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FIBROBLAST-DERIVED EXTRACELLULAR MATRIX: AN ALTERNATIVE CELL
CULTURE SUBSTRATE THAT ALTERS LUNG CANCER CELL LINE PHENOTYPE

By

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B.S., University of Louisville

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FIBROBLAST-DERIVED EXTRACELLULAR MATRIX: AN ALTERNATIVE CELL
CULTURE SUBSTRATE THAT ALTERS LUNG CANCER CELL LINE PHENOTYPE

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ABSTRACT

Poor lung cancer survival can largely be contributed to the metastatic cells that invade and spread throughout the body. The tumor microenvironment (TME) is composed of multiple cell types, as well as non-cellular components. The TME plays a critical role in the development of metastatic cancers by providing migratory cues that change the growing tumor's properties. The Extracellular Matrix (ECM), a main component of the TME, has been shown to change composition during tumor progression, allowing cancer cells to invade tissue and survive away from the primary cancer site. Although the ECM is well-known to influence the fate of tumor progression, little is known about the molecular mechanisms that are affected by the cancer cell-ECM interactions. It is imperative that these mechanisms are understood in order to properly understand and prevent lung cancer dissemination. However, common *in vitro* studies do not incorporate these interactions into everyday cell culture assays. In our lab, we have adopted a model that examines decellularized human fibroblast-derived ECM as a 3D substrate for growth of lung adenocarcinoma cell lines. It is hypothesized that the interactions between lung cancer cell lines and fibroblast-derived ECM will alter phenotypes important for lung cancer progression. Here, we have characterized the effect of various fibroblast-derived matrices on the properties of various lung cancer adenocarcinoma cell lines. Such altered processes include morphology, growth, and migration. This work highlights the significance of the cell-ECM interaction and its requirement for incorporation into *in vitro* experiments. Implementation of a fibroblast-derived ECM as an *in vitro* technique will provide researchers with an important factor to manipulate to better recreate and study the TME.

TABLE OF CONTENTS

ACKNOWLEDGEMENTS.....	iv
ABSTRACT.....	v
TABLE OF CONTENTS.....	vi
NOMENCLATURE.....	viii
LIST OF TABLES.....	ix
LIST OF FIGURES.....	x
I. INTRODUCTION	1
II. METHODS.....	6
A. Cell culture.....	6
B. ECM Production and Decellularization.....	7
C. Phase Microscopy and Fluorescent Microscopy.....	8
D. SDS PAGE and Western Blot.....	9
E. Cell Growth.....	10
F. Trans-well Migration Assay.....	11
G. Microarray.....	12
H. Real Time PCR.....	13
I. Time-Lapse Microscopy.....	14
J. Cell Circularity.....	15
K. Subcutaneous Tumor Formation.....	15
L. Dose Response of Chemotherapeutics.....	16
M. Statistical Analysis.....	16

III. RESULTS AND DISCUSSION.....	16
A. Fibroblasts and Their Derived Extracellular Matrix.....	16
B. Fibroblast-derived Matrices Alter Lung Cancer Cell Morphology.....	18
C. Fibroblast-derived Matrices Alter Lung cancer Cell Growth.....	20
D. Fibroblast-Derived Matrices alter functional protein levels of lung cancer cell lines.....	22
E. Fibroblast Derived Matrices Do not alter Drug Resistance of Lung Cancer Cell Lines...	25
F. Fibroblast Derived Matrices Protect Lung Cancer Cells from Serum Deprivation.....	27
G. Fibroblast-Derived Matrices Change mRNA levels of Various Genes in Lung Cancer Cells.....	29
H. Fibroblast-Derived Matrices can Induce Lung Cancer Cells to Migrate.....	36
I. Fibroblast-derived ECM increase Subcutaneous Xenograft Tumor Formation in Mice...	38
IV. CONCLUSIONS AND FUTURE WORK.....	40
V. APPENDIX.....	42
VI. REFERENCES.....	63

NOMENCLATURE

μm =micrometer
ASNS=Asparagine Synthase
BCA = Bicinchoninic acid
BCAT1 = Branched-chain amino acid transaminase 1
BME = basement membrane extract
CAFs = Cancer-associated fibroblasts
CREB = Cyclic-AMP responsive Element Binding
ECM = Extracellular Matrix
EMT = Epithelial-to-Mesenchymal transition
FAK = Focal Adhesion Kinase
FAP = fibroblast-activated protein
FBS = fetal bovine serum
GAGs = Glycosaminoglycans
IL-8 = Interleukin – 8
KCNMA1 = Potassium Large Conductance Calcium-Activated Channel family M
LOX = Lysyl Oxidase
MAPK = Mitogen-activated Protein Kinase
MMPs = Matrix metalloproteases
MOMP = mitochondrial outer membrane permeablization
MRNA = messenger ribonucleic acid
MTOR = mammalian target of Rapamycin
NH₄OH = Ammonium Hydroxide
NSCLC = Non-small cell lung cancer
NT5E = 5' nucleotidase
PSAT1 = phosphor-serine aminotransferase 1
QPCR = quantitative polymerase chain reaction
TBS-T = Tris-buffered saline with Tween-20
TIMP = tissue inhibitors of metalloproteases
TME = Tumor Microenvironment
VEGF = Vascular Endothelial Growth Factor

LIST OF TABLES

TABLE I: List of Primer Sequences for qPCR.....	13
TABLE II: Common Ranked Down-regulated Genes between Both A549 and H358 on WI38 and IMR90 Fibroblast-derived ECM.....	33
TABLE III: Common Ranked Up-regulated Genes between Both A549 and H358 on WI38 and IMR90 Fibroblast-derived ECM.....	34
TABLE IV: A549 Serum Deprivation Statistical Analysis.....	42
TABLE V: H358 Serum Deprivation Statistical Analysis.....	43
TABLE VI: HPL1D Serum Deprivation Statistical Analysis.....	45
TABLE VII: ECM Protein Statistical Analysis.....	46
TABLE VIII: Time-lapse Statistical Analysis.....	47
TABLE IX: Directionality Statistical Analysis.....	48
TABLE X: Circularity Statistical Analysis.....	49
TABLE XI: A549 Cell Counting Statistical Analysis.....	49
TABLE XII: H358 Cell Counting Statistical Analysis.....	51
TABLE XIII: HPL1D Cell Counting Statistical Analysis.....	52
TABLE XIV: A549 Alamar Blue Statistical Analysis.....	54
TABLE XV: H358 Alamar Blue Statistical Analysis.....	55
TABLE XVI: HPL1D Alamar Blue Statistical Analysis.....	57
TABLE XVII: Xenograft Volume Statistical Analysis.....	58
TABLE XVIII: Xenograft Weight Statistical Analysis.....	59
TABLE XIX: Trans-well Migration Statistical Analysis.....	60

LIST OF FIGURES

FIGURE 1: The Tumor Microenvironment.....	3
FIGURE 2: Schematic of Fibroblast-derived ECM protocol.....	8
FIGURE 3: Depiction of Trans-well Assay.....	12
FIGURE 4: Fibroblasts and their ECM.....	18
FIGURE 5: Fibroblast-derived ECM alter Lung Cancer Cell Line Morphology.....	20
FIGURE 6: Fibroblast-derived ECM alter Lung Cancer Cell Growth.....	22
FIGURE 7: Fibroblast-derived ECM alters Protein Levels of A549 Cells.....	24
FIGURE 8: Fibroblast-derived ECM do does not alter drug response to Gemcitabine and Cisplatin.....	26
FIGURE 9: Fibroblast-derived ECM Protects Lung Cancer Cell Lines from Serum Deprivation.....	29
FIGURE 10: Fibroblast-derived ECM alters mRNA profile of A549 and H358 Cells.....	32
FIGURE 11: Fold Change of mRNA by Quantitative qPCR.....	35
FIGURE 12: Fibroblast-derived ECM alter Cell Migration of Lung Cancer Cell Lines in a Trans- well Chamber.....	36
FIGURE 13: Fibroblast-derived ECM induce Directional Migration.....	38
FIGURE 14: Fibroblast-derived ECM Increase Subcutaneous Tumor Formation of A549 Cells in Mice.....	40

I. Introduction

The five-year survival rate for stage 3 lung cancer patients is only around 15% (American Cancer Society, 2015). This poor survival rate is largely contributed to the metastatic form of the disease, which allows the cancer to become a systemic burden, by infiltrating vital organs. 40%-50% of patients with Non-small-cell lung cancers (NSCLC), which is the classification for nearly 80% of all lung cancers, have metastatic lung cancer at diagnosis (Ihde et al., 1992). Although survival rates improve with early detection, there is a great need for efficacious therapies that treat the metastatic form of lung cancer. There are many FDA approved therapies that are successful for lung cancer patients (eg. surgical resection, local radiation, and chemotherapeutics), but few therapies exist that are effective at specifically targeting cancer cells, while leaving healthy cells untouched, and even fewer that are effective against the metastatic cancers. This failure to produce effective therapies is partly due to false discoveries that are attributed to lack of appropriate models that accurately recapitulate the *in vivo* mechanisms that drive lung cancer and its progression to metastasis (Hoelder et al., 2012). For instance, many cancer therapies are developed from chemicals that illicit a cancer specific cytotoxic response during *in vitro* cell culture environments, but these cell culture environments do not offer the full biological repertoire that is present in the human disease. This means that researchers are limited in the accuracy of their conclusions thus leading them down an incorrect path that may ultimately result in failure in the clinical setting. Although cell culture experiments

are a simple first-line test for new therapies, an improved *in vitro* model could filter out inefficacious treatments before large financial and temporal investments are made.

It is now a fully accepted paradigm that neoplasms are a dynamic environment that host multiple cell types that influences the behavior and outcome of the cancer (Wood et al., 2014). Many studies are being done to determine the exact role the tumor microenvironment (TME) constituents, which include cancer cells, immune cells, fibroblasts, vasculature, and the extracellular matrix (ECM) (Schwendener et al., 2014) (Figure 1). The immune cells have been heavily studied for their role in carcinogenesis because they can support tumor growth and influence angiogenesis by secreting the necessary growth factors, which include VEGF and IL-8 (Wood et al., 2014). Cancer-associated fibroblasts (CAFs), which are fibroblasts that reside within the tumor, but differ from normal resident fibroblasts, have also been shown to be pro-tumorigenic by providing growth factors TGF- β , PDGF, and FGF, as well as cytokines such as IL-8, CXCL14, and IL-6 (Cirri et al., 2011). Many novel therapies have been produced that target the TME. For instance, Bevacizumab is a monoclonal antibody that inhibits lung cancer angiogenesis by targeting VEGF (Das et al., 2012). It is clear that the best way to treat the patient is to treat the entire TME.

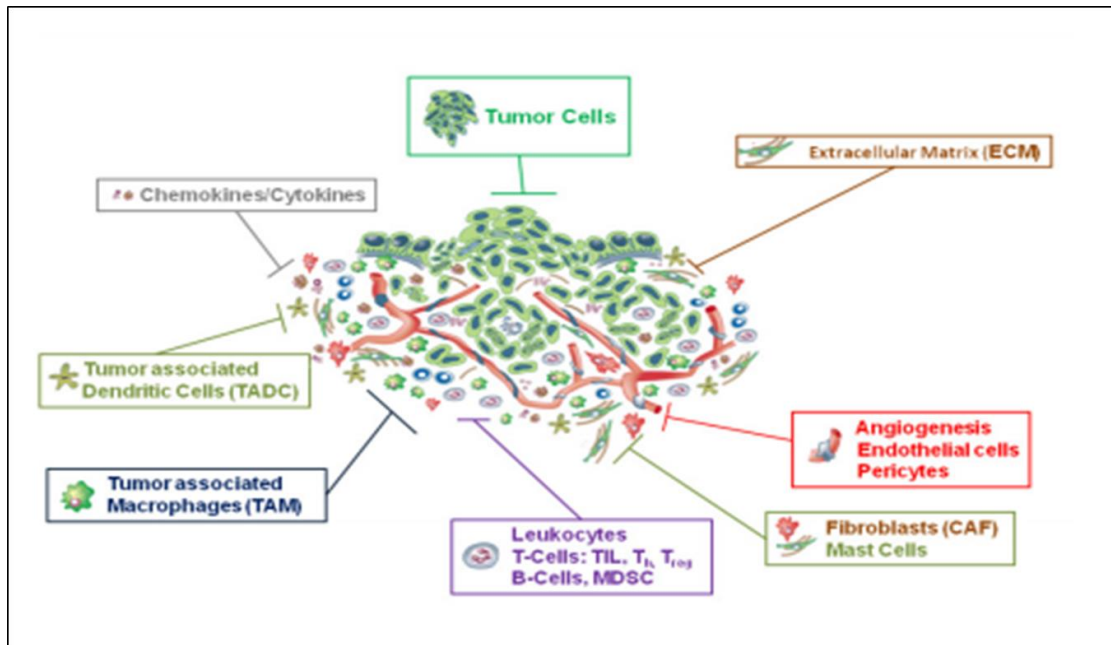


Figure 1- The Tumor Microenvironment

The ECM, an essential constituent of the TME, is a meshwork of protein fibers and glycosaminoglycans (GAGs) that not only provides mechanical support, but also offers growth and migration cues through growth factors, adhesion interactions, and mechano-transduction (Oskarsson et al., 2013). The ECM is generally secreted and organized by fibroblasts, but other cells can contribute to ECM production, such as endothelial and epithelial cells (Lu et al., 2012). Lately, the ECM has been heavily researched for its role in the progression of lung and breast carcinomas (Lu et al., 2012; Oskarsson et al., 2013; Yang et al., 2011). The balance of ECM deposition and ECM degradation can potentiate diseases such as fibrosis and cancer (Cox et al., 2011). Increased production of the high elastic modulus collagen and decreased low elastic modulus elastin expression can stiffen local tissue, therefore altering mechano-transduction pathways (Butcher et al., 2009). Matrix metalloproteases (MMPs) are matrix-degrading enzymes

that can degrade the ECM and alter its elasticity, which can provide cells with important biomechanical stimulation to direct invasion into surrounding tissue and blood vessels, leading to metastasis (Hadler-Olsen et al., 2013). Alternately, ECM can be stiffened by increased matrix production and deposition of collagen via Lysyl Oxidase (LOX) signaling (Gao et al., 2010). For instance, ECM accumulation by increased Collagen deposition has been documented in many tumor cell types, including glioma, breast, and lung cancers (Huijbers et al., et al; Caccavari et al., 2010). This abnormal ECM function can cause changes in the mechano-transduction pathways that regulate growth and migration pathways. Tension-induced signaling has been shown to affect Mitogen-Activated Protein Kinase (MAPK) signaling pathways by p44/42 activation in fetal lung epithelial cell lines (Copland et al., 2007). MAPK signaling is highly affected in cancer that activates many downstream applications. Similarly, focal adhesions are the point of cell-ECM matrix interaction and are comprised of integrins that cluster together and bind the ECM, thus triggering downstream pathways mediated through Focal Adhesion Kinase (FAK) (Caccavari et al., 2010). These downstream signaling pathways have the ability to modulate MMP and tissue inhibitors of metalloproteases (TIMP) that can adjust ECM synthesis and degradation (Caccavari et al., 2010). It is now evident that there exists a complex feedback mechanism between cancer cells and ECM that influences the fate of the tumor (Blauboer et al., 2014). Interference of the cancer-promoting ECM-cell interactions could immobilize cancer cells and inhibit the deadly metastatic form of lung cancer, thus improving patient survival rates. Therefore, more basic research is needed to understand how lung ECM affects lung cancer cells, and vice versa.

To improve the reconstruction of the *in vivo* TME *in vitro*, cancer researchers are growing cancer cells in three-dimensions, rather than the commonly used 2D- tissue cultured-treated dishes. Matrigel™, a basement membrane extract (BME) harvested from EHS mouse sarcoma cells, is a common semi-solid medium used to recapitulate the *in vivo* environment that affects cell morphology, proliferation, migration, and drug response (Benton et al., 2011). There also exists synthetic cell scaffolds such as hyaluronic acid, gelatin, or polyethylene, but these synthetic gels and substrates lack the rich diversity of the stroma-associated ECM that is present *in vivo*. Also, both Matrigel™ and synthetic matrices lack the native fibrillar architecture that may be required for the complex mechanisms of the cancer cell-ECM interaction. Also, BME, like Matrigel™, are not tissue specific, which could make interpreting effects difficult because different ECM protein compositions could have different effects on the biology of the tissue-specific cancer cell. Although suspending cancer cells in 3D protein gels better mimic the native tumor, it does not accurately recapitulate many tumors of the epithelium, including lung adenocarcinoma. Therefore, it is necessary to implement a better tissue-culture technique that incorporates signals from extracellular matrix proteins and structural properties.

Fibroblast-derived ECM could be a simple, cost-effective cell culture technique that recapitulates the ECM-lung cancer cell interactions that have been shown to play a vital role in the progression of cancer. Fibroblasts are the principal contributors to ECM synthesis in most types of tissue, including the lung epithelium. Therefore, an ECM synthesized by fibroblast that retains its structural architecture will more accurately represent the native tumor environment. Edna Cukierman *et al* have published numerous articles describing techniques to harvest mouse fibroblast-derived matrices, and even characterized how these matrices affect the morphology,

growth, and drug response of numerous epithelial tumors cell lines, but no lung adenocarcinoma cells with human lung matrix were tested (Serebriiskii et al., 2008). Further, the same group demonstrated that overexpression of Fibroblast-activated Protein (FAP) in 3T3 cells induced matrix fiber alignment, which increased the growth invasiveness of pancreatic cancer cells (Lee et al., 2011). Still, little is known about how human lung fibroblast-derived matrices alter the morphology, growth, invasiveness, and biochemical properties of lung cancer cell lines.

In this work, we sought to better understand the effect of tissue-specific, human lung fibroblast-derived matrices on two lung adenocarcinoma cell lines, A549, H358, and HPL1Ds in ways previously untested with fibroblast-derived ECM. Techniques such as cell growth assays, drug response, microscopy, western blot, RNA analysis, and migration studies were used to characterize human fibroblast-derived ECM and how they affect lung cancer cell lines. The Objective of this Masters of Engineering project is to characterize the effect of human fibroblast-derived extracellular matrix on lung cancer cell phenotypes. Also, it is a long-term goal of the student to understand how modulation of the fibroblast-derived extracellular matrix alters human lung cancers.

II. METHODS

A. Cell Culture

Three Human fibroblast cell lines were used in this study, IMR-90, WI-38, and HDF, were all purchased from ATCC. IMR-90 Cells are human lung fibroblasts harvested from a female of 16-week gestation. WI-38 cells are also lung fibroblasts harvested from a 12-week old fetus. HDFs are human dermal fibroblast procured from the Patricia Soucy Laboratory. IMR90 and

HDF cells were both cultured in RPMI 1640 (Hyclone SH30027.01) supplemented with 10% FBS (Hyclone SH30070.03), 1% penicillin-streptomycin (Hyclone SV30030), and 1% L-glutamine (Hyclone SH30034.01). WI-38 cells were cultured in MEM media (Gibco 10370-021) supplemented with 10% FBS, 1% penicillin-streptomycin, and 1% L-Glutamine.

Three-epithelial cell lines, A549, H358, and HPL1D, purchased from ATCC, were used to seed onto Fibroblast-derived matrix. A549 and H358 cells are both lung adenocarcinoma cells harvested in 1972, and 1981, respectively. HPL1D cells are normal peripheral lung epithelial cells and were used to compare the cancerous phenotype of A549 and H358 cell lines to normal, steady-state lung epithelial cells. However, HPL1D cells have been shown to form soft-agar colonies in our lab, suggesting they are transformed. All epithelial cells were cultured in RPMI 1640 supplemented with 10% FBS, 1% Glutamine, and 1% penicillin-streptomycin.

B. Extracellular Matrix Production and Decellularization

ECM production and harvesting protocols were adapted from Cukierman *et al* and Soucy *et al* (Soucy et al., 2009). Fibroblasts are seeded at confluence, and cultured for 8 days while changing the media every two days. HDF cells are a little larger than the lung fibroblast cell lines so they are seeded at a lower density than WI-38 and IMR-90 cells. For example, in a 6 well dish, 300,000 IMR90 or WI-38 cells are seeded, but for the same size dish, 270,000 HDF cells were seeded. After 8 days of culture, cells are washed once with PBS (Hyclone SH30028.02), and then incubated at room temperature for 2 minutes in PBS with 50 mM concentration of ammonium Hydroxide (NH₄OH) (Fischer-A669-500) with .05% Triton X-100 (Sigma T8787) solution. Ammonium hydroxide is used to destroy the fibroblasts by creating a hypotonic

solution that lyses the cells. Triton X-100 is a detergent that also aides in cell lysis while leaving the produced matrix undisturbed. Cells are constantly observed under the microscope at 10X to confirm proper removal of all fibroblast debris. After decellularization, the matrices are washed 3 times with PBS. To prevent unwanted fibroblast DNA, which could affect downstream applications, matrices are incubated at for one hour at 37°C in a 20U/mL DNase 1(Thermo Scientific EN0525) concentration in sterile H2O (CellGro 25-055-CV). After DNase 1 treatment, matrices are washed 3 times with PBS and either stored at 4°C, or used immediately (Figure 2).

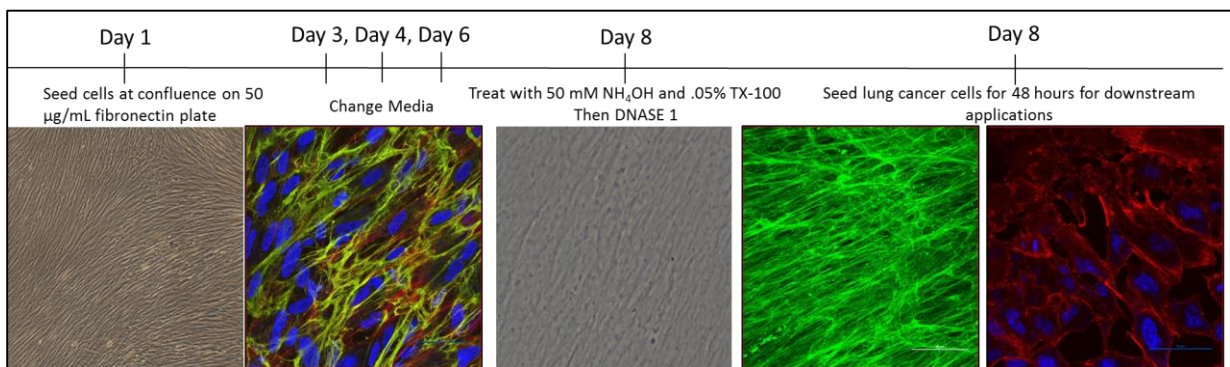


Figure 2- Schematic of Fibroblast-derived ECM protocol

C. Microscopy and Fluorescent Microscopy

For light microscopy, a Zeiss AX10 inverted microscope was used at 10X and 40X. For Immunofluorescence, a Nikon A1R camera mounted on a Zeiss Confocal microscope was used. For immunofluorescence, Matrices and/or cells were fixed with 4% formaldehyde in PBS for 20 minutes at room temperature then washed 3 times with PBS for 5 minutes each. For ECM staining, a succinimidyl ester (NHS ester) conjugated to an Alexa Fluor 488 dye was incubated at a concentration of 10 µg/mL for 30 minutes at 37°C, then washed 3 times with PBS for 5

minutes. The NHS ester binds to any proteins, thus staining the entire extracellular matrix. To ensure no leftover fibroblast DNA is present, DAPI was also added at a concentration of 1:1000 in PBS for 10 minutes at room temperature. To image the cytoskeleton of epithelial cells, Phalloidin was added at a concentration of 1:1000, while incubating with DAPI for 10 minutes at room temperature. After staining, the cells were imaged. The ECM stain fluoresces at 488 nm, while Phalloidin and DAPI fluoresce at 600 nm and 358 nm, respectfully.

D. SDS PAGE and Western Blot

Cells were lysed in CHAPS buffer (1% CHAPS in 150 mM NaCl, 50 mM TRIS pH 7.5, 50 mM EDTA) containing protease and phosphatase inhibitors for 20 minutes on ice. Lysates were centrifuged at 15,000 X G for 8 minutes at 4°C to pellet the insoluble material, which includes the extracellular matrix proteins. Protein concentration was then quantified using a bicinchoninic acid (BCA) assay (Thermo Scientific #23228, #1859078). Briefly, the BCA assay is a 2-step reaction where peptide bonds reduce the copper sulfate in reagent-A, next the reduced copper is chelated by the bicinchoninic acid from reagent-B, forming a purple color, which can be measured at 562 nm by absorption on a spectrophotometer. The amount of reduced copper ions and therefore the absorbance at 562 nm is directly proportional to the amount of protein in the sample. 30 µg of total protein was loaded on a 4-12% bis-tris polyacrylamide gel (Novex). 150 volts was applied across the gel for 60 minutes at room temperature while the gel was submerged in 1X running buffer (Novex NP0002) in water. After protein migration, gels were either stained with colloidal blue (Invitrogen LC6025) for 4 hours to stain nonspecific proteins, or

immunoblotting was performed. For immunoblotting, protein was transferred to a PVDF membrane by applying a voltage of 70 V for 60 minutes in 1X transfer buffer (Novex NP0006) diluted with water and methanol. After protein transfer, membranes were washed for 15 minutes in 5% milk in a mixture of Tris-buffered saline and Tween 20 (TBS-T) to block non-specific binding by the antibody. Primary antibodies specific to human GAPDH, AKT, phosphorylated AKT, p38, phosphorylated p38, p44/42, phosphorylated p44/42, E-cadherin, Vimentin, mTOR, phosphorylated mTOR, and CDC42 (Cell Signaling™) were added at a concentration of 1:10000 in 5% milk and incubated with the membranes overnight at 4°C. After primary antibody incubation, membranes were washed 3 times in 5% milk for 5 minutes. Secondary antibody, conjugated to HRP, specific to target the primary rabbit antibody, was incubated with the membranes in 1.25% milk for 30 minutes at room temperature then washed 2 times for 5 minutes in 1.25% milk. Membranes were then incubated for 20 minutes in TBS-T before adding HRP substrate (Thermo Scientific #34095) for detection by luminescence captured on a photographic film. The amount of luminescence produced by the HRP enzymatic reaction is proportional to the amount of protein tagged by the antibody.

E. Cell Growth

To determine the effect of cell-derived matrices on cell growth and proliferation, two assays were employed. To determine cell number, 30,000 cells were plated in a 12 well dish with a fibronectin coat, or fibroblast-derived ECM. Every 24 hours for 4 days, cells were washed in PBS, trypsinized in Trypsin (Hyclone SH30042.01) for 3 minutes at 37°C, centrifuged at 1200 RPM for 3 minutes and suspended in 1 mL media for counting with Trypan Blue™ (Invitrogen

T10282) in a Countess automated cell counter m (Invitrogen). Trypan blue is a dye used to stain dead cells, which does not stain live cells.

For proliferation, 2,000 cells were seeded into a 96 well-plate, and Alamar Blue™ (Invitrogen 612130) was added each day for 4 days to determine proliferation. Plates were incubated at 37°C until a color change was observed, then read on a spectrophotometer using an excitation of 560 nm and emission of 590 nm. Briefly, Alamar Blue™ is resazurin and enters the cell and gets reduced by cells with an active metabolism. The reaction converts a blue resazurin to a purple resorufin, a fluorescent molecule. Raw fluorescent data were averaged, with the blank reading subtracted, and then normalized to Fibronectin.

F. Trans-well Migration Assay

To determine if fibroblast-derived matrices affected how epithelial cells migrate, trans-well chamber (Fischer 353097) assays were used. Briefly, ECM was produced on the bottom side of a trans-well chamber (Figure 3A-C) and A549 cells were cultured on the top side (Figure 3D). After 24 and 48 hours, non-migrating cells were scraped off the membrane. Migrating cells were then fixed in 4% paraformaldehyde and stain with hemotoxylin and eosin (Protocol 122-911). Membranes were then mounted on a glass slide and migrated cells were counted using ImageJ software analysis. N=4.

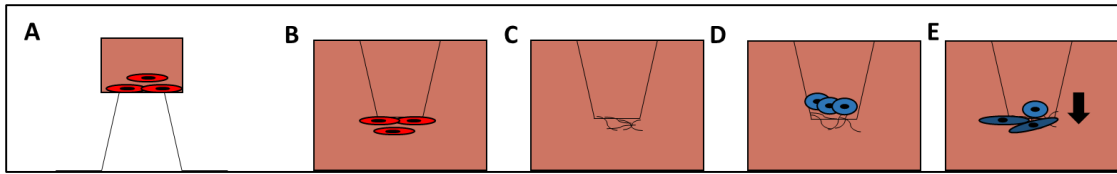


Figure 3- Depiction of Trans-well Assay. (3A) Seed 50,000 fibroblasts on membrane and allow to adhere. (3B) Properly insert chamber 28 hours seeding. (3C) After 8 days, decellularize. (3D) seed 15,000 A549 Cells in upper portion of chamber. (3E) allow cells to migrate for 24 hours.

G. Microarray

RNA was harvested using E.Z.N.A Total RNA Kit (Omega) and analyzed on a nanodrop for quality control. RNA must have a 260/280 absorbance ratio between 1.8 and 1.9, and a 260/230 absorbance ratio between 1.9 and 2.2 to ensure minimal protein contamination then treated with DNase to remove DNA contamination. RNA samples were then converted to cDNA, labeled, and hybridized on an Affymetrix PrimeView Human Array Chip with the aid of the Genomics Core Facility and the assistance of the student. Raw intensity scores were imported into Partek Genomics Suite 6.6 (6.13.0731) and normalized on a gene level using the standard RMA algorithm for normalization and background correction. A 2-way ANOVA was set up and a step-up FDR corrected P-value was included for every P-value calculated. Only significant gene changes with a P-value of $<.05$ and a fold change greater than 1.5 were uploaded into MetaCore™ for pathway and gene ontology analysis.

H. Real Time PCR

New biological replicates were used to generate RNA as previously described for the microarray. Complementary DNA was generated using the RNA to cDNA kit Applied Biosystems). This kit uses thymidine-rich primers that bind to the poly-A tail of mRNA and allow a polymerase to synthesize the complete complimentary strand. Oligonucleotide sequences are as follows:

Table I
List of Primer sequences for Quantitative PCR

Gene	Sense	Antisense
LGALS	GGTCAACCCTGAAGATCACAG	GTCCAATGAGTTGCAGACAATG
PSAT1	AAGGTGTGCTGACTATGTGG	TTGAGGTTCCAGGTGCTTG
ASNS	AGGAGAGTGAGAGGCTTCTG	GGTGGCAGAGACAAGTAATAGG
BCAT1	AATCCCAAGTATGTAAGAGCCTG	AAGAGATGAGCCGTAATTCCC
IL-8	ATACTCCAAACCTTTCCACCC	TCTGCACCCAGTTTTTCCTTG
STAT4	CCTGAAAACCCTCTGAAGTACC	ACCTTTGTCACCCCTTTCTG
C3	AACTACATCACAGAGCTGCG	AAGTCCTCAACGTTCCACAG
C1S	TTTGTAGATGTCCCTTGTAGCC	AATCTCCCCAATCAGTGCAG
MMP7	TTCCAAAGTGGTCACCTACAG	AGTTCCCATACAACTTTCCTG
RND1	ATGTAAGCTCGTTCTGGTGC	CTCTGTTCTCTGTCTCCAAAC
GAPDH	TGCACCACCAACTGCTTAGC	GGCATGGACTGTGGTCATGAG

Primers were designed to be between 18-22 base pairs long, have a 40%-60% GC content, and amplify a product between 60-120 base-pairs. Primer specificity was validated to only amplify one product between at 60°C. Reaction efficiency was also validated to ensure that fluorescence intensity doubled each cycle using a serial dilution of a known cDNA template. Reactions were shown to be efficient between 10 ng and 50 ng concentrations of DNA. To qualify, efficiency must be within 85% and 105% with an R^2 value greater than .98. This quality control confirms that no matter the differences in original starting material between the wells, comparison of threshold values are valid because the samples are within the linear range. To calculate relative fold change compared to a fibronectin-coated dish, the $2^{-\Delta\Delta C_T}$ method. Equations shown below:

$$\Delta\Delta CT = (\Delta CT(ECM) - \Delta CT(FN))$$

$$\Delta CT = \text{Average CT (Target gene)} - \text{Average CT (GAPDH)}.$$

C_T is the PCR cycle number where the fluorescent threshold value is reached. All PCR reactions were analyzed on a biorad CX96.

I. TIME-LAPSE MICROSCOPY

150,000 A549 cells were seeded onto either ECM-coated or Fibronectin-coated 35mm dishes with a coverslip insert for microscopy. 24 hours after cell seeding, dishes were inserted into a live-cell chamber for 24 hours and pictures were taken at 3 different fields every 10 minutes, until 24 hours have past. Cell migration was calculated by tracking 4-8 cells per field throughout the time lapse. Cell tracking was achieved by the MtrackerJ plugin, available for ImageJ

(Meijering et al., 2012). Average point by point velocity for all lines were calculated within MtrackerJ and reported with SEM. Directionality, a measure to determine linearity of migration, was calculated by D/d (Figure 13B).

J. Cell Circularity

After confocal microscopy with Phalloidin and DAPI, images with scale bars were imported in ImageJ and converted from RGB to 32-bit and a threshold was set to create maximum contrast between cell borders and empty space. Built-in functions of ImageJ allowed for easy image segmentation for area and perimeter analysis. Circularity is a value between 0-1 with 1 being a perfect circle (Pasqualato et al., 2013).

$$Circularity = 4\pi \times (Area \div Perimeter^2)$$

K. Subcutaneous Tumor Formation

5×10^6 cells in 500 μ L were mixed with either 500 μ L PBS or 500 μ L of PBS with 615 μ g scraped WI38 derived ECM. 100 μ L was then injected in the subcutaneous tissue of NRG mice. Each mouse received two injections: 100 μ L of A549 cells with ECM on the left flank and 100 μ L of A549 with PBS on the right flank. Tumors were then allowed to form and after 60 days, mice were euthanized by CO₂ and tumors were excised and fixed in 10% buffered formaldehyde. Fixed tissues were sent to HistoServe™ for paraffin embedding, sectioning, and staining. Tumor volume is measured by:

$$Tumor\ Volume = \frac{1}{2}(Lenth\ x\ Width)^2$$

Where length is the greatest longitudinal direction and width is the greatest transverse direction (Jensen et al., 2008).

L. Dose Response of Chemotherapeutics

3,000 A549, H358, and HPL1D cells were seeded in on a 96 well-plate. After 24 hours, Cisplatin and Gemcitabine were added into the media and then incubated for 48 hours before measuring cell growth by Alamar Blue™. N=3.

M. Statistical Analysis

A two-way ANOVA with a Tukey's Test for multiple comparisons were used to determine statistical significance. For *in vivo* data, a unpaired t-test was performed. *, P-value<.05

III. Results and Discussion

A. Fibroblasts and Their Derived Extracellular Matrix

In order to derive accurate conclusions, multiple fibroblast cell lines were procured to produce a variety of ECM from multiple tissue types. Two human fetal lung fibroblasts cell lines, WI38 and IMR-90, and one human dermal fibroblast cell line, HDF, were used. All fibroblast cell lines exhibited the same spindle-shaped morphological features (Figure 4Ai).

Interestingly, all three fibroblast cell lines aligned themselves in a linear pattern, and produced a similar ECM, which also appeared highly linearized, suggesting there is dynamic remodeling after cell seeding (Figure 4Aii and 4Aiii). This linearized ECM could be important for maximizing physiologic relevance to the human disease *in vitro*, which is not available with solubilized ECM, such as Matrigel™. The resultant fibroblast-derived ECM was solubilized in 5M Guanidine-HCL and protein content was measured using a BCA assay (figure 4C). All fibroblast-derived matrix produced similar amounts of ECM per cm², except HDF produces slightly more (P=0.45). WI38 ECM averaged 32.06 µg/cm² +/- 3.961, IMR-90 ECM averaged 32.02 +/- 3.299 µg/cm², and HDF ECM averaged 34.016 +/- 3.990 µg/cm². Results were averaged from three wells of fibroblast-derived matrix solubilized in the same volume. Interestingly, HDF produce an ECM that has significantly (p=0.045) more protein than the other fibroblast-derived ECM (Figure 4B), thus providing suggesting that there are differences in ECM production between tissue-specific fibroblasts. After solubilizing, ECM protein was analyzed by SDS-PAGE and total protein was stained by colloidal blue (Figure 4D). Interestingly, both WI38 and IMR-90 derived matrices have more intense Collagen bands (140 kDa- 175 kDa) compared to HDF. However, HDF-derived ECM has a greater presence of a fibronectin band (260 kDa-280 kDa) compared to WI38 and IMR90 matrices. Collagen and Fibronectin bands are easily identified because they have been previously shown to be the prominent constituent of WI38 derived ECM (Soucy et al., 2009). Also, there appears to be more intermediate protein bands that range between 175 kDa and 250 kDa that are more prominent in HDF derived ECM than WI38 and IMR90 ECM. However, it is not possible to conclusively determine the identity these proteins. It is interesting that there are differences in ECM protein expression between cell lines from two

distinct tissue sources. These subtle differences in protein expression might alter the phenotype of lung cancer cell lines, thus potentially affecting the outcome of the human disease.

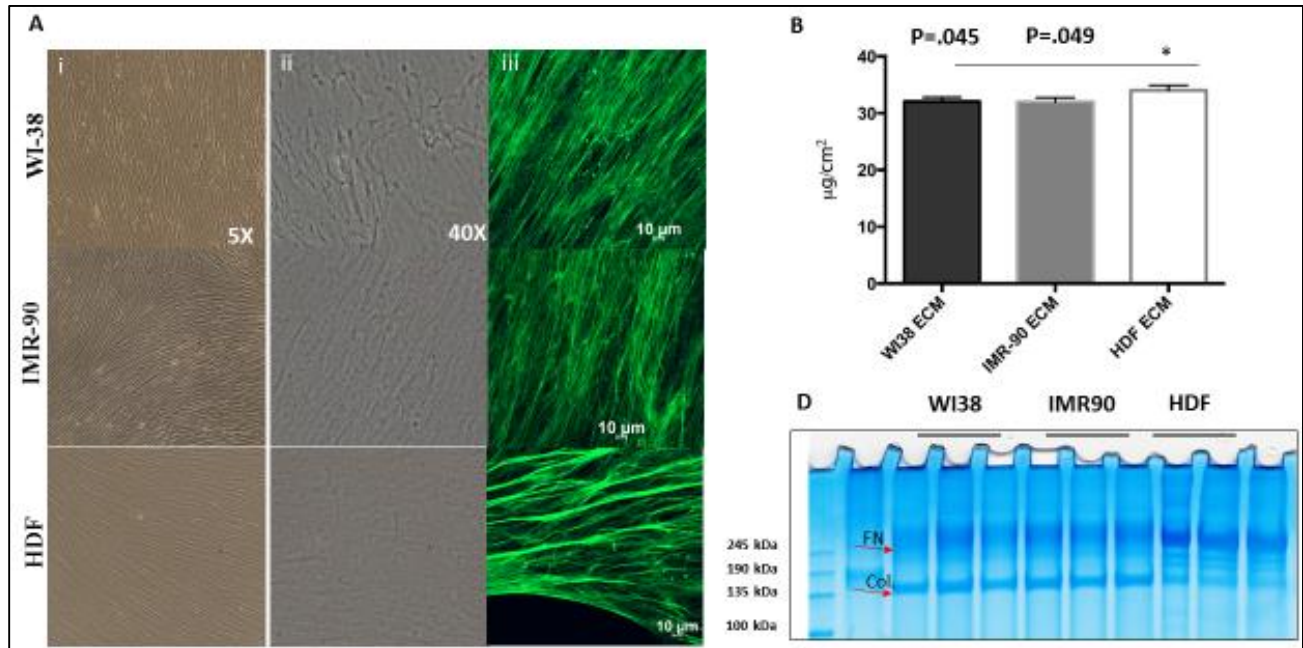


Figure 4-Fibroblasts and their ECM. (Ai) 5X phase contrast microscopy of fibroblasts. (Aii) 40X phase contrast microscopy of decellularized fibroblast-derived ECM. (Aiii) Confocal microscopy of fibroblast-derived ECM with NHS-ester probe. (B) ECM Quantified protein analysis (μg/cm²), N=3 *p<.05 (TABLE VII). (D) SDS PAGE and Colloidal Blue stain of fibroblast-derived ECM. N=3.

B. Fibroblast-derived Matrices Alter Lung Cancer Cell Morphology

The first experiment that was conducted once fibroblast-derived matrices were successfully produced was to determine how cell morphology changes when lung cancer cells

were grown on human lung-specific matrix, human dermal matrix, or a normal fibronectin-coated tissue culture treated dish. Fibronectin-coated dishes were used as a control because it simulates both the traditional 2D cell culture environment as well as having a natural ECM component. After fibroblast extraction and ECM purification, A549, H358, and HPL1D cell lines were seeded onto the fibroblast-derived matrices or FN-coated dish, and then imaged 48 hours later to determine morphology. Observation by phase-contrast microscopy (Figure 5Ai) revealed marked differences in cell shape, especially A549 cells. H358 and HPL1D cells exhibited a slight spindle-shaped morphology. A549 cells showed the biggest change in morphology, thus were analyzed further quantify the change (figure 5B). Phalloidin, a chemical known to have a high specificity for filamentous actin, was used to visualize the cytoskeleton of the cell by confocal microscopy. Simple geometric calculations such as area and perimeter were used to judge cell circularity (Figure 5C). All three fibroblast-derived matrices significantly ($p < .05$, $N=10$) pressured A549 cells to form a more elliptical shape, instead of a more circular shape (Circularity = 1), which is exhibited more on fibronectin (Figure 5C). A549 cells on fibronectin have an average circularity value of $.38 \pm .04$, while the same cells on WI38, IMR-90, and HDF ECM have average circularity values of $.18 \pm .034$ ($p=.004$), $.18 \pm .038$ ($p=.0023$), and $.12 \pm .019$ ($p=.0002$), respectively. Cells settled between matrix fibers and elongated according to the direction of the fibers. However, the type of ECM did not have differential effects on cell morphology, suggesting that the type and content of the ECM does not induce morphological change in lung cancer cell lines. Changes in cell morphology can induce drastic alterations in cell processes such as proliferation and migration, thus influencing the fate of the cancer cells. In fact, it has been previously reported that cell shape can influence the metastatic capabilities of

some cancers (Yin et al., 2013). Cell elongation may contribute to increased cell migration. It is interesting that A549 cells have the most drastic change in morphology, suggesting that this cell line may have altered growth and migration properties, compared to H358 and HPL1D cells.

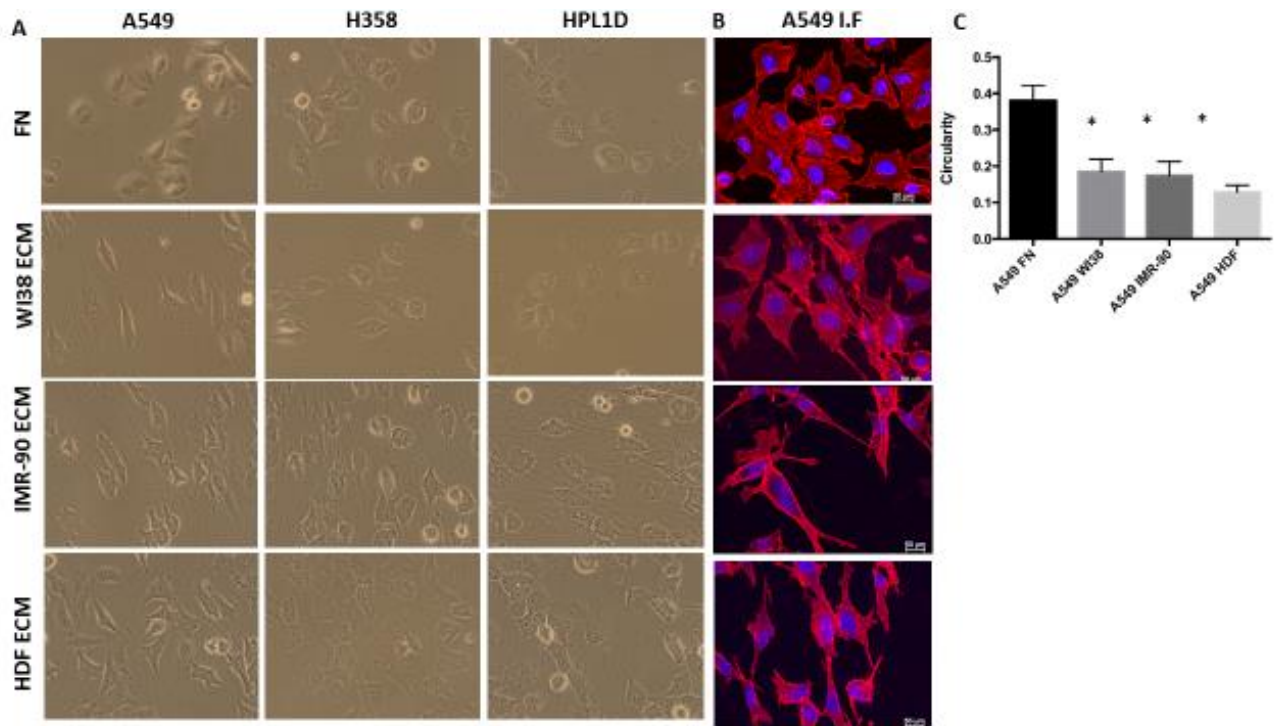


Figure 5- Fibroblast-derived ECM alter Lung Cancer Cell Line Morphology. (A) Phase contrast microscopy photos of A549, H358, and HPL1D cells on FN, WI38 ECM, IMR-90, and HDF ECM. (B) A549 cells on WI38 ECM stained with Phalloidin and DAPI. (C) A549 circularity on FN and WI38 ECM. N=10. *, $p < .05$ (TABLE X)

C. Fibroblast-derived Matrices Alter Lung cancer Cell Growth

Next, cell growth alterations due to interaction with the various fibroblast-derived matrices were examined. Even without employing assays to measure cell number and proliferation, it is clear by microscopy that human fibroblast-derived matrices affect cell growth, because there are

simply fewer cells to observe after several doubling times. To test the hypothesis that lung cancer cells proliferate less when cultured on fibroblast-derived ECM, Alamar Blue™ and cell-counting assays were used. A549, H358, and HPL1D were seeded onto 12-well plates for Trypan blue™ cell counting, or 96-well plates for an Alamar Blue™ assay, and recorded measurements for four consecutive days. Trypan blue™ is a dye that stains cells with a disrupted cell membrane, thus it only accumulates in dead cells. Alternately, Alamar Blue™ is a blue dye that gets reduced to a purple fluorescent molecule by oxidative reaction in the electron transport chain. Thus Alamar Blue™ reduction is directly proportional to cellular metabolism. It is clear that both cell density number (Figure 6A) and subsequent Alamar Blue™ (Figure 6B) metabolism were attenuated when all cell lines were cultured on all matrices. Cukierman et al. showed that various human epithelial cancers such as breast and colon cancer cell lines grow slower when cultured on mouse fibroblast-derived ECM (Serebriiskii et al., 2008). It is the belief in the field that an increased extracellular matrix presence should enhance cancerous properties. However, this data suggests that fibroblast-derived ECM inhibit the growth of lung cancer cell lines. Interestingly, a recent publication has determined that high mechanical strain induces cell cycle entry mediated by E-cadherin (Benham-Pyle et al., 2015). Therefore, this decreased growth could be due to the altered force the cells experience when cultured on fibroblast-derived matrix. Alternately, perhaps lung cancer cells lines grow slower on fibroblast-derived ECM because the ECM creates a barrier between neighboring cells, thus eliminating cell-cell interactions. Nevertheless, more studies are needed that determine exactly how fibroblast-derived ECM slow down the growth of lung cancer cell lines, compared to the same cells on a fibronectin-coated tissue culture dish.

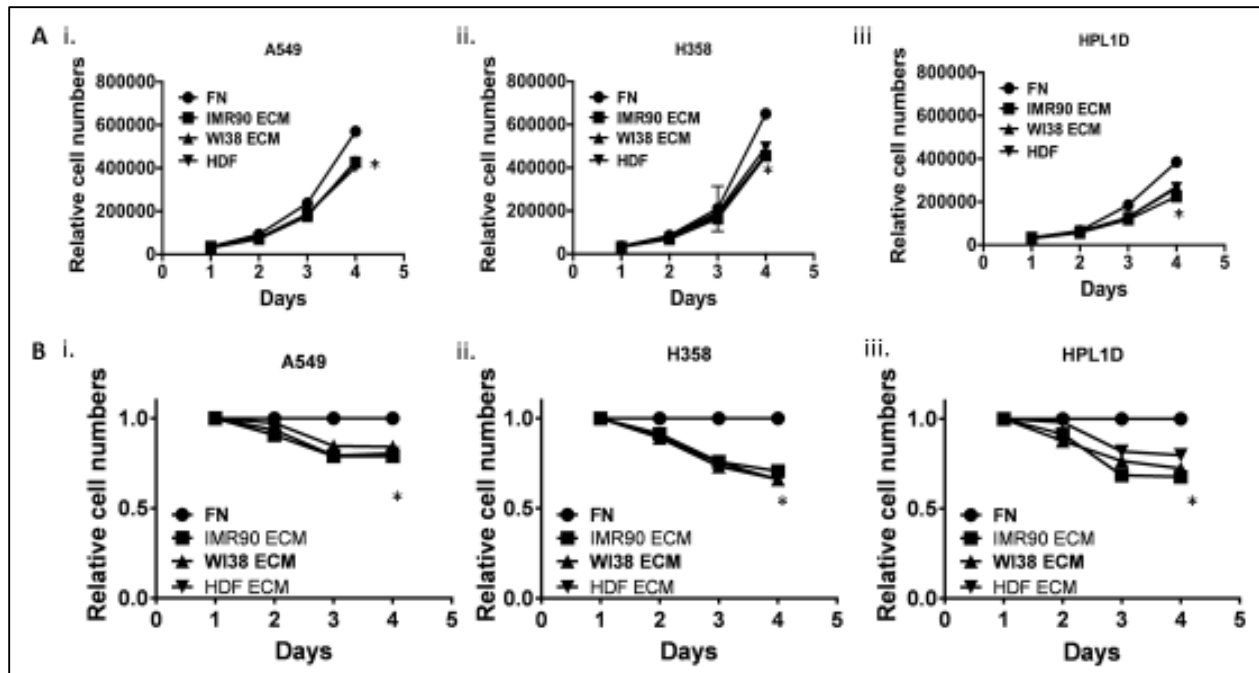


Figure 6- Fibroblast-derived ECM alter Lung Cancer Cell Growth. (A) Manual cell counting. *, $P < .05$. (Ai) A549 on all three fibroblast-derived ECM (TABLE XI). (Aii) H358 on all three fibroblast-derived ECM (TABLE XII). (Aiii) HPL1D on all fibroblast-derived ECM (TABLE XIII). B: Alamar Blue™*, $P < .05$. (Bi) A549 on all three fibroblast-derived ECM. (Bii) H358 on all three fibroblast-derived ECM. (Biii) HPL1D on all fibroblast-derived ECM. N=3. *, P -value $< .05$ (TABLE XIV-XVI)

D. Fibroblast-Derived Matrices alter functional protein levels of lung cancer cell lines

To perhaps understand which growth-related signaling pathways are attenuated when A549 cells interact with various fibroblast-derived ECM, western blot analysis was performed. Western blotting is a common analytical technique that uses antibodies to detect specific protein levels in a cell lysate. Here, A549 cells cultured on FN, WI38 ECM, IMR-90 ECM, and HDF

ECM were lysed, spilling out soluble protein. Antibodies specific for proteins involved in MAPK signaling, Epithelial-to-Mesenchymal Transition (EMT) signaling, as well other proteins known to be involved in cellular proliferation were utilized to help understand why A549 cells have restricted growth when cultured on all fibroblast-derived matrices. Interestingly, many proteins and phosphorylated proteins are decreased at the protein level, as detected by western blot (Figure 7). Phosphorylated MAPK family members such as P38 and P44/42 are decreased when A549 cells are cultured on all fibroblast-derived ECM (Figure 7B). MAPK initiation is known to activate many downstream pathways that affect cell cycle entry and cell growth (Segar et al., 1995). Similarly, AKT, Cyclic-AMP Response Element Binding (CREB) protein, and Mammalian-Target of Rapamycin (MTOR), which are all together involved in several cascades that regulate cell cycle, are decreased when A549 cells are cultured on all fibroblast-derived matrices (Figure 7A and 7B). Also, CDC42, a known cell cycle regulator, and phosphorylated SRC kinase, a known oncogene in lung cancers, are decreased on fibroblast-derived ECM. These data suggest possible effector molecules that are disrupted when A549 cells are cultured on fibroblast-derived ECM, thus providing key insight into the effect of the ECM on lung cancer cell line growth.

As previously stated, metastatic lung cancer is the deadliest and most aggressive form of the disease. EMT is a process initiated by cells in the primary tumor in order to migrate and become metastatic. Biomarkers of EMT are the loss of E-cadherin, and the gain on Vimentin expression in epithelial cancers (Xiao et al., 2010). It is hypothesized that ECM can influence EMT biomarkers, thus inducing cells to undergo EMT. Interestingly, E-cadherin expression is greatly decreased and Vimentin expression is a slightly increased when A549 cells are cultured

on all fibroblast-derived matrices. These data suggest that fibroblast-derived ECM alter EMT markers of A549 cells, which could induce a more migratory phenotype, thus increasing likelihood of metastasis.

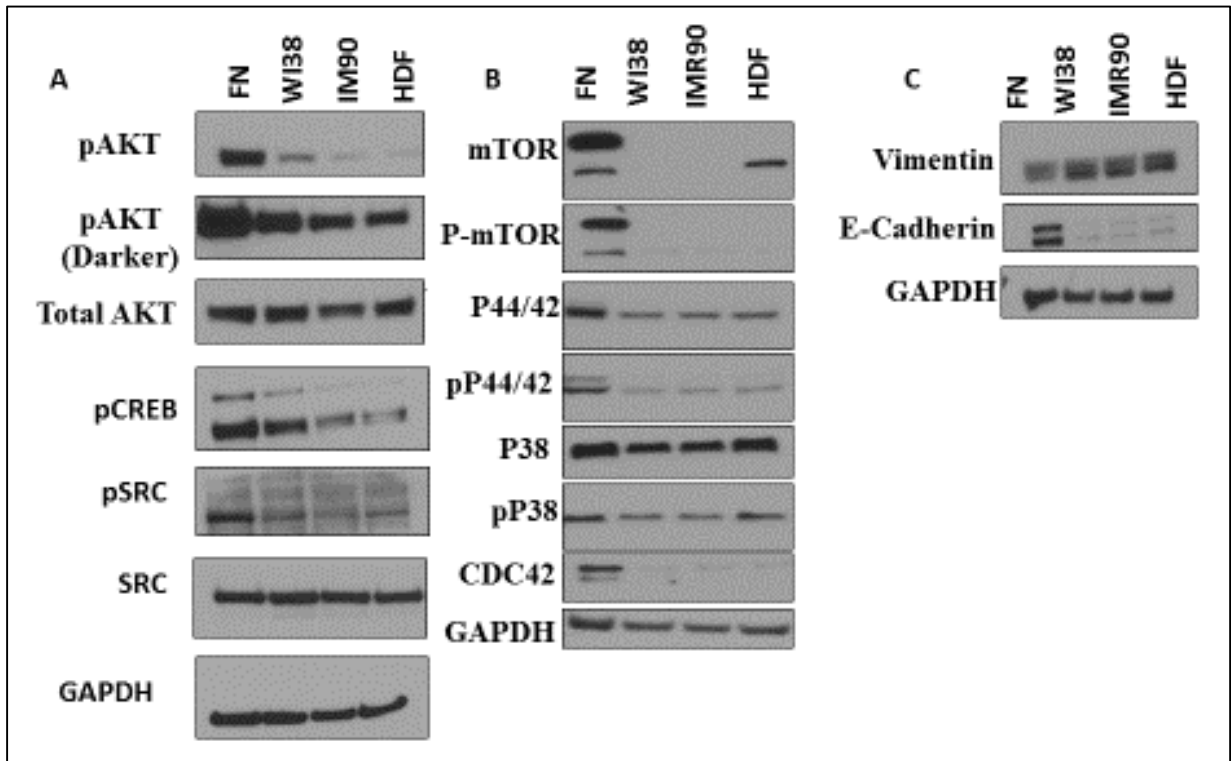


Figure 7- Fibroblast-derived ECM alters Protein Levels of A549 Cells. (A) Western blot of AKT, pAKT, pCREB, pSRC, total SRC, and GAPDH. (B) Western blot of mTOR, Phosphorylated-mTOR, P44/42, Phosphorylated P44/42, p38, Phosphorylated p38, CDC42, and GAPDH. (C) Western blot of Vimentin, E-cadherin, and GAPDH.

E. Fibroblast Derived Matrices Do not alter Drug Resistance of Lung Cancer Cell Lines

Cancer cells grown in three dimensions, such as in Matrigel™, or in a spheroid-forming medium, experience epigenetic changes that alter cancer phenotype. Three dimensional cell culture gives the cell line a TME that influences its response to drug treatment. Similarly, the human TME helps cancer cells evade drug-induced cell death. It was hypothesized that because human lung fibroblast-derived ECM alter lung cancer cell line morphology and growth rates, then it was reasonable to assume drug response would also change. To test this, lung cancer cells cultured on human lung fibroblast-derived ECM were treated with various chemotherapeutic agents often used in the clinic to treat lung cancer patients. Interestingly, culturing A549, H358, and HPL1D cells on WI38 ECM did not alter the dose response of Cisplatin and Gemcitabine (Figure 8A-C), which are common first-line therapies for lung cancer patients. Perhaps these drugs are so cytotoxic that the ECM has little or no effect. However, this model could still be beneficial for future drug studies. For instance, if there is an alteration in a certain drug response on fibroblast-derived ECM, it could provide evidence that the certain therapy is affected by the ECM, which could serve as an indicator of its efficacy in an *in vivo* setting.

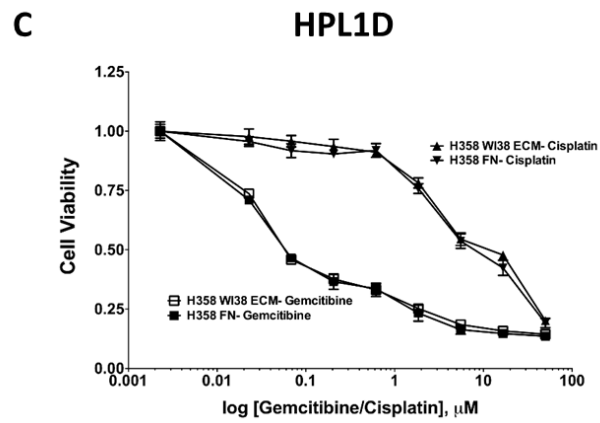
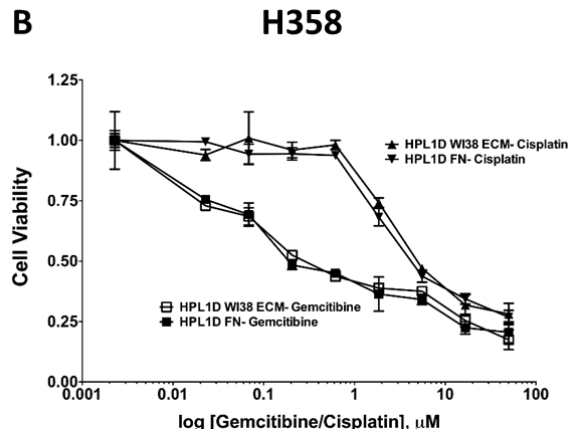
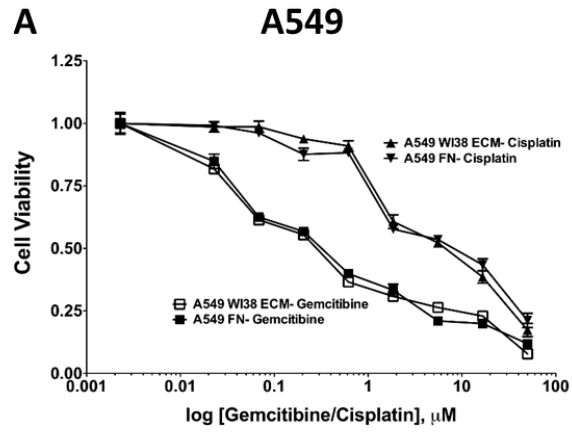


Figure 8- Fibroblast-derived ECM do does not alter drug response to Gemcitabine and Cisplatin.

(A) A549 on WI38 ECM. (B) H358 on WI38 ECM. (C) HPL1D on WI38 ECM. N=3. Not significant ($p > .999$).

F. Fibroblast Derived Matrices Protect Lung Cancer Cells from Serum Deprivation

Although the drug response of lung cancer cell lines is not affected by human fibroblast-derived ECM, other experiments to test the ability of lung cancer cell lines to survive stressful environments when cultured on ECM, such as serum withdrawal, were explored. Serum deprivation induces cellular apoptosis by mitochondrial outer membrane permeabilization (MOMP) in colorectal cancer cells (Braun et al., 2011). Although many cancer cells evade apoptosis by inhibiting MOMP, many cancer cell lines cannot survive prolonged exposure to serum-free media. However, little is known about how the ECM affects serum dependability. It is hypothesized that proteins and growth factors present in the fibroblast-derived ECM can sustain lung cancer cells in an otherwise nutrient-free environment. To test the effects of serum depletion on cells on and off fibroblast-derived ECM, cells were seeded in complete media for 24 hours to allow cells to adhere, media was then removed and cells were washed with PBS, and replenished with serum-free media. Observation by phase contrast microscopy revealed that 48 hours after the removal of serum from the media, there are more A549 cells on ECM compared

to a fibronectin dish, suggesting that the ECM supports cell adhesion and perhaps growth in the absence of serum (Figure 9D). To determine the difference between cell growth on ECM and fibronectin, Alamar Blue™ metabolism was followed each day. All lung cancer cell lines cultured on both human lung and human dermal fibroblast-derived ECM showed an increase in relative viable cells compared to the same cell lines grown on fibronectin plates (Figure 9A, 9B, and 9C). A549 survive serum deprivation the best by Alamar Blue™ (figure 9A), until day 4. HPL1D and H358 cells also tolerated serum withdrawal well until day 4 (Figure 9A, 9B). However, The ECM was not able to sustain cell viability without serum past four days. Perhaps this is due to degradation of the ECM and its growth factors by the lung cancer cell lines. Also, A549 cells cultured on ECM did not change morphology when cultured in serum-free media (Figure 9B), whereas as A549 cells on Fibronectin-coated dish appeared rounded and unhealthy (Figure 9A). Upon western blot analysis, it was clear that fibroblast-derived ECM rescues serum deprivation-induced apoptosis by evading p21 expression and by upregulating BCL-xL expression (Figure 9F). p21, is a cyclin-dependent kinase inhibitor that is known to promote cell-cycle arrest and apoptosis in serum-free conditions (Braun et al., 2011). Interestingly, BCL-xL, an anti-apoptotic protein that protects cells against apoptosis, is slightly increased when A549 cells are cultured on fibroblast-derived ECM with and without serum (figure 9F), thus suggesting it plays a role in the survival of A549 cells in serum-free conditions. These results suggest that the ECM is capable of providing a cell-survival stimulus when the plethora of growth factors and nutrients available in FBS are absent. This observation might be important in understanding how lung adenocarcinoma cells survive in the inner-most core of a solid tumor, where nutrients, as well as oxygen, from the blood supply are scarce.

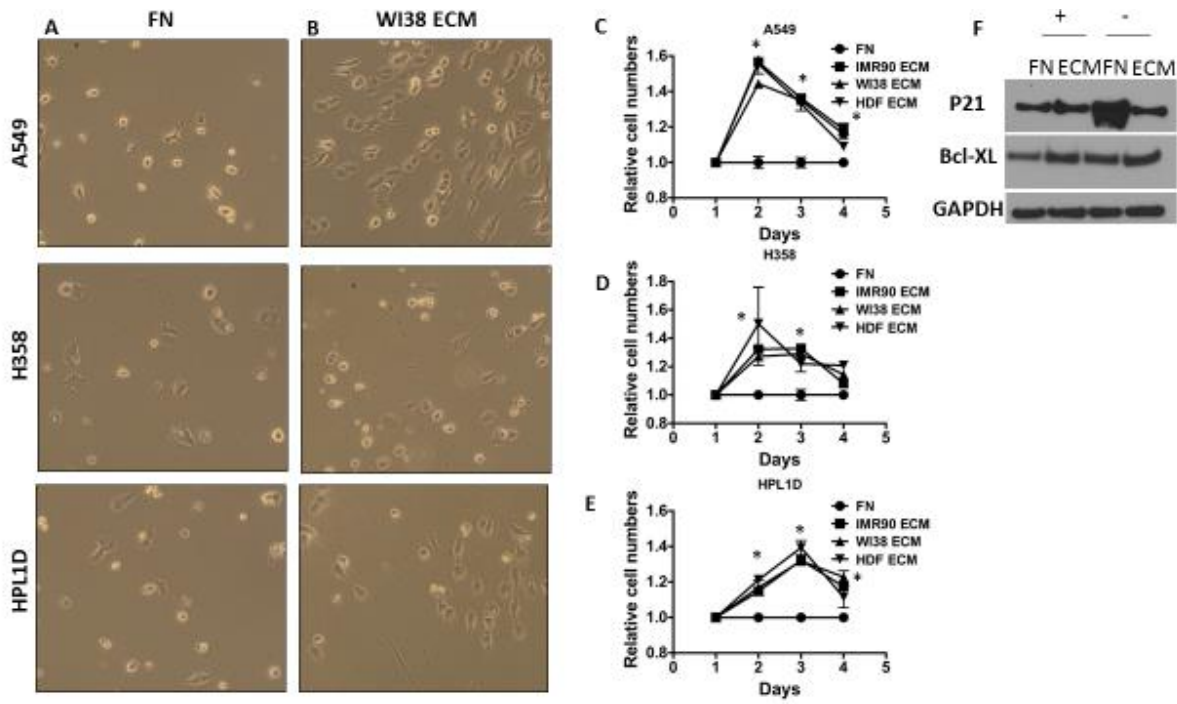


Figure 9- Fibroblast-derived ECM Protects Lung Cancer Cell Lines from Serum Deprivation. (A) A549, H358, and HPL1D cells on fibronectin in serum-free media for 48 hours. (B) A549, H358, and HPL1D cells on WI38 ECM in serum-free media for 48 hours. (C) Relative cell growth of A549 cells. (D) Relative cell growth of H358 cells. (E) Relative cell growth of HPL1D cells. (F) Western blot of serum deprived cells after 48 hours. *, $p < .05$ (TABLE IV-VI)

G. Fibroblast-Derived Matrices Change mRNA levels of Various Genes in Lung Cancer Cells

It is clear that human lung fibroblast-derived ECM alters many processes that influence behaviors like morphology, growth, and serum-deprivation survival. However, little is known about how fibroblast-derived ECM alters the regulation of gene transcription. Transcription

regulation is an extremely important process that influences cellular phenotypes. Therefore, an experiment was designed to determine the effect of gene transcription when A549 and H358 cells interact with human fibroblast-derived ECM. A microarray was used to determine the difference of messenger-RNA (mRNA) copy number when lung cancer cell lines were cultured on fibroblast-derived ECM. mRNA is the result of a transcribed gene, awaiting ribosome entry to be translated into an amino acid sequence, which is then folded to the proper native protein structure. Although mRNA quantity does not directly dictate functional protein quantity, it is directly connected with how much DNA is transcribed. To test the change in mRNA copy number in lung cancer on human fibroblast-derived ECM, A549 and H358 cell lines were cultured for 48 hours on WI-38 ECM, IMR-90 ECM, and Fibronectin-coated plastic. Briefly, RNA was extracted, converted to DNA, labeled, hybridized, and scanned for hybridization quantification. Data was then analyzed by Sabine Waigel and the UofL genomics core facility. A heat map was generated (Figure 10A) to show an unrefined interpretation of each gene probe intensity that was significant changed compared to a fibronectin-coated dish (P-value <.05 and fold change greater than 1.5). It is clear by examining the heat map that there are many gene probes that differ in expression induced by fibroblast-derived ECM. In total, 182 target sequences, which corresponded to 114 gene changes, were shared between both cells lines on both fibroblast-derived ECM compared to fibronectin. Affected Gene IDs were uploaded into Metacore™ for Gene Ontology (GO) analysis. Metacore™ GO process analysis identifies canonical and non-canonical pathways that are likely to be affected by a cluster of significantly changed genes. Top ranked potentially affected pathways include complement-mediated immunity, cytoskeletal remodeling, and ECM remodeling (Figure 10B). Interestingly, 87 of

these significant gene changes were down-regulated, and only 27 were upregulated. Many down-regulated genes coded for proteins that constitute the extracellular matrix, suggesting that when cells are in contact with adequate ECM, they decrease transcription of genes that code for ECM proteins such as fibrinogen alpha, fibrinogen beta, collagen type 4, collagen type5, and villin (Table 2). Also, MMP7, which is a protease that degrades Collagen and Fibronectin, is also down-regulated. The most down-regulated gene, LGALS2 belongs to a family of galectins, which can serve as interacting proteins with ECM proteins and can be deregulated in some cancers, including lung cancer (Liu et al., 2005; Buttery et al., 2004). Also, several genes that regulate the complement-mediated immunity pathway, such as C3, C1S, and bradykinin, are down-regulated when both lung cancer cell lines are cultured on WI38 and IMR90 ECM (Table 2). Complement-mediated immunity pathways have been shown to be activated in lung cancer (Pio et al., 2014). Interestingly, Galectin expression affects also complement activation (Pio et al., 2013). Up-regulated genes include NT5E, IL-8, KCNMA1, BCAT1, ASNS, stanniocalcin-2, and PSAT1 (Table 3). Interestingly, IL-8 is known to promote angiogenesis thus supporting previous research by Patricia Soucy, suggesting a role for ECM in angiogenesis (Luppi et al., 2007; Soucy et al., 2015; Soucy et al., 2009). NT5E, which is also known as ecto-5'-nucleotidase, has been shown to be upregulated in lung and breast cancers where it decreases patient survival rates by attenuating immune responses and promoting environments for vascularization and metastasis (Zhang et al., 2010; Jin et al., 2010). KCNMA1, a gene that reported four probes that had significant expression changes, codes for a protein that complexes to form potassium and calcium ion channels (Table 3). KCNMA1 ion channels have been shown to be more present in metastatic breast cancer cells but little is known of its role in lung cancer

(Khaitan et al., 2009). The ECM is a key player in the initiation of breast cancer cells into metastasis. This result that KCNMA1 mRNA is increased when lung cancer cells are cultured on fibroblast-derived ECM might suggest that ECM can induce KCNMA1 to play a pro-tumorigenic role in lung cancer cells. Similarly, stanniocalcin-2, a gene that had three probes reported from the microarray data, also plays a role calcium regulation, and is also overexpressed in lung cancer and impairs a poor prognosis (Na et al., 2015). ASNS, BCAT1, and PSAT1 all code for enzymes that regulate amino acid synthesis, thus directly affecting cell growth and proliferation. These microarray data reveals the complex mechanisms that the ECM can regulate within lung cancer cells. Thus providing strong evidence that Fibroblast-derived ECM can potentiate the phenotype of lung cancer cell lines.

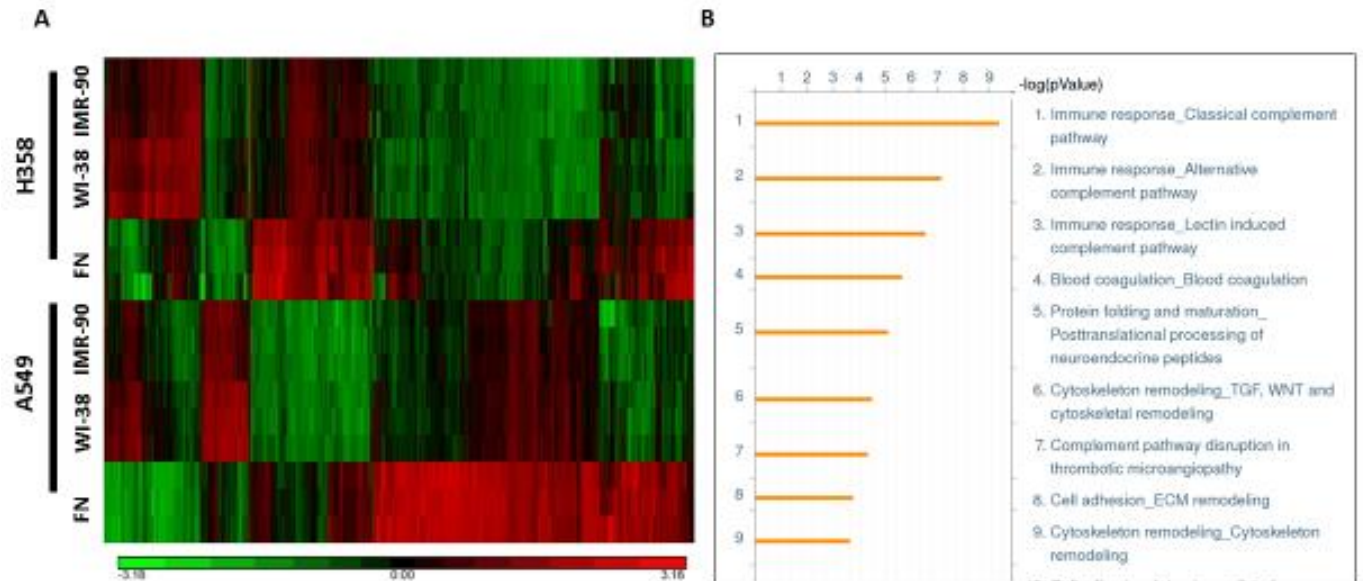


Figure 10- Fibroblast-derived ECM alters mRNA profile of A549 and H358 Cells. (A) heat map of common significant (fold change >1.5 and P-Value<.05). (B) Gene Ontology (GO) Processes.

To validate the microarray results, primers were designed to bind only to the mRNA region of various hits from the microarray data. Separate samples were analyzed to confirm the results. QPCR analysis (figure 11) revealed that indeed, mRNA that was altered on the microarray are also being altered in separate biological replicates. Further, the direction of the fold changes were shared between microarray data and qPCR, thus giving confidence in the highly sensitive microarray data.

TABLE II
COMMON RANKED DOWN-REGULATED GENES BETWEEN A549 AND H358
CELLS ON WI38 AND IMR-90 ECM

Gene Symbol	Fold Change	Gene Symbol	Fold Change
LGALS2	-20.2808	AQP3	-3.72207
PDZK1 /// PDZK1P1	-10.4226	TC2N	-3.60948
KCNJ16	-6.6045	COL12A1	-3.34803
MGAM	-6.26486	BAAT	-3.33715
FGB	-5.69714	CDH1	-3.2636
FGB	-5.60535	RND1	-3.20194
FGG	-5.5835	CP	-3.0807
SLC23A1	-5.33021	COL4A4	-3.00602
CORO2A	-5.31961	KRT4	-2.95014
FGB	-5.2152	CP	-2.9092
CORO2A	-5.18083	SLC17A3	-2.90152
TM4SF20	-5.16027	DDC	-2.87355
FGA	-5.1169	BCAS1	-2.84022
MGAM	-5.00299	RND1	-2.7561
PDZK1 /// PDZK1P1	-4.80629	BDKRB2	-2.72445
SEBOX /// VTN	-4.70113	COL5A1	-2.66928
TM4SF20	-4.53339	ELMO1	-2.3403
SLPI	-4.29842	C3	-2.29049
HGD	-4.25802	MMP7	-2.25888
COL4A4	-4.14101	CLDN2	-2.24846
LCN2	-3.91643	C1S	-2.19828
FGA	-3.90491	CXXC5	-1.80768
VIL1	-3.86278	MYLK	-1.79433
		DNMT3B	-1.62753

TABLE III
COMMON RANKED UP-REGULATED GENES BETWEEN A549 AND H358
CELLS ON WI38 AND IMR-90 ECM

Gene Symbol	Fold Change	Gene Symbol	Fold Change
NT5E	3.07708	FAM129A	2.09353
IL8	2.88492	PHLDA1	2.08875
TMEM158	2.68639	ASNS	2.07363
NT5E	2.60764	SMOX	1.8889
KCNMA1	2.53873	CHAC1	1.88397
KCNMA1	2.4901	SMOX	1.84901
STC2	2.44285	SMOX	1.83655
BCAT1	2.39524	PSAT1	1.83508
KCNMA1	2.3891	CDCP1	1.79694
BCAT1	2.3149	PSAT1	1.7723
ASNS	2.27585	PHGDH	1.76288
STC2	2.2733	PSAT1	1.75183
KCNMA1	2.2436	STMN3	1.74967
CDCP1	2.23108	SMOX	1.73798
ASNS	2.20114	PSAT1	1.7237
STC2	2.18055	PSAT1	1.68015
BCAT1	2.16265	PSAT1	1.67523
ASNS	2.13614	CTSL1	1.66385
FAM129A	2.09353	SFRP1	1.64401
PHLDA1	2.08875	ID4	1.61766
ASNS	2.07363	RAB39B	1.58386

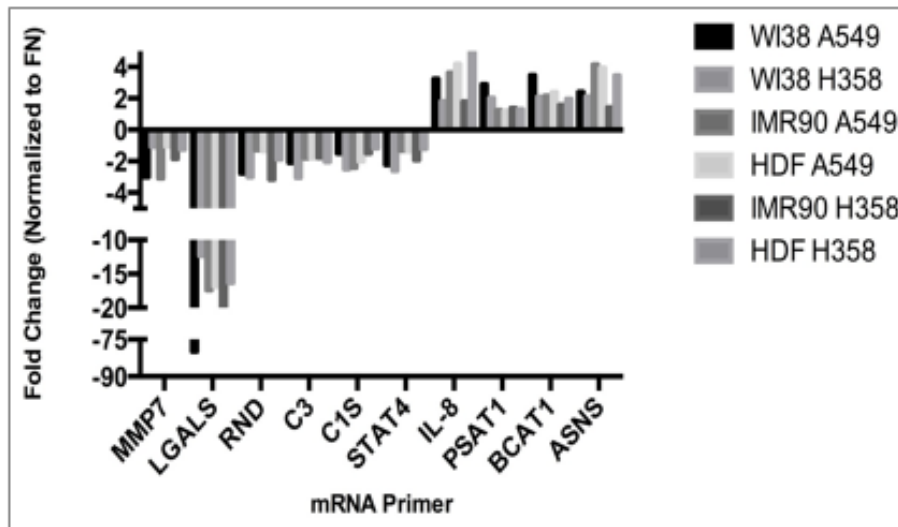


Figure 11- Fold Change of mRNA by Quantitative qPCR

H. Fibroblast-Derived Matrices can Induce Lung Cancer Cells to Migrate

In order to test how human fibroblast-derived ECM affects cellular migration, which is a critical mechanism that lung cancer cells utilize to invade and metastasize, a simple experiment was designed employing a trans-well chamber assay. A trans-well chamber is a tissue culture device that has 0.8 μm pores that allow actively migrating cells to pass through to the bottom. Trans-well chambers are used to model migratory and invasive properties of cells. Many cancer cell biologists employ such chambers to test whether migration is altered following a given treatment. Here, trans-well chambers were used as a substrate to grow ECM, which was then used to recruit lung cancer cells. It is hypothesized that human fibroblast-derived ECM could act as a chemoattractant that could recruit lung cancer cell lines to its side of the Trans-well chamber. Clearly, all fibroblast-derived ECM act as a chemoattractant for A549 cells and induce them to migrate across the chamber (figure 12A). Upon counting, Three times the number of A549 cells were migratory when fibroblast-derived ECM was present, compared to fibronectin. Also, it is clear visually by Hemotoxylin and Eosin staining that more A549 migrated to the fibroblast-derived ECM (Figure 12B). Cell migration is an important process that affects both wound healing and cancer progression. In cancer, local as well as distal ECM environments are equally as important in understanding cancer cell invasion and metastasis. Therefore, incorporation of fibroblast-derived ECM into trans-well migration could provide researchers with a better model to study cancer cell migration.

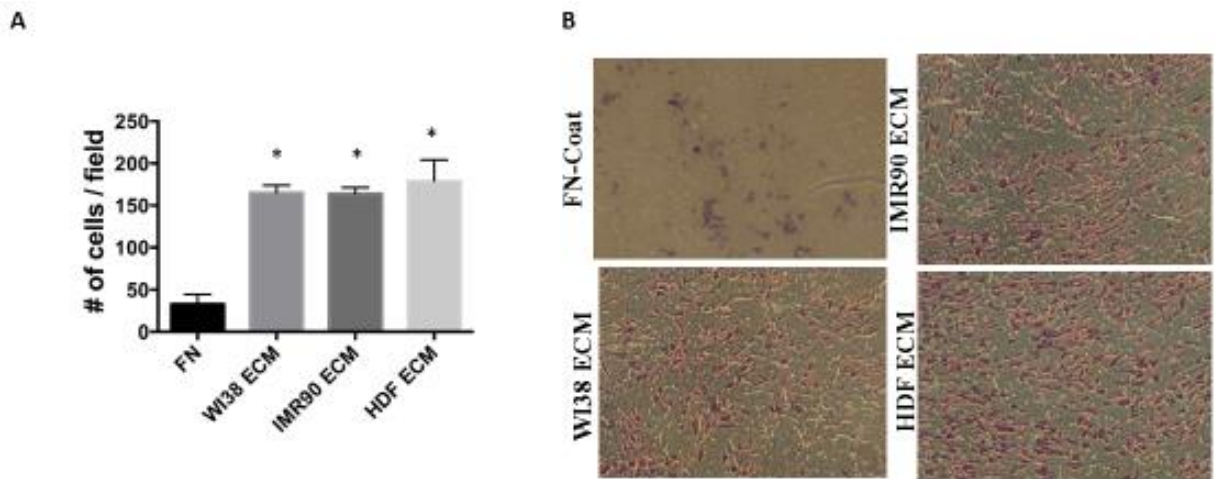


Figure 12- Fibroblast-derived ECM alter Cell Migration of Lung Cancer Cell Lines in a Trans-well Chamber. (A) Average migratory cells per field (4 fields). *, $p < .05$ (TABLE XIX)

It is clear from microscopic observation that lung cancer cell lines on fibroblast-derived ECM do not grow in large clusters like the same cells on fibronectin. It was postulated that cells grown on the 3D fibers of the ECM exhibit migration along the fibers, thus spreading the cells apart from each other. To test this type of migration, A549, H358 and HPL1D cells were seeded onto fibroblast-derived matrices and after 24 hours, and transferred into a live cell microscopy chamber for time lapse imaging (Figure 13). Time lapse shows that cells grown on fibroblast-derived matrix exhibit more migration than on a fibronectin-coated dish. A549, H358, and HPL1D cells all migrated farther distances on fibroblast-derived ECM than on Fibronectin (Figure 13A). Cells were tracked for between 4 and 8 hours (Figure 13A). It is clear that A549, H358 and HPL1D cells migrate farther distances in 4-8 hours when cultured on fibroblast-derived ECM. To characterize this, directionality (D/d) (Figure 13B) is calculated to determine the linearity of the directional migration (Figure 13C). Evidently, A549, H358, and HPL1D cell

lines exhibited a more directional migration on fibroblast-derived ECM compared to control. Also, it is clear by calculating distance traveled by time traveled that the velocity of the cells on fibroblast-derived ECM is greater than the same cells on fibronectin (Figure 13D). It appears that when the lung cancer cell lines interact with a fiber, they travel along that fiber only, rarely migrating towards adjacent fibers. This result will be important for future cell migration studies, encouraging researchers to use ECM that are highly aligned. Cell migration properties are highly adopted in lung cancer cells, which allow them to invade accessible organs. These data indicate that the interaction between fibroblast-derived ECM and lung cancer cell lines induce a migratory phenotype. It is interesting that fibroblast-derived ECM alters EMT properties, thus suggesting that the ECM plays a role in EMT induction. Therefore, fibroblast-derived matrices should be used in the future to study ECM-induced EMT.

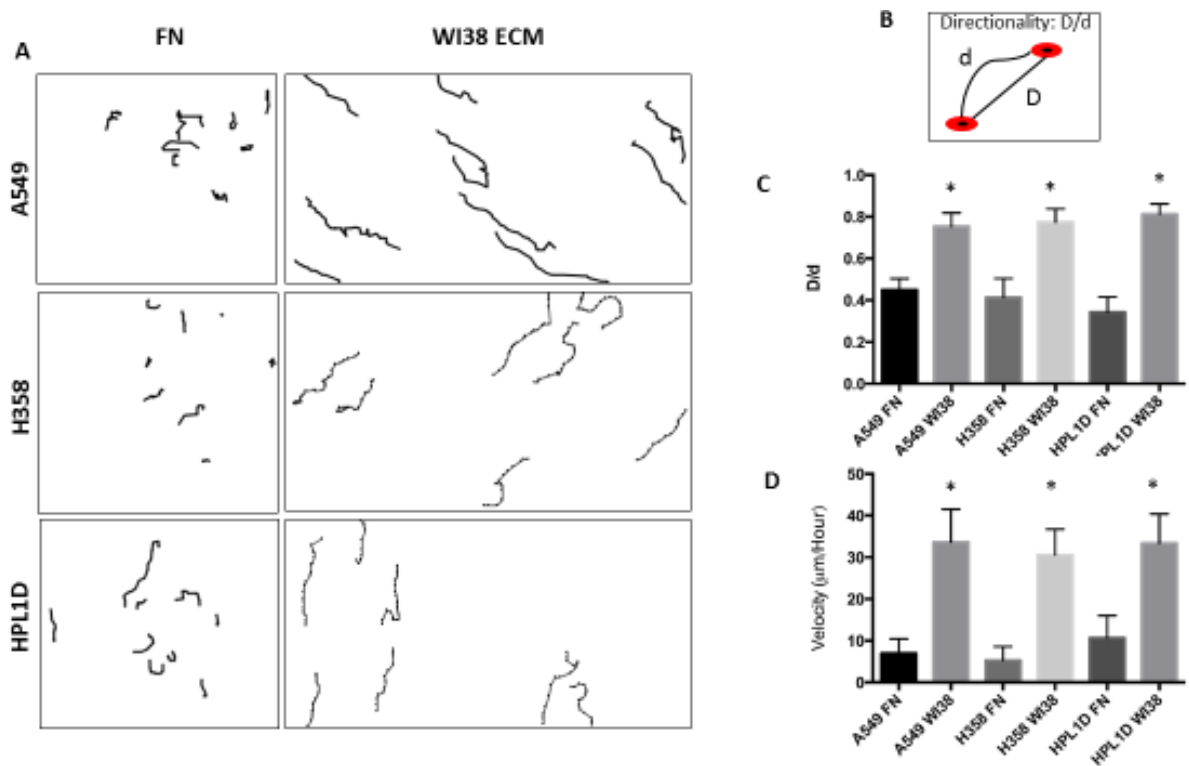


Figure 13- Fibroblast-derived ECM induce Directional Migration. (A) Resultant time-lapse tracks of A549, H358, and HPL1D on WI38 ECM. (B) Depiction of Directionality= D/d. (C) Calculated Directionality A549 p=.033, H358 p=.0067, HPL1D p=.0002). N=7. (D) Average velocity of migrated cells ($\mu\text{m}/\text{hour}$) N=7. A549, H358, HPL1D p<.0001.

I. Fibroblast-derived ECM increase Subcutaneous Xenograft Tumor Formation in Mice

Previously, researchers have mixed cancer cells with Matrigel™ and implant them into the subcutaneous tissue of mice to study tumor development with an extracellular matrix component (Bao et al., 1994). To determine if fibroblast-derived matrix proteins could illicit a similar pro-tumorigenic response, fibroblast-derived ECM was scraped in ice cold PBS and mixed with cells. The same ECM-to-cell ratio of 26,000 cells/cm² that was used *in vitro* was used *in vivo*. A549 cells and matrix or A549 cells without matrix were injected into the subcutaneous tissue of 3 month old NRG mice and let incubate for 60 days. NRG mice are a new breed of mice that harbor mutations in NOD, RAG, and Interleukin-2 gamma receptor, which eliminates the immune system of T and B cells, thus allowing for xenografts without an immune response. A549 cells cultured with the scraped fibroblast-derived ECM grew larger tumors by calculated volume than A549 cells with PBS (figure 14). However, these data are not statically significant, with a p-value of 0.07 (Figure 14C) Further, tumors with ECM showed increased mass (Figure 14D). Although A549 cells mixed with ECM formed larger tumors, hemotoxylin and Eosin staining revealed little differences in the morphology of the tumors (Figure 14Ei and 14Eii). A pathologist might be able to determine differences in the tumor sections, but little is

noticeable to the untrained eye. Although fibroblast-derived ECM decreased lung cancer cell line growth *in vitro*, it appears that they enhance tumor formation *in vivo*. It could be that in the *in vivo* experiments, the native architecture of the fibroblast-derived ECM that is present *in vitro* is absent, thus allowing for unrestricted growth. Further, the same survival signals that the lung cancer cell lines exploit from the ECM *in vitro* when cultured in serum free conditions may be providing growth signals in A549 xenograft. These data are important in elucidating the role of ECM architecture, as well as its composition in cancer signaling.

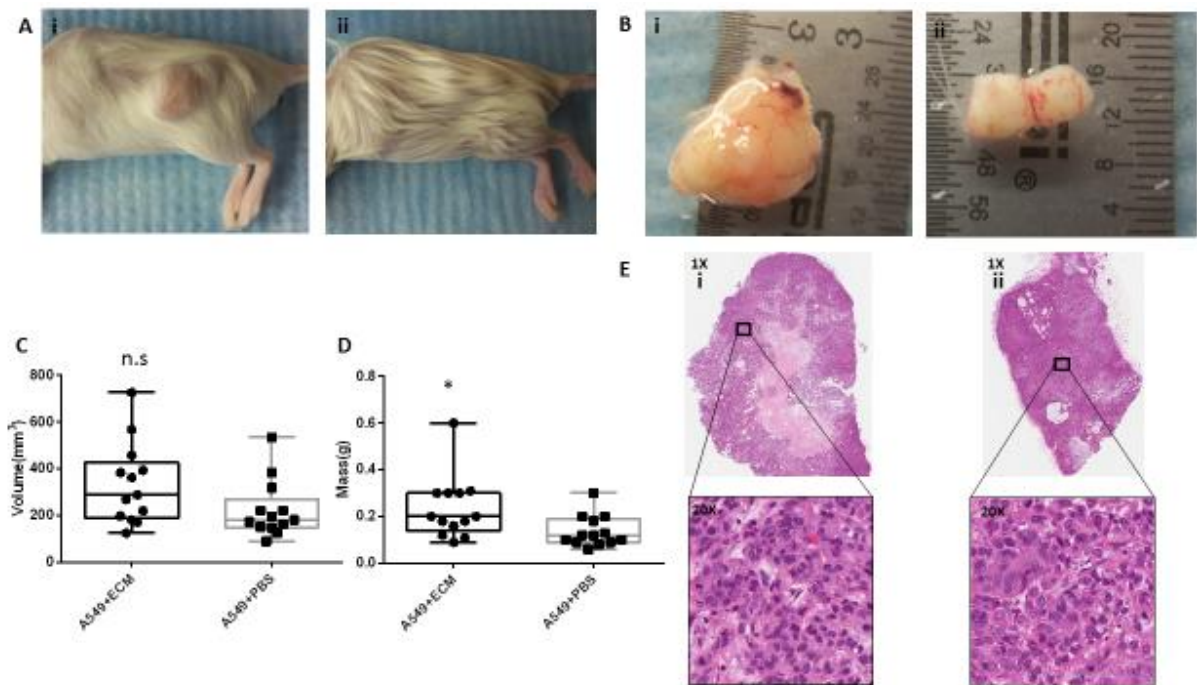


Figure 14- Fibroblast-derived ECM Increase Subcutaneous Tumor Formation of A549 Cells in Mice. Ai: *in vivo* A549 with ECM Xenograft. Aii: *in vivo* A549 with PBS Xenograft. Bi: Picture of removed tumor with ECM. Bii: Picture of removed tumor with PBS. C: Calculated volumes of A549 xenografts with and without ECM (p=.07) D: Measured mass (g) of excised

tumors. Ei: 1X picture of A549 Xenograft with ECM with 20X expanded view. Eii: 1X picture of A549 Xenograft with PBS with 20X expanded view. N=13. *,p=.026.

IV. Conclusions and Future Directions

This work shows that human fibroblast-derived extracellular matrix alters cell morphology, cell growth, proliferation, mRNA expression, functional protein expression, and migration properties of the A549, H358, and HPL1D lung cancer cell lines. It is clear that when cultured on human fibroblast-derived ECM, lung cancer cell lines have less activated MAPK protein expression as well as mTOR and cell cycle regulating pathways, which might contribute to their attenuated growth. Also, lung cancers have altered expression of ECM protein-coding and complement-mediated immunity genes, thus affecting many downstream pathways. As discovered by western blot, lung cancer cell lines also develop an EMT phenotype, which could be responsible for their increased migration. Interestingly, ECM harvested from different fibroblast cell lines originating from different human tissues did not have a noticeable effect in this work. Perhaps a less harsh decellularization technique will leave behind a more unique ECM, thus allowing researchers to better mimic the natural ECM. This work is significant because it provides further evidence that the extracellular matrix has a strong effect on the phenotype of lung cancer cells, which needs to be further studied to successfully treat the entire tumor microenvironment.

In future work, scientists will attempt to alter the influence of the extracellular matrix on lung cancer cells. For instance, the age of an organism significantly increases the probability of cancer and it is known that cancer vulnerability can be influenced by the extracellular matrix.

Fibroblast-derived matrix procured from old and young fibroblasts could differentially regulate the growth and migration properties of lung cancer cells. Also, future researchers could successfully overexpress oncogenes or knock down tumor suppressors in human lung or dermal fibroblasts and determine their matrix components and subsequent effect on lung cancer cell lines. Further, to expand research in different focus groups, future researchers could examine how the ECM affects leukemia cells with regards to homing and drug resistance. This future work could provide valuable knowledge of the feedback mechanism that exists between cancer and its extracellular matrix.

V. Appendix

TABLE IV
A549 SERUM DEPRIVATION STATISTICAL ANALYSIS

A549 Serum deprivation				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Summary	Adjusted P Value
Row 1				
FN vs. IMR90 ECM	0	-0.09736 to 0.09736	ns	> 0.9999
FN vs. WI38 ECM	1.99E-08	-0.09736 to 0.09736	ns	> 0.9999
FN vs. HDF ECM	0	-0.09736 to 0.09736	ns	> 0.9999
IMR90 ECM vs. WI38 ECM	1.99E-08	-0.09736 to 0.09736	ns	> 0.9999
IMR90 ECM vs. HDF ECM	0	-0.09736 to 0.09736	ns	> 0.9999
WI38 ECM vs. HDF ECM	-2E-08	-0.09736 to 0.09736	ns	> 0.9999

Row 2				
FN vs. IMR90 ECM	-0.566	-0.6634 to - 0.4687	****	< 0.0001
FN vs. WI38 ECM	-0.4437	-0.5411 to - 0.3463	****	< 0.0001
FN vs. HDF ECM	-0.5542	-0.6516 to - 0.4569	****	< 0.0001
IMR90 ECM vs. WI38 ECM	0.1223	0.02496 to 0.2197	**	0.0093
IMR90 ECM vs. HDF ECM	0.01178	-0.08558 to 0.1091	ns	0.9876
WI38 ECM vs. HDF ECM	-0.1105	-0.2079 to - 0.01318	*	0.0211
Row 3				
FN vs. IMR90 ECM	-0.3626	-0.4600 to - 0.2653	****	< 0.0001
FN vs. WI38 ECM	-0.3496	-0.4470 to - 0.2523	****	< 0.0001
FN vs. HDF ECM	-0.3315	-0.4289 to - 0.2342	****	< 0.0001
IMR90 ECM vs. WI38 ECM	0.01299	-0.08437 to 0.1103	ns	0.9835
IMR90 ECM vs. HDF ECM	0.03112	-0.06624 to 0.1285	ns	0.8222
WI38 ECM vs. HDF ECM	0.01813	-0.07923 to 0.1155	ns	0.9574
Row 4				
FN vs. IMR90 ECM	-0.1861	-0.2834 to - 0.08870	****	< 0.0001
FN vs. WI38 ECM	-0.159	-0.2563 to - 0.06161	***	0.0006
FN vs. HDF ECM	- 0.08953	-0.1869 to 0.007827	ns	0.0805
IMR90 ECM vs. WI38 ECM	0.02709	-0.07027 to 0.1244	ns	0.8744
IMR90 ECM vs. HDF ECM	0.09652	-0.0008332 to 0.1939	ns	0.0527
WI38 ECM vs. HDF ECM	0.06944	-0.02792 to 0.1668	ns	0.2351

TABLE V
H358 SERUM DEPRIVATION STATISTICAL ANALYSIS

H358 Serum Deprivation				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant?	Adjusted P Value
Row 1				
FN vs. IMR90 ECM	-2E-08	-0.2692 to 0.2692	No	> 0.9999
FN vs. WI38 ECM	-2E-08	-0.2692 to 0.2692	No	> 0.9999
FN vs. HDF ECM	2E-08	-0.2692 to 0.2692	No	> 0.9999
IMR90 ECM vs. WI38 ECM	0	-0.2692 to 0.2692	No	> 0.9999
IMR90 ECM vs. HDF ECM	4E-08	-0.2692 to 0.2692	No	> 0.9999
WI38 ECM vs. HDF ECM	4E-08	-0.2692 to 0.2692	No	> 0.9999
Row 2				
FN vs. IMR90 ECM	-0.3225	-0.5918 to -0.05327	Yes	0.0139
FN vs. WI38 ECM	-0.271	-0.5402 to -0.001707	Yes	0.0481
FN vs. HDF ECM	-0.5012	-0.7705 to -0.2320	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	0.05156	-0.2177 to 0.3208	No	0.954
IMR90 ECM vs. HDF ECM	-0.1787	-0.4480 to 0.09053	No	0.2927
WI38 ECM vs. HDF ECM	-0.2303	-0.4995 to 0.03897	No	0.1153
Row 3				
FN vs. IMR90 ECM	-0.3275	-0.5967 to -0.05824	Yes	0.0122
FN vs. WI38 ECM	-0.2891	-0.5584 to -0.01988	Yes	0.0315
FN vs. HDF ECM	-0.2195	-0.4887 to 0.04979	No	0.1427
IMR90 ECM vs. WI38 ECM	0.03836	-0.2309 to 0.3076	No	0.9801
IMR90 ECM vs. HDF ECM	0.108	-0.1612 to 0.3773	No	0.6998

WI38 ECM vs. HDF ECM	0.06967	-0.1996 to 0.3389	No	0.8959
Row 4				
FN vs. IMR90 ECM	-0.0809	-0.3501 to 0.1883	No	0.8473
FN vs. WI38 ECM	-0.1447	-0.4140 to 0.1245	No	0.4748
FN vs. HDF ECM	-0.2083	-0.4776 to 0.06093	No	0.176
IMR90 ECM vs. WI38 ECM	-0.0638	-0.3331 to 0.2054	No	0.9174
IMR90 ECM vs. HDF ECM	-0.1274	-0.3966 to 0.1418	No	0.5807
WI38 ECM vs. HDF ECM	-0.0636	-0.3328 to 0.2057	No	0.9183

TABLE VI
HPL1D SERUM DEPRIVATION STATISTICAL ANALYSIS

HPL1D Serum Deprivation				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant?	Adjusted P Value
Row 1				
FN vs. IMR90 ECM	-2E-08	-0.1039 to 0.1039	No	> 0.9999
FN vs. WI38 ECM	-2E-08	-0.1039 to 0.1039	No	> 0.9999
FN vs. HDF ECM	-2E-08	-0.1039 to 0.1039	No	> 0.9999
IMR90 ECM vs. WI38 ECM	0	-0.1039 to 0.1039	No	> 0.9999
IMR90 ECM vs. HDF ECM	0	-0.1039 to 0.1039	No	> 0.9999
WI38 ECM vs. HDF ECM	0	-0.1039 to 0.1039	No	> 0.9999
Row 2				
FN vs. IMR90 ECM	-0.1468	-0.2508 to -	Yes	0.003

		0.04287		
FN vs. WI38 ECM	-0.1686	-0.2726 to - 0.06466	Yes	0.0006
FN vs. HDF ECM	-0.2119	-0.3159 to - 0.1080	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	-0.0218	-0.1257 to 0.08215	No	0.9408
IMR90 ECM vs. HDF ECM	-0.06514	-0.1691 to 0.03881	No	0.3414
WI38 ECM vs. HDF ECM	-0.04334	-0.1473 to 0.06061	No	0.6743
Row 3				
FN vs. IMR90 ECM	-0.3222	-0.4261 to - 0.2182	Yes	< 0.0001
FN vs. WI38 ECM	-0.3179	-0.4218 to - 0.2139	Yes	< 0.0001
FN vs. HDF ECM	-0.3933	-0.4973 to - 0.2894	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	0.004302	-0.09964 to 0.1082	No	0.9995
IMR90 ECM vs. HDF ECM	-0.07113	-0.1751 to 0.03282	No	0.2678
WI38 ECM vs. HDF ECM	-0.07543	-0.1794 to 0.02852	No	0.2219
Row 4				
FN vs. IMR90 ECM	-0.1735	-0.2774 to - 0.06951	Yes	0.0004
FN vs. WI38 ECM	-0.2279	-0.3319 to - 0.1240	Yes	< 0.0001
FN vs. HDF ECM	-0.1141	-0.2181 to - 0.01020	Yes	0.027
IMR90 ECM vs. WI38 ECM	-0.05446	-0.1584 to 0.04949	No	0.4968
IMR90 ECM vs. HDF ECM	0.05931	-0.04463 to 0.1633	No	0.423
WI38 ECM vs. HDF ECM	0.1138	0.009828 to 0.2177	Yes	0.0276

TABLE VII
ECM PROTEIN STATISTICAL ANALYSIS

ECM Protein				
Number of families	1			
Number of comparisons per family	3			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Summary	Adjusted P Value
WI38 ECM vs. IMR-90 ECM	0.04534	-1.902 to 1.992	ns	0.9972
WI38 ECM vs. HDF ECM	-1.953	-3.900 to -0.005657	*	0.0495
IMR-90 ECM vs. HDF ECM	-1.998	-3.945 to -0.05100	*	0.0453

TABLE VIII
TIME-LAPSE STATISTICAL ANALYSIS

Time-lapse Migration Velocity				
Number of families	1			
Number of comparisons per family	15			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Summary	Adjusted P Value
A549 FN vs. A549 WI38	-26.64	-35.96 to -17.32	****	< 0.0001
A549 FN vs. H358 FN	1.734	-7.590 to 11.06	ns	0.993
A549 FN vs. HPL1D FN	-3.643	-12.97 to 5.680	ns	0.8454
A549 WI38 vs. H358 FN	28.37	19.05 to 37.70	****	< 0.0001
A549 WI38 vs. H358 WI38	3.07	-6.254 to 12.39	ns	0.918
A549 WI38 vs. HPL1D FN	23	13.67 to 32.32	****	< 0.0001
A549 WI38 vs. HPL1D WI38	0.2967	-9.027 to 9.620	ns	> 0.9999
H358 FN vs. H358 WI38	-25.3	-34.63 to -15.98	****	< 0.0001
H358 FN vs. HPL1D FN	-5.377	-14.70 to 3.947	ns	0.5187
H358 WI38 vs. HPL1D WI38	-2.773	-12.10 to 6.550	ns	0.9453

HPL1D FN vs. HPL1D WI38	-22.7	-32.02 to -13.38	****	< 0.0001

TABLE IX
DIRECTIONALITY STATISTICAL ANALYSIS

Directionality					
Number of families	1				
Number of comparisons per family	15				
Alpha	0.05				
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant ?	Summary	Adjusted P Value
A549 FN vs. A549 WI38	-0.3031	-0.5907 to -0.01558	Yes	*	0.0336
A549 FN vs. H358 FN	0.03643	-0.2511 to 0.3240	No	ns	0.9989
A549 FN vs. H358 WI38	-0.3244	-0.6120 to -0.03687	Yes	*	0.019
A549 FN vs. HPL1D FN	0.1072	-0.1804 to 0.3947	No	ns	0.8735
A549 FN vs. HPL1D WI38	-0.3624	-0.6499 to -0.07483	Yes	**	0.0064
A549 WI38 vs. H358 FN	0.3396	0.05201 to 0.6271	Yes	*	0.0124
A549 WI38 vs. H358 WI38	-0.02129	-0.3088 to 0.2663	No	ns	> 0.9999
A549 WI38 vs. HPL1D	0.4103	0.1227 to	Yes	**	0.0015

FN		0.6978			
A549 WI38 vs. HPL1D WI38	-0.05925	-0.3468 to 0.2283	No	ns	0.9893
H358 FN vs. H358 WI38	-0.3609	-0.6484 to -0.07330	Yes	**	0.0067
H358 FN vs. HPL1D FN	0.07072	-0.2168 to 0.3583	No	ns	0.9765
H358 FN vs. HPL1D WI38	-0.3988	-0.6864 to -0.1113	Yes	**	0.0021
H358 WI38 vs. HPL1D FN	0.4316	0.1440 to 0.7191	Yes	***	0.0008
H358 WI38 vs. HPL1D WI38	-0.03796	-0.3255 to 0.2496	No	ns	0.9987
HPL1D FN vs. HPL1D WI38	-0.4695	-0.7571 to -0.1820	Yes	***	0.0002

TABLE X
CIRCULARITY STATISTICAL ANALYSIS

Cell Circularity					
Number of families	1				
Number of comparisons per family	6				
Alpha	0.05				
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant ?	Summary	Adjusted P Value
A549 FN vs. A549 WI38	0.196	0.05046 to 0.3414	Yes	**	0.0043
A549 FN vs. A549 IMR-90	0.2074	0.06186 to 0.3528	Yes	**	0.0023

A549 FN vs. A549 HDF	0.2528	0.1073 to 0.3982	Yes	***	0.0002
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TABLE XI
A549 CELL COUNTING STATISTICAL ANALYSIS

A549 Count				
Within each row, compare columns (simple effects within rows)				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant?	Adjusted P Value
Row 1				
FN vs. IMR90 ECM	2667	-43697 to 49030	No	0.9986
FN vs. WI38 ECM	6000	-40364 to 52364	No	0.9849
FN vs. HDF	1000	-45364 to 47364	No	> 0.9999
IMR90 ECM vs. WI38 ECM	3333	-43030 to 49697	No	0.9973
IMR90 ECM vs. HDF	-1667	-48030 to 44697	No	0.9997
WI38 ECM vs. HDF	-5000	-51364 to 41364	No	0.9911
Row 2				
FN vs. IMR90 ECM	15667	-30697 to 62030	No	0.7967
FN vs. WI38 ECM	21333	-25030 to 67697	No	0.6026
FN vs. HDF	14667	-31697 to 61030	No	0.8266
IMR90 ECM vs. WI38 ECM	5667	-40697 to 52030	No	0.9872
IMR90 ECM vs. HDF	-1000	-47364 to 45364	No	> 0.9999

WI38 ECM vs. HDF	-6667	-53030 to 39697	No	0.9796
Row 3				
FN vs. IMR90 ECM	51667	5303 to 98030	Yes	0.0243
FN vs. WI38 ECM	60000	13636 to 106364	Yes	0.0071
FN vs. HDF	50000	3636 to 96364	Yes	0.0306
IMR90 ECM vs. WI38 ECM	8333	-38030 to 54697	No	0.9614
IMR90 ECM vs. HDF	-1667	-48030 to 44697	No	0.9997
WI38 ECM vs. HDF	-10000	-56364 to 36364	No	0.9361
Row 4				
FN vs. IMR90 ECM	143333	96970 to 189697	Yes	< 0.0001
FN vs. WI38 ECM	136667	90303 to 183030	Yes	< 0.0001
FN vs. HDF	165000	118636 to 211364	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	-6667	-53030 to 39697	No	0.9796
IMR90 ECM vs. HDF	21667	-24697 to 68030	No	0.5905
WI38 ECM vs. HDF	28333	-18030 to 74697	No	0.3631

TABLE XII
H358 CELL COUNTING STATISTICAL ANALYSIS

H358 Count				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
		95% CI of diff.	Significant?	Adjusted P Value
Tukey's multiple comparisons test	Mean Diff.			

Row 1		-97833 to 105833	No	0.9996
FN vs. IMR90 ECM	4000	-99833 to 103833	No	> 0.9999
FN vs. WI38 ECM	2000	-96833 to 106833	No	0.9991
FN vs. HDF	5000	-103833 to 99833	No	> 0.9999
IMR90 ECM vs. WI38 ECM	-2000	-100833 to 102833	No	> 0.9999
IMR90 ECM vs. HDF	1000	-98833 to 104833	No	0.9998
WI38 ECM vs. HDF	3000			
Row 2		-85333 to 118333	No	0.9712
FN vs. IMR90 ECM	16500	-90833 to 112833	No	0.9911
FN vs. WI38 ECM	11000	-92333 to 111333	No	0.9942
FN vs. HDF	9500	-107333 to 96333	No	0.9989
IMR90 ECM vs. WI38 ECM	-5500	-108833 to 94833	No	0.9977
IMR90 ECM vs. HDF	-7000	-103333 to 100333	No	> 0.9999
WI38 ECM vs. HDF	-1500			
Row 3		-56062 to 147604	No	0.6204
FN vs. IMR90 ECM	45771	-70062 to 133604	No	0.8324
FN vs. WI38 ECM	31771	-85062 to 118604	No	0.9699
FN vs. HDF	16771	-115833 to 87833	No	0.982
IMR90 ECM vs. WI38 ECM	-14000	-130833 to 72833	No	0.8667
IMR90 ECM vs. HDF	-29000	-116833 to 86833	No	0.9781
WI38 ECM vs. HDF	-15000			
Row 4		93167 to 296833	Yes	< 0.0001
FN vs. IMR90 ECM	195000	83167 to 286833	Yes	0.0001
FN vs. WI38 ECM	185000	53167 to 256833	Yes	0.0013
FN vs. HDF	155000	-111833 to 91833	No	0.9933
IMR90 ECM vs. WI38 ECM	-10000	-141833 to 61833	No	0.7133
IMR90 ECM vs. HDF	-40000	-131833 to 71833	No	0.8547

TABLE XIII
HPL1D CELL COUNTING STATISTICAL ANALYSIS

HPL1D Count					
Number of families	4				
Number of comparisons per family	6				
Alpha	0.05				

Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant ?	Summary	Adjusted P Value
Row 1					
FN vs. IMR90 ECM	-500	-20871 to 19871	No	ns	0.9999
FN vs. WI38 ECM	4000	-16371 to 24371	No	ns	0.9506
FN vs. HDF	-1000	-21371 to 19371	No	ns	0.9991
IMR90 ECM vs. WI38 ECM	4500	-15871 to 24871	No	ns	0.9318
IMR90 ECM vs. HDF	-500	-20871 to 19871	No	ns	0.9999
WI38 ECM vs. HDF	-5000	-25371 to 15371	No	ns	0.9094
Row 2					
FN vs. IMR90 ECM	8000	-12371 to 28371	No	ns	0.7135
FN vs. WI38 ECM	9000	-11371 to 29371	No	ns	0.6331
FN vs. HDF	2000	-18371 to 22371	No	ns	0.9933
IMR90 ECM vs. WI38 ECM	1000	-19371 to 21371	No	ns	0.9991
IMR90 ECM vs. HDF	-6000	-26371 to 14371	No	ns	0.8548
WI38 ECM vs. HDF	-7000	-27371 to 13371	No	ns	0.7885
Row 3					
FN vs. IMR90 ECM	67000	46629 to 87371	Yes	****	< 0.0001
FN vs. WI38 ECM	55500	35129 to 75871	Yes	****	< 0.0001
FN vs. HDF	57500	37129 to 77871	Yes	****	< 0.0001
IMR90 ECM vs. WI38 ECM	-11500	-31871 to 8871	No	ns	0.4323
IMR90 ECM vs. HDF	-9500	-29871 to	No	ns	0.5921

		10871			
WI38 ECM vs. HDF	2000	-18371 to 22371	No	ns	0.9933
Row 4					
FN vs. IMR90 ECM	160000	139629 to 180371	Yes	****	< 0.0001
FN vs. WI38 ECM	115000	94629 to 135371	Yes	****	< 0.0001
FN vs. HDF	120000	99629 to 140371	Yes	****	< 0.0001
IMR90 ECM vs. WI38 ECM	-45000	-65371 to - 24629	Yes	****	< 0.0001
IMR90 ECM vs. HDF	-40000	-60371 to - 19629	Yes	****	< 0.0001
WI38 ECM vs. HDF	5000	-15371 to 25371	No	ns	0.9094

TABLE XIV
A549 ALAMAR BLUE STATISTICAL ANALYSIS

A549 Alamar blue				
Within each row, compare columns (simple effects within rows)				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant?	Adjusted P Value
Row 1				
FN vs. IMR90 ECM	1.987E-08	-0.03641 to 0.03641	No	> 0.9999
FN vs. WI38 ECM	-1.987E-08	-0.03641 to 0.03641	No	> 0.9999
FN vs. HDF ECM	1.987E-08	-0.03641 to 0.03641	No	> 0.9999
IMR90 ECM vs. WI38 ECM	-3.974E-08	-0.03641 to 0.03641	No	> 0.9999
IMR90 ECM vs. HDF ECM	0	-0.03641 to	No	> 0.9999

		0.03641		
WI38 ECM vs. HDF ECM	3.974E-08	-0.03641 to 0.03641	No	> 0.9999
Row 2				
FN vs. IMR90 ECM	0.09386	0.05745 to 0.1303	Yes	< 0.0001
FN vs. WI38 ECM	0.02694	-0.009478 to 0.06335	No	0.2077
FN vs. HDF ECM	0.06571	0.02929 to 0.1021	Yes	0.0002
IMR90 ECM vs. WI38 ECM	-0.06693	-0.1033 to - 0.03051	Yes	0.0001
IMR90 ECM vs. HDF ECM	-0.02816	-0.06457 to 0.008258	No	0.1765
WI38 ECM vs. HDF ECM	0.03877	0.002359 to 0.07519	Yes	0.0334
Row 3				
FN vs. IMR90 ECM	0.2109	0.1745 to 0.2474	Yes	< 0.0001
FN vs. WI38 ECM	0.1547	0.1183 to 0.1911	Yes	< 0.0001
FN vs. HDF ECM	0.2077	0.1713 to 0.2441	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	-0.05628	-0.09270 to - 0.01987	Yes	0.0011
IMR90 ECM vs. HDF ECM	-0.003285	-0.03970 to 0.03313	No	0.9948
WI38 ECM vs. HDF ECM	0.053	0.01659 to 0.08941	Yes	0.0022
Row 4				
FN vs. IMR90 ECM	0.2113	0.1749 to 0.2477	Yes	< 0.0001
FN vs. WI38 ECM	0.1586	0.1222 to 0.1950	Yes	< 0.0001
FN vs. HDF ECM	0.1954	0.1590 to 0.2318	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	-0.05269	-0.08910 to - 0.01627	Yes	0.0024
IMR90 ECM vs. HDF ECM	-0.01586	-0.05228 to 0.02055	No	0.6434
WI38 ECM vs. HDF ECM	0.03682	0.0004103 to 0.07324	Yes	0.0467

TABLE XV

H358 ALAMAR BLUE STATISTICAL ANALYSIS

H358 Alamar Blue				
Within each row, compare columns (simple effects within rows)				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple comparisons test	Mean Diff.	95% CI of diff.	Significant?	Adjusted P Value
Row 1				
FN vs. IMR90 ECM	-1.987E-08	-0.06557 to 0.06557	No	> 0.9999
FN vs. WI38 ECM	-1.987E-08	-0.06557 to 0.06557	No	> 0.9999
FN vs. HDF ECM	-1.987E-08	-0.06557 to 0.06557	No	> 0.9999
IMR90 ECM vs. WI38 ECM	0	-0.06557 to 0.06557	No	> 0.9999
IMR90 ECM vs. HDF ECM	0	-0.06557 to 0.06557	No	> 0.9999
WI38 ECM vs. HDF ECM	0	-0.06557 to 0.06557	No	> 0.9999
Row 2				
FN vs. IMR90 ECM	0.08628	0.02071 to 0.1518	Yes	0.0061
FN vs. WI38 ECM	0.1088	0.04320 to 0.1743	Yes	0.0005
FN vs. HDF ECM	0.1004	0.03486 to 0.1660	Yes	0.0013
IMR90 ECM vs. WI38 ECM	0.0225	-0.04307 to 0.08807	No	0.7893
IMR90 ECM vs. HDF ECM	0.01415	-0.05142 to 0.07972	No	0.9359
WI38 ECM vs. HDF ECM	-0.008342	-0.07391 to 0.05723	No	0.9857
Row 3				
FN vs. IMR90 ECM	0.2414	0.1759 to	Yes	< 0.0001

		0.3070		
FN vs. WI38 ECM	0.2667	0.2011 to 0.3322	Yes	< 0.0001
FN vs. HDF ECM	0.2455	0.1800 to 0.3111	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	0.02523	-0.04034 to 0.09080	No	0.726
IMR90 ECM vs. HDF ECM	0.004093	-0.06148 to 0.06966	No	0.9982
WI38 ECM vs. HDF ECM	-0.02114	-0.08671 to 0.04443	No	0.8185
Row 4				
FN vs. IMR90 ECM	0.294	0.2284 to 0.3596	Yes	< 0.0001
FN vs. WI38 ECM	0.3383	0.2727 to 0.4038	Yes	< 0.0001
FN vs. HDF ECM	0.3398	0.2742 to 0.4054	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	0.04428	-0.02129 to 0.1098	No	0.2785
IMR90 ECM vs. HDF ECM	0.0458	-0.01977 to 0.1114	No	0.2514
WI38 ECM vs. HDF ECM	0.001515	-0.06405 to 0.06708	No	> 0.9999

TABLE XVI
HPL1D ALAMAR BLUE STATISTICAL ANALYSIS

HPL1D Alamar Blue				
Within each row, compare columns (simple effects within rows)				
Number of families	4			
Number of comparisons per family	6			
Alpha	0.05			
Tukey's multiple	Mean	95% CI of diff.	Significant?	Adjusted P

comparisons test	Diff.			Value
Row 1				
FN vs. IMR90 ECM	-1.987E-08	-0.05237 to 0.05237	No	> 0.9999
FN vs. WI38 ECM	0	-0.05237 to 0.05237	No	> 0.9999
FN vs. HDF ECM	2.782E-07	-0.05237 to 0.05237	No	> 0.9999
IMR90 ECM vs. WI38 ECM	1.987E-08	-0.05237 to 0.05237	No	> 0.9999
IMR90 ECM vs. HDF ECM	2.98E-07	-0.05237 to 0.05237	No	> 0.9999
WI38 ECM vs. HDF ECM	2.782E-07	-0.05237 to 0.05237	No	> 0.9999
Row 2				
FN vs. IMR90 ECM	0.08531	0.03293 to 0.1377	Yes	0.0006
FN vs. WI38 ECM	0.1229	0.07049 to 0.1752	Yes	< 0.0001
FN vs. HDF ECM	0.0195	-0.03287 to 0.07187	No	0.7455
IMR90 ECM vs. WI38 ECM	0.03755	-0.01482 to 0.08993	No	0.231
IMR90 ECM vs. HDF ECM	-0.06581	-0.1182 to -0.01343	Yes	0.0093
WI38 ECM vs. HDF ECM	-0.1034	-0.1557 to -0.05099	Yes	< 0.0001
Row 3				
FN vs. IMR90 ECM	0.3145	0.2621 to 0.3669	Yes	< 0.0001
FN vs. WI38 ECM	0.2348	0.1824 to 0.2872	Yes	< 0.0001
FN vs. HDF ECM	0.1816	0.1292 to 0.2340	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	-0.07967	-0.1320 to -0.02730	Yes	0.0014
IMR90 ECM vs. HDF ECM	-0.1329	-0.1853 to -0.08051	Yes	< 0.0001
WI38 ECM vs. HDF ECM	-0.05321	-0.1056 to -0.0008385	Yes	0.0453
Row 4				
FN vs. IMR90 ECM	0.3231	0.2708 to 0.3755	Yes	< 0.0001
FN vs. WI38 ECM	0.2765	0.2242 to 0.3289	Yes	< 0.0001

FN vs. HDF ECM	0.2025	0.1501 to 0.2549	Yes	< 0.0001
IMR90 ECM vs. WI38 ECM	-0.04661	-0.09898 to 0.005764	No	0.0952
IMR90 ECM vs. HDF ECM	-0.1206	-0.1730 to -0.06827	Yes	< 0.0001
WI38 ECM vs. HDF ECM	-0.07403	-0.1264 to -0.02166	Yes	0.003

TABLE XVII
XENOGRAFT VOLUME STATISTICAL ANALYSIS

Xenograft Tumor Volume	Data 1
Column B	A549 +PBS
vs.	vs.
Column A	A549+ECM
Unpaired t test	
P value	0.0718
P value summary	ns
Significantly different? (P < 0.05)	No
One- or two-tailed P value?	Two-tailed
t, df	t=1.884 df=24
How big is the difference?	
Mean ± SEM of column A	334.3 ± 48.20, n=13
Mean ± SEM of column B	223.3 ± 33.88, n=13
Difference between means	-111.0 ± 58.92
95% confidence interval	-232.6 to 10.62
R squared	0.1288
F test to compare variances	
F,DFn, Dfd	2.024, 12, 12
P value	0.2364
P value summary	ns
Significantly different? (P < 0.05)	No

TABLE XVIII
XENOGRAFT WEIGHT STATISTICAL ANALYSIS

Xenograft Weights	Data 1

Column B	A549+PBS
vs.	vs.
Column A	A549+ECM
Unpaired t test	
P value	0.0261
P value summary	*
Significantly different? (P < 0.05)	Yes
One- or two-tailed P value?	Two-tailed
t, df	t=2.372 df=24
How big is the difference?	
Mean ± SEM of column A	0.2346 ± 0.03717, n=13
Mean ± SEM of column B	0.1362 ± 0.01849, n=13
Difference between means	-0.09846 ± 0.04151
95% confidence interval	-0.1841 to -0.01278
R squared	0.1899
F test to compare variances	
F,DFn, Dfd	4.043, 12, 12
P value	0.0223
P value summary	*
Significantly different? (P < 0.05)	Yes

TABLE XIX
XENOGRAFT WEIGHT STATISTICAL ANALYSIS

Trans-well Migration					
Number of families	1				
Number of comparisons per family	6				
Alpha	0.05				
		95% CI of diff.	Significant	Summary	Adjusted P Value
Tukey's multiple comparisons test	Mean Diff.				
A549 FN vs. A549 IMR90		-173.8 to -88.15	Yes	**	0.0021
A549 FN vs. A549 WI38	-131	-222.1 to -	Yes	**	0.0081

		69.90			
A549 FN vs. A549 HDF	-146	-162.9 to -101.6	Yes	***	0.0008

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Publications

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Poster Abstracts and Presentations

Michael Scherzer, Morgan Stathem, Patrick Roddy, Leah Siskind, Levi Beverly. An Unbiased Approach to Assess the Role of Sphingolipid Metabolism in Cancer Biology. 7th Symposium: Mechanisms and Models of Cancer, 2013. La Jolla, CA.

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Kumar Saurabh, **Michael Scherzer**, Parag Shah, Alice Mims, Andrew Kraft, Levi Beverly. PIM Kinases: a Small Family with a Big Role in Myeloid Leukemia. Tenth International Workshop on Molecular Aspects of Myeloid Stem Cell Development and Leukemia, 2014. Cincinnati, OH.

Michael Scherzer, Patricia Soucy, Levi Beverly. Using Fibroblast-derived 3D matrices to dissect Micro Environmental Cues Critical for Lung Cancer Progression and Metastasis. Research! Louisville, 2014. Louisville, KY.

Teaching Experience

Artificial Organs (BE 683)

A lecture on hearing physiology and pathology and biomedical devices for therapy

Micro-Electrical-Mechanical-Systems (BE 680)

A lecture on Controlling the Cellular Microenvironment, A topic covered from chapters of George Eng and Milicia Radisic

Biomaterials (BE 650)

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Honors

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