

Transient Suppression of Bacterial Populations Associated with Gut Health is Critical in Success of Exclusive Enteral Nutrition for Children with Crohn's Disease

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Abstract

Background and Aims: Exclusive enteral nutrition [EEN] is a dietary intervention to induce clinical remission in children with active luminal Crohn's disease [CD]. While changes in the gut microbial communities have been implicated in achieving this remission, a precise understanding of the role of microbial ecology in the restoration of gut homeostasis is lacking.

Methods: Here we reconstructed genomes from the gut metagenomes of 12 paediatric subjects who were sampled before, during and after EEN. We then classified each microbial population into distinct 'phenotypes' or patterns of response based on changes in their relative abundances throughout the therapy on a per-individual basis.

Results: Our data show that children achieving clinical remission during therapy were enriched with microbial populations that were either suppressed or that demonstrated a transient bloom as a function of EEN. In contrast, this ecosystem-level response was not observed in cases of EEN failure. Further analysis revealed that populations that were suppressed during EEN were significantly more prevalent in healthy children and adults across the globe compared with those that bloomed ephemerally during the therapy.

Conclusions: These observations taken together suggest that successful outcomes of EEN are marked by a temporary emergence of microbial populations that are rare in healthy individuals, and a concomitant reduction in microbes that are commonly associated with gut homeostasis. Our work is a first attempt to highlight individual-specific, complex environmental factors that influence microbial response in EEN. This model offers a novel, alternative viewpoint to traditional taxonomic strategies used to characterize associations with health and disease states.

Key Words: Gut microbiome; genome-resolved metagenomics; microbial therapeutics

1. Introduction

Crohn's disease [CD], a type of inflammatory bowel disease [IBD], can affect any part of the gastrointestinal tract. Despite novel medical therapies and improved disease surveillance strategies, a medical cure does not exist and the risk of progression to surgery remains high.¹ The incidence of CD continues to rise in countries across the globe,² and data suggest that CD outcomes may be worse for those diagnosed when children.³ The most effective medical therapies work through modification of the host immune system, which may result in sequelae of severe adverse effects and co-morbidities.^{4–6}

For this reason, diet-based therapies, which do not confer the same degree of risk, present an attractive alternative for patients and their families.⁷

One therapy, exclusive enteral nutrition [EEN], has proven efficacy for induction of clinical and endoscopic remission in paediatric subjects with CD. In fact, studies suggest that the rate of mucosal healing may be superior compared to other induction therapies such as systemic steroids.^{8–12}

Additionally, the side effect profile of EEN is limited, and growth, an essential consideration in paediatric CD, is not compromised by this strategy.¹³ The exact mechanism by which EEN works is not known. However, numerous studies

have described dramatic shifts in taxonomic and functional compositions of gut microbial communities during EEN,^{14–20} prompting researchers to suggest a role for the gut microbiome.²¹ Nevertheless, the focus on microbial taxonomy has not yielded a particular set of taxa that is consistently associated with health or disease states. Furthermore, while some studies have identified organisms such as *Faecalibacterium prausnitzii* to have anti-inflammatory properties in CD,²² other investigations of EEN in paediatric CD show that this organism decreases in abundance as a function of successful therapy.^{16,23} These contrasting observations reveal the need for deeper insights into the precise role of microbes in EEN-derived benefits.

A diverse gut microbial community is often associated with healthy gut homeostasis and host immunity,^{24,25} while ‘dysbiosis’ is linked to states of active gut inflammation.^{26,27} However, this wisdom contrasts with successful applications of EEN, where remission in CD is associated with decreasing diversity during EEN therapy and its rebound once a normal diet is resumed.^{14–17} To explain this phenomenon, some have postulated a superior ‘metabolic competence’ allowing certain bacterial populations to succeed in dysbiosis based on their ability to utilize a rapidly cycling carbohydrate input from the host.²⁸ When the diet is converted to a single, consistent carbohydrate this metabolic advantage is lost, creating an environment where other microbial residents may be more competitive.²⁸ This approach essentially shifts the focus from individual taxa to metabolic potential of microorganisms that survive EEN stress. If this is true, it is conceivable to expect that very closely related microbes with comparable metabolisms should respond similarly to EEN in different individuals. Given the interpersonal variation across human gut microbiomes, testing this hypothesis is difficult; however, using time-series data to characterize changes within each individual can shed light on microbiome dynamics as a response to therapy.

Here we apply genome-resolved metagenomics to characterize the response of microbial communities in the gut to EEN, without relying on taxonomy. From a previously established dataset of gut metagenomes from children with CD,²⁹ we reconstructed metagenome-assembled genomes [MAGs] from each individual independently and assigned each MAG a ‘phenotype’ based on their responses to changing diet as indicated by the changes of the relative abundance of each MAG throughout the therapy in a given individual. Our findings demonstrate that positive EEN outcomes, such as clinical remission, are associated with a substantial ecosystem response to the change in diet. Our metagenomic read recruitment analyses using publicly available gut metagenomes showed that microbes that bloomed in children who responded to EEN were typically rare among healthy children and adults, while those that were suppressed in these patients were broadly detected in gut metagenomes of healthy individuals.

2. Materials and Methods

2.1. Shotgun metagenomes

We downloaded gut metagenomes using the accession numbers provided in the original manuscript by Quince *et al.*²⁹ Our dataset included 69 metagenomes 56 of which were from patients who were diagnosed with CD and were sampled four to five times over the course of EEN therapy.²⁹ The characteristics of these children, including exposure to previous

therapy, disease phenotype and clinical response to EEN therapy, are described in the original work.²⁹ We also used publicly available global gut metagenomes to benchmark our findings, accession IDs of which are listed in [Supplementary Tables S1 and S2](#).

2.2. ‘Omics analyses

Whenever applicable, we automated and scaled our ‘omics analyses using the bioinformatics workflows implemented by the program ‘anvi-run-workflow’³⁰ in anvi’o.^{31,32} Anvi’o workflows implement numerous steps of bioinformatics tasks including short-read quality filtering, assembly, gene calling, functional annotation, hidden Markov model search, metagenomic read-recruitment and metagenomic binning. Workflows use Snakemake³³ and a tutorial is available at <http://merenlab.org/anvio-workflows/>.

2.3. Metagenomic assembly and processing of contigs

We individually co-assembled the metagenomes from the same individual in our cohort using IDBA-UD v1.1.3³⁴ with a minimum contig length of 1000 bp. We then converted the resulting FASTA files to anvi’o contigs databases using the program ‘anvi-gen-contigs-database’, which [1] identified open reading frames in assembled sequences using Prodigal v2.6.3,³⁵ [2] predicted gene-level taxonomy using Centrifuge,³⁶ [3] annotated functions by aligning genes to the NCBI Clusters of Orthologous Groups [COG] database³⁷ with DIAMOND v0.9.31,³⁸ and [4] identified single-copy core genes by using HMMER v3.2.1³⁹ to search for matches against a built-in collection of HMM profiles for bacteria and archaea in anvi’o. The average read recruitment rate of contigs back from the metagenomes from which they were assembled remained above 90% for each individual.

2.4. Read recruitment and binning

We used Bowtie2 v2.3.4.3⁴⁰ to recruit short reads from the set of metagenomes from each subject to the co-assembly of that subject’s metagenomes, and samtools⁴¹ to convert the SAM files into BAM files. We then used the programs ‘anvi-profile’ to compute coverage and detection statistics, and ‘anvi-merge’ to merge individual profiles, during which CONCOCT⁴² generated preliminary clusters of contigs based on their differential coverage and sequence composition. We then used the program ‘anvi-refine’ to interactively and manually refine these preliminary clusters as described before.³⁰ We used the program ‘anvi-rename-bins’ to determine bins that were over 70% complete as estimated by our bacterial single-copy core gene collection, assigned taxonomy to these MAGs using GTDB-Tk v0.3.2,⁴³ and used them for downstream analyses. To investigate the distribution of our genomes in publicly available data through metagenomic read recruitment, we used Bowtie2 v2.3.4.3⁴⁰ with default parameters and recovered coverage and detection statistics with anvi’o. We removed any bin from downstream analyses if it was less than 70% complete or more than 10% redundant after refinement based on 71 bacterial single-copy core genes.

2.5. Assignment of microbial population phenotypes

We assigned a ‘phenotype’ to each MAG from a single individual based on their distribution patterns across metagenomes from the same individual as revealed by read-recruitment.

These phenotypes were: [1] Wilt—a decrease in abundance by a factor of 10 from the pre-EEN metagenome to the post-EEN metagenome; [2] Suppressed—a decrease in abundance by a factor 10 in at least one metagenome during therapy that returns to pre-EEN abundance following therapy; [3] Ephemeral bloom—an increase in abundance by a factor of 10 seen in at least one time point during therapy that returns to pre-EEN abundance following therapy; [4] Sustained bloom—an increase in abundance following therapy by a factor of 10 compared with pre-EEN abundance estimates; [5] Stable—no changes in abundance of at least a factor of 10 at any time point throughout the time series; and [6] Unclear—changes by a factor of 10 between values but does not fit into any described pattern above.

2.6. Calculating population relative abundances through read recruitment

We obtained percentage abundance values for each MAG [in the patient it was binned from] by running ‘anvi-summarize’ on each patient’s co-assembly. We then normalized these values by dividing them by the maximum percentage abundance value of the MAG in any sample from its source individual.

2.7. Mapping recovered MAGs in metagenomes of healthy children, mother–infant dyads and adults across countries

Shotgun metagenome sequences of healthy children from the original analysis [$n = 13$],²⁹ publicly available mother–infant dyads [$n = 392$]⁴⁴ and healthy adults across six countries [$n = 939$] were mapped against recovered MAGs to ascertain detection values. These values were then used to generate the box plots in [Figure 3](#), and Wilcoxon signed-rank tests were used in each metagenome pool to determine the statistical difference between ‘ephemeral’ and ‘suppressed’ MAGs. For the healthy adult analysis, ‘ephemeral’ and ‘suppressed’ MAGs from subjects with clinical remission were included [$n = 83$], with three MAGs [two ‘suppressed’; one ‘ephemeral’] from subject G10 due to our initial definition of clinical remission achieved after resuming a regular diet.

2.8. Dereplicating genomes

We dereplicated the 261 MAGs using the program ‘anvi-dereplicate-genomes’, which used pyANI v0.2.9⁴⁵ to estimate the average nucleotide identity [ANI] between each pair of MAGs across all individuals. A pair of MAGs was considered redundant if their ANI was over 97% with the alignment length covering at least 50% of the shorter genome. From each genome cluster that included more than one genome, we selected as a representative genome the one with the highest ‘completion minus redundancy’ value based on the occurrence of bacterial single-copy core genes.

2.9. Genome-level analysis of microbial metabolism

We annotated each MAG with HMM hits to KOfam v94.0,⁴⁶ the database of KEGG Orthologs [KOs], using the program ‘anvi-run-kegg-kofams’, which uses HMMER³⁹ to match gene sequences to the KOfam HMM profiles. The program ‘anvi-run-kegg-kofams’ discarded any KOfam hit that had a bit-score below noise cutoffs set by the KEGG. We then used the program ‘anvi-estimate-metabolism’ to predict the metabolic capabilities of a given MAG. This program uses the KO

annotations to compute the percentage completion of each metabolic pathway in the KEGG MODULE database.⁴⁷ We used the scippy Python library for hierarchical clustering of MAGs according to their KEGG module completion scores.

2.10. Code and data availability

<https://doi.org/10.6084/m9.figshare.15108726.v1> gives reproducible access to the MAGs generated in this study. Metagenomes are available in the European Nucleotide Archive [PRJEB7576].

3. Results

3.1. Genome-resolved metagenomics yields 261 high-quality genomes from 12 children with CD

In our previous study,²⁹ 12 subjects were sampled four to five times before, during and after EEN therapy. Baseline demographic and clinical information, along with disease activity indices were collected and are summarized in [Table 1](#). To characterize the gut microbial ecology of each subject we reconstructed genomes from each individual independently [[Figure 1](#)] which enabled us to produce high-quality MAGs by applying differential coverage signals to minimize genome binning errors and risk of contamination.⁴⁸ Reconstructing genomes directly from gut metagenomes enables accurate tracking of microbial populations over time within a single individual⁴⁹ or across globally distributed human populations,⁵⁰ and offers high-resolution descriptions of ecology beyond marker genes alone.⁵¹

Instead of assembling each metagenome in our dataset and reconstructing genomes independently from each time point, here we elected to co-assemble metagenomes that belonged to the same individual [[Figure 1](#)]. A significant disadvantage of co-assemblies is the increased complexity of metagenomes due to remarkable inter-individual variability of the microbiome when data from multiple individuals are pooled. However, such risk is minimal for time-series data that cover relatively short time frames from a single individual. By contrast, the pooled reads can lead to the recovery of more complete genomes for populations that are rare in every time point. Furthermore, the recovery of genomes from single-assembled metagenomes can yield genomes that are heavily contaminated due to lack of differential coverage signal.⁴⁸ Following the co-assembly of each individual, we used sequence composition and differential coverage signal to cluster resulting contigs into a total of 589 metagenomic bins. We then removed any bins with less than 70% completion or greater than 10% redundancy, resulting in 261 high-quality genomes [[Supplementary Table S1](#)].

Overall, we recovered an average of 22 genomes from each subject that recruited more than 65% of the metagenomic reads [[Supplementary Table S1](#)]. While our average genomes per subject was below the average of 31 genomes per sample in a recent global survey of gut metagenomes,⁵² our high read recruitment rate indicates that we were able to capture a high-quality version of the most abundant MAGs in each subject. There was not a statistically significant difference in the average number of genomes we were able to reconstruct between those children achieving clinical remission and those who did not [25 vs 18 genomes, respectively, p value: 0.1]. The average number of metagenomic reads between samples from those achieving clinical remission did not significantly differ from one another [Wilcoxon, $p = 0.5$], which suggests

Table 1. Demographic data, clinical information and disease activity indices of all 12 paediatric subjects

Subject ID	Age [years]	Sex	Duration of EEN [days]	Crohn's phenotype	New diagnosis	Calprotectin start/end of EEN	PCDAI start/end of EEN	Crohn's medications	Antibiotic exposure
G01	8.4	Female	57	L3L4	Yes	1130/88	20/0	Lactulose [after EEN only]	No
G03	12.7	Female	56	L3L4	No	2076/39	27.5/5	5-ASA	No
G05	12.1	Male	56	L3L4	Yes	2262/2059	70/12.5	Azathioprine	No
G06	7.3	Male	56	L2	No	2102/718	37.5/5	Azathioprine	No
G08	14.6	Male	49	L3	Yes	2272/1686	57.5/10	Probiotic [after EEN only]	No
G09	12.3	Male	56	L3L4	No	2582/2461	25/15	Azathioprine, 5-ASA	No
G10	14.7	Female	40	L3L4	No	2188/2298	20/12.5	5-ASA, methotrexate, folic acid	No
G11	12.9	Male	56	L2L4	Yes	2439/1724	37.5/5	Azathioprine, iron, 5-ASA [after EEN only]	No
G12	13.5	Male	Unknown*	L2L4	Yes	2390/2717	50/37.5	Azathioprine [after EEN]	No
G15	11.5	Male	54	L2L4	Yes	2089/77	30/7.5	None	No
G21	13.9	Female	65	L3	Yes	2474/1865	55/50	None	Yes**
G33	9.2	Male	54	L3L4	Yes	3114/106	42.5/0	Iron, 5-ASA [end of EEN only]	No

Age and sex of the patient along with duration of EEN therapy and Crohn's phenotype based on Montreal classification are shown. Disease activity indices including faecal calprotectin measurements and PCDAI before and after EEN is reported. Medication exposure during therapy and use of antibiotics is also shown. PCDAI = paediatric Crohn's disease activity index; EEN = exclusive enteral nutrition; 5-ASA = 5-aminosalicylate.

*Minimum of 30 days.

**On day 30 of therapy.

that the depth of sequencing was relatively uniform among these samples [an average of ~8 million per sample and 40 million reads per individual via co-assembly used to reconstruct genomes]. However there was a positive correlation [$R^2 = 0.5$] between the number of genomes reconstructed per individual and the fraction of metagenomic reads they recruited [Supplementary Figure S3]. The taxonomy of the MAGs we have reconstructed from these metagenomes matched to the previous taxonomic characterization of these data via 16S rRNA gene amplicons and shotgun metagenomic reads, with 16 of the 20 most common taxa being shared at the family level [Supplementary Table S1].

3.2. Clinical remission in EEN therapy is associated with a rise and fall in microbial population abundances

Our previous analysis of these data using amplicon sequences noted an overall reduction in diversity in microbial community composition in individuals who received EEN therapy.²⁹ Here, instead of describing general patterns emerging from the pooled data as a function of taxonomy, we characterized changes in relative abundance of microbial populations on a host-by-host basis, independent of their taxonomy. For this, we classified the microbial genomes reconstructed from each subject into 'phenotypes' based on changes in their relative abundance throughout EEN. We assumed that the change in relative abundance was significant between two time points if the proportion of reads recruited by the genome differed between the two samples by more than an order of magnitude [see Methods for further detail].

While the relative abundance of some populations increased dramatically during EEN, the relative abundance of others decreased or did not change during therapy [Supplementary Table S1]. We divided these patterns of relative abundance into six 'phenotypes' [Figure 1]: [1] 'suppressed' [population abundance is transiently decreased only during EEN therapy], [2] 'ephemeral' [population abundance is transiently increased only during EEN therapy], [3] 'wilt' [population is abundant only before therapy], [4] 'bloom' [population is abundant only after therapy], [5] 'stable' [the abundance of a population does not fluctuate by more than an order of magnitude at any time point] and [6] 'unclear' [changes in population relative abundance throughout the study do not match to any of these patterns]. For downstream analyses we primarily focused on 92 genomes, of which 40 belonged to the 'suppressed' and 52 belonged to the 'ephemeral' categories, since they represent populations that show the most marked response to EEN therapy [Figure 2].

We observed a significant difference in the proportion of 'ephemeral' and 'suppressed' populations between subjects who achieved clinical remission during EEN (defined as Pediatric Crohn's Disease Activity Index [PCDAI] < 10) and those who did not [Wilcoxon non-parametric rank-sum test, $p = 0.01$] [Figure 2; Supplementary Table S1]. Subject G21 was excluded from this analysis due to insufficient time-series data needed to assess phenotype response.

'Suppressed' and 'ephemeral' populations made up on average 45% of all microbial populations in individuals who showed clinical remission, and 10.1% of the populations were recovered from individuals who did not [Figure 2; Supplementary Table S2].

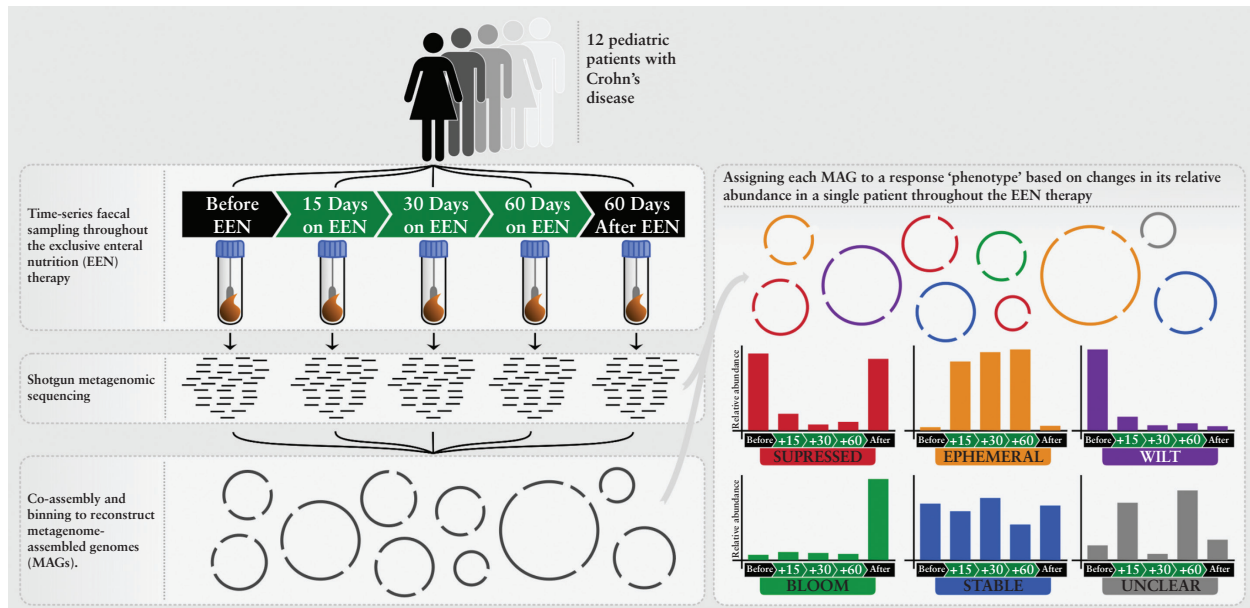


Figure 1. Overview of the sampling and analysis strategy, and the classification of genomes. Our metagenomic assembly and binning-based analysis of the shotgun sequencing of faecal metagenomes generated from 12 paediatric Crohn's disease patients who were sampled before, during and after EEN therapy resulted in microbial genomes reconstructed from each patient independently. We were then able to classify these genomes into distinct 'phenotypic' categories based on the changes in their relative abundance in a given individual over the course of EEN therapy.

3.3. Populations suppressed during EEN are common in healthy infants, children and adults

Next, we sought to understand whether the populations that were suppressed or those that demonstrated an ephemeral bloom during EEN occurred more frequently in healthy infants, age-matched children and healthy adults. First, we used metagenomic read recruitment to quantify the distribution of our genomes in publicly available gut metagenomes from 98 healthy infants who were sampled within days of birth, at 4 months and 1 year of life, and from 98 healthy women who gave birth to the infants.⁴⁴ This analysis revealed stark patterns of detection of our genomes based on phenotype assignment [Figure 3; Supplementary Table S2]. While we were unable to detect most populations in our dataset in infant gut samples that were collected during the first few months of life, there was a marked increase in detection from samples that were collected from the same set of infants as they reached 1 year of life. [Figure 3; Supplementary Table S2]. This increase in detection among infant metagenomes was largely limited to 'suppressed' populations while 'ephemeral' populations remained poorly detected in all infant age groups [Wilcoxon non-parametric rank-sum test, $p = 0.066$]. In the healthy mothers sampled in this dataset, differences in detection between suppressed and ephemeral populations were also significant [Wilcoxon non-parametric rank-sum test, $p = 2.6e-07$].

Next, we performed this analysis in 13 age-matched healthy children²⁹ where an even clearer difference in the detection of 'suppressed' and 'ephemeral' populations was appreciated [Wilcoxon non-parametric rank-sum test, $p = 7.8e-08$]. [Figure 3; Supplementary Table S2].

Finding that microbial populations suppressed during EEN were common in these healthy datasets, we then performed a more comprehensive read recruitment analysis that included 939 additional adult gut metagenomes from six countries [Austria, Denmark, England, Italy, Spain and the

USA]. This more robust analysis confirmed that populations that 'suppressed' during EEN were significantly more often detected in every country studied compared to those that showed an 'ephemeral' phenotype during EEN [Wilcoxon non-parametric rank-sum test, $p = 2.5e-0.6$] [Figure 3; Supplementary Table S2].

3.4. 'Identical' microbial populations in distinct individuals respond differently to EEN

Attempting to bypass the operational assumption of taxonomic assignment which postulates that populations that resolve to the same taxon are comparable across different individuals, or even different studies, we reconstructed genomes on an individual-by-individual basis. From this point, we sought to investigate whether 'identical' MAGs recovered from unique children with CD behaved similarly in response to EEN.

Here, we used a *de novo* clustering strategy that relies solely on whole-genome sequence similarity to identify sequence discrete clusters within the collection of 261 genomes we have reconstructed from 12 individuals. Our clustering analysis using a similarity cutoff of 98% genomic ANI [gANI] resulted in 123 clusters [Supplementary Table S3]. The term 'identical' here does not imply 100% identity, but rather it signifies that two populations have a gANI $\geq 98\%$; we note that this is more highly resolved than taxonomic labels often assumed to confer sameness across individuals and studies.

Our results show that eight clusters contained six or more genomes, meaning that a genome recovered from one child on EEN in the study was more than 98% identical to a genome recovered in at least five other children. In total, 56 clusters contained two or more genomes [recovered from two or more individuals independently] and 67 clusters contained a single genome [i.e. a genome reconstructed from an individual was not more than 98% identical to any other genome reconstructed from any of the individuals].

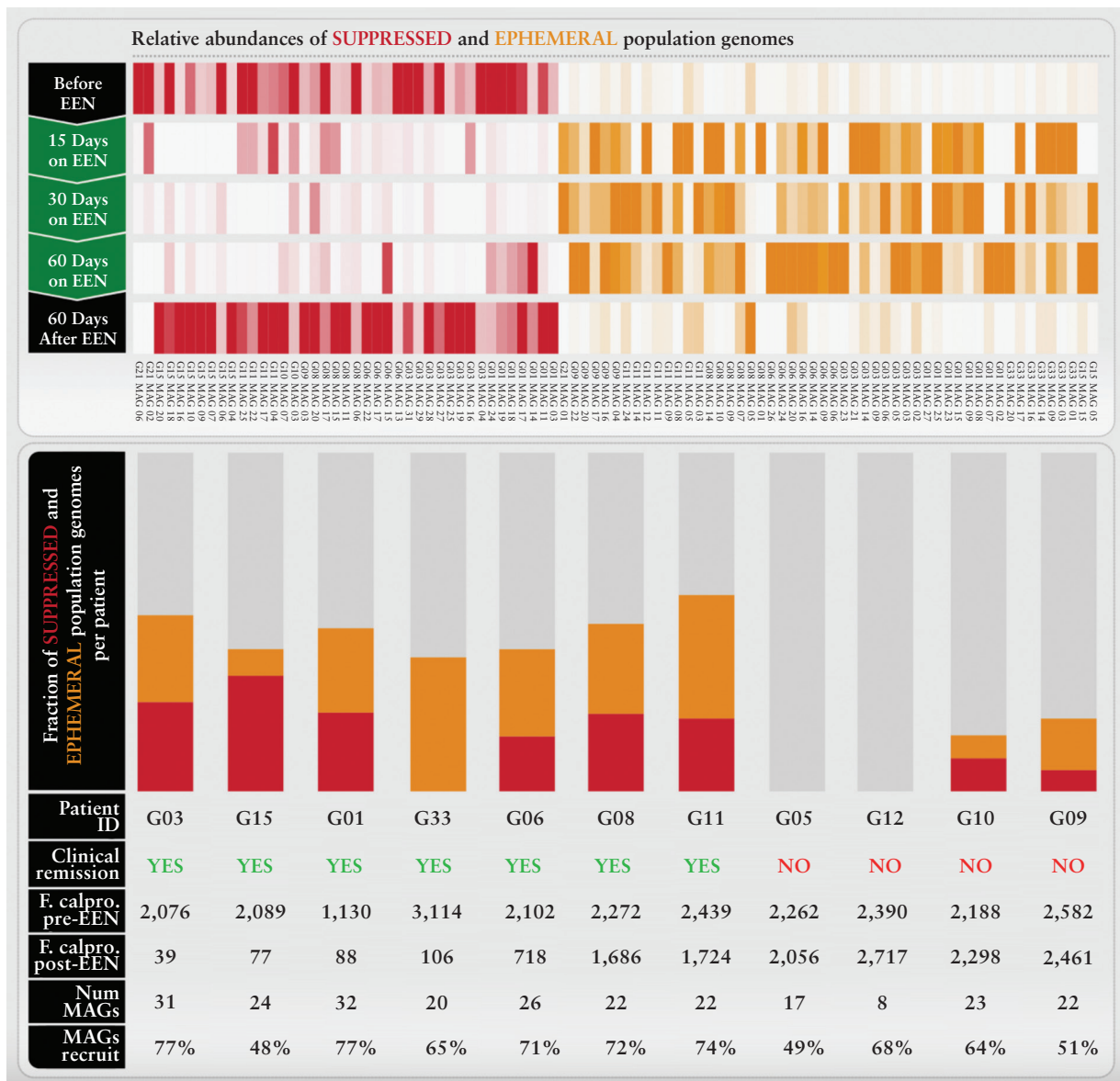


Figure 2. Genomes that showed 'suppressed' and 'ephemeral' phenotypes. Each column in the top panel shows a single genome that is either classified as 'suppressed' [red] or 'ephemeral' [orange], where the intensity of colours indicate its relative abundance across the EEN therapy. The bottom panel shows, for each individual: [1] the proportion of 'suppressed' and 'ephemeral' genomes recovered from each subject, where the grey colour represents the proportion of genomes in a given patient that were classified into other phenotypes [i.e. neither 'suppressed' nor 'ephemeral']; [2] clinical response/remission status [based on PCDAI]; [3] faecal calprotectin measurements before and after the EEN therapy; [4] the total number of genomes recovered; and [5] the percentage of metagenomic short reads MAGs recruit from the metagenomes from a given subject on average. EEN: exclusive enteral nutrition; FCalpro: faecal calprotectin; MAGs: metagenome-assembled genomes. The percentage of reads via both assembly and MAG is given in [Supplementary Figure S3](#).

Each cluster described taxonomically and metabolically homogeneous genomes. Specifically, each genome that occurred in any of the 56 clusters that contained one or more genomes resolved to the same taxon name [[Supplementary Table S3](#)], and the same hierarchical grouping of genomes based on KEGG metabolic modules [[Supplementary Figure 1](#), [Supplementary Table S3](#)].

To our surprise, we found that clusters were rarely composed of genomes that were assigned to the same phenotype [[Figure 4](#) highlights clusters with at least four genomes and [Supplementary Table S3](#) includes full results in 'gANI clusters']. In fact, genome phenotypes within a cluster diverged to the extent that in several instances a population that was 'suppressed' during EEN therapy in one child was 'ephemeral' in another

[[Figure 4](#)]. In total, there were 47 instances where an 'identical' genome was recovered in at least two subjects and did not have a uniform phenotype, and only in nine instances when all phenotypes within a cluster were the same [[Supplementary Table S3](#)]. Overall, the dramatic difference in the phenotypic response of identical populations to EEN therapy suggests that taxonomically identical organisms should not be expected to respond similarly in different individuals.

4. Discussion

Our study offers a new ecological perspective on individual-based microbial responses to EEN therapy, a topic that has

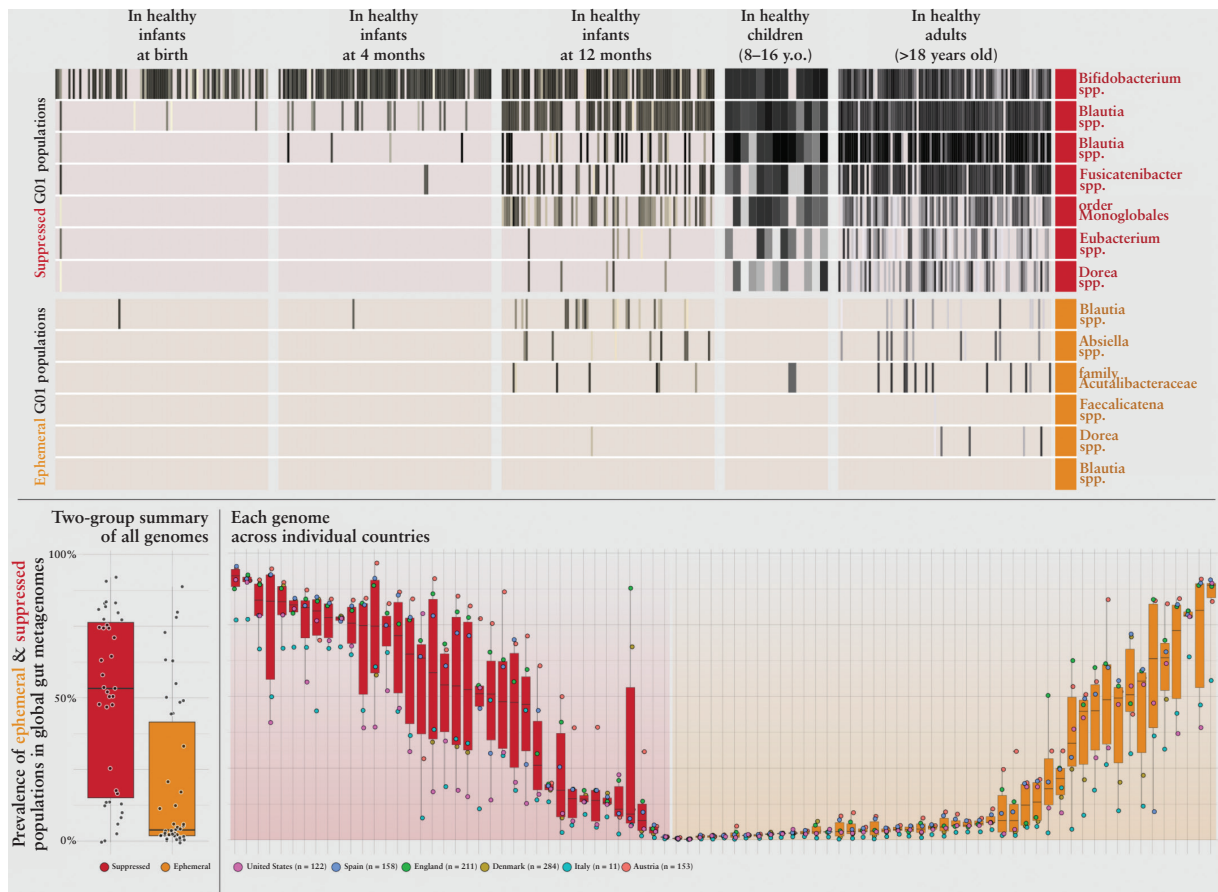


Figure 3. Distribution of ‘suppressed’ and ‘ephemeral’ populations in healthy individuals across age and geography. Each column in the heatmap in the top panel represents a metagenome from: [1] healthy infants at birth, at 4 months and at 12 months; [2] healthy age-matched children to our cohort of paediatric patients; and [3] healthy adults. Data points in this heatmap show the detection of a given G01 ‘suppressed’ [red] or ‘ephemeral’ [orange] genome in a given metagenome. The bottom panel shows the average detection values of all ‘suppressed’ [red] or ‘ephemeral’ [orange] genomes recovered from our patients in our study across 939 healthy adult gut metagenomes from six countries [box-plots on the left], and the detection values of each genome individually across the same set of metagenomes [box-plots on the right].

been traditionally investigated largely through cross-sectional taxonomic profiling of microbial communities. By examining changes in the relative abundance of microbial populations as a function of EEN in each patient, we observed dynamic, community-level shifts correlating with clinical remission in children with CD. Next, mapping to these populations with samples from healthy infants, children and adults, we observed that ‘suppressed’ populations were significantly more often detected than ‘ephemeral’ populations in healthy individuals. This conclusion is anchored in analysis from nearly 1000 healthy adult individuals and this signal is detected as early as 1 year of life, perhaps reflecting the transition to a more established gut constituency.⁵³ This observation gives greater understanding to the influence of EEN on the gut microbiota and raises important questions regarding our understanding of the determinants of successful microbial therapeutics in CD and beyond.

A critical mechanism of EEN’s efficacy is its ability to disrupt a microbial ecosystem, alter the inflammatory state of the gut microbiome and therefore temporarily change the mode of host–microbe interactions. This may be achieved by suppressing resident bacterial populations that are able to adapt and survive in the intestine of children with uncontrolled CD.

The transient emergence of microbes that are not primary members of healthy gut microbiomes and the temporary

suppression of commonly found microbes have implications on how to engineer new therapeutics for CD. While EEN has been primarily studied as a form of induction therapy, others have examined the long-term effects of partial enteral nutrition [around 50% of calories consumed], both as adjunctive therapy in maintenance of remission^{54,55} and as monotherapy.^{56,57} Additionally, new dietary therapies mimicking the carbohydrate, protein and fat composition of traditional EEN have been proposed as effective for both induction and maintenance of remission in children with CD.⁵⁸

Our findings suggest that EEN therapy probably functions less to restore homeostasis and acts more as an environmental stress to the gut community. As EEN temporarily pushes the gut composition away from its stable state, it may disrupt an injurious feedback loop associated with populations more fit to survive in the dysbiotic environment of CD. This concept of metabolic fitness is highlighted in other recent work demonstrating that individuals with IBD are typically depleted of less metabolically competent populations.⁵⁹ Our results characterize these ‘suppressed’ populations as abundant across healthy individuals, tolerant of dysbiosis and effectively disrupted by the stress conferred via EEN.

As such, while EEN offers a highly effective means of induction therapy in paediatric CD, from a purely microbial ecology standpoint its use as a maintenance therapy is less

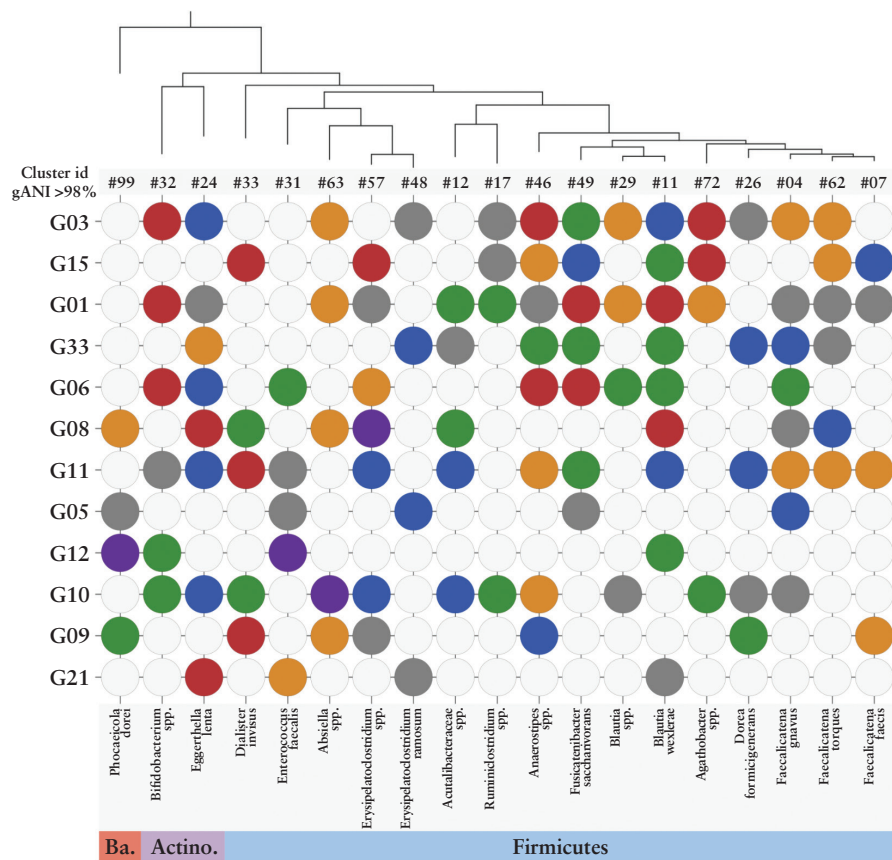


Figure 4. Phenotypes of genomes that are in the same 98% gANI cluster. Every column represents a cluster of genomes where every genome is >98% identical to others in the same cluster. The phylogenetic tree computed from ribosomal proteins of a representative genome from each cluster is shown at the top, organizing genomes based on their evolutionary relatedness. Every row represents a patient, and colours indicate the class of a genome in a given individual as defined in Figure 1, where colours correspond to populations that were suppressed [red], ephemeral [orange], wilt [purple], bloom [green], stable [blue] and unclear [dark grey]. For instance, if all genomes in a single column had the same colour, this would have meant that every near-identical genome had a comparable change in their relative abundance as a function of the EEN therapy in a given individual. The best matching species-level taxon name for a given cluster according to GTDB is shown below, and coloured bars represent bacterial phyla where Ba. stands for Bacteroidetes and Actino. stands for Actinobacteria. Dots in light grey indicate those individuals who did not contribute a genome to a given cluster.

consistent with our current understanding of gut homeostasis. Therefore, if EEN offers long-term benefit to gut health in paediatric CD as an uninterrupted therapy it is probably conveyed via another mechanism such as a change in antigen profile, correction of intestinal permeability or reduction of pro-inflammatory mediators.^{60,61}

However, one means of potentially optimizing the critical microbial disruption highlighted in this study would be to use EEN as an intermittent therapy. This concept is under investigation, examining the potential benefit of ‘cyclic’ EEN as a form of maintenance therapy.⁶² While further data are needed to substantiate this approach, conceptually it agrees with the findings presented here, maximizing the environmental stress of EEN and any benefit derived from a temporary change in the gut microbiota.

While consideration of the taxa present before, during and after EEN has been critical to our understanding of this important dietary intervention, attempting to catalogue bacterial populations as commensal or pathogenic in the metagenomes of children undergoing a microbial therapy may be of limited utility. Here, we observe that a microbe’s behaviour is not uniform across subjects, and in some cases is paradoxical. In other words, learning about a genome/taxon in the global context does not necessarily predict its behaviour in local

contexts. In this case, the sum total of microbes with predicted ‘suppressed’ or ‘ephemeral’ phenotypes before EEN therapy, given all the other members of a single gut community, may suggest that EEN will probably work to induce clinical remission in one individual and be less likely in another. This suggests that using existing gut microbiome data to help personalize therapy and predict outcomes may be more feasible than attempts to fundamentally and permanently change the community structure.

Recent efforts to better understand the probability of response to EEN have suggested that both taxonomic structure²¹ and the metabolome may be predictive of success.^{63,64} Our study offers an interesting association between the success of EEN and the enrichment of ‘suppressed’ and ‘ephemeral’ populations, which indicate a remarkable response to the changing environment within a single individual. Interestingly, we also observe that near-identical populations that are more than 98% identical at the genome level and that share the same species-level taxonomic classification did not respond to the EEN consistently across individuals in our cohort. Our efforts to investigate metabolic differences between such closely related genomes that resolved to the same gANI cluster of over 98% identity yet displayed different phenotypes across individuals did not yield additional insights [Supplementary

Information]. This may serve as evidence that the mechanisms driving microbial responses to change are far more complex than species-level taxonomy, and perhaps mediated by factors including genetic polymorphisms, variations at gene expression and/or metabolic output, and the presence, absence or activity of extrachromosomal genetic elements such as plasmids or viruses.

As with other datasets studying the influence of EEN on the gut microbiome, interpretation of our results is limited by a small sample size, heterogeneous patient population, and the concomitant management strategies applied before and during the study period. We chose to focus on PCDAI as a marker for remission in this work, consistent with the approach taken in the initial publication of this dataset. While decline in the level of calprotectin from baseline may reflect better adherence to the diet,⁶⁵ a recent meta-analysis of EEN studies suggests that PCDAI is also a reliable marker of response.⁶⁶ Here we argue that our host-centric approach offers key insight into the behaviour of microbial populations that are typically inaccessible through commonly used cross-sectional analyses in microbial ecology. However, we note that our phenotype assignment, which relies on a posteriori analyses of data, offers little power to predict the response of a given taxon to EEN. We also acknowledge that similar analyses that rely on ‘absolute abundance’ of populations rather than their ‘relative abundance’ may yield different classes of phenotypes and might support different interpretations. Furthermore, a higher percentage of unmapped reads from metagenomes to the co-assembly in individuals without a significant response to EEN could result in lower MAG recovery. However, the diminished impact of therapy on the gut microbiota of these children may also result in weaker differential coverage and a lower yield of MAGs. Even as sample size makes statistical analysis less applicable for some results, our analysis benefits from time-series sample collection, giving high resolution to the metagenomic-derived data and strength to the conclusions reached.

5. Conclusions

This highly resolved investigation of the gut community response in successful EEN therapy for paediatric CD reveals stereotypical phenotypes of bacterial populations under stress. We were able to validate these phenotypes by highlighting differences in distributions in those achieving clinical remission. Additionally, we demonstrate clear differences in global detection patterns of these phenotypes within healthy gut metagenomes. The ‘omics strategies we have employed were unable to explain why ‘identical’ genomes do not have a uniform phenotypic response to EEN, underlining the complexity of host–microbe interactions that call for host-centric, holobiont approaches to drive much needed therapeutics for CD.

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Conflict of Interest

JR, IV, AW, EF, QC, MY, AS, RV, CQ, KG and AME have no relevant disclosures. DTR has received grant support from Takeda; and has served as a consultant for Abbvie,

Abgenomics, Allergan Inc., Arena Pharmaceuticals, Bellatrix Pharmaceuticals, Boehringer Ingelheim Ltd, Bristol-Myers Squibb, Celgene Corp/Syneos, Check-cap, Dizal Pharmaceuticals, GalenPharma/Atlantica, Genentech/Roche, Gilead Sciences, Ichnos Sciences S.A., InDex Pharmaceuticals, Iterative Scopes, Janssen Pharmaceuticals, Lilly, Materia Prima, Narrow River Mgmt, Pfizer, Prometheus Laboratories, Reistone, Takeda and Techlab Inc. He is also co-founder of Cornerstones Health, Inc. and GoDuRn, LLC; on the Board of Trustees of the American College of Gastroenterology.

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Author Contributions

JR and AME conceived the study. CQ and KG recruited patients, collected clinical and demographic data, and performed metagenomic sequencing. JR, IV and AME analysed data. EF, ARW, QC and MY contributed to data analyses. IV, AS and AME developed research tools. RV, DTR and KG helped with clinical interpretations. JM, IV and AME wrote the paper with critical input from all authors.

Data Availability

<https://doi.org/10.6084/m9.figshare.15108726.v1> gives reproducible access to the metagenome-assembled genomes generated in this study. Metagenomes are available via available in the European Nucleotide Archive [PRJEB7576].

Supplementary Data

Supplementary data are available online at *ECCO-JCC* online.

References

1. Debruyjn JCC, Soon IS, Hubbard J, Wrobel I, Panaccione R, Kaplan GG. Nationwide temporal trends in incidence of hospitalization and surgical intestinal resection in pediatric inflammatory bowel diseases in the United States from 1997 to 2009. *Inflamm Bowel Dis* 2013;19:2423–32.
2. Sýkora J, Pomahačová R, Kreslová M, Cvalínová D, Štych P, Schwarz J. Current global trends in the incidence of pediatric-onset inflammatory bowel disease. *World J Gastroenterol* 2018;24:2741–63.
3. Olén O, Askling J, Sachs MC, *et al.* Increased mortality of patients with childhood-onset inflammatory bowel diseases, compared with the general population. *Gastroenterology* 2019;156:614–22.
4. Kotlyar DS, Osterman MT, Diamond RH, *et al.* A systematic review of factors that contribute to hepatosplenic T-cell lymphoma in patients with inflammatory bowel disease. *Clin Gastroenterol Hepatol* 2011;9:36–41.e1.

5. Ariyaratnam J, Subramanian V. Association between thiopurine use and nonmelanoma skin cancers in patients with inflammatory bowel disease: a meta-analysis. *Am J Gastroenterol* 2014;109:163–9.
6. Long MD, Martin CF, Pipkin CA, Herfarth HH, Sandler RS, Kappelman MD. Risk of melanoma and nonmelanoma skin cancer among patients with inflammatory bowel disease. *Gastroenterology* 2012;143:390–399.e1.
7. Svolos V, Gerasimidis K, Buchanan E, et al. Dietary treatment of Crohn's disease: perceptions of families with children treated by exclusive enteral nutrition, a questionnaire survey. *BMC Gastroenterol* 2017;17:14. doi:10.1186/s12876-016-0564-7.
8. Narula N, Dhillon A, Zhang D, Sherlock ME, Tondeur M, Zachos M. Enteral nutritional therapy for induction of remission in Crohn's disease. *Cochrane Database Syst Rev* Published online 2018. doi:10.1002/14651858.cd000542.pub3.
9. Buchanan E, Gaunt WW, Cardigan T, Garrick V, McGrogan P, Russell RK. The use of exclusive enteral nutrition for induction of remission in children with Crohn's disease demonstrates that disease phenotype does not influence clinical remission. *Aliment Pharmacol Ther* 2009;30:501–7.
10. Borrelli O, Cordischi L, Cirulli M, et al. Polymeric diet alone versus corticosteroids in the treatment of active pediatric Crohn's disease: a randomized controlled open-label trial. *Clin Gastroenterol Hepatol* 2006;4:744–53. doi:10.1016/j.cgh.2006.03.010.
11. Grover Z, Burgess C, Muir R, Reilly C, Lewindon PJ. Early mucosal healing with exclusive enteral nutrition is associated with improved outcomes in newly diagnosed children with luminal Crohn's disease. *J Crohns Colitis* 2016;10:1159–64.
12. Pigneur B, Lepage P, Mondot S, et al. Mucosal healing and bacterial composition in response to enteral nutrition vs steroid-based induction therapy—a randomised prospective clinical trial in children with Crohn's disease. *J Crohns Colitis* 2019;13:846–55.
13. Cohen-Dolev N, Sladek M, Hussey S, et al. Differences in outcomes over time with exclusive enteral nutrition compared with steroids in children with mild to moderate Crohn's disease: results from the GROWTH CD Study. *J Crohns Colitis* 2018;12:306–12.
14. Lionetti P, Callegari ML, Ferrari S, et al. Enteral nutrition and microflora in pediatric Crohn's disease. *JPEN J Parenter Enteral Nutr* 2005;29[4 Suppl]:S173–5; discussion S175.
15. Leach ST, Mitchell HM, Eng WR, Zhang L, Day AS. Sustained modulation of intestinal bacteria by exclusive enteral nutrition used to treat children with Crohn's disease. *Aliment Pharmacol Ther* 2008;28:724–33.
16. Gerasimidis K, Bertz M, Hanske L, et al. Decline in presumptively protective gut bacterial species and metabolites are paradoxically associated with disease improvement in pediatric Crohn's disease during enteral nutrition. *Inflamm Bowel Dis* 2014;20:861–71.
17. Dunn KA, Moore-Connors J, MacIntyre B, et al. Early changes in microbial community structure are associated with sustained remission after nutritional treatment of pediatric Crohn's disease. *Inflamm Bowel Dis* 2016;22:2853–62.
18. Kaakoush NO, Day AS, Leach ST, Lemberg DA, Nielsen S, Mitchell HM. Effect of exclusive enteral nutrition on the microbiota of children with newly diagnosed Crohn's disease. *Clin Transl Gastroenterol* 2015;6:e71.
19. D'Argenio V, Precone V, Casaburi G, et al. An altered gut microbiome profile in a child affected by Crohn's disease normalized after nutritional therapy. *Am J Gastroenterol* 2013;108:851–2.
20. Schwerdt T, Frivolt K, Clavel T, et al. Exclusive enteral nutrition in active pediatric Crohn disease: effects on intestinal microbiota and immune regulation. *J Allergy Clin Immunol* 2016;138:592–6.
21. Jones CMA, Connors J, Dunn KA, et al. Bacterial taxa and functions are predictive of sustained remission following exclusive enteral nutrition in pediatric Crohn's disease. *Inflamm Bowel Dis* 2020;26:1026–37. doi:10.1093/ibd/izaa001.
22. Sokol H, Pigneur B, Watterlot L, et al. *Faecalibacterium prausnitzii* is an anti-inflammatory commensal bacterium identified by gut microbiota analysis of Crohn disease patients. *Proc Natl Acad Sci U S A* 2008;105:16731–6.
23. Jia W, Whitehead RN, Griffiths L, et al. Is the abundance of *Faecalibacterium prausnitzii* relevant to Crohn's disease? *FEMS Microbiol Lett* 2010;310:138–44.
24. Human Microbiome Project Consortium. Structure, function and diversity of the healthy human microbiome. *Nature* 2012;486:207–14.
25. Bäckhed F, Fraser CM, Ringel Y, et al. Defining a healthy human gut microbiome: current concepts, future directions, and clinical applications. *Cell Host Microbe* 2012;12:611–22.
26. Imhann F, Vich Vila A, Bonder MJ, et al. Interplay of host genetics and gut microbiota underlying the onset and clinical presentation of inflammatory bowel disease. *Gut* 2018;67:108–19.
27. Gevers D, Kugathasan S, Denson LA, et al. The treatment-naive microbiome in new-onset Crohn's disease. *Cell Host Microbe* 2014;15:382–92.
28. Britto S, Kellermayer R. Carbohydrate monotony as protection and treatment for inflammatory bowel disease. *J Crohns Colitis* 2019;13:942–8.
29. Quince C, Ijaz UZ, Loman N, et al. Extensive modulation of the fecal metagenome in children with Crohn's disease during exclusive enteral nutrition. *Am J Gastroenterol* 2015;110:1718–29. doi:10.1038/ajg.2015.357.
30. Shaiber A, Willis AD, Delmont TO, et al. Functional and genetic markers of niche partitioning among enigmatic members of the human oral microbiome. *Genome Biol* 2020;21:292.
31. Eren AM, Esen OC, Quince C, et al. Anvi'o: an advanced analysis and visualization platform for 'omics data. *PeerJ* 2015;3:e1319.
32. Eren AM, Kiehl E, Shaiber A, et al. Community-led, integrated, reproducible multi-omics with anvi'o. *Nat Microbiol* 2021;6:3–6.
33. Köster J, Rahmann S. Snakemake—a scalable bioinformatics workflow engine. *Bioinformatics* 2012;28:2520–2.
34. Peng Y, Leung HCM, Yiu SM, Chin FYL. IDBA-UD: a de novo assembler for single-cell and metagenomic sequencing data with highly uneven depth. *Bioinformatics* 2012;28:1420–8.
35. Hyatt D, Chen GL, Locascio PF, Land ML, Larimer FW, Hauser LJ. Prodigal: prokaryotic gene recognition and translation initiation site identification. *BMC Bioinf* 2010;11:119.
36. Kim D, Song L, Breitwieser FP, Salzberg SL. Centrifuge: rapid and sensitive classification of metagenomic sequences. *Genome Res* 2016;26:1721–9.
37. Tatusov RL, Fedorova ND, Jackson JD, et al. The COG database: an updated version includes eukaryotes. *BMC Bioinf* 2003;4:41.
38. Buchfink B, Xie C, Huson DH. Fast and sensitive protein alignment using DIAMOND. *Nat Methods* 2015;12:59–60.
39. Eddy SR. Accelerated profile HMM searches. *PLoS Comput Biol* 2011;7:e1002195.
40. Langmead B, Salzberg SL. Fast gapped-read alignment with Bowtie 2. *Nat Methods* 2012;9:357–9.
41. Li H, Handsaker B, Wysoker A, et al; 1000 Genome Project Data Processing Subgroup. The sequence alignment/map format and SAMtools. *Bioinformatics* 2009;25:2078–9.
42. Alneberg J, Bjarnason B, de Bruijn I, et al. Binning metagenomic contigs by coverage and composition. *Nat Methods* 2014;11:1144–1146. doi:10.1038/nmeth.3103.
43. Chaumeil PA, Mussig AJ, Hugenholtz P, Parks DH. GTDB-Tk: a toolkit to classify genomes with the Genome Taxonomy Database. *Bioinformatics* Published online November 15, 2019. doi:10.1093/bioinformatics/btz848.
44. Bäckhed F, Roswall J, Peng Y, et al. Dynamics and stabilization of the human gut microbiome during the first year of life. *Cell Host Microbe* 2015;17:852.
45. Pritchard L, Glover RH, Humphris S, Elphinstone JG, Toth IK. Genomics and taxonomy in diagnostics for food security: soft-rotting enterobacterial plant pathogens. *Anal Methods* 2016;8:12–24.
46. Aramaki T, Blanc-Mathieu R, Endo H, et al. KofamKOALA: KEGG Ortholog assignment based on profile HMM and adaptive score threshold. *Bioinformatics* 2020;36:2251–2.

47. Muto A, Kotera M, Tokimatsu T, Nakagawa Z, Goto S, Kanehisa M. Modular architecture of metabolic pathways revealed by conserved sequences of reactions. *J Chem Inf Model* 2013;53:613–22.
48. Chen LX, Anantharaman K, Shaiber A, Eren AM, Banfield JF. Accurate and complete genomes from metagenomes. *Genome Res* 2020;30:315–33.
49. Sharon I, Morowitz MJ, Thomas BC, Costello EK, Relman DA, Banfield JF. Time series community genomics analysis reveals rapid shifts in bacterial species, strains, and phage during infant gut colonization. *Genome Res* 2013;23:111–20.
50. Lee STM, Kahn SA, Delmont TO, et al. Tracking microbial colonization in fecal microbiota transplantation experiments via genome-resolved metagenomics. *Microbiome* 2017;5:50.
51. Quince C, Walker AW, Simpson JT, Loman NJ, Segata N. Shotgun metagenomics, from sampling to analysis. *Nat Biotechnol* 2017;35:833–44.
52. Pasolli E, Asnicar F, Manara S, et al. Extensive unexplored human microbiome diversity revealed by over 150,000 genomes from metagenomes spanning age, geography, and lifestyle. *Cell* 2019;176:649–662.e20.
53. Avershina E, Lundgård K, Sekelja M, et al. Transition from infant-to adult-like gut microbiota. *Environ Microbiol* 2016;18:2226–36.
54. Tanaka T, Takahama K, Kimura T, et al. Effect of concurrent elemental diet on infliximab treatment for Crohn's disease. *J Gastroenterol Hepatol* 2006;21:1143–9.
55. Yamamoto T, Nakahigashi M, Umegae S, Matsumoto K. Enteral nutrition for the maintenance of remission in Crohn's disease: a systematic review. *Eur J Gastroenterol Hepatol* 2010;22:1–8.
56. Takagi S, Utsunomiya K, Kuriyama S, et al. Effectiveness of an 'half elemental diet' as maintenance therapy for Crohn's disease: a randomized-controlled trial. *Aliment Pharmacol Ther* 2006;24:1333–40. doi:10.1111/j.1365-2036.2006.03120.x.
57. Davidson N, Rutsky J, Bricker JB, et al. 6 and 12-Month outcomes after 90:10 enteral nutrition induction therapy in pediatric Crohn's disease. *J Pediatr Gastroenterol Nutr* 2022;75:70–75. doi:10.1097/MPG.0000000000003475.
58. Svolos V, Hansen R, Nichols B, et al. Treatment of active Crohn's disease with an ordinary food-based diet that replicates exclusive enteral nutrition. *Gastroenterology* 2019;156:1354–1367.e6.
59. Watson AR, Füssel J, Veseli I, et al. Adaptive ecological processes and metabolic independence drive microbial colonization and resilience in the human gut. *bioRxiv* Published online April 22, 2021:2021.03.02.433653. doi:10.1101/2021.03.02.433653.
60. Johnson T, Macdonald S, Hill SM, Thomas A, Murphy MS. Treatment of active Crohn's disease in children using partial enteral nutrition with liquid formula: a randomised controlled trial. *Gut* 2006;55:356–61.
61. Levine A, Sigall Boneh R, Wine E. Evolving role of diet in the pathogenesis and treatment of inflammatory bowel diseases. *Gut* 2018;67:1726–38.
62. Pigneur Arnaud B, Martinez-Vinson C, Bourmaud A, et al. Cyclic exclusive enteral nutrition to maintain longterm drug-free remission in Paediatric Crohn's Disease: The CD HOPE study of the GETAID pédiatrique. *J Crohns Colitis* 2021;15[Supplement_1]:S015–S015.
63. Diederer K, Li JV, Donachie GE, et al. Exclusive enteral nutrition mediates gut microbial and metabolic changes that are associated with remission in children with Crohn's disease. *Sci Rep* 2020;10. doi:10.1038/s41598-020-75306-z.
64. Ghiboub M, Penny S, Verburgt CM, et al. Metabolome changes with diet-induced remission in pediatric Crohn's disease. *Gastroenterology* Published online June 6, 2022. doi:10.1053/j.gastro.2022.05.050.
65. Mckirdy S, Russell RK, Svolos V, et al. The impact of compliance during exclusive Enteral nutrition on faecal Calprotectin in children with Crohn's disease. *J Pediatr Gastroenterol Nutr* Published online February 22, 2022. doi:10.1097/MPG.0000000000003425.
66. Sohoul MH, Fatahi S, Farahmand F, Alimadadi H, Seraj SS, Rohani PM. Efficacy of exclusive enteral nutrition as induction therapy on disease activity index, inflammation and growth factors in paediatric Crohn's disease. *Aliment Pharmacol Ther* Published online June 24, 2022. doi:10.1111/apt.17109.