

**Title:** Epigenome-wide association study of placental DNA methylation and maternal exposure to night shift work in the Rhode Island Child Health Study

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## ABSTRACT

**Objectives:** Circadian disruption from environmental and occupational exposures can potentially impact health, including offspring health, through epigenetic alterations. Night shift workers experience circadian disruption, but little is known about how this exposure could influence the epigenome of the placenta, which is situated at the maternal-fetal interface. To investigate whether night shift work is associated with variations in DNA methylation patterns of placental tissue, we conducted an epigenome-wide association study (EWAS) of night shift work.

**Methods:** CpG specific methylation genome-wide of placental tissue (measured with the Illumina 450K array) from participants (n=237) in the Rhode Island Child Health Study (RICHs) who did (n=53) and did not (n=184) report working the night shift was compared using robust linear modeling, adjusting for maternal age, pre-pregnancy smoking, infant sex, maternal adversity, and putative cell mixture.

**Results:** Night shift work was associated with differential methylation in placental tissue, including CpG sites in the genes *NAV1*, *SMPD1*, *TAPBP*, *CLEC16A*, *DIP2C*, *FAM172A*, and *PLEKHG6* (Bonferroni-adjusted  $p < 0.05$ ). CpG sites within *NAV1*, *MXRA8*, *GABRG1*, *PRDM16*, *WNT5A*, and *FOXP1* exhibited the most hypomethylation, while CpG sites within *TDO2*, *ADAMTSL3*, *DLX2*, and *SERPINA1* exhibited the most hypermethylation (BH  $q < 0.10$ ). *PER1* was the only core circadian gene demonstrating differential methylation. Functional analysis indicated GO-terms associated with cell-cell adhesion.

**Conclusions:** Night shift work was associated with differential methylation of the placenta, which may have implications for fetal health and development. Additionally, neuron navigator 1 (*NAV1*) may play a role in the development of the human circadian system.

**Keywords:** *Night shift work, epigenetics, epigenome-wide association study, neuron navigator 1, circadian disruption*

## What is already known about this subject?

Night shift work and circadian disruption may play a role in the development and progression of many diseases. However, little is known about how circadian disruption impacts human fetal health and development.

## What are the new findings?

Working the night shift is associated with altered placental methylation patterns, and particularly, neuron navigator 1 (*NAV1*) may play a role in the development of the human circadian system.

## How might this impact on policy or clinical practice in the foreseeable future?

Night shift work prior to or during pregnancy may alter the placental epigenome, which has implications for fetal health. Further studies are needed to evaluate night shift work as a possible risk factor for gestational diabetes and to evaluate the impact of circadian disruption on fetal health and development.

## INTRODUCTION

Disruption of circadian rhythms is an occupational hazard for people who work the night shift and is associated with negative health outcomes such as cancer, metabolic disorders, and neurological disorders <sup>1</sup>. Animal models also demonstrate altered metabolism, hormonal signaling, body temperature rhythms, and adiposity in experiments mimicking shift work exposure <sup>2</sup>. The health risks posed by night shift work may have large public health consequences, as approximately 15% of American employees work outside of the traditional 9AM-5PM work schedule <sup>3</sup>. While some aspects of the circadian system may return to normal after a regular schedule of night shift work, studies suggest the majority of regular night shift workers (~97%) aren't able to fully adapt their endogenous circadian rhythms to their work schedules <sup>4</sup>. People also commonly experience circadian disruption when exposed to light at night (LAN) <sup>5</sup>.

The daily and seasonal patterns of light and dark exposure are an important environmental stimulus. Organisms have evolved an internal timekeeping mechanism, the circadian clock, to generate rhythms of biological activity to adapt to predictable environmental changes, increasing physiological efficiency and fitness. The core circadian clock consists of feedback loops of transcription factors (TF) that generate oscillating cycles of gene transcription and translation. These endogenously generated rhythms rely on cues, such as light, to synchronize patterns of physiological activity with the external environment. When light enters the eye, it activates visual photopigment in photoreceptors and in melanopsin-containing retinal ganglion cells (mRGCs). The light signal is transmitted via the retinohypothalamic tract to the suprachiasmatic nucleus (SCN) of the hypothalamus, the "master clock" that sets the body's peripheral clocks <sup>6</sup>. This elegant system of interdependent signaling ensures processes such as protein synthesis, fatty acid metabolism, and insulin release occur at the appropriate times to adapt to daily and seasonal changes in the environment <sup>7</sup>. However, when circadian rhythms are misaligned, cell signaling becomes inefficient and dissonant, contributing to disease.

Animal studies suggest circadian disruption *in utero* negatively affects the health and development of offspring <sup>8</sup>. Mice exposed to a 22-hour light-dark cycle, instead of the normal 24-hour cycle, had altered methylation patterns in the SCN and altered circadian behavior; differential methylation was also found for genes related to axonal migration, synaptogenesis, and neuroendocrine hormones <sup>9</sup>. Additionally, chronic changes in the photoperiod of pregnant rats caused increased leptin levels, insulin secretion, fat deposition, and decreased glucose tolerance of offspring in adulthood <sup>10</sup>. Little is known about the impact of light or circadian rhythms in human pregnancy or on long-term fetal programming, although there appear to be only small risks of negative reproductive health outcomes associated with shift work <sup>11</sup>. During pregnancy, the placenta acts as a mediator between the maternal and fetal environment to regulate growth and development; yet, little attention has been paid to the impact of circadian disruption on placental function. Variation in DNA methylation patterns in the placenta can affect placental function. Because the placenta is composed of fetal DNA, methylation of placental tissue may reflect fetal exposures and future health effects. In this study, we conducted an epigenome-wide association study (EWAS) to investigate whether night shift work is associated with differences in DNA methylation in the placental epigenome, which can impact long-term health outcomes in the offspring.

## METHODS

### Study population – The Rhode Island Child Health Study (RICHS)

RICHS is a hospital-based cohort study of mothers and infants in Rhode Island, described in detail in <sup>12</sup>. Briefly, from 2009 to 2014, 844 women between the ages of 18-40 and their infants were enrolled at the Women and Infants Hospital of Rhode Island, oversampling for large and small for gestational age (LGA, SGA, respectively) infants and matching each to an appropriate for gestational age (AGA) control by maternal age ( $\pm 2$  years), sex, and gestational age ( $\pm 3$  days). RICHS enrolled only full-term ( $\geq 37$  weeks), singleton deliveries without congenital or chromosomal abnormalities. Demographic information was collected from a questionnaire administered by a trained interviewer and clinical outcome information was obtained from medical records. Information on night shift work was obtained from questionnaire by first asking, "Have you ever worked outside the home? (Yes/No)" and if "Yes", participants were asked "If yes, please list all of the jobs

you have had starting with your current job first. Please indicate whether you worked a swing shift or a night shift on any of these jobs". To indicate shift jobs, the questionnaire included check boxes for "Yes" and "No" under a category for "Night Shift". For this analysis, only the most recently reported job history was used. To adjust for socioeconomic factors while avoiding multicollinearity, we used an adversity score index to adjust for household income, maternal education, marital status and partner support. The cumulative risk score ranged from 0 to 4, with 0 representing the lowest level of adversity and 4 representing the highest level of adversity. A higher risk score was given to women whose median household income (adjusting for the number of people in the household) fell below the federal poverty line for the year the infant was born (+1), to women whose household was larger than 6 (+1), to women who were single and did not receive support from a partner (+1) and to women whose highest level of education was high school or less (+1) <sup>13</sup>.

## Placental sample collection and measurement of DNA methylation

Genome-wide DNA methylation arrays were obtained on 334 placenta parenchyma samples in RICHs as previously described <sup>14</sup>. The QA/QC process has been described elsewhere <sup>15</sup>, including functional normalization, BMIQ, and 'ComBAT' to adjust for technical variations and batch effects in R <sup>14 16</sup>. Briefly, we used the 'minfi' package in R to convert the raw methylation files to  $\beta$  values, a ratio of methylation ranging from 0 to 1, for analysis. Probes associated with the X or Y chromosomes, SNP-associated (within 10bp of the target CpG and with minor allele frequency >1%), identified as cross-reactive or polymorphic by Chen et al <sup>17</sup>, or with poor detection p-values were excluded, yielding 334,692 probes for analysis in this study <sup>14</sup>. DNA methylation array data for RICHs can be found in the NCBI Gene Expression Omnibus (GEO) database with the number GSE75248. Women with missing information on pre-pregnancy smoking status ("No"/"Yes"), defined as smoking 3 months prior to pregnancy, or adversity score were not included in the analysis. Women who did not provide an answer for the nightshift variable (n=16) were recoded to "No". This study included the 237 mother-infant pairs within RICHs for which gene methylation data and the necessary demographic information were available.

## Placental RNA sequencing

Gene expression was measured using the Illumina HiSeq 2500 system in 199 placental samples from RICHs; methods have been previously described <sup>18</sup>. After standard QA/QC procedures, final data were normalized to log2 counts per million (logCPM) values. Raw data is available in the NCBI sequence read archive (SRP095910).

## Statistical analyses

Because a reference panel for placental cell types does not yet exist, we used a validated reference-free method, the 'RefFreeEWAS' package in R, to adjust for heterogeneity in cell-type composition <sup>19 20</sup>. We implemented the RefFree estimation via the same process described in detail in our lab's prior work <sup>21</sup>, and identified 8 components to represent the putative cell mixture in our placental samples. We also examined the outlier screening plots of the cell mixture array for extreme outliers. We then conducted an EWAS using robust linear modeling by regressing CpG methylation  $\beta$ -values on night shift work ("No"/"Yes"), adjusting for putative cell mixture, maternal age (years), pre-pregnancy smoking status ("No"/"Yes") adversity score (0-4) <sup>22</sup>, and sex of the infant ("Female"/"Male"). To adjust for multiple comparisons, we used the Bonferroni method and the Benjamini and Hochberg (BH) false discovery rate (FDR) methods. To evaluate the extent of *in utero* night shift exposure, we compared job and delivery date data. A secondary analysis using data from women who provided night shift job information (n=221) without recoding missing to "No" was also performed. While the distributions of gestational diabetes mellitus (GDM) and BMI differed between non-night shift and night shift workers, they were not adjusted for because they may be part of the causal pathway; numerous studies have found night shift work is associated with the development of obesity and metabolic diseases <sup>23 24</sup>. We also investigated differentially methylated regions (DMRs) using the 'Bumphunter' package in R <sup>25</sup>. We modeled the  $\beta$ -values between non-night shift workers and night shift workers, controlling for the same variables as the

EWAS. CpG sites within 500 base pairs were clustered together and  $\beta$ -values were modeled against a null distribution generated via bootstrapping; sites with differential methylation of 2% or more were considered to be possible DMRs.

To examine the functional implications of night shift work-associated DNA methylation (BH  $q < 0.05$ ), we also conducted an expression quantitative trait (eQTM) analysis using 'MEAL' in R to investigate whether methylation was associated with gene expression in the RICHs samples with methylation and expression data available ( $n = 199$ ). Using robust linear modeling, we regressed the expression levels of genes within a 100kb window of the CpG site on methylation  $\beta$ -values ( $p < 0.05$ ).

## Bioinformatic analyses

To better understand the biological significance of the EWAS results, we analyzed the association of the top 298 CpG sites (BH  $q < 0.10$ ) with GO-terms and KEGG pathway enrichment in R using the 'missMethyl' package <sup>26</sup>. We analyzed the top 298 sites (BH  $q < 0.10$ ) because enrichment analyses generally require a few hundred genes to determine common pathways. We also evaluated genes of CpG sites with BH  $q < 0.05$  for rhythmicity with the CircaDB database <sup>27</sup>. We analyzed all CIRCA experimental data using the JTK filter with a  $q$ -value probability cut-off of 0.05 and a JTK phase range of 0-40 <sup>28</sup>. To investigate whether the EWAS results were associated with previous GWAS findings, the genes of CpG sites and DMRs that were significant after Bonferroni adjustment ( $p < 0.05$ ) were compared to gene results ( $p < 5 \times 10^{-8}$ ) in the GWAS catalog of the National Human Genome Research Institute and the European Bioinformatics Institute (NHGRI-EBI) <sup>29</sup>.

## RESULTS

### Demographic and medical information

Demographic information for the women included in the methylation analysis is provided in **Table 1**.

**Table 1. Demographic characteristics of those included in the EWAS methylation analysis ( $n = 237$ ) by night shift work status. An asterisk (\*) signifies a significant difference ( $p$ -value  $< 0.05$  using either  $\chi^2$  test, Fisher's exact test or 2-sided t-test) between non-night shift and night shift workers.**

	N	Non-night shift ( $n = 184$ )	Night shift ( $n = 53$ )	Statistical significance
<b>Maternal age *</b>	237	28.0/31.0/34.0 30.7+/- 5.4	25.0/29.0/32.0 28.8+/- 5.1	$t = 2.369$ , $df = 87.668$ , $p$ -value = 0.020
<b>Maternal BMI</b>	235			$\chi^2 = 5.925$ , $df = 2$ , $p$ -value = 0.052
Normal		52% (95)	42% (22)	
Overweight		25% (45)	17% (9)	
Obese		23% (43)	40% (21)	
<b>Pre-pregnancy smoking status *</b>	237	12% (22)	25% (13)	$\chi^2 = 4.216$ , $df = 1$ , $p = 0.040$
<b>Gestational diabetes *</b>	234	9% (16)	21% (11)	$\chi^2 = 4.905$ , $df = 1$ , $p = 0.027$
<b>Maternal history of type 2 diabetes</b>	235	1% (2)	0% (0)	Fisher's, $p = 1$
<b>Maternal history of type 1 diabetes</b>	234	1% (1)	2% (1)	Fisher's, $p = 0.396$
<b>Maternal Ethnicity</b>	237			Fisher's, $p = 0.730$
Asian		5% (9)	2% (1)	
Black		7% (13)	9% (5)	
Indian		1% (1)	0% (0)	
More than one		3% (5)	2% (1)	
Other		9% (16)	15% (8)	
Unknown		2% (3)	0% (0)	
White		74% (137)	72% (38)	
<b>Marital status *</b>	237			Fisher's, $p = 0.014$
Single, never married		24% (44)	43% (23)	
Separated or divorced		3% (6)	4% (2)	
Married		73% (134)	53% (28)	
<b>Household income *</b>	229			$\chi^2 = 15.547$ , $df = 4$ , $p = 0.004$
<\$9-14,999		14% (25)	24% (12)	
\$15-29,999		9% (17)	20% (10)	
\$30-49,999		10% (18)	18% (9)	
\$50-99,999		39% (69)	30% (15)	
>\$100,000		28% (50)	8% (4)	
<b>Adversity score *</b>	237			Fisher's, $p = 0.017$
0		78% (143)	58% (31)	

	1	12% (22)	30% (16)	
	2	9% (16)	9% (5)	
	3	1% (2)	2% (1)	
	4	1% (1)	0% (0)	
<b>Maternal education *</b>	237			Fisher's, p = 0.001
<11 <sup>th</sup> grade		5% (10)	4% (2)	
High school		15% (28)	26% (14)	
Junior college or equivalent		22% (40)	40% (21)	
College		36% (67)	26% (14)	
Any post-graduate		21% (39)	4% (2)	
<b>Maternal chronotype</b>	237			$\chi^2 = 8.807$ , df = 4, p = 0.066
Definitely morning		29% (54)	21% (11)	
Somewhat morning		24% (45)	15% (8)	
Neither		8% (15)	15% (8)	
Somewhat evening		12% (22)	8% (4)	
Definitely evening		26% (48)	42% (22)	
<b>Infant sex (male)</b>	237	48% (88)	53% (28)	$\chi^2 = 0.236$ , df = 1, p = 0.627
<b>Infant birthweight (grams)</b>	237	3031/3565/4112	2750/3355/4150	t = 1.097, df = 80.478, p = 0.276
		3560+/- 723	3430+/- 768	
<b>Infant birthweight group</b>	237			$\chi^2 = 0.859$ , df = 2, p = 0.651
<=10% (SGA)		21% (38)	23% (12)	
11-89% (AGA)		47% (86)	40% (21)	
>=90% (LGA)		33% (60)	38% (20)	
<b>Time of delivery (hour)</b>	236	8.0/11.0/13.0	9.0/11.0/12.0	t = 0.528, df = 81.025, p = 0.599
		10.7+/- 2.7	10.4+/- 2.8	
<b>Sample collected (hour)</b>	236	10.0/11.0/14.0	10.0/12.0/14.0	t = 0.129, df = 84.417, p = 0.897
		11.8+/- 2.8	11.8+/- 2.8	

Comparing results from women who provided night shift job information (n=221) without recoding missing to “No” did not indicate any large differences in demographic features (**Tables S1**). Overall, women who reported working the night shift were more likely to be younger, smokers pre-pregnancy, cases of GDM, single and never married, lower household income, and higher adversity (p<0.05). While not statistically significant, women who worked the night shift trended towards a higher BMI and an evening chronotype. Of those included in the analysis, one participant reported taking melatonin and she was not a night shift worker. Additionally, 37 out of the 53 (70%) night shift workers reported working the night shift during pregnancy.

## Epigenome-wide methylation associations

DNA methylation at 298 CpG sites was found to be significantly different in night shift workers after FDR correction at the BH q<0.10 (**Tables S2 and S3**), 57 CpG sites significant at the BH q<0.05 (**Table 2**), and 10 CpG sites at the Bonferroni-corrected p<0.05 (**Table 2**).

**Table 2.** List of differentially methylated CpG sites in night shift workers compared to non-night shift workers after epigenome-wide analysis (BH q<0.05).

UCSC Gene Name	Chromosome	Position	Probe ID	UCSC Gene Group	Enhancer	$\beta_1$	SE	P-value	BH q-value	Bonferroni
NAV1	chr1	201708718	cg14168733	TSS1500	NA	-0.040	0.007	2.53E-08	0.003	0.008
NAV1	chr1	201709135	cg14377596	1stExon	TRUE	-0.040	0.007	2.98E-08	0.003	0.010
SMPD1	chr11	6412852	cg14814323	Body	NA	-0.016	0.003	2.97E-08	0.003	0.010
NAV1	chr1	201709390	cg01411786	Body	TRUE	-0.032	0.006	9.91E-08	0.004	0.033
TAPBP	chr6	33273011	cg03190911	Body	NA	-0.014	0.003	9.94E-08	0.004	0.033
	chr6	27390647	cg06667732		NA	-0.023	0.004	9.35E-08	0.004	0.031
CLEC16A	chr16	11073063	cg08082763	Body	TRUE	-0.023	0.004	7.21E-08	0.004	0.024
DIP2C	chr10	560669	cg21373996	Body	NA	-0.019	0.004	1.06E-07	0.004	0.035
FAM172A	chr5	93076910	cg25342875	Body	NA	-0.024	0.004	9.46E-08	0.004	0.032
PLEKHG6	chr12	6436676	cg14858786	Body	NA	-0.026	0.005	1.42E-07	0.005	0.047
KRT15	chr17	39675154	cg11983245	5'UTR	NA	-0.024	0.005	1.84E-07	0.005	0.062
NAV1	chr1	201709675	cg18539461	Body	TRUE	-0.036	0.007	1.71E-07	0.005	0.057
RHOT2	chr16	717556	cg04365973	TSS1500	NA	-0.019	0.004	2.58E-07	0.007	0.086
NAV1	chr1	201708888	cg13877974	TSS200	NA	-0.043	0.009	4.11E-07	0.010	0.137
ERI3	chr1	44716226	cg24373865	Body	NA	-0.024	0.005	5.66E-07	0.013	0.189
PTPN6	chr12	7060187	cg23147227	TSS1500	NA	-0.020	0.004	8.98E-07	0.019	0.301
EGFL8	chr6	32135718	cg08759957	Body	NA	-0.021	0.004	1.22E-06	0.023	0.407
ZBTB22	chr6	33284168	cg14771240	Body	NA	-0.020	0.004	1.18E-06	0.023	0.396



	chr10	22725309	cg01422243		NA	-0.019	0.004	1.51E-06	0.027	0.504
<i>UBR5</i>	chr8	103344822	cg02530407	Body	TRUE	-0.019	0.004	1.68E-06	0.027	0.561
<i>HDAC4</i>	chr2	240213173	cg23601374	Body	TRUE	-0.017	0.004	1.64E-06	0.027	0.549
	chr7	25702848	cg03700230		NA	0.048	0.010	1.93E-06	0.029	0.645
<i>CYB5R2</i>	chr11	7694163	cg05919312	5'UTR	NA	-0.018	0.004	2.03E-06	0.030	0.679
<i>RPS6KA4</i>	chr11	64139406	cg07425109	3'UTR	NA	-0.016	0.003	2.15E-06	0.030	0.719
<i>MSI2</i>	chr17	55742491	cg07618409	Body	TRUE	-0.020	0.004	2.26E-06	0.030	0.755
<i>CDYL2</i>	chr16	80716710	cg16713168	Body	TRUE	-0.021	0.004	2.33E-06	0.030	0.780
<i>FAM118A</i>	chr22	45705265	cg06575572	5'UTR	NA	-0.020	0.004	2.50E-06	0.030	0.835
<i>LRR2</i>	chr3	46618325	cg07225641	5'UTR	NA	-0.027	0.006	2.61E-06	0.030	0.875
<i>CLDN9</i>	chr16	3063894	cg10492999	1stExon	NA	-0.026	0.006	2.70E-06	0.030	0.905
<i>LOC645323</i>	chr5	87955859	cg13982098	Body	NA	-0.028	0.006	2.65E-06	0.030	0.886
<i>C2orf54</i>	chr2	241827789	cg21333033	Body	NA	-0.019	0.004	2.93E-06	0.032	0.981
<i>SLC41A1</i>	chr1	205780033	cg00762738	5'UTR	NA	-0.017	0.004	3.08E-06	0.032	1.000
<i>MXRA8</i>	chr1	1290712	cg00040588	Body	NA	-0.051	0.011	3.49E-06	0.032	1.000
<i>EGFL8</i>	chr6	32135715	cg12305588	Body	NA	-0.019	0.004	3.35E-06	0.032	1.000
<i>BAIAP2</i>	chr17	79022879	cg12472449	Body	NA	-0.016	0.004	3.43E-06	0.032	1.000
<i>FBXW7</i>	chr4	153437193	cg13536107	5'UTR	TRUE	-0.022	0.005	3.35E-06	0.032	1.000
<i>BAT2</i>	chr6	31599646	cg25371129	Body	NA	-0.005	0.001	3.61E-06	0.033	1.000
<i>MIRLET7A3</i>	chr22	46508563	cg04063235	TSS200	NA	-0.019	0.004	3.71E-06	0.033	1.000
<i>HDLBP</i>	chr2	242174625	cg11221200	Body	NA	-0.014	0.003	3.90E-06	0.033	1.000
	chr11	22454301	cg23181580		TRUE	-0.031	0.007	4.22E-06	0.035	1.000
<i>BATF3</i>	chr1	212874153	cg00168835	TSS1500	NA	0.005	0.001	4.42E-06	0.036	1.000
	chr22	50221949	cg08174792		NA	-0.034	0.007	4.91E-06	0.039	1.000
<i>MFHAS1</i>	chr8	8749074	cg01022370	1stExon	TRUE	-0.023	0.005	5.20E-06	0.040	1.000
<i>ZNF284</i>	chr19	44575547	cg05333740	TSS1500	NA	-0.023	0.005	5.83E-06	0.042	1.000
<i>DPEP2</i>	chr16	68027297	cg06866814	5'UTR	NA	0.002	0.000	5.50E-06	0.042	1.000
<i>GALNTL4</i>	chr11	11438208	cg16337763	Body	TRUE	-0.022	0.005	5.67E-06	0.042	1.000
<i>AZ1</i>	chr17	79184968	cg20296990	Body	NA	-0.020	0.004	5.84E-06	0.042	1.000
<i>GALNTL1</i>	chr14	69725831	cg00080706	TSS1500	NA	-0.019	0.004	5.99E-06	0.042	1.000
<i>MFHAS1</i>	chr8	8749278	cg01784220	1stExon	TRUE	-0.022	0.005	6.21E-06	0.042	1.000
<i>C11orf2</i>	chr11	64863151	cg13626866	TSS1500	NA	-0.026	0.006	6.37E-06	0.043	1.000
<i>BANF1</i>	chr11	65770987	cg17985854	Body	NA	-0.023	0.005	6.49E-06	0.043	1.000
<i>IQGAP2</i>	chr5	75784957	cg23289545	Body	TRUE	-0.019	0.004	6.62E-06	0.043	1.000
	chr17	43222106	cg00625783		TRUE	-0.025	0.006	7.26E-06	0.045	1.000
<i>TUBGCP2</i>	chr10	135120640	cg04070692	5'UTR	NA	-0.019	0.004	7.21E-06	0.045	1.000
<i>BAT1</i>	chr6	31502388	cg10895184	Body	NA	-0.018	0.004	7.53E-06	0.045	1.000
<i>HAPLN1</i>	chr5	83016779	cg18024167	1stExon	NA	-0.023	0.005	7.44E-06	0.045	1.000
<i>PKHD1L1</i>	chr8	110374866	cg19906741	1stExon	TRUE	0.018	0.004	7.77E-06	0.046	1.000

CpG sites for the *NAV1*, *SMPD1*, *TAPBP*, *CLEC16A*, *DIP2C*, *FAM172A*, and *PLEKHG6* genes had genome-wide significance after Bonferroni correction ( $p < 0.05$ ). The *ADAMTS10*, *CLEC16A*, *CTBP1*, *EGFL8*, *GNAS*, *HDAC4*, *HEATR2*, *KCNA4*, *KDEL2*, *MFHAS1*, *MXRA8*, *NAV1*, *PLXND1*, *UBR5*, *WNT5A*, and *ZBTB22* genes had multiple CpG sites represented in the results. There was an overall trend towards hypomethylation (**Figure 1a**). CpG sites for *NAV1*, *MXRA8*, *GABRG1*, *PRDM16*, *WNT5A*, and *FOXG1* were among the 10 sites with the most hypomethylation; CpG sites for *TDO2*, *ADAMTSL3*, *DLX2*, and *SERPINA1* were among the 10 sites with the most hypermethylation (**Table S3**).

The Manhattan plot of the results indicated a number of differentially methylated sites that were distributed across the genome with some occurring in the same regions (**Figure 1b**). To more rigorously examine this finding, we employed a 'Bumphunter' analysis and identified 6584 'bumps', with areas of the *NAV1*, *PURA*, *C6orf47*, and *GNAS* genes as DMRs (BH  $q < 0.10$ ) (**Table 3**).

**Table 3. 'Bumphunter' results of significant DMRs (BH  $q < 0.10$ ).**

Gene	Chromosome	Start	End	$\beta_1$	Area	L	clusterL	P-value	FWER	P-value Area	FWER Area	BH q-value	Bonferroni
<i>NAV1</i>	chr1	201708500	201709675	-0.038	0.452	12	12	3.29E-05	0.166	1.74E-04	0.606	0.054	0.217
<i>PURA</i>	chr5	139493486	139494006	-0.054	0.544	10	10	2.54E-05	0.131	6.80E-05	0.310	0.054	0.167
<i>C6orf47</i>	chr6	31627678	31627678	-0.112	0.112	1	38	3.24E-05	0.163	1.17E-02	1.000	0.054	0.213
<i>GNAS</i>	chr20	57463325	57463725	-0.034	0.482	14	30	9.35E-06	0.050	1.31E-04	0.512	0.054	0.062

Of these, CpGs for the *NAV1* and *GNAS* genes were also differentially methylated in the CpG by CpG analysis (Table S3).

A sensitivity analysis with GDM as a covariate shared many of the top CpG sites with the final results, suggesting GDM is not a main contributor to the findings (Table S4). An additional analysis evaluating GDM as the primary exposure shared no top genes with the EWAS results (BH  $q < 0.10$ , data not shown). In another sensitivity analysis comparing the beta coefficients of those with *in utero* night shift work exposure only ( $n=37$ ) to the beta coefficients of all night shift workers ( $n=53$ ), the differences were small; only 1 CpG site, cg24373865, had an absolute difference in beta coefficients greater than 0.01, at 0.011. We also re-examined our results removing those with missing data on shift work and the findings were substantially similar (Table S5).

## Functional analyses

Comparing the 298 significant CpG sites (BH  $q < 0.10$ ) to the remaining 334,394 CpG sites, there was a higher frequency of top CpG sites within enhancer regions ( $\chi^2 = 13.48$ ,  $df = 1$ ,  $p\text{-value} = 0.0002$ ). Because transcription factors (TFs) can bind to enhancer regions to alter gene expression, we assessed whether CpG methylation was associated with expression levels in nearby genes. The eQTM analysis found the expression of 18 genes to be associated with 14 CpG sites ( $p < 0.05$ ). Of these, the expression levels of *ACBD4* were associated with methylation in cg00625783 ( $\beta_1=2.515$ ,  $p\text{-value}=1.94\text{E-}05$ ) and the expression levels of *KRT15* were associated with methylation in cg11983245 ( $\beta_1=7.895$ ,  $p\text{-value}=8.04\text{E-}05$ ) (Table S6). For both of these genes, increasing methylation of the CpG sites was associated with increased gene expression. cg00625783 is not annotated to a gene but is located within an enhancer region and cg11983245 is annotated to the 5' untranslated region (5'UTR) and 1<sup>st</sup> exon of the *KRT15* gene. Methylation of cg11983245 was also associated ( $p < 0.05$ ) with increased *KRT19* ( $\beta_1=4.404$ ,  $p\text{-value}=3.87\text{E-}03$ ) and *LINC00974* ( $\beta_1=6.011$ ,  $p\text{-value}=3.40\text{E-}02$ ) expression levels.

We analyzed the top 298 CpG sites (BH  $q < 0.10$ ) for enrichment of KEGG pathways and GO-terms. The GO-terms “cell-cell adhesion”, “cell-cell adhesion via plasma-membrane adhesion molecules”, and “hemophilic cell adhesion via plasma membrane adhesion molecules” were found to be significant after FDR correction (BH  $< 0.05$ ) (Table S7). The top KEGG pathway results were “valine, leucine and isoleucine biosynthesis”, “mucin type O-glycan biosynthesis” and “melanogenesis”, but they were not significant after correcting for FDR (Table S7). Surprisingly, *PER1* was the only core circadian gene represented among the 298 CpG sites. However, we evaluated whether the 45 genes of the top 57 CpG sites exhibited circadian rhythmicity with the CircaDB expression database<sup>27</sup> and found 27 out of the 45 genes (60%) displayed rhythmic expression<sup>28</sup> (Table S8). Of these genes, *BAIAP2*, *GALNTL1*, *HDLBP*, *NAV1*, and *TAPBP* displayed rhythmicity in mouse SCN tissue. To explore the physiological role of the Bonferroni ( $p < 0.05$ ) and ‘Bumphunter’ significant genes, we queried the NHGRI-EBI GWAS catalog<sup>29</sup> for gene GWAS results with a  $p\text{-value}$  of  $5 \times 10^{-8}$  or less. The significant genes from the ‘Bumphunter’ analysis were associated with traits such as BMI, blood pressure, the immune system, and autism spectrum disorder or schizophrenia in the GWAS catalog (Table 4).

**Table 4. Query results of EWAS (Bonferroni  $p < 0.05$ ) and ‘Bumphunter’ significant genes in NHGRI-EBI GWAS catalog<sup>29</sup>. The gene name is the gene reported by the author(s) and/or the mapped gene and the listed traits are the traits reported by the author. GWAS results were filtered for significance with a  $p\text{-value}$  of  $5 \times 10^{-8}$ .**

Gene	Number of Associated GWAS Studies	Reported Trait(s) <sup>29</sup>
<i>NAV1</i>	4	BMI, BMI in physically active people, BMI adjusted for smoking, BMI in non-smokers, waist circumference
<i>SMPD1</i>	None	
<i>TAPBP</i>	1	Autism spectrum disorder or schizophrenia
<i>CLEC16A</i>	3	Selective IgA deficiency, sum eosinophil basophil counts, eosinophil counts, eosinophil percentage of white cells, allergic disease (asthma, hay fever, or eczema)
<i>DIP2C</i>	2	Blood metabolite levels, uric acid levels
<i>FAM172A</i>	None	
<i>PLEKHG6</i>	2	Colorectal cancer, primary biliary cholangitis



PURA	None	
C6orf47	4	Ulcerative colitis, inflammatory bowel disease, blood protein levels, autism spectrum disorder or schizophrenia, tuberculosis
GNAS	7	Platelet distribution width, diastolic blood pressure, systolic blood pressure, blood pressure, hypertension, renal function-related traits, BMI-adjusted waist circumference, DNA methylation

## DISCUSSION

We identified a number of CpG sites exhibiting differential methylation associated with night shift work in newborn placental tissue. While the average absolute effect estimates for the 298 CpG site corresponded to a roughly 1.7% change in methylation, even a small change in methylation may have physiologically-relevant effects<sup>30</sup>. The overall trend of hypomethylation with night shift work may be due to increased TF binding to DNA, leading to chromatin changes establishing the hypomethylated state<sup>31</sup>. Light at night (LAN) and night shift work can cause altered hormonal signaling and endocrine disruption; because hormone receptors can act as TFs, it is possible that circadian disruption causes increased hormonal signaling and increased TF binding.

Of the EWAS results, CpG sites for *NAV1* were consistently represented among the top results. The ‘Bumphunter’ analysis also found a DMR in *NAV1*. In general, the functions of *NAV1*, particularly in the placenta, are not well characterized. *NAV1* is homologous to the *unc-53* gene in *C.elegans*, which plays a role in axonal migration<sup>32</sup>. The mouse homolog also appears to play a role in neuronal migration; *NAV1* is enriched in growth cones and associates with microtubule plus ends<sup>33</sup>, and the deficit of *Nav1* causes loss of direction in leading processes<sup>34</sup>. Research has also found increased embryonic lethality, decreased birthweight, and infertility in female offspring for *Nav1*<sup>-/-</sup> mice<sup>35</sup>, suggesting an important role for *Nav1* in fetal development and health. In eye tissue, *Nav1* was associated with mural cells, a precursor of pericytes, and may play a role in angiogenesis<sup>35</sup>. In the embryonic retina, *Nav1* was downregulated in *Math5*<sup>-/-</sup> mice, a TF affecting RGC differentiation, which suggests it may be associated with RGCs<sup>36</sup>. Additionally, during embryonic development, *Nav1* was also found to be regulated by the TF PAX6, which has been implicated in sleep, brain and eye development, and metabolism, with *Pax6*<sup>-/-</sup> mice having significantly lower *Nav1* mRNA expression in lens placode compared to wild type mice<sup>37</sup>. When the CircaDB database of mouse tissue was queried, *Nav1* specifically displayed circadian rhythmicity in mouse SCN tissue (Table S8, JTK q<0.05). This suggests *NAV1* may play a role in the mammalian SCN.

A DMR was also identified in *GNAS*, which is imprinted in the paraventricular nucleus of the hypothalamus and encodes the G<sub>s</sub>α G-protein, which regulates cAMP generation and metabolism. *Gnas* is implicated in REM and NREM sleep and the browning of white adipose tissue for thermogenesis<sup>38</sup>. Additionally, in a microarray analysis of retina samples from an *rd/rd* mouse model, *Gnas* was implicated in melanopsin signaling<sup>39</sup>. Therefore, *GNAS* is likely important in integrating light and metabolic cues.

A possible limitation of this analysis is the moderate sample size of night shift workers (n=53). Additionally, the adjustment for cell-type heterogeneity is an estimation, so there is a possibility of residual confounding by cell type. Additionally, some of the women included as night shift workers did not have *in utero* exposure. Exposure to circadian disruption at different windows of development could have different magnitudes of effect. However, a sensitivity analysis of *in utero* night shift work exposure did not find large effect differences. Prior research has found that shift workers continue to have chronic health effects even after they switch to a day shift schedule. For example, researchers found that a history of shift work was associated with a decrease in cognitive ability that took 5 years or more after cessation of shift work to recover<sup>40</sup>. This suggests recovery from regular shift work may take an extended period of time and a history of shift work may have a prolonged influence on health.

This is the first study to examine the epigenetic impacts of night shift exposure on placental methylation in humans. Methylation of placental tissue, an indicator of the *in utero* epigenetic landscape, reflects functional effects on the placenta, which can impact various aspects of fetal development, including neurodevelopment. The findings that the methylation of *NAV1* differed by night shift work exposure and that *Nav1* is rhythmically expressed in mouse SCN suggests *NAV1* may play a developmental role in the human circadian system. Because the circadian system coordinates an array of physiological systems, alterations to circadian system development

could affect immune response, sleep patterns, behavior, metabolism, and future health status. We have found night shift work to be associated with changes in methylation of placental tissue, which has implications for fetal development and future health. However, these findings may also be relevant for people who experience circadian disruption due to common exposures such as LAN.

## CONCLUSION

Night shift work is associated with differential methylation patterns in placental tissue. NAV1 may be an important component in the development of the human circadian system. Night shift work is a complex exposure encompassing altered hormonal signaling, eating and activity patterns, light exposure, and sleep patterns. Therefore, it is difficult to tease apart which aspects of night shift work contribute to which result. However, night shift work is a prevalent exposure in the workforce and, more generally, circadian disruption is a common facet of modern life. These findings warrant further investigation to evaluate the effects of *in utero* circadian disruption and the epigenetic programming of the circadian system.

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**Data sharing:** The array data for RICHs can be found in the NCBI Gene Expression Omnibus (GEO) database with the number GSE75248 and the raw data for gene expression is available under the NCBI sequence read archive with the number SRP095910.

**Patient consent:** Written informed consent was obtained from each participant for all aspects of the study.

**Ethics approval:** RICHs and this analysis were approved by the Institutional Review Boards for the Women and Infants Hospital of Rhode Island and Emory University.

## BIBLIOGRAPHY AND REFERENCES CITED

1. Bass J, Lazar MA. Circadian time signatures of fitness and disease. *Science* 2016;354(6315):994-99. doi: 10.1126/science.aah4965 [published Online First: 2016/11/26]
2. Opperhuizen AL, van Kerkhof LW, Proper KI, et al. Rodent models to study the metabolic effects of shiftwork in humans. *Frontiers in pharmacology* 2015;6:50. doi: 10.3389/fphar.2015.00050 [published Online First: 2015/04/09]
3. Straif K, Baan R, Grosse Y, et al. Carcinogenicity of shift-work, painting, and fire-fighting. *The Lancet Oncology* 2007;8(12):1065-6. [published Online First: 2007/12/01]
4. Folkard S. Do permanent night workers show circadian adjustment? A review based on the endogenous melatonin rhythm. *Chronobiol Int* 2008;25(2-3):215-24. doi: 10.1080/07420520802106835
5. Chang AM, Aeschbach D, Duffy JF, et al. Evening use of light-emitting eReaders negatively affects sleep, circadian timing, and next-morning alertness. *Proc Natl Acad Sci U S A* 2015;112(4):1232-37. doi: 10.1073/pnas.1418490112
6. Berson DM, Dunn FA, Takao M. Phototransduction by retinal ganglion cells that set the circadian clock. *Science* 2002;295(5557):1070-73. doi: 10.1126/science.1067262
7. Panda S. Circadian physiology of metabolism. *Science* 2016;354(6315):1008-15. doi: 10.1126/science.aah4967 [published Online First: 2016/11/26]
8. Smarr BL, Grant AD, Perez L, et al. Maternal and Early-Life Circadian Disruption Have Long-Lasting Negative Consequences on Offspring Development and Adult Behavior in Mice. *Sci Rep* 2017;7:3326. doi: 10.1038/s41598-017-03406-4
9. Azzi A, Dallmann R, Casserly A, et al. Circadian behavior is light-reprogrammed by plastic DNA methylation. *Nat Neurosci* 2014;17(3):377-82. doi: 10.1038/nn.3651
10. Varcoe TJ, Wight N, Voultsios A, et al. Chronic Phase Shifts of the Photoperiod throughout Pregnancy Programs Glucose Intolerance and Insulin Resistance in the Rat. *PLoS One* 2011;6(4) doi: 10.1371/journal.pone.0018504
11. Palmer KT, Bonzini M, Harris EC, et al. Work activities and risk of prematurity, low birthweight and pre-eclampsia: an updated review with meta-analysis. *Occupational and environmental medicine* 2013;70(4):213-22. doi: 10.1136/oemed-2012-101032
12. Appleton AA, Murphy MA, Koestler DC, et al. Prenatal Programming of Infant Neurobehaviour in a Healthy Population. *Paediatric and perinatal epidemiology* 2016;30(4):367-75. doi: 10.1111/ppe.12294 [published Online First: 2016/03/24]
13. Appleton AA, Armstrong DA, Lesseur C, et al. Patterning in Placental 11-B Hydroxysteroid Dehydrogenase Methylation According to Prenatal Socioeconomic Adversity. *PLoS One* 2013;8(9):e74691. doi: 10.1371/journal.pone.0074691
14. Paquette AG, Houseman EA, Green BB, et al. Regions of variable DNA methylation in human placenta associated with newborn neurobehavior. *Epigenetics* 2016;11(8):603-13. doi: 10.1080/15592294.2016.1195534
15. Maccani JZ, Koestler DC, Lester B, et al. Placental DNA Methylation Related to Both Infant Toenail Mercury and Adverse Neurobehavioral Outcomes. *Environ Health Perspect* 2015;123(7):723-9. doi: 10.1289/ehp.1408561 [published Online First: 2015/03/10]
16. Johnson WE, Li C, Rabinovic A. Adjusting batch effects in microarray expression data using empirical Bayes methods. *Biostatistics (Oxford, England)* 2007;8(1):118-27. doi: 10.1093/biostatistics/kxj037 [published Online First: 2006/04/25]
17. Chen YA, Lemire M, Choufani S, et al. Discovery of cross-reactive probes and polymorphic CpGs in the Illumina Infinium HumanMethylation450 microarray. *Epigenetics* 2013;8(2):203-9. doi: 10.4161/epi.23470 [published Online First: 2013/01/15]
18. Deyssenroth MA, Peng S, Hao K, et al. Whole-transcriptome analysis delineates the human placenta gene network and its associations with fetal growth. *BMC Genomics* 2017;18(1):520. doi: 10.1186/s12864-017-3878-0
19. Houseman EA, Kile ML, Christiani DC, et al. Reference-free deconvolution of DNA methylation data and mediation by cell composition effects. *BMC Bioinformatics* 2016;17(1):259. doi: 10.1186/s12859-016-1140-4
20. Teschendorff AE, Zheng SC. Cell-type deconvolution in epigenome-wide association studies: a review and recommendations. *Epigenomics* 2017;9(5):757-68. doi: 10.2217/epi-2016-0153
21. Everson TM, Punshon T, Jackson BP, et al. Cadmium-associated differential methylation throughout the placental genome: epigenome-wide association study of two US birth cohorts. *bioRxiv* 2017

22. Lam LL, Emberly E, Fraser HB, et al. Factors underlying variable DNA methylation in a human community cohort. *Proc Natl Acad Sci U S A* 2012;109:17253-60. doi: 10.1073/pnas.1121249109
23. Froy O. Metabolism and Circadian Rhythms—Implications for Obesity. *Endocr Rev* 2010;31(1):1-24. doi: 10.1210/er.2009-0014
24. Scheer F, Hilton MF, Mantzoros CS, et al. Adverse metabolic and cardiovascular consequences of circadian misalignment. *Proc Natl Acad Sci U S A* 2009;106(11):4453-58. doi: 10.1073/pnas.0808180106
25. Jaffe AE, Murakami P, Lee H, et al. Bump hunting to identify differentially methylated regions in epigenetic epidemiology studies. *International journal of epidemiology* 2012;41(1):200-9. doi: 10.1093/ije/dyr238 [published Online First: 2012/03/17]
26. Phipson B, Maksimovic J, Oshlack A. missMethyl: an R package for analyzing data from Illumina's HumanMethylation450 platform. *Bioinformatics (Oxford, England)* 2016;32(2):286-88. doi: 10.1093/bioinformatics/btv560
27. Pizarro A, Hayer K, Lahens NF, et al. CircaDB: a database of mammalian circadian gene expression profiles. *Nucleic Acids Res* 2013;41(Database issue):D1009-13. doi: 10.1093/nar/gks1161 [published Online First: 2012/11/28]
28. Hughes ME, Hogenesch JB, Kornacker K. JTK\_CYCLE: an efficient non-parametric algorithm for detecting rhythmic components in genome-scale datasets. *J Biol Rhythms* 2010;25(5):372-80. doi: 10.1177/0748730410379711
29. MacArthur J, Bowler E, Cerezo M, et al. The new NHGRI-EBI Catalog of published genome-wide association studies (GWAS Catalog). *Nucleic Acids Res* 2017;45(Database issue):D896-D901. doi: 10.1093/nar/gkw1133
30. Breton CV, Marsit CJ, Faustman E, et al. Small-Magnitude Effect Sizes in Epigenetic End Points are Important in Children's Environmental Health Studies: The Children's Environmental Health and Disease Prevention Research Center's Epigenetics Working Group. *Environ Health Perspect* 2017;125(4):511-26. doi: 10.1289/EHP595
31. Martin EM, Fry RC. A cross-study analysis of prenatal exposures to environmental contaminants and the epigenome: support for stress-responsive transcription factor occupancy as a mediator of gene-specific CpG methylation patterning. *Environmental epigenetics* 2016;2(1) doi: 10.1093/eep/dvv011 [published Online First: 2016/04/12]
32. Maes T, Barcelo A, Buesa C. Neuron navigator: a human gene family with homology to unc-53, a cell guidance gene from *Caenorhabditis elegans*. *Genomics* 2002;80(1):21-30. [published Online First: 2002/06/25]
33. van Haren J, Draegestein K, Keijzer N, et al. Mammalian Navigators are microtubule plus-end tracking proteins that can reorganize the cytoskeleton to induce neurite-like extensions. *Cell motility and the cytoskeleton* 2009;66(10):824-38. doi: 10.1002/cm.20370 [published Online First: 2009/04/28]
34. Martinez-Lopez MJ, Alcantara S, Mascaro C, et al. Mouse neuron navigator 1, a novel microtubule-associated protein involved in neuronal migration. *Molecular and cellular neurosciences* 2005;28(4):599-612. doi: 10.1016/j.mcn.2004.09.016 [published Online First: 2005/03/31]
35. Kunert S. The role of neuron navigator 1 in vascular development. Humboldt-Universität zu Berlin, Mathematisch-Naturwissenschaftliche Fakultät I, 2014.
36. Mu X, Fu X, Sun H, et al. A gene network downstream of transcription factor Math5 regulates retinal progenitor cell competence and ganglion cell fate. *Dev Biol* 2005;280(2):467-81. doi: 10.1016/j.ydbio.2005.01.028 [published Online First: 2005/05/11]
37. Xie Q, Yang Y, Huang J, et al. Pax6 Interactions with Chromatin and Identification of Its Novel Direct Target Genes in Lens and Forebrain. *PLoS One* 2013;8(1):e54507. doi: 10.1371/journal.pone.0054507
38. Lassi G, Ball ST, Maggi S, et al. Loss of Gnas Imprinting Differentially Affects REM/NREM Sleep and Cognition in Mice. *PLoS Genet* 2012;8(5):e1002706. doi: 10.1371/journal.pgen.1002706
39. Peirson SN, Oster H, Jones SL, et al. Microarray analysis and functional genomics identify novel components of melanopsin signaling. *Current biology : CB* 2007;17(16):1363-72. doi: 10.1016/j.cub.2007.07.045 [published Online First: 2007/08/19]
40. Marquié J-C, Tucker P, Folkard S, et al. Chronic effects of shift work on cognition: findings from the VISAT longitudinal study. *Occupational and Environmental Medicine* 2015;72(4):258-64. doi: 10.1136/oemed-2013-101993

## Figure Legends.

Figure 1a. Volcano plot of effect size (beta coefficients,  $\beta_1$ ) and  $-\log_{10}$  P-values of night shift EWAS, adjusted for maternal age, pre-pregnancy smoking, adversity score, sex of the infant, and estimated cell mixture. Gray dots signify CpG sites with BH  $q < 0.05$  and CpG sites with both absolute beta coefficients of 0.03 or greater and BH  $q < 0.05$  are labelled with UCSC gene names.

Figure 1b. Manhattan plot of placental DNA methylation and night shift work EWAS, adjusted for maternal age, pre-pregnancy smoking, adversity score, sex of the infant, and estimated cell mixture. The dashed upper boundary line denotes p-value of  $1.49 \times 10^{-7}$  as the significance threshold after Bonferroni adjustment ( $p < 0.05$ ), the dashed middle boundary line denotes the p-value of  $7.7 \times 10^{-6}$  as the approximate significance threshold of BH  $q < 0.05$ , and the solid boundary line at denotes the p-value of  $8.8 \times 10^{-5}$  as the approximate significance threshold of BH  $q < 0.10$ .



## Volcano Plot of EWAS Results



