Metagenome Analysis of Premature Birth

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Overview

- Introduction
- 2 Materials
- Methods
- 4 Results
- 5 Discussion

Introduction

Microbiome

- Microbiota: the microorganisms which live inside & on humans (Turnbaugh et al., 2007)
- Microbiome: 10^{13} to 10^{14} microorganisms whose which collective genome (Gill et al., 2006)



Figure: Concept of a core human microbiome (Turnbaugh et al., 2007)

rRNA

- Ribosomal RNA
- Well-known as a key to phylogeny (Olsen & Woese, 1993)

Premature Birth (Preterm Birth; PTB)

PTB:

- PTB < 37 GW (Gestational week)</p>
- ② Normal \geq 37 GW

Detailed PTB:

- Extremely PTB < 28 GW</p>
- 28 GW ≤ Very PTB < 34 GW</p>
- $34 \text{ GW} \leq \text{Late PTB} < 37 \text{ GW}$
- Normal \geq 37 GW
- (J. Tucker & McGuire, 2004; Voronkov, Solonovych, Liashenko, & Revenko, 2018)

Materials

16S rRNA Sequencing

16S rRNA sequencing is the *reference method* for bacterial taxonomy & identification (Mignard & Flandrois, 2006)

- Three main reasons (Janda & Abbott, 2007):
 - 16S rRNA exists in almost all bacteria
 - Functions of the 16S rRNA has not changed over time
 - 16S rRNA is large enough for bioinformatics

Train/Test Data vs. Validate Data

- JBNU/Helixco data
 - First data
 - Second data
 - Stool data

Table: Sample Information

Data	Participants	Samples	Remarks
First	24	107	-
Second	35	288	-
Third	10	106	-
Stool	63	126	Stool

Methods

Methods

Qiime 2 Workflow

Qiime 2 Workflow

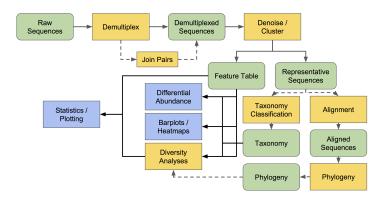


Figure: QIIME 2 workflow (Bolyen et al., 2019; Mandal, Van Treuren, White, Eggesbø, et al., 2015; McDonald et al., 2012)

Filitering with Quality Score

Drawback between:

- Longer sequence read
- Higher quality value
- \therefore Select the maximum length n where:

$$\forall n_i \in \{n_k | \text{MedianQualityScore} \ge 30\}$$

$$\exists! n \in \{n_i\} : n \ge n_i$$
 (1)

Denoising Techniques

- DADA2: Amplicon Sequence Variants (ASVs) (Callahan et al., 2016)
- Deblur: Operational Taxonomic Units (OTUs) (Amir et al., 2017)



Figure: Denoising Algorithms

Taxonomy Classification

- Greengenes (GG) (DeSantis et al., 2006)
- SILVA (Pruesse et al., 2007; Quast et al., 2012)

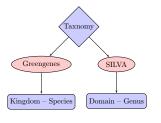


Figure: Taxonomy Classifications

"A **higher** performance at taxonomic levels above *genus level*; but performance appears to **drop** at *species level*" (Gihawi et al., 2019)

Merging Denoising/Taxonomy

Merging multiple IDs (ASVs or OTUs) into one, which have

- Different IDs
- Identified as same taxonomy

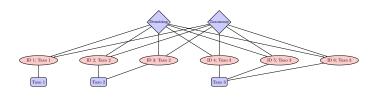


Figure: Example Diagram for Merging Denoising/Taxonomy

Methods

Abundance Test

ANCOM

- Analysis of composition of microbiome (Mandal, Van Treuren, White, Eggesbø, et al., 2015)
- ANCOM detects significantly abundant taxa, while maintain high statistical power
- Find taxa that can divide each classes

Methods

Diversity Indices

Diversity Indices

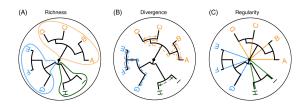


Figure: Three dimensions of phylogenic information (C. M. Tucker et al., 2017)

- A quantitative measure that shows richness, divergence, and regularity (C. M. Tucker et al., 2017)
- Alpha diversity indices: the richness of taxa at a single community
- Beta diversity indices: the taxonomic differentiation between communities

Alpha Diversity Indices

- Evenness index
- Faith's Phylogenetic Diversity (Faith PD) index
- Oberseved Features index
- Shannon's Diversity index

Beta Diversity Indices

- Bray-Curtis distance index
- Jaccard distnace index
- Unweighted UniFrac distance index
- Weighted UniFrac distance index

Methods

Miscellaneous

t-distributed Stochastic Neighbor Embedding (t-SNE)

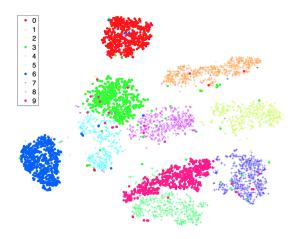


Figure: t-SNE with handwritten data (Maaten & Hinton, 2008)

Python Packages

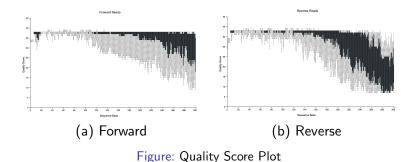
- Pandas (McKinney et al., 2011)
- Scikit-Learn (Pedregosa et al., 2011)
- SciPy (Virtanen et al., 2020)
- Matplotlib (Hunter, 2007)
- Seaborn (Waskom et al., 2020)
- Statannot

Results

Results

Filtering Results

Quality Score from JBNU/Helixco Data



- $\ell_{Forward} = 300$
- $\ell_{Reverse} = 245$

Results

t-SNE with Clinical Information

Workflow for t-SNE with Site/Premature Information



Figure: Workflow of t-SNE for Site/Premature Information

Used Clinical Information

- Diseases
 - Gestational Diabetes
 - Maternal Overweight/Obesity
 - Maternal Weight Gain
 - 4 Hypertension
 - PROM
 - Antibiotic
 - Steroid
- Probing sites
 - Maternal Mouth

Selected t-SNE Plots I

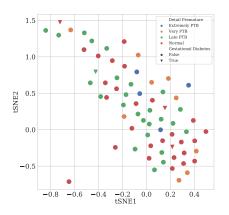


Figure: t-SNE about Gestational Diabetes

Selected t-SNE Plots II

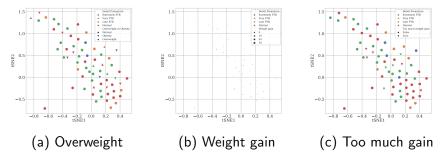


Figure: t-SNE about Maternal Weight

Selected t-SNE Plots III

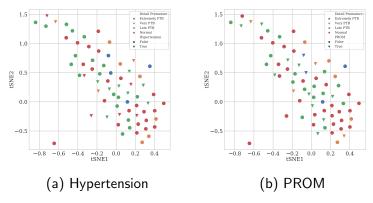


Figure: t-SNE about Disease

Selected t-SNE Plots IV

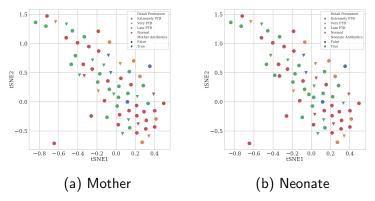


Figure: t-SNE about Antibiotics Usage

Selected t-SNE Plots V

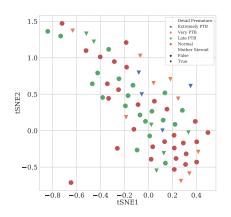


Figure: t-SNE about Steroid Usage

Results

Bacterial Abundance Test with ANCOM

ANCOM?

- Analysis of composition of microbiomes
- ANCOM can be used for analyzing the composition of microbiomes in multiple populations (Mandal, Van Treuren, White, Eggesbø, et al., 2015)
- Differential abundance testing
- clr: Centered log(Ratio)
- W: a count of the number of sub-hypothesis which have passed for given species

ANCOM with ...

Site selection:

- 1 Neonatal mouth: 1-day, 3-day, and 5-day
- Cervix
- Maternal mouth
- Vagina

PTB:

- Premature
- Oetail premature

ANCOM with Neonatal Mouth



Figure: ANCOM with Neonatal Mouth

ANCOM with Maternal Mouth



Figure: ANCOM with Maternal Mouth

 Bacteria Proteobacteria Alphaproteobacteria Rickettsiales mitochondria family

ANCOM with Cervix

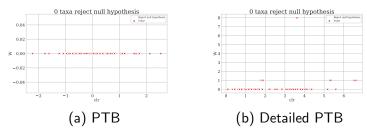


Figure: ANCOM with Cervix

ANCOM with Vagina

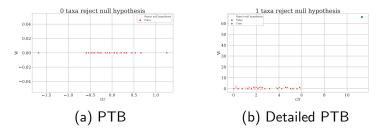


Figure: ANCOM with Vagina

 Bacteria Proteobacteria Epsilonproteobacteria Campylobacterales Campylobacteraceae Campylobacter genus

Results

Diversity Index

Rarefaction

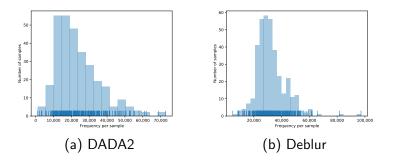


Figure: Rarefaction from the Data

• $min(\ell_{DADA2})$: 1046

• $min(\ell_{Deblur})$: 4864

Diversity Indices

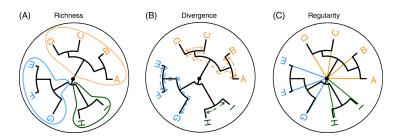


Figure: Dimensions of phylogenetic information (C. M. Tucker et al., 2017)

- Alpha-diversity: the species diversity in a local scale
- Beta-diversity: the species diversity between local scales

Alpha-diversity

- Alpha diveristy indices
 - Faith PD
 - Observed Features
 - Pielou Evenness
 - Shannon Entropy
- Diseases/Conditions
 - Gestational Diabetes
 - 2 Too much Weight Gain
 - Overweight/Obesity
 - 4 Hypertension
 - PROM
 - 6 Antibiotics
 - Steroid
- Site Selection
 - Maternal Mouth
 - Neonatal Mouth 3 day

Alpha-diversity Violin Plots I

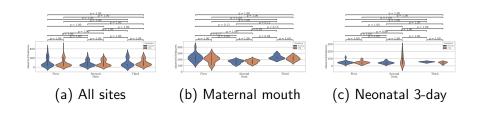


Figure: Observed features by Data

Alpha-diversity Violin Plots II

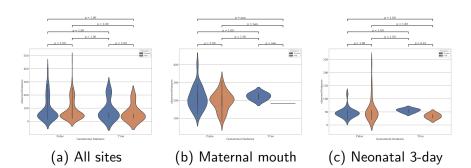


Figure: Observed features by Gestational Diabetes

Alpha-diversity Violin Plots III

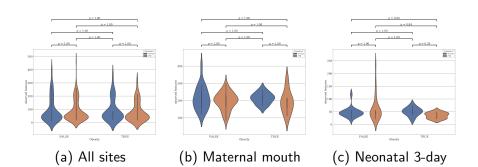


Figure: Observed features by Overweight/Obesity

Alpha-diversity Violin Plots IV

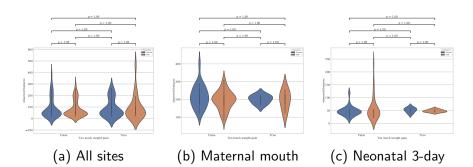


Figure: Observed features by Too much weight gain

Alpha-diversity Violin Plots V

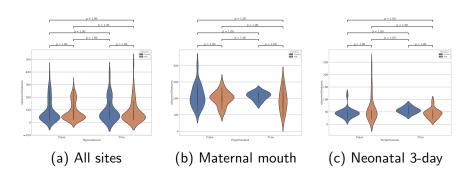


Figure: Observed features by Hypertension

Alpha-diversity Violin Plots VI

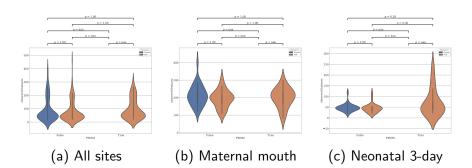


Figure: Observed features by PROM

Alpha-diversity Violin Plots VII

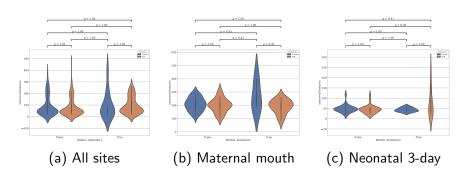


Figure: Observed features by Mother Antibiotics

Alpha-diversity Violin Plots VIII

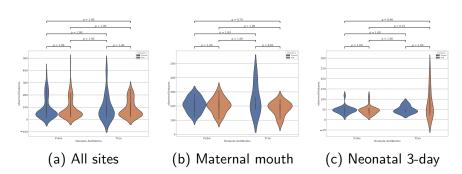


Figure: Observed features by Neonate Antibiotics

Alpha-diversity Violin Plots IX

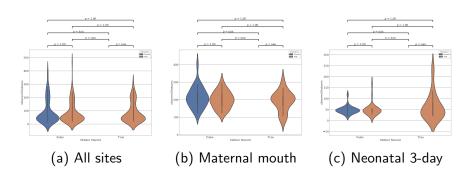
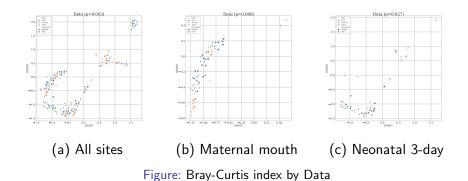


Figure: Observed features by Mother Steroid

Beta-diversity

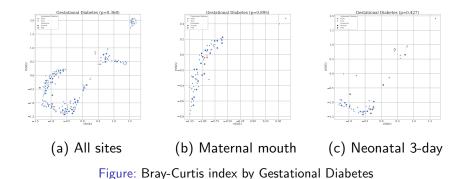
- Beta diversity indices
 - Bray-Curtis distance
 - Jaccard distance
 - Unweighted UniFrac distance
 - Weighted UniFrac distance
- Diseases/Conditions
 - Gestational Diabetes
 - Too much Weight Gain
 - Overweight/Obesity
 - 4 Hypertension
 - PROM
 - Antibiotics
 - Steroid
- Site Selection
 - Maternal Mouth
 - Neonatal Mouth 3 day

Beta-diversity Scatter Plots I



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Beta-diversity Scatter Plots II



Beta-diversity Scatter Plots III

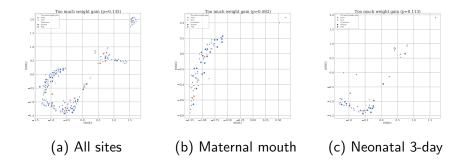
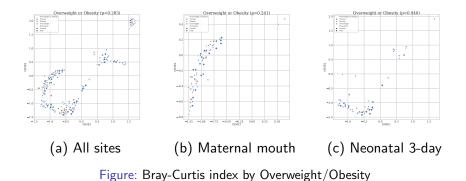


Figure: Bray-Curtis index by Too much weight gain

Beta-diversity Scatter Plots IV



Beta-diversity Scatter Plots V



Figure: Bray-Curtis index by Hypertension

Beta-diversity Scatter Plots VI

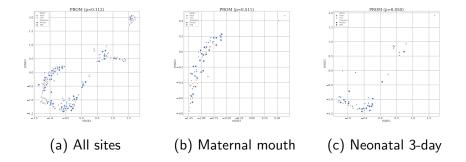
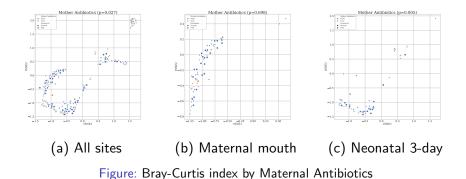
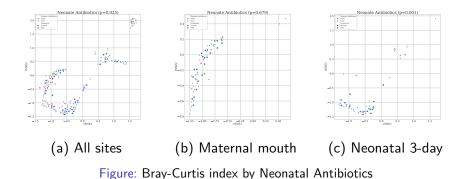


Figure: Bray-Curtis index by PROM

Beta-diversity Scatter Plots VII



Beta-diversity Scatter Plots VIII



Beta-diversity Scatter Plots IX

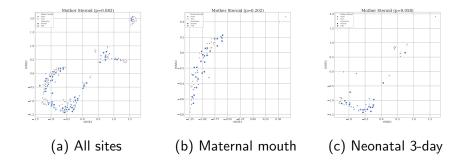


Figure: Bray-Curtis index by Maternal Steroid

Results

Metagenomic Biomarker Discovery

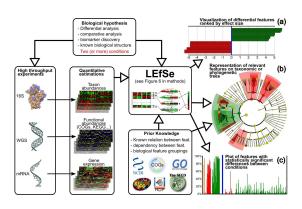


Figure: LEfSe workflow (Segata et al., 2011)

LEfSe Settings

- Analysis Pipeline
 - Algorithm: DADA2 and Deblur
 - Databases: GG, Silva, and HOMD
- Subject Groups
 - PTB
 - Normal
- Sites
 - Cervix
 - Mouth
 - Neonatal Mouth
 - Vagina

LEfSe Results I

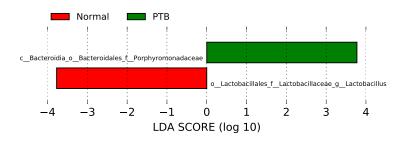


Figure: Deblur+HOMD in Cervix

LEfSe Results II

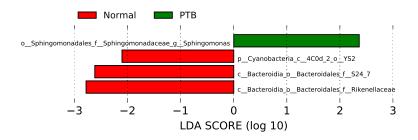


Figure: Deblur+Greengenes in Neonate 1-day

LEfSe Results III

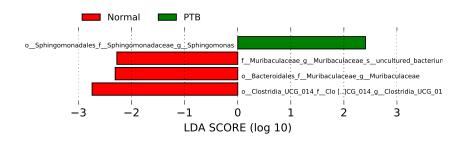


Figure: Deblur+Silva in Neonate 1-day

LEfSe Results IV

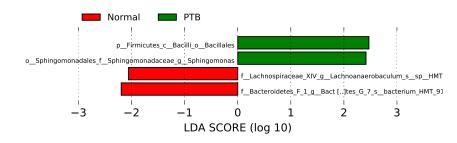


Figure: Deblur+HOMD in Neonate 1-day

LEfSe Results V

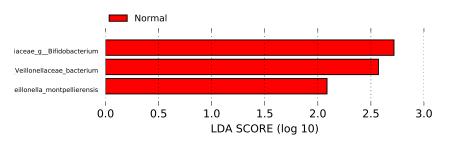


Figure: Deblur+Silva in Vagina

Discussion

To-do List

- Human Oral Microbiome Database
- LEfSe: explain differences between classes for statistical significance

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