Transcriptional changes in prostate of men on active surveillance after a 12-mo glucoraphanin-rich broccoli intervention—results from the Effect of Sulforaphane on prostate CAncer PrEvention (ESCAPE) randomized controlled trial

Maria H Traka,1 Antonietta Melchini,1 Jack Coode-Bate,1,2 Omar Al Kadhi,1,2 Shikha Saha,1 Marianne Defernez,1 Perla Troncoso-Rev,1 Helen Kibblewhite,1 Carmel M O’Neill,1 Federico Bernuzzi,1 Laura Mythen,1 Jackie Hughes,1 Paul W Needs,1 Jack R Dainty,4 George M Savva,1 Robert D Mills,2 Richard Y Ball,3 Colin S Cooper,4 and Richard F Mithen1,5

1Quadram Institute Bioscience, Norwich, United Kingdom; 2Department of Urology, Norfolk and Norwich University Hospitals NHS Foundation Trust, Norwich, United Kingdom; 3Norfolk and Waveney Cellular Pathology Service, Norfolk and Norwich University Hospitals NHS Foundation Trust, Norwich, United Kingdom; 4Norwich Medical School, University of East Anglia, Norwich, United Kingdom; and 5Liggins Institute, University of Auckland, New Zealand

ABSTRACT

Background: Epidemiological evidence suggests that consumption of cruciferous vegetables is associated with reduced risk of prostate cancer progression, largely attributed to the biological activity of glucosinolate degradation products, such as sulforaphane derived from glucoraphanin. Because there are few therapeutic interventions for men on active surveillance for prostate cancer to reduce the risk of cancer progression, dietary approaches are an appealing option for patients.

Objective: We evaluated whether consumption of a glucoraphanin-rich broccoli soup for 1 y leads to changes in gene expression in prostate tissue of men with localized prostate cancer.

Methods: Forty-nine men on active surveillance completed a 3-arm parallel randomized double-blinded intervention study for 12 mo and underwent transperineal template biopsy procedures and dietary assessment at the start and end of the study. Patients received a weekly 300 mL portion of soup made from a standard broccoli (control) or from 1 of 2 experimental broccoli genotypes with enhanced concentrations of glucoraphanin, delivering 3 and 7 times that of the control, respectively. Gene expression in tissues from each patient obtained before and after the dietary intervention was quantified by RNA sequencing followed by gene set enrichment analyses.

Results: In the control arm, there were several hundred changes in gene expression in nonneoplastic tissue during the 12 mo. These were associated with an increase in expression of potentially oncogenic pathways including inflammation processes and epithelial–mesenchymal transition. Changes in gene expression and associated oncogenic pathways were attenuated in men on the glucoraphanin-rich broccoli soup in a dose-dependent manner. Although the study was not powered to assess clinical progression, an inverse association between consumption of cruciferous vegetables and cancer progression was observed.

Conclusion: Consuming glucoraphanin-rich broccoli soup affected gene expression in the prostate of men on active surveillance, consistent with a reduction in the risk of cancer progression. This trial was registered at clinicaltrials.gov as NCT01950143.

Keywords: active surveillance, broccoli, dietary intervention, transcriptome, RNA sequencing, SMCSO, sulforaphane, prostate biopsy, cancer prevention

Introduction

The diagnosis of organ-confined prostate cancer has increased owing to routine prostate specific antigen (PSA) testing and an...
aging population. Up to 48% of cases may exhibit clinical progression after subsequent examination, and a small proportion of these may become metastatic with associated poor prognosis (1, 2). However, owing to the risks associated with curative treatment, men with organ-confined prostate cancer may choose a program of “active surveillance,” in which radical treatment is delayed until there is evidence of cancer progression. Currently, there are no approved therapeutic interventions for men who have chosen a program of active surveillance that may reduce the risk of cancer progression.

Epidemiological studies have suggested a negative association between consumption of cruciferous vegetables and incidence or progression of prostate cancer (3–5). The protective activity has been associated with the biological activity of degradation products of glucosinolates, sulfur-containing glycosides that accumulate in these vegetables. When consumed, glucosinolates are degraded either due to the action of plant-derived thioglcosidases or, if these have been denatured as a result of cooking, by microbial activity in the colon (6). Glucosinolates with aliphatic or aromatic side chains produce isothiocyanates, such as sulforaphane derived from 4-methylsulphonylbutyl glucosinolate (glucoraphanin) that accumulates in broccoli. Glucosinolates with indole side chains produce indole-3-carbinol and associated metabolites (Supplemental Figure 1A and B) (7). These glucosinolate-derived metabolites exhibit a range of biological activity in model systems consistent with the protective effects of cruciferous vegetables (8). However, despite the large number of studies with model systems, there are few examples of human intervention studies with either biological or clinical endpoints to provide further evidence that diets rich in glucosinolates, glucoraphanin, or sulforaphane may prevent prostate cancer progression. Cruciferous vegetables also accumulate S-methyl cysteine sulfoxide (SMCSO) which, in an analogous manner to glucosinolates, degrades to bioactive metabolites (9, 10) (Supplemental Figure 1C).

We report a double-blinded randomized controlled trial to test the hypothesis that a diet rich in glucoraphanin, the glycosylated precursor of sulforaphane, would significantly modify gene and metabolite expression in the prostate of men on active surveillance for prostate cancer. We used broccoli cultivaries specifically developed to have enhanced concentrations of glucoraphanin through the introgression of either 1 or 2 alleles of the Myb28 transcription factor from the wild species Brassica villosa but with otherwise identical chemical profiles (11, 12). The primary outcome of the study was to detect changes in gene expression in response to glucoraphanin-rich diets through RNA sequencing from prostate biopsies, which were collected at the start of the study and after the 12-mo intervention. The secondary outcome was to analyze metabolites from these biopsies.

We analyzed sequential transperineal template prostate biopsy samples from prostate cancer patients immediately before (T0) and after (T12) a 12-mo intervention with a broccoli soup made from 1 of the 3 broccoli genotypes, and reported paired analyses of global gene expression, gene set enrichment analyses (GSEAs), and metabolite profiles (i.e., at T0 and T12) for each of the volunteers. As several studies have reported interactions between diet and Glutathione S-transferase mu1 (GSTM1) genotype, we also investigated whether GSTM1 genotype may affect the response to the dietary intervention. Finally, we quantified the correlation between the intake of individual food components and the clinical parameters of the patient cohort.

Methods

Ethics

The study (NCT01950143) was approved by the Quadram Institute Bioscience Human Research Governance Committee and by the National Research Ethics Service (Research Ethics Committee ref: 13/EE/0110).

Study design

Effect of Sulforaphane on prostate CAnCer PrEvention (ESCAPE) was a randomized, double-blinded 3-arm parallel intervention recruiting men aged 18–80 y with a BMI between 19.5 and 35 kg/m2. The men had a diagnosis of low-risk prostate cancer (PSA < 10 ng/mL, Gleason grade 6; T category T1 or T2) or intermediate-risk prostate cancer (PSA 10–20 ng/mL, Gleason 7, including selected 4 + 3 cases that made informed decisions against radical treatment; T category T1 or T2) and were undergoing active surveillance. Complete eligibility and exclusion criteria are detailed in Supplemental Table 1. The primary outcome was gene expression of prostate tissue obtained before and after a dietary intervention and the secondary outcome was changes in metabolites. The study was powered based upon data obtained from a previous pilot study (13). The number of volunteers necessary to report statistically significant changes in gene expression was calculated by 2 methods: firstly, by using the “Sample Size for Microarray Experiments” tool developed by the Section of Bioinformatics of the University of Texas MD Anderson Cancer Center (https://biostatistics.mdanderson.org/MicroarraySampleSize); and secondly, by reported calculations based on previously published data (14). We estimated that 26 subjects in each of the 3 dietary groups (78 in total) were required to detect 1.5-fold differences with a significant difference (P < 0.02) between any 2 of the 3 dietary groups, with a power of 80% and an SD of 0.66 (based on a log2 scale of gene intensity measurements). However, the accrual rate was below that anticipated and recruitment was stopped before reaching the target sample size goal in order to complete the study within the scheduled date of closure (October, 2016). Patients (n = 61) were recruited through the Urology Department of the Norfolk and Norwich University Hospitals NHS Foundation Trust from October, 2013 to October, 2015.

Study patients were randomly allocated to 1 of 3 dietary arms in which they were required to consume 1 portion of broccoli soup (300 mL) per week as part of their normal diet for 12 mo, with an option to continue the intervention for a further 12 mo. For an optional extension (12–24 mo), patients underwent regular blood analyses, as described below, but no additional study biopsies were collected. Block randomization (www.randomization.com) and blinding were performed by an individual who was not part of the study team. The soups were manufactured by Bakkavor from the commercial cultivar Iron (soup X, genotype Myb28 B/B), the cultivar Beneforte (soup Y, Myb28 B/V, in which V represents an introgressed Myb28 allele from B. villosa), or a noncommercial hybrid cultivar (soup Z, Myb28 V/N). Three-hundred-milliliter portions of soups
manufactured from these 3 genotypes contained 72 ± 2.8 (soup X), 214 ± 7.3 (soup Y), and 492 ± 3.2 (soup Z) μmol 4-methylsulphinylbutyl glucosinolate (glucoraphanin). A previous study reported that these different soup products resulted in contrasting concentrations of sulforaphane in the systemic circulation (12). One weekly portion of soup X was assigned as the control arm, because this soup was manufactured from a commercial cultivar of broccoli and could be expected to be part of a normal diet. This soup provided the lowest concentration of glucoraphanin, which, and on the basis of epidemiological studies (3), would be considered beneath the threshold required for a reduction in cancer progression.

Patients underwent 2 transperineal template biopsy (TTB) procedures, 1 at the start of the intervention (T0) and 1 after 12 mo (T12). Of the 24–56 TTB cores per patient obtained at each TTB, several were individually reserved either in RNA later for RNA sequencing or in extraction solvent (80% HPLC grade methanol:20% water) for targeted and nontargeted metabolite analyses. Two further cores were snap frozen, and the remainder underwent routine histopathological examination. The tissue was examined by a single consultant histopathologist with a special interest in prostate pathology to reduce interobserver error, a potential hazard in diagnosing and grading prostate cancer (15). After histopathology, cores were selected for RNA sequencing analyses that were adjacent to cores that did not contain cancer.

**Dietary analyses**

Patients completed a comprehensive 7-d diet diary immediately before the study to assess their habitual diet, and subsequently at 6 and 12 mo. Diet data were analyzed through DietPlan6 (Forestfield Software Ltd, UK) and combined with additional data on the chemical composition of cruciferous and alliaceous vegetables, obtained from analyses of vegetables purchased in retail outlets in the localities of the volunteers, as previously described (16).

**Gene expression analyses by RNA sequencing**

The primary outcome of the study was to detect changes in gene expression in response to the dietary intervention through RNA sequencing, from prostate biopsies collected at the start of the study and at 12 mo, i.e., after the intervention. Histology of the directly adjacent region confirmed that the prostate biopsies used for RNA sequencing were unlikely to contain neoplastic tissue. Cores of between 3 and 10 mg from each patient were homogenized with a QIAGEN TissueRuptor before total RNA was extracted with the QIAGEN RNEasy Mini kit. The resulting RNA was quality checked with an Agilent Bioanalyzer and samples with RIN values >7 were further processed. Samples were ribodepleted with the Ribo-Zero Magnetic Gold rRNA Removal Kit (Illumina) before constructing Illumina barcoded TruSeq RNA libraries. Sequencing of 98 libraries was performed on an Illumina HiSeq 2500/2000 in high-output mode using 125-bp paired-end reads, generating 50–70 million reads/library. RNA-seq reads were first processed by removing Illumina adapters using TrimmGalore version 0.4.2 (Babraham Bioinformatics) and reads with Phred quality of basecalls >20 and with a length of >60 bp were carried forward. SortMeRNA version 2.1 (17) was used to filter any remaining ribosomal RNA from the adaptor and quality trimmed reads.

Reads were analyzed using the HISAT2-StringTie pipeline (18) aligned to the Ensembl GRCh38.89 reference genome (HISAT2 version 2.0.5 and StringTie version 1.3.3), and gene counts were exported into edgeR in R Bioconductor (19). One patient was removed from all analyses because he underwent prostatectomy at 12 mo, rather than TTB, leaving a total of 48 patients (96 libraries). The complete bioinformatics pipeline, differential gene expression (DGE) analyses, and statistical analysis are available as a GitHub repository (https://github.com/quadram-institute-bio science/ESCAPE_RNAseq_analysis).

In addition, we conducted a query on publicly available RNA sequencing data that were generated using the same Illumina HiSeq 2000/2500 technology as our samples, had ≥100-bp paired reads, and contained benign as well as primary cancer samples. We identified accession GSE80609 in Gene Expression Omnibus (https://www.ncbi.nlm.nih.gov/geo) and downloaded the raw reads (.fastq) before analyzing them with the same pipeline as our samples, described above. When ESCAPE and GSE80609 samples were put together in edgeR we generated multidimensional scaling (MDS) plots for the second and third dimensions, because the first dimension was clearly only differentiating the 2 separate studies (data not shown).

RNA sequencing data from the ESCAPE study have been deposited in ArrayExpress (accession E-MTAB-6525).

**Metabolomics analyses**

Desiccated methanol extracts, derived from 24-h incubation of 1 prostate biopsy per patient, were sent to Metabolon Inc. to undergo ultra-HPLC–mass spectroscopy (MS) and gas chromatography–MS with a high resolution accurate mass (HRAM) platform as previously described (20) (www.metabolon.com). A total of 448 metabolites were semiquantified on the basis of ion count within several biologically relevant classes (amino acids, carbohydrates, vitamins, lipids, nucleotides, peptides, tricarboxylic acid cycle, and xenobiotics). Histology of the core after incubation confirmed the absence of cancerous foci.

**Blood analyses**

Biomarkers of liver and kidney function and full blood count were quantified at T0 and T12 to ensure the glucoraphanin-rich soup had no toxic effects. Fasting blood glucose, PSA, and serum lipid profile (cholesterol, HDL cholesterol, LDL cholesterol, triglyceride) were quantified at 3-mo intervals, up to 24 mo from the start of the study. GSTM1 genotype was quantified as previously described (13).

**Statistical analyses**

Analytical of clinical characteristics.

The difference between clinical parameters (age, BMI, PSA) between the 3 different groups at the start and at the end of the study was assessed either by using ANOVA and correcting for multiple testing by Tukey’s multiple correction test, or by Kruskal–Wallis test corrected for multiple testing by Dunn’s, where appropriate.
DGE analyses.

Paired DGE analysis after calculation of normalized gene counts was undertaken in limma after voom transformation (21). Adjustment for multiple testing was performed using the Benjamin–Hochberg false discovery rate (FDR) method. MDS plots generated in EdgeR were used to determine the variation within different groups at the start and the end of the study. Statistical significance of unadjusted and FDR-adjusted $P$ values was reported for different thresholds.

Functional analyses.

Functional analyses of paired DGE were undertaken by the GSEA software (22) using the Hallmark gene sets (50 gene sets in total) within the available Molecular Signatures Database (MSigDB, version 6.1). DGE lists were ranked according to their $P$ value, modified using the rank–rank hypergeometric overlap (RRHO) algorithm (23). Modified $P$ values were calculated as the signed $\log_{10}$-transformed $P$ value of the paired log fold change over 12 mo for each dietary arm, with the sign denoting the direction of the change: positive for upregulated over time, and negative for downregulated over time. By using the RRHO method we explored the functional consequences of the paired changes in gene expression without being constrained by a given statistical threshold. The ranked DGE list was then submitted to GSEA and statistical significance of enriched pathways was set at an FDR-adjusted $P$ value $<0.05$. Normalized enrichment scores for each individual pathway and their associated FDR-adjusted $P$ value for each diet were reported with and without stratification by $GSTM1$ genotype. An MDS plot was generated in EdgeR to determine the variation in normalized enrichment scores for the different dietary groups stratified by $GSTM1$ genotype.

Metabolomics analyses.

Paired Student’s $t$ tests were undertaken for each individual metabolite within each dietary arm. Adjustment for multiple testing was performed using the Benjamini–Hochberg FDR method. Comparisons of log2-fold changes for each metabolite between dietary arms were made by unpaired $t$ tests with FDR correction.

Exploratory association with clinical outcomes.

Individual dietary components, calculated from the diet diaries of the patients reported at the start and the end of the study, were tested for association with histological and blood markers. This exploratory analysis was undertaken in the R environment (R Foundation) using Pearson correlations. Gleason scores were adjusted to be no lower than any previous biopsy, to correct for the undetected cancers in this cohort, and subsequently were converted to risk groups according to the WHO grade group system (24), to allow differentiation between Gleason 7 scores ($3 + 4$ or $4 + 3$), which occupy different grade groups under the WHO system.

Results

Clinical characteristics and dietary assessment of subjects

Sixty-one men on active surveillance were randomly assigned to 1 of the 3 dietary intervention arms and 49 completed the study (Figure 1). Of the 12 volunteers who did not complete the study, 5 exhibited clinical progression before the start of the dietary intervention after their first TTB, 4 withdrew consent during the study for unknown reasons, and 3 did not undergo a second TTB for either clinical or personal reasons. There were no significant differences in age, BMI, frequency of $GSTM1$ null genotypes, or PSA between the 3 groups at the start or the end of the study (Table 1). Similarly, there were no significant changes in biomarkers of kidney and liver function and metabolism over time or differences between the 3 arms of the study, during the 12-mo study (Table 1; Supplemental Table 2). Fasting blood glucose concentrations showed an initial fall over the first 6 mo of the study, and then a further fall between 12 and 18 mo (Supplemental Figure 2).

At baseline, there were no differences in the habitual diet of the volunteers between the 3 arms, and there were no changes during the 12-mo intervention period with the exception of glucoraphanin intake which, as expected, was significantly different between the 3 arms owing to the provision of the broccoli soups (Supplemental Table 3).

Gene expression profiles of nonneoplastic prostate biopsies

We first compared the transcriptional prostate signature of our cohort with that from a previous study that compared transcriptional profiles from benign prostate hyperplasia and primary prostate cancer (25). Apparent nonneoplastic biopsies from the ESCAPE patients were intermediate between the BPH samples and the prostate cancer samples, although with some overlap with the cancer samples (Figure 2). This indicates that transcriptional changes may be occurring across the whole prostate of patients on active surveillance, with some similarities with those occurring in the cancer lesions themselves, consistent with a “field effect.”

In order to determine whether the transcriptional profiles of the 3 groups at the start or the end of the study differed despite the random group assignment, we used MDS plots to assess variability and found no evidence of outliers, either in single patients or in diet groups (Supplemental Figure 3A and B).

Paired DGE analyses and GSEAs

We investigated the changes in gene expression that occurred within each arm over time. In the control group (i.e., those that received a single portion of broccoli soup made from standard broccoli, genotype Myb28 B/B, soup X) we found significant changes in gene transcription (FDR-adjusted $P < 0.1$, Table 2).

GSEA, with the use of the RRHO method that takes into account all the genes ranked by $P$ value and fold change, identified significant enrichment of pathways associated with the risk of carcinogenesis (FDR-adjusted $P < 0.05$, Table 3), including inflammatory response (Supplemental Figure 4) and epithelial–mesenchymal transition (Supplemental Figure 5).
Among the significantly enriched pathways were also those associated with androgen response (FDR-adjusted $P < 0.001$), angiogenesis (FDR-adjusted $P < 0.001$), and apoptosis (FDR-adjusted $P < 0.002$; Table 3).

When the intake of glucoraphanin was enhanced through the use of broccoli soup with genotype $Myb28 V/B$ (soup Y) or genotype $Myb28 V/V$ (soup Z), the extent of change in gene expression over time was suppressed, even at a low statistical
TABLE 1  Patient characteristics

|                     | Soup X (control; low GR) | Soup Y (intermediate GR) | Soup Z (high GR) |
|---------------------|--------------------------|--------------------------|------------------|
| n                   | 15                       | 17                       | 16               |
| n of GSTM1 (−/−)    | 9                        | 10                       | 12               |
| Age, y              | 68 ± 5                   | 66 ± 6                   | 66 ± 6           |
| BMI, kg/m²          | 26.7 ± 3.1               | 27.6 ± 3.4               | 27.7 ± 2.2       |
| Days from initial diagnosis | 309 ± 232             | 312 ± 356               | 327 ± 292        |

At diagnosis

|                     | Soup X (control; low GR) | Soup Y (intermediate GR) | Soup Z (high GR) |
|---------------------|--------------------------|--------------------------|------------------|
| PSA, µg/L            | 7.7 (5.9–7.9)            | 6.8 (5.6–8.6)            | 7.0 (5.0–9.3)    |
| Gleason score        | 10                       | 10                       | 9                |
| 3 + 3               |                           |                          |                  |
| 3 + 4               |                           |                          |                  |
| 4 + 3               |                           |                          |                  |
| Undetected          |                           |                          |                  |

At 0 mo

|                     | Soup X (control; low GR) | Soup Y (intermediate GR) | Soup Z (high GR) |
|---------------------|--------------------------|--------------------------|------------------|
| PSA, µg/L            | 7.9 (5.9–12.0)           | 7.6 (4.9–9.4)            | 5.8 (4.4–8.7)    |
| PSA density          | 0.10 (0.08–0.18)         | 0.13 (0.09–0.18)         | 0.10 (0.08–0.16) |
| Undetected          |                           |                          |                  |

At 12 mo

|                     | Soup X (control; low GR) | Soup Y (intermediate GR) | Soup Z (high GR) |
|---------------------|--------------------------|--------------------------|------------------|
| PSA, µg/L            | 9.4 (6.6–10.4)           | 7.3 (6.9–10.4)           | 7.5 (5.6–9.3)    |
| PSA density          | 0.13 (0.10–0.02)         | 0.12 (0.08–0.17)         | 0.11 (0.08–0.14) |
| Core ratio           | 7.9 (3.0–14.3)           | 3.7 (0–9.4)              | 11.6 (0–16.1)    |

1GR, glucoraphanin; GSTM1, Glutathione S-transferase M1; PSA, prostate specific antigen.

2Nonsignificant difference (i.e., all P values >0.5) between diets as determined by ANOVA adjusted by Tukey’s multiple correction test. Data shown are mean ± SD.

3Nonsignificant difference (i.e., all P values >0.2) between diets as determined from the Kruskal–Wallis test adjusted by Dunn’s multiple correction test. Data shown are median (IQR).

4PSA at diagnosis missing for 3 patients on Diet X, 1 on Diet Y, and 1 on Diet Z.

5Gleason score at diagnosis missing for 1 patient on Diet Z.

Analyses with GSTM1 stratification

Analyses of the paired gene expression in the 3 dietary arms stratified by GSTM1 null and non-nulls suggested that within intervention arms receiving soups X and Y the changes in gene expression were restricted to GSTM1 non-nulls (FDR-adjusted P < 0.1, Supplemental Table 4). GSEAs for the GSTM1 nulls and non-nulls were very similar in soup X, in terms of direction and magnitude of change of pathways, but exhibited some divergence in soups Y and Z indicative of a possible diet–gene interaction (Supplemental Table 5, Supplemental Figure 6).

Effect of intervention on nuclear factor (erythroid-derived 2)-like 2-regulated genes

Sulforaphane, which would have been derived from the glucoraphanin delivered by the 3 different soups, is a potent inducer of nuclear factor (erythroid-derived 2)-like 2 (NRF2)-regulated genes. We therefore extracted from the RNA sequencing data the expression of previously defined NRF2-target genes (26) (Supplemental Table 6). There was no evidence of a change in expression of any NRF2-regulated genes between the start and end of the dietary intervention (Benjamini–Hochberg FDR-adjusted P value <0.1).
FIGURE 2 Multidimensional scaling plot of the ESCAPE cohort prostate biopsies (ESC) alongside publicly available (GSE80609) BPH and CaP samples. BPH, benign prostatic hyperplasia; CaP, primary prostate cancer; ESCAPE, Effect of Sulforaphane on prostate CaPrention; FC, fold change.

Metabolomics analyses

Paired analyses of metabolites from tissue biopsies did not identify any significant changes in metabolites within any of the 3 dietary intervention arms. There was also no evidence for differences in fold changes in metabolites between dietary arms (data not shown).

Exploratory analysis with clinical parameters of prostate cancer progression

Ten out of 48 patients (28%) exhibited an increase in their cancer grade over the 12 mo of the study. Although our study was not powered to assess a clinical endpoint of prostate cancer progression, we undertook exploratory analysis and observed that the dietary intake of cruciferous vegetables at the start of the study (T0) was significantly inversely correlated with the change in WHO grade over the 12-mo study period (Figure 4A). This association was still apparent at T12, but not significant at \( P < 0.05 \) (data not shown).

The changes in gene expression observed in the control dietary arm were suppressed by the diets with soups with enhanced glucoraphanin, in a dose-dependent manner. Thus, consuming soup Y (\( \text{Myb28} \ B/V \)), delivering intermediate amounts of glucoraphanin, resulted in fewer changes in gene expression, whereas consuming soup Z (\( \text{Myb28} \ V/V \)), delivering the highest amounts of glucoraphanin, entirely suppressed changes in gene expression seen in the control arm (Table 2, Figure 3). Moreover, GSEA functional pathway analyses of Soup Z were markedly different to those of Soup X (Table 3).

| Table 2 | Number of genes changed over time with the different diets\(^1\) |
|---------|---------------------------------------------------------------|
|         | Soup X (\( n = 15 \)) (control; low GR)                     | Soup Y (\( n = 17 \)) (intermediate GR) | Soup Z (\( n = 16 \)) (high GR) |
| FDR-adjusted \( P \) value\(^2\) | 7 (4↑, 3↓)                                      | 0                                      | 0                                      |
| <0.05   | 96 (58↑, 38↓)                                   | 26 (20↑, 6↓)                               | 12 (8↑, 4↓)                              |
| <0.01   | 980 (520↑, 460↓)                                | 331 (224↑, 107↓)                           | 83 (46↑, 37↓)                             |
| <0.001  | 2796 (1460↑, 1336↓)                             | 1359 (783↑, 576↓)                         | 502 (277↑, 225↓)                         |

\(^1\) FDR, false discovery rate; GR, glucoraphanin. \( \uparrow \) indicate increase in gene expression. \( \downarrow \) indicate decrease in gene expression.

\(^2\) Paired \( t \) tests, adjusted for multiple testing correction by Benjamini–Hochberg.

\(^3\) Student’s paired \( t \) tests, unadjusted for multiple testing correction.

Discussion

The primary aim of the study was to test the hypothesis that enhancing glucoraphanin in the diet would result in changes in gene expression in prostate tissue of men on active surveillance that are consistent with reduction in the risk of cancer incidence or progression.

One of the challenges in seeking evidence for the protective effects of dietary components within a complex food matrix is experimental design. To test the effect of glucoraphanin within broccoli, we used 3 broccoli genotypes with contrasting glucoraphanin contents due to their \( \text{Myb28} \) genotype, enabling a randomized double-blinded dietary intervention study.

We analyzed changes in gene expression from sequential prostate tissue biopsies of the same individuals and found that in the control/placebo arm (i.e., commercially available broccoli) several hundred changes in gene expression occurred within the 12-mo period (Table 2, Figure 3). Subsequent GSEA indicated that the tissue within the control arm was likely to be at risk of carcinogenesis, with increased expression of several pathways associated with carcinogenesis or cancer progression (Table 3).

It is notable that these changes may have been occurring in tissue that was histologically normal, consistent with a “field effect” in the prostate gland, and with studies of whole-genome sequencing of noncancerous prostate tissue (27). The amount of broccoli or glucoraphanin consumed by men in this control arm was below the threshold that has been reported in epidemiological studies to reduce the risk of glucoraphanin would have induced these changes.

The changes in gene expression observed in the control dietary arm were suppressed by the diets with soups with enhanced glucoraphanin, in a dose-dependent manner. Thus, consuming soup Y (\( \text{Myb28} \ B/V \)), delivering intermediate amounts of glucoraphanin, resulted in fewer changes in gene expression, whereas consuming soup Z (\( \text{Myb28} \ V/V \)), delivering the highest amounts of glucoraphanin, entirely suppressed changes in gene expression seen in the control arm (Table 2, Figure 3). Moreover, GSEA functional pathway analyses of Soup Z were markedly different to those of Soup X (Table 3).
TABLE 3  GSEA of paired changes over time for the control arm (soup X, low GR) and the experimental arms (soup Y, intermediate GR; soup Z, high GR)\(^1\)

| MSigDb pathway                        | SIZE | NES   | FDR \(P\) value | Soup X (control; low GR) | Soup Y (intermediate GR) | Soup Z (high GR) |
|---------------------------------------|------|-------|------------------|--------------------------|--------------------------|------------------|
| TNFα signaling via NFKB               | 167  | 2.65  | 0.3             | 2.89 0.3                 | 0.3                      | 1.33 0.121       |
| Epithelial–mesenchymal transition     | 176  | 2.70  | 0.3             | 2.30 0.3                 | 0.3                      | 1.03 0.558       |
| Hypoxia                               | 164  | 1.95  | 0.3             | 1.88 0.3                 | 0.3                      | 0.97 0.655       |
| Inflammatory response                 | 152  | 1.95  | 0.3             | 2.36 0.3                 | 0.3                      | 0.95 1           |
| TGF \(\beta\) signaling              | 46   | 2.03  | 0.3             | 1.58 0.007 3            | 0.82 1                   |
| Protein secretion                     | 91   | 1.99  | 0.3             | 1.56 0.012 3            | 1.36 0.095              |
| Androgen response                     | 93   | 2.13  | 0.3             | 1.62 0.014 3            | 1.45 0.055              |
| Myogenesis                            | 173  | 2.01  | 0.3             | 1.51 0.018 3            | 2.30 0.3                 |
| UV response DN                        | 133  | 2.11  | 0.3             | 1.29 0.090              | 0.77 0.953              |
| Angiogenesis                          | 30   | 1.89  | 0.001 3         | 2.06 0.3                 | 1.61 0.016 3           |
| IL2 STAT5 signaling                   | 154  | 1.84  | 0.001 3         | 2.21 0.3                 | 1.25 0.191              |
| Coagulation                           | 89   | 1.76  | 0.001 3         | 2.08 0.3                 | 0.85 0.924              |
| Interferon-\(\gamma\) response       | 162  | 1.81  | 0.001 3         | 2.47 0.3                 | 0.77 0.931              |
| KRAS signaling UP                     | 157  | 1.81  | 0.001 3         | 1.88 0.3                 | 0.81 0.957              |
| Apoptosis                             | 140  | 1.71  | 0.002 3         | 1.99 0.3                 | 1.19 0.259              |
| Notch signaling                       | 29   | 1.68  | 0.003 3         | 1.08 0.344              | 1.10 0.851              |
| Fatty acid metabolism                 | 127  | 1.76  | 0.003 3         | 0.88 0.765              | 1.64 0.016 3           |
| IL6 JAK STAT3 signaling               | 61   | 1.63  | 0.004 3         | 2.13 0.3                 | 0.66 0.989              |
| Unfolded protein response             | 98   | 1.68  | 0.004 3         | 1.49 0.020 3            | 1.58 0.018 3           |
| Cholesterol homeostasis               | 63   | 1.72  | 0.004 3         | 0.73 0.956              | 1.43 0.062              |
| Apical junction                       | 163  | 1.60  | 0.005 3         | 2.04 0.3                 | 1.23 0.206              |
| Peroxisome                            | 82   | 1.63  | 0.008 3         | 0.99 0.504              | 1.02 0.567              |
| Complement                            | 144  | 1.44  | 0.025 3         | 1.96 0.3                 | 0.97 1                  |
| Mitotic spindle                       | 171  | 1.44  | 0.025 3         | 1.23 0.150              | 1.24 0.482              |
| P53 pathway                           | 171  | 1.41  | 0.033 3         | 2.05 0.3                 | 0.97 0.665              |
| Allograft rejection                   | 146  | 1.39  | 0.035 3         | 2.13 0.3                 | 1.28 0.697              |
| MYC targets V1                        | 178  | 1.44  | 0.051           | 1.48 0.021 3            | -2.11 0.03             |
| MTORC1 signaling                     | 179  | 1.42  | 0.054           | 1.35 0.066              | -1.50 0.039 3          |
| Interferon-\(\alpha\) response       | 81   | 1.27  | 0.091           | 2.00 0.3                 | 0.78 0.974              |
| Estrogen response early               | 167  | 1.19  | 0.159           | 1.99 0.3                 | 1.24 0.202              |
| DNA repair                            | 117  | 1.26  | 0.187           | 1.40 0.044 3            | -1.13 0.355             |
| Hedgehog signaling                    | 30   | 1.11  | 0.260           | 1.66 0.003 3            | -0.95 0.651             |
| UV response UP                        | 128  | 1.10  | 0.266           | 1.73 0.001 3            | -1.71 0.009 3           |
| Reactive oxygen species               | 42   | 1.14  | 0.302           | 1.32 0.077              | -1.63 0.014 3           |
| MYC targets V2                        | 51   | 1.14  | 0.323           | 1.82 0.3                 | -1.71 0.011 3           |
| Xenobiotic metabolism                 | 139  | 1.09  | 0.356           | 1.48 0.020 3            | -1.65 0.016 3           |
| Oxidative phosphorylation             | 174  | 1.08  | 0.366           | 0.79 0.907              | -2.16 0.3               |
| Estrogen response late                | 162  | 1.01  | 0.505           | 1.88 0.3                 | -1.28 0.163             |
| Adipogenesis                          | 146  | 0.93  | 0.736           | 1.10 0.328              | -1.86 0.002 3           |
| E2F targets                           | 147  | 0.86  | 0.828           | 1.42 0.036 3            | 0.93 1                  |

\(^1\)FDR, false discovery rate [as described in (22)]; E2F, E2 Factor; GR, glucoraphanin; GSEA, gene set enrichment analysis; IL6, interleukin 6; JAK, Janus Kinase; KRAS, K-Vi-Ras2 Kirsten Rat Sarcoma 2 Viral Oncogene Homolog; MSigDb, Molecular Signature Database; MTORC1, mammalian target of rapamycin complex 1; MYC, myelocytomatosis; NES, normalized enrichment score; NFkB, nuclear factor kappa B1; STAT, signal transducer and activator of transcription; TGF, transforming growth factor; TNFA, tumor necrosis factor alpha; UV, ultra violet. SIZE refers to the numbers of genes in the pathway. UV response DN refers to genes that are down regulated by UV radiation. UV response UP refers to genes that are up regulated by UV radiation.

\(^2\)GSEA by GSEA software version 3.0 on all genes ranked by the significance of fold change (see Methods section for details; http://software.broadinstitute.org/gsea).

\(^3\)Only pathways significant at FDR-adjusted \(P < 0.05\) in \(\geq 1\) of the 3 diets are shown.

Meta-analyses of epidemiological studies have associated the GSTM1 null genotype with enhanced risk of prostate cancer (28, 29) and cancer at other sites (30–32), and several epidemiological studies have reported that the beneficial effect of diets rich in cruciferous vegetables in reducing cancer risk is modified by GSTM1 genotype (33–37). Experimental human dietary intake studies with biological markers have reported greater effect of isothiocyanate intake with GSTM1 null individuals than those with 1 or 2 GSTM1 alleles (38, 39). We undertook exploratory analyses of the possible interaction between GSTM1 genotype and diet by stratifying each arm by genotype and analyzed the changes in paired gene expression and GSEA. We found that changes in gene expression only occurred in GSTM1 non null individuals. One explanation is that, as with previous studies (38, 39), GSTM1 null individuals had a greater response to sulforaphane and thus even with the low-dose glucoraphanin diet there was some attenuation of changes in gene expression. However, after GSEA, enrichment scores of pathways were similar in both GSTM1 genotypes on the control diet (Soup X; Supplemental Figure 6), indicating that if there
Glucoraphanin-rich diet and prostate transcriptome

FIGURE 3 Volcano plots of differentially expressed genes over time. Highlighted are the genes that are significantly upregulated or downregulated over time in each dietary arm, colored in red or blue, respectively. Even at a low statistical threshold we do not observe any changes in gene expression with the high-glucoraphanin soup (Soup Z) (false discovery rate–adjusted \( P \) value < 0.5, paired \( t \) tests adjusted by Benjamini–Hochberg for multiple testing correction). adj\( P \) refers to a probability value that has been adjusted for multiple testing with the use of Benjamini-Hochberg false discovery rate method.

Based upon results from cell and animal model systems we expected an intervention with glucoraphanin (and hence sulforaphane) would induce gene expression in a manner that would reduce the risk of cancer incidence or progression. In contrast, we observed a suppression of changes in gene expression. This finding was only apparent owing to our innovative experimental design, and would not have been evident if we had just compared was any effect of glucoraphanin on gene expression it was insufficient to attenuate oncogenic pathways. Similar results were found after consumption of soups Y and Z, albeit with some indication of an increasing divergence of enrichment of pathways between GSTM1 genotypes with increasing glucoraphanin content of diet, indicative of a possible diet \( \times \) gene interaction.
FIGURE 4 Analysis of dietary components. (A) Correlation matrix of the change in blood and histological markers over 12 mo, and the baseline concentrations of cruciferous vegetables and sulfur-metabolites. (B) Correlation matrix of the change in blood and histological markers over 12 mo, and the mean concentrations of cruciferous vegetables and sulfur-metabolites over the same period. Color denotes the direction of the Pearson correlation and dot size denotes the significance; only significant correlations ($P < 0.05$) are shown; numbers denote the Pearson correlation. ITC, isothiocyanate; PSA, prostate specific antigen; SMCSO, S-methyl cysteine sulfoxide. adjWHO grade refers to a WHO grade that has been adjusted to be no lower than that recorded from a previous biopsy, as described in materials and methods.

individuals at a single time point (e.g., 12 mo). There are few reports of sequential global gene expression in model systems. One example is the suppression of changes in gene expression (“transcriptional drift”) by the antidepressant miaserin in Caenorhabditis elegans (40, 41), in a somewhat analogous manner to the attenuation of changes in gene expression that was observed with the high-glucoraphanin soups. Moreover, miaserin attenuated an oxidative transcriptional signature (41), and this modulation of redox status was considered to be associated with the reduced transcriptional drift associated with ageing of C. elegans. This is analogous to the significant reduction in the enrichment score of the reactive oxygen species pathway induced by the high-glucoraphanin soup (Table 3). It is well established that sulforaphane derived from glucoraphanin induces acute oxidative stress followed by induction of NRF2-regulated genes that modulate cellular redox status (8). In our study, we did not observe any changes in expression of NRF2-regulated genes (Supplemental Table 6). This may be due to the transient nature of the changes in the expression of these genes, with changes only occurring in the few hours directly after consuming the soup. However, the regular (once-weekly) exposure to sulforaphane in the high-glucoraphanin intervention arm may result in the maintenance or improvement of redox status of the prostate tissue that inhibits the changes in gene expression associated with oncogenic pathways that were observed in the control arm.

These data suggest that the putative chemopreventive effects of a diet rich in cruciferous vegetables and glucoraphanin are not mediated by direct effects upon cancerous clones, but through a more generic “antiaging” effect. This would be consistent with the beneficial effect of a diet rich in cruciferous vegetables on other chronic age-related diseases (42, 43). Alternatively, the effects of our intervention could be through epigenetic regulation (44). Broccoli sprouts and sulforaphane have been shown to reduce prostate cancer incidence through reduction of histone deacetylation 3 (HDAC3) in mice, and altering global DNA methylation in prostate cell models (45, 46). Despite these speculations, there is clearly a “mechanistic gap” between the phenomena observed in model systems, that often involve short-term high-dose exposures, and those observed in human studies for which there are several ethical and clinical constraints in study design. This may be partially resolved through improved experimental design in model systems that use longer and lower-dose interventions and sequential analyses of tissues, and more innovative human studies involving analyses of biopsy tissues after precisely timed dietary interventions.

One of the limitations of our trial was the relatively small sample size, resulting from the low accrual rate of eligible patients, and we did not meet our target recruitment to achieve the original power estimation. Having fewer patients in each arm may have decreased the number of genes identified as being differentially expressed at a fixed FDR-adjusted $P$ value and moderated the GSEA, but each arm would have been affected equally and it is not likely that under-recruitment could explain the differences between groups that we observed. Obtaining and analyzing sequential paired prostate biopsy samples from the patients rendered our data less susceptible to interindividual variability, thus partly compensating for the reduced sample size. Another limitation was that the biopsies analyzed were all considered nonneoplastic, based on directly adjacent histology. Although this assumption may be erroneous for some of the biopsies, the global transcriptional profiles of all the biopsies within our cohort were more similar to the profiles of primary prostate cancer, suggesting that the whole prostate undergoes transcriptional changes at the onset of prostate cancer.
In conclusion, our data are entirely consistent with epidemiological studies that inversely correlate diets rich in either cruciferous vegetables or glucosinolates with prostate cancer incidence or progression. We report that an intervention rich in glucoraphanin attenuated the transcriptional changes occurring in prostate of men on active surveillance over a period of 12 mo. Although our study was not designed or sufficiently powered to quantify clinical endpoints, we also observed a negative correlation between the intake of cruciferous vegetables and their associated sulfur-metabolites, and the change in WHO grade over time (Figure 4). Further studies are warranted to explore this association, with sufficient volunteer numbers and appropriate follow-up time to assess clinical endpoints in an active surveillance cohort. The results of the study would support a public health recommendation to include cruciferous vegetables as part of the diet to maintain and promote health.

We thank David Tomlinson and Jocelyn Keshet-Price from the Norfolk and Norwich University Hospitals NHS Foundation Trust Clinical Research and Trials Unit; Tina Brown and Barbara Sparks for help with booking urology patients; Roxanne Brunton-Sim and Susan Steel from the Norwich Biorepository; Hannah Woolley, Anne Vallins, Charlotte Arnah, and Joanne Doleman for their assistance with soup storage and delivery; and the nursing staff Aliceon Blair, Bridget Shobrook, and Anne Legg for their assistance with the study. We also thank Bakkavor for producing the broccoli soups. This research was also supported in part by the NBI Computing infrastructure for Science (CIS) group through High Performance Computing infrastructure, storage, and application support.

The authors’ contributions were as follows—RFM, MHT, CSC, and AM: designed the ESCAPE study; AM: coordinated the ESCAPE study; JC-B, OAK, and RDM: were the urological clinical leads; RYB: examined the prostate biopsies as a consultant histopathologist; MHT: undertook prostate biopsy sample processing; MHT and PF-R: undertook the bioinformatics analysis of the ESCAPE study; FB: genotyped volunteers for GSTM1; SS, PWN, and JC-B: undertook the MS-based quantification of metabolites; MD and RFM: undertook the analysis of metabolomics data; AM, HK, CMO, LM, and JH: liaised with volunteers, and collected and analyzed diet data; JRD and GMS: provided statistical expertise; RFM and MHT: wrote the paper with contributions from all authors; RFM: had primary responsibility for the final content; and all authors: read and approved the final manuscript. RFM is an inventor in a patent for the development of broccoli with elevated glucoraphanin. RFM, MHT, and AM are co-inventors in 2 patents that cover combinations of a composition comprising glucoraphanin and SMCCSO for the treatment or prevention of prostate cancer. None of the other authors reported a conflict of interest related to the study.

References

1. Iremashvili V, Soloway MS, Rosenberg DL, Manoharan M. Clinical and demographic characteristics associated with prostate cancer progression in patients on active surveillance. J Urol 2012;187(5):1594–9.

2. Fleschner NE, Lucia MS, Egerdie B, Aaron L, Eure G, Nandy I, Black L, Rittmaster RS. Dutasteride in localised prostate cancer management: the REDEEM randomised, double-blind, placebo-controlled trial. Lancet 2012;379(9821):1103–11.

3. Richman EL, Carroll PR, Chan JM. Vegetable and fruit intake after diagnosis and risk of prostate cancer progression. Int J Cancer 2012;131(1):201–10.

4. Giovannoni E, Rimm EB, Liu Y, Stampfer MJ, Willett WC. A prospective study of cruciferous vegetables and prostate cancer. Cancer Epidemiol Biomarkers Prev 2003;12(12):1403–9.

5. Liu B, Mao Q, Cao M, Xie L. Cruciferous vegetables intake and risk of prostate cancer: a meta-analysis. Int J Urol 2012;19(2):134–41.

6. Saha S, Hollands W, Teucher B, Needs PW, Narbad A, Ortori CA, Barrett DA, Rossiter JT, Mithen RF, Kroon PA. Isothiocyanate concentrations and interconversion of sulforaphane to erucin in human subjects after consumption of commercial frozen broccoli compared to fresh broccoli. Mol Nutr Food Res 2012;56(12):1906–16.

7. Mithen R. Glucosinolates – biochemistry, genetics and biological activity. Plant Growth Regul 2001;34(1):91–103.

8. Juge N, Mithen RF. Molecular basis for chemoprevention by sulforaphane: a comprehensive review. Cell Mol Life Sci 2007;64(9):1105–27.

9. Edmonds WM, Beckonert OP, Stella C, Campbell A, Lake BG, Lindon JC, Holmes E, Gooderham NJ. Identification of human urinary biomarkers of cruciferous vegetable consumption by metabolomic profiling. J Proteome Res 2011;10(10):4513–21.

10. Marks HS, Hilson JA, Leichtweiss HC, Stoewsand GS. S-methylcysteine sulfoxide in brassica vegetables and formation of methyl methanethiosulfonate from Brussels-sprouts. J Agric Food Chem 1992;40(11):2098–101.

11. Traka MH, Saha S, Huseby S, Kopriva S, Walley PG, Barker GC, Moore J, Mero G, van den Bosch F, Constant H, et al. Genetic regulation of glucoraphanin accumulation in Beneforté broccoli. New Phytol 2013;198(4):1085–95.

12. Sivapalan T, Melchini A, Saha S, Needs PW, Traka MH, Tapp H, Dainty JR, Mithen RF. Bioavailability of glucoraphanin and sulforaphane from high-glucoraphanin broccoli. Mol Nutr Food Res 2018;62(18):1700991.

13. Traka M, Gasper AV, Melchini A, Bacon JR, Needs PW, Frost V, Chantry A, Jones AM, Ortori CA, Barrett DA, et al. Broccoli consumption interacts with GSTM1 to perturb oncogenic signalling pathways in the prostate. PLoS One 2008;3(7):e2568.

14. Wei C, Li J, Bumgarner RE. Sample size for detecting differentially expressed genes in microarray experiments. BMC Genomics 2004;5:87.

15. Goodman M, Ward KC, Osunkoya AO, Datta MW, Luthringer D, Young AN, Marks K, Cohen V, Kennedy JC, Haber MJ, et al. Frequency and determinants of disagreement and error in Gleason scores: a population-based study of prostate cancer. Prostate 2012;72(13):1389–98.

16. Doleman JF, Grisar K, Van Liedekerke L, Barrett CA, Barrett T, Mitchell M, Milgram E. Integrated, non-targeted ultrahigh performance liquid chromatography/electrospray ionization tandem mass spectrometry platform for the identification and relative quantification of the small-molecule complement of biological systems. Anal Chem 2009;81(16):6656–67.

17. Ritchie ME, Phipson B, Wu D, Hu Y, Law CW, Shi W, Smyth GK. limma powers differential expression analyses for RNA-sequencing and microarray studies. Nucleic Acids Res 2015;43(7):e47.

18. Subramanian A, Tamayo P, Mootha VK, Mukherjee S, Ebert BL, Gillette MA, Pomerovich A, Golub TR, Lander ES, et al. Gene set enrichment analysis: a knowledge-based approach for interpreting genome-wide expression profiles. Proc Natl Acad Sci U S A 2005;102(43):15545–50.

19. Robinson MD, McCarthy DJ, Smyth GK. edgeR: a Bioconductor package for differential expression analysis of digital gene expression data. Bioinformatics 2010;26(1):139–40.

20. Evans AM, DeHaven CD, Barrett T, Mitchell M, Miligram E. Integrated, non-targeted ultrahigh performance liquid chromatography/electrospray ionization tandem mass spectrometry platform for the identification and relative quantification of the small-molecule complement of biological systems. Anal Chem 2009;81(16):6656–67.

21. Plaisier SB, Taschereau R, Wong JA, Graebner TG. Rank-rank hypergeometric overlap: identification of statistically significant overlap between gene-expression signatures. Nucleic Acids Res 2010;38(17):e169.

22. Schreiber AD, Zoloria SD, McBride A, Reuter VE, Fine SW, et al. A contemporary prostate cancer grading system: a validated alternative to the Gleason score. Eur Urol 2016;69(3):428–35.

23. Yun SJ, Kim SK, Kim J, Cha EJ, Kim JS, Kim SJ, Ha YS, Kim YH, Jeong P, Kang HW, et al. Transcriptomic features of primary prostate cancer and their prognostic relevance to castration-resistant prostate cancer. Oncotarget 2017;8(70):114845–55.

24. Hayes JD, Dinkova-Kostova AT. The Nrf2 regulatory network provides an interface between redox and intermediary metabolism. Trends Biochem Sci 2014;39(4):199–218.

Reports Glucoraphanin-rich diet and prostate transcriptome 1143
27. Cooper CS, Ecles R, Wedge DC, Van Loo P, Gundem G, Alexandrov LB, Kremeyer B, Butler A, Lynch AG, Camacho N, et al. Analysis of the genetic phylogeny of multifocal prostate cancer identifies multiple independent clonal expansions in neoplastic and morphologically normal prostate tissue. Nat Genet 2015;47(4):367–72.

28. Gong M, Dong W, Shi Z, Xu Y, Ni W, An R. Genetic polymorphisms of GSTM1, GSTT1, and GSTP1 with prostate cancer risk: a meta-analysis of 57 studies. PLoS One 2012;7(11):e50587.

29. Malik SS, Kazmi Z, Fatima I, Shabbir S, Masood N. Genetic polymorphism of GSTM1 and GSTT1 with prostate cancer risk: a meta-analysis of 57 studies. PLoS One 2012;7(11):e50587.

30. Yu P, Kusuma JD, Suarez MAR, Pamela Koong Shiao S-Y. Lung cancer susceptibility from GSTM1 deletion and air pollution with smoking status: a meta-prediction of worldwide populations. Oncotarget 2018;9(57):3120–32.

31. Hussain T, Alrokayan S, Upasna U, Pavithrakumari M, Jayapriya J, Kutala VK, Naushad SM. Meta-analysis of genetic polymorphisms in xenobiotic metabolizing enzymes and their association with breast cancer risk. J Genet 2018;97(2):523–37.

32. Huang M, Zeng Y, Zhao F, Huang Y. Association of glutathione S-transferase M1 polymorphisms in the colorectal cancer risk: a meta-analysis. J Cancer Res Ther 2018;14(1):184–95.

33. Lam TK, Gallicchio L, Lindsley K, Shiels M, Hammond E, Tao XG, Chen L, Robinson KA, Caulfield LE, Herman JG, et al. Cruciferous vegetable consumption and lung cancer risk: a systematic review. Cancer Epidemiol Biomarkers Prev 2009;18(18):9:11156–26.

34. Lin J, Kamat A, Gu J, Chen M, Dinney CP, Forman MR, Wu X. Dietary intake of vegetables and fruits and the modification effects of GSTM1 and NAT2 genotypes on bladder cancer risk. Cancer Epidemiol Biomarkers Prev 2009;18(7):1768–83.

35. Moy KA, Yuan JM, Chung FL, Wang XL, Van Den Berg D, Wang R, Gao YT, Yu MC. Isothiocyanates, glutathione S-transferase M1 and T1 polymorphisms and gastric cancer risk: a prospective study of men in Shanghai, China. Int J Cancer 2009;125(11):2692–9.

36. Wang LI, Giovannucci EL, Hunter D, Neuberg D, Su L, Christiani DC. Dietary intake of cruciferous vegetables, glutathione S-transferase (GST) polymorphisms and lung cancer risk in a Caucasian population. Cancer Causes Control 2004;15(10):1007–17.

37. Joseph MA, Moysich KB, Freudenheim JL, Shields PG, Bowman ED, Zhang Y, Marshall JR, Ambrosone CB. Cruciferous vegetables, genetic polymorphisms in glutathione S-transferases M1 and T1, and prostate cancer risk. Nutr Cancer 2004;50(2):206–13.

38. Yuan JM, Murphy SE, Stepanov I, Wang R, Carmella SG, Nelson HH, Hataksami D, Hecht SS. 2-Phenethyl isothiocyanate, glutathione S-transferase M1 and T1 polymorphisms, and detoxification of volatile organic carcinogens and toxicants in tobacco smoke. Cancer Prev Res (Phila) 2016;9(7):598–606.

39. Navarro SL, Chang JL, Peterson S, Chen C, King IB, Schwarz Y, Li SS, Li L, Potter JD, Lampe JW. Modulation of human serum glutathione S-transferase A1/2 concentration by cruciferous vegetables in a controlled feeding study is influenced by GSTM1 and GSTT1 genotypes. Cancer Epidemiol Biomarkers Prev 2009;18(11):2974–8.

40. Petrascheck M, Ye X, Buck LB. An antidepressant that extends lifespan in adult Caenorhabditis elegans. Nature 2007;450(7169):553–6.

41. Rangaraju S, Solis GM, Thompson RC, Gomez-Amaro RL, Kuriang L, Encalada SE, Niculescu AB, III, Salomon DR, Petrascheck M. Suppression of transcriptional drift extends C. elegans lifespan by postponing the onset of mortality. Elife 2015;4:e08833.

42. Blekkenhorst LC, Bondonno CP, Lewis JR, Devine A, Zhu K, Lim WH, Woodman RJ, Belin LI, Prince RL, Hodgson JM. Cruciferous and allium vegetable intakes are inversely associated with 15-year atherosclerotic vascular disease deaths in older adult women. J Am Heart Assoc 2017;6(10):e006558.

43. Nurk E, Refsum H, Drevon CA, Tell GS, Nygaard HA, Engedal K, Smith AD. Cognitive performance among the elderly in relation to the intake of plant foods. The Hordaland Health Study. Br J Nutr 2010;104(8):1190–201.

44. Gerhauser C. Epigenetic impact of dietary isothiocyanates in cancer chemoprevention. Curr Opin Clin Nutr Metab Care 2013;16(4):405–10.

45. Beaver LM, Löhrl CV, Clarke JD, Glasser ST, Watson GW, Wong CP, Zhang Z, Williams DE, Dashwood RH, Shannon J, et al. Broccoli sprouts delay prostate cancer formation and decrease prostate cancer severity with a concurrent decrease in HDAC3 protein expression in transgenic adenocarcinoma of the mouse prostate (TRAMP) mice. Curr Dev Nutr 2018;2(3):ezy002.

46. Wong CP, Hsu A, Buchanan A, Palomera-Sanchez Z, Beaver LM, Houseman EA, Williams DE, Dashwood RH, Ho E. Effects of sulforaphane and 3,3′-diindolylmethane on genome-wide promoter methylation in normal prostate epithelial cells and prostate cancer cells. PLoS One 2014;9(1):e86787.