A simple liquid 3D cell culture paradigm models oxidative mitochondrial metabolism of epithelial breast cancer cells with relevance for lung metastases

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Running title:
3D culture model for OXPHOS and breast cancer metastasis

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ABSTRACT

Three-dimensional (3D) cell culture systems have emerged as powerful tools to model tumor biology ex vivo. However, the diverse array of 3D culture methods available presents challenges in selecting the most appropriate model for specific research questions. This study provides a comparative analysis of breast cancer cells (SUM149, IBC-3, MDA-MB-468) in the mammosphere culture (SphC) model or an "emboli" culture (EmC) model, which enrich for cancer stem cells and epithelial features, respectively. The EmC model, designed originally for inflammatory breast cancer, is characterized by media viscosity and mechanical rocking of the culture vessel. Notably, cells in EmC showed a distinct and durable reduction in cell proliferation while demonstrating increased capacity to establish experimental lung metastases. Ultrastructural quantitative analysis of electron microscopy images suggested that cells in EmC acquire nuclear and mitochondrial features that resemble those of tumor tissue. Proteomics, single-cell transcriptomics, and metabolic flux analyses showed that cells in EmC and SphC favor mitochondrial oxidative metabolism (OXPHOS) and glycolysis, respectively. EmC rendered cells hypersensitive to OXPHOS inhibition, yet more resistant to oxidative stress. Several genes associated with lung metastasis, including ID1, were specifically enriched in EmC. Given the emerging role of OXPHOS in cancer cell survival during dissemination and as established metastases, we propose that the EmC paradigm is a suitable ex vivo model to study signaling pathways relevant for tumor tissue and to assess drug sensitivities and resistance mechanisms of metastatic breast cancer cells ex vivo.

SIGNIFICANCE

This study provides an in-depth characterization of a resource-efficient yet powerful 3D culture paradigm to improve the physiological relevance of ex vivo approaches. Applicable to epithelial cancers, this model offers a platform to accelerate the discovery of physiologically relevant signaling pathways and specific cancer cell vulnerabilities.

INTRODUCTION

For over a century, cell culture has been a workhorse of biological research, conventionally conducted in two-dimensions (1,2). Although 2D cultures provide ease of use and reproducibility, they are limited by nonphysiological substrate stiffness and favor rapidly proliferating cell lines. To better recapitulate physiological conditions, a variety of three-dimensional culture paradigms are being developed with varying degrees of complexity including scaffolds or matrices, various cell types, and/or dynamic features (3-6). Each system presents distinct limitations and advantages depending on the research question to be addressed.

We previously demonstrated that metastatic breast cancer cells cultured in a liquid 3D paradigm, but not in 2D, recapitulate an *in vivo* signaling pathway that preserves E-cadherin protein expression even when its mRNA is downregulated (7). Epithelial-mesenchymal transition (EMT), a well-studied mechanism of metastasis, involves downregulation of genes for epithelial proteins like E-cadherin and induction of mesenchymal proteins such as vimentin. Driven by EMT transcription factors, it promotes stemness and invasiveness (8,9). EMT gene signatures are notably upregulated in metastatic cancer (10) and contribute to the metastatic process (11,12). However, most invasive ductal carcinomas and overt breast cancer metastases retain E-cadherin protein expression, and several studies have reported a positive correlation between E-cadherin, metastasis, and poor prognosis (13-18). Thus, an ex vivo metastatic breast

cancer culture model that preserves E-cadherin-mediated cell-cell adhesions may provide a more physiologically relevant platform for studying metastatic biology and therapeutic vulnerabilities.

The 3D paradigm used in our prior study (7), originally developed by Lehman et al. specifically to model intralymphatic emboli by inflammatory breast cancer (IBC) cells, incorporates lymph-like media viscosity and slow rocking for mild mechanical stimulation (19). We extended this emboli culture (EmC) model to additional breast cancer cell types, enabling the study of cell phenotypes under conditions where cell-cell adhesions are maintained in all dimensions. Here, we present a systematic characterization of breast cancer cells cultured in EmC compared with 2D cultures, and in some contexts, with tumor tissue. Furthermore, because it is a widely used liquid 3D model, we included sphere culture (SphC), which was developed specifically to select for cancer stem cell phenotypes (20,21). Three basal epithelial cell lines were analyzed: two IBC cell lines, the triple-negative breast cancer (TNBC) SUM149 cells and HER2+ IBC-3 cells, as well as MDA-MB-468 cells, also representing TNBC. Our results indicate that the 3D EmC promotes many features observed in primary tumors and lung metastases in breast cancer. Given the current urgency to increase research efforts with human ex vivo models, we propose EmC as a resource-efficient, yet powerful ex vivo approach to advance the identification of context-specific signaling pathways and actionable vulnerabilities.

MATERIALS AND METHODS

Cell culture and cell death assays

SUM149 and IBC-3 cells originated from Asterand Bioscience and MDACC, respectively. MDA-MB-468 cells were from ATCC. Cells were used at passages 2-30, last authenticated in 2025 by GenePrint®10 (Promega), and routinely *Mycoplasma* tested by qPCR. SUM149 expressing luciferase were as described (7). Cells were cultured at 5% CO_2/37°C in media with 10% FBS, 100 units/ml penicillin and 100 µg/ml streptomycin as follows: MDA-MB-468 in Dulbecco's modified Eagle's medium (GIBCO DMEM#11965-092), SUM149 and IBC-3 in Ham's F-12 media (GIBCO #31765092) with 5 µg/ml hydrocortisone (SIGMA #H-0135) and 1 µg/ml Insulin (SIGMA #I-0516). In addition, all media contained 2.25% PEG8000 (SIGMA # 202452) except for sphere cultures (SphC), which were cultured in Mammocult media (StemCell Technologies #05620) supplemented with 2 µg/ml Heparin (StemCell Technologies #07980) and 0.44 µg/ml hydrocortisone. Emboli culture (EmC) was performed essentially as described (19) except that 6-well or 24-well ultra-low attachment (ULA) dishes were used (Corning, #3471, #3473). Briefly, 2.5 x 10^5 cells were seeded in 6-well plates and gently rocked at approximately 40 rpm for 3 days unless otherwise indicated. To isolate the 3D cell assemblies, cells were centrifuged at 500 rpm for 30 sec, washed with PBS, and treated with TrypLExpress (GIBCO #12604-013) for 10 min with intermittent pipetting before neutralization with cell culture medium. Cell numbers were counted by mixing 10 µl of cell suspension with 10 µl of Trypan Blue dye and analyzing by dye exclusion using a Countess II FL (Life Technologies, USA). For cell proliferation measurements, about 2,000 cells were seeded into 96-well F-bottom black microplate (Griener #655090) and cell numbers were measured at the indicated times using a Celigo imaging cytometer (Nexcelom Biosciences). To assess the number of dying cells, cells were treated with 5 µg/ml propidium iodide (PI; SIGMA #P4130) for 30 min and analyzed by PI-based cell death assay using a Celigo imaging cytometer (Nexcelom).

Flow Cytometry

For cell cycle analysis, cells were cultured and trypsinized as described above. Ice-cold 70% ethanol (~500 μl) was added slowly while vortexing the tubes and kept at 4°C for 2 hours. Cells were washed once with 1X FACS washing buffer (PBS + 0.1% BSA) and 200 μl of the propidium iodide staining solution (50 μg/ml PI + 3.8 mM citrate buffer + 10 ng/ml RNase A) was added and incubated for overnight at 4°C. Samples were washed with FACS washing buffer and analyzed by flow cytometry.

For measurement of reactive oxygen species (ROS), 50,000 cells were trypsin harvested as above and resuspended in 400 μl serum-free medium, followed by incubation with CM-H2DCFDA (5 μM; #6827, ThermoFischer Scientific) for 20 min in the dark at 37°C. After 30 min, 400 μl ice-cold PBS + 0.5% BSA were added to stop the reaction, mixed well, and the cells were pelleted at 1,500 rpm for 5 min at RT. The pellet was resuspended in 400 μl PBS + 0.5% BSA. To the DCF-DA preloaded cells, 100 μM H_2O_2 was added and incubated for 15 min at RT in the dark. The cells were then pelleted (1,500 rpm, 5 min), resuspended in 200 μl PBS + 0.5% BSA, and transferred to a 96-well plate (Griener, #655090). The DCF fluorescence was measured with a Spectramax iD3 plate reader (Molecular Devices, USA) using excitation (492 nm) and emission wavelengths (535 nm) settings.

Seahorse Cell Mito Stress Test

Before the samples were loaded, the seahorse 96-well microplate was coated with 30 μl of Poly-L-Lysine and kept inside the hood for 20 min. After the material was decanted, wells were washed twice with cell culture grade water, incubated inside the hood for 30 min, then placed in a non-CO_2 incubator at 37°C for 30 min. This coated plate was taken for Seahorse Cell Mito Stress Test**.** The Seahorse XF Cell Mito Stress Test (Cat#103015-100, Seahorse Bioscience, MA, USA) uses modulators of respiration that target components of the electron transport chain in the mitochondria to measure key parameters of metabolic functions. The oxygen consumption rate (OCR) was measured using a Seahorse XF-96 analyzer. Before OCR measurement, the sensor cartridge was calibrated with calibration buffer at 37°C in a non-CO_2 incubator overnight. A 72-h culture was used unless indicated otherwise. For drug treatments, drugs were added 12 h before analysis. Cultures were isolated, trypsinized for 5 min using TRYPLExpress, counted and 200,000 Cells were seeded onto the 96 well microplate coated with Poly-L-Lysine, washed with XF assay medium (pH 7.4) supplemented with 1 mM pyruvate, 2 mM glutamine, and 10 mM glucose and placed in a 37°C incubator without CO_2 for 1 h before calibration. 1 μM oligomycin, 2 μM protonophoric uncoupler (FCCP) and 0.5 μM antimycin A + rotenone were preloaded in reagent Ports-A, B and C. Three readings were taken after the addition of each reagent prior to injection with subsequent reagents. OCR and ECAR under the same conditions were automatically calculated and recorded by the sensor cartridge and Seahorse XF-96 software. The numerical data of samples in the test were normalized to the protein amount in each well.

Protein isolation and Western blot analysis

Cells were lysed for 15 min on ice with cell lysis buffer containing 10 mM Tris, 1 mM EDTA, 400 mM NaCl, 0.1% NP-40, 10 μl/ml of phosphatase inhibitors 2 and 3 and 10 μl/ml of protease inhibitor cocktail (Sigma #P8340). Samples were centrifuged at 12,700 rpm for 15 min at 4°C. Protein was quantified by Bradford assay kit (Bio-Rad#5000001), and 20 μg was separated on SDS-PAGE and transferred to Nitrocellulose membranes. After blocking with 5% non-fat dry milk solution for 1 h, the membrane was incubated with the indicated antibodies overnight. Following three washes with TBST (TBS containing 0.05% Tween-20), appropriate secondary antibody was added for 1 h followed by three washes before

scanning with the iBRIGHT 1500 Imaging System (Thermo Fischer Scientific). Antibodies were obtained from the following sources: Cell Signaling Technology (ID1, #23369S; ID3, #9837); Santa Cruz Biotechnology (GAPDH, #sc-47724); DHSB (α-tubulin, # 12G10). IACS-10759 (#S8731), Sodium dichloroacetate (DCA; #S8615) and AGX51 (#E1152) were from Selleck chemicals. DMSO (#D-2650) and Poly-L-Lysine (#P4707) were from Millipore SIGMA.

Xenograft tumors and experimental metastasis assays

To generate primary tumors, SUM149-GFP-Luc cells(7) (2×10^6) were orthotopically injected into NSG mice (8-12 weeks old) and tumors at end point of about 2000 mm³ tumor volume were collected and frozen for proteomics. For electron microscopy and array tomography, a small tumor of about 60 mm³ was collected and fixed immediately (for details see sample preparation for array tomography). For experimental lung metastases, 0.5 million cells in 100 µl PBS were injected by tail vein into 8-12-week-old NSG female mice. Within two hours, bioluminescence was determined to assure successful inoculation (0-day measurement). Bioluminescence was monitored at 2-week intervals for up to 20 weeks using the IVIS Spectrum imager (PerkinElmer Inc.).

All experiments were according to approved protocols. Animal care was provided in accordance with the procedures outlined in the "Guide for Care and Use of Laboratory Animals" including those pertaining to studies of neoplasia (National Research Council, 1996; National Academy Press; Washington, D.C.). NCI-Frederick is accredited by AALACi and follows the Public Health Service Policy for the Care and Use of Laboratory Animals.

Sample preparation for proteomic analysis.

Tissue/Cell lysis, digestion, TMTpro labeling: Three-day cultures from 2D, EmC and SphC from SUM149 and IBC-3 cells were harvested and washed with PBS (2X) and the pellets and/or tissue samples were treated with 300 µl of EasyPep lysis buffer (ThermoFischer Scientific #A45735) and 1 µl of universal nuclease (Thermo #88700). Homogenized with pestle and centrifuged for 5 min at 14,000g. Removed the supernatant to a clean tube. Cell pellets were treated with 100 µl of EasyPep lysis buffer and 1 µl of universal nuclease. Protein concentration was determined using the BCA method. For each sample, the volume corresponding to 50 µg of protein was adjusted to 100 µl total with lysis buffer then treated with 50 µl of reducing and alkylating solutions provided with the EasyPep kit (Thermo #A40006) and incubated at 95°C for 10 min then allowed to cool to room temperature. Samples were then treated with 4 µg of trypsin/LysC provided with the EasyPep kit and incubated at 37°C for 19 h overnight. Samples were then treated with 40 µl of 12.5µg/µl TMTpro in acetonitrile and incubated at room temperature for 1h. Samples were quenched with 50µL of 5% hydroxylamine, 20% formic acid (FA) for 10 min and then combined. Samples were cleaned using spin column provided with the EasyPep kit and eluted peptides were dried in a speedvac.

<u>LC-MS analysis:</u> Dried peptides were resuspended in 0.1%FA and injected in triplicate using a Dionex U3000 RSLC in front of a Orbitrap Eclipse (Thermo) equipped with an EasySpray ion source. Solvent A consisted of 0.1%FA in water and Solvent B consisted of 0.1%FA in 80%ACN. Loading pump consisted of Solvent A and was operated at 7 µl /min for the first 6 min of the run then dropped to 2 µl /min when the valve was switched to bring the trap column (Acclaim™ PepMap™ 100 C18 HPLC Column, 3µm, 75µm I.D., 2cm, PN 164535, Thermo) in-line with the analytical column (EasySpray, 2µm, 75µm I.D., 50cm, PN ES903, Thermo). The gradient pump was operated at a flow rate of 300 nl/min and each run used a linear LC gradient of 5-7%B for 1min, 7-30%B for 134 min, 35-50%B for 35 min, 50-95%B for 4 min, holding

at 95%B for 7 min, then re-equilibration of analytical column at 5%B for 17 min. All MS injections employed the TopSpeed method with a 3 sec cycle time that consisted of the following: Spray voltage was 1800V and ion transfer temperature of 275 ⁰C. MS1 scans were acquired in the Orbitrap with resolution of 120,000, AGC of 4e5 ions, and max injection time of 50ms, mass range of 400-1600 m/z; MS2 scans were acquired in the Ion Trap using turbo scan method with AGC of 1e4, max injection time of 35ms, CID energy of 35%, isolation width of 0.7Da, intensity threshold of 1e4 and charges 2-5 for MS2 selection. Real Time Search (RTS) was enabled for TMTpro labeled peptides using the Uniprot Human Feb 2020 database. RTS settings were Full tryptic peptides, 1 max missed cleavage, fixed carbamidomethyl on Cys, fixed TMTpro on Lys and peptide N-terminus, and variable oxidation on Met. MS3 scans for peptides identified by RTS were acquired in the Orbitrap with resolution of 50,000, AGC of 2.5e5, max injection of 86ms, mass range 100-500 m/z, and isolation width of 2Da. Advanced Peak Determination, Monoisotopic Precursor selection (MIPS), and EASY-IC for internal calibration were enabled and dynamic exclusion was set to a count of 1 for 15sec.

Database searching and post-processing: All three injections were batched together as fractions and all MS files were searched with Proteome Discoverer 2.3 using the Sequest node. Data was searched against the Uniprot Human database from Feb 2020 using a full tryptic digest, 2 max missed cleavages, minimum peptide length of 6 amino acids and maximum peptide length of 144 amino acids, an MS1 mass tolerance of 10 ppm, MS2 mass tolerance of 0.6 Da, variable oxidation on methionine (+15.995 Da) and fixed modifications of carbamidomethyl on cysteine (+57.021), TMTpro (+304.207) on lysine and peptide N-terminus. Percolator was used for FDR analysis and TMTpro reporter ions were quantified using the MS3 Reporter Ion Quantifier node and normalized on total peptide intensity of each channel. TMTpro channel assignment for conditions can be found in Table S2. Statistical analysis was performed within PD2.3 and Log2FC was calculated based on the median of each group and adjusted p-value was calculated using ANOVA. Differentially expressed proteins were set using Log2FC cutoff of +/- 0.6 and adjusted p-value of <0.05. For proteins that were observed in 2 or more replicates of a condition and 0 of the other in a comparison, the Log2FC was set to +/- 0.1 the highest/lowest value in the comparison and adjusted p-value set slightly lower than the lowest value calculated in the comparison. Pathway enrichment analysis was performed using Ingenuity Pathway Analysis and the Log2FC threshold was lowered to +/- 0.1 and the adjusted p-value was kept at <0.05.

Note: Data are being deposited in MassIVE and will be publicly available upon acceptance of the manuscript for publication.

Reversed-Phase Ion-Pairing LC-MS² Assay for Measuring Cell Central Carbon Metabolites

Three-day cultures of SUM149 and IBC-3 cell under 2D, SusC, EmC, and SphC conditions were used. All targeted central carbon metabolite (CCM) reference compounds, including acetyl-CoA and α-ketoglutarate, were purchased from Sigma-Aldrich (St. Louis, MO). The stable isotope-labeled internal standards (SI-CCM) were $^{13}C_3$-lactate and $^{13}C_4$-succinic acid (Cambridge Isotope Laboratory, Andover, MA), as well as $^{13}C_6$-glucose-6-phosphate and $^{13}C_6$-fructose-1,6-diphosphate (Medical Isotopes, Inc., Pelham, NH). All CCM and SI-CCM analytical standards have reported chemical and isotopic purity ≥ 98% and were used without further purification. OmniSolv® LC-MS grade acetonitrile and methanol were obtained from EMD Millipore (Billerica, MA). Tributylamine (TBA), LC-MS grade acetic acid and formic acid were purchased from Fisher Scientific (Hampton, NH). LC-MS grade ammonium formate was obtained from CovaChem (Loves Park, IL). All chemicals and solvents used in this study were HPLC or reagent grade unless otherwise noted. The isotope dilution liquid chromatography-tandem mass spectrometry method adapted

from previous publications (22,23) was employed to measure the concentrations of cell CCMs.In brief, cell pellets were extracted with chilled 80% methanol in water supplemented with appropriate isotopic standards. Reversed-phase ion-pairing LC-MS^2 analysis was performed using Thermo TSQ Quantiva triple quadrupole mass spectrometer (Thermo Scientific, San Jose, CA) coupled with Shimadzu 20AC-XR liquid chromatography system for the CCM measurement. The mass spectrometers were operated in negative ion mode and set to monitor parent-product ion transitions using Selected Reaction Monitoring (SRM). Quantitation of targeted metabolites was performed using Xcalibur™ Quan Browser (Thermo Scientific). Calibration curves for each metabolite were constructed by plotting CCM/SI-CCM peak area ratios obtained from the calibration samples versus their CCM concentrations and fitting the data using linear regression with $1/X$ weighting. The analyte concentrations in experimental samples were then interpolated using the linear function obtained from the calibration curve.

RNA isolation and quantitative RT-PCR

Total RNA was isolated using GeneJET RNA purification kit (Thermo Scientific, #K0732), and 2 µg RNA was taken for cDNA synthesis using Superscript Reverse Transcriptase III (RT) according to manufacturer's instructions (Invitrogen, #18080044). Quantitative PCR was carried out with Fast SYBR Green master mix (#4385612, Applied Biosystems, Foster City, USA) using the 7500 Fast Real-Time PCR instrument (Applied Biosystems) and the relative expression levels were measured using the relative quantitation $\Delta\Delta C$t method and normalized to *actin*. Data represent three independent biological replicates; each measured as technical triplicates. Primers were as follows: *ID1* (5'-aaacgtgctgctctacgaca-3' and 5'-gattccgagttcagctccaa-3'; *MRPS12 (*5'-gctacctgctccatggctac-3' and 5'-cacttgcgattggctgagt-3'); *NDUSF6 (*5'-cacactggccaggtttatga-3' and 5'-tggtgctgtctgaactggag-3'); *RPLP0* (5'-gcaatgttgccagtgtctgtc-3' and 5'-gccttgacctttcagcaagt-3').

Single Cell Library Preparation, Sequencing and Analysis Method

Three-day cultures of SUM149 and IBC-3 cells grown under 2D, EmC, and SphC conditions were harvested, washed with PBS, and had RNA isolated as per manufacturer's instructions and 2 µg RNA was taken. Two biological replicates were labeled with Hashtag1 or Hashtag2 for each experimental condition. The single-cell gene expression libraries were generated employing the 10x Chromium Single Cell 3' Reagent kit v3 (CG000183 Rev A) with Feature Barcode technology in accordance with the guidelines provided by 10X Genomics, US. The TotalSeq -B anti-human Hashtag reagents (BioLegend #394602, 394604, 394606) were used for cell hashing and multiplex samples. These libraries were sequenced on a NovaSeq system utilizing two SP flowcells from Illumina, US. The sequencing run was setup as 28 cycles for Read1, 8 cycles for i7 index, and 75 cycles for Read2 according to protocol recommendation. Sequencing run demultiplexing was performed by using Cellranger mkfastq, with one mismatch allowed in the barcodes. Cell Ranger v4.0.0 was applied for alignment, tagging, gene and transcript counting for Chromium 3' gene expression libraries, and sample demultiplexing was done using Cell Ranger v4.0.0 based on hashtag feature barcodes. The human genome GRCH38 reference with Gencode v32 /Ensembl98 annotation (refdata-gex-GRCh38-2020-A) were used as reference in Cell Ranger analysis. Seurat (v5) (24) R package was used for single cell RNA-seq (scRNA-seq) data analyses, including cell hashtag filtering, quality control, normalization, clustering, visualization, as well as differential expression (DE) analysis. After cell hashtag demultiplexing, only singlets with reasonable expression matrices were kept for downstream analysis. For DE analysis, we conducted pairwise comparisons among the three culture conditions, and we also compared each culture condition versus the other two conditions. Differentially expressed genes (DEGs) were identified with adjusted P value less than 0.05 and average log2 fold change

more than 1 (or less than -1) cutoffs. Genes commonly up- or downregulated in both cell lines were used as inputs for QIAGEN Ingenuity Pathway Analysis (QIAGEN Inc., https://digitalinsights.qiagen.com/IPA, version 01-23-01) (25).

Note: Data are being deposited in GEO and will be publicly available upon acceptance of the manuscript for publication.

Sample preparation for electron microscopy

SUM149 breast cancer cells were cultured as described above, harvested and washed once with PBS, and Karnovsky's fixative (4% formaldehyde and 2% glutaraldehyde in 0.1 M cacodylate buffer) was added and incubated at RT for 5 h with gentle pipetting every hour to prevent clumping. Samples were centrifuged at 500 rpm for 30 sec and the supernatant decanted. 1 ml cacodylate buffer (0.1 M) was added and centrifuged again at 500 rpm for 30 sec. The pellet was suspended in cacodylate buffer and kept at 4°C. Cells from 2D culture were pelleted and embedded in 1% low-melt agarose before continuing with sample preparation. All other cell growth types were left free-floating. Cells were post-fixed in 2% osmium tetroxide and 1.5% potassium ferricyanide in 0.1 M sodium cacodylate for 1 h at room temperature, washed with ultrapure water, and then stained with 1% aqueous uranyl acetate overnight at 4°C. After further washes in ddH$_2$O, the cells were treated with lead aspartate at 65°C for 30 min and washed again. The samples were dehydrated through graded ethanol (35%, 50%, 70%, 95%, and 100% x 5, 10 min. each) and propylene oxide (PO; 5 x 10 min.). The cells were infiltrated with increasing concentrations of Polybed 812 resin (hard formulation) in PO (resin: PO; 1:3, 1:2, 3:1, and 100%), embedded in 100% degassed resin, and cured at 65°C for 48 h in flat BEEM capsules (EMS, Warrington, PA, USA). After polymerization, resin blocks were trimmed with a Leica EM TRIM2 milling system. After trimming the block face down, 100 nm sections were cut with a 45° ultra-diamond knife (Diatome) using a Leica ARTOS ultramicrotome. Sections were collected with a loop and placed on glow-discharged ITO coverslips.

Array tomography SEM imaging and segmentation

Sections on ITO coverslips were mounted onto 4-inch type-p silicon wafers using copper tape and affixed to a 4-inch stage-decel holder in the GeminiSEM 450 (Carl Zeiss). Low-resolution overview scans of individual resin sections were acquired at 150 nm pixel resolution using ATLAS 5 Array Tomography software (Fibics). High-resolution (5 nm pixels) large areas were imaged by stitching multiple fields, enabled by automated stage movement and a four-quadrant backscatter detector (aBSD, Zeiss). Image strips of roughly 40,000 x 8,000 pixels were cropped from the larger images to capture roughly similar numbers of cells. These strips were imported into Napari (26) where segmentation was performed using the empanada plugin. The appropriate deep learning models were deployed in the 2D inference module with image tiling: mitochondria were segmented using MitoNet_v1 (27) lipid droplets with DropNet_base_v1, and nuclei with NucleoNet_base_v1(to be reported elsewhere). Errors in model predications were corrected manually in empanada-napari. To facilitate correction in these large images, the "create tiles" module in empanada was used to generate smaller image tiles to reduce the time needed for manual adjustments. After all corrections were completed on the individual tiles, both the images and their segmentations were reassembled using the "merge tiles" module in empanada. Cell segmentation was performed in 3DSlicer (Ver. 5.8.1; (28) using thresholding to delineate borders for each cell in every image, followed by filling in using the fill tool in Napari. Quantitative measurements of cells, mitochondria, lipid droplets, and nuclei were obtained with the napari-clusters-plotter plugin, which utilizes region props from scikit-image (29,

30). The cell packing ratios were determined by taking the total pixel area of all cells that were segmented in an image divided by the total pixel area of the field of view.

Statistics

Unless stated otherwise, quantitative data were analyzed by the two-tailed unequal variance t-test and are shown as the mean ± S.E.M. Graphs were made using GraphPad Prism10 software. The number of samples (n) refers to biological replicates.

RESULTS

Durable alterations of breast cancer cell proliferation in 3D depending on culture paradigm

In this study, we examined four culture conditions (**Figure 1A)** using three epithelial breast cancer cell lines (SUM149, IBC-3, and MDA-MB-468). Apart from conventional 2D culture of these adherent cells on plastic, all other conditions were in ultra-low attachment plates. Because emboli culture (EmC) requires addition of PEG8000 to standard, cell-line specific media (19), PEG8000 was also added to 2D and suspension culture (SusC) to control for the effects of media viscosity and molecular crowding. Thus, comparison of 2D and SusC reveals the effect of losing substrate attachment. Comparison of SusC and EmC provides information on the effect of the dynamic nature of EmC, which is the only difference between these two types of culture. The media composition is the only difference between SusC and Sphere cultures (SphC) which were in specific medium that enriches stem cell like features (31). EmC and SphC differ in both media composition and dynamic versus static conditions. Bright-field imaging of 3-day cultures revealed characteristic structures formed under each 3D condition (**Figures 1B and S1A**). Cell clusters were smallest in SusC, where loss of substrate attachment limits aggregation, largest in EmC, likely reflecting enhanced aggregation and reinforced cell-cell adhesions under fluid dynamic forces. Comparison of the numbers of cells within 3D structures (isolated by brief low-speed centrifugation), together with flow-cytometry cell cycle analysis, showed that loss of substrate by SusC markedly reduced the cell proliferation rate, whereas the media in SphC rescued proliferation because it is optimized for stemness (**Figure 1C-D**). The mechanical forces in EmC did not change the proliferation rate of SUM149 or MDA-MB-468 cells in suspension, but reduced proliferation of IBC-3 cells even further. Notably, when cells from 3D culture were dissociated and re-plated in 2D, these proliferation rate differences persisted for at least 4 days (**Figure 1E**). Compared to cells never exposed to 3D culture, cells from SphC proliferated more rapidly, whereas cells from SusC, and especially EmC, showed a durable reduction in their proliferation rates. These results indicate that 3D culture conditions impart lasting cell adaptations that are detectable at the level of proliferation. To address molecular differences, we conducted single-cell mRNA-Seq (scSeq) of SUM149 and IBC-3 cells in 2D, SphC and EmC. UMAP cluster analysis revealed that each population of SUM149 cells segregated well with only limited overlap of clusters containing cells of two or more culture conditions (**Figure SB-C**). For IBC-3, cells in 2D and EmC were more similar, while those in SphC segregated the most. Next, we identified differentially expressed genes (DEGs) for each condition in bulk to all others (**Supplemental File 1**) and performed pathway analyses (**Supplemental File 2**). Comparison of the pathways with most significant z-scores validated the observed reduction in cell proliferation in EmC while cholesterol biosynthesis was upregulated in SphC (**Figure 1F**). Increased cholesterol biosynthesis may be consistent with the higher proliferation rate in SphC, although mitosis-related pathways were only

upregulated in IBC-3 in SphC (Supplemental File 2). Collectively, these results demonstrate that cancer cells durably and differentially alter their cell proliferation rate in SphC versus EmC.

Mammosphere culture is known to enrich for cancer stem cells in cell lines, increasing the tumor take rate and experimental metastasis rate (32,33). In contrast, the reduced proliferation rate of cells in EmC could potentially indicate senescence or dormancy. Pilot experiments determining these states were inconclusive, so we next assessed the capacity of SUM149 cells, which express a luciferase reporter, to seed experimental metastasis. Cells from 3D cultures were dissociated into single cells and injected by tail vein into immunocompromised mice; 2D-cultured cells served as controls. Bioluminescence intensity (BLI) was determined on the same day to confirm lung seeding. As anticipated, most injected cells died during the initial phase post-injection, and four weeks later most mice did not show any BLI (**Figure 1G**). However, as expected (33), the frequency of mice with BLI signal was higher in mice which received cells from SphC (34.5%) compared to 2D controls (8.7%). In pilot experiments, we had determined that injection of 10^5 cells was suitable to demonstrate the enhanced colonization by cells from SphC. Strikingly, even more mice (60.7%) injected with cells from EmC exhibited lung colonization, and BLI quantification confirmed that EmC increased lung colonization efficiency relative to both 2D or SphC cells (**Figure 1H**). By week 20, nearly all mice injected with EmC-derived cells developed detectable lung colonies (**Figure 1G**). While we cannot exclude the possibility that re-establishment of cell-cell adhesions between cells from EmC cultures enhanced initial survival after injection (33), these results demonstrate that EmC does not diminish malignancy. Rather, despite their lower proliferation rate in culture, EmC-derived cells exhibit heightened metastatic competence.

SUM149 ultrastructural features in EmC bear most resemblance to primary tumors
Given the distinct morphologies of cell assemblies across the three 3D conditions (Figure 1B), we employed volume electron microscopy, specifically array tomography (34), to assess the ultrastructural adaptations of SUM149 cells and their resemblance to *in vivo* tumor tissue. As shown in **Figure 2A**, electron microscopy images revealed that the cell packing ratio was highest and most similar in EmC and tumor tissue. Cells from 2D culture were included in this analysis, but packing ratio does not apply to this condition. In addition, mitochondria and nuclei appeared largest, and cell-to-cell connectedness most uninterrupted, in EmC and tumors (**Figures S2A-B**). AI-based algorithmic quantification of nuclear and mitochondrial morphometrics in one sample per condition further underscored these similarities. Nuclear analyses showed that area and solidity, signifying mostly convex or positive curvature of the membrane, were highest and most similar in tumor and EmC samples (**Figure 2B**). Nuclear aspect ratios (length/width) and perimeters were similar across culture conditions and tumor tissue (**Figure S2C**). The data indicate that nuclear morphology in EmC resembles the more "normalized" phenotype of tumors, though increased nuclear size has also been correlated to neoplastic malignancy (35). Mitochondrial analysis revealed that cells in EmC and tumor tissue harbored the largest mitochondria by area and major axis length, whereas mitochondria in SphC were the least circular with the highest aspect ratios (**Figures 2C and S2D**). To extend this analysis, we developed new algorithms to examine lipid droplet (LD) morphology. Cells in SphC contained the largest LDs with the greatest area and major axis length, while LDs in EmC more closely resembled those in tumor tissue, showing similarity in major axis length, circularity, and staining intensity (**Figures 2D and S2E-F**). Although the determinants of LD staining intensity remain unclear, we speculate that the lipid composition could be a determining factor. In contrast to a prior ultrastructural study of SUM149 emboli (36), we did not observe many microvilli. Full quantifications and statistical analyses are provided in **Supplemental Files 3 and 4**. Together, these data suggest that SUM149 cells adopt distinct

ultrastructural features depending on culture condition, and that EmC recapitulates certain aspects of mitochondrial, nuclear, and lipid droplet morphology observed in tumor tissue *in vivo*.

Adaptation of mitochondrial metabolism depends on 3D paradigm
To probe molecular adaptations to different growth conditions, we conducted proteomic analysis of the total levels of proteins in 2D, SphC, and EmC compared to xenograft tumor tissues (**Supplemental File 5**). Principal component analysis (PCA) plots of the data from culture conditions (2D, SphC, and EmC) and primary tumors showed well separated clustering of the biological replicates (**Figure S3A**). Using the 2D proteome as reference, we compared the differentially expressed proteins (DEPs) between conditions (**Figures 3A and S3B**). As expected, the more complex and heterogenous tumor tissue was most divergent from cells cultured in 2D (236 up, 576 down). Cells from SphC showed the least DEPs (70 up, 15 down), while cells from EmC were intermediate (148 up, 159 down). Cells from EmC showed the most overlap with tumor tissue (**Figure 3A**). Analysis of the proteins exclusively overrepresented in EmC and tumors by ToppGene (37) showed that 35/39 were part of the Metabolism Reactome (GSEA ID MM14563), while 116/132 underrepresented proteins were part of the RNA Metabolism Reactome (GSEA ID M16843), encompassing splicing and translation. Ingenuity Pathway Analysis (**Figures S3C and Supplemental File 5**) further showed that translational activity was underrepresented in tumors and in EmC consistent with the reduced proliferation rate in EmC compared to 2D (Figure 1D). Regarding metabolic pathways, not surprisingly, tumor contained more proteins in glycolysis and less in oxidative phosphorylation pathways (**Figure 3B**). In contrast, both EmC and SphC showed increases in TCA cycle pathway as well as glycolysis and/or Warburg effect. Collectively, the proteomic analysis showed distinct and common adaptations of cells to different 3D culture conditions and suggest that cells in EmC share more similarities with tumor tissue than do cells in SphC.

To further examine metabolic aspects, we determined the levels of central carbon metabolites in SUM149 cells by mass spectrometry. Four TCA cycle metabolites were enriched in EmC compared to SphC (**Figure 3C**). Similar observations were made at the level of pyruvate and glutamate, which feeds into the TCA cycle, and NADH, which is largely produced by the TCA cycle and fuels the electron transfer chain (ETC) (**Figure S3D**). In contrast, four glycolysis pathway intermediates including lactate were enriched in SphC in comparison to EmC (**Figure 3D**). Other metabolites that were measured were similar between conditions, namely malate, fumarate fructose-6-phosphate, phosphoenolpyruvate, glucose-6-phosphate and glyceraldehyde-3-phosphate. Elevated levels of lactate in SphC over EmC were also observed in IBC-3 cells (**Figure 3E**). While steady state levels of metabolites alone do not allow conclusions on the activity of metabolic enzymes, these data suggested that TCA cycle activity of cells in EmC may drive oxidative phosphorylation (OXPHOS) while cells in SphC may favor glycolysis. To test this hypothesis directly, we proceeded to functionally examine metabolic activities by Seahorse metabolic flux technology. For this, the 3D cell structures were dissociated and single cells seeded for analysis. Comparison of the oxygen consumption rates (OCR), a measure of mitochondrial respiration and oxidative phosphorylation, showed that cells from SUM149 cells in EmC exhibited elevated basal and maximal OCR (**Figure 3F-G**). In contrast, the extracellular acidification rate (ECAR), a measure of glycolytic activity, was reduced in EmC compared to 2D and SphC, and lowest in SusC (**Figures 3H-I**). IBC-3 cells demonstrated overall much higher metabolic activity with modest changes by modification of culture conditions. Nevertheless, EmC exhibited higher OCR and lower ECAR compared to SphC (**Figure S3E-F**). An increase of basal OCR in EmC was also observed in MDA-MB-468 cells (**Figure S3G**).

Collectively, these data suggest that EmC supports oxidative phosphorylation (OXPHOS), while SphC supports glycolytic metabolism.

Specific 3D models reveal distinct drug vulnerabilities

Given the differential metabolic pathway activities between cells in EmC versus SphC, we next examined their responses to pathway-specific inhibitors. For OXPHOS, we used IACS-010759, a complex I inhibitor of the ETC (38). In addition, we used sodium dichloroacetate (DCA) which inhibits pyruvate dehydrogenase kinases (PDK) and thereby promotes the conversion of pyruvate to acetyl CoA, blocking its conversion to lactate (39). Treatment of established 3D cultures of SUM149 or IBC-3 cells with IACS-010759 for three days lead to a dose-dependent reduction in cell numbers in EmC (**Figure 4A**), accompanied by increased cell death (**Figure S4A**), whereas SphC cells were completely resistant. Given that IBC-3 cells showed high levels of basal OCR in all culture conditions, the resistance to IACS-010759 in SphC was surprising. In contrast, SphC cells showed significant sensitivity to DCA (10 mM) while cells in EmC were comparatively more resistant (**Figure 4B and S4B).** These data indicate that cells in SphC rely mostly on aerobic glycolysis, which is consistent with previous reports (39,40). In contrast, cells in EmC are most dependent on OXPHOS for their metabolic sustenance.

Mitochondrial oxidative phosphorylation is a source of reactive oxygen species (ROS) and may trigger oxidative stress (41). We therefore asked whether cells in EmC may exhibit elevated ROS levels or oxidative stress by staining with the cell-permeable fluorogenic dye H_2DCFDA (42). Furthermore, we assessed the response to imminent oxidative stress through exposure to exogenous H_2O_2. As shown in **Figure 4C**, the basal level of H_2DCFDA fluorescence of SUM149 and IBC-3 cells was in fact lower in EmC-derived cells in comparison to 2D and SphC. As expected, H_2O_2 exposure for 15 minutes increased H_2DCFDA fluorescence in all culture conditions but remained lowest in EmC. After two hours of exogenous H_2O_2 exposure, the numbers of viable cells were reduced in 2D and SphC relative to untreated controls but were unchanged in EmC (**Figure 4D**). These data indicate that, despite their reliance on OXPHOS, EmC cells are most protected from oxidative stress..

EmC promotes features associated with lung metastasis

Metabolic plasticity is a hallmark of cancer and the metastatic process (43). In epithelial breast cancer, oxidative mitochondrial metabolism and fatty acid oxidation have been associated with metastatic capacity and especially to the lung (44, 45), which is among the most common sites of breast cancer metastasis (46). In EmC, we observed induction of two OXPHOS related genes that are associated with lung metastasis, *NDUFS6* and *MRPS12* (47) (**Figure 5A**). Furthermore, analysis of the scRNA-Seq data showed that other genes associated with lung metastases were also enriched by EmC in SUM149 and/or IBC-3 cells: *ID1, ID3, SPARC, EREG*, *PTGS2* (48-50), *KRT16* (51,52), *MMP7* (53), and *LCN2* (54) (**Figure 5B**). Validation experiments confirmed the upregulation of ID1 at both the mRNA and protein levels (**Figure 5C**). ID1, an inhibitor of E-box binding transcription factors, is an important regulator of cell cycle, self-renewal, and tumorigenesis (55-57). Consequently, treatment of SUM149 cells in EmC with AGX51, which targets ID proteins for degradation (58), reduced ID1 and ID3 protein levels (**Figure 5D**) and decreased viable cell numbers in a dose-dependent manner (**Figure 5E and S4C**). Similar results were observed in MDA-MB-468 cells (**Figure 5F and S4D)**. Collectively, these results indicate the relevance for ID1/3 upregulation in EmC and suggest a potential mechanism contributing to the high lung colonizing capacity of EmC-derived SUM149 cells (Figure 1G-H).

DISCUSSION

The primary objective of this study was to comprehensively characterize breast cancer cells grown in emboli culture (EmC) and compare the findings with more traditional cell culture models, such as 2D culture and mammospheres (SphC). Our results indicate three distinct advantages of EmC as an *ex vivo* model system: (a) reduced cell proliferation, (b) an oxidative mitochondrial metabolic profile (OXPHOS), and (c) ultrastructural and architectural features reminiscent of tumor tissue *in vivo*. These characteristics are crucial because the high proliferation rates in standard *in vitro* systems often make cells artificially sensitive to pharmacological interventions, leading to false positive results in drug screening and ultimately poor clinical translation (4). In contrast, 3D models, including EmC (7), typically exhibit resistance to drug treatments, making them more appropriate for evaluating therapeutic responses (59).

Notably, we observed that the fluid dynamics inherent to the EmC system promote tissue-like cell-cell contacts and ultrastructural changes. While previous research has explored the impact of fluid shear stress and tissue stiffness on cancer progression (60,61), we do not claim that the specific mechanical forces at play in EmC directly model a particular in vivo scenario. However, tissues are exposed to pulsatile fluid flow in blood and lymph vessels, which are likely providing mechanical cues. For lung cancer, specifically, the lung environment provides significant mechanical forces that shape tumor development and progression (62). Our transcriptomic analysis revealed upregulation of several genes in EmC that associated with breast cancer lung metastasis. Notably, Keratin 16 (*KRT16*) regulates mitochondrial dynamics and is upregulated in lung metastases (51,63). Circulating tumor cells (CTCs) expressing *KRT16* are also associated with shorter relapse-free survival (52). Other lung-metastasis-associated genes upregulated in EmC, including mitochondrial *MRPS12* and *NDUFS6*, as well as *ID1* and *ID3*, are implicated in promoting OXPHOS and lung colonization (51). Higher expression of *MRPS12* and *NDUFS6* correlates with decreased metastasis-free survival in breast cancer (47). Interestingly, despite robust mitochondrial oxidative activity, EmC cells maintained low levels of reactive oxygen species (ROS), consistent with previous findings that active TCA cycling is associated with reduced ROS generation (64). This aligns with reports that an epithelial phenotype is linked with protection from oxidative stress, which may be mediated through ID1 and KRT16 (63,65).

At the level of energy metabolism, our analyses demonstrate that SphC cells predominately rely on aerobic glycolysis, whereas EmC cells favor mitochondrial OXPHOS. Although glycolytically derived pyruvate canonically feeds the TCA cycle and OXPHOS unless shunted to lactate via the Warburg effect (39,66), recent studies suggest that lactate itself can serve as a fuel for mitochondrial metabolism (67,68). However, only EmC cells, but not SphC cells, were sensitive to inhibition of the electron transfer chain, underscoring a reliance on mitochondrial oxidative metabolism. These observations are especially relevant in the context of metastasis, where metabolic reprogramming enables tumor cells to adapt and thrive within new microenvironments (69,70). Studies have shown that OXPHOS is induced in epithelial breast cancer cells and specifically CTCs and metastasis-initiating cells (44,71-73). Elevated OXPHOS gene expression is correlated with decreased patient survival, resistance to chemotherapy, and is a metabolic vulnerability in therapy-resistant triple-negative breast cancers (74-77). Consistent with this, metastatic breast cancer cells in the lung amplify OXPHOS metabolism (78), and inhibition of OXPHOS specifically impedes tumor cell seeding of the lung—even when the primary tumor itself is more glycolytic (51).

Combined with our findings that EmC-derived cells have enhanced colonization of mouse lungs and the prevalence of OXPHOS metabolism, the above reports support the notion that EmC may serve as

a valuable ex vivo paradigm to investigate pathways relevant to lung metastasis. Although most of our analyses were performed using two inflammatory breast cancer (IBC) cell lines, some findings were validated in a non-IBC TNBC line. Additionally, given that our prior work identified a pathway through EmC that was also observed in a non-IBC TNBC PDX model, i.e. COX-2 through AKT/GSK3β to E-cadherin stabilization (7), we postulate that the EmC system may be broadly applicable at least to basal epithelial breast cancers and potentially other subtypes, although this warrants further investigation. Identifying an *ex vivo* culture model that faithfully recapitulates in vivo tumor phenotypes, especially of difficult to study lung metastases, is critical for dissecting the mechanisms that empower metastatic breast cancer cells and for uncovering new therapeutic targets. Our findings establish EmC as a resource-efficient ex vivo model that captures the ultrastructural, metabolic, and functional hallmarks of metastatic breast cancer, providing a valuable tool for probing the biology and therapeutic sensitivities of other epithelial tumor types.

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FIGURE LEGENDS

Figure 1. Cells cultured in EmC reduce cell proliferation ex vivo but acquire lung colonization efficiency in vivo

    A. Overview of cell culture conditions used in this study. See Methods for further details.

    B. Bright filed images of the of indicated cell lines after three days in different 3D cultures (scale bar =1 mm).

    C. Viable cell counts after three days of culture under the indicated conditions and starting with the same number of cells in each (n = 4).

    D. Flow cytometric analysis of DNA content of the cells from SUM149 and IBC-3 cells cultured as in panel C (n = 2).

    E. Quantification of viable cells by imaging cytometer replated in 2D (day 0) after 3 days under the indicated culture conditions (n = 3). Statistical analysis refers to day 4.

    F. Z-scores with P<0.05 derived from IPA analysis of common DEGs in EmC and SphC respectively compared to "all others".

    G. Numbers of mice, from four independent experiments, showing bioluminescence (BLI) in the lungs from SUM149 cells that had been cultured as 2D, SphC and EmC for 3 days before injection by tail vein. Data are for measurements at 4 weeks after injection and cumulative numbers until 20 weeks after injection. *P* values determined by Fischer's exact test as follows:. 2D vs SphC, *P*= 0.045; 2D vs EmC, *P*= 0.00013; SphC vs EmC, *P*= 0.065.

    H. Quantification of bioluminescence of all mice in week 4 after injection. n = 22-29, *P* values were determined by Wilcoxon Test (unpaired, two-sided).

        All quantitative data, except panels G-H, are mean ± SEM, *P<0.05, **P<0.01, ***P<0.001, ****P<0.0001, n.s. not significant.

Figure 2. Ultrastructural analysis of SUM149 cell cultures and tumor tissue.

    A. Representative scanning electron microscope and respective cell segmentation images of SUM149 cells after 3 days under the indicated culture conditions and as xenograft tumor. Scale bar = 10 μm.

    B. Nuclear metrics of the indicated features for each condition (n = 50-68).

    C. Mitochondrial metrics of the indicated features for each condition (n = 1451-1995).

    D. Lipid droplet metrics of the indicated features for each condition (n = 131-654).

        Note: *P<0.01, **P<0.001, ****P<0.00001, n.s. not significant.

Figure 3. Proteomic and metabolic analysis reveals differential cell adaptations to 3D culture.

    A. Venn diagram showing the numbers of SUM149 proteins differentially under- or overrepresented in EmC, SphC and tumor in comparison to 2D.

    B. Most significant z-scores with P<0.05 of pathways in common between primary tumors, EmC, and SphC in comparison to 2D.

    C. Fractional enrichment of the indicated intracellular metabolites of SUM149 cells in SphC and EmC (n = 4).

    D. Fractional enrichment of the indicated intracellular metabolites of SUM149 cells in SphC and EmC (n = 4).

E. Intracellular lactate levels from IBC-3 cells at indicated culture conditions (n = 3).
F. Oxygen consumption rate (OCR) over time of SUM149 cells derived from the indicated culture conditions. Shown are a representative data normalized to protein, n = 3 technical replicates per time point)
G. Basal and maximal OCR, three time points each, as shown in panel F.
H. Extracellular acidification rate (ECAR) over time determined as in panel F.
I. Basal ECAR under indicated conditions for first three time points shown in panel H.
Bar graphs show mean ± SEM, *$P<0.05$, **$P<0.01$, ****$P<0.0001$, n.s. not significant.

Figure 4. EmC renders cells sensitive to ETC inhibitors and protected from oxidative stress
A. Fold change in the numbers of viable SUM149 and IBC-3 cells in EmC and SphC after 72 h treatments with OXPHOS inhibitor IACS-010579, added at the indicated concentrations to the established cultures. DMSO was used as vehicle control (n = 3).
B. Fold change in the numbers of viable SUM149 and IBC-3 cells in EmC and SphC after 72 h treatments with DCA, added at the indicated concentrations to the established cultures. DMSO was used as vehicle control (n = 3).
C. Quantification of reactive oxygen species (ROS) by DCF fluorescence measurements in SUM149 and IBC-3 cells from the indicated culture conditions. H_2O_2 (100 µM) was added to the dissociated cells 15 min prior to analysis (n = 3, mean ± SEM).
D. Fold change in the numbers of viable SUM149 and IBC-3 cells as in panel C after 2 hours of exposure to H_2O_2.
Data are mean ± SEM, *$P<0.05$, **$P<0.01$, ***$P<0.001$, ****$P<0.0001$). n.s. not significant.

Figure 5: EmC leads to induction of genes related to lung metastasis.
A. qRT-PCR of mRNA levels of indicated genes in SUM149 and MDA-MB-468 cells under the indicated culture conditions (n = 3).
B. Differential expression of the indicated genes derived from the scRNA-Seq data comparing EmC versus SphC of SUM149 and IBC-3 cells as indicated.
C. qRT-PCR (top, n = 3) and Western analysis (bottom) of ID1 expression in different SUM149 cultures. Tubulin served as loading control.
D. Western analysis of ID1 and ID3 proteins in established SUM149 EmC that were treated with AGX51 at the indicated concentrations for 3 days. GAPDH served as loading control.
E. Fold change in the numbers of viable SUM149 cells in established EmC after treatment with AGX51 at the indicated concentrations for 3 days (n = 3).
F. Fold change in the numbers of viable MDA-MB-468 cells in established EmC after treatment with 100 µM for 3 days (n = 3).
Data are shown as mean ± SEM, *$P<0.05$, **$P<0.01$, ***$P<0.001$, ****$P<0.0001$. n.s. not significant.

REFERENCES

1.   Benien P, Swami A. 3D tumor models: history, advances and future perspectives. Future Oncol 2014;10(7):1311-27.
2.   Moro LG, Guarnier LP, Azevedo MF, Fracasso JAR, Lucio MA, Castro MV, et al. A Brief History of Cell Culture: From Harrison to Organs-on-a-Chip. Cells 2024;13(24).
3.   Sumbal J, Budkova Z, Traustadottir GA, Koledova Z. Mammary Organoids and 3D Cell Cultures: Old Dogs with New Tricks. J Mammary Gland Biol Neoplasia 2020;25(4):273-88.
4.   Rauner G, Gupta PB, Kuperwasser C. From 2D to 3D and beyond: the evolution and impact of in vitro tumor models in cancer research. Nat Methods 2025.
5.   Wang Y, Jeon H. 3D cell cultures toward quantitative high-throughput drug screening. Trends Pharmacol Sci 2022;43(7):569-81.
6.   Polak R, Zhang ET, Kuo CJ. Cancer organoids 2.0: modelling the complexity of the tumour immune microenvironment. Nature Reviews Cancer 2024;24(8):523-39.
7.   Balamurugan K, Poria DK, Sehareen SW, Krishnamurthy S, Tang W, McKennett L, et al. Stabilization of E-cadherin adhesions by COX-2/GSK3beta signaling is a targetable pathway in metastatic breast cancer. JCI Insight 2023:2022.01.18.476813.
8.   Yang J, Antin P, Berx G, Blanpain C, Brabletz T, Bronner M, et al. Guidelines and definitions for research on epithelial-mesenchymal transition. Nat Rev Mol Cell Biol 2020;21(6):341-52.
9.   Brabletz S, Schuhwerk H, Brabletz T, Stemmler MP. Dynamic EMT: a multi-tool for tumor progression. EMBO J 2021;40(18):e108647.
10.   Jiang B, Mu Q, Qiu F, Li X, Xu W, Yu J, et al. Machine learning of genomic features in organotropic metastases stratifies progression risk of primary tumors. Nat Commun 2021;12(1):6692.
11.   Williams ED, Gao D, Redfern A, Thompson EW. Controversies around epithelial-mesenchymal plasticity in cancer metastasis. Nat Rev Cancer 2019;19(12):716-32.
12.   Bruner HC, Derksen PWB. Loss of E-Cadherin-Dependent Cell-Cell Adhesion and the Development and Progression of Cancer. Cold Spring Harb Perspect Biol 2018;10(3).
13.   Burandt E, Lubbersmeyer F, Gorbokon N, Buscheck F, Luebke AM, Menz A, et al. E-Cadherin expression in human tumors: a tissue microarray study on 10,851 tumors. Biomark Res 2021;9(1):44.
14.   Horne HN, Oh H, Sherman ME, Palakal M, Hewitt SM, Schmidt MK, et al. E-cadherin breast tumor expression, risk factors and survival: Pooled analysis of 5,933 cases from 12 studies in the Breast Cancer Association Consortium. Sci Rep 2018;8(1):6574.
15.   Rodriguez FJ, Lewis-Tuffin LJ, Anastasiadis PZ. E-cadherin's dark side: possible role in tumor progression. Biochim Biophys Acta 2012;1826(1):23-31.
16.   Kowalski PJ, Rubin MA, Kleer CG. E-cadherin expression in primary carcinomas of the breast and its distant metastases. Breast Cancer Res 2003;5(6):R217-22.
17.   Hugo HJ, Gunasinghe N, Hollier BG, Tanaka T, Blick T, Toh A, et al. Epithelial requirement for in vitro proliferation and xenograft growth and metastasis of MDA-MB-468 human breast cancer cells: oncogenic rather than tumor-suppressive role of E-cadherin. Breast Cancer Res 2017;19(1):86.
18.   Kleer CG, van Golen KL, Braun T, Merajver SD. Persistent E-cadherin expression in inflammatory breast cancer. Mod Pathol 2001;14(5):458-64.
19.   Lehman HL, Dashner EJ, Lucey M, Vermeulen P, Dirix L, Van Laere S, et al. Modeling and characterization of inflammatory breast cancer emboli grown in vitro. Int J Cancer 2013;132(10):2283-94.
20.   Nunes AS, Barros AS, Costa EC, Moreira AF, Correia IJ. 3D tumor spheroids as in vitro models to mimic in vivo human solid tumors resistance to therapeutic drugs. Biotechnol Bioeng 2019;116(1):206-26.

21. Dontu G, Wicha MS. Survival of mammary stem cells in suspension culture: implications for stem cell biology and neoplasia. J Mammary Gland Biol Neoplasia 2005;10(1):75-86.
22. Furusawa T, Cavero R, Liu Y, Li H, Xu X, Andresson T, et al. Metabolism-focused CRISPR screen unveils mitochondrial pyruvate carrier 1 as a critical driver for PARP inhibitor resistance in lung cancer. Mol Carcinog 2024;63(6):1024-37.
23. Lu W, Clasquin MF, Melamud E, Amador-Noguez D, Caudy AA, Rabinowitz JD. Metabolomic analysis via reversed-phase ion-pairing liquid chromatography coupled to a stand alone orbitrap mass spectrometer. Anal Chem 2010;82(8):3212-21.
24. Hao Y, Stuart T, Kowalski MH, Choudhary S, Hoffman P, Hartman A, et al. Dictionary learning for integrative, multimodal and scalable single-cell analysis. Nat Biotechnol 2024;42(2):293-304.
25. Kramer A, Green J, Pollard J, Jr., Tugendreich S. Causal analysis approaches in Ingenuity Pathway Analysis. Bioinformatics 2014;30(4):523-30.
26. contributors n. napari: a multi-dimensional image viewer for python. doi:105281/zenodo3555620 2019.
27. Conrad R, Narayan K. Instance segmentation of mitochondria in electron microscopy images with a generalist deep learning model trained on a diverse dataset. Cell Syst 2023;14(1):58-71 e5.
28. Fedorov A, Beichel R, Kalpathy-Cramer J, Finet J, Fillion-Robin JC, Pujol S, et al. 3D Slicer as an image computing platform for the Quantitative Imaging Network. Magn Reson Imaging 2012;30(9):1323-41.
29. van der Walt S, Schonberger JL, Nunez-Iglesias J, Boulogne F, Warner JD, Yager N, et al. scikit-image: image processing in Python. PeerJ 2014;2:e453.
30. Zigutyte L, Savill R, Müller J, Zoccoler M, Wagner T, Haase R. napari-clusters-plotter (v0.9.0). Zenodo https://doiorg/105281/zenodo15685613 2025.
31. Al-Hajj M, Wicha MS, Benito-Hernandez A, Morrison SJ, Clarke MF. Prospective identification of tumorigenic breast cancer cells. Proc Natl Acad Sci U S A 2003;100(7):3983-8.
32. Kong D, Hughes CJ, Ford HL. Cellular Plasticity in Breast Cancer Progression and Therapy. Front Mol Biosci 2020;7:72.
33. Aceto N, Bardia A, Miyamoto DT, Donaldson MC, Wittner BS, Spencer JA, et al. Circulating tumor cell clusters are oligoclonal precursors of breast cancer metastasis. Cell 2014;158(5):1110-22.
34. Peddie CJ, Genoud C, Kreshuk A, Meechan K, Micheva KD, Narayan K, et al. Volume electron microscopy. Nat Rev Methods Primers 2022;2:51.
35. Singh I, Lele TP. Nuclear Morphological Abnormalities in Cancer: A Search for Unifying Mechanisms. Results Probl Cell Differ 2022;70:443-67.
36. Fujii Y, Krishnamurthy S, El-Zein R. Ultrastructural Analysis of Inflammatory Breast Cancer Cell Clusters in an Ex Vivo Environment Mechanically Mimicking the Lymph Vascular System. Breast Cancer (Auckl) 2021;15:11782234211056134.
37. Chen J, Bardes EE, Aronow BJ, Jegga AG. ToppGene Suite for gene list enrichment analysis and candidate gene prioritization. Nucleic Acids Res 2009;37(Web Server issue):W305-11.
38. Kalyanaraman B, Cheng G, Hardy M, You M. OXPHOS-targeting drugs in oncology: new perspectives. Expert Opin Ther Targets 2023;27(10):939-52.
39. Stacpoole PW. Therapeutic Targeting of the Pyruvate Dehydrogenase Complex/Pyruvate Dehydrogenase Kinase (PDC/PDK) Axis in Cancer. J Natl Cancer Inst 2017;109(11).
40. Ciavardelli D, Rossi C, Barcaroli D, Volpe S, Consalvo A, Zucchelli M, et al. Breast cancer stem cells rely on fermentative glycolysis and are sensitive to 2-deoxyglucose treatment. Cell Death Dis 2014;5:e1336.
41. Kuznetsov AV, Margreiter R, Ausserlechner MJ, Hagenbuchner J. The Complex Interplay between Mitochondria, ROS and Entire Cellular Metabolism. Antioxidants (Basel) 2022;11(10).
42. Jakubowski W, Bartosz G. 2,7-dichlorofluorescin oxidation and reactive oxygen species: what does it measure? Cell Biol Int 2000;24(10):757-60.
43. Martinez-Reyes I, Chandel NS. Cancer metabolism: looking forward. Nat Rev Cancer 2021.

44. Young CM, Beziaud L, Dessen P, Madurga Alonso A, Santamaria-Martinez A, Huelsken J. Metabolic dependencies of metastasis-initiating cells in female breast cancer. Nat Commun 2023;14(1):7076.

45. Whitaker-Menezes D, Martinez-Outschoorn UE, Flomenberg N, Birbe RC, Witkiewicz AK, Howell A, et al. Hyperactivation of oxidative mitochondrial metabolism in epithelial cancer cells in situ: visualizing the therapeutic effects of metformin in tumor tissue. Cell Cycle 2011;10(23):4047-64.

46. Jin L, Han B, Siegel E, Cui Y, Giuliano A, Cui X. Breast cancer lung metastasis: Molecular biology and therapeutic implications. Cancer Biol Ther 2018;19(10):858-68.

47. El-Botty R, Morriset L, Montaudon E, Tariq Z, Schnitzler A, Bacci M, et al. Oxidative phosphorylation is a metabolic vulnerability of endocrine therapy and palbociclib resistant metastatic breast cancers. Nat Commun 2023;14(1):4221.

48. Minn AJ, Gupta GP, Siegel PM, Bos PD, Shu W, Giri DD, et al. Genes that mediate breast cancer metastasis to lung. Nature 2005;436(7050):518-24.

49. Gupta GP, Perk J, Acharyya S, de Candia P, Mittal V, Todorova-Manova K, et al. ID genes mediate tumor reinitiation during breast cancer lung metastasis. Proc Natl Acad Sci U S A 2007;104(49):19506-11.

50. Eltarhouny SA, Elsawy WH, Radpour R, Hahn S, Holzgreve W, Zhong XY. Genes controlling spread of breast cancer to lung "gang of 4". Exp Oncol 2008;30(2):91-5.

51. Davis RT, Blake K, Ma D, Gabra MBI, Hernandez GA, Phung AT, et al. Transcriptional diversity and bioenergetic shift in human breast cancer metastasis revealed by single-cell RNA sequencing. Nat Cell Biol 2020;22(3):310-20.

52. Elazezy M, Schwentesius S, Stegat L, Wikman H, Werner S, Mansour WY, et al. Emerging Insights into Keratin 16 Expression during Metastatic Progression of Breast Cancer. Cancers (Basel) 2021;13(15).

53. Han B, Zhou B, Qu Y, Gao B, Xu Y, Chung S, et al. FOXC1-induced non-canonical WNT5A-MMP7 signaling regulates invasiveness in triple-negative breast cancer. Oncogene 2018;37(10):1399-408.

54. Shi H, Gu Y, Yang J, Xu L, Mi W, Yu W. Lipocalin 2 promotes lung metastasis of murine breast cancer cells. J Exp Clin Cancer Res 2008;27(1):83.

55. Zhao Z, Bo Z, Gong W, Guo Y. Inhibitor of Differentiation 1 (Id1) in Cancer and Cancer Therapy. Int J Med Sci 2020;17(8):995-1005.

56. Teo WS, Holliday H, Karthikeyan N, Cazet AS, Roden DL, Harvey K, et al. Id Proteins Promote a Cancer Stem Cell Phenotype in Mouse Models of Triple Negative Breast Cancer via Negative Regulation of Robo1. Frontiers in Cell and Developmental Biology 2020;8.

57. Ling F, Kang B, Sun XH. Id proteins: small molecules, mighty regulators. Curr Top Dev Biol 2014;110:189-216.

58. Wojnarowicz PM, Escolano MG, Huang YH, Desai B, Chin Y, Shah R, et al. Anti-tumor effects of an ID antagonist with no observed acquired resistance. NPJ Breast Cancer 2021;7(1):58.

59. Rao X, Qiao Z, Yang Y, Deng Y, Zhang Z, Yu X, et al. Unveiling Epigenetic Vulnerabilities in Triple-Negative Breast Cancer through 3D Organoid Drug Screening. Pharmaceuticals (Basel) 2024;17(2).

60. Northey JJ, Przybyla L, Weaver VM. Tissue Force Programs Cell Fate and Tumor Aggression. Cancer Discov 2017;7(11):1224-37.

61. Follain G, Herrmann D, Harlepp S, Hyenne V, Osmani N, Warren SC, et al. Fluids and their mechanics in tumour transit: shaping metastasis. Nat Rev Cancer 2020;20(2):107-24.

62. Gargalionis AN, Papavassiliou KA, Basdra EK, Papavassiliou AG. Update on the relevance of mechanobiological mechanisms in lung cancer. Transl Oncol 2025;55:102375.

63. Steen K, Chen D, Wang F, Majumdar R, Chen S, Kumar S, et al. A role for keratins in supporting mitochondrial organization and function in skin keratinocytes. Molecular biology of the cell 2020;31(11):1103-11.

64. Nolfi-Donegan D, Braganza A, Shiva S. Mitochondrial electron transport chain: Oxidative phosphorylation, oxidant production, and methods of measurement. Redox Biol 2020;37:101674.

65. Bensellam M, Montgomery MK, Luzuriaga J, Chan JY, Laybutt DR. Inhibitor of differentiation proteins protect against oxidative stress by regulating the antioxidant-mitochondrial response in mouse beta cells. Diabetologia 2015;58(4):758-70.

66. Zangari J, Petrelli F, Maillot B, Martinou JC. The Multifaceted Pyruvate Metabolism: Role of the Mitochondrial Pyruvate Carrier. Biomolecules 2020;10(7).

67. Cai H, Chen X, Liu Y, Chen Y, Zhong G, Chen X, et al. Lactate activates trained immunity by fueling the tricarboxylic acid cycle and regulating histone lactylation. Nat Commun 2025;16(1):3230.

68. Cai X, Ng CP, Jones O, Fung TS, Ryu KW, Li D, et al. Lactate activates the mitochondrial electron transport chain independently of its metabolism. Mol Cell 2023;83(21):3904-20 e7.

69. Navarro C, Ortega A, Santeliz R, Garrido B, Chacin M, Galban N, et al. Metabolic Reprogramming in Cancer Cells: Emerging Molecular Mechanisms and Novel Therapeutic Approaches. Pharmaceutics 2022;14(6).

70. Welch DR, Hurst DR. Defining the Hallmarks of Metastasis. Cancer Res 2019;79(12):3011-27.

71. Jiang Z, He J, Zhang B, Wang L, Long C, Zhao B, et al. A Potential "Anti-Warburg Effect" in Circulating Tumor Cell-mediated Metastatic Progression? Aging Dis 2024.

72. LeBleu VS, O'Connell JT, Gonzalez Herrera KN, Wikman H, Pantel K, Haigis MC, et al. PGC-1alpha mediates mitochondrial biogenesis and oxidative phosphorylation in cancer cells to promote metastasis. Nat Cell Biol 2014;16(10):992-1003, 1-15.

73. Herst P, Carson G, Lewthwaite D, Eccles D, Schmidt A, Wilson A, et al. Residual OXPHOS is required to drive primary and metastatic lung tumours in an orthotopic breast cancer model. Front Oncol 2024;14:1362786.

74. Pacheco-Velazquez SC, Robledo-Cadena DX, Hernandez-Resendiz I, Gallardo-Perez JC, Moreno-Sanchez R, Rodriguez-Enriquez S. Energy Metabolism Drugs Block Triple Negative Breast Metastatic Cancer Cell Phenotype. Mol Pharm 2018;15(6):2151-64.

75. Uslu C, Kapan E, Lyakhovich A. Cancer resistance and metastasis are maintained through oxidative phosphorylation. Cancer Lett 2024;587:216705.

76. Evans KW, Yuca E, Scott SS, Zhao M, Paez Arango N, Cruz Pico CX, et al. Oxidative Phosphorylation Is a Metabolic Vulnerability in Chemotherapy-Resistant Triple-Negative Breast Cancer. Cancer Res 2021;81(21):5572-81.

77. Baek ML, Lee J, Pendleton KE, Berner MJ, Goff EB, Tan L, et al. Mitochondrial structure and function adaptation in residual triple negative breast cancer cells surviving chemotherapy treatment. Oncogene 2023;42(14):1117-31.

78. Sotgia F, Whitaker-Menezes D, Martinez-Outschoorn UE, Flomenberg N, Birbe RC, Witkiewicz AK, et al. Mitochondrial metabolism in cancer metastasis: visualizing tumor cell mitochondria and the "reverse Warburg effect" in positive lymph node tissue. Cell Cycle 2012;11(7):1445-54.

Figure 1

Figure 2

Figure 3

Figure 4

Figure 5

A. SUM149

NDUFS6 — mRNA (fold); bars for 2D, SusC, EmC, SphC
n.s., n.s., *, **, *

MRPS12 — mRNA (fold); bars for 2D, SusC, EmC, SphC
n.s., n.s., *, *, *

MDA-MB-468

NDUFS6 — mRNA (fold); bars for 2D, SusC, EmC, SphC
n.s., n.s., ***, **, *

MRPS12 — mRNA (fold); bars for 2D, SusC, EmC, SphC
n.s., n.s., ***, ***, *, ***

B.

	avg_log2FC		p_val_adj	
	EmC vs. SphC		EmC vs. SphC	
	SUM149	IBC-3	SUM149	IBC-3
ID1	3.36	1.35	0	9.08E-75
ID3	3.07	0.81	0	3.03E-22
SPARC	0.23	n/a	0.00246	n/a
EREG	0.40	-0.68	1.06E-15	1
PTGS2	0.88	n/a	9.92E-91	n/a
KRT16	1.04	n/a	1.72E-17	n/a
MMP7	2.67	n/a	3.41E-23	n/a
LCN2	1.55	1.86	4.19E-46	0

C. SUM149 *ID1*
mRNA (fold); bars for 2D, SusC, EmC, SphC
n.s., **, *, ****, ****
ID1
Tubulin

D. SUM149
AGX51 (µM)   -   30   100
ID1
ID3
GAPDH

E. SUM149
Cells (fold change); AGX51 (µM): -, 30, 100
*, ***, **

F. MDA-MB-468
Cells (fold change); AGX51 (µM): -, 100
****