

# Neuroinformatics and largescale analysis

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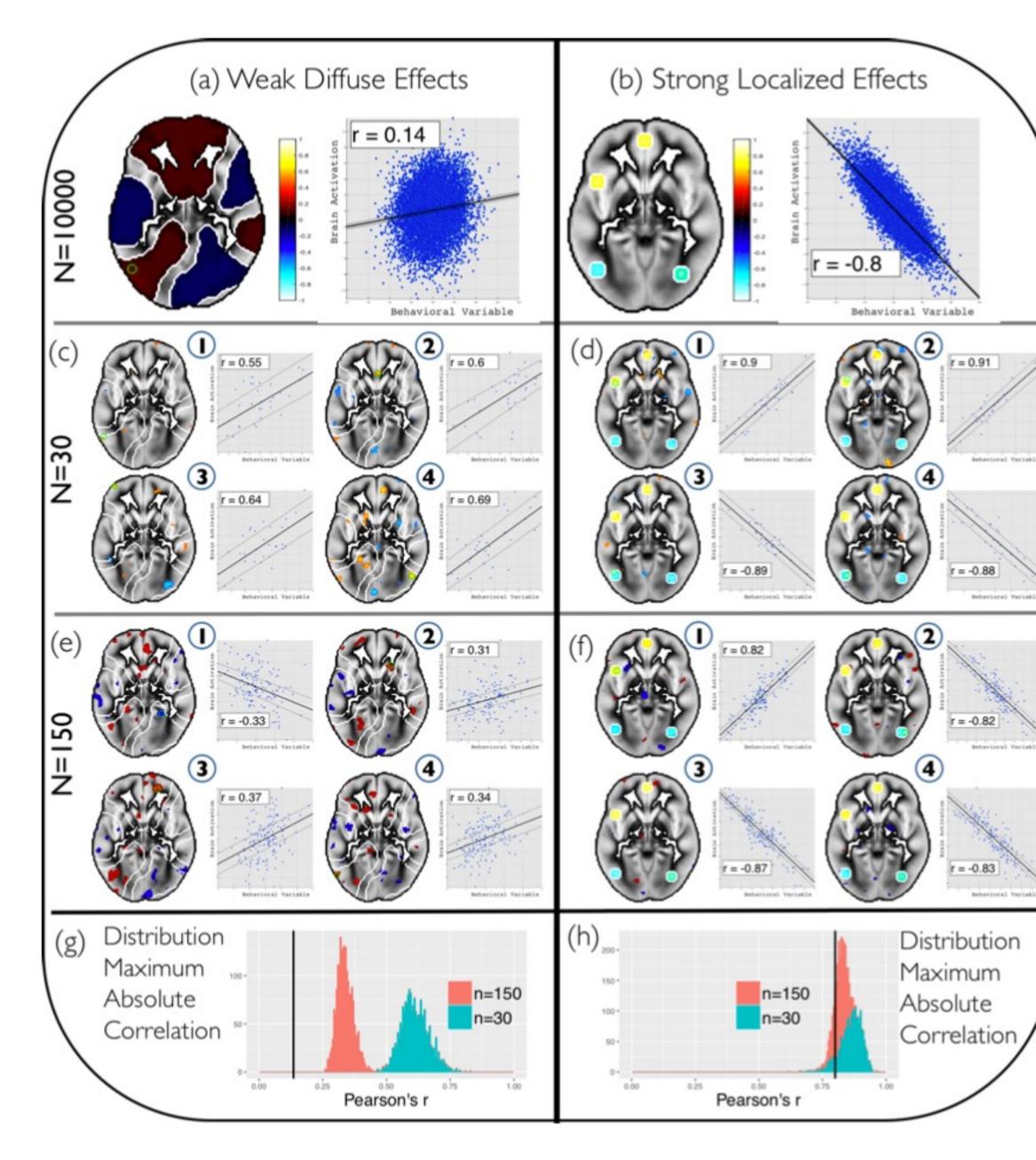




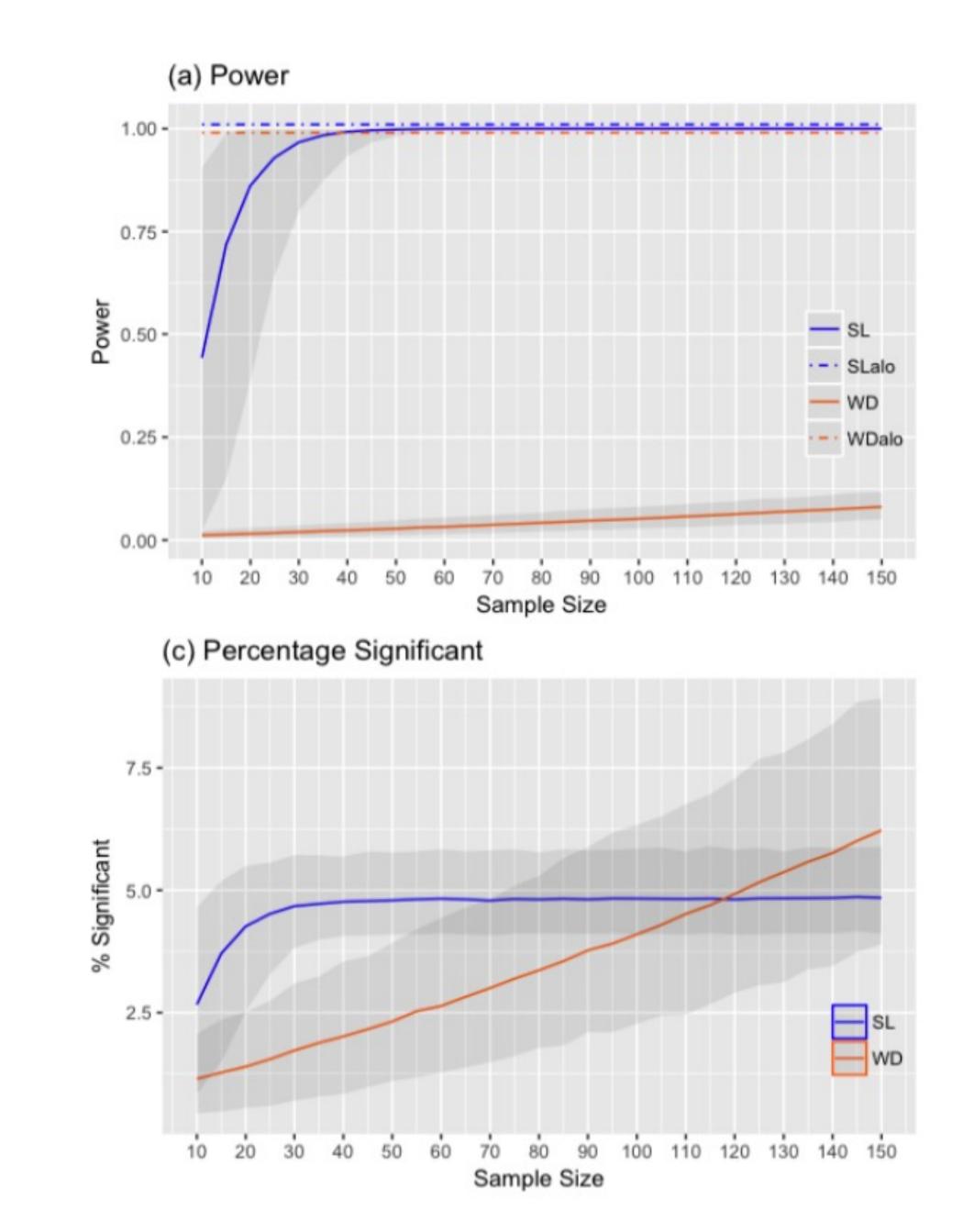


- Imaging is often financially prohibitive (>5000€ / PET scan) thus sample sizes are compromised
- Potential harm to subject needs to be minimised  $\longrightarrow$  sample sizes kept at minimum
- Underpowered studies can result in experimental failure even when experimental design and measurements are otherwise sufficient
- Poor statistical power increases Type 1 & 2 error rates and lead to poor replicability.

### More is more



Cremers et al (2017)



# Basic problems

### Data storage

- Where are my data?
- What if my lab members
  leave?
- How can i reaccess my data?

### Data processing

- What are my postdocs
  doing all day long?
- Are you sure the files are good?

### Data analysis

- Is everything done lege artis?
- How can we reproduce
  our results?

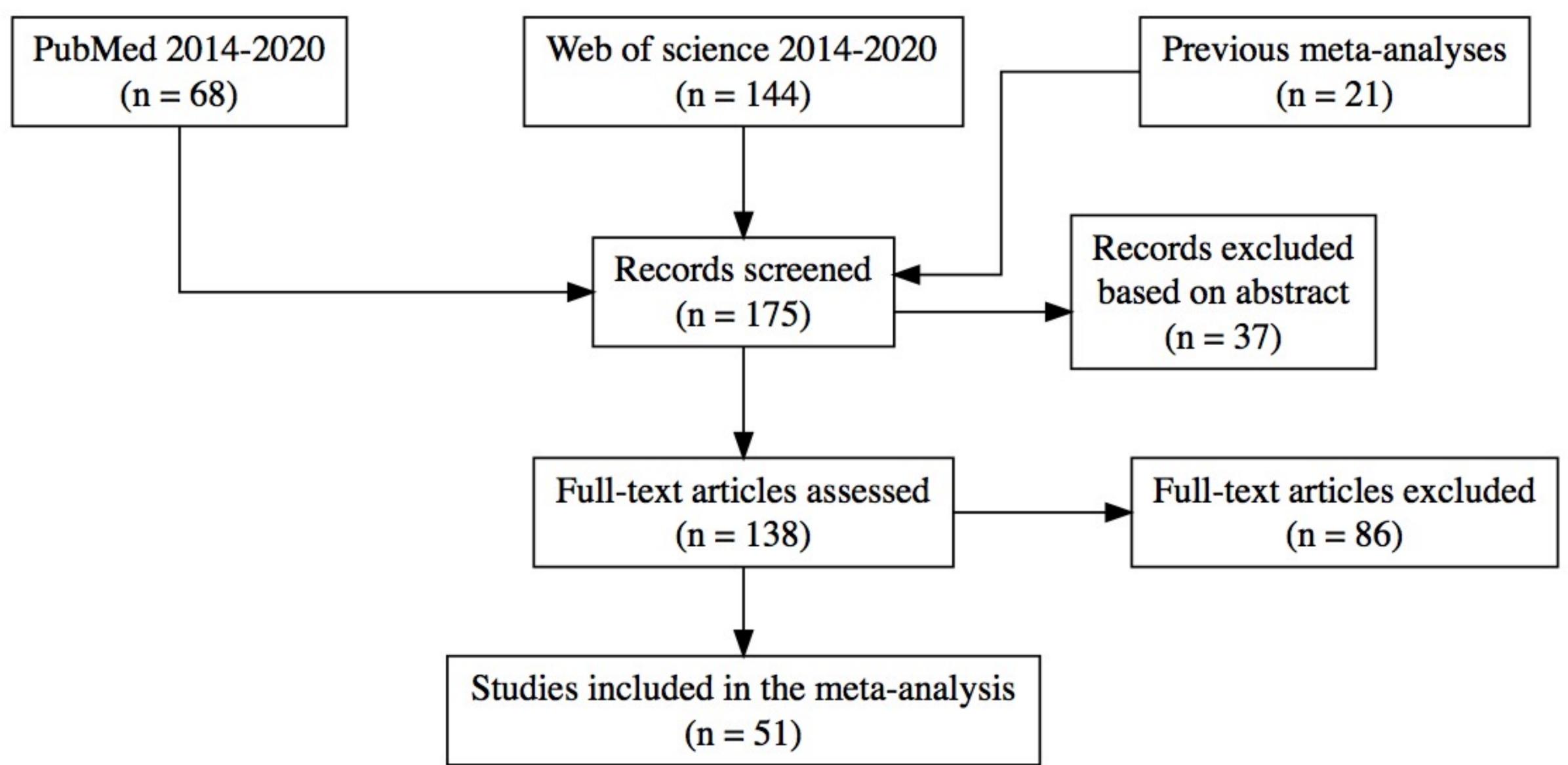
### Data synthesis

- How can we combine
  data?
- What can we combine?

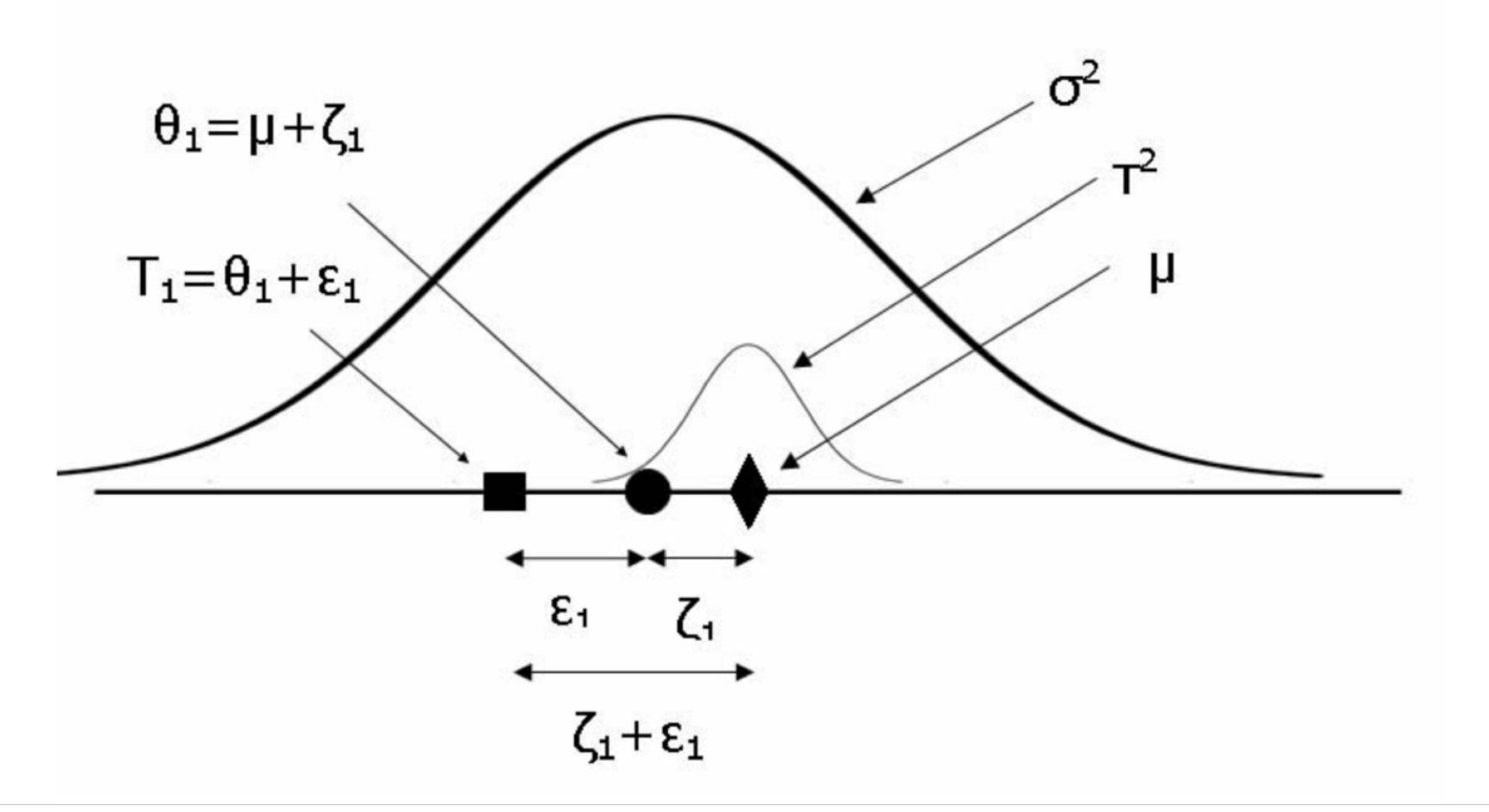
- Meta-analysis: Pooling standardised effect sizes to estimate population effect location and distribution
- For neuroscience, three main approaches
  - ROI level data and classic univariate meta-analysis
  - Coordinate-based data and volumetric meta-analysis
  - Combination of statistical maps from original studies

## Solution 1: Meta-analysis





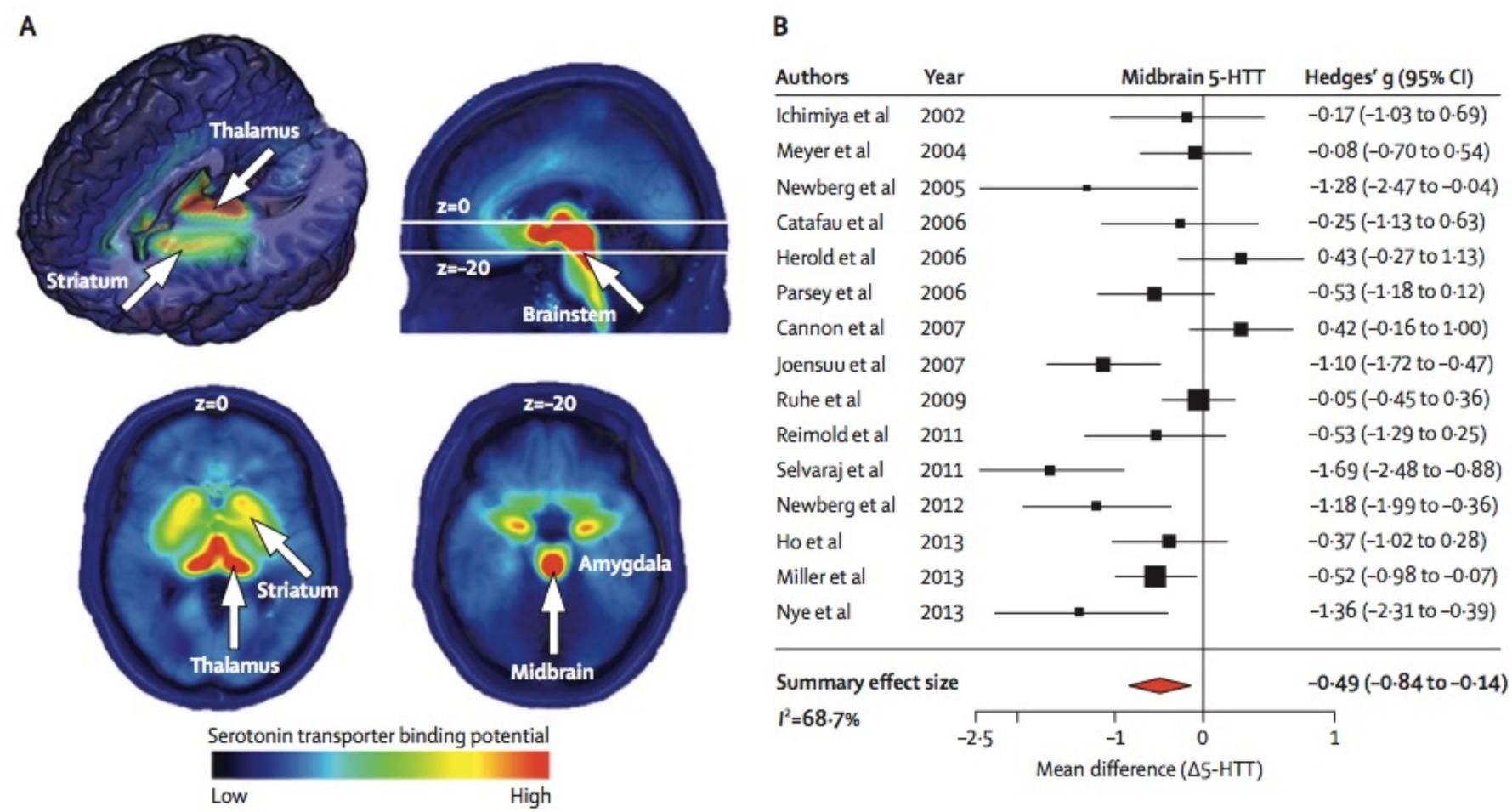
### Classic mixed effects meta-analysis



**The observed effect** T<sub>1</sub> is sampled from a distribution with true effect  $\theta_1$ , and variance  $\sigma^2$ . **This true effect**  $\theta_1$ , in turn, is sampled from a distribution with mean  $\mu$  and variance  $\tau^2$ .



# Approach 1: Regional analysis



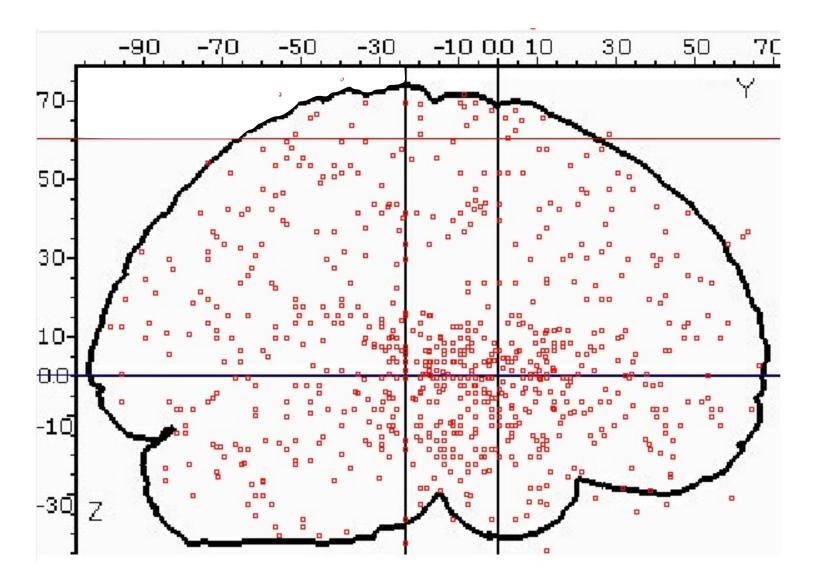
Spies et al (2015)

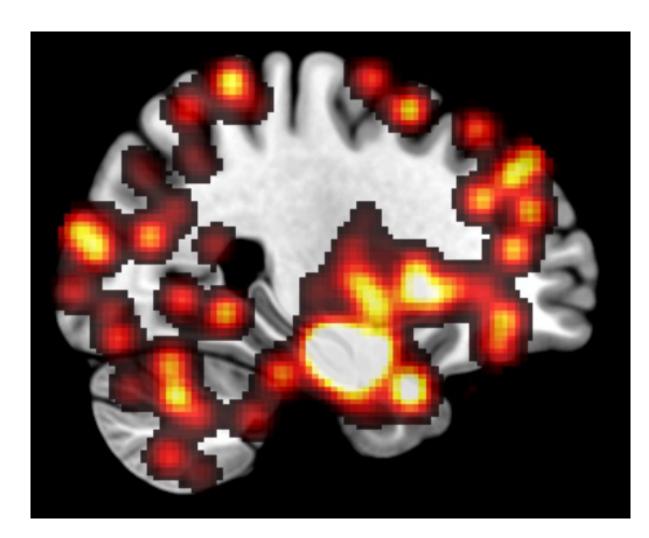
# Regional analysis: Pros and cons

- Easy to analyze and interpret
- Data comparable in statistical terms
- No need to worry about normalization etc.
- Laborious
- Anatomical nomenclature not consisten
- May miss effects outside chosen ROIs

# Approach 2: Peak-based analysis

### Individual foci



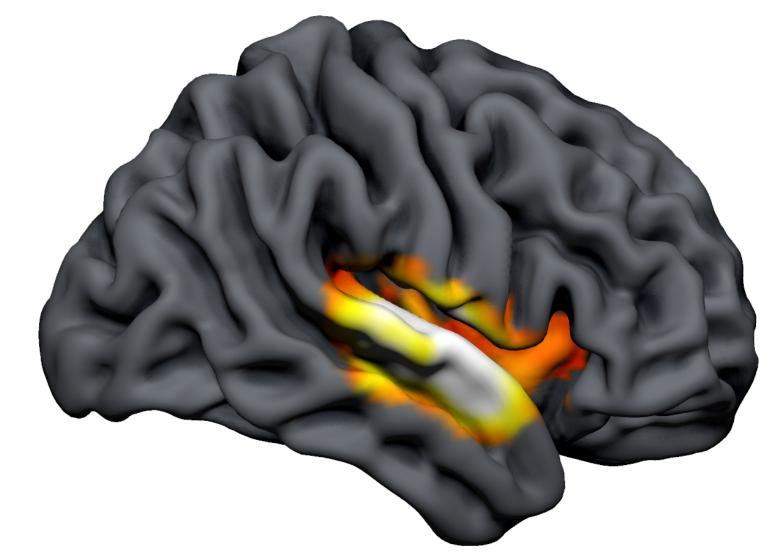


### Convergence of activation locations at given threshold

Activation Likelihood Estimation (Eickhoff et al 2015)

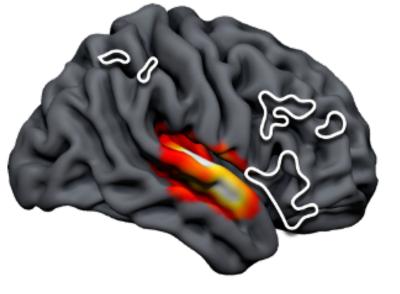
### Permutation

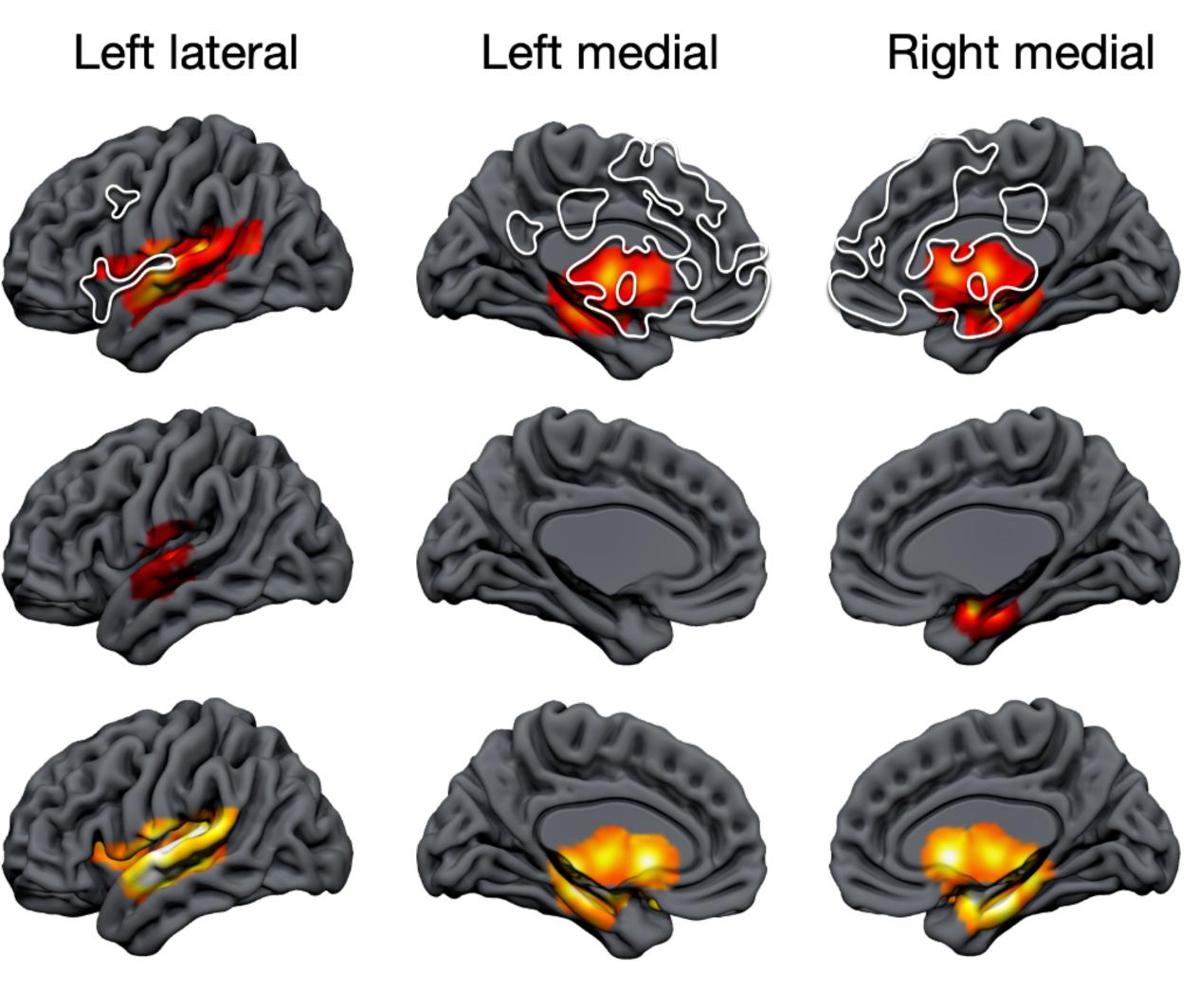
### Thresholding



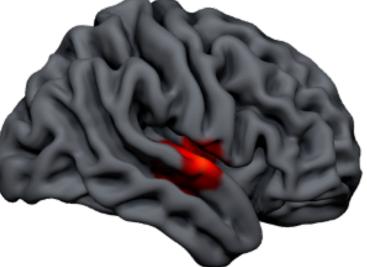
### **Right lateral**

### Positive emotions

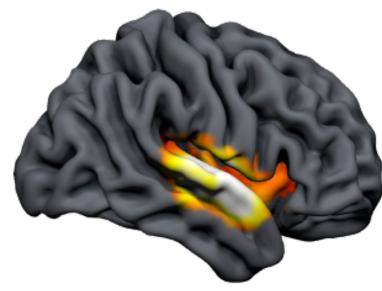


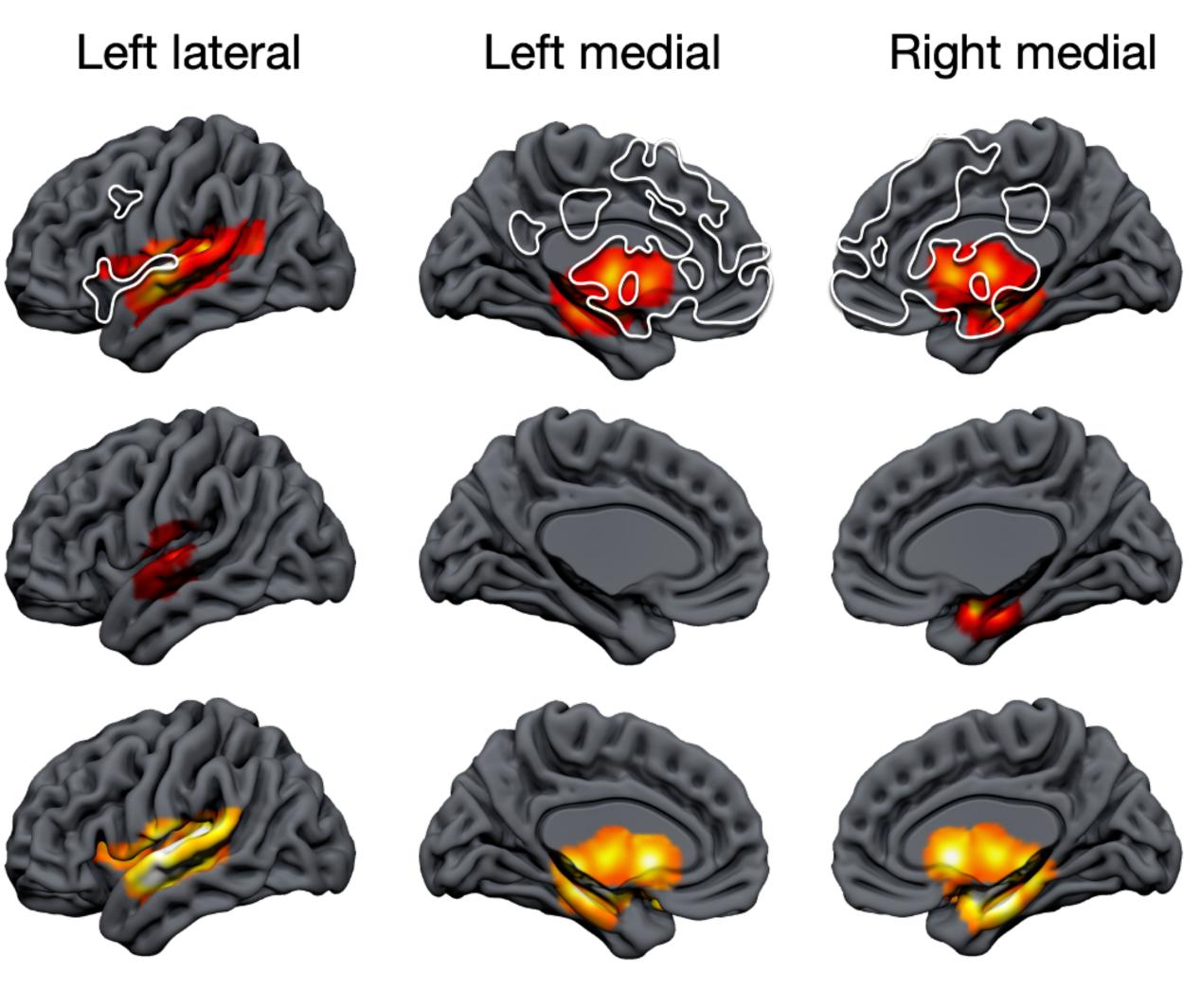


### Negative emotions









0.005

ALE value

0.02

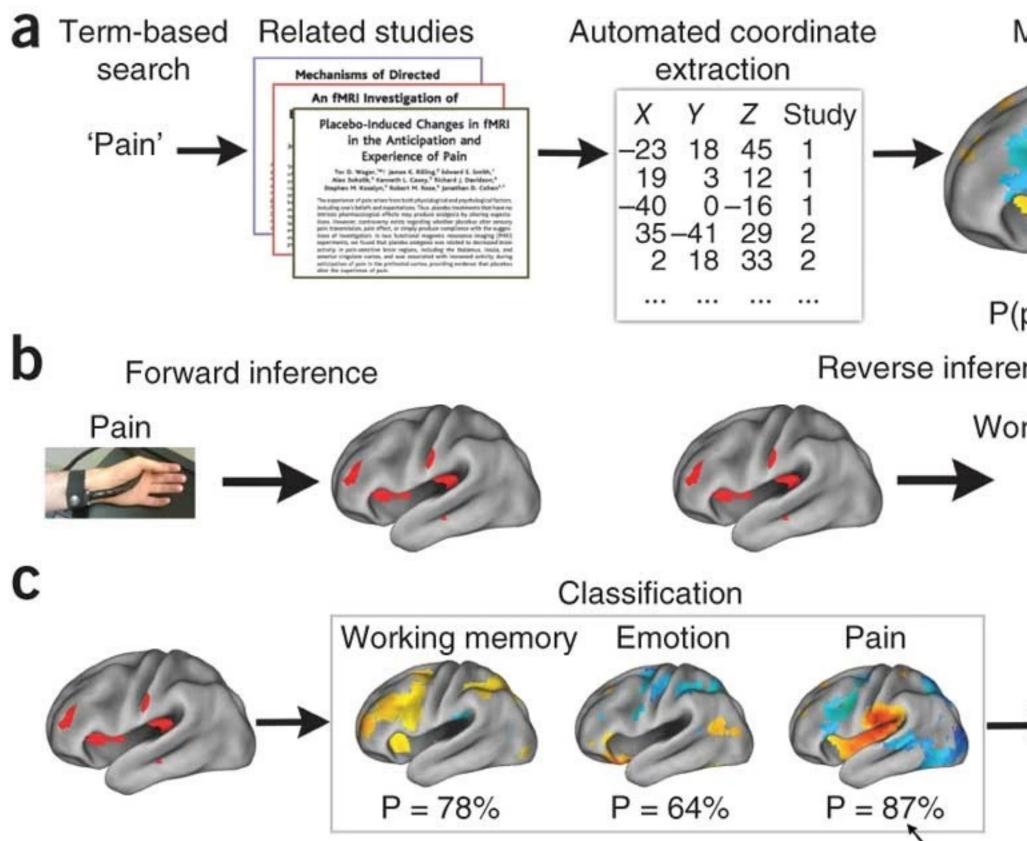
### Nummenmaa, Putkinen & Sams (COiBS 2021)

## ALE Pros and cons

- Relatively easy to analyze and interpret
- Full-volume analysis
- No need to worry about normalization etc.
- Effect sizes scaled only by sample size
- Requires coordinate-levels data

Data modelled per peaks —> cluster size not taken into consideration

### Approach 3: Automated data mining



Select highest pro

NeuroSynth (Yarkoni et al, 2011)

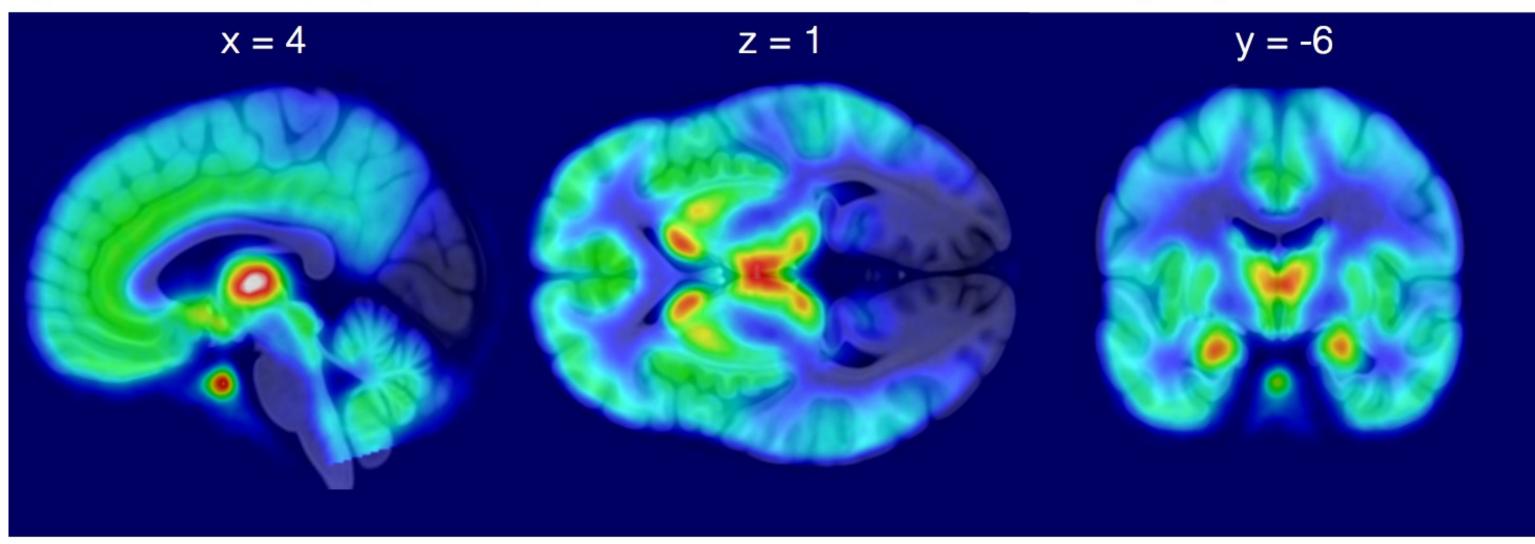
### Meta-analysis

Meta-analysis	Semantic	72																						
	Encoding		68																					
	Executive	70	61	68																				_
	Language	62	68	70	73																			
	Verbal	69	67	53	69	69															ssifie			
	Phonological	68	72	79	64	62	76											(	perc		urac age c		ect)	
	Visual	73	67	63	73	70	67	71																
	Interference	65	62	52	66	57	74	55	66															
(pain activation)	Working memory	72	68	54	71	60	75	68	56	70														
2000	Conflict	77	67	64	74	67	77	75	61	63	73													
ence	Spatial	77	68	63	76	67	76	67	62	64	71	69	_											
orking memory?	Attention	74	65	64	74	69	73	69	56	65	67	54	69	_										
Emotion?	Imagery	69	65	61	68	64	77	61	62	56	72	53	54	67	_									
	Action	75	70	71	77	69	73	65	64	68	72	63	57	54	71									
Pain?	Sensory	74	73	73	74	71	81	76	69	72	73	65	63	56	60	72	_							
?	Perception		69	74	70	72	75	70	68	73	79	69	67	67	65	59	72	_						
	Auditory			73								75	77		-	76			_					
	Pain	85	81	86	84	83	89	83	78	84	78	80	81	80	83	76	77	80	82	_				
	Reward	79	74	76	84	75	88	78	76	79	80	76	74	76	83	76	74	82	80	77				
	Arousal	75	67	76	79	74			68	79	76	75	74	74	76	73	74	77			73	_		
→ 'Pain'	Emotion	75	70	81	81	78	83	78	76	80	83	78	80	78	80	76	77	83				76		
	Social	74	68		75	73	78		74	77	78	75	72	72	74	71	72	77	80			64		
	Episodic	65	61	76	76	75		71	75	76	75	72	74	77	75	78	75	79	88	75			62	
	Retrieval	71	71	66	75	69	76	75	70	72	75	71	73	71	77	79	76	79	85	75	75	76	72	
obability	Recognition	69	60	66	75	68	74	71	65	70	73	70	69	71	75	75	71	80	84	76	70	71	65	
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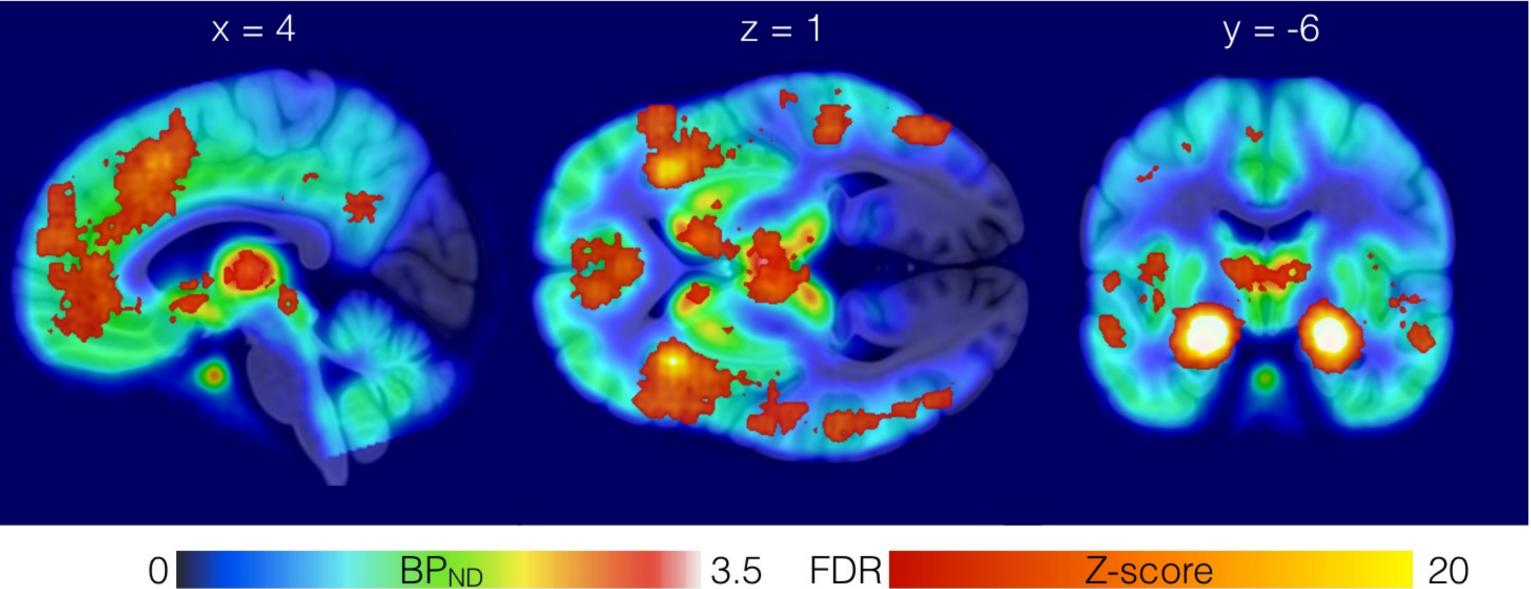




### A) Distribution of µ-opioid receptors in the brain as measured with [11C]carfentanil PET



### B) Overlap between human emotion circuit and the µ-opioid receptor system





 $r_{all} = 0.38$  $r_{\text{pleasure}} = 0.44$ 

Nummenmaa & Tuominen (2018 Br J Pharmac); Kantonen et al (2020 Neurolmage)



- Very easy to analyze and interpret
- Data readily available, allows custom analyses
- Full-volume analysis
- Quality contingent on the parser & reporting in studies
- Currently distinguishing activation / condition direction difficult

### Neurosynth: Pros and cons

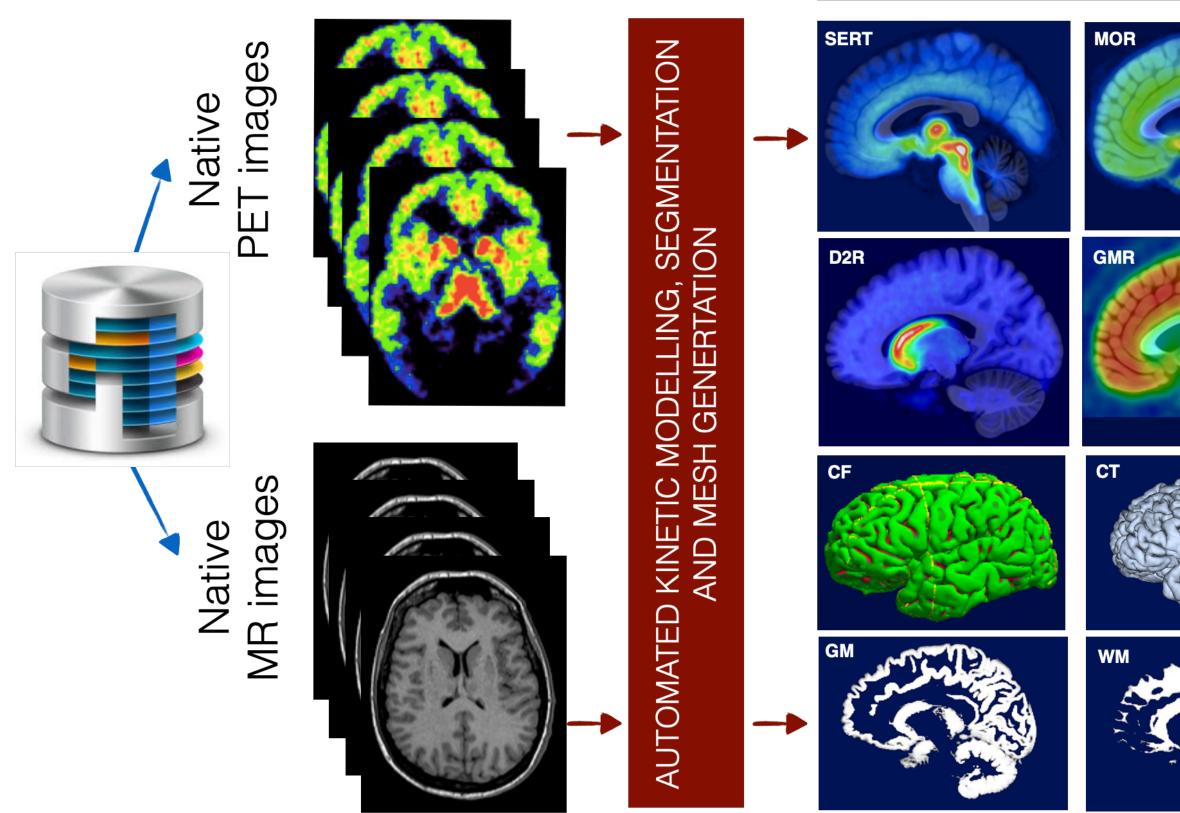
### Solution 2: Large-scale synthesis of old datasets

- Between-study variability and reliance on statistical estimates (rather than raw data) lower the power of meta-analysis
- Existing data is often available and cheap to use given permissions can be reanalysed
- Data however have to be extracted, reprocessed and the metadata needs to be extracted

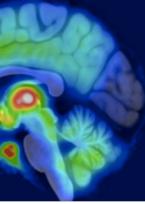
# Integrated approach at PET Centre

### Preprocessed BRAIN data

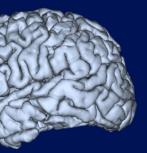
### Hospital **PACS**



Automated and supervised quality control



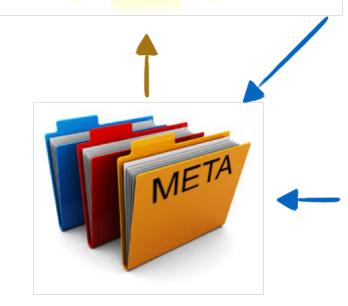






### AIVO database





### **EXTERNAL** register sources

### SOMATIC WELL BEING

- 1. ICD codes for diagnoses
- 2. Laboratory results
- 3. Frequency of hospitalisation & sickness leaves

### **PSYCHOLOGICAL FUCNTIONING**

- 1. Psychopathology (ICD codes)
- 2. Personality structure
- 3. Disorders of cognition (ICD codes)
- 4. Fluid intelligence and school achievement

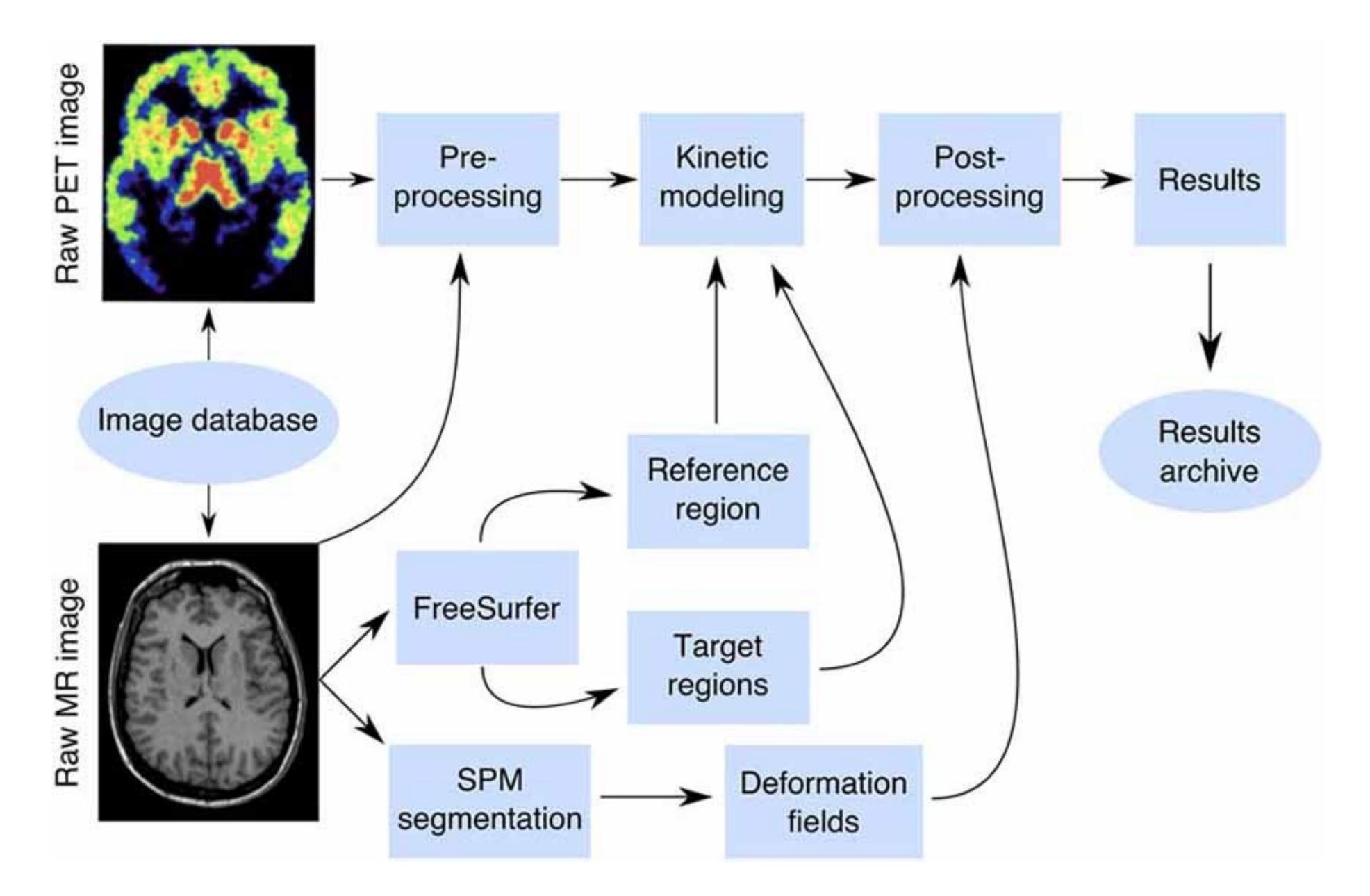
### SOCIOECONOMIC WELL BEING

- 1. Earning and income transfers
- 2. Education and social status
- 3. Labour market attachment
- 4. Developmental socioeconomic stressors

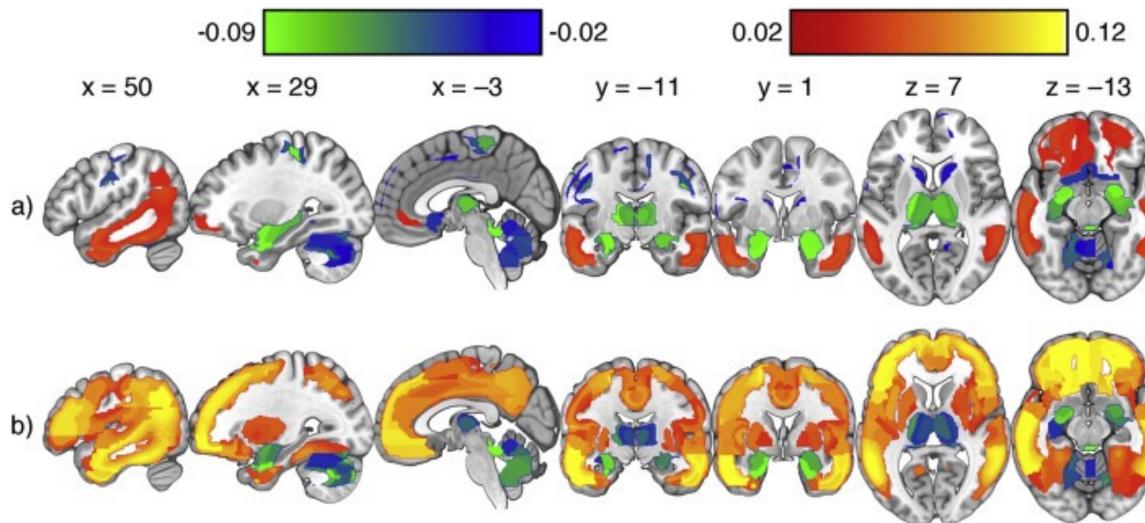
### SOCIAL ATTACHMENT BEHAVIOUR

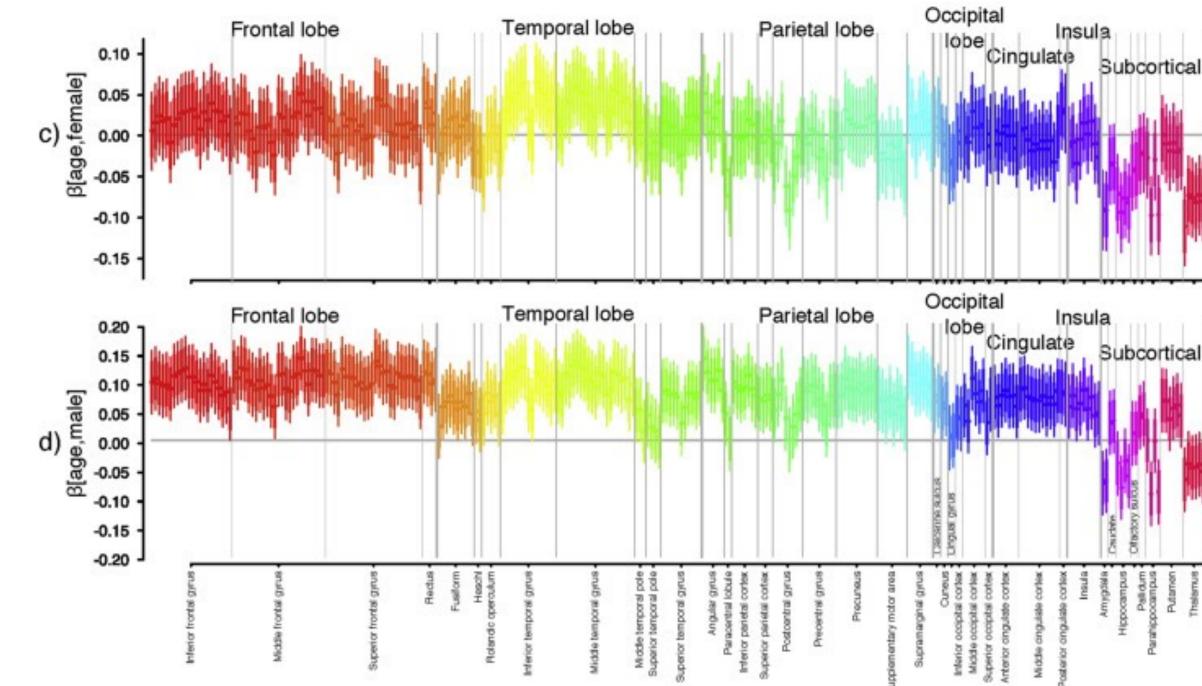
- 1. Marriage and cohabitation
- 2. Family establishment
- 3. Reproduction and family size

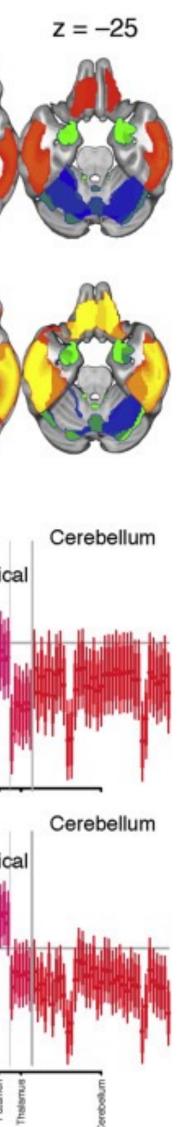


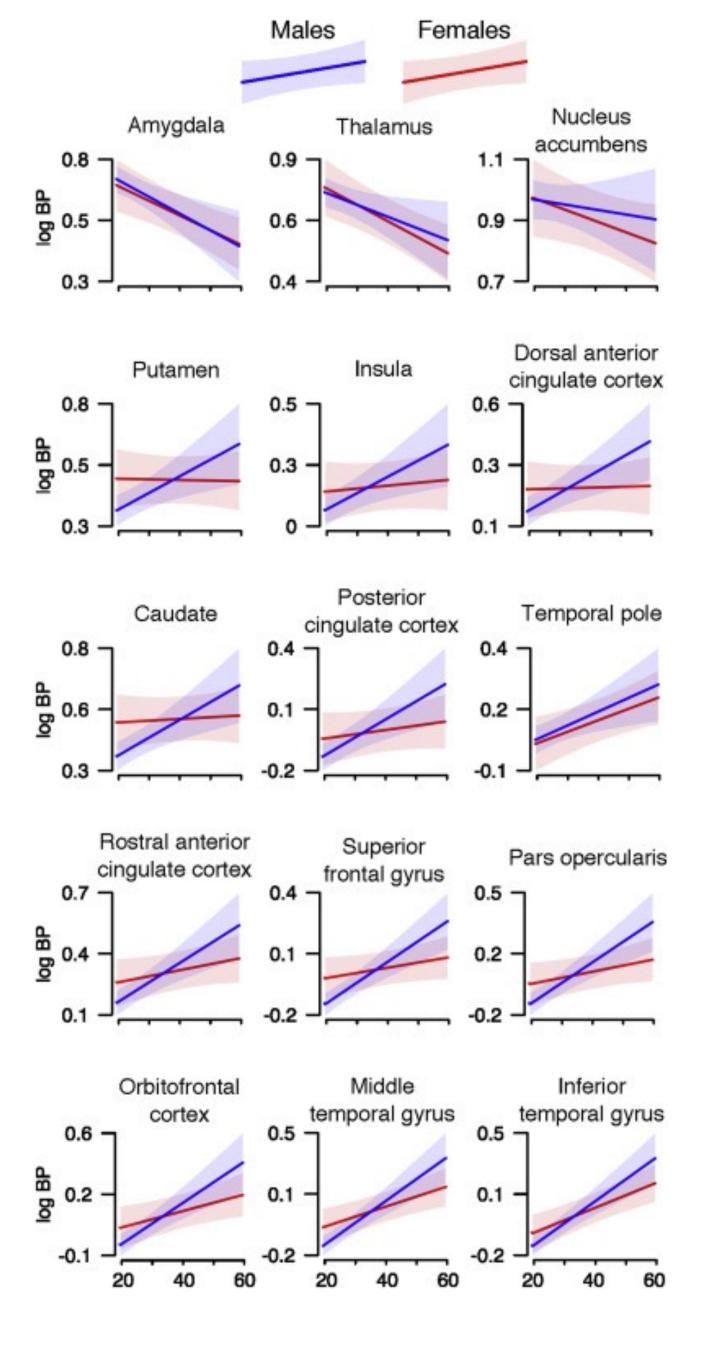


Karjalainen et al (2020)





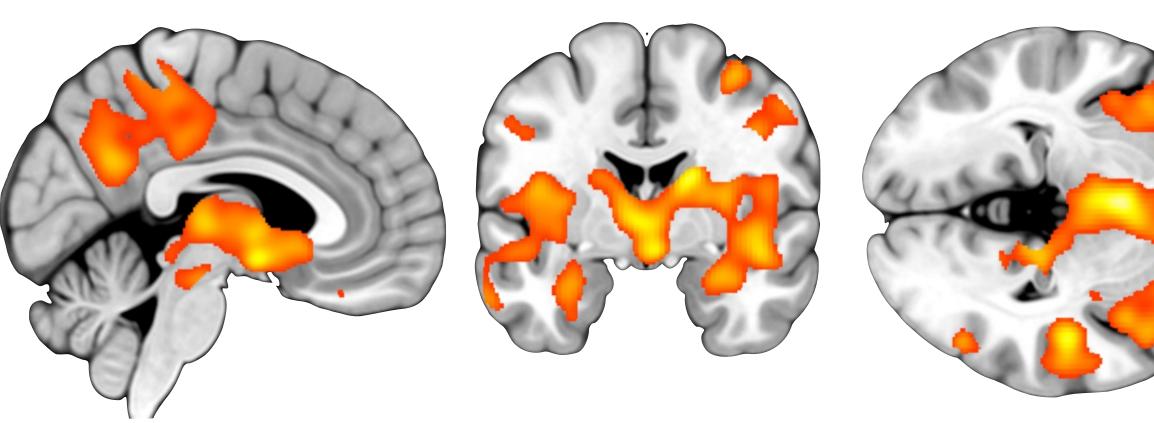




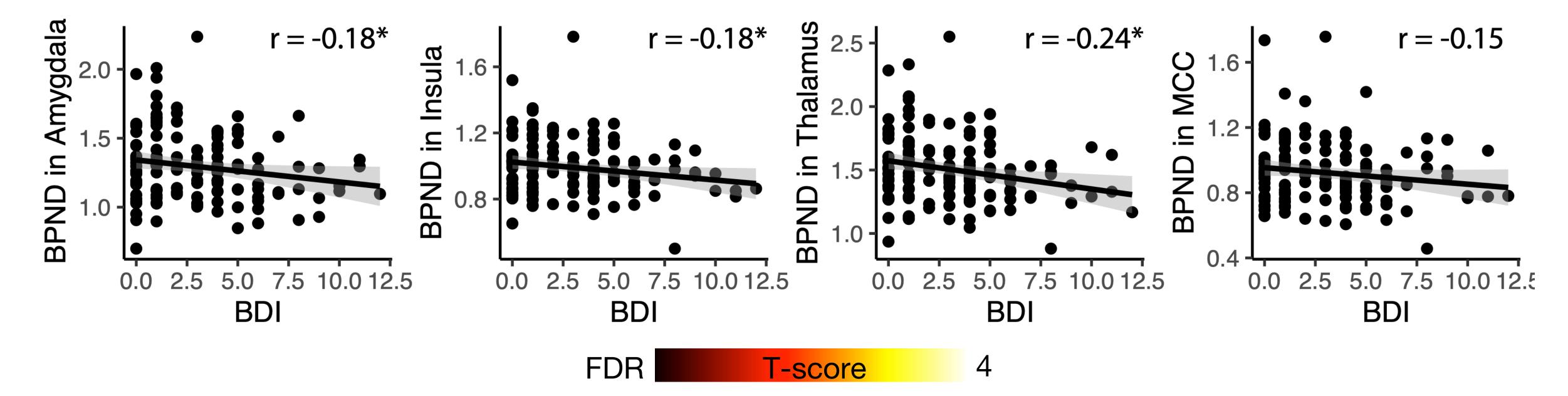
Kantonen et al (2020 Neurolmage)

### Lowered mu-opioid receptor availability in subclinical depression and anxiety

**X** = 4

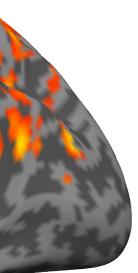


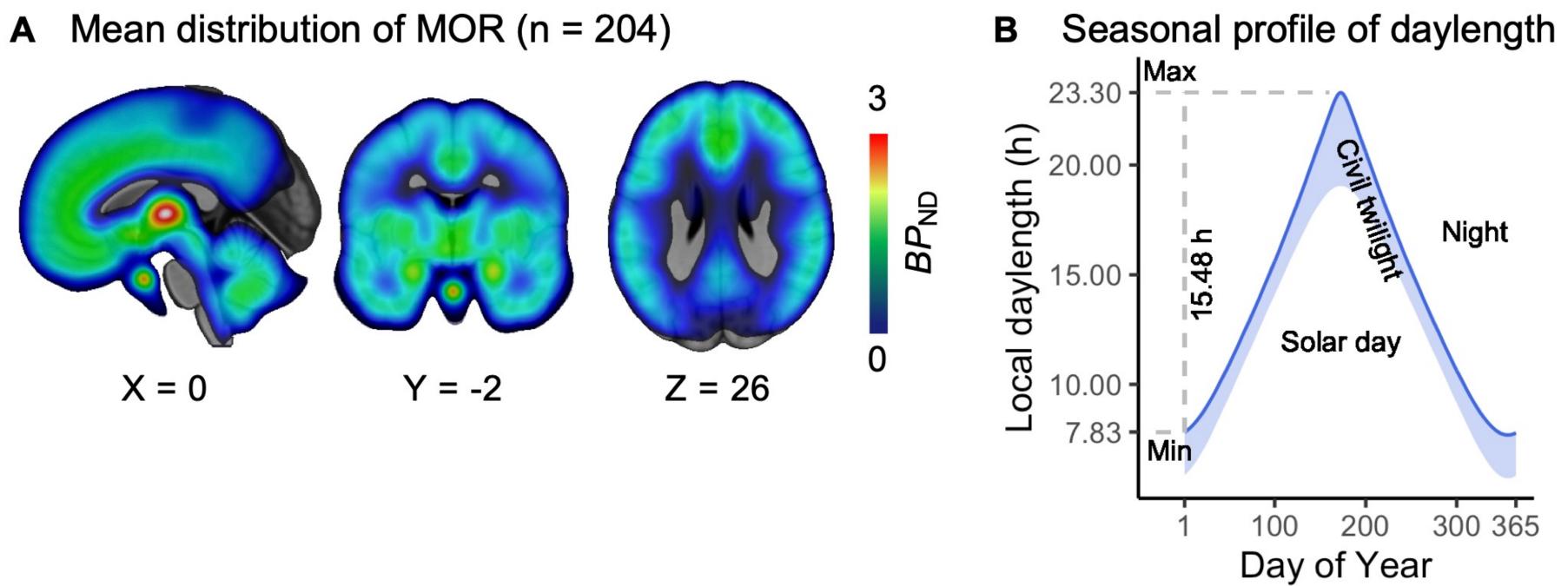
y = -2



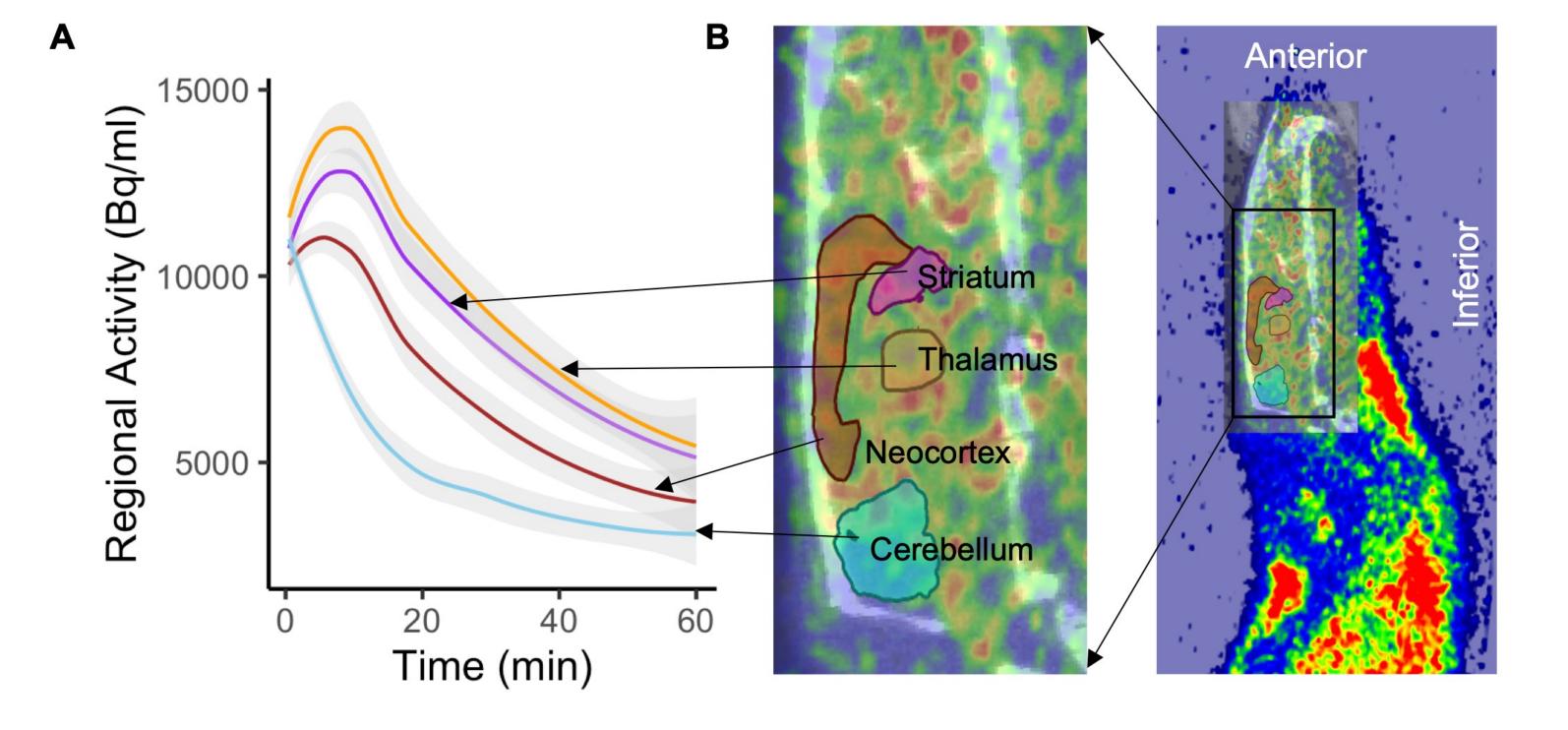
Nummenmaa et al (2020 Neuropsychopharmacology)

Z-1 Left Right





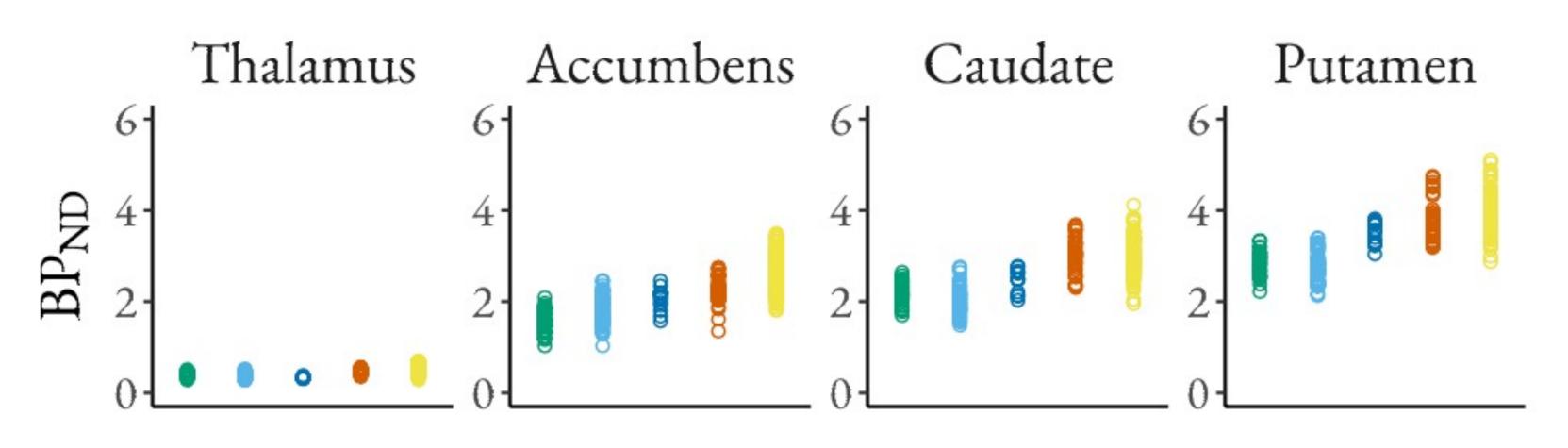
Sun et al (2021 J Neurosci)



Sun et al (2021 J Neurosci)

### Common problems with data integration

- Variable imaging equipment
- Standarization of data acquisition
- Metadata description



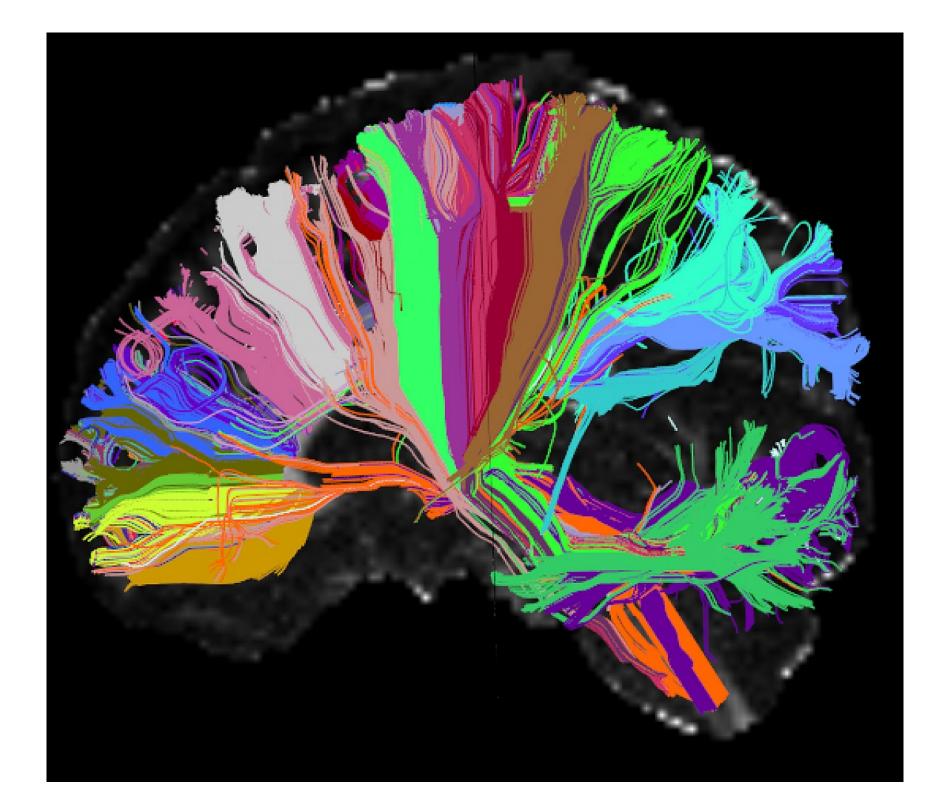
Malen et al (submitted)

- Processing pipelines
- Comparability of conditions
- Specificity of effects

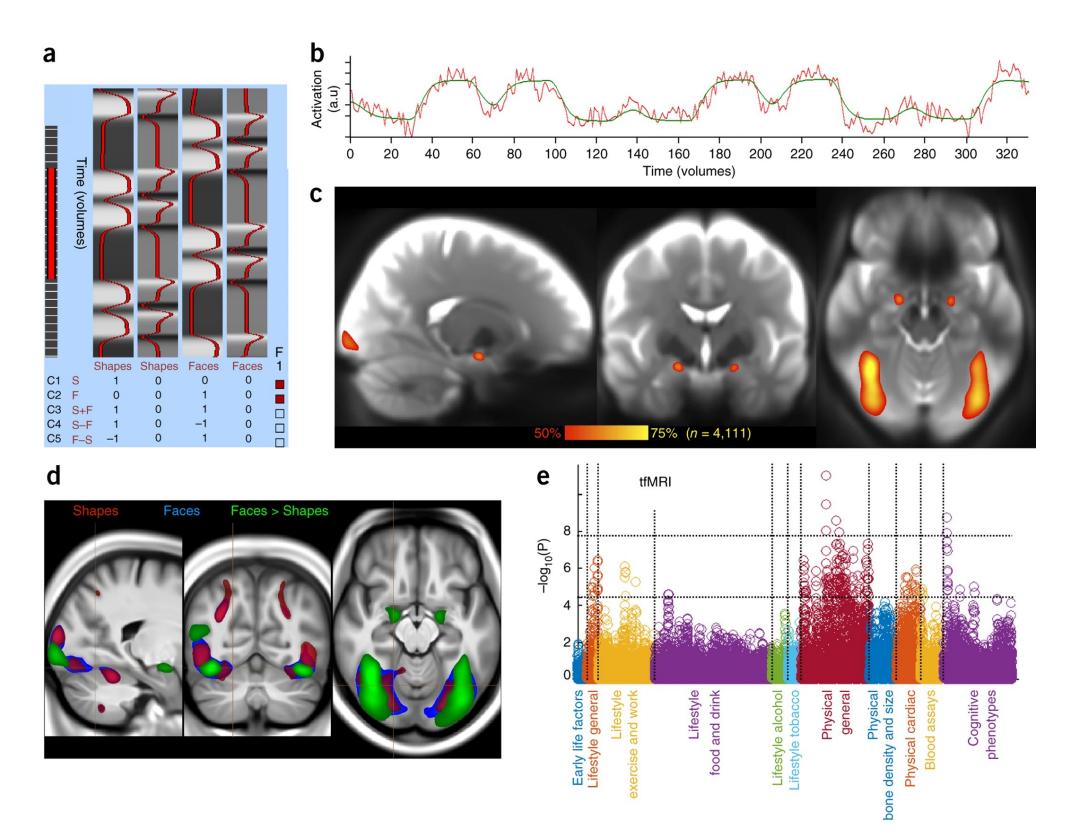
- GE Advance
- HR+
- HRRT
- GE Discovery VCT PET/CT 0
- GE Discovery 690 PET/CT 0



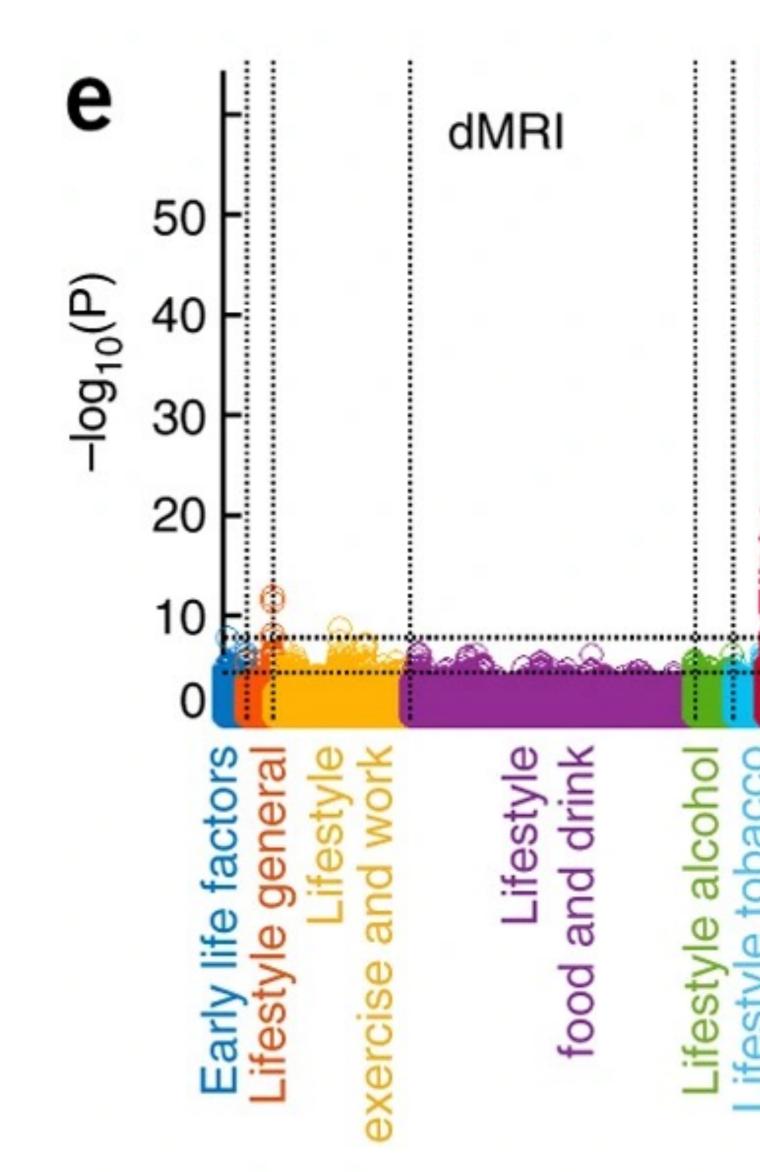
# Solution 3: More is more in the first place



### Human Connectome Project

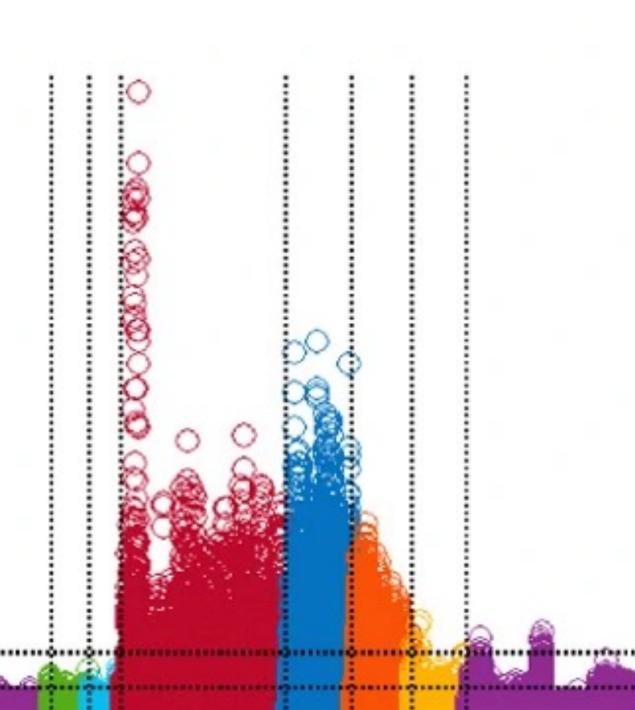


UK Biobank



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Cognitive





# Comparison of the approaches

	<b>Specificity</b>	Price	Computational demands
Meta-analysis	Low-medium	Low	Low
<b>Retrospective reuse</b>	Medium	Medium-high	Moderate
Dedicated large- scale study	High	High-stratospheric	High

