



香港中文大學
The Chinese University of Hong Kong



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Network Analyses in the Microbiome

Joint Graduate Seminar
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A complex network diagram with numerous nodes of varying sizes and colors (grey, white, light blue) connected by thin grey lines, forming a dense web. The word "overview" is centered in the left half of the slide.

overview

1. Microbiology meets network theory
2. Basic network theory concepts
3. Network modelling methods and examples
 - Networks from metagenomics
 - Networks from metatranscriptomics
 - Networks from metaproteomics
 - Networks from metabolomics
 - Networks from multi-omics
4. Challenges
5. Summary



Microbiology meets network theory



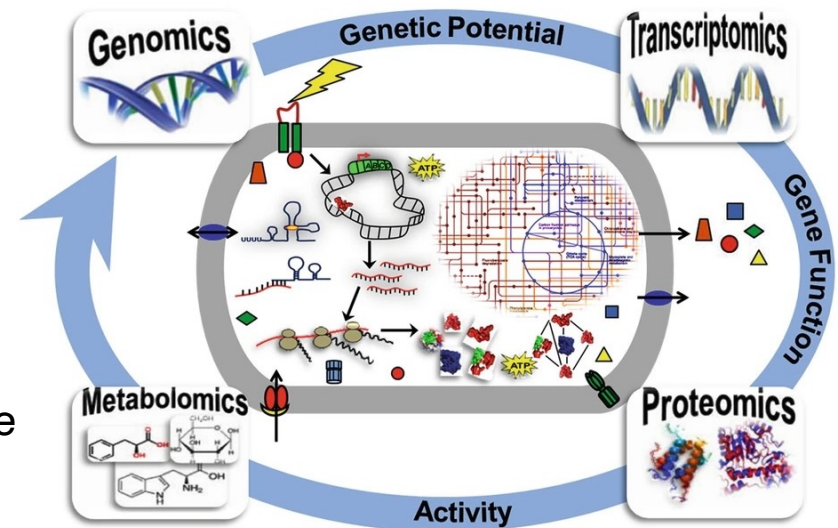
Microbiomes are complex microbial communities. Interactions influence the structure and the function of such communities.

More than 99% of microorganisms in nature could not be cultured alone *in vitro*!

High-throughput Omics technologies:

- high volume of data generation
- high quality of data interpretation
- fairly acceptable cost

Effective computational analysis techniques are necessary!



Dey, D., et al. Integrating Omics Technologies to Understand Microbial Systems. 2021

Microbiology meets network theory

What can we do with a Network?

Obtain info that we wouldn't get by analysing individual components!

- Important protein calls HUBS (have a lot of connections-degree)
- Community of proteins (belongs to the same pathway or Complex)
- Simulate in silico the bacteria behaviour
- Formulate new hypothesis about function

Communities

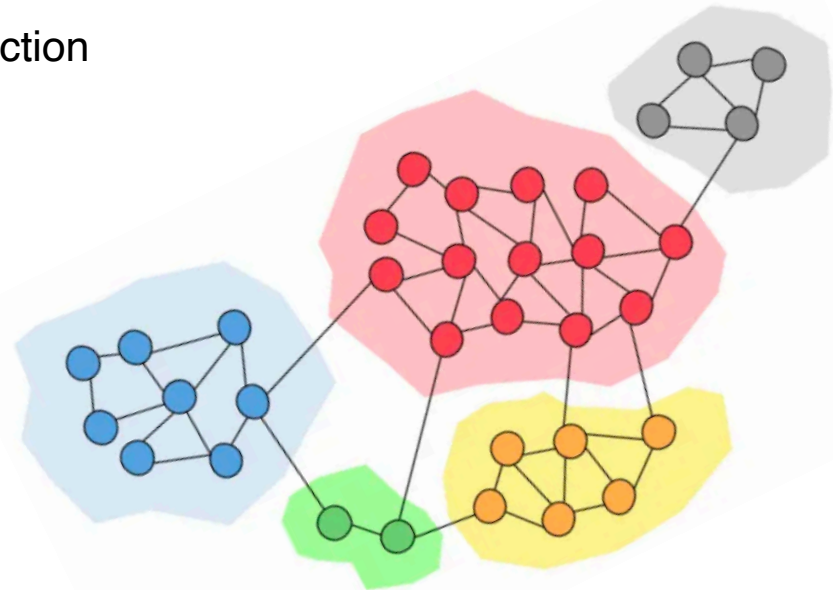
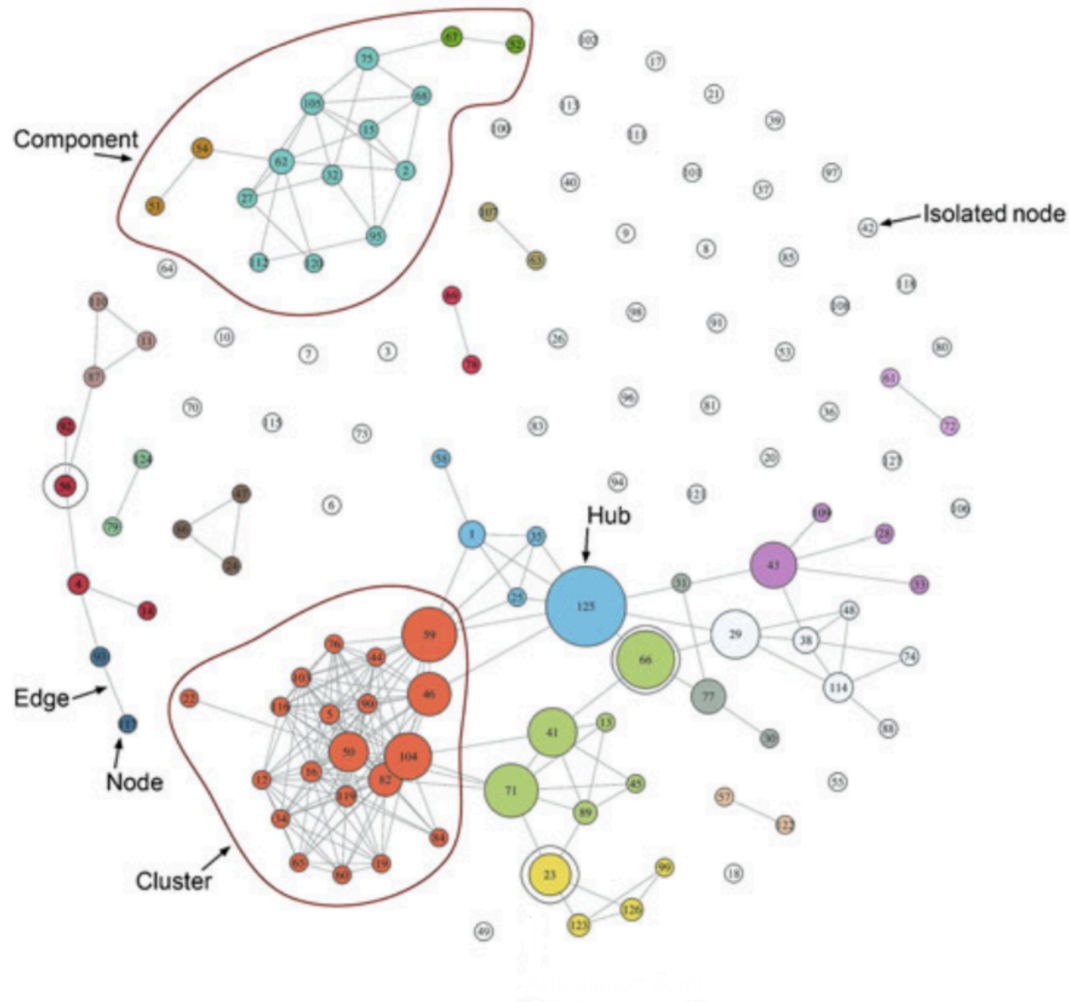


Image by Thamindu Dilshan Jayawickrama



Basic network theory concepts

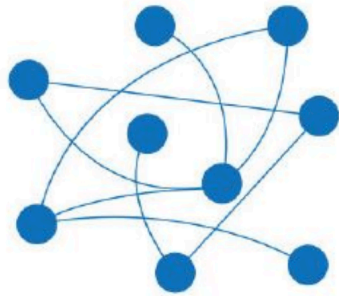


Node
Edge
Hub
Component
Module
Degree

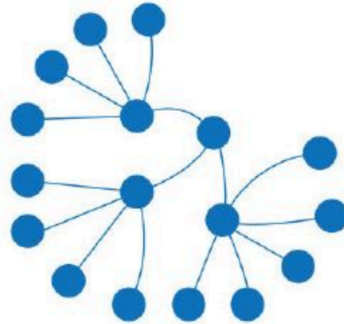
Image by Jean-Claude Walser



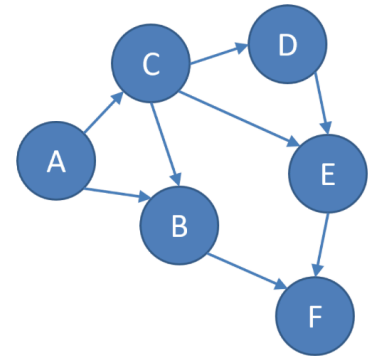
Basic network theory concepts



Random network



Scale-free network



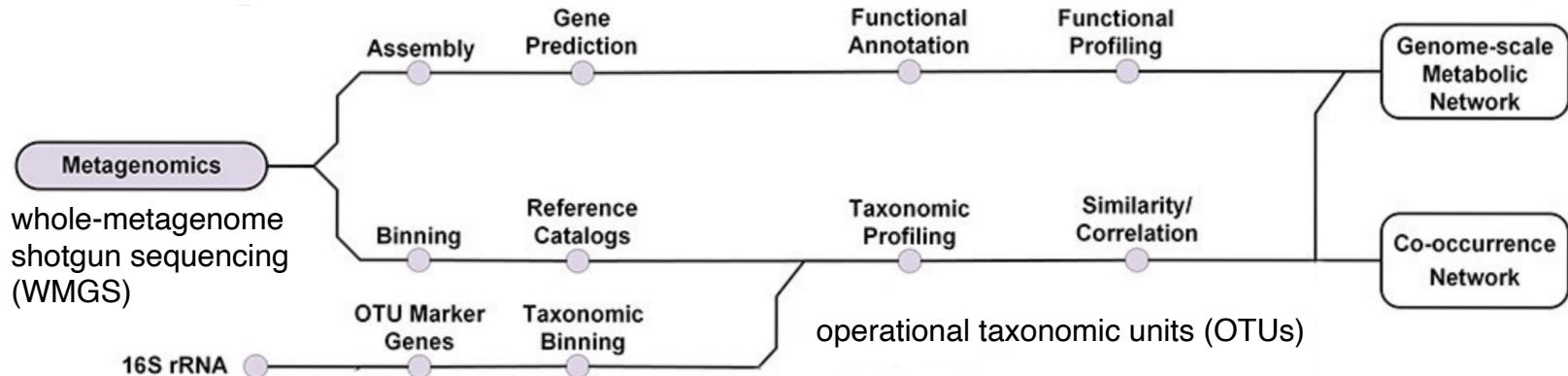
Bayesian network

Scale-free network: most of nodes have only a few connections to other nodes, whereas some hubs are connected to many other nodes in the network. Real-world networks are often claimed to be scale free.

Bayesian network: a probabilistic graphical model that represents a set of variables and their conditional dependencies via a directed acyclic graph (DAG).

Networks from metagenomics

Network modelling methods:



Tools: UPARSE, MetaPhlan4, mOTUs2, QIIME2, HUMAnN3

Taxonomic table

	Sample ID			
	S1	S2	S3	S4
OTU_1				
OTU_2				
OTU_3				
OTU_4				
OTU_5				
⋮				
OTU_n				

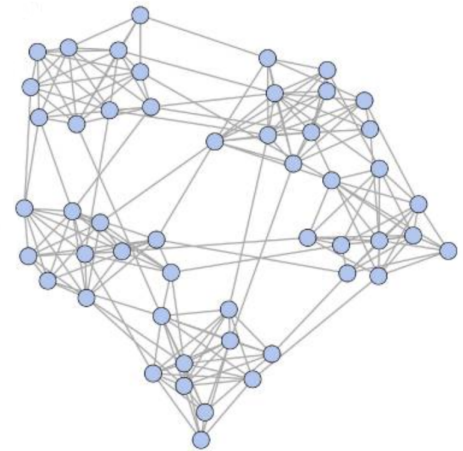
Functional table

	Sample ID			
	S1	S2	S3	S4
KO_01				
KO_02				
KO_03				
KO_04				
KO_05				
⋮				
KO_n				

Constructing Networks



Dissimilarity-Based Methods
Correlation-Based Methods
Regression-Based Methods
Probabilistic Graphical Models



Mehdi Layeghifard, et al. Trends Microbiol. 2017 Mar;25(3):217-228.

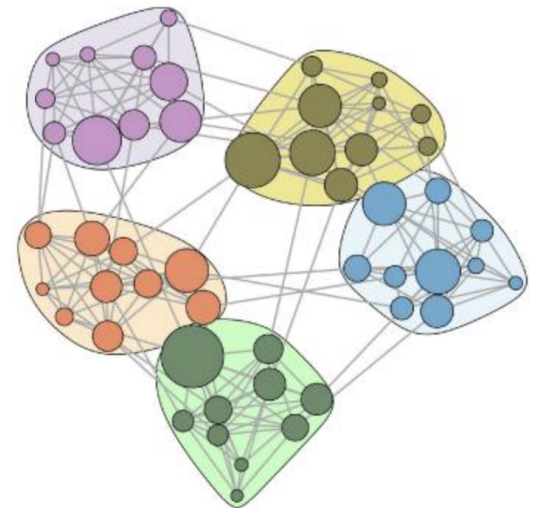
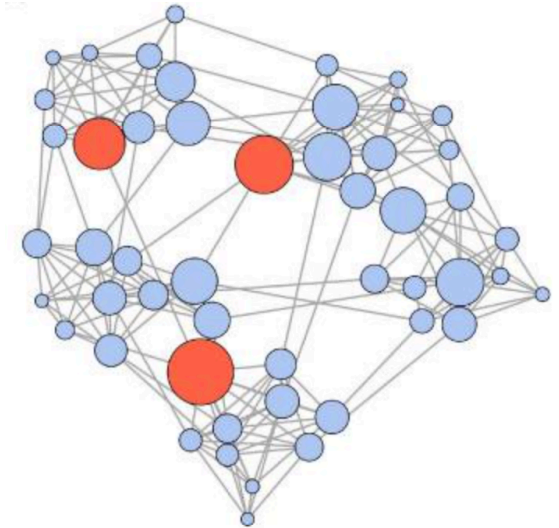
Networks from metagenomics

Detecting Hub node from Networks

important nodes or hubs:
most influential member in the microbial community
most essential microbe for community stability
organism responsible for disease transmission

Detecting Biologically Important Clusters from Networks

Clusters/modules: provides information about the local interaction patterns in the network and their contribution to the overall structure, connectivity, and function of the network



Mehdi Layeghifard, et al. Trends Microbiol. 2017 Mar;25(3):217-228.



Networks from metagenomics

Networks at the taxonomic level

Co-occurrence networks

151 soil samples



metagenome data analyses

2798 OTUs



correlation analysis

Co-occurrence networks
(296 nodes (OTUs) and 679 edges)

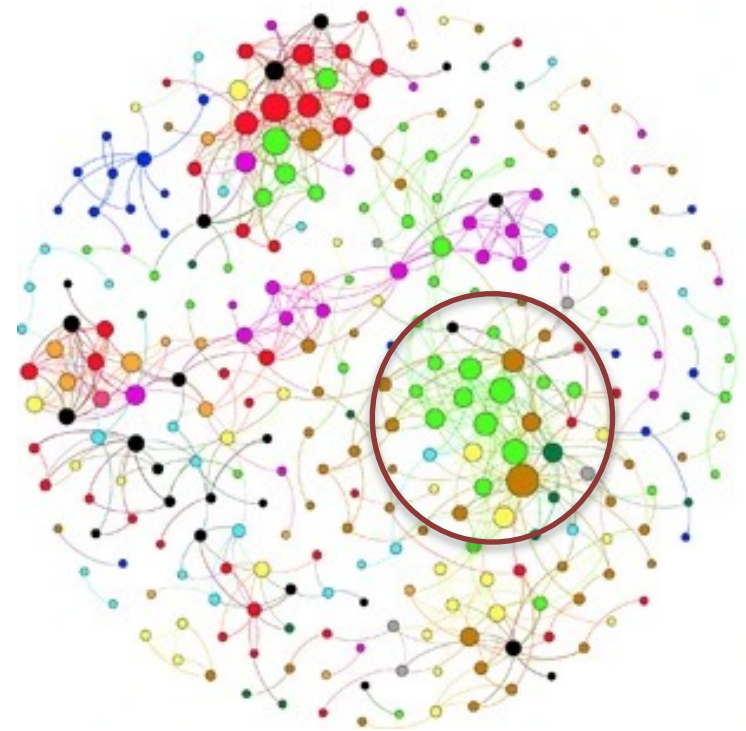
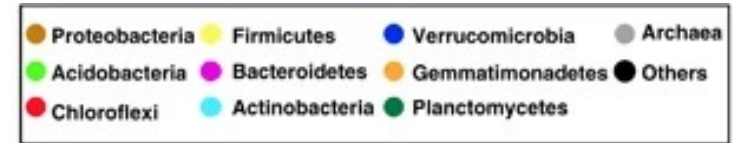


subsequent topological analysis

Modules



OTUs colored by taxonomy



co-occurrence networks in soil microbial communities

Albert Barberán, et al. ISME J. 2012 Feb; 6(2): 343–351.



Networks from metagenomics

Networks at the functional level

50 samples from 10 typical environments



260 ARG subtypes belonging to 18 ARG types
taxonomic abundance



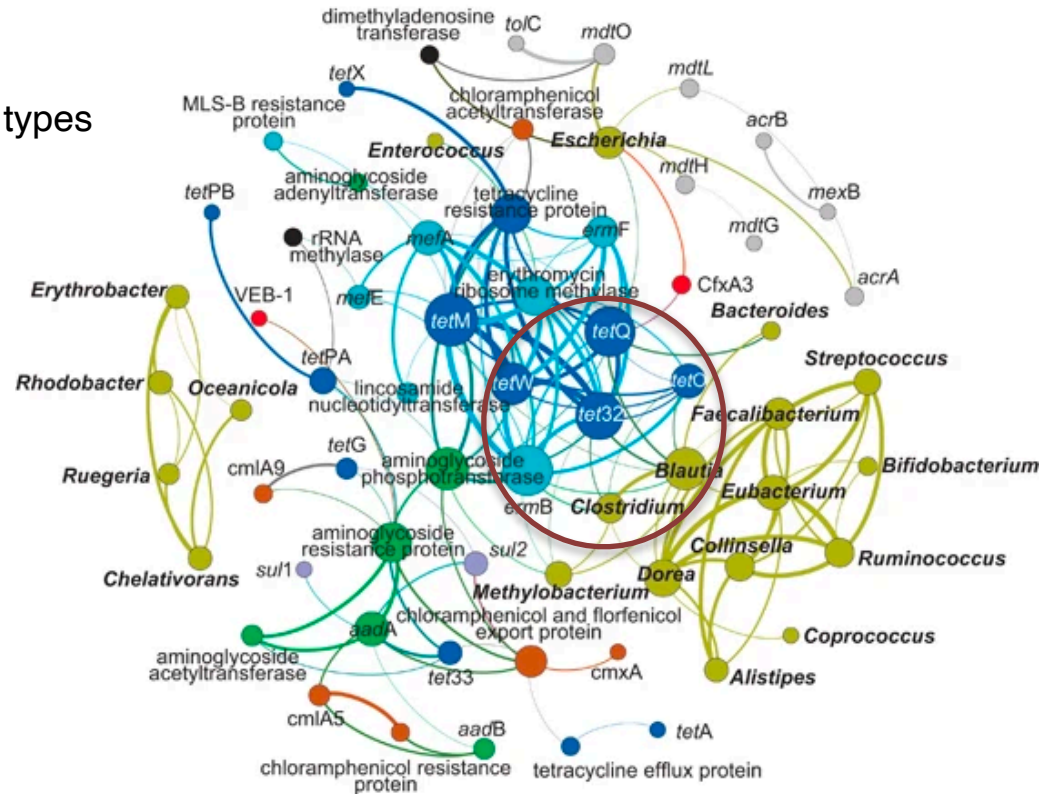
Significant Spearman's rank correlation
between the microbial diversity and the
ARGs diversity



Co-occurrence network



Nodes were coloured according to
ARG types and genus



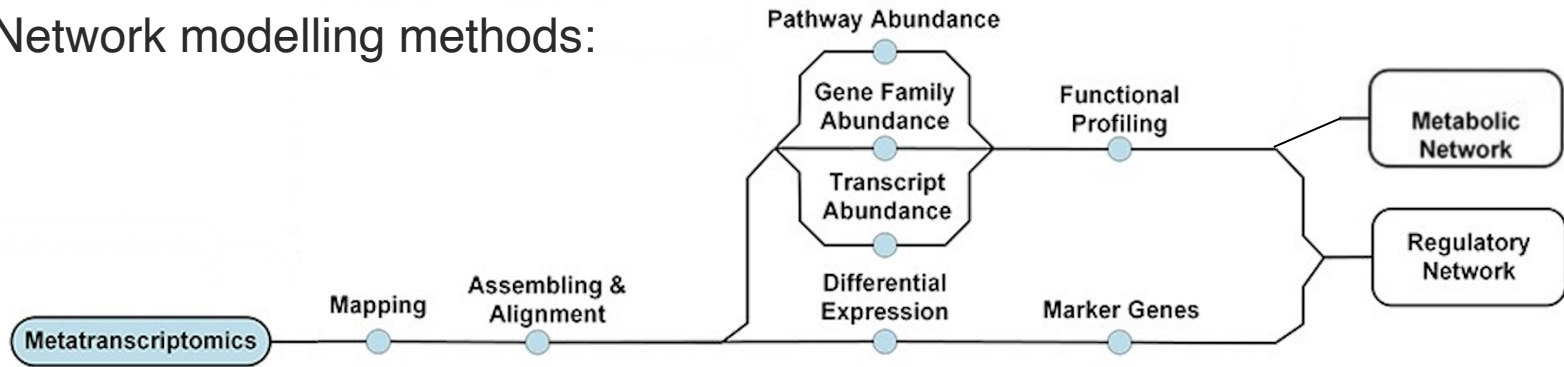
The Network analysis revealing the co-occurrence patterns between antibiotic resistance genes (ARGs) subtypes and microbial taxa in multiple environments.

Bing Li, et al. The ISME Journal, pages 2490–2502 (2015)



Networks from metatranscriptomics

Network modelling methods:



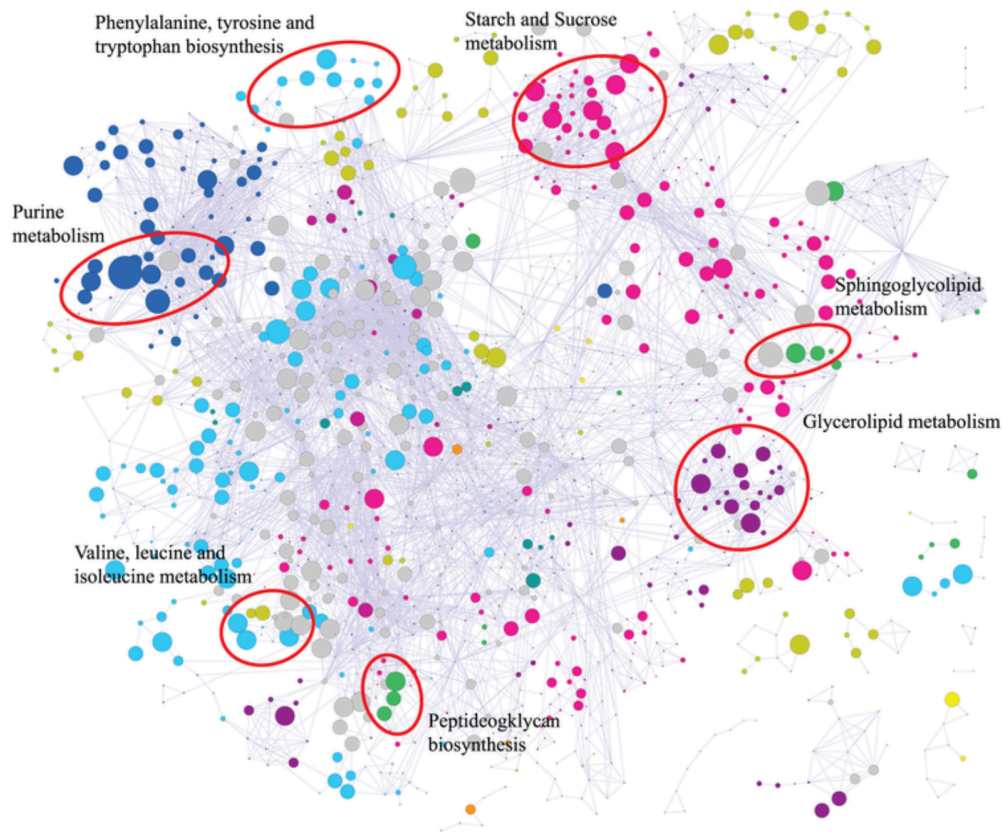
Tools: MetaTrans, SAMSA

More pertinent information on functional activity: metatranscriptomics can reveal details of genes that are transcriptionally active under specific conditions and time

Zhaoqian Liu, et al. Briefings in Bioinformatics, March 2021, Pages 1639–1655



Networks from metatranscriptomics



12 cecal and colon derived samples from mice

236,769 unique transcripts

relative expression of each gene family

Metabolic networks

Colour of node indicates functional category of enzyme as defined by KEGG superclasses

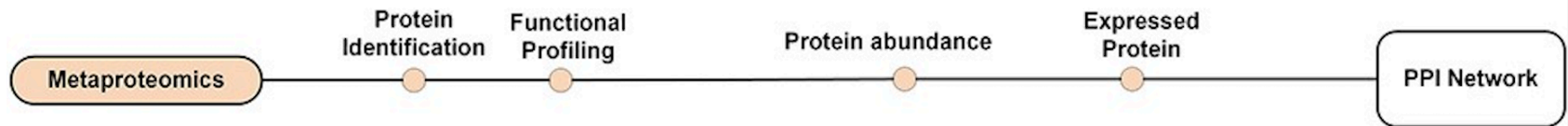
Metatranscriptome data mapped in the context of a global metabolic network

Galeb S, et al. Nature Microbiology volume 3, pages 356–366 (2018)



Networks from metaproteomics

Network modelling methods:

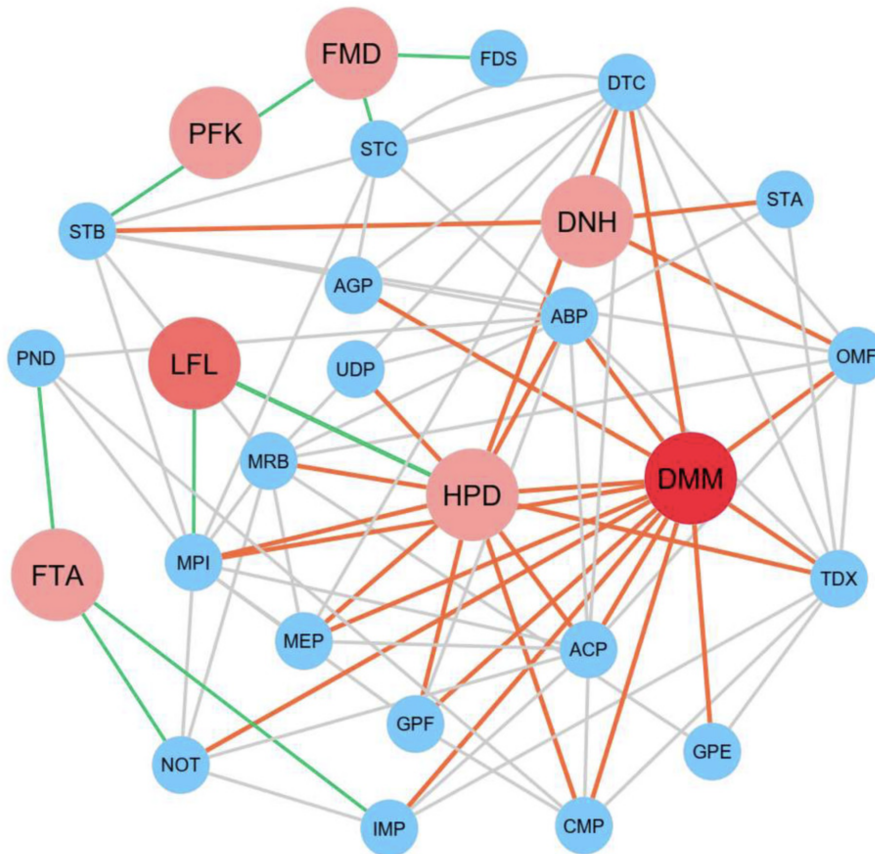


Tools: IdentiPy, Trans-Proteomic Pipeline, compleXView

Protein–protein interaction (PPI) networks describe physical and selective contacts that happen between pairs of proteins, in certain molecular regions and in a defined biological context.



Networks from metaproteomics



18 infants: 11 healthy controls and 7 patients with atopic dermatitis

18 fecal samples

49,973 annotated proteins out of 68,232 total proteins

PPI network

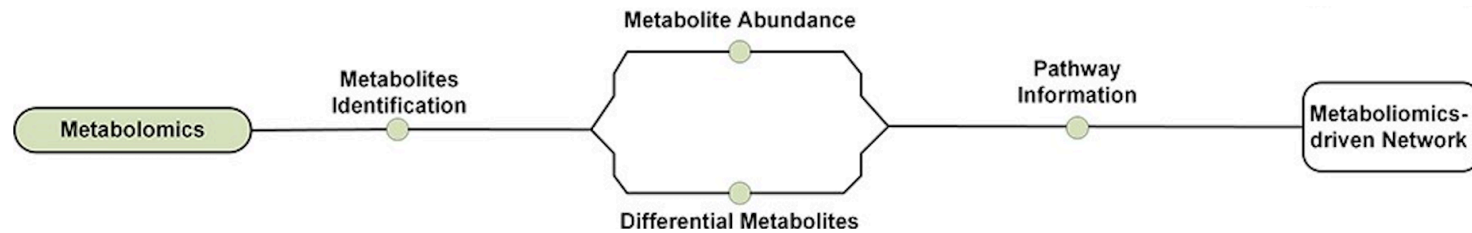
Reporter protein network involving ubiquinone and other quinone biosynthesis as well as energy supply.

Amornthep Kingkaw, et al. PeerJ. 2020 Sep 25;8:e9988.



Networks from metabolomics

Network modelling methods:



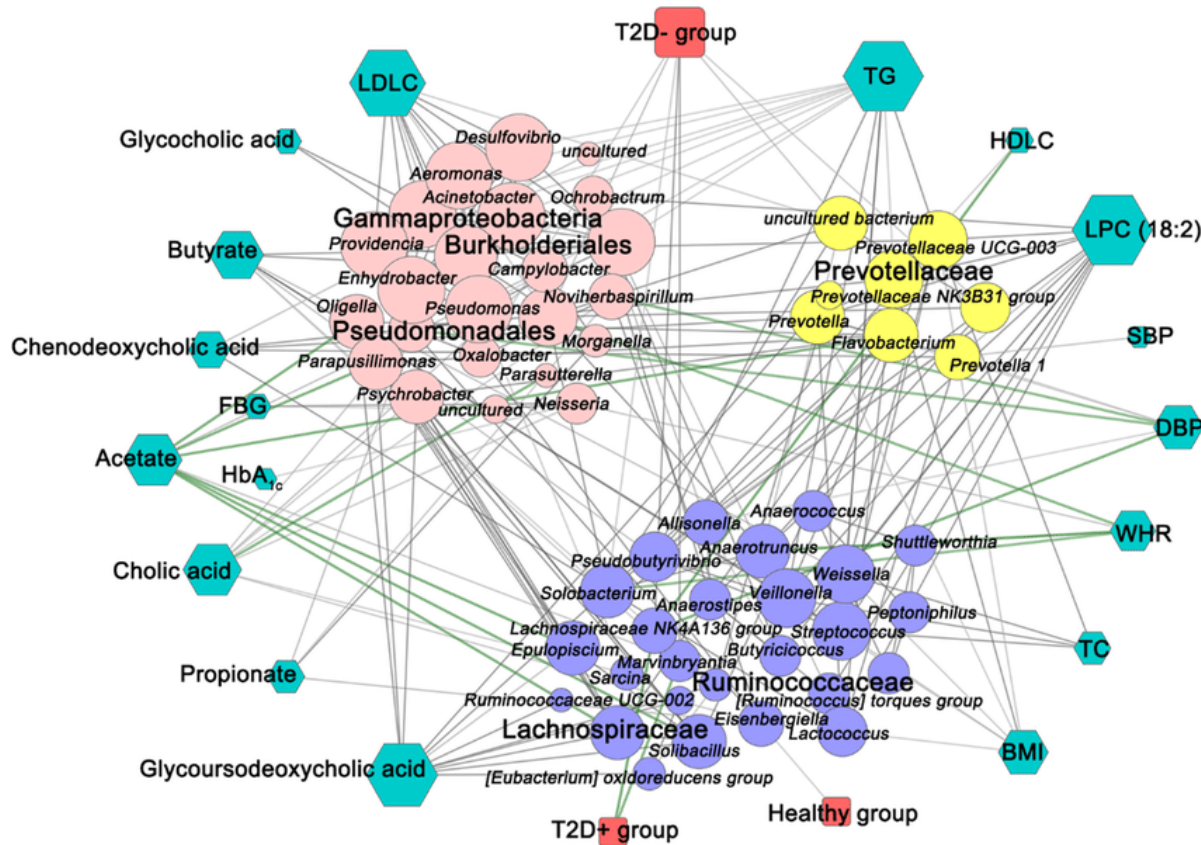
Tools: Pathos, MetaboAnalyst, Netome

Metabolomics-driven networks: consisting of nodes as metabolites and edges as metabolic reactions, provide a comprehensive description of a community's metabolic processes.

Zhaoqian Liu, et al. Briefings in Bioinformatics, March 2021, Pages 1639–1655



Networks from metabolomics



65 T2D patients:
49 with and 16 without
diabetic complications,
and 35 healthy controls

The correlation network between gut microbiota and metabolic traits in T2D and healthy groups

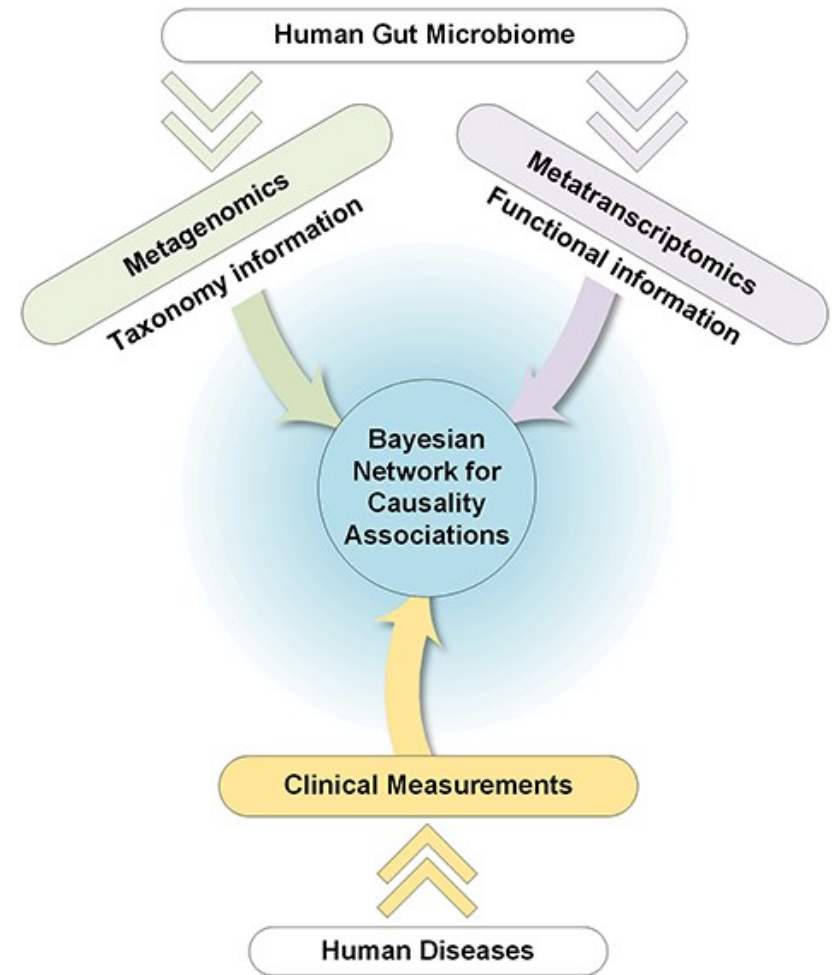
Lijuan Zhao, et al. Endocrine. 2019 Dec;66(3):526-537.



Networks from multi-omics

Each omics analysis has its limitation

1. Integrated metagenomic, metatranscriptomic and metaproteomic data to construct a community-wide metabolic network
2. Apply Bayesian networks, to elucidate the underlying molecular mechanisms of diseases by associating the microbial community, both at the level of taxonomy and functionality, with disease phenotypes and clinical measurements

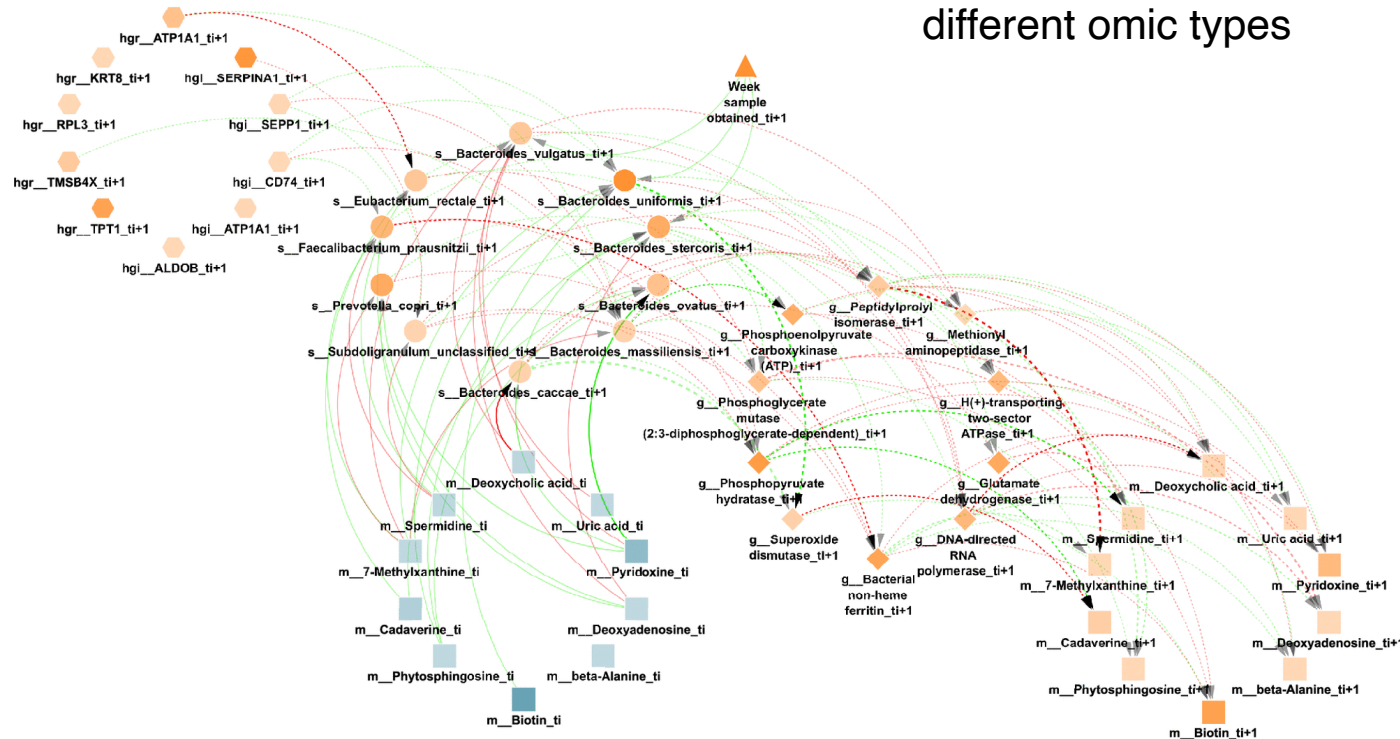


Zhaoqian Liu, et al. Briefings in Bioinformatics, March 2021, Pages 1639–1655



Networks from multi-omics

132 individuals over a year
profiled every 2 weeks, on average, for
different omic types



Dynamic Bayesian network (DBN): a Bayesian network (BN) which relates variables to each other over adjacent time steps.

Daniel Ruiz-Perez, et al. mSystems. 2021 Mar 30;6(2):e01105-20.



Challenges

- What to do with rare taxa?

The majority of taxa in sequencing data are only found in very few samples. This means that a large part of sequencing data consists of zeros.

- How to deal with environmental factors?

It is difficult to determine whether an edge in a microbial network is due to a common response to an environmental factor (or a third taxon) or represents a direct interaction between two taxa.

- How to evaluate microbial network construction in silico?

Evaluations are carried out to assess which tools infer the most accurate networks and to explore how sample number and other data properties affect tool performance.

- How well do microbial networks represent ecosystems?

Assuming that network inference is sufficiently accurate, can network properties such as modularity, and network density give useful information about the ecosystem under study?



Summary

- The microbiome is a complex system of microbes
- With the development of network theories and meta-omics data, network models are widely used to study microbial communities.
- Multi-omics data integration in networks provides a comprehensive view of microbial communities and their interactions with other factors.
- Challenges remain, including detecting rare microbes, incomplete genome annotation, and selecting appropriate network models.



References

Zhaoqian Liu, Anjun Ma, Ewy Mathé, Marlena Merling, Qin Ma, Bingqiang Liu, Network analyses in microbiome based on high-throughput multi-omics data, *Briefings in Bioinformatics*, Volume 22, Issue 2, March 2021

Kaiser, Thomas et al. “Network-based approaches for the investigation of microbial community structure and function using metagenomics-based data.” *Future microbiology* vol. 17 (2022): 621-631.

Jiang, Duo et al. “Microbiome Multi-Omics Network Analysis: Statistical Considerations, Limitations, and Opportunities.” *Frontiers in genetics* vol. 10 995. 8 Nov. 2019

Faust, Karoline. “Open challenges for microbial network construction and analysis.” *The ISME journal* vol. 15,11 (2021): 3111-3118.

Dey, D., Mekala, L.P., Mohammed, M. (2021). Integrating Omics Technologies to Understand Microbial Systems. In: Kumar, A., Kumar, R., Shukla, P., Pandey, M.K. (eds) *Omics Technologies for Sustainable Agriculture and Global Food Security Volume 1*. Springer, Singapore.

Barberán, Albert et al. “Using network analysis to explore co-occurrence patterns in soil microbial communities.” *The ISME journal* vol. 6,2 (2012): 343-51.

Li, Bing et al. “Metagenomic and network analysis reveal wide distribution and co-occurrence of environmental antibiotic resistance genes.” *The ISME journal* vol. 9,11 (2015): 2490-502. doi:10.1038/ismej.2015.59



References

Xiong, Xuejian et al. "Generation and analysis of a mouse intestinal metatranscriptome through Illumina based RNA-sequencing." *PloS one* vol. 7,4 (2012): e36009.

Kingkaw, Amornthep et al. "Analysis of the infant gut microbiome reveals metabolic functional roles associated with healthy infants and infants with atopic dermatitis using metaproteomics." *PeerJ* vol. 8 e9988. 25 Sep. 2020, doi:10.7717/peerj.9988

Zhao, Lijuan et al. "Comprehensive relationships between gut microbiome and faecal metabolome in individuals with type 2 diabetes and its complications." *Endocrine* vol. 66,3 (2019): 526-537.

Ruiz-Perez, Daniel et al. "Dynamic Bayesian Networks for Integrating Multi-omics Time Series Microbiome Data." *mSystems* vol. 6,2 e01105-20. 30 Mar. 2021

Layeghifard M, Hwang DM, Guttman DS. Disentangling Interactions in the Microbiome: A Network Perspective. *Trends Microbiol.* 2017 Mar;25(3):217-228. doi: 10.1016/j.tim.2016.11.008. Epub 2016 Dec 2. PMID: 27916383; PMCID: PMC7172547.



Thank you!

