Identification and Validation of Hub Genes in Acute Pancreatitis and Hypertriglyceridemia

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Background: The pathogenesis of acute pancreatitis (AP) and the relationship between acute pancreatitis and hypertriglyceridemia are complex and not fully understood. The purpose of this study was to identify the hub genes along with common differentially expressed genes (DEGs) between acute pancreatitis and hypertriglyceridemia.

Methods: We downloaded three gene expression profiles of AP and one gene expression profile of hypertriglyceridemia from the Gene Expression Omnibus (GEO) database and filtered the DEGs based on the above four datasets. Next, we identified the hub genes by performing the Gene Ontology (GO) term analysis, Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis, and protein–protein interaction (PPI) construction. We also constructed the miRNA-hub gene network and established mouse models with hypertriglyceridemia and AP using a high-fat diet and injection of caerulein (CAE), respectively. Finally, the immunohistochemical analysis was used to verify the differential expressions of hub genes in AP, hypertriglyceridemia, and normal pancreatic tissue.

Results: A total of 105 DEGs associated with AP and 149 DEGs associated with hypertriglyceridemia were identified. Additionally, we identified six hub genes of AP, all of which were closely related to the cytoskeleton while two DEGs genes were common in both AP and hypertriglyceridemia. We also verified their expression in mouse models. Finally, a network of miRNA-mRNA was also constructed, and the top seven interactive miRNAs (hsa-mir-1–3p, hsa-mir-5195–3p, hsa-mir-145–5p, hsa-let-7b-5p, hsa-mir-10b-5p, hsa-mir-206, and hsa-mir-613) targeting the most hub genes were identified.

Conclusion: Overall, we identified six hub genes associated with AP and two common DEGs associated with AP and hypertriglyceridemia along with seven miRNAs that may regulate AP. This study could provide new ideas for further elucidation of the pathogenesis of hypertriglyceridemia-induced acute pancreatitis in the future.

Keywords: acute pancreatitis, biomarkers, hypertriglyceridemia-induced pancreatitis, bioinformatics

Introduction
Acute pancreatitis (AP) is a common disease found in clinics, and requires urgent Hospital admission. The incidence of AP is increasing in recent years worldwide. The patients with AP increased from 1,727,789.3 to 2,814,972.3 between 1990 and 2019 in 204 countries and territories. Meanwhile, nearly 20% of AP patients will progress to moderate or severe acute pancreatitis. Severe AP is life-threatening and is one of the most common causes of morbidity and mortality in the world. The prevalence rate involves 40 cases per 100,000 adults per year in the United States.

Gallstones, chronic alcohol abuse, hypertriglyceridemia are the major three etiologies of AP. Additionally, the percentage of hypertriglyceridemia-induced pancreatitis (HTGP) is rapidly rising, which ranks 2 in China. The complication rate observed in HTG patients caused by other causes was significantly higher than that of the AP patients. Despite advances in treatment, no significant decline has been observed in the mortality rate of AP patients. Therefore, it is important to further explore the pathogenesis of AP, especially HTGP.

With the progress of the Human Genome Project, gene microarray has been playing an increasingly important role in the identification of genes leading to various diseases and has already been widely used in disease diagnosis.
changes also play an important role in the pathogenesis of AP\textsuperscript{12} and HTG,\textsuperscript{15} but so far, none have studied the common differential genes (DEGs) in hypertriglyceridemia and AP. CAE, a cholecystokinin analog, has been widely used to induce AP in vivo\textsuperscript{16} and in vitro.\textsuperscript{17–19} Animal models of hypertriglyceridemia can also be successfully induced using a high-fat diet.\textsuperscript{20} Therefore, it is feasible to identify differentially expressed genes associated with acute pancreatitis and hypertriglyceridemia by analyzing data related to these two diseases in online databases, and verifying the DEGs using vivo models.

In this study, three gene expression profiles of AP (GSE54129, GSE79973, and GSE13911) and one gene expression profile of hypertriglyceridemia (GSE1010) were downloaded from the Gene Expression Omnibus (GEO) database while the DEGs were filtered based on the above four datasets. Next, the Gene Ontology (GO) term analysis, Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis, and protein–protein interaction (PPI) construction were performed to identify the hub genes. The miRNA-hub gene network was also constructed, and mouse models of hypertriglyceridemia and AP were established using a high-fat diet and injection of CAE, respectively. Finally, immunohistochemistry was used to verify the differential expression of hub genes in AP, hypertriglyceridemia, and normal pancreatic tissue.

**Materials and Methods**

**Downloading and Processing of the Data**

Three microarray datasets of AP (GSE3644, GSE121038, and GSE65146) and one microarray dataset of hypertriglyceridemia (GSE1010) were downloaded from the GEO (https://www.ncbi.nlm.nih.gov/gds/, accessed time: May 16, 2021) and preprocessed using GEOquery\textsuperscript{21} and affy\textsuperscript{22} R packages, respectively. These datasets included the samples of AP or hypertriglyceridemia and their corresponding controls.

**Identification of DEGs**

In the case of the multiple probe sets corresponding to the same gene, the mean value of the expression of those probe sets was calculated. After the probe ID was converted to gene symbol, the differential analysis was performed using the Limma package\textsuperscript{23} of R software. The fold-change was regarded as significant with adjusted P < 0.05 and with threshold values >1.0 or <−1.0. Additionally, the expression of DEGs was visualized through packages pheatmap and ggplot2 of R software.

**Functional Enrichment Analysis**

The Database for Annotation, Visualization and Integrated Discovery (DAVID) (version 6.8, https://david.ncifcrf.gov/, accessed time: May 25, 2021) can provide a comprehensive set of functional annotation tools for investigators to understand the biological meaning behind a large list of genes. We submitted the upregulated and downregulated DEGs to the DAVID online program and performed the GO and KEGG enrichment analyses. Next, we visualized the genes using the ggplot2 package of R software, where P<0.05 was considered statistically significant.

**Construction and Module Analysis of PPI Network of the DEGs**

The PPI network analysis was performed using the Search Tool for the Retrieval of Interacting Genes/Proteins (STRING; https://string-db.org/, accessed time: May 26, 2021) with an interaction score > 0.4. Furthermore, the interaction network between the DEGs and their related genes was presented with the minimum number of two interactions. We used Cytoscape software (version 3.8.2) to visualize the PPI network and analyzed the main modules of the PPI network using the MCODE plugin of Cytoscape. The connectivity scores were calculated using the Cytoscape plugin cytoHubba, and the top six genes were defined as the hub genes.

**Validation of Expression of the Hub Genes in Different Human Organs**

The genotype-tissue expression (GTEx) project is an open access database containing gene expression levels in 328 healthy human pancreatic tissues (https://www.gtexportal.org/home/, accessed time: May 28, 2021). We applied this...
online tool to evaluate the RNA expression level of hub genes in different human organs, especially in the pancreas. After entering the database and searching the gene name in the expression module, the average mRNA expression level of the gene in different tissues can be viewed.

**Construction of miRNA-Hub Gene Network**

The related miRNAs of the six hub genes were predicted through visualization by the CyTargetLinker plugin of Cytoscape software (version 3.8.2). Next, the centrality of networks (degree, betweenness, and closeness) was analyzed, and the hub miRNAs were identified using the CytoNCA plugin.

**Animals**

Twenty-four C57BL/6J male mice, aged 4-weeks, were purchased from Changsha Tianqin Co., LTD (Changsha, Hunan, China) and were randomly divided into the following three groups: the control group, HTG group, and AP group. All mice were kept in a temperature-controlled room, fed freely, and maintained under a humidity of 50–60% with 12-hour light/12-hour dark cycles. Mice in the HTG group were fed a high-fat diet (12451, Boaigang, China), while mice in the non-HTG group were fed a regular diet. When mice were 8-weeks old, AP was induced through seven intraperitoneal injections of CAE (EYS113, Amquar, USA) administered at a dose of 50 µg/kg at an interval of 1 h. Simultaneously, the mice in the control group were injected with the same amount of normal saline. After 24 h of the last injection, blood was collected from the inferior vena cava along with the pancreatic tissues. The obtained blood was used for subsequent detection of triglyceride and amylase levels, while the pancreatic tissue was used for HE and immunohistochemical staining.

**Detection of Triglyceride and Serum Amylase in Mice**

The venous blood of mice was kept at 26 °C for 2 h and then centrifuged (4 °C, 3000 RPM, 15 min). The supernatant was collected for subsequent detection. We used a triglyceride detection kit (c061-a, Changchun Huili Biotech CO., LTD., Changchun, China) and an amylase detection kit (C033, Changchun Huili Biotech CO., LTD., Changchun, China) to detect the levels of triglyceride and serum amylase, respectively. The preparation of the detection reagent was carried out according to the instructions. After adding the detection reagent, the absorbance of the supernatant was read under the microplate reader. The final test results were statistically analyzed by an independent sample t-test using SPSS 23.0.

**HE and Immunohistochemistry**

Pancreatic tissues were immediately soaked in 4% paraformaldehyde solution for 12 h, and then embedded in paraffin, and sectioned (5 µm size). The tissue sections were stained with HE and observed under a microscope for pancreatic pathological scores. The pancreatic histopathological scoring criteria were based on the study of Van Laethem JL.24 The final pancreatic histopathological scores were statistically analyzed by an independent sample t-test. Dewaxing, hydration, and thermal repair were performed, and then the primary antibodies (ACTB, D110001, Sangong Biotech, 1:100; TPM1, D121373, Sangong Biotech, 1:100; DSTN, D222231, Sangong Biotech, 1:200; TPM4, D123959, Sangong Biotech, 1:50; WDR1, D225121, Sangong Biotech, 1:150, WSB1, D123579, Sangong Biotech, 1:50; FBXO4, bs-9087R, Beijing Bioss Biotech, 1:250) were added and incubated overnight at 4 °C. After incubating the tissues with HRP-labeled linked polymer (KIT-5009, MXB biotechnologies) for 40 min at 26 °C, the signal was detected using DAB (P0202, Beyotime). After statistical analysis, the expression of hub genes was measured using the Image-Pro Plus 6.0 and visualized using GraphPad Prism 8.0.

**Results**

**Identification of DEGs**

As shown in Figure 1, 105 DEGs, including 11 down-regulated genes (Figure 1A) and 94 upregulated genes (Figure 1B), were identified from the three microarray profile datasets (GSE121038, GSE3644, and GSE65146) (Table 1). In the three AP datasets and one hypertriglyceridemia dataset, we observed two common upregulated DEGs: WD repeat and SOCS
Additionally, the volcano plots and heatmaps were generated to demonstrate the downregulated and upregulated genes in the GEO datasets (Figures 2 and 3).

**Functional Enrichment Analysis**

In the Biological Process (BP) category, the upregulated DEGs were involved in the movement of cells or subcellular components, locomotion, localization of cells, cell motility, and cell adhesion (Figure 4A), whereas the downregulated DEGs were significantly involved in the cellular responses to an organic substance, topologically incorrect protein and chemical stimulus along with endoplasmic reticulum unfolded protein response (Figure 4E). In the Cellular Component (CC) category, the upregulated DEGs were correlated with the focal adhesion, cell-substrate adherens junction, cell-substrate junction, adherens junction, and anchoring junction (Figure 4B). In the Molecular Function (MF) category, the upregulated DEGs were found to be enriched in the protein complex binding, cell adhesion molecule binding,
macromolecular complex binding, and cadherin-and protein-binding involved in cell-cell adhesion (Figure 4C). In the KEGG pathway enrichment analysis, the upregulated DEGs were found to be enriched in the MAPK signaling pathway, regulation of the actin cytoskeleton, PI3K-Akt signaling pathway, oxytocin signaling pathway, and B-cell receptor signaling pathway (Figure 4D). Detailed results of the enrichment analysis for upregulated and downregulated DEGs are shown in Tables 2 and 3.

Construction and Module Analysis of PPI Network of DEGs
A PPI network consisting of 105 points and 150 edges was constructed using the STRING online tool (PPI enrichment P-value <1.0e-16 (Figure 5). The results of further module analysis are shown in Figures 6A–D. The values of the MCODE score, the number of nodes and edges in each module were as follows: module 1 (Figure 6A), MCODE score = 6.286, 8 nodes and 22 edges; module 2 (Figure 6B), MCODE score = 3.714, 8 nodes and 13 edges; module 3

Figure 2 The Volcano plot distribution of DEGs. The volcano plot of (A) GSE1010, (B) GSE65146, (C) GSE121038, and (D) GSE3644. The green dots indicate the screened downregulated DEGs, Orange dots indicate the screened upregulated DEGs while the gray dots indicate genes with no significant differences. Abbreviation: DEGs, differentially expressed genes.
A total of 6 AP-related hub genes were identified as follows: actin beta (*ACTB*), actin gamma 1 (*ACTG1*), tropomyosin 1 (*TPM1*), tropomyosin 4 (*TPM4*), destrin (*DSTN*), and WD repeat domain 1 (*WDR1*). As shown in Table 4, all the six identified hub genes were upregulated DEGs.

(Figure 6C), MCODE score = 3, 3 nodes and 3 edges; and module 4 (Figure 6D), MCODE score = 3, 3 nodes and 3 edges). A total of 6 AP-related hub genes were identified as follows: actin beta (*ACTB*), actin gamma 1 (*ACTG1*), tropomyosin 1 (*TPM1*), tropomyosin 4 (*TPM4*), destrin (*DSTN*), and WD repeat domain 1 (*WDR1*). As shown in Table 4, all the six identified hub genes were upregulated DEGs.

Figure 3 The heatmap of the top 50 DEGs from GSE1010 (A) and GSE3644 (B). Color change from red to blue indicates a gradual decrease in the expression level of the gene in the sample.
Expression of Hub Genes in Different Human Organs

As shown in Figure 7, the six hub genes were detected in human pancreatic tissue at different RNA expression levels. ACTG1 (Figure 7B, median TPM: 965.3) showed the highest expression level in the pancreas among all hub genes, followed by ACTB (Figure 7A, median TPM: 404.8), DSTN (Figure 7C, median TPM: 100.7), TPM1 (Figure 7D, median TPM: 39.52), and TPM4 (Figure 7E, median TPM: 28.53) while WDR1 (Figure 7F, median TPM: 17.52) showed the lowest levels of RNA expression. Additionally, the expression levels of all these hub genes in the pancreas were lower than those found in most other organs.

Construction of miRNA-Hub Gene Network

We constructed a network consisting of six hub genes and 310 miRNAs using the CyTargetLinker plugin of Cytoscape software (Figure 8). Some miRNAs that regulate multiple genes are speculated to play an important role in the regulation of hub genes. The miRNA hsa-mir-1–3p was predicted to regulate ACTB, WDR1, TPM1, and TPM4; hsa-mir-5195–3p, hsa-mir-145–5p and hsa-let-7b-5p were predicted to regulate ACTB, ACTG1, and TPM4; hsa-mir-10b-5p was predicted to regulate ACTG1, TPM1, and TPM4; hsa-mir-206 and hsa-mir-613 were predicted to regulate ACTB, TMP4, and WDR1. Additionally, details of miRNAs regulating multiple hub genes are shown in Table 5.

Detection of Triglyceride and Serum Amylase Levels in Mice

The serum triglyceride level in the TG group was significantly higher than that of the non-TG group (Figure 9A, HTG vs control, $t = 5.019$, $P < 0.001$; HTG vs AP, $t = 5.338$, $P < 0.001$). However, no significant difference was observed in
| Category          | Term                                      | Count | P-value       | Genes                                                                 |
|-------------------|-------------------------------------------|-------|---------------|----------------------------------------------------------------------|
| BP                | Movement of a cell or subcellular component | 25    | 2.2863E-06    | WDR1, IQGAP1, THBS1, RTN4, GNAI3, EPCAM, GCNT2, TSPAN1, TRPS3INP1, JUN, HSP90AA1, TNFRSF12A, TPM1, ARID5B, ADAM10, RAR2, DSN, EMP2, GTPBP4, RHOB, KLF7, DDIT4, CD9, PAFAH1B2, VCL |
| BP                | Locomotion                                 | 23    | 3.9846E-06    | JUN, HSP90AA1, TNFRSF12A, WDR1, TPM1, ARID5B, RAR2, ADAM10, EMP2, IQGAP1, THBS1, RTN4, GTPBP4, RHOB, GNAI3, RCAN1, KLF7, EPCAM, DDIT4, GCNT2, TSPAN1, VCL, TRPS3INP1 |
| BP                | Localization of cell                       | 21    | 7.7146E-06    | JUN, HSP90AA1, TNFRSF12A, WDR1, TPM1, ARID5B, RAR2, ADAM10, EMP2, IQGAP1, THBS1, RTN4, GTPBP4, RHOB, GNAI3, EPCAM, DDIT4, GCNT2, TSPAN1, VCL, TRPS3INP1 |
| BP                | Cell motility                              | 21    | 7.7146E-06    | JUN, HSP90AA1, TNFRSF12A, WDR1, TPM1, ARID5B, RAR2, ADAM10, EMP2, IQGAP1, THBS1, RTN4, GTPBP4, RHOB, GNAI3, EPCAM, DDIT4, GCNT2, TSPAN1, VCL, TRPS3INP1 |
| BP                | Cell adhesion                              | 22    | 2.4056E-05    | HSP90AA1, TES, TNFRSF12A, TPM1, NEDD9, ADAM10, EMP2, IQGAP1, THBS1, YWHAZ, ACTB, GTPBP4, ACTGI, RHOB, CLDN4, CSRPI, EPCAM, GCNT2, CD9, TAGLN2, CLIC1, VCL |
| BP                | Biological adhesion                        | 22    | 2.6930E-05    | HSP90AA1, TES, TNFRSF12A, TPM1, NEDD9, ADAM10, EMP2, IQGAP1, THBS1, YWHAZ, ACTB, GTPBP4, ACTGI, RHOB, CLDN4, CSRPI, EPCAM, GCNT2, CD9, TAGLN2, CLIC1, VCL |
| BP                | Regulation of cell motility                | 15    | 2.9636E-05    | JUN, TPM1, RAR2, ADAM10, EMP2, IQGAP1, THBS1, RTN4, GTPBP4, RHOB, GNAI3, EPCAM, GCNT2, VCL, TRPS3INP1 |
| BP                | Regulation of locomotion                   | 15    | 4.9058E-05    | JUN, TPM1, RAR2, ADAM10, EMP2, IQGAP1, THBS1, RTN4, GTPBP4, RHOB, GNAI3, EPCAM, GCNT2, VCL, TRPS3INP1 |
| BP                | Regulation of cell cycle                   | 15    | 5.9672E-05    | CAST, HSPA8, JUN, GADD45B, THBS1, GTPBP4, ZWINT, RHOB, NRS, DDIT3, FBXO4, UBA3, EIF4E, CLIC1, TRPS3INP1 |
| BP                | Regulation of cellular component movement  | 15    | 6.9562E-05    | JUN, TPM1, RAR2, ADAM10, EMP2, IQGAP1, THBS1, RTN4, GTPBP4, RHOB, GNAI3, EPCAM, GCNT2, VCL, TRPS3INP1 |
| MF                | Protein complex binding                    | 20    | 4.6349E-08    | HSPA8, HSP90AA1, WDR1, TPM1, KRT8, DSN, EMP2, PPP2R2A, IQGAP1, THBS1, YWHAZ, ACTB, RTN4, RELA, GNAI3, NRS, EPCAM, CD9, KRAS, VCL |
| MF                | Cell adhesion molecule binding             | 13    | 1.6010E-06    | CAST, HSPA8, TES, EMP2, IQGAP1, THBS1, YWHAZ, RTN4, EPCAM, CD9, TAGLN2, CLIC1, VCL |
| MF                | Macromolecular complex binding             | 23    | 1.7411E-06    | TGFI1, HSPA8, JUN, HSP90AA1, WDR1, TPM1, KRT8, DSN, EMP2, PPP2R2A, NMD3, IQGAP1, THBS1, YWHAZ, ACTB, RTN4, RELA, GNAI3, NRS, EPCAM, CD9, KRAS, VCL |
| MF                | Cadherin binding involved in cell-cell adhesion | 10    | 8.3168E-06    | CAST, HSPA8, TES, EPCAM, TAGLN2, IQGAP1, YWHAZ, CLIC1, RTN4, VCL |
| MF                | Protein binding involved in cell-cell adhesion | 10    | 1.0433E-05    | CAST, HSPA8, TES, EPCAM, TAGLN2, IQGAP1, YWHAZ, CLIC1, RTN4, VCL |
| MF                | Protein binding involved in cell adhesion   | 10    | 1.1978E-05    | CAST, HSPA8, TES, EPCAM, TAGLN2, IQGAP1, YWHAZ, CLIC1, RTN4, VCL |
| MF                | Cadherin binding                           | 10    | 1.4855E-05    | CAST, HSPA8, TES, EPCAM, TAGLN2, IQGAP1, YWHAZ, CLIC1, RTN4, VCL |
| MF                | GTP binding                                | 9     | 0.00054152    | GNAI3, NRS, TUB86, HSP90AA1, TUBB2A, RAR2, KRAS, GTPBP4, RHOB |
| MF                | Guanyl ribonucleotide binding               | 9     | 0.00076737    | GNAI3, NRS, TUB86, HSP90AA1, TUBB2A, RAR2, KRAS, GTPBP4, RHOB |
| MF                | Guanyl nucleotide binding                   | 9     | 0.00078006    | GNAI3, NRS, TUB86, HSP90AA1, TUBB2A, RAR2, KRAS, GTPBP4, RHOB |
| CC                | Focal adhesion                             | 16    | 3.8697E-09    | HSPA8, TES, TPM4, FHL2, RAR2, ADAM10, IQGAP1, YWHAZ, ACTB, ACTGI, RHOB, GNAI3, CSRPI, CD9, KRAS, VCL |
| CC                | Cell-substrate adherens junction           | 16    | 4.5967E-09    | HSPA8, TES, TPM4, FHL2, RAR2, ADAM10, IQGAP1, YWHAZ, ACTB, ACTGI, RHOB, GNAI3, CSRPI, CD9, KRAS, VCL |

(Continued)
Table 2 (Continued).

| Category          | Term                           | Count | P-value          | Genes                                                                                           |
|-------------------|--------------------------------|-------|------------------|-------------------------------------------------------------------------------------------------|
| CC                | Cell-substrate junction        | 16    | 5.44675E-09     | HSPA8, TES, TPM4, FHL2, RRAS2, ADAM10, IQGAP1, YWHAZ, ACTB, ACTG1, RHOB, GNAI3, CSRPI, CD9, KRAS, VCL |
| CC                | Adherens junction              | 20    | 6.0427E-09      | CAST, HSPA8, TES, TPM4, FHL2, RRAS2, ADAM10, IQGAP1, YWHAZ, ACTB, RTN4, ACTG1, RHOB, GNAI3, CSRPI, CD9, TAGLN2, KRAS, CLIC1, VCL |
| CC                | Anchoring junction             | 20    | 9.18845E-09     | CAST, HSPA8, TES, TPM4, FHL2, RRAS2, ADAM10, IQGAP1, YWHAZ, ACTB, RTN4, ACTG1, RHOB, GNAI3, CSRPI, CD9, TAGLN2, KRAS, CLIC1, VCL |
| CC                | Extracellular exosome          | 37    | 7.27423E-08     | WDR1, GHITM, IQGAP1, THBS1, MYL12A, ACTB, RTN4, ACTG1, GNAI3, NRAS, TUBB6, CSRPI, EPCAM, MAL2, CLIC1, TSPAN1, EIF4E, HSPA8, HSP90AA1, TPM4, KRT8, ADAM10, RRAS2, DSN, ACSL4, YWHAZ, RHOB, TWSG1, TUBB2A, SERPINB6A, LCN2, CD9, TAGLN2, PLP2, SFT2D2, PAFAH1B2, VCL |
| CC                | Extracellular vesicle          | 37    | 8.43104E-08     | WDR1, GHITM, IQGAP1, THBS1, MYL12A, ACTB, RTN4, ACTG1, GNAI3, NRAS, TUBB6, CSRPI, EPCAM, MAL2, CLIC1, TSPAN1, EIF4E, HSPA8, HSP90AA1, TPM4, KRT8, ADAM10, RRAS2, DSN, ACSL4, YWHAZ, RHOB, TWSG1, TUBB2A, SERPINB6A, LCN2, CD9, TAGLN2, PLP2, SFT2D2, PAFAH1B2, VCL |
| CC                | Extracellular organelle        | 37    | 9.02845E-08     | WDR1, GHITM, IQGAP1, THBS1, MYL12A, ACTB, RTN4, ACTG1, GNAI3, NRAS, TUBB6, CSRPI, EPCAM, MAL2, CLIC1, TSPAN1, EIF4E, HSPA8, HSP90AA1, TPM4, KRT8, ADAM10, RRAS2, DSN, ACSL4, YWHAZ, RHOB, TWSG1, TUBB2A, SERPINB6A, LCN2, CD9, TAGLN2, PLP2, SFT2D2, PAFAH1B2, VCL |
| CC                | Cell junction                  | 24    | 1.74796E-06     | CAST, HSPA8, TES, WDR1, TPM4, KRT8, FHL2, RRAS2, ADAM10, IQGAP1, YWHAZ, ACTB, RTN4, ACTG1, RHOB, GNAI3, CD9, TAGLN2, KRAS, CLIC1, VCL |
| CC                | Membrane-bounded vesicle       | 39    | 5.31841E-06     | WDR1, GHITM, IQGAP1, THBS1, MYL12A, ACTB, RTN4, ACTG1, GNAI3, NRAS, TUBB6, CSRPI, EPCAM, MAL2, CLIC1, TSPAN1, EIF4E, HSPA8, HSP90AA1, TPM4, KRT8, ADAM10, RRAS2, DSN, ACSL4, YWHAZ, RHOB, TWSG1, TUBB2A, SERPINB6A, LCN2, CD9, TAGLN2, PLP2, SFT2D2, PAFAH1B2, VCL |
| PATHWAY           | MAPK signaling pathway         | 10    | 3.33409E-05     | HSPA8, RPS6KA3, NRAS, JUN, GADD45B, DDIT3, RRAS2, KRAS, DUSP16, RELA |
| PATHWAY           | Regulation of actin cytoskeleton | 9    | 6.92899E-05     | GNAI3, NRAS, RRAS2, KRAS, IQGAP1, MYL12A, ACTB, VCL, ACTG1 |
| PATHWAY           | PI3K-Akt signaling pathway     | 10    | 0.000430263     | NRAS, HSP90AA1, DDIT4, PPP2R2A, KRAS, OSMR, THBS1, EIF4E, YWHAZ, RELA |
| PATHWAY           | Oxytocin signaling pathway     | 7     | 0.000434823     | RCAN1, NRAS, RGS2, JUN, KRAS, ACTB, ACTG1 |
| PATHWAY           | B cell receptor signaling pathway | 5    | 0.001139278     | NRAS, JUN, DAPP1, KRAS, RELA |
| PATHWAY           | Proteoglycans in cancer signaling pathway | 7    | 0.00210178    | NRAS, RRAS2, KRAS, IQGAP1, THBS1, ACTB, ACTG1 |
| PATHWAY           | Thyroid hormone signaling pathway | 5    | 0.003917266    | HDAC4, PPP2R2A, MYL12A, ACTB, ACTG1 |
| PATHWAY           | Leukocyte transendothelial migration | 5    | 0.007533064    | HDAC4, MYL12A, ACTB, VCL, ACTG1 |

**Abbreviations:** BP, biological process; CC, cellular component; MF, molecular function.
Amylase levels between AP and non-AP groups (Figure 9B, AP vs control, \( t = 0.819, P = 0.426 \); AP vs HTG, \( t = 1.774, P = 0.978 \)).

**HE Staining and Pancreatic Histopathological Score**

In the AP group, HE staining showed a large number of neutrophil infiltrates in the pancreatic tissue, obvious edema around the pancreatic lobules with significantly widened space between the lobules (Figure 10C). Pancreatic histopathological score in the AP group was significantly higher than that of the control group (Figure 10D, \( t = 29.95, P < 0.001 \)) and HTG group (Figure 10D, \( t = 26.24, P < 0.001 \)). However, no significant difference was observed between the control (Figure 10A and D) and HTG groups (Figure 10B and D, \( t = 1.143, P = 0.255 \)).

**Immunohistochemistry Staining**

We verified the expression levels of five hub genes and two common DEGs by immunohistochemical staining (Figures 11 and 12). Microscopic images showed that the five hub genes in the AP group were significantly more stained than those in the control group (Figure 11A–J). Similarly, the staining degrees of two common DEGs in the AP and HTG groups were significantly higher than those in the control group (Figure 12A–F). Further, the statistical analysis showed significant differences among the groups (Figure 11K, \( t = 22.02, P < 0.001 \); DSTN, AP vs control, \( t = 32.37, P < 0.001 \); TPM1, AP vs control, \( t = 17.71, P < 0.001 \); TPM4, AP vs control, \( t = 9.28, P < 0.0001 \); WDR1, AP vs control, \( t = 18.42, P < 0.0001 \); Figure 12G, FBXO4, HTG vs control, \( t = 22.54, P < 0.0001 \); WSB1, HTG vs control, \( t = 16.82, P < 0.0001 \); FBXO4, AP vs control, \( t = 30.38, P < 0.0001 \); WSB1, AP vs control, \( t = 23.87, P < 0.0001 \)).

**Discussion**

The worldwide incidence of AP is on the rise. Additionally, increasing evidence has been showing that in general, HTGP is more severe than AP of any other etiology, but the reasons for this are not clear.\(^5\,25\) With the advancement of high-throughput sequencing, more and more diseases are being well explained at the genetic level. In recent years, bioinformatics research on the pathogenesis of tumors has made great progress. However, only a few bioinformatics studies are available on AP and hypertriglyceridemia.

HE staining showed that in the AP group, a large number of neutrophil infiltrates were found in the pancreatic tissue along with obvious edema around the pancreatic lobules with significantly widened space between the lobules. The histopathological score of the AP group was significantly higher than that of the non-AP group, with a statistically
significant difference. The serum triglyceride level in the HTG group was significantly higher than that of the non-TG group, with a statistically significant difference. In a study of AP conducted in mice by Xinnong Liu, serum amylase levels began to rise 6 h after the onset of the disease, which peaked at 12 h and dropped to normal levels at 24 h. Our study showed that in the CAE-induced AP mouse model, no significant difference was observed in the serum amylase level after 24 h of the onset of disease in normal mice, which was also consistent with Xinnong Liu’s study. These results confirmed the validity of the CAE-induced AP model as well as the high-fat diet-induced hypertriglyceridemia model in mice.

In this study, 105 DEGs were identified from the three microarray profile datasets of AP, including 11 downregulated and 94 upregulated genes in the AP tissues compared to the normal controls. Subsequently, functional enrichment analysis of the DEGs was performed using the David online tool. In the KEGG pathway enrichment analysis, the upregulated DEGs were found to be enriched in the MAPK signaling pathway, regulation of actin cytoskeleton, and PI3K-Akt signaling pathway. The MAPK-activated protein kinase (MK2) is a key enzyme, which is closely related to p38 and is associated with inflammation. MK2 knockout mice show protection from a wide range of inflammatory/autoimmune diseases, including arthritis, pulmonary inflammation, and myocardial ischemia. P38 MAPK is a signaling molecule downstream of a proinflammatory cytokine receptor, which mediates proinflammatory cytokine synthesis by regulating both transcription and post-transcriptional modification. The inhibition of the MAPK signaling pathway can alleviate lung injury caused by severe pancreatitis. However, whether the inhibition of MAPK signaling alleviates AP is unknown and needs further investigation.

Figure 5 PPI network of 105 DEGs. The criteria of the PPI network are as follows: Confidence score ≥ 0.4 and a maximum number of interactions ≤ 5.

Abbreviations: PPI, protein–protein interaction; DEGs, differentially expressed genes.
The PPI network of the DEGs was constructed in AP based on the information obtained from the STRING database. The PPI network included 105 nodes and 150 edges, and its PPI enrichment P-value was found to be lower than 1.0e-16. The six hub genes showing the highest score (ACTB, ACTG1, TPM1, TPM4, DSTN, and WDR1) were up-regulated in the AP group compared to the normal pancreatic tissue, which was confirmed by immunohistochemistry. ACTB and ACTG1 encode β-actin and γ-actin, respectively. Both TPM1 and TPM4 are members of the tropomyosin family and are involved in the cytoskeletal composition of non-muscle cells. DSTN, also known as an actin-depolymerizing factor, is a member of the ADF family of actin-binding proteins. The main function of the DSTN is to cut the actin filaments (F-actin) and bind to actin monomers (G-actin), while the function of the gene WDR1 is mainly to induce depolymerization of the actin filaments. Changes in the actin cytoskeleton, including myosin activation, actin redistribution, and filamentous actin loss, can be observed in the early stages of CAE-induced

![Figure 6](image1.png)

**Figure 6** The top four modules of the PPI network. (A) module 1, MCODE score = 6.286, 8 nodes and 22 edges; (B) module 2, MCODE score = 3.714, 8 nodes and 13 edges; (C) module 3, MCODE score = 3, 3 nodes and 3 edges; (D) module 4, MCODE score = 3, 3 nodes and 3 edges.

**Abbreviation:** PPI, protein–protein interaction.

| Rank | Name | Score |
|------|------|-------|
| 1    | ACTB | 604   |
| 2    | ACTG1| 516   |
| 3    | TPM1 | 488   |
| 4    | TPM4 | 486   |
| 5    | DSTN | 360   |
| 6    | WDR1 | 240   |

**Table 4** The Top Six Hub Genes in the PPI Network as Ranked by the Degree Method

**Abbreviation:** PPI, protein–protein interaction.
Our study verified that in the AP model induced by CAE, the functions of all up-regulated hub genes were closely related to the actin cytoskeleton, which indicated that the destruction of the actin cytoskeleton might play an important role in the pathogenesis of AP. More severe AP can be caused by the loss of polarity in the pancreatic acinar cells and disruption of the actin network; however, the exact mechanism needs to be clarified. Gene expression levels were evaluated by the online database HPA, and it was found that the expression levels of all hub genes were lower in pancreatic tissues compared to other organs. During the onset of AP, the expression levels of all the hub genes were significantly increased. This further suggested the importance of the role of the identified hub genes in the onset of AP.

In the three AP datasets and one hypertriglyceridemia dataset, two common up-regulated DEGs (WSB1 and FBXO4) were identified. These genes, WSB1 and FBXO4, are substrate recognition proteins within an E3 ubiquitin ligase, which can bind to diverse targets and mediate ubiquitylation and proteolytic degradation. Most of the current studies on WSB1 and FBXO4 have focused on tumors but have not been reported in hypertriglyceridemia and AP. We used bioinformatics methods to screen common DEGs in hypertriglyceridemia and AP and found that WSB1 and FBXO4 were highly expressed in both diseases. This may provide a new idea for us to elucidate the pathogenesis of HTGP at the genetic level in the future.

The miRNAs are small non-coding RNAs that regulate gene expression by interfering with transcription or translation. To further predict miRNAs that might regulate hub genes, a miRNA-mRNA network was constructed.

Figure 7 The comparison of the basic expression of the hub genes between the pancreas and other different human organs. ACTG1 (B, median TPM: 965.3) showed the highest expression level in the pancreas among all hub genes, followed by ACTB (A, median TPM: 404.8), DSTN (C, median TPM: 100.7), TPM1 (D, median TPM: 39.52), and TPM4 (E, median TPM: 29.53) while WDR1 (F, median TPM: 17.52) showed the lowest levels of RNA expression.
Figure 8 Interaction network between the hub genes and targeted miRNAs. Hub genes are presented in gray circles, whereas miRNAs are shown in red circles. The darker the color, the greater the number of hub genes involved in the regulation. The larger the gray circle, the more number of miRNAs regulating that gene. The interaction between the hub genes and related miRNAs is shown in the form of arrows.
Some miRNAs (hsa-mir-1–3p, hsa-mir-5195–3p, hsa-mir-145–5p, hsa-let-7b-5p, hsa-mir-10b-5p, hsa-mir-206, and hsa-mir-613) that have been found to regulate multiple genes are speculated to play an important role in the regulation of hub genes.

**Conclusions and Limitations**
First, we identified 105 DEGs and six hub genes (ACTB, ACTG1, TPM1, TPM4, DSTN, and WDR1) associated with AP. Also, we are the first ones to report the common DEGs (WSB1 and FBXO4) in hypertriglyceridemia and AP. Second, we have validated the expression of six hub genes and two common DEGs by immunohistochemistry. Interestingly, all hub genes were expressed at low levels in normal pancreatic tissues, but were significantly upregulated in AP pathogenesis, and all hub genes functioned in the actin cytoskeleton; hence, we believe that the changes in the actin cytoskeleton caused by these hub genes may have an important role in the pathogenesis of AP. Third, we identified several miRNAs that may regulate hub genes. However, our study has some shortcomings. Firstly, the main data of our study is derived from online databases, and our study also has a small sample size. Secondly, we only identified and validated the hub genes that may be associated with the pathogenesis of AP, along with exploring the common DEGs in AP and hypertriglyceridemia. However, their specific roles in the pathogenesis of acute AP and HTGP need further study. Thirdly, we did not construct HTGP mouse models to verify the differential expression of these genes, which is also one

Table 5 Details of the miRNAs Regulating Multiple Hub Genes

| miRNA       | Targeted Hub Genes            |
|-------------|-------------------------------|
| Hsa-mir-1–3p| ACTB, WDR1, TPM1 and TPM4    |
| Hsa-mir-5195–3p| ACTB, ACTG1, and TPM4       |
| Hsa-mir-145–5p| ACTB, ACTG1, and TPM4       |
| Hsa-let-7b-5p| ACTB, ACTG1, and TPM4       |
| Hsa-mir-10b-5p| ACTG1, TPM1, and TPM4       |
| Hsa-mir-206  | ACTB, TMP4, and WDR1         |
| Hsa-mir-613  | ACTB, TMP4, and WDR1         |
| Hsa-mir-320a | ACTB, ACTG1, and WDR1       |
| Hsa-mir-19a-3p| ACTB, WDR1                 |
| Hsa-mir-19b-3p| ACTB, WDR1                 |
| Hsa-mir-10a-5p| ACTG1, TPM4               |

Figure 9 Differences in serum amylase and triglyceride levels among the groups. (A) No significant difference observed in serum amylase levels between AP and non-AP groups (AP vs control, t = 0.819, P = 0.426; AP vs HTG, t = 1.774, P = 0.978). (B) The serum triglyceride level in the HTG group was significantly higher than that of the non-HTG group (HTG vs control, t = 5.019, P <0.001; HTG vs AP, t = 5.338, P <0.001). ns, no significant difference; ***, P < 0.001.
Figure 10 The HE staining of pancreatic tissues among different groups. Numerous neutrophil infiltrates, perilobular edema, and interlobular space was observed in the acute pancreatitis group (C) and not in the non-pancreatitis group (A, control; B, HTG). (D) The statistical results of pancreatic histopathological score. ns, no significant difference; ***, *P* < 0.001.

Figure 11 Immunohistochemistry and statistical results of the pancreas among different groups. (A–J) The five hub genes in the AP group were significantly more stained than those of the control group. (K) Statistical analysis results of the immunohistochemical staining, which is as follows: ACTB, AP vs control, *t* = 22.02, *P* < 0.001; DSTN, AP vs control, *t* = 32.37, *P* < 0.001; TPM1, AP vs control, *t* = 17.71, *P* < 0.001; TPM4, AP vs control, *t* = 9.28, *P* < 0.001; WDR1, AP vs control, *t* = 18.42, *P* < 0.0001. ****, *P* < 0.0001.
of the limitations of this study. Finally, due to the limited conditions, we could not verify the differential expression of all DEGs, which will be one of our future research works.

**Data Sharing Statement**
All data in the study is available from the submitting author, Shiyu Zhang.

**Ethics Approval**
The study has been approved by the Ethics Committee of the First Affiliated Hospital of Guangxi Medical University (No. 2021 KY-E-187), and followed the principles of ethical animal research outlined in the Basel Declaration and the ethical guidelines of the International Council for Laboratory Animal Science (ICLAS).

**Consent for Publication**
All authors agree to the publication of all texts, images and tables in this study.

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**Disclosure**
All authors report no conflicts of interest for this work and state that there is no potential conflict of interest in this research.
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