Brief Communication High EIF4EBP1 expression reflects mTOR pathway activity and cancer cell proliferation and is a biomarker for poor breast cancer prognosis

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Abstract: Eukaryotic translation initiation factor 4E binding protein 1 (EIF4EBP1) is regulated by the mTOR (mammalian target of rapamycin) signaling pathway. Phosphorylated EIF4EBP1 protein leads to pathway activation and correlates with aggressive breast cancer features. However, the clinical relevance of EIF4EBP1 gene expression as a prognostic biomarker in bulk breast tumors is not understood. In this study, EIF4EBP1 expression was analyzed in over 5000 breast cancers from three large independent cohorts, TCGA, METABRIC, and SCAN-B (GSE96058), and expression was dichotomized into low and high groups by the median. We also performed gene set enrichment analysis (GSEA) and cell cybersorting via the xCell algorithm to investigate EIF4EBP1 biology and expression patterns within the tumor microenvironment (TME). We additionally confirmed EIF4EBP1 expression location in the TME via single cell RNA sequencing. EIF4EBP1 expression was highest in both triple negative and high-grade tumors (both P<0.001), and tumor mutational burden scores were highest in the high EIF4EBP1-expression groups (all P<0.001). High EIF4EBP1 expression significantly correlated to worse overall survival in all three cohorts (hazard ratios (HR) 1.4-1.9), and worse distant relapse-free survival in patients treated with neoadjuvant taxane-anthracycline chemotherapy (HR 2.4). GSEA demonstrated enriched mTOR and cell proliferation-related gene sets, including, MYC, G2M checkpoint, and E2F targets across all three bulk tumor and single cell RNA sequencing cohorts. Phenotypically, these pathways were reflected by increased Ki67 gene expression and signaling via pharmacologically-activated mTOR gene sets in EIF4EBP1 high-expressing tumors (all P<0.001). EIF4EBP1 expression was increased in whole breast tumors compared to normal breast tissue (P<0.001), and was expressed predominantly in cancer epithelial cells, particularly in basal epithelial cell subclasses. EIF4EBP1 expression did not correlate to a consistent immune system phenotype across all three cohorts. Overall, these findings support that high EIF4EBP1 gene expression in bulk breast tumors could represent a poor prognostic marker via mTOR signaling pathways activation and upregulation of cell cycling, ultimately leading to increased tumorigenesis.

Keywords: Bioinformatics, cap-dependent translation, mutational burden, prognostic biomarker, tumor progression, tumor microenvironment

Introduction

With a 1 in 8 lifetime risk, breast cancer is the most common cancer in women, and though five-year survival approaches 99% when managed as a localized condition, 43,000 women still die annually in the United States from this disease [1-3]. The majority of these deaths are

attributable to the evolution of treatment resistance in disease that relapses [4]. Decoding these mechanisms of resistance is a cornerstone of current breast cancer research [5].

One of these pathways implicated in treatment resistance is the phosphoinositide 3-kinase (PI3K)/AKT/mammalian target of rapamycin

(mTOR) pathway, which is a major intracellular system that activates multiple cellular proliferation and metabolism networks in cancer cell signaling [6, 7]. In luminal breast cancers, the PI3K pathway is one of the most altered signaling cascades [8], and is upregulated in over 70% of all breast cancers [9]. At any step within this pathway, either overexpression or mutations result in pathway hyperactivation, which funnels into mTOR complex-1 (MTORC1)-mediated phosphorylation and initiation of multiple downstream transcriptional activators [8]. Specific to breast cancer treatment resistance, PI3K pathway hyperactivation, typically via activating mutations in PI3KCA, the gene encoding the p1110 alpha subunit, will promote escape from hormone dependence in estrogen receptor (ER) positive breast cancer [10] and trastuzumab resistance in human epidermal growth factor (HER2) positive breast cancer [11]. regardless of tumor stage [12, 13]. As such, the mTOR inhibitor everolimus was the first oral targeted therapy widely used in advanced hormone positive breast cancer [14, 15], and is additionally known to extend progression-free survival in patients with trastuzumab-resistant HER2-postive advanced breast cancer [16].

One of the many downstream targets of the PI3K/ATK/mTOR pathway is eukaryotic translation initiation factor (EIF) 4E binding protein 1 (4E-BP1), encoded by the gene EIF4EBP1 (Figure 1) [17]. 4E-BP1, in its unphosphorylated state, functions as a translation suppressor by preventing the eukaryotic translation factor 4E from complexing with a multisubunit scaffold that recruits the 40S ribosomal subunit to the 5' end of mRNA to initiate cap-dependent translation (Figure 1) [18]. The mTORC1 complex, in response to PI3K pathway activation, is the major kinase responsible for phosphorylating 4E-BP1, resulting in release of 4E and subsequent activation of cap-dependent translation [19]. Therefore, unphosphorylated 4E-BP1 can, in principle, act as a tumor suppressor [20]. Clinically however, 4E-BP1 is overexpressed in many tumor types including breast cancer relative to normal and benign tissue, and phosphorylated 4E-BP1 levels positively correlate with increased tumor size, lymph node metastasis, and locoregional breast cancer recurrence [21]. However, as a prognostic biomarker, the clinical relevance of EIF4EBP1 gene expression is poorly understood.

In this study, we investigate the role of EIF4-EBP1 gene expression within the human breast cancer tumor microenvironment using bioinformatical analyses of transcriptomic profiles, as previously published by our group [22-27]. We hypothesize that EIF4EBP1 gene expression in breast cancer is correlated to increased mTORmediated signaling and poor prognosis secondary to increased tumor cell proliferation. By exploring bulk tumor transcriptomics with the aid of gene set enrichment analysis (GSEA) [28] and cell cybersorting with the xCell algorithm [29], we can perform concurrent examinations of multiple cellular markers, functions, and tumor microenvironment (TME) interactions. Combining this data with associated patient clinical outcomes over multiple independent databases validates our conclusions on the clinical significance of EIF4EBP1 gene expression in breast cancer.

Methods

Data acquisition

The primary data in this study was acquired from three main databases: the Cancer Genome Atlas Program (TCGA, whole database n=1090, estrogen-receptor positive and human epidermal growth factor negative tumors (ER+HER2-) n=593, HER2+ n=184, and triple negative breast cancer (TNBC) n=160), the Molecular Taxonomy of Breast Cancer International Consortium (METABRIC) (whole database n=1904, ER+HER2- n=1355, HER2+ n=236, TNBC n=313), and Sweden Cancerome Analysis Network-Breast (SCAN-B, GSE96058) (whole database n=3069, ER+HER2- n=2277, HER2+ n=392, TNBC n=155). TCGA and ME-TABRIC results were obtained via the cBio-Portal (https://www.cbioportal.org), and the expression data for TCGA was log-transformed using "data_mrna_seq_v2_rsem", while the METABRIC data was used as is with the "data expression median". SCAN-B results were downloaded from the Gene Expression Omnibus (GEO) repository of the United States National Institutes of Health (https://www.ncbi. nlm.nih.gov/geo), as described, and the provided normalized expression data from the database was used without any further processing [23, 30]. Data was also obtained from GSE25066 via GEO, a cohort comprised of 508 patients treated with neoadjuvant taxane-



Figure 1. Overview of cellular signaling mediated by 4E-BP1 (4E binding protein 1). The 4E-BP1 protein is encoded by the gene *EIF4EBP1* (eukaryotic translation initiation factor 4EBP1). 4E-BP1 directly interacts with the eukaryotic translation factor 4E (4E), which prevents completion of a multisubunit scaffold that aids in the recruitment of 40S ribosomal subunits to the 5' end of mRNA. In response to mTOR (mammalian target of rapamycin) signaling following activation of the PI3K pathway (phosphoinositide 3-kinase)/AKT pathway, the mTORC1 complex will phosphorylate 4E-BP1. The enables the release of 4E, which is now free to complete the multisubunit scaffold necessary to activate cap-dependent translation of mRNA. Ultimately, this facilitates a broad range of downstream effects resulting in cell cycling and tumorigenesis.

anthracycline chemotherapy [31]. Gene expression data from 114 samples of normal breast tissue from female patients was obtained from the Genotype-Tissue Expression (GTex) Portal (https://gtexportal.org) [32]. Single cell RNA sequencing breast cancer atlas data from female patients was obtained from two large cohorts [33, 34] via the Broad Institute Single Cell Portal (https://singlecell. broadinstitute.org/single_cell), using the accession numbers SCP1039 [33] and SCP1106 [34]. The downloaded features, matrix, and barcode data were integrated and processed using the ReadMtx function in the Seurat package of R-4.2.1 (https://www.R-project.org), as previously described [25]. All R packages used in this project are previously described [25].

As 99.4 percent of all breast cancer cases occur in females [35], and all databases used in the study contain only female patients, male breast cancer was not examined in this study. All data was downloaded in July 2022. Because all data was obtained from deidentified public databases or cohorts, ethics approval requirements were waived by the Roswell Park Institutional Review Board.

Gene set enrichment analysis

Functional enrichment analysis of *EIF4EBP1* was performed by gene set enrichment analysis (GSEA) [28] on the Molecular Signatures Database Hallmark collection (http://www.gsea-msigdb.org) [36]. Gene sets with a false discovery rate (FDR) <0.25 specified enriched signaling [28]. High and low *EIF4EBP1* expression groups were dichotomized by median gene expression. Positive normalized enriched scores (NES) indicate enriched signaling in the *EIF4EBP1*-high expression group.

Other scores

The xCell algorithm (https://xcell.ucsf.edu) [29] was used to correlate EIF4EBP1 expression to the infiltrating fraction of tumor and stromal cells (epithelial cells, fibroblasts, adipocytes, and endothelial cells), and immune cells (CD8+, T helper cell (Th)1 and Th2 cells, T-regulator cells, M1 and M2 macrophages, and dendritic cells) as described [22, 37-39]. The breast cancer mutational landscape (intratumor heterogeneity, homologous recombination defects, fraction genome altered, silent mutation rate, non-silent mutation rate, single-nucleotide neoantigens, and indel mutations) was examined from data derived by Thorsson et al. [40]. Immune cytolytic activity (CYT) in the tumor microenvironment was calculated as the geometric mean of the expression of perforin (PRF1) and granzyme A (GZMA) mRNA expression, which measures the anti-cancer ability of cytotoxic T cells [41].

Statistical analyses

Statistical analyses and figure production were performed with R-4.2.1 and BioRender (https://www.biorender.com). mRNA levels for *EIF4EBP1* were dichotomized into low and high groups based on the median expression level. All results are plotted as box plots, with the lower and upper bounds representing the maximum and minimum values, the upper and lower ends of box representing the 25th and 75th percentile values and the bolded bar within the box representing the median value. For TCGA results (RNA sequencing data), units of expression are log2 transformed RSEM, METABRIC (microarray data), units of expression are log intensity levels, and all GSE results (RNA sequencing data) are log2 transformed CPM. Two group comparisons were performed using the Mann-Whitney U test and multiple group comparisons by the Kruskal-Wallis test. The R survival software package was used to analyze survival based on high or low EIF4EBP1 expression via Cox-proportional hazards regression, and Kaplan-Meier curves were compared by the log rank test. P<0.05 was set for statistical significance.

Results

Demographic data for patients in the three cohorts are presented in **Table 1**. Histograms of EIF4EBP1 gene expression in each of TCGA. METABRIC, and SCAN-B show a relatively bellshaped distribution with a slightly left-shifted curve (Figure 2A). In all three cohorts, EIF4EBP1 expression was significantly highest in TNBCs and lowest in ER+HER2- tumors (all P<0.001, Figure 2B). There was a slight tendency for increased expression from stage I to stage III disease in the TCGA and METABRIC cohorts (staging data not available for SCAN-B) (Figure 2C). Tumors with positive lymph nodes had the same expression levels as node negative tumors (not shown). There was no difference between metastatic and non-metastatic tumors, however, there were only 29 metastatic tumors (stage IV) in TCGA and METABRIC combined (not shown). EIF4EBP1 expression positively correlated with increasing tumor grade in all three cohorts (all P<0.001, Figure 2D).

Because tumor mutational burden is typically a molecular surrogate biomarker of disease aggressiveness [42], we examined a panel of

Cohort	TCGA			METABRIC			SCAN-B		
Group	High	Low	p-value	High	Low	p-value	High	Low	p-value
Median Age (max, min)	58 (49, 69)	58 (48, 66)	0.3	61 (51, 70)	62 (52, 71)	0.14	65 (52, 73)	64 (53, 70)	0.08
Stage, N (%)			0.3			<0.001			N/A
I	79 (15%)	99 (18%)		193 (20%)	282 (30%)				
II	312 (58%)	298 (55%)		427 (45%)	373 (39%)				
III	122 (22%)	123 (23%)		75 (8%)	40 (4%)				
Unknown	26 (5%)	18 (3%)		257 (27%)	257 (27%)		1,535 (100%)	1,534 (100%)	
Lymph Node Involvement, N (%)			0.7			0.049			0.023
Negative	255 (47%)	253 (47%)		475 (50%)	518 (54%)		875 (57%)	936 (61%)	
Positive	269 (50%)	279 (52%)		477 (50%)	434 (46%)		611 (40%)	551 (36%)	
Unknown	15 (3%)	6 (1%)		0 (0%)	0 (0%)		49 (3%)	47 (3%)	
Subtype, N (%)			<0.001			<0.001			<0.001
ER+HER2-	231 (43%)	353 (66%)		555 (58%)	800 (84%)		995 (65%)	1,282 (84%)	
HER2+	92 (17%)	89 (17%)		148 (16%)	88 (9%)		250 (16%)	142 (9%)	
TNBC	130 (24%)	29 (5%)		249 (26%)	64 (7%)		134 (9%)	21 (1%)	
Unknown	86 (16%)	67 (12%)		0 (0%)	0 (0%)		156 (10%)	89 (6%)	
Grade, N (%)			<0.001			<0.001			<0.001
1	23 (4%)	53 (10%)		44 (5%)	121 (13%)		124 (8%)	330 (22%)	
2	96 (18%)	170 (32%)		288 (30%)	452 (47%)		571 (37%)	868 (57%)	
3	161 (30%)	73 (14%)		588 (62%)	339 (36%)		803 (52%)	312 (20%)	
Unknown	259 (40%)	242 (45%)		32 (3%)	40 (4%)		37 (2%)	24 (2%)	

Table 1. Demographic data of patients in the three cohorts

High and low groupings are divided by median EIF4EBP1 gene expression.



Figure 2. *EIF4EBP1* gene expression by breast cancer characteristics. A. Histogram distribution of *EIF4EBP1* expression in breast tumors in TCGA (1090 specimens), METABRIC (1094 specimens), and SCAN-B (3069 specimens). B. Breast cancer subtype. ER+HER2- (estrogen receptor positive, human epidermal growth factor receptor negative

tumors), HER2+, TNBC (triple negative breast cancer). C. Staging according to the American Joint Committee on Cancer (AJCC). Stage is not available for the SCAN-B cohort. D. Grading indicated according to AJCC. Counts for boxplots are indicated in the x-axis. Units of *EIF4EBP1* expression: TCGA - log2 transformed RSEM, METABRIC - log intensity levels, and SCAN-B - log2 transformed CPM. The bolded center bar within the box plots represents the median, the lower and upper box bounds represent the 25th and 75th percentiles, respectively, and the lower and upper tails represent the minimum and maximum values, respectively.

common mutational burden scores dichotomized by median *EIF4EBP1* expression. All scores (intratumor heterogeneity, homologous recombination defects (HRDs), fraction-genome-altered (FGR), silent mutation rate (SMR), non-silent mutation rate (NSMR), single-nucleotide variant (SNV) neoantigens, and indel mutations) were significantly elevated in the *EIF-4EBP1*-high expressing group (all *P*<0.001, **Figure 3**).

We next examined survival trends dichotomized by the median into low and high EIF4EBP1 expression. When comparing disease-free survival (DFS), the hazard ratio was about 1.3-1.4 for the high EIF4EBP1 group compared to the low group in both TCGA and METABRIC. However, this result did not reach statistical significance in the TCGA group (HR 1.39, 95% confidence interval (CI) 0.96-2.04, P=0.08), but with nearly double the number of patients in METABRIC, the HR was 1.30 (1.12-1.52) (P=0.0007, Figure 4A). Disease-specific survival (DSS) was significant in both cohorts, with HR 1.4-1.8 (Figure 4A). Overall survival (OS) was a reported metric in all three large cohorts, and the HR was significant in all cohorts, with a HR of mortality ranging from 1.2-1.9 when comparing the high EIF4EBP1 group to the low group (all P≤0.04, Figure 4A, 4B). We additionally examined distant relapsefree survival (DRFS) for the GSE25066 cohort, which is comprised of 508 patients treated with neoadjuvant taxane-anthracycline chemotherapy [31]. The HR for DRFS was 2.38 (1.61-3.57) when comparing the high EIF4EBP1 group to the low group (Figure 4C).

We used GSEA on the Hallmark pathways to correlate enriched gene signaling to *EIF4EBP1* expression [36]. Gene sets were selected if they were significantly enriched in all three cohorts. Of the 50 gene sets, 8 were enriched, and all of them were enriched in the high *EIF4EBP1* expression group. These included MTORC1 signaling, four cell-cycle related path-

ways (MYC Targets V1 and V2, G2M checkpoint, and E2F targets), unfolded protein response, glycolysis, and DNA repair, with all normalized enrichment scores (NES) in the 1.4-2.2 range (Figure 5A). We then repeated the GSEA analysis on two independent cohorts of single cell RNA sequenced tumors [33, 34]. In both cohorts, gene pathways for MTORC1 signaling, MYC targets V1, E2F targets, G2M checkpoint, and mitotic spindle occupied five of the top six highest NES gene sets (Figure 5B). We then examined whether pathway enrichment in cell cycling and MTORC1 signaling networks correlated to functional tumor biology. In all three cohorts, gene expression of Ki67, a marker of cell proliferation, was significantly increased in the high EIF4EBP1 expression group compared to the lower expression group, consistent with cell cycling pathway gene enrichment (all P<0.001, Figure 5C). Similarly, we examined two gene signatures defined pharmacologically by gene upregulation following mTOR inhibition [43, 44]. These gene signatures were also significantly upregulated in the high EIF4EBP1 expression group in all three cohorts (all P< 0.001, Figure 5D). We also examined the converse, by correlating a gene signature defined pharmacologically by gene downregulation following mTOR inhibition [43]. As predicted, this gene signature was significantly downregulated in the high EIF4EBP1 expression group (all P<0.001, Figure 5D).

EIF4EBP1 expression levels were compared between normal breast tissues and whole breast tumors, and was significantly increased in the breast tumor group in TCGA (P<0.001, **Figure 6A**). We then used the xCell algorithm to examine tumor cell population estimates based on dichotomized *EIF4EBP1* expression. Epithelial cells were significantly increased in the high *EIF4EBP1* group in both TCGA and METABRIC cohorts (P<0.001), but was unchanged in the SCAN-B cohort (P=0.08, **Figure 6B**). Fibroblasts, adipocytes, and endothelial cell populations were decreased in the high







Figure 4. Survival plots for low and high *EIF4EBP1* gene expression in breast tumors. A. Disease-free survival (DFS), disease-specific survival (DSS), and overall survival (OS) for the TCGA and METABRIC cohorts. B. OS for the SCAN-B cohort. C. Distant relapse-free survival (DRFS) for the GSE25066 cohort. This cohort is comprised of 508 patients treated with neoadjuvant taxane-anthracycline chemotherapy [31]. Patients at risk for each time point are listed along the x-axis. *EIF4EBP1* expression is dichotomized into low and high groups by the median. The hazard ratio (HR) compares the high group against the low group. *P* values by log rank test.

EIF4EBP1 groups (**Figure 6B**). On analysis of two independent cohorts of single cell RNA sequencing results [33, 34], the bulk of *EIF-4EBP1* expression was attributable to epithelial/cancer epithelial cell populations (**Figure 6C**). When examined by epithelial cell subclass, the highest *EIF4EBP1* expression was found in basal epithelial cells rather than luminal (mature) epithelial cells (**Figure 6D**). We then finally examined the correlation between immune cell populations and *EIF4EBP1* expression levels. Among anti-cancer immune cells, Th1 cells and M1 macrophages were significantly increased in the high *EIF4EBP1* expression group in all three cohorts (all P<0.001), but there were no consistent trends for CD8+ T-cells or dendritic cells (**Figure 7**). Among procancer immune cells, Th2 were significantly increased in the high *EIF4EBP1* expression group in all three cohorts (all P<0.001), but there were no consistent trends in regulatory T cells or M2 macrophages (**Figure 7**). As a surrogate measure of overall tumor immune response in the TME, immune cytolytic activity

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Figure 5. Gene set enrichment analysis (GSEA) for *EIF4EBP1* in breast cancer, and demonstration of increased Ki67 and mTOR signaling in high expressing *EIF4EBP1*-tumors. (A) GSEA results from the Hallmark gene sets significant in all three cohorts (TCGA, METABRIC, and SCAN-B). A false discovery rate (FDR) of less than 0.25 was considered statistically significant. Dot size represents the FDR value, and they are colored according to the normalized enrichment score (NES). MTORC1 signaling gene set labelled in blue, and cell cycling-related gene sets labelled in red. (B) Most enriched gene sets in single cell RNA sequencing cohorts from cohorts described in [33] (left), and in [34] (right). MTORC1 signaling gene set labelled in blue, and cell cycling-related gene sets labelled in red. (C) Ki67 box plots based on median *EIF4EBP1* expression. (D) MTOR gene set defined by genes upregulated following siroliumus treatment as described in [43] (Left). MTOR gene set defined by genes upregulated following siroliumus treatment as described in [43] (Left). MTOR gene set defined by genes downregulated following siroliumus treatment as described in [43] (Right). For results in (C) and (D), *EIF4EBP1* expression is dichotomized into low and high groups by the median. The bolded center bar within the boxplots represents the median, the lower and upper box bounds represent the 25th and 75th percentiles, respectively, and the lower and upper tails represent the minimum and maximum values, respectively.



Figure 6. *EIF4EBP1* gene expression is enriched in breast tumors compared to normal breast tissues and is expressed primarily in cancer epithelial cells. A. mRNA expression from 114 normal breast tissues is compared to 1090 breast cancer tumors from the TCGA database. B. Box plots of epithelial cells, fibroblasts, adipocytes, and endothelial cells based on the xCell algorithm for the TCGA, METABRIC, and SCAN-B cohorts. *EIF4EBP1* gene expression is dichotomized by the median. The bolded center bar within the boxplots represents the median, the lower and upper box bounds represent the 25th and 75th percentiles, respectively, and the lower and upper tails represent the minimum and maximum values, respectively. C. Single cell RNA sequencing results from cohort described in [33],

comprised of 26 tumors (11 ER+HER2-, 5 HER2+, and 10 TNBC), with a total of 130,246 single cells (Left). Results from cohort described in [34] comprised of 5 TNBC tumors, with a total of 24,271 single cells (Right). D. The dot plot shows the overall percentage of the total *EIF4EBP1* gene expression by cell type and the average expression within each cell type for each cohort. Histogram/violin plot of *EIF4EBP1* gene expression by epithelial cell subclass from cohort described in [33].

(CYT score) showed no consistent pattern across the three cohorts (**Figure 7**).

Discussion

In this investigation, high expression of EIF-4EBP1 was identified as a predictor for worsened breast cancer prognosis and validated in three independent cohorts with over 5000 patients. Breast cancers express increased EIF4EBP1 levels compared to normal tissue. EIF4EBP1 expression occurred primarily in cancer epithelial cells, and high EIF4EBP1 expression correlated with increased tumor grade, increased overall tumor mutational burden. and decreased patient survival metrics. In these high expressing tumors, mTOR signaling and subsequent cell-cycle pathways were the most upregulated gene signatures. These gene signature findings were further supported by increased expression of gene sets characterized by mTOR pharmacological inhibition, and increased Ki67 scores.

EIF4EBP1 amplification has been demonstrated to occur across all breast cancer subtypes in up to 14% of cases [45]. While this might partially account for increased levels in breast tumors over normal tissue, it does not directly explain why high *EIF4EBP1* level correlate with increased tumorigenesis. As previously described, as protein, 4E-BP1 exists essentially in two states: unphosphorylated and phosphorylated. In the unphosphorylated state, 4E-BP1 functions as a both a transcriptional regulator and potential tumor suppressor by blocking cap-dependent translation via sequestration of the 4E transcription factor [20]. Particularly in response to mTOR-mediated signaling, 4E-BP1 becomes phosphorylated, releasing 4E for subsequent function [21]. In multiple cancer sites, including adrenocortical carcinoma, bladder urothelial, breast, ovarian, endometrial, renal, lung, mesothelioma, hepatocellular and acute myeloid leukemia, phosphorylated 4E-BP1 correlates to poor patient outcomes [45-48]. However, notability for gastric adenocarcinomas, phosphorylated 4E-BP1 is known to be highly expressed in early rather than late stage cancers, and is correlated with both prolonged disease-free and overall survival [48, 49].

Though not understood, there are several possible explanations for these discrepant findings among tumor sites. First, 4E-BP1 has at least seven sites of phosphorylation, each of which is likely to have their own influences on protein function regulation [17]. Second, these sites can be targeted by other minor kinases besides mTOR, including GSK-3 beta, ERK, PIM2, ATM, CDK1, and LRKK2, providing an additional level of modulation [17]. Third, there is evidence of spatial regulation of phosphorylated 4E-BP1 function. For example, cytoplasmic phosphorylated 4E-BP1 was observed in invasive breast and ovarian carcinomas as opposed to nuclear localization in normal epithelium and stromal tissues of the same tissue types [21]. However, in other tissue types, particularly endometrial cancer, nuclear rather than cytoplasmic phosphorvlated 4E-BP1, correlated to both increased mTOR pathway activation and worse patient outcomes [50]. In summary, the proteomic, kinomic, and cellular localization aspects of 4E-BP1 regulation are extremely complex.

From this investigation, bulk tumor EIF4EBP1 expression could be a much more straightforward surrogate marker for studying 4E-BP1 tumor biology in future breast cancer studies. as the tumorigenic phenotype exhibited in patients with elevated gene expression levels correlated strongly to known phosphorylated 4E-BP1-mediated characteristics. Mechanistically, this could be explained by the high likelihood that most 4E-BP1 in breast tumors exists in a phosphorylated state due secondary to nearly constitutive PI3K/mTOR pathway activity secondary to activating mutations [51]. Immunohistochemical investigations have showed expression levels of phosphorylated 4E-BP1 in breast tumors to be as high as 87% [21], or as low as 30% [52]. Such variability in part depends on the qualitative nature and even interpersonal variability of immunohistochemistry scoring [53]. To the best of our knowledge,



Figure 7. Anti-cancerous and pro-cancerous immune cell correlation with *EIF4EBP1* expression in breast cancer tumors, and cytolytic (CYT) scores. Box plots are based on the xCell algorithm for the TCGA, METABRIC, and SCAN-B cohorts. *EIF4EBP1* expression is dichotomized into low and high groups by the median. The bolded center bar within the boxplots represents the median, the lower and upper box bounds represent the 25th and 75th percentiles, respectively, and the lower and upper tails represent the minimum and maximum values, respectively.

there are no studies that have conducted patient-matched correlations of 4E-BP1 mRNA and phosphorylated protein levels.

While this study supports the clinical relevance of EIF4EBP1 expression in breast cancer, it does have several limitations. Despite using three very large and widely used independent cohorts to validate our key findings, our study is retrospective in its design, and is therefore prone to selection bias. The absence of granular clinical information requires us to assume that all patients received standard of care treatments; these however are constantly in flux, and therefore patient populations, treatments, and outcomes are heterogeneous. Additionally, although we cannot necessarily deduce mechanism of action from bioinformatic data, these investigations offer powerful insights into the association between EIF4EBP1 expression and patient outcomes, and are conducive to our proteomic understandings of signaling mediated by the pathways this binding protein modulates. Further comparative studies to systematically delineate the mechanistic underpinnings of the genomic, epigenic, proteomic, and kinomic pathways of the PI3K/ mTOR/4E-BP1 network would provide invaluable information to tailor and further design clinical trials for novel adjuvant therapeutics to mitigate breast cancer therapy resistance.

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Disclosure of conflict of interest

None.

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