Reprogramming SREBP1-dependent lipogenesis and inflammation in high-risk breast

with licochalcone A: a novel path to cancer prevention

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Abstract

Background: Anti-estrogens have had limited impact on breast cancer (BC) prevention. Novel

agents with better tolerability, and efficacy beyond estrogen receptor (ER) positive BC are

needed. We studied licochalcone A (LicA) for ER-agnostic BC prevention.

Methods: We evaluated antiproliferative effects of LicA in seven breast cell lines and its

suppression of ER+ and ER- xenograft tumors in mice. High-risk human breast tissue was

treated with LicA *ex vivo*, followed by RNA sequencing and metabolism flux modeling.

Confirmatory testing was performed in an independent specimen set and ER+/- BC cell lines

using NanoString metabolic panel, proteomics, western blots, and spatiotemporally resolved

cholesterol quantification in single cells.

Results: LicA suppressed proliferation *in vitro* and xenograft tumor growth *in vivo*. It

downregulated pivotal steps in PI3K-AKT-SREBP1-dependent lipogenesis, suppressed PI3K

and AKT phosphorylation, SREBP1 protein expression, and cholesterol levels in the plasma

membrane inner leaflet, to the levels in normal breast cells. LicA also suppressed prostaglandin

E2 synthesis and PRPS1-catalyzed *de novo* nucleotide biosynthesis, stalling proliferation;

further evident by reduced MKI67 and BCL2 proteins.

Conclusions: LicA targets SREBP1, a central regulator of lipogenesis and immune response,

reducing pro-tumorigenic aberrations in lipid homeostasis and inflammation. It is a promising

non-endocrine candidate for BC prevention.

Key words: Anti-inflammation, antiproliferation, breast cancer prevention, licochalcone A,

metabolism, PI3K, SREBP

Background

Current recommendations for BC prevention are based on estimates of risk in individual women [1]; for those at markedly increased risk due to high-penetrance germline mutations, risk-reducing bilateral mastectomy can be considered. However, the vast majority of at-risk women are at risk for other reasons, with risk estimates in the 1.5-5-fold range. For them, risk-reducing medications are recommended. At present, the only risk-reducing drugs available for BC are anti-estrogens; aromatase inhibitors (AIs) for postmenopausal women and selective estrogen receptor modulators (SERMs) for pre and postmenopausal women. These drugs can lower the risk of estrogen receptor positive (ER+) BC by 50-65% [2]. However, of the estimated 10 million U.S. women who could benefit from these medications more than 85% decline them, mainly due to their adverse effects: vasomotor symptoms, thrombogenesis/stroke, bone loss, and endometrial cancer [2-7]. In addition, these risk reducing drugs do not prevent estrogen receptor negative (ER-) BC. One of the pathways abnormally active in precancerous lesions such as atypical hyperplasia and ductal carcinoma in situ (DCIS) is PI3K-AKT [8-10]. PI3K pathway inhibitors such as everolimus and alpelisib have been proposed for repurposing as ER- and ER+ BC prevention agents [11-13]; however, their side effects such as hyperglycemia are a major impediment [14]. There have been attempts to repurpose widely used and well-tolerated drugs such as metformin for BC risk reduction; however, clinical trial results have been disappointing [15, 16]. Therefore, there is a clear need to have alternative agents with sufficient efficacy, minimal adverse effects, and greater acceptance.

Upregulated *de novo* fatty acid and cholesterol biosynthesis mediated by the activation of the PI3K-AKT-SREBP axis has been shown to have an important role in BC development and progression through enhanced proliferation, inter and intracellular signaling, and enabling evasion from immune surveillance [17, 18]. Correspondingly, metabolic dysregulation and obesity are associated with the elevated risk: ER+ disease in post-menopausal women and ER- BC in those who are premenopausal [19-24].

The activity of the master regulator of lipogenesis, sterol regulatory element binding protein 1 (SREBP1) has been shown to have prognostic value in BC progression [25, 26]. SREBP1 plays a major role in macrophage polarization and regulation of inflammasome in the tissue microenvironment and forms a complex with NF-kB that regulates the inflammatory response [27, 28]. Excess activation of SREBP1 upregulates the expression of most fatty acid biosynthesis enzymes and SREBP2-dependent cholesterol biosynthesis [29]. This subsequently promotes an inflammatory/oxidative microenvironment, provides the excess cell membrane

structural components necessary for proliferating cells, protects malignant cells from immune

destruction, promotes cancer stem cell viability, and supplements cholesterol and its derivatives

as precursors for estrogen biosynthesis [30-33]. In addition, the role of dysregulated lipogenesis

in promoting ER- BC is associated with the inflammatory cytokines secreted from immune cells

recruited to the stroma consequent to excess free fatty acids and cholesterol in the

microenvironment, and the dysregulation of metabolic hormones and growth factors [33-35].

The cancer-promoting effects of inflammation [36-38] can be tempered through downregulating

SREBP1 and through the activation of NRF2 and suppression of NF-kB-dependent pathways.

[28, 39]. This presents an opportunity for interventions that prevent malignant transformation in

the breast. Importantly, such a strategy will not produce deleterious endocrine responses that

are of concern with presently available risk reduction drugs [39].

Our current research reveals that licochalcone A (LicA), a compound that is already in clinical

trials for the relief of inflammatory skin conditions, is an excellent candidate as a non-endocrine

BC risk-reducing drug with minimal toxicity and better acceptance. It has antioxidant and anti-

inflammatory effects [40-52], moderately inhibits aromatase (CYP19A1), the rate-limiting

enzyme in estrogen biosynthesis, and the target of AIs used for BC prevention [41]. In addition,

LicA blocks estrogen carcinogenic metabolism through the downregulation of CYP1B1, *in vivo*

[45]. LicA has also been reported to relieve conditions such as arthritis, mastitis, and acute lung

injury through anti-inflammatory pathways, and appears to have osteoprotective effects [48, 49,

52-54]. The NF-kB-dependent anti-inflammatory effects of LicA can be partly explained by its

binding to cysteine residues in KEAP1 which simultaneously enhances NRF2-dependent

antioxidant effects [42] while suppressing inflammation [55], but it can also be a consequence of

its effects on SREBP1-dependent pathways [28]. In the current study, we confirm our previous

in vitro and *in vivo* antioxidant observations using LicA in breast microstructures from women at

high risk of BC. We also present novel findings regarding its effects on metabolic pathways

involving PI3K-AKT-SREBP1 which we have elucidated in various breast cell lines, human *ex*

vivo, and *in vivo* models using multiple orthogonal approaches. In addition, we show its

antiproliferative effects resulting from changes in metabolic flux, particularly through decreasing

the *de novo* nucleotide biosynthesis step of the pentose phosphate pathway (PPP). Based on

these results and existing literature, we posit that reducing lipogenesis and inflammation

through suppressing SREBP1 can reduce the risk of BC and that LicA presents as an excellent

candidate for BC prevention and is highly likely to be safe and well-tolerated.

Methods

Chemicals and Materials. All chemicals and reagents were purchased from Sigma-Aldrich (St. Louis, MO), unless otherwise indicated. MammoCult media kit, heparin, and hydrocortisone were purchased from Stem Cell Technologies (Vancouver, BC. Canada). F12-K nutrient mix (Kaighn's) medium was acquired from Gibco (Dublin, Ireland). Fetal bovine serum (FBS) was purchased from Atlanta Biologicals (Norcross, GA). Collagenase I was purchased from Sigma-Aldrich (St. Louis, MO). Licochalcone A (LicA) was acquired from Med Chem Express (Monmouth Junction, NJ). Direct-Zol RNA prep kit was acquired from Zymoresearch (Irvine, CA). TRIzol was obtained from Invitrogen (Waltham, MA). RNAeasy mini prep kit was purchased from QIAGEN (Germantown, MD). RNA concentration and clean up kit was acquired from Norgen Biotek (ON, Canada). KAPA RNA HyperPrep library preparation kit was obtained from Roche (Madison, WI). Unique dual adaptors for next generation sequencing (NGS) and PCR reagents, primers, and master mix were purchased from Integrated DNA Technologies (Coralville, IA). Antibodies were obtained from Abcam (Cambridge, UK). Western blot reagents, buffers, supplies were purchased from Thermo Fisher Scientific (Hanover Park, IL.).

Cell Culture. We grew pre-malignant DCIS.COM and malignant MCF-7 and MDA-MB-231 cells (all obtained from ATCC) in RPMI-1640 supplemented with 10% fetal bovine serum (FBS) and 1% penicillin/streptomycin (Invitrogen Thermo Fisher Scientific, Hanover Park, IL.). We also grew pre-malignant DCIS.COM/ER+ PR+ cells obtained from Dr. Dean Edwards's laboratory (Baylor College of Medicine) [56] in DMEM/F12 supplemented with 10% horse serum, 0.5% penicillin/streptomycin, and 1% HEPES. In addition, we acquired MCF-7 cells overexpressing aromatase (MCF-7aro) from Dr. Shiuan Chen's laboratory (City of Hope) [57] and grew them in MEM supplemented with 10% FBS, 2 mM L-glutamine, 1 mM sodium pyruvate, 1% nonessential amino acids, and 1 U/mL penicillin/streptomycin. BRCA defective malignant breast cell lines HCC-1937 and HCC-3153 were obtained from Dr. Gazdar's group (University of Texas Southwestern Medical Center) and were grown in RPMI-1640 supplemented with 5% FBS and 1% penicillin/streptomycin.

Human breast microstructure preparation and treatment. We prepared breast microstructures from the unaffected contralateral breast of postmenopausal women who underwent bilateral mastectomy for unilateral BC, using the protocol of Tanos et. al., with minor modifications [41, 58, 59]. The tissues were obtained while fresh and diced to approximately 5 mm pieces before digesting at 37°C with 2% collagenase I in F12-K nutrient mix (Kaighn's) medium, overnight. After the completion of digestion, the tissue was centrifuged at 250 x g for 5

min and the supernatant was removed. The pellet was washed with phosphate buffered saline

and was mixed and cultured in MammoCult media supplemented with 0.2% heparin and 0.5%

hydrocortisone. After 24 h incubation at 37°C the treatments were mixed with fresh MammoCult

media and added to the microstructures. **Ethics declarations:** Institutional Review Board (IRB)

approval (IRB STU00202331) was obtained from Northwestern University prior to obtaining

informed consents from patients and collecting samples. All experiments were conducted in

accordance with the approved protocol and guidelines.

RNA sequencing. After 24 h incubation with LicA (5 µM) or DMSO, the microstructure pellets

obtained from 6 subjects were prepared and washed with HBSS. Trizol was added to each

pellet and the protocol of Direct-Zol RNA prep kit was used to extract RNA. The protocol

included a step of DNAse I treatment for removing DNA contamination of the RNA preparations.

RNA was eluted in nuclease free water and quality assessment was performed using Agilent

Bioanalyzer and Qubit. Libraries were made using Roche KAPA Biosystems protocol (KAPA

RNA Hyper Prep Kit Technical Data Sheet, KR1352 – v4.17, Roche Corporate) and the quality

was evaluated. Total RNA sequencing was performed with Illumina NovaSeq 6000 v1.5, 2x100

b, paired end with 40 million reading depth, followed by quality assessment using FASTQC,

alignment to reference genome hg38, raw data processing using STAR, and data analysis to

define differential gene expression, as described previously [60-62]. Significantly (adj P < 0.05)

modulated genes were further analyzed with Enrichr to define highly modulated pathways. The

sample relationships were validated with NGSCheckMate as well.

NanoString metabolism pathway panel. We treated six additional subjects' breast

microstructures and two cell lines MDA-MB-231 (ER-) and MCF-7 (ER+) cells with LicA (5 µM

and 10 µM) or DMSO for 24 h. After extracting RNA and quality assessment we used the

NanoString nCounter Metabolism Panel (NanoString Technologies Inc.) to quantify RNA

expression from 749 genes. Differentially expressed genes and pathway analysis were

determined using ROSALIND platform (ROSALIND, Inc.) with normalization, fold changes (≥ 1.5

fold: upregulated or ≤ -1.5 fold: downregulated), and adjusted p-value < 0.05 using the

Benjamini-Hochberg method.

Genome-scale metabolic network modeling and flux-based analysis (FBA). We calculated

the relative activity of reactions in breast microstructures of 6 women by interpreting gene

expression data using the Recon1 human metabolic model outlined in Shen et al [63-65]. We

then identified a metabolic flux state that is most consistent with gene expression data in control

and LicA treated samples. This was achieved by maximizing the activity of reactions that are

associated with upregulated genes and minimizing flux through reactions that are
downregulated in a condition, while simultaneously satisfying the stoichiometric and
thermodynamic constraints embedded in the model using linear optimization [63-65]. The
glucose, fatty acid, and glutamine levels in the simulations were adjusted based on the growth
media used for culturing the high-risk women's breast microstructures. All p-values were
corrected for multiple comparisons.

Proliferation assay. We seeded pre-malignant DCIS.COM and DCIS.COM/ER+ PR+ cell lines
as well as malignant MCF-7 (ER+ PR+), MDA-MB-231 (ER- PR-), MCF-7aro (ER+ PR+), HCC-
1937 (ER- PR-, BRCA1 mutated), and HCC-3153 (ER- PR-, BRCA mutated1) cells with the
density of 2.5 x 10^3 cells/well in their appropriate media in 96-well plates. We placed the plate in
IncuCyte instrument for continued live cell imaging every 6 h. When cells reached 30%
confluency, eight different concentrations of LicA ranging from 350 nM to 40 µM were added to
the designated wells. We added comparable dilutions of DMSO to the wells allocated for vehicle
control. In the case of MCF-7aro cells which overexpress *cyp19a1* (aromatase), the dosing of
LicA or DMSO was performed in the presence of androstenedione, the substrate for aromatase.
We continued live cell imaging for a minimum of 4 days post treatment. When we found out that
a single low dose is not sufficient for sustained antiproliferation, we added the treatment and
control every 48 h for 6 days and continued live cell imaging for a maximum of 12 days. Data
was analyzed using Zoom software to quantify the images and were plotted as mean ± SEM of
at least two independent measurements.

Proteome Integral Solubility Alterations (PISA) proteomics. MCF-7 and MDA-MB-231 cells
were treated with LicA (10 µM) for 24 h. We used the sample preparation and PISA protocol as
described previously [66], with minor modifications. Cells were detached with non-enzymatic
dissociation buffer and centrifuged at 300 x g for 4 min. The cell pellets were reconstituted in
PBS containing protease and phosphatase inhibitors. This step was repeated. The final pellet
was resuspended in PBS with inhibitors and was distributed into PCR tubes. Thermal
denaturation was performed with a steady temperature at 40, 42, 45, 50, 55, 60, 63, and 65°C
for 3 mins in parallel in a thermal cycler. Immediately after heating, the samples were removed
and incubated at room temperature for 3 min before they were snap-frozen with liquid nitrogen.
Two cycles of freeze-thaw using liquid nitrogen and a heating block set at 25°C were conducted
to ensure a uniform temperature between tubes. After the vortex, the lysates were centrifuged at
20,000 x g for 20 min at 4°C to pellet the debris with precipitated and aggregated proteins.
While on ice, the supernatant with the soluble protein fraction was carefully removed to a new

tube. BCA assay quantified the protein concentration in the supernatant for the two lowest

temperatures. The average of the two lowest temperatures was calculated and the volume

equivalent to 30 µg of protein in the lowest temperature was moved from each temperature

fraction into a clean tube. Following the distribution of protein, each tube was brought to a final

volume of 100 µL by the addition of S-Trap lysis buffer with inhibitors. After acidifying the

samples, they were loaded onto the S-Trap column and centrifuged at 4,000 x g for 1 min to

trap the proteins. The column was washed with an S-Trap binding buffer to remove

contaminants. Trypsin in 50 mM TEAB was added to the S-Trap column and incubated at 37 °C

for 3 h. Peptides were eluted from the S-Trap column and dried using a vacuum concentrator.

The dried material was reconstituted in 100 mM TEAB. TMT reagent was added to the peptide

solution and incubated for 1 h at room temperature. The reaction was quenched with 5%

hydroxylamine. Peptides were cleaned using a desalting column. The labeled peptides were

fractionated using high PH reverse-phase chromatography to reduce sample complexity. The

fractionated peptides were analyzed using LC-MS/MS, and MS2 was used for quantification and

identification of peptides. Proteome Discoverer was used for quality control, statistical analysis,

and differential expression analysis [66]. The differentially expressed proteins were also

analyzed by Enrichr to obtain the significantly stabilized and destabilized pathways.

Western Blot. The Western blot analysis was performed as described before [42]. After 48 h

treatment of each compound, the cells were lysed with RIPA buffer (50 mM Tris-HCl pH 7.5,

0.1% SDS, 1% Triton X-100, 150 mM NaCl, 0.5% Sodium deoxycholate, and 2 mM EDTA)

containing protease/phosphatase inhibitor cocktail (Thermo Fisher Scientific, Waltham, MA,

USA) at 4°C for 1 h. The supernatants were collected and quantified using BCA protein assay

kit (Thermo Fisher Scientific) according to the manufacturer's instructions. To identify the

molecular weight, we used a regular range protein marker and precision plus protein dual color

standard marker. Protein of 10–12 µg was loaded on 4–12% sodium dodecyl sulfate

polyacrylamide gel electrophoresis (SDS-PAGE), and then transferred to polyvinylidene

difluoride membrane. The membranes were blocked with TBS-T buffer (Tris-buffered saline

containing 0.1% Tween 20) containing 5% (w/v) skim milk powder for 1 h at 25°C, and then

incubated with primary antibodies for 18 h at 4°C. After incubation with the secondary HRP-

linked antibody for 1 h at 25°C, the membranes were detected using Clarity™ Western ECL

Blotting Substrates (Bio-Rad, Hercules, CA, USA) and a Bio-Rad imager. Quantification of the

images was performed using the Image J software.

In situ quantitative imaging of cellular cholesterol. Spatiotemporally resolved *in situ* quantification of cholesterol in the inner leaflet of the plasma membrane (IPM) of mammalian cells was performed using a ratiometric cholesterol sensor, WCR-*e*Osh4 as described previously [67, 68]. WCR-*e*Osh4 was prepared and calibrated using giant unilamellar vesicles as described previously [67]. WCR-*e*Osh4 was microinjected into cells, and the cholesterol concentration in the IPM was determined using in-house programs written in MATLAB as described [67]. The three-dimensional display of the local cholesterol concentration profile was calculated using the Surf function in MATLAB.

In vivo studies. Female ovary-intact athymic nude mice were purchased from Jackson Laboratory at 6 weeks of age. Animals were acclimated for a week upon arrival and then were subcutaneously (s.c.) inoculated in flanks with either MCF-7 cells (1 million cells/animal, 12 animals) or MDA-MB-231 (1 million cells/animal, 18 animals). After the xenograft tumors reached the palpable size of 0.8 cm in diameter, the daily administration (s.c) of LicA (80 mg/kg.day) or vehicle (4% DMSO, 6% EtOH + 45% water + 45% PEG-400) was started and continued until day 28 after the first dose. Animals were weighed once per week and the tumor size was measured twice per week using a digital caliper. The size of tumors was calculated using the formula [Tumor volume = (A^2) x B/2] with "A" representing the small diameter and "B" representing the large diameter of the tumor. The tumor volume was plotted for each animal every time the tumor was measured. The growth of tumors in animals treated with LicA was compared with growth of tumors in the animals receiving the vehicle control. After the completion of the treatments, animals were sacrificed and the tumors were harvested, weighed, and preserved for future studies. It should be noted that if an animal met any humane endpoint criteria before the completion of the treatment course, as per the IACUC protocol, we sacrificed the animal. ***Ethics declaration:*** All the in vivo studies and procedures were performed under the approved Northwestern University IACUC protocol # IS00013602.

Statistical analysis. The regular ordinary least square regression (eg. simple linear regression) does not consider heterogeneity across groups or time. Therefore, we fit a linear model using generalized least squares (GLS) to compare the tumor size trend over time between groups using LicA and vehicle for MDA-MB-231 and MCF-7 xenografts. We also compared the marginal tumor size between groups using LicA and vehicle for MDA-MB-231 and MCF-7

280 xenografts through least-squares means. We compared different variance-covariance

281 structures, through ANOVA and selected the best fitted model with the smallest AIC.

Results

Licochalcone A alters gene expression and pathway activity in breast microstructures from high-risk postmenopausal women.

We exposed microstructures produced from fresh surgical breast tissue of 6 postmenopausal

women to LicA (5 µM) or DMSO for 24 h and performed total RNA-seq. LicA produces a

transcriptional profile that is very different from that of DMSO. Differential gene expression

analysis showed that in LicA-treated samples 2341 genes were downregulated (adj P < 0.05)

with 381 genes having logFC < -1. Down-regulated genes include *HMGCR*, *MVD*, *MVK*, *SQLE*,

ACAT2, *SREBF1*, *SREBF2*, *INSIG1*, *LSS*, *PTGST*, *RELA*, *ALDH1A3*, *SERPINB2*, *Wnt*, and

CYP1B1. In these same samples, a total of 1462 genes were upregulated (adj P < 0.05) with

538 genes having logFC > 1. Top ranked genes associated with the upregulated pathways

include *NQO1*, *NQO2*, *HMOX1*, *GST*s, *TXN*, *TXNRD1*, *SLC*s, *G6PD*, and *GCLC*. Pathway

enrichment analysis was performed, and the top downregulated, and upregulated pathways are

shown in Figure 1A and Figure 1B and in the associated supplementary data (S1A and S1B).

The analyses shown in Figure 1A and Figure S1A revealed that metabolic pathways including

lipids and cholesterol metabolism and biosynthesis, PI3K-AKT signaling, and pathways involved

in inflammatory response are among the top downregulated processes. In addition, cholesterol

and acetyl-CoA appeared as the predicted decreased metabolites (Figure 1A). In contrast, as

demonstrated in Figure 1B and Figure S1B, NRF2-dependent antioxidant pathways, pentose

phosphate pathway (PPP), glutathione metabolism, suppression of EGFR signaling, and

ferroptosis were among the top upregulated pathways. Transcription factor analysis (Figure 1C)

revealed that the profoundly upregulated transcription factor in LicA-treated microstructures is

NRF2 which is consistent with our previous findings [42] and the significantly downregulated

transcription factors in these samples are proliferative factors *SP1* and *KLF4*, inflammation

transcription factor *NF-kB* and its subunit *RELA*, and the lipogenic transcription factors sterol

regulatory element binding factor (*SREBF*) *1* and *SREBF2*.

Validation of sequencing results, breast microstructures from 6 additional subjects were

obtained. The NanoString metabolism panel consisting of 749 metabolic genes was employed

and the results were analyzed using ROSALIND software. The data demonstrated significant (P

< 0.05) downregulation of up to 3.5-fold change in *ACAT2*, *CYP1B1*, *L2HGDH*, *ACY1*, *SREBF2*,

NRF1, *AR*, *PLA2G15*, *AKT2*, and *NDC1* expression and significant (P < 0.05) upregulation of up

to 1.9-fold change in *SLC7A11*, *GCLC*, *TXNRD1* expression (Figure 1D). The analysis also

showed that pathways associated with cholesterol homeostasis, mTORC1 signaling, reactive

oxygen species metabolism, and heme metabolism were significantly modulated (adj P < 0.03)

by LicA.

Licochalcone A changes flux through metabolic pathways to facilitate support for

antioxidant effects, reduced lipogenesis, and reduced proliferation.

Further analysis of the sequencing data in the first 6 high-risk women's tissue microstructures

was performed to define flux through metabolic pathways using genome-scale metabolic

modeling (Figure 2) [69, 70]. The analysis showed that flux through 191 reactions was

significantly (p < 0.05) modulated as a function of exposure to LicA. Some of the top modulated

metabolic reactions are key steps in glycolysis/gluconeogenesis, nucleotide biosynthesis and

catabolism, the pentose phosphate shunt, pyruvate metabolism, cholesterol metabolism, fatty

acid elongation, and steroid metabolism (Figure 2A). Interestingly, when the results were further

explored at the metabolic reaction level using the BIGG database [71], the direction of the flux

for many reactions was in favor of generating excess NAD(P)H and creating a

reducing/antioxidant environment (Figure 2B). Examples of these reactions were aldehyde

dehydrogenase in glycolysis/gluconeogenesis, glucose-6-phosphate dehydrogenase and

phosphogluconate dehydrogenase in PPP, 17-beta hydroxysteroid dehydrogenase in steroid

metabolism, and methylene tetrahydrofolate reductase in folate metabolism (Figure 2B). Flux

through the NADPH-generating oxidative branch of PPP which is irreversible was significantly

increased (Figure 2C). Flux through the non-oxidative branch of PPP which is reversible was

toward the generation of ribose 5 phosphate (R5P) (Figure 2C). However, flux through PRPPS,

which is responsible for *de novo* nucleotide biosynthesis, was not in favor of PRPP generation;

rather the production of R5P was favored (Figure 2C) with the subsequent enhanced flux

through PPM-catalyzed reaction which is involved in nucleotide catabolism and salvage

production of nucleotides. Thus, the consumption of R5P and generation of R1P, which

supports efficient recycling of nucleotides for DNA repair or energetics, is favored, as opposed

to R5P formation essential for nucleotide *de novo* biosynthesis and cell proliferation (Figure 2C).

Overall nucleotide flux (Figure 2D) offered further confirmation of the above findings. We

observed the dominance of the salvage nucleotide biosynthesis process and its connection to

antioxidant reactions such as NADPH-dependent formation of reduced thioredoxin (Figure 2D).

This was further made evident when we observed that the formation of ADP-ribose from AMP

and R5P is favored over the formation of R5P which can lead to reduced proliferation (Figure 2B

and 2D). Similarly, we observed that flux through several reactions in cholesterol metabolism

(Figure S2) such as 4-methyl zymosterol, 24-dehydrocholesterol reductase, 7-

dehydrocholesterol reductase, lathosterol oxidase, cholesterol efflux, and several cholesterol

transporters were in a direction to generate less cholesterol and excess NADPH (Figure S2).

These were consistent with the differential gene expression results (Figure 1). In addition, flux

through fatty acid elongation reactions catalyzed by desaturase enzymes such as stearyl CoA

desaturase (SCD) were significantly modulated to enhance the ratio of saturated to

monounsaturated fatty acids, and to generate excess NADPH, as a function of exposure to LicA

(Figure 2E).

Licochalcone A suppresses the proliferation of pre-malignant and malignant breast cells.

Our transcriptomic data and metabolic flux modeling studies suggested LicA has

antioxidant/anti-inflammatory effects along with decreased production of nucleotides which are

required for cell proliferation. These observations encouraged us to evaluate how and at what

concentration(s) LicA affects breast cell proliferation. We selected two pre-malignant breast cell

lines; DCIS.COM (ER- PR-) and DCIS.COM (ER+ PR+), and 5 malignant cell lines MCF-7 (ER+

PR+), MCF-7aro (ER+ PR+ overexpressing CYP19A1), MDA-MB-231 (ER- PR-), HCC1937

(ER- PR-, BRCA1 mutation), and HCC3153 (ER- PR-, BRCA1 mutation). Exposing these cell

lines to various concentrations (range of 350 nM – 40 µM) of LicA we observed that a single

dose at 20 µM or 40 µM was able to exert a sustained suppression of proliferation in DCIS.COM

and DCIS.COM/ER+ PR+ that lasted for at least 4 days post the single dose (Figure 3A). This

observation held true for MCF-7 and MDA-MB-231 cells (Figure 3B, Figure 3C), which also

responded to 10 µM of LicA, although the antiproliferative effect at this dose was not sustained

following a single dose. Next, we tested repeated dosing of LicA, every 48 h for the maximum of

3 doses. Interestingly, two treatments of MDA-MB-231 cells with LicA (10 µM) 48 h apart until

day 4 were able to sustainably retard proliferation until day 11 (Figure 3D) when we stopped our

experiment. We also saw a reduction of proliferation of MCF-7 cells after 3 doses of LicA (10

µM) 48 h apart which was sustained at least until day 12, while MCF-7aro cells, HCC1937, and

HCC3153 cells were also very responsive to LicA (5 µM) (Figures 3E - 3H). Our BrdU assay

confirmed an anti-proliferative rather than a cytotoxic effect and the use of a ferroptosis inhibitor,

ferrostatin, in combination with LicA showed that the antiproliferative effects are not due to

ferroptosis (data not shown).

Supporting evidence for antiproliferative effects was provided with the NanoString data and the

PISA proteomics results. Figure 4A demonstrates that the pro-proliferative genes such as *BCL2*

in MCF-7 cells as well as *RRM2* and *MKI67* in *MCF-7* and MDA-MB-231 cells were significantly

downregulated. Consistent with our metabolism flux data in breast microstructures which

showed reduced *de novo* nucleotide biosynthesis, we also observed significant downregulation

of *PRPS1* in MCF-7 cells (Figure 4A). Proteomics results (Figure 4B) further confirmed that in

these cells, LicA destabilized enzymes essential for proliferation such as several EIFs, and the

proliferation marker MKI67, and that the majority of destabilized proteins were associated with

cell cycle and proliferation (Figure 4C).

LicA downregulates SREBP1-dependent lipogenesis and inflammation in MCF-7 and

MDA-MB-231 cell lines.

The antiproliferative effects in ER+ and ER- BC cell lines (Figure 3) with LicA, along with the

metabolic effects detected in breast tissue microstructures (Figure 1) led us to test if the

antiproliferative effects observed in cell lines could be related to changes in lipogenic

metabolism and inflammation. Our NanoString metabolism panel results (Figure 4A) in MCF-7

and MDA-MB-231 cells were consistent with our observations in high-risk women's breast

microstructures (Figure 1) and clearly showed downregulation of SREBP1-dependent

lipogenesis genes such as *ACAT2*, *FASN*, and *SCD* and the pro-inflammatory *PTGES* (PGE2

synthase), in agreement with the significant upregulation of anti-inflammatory *HMOX1*.

PISA proteomics results were consistent with our NanoString data and confirmed a clear

separation of the LicA-treated and the DMSO-treated cells in both ER+ and ER- cell lines

(Figure S3A and S3B). Destabilization of key proteins is shown in Figure 4B; these include

lipogenic enzymes such as ACAT2, FASN, HMGCS1, and the inflammatory mediator NF-kB1

with a significant suppression of karyopherin β1 (KPNB1) which is needed for the translocation

of NF-kB to the nucleus before its transcriptional activities. LicA also destabilized NEDD8

(Figure 4B) which is important for the stabilization of SREBP1 through the post-translational

neddylation. On the other hand, the anti-inflammatory HMOX1 protein was profoundly stabilized

(Figure 4B).

Analysis of proteomics results with Enrichr further demonstrated significant destabilization of cell

cycle progression pathways (Figure 4C). Interestingly, while lipogenic metabolism was

significantly destabilized, the TCA cycle, branched amino acids degradation, and fatty acids

degradation appeared as the top stabilized pathways (Figure 4C) in both MDA-MB-231 and

MCF-7 cells. Consistent with these observations, western blots (Figure 4D) demonstrated that

SREBP1 expression, the phosphorylation of AKT at Ser 437, and the phosphorylation of PI3K

were suppressed in both MDA-MB-231 and MCF-7 cells treated with LicA. While the levels of

PI3K itself were reduced in MCF-7 cells, it had a slight but detectable increase in the MDA-MB-

231 cells. The reduction in AKT phosphorylation was not accompanied by a change in AKT

expression as a function of LicA treatment in these cell lines.

LicA suppresses local cholesterol levels in MCF-7 and MDA-MB-231 cells.

Transcriptomic data from the high-risk human breast microstructures and MCF-7 and MDA-MB-

231 cells treated with LicA, along with our proteomics observations showed significant

suppression of SREBP1, SREBP2, HMGCR with a shift in metabolism from lipogenesis to TCA

cycle, suggesting that cholesterol biosynthesis could be affected by LicA. Cellular free

(unesterified) cholesterol is mainly found in the plasma membrane in mammalian cells [72, 73].

We thus performed spatiotemporally resolved *in situ* quantification of IPM cholesterol in MDA-

MB-231 and MCF-7 cells before and after LicA treatment [68]. The results showed that the

spatially averaged IPM cholesterol levels were significantly different between the two BC cell

lines (Figure 4E). Importantly, when these cells were treated with LicA, their IPM cholesterol

levels were drastically reduced, i.e., 8-fold in MDA-MB-231 cells (n=14) and 4-fold in MCF-7

cells (n=19) (Figure 4E). These reduced IPM cholesterol levels were comparable to those

reported for unstimulated primary mammalian cells [67], demonstrating the efficacy of LicA in

IPM cholesterol depletion.

Licochalcone A retards luminal and triple negative xenograft tumors in mice.

Based on the antiproliferative effects of LicA in various breast cell lines (Figure 3) and the

downregulation of the proliferation-related genes and pathways (Figure 4), we evaluated the

effects of LicA in xenograft models of ER+ and ER- mammary cancers. In both xenograft

models, analysis of results included all the animals assigned to each experimental group. In

animals bearing MDA-MB-231 (ER-) xenografts, daily treatment with LicA (80 mg/kg.day, s.c.)

for 28 days significantly ($P < 0.000$) reduced the growth rate of tumors in 7/9 (22%) of animals

(Figure 5A). In the animals bearing MCF-7 xenograft tumors, treatment with LicA led to a

significant (P < 0.005) reduction in the rate of tumor growth in all 6/6 (100%) animals, with a

relatively similar responsiveness to treatment (Figure 5B).

Discussion

One of the hallmarks of cancer and an essential step in malignant transformation is metabolic

reprogramming and changes in cellular bioenergetics that facilitate the high energy demand of

hyperproliferating cells and promote their survival [39]. A main mechanism for this effect is the

reprogramming of lipid metabolic pathways [74-79]. Within this major pathway, the *SREBF1*

gene encodes SREBP1a and SREBP1c proteins which are transcription factors that regulate

lipogenesis. *SREBF1* expression is significantly higher in breast tumors compared to normal

tissue adjacent to tumor and in healthy breast tissue of 924 patients in the GTEx and TCGA

datasets [80-83]. Conversely, the downregulation of SREBP1 not only decreases lipogenesis,

but is also associated with bone protection [84-86]. It is known that SREBP1 regulates the

expression of lipogenic factors, fatty acid synthase (FASN), acetyl-CoA acetyltransferase 2

(ACAT2), acetyl CoA carboxylase 1 (ACC1), and ATP citrate lyase (ACLY) which are essential

for the survival of BC cells [25, 87, 88]. There is also a strong association between breast tissue

stiffness (a risk factor for breast carcinogenesis) and SREBP1 activity [26, 89, 90]. SREBP1

activates NF-kB-dependent inflammation by promoting fatty acid desaturation [91, 92]. Recent

studies have demonstrated that activation of NF-kB-dependent inflammation happens in part

through the translocation of the inactive SREBP1-NF-kB complex to the Golgi apparatus and

the subsequent activation of SREBP1 mediated by the SREBP cleavage activating protein

(Scap) [28] suggesting a close connection between SREBP1 metabolism and NF-kB

inflammation. An inflamed environment supports the growth of malignant cells through the

recruitment of macrophages, secretion of inflammatory cytokines, enhancement of fatty acid

oxidation, formation of reactive oxygen species (ROS), change of the hormonal milieu, and

increased proliferation [93]. Upstream of SREBP1, EGFR activation and PI3K-AKT-mTOR

mutation, as well as mTORC1 activity lead to enhanced SREBP1 expression and consequent

increase in SCD activity known to be important for cancer initiation, promotion, and invasion [94-

101].

We used breast microstructures from the opposite breasts of women with breast cancer, since

these represent normal/benign tissue that is known to be at increased risk of a second new

breast cancer. LicA significantly lowered *SREBF1*, as well as the expression of several of its

upstream effectors such as *EGFR*, *PI3K*, *AKT*, and *mTORC1*, in addition to *SCD* which is

downstream of *SREBF1*. These observations extended to MDA-MB-231 and MCF-7 cells,

where transcriptomics and proteomics data clearly demonstrate significant reduction of

lipogenesis signals regulated by SREBP1, in concert with a significant anti-inflammatory

response with suppression of PGE2 biosynthesis and a profound increase in HMOX1

expression. Interestingly, SREBP1 has response elements at the promoter region of *HMOX1*,

i.e. downregulation of SREBP1 is expected to lower *HMOX1* expression; however, in our

results, HMOX1 at gene and protein levels is profoundly enhanced. This could be associated

with the effect of LicA on the KEAP1-NRF2 pathway [42] which activates HMOX1 and

compensates for its likely reduction due to SREBP1 downregulation (Figure 6). In our results,

the significant decrease in the expression of SREBP1 is accompanied by reduced expression of

NEDD8 protein which is important for the stabilization of SREBP1 through posttranslational

neddylation, a process linked to BC aggressiveness [102]. In addition, lowered SREBP1

expression along with the reduced phosphorylation of its upstream effectors PI3K and AKT and

the profound drop in cellular cholesterol levels is aligned with the metabolic shift observed in our

proteomics studies. The combined effects of LicA on stabilizing TCA cycle, degradation of

branched chain amino acids, and fatty acids along with significant destabilization of the cell

cycle show that LicA reprograms cellular metabolism in favor of reducing cell proliferation [103,

104]. These results demonstrate that the axis of PI3K-AKT-SREBP1-NF-kB is the mechanism

by which LicA exerts its antiproliferative and tumor-suppressive effects. LicA's precise target in

this axis is yet to be elucidated.

Another important gene in lipid metabolism is the *SREBF2* encoding the SREBP2 protein which

is under the control of SREBP1 and primarily regulates cholesterol biosynthesis and

homeostasis [105]. It is known that enhanced expression of *SREBF2* promotes the progression

of several types of cancers [76, 106, 107]. *SREBF2* is significantly higher in breast tumors

compared to healthy tissues of 924 patients in GTEx and TCGA samples [80-83]. Among the

genes that are regulated by SREBP2 are members of the mevalonate pathway such as *MVD*,

MVK, *SQLN*, and *HMGCR*. The activity of *SREBF2* is opposed by the major tumor suppressor

P53 through transcriptional upregulation of the ABCA1 transporter and cholesterol efflux [105].

This effect is further reinforced by the activation of LXR and its subsequent regulation of

cholesterol homeostasis and ABCA1 mediated efflux, suppressing *de novo* biosynthesis, and

uptake of cholesterol. In our study in women's breast tissue in addition to lowered *SREBF2*

expression, the effect of LicA on the Mevalonate Pathway resembles the effects of p53 and LXR

activation. This once again shows that LicA reverses the oncogenic direction of the EGFR-PI3K-

AKT-SREBP axis in non-malignant tissue, and thus can protect breast cells by reprogramming

metabolism. It was recently reported that cholesterol ($[Chol]_l$) in the inner leaflet of the plasma

membrane (IPM) of mammalian cells is directly and quantitatively related to the proliferative

activity of cells [68]. The significant reduction in cholesterol biosynthesis in the inner leaflet of

plasma membrane which we depicted by measuring $[Chol]_l$ in MCF-7 and MDA-MB-231 cells

further validated the metabolic reprogramming. The difference we found in the $[Chol]_l$ between

MCF-7 and MDA-MB-231 cells reflects differential reprogramming of cholesterol homeostasis in

these two cell lines.

Interestingly, changes in the SREBP2 expression in women's breast microstructures and in

malignant breast cell lines MCF-7 and MDA-MB-231 differ. In women's breast microstructures

which are nonmalignant and composed of a mixture of various cell types in breast tissue

SREBP2 was profoundly downregulated when the tissue was exposed to LicA; however, in

malignant breast cell lines there was not a significant change in SREBP2 levels at the RNA and

protein levels, although IPM cholesterol was significantly suppressed. Thus, LicA reduces

lipogenesis in non-malignant high-risk breast through reducing the expression of both SREBP1

and SREBP2 while its effects on BC cells are primarily mediated by SREBP1. A recent study

showed that concerted SREBP1-NF-kB modulation might exert metabolic and inflammatory

responses independent of SREBP2 levels or activity, as it is executed through the recruitment of

SREBP1 and NF-kB to the Golgi apparatus mediated by and in complex with Scap [28]. In other

words, it is highly likely that the reduced proliferation in cancer cells are primarily regulated by

the SREBP1-NF-kB complex, through direct interaction with Scap. However, in non-malignant

high-risk breast tissue the additional reduction in SREBP2 levels, reinforces prevention of

cellular transformation. It is also likely that despite the tumor suppressive effects of LicA in both

MCF-7 and MDA-MB-231 xenografts, the precise LicA target(s) could be cell type-dependent,

since there were differences in SREBP1, PI3K, and AKT protein level changes in these cell

lines.

We found that metabolic flux in high-risk women's breast exposed to LicA was not only in favor

of generating excess NADPH to support antioxidant and anti-inflammatory pathways, flux was

also reprogrammed to minimize cell proliferation; the effect that we were able to confirm in a

series of experiments including proliferation assessments, *in vivo* tumor growth studies, and

molecular markers at gene and protein levels.

In addition to the SREBP1 dependent reduction in lipogenesis and inflammation that lead to

reduced cell proliferation, our data once again shows LicA limits oxidative and inflammatory

stress through the activity of NRF2 and suppressing NF-kB pathways in the breast tissue

microstructures of medium-risk women. Such stressors can cause DNA damage, remodel

extracellular matrix, promote cancer cell stemness, and modify paracrine effects to initiate and

promote tumors [55, 93, 108-111]. Moreover, the significant expression of SQSTM1 in the

breast microstructures, as well as in MCF-7 and MDA-MB-231 cells treated with LicA supports

its previously reported bone protective effects [52-54].

LicA-based interventions will effectively reduce BC risk (Figure 6). LicA protects against

oxidative and inflammatory stress, deprives cancer stem cells of their favored inflammatory

environment, while it lowers lipogenesis and proliferation. LicA has compliance with Lipinski's

rule of 5 (RO5) in drug discovery; it has 2 hydrogen bond donors (RO5 standard < 5), has 6

hydrogen bond acceptors (RO5 standard <10), exact molecular mass = 338.15180918 g/mol

(RO5 standard < 500 g/mol), and a calculated octanol-water partition coefficient = 4.9 (RO5

standard < 5). It should be noted that Lipinski's RO5 requires a drug-like compound to comply

with at least 3 out of 4 rules to remain orally active. Based on these physical properties and the

biological activities reported herein and previously, LicA is a good candidate for further

development as a novel BC risk reduction drug with sufficient efficacy, reduced toxicity, and

therefore anticipated to have greater acceptance by women at increased risk. Such an agent

with effects against both ER+ and ER- subtypes, and demonstrated bone protective properties

is predicted to have a better success and a larger impact than the current agents. We are now

planning the next series of studies including evaluations in immunocompetent models of cancer

prevention, and rigorous formulation and pharmacokinetic evaluations for oral delivery, that are

required for translation to clinical testing.

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Author's contributions: AH developed the hypothesis, designed the study, executed the in

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bioinformatics analyses. CHC and SC conducted the metabolism flux analyses and edited the

manuscript, XG, KB, WC conducted the spatiotemporal cholesterol measurements, analysis,

and edited the manuscript, OL assisted with the in vivo studies, RC analyzed in vivo data and

provided biostatistical support. SEC supervised the study design and interpretation of data,

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Figure Legends

Figure 1. Transcriptomics analysis in postmenopausal high-risk women's breast microstructures treated with LicA. Using RNA sequencing, LicA in the first set of 6 subjects' specimens **A)** downregulates cholesterol and lipid metabolism, **B)** upregulates antioxidant pathways, **C)** downregulates transcription factors associated with inflammation, proliferation, lipid and cholesterol metabolism, and upregulates transcription factors associated with antioxidant response: data derived from TRRUST_Transcription_Factors_2019 gene set, **D)** shares significant similarities in downregulated and upregulated transcriptomic signatures with statins: data derived from drug matrix geneset. **E) NanoString metabolism panel analysis of six additional postmenopausal high-risk women's breast microstructures**. LicA downregulates genes associated with lipid and cholesterol metabolism and estrogen carcinogenic metabolism and upregulates antioxidant response genes.

Figure 2. Genome-scale metabolic flux balance analysis in six postmenopausal high risk women's breast microstructures. A) Metabolic flux heat map identifies significant flux (p-value < 0.01) through antioxidant pathways, lipid and cholesterol biosynthesis and metabolism. Magnitude of flux difference between treatment and control are shown. **B)** Metabolic flux in the representative pathways is (B.1) towards the right direction and (B.2) towards the left direction to enhance NAD(P)H generation. **C)** Metabolic flux through pentose phosphate shunt is in the direction generating NAD(P)H without enhancing nucleotide biosynthesis. **D)** Nucleotides essential for bioenergetics and kinetics of the metabolic reactions are salvaged. **E)** Flux through fatty acid elongation drives NAD(P)H generation and formation of saturated fatty acids.

Figure 3. The effect of LicA on proliferation of ER+ and ER- premalignant and malignant breast cell lines. Repeated dosing of LicA exhibits sustained antiproliferative effects in aggressive cells. Red arrows show the time point(s) at which LicA or vehicle control was administered. Data represents mean ± SD of at least three independent replicates.

Figure 4. Changes in PI3K-AKT-SREBP1 lipogenesis in MCF-7 (ER+) and MDA-MB-231 (ER-) treated with LicA (10 µM, 24 h). **A)** NanoString metabolism pathway panel analysis shows the downregulation of lipogenesis, pro-inflammatory, and pro-proliferative genes and

upregulation of antioxidant and anti-inflammatory genes. Data represents the average of six analytical replicates of each cell line. **B)** PISA proteomics volcano plots demonstrate the proteins significantly stabilized or destabilized after the labeled extracted peptides were analyzed using LC-MS/MS. Data represents the average of six analytical replicates of each cell line and was filtered for the PSM numbers > 2 and the $p < 0.05$. **C)** Pathway identification using Enrichr based on the significantly stabilized and destabilized proteins observed in PISA output. **D)** Western blots: quantification of the bands was performed by Image J. Data represent mean ± SD of three independent replicates. **E)** Spatiotemporally resolved quantitative imaging of cholesterol in the inner leaflets of the plasma membranes (IPM). The IPM cholesterol concentrations before and after treatment with LicA (10 µM, 24h) were calculated from the two-channel cross-sectional ratiometric images of representative cells at steady-state. The z-axis scale indicates the cholesterol concentration (mol%). A pseudo-coloring scheme with red and blue representing the highest and the lowest concentration, respectively, is used to illustrate the spatial IPM cholesterol heterogeneity. Scale bars indicate 10 µm. A spatially averaged concentration (average ± SD from triplicate independent determinations with >10 cells per measurement) was calculated for each condition.

Figure 5. The effect of LicA on the rate of tumor growth in vivo. Xenografts of **A)** MDA-MB-231 (n =18 final tumor bearing animals) or **B)** MCF-7 cells (n = 12 final tumor bearing animals) were performed in the right (Rt) and left (Lt) flanks of ovary intact athymic nude mice. After the tumors reached 0.8 cm in diameter, subcutaneous administration of LicA (80 mg/kg.day) commenced and continued for 28 days. A significant reduction in the rate of tumor growth was observed in the LicA treated versus vehicle treated animals.

Figure 6. Suppressing lipogenesis and inflammation, mediated by the central role of SREBP1 regulation by LicA, leads to reduced proliferation in the breast. Colored arrows represent the effects of LicA.

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1 **Figure 1.**

2 **A)**

B)

WikiPathways 2024 Human

Pathway	
NRF2 Pathway WP2884 *2.08e-11	
Ferroptosis WP4313 *5.81e-07	
Amino Acid Metabolism In Triple Negative Breast Cancer Cells WP5213 *9.92e-07	
Photodynamic Therapy Induced NFE2L2 NRF2 Survival Signaling WP3612 *2.e-06	
Nuclear Receptors Meta Pathway WP2882 *1.06e-04	
Transcriptional Activation By NRF2 In Response To Phytochemicals WP3 *4.25e-04	
Pentose Phosphate Metabolism WP134 8.32e-04	
Glutathione Metabolism WP100 1.79e-03	
Oxidative Stress Response WP408 2.13e-03	
Benzo A Pyrene Metabolism WP696 2.66e-03	

$-\log_{10}(\text{p-value})$

C)

D)

Term Name	P-Adj
Hallmark_Cholesterol_Homeostasis	0.03482
Hallmark_MTORC1_Signaling	0.03482
Hallmark_Reactive_Oxygen_Species_Pathway	0.03482
Hallmark_Heme_Metabolism	0.03482
Hallmark_P53_Pathway	0.17471
Hallmark_Bile_Acid_Metabolism	0.17471
Hallmark_Estrogen_Response_Early	0.18615
Hallmark_Glycolysis	0.36128
Hallmark_Fatty_Acid_Metabolism	0.40635
Hallmark_Xenobiotic_Metabolism	0.44873

 Figure 2. A)

B)

B.1)

acald_c + h2o_c + nadp_c ⇌ ac_c + 2.0 h_c + nadph_c

g6p_c + nadp_c ⇌ 6pgl_c + h_c + nadph_c

6pgc_c + nadp_c ⇌ co2_c + nadph_c + ru5p__D_c

g6p_c + nadp_c ⇌ 6pgl_c + h_c + nadph_c

B.2)

adprib_c + h2o_c ⇌ amp_c + 2.0 h_c + r5p_c

h_r + nadph_r + estrone_r ⇌ nadp_r + estradiol_r

h_r + nadph_r + 1.5 o2_r + andrstrn_r ⇌ 2.0 h2o_r + nadp_r + eandrstrn_r

2.0 h_c + mlthf_c + nadph_c ⇌ 5mthf_c + nadp_c

adn_c + pi_c ⇌ ade_c + r1p_c

C)

D)

50   **E)**

Figure 3.

Figure 4.

A)

B)

C)

D)

111 **E)**

112   **Figure 5.**

113   **A)**

115   **B)**

119   **Figure 6.**

123