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TITLE: Identifying Targetable Immune Vulnerabilities in Young Women's Breast Cancer Liver Metastases

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14. ABSTRACT: Background: Patients diagnosed with breast cancer within 10 years of a completed pregnancy have ~1.5-3 fold higher risk of developing metastatic disease. This increased risk occurs in ER+, ER-, and Her-2+ disease, and is not observed in patient age, stage and biologic subtype matched breast cancer patients who have never had a child, nor in women initially diagnosed and treated during their pregnancy, highlighting an unappreciated role of postpartum reproductive biology in breast cancer metastases in young women. Importantly, the only reported site of increased metastases in postpartum women is the liver. This specificity for liver metastasis is the foundation for proposing that the mammary gland and liver are functionally linked postpartum, leading to increased risk for breast cancer metastasis to the liver. To address this novel hypothesis, we turned to rodent models and identified a previously unreported biology, weaning-induced liver involution. In our rodent models we discovered that like the breast, the liver expands to meet the nutrient and metabolic demands of lactation and upon weaning undergoes a developmental program to return to a pre-pregnant like state through an active tissue remodeling process characterized by hepatocyte apoptosis, catabolic metabolism, immune cell influx and ECM remodeling. Further, using our rodent models we found liver involution supports enhanced establishment of metastatic tumors compared to livers of nulliparous hosts. Since programmed cell death is functionally linked to immune tolerance in a variety of biologic contexts, we hypothesize that PPBC liver metastases will be characterized by active immune tolerance and responsivity to checkpoint inhibitors. Objective: We hypothesize that weaning-induce liver involution induces a local immune tolerance phenotype that imprints metastatic lesions with active and sustained immune avoidance capability. These data would implicate normal liver involution in the high rates of liver metastases observed in PPBC patients, and provide novel pathways for targeted immunotherapeutics. Aim 1—Define the immune milieu of non-tumor bearing, involuting murine livers. Aim 2—Define the immune milieu of PPBC and non-PPBC liver metastases in mice and humans. Aim 3—Test efficacy of immunotherapy for PPBC liver metastasis in rodent models.					
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1. INTRODUCTION

Metastasis is associated with worse outcomes and overall poorer survival, but liver metastasis is particularly devastating, having some of the worse outcomes compared to other metastatic sites. Liver metastasis occurs at high frequency in young women diagnosed with breast cancer. In rodent models, we have demonstrated a functional link between the liver and the breast during weaning that transiently transforms the liver into rich “soil” for breast cancer establishment. In addition, our lab has the first-ever evidence that similar post-partum liver biology occurs in women. In this grant, we will expand our molecular understanding of weaning-induced liver involution in mice; investigate how liver involution influences the immune milieu of breast cancer metastasis to the postpartum liver (MLBC) in both rodents and women, and employ rodent models to interrogate immunomodulatory agents to target MLBC.

KEYWORDS: Young women’s breast cancer, pregnancy, lactation, weaning, postpartum breast involution, postpartum breast cancer, liver, breast cancer liver metastasis, metastatic niche, immune milieu, inflammation, myeloid

2. ACCOMPLISHMENTS

What were the major goals of the project and accomplishments?

Aim 1. Define the immune milieu of healthy, involuting rodent livers compared to nulliparous host livers.

- A. Preform RNAseq on livers from non-tumor bearing animals
 - Year 1, months 1-6: Isolation and Quality control of RNA: *100% complete*
 - Year 1, months 7-9: Library preparation: *100% complete*
 - Year 1, months 9-15: Two rounds of sequencing: *100% complete*
 - Year 1, months 16-18: Computational alignment and normalization of data: *100% complete*
 - Year 1, months 19-30: Analysis of RNA expression results: *90% complete*
 - Year 2, manuscript preparation: *90% complete*

- B. Immune cell profiling in non-tumor bearing mouse livers across postpartum developmental spectrum, n = 6 for each time point, Nulliparous, Lactation, Days 2, 4, 6, 8 and 4wks post weaning
 - The method was switched from mIHC to flow cytometry*
 - Year 1, months 1-6: Antibody validation: *50% complete (T cell panel completed, missing myeloid panel)*
 - Year 1, months 7-12, Year 2, months 1-12, Year 3, months 1-2: Acquisition: *50% complete- missing myeloid panel*
 - Year 3, months 5-10: Analysis: *50% complete-missing myeloid panel*

Aim 2. Identify immune milieu of PPBC and non-PPBC liver metastases in rodents and humans.

- A. Perform RNA-Seq on FFPE from postpartum and non-postpartum breast cancer liver metastases human samples (n =16): 0% complete
- B. Multiplex IHC of Human Breast Cancer Liver Metastasis from 16 Non-postpartum and 9 Postpartum and 5 antibody control FFPE specimens: 50% complete.
 - Year 1, months 1-6: Antibody reagent purchase, lot validation and aliquot: *100% complete for Discovery panel, ECM panel and MDSC panel*
 - Year 1, months 7-12: Multiplex IHC staining, image acquisition for 30 cases for 25 analytes: *100% complete*
 - Year 2, months 1-6: Image Alignment, Region Selection and Cell Segmentation: *85% complete*
 - Year 2, months 6-12: Single Cell Image Cytometry Analysis: *75% complete*
 - Year 3, months 1-12: Data Analyses and manuscript preparation: *25% complete*
- C. Multiplex IHC on Murine Postpartum and Nulliparous Liver metastases (6 postpartum, 6 nulliparous)
 - Year 1, months 1-6: Antibody reagent purchase, lot validation and aliquot: *100% complete for Discovery panel and MDSC panel*
 - Year 1, months 7-12: Multiplex IHC image acquisition for 12 slides for 25 analytes: *50% complete*

Year 2, months 1-3: Image Alignment, Region Selection and Cell Segmentation: *50% complete*

Year 2, months 4-6: Single Cell Image Cytometry Analysis: *50% complete*

Year 2, month 6: Completion of immune characterization of murine breast cancer liver metastasis from postpartum and non-postpartum murine models: *35% complete*

Aim 3. Identify efficacy and specificity of immunotherapy for PPBC liver metastasis.

- A. Portal vein implantation of breast cancer cell lines in nulliparous and involution animals with therapeutic administration immunotherapeutics (proposed to do this in 4 cell lines (D2A1, D2OR, E0771, and PY230-have only completed work outlined in this aim for a single cell line: Py230. Year 1, months 1-3: Animal Ordering, Breeding and nursing: *25% complete (Py230 data set is complete)*
- Year 2: tested efficacy of ibuprofen combined with anti-PD1 using portal vein liver Py230 metastasis model. Animal work is *100% completed*.
- Year 2: Monitor metastatic tumor growth in response to immunomodulatory agents, collect samples for flow cytometry, flash frozen, and FFPE tissue banks: *100% complete*
- Year 2: Endpoint analyses: tumor size, Ki67: *25% complete*
- B. Determination of tumor specific T cell response to therapy by flow cytometric analysis of tumor
- Year 1, month 7: Antibody reagent purchase, lot validation and aliquot: *100% complete*
- Months 7, 15, 23, 31: Perform Flow cytometry
- Develop flow panel: *100% complete*
 - Perform flow analysis on study samples: *25% complete overall, 100% complete for Py230 cell line)*
- Year 2, months 8-10: Detailed Analysis of Flow Cytometric Data and manuscript preparation: *0% complete*
- C. Immune focused Multiplex IHC on livers from tumor injected mice comparing involution experienced mice with and without immunomodulatory treatment based upon response to treatment across 2 strains and 4 cell lines. (48 slides)
- Year 1, months 1-6: Antibody reagent purchase, lot validation and aliquot: *60% complete*
- Year 1-3 Rolling: Multiplex IHC image acquisition for 48 slides for 25 analytes, as specimens become available: *0% complete*
- Year 1-3 Rolling: Image Alignment, Region Selection and Cell Segmentation: *0% complete*
- Year 1-3 Rolling: Single Cell Image Cytometry Analysis and manuscript preparation: *0% complete*

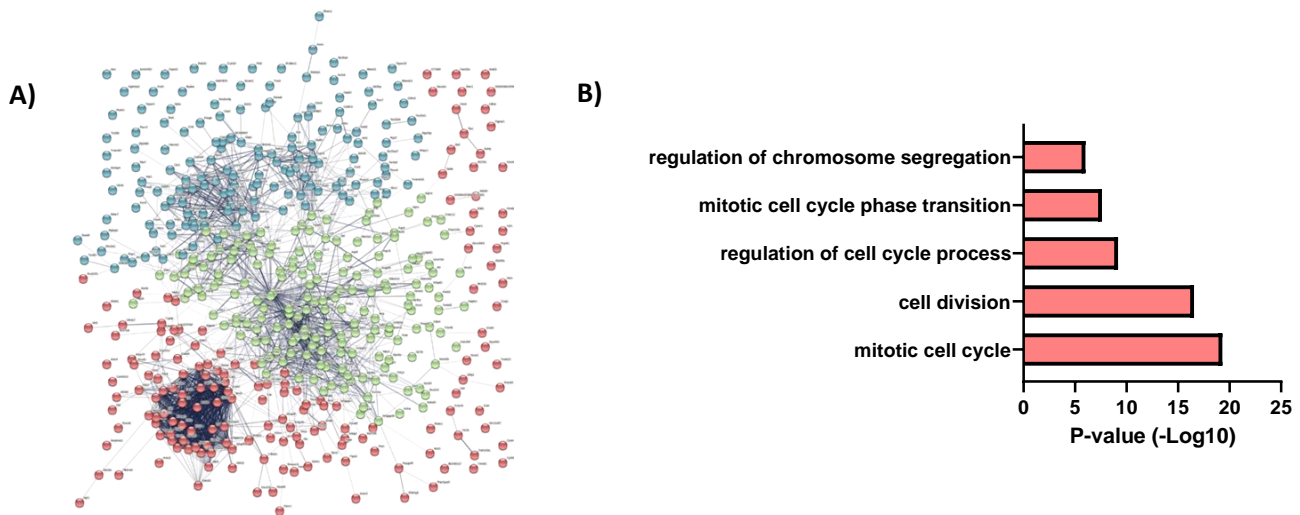


Fig 3. A) STRING analysis of the top 500 most differentially expressed genes between nulliparous and lactation livers. **B)** Panther Go Term Enrichment of the top most significantly enriched family terms for the red cluster.

We next used Gene Set Enrichment Analysis (GSEA) to the fold enrichment differences in these pathways and to identify which group these pathways were enriched in. Using this method of two group comparisons we identified proliferation was upregulated during lactation, fatty acid synthesis was upregulated during lactation, amino acid catabolism was upregulated during early involution, and apoptosis was upregulated during involution day 4 (**Fig 4A**).

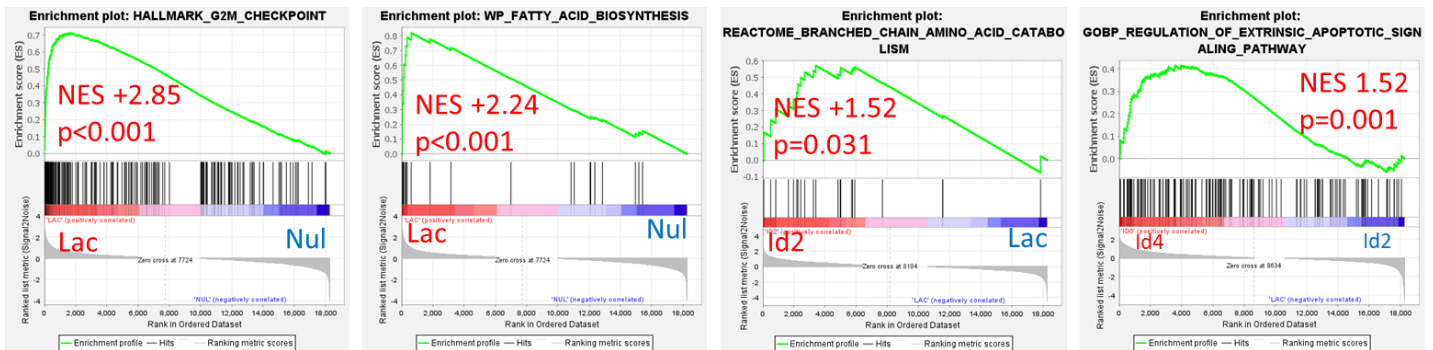


Fig 4. GSEA analysis showing an enrichment of proliferation during lactation, fatty acid synthesis during lactation, amino acid catabolism during involution day 2, and apoptosis during involution day 4.

One of the main objectives from this study was to identify how the immune milieu changes during a reproductive cycle. We hypothesized that we would identify immune signatures in the involution liver that are consistent with tumor promotional gene signatures. First, we used GSEA to compare nulliparous livers to involution day 4 or 6 livers. We identified several cell types were enriched during involution day 4 including monocytes, macrophages, neutrophils, dendritic cells, eosinophils and B cells (**Table 1, white rows**). Interestingly, we were able to detect differences in immune cells between different involution days, seeing an influx of T-cells during involution day 6 (**Table 1, grey rows**). By flow cytometry, and as reported in year 1, we found T cell composition within the liver was reproductive state dependent, with data consistent with immune suppression/tolerance and regulator T cells upregulated in the liver postwean (**Fig 5**)

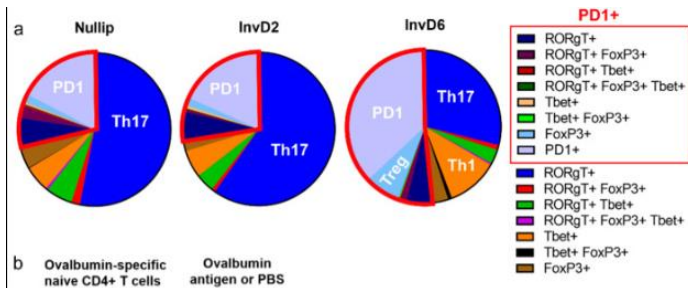


Fig 5: T cell composition in the liver changes with reproductive state. The nulliparous liver is characterized by a TH17 skew, which becomes more regulatory during involution. From Bartlett et al, Cancers 2021

To follow up on these findings, we will employ multiplex immunohistochemistry (mIHC) (Aim 1B) to guide us towards identify specific myeloid populations, which we have yet to characterize. To this end, we have generated a tissue microarray (TMA) of liver samples from mice throughout a full reproductive cycle (n=6 per group) that we are using to optimize our mIHC antibody panels. We will use this TMA as an internal control on each mIHC study slide. Two mIHC panels are in the works (Table 2).

Pan - Immune Cells			
CD45	Pan - Immune Cells		
CD3e	All T cells	Primary T cells subsets	CD79b
CD4	T - helper (Th)		Ly6c/CSF1R
CD8	Cytotoxic T (Tc)	T cell differentiation	Ly6g/CD66b
Tbet	Cytotoxic -effector (Th1/Tc1)		MHCI/HLA-DRPQ
EOMES	Memory T cell - Exhaustion (Tc1)		F480+/CD68
Foxp3	Regulatory T cell		IRF 8
RORgt	Th17 - Tc17 effector	T cell activation	IRF4
GATA3	Th2-Tc2 effector		IRF5
NUR77	Antigen induced T cell activation	T cell activation	B7-2 - CD86
p-ZAP70			PPARG
Granzyme A	Cytolytic Granules	ICAM-1	Inflamed Endothelium
Ki67	Proliferation	COX-2	Immune tolerance enzyme - NSAID Target
CTLA-4	Activation induced Inhibitory	Cleaved Caspase 3	Cell Death
PD-1	Receptor - Checkpoint target	m-mouse	h-human
			Pro - tumor
			Anti-tumor

Aim 2 Objective: Identify immune milieu of PPBC and non-PPBC liver metastases in humans and rodents

Aim 2: Significant Results and Key Conclusions:

Human tissue studies:

- We have obtained demographic and clinical data for 30 breast d metastases to the liver. Staying consistent with our previous human studies, we are defining PPBC as patients who were diagnosed with breast cancer within 10 years after their last pregnancy. In this patient cohort, 18 patients are considered non-PPBC and 12 were identified as PPBC (Table 3).
 - We are currently in the process of obtaining 11 new breast cancer liver metastatic cases and adding them to our cohort.
- We are also in the process of obtaining H&E slides for 10 of our existing 30 cases, so that we can update our differentiation and invasion statistics for this cohort.

GS link to MsigDB	NES	NOM p-val
MONOCYTES	2.65	<0.001
NEUTROPHILS	2.21	<0.001
MACROPHAGES M1	2.2	<0.001
MACROPHAGES M2	2.04	<0.001
MACROPHAGES M0	1.95	<0.001
DENDRITIC CELLS RESTING	1.9	<0.001
DENDRITIC CELLS ACTIVATED	1.74	<0.001
EOSINOPHILS	1.64	0.011
B CELLS NAIVE	1.53	0.014
B CELLS MEMORY	1.5	0.029
MONOCYTES	2.15	<0.001
DENDRITIC CELLS RESTING	2.13	<0.001
NK CELLS ACTIVATED	2.11	<0.001
T CELLS CD4 MEMORY RESTING	2.07	<0.001
T CELLS CD8	2.04	<0.001
MACROPHAGES M1	1.97	<0.001
T CELLS GAMMA DELTA	1.93	<0.001
MACROPHAGES M2	1.93	<0.001
T CELLS CD4 MEMORY ACTIVATED	1.78	<0.001
NK CELLS RESTING	1.78	0.004
T CELLS REGULATORY (TREGS)	1.75	<0.001
MACROPHAGES M0	1.74	<0.001
DENDRITIC CELLS ACTIVATED	1.58	0.002
T CELLS FOLLICULAR HELPER	1.53	0.019
T CELLS CD4 NAIVE	1.49	0.03
EOSINOPHILS	1.47	0.027

Aim 2A. Perform RNA-Seq on FFPE from PPBC and non-PPBC breast cancer liver metastases human samples.

We are currently in the process of updating our IRB to get approval for genetic sequencing. Once this is completed, we will proceed with isolating RNA from the samples currently in our lab. See **4. Challenges/Problems for details.**

Aim 2B. Multiplex IHC of Human Breast Cancer Liver Metastasis from 18 Non-postpartum and 10 Postpartum and 5 antibody control FFPE specimens.

Previously, we developed and mIHC panel to stain for tumor cell markers and host liver cells, which we refer to as the Discovery panel (**Table 4**). We have used these stains, along with H&E slides to determine if our tissue samples contained adjacent normal, tumor border, and tumor stroma tissue regions (**Table 5**). Previous studies in our lab, as well as others, have found immune cell profiles are different between intratumoral regions versus adjacent normal, tumor border, and tumor stroma regions (Jindal, et al., Nature Communications, 2021). Our previous staining panels will be analyzed by these different tissue regions.

Table 4. The Discovery Panel for HUMAN is composed of a variety of markers to look at native liver cell types, as well as markers for breast cancer cells.

Marker	Purpose
CK18	Luminal-origin breast cells, and hepatocytes
Gata3	Luminal-origin breast cells
Aquaporin 1	Bile duct cells, endothelial cells
ER	Estrogen receptor status in liver metastases
CK7	Luminal-origin breast cells, bile duct cells
CK5	Basal-organ breast cells
Ki67	Assess cell proliferation
Col IV	Basement membrane
Heppar	Hepatocytes
E-cad	Junctional complex associated with polarized cells
aSMA	Fibroblast marker

In addition, we developed a human myeloid derived suppressor cell (MDSC) panel this past year (**Table 6**), and have demonstrated that the normal breast has an influx of MDSCs as early as two weeks post wean (**Fig 6**). Because to date, the molecular signatures and immune infiltrates of mammary gland involution have informed our understanding of liver involution, we anticipate an increase in MDSCs in the normal involuting liver, as well as in BCLM that develop in the postpartum period.

Table 6. MDSC panel for Human tissue was developed to investigate the different myeloid immune cell populations that may be changing by parity status.

Cell	Identification
T cells	CD45+CD3+
B cells	CD45+CD3-CD20+
NK	CD45+CD3-CD20-CD56+
Neutrophil	CD45+CD3-CD20-CD56-CD66b+S100A-
DC	CD45+CD3-CD20-CD56-CD66b+S100A-CD11c+
Macrophage	CD45+CD3-CD20-CD56-S100A+CD68+
PMN-MDSC	CD45+CD3-CD56-HLA-S100A+CD66b-CD11b+CD33+CD68-CD14-CD15+
M-MDSC	CD45+CD3-CD56-HLA-S100A+CD66b-CD11b+CD33+CD68-CD14+CD15-

Table 3. Demographics of patients with liver metastases and primary breast cancer.

	N (%)
Age at diagnosis	
<45	14 (50.0%)
45-60	12 (42.9%)
>60	2 (7.1)
PPBC	
Yes	10 (35.7%)
No	18 (64.3%)
Race	
White	26 (92.9%)
Black	0 (0%)
Asian	1 (3.6%)
Hispanic	1 (3.6%)
Estrogen	
Positive	22 (78.6%)
PR+	17 (77.3%)
PR-	5 (22.7%)
Negative	5 (17.9%)
Unknown	1 (3.6%)

Table 5. Cases that have tumor border, adjacent normal tissue, and tumor stroma.

	N (%)
Tumor Border	22 (78.6%)
Adjacent Normal	17 (60.7%)
Tumor Stroma	3 (10.7%)

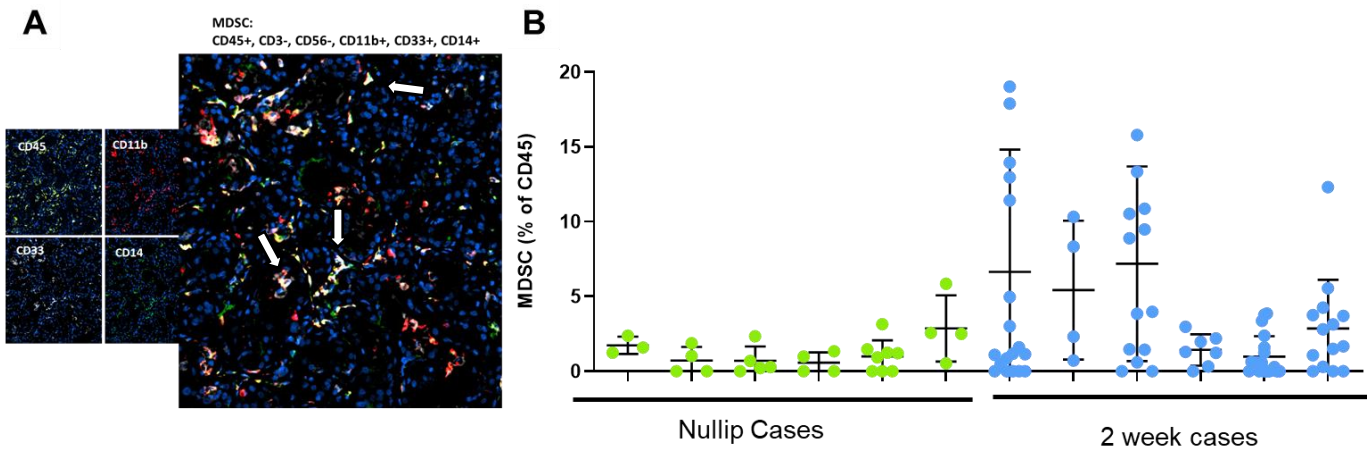


Figure 6. MDSC abundance in normal human breast varies by reproductive state—a model for liver involution. **A)** mIHC staining of normal human breast. MDSC are defined here as CD45+, CD3-, CD56-, CD11b+, CD33+, CD14+, and appear as white cells in the image (arrows). **B)** Quantitation of MDSCs in human breast. Left green, nulliparous (n=6 women) and right blue, 2 weeks postwean biopsies (n=6 women). We provide these data as proof of principal that we have a working protocol to investigate MDSCs in BCLM.

Aim 2C. Multiplex IHC on Murine Postpartum and Nulliparous Liver metastases (6 postpartum, 6 nulliparous).

We have developed a myeloid derived suppressor cell (MDSC) panel for mouse tissue consisting of F480, Cd45, Cd11b, and Ly6G/C. This panel was worked up on murine mammary tumors (**Fig 7**). Staining for this panel is underway.

We have a long standing interest in understanding immune cells in the context of tissue region and neighboring stroma cells. Of particular interest are liver fibroblasts in the tumor microenvironment. We found that within mouse liver metastasis samples that there are differences in fibroblast activation marker, Pdgfra (**Fig 8**). While percent area stained was not different between nulliparous and involution tumors, intensity of stain appeared stronger. We also stained for additional fibroblast markers including Desmin, Gfap, Fabp4, AlphaSMA, PDGFRalpha and PDGFRbeta. These stains will be further analyzed for stain intensity as well as differences in staining patterns. These findings have encouraged us to add above list of fibroblast markers to our mIHC T cell panel (**Table 7**), in order to investigate immune cells in the context of fibroblasts, as fibroblasts and immune cells are known to have crosstalk interactions (Van Linthout, et al., 2014).

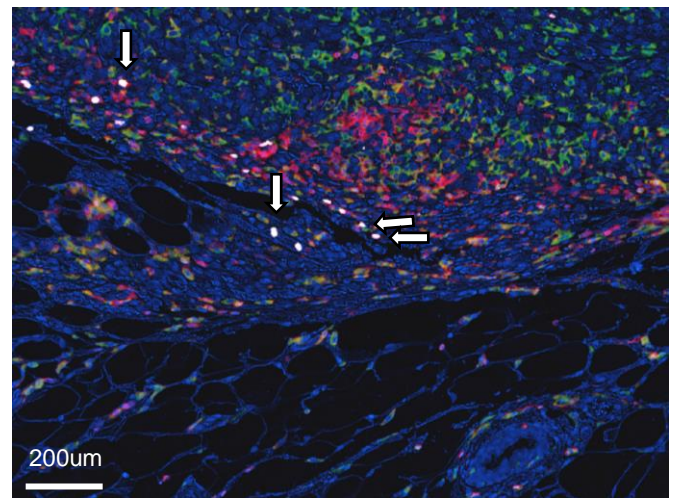


Fig 7. Example pseudo-colored image of a murine mammary tumor border showing presence of MDSC: **F480** (green), **CD45** (red), **CD11b** (orange) and **Ly6G/C** (blue).

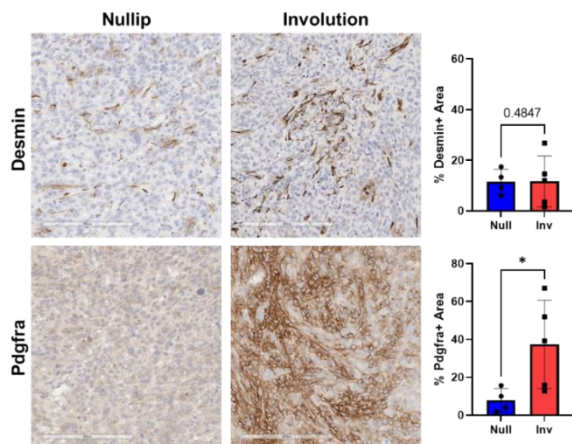


Fig 8. Desmin and Pdgfra immunohistochemistry staining on fixed D2A1 mammary tumors that were injected via portal vein. N=4-5 per group

Aim 3 Objective: Identify efficacy and specificity of immunotherapy for PPBC liver metastasis

Major Activities Overview:

- Expanded our PPBC liver metastasis model to C57/B6 mouse mammary tumor cell line, Py230
- Completed our first immunotherapy treatment of PPBC liver metastasis in our Py230 model

Aim 3: Significant Results and Key Conclusions

Our lab has primarily used BALB/C mice in our postpartum breast cancer studies. BALB/C mice have a Th2 skewed environment, whereas C57/B6 have a Th1 skew. As we investigate the importance of immune cell changes in postpartum breast cancer metastasis, we feel it is important to also investigate our model in C57/B6 mice. Last year we completed our first study in C57/B6 mice, injecting C57/B6 mammary tumor cell line, EO771, via portal vein. We have since introduced an additional C57/B6 mammary cell line, Py230. Similar to our other mouse models of mammary cancer liver metastasis, we find Py230 cells have increased incidence when injected into involution hosts (Fig 9).

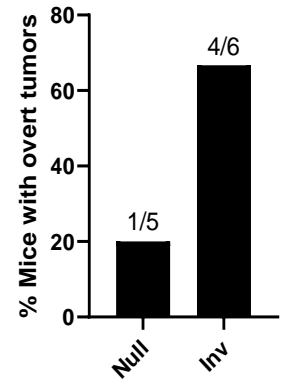


Fig 9. Mice injected with Py230 cells via portal vein, with overt tumors at study end (n=5-per group).

To investigate possible treatments for PPBC liver metastasis we conducted our first PPBC-directed treatment study. We used C57/B6 mice (Nullip, Inv, and Inv with treatment groups) injected with Py230 cells via portal vein, and treated half the mice in the involution group with ibuprofen and anti-PD1 therapy (Fig 10). Tumors were allowed to grow for 4 weeks, and then livers were harvested for flow cytometry analysis and fixed for immunohistochemistry. Two liver lobes were taken for flow cytometry analysis using our Cd4 and Cd8 flow panels that we developed last year (Table 7). Levels of ROR γ T, Tbet, and Gata3 were very low and nearly undetectable in our samples. Due to low signals of markers, we only reported on Cd4, Cd8, and Pd1. We found there were no significant differences between groups in these three T cell markers (Fig 11). However, a potential technical caveat exists, in this study most of the tumors were micrometastases, and thus the immune profiles reflected in our flow data may be more representative of normal liver and may not reflect immune cell responses in or around the tumors. mIHC with the lymphoid and myeloid markers may be a better approach to the question of whether tumor immune infiltrate was impacted by treatment, and will be pursued in year 3.

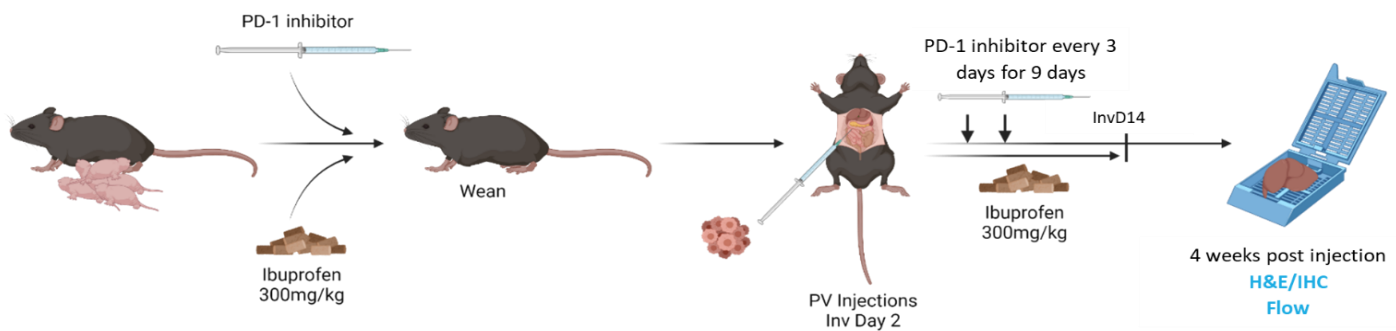


Fig 10. Experimental design for the treatment of PPBC liver metastasis with ibuprofen and anti-Pd1 therapy.

Table 7. CD4 (left) and CD8 (right) panels developed for flow cytometry in murine livers.

Marker	Identification	Marker	Identification
CD45	General immune cells	CD45	General immune cells
CD3	T-cells	CD3	T-cells
CD4	CD4 helper T-cells	CD8	Cytotoxic T-cells
PD1	Activated or exhausted T-cells	PD1	Activated or exhausted T-cells
FoxP3	T-regulatory	CTLA4	Activation
ROR γ T	Th17	CD49b	NK cells
Tbet	Th1	CD11b	Leukocyte and NK maturation
Gata3	Th2	Live/Dead	-
Live/Dead	-		

We quantified tumor burden by H&E. Micrometastases were assessed for a 40 μ m section of liver (**Fig 12**). Tumor burden was quantified, showing involution mice have increased tumor burden, and treatment was able to reduce tumor burden to nulliparous levels. These data suggest that targeting the inflammatory pathways upregulated in the involuting liver are sufficient block the pro-tumor attributes to nulliparous levels. However, these data only 40 μ m of liver per animal. Additional levels will be assessed and added to the quantification.

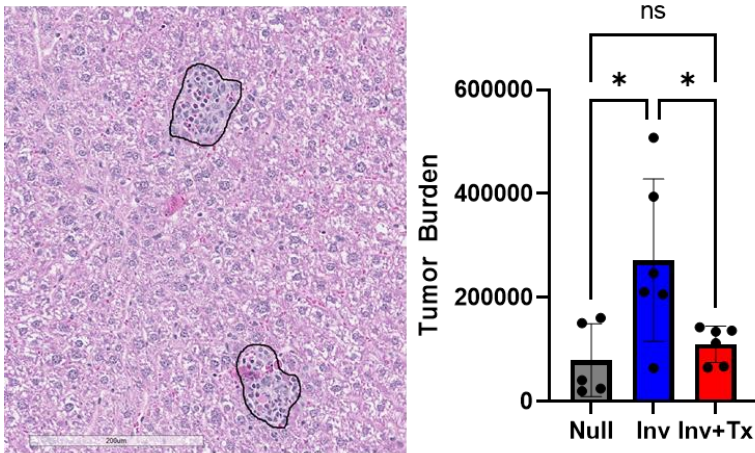


Fig 12. H&E stain of liver showing micrometastases (left) and tumor burden quantified for different groups (right).

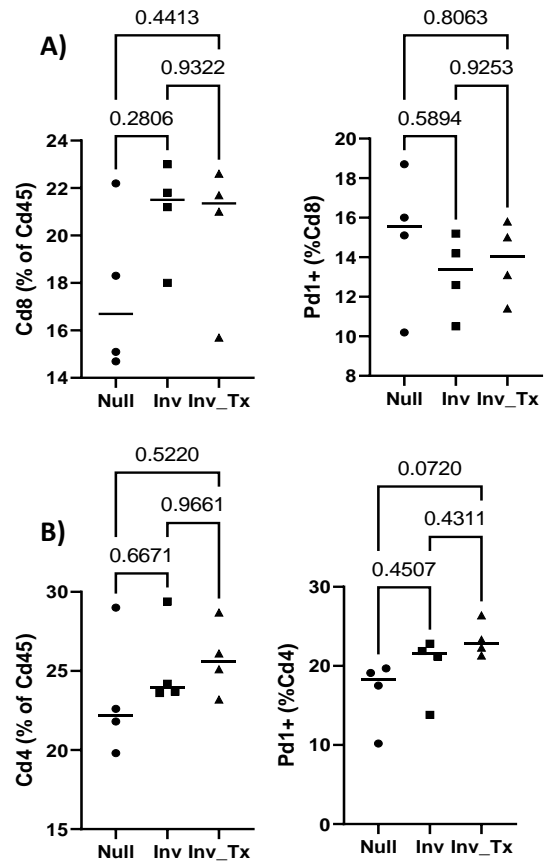


Fig 11. Flow cytometry on liver injected via portal vein with Py230 cells. **A)** Percentage of Cd8 cells (left) and Pd1+ Cd8 cells (right). **B)** Percentage of Cd4 cells (left) and Pd1+ Cd4 cells (right).

What opportunities for training and professional development has the project provided?

Michelle Ozaki: Ms. Ozaki is a PhD candidate who has been working on the aims of this award.

- Developed bench skills in multiplex IHC, flow cytometry, animal handling, tissue staining, and analysis
- Trained in the unique portal vein surgery to model liver metastasis, and expanded the Schedin lab's expertise to include two new murine mammary tumor cell lines required for the completion of this grant.
- Learned big data analysis tools
- Honed data presentation skills across monthly presentations in lab meeting

- Developed written communication skills through the preparation of monthly plans and study summaries, completion of her qualifying exam, and preparation of three grants including the NSG GRFP, and an internal departmental grant
- Data interpretation and study design on the RNAseq portion of the project provided preliminary data that aided in her submission of an F31 grant
- Honed presentation skills through completion of the oral portion of her qualifying exam, monthly presentations at lab meetings, and presentations at joint department lab meetings
- Developed professional network and awareness of the field through participation in conferences including:
 - AACR 2022
 - Gordon Conference on Mammary Gland

Hatun Cete Duran: Dr. Cete Duran is a neonatal surgeon Turkey who has recently immigrated to the US and is interested in pursuing a career in biomedical research, particularly in the area of young women's breast cancer, which is a stigmatized disease in Turkey. Dr. Duran is applying for an internal fellowship to become a postdoctoral trainee.

- Developed animal handling skills
- Has applied her training as surgeon to helping improve BCLM surgeries done in the lab
- Trained in mIHC staining and data analysis

Melody Brizuela: Ms. Brizuela is first generation college applicant, currently in high school, who was a 2022 summer intern in our lab. Through coordination with her high school teachers, Ms. Brizuela continues her training in the lab as a volunteer.

- Developed bench skills in immunohistochemistry
- Honed presentation skills through presenting during Schedin lab meetings and participating in the summer student poster conference, where she won a best poster presentation award

How were the results disseminated to communities of interest?

Pepper Schedin:

National/International Speaking Invitations:

- Keynote Speaker, Launch of the Early Diagnosis Centre; a joint collaboration between The Royal Marsden and The Institute of Cancer Research (RM-ICR), London, UK, September 10, 2021.
- Session Planner, Session 3, ECM Characterization Modalities, The Matrix in Focus: Matrix, Cells, and Interactions in Health, Disease, Aging, and Regeneration, Virtual, St. Louis, MO, Sept 12-15, 2021
- 2021 Session Co-chair, Session 5: The Metastatic Niche, Cold Spring Harbor Laboratory, Biology of Cancer: Microenvironment & Metastasis, Virtual, Cold Spring Harbor, NY, Oct 12-15, 2021.
- 2021 Experienced Investigator Talk, Division of Cancer Prevention New R01 Grantee Workshop, NIH, virtual, November 18, 2021
- 2022 AACR Cancer Progress Report 2022 Steering Committee and Advisors member. Be Cognizant of Reproductive and Hormonal Influences
- Physiologic Tissue Regression and Breast Cancer Metastasis in Young Women. Theresa's Research Foundation's 8th Annual Metastatic Breast Cancer Conference, University of Utah, Salt Lake City, Utah, September 9, 2021
- 2021 Keynote Speaker, Early Detection a Global Perspective, Launch of the Early Diagnosis Centre; a joint collaboration between The Royal Marsden and The institute of Cancer Research (RM-ICR), London, UK, September 10, 2021
- 2021 Breast Cancer Early Detection Challenges & Opportunities, Early Detection of Cancer Conference, Virtual, CRUK, the OHSU Knight Cancer Institute and the Canary Center at Stanford, London UK, October 6, 2021
- 2021 Physiologic Tissue Regression and Breast Cancer Metastasis in Young Women. CSHL Cancer Biology and Metastasis Conference, Cold Spring Harbor, NY, Virtual, October 12, 2021
- 2022 Mammary Collagens and Breast Cancer Metastasis are under Reproductive Control. FASEB Catalyst Conference, Biochemical and Biochemical Impact of Collagens in Cancer, Virtual, March 24, 2022

- 2022 Compromised Myoepithelium Cell Differentiation Correlates with DCIS to IDC Transition. AACR Special Conference: Rethinking DCIS: An Opportunity for Prevention? Philadelphia, Pennsylvania, Sept 9, 2022

Local Speaking Invitations:

- Multidisciplinary study of the intersection between reproduction and breast cancer, Faculty Forum Lunch, CDCB, February 16, 2022

Michelle Ozaki

- Poster presentation:
 - Gordon Research Conference/Gordon Research Seminar—Mammary Gland (May 2022, Lucca, Italy). *RNA-seq analysis of murine liver to identify breast cancer metastatic potential during liver involution*. Michelle Ozaki, Alexandra Bartlett, Zheng Xia, Pepper Schedin
 - The American Association for Cancer Research Conference. (April 2022, New Orleans, LA). *RNA-seq analysis of murine liver to identify breast cancer metastatic potential during liver involution*. Michelle Ozaki, Alexandra Bartlett, Zheng Xia, Pepper Schedin

What do you plan to do during the next reporting period to accomplish the goals?

Aim 1—Define the immune milieu of non-tumor bearing, involuting murine livers.

- Submit RNAseq paper
- Complete MDSC mouse staining in normal liver—lymphoid data was obtained by flow cytometry, but fixed tissues for livers are available and will be used to complete MDSC analyses by mIHC staining

Aim 2—Define the immune milieu of PPBC and non-PPBC liver metastases in mice and humans.

- Complete discovery and MDSC mIHC panels, prepare manuscript
- Complete IRB approval process to sequence human breast cancer liver metastasis cases
- Complete histological analyses of human BCLM cases, prepare manuscript

Aim 3—Test efficacy of immunotherapy for PPBC liver metastasis in rodent models.

- Characterize immune milieu of tumors in the PPBC liver metastasis with ibuprofen and anti-Pd1 therapy study shown in Figs 10, 11, & 12
- Repeat immune therapy treatment study using D2A1 mammary tumor cells in our PPBC BCLM model

3. IMPACT

What was the impact on the development of the principal discipline(s) of the project?

- Our studies are paradigm changing for the metastasis field, which is founded on metastasis being an intrinsic attribute of tumor cells rather than dependent on reproductive biology of the host. Our data shows how a common physiological process, weaning-induced liver involution, remodels the liver microenvironment and makes it susceptible to metastasis.
- Our studies have advanced the early onset breast cancer community by delineating between pregnancy associated breast cancer, a rare and not poor-prognostic cancer, and postpartum breast cancer, a common and deadly cancer. This last year, a group of world thought-leaders came together to write an opinion piece updating the field on PPBC: *Amant F, Lefrère H, Borges VF, Cardonick E, Lambertini M, Loibl S, Peccatori F, Partridge A, Schedin P. The definition of pregnancy-associated breast cancer is outdated and should no longer be used. Lancet Oncol. 2021 Jun;22(6):753-754. doi: 10.1016/S1470-2045(21)00183-2, PMID:34087122*
- Our work has motivated others to collaborate with us to use their young women’s breast cancer cohorts to determine of our observations, i.e. that the postpartum window (weaning-induced involution window) is high risk for metastasis and liver metastasis in particular. This last year we have published a study utilizing the Utah Populating Data Base, confirming the poor prognosis of PPBC and providing supporting data for increased liver

metastasis in PPBC patients: Zhang Z, Bassale S, Jindal S, Fraser A, Guinto E, Anderson W, Mori M, Smith KR, Schedin P. *Young-Onset Breast Cancer Outcomes by Time Since Recent Childbirth in Utah*. *JAMA Netw Open*. 2022 Oct 3;5(10):e2236763. doi: 10.1001/jamanetworkopen.2022.36763. PMID: 36239933

What was the impact on other disciplines?

- Our studies on breast cancer metastasis to the liver have revealed an entire new biology, weaning induced liver involution, which may have implications for both maternal and infant health. This work has led to a new concept paper linking lactating insufficiency to liver biology: Betts, CB, Quackenbush, A, Anderson, W, Marshall, N, Schedin, P. *Mucosal immunity and liver metabolism in the complex condition of lactation insufficiency*. *Journal of Human Lactation*, Aug 14, 2020. PMID: 32795211.
- Further, we have the first-ever data suggesting that liver growth during pregnancy and liver volume loss postpartum are associated with a normal, healthy pregnancy, whereas lack of this growth pattern associated with poorer maternal health including hypertension: Q Bartlett A, Vesco KK, Purnell JQ, Francisco M, Goddard E, Guan X, DeBarber A, Leo MC, Baetscher E, Rooney W, Naugler W, Guimaraes AR, Catalano P, Xia Z, Schedin P. *Pregnancy and weaning regulate human maternal liver size and function*. *Proc Natl Acad Sci U S A*. 2021 Nov 30;118(48):e2107269118. doi: 10.1073/pnas.2107269118. PMID: 34815335

What was the impact on technology transfer?

Nothing to report

What was impact on society beyond science and technology?

- Improving medical providers knowledge of young women's breast cancer (Lancet Oncology 2021 opinion piece)
- Improving public knowledge- the following is from an email to Pepper Schedin, received in August of 2021, from a breast cancer advocate: "Don't know if you've heard, but I've just retired from the KCI Scientific Research Advocate group. I will especially miss the opportunity to continue to work with you. I remember when the MBCA Landscape Report came out. I looked for the names of any OHSU researchers – and there was your name. Shortly after that I heard your lecture – in the old Hatfield conference room. Your work wasn't like anything I'd heard before. Over the years I've learned more about your exciting research. It's not that you think outside of the box. It's more like you're saying, "What's a box?" You've investigated an area full of myths and shown how the breast and other organs adapt to pregnancy and then return to normal. I know you inspire others. It's been a pleasure to know you. Thank you for the work you do."

4. Challenges/Problems

Changes in approach and reasons for change

OHSU and the State of Oregon had some of the country's most restrictive COVID containment policies, resulting in no/or only partial access to the labs from March 2020 through May 2021. Even now, as PI of the lab, as recently as 6/2022, I was considered non-essential and asked to work 100% remotely. After two years of working remotely, I have had 100% lab staff turnover, and new hires are entry-level and cannot pick up where prior staff left off. The training curve is steep. I have returned to working 80% in the lab to better mentor the new lab staff and trainees. Thus, the science outlined in DoD BC191620 has been a difficult to keep on track. None-the-less, I feel the lab members have pivoted well, and in-the-end have been highly productive given the circumstances. While we have not made changes to our original grant aims, we have changed how we have prioritized these aims. Specifically, in year 1, we performed fewer animal studies than expected and performed more multiplex IHC and RNA seq analyses. In year 2, we were able to refocus on animal experiments, however, with the loss of our experienced mIHC technical team, mIHC data analysis has been slower than expected. Also, between years 1 and 2, the clinical collaborators who were front-line for obtaining prospective BCLM (breast cancer liver metastasis) biopsies for our study either left the university for other positions, or were promoted

“up” due to personnel loss, resulting in significantly reduced clinical effort on our project. This led to a search for alternative approaches to obtaining prospective BCLM tissues. In year 2 we teamed up with the OHSU SMMART program (a personalized medicine approach for metastatic patients at OHSU), resulting in the identification of 10 new BCLM cases in hand and 11 cases identified, but not yet received from SMMART. Further, in year 2, we reached out to the Oregon Clinical & Translational Research Institute (OCTRI) to set up a daily screen of scheduled surgeries, to help us identify potential study participants. While we are still modifying our search criteria and the roles of the clinical coordinators, we are very optimistic that this automated approach will help us identify potential participants whom can be approached to determine their interest in joining our study. Finally, in part because of the COVID and science flight-induced delays in research across the board, I delayed submission of our IRB for approval to perform RNAseq on human samples. This IRB submission is now at the top of our priority list.

Actual or anticipated problems or delays and actions or plans to resolve them

See above.

Changes that had a significant impact on expenditures

We have had personnel loss this last year, and have recruited new lab members who are in training phase.

Significant changes in use or care of human subjects, vertebrate animals, biohazards, and/or select agents

No changes/nothing to report

5. Products

Publications, conference papers, and presentations

Journal publications:

Year 1:

- Betts, CB, Quackenbush, A, Anderson, W, Marshall, N, Schedin, P. Mucosal immunity and liver metabolism in the complex condition of lactation insufficiency. *Journal of Human Lactation*, Aug 14, 2020. PMID: 32795211.
- Bartlett, A.Q.; Pennock, N.D.; Klug, A.; Schedin, P. Immune Milieu Established by Postpartum Liver Involution Promotes Breast Cancer Liver Metastasis. *Cancers* 2021, 13, 1698. <https://doi.org/10.3390/cancers13071698>

Year 2:

- Q Bartlett A, Vesco KK, Purnell JQ, Francisco M, Goddard E, Guan X, DeBarber A, Leo MC, Baetscher E, Rooney W, Naugler W, Guimaraes AR, Catalano P, Xia Z, **Schedin P**. Pregnancy and weaning regulate human maternal liver size and function. *Proc Natl Acad Sci U S A*. 2021 Nov 30;118(48):e2107269118. doi: 10.1073/pnas.2107269118. PMID: 34815335
- Zhang Z, Bassale S, Jindal S, Fraser A, Guinto E, Anderson W, Mori M, Smith KR, Schedin P. Young-Onset Breast Cancer Outcomes by Time Since Recent Childbirth in Utah. *JAMA Netw Open*. 2022 Oct 3;5(10):e2236763. doi: 10.1001/jamanetworkopen.2022.36763. PMID: 36239933

Books or other non-periodical, one-time publications:

- Nothing to report

Other publications, conference papers, and presentations:

Conference presentations directly related to grant aims:

Year 1:

- Young Women’s Breast Cancer: Bench to Population Science and Back Again. Genome Science Institute, Graduate Program in Genetics and Genomics, Boston University, Boston MA, September 23, 2000

- Proximity to Pregnancy Determines Outcomes in Young Women’s Breast Cancer. McGill University, Montreal Quebec, February 8, 2021
- Proximity to Pregnancy Determines Outcomes in Young Women’s Breast Cancer. ICR Institute of Cancer Research, London, UK April 28, 2021
- Invited talk: “Multiplex IHC for Extracellular matrix proteins”. American Society of Matrix Biology Biennial Meeting. September 15, 2021, presented by graduate student Alex Quackenbush-Bartlett.

Year 2:

- Session Co-chair, Session 5: The Metastatic Niche, Cold Spring Harbor Laboratory, Biology of Cancer: Microenvironment & Metastasis, Virtual, Cold Spring Harbor, NY, Oct 12-15, 2021.
- Physiologic Tissue Regression and Breast Cancer Metastasis in Young Women. Theresa’s Research Foundation’s 8th Annual Metastatic Breast Cancer Conference, University of Utah, Salt Lake City, Utah, September 9, 2021
- Physiologic Tissue Regression and Breast Cancer Metastasis in Young Women. CSHL Cancer Biology and Metastasis Conference, Cold Spring Harbor, NY, Virtual, October 12, 2021
- Mammary Collagens and Breast Cancer Metastasis are under Reproductive Control. FASEB Catalyst Conference, Biochemical and Biochemical Impact of Collagens in Cancer, Virtual, March 24, 2022

Website(s) or other internet site(s)

Nothing to Report

Technologies or techniques

Nothing to report

Inventions, patent applications, and or/licenses

Nothing to report

Other products

Nothing to report

6. PARTICIPANTS & OTHER COLLABORATING ORGANIZATIONS

What individuals have worked on the project?

Name:	<i>Pepper Schedin</i>
Project Role:	<i>PI</i>
Researcher Identifier (e.g. ORCID ID):	0000-0003-4244-987X
Nearest person month worked:	<i>2.00</i>
Contribution to Project:	<i>Dr. Schedin led all aspects of the project, including scientific focus, experimental design, data analysis, data integrity, budget management, human and animal regulatory aspects, and manuscript writing.</i>
Funding Support:	

Name:	<i>Weston Anderson</i>
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Project Role:	<i>Scientific Writer</i>
Researcher Identifier (e.g. ORCID ID):	
Nearest person month worked:	<i>1.00</i>
Contribution to Project:	<i>Mx. Anderson worked on all manuscript writing associated with this project.</i>
Funding Support:	

Name:	<i>AeSoon Bensen</i>
Project Role:	<i>Animal Husbandry</i>
Researcher Identifier (e.g. ORCID ID):	
Nearest person month worked:	<i>8.00</i>
Contribution to Project:	<i>Ms. Bensen worked with Ms. Michelle to complete the animal surgeries and tissue collection done for Aim 3.</i>
Funding Support:	

Name:	<i>Elise De Wilde</i>
Project Role:	<i>Histotechnician</i>
Researcher Identifier (e.g. ORCID ID):	
Nearest person month worked:	<i>4.68</i>
Contribution to Project:	<i>Ms. De Wilde work on embedding all tissues fixed for Aim 3 and assisted with analysis of Py230 metastases.</i>
Funding Support:	

Name:	<i>Solange Bassale</i>
Project Role:	<i>Biostatistician</i>
Researcher Identifier (e.g. ORCID ID):	
Nearest person month worked:	<i>2.00</i>
Contribution to Project:	<i>Performs association studies between PPBC patient status and tumor characteristics, patients outcomes.</i>
Funding Support:	

Name:	<i>Reuben Hoffmann</i>
Project Role:	<i>Postdoctoral fellow</i>
Researcher Identifier (e.g. ORCID ID):	
Nearest person month worked:	<i>6.00</i>
Contribution to Project:	<i>Mr. Hoffmann is a newly recruited postdoctoral fellow who is learning animal husbandry and mIHC study designs relevant to this DoD project</i>
Funding Support:	

Name:	<i>Michelle Ozaki</i>
Project Role:	<i>Graduate Student</i>
Researcher Identifier (e.g. ORCID ID):	0000-0002-3356-9071
Nearest person month worked:	<i>12.00</i>
Contribution to Project:	<i>Ms. Ozaki carried out analysis and manuscript preparation for Aim 1, performed analysis for Aim 2, coordinated the animal studies, collected tissues, and worked on data analysis for Aim 3.</i>
Funding Support:	NSF GRFP

Has there been a change in the active other support of the PD/PI(S) or senior key personnel since the last reporting period?
Nothing to report

What other organizations were involved as partners?
Nothing to report

7. SPECIAL REPORTING REQUIREMENTS

Nothing to report

8. APPENDICES

- Q Bartlett A, Vesco KK, Purnell JQ, Francisco M, Goddard E, Guan X, DeBarber A, Leo MC, Baetscher E, Rooney W, Naugler W, Guimaraes AR, Catalano P, Xia Z, **Schedin P**. Pregnancy and weaning regulate human maternal liver size and function. *Proc Natl Acad Sci U S A*. 2021 Nov 30;118(48):e2107269118. doi: 10.1073/pnas.2107269118. PMID: 34815335



Pregnancy and weaning regulate human maternal liver size and function

Alexandra Q. Bartlett^a, Kimberly K. Vesco^b, Jonathan Q. Purnell^c, Melanie Francisco^b, Erica Goddard^d, Xiangnan Guan^e, Andrea DeBarber^f, Michael C. Leo^b, Eric Baetscher^g, William Rooney^g, Willscott Naugler^h, Alexander R. Guimaraesⁱ, Patrick Catalano^j, Zheng Xia^{k,e}, and Pepper Schedin^{a,1,m,1}

^aDepartment of Cell, Developmental, and Cancer Biology, Oregon Health & Science University, Portland, OR 97239; ^bCenter for Health Research, Kaiser Permanente Northwest, Portland, OR 97227; ^cKnight Cardiovascular Institute, Oregon Health & Science University, Portland, OR 97239; ^dPublic Health Sciences Division/Translational Research Program, Fred Hutchinson Cancer Research Center, Seattle, WA 98109; ^eComputational Biology Program, Oregon Health & Science University, Portland, OR 97201; ^fDepartment of Chemical Physiology and Biochemistry, Oregon Health & Science University, Portland, OR 97239; ^gAdvanced Imaging Research Center, Oregon Health & Science University, Portland, OR 97239; ^hDepartment of Medicine, Division of Gastroenterology and Hepatology, Oregon Health & Science University, Portland, OR 97239; ⁱDepartment of Diagnostic Radiology, Oregon Health & Science University, Portland, OR 97239; ^jMother Infant Research Institute, Department of Obstetrics and Gynecology, Tufts University School of Medicine, Boston, MA 02111; ^kDepartment of Molecular Microbiology and Immunology, Oregon Health & Science University, Portland, OR 97273; ^lKnight Cancer Institute, Oregon Health & Science University, Portland, OR 97201; and ^mYoung Women's Breast Cancer Translational Program, University of Colorado Anschutz Medical Campus, Aurora, CO 80045

Edited by David D. Moore, University of California, Berkeley, CA, and approved October 14, 2021 (received for review April 16, 2021)

During pregnancy, the rodent liver undergoes hepatocyte proliferation and increases in size, followed by weaning-induced involution via hepatocyte cell death and stromal remodeling, creating a prometastatic niche. These data suggest a mechanism for increased liver metastasis in breast cancer patients with recent childbirth. It is unknown whether the human liver changes in size and function during pregnancy and weaning. In this study, abdominal imaging was obtained in healthy women at early and late pregnancy and postwean. During pregnancy time points, glucose production and utilization and circulating bile acids were measured. Independently of weight gain, most women's livers increased in size with pregnancy, then returned to baseline postwean. Putative roles for bile acids in liver growth and regression were observed. Together, the data support the hypothesis that the human liver is regulated by reproductive state with growth during pregnancy and volume loss postwean. These findings have implications for sex-specific liver diseases and for breast cancer outcomes.

liver | pregnancy | bile acids | maternal health

Sex-specific differences in liver disease have been attributed to sexual dimorphisms in steroid production, metabolic enzymes, and behavior patterns (1). Whether a pregnancy cycle contributes to sex-specific liver disease remains largely unexplored; however, a previously unrecognized liver biology linked to reproductive status has been reported in rodents (2). This rodent study found that during pregnancy and lactation, hepatocytes proliferated and entered a higher anabolic state accompanied by an overall increase in liver size. Upon weaning, hepatocytes rapidly underwent programmed cell death, liver metabolism shifted toward catabolism, and the liver regressed to its pre-pregnant size in a process referred to as weaning-induced liver involution (2). In mice, liver involution promoted breast cancer outgrowth in the liver, suggesting a pathophysiological consequence of liver involution (2, 3).

Notably, young women diagnosed with breast cancer in the postpartum period were found to be at increased risk for liver metastasis (2). Taken together, these findings suggest that weaning-induced liver involution, which we predict would return the enlarged liver to its prepregnant, prelactational state, may create a prometastatic liver niche in women. However, it is unknown whether the human liver changes in size and function across a reproductive cycle, as expected if the liver is tuned to meet the unique metabolic demands of pregnancy, lactation, and weaning. Such evidence would corroborate findings in rodents and would be foundationally important for future studies of liver health in women.

To investigate the impact of reproductive state on liver size and function in women, we conducted a prospective study of healthy pregnant women using magnetic resonance and spectroscopy imaging of the liver and compared findings to a validated rodent model. Here, we show that the human female liver is regulated in both size and function by reproductive state and provide evidence of weaning-induced liver involution in humans. Furthermore, our data provide a hypothesis to explain the increased liver metastasis observed in postpartum breast cancer patients, as well as having potentially broader implications for the understanding of sex-specific liver diseases.

Results

In total, 47 healthy pregnant women completed early (12 to 16 wk gestation) and late pregnancy (32 to 36 wk gestation) study visits (Fig. 1A). Study participants underwent liver MRI (Fig. 1A), provided blood samples, had insulin sensitivity assessed via hyperinsulinemic-euglycemic clamp, and completed body composition analyses. Participant demographics are shown in *SI Appendix, Table S1*.

To assess whether liver size is increased during pregnancy, we measured livers via MRI and found that, on average, liver volumes increased 15% ($182 \text{ cm}^3 \pm 197 \text{ cm}^3$) from early to late

Significance

These human data are consistent with reproductive control of liver size and function in women and concur with recent observations in rodents, suggesting a conserved liver biology. The question of whether this described liver biology has implications for maternal health during pregnancy or sex-specific risk for liver disease remains to be determined. However, our evidence suggestive of weaning-induced liver involution in women may lead to improved understanding of the high rates of liver metastasis observed in young postpartum breast cancer patients.

Author contributions: A.Q.B., K.K.V., J.Q.P., E.G., A.R.G., and P.S. designed research; A.Q.B., K.K.V., J.Q.P., E.G., X.G., A.D., E.B., W.R., W.N., and P.S. performed research; A.D. contributed new reagents/analytic tools; A.Q.B., K.K.V., J.Q.P., M.F., E.G., X.G., M.C.L., E.B., P.C., Z.X., and P.S. analyzed data; and A.Q.B. and P.S. wrote the paper.

The authors declare no competing interest.

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This article contains supporting information online at <http://www.pnas.org/lookup/suppl/doi:10.1073/pnas.2107269118/-DCSupplemental>.

Published November 22, 2021.

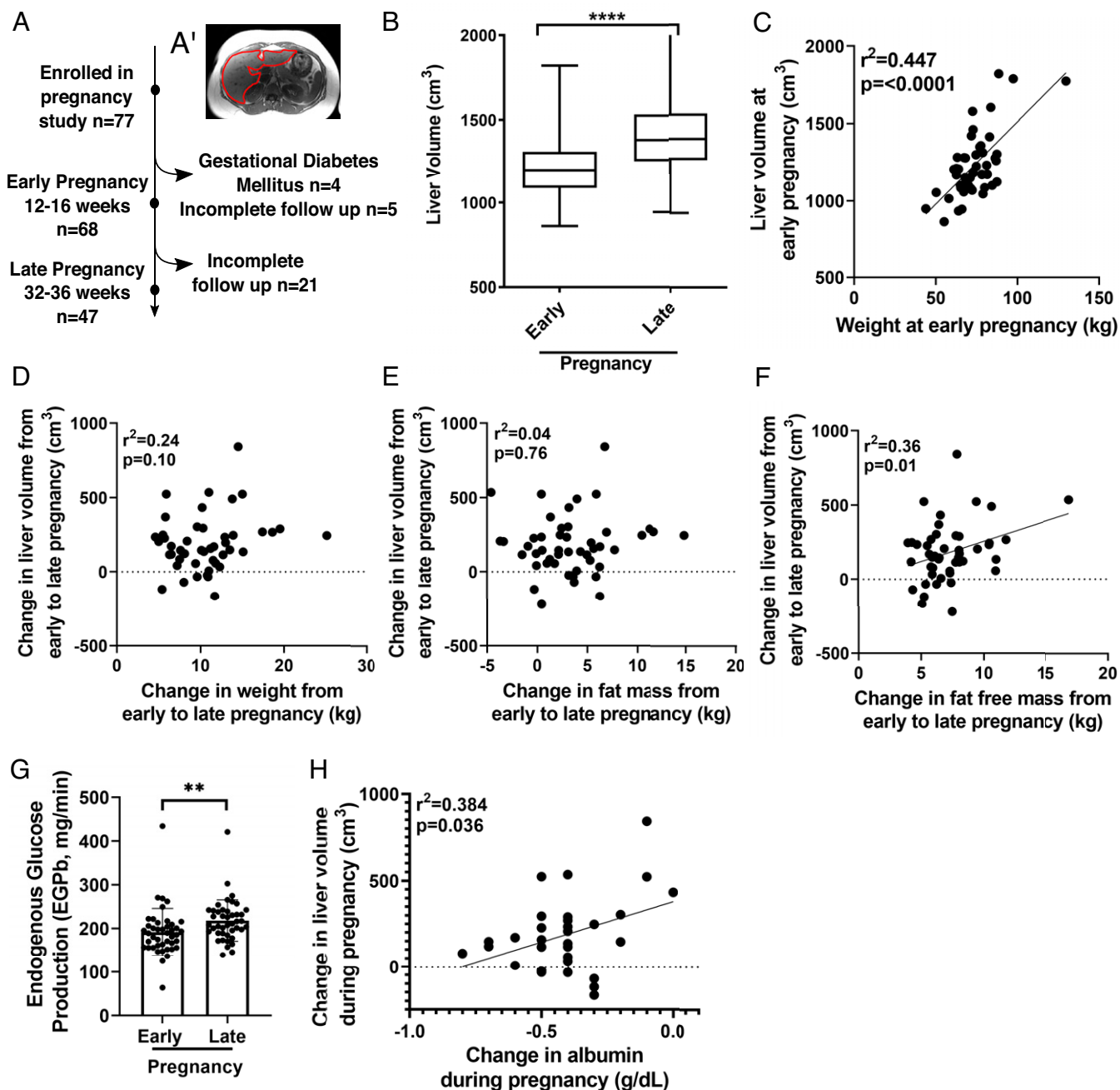


Fig. 1. Liver changes during pregnancy. (A) Diagram of the observational study. (A') Liver MRI cross-section with liver outlined in red. (B) Average liver volume at early and late pregnancy ($n = 47$; **** $P < 0.0001$ by two-tailed paired t test). (C) Pearson's correlation of liver volume and BMI at early pregnancy ($n = 47$). Pearson's correlation of change in liver volume with change in weight (D), fat mass (E), and fat-free mass (F) ($n = 47$). (G) EGP-b at early and late pregnancy ($n = 47$; ** $P < 0.01$ by two-tailed paired t test). (H) Pearson's correlation of change in liver volume and change in albumin ($n = 30$).

pregnancy ($P < 0.0001$) (Fig. 1B). Average liver size at early pregnancy was $1,239 \text{ cm}^3 \pm 220.8 \text{ cm}^3$ and at late pregnancy was $1,421 \text{ cm}^3 \pm 298.6 \text{ cm}^3$ (Fig. 1B).

Because liver size is attuned to overall body size via the “hepatostat” (4), we next determined whether the increase in liver size from early to late pregnancy correlated with increased body mass of pregnancy. First, we investigated the existence of the “hepatostat” at baseline, using body weight at the early pregnancy visit as a baseline surrogate, as pregnancy-related weight gain is minimal at this time point (5). Liver volume at

early pregnancy correlated with body weight (Fig. 1C), confirming previous studies in nonpregnant individuals (4). In contrast, the change in liver volume during pregnancy did not correlate with gestational weight gain (Fig. 1D). Furthermore, we found no relationships between pregnancy liver volume change and change in total fat mass (Fig. 1E), subcutaneous abdominal, or visceral adipose tissue (VAT) volumes (Table 1). However, the change in a woman's fat-free mass, which includes liver, fetal tissue, placenta, and plasma, did correlate with change in liver size (Fig. 1F). The association between change in fat-free mass and

Table 1. Change in liver volume correlated with measures of body composition and metabolism

Variable	Mechanism of collection	Sample size	Pearson correlation coefficient	P value
Body composition				
Change in weight	Scale	47	0.260	0.078
Change in BMI	Scale, stadiometer	47	0.213	0.150
Change in fat mass	BODPOD	47	0.077	0.605
Change in fat-free mass	BODPOD	47	0.335	0.021
Change in SAT	MRI	47	0.123	0.409
Change in VAT	MRI	47	0.245	0.097
Change in IHL	H-MR spectroscopy	47	−0.035*	0.814*
Metabolism				
Change in M value	Hyperinsulinemic-euglycemic clamp	43	−0.015	0.926
Change in EGP	Hyperinsulinemic-euglycemic clamp	43	−0.047*	0.763*
Change in Rd	Hyperinsulinemic-euglycemic clamp	43	0.053	0.736
Change in fasting insulin	Blood draw	45	0.095	0.537
Change in total cholesterol	Blood draw	45	0.062	0.684
Change in triglycerides	Blood draw	45	0.176	0.248
Change in LDL	Blood draw	45	−0.119	0.438
Change in HDL	Blood draw	45	0.103	0.500
Change in very low density lipoprotein	Blood draw	45	−0.103	0.500

Bold text indicates that the change fat free mass was the only variable that reached statistical significance.

*These analyses were done with Spearman Correlation.

liver volume is confounded as fat-free mass is not an independent variable from liver mass. In sum, these data suggest that liver size increase during pregnancy is unlinked to overall body size; that is, it is not controlled by the “hepatostat” mechanism. Rather, these data may reflect an unrecognized, reproductive state-controlled program regulating liver size during pregnancy.

We next asked if metabolic measures were associated with liver volume change and found no relationship with cholesterol concentrations or with measures of insulin sensitivity, that is, endogenous glucose production (EGP) and glucose disposal rate (Rd) (Table 1). We also found no relationship between change in liver volume and change in intrahepatic lipid (IHL) content (Table 1). Assessment of IHL content in rodents also showed no change in IHL during pregnancy (*SI Appendix, Fig. S1*). In sum, we observed that the increase in human liver volume with late pregnancy occurred independent of weight gain of pregnancy, various other measures of body composition, circulating metabolites, and IHL storage.

In rodents, hepatocyte proliferation contributes to increased liver size and metabolic output during pregnancy and lactation (2, 6). Obtaining timed liver biopsies would be the most direct way to investigate hepatocyte proliferation during pregnancy in women; however, liver biopsies were not performed in our study for participant safety. Thus, we indirectly assessed for increased hepatocyte number by evaluating hepatocyte function. We found evidence for increased liver output as measured by increases in EGP (Fig. 1G) and serum albumin concentration (Fig. 1H), two surrogates of liver function (7, 8). Of note, an additional contributor to increased liver volume during pregnancy is increased blood flow, which rises ~50% by late pregnancy (9). However, increased blood flow during pregnancy is not reported to associate with elevated hepatocyte metabolic output. In sum, these data are consistent with an increase in liver size and synthetic capacity during pregnancy, which may be due to increased hepatocyte proliferation as observed in rodents. Additional studies are required to determine if hepatocyte proliferation is increased during pregnancy in women.

We next looked for evidence of weaning-induced liver involution in women, a biology not previously described in humans. Of the 47 women who participated in our pregnancy study,

36% completed a liver MRI >3 mo postwean (median 5.7 mo) (Fig. 2A). Liver volumes trended toward a decrease in size between late pregnancy and postwean (Fig. 2B), and postwean liver volumes were similar to early pregnancy, indicative of a return to baseline (Fig. 2C). These data provide evidence of postpartum liver involution in women.

While our data showed a statistically significant increase in liver size during pregnancy and a trend toward decrease after weaning, there was heterogeneity in how an individual’s liver size changed with pregnancy and postwean. During pregnancy, we found that 72% (34/47) of women had an average increase in liver volume of ~20% (Fig. 2D, black bars). However, 21% of participants (10/47) had no measurable liver volume change and 6% (3/47) had a reduction in liver volume (Fig. 2D, black bars, *SI Appendix, Table S2*). We saw similar heterogeneity with regard to liver volume change from late pregnancy to postwean (Fig. 2D, gray bars).

Considering the heterogeneity in liver volume change and what is known about normal rodent liver biology (i.e., liver weight gain with pregnancy and loss postwean) (2), we performed subgroup analyses. We delineated the participants into two groups: “gain-loss,” the observed pattern in the normal rodent, or “not gain-loss” for those that did not display the rodent pattern. 53% of women displayed the anticipated liver “gain-loss” pattern (Fig. 2E). The “not gain-loss” group comprised heterogeneous patterns and included three women who lost liver volume during pregnancy and regained postwean, three women with no significant liver volume changes, and one woman each with either continuous liver size loss or gain across the three visits (Fig. 2F). Of note, liver volume patterns with pregnancy and postwean did not correlate with a woman’s overall weight gain of pregnancy (*SI Appendix, Fig. S2*).

Upon further exploration, we found that none of the women whose liver changes were similar to the normal rodent pattern of “gain-loss” had gestational hypertension, yet 50% of the “not gain-loss” group did (Fig. 2F, dashed lines). Furthermore, measures of insulin sensitivity differed between these groups. Specifically, we found the “gain-loss” participants had greater EGP at late pregnancy (Fig. 2G), consistent with published data showing elevated EGP in healthy pregnancy (10). We also found greater glucose disposal rates at late pregnancy in the “gain-loss” group (Fig. 2H), consistent with greater insulin

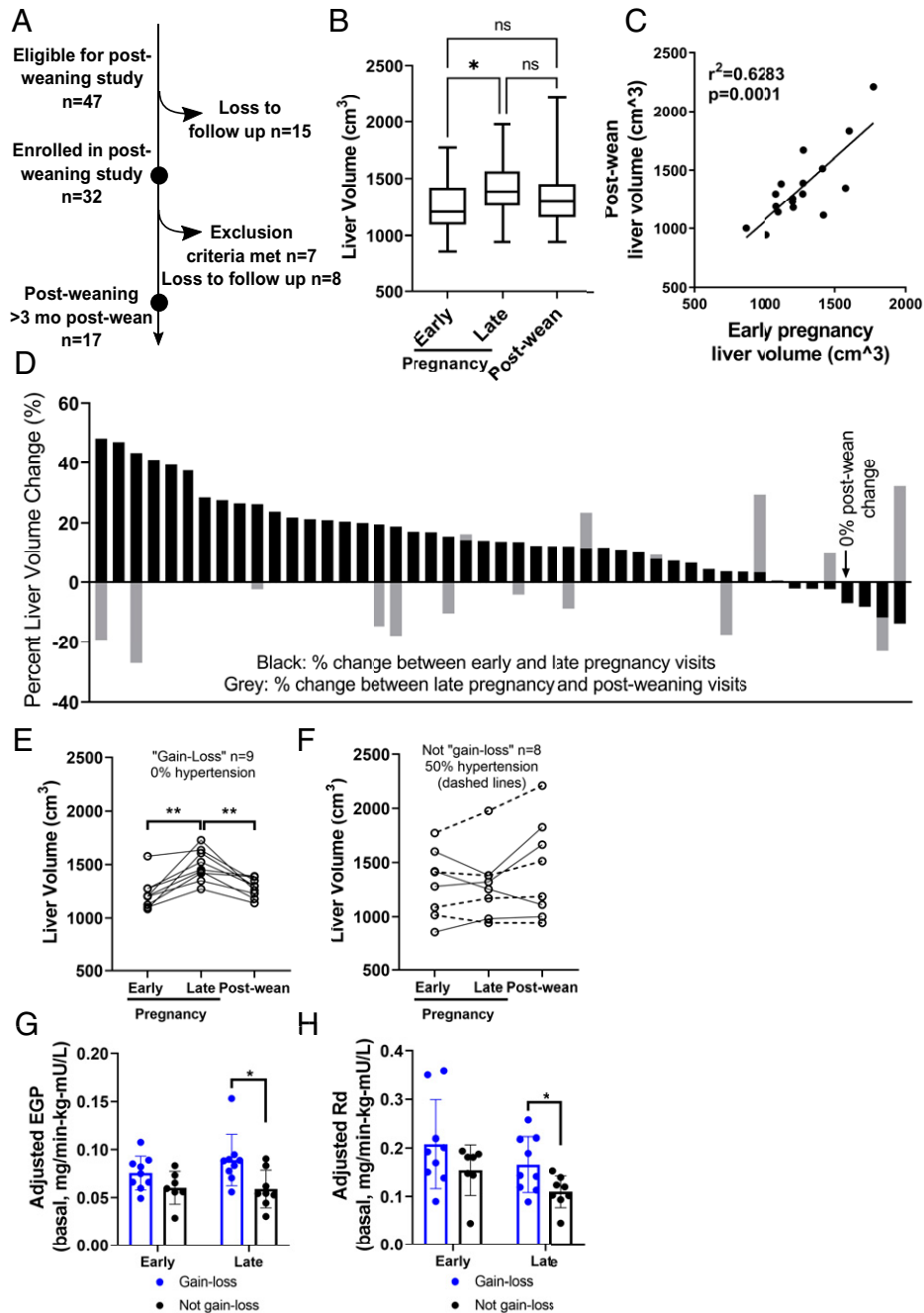


Fig. 2. Human liver volumes postwean. (A) Diagram for postwean observational study. (B) Liver volume at early, late, and postwean time points ($n = 17$). (C) Pearson's correlation of liver volumes at early pregnancy and postwean ($n = 17$). (D) Liver volume change between early and late pregnancy (black bars) and between late pregnancy and postwean (gray bars) per participant. Primary pattern (E) and secondary patterns (F) of liver volume change with pregnancy and postwean. Dashed lines show participants with hypertension (paired t test). EGP (G) and glucose disposal rate, Rd, (H) in women in gain-loss group compared to women not in gain-loss group. Pearson's correlation. P value: * < 0.05 , ** < 0.01 .

sensitivity in the muscle. These data suggest that the “not gain-loss” pattern may be associated with suboptimal gestational metabolic health and gestational hypertension. One question is whether these metabolic parameters impact fetal outcomes. In this cohort, maternal liver size patterns did not correlate with newborn weight, length, or Ponderal index, three common neonatal health measures.

To investigate the mechanistic relationship between reproductive state and liver size, we utilized a rat model, as previously described (2). We found liver weight increases during pregnancy were greater than expected due to gestational weight

gain alone (Fig. 3A). These data suggest rat liver weight during pregnancy is unlinked from the “hepatostat,” corroborating our human data (Fig. 1D). Next, we confirmed maximum hepatocyte proliferation in the rat livers to occur during pregnancy (Fig. 3B), consistent with previous reports (2, 6). Together, these data suggest a physiological model in which increased liver volume of pregnancy is due to increased hepatocyte proliferation that is activated via an unrecognized, pregnancy-mediated developmental program.

As a possible mechanism underlying a pregnancy-associated liver growth program, we investigated bile acid metabolism in

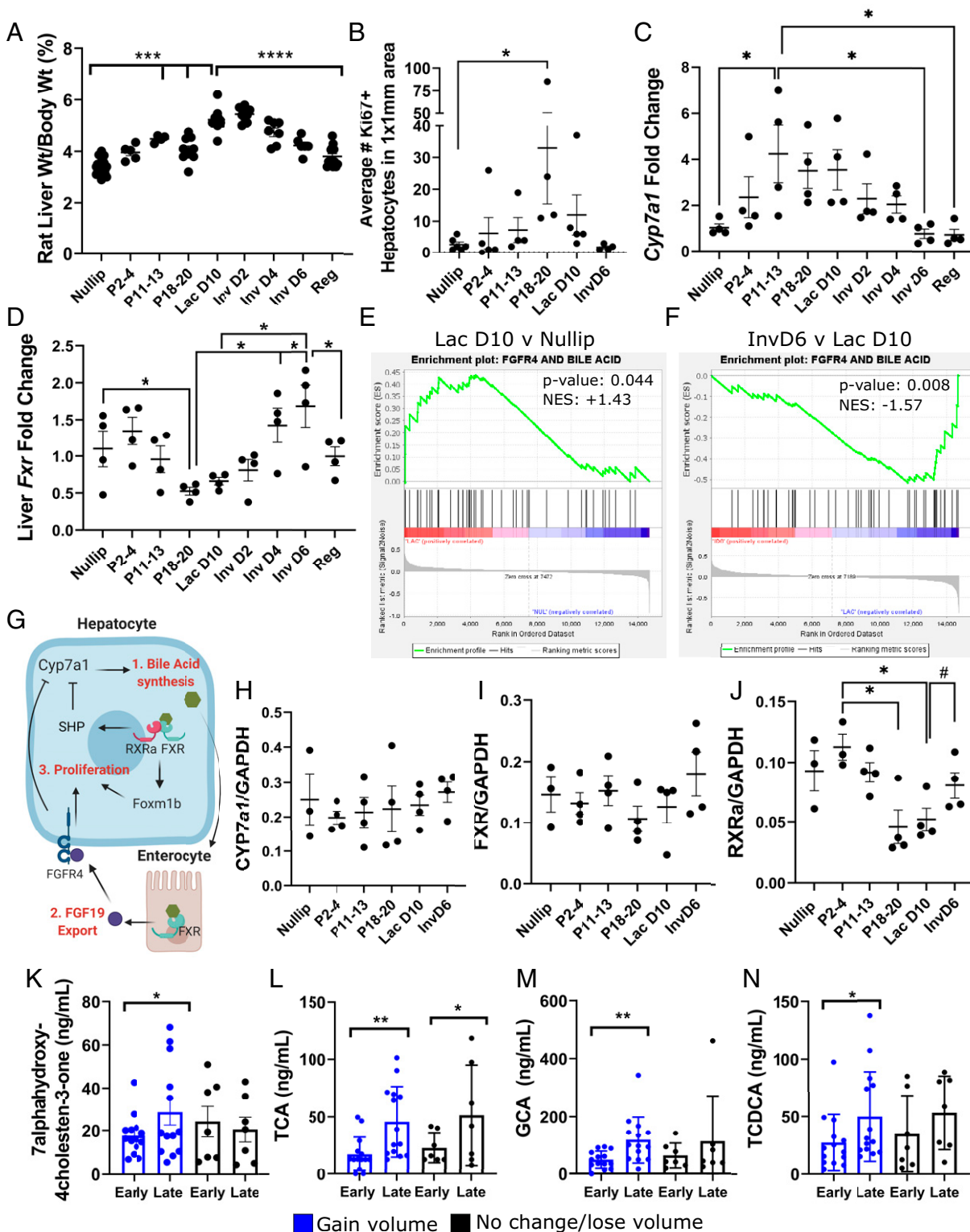


Fig. 3. Hepatic bile acid signaling and liver size. (A) Rat liver weight normalized to body weight: nulliparous (nullip) $n = 24$; early (P2-4) $n = 5$; middle (P11-13) $n = 4$; and late (P18-20) pregnancy $n = 10$; lactation day 10 (Lac D10) $n = 9$; involution (Inv) D2 $n = 9$; InvD4 $n = 7$; InvD6 $n = 6$; Regressed (Reg) $n = 14$; one-way ANOVA. (B) Ki67+ hepatocytes in rat livers, $n = 3$ to 5/group. (C) *Cyp7a1* and (D) *Fxr* mRNA fold change in liver, $n = 4$ per group; one-way ANOVA. Gene set enrichment analysis plots of FGFR4-bile acid gene pathway for (E) lactation day 10 versus nulliparous groups and (F) involution day 6 versus lactation day 10 groups. (G) Model for pregnancy enlargement of liver due to hepatocyte proliferation downstream of bile acid signaling. Protein abundance in whole rat liver of (H) CYP7A1, (I) FXR, and (J) RXR α . Data are normalized to GAPDH protein abundance; nullip $n = 3$, P2-4, P11-13, P18-20, Lac D10, and InvD6 $n = 4$ /group; * $P < 0.05$ by one-way ANOVA; # $P < 0.05$ by Student's t test. (K) Human 7 α -hydroxy-4cholesten-3-one plasma concentrations at early and late pregnancy, separated by liver gain ($n = 14$) and no gain ($n = 7$). Human plasma concentrations of bile acids TCA (L), GCA (M), and TCDCa (N) paired t test, P value: * < 0.05 , ** < 0.01 , *** < 0.001 , **** < 0.0001 .

rodents. Bile acid signaling contributes to liver regeneration following partial hepatectomy and can control liver size independent of body size (11, 12). Furthermore, bile acids have been shown to regulate hepatocyte proliferation in a pathway dependent on enterocyte-derived fibroblast growth factor 15/19 (11, 13, 14). To investigate if the bile acid pool is modulated by reproductive state, we measured liver *Cyp7a1*, the gene that encodes rate limiting enzyme in bile acid synthesis. We found a three- to fourfold increased expression of *Cyp7a1* with pregnancy, which remained elevated during lactation, followed by a rapid decline with weaning (Fig. 3C). Since hepatic FXR signaling acts as a negative regulator of *Cyp7a* (Fig. 3D), we measured hepatic *Fxr*. We found *Fxr* was down-regulated during late pregnancy, when *Cyp7a1* was high, and increased with weaning, when *Cyp7a1* was low (Fig. 3D). To further investigate the hypothesis that elevated bile acids contribute to hepatocyte proliferation through activation of FGFR4 signaling, we analyzed an RNA-sequencing dataset from mouse liver at various reproductive stages. We performed gene set enrichment analysis using a custom gene set composed of genes in bile acid metabolism and FGFR4 downstream signaling pathways. Consistent with our hypothesis, we found an enrichment for bile acid-FGFR4 signaling pathways in lactation stage mice compared to nulliparous (Fig. 3E and *SI Appendix, Fig. S3*). Additionally, at involution day 6 this bile acid-FGFR4 gene signature was significantly decreased compared to lactation (Fig. 3F and *SI Appendix, Fig. S3*). Furthermore, we identified the up-regulation of individual genes involved with bile acid synthesis (*Cyp7a1*, *Cyp8b1*) and proliferation (*Foxm1*) during lactation, which were subsequently down-regulated with involution (*SI Appendix, Fig. S3*). Conversely, genes involved with inhibiting bile acid synthesis (*Rxra*, *Shp*) were reduced during lactation and elevated during involution (*SI Appendix, Fig. S3*). A model depicting a proposed mechanism by which bile acid-FGFR4 signaling increases hepatocyte proliferation is shown (Fig. 3G). Because gene expression and protein concentration are not always linked, we evaluated protein abundance for CYP7A1 and FXR. We were unable to validate reproductive-stage regulation of *Cyp7a1* and *FXR* at the protein level (Fig. 3H and I). However, based on our mouse RNA-sequencing data that showed regulation of *RXRa* by reproductive stage (*SI Appendix, Fig. S3*), we also measured *RXRa* protein concentration. *RXRa* is a known binding partner for FXR that acts as a co-negative regulator of *Cyp7a1* (15, 16). We found that *RXRa* was significantly reduced at pregnancy days 18 to 20 and lactation day 10, then was increased at involution day 6 (Fig. 3J). The decreased gene expression and protein abundance of *RXRa* might be sufficient to activate *Cyp7A1* gene expression without a corresponding decrease in FXR protein. In sum, these data associate increased bile acids with the physiologic expansion of the liver during pregnancy—consistent with a previous report (12)—and extend these observations to suggest a role for bile acids in regulating liver size during pregnancy, lactation, and weaning.

We then examined associations between liver growth and the bile acid pool in pregnant women by measuring a biomarker of bile acid production and serum bile acid concentrations at early and late pregnancy. Serum concentrations of 7 α -hydroxy-4-cholestene-3-one (7 α C4), a readout for cholesterol 7 α -hydroxylase (*Cyp7a1*) enzyme activity (17), were significantly increased at late compared to early pregnancy only in the women who had an increase in liver volume during pregnancy (Fig. 3F). This finding supports the hypothesis that increased bile acid production during pregnancy may be required for liver size increase. Furthermore, among the women whose liver increased in size during pregnancy, we found increases in several primary bile acids and their conjugates (Fig. 3G–I). Of note, changes in secondary bile acids, which are metabolic products of gut bacteria, only weakly

correlated with liver volume change (*SI Appendix, Table S3*). In this human cohort, we did not find associations between concentration of plasma FGF19, an enterocyte product shown to induce hepatocyte proliferation and liver growth in rodents (*SI Appendix, Fig. S4*). One potential caveat to our FGF19 analysis is that plasma concentrations of FGF19 may not reflect concentration in the portal vein that links the gut and liver. In sum, these human data are consistent with an increased bile acid pool during pregnancy, which may contribute to the increased liver size observed in pregnancy.

Discussion

In this study, we find evidence for a previously unreported liver biology in women—namely that during the course of a normal pregnancy cycle liver volume increases during gestation and decreases postwean. Multiple observations and reports demonstrate tight control of liver size in proportion to body size, a phenomenon that has been referred to as the “hepatostat” (4). Yet, in our study, increased liver volume during pregnancy is not accounted for by change in body size. Since liver size is presumed to be directly linked to a physiologic function of the liver (18), our data suggest that a metabolic demand above and beyond body size accounts for increased liver size during pregnancy. Consistent with this hypothesis, in rodents we find that liver size remains elevated through lactation even though body size is reduced compared to late pregnancy.

One potential mechanism controlling liver size during a pregnancy cycle is the circulation of bile acids, which have been shown to modulate liver size independent of the body size hepatostat (19, 20). Such a paradigm where liver size is controlled by bile acid flux would allow for a situation in which body size could become “unlinked” from liver size. The data presented here obtained from rodent models and human correlates support the idea that increased production of primary bile acids during pregnancy and lactation via the *Cyp7a1* synthetic pathway leads to hepatocyte proliferation and thus a larger liver. Since the primary function of bile acids is to facilitate fat absorption in the intestine, an increase in liver parenchyma capable of bile acid synthesis would make sense given increased caloric demand during pregnancy and lactation. In sum, our human data are consistent with dynamic size regulation of the liver to accommodate the unique metabolic demands of pregnancy and lactation. Furthermore, our combined human and rodent data suggest a mechanism whereby physiologically regulated bile acid synthesis underlies liver size changes across a pregnancy cycle. Future studies directly testing this mechanism are needed.

Additional factors that could contribute to increased liver size with pregnancy include hormones, such as thyroid hormone, estrogen, and progesterone. Thyroid hormone can induce hepatocyte proliferation (21), yet thyroid hormone does not associate with significant growth of the liver nor is it required for liver regeneration (22). Furthermore, thyroid hormone is known to decrease between early and late pregnancy (23, 24), which is a pattern of expression inconsistent with a role in hepatocyte proliferation. Additionally, previous work has evaluated the impact of estrogen and progesterone on liver size. Administration of pregnancy-relevant concentrations of estrogen and progesterone did not induce liver growth, suggestive that these hormones are not key for increasing liver size during pregnancy (25). Prolactin, which increases through pregnancy and peaks during lactation (26, 27), is known to stimulate hepatocyte proliferation and is associated with accelerated liver regeneration following partial hepatectomy (28–30). A limitation of our study is lack of identification of the molecular mechanism upstream of increased bile acid production during pregnancy and lactation, of which pregnancy hormones could contribute.

Intriguingly, there is evidence that prolactin can stimulate bile acid synthesis (31). These prolactin data and our observation that RXR α , a member of the steroid receptor super family, is regulated in a manner consistent with bile acid regulation of liver size, may generate new avenues to pursue.

While the pattern of liver size gain with pregnancy and loss upon weaning was observed in the majority of women in this pregnancy study, we also identified a subset of women for whom liver size did not follow a “gain-loss” pattern. Gestational hypertension and reduced liver insulin sensitivity were more common in this subset. Therefore, an intriguing hypothesis is that a facet of healthy pregnancy is the gain-loss pattern of liver volume. A corollary to this hypothesis is that preexisting and/or pregnancy-specific conditions such as gestational hypertension could underlie the inability of the liver to appropriately respond to pregnancy (32). Of note, gestational hypertension affects 5 to 10% of pregnancies and can progress to preeclampsia with known pathogenic liver involvement in ~15 to 20% of cases (33). Gestational hypertension is attributed to a vascular disorder that is initiated at the placental interface, specifically due to incomplete maturation of the maternal spiral arteries (34). If related, the question of whether the same pathophysiology that leads to gestational hypertension can also impair the normal liver response to growth cues of pregnancy, or vice versa, remains unknown. On the other hand, albeit a small study, our data show that liver gain with pregnancy is not a requisite for normal fetal growth, as we observed no differences in newborn size between women who did and did not experience liver gain with pregnancy. Future studies would be required to determine if there are any long-term impacts due to a lack of maternal liver growth during gestation, as has been described for other variations in neonatal nutrition and lifetime risk of disease (35–37).

The data presented here show that the human liver responds to a pregnancy cycle in a similar manner to rodents, namely, increased size with pregnancy and lactation, followed by a decrease in size postwean. The process that returns the rodent liver to its pre-pregnant state, weaning-induced liver involution, promotes breast cancer metastasis to the liver (2, 3). Given that women have an increased risk of liver metastasis if diagnosed with breast cancer within 5 y of pregnancy, we speculate that weaning-induced liver involution creates a prometastatic micro-environment in the liver. Although we cannot definitively demonstrate liver involution in humans, it has recently been demonstrated that the breast undergoes weaning-induced involution in women similar to rodents (38, 39). Therefore, we theorize that there is a conserved mammalian developmental program that links the mammary gland and the liver through a pregnancy cycle, putatively to meet the elevated metabolic demands of pregnancy and lactation. The potential importance of this biology for supporting reproduction and infant health are apparent; however, weaning-induced breast and liver involution may have unanticipated consequences, including the transient increased risk of breast cancer and liver metastasis (2, 40–43). Additional impacts on risk for liver disease may be anticipated given known disparities in liver disease by sex, including increased risk for acute liver failure and autoimmune liver conditions in women (1).

A key strength of our study is that each woman serves as her own control, allowing us to see how an individual’s liver changes during a pregnancy cycle. However, our human cohort study cannot draw mechanistic conclusions because it was purely observational. Additionally, these data were generated in a small, predominately White, non-Hispanic cohort and require validation in a larger study with a diverse population to generalize these findings.

In summary, this work describes an observation in normal women, specifically increased liver size with pregnancy and

decreased size postwean, putatively to accommodate the dramatic changes in metabolic demands across a pregnancy–lactation–wean cycle. These findings demonstrate reproductive control of liver size and function in women and concur with recent observations in rodents, suggesting a conserved liver biology. The question of whether this described liver biology has implications for maternal health during pregnancy or sex-specific risk for liver disease remains to be determined (1, 44, 45). However, our evidence suggest weaning-induced liver involution in women, which if validated, may lead to improved understanding of the high rates of liver metastasis observed in young postpartum breast cancer patients (2).

Materials and Methods

Recruitment.

Prospective cohort. We conducted a prospective cohort study of pregnant women receiving care at Kaiser Permanente Northwest (KPNW) or Oregon Health & Science University (OHSU). All study activities were approved by Institutional Review Boards at KPNW (no. 3993) and OHSU (no. 10438 and 15264). Recruitment started in December 2014 and was completed in August 2017. KPNW members who met the study inclusion criteria were identified weekly using the electronic health record (EHR). Eligible participants were mailed a recruitment letter and received a follow-up phone call a week later. During this telephone call, study personnel conducted additional eligibility screening and scheduled an explanatory visit. If patients consented at the explanatory visit, this was followed by two visits between 12 to 16 wk of gestation and two visits between 32 to 36 wk of gestation. Participants who completed all pregnancy study visits had the option to complete a postwean study visit, between 3–12 mo after weaning.

Inclusion/exclusion criteria. Patients were eligible for the study if they were between 18 to 45 y of age, were less than 12 wk pregnant with a singleton gestation at time of enrollment, had a body mass index (BMI) between 18.5 kg/m² and 38 kg/m², and were fluent English speakers. Participants were excluded if they had any of the following conditions or symptoms: contraindications to MRI study (e.g., claustrophobia, metal implants); pregestational diabetes; gestational diabetes; history of bariatric surgery or other medical conditions requiring specialized nutritional care; anemia; current history of drug, tobacco, or alcohol use; maternal rheumatologic or chronic inflammatory state; or chronic hypertension.

Measures. Data for this paper were collected at three study visits: one at 12 to 16 wk of gestation, one at 32 to 37 wk of gestation, and one between 3–12 mo postwean. Height was measured at the first visit to allow for calculation of BMI; weight was measured using a calibrated scale at each visit. Demographic variables, including parity and preconception BMI, were extracted from the EHR.

Air Displacement Plethysmography. Air displacement plethysmography (BOD POD, COSMED USA, Inc.) was used to determine participants’ fat mass, fat-free mass, and percent body fat at each visit. Participants first changed into a bathing suit or spandex clothing and a swimming cap. They then sat inside the BOD POD while the air displaced by the body was measured. Results included total mass and body density. Fat mass and fat-free mass were estimated using van Raaij’s pregnancy equations to account for changes in the density of fat-free mass during pregnancy (47, 48).

MRI Acquisition. MRI and spectroscopy data were collected using a Siemens Prisma Fit 3T whole-body system (Siemens Healthineers) at the Advanced Imaging Research Center at OHSU. Abdominal MR data were acquired in two stations, the first centered at umbilicus and the second centered over the xyphoid process, to acquire MRI and liver magnetic resonance spectroscopy (MRS). Siemens flexible 18-channel array and spine array receiver coils with body-coil transmission were used. The abdominal MRI protocol included a T1-weighted gradient-echo sequence (TE = 2.5 ms, TR = 140 ms, flip-angle = 90°, (1.25 mm)² in plane resolution, 30 slices with 6 mm thickness) acquired in two-breath holds of ~18 s each. The liver T1-weighted MRI protocol was acquired with identical parameters to the abdominal T1 volume but with a variable number of slices to cover the entire extent of the liver.

MRI Processing. The T1-weighted MRI data sets of abdomen and liver were manually spliced together with affine transformations and overlapping slice elimination. The top of the liver and the L-4/5 intervertebral disk were identified as the upper and lower bounds, respectively, for the segmentation analysis for abdominal visceral and subcutaneous fat volumes.

Abdominal T1-w MRI volumes were segmented into five classes: unlabeled, subcutaneous adipose tissue (SAT), VAT, muscle, and organ (including all other abdominal volume). A custom Python pipeline was used to create an initial automated segmentation using inputs from the umbilicus T1-weighted volumes, the liver T1-weighted volumes, and an 11-slice manual segmentation label map, the merged T1-weighted MRI data set, and the affine transforms that map individual volume acquisitions to the merged image space. Manually generated uterus/placenta and liver masks were created as these two regions have high rates of false positives for classification as adipose tissue.

Processing within the pipeline made use of the following Python libraries: Nipype (49), the Advanced Normalization Tools (50), the Insight Toolkit (51), Scikit-image (52), Scikit-learn (53), and SciPy (54). Following N4 bias field correction, steps in the segmentation pipeline relied upon intensity thresholding and morphological operations. The muscle mask was generated with a compact watershed algorithm seeded with the muscle mask from the 11-slice segmentation. SAT masking made use of the geodesic active contours algorithm (55), coupled with dilation and erosion steps to distinguish the SAT from internal VAT. VAT was taken as the difference between the total adipose mask and the SAT mask. Segmentation masks output from the automated pipeline subsequently underwent slice-by-slice manual review followed by manual refinement by a single analyst (J.Q.P.) using the 3D Slicer software package to ensure accuracy of VAT and SAT masks placements.

Liver segmentation was manually conducted separately using the OsiriX and Image J software programs.

Liver Volume Determination. Image analysis was performed using OsiriX (OsiriX Imaging Software) software and Image J software (NIH). For volume estimation, 3D-VIBE (a T1-weighted FLASH technique with fat selective pre-pulse) sequences were used. The liver was identified on each image, and the outline of the liver tissue annotated by freehand region of interest estimation by operators trained by a body radiologist with over 10 y of experience in MRI of the liver. This allowed for the generation of a liver area on each slice. Liver volume was calculated by multiplying the estimated area of each slice by the interval between slices, summing all volumes containing liver for the total liver volume (46).

Liver volume determinations were performed by two blinded operators. Operators independently measured liver volumes for five cases with two MRI scans per case (early and late pregnancy). The observed interoperator variability (*SI Appendix, Table S2*) was used to benchmark values that are within the range of measurement error, in this case +7 to -7%.

MRS. IHL was measured using 1H single-voxel MRS, following MRI. Liver MRS voxels were positioned within the right lobe with voxel sizes ranging from 18 to 24 cm³.

Liver spectra were collected using a point-resolved spectroscopy single-voxel spectroscopy sequence (TR = 5 s, TE = 30 ms, 1,024 points, 2,000 Hz spectral width). The long repetition time ensured fully relaxed water signal (99.2%), because it serves as an internal standard for quantification. Three separately acquired MRS series were run, each during a 10-s breath hold.

MRS analysis was conducted using the advanced method for accurate, robust, and efficient spectral time-domain fitting module within the jMRUI software program. All spectral fits were inspected and rerun with additional constraints if fitting contained errors. IHL is expressed as a proportion of primary lipid peak to water peak areas.

Hyperinsulinemic-Euglycemic Clamp. Hyperinsulinemic-euglycemic clamp with coinfusion of [6,6-2H₂] glucose was used to determine whole-body and skeletal muscle insulin sensitivity (Rd) and EGP (56, 57). Subjects were advised regarding a standard diet consisting of 30% of total calories from fat sources, 15% from protein, and 55% from carbohydrates for the 3 d before study. Following an 11-h overnight fast, subjects were admitted to the OHSU Clinical and Translational Research Center where a hyperinsulinemic-euglycemic clamp was performed. At 0600, an intravenous catheter was placed in one arm for infusions and in the contralateral hand for blood withdrawal and warmed to 70°C using a warming mitt for sampling of arterialized venous blood. A primed constant infusion of [6,6-2H₂] glucose (Cambridge Isotope Laboratories) was infused at 0.133 mL/min and an enrichment intended to achieve ~1.0 mol percent excess for all subjects. The basal infusion of [6,6-2H₂] glucose was continued for 2 h, and plasma samples were obtained from 90 to 120 min to estimate basal EGP and fasting insulin concentration. Basal EGP was calculated according to the steady-state equations of Steele (58). At the completion of the 2-h infusion glucose isotope, a primed, constant infusion of regular insulin at 40 mU/m²/min was started. Plasma glucose was maintained at 90 mg/dL for the remaining 2 h. During the final 30 min of the clamp, blood samples were obtained every 5 min for isotope analysis. Suppression of EGP by

insulin infusion during the 2-h clamp was estimated using the method developed by Black (59). EGP, Rd, and M value were adjusted for insulin level (mU/L) and fat-free mass (kg).

Labs. Venipuncture was used to obtain blood samples with participants in the fasting state. The following measures were assessed and run in the Laboratory Core of the Oregon Clinical and Translation Research Institute: comprehensive metabolic panel, lipid panel, free fatty acids, liver function tests, glucose, and insulin. Insulin was assessed by radioimmunoassay (Mercodia AB) and glucose by a Hexokinase based colorimetric assay (Stanbio laboratory).

Bile Acid Profiling. Bile acid profiling was performed in the OHSU Bioanalytical Shared Resource/Pharmacokinetics Core. Plasma samples from early and late pregnancy were utilized to quantify plasma bile acids and 7 α -hydroxy-4-cholesten-3-one using liquid chromatography-tandem mass spectrometry (LC-MS/MS). Quantification of plasma bile acids was performed with a 4000 QTRAP hybrid triple quadrupole-linear ion trap mass spectrometer (SCIEX) operating with electrospray ionization (ESI) in the negative mode. The mass spectrometer was interfaced to a Shimadzu high-performance liquid chromatography (HPLC) system consisting of SIL-20AC XR auto-sampler and LC-20AD XR LC pumps. Analyte separation was achieved using a gradient HPLC method and Luna 2.5 μ C₁₈ (2)-HST 50 \times 2 mm column (Phenomenex) kept at 50°C with a Shimadzu CTO-20AC column oven.

The stable isotope dilution LC-MS/MS method to quantify plasma bile acids was previously described (60). In brief, plasma was spiked with internal standards, and bile acids were measured following protein precipitation and extraction with methanol, centrifugation, and filtration of the supernatant. Calibrants were prepared in charcoal stripped matrix (SP1070 from Golden West Biological) using authentic bile acid and conjugate standards (obtained from Toronto Research Chemicals and Cerilliant).

Data were acquired using SCIEX Analyst 1.6.2 and analyzed using SCIEX MultiQuant 3.0.3 software. Sample values were calculated from calibration curves generated from the peak area ratio of the analyte to internal standard versus analyte concentration that was fit to a linear equation with 1/x weighting. The following bile acids were measured: Taurocholic acid (TCA), Glycocholic acid (GCA), Taurochenodeoxycholic acid (TCDC), Glycochenodeoxycholic acid (GCDC), Ursodeoxycholic acid (UDCA), Cholic acid (CA), Chenodeoxycholic acid (CDCA), Deoxycholic acid (DCA), and Lithocholic acid (LCA). Compounds were quantified with multiple reaction monitoring and transitions optimized by infusion of pure compounds.

Plasma 7 α -hydroxy-4-cholesten-3-one was determined by LC-MS/MS following protein precipitation and extraction with acetonitrile. To each 100 μ L sample of EDTA plasma was added 1 ng of internal standard 7 α -hydroxy-4-cholesten-3-one-d₇ (prepared at 0.2 ng/ μ L in methanol) and 300 μ L of acetonitrile. The samples were vortex mixed and centrifuged at 12,000 \times g for 10 min. The supernatant was removed and filtered prior to injection for analysis with LC-MS/MS.

Calibration standards were prepared across the range 1 to 100 ng/mL in charcoal stripped plasma SP1070 (Golden West Biological) using authentic 7 α -hydroxy-4-cholesten-3-one (obtained from Toronto Research Chemicals).

LC-MS/MS was performed using a 5500 QTRAP hybrid triple quadrupole-linear ion trap mass spectrometer (SCIEX) with ESI in the positive mode. The mass spectrometer was interfaced to a Shimadzu HPLC system consisting of SIL-20AC XR auto-sampler and LC-20AD XR LC pumps. The 5500 QTRAP was operated with the following settings: source voltage 4500 kV, GS1 40, GS2 30, CUR 40, TEM 650, and CAD gas high.

Analyte separation was achieved using a Gemini 3 μ C₆-Phenyl 110A 100 \times 2 mm column (Phenomenex) kept at 35°C using a Shimadzu CTO-20AC column oven. The gradient mobile phase was delivered at a flow rate of 0.4 mL/min and consisted of two solvents: A, 0.1% formic acid in water; B, 0.1% formic acid in acetonitrile. The initial concentration of solvent B was 40% followed by a linear increase to 95% B in 10 min, this was held for 2 min, then decreased back to 40% B over 0.1 min, then held for 3 min. The retention time for 7 α -hydroxy-4-cholesten-3-one was 8.2 min.

Data were acquired using SCIEX Analyst 1.6.2 and analyzed using SCIEX MultiQuant 3.0.3 software. Sample values were calculated from calibration curves generated from the peak area ratio of the analyte to internal standard versus analyte concentration that was fit to a linear equation with 1/x weighting.

FGF19 ELISA. Human serum concentration of FGF-19 was determined using the Human FGF-19 Quantikine enzyme-linked immunoassay (ELISA) (R&D Systems, DF1900). Assay was completed according to manufacturer's instructions with samples run in duplicate.

Rodent Studies.

Postpartum rodent model. The University of Colorado Anschutz Medical Campus approved rat procedures. Age-matched Sprague Dawley female rats

(Harlan) were housed and bred as described (61). For tissue collection, rats were euthanized across groups by CO₂ asphyxiation and cardiac puncture. Whole livers were removed, washed 3× in 1× phosphate-buffered saline (PBS), and tissues weighed. Left lobes were fixed in 10% neutral buffered formalin (Anatech Ltd) and processed for formalin fixed, paraffin embedded, and caudate lobes were flash frozen on liquid nitrogen for protein and RNA extraction. Oregon Health & Science University Institutional Animal Care and Use Committees approved mouse procedures. Age-matched Balb/c female mice (Charles River Laboratories, The Jackson Laboratory) were housed and bred as described (61). For tissue collection, mice were euthanized across groups by CO₂ asphyxiation and cardiac puncture. Whole livers were removed, washed 3× in 1× PBS and tissues weighed. Caudate lobes were flash frozen on liquid nitrogen for RNA extraction.

Immunohistochemistry. Immunohistochemical detection was performed as described (62). Briefly, tissues were deparaffinized, rehydrated, and heat-mediated antigen retrieval was performed with EDTA for 5 min at 125°C. Primary antibodies used were as follows: Ki67 (Neomarkers RM-9106-s, 1:50) for 2 h at room temperature (RT) and Adipophilin (LS-B2168/34250 Lifespan Biosciences, 1:400) for 1 h at RT. Secondary antibody was anti-rabbit (Agilent Envision+ K4003, RTU), used for Ki67 at 1 h at RT and for Adipophilin at 30 min at RT. DAB chromogen (Agilent, K346889-2) with hematoxylin counter stain (Agilent, S330130-2) was used to visualize positive stain. Stained sections were scanned using the Aperio AT2 slide scanner (Leica Biosystems). Number of Ki67+ hepatocytes were counted in five 1 × 1 mm areas. Adipophilin signal quantification was performed by Aperio ImageScope version 12.1.0.5029 as described previously (63). All analyses were done by investigators blinded to group.

Real-time qRT-PCR. RNA was isolated from flash frozen rat liver for complementary DNA (cDNA) synthesis and qPCR. One microgram total RNA was used for RT-mediated synthesis of cDNA using SuperScript II RT (Invitrogen) and random hexamer primers for Cyp7a and SuperScript IV (Invitrogen) for FXR. qPCR for rat Cyp7a and reference gene GAPDH was performed using FastStart Essential DNA Green Master (Roche) in an Applied Biosystems thermocycler with 45 cycles of 95°C for 20 s, 60°C for 40 s, and 72°C for 20 s. Rat primer sequences were as follows: Cyp7a, forward CTGTCATACCACAAAGTCTTATG TCA and reverse ATGCTTCTGTGCCAAATGCC; GAPDH forward CGCTGGTG CTGAGTATGTCG and reverse CTGTGTCATGAGCCCTCC.

qPCR for rat FXR and reference gene GAPDH was performed using SsoAdvanced Universal SYBR Green Supermix (BioRad) in the ViiA 7 Real-Time PCR System (Thermo Fisher) with the following times: 95°C for 2 min, 40 cycles of 95°C for 15 s and 56°C for 60 s, then 95°C for 15 s, 61°C for 60 s, and 95°C for 15 s. Rat primer sequences were as follows: FXR, forward AGGCCATGTTCTT CGTTCA and reverse TTCAGCTCCCGACACTTTT; GAPDH, forward ACCACAGT CCATGCCATC and reverse TCCACCACCTGTTGCTGA.

Immunoblotting. Rat liver protein lysates in radioimmunoprecipitation assay buffer were separated by Wes automated gel electrophoresis system (Protein Simple). Primary antibodies and dilutions were as follows: CYP7A1 (Abcam no. ab234982, 1:20), FXR (Thermo Fisher Invitrogen no. 417200, 1:50), RXRα (Abcam no. ab125001, 1:20), and GAPDH (Cell Signaling Technology no. 2118, 1:20). Protein input for CYP7A1 and FXR assays was 0.25 mg/mL. Protein input for RXR assay was 0.5 mg/mL. GAPDH assays used both 0.25 mg/mL and 0.5 mg/mL protein input. Anti-rabbit (Protein Simple no. 042-206, RTU) or anti-mouse (Protein Simple no. 042-205, RTU) horseradish peroxidase-conjugated

secondary antibodies were utilized, followed by chemiluminescent substrate (Protein Simple no. PS-CS01, Luminol-S, Peroxide). Signal was detected using the Wes System camera. Immunoblot electrophoretograms were analyzed by Compass Software (Protein Simple).

RNA-sequencing. RNA was isolated from flash frozen whole murine liver using the Direct-zol RNA MiniPrep kit (Zymo Research no. R2051). An input of 100 ng RNA was used for library preparation. Library construction was performed by Novogene using a NEBNext Ultra RNA Library Prep Kit for Illumina (cat no. E74205, New England Biolabs) according to the manufacturer's protocol. Briefly, messenger RNA (mRNA) was enriched using oligo(dT) beads followed by two rounds of purification and fragmented randomly by adding fragmentation buffer. The first-strand cDNA was synthesized using random hexamers primer, after which a custom second-strand synthesis buffer (Illumina), dNTPs, RNase H, and DNA polymerase I were added to generate the second-strand (double stranded cDNA). After a series of terminal repair, polyadenylation, and sequencing adaptor ligation, the double-stranded cDNA library was completed following size selection and PCR enrichment. The resulting 250- to 350-base pair (bp) insert libraries were quantified using a Qubit 2.0 fluorometer (Thermo Fisher Scientific) and qPCR. Size distribution was analyzed using an Agilent 2100 Bioanalyzer (Agilent Technologies). Qualified libraries were sequenced on an Illumina Novaseq6000 Platform (Illumina) using a paired-end 150 run (2 × 150 bases). The raw fastq files were first quality checked using FastQC (version 0.11.8) software. Fastq files were aligned to mm10 mouse reference genome (GRCm38.39) and per-gene counts quantified by RNA-Seq by Expectation-Maximization (RSEM) (version 1.3.1) based on the gene annotation Mus_musculus.GRCm38.89.chr.gtf. Differential gene expression analysis was performed using DESeq2 (version 1.22.2) (64). Gene expression differences were considered significant if passing the following criteria: adjusted P value < 0.05, log₂(fold change) ≥ 1. Custom gene set for GSEA analysis was built from curated gene lists available from Molecular Signature Database (<http://www.gsea-msigdb.org/gsea/index.jsp>). Specifically, the gene set was composed from the following: REACTOME_DOWNSTREAM_SIGNALING_OF_ACTIVATED_FGFR4 and REACTOME_BILE_ACID_AND_BILE_SALT_METABOLISM. Gene Set Enrichment Analysis (GSEA) analysis was performed with GSEA software developed by the University of California San Diego and Broad Institute (65, 66).

Data Availability. Raw RNA-sequencing data of mouse liver tissues performed in this study have been deposited in the Gene Expression Omnibus database under accession code [GSE188680](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE188680). All other study data are included in the article and/or [SI Appendix](#).

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