Rare disruptive mutations and their contribution to the heritable risk of colorectal cancer

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Colorectal cancer (CRC) displays a complex pattern of inheritance. It is postulated that much of the missing heritability of CRC is enshrined in high-impact rare alleles, which are mechanistically and clinically important. In this study, we assay the impact of rare germline mutations on CRC, analysing high-coverage exome sequencing data on 1,006 early-onset familial CRC cases and 1,609 healthy controls, with additional sequencing and array data on up to 5,552 cases and 6,792 controls. We identify highly penetrant rare mutations in 16% of familial CRC. Although the majority of these reside in known genes, we identify POT1, POLE2 and MRE11 as candidate CRC genes. We did not identify any coding low-frequency alleles (1–5%) with moderate effect. Our study clarifies the genetic architecture of CRC and probably discounts the existence of further major high-penetrance susceptibility genes, which individually account for >1% of the familial risk. Our results inform future study design and provide a resource for contextualizing the impact of new CRC genes.
Colorectal cancer (CRC), a major cause of cancer-related mortality, displays a complex pattern of inheritance. The genetic architecture of CRC susceptibility encompasses a broad spectrum of risk, from rare, highly penetrant germline mutations associated with well-characterized syndromes to common polymorphisms, each individually conferring small risks. However, much of the familial risk remains unexplained but is widely postulated to be enshrined in unidentified, rare, high-impact variants. This class of susceptibility is mechanistically important and highly relevant to the clinical management of both familial and sporadic CRC. Our recent application of next-generation sequencing to familial cases led to the discovery of mutations in POLE and POLED1 that predispose to CRC, thus providing evidence for the existence of hitherto-unidentified, rare, high/moderate-penetrance susceptibility alleles. We therefore implemented whole-exome sequencing (WES) to quantify the contribution of rare disruptive variants to the heritable risk of CRC and identify new susceptibility genes.

The high baseline rate of rare, neutral germline variants makes the identification of rare CRC predisposition alleles problematic. We therefore designed a study for the extreme phenotype of early-onset CRC, as heritability is substantially greater when diagnosed young and/or is familial. In 2007, we established the UK National Study of Colorectal Cancer Genetics (NSCCG) as a bio-repository for studying susceptibility to CRC. From 23,693 CRC cases (diagnosed <70 years) with European Ancestry recruited between 2007 and 2013, we identified 1,143 with early-onset CRC (≥55 years) documented to have at least one first-degree relative with CRC, a highly enriched subset representing <2% of all CRC. One thousand and twenty-eight of the 1,143 Discovery series of cases were subjected to WES. For comparison, we analysed WES data on 1,644 UK population controls from the 1958 Birth Cohort (1958BC) with no history of malignancy. We identify high-penetrant rare mutations in 16% of familial CRC. Although the majority of these reside in known genes, we identify POT1, POLE2 and MRE11 as candidate CRC genes. Our study clarifies the genetic architecture of CRC and probably discounts the existence of further major high-penetrance susceptibility genes.

Results

Whole-exome sequencing. Cases and controls were sequenced using Illumina TruSeq exon capture and Hi-Seq 2000 technology (Fig.1 and Methods). To avoid erroneous findings, we performed alignment and variant calling of all samples simultaneously (Methods). We excluded 57 subjects with low-quality data or non-European ancestry (Supplementary Figs 1 and 2). Each captured base was sequenced to an average depth of 48 × across samples; cases and controls had similar sequencing metrics (Supplementary Table 1). We estimated sensitivity and specificity of calls using a subset of 1,332 samples, which had been genotyped using Illumina HumanExome-12v1_A Beadchips, identifying high levels of concordance (Methods). The final data set comprised 1,006 Discovery cases and 1,609 controls (Fig.1 and Supplementary Fig.1) for which the characteristics are detailed in Supplementary Table 2.

Analysis of recurrent variants. We first examined individual, moderately low-frequency coding variants (minor allele frequency (MAF) 1–5%) for an association with CRC risk. No signal deviated from that expected by chance. To maximize the detection of a statistically significant association, we performed a meta-analysis of our Discovery series with Illumina HumanExome-12v1_A Beadchip genotypes on an additional series of 5,552 cases and 6,792 controls (Fig.1 and Methods).

Gene-based analysis of rare variants. We next examined the impact of rare alleles (MAF<1%) collectively within a gene on CRC risk by aggregating single nucleotide variants (SNVs) and indels (‘T1’ test) in each gene and comparing the counts between cases and controls (Methods). Empirical P-values were obtained using permutation. Acknowledging the limitations of in-silico prediction to enrich for harmful alleles, we considered three sets of variants: Class 1, disruptive mutations (nonsense and frameshift); Class 2, predicted damaging (disruptive plus missense predicted to be damaging and splice donor/acceptor-site); and Class 3, all non-synonymous changes. To account for multiple testing we set the threshold for exome-wide significance at \(P = 8.0 \times 10^{-7}\) (Bonferroni correction for 20,000 genes and 3 classes of variants).

Figure 1 | Study design to investigate the contribution of rare disruptive mutations to the heritable risk of CRC. WES was performed on a highly enriched subset of CRC patients recruited to the NSCCG and a series of UK population controls from the 1958 Birth Cohort. After QC, this Discovery set comprised 1,006 early-onset (diagnosed ≤55 years) familial cases (≥1 first-degree relative with CRC) and 1,609 controls. To test the hypothesis that low-frequency variants confer risk for CRC, we performed a meta-analysis alongside 12,344 UK samples genotyped on the Illumina HumanExome-12v1_A Beadchip. To test the hypothesis that a burden of rare mutations in a gene confers risk for CRC, we performed a burden test on rare (<1% frequency) coding variants and subsequent meta-analysis alongside additional sequencing data from 3,770 UK samples (WGSET). Finally, we performed GSEA to investigate whether the burden of rare variants is overrepresented in a specific biological pathway.

Meta-analyses did not identify a statistically significant association for any single variant (that is, \(P > 4.0 \times 10^{-7}\)).
Class 1 variants showed the strongest enrichment with significant associations being shown for known susceptibility genes: MSH2, MLH1 and APC (Fig. 2 and Supplementary Data 1). T1 association for APC based on Class 2 variants was far less significant, emphasizing the issue of assigning pathogenicity to missense variants (Supplementary Data 2). Moreover, for well-documented CRC genes, including MSH6 and PMS2, irrespective of class, T1 associations did not attain statistical significance, reflecting also their more limited contribution to CRC susceptibility (Fig. 2 and Supplementary Data 1–3). None of the cases were identified as being carriers of multiple Class 1 mutations in a known gene (Supplementary Note 1).

To estimate the maximal contribution of known susceptibility genes to familial CRC, we relaxed sequencing filters and manually assessed additional rare variants incorporating ClinVar/InSight annotations (Methods). Nine per cent of the CRC cases could be ascribed to rare Class 1 mutations in the established high-penetrance genes. Even including the contribution of all rare non-synonymous variants, only 31% of familial CRC is explained (Fig. 3 and Supplementary Table 3).

To search for novel CRC susceptibility genes we confined analysis to the 863 cases, which did not carry a probable disease-causing mutation in an established high-penetrance CRC gene. Although no T1 association was exome-wide significant (Supplementary Data 1), we sought support for promising associations using whole-genome sequencing (WGS) data on an additional 188 familial cases (WG188) in conjunction with WGS data on 3,582 UK10K controls—WGSET (Fig. 1 and Methods). We restricted this analysis to genes with Class 1 nominally significant T1 P-values in the Discovery series. Evaluating candidate genes for support in the WGSET, the presence of recurrent variants and biological plausibility identified three novel potential CRC genes—IL12RB1, LIMK2 and POLE2 (Tables 1 and 2). The recurrent IL12RB1 truncating variants p.Gln542Ter (three WES and two WGSET cases) and p.Gln32Ter (one WES and one 1958BC control) cause recessive IL12-Rβ1 deficiency. Interestingly, another IL12RB1 truncating mutation (Gln376Ter) has been linked with intestinal gastric cancer. In LIMK2, we identified the recurrent variant p.Gln574ArgfsTer12 in five WES cases and one 1958BC control. Positive selection for decreased LIMK2 activity during CRC initiation and progression has been documented, and of the 18 LIMK2 mutations catalogued by COSMIC, 14 were identified in colorectal tumours. The recurrent POLE2 p.Leu469PhedfsTer17 variant was detected in three WES and one WG188 cases; the mutation is rare in Europeans (Exome Aggregation Consortium (ExAC) frequency 10/27,173; Table 1 and Supplementary Table 4). POLE2 is a subunit of the polymerase epsilon enzyme complex. POLE mutation causes polymerase proofreading-associated polyposis; hence, there is a strong likelihood that POLE2 mutation will affect CRC. Support for this comes from the recent identification of POLE2 mutations in two families segregating CRC and polyposis.

**Gene-set enrichment analysis.** As far as they have been deciphered, cancer susceptibility genes for the same tumour type are often implicated in the same biological processes. To complement our T1 analysis of single genes, we conducted a gene-set enrichment analysis (GSEA) based on the GO Biological Process ontology using our entire Discovery series to identify genes, which individually may not be so remarkable as to be detectable at the exome-wide level of significance (Methods). The DNA_REPLICATION gene set was significantly associated with CRC (Q-value < 0.1 in gene-set permutation test; Supplementary Table 5). This enrichment is driven by disruptive mutations in MLH1, MSH2, PMS2 and MSH6, but also POLE2, and POT1 and MRE11A genes hitherto not previously implicated in CRC susceptibility. POT1 and MRE11A represent credible CRC predisposition genes a priori. Both genes participate in telomere length...
Table 1 | Disruptive mutations identified in candidate CRC susceptibility genes with biological relevance.

| Gene | Mutation | Exomes | WGSET | ExAC |
|------|----------|--------|-------|------|
|      | cDNA change | Protein change | Case | Control | $P_{T1}$ | Case | Control | $P_{T1}$ | $P_{T1\text{-META}}$ |
| LI2R81 | c.94C > T | p.Gln32Ter | 4 | 1 | 0.029 | 2 | 0 | 0.001 | 8.7E – 04 |
|       | c.1624C > T | p.Gln542Ter | 3 | 0 | 5 | 2 | 0 | 0.001 | 2.7E – 04 |
| LIMK2 | c.1711_1712insC | p.Gly574ArgfsTer12 | 5 | 1 | 0.012 | 2 | 0 | 0.001 | 8.7E – 04 |
|       | c.1742dupG | p.Cys582LeufsTer4 | 0 | 0 | 1 | 0 | 0 | 151 |
|       | c.2049_2050insA | p.Gln684ThrfsTer16 | 0 | 0 | 1 | 0 | 0 | 0 |
| POLE2 | c.1406dupT | p.Leu469PhefsTer17 | 3 | 0 | 0.021 | 2 | 1 | 0.004 | 9.5E – 04 |

Table 2 | Disruptive mutations identified in novel candidate CRC susceptibility genes.

| Gene | Mutation | Exomes | WGSET | ExAC |
|------|----------|--------|-------|------|
|      | cDNA change | Protein change | Case | Control | Case | Control | Case | Control | $P_{T1\text{-META}}$ |
| MRE11A | c.1726C > T | p.Arg576Ter | 3 | 0 | 0 | 0 | 0 | 4 |
|       | c.1066delC | p.His356ThrfsTer34 | 1 | 0 | 1 | 0 | 0 | 0 |
|       | c.21_26del | p.Leuf7fsTer18 | 1 | 0 | 0 | 0 | 0 | 0 |
| POT1 | c.1851_1852delTA | p.Asp617GlufsTer9 | 2 | 0 | 1 | 0 | 0 | 0 |
|       | c.1087C > T | p.Arg363Ter | 1 | 0 | 0 | 0 | 0 | 0 |
|       | c.219_220insA | p.Asnt57fsTer17 | 0 | 0 | 1 | 0 | 0 | 0 |

Table 1: Genes with biological relevance identified in T1 gene burden analysis of disruptive mutations with $P_{T1} < 0.05$ support in the WG set and/or the presence of rare recurrent variants.

Table 2: Genes identified by GSEA.

POT1 mutations are associated with familial melanoma. The mother of the p.Asp617GlufsTer9 mutation carrier had CRC at 47 and lymphoma at 76 years. Our findings are compatible with a more extensive spectrum of cancer types associated with POT1 mutation.

Search for recessively acting variants. Although most Mendelian CRC genes are dominant, inactivation of MUTYH, one of the three key components of base-excision repair, causes recessive CRC. We looked expressly for evidence of homozygosity or compound heterozygosity for rare damaging variants. Although not highly powered to detect such alleles, we identified compound heterozygosity for the base-excision repair gene NTHL1 p.Gln90Ter/p.Gln287Ter (Supplementary Table 4), supporting the recent observation of NTHL1 as a rare cause of recessive CRC.

Analysis of common variants. Although not the primary purpose of this study, we took the opportunity to search for common (that is, MAF > 5%) SNVs and indels influencing CRC risk—identifying a significant association with ATFI c.327C > T (rs1129406; MAF = 0.39, $P = 2.08 \times 10^{-7}$).

Analysis of non-coding variants. Although our sequencing primarily targeted exons, a proportion of 'off-target' reads are expected to encompass gene regulatory regions (Methods). Acknowledging the limitations of these data we examined such variants, identifying a significant association with a common variant rs749072 (chr3.37096024:g.T > C, MAF = 0.36, $P = 8.0 \times 10^{-8}$), which is associated with MLHI promoter methylation. No rare non-coding variant showed a significant association.

Discussion

Here we have searched for high-impact mutations within the exome, a highly enriched subset of the genome in which it has been argued that disease-causing mutations are most likely to reside. By focusing on the exome, we have limited our ability to identify pathogenic mutations outside of transcribed regions and targeted capture is insufficiently sensitive to detected copy number variation. However, data catalogued by CLINVAR on the known CRC genes suggests copy number variations (1–50 kb) are likely to account for <10% of pathogenic mutations. Accepting these caveats, our findings invite several conclusions. We can confidently ascribe 15% of familial CRC cases to rare
variants in established CRC predisposition genes with a maximum estimate of 31%. We also have found evidence for novel CRC predisposition genes, which merit further functional and/or replication analysis. The biological function of the candidate CRC genes we identified (MRE11, POLE2 and POT1) further underscore the centrality of DNA replication/instability as a cause of heritable CRC. It is however unlikely that further major high-penetration genes with a similar profile to the mismatch repair or APC genes exist. Moreover, we did not identify any coding low-frequency alleles (1–5%) with moderate effect.

Over the range of allele frequencies and effect sizes compatible with the established 2.2-fold sibling relative risk of CRC19, our study had 80% power to identify a new predisposition gene accounting for > 1% of the familial risk (Supplementary Fig.3). Our study thus clarifies the genetic architecture of CRC and probably discounts the existence of further major high-penetration susceptibility genes. For alleles having a more modest impact on overall CRC burden, sample sizes of 10,000 cases are likely to be required for gene discovery initiatives. The identification of missense-driven CRC genes will be hampered by our inability to accurately classify such mutations acting to dilute the association signal and reducing effective population size. Thus, our analysis highlights a major requirement in cancer gene discovery studies, to improve on the ability to accurately assign pathogenicity to sequence changes. In the meantime, additional evidence such as functional data and/or screens of further cases will be required to prioritize candidate pathogenic coding variants. Furthermore, our ability to identify common risk variants, coupled with the prediction from statistical modelling of genome-wide association study (GWAS) data that 17% of the heritable CRC risk can be ascribed to common variants20, provides a rationale for continued implementation of GWAS to identify new risk loci. It is notable that the mutations in the known genes are associated with more profound family history of CRC than in those in whom a genetic diagnosis cannot be made (Supplementary Table 2). Thus, from the clinical perspective, gene testing should not detract from the value of a detailed family history to inform screening requirements in patients and families.

In conclusion, from our analysis it is clear that the identification of additional significant CRC gene associations will require very large-scale sequencing in conjunction with careful statistical analysis. However, such efforts should not detract from the analysis of small but highly informative CRC families ascertained on the basis of highly selected phenotypes and which are likely to continue to prove highly effective in gene discovery.

Methods

Ethics. Written informed consent was obtained from all individuals with ethical review board approval (UK National Cancer Research Network Multi-Research Ethics Committee 02/0/097) and the study was conducted in accordance with the declaration of Helsinki.

Samples and data sets for discovery. The cases comprised 1,028 unrelated patients (559 male) with CRC, aged ≤ 55 years at diagnosis (mean age 48.7, s.d. = 6.0), who had at least one first-degree relative with CRC, ascertained between 2003 and 2011 through the NSCCG1. All the patients were UK residents and had self-reported European ancestry. Germline DNA was isolated from EDTA-venous blood using standard methods and picogreen quantified.

The controls comprised 1,644 healthy individuals from the UK 1958BC21—974 from the ICR1000 data set (EGAD00001001021)22 and an additional 670 individuals, who had at least one first-degree relative with CRC, ascertained between 2011 and 2011 through the NSCCG1. All of the cases were UK residents and had self-reported European ancestry. Germline DNA was isolated from EDTA-venous blood using standard methods and picogreen quantified.

Whole-exome sequencing. A Covaris E Series instrument (Covaris Inc., Woburn, MA, USA) was used to fragment 1 μg of DNA from each individual. Illumina’s TruSeq 62 Mb expanded exome enrichment kit was used to prepare paired-end libraries, as per the manufacturer’s instructions (Illumina, San Diego, CA, USA); 2 × 100 bp sequencing was performed using Illumina HiSeq2000 technology.

Read mapping and variant analysis. CASAVA software (v.1.8.1, Illumina) was used to extract paired-end fastq files. The alignment of reads to build 37 (hg19) of the human reference genome was performed using Stampy25 and BWAMEM software. The Genome Analysis Tool Kit (GATKv3) pipeline was implemented according to best practices26. Analysis was restricted to capture regions defined in the Truseq 62 Mb bed file plus 100 bp padding. The Variant Effect Predictor (VEP; ref. 27) was used to provide annotations on the predicted impact of each variant together with functional classifications from the CONDEL28 algorithm. We additionally annotated with alignability of 100mers and distance from simple repeats defined by UCSC browser tracks.

Sample-level quality control. Using the set of common SNVs defined in dbSNPv138 and performing genotype calling using the Platyplus29 algorithm, we assessed sample quality according to a number of metrics. We identified seven individuals who had non-northern European ancestry, through principal component analysis using EIGENSTRAT30 software, and removed them from further analysis. We used the HapMap Project data. In addition, we excluded individuals with high levels of heterozygosity, sex discrepancy, poor call rate and contamination (Supplementary Fig. 1). During the course of the study, two 1958BC controls were identified as having been diagnosed with cancer and they were excluded. We calculated a similarity metric between all samples to assess identity-by-state; no related individuals were identified. Cases and controls were compared using transition/transversion ratios, dbSNP percentage and number of alternate alleles across different VEP classes called by the GATK pipeline. No substantial difference was observed between these metrics for cases and controls (Supplementary Table 1).

Variant filters. We considered only canonical transcripts and for each variant, assuming the most deleterious predicted effect for each transcript according to VEP. In gene-based analyses, only non-silent variants were considered (that is, transcript ablations, splice donor/acceptors, splice region, stop gain, frameshift, stop lost, initiator codon variants, transcript amplifications, in-frame insertion/ deletions and missense). For all analyses, we imposed GATK internal calling thresholds excluding variants as per the current best practice guidelines—in the 99.5th truth tranche for SNVs and > 99th tranche for indels. To further identify false positives, which would have an adverse impact on the analysis, we adopted an automated approach imposing: GQ ≥ 30; for a heterozygous call, an alternate depth ≥ 3 and GQ2 > 10.83 (that is, P = 0.0001) for the observed versus expected distribution of alternate/reference alleles (alt-ref ratio); UCSC alignability (100 bp window size) = 1, not in simple repeat; Hardy–Weinberg equilibrium test (P > 1.0 × 10−1) in cases and controls; and an overall call rate ≥ 75% in both cases and controls.

Contamination analysis. We used verifyBamId to estimate per sample levels of contamination, identifying a single contaminated sample, which was excluded.

ExomeArray concordance. We evaluated the fidelity of exome sequencing in 1,332 samples, which had also been genotyped using Illumina HumanExome-12v1.1 Beadchip arrays (Illumina). Probes were excluded if monomorphic, call rate < 0.99 in cases/controls, there was a significant difference in genotypes between cases and controls (P < 0.05), there was a significant Hardy–Weinberg equilibrium in controls P < 0.001 and if non-autosomal. Samples were excluded if the call rate was < 0.99 or called variants exhibited outlying heterozygosity (> 3 s.d.). Concordance was assessed for rare bi-allelic SNVs (< 5% frequency); SNVs with unresolvable strands (A/T and C/G) were excluded. We used PLINK-seq to extract a set of 46,811 SNVs seen in the exome array data. Assuming that this set constitutes the set of true positives, we assessed the sensitivity of our exome sequencing data by counting the number of these variants correctly identified through our exome-sequencing protocol. Specificity and sensitivity across all alleles with MAF < 0.05 was > 99.99% and 78.4%, respectively, for filtered variants.

Coverage of non-coding regions. The Illumina expanded TruSeq Exome enrichment captures 62 Mb of the human exonic region, including 28 Mb of 5’- and 3’-untranslated regions. As part of the analysis pipeline, the TruSeq capture regions are padded by an additional 100 bp at either end as per GATK best practices. In total, 46 Mb of intronic sequence and 21 Mb of promoter regions are captured (62% with > 10 × coverage).

Samples and data sets for replication. The WG188 cases comprised 188 CRC cases selected for early age of presentation and family history from the Colorectal Tumour Gene Identification Study (http://public.ukcrn.org.uk/searchStudy/Detail.aspx?StudyId=7590). All of the cases were UK residents and had self-reported European ancestry. All were mutation negative for mismatch repair mutations. WGS data (mean average coverage 58 ×) were obtained using the Complete Genomics technology31. Variant call format files were extracted using CGATools. Outlier variants were considered for this analysis.

UK10K individual-level variant call format files were downloaded from the Sanger ftp site. ALSPAC (1,828) and TWINSUK (1,754) samples that passed

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We have previously reported. We excluded samples overlapping with the Cancer Genome Atlas. European population frequencies passed UK10K quality control were considered. Gene set enrichment analysis was performed on all gene sets stipulating a same GO biological process (c5.bp.v2.5.symbols.gmt provided by GSEA software). Meta-analysis was performed on the two data sets, to combine evidence across study-specific P-values that accounted for each set's enrichment signal (Supplementary Table 5). Calculation of study power. Disease allele frequency in controls was taken as the baseline allele frequency, whereas the frequency in cases was determined by a baseline allele frequency, whereas the frequency in cases was determined by a

P-value in this reduced Discovery series.

% reduction ± 2 × 10^6 times, to ensure recovery of statistical significance for associations. Exome-wide significance was considered to be P = 8.0 × 10^-6 , corresponding to a Bonferroni correction for the testing of ~20,000 genes and three variant sets. The T1 burden test was applied to both Discovery (exome) and WGS sets. For the Discovery set, the T1 test was performed on both the full set of 1,006 cases and a set of cases where CRC could not be confidently ascribed to pathogenic variant in an established CRC predisposition gene. To create this reduced set, we took a conservative criterion to maximize the opportunity of discovering a novel association. Those samples that contained either a Class 1 variant or a variant previously described as being pathogenic or likely to be pathogenic by InSight (The International Society for Gastrointestinal Hereditary Tumours) were removed, leaving 863 samples potentially harbouring novel associations.

The replication analysis, based on the WGSET, was restricted to genes with Class 1 nominally significant T1 P-values in this reduced Discovery series. Meta-analysis was performed on the two data sets, to combine evidence across study-specific P-values using sample size-weighted Z-score method, implemented in METASWEEPCalculated the biological plausibility of candidate genes on the basis of interaction with a known high-penetration gene, being part of a known CRC pathway, mapping to a GWAS signal, being significantly somatically mutated in CRC, established tumour suppressor or oncogenic role.

We used GSEA to assess the potential over-representation of damaging variants within a curated set. Ranking of genes was based their T1 permuted P-values.

A pre-ranked GSEA was then performed using sets defined as being part of the same GO biological process (c5.bp.v2.5.symbols.gmt provided by GSEA software). Set-based permutations (10,000) were performed to determine significance. Leading edge analysis was performed on all gene sets stipulating a Q < 0.25, to identify genes that accounted for each set’s enrichment signal (Supplementary Table 5).

Calculation of study power. Disease allele frequency in controls was taken as the baseline allele frequency, whereas the frequency in cases was determined by a weighted average of the enrichment found in cases with one, two and three affected first-degree relatives. Allele counts were then sampled from frequencies between 0.00001 and 0.01, and relative risks between 1.75 and 4.0. A Fisher’s test was then performed for each sampling of cases and controls. This process was performed 10,000 times for each frequency/relative risk combination and for each instance the frequency of tests that were significant at an exome-wide significance of 8.0 × 10^-7 equated to study power.

Data availability. WES data on the 1,006 CRC cases and 648 controls have been deposited at the European Genome-phenome Archive, which is hosted by the European Bioinformatics Institute, accession numbers EGAS00001001666 and EGAS00001001667, respectively. The remaining data are either contained within the Article and its Supplementary Information or available from the authors on request.

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Author contributions

R.S.H., D.C., P.B., B.K., S.E.D., I.T. and M.G.D. contributed to writing of the manuscript. R.S.H., D.C., P.B., B.K., I.T., S.E.D. and M.G.D. conceived and designed the experiments. D.C., M.N.T., P.B., B.K., I.T., S.E.D., M.F., Y.P.M., A.L.S., M.G.D. and R.S.H. performed the experiments. D.C., P.B., S.E.D., M.F., A.P., A.L.S., Y.P.M. and R.S.H. analysed the data. S.P. and C.P. were involved in study design/sampling/assembly/data collection, collation, curation, sequencing and quality control/data analysis. All authors reviewed the manuscript.

Additional information

Accession codes: The WES data have been deposited at the European Genome-phenome Archive, which is hosted by the European Bioinformatics Institute, under accession codes EGAS00001001666 and EGAS00001001667.

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