Single-base methylome profiling of the giant kelp *Saccharina japonica* reveals significant differences in DNA methylation to microalgae and plants

Xiao Fan\(^{1,2,3}\), Wentao Han\(^1\), Linhong Teng\(^{1,4,5}\), Peng Jiang\(^6\), Xiaowen Zhang\(^1\), Dong Xu\(^1\), Chang Li\(^7\), Matteo Pellegrini\(^8\), Chunhui Wu\(^6\), Yitao Wang\(^1\), Michelle Joyce Slade Kaczurowski\(^9\), Xin Lin\(^{10}\), Leila Tirichine\(^{11}\), Thomas Mock\(^{12}\) and Naihao Ye\(^{1,2}\)

\(^1\)Yellow Sea Fisheries Research Institute, Chinese Academy of Fishery Sciences, Qingdao, China; \(^2\)Function Laboratory for Marine Fisheries Science and Food Production Processes, Qingdao, China; \(^3\)Key Laboratory of Exploration and Utilization of Aquatic Genetic Resources, Ministry of Education, Shanghai Ocean University, Shanghai 201306, China; \(^4\)College of Life Science, Dezhou University, Dezhou 253023, China; \(^5\)Shandong Key Laboratory of Biophysics, Dezhou University, Dezhou 253023, China; \(^6\)Institute of Oceanology, Chinese Academy of Sciences, Qingdao 266071, China; \(^7\)University of Chinese Academy of Sciences, Shenzhen, China; \(^8\)Department of Molecular, Cell and Developmental Biology, Institute for Genomics and Proteomics, University of California, Los Angeles, CA 90095, USA; \(^9\)Biological Sciences, Flinders University, GPO Box 2100, Adelaide, SA 5001, Australia; \(^10\)State Key Laboratory of Marine Environmental Science, College of Ocean & Earth Sciences, Xiamen University, Xiamen, China; \(^11\)CNRS UMR 6286, Faculté des Sciences et des Techniques, Université de Nantes, 2 rue de la Houssinière, 44322, Nantes, France; \(^12\)School of Environmental Sciences, University of East Anglia, Norwich Research Park, Norwich, NR4 7TT, UK

**Summary**
- Brown algae have convergently evolved plant-like body plans and reproductive cycles, which in plants are controlled by differential DNA methylation. This contribution provides the first single-base methylome profiles of haploid gametophytes and diploid sporophytes of a multicellular alga.
- Although only c. 1.4% of cytosines in *Saccharina japonica* were methylated mainly at CHH sites and characterized by 5-methylcytosine (5mC), there were significant differences between life-cycle stages. DNA methyltransferase 2 (DNMT2), known to efficiently catalyze tRNA methylation, is assumed to methylate the genome of *S. japonica* in the structural context of tRNAs as the genome does not encode any other DNA methyltransferases. Circular and long noncoding RNA genes were the most strongly methylated regulatory elements in *S. japonica*.
- Differential expression of genes was negatively correlated with DNA methylation with the highest methylation levels measured in both haploid gametophytes. Hypomethylated and highly expressed genes in diploid sporophytes included genes involved in morphogenesis and halogen metabolism.
- The data herein provide evidence that cytosine methylation, although occurring at a low level, is significantly contributing to the formation of different life-cycle stages, tissue differentiation and metabolism in brown algae.

**Introduction**
Cytosine DNA methylation is a common epigenetic mark essential for genomic imprinting, X-chromosome inactivation and silencing of transposable elements, as well as regulation of gene expression in many species (Meissner et al., 2008; Feng et al., 2010; Law & Jacobsen, 2010; Stelzer et al., 2015). Genome-wide methylation studies with many different plant and animal species but also unicellular eukaryotes (e.g. fungi, algae) over the last few years have revealed significant inter- and intraspecific variations in cytosine DNA methylation (Cokus et al., 2008; Lee et al., 2010; Molaro et al., 2011; Greaves et al., 2012; Ziller et al., 2013). For example, vertebrate genomes are characterized by significant levels of cytosine methylation, whereas genomes of invertebrates, plants and fungi are characterized by sparse DNA methylation (Sturgill et al., 2007; Feng et al., 2010). Generally, the degree of DNA methylation is positively correlated with the complexity of organisms; thus, prokaryotes and unicellular eukaryotes have much lower levels of DNA methylation compared to organisms with complex life cycles, developmental stages and cell-type differentiation (Molaro et al., 2011; Lopez et al., 2015). Furthermore, there are significant differences in establishing, maintaining and modifying DNA methylation in plants vs animals. For instance, in plants, DNA methylation has been observed for GC, CHG and CHH contexts with H being any nucleotide but G. DNA methylation in plants occurs...
predominantly on repetitive DNA elements (e.g. transposons) regardless of their life cycle and developmental stages (Lopez et al., 2015; Takuno et al., 2016). Similar DNA methylation patterns have been observed for unicellular photosynthetic eukaryotes (e.g. microalgae) (Veluchamy et al., 2013). By contrast, DNA methylation in mammals mostly occurs in the GC context throughout the genome except in clusters near promoters (CpG islands) (Molaro et al., 2011; Smith et al., 2012; Hon et al., 2013; Kundaje et al., 2015). Generally, DNA methylation in plants and animals is dynamic with genome-wide reduction during both male and female gametogenesis as well as development (Molaro et al., 2011; Shao et al., 2014; Lopez et al., 2015). However, in plants, many DNA methylation patterns seem to be inherited over many generations whereas transgenerational DNA methylation in animals is much more variable (Hon et al., 2013; Lister et al., 2013; Shao et al., 2014).

Brown algae, although only distantly related to plants and animals, have convergently evolved plant-like body plans and reproductive cycles including male and female gametogenesis (Charrrier et al., 2012). Furthermore, many brown algae have an alternative life cycle with two life stages, termed gametophytes (1N) and sporophytes (2N) (Charrrier et al., 2012; Cock et al., 2014). Gametophytes from some kelp species, such as Saccharina japonica, can even develop into larger multicellular organisms but rarely reach tissue differentiation (Ye et al., 2015). Some other brown algae, such as Ectocarpus siliculosus, have isomorphic life-cycle stages that are filamentous and phenotypically indistinguishable from one another (Luthringer et al., 2014).

Brown algae comprise a group of c. 2000 species, possessing a large variety of phenotypes including the largest multicellular photosynthetic organisms in the ocean with distinct and specialized tissue differentiation, such as holdfast, blade and stripe. Despite their convergently evolved plant-like body plans, their ecological (e.g. main primary producers of temperate and polar rocky shores) and commercial (e.g. alginate, fucoidan) significance, genomes of only three brown algal species are available to date: E. siliculosus (Cock et al., 2010), S. japonica (Ye et al., 2015) and Cladosiphon okamuranus (Nishitsuji et al., 2016). A comparative study based on the genomes of E. siliculosus and S. japonica revealed that they share 4309 gene families, which comprise 17 379 genes in S. japonica and 14 136 genes in E. siliculosus, covering 93% and 86% of the gene content of each genome, respectively. About 40% of the assembled S. japonica genome comprises repetitive elements, which is nearly twice as much as for the E. siliculosus genome (c. 23%) (Ye et al., 2015).

DNA methylation in brown algal genomes was considered to be negligible based on preliminary high performance liquid chromatography (HPLC) analyses of deoxycytosine methylation (5mC) of hydrolyzed DNA from E. siliculosus (Cock et al., 2010). These data indicated that the percentage of 5mC in E. siliculosus is < 0.035%. However, the present study using single-base DNA methylome profiling of a fully developed S. japonica sporophyte (SP), female (FG) and male (MG) gametophyte revealed that c. 1.4% of all cytosines in S. japonica were methylated in GC, CHG and CHH contexts potentially mediated by a DNA methyltransferase 2 (DNMT2) as there is no other DNA methyltransferase encoded in the genome of S. japonica, which is different to most plants, animals and unicellular algae such as the green alga Chlamydomonas reinhardtii (Lopez et al., 2015) and even the more closely related diatom Phaeodactylum tricornutum (Takuno et al., 2016; Tirichine et al., 2017). Among the three life-cycle stages of S. japonica, the highest level of DNA methylation was found in both gametophytes. Furthermore, the highest methylated elements of any life-cycle stage were found to be genes encoding noncoding RNAs (circular and long noncoding), which is different to other photosynthetic eukaryotes as they mostly methylate repetitive elements (e.g. transposons) (Sleutels et al., 2002; Dinger et al., 2008; Law & Jacobsen, 2010). Differentially methylated genes in MG and in FG were significantly enriched for cellular processes, cell–wall organization and cell–cell junctions, whereas genes in SP were more enriched for rRNA modification and DNA methylation. However, for all life-cycle stages, an overall negative correlation was found between DNA methylation and gene expression. Thus, despite an overall low level of cytosine methylation in the genome of S. japonica, there is evidence that it may play a significant role for the development of life-cycle stages and regulation of metabolism via the control of gene expression and noncoding RNAs, which is different to what has been observed so far in plants, animals and many microbes.

Materials and Methods

Strain selection, DNA and RNA extraction and purification

The Saccharina japonica strain Ye-c12 was used for methylated DNA immunoprecipitation (MeDIP), bisulfite and RNA sequencing. The haploid male (MG) and female (FG) gametophytes were collected from a blade of the diploid Ye-c12 sporophyte (SP) and expanded via the application of gametophyte cloning technology (Wang et al., 2013). The diploid sporophyte of Ye-c12 was obtained from sexual reproduction of FG and MG, and collected when the blade length was c. 3 cm. Genomic DNA from haploid gametophytes and diploid sporophytes was extracted using the Plant Genomic DNA kit (Tiangen, Beijing, China), and RNA was removed by incubating the DNA solution at 37°C with a DNase-free RNase A (Tiangen) (Supporting Information Table S1) for 30 min. DNA integrity was assessed using agarose gels and a NanoPhotometer® spectrophotometer (IMPLEN, Westlake Village, CA, USA). DNA was quantified with a Qubit® 2.0 Flurometer (Life Technologies, Carlsbad, CA, USA) using the Qubit® DNA Assay Kit.

Total RNA was extracted using the RNAprep Pure Plant Kit (Tiangen) and DNA was removed using an RNase-Free DNase I treatment according the instructions by the manufacturer (Tiangen). RNA integrity was assessed on 1% agarose gels and by using the RNA Nano 6000 Assay Kit of the Bioanalyzer 2100 system (Agilent Technologies, Palo Alto, CA, USA). RNA was quantified with a Qubit® 2.0. Flurometer (Life Technologies). A total amount of 3 μg RNA per sample was used as input material. Ribosomal RNA was depleted by Epicentre Ribo-zeroTM rRNA Removal Kit (Epicentre, NB, USA), and the RNA-free residue
was cleaned up by ethanol precipitation. Subsequently, libraries were generated using the rRNA-depleted RNA by NEBNext UltraTM Directional RNA Library Prep Kit for Illumina (NEB, USA). The Illumina HiSeq 2500 paired-end platform was used to sequence the libraries with a read length of 125 bp.

Identification of genes encoding circular, long noncoding and transfer RNAs (circRNAs, lncRNAs and tRNAs)

Unmapped RNA-seq reads were kept and 20-mers from 5′ and 3′ ends of these reads were extracted and aligned independently to reference sequences by Bowtie v.2.0.6. Anchor sequences were extended by nd_circ such that the complete reads aligned and the breakpoints were anchored by GU/AG splice sites. The backspliced reads with at least two supporting reads were annotated as circRNAs. CENC (Coding-Non-Coding-Index) (v.2) with default parameters was used to effectively distinguish protein-coding and noncoding sequences using known annotations CPC (Coding Potential Calculator) (0.9-r2) . mainly was used to assess the extent and quality of the ORF in a transcript and search the sequences with the NCBI eukaryotes’ protein database with the e-value of ‘1e-10’. Each transcript was translated in all three possible reading frames and used PfamSCAN (v.1.3) to identify occurrence of any of the known protein family domains documented in the Pfam database (Pfam A and Pfam B), by default parameters of −E 0.001 –domE 0.001. Transcripts predicted with coding potential by either/all of the above were filtered out, and those without coding potential were the candidate set of lncRNAs. Genome-wide tRNA genes were predicted using the rRNAscan-SE web server (Lowe & Chan, 2016). The genomic loci for circRNA, lncRNA and tRNA were deposited at CNGB Nucleotide Sequence Archive (https://db.cngb.org/cnsa/, project CNP0000364).

Determining the level of heterozygosity between gametophytes and the reference genome

Both parental haploid gametophytes were used to perform whole-genome sequencing (WGS) employing an Illumina Hiseq 2500 platform, and the reads of each parent were mapping to the reference S. japonica genome using BWA (Li & Durbin, 2010). The heterozygosity sites and indel sites were called using SAMTOOLS (Li et al., 2009), which also were used to calculate the heterozygosity between individual haplotypes and the reference genome.

Methylated DNA immunoprecipitation (MeDIP)

Extracted genomic DNA was sonicated to obtain fragments from 100 to 500 bp. The fragmented DNA was end-repaired, subjected to A-tailing and PE adapters were ligated. Subsequently, the treated DNA fragments were immunoprecipitated with an antibody that specifically recognizes 5-methylcytosine using the MagMeDIP Kit (Diagenode, Liege, Belgium). The specificity of the enrichment was confirmed by quantitative real-time polymerase chain reaction (qRT-PCR). After PCR amplification of the enriched fragments, they were quantified with the Agilent 2100 Analyzer (Agilent Technologies). Sequencing libraries of MeDIP fragments were constructed by adopting the Illumina paired-end protocol. The negative control (input) MeDIP experiment was performed using the same procedure except for the absence of DNA immunoprecipitation with the 5-methylcytosine recognizing antibody.

MeDIP-Seq sequence alignments and data analysis

Raw sequencing data were filtered by removing the adaptors and discarding low-quality reads. Clean reads were aligned to the reference genome of S. japonica with Bowtie’s (Langmead & Salzberg, 2012) best mode using default parameters. Read depths of each sample were normalized to eliminate the influence via differences of total read numbers between samples (Fig. 1a). After using input reads to eliminate the background noise in three MG samples, peaks were called using ‘MACS2 callpeak’ (Zhang et al., 2008) (Fig. S2A). According to GTF annotation, the normalized absolute read depth (NARD) between genomic components were calculated, including GENOME (genome-wide), TSSUP (transcription start site), TESDOWN (transcript stop site), exon, intron, transposable elements (TEs), genes encoding for lncRNAs, circRNAs and tRNAs. Absolute read depths were normalized to genome read depths in order to be comparable between any two classes. Absolute read depth (ARD) for each class of genomic components was calculated as following: ARD = (absolute read counts (ARC) × 150)/total length of this class. Finally, ARDs of each class were normalized to the NARD of its genome (Fig. S2B). The NARDs across the genomic components were represented via metaplot (Fig. S2C). The regions 2 kbp upstream and downstream of the components were split into 20 nonoverlapping bins, whereas genes were split into 40 equal windows.

Library preparation and quantification of whole-genome bisulfite sequencing (WGBS)

A total amount of 5.2 μg genomic DNA spiked with 26 ng lambda DNA were fragmented to 200–300 bp by sonication (Covaris S220), followed by end-repair and adenylation. Cytosine-methylated barcodes were ligated to sonicated DNA. DNA fragments were treated twice with bisulfite using EZ DNA Methylation-Gold™ Kit (Zymo Research, Irvine, CA, USA), before the resulting single-stranded DNA fragments were PCR-amplified using KAPA HiFi HotStart Uracil + ReadyMix (2×). Libraries were quantified by Qubit® 2.0 Flurometer (Life Technologies) and qRT-PCR, and the insert size was assessed on the Agilent Bioanalyzer 2100 system.

Clustering and data analysis of WGBS

Clustering of the index-coded samples was performed on a cBot Cluster Generation System using TruSeq PE Cluster Kit v.3-cBot-HS (Illumina) according to the manufacturer’s instructions. After cluster generation, the library preparations were sequenced on an Illumina Hiseq 2500 platform and 125-bp paired-end reads were generated. Image analysis and base calling were
performed with the Illumina CASAVA pipeline, and finally 125 bp paired-end reads were generated. BISMARK software (v.0.16.1) (Krueger & Andrews, 2011) was used to perform alignments of bisulfite-treated reads to a reference genome using default parameters. Results of the methylation extractor were transformed to BIGWIG format for visualization using the IGV browser (Thorvaldsdóttir et al., 2013). The bisulfite nonconversion rate was calculated as the percentage of cytosines sequenced at cytosine reference positions in the lambda genome.

Calculation of methylation levels of mCs using WGBS

The methylation level of methylated cytosines (MLmc) was calculated by the read count of methylated cytosines mapped to SP. FG and MG methylation levels were calculated by applying a 10-kbp sliding window. The sliding step length was set at 10 kbp ensuring no region was accounted for twice. The CpG ratio (grey track), CPGo/e (observed value/expected value, dark red track), C + G Ratio (yellow track), CpG ratio (green track), CHH ratio (light red track) and the CHG ratio (blue track) are displayed by the six outermost circles in the graph. Differential transcription (normalized read counts) between the different life-cycle stages in relation to DNA methylation (MLgf) is shown by the six innermost circles.

Fig. 1 Methylome and transcriptome profiles for 31 chromosomes of the Saccharina japonica sporophyte (SP), male (MG) and female (FG) gametophyte genomes. All calculations were conducted by applying a 10-kbp sliding window. The sliding step length was set at 10 kbp ensuring no region was accounted for twice. The CpG ratio (grey track), CPGo/e (observed value/expected value, dark red track), C + G Ratio (yellow track), CpG ratio (green track), CHH ratio (light red track) and the CHG ratio (blue track) are displayed by the six outermost circles in the graph. Differential transcription (normalized read counts) between the different life-cycle stages in relation to DNA methylation (MLgf) is shown by the six innermost circles.
their genomic locus divided by the read count of all cytosines mapped to the same locus:

\[
ML(C) = \frac{\text{reads}(mC)}{\text{reads}(mC) + \text{reads}(C)}
\]

The methylation level of a specific genomic fragment (MLgf) (including every single gene, intron, circRNA, lncRNA, transcriptional start (TSSUP) and stop (TESDOWN) CpG island (CGI), tRNA and 10k bin of chromosomes or the entire chromosome) was calculated based on the methylation levels of all mCs of this genomic fragment divided by the sum of cytosines and guanines of the same fragment (Figs 2f, 3) (Xiang et al., 2010).

In order to identify genomic components that were preferentially methylated, the methylation level of a class of genomic component (MLgc such as tRNAs, lncRNAs) was calculated as following: the sum of MLmc of all mCs belonging to a specific component (e.g. tRNAs, lncRNAs) divided by the sum of cytosines and guanines of this component. The approach also was used to calculate the methylation level of the whole genome (MLwg; Fig. 2b). To remove experimental artifacts, noise and any batch effects, MLgc were normalized according to MLwg, which is defined as the normalized methylation level of genomic components (NMLgc; Fig. 2c) (Xiang et al., 2010).

Differentially methylated sites (DMSs: a DMS corresponds to a single nucleotide), regions (DMRs) and promoters (DMPs) of WGBS

The DMRs were identified using the swDMR software (http://122.228.158.106/swDMR/), which uses a sliding-window approach. The window size of 1000 bp using a step length of 100 bp was used. To avoid overlap of neighbouring DMRs, the sliding-window always started downstream of already identified DMRs. Fisher’s exact test was implemented to detect DMRs. The areaStat value that combines both the \(P\)-values and fold-change between two experimental groups were introduced to identify the DMR (see details for areaStat: http://www.biocond uctor.org/packages/release/bioc/vignettes/DSS/inst/doc/DSS. html#11_background). A larger absolute value of the areaStat is more likely to be a DMR. The minimal absolute value of areaStat in the present study is 40, resulting in a greater than six-fold difference in methylation level between two experimental groups \((n=3)\) with \(P<0.05\). Customized scripts were used to identify DMSs using Fisher’s exact test with FDR multiple test correction. DMSs with \(P<0.05\) and a fold-change of \(>2\) in methylation level between two experimental groups \((n=3)\) were considered candidate DMSs. The identification of DMPs was conducted for 2000 bp upstream of each gene using the same significance criteria as for DMRs.

Validation of DNA methylation by bisulfite sequencing PCR (BS-PCR)

The MLgf of specific genomic regions were validated by BS-PCR. Genomic DNA (500 ng) of MG, FG and SP were converted using EpiTect Fast Bisulfite Conversion Kit (Qiagen) according to the manufacturer’s instruction. CpG islands were identified using the MethPrimer (http://www.urogene.org/meth primer/), which also was used to design specific BSP methylation primers. PCRs were carried out with the following program: initial denaturation at 95°C for 30 s followed by 40 cycles of 95°C for 5 s, 60°C for 10 s, and 72°C for 40 s. The PCR products were purified and recovered by a DNA Universal DNA Purification Kit (Tiangen). Next, purified PCR products were cloned into the pMD18-T vector (Takara, Beijing, China) and transformed into Escherichia coli DH5α-competent cells (Yixin, Shanghai, China) for further replication. Positive clones were screened and sequenced (Sunny, Shanghai, China). Up to 30 randomly picked clones for each amplified locus were sequenced using the ABI3730xl platform.

Phylogenetic analysis

Given the large diversity of DNA methyl transferases (DNMTs), only the DNA methylase motif region (PF00145) was used for constructing phylogenetic trees after sequence alignment using Muscle 3.8.31 (Edgar, 2004). Maximum-likelihood (ML) phylogenetic trees were constructed using MEGA6 with the LG + G + F model, predicted as best model by the ‘Find best DNA/protein models’ module of MEGA6 (Tamura, 2013). Motif structures of genes were visualized using the Interactive Tree of Life (iTOL) (J vic & Peer, 2016). 3D structures of DNMT proteins were visualized using X-ray structure models from Swiss-Model https://www.swiss model.expasy.org/.

Quantitative real-time PCR

Total RNA was extracted using TRIzol (Invitrogen) according to the user manual. Purified RNA was dissolved in diethylpyrocarbonate (DEPC)-treated water. cDNA was synthesized from the total RNA using Moloney murine leukemia virus reverse transcriptase (Promega Biotech). The qRT-PCR reactions were performed with the ABI StepOne Plus Real-Time PCR System (Applied Biosystems, Foster City, CA, USA) using SYBR Green fluorescence (TaKaRa Bio, Kusatsu, Japan) according to the manufacturer’s instruction. The 18S rDNA gene was used as reference. For each of the selected genes, three biological replicates were assayed independently. The qRT-PCR amplifications were carried out in a total volume of 20 μl containing 10 μl of 26 SYBR Premix Ex TaqTM II (TaKaRa Bio), 0.6 μl (10 mM) of each primer, 2.0 μl of the diluted cDNA mix and 6.8 μl de-ionized water. The qRT-PCR protocol was as follows: initial denaturation at 95°C for 30 s followed by 40 cycles of 95°C for 5 s, 60°C for 10 s, and 72°C for 40 s. The \(2^{\Delta\Delta CT}\) method was used to analyze the qRT-PCR data.

Data availability

Datasets S1–S9 and their source Illumina clean data for methylohex have been deposited in NCBI’s Gene Expression Omnibus and are accessible through GEO Series accession no. GSE117191 (https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE117191). The Illumina raw data for transcriptomes have been deposited at the NCBI database (the SRA accession
numbers are SRR5860561–SRR5860568). The Illumina data of the MeDIP-seq experiments and the WGS data were deposited at CNGB Nucleotide Sequence Archive (https://db.cngb.org/cnsa/, with project accession no. CNP0000364), as were the reassembled genome and genomic loci for TEs, CGIs, genes, circRNAs, lncRNAs, tRNAs and the detailed methylation data for genome-wide visualization (visualization data is also deposited in GEO Series accession no. GSE117191). The methylation data based on WGBS and the gene expression based on RNA-seq can be visualized using the online genome browser at the OrcAE database (https://bioinformatics.psb.ugent.be/orcae/).
Results

Assessment of DNA methylation based on an improved genome assembly

In order to obtain evidence of cytosine methylation in the genome of *S. japonica*, MeDIP was performed using a mixture of DNA from all three life stages of the *S. japonica* strain Ye-c12 (SP, FG, MG). Methylated DNA was recognized by a 5-methylcytosine antibody. After PCR amplification, 220–520-bp fragments were selected to construct the sequencing libraries. The libraries revealed a positive signal of the expected length (220–520 bp; Fig. S1A), indicating the presence of 5-methylcytosines in the DNA of *S. japonica*. However, the positive control (Bos DNA) gave near six-fold stronger signals (Fig. S1B), suggesting that the concentration of 5-methylcytosines was significantly higher in Bos DNA (average MLwg of 8.43%) compared to *S. japonica* (Dechow & Liu, 2018).

In order to provide first insights into the methylome landscape on a chromosome-wide scale, pseudo-chromosomes were constructed by combining a genetic linkage map (Zhang et al., 2015) with scaffold information. A total of 1576 scaffolds were anchored to 31 linkages of the genetic map (Fig. 1; Methods S1, part 1), accounting for 64.69% (35 293 Mb) of the assembled kelp scaffolds. The remaining scaffolds were concatenated to construct an artificial chromosome (Figs S2, S3; Methods S1, part 1).

Cytosine methylome landscapes based on MeDIP-seq

In order to obtain insights into the global methylation landscape between life stages, independent MeDIP experiments were performed for FG, MG and SP with three biological replicates each. However, only MG produced enough immunoprecipitated DNA after PCR amplification. The negative results for FG and SP were verified by additional MeDIP experiments, suggesting reduced levels of methylation in FG and SP. Then WGB-seq were performed for three MG replicates and one negative control (also named as input). A total of 20 Gbp paired-end reads were generated from three MG MeDIP-seq libraries and one input library (Table S1). The MeDIP-seq reads were mapped across chromosomal regions in three MG and one control sample (Fig. S2A); 8824 peaks were obtained in a total length (three replicates) of 8 764 952 bp, covering 1.61% of the genome (Table S2). The distribution of MeDIP-seq reads (normalized absolute read depth = NARD) showed that the reads were distributed across all of the genomic feature regions (Fig. S2B). Interestingly, lncRNA genes and circRNA genes had the highest level of NARDs (*P* < 0.05), whereas exons and genes encoding tRNAs had the lowest level of methylation (*P* < 0.05). The distribution of NARDs across lncRNA genes and CpG islands (CGIs) had a higher level of methylation than regions 2 kbp up- and downstream (Fig. S2C).

Cytosine methylome landscapes based on WGBS

Overall, only c. 1.4% of cytosines in the genome of *S. japonica* were methylated based on WGBS. At such low levels of DNA methylation, a few sequencing errors can significantly impact the results. Thus, to minimize false positives and negatives in the present dataset, three biological replicates were sequenced for each life-cycle stage of *S. japonica* using WGBS (Fig. 1; Table S1). By calculating the percentage of cytosines sequenced at cytosine reference positions in the lambda genome, a final conversion rate of > 99.95% was identified for unmethylated cytosines. By combining nine individual methylome profiles, it was possible to obtain a high-quality, high-resolution kelp methylome with a final sequence yield of 92.4 Gbp covering 86% of all cytosines in the genome (Table S1). To minimize the number of false positives, the lower threshold for identification of methylated cytosines was set at four sequence reads per methylated cytosine (Table S1; Datasets S1–S9).

Most of the chromosomes had the highest MLgf in MG with additional hypermethylated chromosomes 4, 17 and regions on chromosomes 9, 12, making MG the most methylated life-cycle stage, followed by FG and SP (Figs 1, S4, S5B; Table S3). Although chromosomes 7 and 29 had a higher MLgf in FG compared to MG (Figs 1, S4, S5B), sex-determining genes were not found to be localized on these chromosomes (Lipinska et al., 2017) (Tables S3–S4). Approximately 57% of the methylated cytosines in the genome were methylated in a CHH context,
Fig. 3 Genomic elements overlapped with differentially methylated regions (DMRs) for all *Saccharina japonica* life-cycle stages (Sporophyte (SP), male (MG) and female (FG) gametophytes) and their relationship to gene expression. (a) Methylation differences of genomic elements (different colours) that overlapped with DMRs (see the Materials and Methods section) for SP, FG and MG. Ternary plots were drawn according to the methylation levels of genomic fragments (MLgf). (b) MLgf of genes for SP, FG and MG including 2-kbp up- (TSSUP) and down-stream (TESDOWN) flanking regions. All 18 733 genes were classified into four quantiles according to their expression levels (high, medium, low and none). The 2-kbp regions up- and downstream of genes were divided into 100-bp intervals for screening. Each gene was divided into 20 intervals (5% of total coding length per interval). Plots show the MLgf of each interval for both coding and noncoding regions up- and down-stream of genes. (c) Heat maps of genes and their promoter regions (TSSUP) with significant negative correlation ($P < 0.05$) between their MLgf and expression levels for pairwise comparisons between all life-cycle stages. Heat maps are coloured according to MLgf whereas gene expression levels are coloured according to log(RPKM+1). Abbreviations used in the figure: methy-level (MLgf), gene-exp (gene expression). (d) Significant negative correlation ($P < 0.05$) between the MLgf and the expression of genes (including their promoter regions) which moved from the sex determination region to autosomal loci. Upper row of panels: transcript read counts of four genes (SJ18945: glycosyl transferase, SJ15975: villin, SJ15874: GTPase activating protein, SJ13722: memo-like protein) from SP, FG and MG. Middle row of panels: real-time (RT)-PCR results of the same four genes using SP as control. Lower row of panels: Methylation levels of the four genes and their promoters. $n = 3$; Statistics were based on Duncan’s and Student’s $t$-tests. Error bars indicate $\pm$ SD.
whereas 19% and 24% were in CHG or CpG contexts, respectively (Fig. 2a; Table S4). Even though more cytosines were methylated in a CHH context, the median methylation level of methylated CpG (MLmc) is highest for all three contexts. (P < 0.05; Fig. S6). Interestingly, MLgfs in 10,000 bps bins of chromosomes show a significantly negative correlation with the CpG O/E (observed/excepted value), CpG ratio (nucleotide counts of all CpGs in bin/bin length) and CGI (CpG Island) ratio (nucleotide counts of all CGIs in bin/bin length), and a positive correlation with the CHH ratio (nucleotide counts of all CHHs in bin/bin length) (Figs S7–12), indicating that the CHH sites are preferentially methylated.

Among all life-cycle stages, SP had the lowest counts of methylated cytosines and the lowest MLwg for the whole genome and MLgc for all classes of components (CpG, CHG and CHH) (P < 0.05; Figs 2a,b, S6, S13). However, NMLgc fluctuated significantly across the genome (Fig. 2c), reflecting a mosaic methylation pattern where nonmethylated regions were interspersed with methylated regions. Notably, the most highly methylated genetic elements in the genome of S. japonica were loci encoding for circRNAs and IncRNAs (Figs 2c, S14; Tables S5–S15) with 81% and 66% of all elements methylated, respectively (Fig. 2f; Tables S4–S14). Coding genes and their regulatory parts (TSSUP and TESDOWN sites) were significantly less methylated than noncoding circRNA and IncRNA genes (P < 0.05; Fig. 2c). Exons had the lowest NMLgc (P < 0.05; Fig. 2d). Most of the exons were methylated in the CHH context with a concave methylation pattern. Introns, however, showed a convex methylation pattern (Figs 2e, S15). Convex methylation patterns also were found in the core regions of genes encoding for IncRNAs and circular RNAs (Fig. S15).

Predicted numbers of tRNA genes for the mitochondrion, chloroplast and nuclear genomes of S. japonica were 24, 29 and 757, respectively. On average, 29.1% of all tRNA genes were methylated (Fig. 2f; Table S14). In plants and algae, repetitive elements such as TEs are often most highly methylated (Cokus et al., 2008; Law & Jacobsen, 2010; Su et al., 2014; Takuno et al., 2016). However, in S. japonica, methylation of TEs does not seem to play a significant role (Figs 2c, S12–S13), as only 47.8% TEs were methylated, which is lower compared to genes (93.1–%

DNA methylation and gene expression

A greater number of highly expressed genes were identified in sporophytes than in gametophytes (RPKM > 10). Generally, there was a negative correlation between gene expression and MLgfs (Figs 3b, S19). In SP, 3884 genes and 3490 TSSUPs were significantly hypomethylated and more highly expressed than in FG, whereas 4132 genes and 4062 TSSUPs were significantly hypomethylated and more highly expressed than in MG (Duncan’s and Student’s t-test, P < 0.05; Fig. 3c; Tables S17–19). Only 51 genes in SP were hypermethylated with significantly lower expression than in either FM or MG (Student’s t-test, P < 0.05; Fig. 3c; Tables S18–20). In particular, some of the hypomethylated and highly expressed genes in SP with their respective GO term categories in comparison to FG and MG revealed some insights into processes that might be important for developing fully grown kelp sporophytes (Tables S21–22). For instance, enriched GO terms for this category in SP vs FG included structural molecule activity (GO:0005198, P < 0.01) and carbohydrate derivative binding (GO:0097367, P < 0.05). The same category for SP vs MG includes the GO term membrane protein complex (GO:0098796, P < 0.01). Interestingly, four genes that were reported to have moved from sex-determining chromosomal regions to auto chromosomes (Lipinska et al., 2017) were more highly expressed in SP and had significantly reduced MLgf in all life-cycle stages (Figs 3d, S20; Table S23; Methods S1, part 2). Other hypomethylated and strongly expressed genes in SP included a cellulose synthase, a mannuronate C5-epimerase (MC5E) and an iodoperoxidase (vIPO) (Fig. S21).
Although there was no significant correlation between MLgf and expression levels of IncRNAs and circRNAs genes (Figs S22–23) (Pearson correlation index \(< 0.01, P > 0.05\)), IncRNAs genes with expression levels in the top 10% had the lowest level of MLgf, and IncRNAs with MLgf in the top 10% had the lowest expression levels. The similar trends also were found in circRNAs. However, there was no significant correlation between MLgf and the expression of TEs (Fig. S24).

A comparison between MeDIP-seq, BS-PCR and WGBS

Evidence for DNA methylation in *S. japonica* although at a relatively low level is provided by MeDIP-seq and WGBS. Furthermore, there is a high consistency between results from both methods (Pearson correlation index \(> 0.5, P < 0.001\); Fig. S25). For example, both methods revealed independently that methylation preferentially took place at genes encoding IncRNAs and circRNAs. These results were confirmed by BS-PCR with subsequent cloning and Sanger sequencing. Ten genes (Figs 3d, S21) were assessed and good agreement was obtained between the WGBS and BS-PCR (Figs S26–35). Thus, the independent validation of DNA methylation based on three different methods suggests that the DNA of *S. japonica* indeed appears to be methylated albeit a low level.

Genome-wide heterozygosity and its influence on estimating the level of DNA methylation

The level of heterozygosity between the reference genome and the DNA obtained from the individuals used for the experiments can have an influence on estimating methylated cytosines. This is particularly critical if the overall level of methylation is low at a considerable level of heterozygosity between the reference genome and the genome of the individuals used for the experiments. To address this potential issue, Illumina WGS was performed for both male (MG) and female (FG) parents to determine differences in the level of heterozygosity. After mapping c. 50 Gbp of FG and MG to the reference genome, it was calculated that the overall level of heterozygosity was \(\leq 0.14\%\) (Table S24). Of all identified heterozygosity sites, C-T/G-A polymorphic sites (168 560 for FG and 118 421 for MG) overlapped with only 0.02%–0.04% of all mCs site for FG and MG samples (Table S24b), indicating that heterozygosity was not impacting \(> 0.04\%\) of all identified mCs in FG and MG.

DNA methyltransferases and demethylases in *S. japonica*

The DNMTs and demethylases (DNDMTs) are involved in the establishment of tissue and cell-type-specific methylation patterns during developmental processes in most multicellular organisms (Meissner et al., 2008; Kohli & Zhang, 2013; Satg et al., 2016; Iurlaro et al., 2017). As the overall DNA methylation was relatively low in *S. japonica*, a homology-based search was performed in the *S. japonica* genome to identify all possible DNMTs and DNDMTs (Figs S36–40). Although the DNA-methylase domain PF00145 is common in all DNMTs, effective DNA methylation requires the following additional domains, which occur only in the DNMT1/3/4/5/6 family: ADD (IPR025766), PWWP (PF00855), PHD (IPR001965) and a bromo-domain (PF00439) (Ponger & Li, 2005). All of the six DNMTs found in *S. japonica* had the highest similarity to members of the DNMT2 family without any of the domains required for efficient DNA methylation (Figs 4a,b, S38). There were no homologues found with significant similarity to any of the other DNMTs such as 3, 4, 5 or 6 (Fig. S37). Interestingly, ADD, PWBP, PHD and bromo-domains were found at other loci in the genome of *S. japonica*, but none of them were found to be DNMT2 (Figs S39–40).

In order to identify proteins involved in active DNA demethylation (DNDMTs), a search was carried out for RRM-fold domains (PF15628), DNA glycosylase domains (IPR011257), DME, DML, ROS, TET and MBD domains, all of which were reported previously to catalyse the removal of the 5mC base (Meissner et al., 2008; Kohli & Zhang, 2013; Su et al., 2014; Lipinska et al., 2017; Teng et al., 2017). Only one potential candidate was found in the genome, which included a DNA-glycosylase and the methyl-CpG-binding domain protein 2 (MBD2, PF01429) (Figs S41–42). The MBD2 domain was reported to catalyse the removal of a methyl group (Wu & Zhang, 2014).

To have six genes from the DNMT2 family encoded in the *S. japonica* genome with all other DNMTs missing may indicate that DNMT2 methylates DNA in the structural context of tRNAs (Kaiser et al., 2017). Thus, the MLgf of tRNA genes encoded in the mitochondrial, chloroplast and nuclear genomes of *S. japonica* were analyzed for all three life stages individually (Fig. S43; Table S25). Overall, the sporophyte had the lowest MLgf of tRNA genes for all three genomes. However, tRNA genes of the nuclear genome of the sporophyte were significantly more methylated than in the organellar genomes, which was not noticeable in the other life stages as tRNA genes in their organellar genomes were much more methylated than their counterparts in sporophytes. The ratio between unmethylated to methylated tRNA genes is highest for SP (Fig. S43). The MLgf of tRNA genes in the nucleus was positively correlated with the MLgf of nuclear genes only for MG. These results suggest possible DNMT2-mediated DNA methylation, with RNAs as natural substrates of DNMT2 acting as guides as shown previously (Kaiser et al., 2017).

Discussion

Differential DNA methylation is the hallmark of cell-type-specific development and life-cycle regulation in multicellular organisms (Law & Jacobsen, 2010; Takuno et al., 2016). For most multicellular organisms, DNMT2 is not considered to be the enzyme for DNA methylation as it primarily catalyzes transfer (t) RNA methylation and only has very weak DNA methyltransferase activity (Kunert et al., 2003; Goll et al., 2006; Shanmugam et al., 2014; Francesca et al., 2015; Kaiser et al., 2017). However, recent data (Kaiser et al., 2017) have provided the first *in vitro* evidence that DNMT2 can efficiently methylate DNA when DNA fragments are presented as covalent DNA-RNA hybrids in
the structural context of tRNAs. Thus, methylated tRNAs can serve as substrates for methylating DNA presented as covalent hybrids with RNA. Although the data herein do not provide direct evidence for this process to be responsible for DNA methylation in vivo in *Saccharina japonica*, it is intriguing to see that there are significantly more methylated tRNA genes in both
gametophytes which might serve as a substrate to methylate genes via DNMT2. In male gametophytes (MG), a significantly positive correlation between the methylation level of a specific genomic fragment (MLgf) of tRNA genes and both the expression level of DNMT2s (Fig. 4c,d) and the MLgf of genes corroborates this assumption (Fig. S43). However, if DNMT2 does indeed methylate the DNA of *S. japonica in vivo*, it is being done inefficiently as the overall DNA methylation level (of methylated cytosines (MLmc) and the whole genome (MLwg)) is still very low even though six different DNMT2s are encoded in the genome and expressed. DNMT2 appears to have evolved from a DNA methyltransferase precursor, which is considered an evolutionary ‘relict’ in terms of biochemical catalysis (Jurkowski & Jeltsch, 2011; Raddatz *et al.*, 2013) and therefore might explain its inefficiency compared to other DNMT enzymes such as 1 and 3. Although it is possible to have missed DNMT5s because of gaps in the genome assembly, the fact that multiple DNMT2 genes are encoded in the *S. japonica* genome, and the positive correlation of their expression with an elevated level of C5 DNA methylation (Pearson correlation test, *P* < 0.05) makes DNMT2s likely candidates for C5 DNA methylation. Support for the role of DNMT2 in C5 DNA methylation comes from reverse-genetics studies with *Drosophila melanogaster* where knockdown and over-expression of DNMT2 resulted in lower and higher levels of 5 methylcytosine in embryos, respectively (Kunert *et al.*, 2003). Likewise, human DNMT2 was shown to have a residual DNA C5 methyltransferase activity in a particular sequence context (Herrmann *et al.*, 2003).

Although the overall DNA methylation level in *S. japonica* is lower than observed for most multicellular organisms with complex life cycles (Takuno *et al.*, 2016), DNA methylation still appears to contribute to the regulation of gene expression and particularly noncoding regulatory RNAs. Furthermore, significant differences between life-cycle stages suggest that differential DNA methylation is involved in their formation, which is similar to other multicellular organisms although most of them have much higher and more dynamic levels of DNA methylation (Lopez *et al.*, 2015; Willing *et al.*, 2015; Gaunt *et al.*, 2016; Kawakatsu *et al.*, 2016; Mayasich *et al.*, 2016; Panikar *et al.*, 2017).

Another striking difference is that genome-wide patterns and mechanisms of cytosine methylation in *S. japonica* are neither more similar to plants or animals, nor similar to other stramenopile species for which single-base methylome data are available such as *Phaeodactylum tricornutum* (Veluchamy *et al.*, 2013). In many photosynthetic organisms including plants, green algae and diatoms, transposable elements (TEs) are the main target of cytosine methylation whereas animals methylate genes and TEs more equally (Cokus *et al.*, 2008; Law & Jacobsen, 2010; Takuno *et al.*, 2016; Tirichine *et al.*, 2017). However, in *S. japonica*, the percentage of methylated TEs is < 48% (Table S5), which is lower than the percentage of methylated genes (93% Table S6). Surprisingly, the highest methylated elements in *S. japonica* were found to be genes encoding circular and long noncoding RNAs (circRNAs and lncRNAs) (*P* < 0.05 in female gametophytes (FG) and MG; Figs 2c, S14). Significant differences in methylation of both groups of genes encoding non-coding RNAs between sporophytes (SP), FG and MG indicate that they appear to have a role in regulating life-cycle stages of *S. japonica*, which, to the best of our knowledge, has not been observed before in any organism (Figs S22–23). Further differences to plants and animals have been revealed by comparing mechanisms of DNA methylation. Animals mainly methylate cytosines in the Cpg context, whereas plants methylate CHG and CHH sites and many algae methylate in the Cpg context (Cokus *et al.*, 2008; Lopez *et al.*, 2015). *Saccharina japonica* methylates cytosines mainly in the CHH context with much lower but almost equal methylation of CHG and Cpg sites.

The fact that a significant number of genes is methylated in *S. japonica*, and that the level of their methylation is negatively correlated with gene expression is in common with many plants and animals. Hence, there is evidence for methylation-regulated gene expression. Furthermore, methylated genes in MG and FG were negatively correlated with gene expression (Fig. 3b). GO enrichment for these genes revealed that many were involved in transport, membrane structure and function (Tables S16, S17), which suggests that these processes are under epigenetic control in MG; and that they are downregulated in comparison to SP and FG. Interestingly, methylated genes in SP and FG were enriched, for example, for small molecule metabolic process and organo-nitrogen compound metabolic processes. Hypomethylated and highly expressed genes involved in morphogenesis (e.g. MC5E, cellulose synthase) and halogene metabolism (e.g. vIPO) were found in SP (Fig. S21), which suggests that cytosine methylation contributes to regulating fundamental processes underpinning cell differentiation, growth and stress response (Peat *et al.*, 1996; Colin *et al.*, 2005; Ye *et al.*, 2015; Inoue *et al.*, 2016). Genes of the sex-determination system do not show a negative correlation between their level of methylation and gene expression (Fig. S20), suggesting that 5mC DNA methylation might not significantly contribute to their regulation. Interestingly, the four genes that are characterized by different protein domains (MEMO, glycosyltransferase, RING-type zinc finger, Rab-GT-Pase-TBC) and which moved from the sex-determination locus in the genome to the autosomal loci were hypomethylated and highly expressed in SP (Lipinska *et al.*, 2017) (Fig. S20).

The first single-base methylome study of a brown alga has revealed that cytosine methylation might be mediated by DNMT2. Although the overall level of DNA methylation is low, it appears to play a significant role in these multicellular algae mainly for controlling regulatory elements (noncoding RNAs) and the expression of genes. Significant differences in overall DNA methylation and methylation-mediated gene expression between life-cycle stages suggests that cytosine methylation is involved in regulating growth, tissue differentiation, and several different physiological responses. However, it remains to be seen if DNMT2 indeed is responsible for DNA methylation in *S. japonica*.

**Acknowledgements**

We are grateful to Jim Mitchell for general advice. This work was supported by the National key research and development
program of China (2018YFD0900703, 2018YFD0901503-8, 2016YFC1402102), National Natural Science Foundation of China (41676145); Shandong key research and Development Plan (2018GKY115010); Qingdao Municipal Science and Technology plan project (17-1-1-96-jch); Special Scientific Research Funds for Central Non-Profit Institutes, Yellow Sea Fisheries Research Institute, Chinese Academy of Fishery Sciences (20063022019006, 20063022016001); Youth Talent Program Supported by Laboratory for Marine Fisheries Science and Food Production Processes, Pilot National Laboratory for Marine Science and Technology (Qingdao) (2018-MFS-01); China Agriculture Research System (CARS-50); Marine S&T Fund of Shandong Province for Pilot National Laboratory for Marine Science and Technology (Qingdao) (no. 2018SDKJ0406-3); Financial Fund of the Ministry of Agriculture and Rural Affairs, China (NFZX2018); Taishan Scholars Funding; Talent Projects of Distinguished Scientific Scholars in Agriculture. TM acknowledges support by the Natural Environment Research Council, UK (NE/R000883/1) and the School of Environmental Sciences at the University of East Anglia, UK. LT acknowledges support from Region Pays de la Loire and Nantes Métropole.

Author contributions

NHY designed the project with contributions from TM, LT and XF; WTH, LHT, PJ, XWZ, DX, CHW, CL and YTW performed the research; XF, NHY, WTH, TM, LT, and XF and LHT analyzed the data; TM wrote the manuscript with contributions from XF, NHY, and LT; MP, MJSK and XL revised and edited the manuscript. All authors read and approved the manuscript before submission. The authors declare no conflict of interest.

ORCID

Xiao Fan https://orcid.org/0000-0002-7093-8868
Chang Li https://orcid.org/0000-0002-7823-3467
Xin Lin https://orcid.org/0000-0001-8323-5056
Thomas Mock https://orcid.org/0000-0001-9604-0362
Matteo Pellegrini https://orcid.org/0000-0001-9355-9564
Linhong Teng https://orcid.org/0000-0002-2245-9563
Xiaowen Zhang https://orcid.org/0000-0001-5338-1837
Naihao Ye https://orcid.org/0000-0002-8294-1044

References

Charrier B, Le BA, De RB. 2012. Plant Proteus: brown algal morphological plasticity and underlying developmental mechanisms. Trends in Plant Science 17: 468–477.
Cock JM, Godfrey O, Macaime N, Peters AF, Coelho SM. 2014. Evolution and regulation of complex life cycles: a brown algal perspective. Current Opinion in Plant Biology 17: 1–6.
Cock JM, Sterck L, Rouze P, Scornet D, Allen AE, Amoutzias G, Anthouard V, Attiguenave F, Auty JM, Badger JH. 2010. The Ectocarpus genome and the evolution of multicellularity in brown algae. Nature 465: 617.
Cokus SJ, Feng S, Zhang X, Chen Z, Merriman B, Haoudienschl CD, Pradhan S, Nelson SF, Pellegrini M, Jacobsen SE. 2008. Shotgun bisulfit sequencing of the Arabidopsis genome reveals DNA methylation patterning. Nature 452: 215–219.
Colin C, Leblanc C, Michel G, Wagner E, Leizewagner E, Van DA, Potin P. 2005. Vanadium-dependent iodoperoxidases in Laminaria digitata, a novel biochemical function diverging from brown algal bromoperoxidases. Journal of Biological Inorganic Chemistry 10: 156.
Dochow CD, Liu WS. 2018. DNA methylation patterns in peripheral blood mononuclear cells from Holstein cattle with variable milk yield. BMC Genomics 19:744.
Dinger ME, Amaral PP, Mercer TR, Pang KC, Bruce SJ, Gardiner BB, Askarian-Amiri ME, Ru K, Solda G, Simons C et al. 2008. Long noncoding RNAs in mouse embryonic stem cell pluripotency and differentiation. Genome Research 18: 1433.
Edgar RC. 2004. MUSCLE: multiple sequence alignment with high accuracy and high throughput. Nucleic Acids Research 32: 1792–1797.
Feng S, Cokus SJ, Zhang X, Chen P, Bostick M, Goll MG, Hetzel J, Jain J, Strauss SH, Halpern ME. 2010. Conservation and divergence of methylation patterning in plants and animals. Proceedings of the National Academy of Sciences, USA 107: 8689–8694.
Francesca T, Friederike H, Nader A, Sebastian B, Oliver P, Giuseppina F, Sonja R, Reinhard L, Georg S, Hermann-Josef G. 2015. The RNA methyltransferase Dnmt2 is required for accurate polyploid synthesis during haematopoiesis. EMBO Journal 34: 2350–2362.
Gaunt TR, Shihab HA, Heman G, Min JL, Woodward G, Lyttleton O, Zheng J, Duggirala A, Meadale WL, Ho K. 2016. Systematic identification of genetic influences on methylation across the human life course. Genome Biology 17: 61.
Goll MG, Kirpekar F, Maggert KA, Yoder JA, Hsieh CL, Zhang X, Golic KG, Jakobsen SE, Bestor TH. 2006. Methylation of tRNA by the DNA methyltransferase homolog Dnmt2. Science 311: 395–398.
Greaves IK, Groszmann M, Ying H, Taylor JM, Peacock WJ, Dennis ES. 2012. Trans chromosomal methylation in Arabidopsis hybrids. Proceedings of the National Academy of Sciences, USA 109: 3570–3575.
Hermann A, Schmitt S, Jeltsch A. 2003. The human Dnmt2 has residual DNA-(Cytosine-C5) methyltransferase activity. Journal of Biological Chemistry 278: 31717–31721.
Hon GC, Rajagopal N, Shen Y, Mccleary DF, Yue F, Dang MD, Ren B. 2013. Epigenetic memory at embryonic enhancers identified in DNA methylation maps from adult mouse tissues. Nature Genetics 45: 1198–1206.
Inoue A, Sato A, Morishita M, Tokunaga Y, Miyakawa T, Tanokura M, Ojima T. 2016. Functional heterologous expression and characterization of mannuronan C5-epimerase from the brown alga Saccharina japonica. Algal Research 16: 282–291.
Iurlaro M, Von MF, Reik W. 2017. DNA methylation homeostasis in human and mouse development. Current Opinion in Genetics & Development 43: 101–109.
Ivica L, Peer B. 2016. Interactive tree of life (iTOL) v3: an online tool for the display and annotation of phylogenetic and other trees. Nucleic Acids Research 44: W242–W245.
Jurkowski TP, Jeltsch A. 2011. On the evolutionary origin of eukaryotic DNA methyltransferases and Dnmt2. PLoS ONE 6: e28104.
Kaiser S, Jurkowski TP, Kellner S, Schneider D, Jeltsch A, Helm M. 2017. The RNA methyltransferase Dnmt2 methylates DNA in the structural context of a tRNA. RNA Biology 14: 1241–1251.
Kawakatsu T, Valdes M, Breakfield N, Schmitz RJ, Nery JR, Urieh MA, Han X, Lister R, Benfey PN. 2016. Unique cell-type-specific patterns of DNA methylation in the root meristem. Nature Plants 2: 16058.
Kohli RM, Zhang Y. 2013. TET enzymes, TDG and the dynamics of DNA demethylation. Nature 502: 472–479.
Krueger F, Andrews SR. 2011. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. Bioinformatics 27: 1571–1572.
Kundaje A, Meuleman W, Ernst J, Bilenky M, Yen A, Heravimoussavi A, Kheradpour P, Zhang Z, Wang J. 2015. Integrative analysis of 111 reference human epigenomes. Nature 518: 317–330.
Kunert N, Marhold J, Stanke J, Stach D, Lyko F. 2003. A Dnmt2-like protein mediates DNA methylation in Drosophila. Development 130: 5083.
Langmead B, Salzberg SL. 2012. Fast gapped-read alignment with Bowtie 2. Nature Methods 9: 357–359.
Law JA, Jacobsen SE. 2010. Establishing, maintaining and modifying DNA methylation patterns in plants and animals. *Nature Reviews Genetics* 11: 204–220.

Lee T, Zhai J, Meyers BC. 2010. Conservation and divergence in eukaryotic DNA methylation. *Proceedings of the National Academy of Sciences, USA* 107: 9027–9032.

Li H, Durbin R. 2010. Fast and accurate long-read alignment with Burrows-Wheeler transform. *Bioinformatics* 26: 589–595.

Li H, Handsaker B, Wysoker A, Fennell T, Ruan J, Homer N. 2009. The sequence alignment/Map format and SAMtools. *Bioinformatics* 25: 2078–2079.

Lipinska AP, Toda NRT, Heesch S, Peters AF, Cock JM, Coelho SM. 2017. The non-coding Air RNA is required for establishment, maintenance and modification of DNA methylation in the *Drosophila* embryo. *Nature* 540: 238.

Su C, Wang C, He L, Yang C, Wang Y. 2014. Shotgun bisulite sequencing of the *Betula platyphylla* genome reveals the tree’s DNA methylation patterning. *International Journal of Molecular Sciences* 15: 22874–22886.

Takuno S, Ran JH, Gaut BS. 2016. Evolutionary patterns of genomic DNA methylation vary across land plants. *Nature Plants* 2: 15222.

Lister R, Mukamel EA, Nery JR, Urich M, Puddifoot CA, Johnson ND, Lucero J, Fan X, Xu D, Zhang X, Mock T, Ye N. 2017. Identification of genes under positive selection reveals differences in evolutionary adaptation across brown-algal species. *Frontiers in Plant Science* 8: 1429.

Evans J, Liu T, Meyer CA, Eckhoute J, Johnson DS, Bernstein BE, Nusbaum C, Myers RM, Brown M, Li W. 2008. Integrative analysis of DNA methylation and histone modification marks in human cell types. *Cell* 132: 909–923.

Ye N, Zhang X, Xiao F, Zhao L, Ma C, Zeng L. 2013. Insights into the role of DNA methylation in diatoms by genome-wide profiling in *Phaeodactylum tricornutum*. *Nature Communications* 4: 2091.

Wang Y, Xu D, Fan X, Zhang X, Wang W, Mao Y, Mou S, Cao S. 2013. Variation of photosynthetic performance, nutrient uptake, and elemental composition of different generations and different thallus parts of *Saccharina japonica*. *Journal of Applied Phycology* 25: 631–637.

Willing EM, Rawat V, Mandá I, Fritsch J, James GV, Nordström KJ, Becker C, Warthmann N, Chica C, Szarzynska B. 2015. Genome expansion of *Arabidopsis alpina* linked with retrotroperonization and reduced symmetric DNA methylation. *Nature Plants* 1: 14023.

Wu H, Zhang Y. 2014. Reversing DNA methylation: mechanisms, genomics, and biological functions. *Cell* 156: 45.

Xiang H, Zhu J, Chen Q, Dai F, Li X, Li M, Zhang H, Zhang G, Li D, Dong Y. 2010. Single base-resolution methylation of the silkworm reveals a sparse epigenetic map. *Nature* 467: 456–459.

Ye N, Zhang X, Xiao M, Zhao L, Zhang Y, Li D, Zeng L. 2015. Construction of a high density SNP linkage map of *Saccharina japonica* by sequencing *Taq 1* site associated DNA and mapping of a sex determining locus. *BMC Genomics* 16: 1–11.

Zhang Y, Liu T, Meyer CA, Eckhoute J, Johnson DS, Bernstein BE, Nusbaum C, Myers RM, Brown M, Li W. 2008. Model-based analysis of ChIP-Seq (MACS). *Genome Biology* 9: R137.

Ziller MJ, Gu H, Müller F, Donaghey J, Tsai I-T, Kohlbacher O, De Jager PL, Rosen ED, Bennett DA, Bernstein BE. 2013. Charting a dynamic DNA methylation landscape of the human genome. *Nature* 500: 477.

Additional Supporting Information may be found online in the Supporting Information section at the end of the article.

**Fig. S1** MeDIP-seq libraries.

**Fig. S2** Methylation profiles of MG based on MeDIP-seq.

**Fig. S3** Profiles of SP, MG and FG methylomes and transcriptomes of Chromosome 0.
Fig. S4 Chromosome-specific profiles for SP, MG and FG methylomes and transcriptomes.

Fig. S5 GC characteristics and methylation levels of chromosomes.

Fig. S6 Comparison between methylation levels of individual cytosines.

Fig. S7 Correlation between CpGo/e and the methylation level.

Fig. S8 Correlation between C+G percentage and the methylation level.

Fig. S9 Correlation between CpG percentage and the methylation level.

Fig. S10 Correlation between CHG percentage and the methylation level.

Fig. S11 Correlation between CHH percentage and the methylation level.

Fig. S12 Correlation between CpG islands ratio and the methylation level.

Fig. S13 Difference in methylation levels for different types of genomic elements between life-cycle stages.

Fig. S14 Normalized methylation level for genomic elements.

Fig. S15 Methylation for all methylation contexts of IncRNA genes and cirRNA genes.

Fig. S16 Most enriched (top 20) KEGG pathways for methylated genes in FG.

Fig. S17 Most enriched (top 20) KEGG pathways for methylated genes in MG.

Fig. S18 Most enriched (top 20) KEGG pathways for methylated genes in SP.

Fig. S19 Genome-wide gene expression profiling and DNA methylation profiles.

Fig. S20 Methylation profiles of the sex-determining region.

Fig. S21 Base-specific methylation and transcriptional coverage of selected metabolic genes.

Fig. S22 Correlation between MLgf and gene expression for IncRNA genes.

Fig. S23 Correlation between MLgf and gene expression for circRNA genes.

Fig. S24 The correlation ship between MLgf of TEs and their expression level.

Fig. S25 Consistency assessment of MeDIP-seq vs WGBS data.

Fig. S26 BS-PCR for SJ18945.

Fig. S27 BS-PCR for SJ15797.

Fig. S28 BS-PCR for SJ15874.

Fig. S29 BS-PCR for SJ13722.

Fig. S30 BS-PCR for SJ21435.

Fig. S31 BS-PCR for SJ22030.

Fig. S32 BS-PCR for SJ15673.

Fig. S33 BS-PCR for SJ19510.

Fig. S34 BS-PCR for SJ22283.

Fig. S35 BS-PCR for SJ20815.

Fig. S36 Phylogeny of methyltransferases based on Pfam PF00145.

Fig. S37 Domain architecture of DNMT4/5/6.

Fig. S38 DNA methylase domain PF000145 and their expression profile.

Fig. S39 Gene structure, expression patterns and chromosomal locations of MBT, PWWD, DNMT1, Zinc finger PHD-type.

Fig. S40 Gene structure, expression patterns and chromosomal locations of ADD, Pre_SET, SET, SNF, helicase and AP2/ERF.

Fig. S41 Gene structure and phylogeny of demethyltransferases.

Fig. S42 Gene structure of MBD domains and their phylogeny.

Fig. S43 Methylation levels of tRNA genes.

Methods S1 1, Constructing the pseudo-chromosomes using a genetic linkage map; 2, methylation profiles of sex-determining regions and their gene expression; 3, DNA methyltransferases evolution and their putative function; 4, other methods in this article.

Table S1 Summary of methylation and RNA sequencing data and mapping results.

Table S2 Peak-specific information.
Table S3 Chromosome-specific gene methylation and expression.

Table S4 Features of methylated sites for all methylation contexts and life-cycle stages.

Table S5 Features of TE methylation.

Table S6 Features of gene methylation.

Table S7 Features of exon methylation.

Table S8 Features of intro methylation.

Table S9 Features of IncRNA gene methylation.

Table S10 Features of circle RNA gene methylation.

Table S11 Features of TSSUP methylation.

Table S12 Features of TESDOWN methylation.

Table S13 Features of CGI methylation.

Table S14 Features of tRNA gene methylation.

Table S15 Information of DMRs.

Table S16 GO enrichment of genes located in DMRs.

Table S17 KEGG enrichment of the genes located in DMRs.

Table S18 Expression and statistical tests of all genes and life-cycle stages.

Table S19 Methylation level and statistical tests of all TSSUPs and life-cycle stages.

Table S20 Methylation level and statistical tests of all genes and life-cycle stages.

Table S21 KEGG enrichment of genes and TSSUPs for Fig. 3(c).

Table S22 GO enrichment of genes and TSSUPs for Fig. 3(c).

Table S23 Methylation and expression of genes located in the sex-determining-region for all life-cycle stages.

Table S24 Heterozygosity between FG and MG vs the reference genome.

Table S25 Methylation levels of tRNA genes.

Please note: Wiley Blackwell are not responsible for the content or functionality of any Supporting Information supplied by the authors. Any queries (other than missing material) should be directed to the New Phytologist Central Office.