

Microbial **Co-Exclusion** and **Co-Occurrence**:

Making it All Add Up

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**Competitive
Exclusion
Principle**

**“Complete
Competitors
Can’t Coexist”**



**They live all
Packed together**

**Or we can have
mutualists that
love to coexist!**

How do they interact?



Consider just
two species...

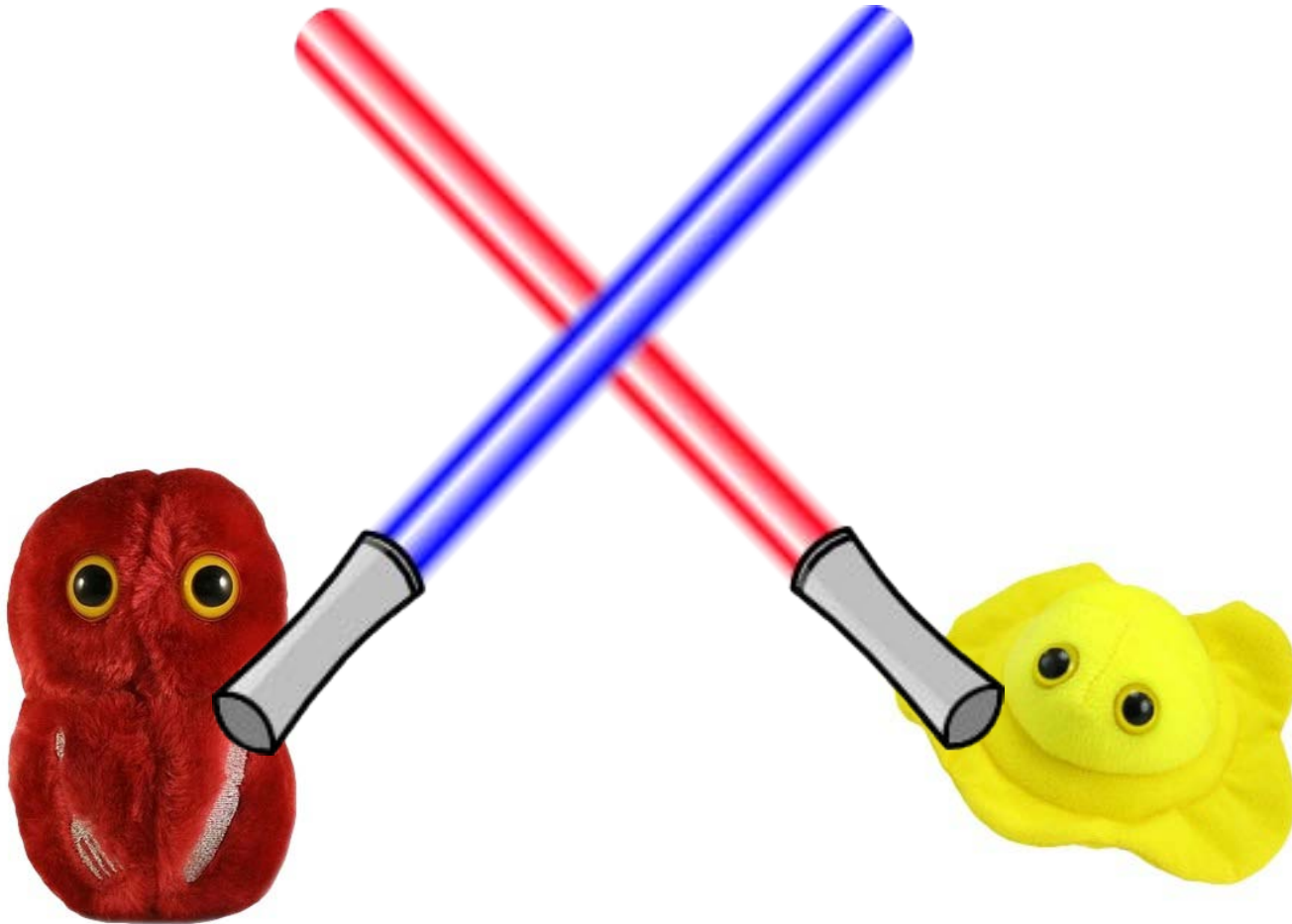
... living together
in an environment.



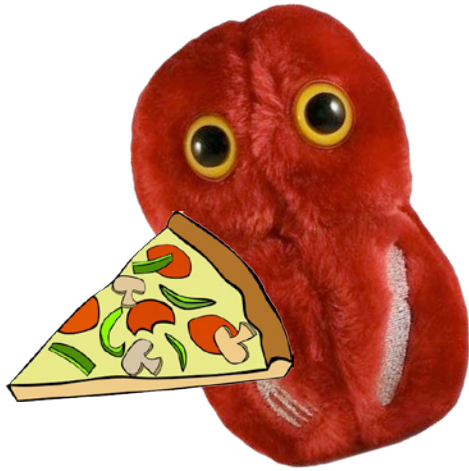
Co-Occurrence?



or **Co-Exclusion?**



Indifference?

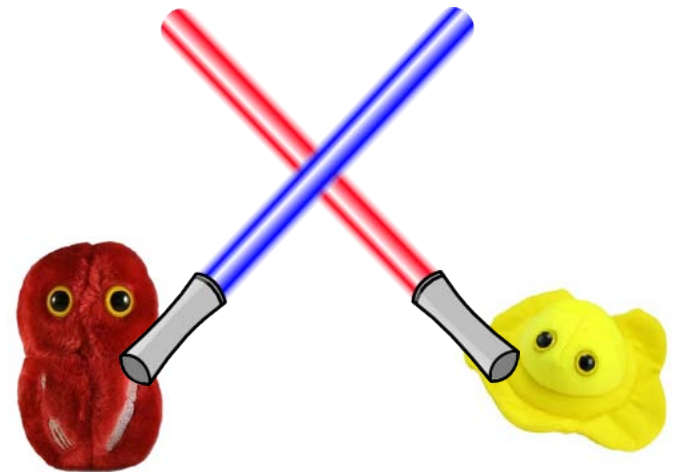


How can we tell which of these is occurring?

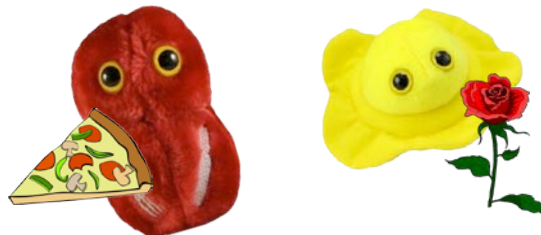
Co-Occurrence



Co-Exclusion



Indifference



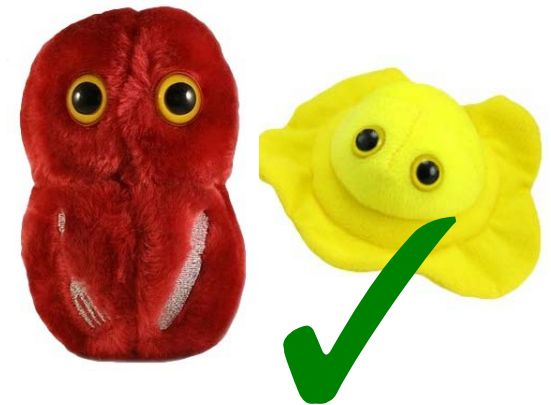
Co-Occurrence



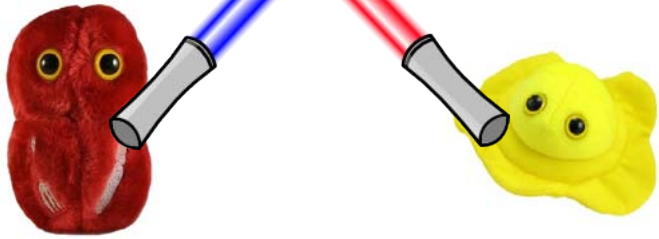
Marginal Relative Abundance



Marginal Relative Abundance



Co-Exclusion



Marginal Relative Abundance

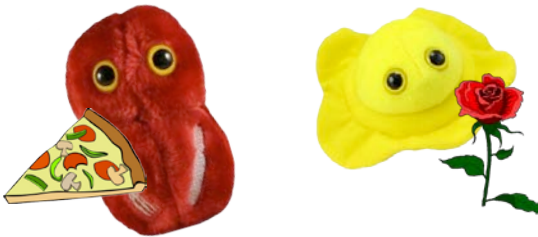


Marginal Relative Abundance



Indifference

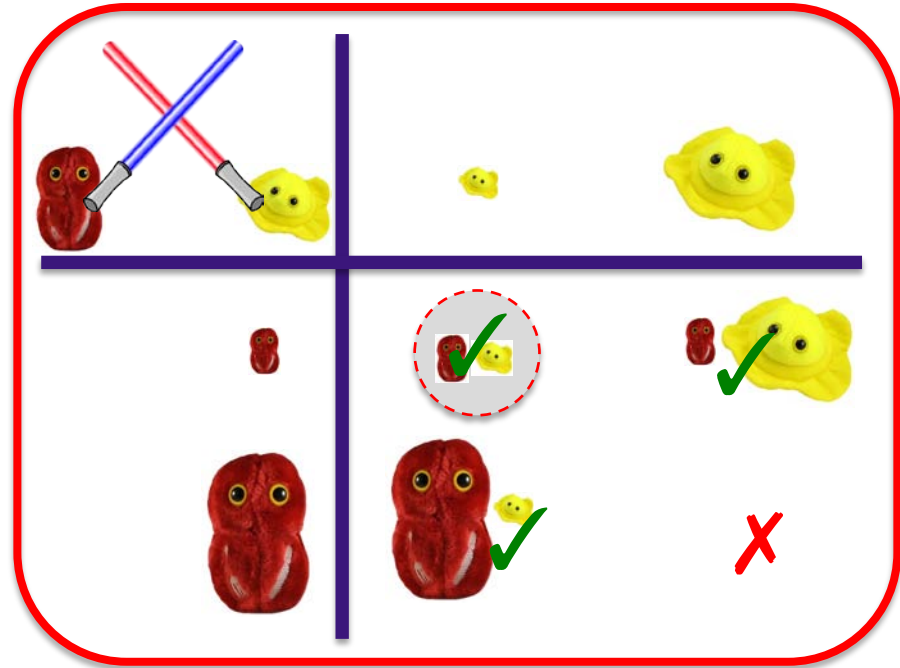
Marginal Relative Abundance



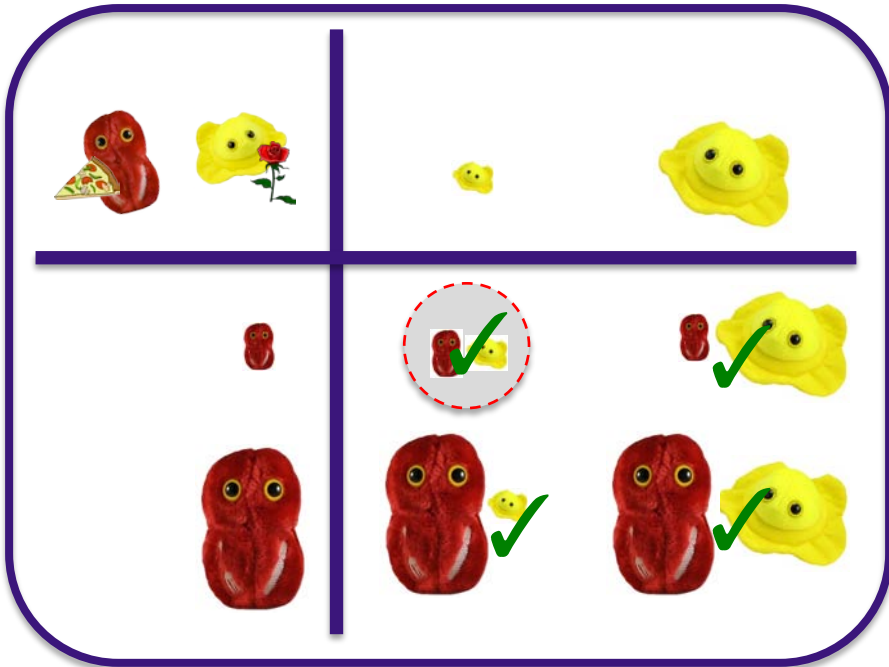
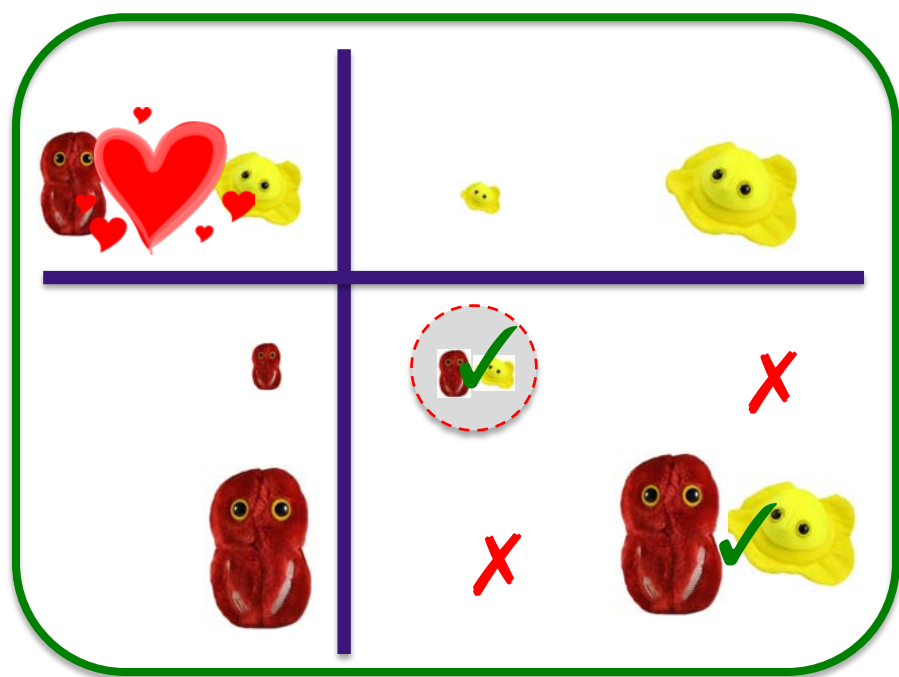
Marginal Relative Abundance



Let's recap...



These are subtle differences!





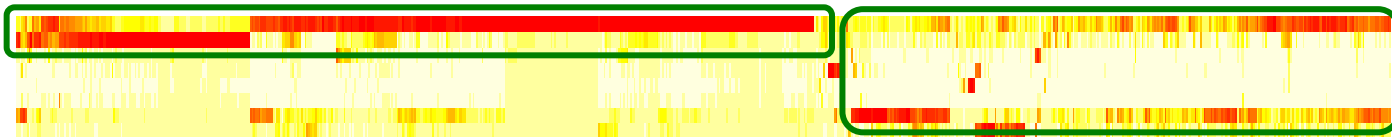
“Hey! Look at these *great* results!
They show **obvious** co-exclusion!” *

Human Vaginal Microbiome

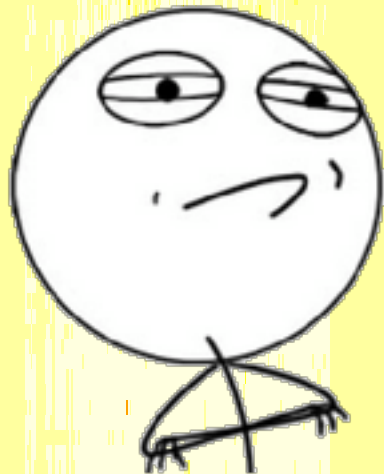
(normal vs. disease state)

Bacterium

seq_0_Lactobacillus_iners
seq_2_Lactobacillus_crispatus
seq_31_Lactobacillus_jensenii
actobacillus_gasseri,johnsonii



CHALLENGE ACCEPTED



“Sorry, boss. No they don’t.”

“What?! Sure they do!”

“Nope.”

“Okay... **prove it!**”

Patients

- Gregory B. Gloor, Ruben Hummelen, Jean M. E. Macklaim, **Andrew D. Fernandes**, and Gregor Reid (2010, *Accepted*) Community Microbiome Profiling by Combinatorial Barcoding with Illumina Sequencing. *PLoS One* (Manuscript PONE-D-10-00044R1). Archived at <http://arxiv.org/abs/1007.5075v1>.
- Ruben Hummelen, **Andrew D. Fernandes**, Jean M. E. Macklaim, Russell J. Dickson, John Changelucha, Gregory B. Gloor, and Gregor Reid (2010, *Accepted*) Deep Sequencing of the Vaginal Microbiota of Women With HIV. *PLoS One* (Manuscript 10-PONE-RA-19937).

***Dramatic Re-interpretation**

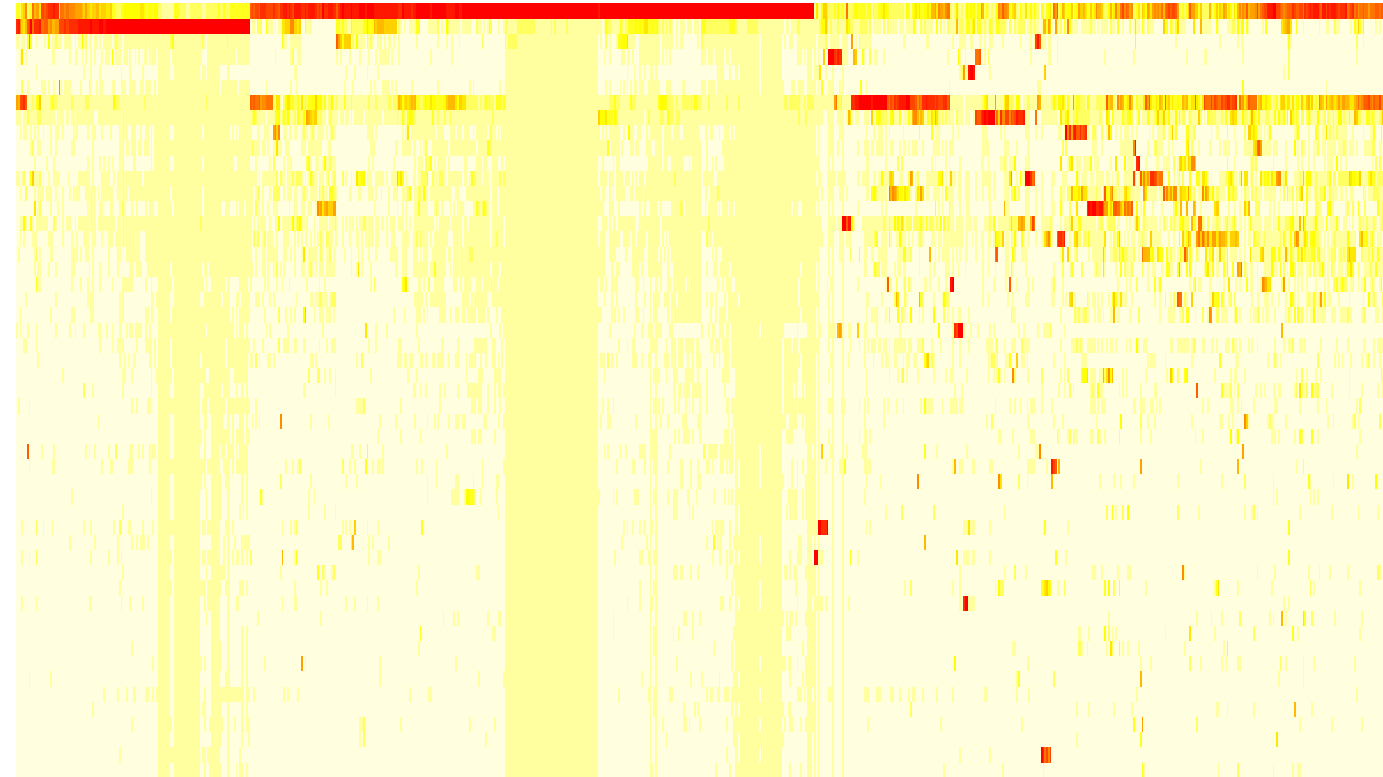
Where do results like this come from?

Claim: Zeros, Logarithms, and PCA!

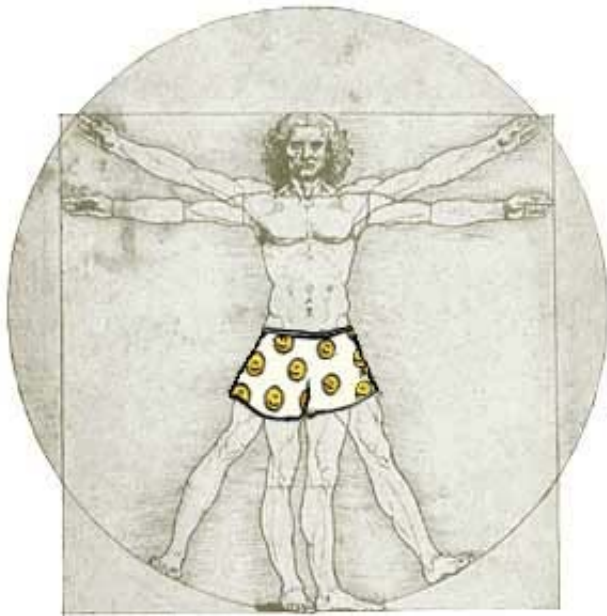
Bacterium

seq_0 Lactobacillus_iners
seq_2 Lactobacillus_crispatus
seq_31 Lactobacillus_jensenii
actobacillus_gasseri.johnsonii

Patients



Zeros



Clinical Subject
(well, female...)



Swab
Sample



PCR



Multiplex
Next-Gen
Sequence



ion torrent



(actually Illumina & Solid...)



	<i>Lactobacillus iners</i>	<i>Lactobacillus crispatus</i>	<i>Lactobacillus jensenii</i>	<i>Lactobacillus gasseri</i>
Patient 1	49679	3177	21389	135
Patient 2	7755	29752	989	368
Patient 3	3286	5955	549	397
Patient 4	2265	3263	13742	148
Patient 5	10239	2926	226	100
Patient 6	16376	20706	1037	79
Patient 7	27313	4878	5320	92
Patient 8	33006	1103	1186	176
Patient 9	20504	1771	346	161

V6 Read Counts

Counts *between samples (patients)* are **meaningless!**

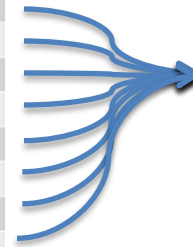
We are interested in sample **proportions**, only!

	<i>Lactobacillus iners</i>	<i>Lactobacillus crispatus</i>	<i>Lactobacillus jensenii</i>	<i>Lactobacillus gasseri</i>	...
Patient 1	49679	3177	21389	135	...
Patient 2	7755	29752	989	368	...
Patient 3	3286	5955	549	397	...
Patient 4	2265	3263	13742	148	...
Patient 5	10239	2926	226	100	...
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Patient 8	33006	1103	1186	176	...
Patient 9	20504	1771	346	161	...

⋮ ⋮

But **proportions** and **counts** are kind-of-almost the same thing, aren't they?

	<i>Lactobacillus iners</i>	<i>Lactobacillus crispatus</i>	<i>Lactobacillus jensenii</i>	<i>Lactobacillus gasseri</i>
Patient 1	49679	3177	21389	135
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$$p_i \approx \frac{n_i}{\sum_j n_j}$$

Nope.



ONLY VALID IF n_i IS NOT SMALL!

So where does $p_i \approx \frac{n_i}{\sum_j n_j}$ come from?

Consider a biome with only **two** species.

$$\Pr(p_H, p_T | n_H, n_T) = \frac{(n_H + n_T)!}{n_H! n_T!} p_H^{n_H} p_T^{n_T}$$

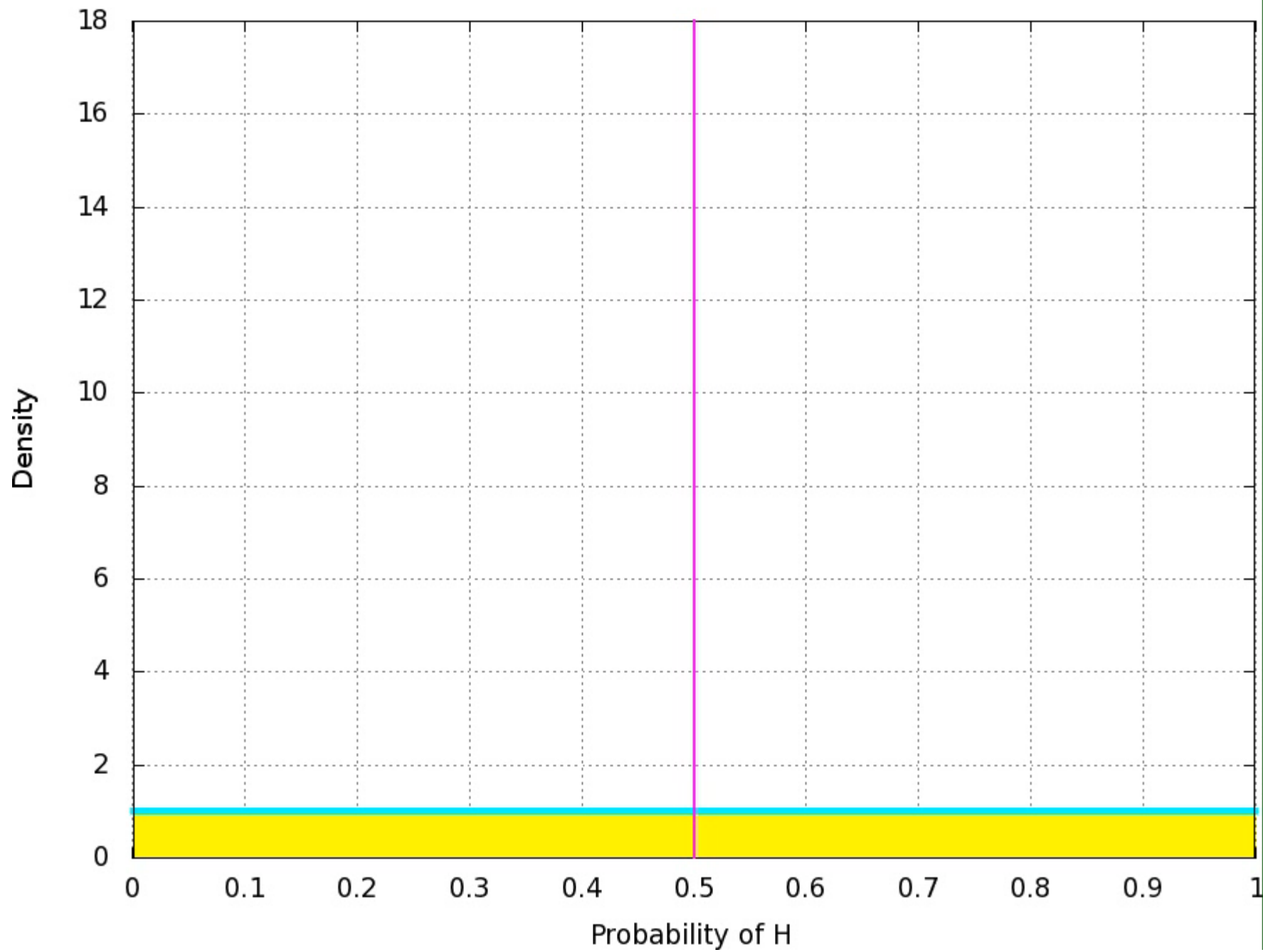
$$\begin{aligned} \log [\Pr(p_H, p_T | n_H, n_T)] &= \log \left[\frac{(n_H + n_T)!}{n_H! n_T!} \right] + n_H \log(p_H) + n_T \log(p_T) \\ &= \log \left[\frac{(n_H + n_T)!}{n_H! n_T!} \right] + n_H \log(p_H) + (n - n_H) \log(1 - p_H) \end{aligned}$$

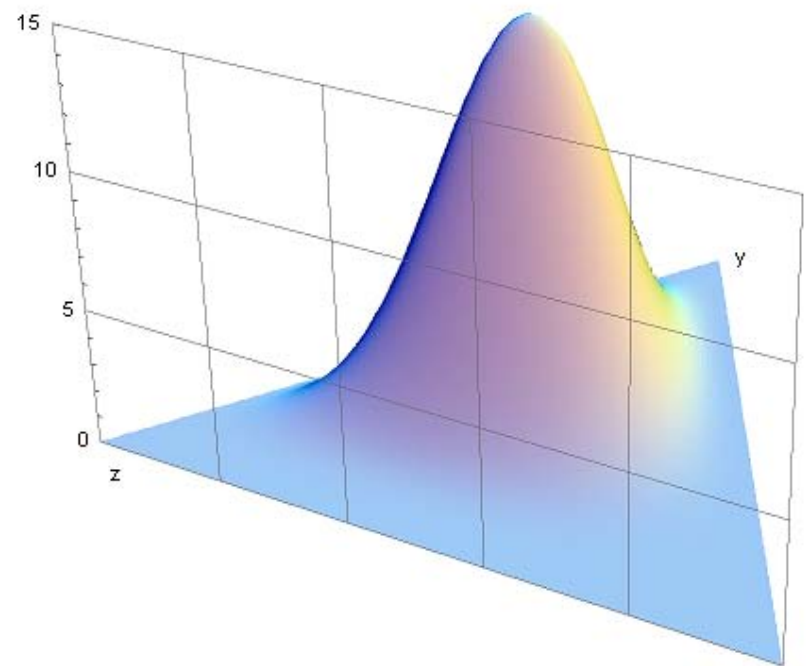
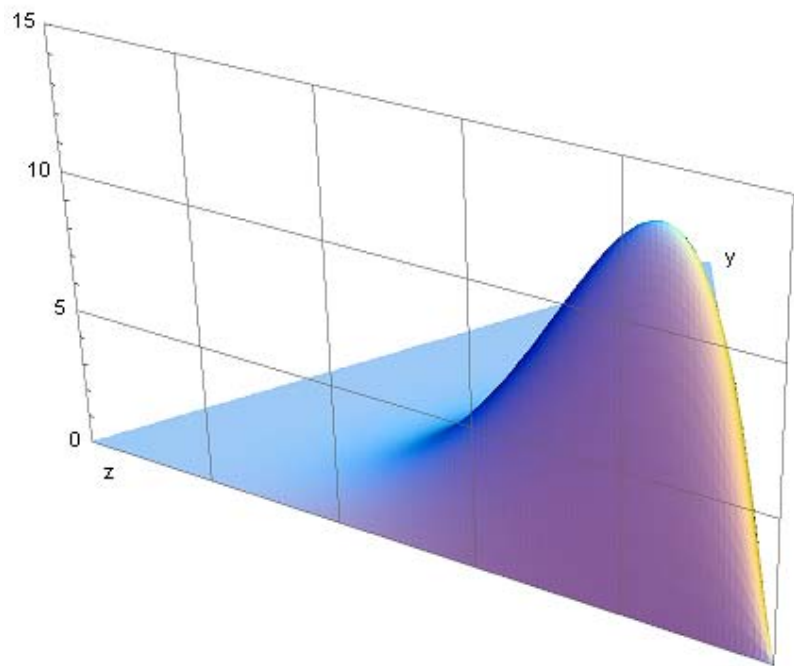
$$\frac{\partial}{\partial p_H} = 0 \quad \Rightarrow \quad \frac{n_H}{p_H} - \frac{n - n_H}{1 - p_H} = 0$$

$$\Rightarrow \quad \Pr(p_H | n, n_H) \text{ is maximized when } p_H = \frac{n_H}{n}$$

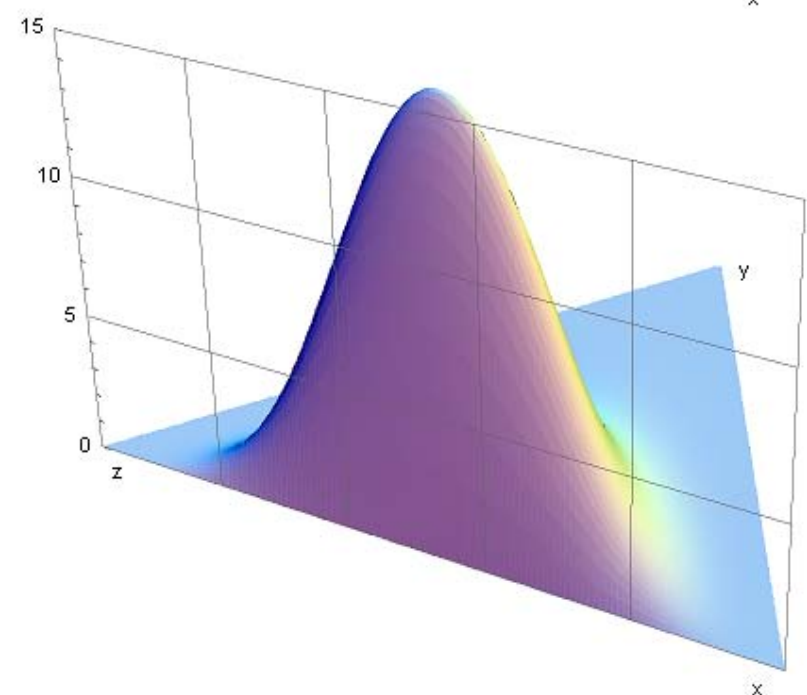
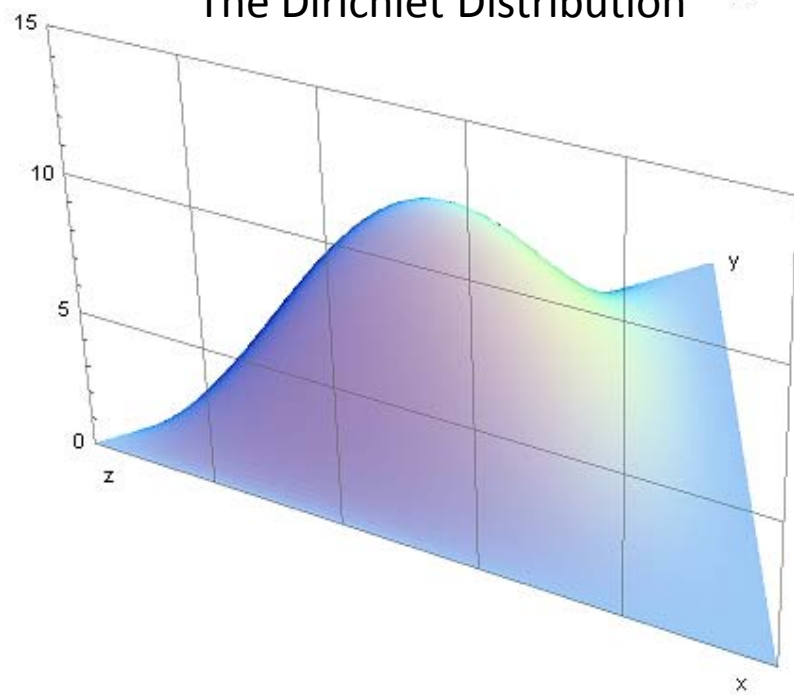
That's a lot of math for something that most people think is rather obvious...

N:000, H:000, T:000





The Dirichlet Distribution



The **point**... mathematically, even though

$$p_i \approx \frac{n_i}{\sum_j n_j} \quad n_i = 0 \not\Rightarrow p_i = 0$$

There is a **enormous** difference between a *biological* zero and a *mathematical* zero!

(Distinguishing “rare” vs. “impossible” is almost the entire basis of Shannon’s Theory of Information...)

$$\mathbb{E} \left[\log(p_i) \right] = \psi(\alpha_i) + \psi\left(\sum_j \alpha_j\right)$$

(Never Zero!)

$$\psi(z) = \frac{d}{dz} \ln \Gamma(z) \quad \alpha_i = n_i + \frac{1}{2}$$

You can think of this as a “pseudo-count”



(You would be completely wrong, but at least you would feel comforted...)

Logarithms

Why use $\log(p_i)$?

Lactobacillus iners

asymptomatic

Both of these are **biologically** wrong!

“0.01% to 0.02% = 100% change!”

“1% abundance of a virulent pathogen is negligible!”

The **“truth”** is likely somewhere in-between, and varies by **specific organism!**

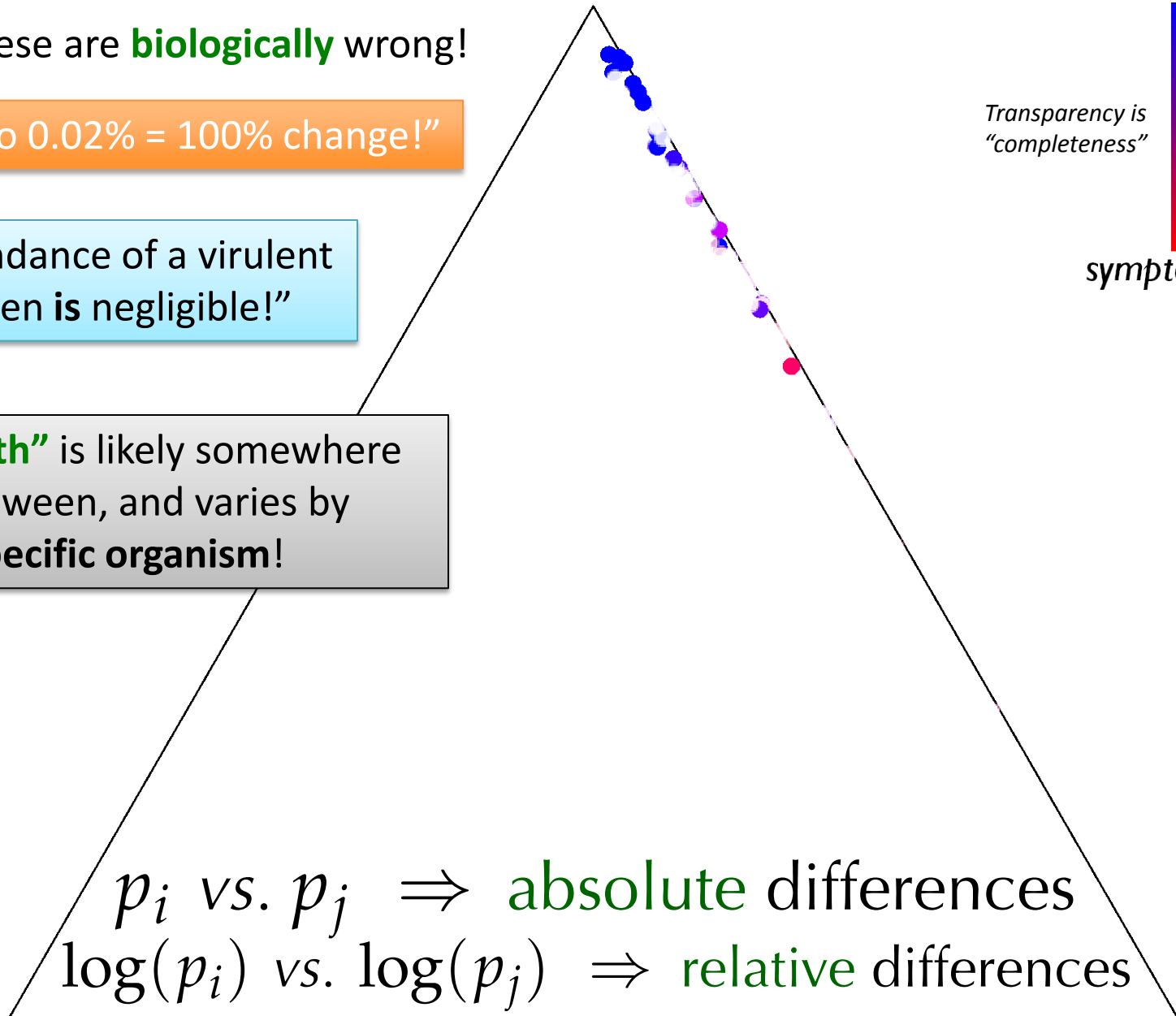
Transparency is “completeness”

symptomatic

p_i vs. $p_j \Rightarrow$ **absolute** differences
 $\log(p_i)$ vs. $\log(p_j) \Rightarrow$ **relative** differences

Prevotella bivia

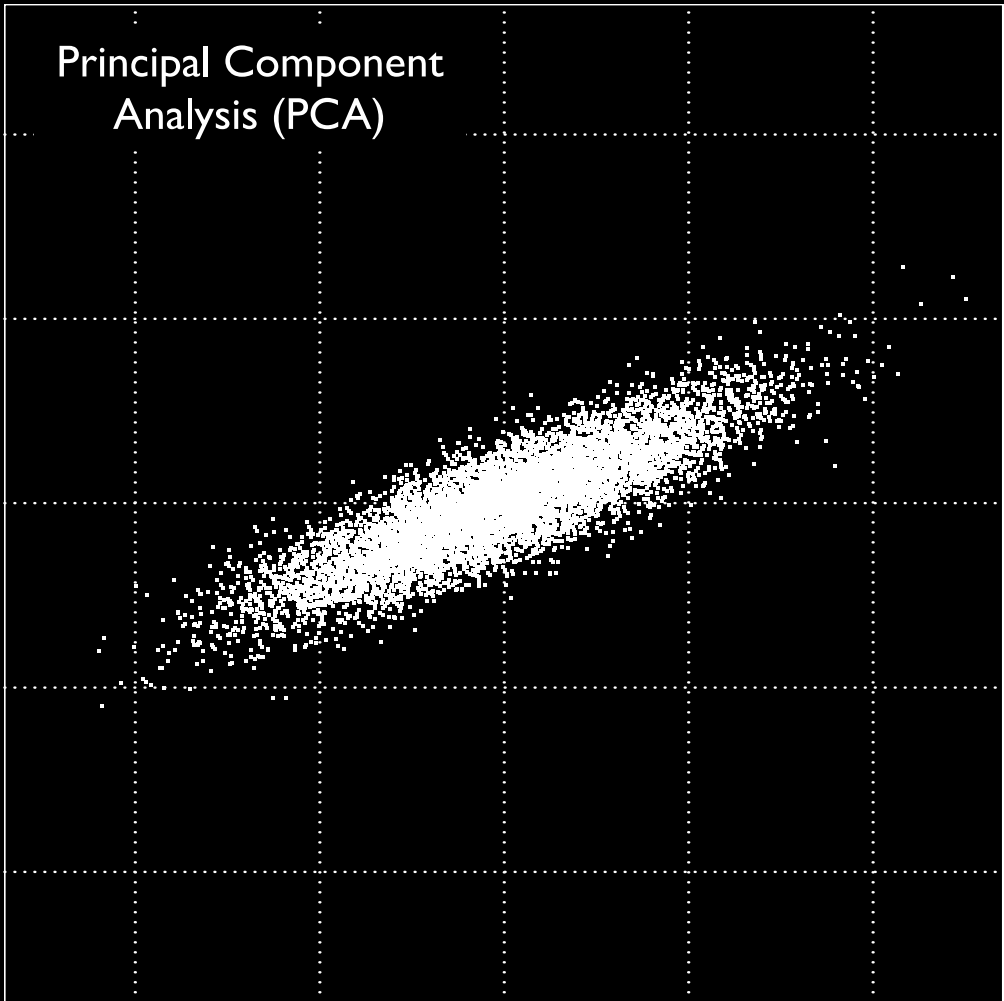
Gardnerella vaginalis



Principal Component Analysis

Why use $\log(p_i)$?

Principal Component
Analysis (PCA)



x_2

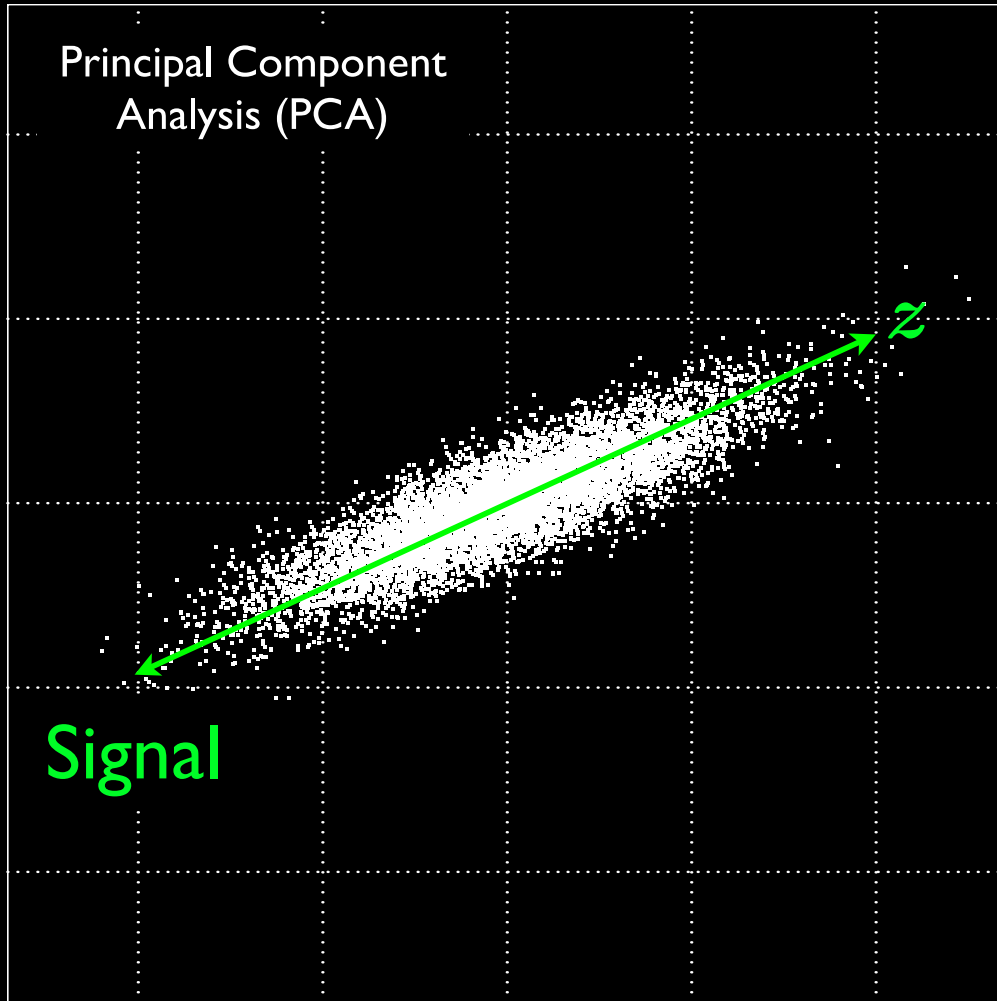
x_1

Principal Component
Analysis (PCA)

x_2

Signal

x_1

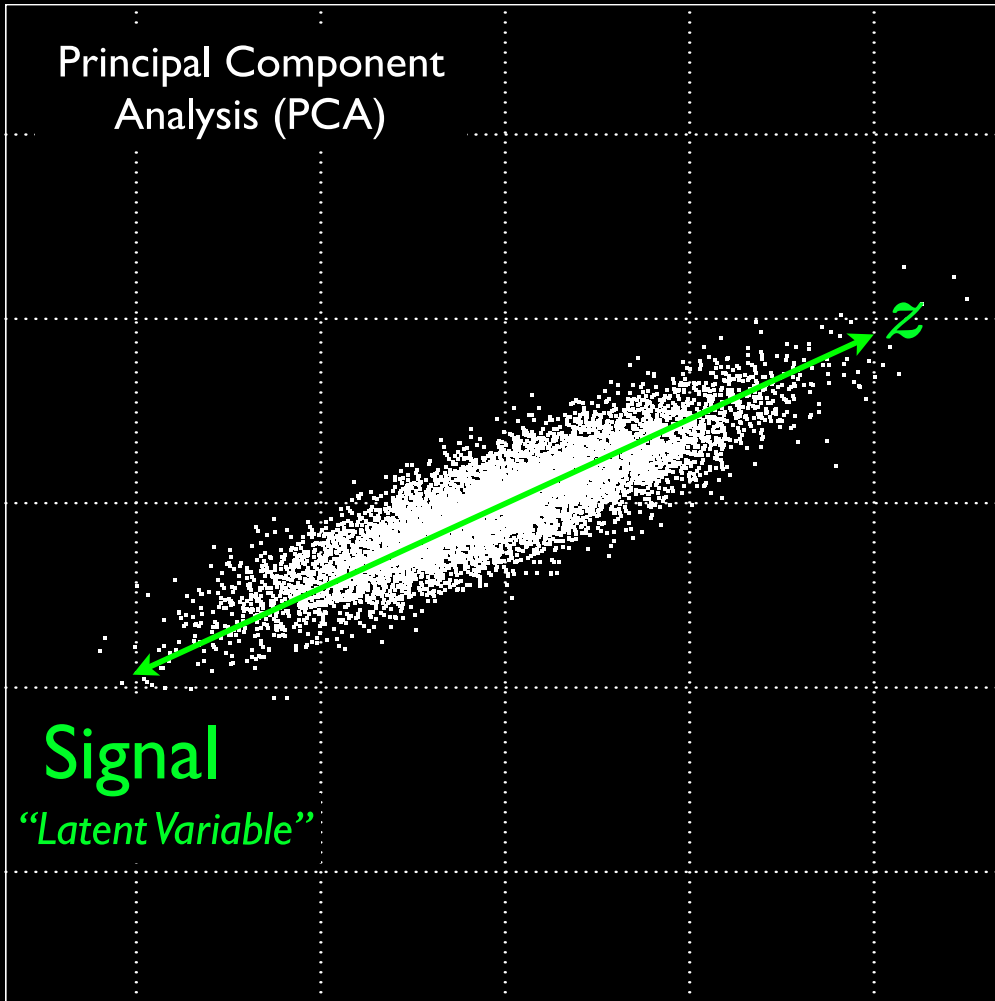


Principal Component
Analysis (PCA)

x_2

Signal
"Latent Variable"

x_1



Principal Component
Analysis (PCA)

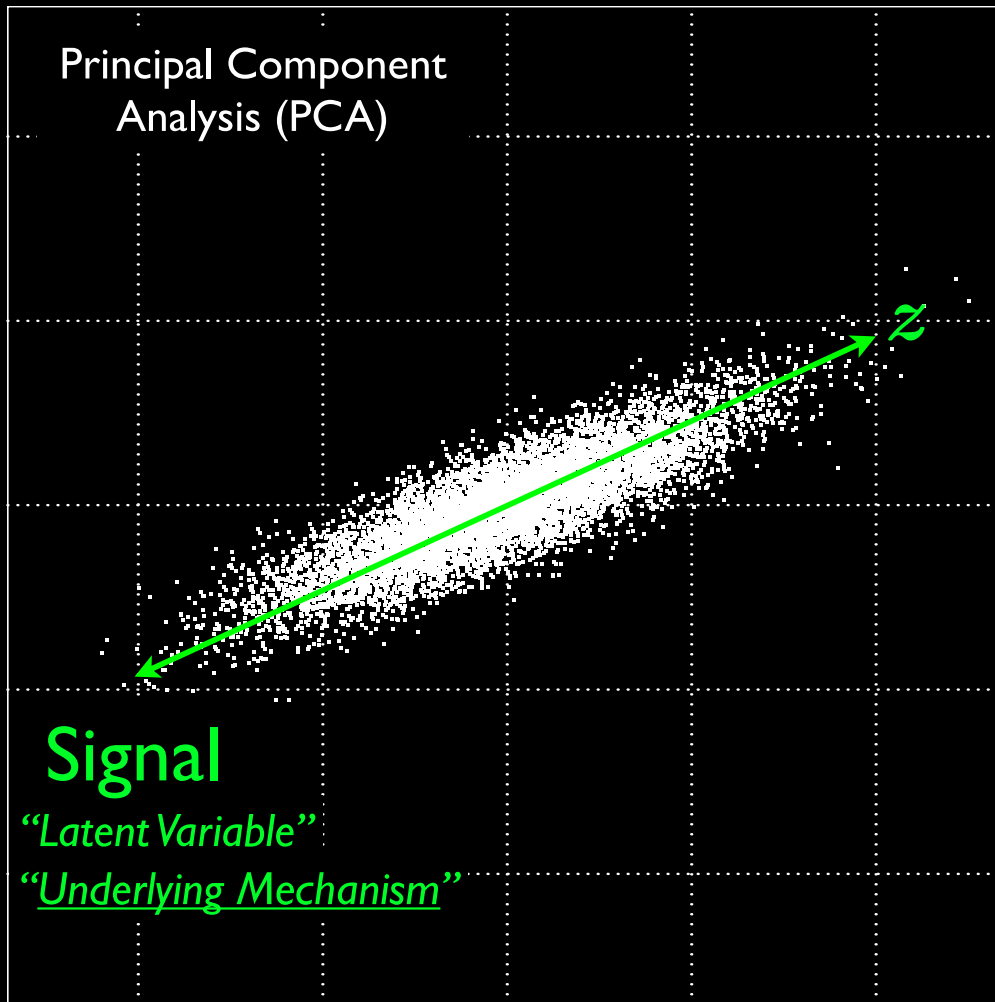
x_2

Signal

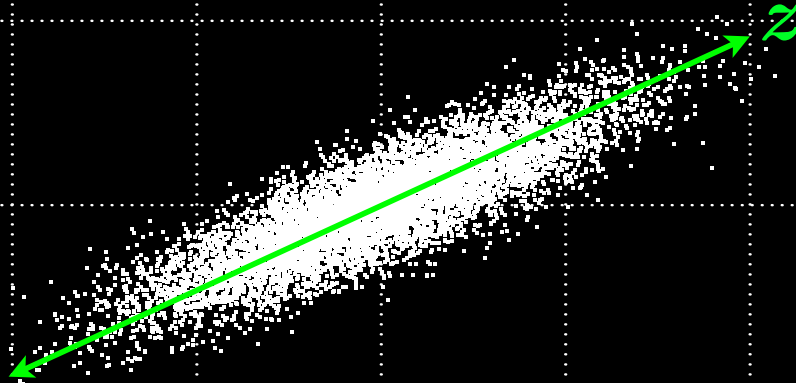
“Latent Variable”

“Underlying Mechanism”

x_1



Principal Component Analysis (PCA)



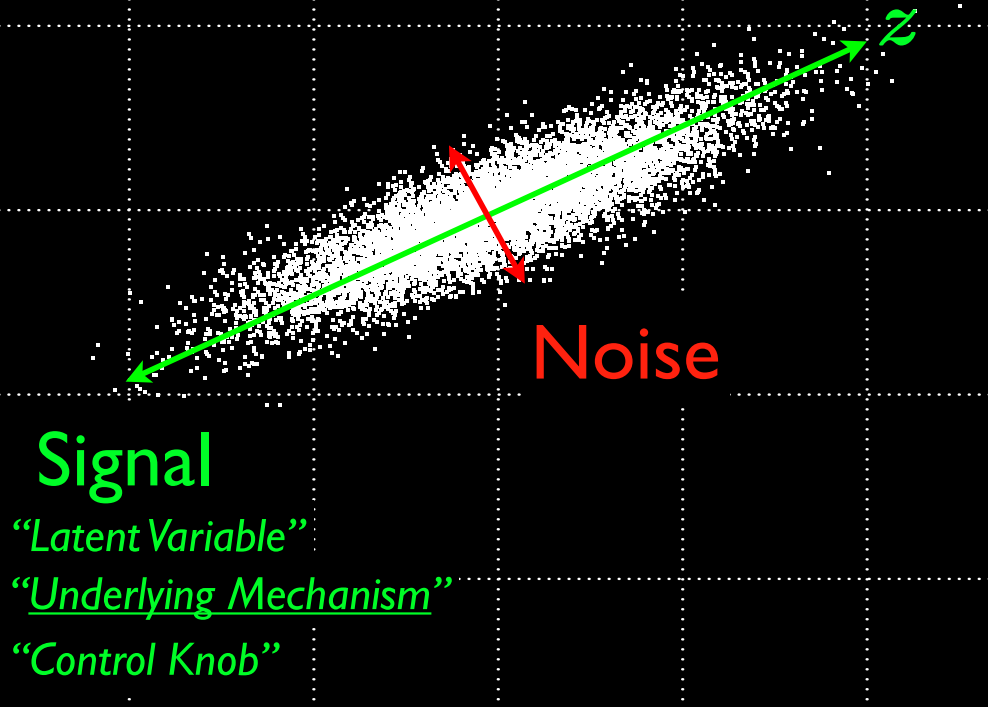
Signal

“Latent Variable”

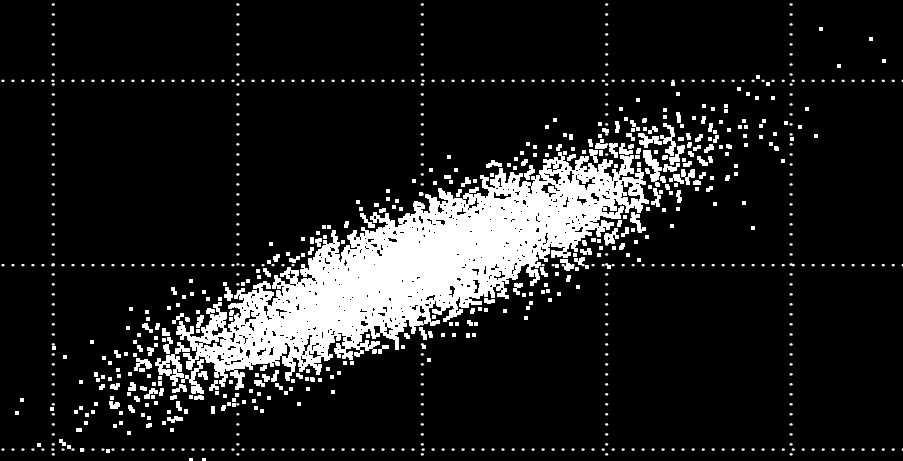
“Underlying Mechanism”

“Control Knob”

Principal Component Analysis (PCA)



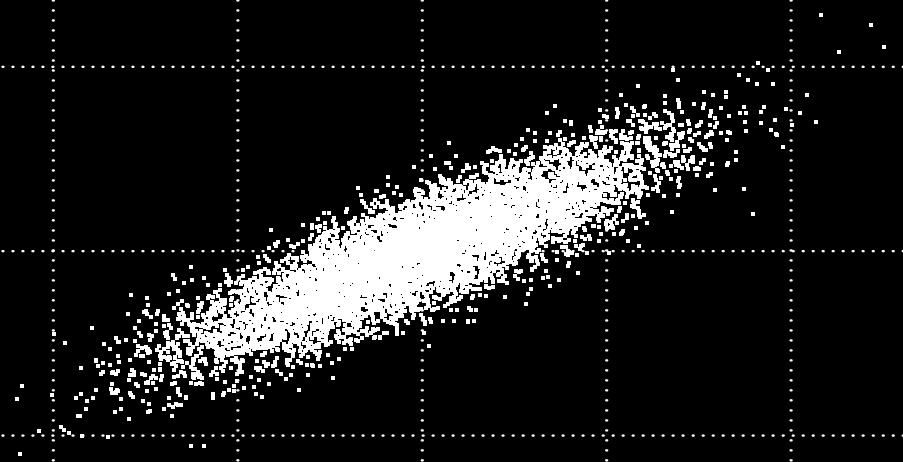
onent
Analysis (PCA)



“Control Knob”



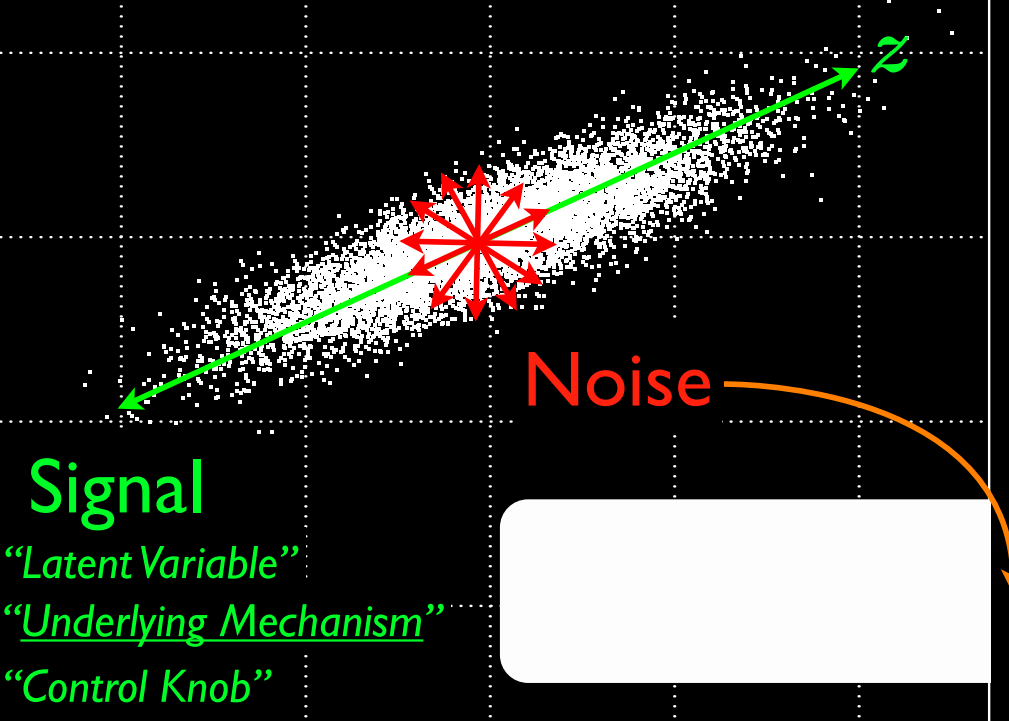
Component Analysis (PCA)



“Control Knob”



Principal Component Analysis (PCA)



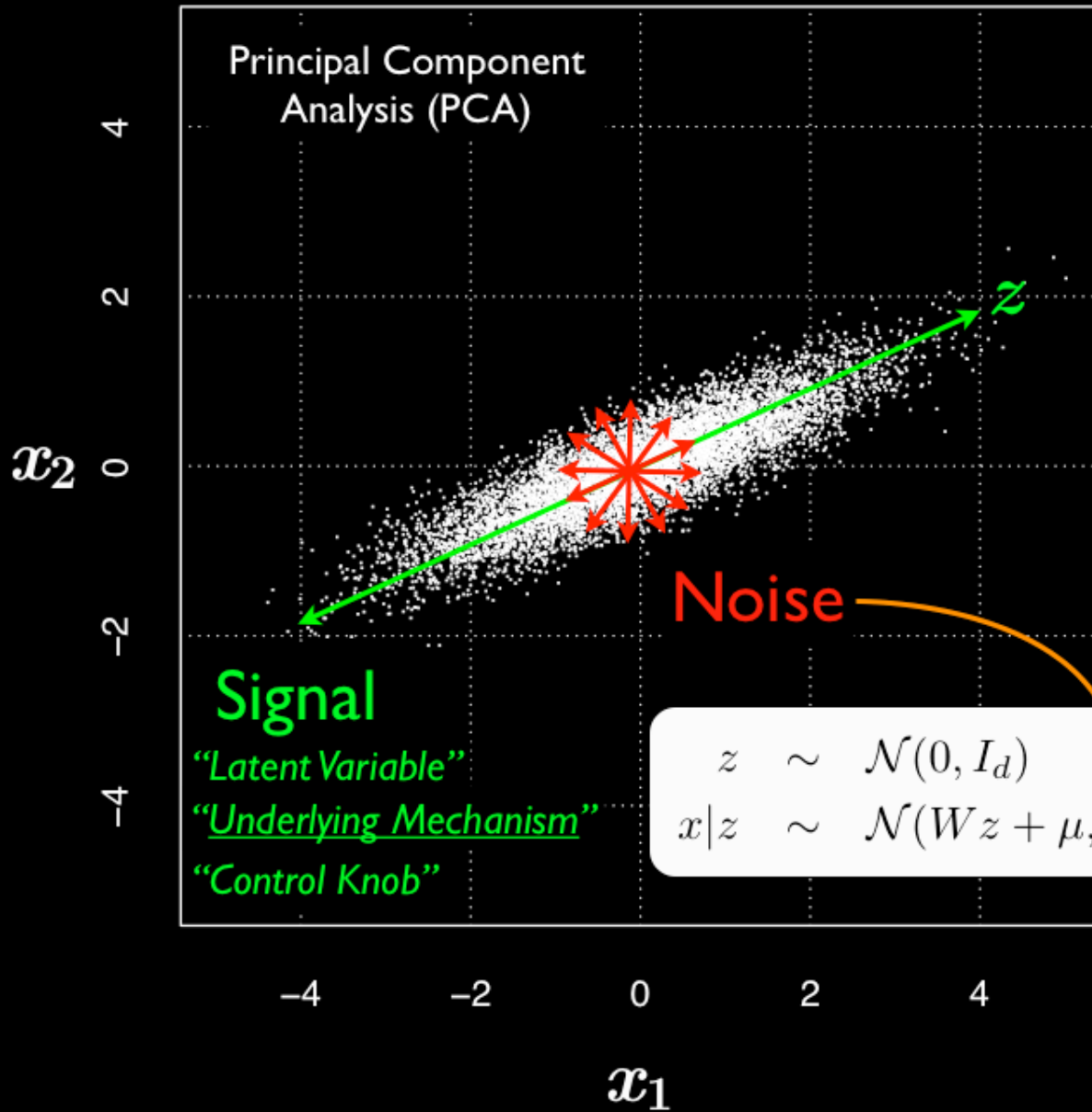
Signal

- “Latent Variable”
- “Underlying Mechanism”
- “Control Knob”

Noise



Principal Component Analysis (PCA)



“Latent Variable”
“Underlying Mechanism”
“Control Knob”

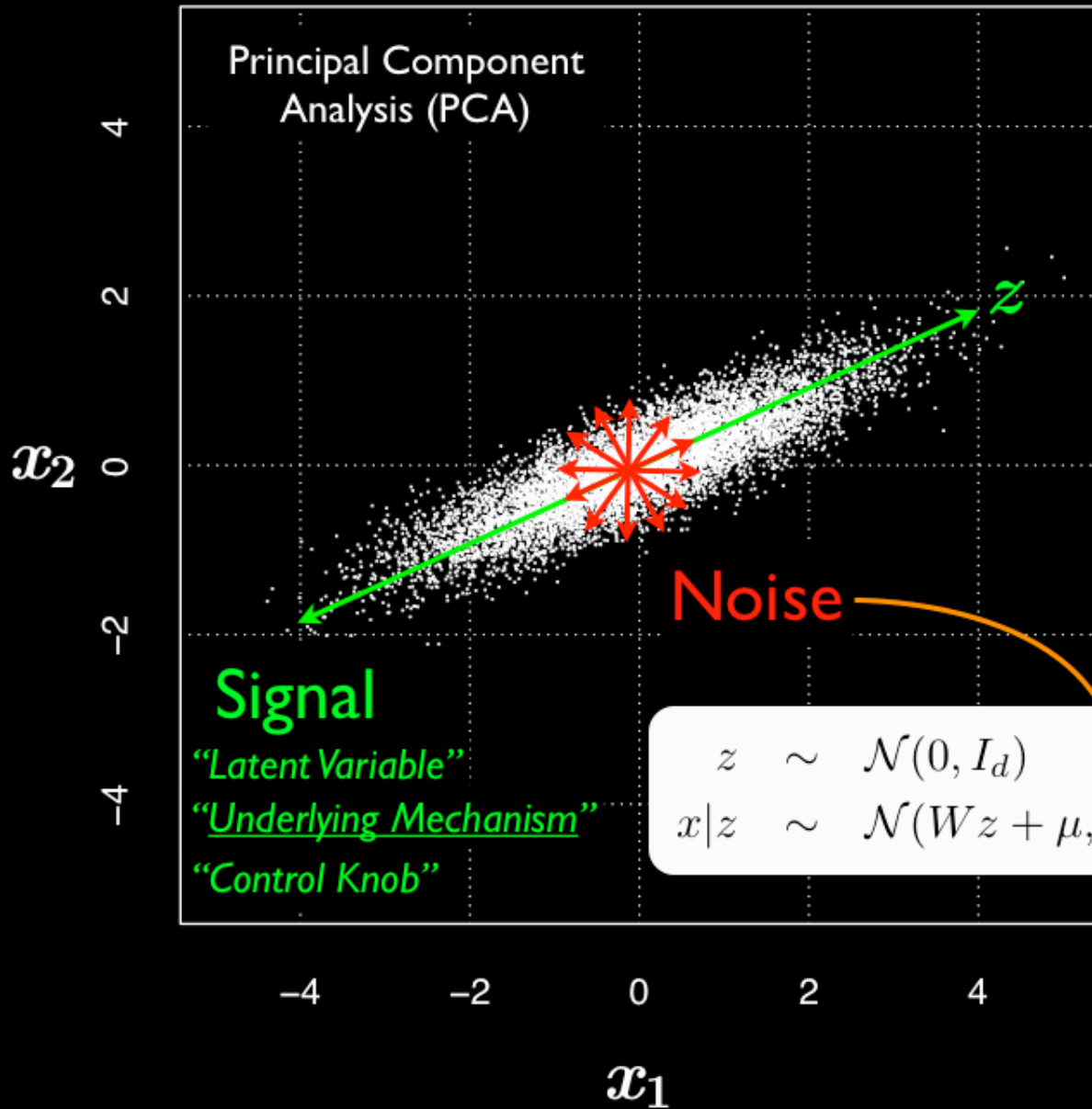
$$z \sim \mathcal{N}(0, I_d)$$
$$x|z \sim \mathcal{N}(Wz + \mu, \sigma^2 I_m), \quad \sigma > 0, \quad W \in \mathbb{R}^{md}$$

Tipping and Bishop (1999)

The PCA variance decomposition has **strange** biological implications:

- ① Every fraction has the **same** proportion of stochastic variation

Principal Component Analysis (PCA)



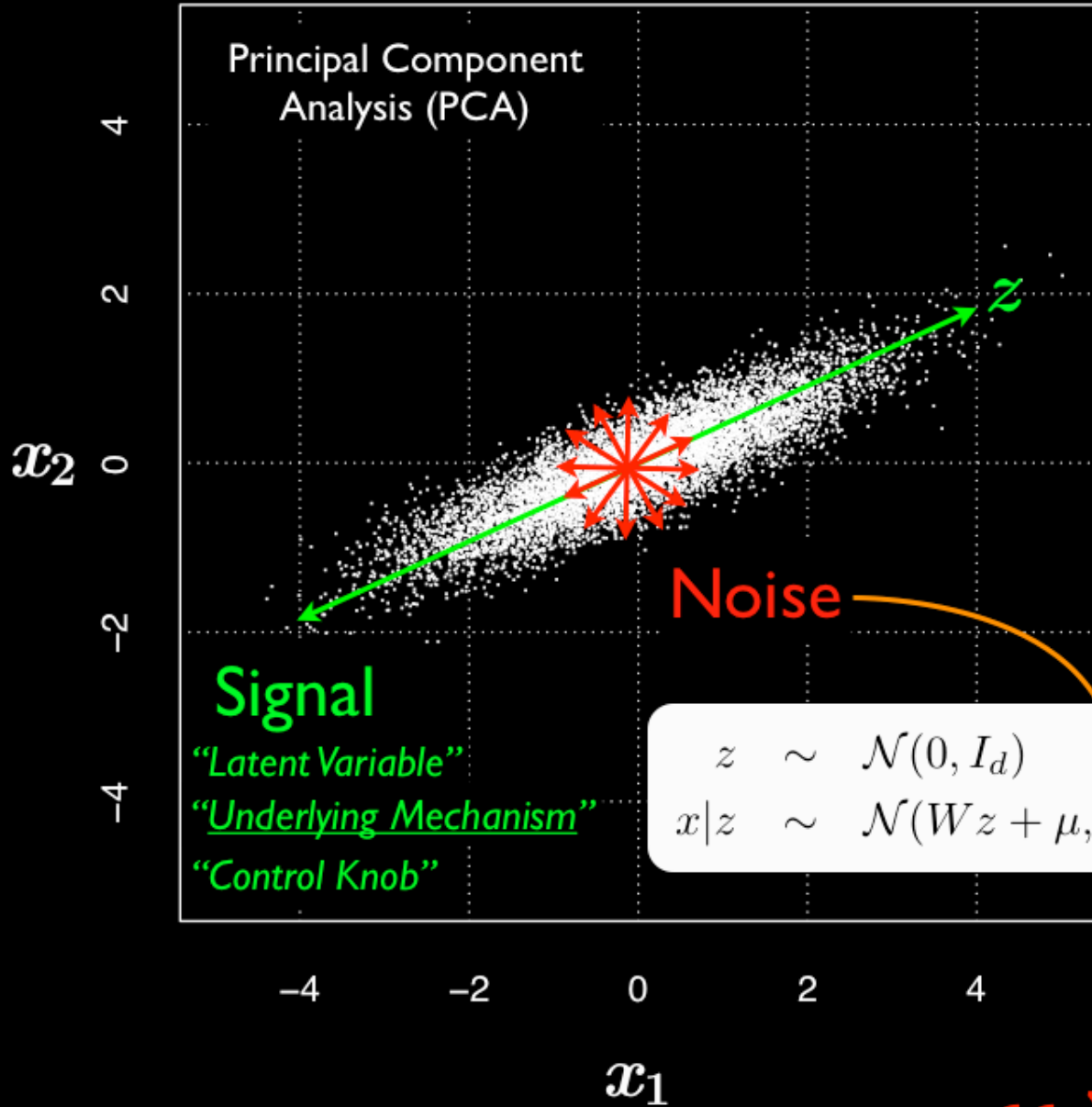
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Principal Component Analysis (PCA)



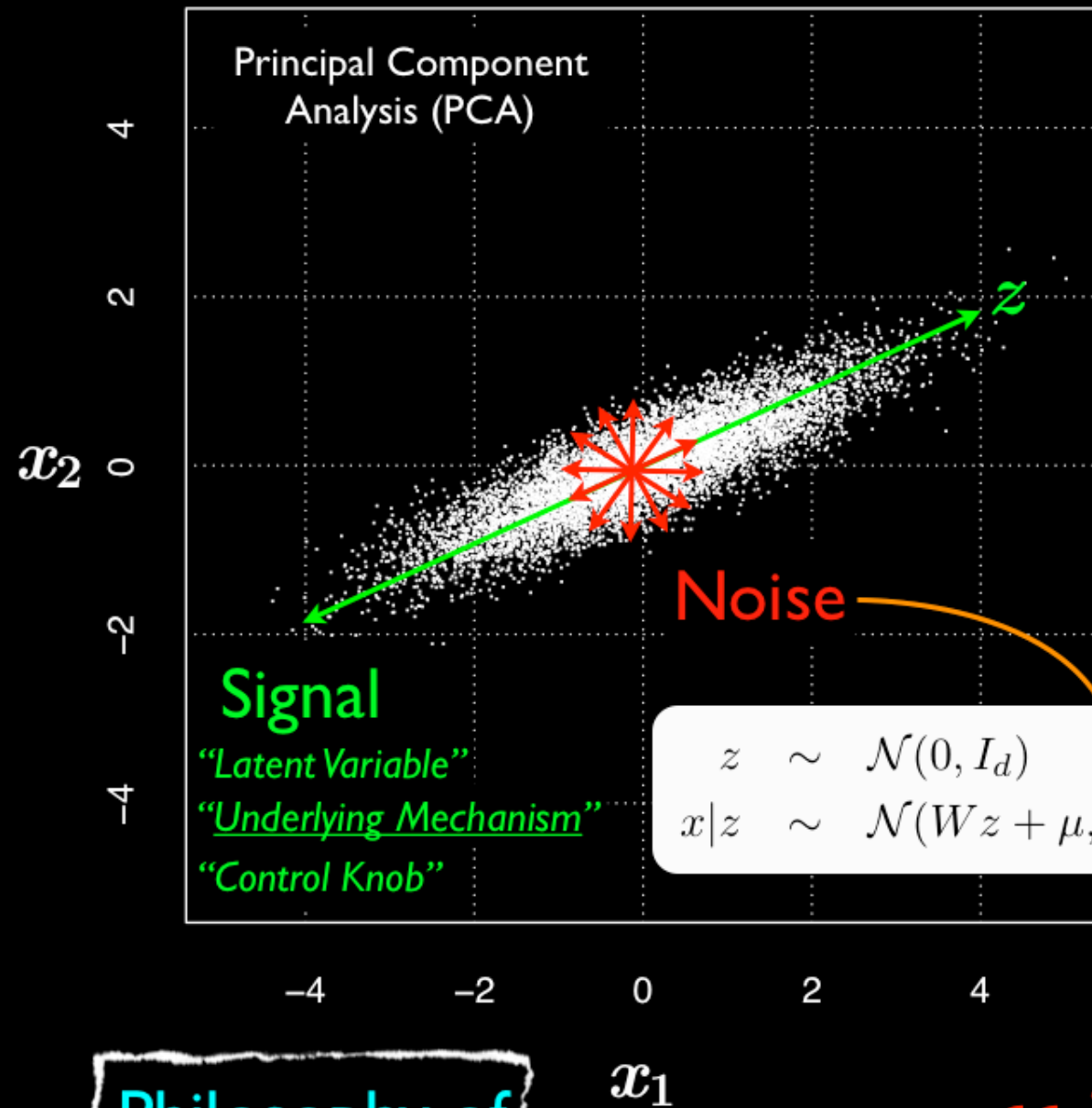
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Tipping and Bishop (1999)

“ Σ or ρ ” ?



The PCA variance decomposition has **strange** biological implications:

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Tipping and Bishop (1999)

Philosophy of Modelling

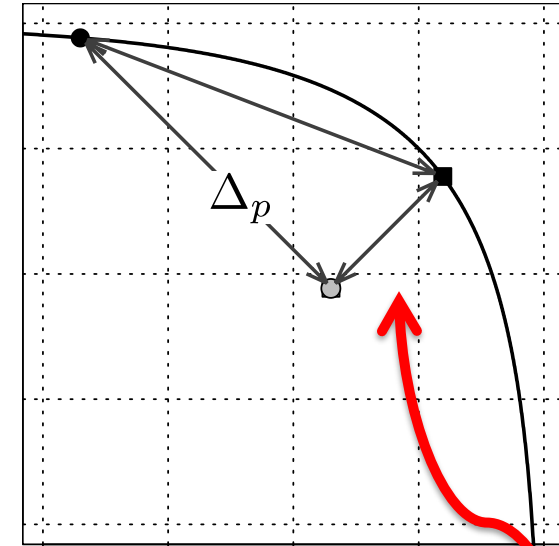
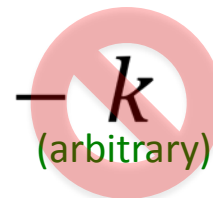
“ Σ or ρ ”?

PCA also introduces systematic distortions in the analysis!

$$\begin{bmatrix} p_i \\ q_i \\ r_i \end{bmatrix} \Rightarrow \begin{bmatrix} p_i / (p_i + q_i + r_i) \\ q_i / (p_i + q_i + r_i) \\ r_i / (p_i + q_i + r_i) \end{bmatrix}$$

$$\begin{bmatrix} \log(p_i / (p_i + q_i + r_i)) \\ \log(q_i / (p_i + q_i + r_i)) \\ \log(r_i / (p_i + q_i + r_i)) \end{bmatrix} \Rightarrow \begin{bmatrix} \log(p_i) - \log(p_i + q_i + r_i) \\ \log(q_i) - \log(p_i + q_i + r_i) \\ \log(r_i) - \log(p_i + q_i + r_i) \end{bmatrix}$$

$$\begin{bmatrix} \log(p_i) - \log(p_i + q_i + r_i) \\ \log(q_i) - \log(p_i + q_i + r_i) \\ \log(r_i) - \log(p_i + q_i + r_i) \end{bmatrix} \Rightarrow \begin{bmatrix} \log(p_i) \\ \log(q_i) \\ \log(r_i) \end{bmatrix}$$



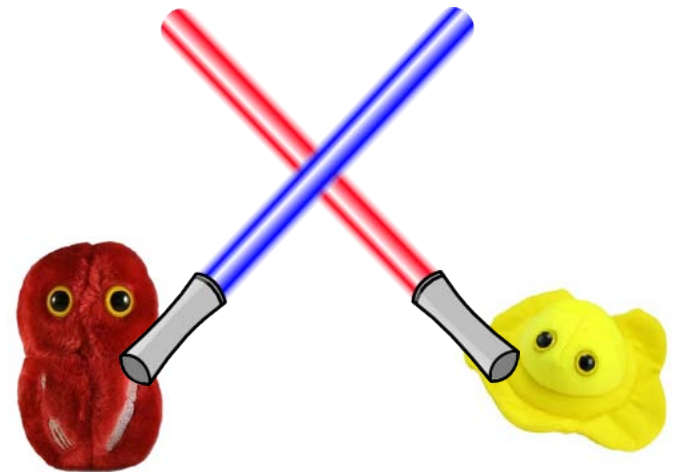
Distortions are routinely between **5% to 50% to 500%** of datum distances!

How can we tell which of these is occurring?

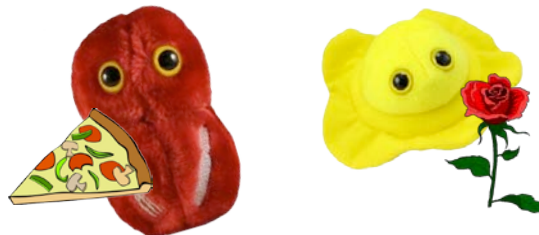
Co-Occurrence



Co-Exclusion



Indifference



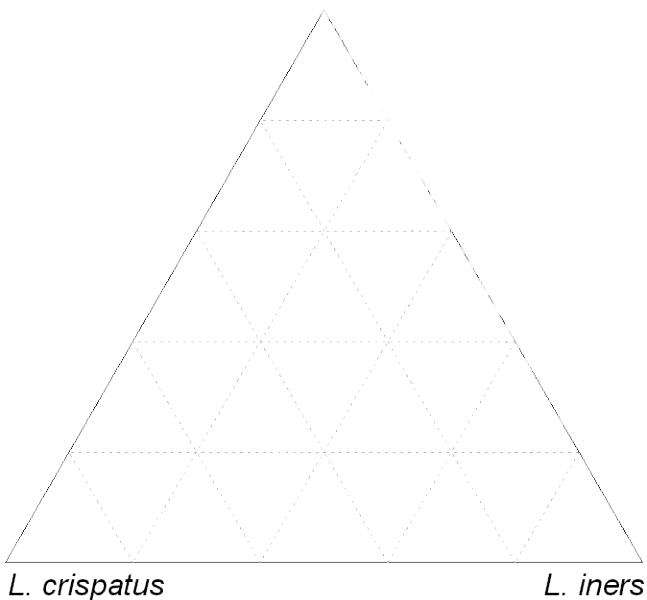
... taking into account ...

- Zero-counts (not zero-proportions)
- Not-really-absolute and not-really-relative
- Removal of systematic distortions
- **Reparameterization invariance**

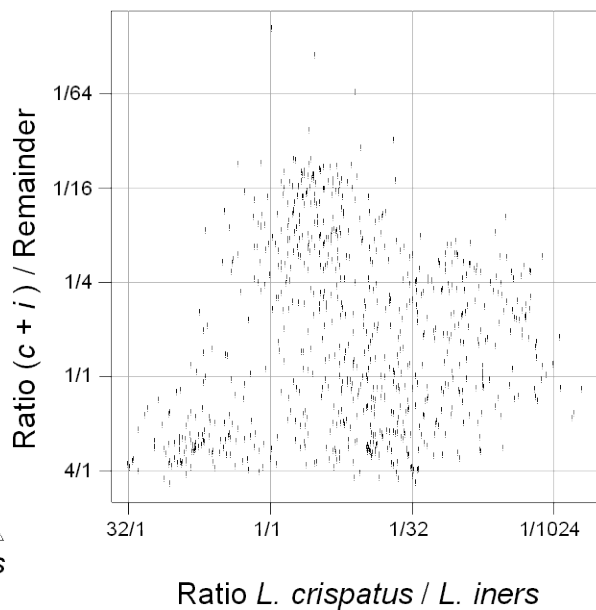
Co-Exclusion



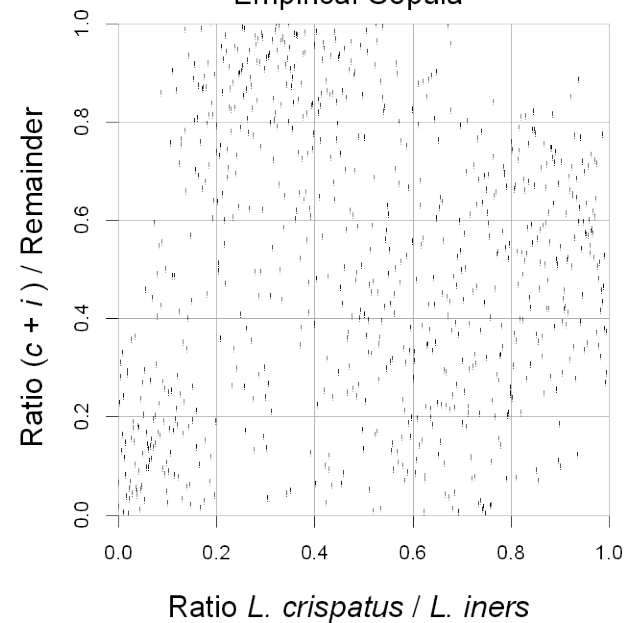
Remainder



Log Base 2

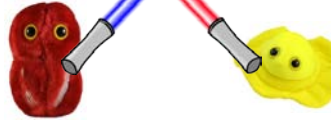


Empirical Copula



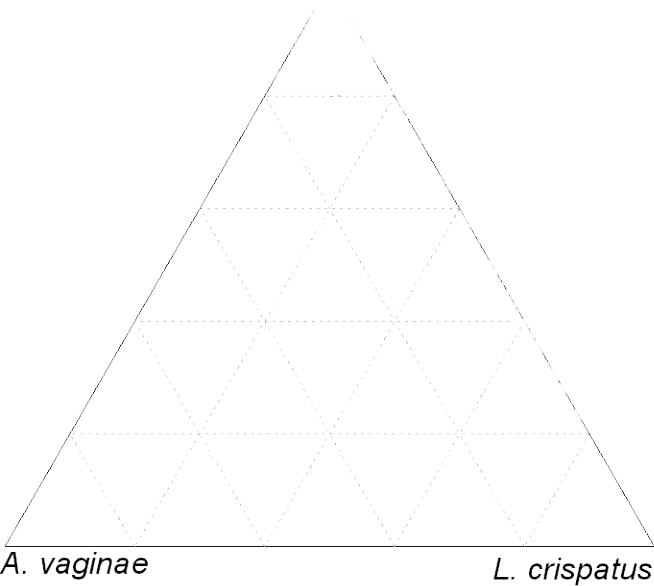
No!

Co-Exclusion

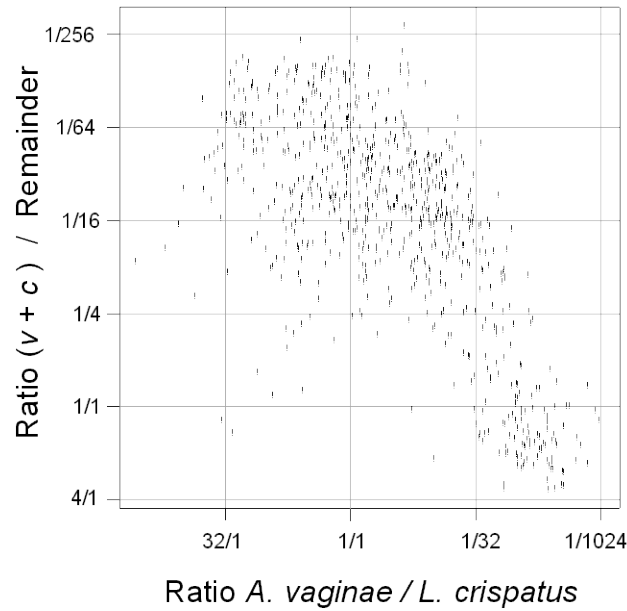


L. crispatus
A. vaginae

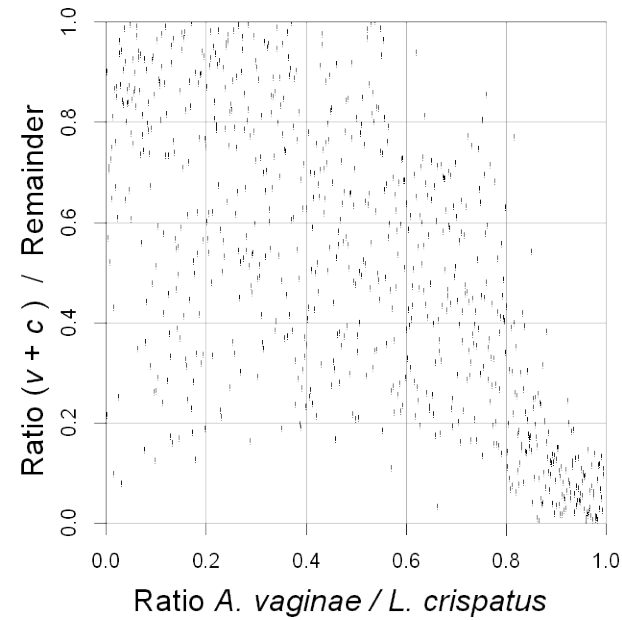
Remainder



Log Base 2



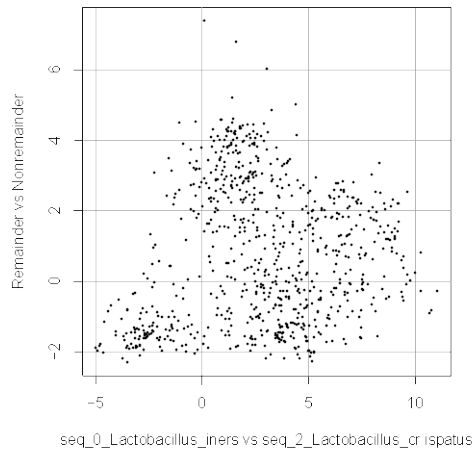
Empirical copula



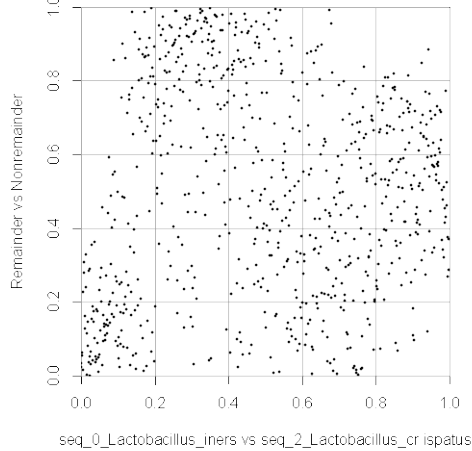
Yes!

(weakly)

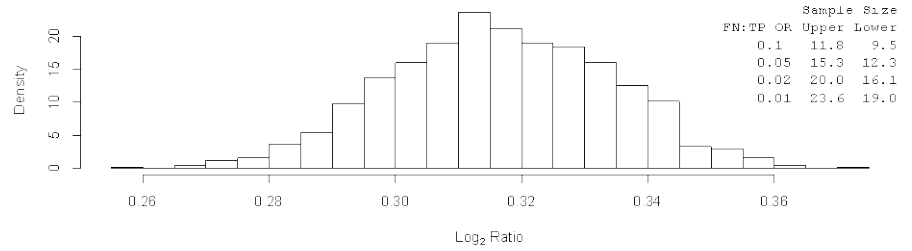
Log₂ Ratios



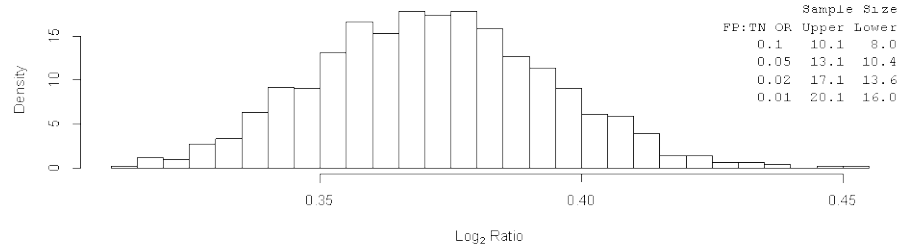
Emperical Copula



Mutual Information (TP:FN)

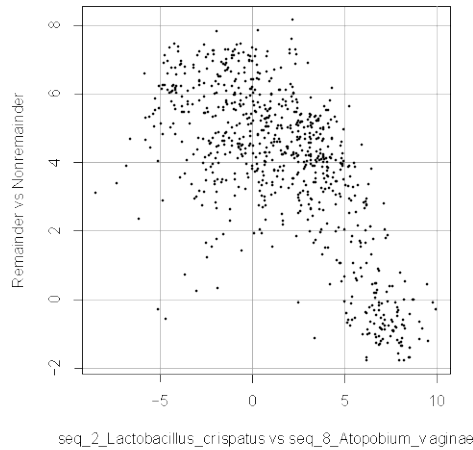


Mutual Information (TN:FP)

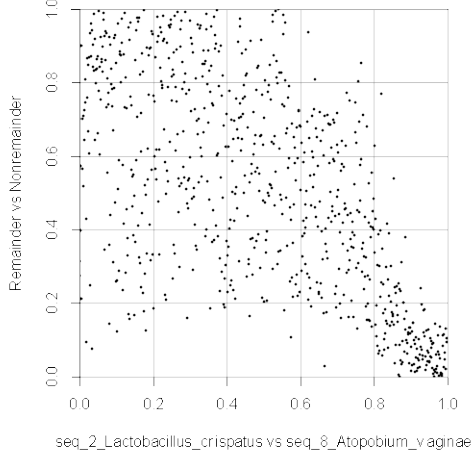


Quantifying "Uniformity"

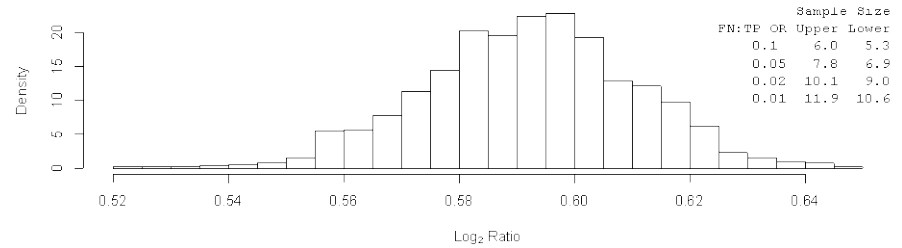
Log₂ Ratios



Emperical Copula



Mutual Information (TP:FN)



Mutual Information (TN:FP)

