#### Check for updates

#### **OPEN ACCESS**

EDITED BY Veronica Anania, Genentech Inc., United States

#### REVIEWED BY

Alexander Mann, Harvard Medical School, United States Jennifer L. Larson-Casey, University of Alabama at Birmingham, United States

\*CORRESPONDENCE Alessandro Venosa Zalessandro.venosa@pharm.utah.edu

RECEIVED 29 April 2024 ACCEPTED 28 June 2024 PUBLISHED 19 July 2024

#### CITATION

Moos PJ, Cheminant JR, Cowman S, Noll J, Wang Q, Musci T and Venosa A (2024) Spatial and phenotypic heterogeneity of resident and monocyte-derived macrophages during inflammatory exacerbations leading to pulmonary fibrosis. *Front. Immunol.* 15:1425466. doi: 10.3389/fimmu.2024.1425466

#### COPYRIGHT

© 2024 Moos, Cheminant, Cowman, Noll, Wang, Musci and Venosa. This is an openaccess article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

# Spatial and phenotypic heterogeneity of resident and monocyte-derived macrophages during inflammatory exacerbations leading to pulmonary fibrosis

Philip J. Moos, Jenna R. Cheminant, Sophie Cowman, Jessica Noll, Qiuming Wang, Teresa Musci and Alessandro Venosa\*

Department of Pharmacology and Toxicology, University of Utah College of Pharmacy, Salt Lake City, UT, United States

**Introduction:** Genetic mutations in critical nodes of pulmonary epithelial function are linked to the pathogenesis of pulmonary fibrosis (PF) and other interstitial lung diseases. The slow progression of these pathologies is often intermitted and accelerated by acute exacerbations, complex non-resolving cycles of inflammation and parenchymal damage, resulting in lung function decline and death. Excess monocyte mobilization during the initial phase of an acute exacerbation, and their long-term persistence in the lung, is linked to poor disease outcome.

**Methods:** The present work leverages a clinical idiopathic PF dataset and a murine model of acute inflammatory exacerbations triggered by mutation in the alveolar type-2 cell-restricted Surfactant Protein-C [SP-C] gene to spatially and phenotypically define monocyte/macrophage changes in the fibrosing lung.

**Results:** SP-C mutation triggered heterogeneous CD68<sup>+</sup> macrophage activation, with highly active peri-injured cells relative to those sampled from fully remodeled and healthy regions. Ingenuity pathway analysis of sorted CD11b<sup>-</sup>SigF<sup>+</sup>CD11c<sup>+</sup> alveolar macrophages defined asynchronous activation of extracellular matrix reorganization, cellular mobilization, and Apolipoprotein E (*Apoe*) signaling in the fibrosing lung. Cell-cell communication analysis of single cell sequencing datasets predicted pro-fibrogenic signaling (*fibronectin/Fn1, osteopontin/Spp1*, and *Tgfb1*) emanating from *Trem2/TREM2*<sup>+</sup> interstitial macrophages. These cells also produced a distinct lipid signature from alveolar macrophages and monocytes, characterized by *Apoe* expression. Mono- and di-allelic genetic deletion of ApoE in SP-C mutant mice had limited impact on inflammation and mortality up to 42 day after injury.

**Abbreviations:** AT2, alveolar type-2 cell; Sftpc<sup>I73T</sup> surfactant protein-C I73T mutant; IPF, idiopathic pulmonary fibrosis; AIE, acute inflammatory exacerbations; AMs, alveolar macrophages; MoDMs, monocyte-derived macrophages; ApoE, apolipoprotein-E; FN1, fibronectin1; SPP1, osteopontin.

**Discussion:** Together, these results provide a detailed spatio-temporal picture of resident, interstitial, and monocyte-derived macrophages during SP-C induced inflammatory exacerbations and end-stage clinical PF, and propose ApoE as a biomarker to identify activated macrophages involved in tissue remodeling.

#### KEYWORDS

alveolar type-2 cell, surfactant protein-C I73T mutant, pulmonary fibrosis, alveolar macrophages, monocyte-derived macrophages, apolipoprotein-E, fibronectin1, osteopontin1

#### Introduction

Acute inflammatory exacerbations represent a key feature in the evolution of interstitial lung diseases (ILD). This cluster of chronic progressive pathologies includes idiopathic pulmonary fibrosis, non-specific interstitial pneumonia, connective tissue disease associated ILD, chronic hypersensitivity pneumonitis, pneumoconiosis, sarcoidosis, and more (1, 2). The exact sequence of events igniting these flare-ups have yet to be fully understood. However, epidemiological and experimental evidence suggest that a combination of genetic susceptibility, preexistent pulmonary and systemic conditions, biological aging/senescence of the alveolar compartment, and external stressors (environmental toxic exposure) contribute to the repeated cycles of focal inflammation, spatially heterogenous injury, and aberrant repair characteristic of fibrotic disease (3–7).

To date, over 60 mutations in the alveolar epithelial type 2 specific gene encoding for the surfactant protein C (SP-C) have been linked to the development of a fibrotic phenotype, with the isoleucine-to-threonine missense substitution at position 73 of the SP-C proprotein (SP-C<sup>I73T</sup>) representing the most common (8–10). Our group has previously characterized endoplasmic reticulum toxicity and macroautophagy block resulting from aberrant processing and trafficking of the SP-C proprotein and described the sequelae of immunological events accompanying tissue remodeling (8, 9, 11, 12). Here, we expand on this prior investigation by defining the spatial and phenotypic distribution of monocytes and macrophages responding to the initial epithelial injury and establish patterns of intercellular communication among cellular species in the lung.

Mounting evidence supports the notion that inflammatory monocyte mobilization in the fibrosing lung represents a valid indicator of poor disease prognosis (13, 14). Yet, experimental modeling and clinical trials designed to non-specifically target inflammation (e.g., corticosteroids, broad spectrum cytokine modulation) have revealed low efficacy or even harmful toxicities (15). Among the reasons for this therapeutic failure is the absence of a nuance approach capable of controlling the maturation, activation, and persistence of ontologically and phenotypically heterogeneous cellular entities, and the relatively fragmented understanding of the spatial distribution of cells and signals in a temporally extended pathology such as fibrosis (16-18). Experimental evidence highlights distinct transcriptomes emanating from tissue-resident and monocyte-derived macrophages (19, 20), with the latter generating a complex fibrogenic signature (10, 21, 22). The fibronectin/FN1, osteopontin/SPP1, tumor growth factor (TGF) \$1, and interleukin 4/ 13 signaling pathways represent the most studied networks mediating tissue remodeling (23-27), while metabolic networks have gained traction as potential targets in chronic inflammatory diseases (28, 29). In vitro systems establish a reliance on glycolysis in response to canonical pro-inflammatory signals (IFNy, LPS), juxtaposed to fatty acid oxidation, tricarboxylic acid cycle, and mitochondrial oxidative phosphorylation following challenge with anti-inflammatory and pro-remodeling signals (IL-10, IL-4/13, TGF $\beta$ 1) (30-32). This evidence emphasizes the importance of factors governing lipid synthesis, handling, and metabolism (PPARy, LXR, FXR, SREBP1) in regulating macrophage function (33, 34). The cholesterol and phospholipid transporter apolipoprotein E (ApoE) has been linked to monocyte-derived macrophage activation in chemical-induced injury, though there is limited evidence that these responses are consistent across the spectrum of fibrosis (35, 36).

Through a combination of bulk, single cell, and spatial transcriptomics we show a hyperactive niche of macrophages surrounding fully remodeled lung regions during fibrogenic exacerbations triggered by mutant SP-C induction. Cellular annotation and communication analysis identify time-related changes in intercellular networks, with alveolar and interstitial/ Trem2<sup>+</sup> macrophages and inflammatory monocytes responsible for Spp1, Fn1, and Tgf\u00c61 pro-fibrotic signaling. Analysis of human idiopathic PF confirms the presence of an interstitial population responsible for pro-fibrotic signaling in the diseased lung. Our findings also highlight disease-related shifts in lipid transcriptional signatures and interstitial/Trem2<sup>+</sup> macrophages as the sole cellular cluster expressing ApoE. While genetic ablation of ApoE in SP-C<sup>I73T</sup> induced injury did not significantly impact fibrotic disease outcome, our results pinpoint this molecule as a potential biomarker identifying fibrogenic myeloid populations. Together, these results reveal temporal and phenotypic heterogeneity in the macrophage compartment and implicates Trem2<sup>+</sup> interstitial macrophages and their monocytic precursors as viable targets for anti-fibrogenic therapy.

### Materials and methods

#### Sex as a biological variable

These studies utilized both male and female mice.

# Murine model of SP-C<sup>173T</sup> induced lung injury

Tamoxifen inducible SP-C<sup>I73T</sup> mice were generated as previously reported (9). For studies investigating the role of the apoliproprotein E in SP-C<sup>I73T</sup> induced injury, a parallel line was crossed with ApoE knock out mice purchased from Jackson laboratories (Strain #002052, The Jackson Laboratory, Bar Harbor, ME). Briefly, an estrogen receptor (ER)-2 controlled Flp-O recombinase strain knocked into the Rosa26 locus (The Jackson Laboratory). Adult homozygote SP-C<sup>I73T</sup>Flp mice (8-12 weeks) received two tamoxifen oral gavages three days apart (90 mg/kg each) to excise a neomycin cassette placed within the Sftpc gene. Both male and female animals were used for the studies. For studies involving apoE mutants, a SP-C<sup>I73T</sup>ApoE<sup>WT/KO</sup> breeding pair was utilized (SP-C<sup>I73T</sup>ApoE<sup>HET</sup>) so as to generate wild type, heterozygous, and homozygous experimental littermates. Control groups mice are represented as pooled data from tamoxifen treated SP-C<sup>I73T</sup> not expressing Flp-O recombinase or oil (vehicle) treated Flp-O expressing SP-C<sup>I73T</sup> mice. All mice were housed under pathogen free conditions in AALAC approved barrier facilities at the Skaggs College of Pharmacy, University of Utah. All experiments were approved by the Institutional Animal Care and Use Committee, University of Utah.

#### Reagents

Tamoxifen (non-pharmaceutical grade) was purchased from Sigma-Aldrich (St Louis, MO). Giemsa cytological stain was purchased from Sigma-Aldrich. Antibody list: Spp1 (RNAscope® Probe Green, Ref#435191); Tgfb1 (RNAscope® Probe Red, Ref# 407751-C2); Apoe (for in situ hybridization - Advanced Cell Diagnostics, RNAscope<sup>®</sup> LS 2.5 Probe #313278; for immunohistochemical - Abcam; Cat #ab183597, 1:500; for western blot - Cell Signaling Technology, Cat #49285, 1:1000), CD68 for immunohistochemistry (Cell Signaling Technology; Cat # 97778; 1:1000). Flow cytometric panel for cell sorting and bulk sequencing of macrophages CD16/32 (clone 93; eBiosciences, San Diego, CA), CD11b (clone # M1/70; eFluo450, eBiosciences); Fixable Viability dye (Cat # 65-0865-14; eFluo780, eBiosciences); SigF (clone S17007L; PE-CF594, BD Biosciences, San Jose, CA); CD45 (clone 30-F11; PerCP5.5, Biolegend, San Diego, CA); CD11c (clone # N418; BV705, Biolegend); Ly6G (clone # 1A8; AF700, Biolegend); CD64 (clone X54-5/7.1; PE/Cy7, Biolegend); CD3 (clone # 17A2; BUV395, Biolegend). All other reagents were purchased from Thermo Fisher Scientific, Inc. (Waltham, MA) or Sigma-Aldrich.

# Bronchoalveolar lavage, cell counts, ELISA, and western blot analysis

Following terminal anesthesia, inert mice were subject to cannulation. Bronchoalveolar lavage (BAL) fluid was collected from mice using 1 mL sterile saline lavage and collected into a microcentrifuge tube. Four additional lavages were performed and collected in a separate container. The two fractions were then spun at  $400 \times g$ , 6 min. Supernatant from the first lavage was collected and immediately frozen at -80°C for ELISA and western blot analysis, while the two cellular pellets were combined and suspended in 1 ml of saline solution for cell counts, flow cytometric, or RNA sequencing analysis. BAL cells were enumerated using a NucleoCounter (New Brunswick Scientific, Edison, NJ). Aliquots of first lavage were analyzed for IL-4 and IL-13 levels using the Luminex platform (Panel MCYTOMAG-70K-17) following Thermo Fisher's protocol. For western blot, equal volumes (15 µl) of thawed BAL fluid were loaded onto 4-12% NuPage Bis-Tris gels (ThermoFisher Scientific) with NuPage 4X LDS sample buffer (ThermoFisher Scientific) and then electrophoresed approximately 90 minutes using a constant voltage of 100V. Proteins from the gels were then transferred to a 0.45  $\mu m$  PVDF membrane at 30V and 4°C for one hour and blocked in 5% non-fat dried milk (NFDM). The membranes were subsequently probed with primary ApoE antibodies. The SuperSignal West Dura Chemiluminescent Substrate detection system was applied before exposing the membrane on the ProteinSimple FluorChem M imager (BioTechne).

# Histology, histochemical and *In situ* hybridization analysis

For histological and histochemical analysis, lungs from unresponsive anesthetized mice were cleared of excess blood through cardiac perfusion with 0.9% sodium chloride solution. A 20-gauge cannula was inserted in the trachea for tissue fixation with 10% neutral buffer formalin at constant pressure (25 cm H<sub>2</sub>O). A suture was used to seal the tracheal opening upon cannula removal, thus avoiding pulmonary deflation during the fixation process. Tissue was placed in a histology cassette and submerged in 10% neutral buffer formalin for 72 h. The suture was then removed, and the lung sequentially moved to a 2% sucrose solution (in PBS, two washes of 5 min) and 70% ethyl alcohol. The submerged cassettes were submitted to the University of Utah histology core (Associated Regional and University Pathologists Inc.) for embedding. Paraffin blocks were sectioned at 6 µm thickness and used for Hematoxylin & Eosin (H&E) or immunohistochemical staining, alone or in combination with in situ hybridization as previously described (37). For protein staining, paraffin was removed using xylene solutions followed by gradient alcohol washes (100-50%). Citrate antigen retrieval (10.2 mM sodium citrate, pH 6.0, for 20 min) and endogenous peroxidase quenching with 3% hydrogen peroxide in methanol (30 min) were performed. Serum-based blocking (10% goat serum in PBS) preceded the overnight incubation with anti-rabbit primary antibody. In all studies, a serum/IgG control was used. During the second day, slides

underwent incubation with a biotinylated secondary antibody (30 minutes, Vectastain Elite ABC kit, Vector Labs, Burlingame, CA) and chromogenic reaction achieved using a Peroxidase Substrate Kit DAB (Vector Labs). Counterstain was accomplished with Harris Modified Hematoxylin (Thermo Fisher Scientific, Inc.). For in situ hybridization studies, after paraffin removal with xylene/alcohol solutions, slides were air dried. Peroxidase quenching (10 min, away from light) was followed by antigen retrieval (20 min, RNAscope® Target Retrieval Reagent, ACD) and incubation with protease IV (30min, RNAscope® Protease IV Reagent, ACD). Excess solution was then washed off. Slides were then incubated for 2 hours in a 40° C hybridization oven with a Spp1, Tgfb1, or Apoe probe. A series of signal amplification steps (6 for single color detection, 10 for doublestaining assay) and washes were followed by chromogen development. At this point the experiment was either concluded with counterstain and toluene-based permount coverslip placement, or the immunohistochemistry protocol resumed from the blocking and primary antibody step as described above.

#### Fluorescence activated cell sorting

In some studies, following cardiac perfusion the left lobe was tied off with a suture and removed for flow cytometric and FACS analysis. Tissue was minced with surgical scissors and transferred into a 50 ml conical tube and incubated (37°C, 30 minutes) in DMEM + 5% FBS + 2 mg/ml Collagenase D (Cat #11088866001, Roche, Indianapolis, IN). Digested lungs were passed through 70-µm nylon mesh to obtain a single-cell suspension, counted and mixed with ACK Lysis Buffer (Thermo Fisher Scientific, Inc.) to remove any remaining red blood cells. The single cell suspension was counted and resuspended to yield  $1 \ge 10^6$  cells per 100µl of flow cytometry staining buffer (PBS+0.1% sodium azide). Cells were then incubated with anti-mouse CD16/32 antibody for 10 min at 4°C to block nonspecific binding. This was followed by 30-minute incubation with fluorescently-tagged antibodies or appropriate isotype controls  $(0.25-1.5 \ \mu g/10^6 \text{ cells})$  for 30 minutes (4°C). Cells were then spun and resuspended in staining buffer for viability staining (30 minutes at 4°C). Cells were fixed in 2% paraformaldehyde and sorted using a FACS ARIA (BD Biosciences). Alveolar macrophages (SigF+CD11b-CD11c<sup>+</sup>) were identified following forward and side scatter selection of singlet CD45<sup>+</sup> viable cells. To ensure cell sorting of a purified population, a series of exclusion gates were designed to remove eosinophils (SigF<sup>int</sup>CD11b<sup>+</sup>CD11c<sup>-</sup>), neutrophils (Ly6G<sup>+</sup>) and lymphocytes (CD3<sup>+</sup>). All analysis was performed using FlowJo software (FlowJo, LLC, Ashland, Oregon).

# Bulk and single-cell RNA sequencing preparation and analysis

For bulk RNA sequencing studies (deposited on NCBI GEO GSE166300), sorted SigF<sup>+</sup>CD11b<sup>-</sup>CD11c<sup>+</sup> macrophages underwent RNA extraction using Qiagen RNeasy Plus Universal mini kit following manufacturer's instructions (Qiagen, Hilden, Germany). Extracted RNA samples were quantified using a Qubit 2.0

Fluorometer (Life Technologies, Carlsbad, CA, USA) and RNA integrity was checked using Agilent TapeStation 4200 (Agilent Technologies, Palo Alto, CA, USA). RNA sequencing libraries were prepared using the NEBNext Ultra RNA Library Prep Kit for Illumina following manufacturer's instructions (NEB, Ipswich, MA, USA). Briefly, mRNAs were first enriched with Oligo(dT) beads. Enriched mRNAs were fragmented for 15 minutes at 94°C. Firststrand and second strand cDNAs were subsequently synthesized. cDNA fragments were end-repaired and adenylated at 3'ends, and universal adapters were ligated to cDNA fragments, followed by index addition and library enrichment by limited-cycle PCR. The sequencing libraries were validated on the Agilent TapeStation (Agilent Technologies, Palo Alto, CA, USA), and quantified by using Qubit 2.0 Fluorometer (Invitrogen, Carlsbad, CA) as well as by quantitative PCR (KAPA Biosystems, Wilmington, MA, USA). The sequencing libraries were pooled and clustered on 1 lane of a flow cell. After clustering, the flowcell was loaded on the Illumina HiSeq4000 instrument according to manufacturer's instructions. The samples were sequenced using a 2x150bp Paired End (PE) configuration. Image analysis and base calling were conducted by the HiSeq Control Software (HCS). Raw sequence data (bcl files) generated from Illumina HiSeq was converted into fastq files and de-multiplexed using Illumina's bcl2fastq 2.17 software. One mismatch was allowed for index sequence identification. Analysis of RNA counts was performed using R (3.6.3) (38). Differential gene expression analysis was conducted using the hciR package (39). Further pathway analysis was conducted using IPA (QIAGEN Inc). Datasets were filtered using log2 fold change (minimum -1 or 1) and p-value cut offs (minimum p<0.05) to ensure an appropriate number of molecules (200-3000) were used in the IPA pipeline.

Single-cell RNA sequencing raw data of SP-C<sup>I73T</sup> model were deposited in NCBI's Gene Expression Omnibus and are accessible through GEO Series accession numbers GSE247520 and GSE196657. Tissue collection and single cell suspension were achieved using mechanical mincing, digestion in Collagenase D, red blood cell lysis, and suspension created by using a 70- $\mu$ m strainer. RNA extraction and library preparation are described in the published manuscript (10). Mining of human control and IPF lungs are accessible through GSE136831 (40). As described by Adams et al., representative apical and basal segments of explanted lungs were minced mechanically, digested [elastase (30 U/ml) + deoxyribonuclease I (0.2 mg/ml) + liberase (0.3 mg/ml) + 1% penicillin/streptomycin], cleared of red blood cells, and single cell suspension created using a 100/70/40 strainers (40). Re-analysis of each single-cell dataset included dimension reduction and clustering by SCTransformation (0.3.5) using the Gamma-Poisson generalized linear model method (glmGamPoi, 1.8.0) and were performed using the Seurat (4.0.4) package (41-45). Multiple levels of resolution were evaluated using Clustree (0.5.0). The data was assessed for cell cycle effects using CellCycleScoring and regressed for uneven cell cycle expression across clusters. Cell types were identified using differential gene expression and all manual annotations were compared to those produced through automated classification using SingleR (1.10.0). A specific R package was used to interface with enrichR database (46). For pseudotime analysis, monocle-3 software was used, while cell-cell communication analysis was conducted with CellChat software (47-51).

#### Spatial transcriptomic analysis

For spatial analysis (deposited on NCBI GEO GSE264128), neutral buffered formalin-fixed lungs were inflated without bronchoalveolar lavage. Paraffin-embedded sections were baked for 1 hour at 60°C and stained according to the NanoString Leica BOND RX RNA FFPE Semi-Automatic protocol. Following Proteinase K (1.0ug/mL, 15 min) and ER2 (20 min) processing, sections were incubated with fluorescently tagged antibodies against PanCytokeratin (PanCK, Novus Biologics, NBP2-33200AF488; 1:400), Syto83 (S11364; 1:10, Thermo Fisher Scientific, Inc.), CD45 (Nanostring Technologies, 121302304; 1:5), and CD68 (Abcam, ab125212; 1:25). Regions of interest (ROI; N = 3, healthy; N = 4, peri-injured, N = 5, injured/remodeled) were selected based on histopathological assessment of inflammation and epithelial thickening and remodeling. Tissue segmentation selectively identified CD45<sup>+</sup>CD68<sup>+</sup> macrophages. The collection plate was then removed from the GeoMx instrument and prepared for sequencing. The GeoMx library was prepared, processed, and sequenced according to the NanoString NGS Readout User Manual Protocol (ref. MAN-10153-04). Sequencing was performed on the Illumina NovaSeq 6000, S2 v1.5 with a 100 cycles flowcell at 27 bp pair-end reads. Generated FASTQ files were then processed to DCC files utilizing the NanoString GeoMx NGS Pipeline according to manufacturer's instructions. Gene expression was analyzed using Nanostring DSP analysis software (NanoString Technologies) with built-in statistical analyses. Raw data counts were run through internal quality control and biological probe quality control.

#### Statistical analysis

Unless otherwise indicated, all data are presented as group mean  $\pm$  SE. Statistical analyses were performed with Prism GraphPad 9.4 (GraphPad Software, San Diego, CA). Student's t-tests were used for paired data; for analyses involving multiple groups, one-way or two-way analysis of variance (ANOVA) was performed with *post hoc* testing as indicated. Survival analyses were performed using Log Rank (Mantel-Cox) test. For spatial analysis, segments were filtered to 55% of the limit of quantitation (LOQ) to render the top 6,000 expressed genes. Filtered genes were normalized to Q3 (3<sup>rd</sup> quartile of all selected targets). Hierarchical clustering was performed as quality control. CD68 ROIs were compared across regions using a linear mixed model (LMM) with Benjamini-Hochberg (BH) correction and a random effect for the region. For RNA sequencing pathway analysis, Wilcoxon rank sum test was used. In all cases, statistical significance was considered at p  $\leq$  0.05.

### Results

# Spatial discrimination of activated macrophages during SP-C<sup>173T</sup>- induced injury

To spatially define the phenotype of macrophages involved in fibrotic lung injury, fluorescent antibodies recognizing DNA (green), CD45 (yellow) and CD68 (red) were used to perform bulk sequencing in the injured/remodeled or peri-injured alveolar regions of the lung 14 days post SP-C<sup>I73T</sup> injury. A control (noninduced) lung was used to define baseline activation. Identification of injured regions was based on histopathological evaluation of all five lobes in the tissue sections. Four peri-injured areas were selected: two in relative proximity to the injury (areas of interest 006 and 010) and two distal region (areas of interest 009 and 012) (Figure 1A). Cell-type deconvolution analysis was used to resolve monocyte/macrophage specific expression signature from the dataset (52) (Supplementary Table 1). Principal component analysis identified clustering of transcriptomes based on sampling annotation, with separation in the peri-injured regions based on proximity to injury (Figure 1B; Supplementary Table 1). Global pathway map showed extensive expression of genes involved in metabolism (including RNA metabolism), signal transduction, transcriptional regulation, immune system function and cell cycle (Supplementary Figure 1A). Reactome-based analysis revealed 20.9% (148/712 pathways) and 34.9% (249/712 pathways) of pathways as differentially expressed when comparing peri-injury vs. healthy macrophages and peri-injury vs. injured cells, respectively. Notably, macrophages isolated form injured regions did not produce a particularly strong signaling signature and demonstrated a considerable degree of similarity to healthy macrophages (≈90% of pathways) (Figure 1C; Supplementary Figures 1B-D). Peri-injured macrophages were enriched in pathways associated with neutrophil degranulation, ROS/RNS production and release, activation in oxidative stress-induced senescence, proremodeling functions (TGF-beta and GPVI cascade), and metabolic alterations (citric acid cycle, gluconeogenesis, lipoprotein assembly/ remodeling/clearance) compared to healthy- and injured-derived CD68<sup>+</sup> macrophages (Figures 1D, E; Supplementary Table 1). When compared to controls, injured region macrophages were defective in programmed cell death signaling, ROS detoxification, and elastic fiber formation, but displayed enhanced DNA damageinduced senescence, degradation of extracellular matrix (ECM), and metalloproteinase function (Figure 1F). Hierarchical gene expression analysis of pathways related to degradation of ECM (Figure 1G), assembly of collagen fibrils (Figure 1H), complement system, adaptive immunity, and L13a-mediated ceruloplasmin expression all showed robust clustering based on region of origin (Supplementary Figures 1B-D, Supplementary Table 1).

Among the top 6,000 genes expressed genes in the dataset, genes associated with complement responses (*C1qa*, *C1qb*, *C1qb*), myeloid cell recruitment (*Cxcl14*, *Cxcl15*), inflammation (*Csf1r*, *Ctss*, *App*) and metalloproteinases (*Mmp13*, *Mmp14*, *Mmp15*) were most abundant in peri-injured macrophages (Figure 1I). Despite increases in normalized counts for fibrogenic genes and significant enrichments in the associated pathways, expression of the individual genes did not meet significance (*Fn1*, *Spp1*) (Figure 1J; Supplementary Figures 1E–G).

#### Temporal heterogeneity of resident alveolar macrophages during SP-C<sup>173T</sup>induced injury

To better understand macrophage behavior in the events leading up to end-stage fibrosis, we sought to examine transcriptional



Spatial analysis of CD68<sup>+</sup> macrophages in SP-C<sup>173T</sup> induced injury. Paraffin-embedded sections from controls or 14 days post SP-C<sup>173T</sup> injury were stained with fluorescent antibodies against CD45, CD68, and DNA. (A) Twelve alveolar regions were selected for CD68<sup>+</sup> macrophage RNA sequencing (cells labeled in red). Regions from an untreated mouse were used as controls (three technical replicates; yellow asterisks). Injured/ remodeled (five technical replicates; white asterisks) and peri-injured (two proximal technical replicates, orange asterisks; two distal technical replicates; red asterisks) region selection was based on histopathological assessment of inflammation and loss of alveolar architecture structure. Insets show representative regions. (B) Three dimensional PCA plot shows the clustering of samples based on sampling regions. Orange asterisks indicate two proximal technical replicates; red asterisks indicate two distal technical replicates. (C) Venn diagram of Reactome-based pathway analysis shows differentially regulated pathways between each pairing. (D−F) Bar graphs representing normalized enrichment scores for selected pathways. (G-H) Hierarchical clustering of 'Degradation of the extracellular matrix' and 'Assembly of collagen fibrils and other multimeric structures' signaling pathways in healthy (dark purple), peri-injured (pink), and injured (green) regions of the lung 14 days after SP-C<sup>173T</sup> induced injury. (I) Box plots of selected genes associated with macrophage activation (chemokine, metalloproteinases, complement cascade). The Y-axis represents normalized counts. A p-value ≤0.05 was considered significant using Linear Mixed Model. (J) Volcano plot comparing gene expression between peri-injured and injured macrophages. Fold changes are represented on log2 scale. Significance is shown as -log10(pvalue) using linear mixed model. In red are representative complement and fibrosis-associated genes.

changes in resident macrophages using a model of fibrogenic injury triggered by mutant SP-C<sup>I73T</sup>. For these studies, we leveraged bulk RNA sequencing of flow cytometry sorted CD11b-SigF<sup>+</sup>CD11c<sup>+</sup>CD64<sup>+</sup> resident mature macrophages isolated from naïve (controls) and inflamed lungs. Exclusion gate ensured no contamination from Ly6G<sup>+</sup> neutrophils, B220<sup>+</sup> B cells, and CD3<sup>+</sup> lymphocytes. Ingenuity Pathway Analysis highlighted distinct gene and signaling signatures in lung macrophages during inflammatory initiation (3 days post mutant induction) and early remodeling (14 days). Principal component analysis highlighted transcriptional variance across the study groups (Figure 2A). Differential gene expression analysis revealed a relatively small set (48 genes) between controls and 3-day injury, while these responses were more pronounced between controls vs. 14-day post induction (3393) or 3-day vs. 14-day comparison (1446) (Figure 2B). Volcano plots showed increases in genes linked to innate immunity (the hematopoietic transcription factor, Gata2; histidine decarboxylase, Hdc; interferon induced transmembrane protein 1, Ifitm1; colony

stimulating factor 1, Csf1; chemokine ligand, Ccl17; immunoglobulin epsilon receptor, Ms4a2) and metabolism (cholesterol side-chain cleavage enzyme, Cyp11a1; ATPase Na+/K+ transporting subunit alpha 3, Atp1a3; adenylate cyclase 6, adcy6; peptidyl arginine deiminase 2, Padi2; and myristoylated alanine rich protein kinase C substrate, Marcks) 3 days post injury (Figure 2C). By comparison, macrophage expression profile was bidirectional at 14 days, and demonstrated more sizable changes (as represented by adjusted pvalues and log fold changes). Immunity, cell cycle, and metabolism genes were among the most significantly altered genes (complement C1q C-chain, C1qc; peptidoglycan recognition protein 1, Pglyrp1; ADAM metallopeptidase domain 19, Adam19; secreted protein acidic and cysteine rich, Sparc; apolipoprotein E, Apoe; insulin like growth factor 1 receptor, Igf1r; toll-like receptor 7, Tlr7; cyclin D2, Ccnd2 (Figure 2D). Ingenuity Pathways Analysis predicted signaling related to activation, proliferation, and apoptosis of leukocyte predominantly at 3 days, while distinct cell movement and chemotaxis pathways, proliferation, angiogenesis and fibrogenesis were projected to be



Transcription analysis of tow solutions between the optimizer of the solution in the control of the solution of the solution

induced 14 days post-induction (Figure 2E; Supplementary Table 2). Specific pathways linked to NFAT dependent regulation of immune responses (directionality/z-score 1.90), Th1 and Th2 activation (no predicted z-score), granulocyte adhesion and diapedesis (no predicted z-score), and STAT3 signaling (z-score 1.13) were noted 3 days post injury. Comparatively, glycoprotein 6 (z-score 5.48), IL-15 production (z-score 5.00), fibrosis (z-score 6.71) and epithelialmesenchymal transition signaling (no z-score), and osteoarthritis (zscore 3.34) were identified in macrophages at 14 days (Figure 2F). Notably, IPA's Upstream Regulators analysis predicted activation of pro-inflammatory signals at 3 days (NFKB1, IL5, IFNG, prostaglandin E2, STAT3 and STAT6), and conventional fibrogenic pathways (TGFB1, TP53, FGF2, and SOX2) 14 days (Figure 2G). Reactome-based breakdown of top differentially expressed genes from the Interferon-y pathway identified distinct gene-sets expressed during inflammatory initiation (Ifi203, Ifi206, Ifi209, Ifi213, Stat4) and 14 days post injury (S100a8, Arg1, Csf1, Alox15, Retnla, Il4) (Figure 2H; Supplementary Table 2). A comparable dual response was observed in the TGF-B1 (3 days: Ms4a2, Ctsk, Clec2i

and Cd55; 14 days: *Il1rl1*, *Arg1*, *Alox15*, *F13a1*, *Itgam*, *Ptgs2*, *C1qc*, and *Anxa8*) and IL-4 signaling pathways (3 days: *Ms4a1*, *Cxcl5*, *Il6*, *Stat4*; 14 days: *S100a8*, *Agf2r*, *Csf1*, *Retnla*, *Il4*) (Figures 2I, J). ELISA-based validation of this established pro-fibrotic pathway confirmed increases in IL-4 and IL-13 expression in the bronchoalveolar lavage fluid 7-14 days post SP-C<sup>173T</sup> injury (Figure 2K). Notably, IL-15 and Glycoprotein-6 signaling, as well as angiogenesis displayed time restricted enrichments (14 days post-induction) (Supplementary Figures 2A, B, and not shown).

#### Single-cell RNA sequencing reveals activation of distinct monocyte/ macrophage clusters responding to SP-C<sup>173T</sup> induced injury

We then employed single-cell RNA sequencing to overcome the constraints (and therefore bias) of antibody-based analysis of macrophages involved in SP-C<sup>173T</sup> induced injury. A 59,440-cell

dataset including healthy controls (29,213 cells), and two injury times representing peak inflammation (14 days - 14,266 cells) and established fibrosis (42 days - 15,961 cells) were studied. Population clustering using SCTransformation with the glmGamPoi method at a resolution of 0.4 yielded 33 clusters (Supplementary Figure 3A). Partition of the clusters based on origin/identity (Ctl, 14 days, 42 days) highlighted shifts in endothelial cells (cluster 25), eosinophils (cluster 8), and macrophages (cluster 9) composition after SP- $C^{I73T}$ injury (Figures 3A, B). A combination of SingleR, manual annotation, and the top three non-redundant genes from each populations was used for identification of epithelial (3), endothelial (6), mesenchymal/stromal (3), megakaryocytes, granulocytes (3), B cells (4), lymphocytes (4), and mononuclear myeloid populations (9) (Figure 3C; Supplementary Figures 3B-I). Macrophage Cluster 19 (identified in only one of the eleven specimens) and Cluster 32 (of low abundance and merged with Cluster 3 after pseudobulk analysis determined high degree of transcriptional overlap).

Uniform Manifold Approximation and Projection (UMAP) analysis for the pan-macrophage gene *Cd68* and the mobilization

marker Itgam/Cd11b were utilized to locate resident and recruited macrophages, monocyte-derived, and by exclusion monocytes and dendritic cells (Figures 3D, E). A curated gene set (mobility, maturity, and activation markers) combined with pseudotime analysis was used to clearly differentiate mononuclear myeloid clusters (Supplementary Table 3). Cluster 3 was annotated as alveolar macrophages since it expressed a combination of Cd68, Itgax, Ear1, Ear2, Siglecf, and moderate levels of Mertk (Figure 3F). Reconstructed trajectory analysis also predicted terminal differentiation within the alveolar macrophage cluster (Figure 3G). Cluster 9, identified as interstitial macrophages (Cd68, Msr1, and C1qa), was labeled based on its specific expression of Trem2 (Figure 3F and not shown). Cluster 21 (macs\_mertk) was identified as a mature subset based on pseudotime analysis and distinctive expression of Mertk alongside Itgax, Mrc1, Atp6v0d2, Kcnip4. This cluster was found solely in controls and 42 days post injury (Figures 3F, G). Two monocytelike subsets were annotated as non-classical/NC (cluster 4 - Csf1r, Cx3cr1) and classical/inflammatory (cluster 11 - Ccr2, Ly6c2, Lyz2, Ms4a6c, F13a1) (Figure 3F; Supplementary Table 3). Cluster 10 was



#### FIGURE 3

Single-cell RNA sequencing of SP-C<sup>I73T</sup> induced injury and fibrosis. Single-cell RNA sequencing analysis was performed on collagenase digested single cell suspensions from controls (Ctl, oil treated SP-C<sup>I73T</sup> mice), 14 days or 42 days post SP-C<sup>I73T</sup> induction. **(A)** UMAP analysis overlaying cells from controls (cI73T, pink), 14 days (t14d I73T, green), and 42 days (t42d I73T, blue). **(B)** Split identify UMAP depiction of all cellular clusters identified in controls (cI73T), 14 days (t14d I73T), and 42 days (t42d I73T) at 0.4 resolution. **(C)** Cluster breakdown identifying 7 endothelial, 3 mesenchymal, 3 epithelial, 2 polymorphonuclear, mast cells, 4 B cell, 4 lymphocytic, and 9 mononuclear myeloid clusters. Note that mononuclear myeloid clusters were renamed based on single R and manual annotation: cluster  $3 = Mac_Alv$ ; cluster  $4 = Mono_non$  classical/NC; cluster  $9 = Mac_Trem2$ ; cluster  $10 = Mono_DC$ ; cluster  $11 = Mono_inflammatory/infl;$  cluster 32 = was combined with cluster  $3/Mac_Alv$  since pseudobulk analysis revealed analogous gene expression and pathway activation. **(D, E)** UMAP analysis for *Cd68* and *Itgam/Cd11b* defines distribution and cellular expression in the SP-C<sup>173T</sup> lung. **(F)** Bubble plot of selected genes associated with macrophage maturation (*Trem2, Itgax, Ear1, Ear2, Atp6v0d2, Siglecf, Mertk*), recruitment and activation (*F13a1, Ly6c2, Ccr2, Cx3cr1, Ctsk, Ly22, H2-eb1, CcL22, Ccl17, Cd80, Cd86*) in all annotated monocytes/macrophages/ dendritic cell clusters. **(H)** Enrichr-based Reactome pathway analysis for the top-15 predicted pathways in the 7 resulting monocyte/macrophage/DC clusters. Single gradient color heatmap shows logarithmic adjusted p-values. Values above the arbitrary threshold (-log of adjusted p-value) of 20 were color-coated in black.

annotated as monocyte-derived DCs based on a mixed myeloid and lymphoid signature (Ccr2, Itgae/Cd103, Cd209/Dc-sign) and major histocompatibility complex genes. A cluster of lymphoid-derived DCs (cluster 24) lacking all classical myeloid features and expressing Cd80, Cd86, Cd40, Il4i, Ccr7, Ccl17, and Ccl22 was also annotated (Figure 3F). Split identity analysis of these curated gene sets also highlighted time related changes in the expression of established macrophage maturity and identity genes (e.g. Atp6v0d2, Itgax, and Cd68 expression increases, downregulation of Siglecf, Ear1, and Ear2) (Supplementary Figure 4A). Similar trends were observed for neutrophils (PMN 1 lost Ptgs2 and Tgm2 in favor of Retnlg and Lcn2), eosinophils (PMN 2 - Siglecf, Itgam, Cxcr2, Cd33, Csf3r, Tgfbr1), epithelial cells (Epi1 - Cldn18, Ager, Hopx, Krt8, Krt19, Sftpb), mesenchymal/stromal cells (Mes 2 -Tgfbr3, Ccn1, Pdgfra, Npnt, Loxl1, Ecm1, Fgf2), and lymphatic endothelial cells (Endo 6 - Nrgn, Itga2b, Gp1bb increased 14 days post injury). By comparison, B and T cells showed limited transcriptional fluctuation over the 42-day injury (Supplementary Figures 4B–G).

Pseudobulk differential expression analysis of Trem2<sup>+</sup> macrophages surveyed gene expression and patterns during injury. Approximately 700 genes demonstrated a transient drop in abundance at 14 days (expression pattern group-1), while the abundance of 621 genes was significantly increased following SP-C<sup>I73T</sup> induction (expression pattern group-2) (Supplementary Figure 5). Smaller gene sets were shown to transiently increase at 14 days (122 genes, expression pattern group-3) or steadily decrease after SP-C induction (61 genes, expression pattern group-4) (Supplementary Figure 5). Notably, Reactome-based analysis predicted no significant pathway to be altered for genes annotated in expression pattern-1, while cytokine and interleukin signaling, immunity, antigen presentation, and lipid and carbohydrate metabolism were among significant pathways for expression pattern-2 genes. By comparison, transcript abundance in inflammatory monocytes followed two expression patterns: transient decrease (237 genes) or increase (158) at 14 days post injury. The latter predicted engagement of pathways related to complement cascade, extracellular matrix organization and immune responses (Supplementary Table 3).

Reactome-based analysis of the top-15 most significantly regulated pathways revealed shared inflammatory signature ('immune system', 'neutrophil degranulation', 'ROS and RNA production in phagocytes') between alveolar macrophages and Trem2<sup>+</sup> interstitial macrophages, though at considerably different adjusted *p*-values (Alv macs *p*-value <10<sup>-25</sup> vs. *Trem2*<sup>+</sup> *p*-value <10<sup>-</sup> <sup>103</sup>) (Figure 3H; Supplementary Table 3). By comparison, Mertk<sup>+</sup> cells were not predicted to engage in innate or adaptive immunity or cytokine mediated signaling, but generated a strong GTPase signature (CDC42, RAC, RHO, and ROBO/SLIT). Examination of monocyte-like clusters highlighted analogous 'cytokine signaling in immune system', 'innate immunity', and 'neutrophil degranulation' engagement. Cx3cr1<sup>+</sup> non-classical cells excelled in 'VEGF signaling', 'signaling by tyrosine kinases', and 'Rho GTPase signaling', while Ccr2<sup>+</sup>Ly6c2<sup>+</sup> classical/inflammatory monocytes produced a signature related to RNA synthesis and translation and pathways involved in 'cellular response to stress'. Lastly, monocyte-derived and

lymphoid DCs produced comparable activation profile (infectious disease response, translational elongation and termination, and ribosomal homeostasis) (Figure 3H).

# Macrophages from human IPF display a comparable phenotype as murine SP-C<sup>173T</sup> injury.

A recent dataset published by Adams and colleagues provided in depth assessment of epithelial, endothelial, mesenchymal, and immune populations in healthy and IPF tissue explants (40). The mononuclear myeloid compartment from this set (44,226 cells; 12,514 from controls and 31,712 IPF, GSE136831) was reexamined to define similarities between clinical and experimental fibrosis. To avoid overfitting the data, the lowest resolution was utilized (0.1), ultimately generating 5 major subsets (Supplementary Figure 6A). UMAP analysis identified clusters 3 and 4 solely in the controls, with an expansion in clusters 0 and 1 was noted in IPF (Figure 4A). Cluster-based analysis of top-5 non-redundant genes revealed differential expression for SPP1, CTSK, MMP7 (cluster 1) and FABP4 (cluster 2) (Figure 4B; Supplementary Table 4). Diseasebased analysis highlighted notable changes in cluster 0 (SPP1, C1QC, APOE) and cluster 1 (THBS1, FN1), consistent with our murine dataset, including, (Figures 4B, C, Supplementary Table 5). A curated set of 25 identity and lipid associated genes, used to annotate this clusters, identified cluster 0 as an interstitial macrophage population (widespread expression of ITGAX, TREM2, MRC1, ITGAM, APOE) and Cluster 1 as alveolar macrophages (ITGAX/CD11C, ITGAM, ABCA1). Notably, cluster 1 also displayed a signature characteristic of a potentially pro-fibrotic monocyte-derived population in IPF lungs (TREM2, MERTK, ATP6V0D2, CTSK, COL4A2, MMP7, MMP9) paired with a shift in metabolic function (FABP5, LPL and LIPA increases). Clusters 2, 3 and 4 did not exhibit distinguishing signatures besides overexpression of major histocompatibility complex genes (HLA-DRB6) (Figure 4C; Supplementary Figure 6B). UMAP analysis of the distribution of MERTK and fibronectin1/FN1 demonstrated widespread presence in cluster 0 and 1 of the IPF lung, while ostepontin1/SPP1 was restricted to cluster 1 (Figures 4D-F). Tumor growth factor  $\beta 1/TGF\beta 1$  was expressed in all cells regardless of disease state (Supplementary Figures 6C, D). Enrichrbased pathway analysis using the Reactome database highlighted IPF induced changes (annotated as "global") in all facets of immune cell behavior, including "Immunity", "Cellular Responses to Stress", "Cytokine Signaling in Immune System", complement activation, and GTPase signaling, and control of transcription and translation. Clusters 0, 1, 2 were responsible for the majority of these signals (Figure 4G; Supplementary Table 4). Direct comparison of Reactome pathways from the spatial analysis and both the murine and human single-cell sequencing datasets highlighted similarities between cluster 1 annotated in the human IPF dataset and inflammatory monocytes from the SP-C<sup>I73T</sup> mouse model. No distinctive similarities were noted with respect to CD68<sup>+</sup> macrophages sampled from peri-injured and injured regions of the lung during our spatial analysis (Figure 4H).



Single-cell RNA sequencing of annotated macrophages from human IPF. Macrophages from a published IPF dataset (GSE136831) were re-mined. (A) Split identity UMAP analysis of clusters annotated at 0.1 resolution using the Seurat analysis package. (B) Cluster-based and disease-based top-5 non-redundant genes from annotated clusters. (C) Split identity UMAP shows expression distribution for *TREM2* in control and IPF lungs. (D) Split identity bubble plot of selected genes associated with macrophage maturation (*TREM2, MERTK, ATP6V0D2, ITGAX*), recruitment and activation (*ITGAM, CD163, HLA-DRB6, CTSK, COL3A1, COL4A2, MMP7, MMP9, FN1, SPP1, TGFB*), and lipid homeostasis (*APOE, CD36, LP1, LPL, LPA, ABCA1, ABCG1, FABP5, PPARG*). (E) Split identity UMAP showing expression distribution for *SPP1* in control and IPF lungs. (H) Enrichr-based Reactome pathway analysis for the top-15 predicted pathways between controls and IPF samples. Single gradient color heatmap shows logarithmic adjusted p-values) of 6 were color-coated in black. (I) Enrichr-based Reactome pathway analysis comparing significant pathways in mononuclear myeloid clusters from the human IPF dataset, SP-C<sup>173T</sup> spatial transcriptomics, and SP-C<sup>173T</sup> single-cell sequencing datasets. Single gradient color heatmap shows logarithmic adjusted p-value) of 10 were color-coated in black.

# Osteopontin1 and fibronectin1 intercellular communication during SP-C<sup>173T</sup> induced injury

CellChat software was utilized to predict communication among cellular clusters during SP-C<sup>I73T</sup> induced fibrotic injury. Initial analysis grouped all 33 clusters into 7 macro-groups (epithelial, endothelial, stromal, B cells, granulocytes, lymphocytes, and Mo/Mac/DCs) (Supplementary Figure 7A) (51, 53). CellChat based aggregation of ligand: receptor expression estimated increases in differential number and strength of interaction between mesenchymal-epithelial and mesenchymal-Mo/Mac/DCs and within epithelial clusters 14 day post-injury, relative to controls. Within the Mo/Mac/DC macro-groups the communication was estimated to increase in strength (Supplementary Figure 7B). Interrogation of the inter-cluster network 42 days post-induction showed increased interactions in stromal and Mo/Mac/DC macro clusters relative to controls and 14 days (Supplementary Figures 7C, D), with incoming and outcoming signals from the Mo/Mac/DC macro-cluster associated with SPP1 and FN1 pathway (Supplementary Figures 7E-G, Supplementary Tables 6-9).

To pinpoint the exact cellular origin of these signals, CellChat analysis was carried out after split of the Mo/Mac/DC macrocluster. This analysis revealed alveolar macrophages as a dynamic cluster at baseline (both incoming/receptor-based and outgoing/ ligand-based interactions), while Trem2<sup>+</sup> macrophages produced high volume of incoming/outgoing signals 14 days post injury and Mertk<sup>+</sup> cells became active at 42 days. Classical/inflammatory monocytes maintained comparable network profile throughout the 42-day time course, with non-classical monocytes effectively reducing outgoing signals after SP-C<sup>I73T</sup> injury (Figure 5A; Supplementary Figures 8A-C). Split identity pathway analysis across the 42-day time course demonstrated the origin of the osteopontin1/SPP1, fibronectin1/FN1, chemokine ligand/CCL, laminin, semaphorin3/SEMA3, galectin, complement, and collagen signaling to be driven by Trem2<sup>+</sup> macrophages at 14 days (Figure 5B). By 42 days, the predicted interaction strength was comparable across the monocyte and macrophage clusters (Supplementary Figures 8D-F). Connectome ring plots were used to capture the changes and directionality of these interactions during injury and fibrosis. Analysis of CCL signaling predicted baseline crosstalk among all annotated mononuclear myeloid cells and granulocytes. While these responses were unaffected 14 days

post SP-C<sup>I73T</sup> injury, by 42 days there was a global activation of this signaling network (Figure 5C). Differential expression of chemokines/cytokines/interleukins and their receptors highlighted distinct signatures in alveolar macrophages (Ccl6, Il18, Cxcl2, and Il1a), Trem2<sup>+</sup> macrophages (Ccl12, Ccl2, Ccl24, Ccl9, Cxcl16, Ccr5, Il10rb, Il11ra1), while Mertk<sup>+</sup> macrophages and monocyte-derived DCs were relatively quiescent. Classical and inflammatory monocytes presented a receptor dominant repertoire (Cx3cr1, Il10ra, Il17ra, Il6ra - inflammatory monocytes exclusively expressed Ccr2) (Figures 5D, E). CellChat-based network analysis for complement signaling predicted outgoing communication from classical monocytes to other myeloid clusters in all conditions, with a transient activation originating from stromal cells 14 days post injury (Figure 5F). Differential expression revealed distinct signatures in trem2<sup>+</sup> macrophages (C1qa, C1qb, C1qc, C3ar1) and stromal cells (stromal 2 - C1qtnf7, C1ra, C2, C3, C4b, C7), with granulocytes also expressing C3ar1, C5ar1, and C5ar2 (Figure 5G).

#### Trem2<sup>+</sup> macrophages coordinate profibrotic communication during SP-C<sup>173T</sup> induced injury and fibrosis

We then query the dataset for distinctive extracellular matrix reorganization gene signatures. Unsurprisingly, mesenchymal cells produced signals from collagen genes, metalloproteinases, laminins, and platelet-derived growth factor receptor alpha and beta (Supplementary Figure 9). Expression of *Timp1*, *Mmp12*, and *Mmp14 were restricted to Trem2*<sup>+</sup> cells, while *Spp1*, *Fn1*, and *Mmp19* were shared with alveolar macrophages (Figure 6A). *Tgfb1* transcripts were maximal in inflammatory monocytes, though UMAP analysis of distribution suggested widespread expression in the lung (Figures 6A, B). Notably, *Spp1*, and *Fn1* expression was also found in the endothelial and mesenchymal compartment (Figures 6C, D). Signaling connectome ring for *Tgfb1* highlighted baseline signals emanating primarily from alveolar



#### FIGURE 5

Cell-cell communication analysis in SP-C<sup>173T</sup> induced injury and fibrosis. CellChat software was used to estimate cell-cell communication in controls (Ctl, oil treated SP-C<sup>173T</sup> mice), 14 days or 42 days post SP-C<sup>173T</sup> induction. **(A)** Ligand:receptor expression analysis examining incoming (y-axis) and outgoing (x-axis) signals in controls, 14 days or 42 days post injury. Note that individual endothelial, mesenchymal, epithelial, granulocyte, B cell, and lymphocyte clusters were combined into "macro-clusters", while mononuclear myeloid cells remained split. The size of the circles is representative of population size. **(B)** Prediction of signaling changes in *Trem2*<sup>+</sup> macrophages between control and 14-day, control and 42-day, and 14-d and 42 d. Plot legend describes directionality (circle - shared by both groups, square - incoming in a specific group, triangle - outgoing in a specific group, diamond - incoming and outgoing in a specific group) and signal specificity (black - shared by both groups, orange - control specific, cyan - injury/ tamoxifen specific). **(C)** Connectome ring plots for chemokine ligand/CCL signaling pathway network predict directionality and communication strength among clusters. **(D, E)** Split identity bubble plot for chemokine/cytokine/interleukin ligands and receptors among mononuclear myeloid clusters. Note that the size of the bubble indicates the relative abundance of the population expressing target gene. Color coating indicates average expression. **(F)** Connectome ring plots for Complement signaling network predicting directionality and strength of communication among clusters. **(G)** Split identity bubble plot for complement associated genes among mononuclear myeloid clusters. Note that the size of the bubble indicates the relative abundance of the population expression.

macrophages, non-classical monocytes, monocyte-derived DCs, and B cells. Despite transient drop 14 days post SP-C<sup>I73T</sup> induced injury, the network returned to control levels by 42 days, with involvement of *Mertk*<sup>+</sup> macrophages (Figure 6E; Supplementary Tables 6-9). Osteopontin1 signaling network revealed alveolar macrophages as the sole driver in control lungs. SP-C<sup>I73T</sup> induced injury produced increases in alveolar and Trem2<sup>+</sup> macrophages (14 days), and ultimately global expression in all lung cells (42 days) (Figure 6F; Supplementary Tables 6-9). Inflammatory monocytes and alveolar macrophages were shown to engage in fibronectin 1 signaling in the control and injured lung. Notably, Trem2<sup>+</sup> macrophages and stromal cells were predicted to partake in FN1 communication 14 days post-injury, while Mertk<sup>+</sup> macrophages were shown at 42 days (Figure 7G; Supplementary Tables 6-9). Although spatial examination did not detect significant differences among CD68<sup>+</sup> macrophages collected from healthy, peri-injured, and injured regions of the lung 14 days post injury, we found that proximity to the remodeled tissue (areas of interest 006 and 010) resulted in higher expression of Tgfb1, Spp1, and Fn1 (Figures 6H-J).

# Genetic ablation of ApoE does not impact SP-C<sup>173T</sup> induced injury

Curated analysis for genes involved in lipid handling and metabolism showed heterogeneous expression among mononuclear myeloid cells during SP-C<sup>173T</sup> induced injury and fibrosis. Monocytes exhibited a limited repertoire of lipid-associated

genes. Alveolar macrophages displayed an extensive signature at baseline (Fabp1, Plin2, Lipa, Lpl, Lpin1, Abcg1, Acox1) which expanded after injury (Nr1h3, Marco, Fabp4). Mertk+ macrophages developed distinct signatures at 14 days (Nr1h3, Marco, Plin2, Abca1) and 42 days (Pparg Ldlr, Lrp1, Lrp12, and Acox1), a time coordinated with their maximal abundance (Figure 7A). Trem2<sup>+</sup> gene expression was restricted to ATP binding cassette transporters (Abca1, Abca9) and exclusive Apoe expression. Analysis of the distribution of Apoe and two of its established receptors (Lrp1, Ldlr) confirmed predominant expression in mononuclear myeloid and mesenchymal clusters (Figures 7B-D) (54). Immunohistochemical analysis of ApoE showed increases in parenchymal and immune cells 7 days and 14 days post induction, in particular within injury foci (Figure 7E). In situ hybridization analysis validated the origin of Apoe in CD68<sup>+</sup> macrophages up to 42 days after SP-C<sup>I73T</sup> injury, while western blot analysis of BAL fluid also confirmed protein increases after SP-C<sup>I73T</sup> induced injury (Figures 7F, G). Our bulk RNA sequencing predicted activation of the ApoE signaling pathway in SigF<sup>+</sup>CD11c<sup>+</sup>CD64<sup>+</sup> alveolar macrophages, however its expression (or that of its receptors) was not spatially restricted (Figures 7H, I). To test the hypothesis that ApoE is directly implicated in Mo/Mac/DC function during pulmonary remodeling, mono- and bi-allelic deletion of ApoE (SP-C<sup>I73T</sup>ApoE<sup>HET</sup> and ApoE<sup>HOM</sup>) was designed (Figure 7J). Neither ApoE hypofunctional or null mice impacted accumulation of Spp1<sup>+</sup> macrophages within areas of injury 14 days post SP-C<sup>I73T</sup> induction (Figure 7K). Similarly, there was no improvements in pulmonary histopathological scoring, BAL cell counts, or mortality (Figures 7L-O).



#### FIGURE 6

Pro-fibrotic cell-cell communication in SP-C<sup>173T</sup> induced injury and fibrosis. CellChat-based analysis of fibrogenic signaling from controls (Ctl, oil treated SP-C<sup>173T</sup> mice), 14 days or 42 days post SP-C<sup>173T</sup> induction. **(A)** Split identity bubble plot for fibrosis-associated genes among mononuclear myeloid clusters. Note that the size of the bubble indicates the relative abundance of the population expressing target gene. Color coating indicates average expression. **(B-D)** UMAP analysis for *tgfb1, spp1, fn1.* **(E-G)** Connectome ring plots for TGFb, osteopontin/SPP1 fibronectin/FN1 signaling pathway network predicting directionality and communication strength among clusters. Note that only Mo/Mac/DC clusters are split. **(H-J)** Box plots for *Tgfb1, Spp1, Fn1.* CD68<sup>+</sup> macrophages isolated from healthy regions are shown in dark purple, peri-injured macrophages in pink, and injured macrophages are shown in green. The Y-axis represents normalized counts.



FIGURE 7

Lipid signature in SP-C<sup>173T</sup> induced injury and fibrosis. Lipid-associated gene expression was assessed in controls (Ctl, oil treated SP-C<sup>173T</sup> mice), 14 days or 42 days post SP-C<sup>173T</sup> induction. (A) Split identity bubble plot for lipid-associated genes among mononuclear myeloid clusters. Note that the size of the bubble indicates the relative abundance of the population expressing target gene. Color coating indicates average expression. (**B-D**) UMAP analysis for *Apoe, Lrp1, Ldlr.* (**E**) Immunohistochemical analysis for ApoE in controls, 14 days or 42 days post SP-C<sup>173T</sup> induced injury. A representative image is shown. (**F**) Combination of immunohistochemistry (CD68, brown) and in situ hybridization (Apoe, magenta). Protein expression was visualized using a DAB Vectastain kit. Original magnification, 400x. Insets, 1000x. (G) ApoE western blot analysis of BAL fluid from SP-C<sup>173T</sup>control (Ctl), 7 days, 14 days, and 42 days post induction (N=3). Red arrowhead indicates molecular weight band for ApoE (34 KDa). (H) Heat maps of significantly altered genes associated with ApoE signaling in flow cytometry sorted CD11b<sup>-</sup>SigF<sup>+</sup>CD11c<sup>+</sup>CD64<sup>+</sup> resident lung macrophages 3- and 14-days after injury. The criteria for significance was a 5% false discovery rate. Note that the orange box highlights signatures specific to a given time point. (I) Box plots for Apoe, Lrp1, Ldlr. The Y-axis represents normalized counts. A p-value <0.05 was considered significant using Linear Mixed Model. (J) ApoE western blot analysis of BAL fluid from control and 14 days SP-C<sup>173T</sup>ApoE<sup>WT</sup> (controls N=2; 14 days N=2), SP-C<sup>173T</sup>ApoE<sup>HET</sup> (controls N=2; 14 days N=2), and SP-C<sup>173T</sup>ApoE<sup>HOM</sup> (controls N=2; 14 days N=3). Red arrowhead indicates molecular weight band for ApoE (34 KDa). Black brackets represent ApoE<sup>WT</sup>, Apoe<sup>HET</sup>, ApoE<sup>HOM</sup> BAL fluid from control SP-C<sup>173T</sup> mice. Red brackets represent ApoE<sup>WT</sup>, Apoe<sup>HET</sup>, Apoe<sup>HOM</sup> BAL fluid 14 days post SP-C<sup>173T</sup> induced injury. **(K)** Duplex *in situ* hybridization analysis for *Tgfb1* (pink) *Spp1* (blue) in SP-C<sup>173T</sup>ApoE<sup>WT</sup>, SP-C<sup>173T</sup>ApoE<sup>HET</sup>, and SP-C<sup>173T</sup>ApoE<sup>HOM</sup> lungs 14 days post injury. Representative images are shown. **(L-M)** H&E staining and histological scoring of SP-C<sup>173T</sup>ApoE<sup>WT</sup>, SP-C<sup>173T</sup>ApoE<sup>HET</sup>, and SP-C<sup>173T</sup>ApoE<sup>HOM</sup> lungs 14 days post injury. Pathological scoring included counting of foci of injury, extent of inflammation, edema, hemorrhage, and alveolar architecture remodeling. Representative images are shown. (N) Bronchoalveolar lavage fluid (BAL) cell counts of SP-C<sup>173T</sup>ApoE<sup>WT</sup>, SP-C<sup>173T</sup>ApoE<sup>HET</sup>, and SP-C<sup>173T</sup>ApoE<sup>HOM</sup> samples (controls and 14 days post injury). Data are presented as mean  $\pm$  SEM (n = 4-9 mice/group), analyzed using two-way ANOVA. A p<0.05 (\*) was considered significant. Lines mark significant groups. (O) Kaplan-Meier survival curve of SP-C<sup>173T</sup>ApoE<sup>WT</sup>, SP-C<sup>173T</sup>ApoE<sup>HET</sup>, and SP-C<sup>173T</sup>ApoE<sup>HOM</sup> lungs 14 days post injury. Analysis includes mice found dead or displaying ≥20% body weight from study initiation. Log-Sum (Mantel-Cox) Rank test was used.

### Discussion

Genetic mutations in key nodes of pulmonary epithelial function are intertwined with familial forms of pulmonary fibrosis (PF) and other interstitial lung diseases (55). Dysbiosis among parenchymal cells promote repeated cycles of injury, inflammation, epithelial-to-mesenchymal transition, and the proliferation of polyclonal fibroblast clusters, leading to spatially and temporally heterogenous tissue remodeling (56-58). In the current paradigm of PF, the role of immune cells remains ambiguous and primarily circumscribed to "acute inflammatory exacerbations", sporadic events that severely worsen disease phenotype and accelerate patient mortality. This paradigm is supported by clinical and experimental evidence linking excess inflammatory monocyte mobilization and their retention as profibrotic monocyte-derived macrophages to poor prognosis (13, 14, 20, 22). Our group has previously defined the involvement of mononuclear myeloid cells at all stages of inflammatory exacerbations triggered by a clinically relevant single point mutation in the alveolar epithelial cell specific gene encoding for the surfactant protein C (SP-C<sup>I73T</sup>) (59). Using this experimental platform and available human IPF datasets, this work aimed to

expand the current understanding of PF and fill an important gap pertaining to macrophage phenotypic heterogeneity and aberrant cell-cell crosstalk in fibrotic disease. Specifically, we spatially characterized the transcriptional profile of  $CD68^+$  macrophages within and surrounding fibrotic foci of injury; defined incoming and outgoing pro-fibrotic communications among pulmonary cells; identified a dynamic lipid signature across all annotated macrophage and monocyte clusters, with a sole population producing *Apoe*; and showed that SP-C<sup>I73T</sup> induced fibrosis is marginally impacted by genetic *Apoe* deletion.

Initial work was designed to spatially localize activated macrophages in the healthy, inflamed, and fibrosing lung. This approach aligns with a handful of recent reports that elegantly resolve epithelial, mesenchymal, and immune cell identity and abundance in clinical and experimental fibrosis (60, 61). Our studies go beyond the focus of these publications by placing emphasis on the distribution of macrophages from spatially diverse regions of the injured lung. CD68 represents a compromise allowing to sample both control/healthy and injured macrophages. Our gene expression and hierarchical clustering analysis demonstrate heterogenous transcriptional signatures among macrophages sampled from the fibrosing lung, thus suggesting a gradient of activation based on spatial localization. Pathway analysis also painted an unexpected picture characterized by transcriptionally quiescent CD68<sup>+</sup> cells sampled from fully remodeled regions, contrasting a hyperactive phenotype emanating from peri-injured macrophages. Within this region of interest our findings also support the notion that macrophages activation is linked to their proximity to the injury. Findings that innate and adaptive immunity, matrix remodeling, senescence, and redox balance signals are predicted to be driven primarily by periinjured macrophages (compared to all other sampled regions) adds depth to previous reports showing unequal expression of inflammatory proteins among immune cells found in healthy and fibrotic regions of the clinically diseased lung (62). Our results also support and complement evidence showing downregulation of inflammatory signaling (TNFa) in IPF immune infiltrates compared to healthy/unaffected regions (63). The comprehensive nature of our transcriptional analysis is novel in the context of fibrosis, particularly in pulmonary injury triggered by genetic susceptibility of the epithelial compartment and could provide insights in the development of more targeted and effective therapeutics in the future.

To further appreciate the role of the resident alveolar compartment during inflammatory exacerbations progressing to fibrosis, flow cytometry-based sorting of CD11b<sup>-</sup>SigF<sup>+</sup>CD11c<sup>+</sup> cells was performed 3 days and 14 days after SP-C<sup>I73T</sup> induced injury. Our results allowed us to place these cells as players of all phases of the SP-C<sup>I73T</sup> injury, through temporally restricted activation of canonical inflammatory pathways (IL6 and prostaglandin E2 exclusively shown 3 days post injury) and pro-fibrotic pathways (glycoprotein/GP6, SOX2, KLF4). At the same time, we show a subset of signaling networks sustained over our 14 day analysis, albeit driven by distinct transcriptional signatures (IFN $\gamma$ , TGF $\beta$ 1, IL-4) (64, 65). While these signals have been previously reported in the fibrotic lung, specific annotation in resident alveolar macrophages strengthens the notion that this population functions as a pleiotropic effector of inflammation and tissue remodeling (66, 67), and further substantiates our prior histopathological investigation of the SP-C<sup>I73T</sup> injury (dysregulated inflammation and increased mortality) following pharmacological depletion of alveolar macrophages with clodronate liposomes (10).

Due to the phenotypic heterogeneity intrinsic of monocytes/ macrophages in chronic disease, we built on our previous characterization of monocyte-derived macrophages through single-cell RNA sequencing (10, 22, 68). Albeit a low-resolution clustering analysis, we identified three macrophage (alveolar macrophages,  $Trem2^+$ , and  $Mertk^+$ ), as well as transcriptionally distinct populations of monocytes (classical - Itgam/cd11b, Cx3cr1; inflammatory - Ccr2, Ly6c, Ly22) (69). Our findings that complement signaling is involved in  $\ensuremath{\text{SP-C}^{\text{I73T}}}$  injury and fibrosis is consistent with evidence that soluble defense collagens support activation, proliferation, and tissue-repair functions of macrophages (70-72). Alongside these signals, our results define inflammatory monocytes and monocyte-derived moieties (interstitial macrophages) as centrally involved in fibronectin/FN1 and osteopontin/SPP1 signaling in the fibrosing lung, a finding consistent with bleomycin-induced injury (42, 73-78).

Mertk has been used to identify alveolar macrophages due to its role in mediating phagocytosis of apoptotic cells (79). Our annotation of a *Mertk<sup>int</sup>* cluster in control lungs is consistent to their identity as alveolar macrophages. However, identification of a second population of *Mertk<sup>high</sup>* cells appearing 42 days post injury and displaying a unique transcriptional signatures suggest these may be monocyte-derived moieties settling within the lung following the end of the inflammatory exacerbation. Further work needs to establish their identity since our pseudotime analysis did not fully recapitulate their origin.

Mining of a publicly available human IPF dataset (GSE136831, www.ipfcellatlas.com) offered the opportunity to add a translational value to these findings (40). Our analysis identified three macrophage clusters in the IPF lung that generate a transcriptional signature comparable to our murine results (innate immunity, interleukin mediated signaling, fibrogenic processes, and heightened transcriptional and translational control). In particular, this work confirmed the presence of TREM2<sup>+</sup> interstitial macrophages (cluster 0) and a population of alveolar macrophages (cluster 1) in late stage IPF (22, 80, 81). Due to the low resolution of the clustering, our findings do show a single cluster co-expressing MERTK and TREM2. At this stage, it is unclear if such population appears in established fibrosis as clinical literature seldom cites both markers. Independent of the nomenclature, our assessment is consistent with reports showing osteopontin/SPP1, fibronectin/FN1, and TGFB1 signaling in the fibrotic niche (73, 82-84).

The importance of defining the metabolic signatures accompanying acute and chronic inflammation has great therapeutic potential (85). Our analysis identifies a robust lipid signature in steady-state alveolar macrophages, a notion consistent with their role in surfactant lipid (fluid) recycling and maintenance (86). The changes observed during SP-C<sup>173T</sup> injury are consistent with the notion that pro-fibrotic reprogramming requires a metabolic shift towards lipid consumption (87, 88).

Therapeutically noteworthy is the transcriptional signature produced by Mertkhigh macrophages 42 days post injury, which included Pparg and several apolipoprotein receptors. Indeed, there is a large body of evidence showing that engaging this transcription factor is effective in attenuating the fibrotic phenotype (89–91). By comparison, this work leveraged the specificity of the Apoe signal originating from Trem2<sup>+</sup> cells and recent experimental evidence linking this lipoprotein to pro-fibrotic monocyte-derived macrophages during chemical-induced fibrosis (36, 92-94). Our findings that ApoE deletion does not provide overt benefits on *Spp1*<sup>+</sup> cell accumulation in the foci of injury, total inflammation, or survival is somewhat surprising. Confounding elements related to the compensatory effects of a global knock out, or the impact of ApoE deletion on a surfactant impaired system may need further examination. Despite these incongruences, our results support the value of ApoE as a biomarker indicative of the presence of Spp1<sup>+</sup> and  $Fn1^+$  pro-fibrotic macrophages in the lung.

Though comprehensive, any experimental modeling of disease has limitations. While pairing murine sequencing data with human singlecell datasets offers translational value, elements related to disease staging (early inflammatory exacerbation vs. end-stage disease) and heterogeneity of human IPF etiology make this assessment less obvious. Furthermore, it is well established that use of antibody-based sorting introduces bias to the analysis (e.g., CD11b<sup>-</sup>SigF<sup>+</sup>CD11c<sup>+</sup> in our bulk sequencing, CD45<sup>+</sup>CD68<sup>+</sup> for spatial analysis). To a lesser degree, clustering analysis of single-cell RNA sequencing data and cellcommunication predictive tools introduce bias related to pathway annotation, and therefore any analysis attempting to describe noncanonical signaling (e.g. ApoE signaling in inflammation rather than lipid homeostasis) may not find a fitting match.

Despite any potential drawback, this work comprehensively assesses the spatial and phenotypic distribution of macrophages in pulmonary fibrosis triggered by a fibrogenic mutation in the alveolar epithelial cell restricted gene encoding for the SP-C. Our data finds peri-injury macrophages to produce an extremely active phenotype, while CD68<sup>+</sup> cells localized within the fibrotic foci appear transcriptionally dormant at the peak of an inflammation exacerbation. Single cell analysis elucidated the intercellular communications occurring in the lung, while identifying Ccr2<sup>+</sup>Ly6c<sup>+</sup> inflammatory monocytes and trem2+ interstitial macrophages as distinct fibrogenic populations in SP-C<sup>I73T</sup> induced injury. Furthermore, our work defined distinct lipid signatures among macrophage populations and propose ApoE as a potential biomarker to identify SPP1- and FN1-producing macrophages. Taken together, this work provides an essential framework for the identification (and future targeting) of deleterious macrophage populations in the early and late stages of the fibrogenic process.

#### 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 in the article/Supplementary Material.

### **Ethics statement**

The animal study was approved by the Institutional Animal Care and Use Committee, University of Utah. The study was conducted in accordance with the local legislation and institutional requirements.

## Author contributions

PM: Data curation, Formal analysis, Methodology, Resources, Software, Visualization, Writing – original draft, Writing – review & editing. JC: Conceptualization, Data curation, Methodology, Validation, Writing – review & editing. SC: Data curation, Formal analysis, Methodology, Software, Visualization, Writing – review & editing. JN: Data curation, Formal analysis, Methodology, Visualization, Writing – review & editing. QW: Data curation, Methodology, Visualization, Writing – review & editing. TM: Formal analysis, Methodology, Visualization, Writing – review & editing. AV: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft, Writing – review & editing.

## Funding

The author(s) declare financial support was received for the research, authorship, and/or publication of this article. Grant Support: NIEHS R01ES032553 (AV); ALSAM Foundation Research Grant (AV).

## **Conflict of interest**

The authors declare the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

## Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

## Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fimmu.2024.1425466/ full#supplementary-material

## References

1. Cottin V, Hirani NA, Hotchkin DL, Nambiar AM, Ogura T, Otaola M, et al. Presentation, diagnosis and clinical course of the spectrum of progressive-fibrosing interstitial lung diseases. *Eur Respir Rev.* (2018) 27:180076. doi: 10.1183/ 16000617.0076-2018

2. Wijsenbeek M, Cottin V. Spectrum of fibrotic lung diseases. N Engl J Med. (2020) 383:958–68. doi: 10.1056/NEJMra2005230

3. Conti S, Harari S, Caminati A, Zanobetti A, Schwartz JD, Bertazzi PA, et al. The association between air pollution and the incidence of idiopathic pulmonary fibrosis in Northern Italy. *Eur Respir J.* (2018) 51(1):1700397. doi: 10.1183/13993003.00397-2017

4. Winterbottom CJ, Shah RJ, Patterson KC, Kreider ME, Panettieri RA Jr., Rivera-Lebron B, et al. Exposure to ambient particulate matter Is associated with accelerated functional decline in idiopathic pulmonary fibrosis. *Chest.* (2018) 153:1221–8. doi: 10.1016/j.chest.2017.07.034

5. Spagnolo P, Kropski JA, Jones MG, Lee JS, Rossi G, Karampitsakos T, et al. Idiopathic pulmonary fibrosis: Disease mechanisms and drug development. *Pharmacol Ther.* (2021) 222:107798. doi: 10.1016/j.pharmthera.2020.107798

6. Cui F, Sun Y, Xie J, Li D, Wu M, Song L, et al. Air pollutants, genetic susceptibility and risk of incident idiopathic pulmonary fibrosis. *Eur Respir J*. (2023) 61(2):2200777. doi: 10.1183/13993003.00777-2022

7. Sesé L, Harari S. Now we know: chronic exposure to air pollutants is a risk factor for the development of idiopathic pulmonary fibrosis. *Eur Respir J.* (2023) 61 (2):2202113. doi: 10.1183/13993003.02113-2022

8. Mulugeta S, Nureki S, Beers MF. Lost after translation: insights from pulmonary surfactant for understanding the role of alveolar epithelial dysfunction and cellular quality control in fibrotic lung disease. *Am J Physiol Lung Cell Mol Physiol.* (2015) 309: L507–25. doi: 10.1152/ajplung.00139.2015

9. Venosa A, Katzen J, Tomer Y, Kopp M, Jamil S, Russo SJ, et al. Epithelial expression of an interstitial lung disease-associated mutation in surfactant protein-C modulates recruitment and activation of key myeloid cell populations in mice. *J Immunol.* (2019) 202:2760–71. doi: 10.4049/jimmunol.1900039

10. Venosa A, Cowman S, Katzen J, Tomer Y, Armstrong BS, Mulugeta S, et al. Role of CCR2(+) myeloid cells in inflammation responses driven by expression of a surfactant protein-C mutant in the alveolar epithelium. *Front Immunol.* (2021) 12:665818–8. doi: 10.3389/fimmu.2021.665818

11. Hawkins A, Guttentag SH, Deterding R, Funkhouser WK, Goralski JL, Chatterjee S, et al. A non-BRICHOS SFTPC mutant (SP-C(I73T)) linked to interstitial lung disease promotes a late block in macroautophagy disrupting cellular proteostasis and mitophagy. *Am J Physiol Lung Cell Mol Physiol.* (2015) 308:L33–47. doi: 10.1152/ajplung.00217.2014

12. Katzen J, Wagner BD, Venosa A, Kopp M, Tomer Y, Russo SJ, et al. An SFTPC BRICHOS mutant links epithelial ER stress and spontaneous lung fibrosis. *JCI Insight*. (2019) 4(6):e126125. doi: 10.1172/jci.insight.126125

13. Teoh AKY, Jo HE, Chambers DC, Symons K, Walters EH, Goh NS, et al. Blood monocyte counts as a potential prognostic marker for idiopathic pulmonary fibrosis: analysis from the Australian IPF registry. *Eur Respir J.* (2020) 55(4):1901855. doi: 10.1183/13993003.01855-2019

14. Kreuter M, Lee JS, Tzouvelekis A, Oldham JM, Molyneaux PL, Weycker D, et al. Monocyte count as a prognostic biomarker in patients with idiopathic pulmonary fibrosis. *Am J Respir Crit Care Med.* (2021) 204:74–81. doi: 10.1164/rccm.202003-0669OC

15. Gibson CD, Kugler MC, Deshwal H, Munger JS, Condos R. Advances in targeted therapy for progressive fibrosing interstitial lung disease. *Lung.* (2020) 198:597–608. doi: 10.1007/s00408-020-00370-1

16. Janssen WJ, Barthel L, Muldrow A, Oberley-Deegan RE, Kearns MT, Jakubzick C, et al. Fas determines differential fates of resident and recruited macrophages during resolution of acute lung injury. *Am J Respir Crit Care Med.* (2011) 184:547–60. doi: 10.1164/rccm.201011-1891OC

17. Gibbings SL, Goyal R, Desch AN, Leach SM, Prabagar M, Atif SM, et al. Transcriptome analysis highlights the conserved difference between embryonic and postnatal-derived alveolar macrophages. *Blood.* (2015) 126:1357–66. doi: 10.1182/blood-2015-01-624809

18. Tan SY, Krasnow MA. Developmental origin of lung macrophage diversity. Development. (2016) 143:1318-27. doi: 10.1242/dev.129122

 Reyfman PA, Walter JM, Joshi N, Anekalla KR, Mcquattie-Pimentel AC, Chiu S, et al. Single-cell transcriptomic analysis of human lung provides insights into the pathobiology of pulmonary fibrosis. *Am J Respir Crit Care Med.* (2019) 199:1517–36. doi: 10.1164/rccm.201712-2410OC

20. Joshi N, Watanabe S, Verma R, Jablonski RP, Chen C-I, Cheresh P, et al. A spatially restricted fibrotic niche in pulmonary fibrosis is sustained by M-CSF/M-CSFR signalling in monocyte-derived alveolar macrophages. *Eur Respir J.* (2020) 55 (1):1900646. doi: 10.1183/13993003.00646-2019

21. Young LR, Gulleman PM, Short CW, Tanjore H, Sherrill T, Qi A, et al. Epithelial-macrophage interactions determine pulmonary fibrosis susceptibility in Hermansky-Pudlak syndrome. *JCI Insight*. (2016) 1:e88947. doi: 10.1172/jci.insight.88947

22. Misharin AV, Morales-Nebreda L, Reyfman PA, Cuda CM, Walter JM, Mcquattie-Pimentel AC, et al. Monocyte-derived alveolar macrophages drive lung fibrosis and persist in the lung over the life span. *J Exp Med.* (2017) 214:2387–404. doi: 10.1084/jem.20162152

23. Takahashi F, Takahashi K, Okazaki T, Maeda K, Ienaga H, Maeda M, et al. Role of osteopontin in the pathogenesis of bleomycin-induced pulmonary fibrosis. *Am J Respir Cell Mol Biol.* (2001) 24:264–71. doi: 10.1165/ajrcmb.24.3.4293

24. Dong J, Ma Q. Osteopontin enhances multi-walled carbon nanotube-triggered lung fibrosis by promoting TGF- $\beta$ 1 activation and myofibroblast differentiation. *Particle Fibre Toxicol.* (2017) 14:18. doi: 10.1186/s12989-017-0198-0

25. Upagupta C, Shimbori C, Alsilmi R, Kolb M. Matrix abnormalities in pulmonary fibrosis. *Eur Respir Rev.* (2018) 27:180033. doi: 10.1183/16000617.0033-2018

26. Gao X, Jia G, Guttman A, Depianto DJ, Morshead KB, Sun K-H, et al. Osteopontin links myeloid activation and disease progression in systemic sclerosis. *Cell Rep Med.* (2020) 1:100140. doi: 10.1016/j.xcrm.2020.100140

27. Hatipoglu OF, Uctepe E, Opoku G, Wake H, Ikemura K, Ohtsuki T, et al. Osteopontin silencing attenuates bleomycin-induced murine pulmonary fibrosis by regulating epithelial-mesenchymal transition. *Biomedicine Pharmacotherapy.* (2021) 139:111633. doi: 10.1016/j.biopha.2021.111633

28. Conroy LR, Clarke HA, Allison DB, Valenca SS, Sun Q, Hawkinson TR, et al. Spatial metabolomics reveals glycogen as an actionable target for pulmonary fibrosis. *Nat Commun.* (2023) 14(1):2759. doi: 10.1038/s41467-023-38437-1

29. Rajesh R, Atallah R, Bärnthaler T. Dysregulation of metabolic pathways in pulmonary fibrosis. *Pharmacol Ther.* (2023) 246:108436. doi: 10.1016/ j.pharmthera.2023.108436

30. Kelly B, O'neill LA. Metabolic reprogramming in macrophages and dendritic cells in innate immunity. *Cell Res.* (2015) 25:771–84. doi: 10.1038/cr.2015.68

31. Liu Y, Xu R, Gu H, Zhang E, Qu J, Cao W, et al. Metabolic reprogramming in macrophage responses. *biomark Res.* (2021) 9:1. doi: 10.1186/s40364-020-00251-y

32. Wang S, Liu G, Li Y, Pan Y. Metabolic reprogramming induces macrophage polarization in the tumor microenvironment. *Front Immunol.* (2022) 13:840029. doi: 10.3389/fimmu.2022.840029

33. Dorotea D, Koya D, Ha H. Recent insights into SREBP as a direct mediator of kidney fibrosis via lipid-independent pathways. *Front Pharmacol.* (2020) 11. doi: 10.3389/fphar.2020.00265

34. Jaroonwitchawan T, Arimochi H, Sasaki Y, Ishifune C, Kondo H, Otsuka K, et al. Stimulation of the farnesoid X receptor promotes M2 macrophage polarization. *Front Immunol.* (2023) 14. doi: 10.3389/fimmu.2023.1065790

35. Baitsch D, Bock HH, Engel T, Telgmann R, Muller-Tidow C, Varga G, et al. Apolipoprotein E induces antiinflammatory phenotype in macrophages. *Arterioscler Thromb Vasc Biol.* (2011) 31:1160–8. doi: 10.1161/ATVBAHA.111.222745

36. Cui H, Jiang D, Banerjee S, Xie N, Kulkarni T, Liu RM, et al. Monocyte-derived alveolar macrophage apolipoprotein E participates in pulmonary fibrosis resolution. *JCI Insight.* (2020) 5(5):e134539. doi: 10.1172/jci.insight.134539

37. Venosa A, Malaviya R, Gow AJ, Hall L, Laskin JD, Laskin DL. Protective role of spleen-derived macrophages in lung inflammation, injury, and fibrosis induced by nitrogen mustard. *Am J Physiol Lung Cell Mol Physiol.* (2015) 309:L1487–1498. doi: 10.1152/ajplung.00276.2015

38. R Core Team. R: A language and environment for statistical computing. *R Foundation for Statistical Computing*. Vienna, Austria (2013). Available at: https://www.R-project.org/.

39. Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. *Genome Biol.* (2014) 15:550. doi: 10.1186/s13059-014-0550-8

40. Adams TS, Schupp JC, Poli S, Ayaub EA, Neumark N, Ahangari F, et al. Singlecell RNA-seq reveals ectopic and aberrant lung-resident cell populations in idiopathic pulmonary fibrosis. *Sci Adv.* (2020) 6:eaba1983. doi: 10.1126/sciadv.aba1983

41. Zappia L, Oshlack A. Clustering trees: a visualization for evaluating clusterings at multiple resolutions. *Gigascience*. (2018) 7(7):giy083. doi: 10.1093/gigascience/giy083

42. Aran D, Looney AP, Liu L, Wu E, Fong V, Hsu A, et al. Reference-based analysis of lung single-cell sequencing reveals a transitional profibrotic macrophage. *Nat Immunol.* (2019) 20:163–72. doi: 10.1038/s41590-018-0276-y

43. Hafemeister C, Satija R. Normalization and variance stabilization of single-cell RNA-seq data using regularized negative binomial regression. *Genome Biol.* (2019) 20:296. doi: 10.1186/s13059-019-1874-1

44. Ahlmann-Eltze C, Huber W. glmGamPoi: fitting Gamma-Poisson generalized linear models on single cell count data. *Bioinformatics*. (2021) 36:5701–2. doi: 10.1093/ bioinformatics/btaa1009

45. Hao Y, Hao S, Andersen-Nissen E, Mauck WM 3rd, Zheng S, Butler A, et al. Integrated analysis of multimodal single-cell data. *Cell.* (2021) 184:3573–3587.e3529. doi: 10.1016/j.cell.2021.04.048

46. Kuleshov MV, Jones MR, Rouillard AD, Fernandez NF, Duan Q, Wang Z, et al. Enrichr: a comprehensive gene set enrichment analysis web server 2016 update. *Nucleic Acids Res.* (2016) 44:W90–97. doi: 10.1093/nar/gkw377 47. Trapnell C, Cacchiarelli D, Grimsby J, Pokharel P, Li S, Morse M, et al. The dynamics and regulators of cell fate decisions are revealed by pseudotemporal ordering of single cells. *Nat Biotechnol.* (2014) 32:381–6. doi: 10.1038/nbt.2859

48. Qiu X, Mao Q, Tang Y, Wang L, Chawla R, Pliner HA, et al. Reversed graph embedding resolves complex single-cell trajectories. *Nat Methods*. (2017) 14:979–82. doi: 10.1038/nmeth.4402

49. Cao J, Spielmann M, Qiu X, Huang X, Ibrahim DM, Hill AJ, et al. The single-cell transcriptional landscape of mammalian organogenesis. *Nature*. (2019) 566:496–502. doi: 10.1038/s41586-019-0969-x

50. Dimitrov D, Türei D, Garrido-Rodriguez M, Burmedi PL, Nagai JS, Boys C, et al. Comparison of methods and resources for cell-cell communication inference from single-cell RNA-Seq data. *Nat Commun.* (2022) 13(1):3224. doi: 10.1038/s41467-022-30755-0

51. Dimitrov D, Schäfer PSL, Farr E, Mier PR, Lobentanzer S, Dugourd A, et al. LIANA+: an all-in-one cell-cell communication framework. *bioRxiv*. (2023) 2023.08.19.553863. doi: 10.1101/2023.08.19.553863

52. Tu J-J, Li H-S, Yan H, Zhang X-F. EnDecon: cell type deconvolution of spatially resolved transcriptomics data via ensemble learning. *Bioinformatics*. (2023) 39:btac825. doi: 10.1093/bioinformatics/btac825

53. Jin S, Guerrero-Juarez CF, Zhang L, Chang I, Ramos R, Kuan C-H, et al. Inference and analysis of cell-cell communication using CellChat. *Nat Commun.* (2021) 12(1):1088. doi: 10.1038/s41467-021-21246-9

54. Lane-Donovan C, Herz J. ApoE, apoE receptors, and the synapse in alzheimer's disease. *Trends Endocrinol Metab.* (2017) 28:273–84. doi: 10.1016/j.tem.2016.12.001

55. Kropski JA, Blackwell TS, Loyd JE. The genetic basis of idiopathic pulmonary fibrosis. *Eur Respir J.* (2015) 45:1717–27. doi: 10.1183/09031936.00163814

56. Cook DN, Brass DM, Schwartz DA. A matrix for new ideas in pulmonary fibrosis. *Am J Respir Cell Mol Biol.* (2002) 27:122–4. doi: 10.1165/ajrcmb.27.2.f245

57. Hardie WD, Hagood JS, Dave V, Perl AK, Whitsett JA, Korfhagen TR, et al. Signaling pathways in the epithelial origins of pulmonary fibrosis. *Cell Cycle*. (2010) 9:2769–76. doi: 10.4161/cc.9.14.12268

58. Winters NI, Burman A, Kropski JA, Blackwell TS. Epithelial injury and dysfunction in the pathogenesis of idiopathic pulmonaryFibrosis. *Am J Med Sci.* (2019) 357:374–8. doi: 10.1016/j.amjms.2019.01.010

59. Nureki SI, Tomer Y, Venosa A, Katzen J, Russo SJ, Jamil S, et al. Expression of mutant Sftpc in murine alveolar epithelia drives spontaneous lung fibrosis. *J Clin Invest.* (2018) 128:4008–24. doi: 10.1172/JCI99287

60. Ciccimarra R, Bolognesi MM, Zoboli M, Cattoretti G, Stellari FF, Ravanetti F. The normal and fibrotic mouse lung classified by spatial proteomic analysis. *Sci Rep.* (2022) 12(1):8742. doi: 10.1038/s41598-022-12738-9

61. Blumhagen RZ, Kurche JS, Cool CD, Walts AD, Heinz D, Fingerlin TE, et al. Spatially distinct molecular patterns of gene expression in idiopathic pulmonary fibrosis. *Respir Res.* (2023) 24:287. doi: 10.1186/s12931-023-02572-6

62. Nuovo GJ, Hagood JS, Magro CM, Chin N, Kapil R, Davis L, et al. The distribution of immunomodulatory cells in the lungs of patients with idiopathic pulmonary fibrosis. *Modern Pathol.* (2012) 25:416-33. doi: 10.1038/modpathol.2011.166

63. Eyres M, Bell JA, Davies ER, Fabre A, Alzetani A, Jogai S, et al. Spatially resolved deconvolution of the fibrotic niche in lung fibrosis. *Cell Rep.* (2022) 40:111230. doi: 10.1016/j.celrep.2022.111230

64. Wu Q, Zhang K-J, Jiang S-M, Fu L, Shi Y, Tan R-B, et al. p53: A key protein that regulates pulmonary fibrosis. *Oxid Med Cell Longevity*. (2020) 2020:6635794. doi: 10.1155/2020/6635794

65. Chandran RR, Xie Y, Gallardo-Vara E, Adams T, Garcia-Milian R, Kabir I, et al. Distinct roles of KLF4 in mesenchymal cell subtypes during lung fibrogenesis. *Nat Commun.* (2021) 12(1):7179. doi: 10.1038/s41467-021-27499-8

66. Lee C-M, Cho SJ, Cho W-K, Park JW, Lee J-H, Choi AM, et al. Laminin  $\alpha l$  is a genetic modifier of TGF- $\beta$ l-stimulated pulmonary fibrosis. *JCI Insight*. (2018) 3(18): e99574. doi: 10.1172/jci.insight.99574

67. Uhl K, Paithankar S, Leshchiner D, Jager TE, Abdelgied M, Dixit B, et al. Differential transcriptomic signatures of small airway cell cultures derived from IPF and COVID-19-induced exacerbation of interstitial lung disease. *Cells.* (2023) 12 (20):2501. doi: 10.3390/cells12202501

68. Nguyen J, Deering-Rice CE, Armstrong BS, Massa C, Reilly CA, Venosa A. Parenchymal and inflammatory cell responses to single and repeated ozone exposure in healthy and surfactant protein-C mutant lung. *Toxicol Sci.* (2022) 189:107–23. doi: 10.1093/toxsci/kfac074

69. Trzebanski S, Jung S. Plasticity of monocyte development and monocyte fates. Immunol Lett. (2020) 227:66–78. doi: 10.1016/j.imlet.2020.07.007

70. Minutti CM, Jackson-Jones LH, García-Fojeda B, Knipper JA, Sutherland TE, Logan N, et al. Local amplifiers of IL-4 $R\alpha$ -mediated macrophage activation promote repair in lung and liver. *Science*. (2017) 356:1076–80. doi: 10.1126/science.aaj2067

71. Casals C, García-Fojeda B, Minutti CM. Soluble defense collagens: Sweeping up immune threats. *Mol Immunol.* (2019) 112:291–304. doi: 10.1016/j.molimm.2019. 06.007

72. Sikkeland LIB, Ueland T, Lund MB, Durheim MT, Mollnes TE. A role for the terminal C5-C9 complement pathway in idiopathic pulmonary fibrosis. *Front Med.* (2023) 10. doi: 10.3389/fmed.2023.1236495

73. Morse C, Tabib T, Sembrat J, Buschur KL, Bittar HT, Valenzi E, et al. Proliferating SPP1/MERTK-expressing macrophages in idiopathic pulmonary fibrosis. *Eur Respir J*. (2019) 54(2):1802441. doi: 10.1183/13993003.02441-2018

74. Matsubara E, Komohara Y, Esumi S, Shinchi Y, Ishizuka S, Mito R, et al. SPP1 derived from macrophages is associated with a worse clinical course and chemo-resistance in lung adenocarcinoma. *Cancers (Basel)*. (2022) 14(18):4374. doi: 10.3390/ cancers14184374

75. Wang J, Zhang L, Luo L, He P, Xiong A, Jiang M, et al. Characterizing cellular heterogeneity in fibrotic hypersensitivity pneumonitis by single-cell transcriptional analysis. *Cell Death Discovery*. (2022) 8:38. doi: 10.1038/s41420-022-00831-x

76. Han H, Ge X, Komakula SSB, Desert R, Das S, Song Z, et al. Macrophage-derived osteopontin (SPP1) protects from nonalcoholic steatohepatitis. *Gastroenterology*. (2023) 165:201–17. doi: 10.1053/j.gastro.2023.03.228

77. Hoeft K, Schaefer GJL, Kim H, Schumacher D, Bleckwehl T, Long Q, et al. Platelet-instructed SPP1(+) macrophages drive myofibroblast activation in fibrosis in a CXCL4-dependent manner. *Cell Rep.* (2023) 42:112131. doi: 10.1016/j.celrep.2023.112131

78. Moore PK, Anderson KC, Mcmanus SA, Tu TH, King EM, Mould KJ, et al. Single-cell RNA sequencing reveals unique monocyte-derived interstitial macrophage subsets during lipopolysaccharide-induced acute lung inflammation. *Am J Physiol Lung Cell Mol Physiol.* (2023) 324:L536–L549. doi: 10.1152/ajplung.00223.2022

79. Chakraborty S, Singh A, Wang L, Wang X, Sanborn MA, Ye Z, et al. Trained immunity of alveolar macrophages enhances injury resolution via KLF4-MERTK-mediated efferocytosis. *J Exp Med.* (2023) 220(11):e20221388. doi: 10.1084/jem.20221388

80. Kotlyarov S. Participation of ABCA1 transporter in pathogenesis of chronic obstructive pulmonary disease. *Int J Mol Sci.* (2021) 22(7):3334. doi: 10.3390/ ijms22073334

81. Yang HQ, Sun H, Li K, Shao MM, Zhai K, Tong ZH. Dynamics of host immune responses and a potential function of Trem2(hi) interstitial macrophages in Pneumocystis pneumonia. *Respir Res.* (2024) 25:72. doi: 10.1186/s12931-024-02709-1

82. Cai B, Dongiovanni P, Corey KE, Wang X, Shmarakov IO, Zheng Z, et al. Macrophage merTK promotes liver fibrosis in nonalcoholic steatohepatitis. *Cell Metab.* (2020) 31:406–421.e407. doi: 10.1016/j.cmet.2019.11.013

83. Lv J, Gao H, Ma J, Liu J, Tian Y, Yang C, et al. Dynamic atlas of immune cells reveals multiple functional features of macrophages associated with progression of pulmonary fibrosis. *Front Immunol.* (2023) 14. doi: 10.3389/fimmu.2023.1230266

84. She Y, Xu X, Yu Q, Yang X, He J, Tang XX. Elevated expression of macrophage MERTK exhibits profibrotic effects and results in defective regulation of efferocytosis function in pulmonary fibrosis. *Respir Res.* (2023) 24:118. doi: 10.1186/s12931-023-02424-3

85. Li J, Zhai X, Sun X, Cao S, Yuan Q, Wang J. Metabolic reprogramming of pulmonary fibrosis. Front Pharmacol. (2022) 13:1031890. doi: 10.3389/fphar.2022.1031890

86. Poelma DL, Ju MR, Bakker SC, Zimmermann LJ, Lachmann BF, Van Iwaarden JF. A common pathway for the uptake of surfactant lipids by alveolar cells. *Am J Respir Cell Mol Biol.* (2004) 30:751–8. doi: 10.1165/rcmb.2003-0127OC

87. Wculek SK, Dunphy G, Heras-Murillo I, Mastrangelo A, Sancho D. Metabolism of tissue macrophages in homeostasis and pathology. *Cell Mol Immunol.* (2022) 19:384–408. doi: 10.1038/s41423-021-00791-9

88. Vassiliou E, Farias-Pereira R. Impact of lipid metabolism on macrophage polarization: implications for inflammation and tumor immunity. *Int J Mol Sci.* (2023) 24(15):12032. doi: 10.3390/ijms241512032

89. Chandran S, Schilke RM, Blackburn CMR, Yurochko A, Mirza R, Scott RS, et al. Lipin-1 contributes to IL-4 mediated macrophage polarization. *Front Immunol.* (2020) 11. doi: 10.3389/fimmu.2020.00787

90. Zhou F, Fan X, Miao Y. LPIN1 promotes triglycerides synthesis and is transcriptionally regulated by PPARG in buffalo mammary epithelial cells. *Sci Rep.* (2022) 12(1):2390. doi: 10.1038/s41598-022-06114-w

91. Zeng Q, Zhou T-T, Huang W-J, Huang X-T, Huang L, Zhang X-H, et al. Asarinin attenuates bleomycin-induced pulmonary fibrosis by activating PPARy. *Sci Rep.* (2023) 13:14706. doi: 10.1038/s41598-023-41933-5

92. Venosa A, Malaviya R, Choi H, Gow AJ, Laskin JD, Laskin DL. Characterization of distinct macrophage subpopulations during nitrogen mustard-induced lung injury and fibrosis. *Am J Respir Cell Mol Biol.* (2016) 54:436–46. doi: 10.1165/rcmb.2015-0120OC

93. Gordon EM, Yao X, Xu H, Karkowsky W, Kaler M, Kalchiem-Dekel O, et al. Apolipoprotein E is a concentration-dependent pulmonary danger signal that activates the NLRP3 inflammasome and IL-1 $\beta$  secretion by bronchoalveolar fluid macrophages from asthmatic subjects. *J Allergy Clin Immunol.* (2019) 144:426–441.e423. doi: 10.1016/j.jaci.2019.02.027

94. Venosa A, Smith LC, Murray A, Banota T, Gow AJ, Laskin JD, et al. Regulation of macrophage foam cell formation during nitrogen mustard (NM)-induced pulmonary fibrosis by lung lipids. *Toxicol Sci.* (2019) 172:344–58. doi: 10.1093/toxsci/kfz187