Endocrine disruptor-induced epimutagenesis *in vitro*: Insight into molecular mechanisms

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Abstract:

Endocrine disrupting chemicals (EDCs) such as bisphenol S (BPS) are xenobiotic compounds that can

disrupt endocrine signaling following exposure due to steric similarities to endogenous hormones within

the body. EDCs have been shown to induce disruptions in normal epigenetic programming

(epimutations) that accompany dysregulation of normal gene expression patterns that appear to

predispose disease states. Most interestingly, the prevalence of epimutations following exposure to

many different EDCs often persists over multiple subsequent generations, even with no further

exposure to the causative EDC. Many previous studies have described both the direct and prolonged

effects of EDC exposure in animal models, but many questions remain about molecular mechanisms by

which EDCs initially induce epimutations or contribute to the propagation of EDC-induced epimutations

either within the exposed generation or to subsequent generations. Additional questions remain

regarding the extent to which there may be differences in cell-type specific susceptibilities to various

EDCs, and whether this susceptibility is correlative with expression of relevant hormone receptors

and/or the location of relevant hormone response elements (HREs) in the genome. To address these

questions, we exposed cultured mouse pluripotent (induced pluripotent stem [iPS]), somatic (Sertoli and

granulosa), and germ (primordial germ cell like [PGCLC]) cells to BPS and measured changes in DNA

methylation levels at the epigenomic level and gene expression at the transcriptomic level. We found

that there was indeed a difference in cell-type specific susceptibility to EDC-induced epimutagenesis and

that this susceptibility correlated with differential expression of relevant hormone receptors and, in

many cases, tended to generate epimutations near relevant HREs within the genome. Additionally,

however, we also found that BPS can induce epimutations in a cell type that does not express relevant

receptors and in genomic regions that do not contain relevant HREs, suggesting that both canonical and

non-canonical signaling mechanisms can be disrupted by BPS exposure. Most interestingly, we found

that when iPS cells were exposed to BPS and then induced to differentiate into PGCLCs, the prevalence

of epimutations and differentially expressed genes (DEGs) initially induced in the iPSCs was largely

retained in the resulting PGCLCs, however, >90% of the specific epimutations and DEGs were not

conserved but were rather replaced by novel epimutations and DEGs following the iPSC to PGCLC

transition. These results are consistent with a unique concept that many EDC-induced epimutations may

normally be corrected by germline and/or embryonic epigenetic reprogramming but that due to

disruption of the underlying chromatin architecture induced by the EDC exposure, many novel

epimutations may emerge during the reprogramming process as well. Thus, it appears that following

exposure to a disruptive agent such as an EDC, a prevalence of epimutations may transcend epigenetic

reprogramming even though most individual epimutations are not conserved during this process.

Introduction

It has now been more than a half-century since Roy Hertz in 1958 first proposed the notion that

certain chemicals, particularly those used in livestock feed at the time, could contaminate food sources

and bioaccumulate in humans mimicking the activity of hormones (Gassner et al., 1958). As the number

of potential chemicals that could exert these effects has grown, so too has the interest in this area to the

point that this is now a major topic of active research. The dangers of many endocrine disrupting

chemicals (EDCs) studied to date have largely been investigated utilizing animal models. A primary

example is the multi-phased Toxicant Exposures and Responses by Genomic and Epigenomic Regulators

of Transcription (TaRGET) program established to determine the contribution of environmental

exposures to disease pathogenesis as a function of epigenome perturbation in mouse models (Wang et

al., 2018). While the use of animal models has been quite informative for determining the various

systemic or organ/tissue-specific disease-related effects associated with exposure to different EDCs, the

molecular mechanisms underlying these phenomena remain poorly understood. Thus, questions persist

regarding i) the mechanisms by which exposure of cells to EDCs results in the initial formation of

epimutations defined here as any change in epigenomic programming, ii) potential cell-type specific

differential susceptibility to epimutagenesis induced by different EDCs, iii) the potential involvement of

relevant hormone receptors and/or genomic hormone response elements in EDC-induced disruption of

the epigenome, and iv) the ability of an EDC-induced epimutated state to persist inter- or

transgenerationally despite generational epigenetic reprogramming (Diaz-Castillo et al., 2019).

Questions about molecular mechanisms responsible for EDC-exposure based disruption of

normal epigenetic programming have been difficult to interpret in whole animal models due, in part, to

the inherent complexity of the intact animal including potential interactions among multiple different

organs, tissues, and cell types at the paracrine, metabolic and/or systemic levels. To circumvent this

challenge, we opted to expose homogeneous populations of specific cell types in culture to doses of the

EDC, BPS, that are below the maximum safe limit previously set by the EPA for exposure of humans to

the similar estrogen mimetic, BPA. In this way we hoped to learn more about the manner in which direct

exposure to a specific EDC induces epimutations and differential gene expression in different individual

cell types including somatic cells, pluripotent cells, and germ cells. This approach was designed to

distinguish the direct effects of an EDC exposure *in vitro* from indirect effects that may accrue as the

result of an EDC exposure on one cell type inducing a secondary effect on a neighboring or related cell

type *in vivo*.

Previous *in vitro* studies have demonstrated direct susceptibility of cultured cell types to EDC-

induced epimutagenesis, however, those studies were focused on immortalized cancer cell lines that do

not necessarily model key cell types involved in normal initial incursion or subsequent intragenerational

propagation or inter- or transgenerational transmission of environmentally-induced epimutations *in vivo*

(Deb et al., 2016; Goodman et al., 2014; Huang et al., 2019; Karaman and Ozden, 2019; Senyildiz et al.,

2017). We chose to examine direct exposure of three key types of cells relevant to environmental

exposures and subsequent propagation and transmission of induced epimutations *in vivo* – somatic cell

types known to be responsive to endocrine signaling (Sertoli cells and granulosa cells), pluripotent cells

mimicking the preimplantation embryo (iPSCs), and germline cells (PGCLCs).

While there have been reports that have clearly indicated the ability of EDCs to disrupt classical

endocrine signaling (Henley and Korach, 2010; Kelce et al., 1995; Swedenborg et al., 2009; vom Saal and

Hughes, 2005; You et al., 1998), there exist other reports indicating EDCs can also disrupt non-classical

endocrine signaling via nuclear receptors (Ozgyin et al., 2015; Song et al., 2011), G protein-coupled

receptors (Thomas and Dong, 2006), and calcium channel signaling (Brenker et al., 2018). This has

further complicated our understanding of the mechanism(s) by which EDC exposure initially induces

epimutations, and has particularly confounded insight into the susceptibility of germ cells to EDC-

induced epimutagenesis given that germ cells are reported to lack expression of classical endocrine

receptors (Meccariello et al., 2014). This is relevant to the effort to understand mechanisms underlying

transgenerational epigenetic inheritance of EDC-induced epimutations (Anway et al., 2006; Guerrero-

Bosagna et al., 2012; Nilsson et al., 2008) which clearly implicates transmission via the germ line.

The concept of germline transmission of EDC-induced epimutations is further confounded by the

known epigenetic reprogramming events that occur in the preimplantation embryo and developing fetal

and neonatal germ line. Thus, regardless of how EDC-induced epimutations become initially manifest in

the germ line, it is not clear how they subsequently persist and are transmitted inter- or

transgenerationally given the large portion of the epigenome that undergoes genome-wide erasure and

resetting of epigenetic programming during each generation (Cantone and Fisher, 2013; Lee et al., 2014;

Santos et al., 2002; Sanz et al., 2010). *In vitro* cell culture systems afford the opportunity to focus on

individual cell types independently, including somatic cells which are likely initially exposed to most

environmental disruptive influences, pluripotent cells which represent the preimplantation embryo in

which the first major reprogramming event occurs during normal development, and early germline cells

where the second major reprogramming event is initiated in primordial germ cells (PGCs). Importantly,

in addition to the potential to study each of these cell types independently, it is possible to induce

transitions in cell fate *in vitro* that model those that occur during normal development, thereby

recapitulating normal reprogramming events in a way that can facilitate high resolution studies of the

fate of EDC-induced epimutations once they have been induced in any of these cell types.

Here we describe our study of the relative susceptibility to induction of epimutations by direct

exposure of four different cell types maintained in culture – somatic Sertoli and granulosa cells,

pluripotent iPSCs, and germ line PGCLCs – to BPS, followed by our analysis of the fate of epimutations

initially induced in mouse iPSCs that are then induced to transition into PGCLCs. We found that there are

cell-type specific differences in susceptibility to epimutagenesis and associated dysregulation of gene

expression patterns following exposure of these different cell types to a similar dose of BPS below the

maximum safe limit established by the EPA for exposure of human cells to BPA (Ribeiro et al., 2019). We

further found that BPS induction of epimutations in both pluripotent and somatic cell types that express

relevant estrogen receptors (ERs), as well as in germ cells which do not express ERs, suggests disruption

of both canonical and non-canonical endocrine signaling. Most interestingly, we found that when iPSCs

exposed to BPS were then induced to undergo a major transition in cell fate and related epigenetic

reprogramming to form PGCLCs, a similar prevalence of epimutations and DEGs was retained, but,

surprisingly, very few (<10%) specific epimutations or DEGs were conserved during this process. This

suggests that the initial EDC exposure induces disruption of the chromatin landscape that subsequent

epigenetic reprogramming is unable to fully restore. Thus, reprogramming during a major cell fate

transition appears to correct many of the initially induced epimutations, but also appears to induce

many *de novo* epimutations, and this imbalance may persist across multiple generations which may

contribute to continued transgenerational epigenetic inheritance of EDC-induced phenotypes during

succeeding generations.

Results

Dose dependent epimutagenesis response to BPS exposure in iPSCs

To initially assess the extent to which epimutations and dysregulated gene expression can be

induced in cells maintained in culture, we exposed mouse iPSCs to three doses of BPS (1, 50, and 100

μM) and measured the impact on the epigenome and transcriptome using the Illumina Infinium Mouse

Methylation BeadChip Array and RNA-seq, respectively (Figure 1). All three doses of BPS induced

individual differentially methylated CpGs (DMCs) as well as differentially methylated regions (DMRs)

(Figure 1a, Supplementary files 1 and 2), and differentially expressed genes (DEGs) (Figure 1b,

Supplementary file 3) when compared to control mouse iPSCs treated with vehicle (EtOH) only. The

extent of this exposure-specific epimutagenesis was correlated with the dose of BPS used. Information

regarding the number and overlap of DMCs/DMRs/DEGs identified for each dose of BPS is shown in

Supplementary files 1-5. Importantly, in many cases we observed a correlation between differential

expression of a gene and the presence of DMCs in the promoter region of that gene (Supplementary file

6). These results provided an initial proof of concept that exposure of cells maintained *in vitro* to an EDC

such as BPS can disrupt the epigenome and transcriptome in a dose-dependent manner. Interestingly,

exposure to 1 µM BPS induced very few DMRs, but did induce widespread DMCs and DEGs. Because 1

µM BPS is below the FDA's suggested safe environmental level established for exposure of humans or

intact animals to BPA (Ribeiro et al., 2019), and was sufficient to induce DMCs and dysregulated gene

expression on all chromosomes in our cultured iPSCs, we utilized this dose and focused solely on DMCs

when assessing epimutations in all subsequent experiments.

Figure 1. Dose dependent impact of epimutagenesis measured in iPSCs exposed to 1, 50, and 100 µM
BPS. a) Ideogram plots displaying chromosomal distribution of genome-wide changes in DNA
methylation caused by BPS exposure. **b)** Mean difference (MD) plots of changes in gene expression
following exposure to increasing doses of BPS. Exposure to increasing doses of BPS induced increasing,
although plateauing numbers of DMCs, DMRs, and DEGs. Blue horizontal lines = hypomethylated DMCs,
red horizontal lines = hypermethylated DMCs, black squares = DMRs.

Cell-type specific susceptibility to induction of epimutations following BPS exposure

We next sought to determine if different key cell types – somatic, pluripotent or germ – are

differentially susceptible to induction of epimutations in response to direct exposure of each to a similar

dose of BPS. Thus, we exposed pluripotent (iPSCs), somatic (Sertoli and granulosa cells), and germ

(PGCLCs) cell types to 1 µM BPS and measured changes in DNA methylation patterns. We identified

exposure-specific DMCs in each exposed cell type relative to its corresponding control (exposed to

carrier only) (Figure 2a). We observed overall differences among the different cell types in total numbers

of DMCs, with iPSCs showing the highest number of DMCs, followed by Sertoli cells and granulosa cells,

and then PGCLCs, respectively (Table 1).

Figure 2. Chromosomal distributions and annotations of BPS-induced epimutations in pluripotent,
somatic, and germ cell types. a) Ideograms illustrating chromosomal locations of DMCs induced by
exposure of each cell type to 1 µM BPS. Blue horizontal lines = hypomethylated DMCs, red horizontal
lines = hypermethylated DMCs. **b)** Enrichment plots indicating feature annotations in genomic regions
displaying prevalent BPS-induced epimutations in each cell type. Dot size = number of overlapping DMCs

with specific annotation, dot color = enrichment score reflecting the relative degree to which
epimutations occurring in a specific annotated class are overrepresented.

Table 1. Treatment-specific differentially methylated sites (DMCs) (treated vs. control)				
DMCs	**Sertoli**	**Granulosa**	**iPSCs**	**PGCLCs**
Hypomethylated[1]	7,385	6,444	9,651	2,315
Hypermethylated[2]	3,022	4,143	4,308	4,785
Total	10,407	10,587	13,959	7,100

[1]A CpG site that was predominantly methylated in the control samples but unmethylated in the exposed samples.
[2]A CpG site that was predominantly unmethylated in the control samples but methylated in the exposed samples.

Interestingly, among the exposure-specific DMCs identified in each cell type, we observed

predominantly hypomethylated DMCs in the somatic and pluripotent cell types, but predominantly

hypermethylated DMCs in PGCLCs (Table 1). These findings confirm that there are both quantitative

differences among cell types in overall susceptibility to epimutagenesis, and qualitative cell-type specific

differences in the prevalence of hypo- versus hypermethylated DMCs following exposure to BPS. The

latter likely reflects the fact that the epigenome in PGCs is naturally more hypomethylated than that in

pluripotent or somatic cell types (Hajkova, 2011), enhancing the likelihood that most changes in DNA

methylation induced in PGCLCs will necessarily involve hypermethylation.

Annotation of BPS-induced epimutations

As we only found an average of 11.05% direct overlap in DMCs between two or more cell types

and no overlapping DMCs shared between all cell types (Supplementary file 4), we next analyzed

annotations associated with genomic sites of BPS exposure-specific DMCs in each cell type. We mined

annotations associated with genomic regions included on the Infinium array and found that a substantial

group of exposure-specific DMCs was associated with enhancer regions in the two somatic (Sertoli and

granulosa) and one pluripotent (iPS) cell types exposed to BPS, while in PGCLCs BPS-induced DMCs were

more prevalent at promoter regions containing transcription start sites (TSSs) and transcription factor

binding sites (Figure 2b), indicative of yet another qualitative cell-type specific difference in induction of

epimutations by the same dose of BPS.

Relationship between susceptibility to BPS-induced epimutagenesis and expression of relevant

hormone receptors

201   As EDCs are thought to induce epimutations via disruption of classical hormonal signaling

(Henley and Korach, 2010; Kelce et al., 1995; Swedenborg et al., 2009; vom Saal and Hughes, 2005; You

et al., 1998), we next sought to determine if the differential extent of BPS induction of epimutations we

observed in different cell types was associated with differential expression of relevant hormone

receptors in each. We performed immunocytochemistry (ICC) to detect the presence of the relevant

estrogen receptors – ERα and ERβ, while co-staining for cell-type specific markers (WT1 [Sertoli cell

marker], FSHR [granulosa cell marker], FUT4 [iPSC marker], and NANOG [PGCLC or endogenous PGC

marker]) to confirm the identity of the four cultured cell types examined in this study (Sertoli cells,

granulosa cells, iPSCs and PGCLCs), as well as endogenous mouse PGCs (Figure 3a). Both somatic cell

types (Sertoli and granulosa) showed positive immunolabeling for both ERα and ERβ, whereas the

pluripotent cells showed positive immunolabeling for ERβ, but negative immunolabeling for ERα, and

the PGCLCs and endogenous PGCs showed negative immunolabeling for both ERα and ERβ. Importantly,

the latter result is consistent with previous reports of lack of expression of either ERα or ERβ at the

protein level in endogenous PGCs (Meccariello et al., 2014).

215   These results demonstrate cell-type specific differences in expression of hormone receptors

(ERα and ERβ) which are potentially relevant to the disruptive action of the estrogen mimetic, BPS. We

noted that expression of at least one potentially relevant hormone receptor (ERβ) in either somatic or

pluripotent cells correlated with a higher incidence of DMCs induced by exposure of somatic or

pluripotent cell types to 1 µM BPS relative to the incidence of DMCs induced by exposure of PGCLCs,

which do not express either estrogen receptor, to the same dose of BPS (Table 1). Nevertheless, we did

still observe induction of DMCs when PGCLCs were exposed to BPS, demonstrating that expression of

relevant canonical hormone receptors is not an absolute requirement for induction of epimutations in

response to an EDC. To exclude the possibility that the observed susceptibility in germ cells could be

correlated with the presence of other endocrine receptors that could interact with BPS, we performed

additional ICC for the presence of AR, PPARγ, and RXRα and found they were all absent as well in

endogenous PGCs (Supplementary file 8). This result and our observation noted above that somatic and

pluripotent cell types showed a higher incidence of epimutations at apparent enhancer regions while

germ cells showed a higher epimutation incidence at apparent promoter regions reveal cell-type specific

differences in susceptibility to induction of epimutations by exposure to the EDC, BPS.

Proximity of BPS-induced epimutations to genomic EREs

The relatively higher incidence of BPS-induced epimutations in cell types expressing one or both

estrogen receptor(s) suggests BPS-induced epimutagenesis may be manifest, at least in part, through

canonical endocrine signaling pathways. If this is the case, we might expect to see elevated induction of

epimutations in genomic regions enriched for relevant HREs which, for interactions with the estrogen

mimetic, BPS, would be EREs. Previous studies have defined a full ERE consensus sequence (Bourdeau et

al., 2004), but other reports have indicated that estrogen receptors can often bind to ERE half-sites

(Mason et al., 2010) (Figure 3b). Indeed, when we mined publicly available ERα ChIP-seq peaks from the

UCSC genome browser database (Dunham et al., 2012; Myers et al., 2011; Sloan et al., 2016; Wang et

al., 2013, 2012) and performed motif enrichment, we identified two distinct ERE half-sites within regions

enriched for DMCs rather than one full sized ERE consensus sequence, with the half-sites appearing to

be more biologically relevant to interaction with estrogen or its mimetics (Figure 3b). We then plotted

the frequency of ERE half-sites in genomic regions within 500 bp of BPS-induced epimutations (Figure

3c,d). We found an increase in the frequency of ERE half-sites identified within 500 bp of BPS-induced

DMCs genome-wide in all four cell types investigated, but the frequency was notably lower in germ cells

(Figure 3c). Thus, this higher frequency of ERE half-sites within 500 bp of BPS-induced DMCs was

conserved in three of the four cell types – somatic (Sertoli and granulosa) and pluripotent (iPSCs) when

we focused solely on apparent enhancer regions, but not in germ cells (PGCLCs) where the majority of

DMCs occurred at apparent promoter regions (Figure 3d). We also mined the UCSC mouse genome

sequence for genome-wide prevalence of ERE half-sites and found that while these sites occur in both

promoter and enhancer regions, they are nearly four-fold more frequent in enhancers (Supplementary

files 9 and 10). Thus, it is perhaps not surprising that BPS-induced epimutations were more prevalent in

enhancer regions in somatic and pluripotent cell types that express one or both ERs, than in germ cells

that do not express either ER.

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Figure 3. Correlation between cell-type specific expression of estrogen receptors and density of
genomic EREs associated with BPS-induced epimutations. a) Assessment of expression of ERα and ERβ
by cell types co-stained for known cell-type specific markers. Somatic cell types express both receptors,
pluripotent cells express ERβ but not ERα, and germ cells do not express either estrogen receptor. **b)**
Motif plots displaying the full ERE consensus sequence and the more biologically relevant ERE half-site
motifs found to be enriched from ERα ChIP-seq. **c,d)** Normalized density plots and box plots displaying
the frequency of ERE half-sites identified c) within 500 bp of all BPS-induced DMCs genome-wide, or d)
within 500 bp of the most enriched categories of BPS-induced DMCs in each cell type (= enhancer
regions for somatic and pluripotent cell types and promoter regions in the germ cell type).

Taken together, these results suggest a relationship between i) expression of either ERβ alone or

ERα and ERβ together, ii) elevated susceptibility to BPS-induced epimutagenesis, and iii) the occurrence

of BPS-induced DMCs in genomic regions – particularly enhancers – containing EREs. These observations

are consistent with the notion that one mechanism contributing to EDC-based induction of epimutations

involves disruption of canonical endocrine signaling pathways. However, the fact that exposure to BPS

also induced epimutations in germ cells, even in the absence of expression of estrogen-related

receptors, and generated many DMCs in regions not inclusive of EREs, suggests that disruption of

canonical endocrine signaling pathways is not the only mechanism by which exposure to an EDC can

induce epimutations.

Cell-type specific features of BPS-induced epimutations

To further interrogate cell-type specific differences in the genesis of epimutations following

exposure to BPS, we compared DNA methylation patterns detected in the control (vehicle only) samples

to identify naturally occurring, cell-type specific DMCs inherently associated with each different cell fate.

Interestingly, of the 297,415 distinct CpGs interrogated by the Illumina Infinium Mouse Methylation

BeadChip Array used for this analysis, >240,000 showed some degree of inherent differential

methylation among the four cell types tested, suggesting they are tied to cell-fate specific differential

DNA methylation (Figure 4a). We next determined the extent to which DMCs induced specifically by

exposure of each cell type to BPS occurred at CpG dinucleotides that were among these naturally

occurring cell-type specific DMCs. We found that a large majority of BPS exposure-specific epimutations

detected in each cell type occurred at CpGs that also show inherent, cell-type specific variation in DNA

methylation (>95% of BPS-induced epimutations in somatic and pluripotent cell types and ~89% in germ

cells) (Figure 4a). As with the overall pattern of BPS-induced epimutations shown in Figure 2, BPS-

induced epimutations at CpGs showing inherent cell-type specific variation were enriched in apparent

enhancer regions that occurred near EREs in somatic and pluripotent cell types, whereas those in germ

cells were found predominantly in promoter regions containing significantly fewer EREs (Figure 4b,d).

However, the low percentage of BPS-induced epimutations that occurred at CpGs that did not show

inherent cell-type specific variation in DNA methylation patterns were found to be enriched in promoter

regions lacking nearby EREs in all four cell types (Figure 4c,d). Thus, this latter group of BPS-induced

epimutations appears to represent a small core group that arises following exposure to BPS via a

mechanism that does not rely upon disruption of canonical endocrine signaling, and so is common to all

cell types, regardless of expression of relevant endocrine receptors or nearby residence of relevant HREs

within the genome.

Figure 4. Direct comparison of BPS exposure-specific and cell-type specific features between cell types. a) Assessment Venn diagrams indicating DMCs that are due either to BPS exposure (top, smaller ovals) or inherent cell-type specific differences (bottom, larger ovals). Numbers of apparent endocrine-signaling related DMCs are shown in the light orange arrow, and apparent endocrine-signaling independent DMCs are shown in the dark orange arrows. Enrichment plots indicating feature annotations in genomic regions displaying b) apparent endocrine-signaling related DMCs occurring predominantly in enhancer regions in somatic Sertoli and granulosa cell types or pluripotent cells

expressing one or more estrogen receptors, or **c)** a smaller set of apparent endocrine-signaling
independent DMCs occurring predominantly in promoter regions in all four cell types regardless of +/-
expression of relevant endocrine receptors. **d)** Normalized density plots and box plots displaying the
frequency of ERE half-sites identified within 500 bp of apparent endocrine-signaling related DMCs
occurring predominantly in enhancer regions and apparent endocrine-signaling independent DMCs
occurring predominantly in promoters.

Impact of BPS exposure on gene expression

We next sought to determine if the cell-type specific differential susceptibility to BPS-induced

epimutagenesis translated to a similar extent of dysregulation of gene expression in each cell type. We

found that there was a similar relationship between the presence of DMCs in the promoter region of a

gene and differential expression of that gene in all four cell types examined (Supplementary file 6).

Somewhat surprisingly, RNA-seq analysis of gene expression patterns in each exposed cell type relative

to its corresponding control (same cell type exposed to carrier only) revealed the greatest number of

dysregulated genes in PGCLCs, despite PGCLCs being the cell type that showed the lowest number of

epimutations following exposure to BPS (Tables 1,2). iPSCs showed the second highest level of

dysregulated genes, while somatic Sertoli and granulosa cells showed relatively low levels of

dysregulated gene expression (Table 2). Indeed, numbers of dysregulated genes were three orders of

magnitude lower in differentiated somatic cells than in either pluripotent cells or germ cells (Table 2).

Because BPS-exposed PGCLCs showed the highest level of dysregulated genes, as well as the

highest enrichment of DMCs occurring primarily at promoters (Figure 2b), we next assessed the general

proximity between DMCs and promoter regions in each cell type to determine if DMCs in promoter

regions may be more likely to predispose dysregulated gene expression than DMCs elsewhere in the

genome (Figure 5a). Indeed, we found that in all four cell types, the smaller the median distance

between BPS-induced DMCs and neighboring promoter regions, the larger the number of BPS-induced

DEGs (Figure 5b). Thus, it appears that while PGCLCs showed the fewest overall BPS-induced DMCs

among the four cell types exposed to BPS, this exposure induced a higher proportion of epimutations in

regions within or adjacent to promoters in this cell type. This, and the relatively high extent of

decondensed chromatin genome-wide in fetal germ cells, appear to have contributed to the higher

incidence of dysregulated gene expression in BPS-exposed PGCLCs than that observed in the other three

BPS-exposed cell types, even though the overall numbers of BPS-induced epimutations were greater in

the other cell types.

Table 2. Exposure-specific differentially expressed genes*				
DEGs	**Sertoli**	**Granulosa**	**iPSCs**	**PGCLCs**
Down regulated	3	0	343	844
Up regulated	32	2	694	1,046
Total	35	2	1,037	1,890

*Genes showing significant differential expression following exposure of each cell type to 1µM BPS relative to
matched control cell types exposed to carrier only.

Figure 5. Correlation between the proximity of DMCs to promoters and dysregulation of gene
expression. a) Proximity plot displaying distances from exposure-specific DMCs to nearest promoter
regions. Dotted lines indicate median points of the data for each cell type. **b)** Correlation plot displaying
a negative relationship between the distance from DMCs to nearest promoters and resulting
dysregulation of gene expression within each cell type.

Potential non-canonical signaling pathways disrupted by BPS exposure

To identify potential mechanisms by which BPS exposure may induce epimutations via

disruption of non-canonical endocrine signaling pathways, we mined our RNA-seq data to identify genes

dysregulated independently of expression of relevant hormone receptors or presence of nearby EREs.

We identified a set of genes in all four cell types tested that displayed promoters enriched for apparent

endocrine-signaling independent DMCs (Supplementary file 12). Gene ontology (GO) analysis indicated

that several of the 1,957 genes we identified were associated with ubiquitin-like protease pathways,

including 124 of 800 genes that regulate protein degradation and 5 of 9 genes involving 1-

phosphatidylinositol-3 kinase activities, linked to the PI3K/AKT signaling pathway (Figure 6a).

Previous reports have established a link between ubiquitin-like protease pathways and classical

ER signaling (Beamish and Frick, 2022; Kabir et al., 2015), suggesting that genes related to ubiquitin-like

protease pathways that have promoters lacking EREs may be regulated by factors that are themselves

encoded by genes with promoters containing EREs, and so could be indirectly activated or repressed by

disruption of classical ER signaling. However, involvement of PI3K/AKT pathway signaling has previously

been linked to estrogen signaling through non-canonical G-protein coupled receptors such as GPER1,

which was sufficient to induce estrogen signaling in ER KO mouse cell lines which lack the capacity for

classical ER signaling (Filardo et al., 2002; Molina et al., 2017). To determine if our data support the

suggestion that the non-canonical BPS-induced changes were linked to involvement of PI3K/AKT

pathway genes, we performed GO analysis on the 1,890 DEGs identified in PGCLCs in which

epimutations appeared to be induced independent of classical ER signaling. Interestingly we detected no

apparent involvement of pathways involving ubiquitin-like proteases, but we did detect differential

expression of 4 of 7 genes associated with a pathway involving phosphatidylinositol-4 phosphatase

signaling which intersects with the PI3K/AKT signaling pathway (Figure 6b). Finally, while we did not

detect expression of *Gper1* transcripts, we did observe differential expression of genes encoding other

less well studied G-protein coupled receptors, including *Gprc5a*, *Gprc5b*, *Gpr89, Gpr107*, and *Gpr161* in

all four cell types exposed to BPS. These receptors have all been previously shown to bind either BPA or

17β-estradiol in rat models and could be potential targets of non-canonical BPS-induced estrogen

signaling (Figure 6c) as described in the following links:

(https://rgd.mcw.edu/rgdweb/report/gene/main.html?id=1305882,

https://rgd.mcw.edu/rgdweb/report/gene/main.html?id=1563245,

https://rgd.mcw.edu/rgdweb/report/gene/main.html?id=1304837,

https://rgd.mcw.edu/rgdweb/report/gene/main.html?id=1310804,

https://rgd.mcw.edu/rgdweb/report/gene/main.html?id=1309510)

Figure 6. Potential involvement of non-canonical estrogen signaling pathways in BPS-induction of
epimutations. Relative expression of genes **a)** enriched for apparent endocrine-signaling independent
promoter-region DMCs found in all cell types or **b)** dysregulated in PGCLCs which lack expression of
estrogen receptors. **c)** Heatmap of relative expression of estrogen receptor genes (*Esr1* and *Esr2*) and G-
coupled protein receptors (*Gprc5a*, *Gpr107*, *Gprc5b*, *Gpr161*, and *Gpr89*) in pluripotent, somatic, and
germ cell types. *Gprc5a*, *Gpr107*, *Gprc5b*, *Gpr161*, and *Gpr89* all have been shown to bind to BPA or 17β-
estradiol in rat models and represent potential G-coupled protein receptors which could lead to the
induction of endocrine-signaling independent DMCs.

Propagation of BPS-induced epimutations during transitions in cell fate

Transitions between pluripotent and germ cell fates, or vice versa, are accompanied by large-

scale epigenetic reprogramming *in vivo* (Cantone and Fisher, 2013; Lee et al., 2014; Santos et al., 2002;

Sanz et al., 2010), and these are recapitulated during similar transitions induced *in vitro* (Ishikura et al.,

2016). To determine the extent to which BPS-induced epimutations persist during a pluripotent to

germline transition *in vitro*, we first exposed iPSCs to 1 μM BPS and then differentiated the exposed

iPSCs first into epiblast-like cells (EpiLCs) and then into PGCLCs, recapitulating the early germline

epigenetic reprogramming event that normally occurs *in vivo* (Kurimoto and Saitou, 2018; Ohta et al.,

2017) (Figure 7a). We then used genome-wide analyses by EM-seq and RNA-seq to compare numbers of

exposure-specific DMCs and DEGs, respectively, in PGCLCs derived from BPS-exposed iPSCs with those in

the directly exposed iPSCs and found lower, but still substantial numbers of both (28,168 vs 38,105

DMCs, and 1,437 vs 1,637 DEGs in the derived PGCLCs compared to the directly exposed iPSCs,

respectively) (Figure 7b,c).

When we compared the specific DMCs and DEGs that were detected in the BPS-exposed iPSCs

with those that were detected in the subsequently derived PGCLCs, we found that >90% of each were

not conserved during the pluripotent to germline transition in cell fate. Specifically, only 3.7% of the

DMCs and 8.4% of the DEGs detected in the BPS-exposed iPSCs were also detected in the PGCLCs

derived from the exposed iPSCs. Among the small portion of specific 138 DEGs (Supplementary file 13)

that were conserved from exposed iPSCs to derived PGCLCs, several (12.32%) (*Cdkn1a*, *Ccnd2*, *Plk2*,

Tgfbr1, *Gadd45g*, *Lck*, *Ltbr*, *Mad2l1*, *Ap3m2*, *Ctsz*, *Tcirg1*, *Gusb*, *Id2*, *Lefty2*, *Gstm7*, *Acsl1*, *Slc39a14*) were

involved in cell cycle and apoptosis pathways which could potentially be linked to cancer development

(Supplementary file 14). Thus, of the 38,105 DMCs and 1,637 DEGs induced by exposure of iPSCs to 1

μM BPS, 36,688 and 1,499, respectively, did not persist during differentiation of iPSCs to form PGCLCs,

and so were apparently corrected by germline reprogramming. Simultaneously however, 26,752 novel

DMCs and 1,299 novel DEGs appeared in the derived PGCLCs that were not present in the BPS-exposed

iPSCs, so were apparently generated *de novo* during the germline reprogramming process. These results

are consistent with the notion that germline epigenetic reprogramming corrected many of the

epimutations that were present in the BPS-exposed pluripotent cells from which they were derived, but

that exposure of cells to EDCs may disrupt the underlying chromatin landscape in a way that then

interferes with subsequent reprogramming such that in addition to correcting many previously induced

epimutations, the germline reprogramming process, acting on a disrupted chromatin landscape, also

generates many novel epimutations *de novo* during the pluripotent to germline transition.

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Figure 7. Persistence of BPS-induced epimutations through recapitulation of early germline reprogramming *in vitro*. a) Schematic illustrating derivation of PGCLCs from iPSCs *in vitro*. iPSCs are first induced to form EpiLCs which are then induced to form PGCLCs. iPSCs were exposed to either ethanol + 1 µM BPS or ethanol (carrier) only, then induced to undergo transitions to form EpiLCs and then PGCLCs. **b)** DNA samples from BPS-exposed or control iPSCs as well as subsequently derived PGCLCs were assessed for exposure-specific DNA methylation epimutations by EM-seq. BPS-treated iPSCs showed 38,105 DMCs and subsequently derived PGCLCs showed 28,169 DMCs. Of those, only 1,417 (3.7%) of the DMCs were conserved from the BPS-exposed iPSCs to the subsequently derived PGCLCs. **c)** RNA samples from BPS-exposed or control iPSCs and subsequently derived PGCLCs were assessed for global gene expression patterns by RNA-seq. BPS-treated iPSCs showed 1,637 exposure-specific DEGs and subsequently derived PGCLCs showed 1,437 exposure-specific DEGs. Of those, only 138 (8.4%) were conserved from the BPS-exposed iPSCs to the subsequently derived PGCLCs.

Discussion

While nearly 20 years of research on the effects of exposure of live animals to various EDCs and

other environmental disruptive influences has clearly established the potential to perturb the

epigenome in ways that can dysregulate normal gene expression patterns and predispose the

development of disease states, the molecular mechanisms underlying these phenomena have remained

largely undefined. Thus, the manner in which environmental exposures introduce biochemical changes

in epigenetic programming, how such disruptions can be propagated within a tissue or throughout the

soma of an exposed individual or enter that individual's germ line, or how these disruptions predispose

disease states in that individual, or how a resulting prevalence of epimutations can be transmitted to

multiple subsequent generations despite generational epigenetic reprogramming, are all mechanistic

questions that have remained unanswered despite more than 5,000 publications on this topic since the

early 2000s. The vast majority of those publications have described experiments in animal models –

typically rodents. While informative, such live animal studies require months to years to complete, are

cost-, labor-, and animal-intensive, and have not yielded substantial insight into the molecular

mechanisms underlying the deleterious effects of environmentally-induced epimutations.

In vitro model systems have proven to be very useful tools for deciphering molecular

mechanisms related to normal development and homeostasis, or disruptions of those processes

predisposing onset of disease. Examples include cell culture systems used to study developmental

processes such as X-chromosome inactivation in mammals (Almeida et al., 2017; Patrat et al., 2009) or

early embryogenesis (Bao et al., 2022; Lau et al., 2022), or to elucidate the cellular and molecular

etiology of many different diseases based on the now popular "disease-in-a-dish" approach (Davaapil et

al., 2020; Song et al., 2023). While *in vitro* cell culture systems are, by nature, devoid of a majority of the

physiological complexities present within the intact organism *in vivo*, this provides an actual advantage

in that it facilitates deconvolution of those complexities while unequivocally focusing on only certain,

specific aspects or cell types. However, this ultimately warrants validation of results discerned from

studies of *in vitro* models to ensure they also reflect functions ongoing in the more complex and

heterogeneous environment of the intact animal *in vivo*.

We assembled an *in vitro* system in which we combined the three cell types normally involved in

i) the initial exposure to disruptive environmental effects (somatic cells), ii) the first phase of

generational epigenetic reprogramming in the preimplantation embryo (pluripotent cells), and iii) the

second phase of generational reprogramming in the developing germ line (germ cells). This system

allowed us to gain insights into multiple mechanistic aspects of environmentally-induced

epimutagenesis including the initial induction of epimutations by exposure of cells to the EDC, BPS, cell-

type specific differences in susceptibility to BPS-induced epimutagenesis, factors that appear to

contribute to that differential susceptibility, and the mechanism by which an epimutated state may

persist across multiple generations despite generational reprogramming.

Here we have shown that exposure of homogeneous populations of cells maintained in culture

to an EDC such as BPS can induce epimutagenesis and dysregulation of gene expression, and that this

occurs in a partially additive dose-dependent manner, although the quantitative effect appears to

plateau at higher doses. This afforded us the opportunity to test the relative effects of direct exposure

of different specific cell types to a similar dose of BPS with no confounding effects imposed by

interactions with other cell types or systemic effects of the sort that normally occur in the *in vivo*

context. We found that somatic, pluripotent, and germ cell types do indeed show differences in direct

susceptibility to induction of epimutations following exposure to BPS, both in terms of the quantity of

epimutations induced and qualities of the resulting epimutations such as a predominance of hyper-

versus hypomethylation, and the genomic locations where such epimutations tend to occur. Our results

suggest that these differences reflect distinctions in the normal status of each cell type at the time of

exposure. Thus, somatic, pluripotent, and germ cell types differ in the normal status of genome-wide

patterns of accessible versus inaccessible chromatin, global DNA methylation, and expression of relevant

endocrine receptors. Interestingly, our data suggest that all of these parameters can influence cell-type

specific susceptibility to induction of epimutations by exposure to BPS, and that different combinations

of these variables ultimately determine the quantity and quality of epimutations induced by exposure to

BPS. We note that while our exposure dose of 1 μM of BPS was below that deemed safe by the EPA for

exposure of humans to the similar EDC, BPA, that same dose may exert greater effects when used to

expose cells in culture in the absence of any sort of mitigating metabolic effects that may accrue in

intact animals or humans. Indeed, the potential to quantify the epimutagenic effects of different doses

of an EDC on different cell types as shown in our study could be used in future studies to assess the

relative effects of a specific dose of an EDC on a specific cell type when that cell type is exposed either in

a homogeneous culture or within an intact animal.

Endogenous fetal germline cells (e.g. PGCs and prospermatogonia in males) normally display

lower levels of global DNA methylation and higher levels of chromatin accessibility than pluripotent or

somatic cell types (Hajkova, 2011), yet PGCLCs developed fewer epimutations following exposure to BPS

than the other cell types. This appears to be explained by the fact that germline cells do not express

either estrogen receptor whereas pluripotent and somatic cell types express one or both of the two

estrogen receptors, apparently rendering them more susceptible to induction of epimutations following

exposure to the estrogen mimetic, BPS. This is consistent with the notion that EDCs impose disruptive

effects by interfering with canonical endocrine signaling pathways, and that notion is further supported

by our observation that, in the cell types that express one or both estrogen receptors, we observed a

prevalence of BPS-induced epimutations in genomic regions containing nearby EREs, whereas in PGCLCs,

which do not express either estrogen receptor, we did not observe a strong correlation between the

locations of BPS-induced epimutations and genomic EREs.

Nevertheless, we did observe induction of epimutations in PGCLCs exposed to BPS, suggesting

that while cell types expressing relevant endocrine receptors may display elevated susceptibility to EDC-

induced epimutagenesis presumably based on disruption of canonical endocrine signaling, this is not the

only mechanism by which EDCs can induce epimutations. Indeed, our results support the suggestion

that there may be at least two types of epimutations induced by exposure to EDCs – one that is

independent of canonical endocrine signaling and so occurs in all cell types regardless of expression of

relevant endocrine receptors or the genomic location of relevant HREs, and another that does involve

disruption of canonical endocrine signaling and so is elevated in cell types expressing relevant endocrine

receptors and predisposes the induction of epimutations in genomic regions near relevant HREs. Beyond

this, a comparison of the quantities of BPS-induced epimutations in pluripotent and somatic cell types,

which both express one or both estrogen receptors, but which differ with respect to elevated genome-

wide chromatin accessibility in pluripotent cells compared to much more limited chromatin accessibility

in somatic cell types (Cantone and Fisher, 2013; Santos et al., 2002), suggests that chromatin

accessibility also contributes to susceptibility to EDC-induced epimutagenesis.

One of our most intriguing findings was that, in addition to estrogen-receptor expressing

somatic and pluripotent cell types developing more epimutations than non-estrogen-receptor

expressing germ cells in response to exposure to BPS, the occurrence of epimutations in the former

(somatic and pluripotent) cell types was predominantly in enhancer regions whereas that in the latter

(germ) cell type was predominantly in gene promoters. As noted above, there was a significant increase

in the occurrence of BPS-induced epimutations near EREs in somatic and pluripotent cell types

compared to germ cells. A genome-wide analysis confirmed that ERE half-sites occur more frequently in

enhancer regions than in promoter regions (Supplementary files 9 and 10) supporting our suggestion

that the higher prevalence of BPS-induced epimutations we observed in enhancer regions in somatic

and pluripotent cell types was due to disruption of canonical hormone signaling.

Our finding that PGCLCs, which do not express endocrine receptors, still accrued epimutations

following exposure to BPS indicates that in addition to inducing epimutations by disrupting canonical

endocrine signaling pathways, exposure to EDCs can induce epimutations via mechanisms that do not

involve disruption of such pathways. This notion is further supported by the occurrence of BPS-induced

epimutations in genomic regions that do not include nearby EREs, even though a prevalence of the

epimutations induced in PGCLCs were located in gene promoters. Pathway analysis of the genes in

which promoter-region epimutations were induced by exposure of PGCLCs to BPS suggests an apparent

endocrine-signaling independent group of genes that can become disrupted by exposure to an EDC via

non-canonical signaling which may occur through the PI3K/AKT pathway via G-protein coupled

receptors. While we observed expression of a number of G-protein coupled receptors known to bind to

either 17β-estradiol or BPA, further studies will be required to determine if any of these contribute to

susceptibility to BPS epimutagenesis.

As expected, disruption of normal epigenetic programming induced by exposure of cells to BPS

also led to dysregulation of gene expression. Thus, RNA-seq analysis detected differentially expressed

genes in all cell types exposed to BPS. Surprisingly, the greatest number of DEGs was detected in PGCLCs

which was the cell type that developed the lowest overall number of epimutations in response to

exposure to BPS. However, PGCLCs showed a higher prevalence of BPS-induced epimutations in gene

promoter regions than was observed in any of the other cell types, so it appears that promoter-region

epimutations impose the most direct impact on gene expression. As noted above, fetal germ cells

normally display the lowest level of global DNA methylation of any cell type at any developmental stage.

This is characteristic of the epigenetic ground state that accrues uniquely in developing germ cells

(Hajkova, 2011), and is believed to reflect a generally more accessible chromatin state genome-wide in

germ cells than in other cell types. Thus, it may be that the epigenetic ground state in developing germ

cells renders this cell type uniquely susceptible to epimutagenesis in gene promoter regions, thereby

predisposing a high level of dysregulation of gene expression in response to exposure to an EDC.

Finally, our *in vitro* system allowed us to follow the fate of BPS-induced epimutations through a

major transition in cell fate from pluripotent to germline cells during which a portion of the normal

generational germline epigenetic reprogramming process is known to occur (Kurimoto and Saitou,

2018). Thus, when we induced iPSCs previously exposed to 1 μM BPS to differentiate into PGCLCs, we

found that a substantial prevalence of epimutations persisted during this transition. However, very few

of the specific DMCs or the associated DEGs detected in the BPS-exposed iPSCs persisted in the

subsequently derived PGCLCs. Specifically, only 3.7 % of DMCs and 8.4 % of DEGs were conserved

between the BPS-exposed iPSCs and the subsequently derived PGCLCs. This suggests that exposure of

cells to EDCs has the potential to disrupt epigenetic programming in a way that then interferes with

subsequent generational reprogramming that normally accompanies either germline to pluripotent or

pluripotent to germline transitions in cell fate. Recent *in vivo* studies have suggested that exposure of

gestating female mice to the EDC tributyltin results in disruption not simply of the pattern of epigenetic

modifications but also of the underlying chromatin landscape (Chamorro-García et al., 2021; Chang et

al., 2022). In this context, our results are consistent with the suggestion that exposure of cells to EDCs

can disrupt the underlying chromatin landscape (e.g. the pattern of A and B chromatin compartments)

such that when the "normal" reprogramming machinery then acts on this disrupted landscape it

corrects many of the originally induced epimutations, but simultaneously induces many novel

epimutations *de novo*. This would predispose ongoing abnormalities in the underlying chromatin

landscape that would, in turn, lead to the recurring correction of many existing epimutations in concert

with the genesis of many novel epimutations during each subsequent generation.

We note that, with the exception of our analysis of granulosa cells, all of our studies were

carried out in "male" XY-bearing cells. It remains possible that XX-bearing cells might differ from XY-

bearing cells in the way(s) in which they respond to exposure to an EDC. However, the similarity we

observed between responses of XX granulosa cells and XY Sertoli cells suggest this may not be the case –

at least in cell types in which dosage compensation has been established by X-chromosome inactivation.

Taken together, our results demonstrate the utility of an *in vitro* cell culture approach for

pursuing molecular mechanisms underlying environmentally induced disruption of normal epigenetic

programming manifest as the formation of epimutations and dysregulated gene expression. This

approach clearly affords unique potential to reveal mechanisms responsible for the initial induction of

epimutations in response to direct exposure of cells to disruptive effects such as EDCs, as well as

offering potential means to elucidate mechanisms by which environmentally induced epimutations are

then propagated within the exposed individual and then to that individual's descendants. Knowledge of

these mechanisms will afford the best opportunity to understand how these defects may, in the future,

be better diagnosed, treated, and/or prevented. With the ever-expanding catalogue of potentially

hazardous man-made compounds permeating our environment, it is increasingly important that we

recognize the potential dangers such compounds may impose and maximize our ability to protect

ourselves from those dangers.

Materials and methods

Key resources table

Reagent type (species) or resource	Designation	Source or reference	Identifiers	Additional information
Genetic	R26^{rtTA}; Col1a1$^{2lox-4F2A}$	The Jackson	011011	

reagent (*M. musculus*)		Laboratory		
Cell line (*M. musculus*)	CF1 Mouse embryonic fibroblasts, MitC-treated	Thermo Fisher Scientific	A34959	
Chemical compound	Dulbecco's Modified Eagle Medium (DMEM)	Thermo Fisher Scientific	10313021	High glucose, pyruvate, no glutamine
Chemical compound	Fetal bovine serum (FBS)	Thermo Fisher Scientific	10439024	Embryonic stem-cell FBS, qualified, USDA-approved regions
Chemical compound	Leukemia inhibitory factor (LIF)	Millipore Sigma	ESG1107	ESGRO® Recombinant Mouse LIF Protein
Chemical compound	DMEM/F12	Thermo Fisher Scientific	21041025	No phenol red
Chemical compound	Insulin	Millipore Sigma	I1882	From bovine pancreas
Chemical compound	Apo-Transferrin	Millipore Sigma	T1147	From human
Chemical compound	Bovine Serum Albumin (BSA)	Thermo Fisher Scientific	15260037	Fraction V (7.5% solution)
Chemical compound	Progesterone	Millipore Sigma	P8783	
Chemical compound	Putrescine dihydrochloride	Millipore Sigma	P5780	
Chemical compound	Sodium selenite	Millipore Sigma	S5261	
Chemical compound	Neurobasal™	Thermo Fisher Scientific		No phenol red
Chemical compound	B-27™	Thermo Fisher Scientific	12587010	(50X), minus vitamin A
Chemical compound	Penicillin-streptomycin	Thermo Fisher Scientific	15070063	(5,000 U/mL)
Chemical compound	GlutaMAX™	Thermo Fisher Scientific	35050061	(100X)
Chemical compound	2-Mercaptoethanol	Thermo Fisher Scientific	21985023	(1000X)
Chemical compound	CHIR99021	BioVision	1677-5	
Chemical	PD0325901	Amsbio	04-0006-02	10 mM in DMSO

compound				
Chemical compound	Recombinant Human/Murine/Rat Activin A	PeproTech	120-14	Insect derived
Chemical compound	Recombinant Human FGF-Basic (FGF-2/bFGF)	Thermo Fisher Scientific	13256-029	
Chemical compound	KnockOut™ Serum	Thermo Fisher Scientific	10828028	
Chemical compound	Glasgow's MEM (GMEM)	Thermo Fisher Scientific	11710035	
Chemical compound	Recombinant human bone morphogenetic protein 4 (BMP-4)	R&D Systems	314-BP-010	
Chemical compound	Recombinant mouse stem cell factor (SCF)	R&D Systems	455-MC-010	
Chemical compound	Recombinant human epidermal growth factor (EGF), carrier free (CF)	R&D Systems	2028-EG-200	
Chemical compound	Dulbecco's phosphate-buffered saline (DPBS)	Thermo Fisher Scientific	14040133	
Chemical compound	Deoxyribonuclease I (DNaseI)	Millipore Sigma	DN25	From bovine pancreas
Chemical compound	Trypsin (2.5%)	Thermo Fisher Scientific	15090046	No phenol red
Chemical compound	Soybean trypsin inhibitor	Thermo Fisher Scientific	17075029	
Chemical compound	Collagenase type IV	Worthington	LS004188	From *Clostridium histolyticum*
Chemical compound	Sertoli Cell Medium	ScienCell Research Laboratories	4521	
Chemical compound	Ethanol (EtOH)	Fisher	BP28184	(200 Proof)
Chemical compound	BSA	Millipore Sigma	A9085	
Chemical compound	Heat inactivated (HI) FBS	Thermo Fisher Scientific	10082147	
Chemical compound	Trypsin-EDTA (0.25%)	Thermo Fisher	25200072	With phenol red

		Scientific		
Chemical compound	Bisphenol S (BPS)	Millipore Sigma	43034-100MG	
Chemical compound	Phenol:Chloroform:Iso amyl Alcohol (25:24:1, v/v)	Thermo Fisher Scientific	15593031	
Chemical compound	TRIzol™	Thermo Fisher Scientific	15596026	
Chemical compound	Isopropanol	Thermo Fisher Scientific	327272500	
Chemical compound	Proteinase K Solution (20 mg/mL)	Thermo Fisher Scientific	25530049	
Chemical compound	MaXtract High Density	Quiagen	129046	Phase lock gel tubes
Chemical compound	Sodium Acetate Solution	Thermo Fisher Scientific	R1181	3M, pH 5.2
Chemical compound	Glycogen (5 mg/ml)	Thermo Fisher Scientific	AM9510	
Chemical compound	NaCl	Thermo Fisher Scientific	J21618.36	
Chemical compound	Tris base	Millipore Sigma	77-86-1	
Chemical compound	Ethylenediaminetetra acetic acid (EDTA)	Millipore Sigma	E9884-100G	
Chemical compound	Sodium dodecyl sulfate (SDS)	Millipore Sigma	151-21-3	
Chemical compound	Triton X-100	Thermo Fisher Scientific	85111	
Chemical compound	RQ1 DNase	Promega	M6101	
Chemical compound	Propidium iodide	BioLegend	421301	FCy 5 µL/10^6 cells
Antibody	ERα	Thermo Fisher Scientific	MA1-310	Host: mouse, ICC 1:100
Antibody	ERβ	GeneTex	GTX70174	Host: mouse, ICC 1:100
Antibody	INHA	Invitrogen	PA5-13681	Host: rabbit, ICC 1:25
Antibody	FSHR	Affinity	AF5477	Host: rabbit,

				ICC 1:250
Antibody	SOX9	Abcam	ab185966	Host: rabbit, ICC 1:100
Antibody	GAPDH	Novus	NB300-221	Host: mouse, ICC 1:100
Antibody	WT1	Novus	NBP2-67587	Host: rabbit, ICC 1:100
Antibody	FUT4	GeneTex	GTX34467	Host: rabbit, ICC 1:50
Antibody	NANOG	Abcam	ab80892	Host: rabbit, ICC 1:100
Antibody	POU5F1	Abcam	ab19857	Host: rabbit, ICC 1:200
Antibody	SOX2	Abcam	ab97959	Host: rabbit, ICC 1:200
Antibody	ID4	Thermo Fisher Scientific	PA5-26976	Host: rabbit, ICC 1:50
Antibody	AR	Santa Cruz	sc-7305	Host: mouse ICC 1:50
Antibody	PPARγ	Santa Cruz	sc-7273	Host: mouse ICC 1:200
Antibody	RXRα	Invitrogen	433900	Host: mouse ICC 1:200
Antibody	PRDM1	Thermo Fisher Scientific	14-5963-82	Host: rat ICC 1:50
Antibody	Goat Anti Mouse Alexa 488	Abcam		Host: goat, ICC 1:1000
Antibody	Goat Anti Rabbit Alexa 647	Abcam	ab150199	Host: goat, ICC 1:200
Antibody	Goat Anti Rat Alexa 647	Thermo Fisher Scientific	A21247	Host: goat ICC 1:200
Antibody	FUT4 (IgM, κ), brilliant violet 421	BioLegend	125614	Host: mouse, FCy 5 µL/10^6 cells
Antibody	ITGB3 (IgG), PE	BioLegend	104307	Host: hamster, FCy 1 µL/10^6 cells
Antibody	IgM, κ Isotype control, brilliant violet 421	BioLegend	562704	Host: mouse, FCy 1.25 µL/10^6 cells
Antibody	IgG Isotype control, PE	BioLegend	400907	Host: hamster, FCy 1 µL/10^6 cells
Commercial assay or kit	RNA Clean & Concentrator-5	Zymo Research	R1016	
Commercial assay or kit	Genomic DNA Clean & Concentrator-10	Zymo Research	D4011	
Commercial	EZ DNA Methylation	Zymo	D5001	

assay or kit	Kit	Research		
Commercial assay or kit	SuperScript™ III One-Step RT-PCR System with Platinum™ Taq DNA Polymerase	Thermo Fisher Scientific	12574026	
Commercial assay or kit	PowerTrack™ SYBR Green Master Mix for qPCR	Thermo Fisher Scientific	A46109	
Commercial assay or kit	Infinium Mouse Methylation BeadChip	Illumina	20041558	
Commercial assay or kit	RNA ScreenTape & Reagents	Agilent	5067-5576	
Commercial assay or kit	DNA ScreenTape & Reagents	Agilent	5067-5583	
Commercial assay or kit	Qubit dsDNA (Broad Range) BR Assay Kit	Thermo Fisher Scientific	Q32850	
Commercial assay or kit	Qubit RNA (high sensitivity) HS Assay Kit	Thermo Fisher Scientific	Q32855	
Commercial assay or kit	NEBNext® Ultra™ II Directional RNA Library Prep Kit for Illumina®	New England BioLabs	E7765	
Commercial assay or kit	NEBNext® Poly(A) mRNA Magnetic Isolation Module	New England BioLabs	E3370	
Software, algorithm	ZEISS ZEN Microscopy Software	https://www.zeiss.com/microscopy/en/products/software/zeiss-zen.html	ZEN 3.7	
Software, algorithm	Primer-BLAST	https://www.ncbi.nlm.nih.gov/tools/primer-blast/		
Software, algorithm	QuantSudtio™ Design & Analysis Software	https://www.thermofisher.com/us/en/home/technical-resources/software-downloads/quantstudio-3-5-real-	QuantSudtio™ v1.5.1	

		time-pcr-systems.html		
Software, algorithm	Fiji	https://fiji.sc/	Fiji v1.54f	(Schindelin et al., 2012)
Software, algorithm	Bfastq2	https://support.illumina.com/downloads/bcl2fastq-conversion-software-v2-20.html	Bcl2fastq2 v2.20	
Software, algorithm	FastQC	https://www.bioinformatics.babraham.ac.uk/projects/fastqc/	FastQC 0.12.0	(Andrews et al., 2023; Smith and de Sena Brandine, 2021)
Software, algorithm	Wg-blimp	https://github.com/MarWoes/wg-blimp	Wg-blimp v0.10.0	(Lehle and McCarrey, 2023; Wöste et al., 2020)
Software, algorithm	R-Project for Statistical Computing	http://www.r-project.org/	R 4.2.1	Packages: Sesame (Ding et al., 2023; Triche et al., 2013; Zhou et al., 2022, 2018), stringr (Wickham and RStudio, 2022), kintr (Xie et al., 2023), SummarizedExperiment (Morgan et al., 2023), ggrepel (Slowikowski et al., 2023), pals (Wright, 2023), wheatmap (Zhou, 2022), tidyr (Wickham et al., 2023) magrittr (Bache et al., 2022), dplyr (Wickham et al., 2023), ggplot2 (Wickham et al., 2023), ggvenn (Yan, 2023), RColorBrewer (Neuwirth, 2022), RIdeogram (Hao et al.,

				2020), AnnotationDbi (Pagès et al., 2023), Mus.musculus (Team BC, 2015), BSgenome.Mmusculus.UCSC.mm10 (Team TBD, 2021), GenomicRanges (Lawrence et al., 2013), universalmotif (Tremblay, 2023), memes (Nystrom, 2023), Plyranges (Lee et al., 2019), rtracklayer (Lawrence et al., 2009), Rsubread (Liao et al., 2019), edgeR (Chen et al., 2016; McCarthy et al., 2012; Robinson et al., 2010)

Animal procedures

Mice were used as the primary source of cells for the establishment of cultures for the analysis

of differential cell-type specific susceptibility to exposure to the EDC BPS at the cellular level. All mice

were euthanized prior to dissection and isolation of the tissue or cell type to be used for experiments.

All animals were bred on-site and maintained in one of the UTSA on-campus vivaria under controlled

temperature and humidity conditions in a fixed 12-hour light, 12-hour dark cycle with free access to

5V5R extruded food and non-autoclaved conventional RO water. 5V5R has been verified to contain a

targeted level of 50 PPM total phytoestrogenic isoflavones (genistein, daidzein, and glycitein) and has

been certified for use with estrogen-sensitive protocols limiting the effect of these plant-derived

compounds. Euthanasia was performed by trained personnel using continuous CO_2 exposure at a rate of

3L/min until one minute after breathing had stopped. Euthanasia was confirmed by cervical dislocation.

Tissue from euthanized mice was used to obtain somatic (Sertoli, granulosa, and mouse embryonic

fibroblast [MEF]) cells, and primordial germ cells [PGCs]. Both Sertoli and granulosa cells were isolated

from respective male and female mice euthanized at postnatal day 20 (P20). MEFs and PGCs were

isolated from fetuses at embryonic day 13.5 postcoitum (E13.5).

***In vitro* generation and/or culture of pluripotent, somatic, and germ cells**

Pluripotent cell culture

Male mouse iPSCs were derived from a transgenic mouse line carrying the *tetO-4F2A* cassette

obtained from The Jackson Laboratories (011011) (The Jackson Laboratory, ME USA). iPSCs can be

induced from essentially any cell type carrying this reprogrammable cassette by exposure to Doxycycline

for one week as previously described (Carey et al., 2010; Hochedlinger et al., 2005). For this project,

iPSCs were reprogrammed from MEFs isolated from a single male (XY) E13.5 mouse fetus.

Reprogrammed colonies were picked and expanded via sub-passaging. Validation of reprogrammed

pluripotency was determined by positive immunocytochemical (ICC) staining for the pluripotent markers

POU5F1, SOX2, NANOG, and FUT4 (Supplementary files 16 and 17). Karyotyping was done on each

candidate iPSC line produced to confirm normal chromosome number and XY sex chromosome

constitution of each line prior to aliquots being prepared for long-term storage in liquid nitrogen

(Supplementary file 18). We selected a male (XY) iPSC line to be used for this project. Upon thawing,

iPSCs were initially maintained on CF-1 feeder cells for a minimum of three passages. Cells were cultured

in DMEM supplemented with 15% fetal bovine serum (FBS) and 1000 U/mL leukocyte inhibitory factor

(LIF). Once pluripotent cells were stabilized in culture, they were transitioned to feeder-free conditions

and were cultured in N2B27 media supplemented with 2i and LIF for two-three passages prior to use in

chemical exposure experiments. A complete list of the media components along with catalog numbers

can be found in Supplementary file 19.

Sertoli cell culture

Primary cultures of Sertoli cells were established as previously described (Karl and Griswold,

1990). Briefly, whole testes were dissected from 5 juvenile male littermates at P20. Following removal of

the tunica albuginea from each individual testis, the bundles of seminiferous tubules were physically

chopped using a sterile razor blade to break down the coiled structure of tubules increasing the surface

area and rendering them easier to digest. These shortened fragments of seminiferous tubules were then

digested in a mixture of 2.5% trypsin and 6.64 mg/ml DNaseI in DPBS for 25 minutes at 37°C. After these

enzymes were inactivated, the tubule sections were washed multiple times, then treated with a final

enzymatic mixture of collagenase IV (0.70 mg/mL) and DNaseI (6.64 mg/mL) for 10 minutes to

permeabilize the thick collagen layer on the exterior of the seminiferous tubules to allow the Sertoli cells

to migrate out away from the tubules in culture. After checking the digested tubules under a microscope

to confirm the collagen layer had been permeabilized, the digested tubules were spun down to wash

away the enzymes, and then resuspended in Sertoli cell media containing retinoic acid from ScienCell^{TM}

(4521) (ScienCell^{TM} Research Laboratories, CA USA). The digested tubules were plated into 6 culture

flasks in order to have 3 replicates of both control and treated cells for each exposure experiment. To

remove contaminating germ cells from this primary culture, the cells were treated with hypotonic shock

treatment on the morning of the second day of culture, and the enriched Sertoli cells were washed and

allowed to recover with fresh media for at least two hours. The enriched primary cultures of Sertoli cells

were then ready to be used for chemical exposure experiments. The estimated purity of the culture was

> 80% based on ICC staining for the Sertoli cell markers SOX9 and WT1 (Supplementary file 17). A full

procedure for the establishment of primary cultures of Sertoli cells can be found in Supplementary file

20.

Granulosa cell culture

Primary cultures of granulosa cells were established by selecting early preantral primary follicles

from enzymatically digested ovary tissue as described previously (Monti and Redi, 2016; Roy and

Greenwald, 1996), followed by breaking selected primary follicles down into a single cell suspension that

could be plated and maintained. To isolate preantral primary follicles we dissected ovaries from 5

female littermates at P20 and then enzymatically digested the ovary using collagenase IV (560 U/pair of

ovaries) for 25 minutes at 37°C with constant agitation. This digestion was stopped by addition of buffer

with 0.5% BSA and the mixture was spun down at 60xg for 5 minutes to pellet the cells. The cells were

resuspended in PBS with 0.5% BSA and transferred to a sterile petri dish under a stereomicroscope.

Primary follicles were individually picked from the solution using a glass needle with suction control and

moved into a clean droplet of PBS containing 0.5% BSA. Selected primary follicles were then spun down

at 1000xg for 5 mins. After spinning, the supernatant was removed, and the cells were finger-vortexed

to resuspend the pellet. The follicles were then digested to single cell suspension by addition of pre-

warmed (37°C) 0.25% trypsin and incubated for 5 mins with regular pipetting followed by pelleting again

at 1000xg for 5 mins. The trypsin-containing supernatant was removed, and the granulosa cells were

resuspended in DMEM supplemented with 15% FBS and plated into 6 culture flasks in order to have 3

replicates of both control and treated cells for each exposure experiment. The oocytes were non-

adherent and were washed away on day 2 when changing the media. The estimated purity of the

culture was > 90% based on ICC staining for the granulosa cell marker FSHR and INHA (Supplementary

file 17). The granulosa cells were then ready to be used for EDC exposure experiments. A full procedure

for the establishment of primary cultures of granulosa cells can be found in Supplementary file 21.

Primordial germ cell like cell culture

Mouse iPSCs were differentiated into PGCLCs as previously described (Hayashi et al., 2011).

Briefly, iPSCs maintained in N2B27 supplemented with 2i and LIF under feeder-free conditions were

differentiated to EpiLCs for two days by the addition of activin A and basic fibroblast growth factor

(bFGF) to the N2B27 media. After two days the EpiLC intermediate cells were passaged to low

adherence round bottom plates for 4 days in GK15 media containing bone morphogenic protein 4

(BMP4), stem cell factor (SCF), epidermal growth factor (EGF), and LIF to induce the differentiation of a

subset of cells (2-3%) in the resulting aggregate to become PGCLCs. 6 batches each being made up of 6

plates were required in order to obtain sufficient cell numbers for 3 replicates of both control and

treated conditions. During this four-day period, cell aggregates were ready for EDC exposure (see

below). Following EDC exposure, cell aggregates were removed from the low adherence round bottom

plates within each batch, dissociated into a single cell suspension, and PGCLCs were fluorescence-

activated cell sorted (FACS) on a BD FACSAria™ II in the UTSA Cell Analysis Core to recover cells that

were double-positive for FUT4 and ITGB3. A full list of the media components and catalog numbers for

the differentiation of iPSCs to PGCLCs along with our FACS gating for double positive FUT/ITGB3 PGCLCs

can be found in Supplementary files 19 and 22, respectively. Additional data demonstrating validation of

PGCLCs by qRT-PCR and ICC staining are shown in Supplementary file 23.

Immunocytochemistry (ICC)

Cells were immunolabeled as previously described (Rodig, 2022) to validate the purity of cell

primary cultures using known cell specific markers (Supplementary file 17) to detect the presence or

absence of relevant endocrine receptors ERα and ERβ at the protein level. Pluripotent (iPS) and somatic

(Sertoli and granulosa) cells were grown on 13mm Thermanox® plastic coverslips (174950) prior to

fixation, permeabilization, and immunolabeling (Nalge Nunc International, NY USA). Non-adherent germ

(PGC and PGCLC) cells were spun down onto poly-L-lysine coated slides (63410-01) at 400 RPM for one

minute using a Thermo Shandon CytoSpin III Cytocentrifuge from Rakin (Electron Microscopy Sciences,

PA USA & Rakin Biomedical Corporation, MI USA). Cells were fixed with 4% formaldehyde for 10 mins at

room temp (RT) and then washed three times for five minutes each with ICC buffer (PBS containing

0.01% Triton X-100 detergent) to permeabilize the cell and nuclear membranes prior to blocking with 5%

goat serum which was added to the ICC buffer and incubated for one hour at RT. Following blocking,

primary antibodies, in ICC buffer, were added to the slides and left to incubate overnight at 4°C. The

following day, slides were washed three times with ICC buffer, and fluorescent secondary antibodies in

ICC buffer were then added to the slides and left to incubate in the dark for 1 hour. After secondary

antibody incubation, cellular nuclei were stained with DAPI at a 1:1000 dilution in ICC buffer for 7

minutes, followed by 3 final washes with ICC buffer for 5 minutes. Coverslips then were transferred to

microscope slides and mounted with 5-10 µL of VECTASHEILD Antifade Mounting Medium (H-1000) and

sealed with clear nail polish before being imaged on a Zeiss AXIO Imager.M1 Fluorescence Microscope

(Vector Laboratories Inc, CA USA & Zeiss Group, Oberkochen DE). Images were processed for contrast

and brightness enhancement and for the addition of scale bars using Fiji (Schindelin et al., 2012).

Information about the primary and secondary antibodies along with catalogue numbers and dilutions

used can be found in the Key resources table.

Quantitative RT-PCR (qRT-PCR)

Total RNA was extracted from 3 replicate preps of cells as described previously (Rio et al., 2010)

and treated with 1.5 U/µg total RNA RQ1 DNase1 (M6101) to remove contaminating genomic DNA

(Promega^{TM} Corporation, WI USA). 50 ng of cleaned RNA was retrotranscribed with SuperScript™ III as

recommended by the manufacturer (Invitrogen, MA USA). Primers were designed using Primer-BLAST

(Ye et al., 2012) from NCBI. A complete list of all primer sequences used in this study can be found in

Supplementary file 24. Relative expression levels of selected genes were assessed by real-time PCR using

the PowerTrack™ SYBR Green Master Mix according to the manufacturer's instructions (Applied

Biosystems, MA USA) then run on a QuantStudio™ 5 Real-Time PCR System (Thermo Fisher Scientific,

MA USA) and analyzed with QuantStudio™ Design & Analysis Software. Each sample was normalized

based on constitutive expression of the *Gusb* reference gene to obtain the ΔCt [(2-[Ctgene-CtGusb)].

BPS exposure design

Despite numerous studies illustrating the dangers of estrogenic mimetic EDCs, the EPA has

currently not published any limits concerning the maximum acceptable dose of BPS below which

exposure on a daily basis is considered to be safe. Thus for this study, we established our initial testing

range of 1 µM, 50 µM, and 100 µM of BPS based on the limit established for BPA at ≤ 4.44 µM (Ribeiro

et al., 2019) selecting 1 dose below the established safe limit and two doses that exceed that limit. For

all cell types except PGCLCs, BPS was dissolved in absolute ethanol and added to media gassed with 5%

CO_2, 5% O_2, and balanced N_2 and then injected into 3 replicate sealed T-25 cell culture flasks and left to

incubate for 24 hours (Supplementary file 25). The media was gassed to ensure enrichment to 5% CO_2,

which is normally regulated by the cell incubator, limiting the potential for the pH to change in the

media during this exposure period due to the absorption of CO_2 by incubating cells. The concentration of

either BPS dissolved in ethanol for treatment groups or ethanol vehicle alone for control groups made

up 0.02% of the total media volume. Following 24 hours of exposure, media containing BPS was

removed and cells were washed and allowed to recover in fresh untreated media (without BPS or EtOH)

for 24 hours prior to harvesting. PGCLCs were not suitable for this exposure paradigm as EpiLCs undergo

differentiation to PGCLCs in cell aggregates formed in low-adherence round-bottom 96 well plates and

cannot be maintained in T-25 sealed flasks. Therefore, diluted BPS was added to PGCLC media and

added to cells in 3 replicate batches of round-bottom 96 well plates to incubate for 24 hours prior to a

shortened wash and "chase" period of 8 hours prior to cell sorting.

Methylation beadchip analysis

A total of 1µg of extracted genomic DNA from each of 3 replicate exposure experiments for each

cell type was bisulfite-converted with the EZ DNA Methylation Kit (D5001) and modified according to the

manufacturer's recommendations (Zymo Research, CA USA & Illumina, CA USA). These samples were

run on the Infinium Mouse Methylation BeadChip Array following the Illumina Infinium HD Methylation

protocol. This beadchip array includes 297,415 cytosine positions within the mouse genome (CpG sites,

non-CpG sites, and random SNPs). The methylation score for each CpG is represented as a β-value which

is a ratio of the fluorescence intensity ranging between 0 (unmethylated) and 1 (methylated). Arrays

were scanned by HiScan (Illumina, CA USA). Quality control (Supplementary file 26) and downstream

data processing of the .idat files was using the Sensible Step-wise Analysis of DNA Methylation

BeadChips (SeSAMe) Bioconductor package (Ding et al., 2023; Triche et al., 2013; Zhou et al., 2022,

2018). DNA methylation levels of differentially methylated cytosines (DMCs) are determined using

mixed linear models. This general supervised learning framework identifies CpG loci whose differential

methylation is associated with known control vs. treated co-variates. CpG probes on the array were

defined as having differential changes that met both p-value and FDR ≤ 0.05 significant thresholds

between treatment and control samples for each cell type analyzed. Additionally, we followed up our

DNA methylation analysis of individual dCpGs by identifying differentially methylated regions (DMRs).

DMRs were created by grouping all CpGs measured on the array into clusters using Euclidean distance

(Ding et al., 2023; Triche et al., 2013; Zhou et al., 2022, 2018). The p-values from the differential

methylation of individual CpGs within the resulting CpG clusters were aggregated, and clusters were

then filtered selecting for regions that contained a p-value ≤ 0.05 (Ding et al., 2023; Triche et al., 2013;

Zhou et al., 2022, 2018).

RNA-seq

Total RNA was extracted from 3 replicate preps of cells using Trizol as previously described (Rio

et al., 2010). Contaminating genomic DNA was removed by RQ1 DNase (M6101) treatment according to

the manufacturer's instructions (Promega Corporation, WI USA). RNA concentration was determined

using Qubit (Q32855) and RNA integrity (RIN) scores were determined using tape station (5067-5576)

according to the manufacturer's instructions (Thermo Fisher Scientific Inc, MA USA & Agilent

Technologies, Inc CA USA). Strand-specific RNA-seq libraries were prepared with the NEBNext® Ultra™ II

Directional RNA Library Prep Kit for Illumina® sequencing (E7760S) according to the manufacturer's

protocol (New England Biolabs, MA USA). Briefly, this process consisted of poly(A) RNA selection, RNA

fragmentation, and double stranded cDNA generation using random oligo(dT) priming followed by end

repair to generate blunt ends, adaptor ligation, strand selection, and polymerase chain reaction

amplification to generate the final library. Distinct index adaptors were used for multiplexing samples

across multiple sequencing lanes. Sequencing was performed on an Illumina NovaSeq 6000 instrument

yielding sequences of paired end 2x50 base pair runs. Demultiplexing was performed with the Illumina

Bcl2fastq2 program (Illumina, CA USA).

RNA-seq expression analysis

The quality of the fastq reads was checked using FastQC (Andrews et al., 2023; Smith and de

Sena Brandine, 2021) (Supplementary file 27). Reads were aligned to the mm10 mouse reference

genome using the Rsubreads package to produce read counts (Liao et al., 2019). These were then used

for differential gene expression analysis using the edgeR package (Chen et al., 2016; McCarthy et al.,

2012; Robinson et al., 2010). Briefly, gene counts were normalized using the trimmed mean of M-value

normalization (TMM) method before determining counts per million (CPM) values (Robinson and

Oshlack, 2010). For a gene to be classified as showing differential gene expression between BPS-treated

and EtOH vehicle-only control samples, a threshold of both a Benjamini-Hochberg adjusted p-value ≤

0.05 and a false discovery rate (FDR) ≤ 0.05 had to be met.

EM-Seq

Genomic DNA was extracted from 3 replicates of cells as previously described (Sambrook and

Russell, 2006). Smaller fragmented DNA (≤ 10 kb) and contaminating RNA were removed by cleaning the

genomic DNA on a genomic DNA clean and concentrator^{TM}-10 column (D4011), according to the

manufacturer's instructions (Zymo Research, CA USA). DNA concentration was determined using a Qubit

(Q32850) and genomic DNA (100 ng) was sheered using a Bioruptor® and the size of the sheered DNA

was determined using a TapeStation 4200 according to the manufacturer's instructions in the UTSA

Genomics Core (Diagenode Inc, NJ USA & Agilent Technologies, Inc CA USA). Large size (470-520 bp) EM-

seq libraries were prepared with the NEBNext® Enzymatic Methyl-seq kit (E7120S) according to the

manufacturer's protocol (New England Biolabs, MA USA). Briefly, this process consisted of A-tailing,

adaptor ligation, DNA oxidation by TET2 initiated by the addition of Fe (II), strand denaturization with

formamide, deamination by APOBEC3A, polymerase chain reaction amplification, and bead selection to

generate the final libraries. Distinct index adaptors were used for multiplexing samples across multiple

sequencing lanes. Sequencing was performed on an Illumina NovaSeq 6000 instrument yielding

sequences of paired end 2x150 base pair runs. Demultiplexing was performed with the Illumina

Bcl2fastq2 program (Illumina, CA USA).

EM-Seq analysis of genome-wide DNA methylation patterns

EM-seq data was processed using the comprehensive wg-blimp v10.0.0 software pipeline (Lehle

and McCarrey, 2023; Wöste et al., 2020). In brief, reads were trimmed prior to initiating the wg-blimp

pipeline using Trim Galore (https://www.bioinformatics.babraham.ac.uk/projects/trim_galore/).

Sequenced reads were aligned to the mm10 genome using gemBS (King et al., 2020; Merkel et al., 2019;

Schilbert et al., 2020). The BAM files from alignment underwent a series of QC tests including read

quality scoring by FastQC (Andrews et al., 2023), overall and per-chromosome read coverage calculation,

GC content, duplication rate, clipping profiles by Qualimap (Okonechnikov et al., 2016), and

deduplication by the Picard toolkit (https://broadinstitute.github.io/picard/) (Supplementary file 28).

Methylation calling was performed by MethylDackel (https://github.com/dpryan79/MethylDackel) and

statistically significant DMC/DMR calling was performed by the metilene (Jühling et al., 2016) and

BSmooth (Hansen et al., 2012) algorithms. Metilene uses a binary segmentation algorithm combined

with a two-dimensional statistical test that allows the detection of DMCs/DMRs in large methylation

experiments with multiple groups of samples. BSmooth uses a local-likelihood approach to estimate a

sample-specific methylation profile, then computes estimates of the mean differences and standard

errors for each CpG to form a statistic similar to that used in a t-test. Finally, potential regulatory regions

were identified through the use of MethylSeekR (Burger et al., 2013). Results from the pipeline were

displayed in the wg-blimp interactive results web browser that was built using the R Shiny local browser

hosting framework (Chang et al., 2023).

Data availability

Methylation beadchip, EM-seq, and RNA-seq data discussed in this publication have been uploaded and

can be obtained from the SRA BioProject PRJNA1026145 and GEO accession GSE252723.

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Ethics

902   All experimental procedures involving animals were approved in advance by the UTSA IACUC (Protocols

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