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Title: Genetic and Epigenetic Characteristics of Inflammatory Bowel Disease–Associated Colorectal Cancer

Genetic and Epigenetic Characteristics of Inflammatory Bowel Disease–Associated Colorectal Cancer

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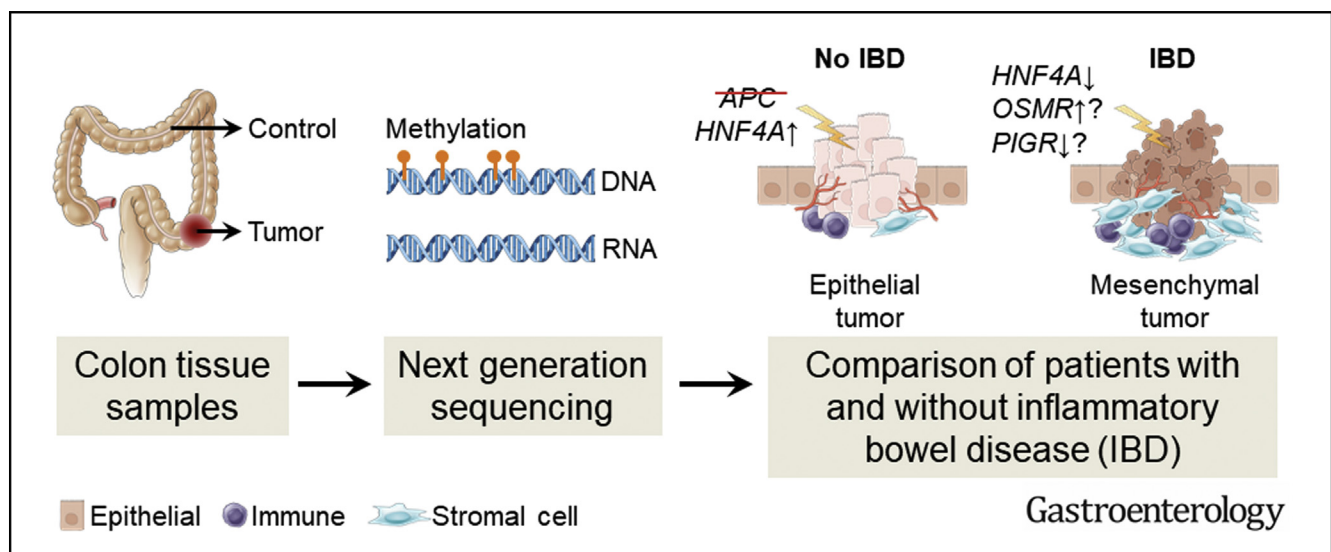
BACKGROUND & AIMS: Inflammatory bowel disease (IBD) is a chronic, relapsing inflammatory disorder associated with an elevated risk of colorectal cancer (CRC). IBD-associated CRC (IBD-CRC) may represent a distinct pathway of tumorigenesis compared to sporadic CRC (sCRC). Our aim was to comprehensively characterize IBD-associated tumorigenesis integrating multiple high-throughput approaches, and to compare the results with in-house data sets from sCRCs. **METHODS:** Whole-genome sequencing, single nucleotide polymorphism arrays, RNA sequencing, genome-wide methylation analysis, and immunohistochemistry were performed using fresh-frozen and formalin-fixed tissue samples of tumor and corresponding normal tissues from 31 patients with IBD-CRC. **RESULTS:** Transcriptome-based tumor subtyping revealed the complete absence of canonical epithelial tumor subtype associated with WNT signaling in IBD-CRCs, dominated instead by mesenchymal stroma-rich subtype. Negative WNT regulators *AXIN2* and *RNF43* were strongly down-regulated in IBD-CRCs and chromosomal gains at *HNF4A*, a negative regulator of WNT-induced epithelial–mesenchymal transition (EMT), were less frequent compared to sCRCs. Enrichment of hypomethylation at HNF4α binding sites was detected solely in sCRC genomes.

PIGR and *OSMR* involved in mucosal immunity were dysregulated via epigenetic modifications in IBD-CRCs. Genome-wide analysis showed significant enrichment of noncoding mutations to 5′untranslated region of *TP53* in IBD-CRCs. As reported previously, somatic mutations in *APC* and *KRAS* were less frequent in IBD-CRCs compared to sCRCs. **CONCLUSIONS:** Distinct mechanisms of WNT pathway dysregulation skew IBD-CRCs toward mesenchymal tumor subtype, which may affect prognosis and treatment options. Increased OSMR signaling may favor the establishment of mesenchymal tumors in patients with IBD.

Keywords: Colorectal Cancer; Inflammatory Bowel Disease; Epithelial–Mesenchymal Transition; DNA Methylation; Consensus Molecular Subtype.

Inflammatory bowel disease (IBD), comprising ulcerative colitis (UC) and Crohn's disease (CD), involves a complex interplay of genetic predisposition and environmental factors that alter host–microbiota interactions causing dysregulation of gut immune responses.[1] The growing prevalence and diminishing age at onset of IBD amplify the risk of comorbities and the associated economic burden.[2] Patients with IBD have an elevated risk of colorectal cancer (CRC)[3] attributed to chronic inflammation, yet the detailed mechanisms remain elusive.[4]

Accumulating evidence suggests that IBD-associated CRC (IBD-CRC) may emerge through a distinct pathway of tumorigenesis compared to sporadic CRC (sCRC). Patients with IBD are younger at CRC diagnosis and the tumors develop at inflamed areas of the colon with characteristic clinicopathologic features.[4,5] IBD-CRCs show lower frequency of somatic *APC* and *KRAS* mutations, whereas *TP53* mutations occur earlier in tumorigenesis compared to sCRCs.[4,6–10] Despite reduced *APC* mutations, nuclear accumulation of β-catenin is prevalent in IBD-CRCs,[11] suggesting an alternative mechanism of WNT pathway activation. Additional suggested driver genes have varied across studies,[6–10] while the frequency of hypermutated[9] and microsatellite unstable tumors,[12] level of somatic copy number alterations,[9,10] and distribution of somatic mutational signatures[6,9,10] appear similar to sCRCs. Studies on DNA methylation patterns characterizing IBD-CRC have focused on a limited number of genes[13–15] or microarray data,[16] warranting further genome-wide analyses.

Transcriptome-based classification of CRCs has emerged as a powerful tool to describe tumor transcriptional, genetic, epigenetic, and microenvironment characteristics.[17,18] A large-scale international effort resulted in amalgamation of 4 consensus molecular subtypes (CMSs).[17] CMS distribution remains unknown in IBD-CRC; in sCRC, epithelial WNT-associated CMS2 is the most common and mesenchymal CMS4 associates with poor prognosis.[17]

Here, we integrate multiple high-throughput sequencing approaches to comprehensively characterize IBD-CRC and to identify differences compared to sCRC. Our most striking finding was the complete absence of CMS2 tumors among IBD-CRCs that were instead skewed toward CMS4.

WHAT YOU NEED TO KNOW

BACKGROUND AND CONTEXT

Inflammatory bowel disease increases the risk of colorectal cancer and the tumors developing in patients may be genetically and epigenetically distinct compared to sporadic tumors.

NEW FINDINGS

Transcriptomic analyses of colorectal cancer specimens from patients with inflammatory bowel disease revealed the absence of canonical epithelial tumor subtype and predominance of mesenchymal tumors associated with oncostatin M receptor overexpression.

LIMITATIONS

The intensive surveillance of colorectal cancer in patients with inflammatory bowel disease makes these tumors rare, limiting the sample size of the study.

IMPACT

The results suggest that colon inflammation may favor the development of mesenchymal tumor subtype, which has previously been associated with poor survival in large colorectal cancer cohorts.

Materials and Methods

Samples

The study adhered to the Declaration of Helsinki and was approved by the local ethics committee (details are in Supplementary Material). CRC patient samples were collected in 1994–2017 at 9 regional hospitals in Finland.[19–21] Thirty-one IBD-CRC cases were identified from our collection comprising approximately 2,500 CRC patients (UC: n = 27, CD: n = 2, unclassified IBD: n = 2 sharing features of UC and CD) (Table 1). The availability of sample material for downstream analyses is summarized in Supplementary Figure 1.

Whole-Genome Sequencing

DNA was isolated from fresh-frozen tumor, normal colon, or blood of 29 patients with IBD-CRC. Libraries were prepared using TruSeq Nano DNA HT or TruSeq PCR-Free Kit (Illumina), followed by paired-end sequencing using Illumina platform (HiSeqXTen/HiSeq2000). sCRCs were sequenced as described previously.[22]

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Table 1. Clinical Characteristics of the Patients With Inflammatory Bowel Disease and Colorectal Cancer

Sample	Sex	IBD diagnosis	Age at IBD diagnosis, y[a]	Age at CRC diagnosis, y	Tumor location	Tumor histology	TNM stage	Grade	MSI status
c174.1T	F	UC	38	54	Cecum	AC	I	2	MSS
c175.1T	M	UC	33	48	Rectum	AC	II	3	MSS
c269.1T	F	UC	41	41	Cecum	ACPM	II	2	MSS
c3.1T	M	IBD-U	27	48	Transverse colon	AC	II	1	MSS
c424.1T1	M	UC	61	72	Rectosigmoid junction	AC	III	3	MSS
c461.1T	M	UC	82	82	Ascending colon	ACPM	III	3	MSS
c492.1T	M	UC	57	57	Rectum	AC	III	2-3	MSS
c589.1T	F	UC	43	50	Transverse colon	AC	III	1	MSS
c596.1T	M	UC	25	51	Rectum	AC	II	2	MSS
c696.1T	F	UC	19	42	Cecum	AC	III	3	MSS
c745.1T	M	CD	25	55	Rectum	AC	II	2	MSS
c989.1T	F	UC	26	62	Cecum	AC	III	3	MSS
s1111.1T	M	UC	36	64	Rectum	AC	II	1	MSS
s1138.1T	M	UC	41	61	Rectum	AC	II	2	MSS
s1170.1T	M	UC	65	65	Sigmoid colon	AC	III	NA	MSS
s1179.1T	M	UC	15	50	Transverse colon	ACPM	III	2	MSS
s205.1T	M	UC	22	60	Descending colon	AC	II	2	MSS
s576.1T	M	UC	22	41	Ascending colon	AC	III	3	NA
s617.1T	M	UC	20	33	Rectum	ACM	IV	NA	MSS
s649.1T	M	UC	20	33	Cecum	ACM	III	NA	MSS
s660.1T	F	IBD-U	56	66	Ascending colon	AC	II	2	MSI
s683.1T	M	UC	8	46	Cecum	ACM	II	NA	MSS
s700.1T	M	UC	41	64	Transverse colon	AC	I	1	MSS
s703.1T	F	UC	11	22	Descending colon	ACPM	III	2	MSS
s750.1T	F	UC	13	30	Cecum	ACPM	III	3	MSS
s751.1T	M	UC	61	61	Sigmoid colon	AC	III	3	MSS
s763.1T	M	CD	18	44	Rectum	AC	I	3	MSS
s814.1T	F	UC	42	72	Rectum	ACPM	I	3	MSS
s842.1T	F	UC	16	53	Rectum	ACPM	III	3	MSS
s85.1T	M	UC	64	64	Cecum	AC	I	NA	MSS
s982.1T1[b,c]	M	UC	34	34	Cecum	ACPM	III	2	MSI
s982.1T2[b,c]	M	UC	34	34	Splenic flexure	ACPM	III	2	MSI

AC, adenocarcinoma; ACM, mucinous adenocarcinoma; ACPM, partially mucinous adenocarcinoma; IBD-U, unclassified IBD; NA, not available.

[a]In cases where IBD was diagnosed together with CRC based on pathologic findings from the surgical resection, there was often a history of several years of undiagnosed gastrointestinal symptoms mentioned in the clinical data.

[b]Two tumors sampled from the same individual.

[c]The individual was also diagnosed with hereditary nonpolyposis CRC.

Somatic Variant Calling and Analysis

Sequence alignment to GRCh38 reference genome, other data preprocessing steps, and somatic variant calling were performed with GenomeAnalysisToolkit GATK4 best practices workflow (version 4.0.4.0.) for all tumor/normal pairs. Gene annotation (Ensembl genes release 89) for somatic single

nucleotide variants and small insertions/deletions was performed with BasePlayer.[23]

Mutational Signatures

Somatic mutational signatures were extracted from whole-genome sequencing (WGS) data of 237 colorectal tumor/normal pairs, including the 27 paired IBD-CRCs. Briefly, single base substitutions (SBSs) were classified based on their flanking sequence context (± 1 bp) into 96 possible mutation types. These mutational spectra were analyzed with standard non-negative matrix factorization, as described previously.[22,24]

OncodriveFML

OncodriveFML (version 2.2.0)[25] was used to analyze the coding sequence, 3′untranslated region (UTR), and 5′UTR for signals of positive selection using the somatic mutations from 27 microsatellite stable (MSS) IBD-CRCs.

Allelic Imbalance

Single nucleotide polymorphism array data were analyzed previously from 1699 colorectal tumor/normal pairs,[26] including 23 IBD-CRCs; allelic imbalance (AI) regions of somatic allelic loss and gain were processed with the same pipeline.[26] Variant calls (HaplotypeCaller) from WGS data were used to calculate AI for 6 additional IBD-CRCs lacking single nucleotide polymorphism array data. The presence of AI in each tumor was evaluated at loci found heterozygous in the corresponding normal sample, as described previously.[26] Analysis was limited to MSS IBD-CRCs (n = 27) and MSS sCRCs (n = 1360; microsatellite instability [MSI] statuses from Palin et al[26]).

RNA Sequencing

Trizol-extracted RNA from 64 CRCs, including 17 IBD-CRCs, underwent HiSeq LncRNA-Seq library preparation and paired-end sequencing using Illumina HiSeqXTen. Raw sequences were mapped onto the human transcriptome (ensembl release 79) using Salmon (version 0.12.0).[27] Gene-level quantification was done with DESeq2 (version 1.18.1),[28] followed by limma (version 3.34.9)[29] correction of sequencing batch effects.

Differential Expression Analysis

Differential gene expression was analyzed using Partek Genomics Suite 6.6 (Partek Inc) Gene Expression workflow, followed by pathway analysis using PANTHER (version 15.0).[30]

Consensus Molecular Subtypes

Random forest classifier of CMSClassifier R package[17] was used to call CMS for each RNA-sequenced tumor, expressed here as the nearest CMS (RF.1) predicted by the classifier.

Deconvolution

The proportions of tumor-infiltrating immune cells were estimated from RNA sequencing data using CIBERSORT.[31,32] Reference was created as described previously[33] by combining bulk RNA sequencing data from isolated blood immune cells (accession GSE60424) and representative median expression profiles from CRCs and normal colon samples from an independent data set.[34] See Supplementary Table 16.

Cell type–specific gene expression profiles were inferred from RNA sequencing data using PRISM,[35] combining single-cell RNA sequencing data of healthy colon[36] and CRCs[37] as reference.

Immune Cell Score

Whole-section slides from 265 formalin-fixed, paraffin-embedded CRCs, including 26 IBD-CRCs, were stained with anti-CD3 (LN10, 1:200; Novocastra) and anti-CD8 (SP16, 1:400; Thermo Scientific) antibodies. Positively stained cells were analyzed using QuPath,[38] as described previously.[39] The immune cell score was formulated as described previously,[40] following the original method by Galon et al.[41]

Nanopore Long-Read Sequencing

Libraries were prepared for 20 IBD-CRCs, 36 sCRCs, and 12 normal colon samples from IBD-CRC patients following Genomic DNA by Ligation (SQK-LSK109) protocol (Oxford Nanopore Technologies). Sequencing and base-calling on PromethION platform employed Live base-calling with MinKNOW. Reads were aligned against the reference genome GRCh38 using minimap2 (version 2.16; preset: map-ont).[42] Structural variants (SVs) were identified using Sniffles (version 1.0.11)[43] and merged together from all tumors and normals with SURVIVOR (version 1.0.6)[44] to filter out SVs found in any normal sample. Genome-wide methylation patterns were obtained using Nanopolish.[45] Differentially methylated loci (DMLs) at autosomal regions were identified and analyzed with R, version 3.5.1 using R packages DSS (version 2.28.0,)[46] bsseq (version 1.16.1), annotatr (version 1.12.1),[47] and Locus Overlap Analysis (version 1.16.0).[48] Chromatin immunoprecipitation sequencing and DNase-sequencing data and 15 chromatin states provided by Roadmap Epigenomics[49] project (Supplementary Table 12) and transcription factor binding sites from CRC cell lines[50] served as region set databases in Locus Overlap Analysis.

Data Availability

Genome-wide somatic single nucleotide variant and insertion/deletion calls (GRCh38) are deposited in the EGA database under accession code EGAS00001004710.

Results

Somatic Point Mutations in Known Colorectal Cancer Driver Genes and Immunity-Related Genes Characterize Inflammatory Bowel Disease–Associated Colorectal Cancers

The study comprised 31 cases of IBD-CRC (Table 1); 29 underwent WGS. A total of 1,104,175 somatic single nucleotide variants and insertion/deletions were identified (Supplementary Figure 2). Two outliers with high somatic variant counts were explained by MSI, a distinct pathway of

CRC tumorigenesis driven by mismatch repair deficiency. We focused on the more typical MSS IBD-CRCs.

The 27 MSS IBD-CRCs harbored a median of 18,194 somatic variants per tumor (3387–73,003), which was highly similar to 259 MSS sCRCs previously whole-genome sequenced in-house (median, 17,319; range, 253–84,646) (Supplementary Figure 2); genome-wide mutation densities showed similar distributions. Coding sequences displayed 4702 variants, with 2816 genes affected by nonsynonymous variants. We ranked the 100 genes mutated in 3 or more tumors by mutation density (Figure 1A). Top 20 genes featured several known CRC driver genes, including TP53 and KRAS having the highest mutation densities. Analysis of mutual exclusivity and co-occurrence of mutations in these genes revealed no significant results (Supplementary Table 1). As expected, KRAS and APC mutations were few in IBD-CRCs compared to sCRCs (22% vs 49% and 22% vs 73%, respectively, Fisher exact test, $P = .0087$ and $P = 2.5 \times 10^{-7}$); TP53 was frequently mutated in both groups (63% vs 62%) (Figure 1A, Supplementary Table 2). A comparison of all variants in COSMIC Cancer Gene Census[51] genes revealed paucity of genes mutated uniquely in IBD-CRC (Supplementary Table 2).

Immunity-related CARD8 and PIGR ranked high by mutation density (Figure 1A). All 3 CARD8 variants were missense, and 2 of the 4 PIGR variants present in 3 of 27 IBD-CRCs (11% vs 1% in MSS sCRCs) were truncating (Supplementary Table 3).

Noncoding TP53 *Mutations Are Enriched in Inflammatory Bowel Disease–Associated Colorectal Cancers*

To discover candidate driver genes, we applied Onco-driveFML[25] to all somatic point mutations from the 27 MSS IBD-CRCs, totaling 622,366 variants. Coding regions of 2 genes showed significant evidence of positive selection after false discovery rate (FDR) correction ($Q < 0.1$), TP53 ($P < 1.1 \times 10^{-6}$, $Q = .00061$) mutated in 16 tumors and GBA2 ($P = .00014$, $Q = .043$) mutated in 2 tumors (Supplementary Figure 2, Supplementary Table 4). We further analyzed 3′UTR and 5′UTR of all protein-coding genes. Overlapping 5′UTR of TP53 and WRAP53 genes was the only region showing significant evidence of positive selection ($P = .00036$, $Q = .022$) in IBD-CRCs, while remaining nonsignificant in sCRCs ($P = .015$, $Q = 1.00$) (Supplementary Figure 2, Supplementary Table 4). This region harbored 4 mutations in 3/27 IBD-CRCs and 6 mutations in 5 of 239 sCRCs, resulting in low TP53 expression (Figure 1B and C).

Uncoupling of Age-Related Mutational Signature From Age in Inflammatory Bowel Disease–Associated Colorectal Cancers

Detection of genome-wide mutational signatures[22,24] in IBD-CRCs revealed 5 somatic SBS signatures that resembled the signatures SBS1 (age-related spontaneous deamination of 5-methylcytosine), SBS8 (unknown), SBS17 (CTCF/

cohesin binding sites), and SBS15 and SBS20 (defective DNA mismatch repair) in human cancers described previously[22,52] (Figure 2A). Compared to MSS sCRCs, MSS IBD-CRCs showed a decreased exposure to SBS1 ($P = .0037$) (Figure 2B) and a significantly lower rate of SBS1 mutations during life before CRC diagnosis (ordinary least squares, $P = .023$). Neither the mean difference nor difference in SBS1 mutation rate could be explained in Bayesian analysis by the younger age at onset in IBD-CRC patients (mean ± SD, 54 ± 11 years vs 69 ± 11 years in IBD-CRC vs sCRC) (Figure 2C). SBS17 exposure was elevated ($P = .0082$) in IBD-CRCs (Figure 2B); tumors dominated by SBS17 showed no obvious correlations to clinical characteristics or known driver mutations.

Analysis of Chromosomal Rearrangements Reveals Tumor Type–Specific Allelic Imbalance

Chromosomal stability of the MSS IBD-CRCs was inspected using both nanopore WGS data (n = 19) and single nucleotide polymorphism array data (n = 27). Numbers of somatic nanopore-detected SVs did not differ between IBD-CRCs and sCRCs (Mann-Whitney U test $P > .76$) (Figure 3A–C). Coding region SVs in IBD-CRC revealed 5 genes with a recurring somatic breakpoint: CCSER1 (n = 2 tumors), FHIT (n = 2), IMMP2L (n = 3), MACROD2 (n = 2), and PIBF1 (n = 2) (Supplementary Table 5), all known fragile site genes.[53,54]

We previously characterized AI in 1699 CRCs[26] and now compared MSS IBD-CRCs (n = 27) to MSS sCRCs (n = 1360). No overall difference between IBD-CRCs (mean, 9.2×10^8 bp of AI per tumor) and sCRCs (mean, 1.1×10^9 bp) was observed (Mann-Whitney U test $P = .29$) (Figure 3D). The only outstanding difference genome-wide (at FDR < 10%) was found at 5p13.1-p12, where IBD-CRC was enriched for gains (10 of 27 tumors (37%); Figure 3E; Supplementary Table 6). Of the genes in this region, OSMR and LIFR cytokine receptors sharing the ligand oncostatin M[55] were significantly up-regulated in IBD-CRCs compared to sCRCs (Supplementary Table 7). Closer inspection of 38 previously characterized AI target genes in CRC[26] revealed 4 genes with differential AI (FDR < 10%): CDKN2B and SMARCA2 enriched for allelic losses, FOXA1 for allelic gains, and HNF4A having fewer gains in IBD-CRC compared to sCRC (Supplementary Table 8).

Differential Gene Expression Highlights Changes Related to Stromal and Immune Cells in Inflammatory Bowel Disease–Associated Colorectal Cancers

Bulk RNA sequencing revealed 870 significantly differentially expressed (DE) genes between MSS IBD-CRCs and MSS sCRCs (FDR < 10%; Supplementary Table 7, Supplementary Figure 3). In IBD-CRCs, pathway analysis showed a strong overenrichment (overexpression) of gene sets related to complement activation and extracellular matrix organization (Figure 4A, Supplementary Table 9). Very few gene sets were significantly underenriched,
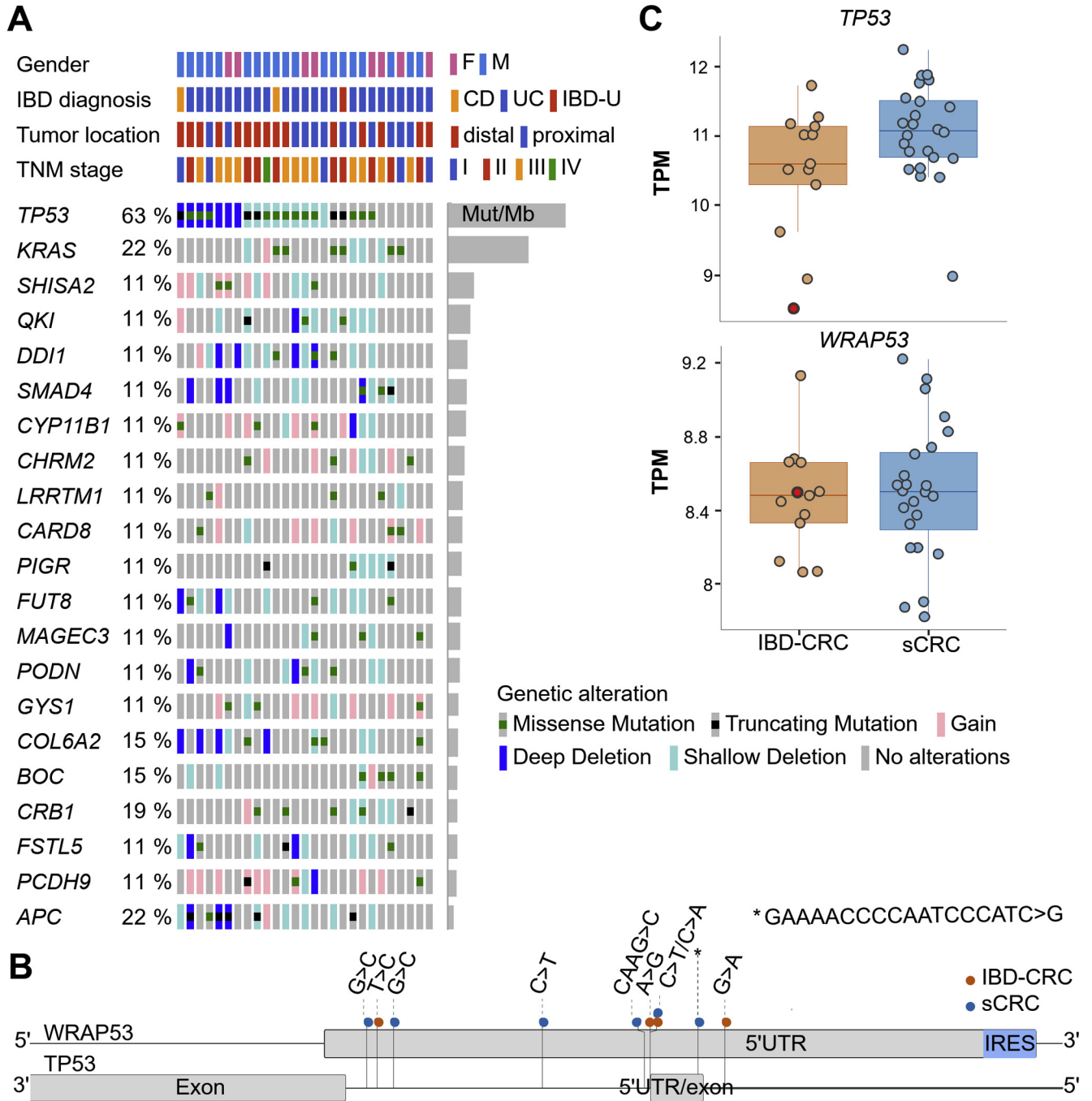
Figure 1. Somatic point mutations. (A) OncoPrint (https://www.cbioportal.org/oncoprinter) showing somatic point mutations (*filled squares*) and copy number alterations (*filled bars*) in genes mutated in 3 or more MSS IBD-CRCs and ranked by mutation density (mutations/Mb; top 20 genes and *APC* ranking no. 41 presented with percentages of mutated tumors). F, female; IBD-U, unclassified inflammatory bowel disease; M, male; TNM, tumor-nodes-metastases. (B) Variants observed in the overlapping 5′UTR of *TP53* and *WRAP53*. IRES, internal ribosome entry site. (C) Gene expression in 64 RNA-sequenced CRCs as transcripts per million (TPM). *Red dot* signifies an IBD-CRC carrying *TP53/WRAP53* 5′UTR variant (no coding *TP53* variants).

including only 1 of 44 significant results for Reactome database, TCF-dependent signaling in response to WNT (R-HSA-201681). Genes in this set included *RNF43* and *AXIN2*, the most significantly down-regulated genes in IBD-CRCs. The most significantly up-regulated gene, *OSMR*, appeared in a large overenriched Immune System gene set (R-HSA-168256) comprising 146 DE genes related to myeloid/lymphoid immune cells, complement, and, interestingly, EMT (*TWIST1*, *ZEB1*, *VIM*) (Supplementary Table 9).

We estimated cell type–specific gene expression profiles for epithelial, stromal, and immune cells using PRISM deconvolution tool35 (Supplementary Figure 4). PRISM-estimated cell-type proportions for each tumor consensus molecular subtype (Supplementary Figure 5) matched those
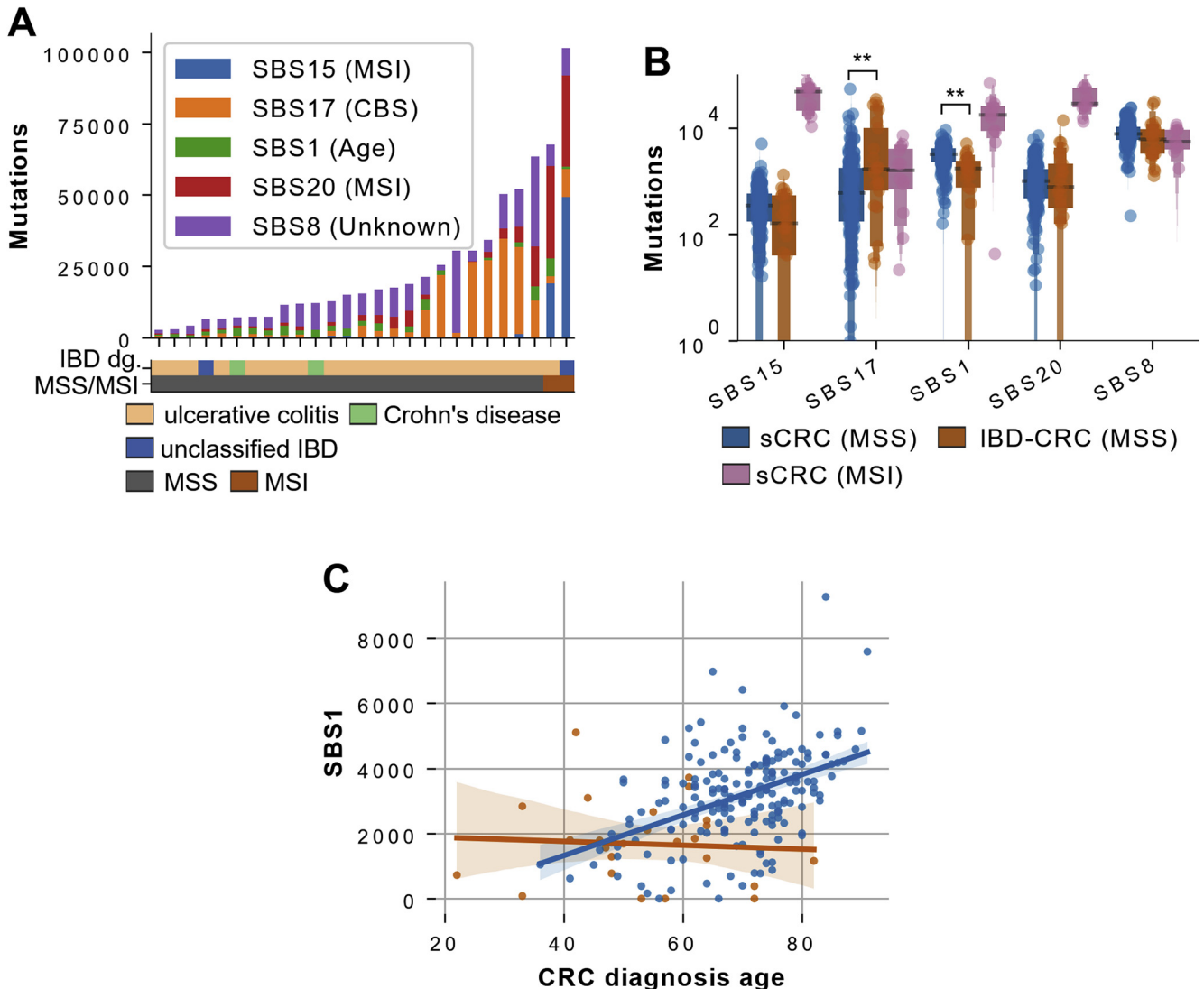
Figure 2. Somatic mutational signatures. (*A*) Contributions of SBS signatures to somatic mutations in IBD-CRCs. CBS, CTCF/cohesin binding site. (*B*) SBS signatures in IBD-CRCs (MSS n = 25) and sCRCs (MSS n = 194, MSI n = 12). Robust linear model regression was used to compare the MSS groups (**P < .01). (*C*) Analysis of age-dependence of SBS1. After Bayesian analysis with and without intercept, depending on the IBD status, the slope for IBD-CRCs for SBS1 exposure is significantly lower than for sCRCs. On average, the MSS IBD-CRCs had 7.4–61 fewer SBS1-associated mutations per year (95% credible interval).

reported previously using a different algorithm.[17] Differential expression analysis between MSS IBD-CRCs and MSS sCRCs yielded 376, 626, and 66 DE genes for epithelial, stromal, and immune cells, respectively (Supplementary Table 7). Pathway analysis of these separate gene lists allowed connecting each pathway change to a particular cell type (Supplementary Table 9). Epithelial DE genes, predominantly down-regulated in IBD-CRCs, were enriched for pathways related to epithelial cell differentiation and development. Stromal DE genes, predominantly up-regulated, were enriched for extracellular matrix organization, integrin interactions, vasculature development, and insulin-like growth factor metabolism. Immune DE genes, predominantly up-regulated, were enriched for classical complement activation and other antibody-mediated immune responses.

Common Epithelial Colorectal Cancer Subtype Associated With WNT Signaling Is Absent in Inflammatory Bowel Disease–Associated Colorectal Cancers

We defined the nearest CMS for each of the 64 RNA-sequenced CRCs, as described previously.[17] As reported, MSI sCRCs were highly enriched for CMS1 (Supplementary Table 10). Based on PRISM deconvolution, immune cell proportion was highest in CMS1, stromal in CMS4, and epithelial in CMS2/CMS3 tumors (Supplementary Figure 5), as reported previously.[17] Comparison of MSS tumors in IBD-CRCs revealed a striking paucity of the canonical epithelial CMS2 subtype associated with WNT and MYC signaling (0% of IBD-CRCs vs 39% of sCRCs; P = .0048, P_{adj} = .019) (Figure 4B, Supplementary Table 10). IBD-CRCs were
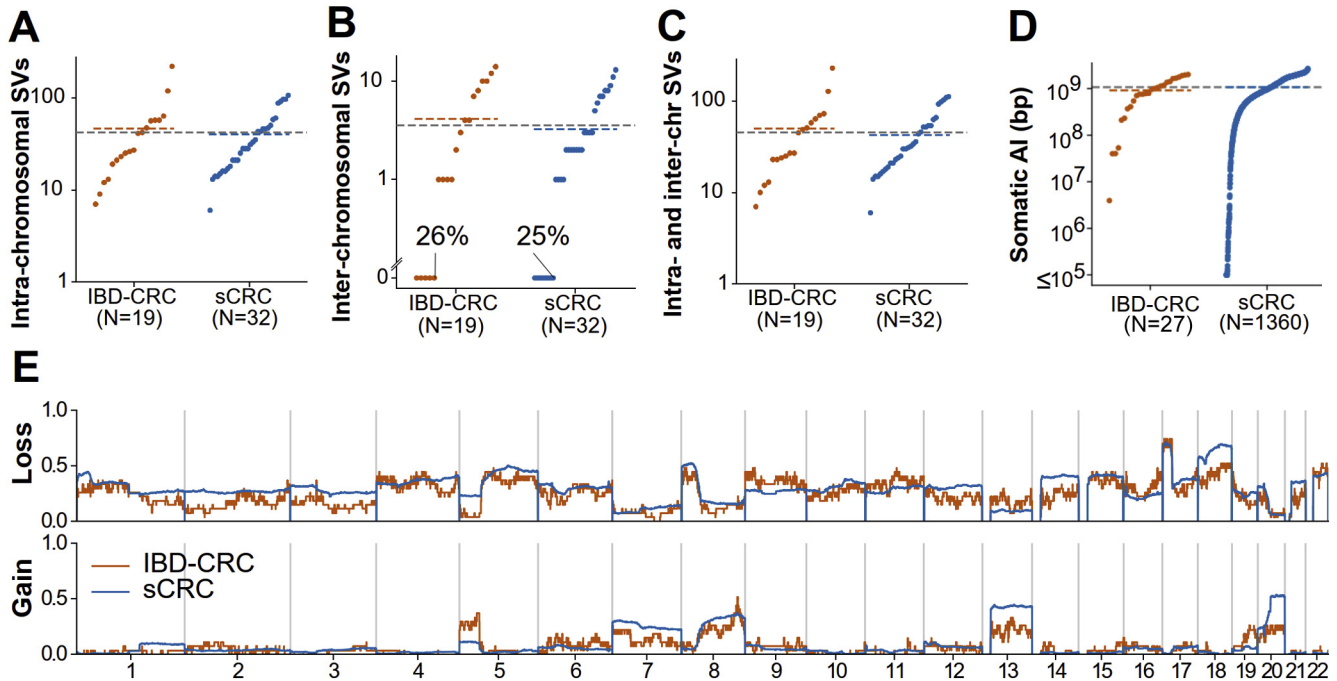
Figure 3. Overview of somatic structural aberrations. Numbers of somatic (A) intra- and (B) inter-chromosomal, and (C) total SVs in 19 nanopore-sequenced MSS IBD-CRCs and 32 MSS sCRCs. The percentages in (B) refer to the proportion of tumors lacking inter-chromosomal SVs. (D, E) Somatic AI in 27 MSS IBD-CRCs and 1360 MSS sCRCs across the autosomes. (D) Total amount of base pairs affected by somatic AI. *Dashed lines* denote the mean per group (*orange/blue*) and overall mean (*gray*). Y-axis is logarithmic and truncated to a minimum 100 kbp. (E) Observed proportion of tumors with allelic loss (or gain) at each genomic position.

dominated by mesenchymal CMS4 (57% vs 21%; $P = .019$, $P_{adj} = .057$), with concomitant up-regulation of transcription factors mediating EMT (*TWIST1, TWIST2, SNAI2, ZEB1, ZEB2*) (Supplementary Table 7).

Mesenchymal Tumors Show a Distinct Pattern of Immune Cell Infiltration

We interrogated the immune cell contexture of RNA-sequenced CRCs using CIBERSORT deconvolution.[31,32] MSI and CMS1 tumors were found in clusters with high estimated proportions of CD8[+] cytotoxic T cells (Supplementary Figure 5), as reported previously.[17] MSS tumors showed 3 clusters dominated by CD4[+] T, CD8[+] T, and B cells, respectively (Figure 4C). IBD-CRCs were divided between the B cell and the CD4[+] T cell clusters, forming in the latter a distinct CMS4-enriched subcluster distinguished by high proportions of monocytes.

We further analyzed immune cell score, a prognostic measure of tumor T cell infiltration reflecting numbers of total (CD3[+]) and cytotoxic (CD8[+]) T cells.[41] Rates of high immune cell score (3–4) were similar between 24 MSS IBD-CRCs and 196 MSS sCRCs (54% vs 43%; Fisher exact test, $P = .39$) (Supplementary Table 11). There were no significant differences in the 4 individual stainings (Figure 4D) or in the ratios of CD8[+] to CD3[+] T cells (Supplementary Figure 5) between these groups (Mann-Whitney U test, Holm-Bonferroni correction).

Similar Genome-Wide Methylation Patterns in Inflammatory Bowel Disease–Associated Colorectal Cancers and Sporadic Colorectal Cancer

Methylation analyses were carried out using whole-genome nanopore sequencing data. Based on methylation values at CpG islands, IBD-CRCs and a pool of nondysplastic normal colon samples from patients with IBD ("IBD-normals") mainly clustered separately from sCRCs (Supplementary Figure 6). All subsequent methylation analyses focused on samples from patients with MSS tumors. IBD-CRCs showed, on average, higher genome-wide methylation compared to sCRCs (Figure 5A). Neither age nor cancer type was significantly associated with the average methylation level (joint model $P = .078$). Differentially methylated loci (DMLs) were studied in autosomes comparing IBD-CRCs to IBD-normals and sCRCs to IBD-normals, resulting in 553,390 DMLs (4.4% hypermethylated) and 2,413,663 DMLs (3.7% hypermethylated), respectively. Five of the 10 IBD-normals were matched with the studied IBD-CRCs, likely decreasing the IBD-CRC DML count, potentially affected also by shared IBD-derived methylation changes.

In both tumor groups, genomic annotation of the DMLs (Figure 5B–D, Supplementary Figure 6) revealed the majority of hypomethylated loci at non-CpG island ("CpG inter") areas and modest enrichment only on quiescent/low chromatin areas, out of 15 chromatin states studied.
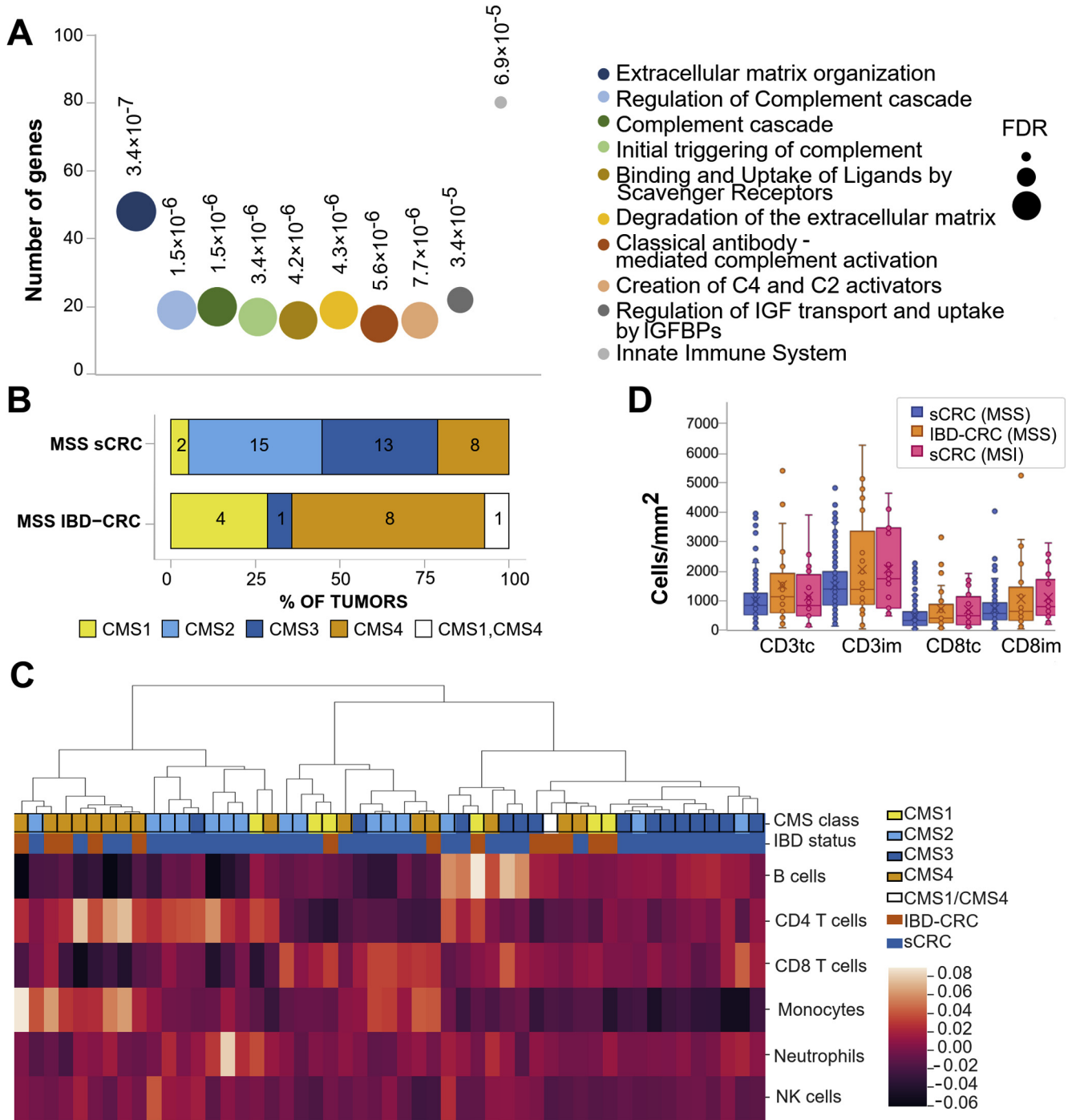
Figure 4. Patterns of gene expression and immune cell infiltration. (*A–C*) RNA sequencing data from 14 MSS IBD-CRCs and 38 MSS sCRCs was compared. (*A*) Top 10 pathways significantly enriched among DE genes (n = 870) in PANTHER Statistical Enrichment Test[30] against the Reactome database. (*B*) CMS[17] presented as the nearest CMS having the highest posterior probability (equal for CMS1 and CMS4 in one tumor). (*C*) Clustering of tumors based on the estimated proportions of immune cells from CIBERSORT deconvolution. NK, natural killer. (*D*) $CD3^+$ and $CD8^+$ T cell numbers determined by immunohisto-chemistry at the tumor core (tc) and invasive margin (im) in 25 MSS IBD-CRCs, 217 MSS sCRCs, and 20 MSI sCRCs.

Conversely, most of the hypermethylated loci were located on CpG islands and intronic areas (Figure 5*B*), showing enrichment on several chromatin states related to bivalent chromatin (Figure 5*C*). From chromatin states with average odds ratio >20, the enrichment of hypermethylation relative to IBD-normals was lower in IBD-CRC compared to

sCRC on bivalent/poised transcription start sites (TSS) (Mann-Whitney U test, $P_{adj} < 1.6 \times 10^{-6}$) and flanking bivalent TSS/enhancers ($P_{adj} = .00011$). In both groups, the enrichment of hypermethylated loci was strongest on H3K27me3 marked chromatin (Figure 5*D*). We further inspected whether the DE genes were commonly annotated

with bivalent TSS, which was indeed the case more often than expected by chance (Supplementary Figure 7).

HNF4A Binding Sites Are Hypomethylated Solely in Sporadic Colorectal Cancer

We studied the enrichment of DMLs in chromatin immunoprecipitation sequencing peaks of 382 transcription factors and DNA binding proteins obtained from LoVo CRC cell line.[50] Significant enrichment was detected at multiple binding sites in both tumor groups compared to IBD-normals (Supplementary Table 13). However, the significance often originated from a small total number of binding sites harboring DMLs, diminishing the biologic relevance, especially in results unique to IBD-CRCs. In sCRCs, the most significant enrichment of hypomethylation relative to IBD-normals was found at HNF4A binding sites, with DMLs detected at >3000 separate binding sites. This enrichment was absent in IBD-CRCs, in agreement with fewer HNF4A gains. As reported previously,[17] the frequency of HNF4A gains and HNF4A expression, together with APC mutations, were highest in CMS2 tumors (Supplementary Figure 8). Interestingly, binding sites of TCF7L2, key transcription factor mediating WNT/β-catenin signaling, showed hypomethylation only in sCRC when compared to IBD-normals, and binding sites of MYC, a critical WNT/β-catenin target gene, were enriched for hypermethylation in IBD-CRCs when compared to sCRCs (Supplementary Table 13).

PIGR and OSMR Involved in Mucosal Immunity Show Dysregulated Promoter Methylation in Inflammatory Bowel Disease–Associated Colorectal Cancers

We next identified genes containing DMLs at promoter region (1 kbp upstream of TSS) in IBD-CRCs, but not sCRCs, compared to IBD-normals. Thirteen and 7 protein coding genes contained >1 hyper- or hypomethylated loci, respectively, at their promoter. Visualization of methylation patterns around these transcripts, together with their expression data (Supplementary Figure 9) revealed distinctly high PIGR promoter methylation in IBD-CRCs, with significantly reduced epithelial expression (Figure 6A and B; Supplementary Table 7; Supplementary Figure 10). An IBD-CRC harboring 2 somatic PIGR mutations displayed lower methylation values than IBD-CRCs with wild-type PIGR (n = 18) (Figure 6A).

DML calling and annotation comparing the 2 tumor groups (Supplementary Figures 11 and 12) revealed multiple hypomethylated DMLs at OSMR promoter region in IBD-CRCs compared to sCRCs (Figure 6C, Supplementary Figures 10 and 12). Stromal OSMR expression was strongly elevated in IBD-CRCs and particularly high in CMS4 tumors (Figure 6D), showing a strong correlation with expression of EMT-related genes especially in IBD-CRCs (Supplementary Figure 13). Inspection of protein coding genes having DMLs at gene bodies (exons, introns, and 5′/3′UTRs) revealed enrichment of WNT pathway genes for gene body hypermethylation in IBD-CRC compared to sCRC

(Supplementary Table 14), including AXIN2 and RNF43 strongly down-regulated in IBD-CRCs (Supplementary Table 15, Supplementary Table 7).

Discussion

Increased risk of CRC in IBD patients has been attributed to long-standing chronic inflammation, yet the underlying mechanisms remain poorly understood.[3,4] Here, we integrated data from multiple experimental approaches to gain a deeper understanding of IBD-CRC. While many tumor characteristics, including somatic AI, SVs, and genome-wide methylation patterns, showed an overall similarity to sCRCs, several distinct features were identified, as summarized in Supplementary Table 17.

Constitutive activation of WNT/β-catenin signaling via loss-of-function mutations in APC is a hallmark of sCRC.[4] The pathway regulates cell fate, proliferation, polarity, and stemness via β-catenin–dependent and –independent mechanisms.[56] The most common transcriptomic CRC subtype, CMS2, displays up-regulation of WNT and MYC target genes and an epithelial differentiation signature.[17] Among our MSS sCRCs, 39% were CMS2. In striking contrast, CMS2 was completely absent among IBD-CRCs, dominated instead by mesenchymal CMS4 previously associated with EMT, matrix remodeling, transforming growth factor β (TGF-β) signaling, complement activation, and depletion of WNT/MYC-related expression signatures.[17] Differential expression analysis reflected this CMS2 vs CMS4 predominance, revealing in IBD-CRCs enrichment of gene sets involved in extracellular matrix organization and complement activation, up-regulation of key transcription factors mediating EMT, and down-regulation of genes related to epithelial differentiation. Moreover, binding sites of TCF7L2 and MYC, mediators of canonical WNT signaling, were hypomethylated in sCRCs relative to IBD-CRCs. Thus, an alternative mode of WNT dysregulation may govern tumorigenesis in IBD-CRCs that still show accumulation of nuclear/cytoplasmic β-catenin.[11]

EMT involves loss of tumor cell polarity and cell–cell adhesion to gain more migratory and invasive properties.[57] WNT signaling is a prominent regulator of EMT in CRC[57] and could thus contribute to the observed mesenchymal skewing of IBD-CRCs. One mechanism could involve the differential activity of HNF4α transcription factor, essential for embryonic development of colonic epithelium.[58] We showed that HNF4α binding sites were highly significantly hypomethylated in sCRCs but not in IBD-CRCs, implying increased binding in sCRC tumorigenesis. Chromosomal gains at HNF4A locus were more common in sCRCs, enriching in CMS2 tumors. In hepatocellular carcinoma, HNF4α maintains epithelial tumor phenotype by facilitating corepression of EMT-related WNT/β-catenin target genes,[59] putting a break on EMT. HNF4A gains could enforce this mechanism in sCRCs with constitutive WNT/β-catenin activation, promoting the establishment of epithelial CMS2 tumors. Conversely, in a mouse model of colitis, inflammation-induced epigenetic change resulted in reduced HNF4α occupancy.[60] Thus, a state of low HNF4α
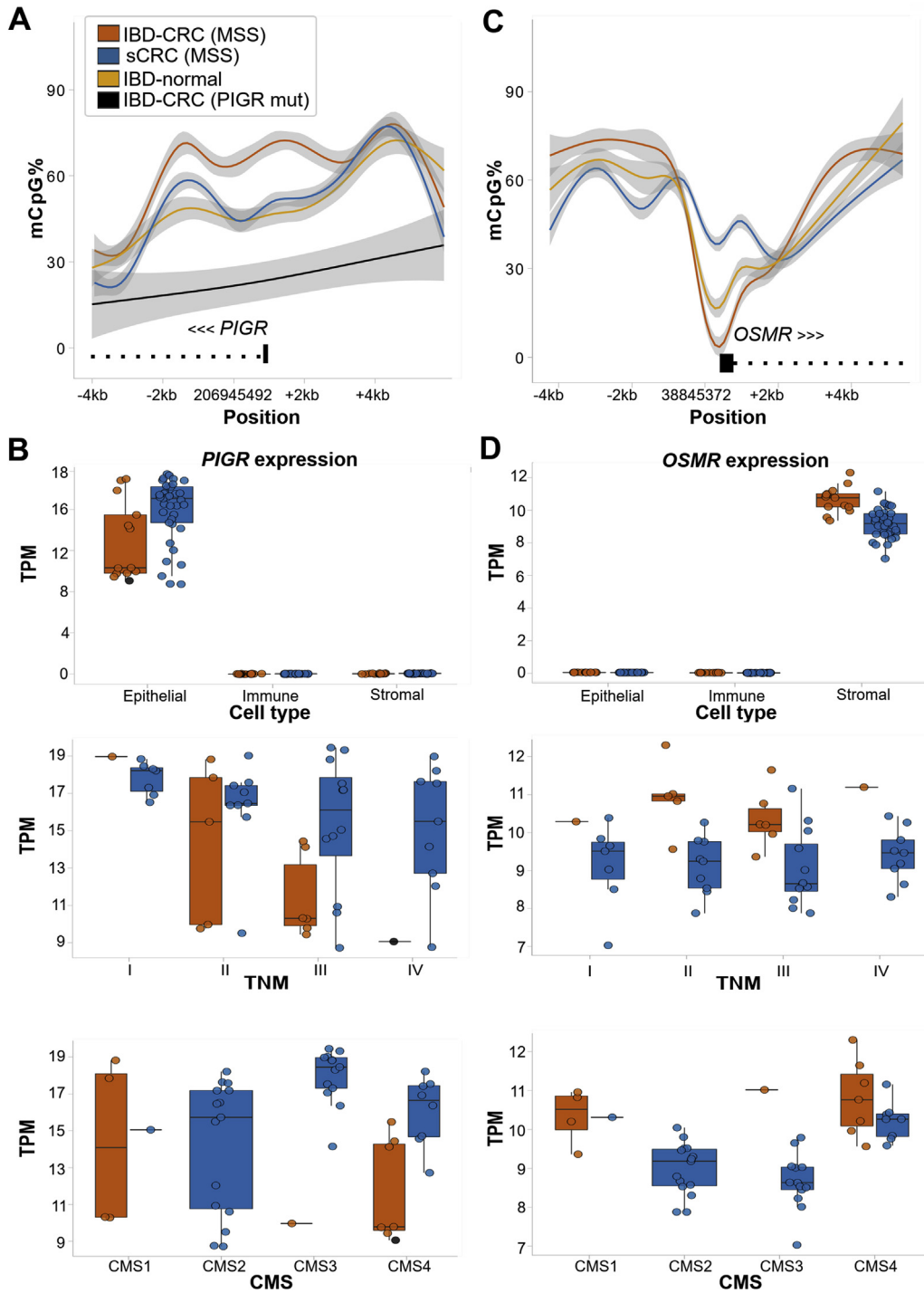
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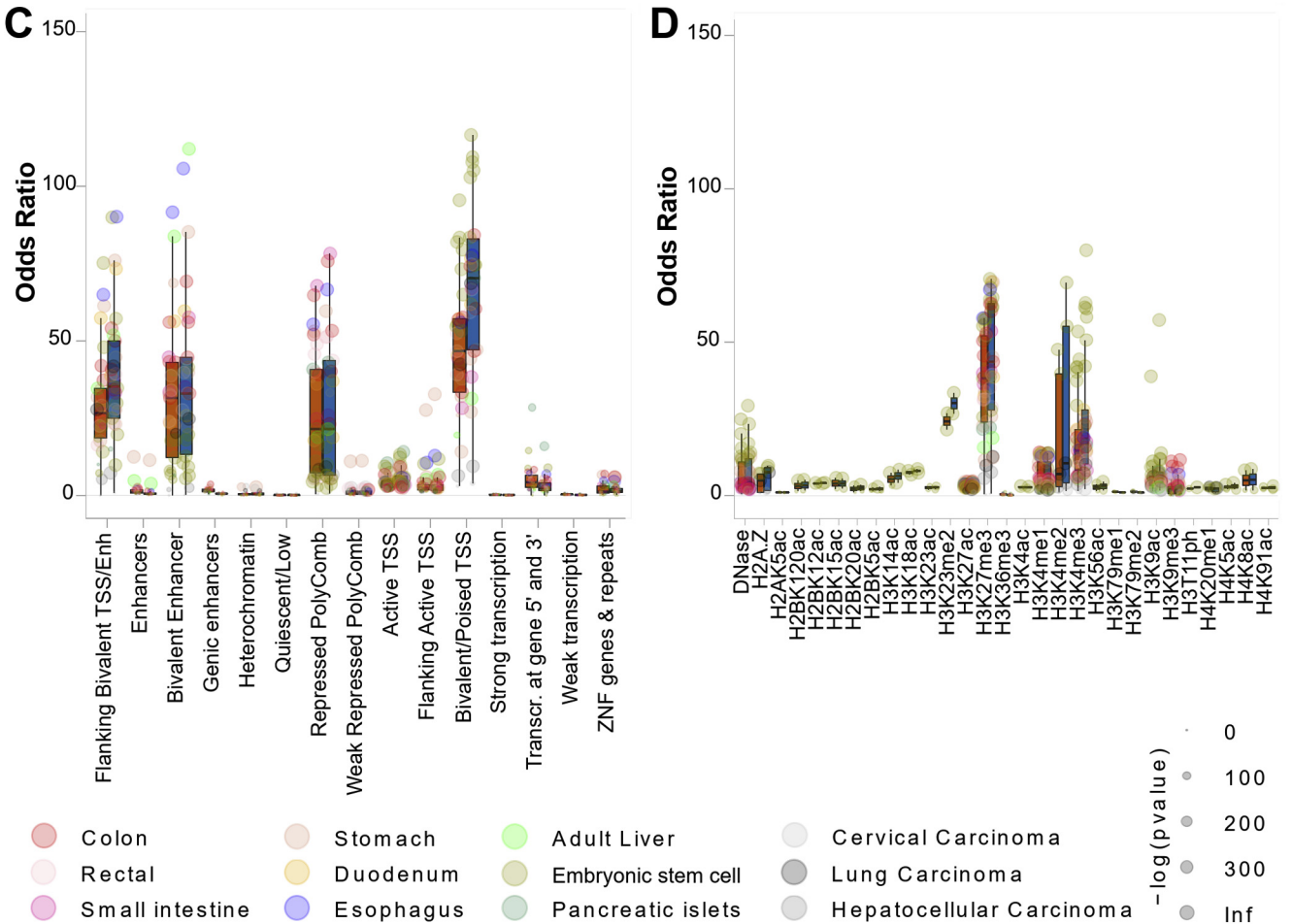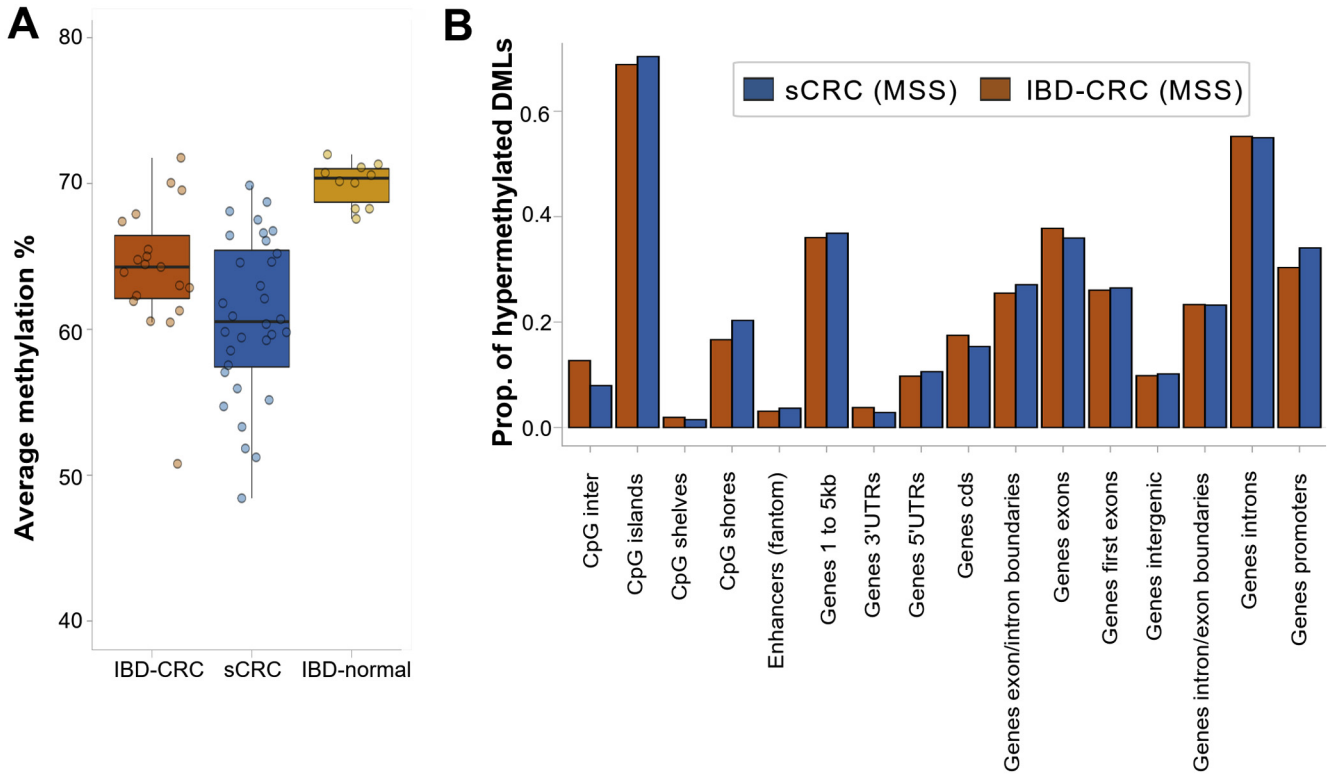
Figure 6. Patterns of PIGR and OSMR methylation and expression. (*A*) Smoothing of the methylation measurements around *PIGR* from 18 *PIGR* wild-type vs 1 *PIGR*-mutated IBD-CRC, 32 *PIGR* wild-type sCRCs, and 10 IBD-normals. (*B*) Deconvoluted *PIGR* expression data; data from epithelial cells are plotted against tumor-node-metastases (TNM) stage and CMS. (*C*) Smoothing of the *OSMR* methylation data. (*D*) Deconvoluted *OSMR* expression data; data from stromal cells are plotted against TNM stage and CMS.

Figure 5. Genome-wide annotation of methylation patterns. (*A*) Average genome-wide methylation levels from 19 MSS IBD-CRCs, 32 MSS sCRCs, and 10 nondysplastic normal colon tissue samples from IBD patients. (*B*) Proportion of hypermethylated DMLs in tumors compared to IBD normals at different genomic regions. (*C, D*) Enrichment analysis of hypermethylated loci using Roadmap Epigenomics chromatin annotations from tissue samples of the normal human gastrointestinal tract and samples relevant to tumorigenesis. Different *colors* depict different tissue origins of the cells (Supplementary Table 12).

occupancy may pre-exist in patients with IBD. Once tumorigenesis is set forth by *TP53* mutations and aberrant WNT signaling, it may favor the emergence of mesenchymal CMS4 tumors through EMT from the earliest stages of tumorigenesis.

The paucity of *APC* and *KRAS* mutations in IBD-CRC has been identified consistently.[4,6–10] Novel driver mutations explaining the relative independence of IBD-CRC from these key sCRC driver events have not emerged. We revealed evidence for positive selection of somatic mutations at noncoding 5′UTR of *TP53* in IBD-CRCs, resulting in low *TP53* expression. A recent pan-cancer analysis described similar variants in this region as a novel rare driver event reducing *TP53* expression,[61] further highlighting the importance of *TP53* loss in IBD-CRC.[4] We could not identify a distinct subgroup of MSS IBD-CRCs with elevated genome-wide mutation densities, as suggested earlier by panel sequencing.[62]

Two negative feedback regulators induced by WNT signaling, *AXIN2* and *RNF43*, were strongly down-regulated in IBD-CRCs compared to sCRCs. *AXIN2* participates with APC in the β-catenin destruction complex,[63] while *RNF43* inhibits upstream WNT signaling by inducing the degradation of frizzled receptors.[64] Both genes are commonly mutated in MSI sCRCs showing mutual exclusivity with APC mutations[63,65] and down-regulated in *APC* wild-type compared to *APC*-mutated MSS sCRCs.[66,67] Thus, loss of WNT regulation by *AXIN2* and *RNF43* could trigger aberrant WNT signaling in *APC* wild-type CRCs enriched among MSI-CRCs and IBD-CRCs. Moreover, disruption of *AXIN2* and *RNF43* in serrated adenomas and carcinomas[66,68,69] may link them with nonconventional types of precursor lesions common in IBD-CRC.[4] Intriguingly, model organoids mimicking sessile serrated adenomas, but not conventional tubular adenomas, are responsive to TGF-β–induced development toward CMS4.[70]

Tumor immune cell infiltrate in CMS4 IBD-CRCs was dominated by CD4$^+$ T cells and monocytes. In agreement, CMS4 has been linked with an "immune inflamed" immunosuppressive microenvironment enriched in CD4$^+$ regulatory T cells and monocyte-derived macrophages.[71] The remaining IBD-CRCs were dominated by B-cell infiltration. While the role of B lymphocytes in cancer remains poorly understood, this could result from an attempt to compensate for reduced *PIGR* expression and disrupted epithelial transport of B cell–derived IgA.

IBD-CRCs tended to show higher genome-wide methylation compared to sCRCs, but decreased exposure to mutational signature linked with age-driven deamination of 5′methylcytosines; the latter was found also when compared with tumors from patients with Lynch syndrome.[62] Neither were explained by the younger age at diagnosis or by differential expression of DNMT or TET enzymes regulating DNA methylation, suggesting inflammation-driven methylation changes.

Aberrant methylation patterns were detected in 2 genes related to mucosal immunity solely in IBD-CRCs. *PIGR* mediates IgA and IgM transport across intestinal epithelium[72]

and accumulates somatic loss-of-function mutations in IBD-affected colon.[73–75] *PIGR* showed promoter hypermethylation in IBD-CRCs compared to IBD-affected non-dysplastic colon, with down-regulated expression. This may imply further selection for *PIGR* loss during tumorigenesis, which could promote epithelial barrier dysfunction and sustained inflammation.[72] Alternatively, the progressive loss of *PIGR* with increasing tumor stage could reflect gradual loss of epithelial properties in tumor cells. *OSMR*, encoding a receptor for cytokine oncostatin M (*OSM*), showed promoter hypomethylation, frequent chromosomal gain, and strong overexpression in IBD-CRCs. Deconvolution analysis pointed toward *OSMR* overexpression originating from stromal cells. In agreement, OSMR-positive inflammatory stromal cells expand in IBD-affected colon.[76] Importantly, elevated intestinal *OSMR* and *OSM* expression is associated with a subgroup of IBD patients showing poor response to TNFα blockers.[76] As OSM stimulation triggers EMT and mesenchymal characteristics in breast cancer cells,[77] increased OSM-OSMR signaling in IBD could favor the establishment of mesenchymal CRC subtype, although this remains to be studied further.

Our study represents one of the largest next-generation sequencing datasets on both IBD-CRC and sCRC patients published so far. Multiple data layers allowed deep characterization of these tumors. Even larger studies would be needed to control for more detailed clinical parameters, such as UC/CD differences, changes in anti-inflammatory medication for IBD over time, and different patient population backgrounds. Further limitations include lack of non-IBD normal colon in methylation analyses, and potential batch effects. The rich layers of data created, focus on results showing connections between different data sets, and rigorous batch-to-batch quality control mitigate many of these effects. For methylation analyses, IBD-CRCs were compared with both IBD normal colon and sCRCs to better capture differences distinguishing IBD-associated tumorigenesis.

Taken together, our results highlight skewing of IBD-associated tumorigenesis toward increased importance of the dynamic process of EMT, driven by distinct mechanisms of WNT pathway dysregulation compared to sCRC. The mesenchymal tumor subtype enriched in IBD-CRCs has been associated with drug resistance[18] and worse relapse-free and overall survival in large CRC patient cohorts,[17] which may affect prognosis and treatment options in IBD-CRC. *OSM-OSMR* signaling could contribute to establishment of this mesenchymal subtype in IBD patients, potentially linking a TNFα blocker–resistant subgroup of patients with increased CRC risk. Cost-effective and clinically feasible alternatives for CMS subtyping, including immunohistochemistry, are reaching high accuracy.[18] First treatment combinations targeted toward CMS4-like TGF-β–activated CRC have entered clinical trials,[18] paving the way for overcoming the tumor-microenvironment crosstalk that drives the mesenchymal phenotype. Further studies are needed to determine whether similar approaches are beneficial in treatment of IBD-CRC.

Supplementary Material

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Päivi Sulo, MSc (Data curation: Supporting; Formal analysis: Supporting; Investigation: Supporting; Methodology: Supporting; Software: Supporting; Visualization: Supporting; Writing – review & editing: Supporting).

Janne Ravantti, PhD (Methodology: Supporting; Resources: Equal; Software: Supporting; Writing – review & editing: Supporting).

Suvi Lehtipuro, PhD (Methodology: Supporting; Software: Supporting; Writing – review & editing: Supporting).

Kirsi J. Granberg, PhD (Methodology: Supporting; Writing – review & editing: Equal).

Matti Nykter, PhD (Methodology: Supporting; Supervision: Supporting; Writing – review & editing: Supporting).

Tomas Tanskanen, MD, PhD (Data curation: Supporting; Investigation: Supporting; Writing – review & editing: Supporting).

Ari Ristimäki, MD, PhD (Data curation: Equal; Resources: Equal; Writing – review & editing: Supporting).

Selja Koskensalo, MD, PhD (Resources: Equal; Writing – review & editing: Supporting).

Laura Renkonen-Sinisalo, MD, PhD (Resources: Equal; Writing – review & editing: Supporting).

Anna Lepistö, MD, PhD (Resources: Equal; Writing – review & editing: Supporting).

Jan Böhm, MD, PhD (Investigation: Supporting; Methodology: Supporting; Writing – review & editing: Supporting).

Jussi Taipale, PhD (Conceptualization: Supporting; Resources: Supporting; Writing – review & editing: Supporting).

Jukka-Pekka Mecklin, MD, PhD (Funding acquisition: Equal; Supervision: Supporting; Writing – review & editing: Supporting).

Mervi Aavikko, PhD (Conceptualization: Equal; Data curation: Equal; Formal analysis: Supporting; Investigation: Equal; Project administration: Equal; Supervision: Lead; Writing – review & editing: Equal).

Kimmo Palin, PhD (Conceptualization: Supporting; Data curation: Equal; Formal analysis: Equal; Investigation: Equal; Methodology: Equal; Software: Equal; Supervision: Lead; Visualization: Supporting; Writing – review & editing: Supporting).

Lauri A. Aaltonen, MD, PhD (Conceptualization: Equal; Funding acquisition: Lead; Project administration: Lead; Supervision: Lead; Writing – review & editing: Equal).