Identification of Liver Immune Microenvironment-Related Hub Genes in Liver of Biliary Atresia

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Background: Biliary atresia (BA) is one of the most common and fatal abnormalities of newborns. Increasing evidences indicated that immunology was the critical part of the etiology. This research used a public gene expression database to explore the immune microenvironment of BA liver.

Methods: The gene expression profiles GSE46960, GSE159720, and GSE15235, containing BA and normal liver gene expression data, were obtained from the Expression Omnibus Gene. We applied CIBERSORTx to quantify 22 subsets of immune cells in BA liver. The differentially expressed genes (DEGs) and immune cells were used to further explore their relationship with liver fibrosis and the inflammation status of BA.

Results: The expression of immune-related genes CXCL6, CXCL8, CXCL10, CCL20, IL32, TGFB2, SPP1, and SLIT2 was significantly different between BA and normal liver, among which CXCL8 was the hub gene. Six of 22 immune cell proportions were significantly different between BA and normal liver. Specifically, M0 macrophages and resting memory CD4⁺ T cells were upregulated in BA liver compared with normal liver. Meanwhile, monocytes, resting natural killer (NK) cells, plasma cells, and regulatory T (Treg) cells were downregulated. A further correlation analysis revealed that SLIT2 and CXCL6 owned high positive correlation coefficients with fibrosis grade, while the proportion of resting NK cells was negatively correlated. Proportions of resting CD4⁺ memory T cells were strongly related to the inflammation grade of BA liver.

Conclusion: Biliary atresia is a disease strongly correlated with immune response. Our results might provide a clue for further exploration of BA etiology, which may promote a potential prediction model based on immune infiltration features.

Keywords: biliary atresia, liver autoimmune disease, immune microenvironment, prognosis, gene expression profile

INTRODUCTION

Biliary atresia (BA) is one of the most common and fatal abnormalities of newborns. Most BA patients cannot survive with their native liver because of severe cholestatic cirrhosis, unless they receive Kasai portoenterostomy to reconstruct the bile drainage (1). However, more than half of the patients cannot survive with their native liver for a long term after receiving Kasai portoenterostomy, which contributes to 75% of pediatric liver transplantation under 2 years old in the world (2).
Although the etiology of BA is still not clear, some theories were related to genetics (3), immunology (4), and toxicology (5). Increasing evidence indicated that immunologic derangement was the core issue that caused BA. The animal model of BA was based on the immune reaction of rhesus rotavirus (RRV) infection in newborn Balb/c mice (6). The abnormality of B cells, T cells, and macrophages in BA patients was also revealed by single-cell RNA-seq in a small population recently, which provided strong evidence that the immune system took part in the occurrence and development of BA to a great extent (7). Therefore, research with a larger sample size were necessary to further prove these findings.

In the past, microarrays and RNA-seq data could only reflect the mRNA expressions of bulk liver tissues composed of hepatocytes, cholangiocytes, endothelial cells, and immune cells, making it impossible to demonstrate the potential functions of each kind of cell. With the development of bioinformatics, powerful deconvolution algorithms now enable us to calculate the immune microenvironment precisely by using the bulk mRNA expression data. Here, we used CIBERSORTx, a tool verified by fluorescence-activated cell sorting, to reveal the relationship between the immune microenvironment of BA liver and the progression of the disease by utilizing the present mRNA expressions data sets of BA (8). Since Rohr-Udilova et al. (9) first used it in exploring the immune cell landscape between healthy liver and hepatocellular carcinoma, CIBERSORT was widely applied in various liver diseases (10–12).

**METHODS**

**Data Source**

Gene expression data GSE46960 (13), GSE159720 (3), and GSE15235 (14) were obtained from the “Gene Expression Omnibus” (GEO) database. The former two data series contained genetic expression profiles of BA liver and normal liver. GSE46960 was constructed by Affymetrix Human Gene 1.0 ST Array. It received liver biopsy samples from 64 infants with BA at the time of intraoperative cholangiogram and seven deceased donor children. GSE159720 was established by Illumina NextSeq 500. Only four BA samples together with three normal samples were found in this data series. GSE15235 was a data series without genetic information of normal liver. However, it had a specific grade of liver inflammation and fibrosis corresponding to the genetic expression, making it possible to analyze the correlation between liver inflammation and fibrosis with the hub genes and liver immune microenvironment.

**Data Processing and Analysis of Differentially Expressed Genes**

After the original data were obtained, the raw data sets were processed according to the platforms of the chip. GSE46960 was processed by the “affy” package first, and the “Limma” package was applied in R to screen the differentially expressed genes (DEGs) of BA liver and normal liver with P-value < 0.05 and \( \log_{2} (\text{fold change}) \geq 1.2 \). GSE159720 was processed by the “oligo” package first, followed by the “DESeq” package in R, and the DEGs were identified under the same criteria as GSE46960. A Venn diagram of DEGs from the former data sets and ImmPort, a data set of genes related to the immune system (15), was made to explore the overlapping genes associated with the immune system.

**Functional Enrichment Analyses of DEGs in BA**

The DEGs of each dataset were processed by gprofiler (https://biit.cs.ut.ee/gprofiler/gost), an online tool for functional enrichment analyses. Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway and Gene Ontology (GO) pathway were selected for enrichment (16). Metascape (https://metascape.org) was an online tool to analyze the DEGs from multiple data sets (17). The enrichment analysis of DEGs from GSE46960 and GSE159720 was processed by Metascape, which exported the most significant GO and KEGG pathways in both data sets.

**Generating the Protein–Protein Interaction Network and Identification of Hub Genes**

The Search Tool for the Retrieval of Interacting Genes (STRING, https://string-db.org) is a biological resource that provides systematic screens of human–protein interactions (18). The overlapping DEGs of GSE46960 and GSE159720 were processed by STRING to get a protein–protein interaction (PPI) network, after which Cytoscape was used to generate a visualized PPI network. Hub genes were identified by CytoHuba, a tool of Cytoscape, with default parameters, through which it turned out that immune-related genes were critical during the disease process.

**Liver Immune Microenvironment Analysis**

CIBERSORTx (https://cibersortx.stanford.edu/) is an analytical tool to provide an estimation of the abundances of member cell types in bulk tissue expression profiles, which is widely used in immune microenvironment analysis (8). CIBERSORTx algorithm under batch mode was applied to compute the immune cell fraction in liver biopsy samples, utilizing the LM22, a validated leukocyte gene signature matrix, as the reference. The differences of immune cell fraction between BA liver and normal liver were analyzed by t-test. A further correlation analysis was evaluated by Pearson correlation coefficient, and P-value under 0.05 was considered significant.

**Exploring the Correlation of Liver Status and Immune Microenvironment**

The matched liver inflammation and fibrosis grades in GSE15235 were obtained from a previous study (14). In brief, inflammation grades were assessed by hematoxylin/eosin stain. According to the hematoxylin/eosin stain, no inflammation was considered grade 0, while portal expansion together with brisk inflammation in >50% portal tracts was considered grade 3. Gomori trichrome stain was used to assess the fibrosis grade, and no fibrosis was rated as stage 0, while portal fibrosis with expansion and bridging in >50% portal tracts or regenerative nodule was rated as stage 3. All the grades can be accessed in the previous study (14). After utilizing CIBERSORTx to decode the immune cell fractions in GSE15235, the expression of overlapping genes generated...
by the Venn diagram and the immune cell fractions were included in the correlation analysis. The Spearman correlation coefficient evaluated the correlations, and P-value under 0.05 was considered significant.

RESULTS

Identification of DEGs in BA

In data set GSE46960, 402 DEGs were identified by limma test, satisfying the criterion of $|\log2FC| \geq 1.2$ and $P$-value $< 0.05$. Under the same criterion, 1,278 DEGs were discovered by DESeq test in GSE159720. By comparing the DEGs in the two data sets, the expression of 128 genes was significantly changed in both data sets. Eight of the 128 duplicated genes were found in ImmPort, suggesting that these DEGs were related to the immune process of BA (Figure 1A). The volcano plots showed the DEGs in each data set (Figures 1B,C), and eight immune-related genes are introduced in Table 1.

Functional Enrichment for BA

Differentially expressed genes in the datasets were selected to perform KEGG and GO functional enrichment analyses, utilizing the gprofiler tool to explore the biological effects, as shown in Figures 2A,B. In GSE46960, the pathways were mainly focused on extracellular matrix in both GO and KEGG enrichment. In GSE159720, besides the pathways related to the extracellular matrix, more immune-related functions were enriched, such as chemokine activity, cytokine activity, and chemokine receptor binding. Figure 2C demonstrates the same genes that are shared by both GSE46960 and GSE159720 with purple arches, and the blue lines link the different genes, where they fall into the same ontology term with $p < 0.05$. To further explore the pathways that were critical in both gene sets, a functional enrichment analysis, which took consideration of GSE46960 and GSE159720 at the same time, was performed by Metascape (Figure 2D). Pathways of leukocyte migration, chemotaxis, and inflammatory response were in the top 20 clusters of the enriched pathways.

Hub Genes of BA

Differentially expressed genes which overlapped in GSE46960 and GSE159720 were analyzed in STRING to evaluate the interaction between these genes further, and a total of 85 nodes and 163 edges were identified from the PPI network; 11 disconnected nodes in the network were hidden, as shown in Figure 3A. CytoHubba identified the top five hub genes
TABLE 1 | List of eight immune-related differentially expressed genes in biliary atresia (BA) and their functions in the progress of BA.

| Gene symbol | Gene name         | Origin                  | Function in BA                          |
|-------------|-------------------|-------------------------|-----------------------------------------|
| CXCL8 (19)  | C-X-C motif chemokine ligand 8 | BECs                  | Promoting ductular reaction and associated liver fibrogenesis |
| CXCL6 (20)  | C-X-C motif chemokine ligand 6  | Unknown               | Stimulating Kupffer cells releasing TGF-β and activating stellate cells |
| CXCL10 (21) | C-X-C motif chemokine ligand 10 | BECs                  | Stimulation of monocytes, natural killer, and T-cell migration |
| CCL20 (22)  | C-C motif chemokine ligand 20  | BECs                  | Th17 cell recruitment                   |
| IL32 (23)   | Interleukin 32     | BECs                  | Amplification and continuance of periductal inflammatory reactions |
| SLIT2 (24, 25) | Slit guidance ligand 2 | Unknown               | Activating hepatic stellate cells       |
| SPP1 (26)   | Secreted phosphoprotein 1 | BECs                  | Stimulating T-cell proliferation and inducing T-cells and macrophages to express other Th1 cytokines |
| TGFβ2 (27, 28) | Transforming growth factor beta 2 | BECs, hepatocytes, and mesenchymal cells | Unknown                                 |

BECs, biliary epithelial cells.

According to the degree algorithms (Figure 3B), and CXCL8 was ranked as number one with 14 degrees, proving that the immune system may play a considerable role in the pathophysiology of BA.

Immune Microenvironment of BA Liver

The CIBERSORTx deconvolution algorithm was applied to access the immune cell compositions of BA liver according to the GSE46960 dataset. Figure 4A summarized the 22 kinds of immune cell compositions from 64 BA liver and seven normal livers. Six of the 22 immune cell proportions were significantly different between BA and normal liver (Figure 4B). As shown in Figure 4B, M0 macrophages and resting memory CD4+ T cells were upregulated in BA liver compared with normal liver. Monocytes, resting natural killer (NK) cells, plasma cells, and regulatory T (Treg) cells were downregulated. The results of the correlation analysis between LM22 (Figure 4C) found that monocytes and Treg cells have the most intense positive relationship with $r = 0.56$ ($p < 0.05$). Except for the strong negative relationship between all kinds of resting cells and activated cells, Treg cells, and resting memory CD4+ T cells had the highest negative correlation coefficients with $r = −0.57$ ($p < 0.05$).

Correlation Between Liver Immune Microenvironment and Disease Status

In GSE15235, we found 47 BA patients with clinical information, such as survival time of native liver, liver function, and the pathology grade of liver inflammation and fibrosis. After excluding patients without pathology grades of liver inflammation and fibrosis, 46 BA patients were included in the correlation analysis. Eight DEGs relating to immune system and immune cell proportions that were significantly regulated in BA liver were incorporated into the analysis to figure out the relationship between these parameters and the pathology grade of liver inflammation and fibrosis. SLIT2 and CXCL6 own high positive correlation coefficients with fibrosis grade, while the proportion of resting NK cells was negatively correlated (Figure 5A). The proportions of resting CD4+ memory T cells are strongly related to the inflammation grade of BA liver (Figure 5B).

DISCUSSION

Biliary atresia is a fatal neonatal cholestatic liver disease with unknown etiology. However, sights were increasingly focused on the immune response to figure this puzzle out. Recently, Wang et al. (7) and Taylor et al. (29) used single-cell RNA sequencing to demonstrate the abnormality of immune cells in BA patients, which further proved the role of immune response in BA pathogenesis. Here, we utilized the public gene expression data set and deconvolution algorithm of BA liver to explore its immune-related genes and immune microenvironment. In this study, eight immune-related DEGs and six subtypes of immune cells were identified as critical factors in the progression of the disease.

Most of the immune-related DEGs that we found in this study were chemokines, which may greatly contribute to the pathogenesis of the BA. CXCL8, a chemotactic factor that guides neutrophils to the site of infection, was identified as the hub gene of BA in this study. The expression of CXCL8 was high in liver and peripheral blood in BA patients. In peripheral blood, it mainly originated from monocytes, neutrophils, T cells, and B cells (30). Meanwhile, it was primarily expressed in the hepatic parenchyma in liver (31). CXCL8 in BA liver was predominantly in cholangiocytes within areas of ductular reaction, which may play an essential role in BA progress by mediating the ductular reaction and liver fibrogenesis (19). It was believed that both expressions of CXCL8 in the liver and peripheral blood were a predominant part in BA progress and can be used as prognosis predictors of BA patients (32).

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SLIT2 is an immune-related gene that regulates cell growth and migration (24). It can recruit immune cells, such as macrophages (33, 34). However, the mechanism may be different in the liver. Former studies found that liver fibrosis was mediated by activating hepatic stellate cells through the Slit2/Robo1 and Slit2/Robo2 signal pathway (24, 25). It has been proved that the hepatic expression of SLIT2 was significantly increased in patients with primary biliary cirrhosis and in bile duct ligation mouse model (35). CXCL6 is a chemokine which can recruit neutrophil, leading to tissue damage and prolonged inflammatory responses (36). Recently, CXCL6 was found to be upregulated in the serum and liver tissue of high-stage liver fibrosis patients (20) and was supposed to participate in fibrogenesis by stimulating Kupffer cells releasing TGF-β and thereby activating stellate cells. This study finds that the expression of SLIT2 and CXCL6 in BA liver is exceptionally high, and it is strongly related to liver fibrosis grade.
The microenvironment of BA was identified by CIBERSORTx. Natural killer cells, monocytes, macrophages, Treg cells, and plasma cells were significantly different in BA and normal livers, as reported in previous research. Moreover, a more specific fraction of immune cells was computed, and correlation analysis was applied to figure out the relationship between microenvironment and liver status of BA patients.

Natural killer cells were found to have adverse effects in BA livers, especially in experimental BA (37–39). Natural killer cells can be activated by dendritic cells in the RRV BA model, thus injuring the biliary trees (40). The deletion of NK cells further proved that the activation of NK cells plays an essential role in bile duct injury (38). Alexandra et al. (41) supposed that CXCL9 and CXCL10 were secreted by cholangiocytes after a virus infection. Natural killer cells were activated via NKG2D ligands expressed by infected cholangiocytes and migrated to the liver and biliary tree along the chemokine gradient of CXCL9 and CXCL10, causing an autoimmune response against biliary epithelium (41). In our research, although the proportion of NK cells activated is similar between normal liver and BA liver, the proportion of NK cells resting is significantly decreased in BA liver. This phenomenon may be caused by NK cells exhausting after immune response in BA liver. The correlation analysis of this research also revealed that NK cells resting was negatively correlated with chronic fibrosis of BA liver in humans, as reported formerly (42), which may give a clue in retarding the fibrosis procedure in BA patients. The inflammation grade of BA liver in our study reflects an acute inflammation status of BA liver. The positive correlation between inflammation grade and NK cell resting may reflect the dysfunction of NK cells in killing activated T cells and other liver-resident cells and promote the inflammation response (43).

The proportion of Treg cells was severely reduced according to our research. Treg cells play an essential role in regulating the immune microenvironment and maintaining immune homeostasis (44). In the murine BA model established by RRV, liver Treg cells decreased in both number and function (45), contributing to the progress of BA. It was proved that the proportion of Treg cells in peripheral blood was reduced in BA patients who were positive for cytomegalovirus (46). Although we found that the proportion of Treg cells was significantly reduced in BA liver, the correlation analysis did not detect the relationship between Treg cells and liver fibrosis grade or liver inflammation grade. On the other hand, the correlation analysis revealed that Treg cells had the widest connection with other immune cells. Both the strongest positive and negative correlation coefficients were related to the Treg cells, reflecting its powerful regulation role in BA liver.
The mononuclear phagocyte system, including macrophages and monocytes, was suppressed in BA liver. Although the total amount of macrophage was significantly high in BA liver, the subtypes of functional macrophages, such as M1 and M2 macrophage, did not show any statistical difference. Previous studies did not have a consensus on the effect of macrophage in BA progression (47–49). The proportion of M0 macrophage in BA liver was remarkably higher than the one in normal liver, indicating that macrophages were inactivated. Besides this, the proportion of monocytes was significantly decreased in BA liver, further demonstrating that the mononuclear phagocyte system was dysfunctional. As Wang
et al. (7) have reported, the macrophages in BA liver were under a hypo-inflammation situation. Meanwhile, the Kupffer cell scavenger function was defective, which may explain why the proportion of M0 macrophages was extremely increased in BA liver.

In this research, we applied CIBERSORTx in the exploration of the microenvironment and correlation between immune cell infiltration and BA status. Meanwhile, several limitations inevitably existed in our exploration. Firstly, this result was calculated by the algorithm, which was an estimation of the actual situation. Despite the fact that LM22, the reference signature of immune cells, was widely used in CIBERSORT-related analysis, including liver diseases, no study has validated the reliability of LM22 in liver research, which may misrepresent the immune cell types present and active in BA or normal livers. Besides this, the present research cohort is limited; a prospective was required to validate the results. In any case, the correlation between immune cell infiltration and BA status not only deserves further research but also provides a potential prediction tool.

CONCLUSION

In conclusion, we depicted the microenvironment of BA liver and the correlation between the microenvironment and patient status. The expression of CXCL6, CXCL8, CXCL10, CCL20, IL32, TGFβ2, SPP1, and SLIT2 was significantly different between BA and normal liver. The abnormal accumulation of six types of immune cells played an essential role in BA, such as the high proportions of M0 macrophages and resting memory CD4+ T cells and the low proportions of monocytes, resting NK cells, plasma cells, and Treg cells. Our results might provide a clue for further exploration of BA etiology and promote a potential prediction model based on immune infiltration features.

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article-supplementary material, further inquiries can be directed to the corresponding author/s.

AUTHOR CONTRIBUTIONS

QX, YL, and JZ designed this research. JZ and MF collected data and wrote the manuscript. JZ and YL analyzed data. All authors read and approved the final manuscript.

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