Systematic analysis of chromatin interactions at disease associated loci links novel candidate genes to inflammatory bowel disease

Claartje A. Meddens¹, Magdalena Harakalova², Noortje A. M. van den Dungen², Hassan Foroughi Asl³, Hemme J. Hijma¹, Edwin P. J. G. Cuppen⁴,⁵, Johan L. M. Björkegren³,⁶,⁷, Folkert W. Asselbergs²,⁸,⁹, Edward E. S. Nieuwenhuis¹ and Michal Mokry¹∗

Abstract

Background: Genome-wide association studies (GWAS) have revealed many susceptibility loci for complex genetic diseases. For most loci, the causal genes have not been identified. Currently, the identification of candidate genes is predominantly based on genes that localize close to or within identified loci. We have recently shown that 92 of the 163 inflammatory bowel disease (IBD)-loci co-localize with non-coding DNA regulatory elements (DREs). Mutations in DREs can contribute to IBD pathogenesis through dysregulation of gene expression. Consequently, genes that are regulated by these 92 DREs are to be considered as candidate genes. This study uses circular chromosome conformation capture-sequencing (4C-seq) to systematically analyze chromatin-interactions at IBD susceptibility loci that localize to regulatory DNA.

Results: Using 4C-seq, we identify genomic regions that physically interact with the 92 DRE that were found at IBD susceptibility loci. Since the activity of regulatory elements is cell-type specific, 4C-seq was performed in monocytes, lymphocytes, and intestinal epithelial cells. Altogether, we identified 902 novel IBD candidate genes. These include genes specific for IBD-subtypes and many noteworthy genes including ATG9A and IL10RA. We show that expression of many novel candidate genes is genotype-dependent and that these genes are upregulated during intestinal inflammation in IBD. Furthermore, we identify HNF4α as a potential key upstream regulator of IBD candidate genes.

Conclusions: We reveal many novel and relevant IBD candidate genes, pathways, and regulators. Our approach complements classical candidate gene identification, links novel genes to IBD and can be applied to any existing GWAS data.

Keywords: Inflammatory bowel disease, Genetics, Epigenetics, Genome-wide association studies (GWAS), Enhancer elements, Chromatin interactions, DNA regulation, Candidate genes
Background

Inflammatory bowel disease (IBD) is an inflammatory disorder of the gastro intestinal tract with an intermittent, chronic, or progressive character. Studies on the pathogenesis of IBD have elucidated the involvement of a broad range of processes that mainly regulate the interaction between the intestinal mucosa, the immune system, and microbiota [1]. A role for genetics in the pathogenesis of IBD has been established through twin-based, family-based, and population-based studies [1]. Subsequently, a substantial effort to identify genetic elements involved in the IBD pathogenesis followed. In this respect, multiple genome-wide association studies (GWASs) have been performed over the past years [2–5]. In these studies, common genetic variants (single nucleotide polymorphisms (SNPs)) are assayed across the whole genome in search of variants that are significantly over-represented or under-represented in patients compared to healthy controls. Although GWASs have revealed many IBD-associated loci, for most loci the causal genes that led to the associations have not been identified. Furthermore, the majority of IBD-associated SNPs are located in non-coding DNA and therefore cannot be causal in the sense that they directly lead to amino acid changes at the protein level [2–4, 6–9]. Therefore, these SNPs are generally thought to be markers for disease-causing variants in nearby genes. This model is used in classical approaches for candidate gene identification. These approaches are mainly based on the selection of genes that have shared functional relationships and are localized in the vicinity of the identified loci [10, 11]. This has led to the identification of crucial genes and pathways involved in the IBD pathogenesis [12]. However, over the past decade it has been established that besides genes, the human genome consists of many other functional elements in the non-protein-coding regions. These regions of the genome can play a role in the pathogenesis of complex diseases. As such, many types of DNA regulatory elements (DRE), especially enhancer elements, are involved in establishing spatiotemporal gene expression patterns in a cell type-specific manner [13]. These elements are crucial in the regulation of developmental processes and in maintaining cell type-specific functionality. It is therefore now widely appreciated that part of the GWAS associations is due to sequence variation in DRE, but this information has largely been ignored in candidate gene identification [9, 14–18].

We have recently shown that 92 of 163 IBD GWAS susceptibility loci localize to DRE (identified through the presence of H3K27Ac in relevant cell types) [9]. DRE are involved in transcription regulation and establishing cell type-specific expression patterns [19]. The genes that are regulated by the IBD-associated elements are likely to play a role in IBD and can therefore be considered as IBD candidate genes. This information has not been used in previous candidate gene approaches, because the identification of these genes comes with several hurdles. Since regulatory elements can regulate genes via chromatin–chromatin interactions that comprise up to 1 Mb [20, 21], these genes cannot be identified based on their linear distance from the regulatory regions. Classical methods for candidate gene identification, that take regulatory mechanisms into account, have mainly been restricted to computational approaches [14, 16, 22, 23]. So far, a limited number of studies have shown the value of using physical interactions between regulatory elements and the genes they regulate through studying the three-dimensional (3D) nuclear conformation chromatin interactions in GWAS interpretation. These studies analyzed either single interactions (3C) or many-vs-many interactions (Hi-C) and were performed in colorectal cancer, auto-immune diseases, and multiple other diseases [24–27]. In contrast to these approaches we make use of circular chromosome conformation capture-sequencing (4C-seq), thereby increasing the number of analyzed interactions compared to 3C and increasing the resolution compared to Hi-C. Our study provides the first systematic analysis of chromatin interactions between disease-associated DRE and candidate genes in IBD. We have identified 902 novel IBD candidate genes, consisting of many noteworthy genes, for example IL10RA, SMAD5, and ATG9A.

Results

Genes interacting with DRE at IBD associated loci

A meta-analysis on GWASs performed in IBD resulted in the confirmation of 163 susceptibility loci [3]. We have recently shown that 92 of these 163 loci overlap with enhancer elements (regulatory elements that enhance transcription) that are active in relevant cell types for IBD (i.e. intestinal epithelial cells and immune cells) [9]. We now use this information to identify novel IBD candidate genes. We do so by identifying the genes that are regulated by these 92 regulatory elements. Since the regulated genes cannot be pinpointed by studying the linear organization of the susceptibility loci, we assayed the 3D conformation of these loci (Fig. 1). The effect of common variants, especially those in regulatory elements, is relatively mild. Therefore, it is very unlikely that a single common variant will ablate or create a whole regulatory region and its 3D interaction [28]. By the same reasoning, we do not expect that the 3D interactions in patients will be fundamentally different compared to healthy controls or cell lines. However, regulation of genes can be genotype specific [16], which demands for the identification of genes that are dysregulated in IBD. For these reasons, we decided on an experimental setup where we assay chromatin conformation in healthy control cells and a cell line,
to identify genes that can be dysregulated in IBD under pathological conditions. Therefore, we have performed 92 high resolution 4C-seq experiments to cover all individual IBD susceptibility loci that overlap DRE in three cell types, thereby creating 276 individual chromatin interaction datasets. This way, we could identify all genes that physically interact with the regulatory elements that are found at IBD associated loci. As the activity of enhancers is known to be cell type-specific [19], we assayed chromatin interactions in monocytes (i.e. CD14+ fraction of PBMCs), lymphocytes (i.e. CD14- fraction of PBMCs), and in an intestinal epithelial cell line (DLD-1, derived from colorectal adenocarcinoma).

4C-seq identifies different sets of candidate genes in different cell types

The candidate genes that we report here all meet the following criteria: (1) the enhancer element physically interacts with the candidate gene ($p > 10^{-8}$); (2) the enhancer element is active in the assayed cell type (i.e. the associated variant or a variant in LD co-localizes with the histone mark H3K27Ac) [9]; and (3) the candidate gene is expressed in the assayed cell type (log2(RPKM) > -0.5). With this approach we identified 1409 candidate genes: 923 genes in monocytes, 1170 in lymphocytes, and 596 in DLD-1 cells, of which 796 were shared by two or more cell types and 810 were found in only one cell type (Fig. 2a and b). We identified 902 IBD candidate genes that have not been reported by GWASs before (Table 1, Additional file 1: Table S2). Of the 92 studied loci, 22 are associated to only one of the IBD subtypes (11 to Crohn’s disease, 11 to ulcerative colitis). The candidate genes that were identified for these loci might contribute to the mechanisms that lead to the subtype-specific phenotypes. Interestingly, for two loci on chromosome 7 that give separate GWAS signals for CD (rs10486483) and UC (rs4722672), the 10 candidate genes that were identified for this CD locus were also found in the UC locus. This implies that in some cases, although the genetic risk factor is different between the subtypes, the mechanisms that underlie the genetic risk can share downstream components. Notably, this UC locus is active in intestinal epithelium, whereas the CD locus is not, which resulted in the identification of additional candidate genes for rs4722672 that are UC-specific (Table 1). Among the identified candidate genes are many noteworthy genes that have been implicated in the IBD pathogenesis, but that were never identified through GWAS associations (Table 2 [29–35]). We have now identified these novel candidate genes that have been missed by classical approaches for candidate gene identification.

As expected, based on their common hematopoietic origin, the two immune cell types show larger overlap compared to DLD-1 cells (Fig. 2b, Additional file 2: Figure S5). With a median enhancer-to-gene distance of
261, 370, and 354 kbp in DLD-1, lymphocytes, and monocytes respectively, a large proportion of the genes we report are found outside the GWAS susceptibility loci (Fig. 2c). Notably, some of the interactions between IBD loci and candidate gene span over 5 Mb. For example, rs925255 shows a significant ($p = 6.068 \times 10^{-9}$) physical interaction with TANK (TRAF family member-associated NF-κB activator), a gene that is localized 30 Mb from this locus (Additional file 1: Table S2).

**Validation and reproducibility of 4C-seq data**

To validate the reproducibility of our data, we prepared a 4C template from lymphocytes from a different donor and performed 4C-seq for the 92 regions on this material. Additional file 2: Figure S4A shows that 91% of the candidate genes that are identified in the replicate dataset were also identified in the dataset that is used throughout this study. This demonstrates the reproducibility of the 4C technique, not only in technical, but also in biological duplicates. These results are in line with studies that have previously shown that in 3C-based methods, results from biological duplicates are highly reproducible [36]. Furthermore, we validated the reproducibility of our data by intersecting the 4C datasets with Hi-C datasets that were created in CD34+ leukocytes and a lymphoblastoid cell line [25]. This confirmed a high reproducibility by showing that 99% (CD34+) and 87% (lymphoblastoid) of the genes that were found by Hi-C were also found in our 4C data (Additional file 2: Figure S4B).

**Identified candidate genes are actively expressed**

We reasoned that genes that are truly regulated by active enhancers in vivo would, on average, be more highly expressed than other genes within the region of the 4C signal. The quantitative examination of expression levels and histone modifications that mark active enhancers and promoters confirmed that the genes that were detected by our method indeed are more actively
transcribed than all other genes (also than genes that were not detected by 4C and are found in the same genomic region, Additional file 2: Figures S6 and S7). These results support the detection of functional interactions by the 4C-seq approach that was executed here. Furthermore, we assessed “possible” insulator elements (i.e., insulators occupied by CTCF protein) between the 92 DRE and the candidate genes. Interestingly, the majority of interactions bypasses several CTCF sites and numerous interactions skip over 50 sites bound by CTCF (Additional file 2: Figure S8). In addition, genes that do not interact with the 4C viewpoint do not seem to have more CTCF sites between the viewpoint and their promoter compared to the interacting genes (Additional file 2: Figure S8).

| Table 1 4C-seq output per locus |
|-------------------------------|
| **Position** Chr. (Mb) | SNP | Key Novel Genes | Novel (total) | **Position** Chr. (Mb) | SNP | Key Novel Genes | Novel (total) |
|-------------------------------|
| IBD 1 8.02 r35675666 | CTNBNP1 | 12(17) |
| IBD 1 151.79 r48456004 | NOTCH2NL | 25(36) |
| IBD 1 155.67 r670523 | MVC1 | 34(55) |
| IBD 1 200.87 r7554511 | 3(7) |
| IBD 1 206.93 r3024505 | CDS5, CD46, IKBKE | 9(12) |
| IBD 1 22.7 r12568930 | CDC42, EPH2B | 5(6) |
| IBD 2 25.12 r65545800 | 4(8) |
| IBD 2 219.14 r2382817 | ATG9A | 14(25) |
| IBD 2 43.81 r10495903 | 9(12) |
| IBD 2 102.86 r971799 | 8(12) |
| IBD 2 191.92 r1517352 | 8(12) |
| IBD 2 28.61 r925255 | TANK | 4(11) |
| IBD 3 48.96 r3197999 | HYAL1, HYAL2, HYAL3 | 43(90) |
| IBD 3 18.76 r4265159 | KAT2B | 7(7) |
| IBD 4 74.85 r2472649 | AREG | 4(10) |
| IBD 5 176.79 r12654812 | FGF8 | 15(30) |
| IBD 5 96.24 r1363907 | 8(13) |
| IBD 5 131.19 r2188962 | SKP1 | 17(32) |
| IBD 5 141.51 r6863411 | HADC3 | 13(18) |
| IBD 5 10.69 r2930047 | 5(6) |
| IBD 5 40.38 r11742570 | RICTOR | 9(11) |
| IBD 6 138 r6920220 | IFNGR1 | 12(13) |
| IBD 6 167.37 r1819333 | DLI1 | 9(13) |
| IBD 6 90.96 r1847472 | MAP3K7 | 7(8) |
| IBD 7 50.245 r1456896 | 3(5) |
| IBD 7 100.34 r1734907 | 45(61) |
| IBD 8 126.53 r921720 | KIAA0196 | 3(5) |
| IBD 8 130.62 r1998166 | MYC | 2(2) |
| IBD 9 139.32 r10781499 | TRAF2 | 12(28) |
| IBD 10 94.43 r7911264 | 3(7) |
| IBD 10 82.25 r6585090 | 6(9) |
| IBD 10 6.08 r12722515 | PRKCO | 10(17) |
| IBD 10 101.28 r4409764 | 1(7) |
| IBD 10 35.3 r11010067 | FZD8 | 3(7) |
| IBD 10 30.72 r1042058 | ITGB1 | 7(9) |
| IBD 10 81.03 r1250546 | DLG5 | 4(8) |
| IBD 10 59.99 r2790216 | 0(4) |
| IBD 11 1.87 r907611 | TOLLIP, DUSP8 | 19(25) |
| IBD 11 65.65 r2231884 | PEI3 | 53(76) |
| IBD 11 61.56 r4246215 | 5(16) |
| IBD 11 76.29 r2155219 | 8(11) |
| IBD 11 118.74 r630923 | IL10RA | 15(28) |
| IBD 11 60.77 r11230563 | 12(20) |
| IBD 12 68.49 r7134599 | MDM2 | 5(6) |
| IBD 12 40.77 r11564258 | 2(3) |
| IBD 12 12.65 r11612508 | 4(12) |

This table shows the 4C output for each associated IBD locus that overlaps an active regulatory element based on the presence of H3K27Ac and is arranged based on the GWAS association of the loci to either Crohn’s disease (CD), ulcerative colitis (UC), or both (IBD). The positions are given as in the GWAS in which the association was found [3] and are relative to human reference genome GRCh37. Depicted are the name of the SNP and the key novel genes that were identified at that locus. The numbers refer to the number of novel genes that were identified at that locus and, between brackets, the total number of candidate genes identified at that locus.
Table 2  Noteworthy novel candidate genes

| Gene Symbol | Gene Name and Description |
|-------------|---------------------------|
| ATG9A       | ATG9A encodes autophagy related protein 9A. Autophagy plays an import role in host defense by eliminating pathogens. ATG-family member ATG16l1 has previously been associated with Crohn’s disease [32]. |
| BATF       | Basic leucine zipper transcription factor ATF-like (BATF) belongs to the activator protein 1 family that is involved in transcription regulation in all immune cells. Batf-deficient mice do not develop Th17 cells and do not produce IL17. Furthermore, BATF regulates cell type specific gene expression in Th2 cells, germinal center B-cells, and T-follicular helper cells [30]. |
| CD46/CD55  | CD46 (also known as MCP) and CD55 (also known as DAF) are regulatory proteins expressed on surface membranes. These proteins protect the host from autologous complement-mediated injury upon activation of the complement cascade. Daf-deficient mice show increased epithelial damage upon induction of colitis, delayed healing, and elevated expression of proinflammatory cytokines [31]. |
| IL10RA     | The IL10-receptor consists of the two subunits IL10RA and IL10RB. Sequence variants in genes encoding these two subunits are known to cause severe very early onset IBD in a monogenic fashion [32]. While the association of IL10RB with the complex form of IBD was reported by GWASs, the link with IL10RA was so far missing. |
| SMAD5      | SMAD5 is a downstream effector in BMP signaling. SMAD5 expression was found to be downregulated in intestinal cells of IBD patients. Furthermore, conditional deletion of Smad5 in mice results in increased susceptibility for development of colitis upon DSS-induction (dextran sulfate sodium) [34]. |

IL10RA interleukin 10 receptor subunit alpha, IL10RB interleukin 10 receptor subunit beta, Th17 cells T-helper 17 cells, Th2 cells T-helper 2 cells, SMAD named after their homologous genes Mothers Against Decapentaplegic (MAD) and the Small Body Size protein (SMA) in Drosophila and C. Elegans, respectively, CD complement-decay accelerating factor, MCP membrane co-factor protein, DAF decay accelerating factor

is in line with observations from Hi-C datasets where 82% of long-range interactions bypass at least one CTCF site [25].

Previously, insulator regions have been shown to prevent enhancer-gene interactions [37]. We therefore investigated whether assessment of the CTCF binding can be used as an alternative to the 4C method by predicting the borders of the regions in which our candidate genes were found. We conclude that CTCF binding information cannot be used as an alternative for the 4C-based candidate gene approach presented here.

4C-seq candidate genes have SNP-dependent expression profiles

We hypothesize that the candidate genes that we identify are contributing to the IBD pathogenesis via impaired transcription regulation caused by variants in DRE. To test this hypothesis, we studied whether 4C-seq candidate genes show different expression profiles in different genetic backgrounds (i.e. in individuals that carry the associated SNP versus individuals that do not) through eQTL analyses [23]. We performed two different analyses in separate databases. First, we used the GTEx database [38] to test whether our approach is able to detect the eQTLs that are present in the intestinal epithelium (colon-sigmoid, colon-transverse, terminal ileum) and whole blood [38]. We performed an eQTL look-up of the 92 IBD-associated SNPs in these tissues and found 50 genes with a SNP-dependent expression profile. Interestingly, all of the 50 genes were identified by our 4C-seq approach (Additional file 3: Table S4). Second, we made use of another eQTL database (STAGE) [39] and explored the presence of candidate genes among the genes that were found to have expression levels that are dependent on the interacting SNP genotype in white blood cells. This revealed 10 candidate genes that have an eQTL in the STAGE database. Next, we analyzed all non-interacting genes within 2 Mb from the 4C viewpoint (Additional file 3: Table S4). In contrast to the interacting genes, none of the non-interacting genes showed genotype-dependent expression in the same database. These findings altogether support the capability of our method to identify the candidate genes of which the expression regulation is dependent on IBD-associated genomic variants.

4C-seq gene set is enriched in genes involved in inflammation in IBD patients

After demonstrating that our method enables the identification of novel IBD candidate genes that are likely subject to SNP-dependent expression levels, we examined whether the genes we report here are involved in the major pathogenic process in IBD, namely intestinal inflammation. To address this, we performed a GSEA [40] in which we used RNA expression data of intestinal biopsies from IBD patients [41]. We compared expression levels in inflamed versus non-inflamed intestinal biopsies and tested whether the 4C-seq candidate genes were enriched among the differentially expressed genes. This analysis shows that all three 4C gene sets (monocytes, lymphocytes, and intestinal epithelium) are highly enriched (p < 0.001) for genes that are upregulated upon intestinal inflammation in IBD patients (Fig. 3). These results support the role of the candidate genes reported here in intestinal inflammation in IBD.

Chromatin interactions reveal IL10RA and ATG9A as novel IBD targets

IL10RA is one of the newly identified candidate genes. Previously, sequence variants in genes encoding the two subunits of the interleukin 10 receptor, IL10RA and IL10RB, were found to cause severe early onset IBD in a Mendelian fashion [32]. Our 4C datasets reveal that IL10RA interacts with an IBD-associated enhancer element in peripheral blood lymphocytes (p = 4.1 × 10^-10). Since IL10RA is located ~1 Mbp upstream of the associated SNP (rs630923) and is separated from the SNP by multiple haploblocks (Fig. 4a), this gene has not been identified through classical candidate gene approaches.
The enhancer element that co-localizes with rs630923 is active in lymphocytes, but not in monocytes and intestinal epithelial cells (i.e. H3K27Ac marks are present only in lymphocytes). These results imply distinctive and cell type-specific regulatory pathways for IL10RA expression in immune cells. Besides IL10RA, we identified 12 candidate genes that are part of the IL10 signaling pathway (Fig. 4b), three of which are novel candidate genes (IL10RA, IKBKE, MAP3K7). These results confirm and further establish the important role of IL10 signaling in IBD.

Furthermore, we identified ATG9A (autophagy-related gene 9A) as a novel candidate gene, as its transcriptional start site is physically interacting with an enhancer element in the proximity of rs2382817 in DLDs and monocytes \( (p = 7.891 \times 10^{-13}\) in monocytes, \( p = 9.787 \times 10^{-12}\) in DLDs, Additional file 2: Figure S9). ATG9A is known to be involved in the generation of autophagosomes. Furthermore, ATG9A has been shown to dampen the innate immune response that occurs in response to microbial dsDNA. ATG9A knockout mice show enhanced expression of IFN-β, IL6, and CXCL10 upon exposure to microbial dsDNA [42]. This gene is furthermore of interest to IBD, because the association of other autophagy genes to IBD is well established [6, 43, 44]. For example, patients that are homozygous for the ATG16L risk allele show Paneth cell granule abnormalities [45]. Based on the role ATG9A plays in responding to microbial dsDNA and the role ATG16L plays in Paneth cell degranulation, it is possible that ATG9A contributes to the IBD pathogenesis in monocytes and intestinal epithelial cells via distinct mechanisms.

**Pathway analysis shows cell type-specific results**

Besides studying the individual associated loci and the genes they regulate, we aimed to elucidate the pathways in which the IBD candidate genes are involved. Since our approach enables us to determine both IBD candidate genes and the cell type in which they are likely dysregulated, we analyzed the pathogenic processes that are possibly involved in monocytes, lymphocytes, and intestinal epithelial cells. Therefore, we performed separate pathway analyses on the datasets generated in these three different cell types. This revealed that the enriched pathways in the two immune cell types are mainly similar to each other, whereas the enrichment in epithelial cells shows different pathways (Fig. 5, Additional file 4: Table S5). Notably, IL10 signaling was found to be highly enriched in the intestinal epithelium dataset. This implies that the members of this pathway are possibly dysregulated in this cell type. As this pathway is also enriched in the immune cells (Additional file 4: Table S5), it is likely that the contribution of IL10 signaling to the IBD pathogenesis can be found in the interplay between the intestinal epithelium and immune cells. Furthermore, several JAK/STAT and Interferon signaling pathways were highly enriched in both monocytes and lymphocytes. JAK-STAT is a common signaling pathway used by many cytokines. Dysregulation of the JAK-STAT pathway can lead to a plethora of immune diseases [46].
For example, tissue specific disruption of STAT3 is known to cause an IBD-like phenotype in mice [46]. The high enrichment of many pathways that are relevant to IBD in the datasets of the separate cell types, supports the relevance of approaches that take cell type-specific role for candidate genes into account.

**Hepatocyte nuclear factor 4α (HNF4α)** is a potential key regulator of the IBD candidate genes

The 4C-seq approach reveals candidate genes based on their physical interaction with active regulatory regions. Transcription factors are important mediators in activating expression from active regulatory regions. Therefore, we aimed to determine which upstream regulators are involved in the regulation of transcriptional activity of the IBD candidate genes. We used an *in silico* analysis that determines which factors regulate expression from the candidate genes and which sets of genes that are regulated by a certain upstream regulator are enriched in our cell type-specific datasets. This analysis shows many significantly over-represented upstream regulators (Fig. 6a, Additional file 5: Table S6), including numerous transcription factors. Notably, HNF4α is highly enriched in all three cell types. HNF4α is a transcription factor that belongs to the nuclear hormone receptor superfamily [47]. Recently, the HNF4α-locus was associated to IBD

---

**Fig. 4** IL10RA is a novel IBD candidate gene. **a** The 4C signal from the rs630923 locus in lymphocytes; signal on the y-axis is depicted as the percentage of fragends covered per pixel. Black bars indicate significant 4C signal (p < 10⁻⁸); all coding genes located in this region are shown. The TSS of IL10RA co-localizes with a distant significant signal (~ 1 Mb from the viewpoint). Rs630923 and IL10RA localize to different haploblocks, meaning these regions do not co-segregate. **b** A network that consists of members of the IL10 signaling pathway. Red dots represent novel IBD candidate genes, orange dots represent candidate genes that were identified by 4C-seq as well as by GWAS, blue dots represent previously reported candidate genes that were not identified in the 4C-seq dataset, and black dots represent members of the IL10 pathway that have not been associated to IBD. Although many genes of the IL10 signaling pathway have been reported previously, we complement the network with three novel candidate genes including IL10RA.
through a GWAS [48]. Mouse studies revealed that during intestinal inflammation, HNF4α has a reduced ability to bind to active enhancers and that Hnf4α knock-out mice spontaneously develop colitis [49, 50].

Our study confirms that many genes that are likely dysregulated in IBD are regulated by HNF4α. Furthermore, HNF4α was found to be one of our candidate genes that was identified by a distal interaction with rs6017342 in intestinal epithelial cells (Additional file 1: Table S2). Upon exposure of intestinal organoids to bacteria lysate, we found that the epithelial response is characterized by a marked upregulation of both the NF-κB pathway and HNF4α (Fig. 6b). The kinetics of HNF4α expression upon epithelial responses and the enrichment of HNF4α-regulated genes among the IBD candidate genes propose HNF4α as a potential key regulator in IBD.

Discussion
This study shows that using chromatin interactions for GWAS interpretation reveals many novel and relevant candidate genes for IBD. Specifically, we have intersected data on chromatin interactions, mRNA expression, and H3K27Ac occupation data (marking active enhancer elements) to identify IBD candidate genes. By applying 4C-seq to cell types involved in IBD, we revealed 902 novel candidate genes, consisting of multiple noteworthy genes like SMAD5, IL10RA, and ATG9A. Notably, many novel genes were located outside the associated loci.

There are multiple ways that can be used to identify significant interactions in 4C-seq datasets and none of these methods offer the ideal solution for all interaction ranges (long, short, inter-chromosomal), resolutions, and dynamic ranges of signal [51, 52]. In this study, we have selected a method that, to our opinion, provides a good balance between the specificity and sensitivity for interactions spanning up to several megabases. In order to reduce the amount of false-positive findings, we chose to use a stringent cutoff ($p \leq 10^{-8}$).

The identification of functional DRE–gene interactions is further established through the overlap of the candidate gene sets identified in the different cell types. Intestinal epithelial cells are developmentally and functionally very distinct from cells with a shared hematopoietic

![Fig. 5 Pathways in IBD. This figure shows the pathways that are most highly enriched among the identified candidate genes in the three separate cell types. Bars correspond with the –Log of the p value, the dashed line indicates the threshold for significance. Numbers between the brackets show (amount of pathway members in dataset/total amount of pathway members). Pathway analyses were performed using Ingenuity Pathway Analysis (IPA, see "Methods"). All significantly enriched pathways can be found in Additional file 3: Table S4](image-url)
origin, in that context monocytes and lymphocytes are more alike. These differences in overlapping background are reflected by the sets of candidate genes identified in the different cell types. Specifically, lymphocytes and monocytes shared a large part of the candidate genes, whereas intestinal epithelial cells showed a more distinct set of genes (for example, monocytes share 42% and 8% of candidate genes with lymphocytes and DLD-1, respectively; Fig. 2a and Additional file 2: Figure S5).

Although this approach gives a general overview of the contribution of lymphocytes to the IBD pathogenesis, it does not enable to discriminate between mechanisms in lymphocyte subsets. Analyzing a pool of cell types also decreases the sensitivity of the detection of candidate genes that are specific to a subset of cells. Therefore, in future approaches, 4C datasets for specific lymphocyte subtypes can provide more insight into the contribution of each of these cell types to the IBD pathogenesis. Furthermore, since UC is limited to the colon and CD can occur throughout the intestine, creating a 4C dataset from epithelium derived from different parts the intestine (i.e. duodenum, jejunum, ileum, and colon) might help to discriminate between the UC and CD specific pathogenic processes.

We examined the presence of eQTLs among the IBD-associated SNPs and the 4C-seq candidate genes. These analyses confirm that our approach is capable to pick up every candidate gene that was found to have SNP-dependent expression levels in tissues relevant for IBD. As expected based on the two eQTL databases that were used, not all 4C-seq candidate genes we found to have a SNP-dependent expression pattern. This is (at least in part) due to the highly context-specific nature of SNP-dependent differential expression of many eQTLs [53]. While eQTLs are usually identified at one specific cell state [53], many SNP-dependent expression patterns are only present under specific conditions (i.e. developmental stages, presence of activating stimuli, etc.), resulting in a high false-negative rate of eQTL detection. For example, many 4C-seq candidate genes might be differentially expressed between genotypes in the presence of pro-inflammatory stimuli. Our findings both confirm that our assay enables to detect genes with a SNP-dependent expression profile and underlines the need of chromatin-based techniques to identify the genes that are missed by eQTL analyses.

By using GSEA we show that the 4C-seq candidate genes are highly enriched among genes that are upregulated in inflamed intestinal biopsies from IBD patients. Since the GSEA compares inflamed versus non-inflamed intestinal tissue within patients, we cannot determine what the baseline difference in expression is between
patients and healthy controls. Although the fact that a gene is upregulated upon inflammation does not show a causal relation between the (dys)regulation of that gene and the IBD phenotype, it shows the involvement of the novel 4C-seq candidate genes in IBD.

We have shown that pathway-enrichment and upstream regulator-enrichment algorithms can be used to interpret and prioritize this large candidate gene dataset. Interpretation of the 4C-seq data can be further optimized by using this data in a quantitative manner (i.e. correlating peak strength instead of using a cutoff value for peak calling). However, as with all approaches for candidate gene identification, further validation is needed to identify the causal genes for IBD. The first step towards this confirmation will in this case consist of revealing the dysregulation of the candidate gene expression upon alteration of the enhancer function in vivo.

We have profiled the chromatin interactions in primary cells from healthy controls and a cell line, to create a profile of the genes that physically interact with the IBD susceptibility loci under normal conditions in peripheral immune cells derived from healthy individuals and in an intestinal epithelium-derived cell line. As the effects of common variants in regulatory regions are relatively mild, it is improbable that a single common variant that is present in an IBD patient will ablate or create a whole regulatory region and its 3D interaction [39]. We therefore do not expect that the identification of candidate genes in cells derived from patients will reveal a substantial number of additional interactions. On the other hand, these variants are expected to cause dysregulation of the candidate genes and thereby contribute to the disease, possibly under very specific conditions, i.e. during certain stages of development or in presence of specific stimuli [16, 53].

Our study provides a proof of principle for the usage of chromatin–chromatin interactions for the identification of candidate genes. The approach presented here complements, but does not replace, previously reported approaches for candidate gene identification [11]. Candidate gene prioritization models for GWASs currently use multiple types of information, for example protein–protein interactions, expression patterns, and gene ontology. We propose that these algorithms should take chromatin interactions into account to optimize gene prioritization.

Conclusions
We have used 4C-seq to study chromatin interactions at loci that have been associated to IBD through GWASs using 4C-seq in cell types that are involved in the pathogenesis of IBD we identified 902 novel candidate genes, consisting of multiple noteworthy genes like SMADS, IL10RA, and ATG9A.

We conclude that 4C-seq and other 3C-derived methods can be applied to candidate gene identification in diseases with a complex genetic background and complement the classical candidate gene identification approaches.

Methods
Cell culture
DLD-1 cells were cultured in RPMI-1640 with 10% FCS and standard supplements. Cells were harvested for 4C template preparation by trypsinization at 60–80% confluence.

Monocyte and peripheral blood lymphocyte (PBL) isolation
Peripheral blood was collected from two healthy donors (one for monocyte isolation, one for PBL isolation) in sodium-heparin tubes. Peripheral blood mononuclear cells (PBMCs) were isolated by Ficoll-Paque gradient centrifugation. PMBCs were incubated with magnetic CD14+ microbeads (Milteny, order no. 130-050-201) according to the manufacturer’s manual. Thereafter cells were magnetically separated by the AutoMACS® Separator; the negative fraction consisted of PBLs, the positive fraction of monocytes.

Circular chromosome conformation capture: sequencing
Template preparation
For each cell type, one 4C-template was prepared. 4C-chromatin preparation, primer design, and library preparation were described previously [54]. 10 × 10^6 cells were used for chromatin preparation per cell type (monocytes, PBLs, and DLD-1). Primer sequences are listed in Additional file 6: Table S1.

Circular chromosome conformation capture: sequencing
Template preparation
For each cell type, one 4C-template was prepared. 4C-chromatin preparation, primer design, and library preparation were described previously [54]. 10 × 10^6 cells were used for chromatin preparation per cell type (monocytes, PBLs, and DLD-1). Primer sequences are listed in Additional file 6: Table S1.

Sequencing
Libraries were sequenced using the HiSeq2500 platform (Illumina), producing single end reads of 50 bp.

Data analysis
The raw sequencing reads were de-multiplexed based on viewpoint-specific primer sequences (the datasets are accessible through GEO Series accession number GSE89441). Reads were then trimmed to 16 bases and mapped to an in silico generated library of fragends (fragment ends) neighboring all DpnII sites in human genome (NCBI37/hg19), using the custom Perl scripts. No mismatches were allowed during the mapping and the reads mapping to only one possible fragend were used for further analysis. To create the 4C signal tracks in the UCSC browser, we have generated the .bed files with information for each mappable fragend on the coordinates and their covered/non-covered (1 or 0) status.
Visualization of the tracks in the UCSC browser was done with the following settings: windowing function: mean; smoothing window: 12 pixels.

Identification of the interacting genes
First, we calculated the number of covered fragends within a running window of \(k\) fragends throughout the whole chromosome where the viewpoint is located. This binary approach (i.e. a fragend is covered or is not covered in the dataset) was chosen to overcome the influence of polymerase chain reaction (PCR)-efficiency-based biases, however this approach decreases the dynamic range of the 4C-seq and may overestimate the strength of distal interactions compared to proximal interactions. The \(k\) was set separately for every viewpoint so it contains on average 20 covered fragends in the area around the viewpoint (+/- 100 kbp), e.g. when 100 out of 150 fragends around the viewpoint were covered the window size was set to 30 fragends. Next, we compared the number of covered fragends in each running window to the random distribution. The windows with a significantly higher number of covered fragends compared to random distribution (\(p < 10^{-8}\) based on binominal cumulative distribution function; \(R_{\text{pbinom}}\)) were considered as a significant 4C signal. The following criteria were defined for the identification of the candidate genes: (1) the transcriptional start site (TSS) co-localizes with a significant 4C-seq signal (\(p < 10^{-8}\)) within 5 kbp; (2) the susceptibility variant or other variant in linkage disequilibrium (LD) co-localizes with the H3K27ac signal (that marks activating regulatory elements) in the cell type from which the 4C signal was obtained (68 loci in monocytes, 73 in lymphocytes, and 52 in intestinal epithelial cells) [9]; and (3) the gene is expressed (log2(RPKM) > -0.5) in the assayed cell type (Additional file 1: Table S2). Datasets used for expression analysis are listed in Additional file 7: Table S3. Quality measures for the 4C library preparation and sequencing can be found in Additional file 2: Supplementary data, Figures S1–S3. The use of single 4C templates per cell type was validated in a biological duplicate of the lymphocyte 4C template that is derived from a different donor (Additional file 2: Figure S4A) and the reproducibility in other chromatin interaction datasets was established by intersecting our findings with two Hi-C datasets [25] (Additional file 2: Figure S4B and Additional file 7: Table S3).

TSS occupancy by H3K27ac and H3K4me3
The publicly available datasets of H3K27ac and H3K4me3 occupancy were accessed from the UCSC/ENCODE browser (http://genome.ucsc.edu/ENCODE/). Datasets are listed in Additional file 7: Table S3. The occupancy around 2 kbp +/- of TSS of was calculated using custom Perl scripts and Cisgenome [55] functions.

eQTL analyses

GTEx
A manual look-up was performed for expression quantitative trait loci (eQTL) in the Genotype-Tissue Expression (GTEx) database (accession dates; eQTL-genes: 05-2016; \(p\) values: 09-2016). The presence of eQTL genes for each of the 92 IBD-associated SNPs was performed in four different tissues: colon-transverse; colon-sigmoid; small intestine-terminal ileum; and whole blood [38]. Next, for each gene for which an IBD-associated SNP turned out to be an eQTL, its presence among the 4C-seq identified genes was evaluated (Additional file 3: Table S4). All transcripts in the GTEx database that were not included in the gene annotation (UCSC genes 2009) that was used for the analysis of the 4C-seq data were removed from the analysis.

STAGE

eQTLs were analyzed using the Stockholm Atherosclerosis Gene Expression (STAGE) [39] dataset (Additional file 2: Supplementary data, Methods). Identified loci from GWAS for IBD were matched with imputed and genotyped SNPs and were selected for eQTL discovery. We compared the amount of eQTLs present in “SNP-candidate gene”-pairs and “SNP-control gene”-pairs. Control genes are genes within the same locus that are not interacting with the IBD-associated locus. An empirical false discovery rate was estimated for each eQTL gene by shuffling patient IDs 1000 times on genotype data as described previously [56].

Gene set enrichment analysis (GSEA)
GSEA [40] was performed using gene expression datasets [41] from intestinal biopsies obtained from ulcerative colitis patients (datasets available at GSE11223). The “normal uninflamed sigmoid colon” and “UC inflamed sigmoid colon” were used and the fold changes in expression were calculated using the GEO2R tool [57] with default settings. Significance of the enrichment was calculated based on 1000 cycles of permutations.

Signaling pathway analysis
The IL10 signaling pathway components were retrieved from Ingenuity Pathway Analysis (IPA®, QIAGEN Redwood City). Genes upregulated upon IL10 signaling (target genes) and genes involved in the bilirubin cascade were removed before further analysis. The interactions between the members of the IL-10 signaling pathway were visualized using the GeneMania tool <http://www.genemania.org/>.

The general pathway analysis was performed with the Ingenuity Pathway Analysis software (IPA®, QIAGEN Redwood City), based on the candidate genes from the three cell types, separately.
Upstream regulators
Upstream regulators that are enriched in the candidate genes in our datasets were identified with the Ingenuity Pathway Analysis software (IPA®, QIAGEN Redwood City), based on the candidate genes from the three cell types separately. The Ingenuity’s Upstream Regulator Analysis algorithm predicts upstream regulators from gene datasets based on the literature and compiled in the Ingenuity knowledge base.

CTCF tracks
CTCF tracks were accessed from the UCSC/ENCODE browser (http://genome.ucsc.edu/ENCODE/). Datasets are listed in Additional file 7: Table S3.

Tracks used for rs630923 and rs2382817
All tracks were accessed from the UCSC/ENCODE browser (http://genome.ucsc.edu/ENCODE/). Datasets are listed in Additional file 7: Table S3. Haploblock structures were visualized with Haplovie [58]; pairwise LD statistics of variants with a distance up to 500 kbp were used in the analyses (Fig. 4, Additional file 2: Supplementary data, Figure S9).

Organoid culture
Colon biopsies were obtained by colonoscopy. The biopsies were macroscopically and pathologically normal. Crypt isolation and culture of human intestinal cells from biopsies have been described previously [59, 60]. In summary, human organoids were cultured in expansion medium (EM) containing RSPO1, noggin, EGF, A83-01, nicotinamide, SB202190, and WNT3A. The medium (EM) containing RSPO1, noggin, EGF, A83-01, nicotinamide, SB202190, and WNT3A. The medium was changed every 2–3 days and organoids were passaged 1:4 every 9 days.

Five to seven days after passaging, the organoids were exposed to 10 μL sterilized E. Coli-lysate (control organoids were not stimulated). After 6 h of exposure, the organoids were not stimulated). After 6 h of exposure, the organoids were not stimulated). After 6 h of exposure, the organoids were not stimulated). After 6 h of exposure, the organoids were not stimulated. The medium was changed and organoids were harvested and RNA was extracted using RNA was extracted using RNA was extracted using RNA was extracted using ATG8A: autophagy related 9A; BP: base pairs; CD: complement-decay accelerating factor; CTCF: CCCTC-binding factor; DAF: decay accelerating factor; DLD-1 cells: D.L. Dexter-1 cells; DRE: DNA regulatory element; E. Coli: Escherichia Coli; ETL: expression quantitative trait loci; FCS: fetal calf serum; GWAS: genome-wide association study; H3K27Ac: acetylation of histone H3 at lysine 27; H3K4me3: trimethylation of histone H3 at lysine 4; HNF4α: hepatocyte nuclear factor 4 alpha; IKKβ: inhibitor of nuclear factor kappa-B kinase subunit epsilon; IL10: Interleukin 10; IL10RA: Interleukin 10 receptor subunit alpha; IL10RB: Interleukin 10 receptor subunit beta; JAK: Janus kinase; Kbp: kilo base pairs; LD: linkage disequilibrium; LMPCs: lamina propria mononuclear cells; MAP3K7: mitogen-activated protein kinase kinase kinase 7; MBp: mega base pairs; MCP: membrane co-factor protein; NFκB: nuclear factor kappa B; PBL: peripheral blood lymphocytes; PBMC: peripheral blood mononuclear cells; PCR: polymerase chain reaction; PIAS1: protein inhibitor of activated STAT 1; PRPM: reads per kilobase of exon per million reads mapped; RPMI medium: Roswell Park Memorial Institute medium; SMAD: named after their homologous genes Mothers Against Decapentaplegic (MAD) and the Small Body Size protein (SMA) in Drosophila and C. Elegans, respectively; SNP: single nucleotide polymorphism; STAT: signal transducer and activator of transcription; TANK: TRAF family member-associated NFκB activator; TGF-β: transforming growth factor beta-1; Th17 cells: T-helper 17 cells; Th2 cells: T-helper 2 cells; TNF: tumor necrosis factor; TSS: transcriptional start site; UCSC: University of California, Santa Cruz

Acknowledgements
We thank Professor J. Cho and Dr. S. Widdendorp for critically reading the manuscript and S. Cardoso for helping with the isolation of the immune cells.

Funding
Claartje A. Meddens is supported by the Alexandre Suerman Stipendium (UWvC, Utrecht). Magdalena Harakalova is supported by NIH Ro1 grant LM010098. Folkert W. Asselbergs is supported by a DeKker scholarship—Junior Staff Member 2014 T001—Netherlands Heart Foundation and UCL Hospitals NIHR Biomedical Research Centre. Michal Moky is supported by OGT/2012 WKZ fund and by the Broad Medical Research Program at CCFA ID.368408.

Availability of data and material
The 4C-seq data discussed in this publication have been deposited in NCBI’s Gene Expression Omnibus and are accessible through GEO Series accession number GSE89441. All other publicly available datasets are listed in Additional file 4: Table S3.

Authors’ contributions
CAM: Study concept and design; acquisition of data; analysis and interpretation of data; drafting of the manuscript. Mh: Drafting of the manuscript; acquisition of data, critical revision of the manuscript for important intellectual content. NAMD: Acquisition of data. HH: Acquisition of data. HFA: Data analysis. EPJGC: Critical revision of the manuscript for important intellectual content; study supervision. MM: Study concept and design; analysis and interpretation of data; drafting of the manuscript; study supervision. All authors read and approved the final manuscript.

Competing interests
The authors declare that they have no competing interests.

Consent for publication
Not applicable.

Additional files

| Additional file 1: Table S2. 4C datasets. (XLSX 43677 kb) |
| Additional file 2: Supplementary data; Figures S1–S9 and Supplementary Methods. (PDF 2295 kb) |
| Additional file 3: Table S4. eQTL analyses. (XLSX 16 kb) |
| Additional file 4: Table S5. Pathway analyses. (XLSX 65 kb) |
| Additional file 5: Table S6. Upstream regulators. (XLSX 135 kb) |
| Additional file 6: Table S1. Primer sequences. (XLSX 17 kb) |
| Additional file 7: Table S3. Publicly available datasets. (XLSX 11 kb) |
Ethics approval and consent to participate

Biopsies were obtained by ileo-colonoscopies that were performed as part of standard diagnostic procedures. Human Material Approval for this study was obtained by the Ethics Committee (Medisch Ethische Toetsings Commissie, METC) of the University Medical Center Utrecht (www.umcutrecht.nl/METC), protocol number 10/402. Blood was obtained from the Mini Donor Dienst (MDD) at the UMCU. The MDD is approved by the Ethics Committee of the UMCU, protocol number 07/125. All patients have given informed consent to participate in this study. All experimental procedures in this study comply with the Declaration of Helsinki.

Author details
1Department of Pediatric Gastroenterology, Wilhelmina Children’s Hospital, University Medical Center Utrecht, Huispostnummer KA.03.019.0, Lundlaan 6, 5557 TB, Utrecht, The Netherlands. 2Department of Cardiology, Division Heart and Lungs, University Medical Center Utrecht, Utrecht, The Netherlands. 3Vascular Biology Unit, Department of Medical Biochemistry and Biophysics, Karolinska Institutet, Stockholm, Sweden. 4Department of Biomedical Genetics, University Medical Center Utrecht, Utrecht, The Netherlands. 5Hubrecht Institute, Utrecht, The Netherlands. 6Department of Genetics and Genomic Sciences, Icahn Institute for Genomics and Multiscale Biology, Icahn School of Medicine at Mount Sinai, New York, NY, USA. 7Department of Physiology, Institute of Biomedicine and Translational Medicine, University of Tartu, Tartu, Estonia. 8Dunner Center for Cardiogenetic Research, Utrecht, The Netherlands. 9Institute of Cardiovascular Science, University College London, London, UK.

Received: 8 June 2016 Accepted: 7 November 2016
Published online: 30 November 2016

References
1. Kaser A, Ziegis S, Blumberg RS. Inflammatory bowel disease. Annu Rev Immunol. 2010;28:573–621.
2. Franke A, McGovern DP, Barrett JC, Wang K, Radford-Smith GL, Ahmad T, et al. Genome-wide meta-analysis increases to 71 the number of confirmed Crohn’s disease susceptibility loci. Nat Genet. 2010;42:1118–25.
3. Justins L, Ripke S, Weirnra RK, Duerr RH, McGovern DP, Hui KY, et al. Host-microbe interactions have shaped the genetic architecture of inflammatory bowel disease. Nature. 2012;491:19–24.
4. Anderson CA, Boucher G, Lees CW, Franke A, D’Amato M, Taylor KD, et al. Meta-analysis identifies 29 additional ulcerative colitis risk loci, increasing the number of confirmed associations to 47. Nat Genet. 2011;43:246–52.
5. Liu JZ, van Someren S, Huang H, Ng SC, Alberts R, Takahashi A, et al. Association analyses identify 38 susceptibility loci for inflammatory bowel disease and highlight shared genetic risk across populations. Nat Genet. 2011;43:979–90.
6. Roux JD, Xavier JR, Taylor KD, Silverberg MS, Goyette P, Huetta A, et al. Genome-wide association study identifies new susceptibility loci for Crohn disease and implicates autophagy in disease pathogenesis. Nat Genet. 2007;39:596–604.
7. Libourel C, Louis E, Hansoul S, Sandor C, Farnir F, Franchimont D, et al. Novel Crohn disease locus identified by genome-wide association maps to a gene desert on 3p13.1 and modulates expression of PTGER4. PLoS Genet. 2007;3:e68.
8. Duerr RH, Taylor KD, Brant SR, Roux JD, Silverberg MS, Daly MJ, et al. A genome-wide association study identifies IL23R as an inflammatory bowel disease gene. Science. 2006;314:461–3.
9. Mokry M, Middendorp S, Wiegersch CL, Witte M, Teunissen H, Meddens CA, et al. Many inflammatory bowel disease risk loci include regions that regulate gene expression in immune cells and the intestinal epithelium. Gastroenterology. 2014;146:1040–7.
10. Raychaudhuri S, Plenge RM, Rossin EJ, Ng AC, International Schizophrenia Consortium, Purcell SM, et al. Identifying relationships among genomic disease regions: predicting genes at pathogenic SNP associations and rare deletions. PLoS Genet. 2009;5:e1000534.
11. Wang K, Li M, Bucan M. Pathway-based approaches for analysis of genomewide association studies. Am J Hum Genet. 2007;81:1278–83.
12. Rivas MA, Beaudoin M, Gardet A, Stevens C, Sharma Y, Zhang CK, et al. Deep resequencing of GWAS loci identifies independent rare variants associated with inflammatory bowel disease. Nat Genet. 2011;43:1066–73.
13. Shlyueva D, Stampfel G, Stark A. Transcriptional enhancers: from properties to genome-wide predictions. Nat Rev Genet. 2014;15:272–86.
14. Mauwanto MT, Humbert R, Ryens E, Thurman RE, Haegnen E, Wang H, et al. Systematic localization of common disease-associate variation in regulatory DNA. Science. 2012;337:6999:1190–5.
15. Kleinjan DJ, Coutinho P. Cis-regulation mechanisms: Disruption of cis-regulatory control as a cause of human genetic disease. Brief Funct Genomic Proteomic. 2009;8:317–32.
16. Schaub MA, Boyle AP, Kundaje A, Fraser KA. Linking disease associations with regulatory information in the human genome Toward mapping the biology of the genome. Genome Res. 2012;22(9):1748–59.
17. McVicker G, van de Geijn B, Degner JF, Cain CE, Banovich NE, Raj A, et al. Identification of genetic variants that affect histone modifications in human cells. Science. 2013;342:747–9.
18. Kilpinen H, Wazsam SK, Geschwind AR, Raghav SK, Witwicki RM, Orrioli A, et al. Coordinated effects of sequence variation on DNA binding, chromatin structure, and transcription. Science. 2013;342:744–7.
19. Heintzman ND, Hon GC, Hawkins RD, Kehadpour P, Stark A, Harp LF, et al. Histone modifications at human enhancers reflect global cell-type-specific gene expression. Nature. 2009;459:108–12.
20. de Laat W, Klous P, Kooon J, Noordermeer D, Palstra RJ, Simonis M, et al. Three-dimensional organization of gene expression in erythroid cells. Curr Top Dev Biol. 2008;82:117–39.
21. Hughes JR, Roberts N, McGowan S, Hay D, Giannoulou E, Lynch M, et al. Analysis of hundreds of cis-regulatory landscapes at high resolution in a single, high-throughput experiment. Nat Genet. 2011;43:620–5.
22. Thurman RE, Ryens E, Humbert R, Vierstra J, Mauwanto MT, Haegnen E, et al. The accessible chromatin landscape of the human genome. Nature. 2012;489:75–82.
23. Nica AC, Dermitzakis ET. Expression quantitative trait loci: present and future. Philos Trans R Soc Lond Ser B Biol Sci. 2013;368:20120362.
24. Wright JB, Brown SJ, Cole MD. Upregulation of c-Myc in cis through a large chromatin loop linked to a cancer risk-associated single-nucleotide polymorphism in colorectal cancer cells. Mol Cell Biol. 2010;30(6):1411–20.
25. Mifsud B, Tavares-Cadete F, Young AN, Sugar R, Schoenfelder S, Ferreira L, et al. Sup Mapping long-range promoter contacts in human cells with high-resolution capture Hi-C. Nat Genet. 2015;47:598–606.
26. Jäger R, Migliorini G, Hensson M, Kandaskwamy R, Speedy HE, Heindl A, et al. Capture Hi-C identifies the chromatin interactome of colorectal cancer risk loci. Nat Commun. 2015;6:6178.
27. Martin P, McGovern A, Ozhazo G, Dufaux K, Yanwood A, Schoenfelder S, et al. Capture Hi-C reveals novel candidate genes and complex long-range interactions with related autoimmune risk loci. Nat Commun. 2015;6:10069.
28. Sadek FM, Zhong W, Lai E, Darrelle J. Liver-enzriched transcription factor HNF-4 is a novel member of the steroid hormone receptor superfamily. Genes Dev. 1990;4:2353–65.
29. UK IBD Genetics Consortium, Barrett JC, Lee JC, Lees CW, Prescott NJ, Anderson CA, et al. Genome-wide association study of ulcerative colitis identifies three new susceptibility loci, including the HNF4A region. Nat Genet. 2009;41:1330–4.
30. Dekkers JF, Wiegerinck CL, de Jonge HR, Bronsveld I, Janssens HM, de Winter de Groot KM, et al. A functional CFTR assay using primary cysctic fibrosis intestinal organoids. Nat Med. 2013;19:939–45.
31. Vemot B, Stergachis AB, Mauwanto MT, Viresta J, Neg D, Thurman RE, et al. Personal and population genomics of human regulatory variation. Genome Res. 2012;22:1689–97.
32. Chahar S, Gandhi V, Yu S, Desai K, Cowper-Sal-lari R, Kim Y, et al. Chromatin profiling reveals regulatory network shifts and a protective role for hepatitis c virus factor 4a during colitis. Mol Cell Biol. 2014;34:3291–304.
33. Saith T, Akira S. Regulation of innate immune responses by autophagy-related proteins. J Cell Biol. 2010;189:925–35.
34. Murphy TL, Tussiwand R, Murphy KM. Specificity through cooperation: BATTLE-IRF interactions control immune-regulatory networks. Nat Rev Immunol. 2013;13:499–509.
35. Darsigny M, Babeu JP, Dupuis AA, Furth EE, Seidman EG, Levy E, et al. Loss of hepatitis c virus factor 4a affects colonic ion transport and causes chronic inflammation resembling inflammatory bowel disease in mice. PLoS One. 2009;4:e7609.
36. de Wit E, Vos ES, Holwerda SJ, Valdes-Quezada C, Verstegen MJ, Teunissen H, et al. CTGF binding polarity determines chromatin looping. Mol Cell. 2015;60:676–84.
37. Raviram R, Rocha PP, Muller CL, Miraldi ER, Badri S, Fu Y, et al. 4C-ker: a method to reproducibly identify genome-wide interactions captured by 4C-Seq experiments. PLoS Comput Biol. 2016;12:e1004780.

38. Fairfax BP, Humborg P, Makino S, Naranbhai V, Wong D, Lau E, et al. In innate immune activity conditions the effect of regulatory variants upon monocyte gene expression. Science. 2014;343:1246949.

39. van de Werken HUG, de Vree PJ, Splinter JE, Holwerda SJ, Klbsus P, de Wit E, et al. 4C technology: protocols and data analysis. Methods Enzymol. 2012;51:89–112.

40. Ji H, Jiang H, Ma W, Johnson DS, Myers RM, Wong WH. An integrated software system for analyzing ChIP-chip and ChIP-seq data. Nat Biotechnol. 2008;26:1293–300.

41. GTEx Consortium. The Genotype-Tissue Expression (GTEx) project. Nat Genet. 2013;45:580–5.

42. Hägg S, Skogsberg J, Lundstrom J, Noori P, Nilsson R, Zhong H, et al. Multi-organ expression profiling uncovers a gene module in coronary artery disease involving transendothelial migration of leukocytes and LIM domain binding 2: the Stockholm Atherosclerosis Gene Expression (STAGE) study. PLoS Genet. 2009;5:e1000754.

43. Foroughi Asl H, Talukdar HA, Kindt AS, Jain RK, Ermel R, Ruusalepp A, et al. Expression quantitative trait loci acting across multiple tissues are enriched in inherited risk for coronary artery disease. Circ Cardiovasc Genet. 2015;8:305–15.

44. Subramanian A, Tamayo P, Mootha VK, Mukherjee S, Elbert BL, Gillette MA, et al. Gene set enrichment analysis: a knowledge-based approach for interpreting genome-wide expression profiles. Proc Natl Acad Sci U S A. 2005;102:15545–50.

45. Noble CL, Abbas AR, Cornelius J, Lees CW, Ho GT, Toy K, et al. Regional variation in gene expression in the healthy colon is dysregulated in ulcerative colitis. Gut. 2008;57:1398–405.

46. NCBI GEO2R. http://www.ncbi.nlm.nih.gov/geo/geo2r/.

47. Barrett JC, Fry B, Maller J, Daly MJ. Haploview: analysis and visualization of LD and haplotype maps. Bioinformatics. 2005;21:263–5.

48. Sato T, Stange DE, Ferrante M, Vries RG, Van Es JH, Van den Brink S, et al. Long-term expansion of epithelial organoids from human colon, adenoma, adenocarcinoma, and Barrett’s epithelium. Gastroenterology. 2011;141:1762–72.

49. Lin F, Spencer D, Hatala DA, Levine AD, Medof ME. Decay-accelerating factor deficiency increases susceptibility to dextran sulfate sodium-induced colitis: role for complement in inflammatory bowel disease. J Immunol. 2004;172:3836–41.

50. Glocker EO, Kolharz D, Boztug K, Gertz EM, Schaffer AA, Noyan F, et al. Inflammatory bowel disease and mutations affecting the interleukin-10 receptor. N Engl J Med. 2009;361:2033–45.

51. Liu B, Tahk S, Yee KM, Fan G, Shuai K. The ligase PIAS1 restricts natural regulatory T cell differentiation by epigenetic repression. Science. 2010;330:521–5.

52. Aliaea JM, Darisigny M, Marcoux SS, Roy SA, Schmouei JF, Umans L, et al. Loss of Smad5 leads to the disassembly of the apical junctional complex and increased susceptibility to experimental colitis. Am J Physiol Gastrointest Liver Physiol. 2011;300:G586–97.

53. Portillo JC, Greene A, Schwartz I, Subauste MC, Subauste CS. Blockade of CD40-TRAF2,3 or CD40-TRAF6 is sufficient to inhibit pro-inflammatory responses in non-haematopoietic cells. Immunology. 2015;144(1):21–33.

54. Raviram R, Rocha PP, Muller CL, Miraldi ER, Badri S, Fu Y, et al. Sequence variants in the autophagy gene IRGM and multiple other replicating loci contribute to Crohn’s disease susceptibility. Nat Genet. 2007;39:830–6.

55. Bell AC, West AG, Felsenfeld G. The protein CTCF is required for the enhancer blocking activity of vertebrate insulators. Cell. 1999;98:387–96.

56. Saitoh T, Fujita N, Hayashi T, Takahara K, Satoh T, Lee H, et al. Atg9a controls dsDNA-driven dynamic translocation of STING and the innate immune response. Proc Natl Acad Sci U S A. 2009;106:20842–6.

57. Hampe J, Franke A, Rosenstiel P, Till A, Teuber M, Huse K, et al. A genome-wide association scan of nonsynonymous SNPs identifies a susceptibility variant for Crohn disease in ATG16L1. Nat Genet. 2007;39:207–11.

58. Parkes M, Barrett JC, Prescott N, Tremling M, Anderson CA, Fisher SA, et al. Sequence variants in the autophagy gene RGM and multiple other replicating loci contribute to Crohn’s disease susceptibility. Nat Genet. 2007;39:830–6.

59. Cadwell K, Liu JY, Brown SL, Miyoshi H, Loh J, Lennern JK, et al. A key role for autophagy and the autophagy gene Atg16L1 in mouse and human intestinal Paneth cells. Nature. 2008;456:259–63.

60. Shuai K, Liu B. Regulation of JAK-STAT signalling in the immune system. Nat Rev Immunol. 2003;3:900–11.