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To link to this article DOI: http://dx.doi.org/10.1111/acel.12030

Publisher: Wiley-Blackwell

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Nutritional factors and gender influence age-related DNA methylation in the human rectal mucosa.

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Running title: Factors influencing age-related DNA methylation

Keywords: DNA methylation; CpG islands; aging; colorectal cancer; nutrition; folate
Summary

Aberrant methylation of CpG islands (CGI) occurs in many genes expressed in colonic epithelial cells, and may contribute to the dysregulation of signalling pathways associated with carcinogenesis. This cross-sectional study assessed the relative importance of age, nutritional exposures and other environmental factors in the development of CGI methylation. Rectal biopsies were obtained from 185 individuals (84 male, 101 female) shown to be free of colorectal disease, and for whom measurements of age, body size, nutritional status and blood cell counts were available. We used quantitative DNA methylation analysis combined with multivariate modelling to investigate the relationships between nutritional, anthropometric and metabolic factors and the CGI methylation of 11 genes, together with LINE-1 as an index of global DNA methylation. Age was a consistent predictor of CGI methylation for 9/11 genes but significant positive associations with folate status and negative associations with vitamin D and selenium status were also identified for several genes. There was evidence for positive associations with blood monocyte levels and anthropometric factors for some genes. In general, CGI methylation was higher in males than in females and differential effects of age and other factors on methylation in males and females were identified. In conclusion, levels of age-related CGI methylation in the healthy human rectal mucosa are influenced by gender, the availability of folate, vitamin D and selenium, and perhaps by factors related to systemic inflammation.

Introduction

Cytosine exhibits a non-uniform pattern of methylation in the human genome. Typically, Cytosine-Guanine dinucleotides (CpGs) distributed within both coding and non-coding regions are methylated, whereas those located within CpG-islands (CGI), sequences with
significant over-representation of CpGs and located within the promoter regions of more than 50% of mammalian genes, are usually unmethylated. There is compelling evidence for an association between colorectal carcinogenesis and aberrant DNA methylation (Lao & Grady 2011), which includes both genome-wide loss of DNA methylation (global hypomethylation) and differential methylation of CGI of specific genes (CGI hypo- or hypermethylation). Global hypomethylation contributes to genome instability, while CGI hypo- and hypermethylation lead respectively to increased and decreased transcription of genes important for regulating processes disrupted during carcinogenesis (Jones & Baylin 2002). Aberrant DNA methylation is observed in the early stages of neoplasia in the human colon (Chan et al. 2002) and also occurs in many apparently healthy tissues (Christensen et al. 2009), including the epithelial cells of the morphologically normal colonic crypt (Belshaw et al. 2010). Methylation of some CGI in the colorectal mucosa increases progressively with age, and this may contribute to field-wide changes in the colorectal mucosa that may increase its vulnerability to neoplastic transformation (Shen et al. 2005). In previous studies we have quantified CGI methylation in biopsies from macroscopically normal colorectal mucosa, and confirmed that field methylation patterns differ significantly between healthy subjects and patients with neoplastic lesions (Belshaw et al. 2008). We have also shown that methylation levels measured within mucosal biopsies reflect highly variable levels of methylation in individual crypts (Belshaw et al. 2010). This mosaic pattern of CGI methylation in the mucosal field is probably a consequence of the unique methylation signatures of the corresponding stem cells.

The origins of aberrant CGI methylation are uncertain but, if the phenomenon plays a functional role in colorectal carcinogenesis, associations between CGI methylation and environmental factors known to influence cancer risk would be anticipated. The most important risk factor for colorectal carcinogenesis is advancing age (Armitage & Doll 1954), and significant associations between age and gene-
specific CGI methylation have been reported (Belshaw et al. 2008). Apart from age, CGI methylation may be influenced by diet, metabolic status or other environmental factors (Christensen et al. 2009; Talens et al. 2012). In a recent study, Wallace et al. measured the methylation status of CGI in ESR1 and SFRP1 in biopsies from the right colon and rectum of 389 polyp patients enrolled in a multicentre trial of aspirin and folic acid for prevention of polyp recurrence (Wallace et al. 2010). For both genes, methylation varied with ethnicity, and with anatomical location, and positive associations were observed with age and red cell folate (RCF) status.

In the cross-sectional study described here we have quantified the CGI methylation status of 11 genes, together with LINE-1 as an index of global DNA methylation, in rectal biopsies from 185 patients undergoing diagnostic colonoscopy, who were shown to be free of gastrointestinal inflammatory or neoplastic disease. The panel of genes was selected on the basis of previous studies showing age and cancer-associated methylation, or a cancer-related function or that methylation status contributed to the discrimination of normal mucosa from patients with and without neoplasia (Belshaw et al. 2008; Lao & Grady 2011). Multivariate modelling using a genetic algorithm approach was used to assess relationships between DNA methylation and age, anthropometry, nutrient status and blood cell counts. We confirmed a strong positive effect of age on CGI methylation levels in the normal rectal mucosa. In addition we identified effects of gender, and showed a significant positive association between CGI methylation levels and folate status and negative associations with selenium and vitamin D. We also obtained evidence for significant effects of white blood cell counts and anthropometric factors.
Results

The participant characteristics are summarised in Table 1 and correlations between each cofactor are provided in Supplementary Table S1. Methylation characteristics of the 9 gene-specific CGI and of LINE-1 (Figure 1) show clear differences in the mean methylation and in the extent of inter-person variability between different CGIs. Methylation of AXIN2 and DKK1 was generally very low and showed no correlation with age (data not shown) so these genes were excluded from subsequent analysis. For the remaining genes, seven showed strong positive inter-correlations (unadjusted p<0.001), indicating concomitant methylation of these genes within a given participant (Supplementary Table S2). An eighth gene, N33, was correlated only with SOX17 and WIF1 (p<0.01) while none of the correlations with APC were significant at the p = 0.001 level. LINE-1 methylation was uncorrelated with all 9 genes (unadjusted p>0.05).

The GA-ANCOVA (genetic algorithm with analysis of covariance) modelling aimed to find subsets of exposure variables that best explained the variation in methylation, allowing particular attention to be given to variables that occurred frequently within the sub-set models. The final models, the frequency of occurrence over 10 GA evaluations, and the square of the correlation coefficient between actual and non cross-validated predictions, which represents the fraction of explained variance, are given in Table 2. The genes for which the measured exposure factors explained the highest and lowest proportion of the variance were WIF1 (>30%) and APC (<5%), with the remaining genes and LINE-1 having values in the range 10 – 20%. The model for the summary statistic PCA1 (the first principal component score vector) explained the largest proportion of the observed variance (38%).
The selection criteria may be susceptible to some degree of over-fitting and hence the $R^2$ figures in Table 2 are optimistic estimates of explained variance. The ability to generalise was therefore evaluated using double cross-validation (2CV). This appraised the overall modelling process as well as the robustness of modelling data associated with a particular gene. The models for predicting methylation of \textit{WIF1}, \textit{HPP1}, \textit{SFRP1}, \textit{SFRP2}, \textit{SOX17}, \textit{ESR1}, \textit{MYOD1} and PCA1 all passed 2CV, suggesting that they could generalise to other datasets. The models for \textit{APC}, \textit{N33} and LINE-1 methylation failed 2CV, probably because of over-fitting, and caution should be exercised in their interpretation. As before, the PCA1 model captured the most methylation variance (21.1%), providing further evidence for concordance in CGI methylation in response to age and other exposures.

For each of the models (Table 2), the corresponding ANCOVA tables, regression vectors and a graphical representation of the models are provided as supplementary material (Supplementary GA-ANCOVA file). Figure 2 summarises the Pearson correlation coefficients between each of the exposure variables and the methylation of each gene, LINE-1 and PCA1. The correlations are shown separately for all study participants, males only, and females only (further details in Supplementary Tables S3a-S3c).

\textbf{Age}

There was a strong positive correlation between participant age and PCA1 methylation ($r = 0.556$), and the GA model also identified an interaction between age and sex. Inspection of the age regression coefficients showed these were positive for both sexes but larger in males. Figure 2 shows that age was also correlated positively in both sexes but stronger in males. This correspondence was found in \textit{HPP1} and \textit{SOX17}, also modelled with sex interactions (Table 2). The difference in the age
regression coefficients may therefore reflect a lower weighting of age in females due to their poorer correlation. The link between methylation and only age and sex was investigated using ANCOVA for PCA1, HPP1 and SOX17. In all cases the sex interaction term was significant, the gradients were positive for both genders, and significantly larger in males.

Figure 3 shows the variation in methylation with age for PCA1 but the same general behaviour was seen for all of the genes studied (Supplementary GA-ANCOVA file). Figure 3 shows a characteristic ‘wedge-shape’, or heteroscedastic, response with age, in which the mean and range of methylation increased with age. This observation violates the assumption of homoscedastic variance made in the regression analysis. The correspondence between the sizes of the correlation and gradient coefficients for gender-related age effects on methylation may therefore be coincidental but, alternatively, could be due to the heteroscedastic relationship between age and methylation. Considering the gradient and correlation findings separately, the steeper gradient in males implies that older males have higher methylation than females of the same age. A lower correlation with age in females suggests other factors have a greater influence on their CGI methylation. Figure 2 shows the stronger correlation of age with methylation in males was also seen to lesser extents in ESR1, APC and SFRP2. For LINE-1 there was a weak negative correlation between age and methylation, consistent with the global loss of DNA methylation during ageing (Iacopetta et al. 2007).

Nutritional status

Plasma folate (PF) status was selected in the PCA1 methylation model (Table 2), and was correlated positively with methylation (Figure 2). Folate status, either PF or RCF, was also selected in the models for WIF1, MYOD1, SFRP1 and N33 methylation.
(Table 2), and for SFRP1 and N33 there was an interaction between folate status and sex. SFRP1 methylation correlated positively with RCF only in males whereas N33 was positively correlated with PF in females and negatively in males. LINE-1 methylation showed a weak negative correlation with folate status in males.

The selection of plasma selenium in the PCA1 methylation model (Table 2) reflected a consistent trend, observed for several of the genes, of a positive correlation with CGI methylation in males and a negative correlation in females (Figure 2). However, for WIF1 methylation, the correlation with plasma selenium was negative in both males and females (Figure 2). There was a positive correlation between LINE-1 methylation and plasma selenium only in females ($r = 0.283$, unadjusted $p<0.01$).

Plasma vitamin D was associated with PCA1 methylation (Table 2). In general, vitamin D status and gene-specific methylation were correlated negatively, with a trend towards stronger associations in females, whereas a small positive correlation with LINE-1 methylation was observed for both sexes (Figure 2).

Neither vitamin B-12 nor homocysteine were selected by any of the models of gene-specific methylation, although there was a weak correlation with MYOD1 methylation in males ($r = 0.223$, unadjusted $p<0.05$). Vitamin B-12 was selected in the model for LINE-1 methylation (Table 2) with a non-significant negative correlation in males ($r = -0.199$) and no association in females ($r = 0.017$) (Figure 2). Homocysteine was not significantly correlated with methylation of any of the genes, or with LINE-1.

**Anthropometry**

Measures of body size including body mass index (BMI), FI (a derived measure of body adiposity), waist and hip circumferences and height did not feature in the model for PCA1 methylation (Table 2), which suggests the absence of a general effect on CGI methylation. However, some gene-specific associations with anthropometric
indices were identified (Table 2). Waist and hip circumference correlated positively with \textit{WIF1} methylation, with stronger correlations in males; positively with \textit{N33} methylation with stronger correlations in females; and negatively with \textit{LINE-1}, with stronger correlations in females (Figure 2). FI was identified by the models for \textit{APC} and \textit{SFRP2} methylation and showed small positive correlations in males and negative correlations in females.

Participant height*sex interactions were selected in the models of methylation for \textit{ESR1} and \textit{LINE-1} (Table 2). In both cases the correlation was stronger in females. For \textit{ESR1}, height correlated positively in females, $r = 0.223$ (unadjusted $p<0.5$) but negatively in males, $r = -0.179$. Converse relationships were found for \textit{LINE-1}, with a negative correlation in females, $r = -0.260$ (unadjusted $p<0.01$) and positive correlation in males, $r = 0.125$ (Figure 2).

**White cell counts**

Although not selected in the model of PCA1 methylation, either monocyte and/or white blood cell counts were selected in models for \textit{WIF1}, \textit{HPP1}, \textit{SFRP1} and \textit{SOX17} methylation. Monocyte count tended to be positively correlated with methylation, particularly in males and white cell count showed weak negative correlations with methylation (Figure 2).

**Discussion**

In this investigation we quantified relationships between DNA methylation in the rectal mucosa of healthy subjects and biological factors implicated in epidemiologic studies as CRC risk factors. We adopted a modelling approach based on genetic algorithms to select variables that are the strongest predictors of gene-specific methylation.
Although the levels of methylation at some of the loci we investigated were relatively low, we validated the robustness of these predictive models and assessed their ability to generalise to other data-sets. Our observations confirm the importance of age as a major determinant of CGI methylation for most of the genes investigated, but also strongly suggest that folate status, and other nutritional and metabolic factors, modulate the methylation of specific genes. Thus these factors contribute to the patterns of methylation acquired across the mucosa, and hence may play a role in the aetiology of CRC.

The single most important CRC risk factor is advancing age, with risk increasing exponentially beyond 50 years (Armitage & Doll 1954). Previous studies have shown the concomitant methylation of multiple genes during ageing, leading to the classification of this subset of genes as ‘Type A’, as distinct from ‘Type C’ genes, where methylation is thought to be cancer-specific (Toyota et al. 1999). However, more recent studies argue against such a rigid classification (Belshaw et al. 2008). The strong association between age and methylation of 8 CGIs observed here is consistent with previous studies (Belshaw et al. 2008). This increases our confidence in the modelling technique, and reinforces the hypothesis that time-dependent silencing of genes that play an important role in regulating tissue renewal and maintaining homeostasis within individual stem cells in the colorectal epithelium contributes to field-wide changes in mucosal vulnerability to neoplasia. However, APC methylation was not significantly correlated with age nor with the methylation of the other genes and was also not strongly associated with the exposure variables. In this context it is interesting to note that CGI methylation has only a limited effect on the expression of APC in human colorectal tumours, and consequently the role of APC methylation in carcinogenesis remains unclear (Segditsas et al. 2008).
The mechanisms leading to age-dependent CGI methylation are unclear. Recent studies showed that susceptible CGI in several cell-types are enriched significantly for targets of polycomb group proteins, which are marked with both active and repressive histone modifications in embryonic stem cells (Teschendorff et al. 2010). This suggests that these CGI are more prone to de novo methylation. Expression of the DNA methyltransferase (DNMT) enzymes responsible for CpG methylation is modulated during ageing in other tissues (Xiao et al. 2008). Increased expression of DNMT3B, a component of the polycomb repressor complex, was associated with sequential DNA methylation changes and colorectal neoplastic progression (Ibrahim et al. 2011), while over-expression of Dnmt3b in the murine colon led to CGI methylation of genes including those shown here to be methylated with age (Steine et al. 2011).

Despite the strong influence of age on DNA methylation, our study shows considerable inter-individual variation in the extent of age-related methylation for the genes investigated here. Previous twin studies have suggested that environmental factors contribute to this variance in age-related methylation for the majority of affected loci, while for a small subset of genes, genetic factors were more important (Bell et al. 2012; Talens et al. 2012). However, there is currently no evidence linking genetic factors to the variance in methylation of this panel of genes and we show that CGI and global DNA (LINE-1) methylation were associated with nutritional status and anthropometric measures, in a manner generally consistent with their impact on CRC risk as shown by epidemiologic studies.

We identified gender-related differences in the rate of age-related CGI methylation such that older males had higher levels of CGI methylation than females of the same age. In addition, for some individual genes, the strength of association between DNA methylation and height and waist circumference differed in males and
females (Table 2). This is consistent with the higher overall risk of CRC in males compared with females (Koo & Leong 2010).

There was an association between increased folate status, determined as either plasma or red cell folate, and increased CGI methylation, with significant positive correlations for MYOD1, SFRP1, SFRP2, WIF1, and HPP1 and for the summary statistic for overall CGI methylation (PCA1). This is consistent with observations showing a positive association between RCF and methylation of SFRP1 and ESR1 in the apparently normal mucosa of previous polyp patients (Wallace et al. 2010). Folate plays a crucial role as a methyl group donor, contributing to the cellular supply of S-adenosylmethionine (SAM), the proximate source of methyl groups for DNA methylation. Indeed mathematical modelling of one-carbon metabolism supports the possibility that increased availability of folate might ‘drive’ DNA methylation through an increased intracellular SAM supply (Nijhout et al. 2006).

Concerns have been expressed that excessive intakes of folate through mandatory food fortification with folic acid may promote the development of CRC (Mathers 2009) through various mechanisms including hypermethylation and silencing of tumour suppressor (and other genome defence) genes. However, epidemiological studies of the relationship between folate intake and risk of adenoma recurrence or CRC are inconsistent (Carroll et al. 2010; Fife et al. 2011; Kennedy et al. 2011) and there is a similar lack of agreement amongst studies investigating the relationship between plasma folate status and risk of CRC. For example, Martinez et al. observed a reduced risk of adenoma recurrence in non-users of multivitamin supplements in the highest quartile of folate status compared with those in the lowest quartile (Martinez et al. 2006), whereas in the Northern Sweden Health and Disease Cohort, higher plasma folate levels were associated with greater CRC risk. Eussen et al. reported no association between plasma folate and risk of CRC (Eussen et al. 2010), but recently
Lee et al. reported lower CRC risk with lower plasma folate status amongst participants in three large prospective trials in the USA (Lee et al. 2012). In view of these uncertainties, further studies are urgently required to understand the relationship between folate status, DNA methylation and colorectal carcinogenesis.

Vitamin B-12 and homocysteine are also important factors in one-carbon metabolism. Vitamin B-12 is a co-factor for methionine synthase responsible for the remethylation of homocysteine to methionine, while homocysteine is produced from S-adenosylhomocysteine, the co-product of methylation reactions and a potent inhibitor of DNA methylation. Al-Ghnaniem et al. reported a negative association between vitamin B-12 and methylation of ESR1 CGI in the normal colonic mucosa, while a positive association was observed for homocysteine (Al-Ghnaniem et al. 2007). In contrast, there was a negative association between plasma homocysteine and global DNA methylation in colon tissue (Pufulete et al. 2005). In this study we found no associations with either vitamin B-12 or homocysteine and DNA methylation.

Vitamin D status was inversely related to CGI methylation in this set of genes, which is consistent with the convincing evidence from epidemiological studies for a protective effect of vitamin D against CRC risk (Lee et al. 2011). Putative mechanisms for this protective effect include inhibition of cell proliferation and promotion of cell differentiation by antagonism of the Wnt signalling pathway, or by induction of the DNA demethylation-dependent expression of E-cadherin (Palmer et al. 2001; Lopes et al. 2012).

Epidemiological evidence suggests a protective effect of selenium against CRC. Selenium modulates DNA methylation in both CRC cell lines and in colon tissue of rats by inhibition of DNMT expression (Davis et al. 2000) and inhibits DNMT activity in vitro (Fiala et al. 1998), which suggest that selenium may decrease CRC risk by preventing aberrant DNA methylation. In the present study, there was a weak negative correlation
between selenium status and methylation of *WIF1* in both sexes. For most of the genes studied, this weak negative correlation was found predominantly in females. In contrast, selenium status correlated positively with LINE-1 methylation in females. This suggests the possibility of a sex-dependent interaction between selenium, DNA methylation (at least for some genes and for LINE-1) and CRC risk. Sex-dependent effects of selenium on cancer risk have been reported for bladder and oesophageal cancers (Amaral *et al.* 2010; Steevens *et al.* 2010) and may reflect differential responses to selenium by sex (Combs *et al.* 2011).

Adult height is positively associated with CRC risk in women but not in men (Park *et al.* 2012). Height is influenced by many genetic and environmental exposures acting *in utero*, and during childhood. In agreement with the specific-link between height and CRC risk, we observed that height was correlated significantly with increased *ESR1* methylation, and with decreased LINE-1 methylation only in females.

Obesity *per se*, and increased central adiposity in particular, are significantly associated with increased CRC risk (Pischon *et al.* 2006). Although the mechanisms remain to be established, increased low-grade, systemic inflammatory signals produced by increased adipose tissue may contribute to greater cancer risk (Johnson & Lund 2007). Higher waist and hip circumference were associated significantly with higher methylation of several CGIs in the present study, and with lower methylation of LINE-1. Further support for an association between systemic inflammation and CGI methylation is suggested by the significant positive association between aberrant DNA methylation and blood monocyte count. This subset of white blood cells is elevated in response to infection and chronic inflammatory conditions such as ulcerative colitis (Mee *et al.* 1980). However no participants in the BORICC Study showed any evidence of inflammatory bowel disease.
In conclusion, the application of a multivariate modelling approach to quantitative methylation analyses of DNA from the normal rectal mucosa of healthy subjects confirmed the importance of age as a major determinant of methylation but also identified significant relationships between DNA methylation and several nutritional, metabolic and anthropometric factors related to CRC. Our panel of genes consisted mainly of genes involved in the Wnt signalling pathway, which plays a central role in controlling tissue renewal in the colon and is over-active in the vast majority of colorectal tumours (Gregorieff & Clevers 2005). Our observations suggest that regulation of this pathway is susceptible to modulation by the combined effects of ageing, nutrition and metabolic status, which would have important implications for the maintenance of colorectal tissue homeostasis. These relationships provide a biologically plausible mechanism for the influence of environmental factors on CRC risk, which requires further investigation. Future studies, utilising this approach and incorporating more environmental and metabolic factors, will further our understanding of how the environment modulates the epigenomes of normal cells, and the consequential influence on CRC risk.

Experimental procedures

Subject recruitment and sample collection

Participants in the BORICC (Biomarkers of Risk of Colorectal Cancer) study were recruited from patients attending Wansbeck General Hospital, Northumberland, UK for diagnostic endoscopy, who had no known familial predisposition to colorectal cancer, and who were shown by flexible sigmoidoscopy to be free from colorectal neoplasia or inflammatory disease. While this procedure is considered sufficient for the clinical
diagnosis of absence of disease, there remains a small possibility that conditions
affecting the proximal colon may be present. Patients were invited to participate in the
study and written informed consent was obtained. Ethical approval for the study was
obtained from the Northumberland Local Research Ethics Committee (Project
reference NLREC2/2001). Nine rectal mucosal samples were taken 10 cm from the
anal verge by pinch biopsy from 185 volunteers (84 males, 101 females; median age
50y, range 17-77 y). Blood was collected by venepuncture, and samples of whole
blood, serum and plasma were prepared, flash-frozen with liquid nitrogen and stored
at -80°C. Anthropometric measurements (height, weight, and waist and hip
circumferences) were made and Fatness Index (FI) was calculated (see
Supplementary Methods).

**DNA methylation analysis**

Biopsies were stored at -80°C before analysis. Genomic DNA was extracted and purified
using a Genelute DNA extraction kit (Sigma-Aldrich, Gillingham, UK). CGI methylation
analysis was carried out using a quantitative methylation-specific PCR assay described
previously (Belshaw et al. 2008), applied to the following genes: *SFRP1, SFRP2, AXIN2,
WIF1, APC, HPP1/TMEFF2, DKK1, SOX17, ESR1, MYOD1* and *N33*. PCR primer
sequences are provided in Supplementary Table S4. Methylation of LINE-1 was measured
as an index of global DNA methylation using the qPCR approach (Iacopetta et al. 2007).

**Nutritional status**

Blood samples were analysed for red cell folate (RCF), plasma folate (PF), vitamin
B12, white blood cells (WBC) and monocytes using standard clinical laboratory
assays. Plasma homocysteine was measured using reverse-phase HPLC as
described by Loehrer et al. (Loehrer et al. 1996). Plasma 25-Hydroxyvitamin D was

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Britain and Ireland
determined using a commercial enzyme immunoassay kit (Immunodiagnostics Systems Limited, Tyne & Wear, UK). Plasma Selenium concentration was measured by ICP MS.

**Statistical analysis**

Data were analysed using Matlab (Mathworks Inc., Cambridge, UK). Pearson correlation was used to investigate relationships of CGI methylation with each independent variable (age, nutritional status and anthropometric indices) and to aid interpretation of the subset regression models (see below).

Data were available for 185 subjects for all variables except for methylation of ESR1, MYOD and N33 where data were available for 174 subjects only due to failure of DNA amplification for some subjects. To facilitate equal weighting for each gene in subsequent analyses, methylation values (expressed initially as %) were auto-scaled to span the range [0.01 – 1]. The first principal component score vector PCA1, based on all 9 genes, was calculated as an overall indicator of CGI methylation status for 174 subjects and was also auto-scaled.

We applied multivariate modelling to predict methylation levels using an approach that selected a subset of explanatory variables, whilst catering for potential gender-differences in each selected cofactor. Each selected variable could be associated with a single ‘term’ or have separate terms for men and women. ANCOVA models of gene methylation were built from predictors comprising 1 categorical variable, ‘Sex’; and 16 continuous variables (anthropometric and blood measurements) using GA (Tapp et al. 2003; Kemsley et al. 2007). For further information on the GA-ANCOVA modelling, see Supplementary Methods.
Acknowledgements

This work was funded by the Food Standards Agency (Grant no. N12015) and by the BBSRC’s core strategic grant to the Institute of Food Research’s Integrated Biology of the Gastrointestinal Health (IFR/08/1) and Gut Health and Food Safety (BB/J004529/1) programmes.

Author contributions

NJB, JCM, ITJ obtained funding and designed the study, DMC, DMB, RA, NJB collected and analysed the samples, HST, ITJ, NJB analysed the data and wrote the manuscript, JCM revised the manuscript.

References


**Supporting information**

Supplementary Tables S1, S2, S3a-S3c and S4.

Supplementary GA-ANCOVA file.

Supplementary Methods file.
### Table 1. Characteristics of the study participants.

<table>
<thead>
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Table 2. Models which predicted methylation of the 9 genes, LINE-1 and PCA1.

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<th>$R^2$</th>
<th>2CV % Explained Variance (95% CI)</th>
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<td>9</td>
<td>0.189</td>
<td>5.0 (0.7 – 12.7)</td>
</tr>
<tr>
<td>APC</td>
<td>185</td>
<td>vitD, FI, age</td>
<td>7</td>
<td>0.048</td>
<td>-</td>
</tr>
<tr>
<td>SFRP1</td>
<td>185</td>
<td>age, sex*RCF, sex, RCF, MC</td>
<td>10</td>
<td>0.197</td>
<td>4.3 (0.4 – 11.6)</td>
</tr>
<tr>
<td>SFRP2</td>
<td>185</td>
<td>age, FI</td>
<td>9</td>
<td>0.104</td>
<td>3.9 (0.3 – 11.0)</td>
</tr>
<tr>
<td>SOX17</td>
<td>185</td>
<td>age, sex*age, sex, WC</td>
<td>9</td>
<td>0.167</td>
<td>5.6 (0.9 – 13.6)</td>
</tr>
<tr>
<td>LINE-1</td>
<td>185</td>
<td>hip, sex*height, Se, BMI, vitB12, vitD, height, sex</td>
<td>8$^a$</td>
<td>0.142</td>
<td>-</td>
</tr>
<tr>
<td>ESR1</td>
<td>174</td>
<td>age, sex*height, height, sex</td>
<td>9</td>
<td>0.181</td>
<td>8.0 (1.9 – 17.1)</td>
</tr>
<tr>
<td>MYOD</td>
<td>174</td>
<td>age, PF, vitD</td>
<td>8</td>
<td>0.170</td>
<td>7.1 (1.5 – 15.9)</td>
</tr>
<tr>
<td>NS3</td>
<td>174</td>
<td>age, sex<em>PF, sex</em>waist, sex, waist, PF</td>
<td>3</td>
<td>0.131</td>
<td>-</td>
</tr>
<tr>
<td>PCA1</td>
<td>174</td>
<td>age, PF, sex*age, sex, Se, vitD</td>
<td>9$^a$</td>
<td>0.380</td>
<td>21.2 (11.2 – 32.5)</td>
</tr>
</tbody>
</table>

n, number of subjects in dataset. ANCOVA, analysis of covariance, an * between co-factors indicates a sex-interaction. Terms ordered by their significance in the model: MC, monocytes; Se, selenium; WC, white cells; BMI, body mass index; PF, plasma folate; vitD, vitamin D; FI, fatness index; RCF, red cell folate; vitB12, vitamin B12.

n1CV, number of occurrences of the model in 10 epochs, $^a$ denotes 1 tie.

$R^2$, square correlation between actual and predicted values (no cross-validation).

2CV % Explained Variance: estimate based on double cross-validation of variance explained if model was applied to new cohort. Missing entries signify the model failed double cross-validation.
Figure Legends

Figure 1. Summary descriptions of methylation values for the 9 genes and for LINE-1.

Figure 2. Pearson correlation coefficients for the associations between the exposure variables and the methylation of the 9 genes, LINE-1 and PCA1 for all subjects (grey), males only (white) and females only (black).

Figure 3. PCA1 methylation in the male (Δ) and female (○) subjects calculated from the CGI methylation values for the 9 genes (open symbols) and predicted from the GA model (filled symbols).
Figure 3

![Graph showing methylation A.U. vs. age years](image-url)