Influence of breast cancer risk factors and intramammary biotransformation on estrogen homeostasis in the human breast

Daniela Pemp1 · Leo N. Geppert2 · Claudia Wigmann2 · Carolin Kleider1 · René Hauptstein1 · Katja Schmalbach1 · Katja Ickstadt2 · Harald L. Esch1 · Leane Lehmann1

Received: 12 February 2020 / Accepted: 15 June 2020 / Published online: 22 June 2020 © The Author(s) 2020

Abstract
Understanding intramammary estrogen homeostasis constitutes the basis of understanding the role of lifestyle factors in breast cancer etiology. Thus, the aim of the present study was to identify variables influencing levels of the estrogens present in normal breast glandular and adipose tissues (GLT and ADT, i.e., 17β-estradiol, estrone, estrone-3-sulfate, and 2-methoxy-estrone) by multiple linear regression models. Explanatory variables (exVARs) considered were (a) levels of metabolic precursors as well as levels of transcripts encoding proteins involved in estrogen (biotrans)formation, (b) data on breast cancer risk factors (i.e., body mass index, BMI, intake of estrogen-active drugs, and smoking) collected by questionnaire, and (c) tissue characteristics (i.e., mass percentage of oil, oil%, and lobule type of the GLT). Levels of estrogens in GLT and ADT were influenced by both extramammary production (menopausal status, intake of estrogen-active drugs, and BMI) thus showing that variables known to affect levels of circulating estrogens influence estrogen levels in breast tissues as well for the first time. Moreover, intratissue (biotrans)formation (by aromatase, hydroxysteroid-17beta-dehydrogenase 2, and beta-glucuronidase) influenced intratissue estrogen levels, as well. Distinct differences were observed between the exVARs exhibiting significant influence on (a) levels of specific estrogens and (b) the same dependent variables in GLT and ADT. Since oil% and lobule type of GLT influenced levels of some estrogens, these variables may be included in tissue characterization to prevent sample bias. In conclusion, evidence for the intracrine activity of the human breast supports biotransformation-based strategies for breast cancer prevention. The susceptibility of estrogen homeostasis to systemic and tissue-specific modulation renders both beneficial and adverse effects of further variables associated with lifestyle and the environment possible.

Keywords Estrogens · Human breast · Multiple linear regression

Introduction
Breast cancer is the most common cancer in women worldwide. Its development is associated with increased levels of circulating estrogens, e.g., 17β-estradiol (E2), estrone (E1), and other endogenous steroids in pre- and postmenopausal women (Endogenous Hormones Breast Cancer Collaborative Group 2002, 2013) over a prolonged period of time. Based on these associations as well as an abundance of experiments in vitro and in animal models, the current understanding of the molecular etiology of breast cancer hypothesizes biotransformation of E2/E1 within the breast tissue to catechols and subsequent oxidation to mutagenic quinones possibly initiating tumor formation. Tumor promotion is then favored by estrogen receptor (ESR)-mediated stimulation of proliferation of the initiated cells (Yager and Davidson 2006). Thus, both tumor initiation and progression would depend predominately on intramammary levels of reactive products of estrogen biotransformation, whereas tumor promotion would depend predominately on levels of E2.
Consequently, commonly accepted risk factors such as early menarche and/or late menopause, late age at first pregnancy, small number of pregnancies, and short or no periods of breastfeeding (Colditz and Bohlke 2014) are supposed to increase the duration or extent of the local exposure of the mammary gland to E2 (biotransformation products) by increasing their systemic production. Concurrently, current risk reduction strategies involve the chemical modulation of ESR as well as systemic inhibition of aromatase or salpingo-oophorectomy (Advani and Moreno-Aspitia 2014) aimed to reduce levels of circulating estrogens. Recently, also modifiable risk factors associated with lifestyle such as (postmenopausal) obesity, alcohol consumption (Colditz and Bohlke 2014), smoking (Gaudet et al. 2017; Jones et al. 2017; Gram et al. 2019), and intake of estrogen-active drugs (EADs) for oral contraception (Grosse et al. 2009) or hormone replacement therapy (Collaborative Group on Hormonal Factors in Breast Cancer 2019) have been associated with both increased breast cancer risk and higher levels of circulating E2 and E1 (Endogenous Hormones Breast Cancer Collaborative Group 2003, 2011, 2013), suggesting that these risk factors also act by affecting intramammary levels of E2/E1 (biotransformation products).

Given the wide range of enzymes present in breast glandular (GLT) and adipose tissues (ADT; Pemp et al. 2019a), additional (biotrans)formation of estrogens within the breast tissue can reasonably be assumed (Labrie 2015; Mueller et al. 2015; Hilborn et al. 2017; Pemp et al. 2019a). Consequently, breast cancer risk factors may also influence tissue levels of E2 and its biotransformation products by affecting estrogen homeostasis in the breast. Thus, to better understand how estrogen homeostasis may affect initiation and promotion of breast cancer, insight into the influence of breast cancer risk factors on both levels of estrogen and estrogen biotransformation in women without breast cancer is needed.

However, only two studies in women without breast cancer have been published in this regard of which only one has performed statistical analyses (Online Resource 1, Savolainen-Peltonen et al. 2014). Even considering those analyzing non-tumor tissue of women with breast cancer and investigating the association of risk factors with tissue levels of estrogens, most studies did not provide information on parameters statistically compared and whether or not all positive/negative correlations were reported (Online Resource 1). Surprisingly, none of the available studies included reproductive history of the participating women in their statistical analyses, or collected data on smoking or analyzed biotransformation products of E2 other than E1 by recommended methods of specific analysis (Online Resource 1).

Only recently, we described quantitative estrogen profiles and transcript levels of enzymes involved in E2 (biotrans)formation in breast GLT and ADT of pre- and postmenopausal women without breast cancer (Pemp et al. 2019a; Fig. 1) providing suitable data to determine variables affecting intramammary levels of estrogens. Furthermore, it was shown that levels of most estrogens and ratios thereof as well as levels of transcripts encoding enzymes involved in their (biotrans)formation differed significantly between GLT and ADT (Pemp et al. 2019a), demonstrating that breast GLT and ADT should be considered separately. Thus, the aim of the present study was to identify variables (reproductive history, lifestyle, and transcript levels of enzymes involved in intracrine activity) influencing levels of estrogens and ratios thereof in breast GLT and ADT.

Materials and methods

Origin of biospecimens

Breast tissue specimens were obtained from 47 adult women without breast cancer undergoing reduction mammoplasty between 2010 and 2015. All women participating in the study gave their written informed consent prior to their inclusion in the study. Women with a personal and/or family history of breast cancer were not eligible for participation. Information on age, height, weight, parity (parous/nulliparous), smoking habits (never smoker, current smoker, current nonsmoker, and the latter two with daily cigarette consumption) was volunteered by 47 women, and information on the intake of EADs by 45 women. Body mass index (BMI) was calculated in kg/m².

Sample preparation and characterization

Biospecimens were prepared as described previously (Pemp et al. 2019a). Briefly, aliquots of apparently plain ADT and GLT with less than 15% adhering ADT were flash-frozen in liquid nitrogen and stored at − 80 °C. From mixed tissues, GLT was isolated from cryosections (40 µm) at maximum − 20 °C using a scalpel. Biospecimens were characterized by their mass percentages of oil (oil%), percentage of area covered by intra- and interstitial adipocytes, and lobule type: oil% in GLT and ADT were determined gravimetrically after extraction with chloroform. Percentage of area covered by intra- and interstitial adipocytes was estimated microscopically (Leica LMD6500) in cryosections (10 µm) of GLT stained with hematoxylin and eosin Y by two different persons and coded slides (Pemp et al. 2019a). The lobule type of GLT was determined microscopically according to Russo and Russo (2004) and Figueroa et al. (2014).
Instrumental analysis of E2, estrone, 2-methoxy-estrone, estrone sulfate, and glucuronide

E2, E1, and 2-methoxy(MeO)-E1 were determined by GC−MS/MS (Varian 450-GC, 300-MS; Bruker Daltonics, Bremen, Germany), whereas E1 sulfate (E1-S) and E1 glucuronide was determined by LC−MS/MS (QTrap® 5500; AB Sciex, Darmstadt, Germany). Tissue levels of E2, E1, 2-MeO-E1, and E1-S were quantified using their respective deuterated derivatives (Pemp et al. 2019a). Data used in statistical analyses are presented in Online Resource 2.

Determination of transcript levels

Analysis of transcript levels of genes encoding enzymes involved in E2 (biotrans)formation and regulation thereof was performed using customized Taqman® Low Density Arrays and Taqman® Gene Expression Assays as described by Pemp et al. (2019a). Data used in statistical analyses are presented in Online Resource 3.

Statistical methods

All statistical analyses were performed with the statistical programming language R [https://www.R-project.org/], version 3.5.2, and all tests of statistical significance were two-sided. Whenever multiple comparisons were performed, p values were adjusted using Holm’s method.

Contingency analyses and linear regression models

Contingency analyses were performed using Chi-square test. In case of categories following a natural order, Chi-square test for trend was used.
To test the association of every possible explanatory variable (exVAR) with the dependent variable, the variable explaining the dependent variable best is chosen by an automatic procedure. Subsequently, all possible exVARs are added one after another to the first one, ultimately choosing the one improving the model most, applying the Akaike information criterion. This is repeated until the model cannot be further improved by adding exVARs. Thus, each exVAR selected into the model contributes to modeling the dependent variable. The significance of the association is expressed by $p$ values and the magnitude of impact is expressed by coefficients of regression. The choice of exVARs is discussed in the results section and more detailed information is given in Online Resource 4. If levels of estrogens or transcripts were below limit of quantification (LOQ) in $> 40\%$ of samples (Online Resource 2 and 3), they were not included as exVARs.

For levels of transcripts or estrogens below the respective limit of detection (LOD) and below the respective LOQ, LOD and LOQ were set, respectively. When levels of transcripts or estrogens were $< \text{LOQ}$ in $> 1$ sample and $\le 40\%$ of samples, the levels of the respective transcript/estrogen were included additionally to the continuous exVAR as the qualitative presence of the exVAR (binary exVARq, compared to levels $< \text{LOQ}$). If, in the computed model, observations with Cook’s Distance $> 1$ appeared, they were removed and the model was computed anew. This process was repeated until no conspicuous observations occurred. To achieve normal distribution, dependent variables were logarithmized. Data distributions were evaluated in Quantile–Quantile plots with simulated confidence bands. Constant standard deviations of the errors were evaluated using scale-location plots. To check the model assumption of independent identically distributed errors, the residual vs. fitted values plot was used.

The adjusted coefficients of determination, the numbers of conspicuous observations removed, the numbers of observations contributing to the final models (maximum of 45 because of two specimens without information on the intake of EADs). In models considering intracrine activity, maximum number of observations was further reduced because of two and one specimens in GLT and ADT, respectively, without information on transcript levels), and the ratio of observations per exVAR of each final model is given. To achieve accurate estimation of regression coefficients, at least two observations per exVAR (Austin and Steyerberg 2015) were aimed for.

In addition, the regression coefficients (which represent the mean changes in the dependent variables for one unit of change in the respective exVAR while holding other predictors in the models constant), their confidence interval, as well as the $p$ values of each exVAR selected are given in Online Resource 4.

**Analyses of independence of variables**

Spearman’s rank correlation analysis was performed to identify collinearity between numerical exVARs which might hinder each other selection and/or influence each other $p$ values within the models. In the case of variables with $> 1$ level below LOD or LOQ, correlation was calculated with randomly distributed ranks for ties, 10,000 times and highest Spearman correlation coefficients and lowest $p$ values were used to rather overestimate collinearity. Relationship between categorical and numerical exVARs was evaluated by comparison of medians using unpaired Wilcoxon tests. Indications for relationships between variables and possible consequences for the selection of exVARs are given for each model in Online Resource 4.

**Results and discussion**

To identify exVARs (e.g., breast cancer risk factors) influencing dependent variables (e.g., tissue levels of E2) by multiple linear regression models, suitable dependent variables were identified first and potential exVARs were chosen subsequently. Then, multiple linear regression models using stepwise forward selection were applied to assess up to 32 exVARs possibly influencing tissue levels of estrogens and ratios thereof in GLT and ADT.

**Identification of dependent variables**

E2, E1, and E1-S were detected in most biospecimens, whereas 2-MeO-E1 was detected predominately in ADT (Online Resource 2). Thus, the influence of exVARs on levels of E2, E1, E1-S, and ratios thereof was analyzed in both GLT and ADT whereas the influence of exVARs on levels of 2-MeO-E1 could be analyzed in ADT only. Furthermore, continuous variables significantly influencing estrogen levels were further analyzed as dependent variables as well.

**Identification of possible exVARs related to the study cohort and to the tissues collected**

First, the study cohort and the tissues collected were characterized and possible exVARs were defined.

**Age, reproductive history, and related variables**

The age of the 47 women participating in the study ranged from 18–66 years. Most tissues were derived from women in the age group of 40–49 years (Online Resource 5).
Menopausal status of the study population was allocated according to the range of age at menopause (46–52 years) observed in German women participating in the EPIC study \((n > 27,000;\) Tsilidis et al. \(2011\)) instead of assessment based on the women’s menstrual cycle characteristics (information not available). Thus, with a high probability, women > 52 and < 46 years old can be assumed to be post- (19\%, Fig. 2) and premenopausal (55\%), respectively. Women between 46–52 years were grouped as perimenopausal (26\%). However, this group is likely to contain pre-, peri-, and postmenopausal women. To reflect continuous influence of age as well as abrupt influence of menopause on dependent variables, both the potential exVARs age and menopausal status were included into the models.

Regarding reproductive history, 40\% of the women participating in the study were nulliparous. In the age class 35–54 years old, 27\% were nulliparous, compared to 21\% in the respective general population (Online Resource 5). The study population was further classified regarding parity and lobule type of the GLT, the latter reflecting age- and parity-related histological changes within the breast. These changes are most obvious in parous women, where lobules type 2 and 3 (Lob2/3), previously induced during pregnancy, regress back to lobule type 1 (Lob1p; Russo and Russo \(2004\)). Nulliparous women (exhibiting lobule type 1, Lob1np) represented the third group of lobule types. 23\%, 36\%, and 40\% of GLT were classified Lob2/3, Lob1p, and Lob1np, respectively (Fig. 2). In the following linear regression models, lobule type was tested as categorical exVAR. Relationship of lobule type with menopausal status and age cannot be completely excluded, but seems to be unlikely (Fig. 2).

**Obesity and related variables**

According to the WHO BMI classification, 53\% and 13\% of women were pre-obese and obese, respectively. The remaining women (34\%) were of normal weight (Online Resource 6). Thus, compared to the German adult female population (29\% pre-obese, 24\% obese, and 45\% of normal weight; Mensink et al. \(2013\)), a higher percentage was pre-obese, but lower percentages were obese and of normal weight. BMI was included as continuous exVAR into the linear regression models.

Median oil% in GLT and ADT were 16\% and 85\%, respectively (Pemp et al. \(2019a\)). In the following linear regression models, oil% were included as continuous exVAR. Directly isolated GLT and GLT isolated from mixed tissue were compared by means of two variables, oil% and relative areas covered by inter- and intrastromal adipocytes in GLT, and \(p\) values were adjusted for two comparisons. Despite detaching adhering ADT, GLT isolated from mixed tissues \((n = 18)\) still exhibited significantly higher oil% (Fig. 3) than GLT which could be isolated directly; indicating a higher number and/or size of adipocytes within intra- and interlobular stroma. Consistently, microscopic comparison of GLT isolated directly and GLT isolated from mixed tissues revealed a significantly larger relative area covered by inter- and intrastromal adipocytes in GLT derived from mixed tissues (Fig. 3). Thus, in the following sections, these specimens will be referred to as large-adipocyte-area (laa)GLT and small-adipocyte-area GLT. The occurrence of laaGLT was statistically independent of BMI classification of the women donating the tissues \((p = 0.40,\) Chi-square test for trend) and lobule types of GLT specimens \((p = 0.28,\) Chi-square test, Online Resource 7). The occurrence of laaGLT was tested as binary exVAR.

**Smoking and intake of EADs**

Twelve women (26\%) declared being current smokers (2–25 cigarettes/day) and EADs were used by eight women. These information were included as categorial exVARs into the linear regression models.

Mosaic plots characterizing the study population used in the linear regression models regarding the exVARs menopausal...
status, lobule type, BMI, intake of estrogen-active drugs, and smoking are depicted in Online Resource 8.

**Identification of possible exVARs related to estrogen biotransformation in tissues**

Further possible continuous exVARs were tissue levels of the direct precursor estrogen(s) and of transcripts encoding enzymes directly forming or further metabolizing the dependent variable according to Fig. 1. In models with ratios of levels of different estrogens as dependent variables, further exVAR considered were levels of transcripts encoding enzymes directly forming or further metabolizing at least one of the estrogens involved in the ratio. If levels of transcripts or precursor estrogens were <LOQ in > 1 sample and ≤40% of samples, the qualitative presence of the respective precursor estrogen or transcript was included as binary exVARq, as well.

**Variables influencing tissue levels of estrogens identified by multiple linear regression models**

Previous studies investigating variables associated with levels of estrogens in breast tissues either performed no statistical analysis at all or univariate analysis (i.e., comparisons of medians in case of categorical variables and correlation analyses in the case of continuous variables, Online Resource 1). In addition, methods nowadays less recommended for biospecimen analysis (Labrie et al. 2015) were applied to determine estrogen levels and/or undefined specimens were used without specifying the presence of GLT or ADT. Because of these differences, the outcome of the present study is only compared with previous ones if at least either a specific method or specifically GLT or ADT was used. Moreover, results observed in previous studies using tissues derived from women with breast cancer or from both women with and without breast cancer together for statistical analyses were included in Online Resource 1, but are not discussed in the following sections.

**Age and menopause**

Cessation of ovarian estrogen production in menopause decreases blood levels of estrogens (Endogenous Hormones Breast Cancer Collaborative Group 2011, 2013) and is thus considered to affect levels of estrogens in breast tissues. Depypere et al. (2015) observed lower median levels of E2 in GLT derived from postmenopausal than from premenopausal women, yet no statistical analysis was performed. In the present study, levels of E2 were not directly influenced by postmenopausal status. However, postmenopausal status influenced levels of E1 in GLT and levels of E1-S in ADT negatively \( (p < 0.05) \). Furthermore, the ratio of E2 levels in ADT to E2 levels in GLT (ADT/GLT of E2) was also influenced negatively by postmenopausal status \( (p < 0.05) \). Interestingly, ADT/GLT of E1 was rather positively influenced by postmenopausal status \( (0.10 > p \geq 0.05, \text{Fig. 4}) \). Thus, menopause seems not to affect levels of all estrogens in GLT and ADT and ratios ADT/GLT of estrogens in the same way which cannot be explained by a mere decrease in systemic delivery of estrogens via plasma. The continuous
Age did not directly influence levels of any estrogen or ratio thereof significantly (Fig. 4).

**Lobule type**

The impact of lobule type on estrogen levels has not been investigated yet. Average levels of E1-S in GLT exhibiting the least developed lobule type 1 (derived from nulliparous women, Lob1np, or from parous women after age-related regression, Lob1p) were lower than levels of E1-S in GLT categorized Lob2/3, yet not significantly ($0.10 > p \geq 0.05$ and $p > 0.1$, respectively; Fig. 4). Furthermore, lobule type 1 influenced levels of E2 and E2/E1 in ADT negatively compared to Lob2/3 ($p < 0.05$, Fig. 4).

**Oil%**

Median oil% in GLT and ADT were 16% and 85%, respectively (Pemp et al. 2019a). Estrogen levels might be affected by oil% via: (i) physicochemical distribution of the lipophilic molecules E2 and E1 and the more hydrophilic E1-S and (ii) cell-specific enzyme expression in stromal adipocytes. Oil% influenced levels of E1 in GLT (positively, $p < 0.05$), levels of E1-S in GLT (negatively, $p < 0.05$), and E1-S/E1 in both GLT and ADT (negatively, $p < 0.05$; Fig. 4). Interestingly, levels of 2-MeO-E1 and E2 (which are more and comparably lipophilic than E1) were not influenced by oil% in ADT (Fig. 4), thereby rendering a mere physicochemical effect less likely. In line with the influence of oil% on the respective estrogens, ADT/GLT of both E2 and E1 were positively influenced by oil% in ADT but negatively in GLT ($p < 0.05$).

Since oil% influenced levels of some estrogens and ratios thereof in ADT and GLT, variables affecting oil% were investigated as well. Besides age, lifestyle factors, and tissue characteristics, oil% may be influenced by estrogens affecting lipogenesis and adipogenesis (Gao and Dahlman-Wright 2013), and thus, levels of E2 were also considered as exVAR. Furthermore, the classification of GLT regarding adipocyte area laaGLT was considered as exVAR as well. Oil% in ADT were positively influenced by laaGLT ($p < 0.05$), but no other exVAR was selected (Fig. 4). Interestingly, oil% in ADT were positively influenced by levels of E2, age, and specimens derived from nulliparous women (compared to parous women prior to

---

**Fig. 4** Influence of various exVARs on levels of estrogens as well as on ratios thereof and oil% in GLT and ADT (dependent variables) identified by multiple linear regression models using stepwise forward selection as detailed in Online Resource 4. For each model, the adjusted coefficient of determination ($R^2$), and ratio of the number of observations (i.e., biospecimens) to exVAR (O/exVAR) after forward selection of variables is given.
age-related regression, all $p < 0.05$), and not significantly, by BMI ($p > 0.1$, with no apparent statistical reason such as collinearity interfering with the exVAR BMI, Online Resource 4).

Notably, both linear regression models exhibited low $R^2$ values (Fig. 4) suggesting either important variables missing in the model (which seems likely in this case) or large variations within the data set.

Lifestyle-associated variables

A positive association between BMI and estrogens in serum has been observed in pre- (E2, E1; Endogenous Hormones Breast Cancer Collaborative Group 2013) and postmenopausal women (E2, E1, E1-S; Endogenous Hormones Breast Cancer Collaborative Group 2003, 2011). The common interpretation is that an increase in BMI leads to an increase in the mass of adipose tissue within the whole body, accompanied by a change in intra- and extramammary function of adipose tissue (Yaghjyan and Colditz 2011; Brown 2014). Consequently, a higher amount of estrogens is produced and distributed within the body via blood (Lønning et al. 2011), contributing to estrogen levels in breast tissue. Yet, no studies investigating associations between BMI and estrogen levels in breast tissues derived from women without breast cancer have been identified. In the present study, BMI influenced levels of E2 in GLT (0.10 $> p \geq 0.05$) and ADT ($p < 0.05$) positively. In contrast, tissue levels of E1-S were not influenced by BMI (Fig. 4).

Besides BMI, smoking has also been associated with higher levels of E2 and E1 in blood of postmenopausal women (Endogenous Hormones Breast Cancer Collaborative Group 2011) but not of premenopausal women (Endogenous Hormones Breast Cancer Collaborative Group 2013). No studies investigating the impact of smoking on levels of estrogens in breast tissues have been identified. In the present study, smoking influenced levels of E1 in ADT positively and, congruently, E2/E1 in ADT negatively (both $p < 0.05$, Fig. 4).

In all studies analyzing estrogens in breast tissues of women without breast cancer, the intake of EADs for oral contraception or hormone replacement therapy was either an exclusion criterion or not considered in statistical analyses (Online Resource 1). In the present study, the intake of ethinyl estradiol did not influence the levels of E2 in GLT or ADT (Fig. 4). Yet, intake of ethinyl estradiol influenced ADT/GLT of E2, levels of E1 in GLT ($p < 0.05$), and E1-S/E1 in ADT negatively ($0.10 > p \geq 0.05$). In contrast, intake of E2-releasing drugs (containing E2 or E2 valerate) used for hormone replacement therapy positively influenced levels of E1-S and E1-S/E1 in ADT ($p < 0.05$), but did not influence levels of E2 or E1 in GLT or ADT.

Biotransformation precursors

No studies investigating correlations among estrogens in breast tissues derived from women without breast cancer have been identified. In the present study, biotransformation precursors of the respective estrogens influenced tissue levels of E2, E1, E1-S, and 2-MeO-E1 positively ($p < 0.05$, except for E1-S as precursor for E1 in ADT $0.10 > p \geq 0.05$; Fig. 4). Thus, exVARs affecting levels of, e.g., E1, may indirectly contribute to levels of E2, E1-S, and 2-MeO-E1. Interestingly, whereas levels of E1 in GLT were positively influenced by levels of E1-S, levels of E1 in ADT were positively influenced by the qualitative presence of both E1-G ($p < 0.05$) and E1-S ($0.10 > p \geq 0.05$).

Transcripts encoding enzymes involved in estrogen (biotrans)formation

Savolainen-Peltonen et al. (2014) observed no correlation of levels of $CYP19A1$, i.e., aromatase, $STS$, and $HSD17B1$, with levels of E2 in ADT and the respective transcript levels did also not significantly influence levels of E2 in ADT in the present study. However, in the present study, levels of E2 in ADT were influenced positively by levels of the transcript encoding GUSB (Fig. 4, $p < 0.05$), the enzyme hydrolizing estrogen glucuronides (Fig. 1). Estrogen glucuronides are mostly associated with elimination from tissues and body but may also contribute to intratissue levels of estrogens, even though GUSB and substrates only meet in a highly controlled manner (Naz et al. 2013). Furthermore, levels of E2 in ADT were influenced negatively ($p < 0.05$) by levels of transcripts encoding HSD17B2, the enzyme forming E1 by oxidation of E2 (Fig. 1). Congruently, levels of E1 in ADT were positively influenced by levels of $HSD17B2$. Moreover, levels of $CYP19A1$ influenced levels of E1 in ADT positively (Fig. 4, $p < 0.05$). The presence of transcripts encoding the conjugating enzyme UGT1A9 and levels of transcripts encoding $CYP1A1$ influenced the levels of the respective substrates (i.e., E1, 2-MeO-E1, and E2, respectively) in ADT negatively ($0.10 > p \geq 0.05$). Likewise, levels of $SULT1A3/4$ influenced E1-S in GLT positively ($0.10 > p \geq 0.05$). Forward selection of exVARs into models describing estrogen levels in ADT and GLT identified levels of further transcripts, yet the associations exhibited $p$ values $\geq 0.10$ (Fig. 4).

Regulation of transcription of genes encoding enzymes by the respective substrates (up-) and products (down-), is a common biochemical feedback mechanism. Furthermore, ligand-activated transcription factors are involved in the regulation of transcription of genes of biotransforming enzyme families; e.g., CYP (Tralau and Luch 2013), SULT (Runge-Morris et al. 2013), and UGT (Hu et al. 2014). Positive associations of levels of $STS$ as well as presence of $UGT1A9$ and $UGT1A10$ were observed with the respective substrates of
the encoded enzymes; i.e., E1-S in ADT, E1 in GLT and E1 in ADT, respectively ($p < 0.05$; Fig. 4). UGT1A9 (Cho et al. 2016) and UGT1A10 (Starlard-Davenport et al. 2008) may be regulated by activated ESR1, whereas STS may be regulated via G-protein-coupled ESR action (Gilligan et al. 2017).

Since levels of CYP19A1, GUSB, and HSD17B2 influenced levels of estrogens in ADT significantly, exVARS influencing these variables were investigated, as well. Besides exVARS associated with physiology and lifestyle, levels of transcripts known to be directly or indirectly involved in regulation of phase I and phase II biotransformation were included, since little is known about specific regulation of the transcription of the genes encoding these enzymes (Naz et al. 2013; Zhao et al. 2016; Hilborn et al. 2017).

Levels of CYP19A1 in ADT were exclusively influenced by the intake of ethinyl estradiol (negatively, Fig. 4, $p < 0.05$). It is known that the ovarian synthesis of estrogens is negatively regulated by estrogen-active compounds (Fleischman et al. 2010). Furthermore, CYP19A1 was detected less frequently in the endometrium of women taking oral contraceptives containing ethinyl estradiol than in non-users (Maia et al. 2013). Low $R^2$ value of the model (Fig. 4) suggests at least one other important variable missing in the model (e.g., transcript levels of glucocorticoid receptor; Zhao et al. 2016).

Levels of GUSB in ADT were significantly influenced by levels of AHR, ESR2 (positively), and qualitative presence of NR1I3 (negatively). Binding sites for transcription factors such as Sp1 and AP-2 (Naz et al. 2013) in the promoter of the GUSB gene provide a link to estrogen signaling (Pellikainen and Kosma 2007; Safe and Kim 2008). BMI influenced levels of GUSB positively ($0.10 > p \geq 0.05$). Most interestingly, levels of HSD17B2 in ADT were significantly influenced by BMI, smoking, and postmenopausal status as well as by levels of ARNT (negatively) and levels of ESR2 and N112 (positively). Transcription of the HSD17B2 gene is regulated by retinoic acid via RAR alpha/RXR alpha tethered to transcription factors Sp1 and Sp3 on the HSD17B2 promoter (Cheng et al. 2008), which provides a link to estrogen-mediated signaling. An obvious link between ARNT or N112 and expression of HSD17B2 has not been described yet.

Concluding this section, it should be emphasized that lack of influence of transcript levels does not exclude the contribution of the respective enzyme activities to estrogen levels.

Relevance

Including E2, E1, E1-S, and 2-MeO-E1, the present study encompasses the major estrogens detectable in breast tissues derived from women without breast cancer. In contrast to all previous studies, the present study fulfilled published prerequisites (Yaghjyan and Colditz 2011; Sherman et al. 2012; Rosner et al. 2013) for data acquisition concerning sample characterization as well as specificity and reliability of estrogen analysis (Pemp et al. 2019a).

However, mammoplasty specimens raise concern regarding sample bias, in particular (i) “young age”, (ii) “obesity”, and (iii) “extremely large fatty breasts” (Sherman et al. 2012), thus putatively reflecting a specific subpopulation. Most specimens were derived from women 40–49, which is also the modal age group of the respective general population (Online Resource 5). Yet, the study population indeed lacked women older than 66 years and the proportion of pre-obese and obese women was higher and lower than in the general population, respectively (Online Resource 6). Moreover, 38% of specimens were classified as laaGLT and exhibited higher oil% than GLT which could be isolated directly (section “Obesity and related variables”). Since no data on the occurrence of laaGLT in the general female population are available, it is currently unknown whether women undergoing mammoplasty are predisposed to laaGLT. Oil% significantly influenced levels of E1, E1-S in GLT, E1-S/E1 in GLT and ADT, as well as ADT/GLT of E1 and E2 and should, therefore, be considered in sample characterization of human breast biospecimens. Of note, oil% in GLT were not influenced by any exVAR deducible by questionnaire.

Intake of exogenous estrogens, menopausal status, BMI, and smoking were previously shown to be associated with levels of estrogens in blood (Endogenous Hormones Breast Cancer Collaborative Group 2003,2011; Fleischman et al. 2010, 2013; Gaudet et al. 2017). The general consensus is that blood levels of estrogens contribute to breast tissue levels, yet whether levels of circulating estrogens serve as surrogate for levels of estrogens in breast tissues, or more precisely, in GLT or ADT, is a matter of debate (Lønning et al. 2011; Bulun et al. 2012; Colditz and Bohlke 2014; Labrie 2015; Stanczyk et al. 2015; Vihma et al. 2016).

In the present study, variables known to affect levels of circulating estrogens indeed influenced estrogen levels in breast tissues, as well (Fig. 5).

However, these exVARS affecting levels of circulating estrogens by targeting ovarian and adiposal production of estrogens did obviously not suffice to explain the variances in estrogen levels within the breast. In addition, estrogen levels in both GLT and ADT were further influenced by levels or the presence of their precursor estrogens and levels of transcripts encoding enzymes involved in estrogen biotransformation. Interestingly, whereas (as expected, Mueller et al. 2015) E1-S seems to represent a source of E1 in breast GLT, intratissue levels of E1 in ADT seem additional to be regulated by glucuronidation via UGT1A9 (Fig. 5). Furthermore, although E2-3-G was not detected in human breast tissues, in the light of the contribution of levels of GUSB to
levels of E2 in ADT and a ten times higher LOD for E2-3-G compared to E1-G (Pemp et al. 2019a), E2 glucuronides cannot be excluded to represent a source of estrogens for the human breast, as well.

Further supporting the role of intratissue biotransformation, levels of E2 and of E1 in ADT were also significantly influenced by levels of HSD17B2. Recently, it was shown that levels of Hsd17b enzymes significantly influenced intramammary levels of E2 in the ACI rat model of tumorigenesis (Pemp et al 2019b), as well, supporting the important role of HSD17Bs in intramammary estrogen homeostasis (Hilborn et al. 2017) across species.

Furthermore, the present study not only supports the commonly accepted importance of CYP19A1 in estrogen homeostasis, but suggests its role within the breast tissue in addition to a systemic effect. Avoiding indirect associations in linear regression models, the influence of exVARs present in ADT on levels of E2 in GLT was not tested in the present study. However, due to correlations between estrogen levels in GLT and ADT (Pemp et al. 2019a) and the influence of precursor estrogens on the respective estrogen levels shown in the present study, levels of E2 in GLT might be affected indirectly by any variable significantly influencing E1 in ADT, i.e., levels of CYP19A1, HSD17B2, presence of UGT1A9, E1-S, and E1-G, as well as BMI and smoking (Fig. 5).

Inhibition of CYP19A1 suggested in breast cancer prevention (Advani and Moreno-Aspitia 2014; Colditz and Bohlke 2014) could thus lower intratissue E2 levels not only by systemic but also by intracrine mechanisms. However, systemic alteration of E2 biosynthesis harbors the risk of losing its beneficial biological effects, e.g., in bone health (Advani and Moreno-Aspitia 2014). Thus, drugs in development for the treatment of endocrine disorders by targeting enzymes involved in more organ-specific E2 homeostasis (e.g., inhibitors of HSD17B2 and STS; Konings et al. 2018) may also be successful in breast cancer chemoprevention.

In conclusion, a thorough characterization of specimens enabled taking into account variables related to obesity and “extremely fatty breasts” during statistical analyses. Tissue characterization of GLT derived from mammoplasty (and possibly also of biopsy) specimens by oil% as well as by lobule type seems to be advisable to prevent sample bias.

Novel insights in estrogen homeostasis in the normal human breast GLT and ADT support contribution of variables affecting both extra- and intratissue (biotrans)formation of estrogens.
of estrogens and suggest a central role of E1 levels in breast ADT homeostasis. The susceptibility of estrogen homeostasis to systemic and tissue-specific modulation renders both beneficial and adverse effects of further variables associated with lifestyle and the environment possible.

**Acknowledgements** Open Access funding provided by Projekt DEAL. The authors would like to thank Prof. Peter Eckert, Prof. Rafael Jakubietz, Dr. Iva Neshkova, and Dr. Ulrike Waldhofen (mammoplasty specimens).

**Funding** This study was funded by the German Research Foundation with a grant to L. Lehmann (DFG LE 1329/10–1) and is part of the joint research project, “Isoflavones: Cross-species comparison of metabolism, estrogen sensitivity, epigenetics and carcinogenesis”.

**Availability of data and material** The datasets generated during and/or analyzed during this study are included in this published article [and its supplementary information files] or are available from the corresponding author on reasonable request.

**Compliance with ethical standards**

**Conflict of interest** The authors declare that they have no conflict of interest.

**Ethics approval** This study was performed in line with the principles of the Declaration of Helsinki and its later amendments. Approval was obtained from the Ethics Committee of the University of Würzburg, Germany (reference numbers 74/10 and 168/10).

**Consent to participate** Informed consent was obtained from all individual participants included in the study.

**Consent for publication** All individual participants consented to publish their anonymized data.

**Open Access** This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article’s Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article’s Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit http://creativecommons.org/licenses/by/4.0/.

**References**

Advani P, Moreno-Aspitia A (2014) Current strategies for the prevention of breast cancer. Breast cancer 6:59–71. https://doi.org/10.2147/BCTTT.S39114

Austin PC, Steyerberg EW (2015) The number of subjects per variable required in linear regression analyses. J Clin Epidemiol 68:627–636. https://doi.org/10.1016/j.jclinepi.2014.12.014

Brown KA (2014) Impact of obesity on mammary gland inflammation and local estrogen production. J Mammary Gland Neoplasia 19:183–189. https://doi.org/10.1007/s10911-014-9321-0

Bulun SE, Chen D, Moy I, Brooks DC, Zhao H (2012) Aromatase, breast cancer and obesity: a complex interaction. Trends Endocrinol Metab 23:83–89. https://doi.org/10.1016/j.tem.2011.10.003

Cheng YH, Yin P, Xue Q, Yilmaz B, Dawson MI, Bulun SE (2008) Retinoic acid (RA) regulates 17beta-hydroxysteroid dehydrogenase type 2 expression in endometrium: interaction of RA receptors with specificity protein (SP) 1/SP3 for estradiol metabolism. J Clin Endocrinol Metab 93:1915–1923. https://doi.org/10.1210/je.2007-1536

Cho SJ, Ning M, Zhang Y, Rubin LH, Jeong H (2016) 17beta-Estradiol up-regulates Udp-glucuronosyltransferase 1a9 expression via estrogen receptor alpha. Acta Pharm Sin B 6:504–509. https://doi.org/10.1016/j.apsb.2016.04.005

Colditz GA, Bohlke K (2014) Priorities for the primary prevention of breast cancer. CA Cancer J Clin 64:186–194. https://doi.org/10.3322/caac.21225

Collaborative Group on Hormonal Factors in Breast Cancer (2019) Type and timing of menopausal hormone therapy and breast cancer risk: individual participant meta-analysis of the worldwide epidemiological evidence. Lancet. https://doi.org/10.1016/S0140-6736(19)31709-X

Deypere HT, Bolca S, Bracke M, Delanghe J, Comhaire F, Blondeel P (2015) The serum estradiol concentration is the main determinant of the estradiol concentration in normal breast tissue. Maturitas 81:42–45. https://doi.org/10.1016/j.maturitas.2015.01.014

Endogenous Hormones Breast Cancer Collaborative Group (2002) Endogenous sex hormones and breast cancer in postmenopausal women: reanalysis of nine prospective studies. J Natl Cancer Inst 94:606–616

Endogenous Hormones Breast Cancer Collaborative Group (2003) Body mass index, serum sex hormones, and breast cancer risk in postmenopausal women. J Natl Cancer Inst 95:1218–1226

Endogenous Hormones Breast Cancer Collaborative Group (2011) Circulating sex hormones and breast cancer risk factors in postmenopausal women: reanalysis of 13 studies. Br J Cancer 105:709–722. https://doi.org/10.1038/bjc.2011.254

Endogenous Hormones Breast Cancer Collaborative Group (2013) Sex hormones and risk of breast cancer in premenopausal women: a collaborative reanalysis of individual participant data from seven prospective studies. Lancet Oncol 14:1009–1019. https://doi.org/10.1016/S1470-2045(13)70301-2

Figueroa JD, Pfeiffer RM, Patel DA, Linville L, Brinton LA, Gierach GL, Yang XR, Papathomas D, Visscher D, Mies C, Degnim AC, Anderson WF, Hewitt S, Khodor ZG, Clare SE, Storniolo AM, Sherman ME (2014) Terminal duct lobular unit involution of the normal breast: implications for breast cancer etiology. J Natl Cancer Inst 106:286. https://doi.org/10.1093/jnci/dju286

Fleischman DS, Navarrete CD, Fessler DM (2010) Oral contraceptives suppress ovarian hormone production. Psychol Sci 21:750–752. https://doi.org/10.1177/0956797610368062

Fleming JM, Miller TC, Quinones M et al (2010) The normal breast microenvironment of premenopausal women differentially influences the behavior of breast cancer cells in vitro and in vivo. BMC Med 8:27. https://doi.org/10.1186/1741-7015-8-27

Gao H, Dahlman-Wright K (2013) Implications of estrogen receptor alpha and estrogen receptor beta for adipose tissue functions and cardiometabolic complications. Horm Mol Biol Clin Invest 15:81–90. https://doi.org/10.1515/hmbci-2013-0021

Gaudet MM, Carter BD, Brinton LA, Falk R, Gram JT, Luo J, Milne RL, Nyante SJ, Weiderrass E, Beane Freeman LE, Sandler DP, Robien K, Anderson KE, Giles GG, Chen WY, Feskanich D, Braaten T, Isaacs C, Butler LM, Koh WP, Wolk A, Adami HO, White E, Margolis KL, Thun MJ, Gaziano SM (2017) Pooled
analysis of active cigarette smoking and invasive breast cancer risk in 14 cohort studies. Int J Epidemiol 46:881–893. https://doi.org/10.1093/ije/dyw288

Gilligan LC, Gondal A, Tang V, Hussain MT, Arvanitis A, Hewitt AM, Foster PA (2017) Estrone sulfate transport and sulfate maturation activity in colorectal cancer: implications for hormone replacement therapy. Front Pharmacol. https://doi.org/10.3389/fphar.2017.00103

Gram IT, Park S-Y, Maskarinec G, Wilkens LR, Haiman CA, Le Grosse Y, Baan R, Straif K, Secretan B, El-Ghissassi F, Bouvard V, Gilligan LC, Gondal A, Tang V, Hussain MT, Arvanitis A, Hewitt AM, Mensink GB, Schienkiewitz A, Haftenberger M, Lampert T, Ziese Lønning PE, Haynes BP, Straume AH, Dunbier A, Helle H, Knapp-Labrie F, Ke Y, Gonthier R, Belanger A (2015) All sex steroids are made intracellularly in peripheral tissues: implications for studies using mammmary biospecimens. Arch Toxicol 93:2823–2833. https://doi.org/10.1007/s00204-019-02564-w

Pemp D, Esch HL, Hauptstein R, Möller FJ, Zierau O, Bosland MC, Geppert LN, Kleider C, Schlereth K, Vollmer G, Lehmann L (2019b) Novel insight in estrogen homeostasis and bioactivity in the ACI rat model of estrogen-induced mammmary gland carcinogenesis. Arch Toxicol 93:1979–1992. https://doi.org/10.1007/s00204-019-02483-w

Rosner W, Hankinson SE, Sluss PM, Vesper HW, Wierman ME (2013) Challenges to the measurement of estradiol: an endocrine society position statement. J Clin Endocrinol Metab 98:1376–1387. https://doi.org/10.1210/jc.2012-3780

Runge-Morris M, Kocarek TA, Falany CN (2013) Regulation of the cytosolic sulfotransferases by nuclear receptors. Drug Metab Rev 45:421–458. https://doi.org/10.3109/03602532.2012.748794

Safe S, Kim K (2008) Non-classical genomic estrogen receptor (ER)/specificity protein and ER/activating protein-1 signaling pathways. J Mol Endocrinol 41:263–275. https://doi.org/10.1677/JME-08-0103

Salovainen-Peltonen H, Vihma V, Leidenius M, Wang F, Turpeinen U, Hamalainen E, Tikkanen MJ, Mikkola TS (2014) Breast adipose tissue estrogen metabolism in postmenopausal women with or without breast cancer. J Clin Endocrinol Metab 99:E2661–2667. https://doi.org/10.1210/jc.2014-2550

Sherman ME, Figueroa JD, Henry JE, Clare SE, Rufenbarger C, Storniolo AM (2012) The Susan G. Komen for the Cure Tissue Bank at the IU Simon Cancer Center: a unique resource for defining the “Molecular Histology” of the breast. Cancer Prev Res (Phila) 5:528–535. https://doi.org/10.1158/1940-6207.CAPR-11-0234

Stanczyk FZ, Mathews BW, Sherman ME (2015) Relationships of sex steroid hormone levels in benign and cancerous breast tissue and blood: a critical appraisal of current science. Steroids 99:91–102. https://doi.org/10.1016/j.steroids.2014.12.011

Starlard-Davenport A, Lyn-Cook B, Radominska-Pandya A (2008) Novel identification of UDP-glucuronosyltransferase 1A10 as an estrogen-regulated target gene. Steroids 73:139–147. https://doi.org/10.1016/j.steroids.2007.09.007

Tralau T, Luch A (2013) The evolution of our understanding of endo-xenobiotic crosstalk and cytoplasmic P450 regulation and the therapeutic implications. Expert Opin Drug Metab Toxicol 9:1541–1554. https://doi.org/10.1517/17440670.11-0234

Tsilidis KK, Allen NE, Key TJ, Dossus L, Lukanova A, Bakken K et al (2011) Oral contraceptive use and reproductive factors and risk of ovarian cancer in the European Prospective Investigation into Cancer and Nutrition. Br J Cancer 105:1436–1442. https://doi.org/10.1038/bjc.2011.371

Vihma V, Wang F, Salovainen-Peltonen H, Turpeinen U, Hamalainen E, Leidenius M, Mikkola TS, Tikkanen MJ (2016) Quantitative determination of estrone by liquid chromatography-tandem mass spectrometry in subcutaneous adipose tissue from the breast in postmenopausal women. J Steroid Biochem Mol Biol 155:120–125. https://doi.org/10.1016/j.jsbmb.2015.10.004
Yager JD, Davidson NE (2006) Estrogen carcinogenesis in breast cancer. N Engl J Med 354:270–282. https://doi.org/10.1056/NEJMr a050776

Yaghjyan L, Colditz GA (2011) Estrogens in the breast tissue: a systematic review. Cancer Causes Control 22:529–540. https://doi.org/10.1007/s10552-011-9729-4

Zhao H, Zhou L, Shangguan AJ, Bulun SE (2016) Aromatase expression and regulation in breast and endometrial cancer. J Mol Endocrinol 57:R19–33. https://doi.org/10.1530/JME-15-0310

Publisher’s Note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.