High-dimensional single-cell analysis delineates radiofrequency ablation induced immune microenvironmental remodeling in pancreatic cancer

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Abstract
Radiofrequency ablation (RFA) is an effective local therapy approach for treating solitary tumor of many types of malignancy. The impact of RFA on the tumor immune microenvironment on distant tumors after RFA treatment is still unclear. In this study, by using syngeneic tumor models and single-cell RNA and T-cell receptor sequencing, we have investigated the alterations of tumor-infiltrating immune cells in distant non-RFA tumors. Single-cell RNA sequencing identified six distinct lymphoid clusters, five distinct monocyte/macrophage clusters, three dendritic cells clusters, and one cluster of neutrophils. We found that RFA treatment reduced the proportions of immunosuppressive cells including regulatory T cells, tumor-associated macrophages and tumor-associated neutrophils, whereas increased the percentages of functional T cells in distant non-RFA tumors. Moreover, RFA treatment also altered gene expressions in single-cell level in each cell cluster. By using pseudo-time analysis, we have described the biological processes of tumor-infiltrating CD8+ T cells and monocytes/macrophages based on the transcriptional profiles. In addition, the immune checkpoints including PD-1 and LAG3 were upregulated in the T cells in distant non-RFA tumors after RFA treatment. In conclusion, our data indicate that RFA treatment induced remodeling of tumor immune microenvironment in distant non-RFA tumors in pancreatic cancer mouse model and suggest that combining RFA with immune checkpoint inhibitors may be an effective treatment approach.

Introduction
Pancreatic ductal adenocarcinoma (PDAC) is the fourth leading cause of cancer-related mortality in the United States1. Most patients diagnosed with an advanced stage with limited effective treatment options2. Immune checkpoint inhibitors have had remarkable efficacy in many malignancies, such as advanced melanoma, non-small-cell lung cancer, renal cell carcinoma, bladder cancer, Hodgkin lymphoma, and head and neck cancer3,4, but not in PDAC. The immunosuppressive tumor microenvironment (TME) in PDAC, characterized by typically poor infiltration of effector T cells and prominent myeloid inflammation, appears to be the main factor for such treatment resistance5. Studies from our laboratory6 and other laboratories4,7–10 have revealed that the combination of immune checkpoint inhibitors and other modalities, including radiotherapy, chemotherapy, oncolytic virus, targeted therapy, and other immunotherapies, may be a promising strategy.

Radiofrequency ablation (RFA) is a local therapy effective for oligometastasis of lung and liver as well as other organs10–12, RFA induced localized coagulative necrosis that may lead to the release of tumor antigens, which could elicit
systemic adaptive immune response against the tumors. It is not clear if the immune response induced by local RFA could generate effective antitumor immune responses in distant non-RFA tumors. Most studies have focused on the changes of immune microenvironment in the RFA-treated tumors after RFA treatment.

We have established a PDAC mouse model with tumor implant on bilateral flanks to recapitulate RFA and non-RFA sides (Fig. 1a). After RFA treatment on one side of tumor, the single-cell RNA sequencing (scRNA-seq) and single-cell T-cell receptor (TCR) sequencing were used to evaluate the cellular compositions and transcriptional landscape of TME on the other side, the non-RFA tumor. We have used the single-cell transcriptomics to characterize the changes of immune cells in response to RFA in the TME.

**Materials and methods**

**Tumor transplantation and treatment**

The mouse pancreatic cancer cell line Panc02 was obtained from Shanghai Aolu Biological Technology Co., Ltd. The Panc02 cells were transplanted into 10-week-old female C57BL/6 mice. A total of 500 tumor cells were injected subcutaneously in the bilateral flanks of the mice. After 2 weeks, when the tumor volumes reached approximately 200-300 mm³, the mice were randomly divided into five groups: control group (n = 10), non-RFA group (n = 10), RFA left group (n = 10), RFA right group (n = 10), and non-RFA right group (n = 10). The RFA group received RFA treatment on the left side tumors, whereas the non-RFA group received no treatment.

**Fig. 1 Changes in distant non-RFA tumor after RFA treatment.** a Diagram of control group and RFA group. RFA treatment was performed on the left side tumors (RFA side). The right side (non-RFA side) tumors were used to further analysis. b H&E staining of tumor tissues on RFA and non-RFA side after RFA treatment. Scale bar = 50 μm (red line at the bottom right). c After 2 weeks, tumors from control group and RFA group were harvested and measured. Tumor volumes were compared to each two groups. p values were calculated based on a Student’s t test (n = 4 per group). d, e Percentage of CD8⁺PD-1⁺ cells in distant non-RFA tumor on day 3, 5, and 8 after RFA treatment. p values were calculated based on a Student’s t test (n = 3 per group).
Company. Tumor cell propagated in vitro in Dulbecco’s Modified Eagle Medium (HyClone) supplemented with 10% fetal bovine serum (FBS, Gibco) were washed extensively, resuspended in endotoxin-free phosphate-buffered saline (PBS) and then 200 µl injected (1 × 10^6 cells) subcutaneously into male C57BL/6 mice on bilateral flanks, respectively. Then randomly separated into two groups and no blinding was done. In this study, 3–5 mice per group were used as indicated in the Figure legend. Treatments were initiated when the tumor volume reached ~500 mm^3.

RFA was performed using an 18-gauge single ablation electrode (STARmed Co., Ltd., Korea) with 7 mm active tip inserted percutaneously and orthogonal to the skin in the center of the tumor on the left flank, whereas the tumor on the right side did not receive any intervention. Treatments were administered for 3.5–4.5 min at the target temperature of 70 °C to ensure complete ablation of the target tumors. The tumor on the right flank (non-RFA side) was used for further analysis (Fig. 1a). All experiments were approved by the Ethics Committee for Animal Research of 900 Hospital of the Joint Logistics Team.

Tumor harvest

Tumor masses on the right flank were removed, diced, digested with tumor dissociation kit (Order no. 130–095–929, Miltenyi Biotec Inc.) at 37 °C for ~40 min. Resuspend sample and applied the cell suspension to a 70 µm MACS SmartStrainer (Order no. 130–098–462) to remove debris and separate cell clumping. Cells were washed twice. Red blood cells were lysed using ACK lysis buffer (G53309). To remove aggregates and clumps, cells were filtered through a 70 µm MACS SmartStrainer. Then cells were washed and resuspended in PBS plus 1% FBS for analysis.

Flow cytometric analysis

Antibodies against CD8a (53–6.7, eBioscience), PD-1 (29 F.1A12, BioLegend) were used for marker staining for 30 min at 4 °C. All flow cytometry was performed on the BD Accuri C6 Plus (BD Biosciences) and analyzed using FlowJo software.

Immunohistochemistry

Immunohistochemical experiment procedure was described previously. The rabbit anti-CD4 antibody (EPR19514, Abcam), anti-CD8 antibody (YTS169.4, Abcam), rabbit anti-FOXP3 antibody (D2W8E, Cell Signaling Technology), rabbit Anti-CD206 antibody (ab64693, Abcam), and rabbit anti-iNOS antibody (ab15323, Abcam) were used to immunohistochemistry staining. The density of the positive cells was calculated with ×400 magnification in five representative fields in the tumor tissues, and the average was calculated. Each section was examined independently by two investigators in a blinded manner.

Intratumoral CD45^+ cell isolation

For separating of intratumoral CD45^+ cells, single-cell suspensions of digested tumor were resuspended in a buffer (PBS, 0.5% bovine serum albumin, and 2 mM thylendiaminetetraacetic acid) and labeled with CD45 (TL) MicroBeads (Order no. 130-110-618, Miltenyi Biotec Inc.) for 15 min in the dark at 4 °C. CD45^+ cells were separated with LS columns (order no. 130-042–401, Miltenyi Biotec Inc.). The purity of the separated cells was used for scRNA-seq analysis.

Single-cell RNA-seq

Single-cell RNA-seq library preparation: according to the manufacturer’s instructions, single-cell RNA-seq was performed by encapsulating separated live CD45^+ tumor-infiltrating cells into droplets and libraries were constructed using Single Cell 5’ Library and Gel Bead Kit and Single Cell V(d)J Enrichment Kit. The libraries were finally sequenced using an Illumina Novaseq6000 (performed by CapitalBio, Beijing).

Cell ranger pipeline

The Cell Ranger software (3.0.1) obtained from 10x Genomics website https://support.10xgenomics.com/single-cell-gene-expression/software/downloads/latest was used to perform sample demultiplexing, alignment, filtering, barcode counting, and UMI counting. Cellranger mkfastq was used to demultiplex raw base call files into sample-specific fastq files. Then, fastq files for each sample were processed with Cellranger count, which was used to align samples to mm10 genome, filter and qualify. For each sample, the recovered-cells parameter was specified as 8000 cells that we expected to recover for each individual library. The data for each sample were aggregated using the cellranger aggr pipeline, which performed a between-sample normalization step and merged two samples into one.

Seurat package analysis

The Seurat pipeline was applied to the analysis of the combined data set. Genes that were expressed in less than three cells and cells that expressed <400 and >5500 genes, were excluded. The function Sctransform was used for normalization. Principle component analysis (PCA) was performed on ~3000 genes with PCA function. A UMAP dimensional reduction was performed on the scaled matrix (with most variable genes only) using first 40 PCA components to obtain a two-dimensional representation of the cell states. Cell clustering was performed using the function FindClusters that implements SNN.
We found that the tumor growth of RFA side in RFA group did not change significantly compared with the control group on day 3 (Fig. 1d, e), began to decrease slightly on day 5 and showed marked decrease on day 8 (Fig. 1d). These results demonstrated that in the early stage (3–5 days) after RFA treatment, the RFA-induced immune responses were observed in the non-RFA distant tumors, but transient and failed to suppress the tumors.

Identification of tumor-infiltrating immune cells by using scRNA-seq

To further characterize the RFA-induced changes on the tumor-infiltrating immune cells of the non-RFA distant tumors, we used high-dimensional scRNA-seq analysis to reveal the complexity of tumor immunity. We obtained the CD45$^+$ immune cells from tumor mass and analyzed them by scRNA-seq and single-cell TCR sequencing with a 10× genomics pipeline (Fig. 2).

We performed scRNA-seq analysis on CD45$^+$ cells isolated from the tumor tissues of control group (n = 10,659 cells sequenced, coverage of 84,655 mean reads per cell) and the non-RFA side of the RFA-treated group (n = 8005 cells sequenced, coverage of 112,967 mean reads per cell). All cells (n = 18,664) were clustered into unbiased cell type classifications (Fig. 3a) by using the Seurat single-cell analysis R package$^{16,17}$. Comparison with the ImmGen database and assessment of known cell type markers identified six distinct lymphoid clusters, five distinct monocyte/macrophage clusters, three clusters of dendritic cells (DCs), and one cluster of neutrophils (Fig. 3a–c). Clusters are annotated as “XXX_s#” where “XXX” represents the cell types, “s” represents the scRNA-seq, and “#” the different clusters.

Use of scRNA-seq and single-cell TCR sequencing to identify RFA-induced changes in the tumor-infiltrating lymphocytes

To elucidate the transcriptional and functional changes in lymphoid cells, we isolated the lymphoid cells from the tumors by separating them from the other immune cells and reanalyzed them at a high resolution (Fig. 4a). This strategy yielded 17 distinct lymphoid subpopulations broadly defined by the distribution of the classical marker genes (Fig. 4b). We then performed an integrative analysis of scRNA-seq and TCR repertoire in the lymphoid subpopulations to examine the cellular context of the immune responses and immune repertoires of T cells. The clonotypes were defined based on the CDR3 sequences of both TCR alpha and beta chains using the cell ranger V(D)J analysis pipeline$^{20}$. We eliminated the non-productive and low-abundance TCRs. Subsequently for defining the clonotype, we strictly defined T cells with at least one pair of identical paired alpha and beta chains to be one clonotype. In total, we detected 641
clonotypes among 1236 cells in control group and 605 clonotypes among 1099 cells in RFA group. The details of clonotypes in each cluster of lymphoid subpopulations were shown in Table S1. Using this analysis approach, RFA-induced changes in the subpopulation proportions and their transcriptional profiles could be exhibited clearly in each cell type.

**Dysfunction of Regulatory T cells (Tregs)**

Tregs express Foxp3 and Cd4, and play immunosuppressive roles in the TME. We found that the number and percentage of Tregs were marginally decreased in RFA group compared with the control group (Fig. 4d). Moreover, by using immunohistochemical staining, we revealed that the number of FOXP3-positive cells in the tumor tissues were similar between the RFA and the control group (Figure S2). After RFA treatment, Tregs displayed increasing expression of Foxp3 and Il2ra (Figure S1), which may stabilize and sustain Tregs by signaling through the IL-2/IL-2R axis, suggesting Tregs' immune suppression was enhanced. However, expressions of Il6ra, Tnfrsf4, Tnfrsf18, and Ccl3 (Figure S1) were also increased in the RFA group. The activation of OX40 (Tnfrsf4), glucocorticoid-induced TNF receptor-related gene (Tnfrsf18) and IL-6/IL-6R axis may adversely affect the stability of the Treg phenotype. In addition, Tregs also require activation via the TCR for full acquisition of their immunosuppressive function. The TCR analysis indicated that the multiformity and quantity of clonotypes were reduced after RFA treatment (Fig. 5a). In sum, the RFA-induced transcriptional and functional changes may reduce the immunosuppressive function of Tregs.

**Changes in CD4⁺ T cells**

Single-cell RNA-seq identified four distinct clusters of FOXP3⁺ CD4⁺ T cells (CD4_s1, CD4_s2, CD4_s3,
Both CD4_s3 and CD4_s4 expressed T-cell activation markers Cd81, Stat3, Rora, and Tgbf1 (Figure S1), however, these two clusters also expressed higher levels of genes associated with the exhausted T cells, such as Pdcd1, Ctla4, and Icos. CD4_s2 had higher expression of Traf1, Ifnrg1, Il2ra, Il17a, Il17f, and Cxcl3 (Figure S1). These genes are part of the TNF-signaling pathway, Th17 cell differentiation, and IL17 signaling pathway (Fig. 4e). CD4_s1 exhibited high expression of cytotoxic molecules, such as Nkg7 and Gzmbl, and also expressed high level of Ifng, Eomes, and Prf1 (perforin), suggesting the immune cells in CD4_s1
Fig. 4 RFA induces remodeling of tumor-infiltrating lymphoid cells.  

a UMAP plot showing the merged data of intratumoral lymphocytes.

b UMAP plot of lymphoid cells displaying select marker-gene expression.

c Comparison of each cluster in two groups.

d Percentage of cells in each cluster by condition.

e Heatmap of gene ontology identifying pathway enrichment by each subpopulation.

f Heatmap of KEGG enrichment analysis enrichment by clusters.
Fig. 5 Quantitative and qualitative changes of lymphocytes in individual clusters. 

a Occupancy of clonotypes in each cluster by condition.
b Pseudo-time trajectory of CD8+ T cells analyzed using Monocle 2. 
c CD8+ T-cell clusters from scRNA-seq overlaid on Monocle 2 pseudo-time plot.
d UMAP plots showing CD4 (blue) and CD8 (red) expression in Mki67hi_s1 by condition.
e Percentage of CD4+ and CD8+ T cells in Mki67hi_s1 by condition.
may be cytotoxic CD4 T cells\textsuperscript{20,24}. In addition, RFA treatment decreased the number of CD4\textsuperscript{+} s3 and CD4\textsuperscript{+} s4 cells and increased the number of CD4\textsuperscript{+} s1 and CD4\textsuperscript{+} s2 cells (Fig. 4d). Using immunohistochemical staining, we also observed that the number of CD4\textsuperscript{+} T cells increased after RFA treatment (Figure S2). The available TCR sequences for these cells revealed that CD4\textsuperscript{+} T-cell clusters had a similar number of clonotypes between control group and the RFA group (Fig. 5a). Based on the different numbers of cells in each clonotype of each cluster, we calculated the percentage and the number of cells in clonotype 1–5 (Table S2). We found that compared with the other clusters, CD4\textsuperscript{+} s1 occupied clonotype 1–5 were highest in both control and RFA group (Fig. 5a and Table S2). These findings demonstrate the augmentation of CD4\textsuperscript{+} T-cell activation, especially cytotoxic CD4 T cells, was triggered by RFA treatment.

**Transformation in CD8\textsuperscript{+} T cells**

CD8\textsuperscript{+} T cells are also known as cytotoxic T cells, which induce apoptosis of target cells by releasing the cytotoxins perforin, granzymes, and granulysin or Fas–Fas ligand signal molecules. In this study, scRNA-seq revealed nine distinct subsets of CD8\textsuperscript{+} T cells (Fig. 4c) and majorities of them were cytotoxic T cells. T cells in the CD8\textsuperscript{+} s1, CD8\textsuperscript{+} s8, and CD8\textsuperscript{+} s9 clusters expressed higher level of several functional markers, like Gzmb, Ppfa, and Fasl (Figure S1). CD8\textsuperscript{+} s3, CD8\textsuperscript{+} s6, and CD8\textsuperscript{+} s7 clusters had high expression of Gzmb and median expression of Fasl. Almost all these cytotoxic T cells expressed exhaustion genes [Haver2 (Tim3), Lag3, Pdcd1, and Cita4]. The other three clusters (CD8\textsuperscript{+} s2, CD8\textsuperscript{+} s4, CD8\textsuperscript{+} s5) expressed naive T-cell markers Sell (Cd62L) and Ccr5\textsuperscript{25}, however, these cells also expressed median levels of Fasl. Following RFA treatment, the proportions of CD8\textsuperscript{+} s1, CD8\textsuperscript{+} s2, CD8\textsuperscript{+} s6, and CD8\textsuperscript{+} s7 cells increased, whereas the number of CD8\textsuperscript{+} s4, CD8\textsuperscript{+} s5, CD8\textsuperscript{+} s8, and CD8\textsuperscript{+} s9 cells decreased (Fig. 4d). Using immunohistochemical staining, we found that the number of CD8\textsuperscript{+} T cells increased after RFA treatment (Figure S2). The TCR analysis for CD8\textsuperscript{+} T cells showed that CD8\textsuperscript{+} s2 had the highest number of clonotypes in both control group and RFA group, and highest percentage of cells were CD8\textsuperscript{+} s1 occupied clonotype 1–5 (Fig. 5a and Table S2). There were three clusters (CD8\textsuperscript{+} s4, CD8\textsuperscript{+} s5, CD8\textsuperscript{+} s9) in RFA group that had several or none clonotypes owing to the lack of cells in these clusters. To gain further insight into the relationship among these nine CD8\textsuperscript{+} T clusters, we used Monocle 2 to analyze scRNA-seq data\textsuperscript{26}. Such pseudo-time analysis is a measurement through biological processes based on transcriptional similarities\textsuperscript{27}. Based on the expression signature, the starting point corresponded to the immune cells within CD8\textsuperscript{+} s4 and CD8\textsuperscript{+} s5, followed by CD8\textsuperscript{+} s9, and then CD8\textsuperscript{+} s2 and CD8\textsuperscript{+} s3, and ended with CD8\textsuperscript{+} s1, CD8\textsuperscript{+} s6, CD8\textsuperscript{+} s7, and CD8\textsuperscript{+} s8 (Fig. 5b, c). Depending on the expression signatures, the four clusters at the late period of pseudo-time were exhausted T cells. CD8\textsuperscript{+} s2 appeared to be a transitional state\textsuperscript{26} between naive and exhausted T cells based on trajectory and TCR analysis. These findings suggest that RFA treatment enhanced the antitumor immune response by promoting functional CD8 accumulation, and these effects were most evident in the antigen-specific CD8\textsuperscript{+} T cells.

**Changes in innate lymphoid cells and Mki67\textsuperscript{high} cells**

Single-cell RNA-seq identified one cluster of innate lymphoid cells (ILC\textsubscript{s}) (Fig. 4c), a cluster of cells lacking antigen-specific receptors and regulating immune response and tissue homeostasis\textsuperscript{28}, and two cluster of Mki67\textsuperscript{high} cells (Mki67\textsuperscript{high} s1 and Mki67\textsuperscript{high} s2) (Fig. 4c), which expressed patterns of genes associated with T-cell proliferation\textsuperscript{18}, including Ezrh2, Hmnr, Cenpa, and Mcm5 (Figure S1). Interestingly, cells of ILC\textsubscript{s} and Mki67\textsuperscript{high} s2 did not express Cd4 and Cdb\textsubscript{+} b1 genes, whereas Mki67\textsuperscript{high} s1 expressed both Cd4 and Cdb\textsubscript{+} b1 genes (Fig. 4b). In ILC\textsubscript{s} cluster, cells showed higher gene expression of I2ra, Icos, Gata3, and Rora (Figure S1), suggesting ILC\textsubscript{s} cluster was likely group 2 ILC\textsubscript{s}\textsuperscript{28,29}. The proportion of ILC\textsubscript{s} cells decreased dramatically in RFA group compared with the control group (Fig. 4d). In Mki67\textsuperscript{high} cluster, Mki67\textsuperscript{high} s1 contained a mixture of CD4\textsuperscript{+} T cells, CD8\textsuperscript{+} T cells, and Tregs based on their profound cell proliferation signature\textsuperscript{18}, whereas Mki67\textsuperscript{high} s2 lacked these cell signatures. The sub-populations of immune cells in Mki67\textsuperscript{high} s1 were also altered after RFA therapy. The number of CD4\textsuperscript{+} T cells and CD8\textsuperscript{+} T cells in Mki67\textsuperscript{high} s1 cluster were increased after RFA treatment (Fig. 5d, e), indicating RFA treatment may induce T-cell proliferation. Gene ontology and KEGG enrichment analysis revealed that Mki67\textsuperscript{high} s1 displayed upregulation of pathways associated with ribosome and T-cell activation, whereas Mki67\textsuperscript{high} s2 was associated with lysosome, T-cell activation and positive regulation of response to external stimulus pathways (Fig. 4e, f). These data suggest that RFA treatment could induce T-cell proliferation within the tumors.

**Changes in monocytes/macrophages**

Intratumoral monocytes/macrophages represented the largest proportion of infiltrative immune cells in both control and RFA treatment group as revealed by scRNA-seq UMAP plots (Fig. 6a). Tumor-associated macrophages that expressed high levels of Mrc1 (CD206) correlated with tumor immunosuppression, angiogenesis, metastasis, and relapse\textsuperscript{30}. In contrast, monocytes/macrophages that expressed high levels of Nos2 (iNOS) were associated with pro-inflammatory and antitumor functions\textsuperscript{18}. In this study, scRNA-seq
Fig. 6 RFA induces intratumoral monocytes/macrophages, DCs, and neutrophils remodeling. 

a UMAP plots from merged data of these subpopulations. 
b UMAP plots with annotated clusters of these cells. 
c Heatmap displaying normalized expression of select markers. 
d Percentage of cells in each cluster by condition. 
e Heatmap of gene ontology enrichment analysis enrichment by clusters.
identified five major monocyte/macrophage subpopulations (Fig. 6b). Mac_s5 were characterized by the absence of classical macrophage maturation markers such as Mertk, but expressed the highest levels of Ccr2, Ccr7, Itga4, and Plac8 (Fig. 6c), consistent with the monocyte phenotype. The percentage of Mac_s5 decreased dramatically after RFA treatment (Fig. 6d). Mac_s1 and Mac_s4 cells were distinguished as having a distinctly anti-inflammatory gene-expression profile. Mac_s1 expressed genes encoding Ccl2, and most strikingly, Cx3cr1 (Fig. 6c) and also expressed signatures concerning the ribosome biogenesis (Fig. 6e). The proportion of Mac_s1 also reduced obviously in RFA group (Fig. 6d). Mac_s4 cluster had higher expression of Ccl2, Ccl7, and Ccl12 genes (Fig. 6c) and increased slightly in RFA group (Fig. 6d). In contrast, Mac_s2 and Mac_s3 were characterized as having distinctly antitumor function markers. Mac_s3 had higher expression of Cxcl3, Ccr2, Il1a, and Il1b genes (Fig. 6c). Mac_s2 expressed high levels of Cxcl10, Gbp2, Irgm1, and Stat1, which are related to the pathways of antigen processing and presentation, as well as the response to interferon-beta and interferon-gamma (Fig. 6e). In addition, the expression of Nos2 was high in Mac_s2 and Mac_s3, whereas Mrc1 was low. (Fig. 6c). The proportion of cells in Mac_s2 and Mac_s3 increased after RFA treatment (Fig. 6d), especially the Mac_s2 cluster. In addition, the number of CD206+ cells had barely changed and iNOS+ cells increased after RFA treatment in Immunohistochemical analysis (Figure S2). Further, we applied the Monocle 2 algorithm to order monocytes/macrophages in pseudo-time to indicate their developmental trajectories. All macrophage clusters formed into a relative process in pseudo-time that began with Mac_s5 cluster (monocytes), followed by Mac_s1 and then become macrophages in Mac_s2, Mac_s3, and Mac_s4 (Fig. 7a, b). Cells in Mac_s2, Mac_s3, and Mac_s4 overlapped with each other in trajectory plot, indicating that these three clusters had transcriptional similarities and might transform into each other based on the intratumoral stimulations. These data demonstrate that RFA treatment could stimulate macrophages to participate in forming an anti-TME.

Changes in DCs and neutrophils

Single-cell RNA-seq revealed three distinct subsets of DCs, including one cluster of conventional DC (cDC_s) and two clusters of monocyte DC (MoDC_s1 and MoDC_s2) (Fig. 6a). cDC_s expressed high levels of Classical type 1 DC markers, including Batf3 and Irf8. Both MoDC_s1 and MoDC_s2 expressed high levels of monocyte DC markers, such as Fcgr1, Cd14, Fcer1g, Itgam (CD11b), and Sirpa (CD172). MoDC_s1 also had higher expression of Il2ra and Il7r, whereas MoDC_s2 expressed high levels of Ccl6, C1qb, and Pf4. All DC clusters displayed upregulation of the costimulatory molecules Cd80/Cd86 and expression of Il12b (IL-12) that drive T-cell proliferation and differentiation (Fig. 6c). The proportions of cDC_s and MoDC_s1 increased after RFA treatment, whereas the proportion of MoDC_s2 decreased (Fig. 6d).

Lately, scRNA-seq also identified a neutrophil cluster (Neu_s) that expressed genes encoding S100a9, Lcn2, G0s2, Ccl3, and Cd24a (Fig. 6a, c). These cells also express high levels of Cxcr2 and Cxcl2 genes, which are associated with tumor progression and metastasis. The percentage of Neu_s was reduced after RFA treatment (Fig. 6d). In summary, these findings demonstrate that RFA treatment increased the numbers of functional DCs, which stimulated T-cell proliferation and differentiation, and suppressed the function of tumor-associate neutrophils.

Discussion

There is an increasing body of evidence that RFA could stimulate transient distant immune response.
Combining this response with immunotherapy may stimulate enhanced antitumor activity. However, most studies focused on the tumor of RFA side to understand the changes in the TME. Our study has demonstrated that RFA could result in a strong T-cell-mediated immune response in distant tumors. To further assess the immune changes, we employed scRNA-seq and TCR profiles analysis approaches to profile immune cells infiltrating in distant non-RFA tumors after RFA treatment. In contrast to some previous studies that showed that incomplete ablation of tumors after RFA treatment. 

In conclusion, our study provides new insight into the transcriptional, molecular, and functional changes in the non-RFA distant tumors induced by RFA treatment, and discovered that T-cell exhaustion after RFA was likely the mechanism for the failure of the RFA to induce effective tumor shrinkage on the distant non-RFA tumors. However, our data suggest that immune checkpoint inhibitors, when combined with RFA, may have the potential to overcome the T-cell exhaustion and enhance the activities of the transiently increased infiltrating cytotoxic T-cell in the distant tumors to exert effective tumor shrinkage.

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**Data and code availability**

Single-cell RNA-seq data that support the findings of this study have been deposited in GEO with the accession code PRJNA611464 and Main R code for the analysis are available at GitHub.

**Conflict of interest**

The authors declare that they have no conflict of interests.

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