



Centre for Molecular Medicine  
and Therapeutics



## Multiple Testing

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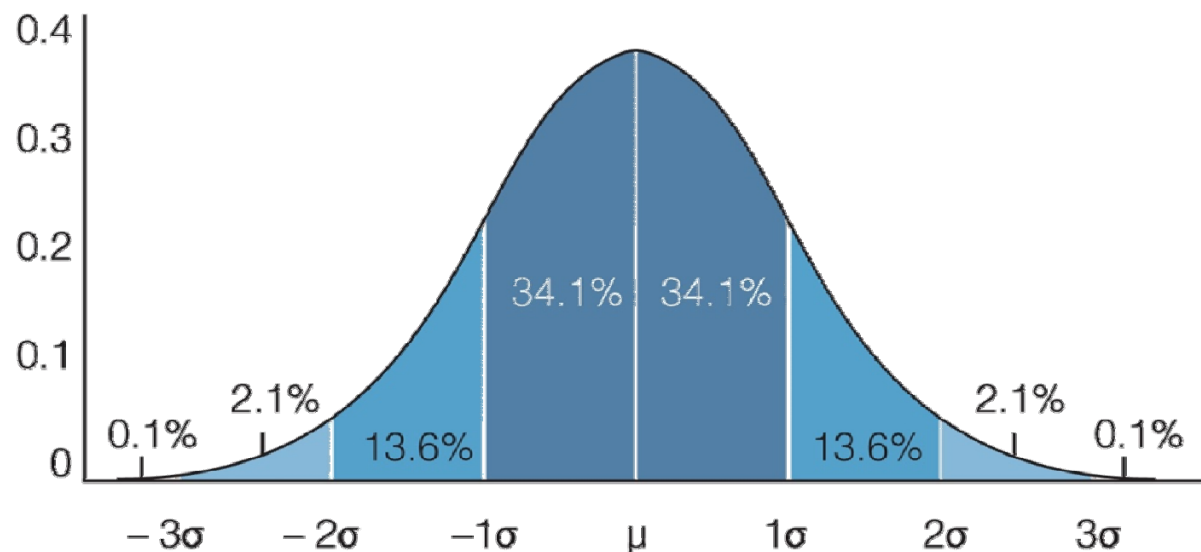
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# Outline

- Single Hypothesis Testing
- Multiple Hypothesis Testing
- Bonferroni Correction
- Step-up Procedure
- False Discovery Rate Correction
- Max-t Permutation Test
- Recent Topics
- Neuroimaging Applications

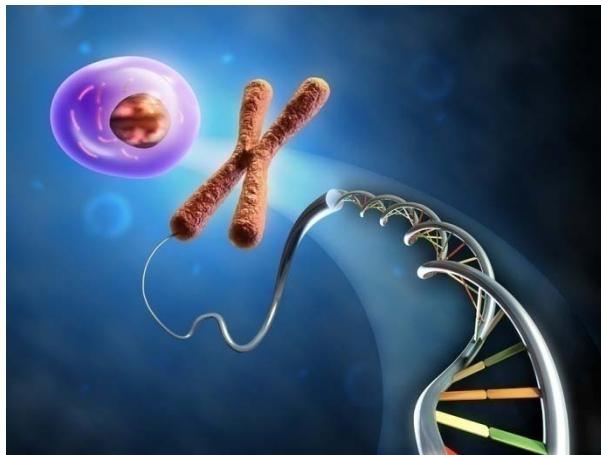
# Single Hypothesis Testing

- In the past, a handful of hypotheses with a lot of samples, e.g. census data.
- $H_0: X = \mu$  vs.  $H_A: X \neq \mu$ 
  - Are girls smarter than guys? => two sample t-test
  - Do last minute studying affect scores? => regression
- Generate statistics e.g.  $z$ ,  $t$ ,  $F$ , ...
- $p < 0.05$

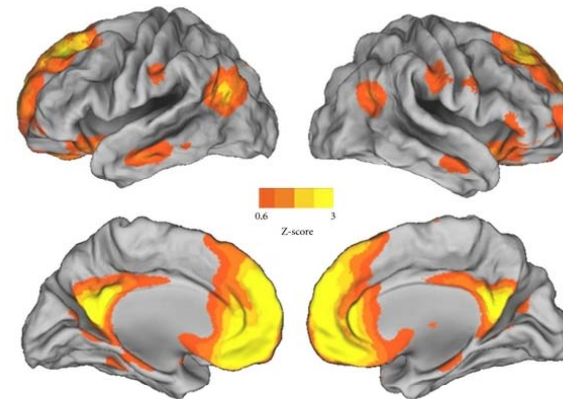


# Multiple Hypothesis Testing

- Nowadays, a lot of (unplanned) hypotheses but not enough samples (for medical research)



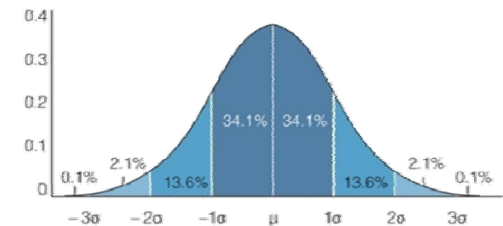
$10^3$  **genomics**  $10^6$



$10^2$  **Neuroimaging**  $10^5$

# Multiple Testing Problem

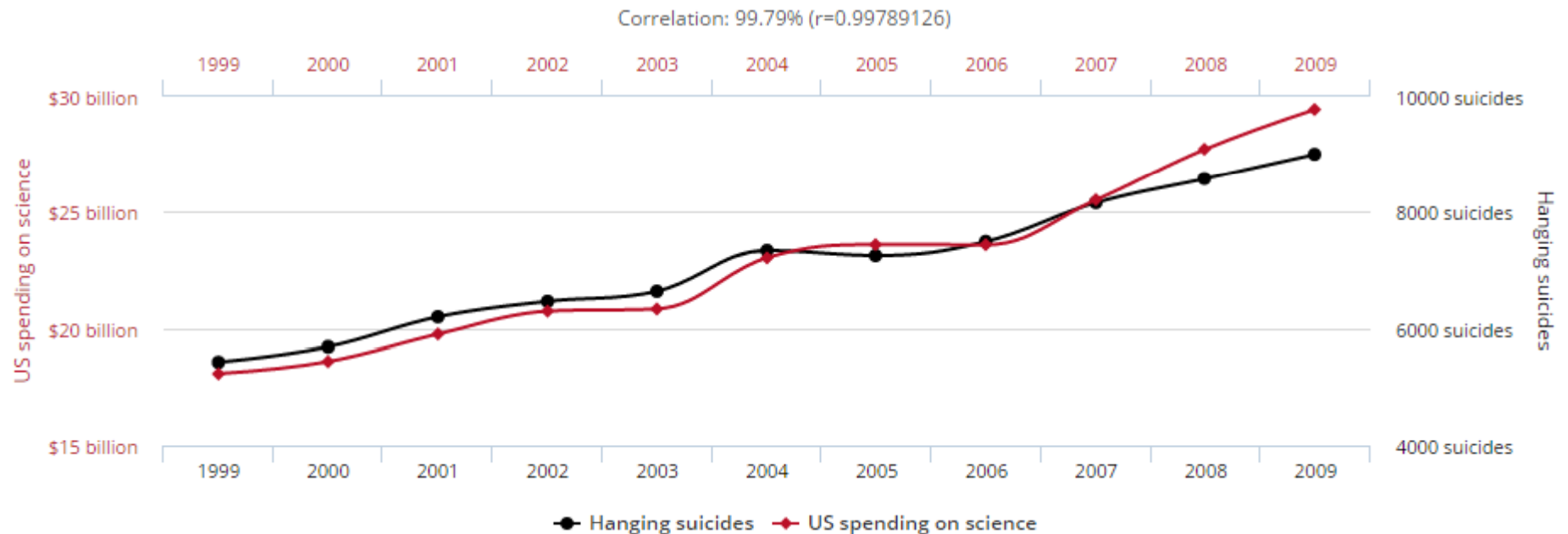
- $P(\text{not rejecting 1 hypothesis}) = 1 - \alpha$
- $P(\text{not rejecting all } n \text{ hypotheses}) = (1 - \alpha)^n$
- $\alpha_{\text{FWER}} = 1 - (1 - \alpha)^n$ 
  - $\alpha = 0.05, n = 10: \alpha_{\text{FWER}} = 0.4013$
  - $\alpha = 0.05, n = 10^2: \alpha_{\text{FWER}} \approx 1$
- So if e.g. run 100 experiments, then  $\alpha_{\text{FWER}} \cdot 100$  of them would have  $\geq 1$  hypothesis falsely rejected.
- Intuition from ML perspective is that the more we sample the variable space, the more “likely” we will get some “extreme” samples.



# MATLAB Demo

# Why Important?

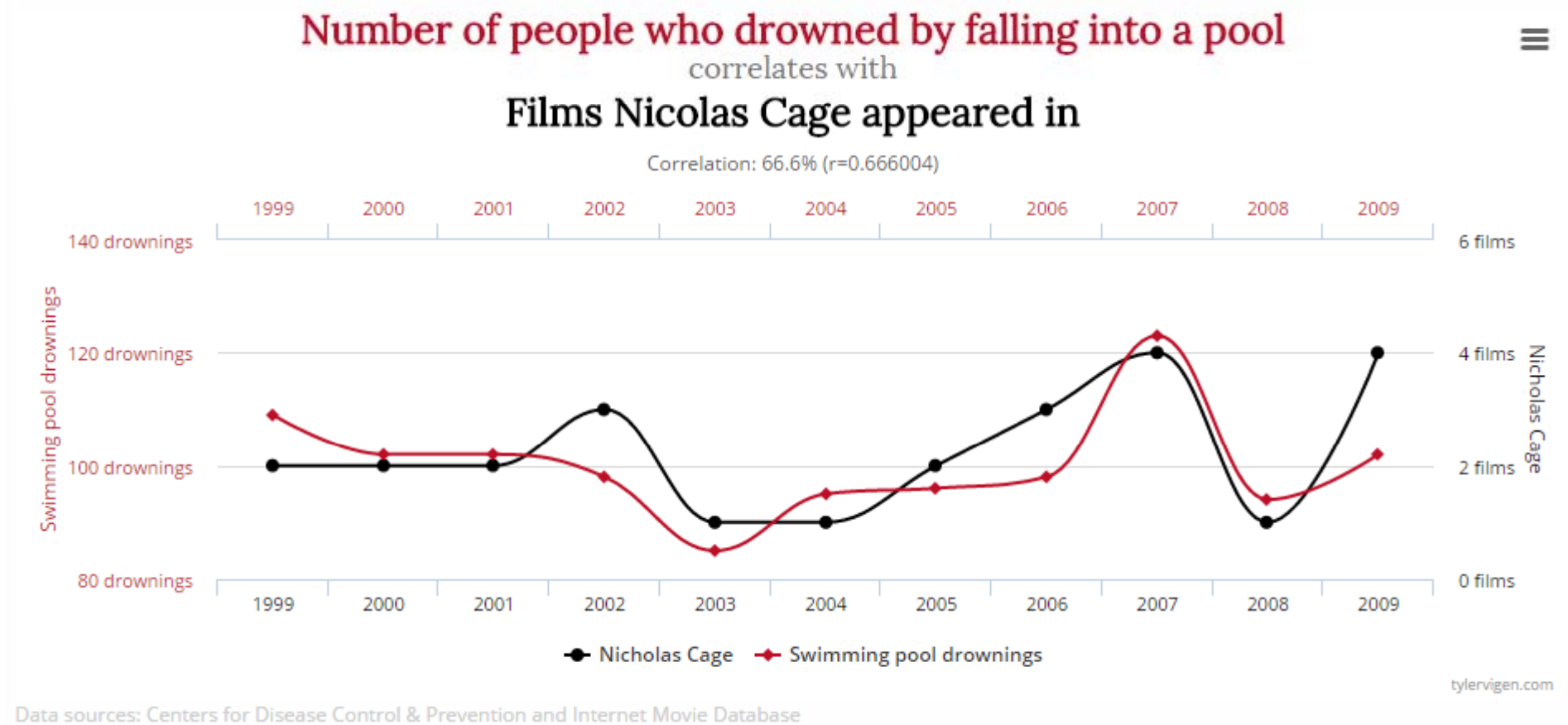
## US spending on science, space, and technology correlates with Suicides by hanging, strangulation and suffocation



tylervigen.com

Data sources: U.S. Office of Management and Budget and Centers for Disease Control & Prevention

# Why Important?

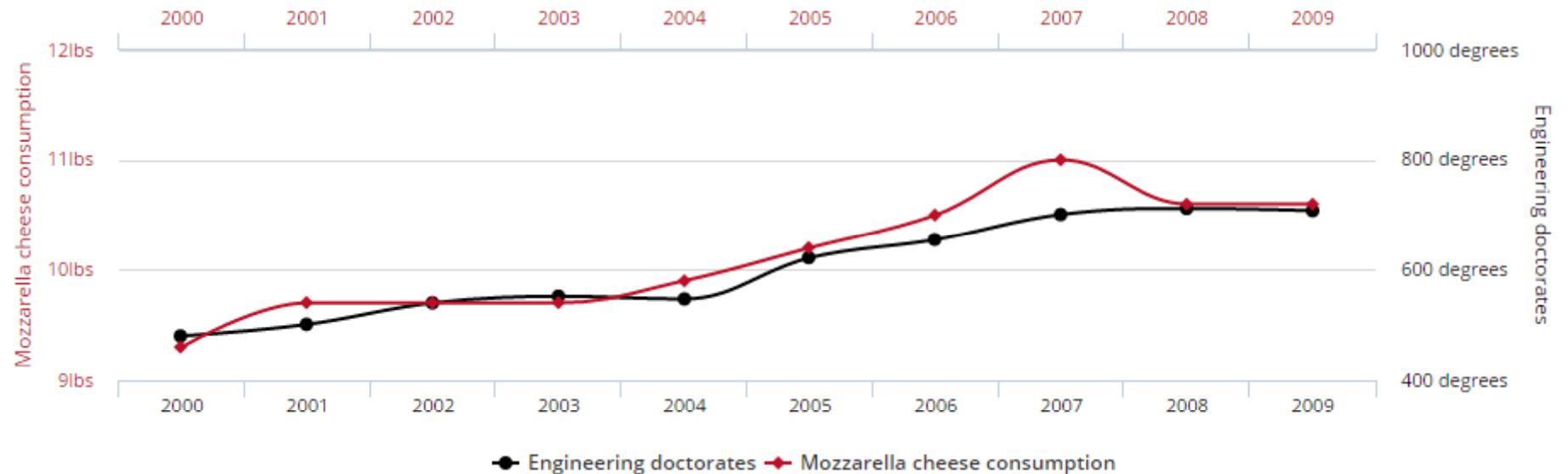




# Why Important?

Per capita consumption of mozzarella cheese  
correlates with  
Civil engineering doctorates awarded

Correlation: 95.86% ( $r=0.958648$ )



Data sources: U.S. Department of Agriculture and National Science Foundation

tylervigen.com

# Why Important?

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## Spurious correlations



Now a ridiculous book!

- Spurious charts
- Fascinating factoids
- Commentary in the footnotes

Amazon | Barnes & Noble | Indie Bound

<http://tylervigen.com/spurious-correlations>

# Why Important?

Problems with scientific research


## How science goes wrong

Scientific research has changed the world. Now it needs to change itself

Last year researchers at one biotech firm, Amgen, found they could reproduce just **six of 53** “landmark” studies in cancer research. Earlier, a group at Bayer, a drug company, managed to repeat just a **quarter of 67** similarly important papers. A leading computer scientist frets that **three-quarters** of papers in his subfield are bunk. In 2000-10 roughly **80,000 patients** took part in clinical trials based on research that was **later retracted** because of mistakes or improprieties.

# Notations and Terminologies

		<i>Predicted</i>		
		<b>True</b>	<b>False</b>	
<i>Ground Truth</i>	<b>True</b>	<div>U</div> <div>True Negative</div>	<div>V</div> <div>False Positive</div>	$n_0$
	<b>False</b>	<div>T</div> <div>False Negative</div>	<div>S</div> <div>True Positive</div>	$n - n_0$
		$n - R$	$R$	$n$

 unobserved

$\text{Sensitivity} = S / (n - n_0)$   
 $\text{Specificity} = U / n_0$

Note: All terms are defined wrt null, e.g. if the ground truth is that the null hypothesis is True, but we claimed it is false, then we have a false positive.

# Bonferroni Correction

## Procedures

- Recall  $\alpha_{\text{FWER}} = 1 - (1 - \alpha)^n$
- Set  $\alpha = 1 - (1 - \alpha_{\text{FWER}})^{1/n} \approx 1 - (1 - \alpha_{\text{FWER}}/n) = \alpha_{\text{FWER}}/n$

## Examples

- $\alpha_{\text{FWER}} = 0.05$  and  $n = 10$ , needs  $\alpha = 0.05/10 = 0.005$
- $\alpha_{\text{FWER}} = 0.05$  and  $n = 10^6$ , needs  $\alpha = 0.05/10^6 = 5 \times 10^{-8}$

## Properties

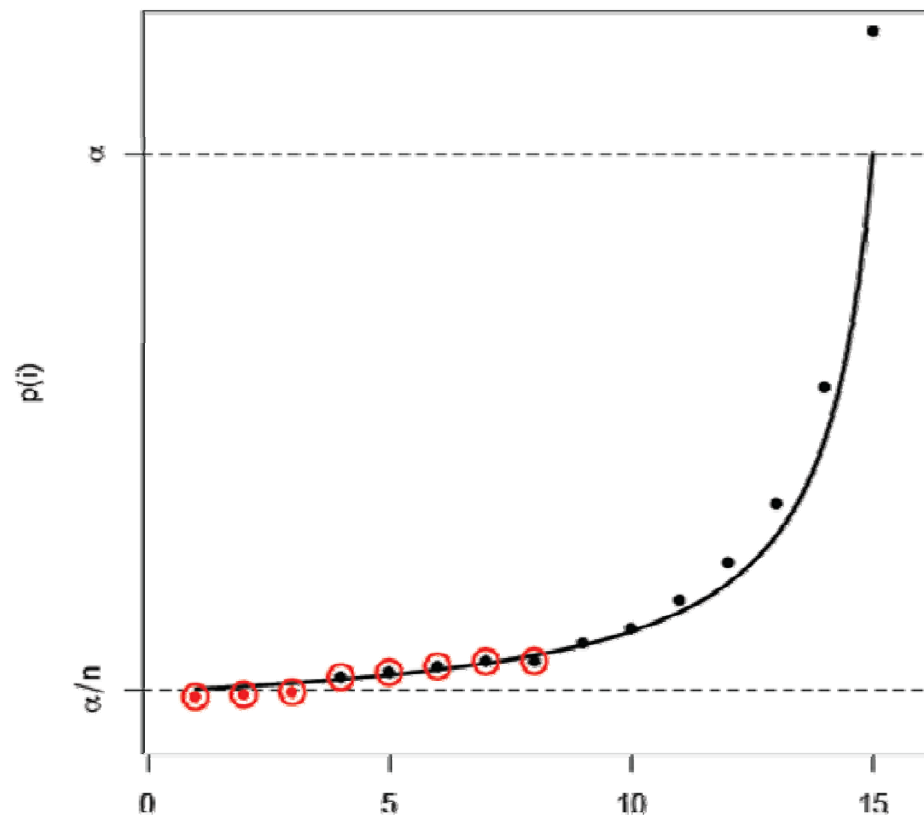
- Controls FWER =  $P(V \geq 1)$  in *strong* sense.
- Can handle correlated hypotheses.
- Very stringent

		Predicted		
		True	False	
Ground Truth	True	U	V	$n_0$
	False	T	S	$n - n_0$
		$n - R$	R	$n$

# MATLAB Demo

# Step-up Procedure

- aka Hochberg's procedure
- Sort  $p$  in descending order
- $p(i) \leq \alpha/(n-i+1)$ 
  - i.  $p(n) \leq \alpha/(n-n+1)$
  - ii.  $p(n-1) \leq \alpha/(n-(n-1)+1)$
  - iii. ...
- Controls FWER in strong sense
- Holm's Step-down procedure uses same threshold but less sensitive.



# MATLAB Demo



# False Discovery Rate

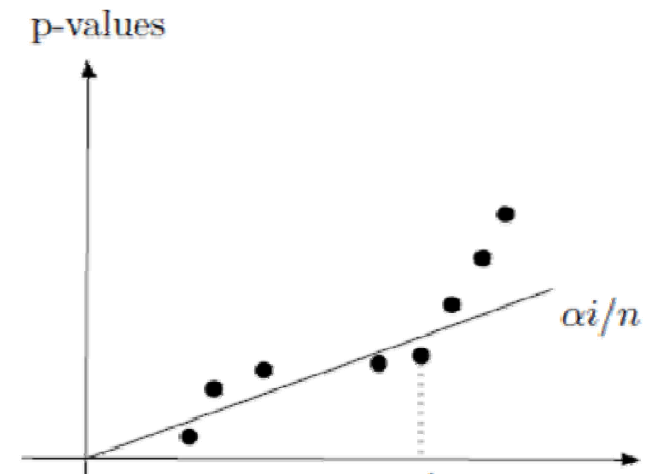
## Idea

- Benjamini & Hochberg, 1995
- Recall  $\text{FWER} = P(V \geq 1)$
- $\text{Fdp} = V/\max(R, 1)$
- But  $V$  unobserved, so:  $\text{FDR} = E(\text{Fdp})$

## Procedures

- Sort  $p$  in ascending order.
- Find  $i_0 = \max i$  s.t.  $p(i) \leq i \cdot q/n$

		Predicted		
		True	False	
Ground Truth	True	U	V	$n_0$
	False	T	S	$n - n_0$
		$n - R$	R	n



# False Discovery Rate

## Properties

- If hypotheses are independent, then  $FDR < q$  for *all* configurations of hypotheses.
- If data are Gaussian and hypotheses are positively correlated, i.e.  $\sum_{ij} \geq 0$ , then  $FDR < q$ .
- If hypotheses are correlated,  
 $FDR < q \cdot (\log(n) + 0.577)$   
 $\Rightarrow p(i) < i \cdot q/n / (\log(n) + 0.577)$   
BUT  $i = 1$ ,  $p(i) < q/n / (\log(n) + 0.577) < q/n$

		Predicted		
		True	False	
Ground Truth	True	U	V	$n_0$
	False	T	S	$n - n_0$
		$n - R$	R	n

# False Discovery Rate

## Properties

- $n=n_0$ , then  $FDR = FWER$  since:  
     $V=R$ , so  $V=0$  iff  $Fdp = 0$  and  $V \geq 1$  iff  $Fdp=1$ ,  
    i.e.  $Fdp$  = indicator variable,  
    thus  $P(V \geq 1) = E(Fdp) = FDR \Rightarrow$  ctrls FWER *weakly*
- $n < n_0$ , controlling FWER controls FDR
- Adaptive: 5/100, 50/1000, ...
- More sensitive than Hochberg  
     $i/n / (1/(n-i+1)) = i \cdot (1-(i-1)/n)$   
    e.g.  $i = n/2 \Rightarrow \sim n/4$  gain

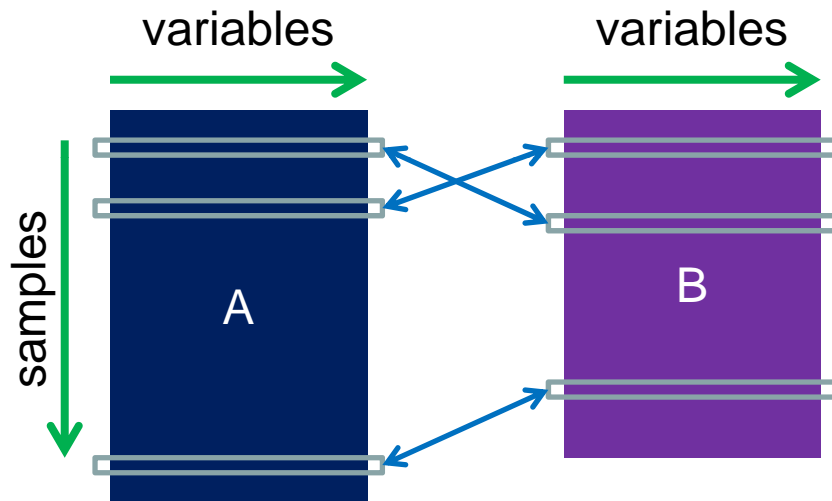
		Predicted		
		True	False	
Ground Truth	True	U	V	$n_0$
	False	T	S	$n-n_0$
		$n-R$	R	n

# MATLAB Demo

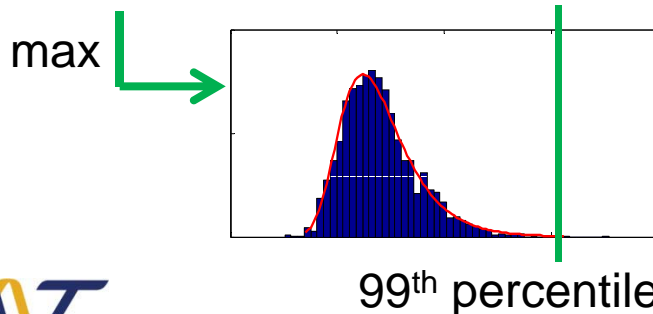
# Max-t Permutation Test

- Strongly controls FWER under any kind of dependence structure under certain assumptions.

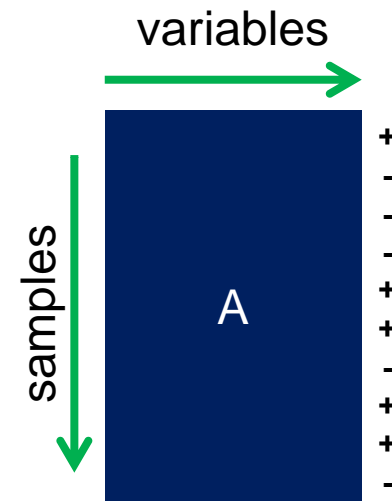
Two Sample



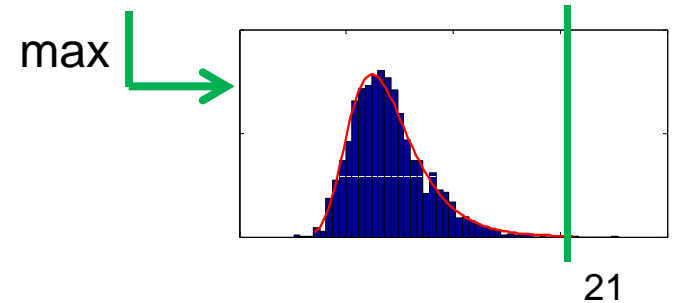
$$t^p = t\text{-test}(A^p, B^p)$$



One Sample



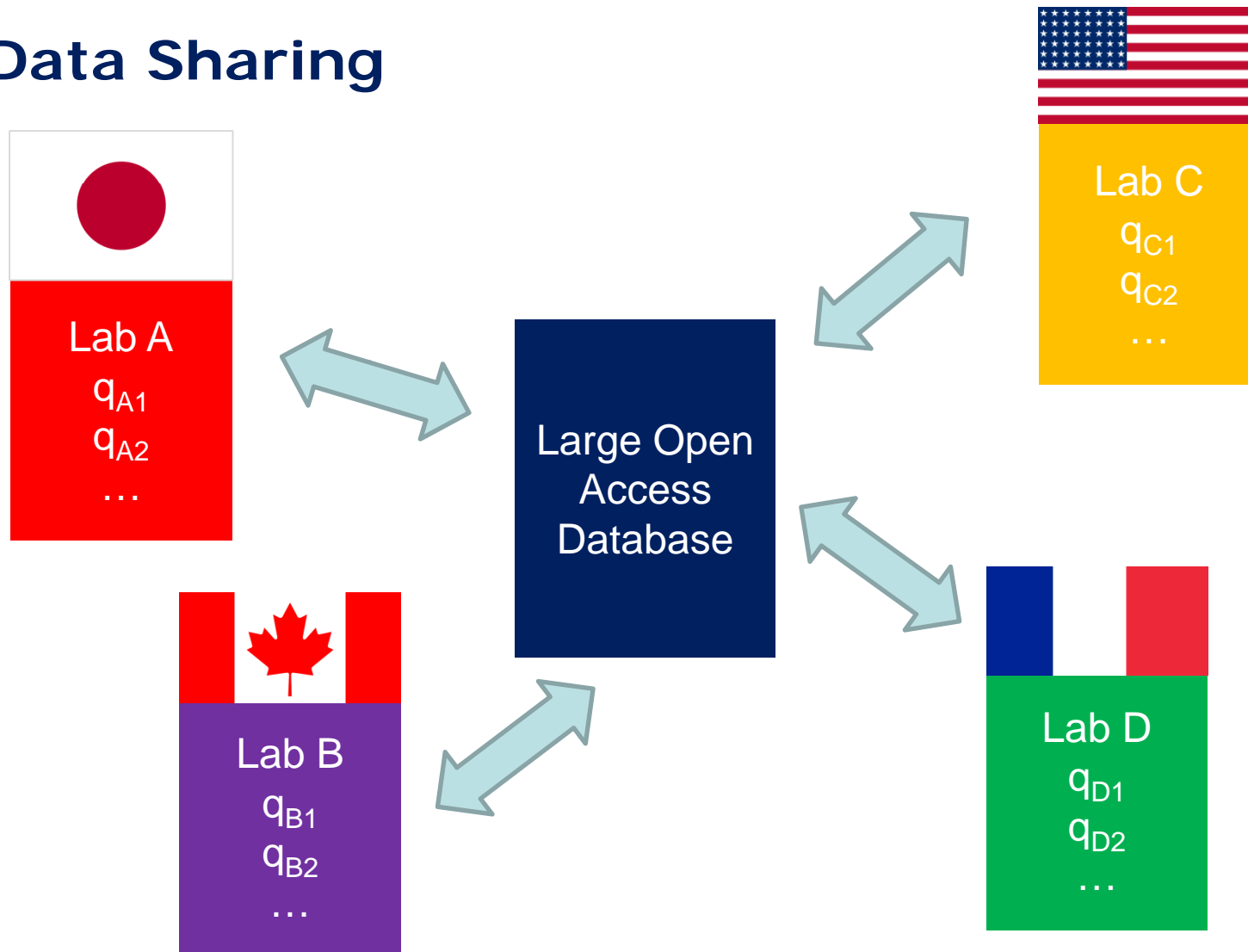
$$t^p = t\text{-test}(A^p)$$



# MATLAB Demo

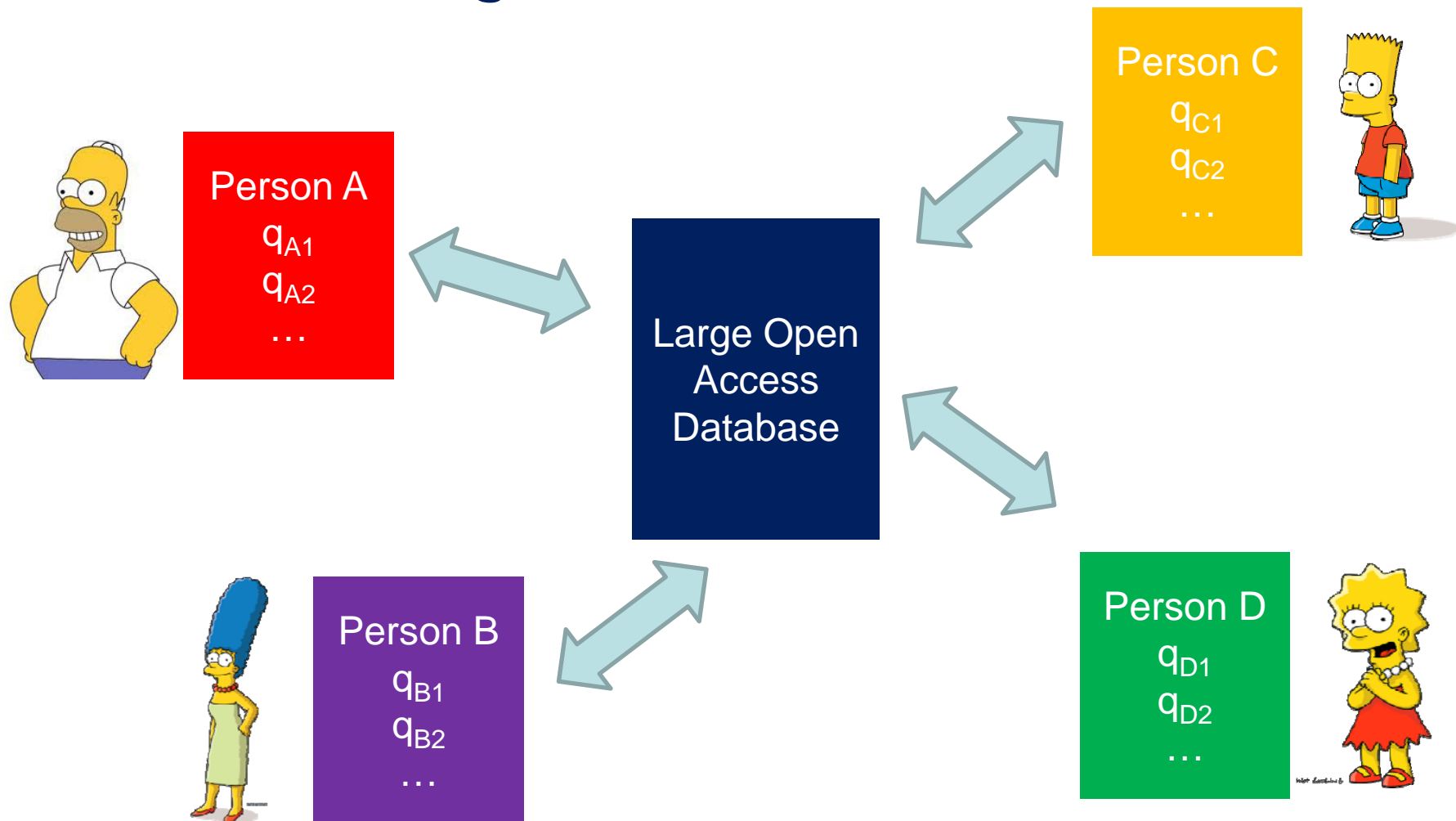
# Recent Topics

- Data Sharing



# Recent Topics

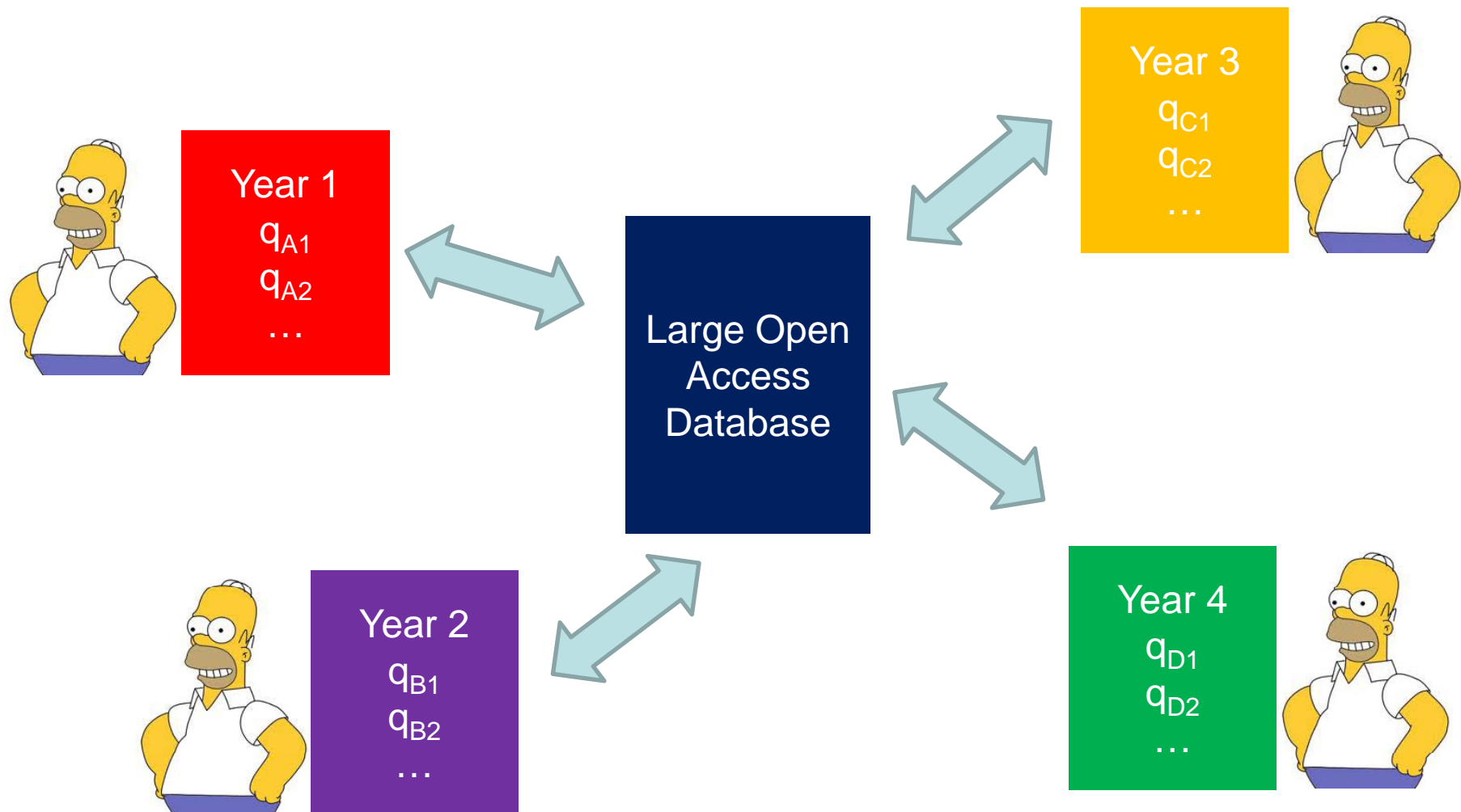
- Data Sharing





# Recent Topics

- Data Reuse



# Online FDR

Javanmard and Montanari, On Online Control of False Discovery Rate, 2015: arXiv:1502.06197

LORD (significance Levels based On Recent Discovery):

- Choose any sequence  $\underline{\beta} = (\beta_i)_{i=1}^{\infty}$ , such that

$$\beta_i \geq 0, \quad \sum_{i=1}^{\infty} \beta_i = \alpha.$$

- Rule is given by

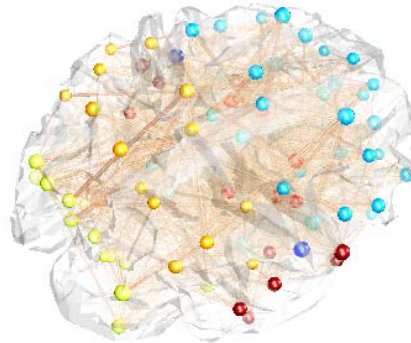
$$\tau_i \equiv \max \left\{ \ell < i, H_{\ell} \text{ is rejected} \right\}.$$

$$\alpha_i = \beta_{i-\tau_i}.$$

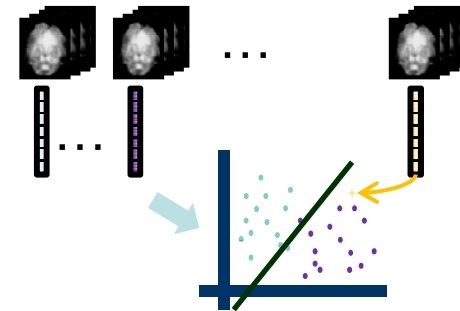
# Neuroimaging Applications



**Activation  
Detection**



**Connectivity  
Estimation**



**Brain  
Decoding**

# Shape vs. Function

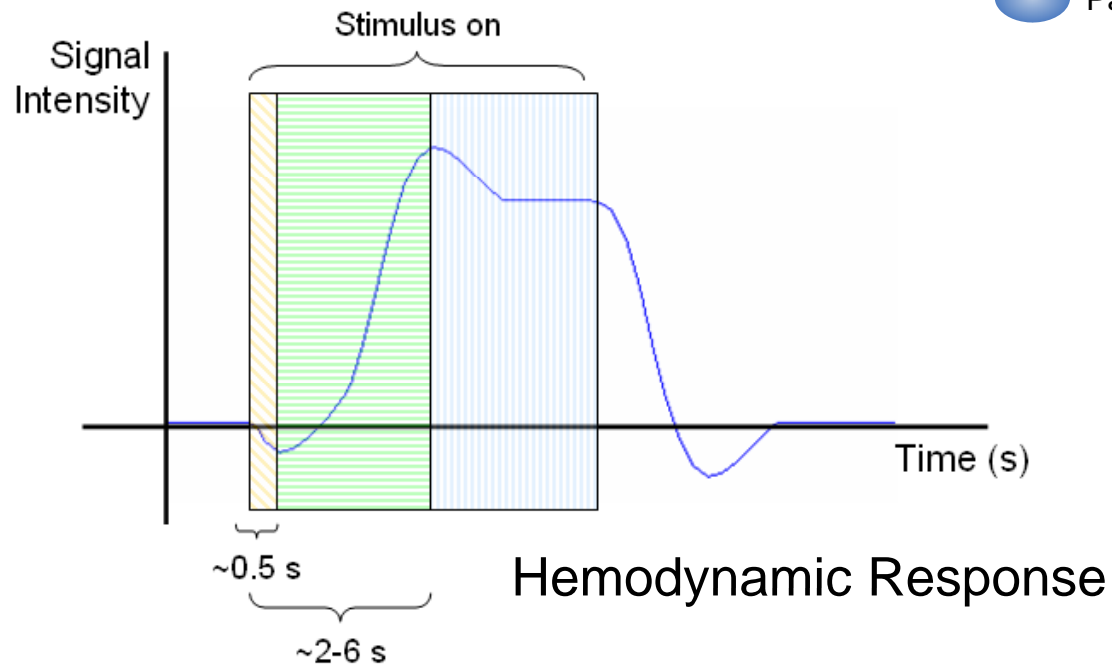
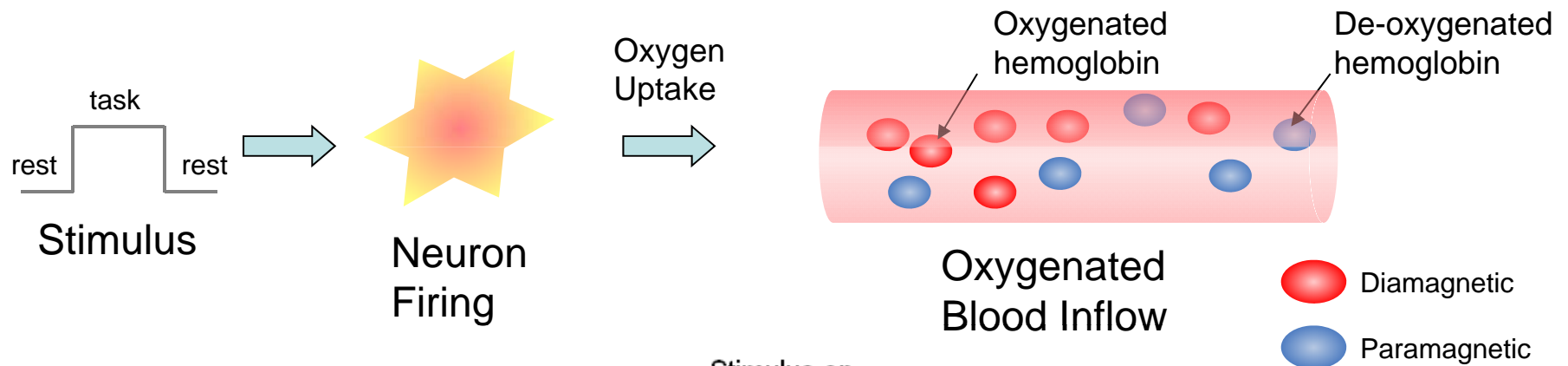


Magnetic Resonance  
Imaging (MRI)

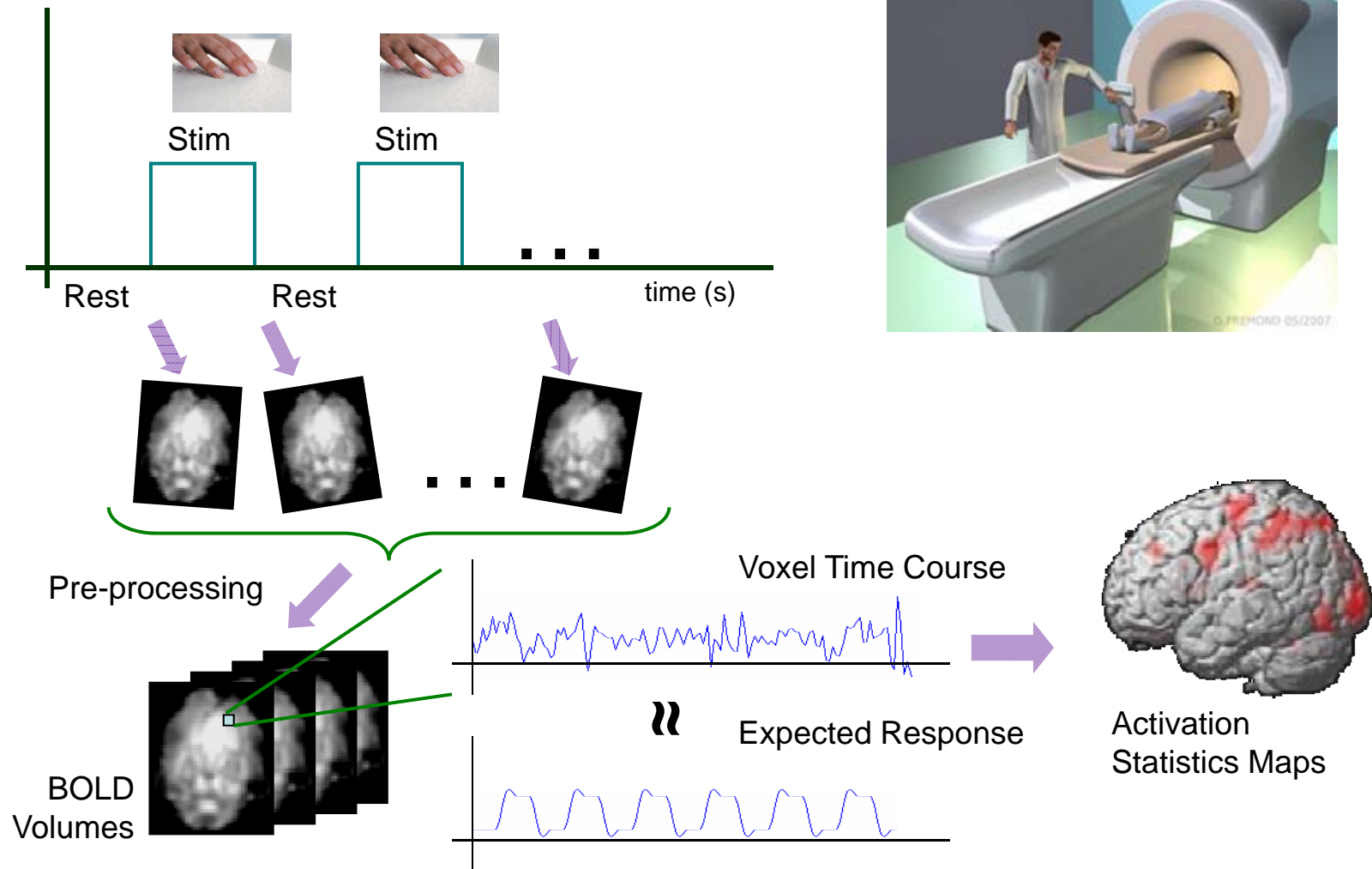


Functional  
MRI (fMRI)

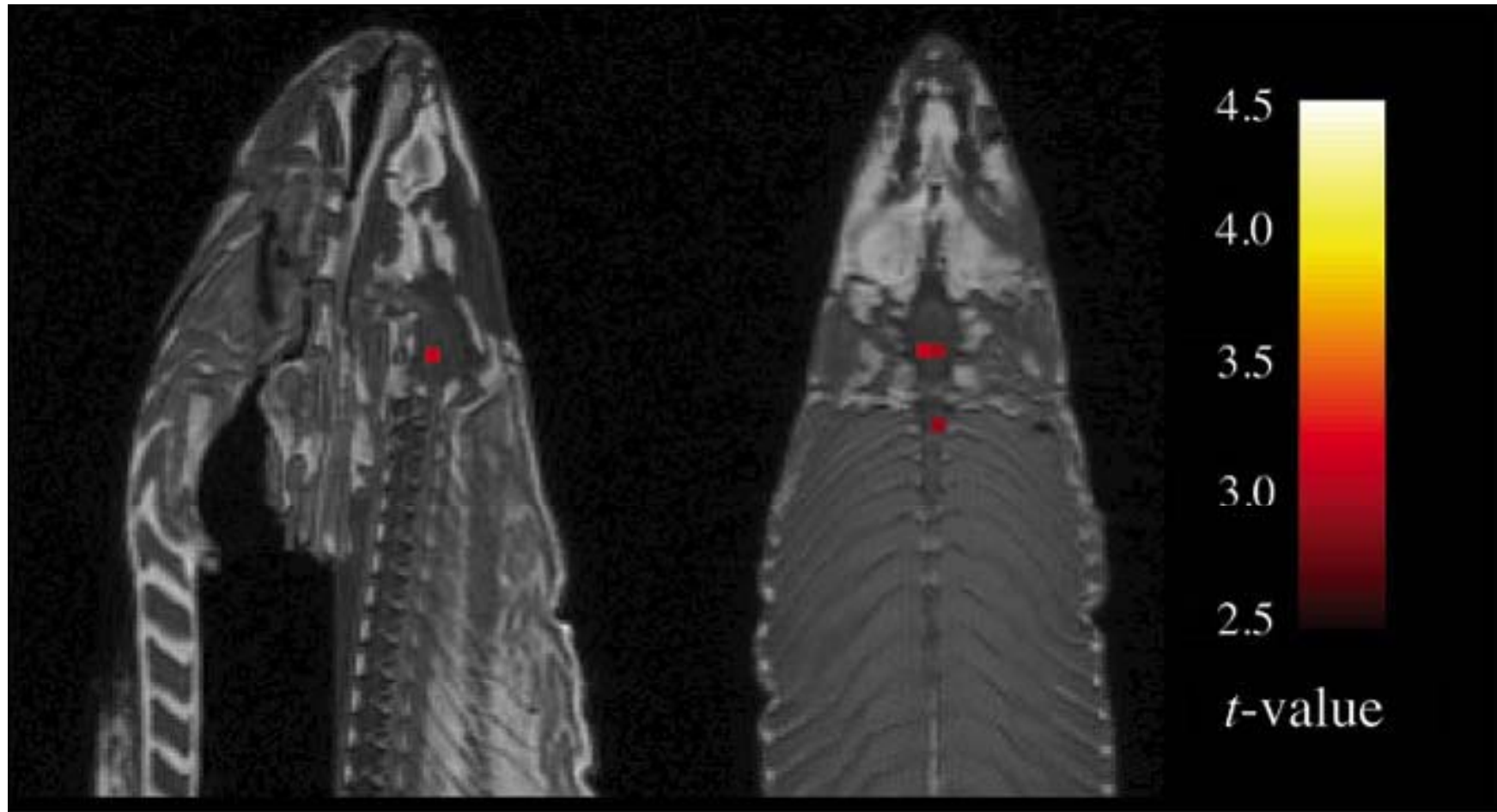
# BOLD Effect



# Task-based fMRI

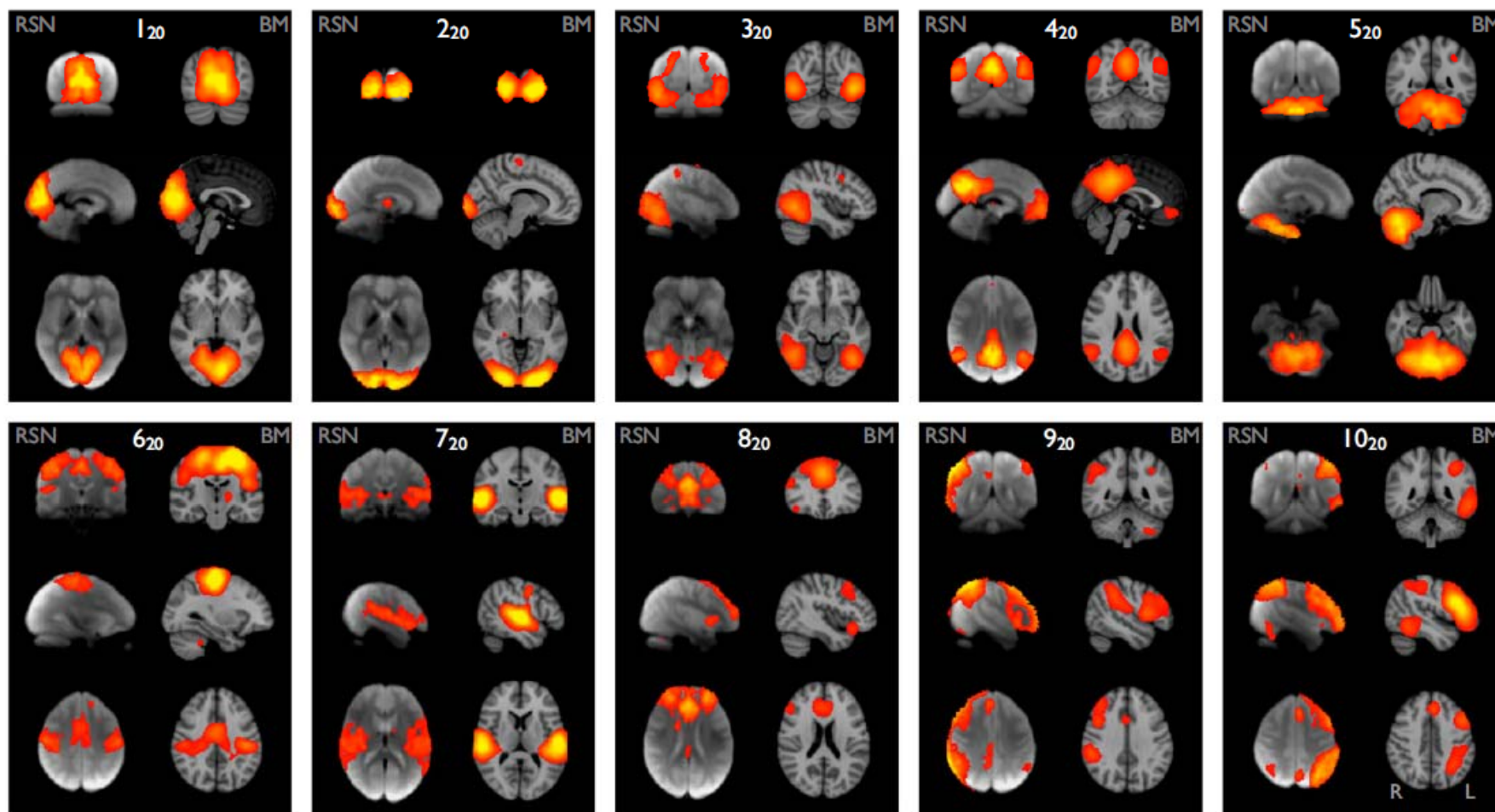


# Famous Dead Salmon Study





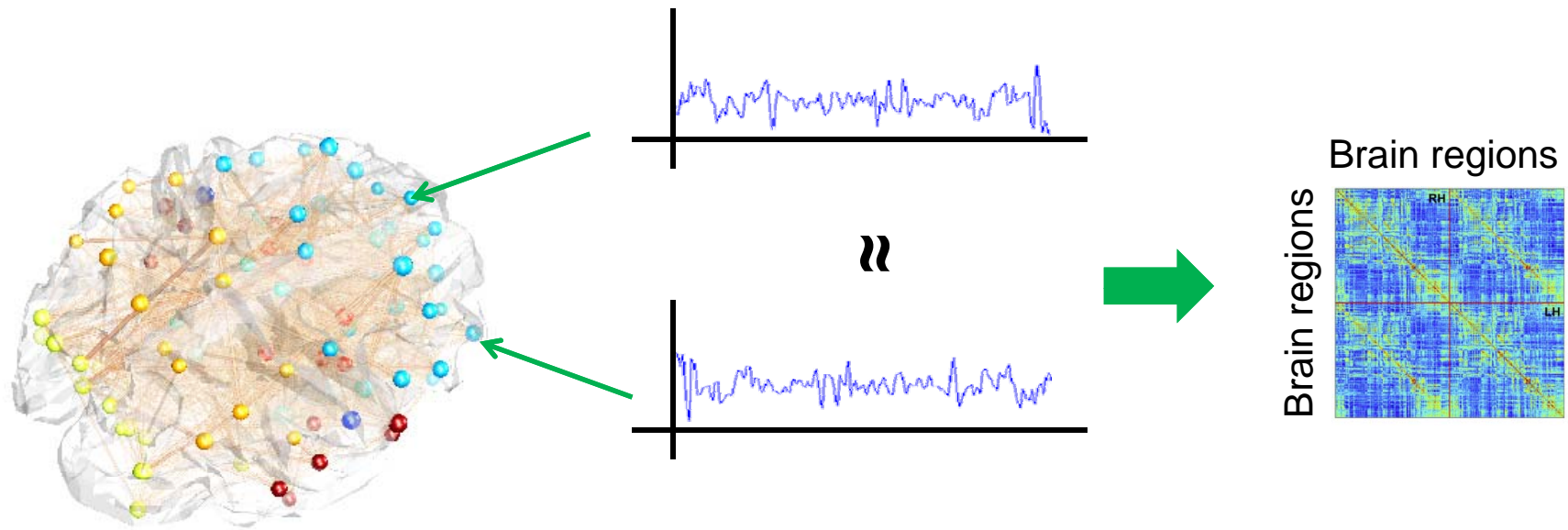
# Resting State fMRI



SM. Smith, P.T. Fox, K.L. Miller, D.C. Glahn, P.M. Fox, C.E. Mackay, N. Filippini, K.E. Watkins, R. Toro, A.R. Laird, and C.F. Beckmann, "Correspondence of the Brain's Functional Architecture During Activation and Rest," Proc. Natl. Acad. Sci., vol.106, pp.13040-13045, 2009

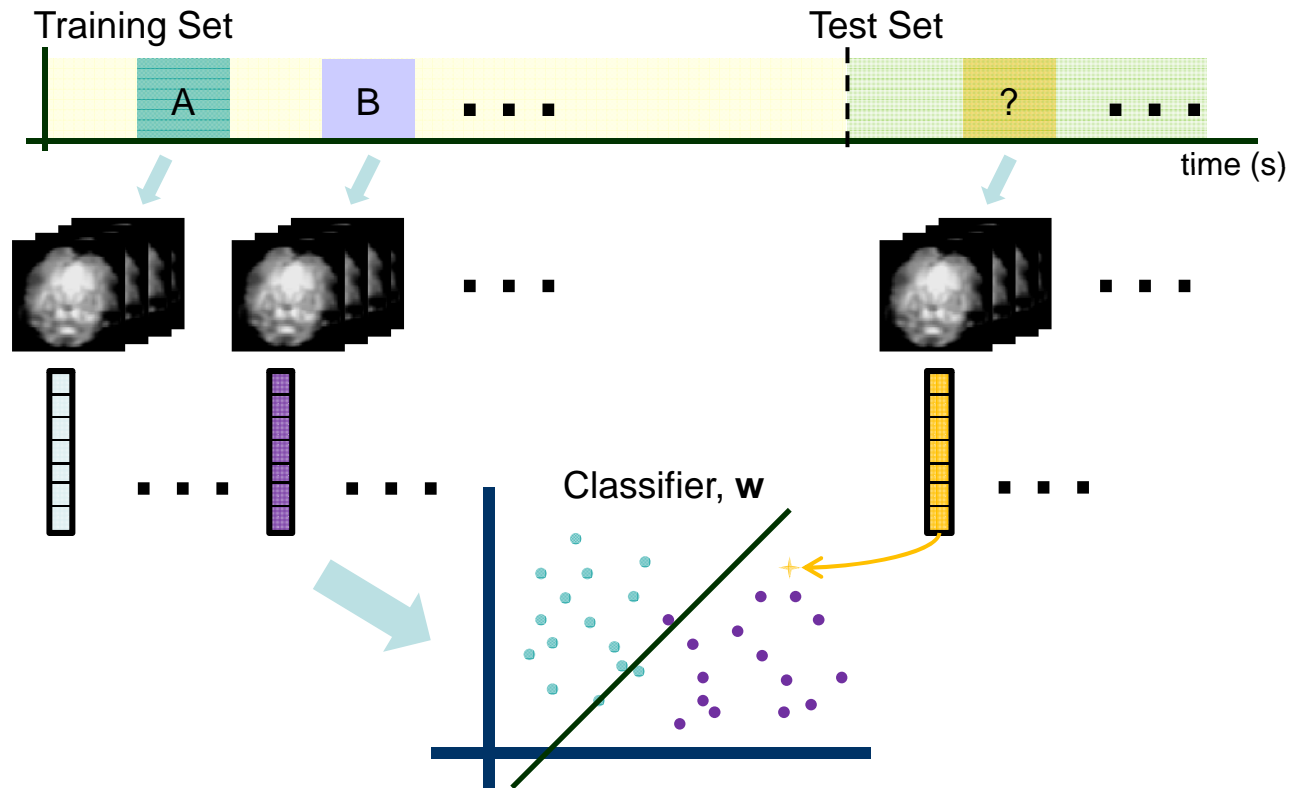


# Connectivity Estimation



Which connection significant?

# Brain Decoding



From  $w$ , which variable significantly drives classification?

# Summary

- Multiple testing can result in many false findings if the number of tests is not accounted for.
- Bonferroni correction is too stringent.
- FDR correction is a good compromise.
- Data sharing is creating new problems.