Cancer cell – fibroblast crosstalk signature promotes macrophage recruitment via HB-EGF and stromal EGFR/MEK signalling

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Summary

Interactions between cells in the tumor microenvironment (TME) shape cancer progression and patient outcomes. To gain new insights into how the TME influences cancer outcomes, we adopt a distinct approach that exploits patterns of gene expression indicative of signalling between stromal fibroblasts and cancer cells, and demonstrate the prognostic significance of these gene signatures in multiple and independent squamous cell carcinoma cohorts. We discover that HB-EGF in cancer cells and EGFR in cancer associated fibroblasts represent a hub of tumor – stroma crosstalk through MAPK signalling and AP-1 transcription factor. We find divergent consequences for pathway activation in cancer cells compared to stromal fibroblasts that favours macrophages recruitment. Together these analyses demonstrate the utility of this new approach to interrogating the extent and consequences of TME crosstalk. By focusing on the transcriptional consequences of cancer cell-fibroblast interactions we derive prognostic signatures and uncover molecular mechanisms promoting fibroblast to macrophage communication.

Keywords

Fibroblasts; Cancer-associated fibroblasts; CAFs; EGFR; HB-EGF; stroma; Macrophages; Immune Microenvironment; gene signature; RAS

Abbreviations

Abbreviations

Activator protein 1 (AP-1)

Cancer associated fibroblasts (CAF)

Cervical squamous cell carcinoma (CESC)

Chemokine (C-C motif) ligand (CCL)

Confidence Interval (CI)

Conditioned medium (CM)

False discovery rate (FDR)

Gene-set enrichment analysis (GSEA)

Hazard ratio (HR)

Head and neck squamous cell carcinoma (HNSCC)

Heparin-binding epidermal growth factor-like growth factor (HB-EGF)

Human papillomavirus (HPV)

Interleukin (IL)

Lung squamous cell carcinoma (LUSC)

Overall Survival (OS)

Mean fluorescent intensity (MFI)

Normalized enrichment score (NES)

Non-treated (NT)

Pancreatic ductal adenocarcinoma (PDAC)

Peripheral blood mononuclear cell (PBMC)

Single-cell RNA sequencing (scRNAseq)

Standard deviation (SD)

Short tandem repeats (STR)

Squamous cell carcinoma (SCC)

The Cancer Genome Atlas (TCGA)

Transcription factor (TF)

Tumor microenvironment (TME)

Tumor necrosis factor (TNF)

Introduction

Cross-talk between cancer cells and non-malignant cells in the tumor microenvironment (TME) influences tumor growth, metastasis and therapy resistance [1]. Cancer-associated fibroblasts (CAFs) promote the invasion of cancer cells, reduce the efficacy of both targeted and cytotoxic therapies and modulate immune cell recruitment and functionality [2]. In addition, CAFs are correlated with a pro-tumorigenic immune landscape, including higher number of tumor-promoting myeloid cells [3], lower numbers of tumor-infiltrating lymphocytes [4] and worse prognosis [5] [6]. Of note, CAFs are linked to poor outcomes in squamous cell carcinoma (SCC) arising at multiple anatomical locations, including lungs (LUSC), cervix (CESC), and head and neck (HNSCC) [6–9].

Indeed, the development of advanced sequencing techniques allows multiple inferences about the type and abundance of different TME components, including CAFS, both from bulk transcriptome and genomic methylation data [7,10–13] . Both methods rely on the identification of cell type specific genes and the application of deconvolution strategies ultimately to infer the abundance of a particular population in a bulk dataset. However, these methods struggle to identify the functionally relevant interactions between cell types, such as signalling events [14,15] and the biological mechanisms associated with cell type crosstalk to be linked to patient outcomes remain incompletely understood.

Here, we propose an alternative approach to identify key players involved in tumor-stroma interaction. Instead of focusing on the abundance of CAFs or specific CAF subpopulations, we identify a signature indicative of signalling between cancer cells and CAFs. This signature is statistically associated with worse overall survival in multiple SCC subtypes. Moreover, we leverage information within the signature to identify a novel mechanism of interaction between cancer cells and CAFs. In co-culture, the RAS / MAPK pathway is strongly activated in both cell types, converging on the upregulation of Activator Protein 1 (AP-1) transcription factor (TF) components. We identify heparin-binding epidermal growth factor-like growth factor (HB-EGF) as a key mediator of cancer cell – CAF cross-talk, primarily expressed by cancer cells and able to upregulate the expression cytokines in CAFs. In turn, we demonstrate that this upregulation can drive attraction of macrophages, ultimately linked to worse overall survival in SCC patients.

Results

Meta-analysis of transcriptomic data of cancer cell and cancer-associated fibroblast co-cultures identifies gene signatures with prognostic value

To identify functionally and clinically relevant gene signatures based on cancer cell – CAF cross-talk, we performed a meta-analysis of transcriptomic datasets that compare co-cultures and mono-cultures of cancer cells and CAFs. The datasets were generated under similar direct co-culture conditions and employed cells derived from different models and cancer types [16,17]. We applied two strategies to derive gene signatures indicative of upregulated cancer cell – CAF signalling: i) selection of the most significantly enriched pathways via gene set enrichment analysis (GSEA) in co-culture for each transcriptomic dataset, followed by the selection of the up-regulated genes most frequently present in each enriched pathway (Figure 1A); ii) selection of the most up-regulated genes in co-culture for each transcriptomic dataset (Figure S1A). Using the first approach, we obtained a list of 5 genes upregulated in cancer cells and of 4 genes upregulated in CAFs upon co-culture, with one present in both. Therefore, this gene signature comprised of 8 genes (named CoCu8) (Figure 1B). The second approach led to a list of 2 genes upregulated in cancer cells and of 29 genes upregulated in CAFs upon co-culture, with one gene in common. Therefore, this gene signature consisted of 30 genes (named CoCu30) (Figure S1B).

We tested CoCu8 and CoCu30 on publicly available dataset of breast cancer – fibroblast co-cultures[18] confirming their relevance (Figure S2A). We also tested whether CoCu8 and CoCu30 are also upregulated when cancer cells are co-cultured with other stromal cell types and for this reason, we analysed a dataset of co-culture between 1205Lu cancer cells and HUVEC endothelial cells[19]: CoCu8 is neither enriched in cancer cells nor in endothelial cells when co-cultured, while CoCu30 shows only a weak correlation with co-culture conditions both in 1205Lu and HUVEC cells (Figure S2B). Thus, we establish new gene signatures specifically indicative of cancer cell - CAF communication.

We next sought to determine the clinical relevance of these signatures by testing in The Cancer Genome Atlas (TCGA). We found that high expression (Q4 vs Q1) of both gene signatures correlated with worse overall survival (OS) in multiple SCC datasets: HNSCC (CoCu8 Hazard Ratio (HR): 1.95, CoCu30 HR: 1.57), cervical squamous cell carcinoma (CESC, CoCu8 HR:2.79, CoCu30 HR: 2.08) and lung squamous cell carcinoma (LUSC, CoCu8 HR: 1.85, CoCu30 HR: 1.78) (Figure 1C; Figure S1C), with CoCu8 consistently showing a slightly higher hazard ratio (HR) compared to CoCu30 in all three tumor types. A multivariate analysis including relevant clinical variables such as age, sex and clinical stage to

evaluate the co-culture signatures effect as a continuous variable, confirmed the relevance of the signatures (Figure S3 for CoCu8, Figure S4 for CoCu30,). Of note, given the clinical and biological differences between HPV positive and negative tumors, we stratified HNSCC patients according to HPV status, observing that the strongest prognostic effect was visible in HPV positive samples both for CoCu8 (Figure 1D) and CoCu30 (HR: 5.47) (Figure S1D).

We validated the association between CoCu8 / CoCu30 and outcome in a second independent cohort of LUSC from the TRACERx study [20] (Figure 1E, Figure S1E). The multi-regional biopsies performed in the study enabled us to ask if the expression of the CoCu8 signature was uniform across tumors. Of the 117 tumors analysed, 86 (74%) showed concordant expression of CoCu8 in all regions, which is significantly greater than would be expected based on chance (Figure 1E). Similar results were observed with CoCu30 (Figure S1E). This indicates that cancer cell-fibroblast crosstalk is typically occurring across the whole tumor. Crucially, this analysis showed that the concordant up-regulation of CoCu8 / CoCu30 across tumour regions is associated with worse prognosis.

The crosstalk gene signature has greater prognostic power than fibroblast abundance

Given that CoCu8 / CoCu30 reflects cancer cell-fibroblast crosstalk, the signature might be predicted to correlate with fibroblast abundance. We performed a correlation analysis of CoCu8 / CoCu30 and methyl CIBERSORT signatures in TCGA datasets, showing that there is a statistically significant correlation between CAFs presence and CoCu8 signature (Figure S5A-B). Similar results were observed with the CoCu30 signature (Figure S5C-D). Then, we sought to validate these results in the TRACERx LUSC cohort and in a second independent UK_HPV positive cohort (Figure S5E-J). As methylome data was not available for these cohorts, we used fibroblast subtype gene signatures defined in a pan-cancer analysis by Galbo et al.[7]. We observed strong positive correlations between CoCu8 / CoCu30 signature and all of the fibroblast subtypes defined both in LUSC and UK_HPV positive HNSCC (Figure S5E-J).

We speculated that our signatures of active cancer cell-fibroblast crosstalk might have better prognostic power than simply CAF abundance. To test this, we analysed the abundance of CAFs using the methyl CIBERSORT deconvolution strategy in TCGA cohorts and probed links with overall survival[10]. This analysis indicated worse OS for HNSCC patients with higher CAF presence, but no significant differences were observed in CESC or LUSC (Figure S6A-B). Of note, both CoCu8 and CoCu30 signatures out-performed the methyl CIBERSORT method for HNSCC, CESC, and LUSC, which adds credence to our method of deriving a signature based on the interaction with fibroblasts, not simply their

abundance. Overall, these data indicate that CoCu8 / CoCu30 signatures correlates with CAF

abundance but have greater prognostic value than gene signatures used to infer CAF abundance.

Pathway enrichment analysis of cancer-associated fibroblast and cancer cell co-culture

reveals a consistent upregulation of AP-1 transcription factor genes

To obtain insight in the molecular basis of the cancer cell-CAF interactions, we analysed the pathways

that were enriched in all the transcriptomic datasets used to generate CoCu8 and CoCu30. This

revealed upregulation of multiple pathways linked to immune regulation, stress response, and

signalling (Figure 2A). Multiple genes belonging to the AP-1 transcription factor complex: *JUNB*, *FOS*,

FOSB were strongly enriched (Figure 2B). Moreover, *PLAUR* is regulated by AP-1 factors. As our meta-

analysis showed the strongest impact in the stratification of OS patients from HPV positive HNSCC, we

decided to explore *JUNB*, *FOS*, *FOSB* expression levels in 4 different co-culture combinations of human

HPV positive HNSCC cell lines, SCC154 and SCC47, and human oral CAFs, OCAF1 and OCAF2. Our results

confirmed that these three AP-1 TFs are upregulated when cancer cells and CAFs are in direct co-

culture, as compared to mono-culture (pooled RNA from both cell lines) (Figure 2C). Analysis of

indirect co-cultures [17] separated by a 0.4μm filter indicated that CoCu8 / CoCu30 are strongly enriched

with direct co-culture also when compared with indirect co-culture, implying that direct contact is

required for increased AP-1 TF expression (Figure S7A).

Interaction between cancer cells and cancer-associated fibroblasts is linked to increased RAS

activity

We next investigated possible mechanisms underlying the upregulation of the multifunctional AP-1

TFs both in cancer cells and CAFs[21]. RAS signalling via MAPK is known to be a major driver of AP-1 gene

expression[22,23]. Accordingly, we found that RAS signalling was strongly up-regulated upon direct co-

culture, as indicated by the enrichment of the KRAS_SIGNALLING_UP signature (Figure 2A) and of the

curated RAS84 gene signature [24], in all our cancer cell-CAF datasets (Figure 3A, Figure S7B).

To interrogate further the linkage between CAFs, the CoCu8 signature, and RAS signalling in HPV

positive patients from the TCGA cohort, we stratified them according to RAS activity. We split the

patient data into three groups (RAS84_0, RAS84_1 and RAS84max) according to the levels of RAS

activity as performed by East et al.[24]. This analysis shows that higher RAS activity (group RAS84_max)

correlates with worse OS, in a similar fashion to the effect observed with CoCu8 stratification (Figure

3B). Indeed, we observed a strong, positive correlation between RAS84 activity and CoCu8 expression
in both HPV positive HNSCC cohorts (TCGA - R = 0.79, Figure 3C and UK_HPV positive cohort - R = 0.84,
Figure S8A). We also observed a statistically significant enrichment in the extent of fibroblasts present
when RAS activity was higher in both cohorts (Figure 3D, Figure S8B).

Activation of the RAS pathway upon cancer cell-CAF co-culture was experimentally validated by co-
culturing SCC154 – OCAF1, which resulted in a strong increase of phosphorylated ERK1/2 (Figure 3E).
To test whether the RAS-MAPK pathway is responsible for the upregulation of *JUNB*, *FOS* and *FOSB*
genes, we used the MEK inhibitor, trametinib, in the SCC154-OCAF1 co-culture. We confirmed that
MEK inhibition downregulates ERK1/2 activation upon co-culture (Figure 3F) and observed a
significant downregulation of *JUNB*, *FOS* and *FOSB* genes (Figure 3G). Thus, multiple genes belonging
to the AP-1 TF complex are upregulated when cancer cells and CAFs are co-cultured and this is
mechanistically linked to the activation of RAS-MAPK kinase signalling.

HB-EGF activation is crucial to trigger RAS pathway signalling.

To explain why AP-1 TFs get upregulated in co-culture, we looked for possible activators of RAS-MAPK
signalling. We noted that *HB-EGF* was among the genes upregulated in all 6 transcriptomic datasets
together with *JUNB*, *FOS* and *FOSB* (Figure 2B). HB-EGF is an EGFR ligand and therefore can activate
MAPK pathway [25,26]. We evaluated the expression levels of all seven EGFR ligands. Importantly, *HB-*
EGF showed a strong and specific activation upon cancer cell - CAF co-culture (Figure 4A). Moreover,
HB-EGF expression strongly correlates with CoCu8 in HPV positive HNSCC patients' data from both
TCGA (R=0.6, p-value=2.2e-16) and UK_HPV positive (R=0.58, p-value=1e-8) datasets (Figure 4B).

HB-EGF can activate EGFR/MAPK as an un-cleaved pro-molecule at the plasma membrane [26,27] and, as
such, signalling by membrane-bound HB-EGF could explain the need of direct cell contact to trigger
the pathway. As membrane-bound HB-EGF should be expressed at about 20-25 kDa, we evaluated the
cellular levels of HB-EGF in OCAF1-SCC154 co-culture and found that it was strongly upregulated at
the protein level at a molecular weight previously reported as un-cleaved protein [28] (Figure 4C). We
also observed EGFR phosphorylation was increased upon direct co-culture (Figure 4C); importantly,
both these effects were abrogated by MEK inhibition (Figure 4D), suggesting the presence of a positive
feedback loop involving HB-EGF / EGFR / MAPK / AP-1 upon direct co-culture. Furthermore, the EGFR
inhibitor, afatinib, blocked AP-1 activation upon co-culture of SCC154-OCAF1 (Figure 4E).

These results suggest that HB-EGF might be the link that activates EGFR upon cancer cell – CAF co-
culture. Therefore, we evaluated the basal expression level of EGFR and HB-EGF in SCC154 and OCAF1
mono-cultures. Interestingly, SCC154 expressed HB-EGF at much higher levels than OCAF1, while both
cell types expressed similar levels of EGFR (Figure 4F). This suggests that both cell types can be reactive
to EGF ligands, but the activation of the positive feedback loop upon direct contact requires higher
levels of HB-EGF, expressed at the membrane of cancer cells. In that case, both cancer cells and CAFs
should be responsive to HB-EGF treatment, albeit with potentially different downstream effects. To
test this hypothesis, we incubated OCAF1 and SCC154 in mono-cultures with different concentrations
of HB-EGF. Firstly, HB-EGF caused an increase in proliferation in both cell types, as shown by a higher
proportion of EdU positive nuclei (Figure 4G). Moreover, SCC154 – but not OCAF1 – showed a scatter-
like phenotype when treated with high doses of HB-EGF, as visible from E-Cadherin staining (Figure
S8C). We then tested the effect of HB-EGF treatment on known CAF markers: after treatment of OCAF1
with HB-EGF for 48h we noticed a slight but significant downregulation of *ACTA2*, a CAF and
myofibroblast marker (Figure S8D). However, no effect was observed for other widely used CAF
markers (*FAP, LRRC15, FN1*) (Figure S8D).

Given these data, we asked whether HB-EGF expression in cancer cells is enough to induce
upregulation of AP-1 genes when cancer cells and CAFs are in co-culture: we therefore performed
knock down of HB-EGF in SCC154 (Figure S8E) and then co-cultured them with OCAF1. Importantly,
the downregulation of HB-EGF in SCC154 is enough to block the upregulation of *JUNB*, *FOS* and *FOSB*
when cancer cells and CAFs are co-cultured (Figure 4H).

These results point to an axis involving HB-EGF in cancer cells and EGFR in CAFs that activates MAPK /
AP-1 inducing a positive feedback loop when cancer cells and CAFs are co-cultured.

Stromal EGFR regulates cytokine expression and macrophage recruitment in tumor/stroma

interaction

To focus on the downstream effects of this crosstalk between cancer cells and CAFs and how HB-EGF
could affect CAFs functions and lead to unfavourable biology, we analysed scRNAseq dataset of HNSCC
with both malignant and non-malignant samples published by Choi et al. [29]. Interestingly, by using
myofibroblast and inflammatory markers (Figure S8F), we found that EGFR is mainly expressed by
inflammatory fibroblasts (iFibroblasts), but not myofibroblastic CAFs (myoFibroblasts), while HB-EGF
is mainly expressed by endothelial and epithelial cells (Figure 5A). This is in line with our preliminary

evidence that HB-EGF treatment in OCAF1 is likely more associated with inflammatory than
myofibroblast signature.

Given that high expression of EGFR is associated with inflammatory fibroblasts, we performed a
cytokine array of conditioned medium from pooled mono-cultures and co-cultures (Figure 5B):
interestingly, we found a strong upregulation of macrophage attraction and differentiation markers
(GM-CSF, CCL5, LIF). By cross-referencing the upregulated cytokines with genes regulated by RAS in
the RAS84 gene signature, we observed that LIF is both upregulated by co-culturing cancer cells and
CAFs and it is linked to RAS activity. We validated by qPCR that *LIF* is transcriptionally upregulated in
co-culture (Figure 5C) and that trametinib, MEK inhibitor, treatment is enough to downregulate its
expression (Figure 5D). Moreover, by reducing HB-EGF expression in cancer cells and then co-culturing
the cells with CAFs, we also observed strong downregulation of *LIF* expression (Figure 5E). Importantly,
HB-EGF treatment in cancer cells and CAFs mono-cultures shows that LIF is strongly upregulated only
by CAFs, indicating that these are the cells responsible for its production when in co-culture (Figure
5F). In line with these data, transcriptomic data of HPV positive HNSCC patients from TCGA show that
there is a strong positive correlation between *LIF* and *HBEGF* mRNA expression (Figure S9A).

To understand if cytokines produced upon cancer cell - CAF direct contact were sufficient to increase
macrophage attraction, we isolated primary monocytes from peripheral blood mononuclear cells
(PBMCs) from healthy donors, differentiated them into macrophages and then performed a migration
assay of macrophages using conditioned medium (CM). Importantly, CM derived from the co-culture
of OCAF1 and SCC154 increased the numbers of migrating macrophages, compared with pooled CM
derived from each cell in monoculture (Figure 5G). We then tested whether MEK inhibition was
enough to decrease the number of macrophages attracted. Indeed, when cancer cells and CAFs are
co-cultured in the presence of trametinib, there is a stark decrease in the number of migrating
macrophages attracted by the CM and this was not the case when MEK inhibitor is freshly added to
CM after it is harvested from the cancer cell-CAF co-culture (Figure 5H).

These data suggest that cancer cell – CAF crosstalk is consistently linked to macrophage recruitment,
therefore we asked whether CoCu8 high patients showed higher levels of macrophages. Of note, when
TCGA patients' data are separated according to CoCu8 expression, we observe a strong enrichment
for cells defined by a CD14-related methylation signature (monocytes/macrophages) (Figure 5I). We
found a similar pattern when patients were separated by fibroblast abundance (Figure S9B) and by
RAS activity (Figure S9C). Importantly, we also observed enrichment for monocyte/macrophage

lineages in our second cohort of 84 HPV positive HNSCC patients when separated for CoCu8 expression

levels (Figure 5I).

In summary, we have demonstrated that cancer cell – CAF cross-talk increases expression of different

cytokines that, in turn, recruit higher numbers of macrophages. This loop is established by HB-EGF

expression in cancer cells that induces a cross-talk with CAFs via stromal EGFR dependent by RAS /

MAPK activity. Activation of this pathway in CAFs is needed to increase the expression of LIF by the

stromal cells. MEK inhibitor, trametinib, is sufficient to reduce the macrophage attraction.

Discussion

The presence of CAFs in tumors correlates with worse patient survival and an immune suppressive TME in multiple tumor types [17,30–34], with recent studies attempting to link different CAF subpopulations to prognosis [35]. However, analysis based on the presence or absence of CAFs does not account for variability in the extent of crosstalk between cancer cells and CAFs. The approach we develop here is based on the selection of genes that are commonly upregulated in both cancer cells and CAFs upon direct cell-to-cell contact, thus focusing on the functional relevance of the cancer cell-CAF interaction, rather than just on the presence of CAFs in the tumor. We applied two different strategies to select genes indicative of cancer cell-CAF interactions. The approach to define CoCu30 enriches for genes that are strongly up-regulated, which has been employed previously [7]. To define the CoCu8 signature, we used a new approach based on the selection of a coherent set of genes linked by function. This method produces a list of genes whose changes in expression levels may not be as dramatic, but that have increased functional relevance. Importantly, CoCu8 and CoCu30 stratification gave rise to similar results, strengthening the observation that cancer cell – CAF crosstalk is associated with worse overall survival.

Comparative analysis of CoCu8 and CoCu30 with annotated gene sets (KRAS SIGNALLING UP from Hallmark signatures and RAS84 signature [24]) suggested a mechanism of cross-talk between cancer cells and CAFs based on the activation of MAPK / AP-1 pathway (Figure S10) that we confirmed through experimentation. These data are consistent with previous literature showing that KRAS is associated with higher stromal presence [36] and with higher cancer cell – stromal interaction [37]. Our analyses indicate that HB-EGF is central to the activation of MAPK signalling upon cancer cell – CAF contact. HB-EGF is the only EGF ligand to be consistently upregulated in co-culture across diverse models. Accordingly, EGFR activation is upregulated in co-culture, suggesting the presence of a positive feedback loop, including HB-EGF / EGFR / MAPK / AP-1. In the future, it will be interesting to explore if proteolytic processing of HB-EGF at the interface between cancer cell and CAFs also plays a role in activation of this feedback loop.

Treatment with HB-EGF allowed us to decipher the cell type-dependent consequences of the activation of the pathway. Both cancer cells and CAFs respond to HB-EGF by activating MAPK and inducing changes in AP-1 TF expression. However, cancer cells exhibited increased proliferation and migration, whereas CAFs showed increased proliferation and a pronounced up-regulation of the MAPK-dependent cytokine LIF. These data are consistent with the finding that AP-1 activation leads to diverse molecular and phenotypic consequences depending on the cell type studied [38]. Ultimately,

we link increased MAPK activity to the chemo-attraction of macrophages. These data provide a
molecular explanation specifically for the correlation of CAFs and macrophages in tumors and, more
generally, for links between CAF and a pro-tumorigenic and immune-suppressive milieu [31]. Notably,
LIF can promote macrophage recruitment and induce a more pro-tumorigenic polarization to alter
immune response during anti PD-1 therapy [39].

CoCu8 and CoCu30 signatures are associated with worse overall survival in multiple SCC datasets from
independent cohorts, suggesting that the mechanism we have identified is broadly applicable. Of note,
the most striking results were observed in HPV positive HNSCC. Although HPV positive HNSCC patients
typically have a better prognosis than HPV negative HNSCC patients, about 25% of these patients still
have poor overall survival [40,41].

We hypothesize that CoCu8 / CoCu30 are particularly effective in HPV positive HNSCC compared to
HPV negative because HPV negative has more frequent mutation and amplification in EGF / RAS /
MAPK pathway [42], hence its worse prognosis and the smaller contribution of CAFs to poor prognosis
(CoCu8 HR is 1.28 in HPV negative HNSCC and 1.85 LUSC compared to 3.19 in HPV positive HNSCC and
2.79 in CESC). In HPV positive disease, the same signalling pathway is driven as the result of paracrine
interactions with CAFs, which occur to varying degrees in different tumors. These findings have clinical
implications for patient stratification and treatment. Our data suggest that the CoCu8 / CoCu30
signatures and CAF abundance could help stratify those patients with worse prognosis within the HPV
positive SCC. This improved patient stratification would be especially relevant in the context of the
recent unsatisfactory efforts to de-escalate and de-intensify treatment for patients with HPV positive
tumors [43] and could help reduce toxicity without compromising outcomes. Moreover, our results
suggest that this subset of patients could benefit from a targeted approach, for example re-purposing
the use of MEK or EGFR inhibitors. Indeed, trametinib – MEK inhibitor – is already used in the
treatment of melanoma [44] and non-small-cell lung cancer [45] and it has been tested in phase I / II oral
cavity SCC patients, showing some reduction in RAS / MAPK activity as neoadjuvant treatment [46]. Our
data support the hypothesis that trametinib or EGFR inhibitors may be beneficial for HPV positive
HNSCC patients with high tumor / stroma content.

In conclusion, our results demonstrate a new approach to detect biologically meaningful stromal
signatures. We show that signatures based on signalling in the TME have the potential to both improve
patient stratification and to identify new mechanisms of cross-talk between cancer cells and CAFs.

Methods

Cell lines and reagents

OCAF1 and OCAF2 human fibroblasts were isolated from patient tissues of oral cancer and immortalized with lentiviral HTERT as described in [47]. These patient samples were collected under the ethical approval REC reference 06/Q0403/125).

CRUK0764 were derived from patients with lung adenocarcinoma. These fibroblasts were established from the tumor tissue. The primary CRUK0764 was immortalised by the following infection with retroviruses expressing human telomerase reverse transcriptase.

PC9 was obtained from the Crick Institute Central Cell Services facility. PC9 were stably transfected with Lipofectamine 2000 Reagent (Thermo Fisher Scientific) according to the manufacturer's instructions. Briefly, PC9 cell line was seeded at 50-70 % confluence in a six-well plate and transfected 2 μg of Piggybac transposase (pPBase-piggyBac) and 2 μg of mEGFP (pPBbsr2-mEGFP) plasmid DNAs. After 24h of incubation, the medium with Lipofectamine/plasmid DNA mix was replaced with a fresh medium. Cells were selected using 2 μg ml -1 blasticidin.

SCC154 (UPCI-SCC154) and SCC47 (UM-SCC47) were purchased from ATCC.

OCAF1, OCAF2, SCC154, SCC47 and CRUK0764 cells were cultured in DMEM (ThermoFisher, #41966052) containing 10% fetal bovine serum (Gibco, #10270-106), 1% penicillin/streptomycin (Invitrogen, #15140122), 1% insulin–transferrin–selenium (Invitrogen, #41400045) and kept at 37°C and 5% CO_2.

PC9 were cultured in RPMI-1640 (Thermo Fisher Scientific, Rockford, IL) supplemented with 10% fetal bovine serum (Gibco, #10270-106), 1% penicillin/streptomycin (Invitrogen, #15140122) and kept at 37 °C in 5% CO_2.

Cells were not allowed to reach more than 90% confluency for routine cell culture cultivation. Cell lines that are not commercially obtainable are available from the authors upon reasonable request.

Routine screening for *Mycoplasma* testing was performed for all cell lines with negative results. STR profiles of human non-commercially available cell lines are included in Supplementary Table 1.

Cell cultures conditions and treatments

Co-cultures and mono-cultures were performed with a ratio of 1:2, typically plating 5.5 x 10^5 CAFs and 2.75 x 10^5 cancer cells for a single well of a 6 well plate for the specified time point. When co-cultures were compared to pooled mono-cultures, for the mono-culture condition, same number of cells was plated but in two separated wells and then lysed together (pooled condition).

When cancer cells and CAFs mono-culture were compared among themselves, 1 x 10^6 OCAF1 and 1

x 10^6 SCC154 cells were plated in a 10cm dish.

For PC9 and CRUK0764 cells monocultures and co-cultures used for RNAseq, following 24 h co-

cultures, the culture media was replaced with fresh medium with DMSO, then harvested after an

additional 24 h. PC9 – CRUK0764 co-cultures were performed in a mixture of RPMI-1640 and DMEM

(1:1) containing 1% fetal bovine serum (Gibco, #10270-106).

For macrophage cultivation, please see "Macrophage migration assay" section.

For cell culture treatments: drugs / factors were added when cells were plated and then added fresh

after 24h. Drugs / factors used: trametinib (Selleckchem, #GSK1120212), afatinib (Selleckchem,

#BIBW2992), human recombinant HB-EGF (Peprotech, #100-47).

For trametinib treatment to collect conditioned medium (CM), in order to control the effect of the

drug presence regardless of its effect on secreted factors, we added fresh trametinib treatment to

DMSO co-culture CM at the same concentration used for the cell co-culture treatment.

All concentrations used are specified in the figures.

RNA interference was performed with Lipofectamine RNAimax reagent from Invitrogen, according to

the manufacturer's instructions. For transient knock down of HB-EGF, cells were subjected to reverse

transfection with 20 nM RNAi oligos plus forward transfection the day after, then analyzed 4 days after

reverse transfection. The following RNAi oligo (Dharmacon) was used: siHB-EGF A (Cat # D-019624-

02), siHB-EGF B (Cat # D-019624-03), as control the following non-targeting siRNA oligo (All Stars

Negative, Quiagen, Cat # 1027281).

Fluorescence-activated cell sorting

For PC9 – CRUK0764 RNAseq experiment, CRUK0764 were labelled with CellVue® Red Mini Kit for

Membrane Labeling (Polysciences, 25567-1) according to the manufacturer's instructions. Briefly, 1 ×

10^7 cells of CRUK0764 were resuspended in the Diluent C and mixed with CellVue® Red working dye

solutions (final concentration: 5 × 10^6 cells/mL, 2 × 10^6 M dye) and then incubated for 5 min at RT.

Cells were washed twice with DMEM, 10% FBS medium to ensure removal of unbound fluorescence

dye.

For fluorescence-activated cell sorting, cells were sorted using a flow cytometer–cell sorter BD

FACSAria™ II. PC9-GFP and CRUK0764 -CellVue Red were sorted by FACS 48 h after seeding them in

monoculture or direct co-culture. The cells were then trypsinised and resuspended in 3% FBS in PBS,

1 mM EDTA in preparation for sorting. Cells were separated into two populations: PC9-GFP and

CRUK0764 with CellVue Red using a 488 nm laser with collection filter 530 nm/30 nm for GFP and 561

nm laser with collection filter 582 nm/20 nm for CellVue Red. Gates were designed on the basis of

negative and single-colour controls. All cell populations were tested for purity, and data were analysed

using FlowJo software.

RNA sequencing analysis for co-cultures

PC9 and CRUK0764 cells were immediately centrifuged at 300 × g for 4 min to remove supernatant

and add 350 µl RLT buffer (Qiagen, 79216) containing 1% β-mercaptoethanol (Sigma, M6250) and total

RNA was extracted using the RNAeasy Mini kit (Qiagen, 74104; n = 3 independent experiments). Prior

to library construction, the quality of total RNA was assessed by Bioanalyzer 2100 (Agilent

Technologies Inc).

For RNAseq analysis: biological replicates libraries were prepared using the polyA KAPA mRNA

HyperPrep Kit and sequenced on the Illumina HiSeq 4000 platform generating ~28 million 75bp single

end reads per sample. FASTQ_files were quality trimmed and adaptor removed using Trimmomatic

(version 0.36) [48]. The RSEM package (version 1.3.30) [49] in conjunction with the STAR alignment

software (version 2.5.2a) [50] was used for the mapping and subsequent gene level counting of the

mapped reads with respect to the Ensembl human GRCh38 (release 89) transcriptome. Normalisation

of raw count data was performed with the DESeq2 package (version 1. 18.1) [51]. All the analysis was

done (version 1. 18.1) [51] within the R programming environment (version 3. 4. 3).

To check the purity of the samples, we analyzed the resulting transcriptomic data for the expression

of 'lineage markers'. CDH1, EPCAM, CD24, and KRT genes were used as markers of carcinoma cells and

for fibroblasts we used COL1A1, COL1A2, DCN, CD248, and PDGFR genes. This revealed high sample

purity for all transcriptomic data, except in the PC9 – CRUK0764 experiment that had variable purity

between samples. Therefore, we estimated the impurity in each sample based on the expression of

the lineage marker genes and calculated the expected level of transcript if the two mono-cultures

(cancer cells alone and CAFs alone) were mixed in proportion with the impurity estimate. The

observed transcript in the co-culture condition was then normalized to account for the effect of

contamination.

EdU proliferation assay

476　The Click-iT Plus EdU Imaging Kit (Invitrogen #c10640) was used to perform the assay. Briefly, 48h after

477　mono-cultures of OCAF1 and SCC154 were seeding, a solution with Edu 20µM was prepared and then

478　diluted 1:1 with cell media to add EdU 10µM final concentration. After 90 minutes incubation, cells

479　were washes twice in PBS, then fixed for 15 minutes with paraformaldehyde 3.7% and then washed

480　twice in BSA 3%. Following this step, cells were incubated for 20 minutes with 0.5% Triton X-100 in

PBS. After two BSA 3% washes, the Click-iT reaction buffer was added for 30 minutes, followed by one wash in BSA 3% and one wash in PBS. Subsequently, nuclei were stained with Hoechst 33342 at 5 µg/mL in PBS incubation for 30 minutes, followed by two PBS washes.

Samples were imaged with Zeiss 980 microscope. Mean fluorescent intensity was analysed using ImageJ.

Immunofluorescence assay

The samples used to perform EdU proliferation assay have been then stained for E-Cadherin. Briefly, samples were washes twice in PBS, followed by incubation for 30 minutes in BSA 3%. Then samples were incubated over night at 4°C. After two washes in BSA 3% of 5 minutes each, samples were incubated with secondary antibody Alexa Fluor 555 in BSA 3% for 45 min. Following this step, samples were washed with PBS twice. Subsequently, nuclei were stained with Hoechst 33342 at 5 µg/mL in PBS incubation for 30 minutes, followed by two PBS washes. Samples were imaged with Zeiss 980 microscope.

Peripheral Blood Mononuclear Cell extraction and monocytes selection

Donations of healthy blood donors were received from the Francis Crick Institute. PBMCs were isolated from whole blood using Lymphoprep (Stemcell Technologies #7811) with SepMate^{TM} density centrifugation tubes in line with manufacturer's instructions (Stemcell Technologies #85450). Freshly isolated PBMCs were then counted before isolation of monocytes (Miltenyi Biotec #130-096-537). Monocytes were then counted for plated in normal plastic dishes.

Cytokine array

Cytokine array used is "Proteome Profiler Human XL Cytokine Array Kit" (R&D Systems, # ARY022B) following manufacturer's instruction. Briefly, conditioned media was isolated and filtered through a 0.4µm low protein binding PVDF Miltex syringe-driven filter (Millipore #SLHV033RS) to remove cellular debris. Media was then concentrated to 4X using Amicon® Ultra-15 and used for subsequent incubation with array.

Macrophage migration assay

Monocytes were plated into 12-well plates ($1x10^5$ cells / well) in RPMI 1640 media (ThermoFisher #12633-012) containing 10% FBS, 1% streptomycin/penicillin and 50ng/mL of M-CSF (Peprotech #300-

25) and kept at 37°C and 5% CO2 for 5 days to allow macrophage differentiation. During incubation

period, OCAF1 – SCC154 mono- and co-cultures were set-up for 48h. Conditioned media was isolated

and filtered through a 0.4μm low protein binding PVDF Miltex syringe-driven filter (Millipore

#SLHV033RS) to remove cellular debris. Media was then concentrated to 4X using Amicon® Ultra-15

centrifugal filter units (Millipore #UFC901024) and frozen into aliquots until needed. Conditioned

media was added to 24-well plates, 8μm hanging cell culture inserts (Millipore #MCEP12H48) were

placed on top of each well. The now differentiated macrophages were seeded inside the hanging cell

culture insert and left to settle for 10 minutes before topping up media. Plates were left in the

incubator for 5 hours to allow macrophages to migrate through membrane pores. After this time, the

inserts were removed and the macrophages sat on top of the membrane were wiped off with a cotton

bud, leaving behind the migrated macrophages at the bottom. Inserts were stained with 0.05% crystal

violet for 30 minutes before washing and then imaged. Inserts were imaged using Zeiss Observer Z1

mounted with a QImaging Colour camera. Quantification of crystal violet staining was carried out using

ImageJ through 'Cell Counter' function.

Gene Set Enrichment Analysis

529   Gene set enrichment analysis was performed with GSEA software v4.1.0. The dataset used to perform

530   the comparative analysis are: RAS84 derived from [24], CoCu8 derived from our own analysis, Hallmarks

531   (h.all.v7.5.symbols.gmt) for all the other analysis. All the parameters have been used as defaults

532   except: permutation type (gene set) and metric for ranking genes (t test). Gene signatures with a false

533   discovery rate < 0.05 were considered as statistically significant.

Transcriptomic data

The transcriptomic data used are: microarray data of A431 / VCAF2b under conditions of mono-

cultures, co-cultures in direct contact and indirect contact are available at the Gene Expression

Omnibus under record GSE121058. The microarray data of MAF2 under conditions of mono-culture

and co-culture in direct contact is available at the Gene Expression Omnibus under record GSE63160.

The microarray data of 5555 under conditions of mono-culture and co-culture in direct contact will be

submitted at the Gene Expression Omnibus.

The RNAseq data of PC9 / CRUK0764 under conditions of mono-cultures and co-cultures in direct

contact will be submitted to the European Genome-Phenome Archive before publication.

The microarray data used for HUVEC – 1205Lu analysis is available at Gene Expression Omnibus under

record GSE8699.

The microarray data used for breast cancer cell lines co-culture with fibroblasts analysis is available at

Gene Expression Omnibus under record GSE41678.

Co-culture gene signature generation

CoCu8 gene signature generation: the A431/VCAF2b, 5555/MAF2, PC9/CRUK0764 co-cultures vs

mono-cultures transcriptional datasets have been analysed with GSEA (see Gene Set Enrichment

Analysis method) to obtain a list of enriched pathways in co-culture with FDR < 0.05 for each condition.

For each cell type, all the genes statistically upregulated have been pulled together. From this list,

genes that were present in 20% or more of the enriched pathways have been selected. The results

obtained for each sample have been merged according to the cell type: the three cancer cells in co-

culture have been pulled together, same for the three CAFs. To select the final list, only genes present

in the three different cancer cells or in the three different CAFs have been selected to generate CoCu8.

CoCu30 gene signature generation: the genes with a fold change upregulation of 1.5 or higher have

been selected for each cell type upon co-culture. The results of the three cancer cells in co-culture

have been pulled together, same for the three CAFs in co-culture. To select the final list, only genes

present in the three different cancer cells or in the three different CAFs have been selected to generate

CoCu30.

TCGA analysis

Clinical data, RSEM (RNA-Seq by Expectation-Maximization) normalised expression data (Illumina

RNASEQ platform) and Methylation data (Illumina Human Methylation 450 platform) for HNSCC, CESC

and LUSC cohorts was downloaded from the Firebrowse website hosed by Broad Institute of MIT and

Harvard. [http://firebrowse.org/]. Data downloads were all version 2016012800.0.0.

De-convolution strategies:

MethylCIBERSORT: signature matrix and mixture files were obtained using MethylCIBERSORT R

package, hosted on Zenodo. The detailed origin of the signatures and the procedure to create the

deconvolution strategy is explained in Chakravarthy et al. [10].

Absolute-CIBERSORT: To calculate the immune infiltrate per sample, the library 'CIBERSORT' (version

1.04 [52]) was run within R version 3.4.3 on the RSEM normalised data and the LM22 signature using the

parameters absolute=TRUE and abs_method ="no.sumto1".

RNA extraction and RT-qPCR

Cells were collected and lysed with RLT buffer and total RNA was extracted using the RNeasy Mini kit (Qiagen, #74104), according to the manufacturer's protocol.

The cDNA was prepared using M-MLV reverse transcriptase (Promega, #M3682), and quantitative PCR was performed using PowerUp™ SYBR™ Green Master Mix (ThermoFisher, #A25778), using the QuantStudio 3 and 7 Real-Time PCR systems (Applied Biosystems).

Custom primers were acquired from Sigma; sequences are available in Supplementary Table 2. RNA levels were normalized using three house-keeping genes using the ΔΔC method and reported as relative fold change compared with Ctr/not treated cells/mono-culture.

scRNAseq analysis

scRNAseq data from Choi et al. [29] was downloaded from GEO (GSE181919) and analysed using Seurat package (version 4) [53].

Protein extraction, quantification and Western Blot analysis

Cells were lysed in RIPA buffer (50 mM TrisHCl, 150 mM NaCl, 1 mM EDTA, 1% Triton X-100, 1% sodium deoxycholate, 0.1% SDS), supplemented with a protease and phosphatase inhibitors (PhosSTOP tablet Roche #04906837001, cOmplete EDTA-free Roche #11873580001, 50 mM NaF). Lysis was performed directly in the cell culture plates using a cell scraper, lysates were kept for 10min on ice and then clarified by centrifugation at 16,000 g for 30 min at 4 °C.

Total protein was quantified using the bicinchoninic acid method in accordance with manufacturer's instructions (ThermoFisher, 23225). Following protein quantification, 20 µg of sample was loaded on a 4–15% gradient Mini-PROTEAN TGX Gels (Biorad, #4561084) and transferred to a Trans-Blot Turbo Mini 0.2 µm PVDF membrane (Biorad, 1704156) for blotting. The membrane was blocked for 1 h in 5% BSA or 5% milk in TBST and then incubated overnight at 4°C or 1 h at room temperature with antibodies. The membrane was then washed before adding the horseradish peroxidase-conjugated secondary antibody (ThermoFisher), and incubating for 1 h at room temperature. The membrane was washed again before developing with Luminata Classico Western HRP substrate (Millipore, #WBLUR0100) Luminata Classico Western HRP substrate (Millipore, # WBLUF0100) and imaging. Antibody information are listed in Supplementary Table 3. All original blots are provided as source data.

Software and visualization

Graphs were generated with Prism software (Graphpad Software v9.4.0) and R (version 4.2.1) using package 'ggplot' except for correlation plot in Figure S9A that was generated with cBioportal [54]. scRNAseq data were analysed with Seurat package (version 4).

Statistics

Statistical analysis was performed using Prism software (Graphpad Software v9.4.0), Excel software (Microsoft Corporation v16.0) and R (version 4.2.1).

All Student's t tests have been performed with two tailed strategy.

P-value information: * is p-value<0.05; ** is p-value<0.01, *** is p-value<0.001, **** is p-value<0.0001.

For GSEA, we used FDR with a threshold below 0.05 to definite the significance.

Kaplan-Meier, Log-Rank and Cox regression on survival data was calculated using the R package 'survminer' using univariable analysis. Correlations were calculated using the Spearman method in R and the package 'corplot' was used to generate the graphs.

UK_HPV positive cohort

FFPE tumor samples from 2 studies formed this cohort:

- INOVATE (MR/R015589/1 ISRCTN32335415), a prospective sample collection study in patients with T1-T2/N1-3 or T3-T4/N0-3 oropharyngeal cancer (AJCC TNM classification 7.0) receiving treatment with radical radiotherapy with or without additional platin-based chemotherapy.
- INSIGHT-2 (C7224/A23275 NCT04242459), a prospective study of optimising radiation therapy in head and neck cancers using functional image-guided radiotherapy and novel biomarkers.

INOVATE was approved by the London -Bloomsbury Research Ethics Committee (19/LO/1558) and INSIGHT-2 was approved by London – Queen Square Research Ethics Committee (19/LO/0638). Written informed consent was obtained from all participants prior to any study procedure.

UK_HPV positive cohort RNAseq and data analysis

Baseline diagnostic biopsies embedded in paraffin blocks were obtained from the above-mentioned cohort. Relevant tumor sections were selected and RNA was extracted from 3-4 slides using the Qiagen AllPrep® DNA/RNA FFPE kit (#80234). Ribosomal RNA was depleted using QIAGEN FastSelect rRNA H/M/R kit (#334375). RNA sequencing libraries were prepared using the NEBNext Ultra II RNA

Library Prep Kit (#E770) for Illumina following manufacturer's instructions. The sequencing libraries

were multiplexed and loaded on the flow cell on the Illumina NovaSeq 6000 instrument according to

manufacturer's instructions. The samples were sequenced using a 2x150 Pair-End (PE) configuration

v1.5 for an estimated output of ~50M paired-end reads per sample. Image analysis and base calling

were conducted by the NovaSeq Control Software v1.7 on the NovaSeq instrument. Raw sequence

data (.bcl files) generated from Illumina NovaSeq was converted into fastq files and de-multiplexed

using Illumina bcl2fastq program version 2.20. One mismatch was allowed for index sequence

identification. Sample adequacy was confirmed using FASTQC, low quality bases and reads were

trimmed using Trimmomatic, we run Hisat2-Stringtie for alignment.

RNAseq was performed on 103 patient samples, of which 7 were from the INSIGHT2 and 96 from

INOVATE. The data from RNAseq was analysed to identify samples with presence of HPV (by aligning

the unmapped sequences to the whole HPV16 genome sequence obtained from GEO using HISAT2

and StingTie) and these samples were classed as HPV positive. 84 samples (77 INOVATE and 7

INSIGHT2) were classified at HPV positive and RNAseq data from these was used for analysis in this

study.

TRACERx cohort

Tumor samples used in this study were collected from LUSC patients enrolled as a part of TRACERx

study (accession code: NCT01888601) which is sponsored by University College London (UCL/12/0279)

and has been approved by an independent research ethics committee (13/LO/1546). Multiple regions

were sampled per tumor and processed as described by Frankell et al. [20] yielding whole-RNA

sequencing data for 295 regions from 117 LUSC patients. Expression count and transcript per million

(TPM) were quantified by the RSEM package [49]. Genes with expression level of at least 1 TPM in at

least 20% of the samples were included. A variance stabilising transformation (VST) was then applied

to filtered count using the DESeq2 package [51].

Supplementary tables

669   Supplementary Tables 1, 2 and 3 are provided with this article.

Data accessibility

672   The transcriptomic data for UK_HPV positive cohort is part of ongoing clinical trials, therefore the data

673   cannot be deposited in a public repository until the trial is finalised. Data can be shared upon

674 reasonable request following corresponding Ethical Research Committee approval following the ICR-
675 Clinical Trials and Statistics Unit policy.
676

Authors contributions

G.G. and A.R. conceptualised and designed the research. GG performed the experimental analysis and
identified the cancer cell – CAF crosstalk gene signature strategy. A.R. performed the clinical data
analysis on TCGA and UK_HPV positive cohorts and performed the deconvolution strategies. G.G.,
A.R., S.H., E.H., Y.N., P.N. and P.C. acquired and analysed the data, except for TRACERx data that were
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and reagents. G.G., A.R. and E.S. wrote the manuscript. All authors reviewed and edited the
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685

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C.S. holds patents relating to assay technology to detect tumor recurrence (PCT/GB2017/053289); to

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Figure legends

Figure 1: Cancer cell / CAFs co-culture gene signature CoCu8 is associated with worse overall survival in multiple squamous cell carcinoma datasets. **A)** Strategy used to obtain CoCu8 gene signature. **B)** Venn diagrams of the genes upregulated in the different datasets (top) and summary table of the genes upregulated in all the datasets (bottom) for cancer cells (right) and CAFs (left). **C)** Kaplan-Meier overall survival analysis of HNSCC (right), CESC (centre), LUSC (left) TCGA datasets stratified for CoCu8 first vs last quartile. Numbers at risk shown in tables below graphs. HNSCC HR=1.95 (95% Confidence Interval (CI) 1.29-2.93), p-value=0.0011. CESC HR=2.79 (95%CI 1.40-5.56), p-value=0.0024, LUSC HR=1.85 (95%CI 1.28-2.70), p-value=0.001. HR and CI were calculated using Cox regression. p-value was calculated using logRank test. **D)** Kaplan-Meier overall survival analysis of HNSCC HPV positive (right) and negative (left) TCGA datasets stratified for CoCu8 first vs last quartile. Numbers at risk shown in tables below graphs. HNSCC HPV positive HR=3.19 (95%CI 1.24-8.18), p-value=0.011. HPV negative HR=1.28 (95%CI 0.84-1.93), p-value=0.25. HR and CI were calculated using Cox regression. p-value was calculated using logRank test. **E)** Kaplan-Meier overall survival analysis of LUSC TRACERx dataset. Individual tumors stratified as high-, discordant or low-risk according to expression profile of CoCu8 signature across multiple regions, as previously described and stratified according to Biswas et al. [55]. Below are shown the numbers at risk in years. HR=1.67 (95%CI 1.13-2.47), p-value=0.0247. HR and CI calculated using Cox regression and are referred to CoCu8 low vs CoCu8 high. p-value was calculated using logRank test.

Figure 2: Cancer cell / CAFs co-culture upregulates AP-1 TF genes. **A)** Bubble plot of the Hallmarks pathways upregulated by cancer cells / CAFs culture. Normalised enrichment score (NES) is depicted as bubble size; False discovery rate (FDR) is depicted as colour intensity. **B)** Heatmap of expression of the genes commonly upregulated upon co-culture in all the tested conditions from the TNFA_SIGNALLING_VIA_NKFB pathway. **C)** qPCR analysis of *JUNB*, *FOS* and *FOSB* genes in OCAF1/OCAF2 with SCC154/SCC47 pooled mono-cultures and co-culture after 24h. mRNA expression is reported as mean ± standard deviation (SD) fold change difference over pooled mono-culture. Genes have been normalized over the average of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n = 4 independent experiments. Paired t-test.

Figure 3: RAS activity is upregulated in cancer cell and CAFs upon co-culture. **A)** Gene set enrichment analysis (GSEA) plot of RAS84 gene signature in mono-culture and co-culture. NES and FDR are specified below each plot. **B)** Kaplan-Meier overall survival analysis of HNSCC HPV positive TCGA

dataset stratified for Ras84 activity according to [24]. Below are shown the numbers at risk in years.

RAS84_0 vs RAS84_1 HR=1.01 (95%CI 0.387-2.62) p-value=0.98. RAS84_0 vs RAS84_max HR=5.85

(95%CI 2.46-13.9) p-value<0.001. HR and CI calculated using Cox regression. **C)** Correlation plot of

RAS84 expression level and CoCu8 expression level in HNSCC HPV positive TCGA dataset. R is

Spearman correlation coefficient. n=97. **D)** Box plot analysis of fibroblast abundance via Methyl

CIBERSORT deconvolution strategy in HNSCC HPV positive TCGA dataset according to RAS84 activity.

Independent t-test. **E)** Western blot analysis of OCAF1 – SCC154 pooled mono-culture vs co-culture

for 48h showing the indicated antibodies. Vinculin is used as loading control. **F)** Western blot analysis

of OCAF1 – SCC154 co-cultures for 48h at the indicated conditions showing the indicated antibodies.

Vinculin is used as loading control. **G)** qPCR analysis of *JUNB*, *FOS* and *FOSB* genes in OCAF1 - SCC154

co-cultures for the indicated treatments after 48h. mRNA expression is reported as mean ± standard

deviation (SD) fold change difference over co-culture DMSO. Genes have been normalized over the

average of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n = 4 independent experiments. Paired t-

test.

Figure 4: HB-EGF / EGFR axis activates AP-1 TF genes in cancer cells and CAFs upon co-culture via RAS

pathway. **A)** Pattern of expression of the 7 EGFR ligands in all the tested transcriptomic datasets. **B)**

Correlation plot of *HB-EGF* expression level and CoCu8 expression level in HNSCC HPV positive TCGA

dataset (left, n=97) and UK_HPV positive cohort (right, n=84). R is Spearman correlation coefficient.

C) Western blot analysis of OCAF1 – SCC154 pooled mono-culture vs co-culture for 48h showing the

indicated antibodies. Vinculin is used as loading control. **D)** Western blot analysis of OCAF1 – SCC154

co-cultures for 48h at the indicated conditions showing the indicated antibodies. Vinculin is used as

loading control. **E)** qPCR analysis of *JUNB*, *FOS* and *FOSB* genes in OCAF1 - SCC154 co-cultures for the

indicated treatments after 48h. mRNA expression is reported as mean ± standard deviation (SD) fold

change difference over co-culture DMSO. Genes have been normalized over the average of *GAPDH*,

ACTB and *RPLP0* housekeeping genes. n = 4 independent experiments. The DMSO treated sample is

the same used for Figure 3G. Paired t-test. **F)** Western blot analysis of OCAF1 and SCC154 mono-

cultures after 48h showing the indicated antibodies. Vinculin is used as loading control. **G)**

Proliferation assay of OCAF1 and SCC154 mono-cultures for the indicated treatments after 48h and

stained with EdU and Hoechst 33342. On the left, a representative image is shown with bar graph. On

the right, dot plot of mean fluorescent intensity (MFI) of EdU over Hoechst 33342 with

mean ± standard deviation (SD) highlighted. Each dot is a field of view. n = 3 independent experiments.

t-test. **H)** qPCR analysis of *JUNB*, *FOS* and *FOSB* genes in OCAF1 - SCC154 co-cultures after 48h with

SCC154 pre-treated with the indicated conditions. mRNA expression is reported as mean ± standard

deviation (SD) fold change difference over siC- condition. Genes has been normalized over the average

of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n ≥ 4 independent experiments. Paired t-test.

1003

Figure 5: Cancer cells – CAFs co-culture induces production of specific cytokines to attract

macrophages. **A)** On the right is shown scRNAseq UMAP analysis from Choi et al. [29]. On the left is

shown violin plot of EGFR and HBEGF mRNA expression levels in the indicated clusters. **B)** Cytokine

array of conditioned medium from pooled mono-culture and co-culture of OCAF1 and SCC154.

Highlighted relevant cytokines. n = 2 independent experiments. **C)** qPCR analysis of *LIF* gene in OCAF1

– SCC154 pooled mono-culture vs co-culture after 24h. mRNA expression is reported as

mean ± standard deviation (SD) fold change difference over pooled mono-culture. Genes have been

normalized over the average of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n = 4 independent

experiments. Paired t-test. **D)** qPCR analysis of *LIF* gene in OCAF1 - SCC154 co-cultures for the indicated

treatments after 48h. mRNA expression is reported as mean ± standard deviation (SD) fold change

difference over co-culture DMSO. Genes have been normalized over the average of *GAPDH*, *ACTB* and

RPLP0 housekeeping genes. n = 4 independent experiments. Paired t-test. **E)** qPCR analysis of *LIF* gene

in OCAF1 - SCC154 co-cultures after 48h with SCC154 pre-treated with the indicated conditions. mRNA

expression is reported as mean ± standard deviation (SD) fold change difference over siC- condition.

Gene has been normalized over the average of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n ≥ 4

independent experiments. Paired t-test. **F)** qPCR analysis of *LIF* gene in OCAF1 and SCC154 mono-

cultures for the indicated treatments after 48h. mRNA expression is reported as mean ± standard

deviation (SD) fold change difference over non-treated (NT) condition for each cell type. Genes have

been normalized over the average of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n = 5 independent

experiments. Paired t-test. **G)** Migration assay of macrophages plated in transwells with conditioned

medium (CM) from OCAF1-SCC154 pooled mono-culture or co-culture. CM have been obtained after

48h culture. On the left, a representative field of view is shown with bar graph. On the right, dot plot

of number of cells per field of view as mean ± standard deviation (SD). Each dot is a field of view

normalized by the average of the pooled mono-culture CM sample. n = 4 different donors. t-test. **H)**

Migration assay of macrophages plated in transwells with CM from OCAF1-SCC154 co-culture with the

indicated treatments. CM have been obtained after 48h culture and for the fresh trametinib sample,

the drug has been added after the CM was collected. Dot plot of number of cells per field of view as

mean ± standard deviation (SD) is shown. Each dot is a field of view normalized by the average of the

co-culture DMSO CM sample. n = 3 different donors. t-test. **I)** At the top, box plot analysis of CD14+

monocytic/macrophage lineage immune cell absolute score via Methyl CIBERSORT deconvolution

strategy in HNSCC HPV positive TCGA dataset separate by first and last quartile of CoCu8 expression.

Independent t-test, Bonferroni correction for multiple comparisons. At the bottom, box plot analysis

of monocyte and macrophage immune cell score using Absolute CIBERSORT deconvolution strategy in

HNSCC HPV positive UK_HPV positive dataset separate by first and last quartile of CoCu8 expression.

Independent t-test.

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Figure S1: Cancer cell / CAFs co-culture gene signature CoCu30 is associated with worse overall survival

in multiple squamous cell carcinoma datasets. **A)** Strategy used to obtain CoCu30 gene signature. **B)**

Venn diagram of the genes upregulated in the different datasets (top) and a table to summarize the

genes constantly upregulated in the datasets (bottom) for cancer cells (left) and CAFs (right). **C)**

Kaplan-Meier overall survival analysis of HNSCC (left), CESC (centre), LUSC (right) TCGA datasets

stratified for CoCu30 first vs last quartile. Below each analysis are shown the corresponding numbers

at risk, time in years. HNSCC HR=1.57 (95%CI 1.06-2.35), p-value=0.024. CESC HR=2.08 (95%CI 1.09-

4.01), p-value=0.024. LUSC HR=1.78 (95%CI 1.23-2.59), p-value=0.0019. HR and CI were calculated

using Cox regression. p-value was calculated using logRank test. **D)** Kaplan-Meier overall survival

analysis of HNSCC HPV positive (left) and HNSCC HPV positive (right) TCGA datasets stratified for

CoCu30 first vs last quartile. Below each analysis are shown the corresponding numbers at risk, time

in years. HPV positive HR=5.47 (95%CI 1.76-17.0), p-value=0.0011. HPV negative HR=1.00 (95%CI 0.67-

1.51), p-value=0.98. HR and CI were calculated using Cox regression. p-value was calculated using

logRank test. **E)** Kaplan-Meier overall survival analysis of LUSC TRACERx dataset. Individual tumors

stratified as high-, discordant or low-risk according to expression profile of CoCu30 signature across

multiple regions, as previously described and stratified according to Biswas et al. [55]. Below are shown

the numbers at risk in years. HR=1.98 (95% CI 1.09-3.6), p-value=0.0152. HR and CI calculated using

Cox regression and are referred to CoCu30 low vs CoCu30 high. p-value was calculated using logRank

test.

1059

Figure S2: Validation of CoCu8 and CoCu30 gene signatures on different co-culture combinations. **A)**

Table with Normalized Enrichment Score (NES) values of different combinations of cancer cells and

CAFs breast cancer cell lines from Rajaram et al. [18]. Negative values represent enrichment towards co-

culture condition. Colour legend is shown. **B)** Table with NES values of CoCu8 and CoCu30 gene

signatures for cancer cells and endothelial cells mono-culture vs co-culture available from Stine et al.

[19]. Negative values represent enrichment towards co-culture condition. Colour legend is shown.

1066

1067 **Figure S3:** Multivariate analysis of CoCu8 overall survival. Forest plot showing Hazard Ratios, 95%

1068 confidence interval and p value calculated using multivariate Cox regression from patients with HNSCC

1069   (HPV positive and negative), LUSC and CESC from the TCGA cohort. Variables include: age (continuous,

1070   years), sex (male vs female, except for CESC as all patients were female), clinical stage (categorical)

1071   and the CoCu8 signature (continuous variable).

1072

1073   **Figure S4:** Multivariate analysis of CoCu30 overall survival. Forest plot showing Hazard Ratios, 95%

1074   confidence interval and p value calculated using multivariate Cox regression from patients with HNSCC

1075   (HPV positive and negative), LUSC and CESC from the TCGA cohort. Variables include: age (continuous,

1076   years), sex (male vs female, except for CESC as all patients were female), clinical stage (categorical)

1077   and the CoCu30 signature (continuous variable).

1078

Figure S5: Fibroblast abundance correlates with CoCu8 / CoCu30 gene signature on different

squamous cell carcinoma datasets. **A)** Box plot analysis of CoCu8 expression in HNSCC, CESC and LUSC

TCGA separated by first and last quartile of fibroblast abundance via Methyl CIBERSORT deconvolution

strategy. Independent t-test. **B)** Box plot analysis of CoCu8 expression in HPV positive and negative

TCGA separated by first and last quartile of fibroblast abundance via Methyl CIBERSORT deconvolution

strategy. Independent t-test. **C)** Box plot analysis of CoCu30 expression in HNSCC, CESC and LUSC TCGA

separated by first and last quartile of fibroblast abundance via Methyl CIBERSORT deconvolution

strategy. Independent t-test. **D)** Box plot analysis of CoCu30 expression in HPV positive and negative

TCGA separated by first and last quartile of fibroblast abundance via Methyl CIBERSORT deconvolution

strategy. Independent t-test. **E-J)** Correlation plot of different fibroblast subpopulations derived from

Galbo et al. [7] with CoCu8 gene signature in LUSC TRACERx **(E)**, HPV positive HNSCC TCGA **(F)**, UK_HPV

positive HNSCC **(G)** patients and with CoCu30 gene signature in LUSC TRACERx **(H)**, HPV positive HNSCC

TCGA **(I)**, UK_HPV positive HNSCC **(J)** patients. The number inside the square represents the R,

Spearman correlation coefficient. The colour legend is shown at the bottom. All correlations are

significant p-value<0.05.

1094

Figure S6: Fibroblast abundance meta-analysis on different squamous cell carcinoma datasets. **A)**

Kaplan-Meier overall survival analysis of HNSCC (left), CESC (centre), LUSC (right) TCGA datasets

stratified for fibroblast abundance first vs last quartile. Below each analysis are shown the

corresponding numbers at risk, time in years. HNSCC HR=1.54 (95%CI 1.05-2.25), p-value=0.025. CESC

HR=1.56 (95%CI 0.77-3.17), p-value=0.21. LUSC HR=1.12 (95%CI 0.71-1.77), p-value=0.62. HR and CI

were calculated using Cox regression. p-value was calculated using logRank test. **B)** Kaplan-Meier

overall survival analysis of HNSCC HPV positive (left) and HNSCC HPV positive (right) TCGA datasets

stratified for fibroblast abundance first vs last quartile. Below each analysis are shown the

corresponding numbers at risk, time in years. HPV positive HR=4.76 (95%CI 1.36-16.5), p-
value=0.0065. HPV negative HR=1.45 (95%CI 0.96-2.18), p-value=0.076. HR and CI were calculated
using Cox regression. p-value was calculated using logRank test.

1106

1107 **Figure S7:** Enrichment of CoCu8, CoCu30 and RAS84 gene signatures in different co-culture conditions.
1108 **A)** Gene set enrichment analysis (GSEA) plot of CoCu8 gene signature (top) and CoCu30 (bottom) in
1109 co-culture indirect vs direct condition in A431 / VCAF2b transcriptomic dataset. NES and FDR are
1110 specified below each plot. **B)** Gene set enrichment analysis (GSEA) plot of RAS84 gene signature [24] in
1111 mono-culture and co-culture. NES and FDR are specified below each plot.

1112

Figure S8: HB-EGF / RAS / MAPK activity in cancer cells and CAFs mono-culture vs co-culture. **A)**
Correlation plot of CoCu8 and RAS84 expression levels in HNSCC HPV positive UK_HPV positive cohort.
R is Spearman correlation coefficient. **B)** Correlation table of different fibroblast subpopulations
derived from Galbo et al. [7] with RAS84 gene signature in the independent cohort UK_HPV positive of
HNSCC HPV positive patients. The number inside the square represents the Spearman R, correlation
coefficient. The colour legend is shown on the top. All correlations are significant at p-value<0.001.
n=97. **C)** Immunofluorescence staining of E-Cadherin and Hoechst 33342 for SCC154 mono-culture for
the indicated treatments after 48h. **D)** qPCR analysis of *ACTA2*, *FAP*, *FN1* and *LRRC15* genes in OCAF1
mono-culture for the indicated treatments after 48h. mRNA expression is reported as
mean ± standard deviation (SD) fold change difference over non-treated (NT) condition. Genes have
been normalized over the average of *GAPDH*, *ACTB* and *RPLP0* housekeeping genes. n = 3 independent
experiments. **E)** qPCR analysis of *HBEGF* gene in SCC154 after 96h of treatment with the indicated
conditions. mRNA expression is reported as mean ± standard deviation (SD) fold change difference
over siC- condition. Gene has been normalized over the average of *GAPDH*, *ACTB* and *RPLP0*
housekeeping genes. n ≥ 4 independent experiments. Paired t-test. **F)** Violin plot of mRNA expression
levels of the indicated genes for each cluster from Choi et al. [29].

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Figure S9: Enrichment of monocyte/macrophage lineage in patients with abundance of cancer cells –
CAFs cross-talk. **A)** Correlation plot of *HBEGF* and *LIF* expression levels in HNSCC HPV positive TCGA
dataset. R is Spearman correlation coefficient. **B)** Box plot analysis of immune cell absolute score via
Methyl CIBERSORT deconvolution strategy in HNSCC HPV positive separated by first and last quartile
of fibroblast abundance. Independent t-test. Bonferroni correction for multiple comparisons. **C)** Box
plot analysis of immune cell absolute score via Methyl CIBERSORT deconvolution strategy in HNSCC

HPV positive separated by RAS84_0 and RAS84_max. t-test Bonferroni correction for multiple

comparisons.

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1139   **Figure S10:** Schematic representation of the current understanding of cancer cells – CAFs co-culture.

1140   Graphics created with BioRender.com.

Figure 1

Figure 2

Figure 3

Figure 4

Figure 5

Figure S1

A

Cancer cell / CAF cell co-culture

NES	CoCu8	CoCu30
Wi38 with MDA-MB-231	-1.99	-2.30
CDD1112Sk with MDA-MB-231	-2.00	-2.12
HFF1 with MDA-MB-231	-2.05	-2.41
HFF2 with MDA-MB-231	-1.88	-2.15
MDA-MB-231 with Wi38	-1.91	-2.39
MDA-MB-231 with CDD1112Sk	-1.26	-1.48
MDA-MB-231 with HFF1	-1.89	-2.22
MDA-MB-231 with HFF2	-1.85	-2.07
Cal51 with Wi38	-1.92	-2.52
Cal51 with CDD1112Sk	-1.36	-1.76
Cal51 with HFF1	-1.97	-2.28
Cal51 with HFF2	-1.79	-1.82
Wi38 with Cal51	-1.33	-1.16
CDD1112Sk with Cal51	-1.56	-1.61
HFF1 with Cal51	-1.68	-1.97
HFF2 with Cal51	-1.69	-1.96

FDR<0.01
FDR<0.05
FDR n.s.

B

Cancer cell / endothelial cell co-culture

NES	CoCu8	CoCu30
HUVEC with 1205Lu	+0.73	-1.46
1205Lu with HUVEC	-1.12	-1.64

FDR<0.01
FDR<0.05
FDR n.s.

Figure S2

CoCu8 Overall Survival

Figure S3

Figure S4

Figure S5

Figure S7

Figure S8

Figure S9

Figure S10