Single-cell sequencing unveils distinct immune microenvironment with CCR6-CCL20 crosstalk in human chronic pancreatitis

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ABSTRACT

Objective: Chronic pancreatitis (CP) is a potentially fatal disease of the exocrine pancreas, with no specific or effective approved therapies. Due to difficulty in accessing pancreas tissues, little is known about local immune responses or pathogenesis in human CP. We sought to characterize pancreas immune responses using tissues derived from patients with different etiologies of CP and non-CP organ donors in order to identify key signaling molecules associated with human CP.

Design: We performed single-cell level cellular indexing of transcriptomes and epitopes by sequencing (CITE-Seq) and T cell receptor sequencing of pancreatic immune cells isolated from organ donors, hereditary, and idiopathic CP patients who underwent total pancreatectomy. We validated gene expression data by performing flow cytometry and functional assay in the second CP patient cohort.

Results: Deep single-cell sequencing revealed distinct immune characteristics and significantly enriched CCR6+CD4+ T cells in hereditary compared with idiopathic CP. In hereditary CP, a reduction in T cell clonality was observed due to the increased CD4+ T (Th) cells that replaced...
tissue-resident CD8⁺ T cells. Shared TCR clonotype analysis among T cell lineages also unveiled unique interactions between CCR6⁺ Th and Th1 subsets, and TCR clustering analysis showed unique common antigen binding motifs in hereditary CP. In addition, we observed a significant upregulation of the CCR6 ligand (CCL20) among monocytes in hereditary CP as compared with those in idiopathic CP. The functional significance of CCR6 expression in CD4⁺ T cells was confirmed by flow cytometry and chemotaxis assay.

**Conclusion:** Single-cell sequencing with pancreatic immune cells in human CP highlights pancreas-specific immune crosstalk through the CCR6-CCL20 axis that might be leveraged as a potential future target in human hereditary CP.

**Significance of this study**

**What is already known about this subject?**

- Chronic pancreatitis (CP) is considered an irreversible fibroinflammatory pancreatic disease and remains a major source of morbidity among gastrointestinal diseases with no active approved therapy.

- Inflammation is a known hallmark and contributor to CP pathogenesis. However, little is known about local immune responses in human CP especially with different etiologies.

**What are the new findings?**

- Single-cell RNA sequencing of pancreatic immune cells from CP patients and organ donors revealed distinct immune transcriptomic features in CP versus non-diseased controls and hereditary versus idiopathic CP.
- Single-cell T cell receptor sequencing unveiled pancreas-specific clonal expansion in CD8$^+$ T cells and CD4$^+$ T cells-driven unique TCR repertoire changes in hereditary CP.

- We identified that the CCR6-CCL20 axis was significantly upregulated in hereditary CP compared with controls or idiopathic CP suggesting a potential future target for human hereditary CP.

**How might it impact on clinical practice in the foreseeable future?**

- The results of this study will improve our understanding of CP heterogeneity and identify distinct immune responses in different types of human CP that could provide novel conceptual directions for therapeutic strategies in treating hereditary and idiopathic CP.

- The CCR6-CCL20 axis found in hereditary CP could be a potential novel targetable signaling pathways in the treatment of hereditary CP.

**INTRODUCTION**

CP is a progressive fibro-inflammatory disease of the exocrine pancreas and remains a major source of morbidity yet remains an untreatable disease so far$^{1,2}$. CP is characterized by morphological changes in the pancreas including acinar cell atrophy, distorted pancreatic ducts, inflammation, and fibrosis$^{3,4}$. Primary risk factors in adult CP include alcohol and smoking, but genetic variants and idiopathic factors are significant contributors for CP of all ages$^{1,5-8}$. Over the past two decades, many animal models have been used to understand disease pathogenic mechanisms in CP$^{9,10}$. However, questions remain regarding the translational accuracy of pre-clinical studies. For instance, most pancreatitis animal studies are dependent on disease inducing-
agents that may not mimic the human CP condition\textsuperscript{11}. Therefore, research with human pancreas tissues is critical to understand human disease-specific pathogenic mechanisms in CP.

Inflammation is a hallmark of CP\textsuperscript{4,12}, and immune cells have emerged as key contributors in CP and its progression\textsuperscript{13,14}. For example, anti-inflammatory (M2) macrophages were observed in different types of CP experimental models\textsuperscript{15,16} and contribute to CP fibrogenesis through the crosstalk with pancreatic stellate cells (PSCs)\textsuperscript{17}. A growing number of studies demonstrated the critical role of immune cells at the different stages and disease progression in CP providing potential therapeutic targets for the disease\textsuperscript{18-20}. However, immune characteristics remain poorly understood in human CP due to the limited access to human pancreatic tissues, and the contribution of disease etiologies to disease heterogeneity remains unexplored.

In collaboration with an institution that performs high volume of total pancreatectomy islet auto-transplantation (TPIAT) in CP patients, our pilot study using flow cytometric analysis and bulk T cell receptor (TCR) sequencing revealed distinct immune responses between hereditary and idiopathic CP implicating different immunopathogenic mechanisms underlying two different subtypes of CP\textsuperscript{21}. Here we performed CITE-seq (combined single-cell antibody-derived tag and RNA sequencing)\textsuperscript{22} and single-cell TCR sequencing (scTCR-seq) of immune cells isolated from human pancreas tissues in CP patients and organ donors for a more in-depth understanding of immune responses associated with the disease. This unbiased systemic analysis of immune signatures that included protein expressions, transcriptomes, and TCR repertoire analyses of pancreatic immune cells revealed distinct and unique disease-specific immune cell signatures and interactions that provide insights into novel immune signaling pathways and offer potential future therapeutic target(s) for human CP.
RESULTS

Human pancreatic immune cell transcriptional atlas reveals unique disease-specific signatures in CP

Exocrine pancreatic tissues collected from the first cohort, organ donors (n=3) and CP (hereditary, n=5 and idiopathic n=4) patients, who underwent total pancreatectomy (table 1, online supplemental table 1,2 and online supplemental figure 1A), were used to isolate immune cells by immune cell enrichment method and fluorescence-activated cell sorting (FACS)\textsuperscript{21}. Next, sorted live CD45\textsuperscript{+} pancreatic immune cells were further stained with 13 different surface protein antibodies for CITE-seq. The stained cells went through a droplet-based gel-bead barcoding system (10x genomics)\textsuperscript{23}, which enables constructing independent maps of immunophenotypes of surface protein expressions, transcriptomes (5’ scRNA-seq), and the TCR repertoire (scTCR-seq) simultaneously in the same cells (figure 1A and online supplemental table 3). After quality control (online supplemental figure 1B) and removing doublets, we retained a total of 28,547 single cells and performed clustering analyses with gene expression data. This revealed 17 different cell populations that were visualized as uniform manifold approximation and projection (UMAP) embeddings (figure 1B and online supplemental data 1). Classification of cell populations was annotated by specific gene expressions that were inferred from various human single immune cell RNA-seq studies (figure 1C and online supplemental figure 2A)\textsuperscript{24-26}. Major pancreatic immune cells that we identified from controls and CP were T cells, myeloid cells, B cells, and mast cells (figure 1B and online supplemental figure 2B). Noticeably, cells did not topologically cluster by experimental batch or individual subject, but cells from each group including control, hereditary, or idiopathic CP grouped together in a distinctive manner,
highlighting the impact of disease and its etiology on the immune transcriptome (figure 1D and online supplemental figure 2C).

We further analyzed the frequency and composition of immune populations within each group or individual subject by identified immune clusters. We found major immune cells that consist of control pancreas tissues were myeloid cells while a higher proportion of T cells contributed to pancreatic immune cells from CP tissues (figure 1E and online supplemental figure 3A, B). Among T cells and myeloid cells, distinct cell clusters were enriched in each group; macrophages including cluster #5, 9, and 11 were enriched in control, and distinct T cell clusters were expanded in CP groups. Cluster #2 was mainly enriched in hereditary CP while cluster #0 was predominant in idiopathic CP (figure 1F and online supplemental figure 3C, D). We also confirmed surface protein marker expression patterns in pancreatic immune cells such as CD3, CD8, CD11B, and HLA-DR by visualizing their expressions in the UMAP created by transcriptome data across all immune clusters (online supplemental figure 4A, B).

Next, we examined disease-specific gene signatures by identifying differentially expressed genes (DEGs) in CP versus control (online supplemental figure 5A, B). The top 20 upregulated genes in CP compared with controls included *HSP90AA1, FTH1, TNFAIP3, NFKBIA, NR4A1, CD69*, and *CCL20*. Furthermore, we performed Gene set enrichment analysis (GSEA) of DEGs in CP versus controls, and the results highlighted strong signatures for inflammatory responses including apoptosis, hypoxia, IL2-STAT5, interferon-gamma, and TNFA signaling (online supplemental figure 5C, D). Overall, single-cell transcriptome data analysis with pancreatic immune cells from control and CP tissues indicate disease-specific immune responses in local pathogenic area and distinct immune transcriptomic and protein expression signatures between hereditary and idiopathic CP.
Distinct transcriptomic characteristics of pancreatic T cells between hereditary and idiopathic CP

One of the major immune populations infiltrating the pancreas of control and CP subjects were T cells. In order to scrutinize T cell transcriptomes in a higher resolution, we analyzed T cell populations (clusters # 0, 1, and 2 of total cell clusters, figure 1B) separately. We complied gene expression data from 15,913 cells for T cell clustering analyses and identified 13 different T cell clusters with population nomenclature annotated by specific marker gene expressions (figure 2A, B and online supplemental figure 6A). UMAP with T cell populations revealed markedly separated cell clusters by three groups, control, hereditary and idiopathic CP (figure 2C). Control cells comprised predominantly GZMA⁺ cytotoxic CD8⁺ T cells while distinct CD4⁺ or CD8⁺ T cell subpopulations constituted CP T cells (figure 2D, E and online supplemental figure 6B, C). Hereditary CP cells mainly consisted of CD4⁺ helper T cell (Th) subpopulations including CCR6⁺, TNF⁺ (Th1), regulatory T (Treg) cells, and HLA-DA⁺ CD8⁺ T cells. However, FTH1⁺ CD4⁺ or CD8⁺ T cells were predominant T cells in idiopathic CP (online supplemental figure 6D, E).

Furthermore, within these T cell subpopulations, we identified DEGs and performed GSEA of upregulated genes in CP versus control (online supplemental figure 7A, B), and in hereditary CP versus idiopathic CP (figure 2F, G). Some of the noticeable genes that were upregulated in hereditary compared with idiopathic CP included chemotactic receptors and ligands such as, CCR6, CXCR4, GPR183, and CCL20, suggesting their involvement in CD4⁺ T cell recruitment in hereditary CP.
CD8+ T cell dependent unique TCR repertoire changes in CP

Next, we examined the TCR repertoire of the same T cells from which we extracted transcriptome data in control and CP. Paired transcripts, TCR alpha (TRA) and TCR beta (TRB), two distally encoded but co-expressed in single cells were sequenced and filtered through a programmed filtering system (Cell Ranger, 10x genomics). After barcode correction and trimming, V(D)J genes in the complementary determining region 3 (CDR3) of TCR transcripts were annotated. We identified a total of 12,856 unique paired ab TCR sequences and 5,180 T cells with paired TRA and TRB CDR3 sequences from the T cell repertoire of 12 control and CP subjects. In control T cells, 15.6 ± 6.0% of unique clonotypes were shared by 2 or more cells, which was significantly higher than that of hereditary CP T cells (4.03 ± 1.6%) (figure 3A). Corresponding to this result, the Gini-coefficient (an index of clonality) of control T cells was also significantly higher compared with the hereditary CP group indicating lower clonal expansion in hereditary CP T cells (figure 3B). Next, we assigned TCR sequences to cells with cluster identities as shown in Fig. 2A. Gini-coefficients of different T cell clusters revealed CD8+ T cell-skewed clonal expansion in control and CP groups (figure 3C). When we split T cells into CD8+ T cells and the rest, the Gini-coefficient of CD8+ T cells was significantly higher than that of the remainder T cells (figure 3D). This result was further confirmed by comparing UMAPs of CD8A gene expression and clonally expanded cell populations (figure 3E). Finally, we observed that Gini coefficients of subjects had a positive correlation with CD8+ T cell frequency in control and CP groups while Gini-coefficients had a negative correlation with CD4+ T cell frequency in all subjects (figure 3F). These data implicate that a significantly higher frequency of CD4+ T cells contributes to the lower Gini-coefficient and clonal expansion in hereditary CP compared with controls.
These scRNA/TCR-seq results revealed that the degree of T cell clonal expansion or diversity in CP is mainly altered by the frequency of cytotoxic CD8$^+$ T cells and infiltration of pathogenic CD4$^+$ T cells. In hereditary CP, infiltrating CD4$^+$ T cells likely dilute the effect of tissue-resident cytotoxic CD8$^+$ T cells on TCR clonal expansion observed in control pancreatic tissues. This may provide key insight on CD4$^+$ T cell-mediated pathogenic response especially in the case of hereditary CP.

**Unique interactions among T cell lineages and shared antigen-binding motifs in CP**

In order to understand the connection and origin of T cell clusters, we examined the unique clonotypes (matched single-cell TCR-αβ profiling) and their overlaps between different clusters by analyzing clonotypes across all groups (figure 4A) or each group (online supplemental figure 8A, B, and C). Although higher Gini-coefficient and expanded clones were observed in control versus CP, no single clonotype was shared between different clusters in controls. Major clonotype overlaps were found in both hereditary and idiopathic CP (online supplemental figure 8A, B, and C). The greatest clonotype overlaps were found between CD8$^+$ cytotoxic T cell clusters (Tc-1 and Tc-3) with 40 shared clonotypes, which were contributed by idiopathic CP. In fact, 63 unique shared clonotypes were found among CD8$^+$ cytotoxic T cell clusters in idiopathic CP (figure 4A and online supplemental figure 8C). The next highest clonotype overlaps were found between CCR6$^+$ CD4$^+$ T cells and Th1 cells with 23 shared clonotypes, and these overlaps mainly occurred in hereditary CP (figure 4A and online supplemental figure 8B). These shared TCR clonotypes elucidated unique interactions among different T cell subtypes and clonal dynamics in each CP group. Specifically, a large number of shared clonotypes between CCR6$^+$ CD4$^+$ T cells and Th1 cells in hereditary CP suggests that
these two Th cell subsets may have a shared origin. Interestingly, our analysis demonstrated overlaps between CD4+ and CD8+ T cell clusters, which may require further analyses regarding the length of CDR3 sequences in overlapped clonotypes as there is a higher probability that significantly short CDR3 sequences to be less impactful in recognizing a particular MHC class.

To examine whether pancreatic TCRs found in controls and CP recognize the same antigens, we performed GLIPH2 analysis which enables clustering TCRs that recognize the same epitope by screening shared antigen-binding motifs on the TCR CDR3β amino acid sequences. A number of motifs shared by at least two different individuals from control donors and CP patients were selected based on final scoring, and among those, three clusters were shown as they also possess a certain level of similarity on their paired CDR3α sequences (figure 4B). The results showed that selected clusters were mainly shared by individuals from hereditary and idiopathic CP, not from controls. Remarkably, most of the TCRs from the three candidate clusters with unique common motifs were mainly shared by individuals from hereditary CP, which indicates T cells in hereditary CP have a higher possibility of reacting against common antigens or epitopes. Consistent with scRNA-seq results of T cells, this scRNA/TCR-seq data also supports that there are distinct immune responses, specifically T cell antigen reactions, between hereditary and idiopathic CP.

**Distinct transcriptional alteration of human pancreatic myeloid cells in CP**

Myeloid cells have been known to play significant roles in CP and interact with non-immune cells such as PSCs. Myeloid cells were another key contributor among pancreatic immune subsets found in the pancreas of control and CP subjects (figure 1B). When we further analyzed myeloid clusters separately (clusters # 3, 4, 5, 6, 9, and 11 in figure 1B), 11 different
myeloid subclusters were identified that included monocytes (Mono), macrophages (Mac), and
dendritic cells (DC) (figure 5A). Remarkably, cells belonging to each group (control, hereditary
CP, or idiopathic CP) gathered together and topologically separated from each other with
minimal overlaps when visualized in UMAP (figure 5B). Myeloid cell clusters were manually
annotated by distinct cell marker gene expressions (figure 5C and online supplemental figure
9A). Importantly, the frequency of CD68+ and CD163+ macrophages and S100A8+ monocytes
(clusters #3, 8) were significantly higher in controls compared with CP (figure 5D, E and online
supplemental figure 9B, C). In CP, distinct monocyte and DC populations contributed to the
myeloid compartment of hereditary or idiopathic CP immune cells. The most noticeable finding
was a significantly higher frequency of CCL20+ monocytes in hereditary CP compared with
control or idiopathic CP. Furthermore, DEG analysis in CP versus control revealed various
inflammatory and chemoattractant molecules such as IL1B, CXCL2, CXCL3, CXCL8, and CCL4,
which were significantly upregulated with enriched inflammatory signaling pathways in CP
myeloid cells (online supplemental figure 9D, E). DEG analysis in hereditary versus idiopathic
CP also identified various genes including HLA-DR, CCL20, and IL1B as significantly
upregulated with enrichment of allograft rejection pathway in hereditary CP compared with
idiopathic CP (figure 5F, G). This data further implicates that hereditary CP with higher HLA
molecule expressions may associate with autoimmune responses although specific HLA typing
would be necessary to confirm this notion29,30. Overall, single-cell transcriptional analysis of
pancreatic myeloid populations revealed distinct myeloid subpopulation enrichment in each
group; macrophages in controls, CCL20+ monocytes and IL1B+ DCs in hereditary CP, and
FTH1+ or S100A8+ monocytes in idiopathic CP. Especially, significantly enriched CCL20+
monocyte population in hereditary CP suggests a potential role as a chemoattractant for the infiltrating CCR6+ CD4+ T cells.

**Functional analysis implicates the CCR6-CCL20 axis in hereditary CP**

Consistent with the transcriptional expression of CCR6-CCL20 axis in CD4+ T cell and monocyte populations from hereditary CP, we confirmed this unique chemokine and receptor axis again by analyzing DEGs between hereditary and idiopathic CP within total immune populations (figure 6A). Accordingly, **CCR6** and **CCL20** were listed on the top 20 genes significantly upregulated in hereditary CP compared to idiopathic CP. Furthermore, to assess the crosstalk between myeloid cells and T cells, we analyzed cytokine or chemokine receptor-ligand interactions between myeloid cells and T cell compartment by combined expression of receptor-ligand (figure 6B). CCR6-CCL20 axis was uniquely expressed only in both CP groups, and hereditary CP had a higher combined expression of CCR6-CCL20 than idiopathic CP. Finally, we assessed the protein expression of CCR6 in T cells isolated from the second cohort of controls and CP subjects by flow cytometry (figure 6C, table 2, and online supplemental table 4,5). First, we were able to confirm a significantly reduced frequency of CD8+ T cells and a significantly higher frequency of CD4+ T cells in hereditary CP compared with controls. Consistent with scRNA-seq results, the frequency of CD8+ or CD4+ T cells in idiopathic CP was in the middle, ranging between that of controls and hereditary CP. More importantly, the percentage of CCR6 expression in CD4+ T cells was significantly augmented in hereditary CP compared with controls or idiopathic CP. After we confirmed the elevated protein expression of CCR6 in CD4+ T cells in hereditary CP, we next assessed the functional significance of the CCR6 expression in T cells by CCL20-mediated cell migration assay (figure 6D). Transwell
chemotaxis was performed, and CXCL12 was used as a strong positive control, which is standard in T cell migration assays\textsuperscript{31}. Pancreatic immune cells from hereditary CP tissues migrated to soluble, recombinant human CCL20 at a concentration comparable to previous reports\textsuperscript{32,33}. The increased chemotactic responsiveness of the pancreas infiltrating T cells to CCL20 confirmed the functional significance of the increased receptor (CCR6) expression in hereditary CP. These data suggest that T cells migrate and infiltrate the diseased pancreas in hereditary CP through the CCR6-CCL20 axis.

DISCUSSION

Human CP studies have been hindered in part due to limited access to clinical specimens. By leveraging the ability to access large fractions of exocrine pancreas tissues from organ donors and CP patients undergoing islet isolation, we were able to perform in-depth single-cell level immune analyses using state of the art technologies. Our studies did not include CP related to alcohol due to an insufficient number of patients undergoing total pancreatectomy with islet transplantation with this etiology\textsuperscript{34}, but we were able to receive a sufficient number of cases with hereditary and idiopathic CP for in-depth immune analyses. Our previous study using flow cytometry and bulk TCR-sequencing revealed distinct immune characteristics between hereditary and idiopathic CP highlighting unprecedented critical insights into distinctive disease pathogenesis of CP with different etiologies\textsuperscript{21}. To further delineate pathogenic signals and uncover differences in immune responses underlying different etiologies of CP, here we used integrative single-cell multi-omics sequencing analyses to assess the TCR repertoire and RNA/protein expressions simultaneously at a single-cell level. These unbiased in-depth analytic technologies enabled comprehensive comparison between control and CP, as well as between
different CP groups and identified novel immune subsets, their crosstalk, and T cell subset based TCR repertoire changes. Consistent with our previous study\(^{21}\), hereditary and idiopathic CP had distinct immune characteristics not only in frequencies of immune subpopulations but also in their unique gene expression patterns and the degree of T cell clonal expansions among different subsets. Furthermore, T cell analyses revealed a significant portion of CD4\(^+\) T cell subsets (\(CCD6^+\) Th, \(TNF^+\) Th1, and Treg) were enriched in the hereditary CP pancreas highlighting the immune signature of hereditary CP pancreas is largely conserved across heterogeneous CD4\(^+\) T cell subsets. However, idiopathic CP showed evenly distributed CD4\(^+\) and CD8\(^+\) T cell subsets uniquely with \(FTH1\) and \(PCBP2\) expressions, which indicate that human idiopathic CP is highly associated with iron metabolism in immune cells and dysregulated iron metabolism has been reported in pancreatitis\(^{35-37}\).

Distinct T cell characteristics between hereditary and idiopathic CP were also seen in the TCR repertoire analysis. Pancreatic T cells revealed an increased clonal expansion in CD8\(^+\) T cell subsets in all control and CP groups, which led to significantly lower T cell clonality in hereditary CP where CD4\(^+\) T cell infiltrates are increased. These data suggest that differences in the TCR repertoire found in control versus CP or hereditary versus idiopathic CP were highly relevant to the distribution of specific T cell subsets. In contrast to reports of TCR repertoires in auto-immune, infectious, and malignant diseases\(^{24,38-40}\) where newly infiltrating immune cells are usually responsible for the increased clonality, here infiltrating pancreatic CD4\(^+\) T cells contribute to the reduced clonality in the diseased states but increased heterogeneity of overall T cell subsets. In addition, shared TCR clone analysis with scRNA/TCR-seq data unveiled the unique interactions between CCR6\(^+\) Th and Th1 subsets in hereditary CP supporting naturally increased CD4\(^+\) T cell heterogeneity in the microenvironment of hereditary CP. The unique
pancreas-specific TCR repertoire found in hereditary and idiopathic CP will contribute to a better understanding of pathogenic T cell infiltration and behavior, which have not been reported previously and likely to provide fundamental clues of pathogenic mechanisms and/or disease progression in CP. Given distinct TCR repertoire changes and common antigen-binding motifs found in hereditary versus idiopathic CP, future studies that include predicting HLA restriction and antigen screening against disease-specific TCR candidates will be important.

One of the striking findings in the single-cell analyses was the upregulation of the CCR6-CCL20 axis in hereditary CP. Upregulation of CCL20 by pro-inflammatory cytokines such as IL-1b and TNF-a and infiltrating CCR6+ lymphocytes have been found in microenvironments of inflammatory, infectious, and malignant states in various organs such as the gut, intestine, liver, and lung. CCL20 expression pattern has been compared in pancreatic cancer versus CP and its tumor-promoting role in pancreatic cancer has been proposed, however, CCR6+ T cell infiltration in hereditary or any type of CP to our knowledge has not been reported. Our findings also support CP-specific crosstalk between CD4+ T cells and monocytes through the CCR6-CCL20 axis in hereditary CP, a novel insight into possible precision-targeting of this disease. Interestingly, among the different causes of CP, hereditary CP has the highest risk for developing pancreatic cancer with a cumulative increased risk of up to 40%. The significant upregulation of the CCL20-CCR6 axis in hereditary as compared to idiopathic CP (and organ donors) might contribute to the increased risk of pancreas cancer in hereditary CP although further studies have to be done to confirm this correlation.

We focused our analysis and reporting on the major immune cell subsets, but we also found that B cells and mast cells were distinctly distributed in hereditary versus idiopathic CP although their frequencies were limited to less than 10% of total immune cells. It will be
important in the future to analyze these cells in-depth and examine their potential functional significance in CP since there are several reports suggesting their potential pathogenic roles in pancreatitis. Collectively, our approaches with integrative single-cell analyses unveiled distinct pancreatic immune signatures and pathways between different etiologies of CP, thereby expanding our understanding of pancreatic immune cell signaling and function in the disease-specific pathogenic microenvironment. In addition, our study contributes to the growing literature of gaining insight into the characterization and function of human pancreas immune cells in health and disease.

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**Author contributions**

B.L. and A.H. designed experiments and wrote manuscript. A.H. provided overall guide and supervision. M.D.B. and D.H. provided CP patients’ tissues and clinical information. G.L.S provided control pancreatic tissues and clinical information. B.L. performed all experiments and
analyzed data, and H.N. performed FACS to prepare single cells for single-cell sequencing, flow cytometry panel design, and data analysis. Y.Y. analyzed CITE-seq and scTCR-seq data. H.H. performed GLIPH2 analysis, interpreted and reviewed scTCR-seq results. M.M.D., S.J.P., and M.D.B reviewed manuscript and participated in the interpretation of data.

**Competing interests**

Authors declare that they have no competing interests.

**Patient and public involvement**

Patients or the public were not involved in the design, conduct, reporting, or dissemination plans of this research.

**Patient consent for publication**

Not required.

**Ethics approval**

This study was approved by Institutional Review Board of the Stanford University. Patients or the public were not involved in the design, conduct, reporting, or dissemination plans of this research.

**Data and materials availability**

All raw and processed sequencing data have been deposited with Gene Expression Omnibus database, GSE165045. Code used for data analysis is available at [https://github.com/yangysheep2018/CITE-seq-TCR_paper](https://github.com/yangysheep2018/CITE-seq-TCR_paper). All other processed data are available in the main text or the supplementary materials.

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**Figure legends**

**Figure 1. Human pancreatic immune cell transcriptional atlas.** (A) Experimental design for single-cell sequencing (CITE-seq and TCR-seq) of pancreatic immune cells from CP patients and non-diseased control donors. (B) UMAP plot of all 28,547 pancreatic immune cells from CP patients (5 hereditary and 4 idiopathic CP) and 3 control donors presenting 17 clusters (Mac, macrophages; Mono/Mac, monocytes and macrophages; pDCs, plasmacytoid dendritic cells). (C) Heatmap of signature gene expression z scores across cells. (D) UMAP plot of pancreatic immune cells including 4,169 cells from control donors, 11,786 cells from hereditary CP, and 12,592 cells from idiopathic CP patients colored based on the group (Ctl, control; Her, hereditary CP; Idio, idiopathic CP). (E) Cell cluster frequency shown as a fraction of clusters from total cells in each patient or donor. (F) The frequency of 17 immune clusters defined by scRNA-seq clustering analysis shown as an average proportion of each cluster within each group. One-way ANOVA Kruskal-Wallis test (*p<0.05, **p<0.01, ***p<0.001). The comparison of each pair was differentiated by color: between Ctl and Her, Pink; between Ctl and Idio, Teal; between Her and Idio, Black.

**Figure 2. Distinct pancreatic T cell transcriptional signatures in hereditary versus idiopathic CP.** (A) UMAP plot of 15,913 pancreatic T cells across controls and CP displaying 13 clusters. (B) Heatmap of signature gene expression z scores across cells. (C) UMAP plot of
pancreatic T cells including 1,012 cells from control donors, 7,271 cells from hereditary CP, and 7,630 cells from idiopathic CP patients colored based on the group (Ctl, control; Her, hereditary CP; Idio, idiopathic CP). (D) Cell cluster frequency shown as a fraction of 13 T cell subclusters in each patient or donor. (E) The frequency of 13 T cell subclusters defined by scRNA-seq clustering analysis shown as an average proportion of each cluster within each group. One-way ANOVA Kruskal-Wallis test (*p<0.05, **p<0.01, ***p<0.001). The comparison of each pair was differentiated by color; between Ctl and Her, Pink; between Ctl and Idio, Teal; between Her and Idio, Black. (F) DEGs in between hereditary CP and idiopathic CP. Each dot represents a gene, with significantly up-regulated top 20 genes and down-regulated top 20 genes in hereditary CP versus idiopathic CP colored blue and yellow, respectively. (F) Functional enrichment analysis of significant hallmark gene sets in total T cells from hereditary CP versus idiopathic CP. NES, normalized enrichment score.

Figure 3. scRNA/TCR-seq unveils CD8+ T cell subset dependent unique TCR repertoire changes in CP. (A) Bar graph shows the percentage of unique paired TRA and TRB sequences that are shared by one cell (blue), by two cells (green), or three or more cells (orange) in each donor or CP patient. (B) Gini coefficients of control and CP group across all pancreatic T cell populations. For each group, a box and whisker plot is shown with the median and all values from minimum to maximum. One-way ANOVA with Tukey’s multiple comparison test, *p<0.05. (C) Heatmap showing Gini coefficients of pancreatic T cell clusters in each donor or patient. (D) Gini coefficients of pancreatic CD8+ T cells and other T cells in control and CP groups. (E) UMAP plot of CD8A gene expression in control or CP groups (top). TCR clonality on expression-driven UMAP overlay showing the distribution of clonal and unexpanded cell
populations in cell clusters in control and CP groups (bottom). Q0, unexpanded; Q1, cells encompassing the top 20% of the most expanded clones in the group; Q2-4, cells representing the middle three quantiles; Q5, cells encompassing the bottom 20% of the most expanded clones.

(F) Correlation analysis of Gini coefficients for pancreatic T cells and the frequency of CD8+ T cells (top) or CD4+ T cells (bottom) in control donors and CP patients.

Figure 4. scRNA/TCR-seq uncovers unique interactions in T cell lineages and shared antigen binding motifs in CP. (A) Upset plots displaying TCR clonotypes shared among T cell clusters from T cells across control and CP groups. Each shared clonotype was indicated by black dots with a connected black line. The horizontal bar graph indicates the total number of shared TCR clonotypes for cluster intersections, and the vertical bar graph indicates the number of unique clonotypes found in a single cluster. (B) Representative TCR specificity groups and potential antigen binding motifs by GLIPH2 cluster analysis of TCR CDR3b sequences from pancreatic T cells across control, hereditary and idiopathic CP. Candidates were selected from the clusters with a final score of less than 10^-8 and shared by at least two different individuals.

Figure 5. Distinct pancreatic myeloid cell transcriptional signatures between hereditary and idiopathic CP. (A) UMAP plot of 8,590 pancreatic myeloid cells presenting 11 clusters (DC, dendritic cells; Mac, macrophages; Mono, monocytes). (B) UMAP plot of pancreatic myeloid cells including 2,998 cells from control donors, 2,769 cells from hereditary CP, and 2,823 cells from idiopathic CP patients colored based on group (Ctl, control; Her, hereditary CP; Idio, idiopathic CP). (C) Heatmap of signature gene expression z scores across cells. (D) Cell cluster frequency shown as a fraction of 11 myeloid subclusters in each patient or donor. (E) The
frequency of 11 myelo
id subclusters shown as an average frequency of each cluster within each
group. One-way ANOVA Kruskal-Wallis test (*p<0.05, **p<0.01). The comparison of each pair
was differentiated by color; between Ctl and Her, Pink; between Ctl and Idio, Teal; between Her
and Idio, Black. (F) DEGs between hereditary and idiopathic CP groups. Each dot represents a
gene, with significantly up-regulated top 20 genes and down-regulated top 20 genes in hereditary
CP versus idiopathic CP colored blue and yellow, respectively. (G) Functional enrichment
analysis of significant hallmark gene sets comparing myeloid cells from hereditary CP with
idiopathic CP. NES, normalized enrichment score.

Figure 6. Functional analysis implicates CCR6-CCL20 axis as hereditary CP-specific
pancreatic immune crosstalk. (A) Volcano plot of DEG analysis between hereditary and
idiopathic CP groups with total immune populations. Each dot represents a gene, with
significantly up-regulated top 20 genes and down-regulated top 20 genes in hereditary CP versus
idiopathic CP colored blue and yellow, respectively (left). DEGs ranked by fold changes in
hereditary CP versus idiopathic CP (right). (B) UMAP plot of total pancreatic immune cells
(left). Heatmap of combined expression of selected cytokine/chemokine and their receptor
displaying the interactions between myeloid cells and T cells discovered by CellphoneDB
(right). (C) Flow cytometry analysis of CD8\(^+\), CD4\(^+\), or CCR6\(^+\)CD4\(^+\) T cells (Kruskal-Wallis
test with Dunn’s multiple comparisons test, *p<0.05, **p<0.01). (D) Chemotaxis assay with
pancreatic immune cells from hereditary CP (left, One-way ANOVA with Tukey’s multiple
comparisons test; right, paired t-test; **p<0.01, ***p<0.001).
Tables.

Table 1. Statistical comparisons of demographic and characteristics between control and CP subjects, a cohort for single-cell sequencing analysis.

| Characteristics                                  | Control (n=3) | CP (n=9)   | P value |
|--------------------------------------------------|---------------|------------|---------|
| Age (years, mean ± SD)                           | 44.3 ± 8.2    | 26.2 ± 14.5| 0.08    |
| Sex, m/f                                        | 1/2           | 3/6        | 1       |
| Height (meters, mean ± SD)                       | 1.70 ±0.41    | 1.62 ±0.13 | 0.31    |
| Weight (kg, mean ± SD)                           | 76.1 ± 18.5   | 61.6 ± 15.9| 0.24    |
| Body mass index (kg/m², mean ± SD)               | 26.1 ± 5.1    | 23.2 ± 4.1 | 0.37    |
| HbA1c (% , mean ±SD)                             | 5.1 ± 0.1     | 5.2 ± 0.5  | 0.77    |

SD, standard deviation; IQR, interquartile range.
^Unpaired two-tailed t-test; ^Chi-square was used for p-values, Bold p value, p<0.05.
Table 2. Statistical comparisons of demographic and characteristics between control and CP subjects, a cohort for functional validation analyses.

| Characteristics                        | Control (n=7) | CP (n=12) | P value |
|----------------------------------------|--------------|-----------|---------|
| Age (years, mean ± SD)                 | 31.1 ± 9.6   | 27.5 ± 12.3 | 0.51    |
| Sex, m/f                               | 4/3          | 6/6       | 0.76    |
| Height (meters, mean ± SD)             | 1.69 ±0.41   | 1.67 ±0.90 | 0.58    |
| Weight (kg, mean ± SD)                 | 84.8 ±11.2   | 70.2 ±19.7 | 0.09    |
| Body mass index (kg/m$^2$, mean ± SD)  | 29.6 ±3.5    | 24.8 ±5.3 | 0.05    |
| HbA1c (%)                              | 5.3 ± 0.6    | 5.3 ± 0.3  | 0.97    |

SD, standard deviation; IQR, interquartile range.

*A* Unpaired two-tailed t-test; *B* Chi-square was used for p-values, Bold p value, p<0.05.
Figure 1.

A) Exocrine tissue dissociation & Immune cell enrichment
   - Sorting live CD45+ immune cells
   - Staining immune cells with 13 feature surface antibodies
   - Single-cell analysis of surface proteins (13 feature antibodies) transcriptome (5' scRNA-seq)
   - TCR repertoire (scTCR-seq)
   - Validation with 2nd cohort (Flow cytometry, Chemotaxis assay)

B) UMAP plot showing clusters of different cell types:
   - T cells
   - Myeloid cells
   - B cells
   - Mast cells

C) Heatmap and UMAP plot for T cells and Myeloid cells:
   - Expression of surface proteins
   - Identity
   - Expression

D) UMAP plot showing fraction of total cell clusters in each group:
   - Ctl
   - Her
   - Idio

E) Bar chart showing fraction of clusters within each sample:

F) Line graph showing average frequency of total cell clusters in each group:
   - Ctl
   - Her
   - Idio

Legend:
- Ctl
- Her
- Idio

Additional details:
- Total pancreatectomy with islet isolation
- First cohort (n=12)
- Second cohort (n=19)
- Single-cell analysis of surface proteins (13 feature antibodies)
- Transcriptome (5' scRNA-seq)
- TCR repertoire (scTCR-seq)
- Validation with 2nd cohort (Flow cytometry, Chemotaxis assay)
- Flow cytometry
- Chemotaxis assay
- Exocrine tissue dissociation & Immune cell enrichment
- Sorting live CD45+ immune cells
Figure 2.

A. UMAP plot showing clusters of T cells labeled with different cell types:
- CD4 T cells (CCR6)
- CD8 Tc-1
- CD4 T cells (FTH1)
- CD4 T cells (BAG3)
- CD4 T cells (Th1)
- CD8 Tc-2 (HLA-DR, GZMK)
- CD8 Tc-3 (BAG3)
- CD8 Tc-4
- NK cells
- MALAT1+ cells
- Treg
- Naïve CD4 T cells
- CD8 Tc-5 (NK-like)

B. Heatmap showing expression levels of various genes across different clusters.

C. UMAP plot showing clusters of cells labeled with different samples:
- Ctl
- Her
- Idio

D. Bar graph showing the fraction of clusters within each sample.

E. Graph showing the average frequency of T cell sub-clusters in each group.

F. Scatter plot comparing Her versus Idio, highlighting upregulated and downregulated genes.

G. Table listing pathways with associated gene ranks, NES, pval, and padj values.

Pathways:
- MTORC1_SIGNALING
- UV_RESPONSE_UP
- TNFA_SIGNALING_VIA_NFKB
- ESTROGEN_RESPONSE_LATE
Figure 3.
**Figure 4.**

### A

The figure shows a bar graph with the x-axis labeled as "# of unique TCR clonotypes found in each cluster." The y-axis represents different cell types and subtypes, including Naïve CD4 T cells, MALAT1+ cells, Treg, NK cells, CD8 Tc-1, CD8 Tc-4, CD8 Tc-2 (HLA-DR, GZMK), CD8 Tc-3 (BAG3), Treg, Naïve CD4 T cells, MALAT1+ cells, and CD8 Tc-5 (NK-like). Each bar represents the number of unique TCR clonotypes found in each cluster.

### B

**Table:**

| Candidates | Pattern | Final score | CDR3β | Vβ | Jβ | CDR3α | Sample:Clonotype ID | Freq |
|------------|---------|-------------|-------|----|----|-------|---------------------|------|
| #1         | global-S%GLADT | 6.90E-10    | CAS$^+$GLADTQYF | TRBV19 | TRBJ2-3 | CAVSDRSNYQLIW | Her3:clonotype1052 | 1    |
|            |         |             | CAS$^+$GLADTQYF | TRBV19 | TRBJ2-3 | CAVSERSNYQLIW | Her1:clonotype138  | 1    |
| #2         | motif-DGGNT global-S%DGGNT | 2.80E-09    | CAS$^+$DGGNTAFF | TRBV5-1 | TRBJ1-1 | NA | Her4:clonotype39   | 2    |
|            |         |             | CAS$^+$DGGNTAFF | TRBV5-1 | TRBJ1-1 | CAETANAGKSTF | Her4:clonotype437  | 1    |
| #3         | global-SL%DRGRNE | 1.20E-09    | CAS$^+$S$^+$DRGRNEQFF | TRBV5-4 | TRBJ2-1 | CATAETGFQKLVF | Her4:clonotype1859 | 1    |
|            |         |             | CAS$^+$S$^+$DRGRNEQFF | TRBV5-4 | TRBJ2-1 | CAEKFTDKLIF | Her5:clonotype14    | 4    |
Figure 5.

A Myeloid cells

B

C

D

E

F

G

H

I

J

K

L

M

N

O

P

Q

R

S

T

U

V

W

X

Y

Z

average frequency of sub-clusters in each group

Figure 5.
Figure 6.

A

Her versus Idio (in total cells)

B

B cells

Myeloid cells

T cells

Mast cells

C

C

C

C

D

Migration to CCL20

% of CCR6+ cells in CD4+ T cells

% of CD8+ T cells in CD3 T cells

% of CD4+ T cells in CD3 T cells

Migration to CCL20 (relative to vehicle)

% of migration to CCL20 (relative to vehicle)