SINGLE CELL COMPENDIUM OF THE MUSCLE MICROENVIRONMENT IN PERIPHERAL ARTERY DISEASE REVEALS CAPILLARY ENDOTHELIAL HETEROGENEITY AND ACTIVATION OF RESIDENT MACROPHAGES

Short title: scRNAseq of human skeletal muscle in PAD

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ABSTRACT

Background: Peripheral artery disease (PAD) is caused by atherosclerosis and chronic narrowing of lower limb arteries leading to decreased muscle perfusion and oxygenation. Current guidelines for treating PAD include endovascular strategies or bypass surgery but long-term outcomes have been suboptimal. This is likely due to our limited understanding of the contribution of the microvasculature as well as other cell types, in particular macrophages, to PAD skeletal muscle pathophysiology. We used single cell sequencing to investigate cellular and transcriptional heterogeneity of the skeletal muscle microenvironment in PAD.

Methods: Samples from the medial head of the *gastrocnemius* muscle of individuals undergoing either lower limb aneurysm surgery (controls) or PAD bypass surgery (PAD) were collected. Samples were either frozen for histological evaluation (control: n=4; PAD: n=6) or were immediately processed for single cell RNA sequencing of mononuclear cells (control: n=4; PAD: n= 4). Bioinformatic tools were used to annotate cell types and their subpopulations, to study transcriptional changes and to analyze cellular interactions.

Results: We generated a dataset comprised of 106,566 high-quality, deep-sequenced cells that compose the muscle microenvironment. Focusing on endothelial cells (ECs) and macrophages, we confirmed the presence of ATF3/4$^+$ ECs with angiogenic and immune regulatory capacities in human muscle and found that their transcriptional profile profoundly alters during PAD. Also, capillary ECs display features of endothelial to mesenchymal transition. Furthermore, we identified *LYVE1*^{hi}MHCII^{low} resident macrophages as the dominant macrophage population in human muscle, even under a chronic inflammatory condition such as PAD. During PAD, *LYVE1*^{hi}MHCII^{low} macrophages get activated and acquire a more pro-inflammatory profile. Finally, we map strong intercellular communication in the muscle microenvironment, which is significantly altered in PAD.

Conclusions: The dataset we present here provides a highly valuable resource for gaining deeper insights into the critical roles that cells in the muscle microenvironment may play in PAD skeletal muscle pathology. We propose that targeting the crosstalk between ECs and macrophages could provide novel insights for developing effective treatments against this disease.

KEYWORDS

1. Peripheral artery disease, human, skeletal muscle, endothelial cells, macrophages, single cell RNA sequencing

Nonstandard Abbreviations and Acronyms

PAD, peripheral artery disease; IC, intermittent claudication; CLI, critical limb ischemia; ABI, ankle brachial index; scRNAseq, single cell RNA sequencing; EC, endothelial cell; MoMac; monocytes/macrophages.

Introduction

Peripheral artery disease (PAD) is a common disorder caused by atherosclerosis and chronic

narrowing of lower limb arteries leading to decreased muscle perfusion and oxygenation. Worldwide,

more than 200 million people are diagnosed with PAD and numbers are still rising[1–3]. PAD patients

suffer from pain, fatigue, and discomfort in the limb muscles due to exercise-induced ischemia which

resolves after a short period of rest, a syndrome termed intermittent claudication (IC)[4,5]. Moreover, if

left untreated, persistent lack of lower limb perfusion can lead to critical limb ischemia (CLI), which

comprises rest pain, tissue loss and necrosis, ulcerations, and gangrene. In fact, PAD is the most

common cause of lower limb amputations in the western world[1]. The ultimate goal of PAD treatment

is to restore perfusion of the ischemic limb. Current guidelines for treating PAD mostly target large

atherosclerotic lesions in the feeding arteries of the lower limb and include endovascular strategies or

bypass surgery, but long-term clinical outcomes have been suboptimal, and many patients still suffer

from adverse events[6–9].

Another proposed strategy to improve perfusion of the ischemic limb is to stimulate blood

vessel growth (therapeutic angiogenesis or revascularization). Medical or cell therapies that deliver

VEGF, the main angiogenic growth factor, and other angiogenic growth factors have been developed,

but despite great effort, no effective treatment is currently available[10,11]. This is largely due to

challenges in the dosing of growth factor delivery, due to our lack of knowledge on how to form

functional and *perfused* blood vessels in the muscle and due to our limited understanding of the

underlying skeletal muscle pathology in PAD[5]. Indeed, microvascular dysfunction is increasingly being

reported in PAD, with abnormal microvascular architecture and fibrosis in IC and CLI patients and few

studies recently reported partial endothelial-to-mesenchymal transition in PAD patients[12,13] and

capillaries of end-stage patients[13] . Also, impaired vasoreactive responses might impair oxygen and

nutrient delivery into the ischemic muscle[14]. In this respect, increased vascular density has been

observed in some[15–18] but not all[19,20] studies. While the discrepancy underlying these conflicting

observations is unclear, persistent ischemia in PAD patients despite increased vascularization

nonetheless suggests that an active angiogenic response does not restore microvascular perfusion nor

does it resolve tissue ischemia.

Several studies using preclinical models of muscle injury have shown that microvascular

endothelial cells (ECs) are not only essential for inducing muscle revascularization, but at the same

time steer muscle regeneration and tissue repair through their interactions with other cells in the

damaged or ischemic muscle microenvironment. ECs induce activation of muscle stem cells[21], control

vascular tone through interaction with smooth muscle cells[22] and have strong immune regulatory

properties by controlling the entry of immune cells into the tissue, their differentiation into specialized

immune effectors as well as (co-)defining the functional properties thereof[23,24]. However, our

understanding on how ECs interact with other cell types and might contribute to skeletal muscle

pathology in PAD, remains largely unknown.

Macrophages are key regulators of muscle regeneration in many settings[25] but also stimulate

muscle angiogenesis as well as arteriogenesis in preclinical models of PAD, suggesting that they are

key regulators of vascular function[23,26,27]. There is however very limited information on the genetic nor

functional properties of macrophages in skeletal muscle from patients suffering from PAD.

Inflammatory cytokine profiles of PAD muscle showed that higher levels of inflammatory cytokines

correlate with shorter onset claudication pain[28]. In PAD muscle, one manuscript showed that the

majority of macrophages express the pro-inflammatory marker CD80 whereas the fraction of

regenerative macrophages (using markers such as CD206, CD163 and Arg1) did not differ between PAD

patients and healthy controls[29]. In contrast, Kosmac et al[30] reported an increase in the fraction of

CD11b$^+$CD206$^+$ regenerative macrophages using a histological approach, and those macrophages were

positively correlated with fiber size and satellite cell number. On the other hand, less CD11b$^+$CD206$^-$

inflammatory macrophages was correlated with improved walking performance.

In this study, we have applied single cell RNA sequencing (scRNAseq) on muscle samples from

PAD patients and non-PAD controls, generating a comprehensive dataset containing over 100,000 cells

that provides new insights into the cellular complexity of muscle tissue under both healthy as well as

PAD conditions. We confirmed the presence of ATF3/4$^+$ ECs[31] with angiogenic and immune regulatory

capacities in human muscle and found that their transcriptional profile is profoundly altered during

PAD progression. Furthermore, we identified that $LYVE1$^{hi}MHCII^{low} resident macrophages are the

dominant macrophage population in human muscle, even under a chronic inflammatory condition

such as PAD. This population exhibits a high and active communication with ECs, which is significantly

altered in PAD. During PAD, $LYVE1$^{hi}MHCII^{low} macrophages get activated and acquire a more pro-

inflammatory profile. We hypothesize that resident macrophage activation could contribute to the

chronic inflammatory state observed in PAD[32–34] but their contribution to PAD muscle pathology needs

further study.

Methods

Patients and study design

Control and PAD patients were recruited from the Department of Angiology, Cantonal Hospital Baden, Baden, Switzerland. The general **inclusion criteria for both control and PAD patients** were: scheduled for surgery (bypass surgery in PAD patients or lower limb aneurysm surgery in control samples); Age ≥45 years; ability to understand German; signed informed consent to participate in the study. **Inclusion criteria specific for PAD patients**: diagnosis of PAD ranging from exercised limited by claudication (Fontaine classification: IIa or IIb) to CLI (Fontaine III-IV); ankle-brachial index (ABI) ≤ 0.90 at rest unless diagnosed with media sclerosis; stable medicament regimen (i.e., statin, anti-platelet, anti-hypertensive regimen). **Inclusion criteria specific for control patients**: diagnosis lower limb aneurysm; ABI > 1 at rest. The general **exclusion criteria for both control and PAD patients** were: severe peripheral neuropathy (Total Neuropathy Score ≥3); unstable angina or severe (>70%) three coronary vessel disease; significant liver dysfunction (Model for End-Stage Liver Disease Score) or known liver cirrhosis (Child-Turcotte-Pugh score >6 points); treated for cancer in the last 2 years; diagnosis of Parkinson's disease. **Exclusion criteria specific for PAD patients**: above/below-knee amputation. **Exclusion criteria specific for control patients**: peripheral artery disease. The study was conducted according to the Declaration of Helsinki, the Human Research Act (HRA) and the Human Research Ordinance (HRO) and the protocol was approved by the Ethics Committee of the Canton of Zurich (KEK number: 2020-01393).

Muscle samples from the medial head of *gastrocnemius* muscle were collected during lower limb aneurysm surgery (control samples) or bypass surgery (PAD samples) and immediately transferred to an ice-cold hypothermic preservation solution (HypoThermosol® FRS Preservation Solution, H4416 Sigma-Aldrich) for optimal transport and conservation until further processing[35]. From 4 PAD and 4 control samples, samples were processed for scRNAseq analysis. An independent set of samples (n= 6 for PAD; n=4 for control) was embedded in optimal cutting temperature medium, frozen in liquid nitrogen-cooled isopentane and stored at -80°C for histological analysis. Transversal 10-µm cryosections of the samples were prepared at -20°C, air dried for ~30 min and stored at -80°C until further processing.

Isolation of cells from human skeletal muscle for single cell RNA sequencing

For isolation of cells for scRNAseq, and in less than 2 hours after collection and preservation in

ice-cold hypothermic preservation solution (HypoThermosol® FRS Preservation Solution, H4416 Sigma-

Aldrich), samples were minced in a petri dish on ice using scissors and a surgical blade. Minced samples

were transferred into a digestion buffer (4 mL digestion buffer per 100 mg of muscle) containing 2

mg/ml Dispase II (D4693 Sigma-Aldrich) and 2 mg/ml Collagenase IV (17104019 Thermo Fisher

Scientific) in HBSS buffer (14025100 Thermo Fisher Scientific) at 37°C for 25-30 minutes, with gentle

shaking every 3-5 minutes. After digestion, the reaction was stopped by adding an equal volume of

20% heat inactivated FBS (10500064 Thermo Fisher Scientific) 1mM EDTA (E177 VWR Life Science) in

HBSS (14025100 Thermo Fisher Scientific) and the suspension was passed through a 100-μm cell

strainer (800100 Bioswisstec) and a 40-μm cell strainer (352340 Falcon). Cell suspension was

centrifuged at 500 g for 5 minutes at 4°C and the supernatant discarded. The pellet was resuspended

in 1 mL of ACK Lysing Buffer (A1049201 Thermo Fisher Scientific) for red blood cell lysis for 2 minutes

at room temperature. After incubation, the reaction was stopped with 4 mL of 10% heat inactivated

FBS (10500064 Thermo Fisher Scientific) in HBSS buffer (14025100 Thermo Fisher Scientific). Cell

suspension was centrifuged at 500 g for 5 minutes at 4°C and the pellet resuspended in 0.5% BSA

(A1391 PanReac AppliChem) in DPBS (14190250 Thermo Fisher Scientific) containing 1mM EDTA (E177

VWR Life Science) for FACS. Calcein Violet (65-0854-39 Thermo Fisher Scientific) was added directly to

the cell suspension at 1:1000 dilution 10 minutes before sorting.

Human scRNAseq analysis

Metabolically active (Calcein$^+$) human muscle mononuclear cells were FACS sorted and loaded

into the 10X Chromium Next GEM Chip G, aiming the recovery of ~10000 cells per sample. scRNAseq

libraries were generated according to 10X Chromium Next Gem Single Cell 3' Reagent Kits v3.1

(CG000204 Rev D) and sequenced on the Illumina Novaseq 6000 system. We sequenced a total of 13

samples from 8 different patients (Table 1). Sequencing reads were aligned to the human genome

using CellRanger v5.0.0-7.0.0 and the genome assembly GENCODE GRCh38.p13 Annotation Release

37. Intron reads were not included during the alignment. Each individual sample was independently

processed for quality control (QC) and log-normalization using *scran* 1.24.1 and *scater* 1.24.0

packages[36,37] in R (version 4.2.0). High quality cells were selected based on library size, number of

detected genes and percentage of reads mapped to mitochondrial genes, independently adjusting the

thresholds for each sample. Raw UMI counts were log-normalized based on library size factors.

Doublets were detected and removed in every individual sample using the default parameters of the

function *runDoubletFinder* from *singleCellTK* 2.6.0 package[38]. After individual QC and log-

normalization, all samples were batch-corrected and merged using *Harmony* 0.1.1 package[39].

Dimensionality reduction by t-Distributed Stochastic Neighbor Embedding (TSNE) was performed

based on the dimensions generated by *Harmony* integration. Clustering was performed by building a

nearest neighbor graph (k = 20) and then applying the Louvain algorithm as implemented in *Seurat*

4.1.1 package[40]. For individual analysis and reclustering of the EC and monocyte/macrophage

populations, cells were selected by tracing back the clustering of the complete dataset in every

individual sample. Raw counts from selected cells, either ECs or monocyte/macrophage, were log-

normalized in every individual sample and batch-corrected and merged as described above for the

complete dataset. All dotplots showing gene expression were generated using the *plotDots()* function

as implemented in scater 1.24.0 package. Cell type proportions in each condition were calculated by

dividing the number of cells in each cluster by the total number of cells in that respective condition.

Differentially expressed genes (DEGs) were detected by performing a Wilcoxon Rank Sum test

as implemented in *Seurat* 4.1.1 package. DEGs were selected based on a minimum $\log_2$ fold change of

0.25 and an adjusted p-value < 0.05. For over-representation analysis (ORA), selected DEGs were

ordered based on increasing adjusted p-value and given as an input in the web-based implementation

of g:Profiler[41], selecting the option "Ordered query" before running the analysis. ORA results were

represented as the -log10 of the p-values obtained for each process. ORA analysis results were

simplified by using *rrvgo* 1.10.0 package[42]. Redundant terms were grouped and only the most

significant term of each group was conserved. Schematic visualization of KEGG pathways from ORA

results were generated using *pathview* 1.38.0 package[43].

For pseudobulk analysis, lowly expressed genes were removed if they were not detected in at

least 10 cells. Raw UMI counts were aggregated per patient (total sum of counts for every gene from

all cells in every patient). For differential expression analysis in the pseudobulk dataset, we applied

DESeq2 1.38.3 package[44] and performed the comparison between control and PAD patients. All

significant DEGs (adjusted p-value <0.05) were reported in a heatmap (Figure 1J) by using the

normalized counts (by size factors) in each patient and scaled and centered based on gene expression

per patient. For Gene Set Enrichment Analysis (GSEA) in the pseudobulk dataset we applied *fgsea*

1.24.0 package[45]. From the differential expression analysis results performed with DESeq2 we selected

the *stat* value (Wald statistic: the log2FoldChange divided by lfcSE) for each gene and performed GSEA

analysis using the default settings of *fgsea()* function from *fgsea* 1.24.0 package. Significant pathways

were selected based on an adjusted p-value < 0.05. Human Hallmark pathways from MsigDB[46] were

used for this analysis.

Gene Set Variation Analysis was performed using *GSVA* 1.46.0 package[47]. Log-normalized

counts from the corresponding scRNAseq dataset (either ECs or Monocytes/Macrophages) were given

as an input for the *gsva()* function as implemented in GSVA 1.46.0 package. The gene sets to perform

GSVA were obtained from Fan et al[31] (ATF4 dependent genes as defined by genes with logFC < -0.1 and

adjusted p-value < 0.05 after comparing mouse muscle ECs in wild-type and EC-specific inducible ATF4

knockout mice) or Sanin et al[48] (macrophage activation processes). For GSVA of M1-M2 signatures, we

obtained the genesets from Varga et al[49], which reported a set of M1-M2 genes derived from *in vitro*

stimulation and generated genesets for pro-inflammatory ($Ly6C^+$) and pro-regenerative ($Ly6C^-$)

macrophages at different timepoints after cardiotoxin-induced muscle injury.

Bulk RNAseq data from muscle ECs in WT and EC-specific ATF4 KO mice were obtained from

Fan et al[31] and processed as described by the authors. Only WT and ATF4 KO red muscle ECs (RmECs)

were used in this study.

Transcription factor inference analysis was performed using the Python implementation of

SCENIC package (pySCENIC)[50,51] and following the script provided by the authors in their GitHub

repository (https://github.com/aertslab/pySCENIC). Cell communication analysis on the complete

dataset (all cell types in control and PAD samples) was performed using *CellChat* 1.6.1 package[52]

following the script provided by the authors in their GitHub repository (*Full tutorial for comparison*

analysis of multiple datasets section in https://github.com/sqjin/CellChat). Cell communication

analysis between ATF3/4$^+$ ECs and *LYVE1*^{hi}MHCII^{low} macrophages was performed using *nichenetr* 1.1.0

package[53] following the script provided by the authors in their GitHub repository (*Differential NicheNet*

analysis between conditions of interest section in https://github.com/saeyslab/nichenetr).

All raw and processed (10X CellRanger output files) sequencing data are available at

GSE235143.

Tissue Immunofluorescence and Histology

Human muscles were harvested and embedded in OCT embedding matrix (6478.1, Carl Roth)

and frozen in liquid N_2-cooled isopentane. Frozen sections (10 µm) were made using a cryostat (Leica

CM 1950). The following antibodies were used: anti-LYVE1 (1:200 dilution, ab33682, Abcam), anti-

CD68 (1:100 dilution, ab955, Abcam), anti-CD31 (1:200 dilution, M082329-2, Dako), Goat anti-Rabbit

IgG Cross-Adsorbed Secondary Antibody, Alexa Fluor 568 (1:200 dilution, A-11001, Thermofisher),

Goat anti-Mouse IgG Cross-Adsorbed Secondary Antibody, Alexa Fluor 488 (1:200 dilution, A-1101,

Thermofisher), Alexa Fluor 647 conjugated wheat germ agglutinin (WGA, 1:50 dilution, W32466,

ThermoFisher). Muscle sections were fixed in 4% paraformaldehyde (PFA) followed by epitope

retrieval using sodium citrate (10 mM, pH 6.5) at 92 °C for 20 min. Endogenous peroxidase activity was

blocked with 3% hydrogen peroxide in phosphate-buffered saline followed by 10 min of

permeabilization (PBS + 0.5% Triton X-100) and 60 min of blocking at room temperature in blocking

buffer (PBS + 1% BSA). After blocking, sections were stained overnight at 4°C with primary antibodies,

followed by 1 hour incubation of secondary antibodies at room temperature. Extracellular matrix was

stained using wheat germ agglutinin (WGA). Nuclei were detected using Hoechst diluted 1:5000 in PBS

and for 5 minutes at room temperature. Slides were then washed with PBS and mounted using Immu-

Mount (9990412, Thermofisher). Slides were imaged on an AxioObserver.Z1 fluorescence microscope

using a 20x and 40x objective (Carl Zeiss, Oberkochen, Germany). For the LYVE1 and CD68 staining,

315 ± 106 (mean ± standard deviation [SD]) fibers were counted per patients, and for the CD31

staining, 409 ± 80 (mean ± SD) fibers were counted. Different regions over the whole tilescan muscle

section were analyzed with the Zen software (ZEN 2011 imaging software, Zeiss) and Fiji. General tissue

morphology was evaluated using haematoxylin and eosin (H&E) staining. H&E images were acquired

with an Eclipse Ti2 inverted microscope (Nikon) using a 20x objective.

Statistics

The histology images presented in the manuscript are representative of the data

(quantification of images is approximately the group average) and the staining quality. All imaging data

are represented as violin plots, where the three horizontal lines represents the median and quartiles,

and the width of the plot indicates the distribution of the data. GraphPad Prism software (version

9.2.0) was used for statistical analyses. Shapiro-Wilk test was performed to analyze data distribution.

Non-normally distributed data were analyzed by non-parametric Mann-Whitney U test to compare

two groups. Normally distributed data were analyzed using Student's t-test in an unpaired two-tailed

fashion to compare two groups. $P > 0.05$ is considered non-significant (ns).

Results

scRNAseq reveals a high degree of cellular heterogeneity in human skeletal muscle

To investigate the cellular landscape of patients who have been diagnosed with PAD, we
collected samples from the medial head of the *gastrocnemius* muscle of individuals undergoing either
lower limb aneurysm surgery (controls) or PAD bypass surgery (PAD) (see Methods). A subset of patient
samples was used for scRNAseq (4 control vs 4 PAD) while an independent subset was frozen down for
histological analysis (4 control vs 6 PAD (Figure 1A, Table 1). Histological evaluation revealed that
myofiber size nor diameter was affected in PAD patients (Figure 1B-D). Also, we did not observe signs
of muscle necrosis nor regenerating fibers characterized by centrally localized nuclei. Since there have
been conflicting observations with some studies reporting reduced[19,20] or higher[15–18] vascular density
in PAD, we also stained for CD31[+] ECs and found increased vascular density in our specific cohort
(Figure 1E-F).

To obtain better insight into the molecular characteristics of PAD muscle, we performed
scRNAseq in an independent subset of patients (Table 1). Those samples were processed for tissue
digestion, and metabolically active mononuclear cells (Calcein[+]) were isolated using fluorescence
activated cell sorting (FACS). Sorted cells were then loaded into the 10X Chromium technology to
perform scRNAseq (Figure 1A). Following sequencing, each individual sample underwent quality
control (QC) filtering to select high quality cells, and raw counts were log normalized. The resulting QC
filtered and log normalized samples were pooled together and batch corrected. The final dataset was
comprised of 106,566 high-quality, deep-sequenced cells, as indicated by an average of 2359 genes
detected per cell and an average of 10.84% of reads mapped to mitochondrial genes. After initial
bioinformatic analysis, clustering, and t-Distributed Stochastic Neighbor Embedding (TSNE) we
detected 9 major cell populations (Figure 1G). As revealed by unbiased marker gene detection, the
dataset comprised heterogeneous and defined cell populations that we manually annotated based on
the expression of well-known marker genes[54–56] (Figure 1H). We still detected approximately 10-15%
of myonuclei, similar to other datasets[57,58], but our dataset was nicely enriched for mononuclear cells.
We identified the main cell types found in skeletal muscle, including several immune cell populations
(macrophages/monocytes, mast cells, neutrophils, NK/T/B cells), endothelial cells (ECs), mural cells
(smooth muscle cells (SMCs)/pericytes), stem cells (fibro-adipogenic progenitors (FAPs), and muscle
stem cells (MuSCs)/satellite cells). All cell types were present in both control and PAD patients and no
obvious differences in the fractional contribution of a cell type to the overall sample were detected
between groups (Figure 1I).

To get insight into the general transcriptional differences between conditions, we compared

the transcriptomes of all patients by a pseudobulk analysis[59]. As shown in Figure 1J (and Supplemental

File 1), control and PAD samples were separated in two distinct clusters based on differential

expression of the top high variable genes in the dataset, showing a clear differential transcriptomic

signature between both conditions. Gene Set Enrichment Analysis (GSEA) using Hallmark pathways[46]

over this comparison revealed an increase in tumor necrosis factor alpha (TNFA) signaling and

interleukin-6 (IL6) signaling, both associated with poor prognosis and reduced walking performance in

PAD patients[60,61] (Figure 1K). Beyond inflammatory processes, GSEA also showed higher enrichment

score of several metabolic pathways such as oxidative phosphorylation, glycolysis and mTORC1

signaling in PAD patients, while bile acid metabolism was repressed (Figure 1K). Interestingly, several

of the identified pathways such as the inflammatory activation, unfolded protein response and

metabolism were also differentially regulated in previously published bulk RNAseq experiments of PAD

muscle[62,63]. In contrast to our mononuclear single cell approach, those bulk tissue datasets are

commonly dominated by myonuclear gene pools, typically >70% of nuclei in skeletal muscle are of

myonuclear origin[64], thus suggesting that common pathways might be affected in different cell types.

Our initial clustering (Figure 1G) only partially addressed the high degree of heterogeneity of

the dataset and provided a significant source for exploration of the molecular differences between

PAD and control samples, thereby potentially unravelling molecular mechanisms involved in PAD

progression. All transcriptomic data are made available for download (GEO Accession Code:

GSE235143). Moreover, further subcellular clustering will identify many subpopulations within each

major cell type. For the purpose of this manuscript, we decided to conduct a comprehensive analysis

on ECs and macrophages because of their crucial role in angiogenesis, inflammation, and tissue repair,

which are key factors in the pathogenesis and progression of PAD[29,65,66].

Human skeletal muscle contains ATF3/4+ endothelial subpopulations with

immunoregulatory capacities

Using our initial clustering, we selected ECs from every individual QC filtered and log

normalized sample. Following the same approach we used for the complete dataset, we batch

corrected and pooled all the samples together. The dataset comprised 14355 high-quality ECs with an

average of 2120 genes detected per cell and an average of 10.54% of reads mapped to mitochondrial

genes. Clustering and TSNE projection revealed 6 different EC clusters (Figure 2A). Previous scRNAseq

studies in skeletal muscle[31,67] have established marker genes for the main EC types (arteries, capillaries,

veins, and lymphatic ECs) which we used to annotate the EC populations (Figure 2B). Additionally,

unbiased marker gene detection revealed specific gene sets for every individual population,

demonstrating that we identified well defined EC subtypes (Figure 2C and Supplemental File 1). Our

clustering however revealed two distinct populations of venous and capillary ECs that could not be

distinguished by general endothelial markers (Figure 2B) but clearly separated in the TSNE graph and

in the marker gene heatmap (Figure 2A,C). Interestingly, differential expression analysis between both

venous clusters detected a common set of genes significantly upregulated in venous 1 and capillary 1

populations but not in the other EC subtypes (Figure S1A). For a functional characterization of these

venous 1 and capillary 1 populations, we then performed an over-representation analysis (ORA)[41] to

identify the main processes in which these genes are involved. Interestingly, the top enriched pathways

identified by the ORA analysis using the KEGG pathway database[68] yielded a collection of processes

that have previously been linked to recruitment and functional modulation of immune cells by ECs

including TNF signaling, IL-17 signaling and NF-kappa B signaling pathways[24,69,70] as well as a lipid and

atherosclerosis signaling (Figure 2D). A deeper analysis of the TNF signaling and lipid and

atherosclerosis signaling pathways revealed that the upregulated genes in these ECs are immune

regulatory genes directly involved in leukocyte recruitment, leukocyte activation, cell adhesion and

cytokine signaling (Figure S1B-C).

ORA analysis for transcription factor (TF) motif enrichment identified ATF4 as a top TF

regulating the expression of these genes (Figure 2E). In a prior study, we identified a subpopulation of

Atf3/4+ muscle ECs in mouse using scRNAseq and subsequently found that ATF4 is functionally relevant

by showing its requirement for exercise-induced angiogenesis by controlling a set of genes involved in

amino acid uptake and metabolism[31]. Similarly, ATF3 is required for regeneration of the aortic

endothelium[71]. First, direct comparison of the marker genes defining the mouse *Atf3/4+* subpopulation

based on the scRNAseq experiment with the current human dataset showed that those were highly

enriched in both the venous 1 and capillary 1 clusters from human skeletal muscle (Figure S1D). In our

previous work, we also performed a bulk RNAseq experiment in ECs isolated from mouse skeletal

muscle of wild type and EC-specific inducible ATF4 knockout (KO) mice. This experiment revealed a

specific set of endothelial ATF4-dependent genes that were downregulated upon EC-specific loss of

ATF4[31]. When mapping this ATF4-dependent gene set against the genes that we found upregulated in

the identified human venous 1 and capillary 1 clusters (Figure S1A), we noticed that the majority

thereof were controlled by ATF4 (Figure S1E). We also used the ATF4-dependent gene set to perform

gene set variation analysis (GSVA)[72] which calculates a score for each cell based on the expression

levels of an input gene set (e.g. ATF4-dependent genes) relative to the whole transcriptome of the cell.

Within our human dataset, the ATF4-dependent GSVA score was higher for venous 1 and capillary 1

clusters compared to the other clusters (Figure 2F). Furthermore, we performed unbiased TF regulator

network inference using SCENIC[50,51] which generates a *TF regulon activity score* for each cell based on

the coexpression of TFs and their target genes that are identified by the presence of TF binding motifs

in their promoter regions. SCENIC revealed that ATF4-dependent transcription factors (such as ATF3,

FOS, FOSB, JUN, JUNB and JUND) (Figure S1E) showed the highest transcriptional activity in ECs and

their regulons were specifically active in venous 1 and capillary 1 clusters (Figure S1F). In conclusion,

venous 1 and capillary 1 populations are ATF3/4+ ECs and we thus annotated these populations as

ATF3/4+ venous ECs and ATF3/4+ capillary ECs. Interestingly, when looking at the fractional

contribution of EC subpopulations, we found a reduced fraction of ATF3/4+ capillaries in PAD samples

when compared to control (Figure 2G), suggesting that these cells could contribute to the progression

of the disease.

ECs undergo profound transcriptomic rewiring during PAD

Next, we assessed whether the different EC populations showed transcriptomic changes

between control and PAD samples. Therefore, we calculated the number of genes that are

differentially expressed between conditions for each individual population and, at the same time,

specific for that cluster compared to the rest (Figure 2H and Supplemental File 1). Both ATF3/4+ venous

and capillary ECs showed the highest number of differentially expressed genes between control and

PAD samples (Figure 2I) indicating that these two populations are profoundly modifying their

transcriptome during PAD progression. To further investigate the transcriptomic switch in PAD samples

at the pathway level, we selected genes that were up- or downregulated in both ATF3/4+ venous as

well as capillary ECs (Figure S2A and Supplemental File 1) and performed an ORA analysis using the

biological process category from the Gene Ontology (GO) database[73,74]. Interestingly, while few of

enriched processes were activated in the PAD ATF3/4+ populations (Figure S2B and Supplemental File

1), most of them were downregulated indicating that many of the biological functions exerted by the

ATF3/4+ ECs are repressed during the progression of PAD. The PAD populations showed an increase in

processes related to platelet activation and wound healing (Figure 2J). Although these results suggest

that ATF3/4+ ECs are actively trying to induce a regenerative response, it is likely not effective as they

at the same time downregulated many processes related to angiogenesis, immune regulatory

functions, interaction with leukocytes as well as metabolism (Figure 2J and S2C). Also, ATF3/4+ ECs

upregulated a set of genes related to the cellular response to metal ions such as cadmium, copper, and

zinc in PAD samples (Figure 2J). This is interesting since several studies have confirmed elevated levels

of these ions in the blood and urine of patients with PAD and atherosclerotic lesions, suggesting they

might be risk factors for the disease[75–79], and have linked those to endothelial dysfunction[80–82].

The activation of epithelial-to-mesenchymal transition in ATF3/4+ capillary ECs (Figure 2J)

prompted us to investigate whether capillary ECs undergo Endothelial-to-mesenchymal (EndoMT) in

PAD. Recent manuscripts reported abnormal microvascular architecture with increased perivascular

collagen deposition[13]. Increased markers of EndoMT were also reported in a fraction of the small

arterioles in CLI patients, likely due to increased TGB signaling via pericytes[12]. Focusing on capillary ECs,

we noticed a characteristic EndoMT pattern, as described in Tombor *et al.*[83], which includes the

downregulation of fatty acid transport and metabolism genes versus the partial activation of a

mesenchymal signature (Figure 2K). Interestingly, some 'canonical' EndoMT genes like *SNAI1, S100A4*

and *SERPINE1* were not affected in PAD while several genes related to extracellular matrix (ECM)

composition, such as *COL1A1*, *COL1A2* and *FN1* were highly upregulated (Figure 2K). This phenotype

was present in both capillary subpopulations but was more pronounced in the ATF3/4[-] ECs (annotated

as Capillary). Also, capillary ECs from PAD patients lowered CDH5 expression but did not completely

lose it (Figure S2D), suggesting that they at least partially undergo EndoMT. Because of the strong

activation of ECM genes, we investigated how ECs in PAD rewire their matrisome, a set of genes that

define ECM composition and function[84]. Capillary ECs in PAD only changed 11.7% (32/274) of the core

matrisome genes, but showed significant activation of specific ECM genes such as collagens, *FN1,*

FBN1, as well as alterations in components of the basement membrane such as laminins, *HSPG2*

(perlecan), and *NID2* (nidogen) (Figure 2L). Taken together, ECs in PAD undergo profound

transcriptomic alterations that impair immune regulatory functions but promote EndoMT.

397   LYVE1[hi]MHCII[low] resident macrophages are the dominant macrophage population in

398   PAD

Blood vessel formation and remodeling is heavily influenced by resident and recruited cells. In

experimental models of PAD, monocyte-derived macrophages contribute to arteriogenesis as well as

angiogenesis[11,23,26,27,29,65,85]. Macrophages also control muscle regeneration: under settings of acute

muscle ischemia or muscle damage, they initially have a pro-inflammatory M1-like phenotype, which

is associated with the expression of several cytokines that further promote inflammation[86]. Rapidly

thereafter, those M1-like macrophages repolarize and change their phenotype towards an anti-

inflammatory restorative/regenerative M2-like phenotype[86–88]. This M1- to M2-like repolarization is

required for optimal muscle regeneration[87,89,90]. M2-like macrophages also have an important pro-

angiogenic role since they are a main source of VEGF. Preventing macrophage infiltration in the injured

muscle is also associated with decreased angiogenesis[91] and inhibiting M2-like macrophage

polarization fails to remodel collateral arteries and to induce angiogenesis following ischemic

damage[92]. However, acute mouse models do not reflect the chronic pathophysiology of PAD, so the

contribution of macrophages to human PAD is poorly understood. We thus decided to further

investigate the Monocytes/Macrophages cluster (MoMac). Similar to our approach with ECs, we

selected MoMac cells from each QC filtered and log normalized sample, performed batch correction,

and merged them. The final dataset was comprised of 10783 high-quality MoMac cells as indicated by

an average of 2743 genes detected per cell and an average of 11.94% of reads mapped to

mitochondrial genes. Clustering and TSNE revealed 5 distinct clusters (Figure 3A). Previous research

has established an extensive set of marker genes for different MoMac populations[93–96] that allowed us

to clearly annotate the different populations in our dataset (Figure 3B). Unbiased marker gene

detection confirmed that we identified well defined macrophage and monocyte populations (Figure

3C and Supplemental File 1). The dataset was comprised of classical monocytes (cMonocytes),

CD16(*FCGR3A*)$^+$ monocytes, dendritic cells (DCs) and two macrophage populations. Alternatively, we

could also classify the two monocytes and DC populations as CCR2$^+$CX3CR1^{low}CD14^{hi}CD16^{low}

(cMonocytes and DCs) and CCR2$^-$CX3CR1^{hi}CD14^{low}CD16^{hi} (CD16$^+$ Monocytes) (Figure S3A). The

CD14^{hi}CD16^{low} and CD14^{low}CD16^{hi} human monocyte populations are the equivalent to Ly6C^{hi} and

Ly6C^{low} monocytes in mice, respectively[97]. Ly6C^{hi} monocytes are recruited monocytes that rapidly

infiltrate the tissue upon injury or infection[98] while Ly6C^{low} cells are patrolling monocytes that crawl on

the inside of blood vessels onto endothelial cells and only extravasate in response to tissue damage or

infection after which they differentiate into macrophages[99].

The macrophage subpopulations were characterized by different expression levels of *LYVE1*

and genes of the Major Histocompatibility Complex Class II (MHCII), and therefore annotated as

LYVE1^{hi}MHCII^{low} macrophages and *LYVE1*^{low}MHCII^{hi} macrophages (Figure S3B). Both populations have

previously   been   identified[100]   in   various   tissues   as   *LYVE1*^{hi}MHCII^{low}*CX3CR1*^{low} and

LYVE1^{low}MHCII^{hi}*CX3CR1*^{hi}, which is consistent with our dataset (Figure S3B)[100]. *LYVE1*^{hi} and *LYVE1*^{low}

macrophages derive from different origins[101], with the former being tissue-resident macrophages

which are seeded in the tissue during embryonic development and locally self-renew throughout life

while the latter are mainly derived from recruited monocytes during adulthood. Additionally,

LYVE1^{hi}MHCII^{low} macrophages have a perivascular localization[100], but how they interact with blood

vessels is poorly studied. Consistent with literature, *LYVE1*^{hi}MHCII^{low} macrophages in our human

muscle samples expressed high levels of tissue-resident macrophage marker genes including *LYVE1*,

FOLR2, *CD209*, *MERTK*, *SELENOP*, *STAB1* and *MAF*[94–96], confirming that those are resident macrophages

(Figure 3B). As expected, *LYVE1*^{hi} macrophages were the largest population in the control samples,

accounting for 70 ± 6% of the MoMac dataset (Figure 3D). And even though the fractional contribution

of *LYVE1*^{hi} resident macrophages was slightly lower in PAD samples, they still remained the largest

population (40 ± 3% of the PAD MoMac dataset). Immunostainings for LYVE1 in combination with the

pan-monocyte/macrophage marker CD68 confirmed the dominant fractional contribution of *LYVE1*^{hi}

resident macrophages in human muscle, both under control as well as PAD settings (Figure 3E-F)

despite a clear increase in the number of CD68$^+$ monocyte/macrophages in PAD (Figure 3G), consistent

with previous work[29]. We did however not observe a reduction in the fraction of *LYVE1*^{hi}*CD68*$^+$ cells,

likely due to thresholding differences or potential differences in cell size. Finally, in our scRNAseq

dataset, the fraction of cMonocytes, but not monocyte-derived *LYVE1*[low] macrophages, went up in PAD

(Figure 3D), suggesting that monocyte-to-macrophage differentiation is impaired in PAD.

Macrophages and monocytes are heterogeneous cells with a wide range of functionalities

depending on the tissue environment and type of stimuli. Sanin et al[48] established a framework for

macrophage activation by analyzing macrophage heterogeneity across different tissues and

inflammatory conditions and reported a set of common initial and early stages of macrophage

activation that further divided into different functional categories defined as phagocytic, oxidative

stress, inflammatory or remodeling (regenerative) paths. We utilized this framework to characterize

the functional differences between the MoMac populations in our dataset. Classical monocytes, CD16[+]

Monocytes and DCs shared a set of common processes related to an early stage of activation including

ECM and actin regulation, cytoskeleton organization and ECM-receptor interaction (Figure 3H).

Additionally, these cells showed processes related to the inflammatory and oxidative stress paths

including innate immune responses and phagosome, respectively (Figure 3H). Interestingly, only

LYVE1[low] macrophages, along with DCs, showed a profile of antigen presentation, in line with their

higher MHCII levels (Figure S3B) and consistent with previous characterization of these cells[100]. Both

LYVE1[hi] and *LYVE1*[low] macrophages were characterized by a specific phagocytic profile as they showed

processes related to this path including complement signaling, phagocytosis, coagulation cascades and

regulation of myeloid leukocyte differentiation. Interestingly, Sanin et al[48] reported that phagocytic

path macrophages are the dominant population in a dataset from mouse aorta atherosclerotic

plaques[102] and interventions promoting plaque regression accelerated the transit of macrophages

along the phagocytic path[48].

The function and contribution of resident macrophages in muscle upon loss of muscle

homeostasis, and in particular in PAD, is poorly understood. Moreover, no genetic data exploring the

transcriptional changes of these cells upon loss of muscle homeostasis are available. We thus decided

to further explore whether resident *LYVE1*[hi] and monocyte-derived *LYVE1*[low] macrophages exhibit

features corresponding to the pro-inflammatory (M1) versus alternative pro-regenerative

macrophages (M2)[103,104]. We used published gene sets associated with M1 or M2 polarization[49] under

in vitro (LPS+IFN-γ (M1) versus IL-4 (M2) stimulation). Since those do not properly reflect the *in vivo*

functional polarization state, we also used gene sets from macrophages that were isolated from

skeletal muscle upon cardiotoxin injury[49] (Ly6C[+] at day 2 after injury are pro-inflammatory (M1-like)

macrophages whereas Ly6C[-] at day 4 after injury are pro-regenerative (M2-like) macrophages). GSVA

using these gene sets showed that *LYVE1*[hi] resident macrophages expressed lower levels of both

M1/inflammatory as well as M2/regenerative related genes (Figure 3I), consistent with their proposed

'homeostatic' function. Additionally, *LYVE1*[low] monocyte-derived macrophages showed higher

expression of both M1 and M2 signatures, as shown in previous studies that report hybrid phenotypes

of *in vivo* macrophages[105], although they exhibited a more pronounced pro-inflammatory M1 signature

(Figure 3I). Thus, *LYVE1*[hi] resident macrophages do not reflect the genetic fingerprint of

M1/inflammatory nor M2/regenerative macrophages.

LYVE1[hi]MHCII[low] macrophages get activated during PAD by increasing their inflammatory and migratory responses

To assess whether the transcriptomic profiles of the MoMac populations are altered upon PAD,

we detected the differentially expressed genes between conditions for every individual population and

observed significant alterations in gene expression across all MoMac populations during PAD

progression (Figure 3J-K and Supplemental File 1). However, since the role of perivascular resident

muscle macrophages is poorly understood, and because it comprises the dominant macrophage

population, even in PAD muscle, we specifically focused on the resident *LYVE1*[hi]MHCII[low] macrophages

where a considerable set of genes was activated in PAD, but we also detected several downregulated

genes (Figure S3C and Supplemental File 1). We performed ORA analysis on the DEG between control

and PAD *LYVE1*[hi]MHCII[low] macrophages (Figure S3D and Supplemental File 1) and found that most of

the enriched biological processes were associated with the genes upregulated during PAD (Figure S3D).

The top enriched processes in PAD *LYVE1*[hi]MHCII[low] macrophages were mainly associated with cell

migration, inflammatory and immune responses, increased cytokine production and macrophage

activation (Figure S3E), suggesting that resident *LYVE1*[hi]MHCII[low] macrophages get activated in PAD.

CCL18, *NAMPT* and *CCL3* were the top upregulated cytokines in *LYVE1*[hi]MHCII[low] macrophages during

PAD. High expression of these cytokines has been detected in atherosclerotic lesions and patients with

cardiovascular diseases[106–109] with *NAMPT* and *CCL3* having an active role in atherosclerotic plaque

formation[110,111], but a role in skeletal muscle of PAD has not been studied. Interestingly, the top

repressed processes in PAD were related to heat responses and protein folding (Figure S3E). ORA

analysis revealed that the heat shock proteins *HSPA1A* and *HSPA1B* were the main drivers for the

repressed processes. *HSPA1A* and *HSPA1B* have been associated with atheroprotective and anti-

inflammatory roles in macrophages[102,112,113], and the loss of this anti-inflammatory role in PAD is

consistent with the activation of inflammatory processes. Altogether, resident *LYVE1*[hi]MHCII[low]

macrophages get activated and acquire pro-inflammatory properties during PAD.

Endothelial-macrophage crosstalk modifies the functional properties of LYVE1[hi]MHCII[low] macrophages during PAD

To study the origin and strength of cellular communication networks within the muscle

microenvironment, we subsequently applied CellChat[52] in the complete dataset. CellChat revealed a

complex communication network between all the cell types (Figure 4A), but while the number of

interactions did not change between control and PAD (Figure 4A), the interaction weight, defined as

the communication probability based on ligand-receptor pairs expression, was reduced in PAD (Figure

4B). Focusing on ECs, they had more interactions in PAD (Figure 4C), specifically with other components

of the vascular niche such as FAPs, monocytes/macrophages and smooth muscle cells/pericytes,

although the strength of these interactions was slightly reduced (Figure 4D). Monocytes/macrophages

showed the highest increase in the interaction strength during PAD, particularly with neutrophils.

When analyzing all outgoing and incoming signaling, FAPs were the main senders in the muscle

whereas ECs, monocytes/macrophages, NK/T/B cells and smooth muscle cells/pericytes were the main

receptors for ligands (Figure 4E). In PAD, the overall outgoing and incoming signals were reduced,

although monocytes/macrophages and ECs remained the top receivers. Next, we analyzed which

signaling networks were altered between control and PAD. As expected, most pathways were reduced

in PAD (Figure 4F). However, pro-inflammatory IL1 signaling was one of the top enriched pathways in

PAD samples. On the other hand, MHC-II signaling was one of the top signaling pathways reduced in

PAD (Figure 4F). Monocytes/Macrophages were the main receiver of these ligands whereas ECs were

the main senders (Figure 4G). Additionally, VEGF signaling was also reduced in PAD (Figure 4H), with

ECs as the specific receiver of this communication. The reduction of MHC-II and VEGF signaling in ECs

is consistent with the loss of immune regulatory and angiogenic properties of ATF3/4$^+$ ECs in PAD

(Figure 2J).

Finally, to gain a better understanding on how EC-macrophage interactions are affected in PAD

at a deeper level, we selected both cell types from our initial clustering (Figure 1G) in every individual

QC filtered and log normalized sample. After pooling and batch correcting all the samples, we applied

the individual clustering of ECs and MoMac datasets (Figure 2A and 3A) to the combined dataset. We

applied NicheNet[53] to predict ligand-receptors pairs that are both differentially expressed and active

between control and PAD samples. Compared to CellChat, NicheNet can additionally link the ligand-

receptor pairs to the (in)activation of their downstream targets, and thus infer signaling activity, by

taking advantage on prior knowledge on signaling and gene regulatory networks. We focused on

identifying which ligands expressed by ATF3/4$^+$ venous and capillary ECs are recognized by receptors

expressed on *LYVE1*^{hi}MHCII^{low} macrophages. Strikingly, NicheNet identified a set of ligand-receptor

pairs specifically upregulated in control samples (Figure 5A) as well as a completely different set

upregulated in PAD samples (Figure 5B), indicating a profound rewiring of the communication network

between ATF3/4$^+$ ECs and *LYVE1*^{hi}MHCII^{low} macrophages in PAD. Some of the top reduced signals

during PAD were related to GAS6 and CSF1 signaling networks (Figure 5A). Interestingly, CSF1 is the

primary growth factor responsible for macrophage differentiation, survival, and proliferation[114], and

endothelial CSF1 contributes to macrophage differentiation[115]. Furthermore, inhibition of GAS6

signaling in macrophages increases their inflammatory responses and cytokine production[116,117].
Inhibition of GAS6 would be consistent with the activated state of *LYVE1*hiMHCIIlow macrophages in
combination with their increased cytokine production (Figure S3E). On the other hand, some signaling
networks were enriched during PAD including IL6 signaling (Figure 5B), which has classically been
associated with pro-inflammatory functions, but several studies also indicated an anti-inflammatory
and atheroprotective role for IL6[118,119]. These results further highlight a complex communication
network between ATF3/4+ ECs and *LYVE1*hiMHCIIlow macrophages, which is profoundly altered during
PAD.

We also used NicheNet to detect which significantly upregulated genes in *LYVE1*hiMHCIIlow
macrophages during PAD could be downstream targets of ligands produced by ATF3/4+ ECs (Figure 5C).
These target genes included *CCL18*, *SGK1*, *PLAUR*, *CCL3*, *LMNA*, *HBEGF*, *IGFS6*, *FTH1* and *CTSB* (Figure
5C). An ORA analysis on these target genes predicted their association with migration, cytokine
response, inflammatory response, and immune related processes (Figure 5D). Since we already
showed that *LYVE1*hiMHCIIlow macrophages significantly upregulate genes that are associated with
migration, cytokine signaling, inflammatory and immune responses (Figure S3E), the results from this
cell communication analysis indicate that the ligands expressed by ATF3/4+ ECs during PAD might play
a role in the activation of *LYVE1*hiMHCIIlow macrophages.

In conclusion, cellular crosstalk in the muscle microenvironment is significantly affected in
PAD. For instance, high and active communication between ATF3/4+ ECs and *LYVE1*hiMHCIIlow
macrophages is considerably altered in a PAD context and some of the ligands produced by ATF3/4+
ECs could be directly involved in macrophage activation during PAD by modulating signaling networks
related to migration, cytokine signaling and inflammation.

Discussion

Using single cell RNA sequencing approaches, we sequenced more than 100'000 cells and generated a comprehensive dataset of the muscle microenvironment of patients with PAD as well as healthy controls (undergoing angioplasty for an aneurysm). We chose to go for a single cell, and not single nuclear approach, because of our specific interest in characterizing the muscle microenvironment in PAD. Using this approach, 90% of the high-quality cells were mononuclear cells, which allowed us to map the composition and transcriptional signature of the muscle microenvironment in PAD with unprecedented detail. Our dataset provides a toolbox for the community that allows to investigate the transcriptomic alterations in the PAD microenvironment in a cell-type dependent fashion. All data are available at GSE235143. Within this manuscript, we chose to focus on endothelial cells (ECs) and macrophages.

We identified common EC subpopulations such as arterial, venous, capillary and even lymphatic populations which we were able to annotate using previously published markers[31,67]. We also found two subpopulations of venous and capillary ECs in human skeletal muscle that are characterized by high levels and activity of the transcription factors ATF3/4. The presence of the ATF3/4[+] subset confirmed earlier work in mouse models which showed that ATF3/4[+] ECs have a higher angiogenic capacity when compared to ATF3/4[-] ECs and are metabolically prepared for angiogenesis. As a consequence, removing ATF4 from ECs prevented exercise-induced angiogenesis[31]. Similarly, deletion of ATF3 reduced endothelial proliferation and impaired aorta regeneration[71], suggesting that ATF3/4[+] ECs have distinct and specific characteristics. Our scRNAseq data further showed that the fraction of ATF3/4[+] ECs was lower in PAD. Also, these ECs changed their ATF3/4 transcriptomic signature and downregulated angiogenic as well as immune regulatory pathways, suggesting that the immune regulatory role of ECs gets (at least partially) lost during PAD.

We measured an almost 3-fold rise in the number of CD68[+] cells on tissue sections. Interestingly, most of those were *LYVE1*[+] resident macrophages and not recruited monocyte-derived macrophages. In fact, while the fraction of monocytes went up in PAD, this did not translate into increased monocyte-derived macrophage number and suggests impaired differentiation of recruited macrophages. Preclinical mouse models for muscle ischemia have unequivocally supported the requirement of recruited monocyte-derived macrophages for muscle regeneration. Injections of clodronate loaded liposomes[120] or genetic depletion of CCR2[121] and other strategies that prevent monocyte-derived macrophage accumulation, impair revascularization, regeneration, and/or reperfusion of the ischemic limb. It is unclear whether the failed accumulation of monocyte-derived macrophages is a result of the chronic nature of PAD or contributing to the impaired regenerative response.

The majority of the muscle macrophage population, even under a setting of chronic

inflammation such as PAD, were LYVE1[hi]MHCII[low] macrophages. LYVE1[hi]MHCII[low] macrophages are

localized close to blood vessels (perivascular macrophages) in most tissues and have a distinct gene

expression profile, phenotype and function compared to monocyte-derived macrophages[100,122,123].

Single cell analysis in murine skeletal muscle previously identified the presence of Lyve1[low]MHCII[hi] and

Lyve1[hi]MHCII[low] subpopulations in the muscle[101]. Interestingly, Lyve1[hi]MHCII[low] macrophages are

embryonically derived, proliferate locally and self-renew with minimal monocyte input, showing that

they are resident muscle macrophages[94]. The role (and even composition) of the resident macrophage

population(s) in muscle is still poorly understood. They cloak microlesions to prevent widespread

neutrophil recruitment[124]. Upon acute damage, they crucially contribute to the clearance of damage-

induced apoptotic cells[96], but there is no evidence that they actively contribute to PAD muscle

pathology. Resident macrophages express FOLR2 and other markers which are typically associated

with M2-like regenerative macrophages such as CD163 and MRC1 (CD206), but we found that during

PAD, they also activate a pro-inflammatory gene program. The co-expression of genes that are typically

associated with an M1 as well as M2-like state, might potentially explain the contrasting findings that

have been reported in literature[29,30]. Recently, TIMD4 was proposed as a marker to isolate resident

macrophages from mouse muscle[96], but we could not detect *TIMD4* expression in human

macrophages. Whether and how resident macrophages contribute to skeletal muscle pathology in PAD

remains to be investigated. Previous work has however linked resident macrophages with vascular

permeability and increased vessel density in non-muscle tissues, though also non-resident

macrophages were depleted in those studies[95,125]. Also, perivascular resident skin macrophages

protrude into the vascular lumen and are skewed towards an anti-inflammatory phenotype while

insensitive to pro-inflammatory stimuli and play distinct roles in skin regeneration when compared to

recruited monocyte-derived-macrophages[126]. In the lung, depletion of LYVE1[hi] macrophages leads to

vessel leakiness and fibrosis, suggesting widespread activation of fibroblasts[100]. Future research needs

to unravel whether the reduced (and altered) crosstalk between ECs and perivascular macrophages

contributes to PAD muscle pathology.

In depth analysis of the transcriptomic changes in microvascular capillary ECs revealed that

PAD is associated with a partial EndoMT signature. EndoMT has already been observed in coronary

artery ECs from patients with atherosclerosis[127]. Also, a fraction of arterioles of patients with CLI

undergo partial EndoMT and retain endothelial markers while activating SMAD2/3 signaling[12]. It has

been reported before that ischemia promotes the activation of EndoMT in clonally expanding ECs in

several pathological settings[128]. Also, in myocardial infarction, partial activation of an EndoMT state is

required for optimal angiogenesis[83]. We did not observe any proliferating ECs in PAD patients (data

not shown). However, it is possible that prolonged states of hypoxia contributed to a state of partial,

and chronic EndoMT. While we did observe activation of some, but not all, markers of EndoMT in PAD

capillary ECs, genes involved in ECM were significantly activated. Also, the endothelial marker *CDH5*

was reduced, but not lost, in PAD. Interestingly, this state of partial EndoMT might be reversible[83].

646   Despite the pronounced increase in vascular density, all patients enrolled in the study reported

intermittent claudication during exercise (Fontaine IIb) or even under resting conditions (Fontaine III-

IV). Potentially, the presence of a thickened ECM and/or basement membrane might limit oxygen and

nutrient diffusion into the ischemic myofibers. In fact, increased thickness of Collagen IV[+] basement

membrane was already shown in PAD[13]. Also, as part of the EndoMT signature, ECs in PAD

downregulated genes involved in lipid transport and trafficking. Beyond reducing endothelial lipid

metabolism, this might affect lipid transport since loss of FABP4 in ECs has been shown to reduce lipid

transport into lipid consuming muscles[129]. It will be exciting to investigate whether reversing EndoMT

in PAD patients could affect capillary ECs properties and restore oxygen and nutrient delivery to the

muscle.

656   Monocyte-derived   macrophages   contribute   to   arteriogenesis   as   well   as

angiogenesis[11,23,26,29,65,85] and control perfusion upon hindlimb ischemia[27]. On the other hand,

endothelial cells also control macrophage differentiation[130], and fate specification via direct contact[131]

or via the secretion of angiocrine factors[27]. We detected strong cellular crosstalk between different

cell types in the muscle microenvironment that was significantly altered in PAD. Interestingly, FAPs

turned out to be main players controlling muscle homeostasis, but how FAPs contribute to the muscle

pathology in PAD needs further investigation. We found that IL1B signaling was one of the top

activated pathways in PAD. IL1B is a known regulator of EndoMT[132]. But also, a study using

canakinumab, an IL1B blocker, improved maximum and pain-free walking distance in patients with

symptomatic PAD while it failed to reach its primary endpoint being an effect on plaque progression

in the superficial femoral artery[133]. Unfortunately, muscle pathology was not further investigated in

this study. Finally, we found that the activation of LYVE1[hi]MHCII[low] resident macrophages could be

driven by ECs. Beyond having a perivascular localization, these data imply that ECs might control the

fate of resident macrophages, an exciting hypothesis that requires further investigation.

670   In conclusion, the contribution of skeletal muscle pathology to PAD progression is poorly

understood. We have used a single cell approach to characterize the composition and transcriptional

changes of the skeletal muscle microenvironment during PAD. We identify profound transcriptomic

alterations in endothelial cells and found that resident macrophages are the main macrophage

population in PAD muscle. Our dataset provides unprecedented insight into the skeletal muscle

microenvironment during PAD. We propose that unravelling the (impaired) cellular crosstalk in PAD

muscle could potentially offer novel targets for therapeutic intervention.

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DISCLOSURES

684   None.

SUPPLEMENTAL MATERIAL

686   Figures S1-S3

687   Supplemental File 1

References

1. Fowkes FGR, Aboyans V, Fowkes FJI, McDermott MM, Sampson UKA, Criqui MH. Peripheral
artery disease: epidemiology and global perspectives. *Nat Rev Cardiol*. 2017;14:156–170.

2. Fowkes FGR, Rudan D, Rudan I, Aboyans V, Denenberg JO, McDermott MM, Norman PE,
Sampson UKA, Williams LJ, Mensah GA, et al. Comparison of global estimates of prevalence
and risk factors for peripheral artery disease in 2000 and 2010: a systematic review and
analysis. *Lancet*. 2013;382:1329–40.

3. De Stefano F, Rios LHP, Fiani B, Fareed J, Tafur A. National Trends for Peripheral Artery
Disease and End Stage Renal Disease From the National Inpatient Sample Database. *Clin Appl*
Thromb Hemost. 2021;27:10760296211025624.

4. ROSE GA. The diagnosis of ischaemic heart pain and intermittent claudication in field surveys.
Bull World Health Organ. 1962;27:645–58.

5. Bethel M, Annex BH. Peripheral arterial disease: A small and large vessel problem. *American*
Heart Journal Plus: Cardiology Research and Practice. 2023;28:100291.

6. Cooke JP, Meng S. Vascular Regeneration in Peripheral Artery Disease. *Arterioscler Thromb*
Vasc Biol. 2020;40:1627–1634.

7. Annex BH. Therapeutic angiogenesis for critical limb ischaemia. *Nat Rev Cardiol*. 2013;10:387–
396.

8. West AM, Anderson JD, Epstein FH, Meyer CH, Hagspiel KD, Berr SS, Harthun NL, Weltman AL,
Annex BH, Kramer CM. Percutaneous intervention in peripheral artery disease improves calf
muscle phosphocreatine recovery kinetics: A pilot study. *Vascular Medicine*. 2012;17:3–9.

9. Kodama A, Takahara M, Iida O, Soga Y, Terashi H, Kawasaki D, Izumi Y, Mii S, Komori K, Azuma
N. Ambulatory Status Over Time after Revascularization in Patients with Chronic Limb-
Threatening Ischemia. *J Atheroscler Thromb*. 2022;29:62892.

10. Gianni-Barrera R, Bartolomeo M, Vollmar B, Djonov V, Banfi A. Split for the cure: VEGF, PDGF-
BB and intussusception in therapeutic angiogenesis. *Biochem Soc Trans*. 2014;42:1637–42.

11. Simons M, Ware JA. Therapeutic angiogenesis in cardiovascular disease. *Nat Rev Drug Discov*.
2003;2:863–71.

12. Chevalier J, Yin H, Arpino J-M, O'Neil C, Nong Z, Gilmore KJ, Lee JJ, Prescott E, Hewak M, Rice
CL, et al. Obstruction of Small Arterioles in Patients with Critical Limb Ischemia due to Partial
Endothelial-to-Mesenchymal Transition. *iScience*. 2020;23:101251.

13. Mietus CJ, Lackner TJ, Karvelis PS, Willcockson GT, Shields CM, Lambert NG, Koutakis P,
Fuglestad MA, Hernandez H, Haynatzki GR, et al. Abnormal Microvascular Architecture,
Fibrosis, and Pericyte Characteristics in the Calf Muscle of Peripheral Artery Disease Patients
with Claudication and Critical Limb Ischemia. *J Clin Med*. 2020;9:2575.

14. Coats P, Hillier C. Differential Responses in Human Subcutaneous and Skeletal Muscle Vascular
Beds to Critical Limb Ischaemia. *European Journal of Vascular and Endovascular Surgery*.
2000;19:387–395.

15.   Hammarsten J, Bylund-Fellenius AC, Holm J, Scherstén T, Krotkiewski M. Capillary supply and
727   muscle fibre types in patients with intermittent claudication: relationships between
728   morphology and metabolism. *Eur J Clin Invest*. 1980;10:301–5.

729   16.   Ho TK, Rajkumar V, Ponticos M, Leoni P, Black DCM, Abraham DJ, Baker DM. Increased
730   endogenous angiogenic response and hypoxia-inducible factor-1α in human critical limb
731   ischemia. *J Vasc Surg*. 2006;43:125–133.

732   17.   McGuigan MR, Bronks R, Newton RU, Sharman MJ, Graham JC, Cody D V, Kraemer WJ. Muscle
733   fiber characteristics in patients with peripheral arterial disease. *Med Sci Sports Exerc*.
734   2001;33:2016–21.

735   18.   Tsui JCS, Baker DM, Biecker E, Shaw S, Dashwood MR. Potential role of endothelin 1 in
736   ischaemia-induced angiogenesis in critical leg ischaemia. *Br J Surg*. 2002;89:741–7.

737   19.   Clyne CA, Mears H, Weller RO, O'Donnell TF. Calf muscle adaptation to peripheral vascular
738   disease. *Cardiovasc Res*. 1985;19:507–12.

739   20.   Robbins JL, Jones WS, Duscha BD, Allen JD, Kraus WE, Regensteiner JG, Hiatt WR, Annex BH.
740   Relationship between leg muscle capillary density and peak hyperemic blood flow with
741   endurance capacity in peripheral artery disease. *J Appl Physiol (1985)*. 2011;111:81–6.

742   21.   Christov C, Chrétien F, Abou-Khalil R, Bassez G, Vallet G, Authier F-J, Bassaglia Y, Shinin V,
743   Tajbakhsh S, Chazaud B, et al. Muscle satellite cells and endothelial cells: close neighbors and
744   privileged partners. *Mol Biol Cell*. 2007;18:1397–409.

745   22.   Méndez-Barbero N, Gutiérrez-Muñoz C, Blanco-Colio L. Cellular Crosstalk between Endothelial
746   and Smooth Muscle Cells in Vascular Wall Remodeling. *Int J Mol Sci*. 2021;22:7284.

747   23.   Krishnasamy K, Limbourg A, Kapanadze T, Gamrekelashvili J, Beger C, Häger C, Lozanovski VJ,
748   Falk CS, Napp LC, Bauersachs J, et al. Blood vessel control of macrophage maturation
749   promotes arteriogenesis in ischemia. *Nat Commun*. 2017;8:952.

750   24.   Amersfoort J, Eelen G, Carmeliet P. Immunomodulation by endothelial cells — partnering up
751   with the immune system? *Nat Rev Immunol*. 2022;22:576–588.

752   25.   Dort J, Fabre P, Molina T, Dumont NA. Macrophages Are Key Regulators of Stem Cells during
753   Skeletal Muscle Regeneration and Diseases. *Stem Cells Int*. 2019;2019:4761427.

754   26.   Ganta VC, Choi MH, Kutateladze A, Fox TE, Farber CR, Annex BH. A MicroRNA93–Interferon
755   Regulatory Factor-9–Immunoresponsive Gene-1–Itaconic Acid Pathway Modulates M2-Like
756   Macrophage Polarization to Revascularize Ischemic Muscle. *Circulation*. 2017;135:2403–2425.

757   27.   Zhang J, Muri J, Fitzgerald G, Gorski T, Gianni-Barrera R, Masschelein E, D'Hulst G, Gilardoni P,
758   Turiel G, Fan Z, et al. Endothelial Lactate Controls Muscle Regeneration from Ischemia by
759   Inducing M2-like Macrophage Polarization. *Cell Metab*. 2020;31:1136-1153.e7.

760   28.   Casale GP, Thompson JR, Carpenter LC, Kim J, Lackner TJ, Mietus CJ, Ha DM, Myers SA,
761   Brunette KE, Li S, et al. Cytokine signature of inflammation mediated by autoreactive Th-cells,
762   in calf muscle of claudicating patients with Fontaine stage II peripheral artery disease.
763   *Translational Research*. 2021;228:94–108.

764   29.   Ganta VC, Choi M, Farber CR, Annex BH. Antiangiogenic VEGF $_{165}$ b Regulates Macrophage
765   Polarization via S100A8/S100A9 in Peripheral Artery Disease. *Circulation*. 2019;139:226–242.

766   30.   Kosmac K, Gonzalez-Freire M, McDermott MM, White SH, Walton RG, Sufit RL, Tian L, Li L,
767   Kibbe MR, Criqui MH, et al. Correlations of Calf Muscle Macrophage Content With Muscle
768   Properties and Walking Performance in Peripheral Artery Disease. *J Am Heart Assoc*. 2020;9.

769   31.   Fan Z, Turiel G, Ardicoglu R, Ghobrial M, Masschelein E, Kocijan T, Zhang J, Tan G, Fitzgerald G,
770   Gorski T, et al. Exercise-induced angiogenesis is dependent on metabolically primed ATF3/4+
771   endothelial cells. *Cell Metab*. 2021;33:1793-1807.e9.

772   32.   Brevetti G, Giugliano G, Brevetti L, Hiatt WR. Inflammation in Peripheral Artery Disease.
773   *Circulation*. 2010;122:1862–1875.

774   33.   Silvestro A, Scopacasa F, Ruocco A, Oliva G, Schiano V, Zincarelli C, Brevetti G. Inflammatory
775   status and endothelial function in asymptomatic and symptomatic peripheral arterial disease.
776   *Vasc Med*. 2003;8:225–32.

777   34.   McDermott MM, Guralnik JM, Corsi A, Albay M, Macchi C, Bandinelli S, Ferrucci L. Patterns of
778   inflammation associated with peripheral arterial disease: the InCHIANTI study. *Am Heart J*.
779   2005;150:276–81.

780   35.   Madissoon E, Wilbrey-Clark A, Miragaia RJ, Saeb-Parsy K, Mahbubani KT, Georgakopoulos N,
781   Harding P, Polanski K, Huang N, Nowicki-Osuch K, et al. scRNA-seq assessment of the human
782   lung, spleen, and esophagus tissue stability after cold preservation. *Genome Biol*. 2020;21:1.

783   36.   Lun ATL, McCarthy DJ, Marioni JC. A step-by-step workflow for low-level analysis of single-cell
784   RNA-seq data with Bioconductor. *F1000Res* [Internet]. 2016;5:2122. Available from:
785   https://f1000research.com/articles/5-2122/v2

786   37.   McCarthy DJ, Campbell KR, Lun ATL, Wills QF. Scater: Pre-processing, quality control,
787   normalization and visualization of single-cell RNA-seq data in R. *Bioinformatics*.
788   2017;33:1179–1186.

789   38.   Hong R, Koga Y, Bandyadka S, Leshchyk A, Wang Y, Akavoor V, Cao X, Sarfraz I, Wang Z,
790   Alabdullatif S, et al. Comprehensive generation, visualization, and reporting of quality control
791   metrics for single-cell RNA sequencing data. *Nat Commun*. 2022;13:1688.

792   39.   Korsunsky I, Millard N, Fan J, Slowikowski K, Zhang F, Wei K, Baglaenko Y, Brenner M, Loh P,
793   Raychaudhuri S. Fast, sensitive and accurate integration of single-cell data with Harmony. *Nat
794   Methods*. 2019;16:1289–1296.

795   40.   Hao Y, Hao S, Andersen-Nissen E, Mauck WM, Zheng S, Butler A, Lee MJ, Wilk AJ, Darby C,
796   Zager M, et al. Integrated analysis of multimodal single-cell data. *Cell*. 2021;184:3573-
797   3587.e29.

798   41.   Raudvere U, Kolberg L, Kuzmin I, Arak T, Adler P, Peterson H, Vilo J. g:Profiler: a web server for
799   functional enrichment analysis and conversions of gene lists (2019 update). *Nucleic Acids Res*.
800   2019;47:W191–W198.

801   42.   Sayols S. rrvgo: a Bioconductor package for interpreting lists of Gene Ontology terms.
802   *MicroPubl Biol*. 2023;2023.

803   43.   Luo W, Brouwer C. Pathview: an R/Bioconductor package for pathway-based data integration
804   and visualization. *Bioinformatics*. 2013;29:1830–1.

805   44.   Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq
806   data with DESeq2. *Genome Biol*. 2014;15:550.

807   45.   Korotkevich G, Sukhov V, Budin N, Shpak B, Artyomov MN, Sergushichev A. Fast gene set
808   enrichment analysis. *bioRxiv* [Internet]. 2021;060012. Available from:
809   http://biorxiv.org/content/early/2021/02/01/060012.abstract

810   46.   Liberzon A, Birger C, Thorvaldsdóttir H, Ghandi M, Mesirov JP, Tamayo P. The Molecular
811   Signatures Database Hallmark Gene Set Collection. *Cell Syst*. 2015;1:417–425.

812   47.   Hänzelmann S, Castelo R, Guinney J. GSVA: gene set variation analysis for microarray and
813   RNA-seq data. *BMC Bioinformatics*. 2013;14:7.

814   48.   Sanin DE, Ge Y, Marinkovic E, Kabat AM, Castoldi A, Caputa G, Grzes KM, Curtis JD, Thompson
815   EA, Willenborg S, et al. A common framework of monocyte-derived macrophage activation.
816   *Sci Immunol*. 2022;7.

817   49.   Varga T, Mounier R, Horvath A, Cuvellier S, Dumont F, Poliska S, Ardjoune H, Juban G, Nagy L,
818   Chazaud B. Highly Dynamic Transcriptional Signature of Distinct Macrophage Subsets during
819   Sterile Inflammation, Resolution, and Tissue Repair. *The Journal of Immunology*.
820   2016;196:4771–4782.

821   50.   Aibar S, González-Blas CB, Moerman T, Huynh-Thu VA, Imrichova H, Hulselmans G, Rambow F,
822   Marine J-C, Geurts P, Aerts J, et al. SCENIC: single-cell regulatory network inference and
823   clustering. *Nat Methods*. 2017;14:1083–1086.

824   51.   González-Blas CB, De Winter S, Hulselmans G, Hecker N, Matetovici I, Christiaens V,
825   Poovathingal S, Wouters J, Aibar S, Aerts S. SCENIC+: single-cell multiomic inference of
826   enhancers and gene regulatory networks. *bioRxiv* [Internet]. 2022;2022.08.19.504505.
827   Available from: http://biorxiv.org/content/early/2022/08/19/2022.08.19.504505.abstract

828   52.   Jin S, Guerrero-Juarez CF, Zhang L, Chang I, Ramos R, Kuan C-H, Myung P, Plikus M V., Nie Q.
829   Inference and analysis of cell-cell communication using CellChat. *Nat Commun*. 2021;12:1088.

830   53.   Browaeys R, Saelens W, Saeys Y. NicheNet: modeling intercellular communication by linking
831   ligands to target genes. *Nat Methods*. 2020;17:159–162.

832   54.   Jones RC, Karkanias J, Krasnow MA, Pisco AO, Quake SR, Salzman J, Yosef N, Bulthaup B,
833   Brown P, Harper W, et al. The Tabula Sapiens: A multiple-organ, single-cell transcriptomic
834   atlas of humans. *Science (1979)*. 2022;376.

835   55.   de Micheli AJ, Spector JA, Elemento O, Cosgrove BD. A reference single-cell transcriptomic
836   atlas of human skeletal muscle tissue reveals bifurcated muscle stem cell populations. *Skelet
837   Muscle*. 2020;10:19.

838   56.   Rubenstein AB, Smith GR, Raue U, Begue G, Minchev K, Ruf-Zamojski F, Nair VD, Wang X, Zhou
839   L, Zaslavsky E, et al. Single-cell transcriptional profiles in human skeletal muscle. *Sci Rep*.
840   2020;10:1–15.

841   57.   Cameron A, Wakelin G, Gaulton N, Young L V., Wotherspoon S, Hodson N, Lees MJ, Moore DR,
842   Johnston AP. Identification of underexplored mesenchymal and vascular-related cell
843   populations in human skeletal muscle. *American Journal of Physiology-Cell Physiology*.
844   2022;323:C1586–C1600.

845   58.   De Micheli AJ, Laurilliard EJ, Heinke CL, Ravichandran H, Fraczek P, Soueid-Baumgarten S, De
846   Vlaminck I, Elemento O, Cosgrove BD. Single-Cell Analysis of the Muscle Stem Cell Hierarchy

Identifies Heterotypic Communication Signals Involved in Skeletal Muscle Regeneration. *Cell*
Rep. 2020;30:3583-3595.e5.

59.   Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq
data with DESeq2. *Genome Biol*. 2014;15:550.

60.   Pande RL, Brown J, Buck S, Redline W, Doyle J, Plutzky J, Creager MA. Association of monocyte
tumor necrosis factor α expression and serum inflammatory biomarkers with walking
impairment in peripheral artery disease. *J Vasc Surg*. 2015;61:155–61.

61.   Signorelli SS, Mazzarino MC, Di Pino L, Malaponte G, Porto C, Pennisi G, Marchese G, Costa
MP, Digrandi D, Celotta G, et al. High circulating levels of cytokines (IL-6 and TNFalpha),
adhesion molecules (VCAM-1 and ICAM-1) and selectins in patients with peripheral arterial
disease at rest and after a treadmill test. *Vasc Med*. 2003;8:15–9.

62.   Ferrucci L, Candia J, Ubaida-Mohien C, Lyaskov A, Banskota N, Leeuwenburgh C, Wohlgemuth
S, Guralnik JM, Kaileh M, Zhang D, et al. Transcriptomic and Proteomic of Gastrocnemius
Muscle in Peripheral Artery Disease. *Circ Res*. 2023;

63.   Ryan TE, Yamaguchi DJ, Schmidt CA, Zeczycki TN, Shaikh SR, Brophy P, Green TD, Tarpey MD,
Karnekar R, Goldberg EJ, et al. Extensive skeletal muscle cell mitochondriopathy distinguishes
critical limb ischemia patients from claudicants. *JCI Insight*. 2018;3.

64.   Santos M Dos, Gioftsidi S, Backer S, Machado L, Relaix F, Maire P, Mourikis P. Extraction and
sequencing of single nuclei from murine skeletal muscles. *STAR Protoc*. 2021;2:100694.

65.   Tan RP, Ryder I, Yang N, Lam YT, Santos M, Michael PL, Robinson DA, Ng MK, Wise SG.
Macrophage Polarization as a Novel Therapeutic Target for Endovascular Intervention in
Peripheral Artery Disease. *JACC Basic Transl Sci*. 2021;6:693–704.

66.   Kavurma MM, Bursill C, Stanley CP, Passam F, Cartland SP, Patel S, Loa J, Figtree GA, Golledge
J, Aitken S, et al. Endothelial cell dysfunction: Implications for the pathogenesis of peripheral
artery disease. *Front Cardiovasc Med*. 2022;9:1054576.

67.   Kalucka J, de Rooij LPMH, Goveia J, Rohlenova K, Dumas SJ, Meta E, Conchinha N V, Taverna F,
Teuwen L-A, Veys K, et al. Single-Cell Transcriptome Atlas of Murine Endothelial Cells. *Cell*.
2020;180:764-779.e20.

68.   Kanehisa M, Sato Y, Kawashima M, Furumichi M, Tanabe M. KEGG as a reference resource for
gene and protein annotation. *Nucleic Acids Res*. 2016;44:D457–D462.

69.   Wedgwood JF, Hatam L, Bonagura VR. Effect of interferon-gamma and tumor necrosis factor
on the expression of class I and class II major histocompatibility molecules by cultured human
umbilical vein endothelial cells. *Cell Immunol*. 1988;111:1–9.

70.   Mai J, Nanayakkara G, Lopez-Pastrana J, Li X, Li Y-F, Wang X, Song A, Virtue A, Shao Y, Shan H,
et al. Interleukin-17A Promotes Aortic Endothelial Cell Activation via Transcriptionally and
Post-translationally Activating p38 Mitogen-activated Protein Kinase (MAPK) Pathway. *J Biol*
Chem. 2016;291:4939–54.

71.   McDonald AI, Shirali AS, Aragón R, Ma F, Hernandez G, Vaughn DA, Mack JJ, Lim TY, Sunshine
H, Zhao P, et al. Endothelial Regeneration of Large Vessels Is a Biphasic Process Driven by
Local Cells with Distinct Proliferative Capacities. *Cell Stem Cell*. 2018;23:210-225.e6.

72.   Hänzelmann S, Castelo R, Guinney J. GSVA: gene set variation analysis for microarray and
888   RNA-Seq data. *BMC Bioinformatics*. 2013;14:7.

889   73.   Ashburner M, Ball CA, Blake JA, Botstein D, Butler H, Cherry JM, Davis AP, Dolinski K, Dwight
890   SS, Eppig JT, et al. Gene ontology: tool for the unification of biology. The Gene Ontology
891   Consortium. *Nat Genet*. 2000;25:25–9.

892   74.   Gene Ontology Consortium. The Gene Ontology resource: enriching a GOld mine. *Nucleic
893   Acids Res*. 2021;49:D325–D334.

894   75.   Tellez-Plaza M, Guallar E, Fabsitz RR, Howard B V, Umans JG, Francesconi KA, Goessler W,
895   Devereux RB, Navas-Acien A. Cadmium exposure and incident peripheral arterial disease. *Circ
896   Cardiovasc Qual Outcomes*. 2013;6:626–33.

897   76.   Tinkov AA, Filippini T, Ajsuvakova OP, Skalnaya MG, Aaseth J, Bjørklund G, Gatiatulina ER,
898   Popova E V, Nemereshina ON, Huang P-T, et al. Cadmium and atherosclerosis: A review of
899   toxicological mechanisms and a meta-analysis of epidemiologic studies. *Environ Res*.
900   2018;162:240–260.

901   77.   Stadler N, Stanley N, Heeneman S, Vacata V, Daemen MJAP, Bannon PG, Waltenberger J,
902   Davies MJ. Accumulation of Zinc in Human Atherosclerotic Lesions Correlates With Calcium
903   Levels But Does Not Protect Against Protein Oxidation. *Arterioscler Thromb Vasc Biol*.
904   2008;28:1024–1030.

905   78.   Bagheri B, Akbari N, Tabiban S, Habibi V, Mokhberi V. Serum level of copper in patients with
906   coronary artery disease. *Niger Med J*. 2015;56:39–42.

907   79.   Navas-Acien A, Silbergeld EK, Sharrett R, Calderon-Aranda E, Selvin E, Guallar E. Metals in
908   urine and peripheral arterial disease. *Environ Health Perspect*. 2005;113:164–9.

909   80.   Knoflach M, Messner B, Shen YH, Frotschnig S, Liu G, Pfaller K, Wang X, Matosevic B, Willeit J,
910   Kiechl S, et al. Non-toxic cadmium concentrations induce vascular inflammation and promote
911   atherosclerosis. *Circ J*. 2011;75:2491–5.

912   81.   Liang H, Yue R, Zhou C, Liu M, Yu X, Lu S, Zeng J, Yu Z, Zhou Z, Hu H. Cadmium exposure
913   induces endothelial dysfunction via disturbing lipid metabolism in human microvascular
914   endothelial cells. *J Appl Toxicol*. 2021;41:775–788.

915   82.   Yeh S-C, Tsai F-Y, Chao H-R, Tsou T-C. Zinc ions induce inflammatory responses in vascular
916   endothelial cells. *Bull Environ Contam Toxicol*. 2011;87:113–6.

917   83.   Tombor LS, John D, Glaser SF, Luxán G, Forte E, Furtado M, Rosenthal N, Baumgarten N,
918   Schulz MH, Wittig J, et al. Single cell sequencing reveals endothelial plasticity with transient
919   mesenchymal activation after myocardial infarction. *Nat Commun*. 2021;12:681.

920   84.   Hynes RO, Naba A. Overview of the matrisome--an inventory of extracellular matrix
921   constituents and functions. *Cold Spring Harb Perspect Biol*. 2012;4:a004903.

922   85.   Hamm A, Veschini L, Takeda Y, Costa S, Delamarre E, Squadrito ML, Henze A, Wenes M,
923   Serneels J, Pucci F, et al. PHD2 regulates arteriogenic macrophages through TIE2 signalling.
924   *EMBO Mol Med*. 2013;5:843–857.

925   86.   Mosser DM, Edwards JP. Exploring the full spectrum of macrophage activation. *Nat Rev
926   Immunol*. 2008;8:958–69.

87.   Arnold L, Henry A, Poron F, Baba-Amer Y, van Rooijen N, Plonquet A, Gherardi RK, Chazaud B.
Inflammatory monocytes recruited after skeletal muscle injury switch into antiinflammatory
macrophages to support myogenesis. *J Exp Med*. 2007;204:1057–69.

88.   Raes G, De Baetselier P, Noël W, Beschin A, Brombacher F, Hassanzadeh Gh G. Differential
expression of FIZZ1 and Ym1 in alternatively versus classically activated macrophages. *J
Leukoc Biol*. 2002;71:597–602.

89.   Deng B, Wehling-Henricks M, Villalta SA, Wang Y, Tidball JG. IL-10 triggers changes in
macrophage phenotype that promote muscle growth and regeneration. *J Immunol*.
2012;189:3669–80.

90.   Tidball JG, Wehling-Henricks M. Macrophages promote muscle membrane repair and muscle
fibre growth and regeneration during modified muscle loading in mice in vivo. *J Physiol*.
2007;578:327–36.

91.   Ochoa O, Sun D, Reyes-Reyna SM, Waite LL, Michalek JE, McManus LM, Shireman PK. Delayed
angiogenesis and VEGF production in CCR2-/- mice during impaired skeletal muscle
regeneration. *Am J Physiol Regul Integr Comp Physiol*. 2007;293:R651-61.

92.   Zhu H, Zhang M, Liu Z, Xing J, Moriasi C, Dai X, Zou M-H. AMP-Activated Protein Kinase α1 in
Macrophages Promotes Collateral Remodeling and Arteriogenesis in Mice In Vivo. *Arterioscler
Thromb Vasc Biol*. 2016;36:1868–78.

93.   Mulder K, Patel AA, Kong WT, Piot C, Halitzki E, Dunsmore G, Khalilnezhad S, Irac SE,
Dubuisson A, Chevrier M, et al. Cross-tissue single-cell landscape of human monocytes and
macrophages in health and disease. *Immunity*. 2021;54:1883-1900.e5.

94.   Dick SA, Wong A, Hamidzada H, Nejat S, Nechanitzky R, Vohra S, Mueller B, Zaman R, Kantores
C, Aronoff L, et al. Three tissue resident macrophage subsets coexist across organs with
conserved origins and life cycles. *Sci Immunol*. 2022;7.

95.   Moura Silva H, Kitoko JZ, Queiroz CP, Kroehling L, Matheis F, Yang KL, Reis BS, Ren-Fielding C,
Littman DR, Bozza MT, et al. c-MAF–dependent perivascular macrophages regulate diet-
induced metabolic syndrome. *Sci Immunol*. 2021;6.

96.   Babaeijandaghi F, Cheng R, Kajabadi N, Soliman H, Chang C-K, Smandych J, Tung LW, Long R,
Ghassemi A, Rossi FM V. Metabolic reprogramming of skeletal muscle by resident
macrophages points to CSF1R inhibitors as muscular dystrophy therapeutics. *Sci Transl Med*.
2022;14.

97.   Ziegler-Heitbrock L, Ancuta P, Crowe S, Dalod M, Grau V, Hart DN, Leenen PJM, Liu Y-J,
MacPherson G, Randolph GJ, et al. Nomenclature of monocytes and dendritic cells in blood.
Blood. 2010;116:e74-80.

98.   Geissmann F, Jung S, Littman DR. Blood Monocytes Consist of Two Principal Subsets with
Distinct Migratory Properties. *Immunity*. 2003;19:71–82.

99.   Auffray C, Fogg D, Garfa M, Elain G, Join-Lambert O, Kayal S, Sarnacki S, Cumano A, Lauvau G,
Geissmann F. Monitoring of Blood Vessels and Tissues by a Population of Monocytes with
Patrolling Behavior. *Science (1979)*. 2007;317:666–670.

100.   Chakarov S, Lim HY, Tan L, Lim SY, See P, Lum J, Zhang X-M, Foo S, Nakamizo S, Duan K, et al.
Two distinct interstitial macrophage populations coexist across tissues in specific subtissular
niches. *Science (1979)*. 2019;363.

101.   Wang X, Sathe AA, Smith GR, Ruf-Zamojski F, Nair V, Lavine KJ, Xing C, Sealfon SC, Zhou L.
Heterogeneous origins and functions of mouse skeletal muscle-resident macrophages.
Proceedings of the National Academy of Sciences. 2020;117:20729–20740.

102.   Lin J-D, Nishi H, Poles J, Niu X, Mccauley C, Rahman K, Brown EJ, Yeung ST, Vozhilla N,
Weinstock A, et al. Single-cell analysis of fate-mapped macrophages reveals heterogeneity,
including stem-like properties, during atherosclerosis progression and regression. *JCI Insight*.
2019;4.

103.   Martinez FO, Gordon S. The M1 and M2 paradigm of macrophage activation: time for
reassessment. *F1000Prime Rep*. 2014;6.

104.   Mantovani , Alberto, Sica , Antonio, Locati M. Macrophage Polarization Comes of Age.
Immunity. 2005;23:344–346.

105.   Palma A, Jarrah AS, Tieri P, Cesareni G, Castiglione F. Gene Regulatory Network Modeling of
Macrophage Differentiation Corroborates the Continuum Hypothesis of Polarization States.
Front Physiol. 2018;9:1659.

106.   Orekhov AN, Nikiforov NG, Elizova N V, Korobov GA, Aladinskaya A V, Sobenin IA, Bobryshev Y
V. Tumor Necrosis Factor-α and C-C Motif Chemokine Ligand 18 Associate with
Atherosclerotic Lipid Accumulation In situ and In vitro. *Curr Pharm Des*. 2018;24:2883–2889.

107.   Hägg DA, Olson FJ, Kjelldahl J, Jernås M, Thelle DS, Carlsson LMS, Fagerberg B, Svensson P-A.
Expression of chemokine (C-C motif) ligand 18 in human macrophages and atherosclerotic
plaques. *Atherosclerosis*. 2009;204:e15-20.

108.   Halvorsen B, Espeland MZ, Andersen GØ, Yndestad A, Sagen EL, Rashidi A, Knudsen EC,
Skjelland M, Skagen KR, Krohg-Sørensen K, et al. Increased expression of NAMPT in PBMC
from patients with acute coronary syndrome and in inflammatory M1 macrophages.
Atherosclerosis. 2015;243:204–10.

109.   de Jager SCA, Bongaerts BWC, Weber M, Kraaijeveld AO, Rousch M, Dimmeler S, van Dieijen-
Visser MP, Cleutjens KBJM, Nelemans PJ, van Berkel TJC, et al. Chemokines CCL3/MIP1α,
CCL5/RANTES and CCL18/PARC are independent risk predictors of short-term mortality in
patients with acute coronary syndromes. *PLoS One*. 2012;7:e45804.

110.   Li S, Wang C, Li K, Li L, Tian M, Xie J, Yang M, Jia Y, He J, Gao L, et al. NAMPT knockdown
attenuates atherosclerosis and promotes reverse cholesterol transport in ApoE KO mice with
high-fat-induced insulin resistance. *Sci Rep*. 2016;6:26746.

111.   de Jager SCA, Bot I, Kraaijeveld AO, Korporaal SJA, Bot M, van Santbrink PJ, van Berkel TJC,
Kuiper J, Biessen EAL. Leukocyte-Specific CCL3 Deficiency Inhibits Atherosclerotic Lesion
Development by Affecting Neutrophil Accumulation. *Arterioscler Thromb Vasc Biol*. 2013;33.

112.   Brykczynska U, Geigges M, Wiedemann SJ, Dror E, Böni-Schnetzler M, Hess C, Donath MY,
Paro R. Distinct Transcriptional Responses across Tissue-Resident Macrophages to Short-Term
and Long-Term Metabolic Challenge. *Cell Rep*. 2020;30:1627-1643.e7.

113.   Dulin E, García-Barreno P, Guisasola MC. Genetic variations of HSPA1A, the heat shock protein
1007   levels, and risk of atherosclerosis. *Cell Stress Chaperones*. 2012;17:507–16.

1008   114.   Sehgal A, Irvine KM, Hume DA. Functions of macrophage colony-stimulating factor (CSF1) in
1009   development, homeostasis, and tissue repair. *Semin Immunol*. 2021;54:101509.

1010   115.   Sinha SK, Miikeda A, Fouladian Z, Mehrabian M, Edillor C, Shih D, Zhou Z, Paul MK,
1011   Charugundla S, Davis RC, et al. Local M-CSF (Macrophage Colony-Stimulating Factor)
1012   Expression Regulates Macrophage Proliferation and Apoptosis in Atherosclerosis. *Arterioscler*
1013   *Thromb Vasc Biol*. 2021;41:220–233.

1014   116.   Deng T, Zhang Y, Chen Q, Yan K, Han D. Toll-like receptor-mediated inhibition of Gas6 and
1015   ProS expression facilitates inflammatory cytokine production in mouse macrophages.
1016   *Immunology*. 2012;135:40–50.

1017   117.   Chiu K-C, Lee C-H, Liu S-Y, Chou Y-T, Huang R-Y, Huang S-M, Shieh Y-S. Polarization of tumor-
1018   associated macrophages and Gas6/Axl signaling in oral squamous cell carcinoma. *Oral Oncol*.
1019   2015;51:683–9.

1020   118.   Fuster JJ, Walsh K. The Good, the Bad, and the Ugly of interleukin-6 signaling. *EMBO J*.
1021   2014;33:1425–1427.

1022   119.   Fernando MR, Reyes JL, Iannuzzi J, Leung G, McKay DM. The pro-inflammatory cytokine,
1023   interleukin-6, enhances the polarization of alternatively activated macrophages. *PLoS One*.
1024   2014;9:e94188.

1025   120.   Vågesjö E, Parv K, Ahl D, Seignez C, Herrera Hidalgo C, Giraud A, Leite C, Korsgren O, Wallén H,
1026   Juusola G, et al. Perivascular Macrophages Regulate Blood Flow Following Tissue Damage. *Circ*
1027   *Res*. 2021;128:1694–1707.

1028   121.   Lu H, Huang D, Saederup N, Charo IF, Ransohoff RM, Zhou L. Macrophages recruited via CCR2
1029   produce insulin-like growth factor-1 to repair acute skeletal muscle injury. *FASEB J*.
1030   2011;25:358–69.

1031   122.   Schyns J, Bai Q, Ruscitti C, Radermecker C, De Schepper S, Chakarov S, Farnir F, Pirottin D,
1032   Ginhoux F, Boeckxstaens G, et al. Non-classical tissue monocytes and two functionally distinct
1033   populations of interstitial macrophages populate the mouse lung. *Nat Commun*.
1034   2019;10:3964.

1035   123.   Ural BB, Yeung ST, Damani-Yokota P, Devlin JC, de Vries M, Vera-Licona P, Samji T, Sawai CM,
1036   Jang G, Perez OA, et al. Identification of a nerve-associated, lung-resident interstitial
1037   macrophage subset with distinct localization and immunoregulatory properties. *Sci Immunol*.
1038   2020;5.

1039   124.   Uderhardt S, Martins AJ, Tsang JS, Lämmermann T, Germain RN. Resident Macrophages Cloak
1040   Tissue Microlesions to Prevent Neutrophil-Driven Inflammatory Damage. *Cell*. 2019;177:541-
1041   555.e17.

1042   125.   He H, Mack JJ, Güç E, Warren CM, Squadrito ML, Kilarski WW, Baer C, Freshman RD,
1043   McDonald AI, Ziyad S, et al. Perivascular Macrophages Limit Permeability. *Arterioscler Thromb*
1044   *Vasc Biol*. 2016;36:2203–2212.

1045   126.   Barreiro O, Cibrian D, Clemente C, Alvarez D, Moreno V, Valiente Í, Bernad A, Vestweber D,
1046   Arroyo AG, Martín P, et al. Pivotal role for skin transendothelial radio-resistant anti-
1047   inflammatory macrophages in tissue repair. *Elife*. 2016;5.

1048   127.   Chen P-Y, Qin L, Baeyens N, Li G, Afolabi T, Budatha M, Tellides G, Schwartz MA, Simons M.
1049   Endothelial-to-mesenchymal transition drives atherosclerosis progression. *J Clin Invest*.
1050   2015;125:4514–28.

1051   128.   Manavski Y, Lucas T, Glaser SF, Dorsheimer L, Günther S, Braun T, Rieger MA, Zeiher AM, Boon
1052   RA, Dimmeler S. Clonal Expansion of Endothelial Cells Contributes to Ischemia-Induced
1053   Neovascularization. *Circ Res*. 2018;122:670–677.

1054   129.   Iso T, Maeda K, Hanaoka H, Suga T, Goto K, Syamsunarno MRAA, Hishiki T, Nagahata Y, Matsui
1055   H, Arai M, et al. Capillary endothelial fatty acid binding proteins 4 and 5 play a critical role in
1056   fatty acid uptake in heart and skeletal muscle. *Arterioscler Thromb Vasc Biol*. 2013;33:2549–
1057   57.

1058   130.   Baer C, Squadrito ML, Iruela-Arispe ML, De Palma M. Reciprocal interactions between
1059   endothelial cells and macrophages in angiogenic vascular niches. *Exp Cell Res*. 2013;319:1626–
1060   34.

1061   131.   He H, Xu J, Warren CM, Duan D, Li X, Wu L, Iruela-Arispe ML. Endothelial cells provide an
1062   instructive niche for the differentiation and functional polarization of M2-like macrophages.
1063   *Blood*. 2012;120:3152–62.

1064   132.   Maleszewska M, Moonen J-RAJ, Huijkman N, van de Sluis B, Krenning G, Harmsen MC. IL-1β
1065   and TGFβ2 synergistically induce endothelial to mesenchymal transition in an NFκB-
1066   dependent manner. *Immunobiology*. 2013;218:443–54.

1067   133.   Russell KS, Yates DP, Kramer CM, Feller A, Mahling P, Colin L, Clough T, Wang T, LaPerna L,
1068   Patel A, et al. A randomized, placebo-controlled trial of canakinumab in patients with
1069   peripheral artery disease. *Vasc Med*. 2019;24:414–421.

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Figure 1. scRNAseq reveals cell heterogeneity in PAD. A, Graphical abstract of the experimental

approach used in this study. **B**, Representative images of haematoxylin and eosin (H&E)-stained

sections of gastrocnemius muscle from control (n=4) and PAD (n=6) patients (Scale bar: 50 µm). **C,D**,

Quantification of fiber size in $µm^2$ (**C**) and minimum feret diameter in µm (**D**) from panel B. **E**,

Representative images of gastrocnemius muscle from control (n=4) and PAD (n=6) stained for the EC

marker (CD31, green) and cell nuclei (Hoechst, blue) (Scale bar: 50 µm). Muscle fibers are represented

with white dashed lines. **F**, Quantification of capillary density from panel E measured as the number of

CD31$^+$ cells per mm^2. **G**, TSNE plot of cell populations identified from human gastrocnemius muscle,

color-coded by the identified populations. **H**, Dot plot of centered logcounts values from cell type-

specific marker genes. Color and size of the dots indicate the centered logcount value and the

proportion of cells that express the gene, respectively. **I**, Stacked bar plots showing cluster percentage

in each condition, color-coded by cluster. **J**, Heatmap of centered normalized values from top high

variable genes (adjusted p-value < 0.05) in each patient based on pseudobulk analysis between control

and PAD conditions, color indicates the centered normalized values. **K**, Bar plots showing Normalized

Enrichment Score of significant (adjusted p-value < 0.05) Hallmark pathways (MsigDB) from GSEA

analysis on the pseudobulk dataset, color indicates the condition. Each dot represents a single patient

in panels C,D and F. Student's t test (two-tailed, unpaired, parametric, ns > 0.05, $^{**}p$ < 0.01) was used

in C,D and F.

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Figure 2. Human skeletal muscle contains several EC subtypes with different functional properties.

A, TSNE plot of EC subtypes identified from human gastrocnemius muscle, color-coded by the

identified subtypes. **B**, Dot plot of centered logcounts values from EC subtype-specific marker genes.

Color and size of the dots indicate the centered logcount value and the proportion of cells that express

the gene, respectively. **C**, Heatmap of centered logcounts of DEGs (adjusted p-value <0.05) between

EC subtypes, color indicates the centered logcount value. **D,E**, ORA analysis from genes upregulated in

Venous 1 compared to Venous 2 (see Figure S1A) using KEGG pathways (**D**) or TF motif enrichment (**E**),

color and size of the dots indicates the -log10 adjusted p-value and the number of genes in each

category (intersection size), respectively. **F**, TSNE plot of GSVA scores using the ATF4 dependent

geneset from Fan et al (see Methods), color-coded by the GSVA score. **G**, Stacked bar plots showing

cluster percentage in each condition, color-coded by cluster. **H**, Heatmap of centered logcounts of

DEGs (adjusted p-value < 0.05) specific for each EC subtype in each condition, color indicates the

centered logcount value. **I**, Bar plots showing the number of DEGs for each EC subtype in each

condition, color indicates the condition. **J**, Heatmap of centered values from ORA analysis over the

DEGs in ATF3/4$^+$ venous and capillary ECs between conditions, color indicates the centered values. **K,L**,

Heatmap of centered logcounts values of fatty acid metabolism and mesenchymal markers as defined

in Tombor et al (**K**) and core matrisome genes as defined in Hynes et al (**L**), color indicates the centered

values.

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Figure 3. **LYVE1^{hi}MHCII^{low} macrophages are the dominant macrophage population in PAD. A**, TSNE plot of Monocytes/Macrophages subtypes identified from human gastrocnemius muscle, color-coded by the identified subtypes. **B**, Dot plot of centered logcounts values from MoMac subtype-specific marker genes. Color and size of the dots indicate the centered logcounts values and the proportion of cells that express the gene, respectively. **C**, Heatmap of centered logcounts of DEGs (adjusted p-value <0.05) between MoMac subtypes, color indicates the centered logcount value. **D**, Stacked bar plots showing cluster percentage in each condition, color-coded by cluster. **E**, Representative images of gastrocnemius muscle from control (n=4) and PAD (n=6) stained for a pan-macrophage marker (CD68, green), LYVE1 (red), cell nuclei (Hoechst) and WGA (white) (Scale bar: 50 µm). **F-G**, Quantification of the percentage of LYVE1$^+$CD68$^+$ cells from the total of CD68$^+$ cells (**F**) and the total number of macrophages measured as the number of CD68$^+$ cells per mm^2 (**G**). **H**, TSNE plots showing GSVA scores for different macrophage activation processes as defined by Sanin et al (see Methods), color-coded by GSVA score. **I**, GSVA scores from different pro-inflammatory and pro-regenerative signatures as defined by Varga et al between LYVE1^{hi} and LYVE^{low} macrophages, color-coded by cluster. **J**, Heatmap of centered logcounts of DEGs (adjusted p-value < 0.05) specific for each MoMac subtype in each condition, color indicates the centered logcount value. **K**, Bar plot showing the number of DEGs for each MoMac subtype in each condition, color indicates the condition. Each dot represents a single patient in panels F and G. Student's t test (two-tailed, unpaired, parametric, ns > 0.05) was used in F and Mann-Whitney U test (two-tailed, unpaired, non-parametric, $^*p$ < 0.05). was used in G.

Figure 4. Cellular communication in human skeletal muscle is profoundly altered in PAD. A-B, Circle

plots showing the number (**A**) and weight (**B**) of inferred interactions between cell types in control and

PAD (left and middle plots) and bar plots quantifying these parameters (right plots). Color and width

of the edges indicates the sender and number or weight of interactions, respectively. **C-D**, Heatmaps

showing increased (red) or decreased (blue) number (**C**) and weight (**D**) of inferred interactions in PAD.

Labels in the left side indicates the cell type expressing the ligands and bottom labels the cell types

receiving them. **E**, Scatter plots showing the strength of the outgoing (x-axis) and incoming (y-axis)

interactions for each cell type in control (left) and PAD (right). Color and size of the dots indicate the

cell type and the number of inferred interactions, respectively. **F**, Bar plots showing the relative

information flow for significant communication pathways between control and PAD. **G-H**, Circle plots

showing MHC-II (**G**) and VEGF (**H**) communication in control (left) and PAD (right). Color and width of

the edges indicates the sender and weight of interactions, respectively.

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Figure 5. Cellular communication between ATF3/4$^+$ ECs and LYVE1^{hi}MHCII^{low} macrophages is altered during PAD. A-B, Heatmaps showing downregulated (**A**) and upregulated (**B**) ligand-receptor (LR) pairs in PAD. In each panel, left part indicates the ligand log fold change (LFC) in ATF3/4$^+$ ECs and right part the receptor LFC in LYVE1^{hi}MHCII^{low} macrophages from each LR pair. Color indicates the LFC. **C**, Heatmap showing the regulatory potential of ATF3/4$^+$ ECs ligands upregulated in PAD (left labels) over the downstream target genes upregulated in LYVE1^{hi}MHCII^{low} macrophages (top labels). Color indicates the regulatory potential. **D**, Dot plot showing ORA results from the predicted targets in panel C. Color and size of the dots indicate -log10 adjusted p-value and number of genes in each category (intersection size), respectively.

1216 **TABLES**

	Control	PAD
Number of subjects	7	10
Average age (years)	71.6±2.6	74.1±2.7
Ankle Brachial Index	1.16±0.08	0.78±0.13*
Fontaine classification I/IIa/IIb/III/IV	7/0/0/0/0	0/0/6/2/2
scRNAseq/Histology	4/4[†]	4/6
Gender (male/female)	7/0	10/0
Smoking (never/current/former)	2/1/4	3/5/2
Obesity (BMI > 30 kg/m^2)	1	0
Diabetes Mellitus	2	3
Hypertension	7	8
Hyperlipidemia	7	8
Statins	7	7
Anti-platelet agents	7	8
Antihypertensive drugs	7	7
Antidiabetic agents	2	3
Involved side (left/right)	3/4	5/5

1217 *Table 1.Patient demographics. Continuous values are presented as mean ± standard error of the mean. BMI indicates Body*
1218 *Mass Index. *p < 0.05, Student's t test (unpaired, two-tailed). [†] Sample from one subject was used for both scRNAseq and*
1219 *histology.*