

# Mutual information for functions, maybe even ERPs.

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# Shannon's Entropy 1

$$H(X) = - \sum_x p(x) \log_2 p(x)$$

## Shannon's Entropy 2

1/2	1/4	1/8	1/16	1/32	1/64	1/128	1/128
000	001	010	011	100	101	110	111
0	10	110	1110	11110	111110	1111110	1111111

$$\text{average code length} = \frac{1}{2} + \frac{1}{4}2 + \frac{1}{8}3 + \frac{1}{16}4 + \dots = H(X) \approx 1.98 < 3$$

# Mutual Information 1

$$I(X, Y) = \sum_{x,y} p_{X,Y}(x, y) \log_2 \frac{p_{X,Y}(x, y)}{p_X(x)p_Y(y)}$$

# Mutual Information 2

$$I(X, Y) = H(X) - H(X|Y) = H(Y) - H(Y|X)$$

## Mutual Information 3

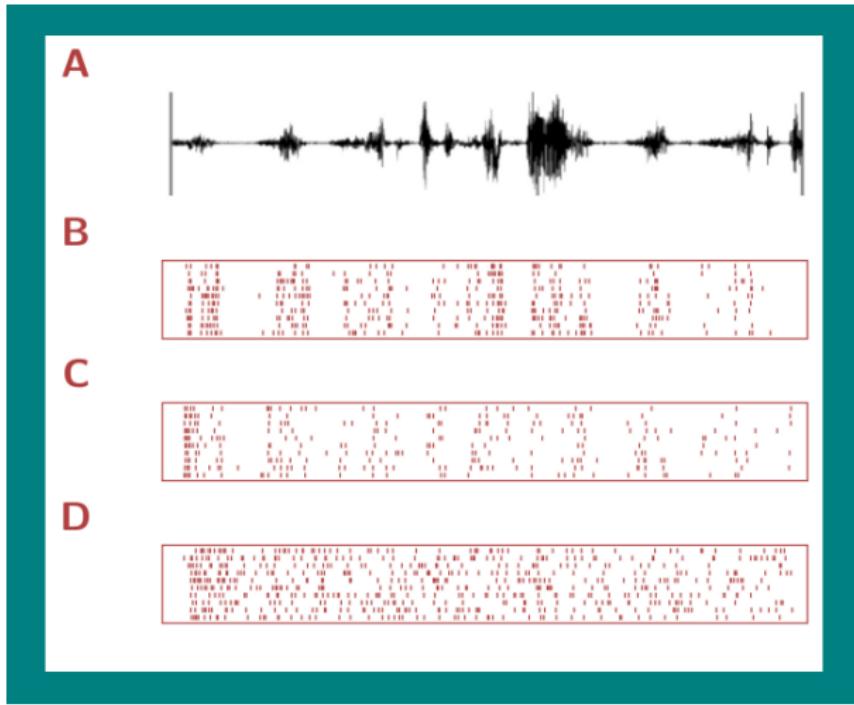
$$I(X, Y) = H(X) - H(X|Y) = H(Y) - H(Y|X)$$

or, for example

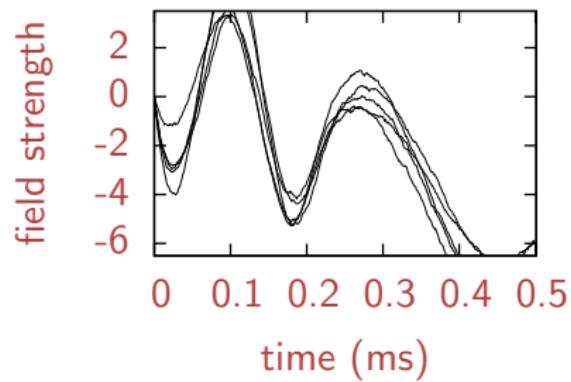
$$H(Y) = H(Y|X) + I(X, Y)$$

info in  $Y$  = (info remaining in  $Y$  if you know  $X$ ) + (mutual info)

# Spike trains



# Functions



# A dart board 1



photo from ebay (£4.20 +p.p.)

## A dart board 2



# Probability mass function



$$\text{prob(dart lands in } B) = \int_B p(x) dV$$

# Estimating using the number of holes 1



$$\langle \text{number of holes in } B \rangle = \int_B p(x) dV \times (\text{total number of holes})$$

where the total volume is normalized.

## Estimating the probability mass function

If the mass function varies slowly:

$$\int_B p(\mathbf{x}) dV \approx p(\mathbf{x}_0) \times \text{vol } B$$

so

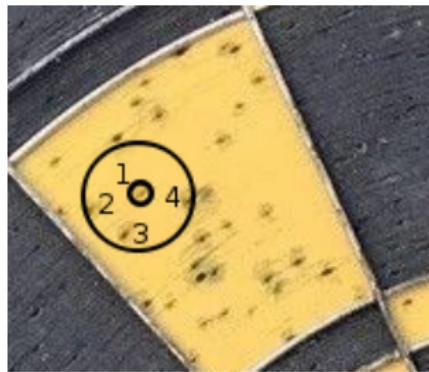
$$\text{number of holes in } B \approx p(\mathbf{x}_0) \times \text{vol } B \times (\text{total number of holes})$$

Using this to find the mutual information gives a *Kozachenko-Leonenko* estimator.

## Estimating using the number of holes 3

$$p(x_0) \approx \frac{\#B}{n \times \text{vol } B}$$

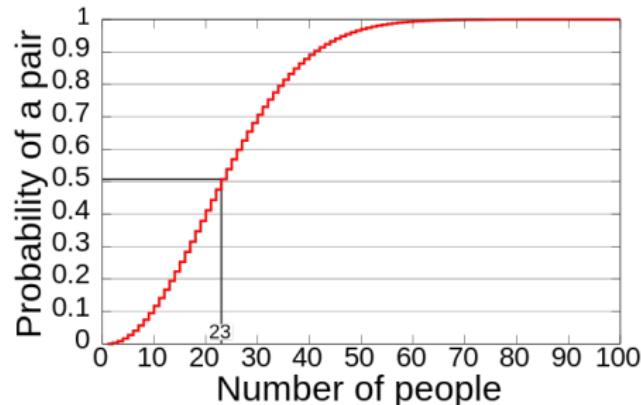
where  $n$  is the total number of points and  $\#B$  is the number of points in  $B$ .



so

$$p(\circ) = \frac{4}{n \text{vol } B}$$

# Kozachenko-Leonenko estimators are very good



graph from wikipedia article on the birthday paradox

# Problem

How do we work out the volume in the space of functions? We have no coordinates xyz to do

$$\text{vol } B = \int_B dx dy dz$$

# Use the mass function as a measure!

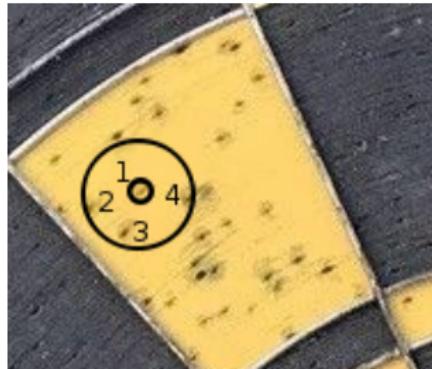


$$\text{vol } B = \int_B p(\mathbf{x}) dV$$

# Volume by counting holes 1

$$\text{vol } B \approx \frac{\text{number of holes in } B}{\text{total number of holes}}$$

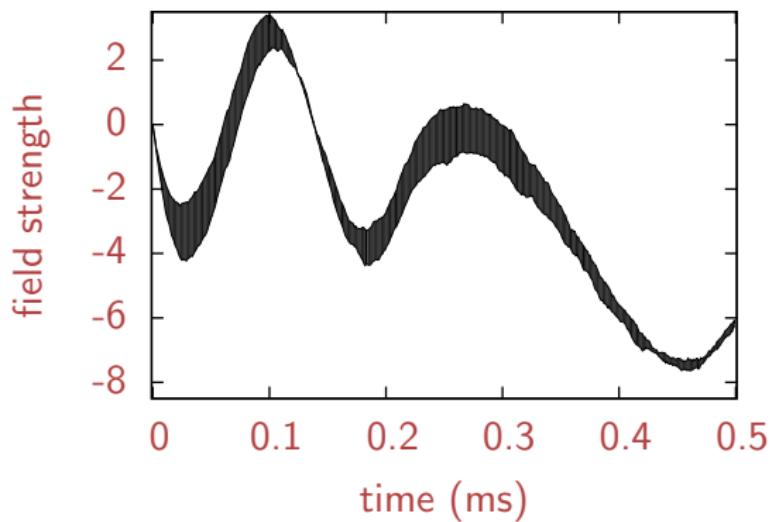
## Volume by counting holes 2



A ball with volume  $h/n$  around the circled point, where  $n$  is the total number of holes and  $h = 4$ .

## Metric

To make a ball you need a metric; not to measure the radius since the size is being defined by the volume, but to define ‘the nearest  $h$  points’.



Oh no

$$p(\mathbf{x}_0) \approx \frac{\#B}{n \times \text{vol } B} = \frac{h}{nh/n} = 1$$

and using this measure gives  $H(X) = 0$ ; in fact the differential entropy is not well-defined. However the mutual information is!

# Mutual information

$$I(X, Y) = H(Y) - H(Y|X)$$

has two probability distributions:  $p_Y(y)$  and  $p_{Y|X}(y|x)$ !

IDEA: use one to estimate volume, the other can then be estimated by counting!

## Formula 1

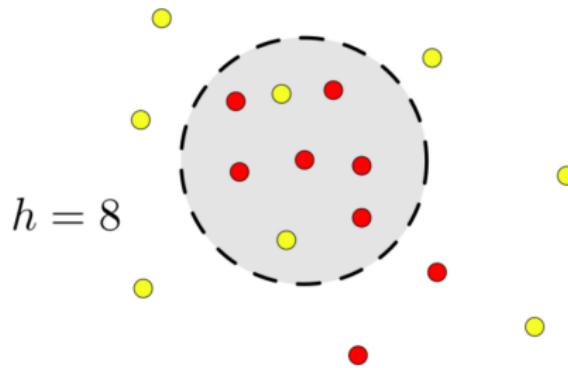
This is for the case where  $X$  is a discrete random variable and everything exciting is happening in  $Y$  space.

$$I(X, Y) = \frac{1}{n} \sum_{y_i} \log_2 \frac{n \#_{y_i} B}{h}$$

where  $\#_{y_i} B$  are the number of points in  $B$  that correspond to the  $X$  value as  $y$  and  $n_s$  is the number of stimuli.

## Formula 2

$$I(X, Y) = \frac{1}{n} \sum_{y_i} \log_2 \frac{n_s \#_{y_i} B}{h}$$



There are two approximations:

$$\int_B p(x)dV \approx \#B \times \text{vol } B$$

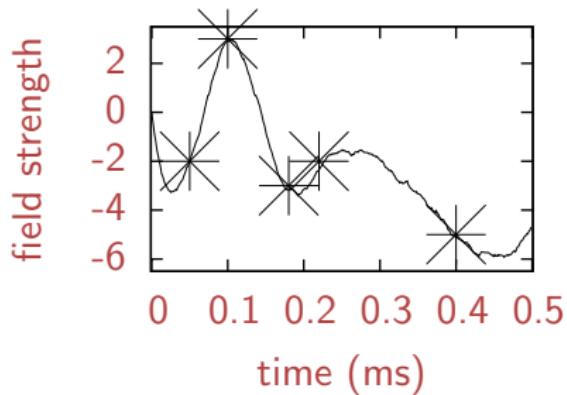
and

$$\int_B p(x)dV \approx V \times p(x_0)$$

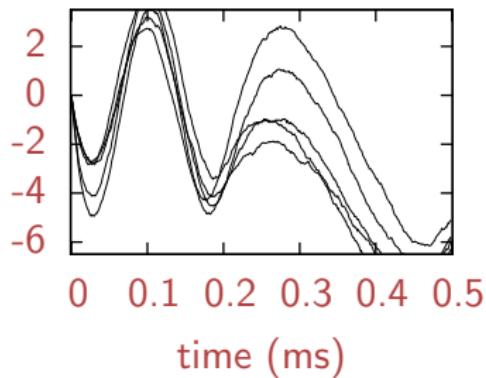
The first approximation gets better if the volume is bigger, the second gets worse; the correct choice of  $h$  is a compromise between these two. There is actually a clever approach to picking  $h$  that seems to work, based on the bias.

# Pretend ERPs 1

A



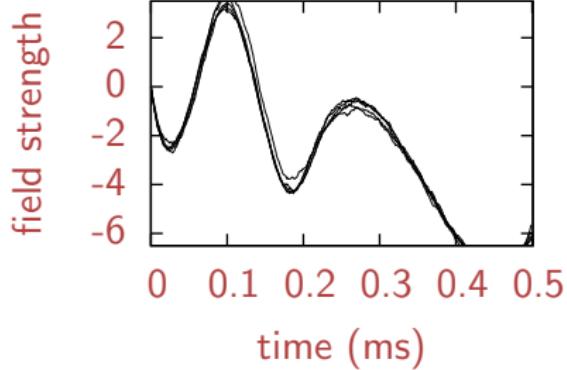
B



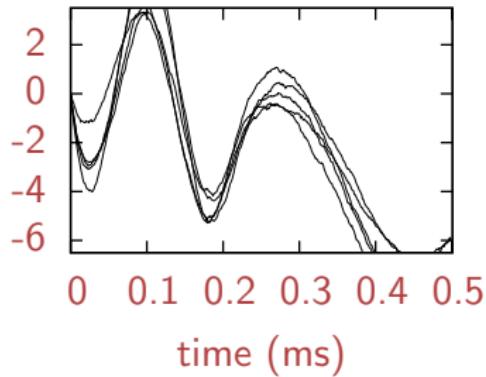
**Fictive event related potentials.** The 'stimuli' are random 5-vectors of landmarks; an ERP is produced by perturbing the landmarks, interpolating with splines and adding noise. **B** shows responses to different stimuli.

## Pretend ERPs 2

A

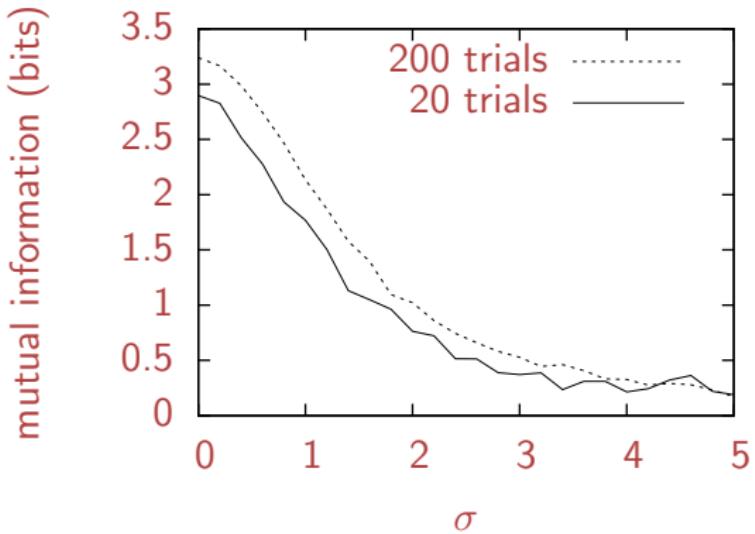


B



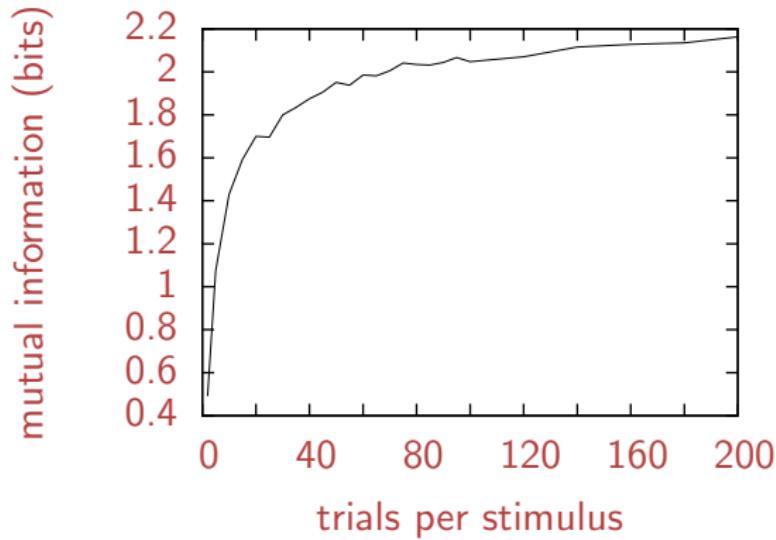
**Multiple trials to the same stimulus.** The amount the landmark points are moved is determined by  $\sigma$ ; for **A**  $\sigma = 0.0$ , for **B**  $\sigma = 2$ .

# Results 1



**Estimated mutual information.**

## Results 2



**Estimated mutual information.** Here  $\sigma = 1$ .

# The End

THANK YOU!