1 Genome-wide disruption of DNA methylation by 5-aza-2'-deoxycytidine in

2 the parasitoid wasp Nasonia vitripennis

- 3 Cook N¹*, Parker DJ^{2,3}*, Turner F⁴, Tauber E⁵, Pannebakker BA⁶ and Shuker DM¹
- 4
- 5 *These authors contributed equally
- 6
- 7 1. School of Biology, University of St Andrews, UK
- 8 2. Department of Ecology and Evolution, University of Lausanne, Switzerland
- 9 3. Swiss Institute of Bioinformatics, Lausanne, Switzerland
- 10 4. Edinburgh Genomics, School of Biological Sciences, University of Edinburgh, UK
- 11 5. Faculty of Natural Sciences, University of Haifa, Israel
- 12 6. Laboratory of Genetics, Wageningen University, The Netherlands
- 13
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18 Abstract

19 DNA methylation of cytosine residues across the genome influences how genes and phenotypes are 20 regulated in a wide range of organisms. As such, understanding the role of DNA methylation and other 21 epigenetic mechanisms has become very much a part of mapping genotype to phenotype, a major 22 question in evolutionary biology. Ideally, we would like to manipulate DNA methylation patterns on a genome-wide scale, to help us to elucidate the role that epigenetic modifications play in phenotypic 23 24 expression. Recently, the demethylating agent 5-aza-2'-deoxycytidine (5-aza-dC; commonly used in the epigenetic treatment of certain cancers), has been deployed to explore the epigenetic regulation 25 26 of a number of traits of interest to evolutionary ecologists, including facultative sex allocation in the 27 parasitoid wasp Nasonia vitripennis. In a recent study, we showed that treatment with 5-aza-dC did 28 not ablate the facultative sex allocation response in Nasonia, but shifted the patterns of sex allocation 29 in a way predicted by genomic conflict theory. This was the first (albeit indirect) experimental evidence 30 for genomic conflict over sex allocation facilitated by DNA methylation. However, that work lacked 31 direct evidence of the effects of 5-aza-dC on DNA methylation, and indeed the effect of the chemical 32 has since been questioned in *Nasonia*. Here, using whole-genome bisulphite sequencing of more than 4 million CpGs, across more than 11,000 genes, we demonstrate unequivocally that 5-aza-dC disrupts 33 methylation on a large scale across the Nasonia vitripennis genome. We show that the disruption can 34 35 lead to both hypo- and hyper-methylation, may vary across tissues and time of sampling, and that the 36 effects of 5-aza-dC are context- and sequence specific. We conclude that 5-aza-dC does indeed have 37 the potential to be repurposed as a tool for studying the role of DNA methylation in evolutionary 38 ecology, whilst many details of its action remain to be discovered.

40 Introduction

41 The genomics revolution has transformed our understanding of how genomes build phenotypes. The 42 extraordinary diversity of morphologies, behaviours, and physiological adaptations we see across the natural world are not necessarily the product of very different sets of genes, or very different numbers 43 44 of genes in organisms, but rather differences in how, where and when genes are expressed during 45 development. The genomics revolution has therefore been as much about understanding gene 46 regulation as it has been about identifying allelic variants at genetic loci (Ashbrook et al. 2017, Biémont 47 et al. 2010). At the forefront of our growing knowledge of gene regulation has been the role of 48 epigenetic modification of DNA, including DNA methylation. In the vast majority of cases, DNA 49 methylation occurs at the C5 position of the cytosine ring in CpG dinucleotides and is catalysed by 50 DNA methyltransferase enzymes (Dnmts). Three types of Dnmt are known in eukaryotes; Dnmt3 51 establishes DNA methylation patterns de novo, Dnmt1 maintains these patterns, and Dnmt2 is 52 involved in tRNA methylation (Lyko 2018). In mammals, broadly, DNA methylation is found throughout 53 the genome (except at CpG islands near the promoters of genes) and is associated with transcriptional 54 repression (Feng et al. 2010, Zemach et al. 2010). By contrast, in insects, DNA methylation is 55 concentrated in gene bodies and is associated with more stable patterns of gene expression (Bewick 56 et al. 2017). Notably, vertebrate and invertebrate DNA methylation varies greatly in its extent 57 throughout the genome, with the proportion of methylated CpGs much lower in invertebrates; 0-14% 58 as compared to 60-90% in mammals (Bewick et al. 2017, Glastad et al. 2011). The evolutionary reasons 59 for these differences are not yet clear, although it is thought that gene body methylation is an ancient 60 feature of eukaryote genomes (Zemach et al. 2010). However, DNA methylation status influences how 61 genes (and phenotypes) are expressed in many organisms, including reproductive vs non-reproductive 62 worker bumblebees (Amarasinghe et al. 2014), caste differentiation (Kucharski et al. 2008, Lyko et al. 63 2010), and learning and memory (Li et al. 2017) in honey bees, and behavioural division of labour in 64 ants (Pogonomyrmex barbatus; Smith et al. 2012; but see Libbrecht et al. 2016 for a robust critique of 65 earlier studies of DNA methylation and caste determination in social insects). Therefore, DNA 66 methylation patterns are very much part of mapping genotypes to phenotypes.

The flexible control of gene expression provided by mechanisms such as DNA methylation has also caught the interest of biologists interested in the underlying mechanisms of phenotypic plasticity, and indeed for evolutionary biologists interested in how such plasticity may evolve (Bonduriansky and Day 2009, 2018). Moreover, there has also been great interest in the extent to which epigenetic modifications may be inherited across generations (Lind and Spagopoulou 2018, Burggren 2016, Bondurianksy and Day 2009). Transmission of epigenetic state allows transmission of parent-of-origin information, or so-called genomic imprinting, setting the scene for genomic conflict (Patten *et al.*

74 2014, Haig 2002). Genomic conflict arises when maternally- and paternally- inherited alleles have 75 different evolutionary optima for traits expressed in an individual (Haig 1997, 2002). A classic example 76 is the expression of maternally- or paternally- inherited alleles during embryonic development in 77 mammals: paternally-inherited alleles are selected to extract more resources from the mother in utero 78 compared to maternally-inherited alleles (Moore and Haig 1991). This conflict is now known to be, at 79 least in part, facilitated by parent-of-origin specific gene expression of insulin-like growth factor II 80 (Igf2) and Igf2r. In mammals, the Igf2 gene promotes growth and cellular differentiation during 81 development and also regulates the placental supply of nutrients and the demand of nutrients by the 82 foetus (Constância et al. 2002).

To further understand the role epigenetic modifications such as DNA methylation have in 83 84 influencing phenotypes and their evolution, we need robust ways of manipulating them. Here we 85 consider the efficacy of the chemical 5-aza-2'-deoxycytidine (5-aza-dC) as a way of manipulating DNA methylation, using the parasitoid wasp Nasonia vitripennis. The extent and nature of DNA methylation 86 87 across the *N. vitripennis* genome has recently been characterised (Wang et al. 2013; Pegoraro et al. 88 2016), and the presence in the genome of a full "methylation toolkit" has also been confirmed (Werren 89 et al. 2010). 5-aza-dC is well-known from the cancer literature as a hypomethylating agent, and is used 90 in the treatment of acute myeloid leukaemia and myelodysplastic syndrome (Seelan et al. 2018, Bryan 91 et al. 2011, Momparler 2012). In cancer patients, many genes that suppress leukemogenesis in 92 humans are silenced by aberrant DNA methylation. 5-aza-dC acts as a demethylating agent to 93 reactivate these genes (see below for details of action). In recent years, 5-aza-dC has been used across 94 a range of species as an experimental tool to explore how DNA methylation influences phenotypes 95 including diapause in Nasonia vitripennis (Pegoraro et al. 2016) and worker reproductive status in 96 bumble bees (Amarasinghe et al. 2014). Previously, we used 5-aza-dC to explore whether DNA 97 methylation was involved in the control of facultative sex allocation in N. vitripennis (Cook et al. 98 2015a).

99 Like all Hymenoptera, N. vitripennis is haplodiploid, with males arising from unfertilised 100 haploid eggs and females arising from fertilised diploid eggs. Females have putative control over the 101 sex of their offspring, releasing sperm to produce daughters or not to produce sons. Female Nasonia 102 allocate sex broadly in line with the predictions of Local Mate Competition theory (LMC: Hamilton 103 1967, Taylor and Bulmer 1980, Werren 1980, 1983, Shuker and West 2004, Burton-Chellew et al., 104 2008; reviewed by West 2009). Females laying eggs alone on their blowfly pupae hosts produce very 105 female biased offspring sex ratios to minimise competition amongst her sons for mating opportunities. 106 As more females (foundresses) lay eggs on the same host(s) together, females produce less female-107 biased sex ratios, as LMC is reduced and the fitness gained through the production of sons and

108 daughters moves towards equality (Hamilton 1967). When treated with 5-aza-dC, the facultative sex 109 allocation response in female Nasonia was maintained, suggesting that this plastic behaviour in and 110 of itself did not require certain patterns of methylation. Rather, the sex ratios (as proportion male) 111 produced by females were shifted slightly upwards, with more sons produced by treated females than 112 by controls. This subtle shift is as predicted by genomic conflict over sex allocation theory, which 113 requires some form of genomic imprinting (Wild and West 2008). Assuming that DNA methylation is 114 the mechanism by which parent-of-origin allelic information is carried in Nasonia, our data therefore 115 provided the first (albeit indirect) evidence for genomic conflict influencing sex allocation (Cook et al. 116 2015a).

Recently, Ellers et al. (2018) challenged the effects of 5-aza-dC as a general modifier of 117 118 methylation status, considering CpGs in ten genes in Nasonia vitripennis. Ellers et al. (2018) failed to find changes in methylation status across any of the 155 CpGs in their study, questioning the action of 119 120 5-aza-dC in Nasonia, and hence of course the interpretation of the effects of 5-aza-dC in our earlier 121 study (Cook et al. 2015a). Fully negative results are of course very hard to interpret, but Ellers et al. 122 (2018) nonetheless raised valid concerns of the efficacy of the chemical. As noted above, levels of 123 methylation in insects is generally low, and a genome-wide study would provide greater resolution to 124 explore the effects of this chemical in an ecologically valuable model system. Here we consider the 125 effects of 5-aza-dC across the whole genome using a Bisulfite sequencing (BS-Seq) experiment. There 126 is evidence to suggest that the effects of 5-aza-dC are unlikely to be random across methylated CpGs 127 in a given genome. For instance, in mammals, genomic context can influence the extent to which 5-128 aza-dC is effective as a demethylating agent. DNA sequence context, the distribution of transcription 129 factor binding sites, and chromatin structure are known to influence the DNA methylation patterns 130 produced by 5-aza-dC exposure (Ramos et al. 2015, Mossmann et al. 2010, Hagemann et al. 2011). 131 Although patterns of methylation differ among eukaryotes, the epigenetic modification itself is 132 chemically identical. Here, we verify that 5-aza-dC influences patterns of methylation across the 133 genome in N. vitripennis. Altered methylation patterns produced by 5-aza-dC exposure were non-134 random, with enrichment in genes associated with transcription-factor activity and sequence specific 135 binding.

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138 Materials and Methods

139 Study species

Nasonia vitripennis (Hymenoptera, Chalcicoidea) is a generalist parasitoid of large dipteran pupae 140 141 including species of Calliphoridae. Females oviposit between 20 and 50 eggs in an individual host, with 142 male offspring emerging just before females (after approximately 14 d at 25 °C: Whiting 1967). Males 143 are brachypterous and unable to fly, remaining close to the emergence site where they compete with 144 each other for emerging females, including their sisters. Females disperse after mating to locate new 145 hosts. The females used in this experiment were from the wild-type AsymC strain, originally isolated 146 in 1986 by curing the wild-type strain LabII of Wolbachia. Wasps have been maintained on Calliphora vomitoria or C. vicina hosts at 25 °C, 16L:8D light conditions ever since, and AsymC is also the 147 reference genome strain (Werren et al. 2010). The quantitative genetic basis of sex ratio variation has 148 been repeatedly quantified (reviewed in Pannebakker et al. 2008, 2011) whilst whole-genome 149 150 transcription studies have confirmed that facultative sex allocation takes place without discernible 151 changes in gene expression (Cook et al. 2015b, 2018a).

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153 5-aza-2'-deoxycytidine

We used 5-aza-dC (Sigma Aldrich Company Ltd., Gillingham, UK) at a non-lethal concentration ($10 \mu M$). 154 155 Widely used in epigenetic anticancer treatments (Seelan et al. 2018), 5-aza-dC is a nucleoside analogue that works by inhibiting DNA methyltransferase activity (Piskala and Sorm 1964, Jones and 156 157 Taylor 1980, Momparler 1985, Christman 2002). When initially introduced to a cell, 5-aza-dC is 158 essentially inert but is then converted by processes inherent in the cell to its active form 5-aza-dCTP (Seelan et al. 2018). This active form is readily incorporated into DNA during S-phase in the place of 159 160 cytosine. In the daughter strands arising from a replication event, DNA methyltransferase 1 (Dnmt1) 161 will attempt to restore the patterns of methylation that were present in the parent strand. However, Dnmt1 becomes irreversibly bound to the 5-aza-dCTP residues in the daughter strands, resulting in a 162 diminishing pool of Dnmt1. Thus, administering 5-aza-dC results in a passive loss of DNA methylation; 163 164 the key effect of 5-aza-dC (for a fuller description, see Cook et al, 2018b).

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166 Experimental design

To control for possible host and other maternal effects, experimental females were not drawn straight
 from stock populations. Instead, two-day-old, mated, wild-type AsymC females were isolated from

169 the mass cultures into individual glass vials. Each female was provided with three hosts and allowed 170 to oviposit. Experimental females were drawn from the F1 generation, two days after emergence, one 171 female per "grandmother". We then employed a three by three factorial design whereby females 172 were allocated to one of three 5-aza-dC exposure regimes and were harvested post-feeding at three 173 time points. Exposure regimes were a) 20% sucrose for 24h, b) 20% sucrose supplemented with 10 174 μ M 5-aza-dC for 24h and c) 20% sucrose supplemented with 10 μ M 5-aza-dC for 48h. For each of the 175 three regimes, females were harvested at 0, 24 or 48h post-exposure. Throughout the experiment, 176 females were kept individually in glass vials and incubated at 25 °C, 16L :8D. Solutions (200 µl) were provided in the lid of a 2 ml microcentrifuge tube. For individuals exposed to 5-aza-dC for 48 h, the 177 solutions were topped up by 100µl half-way through the exposure period to compensate for 178 179 evaporation. All solutions were stained green with food colouring such that a subset of 10 females 180 from each treatment combination could be dissected to check that feeding had occurred. At the point 181 of harvest, females were flash-frozen in liquid nitrogen and stored at -80 °C prior to DNA extraction.

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183 DNA extraction and bisulphite-conversion

184 Prior to DNA extraction, individuals were removed from -80 °C storage and heads were quickly excised 185 from bodies using a sterile scalpel and a cryloyser to keep samples cold. Heads (and likewise bodies) 186 from the same treatment combination were randomly pooled into groups of 10 to give a maximum of 187 3 biological replicates for each treatment combination giving a total of 54 DNA samples. Briefly, tissue 188 was homogenised using a micropestle in 350 µl CTAB (Hexadecyltrimethylammonium Bromide) buffer 189 with subsequent overnight proteinase K digestion (400 µg) at 56 °C. Samples were cooled to room 190 temperature prior to RNase A digestion for 1 hour at 37 °C. A chloroform: isoamyl-alcohol wash (300 191 µl) was performed twice before ethanol/sodium acetate precipitation of the DNA. DNA purity was 192 analysed using the Nanodrop spectrophotometer with all samples having a 260/280 ratio of ≥ 1.8 . DNA 193 integrity was confirmed using agarose gel electrophoresis.

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195 Bisulphite sequencing

Bisulphite conversion, library preparation and sequencing were carried out by Edinburgh Genomics.
Briefly, bisulphite conversion was carried out using the EZ DNA Methylation Kit (Zymo Research, Irvine
USA), designed to reduce template degradation during the harsh bisulphite treatment whilst ensuring
complete conversion of the DNA. Library preparation was carried out using the TruSeq DNA
methylation Kit which takes the single-stranded DNA that results from bisulphite treatment and

converts it into an Illumina sequencing library. Sequencing of bisulphite-treated DNA was carried out
 on the Illumina HiSeq 2500 in high output mode with 125bp paired-end reads. Library preparation
 failed for three of the samples due to low levels of input DNA, leaving only two biological replicates
 for three out of 18 treatment combinations and a total of 51 Bisulphite-sequenced samples (Table 1).

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206 Raw data processing

Initial data processing was carried out by Edinburgh Genomics. Briefly, reads were trimmed using 207 208 Cutadapt version 1.12 removing any adapter sequences and for quality using a threshold of 30. After 209 trimming reads were required to have a minimum length of 50 bp. Nasonia vitripennis genome version 210 2.1 and the associated annotation, available for download at Ensembl (http://metazoa.ensembl.org/Nasonia vitripennis/Info/Index), were used as a reference. Reads were 211 212 aligned to the reference genome using Bismark (version 0.16) with the parameter "bowtie2". PCR 213 duplicates were removed from the resulting BAM files using samtools with the parameter "view -F 214 1024". Data for each CpG was then extracted from the BAM files using the Bioconductor package methylKit (version 0.99.3) with R (version 3.3.1). The methylation status of each CpG for all samples 215 216 was calculated using the function "processBismarkAln" using the default parameters, except 217 "min.cov=1 no;ap=TRUE", which excludes CpG's not covered by any reads across all samples and does 218 not count CpGs covered by overlapping ends of a paired read as covered twice. The 1% of bases with 219 the highest coverage were removed using the function "filterByCoverage" to eliminate potential PCR 220 bias. After initial processing as outlined above, the raw data comprised percentage methylation at 221 4,150,376 CpG loci (no. of reads in which the CpG is methylated/total no. of reads covering the CpG) 222 for each of 51 samples. CpG read counts were then summed by gene, (including 1Kb up and 223 downstream) using gene annotation from Ensembl (version 2.1. available from: 224 ftp://ftp.ensemblgenomes.org/pub/metazoa/release-40/gff3/nasonia vitripennis). Around half 225 (2,119,643) of the CpG's in the full dataset fell within genic regions with 14,765/17,279 genes (85%) 226 having at least one CpG present.

227

228 Statistical analysis

The overall structure of our dataset in terms of the proportion of methylated reads per gene and per loci was visualised with two complementary analyses: a principle component analysis (PCA), which allows samples to be visualised in relation to the two major axes of variation in our experiment, and a T-distributed Stochastic Neighbour Embedding (t-SNE) analysis. t-SNE is a machine-learning algorithm

for data visualisation well-suited for high-dimensional datasets (van der Maaten and Hinton 2008). The PCA was conducted using the R base function prcomp with default values. To perform the t-SNE analysis we used the R package Rtsne (Krijthe 2015), with default values and perplexity set to 15. For the PCA and t-SNE analyses any genes or loci with 0 coverage in any one sample were discarded prior to analysis.

238 To examine the effect of 5-aza-dC exposure on individual genes, we modelled the number of 239 methylated and unmethylated reads from a gene as a generalized linear model (GLM) with a binomial 240 distribution with the following terms: exposure, tissue-type, collection time and their interactions. 241 Genes with 0 coverage for all samples in any particular treatment group were excluded from our analysis (leaving a total of 11582 genes). Significance of each term was determined by a likelihood 242 243 ratio test, which was corrected for multiple testing using Benjamini and Hochberg's algorithm 244 (Benjamini and Hochberg 1995), with statistical significance set to 5%. Coefficients (log odds of methylation) for each term were extracted from the full model. All statistical analyses were conducted 245 246 in R (v. 3.4.1; R Core Team 2017).

247

248 Functional enrichment

249 Ν. vitripennis GO-terms downloaded "hymenopteramine" were from (http://hymenopteragenome.org/hymenopteramine/begin.do) on 27th September 2018. In addition, 250 we used hymenopteramine to firstly obtain *D. melanogaster* orthologs for each *N. vitripennis* gene, 251 from which we could then obtain the associated *D. melanogaster* GO terms. We used both of these 252 253 GO-term sets for enrichment analyses (below). Some N. vitripennis genes had multiple D. 254 melanogaster orthologs. In these cases, all GO-terms from each D. melanogaster ortholog were 255 combined together. To ensure this did not bias our results we repeated our analyses, but instead of 256 combining GO-terms we kept only GO-terms that were shared between all orthologs.

To examine which processes are most affected by 5-aza-dC exposure, we conducted gene set enrichment analyses using the R package TopGO (v. 2.28.0) (Rahnenfuhrer 2016) using the elim algorithm to account for the GO topology. Gene set enrichment analyses identify enriched GO terms in a threshold-free way, by finding GO-terms that are overrepresented at the top of a ranked list of genes. Here we ranked genes by FDR for the exposure effect. GO terms were considered to be significantly enriched when p < 0.05.

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264 Results

Median coverage across the whole experiment (N = 51 BS-seq libraries) was 916X, but coverage was heterogeneous between loci (Figure 1) and samples (Table S1). Mean read coverage per sample ranged from 2.7X to 49.4X. Four samples had less than 10X mean coverage per CpG, however, these samples were distributed across treatment groups.

269 Treatment with 5-aza-dC had wide-ranging effects on DNA methylation across the genome. 270 First, samples clustered strongly by exposure regime at the gene-level in both the t-SNE visualisation 271 (Figure 2) and in the principle component analyses (Figure 3), indicating that 5-aza-dC has a strong 272 effect on the methylation of genes. Both analyses show the strong clustering at the gene loci, and that 273 wasps exposed for 24 hours are more like controls than wasps treated for 48 hours. In addition, the 274 PCA analysis showed clear separation along PCA1 (which presumably is heavily associated with exposure regime; Figure 3). Importantly though, this analysis also emphasises the extent of variation 275 276 in response across the three treatment regimes, which is smallest for the controls, and largest for the 48 hour-exposed wasps (note the spread along PCA axis 2 in Figure 3). In addition, both cluster 277 278 analyses placed the controls as some way intermediate between the two 5-aza-dC exposed 279 treatments, which we consider further in the next two paragraphs. Both analyses also show that exposure regime dominated the clustering, over and above of any effects of either tissue-type or 280 collection time, suggesting that this is the key effect. When we consider clustering at the level of 281 282 individual CpG loci, clustering is much less obvious, using both t-SNE and PCA. However, the control 283 treatment appears to cluster within the others in the t-SNE analyses, and the 24 hour and 48 hour 284 treatments do separate to some extent, e.g. with the 24h samples more the bottom and left of Figure 285 S1 and the 48h samples more to the top and right (see Figures S1 and S2). The differences between the individual CpG and gene-level cluster analyses are likely to be due to a combination of low 286 287 coverage and a greater stochasticity of methylation at the individual CpG level (see discussion).

288 Second, a more in-depth examination of the data using a generalised linear modelling (GLM) 289 approach to examine differential methylation at the gene level found that 5-aza-dC exposure had a 290 significant effect on the methylation status of the majority of genes examined (exposure treatment was significant as a main effect at an FDR level of 0.05 for 8,556 out of 11,582 genes, or 73.9% of 291 292 genes). Moreover, both tissue-type and collection time also influenced methylation status across a 293 large number of genes (significant as main effects in the GLMs for 3483 and 3134 genes respectively). 294 Significant interactions between factors were also present for the majority of genes (exposure regime 295 * collection time = 10,547 genes, exposure regime * tissue type = 9,042 genes, collection time * tissue 296 type = 5,011 genes, and exposure * collection time * tissue type = 10,234 genes), showing widespread

evidence for significant interplay between these factors. These results emphasise the context-dependent nature of the action of 5-aza-dC. Full results from the GLMs are presented in Table S2.

299 Interestingly, it appears that methylation tends to increase initially in response to exposure to 300 5-aza-dC, an effect that is noticeable after 24h of exposure, before decreasing after 48h of exposure 301 (Figure 4A). This means that how long individuals are exposed to the chemical may well influence what 302 kind of methylation changes are observed; in other words, both hyper- and hypo-methylation can 303 result from 5-aza-dC treatment. In terms of when wasps were collected after exposure, 3134 genes 304 displayed significantly decreased methylation in association with collection time, with a more 305 pronounced decrease in methylation in samples collected 48h post-exposure relative to 24h post-306 exposure (when compared to 0h post-exposure; Figure 4B). This suggests that the effect of 5-aza-dC 307 persists at least 48h after exposure has stopped. Finally, 3483 genes displayed significantly altered 308 methylation patterns in association with tissue type (Figure 4C), with lower methylation overall in the 309 head. However, we re-iterate that these main effects need to be contextualised within the pattern of 310 widespread interactions seen in the data, with 88% of genes exhibiting a significant second-order 311 interaction between exposure, tissue type and collection time.

312 Gene-set enrichment analysis, using *N. vitripennis* GO terms, when genes were ranked by their response to exposure regime is presented in Table S3. Broadly, we identified 129 significantly 313 enriched GO terms, 52 of these terms fell under the "Biological process" (BP) category, 16 under 314 315 "Cellular component" (CC) and 61 under "Molecular function" (MF). The top three enriched GO terms were "sequence-specific DNA binding" (GO:43565), "regulation of transcription, DNA 316 317 templated" (GO:6355) and "transcription factor activity, sequence-specific DNA binding" (GO: 318 43565) (Table S3). When the analysis was repeated using GO terms for the D. melanogaster 319 orthologs, a higher number of significantly enriched GO terms was obtained (686; 526 BP, CC 59, MF 320 89) of significantly enriched GO terms was obtained (Table S4). The top three GO terms for the D. 321 melanogaster analysis were "nucleus" (GO:0005634), "positive regulation of transcription from RNA 322 polymerase II promoter" (GO:45944), and, making a reappearance, "sequence-specific DNA binding" 323 (GO:43565). This analysis combined GO-terms when N. vitripennis genes had multiple D. 324 melanogaster orthologs, which could lead to bias, however, we retrieve very similar enriched terms 325 when we used only the GO terms that were shared between the multiple orthologs (Table S5), 326 suggesting any bias caused by this approach is small.

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329 Discussion

330 A genome-wide analysis of CpGs across more than 11,000 genes, using 51 biological replicates across 331 our treatment combinations, has shown that the demethylating agent 5-aza-dC has very widespread effects on patterns of DNA methylation in the Nasonia vitripennis genome. The methylome has been 332 333 characterised in Nasonia vitripennis by Wang et al (2013) and Pegoraro et al (2016), and in common 334 with other insect species, DNA methylation appears to be primarily associated with gene bodies 335 (including intron-exon boundaries), and a rather low proportion of CpGs are methylated. As 336 mentioned above, previous work that has attempted to confirm the effects of 5-aza-dC on DNA 337 methylation have had mixed results. Pegoraro et al (2016) confirmed the effects of 5-aza-dC across five genes (four via a qPCR MethylQuant assay and one via cloning a fragment of the msn gene from 338 339 treated versus untreated individuals), whilst Ellers et al (2018) have recently failed to show any 340 effects of 5-aza-dC on the methylation status of 155 CpGs across ten genes (nine of which provided 341 libraries; they used high-throughput bisulfite amplicon sequencing: BSAS; see Redshaw et al. 2014 342 for a discussion of different methylation quantification assays). Our larger-scale study has both 343 confirmed the action of 5-aza-dC as a disruptor of methylation in *Nasonia*, but also suggested why it 344 might be so difficult to pin down those effects. In particular, we find that whilst we see statistically 345 significant main effects of exposure regime on DNA methylation (which includes differences in 346 length of exposure), we also see effects of tissue type (albeit at the crude level of heads versus 347 bodies), and collection time. Moreover, we also see many statistically significant interactions 348 between these factors; all told, the action of 5-aza-dC is likely to be context-dependent, both in 349 terms of the sequences more likely to be targeted but also in terms of the experimental details, 350 including target tissues. Nonetheless, our results strongly suggest that 5-aza-dC is not chemically 351 inert in terms of influencing DNA methylation in *Nasonia* – indeed, quite the opposite.

352 That the effects of 5-aza-dC are non-random in terms of the genes it influences are further 353 confirmed by our gene ontology enrichment analyses. We see enrichment for genes associated with 354 DNA binding, regulation of transcription, and transcription factor activity. These processes are of 355 interest given that we might expect the regulation of transcription to be a crucial component of the epigenetic role of DNA methylation, given what we know from both mammals and insects, even 356 357 though the targets (throughout the genome versus exons) and outcomes (transcriptional silencing 358 versus stable gene expression) are typically different in the two taxa. Our analysis certainly confirms the sequence-specific action of 5-aza-dC seen in other species. 359

A further important effect we see is that 5-aza-dC exposure can result in hyper-methylation
 as well as hypo-methylation. This means that the characterisation of 5-aza-dC as a strictly

362 demethylating agent may be somewhat misleading. The complexity of outcomes of 5-aza-dC in 363 mammals, in particular in terms of trying to understand its action for cancer treatments, is becoming 364 increasingly well-appreciated (e.g. Seelan et al 2018). As such, our data confirm in an insect what is 365 being seen in mammalian cells. We are not well placed to speculate on the mechanism(s) by which 366 5-aza-dC may lead to hyper-methylation (see instead Seelan et al. 2018), even though its 367 hypomethylating effect is well characterised chemically (see Introduction). However, if 5-aza-dC 368 disrupts the regulation of the DNA methylation machinery itself, for instance in terms of how the 369 Dnmt genes are expressed and act, then the downstream outcomes could be both hyper- and hypo-370 methylation. Perhaps related to this, in their study of DNA methylation and the control of the 371 diapause response in Nasonia, Pegoraro et al (2016) used both 5-aza-dC and RNA interference 372 (RNAi) to manipulate methylation, in the latter case disrupting DNA methylation by knocking down 373 both Dnmt1a and Dnmt3. Whilst both 5-aza-dC and the RNAi constructs influenced how diapause 374 was influenced by exposure to short or long day length, they did so in different ways. Whilst both 375 manipulations abolished the day-length response, the RNAi of Dnmt1a and Dnmt3 resulted in a 376 change in how wasps responded to long-day length (increasing the diapause response), whilst 5-aza-377 dC led to a change on response to short-day length (decreasing the diapause response). As such, 378 different manipulations of DNA methylation may lead to different phenotypic outcomes, depending on the regulatory causes and consequences of the methylation pattern itself. More generally then, it 379 380 is clear that there is much that remains to be elucidated in how 5-aza-dC shapes methylation 381 patterns, in both insects and mammals (Seelan et al. 2018).

382 On a more technical note, perhaps unsurprisingly we found more significant genes in the GO 383 enrichment analysis using the GO terms for Drosophila melanogaster orthologs. This is perhaps 384 because of the more detailed gene ontology information available for the *D. melanogaster* genome. 385 Nonetheless, the two analyses provide a similar story in terms of GO enrichment. In addition, the 386 majority of our analyses were undertaken at the level of the gene. Perhaps given the variation in 387 coverage at the level of individual CpGs, the clustering analyses were not by any means as clear cut 388 as the gene-level results. However, in the t-SNE analysis (Figure S1), the control replicates cluster together, albeit within the broader distribution of the other two exposure treatment groups. The 389 390 effect of genomic and sequence context may be particularly relevant here: our analysis at the level 391 of the gene (see Methods for details) brings together CpGs that share a lot of context, by virtue of 392 being associated with a given genetic locus. The differences in the clustering at the individual CpG 393 versus gene levels may therefore represent the fact that the action of 5-aza-dC does appear to be 394 non-random, and that we gain clarity when we consider closely associated CpGs whose patterns of

methylation, and responses to 5-aza-dC, are more similar than randomly chosen CpGs across thegenome.

397 In conclusion, our results suggest that 5-aza-dC remains a potential tool for evolutionary 398 ecologists wishing to explore how DNA methylation influences phenotypic expression. As with all 399 genomic approaches however, no one approach gives all the answers. 5-aza-dC may prove a very 400 useful, if rather blunt, tool for first assessing the range of phenotypes that are influenced by 401 methylation status, whilst no doubt more targeted knock-downs and genetic manipulations will be 402 needed to link phenotypes to causal CpGs.

403

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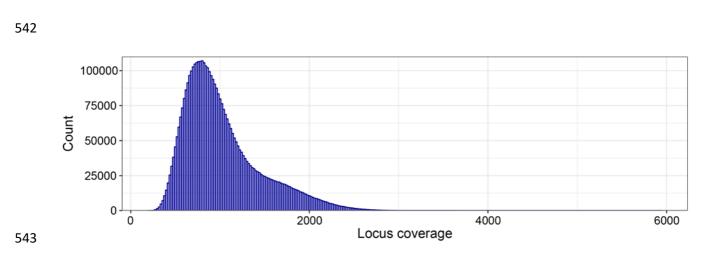
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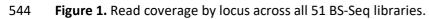
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536 Data and code availability

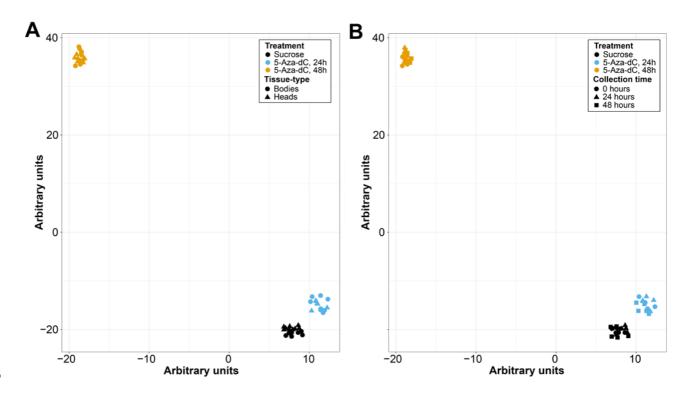
- 537 Raw reads will be deposited in the SRA and accession codes will be made available upon publication.
- 538 Scripts for the analyses in this paper are available at: https://github.com/DarrenJParker/5-Aza-
- 539 dC_methylation

541 Tables and Figures





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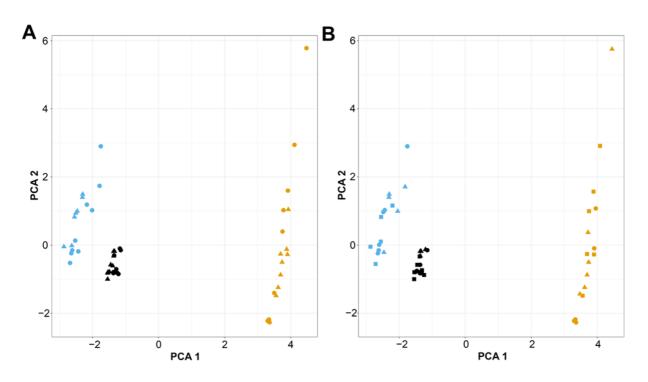


548 regime and tissue type and (B) Exposure regime and harvest time.

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Figure 3. PCA of the proportion of methylated reads per gene, showing the effects of (A) Exposure regime and tissue type and (B) Exposure regime and harvest time.

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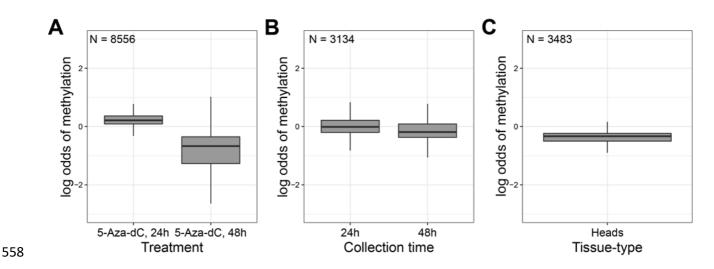


Figure 4. A significant effect on gene methylation status (displayed as log odds of methylation) was
 noted in response to (A) 5-aza-dC exposure regime, (B) harvest time after exposure and (C) tissue type. N= number of genes where significantly differentially methylated CpGs in association with each
 factor.

564 **Table 1.** Experimental design. Female wasps were exposed to 1 of 3 5-aza-dC exposure regimes

565 including a control. Females were then harvested at 1 of 3 timepoints after the exposure period

heads were excised from bodies to give a degree of tissue-specificity to the analysis. Three biological

replicates were available for almost all treatment combinations with each biological replicate

568 consisting of tissue from three *N. vitripennis* females.

Treatment	5-aza-dC exposure regime	Harvest	Tissue type	No. replicates
Combination		time		(1 x rep. = 1 pool of 10)
1	20% sucrose for 24 h	0h	Bodies	3
2	20% sucrose for 24 h	0h	Heads	3
3	20% sucrose for 24 h	24h	Bodies	3
4	20% sucrose for 24 h	24h	Heads	3
5	20% sucrose for 24 h	48h	Bodies	3
6	20% sucrose for 24 h	48h	Heads	3
7	10µM 5-aza-dC for 24 h	0h	Bodies	3
8	10µM 5-aza-dC for 24 h	0h	Heads	3
9	10µM 5-aza-dC for 24 h	24h	Bodies	3
10	10µM 5-aza-dC for 24 h	24h	Heads	2
11	10µM 5-aza-dC for 24 h	48h	Bodies	3
12	10µM 5-aza-dC for 24 h	48h	Heads	2
13	10µM 5-aza-dC for 48 h	0h	Bodies	3
14	10µM 5-aza-dC for 48 h	0h	Heads	2
15	10µM 5-aza-dC for 48 h	24h	Bodies	3
16	10µM 5-aza-dC for 48 h	24h	Heads	3
17	10µM 5-aza-dC for 48 h	48h	Bodies	3
18	10µM 5-aza-dC for 48 h	48h	Heads	3

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571 Supplementary Figures

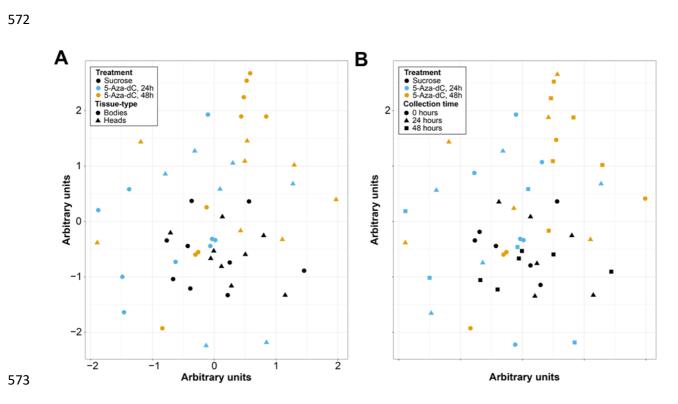
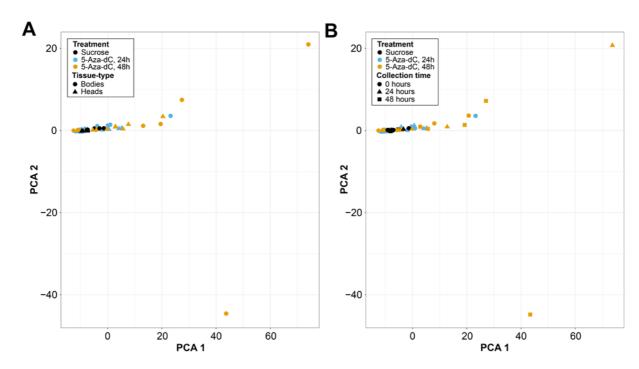
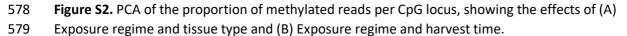


Figure S1. t-SNE of the proportion of methylated reads per CpG locus, showing the effects of (A)
 Exposure regime and tissue type and (B) Exposure regime and harvest time.







580

581 Supplemental Table Information

- 582
- 583 **Table S1.** Number of reads and coverage of each sample
- 584 **Table S2.** Output of coefficients and significance values from the GLM analysis
- 585 **Table S3.** Enriched GO terms for treatment using *Nasonia vitripennis* GO terms
- Table S4. Enriched GO terms for treatment using *D. melanogaster* GO terms, when GO-terms from
 multiple orthologs are combined together
- 588 **Table S5.** Enriched GO terms for treatment using *D. melanogaster* GO terms, when only GO-terms
- 589 which are the same are kept (when a *Nasonia vitripennis* gene has multiple *D. melanogaster*
- 590 orthologs)