Microbiome Premature

Jaewoong Lee

Ulsan National Institute of Science and Technology jwlee230@unist.ac.kr

2020-09-25

Overview

- Introduction
- 2 Materials
- 3 Literature Survey I (Dominguez-Bello et al., 2016)
- 4 Literature Survey II (Fettweis et al., 2019)
- Methods
- 6 Results
- Proceedings
 References

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 (Preterm Birth)



Figure: Definitions of Premature (Tucker & McGuire, 2004)

∴ Hence, in this study,

• Premature: < 37 weeks

• Normal: \geq 37 weeks

Materials

16S rRNA Sequencing

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

- Train/Test data
 - Helixco: Data collected by Helixco
- Validate data
 - EBI (European Bioinformatics Institute): Data collected by Dominguez-Bello et al., 2016
 - HMP (Human Microbiome Project): Data collected by Fettweis et al., 2019

Table: Metadata of Data

Data	Participants	Samples	Remarks
Helixco	24	107	-
EBI	18	1016	Only Normal
HMP	1572	9205	Only Premature

Literature Survey I (Dominguez-Bello et al., 2016)

BRIEF COMMUNICATIONS



Partial restoration of the microbiota of cesarean-born infants via vaginal microbial transfer

Maria G Dominguez-Bello^{1,2}, Kassandra M De Jesus-Laboy², Nan Shen³, Laura M Cox¹, Amnon Amir⁴, Antonio Gonzalez⁴, Nicholas A Bokulich¹, Se Jin Song^{4,5}, Marina Hoashi^{1,4}, Juana I Rivera-Vinas⁷, Keimari Mendez⁷, Rob Knight^{4,8} & Jose C Clemente^{3,9} estimated 15% of births that require C-section delivery to protect the health of the mother or baby 11 .

Here we exposed C-section-delivered infants to their maternal vaginal fluids at birth and longitudinally determined the composition of their microbiota to assess whether it developed more similarly to vaginally born babies than to unexposed C-section-delivered infants. We collected samples from 18 infants and their mothers, including 7 born vaginally and 11 delivered by scheduled C-section, of which four were exposed to the maternal vaginal fluids at birth (Supplementary Table 1). Birely, the microbial restoration procedure, or vaginal microbial transfer, consists of incubating sterile gauze in the vagina of moth-

EBI Data II

- Study Objectives
 - Compare Vaginally vs. Cesarean-section (C-section)
 - 2 Restore the microbiota of C-section
- Microbial restoration procedure
 - Measure maternal vaginal pH
 - Put sterile gauze with saline solution in vagina for 1 hour
 - Swab the infant with the gauze
- Sample collection procedure
 - Sample at right after birth, day 3 and weekly for the first month
 - Sample from oral, forehead, arm, foot and anal
- Notable Methods/Results
 - Using distance methods: e.g. UniFrac distance, Hamming distance

Literature Survey II (Fettweis et al., 2019)

HMP Data I





OPEN

The vaginal microbiome and preterm birth

Jennifer M. Fettweis 1.23, Myrna G. Serrano 1.3, J. Paul Brooks 3.4, David J. Edwards 3.5, Philippe H. Girerd 3.4, Hardik I. Parikh¹, Bernice Huang¹, Tom J. Arodz 3.6, Laahirie Edupuganti 1.3, Abigail L. Glascock³, Jie Xu³.8.9, Nicole R. Jimenez 1.3, Stephany C. Vivadelli 1.3, Stephen S. Fong 3.30, Nihar U. Sheth¹, Sophonie Jean¹, Vladimir Lee¹, Yahya A. Bokhari6, Ana M. Lara¹, Shreni D. Mistry¹, Robert A. Duckworth III¹, Steven P. Bradley¹, Vishal N. Koparde¹, X. Valentine Orenda 0¹, Sarah H. Milton², Sarah K. Rozycki¹², Andrey V. Matveyev¹, Michelle L. Wright 0¹3.14,15, Snehalata V. Huzurbazar¹6, Eugenie M. Jackson¹6, Ekaterina Smirnova 0³1.8, Jonas Korlach¹9, Yu-Chih Tsai 0³9, Molly R. Dickinson¹, Jamie L. Brooks¹, Jennifer I. Drake¹, Donald O. Chaffin²0, Amber L. Sexton²0, Michael G. Gravett²0.2¹, Craig E. Rubens²0, N. Romesh Wijesooriya°, Karen D. Hendricks-Muñoz 8.8, Kimberly K. Jefferson¹³, Jerome F. Strauss III²³ and Gregory A. Buck 0³13.6*

HMP Data II

- Study Objectives
 - Predicting & Preventing premature
 - Report community resources
 - Provide an analysis of the longitudinal, comprehensive, multi-omic profiling of vaginal samples
- Sample collection Procedure
 - 1 Premature birth vs. Matched normal birth
 - 2 Ethnically diverse cohort
- Notable Methods/Results
 - Imitate figures

HMP Data III

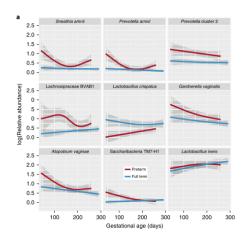


Figure: Microbiome Composition during Pregnancy

Methods

Qiime 2 Workflow

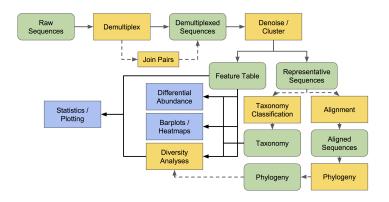


Figure: QIIME 2 workflow (Bolyen et al., 2019; Mandal et al., 2015; McDonald et al., 2012)

Filitering with Quality Score I

Drawback between:

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

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

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

Filitering with Quality Score II

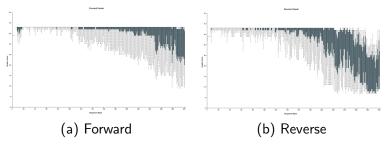


Figure: Sequence Quality Plot from Helixco Data

Maximum Length: 265

Filitering with Quality Score III

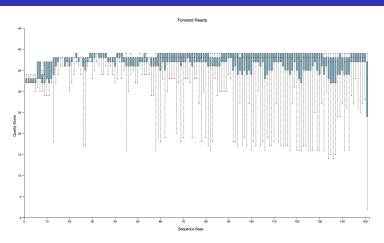
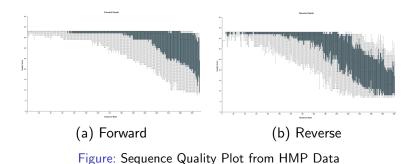


Figure: Sequence Quality Plot from EBI

Maximum Length: 150

Filitering with Quality Score IV



Maximum Length: 226

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)

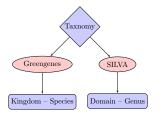


Figure: Taxonomy Classifications

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

Mothur



Figure: Mothur

Note: Still in progress

t-distributed Stochastic Neighbor Embedding (t-SNE)

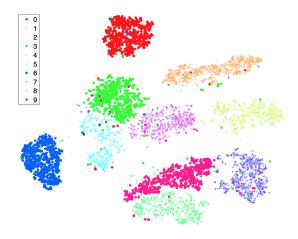


Figure: t-SNE with handwritten data (Maaten & Hinton, 2008)

- 4 ロ ト 4 週 ト 4 夏 ト 4 夏 ト 9 Q Q

26 / 50

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)

Results

t-SNE for Brief Information I

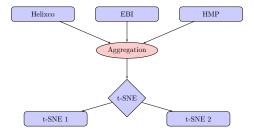


Figure: Workflow of t-SNE for Brief Information

t-SNE for Brief Information II

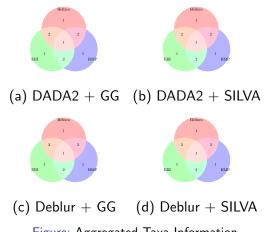


Figure: Aggregated Taxa Information

Note: need to be updated

t-SNE for Brief Information III





(a) DADA2 + GG (b) DADA2 + SILVA





(c) Deblur + GG (d) Deblur + SILVA

Figure: t-SNE for Brief Information

t-SNE with Site I

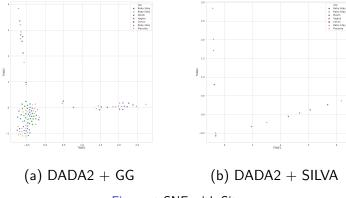
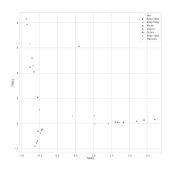
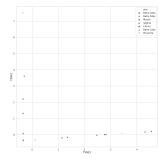


Figure: t-SNE with Site

t-SNE with Site II





(c) Deblur + GG

(d) Deblur + SILVA

Figure: t-SNE with Site

t-SNE with Site + Premature I

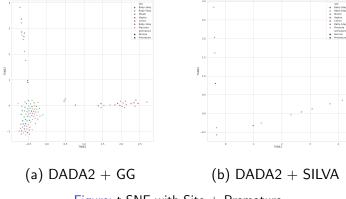
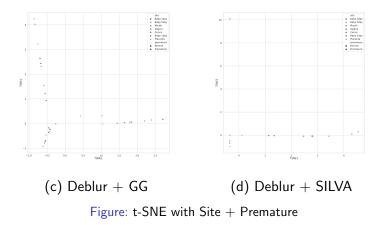


Figure: t-SNE with Site + Premature

t-SNE with Site + Premature II



Histogram with Clinical Information

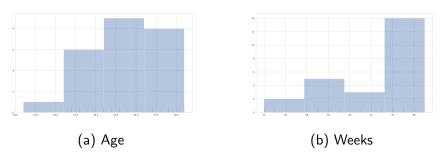


Figure: Histogram with Clinical Information

Random Forest Classifier I

Input Data was treated with **Deblur** and **SILVA**.

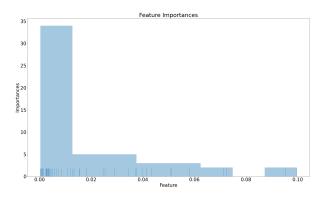


Figure: Feature Importance derived by Random Forest Classifier

Jaewoong Lee (UNIST)

Random Forest Classifier II

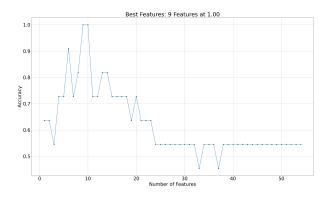


Figure: Number of Features vs. Accuracy

Random Forest Classifier III

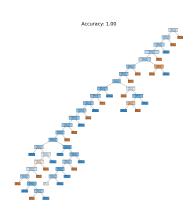


Figure: Random Forest Classifier

Random Forest Classifier IV

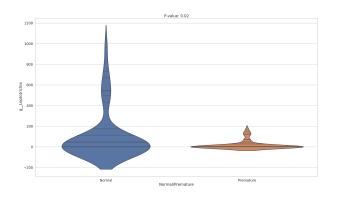


Figure: Violin Plot of Leptotrichia

Fusobacteriota Fusobacteriia Fusobacteriales Leptotrichiaceae Leptotrichia

Proceedings

Yields I

- t-SNE with databases
- Random Forest Classifier

Requirements I

- More data
- Metadata for databases
- Mothur pipeline

Expectations I

Classifier result (Statistical values)

References I

- Amir, A., McDonald, D., Navas-Molina, J. A., Kopylova, E., Morton, J. T., Xu, Z. Z., . . . others (2017). Deblur rapidly resolves single-nucleotide community sequence patterns. *MSystems*, *2*(2).
- Bolyen, E., Rideout, J. R., Dillon, M. R., Bokulich, N. A., Abnet, C. C., Al-Ghalith, G. A., . . . Caporaso, J. G. (2019). Reproducible, interactive, scalable and extensible microbiome data science using qiime 2. *Nature Biotechnology*, *37*(8), 852-857. Retrieved from https://doi.org/10.1038/s41587-019-0209-9 doi: 10.1038/s41587-019-0209-9
- Callahan, B. J., McMurdie, P. J., Rosen, M. J., Han, A. W., Johnson, A. J. A., & Holmes, S. P. (2016). Dada2: high-resolution sample inference from illumina amplicon data. *Nature methods*, *13*(7), 581–583.

References II

- DeSantis, T. Z., Hugenholtz, P., Larsen, N., Rojas, M., Brodie, E. L., Keller, K., . . . Andersen, G. L. (2006). Greengenes, a chimera-checked 16s rrna gene database and workbench compatible with arb. *Applied and environmental microbiology*, 72(7), 5069–5072.
- Dominguez-Bello, M. G., De Jesus-Laboy, K. M., Shen, N., Cox, L. M., Amir, A., Gonzalez, A., ... others (2016). Partial restoration of the microbiota of cesarean-born infants via vaginal microbial transfer. *Nature medicine*, *22*(3), 250.
- Fettweis, J. M., Serrano, M. G., Brooks, J. P., Edwards, D. J., Girerd, P. H., Parikh, H. I., . . . others (2019). The vaginal microbiome and preterm birth. *Nature medicine*, *25*(6), 1012–1021.

- Gihawi, A., Rallapalli, G., Hurst, R., Cooper, C. S., Leggett, R. M., & Brewer, D. S. (2019). Sepath: benchmarking the search for pathogens in human tissue whole genome sequence data leads to template pipelines. *Genome biology*, 20(1), 1–15.
- Gill, S. R., Pop, M., DeBoy, R. T., Eckburg, P. B., Turnbaugh, P. J., Samuel, B. S., ... Nelson, K. E. (2006). Metagenomic analysis of the human distal gut microbiome. *science*, *312*(5778), 1355–1359.
- Hunter, J. D. (2007). Matplotlib: A 2d graphics environment. *Computing in science & engineering*, *9*(3), 90–95.
- Janda, J. M., & Abbott, S. L. (2007). 16s rrna gene sequencing for bacterial identification in the diagnostic laboratory: pluses, perils, and pitfalls. *Journal of clinical microbiology*, 45(9), 2761–2764.
- Maaten, L. v. d., & Hinton, G. (2008). Visualizing data using t-sne. Journal of machine learning research, 9(Nov), 2579–2605.

References IV

- Mandal, S., Van Treuren, W., White, R. A., Eggesbø, M., Knight, R., & Peddada, S. D. (2015). Analysis of composition of microbiomes: a novel method for studying microbial composition. *Microbial ecology in health and disease*, *26*(1), 27663. doi: 10.3402/mehd.v26.27663
- McDonald, D., Clemente, J. C., Kuczynski, J., Rideout, J. R., Stombaugh, J., Wendel, D., . . . Caporaso, J. G. (2012). The biological observation matrix (biom) format or: how i learned to stop worrying and love the ome-ome. *GigaScience*, 1(1), 7. doi: 10.1186/2047-217X-1-7
- McKinney, W., et al. (2011). pandas: a foundational python library for data analysis and statistics. *Python for High Performance and Scientific Computing*, 14(9).
- Mignard, S., & Flandrois, J.-P. (2006). 16s rrna sequencing in routine bacterial identification: a 30-month experiment. *Journal of microbiological methods*, 67(3), 574–581.

References V

- Olsen, G. J., & Woese, C. R. (1993). Ribosomal rna: a key to phylogeny. *The FASEB journal*, 7(1), 113–123.
- Pedregosa, F., Varoquaux, G., Gramfort, A., Michel, V., Thirion, B., Grisel, O., . . . others (2011). Scikit-learn: Machine learning in python. the Journal of machine Learning research, 12, 2825–2830.
- Pruesse, E., Quast, C., Knittel, K., Fuchs, B. M., Ludwig, W., Peplies, J., & Glöckner, F. O. (2007). Silva: a comprehensive online resource for quality checked and aligned ribosomal rna sequence data compatible with arb. *Nucleic acids research*, *35*(21), 7188–7196.
- Quast, C., Pruesse, E., Yilmaz, P., Gerken, J., Schweer, T., Yarza, P., ... Glöckner, F. O. (2012). The silva ribosomal rna gene database project: improved data processing and web-based tools. *Nucleic acids research*, *41*(D1), D590–D596.
- Tucker, J., & McGuire, W. (2004). Epidemiology of preterm birth. *Bmj*, 329(7467), 675–678.

References VI

- Turnbaugh, P. J., Ley, R. E., Hamady, M., Fraser-Liggett, C. M., Knight, R., & Gordon, J. I. (2007). The human microbiome project. *Nature*, 449(7164), 804–810.
- Virtanen, P., Gommers, R., Oliphant, T. E., Haberland, M., Reddy, T., Cournapeau, D., . . . others (2020). Scipy 1.0: fundamental algorithms for scientific computing in python. *Nature methods*, 17(3), 261–272.
- Waskom, M., Botvinnik, O., Ostblom, J., Gelbart, M., Lukauskas, S., Hobson, P., ... Brian (2020, April). mwaskom/seaborn: v0.10.1 (april 2020). Zenodo. Retrieved from https://doi.org/10.5281/zenodo.3767070 doi: 10.5281/zenodo.3767070