# Comparative Proteome Analysis of Breast Cancer Tissues Highlights the Importance of Glycerol-3phosphate Dehydrogenase 1 and Monoacylglycerol Lipase in Breast Cancer Metabolism

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Abstract. Background/Aim: Breast cancer (BC) incidence and mortality rates have been increasing due to the lack of appropriate diagnostic tools for early detection. Proteomicsbased studies may provide novel targets for early diagnosis and efficient treatment. The aim of this study was to investigate the global changes occurring in protein profiles in breast cancer tissues to discover potential diagnostic or prognostic biomarkers. Materials and Methods: BC tissues and their corresponding healthy counterparts were collected, subtyped, and subjected to comparative proteomics analyses using two-dimensional gel electrophoresis (2-DE) and twodimensional electrophoresis fluorescence difference gel (DIGE) coupled to matrix-assisted laser desorption/ ionisation-time of flight mass spectrometry (MALDI-TOF/TOF) to explore BC metabolism at the proteome level. Western blot analysis was used to verify changes occurring at the protein levels. Results: Bioinformatics analyses performed with differentially regulated proteins highlighted the changes occurring in triacylglyceride (TAG) metabolism, and directed our attention to TAG metabolism-associated proteins, namely glycerol-3-phosphate dehydrogenase 1 (GPD1) and monoacylglycerol lipase (MAGL). These proteins were downregulated in tumor groups in comparison to controls.

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*Key Words:* Breast cancer proteomics, monoacylglycerol lipase, glycerol-3-phosphate dehydrogenase.

Conclusion: GPD1 and MAGL might be promising tissuebased protein biomarkers with a predictive potential for BC.

Despite major advances in health sciences, breast cancer (BC) is still one of the most life-threatening diseases for women. Unfortunately, BC mortality rates have been drastically increasing around the world reaching up to 522,000 deaths in 2017 alone (1). Luckily, however, not all women who have BC expire. In fact, the overall survival rate is highly dependent on the stage of the cancer. For the early detected and localized tumors (stage I and II) the chance of survival can reach up to 80-90%, while for the late detected tumors (stage III and IV) the survival rate drops dramatically to as low as 24% (1). Similarly, when 5-year survival rates were analysed, 92% of the stage 0 patients are not affected by the disease (2). While this rate for stage II patients is 75%, it goes down to 13% in stage IV patients (2). Overall, these numbers are indicative of the significance of early BC diagnosis and getting appropriate cancer treatment to prevent deaths.

Early diagnosis of BC is depended on two different tests. One of the tests relies on physical examination and imaging of the breast tissue by a physician or the patient herself. If any suspicious lump is detected, an image of the breast *via* mammography may be examined to assess the presence of the tumor (3). The other test may be performed directly on the biopsy material taken from the breast tissue or on the serum sample of the suspected BC patient. For BC, the CA 15-3, CA 27-29, carcinoembryonic antigen (CAE), estrogen receptor (ER), progesterone receptor (PR), human epidermal growth factor 2 (HER2) are currently the most commonly used biomarkers in the clinic (4). Each biomarker is a reporter of a special status of breast tumor. For example, CA 15-3 and CA 27-29 are glycoproteins encoded by the genes

of the MUC family (5, 6); their levels are increased in BC patient serum, thus representing valuable markers for BC prognosis, therapy and monitoring. However, none of these biomarkers are useful for the early diagnosis (7). Recent efforts to find a biomarker or a group of biomarkers for the early BC detection were not successful, although some novel prognostic biomarkers were proposed (8).

To discover useful biomarkers that are BC-specific and allow early diagnosis, the molecular mechanisms causing the conversion of cells from normal to cancerous state have to be elucidated in detail. Migration properties and growth rates of the aggressive tumor types, as well as their genetic and biochemical changes should be determined and compared to their non-aggressive counterparts. The answers to the question of how a dysregulated protein expression in cancer cells contributes to ungoverned tumor growth should provide a list of candidate biomarkers that will have the power to specifically diagnose BC at an early stage.

In this study, we first investigated the global changes in protein profiles using protein extracts from human BC tissues and their healthy counterparts *via* two-dimensional gel electrophoresis (2-DE). Then fluorescence difference gel electrophoresis (DIGE) analysis was performed to verify 2-DE results and generate a list of differentially regulated proteins. The changes in the expression levels of two proteins namely MAGL and GPD1 were further analysed in the BC molecular subtypes and the healthy controls. A bioinformatics analysis was carried out to associate the changes in protein abundance with the relevant biological processes. The data generated in this study represented the first detailed report for a comprehensive proteomic study using fresh-frozen BC tissue samples classified into the BC subtypes.

#### Materials and Methods

Ethical approval. This study has been approved by the Ethics Committee of Kocaeli University (approval numbers: KOU KAEK 2015/274 and KU GOKAEK 2018/49). Informed consents, approved by the institutional ethics committee, were obtained from each patient.

Procurement and storage of the tissues. The BC and corresponding healthy tissues were collected from the patients who were diagnosed with BC at Kocaeli University Medical School General Surgery Department and underwent lumpectomy or mastectomy between the years of 2015 and 2017. Immediately after the removal, the breast tissues were washed with ice-cold 10 mM Tris buffer containing 250 mM sucrose pH 7.4, snap frozen in liquid nitrogen and stored at -80°C.

Patient characteristics. One hundred female BC patients aged between 30-60 years old and diagnosed with invasive or infiltrating ductal carcinoma (IDC) have been selected, and incorporated into this study. The patients did not receive prior treatment such as endocrine therapy and/or chemotherapy.

Molecular subtyping of tumor samples. Molecular subtyping of tumor samples was performed at Kocaeli University Medical School Pathology department by analysing expressions of ER, PR, HER2, and Ki-67 proteins. The expressions of these proteins were evaluated using routine immunohistochemical methods (9). Based on the expression patterns, the tumor tissues were grouped into five different subtypes, namely Luminal A (ER+ and/or PR+, HER2−, Ki-67<15%); Luminal B (Her2−) (ER+ and/or PR+, HER2−, Ki-67≥15%); Luminal B-like (HER2+) (ER+ and/or PR+, HER2+, Ki-67≥15%); Her2-enriched (ER−, PR−, HER2+); triple negative BC (TNBC) (ER−, PR−, HER2−) (Table I) (10, 11). In overall, Luminal A, Luminal B, Luminal B-like, Her2-enriched, and TNBC groups were consisted of 34, 33, 14, 8, and 11 patients, respectively (Table II). The control group consisted of 100 samples obtained from healthy breast tissue.

Preparation of protein extracts. Tissue samples were thawed on ice and cut into small pieces (<1.00 mm) before centrifuged for 5 min at  $10,000 \times g$  to remove blood and the wash solution. Approximately 200 µl lysis buffer (7 M urea, 2 M thiourea, 10 mM Tris-HCl pH 8.0, 5 mM magnesium acetate, 4% CHAPS, pH 8.0) was added onto 0.15 g minced tissues and homogenized with Scilogex homogenizer (Rocky hill, CT, USA) for about 10 s at 20,000 rpm, on ice. Further homogenization was achieved with a bead-beater (Blut blender, Next Advance, Troy, NY, USA) using 0.2 mm stainless steel beads. The homogenates were centrifuged for 15 min at  $10,000 \times g$  to obtain a crude supernatant which was further centrifuged for 45 min at  $15,000 \times g$  to obtain a clear protein extract. Protein concentrations were determined by a modified Bradford assay (Bio-Rad, Hercules, CA, USA).

Two-dimensional gel electrophoresis (2-DE). Protein extract (500 µg) was loaded onto 11 cm non-linear pH 3-10 immobilized pH gradient (IPG) strips (Bio-Rad) by passive rehydration. First dimension separation based on isoelectric points was performed with a Protean isoelectric focusing cell (Bio-Rad) using the following conditions: 20 min at 250 V with rapid ramp, 2 h at 4,000 V with slow ramp, and 8,000 V with rapid ramp until a total of 25,000 V/h was reached, at 20°C. Following isoelectric focusing, strips were washed with buffer I (6 M Urea, 375 mM Tris-HCl pH 8.8, 2% SDS, 20% glycerol, 2% w/v DTT) for 15 min and then with buffer II (6 M Urea, 375 mM Tris-HCl pH 8.8, 2% SDS, 20% glycerol, 2.5% w/v iodoacetamide) for 15 min at room temperature. Second dimension separation based on molecular weights was achieved by sodium dodecyl sulphate-polyacrylamide gel electrophoresis (SDS-PAGE; 12%) using Criterion Dodeca gel running system (Bio-Rad). Gels were stained with colloidal Coomassie Blue and visualized by VersaDoc4000 MP (Bio-Rad) using QuantityOne software (Version 4.6.7, Bio-Rad).

Fluorescence difference gel electrophoresis (DIGE). Protein extracts were pooled in equal amounts and were labelled with CyDye minimal fluors using the CyDye<sup>TM</sup> DIGE Fluor Minimal Labeling Kit (GE Healthcare, Chicago, IL, USA); the experiments were carried out in dark. In brief, 50 μg of protein sample was used in each experiment. The pH of the extracts was adjusted to 8.5, and then protein pools were incubated with the dyes Cy2, Cy3, and Cy5 at 4°C for 30 min. Labelling reactions were stopped by adding 10 mM lysine. Labelled protein samples were then combined and a standard 2-DE experiment was performed.

Analysis of 2-DE and DIGE gel images. The Coomassie-stained 2-DE gels were analysed with PDQuest Advance 2-DE Analysis Software (Version 8.0.1 build 055, Bio-Rad) which provided the tools for spot intensity calibration, spot detection, and background subtraction. Two separate gels were run for each pooled group to minimize the experimental variation. Stain speckles were filtered and the standardized areas of interest from all gels were warped and matched and the quantity of each spot was normalized by the total valid spot intensity using linear regression model. The statistical significance of image analysis was determined by the Student's t-test (statistical level of p<0.05 was considered significant). Protein spots with differences in expression more than 2-fold were selected and excised from gels using ExQuest Spotcutter (Bio-Rad).

DIGE gels were also visualized with VersaDoc MP4000 (Bio-Rad) using three different light sources. After automatic spot detection by using PDQuest Advance Analysis Software, spots were manually revised with edition tools to prevent mismatches. Spot picking was performed using a preparative 2-DE gel, which was stained with colloidal Coomassie blue. The preparative gel was aligned with the DIGE reference image to outline the spots of interest selected during DIGE analysis. All selected spots were excised as described above.

In-gel tryptic digestion and matrix-assisted laser desorption/ionisation-time of flight mass spectrometry (MALDI TOF/TOF) analysis. In-gel digestion was performed using the in-gel tryptic digestion kit (Thermo Fisher Scientific) according to the manufacturer's instructions. Digested peptides were then desalted and concentrated using C18 ZipTips (Millipore, Burlington, MA, USA). Peptides were eluted with a matrix solution containing 50% (v/v) acetonitrile (Sigma Aldrich, St. Louis, MO, USA) 0.1% (v/v) trifluoroacetic acid (Sigma Aldrich) and 10 mg/ml alpha-cyano-4-hydroxycinnamic acid (α-CHCA) (Sigma Aldrich), and were spotted onto MALDI target plate. An AB SCIEX TOF/TOF 5800 instrument (AB Sciex, Framingham, MA, USA) was used to generate peptidemass fingerprints (PMFs) for protein identification (12).

Tandem mass spectrometry (MS/SM) data analysis. PMFs were searched in the MASCOT database (Matrix Science, Boston, MA, USA) using ProteinPilot software 4.0.8085 revision 148085 (AB Sciex). Search parameters included; human as taxonomy, trypsin as the enzyme,  $\pm 50$  ppm as the precursor tolerance,  $\pm 0.4$  Da as the MS/MS fragment tolerance, and carbamidomethyl and MetOxidation as the variable modifications. Only hits with p < 0.05 value were accepted for protein determination.

Western blot (WB) analysis. Proteins were separated by SDS-PAGE (12%) and transferred onto a nitrocellulose membrane (GE Healthcare, Chicago, IL, USA) using a Trans-Blot SD Semi-Dry Transfer Cell (Bio-Rad). The membrane was blocked in TBS-T buffer (Tris-HCl 25 mM pH 7.2, NaCl 150 mM, and 0.1% Tween 20) containing 5% non-fat dry milk for 1 h at room temperature and washed with TBS-T for three times before incubation of primary antibody diluted in TBS-T for 1 h at room temperature or overnight at 4°C. The membrane was then washed three times with TBS-T and incubated with goat antimouse HRP-labelled secondary antibody (#170-5047; Bio-Rad) for 1 h at room temperature. Following subsequent three washes with TBS-T, protein bands were visualized with an enhanced chemiluminescence detection system (Bio-Rad). Monoclonal anti-β-actin antibody (sc-81178; Santa Cruz Biotechnology, Santa Cruz, CA, USA), monoclonal

Table I. Surrogate definitions of intrinsic sub-types of breast cancer. (Her2-: Score 1, score 2; Her2+: Score 3).

	IHC status						
Intrinsic subtype	ER	PR	Her2*	Ki-67 (%)			
	+	+	_	<15			
Luminal A	+	_	_	<15			
	_	+	_	<15			
	+	+	_	≥15			
Luminal B	+	_	_	≥15			
	_	+	_	≥15			
	+	+	+	≥15			
Luminal B-like Her2+	+	_	+	≥15			
	_	+	+	≥15			
TNBC	_	_	_	any			
Her2-enriched	-	-	Over-expressed	any			

IHC, Immunohistochemistry; ER, estrogen receptor; PR, progesterone receptor; TNBC, triple-negative breast cancer. \*Her2-: score 1 or score 2; Her2+: score 3.

Table II. Clinical parameters of breast cancer patients.

	n	
Molecular subtype		
Luminal A	34	
Luminal B	33	
Luminal B-like Her2+	14	
TNBC	11	
Her2-enriched	8	
Clinical variables		
Histological tumor type		
Invasive ductal	93	
Infiltrative ductal	7	
TNM staging*		
I	13	
II	63	
III	22	
IV	2	
Tumor grade**		
1	17	
2	56	
3	27	

TNBC, Triple-negative breast cancer.

anti-glycerol-3-phosphate dehydrogenase NAD+, cytoplasmic antibody (anti-GPD1) (sc-376219; Santa Cruz Biotechnology), and monoclonal anti-monoglyceride lipase antibody (anti-MAGL) (sc-398942; Santa Cruz Biotechnology, USA) were used at the respective dilutions of 1:1000, 1:750, and 1:750. Either QuantityOne (Bio-Rad) or ImageJ (https://imagej.nih.gov/ij/index.html) was used for relative quantitative assessments of the band intensities when necessary.

Cell culture. CHO (Chinese hamster ovary cells), SH-SY5Y (human neuroblastoma cells), HeLa (human cervical cancer cells) and MCF-7

(human breast adenocarcinoma cells) cell lines were purchased from ATCC (Manassas, VA, USA). DP (dental pulp derived mesenchymal stem cells) was purchased from Kocaeli University Stem Cell Research Center (KOGEM, Kocaeli, Turkey). The cells were cultured in DMEM (Lonza, Basel, Switzerland) low-glucose medium supplemented with 10% fetal bovine serum (FBS) (Lonza), 100 µg/ml penicillin/streptomycin (Lonza) and 2 mM L-glutamine (Lonza) at 37°C in a humidified 5% CO2 atmosphere. Primary breast adenocarcinoma cells (PCL) were isolated using a commercial kit (Cancer Cell Isolation Kit, Thermo Scientific) from a TNBC tissue sample and grown in DMEM high-glucose medium supplemented with 10% heat-inactivated FBS, 100 µg/ml penicillin-streptomycin, 2 mM L-glutamine, and human epidermal growth factor (EGF; 10 ng/ml) at 37°C in a humidified 5% CO<sub>2</sub> atmosphere. Confluent cells (70-80%) were washed with PBS for three times and removed from the plates by scraping. Cell pellet was created by centrifugation at  $1,500 \times g$  for 10 min at 4°C. Mammalian protein extraction reagent (M-PER, Thermo Scientific) containing protease inhibitor cocktail (Roche, Indianapolis, IN, USA) was added over each cell pellet and cells were homogenized with a bead-beater using stainless steel beads. The soluble protein fraction was obtained by centrifugation at 10,000 × rpm for 10 min and 15,000 × rpm for 45 min at 4°C. Protein concentrations were measured as described above. Protein extracts were stored at -80°C after snap-frozen in liquid nitrogen.

Immunofluorescence (IF) microscopy. Cells were also grown on culture plates containing 12 mm poly-D-lysine-coated cover slips (Corning Inc., Corning, NY, USA) for IF microscopy. Cells on cover slips were fixed with 3% formaldehyde and permeabilized with PBS containing 0.5% Triton X-100. Anti-GPD1 and anti-MAGL antibodies (Santa Cruz) were used at the dilution of 1:100. Goat anti-mouse Texas Red antibody (T-6390; Thermo Scientific) was used as the secondary antibody with a dilution of 1:100. The nuclei were stained with DAPI (Molecular Probes, Eugene, OR, USA) (0.5% in PBS) and cover slips were mounted on glass slides using Mowiol<sup>®</sup>. Cells were observed with an inverted Olympus CKX41 microscope (Olympus, Tokyo, Japan) with appropriate filters. Images were processed using ImageJ software and localization of the proteins was assessed.

Bioinformatics analysis. STRING analysis (https://string-db.org) was carried out using the UniProt accession numbers of the regulated proteins. The search engine option was set to "multiple proteins by names/identifiers" and the organism was specified as Homo sapiens. The retrieved proteins were manually checked to assure that they were all correctly retrieved from the database. Whole genome analysis was the preferred choice. The setting tab was used to change the stringency of the analysis. The parameters for STRING search included interaction score of medium confidence, selected maximum number of interactors for the 1st and 2nd shells of no more than 5 interactors, active interaction sources mining, experiments, databases, co-expression, neighbourhood, gene fusion and co-occurrence. The hits considered in this study had false discovery rates (FDR) smaller than 1e-04. DAVID analysis (version 6.8; David Bioinformatics Resources) (https://david.ncifcrf.gov) was carried out using the UniProt accession numbers of the regulated proteins. Principal component analysis (PCA) was performed using the add-in statistical software XLSTAT (Addinsoft, New York, NY, USA). Average spot intensities were exported from PDQuest Advance as tab delaminated Excel data. Missing intensities were estimated using mean mode approach in XLSTAT and descriptive statistics and correlations were calculated. Venn diagrams were created using the on-line tool available at: http://bioinformatics.psb.ugent.be/webtools/Venn/.

Statistical analysis. 2-DE experiments were performed in duplicate while WB experiments were performed in triplicate. When it was required quantitative results were expressed as means±standard deviation (SD). The *p*-values were calculated using one-way analysis of variance (ANOVA) or Student *t*-tests in XLSTAT software (Addinsoft, New York, USA). All *p*-values <0.05 were considered statistically significant. Only significant hits, as defined by the MASCOT probability analysis (*p*<0.05), were accepted.

#### Results

Preparation of protein pools. SDS-PAGE gels were run to assess the quality and the quantity of each protein extract after measurement of protein concentrations. In terms of quality, each sample displayed distinct bands without any evidence of smear indicating that the protein extracts required no further clean up (Figure 1A). Moreover, WB analysis was performed using anti-beta-actin antibody to ensure equal protein loading (13). The protein pools were accurately recreated according to WB quantification results (Figure 1B).

Comparative 2-DE analysis revealed 13 proteins that were differentially regulated in the tumor groups compared to the control. Well-resolved and reproducible 2-DE gels were produced and subjected to spot detection (Figure 2). An average of 750 spots per analytical gel was detected. The overall mean coefficient for spot matching was 24%, indicating that the protein distribution patterns among the gels were highly similar. Protein spots that significantly differed in expression (more than 2-fold) were selected and identified by MALDI-TOF/TOF analysis. The changes in spot intensities were compared for each tumor protein pool in comparison to the control pool (e.g., Luminal A versus control or Luminal B versus control), while comparisons between tumor groups (e.g., Luminal A versus Luminal B) were not considered in this study. Lists of the identified proteins with their respective MALDI scores and their corresponding regulation ratios among the tumor groups over the control group are given in Tables III and IV, respectively.

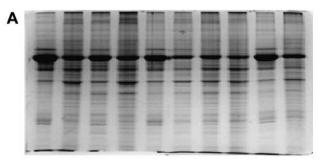
A Venn diagram was created using the accession numbers of differentially regulated proteins in each comparative analysis (tumor groups *versus* control group) (Figure 3). Thirteen shared proteins were differentially regulated in each tumor group in comparison to the control (Figure 4). Those proteins were elongation factor 2 (EEF2), 60 kDa heat shock protein (HSPD1), heat shock protein 90-alpha (HSP90AA1), monoacylglycerol lipase (MAGL), pyruvate kinase (PK), protein disulfide-isomerase (PDIA1), elongation factor 1-gamma (EEFG1), L-lactate dehydrogenase B chain (LDHB),

carbonic anhydrase 1(CA1), aldo-keto reductase family 1 member C2 (AKR1C2), cytoplasmic glycerol-3-phosphate dehydrogenase [NAD+] (GPD1), keratin type II cytoskeletal 8 (KRT8), and keratin type I cytoskeletal 19 (KRT19). When regulation ratios of these thirteen proteins were analysed, MAGL, LDHB, CA1, AKR1C2 and GPD1 displayed low expression levels while EEF2, HSPD1, HSP90AA1, PKM, PDIA1, EEFG1, KRT8 and KRT19 displayed high expression levels in tumor groups in comparison to the control group.

Comparative DIGE analysis revealed 5 proteins downregulated in tumor tissues compared to the control. For DIGE analysis, a separate protein pool representing each tumor subtype was formed. This protein pool was then compared with the control protein pool (Figure 5). Based on the 2-fold change criteria, differentially regulated proteins were detected and identified by MALDI-TOF/TOF (Table V). GPD1 received our attention since it was highly downregulated in the tumor tissues in comparison to control tissues in the DIGE experiment with a similar trend observed in 2-DE experiments. Among the proteins that were differentially regulated in the tumor groups compared to the control, we focused on two that showed down-regulation and were involved in the same pathways. Specifically, GPD1 and MAGL, which are involved in triacylglycerol metabolism (14) were selected for further verification experiments.

WB analysis verified the changes in GPD1 and MAGL protein levels. The protein pools from the tumor and the control groups were analysed by WB using anti-GPD1 and anti-MAGL monoclonal antibodies. The results indicated low levels of both proteins in the tumor pools in comparison to the control, thus verifying the findings of 2-DE and DIGE experiments. Moreover, comparisons among the tumor groups revealed that the TNBC group had the lowest level of GPD1 and MAGL expression, while the other tumor groups displayed similar GPD1 expression levels (GPD1: >1000-fold difference, p=0.03; MAGL: >100-fold difference, p=0.04) (Figure 6A and B).

GPD1 protein levels varied among the TNBC samples. To further determine the variation in GPD1 expression levels among individual patients in the tumor groups, eight samples from each tumor subtype were selected and subjected to WB analyses (Figure 7A). Except five samples of the TNBC and three samples of the Her2<sup>+</sup> group, GPD1 expression was detected in each sample at a similar level. However, GPD1 levels in each sample belonging to tumor groups were always lower than the GPD1 levels in the control group. For the samples belonging to the TNBC group, 5/8 samples displayed extremely low/non-detectable the GPD1 expression. Thus, WB experiments were repeated to screen all the available



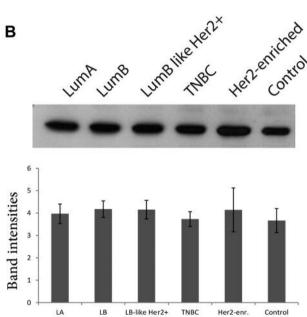


Figure 1. Normalization of protein concentrations in protein pools. Selected protein extracts were analyzed by SDS-PAGE (A). Western blot (WB) analysis of the protein pools of each tumor and control groups after normalization with actin. The band intensities were measured by performing densitometric analysis using ImageJ software (B).

TNBC samples (n=11). The results indicated that nearly half of the TNBC samples had non-detectable GPD1 expression. Each control sample belonging to the TNBC group (peripheral healthy breast tissues collected from the same TNBC patients) was also subjected to WB analyses to assess the changes in GPD1 levels (Figure 7B). Two out of the 11 control samples also displayed extremely low GPD1 levels. Regarding Her2<sup>+</sup> samples, there was also variation in GPD1 levels but it was not as significant (>2-fold change) as that observed in the TNBC samples (data not shown).

Based on these findings, all TNBC samples (11 patients) were selected for further analysis by WB using anti-MAGL antibody, in order to determine the variation in expression levels of MAGL among individual patients in the tumor groups. Two samples (P2 and P9) displayed relatively low levels of MAGL

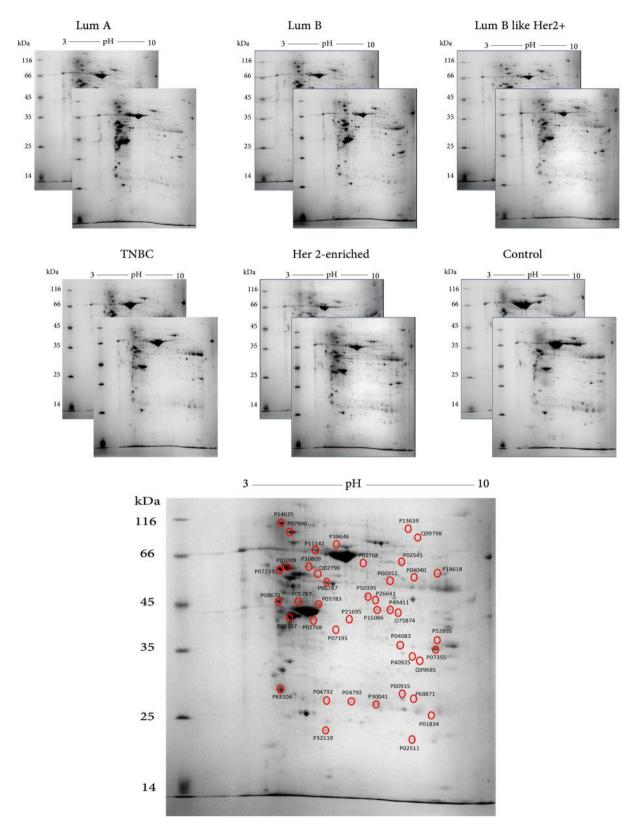


Figure 2. Images of the two-dimensional electrophoresis (2-DE) gels for tumor and control groups. The representative gel marked with hollow circles and labelled with Uniprot accession numbers indicates the positions of the identified protein spots on the gel. Lum, Luminal; TNBC, triplenegative breast cancer.

Table III. List of the differentially regulated proteins identified by two-dimensional gel electrophoresis coupled to matrix-assisted laser desorption/ionisation-time of flight mass spectrometry.

	Accession number	Protein description	Protein mass	Protein score	Expect	Matches	Theoretical pI	Seq. cov.
1	P02768	Serum albumin	69321	272	1.3e-023	37	5.92	37
2	P01009	Alpha-1-antitrypsin	46707	335	6.4e-030	37	5.37	55
3	P30041	Peroxiredoxin-6	25019	414	8.1e-038	31	6	75
4	P05787	Keratin, type II cytoskeletal 8	53671	594	8.1e-056	53	5.52	64
5	P32119	Peroxiredoxin-2	21878	366	5.1e-033	24	5.66	74
6	P07355	Annexin A2	38580	508	3.2e-047	38	7.57	58
7	P49411	Elongation factor Tu, mitochondrial	49510	337	4E-30	34	7.26	59
8	P63104	14-3-3 protein zeta/delta	27728	205	6.4e-017	28	4.73	61
9	P08670	Vimentin	53619	566	5.1e-053	52	5.06	60
10	P04792	Heat shock protein beta-1	22768	293	1E-25	23	5.98	67
11	P50395	Rab GDP dissociation inhibitor beta	50631	401	1.6e-036	45	6.11	62
12	P13639	Elongation factor 2	95277	301	1.6e-026	48	6.41	39
13	P08727	Keratin, type I cytoskeletal 19	44065	590	2e-055	45	5.04	69
14	P15086	Carboxypeptidase B	47338	381	1.6e-034	30	6.16	59
15	P10809	60 kDa heat shock protein, mitochondrial	61016	480	2E-44	41	5.7	49
16	P05783	Keratin, type I cytoskeletal 18	48029	563	1E-52	43	5.34	56
17	P02545	Lamin-A/C	74095	304	8.1e-027	42	6.57	47
18	Q02790	Peptidyl-prolyl cis-trans isomerase FKBP4	51722	153	1E-11	31	5.35	49
19	P14625	Endoplasmin	92411	373	1e-033	45	4.76	36
20	P07900	Heat shock protein (HSP) 90-alpha	84607	288	3.2e-025	42	4.94	49
21	P11142	Heat shock cognate 71 kDa protein	70854	341	1.6e-030	39	5.37	40
22	Q99685	Monoglyceride lipase	33240	183	1E-14	25	6.49	62
23	Q99798	Aconitate hydratase, mitochondrial	85372	303	1E-26	36	7.36	39
24	P14618	Pyruvate kinase isozymes M1/M2	57900	269	2.6e-023	41	7.96	61
25	P07237	Protein disulfide-isomerase	57081	329	2.6e-029	34	4.76	49
26	P26641	Elongation factor 1-gamma	50087	132	1.3e-009	22	6.25	32
27	P00352	Retinal dehydrogenase 1	54827	92	1.2e-005	16	6.3	30
28	P38646	Stress-70 protein, mitochondrial	73635	80	0.00018	18	5.57	27
29	O75874	Isocitrate dehydrogenase [NADP] cytoplasmic	46630	325	6.4e-029	41	6.53	57
30	P07195	L-lactate dehydrogenase B chain	36615	269	2.6e-023	28	5.71	48
31	P00915	Carbonic anhydrase 1	28852	258	3.2e-022	20	6.59	70
32	P04083	Annexin A1	38690	521	1.6e-048	36	6.57	63
33	P02511	Alpha-crystallin B chain	20146	268	3.2e-023	23	6.76	51
34	P01834	Ig kappa chain C region	11602	120	0.00000002	10	5.58	88
35	P68871	Hemoglobin subunit beta	15988	211	1.6e-017	17	6.75	83
36	P52895	Aldo-keto reductase family 1 member C2	36712	209	2.6e-017	25	7.13	56
37	P21695	Glycerol-3-phosphate dehydrogenase [NAD+], cytoplasmic	37543	373	1E-33	37	5.81	75
38	P40925	Malate dehydrogenase, cytoplasmic	36403	204	8.1e-017	21	6.91	47
39	P04040	Catalase	59719	290	2e-025	28	6.9	38

Seq. cov., Sequence coverage.

expression, while repeated WB experiments using all the TNBC samples confirmed the findings and indicated that one fifth of the TNBC group displayed low level MAGL expression. Each control sample belonging to the TNBC group was also subjected to WB analyses to assess the changes in MAGL levels. One of the control samples (P9) displayed low level of MAGL expression. The corresponding tumor sample of P9 also displayed low level of MAGL expression (Figure 7C).

PCA of the tumor groups and the control group. By taking 2-DE spot intensities on every replicate tumor group and the control

group, the variations were projected onto a 2-dimensional space. Therefore, we were able to determine which groups were similarly contributing to the variance and which groups appeared in a separate region (Figure 8A). PCA succeeded to differentiate the tumor groups from the control group. PCA also provided the evidence for the separation of TNBC from the other BC subtypes, which was indicative of variation in their respective protein expression profiles.

Comparative DIGE analysis for the GPD1-positive and -negative TNBC samples. Analysis of the WB results revealed

Table IV. The list of differentially regulated proteins and their corresponding regulation ratios among the tumour groups over the control group.

No.	Accession number	SSP no.	Protein description	Luminal A vs. Control	Luminal B vs. Control	Luminal B-like Her2 +	TNBC vs. Control	Her2-enriched vs. Control
						vs. Control		
1	P02768	5603	Serum albumin	NR	0.21	0.33	NR	NR
2	P01009	1504	Alpha-1-antitrypsin	NR	NR	0.29	NR	0.49
3	P30041	6001	Peroxiredoxin-6	NR	NR	NR	NR	NR
4	P05787	4505	Keratin, type II cytoskeletal 8	8.81	9.53	10.12	3.16	3.7
5	P32119	4002	Peroxiredoxin-2	NR	NR	NR	NR	NR
6	P07355	8104	Annexin A2	NR	NR	2.52	NR	2.81
7	P49411	6202	Elongation factor Tu. mitochondrial	2.84	3.13	2.02	NR	NR
8	P63104	1001	14-3-3 protein zeta/delta	3.26	NR	2.62	2.11	3.01
9	P08670	0204	Vimentin	NR	NR	NR	NR	NR
10	P04792	5002	Heat shock protein beta-1	NR	2.18	NR	NR	NR
11	P50395	5301	Rab GDP dissociation inhibitor beta	2.5	2.53	2.34	2.22	NR
12	P13639	7805	Elongation factor 2	3.12	5.93	10.41	3.23	4.39
13	P08727	2202	Keratin. type I cytoskeletal 19	14.74	12.07	10.26	3.12	5.57
14	P15086	6203	Carboxypeptidase B	5.24	7.4	NR	NR	NR
15	P10809	3602	60 kDa heat shock protein. mitochondrial	4.7	6.52	5.3	2.64	6.07
16	P05783	4201	Keratin. type I cytoskeletal 18	6.33	7.69	8.96	NR	NR
17	P02545	7602	Lamin-A/C	2.5	3.18	3.81	NR	NR
18	Q02790	4502	Peptidyl-prolyl cis-trans isomerase FKBP4	NR	4.69	4.18	0.26	NR
19	P14625	1801	Endoplasmin	2.38	2.59	NR	2.11	2.12
20	P07900	1802	Heat shock protein HSP 90-alpha	5.14	6.58	4.01	3.11	4.81
21	P11142	3703	Heat shock cognate 71 kDa protein	NR	2.08	2.32	NR	2.71
22	O99685	8110	Monoglyceride lipase	0.18	0.05	0.18	0.01	0.05
23	Q99798	8802	Aconitate hydratase. mitochondrial	NR	2.21	2.31	NR	NR
24	P14618	8603	Pyruvate kinase isozymes M1/M2	6.14	11.26	9.98	6.94	7.02
25	P07237	0504	Protein disulfide-isomerase	3.14	2.84	3.92	4.1	6.73
26	P26641	6302	Elongation factor 1-gamma	6.01	7.22	7.88	8.4	8.38
27	P00352	6502	Retinal dehydrogenase 1	NR	NR	NR	NR	NR
28	P38646	4708	Stress-70 protein. mitochondrial	NR	0.24	0.16	NR	2.69
29	O75874	7203	Isocitrate dehydrogenase [NADP]. cytoplasmic	: NR	NR	NR	NR	2.09
30	P07195	4109	L-lactate dehydrogenase B chain	0.03	0.01	0.04	0.01	0
31	P00915	7002	Carbonic anhydrase 1	0.35	0.21	0.34	0.19	0.4
32	P04083	6101	Annexin A1	NR	NR	NR	NR	NR
33	P02511	8012	Alpha-crystallin B chain	NR	0.45	0.4	NR	NR
34	P01834	8004	Ig kappa chain C region	0.27	0.21	NR	2.04	NR
35	P68871	8006	Hemoglobin subunit beta	0.46	NR	0.26	NR	NR
36	P52895	8206	Aldo-keto reductase family 1 member C2	0.06	0.06	0.03	0.06	0.07
37	P21695	5101	Glycerol-3-phosphate dehydrogenase	0.06	0.06	0.02	0.001	0.03
			[NAD+]. cytoplasmic					
38	P40925	7106	Malate dehydrogenase. cytoplasmic	NR	NR	NR	0.4	NR
39	P04040	8506	Catalase	NR	NR	NR	NR	NR

SSP no., Standard spot number; TNBC, triple-negative breast cancer; NR, non-relevant (Indicates less than two-fold up/down regulation).

two subgroups with different GPD1 expression levels in the TNBC group. The first group consisted of the samples with low/non-detectable GPD1 expression (GPD1<sup>-</sup>) and the second of those with moderate GPD1 expression (GPD1<sup>+</sup>). This finding led us perform DIGE analysis to compare GPD1<sup>+</sup> and GPD1<sup>-</sup> samples at the proteome level (Figure 8B). Sixteen proteins displayed regulation in their expression levels (Table VI).

GPD1 and MAGL expression in different cell lines. The expression levels of GPD1 and MAGL were further

examined *in vitro*, in CHO, SH-SY5Y, HeLa, DP, MCF-7 and PCL cell lines. All cell lines expressed both proteins when cultured. Among the cell lines, the lowest expression level of GPD1 was observed in the primary PCL cells which were obtained from a TNBC patient. The low GPD1 expression level in this sample was in agreement with our previous findings and demonstrated that TNBC samples either lacked or had low level of GPD1 expression (Figure 7A). As to MAGL, its expression levels fluctuated among the cell lines (Figure 9A). The expressions of the two

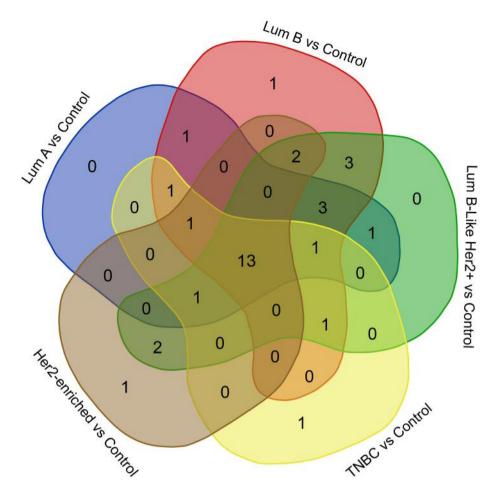


Figure 3. Venn diagram created to reveal identities of the shared proteins among the differentially regulated proteins in each comparative analysis of tumor groups against the control group. Lum, Luminal; TNBC, triple-negative breast cancer.

proteins in the cell lines were also confirmed by IF imaging. Results showed a uniform cytoplasmic distribution of both proteins within the cells (Figure 9B).

# Discussion

STRING analysis using differentially regulated proteins listed in Table IV created an interactome which led to the pathways associated with protein folding and unfolded protein response (UPR). STRING analysis of the differentially regulated proteins determined by DIGE also confirmed the findings of 2-DE, pointing to the pathways associated with protein folding and UPR. Similar to our findings, Pozniak *et al.* (2016) reported changes in protein levels associated with UPR. In their study, formalin-fixed paraffin-embedded BC tissue samples were analyzed and down-regulation of UPR-related proteins was found in tumors compared to normal tissue (15). UPR activation

indicates the presence of cellular stress related to endoplasmic reticulum due to accumulation of unfolded or misfolded proteins (16). Sustained over-activation of the UPR has also been implicated in several different human solid tumors (17-21). In general, UPR is expected in solid tumors because human solid tumors contain areas of pure vascular perfusion which causes inadequate blood circulation in the tumor microenvironment and severe oxygen and glucose deprivations. UPR is expected to protect the tumor cells in response to adverse conditions (22). It has been demonstrated that in antiestrogen-treated cancer cells, and in cells that have increased protein production and secretion, UPR activation is mild and protective (23). UPR activation has also been linked with inhibition of chemosensitivity to drugs (24). Therefore, markers of UPR activation may provide clinical relevance for BC therapeutic approaches.

Some of the differentially regulated proteins of this study can be linked to carbon metabolism. Those proteins include

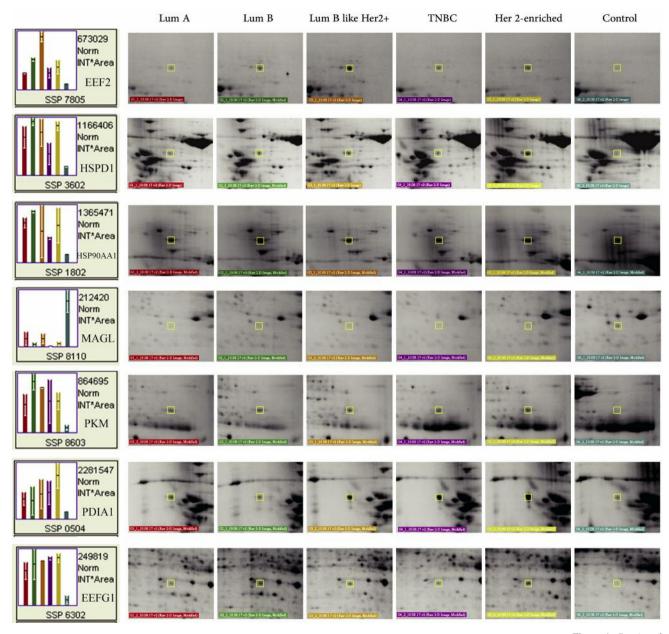


Figure 4. Continued

aconitate hydratase (Q99798), malate dehydrogenase (P40925), pyruvate kinase (P14618) and lactate dehydrogenase (P07195). It has been long known that cancer cells can reprogram their central carbon metabolism and maintain elevated level of glycolysis to sustain their energy needs and convert glucose to lactate (25-28). During this conversion, pyruvate kinase (PK) catalyses the last step of glycolysis converting phosphoenolpyruvate to pyruvate. Pyruvate is then converted to lactate by lactate dehydrogenase (LDH) to provide sufficient reducing power for energy production (29). In our study, we

observed up-regulation in PK in all BC subtypes in comparison to the control. The up-regulation of PK in cancer cells can be linked to the higher cell proliferation ability and increased anabolic functions (30, 31). In literature, there is evidence indicating that the expression of LDH isoforms differs in healthy and cancerous tissues and LDH-A is the isoenzyme which plays the key role in regulating glycolysis in tumor cells. In fact, carcinomatous tissues possess high LDH-A isoform (32-34). However, the LDH isoform identified in this study was, LDH-B (48% sequence coverage). So far, there was no

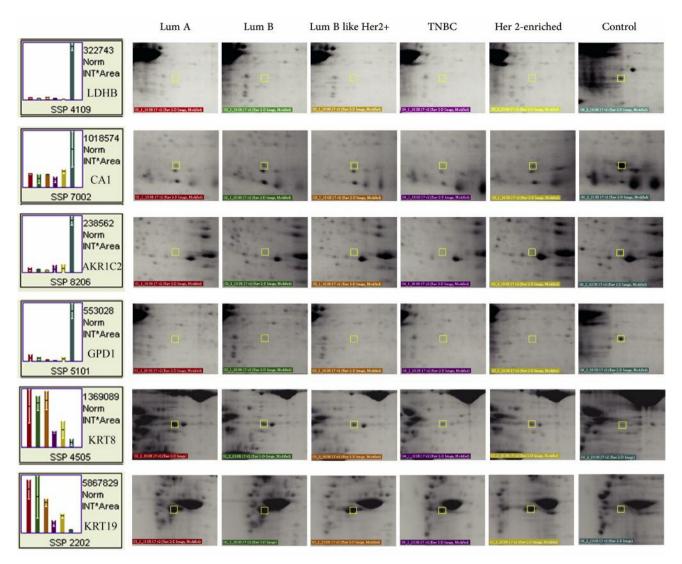


Figure 4. Representative close-up images of the differentially regulated proteins. The proteins shared in the Venn diagram (n=13) were determined by comparing tumor groups with the control group. The determined the proteins were used to create the close-up images. Lum, Luminal; TNBC, triple-negative breast cancer.

report emphasizing the changes occurring at LDH-B levels. To the best of our knowledge, this is the first study reporting regulation in LDH-B levels in all subtypes of BC.

The role of lipid metabolism has been conferred to explain the aggressive properties of malignant cancers (35). In the literature, dysregulated core enzymes in lipid metabolism and their contribution to the cancer cell migration, invasion and metastasis have been well established (35-37). The genes involved in cellular fatty acid uptake and *de novo* lipogenesis are amplified in metastatic tumors and fatty acid synthesis plays an important role in cancer pathogenesis (38, 39). The liberated stored fatty acids and the newly-synthesized fatty acids are rapidly incorporated into neutral-lipid and phospholipid stores.

Herein, two differentially regulated proteins, namely MAGL and GPD1 helped us establish an association between triglyceride metabolism (TGM) and BC.

GPD1 is a member of the NAD<sup>+</sup>-dependent glycerol-3-phosphate dehydrogenase family and plays a critical role in carbohydrate and lipid metabolism by catalysing the reversible conversion of dihydroxyacetone phosphate (DHAP) and reduced nicotinamide adenine dinucleotide (NADH) to glycerol-3-phosphate (G3P) and NAD<sup>+</sup> (40, 41). Based on the Human Protein Atlas, GPD1 expression can be detected in most malignant tumors including renal and hepatocellular carcinomas, malignant gliomas and prostate cancers (http://v15.proteinatlas.org/ENSG00000167588-GPD1/cancer).

Table V. The list of differentially regulated proteins identified by two-dimensional electrophoresis fluorescence difference gel coupled to matrix-assisted laser desorption/ionisation-time of flight mass spectrometry.

	Accession number	Protein description	Protein mass	Protein score	Expect	Matches	Theoretical pI	Seq. cov.
1	P09211	Glutathione S-transferase P	23341	68	0.0031	7	5.43	31
2	P04792	Heat shock protein beta-1	22768	214	8.1E-18	15	5.98	46
3	A6NL28	Putative tropomyosin alpha-3 chain-like protein	26253	97	0.0000039	5	4.47	24
4	P62258	14-3-3 protein epsilon	29155	80	0.0002	14	4.63	44
5	P01834	Ig kappa chain C region	11602	80	0.00019	6	5.58	76
6	P02511	Alpha-crystallin B chain	20146	68	0.0031	7	6.76	33
7	P60174	Triosephosphate isomerase	26653	66	0.0055	9	6.45	35
8	P35232	Prohibitin	29786	101	0.0000016	8	5.57	19
9	P62937	Peptidyl-prolyl cis-trans isomerase A	18001	138	3.2E-10	16	7.68	58
10	P63261	Actin. cytoplasmic 2	41766	92	1.4e-005	12	5.31	33
11	P05787	Keratin. type II cytoskeletal 8	53671	252	1.3e-021	24	5.52	45
12	P30101	Protein disulfide-isomerase A3	56747	147	4E-11	17	5.98	33
13	P05783	Keratin. type I cytoskeletal 18	48029	192	1.3e-015	22	5.34	30
14	P10809	60 kDa heat shock protein. mitochondrial	61016	95	0.0000069	12	5.70	22
15	P20774	Mimecan	33901	99	0.0000023	8	5.46	29
16	Q15084	Protein disulfide-isomerase A6	48091	151	1.6E-11	12	4.95	24
17	P08727	Keratin. type I cytoskeletal 19	44065	449	2.6e-041	29	5.04	46
18	P11021	78 kDa glucose-regulated protein	72288	190	2e-015	15	5.07	20
19	P08670	Vimentin	53619	209	2.6E-17	21	5.06	30
20	P07237	Protein disulfide-isomerase	57081	160	2E-12	12	4.76	22
21	P27797	Calreticulin	48112	116	0.000000051	12	4.29	25
22	P04083	Annexin A1	38690	266	5.1E-23	15	6.57	40
23	P04264	Keratin. type II cytoskeletal 1	65999	101	0.0000016	14	8.15	27
24	P60709	Actin. cytoplasmic 1	41710	96	0.0000049	9	5.29	16
25	P23141	Liver carboxylesterase 1	62841	76	0.00049	8	6.15	13
26	P21695	Glycerol-3-phosphate dehydrogenase [NAD+]. cytoplasmic	37543	150	2e-011	22	5.81	43
27	Q13228	Selenium-binding protein 1	52358	114	0.000000081	15	5.93	20

Seq. cov., Sequence coverage.

Table VI. List of differentially expressed proteins of glycerol-3-phosphate dehydrogenase 1+/- group by two-dimensional electrophoresis fluorescence difference gel coupled to matrix-assisted laser desorption/ionisation-time of flight mass spectrometry.

No.	Accession number	Protein description	Protein mass	Protein score	Expect	Matches pI	Theoretical (%)	Seq. cov.
1	P14618	Pyruvate kinase isozymes M1/M2	57900	178	3.2e-014	22	7.96	31
2	P06733	Alpha-enolase	47139	138	3.2e-010	18	7.01	28
3	P00558	Phosphoglycerate kinase 1	44586	119	2.6e-008	24	8.3	52
4	P04083	Annexin A1	38690	96	5e-006	22	6.57	46
5	P14550	Alcohol dehydrogenase [NADP+]	36550	90	1.9e-005	21	6.32	39
6	P00338	L-lactate dehydrogenase A chain	36665	65	0.0063	15	8.44	24
7	P02647	Apolipoprotein A-I	30759	81	0.00016	17	5.56	32
8	P08758	Annexin A5	35914	224	8.1e-019	24	4.94	43
9	P07237	Protein disulfide-isomerase	57081	161	1.6e-012	22	4.76	20
10	P08670	Vimentin	53619	133	1.0e-008	25	5.06	44
11	P13645	Keratin. type I cytoskeletal 10	58792	80	0.00022	21	5.13	24
12	P60709	Actin. cytoplasmic 1	41710	172	1.3e-013	26	5.29	44
13	P02787	Serotransferrin	77000	259	2.6e-022	47	6.81	45
14	P52565	Rho GDP-dissociation inhibitor 1	23193	152	1.3e-011	20	5.02	40
15	P63104	14-3-3 protein	27728	106	5.1e-007	20	4.73	43
16	P62258	14-3-3 protein epsilon	29155	69	0.0027	21	4.63	41

Seq. cov., Sequence coverage.

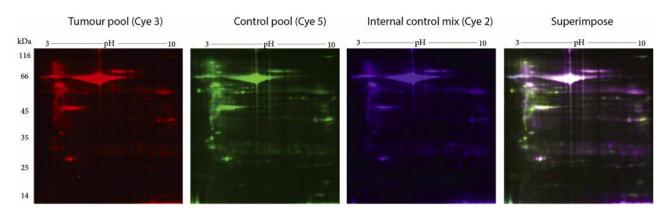


Figure 5. Representative images of the two-dimensional electrophoresis fluorescence difference (DIGE) gels used for comparative analysis. Pooled protein extracts from the tumor and control samples were labelled with Cye3 and Cye5, respectively. The internal control mix (Cye2-labeled), used for normalization, was formed by mixing equal amounts of protein from the tumor and the control pools. The superimposed image was the product of superimposing all three Cye dye images; images were pseudo-coloured.

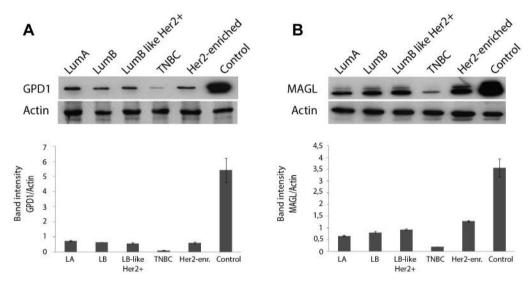
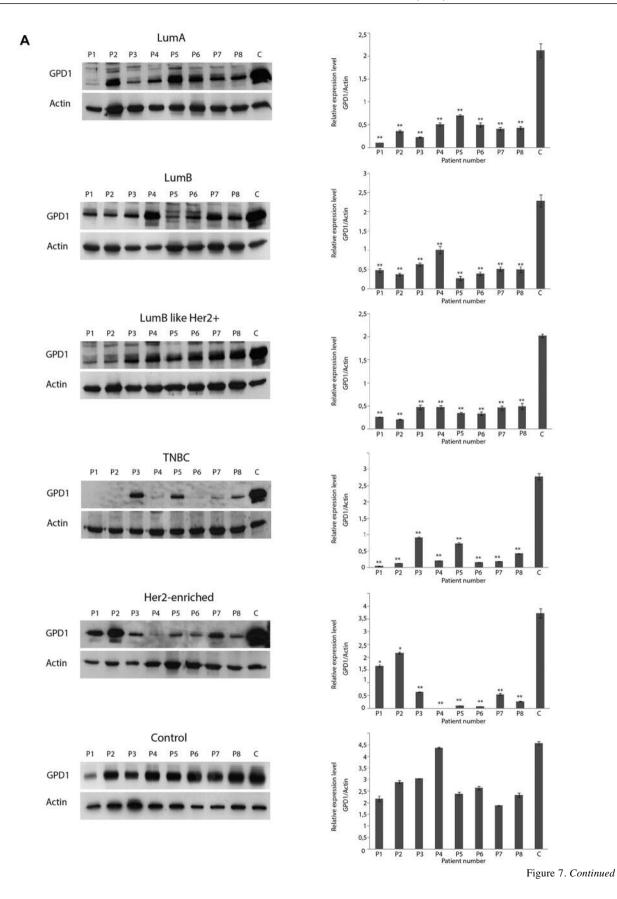


Figure 6. Western blot analysis of the tumor groups and the control group using anti- glycerol-3-phosphate dehydrogenase 1 (anti-GPD1) antibody (A) and anti-monoacylglycerol lipase (anti-MAGL) antibody (B) to determine the changes on the protein expression levels. The band intensities were measured by performing densitometric analysis using ImageJ software. Lum, Luminal; TNBC, triple-negative breast cancer.

The relative expression levels of GPD1 among 20 different cancer types indicate that GPD1 is moderately expressed in BC tissues compared to other cancer types. In the present study, we found that GPD1 was highly down-regulated in all BC subtype groups in comparison to the control. In addition, the lowest GPD1 expression was detected in the TNBC samples. In fact, in some TNBC samples, GPD1 expression could not be detected at all. This was in line with a recent study by Zhou *et al.* demonstrating that the mRNA expression level of GPD1 was found to be significantly down-regulated in BC patients (42). The same study revealed several important findings including higher GPD1 expression levels in receptor (ER and PR) positive

and Her2 negative subtypes, shorter survival time in patients with low GPD1 expression, prognostic GPD1 value for ERpositive tumors, significant down-regulation at GPD1 mRNA levels in BC cell lines and significant down-regulation in GPD1 protein levels in BC cell lines and tissues. In the present study, we have observed similar GPD1 protein expression levels in all BC subtypes, except the TNBC. Luminal A displayed the highest GPD1 expression at the protein level. The lowest level of GPD1 protein expression was observed in TNBC samples, implicating that not only Her2 reactivity but also ER/PR positivity is associated with GPD1 expression levels. Analysis of GPD1 expression levels in each sample revealed some



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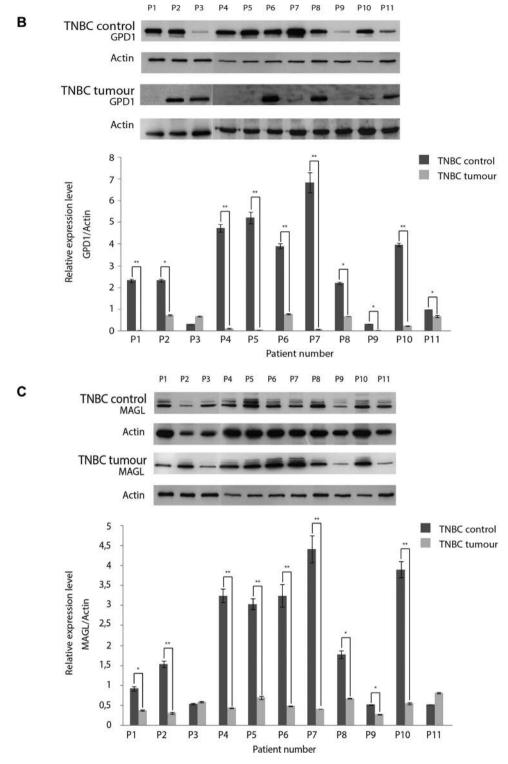
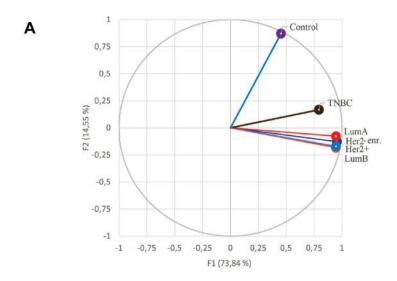


Figure 7. Glycerol-3-phosphate dehydrogenase 1 (GPD1) expression levels were determined by western blot (WB) in eight individual samples from each tumor group and the control group. The blot was stripped and re-probed with an anti-actin antibody for normalization of the expression levels. The band intensities were measured with the QuantityOne software. The letters "P" and "C" designate the patient number and the control pool formed by the combination of all individual controls, respectively. \*p<0.05 and \*\*p<0.01 compared to the control pool (A). WB analysis was also performed for all individual triple negative breast cancer (TNBC) samples and their respective controls to determine the changes in GPD1 (B) and monoacylglycerol lipase (MAGL) (C) levels. \*p<0.05 and \*\*p<0.01 for within subject TNBC control. Data presented as mean $\pm$ SD (n=3).



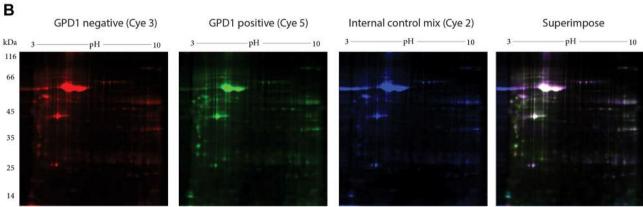


Figure 8. Principle component analysis (PCA) was performed using XLSTAT, an add-in software developed by Addinsoft. Average spot intensities were exported from the PDQuest Advance as tab delaminated Excel data (A). DIGE images of the pooled glycerol-3-phosphate dehydrogenase 1 (GPD1-) (Cye3) and pooled GPD1+ (Cye5) triple negative breast cancer (TNBC) samples. The internal control mix (Cye2), used for normalization, was formed by mixing equal amount of protein from the pooled GPD1+ and the pooled GPD1- TNBC samples. The superimposed image was the product of superimposing all three Cye dye images; images were pseudo coloured (B). Lum, Luminal.

variation among the individual samples, mainly in TNBC and Her2<sup>+</sup> groups. This variability is an expected result considering the heterogeneous nature of BC subtypes. BC is a heterogeneous disease that differs greatly among different patients (inter-tumor heterogeneity) and even within the same tumor sample (intra-tumor heterogeneity) (43). Some of the samples in the TNBC group had either no or barely detectable GPD1 expression; thus, TNBC samples were suggested to be divided into two different groups, GPD1<sup>+</sup> and GPD1<sup>-</sup>.

TNBC represents 12-15% of all BCs (44). Its molecular heterogeneity is reflected onto a significantly different clinical outcome among other subtypes. Very often, this cancer is much more aggressive causing distant metastasis (45). Therefore, it is necessary to establish a detailed classification for the TNBC sub-group. Statistically, high

percentages of TNBCs belong to the basal-like BC group which has high mitotic index and high histological grade (46, 47). Basal-like BCs are identified by a variety of immunohistochemical markers, such as cytokeratins (CK5, CK6, CK14 or CK17), EGFR, SMA, p-cadherin, p63 or c-kit antigen (48). Herein, we were curious about whether GPD1 negative samples represented the basal-like BCs. We thus determined CK5 and CK6 expressions in eight of those samples and evaluated the coherence between CK5/6 and GPD1 expressions. Except one sample, there was an inverse correlation between CK5/6 status and GPD1 expression levels. Samples which lacked GPD1 expression had positive immunohistochemical staining for CK5/6, while the samples which expressed GPD1 were negative for CK5/6. This finding indicated that GPD1 may be a good candidate

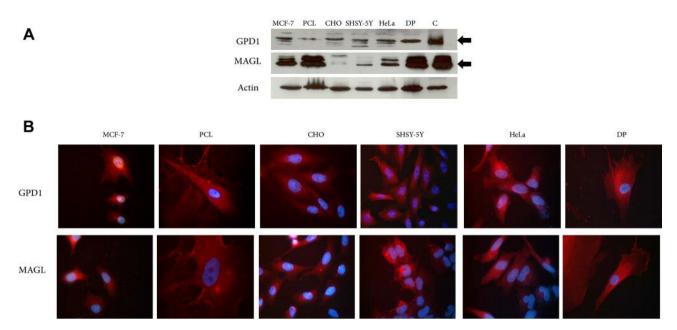


Figure 9. Protein extracts from MCF-7, primary breast cancer cell line (PCL), CHO, SHSY-5Y, HeLa and dental pulp-derived mesenchymal stem cells (DP) were analyzed by western blot for glycerol-3-phosphate dehydrogenase 1 (GPD1) and monoacylglycerol lipase (MAGL) expressions; protein pools prepared from the control breast tissues were used as positive control (C) (A). Immunofluorescence of the cells to demonstrate expression of GPD1 and MAGL. Red color depicted the distributions of GPD1 and MAGL. Cell nuclei were stained with DAPI (B).

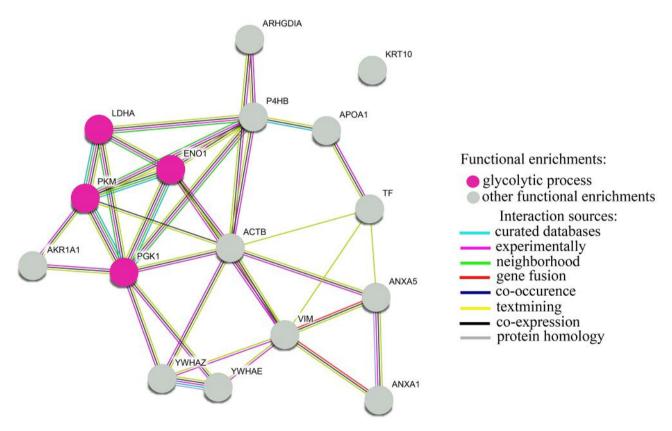


Figure 10. STRING analysis of the differentially regulated proteins identified in two-dimensional electrophoresis fluorescence difference performed to compare GPD1+ and GPD1- groups.

biomarker to differentiate basal-like BCs from other TNBC subtypes (47). The results of comparative DIGE experiments with GPD1<sup>+</sup> and GPD1<sup>-</sup> samples revealed the presence of 16 differentially regulated proteins. STRING analysis with the differentially regulated proteins listed in Table VI indicated a significant change in glycolytic process (FDR of 8.56e-05) (Figure 10). The glycolytic proteins that displayed downregulation in GPD1+ samples in comparison to GPD1samples were lactate dehydrogenase, pyruvate kinase, phosphoglycerate kinase 1 and alpha-enolase. It is likely that the heterogenic complex nature of TNBC may require flexible cell metabolism in which non-enzymatic functions of most glycolytic enzymes may offer important contributions (49). Future studies should consider whether the changes in the levels of these glycolytic enzymes may help TNBC cells to sustain their growth.

MAGL is the enzyme responsible for the breakdown of monoacylglycerols (MAG) to free fatty acids (FFA) (50). It is best recognized for its role in degrading endogenous 2-arachidonoylglycerol (2-AG) in the brain and the peripheral tissues (51). MAGL is considered to be one of the key enzymes which regulate the network of FFAs in numerous aggressive tumors such as colorectal cancer, neuroblastoma, and nasopharyngeal carcinoma (52-54). Increased aggressiveness requires increased lipid synthesis by lipolytic pathway to liberate stored fatty acids by MAGL. This was demonstrated in a seminal paper by Nomura et al. (55) who showed that MAGL was highly expressed in aggressive human cancer cells and primary tumors, where it regulated a fatty acid network enriched in oncogenic signalling lipids. Aggressive cell lines displayed much greater in vitro migration and in vivo tumor growth rates in comparison to their non-aggressive counterparts. Moreover, inhibition of MAGL demonstrated that aggressive cancer cells display highly elevated MAG activity, most of which comes from the MAGL enzyme. Blocking MAGL by either siRNA or an inhibitor reduced in vitro migration, invasion, and cell survival under serum deprivation; while, overexpression of MAGL in non-aggressive cancer cells increased their pathogenicity. The work by Nomura et al. opened a way to investigate whether MAGL could be a potential therapeutic target and prognostic indicator for various cancer types (50). In our study, we have identified MAGL as one of the regulated proteins in BC subtypes. Interestingly, our 2-DE experiments demonstrated that MAGL was down-regulated in all BC subtypes in comparison to the control. More specifically, the TNBC subtype showed the lowest MAGL expression compared to the other BC subtypes. Some individual patient samples had much less MAGL levels in comparison to the other patient samples indicating variability in MAGL levels amongst the patients. Similarly, there was also variation in MAGL levels among the controls.

Contrary to the results by Nomura et al., the current study demonstrated a decrease in the MAGL expression level in the tumor tissues compared to their counterparts. This conundrum cannot be explained with the current knowledge about the known function of MAGL. The complex interplay between oncogenic signalling and lipid metabolism and the large spectrum of lipid functions at both the cellular and organismal level requires more detailed understanding of the alterations in lipid metabolism in cancer. A possible reason to explain the discordance between our data and that by Nomura et al., is that they performed activity-based proteomic analysis and did not monitor the changes in MAGL protein levels. On the other hand, we performed non-functional proteomic analysis and only determined the changes in MAGL protein levels. MAGL overexpression causes significant reduction in MAGs and elevation in FFAs (50). Based on this knowledge, we propose that the in vivo aggressiveness of cancer cells depends on the level of MAGs rather than the level of FFAs. However, currently there is no supporting data for our proposal in the literature. Despite the recent research indicating the relationship between MAGL and the tumor progression, the mechanism of pro-tumor activity of MAGL is still unknown. The data presented here raises a point for MAGL to be a candidate therapeutic target and prognostic indicator for BC. Histological examination of tissue samples may be useful for clinical application of MAGL for BC diagnosis and subtyping. It may be worth investigating the changes in serum levels of MAGL to develop a more reliable and convenient clinical test for BC diagnosis and prognosis.

Cancerous state is acquired through reprogramming the metabolic pathways to deviate cells from the normal state (56). There is a growing appreciation that the regulation of metabolism defines the features of cancer (57). In this context, the work by Tyanova et al. shed some light onto the changes in metabolic events occurring in BC tissues (58). They analyzed changes in a vast number of proteins using LC-MS/MS technology and described significant changes occurring in energy metabolism (glycolysis and gluconeogenesis), amino acid metabolism, replication/repair, cell growth/death and translation. In another recent study, Yanovich et al. aimed at detailed analysis of BC subtypes and reported changes in the levels of vast number of proteins (59). Their study also involved comparison of their findings with the past studies and concluded that certain metabolic pathways display changes in certain BC subtypes such as changes in galactose metabolism, oxidative phosphorylation and immune activity in Her2+, ER+ and TNBC BC subtypes, respectively.

#### Conclusion

The currently used diagnostic biomarkers for BC are either not specific or do not allow early detection of the disease. The classical radiological scans and the other clinical approaches do

not appear to be powerful and convenient enough to allow early detection. During the last decade, novel biomarker candidates have emerged, however they also lack sensitivity and specificity (60-62). On the other hand, panels of gene/protein biomarkers have been proposed, but their use is limited due to their high cost and availability issues. The difficulty in finding a biomarker for early detection lies in the complexity of pathways that transform cells into cancerous state. Understanding the complex interplay among these pathways will ultimately lead to discovery of a simple and effective biomarker/biomarker panel. However, this requires a detailed analysis of the genome, transcriptome, metabolome and proteome. Herein, the changes occurring in the proteome profile of BC subtypes with respect to their healthy counterparts were analyzed. The findings placed an emphasis on the relationship among triacylglycerol metabolism and BC-subtypes and suggested that the alterations occur in two key proteins, GPD1 and MAGL. These proteins may have high potentials to become diagnostic /prognostic /predictive biomarkers.

## **Authors' Contributions**

KKY contributed to experiment conduction and prepared the data for publication. MK and GA contributed to experiment conduction and manuscript preparation. AG performed the surgeries, collected and classified the tissue samples. BG performed the pathological examinations of tissue samples and helped their classification. NZU performed the surgeries and contributed to tissue sample collection and clinical evaluation of the data.

# **Ethics Approval and Consent to Participate**

This study was approved by the Ethics Committee of Kocaeli University (approval numbers: KOU KAEK 2015/274 and KU GOKAEK 2018/49). Informed consents, approved by the institutional ethics committee, were obtained from each patient.

## **Conflicts of Interests**

The authors declare that they have no competing interests.

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