

pepFunk: peptide-centric functional enrichment for metaproteomic gut microbiome data

useR2020!

Caitlin Simopoulos, PhD



caitsimop

csimopou@uottawa.ca

Postdoctoral Associate

Department of Biochemistry, Microbiology and Immunology

Ottawa Institute of Systems Biology

Faculty of Medicine, University of Ottawa, Ontario, Canada

1. Brief intro to the gut microbiome
2. Challenges with metaproteomics
3. Following the pepFunk methodology with real data
4. pepFunk as a Shiny app integrated with MetaLab

The gut microbiome is associated with both health and disease

The function (and dysfunction) of the gut microbiome can influence our health and well-being.

Some examples of diseases and conditions that have an association with the gut microbiome are:

- Immune-system-associated:
 - ▶ Inflammatory Bowel Disease (IBD)
 - ▶ Asthma
 - ▶ Multiple Sclerosis
- Metabolic disorders
 - ▶ Diabetes
 - ▶ Obesity
- Cardiovascular disease
- Mental health
 - ▶ Anxiety
 - ▶ Depression

Gut microbiomes are influenced by our environment

The composition of our microbiomes can be influenced by our genetics, but are mostly influenced by our environment and lifestyles.

Geographic location



Genes



Exercise level



Mode of delivery



Stress



Diet



Age



Medication



1

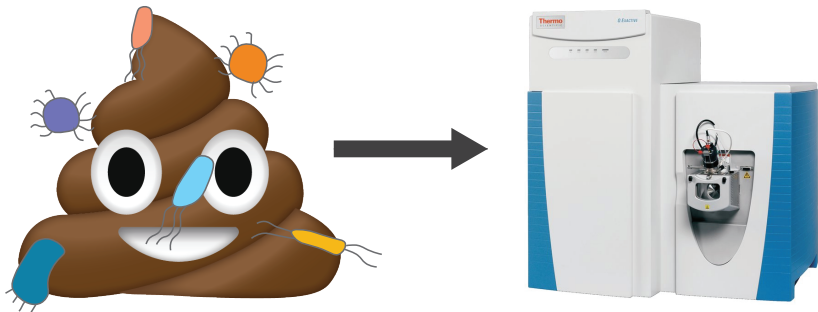
¹Figure inspired by Clarke et al 2019 Pharm Rev doi:<https://doi.org/10.1124/pr.118.015768>

Gut microbiome research through fecal samples



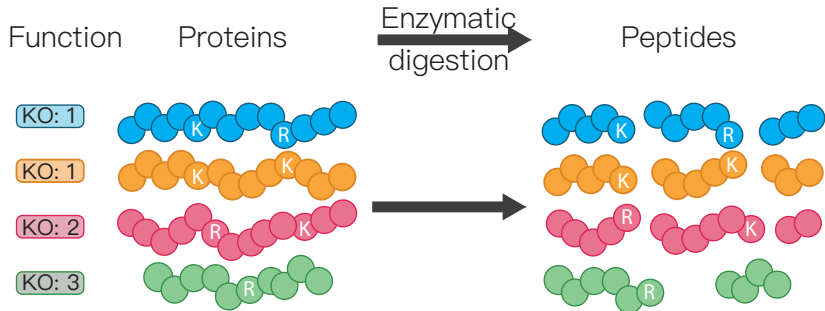
- We study the gut microbiome through **fecal samples**.
- We extract microbial **proteins** out of fecal samples.
 - ▶ **Metaproteomics** = study of proteins from a community (multiple species).
- Proteins are the **functional** players of the microbiome.
- Metaproteomics allows us to look at **what gut microbes are doing**.

Metaproteomics and microbiome research



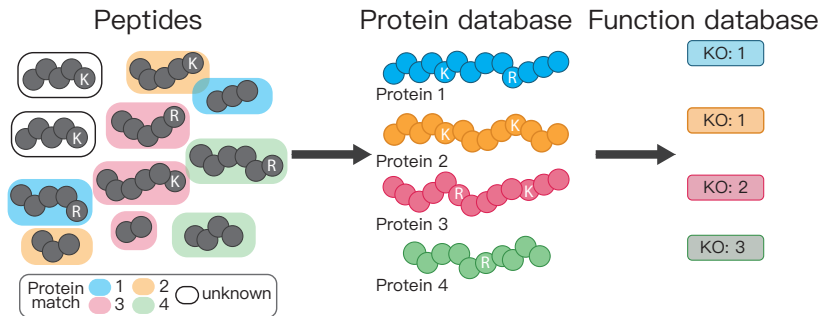
- Proteins are analyzed on a **mass spectrometer**.
- The **mass spectrometer** returns hundreds of thousands of spectra that we match to a database
- Using this database, we can computationally infer which protein was identified, from which microbial species, and how much of protein is in a sample.
 - ▶ We can analyze **microbiome composition** and **function**

Challenges with identifying function in metaproteomics



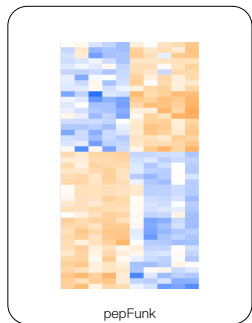
- Proteins are too large to be measured by a mass spectrometer and must be cut into smaller **peptides**.
- We use enzymes to cut proteins at **predictable sites** (K and R).
- The same peptide can be found in multiple proteins.
- **It is impossible to match some peptides back to the original protein.**

Why do we match peptides back to their parent protein?



A typical work flow matches each peptide back to a parent protein, and uses proteins for functional enrichment analysis. **What if we skipped this step and looked at functional enrichment of the identified peptides themselves?**

pepFunk: a peptide-centric functional enrichment methodology



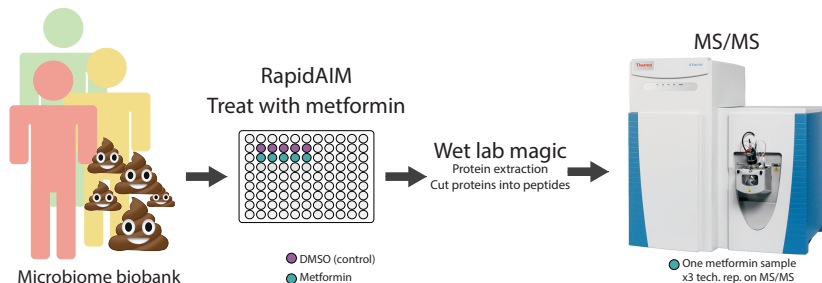
- **pepFunk highlights molecular functions that are over- or under-represented** in microbiome samples using Gene Set Variation Analysis (GSVA) adapted for use with **peptide** intensity data
- pepFunk can handle peptides with multiple functional annotations by intensity weighting
- pepFunk is also available as a **Shiny app** for simplified analysis

1. **Subset peptides into “peptide sets”**
 - ▶ KEGG pathways (gives information on functions)
2. **Estimate peptide intensity statistics** using a non-parametric kernel estimation of the cumulative density function
3. **Rank each peptide** by the expression statistic in every sample
4. **Calculate a Kolmogorov-Smirnov (KS)-like rank statistic** for each peptide set in each sample
5. **Calculate “GSVA” enrichment scores** for each peptide set²

I will illustrate these steps with an example from real data.

²Hanzelmann et al 2013 BMC Bioinformatics <http://www.biomedcentral.com/1471-2105/14/7>

using pepFunk on real data



We treated a single microbiome with Metformin for 24 hours (using **RapidAIM**³). The microbial proteins were extracted and analyzed on a **tandem mass spectrometer (MS/MS)**.

³Li et al 2020 Microbiome <https://doi.org/10.1186/s40168-020-00806-z>

Peptide sets organized by function

1. Metabolism

1.0 Global and overview maps

01100 Metabolic pathways *Major update!*
01110 Biosynthesis of secondary metabolites
01120 Microbial metabolism in diverse environments
01130 Biosynthesis of antibiotics
01200 Carbon metabolism
01210 2-Oxocarboxylic acid metabolism
01212 Fatty acid metabolism
01230 Biosynthesis of amino acids
01220 Degradation of aromatic compounds

1.1 Carbohydrate metabolism

00010 Glycolysis / Gluconeogenesis
00020 Citrate cycle (TCA cycle)
00030 Pentose phosphate pathway
00040 Pentose and glucuronate interconversions
00051 Fructose and mannose metabolism
00052 Galactose metabolism
00053 Ascorbate and aldarate metabolism
00500 Starch and sucrose metabolism
00520 Amino sugar and nucleotide sugar metabolism
00620 Pyruvate metabolism
00630 Glyoxylate and dicarboxylate metabolism
00640 Propanoate metabolism
00650 Butanoate metabolism
00660 C5-Branched dibasic acid metabolism
00562 Inositol phosphate metabolism

1.2 Energy metabolism

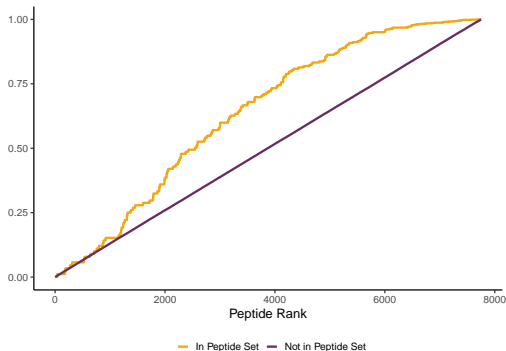
00190 Oxidative phosphorylation
00195 Photosynthesis
00196 Photosynthesis - antenna proteins
00710 Carbon fixation in photosynthetic organisms
00720 Carbon fixation pathways in prokaryotes
00680 Methane metabolism
00910 Nitrogen metabolism
00920 Sulfur metabolism

1.3 Lipid metabolism

00061 Fatty acid biosynthesis
00062 Fatty acid elongation
00071 Fatty acid degradation

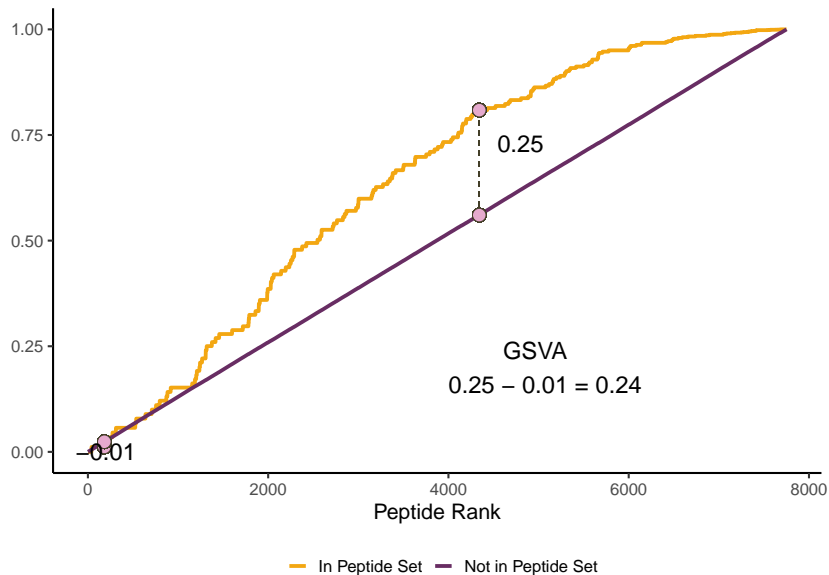
- We created peptide sets according to KEGG terms
 - ▶ Describes function
- Peptides are assigned to functions according to our curated database
- Peptides with multiple functions are weighted according to confidence in functional assignment
- Calculate potential enrichment of functions in each sample

We calculate a KS-like statistic for **each** peptide set in **each** sample

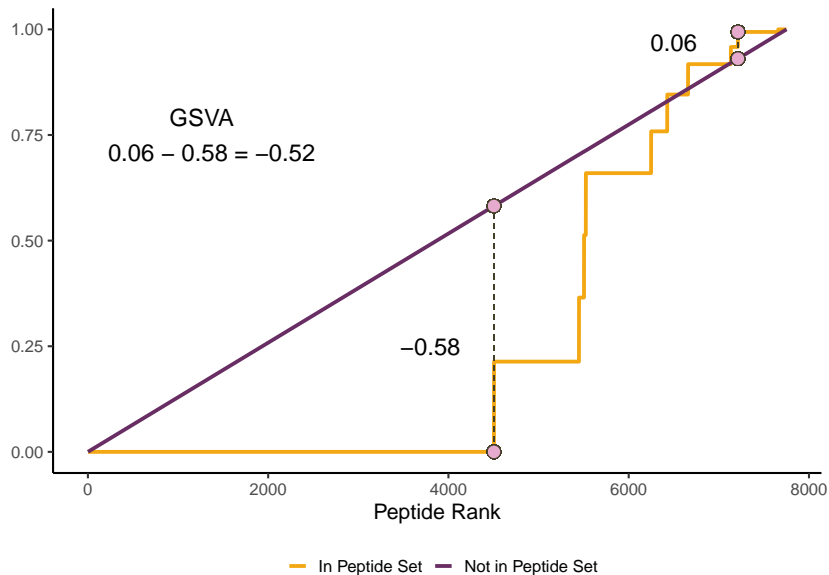


We use a **random walk** along the peptide rankings (from highest to lowest intensity) to determine if the peptide set is **more highly or lowly ranked than expected**. If a peptide is in a peptide set, **we rise according to its previously calculated peptide expression statistic**. We do the same for all peptides not in our set of interest.

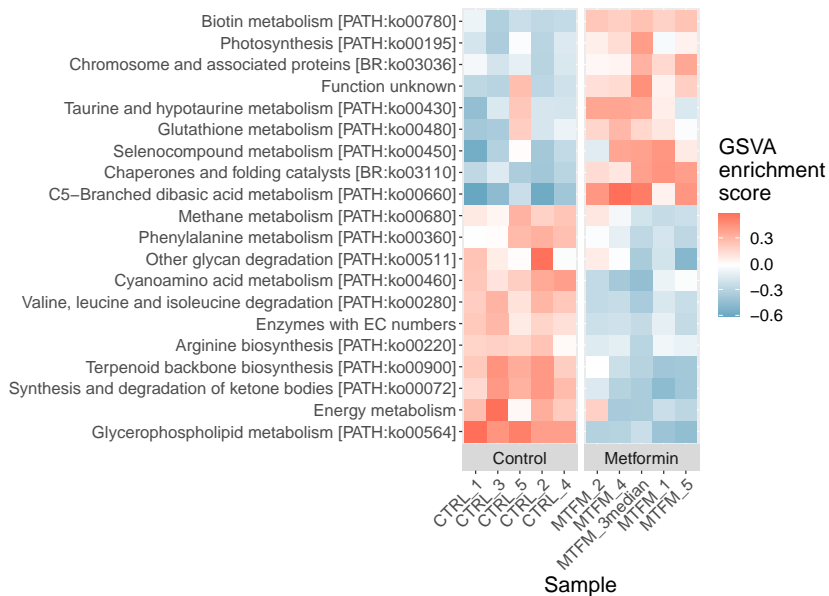
Positive GSVA score, higher intensity than expected



Negative GSVA score, lower intensity than expected



Adapted GSVA estimation - subset of results




pepFunk as part of MetaLab suite



- MetaLab/iMetLab⁴ is an entire platform dedicated to metaproteomic data analysis
- pepFunk is included in the iMetaLab platform along with other Shiny apps for metaproteomic data analysis

⁴Cheng et al 2020 J. Am. Soc. Mass Spectrom doi:10.1021/jasms.0c00083

The pepFunk Shiny app

☰✖

- Upload Data
- Analysis
- Gallery
- About
- MetaLab
- pepFunk on GitHub

1. Data input

A. Import peptide file:

Input data type:

Upload your own data
 Use our sample data

Choose the peptide intensity file to be analysed

Browse... No file selected

File format:

peptide.txt
 CSV with peptide sequences and intensities

File format: The peptide.txt output file from MaxQuant or MetaLab.

B. Add treatment information

Manual or auto condition formatting?

Manual
 Auto

Auto condition formatting will try and match your treatment conditions with your samples. If your sample names contain your treatment, this is a great option. Try typing your control and treatment names in the boxes above. If you have more than one treatment, please push the button to add another treatment option.

Input control condition

Enter control/reference condition name

Input condition 1

Enter test condition name

Add additional condition Remove added condition

2. Check sample names and sample conditions

Please upload a file of peptide intensity values.

Note: you can update your sample names here. Condition names are either auto filled or can be typed in. Please use the drop down options for conditions.

3. Analysis options

A. Data Normalization

Would you like pepFunk to normalize your data by depth?

Yes
 No

Note: if you opt for no normalization, we highly recommend you normalize your data using your own methods before uploading to pepFunk. If you'd like to know more about our normalization technique, please see our manuscript (<https://doi.org/10.1093/bioinformatics/btau209>).

B. Choose log transformation

Transform intensity values using:

Log2
 Log
 No transformation

C. Choose peptide-to-KEGG database

Peptide-to-KEGG database:

Curated human microbiome
 Upload your own database

Please upload a file of peptide intensity values.

- We can use pepFunk to look at functional changes of a microbiome:
 - ▶ after drug treatment, better understanding of side effects
 - ▶ because of diet
 - ▶ because of disease
 - ▶ other lifestyle changes
- pepFunk is **not limited to gut microbiomes**
 - ▶ **soil microbes** (agriculture)
 - ▶ **water microbes** (water treatment)
 - ▶ ...your sourdough starter? (the ultimate sourdough bread)

<https://github.com/northomics/pepFunk>

- **Dr Daniel Figeys**
 - **Dr Mathieu Lavallée-Adam**
 - Dr Zhibin Ning
 - Dr Xu Zhang
 - Dr Leyuan Li
 - Krystal Walker
- Patrick Smyth

[Link to Shiny app](#)
[Link to publication](#)



TECHNOMISE



Natural Sciences and Engineering
Research Council of Canada

Conseil de recherches en sciences
naturelles et en génie du Canada



uOttawa

Canada