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# Extracellular vesicles promote activation of pro-inflammatory cancer-associated fibroblasts in oral cancer 

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#### Abstract

Introduction: Oral squamous cell carcinoma (OSCC) is the most common form of head and neck cancer and has a survival rate of $\sim 50 \%$ over 5 years. New treatment strategies are sorely needed to improve survival rates-and a better understanding of the mechanisms underlying tumorigenesis is needed to develop these strategies. The role of the tumor microenvironment (TME) has increasingly been identified as crucial in tumor progression and metastasis. One of the main constituents of the TME, cancer-associated fibroblasts (CAFs), plays a key role in influencing the biological behavior of tumors. Multiple mechanisms contribute to CAF activation, such as TGF $\beta$ signaling, but the role of extracellular vesicles (EVs) in CAF activation in OSCC is poorly understood. Assessing the impact of oral cancer-derived EVs on CAF activation will help to better illuminate OSCC pathophysiology and may drive development of novel treatments options.


Methods: EVs were isolated from OSCC cell lines (Cal 27, SCC-9, SCC-25) using differential centrifugation. Nanoparticle tracking analysis was used for EV characterization, and Western blot to confirm the presence of EV protein markers. Oral fibroblasts were co-cultured with enriched EVs, TGFß, or PBS over 72 h to assess activation. Flow cytometry was used to evaluate CAF markers. RNA collected from fibroblasts was extracted and the transcriptome was sequenced. Conditioned media from the co-cultures was evaluated with cytokine array profiling.

Results: OSCC-derived EVs can activate oral fibroblasts into CAFs that are different from those activated by TGF $\beta$, suggesting different mechanisms of activation and different functional properties. Gene set enrichment analysis showed several upregulated inflammatory pathways in those CAFs exposed to OSCC-derived EVs. Marker genes for inflammatory CAF subtypes were also upregulated, but not in CAFs activated by TGF $\beta$. Finally, cytokine array analysis on secreted proteins revealed elevated levels of several pro-inflammatory cytokines from EV-activated CAFs, for instance IL-8 and CXCL5.

Discussion: Our results reveal the ability of OSCC-derived EVs to activate fibroblasts into CAFs. These CAFs seem to have unique properties, differing from TGF $\beta$-activated CAFs. Gaining an understanding of the interplay between

EVs and stromal cells such as CAFs could lead to further insights into OSCC tumorigenesis and potential novel therapeutics.

KEYWORDS<br>cancer-associated fibroblast, tumor microenvironment, oral squamous cell carcinoma, extracellular vesicles, inflammation, tumorigenesis

## 1 Introduction

Cancer represents one of the leading causes of mortality and morbidity among humans. Oral squamous cell carcinoma (OSCC) affects mainly the floor of the mouth and the tongue, and has a 5 -year survival rate of only $50 \%$. Frequent late-stage diagnoses and high rates of recurrence are key drivers of this poor prognosis. There remains a clear need to develop new approaches for treating this disease. A better understanding of the molecular basis of OSCC progression and metastasis is needed to provide new disease management options and treatments strategies.

Tumor progression is no longer recognized as relating only to genetic mutation and uncontrollable growth in cancer cells. Different types of growth factors and cytokines secreted by tumor cells and stromal cells within a tumor, as well as signaling pathways induced by cell-cell interactions, are all thought to play key roles in tumorigenesis and metastasis (Meurette and Mehlen, 2018; Vitale et al., 2019). The tumor microenvironment (TME) is a dynamic space composed of cellular components, including fibroblasts, immune cells, inflammatory cells, epithelial cells, endothelial cells, and the extracellular matrix (ECM). The TME, ECM, and cellular components can regulate and influence each other (Zhou and Lu, 2017; Meurette and Mehlen, 2018; Roswall et al., 2018; Vitale et al., 2019; Vuong et al., 2019; Osman et al., 2020). Specifically, cancer-associated fibroblasts (CAFs) are one of the main cell types in the TME and are known to play a key role in influencing the biological behavior of tumors (Shoucair et al., 2020). Hence, CAFs have recently become of great interest in targeted medicine and tumor immunotherapy.

CAFs are a group of activated fibroblasts with a mesenchymal cell lineage and heterogeneity in the TME. They secrete a variety of active factors to participate in the generation and maintenance of cancer cell stemness, immune regulation, angiogenesis, metabolic response, ECM remodeling, internal environment stability, therapeutic resistance, and other biological processes (Shoucair et al., 2020). Fibroblasts can be activated and differentiate into a CAF phenotype through growth factors (TGF $\beta$, HGF, FGF, PDGF), transcription factors (NF-kB and HSF-1), metalloproteinases, signal proteins, reactive oxygen species secreted by tumor cells or immune cells and by extracellular vesicles released by tumor cells (Dai et al., 2018; Ringuette Goulet et al., 2018; Ping et al., 2021). EVs are small membraneencapsuled structures released by cells and serve as key mediators in cell-cell communication. Stromal cells, particularly fibroblasts, take up large numbers of cancer cell-derived EVs when compared to epithelial cells, making them important targets of EV-mediated cross-talk (Lee et al., 2016). CAF activation through cancer cell-
derived EVs has however scarcely been shown in oral cancer (Ding et al., 2018; Jiang et al., 2019). We demonstrate the ability of OSCC-derived EVs to shape fibroblasts into CAFs that are different from canonical TGF $\beta$-activated CAFs. CAFs activated by OSCC tumor-derived EVs have a distinct expression and secretome profile as compared to both non-activated fibroblasts or those activated by TGF $\beta$, a finding that is suggestive of a unique function for these activated CAFs.

## 2 Materials and methods

### 2.1 Cell lines and cell culture

The oral cancer cell lines Cal 27 (ATCC ${ }^{\oplus}$ CRL $2095^{\text {TM }}$ ), SCC-9 (ATCC ${ }^{\ominus}$ CRL1629 ${ }^{\text {TM }}$ ), and SCC-25 (ATCC ${ }^{\ominus}$ CRL1628 ${ }^{\text {TM }}$ ) were obtained from the American Type Culture Collection (ATCC). Cal 27 cells were cultured in DMEM media supplemented with $10 \%$ FBS. SCC-9 and SCC-25 cells were cultured in DMEM/F12 media supplemented with $10 \%$ FBS and $400 \mathrm{ng} / \mathrm{mL}$ hydrocortisone. Cancer cell lines were used for no more than 2 months after thawing to prevent genetic drift. The human oral fibroblast line (HOrF \#2640) was obtained from ScienCell ${ }^{\text {TM }}$ Research Laboratories. Cells were cultured on coated surfaces using poly-L-lysine (ScienCell ${ }^{\text {TM }}$ Research Laboratories Cat \#0413) and in complete Fibroblast Medium (ScienCell ${ }^{\text {TM }}$ Research Laboratories Cat \#2301) containing $2 \%$ FBS. The HOrFs were used for no more than passage ten. All cells were cultured in a $37^{\circ} \mathrm{C}, 5 \% \mathrm{CO}_{2}$ incubator.

### 2.2 EV isolation

Cancer cell lines (Cal 27, SCC-9, and SCC-25) were seeded into 15 cm plates. Forty-eight hours before reaching $90 \%$ confluency, normal media was replaced after PBS wash with media containing $1 \%$ FBS without noticeably affecting growth. The FBS was depleted of EVs by ultracentrifugation at $110,000 \mathrm{~g}$ for 16 h in order to reduce contamination from bovine EVs. Cells were allowed to excrete EVs into the media for 48 h , after which the conditioned media was collected and subjected to multiple rounds of centrifugation as previously described (Théry et al., 2018). Dead cells and cellular fragments were removed from media with $4^{\circ} \mathrm{C}$ centrifugation at 300 g for $10 \mathrm{~min}, 2,000 \mathrm{~g}$ for 20 min , and $10,000 \mathrm{~g}$ for 30 min , with the precipitate being discarded at each interval. EVs were precipitated using an ultracentrifuge at $110,000 \mathrm{~g}$ at $4^{\circ} \mathrm{C}$ for 90 min , after which the supernatant was removed, and the pellet was rinsed with PBS. EVs were re-precipitated with an additional $110,000 \mathrm{~g}$ spin for 90 min . NanoSight (Malvern, United Kingdom) nanoparticle tracking analysis was assessed to determine EV particle data. Vesicle flow was analyzed in a diluted sample for optimal
particle range for the machine, and each sample was analyzed in triplicate for 60 s .

### 2.3 Fibroblasts co-cultured with EVs

HOrFs were seeded in six well plates with 2 mL of media. EVs, PBS, or TGF $\beta$ was then added three times with 24 h intervals in between (total time of co-culture: 72 h ). The amount of EVs added at each timepoint was determined after titration assays evaluated with microscopy (to determine cell death) and flow cytometry (using the same panel of CAF markers as described below to assess CAF activation). In detail, when HOrFs had reached $30 \%$ confluency, filtered EVs isolated from oral cancer cell lines and resuspended in PBS ( $25 \mu \mathrm{~L}$ per well) were added to the culture media ( $22.6 \times 10^{9} \mathrm{EV}$ particles per well) while $25 \mu \mathrm{~L}$ of PBS was added per well to the PBS controls and $20 \mathrm{ng} /$ well of TGF-beta 1 protein with carrier (R\&D systems, final concentration $10 \mathrm{ng} / \mu \mathrm{L}$ after reconstitution) resuspended in $25 \mu \mathrm{~L}$ PBS was added per well to the TGF $\beta$ controls. After an additional 24 h and 48 h , the media was replaced with $2 \mathrm{~mL} /$ well of EV supplemented fresh media ( $45.2 \times 10^{9} \mathrm{EV}$ particles per well) for samples co-cultured with EVs, $2 \mathrm{~mL} /$ well of PBS-supplemented fresh media to the PBS controls, and $20 \mathrm{ng} /$ well TGF $\beta$ (final concentration $10 \mathrm{ng} / \mathrm{mL}$ ) resuspended in $25 \mu \mathrm{~L}$ PBS supplemented fresh media ( $2 \mathrm{~mL} /$ well) for the TGFß controls.

### 2.4 Flow cytometry

HOrFs were lifted 24 h after the third treatment of EVs or TGF $\beta$ had been added, using EDTA $0.04 \%$. Cells were stained with the following monoclonal antibodies: anti-alpha-Smooth Muscle Actin (aSMA) (Novus Biologicals Cat \#NBP2-34760APC), anti- Human Fibroblast Activation Protein alpha (FAP) (R\&D Systems Cat \#FAB3715P-100), anti-Podoplanin (PDPN) (BioLegend Cat \#337006), anti-CD29 (BioLegend Cat \#303020), and anti-PDGFR $\beta$ (Cell Signaling Technology Cat \#3169S) along with Brilliant Violet $421^{\mathrm{TM}}$ Donkey anti-rabbit IgG (BioLegend Cat \#406410). All antibodies were titrated for optimum concentration before use, with dilutions ranging from 1:50 to 1:400. eBioscience ${ }^{\mathrm{TM}}$ Foxp3/Transcription Factor Staining Buffer Set (Invitrogen) was used for staining intracellular protein. Cells were identified based on forward and side scatter properties and singlets. Dead cells were excluded using Fixable Viability Dye eFluor (ThermoFisher Cat \#65-0865-14). Gates on antibodies were performed using fluorescence minus one (FMO) controls. Cells were analyzed on an LRSFortessa analyser (BD Biosciences), and data were processed using FlowJo software (Tree Star, Inc., Ashland, United States). EV co-cultures for each of the oral SCC cell lines were performed on different days, each with their own controls (PBS and TGF $\beta$ ), hence we observe some variability among the various control samples when analyzed with flow cytometry.

### 2.5 RNA extraction from HOrFs and RNA sequencing

RNA from HOrFs co-cultured with EVs, TGF $\beta$, or PBS (three replicates per group) was extracted in TRIzol per manufacturer
instructions, 24 h after the third time EVs had been added. RNA expression profiling was performed at the Biomedical Research Centre at the University of British Columbia with 7,000$10,000 \mathrm{ng}$ of total RNA as input. Sample quality control was performed using the Agilent 2,100 Bioanalyzer yielding RNA integrity number (RIN) 10 for all samples. Qualifying samples were then prepped following the standard protocol for the NEBnext Ultra ii Stranded mRNA (New England Biolabs). Sequencing was performed on the Illumina NextSeq2000 with Paired End $61 \mathrm{bp} \times 61 \mathrm{bp}$ reads. Sequencing data was demultiplexed using Illumina's bcl2fastq2. De-multiplexed read sequences were then aligned to the Homo sapiens/Mus musculous (PAR-masked)/hg19 or mm10) reference sequence using STAR aligner (https://www.ncbi.nlm.nih.gov/pubmed/ 23104886).

### 2.6 Data analysis of RNA sequencing

Sequencing analysis resulted in an average of 26 million reads per sample. Assembly and differential expression were estimated using Cufflinks (http://cole-trapnell-lab.github.io/cufflinks/) through bioinformatics apps available on Illumina Sequence Hub, using FPKM values and comparing treatment group (EV or TGFß treatment) versus control (PBS treatment). Corrected $p$-value $<0$. 05 was considered significant. Pre-ranked Gene set enrichment analysis (GSEA) was performed on the counts expressed genes in each treatment group, using chip platform Human Gene Symbol with Remapping and Hallmarks and Biocarta gene sets from MSigDB (Liberzon et al., 2011). Enrichment map visualization in GSEA was used together with Cytoscape 3.9.1. Expression-based heat map of signature marker genes for CAF subtypes was performed according to instructions (Babicki et al., 2016).

### 2.7 Cytokine array and ELISA

HOrFs were co-cultured with EVs for 72 h as previously described. Twenty-four hours after the third time EVs had been added, the media was removed and replaced with serum-free media (without FGS and FBS). Forty-eight hours later, the conditioned serum free media was collected, spun at $2,000 \mathrm{rpm}$ for 10 min at $4^{\circ} \mathrm{C}$. The supernatant was then analyzed for 80 cytokines using the Human Cytokine Array C5 (RayBiotech) as per manufacturer instructions. Upon 30 min of blocking the arrays at RT, 1 mL of undiluted conditioned media was added to each well and incubated overnight at $4^{\circ} \mathrm{C}$. After the first washing step, 1 mL of Biotinylated Antibody Cocktail was added to each well and incubated for 2 h at room temperature (RT). Upon second washing step, 2 mL of 1 X HRP-Streptavidin was added to each well and incubated for 2 h at RT. Membranes were then washed a third time and then arranged for Chemiluminescence Detection using Detection Buffer using the ChemiDoc MP Imaging system (BioRad). Background response was corrected using ImageJ Protein Array Analyzer (http://image.bio.methods.free.fr/dotblot.html). Dot values were measured using the same program. Normalization was performed per array manufacturer recommendations. Protein levels of CXCL5 and IL-8 in the cell

culture supernatant were determined using enzyme linked immunosorbent assays (ELISA) assays ( R and D Systems, Human IL-8 ELISA/CXCL8 and CXCL5/ENA-78 Quantikine ELISA Kits) as per manufacturer's instructions.

### 2.8 Western blot

Protein was collected from EVs, cells or cell lysates through lysis with radioimmunoprecipitation assay (RIPA) buffer supplemented with 1:100 Protease Inhibitor Cocktail Set III and Phosphatase Inhibitor Cocktail 1 \& 2 (Sigma-Aldrich). Protein was quantified using a Pierce BCA Protein Assay Kit (ThermoFisher). Ten $\mu \mathrm{g}$ of protein was separated using NuPAGE $4 \%-12 \%$ Bis-Tris Mini Protein Gels (ThermoFisher) and transferred to polyvinylidene difluoride membranes (Millipore). Membranes were blocked in $5 \%$ BSA, 1 X TBS, and $0.1 \%$ Tween- 20 at RT for 1 h . Membranes were then incubated overnight at $4^{\circ} \mathrm{C}$ with the appropriate primary
antibody: 1:1000 diluted anti-Alix (sc-53540), anti-Hsp-70 (sc-24), anti-Flotillin-1 (sc-74566), anti-CD81 (sc-166029), anti-CD63 (sc5275), and anti-GRP 94 (sc-393402) (all from Santa Cruz), and antiTGF $\beta$ (\#3711, Cell Signaling). After washing the membranes, they were subsequently incubated with peroxidase-conjugated 1 : 2000 Anti-Mouse (NXA931, GE Healthcare) or Anti-Rabbit (\#7074, Cell Signaling). Detection was performed using Amersham ECL Western Blotting Detection Kit (GE Healthcare) and Chemidoc MP Imaging System (Bio-Rad).

### 2.9 Statistics

For flow cytometric data: a $p$-value of 0.05 or less was considered statistically significant $\left(^{*} p<0.05,{ }^{* *} p<0.01,{ }^{* * *} p<0.001,{ }^{* * * *} p<\right.$ 0.0001 ). Statistical analyses were performed using GraphPad Prism software (version 9.0, GraphPad Software, La Jolla, CA). All data are shown as mean $\pm$ S.E.M. For cell culture data, $n$ equals the number of
separate wells within the same experiment. For differential expression analysis of RNA sequencing data, corrected $p$-value $<0.05$ was considered statistically significant. For gene set enrichment, FDR q-value $<0.25$ along with nominal $p$-value $<0.01$ was considered enriched. For enrichment map visualisation, $p$-value cutoff: 0.05 ; FDR q-value cutoff: 0,1 ; overlap coefficient 0.5 . Heatmap on specific CAF genes was done using z-scores calculated from FPKM values of differential expression analysis.

## 3 Results

### 3.1 EV characterization reveals high presence of small extracellular vesicles

To understand the impact of oral tumor-derived EVs on surrounding fibroblasts, EVs from oral cancer cell lines Cal 27, SCC-9, and SCC-25 were first collected by repeated ultracentrifugation (Théry et al., 2018). The majority of EVs finally pelleted at $100,000 \mathrm{~g}$ consisted of exosomes ranging from 100 to 135 nm in size according to NanoSight analysis (Figure 1A). A peak was also seen around 180 nm , possibly partially containing microvesicles (Gould and Raposo, 2013). Western blot on isolated EVs as well as whole cell protein (the latter being source material) was performed confirming the presence of EV protein by EV markers Alix, HSP70, Flotillin-1, and CD81 (Figure 1B). GRP94 (also called HSP90B1) is a protein found in the endoplasmic reticulum. This protein is not present in EVs smaller than 200 nm and therefore was used as a control. GRP94 could not be detected in the EV protein (Figure 1B). Taken together, the collection method, size distribution, and protein content revealed the presence of EVs of the exosome type in the precipitate. This was used for further experiments.

### 3.2 EVs from oral cancer cells activate fibroblasts into cancer-associated fibroblasts (CAFs)

The currently accepted method for defining CAFs requires a combination of morphological features and biomarkers (Sahai et al., 2020). Various subtypes of CAFs have been identified expressing different CAF marker patterns (Costa et al., 2018; Givel et al., 2018; Galbo et al., 2021). We hypothesized that EVs from oral cancer cells could induce CAF activation and would give rise to CAFs with a unique expression profile.

To test this, human oral fibroblasts (HOrFs) were co-cultured with either PBS, TGF $\beta$, or EVs isolated from oral cancer cell lines Cal 27, SCC-9, and SCC-25 cells for 72 h . Cell populations isolated from tumor tissue samples are defined as CAFs when they have an elongated spindle morphology, negative staining for nonmesenchymal biomarkers, and positive staining for mesenchymal biomarkers such as podoplanin (PDPN), CD29, alpha-smooth muscle actin ( $\alpha$ SMA), fibroblast activation protein (FAP), and platelet-derived growth factor beta (PDGF $\beta$ ) (Costa et al., 2018; Han et al., 2020). We analyzed CAFs by flow cytometry using antibodies for known CAF markers (FAP, PDPN, CD29, PDGFR $\beta$, and $\alpha$ SMA). Flow cytometry analysis revealed a
statistically significant increase in all three of the EV treatment groups compared to PBS for FAP, PDPN, and CD29 (Figures 1C-E). No difference was observed between the PBS and EV groups for PDGFR $\beta$ or aSMA (Figures 1F,G). TGF $\beta$, a known activator of CAFs, was used as a positive control, resulting in an upregulation of FAP, PDPN, CD29, and PDGFR $\beta$ when compared to PBS (Figures $1 \mathrm{C}-\mathrm{F}$ ). The level of $\alpha$ SMA was generally high in all fibroblasts (cocultured with EVs, TGF $\beta$, or PBS) revealing no or week significances between groups (Figure 1G). This was possibly due to fibroblast activation linked to in vitro settings with cells cultured on stiff plastic surfaces, as previously experienced (Landry et al., 2019). See full gating procedures and representative dot blots in Supplementary Figure S1 and MFI data in Supplementary Figure S2. Western blot for TGF $\beta$ was performed on lysates obtained from the oral cancer cell lines as well as corresponding EVs to determine if TGF $\beta$ present in EVs could be a source of CAF activation (Supplementary Figure S2). Latent TGF $\beta$ was detected in both EVs and cell line lysates whereas activated TGF $\beta$ was only detected in EV protein from SCC9 and SCC-25 cells, indicating a possible source of CAF activation.

### 3.3 RNA sequencing reveals different CAF signatures depending on activation pattern

To determine if oral fibroblasts activated by TGF $\beta$ differed from those activated by oral tumor EVs, RNA from the activated fibroblasts was collected and RNA sequencing was performed. RNA was extracted from the activated fibroblasts using the same co-culture protocol as described above. Differential expression analysis was performed using Cufflinks confirming an upregulation in the RNA sequencing data of the CAF markers CD29, FAP, and PDPN in fibroblasts co-cultured with EVs seen with flow cytometry (except CD29 for fibroblasts co-cultured with EVs from SCC-25 cells) (Table 1). As expected, several additional known CAF markers (Öhlund et al., 2014; Gascard and Tlsty, 2016; Costa et al., 2018) were upregulated in all treatment groups (TGF $\beta$ and EVs) compared to PBS control, although overexpression was not necessarily observed across all samples for all markers, indicating different mechanisms of activation and/or function for resulting CAFs (Table 1). Similarly, expression of Caveolin-1 and Cavin-1 was not observed in CAFs activated by OSCC-derived EVs, as expected when fibroblasts are activated into CAFs (Table 1) (Öhlund et al., 2014; Gascard and Tlsty, 2016; Costa et al., 2018). In looking at the differences in the number of highly differentially expressed (DE) genes ( $q$-value $>0.05$ and $\log 2$ fold change $>1$ ) compared to control cells treated with PBS, we see marked differences in the profile of TGF $\beta$-induced cells with a higher number of genes showing repression compared to the cells treated with EVs (Figure 2A), whereas the number of upregulated genes was relatively similar. The majority of differentially expressed genes in the cells with TGF $\beta$ treatment were unique to this treatment ( $88 \%$ of upregulated and $86 \%$ of repressed genes) (Figures 2B,C). Conversely, the majority of highly upregulated genes in the different cell line EV treatments showed an overlap in at least one other EV treatment, with $68 \%, 73 \%$, and $75 \%$ of genes upregulated in Cal 27, SCC-9, and SCC-25 EV treatments, respectively, showing an overlap. Taken together, while TGF $\beta$

TABLE 1 Differential expression analysis of CAF associated genes and whether genes where upregulated ( $\boldsymbol{p}<\mathbf{0 . 0 5}$ ) (gain) or downregulated ( $\boldsymbol{p}<\mathbf{0 . 0 5 )}$ (loss) or not (-) (see heading Affect), p-value, and relative fold change compared to PBS control (heading FC (log2)) in treatment groups with TGF $\beta$ or EVs compared to PBS. No upregulation in any treatment groups for the following genes: VIM, ACTA2, S100A4, NG2, FN1, MFAP5, ASPN, OGN, and SPARC. No downregulation in any treatment groups for the following gene: CD36.

| CAF associated gene | Genes expected to show increased expression in CAFs |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | HOrFs co-cultured with TGF $\beta$ |  |  | HOrFs co-cultured with cal 27 EVs |  |  | HOrFs co-cultured with SCC-9 EVs |  |  | HOrFs co-cultured with SCC-25 EVs |  |  |
|  | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ |
| FAP | Gain | 0.0003 | 2.035 | Gain | 0.0007 | 0.540 | Gain | 0.0003 | 0.479 | Gain | 0.0005 | 0.371 |
| PDPN | Gain | 0.0003 | 1.679 | Gain | 0.0007 | 1.909 | Gain | 0.0003 | 2.242 | Gain | 0.0005 | 1.426 |
| ITGB1 (CD29) | Gain | 0.0003 | 0.544 | Gain | 0.0182 | 0.264 | Gain | 0.0068 | 0.263 | - | 0.9247 | 0.014 |
| PDGFR $\beta$ | Gain | 0.0003 | 0.711 | Loss | 0.0177 | -0.253 | Loss | 0.0003 | -0.415 | - | 0.4460 | -0.094 |
| COL1A1 | Gain | 0.0003 | 0.873 | Loss | 0.0020 | -0.304 | - | 0.1616 | 0.150 | Gain | 0.0005 | 0.695 |
| COL1A2 | Gain | 0.0003 | 0.917 | - | 0.1824 | -0.172 | Gain | 0.0003 | 0.452 | Gain | 0.0005 | 0.793 |
| P4HA1 | - | 0.8021 | -0.031 | Gain | 0.0007 | 0.342 | Gain | 0.0003 | 0.434 | Gain | 0.0005 | 0.402 |
| P4HA2 | Gain | 0.0003 | 0.653 | - | 0.5734 | 0.099 | Gain | 0.0145 | 0.243 | Gain | 0.0005 | 0.349 |
| P4HB | Gain | 0.0066 | 0.257 | - | 0.3061 | 0.143 | Gain | 0.0247 | 0.218 | - | 0.2010 | 0.141 |
| TNC | Gain | 0.0003 | 0.403 | Gain | 0.0007 | 0.617 | Gain | 0.0003 | 0.364 | - | 0.4142 | 0.097 |
| COL11A1 | Gain | 0.0003 | 0.634 | - | 0.2957 | -0.189 | Gain | 0.0207 | 0.285 | Gain | 0.0005 | 0.504 |
| ZEB1 | Gain | 0.0003 | 0.450 | Gain | 0.0166 | 0.263 | - | 0.9907 | -0.002 | - | 0.1205 | -0.169 |
| MCT4 | - | 0.7737 | 0.034 | Gain | 0.0007 | 0.662 | Gain | 0.0003 | 0.645 | Gain | 0.0074 | 0.263 |
| MMP1 | Loss | 0.0003 | -1.375 | Gain | 0.0007 | 0.365 | Loss | 0.0379 | -0.205 | Loss | 0.0005 | -0.447 |
| MMP2 | Gain | 0.0003 | 1.147 | Gain | 0.0464 | 0.226 | Gain | 0.0063 | 0.259 | Gain | 0.0409 | 0.210 |
| MMP13 | Gain | 0.0069 | 1.970 | - | 1 | -0.178 | - | 1 | -1.092 | - | 1 | 0.024 |
| MMP14 | Gain | 0.0003 | 0.921 | - | 0.9571 | 0.016 | - | 0.1798 | -0.141 | - | 0.3165 | -0.114 |
| MMP19 | Gain | 0.0005 | 0.880 | - | 0.9960 | -0.004 | - | 0.0620 | -0.530 | - | 0.1941 | -0.395 |
| FOXF1 | Loss | 0.0003 | -0.333 | Gain | 0.0036 | 0.293 | - | 0.4194 | -0.095 | Loss | 0.0099 | $-0.281$ |
|  | Genes expected to show decreased expression in CAFs |  |  |  |  |  |  |  |  |  |  |  |
|  | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & (\log 2) \end{aligned}$ | Affect | $p$-value | $\begin{aligned} & \text { FC } \\ & \text { (log2) } \end{aligned}$ |
| Caveolin-1 (CAV1) | Loss | 0.0003 | -0.626 | Loss | 0.0007 | -0.838 | Loss | 0.0003 | -0.931 | Loss | 0.0005 | -0.411 |
| Cavin-1/PTRF | Gain | 0.0003 | 0.454 | Loss | 0.0007 | -0.545 | Loss | 0.0003 | -0.785 | Loss | 0.0005 | -0.551 |

was detected in the EVs and may be a source of activation the differences between the TGF $\beta$ and EV treatments give evidence to distinct mechanisms of activation (Figures 2B,C). This is especially true for EVs from the Cal 27 cell line as TGF $\beta$ was not detected in the Cal 27 EVs and CAF activation was still observed and the expression profiles from all the EV treated groups clustered together and separated from the TGF $\beta$ treatment (Figures 2B,C).

Interestingly, when looking at the most highly differentiallyexpressed genes (Figures 2D,E), two of the OSCC cell lines EVs (SCC-9 and Cal 27) showed multiple cytokine/chemokine genes being upregulated in the CAFs after treatment with their EVs.

Although SCC-25 did not have the same cohort of genes as highly differentially expressed, they still for the most part showed an obvious difference in the same genes, thus indicating that EV content plays a role in CAF expression. TGF $\beta$ treatment did not show this same trend (Figure 2E). Furthermore, when comparing the genes with the most highly differential expression (indicated in Figure 2D) for each treatment across all treatment types, the results of the EV-induced cells are much more similar to each other than the TGF $\beta$-induced cells. Choosing four genes (CXCL5, IL-8, POSTN, and PMEPA1), $q P C R$ validated the RNA sequencing data (Supplementary Figure S3). The increase in cytokine/chemokine expression and difference in expression from TGF $\beta$ treated

fibroblasts indicates that these fibroblasts have likely polarized into an inflammatory-like fibroblast, analogous to the "iCAF" phenotype found in tumors (Biffi et al., 2019; Galbo et al., 2021). For full differential expression analysis, please see Supplementary Table S1.

### 3.4 CAFs activated through EVs show unique pro-inflammatory properties

Pathway analysis was performed to determine the impact of EV activation compared to activation by TGF $\beta$, and to gain insight into the function of EV activated CAFs. Gene set enrichment analysis (GSEA) using Hallmarks gene sets was preformed revealing several enriched pathways (Supplementary Table S2), showing gene sets
significantly enriched (FDR q-value $<0.25$ and nominal $p$-value $<$ 0.01 ). Several of the significantly ( $p$-value $<0.01$ and FDR $>25 \%$ ) enriched pathways in the CAFs activated by the three types of EVs were common in all three EV conditions and were distinct from the TGF $\beta$ group except for the EMT pathway, which was found to be enriched in all groups. Interestingly, pathways linked to inflammation were upregulated in all three cultures activated by the EVs (Supplementary Table S2, red box). All pathways involved in inflammation are illustrated in enrichment blots (Supplementary Figure S4) and enriched pathways are visualized in bar plots (Supplementary Figure S4).

To determine possible interactions between different pathways, GSEA was performed using the more suitable Biocarta gene sets, since Hallmarks gene sets are created to avoid gene overlaps. An enrichment

## HOrFs co-cultured with Cal 27 EVs



HOrFs co-cultured with SCC-25 EVs


## HOrFs co-cultured with SCC-9 EVs



HOrFs co-cultured with TGF $\beta$


FIGURE 3
Enrichment map visualization. Commonalities between different enriched gene sets visualized through network analysis. Nodes are gene sets enriched (red circles) or depleted (blue circles). Lines link nodes with overlapping genes (bold line: high significance, thin line: moderate significance). Enrichment map visualization in GSEA was used together with Cytoscape 3.9.1. Enrichment Map Parameters: p-value cutoff: 0.05 ; FDR $Q$-value cutoff: 0.1; overlap coefficient: 0.5.
map was done visualizing enriched gene sets as a network, disclosing several overlapping genes between inflammatory pathways (Figure 3) (pathways described briefly in Supplementary Table S3). These maps emphasize the pro-inflammatory properties of CAFs activated through EVs from oral cancer cells. This was not seen in fibroblasts activated through TGF $\beta$ (Figure 3).

Several authors have described CAF subtypes in head and neck tumors (Costea et al., 2013; Patel et al., 2018; Galbo et al., 2021). Using several gene expression patterns, Galbo et al. defined six CAF subtypes (referred to as pan-CAF) (Galbo et al., 2021). To determine if the CAFs activated by oral cancer-associated EVs produced a specific CAF subtype (as defined by Galbo et al.) we analyzed the expression of gene sets that comprised each of the Galbo subtypes (Galbo et al., 2021). CAFs activated through co-culture with oral cancer EVs showed high presence of CAFs with pan-iCAF and pan-iCAF-2 signature (Figure 4). These two subsets have been described to have high expression of genes related to inflammation (Galbo et al., 2021). This was most evident in fibroblasts activated through co-culture with EVs from Cal 27 and SCC9 cells (Figure 4), and not seen in fibroblasts activated through TGF $\beta$.

CAFs activated with SCC-25-derived EVs where more similar to the other EV groups than to TGF $\beta$ activated CAFs but appear to have a distinct expression pattern. This is not surprising as the EV miRNA content of EVs from SCC-25 cells is also distinct from the EV content from SCC-9 and Cal 27 cells (see Supplementary Figure S5) (Dickman et al., 2017).

Taken together, pathway analysis, enrichment map visualization, and subtype analysis all suggest that oral CAFs activated through EVs from oral cancer cells show proinflammatory properties unique from CAFs activated through TGF $\beta$.

### 3.5 CAFs activated through EVs secrete the pro-inflammatory cytokines IL-8 and CXCL5

Cells with pro-inflammatory properties are known to secrete a distinct pattern of cytokines that promote inflammation (Kopf et al., 2010; Deckers et al., 2023). To further understand


FIGURE 4
Specific marker genes related to pan-CAF subsets (right). Expression high (yellow), middle (white), low (blue), missing data (grey). Marker genes linked to pan-iCAF and pan-iCAF-2 marked with red rectangle. Z-scores calculated from FPKM values in differential expression analysis of fibroblasts cocultured with TGF 3 (TGFb), EVs from Cal 27 (CAL27), SCC-9 (SCC9) or SCC-25 (SCC25) cells, compared to fibroblasts co-cultured with PBS.
the role of EV-activated CAFs, conditioned media from CAFs activated by oral cancer cell EVs was collected for cytokine analysis using a screening membrane-based antibody array analyzing 80 different cytokines. Out of a total of 80 cytokines profiled on the array, 76 were detected in at least one of the samples (Supplementary Table S4). The majority of cytokines were detected at similar levels compared to the PBS control; however, 16 were differentially expressed in conditioned media from at least 1 cell line cultured with EVs as compared to the PBS control. The majority of proteins were over-expressed (14/16) and only two proteins were under-expressed compared to the PBS control in one of the cell lines (Cal 27) (Table 2). The most highly differentially expressed proteins in the conditioned media from cells cultured with EVs were IL-8 (CXCL8) and CXCL5 (ENA-78). These two factors had increased relative expression in two out of the three EV treated cell lines (Cal 27 and SCC-9) (Figures 5A-F). To validate the data, an ELISA was performed on the conditioned media from CAFs activated by oral cancer cell EVs or TGF $\beta$, displaying CAFs activated by Cal 27 or SCC-9derived EVs to secrete upregulated levels of IL-8 and CXCL5 compared to PBS treated fibroblasts (Figures 5G, H). Correspondingly, IL-8 and CXCL5 were significantly upregulated in CAFs activated through oral cancer cell EVs (from Cal 27, SCC-9, and SCC-25 cells) according to differential expression analysis of RNA sequencing data (Supplementary Table S1).

## 4 Discussion

Several subtypes of CAFs have been identified (Costea et al., 2013; Costa et al., 2018; Givel et al., 2018; Patel et al., 2018; Galbo et al., 2021). However, their specific function and how they are generated is not fully understood.

TGF $\beta$ family ligands constitutes well-established activating signals for myofibroblast-like fibroblasts (Sahai et al., 2020). In this study we demonstrate that CAFs activated through oral cancer-derived EVs induce an iCAF phenotype. GSEA of CAFs activated by oral cancer-derived EVs revealed upregulation of several pathways linked to inflammation including TNFa, IL6_JAK_STAT3 pathway, and IL2_ STAT5 signaling pathway (Supplementary Table S2, red box). The NF-кB family of transcription factors are widely described as central for inducing genes regulating an inflammatory response, both through the innate and adaptive immune system (Liu et al., 2017). The IL-6 pathway stimulates acute phase responses and immune reactions. Elevated expression of IL-6 leads to activation of JAK-STAT signaling, a cornerstone to cancer progression in particular linked to STAT3 activation, both through tumor growth and metastasis as well as modulation of immune surveillance (Brooks and Putoczki, 2020). CAFs activated by oral cancer-derived EVs revealed upregulation of the pro-inflammatory gene IL-6, not seen in

TABLE 2 Relative fold change values for differentially expressed proteins on cytokine antibody array.

| Upregulated proteins | Relative fold change compared to PBS control |  |  |
| :---: | :---: | :---: | :---: |
|  | Cal27 | SCC-9 | SCC-25 |
| IL-8 | 4.480 | 6.044 | 1.502 |
| CXCL5 | 3.136 | 4.997 | 1.342 |
| GRO-alpha | 1.272 | 5.170 | 1.668 |
| MCP-1 | 0.813 | 2.047 | 1.086 |
| GRO | 1.384 | 2.824 | 0.903 |
| IL-1alpha | 2.235 | 1.612 | 1.423 |
| IL-10 | 2.390 | 1.850 | 1.233 |
| TNF-beta | 1.001 | 0.865 | 7.263 |
| TNF-alpha | 0.881 | 0.865 | 4.272 |
| EGF | 1.178 | 0.958 | 3.486 |
| IGF-1 | 1.850 | 1.026 | 3.292 |
| TGF-beta 1 | 1.042 | 0.989 | 3.288 |
| Eotaxin-3 | 1.491 | 0.857 | 3.229 |
| PIGF | 1.155 | 1.011 | 2.067 |
| Downregulated Proteins | Relative Fold Change Compared to PBS Control |  |  |
|  | Cal27 | SCC-9 | SCC-25 |
| HGF | 0.37 | 0.86 | 0.71 |
| Osteoprotegerin | 0.48 | 1.11 | 0.96 |

CAFs activated by TGF $\beta$ (Supplementary Table S1) and iCAFs are known to secrete IL-6 (Biffi et al., 2019).

CAFs activated by EVs also showed upregulated genes in response to low oxygen levels (hypoxia pathway), not seen in CAFs activated by TGF $\beta$. Hypoxia related genes NF-кB, HIF1 $\alpha$, HK2, and PFKL (Ikeda et al., 2020; Baqlouq et al., 2021) were upregulated in CAFs activated by oral cancer-derived EVs (Supplementary Table S1). Only NF- $\kappa$ B was upregulated in CAFs activated by TGF $\beta$ (Supplementary Table S1). Tumors with high proportion of hypoxic cells have worse therapeutic outcome linked both to surgery and radiation, and patients with hypoxic tumors have increased risk for metastases and worse overall survival (Wadsworth et al., 2022). Hypoxia has previously been linked to IL-1 induced iCAFs (Biffi et al., 2019; Schwörer et al., 2023).

The Interferone_gamma_response and the complement pathways were also upregulated in CAFs activated by EVs. In the TME, IFN $-\gamma$ and the complement system consistently orchestrate both pro-tumorigenic and antitumor immunity overall (Jorgovanovic et al., 2020; Revel et al., 2020). The IFN- $\gamma$ pathway has been shown to be enriched in iCAFs, with known roles in cancer progression and disease pathophysiology (Öhlund et al., 2014).

GSEA revealed several overlapping genes between inflammatory pathways, strengthening the suggested pro-inflammatory profile of CAFs activated through oral cancer cell EVs. In line with these
results, cytokine analysis of the CAF secreted proteins revealed 14 upregulated cytokines in EV activated CAF supernatant compared to supernatant from fibroblasts that received only PBS, with IL-8 and CXCL5 being the cytokines most upregulated (Figure 5). IL-8 and CXCL5 were also among the most highly over-expressed genes in the EV activated CAFs compared to the PBS treatment (Supplementary Table S1). The possible importance of IL-8 can not be underestimated. IL-8 (CXCL8) is a proinflammatory chemokine that induces neutrophil degranulation and is vital during systemic inflammation (Bernhard et al., 2021). In cancer, it enhances tumor cell growth and metastasis through epithelial to mesenchymal transition (Fernando et al., 2011), formation of neutrophil extracellular traps (Alfaro et al., 2016), angiogenesis and infiltration of inflammatory cells that suppress antitumor CD3+/CD8+ T cell functions (David et al., 2016). Cancer immunology has to a large extent been focused on T cells and CD3+/ $\mathrm{CD} 8+\mathrm{T}$ cell response is linked to cancer survival (Fridman et al., 2012; Jimenez et al., 2021). The role of neutrophils in cancer is heterogenous. However, most studies support a pro-tumoral role for these abundant myeloid cells (Millrud et al., 2017; Hedrick and Malanchi, 2022). The effects of IL-8 linked to neutrophils and CD3+/CD8+ cells are thus of great importance in tumor biology. Additionally, high plasma levels of IL-8 have been associated with decreased efficiency of PD-L1 blockade in metastatic urothelial and


FIGURE 5
Cytokine antibody array profiles (A-F) and ELISA (G-H) of conditioned media from HOrF cell culture media upon 72 h of coculture. (A) Co-culture with PBS. (B) Co-culture with EVs derived from Cal 27 cells. (C) Co-culture with EVs derived from SCC-9 cells. (D) Co-culture with EVs derived from SCC-25 cells. Positive control spots are outlined in green and negative in gray. Spots corresponding to IL-8 are highlighted in blue and CXCL5 in red. (E) Bar graphs indicating increased protein expression of IL-8 and (F) CXCL5. ELISA on (G) IL-8 and (H) CXCL5.
renal cell carcinoma (Yuen et al., 2020). The small inflammatory cytokine CXCL5 activates neutrophils and mediates neutrophil infiltration. CXCL5 directly induces cancer cell proliferation and invasion and has been reported to be over-expressed in several cancer forms, including squamous cell cancer (Miyazaki et al., 2006; Wente et al., 2006; Park et al., 2007; Wong et al., 2007; Kuo et al., 2011). Altogether, this suggests CAFs activated by tumor EVs can shape the immunosuppressive microenvironment.

CAFs have been shown to have both pro-tumorigenic as well as anti-tumorigenic effects. The presence of CAF subtypes with different marker patterns at least partially explains this. There are
several studies describing CAF subtypes in cancer. In the areas of breast and ovarian cancer, CAF subtypes have been defined based on multicolor flow cytometry yielding the four subtypes depending on how they stained for the fibroblast markers FAP, CD29, aSMA, S100-A4, PDGFR $\beta$, and CAV1, accumulating differently in different tumors (Costa et al., 2018; Givel et al., 2018). The oral fibroblasts we describe in this manuscript, that were activated into CAFs through oral cancer cell-derived EVs, most likely constitute the CAF-S1 subtype, positive for both CD29 and FAP, while lacking expression of CAV1 (Table 1). CAF-S1 is described as having immunosuppressive functions by attracting CD25+FOXP3+ T cells. This effect was shown to be dependent on CXCL12 $\beta$ isoform mediated by IL-6 (Givel et al., 2018). CAFs activated by OSCC-derived EVs revealed in line with this upregulated IL-6 (Supplementary Table S1). However, CXCL12 showed no upregulation, possibly due to inability at RNA sequencing and cytokine profiling to distinguish the $\beta$-isoform from the $\alpha$ isoform (Supplementary Table S1-S4). Further functional studies investigating what effects oral CAFs activated through oral cancerderived EVs have on CD25+FOXP3+ T cells would hence be highly interesting. As suggested by Costa et al., targeting CAF subtypes might be a promising therapeutic approach as a complement to established treatment management.

CAF subtypes have also been described in oral cancer (Costea et al., 2013; Patel et al., 2018; Galbo et al., 2021). Using RNA sequencing, Galbo et al. (Galbo et al., 2021) compared CAFs across multiple cancer types, including head and neck squamous cell carcinoma, to identify CAF subtypes characterized by different cell-surface pattern and with diverse critical molecular pathways activated. According to our data, CAFs activated through oral cancer-derived EVs showed similar gene expression patterns as previously described pan-iCAFs and pan-iCAF-2 with high expression of genes related to inflammation (Figure 4). This was not seen in CAFs activated through TGF $\beta$, displaying more similarities with pan-pCAFs and pan-dCAFs. Pan-iCAFs and pan-iCAF-2 have low expression of collagens but high expression of proinflammatory molecules and the transcription factors NR1H3 and the NFkB subunit RELB, both promoting inflammatory transcriptional programs (Galbo et al., 2021). These results suggests that pan-iCAFs and pan-iCAFs-2 are potentially associated with a molecular microenvironment with enhanced immune cell activation. CAFs activated by SCC-25-derived EVs did not display pan-iCAFs and pan-iCAFs-2 as distinct as CAFs activated through EVs from Cal 27 or SCC-9 cells (Figure 4). This suggest the cell specific EV content may be crucial for CAF subtype differentiation. Several components in EV cargo have been shown to affect stromal cells including miRNAs, TGF $\beta$, cells surface proteins, and other proteins, as well as mRNAs and lncRNAs (Dai et al., 2018; Shoucair et al., 2020; Zhang et al., 2022). The EV content that drives CAF activation is still under investigation, however it is tempting to speculate miRNA cargo to be of importance (Supplementary Figure S4). Intercellular differences were seen regarding EV miRNA content that could explain the small differences seen in activation of inflammatory CAFs. Future studies looking at key players in EV cargo activating inflammatory CAFs and at tissue samples for the various CAF
subtypes that are actually present in patients and in what frequency would be crucial in understanding the role for specific CAF subtypes.

## Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found below: https://www.ncbi.nlm.nih.gov/geo/ query/acc.cgi?acc=GSE233618.

## Ethics statement

Ethical approval was not required for the studies on humans in accordance with the local legislation and institutional requirements because only commercially available established cell lines were used.

## Author contributions

$\mathrm{JA}, \mathrm{KB}$, and CG contributed to the conception and design of the study. JA, RT, and C-ML performed the experiments. JA and RT performed the statistical analysis. JA wrote the first draft of the manuscript. JA, RT, C-ML, KB, and CG wrote sections of the manuscript. All authors contributed to the article and approved the submitted version.

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## References

Alfaro, C., Teijeira, A., Oñate, C., Pérez, G., Sanmamed, M. F., Andueza, M. P., et al. (2016). Tumor-produced interleukin-8 attracts human myeloid-derived suppressor cells and elicits extrusion of neutrophil extracellular traps (NETs). Clin. Cancer Res. 22 (15), 3924-3936. doi:10.1158/1078-0432.ccr-15-2463

Babicki, S., Arndt, D., Marcu, A., Liang, Y., Grant, J. R., Maciejewski, A., et al. (2016), Heatmapper: web-enabled heat mapping for all. Nucleic Acids Res. 44 (W1), W147-W153. doi:10.1093/nar/gkw419
Baqlouq, L., Zihlif, M., Hammad, H., and Thaib, T. M. A. (2021). Determining the relative gene expression level of hypoxia related genes in different cancer cell lines. Curr. Mol. Pharmacol. 14 (1), 52-59. doi:10.2174/1874467213666200521081653
Bernhard, S., Hug, S., Stratmann, A. E. P., Erber, M., Vidoni, L., Knapp, C. L., et al. (2021). Interleukin 8 elicits rapid physiological changes in neutrophils that are altered by inflammatory conditions. J. Innate Immun. 13 (4), 225-241. doi:10. 1159/000514885

Biffi, G., Oni, T. E., Spielman, B., Hao, Y., Elyada, E., Park, Y., et al. (2019). IL1Induced JAK/STAT signaling is antagonized by TGF $\beta$ to shape CAF heterogeneity in pancreatic ductal adenocarcinoma. Cancer Discov. 9 (2), 282-301. doi:10.1158/2159-8290.cd-18-0710

Brooks, A. J., and Putoczki, T. (2020). JAK-STAT signalling pathway in cancer. Cancers (Basel) 12 (7), 1971. doi:10.3390/cancers12071971
Costa, A., Kieffer, Y., Scholer-Dahirel, A., Pelon, F., Bourachot, B., Cardon, M., et al. (2018). Fibroblast heterogeneity and immunosuppressive environment in human breast cancer. Cancer Cell 33 (3), 463-479. doi:10.1016/j.ccell.2018. 01.011

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## Conflict of interest

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

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## Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fcell.2023.1240159/ full\#supplementary-material

Costea, D. E., Hills, A., Osman, A. H., Thurlow, J., Kalna, G., Huang, X., et al. (2013). Identification of two distinct carcinoma-associated fibroblast subtypes with differential tumor-promoting abilities in oral squamous cell carcinoma. Cancer Res. 73 (13), 3888-3901. doi:10.1158/0008-5472.can-12-4150

Dai, G., Yao, X., Zhang, Y., Gu, J., Geng, Y., Xue, F., et al. (2018). Colorectal cancer cell-derived exosomes containing miR-10b regulate fibroblast cells via the PI3K/Akt pathway. Bull. Cancer 105 (4), 336-349. doi:10.1016/j.bulcan.2017.12.009
David, J. M., Dominguez, C., Hamilton, D. H., and Palena, C. (2016). The IL-8/IL-8R Axis: A double agent in tumor immune resistance. Vaccines (Basel) 4 (3), 22. doi:10. 3390/vaccines4030022
Deckers, J., Anbergen, T., Hokke, A. M., de Dreu, A., Schrijver, D. P., de Bruin, K., et al. (2023). Engineering cytokine therapeutics. Nat. Rev. Bioeng. 1 (4), 286-303. doi:10. 1038/s44222-023-00030-y

Dickman, C. T., Lawson, J., Jabalee, J., MacLellan, S. A., LePard, N. E., Bennewith, K. L., et al. (2017). Selective extracellular vesicle exclusion of miR-142-3p by oral cancer cells promotes both internal and extracellular malignant phenotypes. Oncotarget 8 (9), 15252-15266. doi:10.18632/oncotarget. 14862
Ding, L., Ren, J., Zhang, D., Li, Y., Huang, X., Hu, Q., et al. (2018). A novel stromal lncRNA signature reprograms fibroblasts to promote the growth of oral squamous cell carcinoma via LncRNA-CAF/interleukin-33. Carcinogenesis 39 (3), 397-406. doi:10. 1093/carcin/bgy006

Fernando, R. I., Castillo, M. D., Litzinger, M., Hamilton, D. H., and Palena, C. (2011). IL-8 signaling plays a critical role in the epithelial-mesenchymal transition of human carcinoma cells. Cancer Res. 71 (15), 5296-5306. doi:10.1158/0008-5472.can-11-0156

Fridman, W. H., Pagès, F., Sautès-Fridman, C., and Galon, J. (2012). The immune contexture in human tumours: impact on clinical outcome. Nat. Rev. Cancer 12 (4), 298-306. doi:10.1038/nrc3245
Galbo, P. M., Jr., Zang, X., and Zheng, D. (2021). Molecular features of cancerassociated fibroblast subtypes and their implication on cancer pathogenesis, prognosis, and immunotherapy resistance. Clin. Cancer Res. 27 (9), 2636-2647. doi:10.1158/1078-0432.ccr-20-4226

Gascard, P., and Tlsty, T. D. (2016). Carcinoma-associated fibroblasts: orchestrating the composition of malignancy. Genes Dev. 30 (9), 1002-1019. doi:10.1101/gad. 279737.116

Givel, A. M., Kieffer, Y., Scholer-Dahirel, A., Sirven, P., Cardon, M., Pelon, F., et al. (2018). miR200-regulated CXCL12 $\beta$ promotes fibroblast heterogeneity and immunosuppression in ovarian cancers. Nat. Commun. 9 (1), 1056. doi:10.1038/ s41467-018-03348-z

Gould, S. J., and Raposo, G. (2013). As we wait: coping with an imperfect nomenclature for extracellular vesicles. J. Extracell. Vesicles 2, 20389. doi:10.3402/ jev.v2i0. 20389

Halim, C. E., Deng, S., Ong, M. S., and Yap, C. T. (2020). Involvement of STAT5 in oncogenesis. Biomedicines 8 (9), 316. doi:10.3390/biomedicines8090316

Han, C., Liu, T., and Yin, R. (2020). Biomarkers for cancer-associated fibroblasts. Biomark. Res. 8 (1), 64. doi:10.1186/s40364-020-00245-w

Hedrick, C. C., and Malanchi, I. (2022). Neutrophils in cancer: heterogeneous and multifaceted. Nat. Rev. Immunol. 22 (3), 173-187. doi:10.1038/s41577-021-00571-6

Ikeda, S., Abe, F., Matsuda, Y., Kitadate, A., Takahashi, N., and Tagawa, H. (2020) Hypoxia-inducible hexokinase-2 enhances anti-apoptotic function via activating autophagy in multiple myeloma. Cancer Sci. 111 (11), 4088-4101. doi:10.1111/cas. 14614

Jiang, E., Xu, Z., Wang, M., Yan, T., Huang, C., Zhou, X., et al. (2019). Tumoral microvesicle-activated glycometabolic reprogramming in fibroblasts promotes the progression of oral squamous cell carcinoma. Faseb J. 33 (4), 5690-5703. doi:10 1096/fj.201802226R

Jimenez, D. G., Sobti, A., Askmyr, D., Sakellariou, C., Santos, S. C., Swoboda, S., et al. (2021). Tonsillar cancer with high CD8(+) T-cell infiltration features increased levels of dendritic cells and transcriptional regulation associated with an inflamed tumor microenvironment. Cancers (Basel) 13 (21), 5341. doi:10.3390/cancers13215341
Jorgovanovic, D., Song, M., Wang, L., and Zhang, Y. (2020). Roles of IFN- $\gamma$ in tumor progression and regression: A review. Biomark. Res. 8, 49. doi:10.1186/s40364-020-00228-x

Kopf, M., Bachmann, M. F., and Marsland, B. J. (2010). Averting inflammation by targeting the cytokine environment. Nat. Rev. Drug Discov. 9 (9), 703-718. doi:10.1038/nrd2805
Kuo, P. L., Chen, Y. H., Chen, T. C., Shen, K. H., and Hsu, Y. L. (2011). CXCL5/ ENA78 increased cell migration and epithelial-to-mesenchymal transition of hormoneindependent prostate cancer by early growth response-1/snail signaling pathway. J. Cell Physiol. 226 (5), 1224-1231. doi:10.1002/jcp. 22445

Landry, N. M., Rattan, S. G., and Dixon, I. M. C. (2019). An improved method of maintaining primary murine cardiac fibroblasts in two-dimensional cell culture. Sci. Rep. 9 (1), 12889. doi:10.1038/s41598-019-49285-9

Lee, T. H., Chennakrishnaiah, S., Meehan, B., Montermini, L., Garnier, D., D’Asti, E., et al. (2016). Barriers to horizontal cell transformation by extracellular vesicles containing oncogenic H-ras. Oncotarget 7 (32), 51991-52002. doi:10.18632/ oncotarget. 10627

Liberzon, A., Subramanian, A., Pinchback, R., Thorvaldsdóttir, H., Tamayo, P., and Mesirov, J. P. (2011). Molecular signatures database (MSigDB) 3.0. Bioinformatics 27 (12), 1739-1740. doi:10.1093/bioinformatics/btr260

Liu, T., Zhang, L., Joo, D., and Sun, S. C. (2017). NF-кB signaling in inflammation Signal Transduct. Target Ther. 2, 17023-. doi:10.1038/sigtrans.2017.23

Meurette, O., and Mehlen, P. (2018). Notch signaling in the tumor microenvironment. Cancer Cell 34 (4), 536-548. doi:10.1016/j.ccell.2018.07.009

Millrud, C. R., Kagedal, A., Kumlien Georén, S., Winqvist, O., Uddman, R., Razavi, R., et al. (2017). NET-producing CD16(high) CD62L(dim) neutrophils migrate to tumor sites and predict improved survival in patients with HNSCC. Int. J. Cancer 140 (11), 2557-2567. doi:10.1002/ijc. 30671
Miyazaki, H., Patel, V., Wang, H., Edmunds, R. K., Gutkind, J. S., and Yeudall, W. A. (2006). Down-regulation of CXCL5 inhibits squamous carcinogenesis. Cancer Res. 66 (8), 4279-4284. doi:10.1158/0008-5472.can-05-4398

Öhlund, D., Elyada, E., and Tuveson, D. (2014). Fibroblast heterogeneity in the cancer wound. J. Exp. Med. 211 (8), 1503-1523. doi:10.1084/jem. 20140692

Osman, A., Afify, S. M., Hassan, G., Fu, X., Seno, A., and Seno, M. (2020). Revisiting cancer stem cells as the origin of cancer-associated cells in the tumor microenvironment: A hypothetical view from the potential of iPSCs. Cancers (Basel) 12 (4), 879. doi:10.3390/cancers12040879

Park, J. Y., Park, K. H., Bang, S., Kim, M. H., Lee, J. E., Gang, J., et al. (2007). CXCL5 overexpression is associated with late stage gastric cancer. J. Cancer Res. Clin. Oncol. 133 (11), 835-840. doi:10.1007/s00432-007-0225-x

Patel, A. K., Vipparthi, K., Thatikonda, V., Arun, I., Bhattacharjee, S., Sharan, R., et al. (2018). A subtype of cancer-associated fibroblasts with lower expression of alphasmooth muscle actin suppresses stemness through BMP4 in oral carcinoma. Oncogenesis 7 (10), 78. doi:10.1038/s41389-018-0087-x
Ping, Q., Yan, R., Cheng, X., Wang, W., Zhong, Y., Hou, Z., et al. (2021). Cancerassociated fibroblasts: overview, progress, challenges, and directions. Cancer Gene Ther. 28 (9), 984-999. doi:10.1038/s41417-021-00318-4
Revel, M., Daugan, M. V., Sautés-Fridman, C., Fridman, W. H., and Roumenina, L. T. (2020). Complement system: promoter or suppressor of cancer progression? Antibodies (Basel) 9 (4), 57. doi:10.3390/antib9040057

Ringuette Goulet, C., Bernard, G., Tremblay, S., Chabaud, S., Bolduc, S., and Pouliot, F. (2018). Exosomes induce fibroblast differentiation into cancer-associated fibroblasts through TGF $\beta$ signaling. Mol. Cancer Res. 16 (7), 1196-1204. doi:10.1158/1541-7786. mcr-17-0784
Roswall, P., Bocci, M., Bartoschek, M., Li, H., Kristiansen, G., Jansson, S., et al. (2018). Microenvironmental control of breast cancer subtype elicited through paracrine platelet-derived growth factor-CC signaling. Nat. Med. 24 (4), 463-473. doi:10.1038/ nm. 4494

Sahai, E., Astsaturov, I., Cukierman, E., DeNardo, D. G., Egeblad, M., Evans, R. M., et al. (2020). A framework for advancing our understanding of cancer-associated fibroblasts. Nat. Rev. Cancer 20 (3), 174-186. doi:10.1038/s41568-019-0238-1

Schwörer, S., Cimino, F. V., Ros, M., Tsanov, K. M., Ng, C., Lowe, S. W., et al. (2023). Hypoxia potentiates the inflammatory fibroblast phenotype promoted by pancreatic cancer cell-derived cytokines. Cancer Res. 83 (10), 1596-1610. doi:10.1158/0008-5472. can-22-2316

Shoucair, I., Weber Mello, F., Jabalee, J., Maleki, S., and Garnis, C. (2020). The role of cancer-associated fibroblasts and extracellular vesicles in tumorigenesis. Int. J. Mol. Sci. 21 (18), 6837. doi:10.3390/ijms21186837

Théry, C., Witwer, K. W., Aikawa, E., Alcaraz, M. J., Anderson, J. D., Andriantsitohaina, R., et al. (2018). Minimal information for studies of extracellular vesicles 2018 (MISEV2018): A position statement of the international society for extracellular vesicles and update of the MISEV2014 guidelines. J. Extracell. Vesicles 7 (1), 1535750. doi:10.1080/20013078.2018.1535750
Vitale, I., Manic, G., Coussens, L. M., Kroemer, G., and Galluzzi, L. (2019). Macrophages and metabolism in the tumor microenvironment. Cell Metab. 30 (1), 36-50. doi:10.1016/j.cmet.2019.06.001
Vuong, L., Kotecha, R. R., Voss, M. H., and Hakimi, A. A. (2019). Tumor microenvironment dynamics in clear-cell renal cell carcinoma. Cancer Discov. 9 (10), 1349-1357. doi:10.1158/2159-8290.cd-19-0499

Wadsworth, B. J., Lee, C. M., and Bennewith, K. L. (2022). Transiently hypoxic tumour cell turnover and radiation sensitivity in human tumour xenografts. Br . J. Cancer 126 (11), 1616-1626. doi:10.1038/s41416-021-01691-5

Wente, M. N., Keane, M. P., Burdick, M. D., Friess, H., Büchler, M. W., Ceyhan, G. O., et al. (2006). Blockade of the chemokine receptor CXCR2 inhibits pancreatic cancer cell-induced angiogenesis. Cancer Lett. 241 (2), 221-227. doi:10.1016/j. canlet.2005.10.041
Wong, Y. F., Cheung, T. H., Lo, K. W., Yim, S. F., Siu, N. S., Chan, S. C., et al. (2007). Identification of molecular markers and signaling pathway in endometrial cancer in Hong Kong Chinese women by genome-wide gene expression profiling. Oncogene 26 (13), 1971-1982. doi:10.1038/sj.onc. 1209986

Yuen, K. C., Liu, L. F., Gupta, V., Madireddi, S., Keerthivasan, S., Li, C., et al. (2020). High systemic and tumor-associated IL-8 correlates with reduced clinical benefit of PD-L1 blockade. Nat. Med. 26 (5), 693-698. doi:10.1038/s41591-020-0860-1
Zhang, X., Zhang, Y., Qiu, X., Cai, J., Yang, Z., and Song, F. (2022). Extracellular vesicles derived from lung cancer cells induce transformation of normal fibroblasts into lung cancer-associated fibroblasts and promote metastasis of lung cancer by delivering lncRNA HOTAIR. Stem Cells Int. 2022, 3805013. doi:10.1155/2022/ 3805013

Zhou, Z., and Lu, Z. R. (2017). Molecular imaging of the tumor microenvironment. Adv. Drug Deliv. Rev. 113, 24-48. doi:10.1016/j.addr.2016.07.012

