Microbial and metabolic features associated with outcome of infliximab therapy in pediatric Crohn’s disease

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ABSTRACT

Gut microbial dysbiosis and altered metabolomics have been implicated in the pathogenesis of Crohn’s disease (CD). The aim of our study was to characterize the gut microbiome structure and metabolic activities in pediatric CD patients with different clinical outcomes after infliximab (IFX) therapy. Fecal samples were collected from 20 healthy children and 29 newly diagnosed pediatric CD patients. 16S rRNA/ITS2 gene sequencing and targeted metabolomics analysis were applied to profile the gut bacterial microbiome, mycobiome, and metabolome, respectively. Pediatric CD patients exhibited lower relative abundances of short-chain fatty acids (SCFAs)-producing bacteria including Faecalibacterium, Clostridium clusters IV and XIVb, Roseburia, and Ruminococcus, which were correlated with reduced fecal levels of SCFAs. Decreased unconjugated bile acids (BAs) pool size and a lower unconjugated/conjugated BAs ratio were associated with reduced relative abundances of Bifidobacterium and Clostridium clusters IV and XIVb which contain bile salt hydrolases (BSH) genes. IFX treatment enriched the BSH-producing bacteria in CD subjects, which may explain a decreased level of conjugated BAs and an increase in unconjugated BAs as well as the unconjugated/conjugated BAs ratio. Furthermore, a sustained response (SR) of IFX therapy was associated with higher abundances of Methylobacterium, Sphingomonas, Staphylococcus, and Streptococcus, and higher fecal concentrations of amino acids, including L-ascorbic acid, linoleic acid, and L-lactic acid at baseline. Our study suggests that the effects of IFX might be partially mediated by enriching bacteria taxa that producing SCFAs and BSH thereby inhibiting inflammation and restoring the BA metabolism. Some fecal bacteria and metabolites may be predictive outcomes of IFX therapy for pediatric CD patients.

INTRODUCTION

Crohn’s disease (CD) is a major form of inflammatory bowel diseases (IBD) characterized by chronic transmural inflammation, mucosal aphthous ulcers, and noncaseating granulomas in the gastrointestinal (GI) tract. The incidence rate of CD in the pediatric population is steadily increased in past two decades, particularly in the newly developed countries. Children with CD usually suffered episodes of disease remission and relapse that significantly impair their growth and development. The exact etiology and pathogenesis of CD are still unclear, but emerging evidence indicates that CD is strongly associated with disturbances to the gut microbiota.

Currently, there are several studies reported the alterations of gut bacterial microbiota in pediatric IBD patients. The numbers of Coriobacteriaceae and Lachnospiraceae, and the diversity of Ruminococcaceae and Bifidobacterial population were lower in pediatric IBD patients as compared with the healthy children. It was reported that the abundances of bacteria within families Enterobacteriaceae, Fusobacteriaceae, Pasteurellaceae, and Veillonellaceae were higher, and abundances in orders Bacteroidales, Clostridiales, and Erysipelotrichaceae were lower in newly diagnosed pediatric CD patients than the healthy children. Our previous study showed that the fecal microbiota of pediatric CD patients was represented by a lower biodiversity,

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Supplemental data for this article can be accessed on the publisher’s website.

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a gain in Enterococcus, and a significant loss in multiple short-chain fatty acids (SCFAs)-producing bacteria. Beyond bacteria, other microorganisms present in gut microbiota, such as fungi, have been suggested to play a role in IBD pathogenesis. The dominating fungal microbiota was species with Basidiomycota in de-novo pediatric IBD patients, while Ascomycota was predominated in healthy subjects. Chehound et al. reported a lower diversity of mycobacteria and an increasing abundance of Candida spp. in pediatric IBD patients. Compared to the healthy children, Pseudoxanthomonas, Cortinariaeaceae, Pseudophyllaca, and Gymnopus ilus were significantly enriched, whereas Monilinia was significantly depleted in CD children. However, the prior results on this topic have been inconsistent.

The gut microbiota interacts with the host mainly through signals triggered by microbial metabolites. Changes in the gut microbiota correspond with altered microbial metabolic functions have been implicated in the pathogenesis of IBD. Metabolomic profiling of biofluids in IBD patients may help to find novel biomarkers for disease diagnosis and monitoring, and reveal the role of metabolic alterations in disease development and progress. Alterations in a number of metabolites in patients with IBD were confirmed by metabolomics studies, such as fecal bile acids (BAs) and SCFAs. Metabolic profiling of serum samples revealed that numbers of chemically annotated metabolites belong to phospholipids were downregulated in CD children. Kolho et al. showed that serum and fecal metabolite profiles in newly diagnosed pediatric IBD patients were different from healthy children, neopterin was the top metabolite in the discriminant analysis with a high level in serum of CD. Nonetheless, the microbial origins of CD-associated metabolites and the mechanisms underlying gut microbiota-mediated changes in CD metabolome have not been fully investigated.

A prolonged treatment with anti-tumor necrosis factor (TNF) agents, such as infliximab (IFX), is frequently advised for moderate to severe CD patients. Despite high primary response rates of 70–90%, the long-term outcome of CD patients treated with IFX is still suboptimal. Many CD patients with an initial clinical response have recurrent symptoms during IFX maintenance therapy, and nearly 40% of them lost the clinical responses eventually. Some biomarkers for predicting response to anti-TNF treatment have been identified, including age, body mass index (BMI), duration of disease, fecal calprotectin, and serological anti-bodies. Recent studies have suggested that assessing the gut microbial community and metabolomics may provide new insights into anti-TNF treatment optimization. It was reported that increased Bifidobacterium, Collinsella, Lachnospira, Lachnospiraceae, Roseburia, Eggerthella taxa and reduced Phascolarctobacterium were associated with treatment success of anti-TNF therapy in CD patients. F. prausnitzii represented a predictive factor for recurrence after IFX discontinuation in CD. We previously showed that a sustained response (SR) of IFX therapy in CD children was positively associated with an expansion of SCFA-producing bacteria, especially the Blautia, Faecalibacterium, Lachnospira, and Roseburia. Metabolite profiling of serum and urine identified histidine and cysteine as biomarkers of response to anti-TNF therapy in adult CD. Furthermore, fecal levels of butyrate and substrates involved in butyrate synthesis were significantly associated with clinical remission of CD patients following anti-TNF therapy.

In the current study, we aim to investigate the characteristics of fecal microbiome (bacteria and fungi) and metabolome, and attempt to explore their relationships and functional roles in clinial response of IFX therapy in a group of pediatric CD patients.

Results

Responses of pediatric CD to IFX therapy

Demographic and clinical characteristics of the 29 pediatric CD patients included in this study were shown in (Table 1). The median age of the 29 CD patients was 13 y (interquartile range (IQR), 11–14 y), and 21 were boys. The majority of patients (25/29, 86.2%) having ileocolonic disease (L3). These CD patients had a median baseline PCDAI score of 32.5 (IQR, 20.0–37.5) and with a median CRP of 41 mg/L (IQR, 15–56 mg/L). During the study period, 18 patients received 3–6 times of IFX therapy. A total of 11 patients achieved SR, defined as (1) a PCDAI ≤10 at each follow-up time point, or (2) a ≥10-point reduction in the PCDAI from baseline to after the third IFX infusion and PCDAI ≤10 after the sixth IFX infusion. Furthermore, significant decreases in WBC, ESR, PLT, and CRP values were observed in SR patients after IFX therapy (Table S1). Among seven patients with non-sustained response (NSR, PCDAI >10 after sixth IFX
therapy), two patients (CD06, CD12) had early clinical response after third IFX therapy; however, they lost response in the subsequent IFX maintenance therapy.

**Pediatric CD is linked to specific fecal microbiome and mycobiome fingerprints**

Regarding the 16S rRNA data, the read counts per sample are 186, 189 in average, ranging from 36,481 to 525,028. After sequence processing and filtering, we obtained a total of 1,628,685 16S rRNA gene sequences spanning a total of 2,872 bacterial OTUs. No significant difference in the alpha diversity of bacterial communities was identified between CD and HS, except Chao 1 (Figure S1a). The bacterial microbiota was dominated by phyla Actinobacteria, Bacteroidetes, Firmicutes, and Proteobacteria in both pediatric CD patients and HS (Figure S1b). The top three most abundant genera in HS were Bacteroides, Bifidobacterium, and Blautia, whereas potentially pathogenic genus Escherichia/Shigella was more prevalent in the pediatric CD patients at baseline (Figure S1c). Multidimensional scaling revealed a significant clustering of samples according to the occurrence of CD (Figure 1a). Inter-group comparisons of taxonomic profiles at the genus level revealed that fecal samples from pediatric CD patients exhibited lower relative abundances of Acidainococcus, Bifidobacterium, Blautia, Clostridium IV, Clostridium XIVb, Faecalibacterium, Fusibacter, Gemmiger, Parasutterella, Romboutsia, Roseburia, Ruminococcus, and Turicibacter, and higher relative abundances of Clostridium XI, Enterococcus, Escherichia/Shigella, and Peptostreptococcus as compared to that of HS (Figure 1b).

As to the ITS data, the average counts per sample are 278,523, ranging from 6,955 to 117,618. We obtained a total of 828,553 ITS2 gene sequences spanning a total of 1,250 fungal OTUs. No statistically significant difference in the alpha diversity of fungal community was detected between samples from pediatric CD patients and HS (Figure S1d). In addition, no clear separation was observed between fungal community structure of CD and HS (Figure 1c). The fungal microbiome was dominated by phyla Ascomycota and Basidiomycota in both the pediatric CD patients and HS (Figure S1e), and the most commonly observed genera were Saccharomyces, Candida, Aspergillus, and Gibberella (Figure S1f). Compared to that of HS, fecal samples from the pediatric CD patients exhibited higher relative abundances of Alternaria and Thielavia, and lower relative abundances of Botryotinia, Coprinellus, Eupenicillium, Exophiala, Geosmithia, Hyphopichia, Myxomucidium, Phellinidium, Trichosporon, and Wickerhamomyces (Figure 1d).

**Dysmetabolism in the pediatric CD is characterized by reduced SCFA production and altered BA profile**

A total of 215 metabolites were detected and quantified in all the fecal samples, and different distribution patterns of metabolites between CD and HS were obser
Among the ten metabolites having the most discriminative (i.e., the highest loading values according to VIP) for classification between the two groups, CD patients showed higher fecal levels of two amino acids (L-leucine and L-norleucine) and two organic acids (methylmalonic acid, succinic acid), and lower levels of all major SCFAs (acetic acid, butyric acid, and propanol acid) and three BAs (deoxycholic acid, hyodeoxycholic acid, and lithocholic acid) (Figure 2c). Notably, the reduced levels of SCFAs in the CD patients could be well correlated with the lower abundances of multiple SCFA-producing taxa assigned to Firmicutes phylum, such as Faecalibacterium, Clostridium clusters IV and XIVb, Roseburia, and Ruminococcus (Figure 1b). ROC analysis showed methylmalonic acid, L-proline, L-phenylalanine, and L-leucine presenting the highest discrimination value (Figure 2d-g; AUC > 0.9).

Alterations in BAs metabolism are often seen as a consequence of changes in the gut microbial community by affecting the bile salt biotransformation. Our data showed that total BAs were not significantly different between CD and HS (Figure S2a). However, the pediatric CD group showed a higher level of conjugated and lower level of unconjugated BAs (Figure S2b,c). Along this line, the ratio of unconjugated/conjugated BAs was also lower in the CD samples (Figure S2d). Moreover, the level of primary BAs was higher, and secondary BAs was lower in the CD children, which led to a lower secondary/primary BA ratio than that of HS (Figure S2e-g).
Putative mechanistic associations between CD-related gut microbes and metabolites

In order to couple the alterations of the gut microbiome and dysmetabolism, Spearman rank correlations were calculated between the fecal concentrations of metabolites, and the abundances of bacterial/fungal taxonomic groups. A total of 399 significant bacteria-metabolite correlations were identified at the genus level (Figure 3). With a cutoff value of 0.6 for Spearman correlation co-efficient, we further identified 80 strong correlations between 26 metabolites (8 amino acids, 8 fatty acids, 6 BAs, 3 organic acids, and 1 benzenoid) and 20 bacterial genera (12 belong to the Firmicutes phylum, Table S2).

Gut bacterial metabolism plays vital roles in determining the intestinal pool of bile salts. In the intestine, the primary BAs, cholic acid and chenodeoxycholic acid are converted by intestinal bacteria to form the secondary BAs, deoxycholic acid, and lithocholic acid, respectively. This transformation is processed via deconjugation that is catalyzed by bile salt hydrolases (BSH), an enzyme expressed predominantly by Bacteroides, Bifidobacteria, Clostridium, and Lactobacillus, and followed by 7α-dehydroxylase (mainly expressed by Clostridium and Eubacterium). The gut microbiome analysis revealed lower abundances of genera Bifidobacteria, and Clostridium (clusters IV and XI) in the pediatric CD patients (Figure 1b), which may explain the cause of reduced unconjugated BAs pool and lower unconjugated/conjugated BAs ratio (Figure S2c, d).

IFX modified the gut bacterial, fungal microbiome, and metabolome

We next looked at the variations of bacterial, fungal communities, and metabolome in feces by discrimina

Figure 2. Altered fecal metabolomics in the CD patients compared with the HS. (a) Heat map showing the top 50 metabolites that different between the CD and HS, and the significance was determined by using Wilcoxon rank-sum test (***P < .001). (b) Supervised clustering of fecal metabolites using sPLS-DA. (c) The top metabolites ranked by VIP scores. (d–g) ROC analysis of potential biomarkers (AUC>0.9) for differentiating the CD patients from HS. CD, Crohn’s disease; HS, healthy subjects; sPLS-DA, sparse partial least squares discriminant analysis; VIP, variable importance in projection.
Figure 3. Potentially mechanistic associations between CD-linked microbes and metabolites. Network of the top 100 pairwise metabolite-microbe correlations. Metabolites and microbes are represented as red circles and blue ovals. Their positive and negative correlations are indicated using red and green color, respectively. Differential metabolites and bacteria between the CD and HS at baseline (Mann–Whitney U test, \( P < .01 \)) were selected for the Spearman correlation coefficients. Coefficient values \( R \leq -0.5 \) and \( \geq 0.5 \) with \( P \) value \( < .01 \) were considered statistically significant, and were plotted in network. CD, Crohn’s disease; HS, healthy subjects.

...tting samples before and after IFX treatment. To avoid the effect of inter-individual differences on the interpretation of the results, only CD subjects provided both baseline and post-treatment fecal samples were included in the analysis. The bacterial community diversity was not significantly different in the CD patients before and after IFX treatment (Figure S3a). PCoA showed that the IFX-treated samples did not form a cluster of samples that distinguished from the treatment-naive samples (Figure 4a). Comparisons of the bacterial microbiome profile revealed that the relative abundances of Blautia, Clostridium IV, Collinsella, Eubacterium, and Ruminococcus were increased, whereas Abitoraphia and Lactococcus were decreased in the CD patients after IFX treatment (Wilcoxon on rank-sum test \( P < .05 \); Figure 4b). Regarding the fecal mycobionta, no statistically significant change in alpha diversity was detected after IFX treatment (Figure S3b), and PCoA did demonstrate clustering of post-IFX samples (Figure 4c). At the genus level, an increased relative abundance of Galactomyces was observed after IFX therapy (Wilcoxon rank-sum test \( P < .05 \); Figure 4d).

By using Spearman rank correlation, we identified 21 bacterial genera with significant correlations with the severity of pediatric CD (Figure S4a). In specific, the PCDAI score and some serum indexes of inflammation, including CRP, WBC, PLT, and ESR, were positively correlated with the relative abundances of Mycoplasma, Parasutterella, Finegoldia, and Haemophilus, whereas negatively correlated with Blautia, Egg erthella, and Eubacterium. Regarding the fungal microbiota, we found that PCDAI score was negatively correlated with the relative abundances of Galactomyces, while positively correlated with Rhodospiridium and Kabatiella (Table S3).

With respect to fecal metabolome, profound metabolic changes were detected in the pediatric CD patients after IFX treatment (Figure 5a, b). We further identified ten metabolites that were most responsible for separation of the IFX-treated samples from that of baseline. In specific, the concentrations
of amino acids L-aspartic acid, glycine, and organic acid, arachidonic acid, docosahexaenoic acid, docosapentaenoic acid 22n6, eicosapentaenoic acid, L-lactic acid, and threonic acid were decreased, while azelaic acid was increased in the IFX-treated samples (Figure 5c).

In the fecal BA pool (Figure S5), the size of conjugated BAs was significantly decreased (Figure S5b), while the ratio of unconjugated/conjugated BAs ratio was significantly uplifted after IFX treatment (Figure S5d), which was in accord with an increase in the relative abundances of Blautia and Collinsella (Figure 4b) that containing BSH-producing species.

Moreover, IFX treatment induced a significant increase in the ratio of secondary/primary BAs (Figure S5g), albeit an increase in the level of secondary BAs did not achieve a statistical significance (Wilcoxon rank-sum test, P = .078; Figure S5f). This phenomenon may be explained by an increase in 7α-dehydroxylating genera including Clostridium IV and Eubacterium (Figure 4b).

Spearman rank correlation analysis found 13 amino acids (L-alanine, L-aspartic acid, L-cystine, L-glutamic acid, L-histidine, L-kynurenine, L-leucine, L-nor leucine, L-proline, L-phenylalanine, glycine, glycyll-proline, and N-acetyl-L-aspartic acid), 3 fatty acids (docosapentaenoic acid 22n6, arachidonic acid, and eicosapentaenoic acid), 1 BA (glycochenodeoxycholic acid), 2 dicarboxylic acids (methylmalonic acid and succinic acid), 4 organic acids (cis-aconitic acid, trans-aconitic acid, citric acid, and isocitric acid) were positively correlated with PCDAI and other serum indexes of inflammation (Figure S4b). In contrast, azelaic acid, butyric acid, deoxycholic acid, isovaleric acid, valeric acid, and sebamic acid were negatively correlated with the CD disease severity (Figure S4b). Moreover, the PCDAI-correlated bacteria and metabolites were also strongly correlated with each other (Figure S4c), indi-
cating their interactions were closely associated with the severity of inflammation in the CD patients and the outcomes of IFX therapy.

**Microbial and metabolic features indicating therapeutic outcomes of IFX**

We next tried to identify specific features of the gut bacterial, fungal microbiome, and metabolome that indicative of therapeutic outcomes of IFX. The stool samples classified for the SR patients who did not immediately achieve PCDAI ≤10 (second sample for CD15, CD17, and CD19) were treated as response samples, and stool samples for NSR patients who initially responded (second stool sample for CD12) were treated as non-response in the data analysis (Table S1). No difference in the gut bacterial alpha diversities between SR and NSR patients was observed before (Figure 5a) and after (Figure 5b) IFX treatment. Comparisons of taxonomic profiles of samples from the CD patients before IFX treatment showed that the relative abundances of several bacterial genera of the NSR subjects were significantly higher than that of SR, including *Clostridium XI, Clostridium XVIII, Eggerthella, Lachnospiracea incertae sedis, Parabacteroides*, and *Peptococcus*, while the SR patients had higher abundances of *Methylobacterium, Sphingomonas, Staphylococcus*, and *Streptococcus* (Wilcoxon rank-sum test, P < .05; Figure 6a). After IFX therapy, the genera *Actinomyces, Atopobium*, and *Parabacteroides* were more abundant in the SR patients, whereas *Dorea* and *Holdemania* were found to be more abundant in the NSR patients (Wilcoxon rank-sum test, P < .05; Figure 6b).

No significant difference in the gut fungal alpha diversity was identified between the SR and NSR patients before (Figure 6c) and after (Figure 6d) IFX treatment. Compositionally, the NSR subjects showed a higher relative abundance of *Chaetomium* and a lower abundance of *Malassezia* after IFX treatment (Figure 6c).

Interestingly, we observed distinct fecal metabolome profiles between the pediatric CD patients with different therapeutic outcomes before IFX treatment (Figure 7a, b). There was a pattern in fecal metabolome profiles of the pediatric CD patients with SR in prior to IFX treatment (cluster M1 in Figure 7a), which was featured by relatively higher fecal concentrations of glycine, linoleic acid, and L-lactic acid (Figure 7c). Among the metabolites that enriched in the NSR patients (cluster M2 in Figure 7b), the characteristic metabolites included N-acetylserotonin, met hyglutaric acid, adipic acid, 4-aminohippuric acid, citramalic acid, isovaleric acid, and nicotinic acid (Figure 7c).
We further identified distinct fecal metabolome profiles between the pediatric CD patients with different therapeutic outcomes after IFX treatment (Figure 7d, e). Higher concentrations of glycine and L-lactic acid were found in the SR patients, whereas the NSR patients showed higher fecal levels of azelaic acid, N-acetylserotonin, sebacic acid, pimelic acid, isovaleric acid, nicotinic acid, adipic acid, and 4-aminohippuric acid (Figure 7f).

**Discussion**

The components of microbial communities and their metabolites regulate host energy metabolism, immune homeostasis, and mucosal integrity. The alterations in gut microbiome and metabonomics have been described in many immune-mediated diseases, including IBD. In this study, we characterized the gut microbiota dysbiosis and the altered metabolome in pediatric CD patients, and explored their interactions involved in the therapeutic effects of IFX. The gut microbiota dysbiosis in children with CD was characterized by specific taxonomic alterations in the fecal microbial community of both bacteria and fungi. Agreeing with previous findings, the CD children showed reduced relative abundances of *Bifidobacterium*, *Blautia*, *Clostridium IV*, *Faecalibacterium*, *Clostridium XIVb*, *Romboutsia*, *Roseburia*, and *Turicibacter*, and enhanced relative abundances of *Clostridium XI*, *Enterococcus*, *Escherichia Shigella*, and *Peptostreptococcus*. In addition, emerging evidence suggests a role of fungal microbiota in CD pathogenesis in both adult...
and pediatric populations. Several studies that included small number of patients reported a mycobiota dysbiosis in pediatric CD patients. Our results showed that the mycobiota of pediatric CD patients were characterized by higher relative abundances of *Alternaria* and *Thielavia*, and lower relative abundances of *Botryotinia*, *Coprinellus*, *Eupenicillium*, and *Wicherhamomyces* than that of HS.

Metabolites, small molecules that are derived from bacterial metabolism of dietary substrates, modification of host molecules, or directly from bacteria, are key mediators of microbiota–host interactions.
Microbial metabolites are detectable in different types of biofluids, such as serum, urine, and blood. Previous metabolomics studies have revealed specific metabolic signatures that discriminate between IBD patients and healthy individuals. However, these studies were heterogeneous for including different subtypes of IBD and using different metabolomics approaches. So far, limited metabolomics studies were performed in pediatric IBD patients. Previous fecal metabolomic profiling studies showed that the metabolotype associated with pediatric CD was characterized by increased BAs, taurine, and tryptophan. It has been shown that the serum levels of neopterin, L-arginine, and dimethylglycine were upregulated, while kynurenic acid and trimethylamine-N-oxide were downregulated in pediatric CD patients as compared to healthy children. Ni et al. showed that fecal amino acids and their derivatives were positively correlated with the intestinal inflammation in children with CD, including proline, phenylalanine, and leucine that confirmed in this study. Overall, our data revealed that the fecal metabolome of the pediatric CD patients was remarkably altered, which were featured by higher levels of several amino acids (L-leucine, L-norleucine) and organic acids (succinic acid, methylmalonic acid), lower levels of SCFAs, and reduced unconjugated/conjugated BAs ratio.

SCFAs, such as acetate, propionate, and butyrate, are an important fuel for intestinal epithelial cells and play a vital role in maintaining the gut barrier function. Moreover, SCFAs, especially butyrate, are known to exert immunomodulation effects via expanding the pool of intestinal regulatory T cells, and maintain epithelial homeostasis through the production of interleukin (IL)-18 via inflammasome activation. The fecal level of SCFAs and gut bacteria fermentation fibers that produce SCFAs were typically reduced in IBD patients. As expected, our data demonstrated that all the major SCFAs including acetic acid, butyric acid, propanoic acid, and some SCFA-producing bacteria taxa, such as genera Faecalibacterium and Roseburia, were reduced in the pediatric CD patients compared to the HS. The deficiency of SCFAs made them to be tested as potential therapeutic agents in treating IBD. However, different approaches, including enemas of butyrate, or mixtures of acetate, propionate, and butyrate resulted in diverse clinical outcomes in IBD patients.

There was a dysmetabolism of BA in the pediatric CD featured by a significant reduction in the secondary BA concentrations and unconjugated/conjugated BA ratio, which were likely caused by a loss of bacteria expressing BSH and 7a-dehydroxylase from our 16S rRNA data. This phenomenon agrees with the analysis of metagenomic samples from the Human Microbiome Project and MetaHit, in which reduction in the abundance of a cluster of bsh genes was identified in IBD that associated with certain members of Firmicutes phylum. In fact, BA and the gut microbiota exert bidirectional effects on each other. BA can influence the mucosal barrier integrity, and perform antimicrobial activity by inducing genes that encode anti-bacterial peptides and lectins via farnesoid X receptor (FXR). Alteration in BA metabolism can cause membrane damage to both microbial cells and the intestinal barrier, and impair the development of colonic RORy+ regulatory T cells thereby promoting intestinal inflammation. Taken together, our data indicate that gut microbiota dysbiosis may contribute to the inflammation and mucosal barrier damage in CD by altering the intestinal metabolome. How this altered metabolic state interacts with the gut microbiota (e.g., maintain microbiota dysbiosis by suppressing certain taxa through anti-bacterial peptides) remains to be illuminated.

It is of clinical value to identify biomarkers for predicting the response of anti-TNF therapy in the clinical management of CD. Recent studies suggested that the changes of gut microbiota were associated with treatment success of anti-TNF therapy in CD patients. It was shown that elevated proportions of Lachnospiraceae and Blautia in feces at week 6 after IFX treatment were associated with the clinical and endoscopic response to IFX in adult CD patients. Clostridiales was found to be significantly increased in CD patients responding to IFX therapy, which can predict the treatment effectiveness with 86.5% accuracy. Gut microbiome analysis from mucosal biopsy samples of adult CD patients showed a high abundance of Hungatella, Ruminococcus gnavus, and Parvimonas at baseline typifies responsive patients to IFX, whereas high abundances of Blautia, Faecalibacterium, Roseburia, and Negativibacillus genera are associated with disease refractory. Our data showed that the IFX responders were characterized by higher abundances of Methylobacterium, Sphingomonas, Sta...
phylococcus, and Streptococcus at baseline, while Clostridium XI, Clostridium XVIII, Eggerthella, Lachnospira racerea incertae sedis, Parabacteroides, and Peptococcus were more abundant in the non-responders. In addition, we further confirmed that the abundances of several SCFA-producing bacteria were increased after IFX treatment in the pediatric CD patients. Above inconsistent results generated from different studies may be due to the limited size, different study population, and large age differences across the study subjects.

In addition to gut bacteria, mycobacteria analysis revealed that two fungal genera were potentially associated with the IFX response. A lower relative abundance of Chaetomium and a higher abundance of Malassezia were identified in the SR patients after IFX treatment. Malassezia genus has recently been associated with IBD. For example, Malassezia restricta was found to be enriched in the mucosa of CD patients, which may exacerbate colitis via producing inflammatory factors including TNF-α and IL-8. Malassezia symphodialis has been found with a decrease in abundance in flare status of CD patients and increased in those with remission. Identifying specific species of Malassezia may further our understanding of their roles in CD and associations with response to IFX.

Recent studies have suggested that metabolites involving lipid, BAs, and amino acid pathways may contribute to predict response of anti-TNF therapy in adult IBD. Bjerrum et al. found distinct fecal metabolome profiles between IBD patients with different IFX outcomes, however, no applicable response biomarkers could be identified. Ding et al. demonstrated that a range of metabolic biomarkers involving amino acid, BA, and lipid pathways may have a potential to predict the response of CD patients to anti-TNF therapy. In particular, serum levels of phosphocholines, ceramides, sphingomyelins, triglycerides, serum and fecal BA, histidine, and urinary cysteine were found to be associated with non-response to anti-TNF therapy in adult CD patients. However, none of the proposed metabolic biomarkers were identified in our data. The inconsistent findings may be partially explained by a great difference in the age of patients between the two studies. We showed that metabolic phenotypes associated with SR were characterized by higher levels of glycine, linoleic acid, and L-lactic acid in prior to IFX treatment compared to NSR. Nevertheless, due to the inconsistent results of limited available studies and small size of our study, the roles of metabolites in predicting anti-TNF therapy in IBD are needed to be further investigated.

There are several limitations that exist in this study. First, although this is the largest longitudinal cohort study of pediatric CD using metabonomic and metatranscriptomic profiling from China to date, the subgroups within this cohort were small, which may lead to unintentional bias, a discovery-validation cohort study is needed to validate the findings in the future. Second, metabolome and microbiome analysis performed only on fecal samples collected from two time points after IFX treatment that may dilute the time-dependent effect of IFX, and the inadequate sequencing of samples from patients with severe diarrhea may cause a possible bias despite no significant difference was obtained by comparing the serum inflammatory factors in patients with diarrhea to those with normal/lose stools at baseline (data not shown). Third, metatranscriptomic analysis is needed to further investigate the proposed mechanistic relationships between gut microbial community dynamics and metabolic changes.

In summary, the gut microbiota dysbiosis in pediatric CD likely contribute to inflammation and mucosa damage by altering the intestinal metabolism featured by reductions in SCFAs concentrations and an imbalance of unconjugated/conjugated BA ratio. The severity of CD and the outcomes of IFX therapy are correlated with the abundances of certain gut bacteria genera and levels of metabolites. Higher fecal levels of multiple amino acids are a common feature shared by the pediatric CD patients with SR before IFX treatment, which may further be assessed as potential prognostic biomarkers with a large study cohort.

Methods

Study design, subjects, and samples collection

This study was approved by the Regional Ethical Review Board in the Shanghai Children’s Hospital and the Regional Ethical Review Board in the Shanghai Tenth People’s Hospital of Tongji University. Written informed consent was obtained from parents or legal guardians of all pediatric participants. Twenty-nine children with CD were prospectively recruited to the stu
Genomic DNA extraction

Genomic DNA was extracted using the QIAamp DNA Stool Mini Kit (Qiagen, Germany) combined with the bead-beating method as suggested in the International Human Microbiome Standards (IHMS) protocol Q. In brief, 250 μL of the fecal samples was transferred to a 2 mL tube containing 0.3 g of 0.1 mm zirconia beads (BioSpec, USA), and homogenized for 5 min at 1,500 g on a Scientz-48 High-throughput Tissue Grinder (Scientz, China). The eluted DNA was quantified using a dsDNA HS assay on a Qubit 3.0 (Thermo Fisher Scientific, USA). The DNA concentration of each sample was adjusted to 50 ng/μL for subsequent 16S rRNA/ITS2 gene sequencing.

PCR amplification

Isolated genomic DNA was amplified for the 16S rRNA V3-V4 hypervariable regions (approximately 465 bases) using the universal primer set 319 F 5′-GCTACGGGAGGCAGCAG-3′/806 R 5′-GGACTACHVGGGTATCTAAT-3′. ITS2 primers (Forward 5′-GTGARTTCATCGAATCTTT-3′/Reverse 5′-GATATGCTTAAGTTCAGCGGT-3′) were used to amplify the ITS2 gene fragment (approximately 350 bases). For each sample, 25 μL PCR reactions were performed using 10 ng template genomic DNA, 0.5 μM of forward/reverse primers, and 1X Phusion® High Fidelity Buffer. Genomic DNA was initial denatured at 98°C for 1 min, followed by 30 cycles of denaturation at 98°C for 10 s, annealing at 50°C for 30 s, elongation at 72°C for 30 s, and a final extension step at 72°C for 5 min. The PCR products were purified using QIAquick Gel Extraction Kit (Qiagen, Hilden, Germany). Bar-coded libraries were generated using TruSeq® DNA PCR-Free Sample Preparation Kit (Illumina, USA) as per the manufacturer’s recommended protocol. Library quality was assessed on a Qubit 2.0 Fluorometer (Thermo Scientific, USA), and the library concentration was estimated on an Agilent 2100 Bioanalyzer (Agilent Technologies, USA). Each library was diluted to a final concentration of 12.5 nM and pooled in an equimolar ratio prior to clustering. Sequencing was performed using an Illumina NovaSeq platform (Illumina, USA), and 2 × 250 bp paired-end reads were generated.
Gene sequencing analysis

Quality control check on raw sequence data was performed using FastQC (Babraham Bioinformatics, UK). Post-processing of the sequencing reads was performed using USEARCH (version 10.0.240).\textsuperscript{56} Raw forward and reverse reads were joined, and quality trimmed by using -fastq_filter command with -fastq_maxee 1.0. The unique sequences in FASTA format were generated by using fastx_uniques command. NOISE3 algorithm was used to remove chimeras in the unique sequences, and produce representative zero-radius operational taxonomic units (OTU) for subsequent analyses.\textsuperscript{57} Ribosomal Database Project (RDP) reference database was employed for taxonomic assignment.

Calypso (version 8.84) was used to analyze the bacterial and fungal community composition data.\textsuperscript{58} The OTU counts were normalized by total sum normalization (TSS). Taxa have less than 0.02% relative abundances were excluded from the following analysis. Cumulative-sum scaling (CSS) was applied for correcting biases introduced by TSS, followed by log2 transformation to account for the non-normal distribution of the taxonomic counts data. The microbiome alpha diversity was measured using the Shannon’s, Chao1, Simpson’s, and Inverse Simpson’s indexes. The beta diversity among samples were calculated through Principal coordinate analysis (PCoA) to Bray–Curtis distance based on the OTU abundance. Permutational multivariate analysis of variance (PERMANOVA) was carried out to test whether the gut microbiome structure was significantly different between two groups. Statistical differences in alpha diversity and the abundances of taxa were assessed by Wilcoxon rank-sum test for comparison between two groups with $P$ values adjusted for False Discovery Rate (FDR).

Quantification of fecal metabolites

Targeted metabolomics analysis was performed to measure the absolute concentrations of fecal metabolites as previously described.\textsuperscript{39,60} For each sample, 10 mg of lyophilized homogenized feces was dissolved in 40 μL of water using a Bullet Blender Tissue Homogenizer (Next Advance, Inc., Averill Park, NY). A 200 μL aliquot of methanol containing 50 internal standards was homogenized. After centrifugation, 40 μL supernatant from each sample was transferred to a 96-well plate, and freshly prepared derivative reagents were added to each well. The plate was sealed and the derivatization was carried out at 30°C for 60 min. After derivatization, 300 μL of ice-cold 50% methanol solution was added to dilute the sample. Then, the plate was stored at −20°C for 20 min and followed by 4,000 g centrifugation at 4°C for 30 min. A 150 μL of supernatant was transferred to a new 96-well plate for injection. A 5 μL aliquot of sample was injected for liquid chromatography–tandem mass spectrometry analysis measurement, and the flow rate was 0.4 mL/min.

The fecal metabolites were quantified with an ultra-performance liquid chromatography (UPLC) coupled to tandem mass spectrometry (UPLC-MS/MS) system (ACQUITY UPLC-Xevo TQ-S, Waters Corp., Milford, MA, USA). The MassLynx software (version 4.1, Waters, Milford, MA) was used for instrument control, data acquisition and processing. Chromatographic separation was achieved by an ACQUITY UPLC BEH C18 column (2.1 mm × 100 mm, 1.7 μm). UPLC-MS raw data obtained with both positive and negative ionization mode were analyzed using TargetLynx™ application manager (Waters Corp., Milford, MA, USA) to obtain calibration equations and the quantitative concentration of each metabolite.

MetaboAnalyst 4.0.\textsuperscript{61} was used for statistics and feature selection of the metabolomic data. Sparse partial least squares discriminant analysis (sPLS-DA) and random forest model was employed for classification and feature selection. Univariate statistical analysis was performed using Wilcoxon rank-sum test with $P$ values adjusted with Bonferroni correction procedure.

Metabolic function network analysis

Metabolic function network analysis was performed in the M2IA server.\textsuperscript{62} Differential metabolites and bacteria between CD and HS at baseline were selected at first according to univariate analysis (Mann–Whitney $U$ test, $P$ value$<.01$). The Spearman correlation coefficients were then calculated between the differential metabolites and bacteria using pairwise correlation analysis method. Coefficient values $R ≤ −0.5$ and $≥0.5$ with $P$ value $< .01$ were
considered statistically significant and were plotted in the network.

**Abbreviations**

- AUC, area under the curve
- BA, bile acids
- BSH, bile salt hydrolases
- CA, cholic acid
- CD, Crohn’s disease
- CDCA, chenodeoxycholic acid
- CRP, C-reactive protein
- ESR, erythrocyte sedimentation rate
- HB, hemoglobin
- HS, healthy subjects
- IBD, inflammatory bowel disease
- IFX, infliximab
- IQR, interquartile range
- NSR, non-sustained response
- OTU, operational taxonomic units
- PCDAI, Pediatric Crohn’s Disease Activity Index
- PCoA, principal coordinate analysis
- PLS-DA, partial least squares discriminant analysis
- PLT, platelets
- RDP, ribosomal database project
- ROC, receiver operating characteristic
- SCFAs, short-chain fatty acids
- SR, sustained response
- TNF, tumor necrosis factor
- UC, ulcerative colitis
- UPLC, ultraperformance liquid chromatography
- VIP, variable importance in projection
- WBC, white blood cells

**Author contributions**

YW, ZL, and TZ designed the study. XG analyzed the omics data and created figures. YW and XG interpreted the data, and wrote the manuscript. XZ, FX, HH, XL, FD, MS, YX, TG, and DL analyzed and interpreted the patient data regarding the CD. All authors read and approved the final manuscript.

**Data availability**

Raw metabolomic data for this study are available at the MetaboLights metabolomics repository under study accession number MTBLS1753. Raw sequencing data are available at the European Nucleotide Archive server under study accession number PRJE8210.

**Funding**

This work was supported by the grants from the National Natural Science Foundation of China (grant number 81870373, 81900474, and 81900472), Shanghai Hospital Development Center New Frontier Technology Joint Research Project (grant number SHDC12017115), and Shanghai Science and Technology Committee (grant number 1914094300). The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

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