

# Learn to live with lack of rejection

Einar Holsbø & Kajsa Møllersen, 21. Oct 2019

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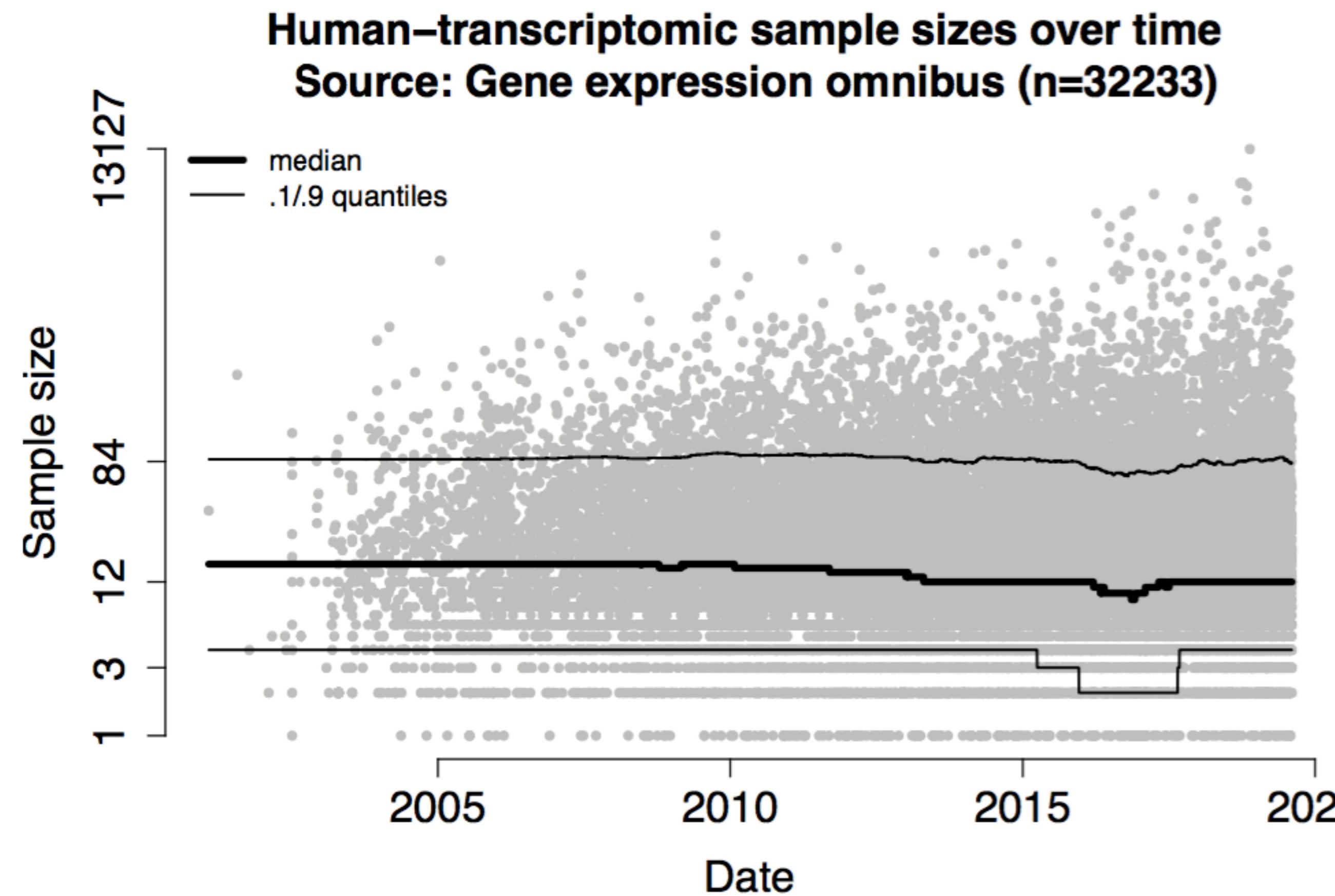
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- Typical data sizes limit the usefulness of hypothesis tests
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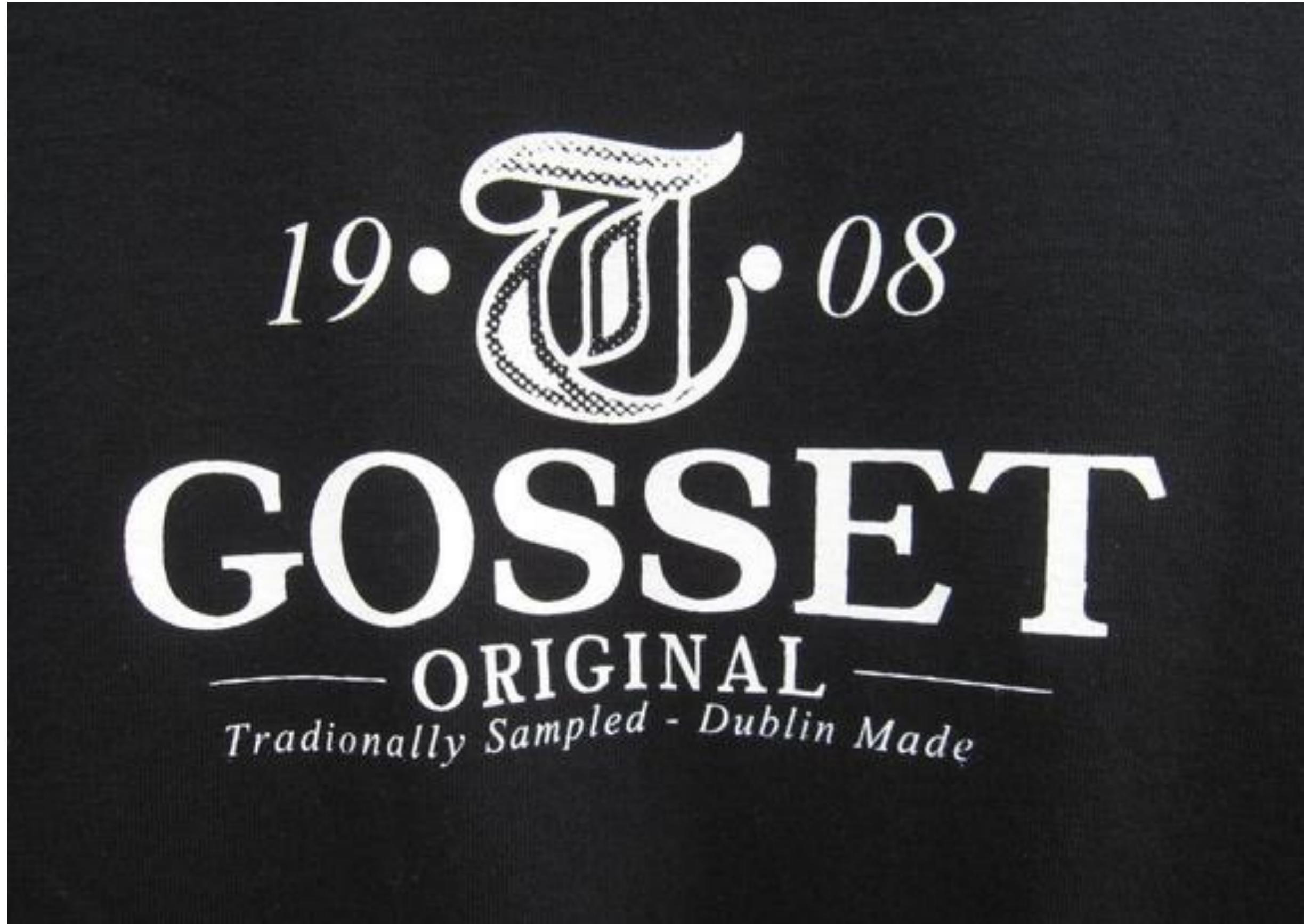
- Your datasets are tiny
- Typical data sizes limit the usefulness of hypothesis tests
- Typical avg. fold changes fall outside those limits
- Leads to all kinds of errors
- There is a better way; Kajsa will explain

More than 90% of datasets comprise under 100 observations



# Components of a hypothesis test

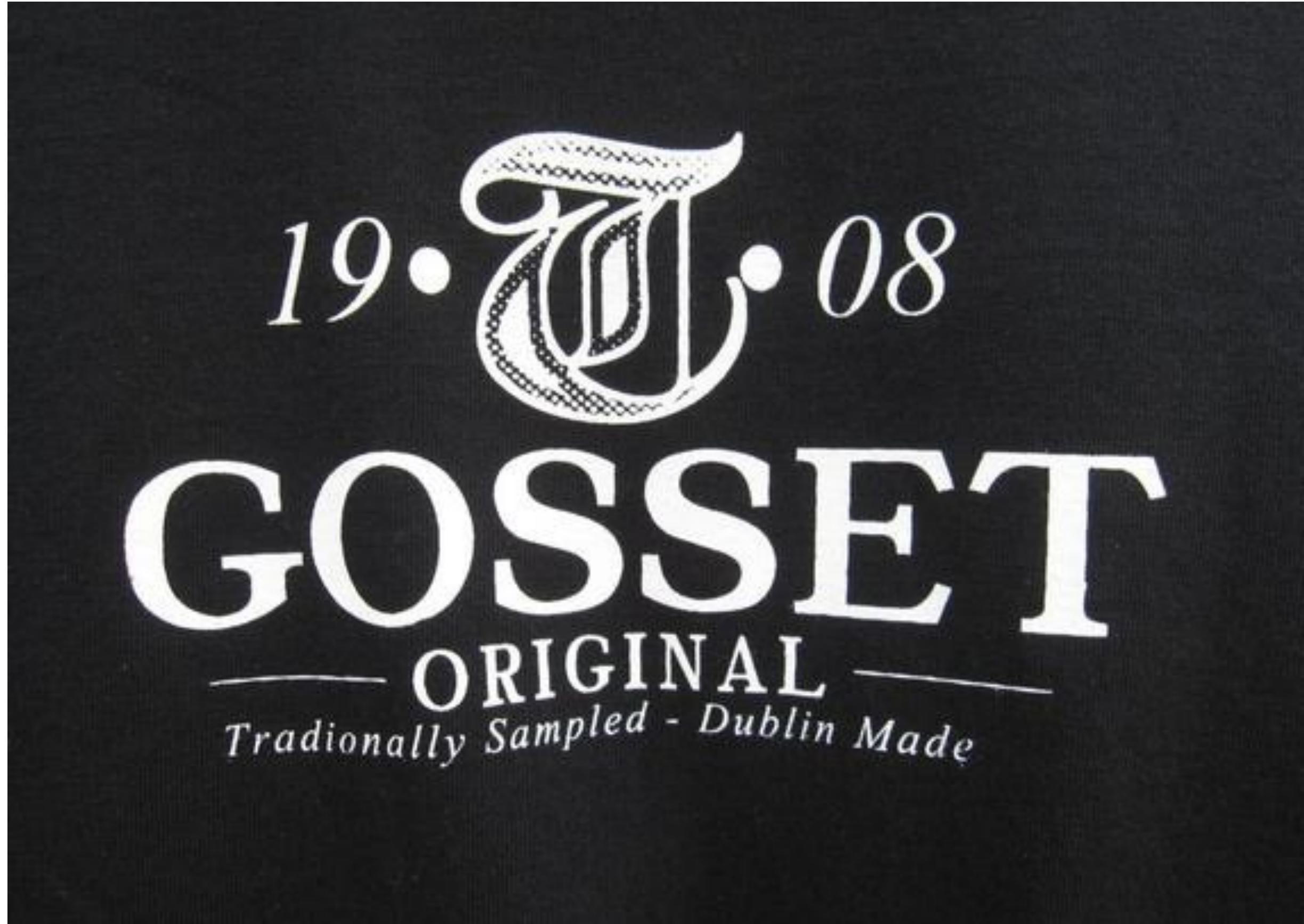
1. Mathematical model of what “nothing” looks like: the null model
2. Mathematical way of comparing data to “nothing”: a statistic
3. A decision rule for what is far enough away from “nothing” to be interesting



$$\frac{\bar{x} - \mu}{\text{s.e.}(\bar{x})}$$



“Signal”  
—  
S.e.( $\bar{x}$ )



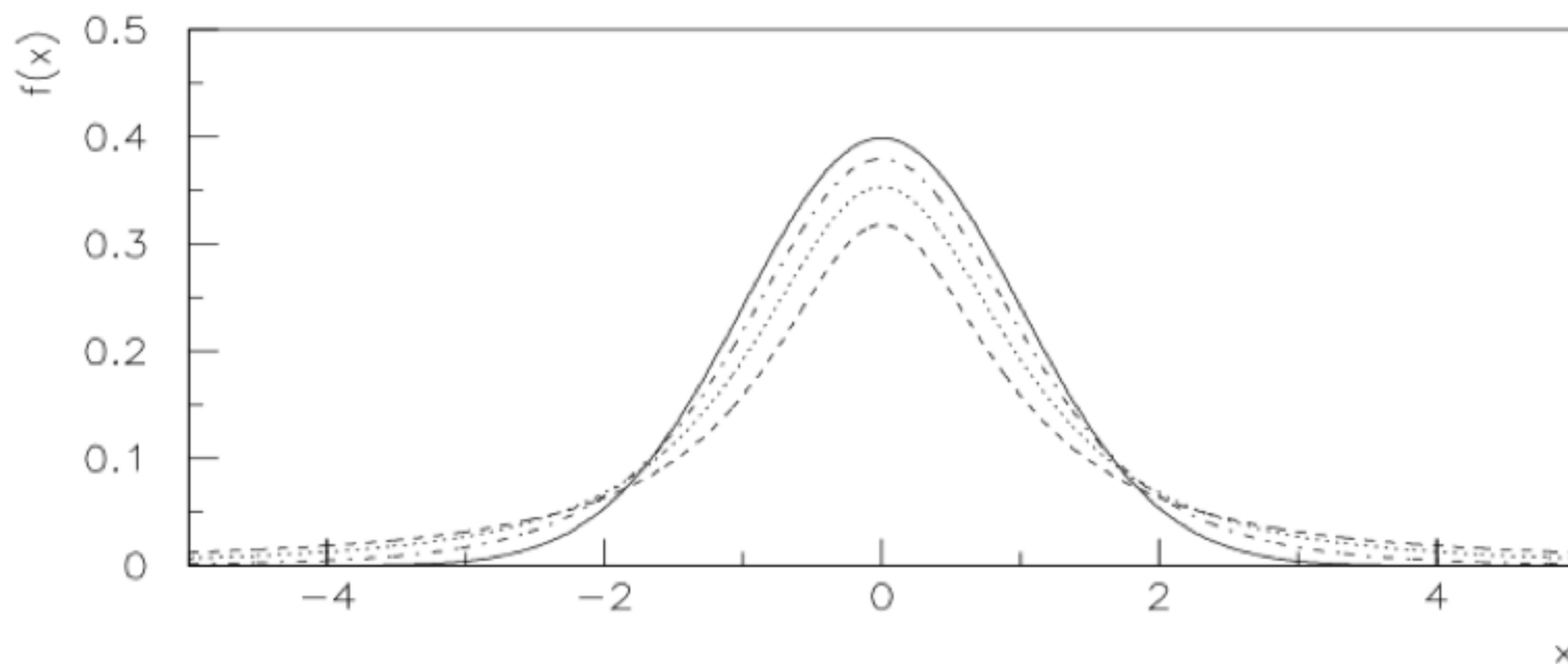
“Signal”

“Noise”

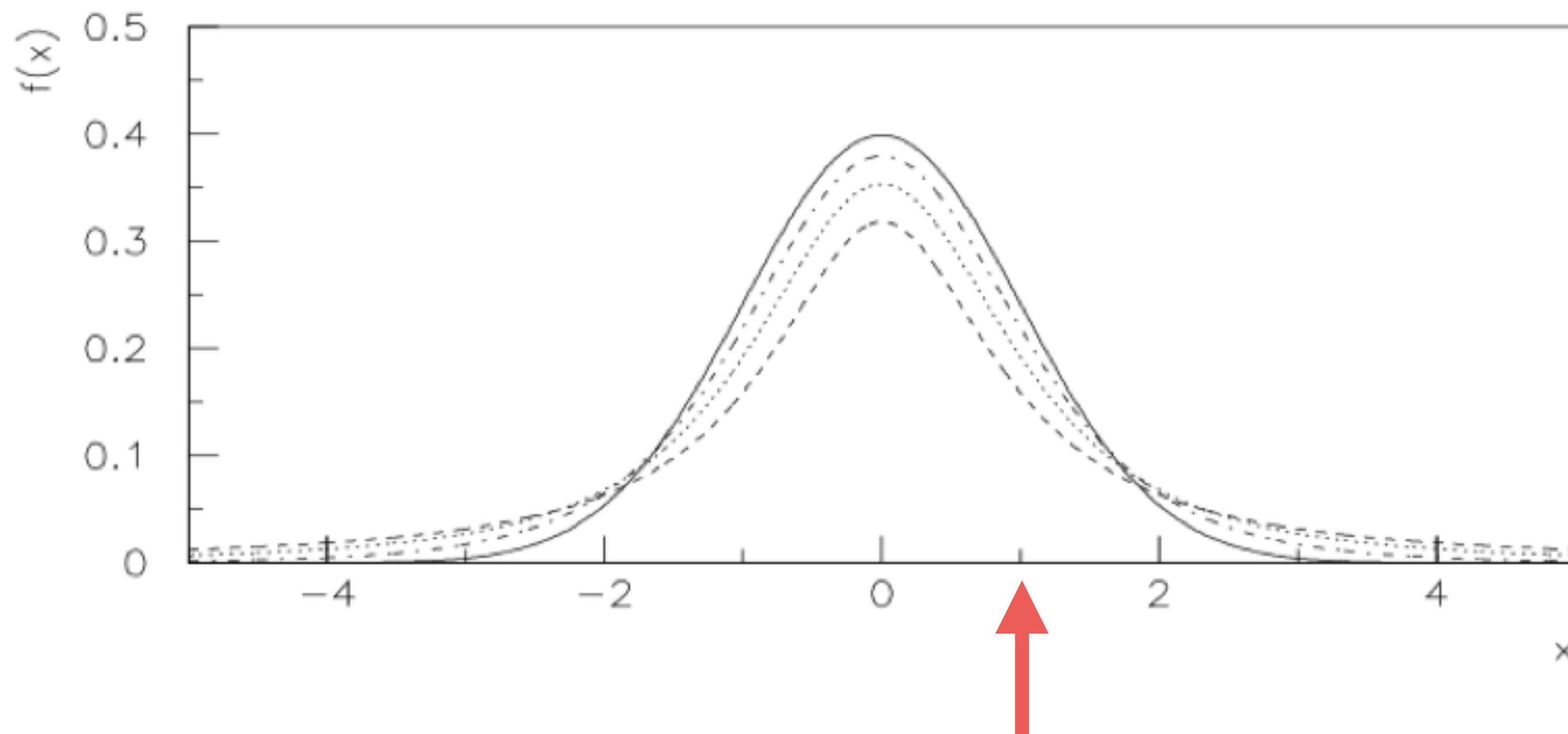
How many times  
larger is the signal  
than the noise?

**“Signal”**  
—  
**“Noise”**

# Schematic representation of “nothing”

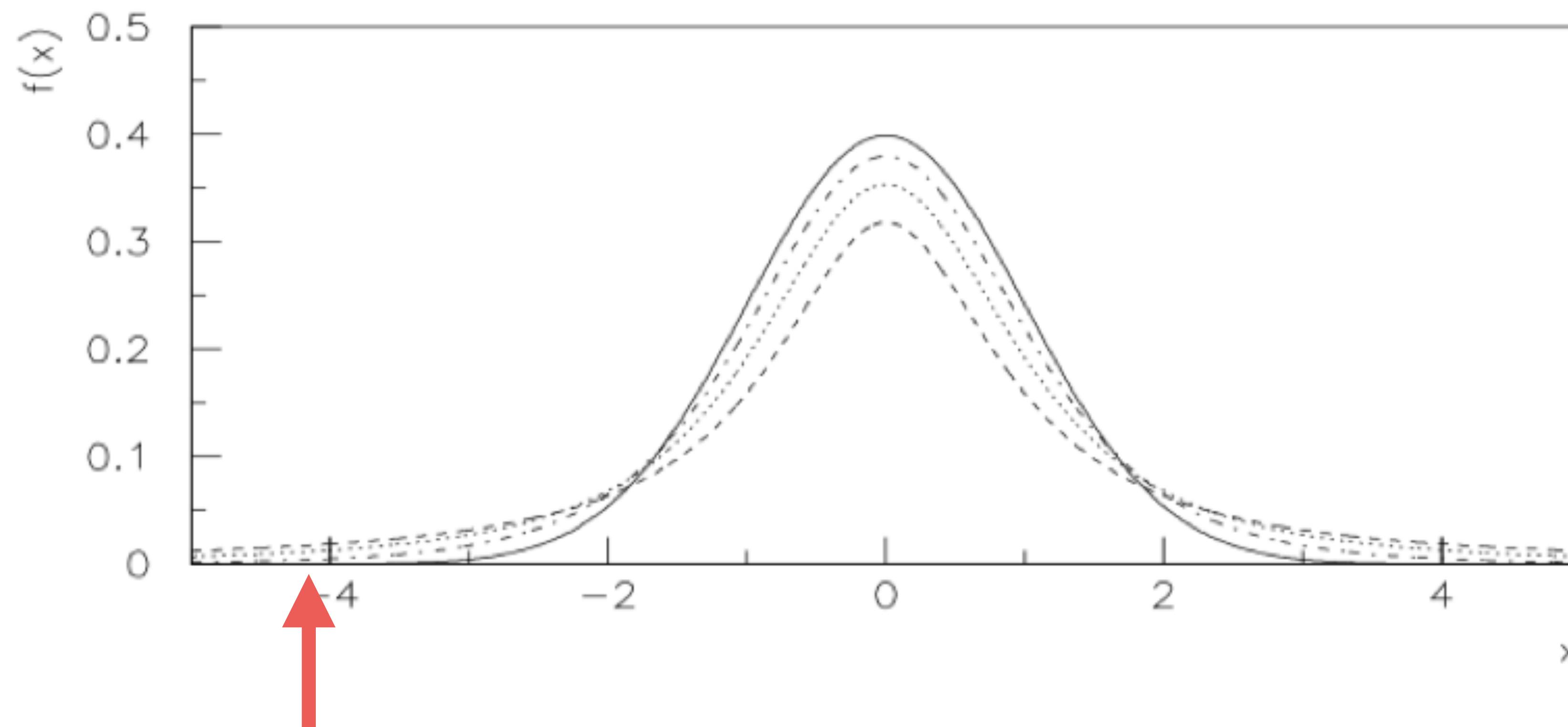


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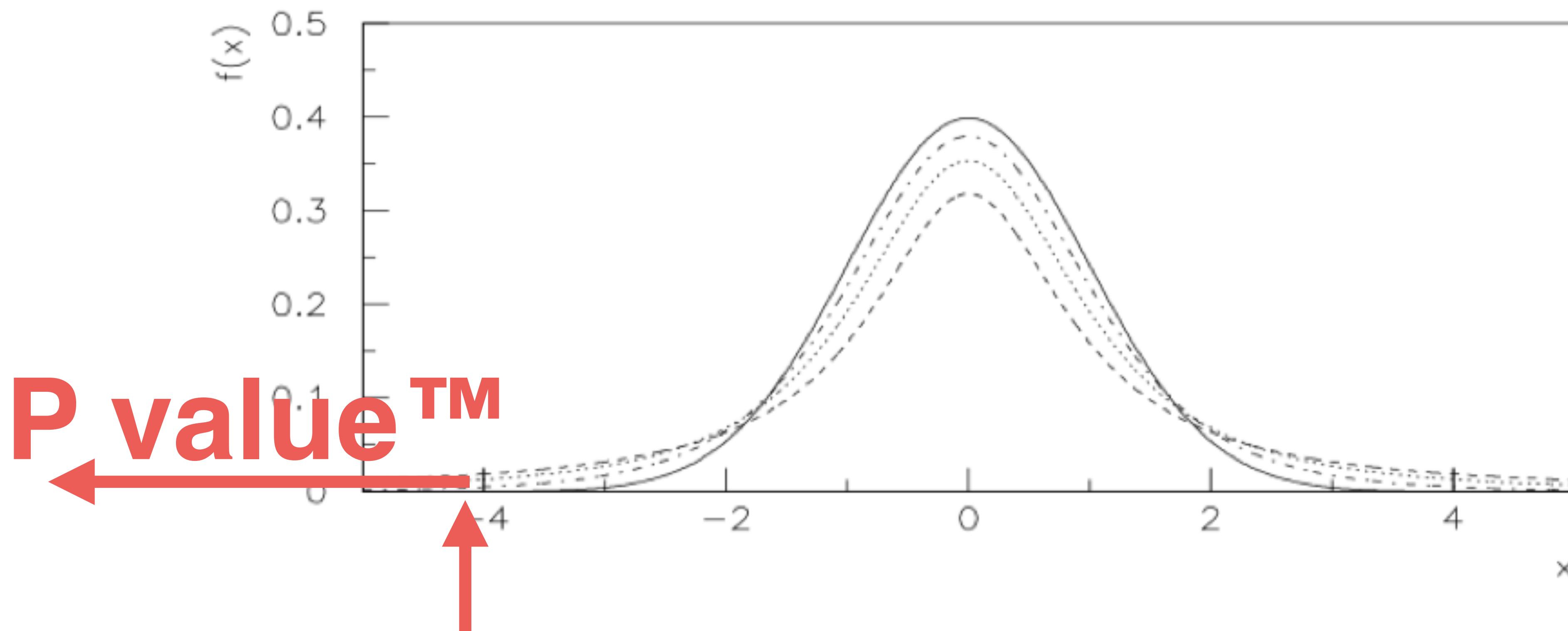
Observed s/n quite likely

# Schematic representation of “nothing”



Observed s/n not very likely

# Schematic representation of “nothing”



Observed s/n not very likely

# What is a good p-value?



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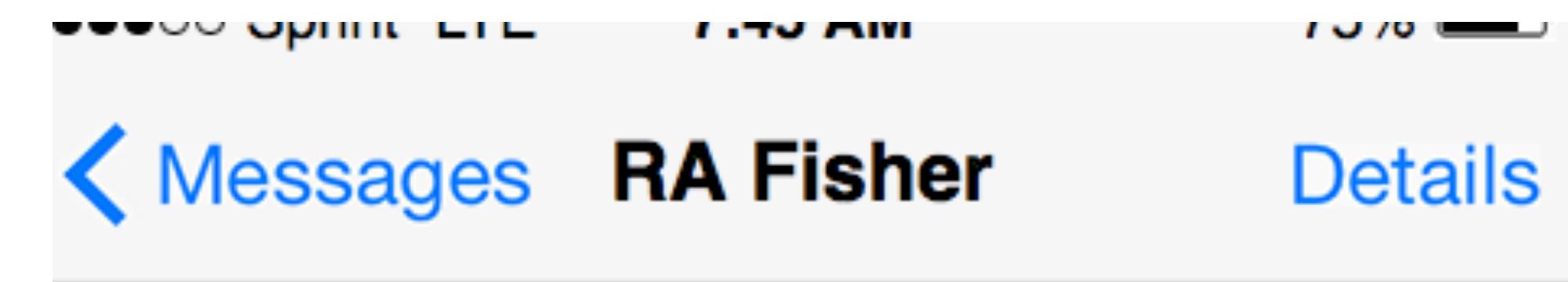
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– RA Fisher  
(Statistical Methods for Research Workers)



Use  $p < .05$ , it's the easiest

OK I will do this for  
always and ever

xoxo from the scientific  
community

# Back to the 100 observations

- If we observe fold change so that it is 2 times the noise, we “win”
- Implies that true fold change should at least be 4 times the noise if we want to almost surely win
- Noise shrinks with the inverse of square root of number of observations
- Details in **The Book**

# Back to the 100 observations

$$\frac{\mu}{\text{s.e.}(\bar{x})} > 4 \iff \mu > 4\text{s.e.}(\bar{x}),$$

$$\text{s.e.}(\bar{x}) = \frac{\sigma}{\sqrt{n}}$$

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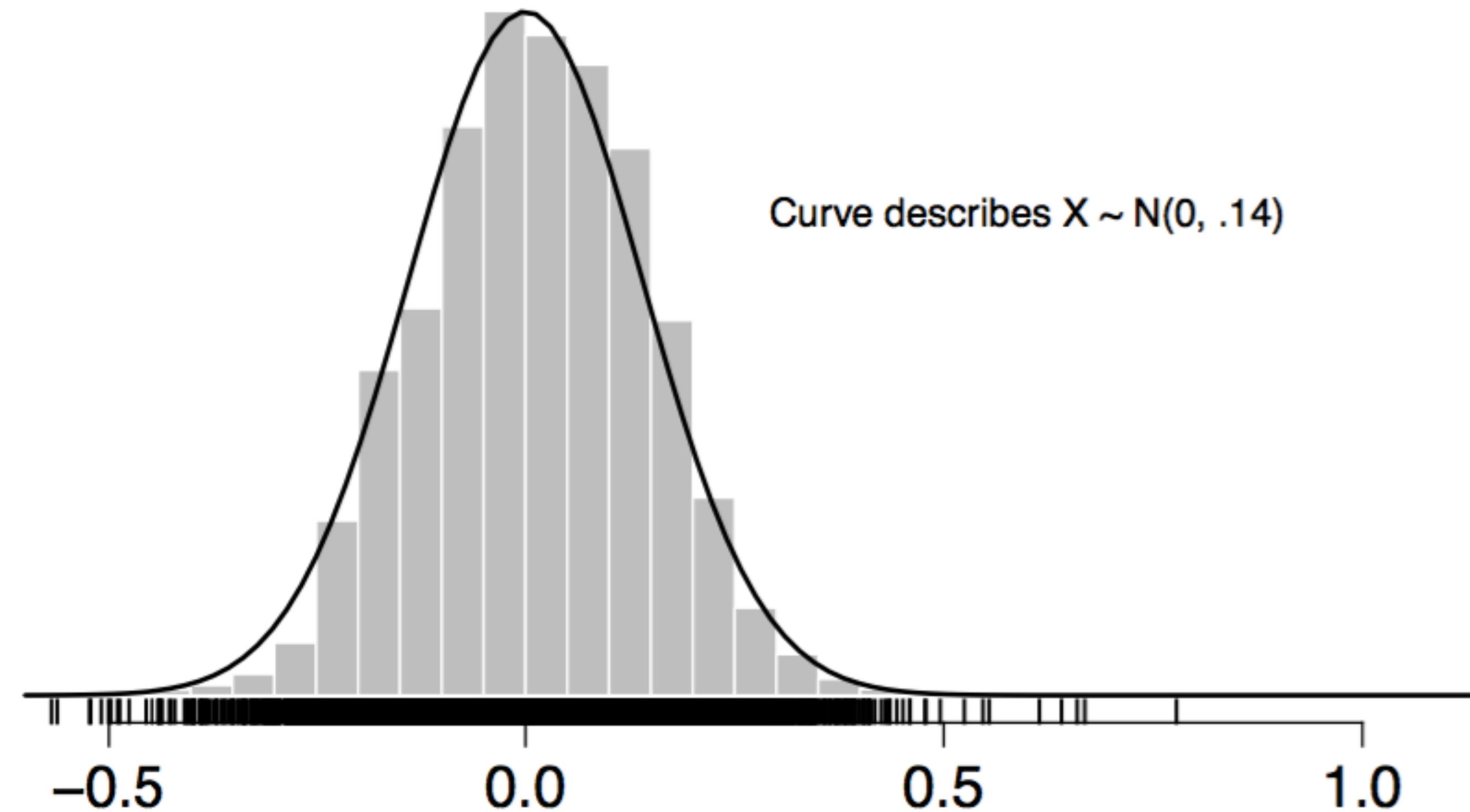
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# Smallest detectable means in number of standard deviations

Difference from zero:	$.4\sigma$
Difference between two groups of 50:	$.8\sigma$
Subgroups of 25, diff from zero:	$1.1\sigma$
Difference between subgroups:	$1.6\sigma$

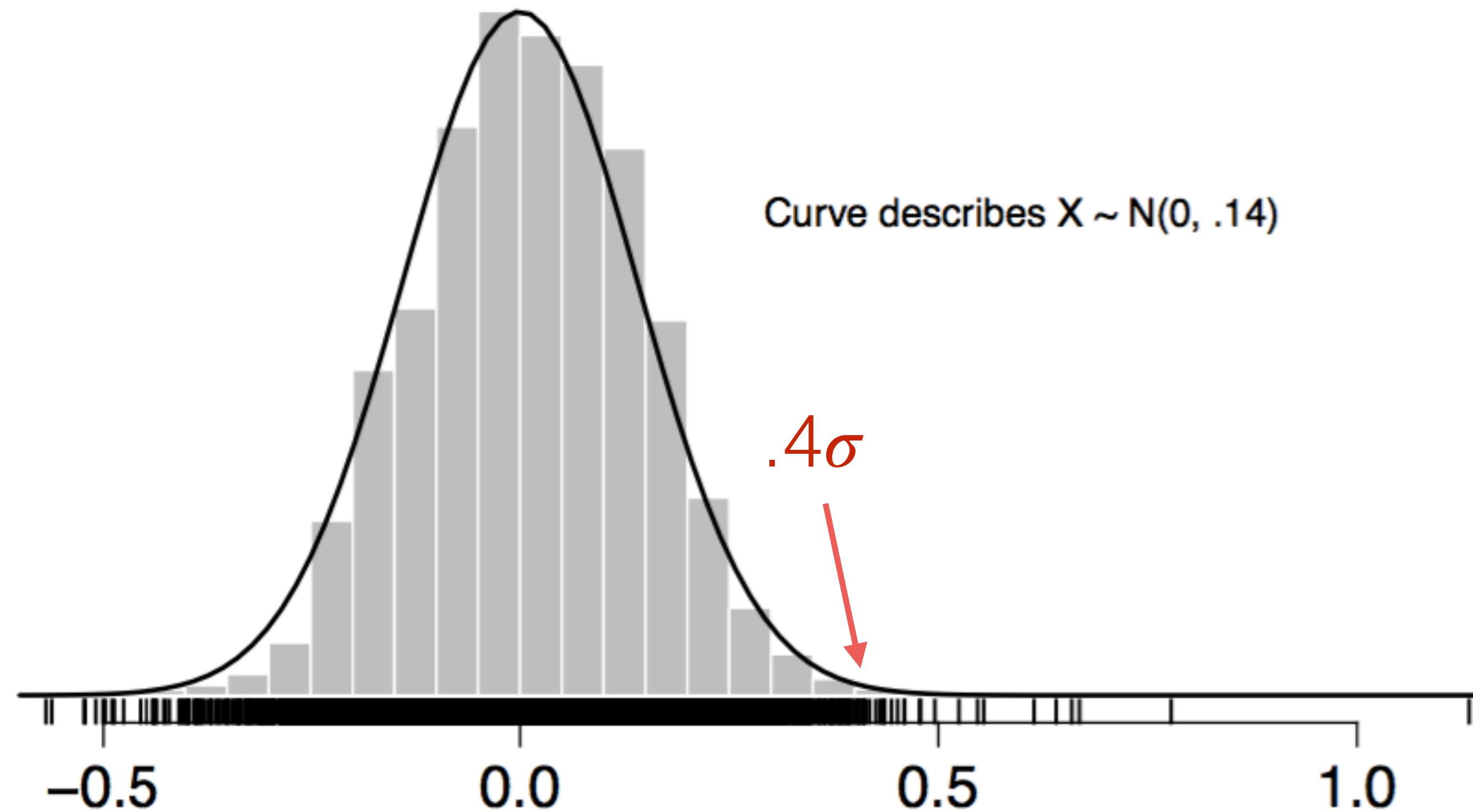
# Reality check: smokers vs nonsmokers

**Example of  $\mu$  -size in number of  $\sigma$ 's**



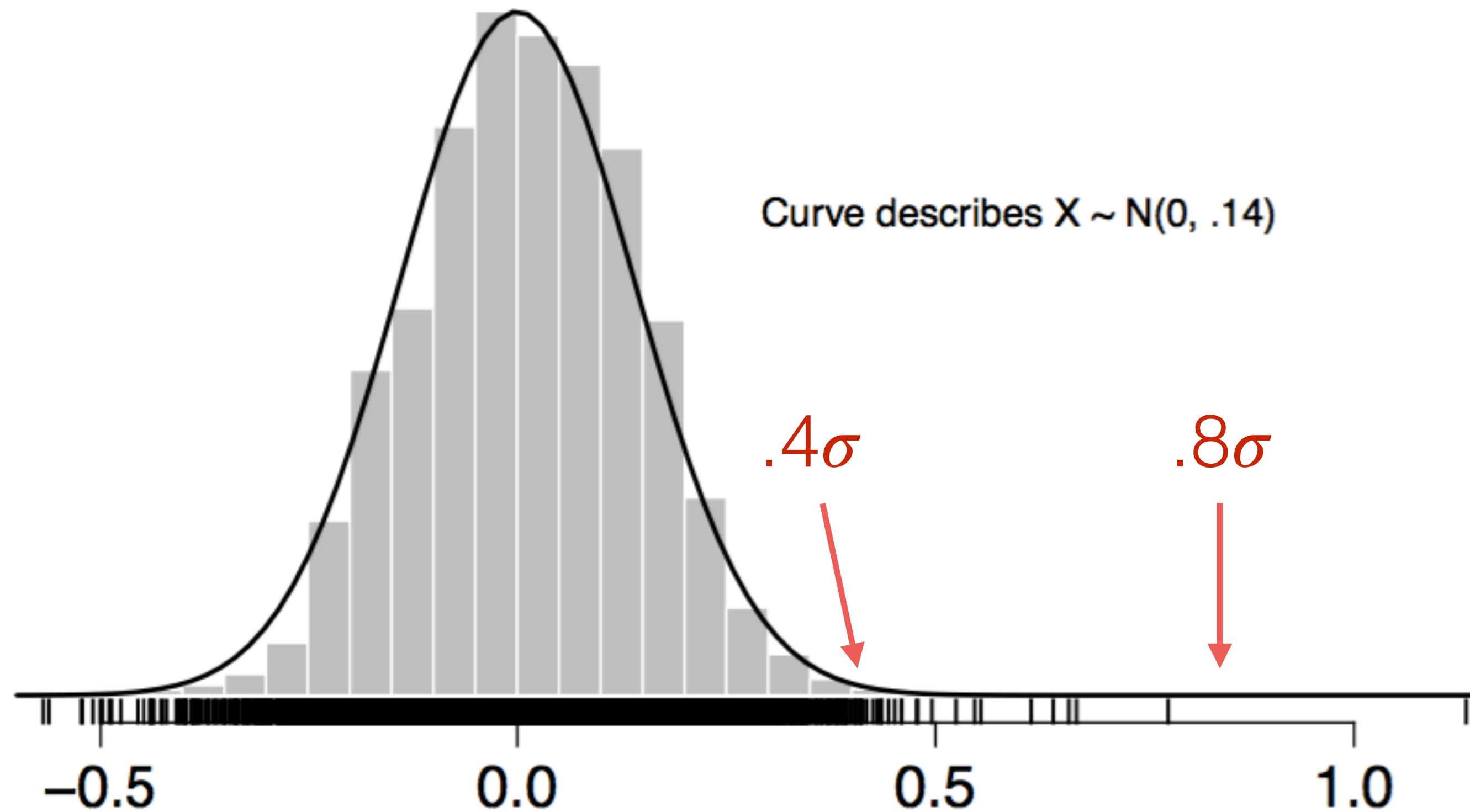
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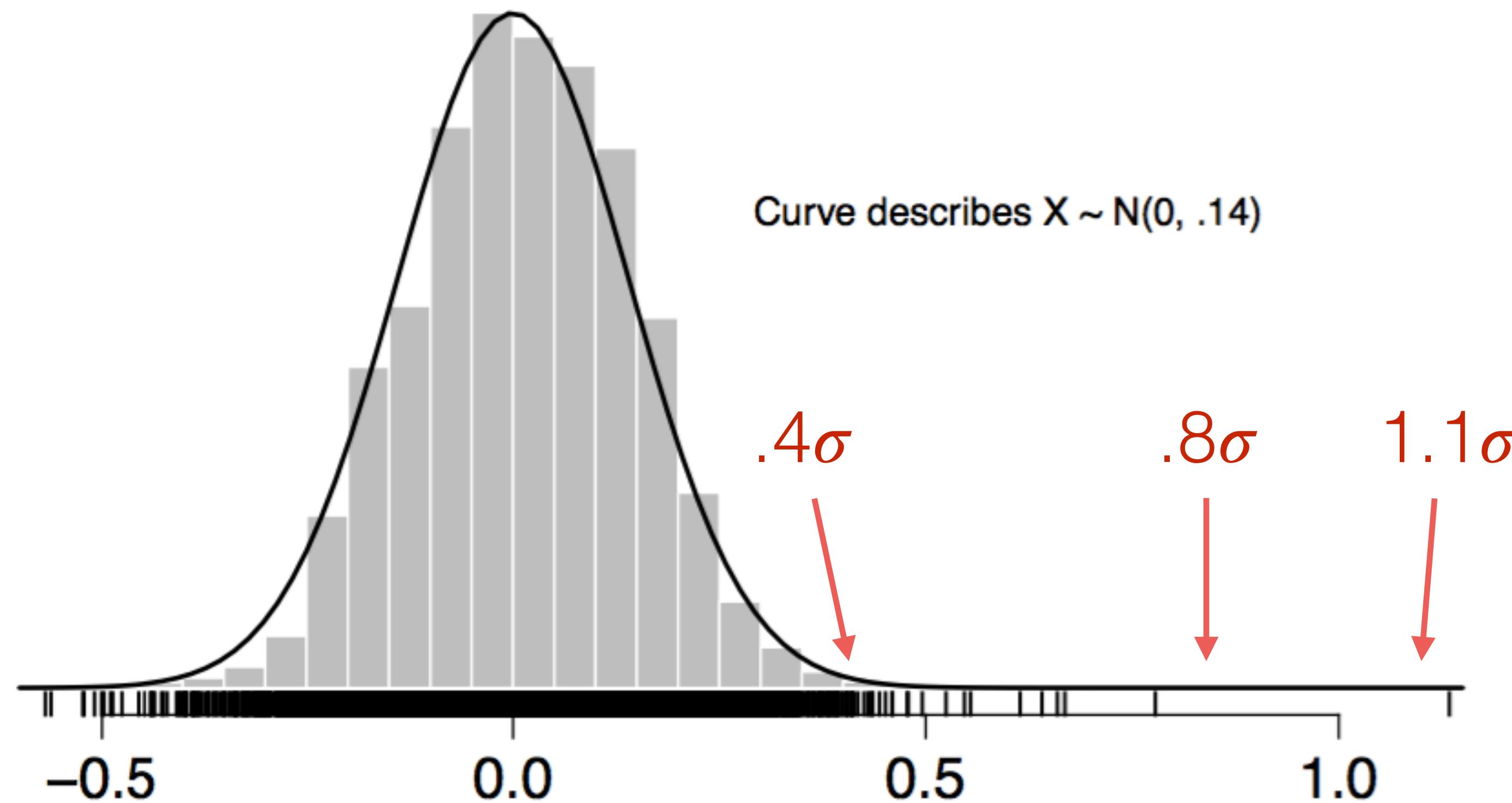
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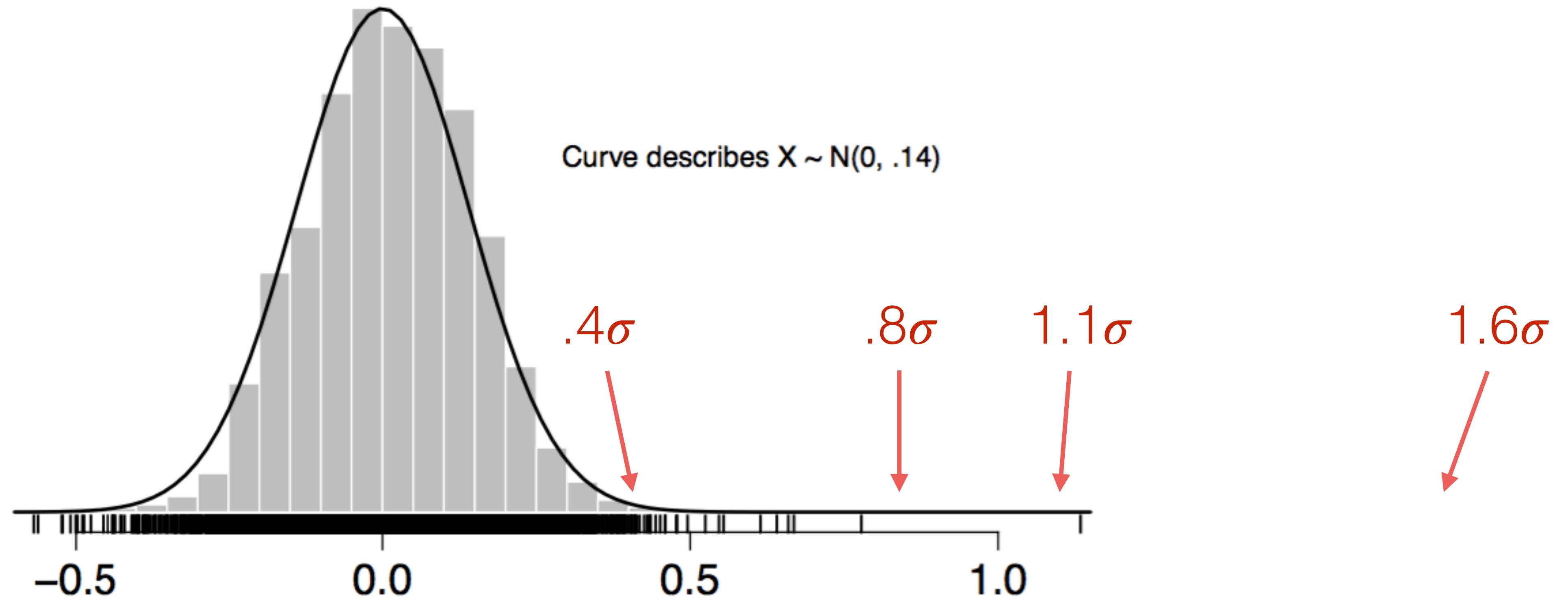
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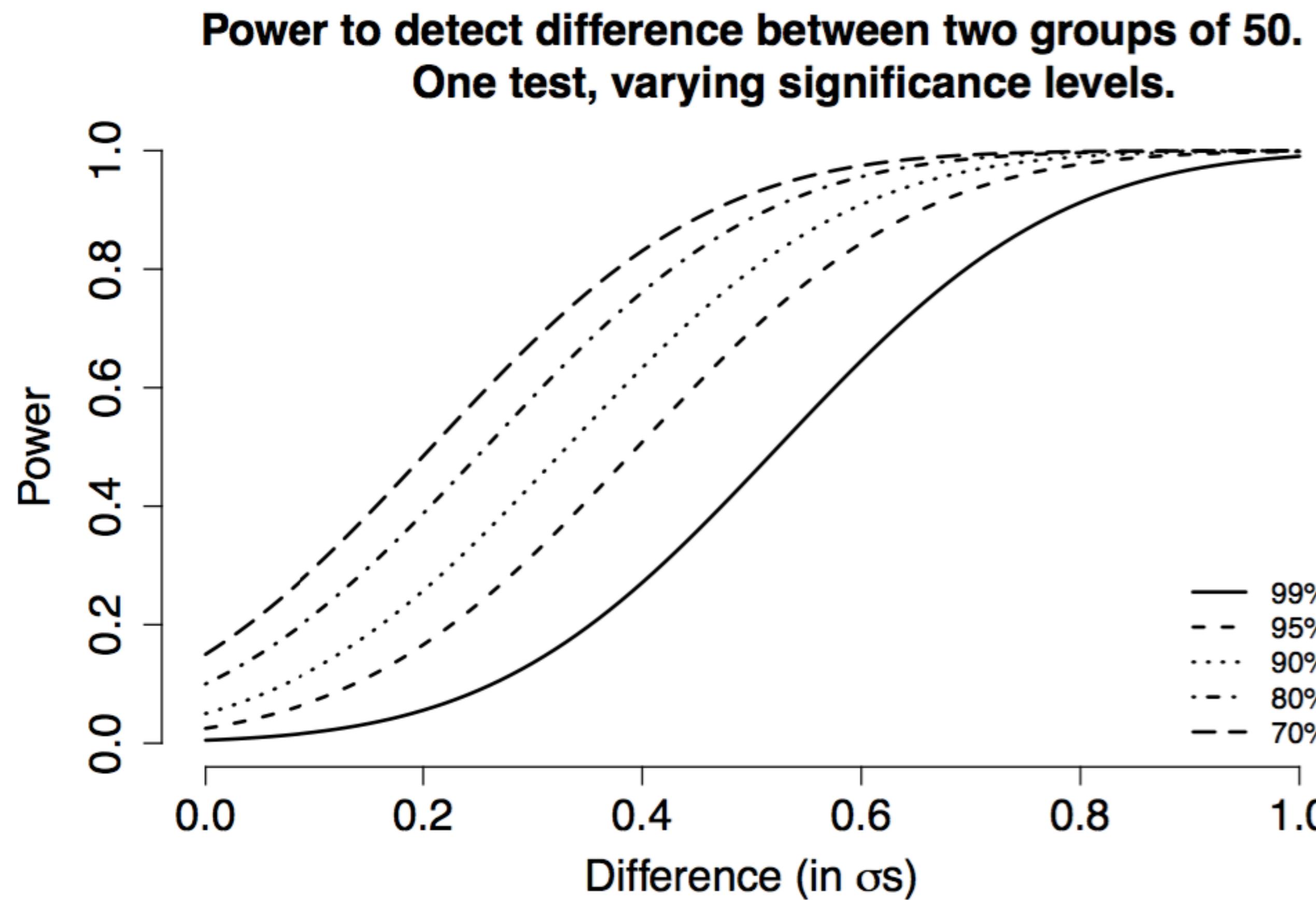


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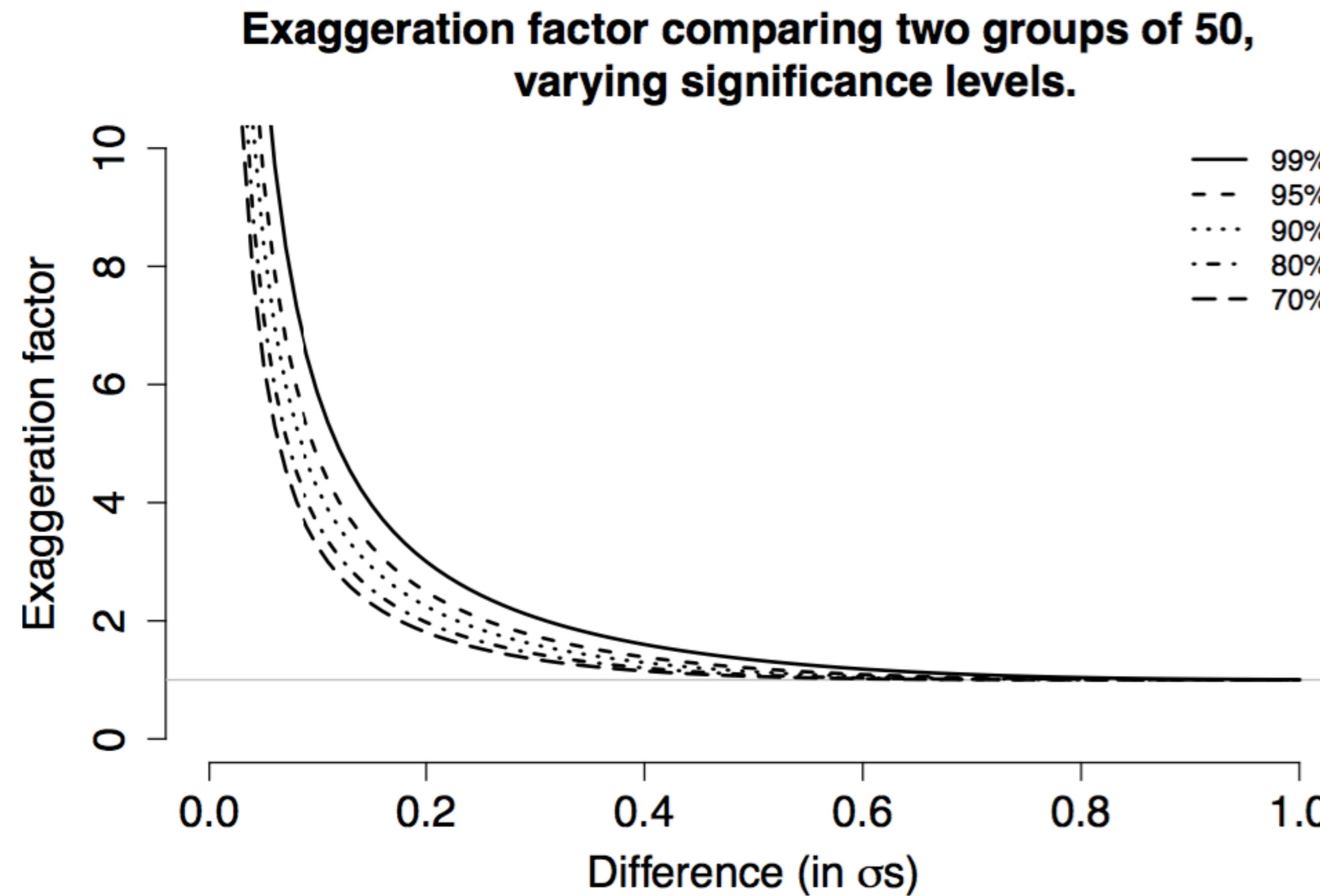
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# Statistical power is tragic in data such as these



The “wins” are overestimated by like 200%



# What to expect

<b>Confidence</b>	99%	95%	90%	80%	70%
<b>Magnitude error</b>	2.3	2.0	1.8	1.6	1.5
<b>Power</b>	12%	28%	39%	53%	63%

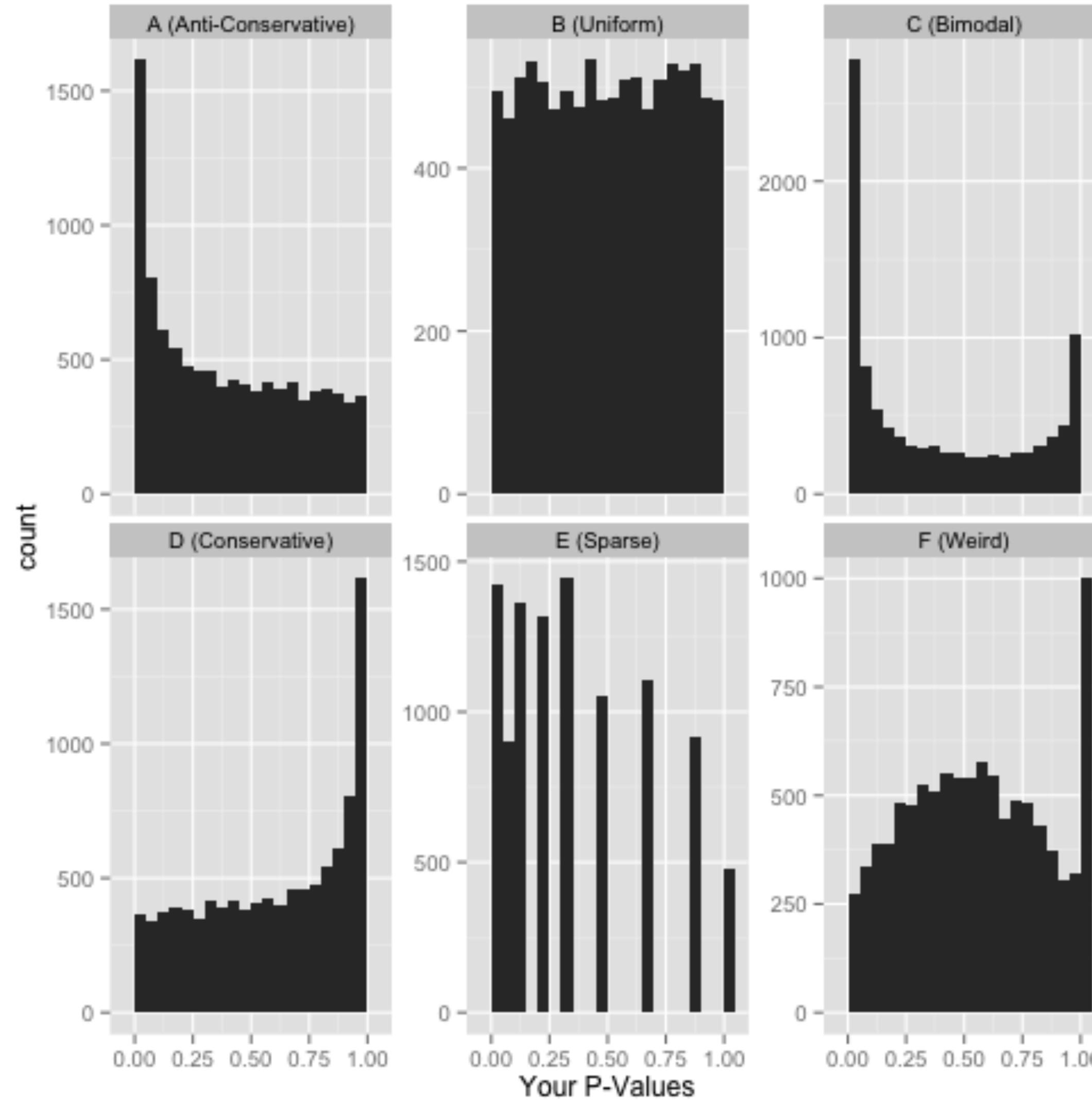
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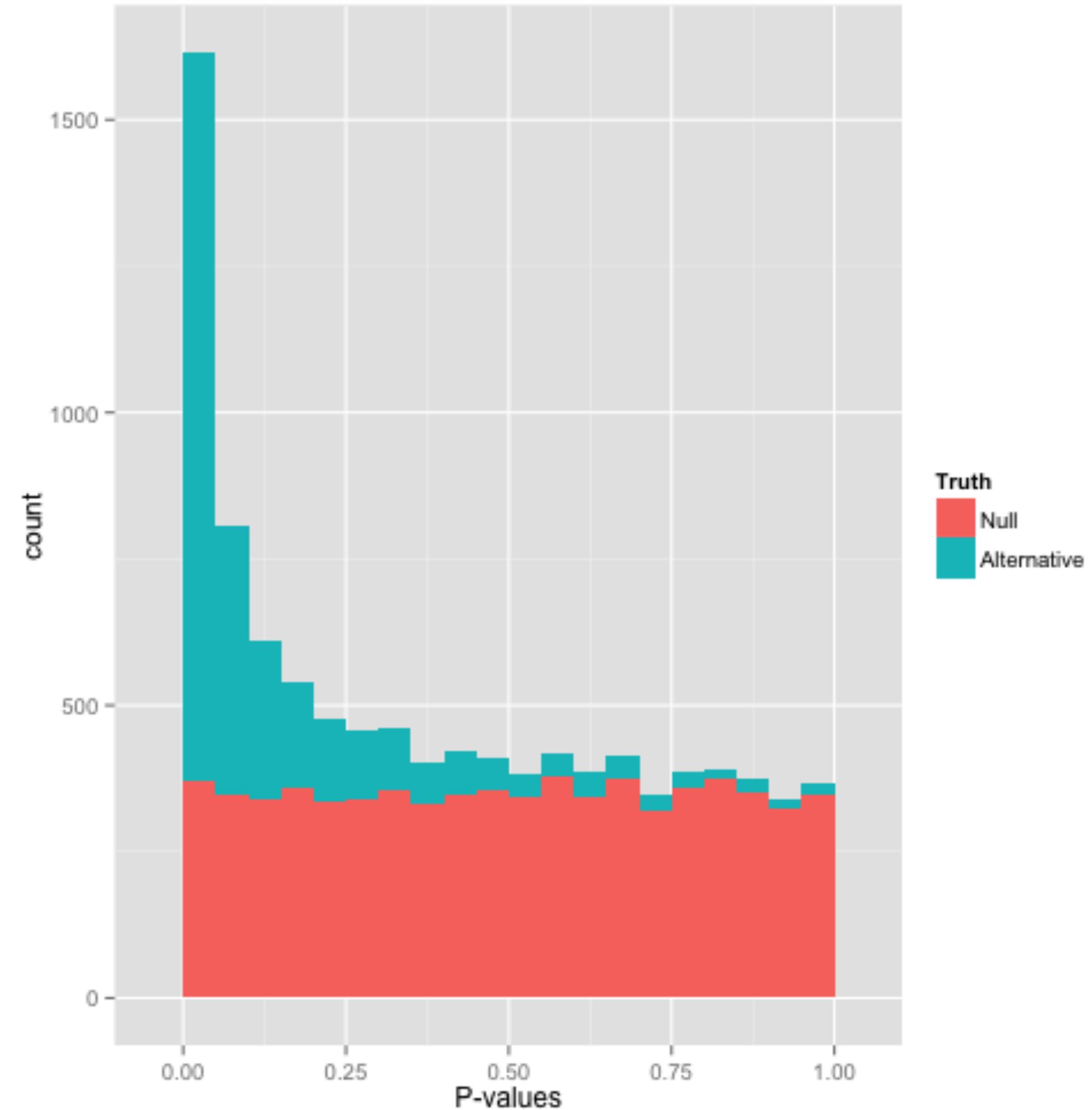
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Food for thought: are you sure hypothesis testing is for you?



# Humble scholars of uncertainty





David Robinson

# BE A BIOLOGIST!

- Ask a biology question
- Let the data answer
- Use a statistical tool to quantify the uncertainty

# BE A BIOLOGIST - USE FAMILIAR STATISTICS TOOL

- Choose an  $\alpha$ .
- Calculate the mean difference - which ones are interesting?  
Sort accordingly.
- Confidence intervals for those.
- Pick those with a low  $p$ -value,  
not necessarily lower than  $\alpha$ .

