

Meta-analysis techniques

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Why meta-analysis?

“Why Most Published Research Findings Are False” (Ioannidis, 2005):

“There is increasing concern that in modern research, false findings may be the majority or even the vast majority of published research claims.”

“The greater the flexibility in designs, definitions, outcomes, and analytical modes in a scientific field, the less likely the research findings are true.”

Is the last quote especially true for neuroimaging?

Why meta-analysis?

“The Difference Between ‘Significant’ and ‘Not Significant’ is not Itself Statistically Significant” (Gelman and Stern, 2006)

Two apparently conflicting studies—one significant, another not significant—may not necessarily be conflicting. One may simply not have enough power.

Information increase

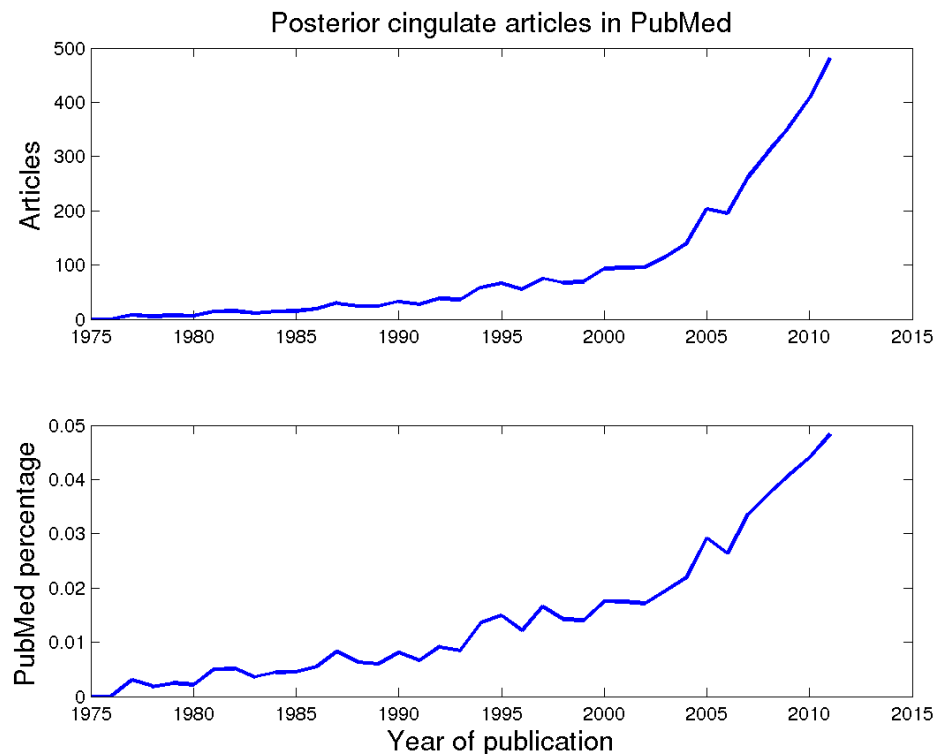


Figure 1: Increase in the number of articles in PubMed which are returned after searching on posterior cingulate and related brain areas.

There are too much data for one person to grasp

The results across experiments are too conflicting

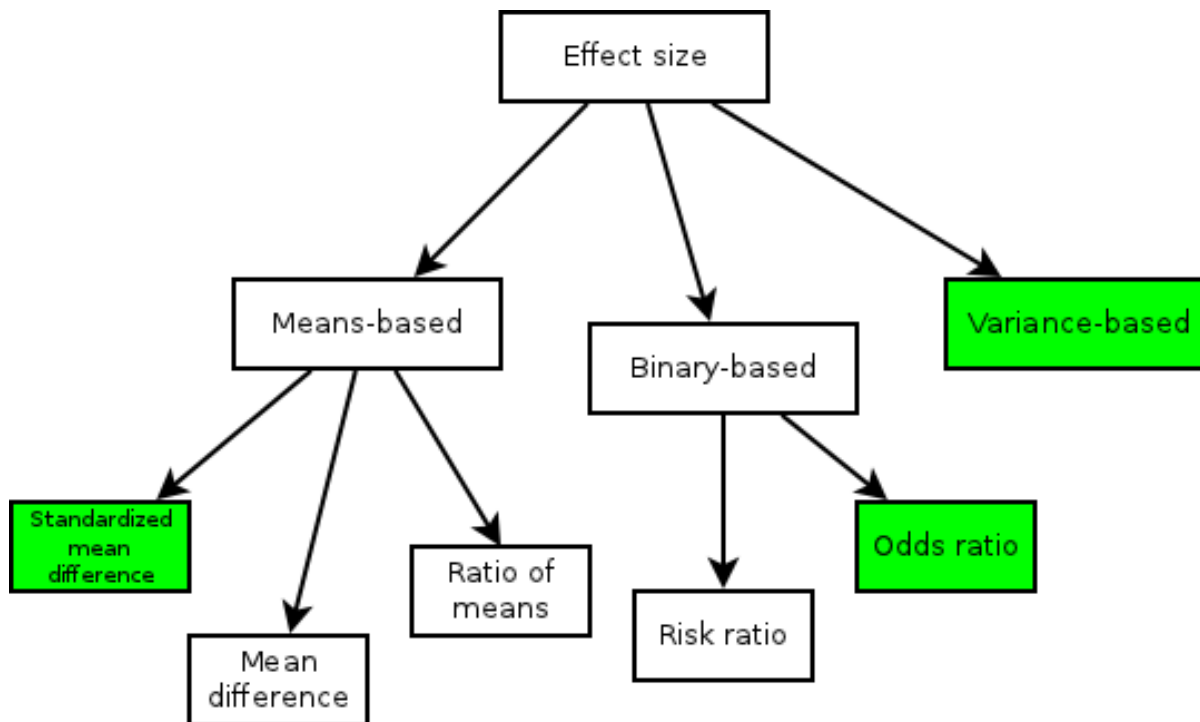
Need for tools that collect data across studies, bring order to data, make search easy and automate analyses to bring out consensus results: **meta-analysis and meta-analytic databases**

Meta-analysis

The page-one definition (Hartung et al., 2008):

The statistical analysis of a large collection of analysis results from individual studies for the purpose of integrating the findings.

Effect sizes



The *effect size* is the central measure in ordinary meta-analysis.

The mean effect size is (usually) independent of the number of subjects.

Effect sizes can be formed from other variables than those shown in the figure: correlation, proportions.

Standardized mean difference (SMD)

For continuous data with (approximately) normal distribution. Example:

Study	Bipolars			Controls		
	N	Mean	SD	N	Mean	SD
Strakowski SM, 1999	24	7.1	1.1	22	6.3	0.8
Altshuler LL, 2000	24	3825.9	695	18	3375	639
...						

Table 1: Data for meta-analysis with SMD. Amygdala volume from bipolar patients and controls.

Take the difference between the means of the two groups (experimentals e and controls c) and divide by the pooled standard deviation

$$g_{\text{smd}} = \frac{\bar{x}_e - \bar{x}_c}{s_{\text{pooled}}}$$

g_{smd} independent of unit of the original study, e.g., whether a brain volume was reported in cubic millimeters or cubic centimeters. it is also independent of the number of subjects (n_e in experimental group)

SDM — details & inference

With, e.g., s_e as the standard deviation for the experimental group:

$$s_{\text{pooled}} = \sqrt{\frac{(n_e - 1)s_e^2 + (n_c - 1)s_c^2}{n_e + n_c - 2}} \quad (1)$$

$$d_{\text{smd}} = E[g_{\text{smd}}] \approx \left(1 - \frac{3}{4(n_e + n_c) - 9}\right) g_{\text{smd}} \quad (2)$$

$$\widehat{\text{Var}}[g_{\text{smd}}] \approx \frac{1}{\tilde{n}} + \frac{g_{\text{smd}}}{2(n_e + n_c - 3.94)} \quad \text{where } \tilde{n} = \frac{n_e n_c}{n_e + n_c}. \quad (3)$$

If the effect size is small ($g_{\text{smd}} \rightarrow 0$) and the two groups are of the same size ($n_e = n_c$) then the variance becomes proportional to the number of subjects in the groups

$$\widehat{\text{Var}}[g_{\text{smd}}] \approx 2/n_e = 2/n_c, \quad (4)$$

i.e., the more subjects the better the effect size is determined.

Odds ratio

For binary data we can construct a contingency table for the results:

	“Success”	“Failure”	total
“Experimentals”	n_{es}	n_{ef}	n_e
“Controls”	n_{cs}	n_{cf}	n_c
Total	n_s	n_f	n

One effect size for binary data is the (natural) logarithm of the odds ratio (Hartung et al., 2008, p. 20)

$$d_{or} = \ln \left[\frac{c(n_{es})/c(n_e - n_{es})}{c(n_{cs})/c(n_c - n_{cs})} \right], \quad \text{where e.g., } c(x) = x + 0.5$$

Addition of 0.5 to get around a problem if there is zero count in any of the cells (Hartung et al., 2008, p. 117)

An estimate of its variance as an estimator is

$$\widehat{\text{Var}}[d_{or}] = \frac{1}{c(n_{es})} + \frac{1}{c(n_e - n_{es})} + \frac{1}{c(n_{cs})} + \frac{1}{c(n_c - n_{cs})}.$$

Variance ratio

Example claim: Men have higher variation in intelligence than women.

We should test this

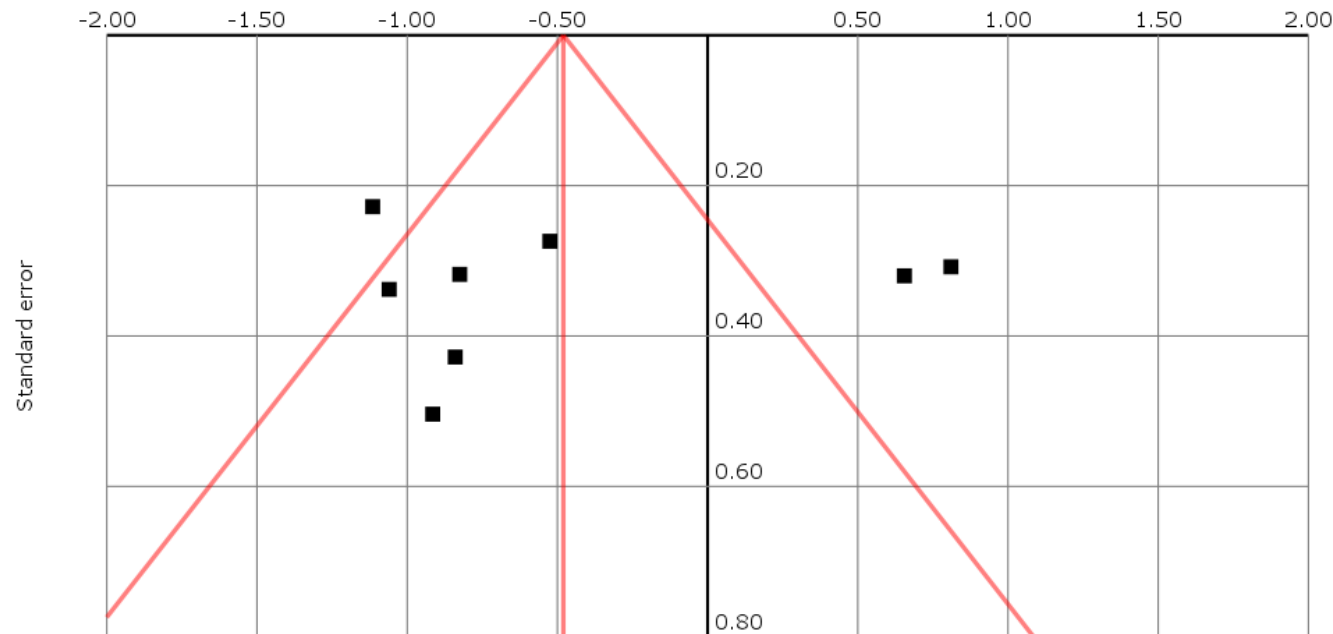
$$\sigma_{\text{men}}^2 > \sigma_{\text{women}}^2$$

The logarithm of the ratio between the two variations (Invr) results in a good statistics (Shaffer, 1992)

$$d_{vr} = \ln \left(\frac{s_e^2}{s_c^2} \right) \quad (5)$$

This is better than the variance ratio s_e^2/s_c^2 (or standard deviation difference $s_e - s_c$)

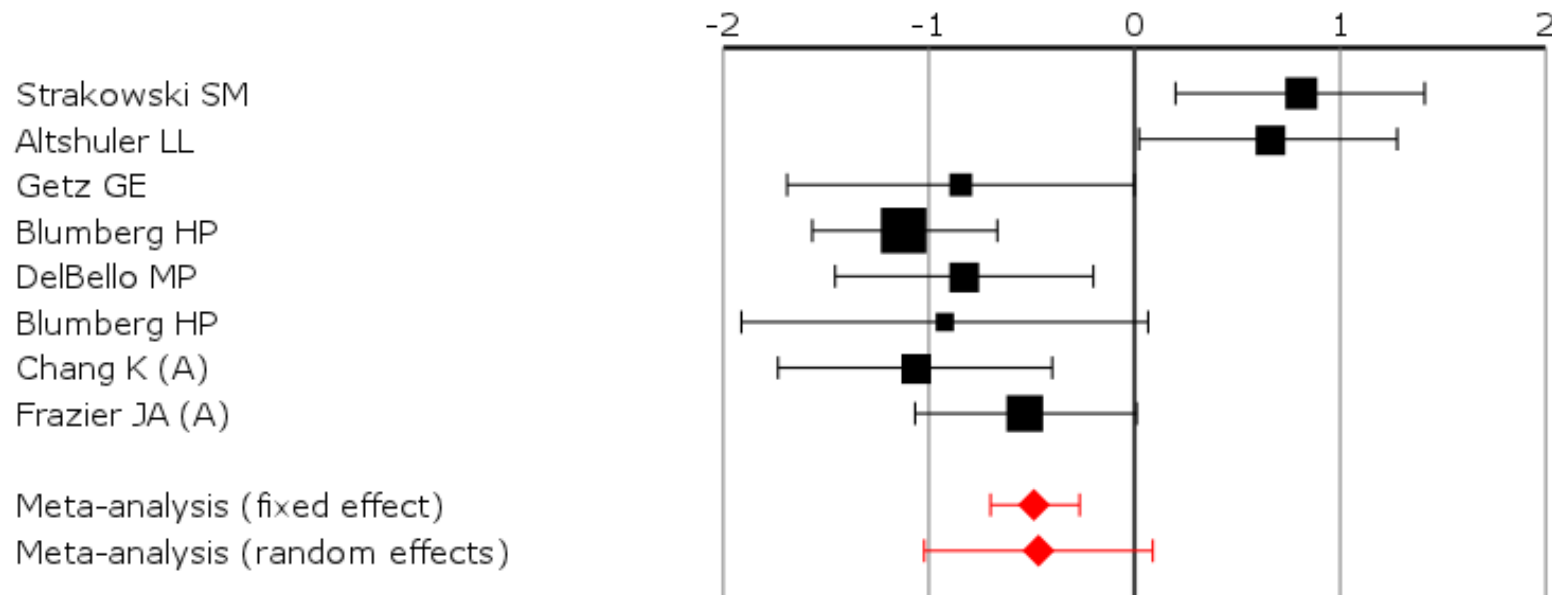
Funnel plot with multiple studies



Scatter plot of the effect sizes and their standard errors (related to variance and sample size) for multiple studies.

May indicate publication bias, if researchers of small studies only publish if they see an effect then the funnel plot becomes asymmetric.

Forrest plot with multiple studies



Forest plot shows the effect size for 8 different studies (Amygdala volume in bipolar disorder).

The squares are the 8 effect sizes (d_i) and the lines indicate 0.05-confidence interval: $d_i \pm 1.96\sqrt{\text{Var}[d_i]}$

Combining effect sizes across studies

Meta-analytic effect size: Inverse variance-weighting (in so-called fixed effect) for weighted averaging of studies (Hartung et al., 2008, p. 36)

$$d_{\text{meta}} = \frac{\sum_i w_i d_i}{\sum_i w_i} \quad (6)$$

where d_i is the effect size for the i th study and the weight for the i th study is determined as the inverse variance

$$w_i = 1 / \text{Var}[d_i] \quad (7)$$

Variance:

$$\text{Var}[d_{\text{meta}}] \approx \frac{1}{\sum_i w_i} \quad (8)$$

when the number of subject for study i increases ($n_i \rightarrow \infty$), then the variance decreases ($\text{Var}[d_i] \rightarrow 0$), the weight increases ($w_i \rightarrow \infty$) and the meta-analytic variance decreases ($\text{Var}[d_{\text{meta}}] \rightarrow 0$)

Random effects analysis

“Random effects” in meta-analysis adds an extra parameter that models the variation between studies.

One approach is the so-called DerSimonian-Laird (Hartung et al., 2008, p. 108)

Free tools for meta-analysis

R with `meta` package by Guido Schwarzer

RevMan and Archie of the Cochrane Library (Elamin et al., 2009)

Open science meta-analysis

Brede Wiki & its meta-analysis service

Brede Wiki for personality genetics

Online data and meta-analyses

Brede Wiki with data

File Edit View History Bookmarks Tools Help

neuro.imm.dtu.dk/wiki/Major_Depressive_Disorder_Neuroimaging_Database_-_Amygdala,_total_-_Statistics

matthew J. kempton

Fnielsen my talk my preferences my watchlist my contributions log out

page discussion edit history delete move protect unwatch

Major Depressive Disorder Neuroimaging Database - Amygdala, total - Statistics

Major Depressive Disorder Neuroimaging Database - Amygdala, total

Major Depressive Disorder Neuroimaging Database - Amygdala, total - Statistics.csv

First Author	Year	Subgroup	Number of Patients	Number of Controls	Patient Mean	Patient SD	Control Mean	Control SD	Patient Age	% Female Patients	Age of Onset	Ham-D rating scale	Antidepressants %	Mood stabilizers %	Antipsychotics %	Drug Free %	Imaging	MRI Field Strength (T)	Slice Thickness (mm)	PMID	
Sheline YI	1998		20	20	3974	562.4946352	3534	560.6719183	54	100.0	5	70.0					MRI	1.5	1.25	9674587	
Bremner JD	2000		16	16	1676	474	1341	449	43	37.5		100.0	0.0	0.0	0.0		MRI	1.5	3	10618023	
Frod T	2003	first episode	30	30	8895	525.738338	9551	541.7137621	40.3	56.7	40	24.8				12.3	MRI	1.5	1.5	12586453	
Frod T	2003	multiple episode	27	27	3542	458.2534233	3556	530.3728877	49.1	48.1	37.4	21.3				12.3	MRI	1.5	1.5	12586453	
Caetano SC	2004		31	31	3.87	0.816259763	4.2	0.78054774	39.2	77.4	27.9	11.8	0.0	0.0	0.0	100.0	MRI	1.5	1.5	15598548	
Hastings RS	2004		10	10	2865	531.5370166	3563	536.2704542	38.9	100.0		23	0.0	0.0	0.0	100.0	MRI	1.5	1.5	14997169	
Lange C	2004		17	17	2.55	0.49	2.26	0.33	34	100.0	29	22	100.0			0.0	MRI	1.5	1.3	15554576	
Xia J	2004		22	13	4477.23	247.7155727	4629.23	84.87261137	39.5	45.5		21.45					MRI	1.5	1.2	15641704	
Rosso IM	2005		20	24	4.62	0.76	5.26	0.735	15.35	85.0	12.8	16.55	0.0	0.0	0.0	100.0	MRI	1.5	1.5	15607296	
Velakoulis D	2006		12	87	3508	593.4409827	3010	461.1333863	22.6	41.7	21.5						MRI	1.5	1.5	16461856	
Werniger G	2006		21	23	2.6	0.569209979	2.3	0.379473319	34	100.0	28	23	100.0		0.0	0.0	MRI	1.5	1.3	16740316	
Caetano SC	2007		19	24	3.02	0.426919196	3.16	0.502891638	13	31.6	10.3	47.4			52.6		MRI	1.5	1	17949901	
Hickie IB (A)	2007		45	16	3.1	0.6	3.4	0.5	52	66.7	36.1	26.8	64.4				MRI	1.5	1.5	16930719	
Munn MA	2007		26	18	28943.6	3425.699339	28577.87	3372.593947	20.54	100.0	15.58						MRI	1.5	1	17511971	
Andreescu C	2008		71	32	0.22	0.04	0.26	0.04	72.2	69.0	52.3	18.3	16.9		1.4		MRI	1.5	1.5	18075490	
Keller J	2008	psychosis	23	11	4.85	0.939627586	5.2	0.854025761	36.5	47.8	27.6	30.5	65.2	17.4	70.0		MRI	3	1.5	18450991	
Keller J	2008	no psychosis	19	11	5.38	0.977189848	5.2	0.854025761	36.6	63.2	27	23.7	57.9	10.5	0.0	42.1		MRI	3	1.5	18450991
MacMaster FP (B)	2008		32	35	3.01	0.598347725	2.72	0.550236313	14.08	62.5	11.77		0.0	0.0	100.0		MRI	1.5	1.5	17640621	
Tamburo RJ	2008		14	11	2728	692.0411837	3100	590.1908166	69.8	35.7		13.8					MRI	1.5	1.5	19085964	
Kronenberg G	2009		24	14	3.45	0.579120022	3.94	0.51232802	54.5	62.5		25.3	0.0	0.0	0.0		MRI	1.5	1.05	19394960	
Lorenzetti V (B)	2009	depressed	29	15.5	3263.63	324.8962542	3206.12	289.1374099	35.52	75.9	21.07					16.1	MRI	1.5	1	19464062	
Lorenzetti V (B)	2009	remitted	27	15.5	3309.7	368.1797727	3206.12	289.1374099	35.07	66.7	26.04					16.1	MRI	1.5	1	19464062	
van Eijndhoven P	2009	depressed	20	10	4747	515.6240879	4375	766.5528031	34.1	65.0	34.1	21.08	0.0				MRI	1.5	1	19028381	
van Eijndhoven P	2009	remitted	20	10	4086	561.6262102	4375	766.5528031	35.8	70.0	33.4	3.4	0.0				MRI	1.5	1	19028381	
Weber K	2009		38	62	1.67	0.25	1.68	0.27	66.11	81.6	37.76		47.4				MRI	3	0.9	20018381	

Download data as CSV | Edit data as CSV | Meta-analysis

Category: Includes CSV

Find: simple Previous Next Highlight all Match case

Store numerical data on a spreadsheet-like page in a MediaWiki. Describe the data in structured format on the wiki.

Data in the Brede Wiki primarily from large meta-analyses by Matthew Kempton, Institute of

Psychiatry, and his coworkers (Kempton et al., 2008; Kempton et al., 2010; Kempton et al., 2011)

Meta-analysis with the Brede Wiki

Major depressive disorder

Major depressive disorder
See also:
[Major Depressive Disorder](#)
[Neuroimaging Database](#)

Papers [edit]

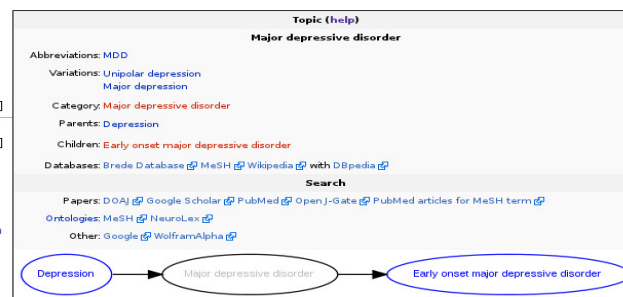
Neuroimaging [edit]

1. Brain volume abnormalities in major depressive disorder: a meta-analysis of magnetic resonance imaging studies
2. Structural neuroimaging studies in major depressive disorder: meta-analysis and comparison with bipolar disorder
3. Hippocampal volume in primary unipolar major depression: a magnetic resonance imaging study
4. Reduced caudate and nucleus accumbens response to rewards in unmedicated individuals with major depressive disorder
5. Regional cerebral blood flow abnormalities in depressed patients with cognitive impairment

Treatment response and genetics [edit]

1. 5-HTT1A, 5-HTT2A, 5-HTT6, TPH1 and TPH2 polymorphisms and major depression
2. Genetic association analysis of serotonin 2A receptor gene (HTR2A) with bipolar disorder and major depressive disorder in the Japanese population

Categories: [Topics](#) | [Topics in Brede Database](#) | [Topics in MeSH](#) | [Topics in Wikipedia](#)



Meta-analyses

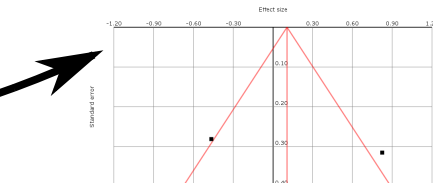
Topic	Data	Raw data	Meta-analysis
Amygdala — Major depressive disorder — MaND	Data	CSV	Meta-analysis
Amygdala — Bipolar disorder — BiND	Data	CSV	Meta-analysis
Amygdala — Obsessive-compulsive disorder — ObND	Data	CSV	Meta-analysis

Study	Experiments	Controls	Effects	Weight	Weighted
Gruber 1993	Mean SD Events N	Mean SD Events N	Effect SE CI		Weighted
Gruber 1993	0.00000 0.00000 0.00000 0.00000	0.00000 0.00000 0.00000 0.00000	0.00000 0.00000 0.00000 0.00000	1.00000	0.00000
Gruber 2003	0.00000 0.00000 0.00000 0.00000	0.00000 0.00000 0.00000 0.00000	0.00000 0.00000 0.00000 0.00000	1.00000	0.00000
Meta-analysis (fixed effects)			0.00000 0.00000 0.00000 0.00000	100%	0.00000
Meta-analysis (random effects, DGL)			0.00000 0.00000 0.00000 0.00000	100%	0.00000

Forest plot



Funnel plot



Obsessive-compulsive disorder Neuroimaging Database - Amygdala

ObND for amygdala

Obsessive-compulsive disorder Neuroimaging Database - Amygdala.csv

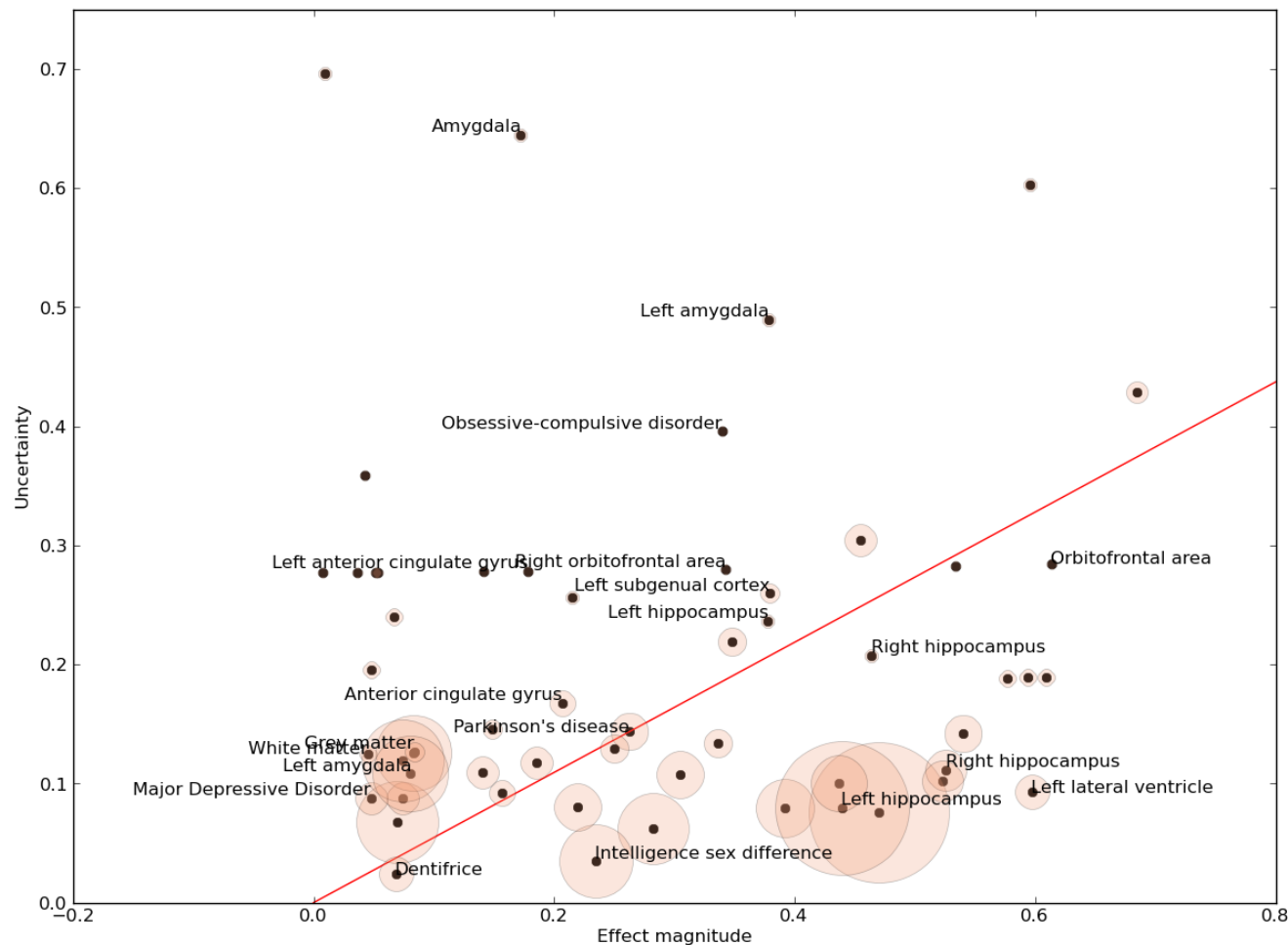
Author	Year	Region	Patient mean	Patient SD	Number of patients	Controls mean	Controls SD	Number of controls	Patient Y-BOCS mean	Patient Y-BOCS SD	Controls Y-BOCS mean	Controls Y-BOCS SD	Patient mean age	Patient SD age	Controls mean age	Controls SD age	Notes	PDB	Title
Szeisler	1999	Amygdala	1.80	0.3115	22	1.25	0.2851	22					32.2	8	29.8	6.9		10530639	Orbital frontal and amygdala volume reductions in obsessive-compulsive disorder
Kwon	2003	Amygdala	1.50	0.3115	22	1.25	0.2851	22					26.7	7.2	26.2	6.1	Values computed from left/right	12810750	Similarity and disparity of obsessive-compulsive disorder and schizophrenia in MR volumetric abnormalities of the hippocampus-amygdala complex

Download data as CSV | Edit data as CSV | Meta-analysis

Mass meta-analysis

With numerical data and information about it in the wiki it is possible to download and analyze all meta-analysis together.

Here a L'Abbé-like plot of many of the meta-analyses in the Brede Wiki with effect magnitude on the x-axis and its uncertainty on the y-axis.



Brede Wiki for personality genetics

	Effect	Std	P	Studies	Subjects	Gene	Polymorphism	Trait
1	0.854	0.223	0.00013	2	107	ESR1	TA repeat	Harm avoidance
2	-1.102	0.289	0.00014	2	245	HTR3A	C178T	Harm avoidance
3	-0.779	0.220	0.00039	1	90	ESR1	TA repeat	Anxiety
4	-0.445	0.135	0.00098	1	247	TH	TCAT repeat	Extraversion
5	-0.401	0.123	0.00108	1	315	DRD4	Exon 3 VNTR	Positive emotions
6	0.165	0.051	0.00118	13	1747	MAOA	uVNTR	Reward dependence
7	-0.393	0.123	0.00135	1	315	DRD4	Exon 3 VNTR	Extraversion
8	-1.355	0.427	0.00152	1	125	HTR3A	C178T	Nonconformity
9	-0.758	0.240	0.00161	1	122	SLC6A4	5-HTTLPR	Activity
10	-0.174	0.055	0.00163	16	1791	SLC6A4	5-HTTLPR	Agreeableness

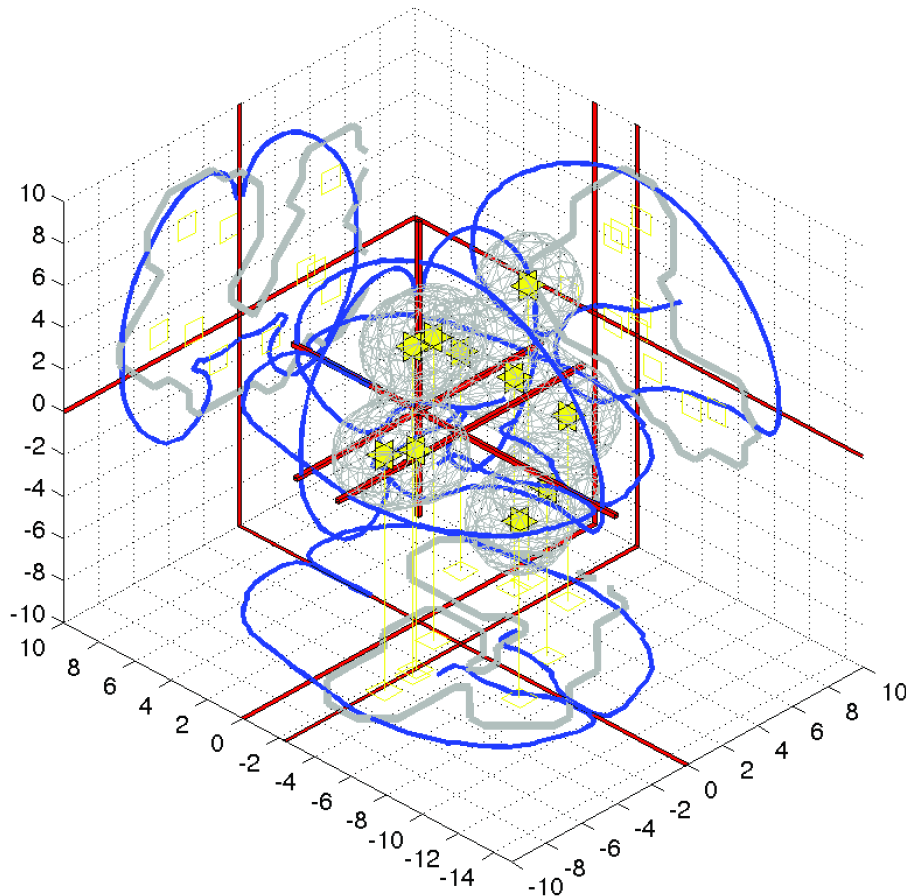
Meta-analysis across traits and polymorphisms

Large-scale data mining across all recorded personality traits and genetic polymorphisms and present the result on the wiki.

Order meta-analytic results, e.g., with respect to P -value

Neuroimaging meta-analysis

Image-based meta-analysis if you got the summary images (Salimi-Khorshidi et al., 2009).

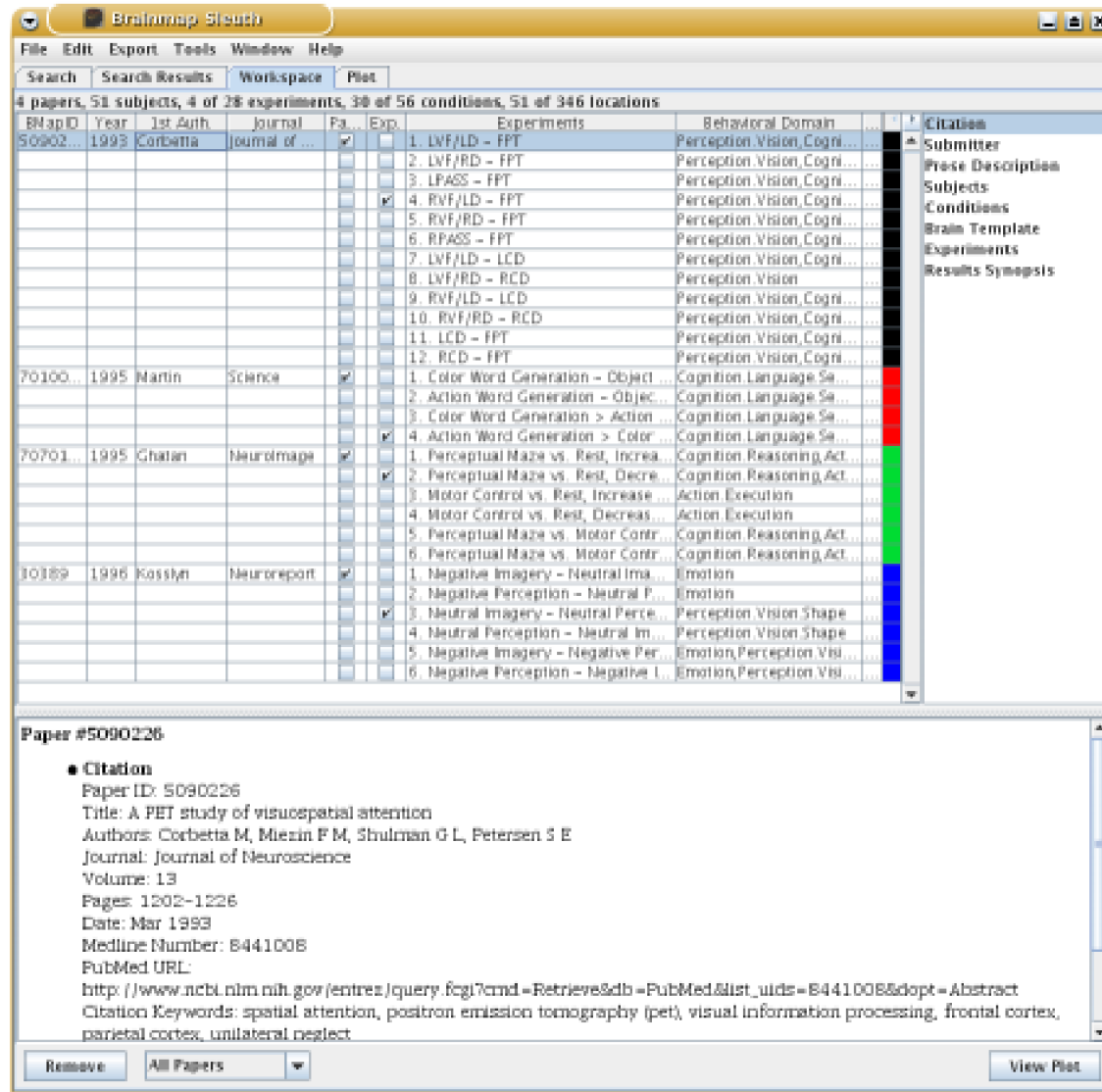


Coordinate-based meta-analysis if you got the stereotaxic coordinates (Fox et al., 1997; Nielsen and Hansen, 2002; Turkeltaub et al., 2001).

Convolve a smooth kernel on its stereotaxic coordinate

Tools: BrainMap's GingerALE, Brede Toolbox

BrainMap

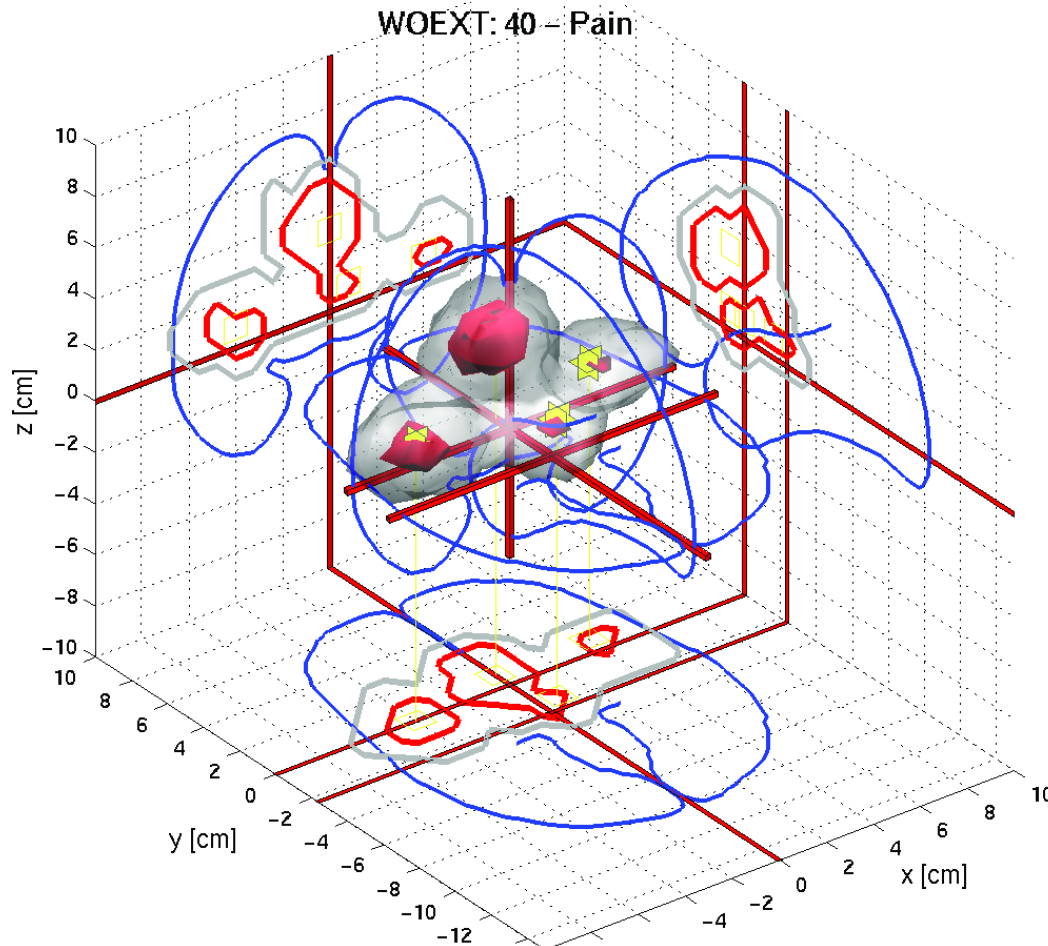


One of the first and most comprehensive databases (Fox et al., 1994; Fox and Lancaster, 2002)

Presently 85007 locations from 2238 papers (2012 September)

Graphical Internet-based interface in Java, *sleuth*, with search facilities, e.g., on author, 3D coordinate, an others and *GingerALE* meta-analysis

Example on meta-analysis



Coordinate-based meta-analysis with the Brede Toolbox on pain studies

Volume threshold at statistical values determined by re-sampling statistics (Nielsen, 2005).

Red areas are the most significant areas: Anterior cingulate, anterior insula, thalamus. In agreement with “human” reviewer (Ingvar, 1999).

Semantic MediaWiki for databasing papers

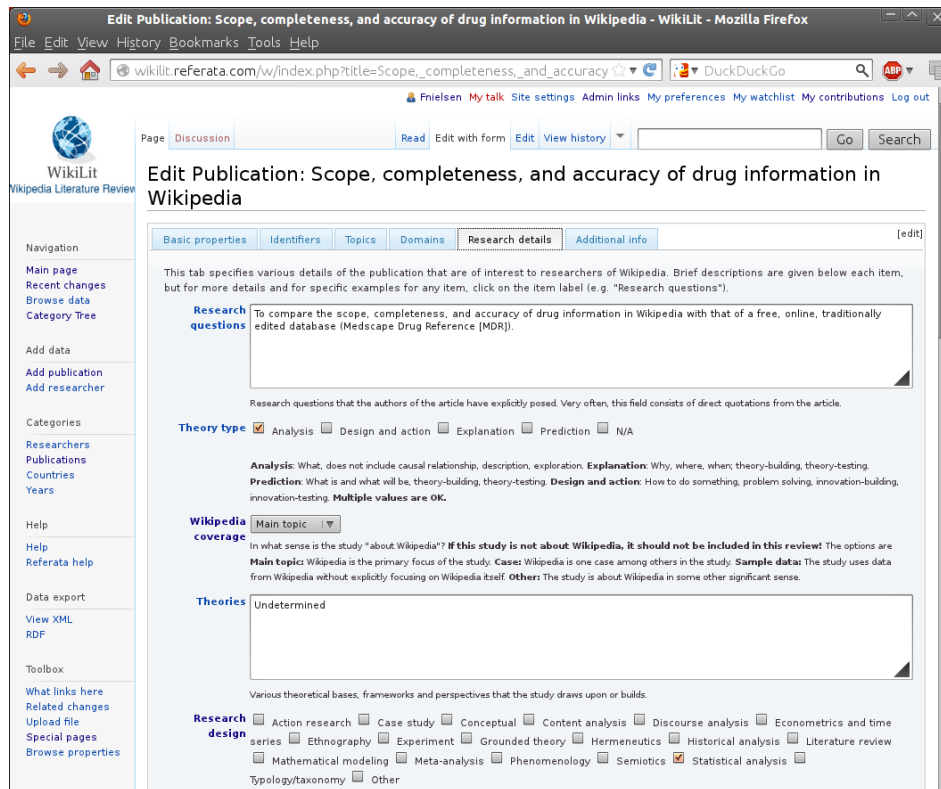


The Semantic MediaWiki allows you to construct a database on the Web without having to setup a standard database management system :-)

Semantic MediaWiki has a flexible schema: You can add fields after you have setup “table”, e.g., add a “is peer reviewed” field or “imaging modality” field. :-)

A full setup of a Semantic MediaWiki with forms and templates requires mastering of the somewhat obscure MediaWiki template language :-)

Semantic MediaWiki



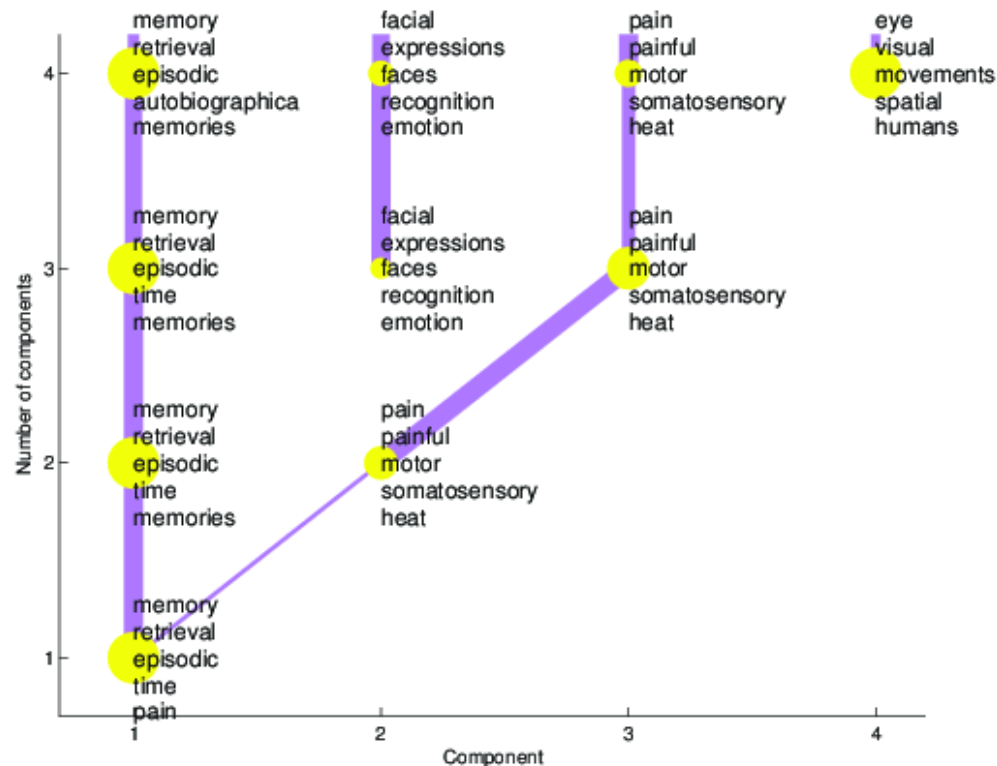
Our Semantic MediaWiki instance
<http://wikilit.referata.com/> setup
for a systematic review

Semantic MediaWiki allows you to
define forms for input: text and cat-
egorical.

Semantic queries can be made so
you can get the inputted data
in comma-separated values suitable
for further numerical processing

... and you can filter with the queries, e.g., peer-reviewed publications if
that category is setup

Automated literature reviews with text mining

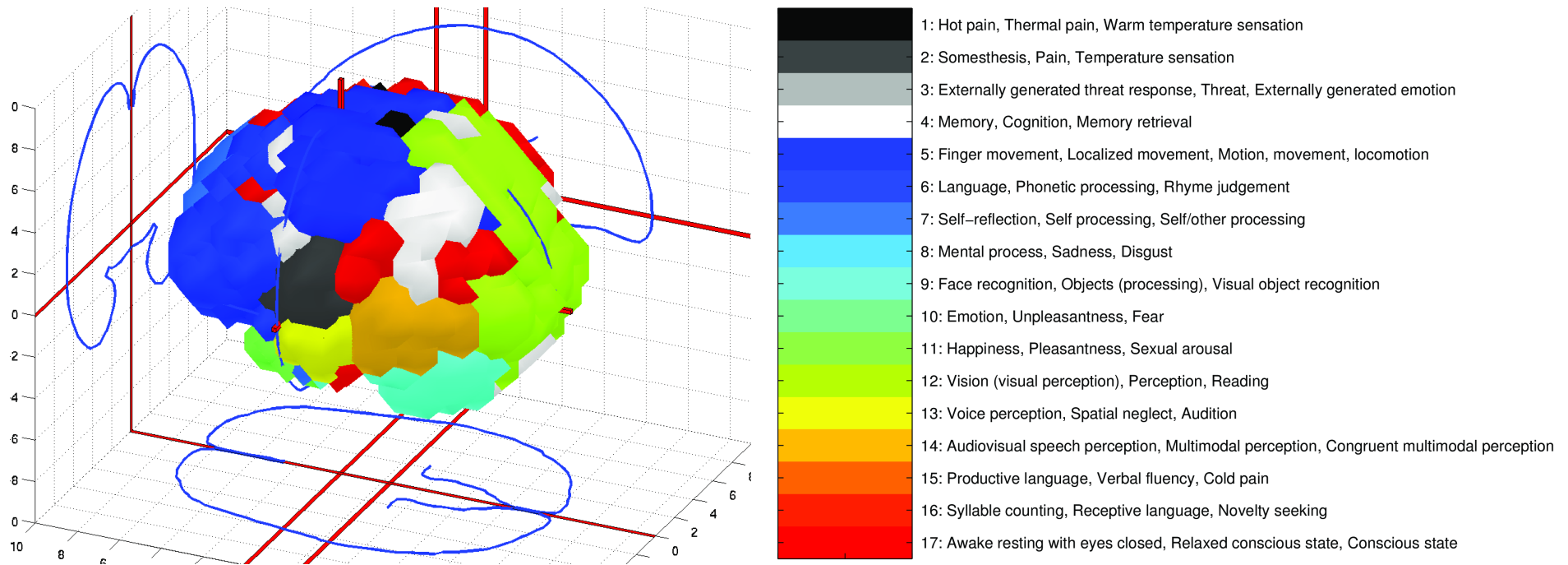


Download papers, extract words and represent them in a bag-of-words matrix, perform topic mining with an unsupervised multivariate analysis method, e.g., non-negative matrix factorization, to find themes (Nielsen et al., 2005).

For a recent example with NeuroSynth see (Poldrack et al., 2012)

Figure 2: Some of the topics found in a corpus on posterior cingulate neuroimaging.

Combining ontologies and coordinates



Combining the cognitive ontology in Brede with the coordinate-based meta-analysis to constructed a functional atlas

What can you do?

Report the summary statistics: mean, standard deviation and number of subjects.

Report summary statistics for all groups

Open Science

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