

DTU





**DTU Health Technology
Bioinformatics**

Quantitative metagenomics

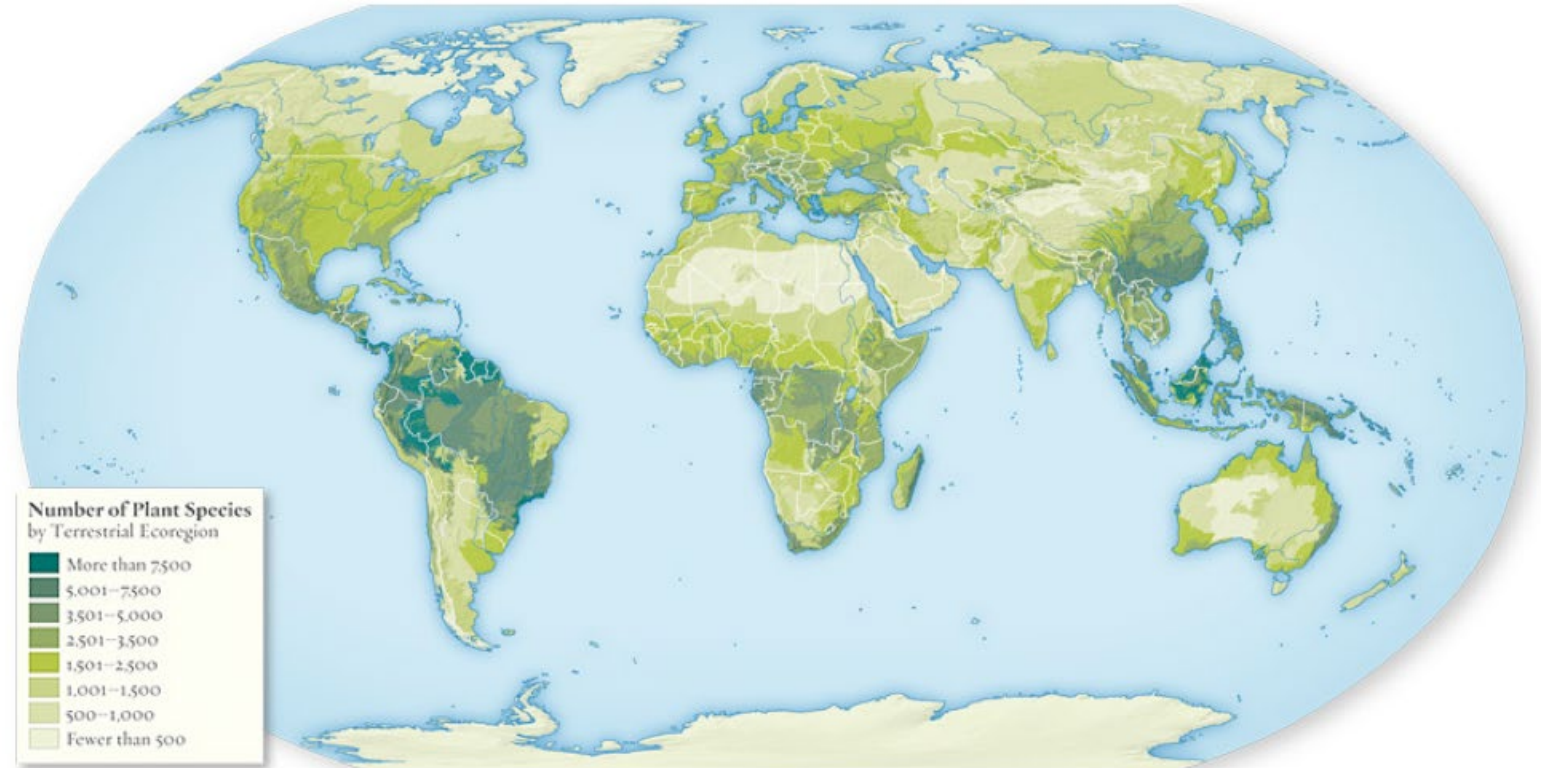
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Menu

- Diversity measurements
 - Alpha & beta diversity
- Normalization
- Sequencing data is compositional data

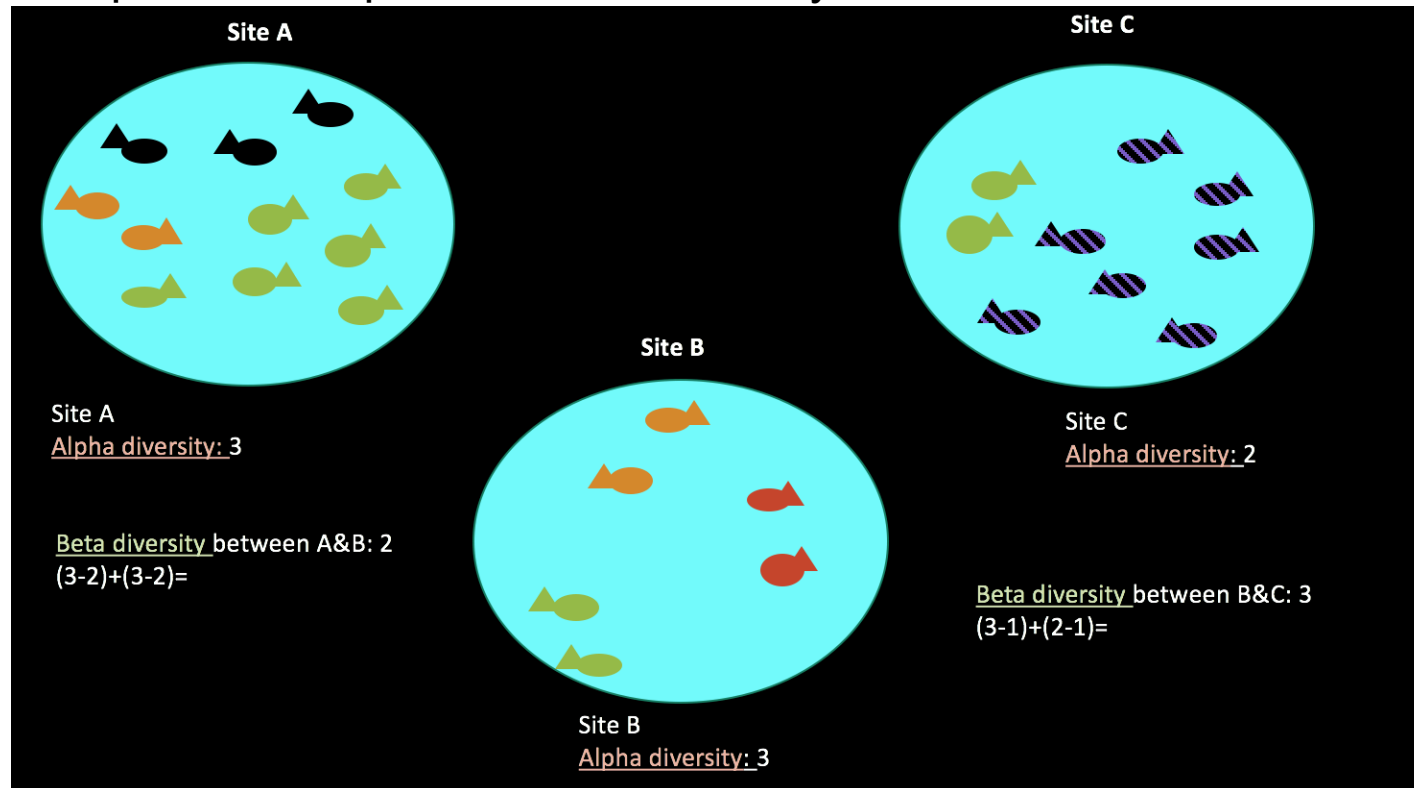
Classical measures

- Abundance
- Richness
- Rarefaction
- Diversity
 - Alpha
 - Beta

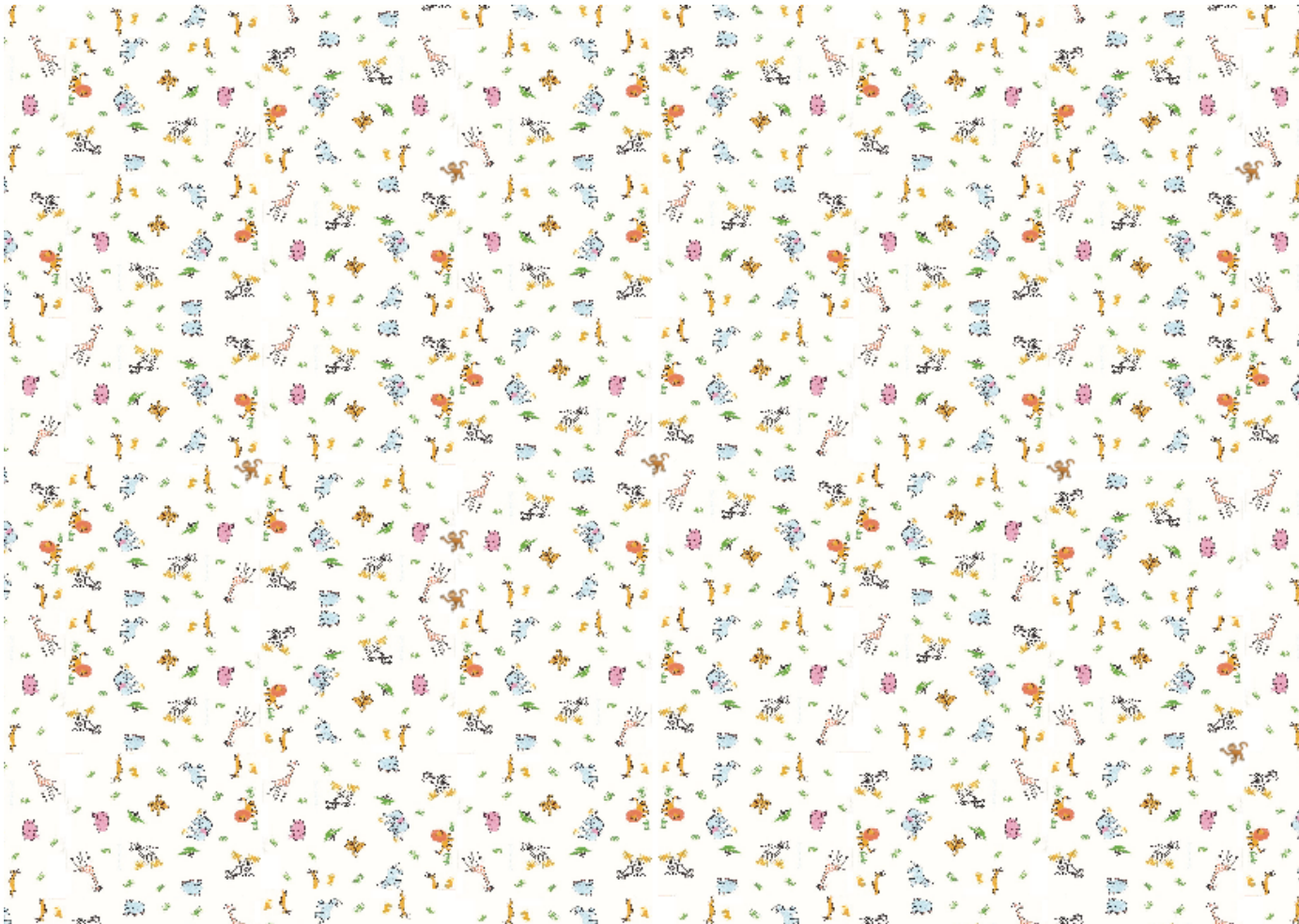


Describing the spatial component of biodiversity

- Alpha diversity (within sample)
- Beta diversity (between samples)
- We can compare both alpha and beta diversity



Abundance (counts)



| | |
|--------------|-----|
| Lion | 64 |
| Zebra | 128 |
| Giraffe | 64 |
| leopard | 64 |
| rhinoceros | 64 |
| hippopotamus | 128 |
| gazelle | 128 |
| elephant | 64 |
| monkey | 9 |

Species richness

- The number of different species in a system

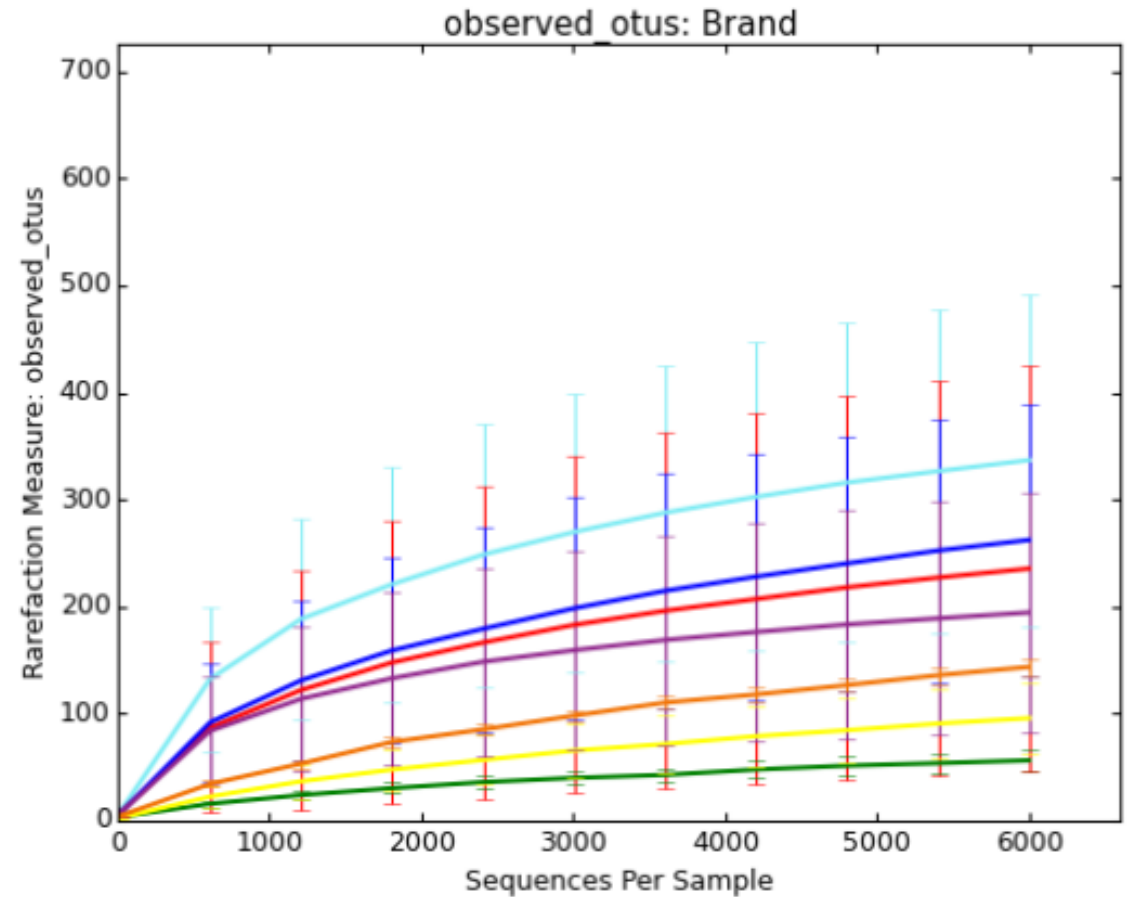
| | |
|--------------|-----|
| Lion | 64 |
| Zebra | 128 |
| Giraffe | 64 |
| leopard | 64 |
| rhinoceros | 64 |
| hippopotamus | 128 |
| gazelle | 128 |
| elephant | 64 |
| monkey | 9 |



9 observed species

Rarefaction

- Species richness is a function of our no. observations
- When have we sampled enough?
- Mostly used for 16s rRNA amplicons...why?



Shannon index

- Incorporates species richness & evenness
- Quantify the entropy (information content)
- Quantifies the uncertainty (degree of surprise) associated with a prediction
- The Shannon index increases as both the richness and the evenness of the community increase
- Typical values are generally between 1.5 and 3.5 in most ecological studies, and the index is rarely greater than 4

$$H' = - \sum_{i=1}^R p_i \ln p_i \qquad H' = -(\ln p_1^{p1} + \ln p_2^{p2} + \ln p_3^{p3} + \dots + \ln p_R^{pR})$$

P_i = species proportion

R = observed species

Alpha diversity



| | |
|--------------|---|
| Lion | 1 |
| Zebra | 2 |
| Giraffe | 1 |
| Leopard | 1 |
| Rhinoceros | 1 |
| Hippopotamus | 2 |
| Gazelle | 2 |
| Elephant | 1 |
| Monkey | 0 |

$$H' = -(\ln p_1^{p_1} + \ln p_2^{p_2} + \ln p_3^{p_3} + \dots + \ln p_R^{p_R})$$

11 animals (NOT species) meaning each animal is 0.09 of the total abundance

$$H' = -(\ln(0.09^{0.09}) + \ln(0.18^{0.18}) + \dots) = 2.0$$

Bray-curtis dissimilarity

$$0 \leq B \leq 1$$

$$B_{ij} = 1 - 2C_{ij} / (S_i + S_j)$$

C = sum of the lowest count of all common species

S = total count of the sample

1 means that they do not share anything

$$B_{s_1s_2} = 1 - 2*(2+1) / (9 + 13) = 0.73$$

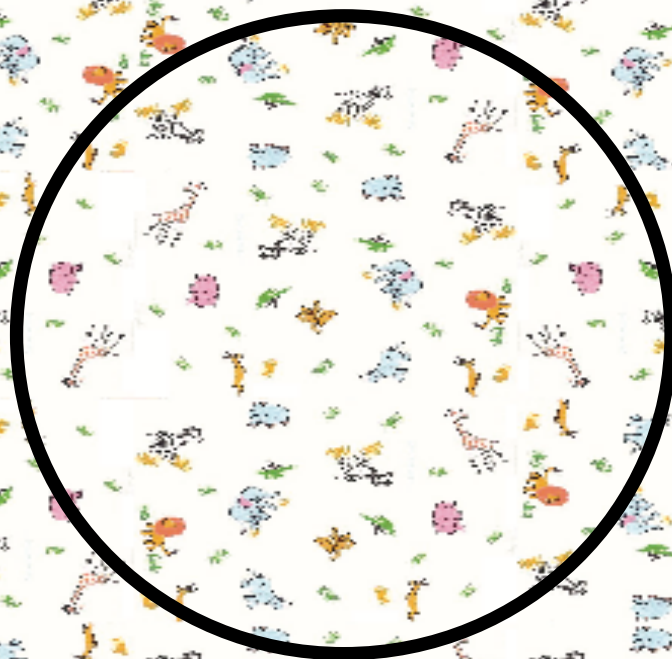
| | | |
|--------------|----------|-----------|
| Lion | 0 | 2 |
| Zebra | 3 | 2 |
| Giraffe | 0 | 4 |
| Leopard | 0 | 2 |
| Rhinoceros | 1 | 2 |
| Hippodrome | 4 | 0 |
| Gazelle | 0 | 1 |
| Elephant | 1 | 0 |
| Total | 9 | 13 |

Sampling effect

- To be fair we should sample equally in the systems we investigate



Sample sizes



Sample sizes

- Accounting for different sample sizes:
 - Normalise to sample size
 - Rarefy (downsize) samples
 - Statistically model the variance

Normalizing

$$N = n_i/n_{tot}$$

| | | |
|--------------|------------|-----------|
| Lion | 64 | 1 |
| Zebra | 128 | 2 |
| Giraffe | 64 | 1 |
| Leopard | 64 | 1 |
| Rhinoceros | 64 | 1 |
| Hippopotamus | 128 | 2 |
| Gazelle | 128 | 2 |
| Elephant | 64 | 1 |
| Monkey | 9 | 0 |
| Total | 713 | 11 |

| | | |
|--------------|------------|------------|
| Lion | 8.98 | 9.09 |
| Zebra | 17.95 | 18.18 |
| Giraffe | 8.98 | 9.09 |
| Leopard | 8.98 | 9.09 |
| Rhinoceros | 8.98 | 9.09 |
| Hippopotamus | 17.95 | 18.18 |
| Gazelle | 17.95 | 18.18 |
| Elephant | 8.98 | 9.09 |
| Monkey | 1.26 | 0 |
| Total | 100 | 100 |

Issue with different sampling power (higher chance of observing rare species) and does not take compositional nature into account

Downsize / rarefy

Resample x amount of observations

| | | |
|--------------|------------|-----------|
| Lion | 64 | 1 |
| Zebra | 128 | 2 |
| Giraffe | 64 | 1 |
| Leopard | 64 | 1 |
| Rhinoceros | 64 | 1 |
| Hippopotamus | 128 | 2 |
| Gazelle | 128 | 2 |
| Elephant | 64 | 1 |
| Monkey | 9 | 0 |
| Total | 713 | 11 |

| | | |
|--------------|-----------|-----------|
| Lion | 2 | 1 |
| Zebra | 3 | 2 |
| Giraffe | 0 | 1 |
| Leopard | 1 | 1 |
| Rhinoceros | 0 | 1 |
| Hippopotamus | 3 | 2 |
| Gazelle | 1 | 2 |
| Elephant | 0 | 0 |
| Monkey | 0 | 0 |
| Total | 10 | 10 |

Downsize / rarefy

- Select the target depth carefully
- The more reads we keep the more sensitive
- We may have to remove samples with few counts
- We might throw away a lot of data
- Still does not take compositional nature of data into account

Compositional data

- Arbitrary total
- Sequencing depth never 100%
- Species can co-exist without abundance inter-influences
- Independence between abundance is affected by the capacity of the sequencing instrument
- Sequencing instrument has fixed number of slots

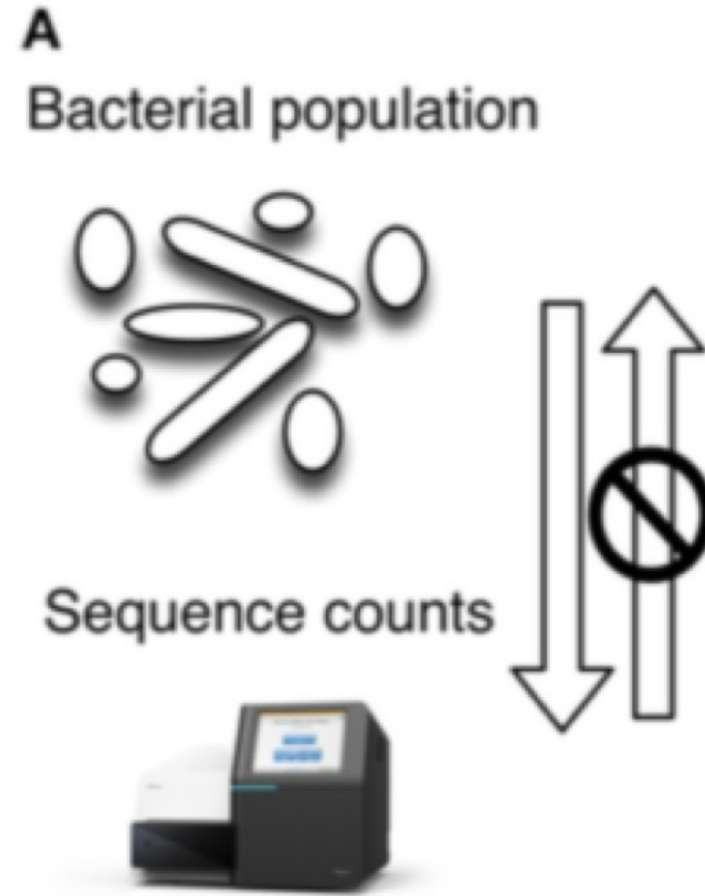
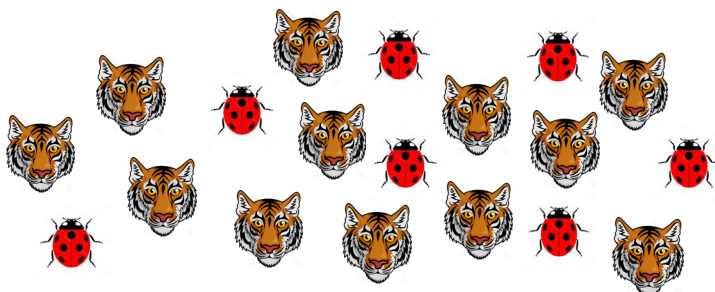


Figure from: Gloor, Gregory B. *et al.*, Microbiome Datasets Are Compositional: And This Is Not Optional. *Frontiers in Microbiology* **8** (2017)

Compositional data problems

- **Example:** an environment containing both tigers and ladybugs
 - The abundances of the two are not affected by each other
 - If the abundance of the ladybugs increases some of the slots with tigers must instead be filled by ladybugs
 - i.e. the two environmentally independent species are affecting the read count of each other

Population: 12 tigers and 8 ladybugs



Count: 6 tigers and 4 ladybugs



Increase in abundance of ladybugs,
no change in abundance of tigers



Population: 12 tigers and 10 ladybugs



Count: 5 tigers and 5 ladybugs



Dealing with compositional data

- Statistically model the variance & heteroscedasticity
- Use packages developed for RNA-seq such as DESeq2 and edgeR
- DESeq2 takes raw counts divided by sample-specific size factors determined by median ratio of gene counts relative to geometric mean per gene
([See this link for a brilliant explanation](#))