SCIENTIFIC REPORTS

Received: 23 February 2016 Accepted: 19 September 2016 Published: 07 October 2016

OPEN Lipid metabolism is associated with developmental epigenetic programming

Elizabeth H. Marchlewicz¹, Dana C. Dolinoy^{1,2,3}, Lu Tang⁴, Samantha Milewski⁵, Tamara R. Jones¹, Jaclyn M. Goodrich¹, Tanu Soni⁶, Steven E. Domino⁷, Peter X. K. Song⁴, Charles F. Burant^{2,6,8,9} & Vasantha Padmanabhan^{1,3,5,7,9}

Maternal diet and metabolism impact fetal development. Epigenetic reprogramming facilitates fetal adaptation to these in utero cues. To determine if maternal metabolite levels impact infant DNA methylation globally and at growth and development genes, we followed a clinical birth cohort of 40 mother-infant dyads. Targeted metabolomics and guantitative DNA methylation were analyzed in 1st trimester maternal plasma (M1) and delivery maternal plasma (M2) as well as infant umbilical cord blood plasma (CB). We found very long chain fatty acids, medium chain acylcarnitines, and histidine were: (1) stable in maternal plasma from pregnancy to delivery, (2) significantly correlated between M1, M2, and CB, and (3) in the top 10% of maternal metabolites correlating with infant DNA methylation, suggesting maternal metabolites associated with infant DNA methylation are tightly controlled. Global DNA methylation was highly correlated across M1, M2, and CB. Thus, circulating maternal lipids are associated with developmental epigenetic programming, which in turn may impact lifelong health and disease risk. Further studies are required to determine the causal link between maternal plasma lipids and infant DNA methylation patterns.

Fetal development is a sensitive window for epigenetic reprogramming that can have lifelong impacts on gene expression and subsequent disease risk. Demethylation and subsequent reprogramming of epigenetic marks occurs in the mammalian preimplantation embryo and developing fetus¹. Increasing evidence shows that maternal nutrition impacts uterine metabolic environment and successive fetal epigenetic programming²⁻⁴. For example, studies in animal models demonstrate that maternal diet modulation, including caloric^{5,6} and protein restriction^{7,8}, over-nutrition^{9,10}, and micronutrient supplementation¹¹, alters DNA methylation at candidate gene loci and globally at cytosine-guanine (CpG) dinucleotides and repetitive elements, including long interspersed element 1 (LINE1).

Plasma metabolite levels provide a more objective measure of the maternal environment to which the developing fetus is exposed, compared to self-reported dietary data. The metabolome is relatively stable over time in non-pregnant adults¹²⁻¹⁴, and can be used as a proxy for dietary behavior of individuals interacting with genetic factors^{15,16}. In contrast, circulating nutrient levels vary across pregnancy depending on the stage of fetal growth. In the first trimester of pregnancy, increased adipose tissue response to insulin results in anabolic storage of maternal fatty acids; in the third trimester fetal nutrient demands intensify, driving a catabolic release of maternal lipid stores and a subsequent rise in maternal plasma lipid levels¹⁷⁻¹⁹. Despite third trimester catabolism, fetal demand for long chain polyunsaturated fatty acids (LCPUFA) overwhelms maternal desaturase activity²⁰; thus,

¹Department of Environmental Health Sciences, University of Michigan School of Public Health, Ann Arbor, Michigan, 48109, USA. ²Department of Nutritional Sciences, University of Michigan School of Public Health, Ann Arbor, Michigan, 48109, USA. ³Reproductive Sciences Program, Department of Obstetrics and Gynecology, University of Michigan Medical School, Ann Arbor, Michigan, 48109, USA. ⁴Department of Biostatistics, University of Michigan School of Public Health, Ann Arbor, Michigan, 48109, USA. ⁵Department of Pediatrics, University of Michigan Medical School, Ann Arbor, Michigan, 48109, USA. ⁶Michigan Regional Comprehensive Metabolomics Resource Core, University of Michigan, Ann Arbor, Michigan, 48109, USA. ⁷Department of Obstetrics and Gynecology, University of Michigan Medical School, Ann Arbor, Michigan, 48109, USA. ⁸Department of Internal Medicine, University of Michigan Medical School, Ann Arbor, Michigan, 48109, USA. ⁹Department of Molecular and Integrative Physiology, University of Michigan Medical School, Ann Arbor, Michigan, 48109, USA. Correspondence and requests for materials should be addressed to V.P. (email: vasantha@umich.edu)

maternal dietary intake of LCPUFA is critical for maintenance of adequate fetal levels^{21,22}. The use of metabolomics to characterize maternal metabolites across pregnancy is only beginning to be realized^{23,24}.

This study investigates the changes in the maternal metabolome and epigenomic state from pregnancy to delivery and the extent to which maternal states compare to those of their infants at birth. The impact of the maternal metabolites on the epigenetic patterning in offspring is likely dependent on the timing, duration, and magnitude of exposure. We performed targeted metabolomics profiling of 40 mother-infant dyads in 1st trimester maternal plasma (M1) as well as delivery maternal plasma (M2) and infant umbilical cord blood plasma (CB). We then correlated metabolites to global and candidate loci DNA methylation profiles of the maternal and fetal peripheral blood cells to address the following questions: 1) Do maternal metabolomics profiles change from trimester 1 to delivery? 2) Do cord blood metabolome and epigenome patterns mirror those of 1st trimester and term blood? 3) Are specific metabolites associated with DNA methylation status in cord blood? The results suggest a relationship between maternal lipid-derived metabolites and cord blood DNA methylation status, which may lead to the identification of predictive, and modifiable, biomarkers of epigenetic programming of offspring.

Results

Maternal-infant characteristics. The 40 women included in this study were predominantly white (85%), non-Hispanic (95%), married (80%), and non-smokers (92.5%), with ages ranging from 21–42. All infants included in the study were from full-term pregnancies (>37 weeks), most were delivered vaginally (85.0%), no babies were classified as low birth weight (<2,500 g), and 50% were female. The majority of moms had Trimester 1 body mass index (BMI) in the healthy or overweight range (80%); maternal BMI was not correlated with pregnancy weight gain (p = 0.798), but tended to suggest an association with infant birth weight (p = 0.073). Additional demographic and physiologic measures are presented in Table 1.

Maternal metabolomics and epigenetic profiles from pregnancy to delivery. Plasma long-chain fatty acids (LCFA) and corresponding long-chain acylcarnitines (LCAC) increased (p < 0.001) by 1.4–2.6 fold (Fig. 1A,B, Table S1) from M1 to M2, consistent with the known increase in maternal fatty acid concentration and utilization, thought to spare glucose and amino acid (AA) use for the fetus²⁵. In contrast, branched chain amino acids (BCAA) and short chain acylcarnitines (SCACs) decreased (p < 0.001) by 0.6–0.7 fold (Fig. 1B,C) from M1 to M2, consistent with an increase in nonoxidative utilization of BCAA later in pregnancy²⁶. Forty-two of the 75 measured metabolites, including essential fatty acids (EFA, linoleic acid (LA) and α -linolenic acid (ALA)) and conditionally essential fatty acids (CEFA, derived from LA and ALA), LCACs, and AAs were significantly correlated (adjusted p < 0.05) in mothers from M1 to M2 (Fig. 2, Table S2).

Maternal global DNA methylation levels increased slightly, by 1.02 fold (1.86%, p = 0.010), from MI to M2, as measured via the Luminometric Methylation Assay (LUMA) at CCGG sites throughout the genome. Mean DNA methylation at maternal Estrogen Receptor 1 (*ESR1*) decreased by 0.83 fold (-19.83%, p = 0.001) from M1 to M2 (Table S3). Methylation at LINE1, Peroxisome Proliferator-Activated Receptor Alpha (*PPAR* α) and the imprinted loci Insulin-Like Growth Factor-2 (*IGF2*) and *H19*, was unchanged across pregnancy. Whether or not there was a change over time, the levels of methylation were significantly correlated (adjusted p < 0.05) between M1 and M2 (Fig. 3, Table S4).

Infant metabolome and methylome mirror their mothers. Of the 75 metabolites measured, levels of 24 metabolites were positively correlated (adjusted p < 0.05) between infant (CB) and maternal plasma at both M1 and M2. (Fig. 2, Table S2). These include the EFA critical for brain development $(18:3 (n-6))^{17}$ and the conditionally essential derivatives (20:3, 22:4, 22:5, 22:6)^{27,28}. Forty-six metabolites significantly correlated between mothers at either time point (M1 or M2) and infants (CB); 23 (50%) of these were also stable in maternal plasma from M1 to M2, as reported above. This consistency suggests tight regulation of these metabolites, especially those involved in fatty acid biosynthesis and elongation pathways, during pregnancy.

DNA methylation levels at LUMA, *LINE1*, and *PPAR* α were positively correlated (adjusted p < 0.001, p < 0.05, and p < 0.05, respectively) within maternal-infant dyads (M1, M2, CB) (Fig. 3, Table S4). M1 *H19* DNA methylation positively correlated (adjusted p < 0.05) with both M2 and CB. However, M1 *IGF2* DNA methylation was significantly correlated (p < 0.05) only with M2, not CB. Significant correlations between maternal and infant DNA methylation were CpG site specific within *ESR1* (sites 1 and 3, p < 0.05) (Table S4). Percent DNA methylation did not differ by infant sex at any CpG sites assessed.

Metabolites associated with DNA methylation status in infant cord blood leukocytes. The correlation among maternal metabolites and infant DNA methylation depended on genomic region. At both M1 and M2, there was a significant correlation (p < 0.05) of EFAs, CEFAs, medium chain acylcarnitines (MCAC) and LCACs with DNA methylation levels and at LUMA, LINE1, *ESR1*, and *PPAR* α sites, although the direction differed – with positive correlations at LINE1, *ESR1*, and *PPARa* and a negative correlation at LUMA. At M1, there were fewer significant correlations between metabolites and imprinted genes, *IGF2* and *H19*, compared to non-imprinted genes and measures of global methylation. At M1, the LCFA 20:3 and the LCAC C20:4 were significantly correlated (p < 0.05) to methylation of the imprinted gene *IGF2*; by M2, the EFA linoleic acid and CEFAs (20.1, 20.3, 20.4) were significantly correlated (p < 0.05). In contrast, AAs were most highly correlated with the imprinted gene *H19*, but only valine was significant (p < 0.05) at M1 (Table 2, Figure S2).

Certain maternal metabolites were commonly correlated to infant DNA methylation at many genic loci, when ranked by p-value (Table 2). Nervonic acid (24:1) and decenoylcarnitine (C10:1) were the top metabolites, significantly correlating to 8/12 and 7/12 gene loci at M1 (p < 0.05) or M2 (p < 0.05), followed by DHA (22:6, 6/12), erucic acid, DPA and lignoceric acid (22:1, 22:5 & 24:0, 5/12) supporting the premise that LCFAs derived from EFAs could play a role in modulation of the infant epigenome. When multivariate regression analyses

Cohort Characteristics	Number (n)	Percent (%)
Maternal Characteristics	l	
Race/Ethnicity		
Asian	0	0.0
African American	2	5.0
Hispanic	2	5.0
Native American	1	2.5
White	34	85.0
More than one race	1	2.5
Marital Status		
Single	8	20.0
Married	32	80.0
Smoking		
Never smoked	28	70.0
Quit before pregnancy	9	22.5
Current smoker	0	0.0
Missing data	3	7.5
Body Mass Index (BMI, kg/m ²)		
<18.5	0	0.00
18.5–24.9	24	57.5
25.0–29.9	9	22.5
30.0-34.9	3	7.5
>35.0	4	10.0
Newborn Characteristics		
Newborn Gender		
Male	20	50.0
Female	20	50.0
Preterm		
Yes	0	0.0
No	40	100.0
Delivery Mode		
Vaginal	34	85.0
C-Section	6	15.0
Apgar Score		
Severely Depressed (1-3)	0	0.0
Moderately Depressed (4-6)	2	5.0
Excellent Condition (7-9)	38	95.0
Physiologic Measures	Mean (SD)	Range
Maternal Characteristics		
Age (years)	32.28 (5.06)	21-42
Pregnancy weight gain (kg)	13.69 (4.61)	6-24
Height (cm)	167.10 (6.15)	155-177
Gestational age at delivery (days)	277.43 (6.41)	262-292
Newborn Characteristics		
Birth weight (g)	3533.38 (529.50)	2630-4575

Table 1. Cohort Characteristics and Physiologic Measures (N = 40).

were performed (Fig. 4), EFA and CEFAs associated significantly with CB methylation. Many of the same M1 and M2 metabolites significantly associated with changes in global methylation at both LUMA and LINE1 sites (Fig. 4, left columns); however, the direction of association differed (e.g. nervonic acid is negatively associated with LUMA methylation, but positively associated with LINE1 methylation). Because the two global methylation measures represent different aspects of global methylation (LUMA assesses methylation at every CCGG region and LINE1, a specific repetitive element), the different outcomes are not surprising, although it emphasizes the need to choose a global methylation measure that is relevant for each study. Lastly, of the potential confounding variables included in the regression models (maternal smoking, pre-pregnancy BMI, delivery mode, and infant birth weight), only infant birth weight was significantly associated with DNA methylation and only at imprinted gene loci (Fig. 4, middle columns). Methylation status of *IGF2* and *H19* is related to fetal growth²⁹, so an affect on birth weight is expected.





Correlation Coefficient	

		-1	+1
Metabolite	M1-M2	M1-CB	M2-CB
L-Carnitine	*	*	*
Acetylcarnitine (C2.0)	*		
Propionylcarnitine (C3.0)			
Isobutyryicarnitine (C5.0)			
Glutarylcarnitine (C5.0DC)			
Hexanoylcarnitine (C6.0)	*		
Octanoylcarnitine (C8.0)			*
Octenoylcarnitine (C8.1)	*	*	*
Decanoylcarnitine (C10.0)	*	*	*
Decenoylcarnitine (C10.1)			
3-OH-Dodecanovicarnitine (C12.0)			*
Dodecenovicarnitine (C12.1)			
Tetradecanoylcarnitine (C14.0)			
3-OH-Tetradecenoylcarnitine (C14.0OH)			
Tetradecenoylcarnitine (C14.1)			*
Tetradecadienoylcarnitine (C14.2)			
Hexadecanoylcarnitine (C16.0)			
3-OH-Hexadecanoylcarnitine (C16.0OH)			•
Hexadecanoylcarnitine (C16.1)		•	*
Octadecenovicarnitine (C18.0)			*
Octadecadienov/carnitine (C18.2)	*	•	*
3-OH-Octadecadienoylcarnitine (C18.2OH)			
Long chain acylcarnitine (C20.0)			*
Long chain acylcarnitine (C20.1)	*		
Long chain acylcarnitine (C20.2)			*
Long chain acylcarnitine (C20.3)			*
Long chain acylcarnitine (C20.4)			
Release Acid (14.0)			*
Palmitic Acid (16.0) Palmitoleic Acid (16.1)			*
Stearic Acid (18.0)			
Vaccenic Acid (18.1 n-7)			
Oleic Acid (18.1 n-9)			
Linoleic Acid (18.2)	*		
γ-Linolenic Acid (18.3 n-6)	•	•	•
α-Linolenic Acid (18.3 n-3)		*	
Arachidic Acid (20.0)			
Eicosadienoic Acid (20.2)	*	*	
Mead Acid (20.3)	*	•	*
Arachidonic Acid (20.4)	*		
Eicosapentaenoic Acid (20.5)		*	
Behenic Acid (22.0)	*	*	
Erucic Acid (22.1)		•	*
Adrenic Acid (22.4)			
Docosabexaenoic Acid (22.5)	*	•	*
Lignoceric Acid (24.0)	*	•	*
Nervonic Acid (24.1)	*	*	*
a-Aminoisobutyric Acid (AIB)	*		*
Alanine (Ala)	.*		
Asparagine (Asn)		•	
Aspartic Acid (Asp)		•	*
Cysteine (Cys)	•	*	*
Glutamic Acid (Glu)			
Glycine (Gly)		•	*
Histidine (His)	*	*	*
4-OH-Proline (4-OH-Pro)			
Isoleucine (Iso)			
Leucine (Leu)			*
Lysine (Lys)	*		*
Methionine (Met)			
Phenylalanine (Phen)	*		
Proline (Pro)	*		
Sarcosine (Sar)			
Serine (Ser)	*		*
Threonine (Thr)			*
Tryptophan (Trp)	*	*	*
Tyrosine (Tyr)	•		
Valine (Val)			

Figure 2. Significant Correlations between Maternal Metabolites at M1 and M2 and Infant Delivery Cord Blood (CB) Metabolites. Pearson's Correlations were performed on metabolite data and plotted into a heatmap. The darker red boxes indicate a larger positive correlation coefficient; smaller and negative coefficients are yellow to green. Significant correlations (adjusted p < 0.05) are denoted with a white asterisk.

Correlation Coefficient -1 +1

DNA Methylation M1-M2 M1-CB M2-CB Loci * LUMA LINE1 Mean LINE1 CpG1 * LINE1 CpG2 LINE1 CpG3 LINE1 CpG4 IGF2 Mean IGF2_CpG1 * IGF2 CpG2 IGF2 CpG3 H19 Mean H19 CpG1 H19 CpG2 H19 CpG3 H19 CpG4 * ESR1 Mean * * * ESR1_CpG1 ESR1 CpG2 ESR1_CpG3 PPARα Mean * PPARa_CpG1 * PPARα CpG2 * PPARα CpG3

Figure 3. Correlations between Maternal DNA Methylation at M1 and M2 and Infant Delivery Cord Blood (CB) Methylation. Pearson's Correlations were performed on global (LUMA, LINE1) and candidate gene (*IGF2*, *H19*, *ESR1*, and *PPAR alpha*) methylation data and plotted into a heatmap. The darker red boxes indicate a larger positive correlation coefficient, smaller and negative coefficients are yellow to green. Significant correlations (p < 0.05) are denoted with a white asterisk.

Top 10% M1 Metabolites Correlated with CB Mean DNA Methylation (n = 32)								
Infant Cord Blood - Mean DNA Methylation								
Metabolite Rank	LUMA	LINE-1	IGF2	H19	ESR1	PPARa		
1	C16.1*	24.1*	20.3*	Val*	20.0*	22.6*		
2	22.1*	His*	C20.4*	Iso	22.6*	C20.0*		
3	24.1*	24.0*	20.4	14.0	22.5*	24.1*		
4	C10.1*	C16.1*	C10.0	Cys	18.3n-6*	22.5		
5	C18.0*	Trp*	C14.0OH	Pro	His*	18.3n-6		
6	24.0*	C10.1*	C10.1	Phe	220*	Val		
7	His*	22.1*	C12.0OH	Leu	24.1*	C18.2		
8	22.4*	C18.0*	C16.1	C14.0	C10.1*	C18.1		
Top 10% M2 Metabolites Correlated with CB Mean DNA Methylation (n = 32)								
1	22.1*	24.0*	18.2*	18.3n-6*	C10.1*	24.1*		
2	24.0*	C10.1*	Car*	C6.0*	22.5*	C10.0*		
3	24.1*	22.1*	20.3*	C14.0OH*	22.6*	C12.0*		
4	C10.1*	C16.1*	20.4*	Trp*	24.0*	22.5*		
5	22.5*	24.1*	20.1*	C18.2OH*	24.1*	22.6*		
6	22.6*	22.5*	SumFFA*	18.0*	20.3*	C8.0*		
7	C16.1*	22.6*	C3.0*	Pro	C12.OH*	C10.1*		
8	C18.0*	Trp*	16.1	C5.0DC	22.1*	C12.0OH*		
Top 1	0% CB Metab	olites Correlat	ed with CB M	ean DNA Methy	lation ($n = 33$))		
1	24.1***	C18.0***	His*	Cys*	His**	C8.1		
2	22.1***	24.1***	Trp*	C5.0	18.3n6**	C18.0		
3	C10.1***	22.5***	22.5	C20.4	C10.1**	Sar		
4	22.5***	24.0***	20.3	C20.0	22.6**	C18.1		
5	24.0***	22.6***	Glu	C20.3	22.5*	24.1		
6	22.6***	22.1***	Sar	20.0	24.1*	C20.0		
7	C16.1***	C10.1***	OH-Pro	X18.3n6	C18.0*	C20.3		
8	Val**	His***	24.0	Trp	Sar*	Lys		

Table 2. Ranked Correlation Analysis of Metabolites (M1, M2, CB) and Mean Infant (CB) DNA Methylation. Kendall's Tau, non-parametric correlation analyses comparing metabolites (M1, M2, CB) to CB DNA methylation were ranked by p-value. Significant correlations are denoted by asterisks: *p < 0.05, **p < 0.01, ***p < 0.001. The metabolite with the most significant p-value is ranked as number 1. **Bold** font indicates positive correlation coefficients; standard font indicates negative correlation coefficients. SumFFA is a summation measure of the 22 free fatty acids measured in this study. To differentiate fatty acids from acylcarnitines, acylcarnitines have a 'C' placed in front, e.g. C10.1; whereas fatty acids begin with the number of carbons in their chain, e.g. 24.1.

Significant correlations between CB metabolites and CB DNA methylation followed similar trends to those of maternal metabolites and CB DNA methylation correlations (Table 2). For example, CB CEFAs negatively correlated (p < 0.001) with CB LUMA methylation and positively correlated (p < 0.001) at LINE1 and *ESR1* sites. The few CB metabolites that significantly correlated (p < 0.05) with CB *IGF2* and *H19* were amino acids (His, Trp, and Cys). No CB metabolites appeared to relate to *PPAR* α methylation.

Discussion

This is the first study in human subjects to investigate the role of the maternal metabolic environment in fetal epigenetic reprogramming. Most human epidemiologic birth cohorts rely on maternal diet as an indicator of metabolic environment^{30–32}. However, compared to maternal self-reported dietary data, examining maternal metabolomics profiles provides a more precise, non-biased, and comprehensive measure of the metabolic environment to which a developing fetus is exposed. The study design involving maternal sampling from the same subject dyads during first trimester and delivery also provided an opportunity to investigate whether the timing and duration (from M1 to M2) of maternal exposure impact infant DNA methylation. This study design revealed the following novel findings that very long chain fatty acids (22:1, 22:5, 22:6, 24:0, 24:1), medium chain acylcarnitines (C10.1 and C16.1), and histidine were: (1) stable in maternal plasma across pregnancy, (2) significantly correlated between maternal and infant plasma, and (3) in the top 10% of maternal metabolites correlated with infant DNA methylation. This suggests that these highly regulated metabolites may have a potential role in determining the levels of DNA methylation in the offspring.

Regarding maternal metabolite changes across pregnancy, we found higher circulating LCFA (16s–18s) levels, including EFAs, at M2 compared to M1, but no change in circulating VLCFAs (20s–24s), MCAC (C10.1), LCACs (C16-C20), or AAs. Traditionally, maternal lipolysis coupled with a decrease in maternal lipoprotein lipase activity and a metabolic state similar to starvation have characterized the third trimester of pregnancy^{33,34}. The higher





Figure 4. Multivariate Regression Analyses Illustrating the Maternal Metabolites Associated with Cord Blood DNA Methylation. Multivariate regression was used to investigate the joint associations between all metabolites and mean DNA methylation, adjusting for confounders: maternal smoking, maternal prepregnancy BMI, delivery mode, and infant birth weight. M1 and M2 metabolites that were significantly associated with global or candidate gene loci are plotted in the heatmap. Negative associations are denoted in blue, positive associations in red.

circulating LCFAs at M2 vs. M1 we observed are consistent with this established trend, supporting the theory that maternal third trimester hypertriglyceridemia fuels the rapid fetal growth characteristic of this trimester^{33,34}. Consistent with the increase in the levels of LCFA oxidation is the increased levels of MCAC. Interestingly, while the levels of the acylcarnitines derived from the 2 major fatty acid species, C16:0 and C18:1 were unchanged the levels of the VLCAC derived from LA and ALA (C20:3 and C20:4) were reduced. We speculate that this may be due to diversion of these CEFA into biosynthetic pathways.

The conservation of many VLCFAs, MCAC, LCACs and AA levels in maternal plasma across pregnancy is a novel finding. This may occur to support healthy fetal neurodevelopment. Fatty Acid Desaturase 1 and 2 (*FADS1*, *FADS2*) encode delta-5 and delta-6 desaturases, which are the rate-limiting steps in the conversion of LA and ALA to the longer, more unsaturated and biologically active EPA, DHA and arachidonic acid; *FADS1* and *FADS2* are also related to levels of VLCFA in blood cells and plasma³⁵. Fetal delta-5 and delta-6 desaturase activity are negligible prior to birth, so fetal growth is dependent on placental transfer of LCPUFA²¹. Thus, maternal dietary intake of LCPUFA is critical to achieve adequate fetal levels^{20,22}, suggesting maternal diet may drive the fetal programming effect observed as differential DNA methylation in cord blood at birth.

Comparing maternal and infant plasma metabolite levels also demonstrated highly correlated levels of MCACs (C8 s, C10 s), LCACs (C16 s, C18 s), EFA (X18.3 n-6), VLCFA (X22 s, X24 s), and certain AAs (Cys, Gly, His, Trp). The fact that these are the same metabolites that were highly conserved from M1 to M2 further supports the developmental importance of lipid metabolism. Previous studies have reported fetal LCFA levels

depend on circulating maternal LCFAs^{17,36}, and observed preferential transport of DHA from maternal plasma to placental and fetal tissues occurs in the third trimester of pregnancy³⁷.

Thus, maternal regulation of lipid metabolism during pregnancy is likely critical for fetal health. VLCFAs, especially DHA (22.6) and nervonic acid (24.1), are vital components of myelin sheaths; adequate levels of these VLCFAs stimulate neuronal growth^{27,28,36,38}. VLCFAs are likely important components of fetal neurodevelopment, and thus maternal-fetal levels of these metabolites may be tightly regulated across pregnancy to ensure fetal health.

Surprisingly, the comparison of maternal-infant epigenomes revealed similarities in global and candidate gene DNA methylation levels between the dyads. Erasure of parental DNA methylation marks in mammals occurs early in the first trimester, prior to implantation³⁹. Fetal epigenetic reprogramming follows in two waves (1) in preimplantation zygotes during early embryogenesis and (2) in primordial germ cells (PGC) of the developing fetus, differentially by offspring sex^{40–42}. Repetitive elements, including LINE1, are known to remain highly methylated in PGCs^{42,43} and preimplantation embryos⁴³, suggestive of the need to repress transposon movement in order to maintain genomic stability. The significant correlation between maternal-infant LUMA, LINE1 and candidate gene methylation, despite global reprogramming in early development³⁹, implies that methylation status at some CpG sites may be influenced by maternal genetics and/or shared environment.

The maternal metabolites that were significantly correlated to infant DNA methylation, largely overlapped the metabolites that remained constant from M1 to M2 and the metabolites that correlated between mother-infant pairs, signifying a potential role in epigenetic programming. The observed positive correlation of VLCFAs with methylation at LINE1 retrotransposons and in the promoter regions of candidate genes (*ESR1, PPAR* α) is supported by previous findings that DHA exposure affects DNA methylation globally and at metabolic and inflammatory candidate gene loci in *in vitro*⁴⁴, murine⁴⁵, and human⁴⁶⁻⁴⁸ studies. In the Southampton Women's Survey birth cohort, maternal third trimester plasma total n-6 PUFAs positively associated with child adipose mass while total n-3 PUFAs positively correlated with lean mass at four and six years of age, suggestive of age-specific metabolic reprogramming effects of prenatal PUFA exposure during human fetal development⁴⁹.

Our study also found that infant *H19* methylation correlated most strongly with maternal amino acid metabolites indicative of maternal protein playing a critical role in DNA methylation patterns at imprinted gene loci. Imprinted genes maintain allele-specific epigenetic marks based on the parent of origin of the transmitted allele⁵⁰. Altered DNA methylation and gene expression of imprinted *IGF2* and *H19* loci in offspring following perinatal protein restricted diets are well documented in animals⁵¹⁻⁵³ and humans⁵⁴. Hypermethylation at *H19* positively correlates to weight-for-age greater than 85th percentile in infants⁵⁵ and greater skinfold thickness measures and subcutaneous adiposity in teens⁵⁶, demonstrating metabolic effects on offspring from circulating maternal amino acids in addition to circulating VLCFAs.

These study findings provide important insight into understanding the sensitivity of the fetal epigenome to maternal metabolic environment. Additional studies are needed, however, to further characterize the maternal-fetal metabolic milieu in a larger sample size, with more frequent time-points throughout pregnancy and with longitudinal follow-up of infants into childhood to assess potential alterations in growth and development resulting from differential epigenetic reprogramming. Maternal blood collection at delivery may not be representative of the metabolic and epigenomic states during the third trimester due to the stress of parturition; collection earlier in the third trimester would improve characterization of the changing metabolic environment. The investigation of other epigenetic mechanisms (e.g. histone modifications, miRNAs) would also provide a more comprehensive evaluation of pregnancy-related impacts on fetal programming.

Conclusion

Understanding the role of maternal metabolomics across pregnancy and the impact of metabolites on fetal epigenetic programming in healthy moms with uncomplicated pregnancies, as examined in this study, provides a point of comparison for the numerous studies now being conducted on the impact of maternal conditions, such as obesity, gestational diabetes, and non-alcoholic fatty liver disease, on fetal metabolic programming and birth outcomes. The study design allowed for the novel comparison of metabolites and epigenetic marks from pregnancy to delivery in the mother and the comparison of maternal measures to infant levels. Significant findings include the tight regulation of LC/VLCFA and MC/LCAC levels in maternal blood across pregnancy and in cord blood at delivery. The fact that the same LC/VLCFAs, MC/LCACs, and histidine correlate with infant cord blood DNA methylation is noteworthy since these maternal metabolites have previously been associated with fetal neuro-visual development (LC/VLCFA) and general fetal growth (histidine and other amino acids). DNA methylation at CCGG sequences, LINE1 repetitive elements and some genic loci were highly correlated between maternal blood and infant cord blood. Thus, the shared genetic and/or environmental factors of mother-infant dyads impact DNA methylation, potentially contributing to the developmental origins of health and disease risk. This study is the first to examine the impact of individual plasma metabolites on infant DNA methylation globally and in candidate imprinted and non-imprinted genes, which provided more detailed insights into fetal responses to the maternal metabolic milieu. Once fully characterized, a metabolomics assessment from maternal blood drawn during prenatal clinic visits may serve as useful screens for fetal nutritional sufficiency and subsequent disease risk, thus providing unprecedented benefit for early preventive care and lifestyle recommendations to decrease disease incidence beginning at birth.

Materials and Methods

Subject Population. Forty women were recruited from the community surrounding the University of Michigan Von Voigtlander Women's Hospital, after review and approval by the University of Michigan Institutional Review Board (IRBMED: HUM00017941). The study was carried out in accordance with the approved guidelines and regulations, including appropriate informed consent. Women were contacted during their 1st trimester prenatal clinic visit and were included in the study if they fit the following inclusion criteria: 18



Figure 5. Timeline of Clinical Study Visits. Mothers were recruited, consented, and blood was drawn during a 1st trimester prenatal clinic visit. Maternal blood was also drawn when women were admitted to the hospital, prior to delivery. Infant cord blood was collected immediately after delivery.

years or older, natural conception (no infertility treatment), singleton pregnancy, 8 to 14 weeks pregnant (Fig. 5). At this initial study visit, participants had blood drawn via venipuncture and filled out a simple, one-page study questionnaire. The questionnaire included items on race, ethnicity, income, and smoking status. Women were re-contacted between 34 to 38 weeks of pregnancy and provided study materials to ensure they understood sample collection procedures during delivery. Upon admission for delivery, venous blood samples were drawn. After delivery of the placenta, cord blood samples were collected via venipuncture from the umbilical cord, which was clamped proximal to the placenta. Due to limited availability of biological samples, epigenetic and metabolomics analyses were performed on a subset of the maternal-infant dyads: M1-M2 n = 37, M1-CB n = 32, M2-CB n = 32, CB-CB n = 33.

Mass-Spectrometry Based Metabolomics Analysis. Directed metabolomics analyses (76 metabolites) were conducted on M1, M2, and CB plasma samples. Twenty-three amino acids (AA) and 23 free fatty acids (FFA) were measured via gas chromatography-mass spectrometry (GC-MS) on an Agilent 7890A-5975C instrument. Agilent CHEMSTATION software was used to analyze the MS spectral data. Thirty acyl carnitines (AC) were measured via reverse-phase liquid chromatography-mass spectrometry (LC-MS) on an Agilent 1200 LC/6400 series triple quadrupole mass spectrometer with electrospray ionization source (ESI), which is operated in positive mode. Data were processed by MassHunter Quantitative software; metabolite levels are presented as absolute measures, not relative amounts. Rationale and quality control measures for metabolomics analyses are explained in SI Materials and Methods.

DNA Methylation Analysis. Fifty ng of bisulfite converted DNA (SI Materials and Methods) from each sample was used in each PCR reaction mixed with HotStarTaq master mix (Qiagen), 10 pmol forward primer and 5 pmol biotin-labeled reverse primer followed by assay specific PCR cycling parameters. Primer sequences and PCR cycling parameters for the following assays are provided in the Supporting Information (LINE1, *IGF2, H19, ESR1, PPAR* α). Products amplified by PCR were verified via gel electrophoresis. The PyroMark TMQ96 MD pyrosequencing system (Qiagen) uses a predetermined sequence-to-analyze to dispense nucleotides. PyroMark Q-CpG software (Qiagen) computes the ratio of cytosine to thymine signal intensity at each CpG site in the assays; data outputs as percent methylation. The Luminometric Methylation Assay (LUMA) uses restriction enzymes to digest genomic DNA and detects methylation at 5'-CCGG-3' sequences across the genome. A detailed description of LUMA analysis via pyrosequencing has been described previously^{57,58}; briefly, global DNA methylation is calculated by the ratio of normalized product signal between methylation sensitive and insensitive digestions from the same sample: $1-[(HpaII/EcoR1)/(MspI/EcoR1)] \times 100$.

The Luminometric Methylation Assay (LUMA) provides a global measure of DNA methylation at CCGG sites and has been used in both clinical and longitudinal epidemiologic studies⁵⁷. Altered global methylation assessed via LUMA has been associated with environmental exposures in humans⁵⁹. LINE1, a measure of long interspersed retrotransposons across the genome, is another method of measuring global methylation⁶⁰ and has been shown to be a useful marker of chemical exposure⁶¹. While LINE1 methylation is a convenient way to measure global methylation, gene-specific promoter regions may methylate differently in response to exposures, making it essential to measure both global and candidate genes of interest⁶².

Candidate loci were selected for investigation based on their role in fetal growth and development and previously reported associations between loci and fetal programming effects. A significant body of research has investigated the importance of imprinted genes, *IGF2* and *H19*, in healthy fetal development^{63,64}. Imprinted genes maintain allele-specific epigenetic marks based on the parent of origin of the transmitted allele⁵⁰. Maternal nutrient intake^{53,55,65,66} has been associated with alterations in methylation at the imprint control regions (ICRs) for *IGF2* and *H19*. Expression of these imprinted genes, critical for fetal development and rapid somatic growth in the early postnatal period, decreases after this initial growth phase, coordinating growth deceleration across tissues to limit adult body size⁶⁷.

For non-imprinted genes, estrogen receptor 1 (*ESR1*) is an estradiol-sensitive signal transducer important in the progression of labor and delivery⁶⁸ and is associated with development of obesity and non-alcoholic fatty liver disease later in life⁶⁹⁻⁷¹. Similarly, DNA methylation of *PPAR* α has been implicated as a fetal programmer of metabolic health and disease⁷². Early life exposure to leptin increases *PPAR* α variant transcription in the liver; transcriptional changes are sustained into adulthood⁷³. Many other fetal candidate genes may be impacted by the maternal metabolome; this study served as a proof of concept that both imprinted and non-imprinted genes could be affected. Additional studies in this human birth cohort plan to expand the candidate gene panel to investigate the impact on other gene loci. PCR parameters and PCR and pyrosequencing primers used in optimized assays to identify DNA methylation at each candidate locus are provided in Table S5.

Statistical Analysis. Clinical data collected during the 1st trimester and delivery clinic visits were combined with metabolomics and DNA methylation data into one large dataset. No outliers, defined by a cutoff of more than three standard deviations, were identified in epigenetics or metabolomics data. DNA methylation measures were normally distributed, so values are presented as mean and standard deviation (SD); paired t-tests were used to compare maternal methylation at M1 and M2. Despite no significant outliers, metabolite levels were all right skewed, so results are presented as median and interquartile range (IQR); non-parametric tests (Kendall's rank correlation and Kruskal-Wallis tests) were used to carry out bivariate analyses. In order to correct for multiple testing and control false discovery rate (FDR), the Benjamini-Hochberg method was used to calculate adjusted p-values to determine significance of results⁷⁴.

Multivariate regression models were fitted to simultaneously investigate the adjusted association between all metabolites and mean DNA methylation at all loci, with adjustment for confounders. Using a penalized multivariate linear regression model, coefficients of unimportant predictors and confounders are estimated to exactly zero (fitting is done using the remMap package in R⁷⁵). The important non-zero associations from penalized regression were then plotted via heat maps based on the magnitudes of the beta coefficients. Additional details regarding bivariate analyses and false discovery rate are provided in Supporting Information. All analyses were performed in R Studio, Version 3.0.2.

References

- 1. Cantone, I. & Fisher, A. G. Epigenetic programming and reprogramming during development. *Nat. Struct. Mol. Biol.* **20**, 282–289 (2013).
- 2. Bateson, P. et al. Developmental plasticity and human health. Nature 430, 419-421 (2004).
- 3. Lillycrop, K. a & Burdge, G. C. The effect of nutrition during early life on the epigenetic regulation of transcription and implications for human diseases. *J. Nutrigenet. Nutrigenomics* **4**, 248–260 (2011).
- 4. Canani, R. B. et al. Epigenetic mechanisms elicited by nutrition in early life. Nutr. Res. Rev. 24, 198–205 (2011).
- Gluckman, P. D. et al. Metabolic plasticity during mammalian development is directionally dependent on early nutritional status. Proc. Natl. Acad. Sci. USA 104, 12796–12800 (2007).
- Palou, M., Priego, T., Sánchez, J., Palou, A. & Picó, C. Sexual dimorphism in the lasting effects of moderate caloric restriction during gestation on energy homeostasis in rats is related with fetal programming of insulin and leptin resistance. *Nutr. Metab. (Lond)*. 7, 69–79 (2010).
- Altobelli, G., Bogdarina, I. G., Stupka, E., Clark, A. J. L. & Langley-Evans, S. Genome-wide methylation and gene expression changes in newborn rats following maternal protein restriction and reversal by folic acid. *PLoS One* 8, e82989 (2013).
- Oster, M., Murani, E., Metges, C. C., Ponsuksili, S. & Wimmers, K. A low protein diet during pregnancy provokes a lasting shift of hepatic expression of genes related to cell cycle throughout ontogenesis in a porcine model. *BMC Genomics* 13, 93 (2012).
- Burgueño, A. L., Cabrerizo, R., Gonzales Mansilla, N., Sookoian, S. & Pirola, C. J. Maternal high-fat intake during pregnancy programs metabolic-syndrome-related phenotypes through liver mitochondrial DNA copy number and transcriptional activity of liver PPARGC1A. J. Nutr. Biochem. 24, 6–13 (2013).
- 10. Samuelsson, A.-M. *et al.* Diet-induced obesity in female mice leads to offspring hyperphagia, adiposity, hypertension, and insulin resistance: a novel murine model of developmental programming. *Hypertension* **51**, 383–392 (2008).
- 11. Gao, R. *et al.* Effect of folate deficiency on promoter methylation and gene expression of Esr1, Cdh1 and Pgr, and its influence on endometrial receptivity and embryo implantation. *Hum. Reprod.* **27**, 2756–2765 (2012).
- 12. Breier, M. *et al.* Targeted metabolomics identifies reliable and stable metabolites in human serum and plasma samples. *PLoS One* **9**, e89728 (2014).
- 13. Bernini, P. et al. Individual Human Phenotypes in Metabolic Space and Time. J Proteome Res. 8, 4264-4271 (2009).
- 14. Kim, K. et al. Mealtime, temporal, and daily variability of the human urinary and plasma metabolomes in a tightly controlled environment. PLoS One 9, e86223 (2014).
- 15. Swann, J. & Claus, S. Nutrimetabonomics: nutritional application of metabolic profiling. Sci. Prog. 97, 41-47 (2014).
- 16. Shin, S.-Y. et al. An atlas of genetic influences on human blood metabolites. Nat. Genet. 46, 543–550 (2014).
- 17. Cetin, I., Alvino, G. & Cardellicchio, M. Long chain fatty acids and dietary fats in fetal nutrition. J. Physiol. 587, 3441–3451 (2009).
- Al, M. D. M. et al. Maternal essential fatty acid patterns during normal pregnancy and their relationship to the neonatal essential fatty acid status. Br. J. Nutr. 74, 55–68 (1995).
- Butte, N. F. Carbohydrate and lipid metabolism in pregnancy : normal compared with gestational diabetes mellitus 1–5. Am. J. Clin. Nutr. 71, 1256–1261 (2000).
- Hoving, E. B., van Beusekom, C. M., Nijeboer, H. J. & Muskiet, F. A. Gestational age dependency of essential fatty acids in cord plasma cholesterol esters and triglycerides. *Pediatr. Res.* 35, 461–469 (1994).
- Hanebutt, F. L., Demmelmair, H., Schiessl, B., Larque, E. & Koletzko, B. Long-chain polyunsaturated fatty acid (LC-PUFA) transfer across the placenta. *Clin. Nutr.* 27, 685–693 (2008).
- 22. Haggarty, P. Fatty Acid Supply to the Human Fetus. Annu. Rev. Nutr. 30, 237–255 (2010).
- 23. Luan, H. et al. Pregnancy-induced metabolic phenotype variations in maternal plasma. J. Proteome Res. 13, 1527-1536 (2014).
 - 24. Pinto, J. et al. Following healthy pregnancy by NMR metabolomics of plasma and correlation to urine. J. Proteome Res. 14, 1263–1274 (2015).
 - Ghio, A., Bertolotto, A., Resi, V., Volpe, L. & Di Cianni, G. Triglyceride metabolism in pregnancy. Adv. Clin. Chem. 55, 133–153 (2011).
 - 26. Jolly, M. *et al.* Increased Leucine Turnover in Women during the Third Trimester of Uncomplicated Pregnancy. *Metabolism.* 53, 545–549 (2004).
 - Martinez, M. & Mougan, I. Fatty Acid Composition of Human Brain Phospholipids during Normal Development. J. Neurochem. 71, 2528–2533 (1998).
 - 28. Innis, S. Dietary (n-3) Fatty Acids and Brain Development. J. Nutr. 137, 855-859 (2007).
 - 29. St-Pierre, J. et al. IGF2 DNA methylation i a modulator of newborn's fetal growth and development. Epigenetics 7, 1125–1132 (2012).
 - Godfrey, K. M. et al. Epigenetic gene promoter methylation at birth is associated with child's later adiposity. Diabetes 60, 1528–1534 (2011).

- Lumey, L. H., Stein, A. D., Kahn, H. S. & Romijn, J. A. Lipid profiles in middle-aged men and women after famine exposure during gestation: the Dutch Hunger Winter Families Study. Am. J. Clin. Nutr. 89, 1737–1743 (2009).
- 32. Desgagné, V. et al. Epigenetic dysregulation of the IGF system in placenta of newborns exposed to maternal impaired glucose tolerance. *Epigenomics* **6**, 193–207 (2014).
- Felig, P. & Lynch, V. Starvation in human pregnancy: hypoglycemia, hypoinsulinemia and hyperketonemia. Science (80-.). 170, 990–992 (1970).
- Herrera, E. et al. Role of lipoprotein lipase activity on lipoprotein metabolism and the fate of circulating triglycerides in pregnancy. Am J Obs. Gynecol 158, 1575–1583 (1988).
- 35. Standl, M. *et al.* FADS1 FADS2 gene cluster, PUFA intake and blood lipids in children: results from the GINIplus and LISAplus studies. *PLoS One* 7, e37780 (2012).
- 36. Bernard, J. Y. *et al.* The Dietary n6 : n3 Fatty Acid Ratio during Pregnancy Is Inversely Associated with Child Neurodevelopment in the EDEN Mother-Child Cohort. *J Nutr* **143**, 1481–1488 (2013).
- 37. Gil-sanchez, A. *et al.* Maternal-fetal *in vivo* transfer of [13 C] docosahexaenoic and other fatty acids across the human placenta 12 h after maternal oral intake. *Am J Clin Nutr* **92**, 115–122 (2010).
- Guesnet, P. & Alessandri, J. M. Docosahexaenoic acid (DHA) and the developing central nervous system (CNS) Implications for dietary recommendations. *Biochimie* 93, 7–12 (2011).
- 39. Feng, S., Jacobsen, S. E. & Reik, W. Epigenetic reprogramming in plant and animal development. Science 330, 622-627 (2010).
- Bernal, A. J. & Jirtle, R. L. Epigenomic disruption: the effects of early developmental exposures. Birth Defects Res. A. Clin. Mol. Teratol. 88, 938-944 (2010).
- Yamazaki, Y. et al. Reprogramming of primordial germ cells begins before migration into the genital ridge, making these cells inadequate donors for reproductive cloning. Proc. Natl. Acad. Sci. 100, 12207–12212 (2003).
- 42. Hajkova, P. et al. Epigenetic reprogramming in mouse primordial germ cells. Mech. Dev. 117, 15–23 (2002).
- 43. Guibert, S., Forne, T. & Weber, M. Global profiling of DNA methylation erasure in mouse primordial germ cells. *Genome Res.* 22, 633–641 (2012).
- 44. Cho, Y. *et al.* Colon cancer cell apoptois is induced by combined exposure to the n-3 fatty acid docosahexaenoic acid and butyrate through promoter methylation. *Exp Biol Med* **239**, 302–310 (2014).
- 45. Kulkarni, S. et al. Maternal lipids are as important as glucose for fetal growth. Diabetes Care 36, 2706–2713 (2013).
- Lee, H. S. et al. Modulation of DNA methylation states and infant immune system by dietary supplementation with v-3 PUFA during pregnancy in an intervention study. Am. J. Clin. Nutr. 98, 480–487 (2013).
- 47. Howard, T. D. *et al.* DNA methylation in an enhancer region of the FADS cluster is associated with FADS activity in human liver. *PLoS One* **9**, 1–7 (2014).
- Aslibekyan, S. *et al.* DNA Methylation Patterns Are Associated with n 3 Fatty Acid Intake in Yup'ik People. J Nutr 144, 425–430 (2014).
- 49. Moon, R. J. et al. Maternal plasma polyunsaturated fatty acid status in late pregnancy is associated with offspring body composition in childhood. J Clin Endocrinol Metab 98, 299–307 (2013).
- 50. Hackett, J. & Ssurani, M. DNA methylation dynamic during the mammalian life cycle. *Philos. Trans. R. Soc. London, Biol. Sci.* 368, 20110328 (2013).
- Micke, G. C., Sullivan, T. M., Mcmillen, I. C., Gentili, S. & Perry, V. E. A. Heifer nutrient intake during early- and mid-gestation programs adult offspring adiposity and mRNA expression of growth-related genes in adipose depots. *Reproduction* 141, 697–706 (2011).
- 52. Gao, H. et al. Maternal protein restriction regulates IGF2 system in placental labyrinth. Front Biosci (Elite Ed) 4, 1434-1450 (2013).
- Gong, L., Pan, Y.-X. & Chen, H. Gestational low protein diet in the rat mediates Igf2 gene expression in male offspring via altered hepatic DNA methylation. *Epigenetics* 5, 619–626 (2010).
- 54. Heijmans, B. T. *et al.* Persistent epigenetic differences associated with prenatal exposure to famine in humans. *Proc. Natl. Acad. Sci. USA* **105**, 17046–17049 (2008).
- 55. Perkins, E. *et al.* Insulin-like growth factor 2/H19 methylation at birth and risk of overweight and obesity in children. *J. Pediatr.* 161, 31–39 (2012).
- Huang, R.-C. *et al.* DNA methylation of the IGF2/H19 imprinting control region and adiposity distribution in young adults. *Clin. Epigenetics* 4, 21–31 (2012).
- 57. Karimi, M., Johansson, S. & Ekstrom, T. J. Using LUMA: a Luminometric-based assay for global DNA-methylation. *Epigenetics* 1, 45–48 (2006).
- Karimi, M. et al. LUMA (LUminometric Methylation Assay) A high throughput method to the analysis of genomic DNA methylation. Exp. Cell Res. 312, 1989–1995 (2006).
- Lind, L. et al. Global DNA hypermethylation is associated with high serum levels of persistent organic pollutants in an elderly population. Environ. Int. 59, 456–461 (2013).
- 60. Yang, A. *et al.* A simple method for estimating global DNA methylation using bisulfite PCR of repetitive DNA elements. *Nucleic Acids Res.* **32**, e38 (2004).
- 61. Florea, A. for the Assessment of Epigenetic Active Environmental or Clinical Relevant Chemicals. 2013 (2013).
- 62. Price, E. M. & Robinson, W. P. Response to: "Response to Different measures of 'genome-wide' DNA methylation exhibit unique properties in placental and somatic tissues. *Epigenetics* 7, 965 (2012).
- 63. Wutz, a et al. Imprinted expression of the Igf2r gene depends on an intronic CpG island. Nature 389, 745-749 (1997).
- 64. Thorvaldsen, J. L., Fedoriw, A. M., Nguyen, S. & Bartolomei, M. S. Developmental Profile of H19 Differentially Methylated Domain (DMD) Deletion Alleles Reveals Multiple Roles of the DMD in Regulating Allelic Expression and DNA Methylation at the Imprinted H19/Igf2 Locus Developmental Profile of H19 Differentially Met. *Mol Cell Biol* 26, 1245 (2006).
- 65. Kwong, W. Y. et al. Imprinted gene expression in the rat embryo-fetal axis is altered in response to periconceptional maternal low protein diet. Reproduction 132, 265–277 (2006).
- 66. Hoyo, C. et al. Methylation variation at IGF2 differentially methylated regions and maternal folic acid use before and during pregnancy. Epigenetics 6, 928-936 (2011).
- Lui, J. C., Finkielstain, G. P., Barnes, K. M. & Baron, J. An imprinted gene network that controls mammalian somatic growth is down-regulated during postnatal growth deceleration in multiple organs. 1103, 189–196 (2008).
- 68. Schroeder, J. W. et al. Neonatal DNA methylation patterns associate with gestational age. Epigenetics 6, 1498–1504 (2011).
- 69. Cordero, P., Gomez-Uriz, a M., Campion, J., Milagro, F. I. & Martinez, J. a. Dietary supplementation with methyl donors reduces fatty liver and modifies the fatty acid synthase DNA methylation profile in rats fed an obesogenic diet. *Genes Nutr.* **8**, 105–113 (2013).
- Mori, H., Matsuda, K. I., Tsukahara, S. & Kawata, M. Intrauterine position affects estrogen receptor α expression in the ventromedial nucleus of the hypothalamus via promoter DNA methylation. *Endocrinology* 151, 5775–5781 (2010).
- Sertic, J. et al. Variants of ESR1, APOE, LPL and IL-6 loci in young healthy subjects: association with lipid status and obesity. BMC Res. Notes 2, 203 (2009).
- 72. Rees, W. D., McNeil, C. J. & Maloney, C. a. The Roles of PPARs in the Fetal Origins of Metabolic Health and Disease. *PPAR Res.* **2008**, 459030 (2008).

- 73. Garratt, E. S. *et al.* Tissue-specific 5' heterogeneity of PPARα transcripts and their differential regulation by leptin. *PLoS One* **8**, e67483 (2013).
- 74. Benjamini, Y. & Hochberg, Y. Controlling the False Discovery Rate : a Practical and Powerful Approach to Multiple Testing. J. R. Stat. Soc. 57, 289–300 (1995).
- Peng, J. et al. Regularized Multivariate Regression for Identifying Master Predictors with Application to Integrative Genomics Study of Breast Cancer. ARXIV eprint arX 1–53 (2009).

Acknowledgements

Funding support for this study includes: NIH Grants R01 ES017005, P01 ES02284401/U.S. EPA RD83543601, P30 DK089503, P30 ES017885, T32 ES007062, U24 DK097153, UL1 RR024986, the Robert C. and Veronica Atkins Foundation, and the A. Alfred Taubman Medical Research Institute. The contents of this publication are solely the responsibility of the grantee and do not necessarily represent the official views of the US EPA or the NIH. Further, the US EPA does not endorse the purchase of any commercial products or services mentioned in the publication.

Author Contributions

V.P., D.C.D. and C.B. conceived and designed the study; S.E.D. provided clinical oversight; P.X.K.S. advised on or performed statistical analyses; S.M. retrieved clinical data and maintained data base, E.H.M. and T.R.J. performed epigenetic analyses; C.B. and T.S. performed metabolomics analyses; L.T. performed statistical analyses under the direction of P.K.K.S. with input from E.H.M., J.M.G. and T.S. E.H.M., D.C.D., L.T., P.X.K.S., C.B. and V.P. drafted the manuscript and all authors critically revised and approved the manuscript.

Additional Information

Supplementary information accompanies this paper at http://www.nature.com/srep

Competing financial interests: The authors declare no competing financial interests.

How to cite this article: Marchlewicz, E. H. *et al.* Lipid metabolism is associated with developmental epigenetic programming. *Sci. Rep.* **6**, 34857; doi: 10.1038/srep34857 (2016).

This work is licensed under a Creative Commons Attribution 4.0 International License. The images or other third party material in this article are included in the article's Creative Commons license, unless indicated otherwise in the credit line; if the material is not included under the Creative Commons license, users will need to obtain permission from the license holder to reproduce the material. To view a copy of this license, visit http://creativecommons.org/licenses/by/4.0/

© The Author(s) 2016