Glutamate-Handling Proteins Associate with Adverse Clinicopathologic Features and Comorbidities in Invasive Lobular Carcinoma

Shaymaa Bahnassy[1,9]*, Todd A. Young**[1]***, Theresa C. Abalum[1,2], Eden A. Pope[1,3], Amanda Torres Rivera[1], Aileen I. Fernandez[1,4], Ayodeji O. Olukoya[1], Dua Mobin[1], Suman Ranjit[5], Nicole E. Libbey[1], Sonali Persaud[1], Aaron M. Rozeboom[1], Krysta Chaldekas[1], Brent T. Harris[1,6], Zeynep Madak-Erdogan[7], Joseph L. Sottnik[8], Matthew J. Sikora[8], Rebecca B. Riggins[1,9]

[1]Department of Oncology, Lombardi Comprehensive Cancer Center, Georgetown University, Washington, DC 20057
[2]Towson University, Towson, MD 21252
[3]Wake Forest University, Winston-Salem, NC 27109
[4]Department of Pathology, Yale School of Medicine, New Haven, CT 06520
[5]Department of Biochemistry and Molecular & Cellular Biology, Georgetown University Medical Center, Washington, DC 20057
[6]Departments of Neurology and Pathology, Georgetown University Medical Center, Washington, DC 20057
[7]Department of Food Science and Human Nutrition, Cancer Center at Illinois, Division of Nutritional Sciences, University of Illinois Urbana-Champaign, Urbana, IL 61801
[8]Department of Pathology, University of Colorado Anschutz Medical Campus, Aurora, CO 80045

[9]Corresponding authors: Shaymaa Bahnassy, PhD, and Rebecca B. Riggins, PhD

Email: sb1966@georgetown.edu, and rbr7@georgetown.edu

Phone: 202 687 1260

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*These authors contributed equally to this work.

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Running title: Glutamate transport and metabolism in ILC

Abstract

Invasive Lobular Carcinoma (ILC) is a subtype of breast cancer characterized by distinct biological features, and limited glucose uptake coupled with increased reliance on amino acid and lipid metabolism. Our prior studies highlight the importance of glutamate as a key regulator of ILC tumor growth and therapeutic response. Here we examine the expression of four key proteins involved in glutamate transport and metabolism – SLC3A2, SLC7A11, GPX4, and GLUD1/2 – in a racially diverse cohort of 72 estrogen receptor-positive (ER+) ILC and 50 ER+ invasive ductal carcinoma, no special type (IDC/NST) patients with primary disease. All four proteins associate with increased tumor size in ILC, with three showing stronger associations in Black women, but not in IDC/NST. Among these three proteins in ILC, GLUD1/2 uniquely associates with ER expression in all women, while GLUD1/2 and SLC3A2 are enriched in hypertensive women. GLUD1/2 and GPX4 are upregulated in endocrine therapy-resistant ILC cell lines, and pharmacological inhibition of GLUD1 reduces ER protein levels and cell viability. Together, these findings support a potentially important role for glutamate metabolism in ILC and suggest GLUD1 and other glutamate-handling proteins as candidate targets for therapeutic intervention in ILC.

Keywords

Invasive lobular carcinoma, glutamate metabolism, GLUD1, disparities

Introduction

ILC (Invasive Lobular Carcinoma) is a special histological subtype of breast cancer that accounts for 10-15% of cases diagnosed annually[1,2]. ILC has unique genetic, transcriptomic, and biological features as compared to the more common invasive ductal breast cancer, or breast cancer of no special type (IDC/NST). Nearly all ILC is estrogen receptor-positive (ER+), and characterized by mutational inactivation of *CDH1*, the gene encoding E-cadherin[3], which is associated with the diffuse growth pattern of ILC tumors. Additionally, ILC metastasizes to the lung, bone, and brain like other ER+ breast malignancies, but it also has a propensity to spread to the digestive tract, ovaries, and the peritoneal cavity[1,2]. Complicating this atypical dissemination pattern, advanced or metastatic ILC may not be as easily detected by positron emission tomography using the glucose analog tracer ^{18}F-fluorodeoxyglucose (i.e. [^{18}F]FDG-PET)[4], as several clinical studies have shown ILC has limited FDG avidity. While these observations complicate care of patients with metastatic ILC, they also suggest that ILC has a distinct metabolic phenotype among breast cancers that is less reliant on glucose uptake and metabolism.

Other PET tracers provide opportunities to understand differential ILC metabolism. Fluciclovine is an amino acid analog of leucine[5], and prospective clinical trials demonstrate that ILC shows substantially higher [^{18}F] Fluciclovine-PET uptake compared to [^{18}F]FDG-PET][6–8]. Among ILC, 100% of lesions were fluciclovine-positive while only 43% of lesions were FDG-positive. These findings are consistent with clinical and preclinical research, in which we and others show that ILC is more dependent on lipid and amino acid metabolism compared to glucose metabolism[9–12]. Our work suggests that the amino acid glutamate may be particularly important in ILC. Endocrine therapy-resistant ILC models upregulate multiple metabotropic glutamate receptors (mGluRs)[13], and the small molecule Riluzole (a modulator of glutamate metabolism FDA approved for amyotrophic lateral sclerosis) is efficacious against *in vitro, in vivo*, and *ex vivo* models of ILC[14,15], (discussed in[16]). Therefore, understanding glutamate metabolism in ILC can provide important opportunities to advance imaging and clinical management, as well as precision treatment paradigms.

For this study, we assembled a cohort of 72 ER+ primary ILCs and 50 ER+ primary IDC/NSTs with well-annotated clinical, pathological, and demographic data, and used multiplex immunohistochemistry (IHC) to determine the prognostic value of four key proteins in glutamate transport and metabolism (**Figure 1A**). Solute carriers (SLC) 3A2 (SLC3A2 or CD98) and 7A11 (SLC7A11) heterodimerize to form System xCT, an antiporter that imports cystine and exports glutamate. Downstream of xCT, glutathione peroxidase 4 (GPX4) protects cells from lipid peroxidation by catalyzing the reduction of lipid hydroperoxides and converting the glutamate-glycine-cysteine tripeptide glutathione (GSH) to glutathione disulfide (GSSG), thereby inhibiting ferroptosis (an iron-dependent mechanism of cell death). Finally, glutamate dehydrogenase 1/2 (GLUD1/2 or GDH1/2) is a bidirectional metabolic enzyme that in the forward direction converts glutamate to alpha-ketoglutarate to replenish the tricarboxylic acid (TCA) cycle, while in the reverse direction it recycles ammonia to bolster tumor cell growth and amino acid synthesis[17,18].

xCT promotes metastasis and therapy resistance in multiple cancers, including breast cancer[19]. As a negative regulator of ferroptosis downstream of xCT, GPX4 expression in cancer and stromal cells is also associated with disease progression and invasiveness[20–22]. GLUD1/2 has recently emerged as a key node connecting glutamate metabolism, xCT function, and ferroptosis[23–25]. However, the contribution of these proteins to ILC biology, prognosis, and clinicopathologic features including tumor characteristics and comorbidities, is not known, despite mounting evidence highlighting the importance of glutamate metabolism in this breast cancer subtype.

Methods

Use and Re-Use of Publicly Available and Previously Published Data. mRNA expression data from invasive lobular breast cancer (ILC), mixed ductal-lobular breast cancer (mDLC), and breast cancer of no special type (NST) from The Cancer Genome Atlas (TCGA, RRID:SCR_003193) pan-cancer clinical data resource[26], METABRIC[27], and the Sweden Cancerome Analysis Network - Breast (SCAN-B)[28,29] shown in **Figure S1A** were accessed via Gene Expression eXPLORER (GEXPLORER) at https://leeoesterreich.org/resources on 6/4/2024. mRNA expression data and associated clinicopathological data from ILC and IDC/NST from TCGA shown in **Figure 4C** were accessed via cBioPortal[30,31] (RRID:SCR_014555) on 6/13/2024. mRNA expression data for *GLUD1* in breast cancer cell lines from DepMap release 25Q2 (https://depmap.org/portal) in **Figure S2A** were accessed on 8/27/2025. GLUD1 peptide counts from ER rapid immunoprecipitation mass spectrometry of endogenous proteins (RIME) in ILC cell lines shown in **Figure 6B** were graphed from Supplemental File 6 of Sottnik et al[32].

Primary Breast Cancer Cohorts and Tissue Microarray (TMA) Construction. The ER+ IDC/NST cohort and TMA construction has been previously described[33]. For the ILC cohort, we similarly constructed a TMA with $\geq$2 cores per patient from n=72 patients with primary ER+ ILC (Inclusion criteria: female, $\geq$ 10% ERα positivity, PR+/−, HER2 amplification negative, and had surgery for primary breast cancer at MedStar Georgetown University Hospital). Exclusion criteria were patients diagnosed with carcinoma *in situ* only, known *BRCA* or other familial mutation carriers, and evidence of neoadjuvant therapy. All patients were consented through the Histopathology and Tissue Shared Resource, (HTSR), the Survey, Recruitment, and Biospecimen Shared Resource (SRBSR), or IndivuMed under the following respective Georgetown University IRB protocols: 1992-048, Pr0000007, and 2007-345. Demographic, clinical, and pathological data, vital status, and follow-up time (vital status and follow-up time updated July 2021) are shown in **Table 1**.

OPAL Multiplex IHC and Image Processing. Duplicate sections of each TMA, taken from different depths, were subjected to multiplex IHC conducted on the Vectra3 multispectral imaging platform (Akoya Biosciences, Marlborough, MA) using OPAL chemistry as previously described[14]. Antibody/OPAL pairing and staining order was empirically determined by several criteria, including spectrally separating co-localized markers and separating spectrally adjacent dyes. We first performed singleplex IHC with the chosen antibody/OPAL dye pair to optimize signal intensity values and gauge proper cellular expression, followed by optimizing the entire multiplex assay. Primary antibodies in the panel, OPAL pairing, and other methodological details are shown in **Supplemental Table 1**. Image scanning, spectral unmixing, tissue/cell segmentation, phenotyping, and analysis on the Vectra 3.0 Automated Quantitative Pathology Imaging System using Phenochart and inForm 2.4.6 (PerkinElmer/Akoya) was conducted as previously described[14]. Data (percent cell positivity) from replicate cores and duplicate TMA sections for each tumor were averaged.

Cell Culture and Reagents. Cell lines were maintained at 37°C in a humidified incubator with 95% air and 5% carbon dioxide (CO_2). Cell lines were authenticated by short tandem repeat (STR) profiling, and regularly tested to ensure they remained free of *Mycoplasma spp*. contamination, by the Tissue Culture and Biobanking Shared Resource (TCBSR). The following ER+ ILC cell lines were cultured as previously described[14,34,35]: SUM44PE (RRID:CVCL_3424) and its tamoxifen-resistant derivative LCCTam; and MDA-MB-134VI (RRID:CVCL_0617) and its long-term estrogen-deprived (LTED) derivatives LTED-A, -B, -D, and -E. General cell culture supplements and reagents were purchased from ThermoFisher (Grand Island, NY) or Sigma

Aldrich (St. Louis, MO). The glutamate dehydrogenase inhibitor R162 (Cat# 538098) and Poly-L-
lysine hydrobromide (PLL; Cat# P6282) were purchased from Sigma Aldrich.

Immunoblotting. Assays were conducted as previously described[14] and whole-cell lysates were
probed with the following primary antibodies: SLC3A2 (1:1000, Santa Cruz Biotechnology Cat#
sc-376815, RRID:AB_2938854); SLC7A11 (1:500, Abcam Cat# ab175186, RRID: AB_2722749);
GLUD1/2 (1:1000, Cell Signaling Technology Cat# 12793, RRID:AB_2750880); GPX4 (1:1000,
Abcam Cat# ab125066, RRID: AB_10973901); ER alpha (clone D8H8; 1:1000, Cell Signaling
Technology Cat# 8644, RRID:AB_2617128); and beta-actin (1:2000, Cell Signaling Technology
Cat# 3700, RRID:AB_2242334). ImageJ[36] was used to perform densitometry analysis.

Cell Viability Assays. Cell viability was assessed using the trypan blue exclusion assay. MM134
parental and LTED-D cells were seeded at a density of 100,000 cells/well in 24-well plates and
incubated for 24 and 48 hours, respectively, before treatment with the indicated concentrations of
R162 for 72 hours. After treatment, cells were trypsinized, stained with trypan blue, and live cells
were counted using the Countess II Automated Cell Counter (ThermoFisher Scientific).

Metabolic Imaging of NADH Autofluorescence, Fluorescent Lifetime Imaging (FLIM)
Instrumentation, and Phasor Analysis. MM134 and MM134 LTED-D cells were seeded on PLL-
coated 22 mm^2 diameter glass coverslips (VWR, Cat# 48380-046) in 6-well plates at 150,000
cells/well. Forty-eight hours later, cells were treated with DMSO or 10 µM R162 for 12 hours
before transferring coverslips to 35 mm^2 diameter glass bottom dishes (MatTek, Cat# P35GCOL-
1.5-14-C) and performing live-cell imaging on the FVMPE-RS multi-photon, laser-scanning
microscope (Olympus, Waltham, MA) with deep imaging via emission recovery (DIVER) detector
in the Microscopy and Imaging Shared Resource (MISR). Cells were imaged with a 40X water
objective (Olympus, Waltham, MA) using a 740 nm laser from an InsightX3 laser. The signal was
collected using the DIVER detector in the forward direction that is connected to a FastFLIM
acquisition card (ISS, Champaign IL).

The FLIM data acquisition and analysis was carried out using GSlab[37]. The fluorescence lifetime
of each pixel in the autofluorescence intensity image was transformed into a phasor plot. These
phasor plots were color-coded based on the phase angle, with longer lifetimes appearing in red–
yellow hues and shorter lifetimes in blue–cyan (**Figure S2B**).

To isolate cytoplasmic and mitochondrial regions, a histogram-based intensity correction was
applied, excluding nuclear areas due to their lower fluorescence intensity. Fractional intensity of
free NADH was calculated using a multiplexing algorithm previously described[14,37–41]. Three
phasor components were selected for this analysis: free NADH (0.4 ns), bound NADH (3.4 ns),
and background (0.0 ns)[42,43]. Following calculation, individual fractional intensity maps were
exported both as data files and as binary images, where fractions ranging from 0 to 1 were scaled
to a grayscale range of 0–255. Bound and free NADH fraction images were then merged using
ImageJ[36] to visualize metabolic activity distribution. For enhanced visualization, the color scale
was adjusted so that bound NADH (purple) was restricted to a range of 140–255, and free NADH
(yellow) to a range of 1–35. This adjustment was necessary due to the predominance of bound
NADH in these cell types.

Metabolic Assays. Non-mitochondrial oxygen consumption rate was determined using Seahorse
assays. MM134 parental and LTED-D cells were seeded at 80,000 cells/well in PLL–coated XF96
cell culture microplates (Agilent, Cat# 103793-100). After 24 h, cells were treated with DMSO or
R162 (10 or 20 µM) for 12 hours. Media was then replaced with Seahorse XF DMEM

supplemented with 10 mM glucose, 2 mM glutamine, and 1 mM sodium pyruvate (Agilent, Cat#
103680-100), followed by 1 hour incubation at 37 °C in a non-CO_2 incubator. Metabolic flux
measurements were collected under basal conditions and after sequential injection of the
Seahorse Mito Stress Test reagents (Agilent, Cat# 103015-100) at final concentrations of
oligomycin (1.5 µM), FCCP (2 µM), and rotenone/antimycin A (0.5 µM), using the Seahorse
XFe96 Analyzer (Agilent).

Extracellular lactate was measured using the Lactate-Glo Assay kit (Promega, Cat# J5021) per
manufacturer's instructions. MM134 parental and LTED-D cells were seeded at 25,000 cells/well
in 96-well plates precoated with PLL. Parental cells were treated 24 hours after seeding, whereas
LTED-D cells were treated 72 hours after seeding, with DMSO or R162 at 10 µM or 20 µM. Media
samples were collected 12 hours after treatment. For analysis, 3 µL of media was diluted in PBS
(1:200 for parental, 1:300 for LTED-D) and samples were mixed 1:1 with the lactate detection
reagent and read on a microplate reader (EnSpire, Perkin Elmer).

Statistical Analyses. GraphPad Prism 10 (Boston, MA, RRID:SCR_002798) was used for
statistical analyses. Mann Whitney tests were used to analyze and compare protein expression
in lobular vs. ductal tumor (panCK+) and stromal cells (panCK-), and by lymph node-positive vs.
negative status. Simple linear regression and Spearman ρ correlation analyses were used to
compare protein expression versus tumor size or patient age, and co-expression between
proteins in tumor and stroma. Kaplan-Meier analyses and log-rank tests were used for overall
survival analysis. Fisher's exact and Chi-squared tests were used to analyze demographic,
clinical and pathological data where appropriate, and the relationship between comorbidity or race
and high vs. low (above vs. below the median) protein expression. Unpaired t-tests and one-way
analysis of variance (ANOVA) with Dunnett's multiple comparison tests were used to analyze cell
viability, bound/total NADH from FLIM, Seahorse, and extracellular lactate assays, as appropriate.
All relevant tests were two-sided and statistical significance was defined as $p \leq 0.05$. Asterisks
denote statistical significance as follows: *, $p \leq 0.05$; **, $p \leq 0.01$; ***, $p \leq 0.001$; ****, $p \leq 0.0001$.

Results

Study population and cohort comparisons. We constructed a series of TMAs of ER+ ILC (n=72) and IDC/NST (n=50) for staining with our custom multiplex IHC panel (**Figure 1B**). Our cohort is unique in having 34% of patients self-report their race as non-white, including 25% identifying as Black or African American, with equal representation in the ILC and IDC/NST cohorts **(Table 1)**. There has historically been a paucity of data on ILC from non-white patient populations. However, recent studies report that the average annual percent change in ILC incidence has increased across all racial groups[44], with a higher increase among non-Hispanic Black women than for non-Hispanic white women in the past few decades[45]. Preliminary evidence suggests that Black women with ILC have significantly lower rates of 5-year breast cancer-specific survival and are less likely to undergo surgery, despite clinically presenting with locally advanced disease[46].

Prior retrospective analyses suggest that ILC occurs primarily in women >60y, which may reflect challenges in detecting this histological subtype. Another distinct feature of our cohort is that patients with ILC were <60y and younger than those with IDC/NST at diagnosis (54 vs 59y, **p=0.003). Likely consistent with younger age and a numerical trend towards shorter duration of follow-up, a significantly greater proportion of ILC patients were alive at last contact (85%, vs 64% for IDC/NST, ***p=0.001). Other clinicopathological features including nodal status and treatment modality were similar across cohorts. Although all tumors met our inclusion criteria for ER+ and HER2 amplification negative, we further assessed ER, progesterone receptor (PR), and HER2 expression by IHC (Ki67 data were not available) from the original pathology reports. HER2 protein expression (IHC 0 vs. 1+ to 3+) was more common in ILC vs. IDC/NST (Fisher's exact test *p=0.02). In ILC, percent PR+ was higher but not significant (71% vs. 45% for IDC/NST, p=0.07), whereas percent ER+ was lower (83.5% vs. 91% for IDC/NST, *p=0.02), consistent with a prior study that demonstrated differential ER protein turnover in ILC vs. IDC/NST[47].

Expression of glutamate-handling proteins in ILC vs. IDC/NST. We used three well-characterized public datasets (TCGA, METABRIC, and SCAN-B)[26–29] to compare mRNA expression of the four target proteins in our multiplex IHC panel in ILC, IDC/NST, and (where reported) mixed mDLC (**Figure S1A**). *SLC7A11* and *SLC3A2* expression are both significantly greater in IDC/NST compared to ILC in multiple datasets. This may in part be due to the fact that these analyses do not stratify by hormone receptor status, and xCT is often overexpressed in triple negative breast cancer[48], which is more prevalent in IDC/NST. By contrast, *GPX4* expression is significantly greater in ILC vs. IDC/NST in the METABRIC dataset, while *GLUD1* expression is significantly greater in ILC vs. IDC/NST in both the METABRIC and SCAN-B datasets. Prior studies in models of mammary epithelial cells show GLUD1 expression is elevated under quiescent and slower growth conditions[49], which is generally consistent with ILC's more indolent nature.

In our TMAs, we compared the abundance of each target protein expressed, as a percentage of positive cells, in ILC vs. IDC/NST within the tumor (pan-cytokeratin-positive, panCK+ cells) and stromal (panCK- cells) compartments (**Figure 1C**). Consistent with public mRNA data, but despite all tumors being ER+, SLC7A11 and SLC3A2 protein expression in tumor cells is more abundant in IDC/NST (40-45%) compared to ILC (<25%). In our cohort, GLUD1/2 expression in tumor cells is also significantly greater in IDC/NST, though expression in both cohorts is $\geq$50%, while there is no significant difference for GPX4 (~50% positive cells in both cohorts). In stromal or non-epithelial cells, SLC3A2 expression is significantly more abundant in IDC/NST, while GPX4 expression is significantly more abundant in ILC. Together these data suggest that while all four

glutamate-handling proteins are expressed in ILC across multiple datasets and sample types, the
intracellular enzymes GLUD1/2 and GPX4 are generally more prevalent in ILC vs. IDC/NST.

All four glutamate-handling proteins are each associated with increased tumor size in ILC.
We assessed the relationship between the abundance of each protein in the tumor and stromal
compartment, and known breast cancer prognostic factors, including lymph node status, age, and
tumor size at diagnosis, using linear regression and Spearman ρ. We find no statistically
significant relationship between any of these glutamate handling proteins and either lymph node
status or age as a continuous variable (data not shown). However, SLC7A11, SLC3A2, GLUD1/2,
and GPX4 expression in tumor cells is each positively and significantly associated with increased
tumor size in ILC, but not IDC/NST (**Figure 2A, B**). GLUD1/2 and GPX4 expression in the stroma
is also significantly correlated with larger tumors in the ILC cohort. Since larger tumor size is linked
to shorter overall survival (OS) in breast cancer (e.g.[50]), we tested whether high vs. low (above
vs. below median) protein expression for each target is associated with OS in our ILC and
IDC/NST cohorts, which have a median follow-up time of 8.4 and 10.4 years, respectively. Neither
individual (data not shown) nor combined expression of the four proteins is significantly
associated with OS in either cohort (**Figure 2C**). Together these data highlight a specific
association between each member of this network of glutamate-handling proteins and increased
tumor size in ILC, but not IDC/NST.

Co-expression of glutamate-handling proteins in ILC. Since SLC3A2, SLC7A11, GPX4, and
GLUD1/2 are all functionally related (Figure 1A), we assessed whether they are co-expressed –
e.g. do ILC with a high percentage of cells expressing one of the proteins also have a high
percentage of cells expressing one or more of the others? There is moderate to strong significant
co-expression amongst all four proteins in ILC (**Figure 3A**). For example, in ILC, tumor cell GPX4
abundance is significantly associated with expression of every other protein in both the tumor and
the stroma. There is also a significant positive correlation between tumor or stromal GLUD1/2 and
ER expression. By contrast, in IDC/NST, tumor cell GPX4 abundance is only significantly
positively associated with itself and SLC7A11 in stromal cells, and there is no correlation of any
protein with either ER or PR expression (**Figure 3B**). These data show that the known functional
relationships between these glutamate-handling proteins translate to increased co-expression in
ILC, where all members of this network are significantly associated with increased tumor size.

GLUD1/2, GPX4, and SLC3A2 are strongly associated with increased tumor size in Black
women with ILC. Racial disparities in breast cancer outcomes are widely discussed in the context
of triple negative breast cancer. However, Parab et al.[51] recently published that non-Hispanic
Black or African-American women are significantly more likely to have aggressive ER+ breast
tumors (defined by a high-risk recurrence score on the Oncotype DX panel) when compared to
non-Hispanic white women. Van Alsten et al.[52] show similar data for the PAM50-based risk of
recurrence score, wherein young (<50 years of age) Black women with ER+ breast cancer are
more likely to have intermediate to high scores as compared to young white women. Consistent
with this, Rauscher et al.[53] report the risk of death from ER+/PR+ breast cancer is more than four
times higher for Black women as compared to white women, even after adjustment for tumor
stage, grade, and delays in initiation of treatment, while a meta-analysis by Torres et al. shows
Black women have a 50% higher relative risk of death from ER+ breast cancer[54].

Unfortunately, these (and most other) studies of racial disparities in ER+ breast cancer do not
report the presence or absence of lobular histology, a point that is complicated by our poor

understanding of the true prevalence of ILC in Black women. Analysis of publications that provide
such data[55–60] indicates that the prevalence of ILC in Black women ranges from 5.5% to 16% of
breast cancer cohorts – a rate that, in some of these studies, is significantly lower compared to
non-Hispanic white women within the same cohort (**Supplemental Table 2**). Consistent with this,
Giaquinto et al. report using Surveillance, Epidemiology, and End Results (SEER) data that the
age-adjusted incidence rate of ILC for Black women is 11 per 100,000 breast cancer cases vs.
14.7 for non-Hispanic white women[44]. It is important to note that all of these studies may be
confounded by variability in diagnostic criteria for ILC in all populations[61,62].

Nevertheless, with sixteen out of 72 (22%) and fifteen out of 50 (30%) women in our ILC and
IDC/NST cohorts, respectively, self-reporting their race as Black or African American (none self-
report Hispanic ethnicity), we assessed the relationship between the abundance of each protein
in the tumor and stromal compartments with tumor size specifically in Black women. Tumor and
stromal cell GLUD1/2, GPX4, and SLC3A2 abundance are each positively and significantly
associated with increased tumor size in ILC, but not IDC/NST (**Figure 4A**). For all three proteins,
the association with tumor size is markedly stronger in Black women with ILC than in the full ILC
cohort (**Figure 4A and 2B**). Although not statistically significant, high individual (data not shown)
and combined expression of the four glutamate-handling proteins shows a trend toward worse
OS in Black women with ILC (**Figure 4B**).

Adding to our poor understanding of the ILC biology in Black women is the limited racial diversity
in publicly available datasets. METABRIC and SCAN-B report zero Black women with ILC, and
current TCGA data report only thirteen (6.5% of the ILC cohort). We first analyzed *GLUD1* mRNA
expression in relation to tumor stage in Black women with ILC or with luminal IDC/NST in TCGA
(**Figure 4C**), since this enzyme has the strongest correlation with tumor size (Spearman ρ=0.61)
and is uniquely associated with ER expression in our ILC cohort (**Figure 3A**). Despite the low
sample number, there is a clear, stage-dependent increase in *GLUD1* expression in tumors from
Black women with ILC, though not statistically significant (**Figure 4C**). By contrast, there is a
significant decrease in *GLUD1* expression in Stage II vs. Stage I luminal IDC/NST in Black
women. These stage-dependent trends are not observed for *GPX4* or *SLC3A2* mRNA expression
in Black women of either cohort (data not shown). Gene set enrichment analysis (using Enrichr[63–
65]) of the top 100 genes positively correlated with *GLUD1* expression in TCGA tumors from Black
women with ILC show an enrichment for glutamine and glutamate metabolism pathways (KEGG,
q=0.011). Tumors from Black women with ILC also show significant enrichment for the synthesis
of arginine (KEGG, q=0.011), N-glycans (KEGG, q=0.046), and uridine diphosphate N-
acetylglucosamine (UDP-GlcNAc, Metabolomics Workbench Metabolites, q=0.018). With each of
these ILC-enriched pathways independently linked to breast cancer progression and metastatic
potential (e.g.[66,67]), this analysis further highlights the potential importance of GLUD1/2 for Black
women with ILC.

Enrichment of GLUD1/2 and SLC3A2 in women with ILC and hypertensive comorbidities.
Multiple comorbid conditions like hypertension, obesity and metabolic syndrome are linked to poor
outcomes in breast cancer, and there are racial disparities in the prevalence of these
comorbidities[68–71]. Abstraction of electronic health records show no difference in the presence or
absence of total comorbidities by ICD-9-CM diagnosis codes (details in **Supplemental Table 3**).
However, obesity is significantly enriched among patients with IDC/NST , while hypertension is
the most prevalent comorbidity in both the ILC and IDC/NST cohorts, with no statistically
significant difference between them (**Figure 5A**). Next, we compared comorbidity presence or

absence between Black and white women. Significantly more Black women have at least one comorbid condition in both cohorts (**Figure 5B**). Of note, hypertension is significantly enriched among Black women with ILC, but not IDC/NST (**Figure 5C**). Among women with ILC, high expression of GLUD1/2 and SLC3A2 (above median) is enriched in hypertensive patients in both the full and Black cohorts (**Figure 5D** and **5E**), indicating an association between hypertension and elevated expression of glutamate-handling proteins in ILC.

Pharmacological inhibition of GLUD1 reduces ER protein levels, cell viability and oxidative phosphorylation in ILC cell lines. Breast cancer preclinical models are not racially or ancestrally diverse. Nevertheless, we used two of the best-characterized ER+ ILC cell line models and their endocrine therapy-resistant derivatives[34,35], to evaluate the expression of the four target proteins by immunoblotting (**Figure 6A**). Tamoxifen-resistant LCCTam cells show modest (<2-fold) upregulation of SLC7A11, GLUD1/2, and GPX4 compared to the parental SUM44 cells. However, three of the four long-term estrogen-deprived cell lines (LTED, mimicking aromatase inhibitor resistance) derived from parental MDA-MB-134VI (MM134) cells exhibit robust, 2- to 6-fold upregulation of GPX4 and GLUD1/2, suggestive of a role for these enzymes in aggressive or more advanced ILC.

Given this pattern, we next focused on GLUD1 to extend our novel observations that GLUD1/2 expression is: 1) markedly increased in endocrine therapy-resistant variants of ER+ ILC cells (**Figure 6A**); 2) significantly associated with increased tumor size and ER expression in our ILC patient cohort (**Figures 2-4**); and **3**) upregulated with increasing stage in tumors from Black women with ILC in TCGA (**Figure 4C**). Among breast cancer cell lines profiled by DepMap[72], ER+ ILC and ILC-like cell lines (as defined in[73]) rank among the highest for *GLUD1* mRNA expression (**Figure S2A**). Given its dual roles in TCA cycle anaplerosis and ammonia recycling, GLUD1/2 is predominantly localized in the mitochondria. However, a sub-population of GLUD1 has been reported in the nucleus, where it is involved in chromatin regulation and gene transcription in a catalytic activity-dependent manner[74,75]. Proteomic profiling of the nuclear ER interactome with ER RIME (rapid immunoprecipitation mass spectrometry of endogenous proteins) by Sottnik et al.[32] identified GLUD1 as one of the top ILC-specific ER-associated proteins (**Figure 6B**) that was not identified in IDC/NST cells[76]. To examine the potential role of GLUD1 in regulating ER expression, we treated four ER+ ILC cell lines with the GLUD1 inhibitor R162. Short-term treatment with R162 reduces ER expression by 35-86% in all cell lines tested (**Figure 6C**), including endocrine therapy-resistant cell lines.

To assess whether GLUD1 can be a useful target for therapeutic intervention in ILC, we performed cell viability assays and show that R162 significantly reduces cell viability of parental MM134 and its endocrine therapy-resistant derivative LTED-D cells in a dose-dependent manner (**Figure 6D**). Previous studies showed that R162 treatment attenuates GLUD1's pro-tumorigenic functions in multiple malignancies by reducing alpha-ketoglutarate levels and inducing cell death by blocking mitochondrial oxidative phosphorylation (OXPHOS) and/or increasing reactive oxygen species (ROS)[17,77–80]. To evaluate potential metabolic reprogramming in response to GLUD1 inhibition in ILC cells, we used an innovative metabolic imaging approach that combines fluorescence lifetime imaging (FLIM) with phasor analysis (**Figure 6E, S2B**). FLIM leverages autofluorescence of the reduced form of nicotinamide adenine dinucleotide (NADH) and permits separation of protein-bound vs. free NADH, which correlates with OXPHOS vs. glycolytic metabolism, respectively[14,39,81]. In both parental MM134 and LTED-D cells, R162 significantly reduces the proportion of protein-bound NADH, suggestive of reduced OXPHOS (**Figure 6F**). To confirm these results, we performed conventional metabolic assays. R162 concomitantly increases non-

410   mitochondrial oxygen consumption and extracellular lactate production (**Figure 6G, 6H**), each of
411   which can be a measure of reduced OXPHOS and increased ROS associated with mitochondrial
412   dysfunction[82–84]. Altogether, GLUD1 pharmacological inhibition suppresses cell viability, reduces
413   ER protein expression, and reprograms metabolism in multiple ILC cell lines, providing preliminary
414   evidence that inhibition of GLUD1 may offer therapeutic benefit in ILC.

Discussion

ILC has emerged as a breast cancer histologic subtype that is more reliant on amino acid and lipid metabolism than glucose metabolism. Our prior studies specifically highlight the importance of glutamate signaling in ILC[13,14]. Here, we provide evidence for a relationship between a focused network of four glutamate-handling proteins (SLC3A2, SLC7A11, GPX4, and GLUD1/2) and increased tumor size in primary ILC, but not primary IDC/NST. GLUD1/2, GPX4, and SLC3A2 expression in tumor and stroma show a markedly stronger relationship with increased tumor size in Black women with ILC. We further demonstrate that GLUD1 and GPX4 are upregulated in ILC endocrine therapy-resistant preclinical models, and that pharmacological inhibition of GLUD1 reduces cell viability, in part by reducing ER protein levels and OXPHOS.

All four glutamate metabolism and transport proteins measured here each have a significant positive relationship with larger tumor size, and with each other, in ILC. Their high degree of co-expression results in multicollinearity that confounds multiple linear regression models, therefore, we do not report the results of these analyses. The strong association of individual and combined protein expression with tumor size, but lack of a significant association with overall survival in our cohort, could be explained by limited statistical power to detect survival differences in a cohort of this size. Larger studies are needed to better assess these relationships. Future comprehensive *in vitro* and *in vivo* preclinical studies will also be necessary to determine if targeting one protein impacts expression or function of the others in ILC.

This interdependency is particularly relevant for assessing the role of SLC7A11 in ILC, since it not only heterodimerizes with SLC3A2, but this heterodimer acts directly upstream of the ferroptosis suppressor GPX4 (discussed further below). ER can transcriptionally upregulate SLC7A11, which is further elevated in endocrine therapy-resistant ER+ breast cancer[85]. However, these studies by Cao et al. were conducted in preclinical models of IDC/NST, and future studies will be needed to specifically address the role of SLC7A11 in ILC.

In addition to its association with tumor size, high expression of SLC3A2 is enriched among hypertensive women in our full and Black ILC cohort. SLC3A2 heterodimerizes with SLC7A8 (LAT2), and genetic variation in these genes has been linked to hypertension risk, supporting a role for altered amino acid transport in regulating renal dopamine synthesis, sodium handling, and blood pressure[86]. Upregulation of SLC3A2 is also associated with poor prognosis in multiple cancers[21,87,88], including certain highly proliferative, c-Myc-driven breast cancers[22]. Recent studies show SLC3A2 is required for the growth of tamoxifen-resistant ER+ breast cancer cells[89]. Importantly, SLC3A2 is an obligate heterodimeric partner for multiple L-type transporters beyond SLC7A11 and SLC7A8, positioning this protein as a crucial node in a broader amino acid transport network. For example, heterodimerization of SLC3A2 with SLC7A5 (LAT1) is essential for pro-survival amino acid transport in multiple preclinical cancer models[90]. Independent of amino acid transport, SLC3A2 acts as a beta1 and beta3 integrin co-receptor that is implicated in cell adhesion and mechanotransduction[91], both of which are compromised in ILC as a result of *CDH1* mutational inactivation. The relative contribution of SLC3A2 to amino acid transport-dependent and -independent functions in ILC remains to be determined.

GPX4 plays a key role in detoxifying lipid peroxides that accumulate in cells, thereby preventing ferroptosis[92,93]. Recent work by Hu et al. shows that endocrine therapy can sensitize ER+ IDC/NST cell lines to ferroptosis induction by GPX4 inhibitors[94], while in tamoxifen-resistant IDC/NST cell lines, activation of the RelB arm of the NF-kB pathway inhibits ferroptosis by

upregulating GPX4[95]. It is not known whether RelB is a transcriptional regulator of GPX4 in ILC,
but we do see an increase in GPX4 expression in tamoxifen-resistant and LTED ILC cells (**Figure**
6A). Of note, GPX4 sits at the intersection of multiple pathways that confer heightened sensitivity
to ferroptosis. Cell-cell interactions mediated by E-cadherin suppress ferroptosis due to activation
of the Hippo pathway, and loss or inhibition of E-cadherin leads to increased vulnerability to
ferroptosis[92,96]. Cell density also influences ferroptosis sensitivity in the presence of wild type E-
cadherin. Cells cultured at low density with few cell-cell contacts are more susceptible to cell
death caused by ferroptosis via GPX4 inhibition than those cultured at high density[93]. In the
stroma, ferroptosis and GPX4 are regulated by metabolic pathways that influence tumor crosstalk
with the potential to alter cancer progression and response to treatment (discussed in[97,98]). With
E-cadherin loss via *CDH1* mutational inactivation and a discohesive growth pattern with few cell-
cell contacts both being hallmarks of ILC that contribute to ferroptosis sensitivity, our data strongly
implicate GPX4 as an attractive target for therapeutic intervention in ILC that requires further
exploration.

GLUD1/2 protein expression in both the tumor and stromal compartments of ILC is strongly
correlated with increased tumor size in Black women in our cohort, while *GLUD1* mRNA increases
with tumor stage in ILC from Black women in TCGA (**Figure 4**). By contrast, there is an inverse
or negative relationship between GLUD1/2 protein expression and tumor size, and *GLUD1* and
tumor stage in luminal IDC/NST, in our cohort and in TCGA data, respectively. Our findings in ILC
contradict the observations of Craze et al.[99], who showed that *GLUD1* mRNA decreases with
tumor grade, while both mRNA and protein are associated with better breast cancer-specific
survival. Their study included ILC cases (~8%), but the analyses did not separate or compare ILC
with IDC; we therefore posit that the reduction of GLUD1 seen with increasing tumor grade in their
study stems from the IDC/NST cases. Also in that work, the IDC/NST category explicitly included
mDLC, which we now appreciate as highly heterogeneous and perhaps a distinct molecular
entity[100]. The strong association of GLUD1/2 with tumor size and stage in Black women, but lack
of a significant association with overall survival in our cohort, could be explained by limited
statistical power to detect survival differences, given the relatively small subset of Black women
with ILC in our cohort. Larger studies will be needed to better assess these relationships.

Findings from our preclinical studies further support the conclusion that GLUD1/2 expression may
confer poor prognosis in ILC. GLUD1/2 is increased in tamoxifen-resistant and LTED ILC cells,
has novel interaction with nuclear ER in ILC cell lines (as demonstrated by RIME[32]), and
pharmacological inhibition of GLUD1 markedly reduces ER protein expression in multiple ILC cell
lines (**Figures 6A-6C**). Placed in the context of current literature demonstrating ER's novel
transcriptional regulatory functions and coregulators in ILC[32,35,101,102], and the role of a nuclear
GLUD1 sub-population in chromatin remodeling and gene transcription[74,75], we propose that
GLUD1:ER interactions may confer ILC-specific ER activities that could be targeted by GLUD1
inhibition, alone or in combination with endocrine therapies. Accordingly, testing whether GLUD1
inhibition synergizes with endocrine therapy and/or reverses endocrine resistance in ILC is
warranted. It is important to note that while we show through biochemical and advanced
microscopy approaches that pharmacological inhibition of GLUD1 in ILC cells leads to a decrease
in OXPHOS, further studies are needed to distinguish these well-studied mitochondrial functions
of GLUD1 from its still-emerging roles in the nucleus. GLUD1 inhibition has also been reported to
induce ferroptosis in other cancers[17]. The simultaneous R162-mediated increase in extracellular
lactate and non-mitochondrial oxygen consumption we observe in ILC cells is suggestive of
oxidative stress, which can lead to ferroptosis[103]. Therefore, like GPX4 (discussed above), we

posit that targeting GLUD1 may also represent a therapeutic opportunity to capitalize on ILC's
hypothesized increased sensitivity to ferroptosis.

There are several limitations to consider in the interpretation of this study. This is a single-
institution, relatively small cohort, despite better proportional representation of Black or African
American women than many other published studies of ILC. Self-reported race is an important
social construct, but ancestry informative marker-guided analyses will be integral to our follow-up
studies, as will access to more comprehensive data on comorbidities, treatments received, and
systemic and social determinants of health that are potential mediators of breast cancer
disparities. This latter point is particularly relevant because several of the glutamate-handling
proteins studied here have mechanistic connections to biological readouts for stress (such as
allostatic load) that is disproportionately experienced by minoritized or marginalized communities,
and/or associated with higher breast cancer risk or poorer outcomes[104–107]. To our knowledge just
two studies have specifically examined systemic and social determinants of health in ILC. Kaur
et al.[108] report that lower socioeconomic status is associated with larger tumor size, higher tumor
grade, and an increased risk of death in ILC, while the increased incidence of ILC reported by
Quinn et al.[45] is particularly acute in areas with the lowest attainment of high school education
(<10%). Our ongoing studies seek to integrate these systemic and social determinants of health
with genomic and transcriptomic patterns in large, diverse, multi-institutional ILC cohorts.

In conclusion, our novel findings implicate an important role for glutamate handling in ILC and
suggest GLUD1 and potentially other glutamate-handling proteins as candidate targets for
therapeutic intervention, warranting further investigation in diverse preclinical and translational
studies.

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Conception and design: TAY, SB, RBR
Provision of study materials: AMR, KC, BTH
Collection and assembly of data: TAY, SB, TCA, EAP, ATR, NH, SP, AIF, DM, AOO, SR, JLS, MJS
Data analysis and interpretation: TAY, SB, TCA, EAP, ATR, AOO, SR, JLS, MJS, ZME
Manuscript writing: All authors.
Final approval of manuscript: All authors.
Accountable for all aspects of the work: All authors.

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Abbreviations

[^{18}F]FDG-PET, 18F-fluorodeoxyglucose positron emission tomography; Asp, aspartic acid; Cys, cyst(e)ine; DIVER, Deep Imaging Via Emission Recovery; ER+, Estrogen Receptor Positive; FLIM, Fluorescent Lifetime Imaging; GEXPLORER, Gene Expression eXPLORER; Gln, glutamine; Glu, glutamate; GLUD1/2 or GDH1/2, Glutamate Dehydrogenase 1/2; Gly, glycine;

GPX4, Glutathione Peroxidase 4; GSH, glutathione; GSSG, glutathione disulfide; HTN,
hypertension; IDC/NST, Invasive Ductal, or Breast Cancer of No Special Type; IHC,
Immunohistochemistry; ILC, Invasive Lobular Breast Cancer; LCCTam, Tamoxifen-resistant
Derivative of SUM44PE; Leu, leucine; LTED, Long-term Estrogen-deprived Derivative of MDA-
MB-134VI; mDLC, Mixed Ductal-lobular Breast Cancer; mGluRs, Metabotropic Glutamate
Receptors; OXPHOS, Oxidative Phosphorylation; panCK, Pan-cytokeratin; Pro, proline; R162,
GLUD1 inhibitor; RIME, Rapid Immunoprecipitation Mass Spectrometry of Endogenous Proteins;
ROS, Reactive Oxygen Species; SCAN-B, Sweden Cancerome Analysis Network – Breast;
SEER, Surveillance, Epidemiology, and End Results; SLC, solute carrier; TCA, tricarboxylic acid;
TCGA, The Cancer Genome Atlas; TMA, Tissue Microarray.

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Table 1. Clinical and pathological characteristics of TMA cohorts					
	ILC		**IDC/NST**		**P value**
	Count	Percent	Count	Percent	
Cohort period	2003-2014	-	2004-2011	-	
Number of patients	72	-	50	-	
Number of surgical events	78	-	50	-	
Sex					
Female	72	100%	50	100%	
Age					
Under 40	0	0%	3	6%	
40-55	37	51%	17	34%	
Over 55	35	49%	30	60%	
Age, initial diagnosis (years)	40.23 - 88.05	-	30.7-90.14	-	
Median age, initial diagnosis (years)	54.05	-	59	-	**p = 0.003
Race					
Black	16	22%	15	30%	p = 0.436
Other+Unknown	6	8%	4	8%	
White	50	70%	31	62%	
Lymph Node Status					
Positive	23	32%	18	36%	p = 0.655
Negative	49	68%	32	64%	
Vital Status (as of 7/2021)					
Alive	61	85%	32	64%	***p = 0.001
Deceased	11	15%	18	36%	
Duration of Follow-up (as of 7/2021)					
Range (years)	1.49 - 17.23	-	1.06-17.24	-	
Median (years)	8.42	-	10.36	-	p = 0.070
Treatment Received					
Chemotherapy	33	46%	29	59%	p = 0.202
Radiotherapy	34	47%	21	42%	p = 0.585
Endocrine Therapy	51	71%	35	71%	p > 0.999
Immunotherapy	1	1%	2	4%	p = 0.567

Figure Legends

Figure 1. Expression of glutamate-handling proteins in ILC and IDC/NST. A, Schematic showing functional relationships between the four glutamate-handling proteins included in our multiplex IHC panel. Gln, glutamine; Glu, glutamate; Cys, cyst(e)ine; Gly, glycine; Pro, proline; Asp, aspartic acid; Leu, leucine; GSH, glutathione; R162, GLUD1 pharmacological inhibitor. **B**, Schematic showing cohort assembly and study workflow. A representative TMA core stained for the four target proteins and nuclear staining (DNA) is shown, along with its corresponding images processed for tissue segmentation, indicating areas of panCK+ (yellow, epithelial) and panCK- (aqua, stromal) cells, followed by phenotype mapping for (as an example) GLUD1/2+ (magenta) and GLUD1/2- (blue) cells. **C**, Comparison of protein expression for GPX4, GLUD1/2, SLC3A2 and SLC7A11 in the multiplex IHC panel between ILC and IDC/NST in panCK+ tumor cells (top row) and panCK- stromal cells (bottom row). Data are presented as scatter plots, with each dot representing the mean of percent positive cells for each patient tumor and the solid line indicating the median. Data are analyzed by Mann-Whitney U test. Panels A and B were created in BioRender (Riggins, R. (2026)

https://app.biorender.com/illustrations/690fae51cfb5598c629cb341?slideId=c66ed237-665a-43d1-8a8d-0e63a54e1307).

Figure 2. Glutamate-handling proteins are associated with increased tumor size in ILC. A, Graphs illustrating the relationship between tumor size (cm) and percent positive cells for tumor expression of GPX4, GLUD1/2, SLC3A2 and SLC7A11 in the ILC cohort. **B**, Spearman correlation coefficient (ρ) and p value for each protein's relationship to tumor size in panCK+ tumor cells and panCK- stromal cells for the ILC and IDC/NST cohorts. **C**, Kaplan-Meier survival analysis, log-rank p value, hazard ratio (HR), and 95% confidence interval (CI) for high (above median) versus low (below median) tumor expression (percent positive cells) of the combined four proteins in the ILC and IDC/NST cohorts.

Figure 3. Increased co-expression of glutamate transport proteins and metabolic enzymes in ILC. Heatmaps showing Spearman correlation coefficients (ρ, ranging from +1.0 to -1.0) for each protein's relationship to one another, and hormone receptor expression (percent positive cells), in the ILC (**A**) and IDC/NST (**B**) cohorts. Asterisks denote statistical significance as defined in the Methods section.

Figure 4. High GLUD1/2 expression is associated with increased tumor size and stage in Black women with ILC. A, Spearman correlation coefficient (ρ) and p value for each protein's relationship to tumor size in panCK+ tumor cells and panCK- stromal cells for Black women in both the ILC and IDC/NST cohorts. **B**, Kaplan-Meier survival analysis, log-rank p value, hazard ratio (HR), and 95% confidence interval (CI) for high (above median) versus low (below median) tumor expression (percent positive cells) of the combined four proteins in Black women with ILC and IDC/NST cohorts. **C**, *GLUD1* mRNA expression (RSEM, RNA-Seq by Expectation-Maximization) by tumor stage from TCGA for Black women with ILC (left) and Luminal A and Luminal B IDC/NST (right) breast cancer. Data are presented as a scatter plot, with the solid line indicating the median. *GLUD1* expression in Stage I vs. Stages II and III by unpaired t-test (*, p<0.05).

Figure 5. Enrichment of GLUD1/2 and SLC3A2 in tumors from women with hypertensive comorbidities in ILC. A, Fisher's Exact Test of the presence or absence of any comorbidity, or

specific comorbidities, in ILC compared to IDC/NST. **B-C**, Fisher's Exact Test of the presence or
absence of any comorbidity (B) or hypertension (HTN; C) in Black women compared to white
women across the ILC and IDC/NST cohorts. **D-E**, Fisher's Exact Test of the proportion of high
(above median) and low (below median) GLUD1/2 and SLC3A2 expression in women with versus
without HTN in the entire ILC cohort (left), and in hypertensive Black (HTN Blk) vs White (HTN
Wht) ILC cohorts (right).

Figure 6. Pharmacological GLUD1 inhibition reduces ER protein levels, cell viability and
oxidative phosphorylation in ILC cell lines. A, Left, immunoblot analysis of the expression of
proteins in ILC breast cancer cell lines. SUM44 and LCCTam, pair of tamoxifen-sensitive and -
resistant (respectively) cell lines. MM134 and LTED-A through E, series of parental (MM134) and
long-term estrogen-deprived (LTED, mimicking aromatase inhibitor resistance) cell lines. Actin
serves as a loading control. Right, densitometry plot showing fold change in expression
(target:actin ratio) for each resistant cell line normalized to its parental control. **B**, Heatmap of
GLUD1 peptide count from ER RIME assays published by Sottnik et al[32]. Data are presented as
the mean peptide count from two biological replicates of the SUM44, MM134, and BCK4 ILC cell
lines. **C**, Immunoblot analysis of the expression of ER in ILC breast cancer cell lines cultured in
the absence (-) and presence (+) of treatment with 10 μM GLUD1 inhibitor R162 for 18-24 hours.
Actin serves as a loading control. ImageJ was used to quantify relative expression in R162- vs.
DMSO-treated cells. **D**. Cell viability after treatment of ILC cells with increasing doses of R162 for
3 days. Data were analyzed from three independent experiments, each with 4-8 technical
replicates. **E**, Fluorescence lifetime imaging microscopy (FLIM) of NADH in MM134 parental and
MM134 LTED-D cells under vehicle and R162 treatment conditions. The top row displays
autofluorescence intensity images in the NADH channel. Row 2 shows phase lifetime–color-
coded images, where shorter lifetimes (blue–cyan) are predominantly nuclear and longer lifetimes
(yellow–red) are cytoplasmic. Row 3 presents histogram-corrected images highlighting the
distribution of lifetimes in only the cytoplasm, with a notable shift toward shorter lifetimes (blue–
green) in R162-treated parental cells. Row 4 shows color-coded maps of bound/free NADH
fractions, indicating increased free NADH (yellow) and decreased bound NADH (purple) upon
R162 treatment, especially in parental cells. Scale bar = 10 μm. **F**, Quantification of bound/total
NADH fractional intensity ratios derived from FLIM images in A. Data are presented as a box and
scatter plot, with each point representing a single image (16-18 per treatment group) and the solid
line indicating the median. Data are analyzed by unpaired t-test. **G**, Non-mitochondrial oxygen
consumption in ILC cells treated with vehicle or R162. Data were analyzed from two independent
passages for parental cells and four independent passages for LTED-D cells, each with 6-8
technical replicates. **H**, Extracellular lactate production in culture media from ILC cells treated with
vehicle or R162. Data were analyzed from two independent experiments, each with three
technical repeats, and a representative graph from one experiment is shown. Data in D, G and H
are presented as mean ± SEM and analyzed by one-way ANOVA with Dunnett's multiple
comparison test. Asterisks denote statistical significance as defined in the Methods section.

Figure 1

Figure 2

A.

ILC Tumor GPX4

Spearman ρ = 0.3153
**P = 0.0099

ILC Tumor GLUD1/2

Spearman ρ = 0.2933
*P = 0.0168

ILC Tumor SLC3A2

Spearman ρ = 0.2951
*P = 0.0162

ILC Tumor SLC7A11

Spearman ρ = 0.2719
*P = 0.0272

B.

Expression vs Tumor Size				
Tumor				
Protein	ILC Spearman ρ	ILC P Value	IDC/NST Spearman ρ	IDC/NST P Value
GPX4	0.3153	0.0099**	0.02268	0.8784
GLUD1/2	0.2933	0.0168*	0.007924	0.9574
SLC3A2	0.2951	0.0162*	-0.03899	0.7925
SLC7A11	0.2719	0.0272*	-0.02526	0.8647
Stroma				
Protein	ILC Spearman ρ	ILC P Value	IDC/NST Spearman ρ	IDC/NST P Value
GPX4	0.2457	0.0467*	0.09263	0.5312
GLUD1/2	0.2484	0.0443*	-0.04844	0.7437
SLC3A2	0.03266	0.7946	-0.2197	0.1334
SLC7A11	0.2152	0.0827	-0.1506	0.3069

C.

ILC Tumor

Low
High

logrank P = 0.595
HR = 0.7
95% CI = 0.188 to 2.60

ns

IDC/NST Tumor

Low
High

logrank P = 0.651
HR = 0.804
95% CI = 0.311 to 2.08

ns

Figure 3

A. ILC

B. IDC/NST

Figure 4

A.

Expression vs Tumor Size, Black women				
Tumor				
Protein	ILC Spearman ρ	ILC P Value	IDC/NST Spearman ρ	IDC/NST P Value
GPX4	0.5873	0.0185*	-0.2953	0.2819
GLUD1/2	0.6097	0.0139*	-0.1650	0.5538
SLC3A2	0.5803	0.0204*	-0.2280	0.4102
SLC7A11	0.3729	0.1543	-0.3564	0.1906
Stroma				
Protein	ILC Spearman ρ	ILC P Value	IDC/NST Spearman ρ	IDC/NST P Value
GPX4	0.5729	0.0223*	-0.4272	0.1127
GLUD1/2	0.5940	0.0170*	-0.2083	0.4528
SLC3A2	0.5954	0.0167*	-0.5220	0.0481*
SLC7A11	0.3314	0.2087	-0.4789	0.0727

B.

C.

Figure 5

A.

Cohort Comorbidities					
	ILC		**IDC/NST**		**P value**
Comorbidity	Count	Percent	Count	Percent	
Any	19	26%	17	34%	0.4216
Obesity	0	0%	7	14%	0.0015**
Hypertension	19	26%	12	24%	0.8345

B.

C.

D. ILC, All women

E. ILC, Black and White women

Figure 6

Supplemental Figure Legends

Figure S1. **Expression of glutamate-handling proteins in publicly available breast cancer**
datasets. Violin plots of mRNA expression data from invasive lobular breast cancer (ILC), mixed
ductal-lobular breast cancer (mDLC), and breast cancer of no special type (NST) for each target
in the multiplex IHC panel from TCGA, METABRIC, and SCAN-B. Data were analyzed and
graphed using Gene Expression eXPLORER (GEXPLORER) at
https://leeoesterreich.org/resources.

Figure S2. Expression *GLUD1* mRNA in ILC cell lines and phasor-FLIM metabolic imaging
workflow. **A**, Waterfall plot of *GLUD1* mRNA expression data from all breast cancer cell lines in
DepMap release 25Q2 (left), and boxplot of *GLUD1* mRNA expression for ER+ ILC/ILC-like vs.
IDC/NST cell lines in this dataset (right, line at median, error bars minimum to maximum). Data
are analyzed by unpaired t-test. Cell lines are assigned as ILC/ILC-like according to Shah et al[1].
B, Workflow for analyzing autofluorescence FLIM images of NADH. The fluorescence lifetime of
each pixel in the intensity image (left) is transformed into a phasor plot (right). Using a phase
lifetime color map (rainbow scale above the phasor plot), phasor points are color-coded and
mapped back onto the original image to generate lifetime-coded autofluorescence images. These
images are then restricted to the cytoplasm using intensity histogram correction. Bound/free
NADH fractions are calculated using a multicomponent phasor approach, where purple indicates
fully bound NADH and yellow indicates fully free NADH (binary scale: 0–255). To enhance
visualization of sample differences, the color scale is adjusted to 140–255 for bound NADH
(purple) and 1–35 for free NADH (yellow). These scaled images are presented in Figure 6.

Supplement References

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Figure S1

Figure S2

A.

B.

Supplemental Table 1: Primary Antibody/OPAL Dye Pairings and Incubation Conditions

	Antibody 1	Antibody 2	Antibody 3	Antibody 4	Antibody 5
Antigen	CD98	GLUD1/2	GPX4	panCK	SLC7A11
Company	Santa Cruz	Cell Signaling	Abcam	Agilent	Abcam
Cat#, RRID	Cat# sc-376815, RRID:AB_2938854	Cat# 12793, RRID:AB_2750880	Cat# ab125066, RRID:AB_10973901	Cat# M3515, RRID:AB_2132885	Cat# ab175186, RRID:AB_2722749
Species	Mouse	Rabbit	Rabbit	Mouse	Rabbit
Dilution	1:200	1:200	1:300	1:300	1:800
Incubation Time	1 hr	1 hr	1 hr	1 hr	1 hr
Incubation Temp.	RT	RT	RT	RT	RT
Control Tissue	Breast Cancer	Breast Cancer	Liver Cancer	Breast Cancer	Glioma
OPAL Fluor.	520	650	570	690	620
OPAL Conc.	1:150	1:150	1:400	1:30	1:300
Antigen Retrieval	AR6	AR6	AR6	AR6	AR6

Supplemental Table 2: Literature estimates of ILC prevalence in Black women			
Author, Year (number of Black women)	Black and Lobular (%)	Significant vs NHW?	Reference
^Baquet, 2008 (nr)	nr (5.5%)	yes, ↓	55
Elledge, 1994 (1016)	nr	no	56
Gallagher, 2022 (295)	25 (8.5%)	yes, ↓	57
^Han, 2021 (7292)	541 (7.4%)	na	58
Oesterreich, 2022 (3022)	276 (7.6%)	yes, ↓	59
Williams, 2018 (1393)	126 (16%)	yes, ↓	60

n, number
NHW, non-Hispanic white
^Surveillance, Epidemiology, and End
Results (SEER) data

nr, not reported
na, not applicable (Black women only)

References numbered according to
main manuscript text

Supplemental Table 3: Comorbidity Analysis
Two-sided Fisher's exact test was used for statistical analysis.

A) Overall comorbidity analysis for ILC and IDC/NST

	ILC	IDC	P-value
Comorbidity	19 (26.389%)	17 (34%)	0.4216 (NS)

- frequency and corresponding % represent the presence or absence of any comorbidity
- ILC: 9 out of 72 had multiple comorbidities
- IDC: 6 out of 50 had had multiple comorbidities

B) Categorical comorbidity analysis for ILC and IDC/NST

	ILC	IDC	P-value
Diabetes mellitus	5 (6.944%)	4 (8%)	>0.9999 (NS)
Hypercholesterolemia and/or hyperlipidemia	8 (11.111%)	1 (2%)	0.0804 (NS)
Obesity	0 (0%)	7 (14%)	0.0015 (**p<0.01)
Hypertension	19 (26.389%)	12 (24%)	0.8345 (NS)
Chronic kidney disease	0 (0%)	1 (2%)	0.4050 (NS)