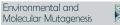
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Intergenerational arsenic exposure on the mouse epigenome and metabolic physiology

Mathia L. Colwell^{1,2} | Nicole Flack³ | Amanda Rezabek² | Christopher Faulk²

¹Department of Environmental Health Sciences, School of Public Health, University of Michigan, Ann Arbor, Michigan, USA

²Department of Animal Science, University of Minnesota College of Food, Agricultural and Natural Resource Sciences, Falcon Heights, Minnesota, USA

³Department of Veterinary and Biomedical Sciences, University of Minnesota College of Veterinary Medicine, Falcon Heights, Minnesota, USA

Correspondence

Mathia L. Colwell, Department of Environmental Health Sciences, School of Public Health, University of Michigan, Ann Arbor MI 48109 USA Email: mathiac@umich.edu

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Abstract

Inorganic arsenic (iAs) is one of the largest toxic exposures to impact humanity worldwide. Exposure to iAs during pregnancy may disrupt the proper remodeling of the epigenome of F1 developing offspring and potentially their F2 grand-offspring via disruption of fetal primordial germ cells (PGCs). There is a limited understanding between the correlation of disease phenotype and methylation profile within offspring of both generations and whether it persists to adulthood. Our study aims to understand the intergenerational effects of in utero iAs exposure on the epigenetic profile and onset of disease phenotypes within F1 and F2 adult offspring, despite the lifelong absence of direct arsenic exposure within these generations. We exposed FO female mice (C57BL6/J) to the following doses of iAs in drinking water 2 weeks before pregnancy until the birth of the F1 offspring: 1, 10, 245, and 2300 ppb. We found sex- and dose-specific changes in weight and body composition that persist from early time to adulthood within both generations. Fasting blood glucose challenge suggests iAs exposure causes dysregulation of glucose metabolism, revealing generational, exposure, and sex-specific differences. Toward understanding the mechanism, genome-wide DNA methylation data highlights exposure-specific patterns in liver, finding dysregulation within genes associated with cancer, T2D, and obesity. We also identified regions containing persistently differentially methylated CpG sites between F1 and F2 generations. Our results indicate the F1 developing embryos and their PGCs, which will result in F2 progeny, retain epigenetic damage established during the prenatal period and are associated with adult metabolic dysfunction.

KEYWORDS

arsenic, DNA methylation, DOHaD, germ cell, intergenerational, mouse

1 INTRODUCTION

Arsenic (iAs) exposure via drinking water is one of the largest global exposures affecting prenatal health, and has a growing body of evidence that suggests gestational iAs exposure influences intergenerational epigenetic inheritance (Hossain et al., 2017; National Research Council (U.S.). Subcommittee on Arsenic in Drinking Water, 1999; Nohara et al., 2011; Nohara et al., 2020). Data from mother-infant pairs show prenatal iAs exposure alters DNA methylation (Tsang et al., 2012; Xie et al., 2007; Zhao et al., 2002) and birth outcomes (Fei et al., 2013; Gilbert-Diamond et al., 2016; Laine et al., 2015; Marie et al., 2018) in newborns. Studies using rodent models find prenatal iAs exposure causes type 2 diabetes (T2D) (Liu et al., 2014; Young et al., 2018), obesity (Rivas-Santiago et al., 2019; Rodriguez et al., 2016), and differential methylation (Bailey et al., 2013; Martin et al., 2017; Rojas et al., 2015) in adult mice. However, most rodent

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studies use doses of iAs that are considered carcinogenic (Toxicological Profile For Arsenic|Enhanced Reader, n.d.); thus, the role of methylation and disease may be skewed, and methylation data reported from epidemiological studies are confounded by co-exposures. Due to its prevalence as a health threat in the United States and around the world, it is important to characterize the multigenerational epigenetic damage caused by iAs exposure and its long-term health effects in adult offspring. However, there are no studies to date investigating the intergenerational impact of maternal iAs exposure inherited through the female germline.

Epigenetic inheritance occurs by escape from complete epigenetic reprogramming. Epigenetic reprogramming is critical for regulation of gene expression and tissue differentiation in the developing embryo, and imperative for sex differentiation within primordial germ cells (PGCs) (Cedar & Bergman, 2012; Feng et al., 2010). Reprogramming occurs during early and late embryogenesis, where DNA methylation is erased then re-established in two waves of demethylation/re-methylation. The first reprogramming event takes place within the PGCs (F2, PGCs that result in the F2 progeny) of F1 offspring, and the second in the somatic F1 post-fertilization pluripotent stem cells (Sasaki & Matsui, 2008). Since the epigenome is vulnerable during development, disruption of epigenetic reprogramming by environmental exposures impacts both the developing zygote (F1 generation) and the PGCs (F2 generation), and is known to cause the onset of disease in subsequent generations (Painter et al., 2008; Skinner & Guerrero-Bosagna, 2014; Titus-Ernstoff et al., 2008). When the F1 PGCs develop into oocytes that will give rise to the F2 progeny, become fertilized, and undergo the waves of somatic demethylation/re-methylation as the developing F2 zygote during pregnancy, it is unclear if grand-maternal epigenetic damage caused by in utero iAs exposure will persist in adult F2s, despite the reprogramming event. Thus, we designed an exposure paradigm to many (1) identify the intergenerational epigenetic scarring caused by in utero iAs exposure and (2) assess the contribution to the onset of disease in adulthood using a mouse model.

The epigenetic damage of iAs exposure during pregnancy is hypothesized to be associated with iAs metabolism and the 1-carbon pathway. iAs metabolism acts as an inhibitor of DNA methylation due to competition for DNA methylation precursors. Biotransformation of iAs utilizes the one carbon metabolism pathway, the same metabolic pathway that generates methyl groups for the synthesis of DNA. The universal methyl donor, S-aenosyl-I-methione (SAM), is utilized by iAs-3-methyltransferase (AS3MT) to generate metabolites monomethylarsonic acid (MMA) and dimethylarsonic acid (DMA) for urinary excretion (Spratlen et al., 2017). SAM is also used by DNMTs to add methyl groups to convert unmodified cytosines to 5-methylcytosine (5-mC) (Thomas et al., 2004). Since both AS3MT and DNMT1 compete for SAM, evidence suggests utilization of SAM is responsible for the lower global methylation changes seen in exposed individuals(Nohara et al., 2011; Xie et al., 2007; Zhao et al., 2002). Thus, in utero iAs exposure has the potential to disrupt normal methylation patterning during epigenetic reprogramming.

We hypothesized the disruption of DNA methylation by in utero iAs exposure would target both the developing fetus (F1) and PGC (F2) epigenomes leading to the onset of adulthood diseases. We exposed F0 females to human-relevant doses of iAs throughout pregnancy that represented the WHO limit (10 ppb) (National Research Council (U.S.). Subcommittee on Arsenic in Drinking Water, 1999), the highest average global exposure (245 ppb) (Argos, 2015), and the known prenatal carcinogenic exposure in mice (2300 ppb) (Waalkes et al., 2004). Our gestational exposure window was chosen to restrict the impact of iAs to development and causally link early exposure to later-in-life metabolic diseases. We assess the effects of maternal in utero iAs exposure on eight, total body fat, behavior, and glucose tolerance at multiple time points into adulthood in both sexes of the F1 and F2 generations. We found arsenic causes physiological metabolic changes in a sex-, dose-, and generation-specific manner. Further, we found differentially methylated CpGs (DMCs) and differentially methylated regions (DMRs) that were unique or persisted between sex. dose. and generations, identifying DMCs within genes associated T2D or obesity.

MATERIALS AND METHODS 2

Mouse model, exposure, and breeding 2.1 paradigm

F0 C57BL/6J mice (n = 16, 8 weeks of age) were purchased from Jackson Laboratory. Mice were maintained on a 12:12 light/dark cycle and had ad libitum access to the standard lab diet Envigo 2018 (18% protein) chow and tap water (tested at 1 ppb iAs) until the beginning of the exposure. All animals were individually housed until breeding. Tap water from the University of Minnesota was treated with iAs^{III} (Millipore Sigma) at the following doses: 10, 245, and 2300 ppb. Water quality and arsenic concentrations were tested at Tri-City/ William Lloyd Analytical Lab. Female mice from the F1 generation (n = 4 per treatment) were used to produce the F2 generation. Mice used for glucose tolerance were selected at random for the initial 8-week measurement and remained enrolled for glucose tolerance at the 25- and 38-week time points throughout the study. The following quantity of mice survived the 40 weeks without health confounding factors, and were used for behavior testing and weekly body weight; (F1 1 ppb: male n = 10, female n = 7; F1 10 ppb: male n = 9, female n = 4; F1 245 ppb: male n = 8, female n = 4; F1 2300 ppb: male n = 7, female n = 7; F2 1 ppb: male n = 10, female n = 8; F2 10 ppb: male n = 10, female n = 12; F2 245 ppb: male n = 10, female n = 11; F2 2300: male n = 6, female n = 5). Total mice, litter size, and sex distribution are found in Table 1. Mice were treated humanely and in accordance with the Guidelines for the Care and Use of Laboratory Animals. The study protocol was approved by the University of Minnesota Institutional Animal Care and Use Committee (IACUC).

F1 offspring				F2 offspring			
F0 female breeders	Female	Male	Total	F1 female breeders	Female	Male	Total
1 ppb Breeder 1	2	1	3	1 ppb Breeder 1	1	2	3
1 ppb Breeder 2	3	2	5	1 ppb Breeder 2	4	3	7
1 ppb Breeder 3	2	4	6	1 ppb Breeder 3	3	5	8
1 ppb Breeder 4	4	3	7	1 ppb Breeder 4	-	-	-
10 ppb Breeder 1	3	4	7	10 ppb Breeder 1	4	2	6
10 ppb Breeder 2	3	4	7	10 ppb Breeder 2	5	4	9
10 ppb Breeder 3	2	1	3	10 ppb Breeder 3	3	4	7
245 ppb Breeder 1	3	4	7	245 ppb Breeder 1	5	1	6
245 ppb Breeder 2	3	3	6	245 ppb Breeder 2	4	4	8
245 ppb Breeder 3	3	0	3	245 ppb Breeder 3	2	5	7
2300 ppb Breeder 1	2	4	6	2300 ppb Breeder 1	4	2	6
2300 ppb Breeder 2	6	2	8	2300 ppb Breeder 2	1	4	5
2300 ppb Breeder 3	3	1	4	2300 ppb Breeder 3	-	-	-

2.2 | Glucose tolerance testing

At ages 8, 25, and 38 weeks, mice (n = 4 per treatment) underwent a 6-h fast before receiving a 100 µl glucose bolus by oral gavage. Blood glucose was monitored before glucose administration (0 min) and recorded every 15 min for 1.5 h by Accu-Check Blood Glucose Meter (Roche Diabetes Care). Statistical significance was determined by a two-way ANOVA (p < .05), and area under the curve (AUC) was calculated using the AUC function in Prism 9.

2.3 | Behavior testing

To assess anxiety and compulsive behavior, we performed marble burying in a standard marble burying condition. A set of 20 glass marbles were evenly spaced in a rat cage on sawdust bedding. For the duration of 20 min, mice explored and buried marbles undisturbed. Marbles that were at least two-thirds covered by corn cob bedding at the end of the testing period were counted as buried. The number of marbles buried was scored by a single skilled observer; inter-group differences were assessed by one-way ANOVA (p value < .05).

To assess spatial working memory and cognitive integrity, we performed a Y-maze behavior test (Hughes, 2004; Purpose and Cognition: The Determiners of Animal Learning, n.d.). Mice were placed in one arm of a standard Y-maze (MazeEngineers, Boston, MA), consisting of a high-walled chamber with three arms connected at 120°. Mice were video-recorded exploring the maze undisturbed for 10 min. After each test, the maze was cleaned with 70% ethanol and then water to eliminate confounding scents for the subsequent trials. The total number of maze arm entries and spontaneous alterations (SAs) were scored by a treatment-blinded skilled observer. SA percentage was calculated as the number of spontaneous alterations \div (number of entries – 2) × 100. One SA was counted when three consecutive entries into unique arms (e.g., A, B, and C) were recorded on video. Arm entries were recorded when the test mouse was positioned with all 4 ft inside the maze arm. The total number of arm entries was also recorded as a measure of exploration. Results were determined significantly by a one-way ANOVA (*p* value < .05).

2.4 | Body composition

After weaning (3 weeks of age), F1 and F2 mice were weighed weekly for 40 weeks. Final body weight was recorded at the time of euthanasia. Wet weight was recorded for adipose deposits on the right abdomen and the left kidney. To determine differences in body composition, ratios of body weight to adipose mass (body weight/fat pad mass) and body weight to kidney mass (body weight/kidney weight) were calculated and compared using a one-way ANOVA (*p* value < .05).

2.5 | DNA extractions and reduced represented bisulfite sequencing data analysis

Reduced represented bisulfite sequencing (RRBS) Genomic DNA from F1 and F2 liver was isolated using the Zymo Quick-DNA Mini-Prep Kit (Zymo Research). Upon submission, the gDNA samples were quantified using a fluorometric PicoGreen assay and analyzed for quality using the Nanodrop. All samples passed the Input DNA Requirements outlined in Tecan's Ovation RRBS Methyl-Seq System (Publication number: M01394v7). gDNA samples were converted to sequencing libraries using Tecan's Ovation RRBS Methyl-Seq with TrueMethyl oxBS System following the manufacturers' protocols (Part # 9522-A01). In summary, extracted gDNA was digested by Mspl and then ligated to indexed sequencing adapters. Libraries were bisulfite converted (oxBS module was not used) and amplified. Final library size distribution was validated using capillary electrophoresis and quantified using fluorimetry (PicoGreen). Libraries were pooled and sequenced on one lane of an S4 2×150 paired-end run on the Illumina NovaSeq 6000 System.

R (R Core Team (2021). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. https://www.R-project.org/.) and Linux command line tools were used for RRBS analysis. FastQC (version 11.3) assessed overall sequencing quality, and TrimGalore (Krueger, 2015) (version 4.5) with default settings trimmed low-quality bases, adapter sequences, and end-repair bases from 3' ends. Bismark (Krueger & Andrews, 2011) (version 19.0) aligned and called methylation. Reads were aligned to mm10 mouse reference genome using Bowtie2 (Langmead & Salzberg, 2012) (version 2.3.4) using default parameters. Methylation calls were reported for all nucleotides with the minimum read depth of 5. Paired raw fastq.gz reads were uploaded to the Sequence Read Archive and can be accessed under Bioproject number (PRJNA881737). The DSS package in R (version 2.32.0) detected differential methylation in liver for F1 and F2 females versus sex- and generation-matched 1 ppb controls (n = 3 per group) for sites with a minimum coverage of 10 reads. The DMLtest and callDML functions were used to detect DMCs. Smoothing was set to FALSE in DMLtest as is recommended for sparse data. DMRs were identified with callDMR. The p value threshold was set to FDR < 0.001 for both tests with delta > 0.01 required to call DMCs. A delta value was not enforced for DMRs; all CpGs within a potential DMR need not exhibit the same direction of change, meaning a delta threshold could lead to false negatives. The Annotatr R package was used to annotate DMCs and DMRs to CpG islands, genes, and intergenic regions in the mm10 genome (Cavalcante & Sartor, 2016). The COLWELL ET AL.

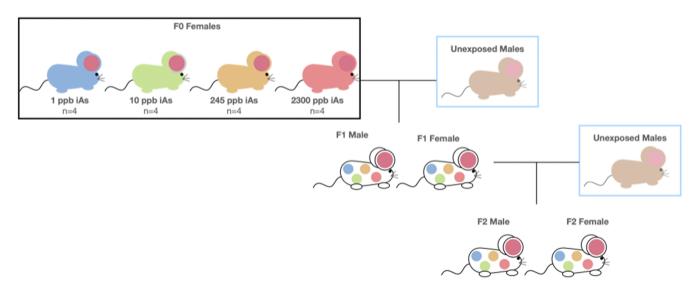
functions plot_annotation and plot_categorical functions were used to generate figures.

RESULTS 3 |

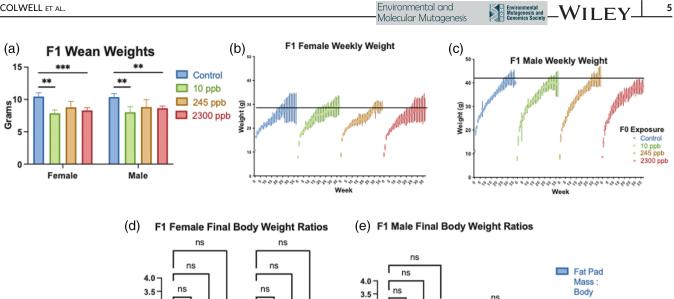
Maternal iAs exposure alters wean weights. 3.1 adult weight gain, and final body weight ratios in F1 offspring

To understand the effects of in utero iAs exposure on the developing embryo and PGCs, we exposed F0 females to iAs treated water for 2 weeks before mating (Figure 1). To ensure intergenerational effects were passed through the maternal line, unexposed male mice were placed in female cages to generate the F1 population. Male mice were separated from females after 48 h to avoid confounding paternal exposure. Treated water was removed within 24 h after the F1 offspring were born. F1 mice were weaned at 3 weeks of age and weekly weights were recorded until 40 weeks of age. F1 female mice (n = 4 per treatment, 8 weeks of age) were bred with unexposed, unrelated C57BL/6J male mice (8 weeks of age) to produce the F2 generation. F2 mice were weaned at 3 weeks of age. None of the F1 or F2 mice were directly exposed to iAs drinking water during their lifetime.

Male and female F1 mice born to mothers exposed to 10 and 2300 ppb had lower wean weights compared to control, where there were no significant differences in body weight to fat pad mass ratio (Figure 2a-e). Over a period of 40 weeks, male F1 offspring from 10 and 2300 ppb dams had lower final body weight compared to the control average (Figure 2c). Despite low final body weight, F1 male mice had no significant difference in weight gain for the 40 weeks, or body weight to fat pad mass ratio (Figure 2d,e). Thus, maternal exposure to iAs at three ranges alters wean weight and final body weight in a sex-specific manner.



Breeding and exposure paradigm F0 females (n = 4 per treatment) were exposed to iAs treated water 2 weeks before breeding FIGURF 1 and throughout pregnancy. FO females were bred with unexposed males to produce the F1 generation. F1 females were mated with unexposed males to produce the F2 generation. F1 and F2 mice did not consume iAs treated water.



3.0

2.5

2.0

1.5

1.0

0.5

0.0

control

Percent

Environmental and

FIGURE 2 F1 body weight and composition. (a) Wean weight of in utero exposed F1 mice was compared to F1 control mice. (b) F1 female weekly weight over a 40-week time period. (c) F1 male weekly weight over a 40-week time period. Fat pad mass to body weight ratio and kidney mass to body weight ratio percentages at 40 weeks of age. (c) Final body composition of F1 female mice. (d) Final body composition of F1 male mice. All data were compared with control, presented as mean ± SEM and determined significantly by a one-way ANOVA. ***p < .0001; standard error of mean (F1 1 ppb: male n = 10, female n = 7; F1 10 ppb: male n = 9, female n = 4; F1 245 ppb: male n = 8, female n = 4; F1 2300 ppb: male n = 7, female n = 7).

3.2 F1 females, but not males, show delayed glucose response similar to type II diabetes

3.0

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F0 Treatment

2300 Ppb

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Blood glucose was measured at three lifetime points (8, 25, and 38 weeks) to understand if in utero iAs exposure causes type II diabetes (T2D) associated phenotypes in F1 offspring (Figure 3a,b). Throughout the study, F1 female offspring of 10 ppb dams exhibited consistent (i.e., multi-timepoint) dysregulated metabolism of glucose during the 25- and 38-week oral glucose challenges (Figure 3a). F1 male mice exhibited no differences in glucose metabolism (Figure 3b). No significant differences were identified in AUC analysis. Our data indicate a low exposure of 10 ppb iAs during pregnancy contributes to irregular metabolism of glucose in late-life F1 females.

3.3 Grandmaternal iAs exposure alters lifelong body composition in male and female F2 offspring

Wean weights, weekly body weight, and final body composition in F2 offspring were altered by grandmaternal exposure (Figure 4a-e). Wean weights in F2 females were normal, whereas F2 10 ppb male mice had significantly higher wean weight compared to the F2 control (Figure 4b,c). F2 females had no significant differences in body weight through 40 weeks of age (Figure 4b). Remarkably, F2 male mice in all treatment groups had significantly lower body weight compared to control starting at Week 28 that persisted through 40 weeks of age (Figure 4c). Similar results were seen in body mass to fat pad mass ratios (Figure 4d,e). No significant differences were seen in F2 female final body composition (Figure 4d). Consistent with low body weight, F2 males in the 10 and 2300 ppb iAs exposure groups had significantly lower body mass to fat pad mass ratio (Figure 4e).

ns

ns

10 PPD

F0 Treatment

245 pp

2300 PPb

245 000

10 000

Weight

Kidney

Mass

Body

Weight

Dysregulated glucose metabolism in adult F2 3.4 male and F2 female mice

We measured blood glucose at three lifetime points (8, 25, and 38 weeks) within the F2 exposure groups to understand the effects of grandmaternal exposure on adult F2 diabetic phenotypes (Figure 5a,b). The FO exposure of 245 ppb causes a T2D phenotype in F2 female mice starting at an early age (3 weeks) that is persistent through adulthood (38 weeks) (Figure 5a). In addition, F2 male mice developed dysregulated glucose metabolism only in late life (38 weeks) (Figure 5b). We identified no significant changes in AUC analysis. These results indicate PGCs are

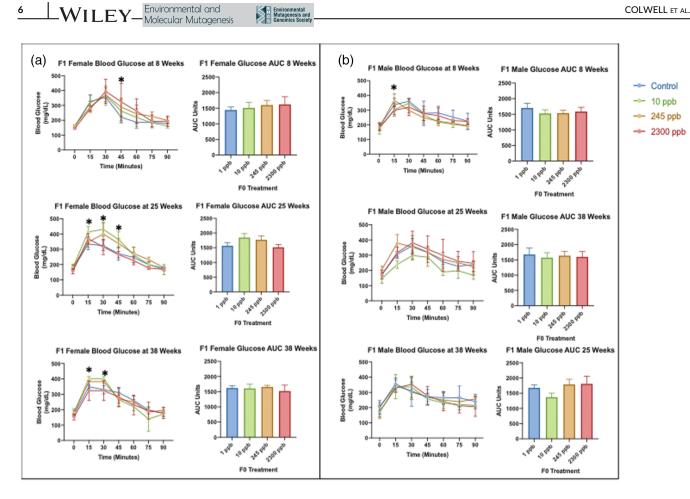


FIGURE 3 Blood glucose metabolism in F1 female and male mice. At ages 8, 25, and 38 weeks, the glucose tolerance test was performed after a 6-h fasting period followed by an oral administration of glucose. Tail vein blood glucose concentrations were measured at 0, 15, 30, 45, 60, 75, and 90 min. (a) F1 female mouse blood glucose data with area under the curve analysis. (b) F1 male blood glucose data with area under the curve analysis. All exposed data were compared to the control values to determine significant differences. Statistical significance between control and exposure groups were determined using a two-way ANOVA. *p < .05; standard error of mean (SEM) (n = 4 per sex, per treatment).

damaged by grandmaternal iAs exposure and result in adult metabolic disruption that is dose-specific.

3.5 | Maternal iAs exposure does not influence anxiety or memory behaviors in F1 or F2 offspring

We performed marble burying to determine whether gestational iAs exposure induced anxiety and a Y-maze assessment for spatial memory within adult F1 and F2 mice (Figure 6a,b). We identified no significant differences between exposure groups and control for either generation in both the marble burying and SA behavior in the Y-maze. These results suggest that in utero iAs exposure does not cause anxiety or impair spatial memory in F1 and F2 adult mice.

3.6 | Preferential hypermethylation in DMCs and DMRs of F1 generation

To identify a potential mechanism for epigenetic damage, we assessed DMCs and DMRs in the liver of 10 and 245 ppb F1 mice (Table 2)

since these dose levels are most relevant to human exposures. In the F1 animals of both sexes, we found an excess of hypermethylated CpG sites at both doses. Specifically, in F1 females, we found 259 (66%) hypermethylated and 132 (34%) hypomethylated DMCs within the 10-ppb group, and 157 (61%) hypermethylated and 99 (39%) hypomethylated DMCs within the 245-ppb group. Only 4 (80%) DMRs were hypermethylated and 1 DMR hypomethylated in the 10-ppb group, whereas no DMRs were identified in the 245-ppb group. In F1 males, we identified 299 (65%) hypermethylated and 196 (42%) hypomethylated DMCs within the 10-ppb group, and 265 (64%) hypermethylated and 151 (36%) hypomethylated DMCS in the 245-ppb group. Only 3 (43%) hypermethylated and 4 (57%) hypomethylated DMRs were identified in the 10 ppb iAs group, whereas 9 (82%) hypermethylated and 2 (18%) hypomethylated DMRs were identified in the 245-ppb group. Our results show F1 males have more dysregulated methylation in DMCs and DMRs compared to F1 females, regardless of the exposure group.

The F2 mice sustained similar overall numbers of DMC sites and recapitulated the hypermethylation bias seen in F1 animals. Our F2 results indicate higher DMC and DMR content within F2 females compared to F2 males, a contrast to the F1 generation (Table 2).

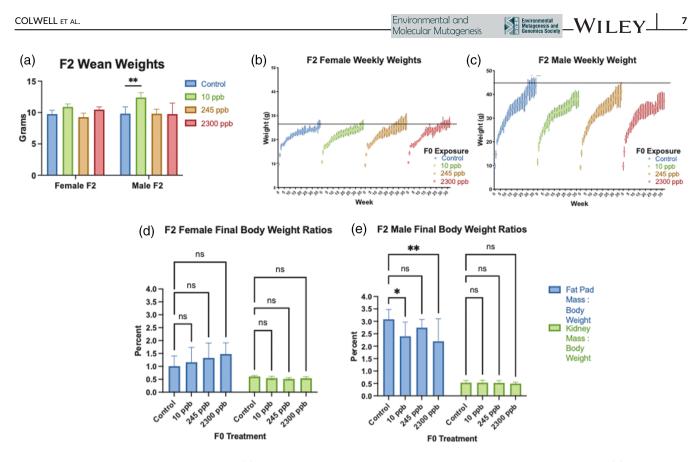


FIGURE 4 F2 body weight and composition. (a) Wean weight of in utero exposed F2 mice was compared to F2 control mice. (b) F2 female weekly weight over a 40-week time period. F2 male weekly weight over a 40-week time period. F2 male weekly weight over a 40-week time period. Fat pad mass to body weight ratio and kidney mass to body weight ratio percentages at 40 weeks of age. (c) Final body composition of F2 female mice. (d) Final body composition of F2 male mice. All data were compared to control, presented as mean ± SEM, and were calculated as significantly different by a one-way ANOVA. *p < .05. **p < .01; standard error of mean (F2 1 ppb: male n = 10, female n = 8; F2 10 ppb: male n = 10, female n = 12; F2 245 ppb: male n = 10, female n = 11; F2 2300: male n = 6, female n = 5).

Females of the 10-ppb group had preferential DMC hypermethylation in the liver, where 320 (60%) hypermethylated and 217 (40%) hypomethylated DMCs were identified in the 10-ppb group. We identified 9 (56%) hypermethylated and 7 (44%) hypomethylated DMRs within the 10-ppb group. Females of the 245-ppb group had a slight bias toward hypomethylation, where 208 (57%) hypomethylated and 156 (43%) hypermethylated DMCs were identified. We found 1 (20%) hypermethylated and 4 (80%) hypomethylated DMRs in the 245 ppb F2 females. Within the 10-ppb F2 males, we identified 140 (54%) hypermethylated and 118 (46%) hypomethylated DMCs, detecting 1 hypermethylated DMR. F2 males in the 245-ppb group had 144 (52%) hypermethylated and 133 (48%) hypomethylated DMCs, and only 1 hypermethylated DMR. Our findings show female DMCs and DMRs within the liver are more reactive to prenatal exposure to iAs in both the 10 and 245 ppb groups when compared to males.

3.7 | Generational intersect of DMCs and DMRs and top DMC-containing genes

The overlap of DMC-containing genes was assessed as no single-nucleotide positional DMCs were shared between generations. Overall, few DMC-containing genes were shared between F1 and F2,

with most overlapping genes occurring between iAs doses withingeneration (Figure 7). The top three greatest DMC-containing gene overlap was identified within generations and sex (Male F110 ppb vs. 245 ppb, Female F2 10 ppb vs. 245 ppb, and F1 Female 10 ppb vs. 245 ppb). We identified generational DMCs, finding males exposed to 245 ppb in utero had the highest overlap (eight DMCs) compared to other generational comparisons.

No DMR-containing genes were shared intergenerationally, although the genes Rlbp1 and Tcf4 both contained DMRs in 10 and 245 ppb F1 males (Data S1 "DMR_all_hits.xlsx"). The Rlbp1 DMRs were hypomethylated and the Tcf4 DMRs were hypermethylated in both doses relative to control. One hypermethylated intronic DMC was identified in Tcf4 in F2 females at the 245 ppb dose but not in any other F2 animals (Data S1 "DML_all_hits.xlsx"). No Rlbp1 DMCs were observed in the F2 generation. This data show limited persistence of DMCs and DMRs through the F1 to F2 generation.

3.8 | Differentially methylated genes and targeted genic regions for dysregulation

To evaluate the potential molecular consequences of iAs exposure, we identified genes containing DMCs and DMRs Identified genes 8

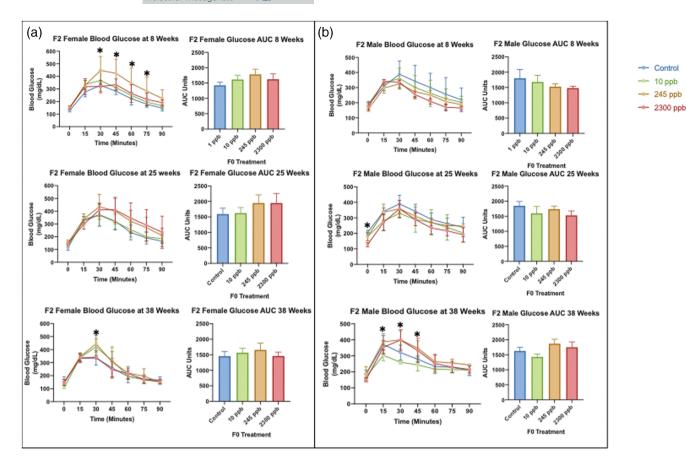


FIGURE 5 Blood glucose metabolism in F2 female and male mice. At ages 8, 25, and 38 weeks, the glucose tolerance test was performed after a 6-h fasting period followed by an oral administration of glucose. Tail vein blood glucose concentrations were measured at 0, 15, 30, 45, 60, 75, and 90 min. (a) F2 female mouse blood glucose data and area under the curve analysis. (b) F2 male blood glucose data and area under the curve analysis. All exposed data were compared to the control values to determine significant differences. Statistical significance between control and exposure groups were calculated by a two-way ANOVA, *p < .05; standard error of mean (SEM) (n = 4 per sex, per treatment).

were cataloged based on DMC content (Table 3) and the degree of differential methylation (Table 4). Several top DMC and DMR contain genes that were associated with neurological function, retina or cornea function, and anatomy, along with cell junction and membrane integrity. Genes identified with the greatest degree of differential methylation were associated with DNA helicase activity, cell adhesion, signal receptor binding, cell growth, muscle development, and nervous system development. In relation to the specific diseases connected to iAs exposure, we identified genes essential to liver function and associated diseases. For example, our data included genes associated with T2D and insulin resistance (Ali, 2013) such as Pik3c2g (Saeed, 2018) (F2 female 245 ppb), Ptprn2 (Ouni et al., 2020) (F2 male 245 ppb), Adcy5 (Sommese et al., 2018) (F1 male 10 ppb and F2 female 10 ppb), and Slc27a4 (Rowlands et al., 2014) (F1 male 10 ppb). We also found dysregulated methylation within Slc43a2 (Owaydhah et al., 2021) (F2 female 245 ppb) and Bicc1 (Park et al., 2016) (F1 male 245 ppb), genes that are associated with placental development and embryogenesis. Within the generational overlap, we found several genes associated with T2D such as Irs2 (10 ppb females), Tcf12 (245 ppb males), Jazf1 (10 ppb females), Adcy5 (10 ppb males), and Slc27a

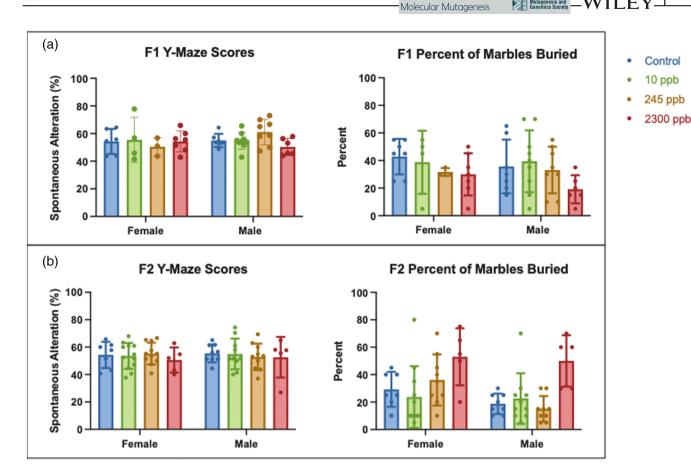
(245 ppb females) (Ali, 2013). We also identified genes associated with obesity, such as *Adipor1* (F1 female 10 ppb), *Fto* (F1 male 10 ppb), *Pparg* (F2 male 245 ppb), and *Adcy3* (F1 male 245 ppb) within our DMC list (Loos & Yeo, 2021).

Using AnnotatR, we determined introns, 1–5 kb regions, were the most sensitive while intergenic regions were most protected from iAs exposure by using the randomize regions function (Figure 8a–c). This function determines if differential methylation happens by chance or if specific regions are more responsive to exposure compared to others. Across sex, dose, and generation, we found introns and 1–5 kb regions contained more DMCs compared to other genic regions. In addition, intergenic regions were less sensitive to iAs exposure, indicating intergenic regions may be protected from dysregulation caused by iAs exposure.

4 | DISCUSSION

There is increasing evidence suggesting the early life environment has an impact on multigenerational health. This study characterizes the

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FIGURE 6 F1 male and female behavior data. (a) F1 Y-maze scores and percent of marbles buried. (b) F2 Y-maze scores and percent of marbles buried. Control values were compared to exposed data to determine significance (F1 1 ppb: Male n = 10, female n = 7; F1 10 ppb: Male n = 9, female n = 4; F1 245 ppb: Male n = 8, female n = 4; F1 2300 ppb: Male n = 7, female n = 7; F2 1 ppb: Male n = 10, female n = 8; F2 10 ppb: Male n = 10, female n = 12; F2 245 ppb: Male n = 10, female n = 11; F2 2300: Male n = 6, female n = 5).

intergenerational effects of iAs on offspring health, as inherited through the female germline. We provide evidence that iAs has sexspecific effects, such that dysregulated glucose metabolism affects both F1/F2 females and F2 males. More specifically, these effects were dose-specific with a late-life onset. We show exposed F1 and F2 males have impaired growth as indicated by weekly body weight, where only a subset of the exposed F2 female lineage had surpassed the weight gain of F2 female control mice. We probed DNA methylation changes as a potential mechanism for intergenerational damage. We found DMCs and DMRs occur across sex and generations, identifying genes with high DMC/DMR content. Together, this data suggest that in utero arsenic exposure alters metabolic phenotypes into adulthood despite the absence of iAs exposure during adult lifetimes though with limited recapitulation of methylation changes across generations. The effects presented are sex-, dose-, and generationspecific indicating the prenatal environment influences the onset of disease that persists throughout generations; more studies exploring the intergenerational effects on epigenetic inheritance and the associated functionality are needed to link iAs exposure to changes in gene expression and protein function.

It has been well-established that maternal in utero iAs exposure is associated with dysregulated glucose metabolism in the F1 populations (Navas-Acien et al., 2019; Tinkelman et al., 2020; Young et al., 2018). In a rat model, the maternal exposure of 500 ppb and 50 ppm throughout gestation and 2 months post-partum resulted in the onset of T2D phenotype in F1 offspring (Bonaventura et al., 2017). Mouse studies investigating in utero iAs of 100 ppb also identify diabetes-related phenotypes such as increased plasma insulin, decreased pancreatic insulin production, or the onset of non-alcoholic fatty liver disease within offspring (Ditzel et al., 2016; Huang et al., 2018; Martin et al., 2017). Unlike our study using standard chow, these studies are coupled with nutritional arms (total western diet or supplementation with folate) and doses of iAs that exceed doses found in drinking water. Our exposure paradigm was designed to identify if iAs targeted the developing fetus and germ cells using human relevant exposures. Thus, our results indicate the dysregulated glucose metabolism phenotype in F1 and F2 generations are present at doses relevant to the human population.

Our findings on generational impacts on body weight and fat mass contribute to the growing body of in utero iAs exposure on offspring metabolism. Our data show maternal exposure causes sex- and dose-specific weight changes in F1 offspring. Additionally, grand maternal exposure alters the lifelong body composition in both male and female F2 offspring. Previous studies using exposures ranging ons (http

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F1 generation: DMC a								
	10 ppb i	As			245 ppb	iAs		
5mC	Female		Male		Female		Male	
Differentially methyla	ated CpGs (DI	MCs)						
Hypermethylated	259	66%	299	65%	157	61%	265	64%
Hypomethylated	132	34%	196	42%	99	39%	151	36%
Total	391		462		256		416	
Differentially methyla	ated regions (DMRs)						
Hypermethylated	4	80%	3	43%	-	0%	9	82%
Hypomethylated	1	20%	4	57%	-	0%	2	18%
Hypomethylated Total	1 5	20%	4 7	57%	-	0%	2 11	18%
	5		7		-	0%		18%
Total	5	int in female a	7		- - 245 ppb			18%
Total	5 and DMR cou	int in female a	7		- 245 ppb Female			18%
Total F2 generation: DMC a	5 and DMR cou 10 ppb i Female	nt in female a	7 Ind male gran				11	18%
Total F2 generation: DMC a 5mC	5 and DMR cou 10 ppb i Female	nt in female a	7 Ind male gran				11	52%
Total F2 generation: DMC a 5mC Differentially methyla	5 and DMR cou 10 ppb i Female ated CpGs (DI	As MCs)	7 Ind male gran Male	d-offspring	Female	iAs	11 Male	
Total F2 generation: DMC a 5mC Differentially methyla Hypermethylated	5 and DMR cou 10 ppb i Female ated CpGs (DI 320	As MCs) 60%	7 and male gran Male 140	d-offspring	Female 156	iAs 43%	11 Male 144	52%
Total F2 generation: DMC a 5mC Differentially methyla Hypermethylated Hypomethylated	5 and DMR cou 10 ppb i Female ated CpGs (DI 320 217 537	MCs) 40%	7 and male gran Male 140 118	d-offspring	Female 156 208	iAs 43%	11 Male 144 133	52%
Total F2 generation: DMC a 5mC Differentially methylated Hypomethylated Total	5 and DMR cou 10 ppb i Female ated CpGs (DI 320 217 537	MCs) 40%	7 and male gran Male 140 118	d-offspring	Female 156 208	iAs 43%	11 Male 144 133	52% 48%
Total F2 generation: DMC a 5mC Differentially methylated Hypomethylated Total Differentially methylated	5 and DMR cou Female ated CpGs (DI 320 217 537 ated regions (MCs) 60% 40%	7 and male gran Male 140 118 258	54% 46%	Female 156 208 364	iAs 43% 57%	11 Male 144 133 277	52%

TABLE 2Differential methylationacross and generation sex and dose inliver tissue from F1 and F2 offspring andexposures 10 ppb and 245 ppb wereassessed for differential DNAmethylation as a result of in uteroexposure

Note: RRBS differential methylation was identified by DSS package in R studio. Statistical significance was determined by running a
Wald test in DSS, where data displayed have a p value threshold set to FDR < 0.001 ($n = 3$ per sex, per treatment).

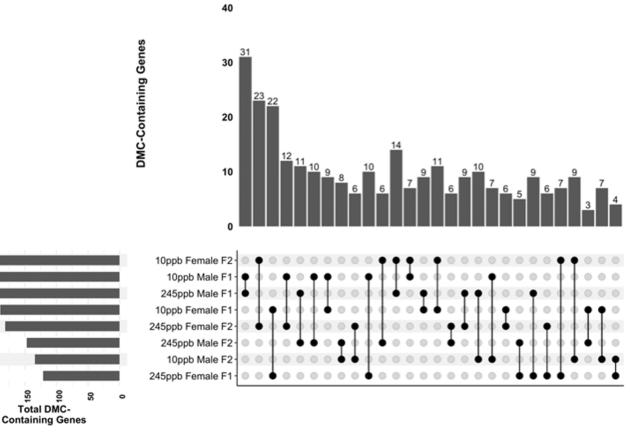


FIGURE 7 Shared DMC containing genes across sex and generation. DMC-containing genes were compared across sex and generation. Comparisons are shown from greatest gene overlap to least overlap (n = 3 mice per sex, per treatment).

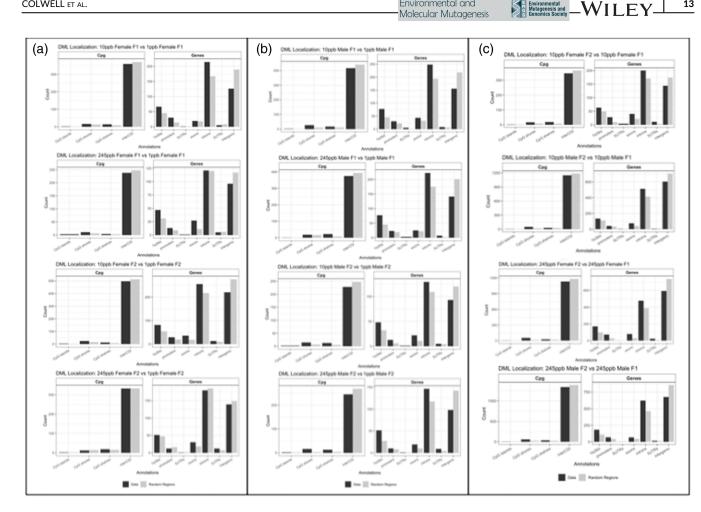
Top 10 D	MC-containi	Top 10 DMC-containing genes: RRBS													
F1 female 10 ppb	10 ppb			F2 female 10 ppb	10 ppb		ĺ	F1 female 245 ppb	245 ppb			F2 female 245 ppb	145 ppb		
Gene name	Number of DMCs	Mean differential methylation	ß	Gene name	Number of DMCs	Mean differential methylation	ß	Gene name	Number of DMCs	Mean differential methylation	ß	Gene name	Number of DMCs	Mean differential methylation	ß
Lrp1	8	62.89	5.28	Dok7	5	-62.24	2.73	Mad111	9	48.44	5.96	Tmem116	5	-55.60	7.31
Nav2	S	68.59	7.53	Eda	4	47.12	0.10	Nav2	4	69.01	8.64	Slc43a2	ო	55.32	0.85
Dock9	4	70.26	10.16	Fgf12	4	-75.69	4.59	Abcb10	2	-59.89	2.22	Cacna2d3	2	62.62	8.56
St3gal3	4	-66.09	7.30	Sox5	4	76.07	5.24	Dock9	2	72.59	15.58	Dip2c	2	-68.37	7.87
Tfr2	4	-48.99	1.16	Cacna1f	ю	47.52	5.61	Gpx1	2	61.04	4.50	Fam13c	2	64.70	0.00
Adgrg4	2	-59.26	2.03	CdkI5	e	46.04	4.35	Pde10a	2	83.57	11.72	Gipc3	2	-61.91	0.73
Afap112	2	-66.62	0.19	Entpd8	e	60.13	1.78	Pelp1	2	71.91	7.09	lgsf21	2	-54.54	9.16
Ak5	2	-3.46	88.54	Eya2	С	60.24	8.43	Shroom3	2	60.72	0.00	Kank1	2	-2.66	71.40
Cnksr2	2	76.22	0.62	Srgap1	ю	-27.63	88.53	Acacb	1	-67.30	AN	Pik3c2g	2	-40.53	2.78
Cpne8	2	-77.88	0.20	Ar	2	45.33	1.44	Acvr2b	1	73.53	AN	Slc7a5	2	-69.36	6.40
F1 male 10 ppb	0 ppb			F2 male 10 ppb	l0 ppb			F1 male	F1 male 245 ppb			F2 male 245 ppb	:45 ppb		
Gene name	Number of DMCs	Mean differential methylation	SD	Gene name	Number of DMCs	 Mean differential methylation 	S	Gene name	Number of DMCs	Mean differential methylation	S	Gene name	Number of DMCs	Mean differential methylation	ß
Tcf4	9	54.21	2.23	Agxt	2	-60.46	5.95	Acot7	9	57.15	2.33	Katna1	4	73.97	8.40
Rlbp1	4	-68.24	4.62	Btla	2	-76.68	00:0	D Tcf4	9	64.40	4.32	Pou2f1	4	0.19	79.44
Smarca2	4	-73.41	3.83	Clmn	2	65.79	2.91	. Bicc1	4	56.25	2.37	Abca4	e	-80.70	1.48
Atp8b4	С	48.99	5.33	Pcdhb16	2	57.51	3.38	lrag1	4	70.52	1.69	Ddx46	ю	-69.67	0.00
Cdk7	ю	51.65	2.05	Sptb	2	-66.01	2.31	. Cdk7	ю	50.44	0.68	Ptprn2	ю	-22.65	67.25
Gzmk	e	56.50	6.98	Acot7	1	69.40	NA	Erc2	ю	58.84	9.66	Spidr	2	63.63	1.33
Pkd1l3	з	44.99	2.86	Acox1	1	-71.26	NA	Pick1	ю	61.41	9.90	Zbtb12	2	-51.57	2.22
Slc27a4	ю	66.19	8.23	Adamts14	4 1	46.30	NA	Rlbp1	ю	-64.19	0.13	Acly	1	69.99	NA
Acp7	2	63.06	2.36	Aff3	1	-64.38	NA	Scpep1	ю	66.22	15.71	Acot7	1	70.18	NA
Adcy5	2	60.53	1.18	AI314278	8 1	68.79	NA	Atp8b4	2	49.51	2.14	Actr2	1	-59.52	NA
<i>Note</i> : Genes FDR < 0.00:	s are categori $1 (n = 3 per s)$	Note: Genes are categorized first by number of DMCs then in alphabetical or FDR < 0.001 ($n = 3$ per sex, per treatment).	ber of DMC t).	Cs then in all	phabetical or	der. Statistical sig	gnificance v	was determi	ned by runnir	ng a Wald test in I	DSS, wher	e data display	/ed have a <i>p</i>	der. Statistical significance was determined by running a Wald test in DSS, where data displayed have a p value threshold set to	it to

TABLE 3 Top 10 DMC-containing genes

	ndd				F2 fem	F2 female 10 ppb					F1 femak	F1 female 245 ppb					F2 fema	F2 female 245 ppb				
Genes with hypomethylated CpGs	ed CpGs	Genes with hypermethy	Genes with hypermethylated CpGs	GS	Genes with hypomethy	Genes with hypomethylated CpGs	CpGs	Genes with hypermeth	Genes with hypermethylated CpGs	pGs	Genes with hypomethy	Genes with hypomethylated CpGs	Gs	Genes with hypermeth	Genes with hypermethylated CpGs	pGs	Genes with hypomethy	Genes with hypomethylated CpGs	ő	Genes w	vith hyper	Genes with hypermethylated CpGs
Number Gene of name DMCs	ber Degree of differential S methylation	Gene name	Number of DMCs	 Degree of differential methylation 	Gene name	Number of DMCs	r Degree of differential methylation	Gene name	Number of DMCs	Degree of differential methylation	Gene name	Number of DMCs	Degree of differential methylation	Gene name	Number of DMCs	r Degree of differential methylation	Gene name	Number of DMCs	r Degree of differential methylation	Gene name	Number of DMCs	 Degree of differential methylation
Cpne8 2	-77.88	Trim33	1	92.35	Nub1	1	-83.85	Copz2	4	84.71	Mrc1	1	-77.94	Catspere1	1 1	92.43	Tenm2	7	-84.28	Ltbp1	1	85.44
Afap1 1	-76.12	Clec4a2	1	83.79	Kif4	2	-77.74	Myo1d	1	82.42	Slc14a2	1	-74.93	Pde10a	2	83.57	Fbxo42	1	-78.12	Mpzl3	1	78.48
Finc 1	-75.46	Qpctl	1	83.79	Clpx	1	-77.28	Ubr4	4	81.19	Ccni	1	-73.53	Fh11	Ļ	82.64	Srrm4	7	-77.31	lft43	1	78.05
Tox2 1	-74.79	Snrpd2	1	83.79	Tspan9	0 1	-77.22	Atp8a2	ر ا	78.00	Ptprd	1	-69.60	Rad54I2	1	80.39	Prickle2	1	-75.88	Ttc41	1	76.07
Usp7 1	-74.12	Glmn	1	83.78	Acacb	1	-76.50	Sox5	4	76.07	Camk 2a	1	-69.41	Myo18a	4	79.17	Trim11	Ļ	-75.71	Pitpnc1	1	75.98
Snf8 1	-74.08	Rad54l2	1	83.27	Fgf12	4	-75.69	Slx4ip	1	75.51	Gucy1a2	1	-68.64	Atp9b	1	77.84	Arhgap44	4 1	-75.54	Wdr26	7	75.74
Hpcal4 1	-73.40	Ppm1e	1	82.62	Rtel1	1	-75.64	Emilin1	1	73.99	Arhgef10	1	-68.39	Tmem94	1	75.28	Myo6	1	-75.47	Prorp	7	75.60
Cdr1 1	-73.05	Ggcx	1	82.03	Hdnr	1	-75.57	Recql5	4	73.69	Pde8b	1	-68.14	Agbl1	1	75.21	Alpk2	1	-74.79	Apcdd1	1	75.22
Cdr1os 1	-73.05	Catspere1	1 1	79.40	Zfp469	1	-75.48	Ptpn14	1	73.48	Muc3	1	-67.62	Sort1	1	74.55	Zmat4	1	-74.67	Tspear	7	74.55
Runx1 1	-72.18	Nsd2	1	78.53	Rps25	1	-74.92	Thap4	ب ا	73.33	Sorcs1	1	-67.42	Mtor	4	73.78	Gsk3b	4	-74.50	Snord92	1	72.71
F1 male 10 ppb	ą				F2 mak	F2 male 10 ppb					F1 mal	F1 male 245 ppb					F2 male	F2 male 245 ppb				
Genes with hypomethylated CpGs	ed CpGs	Genes with hypermethy	Genes with hypermethylated CpGs	CpGs	Genes with hypomethy	Genes with hypomethylated CpGs	CpGs	Genes with hypermethy	Genes with hypermethylated CpGs	pGs	Genes with hypomethy	Genes with hypomethylated CpGs	CpGs	Genes with hypermethy	Genes with hypermethylated CpGs	CpGs	Genes with hypomethy	Genes with hypomethylated CpGs	ge	Genes w	vith hyper	Genes with hypermethylated CpGs
Numbe Gene of name DMCs	Number Degree of of differential Gene DMCs methylation name	al Gene on name	Number of DMCs	r Degree of differential methylation	Gene name	Number of DMCs	r Degree of differential methylation	Gene name	Number of DMCs	r Degree of differential methylation	- Gene	Number of DMCs	er Degree of differential methylation	Gene n name	Number of DMCs	r Degree of differential methylation	Gene	Number of DMCs	r Degree of differential methylation	Gene name	Number of DMCs	r Degree of differential methylation
Foxi2 1	-83.85	lfitm3	4	79.14	Cdh13	7	-84.41	Scn8a	Ļ	84.25	Kcnn3	1	-76.86	Fam102a	la 1	77.90	Abca4	ю	-80.70	Smg6	1	76.13
Senp5 1	-76.06	Camk2b	b 1	77.24	Pum1	1	-80.87	Ddx39b	1	75.64	Sirt1	1	-76.16	Slc22a14	4 1	75.62	Smc1b	1	-77.68	Dapl1	1	76.06
Col23a1 1	-75.85	Zmiz1	2	74.89	Btla	2	-76.68	Mir8094	7	75.64	Sh3gl2	1	-76.06	lgsf3	1	75.46	Gse1	1	-77.15	Slc20a2	1	74.82
Cds1 1	-75.53	Hoxb7	1	74.89	Msi2	1	-76.25	H2-Aa	1	73.38	Rimklb	1	-75.43	Adam9	1	75.26	Lipc	1	-76.49	Poglut1	1	74.49
Fgfr4 1	-75.47	Hoxb8	1	74.89	Cep85I	1 1	-75.75	Kalrn	7	73.18	Fbxw10	0 1	-74.42	Katna1	1	74.04	Ehmt2	1	-75.52	Kpna4	1	74.24
Celf1 1	-75.22	Tesmin	1	74.89	Rpn2	1	-75.15	Pias2	1	73.18	Gba2	1	-72.95	Stip 1	1	73.62	Nub1	1	-70.93	Katna1	4	73.97
Scfd2 1	-73.45	lqschfp	1	73.42	Pgbd1	1	-74.88	Pdzd2	1	72.96	Pecr	1	-71.68	Recq15	1	72.25	Ddx46	ю	-69.67	Tspan15	1	73.44
Zdhhc17 1	-73.45	Ubac2	1	73.42	Deup1	1	-73.77	Nexmif	1	72.14	Grm1	1	-71.19	Smim5	1	72.25	Hs 6st3	1	-69.63	Hdac5	1	73.34
Smarca2 4	-73.41	Ccdc107	17 2	73.00	Hs3st6	1	-73.72	Tmem247	7 1	71.72	Ube2e1	1 1	-69.02	Irag1	4	70.52	Maml3	1	-68.20	Ndufaf2	1	72.74
Fbxw10 1	-72.11	Myo1d	1	72.75	Atp8a1 1	1 1	-73.34	Vps8	1	71.53	Dek	1	-68.96	Elob	1	70.49	Arhgap28	8 1	-68.18	Smim15	1	72.74

TABLE 4 Degree of differential methylation in DMC-containing genes

Environmental Mutagenesis and



FIGURF 8 Differentially methylated loci localization. (a) DML comparisons in females within generations. (b) DML comparisons in males within generations. (c) DML comparisons within sex across generations. Statistical significance was determined by running a Wald test in DSS, where data displayed have a p value threshold set to FDR.

from 10 ppb to 42.5 ppm show maternal iAs exposure contributes to higher body weight in female offspring (Rodriguez et al., 2016). Additionally, a study using a similar exposure of 250 ppb found no changes in body weight within the F1 offspring, early life stunted growth in F2 offspring that was restored in late life, and increased adiposity in F3 males (Gong et al., 2021). These studies focusing on rodent models are but a few indicating doses of iAs and sex have a role in adiposity. Further research is needed to identify the influence of iAs early and late-life body composition and the implications on human metabolic health.

Our data show glucose metabolism and body composition are dysregulated by in utero iAs exposure, yet it is unclear if the onset of disease in our study is influenced by epigenetic alterations or other factors. The onset of T2D in adults chronically exposed to iAs may be a result of non-epigenetic mechanisms such as inhibition of a GLUT4, damage to ß-cells by ROS production, inhibition of glucose-stimulated insulin secretion, or increased stimulation of liver gluconeogenesis (Martin et al., 2017). Chronic iAs exposure is known to cause increased body weight and alterations in lipid metabolism (Castriota et al., 2018; Rivas-Santiago et al., 2019). Additionally, studies show the maternal metabolic milieu influences the onset of obesity in

offspring ranging from adolescence and into adulthood (Derraik et al., 2015; Tequeanes et al., 2009). Despite the role of iAs in direct damage to tissues and the maternal metabolism, increasing evidence from mouse and human studies shows developmental iAs exposure alters DNA methylation in offspring with associations to metabolic phenotypes. In a longitudinal mother-child cohort in Bangladesh, differential methylation of CpG sites within in blood mononuclear cells were characterized, where 12 hypermethylated CpG sites were associated with prenatal iAs exposure (Gliga et al., 2018). Of the DMC sites, several were associated with proteins essential to insulin secretion by ß-cells in the pancreas. Similar to previously reported studies, our data finds differential methylation within genes associated with T2D (Ali, 2013) and obesity (Loos & Yeo, 2021)-Pik3c2g, Ptprn2, Adcy5, Slc27a4, Irs2, Tcf12, Jazf1, Adcy5, Slc27a, Adipor1, Fto, Pparg, and Adcy3. While in utero exposure is known to cause long-term health effects in developing fetuses and young children, often impacting the establishment of DNA methylation (Smeester & Fry, 2018), further research is needed to explore the association between our DMCs and the onset of disease. It is clear the identification of differential methylation in offspring is associated with maternal iAs exposure, where differential methylation may be a plausible contributor to

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the onset of metabolic phenotypes. Although our DMC and DMR data provide insight on genes associated with disease, the exploration of differential methylation on gene expression is still necessary. The correlation between DMCs and DMRs with dysregulated gene expression would potentially link our data with the characterized disease phenotypes. Additionally, the exploration of the physiological state of the liver, such as morphology, lipid accumulation, or RNA expression, could provide context if differential methylation is an artifact of cell composition. While levels of iAs that exceed our dose regimen (up to 85 ppm iAs) are known to cause intergenerational carcinogenesis targeting the liver, we found no visible carcinomas or immediate morphological changes such as color across treatment groups. Further investigation of liver morphology, cell composition, and RNA transcription is needed to understand the effects of iAs on hepatotoxicity and metabolism.

Sex-, dose-, and generation-specific effects are common among prenatal environmental exposures. Previous studies have reported similar effects of paternal iAs exposure on glucose homeostasis through the F1 and F2 generations (Gong et al., 2021). Glucose intolerance and hepatic insulin resent were present in F1 females, not F1 males, without altering body weight. Despite low body wean weights, body weight was restored into adulthood within the F2 generation. In contrast to our study, Gong et al. explore the transmission of metabolic phenotypes through the paternal lineage (mating F1 male offspring), not the maternal lineage (mating F1 female offspring). These results indicate a paternal epigenetic influence on adult F1 and F2 offspring metabolic disease phenotypes. Another study exploring both maternal and paternal exposure on transgenerational health found altered DNA methylation and reproductive phenotypes (Nava-Rivera et al., 2021). Methylation in the ovary and testes was significantly lower in the F1 generation, remained unchanged in the F2 generation. and increased within the F3 generation. Sperm quality parameters were also lower in the F1 and F3 male offspring, not within the F2 offspring. Results from these two intergenerational studies are foundational for the understanding of iAs exposure and epigenetic reprogramming. Our shared DMC containing genes across sex and generation provides an exciting foundation for genes targeted by prenatal iAs exposure. Despite the absence of overlap between generations, it is possible CpG methylation within the detected DMCs was re-established during F2 embryogenesis. Our findings could indicate female PGCs may retain epigenetic damage from FO exposure that persists regardless of epigenetic reprogramming. Our results join the few novel studies indicating intergenerational iAs exposure alters metabolic phenotypes and the epigenome in a sex-, dose-, and generation-specific manner.

AUTHOR CONTRIBUTIONS

Mathia L. Colwell and Christopher Faulk designed the study. Mathia L. Colwell, Nicole Flack, Amanda Rezabek, and Christopher Faulk performed the animal experiments. Mathia L. Colwell performed the molecular experiments. Nicole Flack performed the bioinformatics. Mathia L. Colwell and Nicole Flack wrote the manuscript. Christopher Faulk edited the manuscript. All authors have read and approved the manuscript prior to submission.

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CONFLICT OF INTEREST

The authors declare no conflict of interest.

ORCID

Mathia L. Colwell (D https://orcid.org/0000-0001-5008-8520) Nicole Flack D https://orcid.org/0000-0002-4213-5195 Christopher Faulk D https://orcid.org/0000-0002-9749-9658

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SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

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