Title: HBEGF⁺ macrophages identified in rheumatoid arthritis promote joint

2 tissue invasiveness and are reshaped differentially by medications

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13	One Sentence Summary:
14	A newly identified human macrophage phenotype from patients with the autoimmune condition
15	RA is found to promote joint tissue invasiveness and demonstrates variable sensitivities to anti-
16	inflammatory medications used to treat the disease.
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1 Abstract:

2 Macrophages tailor their function to the signals found in tissue microenvironments, taking on a wide spectrum of phenotypes. In human tissues, a detailed understanding of macrophage 3 phenotypes is limited. Using single-cell RNA-sequencing, we define distinct macrophage subsets 4 in the joints of patients with the autoimmune disease rheumatoid arthritis (RA), which affects 5 $\sim 1\%$ of the population. The subset we refer to as HBEGF⁺ inflammatory macrophages is 6 enriched in RA tissues and shaped by resident fibroblasts and the cytokine TNF. These 7 macrophages promote fibroblast invasiveness in an EGF receptor dependent manner, indicating 8 9 that inflammatory intercellular crosstalk reshapes both cell types and contributes to fibroblastmediated joint destruction. In an *ex vivo* tissue assay, the HBEGF⁺ inflammatory macrophage is 10 targeted by several anti-inflammatory RA medications, however, COX inhibition redirects it 11 towards a different inflammatory phenotype that is also expected to perpetuate pathology. These 12 13 data highlight advances in understanding the pathophysiology and drug mechanisms in chronic inflammatory disorders can be achieved by focusing on macrophage phenotypes in the context of 14 complex interactions in human tissues. 15 16 17 18

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1 Introduction

Macrophage plasticity provides tailored homeostatic, immunologic and reparative mechanisms in 2 a wide-range of tissues (1, 2). Their transcriptional, epigenetic and functional versatility allow 3 macrophages to conform to tissue- and disease-specific factors, resulting in phenotypes 4 indicative of the type of tissue and physiologic state (3-9). While macrophages are a unifying 5 feature in chronic human diseases such as atherosclerosis, autoimmunity and granulomas (2, 10, 6 11), little is known about macrophage phenotypes in the context of human tissue pathology— 7 particularly at the single-cell level. Furthermore, while in vitro studies have provided valuable 8 9 insights into the range of macrophage polarization states (12, 13), the relevance of these wellcharacterized responses has been difficult to document in human tissues. 10

A precise understanding of human tissue macrophages may enable more effective 11 therapeutic decisions for inflammatory diseases, where it has often been difficult to discern 12 which molecular pathways to target. For example, in inflammatory bowel diseases cytokines 13 such as IL-17 and IFN- γ have been implicated in the pathophysiology, yet blockade of these 14 factors have produced variable results and in some cases worsens symptoms, whereas anti-TNF 15 therapies are commonly effective (14). In the autoimmune disease rheumatoid arthritis (RA) 16 nonsteroidal anti-inflammatory drugs (NSAIDS) treat pain and components of inflammation, but 17 for unclear reasons do not curb joint erosion (15), whereas anti-TNF therapies have proven 18 highly effective on both fronts. Macrophages, as innate immune cell types common to tissues 19 20 that are heavily influenced by microenvironmental factors, could serve in affected tissues as indicators of disease pathways and as targets of tailored treatments. Furthermore, in cancer, new 21 therapeutic strategies not only disrupt support of tumor growth and metastasis by macrophages, 22 23 but aim to repolarize them towards pro-resolution anti-tumor states (16, 17). To develop such

therapeutics for autoimmune and inflammatory conditions, an in-depth classification is needed for tissue macrophages and how medications redirect them, amidst the complexity of tissue and disease signals, towards beneficial or a differing yet pathologic state.

In the chronically inflamed RA joint tissue, macrophages are understood to be a source of TNF, a well-established driver of RA (*18-21*). However, the precise nature and variety of macrophages within the RA synovium is not yet defined, nor is the cumulative impact of surrounding intercellular interactions and medication-induced perturbations on macrophage responses in RA tissue pathology.

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10 **RESULTS**

Single-cell RNA sequencing detects HBEGF⁺ inflammatory macrophages in RA synovial tissue

To define the spectrum of macrophage phenotypes in a human tissue affected by autoimmunity, 13 14 as reported (22), we sorted synovial cells from ten RA patients and two osteoarthritis (OA) disease control patients by CD14⁺ cell surface protein expression and applied single-cell RNA 15 sequencing (scRNA-seq, CEL-Seq2). After stringent quality control filtering, 940 CD14⁺ single 16 cells clustered into four major CD14⁺ synovial cell subsets based on Canonical Correlation 17 Analysis (CCA) as described (22)(Fig. 1A). Cells from all four clusters expressed genes that 18 19 define the myeloid lineage, such as CD68, CD163 and C1QA and based on their high levels of 20 CD14 (Fig. S1A), we designated these cells macrophages rather than dendritic cells (7, 23). Clusters 1 and 2 contained the majority of the CD14⁺ synovial cells (45% and 30%, respectively) 21 and the largest number of genes that distinguished them from the other clusters, such as PLAUR 22 23 and *HBEGF* for Cluster 1 (referred to as the 'Cluster 1 HBEGF⁺) and *ADORA3* and *MERTK* for

Cluster 2 (Fig. 1B, Fig. S1B, 125 and 193 genes >2-fold, respectively). Cluster 3 appeared less
well defined by positive markers, whereas Cluster 4 had robust markers including IFN-target
genes such as *IFI6* and *IFI44L* and herein is referred to as the 'IFN/STAT' cluster (Fig. 1B, Fig.
S1B).

To estimate the abundance of the macrophage clusters in RA tissue, we sorted $CD14^+$ 5 synovial cell populations from 18 RA and 14 OA patients and analyzed them using bulk RNA-6 seq (~1,000 CD14⁺ cells from each patient) as reported (22). We detected 1,726 genes with 7 distinct expression patterns between the disease states (FDR adjusted p value <0.1). Enriched in 8 RA CD14⁺ cell populations, we detected elevated levels of genes and pathways that associate 9 with RA disease such as TNF, STAT1 and STAT4 and the response to TNF, Interferon and 10 Inflammation (Gene Set Enrichment Analysis (GSEA)) (Fig. 1C, left and right panel, 11 12 respectively) (24-27). Overlaying the RA versus OA comparison with markers from each of the single-cell clusters, we found that genes defining Cluster 1 HBEGF⁺ and Cluster 4 IFN/STAT 13 were consistently more abundant in RA CD14⁺ populations, suggesting an enrichment of these 14 cell subsets in RA (Fig. 1D). Cluster 2 genes were in higher ratios in OA tissues, while Cluster 3 15 markers showed no consistent association with either disease (Fig. 1D). Furthermore, RA CD14⁺ 16 populations were positively enriched in genes sets from Cluster 1 HBEGF⁺ and Cluster 4 17 IFN/STAT while negatively enriched in Cluster 2 genes (GSEA, FDR adjusted p-value 18 <0.001)(Fig. 1E, light grey bars). 19

To identify factors from the microenvironment that shape synovial macrophage phenotypes, the synovial CD14⁺ single-cell gene sets were compared to human blood-derived macrophages activated by diverse stimuli (Fig. 1E, colored bars, ranked gene lists). Cluster 1 HBEGF⁺ and Cluster 4 IFN/STAT genes were positively enriched in pro-inflammatory M1-like

Fig. 1



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Fig. 1. HBEGF⁺ inflammatory macrophage identification in RA joints by single-cell RNA-2 sequencing. (A) Synovial CD14⁺ single-cell RNA-seq clusters (940 cells) identified by 3 4 Canonical Correlation Analysis (22). (B) CD14⁺ single-cell cluster marker genes. Median expression indicated by color and percentage of expressing cells indicated by size. (C) 5 Differential gene expression for bulk CD14⁺ synovial cells from RA (n=16) versus OA (n=13) 6 patients plotted as log_2 fold-change with $-log_{10}$ FDR adjusted p-value; dark grey <0.1. Right: 7 Positively enriched pathways in RA bulk CD14⁺ cells. (**D**) CD14⁺ single-cell cluster genes (up to 8 100) highlighted on the bulk RA v. OA plot from b (Bonferroni corrected p-value < 0.1, 9 expressed in >30% of cells). Hypergeometric test: ***,* represent p $<10^{-6}$, $<10^{-3}$, respectively. 10 Right: Number of cluster genes higher in RA or OA bulk comparison or not significant (ns) 11 (FDR adjusted p-value < 0.1). (E) GSEA using the CD14⁺ single-cell markers as Gene Sets and 12 ranked gene lists from human blood-derived macrophages exposed to various stimuli (colored 13 bars) or the bulk CD14⁺ RA v. OA analysis (background grey bars). Normalized enrichment 14 score; |NES| > 2.5 were significant at FDR adjusted p < 0.001. (F) Gene expression level for each 15 cell, plotted as *log*₂ counts per million (CPM)+1. (G) *HBEGF* expression in patient CD14⁺ bulk 16 populations, plotted as log₂ transcripts per million (TPM)+1. n=16 RA, 13 OA samples. *, 17 Bonferroni corrected p value $<10^{-3}$. 18 19 20

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macrophage genes induced by TNF and anti-correlated with the IL-4 driven M2 antiinflammatory phenotype, indicating these two macrophage subsets are activated by proinflammatory factors (Fig. 1E, red v yellow bars). Conversely, Cluster 2 genes were negatively enriched for TNF-induced inflammatory response suggesting an anti-inflammatory phenotype, while the lack of enrichment with the IL-4 induced M2 state potentially indicates a novel tissue macrophage phenotype. The limited number of Cluster 3 positive markers did not significantly associate with any of the stimulated macrophage states.

As the most abundant cell type in the RA synovium (28) (Fig. S2A), synovial fibroblasts 8 can evoke large shifts in macrophage gene expression profiles (29). Importantly, the combination 9 of synovial fibroblasts together with TNF (TNF+Fib) generated a macrophage phenotype that 10 aligned with Cluster 1 HBEGF⁺ macrophages more than the other stimuli (Fig. 1E). This was 11 12 unlikely due to an additive effect that exacerbates the TNF response, as synovial fibroblasts and TNF can independently induce robustly opposing effects, for example in Cluster 4 genes (Fig. 13 1E). Thus, we posited that the abundant Cluster 1 HBEGF⁺ macrophages identified by single-cell 14 analyses from affected human tissue are driven by the combination of tissue-specific factors 15 from the resident synovial fibroblasts and inflammatory signals such as TNF found in the RA 16 synovium. 17

Consistent with enrichment in RA, the CD14⁺ Cluster 1 HBEGF⁺ single-cells expressed high levels of classic inflammatory genes like *IL1B* and low levels of M2-associating genes such as *MERTK* (Fig. 1B, F)(*13*). However, Cluster 1 HBEGF⁺ cells were also distinctively high in the expression of genes such as *HBEGF* (Heparin Binding EGF Like Growth Factor) and *PLAUR* (plasminogen activator, urokinase receptor) (Fig. 1F), which have largely been reported as marking anti-inflammatory phenotypes or cells derived from a combination of pro- and anti-

inflammatory triggers (*30, 31*). Considering this distinct program (Fig. 1G), we designate Cluster
1 macrophages 'HBEGF⁺ inflammatory macrophages'.

Importantly, in an independent study using a droplet-based single-cell RNA-seq platform 3 (Drop-seq) applied to five RA patient synovial tissues (Fig. S2A)(28), we have detected an 4 abundant macrophage subset with considerable overlap of marker genes with the HBEGF⁺ 5 inflammatory macrophages (Fig. S2B, 62%), including HBEGF, PLAUR, IL1B and CREM (Fig. 6 S2C). Overlap with the other CD14⁺ single-cell clusters was less clear, potentially due to 7 differences in the patient populations in the two studies. Nonetheless, through independent 8 single-cell platforms and patient cohorts, synovial macrophages of the HBEGF⁺ inflammatory 9 10 macrophage phenotype can be robustly detected in human tissue affected by RA.

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12 Tissue-resident synovial fibroblasts shape HBEGF⁺ inflammatory macrophages

To further define the HBEGF⁺ inflammatory phenotype, we compared the transcriptome of 13 macrophages exposed to TNF and synovial fibroblasts versus TNF alone. In a transwell co-14 culture system where both cell types were exposed to TNF (24h, n=4 donors), synovial 15 fibroblasts altered the expression of 3,709 macrophage genes (FDR adjusted p value < 0.1)(Fig. 16 2A). The majority of fibroblast-mediated effects opposed the direction of change brought on by 17 TNF alone (Fig. 2A, 69% of fibroblast-effects anti-correlated with TNF effects, upper left + 18 lower right). These changes included downregulation of TNF-inducible pro-inflammatory 19 20 mediators such as CFB, CXCL13, CCL8, MT1H, MMP2 and SLAMF7 (Fig. 2A, upper left, genes not labeled). Alternatively, fibroblasts upregulated M2-like anti-inflammatory factors, otherwise 21 suppressed by TNF, including MRC1, MSR1, and TREM1 (Fig. 2A, lower right, genes not 22 23 labeled). Pathway analysis indicated fibroblasts likely altered the metabolic state of TNF-treated

macrophages, indicated by a collective suppression of factors involved in Oxidative
Phosphorylation (Fig. S3A).

Despite the largely opposing effects of synovial fibroblasts on the macrophage TNF 3 response, fibroblasts enhanced or maintained induction of a substantial portion of TNF-induced 4 HBEGF⁺ inflammatory Cluster 1 genes, including *PLAUR* and *IL8* (Fig. 2A, upper right, red 5 dots). Synovial fibroblasts also induced a portion of HBEGF⁺ inflammatory genes that were 6 suppressed by TNF treatment alone, such as HBEGF, RGS2 and CREM (Fig. 2A, lower right, red 7 dots, and 2B). This indicates within synovial tissue, fibroblasts can also modulate macrophage 8 9 polarization by inducing a portion of Cluster 1 genes that would otherwise be downregulated upon TNF exposure. The induction of *HBEGF* by synovial fibroblasts peaked around 12 hour 10 and lasted over the course of days (Fig. 2C; qPCR representative of n=6 donors). The expression 11 of PLAUR and the RA-associated transcription factor STAT4, while elevated transiently by TNF 12 alone, was hyperinduced by synovial fibroblasts, resulting in a second elongated wave over days 13 (Fig. 2C). Enhanced gene expression correlated with an increase at the protein level for cell 14 surface PLAUR and intracellular STAT4 (Fig. 2D, E, respectively). 15

Cross-referencing the fibroblast-induced HBEGF⁺ macrophage profile with a panel of 16 previously reported macrophage polarization states (12) demonstrated a strong correlation with 17 macrophages treated with TNF and prostaglandin E2 (PGE₂) and/or the Toll-like receptor 2 18 (TLR2) ligand Pam3-Cys (referred to as 'TPP' (12, 32))(Fig. S3B, red text). This strong 19 20 correlation differed from the weaker or negative correlations with canonical M1 and M2 polarization states (induced by TNF or IFN- γ and IL-4 or IL-13, respectively)(Fig. S3B). From 21 these data, we hypothesized that PGE₂ mediated a considerable portion of the synovial fibroblast 22 23 effect on macrophages. Indeed, TNF-stimulated synovial fibroblasts (Fig. 2F, F+T) produced



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Fig. 2. HBEGF⁺ inflammatory macrophages polarization by tissue fibroblasts. (A) Human 2 blood-derived macrophage genes regulated by synovial fibroblasts and TNF (3,709 genes, FDR 3 4 adjusted p < 0.1; n=4 donors). Expression changes plotted as log_2 -fold. x-axis plots Fibroblasts+TNF v. TNF; y-axis plots TNF v. untreated. HBEGF⁺ Cluster 1 single-cell markers 5 labeled in red (expressed in >55% of cells, Bonferroni corrected $p < 10^{-6}$). (B) Expression of 6 7 select synovial CD14⁺ Cluster 1 genes in the blood-derived macrophages exposed to TNF (T) or fibroblasts + TNF (F+T). n=3 donors. (C) qPCR of blood-derived macrophages overtime, plotted 8 as percent (%) of GAPDH. Mean, standard error of the mean (SEM). Representative, n=4 9 donors. (D) PLAUR (CD87) surface protein detected by flow cytometry in blood-derived 10 macrophages, 24h. Representative, n=3 donors. (E) Western blot of STAT4 in blood-derived 11 12 macrophages, 24h. Representative, n=4 donors. *, non-specific band. Hsp90, loading control. (F) Prostaglandin E2 (PGE₂) ELISA on supernatants from blood-derived macrophages (M ϕ), 24h. 13 n=4 donors. Mean with SEM. (G) qPCR of blood-derived macrophages, 24h. prostaglandin E2, 14 (PGE₂). n=8 donors. (H) qPCR of blood-derived macrophages, 24h. Nap, COX inhibitor 15 naproxen (150nM). (I) ATAC-seq tracks from PLAUR gene promoter regions in blood-derived 16 macrophages, 3h. Change in open chromatin, orange bar. (J) FAIRE-qPCR of open chromatin in 17 region highlighted in subpanel I, % total input reported as mean SEM. n=4 donors. *, **, **** 18 represent p<0.05, 0.01, 0.0001, respectively. 19 20

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1	large amounts of PGE ₂ , whereas macrophages produced no detectable PGE ₂ irrespective of TNF
2	exposure (Fig. 2F, M Φ and M Φ +T). Furthermore, induction of HBEGF ⁺ genes such as <i>PLAUR</i>
3	and <i>HBEGF</i> in blood-derived macrophages was recapitulated by exogenous PGE ₂ and TNF, with
4	the fibroblast-mediated induction blocked by the COX enzyme inhibitor naproxen (Fig. 2G, H,
5	respectively). The transcriptional induction of PLAUR expression associated with chromatin
6	opening in the PLAUR promoter region, whereby TNF and prostaglandin synergized to open a
7	new region, detected by ATAC-seq and verified by FAIRE-qPCR (Fig. 2I, J, respectively).

8 These data suggest that in the RA synovium, chronic exposure to the pro-inflammatory 9 environment results in heightened synovial fibroblast production of prostaglandins that, together 10 with inflammatory factors, drive macrophages towards a state distinct from classical M1 and M2 11 polarization. Unique transcriptional regulators, cell surface markers, metabolic pathways and 12 chromatin modifications mark this HBEGF⁺ inflammatory state.

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14 **HBEGF⁺** inflammatory macrophages promote fibroblast invasiveness

To examine how HBEGF⁺ inflammatory macrophages may impact RA pathology, we focused on 15 their intercellular relationship with synovial fibroblasts. In an RNA-seq analysis, exposure to 16 HBEGF⁺ inflammatory macrophages altered 855 synovial fibroblast genes (FDR adjusted p-17 value <0.1), including induction of IL11, LIF, CSF3 (GCSF), IL33, IL6 and PLAU (Fig. 3A). 18 Pathway analyses revealed an upregulated EGFR response in the fibroblasts, observed with two 19 20 gene sets (one induced by EGFR ligands and a second blocked by an EGFR inhibitor)(Fig. 3B). Similar to the gene expression changes, we also detected increased GCSF and IL-33 protein 21 secretion in the presence of HBEGF⁺ macrophages (Fig. 3C). The increased molecular 22 production of neutrophil chemoattractants and growth factors, such as GCSF, led us to 23

investigate potential cellular effects of HBEGF⁺ inflammatory macrophages. We report neutrophil accumulation—a well-documented feature of RA joint fluid (*33, 34*) following coculture of HBEGF⁺ macrophages with synovial fibroblasts (Fig. S4A). Lastly, we found the expression of GCSF and IL-33 was sensitive to EGFR inhibition (AG-1478 (AG))(Fig. 3D), further implicating EGFR involvement in this inflammatory macrophage-fibroblast crosstalk system and identifying an additional therapeutic avenue to target HBEGF⁺ macrophage driven inflammation.

To further understand the EGFR-mediated gene expression in the synovial fibroblasts, we 8 9 investigated the levels of EGF ligands in this system. Synovial fibroblasts expressed only trace amounts of seven EGF ligands with no change in response to TNF and HBEGF⁺ macrophages 10 (Fig. S4B). In notable contrast, macrophages induced HBEGF and a second EGF ligand EREG 11 (epiregulin) upon combined exposure of TNF and synovial fibroblasts (Fig. S4B). While EREG 12 expression in the blood-derived macrophage system was higher than HBEGF (Fig. S4B), we 13 have placed more emphasis on HBEGF as the RA patient synovial macrophages exhibit 14 considerably higher expression of HBEGF (Fig. 1B). For EGF receptor expression, synovial 15 fibroblasts expressed two subunits (EGFR and ERBB2), while neither blood-derived 16

macrophages nor RA synovial macrophages expressed EGFR subunits (Fig. S4C, D).
Considering these data, we posit synovial fibroblast EGFR responses correspond to production
of EGF ligands by fibroblast-entrained inflammatory HBEGF⁺ macrophages.

EGFR signaling is a robust regulator of fibroblast motility (*35*) and in RA, synovial fibroblasts invade and destroy cartilage and bone (*36-39*). Thus, we next tested whether the heightened EGF response influenced fibroblast migration through extracellular matrix. Indeed, pre-incubation of fibroblasts with TNF and macrophages induced a pronounced increase in





MT

MT+AG





Fig. 3

2	Fig. 3. HBEGF ⁺ inflammatory macrophages promote EGFR-dependent synovial fibroblast
3	pathologic activity. (A) Human synovial fibroblast genes altered by blood-derived macrophages
4	during a TNF response (885 genes differentially expressed), RNA-seq. 48h. n=2 donors. x-axis:
5	log_2 fold-change by macrophages. y-axis: significance as the $-log_{10}$ FDR adjusted p-value. (B)
6	Fibroblast pathways regulated by macrophages under TNF conditions (GSEA and Ingenuity
7	Pathways Analysis (IPA). Macrophage-induced, brown. Macrophage-downregulated, maroon.
8	(C) ELISA assay on supernatants of synovial fibroblast (F) cultures with or without
9	macrophages (M) and TNF (T) for 48h. n=4 donors for both, reported as mean with SEM. (D)
10	qPCR on fibroblasts, 32h. EGF receptor inhibitor, AG 1478 (AG). Mean, SEM. n=6 donors. (E)
11	Synovial fibroblast Matrigel invasion assays, 18h after a 24h pre-incubation in with macrophages
12	(M), TNF and the EGFR inhibitor (AG, 4 μ M). n=3 donors. *, **, *** represent p <0.05, 0.01,
13	0.001 by paired Student's t-test, respectively. (F) Unique inflammatory response arises for
14	nearby synovial macrophages and fibroblasts, wherein [1] TNF induces [2] prostaglandin
15	production by fibroblasts that together with TNF in macrophages drives [3] HBEGF+
16	inflammatory phenotype and EGF ligand production that then feeds back to induce [4] fibroblast
17	invasive behaviors.
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fibroblast invasiveness that was mitigated by EGF receptor inhibition (Fig. 3E, F). These data
demonstrate that the HBEGF⁺ macrophage-dependent EGFR response may promote pathologic
fibroblast-mediated and pannus-associated tissue destructive behaviors in RA joints.

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5 RA medications target HBEGF⁺ inflammatory macrophage polarization

We next examined how clinically effective RA medications impact the polarization of HBEGF⁺ 6 inflammatory macrophages, using human blood-derived macrophages exposed to TNF and 7 synovial fibroblasts. In comparison to macrophages exposed only to TNF, the presence of 8 9 synovial fibroblasts substantially altered the impact of most medications (Fig. S5A). For example, the majority of auranofin targets were lost when macrophages were polarized towards 10 the HBEGF⁺ phenotype (1,300 genes, 67%) (Fig. S5A, blue versus yellow). In contrast, several 11 medications gained more than 1,000 gene targets in HBEGF⁺ inflammatory macrophages, 12 including naproxen (Nap) and leflunomide (active metabolite A77) (Fig. S5A). Thus, the 13 pronounced impact synovial fibroblasts confer onto macrophages includes alteration of drug 14 responsiveness. 15

Several medications reversed a large portion of fibroblast-induced effects (~1,000 genes), 16 suggesting the efficacy of these treatments could involve suppressed generation of HBEGF⁺ 17 inflammatory macrophage in RA synovium (Fig. 4A, black). This included leflunomide (A77), 18 dexamethasone (steroid), naproxen and Triple therapy (Tri, hydroxychloroquine + sulfasalazine 19 20 + methotrexate)(Fig. 4A, black). As naproxen inhibits COX enzyme-mediated prostaglandin production and affected the majority (~80%) of genome-wide fibroblast-mediated effects (Fig. 21 4A and Fig. S5A-C), this further supports prostaglandins as a dominant fibroblast product 22 23 driving HBEGF⁺ inflammatory macrophages. The drug-induced inhibition of HBEGF⁺

inflammatory macrophages, however, resulted in at least two functionally distinct states. 1 Naproxen, for example, blocked the prostaglandin arm driving HBEGF⁺ inflammatory 2 macrophages, but still permitted TNF polarization of the macrophages towards an M1-like pro-3 4 inflammatory phenotype, which presumably also functions pathologically in RA synovium (Fig. 5 S5A-C). Medications like dexamethasone were capable of blocking large portions of both programs (Fig. S5A). This result highlights the importance of understanding the resulting 6 phenotypes of macrophages targeted by medications, particularly in complex tissue settings 7 where multiple factors could redirect macrophages towards a differing yet pathologic state. 8

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RA medications target the HBEGF⁺ inflammatory macrophage-synovial fibroblast axis in patient samples

To examine the impact of therapeutically targeting HBEGF⁺ macrophages in RA joints, we 12 directly assayed synovial tissue from RA patients with definitive RA diagnoses (40, 41) and 13 extensive joint inflammation confirmed by histologic scoring (42, 43) (Table S1, n=8). 14 Specifically, in an *ex vivo* tissue assay, synovium was dissociated into cell suspensions (44) and 15 cultured with FDA-approved RA medications (Fig. 4B). Similar to the seminal assays that laid 16 17 the precedent for anti-TNF therapies in RA (45), as well as the documented clinical responses (46, 47), exposure to anti-TNF antibodies (α TNF, adalimumab) suppressed production of the 18 19 inflammatory mediator IL-1ß (Fig. 4C). Furthermore in this assay, the JAK inhibitor tofacitinib imparted striking reductions in genome-wide IFNa responses and protein levels for the 20 JAK/STAT target CXCL10 (IP10) (Fig. 4C). 21

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Fig. 4. Clinically effective RA medications and an FDA-approved EGFR inhibitor target the HBEGF⁺ inflammatory macrophage-fibroblast crosstalk signatures in RA tissue.

(A) Number of blood-derived macrophage genes affected by RA medications in the presence of 4 TNF and synovial fibroblasts, 24h. Black, genes opposed by drug. Grey, all other genes 5 regulated by drug. FDR adjusted p<0.1. n=2 to 4 donors. (B) RA patient synovial tissue ex vivo 6 drug response assay using highly inflamed synovium. Dissociated cells were exposed in culture 7 to a panel of medications, 24h. (C) ELISA on supernatants from RA tissue ex vivo assay. n=8 8 donors. *, **: p<0.05, 0.01, respectively by Wilcoxon signed-rank test. Lower panel: IFNa 9 response upon tofacitinib exposure, bulk RNA-seq and GSEA. n=2 donors. (D) Normalized 10 enrichment scores (GSEA) for drug-induced gene expression in the RA patient ex vivo assay for 11 12 $CD14^+$ single-cell Cluster markers. (E,H) ELISAs as described in c. (F) PGE₂ ELISA on supernatants. Mean, SEM. n=7 donors. *, p<0.05, paired Student's t-test. (G) qPCR on ex vivo 13 synovial cells treated with naproxen, plotted as percent (%) of TBP. n=7 donors. Mean, SEM. *, 14 p < 0.05, paired Student's t-test. (I) Schematic of differential effects of anti-TNFs and the COX 15 inhibitor naproxen on the two RA-enriched synovial macrophage subsets. (J) RA patient ex vivo 16 17 synovial expression changes with EGFR inhibitor AG-1478 (y-axis) compared to the synovial fibroblast gene regulation by macrophage and TNF (x-axis, data from Fig. 3A); n=2 donors. 18 FDR adjusted p < 0.1, plotted as a log_2 fold-change. Highlighted genes from Fig. 3A. 19

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1 Notably in this tissue explant assay, anti-inflammatory therapies such as anti-TNF, tofacitinib and dexamethasone inhibited the HBEGF⁺ inflammatory synovial macrophage 2 Cluster 1 gene set (Fig. 4D). Conversely, the panel of anti-inflammatory medications upregulated 3 4 the Cluster 2 gene set (with one exception, A77/leflunomide) (Fig. 4D), further suggesting Cluster 2 cells possesses anti-inflammatory features, which also aligns with high levels of anti-5 inflammatory genes MERTK, TGFBI and MARCO (48-50)(Fig. 1F, Fig. S5D). Tofacitinib was 6 the most potent inhibitor of the Cluster 4 'IFN/STAT' gene set, further implicating a robust 7 JAK/STAT response in this RA-associated macrophage subset (Fig. 4C). Interestingly, the 8 JAK/STAT response in Cluster 4 was induced by naproxen, as shown by enrichment in Cluster 4 9 genes and an elevated CXCL10/IP10 protein production in naproxen-treated RA synovial cells 10 (Fig. 4D-F). This correlated with reduced prostaglandins levels with naproxen and is in accord 11 12 with the established inhibitory effects of prostaglandins on IFN responses (51-53).

To test for evidence of the crosstalk module we have identified between HBEGF⁺ 13 inflammatory macrophages and synovial fibroblast EGFR responses (Figs. 2, 3), we further 14 examined the effects of naproxen in our tissue explant system. In addition to blocking 15 prostaglandin levels in the RA synovial explants and reducing HBEGF expression (Fig. 4G), 16 naproxen also reduced GCSF (CSF3) secretion, indicating a decoupling of the macrophage-17 fibroblast circuit (Fig. 4H). Importantly, as mentioned above and in Fig. S5 and B, naproxen 18 does not interfere with TNF-mediated effects and thus despite dampening prostaglandin levels in 19 20 the tissue assay (Fig. 4F), naproxen treatment did not suppress IL-1b protein production in the ex vivo synovial cell assay (data not shown). TNF blockade, however, was effective in blocking 21 HBEGF expression, as well as, IL-1b and GCSF protein levels (Fig. 4C, D) and thereby derails 22

generation of both the HBEGF⁺ inflammatory macrophages-synovial fibroblast axis and pro inflammatory M1-like macrophages (Fig. 4I).

To directly assay EGFR-dependent fibroblast gene expression changes consistent with 3 4 HBEGF⁺ inflammatory macrophage influence, we introduced an EGFR inhibitor into the *ex vivo* 5 assay. Importantly, the EGFR inhibitor reversed the majority of the synovial fibroblast gene expression effects elicited by HBEGF⁺ inflammatory macrophages (77%), including a 6 suppression of CCL20, CSF3, LIF, IL33, IL11, and PLAU (Fig. 4J, 3A). Together, these data 7 provide evidence within patient tissues for the crosstalk module we identified between HBEGF⁺ 8 9 inflammatory macrophages and tissue-resident fibroblasts and furthermore indicates blockade of 10 EGFR responses may provide a non-immunosuppressive therapeutic approach for RA.

11

12 **Discussion**

Challenges in treating autoimmune and inflammatory disorders have led to strategies including 13 administering medications sequentially until one is found to improve clinical symptoms. In an 14 effort to improve a priori knowledge of which medication will be most effective, we have taken 15 into consideration the complexity of cellular interactions in affected tissues and identified a 16 population of HBEGF⁺ inflammatory macrophages in inflamed RA joints and delineated their 17 deterministic tissue-resident and disease-specific factors. Furthermore, we have identified 18 medications that successfully interfere with the generation of these macrophages and the support 19 20 they provide towards fibroblast invasiveness, which contributes to irreversible joint destruction in RA. 21

The crosstalk mechanisms and functional output between macrophages and fibroblasts in the RA tissue environment differ from other pathologic states driven by these two cell types. In

1 fibrosis, macrophages and fibroblasts drive excessive collagen deposition blocking tissue function in part by static physical barriers rather than invasion and erosion (2, 54, 55). We find in 2 RA, fibroblast products such as prostaglandins combine with chronic inflammatory signals such 3 as TNF to polarize macrophages into a state that is distinct from inflammatory M1, anti-4 inflammatory wound healing M2 and pro-fibrotic TGF\beta-expressing macrophages. The RA-5 enriched HBEGF⁺ inflammatory macrophages produce a defined subset of inflammatory 6 products such IL-1 and the EGF growth factors HBEGF and epiregulin. As HBEGF⁺ 7 8 inflammatory macrophages subsequently induce invasiveness in synovial fibroblasts (Fig. 3F), we classify them as 'pro-invasive macrophages'. 9

10 While leukocyte-derived growth factors have emerged as critical components in tissue 11 homeostasis and repair (1, 56), our data uncover a pathologic correlate. In the RA tissue 12 environment, macrophage-produced EGF ligands lead to EGFR-dependent synovial fibroblast invasiveness and GCSF production concomitant with enhanced neutrophil accumulation (Fig. 13 3)—dominant features of RA joint pathology (33, 34, 37). In inflammatory arthritis in vivo 14 15 models, disease progression is driven by EGF ligands, EGF receptor activity and the enzyme iRhom2, which mediates release of both TNF and HBEGF from immune cells (57-60). Our data 16 provide relevance to these findings in human disease and a mechanistic understanding of the 17 cellular crosstalk program involving growth factors in RA joints. Furthermore, considerable 18 evidence implicates TNF and HBEGF as pathologic drivers of kidney disease in the autoimmune 19 20 condition lupus (61, 62). Thus, linked TNF and EGFR responses may be a unifying and targetable feature in tissues affected by disparate autoimmune conditions. 21

Using perturbations relevant to human disease, namely FDA-approved medications, we have detailed the disruption of intercellular interactions in patient samples and the resulting

1 phenotypes, thereby gaining insights into the complex consequences of medication in RA joint tissue. In particular, our data has yielded insights into a longstanding question of why anti-2 inflammatory COX inhibitors known as NSAIDs are not "disease-modifying" in RA (15). Our 3 data suggest NSAIDs block the prostaglandin-mediated arm in HBEGF⁺ inflammatory 4 polarization, but still permit macrophage TNF responses (Fig. 4I, Fig. S5B-C). Thus, NSAID 5 therapy in RA likely redirects HBEGF⁺ inflammatory macrophages towards a classic pro-6 inflammatory M1-like phenotype, which would presumably perpetuate inflammation albeit 7 through a different pathway. In that regard, NSAIDS may best be used in combination with 8 medications like anti-TNFs, in order to target two arms of HBEGF⁺ inflammatory macrophage 9 polarization. However COX inhibition by naproxen also proved problematic in the other RA-10 enriched macrophage phenotype, inducing the Cluster 4 'IFN/STAT' response (Fig. 4I), 11 12 consistent with robust suppression of IFN responses by prostaglandins (51, 53). Thus for two RA-enriched macrophage phenotypes, NSAIDS are permissive of pathologic responses. These 13 data speak to a greater need in understanding tissue microenvironment factors that polarize 14 macrophages and how in the presence of therapeutics the intercellular communication networks 15 are rewired and subsequently repolarize macrophages into states that either resolve or perpetuate 16 17 pathology.

Along with the seminal synovial explant assay that incited the use of TNF therapies in RA (45), our work impels the implementation of *ex vivo* assays to better understand and treat patients with autoimmune and inflammatory disorders. Specifically, in addition to detecting genome-wide established targets of anti-TNF and tofacitinib therapies, our RA patient *ex vivo* synovial tissue bioassay provides a human- and disease-relevant system that unmasked the interconnectivity and drug responsiveness of the synovial macrophage-fibroblast interaction we

1	identified. Human tissue-based therapeutic testing could offer guidance in relevant therapeutic
2	choices based directly on the response of the unique cellular composition of a patient's tissue.
3	For RA, this could be accomplished with the expanding use of synovial biopsies (21, 22, 63) and
4	ultimately over time with the identification of circulating biomarkers that correlate with tissue-
5	based assays. Lastly, effective blockade of the macrophage-induced fibroblast response in the
6	RA tissue ex vivo assay with an EGFR inhibitor developed for cancer warrants testing of this as a
7	new treatment direction, particularly as it could represent a non-immunosuppressive treatment
8	option.
9	
10	
11	
12	MATERIAL and METHODS
13	Study Design
14	The experimental objectives for this study involved directly analyzing primary human tissues to
15	understand how tissue and disease factors influence macrophage responses in the autoimmune
16	disease RA. This study included the use of tissue samples from patients treated for RA and OA,
17	in addition to healthy human blood donors. Inclusion and exclusion criteria for the arthritis
18	patients were based on standard clinical diagnostic criteria. Perceived outliers in sequencing
19	datasets remained in the analyses upon application of surrogate variable analysis by svaseq
20	3.26.0. Human donor blood-derived macrophage biologic replicates used for various assays
	depended on the robustness of the response for each type of assay and therein the amount of
21	
21 22	variability seen across donors, whereby for robust assay responses typically \sim n=4 donors were
21 22 23	variability seen across donors, whereby for robust assay responses typically \sim n=4 donors were used but for assays with higher variability up to n=8 donors were used.

1 Patient recruitment and CD14+ synovial cell sorting for RNA-seq

The multicenter RA/SLE Network of the Accelerating Medicines Partnership (AMP) consortium 2 enrolled individuals meeting the ACR 2010 RA classification according to protocols approved 3 by the institutional review board at each site (22, 44). Synovial tissues were collected from 4 ultrasound-guided biopsies or joint replacement surgery and viably frozen in Cryostor CS10 5 cryopreservation media (Sigma-Aldrich). At a central processing site, tissues were dissociated 6 and cells were FACS sorted (BD FACSAria Fusion) into fibroblast, macrophage, B cell and T 7 cell populations. Macrophages were sorted based on CD14+CD45+ cell surface expression. For 8 9 bulk CD14+ synovial cell population RNA-seq, ~1,000 cells were sorted directly into RLT buffer (Qiagen). For CD14+ synovial single-cell RNA-seq, ~100 live cells per patient were 10 individually plated and lysed in 384 well plates. 11

12

13 Bulk RNA-seq of sorted CD14+ synovial cell population

Full-length cDNA and sequencing libraries were generated using Illumina Smart-Seq2 protocol
 (64). Libraries were sequenced on a MiSeq System (Illumina) to generate 35 length base pair,
 paired-end reads.

17

18 Single-cell RNA-seq of sorted CD14+ synovial cells

Single-cell RNA-seq was performed on sorted macrophage and synovial fibroblast using the
CEL-Seq2 protocol (65) in 384 well plates. Libraries were sequenced on a HiSeq 2500
(Illumina) in Rapid Run Mode to generate 76 length base pair, paired-end reads.

22

1 CD14+ single-cell RNA-seq read alignment and differential gene expression analysis

RNA-seq reads were aligned with STAR version 2.5.2b (66) to the hg19 reference genome. 2 Transcript levels were quantified as Counts Per Million using the GENCODE Release 24 3 annotation. The single cell gene expression matrix was clustered based on a canonical correlation 4 analysis (CCA) methodology (22). Briefly, highly variable genes identified from single-cell 5 6 RNA-seq and bulk RNA-seq datasets were integrated based on genes that maximized the correlation between the two datasets. The correlated canonical variates were then used to 7 construct a nearest neighbor network thereby generating clusters that are verified to be present in 8 9 the bulk data. Using the four clusters identified from the CCA method, the top ten canonical coordinates were used to generate a Euclidean distance matrix for tSNE visualization using a 10 perplexity parameter of 40. Positive and negative cluster markers were identified using the 11 Wilcoxon rank sum test with a Bonferroni correction for multiple testing. 12

13

14 CD14+ bulk RNA-seq read alignment and differential gene expression analysis

Reads were aligned and quantified with STAR version 2.4.2a (*66*) against the GRCh38 genome and GENCODE Release 27 annotation, respectively. Differential expression analyses and batchcorrection were performed using DESeq2 (*67*) version 1.18.1 and svaseq (*68*) version 3.26.0, respectively.

19

20 Pathway analysis

Gene lists were processed for GSEA(*69*) version 3.0 by taking the inverse of the FDR adjusted p-value for each gene and multiplying it by the sign of the log2 fold change relative to the baseline conditions. GSEA was run under the "pre-rank" mode with 1,000 permutations for each

1 of the gene sets available in MSigDB and ImmunSigDB Version 5.2. Additional pathway analysis was performed using QIAGEN's Ingenuity® Pathway Analysis (IPA®, QIAGEN 2 Redwood City, www.giagen.com/ingenuity). Reported pathways were referenced as follows with 3 the MSigDB systematic name in parentheses: TNF-a Signaling via NFkB (M5890), Interferon-g 4 Response (M5913), Interferon-a Response (M5911), Inflammation (M5932), MYC Targets 5 (M5926, M5928), Oxidative Phosphorylation (M5936), Translation (M11989), Cell Cycle 6 (M543), Induced by EGF (M2613), Hypoxia Induced (M5891), Interferon Responsive Genes 7 (M9221), EGFR Inhibitor (M16010), and KEGG Ribosome Pathway (M189). Gene sets derived 8 9 from the CD14+ synovial single-cell RNA-seq markers were composed of up to 500 genes that exhibited $>0.5 \log_2$ fold change (positive marker) or $<-0.5 \log_2$ fold change (negative marker) 10 genes) relative to all other clusters, sorted by their fold change. 11

12

13 Independent RA arthroplasty cohort analyzed by droplet-based single-cell RNA-seq

In an independent analysis, we collected synovial tissue from five RA patients consented under 14 HSS RA Studies (IRB#2014-317 and 2014-233) during arthroplasty and synovectomy 15 procedures. Tissues were dissociated into single cell preparations and all cells were run through 16 17 Drop-seq protocol and sequenced on a HiSeq 2500 (Illumina)(28). Following cell and gene filtering(28), we applied Seurat version 2.3.0 to generate PCA based single cell clusters, which 18 were labeled based on cell-type markers. 20,031 single cells were visualized using the t-SNE 19 20 implementation in Seurat using a perplexity parameter of 20 and 13 principal components. After identifying a macrophage cluster consistent with our previous results (4,212 single cells), we re-21 applied Seurat and identified distinct subpopulations and visualized in t-SNE space. 22

1 Cell culture for human blood-derived macrophages and synovial fibroblasts

Human CD14+ monocytes were purified from leukocyte preparations purchased from the New 2 York Blood Center and differentiated into blood-derived macrophages for 1–2 d in 10 ng/ml M-3 CSF (PeproTech) and RPMI 1640 medium (Life Technologies)/10% defined FBS (Hyclone). 4 Cells were stimulated with 20 ng/ml recombinant human TNF (PeproTech). Drug treatments 5 were administered 15 minutes after TNF exposure to both the top and bottom wells of the 6 transwell system to achieve the stated final concentrations. After suspending in DMSO according 7 to the company's instructions, auranofin (Sigma Aldrich) was added to cells at 500 nM, A77 8 1726 (Santa Cruz) was added at 50 mM, Tyrphostin AG 1478 (Sigma Aldrich) at 4 µM, GW 9 627368X (Cayman Chemical) at 10 μ M, sulfasalazine (Sigma Aldrich) at 3 μ M, 10 hydroxychloroquine sulfate (Sigma Aldrich) at 50 µM, methotrexate (Cayman Chemical) at 110 11 µM, tofacitinib citrate (Cayman Chemical) at 1µM ENREF 61, and Pam3CSK4 (InvivoGen) at 12 1000 µg/mL. Dexamethasone (Sigma Aldrich), and prostaglandin E2 (Sigma) were first 13 suspended in absolute ethanol and then added to cells at 100 nM, and 280 nM, respectively. 14 Naproxen was suspended in RPMI supplemented with 10% FBS and then added to cells at a final 15 concentration of 100 uM. Anti-TNF was provided as adalimumab and was added to cells at a 16 17 final concentration of 50 μ g/mL.

18

Human synovial fibroblasts derived from de-identified synovial tissues of RA patients undergoing arthroplasty (HSS IRB#14-033)_ENREF_60. Dissociated cells were plated in aMEM based media up to 10 days, washing the media numerous times to remove dying blood cell components. Synovial fibroblasts at passages 4-6 were used for experiments. The diagnoses of RA were based on the ACR 2010 criteria. For Transwell culture experiments, synovial

fibroblasts adhered to polyester chambers with 0.4-µm pores (Corning) and were suspended above the wells containing macrophages, with a ratio of fibroblasts to macrophages of 1:16 based on the size of the cells and their coverage of the culture well surface. The number of donors used for each experiment is listed in the figure legend and refers to unique donors for both the blood-derived macrophages and for the synovial fibroblast lines.

6

7 RNA-seq for human blood-derived macrophages and synovial fibroblasts

Total RNA was first extracted using RNeasy mini kit (Qiagen). Tru-seq (non-stranded and PolyA 8 9 selected) sample preparation kits (Illumina) were then used to purify poly-A transcripts and generate libraries with multiplexed barcode adaptors. Single-end libraries were multiplexed, 10 pooled, and sequenced using the Single Read Clustering with 100 cycles for 3 lanes and 50 11 cycles for 6 lanes on an Illumina HiSeq 4000. Paired end libraries were multiplexed, pooled and 12 sequenced using the Paired End Clustering protocol with 51x2 cycles sequencing for 4 lanes on 13 an Illumina HiSeq 2500. Sequencing was performed by the Weill Cornell Medical College 14 Genomics Resources Core Facility. RNA-seq read alignment, quantification, differential testing, 15 and pathway analysis was performed as previously described. 16

17

18 Real-time PCR

RNA obtained using RNAeasy Mini kit (Qiagen) with DNAse treatment was reverse transcribed
into cDNA (Fermentas) and analyzed by real-time quantitative PCR (Fast SYBR Green; Applied
Biosystems) on a 7500 Fast Real-Time PCR System (Applied Biosystems). Gene-specific primer
sequences were as previously described_ENREF_63 or listed in the Table below. Expression
levels were normalized to GAPDH or TBP.

1

Gene		Forward Primer	Reverse Primer
AREG		GCTGCGAAGGACCAATGAGA	CCCCAGAAAATGGTTCACGC
BTC		GCTTGGCATTCCTTAAGCCC	GGTCCCTACCTGGTCTCTCC
CSF3		CCAGGAGAAGCTGGTGAGTG	GAAAAGGCCGCTATGGAGTT
EGF		GGATGTGCTTGATAAGCGGC	ACGGTCACCAAAAAGGGACA
EPGN		CATCAACGGTGCTTGTGCAT	ACAAAGGCCTCACAGTGGTC
EREG		ATCACAGTCGTCGGTTCCAC	AGGCACACTGTTATCCCTGC
GAPDH		ATCAAGAAGGTGGTGAAGCA	GTCGCTGTTGAAGTCAGAGGA
HBEGF		AGGAGCACGGGAAAAGAAA	CTCAGCCCATGACACCTCTC
IL1A		AGTAGCAACCAACGGGAAGG	AAGGTGCTGACCTAGGCTTG
IL33		TGAATCAGGTGACGGTGTTG	TGAAGGACAAAGAAGGCCTG
IP10		ATTTGCTGCCTTATCTTTCTG	TCTCACCCTTCTTTTTCATTGT
STAT4		GAGACCAGCTCATTGCCTGT	CAATGTGGCAGGTGGAGGAT
TGFa		CTCCTGAAGGGAAGAACCGC	CAGGCCAAGTAGGAAGGTCTG
PLAUR		GCTGCAACACCACCAAATGC	TTTTCGGTTCGTGAGTGCCG
PLAUR I	FAIRE	TCACTCTGTCACCCAGGCTA	GTGCCCCTGTAATCCCAGTT

2

3 Western blots on human blood-derived macrophages

4 Western blot analyses using a STAT4 (Santa Cruz) and HSP90 (Cell Signaling Technology)

5 antibody were performed using standard procedures with the additional step of adding Pefabloc

- 6 (Sigma-Aldrich) to macrophage cultures before cell lysis to prevent STAT protein degradation.
- 7

8 Flow cytometry on human blood-derived macrophages

9 Blood-derived macrophages were resuspended in 100uL FACS buffer (PBS with 1% FBS) and

10 stained with FITC-conjugated CD87/PLAUR (VIM5-Miltenyi Biotec), acquired by FACSCanto

11 (BD Biosciences) and analyzed using FlowJo (Tree Star, Inc.) software.

12

13 Assay for Transposase-Accessible Chromatin (ATAC)-Seq

- 14 ATAC-seq was performed as previously described(70). Human blood-derived macrophages were
- treated for 3 hours with 20ng/ml TNF (Peprotech) and/or 280nM PGE2 (Sigma-Aldrich). 50,000

1 cells were washed in PBS, lysed (lysis buffer: 10 mM Tris-HCl, pH 7.4, 10 mM NaCl, 3 mM MgCl2 and 0.1% IGEPAL CA-630) and centrifuged immediately at 500g for 10 min to obtain 2 nuclei. The pellet was resuspended in the transposase reaction mix (25 μ l 2× TD buffer, 2.5 μ l 3 transposase (Illumina) and 22.5 µl nuclease-free water) and incubated at 37°C for 30 min. DNA 4 was purified using a Qiagen MinElute kit and library fragments amplified for 13 cycles using 1X 5 NEB next PCR master mix and custom Nextera PCR primers(70). The libraries were purified 6 using Agencourt AMPure XP PCR Purification kit (Beckman Coulter) and single-end sequenced 7 on a HiSeq 2500 (Illumina). 8

9

10 FAIRE qPCR

FAIRE experiments were performed as previously described (71). Chromatin was crosslinked by 11 treating cells with 1% formaldehyde for 7 min and the reaction quenched with 0.125 M glycine 12 for 5 min. Cells were washed with cold PBS and scraped, followed by a second wash. Fixed cells 13 were lysed in buffer LB1 (50 mM HEPES-KOH, pH 7.5, 140 mM NaCl, 1 mM EDTA, 10% 14 glycerol, 0.5% NP-40, 0.25% Triton X-100, and protease inhibitors) for 10 min. Pelleted nuclei 15 were resuspended in buffer LB2 (10 mM Tris-HCl, pH 8.0, 200 mM NaCl, 1 mM EDTA, 16 17 0.5 mM EGTA, and protease inhibitors) and incubated on a rotator for 10 min. The nuclei were pelleted and lysed in buffer LB3 (10 mM Tris-HCl, pH 8.0, 100 mM NaCl, 1 mM EDTA, 18 0.5 mM EGTA, 0.1% Na-deoxycholate, 0.5% N-lauroylsarcosine, and protease inhibitors). 19 20 Chromatin was sheared using a Bioruptor Pico device (Diagenode). A total of 10% of sonicated nuclear lysates were saved as input. Phenol-chloroform-purified nuclear lysates and de-21 crosslinked input DNA were used for qPCR analysis using specific primers for the upstream 22 23 region. Chromatin accessibility is displayed relative to total input.

1 **Prostaglandin E2 EIA Kit**

Culture supernatants were collected from the culture after 24 hours, diluted 1:50 in plain RPMI,
and prostaglandin concentrations measured using the Prostaglandin E2 EIA Kit-Monoclonal
(Cayman Chemical). ELISA plates were read on Varioskan Flash Multimode Reader
(ThermoFisher Scientific).

6

7 Neutrophil viability assay

Supernatants were collected from cultures of human macrophages and synovial fibroblasts after 24 hours and frozen at -80 °C. Whole blood from healthy human subjects was collected into heparin coated tubes. Neutrophils were isolated with the EasySep Direct Human Neutrophil Isolation Kit (Stemcell Technologies) with "The Big Easy" EasySep Magnet. In a 12-well plate, 600,000 neutrophils were plated in 800 µL of supernatant for 24 hours. Neutrophil viability was measured using Muse Annexin V and Dead Cell Assay Kit on a Muse Cell Analyzer mini-flow cytometer (EMD Millipore).

15

16 Fibroblast invasion assay

Human synovial fibroblasts were plated in transwells with macrophages as in the transwell experiments. AG 1478 (Sigma Aldrich) was added at 4 μ M for 32 hours. Fibroblasts were trypsinized and re-plated in 500 μ L plain alpha-MEM with 10 ng/ml M-CSF at 0.1 x 106 cells per well into 24-well Corning BioCoat Matrigel Invasion Chambers. Macrophages were resuspended in 750 μ L plain alpha-MEM and seeded underneath invasion transwells in the appropriate conditions. AG 1478 was added at a concentration of 4 μ M; after 18 hours, the fibroblasts were

fixed for 10 minutes in ice-cold methanol and stained using crystal violet. Invasive fibroblast
numbers were quantified via light microscopy.

3

4 Human RA synoviocyte cultures

5 For synoviocyte cultures, RA patient synovial tissue was obtained from patients consented into the HSS FLARE study (IRB# 2014-233). Tissues were digested with Liberase TL (100 µg/mL, 6 Roche) and DNaseI (100 µg/mL, Roche) for 15 minutes and passed through three 70 µM cell 7 strainers. Cells were then suspended in 1 mL RBC Lysis Buffer (gift of J. Lederer, BWH) for 3 8 9 minutes followed by addition of RPMI/10%FBS/1%Glutamine to quench the reaction. Disaggregated synoviocytes were plated in RPMI/10%FBS/1%Glutamine at 0.2 x 106 in 96-well 10 plates. Cells were treated with drugs at aforementioned concentrations for 24 hours. Supernatants 11 and RNA were collected for Luminex experiments and quantitative PCR, respectively. For RNA-12 seq, the samples were multiplexed in eight samples per lane, 50 cycles, single-end reads, with 13 Truseq (Illumina) for library prep and a HiSeq 4000 (Illumina) ran in the Weill Cornell Medical 14 College Genomics Resources Core Facility. RNA-seq read alignment, quantification, differential 15 testing, and pathway analysis was performed as previously described(29). 16

17

18 Luminex

Supernatants were collected from macrophage-fibroblast cultures or synoviocyte cultures at 48
hours. Customized Luminex panels were ordered from R&D Systems. Protein concentrations
were read by a MAGPIX from EMD Millipore using xPONENT 4.2.

22

1 Statistical analysis

Data are shown as mean and standard error unless stated otherwise. Two-tailed paired t-tests 2 were performed for human sample derived qPCR and ELISA data using GraphPad Prism version 3 5.04 for Windows. Luminex multiplex ELISA experimental data was tested for normality by the 4 Shapiro-Wilks test with a significance threshold of p < 0.05 and was found to not follow a 5 Gaussian distribution. Subsequent statistical analysis was thus performed by Wilcoxon signed 6 rank tests, a non-parametric method, using GraphPad Prism. Single-cell RNA-seq clusters were 7 identified using a Canonical Correlation Analysis (CCA)(22). Markers for different clusters were 8 9 determined by Bonferroni corrected Wilcoxon rank sum tests implemented in Seurat version 2.3.0. Visualization of intersecting sets was performed using UpSetR version 1.3.3(72). Testing 10 for differentially expressed genes from bulk RNA-seq count data was performed using DESeq2 11 version 1.18.1 and surrogate variable analysis was performed using svaseq version 3.26.0, both 12 run on the R version 3.3.2. All RNA-seq significance levels are reported as False Discovery Rate 13 (FDR) adjusted p-values. Spearman and Pearson correlation analysis was performed using the 14 seaborn statistical data visualization package version 0.8.1 run on Python version 3.6.4. 15 Statistical significance of pathway analysis reported as the Normalized Enrichment Score and 16 FDR adjusted p-values determined from testing of 1,000 permutations from GSEA version 3.0. 17 Pathway z-scores for upstream regulatory analyses were performed using Ingenuity® Pathway 18 Analysis (IPA®, QIAGEN). All data transformations for visualization purposes, replicate 19 20 numbers, and statistical significance thresholds are reported in the Methods and Figure Legends.

21

22

1 Data Availability

RNAseq data have been deposited to database of Phenotypes and Genotypes (dbGaP) with the 2 accession numbers phs001340.v1.p1 and phs001529.v1.p1 as well as the Gene Expression 3 Omnibus (GEO) database with the accession numbers GSE57723, GSE95588, and GSE100382. 4 The CD14+ synovial single-cell data generated by the AMP consortium is housed at 5 http://immport.org with the study accession numbers SDY998 and SDY999. 6 7 **Supplementary Materials** 8 **Fig. S1.** Gene markers for synovial CD14⁺ single-cell clusters. 9 **Fig. S2.** Identification of synovial HBEGF⁺ inflammatory macrophages in an independent RA 10 11 patient study. Fig. S3. The transcriptome of human blood-derived macrophage exposed to synovial fibroblasts 12

13 and TNF compared with published pathways and macrophage polarization phenotypes.

Fig. S4. Synovial fibroblasts express EGF receptors while HBEGF+ inflammatory macrophages
 express two EGF ligands.

Fig. S5. RA medications impose highly variable effects depending on inflammatory
 macrophages depending on the presence of synovial fibroblasts.

18 **Table S1.** Baseline Characteristics.

19 **Table S2.** Participants Characteristics.

20

21

22

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