Supplementary Materials for "Characterization of the Clinical Value of Alpha-diversity metrics in Microbiome Studies"

Nataša Mortvanski^{1,2,*}, José Luis Villanueva-Cañas^{2,3} and Climent Casals-Pascual^{1,4,5}

¹Barcelona Institute for Global Health (ISGlobal), Barcelona, Spain.

²Pompeu Fabra University (UPF), Barcelona, Spain.

³Molecular Biology CORE (CDB), Hospital Clínic of Barcelona, Barcelona, Spain.

⁴Department of Clinical Microbiology, Hospital Clínic of Barcelona, Barcelona, Spain.

⁵University of Barcelona (UB), Barcelona, Spain.

*Corresponding author: natasa.mortvanski01@estudiant.upf.edu

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1 Supplementary Notes

1.1 Data processing in QIIME2

All QIIME2 artefacts analysed on Qiita platform went through the same standardised preprocessing procedure. Trimmed to a length of 100bp, quality control was done by Deblur 2021.09, Greengenes database was used as reference phylogeny. In order to avoid doing data preparation for all these studies from scratch, we decided to keep working with these artefacts.

However, Deblur is not the most compatible with analysing paired-end sequencing data, such as data obtained from Hospital Clínic (paired reads need to be joined before denoising). In this case it was easier to use the DADA2 pipeline for quality control. CDI data from BioProject database (Khanna et al. 2016) on the other hand is produced by single-end sequencing technology. Since it was being compared with American Gut Project data (McDonald et al. 2018) and other CDI datasets (Weingarden et al. 2015, Khanna et al. 2017) obtained from Qiita, we wanted to analyse these datasets in the same way. That is why we processed it using Deblur.

Pre-fitted sklearn-based taxonomy classifier scikit-learn_0.24.1 was used for taxonomy assignment on Qiita platform (the most recent update of QIIME2 - qiime2 2022.2.1), while scikit-learn_0.23.1 was used for Khanna et al. (2016) CDI dataset and Hospital Clínic's data (analysed on local machine with QIIME2 version 2020.8.0).

1.2 Overview of different alpha diversity metrics

We conducted a literature search to obtain the definitions of different alpha metrics that were used in the analysis. There are two main aspects on which those indices are based, namely richness and evenness. Some of the indices are based on a combination of both. Below are the results sorted accordingly.

1.2.1 Richness metrics

Richness indices estimate the number of different species in a sample. The simplest measure for richness is the number of species or Operational Taxonomic Units (OTU). However, simply counting the number of present species is strongly affected by the bias introduced by undersampling and sequencing. This bias is even worse when the species evenness is low. There are numerous different metrics that are trying to capture or estimate richness. Here are some of the metrics that can be calculated using QIIME2 that we selected to use in our analysis:

Chao1 index is a nonparametric estimator of species richness that is correcting the observed richness for the number of lost species, estimated considering the distribution of the rarest species (Bent 2008, Finotello 2018). Chao's index for estimation of species richness is given by the equation (Thukral 2017, Website: CD Genomics):

$$S_{(max)Chao} = S_{obs} + (a^2 + b^2)$$

, where Smax = maximum no. of species, Sobs = number of species observed in different samples, a = singletons (number of species represented by one individual each), b = doubletons (number of species represented by two individuals each).

Margalef's index measures the species richness in a given area or community. It is defined as:

$$R_{MAR} = \frac{S-1}{\ln N}$$

, where, S is the total number of species and N is the total number of individuals in the sample (Thukral 2017).

Menhinick's index is defined as the ratio of the total number of species (S) to square root of number of individuals in the sample (N) (Thukral 2017):

$$R_{MEN} = \frac{S}{\sqrt{N}}$$

Fisher alpha is measuring the relationship between the number of species and the relative abundance of each species (Finotello 2018). This index is based upon the logarithmic distribution of number of individuals of different species:

$$S = \alpha ln(1 + \frac{N}{\alpha})$$

where, S is the total number of species and N is the total number of individuals in the sample. The value of Fisher's alpha is computed by iteration (Thukral 2017).

Faith's phylogenetic diversity is the sum of OTU branch lengths. It takes into account phylogenetic distance between OTUs. The greater the number of unique, phylogenetically more distant OTUs, the higher this index will be (Finotello 2018).

1.2.2 Evenness metrics

Evenness indices measure how evenly the relative abundances are distributed across the different species. Besides being a valuable indicator of biodiversity, evenness also determines the stability and resilience of an ecosystem. Some indices estimate unevenness or dominance, which is complementary to evenness.

Gini index (Bendel 1989, Zheng 2008) is also defined in reference to the Lorenz curve which results from a plot of the cumulative proportion of the population to the cumulative proportion of the variable. The Gini coefficient can be, as in the figure, defined geometrically as the ratio of two geometrical areas in the unit box: (a) the area between the line of perfect equality (45 degree line in the unit box) and the Lorenz curve, which is called area A and (b) the area under the 45 degree line, or areas A + B. Because areas A + B represents the half of the unit box, that is, A+B = 1/2, the Gini Coefficient, G, can be written as:

$$G = \frac{A}{A+B} = 2A = 1 - 2B$$

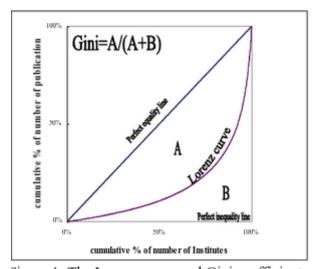


Figure 1: The Lorenz curve and Gini coefficient

Therefore a Gini coefficient is a number between 0 and 1 that measures the degree of inequality with 0 being maximum equality and 1 being maximum inequality.

Strong's index (Strong 2002) assesses species relative dominance concentration. It was, in part, based on the Gini index and Lorenz curve or partial order approach, but without the need to calculate area. It is defined as:

$$D_{W} = \max_{i} [(b_{i}/Q) - i/R]$$

, where R = the number of species in the sample, i = the i-th species in the data set (i = 1 through R), b = the sequential cumulative totaling of i-th species abundance values (a_k) ranked largest to smallest (i.e., b = largest a_k , b = b + second largest a_k , b = b + third largest a_k ...), Q = sum of species abundance values (Σa_k) , where k = 1 through R;

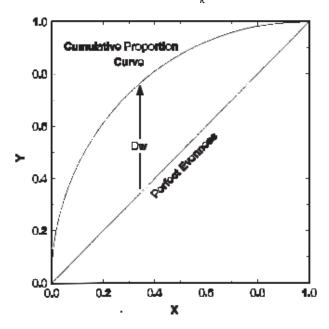


Figure 1. A cumulative proportion graph with an example of a dominance (D_W) measure. The x-axis (i/R) and y-axis (b_i/Q) of the diagram represent the right and left halves of the D_W equation. See Methods for definitions of individual parameters.

Pielou evenness is calculated as the ratio of the observed diversity to the maximum possible diversity having the same number of species (Pielou 1966). The formula is:

$$J' = H'/H'max$$

It has Shannon's formula in both numerator and denominator. H'max = logS, where S is the number of species.

1.2.3 Both richness and evenness

Shannon's index (Shannon–Weaver index, Shannon entropy) (Thukral 2017) was originally developed for communication systems and is based on information theory, however, it can also be used to define the biological diversity of the communities. The information content, H, therefore can be written as a function of probability:

$$H' = -\sum \frac{n_i}{N} ln \frac{n_i}{N} = -\sum p_i ln p_i$$

, where a message consists of N number of alphabets (number of species), pi is the probability of each letter in a message consisting of m alphabets (species), ni is the number of individuals of the i th letter (specie).

The minimum value of H' is 0 when all the individuals in the sample belong to the same species. This community has minimal redundancy and therefore maximum entropy (Bent 2008). This index is reaching the maximum if all the species in the sample are represented by equal number of individuals:

$$H'max = logS$$

Simpson's index reflects the probability that any two organisms sampled will be the same phylotype by capturing both richness and relative abundance (Finotello 2018). If there are k species consisting of n individuals, distributed among different species as n1, n2, n3, ...nk, then the probability (p1) of the first individual belonging to a species will be:

$$p_i = \frac{n_i}{n}$$

Probability (p1,2) that the second individual drawn from the sample without replacement also belongs to the same species will be:

$$p_{1,2} = (\frac{n_1}{n} * \frac{n_1 - 1}{n - 1})$$

The sum of the probabilities for all the species is a measure of the concentration (or abundance) (C) of the species:

$$C_{Simpson} = \left(\frac{\sum n_1(n_i - 1)}{n(n - 1)}\right)$$

If the sample size is large, then the probability (p1,2) that the second individual drawn from the sample with replacement also belongs to the same species (C') will be:

$$C_{Simpson} = \sum p_i^2 = \frac{\sum n_i^2}{n^2}$$

The maximum value of Simpson's concentration is 1 when all the individuals in the sample belong to the same species. It follows logarithmic distribution which means that similar increments on the assessment scale do not represent equal changes in dominance concentration.

1.3 Study design

3.1 AGP characterisation

filters:

- 20-69 years old
- 18,5 < BMI < 25
- no reported IBD, IBS, CDI
- no antibiotics used

3.2.1 IBD samples and control

Comparisons:

- IBD and UC dataset vs AGP dataset
- Longitudinal CD (cases vs control)

3.2.2 CDI samples and control

Comparisons:

- Bio Project CDI vs AGP
- Longitudinal CDI dataset during time after FMT
- CDI and IBD dataset (different combinations of conditions)

3.2.3 Hospital Clínic CDI vs control

Comaprisons:

- Hospital Clínic's dataset (difference between pre-FMT, post-FMT, donors)
- Hospital Clínic's vs AGP dataset

Datasets used for analysis



AGP dataset (McDonald et al. 2018) (n = 1470)

B

IBD dataset (Lloyd-Price et al. 2019) (26 CD, 7 UC)

UC dataset (Qiita ID 11549) (n = 33)

Longitudinal CD dataset (Vázquez-Baeza et al. 2018) (293 CD, 353 control)

(E)

Longitudinal CDI dataset (Weingarden et al. 2015) (n = 92)

F

Bio Project CDI dataset (Khanna et al. 2016) (n = 73)

G

CDI and IBD dataset (Khanna et al. 2017) (27 CDI, 6 CDI+CD, 6 CDI+UC, 1 donor)

Hospital Clínic's dataset (Aira et al. 2022) (38 pre-FMT, 18 post-FMT, 151 donors)

3.3 Statistical power analysis

Wilcox statistical power for difference between healthy and unhealthy samples:

- healthy (n =1823) → A + D controls
- unhealthy (n =432) → B © D

3.4 Random forest classification

All datasets (except from (E) (G) (H)):

train → 200 healthy, 204 unboolthy.

 $train \rightarrow 290$ healthy, 304 unhealthy test \rightarrow 546 healthy, 128 unhealthy

- IBD and healthy:
 - ABCD

train \rightarrow 236 healthy, 225 CD, 24 UC test \rightarrow 546 healthy, 94 CD, 16 UC

- CDI and healthy:
 - (A)(F)

train \rightarrow 41 healthy, 46 CDI test \rightarrow 546 healthy, 27 CDI

- Hospital Clínic:
 - H

train \rightarrow 15 pre-FMT, 77 donors test \rightarrow 5 pre-FMT, 36 donors

3.5 Modified t-test

Model 1 (AF):

Control \rightarrow 70% of AGP dataset Test \rightarrow 437 AGP, 73 CDI

Model 2 (H):

Control \rightarrow 77 donors Test \rightarrow 36 donors, 18 pre-FMT, 36 post-FMT

2 Supplementary Tables

| Reference | Qiita ID | Study title | 16S reg. | Raref. depth | N (used in analysis) | Meta data | Country | Techn. |
|--|-------------|--|-------------|-----------------|--|--------------|--------------------|---|
| McDonald et al. 2018 | 10317 | American Gut: an Open Platform for Citizen Science Microbiome Research | V4 | 5000 | 1470 (healthy) | Yes | America/ Europe | Illumina MiSeq |
| Lloyd-Price et al. 2019 | 11484 | Multi-omics of the gut microbial ecosystem in inflammatory bowel diseases | V4 | 1000 | 33 (26 CD, 7 UC) | Yes | America | Illumina MiSeq HiSeq2000 or 2500 2x101 |
| No publication (PI: Robert Knight) (Website: Qiita) | 11549 | Metaomics Reveals Microbiome Based Proteolysis as a Driver of Ulcerative Colitis Severity | V4 | 4500 | 33 (UC) | Yes | America (UCSD) | Illumina MiSeq |
| Vázquez-Baeza et al. 2018 | 2538 | Guiding longitudinal sampling in IBD cohorts | V4 | 7000 | 646 (293 CD, 353 control) | Yes | America (UNC) | Illumina HiSeq 2000 |
| Weingarden et al. 2015 | 1924 | Dynamic changes in short- and long-term bacterial composition following fecal microbiota transplantation for recurrent Clostridium difficile infection | V4 | 15000 | 92 (CDI) | Yes | America (UMN) | Illumina MiSeq 2 × 150 bp |
| Khanna <i>et al</i> . 2017 | 10057 | Changes in microbial ecology after fecal microbiota transplantation for recurrent C. difficile infection affected by underlying inflammatory bowel disease | V4 | 30000 | 40 (27 CDI, 6 CDI+CD, 6 CDI+UC, 1 donor) | Yes | America | Illumina MiSeq |

Supplementary Table S1. Properties of selected studies from Qiita repository. All artefacts from Qiita had the same preprocessing steps (*Deblur 2021.09* (*Reference phylogeny for SEPP: Greengenes_13.8, BIOM: all.biom*) | *Trimming (length: 100)*)

| Study | Study Accession | 16S reg. | N (used in analysis) | Rarefact. depth | Metadata | Country | Techn. |
|------------------------------|--------------------|----------|----------------------------|--------------------|----------|---------|---------------------------------|
| Khanna <i>et al.</i> 2016 | PRJNA34234 7 | V4 | 73 (CDI) | 6260 | Yes | America | Illumina MiSeq (Single-Read) |

Supplementary Table S2. Properties of selected study from BioProject repository.

| Dataset | 16S region | n (Pre-FMT) | n (Post-FMT) | n (donors) | Rarefaction depth | Technology |
|-----------------|------------|-------------|--------------|------------|-------------------|----------------|
| CDI samples | V3-V4 (?) | 38 | 18 | 38 | 5500 | Illumina MiSeq |
| Catalan biobank | V3-V4 (?) | / | 1 | 113 | 15000 | Illumina MiSeq |

Supplementary Table S3. Properties of data from Hospital Clínic.

| metric | statistic | p.value | skewness | kurtosis |
|-----------------|-----------|----------|----------|----------|
| Faith PD | 0.9937 | 4.46e-12 | 0.2673 | 2.7841 |
| Margalef | 0.9915 | 1.83e-07 | 0.3456 | 2.9911 |
| Menhinick | 0.9915 | 1.83e-07 | 0.3456 | 2.9911 |
| Chao1 | 0.9831 | 1.51e-15 | 0.5375 | 3.3656 |
| Fisher alpha | 0.9742 | 7.79e-06 | 0.6494 | 3.4489 |
| Gini index | 0.9604 | 1.38e-19 | -0.7684 | 3.6214 |
| Strong | 0.9559 | 1.21e-20 | 0.8152 | 3.4930 |
| Shannon entropy | 0.9454 | 1.14e-30 | -0.8837 | 3.4804 |
| Pielou evenness | 0.8938 | 6.89e-23 | -1.2674 | 4.4065 |
| Simpson | 0.7173 | 2.53e-44 | -2.2548 | 8.2933 |

Supplementary Table S4. American Gut Project data alpha metrics' distribution statistics. Normality test (Shapiro-Wilk), skewness and kurtosis of different metrics

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|--------|---------|----------|------------|
| Chao1 | CD | healthy | 2.10e-18 | 3.16e-17 |
| Margalef | CD | healthy | 3.15e-09 | 1.57e-08 |
| Faith PD | CD | healthy | 4.41e-08 | 1.70e-07 |
| Gini index | CD | healthy | 9.39e-08 | 2.82e-07 |
| Strong | CD | healthy | 4.32e-06 | 1.08e-05 |
| Fisher alpha | CD | healthy | 0.00007 | 0.00015 |
| Pielou evenness | CD | healthy | 0.00831 | 0.01558 |
| Menhinick | CD | healthy | 0.07199 | 0.10798 |
| Shannon entropy | CD | healthy | 0.14985 | 0.20434 |
| Simpson | CD | healthy | 0.75357 | 0.76939 |

Supplementary Table S5. Results of the Mann-Whitney-Wilcoxon test for difference in means of healthy population and Crohn's disease samples from Lloyd-Price et al. 2019

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|--------|---------|----------|------------|
| Chao1 | UC | healthy | 3.27e-27 | 9.82e-26 |
| Margalef | UC | healthy | 1.40e-14 | 1.40e-13 |
| Menhinick | UC | healthy | 1.41e-12 | 1.06e-11 |
| Fisher alpha | UC | healthy | 9.57e-12 | 5.74e-11 |
| Faith PD | UC | healthy | 4.52e-08 | 1.70e-07 |
| Gini index | UC | healthy | 8.05e-08 | 2.68e-07 |
| Strong | UC | healthy | 0.00072 | 0.00144 |
| Shannon entropy | UC | healthy | 0.01894 | 0.03157 |
| Pielou evenness | UC | healthy | 0.20476 | 0.26708 |
| Simpson | UC | healthy | 0.48789 | 0.56295 |

Supplementary Table S6. Results of the Mann-Whitney-Wilcoxon test for difference in means of healthy population and Ulcerative colitis samples (Lloyd-Price et al. 2019 and Qiita ID: 11549)

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|--------------|-----------|-----------|------------|
| Fisher alpha | control(AGP) | control_2 | 1.93e-187 | 1.93e-185 |
| Gini index | control(AGP) | control_2 | 2.40e-30 | 2.66e-29 |
| Margalef | control(AGP) | control_2 | 4.85e-13 | 1.87e-12 |
| Menhinick | control(AGP) | control_2 | 5.61e-10 | 1.44e-09 |
| Chao1 | control(AGP) | control_2 | 7.50e-06 | 1.32e-05 |
| Faith PD | control(AGP) | control_2 | 0.00045 | 0.00066 |
| Simpson | control(AGP) | control_2 | 0.20806 | 0.22615 |
| Pielou evenness | control(AGP) | control_2 | 0.4135 | 0.43526 |
| Shannon entropy | control(AGP) | control_2 | 0.52903 | 0.54539 |
| Strong | control(AGP) | control_2 | 0.77432 | 0.77432 |

Supplementary Table S7. Results of the Mann-Whitney-Wilcoxon test for difference in means of controls from longitudinal CD study (Vázquez-Baeza et al. 2018) and AGP controls

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|--------|--------|----------|------------|
| Faith PD | CD_2 | CD_1 | 2.01e-14 | 3.01e-13 |
| Gini index | CD_2 | CD_1 | 2.14e-09 | 5.35e-09 |
| Menhinick | CD_2 | CD_1 | 6.30e-08 | 1.35e-07 |
| Strong | CD_2 | CD_1 | 9.74e-07 | 1.83e-06 |
| Pielou evenness | CD_2 | CD_1 | 0.00015 | 0.00024 |
| Chao1 | CD_2 | CD_1 | 0.00586 | 0.00732 |
| Margalef | CD_2 | CD_1 | 0.01065 | 0.01278 |
| Simpson | CD_2 | CD_1 | 0.22596 | 0.2421 |
| Fisher alpha | CD_2 | CD_1 | 0.42371 | 0.43832 |
| Shannon entropy | CD_2 | CD_1 | 0.68844 | 0.68844 |

Supplementary Table S8. Results of the Mann-Whitney-Wilcoxon test for difference in means of CD samples from longitudinal study (Vázquez-Baeza et al. 2018) and CD samples from first data set (Lloyd-Price et al. 2019)

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|--------|---------|---------|------------|
| Simpson | crohns | control | 0.00007 | 0.00010 |
| Pielou evenness | crohns | control | 0.00010 | 0.00014 |
| Shannon entropy | crohns | control | 0.00067 | 0.00091 |
| Chao1 | crohns | control | 0.00074 | 0.00097 |
| Strong | crohns | control | 0.00087 | 0.00109 |
| Gini index | crohns | control | 0.00926 | 0.01068 |
| Faith PD | crohns | control | 0.04426 | 0.04918 |
| Margalef | crohns | control | 0.06702 | 0.06702 |
| Menhinick | crohns | control | 0.06702 | 0.06702 |
| Fisher alpha | crohns | control | 0.06702 | 0.06702 |

Supplementary Table S9. Results of the Mann-Whitney-Wilcoxon test for difference in means of controls and Crohn's samples in Vázquez-Baeza et al. 2018

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|------------------|---------|----------|------------|
| Chao1 | crohns (surgery) | control | 1.39e-24 | 4.17e-23 |
| Margalef | crohns (surgery) | control | 2.14e-23 | 3.22e-22 |
| Menhinick | crohns (surgery) | control | 4.20e-23 | 4.20e-22 |
| Fisher alpha | crohns (surgery) | control | 1.23e-21 | 6.15e-21 |
| Faith PD | crohns (surgery) | control | 1.23e-21 | 6.15e-21 |
| Gini index | crohns (surgery) | control | 1.23e-21 | 6.15e-21 |
| Strong | crohns (surgery) | control | 1.04e-19 | 4.47e-19 |
| Pielou evenness | crohns (surgery) | control | 1.27e-19 | 4.77e-19 |
| Shannon entropy | crohns (surgery) | control | 1.65e-17 | 5.50e-17 |
| Simpson | crohns (surgery) | control | 9.04e-11 | 2.26e-10 |

Supplementary Table S10. Results of the Mann-Whitney-Wilcoxon test for difference in means of controls and Crohn's samples that undergone surgery in Vázquez-Baeza et al. 2018

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|---------|--------|----------|------------|
| Chao1 | healthy | CDI | 2.92e-47 | 2.92e-46 |
| Faith PD | healthy | CDI | 9.10e-47 | 4.55e-46 |
| Fisher alpha | healthy | CDI | 2.69e-45 | 8.97e-45 |
| Gini index | healthy | CDI | 1.41e-44 | 3.52e-44 |
| Margalef | healthy | CDI | 8.07e-44 | 1.61e-43 |
| Menhinick | healthy | CDI | 1.06e-43 | 1.77e-43 |
| Shannon entropy | healthy | CDI | 9.76e-21 | 1.39e-20 |
| Simpson | healthy | CDI | 7.75e-12 | 9.69e-12 |
| Pielou evenness | healthy | CDI | 0.00004 | 0.00004 |
| Strong | healthy | CDI | 0.99646 | 0.99646 |

Supplementary Table S11. Results of the Mann-Whitney-Wilcoxon test for difference between controls and CDI samples in Khanna et al. 2016

| parameter | group1 | group2 | p.value | p.adjusted |
|-----------------|--------|---------|----------|------------|
| Chao1 | donor | CDIpost | 2.84e-20 | 8.52e-19 |
| Chao1 | donor | CDIpre | 1.46e-15 | 2.18e-14 |
| Faith PD | donor | CDIpost | 6.45e-15 | 6.45e-14 |
| Faith PD | donor | CDIpre | 1.34e-14 | 1.00e-13 |
| Fisher alpha | donor | CDIpost | 7.26e-13 | 4.36e-12 |
| Fisher alpha | donor | CDIpre | 1.18e-12 | 5.88e-12 |
| Gini index | donor | CDIpost | 5.29e-12 | 2.27e-11 |
| Gini index | donor | CDIpre | 1.04e-11 | 3.90e-11 |
| Margalef | donor | CDIpost | 3.19e-11 | 1.06e-10 |
| Margalef | donor | CDIpre | 6.77e-11 | 1.85e-10 |
| Pielou evenness | donor | CDIpost | 6.77e-11 | 1.85e-10 |
| Pielou evenness | donor | CDIpre | 9.17e-11 | 2.29e-10 |
| Shannon entropy | donor | CDIpost | 2.67e-10 | 6.16e-10 |
| Shannon entropy | donor | CDIpre | 9.59e-10 | 2.05e-09 |
| Simpson | donor | CDIpost | 2.05e-09 | 4.11e-09 |
| Simpson | donor | CDIpre | 7.63e-09 | 1.43e-08 |
| Strong | donor | CDIpost | 8.33e-08 | 1.47e-07 |
| Strong | donor | CDIpre | 2.16e-07 | 3.59e-07 |
| Menhinick | donor | CDIpre | 0.00001 | 0.00002 |
| Gini index | CDIpre | CDIpost | 0.00876 | 0.01314 |
| Shannon entropy | CDIpre | CDIpost | 0.00925 | 0.01321 |
| Chao1 | CDIpre | CDIpost | 0.01016 | 0.01385 |
| Fisher alpha | CDIpre | CDIpost | 0.01271 | 0.01469 |
| Margalef | CDIpre | CDIpost | 0.01271 | 0.01469 |
| Menhinick | CDIpre | CDIpost | 0.01271 | 0.01469 |
| Simpson | CDIpre | CDIpost | 0.01273 | 0.01469 |
| Menhinick | donor | CDIpost | 0.09159 | 0.10177 |
| Pielou evenness | CDIpre | CDIpost | 0.09762 | 0.10459 |
| Faith PD | CDIpre | CDIpost | 0.14901 | 0.15415 |
| Strong | CDIpre | CDIpost | 0.31623 | 0.31623 |

Supplementary Table S12. Results of the Mann-Whitney-Wilcoxon test for difference in means of different conditions (healthy donors, CDi pre-FMT, CDI post-FMT) in Hospital Clínic's dataset (Aira et al. 2022)

| accuracy_condition | accuracy_healthy_or_not |
|--------------------|---|
| 0.88 | 0.89 |
| 0.85 | 0.86 |
| 0.84 | 0.85 |
| 0.84 | 0.87 |
| 0.83 | 0.85 |
| 0.82 | 0.85 |
| 0.72 | 0.74 |
| 0.72 | 0.74 |
| 0.72 | 0.74 |
| 0.71 | 0.73 |
| 0.71 | 0.72 |
| 0.71 | 0.74 |
| 0.71 | 0.73 |
| 0.70 | 0.74 |
| 0.70 | 0.73 |
| 0.70 | 0.73 |
| 0.69 | 0.70 |
| 0.69 | 0.73 |
| 0.69 | 0.71 |
| 0.68 | 0.71 |
| 0.68 | 0.72 |
| 0.68 | 0.71 |
| 0.68 | 0.68 |
| 0.68 | 0.70 |
| 0.67 | 0.71 |
| 0.64 | 0.66 |
| | 0.88 0.85 0.84 0.84 0.83 0.82 0.72 0.72 0.72 0.71 0.71 0.71 0.71 0.70 0.70 0.70 0.69 0.69 0.69 0.68 0.68 0.68 0.68 0.68 |

Supplementary Table S13. Accuracy of prediction of different models of random forest classifier trained on a dataset consisting of IBD (CD and UC), CDI and healthy samples

| model | accuracy_condition | accuracy_healthy_or_not |
|-------------------------|--------------------|-------------------------|
| all alpha metrics | 0.87 | 0.88 |
| Faith + Gini | 0.86 | 0.86 |
| Menhinick + Gini | 0.85 | 0.86 |
| Fisher + Gini | 0.84 | 0.86 |
| Margalef + Gini | 0.83 | 0.85 |
| Chao1 + Gini | 0.82 | 0.84 |
| Fisher + Simpson | 0.70 | 0.70 |
| Menhinick + Pielou | 0.70 | 0.69 |
| Chao1 + Pielou | 0.69 | 0.70 |
| Margalef + Pielou | 0.69 | 0.70 |
| Fisher + Pielou | 0.69 | 0.69 |
| chao1 + shannon_entropy | 0.69 | 0.70 |
| margalef + simpson | 0.68 | 0.70 |
| chao1 + simpson | 0.68 | 0.69 |
| Menhinick + Simpson | 0.68 | 0.68 |
| Fisher + Shannon | 0.67 | 0.67 |
| Menhinick + Strong | 0.67 | 0.66 |
| Menhinick + Shannon | 0.67 | 0.66 |
| Margalef + Shannon | 0.66 | 0.66 |
| Chao1 + Strong | 0.65 | 0.67 |
| Margalef + Strong | 0.64 | 0.65 |
| Fisher + Strong | 0.64 | 0.65 |
| Faith + Simpson | 0.63 | 0.63 |
| Faith + Pielou | 0.62 | 0.61 |
| Faith + Shannon | 0.61 | 0.62 |
| Faith + Strong | 0.59 | 0.62 |

Supplementary Table S14. Accuracy of prediction of different models of random forest classifier trained on dataset consisting of IBD (CD and UC) and healthy samples

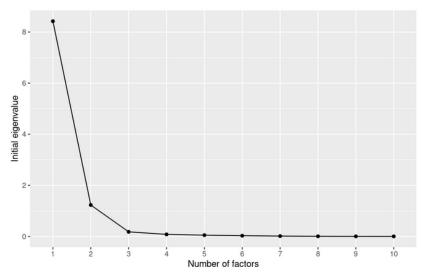
| model | accuracy |
|---------------------|----------|
| Chao1 + Gini | 1.00 |
| Margalef + Gini | 1.00 |
| Menhinick + Gini | 1.00 |
| Fisher + Gini | 1.00 |
| Faith + Gini | 0.99 |
| Faith + Strong | 0.98 |
| Faith + Pielou | 0.98 |
| Faith + Shannon | 0.98 |
| Faith + Simpson | 0.98 |
| All alpha metrics | 0.98 |
| Chao1 + Simpson | 0.97 |
| Menhinick + Pielou | 0.97 |
| Menhinick + Simpson | 0.97 |
| Chao1 + Strong | 0.97 |
| Chao1 + Pielou | 0.97 |
| Chao1 + Shannon | 0.97 |
| Menhinick + Shannon | 0.97 |
| Menhinick + Strong | 0.96 |
| Margalef + Strong | 0.96 |
| Margalef + Pielou | 0.95 |
| Fisher + Strong | 0.95 |
| Fisher + Pielou | 0.95 |
| Fisher + Shannon | 0.95 |
| Fisher + Simpson | 0.95 |
| Margalef + Shannon | 0.95 |
| Margalef + Simpson | 0.95 |

Supplementary Table S15. Accuracy of prediction of different models of random forest classifier trained on dataset consisting of CDI and healthy samples

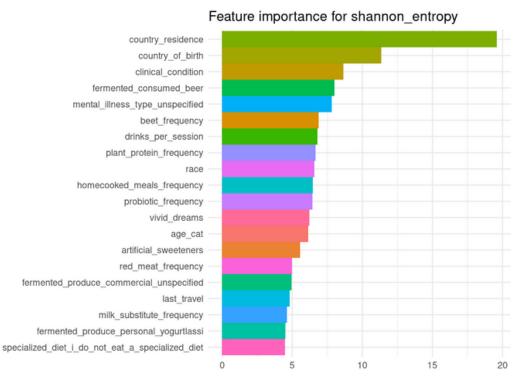
| model | accuracy |
|---------------------|----------|
| Chao1 + Gini | 1.00 |
| Margalef + Gini | 1.00 |
| Menhinick + Gini | 1.00 |
| Menhinick + Strong | 1.00 |
| Menhinick + Shannon | 1.00 |
| Fisher + Gini | 1.00 |
| Faith + Gini | 1.00 |
| All alpha metrics | 1.00 |
| Chao1 + Strong | 0.97 |
| Menhinick + Pielou | 0.97 |
| Menhinick + Simpson | 0.97 |
| Fisher + Shannon | 0.97 |
| Fisher + Simpson | 0.97 |
| Faith + Strong | 0.97 |
| Faith + Pielou | 0.97 |
| Faith + Shannon | 0.97 |
| Faith + Simpson | 0.97 |
| Chao1 + Pielou | 0.95 |
| Chao1 + Simpson | 0.95 |
| Margalef + Strong | 0.95 |
| Margalef + Pielou | 0.95 |
| Fisher + Pielou | 0.95 |
| Chao1 + Shannon | 0.92 |
| Margalef + Shannon | 0.92 |
| Margalef + Simpson | 0.92 |
| Fisher + Strong | 0.92 |

Supplementary Table S16. Accuracy of prediction of different models of random forest classifier trained on dataset consisting of Hospital Clínic's CDI samples before FMT and healthy donor samples

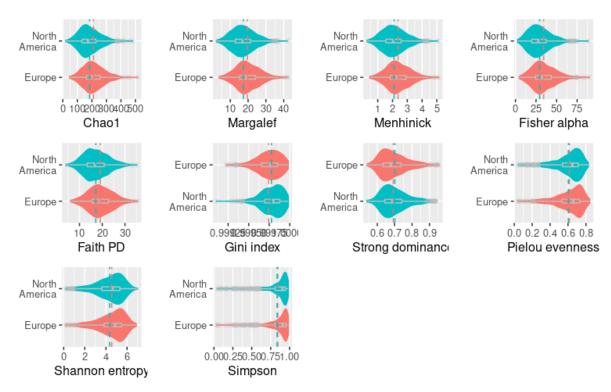
3 Supplementary Figures



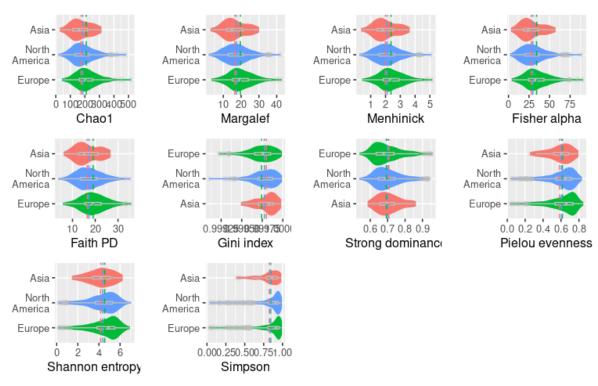
Supplementary Figure S1. Scree plot based on exploratory factor analysis of 10 alpha diversity indices computed on AGP dataset



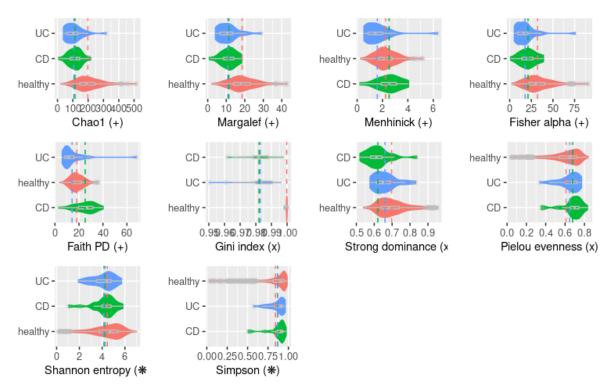
Supplementary Figure S2. Feature importance of AGP metadata categories for estimating alpha diversity metrics (in this case Shannon entropy) obtained by Random Forest classifier



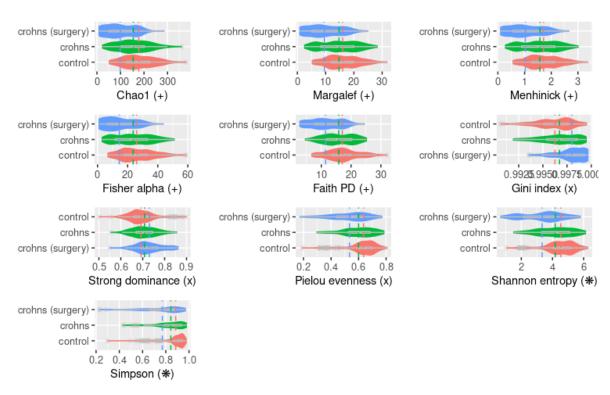
Supplementary Figure S3. Difference in distributions and means of alpha metrics in groups of AGP samples with different countries of residence (grouped by continents)



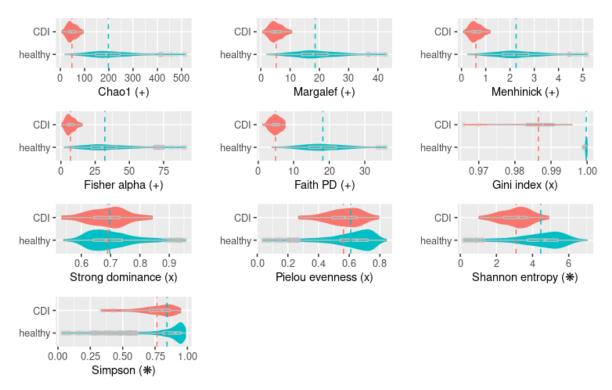
Supplementary Figure S4. Difference in distributions and means of alpha metrics in groups of AGP samples with different countries of birth (grouped by continents)



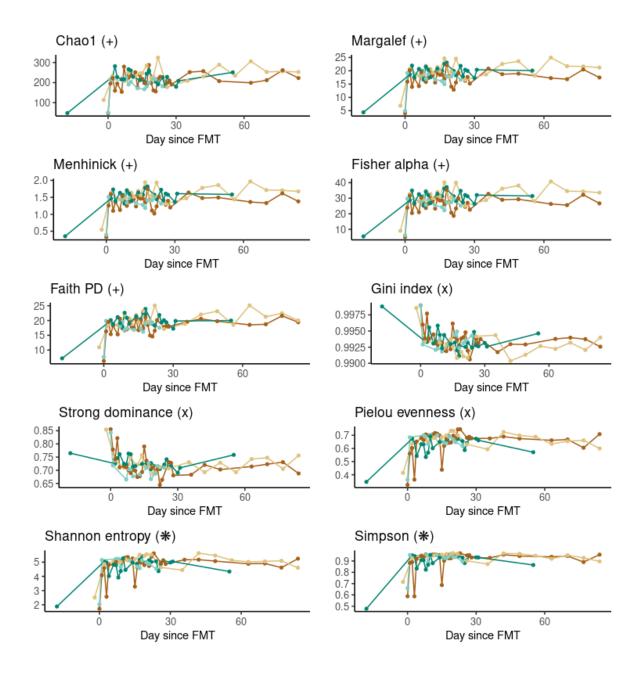
Supplementary Figure S5. Difference in distributions and means of different alpha metrics between healthy and IBD dataset (Lloyd-Price et al. 2019 and Qiita ID: 11549)



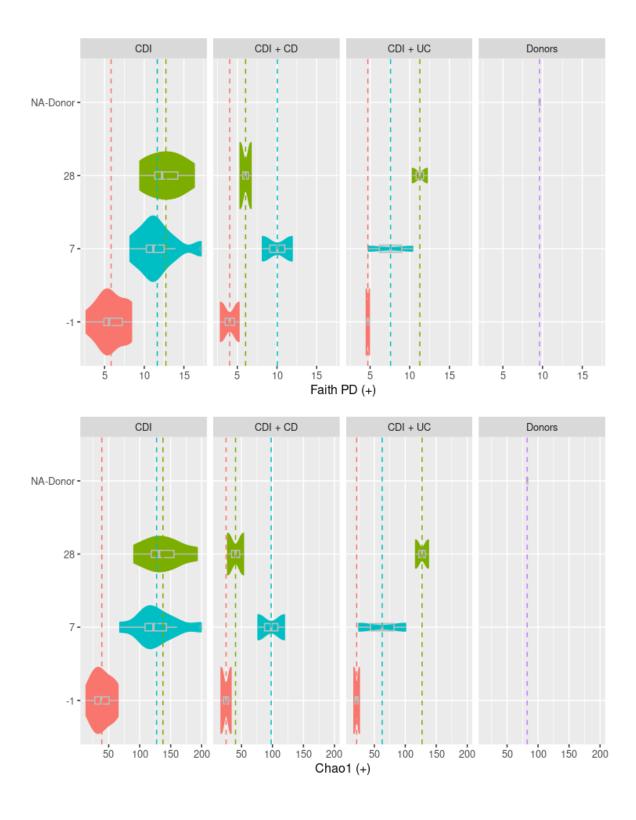
Supplementary Figure S6. Difference in distributions and means of different alpha metrics between controls and Crohn's samples (Vázquez-Baeza et al. 2018)

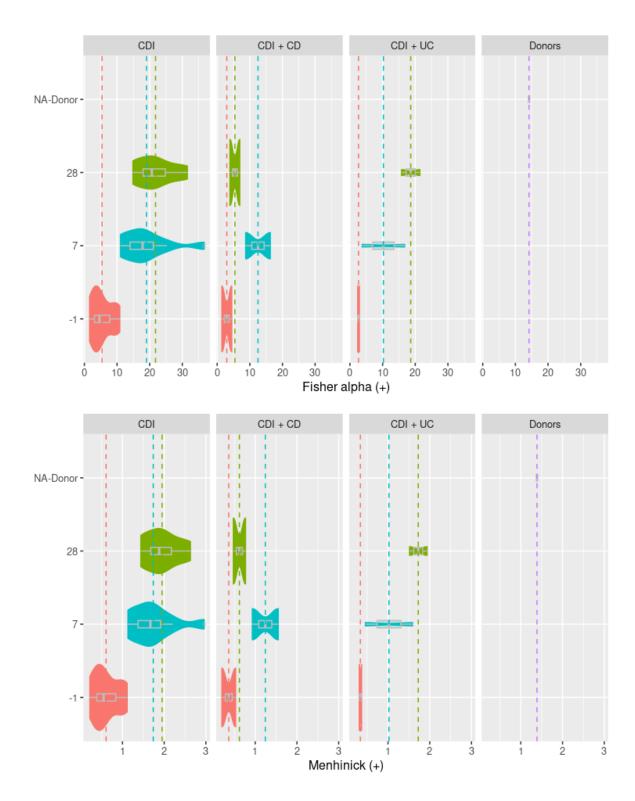


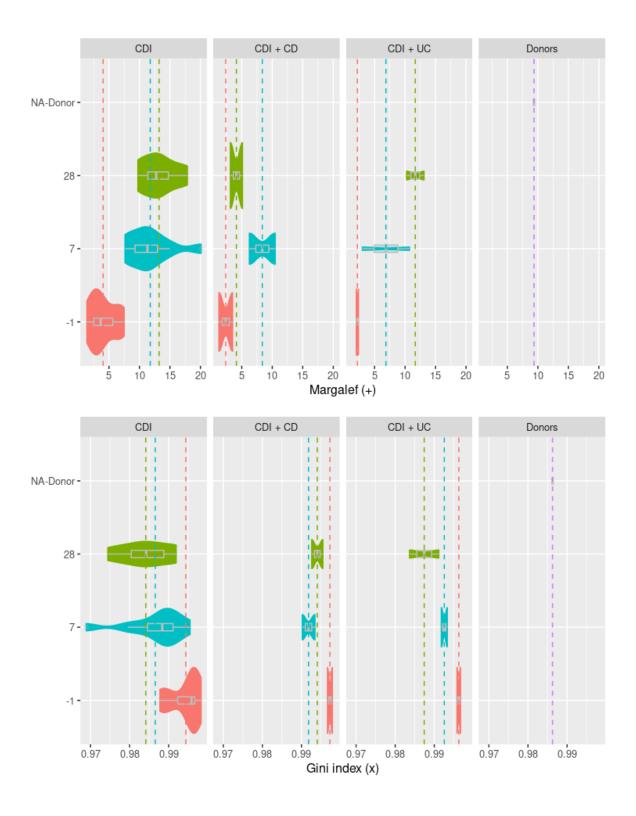
Supplementary Figure S7. Difference in distributions and means of different alpha metrics between controls and CDI samples (Khanna et al. 2016)

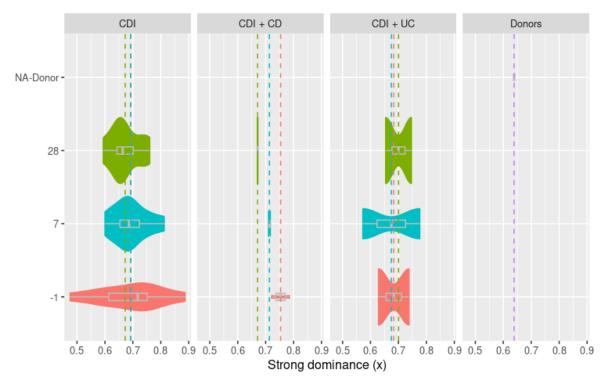


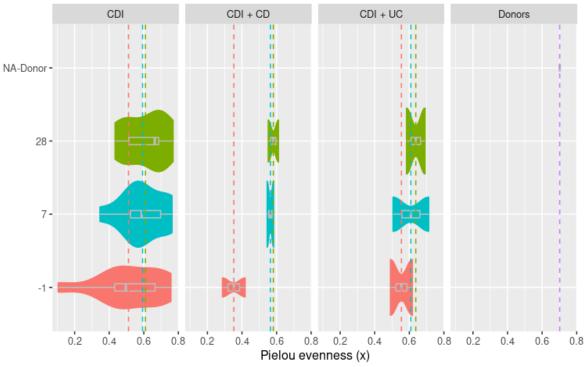
Supplementary Figure S8. Progression of alpha diversity metrics' value in time of CDI samples after FMT (Weingarden et al. 2015)

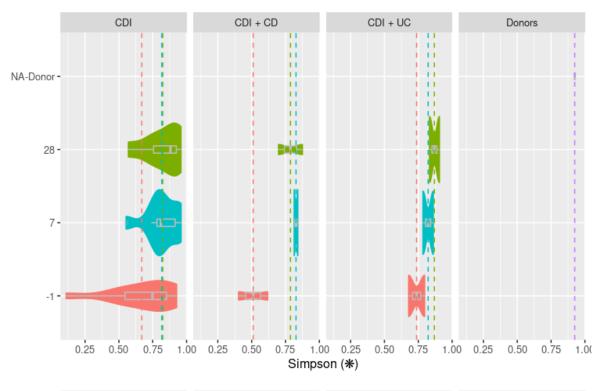


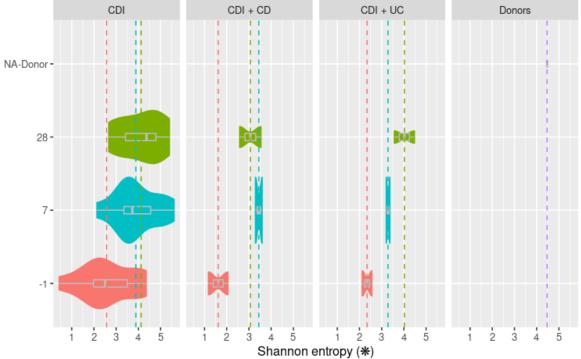












Supplementary Figure S9. Difference of different alpha diversity value depending on day since FMT in CDI patients with different underlying conditions (none, CD or UC) from Khanna et al. 2017

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