ji J-T, Li Z-S. Potential application of miRNAs as diagnostic and therapeutic tools in chronic pancreatitis. J Cell Mol Med 2015;19:2049–57.
- [55] Charrier A, Chen R, Chen L, et al. Connective tissue growth factor (CCN2) and microRNA-21 are components of a positive feedback loop in pancreatic stellate cells (PSC) during chronic pancreatitis and are exported in PSC-derived exosomes. J Cell Commun Signal 2014;8: 147–56.