

# Priority Research Programme

## Foods for improving gut function and comfort

HIGH-VALUE  
NUTRITION

Ko Ngā Kai  
Whai Painga

***Are microbes the missing piece of the  
gut comfort puzzle?***

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Senior Scientist, AgResearch

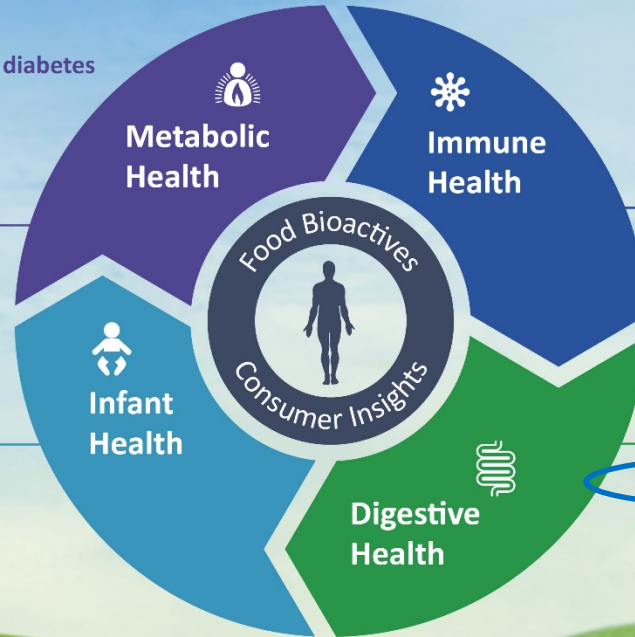
Host institution



# High-Value Nutrition National Science Challenge

## research themes and projects

- TOFI: Thin on the outside, fat inside: preventing diabetes
- Kiwifruit for glucose control
- Combined proteins for lean body mass
- Grass-fed beef for cholesterol control

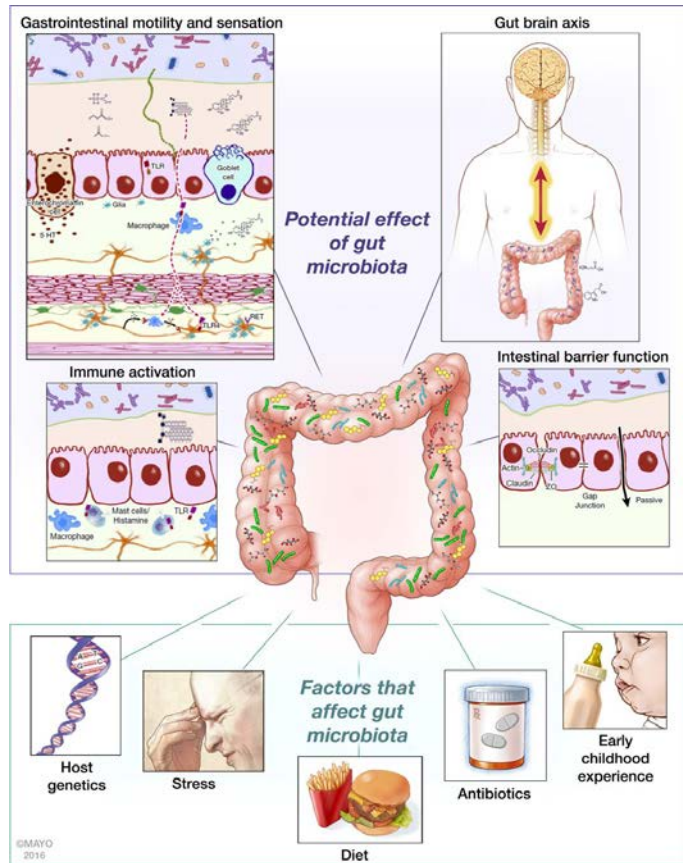


- Building immune defence
- Natural milk for allergy management
- Greenshell™ mussels to manage inflamed joints

- Complementary feeding for immune protection
- Fibres for sustained energy release

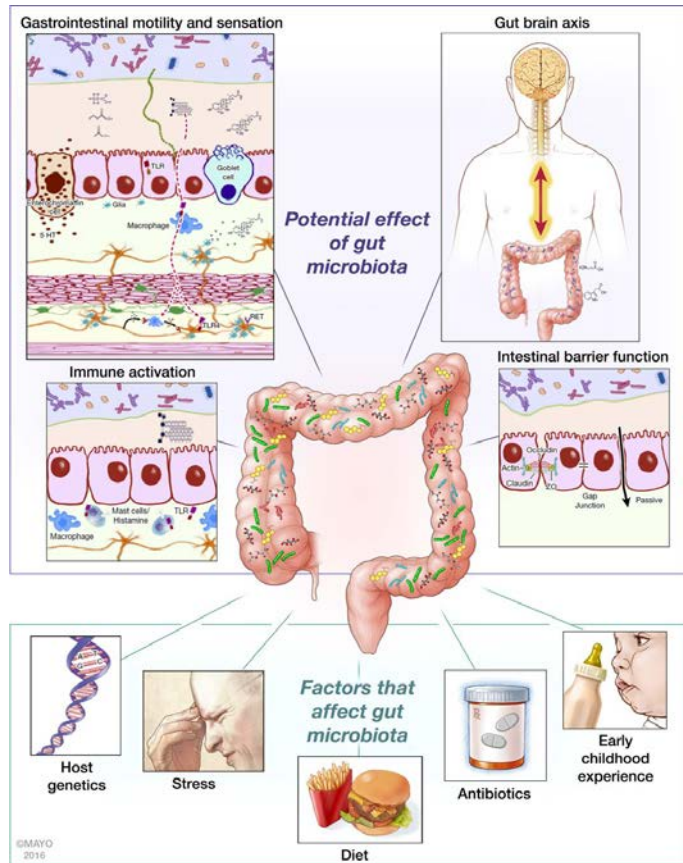
- Foods for gut function and comfort
- A2 Milk for gut comfort

**Irritable Bowel Syndrome is the ideal  
model for developing future foods with  
clinical evidence to support claims for  
healthy people**



# Microbiome and gut comfort

- Gut microbiota modulates mechanisms underlying gut comfort;
  - Motility
  - Immune system
  - Barrier function
  - Gut-brain axis
- Food is the obvious choice to fine-tune the microbiome for health conscious consumers



# Microbiome and gut comfort

- Altered microbiota associated with perturbed gut function, including irritable bowel syndrome (IBS)
- Changes differ between studies, e.g. Firmicutes, *Faecalibacterium*, *Blautia*, bifidobacteria, methanogens, *Prevotella*<sup>1-4</sup>
- For food to be effective, we need to know where we are coming from and where we are going

Bhattarai *et al* (2017). *Am J Physiol Gastrointest Liver Physiol*. 312:G52-G62

1. Jeffery *et al* (2012); 2. Tap *et al* (2017);  
3. Labus *et al* (2017); 4. Malinen *et al* (2005)

National  
**Science**  
Challenges

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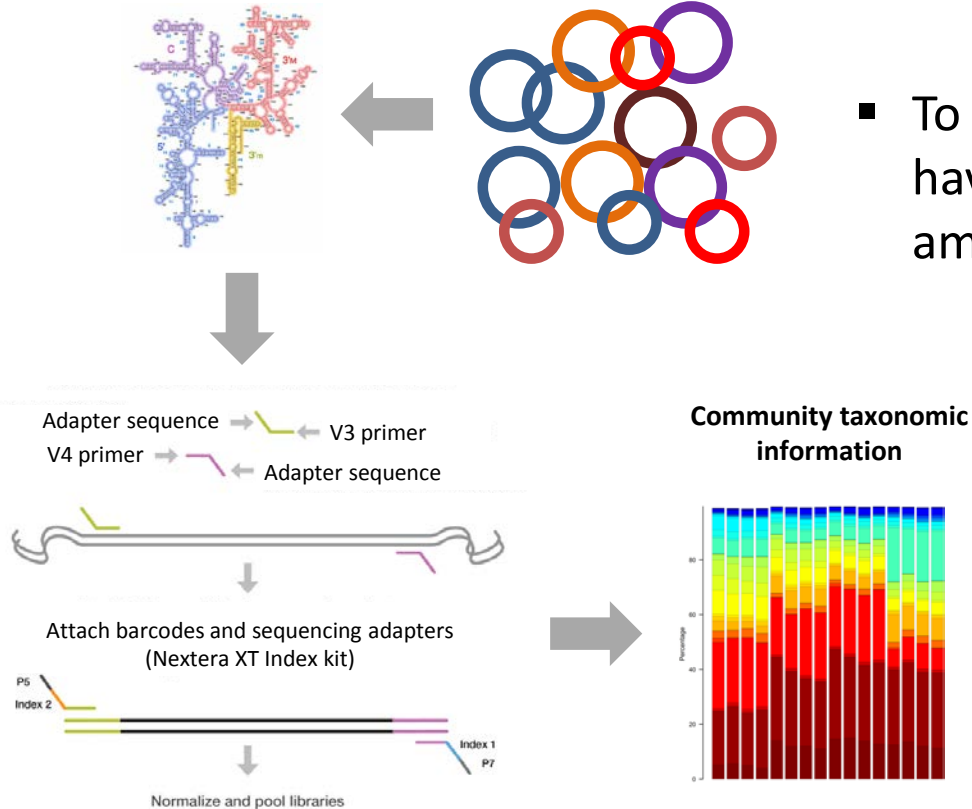


PCR amplify regions of bacterial 16S rRNA gene

Extract metagenomic DNA from faecal sample

# Which bacteria are present?

- To date, most IBS microbiome studies have been based on 16S rRNA gene amplicon sequencing



- Relatively easy to do
- **Only provides taxonomy** (who is there?)

# Our whole systems approach to High-Value Nutrition science

## Our biology

Organ networks  
Cellular networks  
Molecular networks  
Genetic interaction



## Our environment

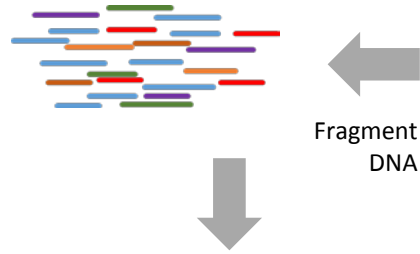
Where we live  
Cultural backgrounds  
Social networks  
Food choices

Our research focuses on understanding biological processes as complex integrated systems.  
Nutrition to keep us healthy and well requires an holistic approach.

## Shotgun metagenomic sequencing

## Extract metagenomic DNA from faecal sample

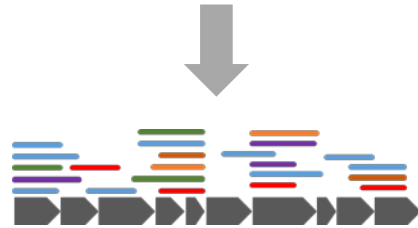
# What are the bacteria doing?



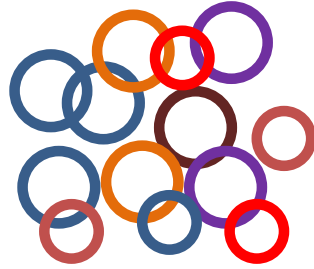
Fragment DNA

Attach barcodes and sequencing adapters  
Nextera XT Index kit/Rubicon ThruPLEX kit

Normalise and pool libraries  
Illumina HiSeq paired-end sequencing

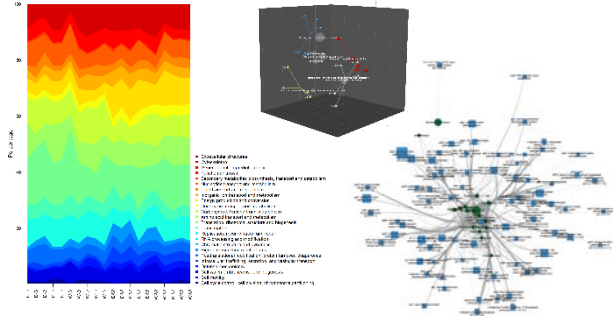


Align DNA sequences against reference  
genomes using MG-RAST/IMG



- We want to know what the bacteria are capable of doing
- Sequencing all DNA in the community (the **metagenome**) provides this insight

## Community taxonomic and functional gene abundance information



**Metagenomic sequencing gives insight into what the community is doing**

National  
**Science**  
Challenges

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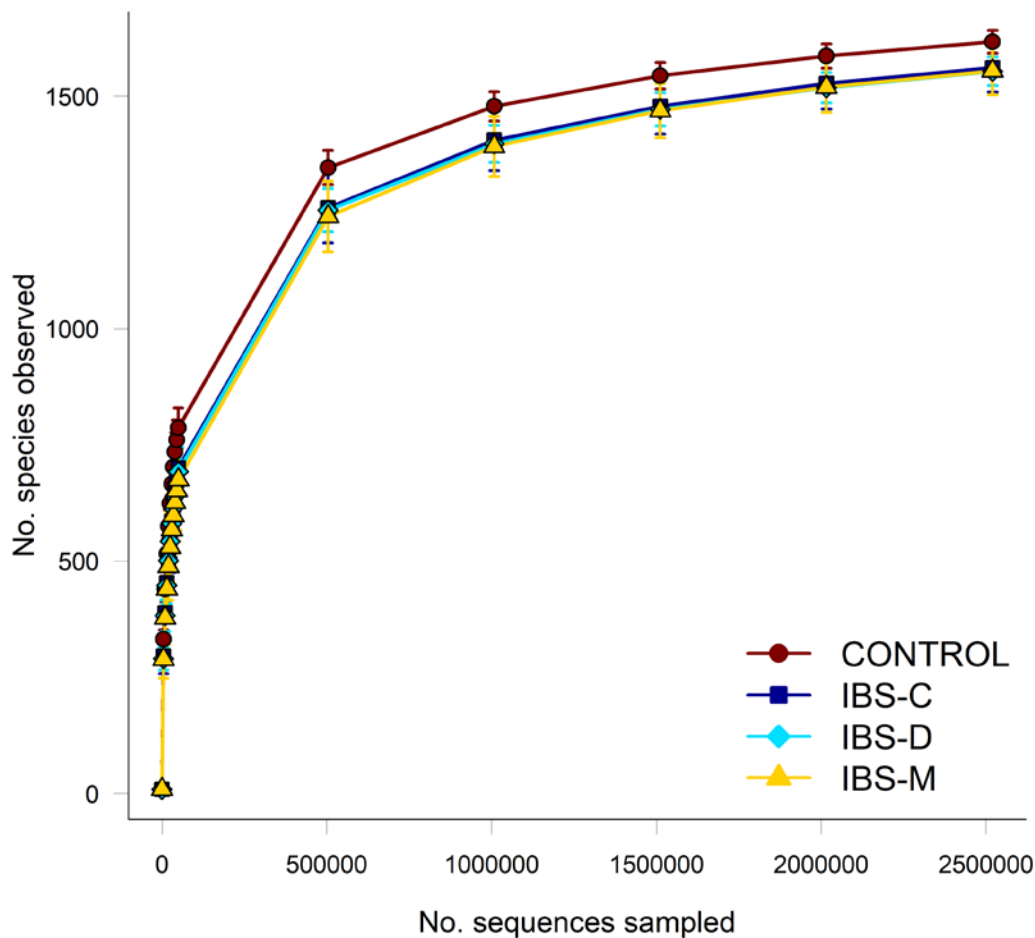
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# Metagenomic sequencing of COMFORT cohort faecal microbiome

- Faecal microbiome analysed by shotgun metagenome sequencing using Illumina NextSeq 500 paired-end 2x150 bp (APC/Teagasc)
- 112 samples sequenced
  - 41 case-controls
  - 9 IBS constipation (IBS-C)
  - 22 IBS diarrhoea (IBS-D)
  - 10 IBS mixed (IBS-M)
  - 16 functional constipation (FC)
  - 5 functional diarrhoea (FD)
  - 9 not determined

# What did we find?

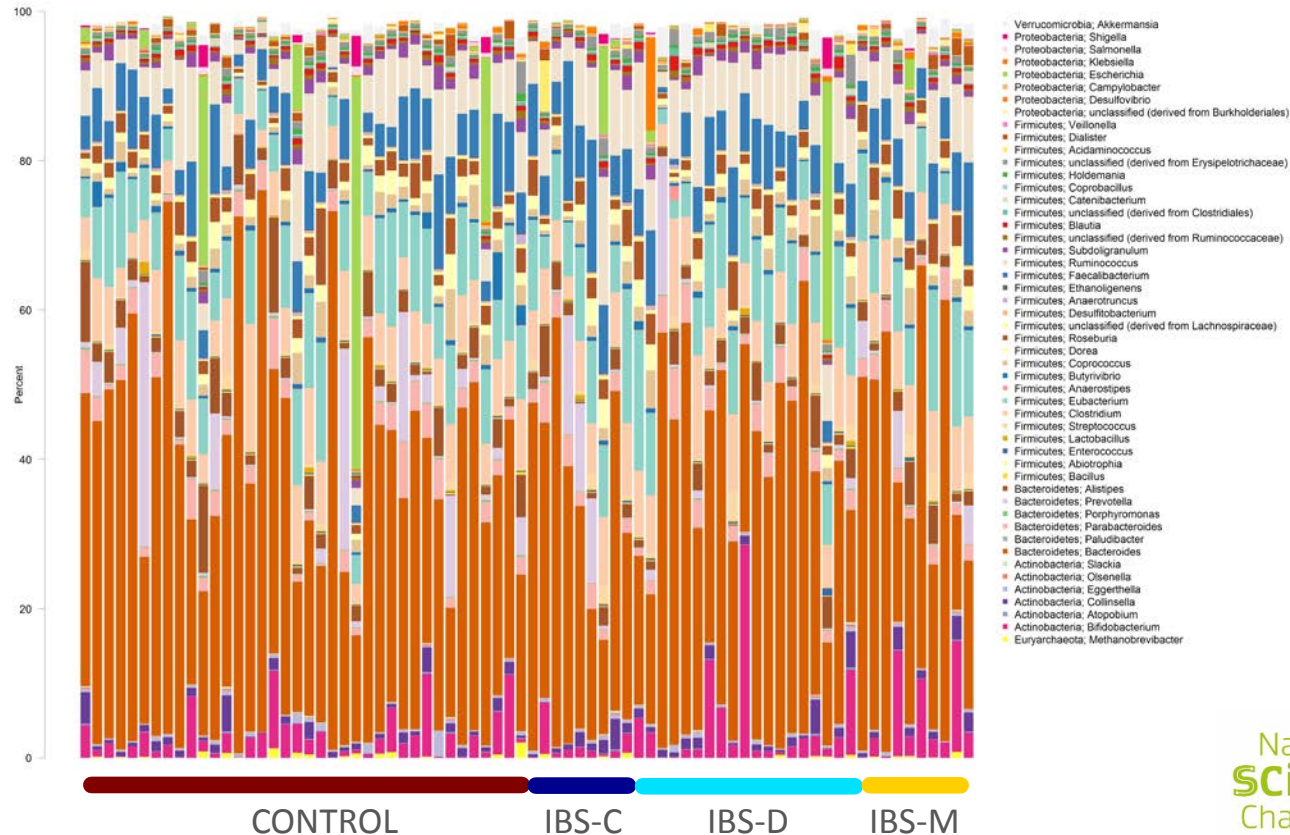


## Alpha diversity (how many different types)

- IBS groups have lower alpha diversity than controls ( $P=0.05$ )

# Microbiome composition

## Genus level

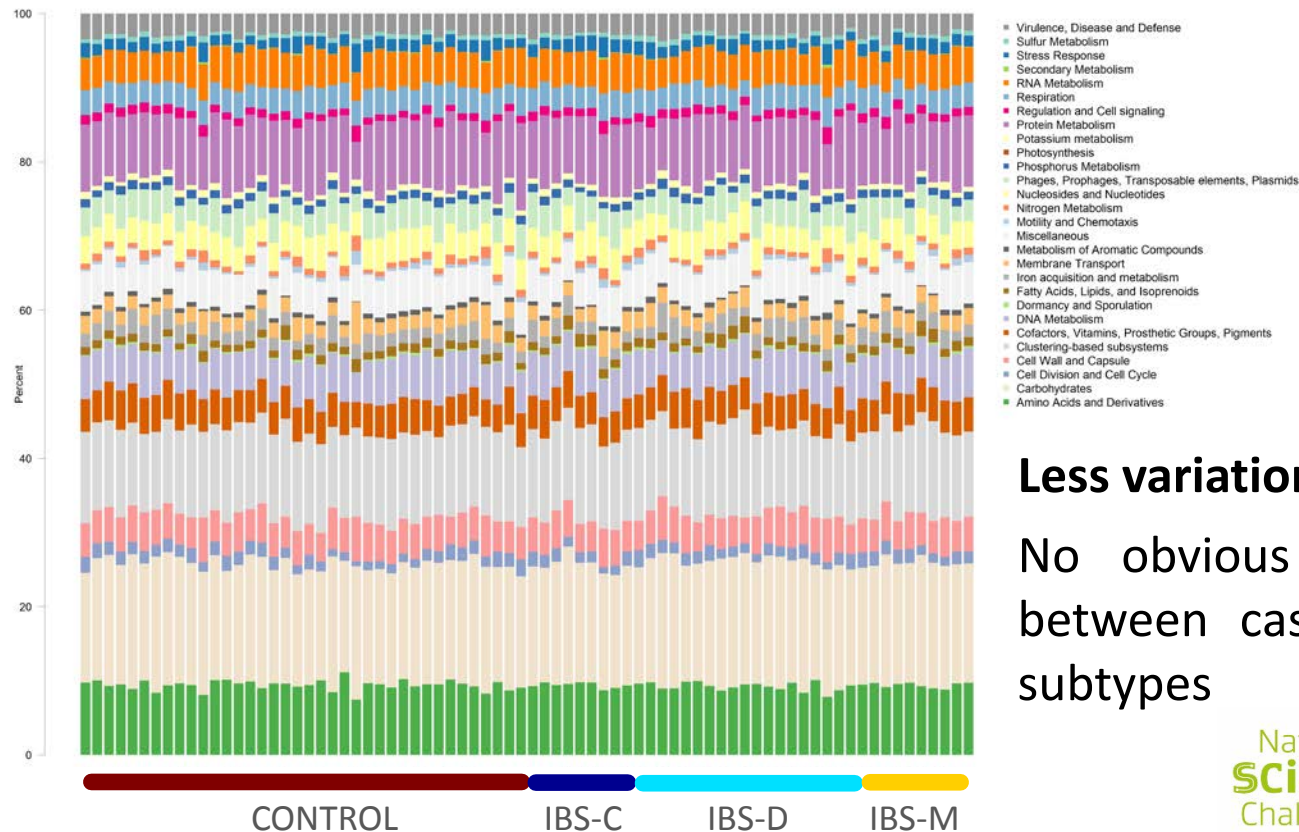


Highly variable

No obvious division immediately apparent between case-controls and IBS subtypes

# Gene functions

## Level 1



**Less variation between subjects**

No obvious division apparent between case-controls and IBS subtypes

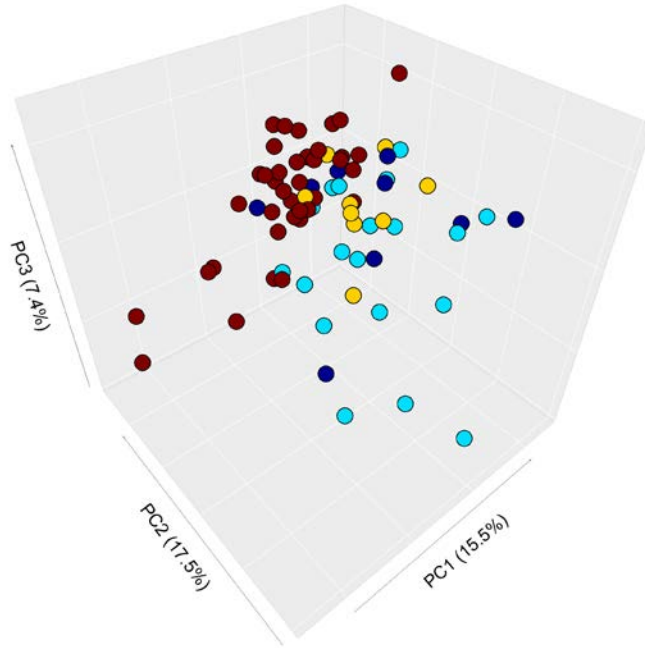
# Are there differences in microbiome composition and function?

- Standard statistical methods show few differences between case-controls and IBS subtypes
- Machine learning methods (e.g. PLS-DA, SVM, random forest) could help to separate groups

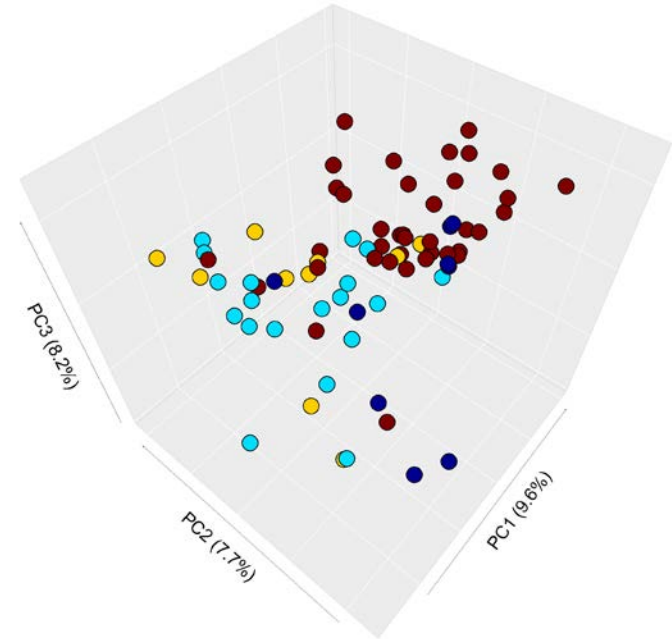


# Partial least squares discriminant analysis (PLS-DA)

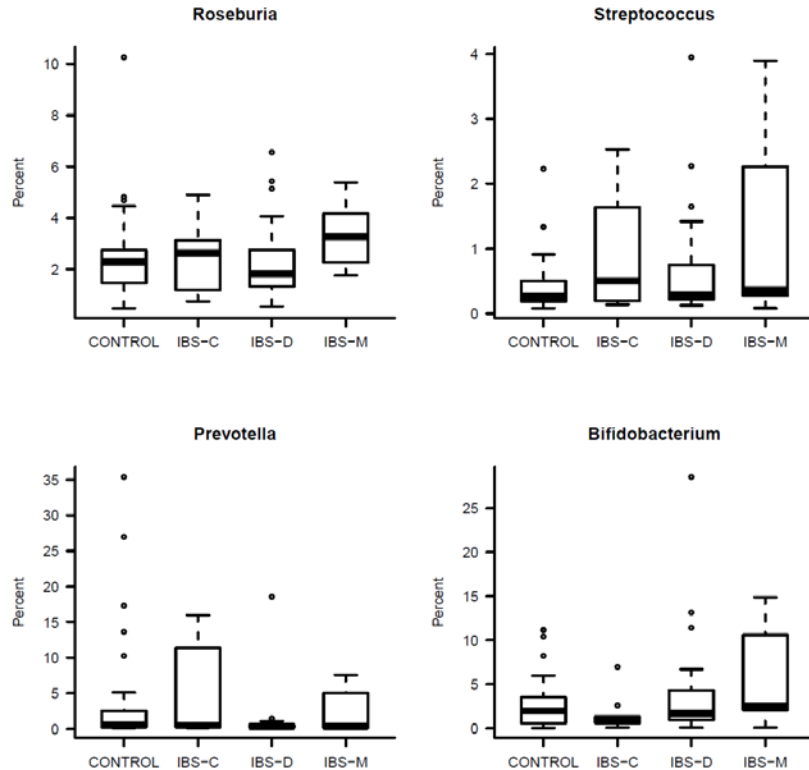
## Taxonomy



## Gene function



- IBS-C
- IBS-D
- IBS-M
- CONTROL

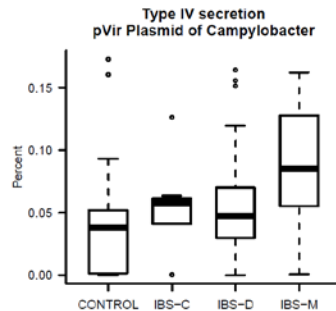
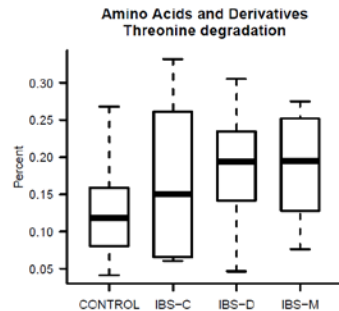
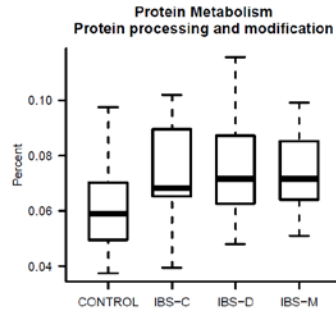
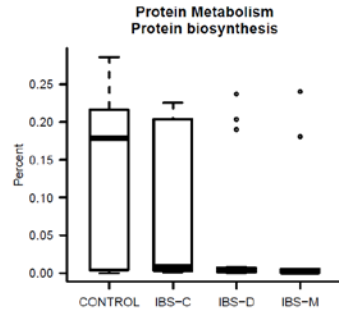
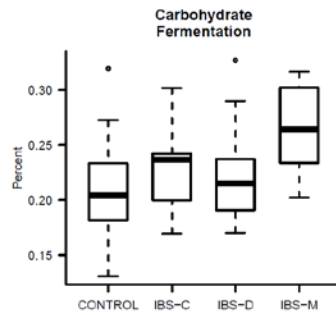
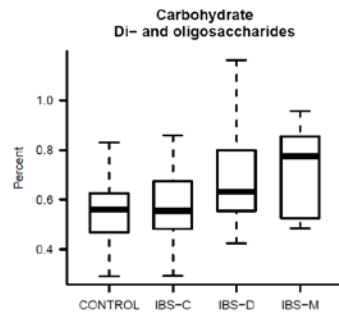


## What are the taxa and gene functions that lead to separation in PLS-DA models?

Genera include *Roseburia*<sup>1</sup>, *Streptococcus*<sup>1</sup>, *Prevotella*<sup>1,2</sup>, *Bifidobacterium*<sup>3</sup>

- Complex carbohydrate fermenters.
- Some generally thought of as “good” bacteria, but appear to be higher in IBS groups

1. Rigsbee *et al* (2012) Am J Gastrol 107:1740-71
2. Jeffery *et al* (2012) Gut 61:997-1006
3. Tap *et al* (2017) Gastroenterology 152:111-123



# What are the taxa and gene functions that lead to separation in PLS-DA models?

- Functions include those related to carbohydrate and protein metabolism
- Aligns with taxonomy data

# How robust is the separation between groups?

## Support vector machines (SVM)

### Taxonomy

| predict | CONT      | IBS       |
|---------|-----------|-----------|
| CONT    | <b>35</b> | 7         |
| IBS     | 3         | <b>31</b> |

### Gene function

| predict | CONT      | IBS       |
|---------|-----------|-----------|
| CONT    | <b>38</b> | 3         |
| IBS     | 0         | <b>35</b> |

# How robust is the separation between groups?

## Support vector machines (SVM)

### Taxonomy

| predict | CONT      | IBS-C    | IBS-D    | IBS-M    |
|---------|-----------|----------|----------|----------|
| CONT    | <b>38</b> | 7        | 10       | 9        |
| IBS-C   | 0         | <b>2</b> | 0        | 0        |
| IBS-D   | 0         | 0        | <b>9</b> | 0        |
| IBS-M   | 0         | 0        | 0        | <b>1</b> |

### Gene function

| predict | CONT      | IBS-C    | IBS-D     | IBS-M    |
|---------|-----------|----------|-----------|----------|
| CONT    | <b>38</b> | 7        | 6         | 8        |
| IBS-C   | 0         | <b>2</b> | 0         | 0        |
| IBS-D   | 0         | 0        | <b>13</b> | 0        |
| IBS-M   | 0         | 0        | 0         | <b>2</b> |

Improved predictive power with gene function

# How robust is the separation between groups?

## Support vector machines (SVM)

### Combined taxonomy and gene function

| predict | CONT      | IBS       |
|---------|-----------|-----------|
| CONT    | <b>37</b> | 2         |
| IBS     | 1         | <b>36</b> |

| predict | CONT | IBS-C    | IBS-D     | IBS-M    |
|---------|------|----------|-----------|----------|
| CONT    | 38   | 6        | 2         | 7        |
| IBS-C   | 0    | <b>3</b> | 0         | 0        |
| IBS-D   | 0    | 0        | <b>17</b> | 2        |
| IBS-M   | 0    | 0        | 0         | <b>1</b> |

**Predictive power further improved by combining taxonomy with gene function**

# Can we classify the undefined samples?

|       |       |       |       |       |       |       |       |       |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| 90040 | 90050 | 90055 | 90074 | 90075 | 90111 | 90121 | 90123 | 90126 |
| CONT  | CONT  | IBS   | CONT  | IBS   | IBS   | CONT  | IBS   | IBS   |

**Time will tell if these predictions are accurate**



# Conclusions

- Microbiome composition and function appear be different between controls and IBS subtypes, more pronounced with IBS-D
  - Add to predictability of existing biomarkers?
  - Carbohydrate fermentation appears to play a role
  - Some ostensibly “good” bacterial increased in IBS
  - What we need to know is “what they are doing”?
- *Clinical and systems approach will de-risk developing new foods with validated gut health benefits that will be highly desirable and sought after by healthy consumers*

# Research Team and collaborators

## Programme Leader/Principal Investigator

- Nicole Roy, AgResearch, Riddet Institute



## Principal Investigator, clinical

- Richard Gearry, University of Otago



## Associate Investigators, biomarkers

- Metabolites: Karl Fraser, AgResearch
- Microbiota: Wayne Young, AgResearch
- Immune: Oliver Grasser, Malaghan Institute
- Proteins: Janine Cooney, Plant and Food Research



## Collaborator

- Sequencing: Paul Cotter, APC/Teagasc



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