ORIGINAL ARTICLE

Longitudinal study of murine microbiota activity and interactions with the host during acute inflammation and recovery

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Although alterations in gut microbiota composition during acute colitis have been repeatedly observed, associated functional changes and the recovery from dysbiosis received little attention. In this study, we investigated structure and function of the gut microbiota during acute inflammation and recovery in a dextran sodium sulfate (DSS)-colitis mouse model using metatranscriptomics, bacterial 16S rRNA gene amplicon sequencing and monitoring of selected host markers. Parallel to an increase of host markers of inflammation during acute colitis, we observed relative abundance shifts and alterations in phylotype composition of the dominant bacterial orders *Clostridiales* and *Bacteroidales*, and an increase of the low abundant *Enterobacteriales*, *Deferribacterales*, *Verrucomicrobiales* and *Erysipelotrichales*. During recovery, the microbiota began to resume, but did not reach its original composition until the end of the experiment. Microbial gene expression was more resilient to disturbance, with pre-perturbation-type transcript profiles appearing quickly after acute colitis. The decrease of *Clostridiales* during inflammation correlated with a reduction of transcripts related to butyrate formation, suggesting a disturbance in host-microbe signalling and mucosal nutrient provision. The impact of acute inflammation on the *Clostridiales* was also characterized by a significant downregulation of their flagellin-encoding genes. In contrast, the abundance of members of the *Bacteroidales* increased along with an increase in transcripts related to mucin degradation. We propose that acute inflammation triggered a selective reaction of the immune system against flagella of commensals and temporarily altered murine microbiota composition and functions relevant for the host. Despite changes in specific interactions, the host–microbiota homeostasis revealed a remarkable ability for recovery.

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Introduction

The mammalian intestine is densely populated by billions of bacteria belonging to an estimated 500–1000 different species predominantly classified to the phyla *Firmicutes* and *Bacteroidetes*. Under conditions of health, host and microbiota live in a mutualistic relationship. The host supplies energy sources, and the key roles of the microbiota are the breakdown of nutrients, the synthesis of hormones and vitamins and colonization resistance to pathogens (Nicholson et al., 2012; Stecher et al., 2013). Short chain fatty acids (SCFAs such as butyrate, propionate and acetate) synthesized during metabolism of diet- or host-derived carbohydrates are actively absorbed and contribute to host energy maintenance (Bergman, 1990). Host–microbiota interactions are also important at the immunological level (Saleh and Trinchieri, 2011). Members of the
microbiota or their metabolites, for example, *Clostridium* clusters IV and XIV can induce regulatory T-cells, thereby promoting an anti-inflammatory immune response (Atarashi et al., 2011). Butyrate, which is predominantly produced by *Clostridium* clusters IV and XIV, is a main energy source of colonocytes, impacts cell proliferation and differentiation and lowers the risk of colitis and colorectal cancer (Wang et al., 2006; Ploeger et al., 2012).

The human intestine has the highest immune activity in the body (Vickery et al., 2011) maintaining homeostasis with commensals but also responding to the invasion of pathogens. In the colon, this homeostasis is guaranteed by the combined action of a mucin layer, which acts as a first physical line of defense, and a tightly regulated network of innate and adaptive immunity. Sensing of microbe-associated molecular patterns activates host-signalling pathways via pattern recognition receptors, including various Toll-like receptors (TLRs). TLR signals activate the synthesis of antimicrobial peptides, maintain tight junctions and trigger cytokine production in intestinal epithelial cells (Kumar et al., 2009).

Disturbances of the intestinal homeostasis caused by external factors (e.g., dietary shifts and antibiotics treatment) can lead to an imbalance in microbiota composition referred to as dysbiosis (Honda and Littman, 2012). Dysbiosis is also associated with inflammatory bowel disease (IBD), a chronic inflammation that is characterized by phases of severe (flare-ups) or no symptoms (remission) (Berry and Littman, 2012). Dysbiosis is also associated with inflammatory bowel disease (IBD), a chronic inflammation that is characterized by phases of severe (flare-ups) or no symptoms (remission) (Berry and Littman, 2012). Although the exact etiology of IBD remains unclear, several studies suggest that genetic predisposition, environmental factors, an overreacting immune system and the intestinal microbiota contribute to the development of chronic inflammation.

Microbiota alterations frequently observed in chronic colitis are an increase of *Enterobacteriaceae*, shifts in abundance of *Bacteroidetes* and *Firmicutes*, and within-group alterations that can extend to the family level or below (Frank et al., 2007; Berry et al., 2012). Recently, studies applying metagenomics or metaproteomics also investigated functional variations and concordantly reported alterations in metabolic pathways that might affect host–microbiota interactions (Erickson et al., 2012; Morgan et al., 2012). However, investigations on human subjects are mostly limited to the acute phase of inflammation.

With this study we aimed to obtain a more holistic picture of the functional links between host and microbiota during the time course of colitis development and recovery and to investigate microbial functions affected by the inflammation event. We used dextran sodium sulfate (DSS) to induce a reversible form of colitis (Okayasu et al., 1990). Metatranscriptomics was applied to concurrently investigate the active microbial community and its functional properties, and to link functions to bacterial taxa (Berry et al., 2012; Helbling et al., 2012). To investigate abundance shifts of species-level phylotypes, 16S rRNA gene amplicon sequencing was used. Pro- and anti-inflammatory cytokines were monitored as an indicator of the state of the immune system. Finally, correlation networks were generated applying metatranscriptome, amplicon sequence and host inflammation marker data in order to identify associations between host and microbiota.

**Materials and methods**

**Animal experiments**

Wild-type C57BL/6 mice originating from three litters and 6–8 weeks of age were provided with 2% DSS (molecular weight: 36–50 kDa, MP Biomedicals, Eschwege, Germany) in autoclaved drinking water *ad libitum* for 5 days, after which they received DSS-free drinking water. The DSS dose chosen caused a moderate to severe colitis. Three mice, which received water throughout the time course of the experiment, were used as controls for weight determination. Each sampling day, representatives from at least two litters were killed. The cecum and colon were removed and flushed with phosphate-buffered saline to collect luminal contents (Berry et al., 2012). Flushed lumen contents were homogenized, collected by centrifugation and either snap-frozen for nucleic acid isolation or fixed in 2% paraformaldehyde overnight at 4 °C and then stored in 70% ethanol/30% phosphate-buffered saline at −20 °C until analysis. The intestine was fixed in 2% paraformaldehyde and prepared for fluorescence *in situ* hybridization (FISH), immunohistochemistry and pathology evaluation using established methods (Horino et al., 2008). Hematoxylin and eosin-stained intestinal colon tissue (photomicrographs taken at 100 × magnification) was scored in blind evaluation. For the inflammation score, three subscores (inflammation severity, inflammation extent and crypt damage) ranging from 0–4 were given and added together which could give a maximum possible score of 12. Animal experiments were approved by the institutional ethics committee and conducted in accordance with protocols approved by the Austrian laws (BMWF-66.006/0002-II/10b/2010).

**DNA and RNA purification and preparation of cDNA libraries**

Nucleic acid from lumen contents were extracted with a phenol-chloroform bead-beating procedure (Griffiths et al., 2000) and separated into DNA and RNA using the AllPrep DNA/RNA Mini kit (Qiagen, Hilden, Germany), as described before (Berry et al., 2012). DNA and RNA purified from the same extractions were used for 16S rRNA gene amplicon sequencing and cDNA generation, respectively. For double-stranded cDNA synthesis, total RNA was reverse-transcribed using the SuperScript Double-Stranded cDNA Synthesis Kit (Invitrogen, Lofer,
Austria) with modifications as described previously (Berry et al., 2012). Single-stranded cDNA was produced from total RNA using SuperScript III reverse transcriptase according to instructions of the supplier (Invitrogen).

RNA isolation and cDNA synthesis from colon tissue
For RNA preparation from colon tissue, pieces were homogenized in 700 µl RA1 buffer of the NucleoSpin II RNA isolation kit (Macherey–Nagel, Graz, Austria) and processed according to manufacturer’s instructions. cDNA was prepared as described previously (Stockinger et al., 2004).

Metatranscriptome sequencing and data analysis
Double-stranded cDNA libraries were paired-end sequenced using an Illumina HiSeq (Campus Science Support Facilities GmbH, Vienna, Austria). Read pairs were overlapped using FLASH (Magócs and Salzberg, 2011), which yielded between 4.1 and 15.6 Mio reads of ~170 bp. Metatranscriptomic sequencing data were analyzed following an established double RNA analysis pipeline (Urich et al., 2008). More than 90% of reads were derived from rRNA; the number of putative mRNA reads per sample ranged from 75 305 to 1 166 930 (Supplementary Table S1). Community composition was determined from 100 000 randomly chosen rRNA reads per sample, which were taxonomically assigned using CREST (Lanzén et al., 2012) (bit score = 150, top percent = 10, minimal support = 5). Putative mRNA tags were compared against the NCBI RefSeq database using BlastX and functionally and taxonomically classified using MEGAN and the SEED functional classification scheme therein (bit score = 40, top percent = 10, minimal support = 1) (Mitra et al., 2011). This resulted in 24 190 to 368 950 functionally annotated mRNAs. Metatranscriptomes were generated from three replicates collected on days 1, 5 and 14 and from four replicates collected on days 8 and 25 (Supplementary Table S2). PAST (Hammer et al., 2001) was used for multivariate analysis of metatranscriptome data. Principal component analysis (PCA) was done by eigenvalue decomposition of a data correlation matrix. Unpaired t-test and Pearson Moment analysis in SigmaPlot 11 (Systat) were applied for statistical analysis of variance and correlations, respectively.

16S rRNA gene amplicon pyrosequencing
Amplicon libraries were obtained using barcoded pyrosequencing primers 909F and 1492R, pooled, purified and quantified as described before (Berry et al., 2012). Pyrosequencing was performed with Titanium reagents on a 454 genome sequencer FLX (Roche, Vienna, Austria). All 20 samples collected on 5 sampling days were subjected to 16S rRNA amplicon sequencing, but one sample from day 1 was excluded from analysis due to low sequencing depth. Reads were quality-filtered using the amplification pipeline of the GS Run Processor (Roche) and the PyroNoise algorithm in mothur (Schloss et al., 2011). Operational taxonomic units were generated at 97% sequence identity using UCLUST, as described (Berry et al., 2012). Amplicon sequencing libraries, which had a mean number of 17 999 reads (range [4415–26 495]) of an average length of 269 nt were re-sampled at 3500 reads for α- and β-diversity analyses using QIIME (Caporaso et al., 2010).

FISH analysis
FISH was performed on parafomaldehyde-fixed biomass, as previously described (Berry et al., 2012). Probes specific for Bacteroides acidificiens (OTU 1812) and Bacteroidales OTU 183 were designed based on near full-length 16S rRNA genes recovered from screening of clone libraries established with primer pair 8F and 1492R (Lane, 1991). Probe specificity was evaluated by checking the probes against the SILVA SSU NR108 database (Quast et al., 2013) and stringency of hybridization conditions was optimized using either clone-FISH (Schramm et al., 2002) or environmental samples containing abundant target populations. For Bacteroides acidificiens (OTU 1812) probes Bacid87 I (S-S-Bac-0087-a-A-24, 5’-GGCGGCGGTCGGCCATCAAA AAGTTTG-3’) and Bacid87 II (S-S-Bac-0087-a-A-22, 5’-GCGGCGGTCGGCCATGGAAGTTTG-3’) were used with 35% formamide. For Bacteroidales OTU 183, probe Bada183_437 (S-S-Bda-0437-a-A-22, 5’-CGGCGCCCTTTGCTCCCTGACAAAA-3’) was used with 30% formamide. Probes were deposited at probeBase (Loy et al., 2003).

Quantification of selected genes and transcripts by qPCR
Copies of 16S rRNA genes from selected taxa and selected microbial transcripts were quantified by qPCR using a Mastercycler ep realplex (Eppendorf, Vienna, Austria). Reaction mixtures (20 µl) contained 10 µl QuanttiFast SybrGreen (Qiagen), 1 µl of each of the specific primers (final concentration 0.25 µM, Supplementary Table S2) and 1 µl of template. Running conditions were 95 °C for 5 min, followed by 40 cycles of 95 °C for 10 s, annealing for 15 s with the temperature specified in Supplementary Table S2, and 72 °C for 30 s followed by melting curve analysis. Standard curves were generated as described before (Berry et al., 2012). Gene and transcript copy numbers were calculated per µg DNA or RNA, respectively. Flagellin expression levels were calculated as the ratio of transcript copies and gene copies per µg nucleic acid preparation. Unpaired t-test and one-way ANOVA combined with an all pairwise multiple comparison procedure (Student–Newman–Keuls Method) were applied for statistical analysis of variance (SigmaPlot 11).
For mouse gene expression analysis, qPCRs were run with 60°C annealing temperature on an Eppendorf realplex cycler in a final volume of 15 μl containing 1.5 μl cDNA and 0.3 μM primer using GoTaq qPCR Master Mix (Promega, Mannheim, Germany). Primers used are summarized in Supplementary Table S3. Significance between control (day 1) and test conditions (days 5, 8, 14 and 25) was calculated using the Relative Expression Software Tool (Pfaffl et al., 2002) software via a bootstrapping approach (Pairwise Fixed Reallocation Randomisation Test). 2000 random reallocations of samples were performed; glyceraldehyde 3-phosphate dehydrogenase was employed as a housekeeping gene; comparable results were obtained using hypoxanthine-guanine phosphoribosyltransferase (data not shown).

**Correlation networks**

To construct correlation networks, qPCR data from host markers, OTU relative abundances from 16S rRNA gene amplicon pyrosequencing and the relative abundance of transcripts from SEED categories level 1 from metatranscriptome libraries were combined. For sequencing-based data, any OTU or transcript class that did not reach at least 1% relative abundance in any sample was excluded from the analysis. Data was log-normalized and the Pearson correlation coefficient was generated. Pearson correlation coefficients $r > 0.6$ or $r < -0.6$ were used to construct the correlation networks in R software environment, which were then visualized using Cytoscape (Shannon et al., 2003).

**Data access**

Amplicon and metatranscriptomic data were archived at NCBI Sequence Read Archive under SRP008057 and SRX314796, respectively.

**Results**

**Host markers of acute colitis**

Development and remission of DSS-induced colitis in mice was determined by weight measurements and histology scores, and by quantification of expression of pro- (IL-1β, IL-6 and IFNγ) and anti-inflammatory (IL-10) cytokines, the chemokine CXCL1 and the inducible nitric oxide synthase. DSS-treated mice started to lose weight at day 5 and reached their lowest weight at days 8 and 9 (Figure 1a). At day 14, weight had returned to day 1 levels, while at day 25 mice had gained 15% compared to their starting weight. This weight gain

![Figure 1](image)

**Figure 1** Host parameters during colitis and recovery. (a) Weight curves of untreated (control) and DSS-treated mice. Colitis was triggered by the addition of DSS (2%) to drinking water for 5 days (red box); control mice received un-amended drinking water during the course of the experiment. Samples were taken before (day 1) and after DSS treatment (day 5), during acute colitis (day 8) and recovery (days 14 and 25) (green arrows). (b) Inflammation scores (calculated as the sum of inflammation severity and extent, and crypt depth) and area of inflammation (%). (c) Representative photomicrographs (100× magnification) of hematoxylin and eosin-stained colon tissues obtained from each sampling date. (d) Expression ratios of pro- (IL-6, IL1β, IFNγ and CXCL1) and anti- (IL-10) inflammatory cytokines and chemokines and markers of inflammation (iNOS) calculated relative to expression at day 1 and GAPDH. Expression ratios were calculated using REST, asterisks (*) indicate significance of expression ratios as determined using Pair Wise Fixed Reallocation Randomisation Test (Pfaffl et al., 2002). NS, not significant; ND, not detectable.
was comparable to the weight gain of control mice not receiving DSS (Figure 1a). Tissue damage peaked at day 8 and was slightly decreased at day 14; the proportion of affected area declined even further to day 25 (Figures 1b and c). IL-10 was significantly downregulated only at day 5; IL1-β, IL-6 and CXCL1 were upregulated (P < 0.05) at day 8 and remained at significantly higher transcriptional level at days 14 and 25 (Figure 1d). IFNγ and inducible nitric oxide synthase were significantly higher expressed at days 8 and 14 (inducible nitric oxide synthase only), but not detectable or not significantly different at day 25, indicating resolution of acute inflammation (Figure 1c).

**Temporal changes in microbiota composition**

16S rRNA transcripts of the metatranscriptomes were used to investigate temporal alteration in microbial community composition. The bacterial phyla Firmicutes and Bacteroidetes contributed the majority of 16S rRNA transcripts (together 96–97%); with Clostridiales and Bacteroidales being the two dominant orders (together 94–95%, Supplementary Figure S1A). We applied the 16S rRNA transcripts as a measure of relative abundance of bacterial groups even though the rRNA content does not necessarily reflect cellular abundance and is affected in complex ways by the physiological state of the cell. PCA of the relative abundance of bacterial groups from 16S rRNA transcripts revealed that microbiota shifts appeared at the end of DSS treatment (day 5) and further alterations occurred at the peak of inflammation (day 8) (Figure 2a). At day 14, the microbial community began to resemble the community before inflammation, but did not reach its original composition even by day 25. PCA furthermore indicated a higher degree of community variability between replicates in the inflammation and recovery phases (Figure 2a). Shifts from the healthy to an inflamed and recovering microbiome occurred primarily along the first principal component; the clostridial family Lachnospiraceae was associated with the healthy status, while members of the Bacteroidetes (Bacteroidaceae, Prevotellaceae, Porphyromonadaceae and Rikenellaceae) were indicative of inflammation (Figure 2b). The increase of Bacteroidetes 16S rRNA occurred in two stages: unclassified Bacteroidales peaked at day 5, while members of the Bacteroidaceae, Porphyromonadaceae and Prevotellaceae were most abundant at day 8 (Supplementary Figure S1C). Some bacterial families low in relative 16S rRNA abundance, like the Enterobacteriaceae, Verrucomicrobiaceae (mainly Akkermansia spp.), Erysipelotrichaceae and Deferribacteraceae (mainly Mucispirillum spp.), also increased during acute inflammation, but 16S rRNA reads of these families returned to abundance levels of healthy mice at day 14, with the exception of the Erysipelotrichaceae (Supplementary Figure S1D), which only decreased by day 25. Generally, shifts in relative abundance of 16S rRNA of Clostridiales and Bacteroidales and the increase of Deferribacteraceae and Enterobacteriaceae 16S rRNA were most pronounced in mice that suffered from greatest weight loss and the most severe inflammation (Supplementary Table S4). During recovery, the ratio of Clostridiales to Bacteroidales 16S rRNA significantly increased compared to day 1 (Supplementary Figure S1B).

In a previous study (Berry et al., 2012), we showed that many microbiota rearrangements during acute colitis are hidden when 16S rRNA (gene) sequences are grouped on phylum, order or family level. To also determine composition shifts of roughly species-level phylotypes, which is not possible with rRNA reads derived from shotgun metatranscriptomes, 16S rRNA gene sequences derived from amplicon pyrosequencing...
were analysed. Similar to the metatranscriptomic data, principal coordinates analysis based on weighted and unweighted Unifrac distance indicated a health-state dependent clustering of bacterial communities (Supplementary Figure S2A–D). Phylotype richness had already decreased before the onset of inflammation (day 5) and only started to recover at day 14, whereas community evenness was generally not affected (Supplementary Figure S3).

To investigate compositional dissimilarity within bacterial groups across sampling days, Bray–Curtis dissimilarity index was calculated based on relative abundances or presence and absence of phylotypes (Supplementary Figure S4). In general, a shuffling of abundances rather than a replacement of community members was the cause for the observed community shifts. Shifts were more pronounced for members of the Bacteroidetes (Bacteroidales, unclassified Bacteroidales and Bacteroidaceae) than for the Firmicutes, and appeared already at day 5. Bray–Curtis dissimilarity based on relative abundances of phylotypes assigned to Bacteroidaceae and Lachnospiraceae varied to a greater extent at days 14 and 25 than in healthy mice (day 1).

**Bacterial gene expression during colitis and recovery**

In healthy mice, 69–70% of all putative bacterial mRNAs were assigned to Firmicutes (44–47%) and Bacteroidetes (23–26%) (53–56% to Clostridiales and Bacteroidales), ‘Protein Metabolism’ and ‘Carbohydrate Metabolism’ were the most abundant SEED categories; ‘Virulence’ and ‘Motility and Chemotaxis’ were the third major categories in Bacteroidetes and Firmicutes metatranscriptomes, respectively (Supplementary Figures S5–7). PCA generally clustered mRNA profiles according to the day of treatment (Figure 3a). Early changes in gene expression were already observed at day 5 but became most extensive on day 8 (Figure 3a). mRNA reads assigned to the categories ‘Motility and Chemotaxis’ and ‘Regulation and Cell Signalling’ were associated with healthy and recovering microbiomes (Figure 3b) and the former category correlated positively with Clostridiales and Lachnospiraceae ($r$: 0.516 and 0.528, $P<0.05$). Associated with inflammation were transcripts of ‘Virulence’, ‘Sulfur Metabolism’, ‘Stress Response’ and ‘RNA Metabolism’ (Figure 3b), all of which positively correlated with the Bacteroidales ($r$: 0.522, 0.625, 0.555 and 0.498, respectively, all $P<0.05$). During acute colitis, transcripts of ‘Motility and Chemotaxis’ were significantly reduced compared to day 1, while relative abundances of mRNA reads affiliated with ‘RNA Metabolism’, ‘Clustering-based subsystems’, ‘DNA Metabolism’, ‘Cofactors, Vitamins, Prosthetic Groups, Pigments’ and ‘Fatty Acids’ were significantly increased (Supplementary Figure S5).

**Acute colitis is associated with decreased expression of butyrate synthesis genes**

As carbohydrates are major energy sources of the gut microbiota, we more closely analysed transcripts of the category ‘Carbohydrate Metabolism’, which contributed ~20–25% of all transcripts to the metatranscriptome (Supplementary Figure S5). The relatively high abundance of transcripts assigned to ‘Fermentation’ and its subcategory ‘Acetyl-CoA-fermentation-to-butyrate’ and ‘Organic acids’ (sub-category ‘Propionyl-CoA to Succinyl-CoA Module’) reflected the major fermentative pathways of Clostridiales and Bacteroidales, that is, butyrate production and succinate-propionate conversion (Macy et al., 1978; Pryde et al., 2006), respectively (Supplementary Figure S8). Yet, whereas the relative abundance of transcripts related to succinate-pro-
pionate conversion remained unaffected, those associated with butyrate formation were significantly reduced on day 8 (Figure 4, Supplementary Figure S8). This decrease occurred within the entire metatranscriptome as well as within the metatranscriptome of Firmicutes (Figure 4, Supplementary Figure S8). The key enzymes acetyl-CoA acetyltransferase, which converts acetyl-CoA to acetoacetyl-CoA, 3-hydroxybutyryl-CoA dehydratase catalyzing the dehydration of 3-hydroxybutyryl-CoA to crotonyl-CoA and butyryl-CoA dehydrogenase converting acetacetyl-CoA to butyryl-CoA, were significantly reduced on day 8 compared to day 1 (Supplementary Table S5). Clostridiales possess two pathways for the final conversion of butyryl-CoA to butyrate (Louis and Flint, 2009). Butyrate kinase and phosphate butyryltransferase, which form butyrate from butyryl-CoA via the intermediate butyryl-phosphate, were not affected by acute colitis (Supplementary Table S5). Transcripts of butyryl-CoA:acetate CoA-transferase (ButCoA transferase) transferring the CoA of butyryl-CoA to external acetate were not detected within the metatranscriptomes. However, a significant decrease could be determined by quantitative PCR (day 1: log4.58 ± 0.59, day 8: log3.06 ± 0.35, transcripts μg⁻¹ RNA). In confirmation, a decrease of both ButCoA transferase transcripts and genes was also observed in independent additional experiments (Supplementary Figure S9). Additionally, transcripts associated with the category ‘Butanol Biosynthesis’ were significantly lower on day 8 (Figure 4). This category was mainly represented by pyruvate-formate lyase transcripts of Firmicutes. Also significantly reduced was ‘Polysaccharide Metabolism’ due to decreased contribution of Bacteroidetes transcripts (Figure 4, Supplementary Figure S8).

Decrease of clostridial flagellin transcripts and genes
The SEED category ‘Motility and Chemotaxis’ consisted to a large part (80%) of transcripts of clostridial ‘Flagellar Motility’. ‘Motility and Chemotaxis’ dominated the separation of functional clusters along the first principal component of the PCA plot and was also responsible for the large spread of data points during recovery phase (day 25) (Figure 3). At days 5 and 8, there were 21 and 49% less transcripts assigned to ‘Motility and Chemotaxis’ compared to day 1. (Supplementary Figure S5). Likewise, the relative abundance of ‘Motility and Chemotaxis’ was reduced in the metatranscriptomes of Firmicutes (Supplementary Figure S7). To determine whether this reduction that mirrored the decrease of Clostridiales and/or Lachnospiraceae 16S rRNA was due to species reorganization within the phyla Firmicutes or was caused by a downregulation of flagellin gene expression, we additionally determined abundances of clostridial flagellin genes and transcripts by qPCR (Figure 5). Flagellin genes and transcripts were reduced at day 8 compared to day 1, and flagellin transcripts were also lower at day 25 than at the start of the experiment. Yet, whereas at days 1 and 5 the ratio of transcripts to genes remained fairly stable, it was significantly reduced at days 8 and 25, and there was a strong variation in replicates at day 14 (Figure 5) hinting that the observed reduction of transcripts during acute colitis and at day 25 was not only caused by the reduced abundance of Lachnospiraceae but was also caused by a downregulation of flagellin transcription.

To further validate the decreased expression of clostridial flagellin genes during acute colitis, we re-analyzed data from a previous experiment in a different genotype (STAT1⁻/⁻ mice on C57BL/6 background; Berry et al., 2012), during which we had also

Figure 4 Shifts in ‘Carbohydrate metabolism’-related transcripts. Shown are relative abundances of transcripts of the categories (a) ‘Carbohydrate metabolism’ and (b) ‘Fermentation’ assigned with MEGAN at days 1, 5, 8, 14 and 25 in relation to all mRNA reads. Means at days 1 and 8 were compared with unpaired t-test, P<0.05 was considered significantly different (*).
observed a decrease of flagellins in metatranscriptomes. Also here, the ratio of flagellin transcripts to genes was reduced ~fourfold ($P<0.05$) in mice suffering from acute colitis compared to a control group (control group ($n=5$): $0.09 \pm 0.18$, colitis group ($n=5$): $-0.34 \pm 0.31$ [log ratio flagellin transcripts/genes]).

**Mucin degradation and utilization during acute colitis**

The SEED category ‘Aminosugars’, which includes β-hexosaminidases capable of releasing N-acetylglucosamines and N-acetylgalactosamines from O-linked mucins, was significantly more abundant in inflamed mice than in healthy or recovering animals (Figure 4). That led us to investigate whether the transcription of glycoside hydrolases putatively involved in mucin degradation and sulfatases catalyzing the rate-limiting step of mucin degradation (Wright *et al.*, 2000) were affected by inflammation. Beta-mannosidase, neuraminidase/sialidase, β-galactosidase and sulfatase transcripts were more abundant at day 8 compared to day 1; α-mannosidase and β-hexosaminidase transcripts were significantly enriched (Figure 6). Taken together, transcripts of all genes presumably involved in mucin degradation were significantly higher on day 8 compared to day 1 (Figure 6). This increase corresponded to a significant increase of transcripts of mucin-degrading enzymes within the *Bacteroidetes* transcriptomes during inflammation (day 1: $1.36 \pm 0.05\%$, day 5: $1.39 \pm 0.26\%$, day 8: $2.03 \pm 0.28\%$, day 14: $1.31 \pm 0.44\%$ and day 25: $1.33 \pm 0.11\%$).

Enhanced degradation of mucin could increase the availability of mucin components. Indeed, transcripts of fucose uptake and utilization were slightly more abundant while mannose and N-hexosaminidases uptake and utilization remained unaffected (Supplementary Table S6). Galactose uptake and metabolism was significantly reduced due to the decrease of the galactose/methyl galactoside ABC transport system MglABC (Supplementary Table S3). MglABC is a high-affinity ABC transporter acting as a scavenging system in case of low substrate concentration (Wilson, 1974), thus its downregulation suggested increased substrate availability.

**Microbiota-host networks in acute colitis and recovery**

We used correlation networks to identify bacterial phylotypes associated with functional transcripts and host parameters. IFN$_\gamma$, IL-6, IL-1β and CXCL1 expression levels correlated with pathology scores,
and members of this cluster correlated positively with *Turicibacteraceae* OTU 2651 and negatively with *Bacteroidales* OTU 102 (Figure 7a). Additionally, inducible nitric oxide synthase expression was negatively correlated with *Lachnospiraceae* OTU 2896 and *Bacteroidales* OTU 1366, and IL10 expression was positively correlated with *Ruminococcaceae* OTU 1695 and *Akkermansia* OTU 711. Examination of the first neighbours of phylotypes that correlated with host markers revealed extensive intra-phylum discordance in phylotype abundances in the *Bacteroidetes* and *Firmicutes*, for example, *Bacteroidales* OTU 183 decreased during inflammation and recovered by

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**Figure 7** Correlation networks of 16S rRNA gene amplicon defined OTUs, functional categories (SEED subsystems level 1) and host parameters. (a) Sub-network of inflammation or pathology-associated host markers and their first and second neighbours in microbiota data sets. (b) Relative abundance of the inflammation-associated *Bacteroides acidifaciens* OTU 1812 and the health-associated *Bacteroidales* OTU 183 in amplicon sequencing data and representative fluorescence *in situ* hybridization images of the two phylotypes. (c) Sub-network of first neighbours of TLR5. Black and red edges indicate positive and negative correlations, respectively ($r > 0.6$ or $r < -0.6$).
day 25, while Bacteroides acidifaciens OTU 1812 increased transiently during inflammation (Figure 7b), which highlights the inner-group dynamics occurring during dysbiosis. As clostridial flagellin transcription was heavily impacted during inflammation, we also quantified the expression of TLR5, the pattern recognition receptor of flagellins (data not shown). TLR5 correlated with ‘Motility and Chemotaxis’ and the anti-inflammatory cytokine IL-10 (Figure 7c).

**Discussion**

A global perspective on microbial gene expression pattern before, during and after acute intestinal inflammation has been lacking. We thus performed a longitudinal study of the intestinal metatranscriptome and bacterial community dynamics, and selected host parameters in DSS-treated mice to address the following questions: How resistant and resilient is the gut microbiota to acute intestinal inflammation? Which microbial activities are affected during acute colitis and recovery? How are microbial energy and carbon metabolism impacted by the inflammation? What are the impacts of colitis on host–microbiota interactions?

**Host and microbiota during progression of DSS-triggered acute colitis and recovery**

DSS destroys the mucus biopolymer network resulting in enhanced mucus permeability; however, mucus biosynthesis is not affected (Johansson et al., 2010). The loss of barrier function allows the translocation of microbes, leading to increased stimulation of epithelial immune cells and the induction of inflammatory responses (Kitajima et al., 1999). We started to observe weight loss, tissue inflammation and the downregulation of IL-10 known to limit the expression of cytokines from dendritic cells and macrophages after activation by TLRs (Murray, 2005) at day 5 (Figure 8). Microbiota alterations occurred concurrently with the onset of inflammation. Members of the Bacteroidetes were identified as possible indicators of disease onset in this study, and were previously shown to be capable of inducing colitis in antibiotic-pretreated mice (Bloom et al., 2011). However, abundance shifts of Firmicutes and Bacteroidetes and intra-phylum variations only had a minor impact on microbiota gene expression at day 5, highlighting a certain degree of functional redundancy of the microbial community in the face of slight perturbations. The significant upregulation of markers of inflammation, extensive weight loss and high inflammation scores marked the onset of full-blown colitis (day 8).

Confirming previous studies, acute colitis was characterised by a distinctive dysbiosis (Frank et al., 2007; Nagalingam et al., 2011; Berry et al., 2012) manifested by an increase of low abundant families and concurrent intra- and inter-phyla abundance shifts of the Bacteroidetes and Firmicutes.

Acute colitis likewise had major impacts on gene expression indicated by temporal abundance

![Figure 8](image-url) Timeline of the interplay of host and microbiota during colitis and recovery.
changes of mRNA belonging to broad functional categories. With the recovery of the host from acute inflammation, severely disturbed microbial traits contributing to host–microbiota interactions, such as butyrate formation and mucin utilization recovered quickly to pre-inflammation levels (day 14). In contrast, community composition was characterised by a memory of disturbance comparable to that observed in studies using antibiotics as disturbing agents (De La Cochetière et al., 2005; Dethlefsen and Relman, 2011). Upregulation of pro-inflammatory cytokines, inflamed tissue and a compositionally altered microbiota persisted up to day 25. In this state, the interdependent host–microbiota system might remain sensitive towards relapsing flares of inflammation.

Correlation networks were produced in order to compare changes in the microbiota with inflammation/immune state. This analysis revealed that pathology scores and expression of pro-inflammatory cytokines correlated and also were correlated (both positively and negatively) with several bacterial phylotypes (Figure 7a). With future validation, this approach of identifying gut bacterial phylotypes that correlate with host health markers may be a valuable tool for the development of novel non-invasive bacterial biomarkers of intestinal disease and may reduce the need for endoscopy.

**Impact of acute colitis on Clostridiales–host interactions**

Increasing evidence points to an errant immune response directed at commensals during acute inflammation (Hand et al., 2012); one frequently identified potential target molecule is the flagellin of *Clostridiales* (Lodes et al., 2004; Duck et al., 2007; Hand et al., 2012). A decrease of *Clostridiales* during acute colitis and the concurrent reduction of clostridial flagellins as described in this study has been repeatedly observed in humans and in animal models (Frank et al., 2007; Erridge et al., 2010; Berry et al., 2012), but the underlying causes have received little attention. Upon disruption of the mucus layer and epithelium by DSS treatment, the basolateral flagellin receptor TLR5 (Hayashi et al., 2001) and intracellular flagellin sensor NLRC4 (Lightfield et al., 2008) can be stimulated (Figure 8, Rhe et al., 2005), eventually leading to specific antibody production (Vijay-Kumar et al., 2010b). Indeed, we observed a correlation between relative abundance of flagellin transcripts and expression of TLR5 (Figure 7c). Vijay-Kumar et al., 2010a previously showed the regulatory impact of TLR5 on the gut microbiota as its inactivation resulted in increased bacterial load and changes in *Firmicutes* and *Bacteroidetes* species composition. In colitic mice and human IBD patients, antibodies targeting flagellins of commensal strains were dominant antigens (Lodes et al., 2004; Duck et al., 2007). This directed immune response against flagellated members of the major intestinal phylum *Firmicutes* might be one reason for the observed abundance shift of *Clostridiales* and *Bacteroidetes* in acute colitis.

The downregulation of flagellin transcription (Figure 5) observed here and in a previous study hints that members of *Lachnospiraceae* have developed a strategy to avoid, at least to some extent, attacks by the host in a similar way as pathogens that tightly regulate flagellin transcription to maintain low metabolic costs and to escape recognition by the host immune system (Winter et al., 2010). In *Bacteroides thetaiotaomicron*, the expression of a carbohydrate epitope was observed to negatively correlate with IgA levels, indicating that commensals are capable of modifying their surface structure in response to host activity (Peterson et al., 2007). Disturbances in flagellin expression could still be observed during the recovery phase. This is in line with observations that clostridial flagellin-specific memory T-cells that were initiated during acute colitis caused by *T. gondii* infection or DSS, persisted for more than two months and could be rapidly activated in case of another challenge (Hand et al., 2012).

**Impact of acute colitis on Bacteroidetes-host interactions**

In healthy mice, the phyla *Firmicutes* and *Bacteroidetes* both contributed to transcripts associated with enzymes involved in degradation of mucin glycoproteins. During inflammation, the *Bacteroidetes* might be the numerically abundant group of mucin degraders, as judged from an increase in their relative abundance and the expression of the respective transcripts. Increased abundance of transcripts related to mucus degradation was contrasted by decreased abundance of ‘Polysaccharide Metabolism’, possibly indicating a shift in substrate preference within the *Bacteroidetes*. This shift was likely due to shifting intra-phylum composition because Bray–Curtis dissimilarity, correlation network analysis and FISH revealed that this shift was not a common event in all *Bacteroidetes* phylotypes, but rather that a subgroup profited from the inflammation-induced changes in the intestinal milieu. One of the phylotypes enriched during acute colitis was *Bacteroides acidifaciens* (OTU 1812, Figure 7b), a species originally isolated from mouse cecum (Miyamoto and Itoh, 2000). *B. acidifaciens* was also previously identified as an indicator phylotype in acute colitis (Berry et al., 2012). *B. acidifaciens*, together with *Akkermansia*, was recently described as an important mucin-degrader in vivo (Berry et al., 2013). In the present study, the increase of *B. acidifaciens* was correlated positively with several mucin glycoprotein-degrading enzyme transcripts (α-1,2-mannosidase, r: 0.683, β-hexosaminidase, r: 0.713, sialidase, r: 0.551, P<0.05), suggesting that this strain, as a
mucin degrader, might have profited from the changed microenvironment during acute colitis.

In contrast to *B. acidifaciens*, the metatranscriptomic analysis did not provide indications for *Akkermansia* as mucus degrader during inflammation, although this organism possesses all genomic features to degrade complex mucus components (van Passel et al., 2011). Possibly, the sequencing depth of the metatranscriptome was not sufficient for such an identification, since *Akkermansia* abundances were very low in the mice analyzed.

Temporary alterations of metabolic interactions between microbiota and host during acute colitis

Abundance and intra-taxon shifts of *Clostridiales* and *Bacteroidales* during the acute phase of inflammation have far-reaching implications on metabolite-related host–microbiota interactions. The decrease in butyrate-producing clostridia indicated by the decrease of clostridial transcripts of pyruvate-formate lyase and key enzymes of butyrate fermentation pathways, potentially resulted in a reduction of butyrate production (Supplementary Figure S1, 4B, Supplementary Table S4). Concurrently, *Bacteroidetes* such as the mucolytic *B. acidifaciens* form acetate, succinate and potentially propionate as metabolic end products (Macy and Probst, 1979; Miyamoto and Itoh, 2000) shifting the balance of SCFA towards acetate and propionate. As acetate is absorbed and transported to the liver, a shifted SCFA balance can lead to a decrease in available butyrate as an energy source for colonic epithelial cells, but might also affect regulation of colonic cell proliferation and differentiation (Wong et al., 2006; Plöger et al., 2012). The high abundance of transcripts of pyruvate-formate lyase, an oxygen-sensitive enzyme that reversibly converts pyruvate to acetyl-CoA and formate in clostridia (Thauer et al., 1972) emphasizes its role as a second major pyruvate metabolism pathway beside pyruvate: ferrooxido reductase (Uyeda and Rabinowitz, 1971). Pyruvate-formate lyase was also prominent in the human intestinal metaproteome (Kolmeder et al., 2012).

Conclusion

In this study, gut microbiota activity and interactions with the host during acute inflammation and recovery were investigated. With the information gained, we proposed a time line of events that determined the interplay of host and microbiota (Figure 8). Inflammation temporarily had a large impact on the microbiota and the host; however, the host–microbiota homeostasis revealed a significant resilience, particularly evident in the transcriptional activity of the gut microbiota. While acute DSS colitis is certainly different from IBD, there are some similarities such as shift in microbiota composition with decrease of flagellated, butyrate producing *Lachnospiraceae* and an increase of *Enterobacteriaceae* (Frank et al., 2007), as well as changes in mucus degrader activity (Png et al., 2010). Therefore, the mechanistic outcome of this study might be relevant for research on human IBD.

Conflict of Interest

The authors declare no conflict of interest.

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