Role of gene body methylation in coral acclimatization and adaptation.

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One-sentence summary:

Genome-wide shifts in gene body methylation predict gene expression and fitness during

acclimatization but do not contribute to epigenetic divergence between populations.

Abstract

Gene body methylation (GBM) has been hypothesized to modulate responses to environmental

change, including transgenerational plasticity, but the evidence thus far has been lacking. Here

we show that coral fragments reciprocally transplanted between two distant reefs respond with

genome-wide increase or decrease in GBM disparity among genes. Surprisingly, this simple

genome-wide adjustment predicted broad-scale gene expression changes and fragments' fitness

in the new environment. This supports GBM's role in acclimatization, which may consist in

modulating the expression balance between environmentally-responsive and housekeeping

genes. At the same time, constitutive differences in GBM between populations did not align

with plastic GBM changes upon transplantation and were mostly observed among F_{ST} outliers,

indicating that they arose through genetic divergence rather than through transgenerational

inheritance of acquired GBM states.

GBM is a taxonomically widespread epigenetic modification the function of which

remains unclear (*1*, *2*). GBM is bimodally distributed among genes: it is high in ubiquitously

expressed housekeeping genes and low in inducible genes (*2*, *3*). Only the detrimental effect of

GBM is well understood: GBM causes hypermutability in protein-coding regions (*4*). Indeed, in

humans, GBM is the primary driver of deleterious parent-age-related mutations (*5*). To merit

pervasive evolutionary conservation, the fundamental biological function of GBM must be

important enough to outweigh this risk (*1*). Among putative cellular functions of GBM,

suppression of intragenic transcription initiation is currently the most studied (*6*, *7*) but still

remains controversial (*8*). The ecological genomic literature has long expected that GBM (as well

as other epigenetic marks) respond to the environment and assist acclimatization by modulating

gene expression, possibly across generations (*9–11*). Still, thus far the only well-documented

case of GBM responding to the environment and resulting in phenotypic change is caste

determination in social insects depending on diet (*12*). Beyond this highly taxon-specific

example, GBM response to the environment has been reported in a recent paper on coral

acclimatization to acidic conditions (*13*); notably, GBM change did not specifically affect genes

expected to be involved in such acclimatization. One way to address the role of GBM in

adaptation would be to examine differences between populations, as has been recently done in

plants (*14*, *15*); however, this interpretation requires disentangling the effect of environment

from the effect of genetic divergence among populations (*16*). As for transgenerational

plasticity, for GBM to be involved in this process it must be both responsive to the environment

and heritable. This somewhat contradictory combination of properties has not yet been

demonstrated for GBM in any study system.

Here, we used a reciprocal transplantation framework (*17*) to test for the roles of GBM,

gene expression, and genetics in local adaptation and acclimatization in a reef-building coral

Acropora millepora. In corals, the possibility of clonal replication makes disentangling the

effects of genotype and environment very straightforward, solving a major problem of

ecological epigenetics (*16*). We compared corals from two reefs, Orpheus Island in the central

sector of the Great Barrier Reef (GBR) and Keppel Island in the south of GBR (Fig. 1 A). These

reefs are notably different in temperature (Fig. 1 B) as well as several other abiotic parameters

(*18*) and host slightly genetically divergent *A. millepora* populations (*19*, *20*). Fifteen coral

colonies from each reef were halved and reciprocally transplanted between the sites for three

winter months (Fig. 1 A, B). In this way, each individual genotype was simultaneously exposed to

two distinct reef conditions. After three months in the field, tissue samples were collected from

each fragment and assayed for genome-wide gene expression using TagSeq (*21*) and for DNA

methylation using MBD-seq (*22*). These data were analyzed in the context of several fitness

proxies and genetic distances between individuals (based on genetic polymorphisms detected in

the MBD-seq data). All experimental corals were dominated by *Symbiodinium* clade C, which

was established by analyzing *Symbiodinium*-matching MBD-seq reads.

For both Keppel- and Orpheus-origin corals fitness proxies were higher at Keppel,

possibly due to higher concentration of inorganic nutrients there (*18*, *19*). For Keppel-origin

corals the difference in performance was predominantly in weight gain (i.e., skeletal growth)

while for Orpheus-origin corals the difference was mostly in internal stores (lipids, proteins and

carbohydrates, Fig. 1 C). This indicates that (i) Keppel and Orpheus environments are indeed

different from corals' perspective, (ii) both populations are capable of plastic responses to these

differences, and (iii) there is also constitutive divergence in physiology between populations.

<u>Figure 1</u>. Overview of the experiment. A: Design of reciprocal transplantation. Inset – *Acropora millepora*.

B: Temperature profiles at the transplantation sites, historical (thick lines) and during the year of

experiment (thin lines). C: Principal component analysis of the four fitness proxies: weight gain, lipid,

protein and carbohydrate content. In the sample names, the first letter is the origin location and the

second letter is transplantation location. D: Scatter plot of the measure of depletion of CpG dinucleotides

(ratio of observed vs. expected number of CpGs, $CpG_{O/E}$) against MBD-score ($\log_2$ of fold-difference in

read counts between MBD-captured and flow-through fraction) for each gene. The MBD-score threshold

defining the two GBM classes is indicated. E and F: Principal coordinate analysis of gene expression (E)

and gene body methylation (F). Lines connect fragments of the same original colony; the color scheme

follows the legend on panel C. For GE, PC1 corresponds to the colony's origin and PC2 – to the

transplantation site; no such partitioning is visible for GBM. G: Percentage of variation in GE and GBM

explained by colony identity (i.e., by similarity among clonal fragments, or broad-sense heritability),

colony's origin and transplantation site. *** $p < 0.001$; ** $p < 0.01$; * $p < 0.05$.

To determine the characteristic level of GBM for each gene we used MBD-seq results for

12 samples (three per each experimental group, KK, KO, OK, and OO) for which both MBD-

captured and flow-through fraction were sequenced. Logarithm with the base 2 of each gene's

abundance in captured vs. flow-through fractions ("MBD-score" (*22*)) exhibited the expected

bimodal distribution (Fig. S1 A) and showed strong correlation (Fig. 1 D) with the established

proxy of historical methylation level, the ratio of observed to expected numbers of CpG

dinucleotides ("CpG$_{O/E}$" (*3*)). In addition to this genome-wide validation of our methylation data,

we validated GBM results by bisulfite-sequencing 13 amplicons representing different MBD-

score classes (Fig. S1 B, C). Genes were assigned to either low-methylated class (class 1) or high-

methylated class (class 2) based on the MBD-score threshold indicated on Fig. 1 D. Relative

quantification results based only on MBD-captured read data were nearly identical to those

obtained when using both captured and flow-through fractions (Fig. S2). For the remaining 32

samples, we sequenced only the MBD-captured fraction.

Multivariate analysis of gene expression (GE, Fig. 1 E) and GBM (Fig. 1 F) revealed that

GBM is more consistent across fragments of the same original colony than GE, resulting in the

estimate of broad-sense heritability of 0.79 compared to 0.64 for GE (Fig. 1G, Fig. S3 and S4). In

addition, significant effects of origin and transplantation site were observed for GE and GBM,

both more pronounced for GE (Fig. 1 E and G).

Surprisingly, the change in GBM in response to transplantation from Orpheus to Keppel

consisted mainly in genome-wide reduction of disparity between high- and low-methylated

genes: highly-methylated genes became less methylated and low methylated became more

methylated (Fig. 2 A). This change was mirrored by less pronounced but clearly reciprocal GBM

adjustment in Keppel corals transplanted to Orpheus (Fig. 2 B, D) and recapitulated the

difference between fragments planted in their native environment (Fig. 2 C). Both exons and

introns underwent this change (Fig. S5). Absolute changes in methylation included both positive

and negative shifts (Fig. S6 A, B), indicating that the change in relative GBM levels among genes

is not due to biased genome-wide methylation increase or decrease. Methylation levels of

intergenic regions and repeated elements remained relatively constant compared to genic

regions (Fig. S6 C, D).

GE response was also reciprocal between transplantation directions but demonstrated

much broader range of fold-changes compared to GBM response (Fig. 2 D). Remarkably, gene

expression changes correlated with low-and high-methylated gene classes and were in the

opposite direction relative to GBM changes (Fig. 2 F, G). Although at the gene level the

correlation between GE and GBM was weak (but still highly significant, Fig. S7), it was very

strong when comparing broader functional groups of genes (Gene Ontology categories, Fig. 2 H

and Fig. S8).

To characterize GBM disparity among genes in each sample, we computed GBM class

difference ("GBMcd") as difference in mean $\log_2$(GBM) between high- and low-methylated gene

classes (Fig. 2 I). This measure aligned nearly perfectly with the first principal component of

GBM variation for the whole experiment, explaining 33% of total variation (Fig. 1 F and Fig. 2 J,

which show the same PCA ordination but are colored differently). For comparison, the next

principal component explains only 8% of GBM variation.

<u>Figure 2</u>. Changes in gene body methylation (GBM) and gene expression (GE) in response to

transplantation ("away" relative to "home"). On all panels, GE and GBM measures are $\log_2$ fold-changes.

A though C: Differences in GBM in fragments transplanted from Orpheus to Keppel (A), from Keppel to

Orpheus (B), and fragments planted in their home environment (C) plotted against MBD-score for each

gene. Red and blue curves near the x-axis are density plots of significantly up- and down-regulated genes

(there were none for the K→O experiment). D and E: Correlation between GBM (D) and GE (E) changes in

reciprocally transplanted fragments. F and G: Change in GE in fragments transplanted from Orpheus to

Keppel (F) and from Keppel to Orpheus (G) plotted against MBD-score for each gene (compare to panels

A and B). H: Plot of delta-ranks for 20 Gene Ontology categories that were significantly enriched with

either up-regulated genes (positive delta-rank) or down-regulated genes (negative delta-rank) in both GE

and GBM analyses. I: Calculation of GBM class difference (GBMcd), a proxy of genome-wide GBM

disparity among genes. Each panel shows two density plots of GBM levels for the two GBM gene classes

(see Fig. 1 D), with the mean for each class indicated by a dashed line. The top panel shows the sample

with the lowest GBMcd, bottom panel – the sample with the highest GBMcd. J: Principal coordinate

biplot (the same ordination as on Fig. 1 F) colored according to GBMcd. The purple vector shows the

direction of correlation of GBMcd with the ordination. All correlations with reported R^2 are significant at

the $p < 0.0001$ level.

Our data do not support transgenerational inheritance of acquired GBM changes. If

plastic changes in GBM were transmitted across generations, one would expect constitutive

GBM differences between Orpheus- and Keppel-origin corals to parallel the GBM response

induced upon transplantation. Instead, constitutive differences between Orpheus and Keppel

(differences between all Orpheus-origin and all Keppel-origin fragments, irrespective of

transplantation site) were in the opposite direction compared to plastic changes (compare Fig. 3

A and Fig. 2 C), did not predominantly affect genes at the opposite ends of MBD-score spectrum

(compare density curves on Fig. 2 A and Fig. 3 A), and were positively rather than negatively

associated with differences in GE (Fig. 3 B). In addition, genes showing significant constitutive

differences in GBM also showed elevated F_{ST} (Fig. 3 C) unlike genes undergoing significant plastic

changes (Fig. S9). This indicates that the mechanism giving rise to constitutive GBM changes

between populations is unrelated to the mechanism of plasticity and is most likely genetic

divergence.

<u>Figure 3</u>. Constitutive log$_2$ fold-differences in gene body methylation (GBM) and gene expression (GE)

between populations (Orpheus relative to Keppel). A: Difference in GBM plotted against MBD-score for

each gene. Red and blue curves by the x-axis are density plots of significantly up- and down-regulated

genes. B: Correlation between GBM and GE differences. Red points are genes significantly different in

GBM at 10% FDR level. This correlation is positive, while correlations between plastic GBM and GE

changes are negative (Fig. 2 H and Fig. S7). E: Density plot of between-population F_{ST} for genes showing

significant constitutive difference in GBM (red line) compared to 500 randomly chosen non-significant

genes (black line). Dashed vertical lines mark the mean of each dataset. All correlations with reported R^2

are significant at p<0.0001 level.

To see whether plastic GBM changes were related to acclimatization (rather than, for

example, overall stress in response to transplantation) we used Differential Analysis of Principal

Components (DAPC, Fig. 4 A-C) to test whether GBM similarity to the local population (Fig. 4 D)

predicted the transplanted fragment's fitness in the new environment. The resulting "GBM

similarity to locals" was highly correlated with the first principal component of measured fitness

proxies, the most strongly with the weight gain (Fig. 4 E). This correlation held both before and

after (as shown on Fig 4 E) controlling for overall higher fitness of Orpheus corals transplanted

to Keppel. This result indicates that a coral's GBM profile, despite being highly variable among

individuals (Fig 1 F, G), also reflects adaptation and acclimatization to local environmental

conditions. Interestingly, the same analysis performed for GE and genotypes (GT) revealed no

relationships with fitness (Fig. 4 E). One explanation of this could be the difference in time scales

on which the effects of GE and GT on fitness would be detectable: compared to the three-

month length of our experiment, GE might be linked to fitness on shorter time scales, and GT –

on longer time scales.

Figure 4. Relationship between gene body methylation (GBM), gene expression (GE), and genotype (GT)

with fitness proxies in corals transplanted into novel environment. A-C: Density plots of discriminant

function values for each variable. The functions were developed to distinguish between coral fragments

in their native environment (KK vs. OO) and applied to transplanted fragments (KO and OK) to quantify

their adaptive plasticity (ability to shift towards local values). Arrows indicate mean plasticity. D:

Calculation of "similarity to locals" for each transplanted fragment. E: Principal components biplot of

fitness proxies in fragments transplanted to another reef, corrected for the overall higher fitness of

Orpheus corals transplanted to Keppel. Fit of "similarity to locals" for GBM, GE and GT onto this

ordination is shown as purple vectors scaled by correlation coefficient. Only GBM fit is statistically

significant.

Association between plastic GBM changes and gene expression changes (Fig. 2 F - H)

prompts the question: which one is likely to be the leading cause? The magnitude of GBM

changes is much lower than changes in GE (Fig. 2 D, E), which suggests that GE might be the

primary driver. However, considering the simplicity of plastic GBM changes - genome-wide

exaggeration or mitigation of GBM disparity among genes - it is more parsimonious to assume

that this is the leading change driven by the environment, which then translates into broad-

scale functional shifts in gene expression (Fig. 2 H and Fig. S8). In this way, genome-wide

adjustment of gene activity could be achieved by modulating a single process that governs the

disparity in GBM among genes.

Our results suggest that GBM disparity controls the balance between expression of two

broad gene classes: the low-methylated environmentally-responsive genes and highly-

methylated housekeeping genes. In the higher-quality environment of the Keppel island, where

corals attain higher fitness (Fig. 1C), GBM disparity change is associated with suppression of the

environmentally responsive genes and up-regulation of housekeeping genes (Fig. 2 F and Fig.

S8). Conversely, at the low-quality Orpheus Island the balance shifts the other way, from

housekeeping to environmental responsiveness (Fig. 2G and Fig. S8). We propose that mediating

such coarse adjustment of physiology in response to the environment is the ecological function

of GBM.

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222   **References** (Main paper)
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Materials and Methods

Sample sizes

The experiment started with 15 colonies from each of the two sites, which were halved

and transplanted, resulting in a total of 60 fragments distributed across four study groups 15

samples each: two "natives" (KK and OO), and two "transplants" (OK and KO, Fig. 1C). Although

not all original samples were successfully analyzed using all approaches (Table S1), in the end

each of these four groups included 11 fragments representing unique coral genotypes and

paired by genotype between "home" and "away" groups that were analyzed for GBM, gene

expression, genotype, and fitness proxies.

Reciprocal Transplantation Experiment

Field work was conducted with permission from the Great Barrier Reef Marine Park

Authority (Research permit G09/29894.1) as described previously (*23*). Reciprocal

transplantations were undertaken between two environmentally distinct study sites (Miall

Island in the Keppel Island group: 23°09S 150°54E and Hazard Bay on Orpheus Island 18°37S

146°29E) separated by 4.5 degrees of latitude on the Great Barrier Reef (Figure 1A). On the

23rd of April (Orpheus) and 4th of May (Keppels) 2010 fifteen colonies were collected from wild

populations from each site and split in two. One half of each colony was replaced in its native

habitat, while the second half was transplanted to the alternate study site. Samples from all

coral fragments were collected at midday after three months (9th July 2010 at Orpheus,

14th July at Keppels) frozen in liquid nitrogen, then transferred into RNAlater (Ambion, Austin,

TX, USA) for gene expression and DNA methylation profiling.

Coral growth rate

Total colony skeletal growth was evaluated using wet buoyant weight as in Jokiel,

Maragos, & Franzisket (1978). Measurements from beginning and end of the experiment were

expressed as percent daily weight gain.

Coral condition

The energetic condition of corals was determined by the analysis of total protein,

carbohydrate and lipid content. Coral tissue was removed from an average of 8.6 cm^2 of coral

branch (mean $\pm$ SE = 0.18) using an air gun in 12 – 15 ml of 0.2 µM filtered sea-water (FSW). The

coral slurry was homogenized and centrifuged at 1500 g for 5 min at $4^{o}C$ to separate coral and

endosymbiotic algal components. The coral slurry was aliquoted for protein, carbohydrate and

lipid assays and frozen at $-20^{o}C$ (10 ml for lipid, 0.5 ml for protein and carbohydrate). The algal

component was re-suspended in 2.75 ml of 0.2 µM FSW. Coral tissue surface area was

determined from cleaned (10% bleach for 24 hrs) skeletons of water-blasted samples using a

twice dip paraffin wax method (*25*). Surface area of coral branches was determined by the

weight gain of the second wax layer from a session specific standard curve (R^2 were 0.97 – 0.98)

using seven nylon cylinders of known surface area (5 – 48 cm^2).

Protein analysis

Total protein was extracted from a 0.5 ml aliquot using 0.5 ml 1M NaOH. Samples were incubated at 90°C for one hour and spun at 3000 g for 5 minutes to separate cell-debris from the solution and 500 µL clear supernatant was transferred onto a 96-well format. Protein concentration was quantified in three technical replicates of 50 µl of coral protein extract using a microplate Peterson – Lowry assay following the manufacturer's recommendations (Sigma: TP0300). Blank 0.5 M NaOH samples and BSA protein standards (50, 100, 200, 300 and 400 µg / ml) were run in duplicates on each plate and absorbance was read at 595 nm on a spectrophotometer. Standard curves had R^2 of 0.97 – 0.99. The coefficient of variation (CV) was calculated among technical replicates and deemed acceptable when <10%. For samples with a CV > 10%, single deviant outliers (>1 SD deviant from other two measurements) were identified and omitted. Total protein content per sample was expressed as average content (mg) of the technical replicates per cm^2 of coral branch surface area.

Carbohydrate analysis

Total carbohydrate content estimates were obtained from the average of three technical replicates of 50 µL coral slurry using D-glucose as a standard (*26*). Blank water samples and BSA protein standards (20, 50, 100, 200, 500 and 1000 µg/ml) were run in triplicate on each plate and absorbance was read at 485 nm on a spectrophotometer. Standard curves had R^2 of 0.96 – 0.99 and total carbohydrate content per sample was expressed in average mg of the technical replicates per cm^2 of coral branch surface area after quality control as described for proteins.

Lipid analysis

Lipids were extracted using a modified version of Harland et al. (1993). Ten ml of coral

slurry was freeze dried overnight then homogenised by vortexing with 2 x 5 mL of 2:1

dichloromethane:methanol (DCM:MeOH), each left for 24 hours at 4^oC. Extracts were filtered to

remove any debris using a glass filter (Whatman GF/C) and purified once with 5 ml 0.88%

KCl:H_2O and thrice with 1:1 MeOH:H_2O. Total lipid content was determined gravimetrically from

dried samples (60^oC overnight) in pre-weighed acetone washed aluminum trays. Total lipid

content was expressed in mg per cm^2 of coral branch surface area.

MBD-seq library preparation

DNA was isolated from adult holobiont tissue using dispersion buffer (4 M guanadine

thiocyanate, 30 mM sodium citrate, 30 mM β-mercaptoethanol) followed by phenol chloroform

purification and a final cleanup with Zymo Genomic DNA Clean and Concentrator-10 kit (Catalog

No D4011). Genomic DNA was sheared using a Misonix Sonicator 3000 to a size range of ~200 to

800 bp checked by gel electrophoresis. Enrichment reactions were performed using the

MethylCap kit (Diagenode Cat. No. C02020010) with an initial input of 2 μg of sheared DNA per

reaction. The methylated fraction was eluted from the capture beads in a single step using High

Elution Buffer. Library preparation using NEBNext® DNA library Prep Master Mix Set (E6040L)

and sequencing on an Illumina HiSeq4000 was performed at the University of Texas Genome

Sequencing and Analysis Facility. Our total sample size was N=44 (22 colonies divided in half

giving 11 samples per treatment group). For the majority of these we sequenced only the

enriched library eluted from the capture beads. Fold coverages from these captured libraries

were used to estimate relative differences in GBM between samples. For a subset of 12 of the

44 samples, we sequenced both the captured and flow-through fractions. Fold differences

between these captured and flow-through libraries were used to estimate absolute levels of

methylation across genes. We did this for only a subset of samples because we were primarily

interested in relative differences in GBM between groups. As relative differences could be

reasonably well assessed without sequencing the flow-through (Figure S2), we chose to focus

our sequencing resources on increasing sample size rather than more thorough estimates of

absolute methylation levels.

Repeated elements annotation

De novo identification of repeated elements in *A. digitifera* genome was performed using

RepeatModeler (http://www.repeatmasker.org/RepeatModeler/, version 1.0.10).

RepeatModeler launches two repeat finding programs, RECON 1.05 (*28*)

(http://eddylab.org/software/recon/) and RepeatScout 1.0.5 (*29*)

(https://bix.ucsd.edu/repeatscout/), to generate a genomic database and upon which to build,

refine and classify consensus models of putative interspersed repeats. This pipeline produced a

total of 2,858 non-redundant repeat families in recent *A. digitifera* genome sequence. These

sequences were then used as the custom library to search the *A. digitifera* genome using

RepeatMasker (http://www.repeatmasker.org/). Approximate 7.33% and 5.47% of the genome

was annotated to be composed of retrotransposon and DNA transposon elements, respectively.

MBD-seq data processing

Forty four MBD-seq libraries were prepared from reciprocally transplanted coral

fragments. Sequencing produced 980 million raw reads with a mean of 16 ± 0.65 SEM million

reads per sample. Raw reads were trimmed of non-template sequence using Cutadapt (*30*) and

quality filtered using Fastx toolkit (http://cancan.cshl.edu/labmembers/gordon/fastx_toolkit/).

Adapter trimming and quality filtering reduced these the total read count 940 million, mean =

15 ± 0.64 SEM million reads per sample. The reference genome and annotations (version 1.1)

for *Acropora digitifera* (*31*) were downloaded from NCBI. Trimmed and filtered reads were

mapped to this concatenated reference using Bowtie2 (*32*). To ensure that using a reference

from an alternate species did not severely impair mapping, we compared the mean mapping

efficiency against the *A. digitifera* reference with that of a draft genome sequence for *A.*

millepora produced by David Miller and coworkers (James Cook University). Mean mapping

efficiency against the *A. digitifera* ($78.4 \pm 0.7\%$ SEM) reference was only 5.2% lower than that of

A. millepora ($83.6 \pm 0.8\%$ SEM), hence sequence differences between the two species did not

appear to substantially impair read mapping. Following mapping, PCR duplicates were removed

using Picard (https://broadinstitute.github.io/picard/). Mean duplication frequency was $14.3 \pm$

0.5% SEM. The BAM files were filtered to retain only highly uniquely aligned reads (mapping

quality 30, or 0.1% chance of non-unique alignment) using samtools (*33*), after which the

number of reads overlapping various genomic regions - exon, intron, genic (exon + intron),

repeated elements longer than 500b, and repeat-free intergenic regions at least 1 kb long and at

least 2 kb away from any gene - were counted using BEDtools (*34*).

MBD-score and GBM gene classes

For 12 samples (3 per each experimental group) we quantified absolute levels of GBM as

the $\log_2$ fold difference in coverage between captured and flow-through DNA fractions while

controlling for genotype, as described in Dixon et al. (2016). These values were used as gene-

specific "MBD scores" throughout the study. As expected based on previous studies (*3*, *35*),

MBD-scores were bimodally distributed (Fig. S1 A) and negatively correlated with $CpG_{o/e}$ (Figure

1 D) providing genome-wide validation for our MBD-seq technique. The MBD-score threshold

separating the two clouds of high point density in the scatter plot on Figure 1 D was used to

delineate the two gene classes for computation of the "GBM class difference" (GBMcd) for each

sample (Fig. 2 I, J).

Tag-seq data processing

Transcription was quantified using Tag-seq (Meyer et al. 2011; Lohman et al. 2016). The

Tag-seq reads were downloaded from the SRA database (accession SRP049522; (*23*)) and

mapped against the *A.digitifera* genome using SHRiMP (*37*). Mapped reads overlapping

annotated coding sequences were counted using intersection-nonempty method in HTseq

version 0.6.1p1 (*38*). Normalization of raw counts and statistical analyses were performed using

DESeq2 (*39*).

Assessing variation in gene body methylation and transcription

423   As with gene expression analyses, normalization and statistical analyses of MBD-seq

reads were performed with DESeq2 (*39*). To test for origin effects, we performed two tests that

compared the two groups placed at Keppel to each other (OK vs KK) and the two groups placed

at Orpheus to each other (OO vs KO). These tests were intended to identify effects of origin

while controlling for environmental conditions experienced during the experiment. To assess

effects of transplantation, we compared groups that originated from Orpheus to each other (OO

vs OK) and groups that originated from Keppel to each other (KK vs KO). For these tests, we

included an additional parameter of colony identity to identify effects of transplantation while

controlling for genotype. For TagSeq, only genes with mean read count ≥3 were considered for

analysis (19706 genes), and for MBD-seq we chose genes with mean read count ≥20 (27084

genes).

Discriminant analysis of principal components

436   GBM data were further analyzed using discriminant analysis of principal components

(DAPC) implemented in the R package adegenet (*40*), following the procedures outlined in

Kenkel & Matz (2016). DAPC is a multivariate analysis method designed to identify between-

group variation while neglecting within-group variation. We used this method to distill our

multivariate MBD-seq dataset into single axes that maximized discrimination between natives

from the two experimental sites (KK and OO samples). The result was a discriminant axis

contrasting the GBM variation of the two native populations—with one pole designating native

Keppel-like GBM patterns and the opposite pole designating native Orpheus-like patterns. The

function was then applied to data from the transplanted samples, so that their positions along

discriminant axis described their similarity in GBM patterns to the contrasted native

populations. It seems plausible that, either due to natural selection or plasticity, native corals

would possess patterns optimal for their particular site. If this is true, then transplants with

DAPC values closer those of local corals would be expected to show improved fitness proxies

(eg: samples transplanted to Keppel with DAPC values closer to the 'Keppel-like' pole of the

discriminant axis would be expected to show higher fitness proxies and vice versa for samples

transplanted to Orpheus). 'DAPC Similarity' values describing each transplant's proximity along

the discriminate axis to natives of their transplantation site (Fig. 4 D) were computed by taking

the absolute value of the difference between the transplants' DAPC values and the mean value

for natives of their respective transplant sites, converting these distances into z-scores, and

multiplied them by -1. The DAPC Similarity value for sample X was:

$$DAPC\ Similarity_x = -1 \times (|D_x| - \overline{D})/\sigma$$

where, D_x is its distance along the discriminant axis between sample X and the mean DAPC

value for natives of its transplantation site, $\overline{D}$ is the mean absolute distance for all transplants,

and σ is the standard deviation of absolute distance for all transplants. Correlations of these

values with fitness characteristics of transplanted fragments (Fig. 4 E) were computed using

function *envfit* (R package *vegan* (*42*))

Validation of MBD-seq results by bisulfite amplicon sequencing

To validate our MBD-seq results, we used targeted bisulfite sequencing. Genomic DNA

for this procedure was extracted and purified as described above. It should be noted that as

with the MBD-seq and Tag-seq, targeted bisulfite sequencing was performed from separate

isolations from the same or closely adjacent branch. We performed bisulfite conversion using an

EZ DNA Methylation-Gold kit (Zymo Research; cat. no. D5005). Thermocycler conditions for

conversion were 98°C for 10 minutes followed by 53°C for 4 hours. Primers for post-conversion

amplification were designed using Bisulfite Primer Seeker (Zymo Research;

http://www.zymoresearch.com/tools/bisulfite-primer-seeker). Primer sequences are given in

Table S1. We designed 13 primer sets to target coding sequences (excised from the *A. digitifera*

reference genome) of 13 separate loci. Primer sequences included 5' tails for downstream

addition of barcodes and Illumina adapters by PCR. Target loci were selected based on either

strong origin or transplant effects in the DESeq2 analyses. Amplification from bisulfite converted

samples was performed using EpiMark Hot Start Taq DNA Polymerase (New England Biolabs;

cat. no. M0490S). Thermocycler conditions were 95°C for 30 seconds followed by 35-40 cycles

of 95°C for 15 seconds, 53°C for 30 seconds, 68°C for 30 seconds. Barcodes and adapters for

multiplexed sequencing were added to the resulting PCR products in a second PCR. The

oligonucleotide sequences used for barcoding were the same as those given in the Tag-seq

library preparation (Meyer et al. 2011; https://github.com/z0on/tag-based_RNAseq).

Sequencing was performed using paired-end 600 cycle runs on the Illumina Miseq. Resulting

reads were quality trimmed using cutadapt (*30*). Quantification of CpG methylation was

performed using Bismark (*43*) using a fasta file of the 13 exon sequences used to design the

primers as a reference. For a given sample, CpG sites represented by fewer than 50 reads were

excluded from the analysis (2022 low coverage calls excluded). The difference in means based

on origin for all CpGs in each gene correlated with the difference in mean normalized counts

from the MBD-seq reads (Figure 1 D).

Primers used for targeted bisulfite sequencing. Each primer includes a 5' tail (bold) for

amplification with barcoded primers for multiplex Illumina sequencing included in the Tag-seq

library preparation.

Primer	Sequence (5'-3')
LOC107327073_bsfTag_For	**CTACACGACGCTCTTCCGATCT**TTTTTGYGAGGTTGATTTTGTTATTATG
LOC107327073_bsfTag_Rev	**ACGTGTGCTCTTCCGAT**TTTCCAAACATATTCCTTTCCATAACATTC
LOC107356898_bsfTag_For	**CTACACGACGCTCTTCCGATCT**TAGATTTYGTTATAATGTTATTAAGAAGTGAAGG
LOC107356898_bsfTag_Rev	**ACGTGTGCTCTTCCGAT**AATAATATTAAACATTCTCTCTACAAATCTACCAC
LOC107358158_bsfTag_For	**CTACACGACGCTCTTCCGATCT**TGGTYGGATTGTTGAAGAGTTTAAGTAG
LOC107358158_bsfTag_Rev	**ACGTGTGCTCTTCCGAT**ACACCCAAATCACCCATCTCATTAAC
LOC107356899_bsfTag_For	**CTACACGACGCTCTTCCGATCT**GTTTTTAAAATTTGAAGGATTTGGTTTTGTTG
LOC107356899_bsfTag_Rev	**ACGTGTGCTCTTCCGAT**TTACATTATATTTTCCAAACATATTTCATACCATAAC
LOC107358871_bsfTag_For	**CTACACGACGCTCTTCCGATCT**GATATYGGGTTTTTAATAATAATTGTATGTTGGTTG
LOC107358871_bsfTag_Rev	**ACGTGTGCTCTTCCGAT**TTTTCCTTTAAAATTATTTCCACCCAACTCC
LOC107334334_bsfTag_For	**CTACACGACGCTCTTCCGATCT**ATAAGGATATGTAGGGTTTTGGTAAGG
LOC107334334_bsfTag_Rev	**ACGTGTGCTCTTCCGAT**AATTCATAATAACTACCCTACAACAAAAAATCCC
LOC107347512_bsTag_For	**CTACACGACGCTCTTCCGATCT**ATYGATAGATTAAAAGAAGTTGGAGTATTG

LOC107347512_bsTag_Rev	**ACGTGTGCTCTTCCGAT**CCTAAACAATCCATAAAAACCTTCCTACAATTC
LOC107350794_bsTag_For	**CTACACGACGCTCTTCCGATCT**GAYGTGTTTTGTATTTAGTTATTGGATATTTGG
LOC107350794_bsTag_Rev	**ACGTGTGCTCTTCCGAT**ATCTTCCRCAATTAAAAAAACATACTTAAAATCTTCAC
LOC107339795_bsTag_For	**CTACACGACGCTCTTCCGATCT**TGAATATAGTTAAGGTAAATGGATGAGTTATATATGG
LOC107339795_bsTag_Rev	**ACGTGTGCTCTTCCGAT**CAAAAATAAACRAATCAAACAAAACAATCATTAC
LOC107336909_bsTag_For	**CTACACGACGCTCTTCCGATCT**AGAGATGYGGAATAGATATTTTTGGTTGG
LOC107336909_bsTag_Rev	**ACGTGTGCTCTTCCGAT**AATAACRAACATTACATCTTATTTCTCTAAAATAATAC
LOC107352877_bsTag_For	**CTACACGACGCTCTTCCGATCT**TGGTGTYGTGAATTTGTTTAAATAATTTATG
LOC107352877_bsTag_Rev	**ACGTGTGCTCTTCCGAT**CATTCCATCRACACAACTAAATAAAAAAC
LOC107351808_bsTag_For	**CTACACGACGCTCTTCCGATCT**GTYGAGATATGTTAAAATTTGAGGATGTGAGTTTG
LOC107351808_bsTag_Rev	**ACGTGTGCTCTTCCGAT**ATCCACCATTCRCCAACTTACTAATAAATCTCCC
LOC107327285_bsTag_For	**CTACACGACGCTCTTCCGATCT**ATTATGTTAGYGTTTTTGTTATTTTGGATTGTGG
LOC107327285_bsTag_Rev	**ACGTGTGCTCTTCCGAT**CCTCATTCACAAAAACAACCTTTAATC

Analysis of genetic differentiation

The MBD-seq reads were also used to analyze genetic distances among the 22

sequenced colonies. For this procedure, we concatenated reads from clone pairs into single

files, and mapped the reads to the *A. digitifera* reference genome (*31*) using Bowtie2 (*32*). For

the subset of 12 samples where we sequenced both captured and flow-through fractions, both

sets of reads were concatenated for genotyping. Between-individual genetic distances for DAPC

analysis were determined as (1 - Identity-By-State) using ANGSD (*44*) with the following filter

settings: -uniqueOnly 1 -remove_bads 1 -minMapQ 20 -minQ 30 -baq 1 -minInd 10 -snp_pval 1e-

2 -minMaf 0.1 (to summarize, only high-quality and uniquely-mapped reads were taken into

account; only sites covered in a minimum of 10 individuals were considered; the cutoff for the

p-value for the SNP being true was set at 0.01; and only alleles with frequency equal or

exceeding 0.1 were used). This resulted in genotyping information for roughly 507,000

polymorphic sites. The between-population F_{ST} for genes passing 10% FDR cutoff for either for

origin or transplant effect and a random subset of 500 non-significant genes (rather than whole

genome, to speed up computation) were determined using ANGSD based on site frequency

spectra as priors, which were generated for every gene subset without using filters that would

affect representation of alleles of different frequencies (-uniqueOnly 1 -remove_bads 1 -

minMapQ 20 -minQ 30 -baq 1 -minInd 4).

Determination of relative proportions of Symbiodinium *clades*

The MBD-seq reads were mapped to the concatenated reference including A. digitifera

genome and four *Symbiodinium* sp. trasncriptomes from four different genotypic groups

("clades"). Transcriptomes for *Symbiodinium* clades A and B were from (*45*) and transcriptomes

for clades C and D were from (*46*). We then counted the relative proportions of reads producing

highly unique matches (mapping quality 40 or higher) to each *Symbiodinium* transcriptome,

using a custom perl script *zooxType.pl*. All corals were found to be dominated by *Symbiodinium*

clade C.

Data and scripts availability

TagSeq, MBD-seq, and amplicon-bisulfite sequencing datasets are available through NCBI

Short Read Archive (Project Accession: SRP049522). Scripts and traits data are available in a

GitHub repository dedicated to this paper

(https://github.com/grovesdixon/reciprocal_transplant_methylation).

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Supplementary Tables and Figures

Table S1. Sample sizes

Sample group	N (TagSeq)	N (MBD-seq, captured and flow-through fractions)	N (MBD-seq, only captured fraction)
KK	14	3	8
KO	15	3	8
OK	15	3	8
OO	15	3	8

Figure S1. Distribution of MBD-scores (A) and validation by bisulfite sequencing of amplicons (B, C). B: Mean percent methylation was calculated as the proportion of methylated CpG sites within each gene averaged across all samples. C: Correlation between MBD-based and bisulfite-based measurement of the effects of each coral individual on methylation for each of the assayed genes.

Figure S2 Correlation between inferred changes in GBM based on comparison of MBD-captured

and flow-through libraries (x-axis) and only on MBD-captured fraction (y-axis). The red line is 1:1

correspondence. (A) Orpheus to Keppel transplantation; (B) Keppel to Orpheus transplantation.

These plots were generated for a subset of 12 samples (three for each of the four experimental

groups, KK, KO, OK, and OO) for which both the captured and the flow-through fractions were

sequenced.

Figure S3 Heatmap of GBM correlations among samples, illustrating strong dependence of GBM

on genotype. Colors indicate Spearman's rank correlations for normalized MBD-seq read counts

across all coding genes (N = 24853); the lowest observed correlation value was 0.9. Samples

were clustered by maximum distance method. First letter of sample names indicates sample

origin, second letter indicates transplantation site, and number indicates replicate. Samples

sharing the same first letter and the same number are clonal fragments from the same colony.

All of the 22 clone pairs except one (sample pair KK4 and KO4) clustered together.

Figure S4 Heatmap of overall correlations in transcription illustrating strong dependence of

transcription on genotype. Colors indicate Spearman's correlations for normalized Tag-seq read

counts across all coding genes (N=19706). Samples were clustered by maximum distance

method. First letter of sample names indicates sample origin, second letter indicates

transplantation site, and number indicates replicate. Samples sharing the same first letter and

number are clonal fragments from the same colony. All 24 available clone pairs clustered

together (six samples lacked data for clone pairs).

Figure S5. GBM log$_2$ fold-changes change in response to transplantation in exons and introns

plotted against MBD-score for each gene. Red and blue curves near the x-axis are density plots

of significantly up- and down-regulated genes.

Figure S6. Absolute methylation changes in response to transplantation ("away" vs. "home")

can be positive or negative across genes (A, B) and affect mostly protein-coding regions (C, D).

Plots on A and B are based on comparison between captured and flow-through fractions for 12

samples (three for each of the four experimental groups, KK, KO, OK, and OO). Plots on panels C

and D were generated based on captured-only analysis of all 44 MBD-seq samples (Table S1).

Figure S7. Gene expression changes in response to transplantation are negatively correlated

with GBM changes. Both correlations are highly significant (p<0.0001). Red points on panel A

represent genes with statistically significant (FDR<10%) change in GBM.

Figure S8. Gene ontology categories significantly enriched with up- or down-regulated genes

(red and blue, respectively) in GE and GBM comparisons. The Benjamini-Hochberg corrected

significance of enrichment is indicated by the size and type of the font (see legends). Note that

for plastic GBM the significance thresholds are ten times more stringent, to fit the GO plot onto

one page. The fraction preceding the category name is the number of genes exceeding raw p-

value of 0.05 relative to the total number of genes in the category. The dendrograms are

hierarchical clustering of GO categories based on proportion of shared genes (*47*). Genes

comprising the branch labelled by the yellow circle (transcription factors) show plastic changes

(D) that are opposite to the constitutive differences between populations (B).

651   **Figure S9**. Density plots of F_{ST} for genes that show significant (FDR <0.1) GBM change in

652   response to transplantation from Orpheus to Keppel ("Transp", red curve) compared to 500

653   randomly picked non-significant genes ("NS", black curve). Dashed vertical lines mark the mean

654   of each dataset.