Introduction

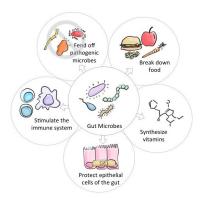
A Novel Graph-Theoretical Approach for Identifying Inter-correlations and Functional Pathways in Microbiome Data

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What do our gut microbes do for us?

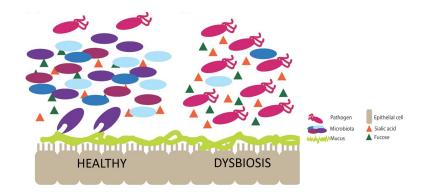


 The human microbiome represents a vastly complex ecosystem that is tightly linked to our development, physiology, and health.



Introduction 000000

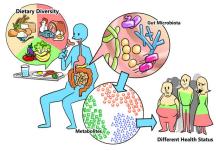
Microbiome, key controller of health



 Several studies highlighted the role of microbiome in human health and disease.

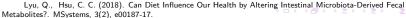


Gut microbiome and Disease



Examples of disease associated with altered gut microbiota diversity:

- Inflammatory bowel disease (IBD)
- Parkinson's disease
- Depression and Anxiety
- Diabetes and obesity
- Colorectal cancer
- Autism
- Irritable bowel syndrome
- C.difficile infection



Case Study of Human Microbiome

Motivation

- Interactions between species shape its host microbial community.
- These interactions are important to understand how microbiome related diseases.
- Metagenomics-based studies of microbiome
 - Characterize the composition of species in a microbiome
 - Complex co-occurrence patterns
- Host and microbiota do not operate alone
 - The interactions between the host and bacterial systems need to be considered through a more holistic, systematic approach

Literature Review

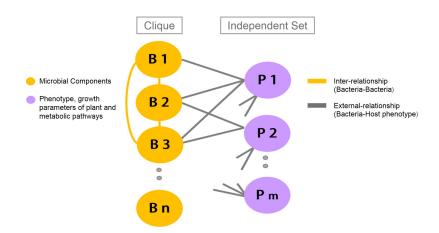
- Multi high-throughput methods for studying the human microbiome are available. [Buschart, A et al., 2017, Borenstein, E et al., 2013]
- With the advanced high throughput technques, it calls for the development of systems-level methods in microbiome study. [Borenstein, E et al., 2013]
- Difficult to identify the actual contribution of microbial traits to human physiology with metagenomics study alone. [Greenblu, S et.al 2013, Buschart, A et al., 2017]
- Preliminary systems-level modeling show promise but many challenges remain. [Greenblum, S et al., 2017]

Goal of the project

Developing a comprehensive framework that integrates microbe-microbe and microbe-host interactions:

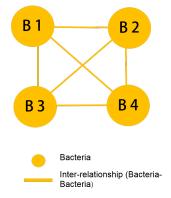
• To create a graph theoretic pipeline for identifying bacterial inter-relations and functional pathways in microbial ecosystem.

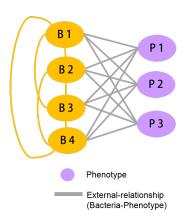
Proposed Solution: Split Graph Model



Advantages of Split Graph

Two different relationships in split graph

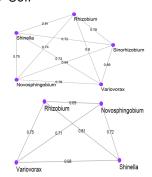




Case Study of Human Microbiome

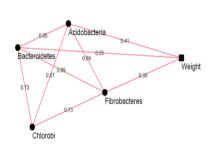
Split graphs in multiple ecosystems

Soil



Healthy (upper) vs. Unhealthy(bottom)

Fish



FCB group with Acidobacteria in fish gut

Microbiome in Human gut

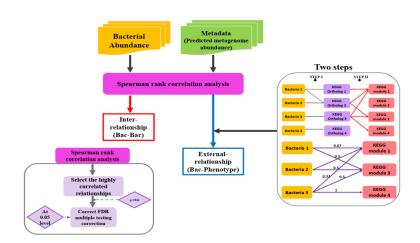
- Human gut microbiome data
- Bacterial relative abundance (family level)
 - All samples from a Korean population

Control group	Crohn's Disease patient (mucosal tissue)	Crohn's Disease patient (fecal)
10	36	9

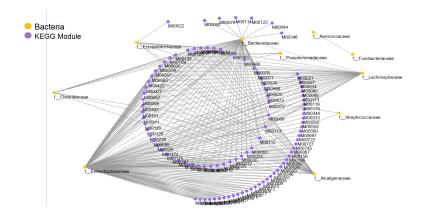
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Case Study of Human Microbiome

Split Graph Analysis



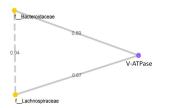
Integrated Correlation Network (CDS)

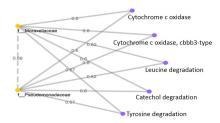


Case Study of Human Microbiome

Introduction

Split graph in Crohn's disease (stool and tissue)





- All Bacteria are reported to increased risk of IBD [Hasler et al., 2016]
- V-ATPase is involved in a number of human diseases [Hinton et al., 2009]
- Modules/Pathways in CDT are closely associated with energy metabolism, amino acid degradation, and energy deficiency (low ATP levels).



Results

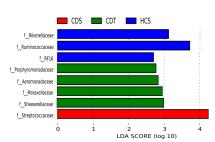
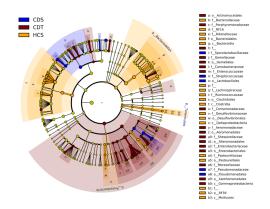


Fig: Differentially abundant bacterial taxa

TABLE I: Inter-bacteria correlations in all sample groups

FAerococcaceae fFusobacteriaceae HCS fPrevotellaceae fRF16	R ² 0.94 0.98 0.98 0.98 0.97 0.81
f_Aerococcaeae f_Fusobacteriaceae HCS f_Prevotellaceae f_RF16 f_Bacillaceae f_Staphylococcaeae f_Rikenellaceae f_Ruminococcaeae f_Ruminococcaeae	0.98 0.98 0.98 0.97 0.81
HCS f_Prevotellaceae f_RF16 (f_Bacillaceae f_Staphylococcaceae (f_Rikenellaceae f_Rminococcaceae (CDT f_Aeromonadaceae f_Shewanellaceae (0.98 0.98 0.97 0.81
f_Bacillaceae f_Staphylococcaeae f_Rikenellaceae f_Ruminococcaeae CDT f_Aeromonadaceae f_Shewanellaceae f_Shewanellaceae	0.98 0.97 0.81
f_Rikenellaceae f_Ruminococcaceae (CDT f_Aeromonadaceae f_Shewanellaceae (0.97
CDT f_Aeromonadaceae f_Shewanellaceae (0.81
J	
f RA059 f Syntrophobacteraceae (77
j_bilos) j_bynirophobacieraceae	1.12
f_Planococcaceae f_Gallionellaceae (0.72
f_Porphyromonadaceae f_Pseudomonadaceae (0.71
f_Carnobacteriaceae f_Streptococcaceae (0.70
f_Moraxellaceae f_Pseudomonadaceae (0.68
f_Microbacteriaceae f_Spirochaetaceae (0.68
f_BA059 f_Gallionellaceae (0.68
f_Peptococcaceae f_Alteromonadaceae (0.68
f_Peptococcaceae f_Sinobacteraceae (0.68
f_Nitrospiraceae f_Syntrophobacteraceae (0.68
f_Procabacteriaceae f_Halomonadaceae (0.68
f_Veillonellaceae f_Pseudomonadaceae -(0.66
f_Porphyromonadaceae f_Shewanellaceae	0.66

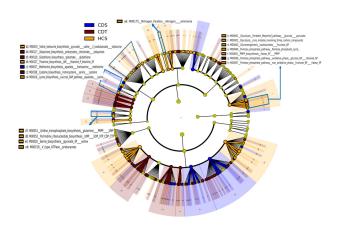
Identify candidate taxonomic biomarkers between Control and CD groups



 Some of the bacterial taxa identified by the split graph are candidate biomarkers.



Identify candidate functional biomarkers between Control and CD group



 Some of the functional features(phenotype) identified by the split graph are also candidate biomarkers.



Key findings

- The split graph model easily captured the structure and function of microbial communities in Crohn's disease patients and healthy individuals.
- The taxonomic and metabolic biomarkers in different conditions were also highlighted in split graph analysis.

Conclusion

- Provided new insights into identifying two distinct relationships with split graph analysis.
- Microbial elements within the relationship in CDT are closely associated with the metabolic mechanisms that lead to the Crohn's disease.
- Identified the distinct structure or function of microbial community in different health conditions.
- Pipeline was extended to allow additional microbial study by overlaying of multi-omics data.