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EDITED BY

Alexandre Todorovic Fabro,
University of São Paulo, Brazil

REVIEWED BY

Peter J. van der Spek,
Erasmus University Rotterdam,
Netherlands
Sara Gangi,
University of Siena, Italy

*CORRESPONDENCE

Ivana V. Yang
✉ ivana.yang@cuanschutz.edu

[†]These authors have contributed
equally to this work and share
first authorship

[‡]These authors have contributed
equally to this work and share
senior authorship

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Single cell transcriptome signatures and cell-cell interactions associated with sarcoidosis in lung immune cell populations

Camille M. Moore^{1,2†}, Shu-Yi Liao^{3,4†}, Cheyret Wood²,
Arunangshu Sarkar², Jonathan H. Cardwell⁴, Kristyn MacPhail³,
Margaret M. Mroz³, Christina Riley³, Kara Mould^{3,4},
Clara I. Restrepo^{3,4}, Li Li^{3,4}, Lisa A. Maier^{3,4‡} and Ivana V. Yang^{4,5*‡}

¹Center for Genes, Environment and Health and Department of Immunology and Genomic Medicine,
National Jewish Health, Denver, CO, United States, ²Department of Biostatistics and Informatics,
Colorado School of Public Health, University of Colorado Anschutz Medical Campus, Aurora, CO,
United States, ³Department of Medicine, National Jewish Health, Denver, CO, United States,

⁴Department of Medicine, School of Medicine, University of Colorado Anschutz Medical Campus,
Aurora, CO, United States, ⁵Department of Biomedical Informatics, School of Medicine, University of
Colorado Anschutz Medical Campus, Aurora, CO, United States

Background: Sarcoidosis is a complex, multi-system granulomatous disease characterized by immune dysregulation and chronic inflammation, primarily affecting the lungs. To identify cell specific molecular changes associated with sarcoidosis development and progression, we aimed to characterize cellular composition, gene expression patterns, and cell-cell interactions in bronchoalveolar lavage (BAL) cells from patients with pulmonary sarcoidosis and healthy controls.

Methods: Single cell RNA-seq on 16 sarcoidosis cases (8 progressive and 8 non-progressive) and 14 controls were analyzed to identify differences in cell proportions by disease group using F-tests and differential gene expression (DE) using pseudobulk analysis. Enriched pathways and upstream regulators were identified using Ingenuity Pathway Analysis (IPA). Cell-to-cell communication and ligand-receptor interaction analyses were performed using CellChat.

Results: We identified significant DE of genes and pathways associated with sarcoidosis in resident macrophages (upregulation of *IL1R1*, *PSTPIP2*, *TAPBP*), recruited macrophages (downregulation of *AKT1*, *ACKR3*, *AZU1*), and proliferating macrophages (upregulation of *CCL4*). We also observed a limited number of DE transcripts associated with disease progression in resident and recruited macrophages. In non-macrophages cells, we observed a significant reduction in the number of B cells in sarcoidosis. Among T cell populations, we identified specific transcriptional alterations at gene and pathway levels. Most changes in upstream regulators were observed in CD4+ T cells, including activation of *TNF*, *IFNG*, and *IL1B*. We observed distinct differences in cell-to-cell interactions of macrophages and T cells between sarcoidosis patients and controls. While overall cell interactions were reduced in sarcoidosis, there was a relative increase in CD4+ T cell interactions, indicating a potential shift in immune dynamics. Key disruptions observed included downregulation of LGALS9-CD45 signaling.

Conclusions: These findings underscore the complexity of immune cell involvement in pulmonary sarcoidosis and highlight potential cellular and molecular targets for further investigation.

KEYWORDS

lung macrophages, pulmonary sarcoidosis, sarcoidosis progression, single cell molecular profiles, transcriptomics

Introduction

Sarcoidosis is a complex, multi-system granulomatous disease characterized by immune dysregulation and chronic inflammation, primarily affecting the lungs (1). Its exact etiology remains unclear; however, it is believed to result from an exaggerated immune response to environmental or infectious triggers in genetically predisposed individuals. The hallmark of sarcoidosis is the formation of non-necrotizing granulomas, primarily composed of activated macrophages and T lymphocytes (2), contributing to tissue damage and in some fibrosis. We and others have identified epigenetic (3), transcriptional (4, 5), proteomic (6), and multi-omic (7) signatures of sarcoidosis in bronchoalveolar lavage (BAL) cells. Sampling BAL cells allows us to capture molecular signatures of disease activity in immune cell populations from the target organ, the lung. However, these past studies used bulk profiling of all BAL cells and did not identify molecular signatures in specific cell populations (macrophages, T lymphocytes, etc).

Because of its heterogeneous presentation, distinguishing between progressive and non-progressive or resolving disease remains challenging clinically, emphasizing the need for better diagnostic biomarkers and therapeutic targets (8). We have used clinical features, specifically acuity of presentation, organ involvement, pulmonary function tests, chest imaging and immunosuppressive treatment at time of BAL and up to two years after BAL to define progressive and non-progressive disease for our proteomic (6), multi-omic (7), biomarker discovery (9), and single cell pilot (10) studies.

Recent advances in single-cell RNA sequencing (scRNA-seq) have provided an unprecedented level of detail in understanding the cellular landscape and transcriptional profiles associated with sarcoidosis. For example, a recent study used scRNA-seq together with spatial transcriptomics to identify an immune stimulatory environment in granulomas that repurposes transcriptional programs associated with lymphoid organ development (11). This work highlighted the role of monocytes in driving granuloma formation in skin and lung. However, the only study that has addressed disease development and progression in lung in bronchoalveolar lavage (BAL) cells was our pilot study with a limited number of samples (10). Leveraging this knowledge gap, we aimed to better characterize the cellular composition and gene expression patterns in BAL cells from patients with sarcoidosis (both progressive and non-progressive) and healthy controls. We integrated our current dataset with our previously published data (10, 12) to increase statistical power. By focusing on immune cell subsets, particularly macrophages and T cells, we sought to unravel potential molecular mechanisms driving disease progression and identify novel targets for future functional studies.

Materials and methods

Sample selection and single cell RNA-seq data collection

Sample Sources. Sarcoidosis cases and non-diseased controls were enrolled at National Jewish Health. 12 sarcoidosis and 4 control participants were enrolled for the purpose of the current study. We combined scRNA-seq data from these participants (GSE288459) with our previously collected data on 4 sarcoidosis (GSE184735) and 10 control (GSE151928) participants for a final sample size of 16 sarcoidosis cases and 14 controls. All sarcoidosis cases were phenotyped using the same criteria (7). Bronchoscopies, BAL cell processing, and scRNA-seq data collection were performed using uniform protocols for samples from our current and previous studies.

Clinical Phenotyping. For all sarcoidosis cases, medical records were reviewed for clinical features including acuity of presentation (acute/non-acute), organ involvement, pulmonary function tests (PFT), chest imaging and immunosuppressive treatment at time of BAL and up to two years after BAL (7). All sarcoidosis subjects met the ATS/ERS criteria (13) for tissue biopsy confirmation of diagnosis of sarcoidosis. The non-progressive phenotype was defined as having either acute (i.e. consistent with Löfgren's syndrome) or non-acute disease presentation, no new organ involvement, lung function testing with <10% decline in FVC or FEV₁, <15% decline in DL_{CO}, and stable chest imaging within 2 years after BAL. The progressive phenotype had a non-acute disease presentation; pulmonary function testing (PFT) with ≥10% decline in FVC or FEV₁; or ≥15% decline in DL_{CO}; worsening chest imaging; and/or if they required initiation of systemic immunosuppressive treatment any time up to 2 years after BAL. 8 of our sarcoidosis cases had progressive and 8 non-progressive disease. 8 progressive cases were defined based on the following: chest imaging, PFT, and treatment (N = 1); chest imaging and treatment (N = 4); chest imaging and PFT (N=1), chest imaging (N=1), and treatment (N=1).

Bronchoalveolar lavage (BAL). Bronchoscopy with BAL was performed as previously described. (14, 15). Briefly, four to eight 60 ml aliquots of sterile normal saline at room temperature were instilled into the airway with a syringe then aspirated using low suction. For each collection from an individual patient, the aspirated BAL specimens were pooled, kept on ice, and processed within one hour of collection. The BAL cells were obtained by centrifugation and cryopreserved.

Single-cell RNA-Seq Data Collection. Cryopreserved cells were thawed and captured on 10X Chromium in the Genomics

Core at National Jewish Health using 10x Genomics Single Cell 3' v3 chemistry, followed by sequencing on Illumina NovaSeq.

Single-cell RNA-seq computational pipeline and preliminary quality control

Initial pre-processing of the 10x Genomics scRNA-Seq data from 30 participants, including demultiplexing, alignment to the hg38 human genome, and unique molecular identifier-based (UMI-based) gene expression quantification, was performed using Cell Ranger (Mould samples: version 3.0.2, this study and Liao samples: version 3.1.0, 10x Genomics). Data were available on 171,003 cells from 30 samples. We filtered out low-quality cells with fewer than 100 genes detected or with greater than 25% of mapped reads originating from the mitochondrial genome. We additionally safeguarded against doublets by removing cells with a UMI count greater than the 98th percentile of UMI counts for each sample. Prior to downstream analysis, select mitochondrial and ribosomal genes (genes beginning with MT-, MRPL, MRPS, RPL, or RPS) were removed. The preliminary quality-controlled data set consisted of 150,176 cells and 20,491 genes. To account for differences in coverage across cells, we normalized and variance stabilized UMI counts using the SCTransform method in the Seurat version 5.1.0 R package. In addition to adjusting for sequencing depth, we also adjusted for the proportion of mitochondrial reads.

Data integration, dimensionality reduction, and clustering

Data from the 30 participants were combined using single-cell integration implemented in Seurat v3 utilizing reference samples (4 from this study, 2 from Mould, and 1 from Liao). Integration was carried out using the top 30 dimensions from a canonical correlation analysis based on SCTransform-normalized expression of the top 3,000 most informative genes, defined by gene dispersion using the Seurat's SelectIntegrationFeatures function. Integrated data were then clustered and visualized using the top 26 principal components. For visualization, we reduced variation to 2 dimensions using UMAP (n.neighbors = 50, min.dist = 0.3). Unsupervised clustering was performed by using a shared nearest neighbor graph based on 50 nearest neighbors and then by determining the number and composition of clusters using a smart local moving algorithm (resolution = 0.6). This algorithm identified 26 preliminary clusters.

Cluster markers

To identify cluster markers, we carried out pairwise differential expression analysis comparing log-normalized expression in each cluster to all others using a Wilcoxon rank sum test. Markers were identified as genes exhibiting significant upregulation when compared against all other clusters, defined by having a Bonferroni-adjusted $P < 0.05$, a log fold change > 0.25 , and $> 10\%$ of cells with detectable expression. This analysis was then performed separately for each participant using Seurat's FindConservedMarkers function to determine if marker genes were consistent across participants. Using these markers, 23 clusters were identified as either macrophage or nonmacrophage cells. The remaining three clusters were

unable to be identified and were excluded; all three had low number of cells (≤ 330), one had high number of mitochondrial reads, the other two had high amounts of ribosomal RNA. The dataset was divided into macrophage (72,935 cells) and nonmacrophage (23,806 cells) data sets for further analysis.

Additional quality control and re-clustering

Dimensionality reduction and clustering were performed separately for the macrophage and non-macrophage datasets as described above, resulting in 17 macrophage and 16 non-macrophage clusters. Potential doublets were assessed with the scds R package, which calculates a hybrid of the co-expression based and binary classification doublet scores to for each cell. Cells with high hybrid doublet scores (hybrid > 0.5) were excluded from further analysis in the macrophage dataset (this step was not necessary in non-macrophage cells). We also examined marker expression and excluded macrophage cluster 11 based on low expression of *CD68*, *FABP4*, *FCN1*, and *VCAN*. Similarly, non-macrophage clusters 10 and 14 were determined to be composed of a mixture cells, so these cells were excluded from further analysis. This resulted in 71, 616 cells in the macrophage and 22, 810 cells in the non-macrophage datasets.

Cell proportion analysis

We tested differences in cell proportions by disease group using F-tests on cell proportions calculated by the propeller R package. Propeller calculates cell type proportions for each biological replicate, performs a variance stabilizing transformation (logit) on the matrix of proportions, and fits a linear model for each cell type or cluster using the limma framework.

Pseudo-bulk differential expression

To identify differentially expressed genes (DEGs) between control and sarcoidosis samples while accounting for clustering of cells within participants, we performed pseudo-bulk differential expression analysis separately for each cell type. Within each cell type, expression counts were summed across all cells for a participant, resulting in a single expression measurement for each gene for each participant. Low expressing genes were removed as defined of having less than 1 count per million in 1/3 of the samples. Pseudo-bulk expression was compared between participants with sarcoidosis and controls using bulk RNA-Seq analysis methods with the DESeq2 R package. Genes with Benjamini-Hochberg adjusted P values less than 0.05 were considered differentially expressed. This analysis was repeated to identify genes differentially expressed between progressive sarcoidosis and non-progressive sarcoidosis.

Cell-level differential expression

Linear mixed models were used to compare cell level gene expression between sarcoidosis and control for each cell type. Genes were included in the analysis if they were expressed in at least 1% of the cells. DESeq2's variance stabilizing transformation (VST) was used to normalize the UMI count data prior to modeling. The linear

mixed model was fit with diagnosis (sarcoidosis or control) as a fixed effect and a random intercept for each subject to account for the clustering of cells within each subject.

Cell-cell interaction analysis

Cell to cell communication and pathway analyses were performed using CellChat version 2.1.2. Communication was estimated separately for the sarcoidosis samples and controls samples for the following cell types; CD4+ T-cells, CD8+ T-cells, Resident Macrophages, High MT Resident Macrophages, Recruited Macrophages, Pro-fibrotic Recruited Macrophages, and Proliferating Macrophages. The human CellChat database was used to define the ligand-receptor interaction pairs and to group these pairs into functionally related pathways. Average gene expression was computed using the 25% truncated mean. 10,000 permutations were used to estimate cell communication probability. We filtered out communications that were not present in at least 75% of the samples. To account for multiple testing of the pathways, statistically significant communication was defined as having a Bonferroni corrected p-values less than 0.05. Interaction strength and number of interactions between cell types were compared between the sarcoidosis and control groups.

Results

Enrolled participants

Our single cell RNA-Seq dataset consisted of BAL cells from 30 participants: 16 sarcoidosis (8 progressive [SarcP] and 8 non-progressive [SarcNP]) and 14 non-diseased controls. Characteristics of participants included in the study are summarized in Table 1. Sarcoidosis patients were older than controls and included some former smokers, while there were no former smokers in the control

group. 14 samples were from our previous studies, including 4 sarcoidosis cases (2 SarcP and 2 SarcNP) (10) and 10 control (12) samples collected using the same protocols for sarcoidosis phenotyping, BAL collection, and scRNA-seq data collection at NJH.

Transcriptional signatures of disease in macrophage cell populations

Following quality control and clustering described in Methods, we split our dataset into macrophage and non-macrophage cells to enable more refined clustering. The macrophage dataset comprised resident macrophages (64,143 cells; high expression of *FABP4*, *CD68*, *MRC1*), of which 2,764 were marked by high metallothionein (MT) gene expression; recruited monocyte-derived macrophages (6,474 cells; high expression of *FCN1*, *VCAN*, *CCL2*), of which 1,263 were pro-fibrotic recruited macrophages (high expression of *CHI3L1*, *SPP1*, *MMP9*, *MARCKS*, *PLA2G7*); and proliferating macrophages (999 cells; high expression of *MKI67*, *MCM2*, *PCNA*) (Figure 1A). We first examined the proportion of these five macrophage populations across the samples (Figure 1B) and determined that there were no significant differences in sarcoidosis samples compared to controls (Supplementary Table 1A). Differential gene expression analysis on pseudobulk counts identified significant differences (FDR-adjusted p-value < 0.05) associated with sarcoidosis in 16 genes in resident macrophages, 3 in high MT resident macrophages, 28 in recruited, 2 in profibrotic recruited, and 1 in proliferating macrophage populations (Supplementary Table 2). We also tested for differences in cell-level gene expression using linear mixed-effects models. While that analysis did not result in significant differences at the stringent FDR-adjusted p < 0.05, top genes were similar to those identified by pseudobulk testing (Supplementary Table 3). Representative violin plots for significant DEGs by pseudobulk analysis are shown in Figure 1C. Among transcripts associated with sarcoidosis, we identified increased gene expression in resident macrophages in interleukin 1 receptor type (*IL1R1*),

TABLE 1 Demographic and clinical characteristics of the control, progressive sarcoidosis, and non-progressive sarcoidosis subjects.

Characteristic	Controls	Non-progressive sarcoidosis	Progressive sarcoidosis	P value
N	14	8	8	
Age at sampling (median [IQR])	32 [28, 44]	50 [43, 55]	58 [46, 59]	<0.001 [#]
Birth Sex = Male (%)	8 (57%)	2 (25%)	7 (88%)	0.04*
Race/Ethnicity (N (%))				0.71*
Non-Hispanic White	9 (64%)	7 (88%)	6 (75%)	
Hispanic	3 (21%)	0 (0%)	1 (13%)	
Non-White/Non-Hispanic	2 (14%)	1 (13%)	1 (13%)	
Smoking Status: Former Smoker (%)	0 (0%)	3 (38%)	2 (25%)	0.06*
FVC % predicted (mean (SD))	NA	86 (14)	92 (14)	0.50 [^]
Z score [†] (mean (SD))	NA	-0.91 (1.16)**	-1.16 (1.26)	1.00 ^{\$}
FEV1% predicted (mean (SD))	NA	82 (17)	78 (19)	0.66 [^]
Z score [†] (mean (SD))	NA	-0.53 (1.02)**	-0.34 (1.11)	0.90 ^{\$}
DLCO % predicted (mean (SD))	NA	114 (8)	104 (21)	0.29 [^]
Z score [†] (mean (SD))	NA	1.08 (0.43)**	0.19 (1.17)	0.25 ^{\$}

*Pearson's Chi-squared test, [#]Linear Model ANOVA (3-Group), [^]Student's T-test ^{\$}Kruskal-Wallis test, **Calculated from 6 patients (2 missing pulmonary function test). [†] Z score was calculated based on the Global Lung Function Initiative race-neutral formula.

FVC, forced vital capacity; FEV1, forced expiratory volume over 1 second; DLCO, diffusing capacity of the lung for carbon monoxide.

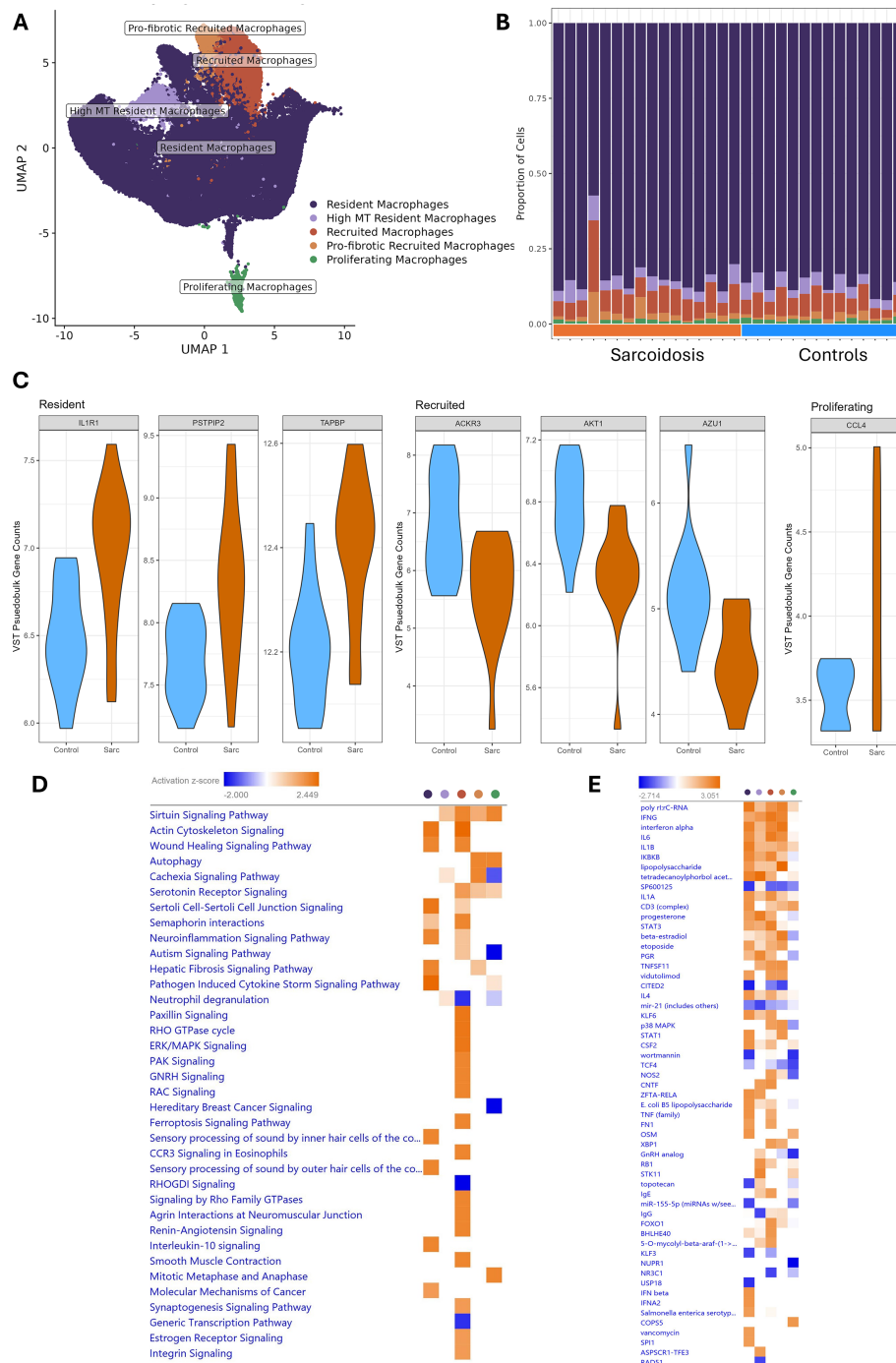


FIGURE 1

Sarcoidosis transcriptome signatures in macrophage cell populations. (A) UMAP projections of five macrophage clusters. (B) Proportion of cells in each of the five macrophage clusters across 30 individuals included in the analysis. (C) Violin plots of select genes differentially expressed in sarcoidosis (N = 16) compared to controls (N = 14) in specific macrophage populations. (D, E) Ingenuity pathway (D) and upstream regulator (E) analysis of top 100 DEGs in sarcoidosis compared to controls. Colors in Panels D and E indicate cell populations using the same color legend as the UMAP in Panel A.

receptor for IL-1 α , IL-1b β , and interleukin-1 receptor antagonist (ILRAP); a novel macrophage actin protein proline-serine-threonine phosphatase interacting protein 2 (*PSTPIP2*), and TAP binding protein (*TAPBP*) that is essential for optimal peptide loading on the MHC class I molecule.

In recruited macrophages, three downregulated transcripts, atypical chemokine receptor 3 (*ACKR3*), AKT serine/threonine

kinase 1 (*AKT1*), and azurocidin 1 (*AZU1*) were among the DEGs. Furthermore, we highlight upregulation of C-C motif chemokine ligand 4 (*CCL4*, also known as *MIP1 β*), ligand for the CCR5 receptor, in proliferating macrophages. To identify DE changes associated with sarcoidosis beyond the few highly statistically significant genes, we performed pathway and upstream regulator analysis of top 100 most significant genes (by nominal p value)

in the five macrophage clusters. Generally, we observed cell population specific pathway activation with recruited macrophages having the most pronounced changes (Figure 1D). Among pathways activated in recruited macrophages were ERK/MAPK, RHO GTPase, CCR3, ferroptosis, and integrin signaling. On the other hand, resident macrophages were marked by activation of IL10 signaling. A few pathways were also activated across macrophage populations; including sirtuin signaling in all but resident macrophages, wound healing in resident and recruited macrophages, and autophagy in profibrotic and proliferating macrophages. Upstream regulator analysis identified activation of many shared and unique transcriptional regulators (Figure 1E). All but proliferating macrophages were marked by activation of cytokines IFNG, IL6, IL1, and IL4.

To assess relevance of our findings to granuloma formation, we compared lists of sarcoidosis-associated genes in macrophages to the Krausgruber et al. gene list comparing macrophages found in cutaneous sarcoidosis granulomas to homeostatic macrophages in non-lesional skin (11). This analysis revealed that 13 genes in our analysis, including *IL1R1*, are also associated with cutaneous granulomas (Supplementary Table 4A). At the pathway level, the most striking overlap between lung and skin is in activation of IFNG signaling. We also compared our data to monocyte gene expression from the Garman et al. peripheral blood dataset (16). Four sarcoidosis-associated genes in macrophages in our studies are also associated with sarcoidosis in CD14+ monocytes, including *IRF1*, and an additional two in CD16+ monocytes (Supplementary Table 4B).

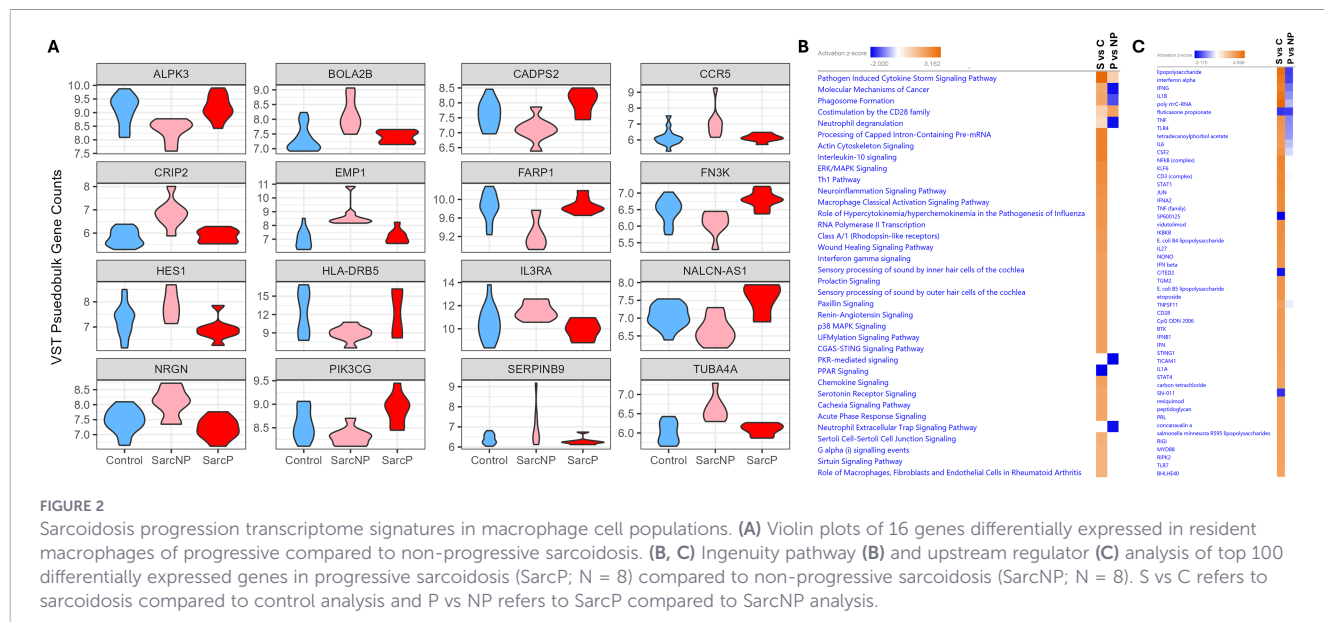
Transcriptional signatures of disease progression in macrophage cell populations

Focusing on disease progression, we did not observe any significant changes in cell proportions across macrophage populations (Supplementary Table 1B). Pseudobulk analysis identified 16 transcripts in resident macrophages significantly (FDR-adjusted $p < 0.05$) differentially expressed in progressive compared to non-

progressive sarcoidosis (Figure 2A); another 14 were significant at FDR-adjusted $p < 0.1$ (Supplementary Table 5). Surprisingly, the majority of transcript levels for the majority of significant DE genes were similar in SarcP and controls, with SarcNP cases demonstrating different levels compared to controls and Sarc P. Minimal significant changes (1–2 genes each) were identified in the remaining three macrophage populations, likely due to lower power to detect changes in these smaller cell clusters. Among transcripts upregulated in resident macrophages from participants with progressive disease was the MHC class II molecule *HLA-DRB5* (also upregulated in other macrophage populations), while downregulated transcripts include C-C motif chemokine receptor 5 (*CCR5*) and Interleukin 3 Receptor Subunit Alpha (*IL3RA*). We compared pathways enriched in top 100 genes associated with sarcoidosis and progression in resident macrophages and found very few pathways dysregulated in SarcP compared to SarcNP; this is in contrast to pronounced dysregulation, mainly activation, we observed in sarcoidosis compared to controls (Figure 2B). We observed activation of Pathogen-induced cytokine storm and Costimulation by CD28 family pathways in both disease and progression analyses while pathways such as Phagosome formation and Neutrophil deregulation demonstrated opposite effects in the two comparisons (activation in disease but deactivation in SarcP compared to SarcNP). Similarly, most upstream regulators were activated in disease but deactivated in SarcP compared to SarcNP (Figure 2C).

Transcriptional signatures of disease in non-macrophage cell populations

As expected, the majority of non-macrophage cells were comprised of T-cells (18,047 cells), of which 6,399 cells were CD4+ T-cells, 8,800 were CD8+ T-cells, and the remaining 2,848 were mixed T-cells. Other non-macrophage clusters included dendritic cells (DC; 2,597 cells), B-cells (823 cells), natural killer (NK) cells (757 cells), plasmacytoid dendritic cells (pDC; 154 cells), and epithelial cells (432) (Figure 3A). We first examined proportions of these cell populations across samples (Figure 3B) and observed a significant



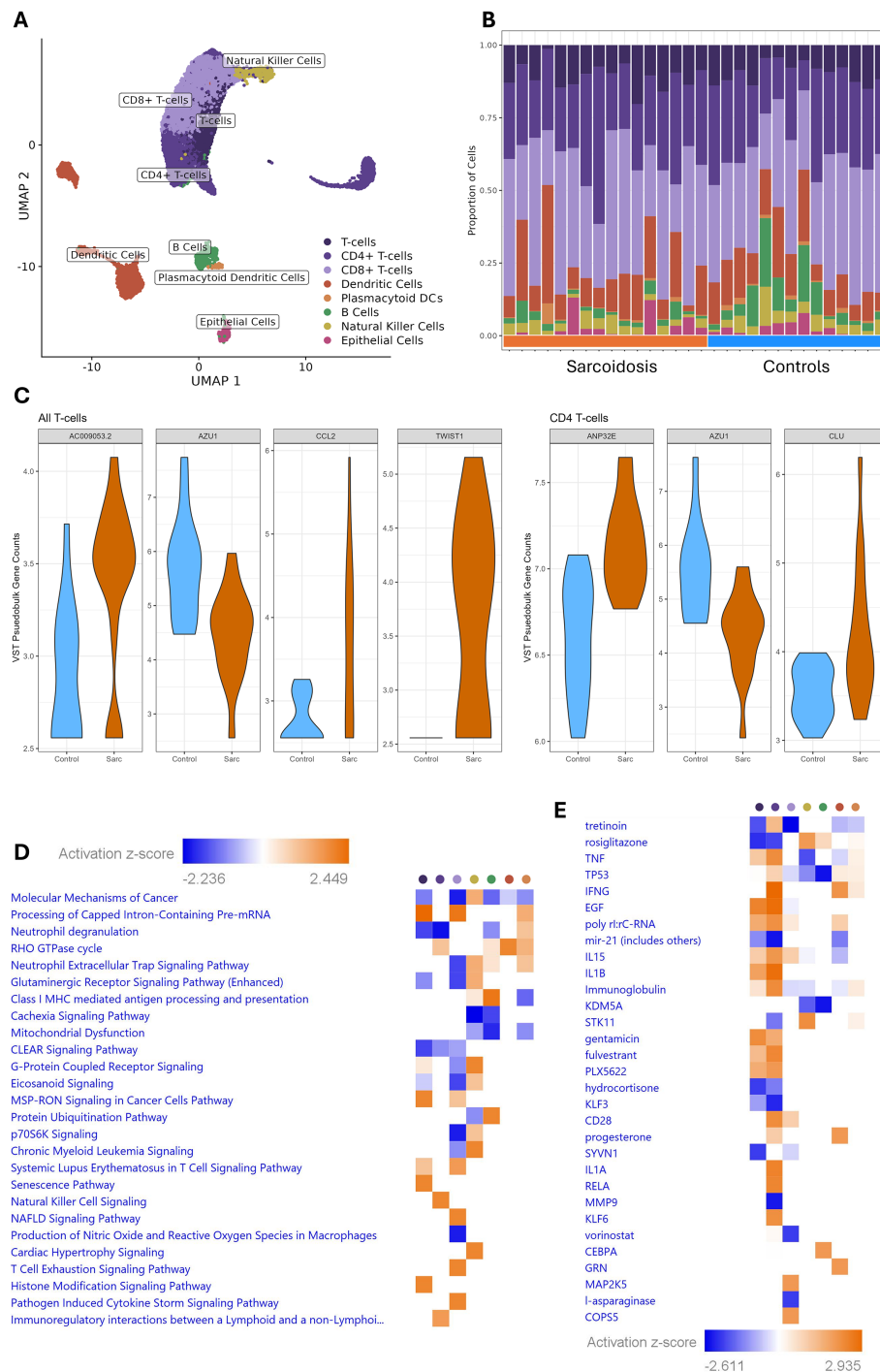


FIGURE 3

Sarcoidosis transcriptome signatures in non-macrophage cell populations. **(A)** UMAP projections of cell clusters. **(B)** Proportion of cells in each of the clusters across 30 individuals included in the analysis. **(C)** Violin plots of select genes differentially expressed in sarcoidosis ($N = 16$) compared to controls ($N = 14$) in specific cell populations. **(D, E)** Ingenuity pathway **(D)** and upstream regulator **(E)** analysis of top 100 differentially expressed genes in sarcoidosis compared to controls. Colors in Panels D and E indicate cell populations using the same color legend as the UMAP in Panel A.

decrease ($p=0.01$) in B cells in sarcoidosis (2.6% of non-macrophage cells) compared to controls (7.5% of non-macrophage cells) (Supplementary Table 6). The DEG analysis on pseudobulk counts identified significant differences (FDR-adjusted p -value <0.05) associated with sarcoidosis in five transcripts in all T cells, 8 when subset to CD4+ T cells, and one in dendritic cells (Supplementary Table 7). Representative violin plots for

significantly DEGs by pseudobulk analysis are shown in Figure 3C. Among transcripts associated with sarcoidosis, we highlight upregulation of unknown transcript AC009053.2, chemokine ligand 2 (*CCL2*) and transcription factor *TWIST1* in combined T cells, upregulation of acidic nuclear phosphoprotein 32 family member E (*ANP32E*) and clusterin (*CLU*) in CD4+ T cells, and downregulation of *AZU1* and *LYZ* in all T cells and when focused to

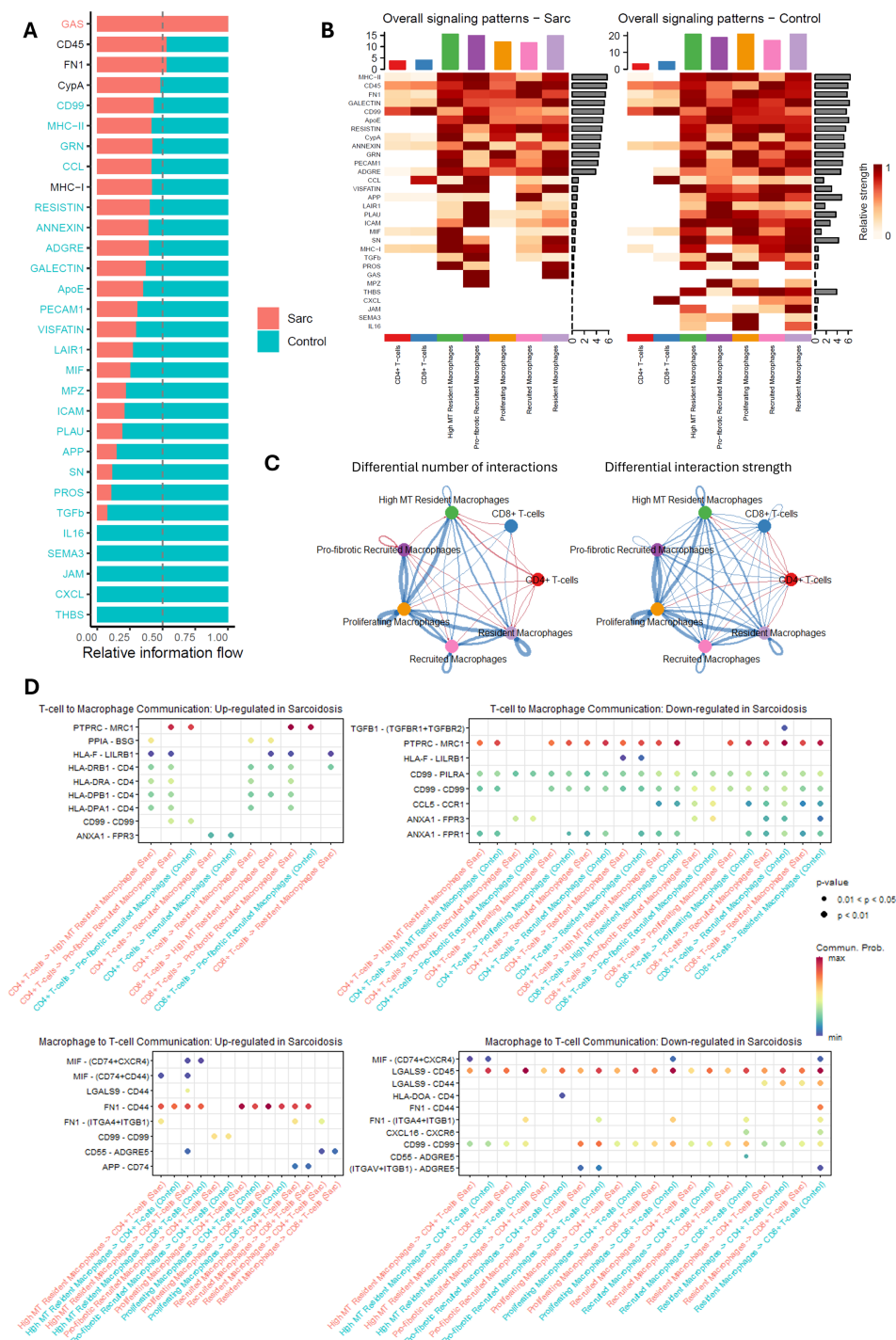


FIGURE 4

Cell-cell interaction analysis of macrophage and T cell populations. **(A)** Global analysis of ligand-receptor pathways between sarcoidosis and controls depicted as the colored bar plot comparing information flow for each pathway. Pathway labels are colored according to significant difference in signaling information, with enrichment in sarcoidosis (salmon) or controls (light blue). Black labels indicate no significant difference in signaling information between sarcoidosis and control. **(B)** Heatmaps for signaling in ligand-receptor pathway at the cell type level sarcoidosis (left) and controls (right). **(C)** Circle plots of differential number of interactions of interaction strength among cell populations. Blue lines indicate enrichment in interactions in controls and red in sarcoidosis. Thickness of the line indicates the strength of the differential enrichment between sarcoidosis and controls. **(D)** Comparison of specific ligand-receptor interactions in macrophage-T cells populations that are down- (left) and up-regulated (right) in sarcoidosis.

CD4+ T cells. Pathway analysis of top 100 transcripts (Figure 3D) highlighted a few shared and mostly unique pathways across the cell populations we examined; activation of T cell exhaustion pathway in CD8+ T cells is noteworthy. Most changes in upstream regulators

were observed in CD4+ T cells, including but not limited to activation of TNF, IFNG, and IL1B; other cell populations demonstrated less prominent transcriptional changes (Figure 3E). Given the small sample sizes, we did not expect to identify many changes

associated with disease progression. Analysis of disease progression in non-macrophage cell clusters identified only a few significant differences at $FDR\text{-}adj < 0.05$, likely because of a small number of cells and samples.

Cell-cell interaction analysis of macrophage and T cell populations

In the final set of investigations, we performed ligand-receptor analysis on sarcoidosis and control single cell datasets to study the crosstalk between macrophage and T cell populations; we excluded other cell types due to small numbers. Globally, we observed reduced signaling in sarcoidosis compared to controls (Figure 4A), with only one pathway (GAS signaling) exhibiting stronger communication in sarcoidosis and 25 pathways with stronger communication in controls; these include immune pathways such as CCL, CXCL, and MHC class II, as well as pro-fibrotic pathways such as ICAM, PECAM, and TGF β . We observed similar patterns of reduced ligand-receptor signaling in sarcoidosis compared to controls at the cell type level (Figure 4B). We also observed a larger number of changes in receiving or sending signals in macrophage than T cell populations (Supplementary Figure 1A). Considering interactions across macrophage and T cell populations (Figure 4C), we observed a higher number of interactions and interaction strength among most cell populations in controls (blue lines), however, the number of interactions and interaction strength of CD4 $^{+}$ T cells with other cell types were higher in sarcoidosis (red lines). We then focused on cell-specific ligand-receptor analysis to identify dysfunctional signaling in sarcoidosis compared to controls. Comparing the communication probabilities, we observed many interactions decreased and downregulated in sarcoidosis, and a few increased and upregulated in sarcoidosis (Supplementary Figure 2). Focusing on only interactions between macrophage and T cell populations, we observed differences in interactions of all macrophage populations with CD4 $^{+}$ and CD8 $^{+}$ T cells (Figure 4D, Supplementary Figure 1B). This includes upregulation of FN1-CD44 signaling from high MT resident and proliferating macrophages to CD4 $^{+}$ and CD8 $^{+}$ T cells as well as downregulation of LGALS9-CD45 signaling from all macrophage populations to CD4 $^{+}$ and CD8 $^{+}$ T cells. When looking at T cell to macrophage signaling, more specific downregulation of HLA-DRB1-CD4 and HLA-F-LILRB1 signaling was observed from CD4 $^{+}$ and CD8 $^{+}$ T cells to high MT resident and profibrotic recruited macrophages. On the other hand, we observed upregulation of HLA-DRA-CD4, HLA-DPA1-CD4, and HLA-DPB1-CD4 signaling from CD4 $^{+}$ and CD8 $^{+}$ T cells to high MT resident and profibrotic recruited macrophages. These results, while initially surprising in directionality, are consistent with expression patterns of these genes in macrophage and T cell populations we examined (Supplementary Figure 3).

Discussion

Our single-cell RNA-Seq analysis of BAL cells from sarcoidosis patients and healthy controls is the largest to date, revealing distinct macrophage subpopulations and gene expression patterns associated

with sarcoidosis and disease progression. Specifically, we identified five macrophage populations: resident, high metallothionein (MT) resident, recruited, profibrotic recruited, and proliferating macrophages. Each subpopulation displayed unique gene expression profiles, with notable DEGs and pathways linked to sarcoidosis in resident macrophages, recruited macrophages, and proliferating macrophages. We also observed changes in gene expression associated with disease progression in resident and recruited macrophages, albeit in limited number of genes. In non-macrophages cells, we observed a significant reduction in the number of B cells in sarcoidosis patients compared to controls. Among T cell populations, we identified specific transcriptional alterations at gene and pathway levels. Additionally, we observed distinct differences in cell-to-cell interactions of macrophages and T cells between sarcoidosis patients and healthy controls. These findings support the immune dysfunction associated with sarcoidosis and the complexity of immune cell involvement in sarcoidosis, while highlighting potential cellular and molecular targets for further investigation. Our limited findings related to disease progression highlight the need for larger sample sizes and potentially alternative phenotyping approaches.

In our study, the overall cell proportions in BAL were similar between sarcoidosis patients and healthy controls, with the exception of the reduced number of B cells. A prior study (17) reported increased proportions of macrophages, NK/T cells, and endothelial cells in cardiac sarcoidosis, and elevated HLA-DR $^{+}$ macrophages and epithelial cells in pulmonary sarcoidosis, with no significant changes in other cell types. While our analysis did not specifically focus on HLA-DR $^{+}$ macrophages and we did not capture epithelial cells in the BAL, our findings are consistent with theirs in that the proportions of broader cell populations - such as total macrophages and T cells - did not differ significantly between sarcoidosis and control groups. As in our prior work (12), we identified a population of resident macrophages marked by high expression of metallothionein (MT) genes that encode a family of proteins crucial for metal homeostasis and defense against heavy metals, oxidative stress, and DNA damage (18). This population has similarities to splenic metallophilic marginal macrophages that internalize blood-borne pathogens (19), cross-prime CD8 $^{+}$ T cell-mediated protective immunity against blood-borne tumors (20) and contribute to wound healing (21).

In our study, sarcoidosis resident macrophages exhibit upregulation of several genes, including *IL1R1*, *TAPBP*, and *PSTPIP2*. In pulmonary sarcoidosis, IL-1 signaling recruits more immune cells to granulomas, driving chronic inflammation and disease persistence. (22). IL1R serves as a receptor for IL-1 β and IL-1 receptor antagonist (IRAP), both upregulated in sarcoidosis BAL cells, with increased IL-1 β secretion. (23, 24). IL-1 β expression increases in sarcoidosis granulomas and is critical for granuloma formation in the trehalose 6, 6'-dimycolate mouse model. (23) and IRAP has also been localized to the sarcoidosis granuloma in lung tissue (24). *TAPBP* is involved in the antigen processing and presentation pathway; while genetic variation in *TAP* genes has not been associated with sarcoidosis (25, 26), antigen processing and presentation pathways have been associated with sarcoidosis (6) and beryllium-induced lung disease (27–29). Importantly, a recent meta-analysis of 21 gene expression datasets across multiple tissues identified upregulation of *TAP1* in sarcoidosis (30), providing additional support for our findings.

PSTPIP2 has not appeared in prior sarcoidosis literature but is a significant novel target. *PSTPIP2* negatively regulates macrophage activation, neutrophil migration, cytokine production, and osteoclast differentiation, linking it to innate immune and autoinflammatory diseases. (31).

Sarcoidosis recruited macrophages display downregulation of genes such as *AKT1*, *ACKR3*, and *AZU1*. PI3K/AKT1 pathway plays a key role in sarcoidosis, influencing T cell function, with its deregulation linked to disease progression (6, 7). The PI3K/AKT1 pathway has also been implicated in activation of mTOR (6), which has been associated with granuloma formation (32). Previous studies demonstrated reduced T cell proliferation and exhaustion in progressive sarcoidosis, driven partly by PI3K/AKT1 signaling inhibition (33). Unlike *AKT1*, *ACKR3* and *AZU1* have not been previously linked to sarcoidosis, although they are likely important in regulation of inflammation. Specifically, *ACKR3* (CXCR7) regulates chemokines, including CXCL12 and CXCL11, which are essential for immune cell recruitment (34, 35); CXCL11 is upregulated in serum of patients with sarcoidosis (36) and CXCL12/CXCR4 is increased in sarcoidosis granulomas (37). *ACKR3* enhances inflammation by promoting pro-inflammatory leukocyte phenotypes and angiogenesis while acting as a chemokine scavenger to maintain immune balance (38). *ACKR3*'s ability to modulate chemokine gradients suggests that it might influence granuloma architecture (39). Our finding of downregulation of *AZU1*, which encodes an antimicrobial protein, may indicate an impaired immune response to microbial or foreign antigens in sarcoidosis; however, its specific role in sarcoidosis, especially regarding immune responses to proposed triggers, remains unclear.

Proliferating macrophages in sarcoidosis show upregulation of *CCL4*. *CCL4* is a chemokine that binds to CCR5 and recruits additional immune cells to the site of inflammation. *CCL4* was previously shown as upregulated in the BAL of sarcoidosis patients (40) and a mouse model of pulmonary granulomatosis (41). Moreover, *CCL4* has been shown to function as a chemokine neoantigen that induces beryllium-specific CD4⁺ T cells in CBD (42). Its receptor CCR5 is also upregulated in sarcoidosis (43), and genetic variants in *CCR5* (44) have been associated with sarcoidosis susceptibility.

Besides its link to disease susceptibility, genetic variation in CCR5 is also associated with persistent lung involvement in sarcoidosis (45). We now identify expression of this gene as associated with disease progression, specifically in resident macrophages. Similarly, *HLA-DRB5*, whose expression is increased in progressive compared to non-progressive sarcoidosis in all macrophage populations, is also a genetic locus for sarcoidosis (46). Interestingly, we show that level of expression of many genes in progressive sarcoidosis resembles control transcripts (also observed at the pathway and upstream regulator level), suggesting that dysregulation is important in earlier stages of disease onset; this finding aligns with immune exhaustion observed in sarcoidosis. Because of this, some genes such as CCR5 are upregulated in sarcoidosis compared to controls but downregulated in progressive vs non-progressive comparison. Another plausible explanation for this observation is genetic variants that allow for disease progression, as may be the case for CCR5. These results need further

investigation but underscore the importance of analysis stratified by disease phenotypes.

In the non-macrophage population, we identified increased *CCL2* and *TWIST1* in all T cells, which is consistent with previous findings in sarcoidosis. For example, increased levels of the secreted monocyte chemoattractant CCL2/MCP1 protein have been observed in sarcoidosis BAL (47, 48) and transcription factor TWIST1 (49) in sarcoidosis BAL cells. We also identified activation of T cell exhaustion pathway in CD8⁺ T cells. In a previous study, exhausted CD8⁺ T cells were found in sarcoidosis and restoration of CD8⁺ T cell function led to disease resolution (50). We did not identify significant differences in other non-macrophage cells due to small cell numbers, demonstrating the need for larger sample sizes or cell enrichment in future studies.

In the cell-cell interaction analysis, while overall cell interactions were reduced in sarcoidosis, there was a relative increase in CD4⁺ T cell interactions, indicating a shift in immune dynamics. Key disruptions included downregulation of LGALS9-CD45 signaling from macrophages to T cells, which could impair the stability and function of adaptive regulatory T cells (51). Somewhat surprisingly, we observed an increase in signaling from HLA molecules on CD4⁺ and CD8⁺ T cells to CD4⁺ on highMT resident and profibrotic recruited macrophages, opposite to our expectation. However, it has been shown that HLA molecules are expressed on activated T cells, including patients with active tuberculosis (40, 41), and we now demonstrate their expression on T cells in sarcoidosis. Similarly, CD4⁺ expression has been documented on human, but not mouse, monocytes (52) and macrophages (53). These intriguing findings are preliminary in nature and need further investigation, including experimental validation.

While we are limited in our ability to directly compare our findings to those from single-cell studies in other organs, there are notable consistencies. A spatial transcriptomic study of cutaneous sarcoidosis (11) identified granuloma-associated (GA) macrophages in lesional skin that highly expressed IFNGR1 and showed upregulation of genes associated with IFN- γ -activated macrophages. This aligns with our pathway analysis across different macrophage subtypes, suggesting that IFN- γ -mediated activation may represent a shared immunological feature across sarcoidosis-affected organs, including both skin and lung (BAL). Additionally, a separate study investigating the effects of tofacitinib (54) demonstrated clinical improvement in both cutaneous and pulmonary sarcoidosis. Their single-cell RNA-seq analysis identified CD4⁺ T cell-derived IFN- γ as a key cytokine driving macrophage activation, further supporting the central role of IFN- γ signaling in sarcoidosis pathogenesis. Our analysis also identified other sarcoidosis-associated gene expression changes that are shared in skin macrophages and peripheral blood monocytes.

Conclusions

In summary, we identified immune cell-specific changes in gene expression and pathways related to disease and its progression. Our analysis revealed potential disruptions in communication between antigen-presenting macrophages and T cells, along with unexpected signaling from T cells to macrophages. Future research

should validate these preliminary findings and explore how these genes affect immune dysregulation in sarcoidosis, possibly leading to more effective treatments. This study is the first to evaluate BAL data by phenotypes to this degree but has limitations, including a small sample size in certain cell subclusters, which may reduce the power to detect significant changes and limit generalizability. Interpreting results from pseudobulk analyses can be challenging, as these may mask cell-specific nuances. Because of small group sizes, we did not adjust for covariates such as age, sex, smoking history, and treatment that may be confounders in our analysis. Nonetheless, these methods are more rigorous than earlier studies that likely overestimated gene expression changes in single cell datasets, including our pilot study (10). As new analytical tools become available, additional analyses of this dataset should be considered. Importantly, our findings in specific cell populations as well as cell-cell interaction analyses will require experimental validation. However, this study provides important targets for future functional studies in cell culture models.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found below: Gene Expression Omnibus (GEO) accession GSE288459.

Ethics statement

The studies involving humans were approved by National Jewish Health BRANY Institutional Review Board. The studies were conducted in accordance with the local legislation and institutional requirements. The participants provided their written informed consent to participate in this study.

Author contributions

CM: Formal Analysis, Supervision, Methodology, Writing – review & editing, Writing – original draft, Software, Visualization. SL: Methodology, Validation, Data curation, Writing – original draft, Investigation, Writing – review & editing. CW: Writing – review & editing, Software, Formal Analysis, Methodology, Visualization. AS: Writing – review & editing, Software, Formal Analysis, Methodology. JC: Formal Analysis, Software, Writing – review & editing, Methodology. KMa: Investigation, Writing – review & editing, Methodology. MM: Data curation, Writing – review & editing, Investigation. CR: Writing – review & editing, Data curation, Investigation. KMo: Resources, Data curation, Writing – review & editing. CR: Writing – review & editing, Investigation, Data curation. LL: Methodology, Investigation, Writing – review & editing, Validation, Conceptualization. LM: Methodology, Writing – original draft, Project administration, Writing – review & editing,

Investigation, Conceptualization, Supervision, Funding acquisition. IY: Writing – review & editing, Supervision, Funding acquisition, Project administration, Writing – original draft, Visualization, Conceptualization, Data curation, Investigation.

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Conflict of interest

CM, SL, KMa, CR, LL, LM, and IY report NIH grants to their institutions. LL, LM, and IY report funding from Anne Theodore Foundation, and LM reports additional funding from The Foundation for Sarcoidosis, Mallinckrodt Pharmaceuticals, and Boehringer Ingelheim. CR reports consulting fees from Xentria, ATyr Pharma, and Kinevant. LM serves on the Scientific Advisory Board for for FSR. IY reports consulting fees from Eleven P15.

The remaining author(s) declared that that this work was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

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SUPPLEMENTARY FIGURE 1

(A) Scatterplots of cell populations with significant changes in sending or receiving signals. (B) Scatterplots of differential incoming and outgoing interaction strength for sarcoidosis compared to controls in five cell population (high MT resident macrophages, profibrotic recruited macrophages, proliferating macrophages, CD4+ and CD8+ T cells).

SUPPLEMENTARY FIGURE 2

Comparison of specific ligand-receptor interactions in specific cell types that are (A) increased or decreased in sarcoidosis, and (B) down- and up-regulated in sarcoidosis.

SUPPLEMENTARY FIGURE 3

Violin plots of select transcripts identified by CellChat interaction analysis between macrophage and T cell populations.

SUPPLEMENTARY TABLE 1

Tests for differences in cell proportions across macrophage clusters in (A) sarcoidosis compared to controls and (B) progressive sarcoidosis, non-progressive sarcoidosis, and controls.

SUPPLEMENTARY TABLE 2

Differentially expressed transcripts in sarcoidosis (N = 16) compared to controls (N = 14) across macrophage clusters, identified by pseudobulk

analysis. (A) Resident macrophages, (B) high MT resident macrophages, (C) recruited macrophages, (D) pro-fibrotic recruited macrophages, and (E) proliferating macrophages.

SUPPLEMENTARY TABLE 3

Differentially expressed transcripts in sarcoidosis (N = 16) compared to controls (N = 14) across macrophage clusters, identified by mixed-effects linear modeling of individual cells. (A) Resident macrophages, (B) high MT resident macrophages, (C) recruited macrophages, and (D) pro-fibrotic recruited macrophages.

SUPPLEMENTARY TABLE 4

Overlap of genes associated with sarcoidosis in macrophages with published datasets.

SUPPLEMENTARY TABLE 5

Differentially expressed transcripts in progressive sarcoidosis (N = 8) compared to non-progressive disease (N = 8) in (A) resident macrophages and (B) recruited macrophages, identified by pseudobulk analysis.

SUPPLEMENTARY TABLE 6

Tests for differences in cell proportions across non-macrophage clusters in (A) sarcoidosis compared to controls and (B) progressive sarcoidosis, non-progressive sarcoidosis, and controls.

SUPPLEMENTARY TABLE 7

Differentially expressed transcripts in sarcoidosis (N = 16) compared to controls (N = 14) across non-macrophage clusters, identified by pseudobulk analysis. (A) CD4+ T cells, (B) CD8+ T cells, (C) mixed T cell cluster, (D) B cells, (E) dendritic cells (DCs), and (F) epithelial cells.

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