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1 **Widespread conservation and lineage-specific diversification of genome-wide DNA**
2 **methylation patterns across arthropods**

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21 **Abstract**

22 Cytosine methylation is an ancient epigenetic modification yet its function and extent within genomes
23 is highly variable across eukaryotes. In mammals, methylation controls transposable elements and
24 regulates the promoters of genes. In insects, DNA methylation is generally restricted to a small subset
25 of transcribed genes, with both intergenic regions and transposable elements (TEs) depleted of
26 methylation. The evolutionary origin and the function of these methylation patterns are poorly
27 understood. Here we characterise the evolution of DNA methylation across the arthropod phylum.
28 While the common ancestor of the arthropods had low levels of TE methylation and did not methylate

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1 promoters, both of these functions have evolved independently in centipedes and mealybugs. In
2 contrast, methylation of the exons of a subset of transcribed genes is ancestral and widely conserved
3 across the phylum, but has been lost in specific lineages. Remarkably the same set of genes are
4 likely to be methylated in all species that retained exon-enriched methylation. We show that these
5 genes have characteristic patterns of expression correlating to broad transcription initiation sites and
6 well-positioned nucleosomes, providing new insights into potential mechanisms driving methylation
7 patterns over hundreds of millions of years.

8 **Author Summary**

9 Animals develop from a single cell to form a complex organism with many specialised cells.
10 Almost all of the fantastic variety of cells must have the same sequence of DNA, and yet
11 they have distinct identities that are preserved even when they divide. This remarkable
12 process is achieved by turning different sets of genes on or off in different types of cell using
13 molecular mechanisms known as "epigenetic gene regulation".

14 Surprisingly, though all animals need epigenetic gene regulation, there is a huge diversity in
15 the mechanisms that they use. Characterising and explaining this diversity is crucial in
16 understanding the functions of epigenetic pathways, many of which have key roles in human
17 disease. We studied how one particular type of epigenetic regulation, known as DNA
18 methylation, has evolved within arthropods. Arthropods are an extraordinarily diverse group
19 of animals ranging from horseshoe crabs to fruit flies. We discovered that the levels of DNA
20 methylation and where it is found within the genome changes rapidly throughout arthropod
21 evolution. Nevertheless, there are some features of DNA methylation that seem to be the
22 same across most arthropods- in particular we found that there is a tendency for a similar
23 set of genes to acquire methylation of DNA in most arthropods, and that this is conserved
24 over 350 million years. We discovered that these genes have distinct features that might
25 explain how methylation gets targeted. Our work provides important new insights into the
26 evolution of DNA methylation and gives some new hints to its essential functions.

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1 **Introduction**

2 In most organisms DNA bases are adorned with a variety of chemical modifications. Amongst the
3 most common of these is methylation at the 5 position of cytosine (C5me), which is present from
4 bacteria to humans (Ponger and Li, 2005; Casadesús and Low, 2006; Jurkowski and Jeltsch, 2011).
5 In eukaryotes, a key property of cytosine DNA methylation is its ability to act epigenetically — that is,
6 once introduced, methylation at specific cytosines can remain in place through cell division (Holliday
7 *et al.*, 1987; Holliday, 2006). This relies on the activity of "maintenance" methyltransferases, DNMT1
8 in animals (Law and Jacobsen, 2010), which recognise CG dinucleotides (CpG sites) where one
9 strand is methylated and one strand unmethylated and catalyse the introduction of methylation on the
10 unmethylated strand (Jeltsch, 2006). Meanwhile "de novo" methyltransferases act on unmethylated
11 DNA. In animals this role is performed by DNMT3 enzymes, which introduce 5meC predominantly
12 within CpG sites (Jeltsch, 2006). Mechanisms also exist to remove methylation from DNA, including
13 the TET family of enzymes, which convert 5meC to a hydroxymethylated intermediate which can be
14 removed by base excision repair or diluted out through cell division (Nashun, Hill and Hajkova, 2015).
15 As the maintenance and de novo methylation of CG sequences occurs through the activity of
16 homologous enzymes in plants and animals (Ponger and Li, 2005), CpG methylation was likely
17 present among the earliest eukaryotic organisms.

18 In mammals, a key function of CpG methylation is to defend the genome against transposable
19 elements (TEs) by preventing their transcription and transposition (Bird, 2002), and loss of DNA
20 methylation leads to reactivation of TEs (Walsh, Chaillet and Bestor, 1998). CpG methylation targeted
21 to the promoters of genes can also suppress transcription, typically resulting in stable silencing (Bird,
22 2002). Another notable genome-wide pattern is the enrichment of CpG methylation within the exons
23 of transcribed genes (Suzuki and Bird, 2008). In contrast to TE and promoter methylation, this is not
24 associated with transcriptional silencing.

25 Whilst CpG methylation at both TEs and gene bodies is present in both plants and animals (Law and
26 Jacobsen, 2010), across eukaryotic species both DNA methylation levels and the targets of
27 methylation have evolved rapidly (Feng *et al.*, 2010; Zemach *et al.*, 2010). Most strikingly, CpG
28 methylation has been independently lost altogether several times, coinciding with the loss of DNMT1
29 and DNMT3 DNA methyltransferase enzymes (Ponger and Li, 2005; Feng *et al.*, 2010; Zemach *et al.*,

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1 2010). Across eukaryotes, loss of CpG methylation tends to be accompanied by loss of the DNA
2 alkylation repair enzyme ALKB2, which repairs damage caused by DNMTs introducing 3-
3 methylcytosine into DNA. This suggests that some species correct DNA alkylation using ALKB2, and
4 others avoid it altogether by losing the DNA methylation pathway (Rošić *et al.*, 2018). Even within
5 species that retain DNA methyltransferases, the genomic distribution of CpG methylation differs
6 widely (Feng *et al.*, 2010; Zemach *et al.*, 2010; Bewick *et al.*, 2017, 2019; Rošić *et al.*, 2018; de
7 Mendoza *et al.*, 2019; de Mendoza, Pflueger and Lister, 2019). Such variability is surprising given the
8 essential role of CpG methylation in genome regulation in mammals and plants, and there are few
9 clues as to what factors drive the changes. Tracing the evolution of CpG methylation is currently
10 challenging because detailed descriptions of DNA methylation patterns are patchy and focussed on
11 model systems, leaving large parts of the phylogenetic tree underexplored.

12 Here we study CpG methylation patterns across arthropods. Arthropods have been suggested to
13 represent a very different system of CpG methylation from mammals (Keller, Han and Yi, 2016).
14 Whilst the well-characterised model organism *Drosophila melanogaster* lacks DNA methylation
15 altogether, DNA methyltransferases 1 and 3 were found in the honey bee *Apis mellifera* (Wang *et al.*,
16 2006). Genome-wide CpG methylation mapping demonstrated that methylation was globally
17 extremely low, and restricted to the bodies of a subset of transcribed genes (Lyko *et al.*, 2010;
18 Zemach *et al.*, 2010). Subsequently, similarly restricted patterns of DNA methylation were found in
19 other insects (Lyko *et al.*, 2010; Xiang *et al.*, 2010; Bonasio *et al.*, 2012; Wang *et al.*, 2013). Such
20 patterns support the proposal that gene body methylation is conserved across eukaryotes while TE
21 methylation has been lost altogether in arthropods (Zemach *et al.*, 2010; Keller, Han and Yi, 2016).
22 However some insects show considerably higher levels of genome-wide methylation (Bewick *et al.*,
23 2017), and variation in arthropod methylation levels also exists outside of insects (Kao *et al.*, 2016;
24 Kvist *et al.*, 2018; de Mendoza, Pflueger and Lister, 2019; Liu *et al.*, 2019). There is also evidence of
25 TE methylation in the desert locust *Schistocerca gregaria* (Lyko *et al.*, 2010). A thorough
26 reconstruction of the evolution of methylation across the phylum is still lacking.

27 We set out to explore the evolution of arthropod methylation patterns by characterising genome-wide
28 CpG methylation across the phylum. We show that TE methylation was ancestral to arthropods,
29 although at a relatively low level. Methylation of protein-coding genes was also ancestral to
30 arthropods, with similar subsets of genes being targeted for methylation across arthropods. Despite

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1 these conserved features, we find many examples of diversification in methylation patterns across
2 arthropods, in particular loss of gene methylation in crustaceans and gain of both promoter
3 methylation and genome-wide TE methylation in the myriapod *Strigamia maritima* and the insect
4 *Planococcus citri*. We find that methylation at genes, enriched within exons, is the most widely
5 conserved feature of arthropod methylomes and we use comparative analysis to identify a link
6 between exon methylation and nucleosome positioning. Overall, our findings demonstrate that while
7 key features of global methylation patterns have been conserved across millions of years of arthropod
8 evolution, the targets of DNA methylation can rapidly diverge within individual lineages.
9

10 **Results**

11 **Genome-wide levels of CpG methylation vary widely across the arthropods**

12 We carried out high-coverage whole-genome bisulphite sequencing (WGBS) on 14 species of
13 arthropod and quantified the levels of DNA methylation with base-pair resolution. To examine
14 genome-wide methylation levels we combined this data with published results from 15 additional
15 species (Bewick *et al.*, 2017; Wu *et al.*, 2017; Kvist *et al.*, 2018) which were mostly sequenced at
16 lower coverage. Estimates of genome-wide CpG methylation were then used to reconstruct ancestral
17 methylation levels across the arthropod phylogeny. All 18 species of holometabolous insects had low
18 levels of CpG methylation, and this was likely the ancestral state of this clade (Figure 1A and 1B).
19 While CpG methylation rates in other arthropod clades tended to be higher, they varied considerably
20 (Figure 1A and 1B). The ancestral arthropod likely had moderate methylation levels (8.59±4.8%;
21 Figure 1A) but higher methylation levels evolved in *S. maritima*. Similarly, the ancestor of insects had
22 methylation levels lower than some taxa such as *B. germanica* (3.9±3.3% versus 12%) indicating that
23 methylation level fluctuated throughout arthropod evolution.

24 To investigate the evolution of the DNA methylation machinery across arthropods, we searched the
25 genomes of these species for homologues of the genes encoding the methyltransferases DNMT1-3.
26 We confirmed the genes all encoded a full cytosine methyltransferase domain, and where we did not
27 find annotated homologues we directly search the genomic DNA for unannotated genes. In each
28 species we found a single copy of DNMT2, which methylates tRNAs (Goll *et al.*, 2006) (Figure 1C).
29 DNMT1 was present in all species apart from the five Diptera (Figure 1C). The loss of this gene was

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1 associated with the loss of CpG methylation (Figure 1c), with methylation rates in *D. melanogaster* not
2 significantly different from the unmethylated DNA spike-in included in each reaction. DNMT3 was
3 absent from the genomes of 14 species, with inspection of the tree suggesting at least eight
4 independent losses (Figure 1C). Several of these species possessed moderate levels of CpG
5 methylation (Figure 1B), indicating that DNMT1 alone can be sufficient for introducing genome-wide
6 DNA methylation, consistent with earlier studies in arthropods and nematodes (Xiang *et al*., 2010;
7 Bewick *et al*., 2017; Rošić *et al*., 2018).

8 Across the eukaryotes ALKB2, which repairs DNA alkylation damage introduced by DNMTs, tends to
9 be lost from the same taxa as DNMT1 and 3 (Rošić *et al*., 2018). Arthropods exhibited many
10 exceptions to this general rule—there have been at least five losses of ALKB2 but only one of these is
11 associated with the loss of DNMT1 and 3 (Figure 1C). However, we found that species with ALKB2
12 possessed higher methylation levels (Figure 1 Supplement; phylogenetic mixed model: p=0.0182),
13 suggesting ALKB2 is dispensable in species with low levels of DNA methylation.

14 **Rapid loss and gain of TE methylation across arthropods**

15 In mammals, plants and nematodes, transposable elements (TEs) are preeminent targets of DNA
16 methylation, but previous studies have shown that TE methylation is rare in holometabolous insects
17 (Feng *et al*., 2010; Lyko *et al*., 2010; Zemach *et al*., 2010; Bonasio *et al*., 2012; Wang *et al*., 2013).
18 However, DNA methylation has been found at TEs in some arthropods (Falckenhayn *et al*., 2013; Kao
19 *et al*., 2016; de Mendoza, Pflueger and Lister, 2019; Liu *et al*., 2019). To explore the distribution of TE
20 methylation across arthropods we annotated transposable elements using RepeatMasker analysis of
21 the entire genome, and removed annotations that did not contain Pfam domains derived from
22 transposable elements. We focused on 14 species that represent the diversity of arthropods, and
23 have assembled and annotated genomes (see Fig 2B).

24 Compared to unannotated regions of the genome, TEs were strongly enriched for methylation in *S.*
25 *maritima* and *P. citri,* and weakly enriched in several other species (Figure 2B,C). This pattern is
26 reflected in the distribution of methylation across TEs — this is skewed towards 0% for most species,
27 but in *S. maritima* and *P. citri* the large majority of TEs are methylated (Figure 2A,B; Figure 2
28 Supplement). In these two species there was a sharp drop in methylation rates at the boundary of the
29 TE (Figure 2D). In agreement with earlier studies (Lyko *et al*., 2010; Bonasio *et al*., 2012), the

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1 methylation rate of TEs was low in holometabolous insects. However, outside of this group there was
2 moderate methylation of TEs in chelicerates (*L. polyphemus* and *P. tepidariorum*), the crustacean *P.*
3 *hawaiensis* and hemimetabolous insects (*B. germanica* and *A. pisum*) (Figure 1A,C). To further
4 quantify the extent of TE methylation, we clustered TEs into highly- and lowly-methylated groups in
5 each species separately, and calculated the proportion of TEs that were assigned to the highly-
6 methylated group (Table 1). The large majority of TEs were targeted by methylation in *S. maritima*
7 and *P. citri,* while in all other species under 15% of TEs were methylated. Ancestral state
8 reconstruction suggested that a low level of TE methylation was present in the ancestral arthropod,
9 but was lost in the ancestor of holometabolous insects (Figure 2A).

10

11 **Table 1. Proportion of Genes and TEs that are highly methylated**

	TEs[a]		Genes	
Species	**Number**	**Proportion methylated[b]**	**Number**	**Proportion methylated[b]**
Acyrthosiphon pisum	293	0.017	13147	0.171
Apis mellifera	7	0.143	10066	0.272
Armadillidium vulgare	655	0.020	4703	0.019
Blattella germanica	276	0.145	9272	0.387
Bombus terrestris	78	0.128	8550	0.069
Heliconius melpomene	34	0.088	11583	0.077
Ixodes scapularis	212	0.033	5775	0.219
Limulus polyphemus	342	0.117	7227	0.265
Nicrophorus vespilloides	9	0.111	12305	0.032
Parasteatoda tepidariorum	622	0.032	9742	0.243
Parhyale hawaiensis	89	0.079	3302	0.028
Planococcus citri	361	0.751	34044	0.099
Strigamia maritima	719	0.758	12898	0.326

12 [a] TEs with annotated TE-associated domains (see Methods); [b] the proportion falling into the highly
13 methylated group after clustering each feature type within each species

14

15 **Methylation at exons is conserved across most arthropods**

16 We next investigated methylation at genes across arthropods. In all but one of the species we tested,
17 mean methylation levels across exons were significantly higher than unannotated regions of the
18 genome (Figure 3B). The exception was *P. hawaiensis,* where exons are significantly less methylated
19 than unannotated regions of the genome (Figure 3B). There is a significant difference between
20 methylation at exons and introns in *P. hawaiensis* (p=0.001, paired t test). In the species with exon

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1 methylation, the distribution of methylation suggested that a subset of genes is targeted for
2 methylation (Figure 2C). When clustered into highly and lowly methylated genes, the proportion of
3 methylated genes varied similarly to mean methylation across genes (Table 1).

4 To investigate the distribution of methylation within genes, we compared the methylation levels at
5 exons and introns in each species. Methylation was higher at exons in the majority of species,
6 suggesting that the gene body methylation in arthropods is due to targeting of methylation to exons.
7 However, there was little difference between exons and introns for the two crustaceans, *P.*
8 *hawaiensis* and *A. vulgare* (Figure 3C; supplemental Figure S3). Given that *P. hawaiensis* exons are
9 depleted for methylation relative to the genome-wide background while *A. vulgare* exons are only
10 slightly greater than the background, this may reflect an ancient loss of gene body methylation in the
11 ancestor of these species. Among species with exon methylation, there were differences in how
12 methylation levels changed across the gene (Figure 3C). For example, methylation was largely
13 confined to the first three exons of *P. citri* and *N. vespilloides,* while methylation in *B. germanica* is
14 largely found from exon four onwards (Figure 3C). Together these data suggest that exon-enriched
15 methylation was an ancestral property of arthropod methylomes which is largely conserved across the
16 phylum.

17 **Independent acquisition of promoter methylation in arthropod lineages**

18 In mammals, methylation of regions immediately upstream of genes, often at CpG islands, is
19 associated with gene silencing. However, there is no evidence of promoter methylation in insects
20 (Lyko *et al*., 2010; Xiang *et al*., 2010; Bonasio *et al*., 2012). To examine promoter methylation
21 associated with gene silencing across arthropods, we extracted 1kb upstream of genes for all
22 species. In most species there was little difference in upstream methylation between high and low
23 expression genes; however, low expression genes in *P. citri* and *S. maritima* had significantly higher
24 upstream methylation than high expression genes (Figure 4A). In *S. maritima* only genes with very
25 high upstream methylation showed clearly reduced gene expression (p=1e-15, Kruskal Wallis test),
26 whilst in *P. citri* there was a positive correlation between upstream methylation and gene expression
27 across a wider range of upstream methylation levels (Figure 4B). The different relationship between
28 upstream methylation and gene expression between *S. maritima* and *P. citri* and the lack of a similar

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1 relationship in other arthropod species suggests that promoter methylation associated with gene
2 silencing may have evolved independently in these two species.

3 **Methylated genes are conserved and have moderate to high expression**

4 Our results suggest that the most highly conserved feature of arthropod methylomes is enrichment of
5 methylation at the exons of a subset of genes. Across species, we asked whether there was any
6 tendency for orthologous genes to be methylated in different species. We ranked orthologous genes
7 by relative methylation levels across species and observed that there was a clear tendency for
8 orthologs to have similar levels of methylation in different species (Figure 5A). The observation that
9 the same genes are methylated in different species raised the question of what determines which
10 genes acquire methylation. We used comparative analysis to investigate this across the phylum.

11 Methylation has been shown to be enriched at alternatively spliced genes in some insects (Lyko *et al.*,
12 2010; Bonasio *et al.*, 2012). To test for a link between methylation and splicing across arthropods, we
13 compared the level of methylation between genes with one exon (which cannot undergo splicing) and
14 genes with two or more exons (which may undergo splicing). We found no clear difference in any
15 species (Figure S5), suggesting that splicing does not explain the propensity of genes to acquire
16 methylation across arthropods.

17 Previously, methylation of genes in individual insect species has been correlated to higher levels of
18 expression (Xiang *et al.*, 2010; Bonasio *et al.*, 2012). We find a statistically significant tendency for
19 genes with high methylation to have higher expression across most species (Table S2). However,
20 many highly expressed genes are not methylated. Instead a more prominent trend is for methylated
21 genes to have more focussed levels of gene expression such that genes with very low expression
22 levels are rarely methylated (Figure 5B,C; Figure 5 supplement 2). Curiously, this pattern is reversed
23 in *P. citri,* where the exons of methylated genes tend to have low expression (Figure 5 supplement 2).

24 Previously it has been noted that methylated genes are more likely to perform conserved
25 “housekeeping” functions (Hunt *et al.*, 2013). We clustered genes into orthologous groups across
26 species and examined genes that were conserved across all species compared to species-specific
27 genes. Across all species carrying gene body methylation, conserved genes with moderate to high
28 expression were more likely to be methylated (Figure 5C; Figure S5). Nevertheless many conserved

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1 and highly expressed genes lacked methylation suggesting that neither conservation nor expression
2 is sufficient to explain gene body methylation.

3 **Nucleosome positioning influences DNA methylation levels across arthropods**

4 In order to investigate molecular mechanisms that might be responsible for influencing DNA
5 methylation we examined how the correlation in methylation between pairs of CpGs varied with
6 increasing separation. In many species with exon-enriched methylation the correlation coefficient
7 between methylation levels of individual CpGs oscillated periodically (Figure 6A,B). Fourier analysis
8 showed that the period of oscillation was ~160 nucleotides, roughly corresponding to the average
9 nucleosome repeat length (Figure 6A,B; Figure S6-1). We quantified this nucleosome-length
10 periodicity within exons across all species. While the majority of species with exonic methylation
11 displayed a nucleosome periodicity signal, its magnitude varied greatly – for example *H. melpomene*
12 has gene methylation but less apparent periodicity (Figure 6B). Interestingly a clear signal of
13 periodicity was also seen for TE methylation in *S. maritima* and *P. citri*, both of which have high levels
14 of TE methylation (Figure S6-1).

15 We wondered whether the periodicity in correlation between methylated DNA might reflect an
16 influence of nucleosome positioning on DNA methylation, as has been shown in plants (Chodavarapu
17 *et al.*, 2010) and inferred from analysis of mammalian DNA methylation profiles. In the absence of
18 genome-wide nucleosome positioning data for the majority of species, we investigated nucleosome
19 positioning from *Drosophila* (Ho *et al.*, 2014), examining orthologues of genes either enriched or
20 depleted for DNA methylation across arthropods. The promoters of methylated genes possessed high
21 nucleosome occupancy overall and strongly positioned nucleosomes just upstream (-1) and
22 downstream (+1) of the transcription start site (TSS) (Figure 6C). The promoters of unmethylated
23 genes showed lower nucleosome occupancy overall and demonstrated weaker positioning of the -1
24 and +1 nucleosome. Previous analyses of promoter types across eukaryotes have indicated that
25 promoters with strong positioning of nucleosomes lead to initiation of transcription across a broad
26 region (broad TSS) whilst promoters with weaker nucleosome positioning tend to have a much
27 narrower TSS focussed around a dominant initiation site(Haberle and Lenhard, 2016). Using cap
28 analysis of gene expression (CAGE) data from *D. melanogaster* we found that the TSS of *D.*

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melanogaster orthologs of methylated genes was broader than the TSS of orthologs unmethylated genes (Figure 6C).

Further evidence for a connection between nucleosome occupancy and a periodic signal in the correlation between methylation sites comes from a comparison of exons and introns. Exons are known to have much higher nucleosome occupancy than introns and accordingly the periodic signal of methylation correlation is markedly weaker in introns than in exons (Figure S6-2). Together this supports a potential role for nucleosome occupancy in shaping CpG methylation patterns in arthropods.

The alternative patterns of nucleosome occupancy and transcription initiation corresponded to previous analyses across organisms demonstrating that housekeeping genes tend to have well positioned nucleosomes just downstream of promoters and broad TSS whereas tissue-specific genes tend to have less well-defined nucleosome positions at promoters and narrow TSS (Carninci *et al.*, 2006; Hoskins *et al.*, 2011; Lenhard, Sandelin and Carninci, 2012; Haberle *et al.*, 2014). We therefore tested whether methylated genes were more likely to have tissue-specific or global gene expression using RNAseq data from different tissue types. In every species with gene body methylation, we found that methylated genes tended to have less variable expression across different tissues (Figure 6D). Altogether this suggests that across arthropods conserved genes with strongly positioned nucleosomes, broad TSS and housekeeping functions are targeted for methylation whilst tissue-specific genes with opposite patterns of nucleosome occupancy and TSS width tend to be depleted of methylation.

Discussion

Molecular pathways involved in epigenetic gene regulation evolve surprisingly rapidly and DNA methylation is no exception. Our work adds to the complex picture of how DNA methylation patterns change across evolutionary time and offers new insight into potential factors influencing the distribution of DNA methylation within genomes.

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1 **Plasticity of DNA methylation landscapes**

2 Prior to this study, DNA methylation had been characterised across insects (Bewick *et al*., 2017) but
3 only isolated species from more basal arthropod clades had been studied (Falckenhayn *et al.*, 2013;
4 Kao *et al*., 2016; Kvist *et al*., 2018; de Mendoza, Pflueger and Lister, 2019; Liu *et al*., 2019). By
5 examining a phylogenetically broad range of arthropod methylomes we reconstructed the trajectory of
6 DNA methylation patterns across the phylum. Our data show that ancestral arthropods likely had
7 moderate genome-wide methylation including methylation of a small number of transposable
8 elements. Methylation of genes was also prominent and was enriched in exons over introns; however,
9 the magnitude of the difference between exonic and intronic methylation was not as striking as in
10 insects such as *A. mellifera* reflecting the presence of a higher background genomic methylation.
11 Crucially our data also show that changes in methylation patterns can evolve rapidly within individual
12 lineages. Most strikingly, we find strong enrichment of TE methylation evolved independently in the
13 centipede *S. maritima* and the mealybug *P. citri*, which very likely occurred independently. This
14 enrichment does not correlate to any obvious change in genome structure such as increased TE
15 proportion or genome size, however it is interesting that a recent paper reported acquisition of a
16 relatively recent TE family in *S. maritima* that acquires high levels of methylation (de Mendoza,
17 Pflueger and Lister, 2019), which may underpin gain of TE methylation in that species.

18 It is intriguing that the two species with high TE methylation had independently acquired methylation
19 of promoters of silent genes, whilst the exons of these genes are devoid of methylation. Gene
20 regulation by promoter methylation is also found in mammals and was likely acquired independently
21 in the sponge *Amphimedon queenslandica* (de Mendoza *et al*., 2019). In all these cases TE
22 methylation is also prominent so it is possible that the two are linked, perhaps relating to a
23 requirement to control TE-derived promoter regions; however testing this hypothesis would require
24 experimental manipulation of methylation in *P. citri* or *S. maritima* which is currently not possible.

25 It is curious that repeated acquisition of similar types of DNA methylation occurs across phylogenies.
26 This may indicate that targeting of DNA methylation to new regions can be achieved with very few
27 genetic changes. In vertebrates, a possible example is the KRAB-Zinc finger proteins which can
28 recruit DNA methylation to TEs through sequence-specific binding (Quenneville *et al*., 2012). Further

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1 work to identify potential “pioneer” factors that recruit DNMTs to specific regions and underlie the
2 divergence of methylation patterns between species will be of great interest.

3 **Potential factors influencing methylation of genes**

4 Our study confirms earlier speculation that the most widely conserved feature of arthropod
5 methylomes is methylation of genes, biased towards exon methylation (Keller, Han and Yi, 2016).
6 Additionally, we confirm insights from insects that broadly expressed, housekeeping genes are more
7 likely to be targeted for methylation than tissue-specific genes (Hunt *et al.*, 2013). This is strikingly
8 similar to observations in plants and other animal groups, suggesting an ancient evolutionary origin
9 (Bewick and Schmitz, 2017; Zilberman, 2017). Exactly what the function of this modification is
10 remains to be elucidated. It is clearly dispensable under some circumstances as, in addition to the
11 complete loss of DNA methylation in *Drosophila*, we found that DNA methylation at genes has been
12 lost in both the crustaceans we examined, suggesting that even in species where DNA methylation is
13 present in the genome, enrichment of DNA methylation at exons is not essential for viability.

14 Whilst we cannot decipher the function of exon-enriched DNA methylation, our analyses potentially
15 offer new insights into the molecular mechanisms whereby DNA methylation might be deposited. We
16 identify a remarkable methylation pattern across many arthropods such that methylation levels vary
17 periodically with the nucleosome-repeat length. This striking genome-wide pattern that we observe in
18 some species, in particular *S. maritima*, has not been observed to our knowledge in any species
19 previously. However, there are specific regions within the human genome that display apparently
20 nucleosome length periodicity in the correlation between adjacent sites (Zhang *et al.*, 2017);
21 furthermore the influence of nucleosomes on methylation by DNMT3B was observed in human and
22 yeast cells(Baubec *et al.*, 2015; Morselli *et al.*, 2015). Moreover, DNA methylation levels show a 10bp
23 periodicity in *Arabidopsis*, corresponding to methylation targeting nucleotides on the same face of the
24 nucleosome (Chodavarapu *et al.*, 2010). Together these observations reflect a positive correlation
25 between nucleosome occupancy and DNA methylation in *Arabidopsis* and mammals (Chodavarapu *et*
26 *al.*, 2010). Exons are known to have better positioned nucleosomes than introns (Schwartz, Meshorer
27 and Ast, 2009; Tilgner *et al.*, 2009) which might explain why exons are enriched in methylation across
28 species. We also find that promoters of genes with high levels of methylation tend to carry a clear
29 nucleosome positioning pattern, typical of housekeeping genes, where nucleosome occupancy is high

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1 upstream and just downstream of the TSS with a nucleosome-free region between the two (Lenhard,
2 Sandelin and Carninci, 2012; Haberle *et al*., 2014). Both nucleosome positioning and DNA
3 methylation could be linked to transcription. Since tissue-specific genes are highly expressed in only a
4 few cell types, this might explain why they do not appear methylated in whole animal bisulphite
5 sequencing. This would also explain why across all species genes with very low expression are
6 depleted of methylation (Figure 4D). Alternatively, nucleosomes themselves could dictate where DNA
7 methylation takes place. Supporting this point there is little periodicity in DNA methylation in introns
8 compared to exons (Figure S6-2), suggesting that transcription itself is insufficient to account for this
9 effect.

10 Importantly, the fact that we see these patterns based on nucleosome positioning in *Drosophila* where
11 DNA is not methylated suggests that nucleosome positioning may cause differences in DNA
12 methylation. Thus, we suggest that nucleosome positioning may be a primary determinant of variation
13 in DNA methylation across arthropod genomes. Our analyses may therefore prompt a search for how
14 nucleosome occupancy might determine methylation patterns across eukaryotes.

15

16

17 **Methods**

18 **DNMT identification**

19 To identify species that have retained or lost the DNA methylation pathway, we searched for
20 homologues of DNMT. For each species, we used DIAMOND (Buchfink, Xie and Huson, 2015) to
21 perform BLASTp searches against all annotated proteins, with *A. mellifera* DNMT1 (NM001171051),
22 DNMT2 (XM006562945) and DNMT3 (NM001190421) as query sequences. We used InterProScan to
23 screen out hits that lacked the C-5 cytosine-specific DNA methylase domain, and NCBI BLASTP to
24 screen out bacterial contaminants (i.e. hits that were more similar to bacterial DNMTs than eukaryotic
25 DNMTs). To classify DNMTs into subclades (DNMT1, 2 & 3) we aligned all homologues with MAFFT,
26 screened out badly-aligned regions with Gblocks (Castresana, 2000), and inferred a neighbour-joining
27 phylogenetic tree under the Jukes-Cantor model using Geneious v10.1.3 (https://www.geneious.com).

28

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1 **Genome annotation**

2 To annotate exons in each genome we used existing annotations, excluding genes that were split
3 across multiple contigs. To annotate regions which may contain promoters or enhancers, we took
4 1,000 bases upstream of each gene, excluding genes where this exceeded the contig start or end
5 point. We annotated introns based on the position of exons, excluding genes that were split across
6 multiple contigs (using intron_finder.py script available at
7 https://github.com/SamuelHLewis/BStoolkit/). To annotate TEs, we used RepeatModeller v1.0.8 to
8 generate a model of TEs for each genome separately, and then RepeatMasker v4.0.6 to annotate
9 TEs based on the model for that genome. Within each TE, we used interproscan (Jones *et al.*, 2014)
10 to search for the following TE-associated domains: PF03184, PF02914, PF13358, PF03732,
11 PF00665 & PF00077.

12 To annotate rRNA, we either used existing annotations or RNAmmer v1.2 (Lagesen *et al.*, 2007). To
13 annotate tRNA, we either used existing annotations or tRNAscan-SE v1.3.1 (Lowe and Eddy, 1997).
14 To avoid ambiguous results caused by overlapping features, we screened out any TE annotations
15 that overlapped any rRNA, tRNA or exon, and any upstream regions which overlapped any TE, rRNA,
16 tRNA or exon.

17 **Whole genome bisulphite sequencing**

18 To measure DNA methylation on a genome-wide scale, we carried out whole-genome bisulphite
19 sequencing. We used the DNeasy Blood and Tissue kit (QIAGEN) according to the manufacturer's
20 protocol to extract DNA from adult somatic tissues of the following species: *L. polyphemus*, *P.*
21 *tepidariorum*, *S. maritima*, *A. vulgare*, *B. germanica*, *A. pisum*, *B. terrestris*, *N. vespilloides*, *H.*
22 *melpomene* and *D. melanogaster*. For *I. scapularis*, we used the same method to extract DNA from
23 the IDE2, IDE8 and ISE18 cell culture. To estimate bisulphite conversion efficiency, we added a
24 spike-in of unmethylated DNA (P-1025-1, EpiGentek) equal to 0.01% of the sample DNA mass to
25 each sample. We then prepared whole-genome bisulphite sequencing libraries from each DNA
26 sample using the Pico Methyl-Seq Library Prep Kit (Zymo Research), according to the manufacturer's
27 protocol (see Supplementary Table 1 for detailed sample metadata and sequence accession codes).
28 We sequenced these libraries on an Illumina HiSeq 2500 instrument to generate 100bp paired-end

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1 reads. We used pre-existing whole-genome bisulphite sequencing datasets for *P. hawaiensis*
2 (SRR3618947, (Kao *et al.*, 2016)) and *A. mellifera* (SRR1790690, (Galbraith *et al.*, 2015)).

3 To generate bisulphite sequencing data for *P. citri*, we extracted DNA from adult males using the
4 DNeasy Blood and Tissue kit (QIAGEN) according to the manufacturer's protocol. To estimate
5 bisulphite conversion efficiency, we included a spike-in of non-methylated *Escherichia coli* lambda
6 DNA (isolated from a heat-inducible lysogenic *E. coli* W3110 strain, provided by Beijing Genomics
7 Institute (BGI), GenBank/EMBL accession numbers J02459, M17233, M24325, V00636, X00906).
8 Sequencing of bisulphite libraries was carried out by BGI on an Illumina HiSeq 4000 instrument to
9 generate 150bp paired-end reads.

10 **Bisulphite sequencing data analysis**

11 Before mapping reads to the genome, we trimmed sequencing adapters from each read, and then
12 trimmed 10 bases from the 5' and 3' end of each read (using the script
13 https://github.com/SamuelHLewis/BStoolkit/blob/master/BStrim.sh). We aligned bisulphite sequencing
14 reads to each genome using Bismark v0.19.0 (Krueger and Andrews, 2011) in --non_directional mode
15 with default settings. We used MethylExtract v1.9.1 (Barturen *et al.*, 2014) to estimate the level of
16 methylation at each CpG site, calculated as the number of reads in which the cytosine is methylated
17 divided by the total number of reads covering the cytosine, excluding sites covered by fewer than 10
18 reads on each strand. Due to the large number of contigs in their genome assemblies exceeding the
19 memory limit for MethylExtract, we split the genomes of *I. scapularis*, *L. polyphemus* and *P.*
20 *hawaiensis* into individual contigs, ran MethylExtract on each contig separately, and concatenated the
21 resulting output files into one file for each genome.

22 To estimate the genome-wide background level of CpG methylation, we calculated the mean
23 methylation for all CpGs outside annotated features (exon, intron, upstream region, TE, rRNA &
24 tRNA). To gain an accurate estimate of the methylation level of each feature, we calculated the mean
25 methylation level of all CpGs within that feature, excluding any feature with fewer than 3 sufficiently-
26 covered CpGs (only CpGs covered by >10 reads are analysed). We estimated 95% confidence
27 intervals for the mean methylation of genes and TEs within each species using 1000 nonparametric
28 bootstrap replicates (i.e. genes or TEs were resampled with replacement 1000 times to generate an
29 empirical distribution of the mean).

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Phylogenetics and ancestral state reconstruction

To infer the ancestral levels of genome-wide methylation across 29 species of arthropods with newly-produced or publicly-available methylation data (Figure 1), we obtained a time-scaled species tree from TimeTree (www.timetree.org, accessed 12.03.2019). We then used a maximum-likelihood approach to infer the genome-wide methylation level at all internal nodes of this tree based on the levels at the tips, using the fastAnc function within phytools (Revell, 2012).

To infer the ancestral levels of gene-body and TE methylation for the 14 focal species, we constructed a Bayesian time-scaled species tree for 14 focal species (Figure 2 & 3). We first identified 236 proteins present as 1:1:1 orthologues across our species set, concatenated the protein sequences together, and aligned them using MAFFT v7.271 (Katoh and Standley, 2013) with default settings. We then screened out poorly-aligned regions using Gblocks (Castresana, 2000) with least stringent settings. Using this alignment, we constructed a phylogenetic tree using BEAST v1.8.4 (Drummond *et al.*, 2012) to infer branch lengths. We specified a strict molecular clock, gamma-distributed rate variation, no invariant sites, and a birth-death speciation process. We fixed the topology and set prior distributions on key internal node dates (Arthropoda = 568 ± 29, Insecta–Crustacea = 555 ± 33, Insecta = 386 ± 27, Hymenoptera–Coleoptera–Lepidoptera–Diptera = 345 ± 27, Coleoptera–Lepidoptera–Diptera = 327 ± 26), deriving these values from an existing phylogenetic analysis of arthropods (Misof *et al.*, 2014). We ran the analysis for 10 million generations, and used TreeAnnotator (Drummond *et al.*, 2012) to generate a maximum clade credibility tree. We then used a maximum-likelihood approach to infer the gene-body and TE methylation levels (separately) at all internal nodes of this tree, using the fastAnc function within phytools (Revell, 2012).

To test whether genome-wide methylation levels differ between species with and without ALKB2, we fitted a phylogenetic mixed model using MCMCglmm (Hadfield, 2010). To account for phylogenetic non-independence caused by sampling species with different levels of relatedness, we used the branch lengths of the time-scaled (ultrametric) species tree (see above) to calculate a genetic distance matrix, and included this in the model as a random factor. We ran the analysis for 6 million iterations, with a burn-in of 1 million iterations and thinning of 500 generations.

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2 **RNA-Seq data analysis**

3 To investigate the link between DNA methylation and transcription, we used RNA-Seq data generated
4 previously for arthropod somatic tissue (NCBI PRJNA386859, (Lewis *et al*., 2018) and the *I.*
5 *scapularis* IDE-8 cell line (SRR1756347, Arthropod Cell Line RNA Seq initiative, Broad Institute,
6 broadinstitute.org). To measure the expression of each feature, we trimmed adaptors and low-quality
7 ends using Trim Galore with default settings, and mapped RNA-Seq reads to the genome of each
8 species using TopHat2 v2.1.1 (Kim *et al.*, 2013) with default settings for strand-specific libraries (--
9 library-type fr-firststrand mode). We counted the number of reads overlapping each feature using
10 BEDTools coverage v2.25.0 in strand-specific mode, and divided the number of reads by the feature
11 length to generate expression level estimates in fragments per per kilobase million (FPKM).

12 To test whether variation in tissue-specific expression differs between highly- and lowly-methylated
13 genes, we calculated the coefficient of variation for expression of each gene in each species with
14 RNA-Seq data (i.e. excluding *B. germanica*, *I. scapularis* & *P. hawaiensis*). For *S. maritima* we used
15 RNA-Seq data for fat body and nerve chord; for *P. citri* & *A. pisum* we used RNA-Seq data for female
16 soma and germline; and for all other species we used RNA-Seq data for female and male soma and
17 germline.

18 **Periodic correlation in methylation levels**

19 To obtain an estimate of how the correlation between the methylation levels of sites varied with
20 distance between the sites, we collected all pairs of sites separated by *d* nucleotides where *d* could
21 vary between 3 and 500 nucleotides within the same exon. For each separate *d* we then computed
22 the correlation coefficient across all the pairs. To quantify the periodic component of the signal we
23 subtracted any gradual change in correlation across the entire window by calculating the residuals of
24 a linear model. This signal was subjected to Fourier analysis using the fast Fourier transform
25 algorithm implemented in R. A linear model was used to subtract the baseline across the 500bp and
26 the residuals were used as a time series for input into the algorithm, with 50000 0 values ended on to
27 the end of the series to increase the resolution of the algorithm. The total intensity of the components
28 between 140 and 200 base pairs was calculated to give the nucleosome periodicity for each species.

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1 **Nucleosome positioning analysis**

2 The genomic coordinates of the *D. melanogaster* members of orthogroups conserved across all
3 species were extracted and the top 20% (high methylation) and bottom 20% methylation (low
4 methylation) levels selected. Nucleosome positioning data from the *D. melanogaster* S2 cell line was
5 downloaded from Modencode (Ho *et al.*, 2014). The average signal was computed across 200bp
6 windows spanning 2kb either side of the annotated transcription start site for each gene. The mean
7 signal was computed within the high methylation and low methylation sets separately and a loess fit
8 performed. To obtain confidence intervals, the mean signal was computed on 100 random samples
9 containing 90% of the data and a loess fit calculated on the lowest and highest values obtained for
10 each 200bp window.

11 **CAGE data analysis**

12 Total body RNA was extracted from L3 *Drosophila melanogaster* (w^{1118}) larvae using the Qiagen
13 RNeasy kit. CAGE library preparation was performed using the nAnT-iCAGE protocol (Murata *et al.*,
14 2014). Two biological replicates were prepared from 5 ug of total RNA each. The libraries were
15 sequenced in single-end 50 bp-pair mode. CAGE tags (47 bp) were mapped to the reference *D.*
16 *melanogaster* genome (assembly Release 6) using Bowtie2 (Langmead and Salzberg, 2012) with
17 default parameters. Uniquely mapped reads were imported into R (http://www.R-project.org/) as bam
18 files using the standard workflow within the CAGEr package (Haberle *et al.*, 2015). The 5' ends of
19 reads are CAGE-supported transcription start sites (CTSSs) and the number of tags for each CTSS
20 reflects expression levels. Raw tags were normalised using a referent power-law distribution and
21 expressed as normalized tags per million (TPMs). Biological replicates were highly correlated (r^2 =
22 0.99) and were therefore merged prior to downstream analyses using standard Bioconductor
23 packages (http://www.bioconductor.org/) and custom scripts.

24 CTSSs were clustered together into tag clusters, a single functional transcriptional unit, using
25 distance-based clustering, with the maximum distance allowed between adjacent CTSSs being 20 bp.
26 For each tag cluster, the interquantile width was calculated as the distance between CTSSs at the
27 10th and 90th quartile of the cumulative distribution of expression across the cluster. The interquartile
28 range of each gene within the top 20% and bottom 20% of methylation levels was extracted and
29 compared.

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2 **Availability of scripts and data**

3 Sequence data that was newly-generated for this project have been deposited in the NCBI Short
4 Read Archive under the BioProject accession code PRJNA589724. The source code, input data and
5 newly-identified DNMT & ALKB2 gene sequences are available from the Cambridge Data Repository
6 (https://doi.org/10.17863/CAM.45964).

7

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Figure Legends

Figure 1. Genome-wide CpG methylation across the arthropod phylogeny. (A) A phylogeny of 29 arthropod species that have publicly available or newly computed genome-wide methylation estimates, with branches coloured to show an ancestral state reconstruction of the percentage of CpG sites that are methylated in the genome. (B) The percentage of CpG sites that are methylated genome-wide. (C) The number of *DNMT* and *ALKB2* homologues in the genomes of each species.

Figure 2. Methylation of transposable elements. For 14 diverse arthropod species with annotated genomes, we explored methylation characteristics of genomic features. (A) Density plot of the mean % CpG methylation per gene and per TE. (B) Ancestral state reconstruction of the mean % methylation of CpGs within TEs. (C) Mean % methylation of CpGs within TEs with 95% bootstrap confidence intervals. Red points are CpGs >1kB from annotated regions of the genome. (D) Metagene plot of methylation within TEs (pink) and in flanking sequence for *S. maritima* and *P. citri*.

Figure 3 Gene body methylation. (A) Ancestral state reconstruction of the mean % methylation of CpGs within exons. (B) Mean % methylation of CpGs within exons with 95% bootstrap confidence intervals. Red points are CpGs >1kB from annotated regions of the genome. (C) Metagene plot of methylation across introns (white), exons (pink), UTRs (blue) and 1kB of flanking sequence (white).

Figure 4 Promoter methylation. (A) Methylation across upstream regions for highly expressed genes (top 20%) and lowly expressed genes (bottom 20%). *P. hawaiensis* is omitted due to lack of gene expression data. Expression of genes across bins of decreasing upstream methylation in *S. maritima* (B) and *P. citri* (C).

Figure 5 The expression and conservation of methylated genes. (A) Methylation of orthologous genes in different species. Only genes with orthologs in all species are shown, and in species with multiple paralogs the mean % CpG methylation is shown. Genes are ranked by their mean methylation. (B) Histogram of gene expression estimated from RNAseq data for methylated and unmethylated genes in *L. polyphemous* (FPKM: fragments per kilobase million). (C) The relationship between gene expression and CpG methylation for genes that are conserved across all species and species-specific genes. To combine data across species, the methylation rate was normalised by taking the Z-score of methylation and expression of each gene within each species. Each point is a gene from a single species, and the colour represents the density of overlaid points.

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Figure 6. Nucleosome occupancy and DNA methylation. The Pearson correlation coefficient in DNA methylation levels between pairs of CpG at different distances apart in (A) *S. maritima* and (B) *H. melpomene*. (C) Nucleosome occupancy in *D. melanogaster* orthologues of genes that are either highly methylated (grey) or unmethylated (red) in arthropods. Shaded area is a 95% bootstrap confidence interval. Across all species in the dataset, mean methylation levels were estimated for each group of orthologous genes using a general linear mixed model. The top and bottom 20% were classified as methylated and unmethylated respectively. Only genes with orthologs in all species are shown. (D) Interquartile range of the TSS window for the *D. melanogaster* orthologues of highly methylated orthogroups (top 20%) and lowly methylated orthogroups (bottom 20%). (E) The coefficient of variation in expression of genes with high (top 20%) and low (bottom 20%) methylation across different tissues estimated using RNAseq data. *P. hawaiensis* is omitted because no tissue-specific data is available for this species.

Figure 6 supplement 1: estimation of nucleosomal periodicity signal for methylation of exons and transposable elements across all species. (A) workflow using *S. maritima* exons as an example for how baseline correction and fast fourier transform were used to obtain a nucleosome signal. (C) Nucleosome signal as a fraction of total signal for exonic methylation across arthropods. (D) Nucleosome signal as a fraction of total signal for TE methylation across arthropods.

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1 **Supplementary Information**

2 **Figure 1-Supplement: ALKB2 DNA repair is associated with high levels of DNA methylation**
3 **across arthropods**

4 Boxplot showing genome-wide methylation levels in 29 arthropod species with and without ALKB2.

5 **Figure 2 supplement: Metagene plot of methylation within TEs and in flanking sequence for all**
6 **species.** TEs are shown in pink, flanking sequence in white.

7 **Figure 5 supplement 1: Expression patterns of methylated and unmethylated genes for all**
8 **species (cf Figure 5B)**

9 **Figure 5 supplement 2: Methylation of single exon and multi-exon genes for all species in**
10 **which we see gene body methylation**

11 **Figure 6 Supplement 2**: Intron periodicity is markedly less apparent than exon periodicity. *S.*
12 *maritima* exons 1 to 4 (A) and introns 1 to 4 (B) are shown for comparison.

13

14 **Supplementary Table 1:** Details of the tissue type, sex, caste, BioSample Accession and SRA
15 Accession of each sample that was newly-sequenced in this study.

16

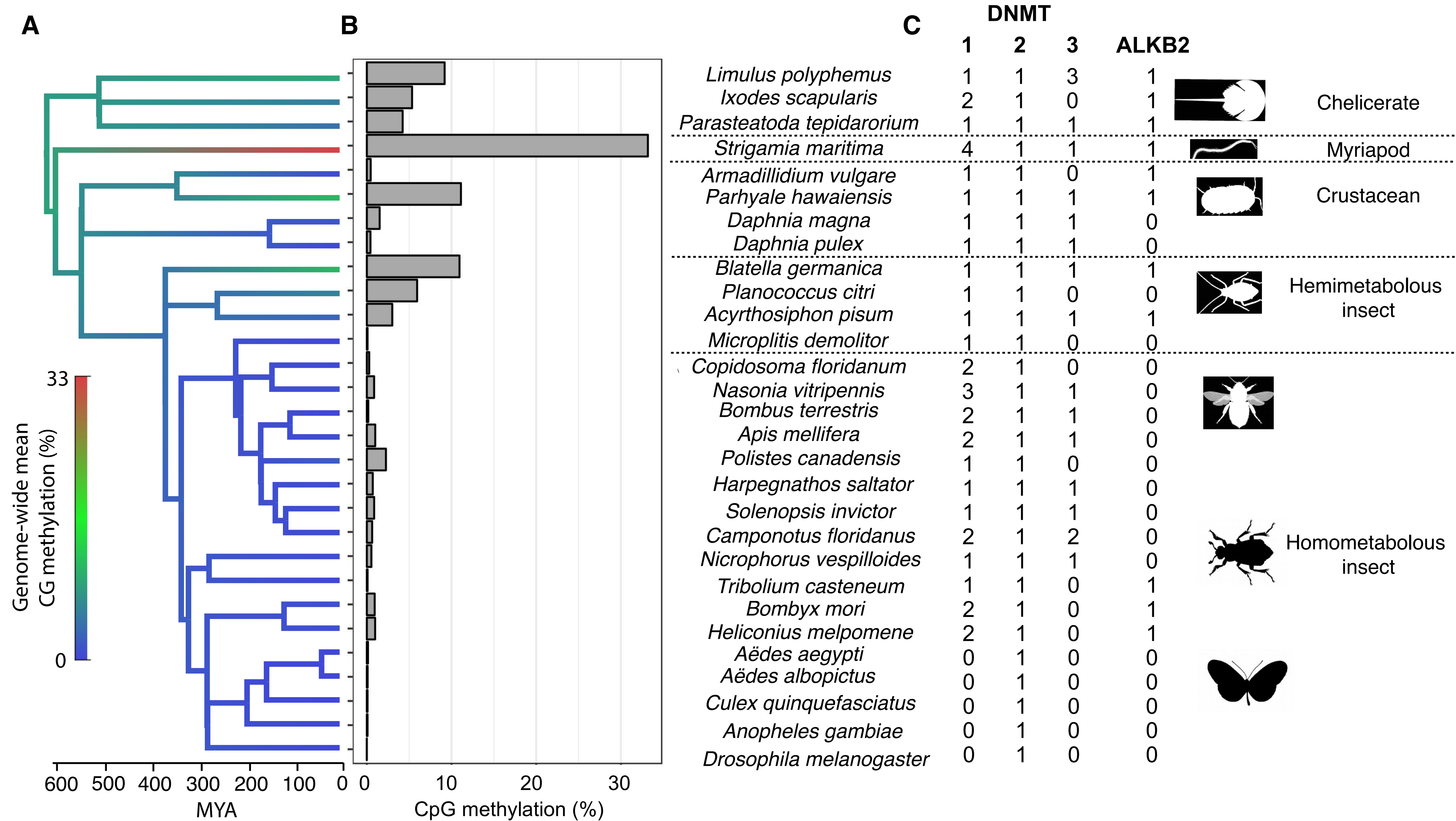

Figure 1

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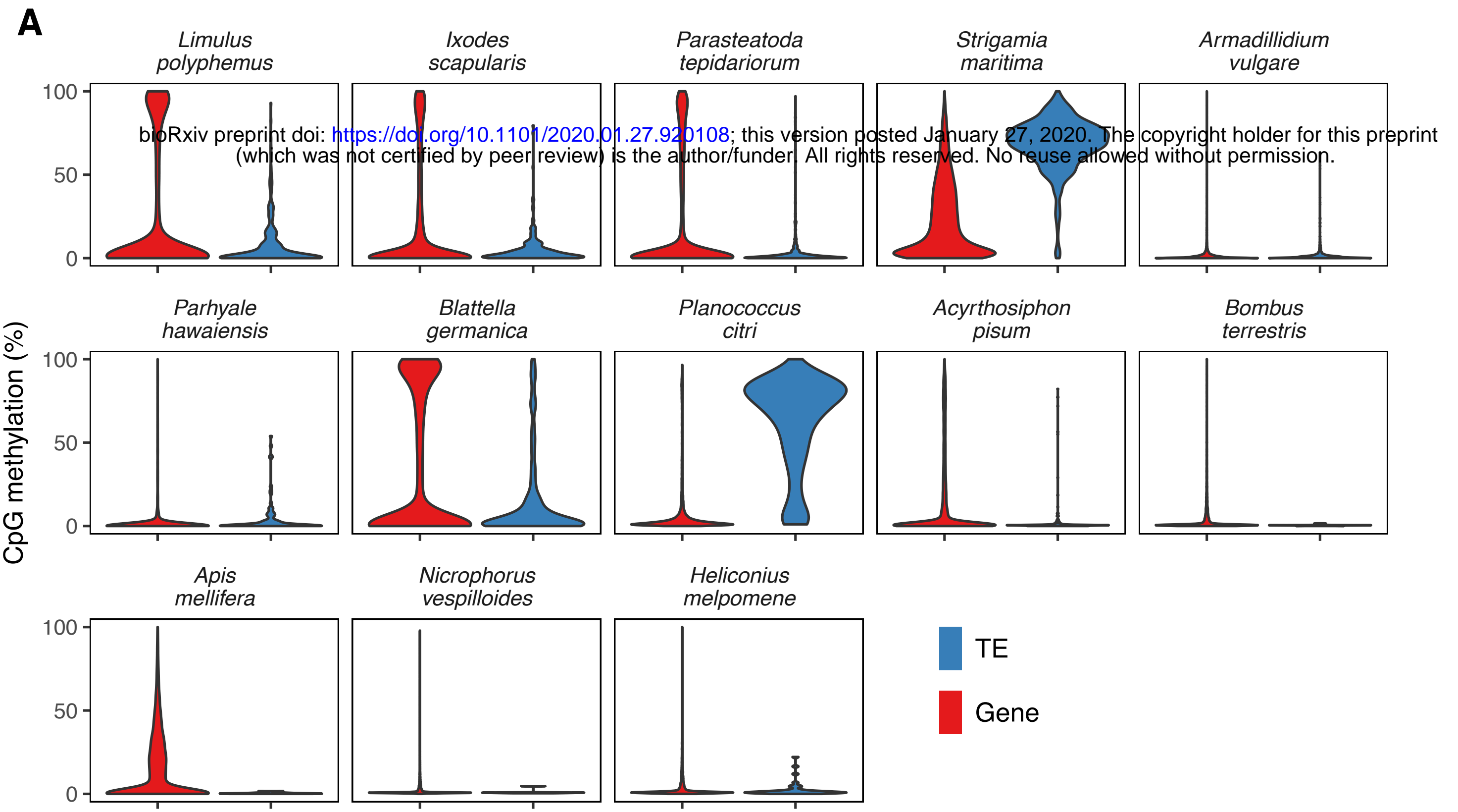

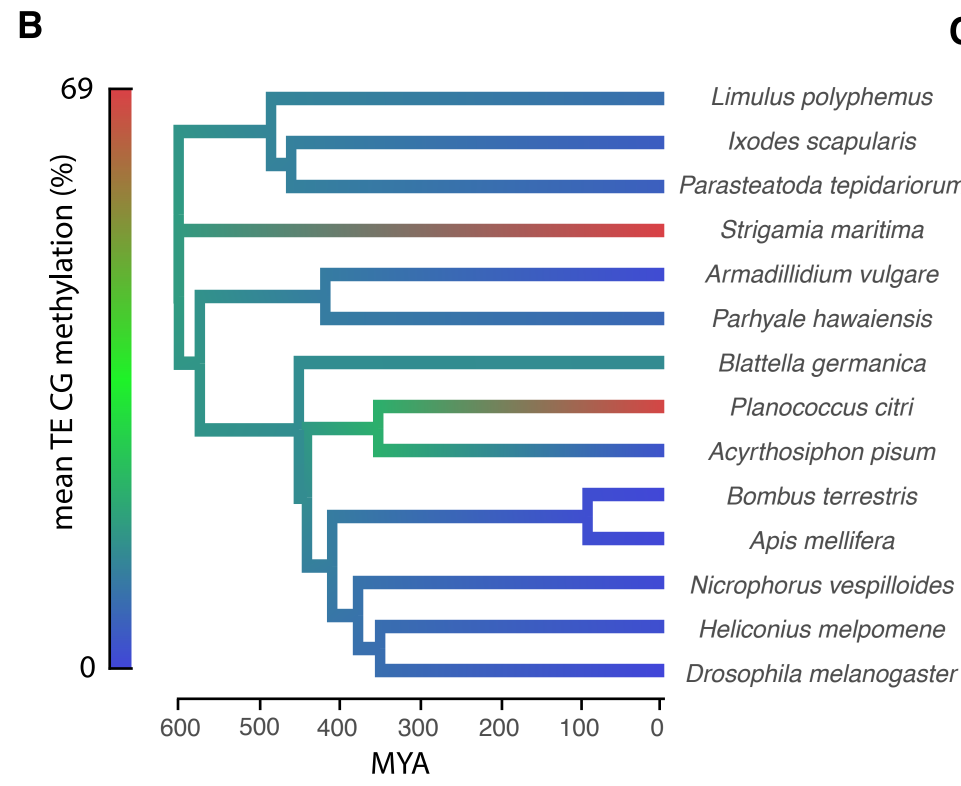

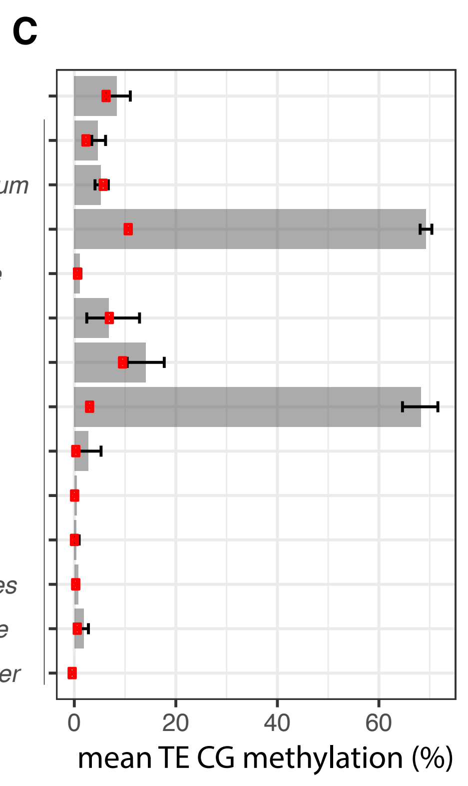

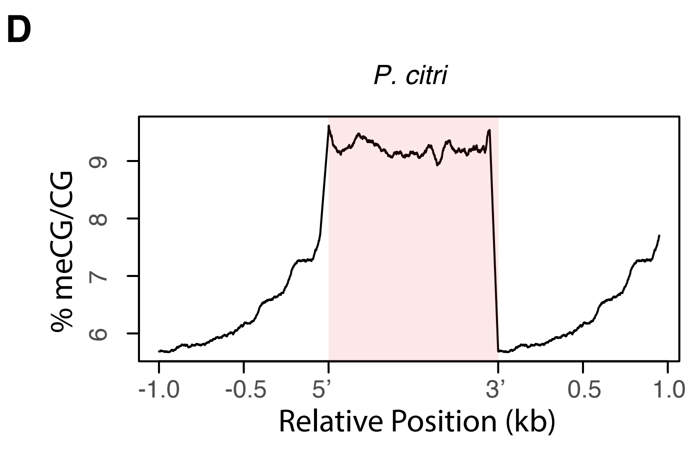

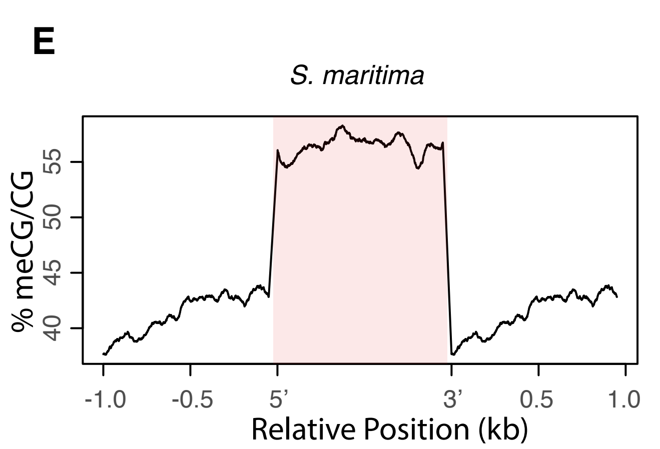

Figure 2

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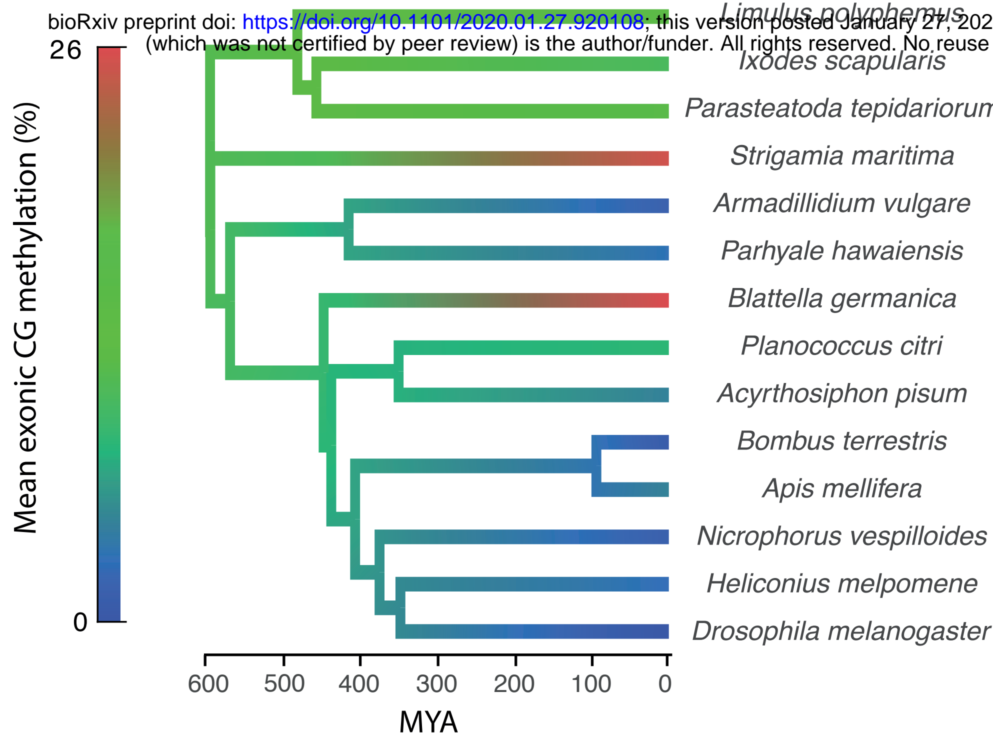

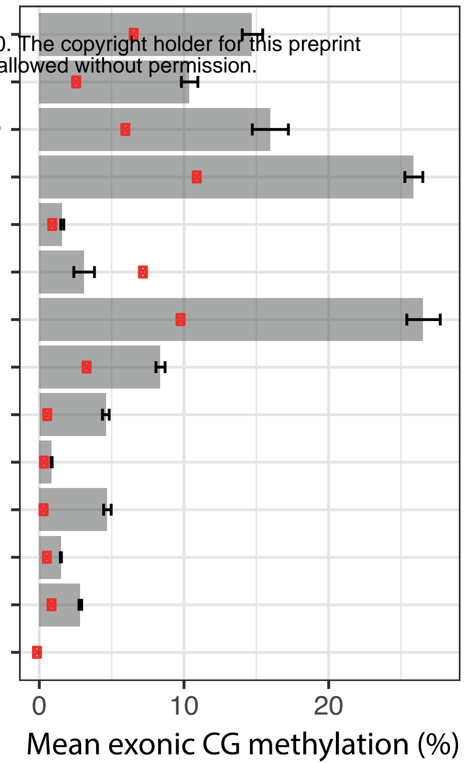

C

Limulus polyphemus

Ixodes scapularis

Parasteatoda tepidariorum

Strigamia maritima

Armadillidium vulgare

Parhyale hawaiensis*

Blattella germanica

Acyrthosiphon pisum

Planococcus citri*

Bombus terrestris

Apis mellifera

Nicrophorus vespilloides

Heliconius melpomene

mean CG methylation (%)

UTR Exon Intron

Figure 3

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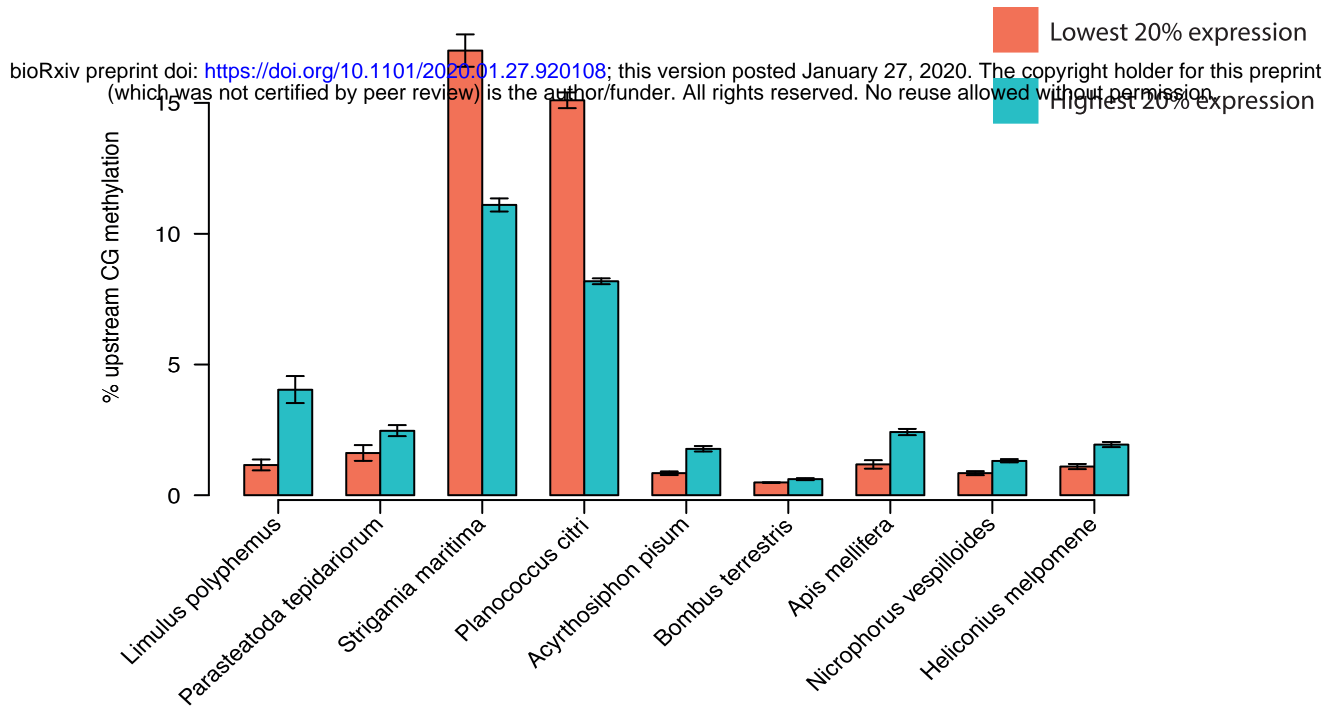
A



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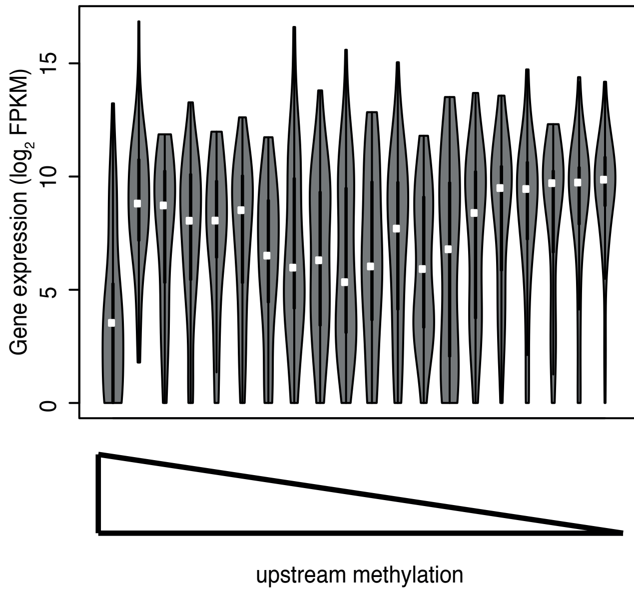

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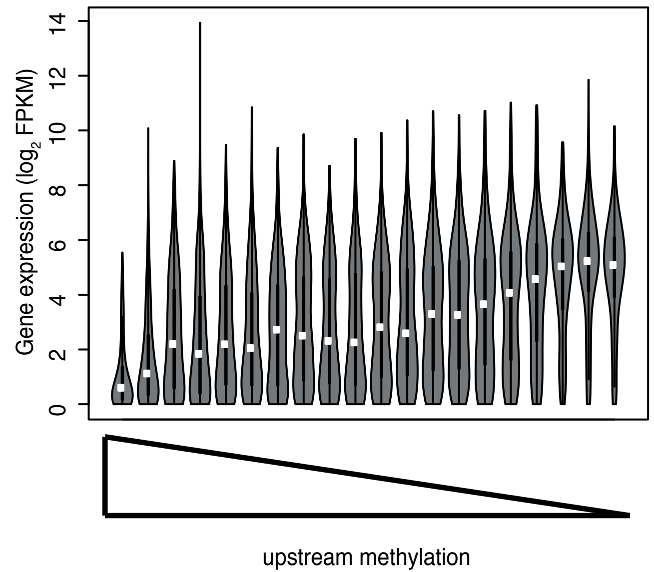

Figure 4

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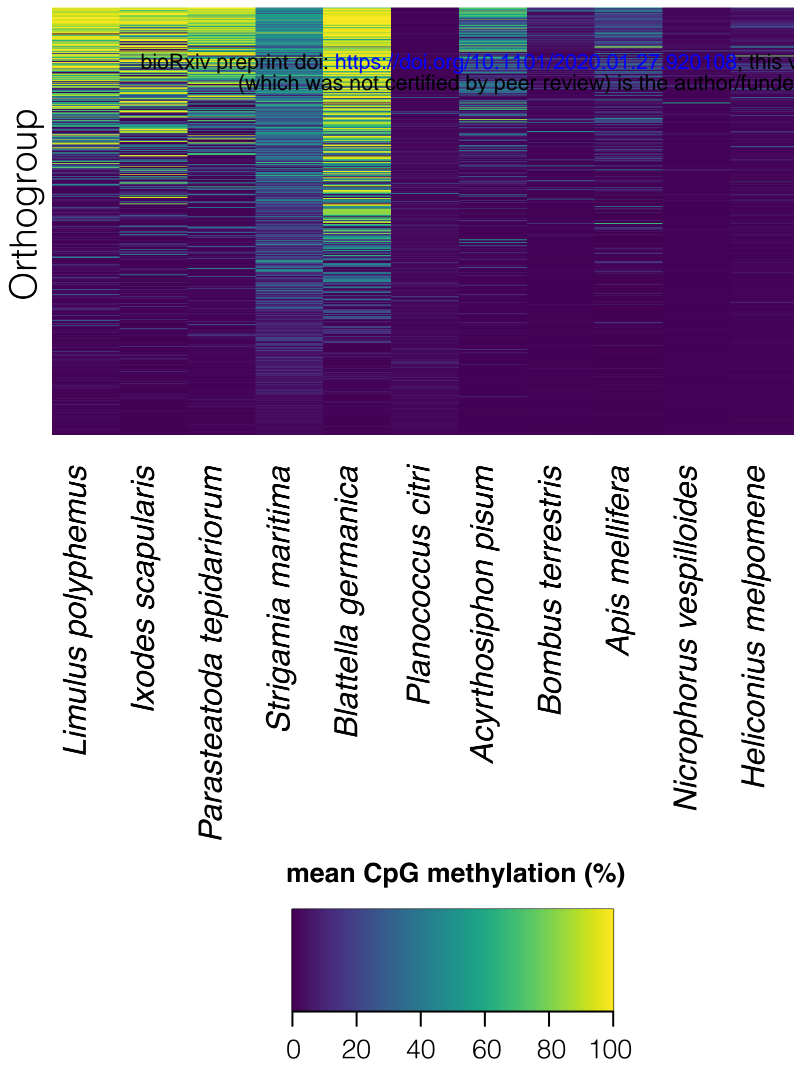

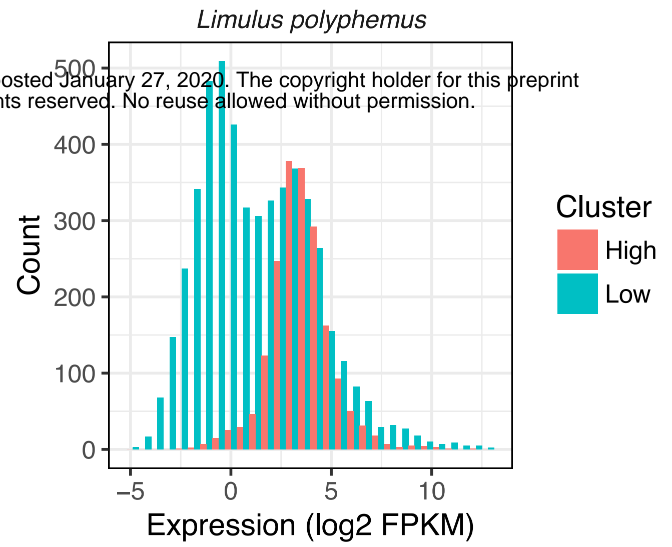

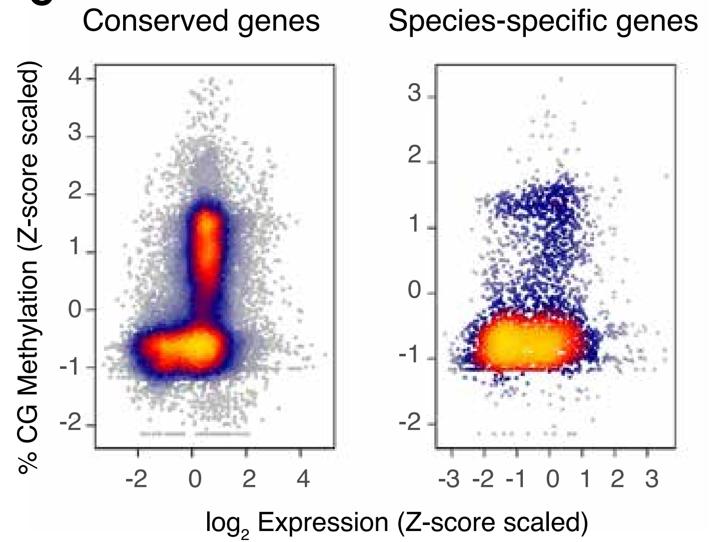

Figure 5

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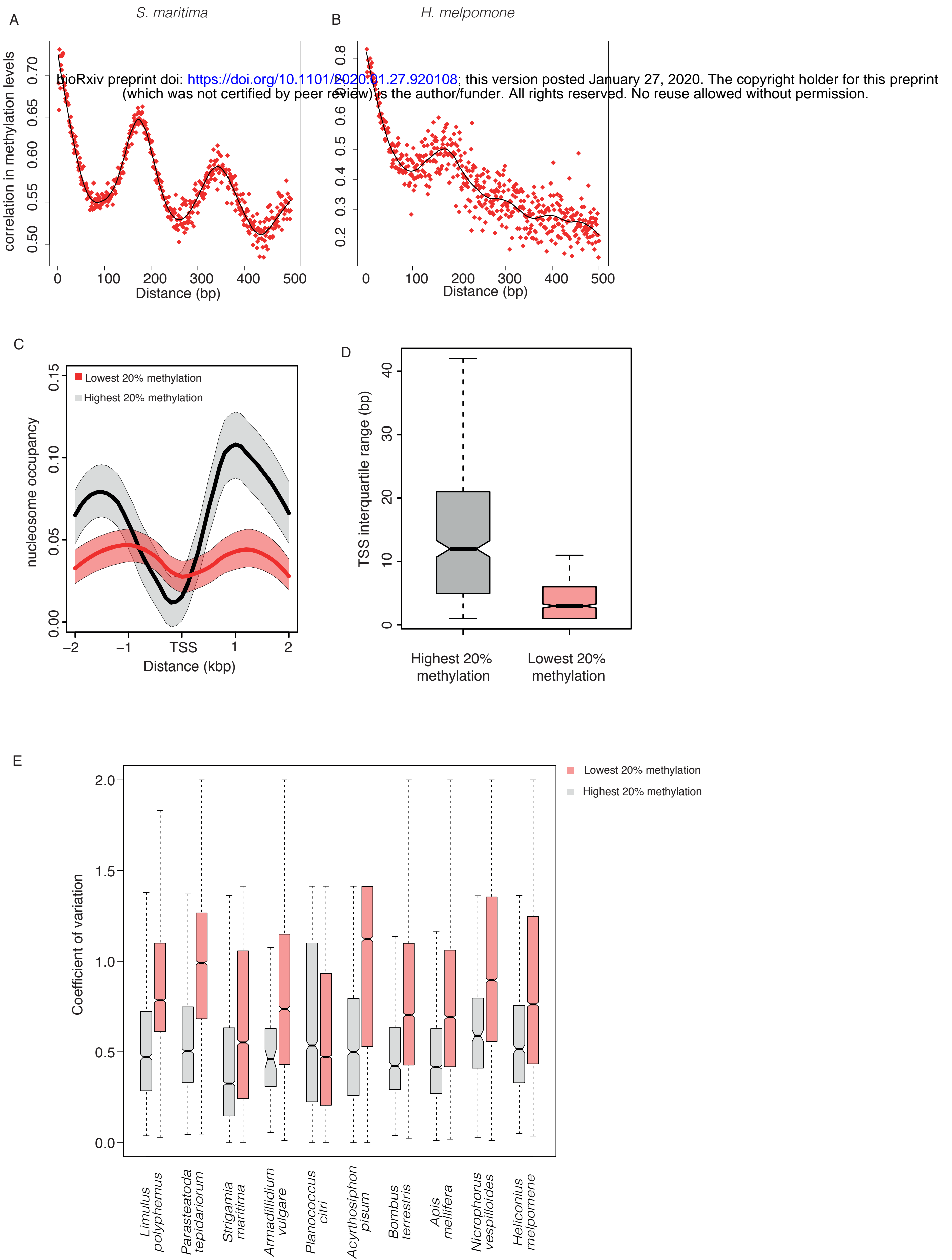

Figure 6