## Metagenome Analysis of Premature Birth

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### Overview

- Introduction
- 2 Materials
- Methods
- 4 Results
- 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
- 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

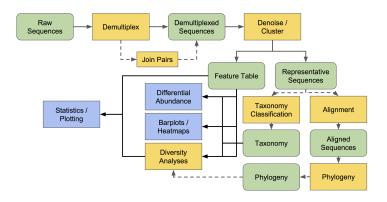


Figure: QIIME 2 workflow (Bolyen et al., 2019; Mandal, Van Treuren, White, Eggesbø, et al., 2015; McDonald et al., 2012)

### 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

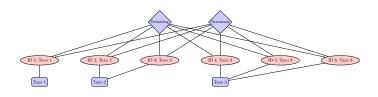


Figure: Example Diagram for Merging Denoising/Taxonomy

### 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

### **M**ethods

**Diversity Indices** 

### **Diversity Indices**

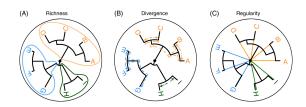


Figure: Three dimensions of phylogenic information (C. M. Tucker et al., 2017)

- 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)

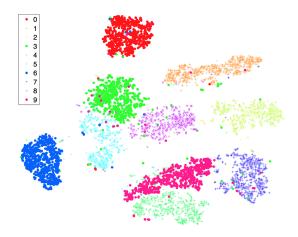


Figure: t-SNE with handwritten data (Maaten & Hinton, 2008)

# 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



Figure: Quality Score Plot

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

### 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

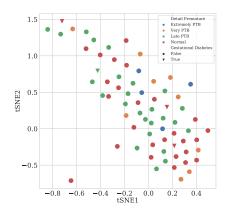


Figure: t-SNE about Gestational Diabetes

#### Selected t-SNE Plots II

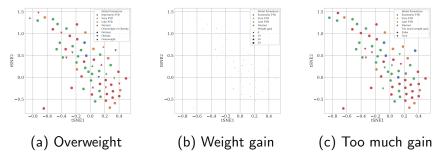


Figure: t-SNE about Maternal Weight

#### Selected t-SNE Plots III

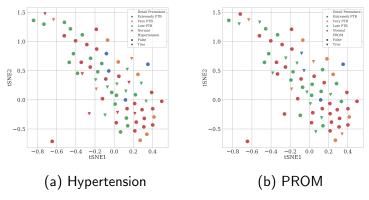


Figure: t-SNE about Disease

### Selected t-SNE Plots IV

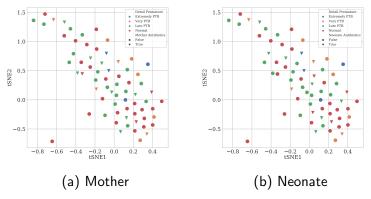


Figure: t-SNE about Antibiotics Usage

### Selected t-SNE Plots V

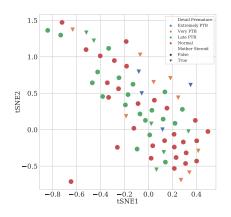


Figure: t-SNE about Steroid Usage

### 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

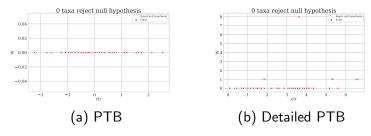
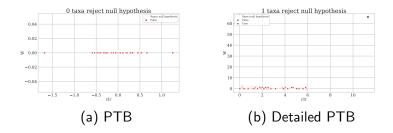


Figure: ANCOM with Cervix

# ANCOM with Vagina



Ç .

 Bacteria Proteobacteria Epsilonproteobacteria Campylobacterales Campylobacteraceae Campylobacter genus

Figure: ANCOM with Vagina

### Results

Diversity Index

### Rarefaction

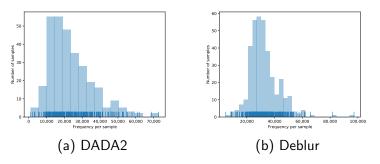


Figure: Rarefaction from the Data

•  $min(\ell_{DADA2})$ : 1046

•  $min(\ell_{Deblur})$ : 4864

## **Diversity Indices**

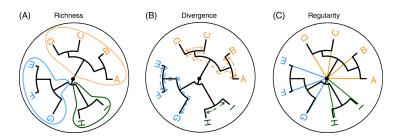


Figure: Dimensions of phylogenetic information (C. M. Tucker et al., 2017)

- 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
  - Antibiotics
  - Steroid
- Site Selection
  - Maternal Mouth
  - Neonatal Mouth 3 day

# Alpha-diversity Violin Plots I

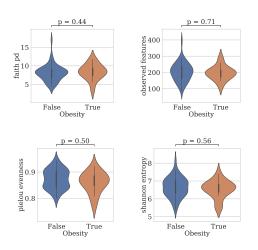


Figure: Alpha-diversities by DADA2 & Overweight/Obesity in Maternal Mouth

# Alpha-diversity Violin Plots II

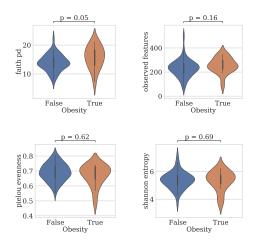


Figure: Alpha-diversities by Deblur & Overweight/Obesity in Maternal Mouth

# Alpha-diversity Violin Plots III

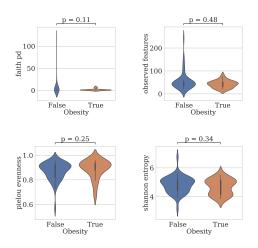


Figure: Alpha-diversities by DADA2 & Overweight/Obesity in Neonatal-3day

## Alpha-diversity Violin Plots IV

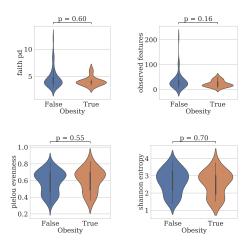
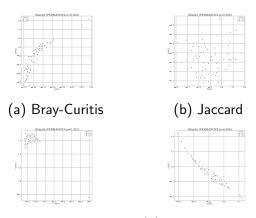


Figure: Alpha-diversities by Deblur & Overweight/Obesity in Neonatal-3day

### 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

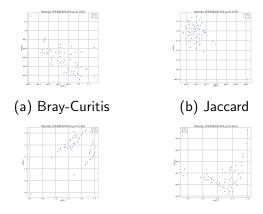
# Beta-diversity Scatter Plots I



(c) Unweighted UniFrac (d) Weighted UniFrac

Figure: Beta-diversities by DADA2 & Overweight/Obesity in Maternal Mouth

# Beta-diversity Scatter Plots II



(c) Unweighted UniFrac (d) Weighted UniFrac

Figure: Beta-diversities by Deblur & Overweight/Obesity in Maternal Mouth

### Beta-diversity Scatter Plots III





(a) Bray-Curitis



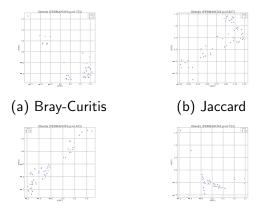
(b) Jaccard



(c) Unweighted UniFrac (d) Weighted UniFrac

Figure: Beta-diversities by DADA2 & Overweight/Obesity in Neonate 3-day

# Beta-diversity Scatter Plots IV



(c) Unweighted UniFrac (d) Weighted UniFrac

Figure: Beta-diversities by Deblur & Overweight/Obesity in Neonate 3-day

### Results

Metagenomic Biomarker Discovery

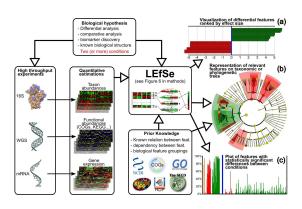


Figure: LEfSe workflow (Segata et al., 2011)

# 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 with PTB I

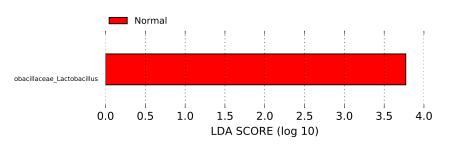


Figure: Deblur+HOMD in Cervix

### LEfSe Results with PTB II

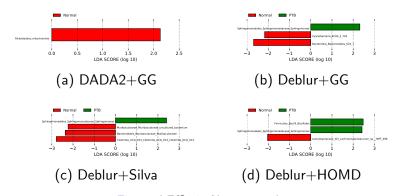


Figure: LEfSe in Neonate 1-day

### LEfSe Results with PTB III

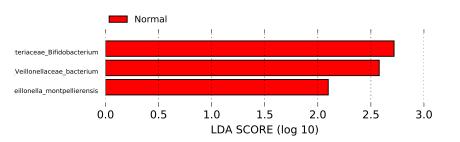
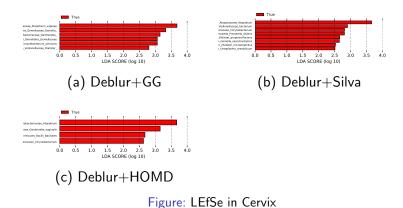


Figure: Deblur+Silva in Vagina

# LEfSe Results with Obesity I



# LEfSe Results with Obesity II

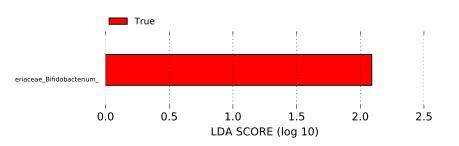


Figure: DADA2+GG in Vagina

### Discussion

#### To-do List

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

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