Single-cell transcriptomics identifies pathogenic T-helper 17.1 cells and pro-inflammatory monocytes in immune checkpoint inhibitor-related pneumonitis

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ABSTRACT

Background Immune checkpoint inhibitor (ICI)-related pneumonitis is the most frequent fatal immune-related adverse event associated with programmed cell death protein-1/programmed death ligand-1 blockade. The pathophysiology however remains largely unknown, owing to limited and contradictory findings in existing literature pointing at either T-helper 1 or T-helper 17-mediated autoimmunity. In this study, we aimed to gain novel insights into the mechanisms of ICI-related pneumonitis, thereby identifying potential therapeutic targets.

Methods In this prospective observational study, single-cell RNA and T-cell receptor sequencing was performed on bronchoalveolar lavage fluid of 11 patients with ICI-related pneumonitis and 6 demographically-matched patients with cancer without ICI-related pneumonitis. Single-cell transcriptomic immunophenotyping and cell fate mapping coupled to T-cell receptor repertoire analyses were performed.

Results We observed enrichment of both CD4+ and CD8+ T cells in ICI pneumonitis bronchoalveolar lavage fluid. The CD4+ T-cell compartment showed an increase of pathogenic T-helper 17.1 cells, characterized by high co-expression of TBX21 (encoding T-bet) and RORC (ROR-γ), IFN-γ (IFN-γ), IL-17A, CSF2 (GM-CSF), and cytotoxicity genes. Type 1 regulatory T cells and naive-like CD4+ T cells were also enriched. Within the CD8+ T-cell compartment, mainly effector memory T cells were increased. Correspondingly, myeloid cells in ICI pneumonitis bronchoalveolar lavage fluid were relatively depleted of anti-inflammatory resident alveolar macrophages while pro-inflammatory ‘M1-like’ monocytes (expressing TNF, IL-1β, IL-6, IL-23A, and GM-CSF receptor CSF2RA, CSF2RB) were enriched compared with control samples. Importantly, a feedforward loop, in which GM-CSF production by pathogenic T-helper 17.1 cells promotes tissue inflammation and IL-23 production by pro-inflammatory monocytes and vice versa, has been well characterized in multiple autoimmune disorders but has never been identified in ICI-related pneumonitis.

WHAT IS ALREADY KNOWN ON THIS TOPIC

⇒ Immune checkpoint inhibitor (ICI)-pneumonitis is a frequent serious adverse event of cancer immunotherapy hinging on a cell-mediated immune response, though the exact pathophysiology is currently unknown.

WHAT THIS STUDY ADDS

⇒ Single-cell transcriptomics identifies enrichment of pathogenic (TBX21, RORC, IFNG, IL-17A, CSF2 expressing) T-helper 17.1 cells and pro-inflammatory (TNF, IL-1β, IL-6, IL-23A expressing) monocytes in ICI-pneumonitis bronchoalveolar lavage fluid, putatively engaging in a feedforward inflammatory loop.

HOW THIS STUDY MIGHT AFFECT RESEARCH, PRACTICE OR POLICY

⇒ The data yield several novel potential therapeutic targets for ICI-pneumonitis. Notably anti-IL-23 holds promise as an efficacious and safe treatment for ICI-pneumonitis that deserves further research.

Conclusions Using single-cell transcriptomics, we identified accumulation of pathogenic T-helper 17.1 cells in ICI-pneumonitis bronchoalveolar lavage fluid—a phenotype explaining previous divergent findings on T-helper 1 versus T-helper 17 involvement in ICI-pneumonitis—putatively engaging in detrimental crosstalk with pro-inflammatory ‘M1-like’ monocytes. This finding yields several novel potential therapeutic targets for the treatment of ICI-pneumonitis. Most notably repurposing anti-IL-23 merits further research as a potential efficacious and safe treatment for ICI-pneumonitis.

BACKGROUND

Immune checkpoint inhibitors (ICIs) have been approved for the treatment of many tumor types in various disease stages, offering
more durable responses and better tolerability compared with conventional cytotoxic treatments. However, with the expanding use of ICIs, clinicians are increasingly confronted with immune-related adverse events (irAEs). While most of these side effects present with mild symptoms, some patients experience severe immune-mediated toxicity and require immunosuppressive therapy. ICI-related pneumonitis (ICI-pneumonitis), most often occurring in patients with non-small cell lung cancer (NSCLC) treated with programmed cell death protein-1/programmed death ligand-1 (PD-1/PD-L1) blockade, is an irAE deserving special consideration. Its real-world incidence is up to 19%, and it is the most frequent fatal toxicity and require immunosuppressive therapy. Some patients experience severe immune-mediated pneumonitis and require immunosuppressive therapy. ICI-related pneumonitis (ICI-pneumonitis), most often occurring in patients with non-small cell lung cancer (NSCLC) treated with programmed cell death protein-1/programmed death ligand-1 (PD-1/PD-L1) blockade, is a fatal irAE deserving special consideration. Its real-world incidence is up to 19%, and it is the most frequent fatal adverse event in lung cancer patients, responsible for 35% of iatrogenic deaths. This reflects how difficult it remains to diagnose and effectively treat ICI-pneumonitis, as a result of the scarcity of available pathophysiological knowledge.

Using flow cytometry, Suresh et al analyzed immune perturbations within the ICI-pneumonitis alveolar compartment, comparing bronchoalveolar lavage fluid (BALF) of 12 patients with ICI-pneumonitis to 6 ICI-treated patients without ICI-pneumonitis. They predominantly observed an increase of T cells in BALF. Particularly, CD4+ central memory T cells were increased, which on in vitro stimulation displayed increased tumor necrosis factor alpha (TNF-α) and interferon-gamma (IFN-γ) secretion, suggesting a T-helper 1 (Th1)-mediated immune response to underlie ICI-pneumonitis. Moreover, they observed that regulatory T cells (Treg) had decreased immunoregulatory capacity, leading them to speculate that diminished immunosuppression by Treg might trigger an exuberant Th1 immune response in ICI-induced pneumonitis. Myeloid cells also gained a more inflammatory phenotype in this study, expressing higher levels of interleukin (IL)-1β and TNF-α. In contrast, Wang et al profiled by flow cytometry longitudinally-collected peripheral blood mononuclear cells of 13 patients with ICI-pneumonitis and proposed a different mechanism. Particularly, they observed an increase in circulating T-helper 17 cells (Th17), as well as an increase in the levels of IL-17 in plasma and BALF, suggesting an important role of Th17 in mediating ICI-pneumonitis. A recent bulk transcriptomic study on surgical biopsies of 29 patients with NSCLC and normal lung samples profiled by single-cell RNA-sequencing (scRNA-seq) data of eight early-stage NSCLC tissue samples (~338 million unique transcripts from 141,056 cells). Dimensionality reduction and clustering was performed using Seurat V4.1.0. Main cell types were annotated according to established marker gene panels (figure 1A, B). There was no evidence of clustering bias (ICI-pneumonitis vs control, or across individual patients; online supplemental figure S1).

To refine our subclustering efforts, we used additional samples profiled by scRNA-seq. Specifically, we pooled 3’ scRNA-seq data of eight early-stage NSCLC tissue samples (~338 million unique transcripts from 72,170 cells) and seven normal lung samples (~134 million unique transcripts from 29,616 cells; online supplemental figure S2A). Assignment of main cell types was done in a similar fashion as for the BALF data, without evidence of batch effects (NSCLC vs normal lung, or across individual patients; online supplemental figure S2B–D). Subsequently, data from each main cell type (derived from BALF, NSCLC and normal lung samples) were integrated separately using canonical correlation analysis, as described previously, without signs of batch effects across sequencing technologies, data sets or individual patients (online Supplemental figure S3). After integration, subclustering of main cell types was done based on differential expression of marker genes.

**METHODS**

We prospectively collected BALF from 11 patients (10 patients with NSCLC and 1 patient with melanoma) receiving anti-PD-1/PD-L1 treatment and developing ICI-pneumonitis. Diagnosis of ICI-pneumonitis was based on clinical, laboratory, radiographic and (invasive) microbiological findings. Grade was moderate to severe in all patients (grade 2–3 according to Common Terminology Criteria for Adverse Events (V.5.0), with only two patients requiring supplemental oxygen, and all patients received first-line treatment according to European Society for Medical Oncology guidelines. Importantly, BALF was collected prior to the start of this treatment. We also analyzed BALF data from six control patients. These patients were demographically matched and underwent bronchoalveolar lavage of an unaffected (contralateral) lobe, solely for research purposes, during bronchoscopy with transbronchial biopsy of a newly diagnosed lung tumor. Detailed inclusion criteria as well as demographic and clinical data of the patient cohort are summarized in table 1, online Supplementary Table S1 and in the Supplementary Methods section.

Immediately after collecting BALF, cells were subjected to single-cell profiling using the 5’ scRNA-seq kit from 10x Genomics. After quality control and filtering (online supplemental methods), we obtained ~775 million unique transcripts from 141,056 cells. Dimensionality reduction and clustering was performed using Seurat V4.1.0. Main cell types were annotated according to established marker gene panels (figure 1A, B). There was no evidence of clustering bias (ICI-pneumonitis vs control, or across individual patients; online supplemental figure S1).

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To gain information on clonotype distribution and dynamics, we performed scTCR-seq on these samples and considered all productive TCRs, which we defined as T cells with TCRs that can be joined in a reading frame by V(D)J recombination without premature stop codons, enabling assessment of a complete TCR α or β chain for downstream analysis. Relative clonotype richness was defined as the number of unique TCRs divided by the total number of cells with a unique TCR. Clonotype evenness was assessed by calculating the inverse Simpson index divided by the number of unique clonotypes. Gini coefficient, a summary metric of inequality of clonotype distribution within a repertoire, was calculated using the ‘Gini’ function from the DescTools R package.15 16

We used the CellRank algorithm to functionally characterize CD4+ T-cell state transitions, based on transcriptional similarity between cells and RNA velocity.17

A detailed workflow of downstream analyses is provided in the online supplemental methods section.

### RESULTS

**T cells dominate the bronchoalveolar space in ICI-pneumonitis**

After quality control and filtering (online supplemental methods), we obtained 141,056 cells of which 85,481 were derived from ICI-pneumonitis BALF and 55,575 from demographically-matched control BALF. We identified several clusters, which we linked to cell types (figure 1A) based on canonical marker gene expression (figure 1B), and we then evaluated changes in immune cell proportions (figure 1C). T cells made up more than half of all immune cells in ICI-pneumonitis BALF, a striking fourfold increase compared with control BALF. Conversely, we observed a clear relative depletion of monocytes/macrophages in ICI-pneumonitis BALF, while dendritic cells (DCs) were threefold increased. For B cell, neutrophil and mast cell abundance, no statistically significant differences between ICI-pneumonitis and control BALF were observed.

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**Table 1**  Demographics and characteristics of study cohort

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<tr>
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<th>ICI-pneumonitis (n=11)</th>
<th>Control (n=6)</th>
<th>P value</th>
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<tr>
<td>Age, years</td>
<td>68.5 (61.7–72.8)</td>
<td>65.7 (53.0–75.9)</td>
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<tr>
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<td></td>
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<tr>
<td>Men</td>
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<td>3 (50.0)</td>
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</tr>
<tr>
<td>Women</td>
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<td>3 (50.0)</td>
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<tr>
<td>Smoking status</td>
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<tr>
<td>Never</td>
<td>1 (9.1)</td>
<td>3 (50)</td>
<td>0.11</td>
</tr>
<tr>
<td>Former</td>
<td>9 (81.8)</td>
<td>2 (33.3)</td>
<td></td>
</tr>
<tr>
<td>Active</td>
<td>1 (9.1)</td>
<td>1 (16.7)</td>
<td></td>
</tr>
<tr>
<td>COPD</td>
<td>3 (27.3)</td>
<td>2 (33.3)</td>
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<td>Underlying malignancy</td>
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<tr>
<td>Melanoma</td>
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<td>0 (0)</td>
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<td>NA</td>
</tr>
<tr>
<td>3</td>
<td>2 (18.2)</td>
<td>NA</td>
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</tbody>
</table>

Data are median (IQR) or n (%). The p values are from Mann-Whitney U test for continuous data and Pearson’s χ² for non-ordered categorical data, all based on a two-sided hypothesis.

*Last administration of chemotherapy at least 4 months prior to sampling.
†Radiotherapy completed at least 3 months prior to sampling.
‡Grading according to Common Terminology Criteria for Adverse Events V.5.0.

COPD, chronic obstructive pulmonary disease; ICI, immune checkpoint inhibitor; NSCLC, non-small cell lung cancer.
Altogether, these broad immune cell typing efforts confirm previous reports highlighting prominent T-cell involvement in ICI-pneumonitis, but they fail to point towards neutrophil or B cell involvement. Furthermore, we describe enrichment of DCs as a novel characteristic of BALF from ICI-induced pneumonitis.

**Pathogenic CD4+ T\textsubscript{H17.1} and effector memory CD8+ T cells in ICI-pneumonitis BALF**

To better characterize the T-cell response, we subclustered all 65,293 T cells and performed an in-depth analysis of T-cell subclusters. Overall, we identified seven CD4+ T-cell subtypes; naïve-like (T\textsubscript{N}, with CCR7, LEF1 and TCF7 as marker genes), effector memory (T\textsubscript{EM}; ANXA1, CXCR4, IL-2), T follicular helper (T\textsubscript{FH}; BCL6, CXCR5, ICA1), two clusters of T\textsubscript{REG}, and two clusters of CD4+ T cells with both (T\textsubscript{H17}-like; RORC, IL-23R, CCR6) and (T\textsubscript{H1}-like; TBX21, IFNG, CXCR3) properties, termed T\textsubscript{H17.1} cells (figure 2A,B; online supplemental figure S4A). T\textsubscript{REG} consisted of a FOXP3+ T\textsubscript{REG} cluster (T\textsubscript{REG}; FOXP3, IL-2RA, IL-1R2) and FOXP3– regulatory type 1 T-cell cluster (T\textsubscript{R1}; CRTAM, IL-10, LAG3), which derive from T-helper, CD4+ naïve or memory T cells in inflamed peripheral organs where they induce tolerance, mainly by producing IL-10 and by selective killing of antigen-presenting cells (APCs) through granzyme B and perforin secretion. Indeed, the cytotoxic capacity of these CD4+ T cells explains why they cluster in close proximity to CD8+ T cells on the Uniform Manifold Approximation and Projection plot (online supplemental figure S4B). T\textsubscript{H17.1} cells finally constituted a large cluster, termed T\textsubscript{H17.1}\textsubscript{RORC} cells, with balanced expression of T\textsubscript{H1}-related and T\textsubscript{H17}-related genes, and a smaller cluster, termed T\textsubscript{H17.1}\textsubscript{TBX21} cells, with a pronounced pathogenic phenotype evidenced by retained T\textsubscript{H17}-related gene expression (RORC, CCR6, IL-17A, IL-23R), but...
Figure 2  T-cell phenotypes and relative abundance in ICI-pneumonitis and control bronchoalveolar lavage fluid. (A) Uniform Manifold Approximation and Projection plot of 65,293 T cells, (B) annotated according to canonical marker gene expression of CD4+ (left panel) and CD8+ T cells, MAIT, and NK-cells (right panel). (C) A comparison of relative T-cell subtype abundance, revealed accumulation of CD4+ T-helper 17.1 cells with a pathogenic phenotype (T_{\text{Th17.1,TEM}}) in ICI-pneumonitis, as well as enrichment of CD4+ regulatory type 1 T cells (T_{\text{Tr1}}), CD4+ naive-like T cells (T_{\text{N}}), and CD8+ effector memory (T_{\text{EM}}) T cells. Wilcoxon rank-sum test was used; significance is shown as *p<0.05. (D) Volcano plot showing differentially expressed genes in T cells comparing ICI-pneumonitis and control T cells. P values were obtained by the model-based analysis of single-cell transcriptomics (MAST) test and Bonferroni-corrected (see online supplemental table S1 for all differentially expressed genes). (E) Differential gene set enrichment analysis (DGSEA) on differentially expressed genes for ICI-pneumonitis versus control T cells using the R package hypeR. Only significant genes (adjusted p value<0.05) and genes with a log-fold change higher than 0.5 or lower than −0.5 were used (see online supplemental tables S2 and S3 for all differentially expressed gene sets in ICI-pneumonitis and control T cells, respectively). CD4_N, CD4+ naive-like T cells; CD4_EM, CD4+ effector memory T cells; CD4_TR1, CD4+ regulatory type 1 T cells; CD4_Th17.1_RORC, CD4+ T-helper 17.1 lymphocytes with predominant non-pathogenic features; CD4_Th17.1_TXB21, CD4+ T-helper 17.1 lymphocytes with predominant (pathogenic) T-helper 1-like features; CD4_FH, CD4+ follicular helper T-cells; CD4_REG, CD4+ regulatory T cells; CD8_N, CD8+ naive-like T cells; CD8_EM, CD8+ effector memory T cells; CD8_RM, CD8+ resident memory T cells; CD8_EMRA, CD8+ recently activated effector memory T cells; CD8_EX, CD8+ exhausted T cells; CD8_gd, CD8+ gamma delta T cells; DGEA, differential gene expression analysis; FDR, false discovery rate; ICI, immune checkpoint inhibitor; MAIT, mucosal associated invariant T cells; NK, natural killer; NK_cyto, cytotoxic NK-cells; NK_infla, inflammatory NK-cells.

upregulation of T_{\text{H1}}-related (TBX21, IFNG), cytotoxicity-related (GZMB, PRFI, GNLY), exhaustion-related (PDCD1, CTLA4, LAG3) and monocyte activation-related genes (CSF2, encoding granulocyte-macrophage colony-stimulating factor (GM-CSF)).^{10}

Next, we identified seven CD8+ T-cell subtypes, namely (T_{\text{N}}, CCR7, LEF1, TCF7), (T_{\text{EM}}, GZMK, GZMM, CXCR4), resident-memory (ZNP683, ITGAE, ITGA1), recently activated effector memory (CX3CR1, FGBP2, FCGR3A) and experienced (co-expressing effector markers GZMB,
**Cell fate mapping and TCR analysis reveals pathogenic phenotype shift of T-helper cells**

The above analyses suggested a central role for a pathogenic subset of Th17.1 cells in ICI-pneumonitis pathophysiology. To more comprehensively characterize the Th17.1 immune response, we performed single-cell fate mapping and continuous gene expression analysis by running the CellRank algorithm on CD4+ T-cells. This algorithm ranks individual cells based on similarity in gene expression and RNA velocity along probabilistic state-change trajectories. It allows to study and compare relatedness and gene expression dynamics of cell subtypes between different conditions in a continuous manner. After we excluded Treg from these analyses given their distinct developmental paths, we identified a trajectory in which Tem were connected to the Th17.1_RORC cluster, which then branched into either Th1 or Th17 cells, a regulatory cell type relying mainly on IL-10 secretion and APC killing to exert its function, suggesting they represent a negative feedback mechanism counterbalancing Th17.1-mediated inflammation. Indeed, although enrichment of an IL-10 secreting regulatory cell type may seem counterintuitive, increased systemic levels of IL-10 during ICI-pneumonitis development have been reported previously.

**Cell fate mapping** and **TCR analysis** reveal pathogenic phenotype shift of T-helper cells...
transcription factors, for example, \textit{ZFX} (driving a TH17 self-renewal program) and \textit{FOXO1} (downregulating inflammatory cytokine secretion), coupled to expression of mostly non-pathogenic surface molecules and cytokines; \textit{PROCR} (which negatively regulates pro-inflammatory cytokine secretion), \textit{IL-1RN} (encoding IL-1 receptor antagonist, unexpectedly co-expressed with IL-1 receptor \textit{IL-1RI}), \textit{IL-6R} (which ‘stabilizes’ non-pathogenic TH17-like

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{image.png}
\caption{Cell fate mapping and T-cell receptor repertoire analysis of CD4+ T cells in ICI-pneumonitis and control bronchoalveolar lavage fluid. (A) Uniform Manifold Approximation and Projection plot of 11,570 CD4+ T cells with cell fate trajectories and (B) latent time as calculated by the CellRank algorithm, showing \textit{T}_{EM} were connected to the \textit{T}$_{\text{TH17.1_RORC}}$ cluster, which branched into either \textit{T}$_{\text{TH17.1_TBX21}}$ or \textit{T}$_{\text{TH17.1_TR1}}$ as terminal states, or formed a terminal state itself. (C) Continuous gene expression profiling along the \textit{T}$_{\text{TH17.1_TBX21}}$ trajectory. (D) Barplot of relative cell abundance in ICI-pneumonitis and control BALF along the \textit{T}$_{\text{TH17.1_TBX21}}$ trajectory. (E) Line graph of the Gini coefficient for ICI-pneumonitis and control BALF T-cell receptor repertoire along the \textit{T}$_{\text{TH17.1_TBX21}}$ trajectory, as calculated by the DescTools algorithm. CD4_\text{N}, CD4+ naïve-like T cells; CD4_\text{EM}, CD4+ effector memory T cells; CD4_\text{TR1}, CD4+ regulatory type 1 T cells; CD4_\text{Th17.1_RORC}, CD4+ T-helper 17.1 lymphocytes with predominant non-pathogenic features; CD4_\text{Th17.1_TBX21}, CD4+ T-helper 17.1 lymphocytes with predominant (pathogenic) T-helper 1-like features; CD4_FH, CD4+ follicular helper T cells; ICI, immune checkpoint inhibitor.}
\end{figure}
Further along the trajectory, transcription factors associated with a pathogenic phenotype are upregulated, most notably TBX21 (inducing a.o. IFN-γ secretion) and PRDM1 (encoding BLIMP-1, a transcription factor promoting a.o. GM-CSF secretion). As a result, pro-inflammatory chemokines (CCL4, CCL5), cytotoxicity genes (GZMB, GZNY), and especially key genes mediating T_h17,1 pathogenicity, namely IFNG and CSF2, are progressively upregulated near the end of the T_h17,1 trajectory, where cell density peaks for ICI-pneumonitis BALF samples.32,33

In a similar fashion, we comparatively examined T-cell clonality along the T_h17,1_TBX21 trajectory, obtaining 11,570 CD4+ T cells with a TCR sequence. To assess clonotype diversity and distribution, we calculated TCR richness and evenness as well as the Gini coefficient (figure 3E; online supplemental figure S4F). While low TCR richness indicates a limited number of T-cell specificities, low evenness marks a skewed distribution of these specificities. The Gini coefficient is a summary metric of inequality of clonotype distribution within a repertoire, such that a high Gini coefficient and low TCR richness and evenness reflect clonal expansion of specific T-cell clones following cognate antigen recognition.15 We observed a progressive increase of the Gini coefficient, and a decline of TCR richness and evenness, along the T_h17,1_TBX21 trajectory in ICI-pneumonitis, which was not seen in control BALF. This suggests strong antigen-driven clonal proliferation of T_h17,1 cells, another established feature of pathogenic T_h17,1 cells.33 Overall, cell fate mapping coupled to single-cell TCR repertoire analysis unraveled the plasticity of T_h17,1 cells in ICI-pneumonitis and control BALF. These cells are skewed towards a pathogenic IFNG_{high} CSF2_{high} phenotype in ICI-pneumonitis BALF, regulated by transcription factors TBX21 and PRDM1, respectively. Importantly, T-cell clonality analyses showed pronounced antigen-driven clonal expansion of this phenotype, further supporting the notion that these are not just bystander activated T cells but central players in ICI-pneumonitis pathophysiology.33

Crosstalk between pathogenic T_{h17,1} cells and pro-inflammatory monocytes

We next wondered how innate immune cells are involved in ICI-pneumonitis immunopathology, and subclustered all 63,330 monocytes and macrophages (figure 4A–C). First, monocytes were separated from macrophages based on differential canonical monocyte (FCN1, LILRB2, LILRA5) or macrophage marker gene expression (PPARG, FABP4, MARCO), respectively. Monocytes were then subclustered into classical FCN1_{high} (FCN1, S100A8, S100A9) and inflammatory IL-1B_{high} monocytes (IL-1B, CCL20, IL-6). Within the macrophages, we identified monocyte-derived (expressing LGMN, CH3LI and MERTK) and tissue-resident FABP4_{high} alveolar macrophages (FABP4, PPARG, MARCO), which formed a non-proliferating and a proliferating subcluster (MKI67, TOP2A, CDK1). We observed a significant decrease of anti-inflammatory alveolar macrophages in ICI-pneumonitis BALF, and a corresponding increase of monocyte-derived macrophages and pro-inflammatory IL-1B_{high} monocytes (figure 4D).

Besides IL-1B, these pro-inflammatory monocytes show high expression of other pro-inflammatory genes, most notably TNF (encoding TNF-α), IL-6, IL-23A, CSF2RA, CSF2RB, CCL20 and the interferon-induced chemokines CXCL9/10, CXCL9/10 and CCL20 are potent chemoktractants of CXCR3+ and CCR6+ cells, respectively, with presence of both chemokinesis receptors defining T_h17,1 cells.34 IL-1 and IL-23 moreover are critical cytokines inducing pathogenic features (notably GM-CSF secretion) in T_h17,1 cells.19 Reciprocally, GM-CSF signaling through the GM-CSF receptor (CSF2RA, CSF2RB) is known to induce this pro-inflammatory monocyte phenotype.19 DGEA and DGSEA across all monocytes/macrophages comparing ICI-pneumonitis and control BALF, confirmed that pro-inflammatory IL-1B_{high} monocytes are involved in the innate immune response in ICI-pneumonitis (figure 4E,F). Specifically, expression of pro-inflammatory ‘M1-like’ genes (CCL3, CCL4, IL1B, TNF, NFKBIA) was upregulated in ICI-pneumonitis monocytes/macrophages, while anti-inflammatory ‘M2-like’ genes (IGF1, MARCO) were relatively enriched in control monocytes/macrophages.

We also analyzed the DC compartment in a similar fashion, as these cells have a central role in directing T-cell phenotypes. Subclustering of DCs revealed classical type I DCs (CLEC9A, XCR1, CPNE), type II DCs (CLEC10A, FCGR2B, FCR16), plasmacytoid DCs (LILRA4, CLEC4C, IRF7), and migratory DCs (CCR7, LAMP3, FCN1) (figure 5A,B). Although DCs were overall enriched in ICI-pneumonitis BALF, we observed no significant differences for DC subtype abundance between ICI-pneumonitis and control BALF (figure 5C). Hence, the role of DCs in shaping the ICI-pneumonitis immune response remains unclear. Our data do suggest, however, that myeloid cells, specifically pro-inflammatory monocytes, might not only directly mediate tissue damage in ICI-pneumonitis but also sustain tissue inflammation by recruiting and shaping pathogenic T_{h17,1} cells under the influence of GM-CSF.10

DISCUSSION

Here, we shed light on the mechanisms underlying pneumonitis occurring in patients with cancer during anti-PD-1/PD-L1 therapy. These findings are important, as they are a necessary first step to identifying immunomodulatory treatments effectively targeting the root cause of ICI-pneumonitis. These are currently lacking, making ICI-pneumonitis the most frequent fatal anti-PD-1/PD-L1-related adverse event. For the first time, we present a single-cell transcriptomic atlas of the bronchoalveolar immune landscape in ICI-pneumonitis, comparing the relative abundance of immune cell types as well as their precise phenotype and clonotype distribution to those in BALF of demographically-matched controls.
First and foremost, we confirm that T-cell accumulation, of both CD4+ and CD8+ T cells, is a hallmark of ICI-pneumonitis. T cells constitute about 10% of immune cells in control BALF (in line with historical findings in healthy BALF), whereas they make up more than half of all immune cells in ICI-pneumonitis BALF. T_{H17} cells, which are T_{H1} cells that have gained T_{H1} features such as expression of transcription factor T-bet (encoded by TBX21) and IFN-γ (IFNG), make up the bulk of T cells in both ICI-pneumonitis and control BALF (which...
corresponds to a fourfold increase of the ratio of $T_{H17.1}$ cells over all immune cells in ICI-pneumonitis). More importantly we show, both by subclustering and by fate mapping of $T_{H17.1}$ cells coupled to TCR repertoire analysis, that they display a more pathogenic phenotype evidenced by higher expression of IFNG, CSF2 (encoding GM-CSF) and cytotoxicity genes and by strong antigen-driven clonal proliferation. This cell type is increasingly being recognized in autoimmune processes, such as IBD, multiple sclerosis and sarcoidosis, previously thought to have been mediated by $T_{H1}$ cells. The type 1 immune response conveyed by $T_{H17.1}$ cells is probably strengthened by CD8+ effector memory T cells, which were more abundant in ICI-pneumonitis BALF. Earlier studies on

Figure 5  Dendritic cell phenotypes and relative abundance in ICI-pneumonitis and control bronchoalveolar lavage fluid. (A) Uniform Manifold Approximation and Projection plot of 2642 dendritic cells, (B) annotated according to canonical marker gene expression. (C) A comparison of relative dendritic cell subtype abundance showed no statistically significant differences between ICI-pneumonitis and control bronchoalveolar lavage fluid. Wilcoxon rank-sum test was used. cDC1, classical type I dendritic cell; cDC2, classical type II dendritic cell; ICI, immune checkpoint inhibitor; migDC, migratory dendritic cell; pDC, plasmacytoid dendritic cell.
ICI-pneumonitis also observed prominent lymphocytophagia, with higher abundance of TNF-α^{high} CD8+ T cells, but contradictory findings were reported on T_{HI} versus T_{HI}, involvement. We argue this might be due to limited flow cytometry panels used in these studies (compared with whole-transcriptome scRNA-seq data) that did not include the canonical T_{HI} transcription factor antibody anti-RORγt, hence precluding the identification of a hybrid T_{HI}, phenotype.  

Second and somewhat unexpectedly, we observed a higher abundance of FOXP3^{+} IL-10^{high} T_{R1} in ICI-pneumonitis BALF. In preclinical models, it has been shown that in pro-inflammatory conditions, in the presence of TGF-β/IL-6/IL-23, T_{HI} cells are skewed towards a pathogenic T_{HI}, phenotype but that a small proportion of these T_{HI} cells gives rise to T_{R1} cells, in line with our findings. Indeed, while IL-10 is the prototypical anti-inflammatory IL, increased serum IL-10 levels have been reported during ICI-pneumonitis.  

How innate immune cells interact with T_{HI} cells in ICI-pneumonitis remains speculative, but our scRNA-seq data provide a solid basis for hypothesis generation. First, we show that anti-inflammatory phagocytosing alveolar resident macrophages are depleted, while monocyte-derived macrophages and pro-inflammatory IL-1^{high} monocytes are relatively enriched in ICI-pneumonitis BALF. A detrimental feedforward loop in which pathogenic T-helper cell-derived GM-CSF instructs GM-CSF receptor expressing monocytes to orchestrate tissue inflammation (TNF-α, IL-6) and tissue damage (reactive oxygen species) while sustaining pathogenic T-helper cells (through IL-23 secretion, intensified by IL-1β secretion), is a well-established mechanism of auto-inflammation, best characterized in multiple sclerosis but also discerned in one patient with inherited PD-1 deficiency. In line with this mechanism, is a prior study in which pathogenic T-helper cell-derived GM-CSF instructs GM-CSF receptor expressing monocytes to orchestrate tissue inflammation (TNF-α, IL-6) and tissue damage (reactive oxygen species) while sustaining pathogenic T-helper cells (through IL-23 secretion, intensified by IL-1β secretion), is a well-established mechanism of auto-inflammation, best characterized in multiple sclerosis but also discerned in one patient with inherited PD-1 deficiency. In our study cohort (also with regards to smoking status, lung disease and the presence of a thoracic malignancy, importantly), these patients did not receive prior anti-PD-1/PD-L1 treatment. All BALF samples were taken during routine bronchoscopy so we could minimize patients’ exposure to the risks of this invasive procedure, knowing that anti-PD-1/PD-L1 does not seem to induce changes in the immune compartment of non-inflamed organs. While scRNA-seq and scTCR-seq data allowed us to put forward several interesting mechanistic hypotheses, further research is needed to validate these hypotheses. Finally, our data cannot provide an answer to the question why some patients develop ICI-pneumonitis and others do not. Indeed, while we provide data on the TCR repertoire in ICI-pneumonitis BALF, larger studies combining in-silico TCR repertoire with in vitro TCR reactivity data (examining (cross-)reactivity of T cells in ICI-pneumonitis lesions) are presumably key to identify the substrate of the immune reaction in ICI-pneumonitis. In parallel, genomic studies (eg, examining IL-23R or IL-10R polymorphisms) will help to uncover susceptibility factors to the development of irAEs. 

In conclusion, we use scRNA-seq and scTCR-seq to perform a deep immune profiling effort and characterize the immune response in ICI-pneumonitis, thereby yielding novel pathophysiological insights. We offer a clear rationale for a novel targeted treatment approach for ICI-pneumonitis to improve outcome of this potentially fatal irAE while maintaining tumor control.
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Contributors AF contributed to methodology and performed formal analysis, visualization, investigation and writing (original draft and revision). PVM contributed to methodology and project administration, formal analysis, investigation, resource gathering and data curation and performed writing (original draft and revision). SV and ED contributed to investigation and provided resources. RS and TVB performed experiments. CD, JY, NDC, DT,WDW, KN, JV, RV and SH-B provided resources and performed investigation. DL and EW conceptualized the study, provided resources, performed investigation and writing (original draft and revision); supervised the study, and were responsible for project administration and funding acquisition. EW accepts responsibility for the work as guarantor.

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