

Neuroinformatics techniques for provenance & data sharing

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Neuroinformatic techniques for provenance & data sharing

Camille Maumet

GlaxoSmithKline - Neurophysics Workshop on Skeptical Neuroimaging

January 14th, 2014



Outline

- 1. Data sharing: current practice in neuroimaging
- 2. How to become less skeptical?
- 3. Neuroinformatics techniques for provenance and data sharing

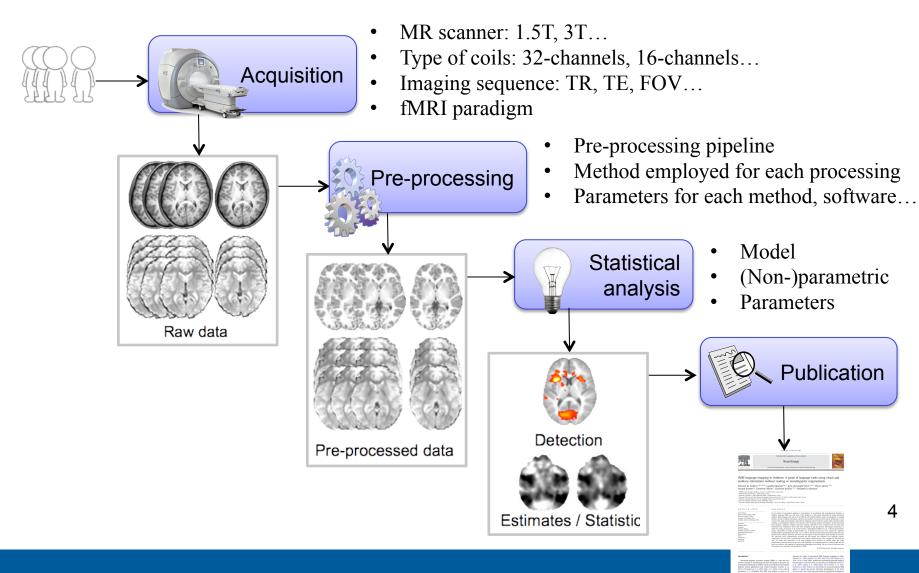


Outline

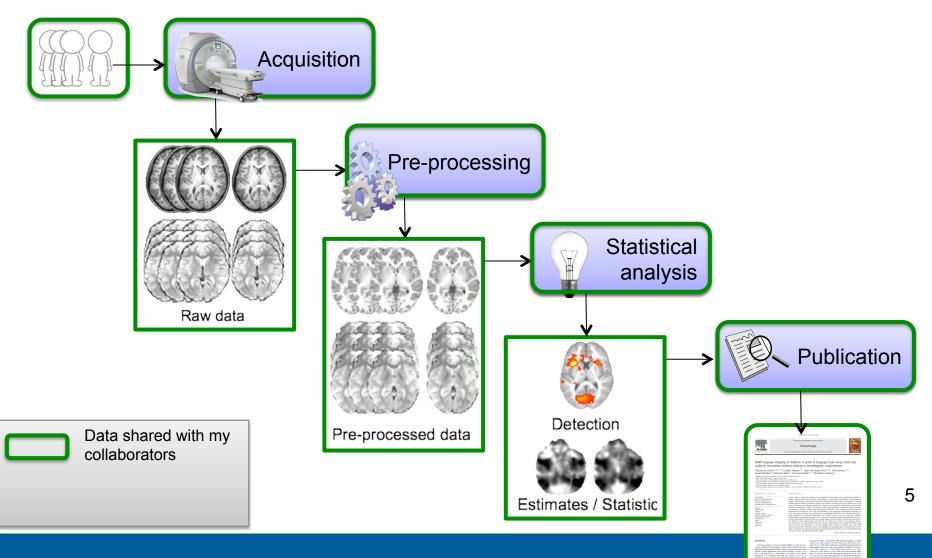
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Overview of a neuroimaging study

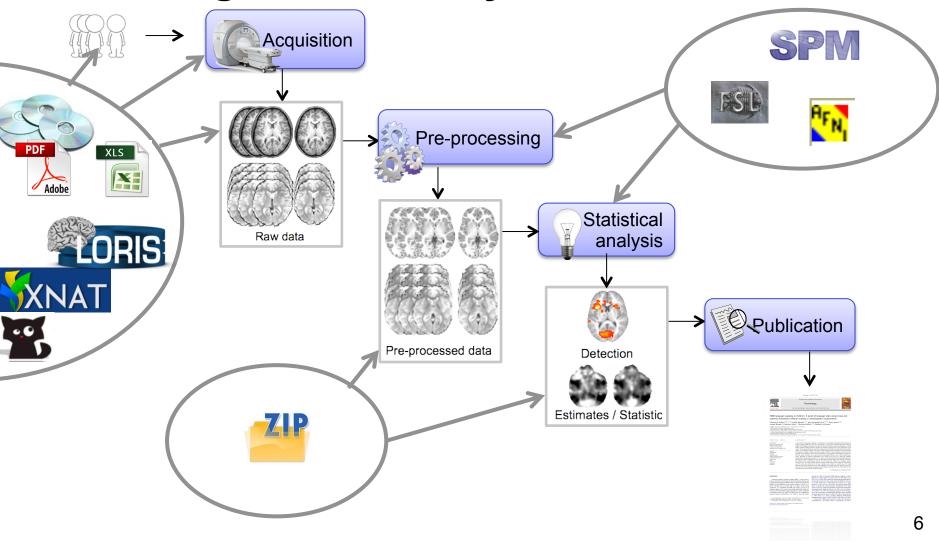


Neuroimaging and data sharing



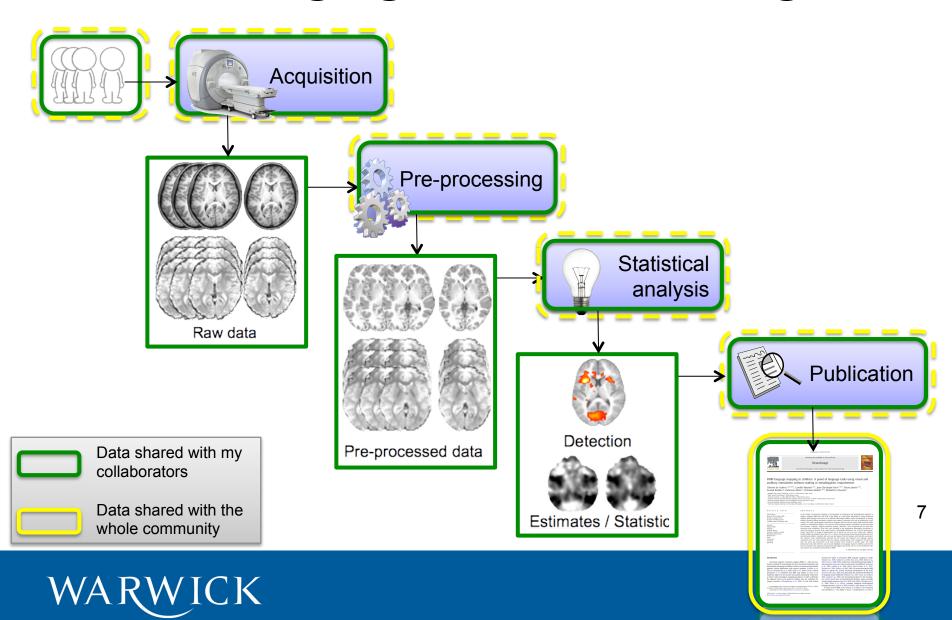


Sharing data with my collaborators





Neuroimaging and data sharing



A neuroimaging publication

Methods section: metadata in free-form text.

General technical implementation

A single scanner session included the four paradigms separately implemented with the same parameters; a simple block design alternated a rest condition as control and the language task, starting with rest, with a preliminary period of signal acquisition for MRI signal stabilization which was later discarded during data processing. Each paradigm included three 27-s blocks of each condition and had a total duration of 2 min 48 s. The scanner session, including the anatomical acquisition and the four language paradigms, had a duration of about 30-35 min. All subjects performed the tasks in the same order, as during the preparation step, in order to avoid the mix of auditory and visual tasks and the resulting complication for the child. Words required by the tasks were one-to-three-syllable words highly frequent in the lexicon of French 8 years old children (Lambert and Chesnet, 2001).

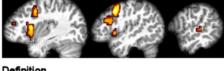
During the rest condition, a red cross was displayed on the projection screen and children were asked "not to work" to "think

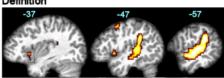
	Auditory langua	
	Categ>Def	
Left Hemisphere		
Inf frontal-Oper		
Precentral	18-3.38 ⁽⁵⁾	
Mid frontal	33-3.66	
SMA		
Cingulate		
Med sup frontal	174-4.69	
Rol operculum		
Insula		
Sup temporal		
Mid temporal		
Inf parietal		
Sup parietal		
Postcentral		
Sun occinital		

Table

Results section:







2D plot(s) of the detections

General technical implementation

A single scanner session included the four paradigms separately implemented with the same parameters: a simple block design alternated a rest condition as control and the language task, starting with rest, with a preliminary period of signal acquisition for MRI signal stabilization which was later discarded during data processing. Each paradigm included three 27-s blocks of each condition and had a total duration of 2 min 48 s. The scanner session, including the anatomical acquisition and the four language paradigms, had a duration of about 30-35 min. All subjects performed the tasks in the same order, as during the preparation step, in order to avoid the mix of auditory and visual tasks and the resulting complication for the child. Words required by the tasks were one-to-three-syllable words highly frequent in the lexicon of French 8 years old children (Lambert and Chesnet, 2001).

During the rest condition, a red cross was displayed on the projection screen and children were asked "not to work", to "think about nothing" and, because of the complexity of this instruction, to listen to the noise of the scanner and fix attention on the red cross

Description of the detections

	Auditory language tasks			Visual language tasks	
	Categ>Def	Def>Categ	Categ C Def	Ph-s>Ph-d	Ph-
Left Hemisphere					
Inf frontal-Oper			348-4.10(4)		825
Precentral	18-3.38 ⁽⁵⁾		348-5.09		825
Mid frontal	33-3.66				
SMA			1433-5.48		357
Cingulate			1433-5.08(3)		
Med sup frontal	174-4.69				
Rol operculum				36 - 4.31	
Insula			396-4.87(8)		58-
Sup temporal			351-3.81(1)		91-
Mid temporal		1658-4.67 ⁽³⁾	351-5.61 ⁽²⁾		10-
Inf parietal		1658-5.18 ⁽⁶⁾			
Sup parietal					97€
Postcentral					976
Sup occipital					
Mid occipital				146-4.43	146
Inf occipital					146
Fusiform				207 5 44	1.46

Task comparisons (>) and conjunctions (C). Peak locations, cluster extent-Z-score (p<0.001 unc.; k=10)

Table of local maxima



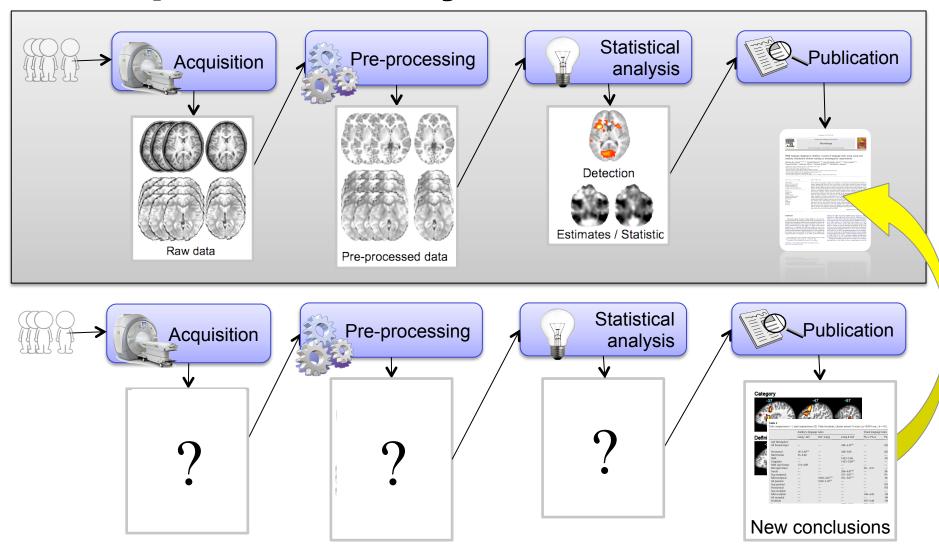


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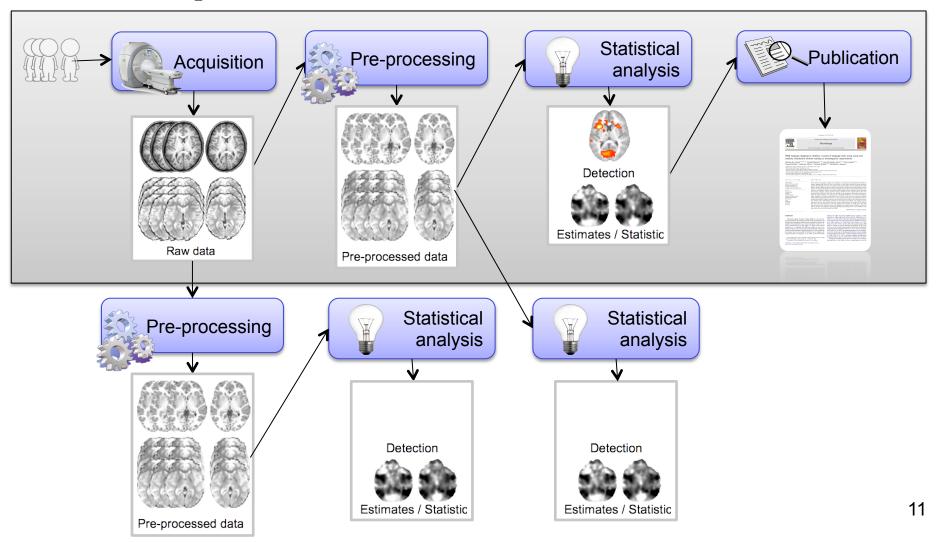


Reproducibility





Full provenance





Meta-analysis: analyzing the analyses

Coordinate-Based Meta-Analysis (CBMA)



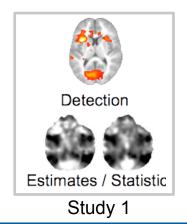


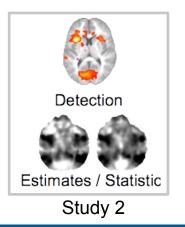


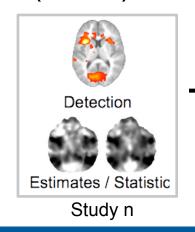


New results!

Image-Based Meta-Analysis (IBMA).









New results!



How to become less skeptical?

- Reproducibility
 - Confirm results by re-running an analysis
- Provenance
 - Needed for reproducibility
 - Avoid selection bias.
- Meta-analysis
 - Strengthen results by combining studies.
- What do we need?
 - Sharing data, meta-data and provenance.



Data sharing: obstacles

- Psychological
 - "My" data
- Ethical constraints
- Technical: difficulties to share data with enough metadata to be really useful
 - Available data versus usable data.

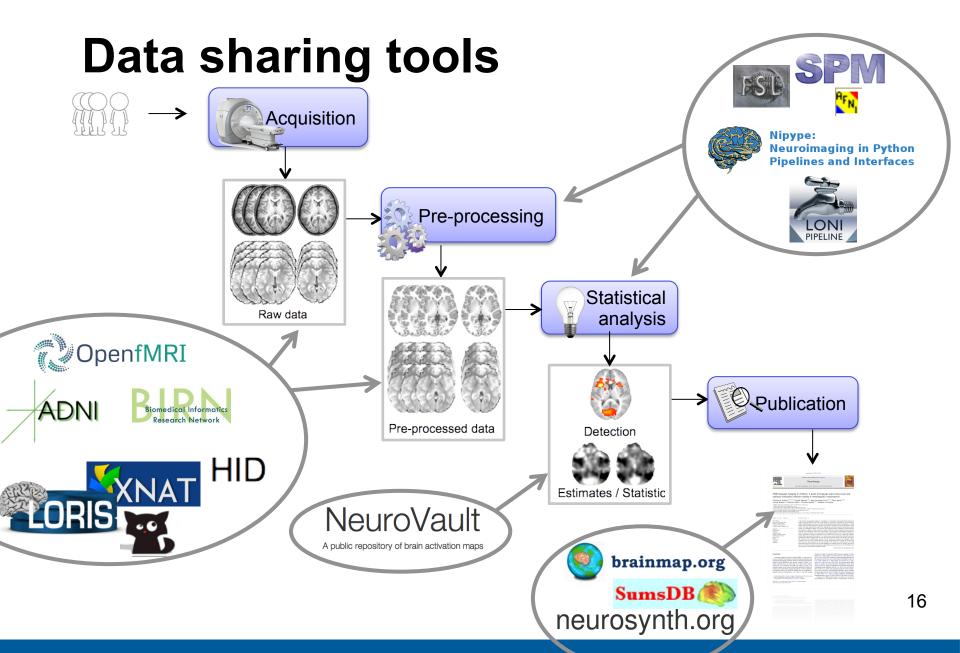
"Less than a few percents of acquired neuroimaging data is available in public repositories" [Poline 2012]



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A standard format for meta-data

- Sharing data across the data sharing tools...
- First attempt of an agnostic format: XML-Based
 Clinical Experiment Data Exchange Schema
 (XCEDE): www.xcede.org
 - Describes subject, study, activation
 - Limited provenance encoding
 - Initiative of the BIRN
- Neurolmaging Data Model NI-DM:

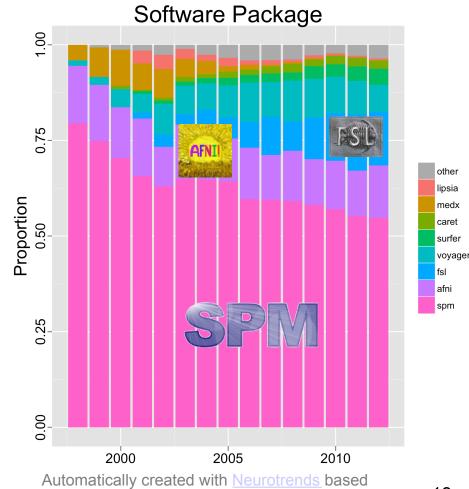
www.nidm.nidash.org

- Based on web-semantic tools.
- Initiative of the BIRN and INCF



Three major players

- Bottom-up approach.
- Lean on existing analysis software (SPM, FSL, AFNI) to disseminate the standard.



Automatically created with <u>Neurotrends</u> based on over 16 000 journal articles

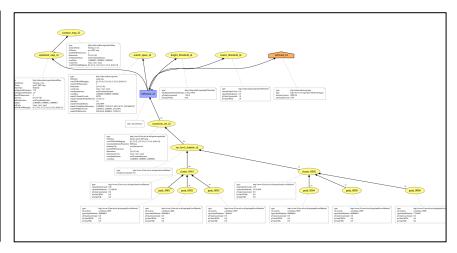




Work in progress

 Define a format to represent the results of a neuroimaging study with a focus on meta-analysis.

Term name	Definition	Example	BIRNLex or NIDM Concept ID
BonferroniCorrection	Bonferroni correction for multiple statistical tests		nidm:nidm_80
chi-squareStatistic	A statistical parameter drawn from a chi-square statistic		nidm:nidm_81
FDR	False Discovery Rate correction		nidm:nidm_82
FWER	Family-wise Error Rate correction		nidm:nidm_83
Scan Image	An image that is the ouput of an MRI or CT or PET scan		nidm:nidm_84
SliceOrder	The temporal order in which the 2D slices were acquired by the imaging systems		nidm:nidm_85
Voxel	volumetric pixel		nidm:nidm_86
Z-Statistic	A statistical parameter drawn from a normal or z distribution		nidm:nidm_87
extentThresh	Minimum cluster size used when thresholding a statistic image	5voxels	nidm:nidm_88
errorDegreesOfFreedom	Degrees of freedom of the error.	73	nidm:nidm_89
effectDegreesOfFreedom	Degrees of freedom of the effect.	1	nidm:nidm_90
StatisticMap	A map (2D or 3D structured dataset) whose value at each location is a statistic.		nidm:nidm_91
voxelSize	3D size of a voxel measured in voxelUnits.	[2 2 4]	nidm:nidm 92
cluster	A group of neighboring image elements (voxels or vertices)		nidm:nidm_93
qValueFDR	p-value adjusted for the search volume, controlling for the False Discovery Rate	0.000154	nidm:nidm_94
pValueFWE	p-value adjusted for the search volume, controlling for the Famillywise Error Rate	0.00554	nidm:nidm_95
pValueUncorrected	Uncorrected p-value	0.0542	nidm:nidm_96



Vocabulary

Data model



Neuroimaging terms

Define a vocabulary to support the format.

Term name	Definition	Example	BIRNLex or NIDM Concept ID	synonyms and related urls	Parent term
cluster	A group of neighboring image elements (voxels or vertices)		nidm:nidm_93	http://ncicb.nci.nih.gov/xml/owl/EVS/Thesaurus.owl#C4 or http://purl.obolibrary.org/obo/OBI_0000251	3
qValueFDR	p-value adjusted for the search volume, controlling for the False Discovery Rate	0.000154	nidm:nidm_94	http://purl.obolibrary.org/obo/OBI_0001442	p-value i.e. nidm:nidm_0011
3 4 1 203 4 200	p-value adjusted for the search volume, controlling	0.00554	nidas of	http://p.ud.phalibeau.com/pha/QDI_0004005	a color i a midrocoldes 0044
pValueUncorrected	Uncorrected p-value	0.0542	nidm:nidm_96	http://purl.obolib p-value i.e.	nidm:nidm_0011
SelectionProcedure	reported		nidm:nidm_97		
clusterSizeInVoxels	Number of voxels contained in a cluster.	18 nidm:nidm_98			
softwareVersion	Name and Number specifying software version.	SPM99, SPM2, SPM5, SPM8, SPM12b, FSL5.0.0	nidm:nidm_99		nidm:Software
softwareRevision	Software revision number.	v5417	nidm:nidm_100		
clusterSizeInVertices	Number of vertices contained in a cluster.	10 nidm:nidm_101			
clusterSizeInResels	Number of resels contained in a cluster.	13	nidm:nidm_102		
voxelUnits	Units associated to each dimensions of some N- dimentional data.	{'mm' 'mm' 's'}	nidm:nidm_103		
ReselSizeInWorldUnits	Volume of a resel, a resolution element, expressed in units. It expresses the smoothness of the noise, with smoother images having larger resels.		nidm:nidm_104	http://en.wikipedia.org/wiki/Resel	
			nidm:nidm_105		
Мар	2D or 3D structured dataset.		nidm:nidm_106		
fileName	Name associated with a file (without path).		nidm:nidm_107		
numberOfDimensions	Number of Dimensions of some N-dimentional data.	3	nidm:nidm_108		
dimensions	Dimensions of some N-dimentional data.	[64 64 20]	nidm:nidm_109		
coordinateSystem	Type of coordinate system.	nidm:mniCoordinateSystem	nidm:nidm_110		
searchVolumeInVoxels	Total number of voxels within the search volume.	68656	nidm:nidm_111	Synonyms of nidm:volumeInVoxels	
searchVolumeInResels	Total number of resels within the search volume.	151.3	nidm:nidm_112	Synonyms of nidm:volumeInResels	

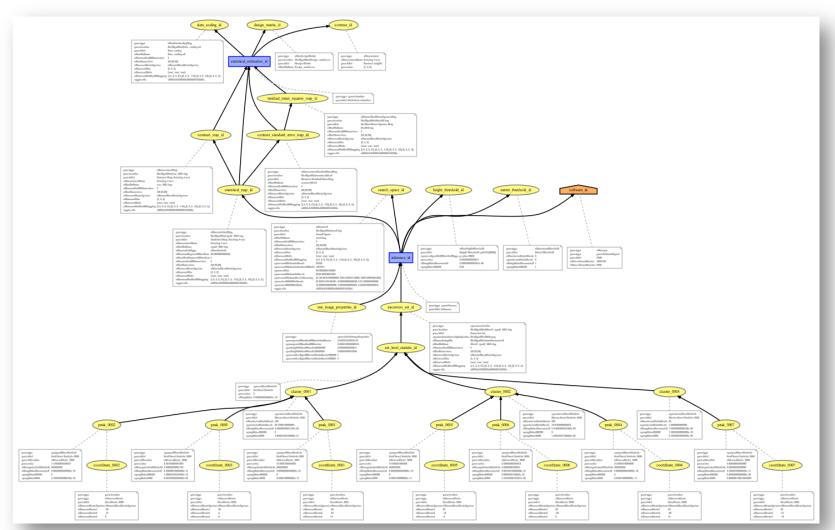


Data model

Based on PROV-DM a W3C recommendation to encode wasDerivedFrom provenance. www.w3.org/TR/prov-dm/ **Entity** wasAttributedTo wasGeneratedBy **Agent** used wasAssociatedWith **Activity**



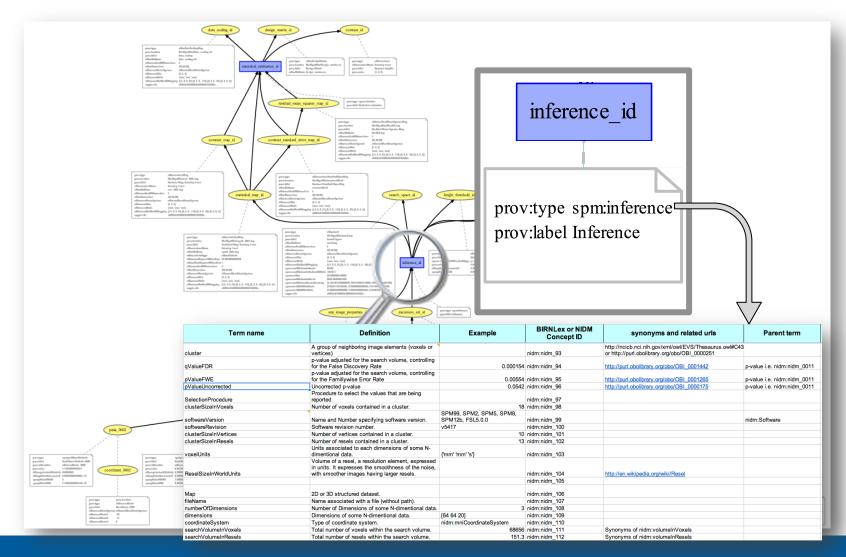
Data model





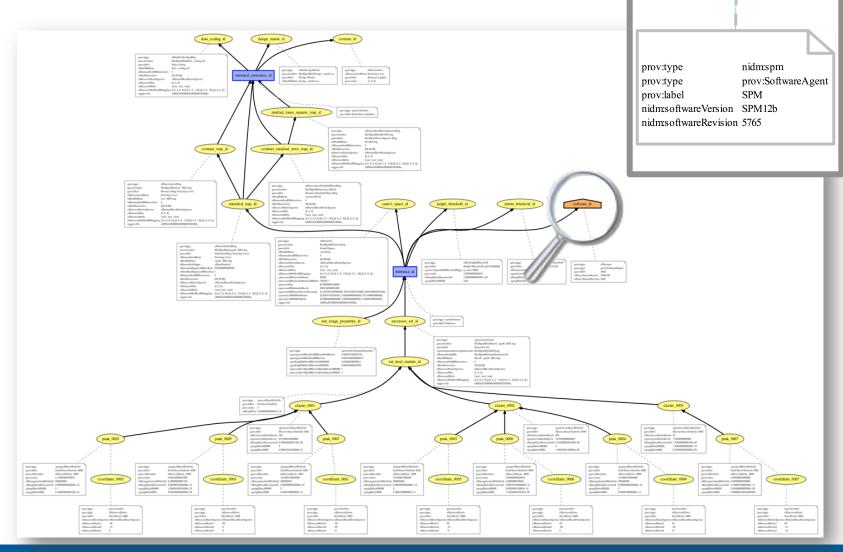
22

Data model: activities





Data model: agent

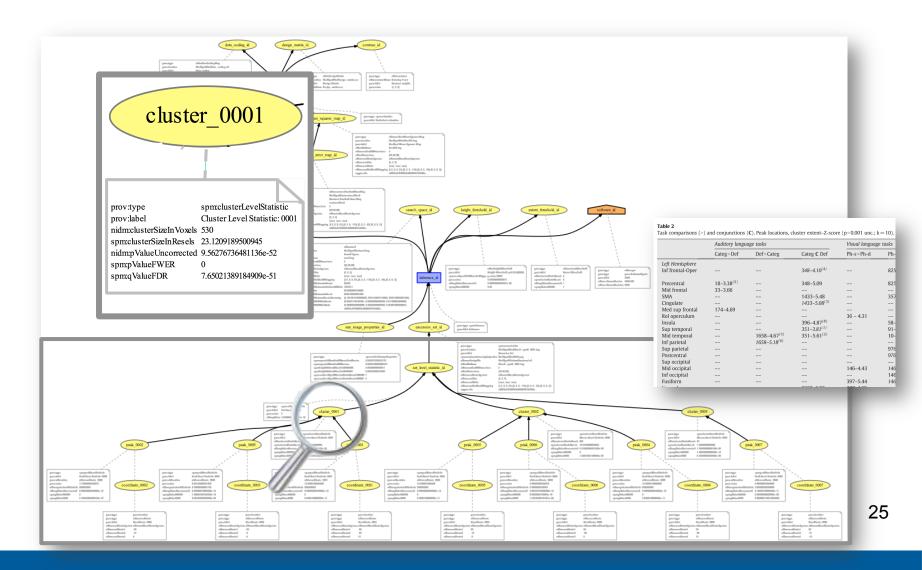




24

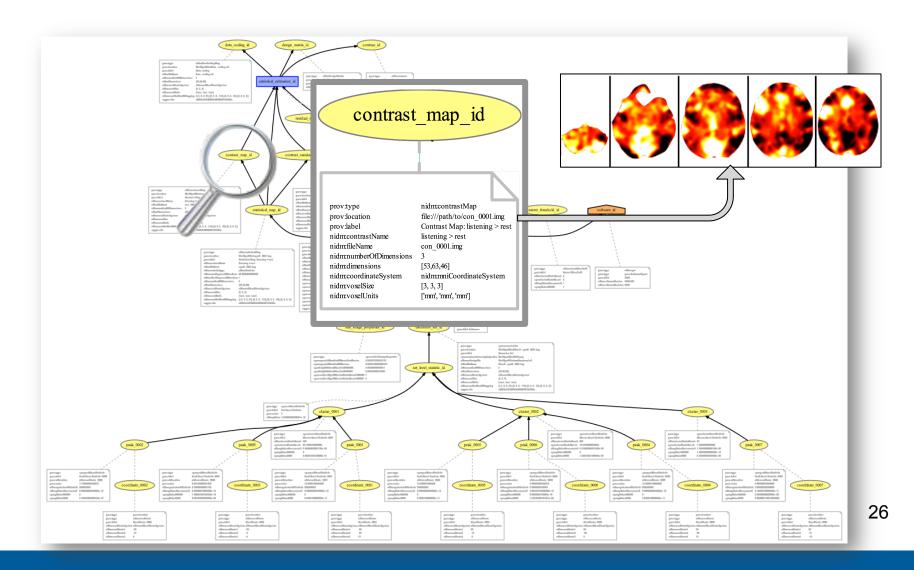
software_id

Data model: entities





Data model: entities





Conclusion

- Data sharing is one key to reduce skepticism.
- There is already a number of technical solutions for data sharing in neuroimaging.
- A meta-data standard would beneficiate to all of these efforts
 - NI-DM: http://nidm.nidash.org



Q & A

This work is supported by the **Welcome**trust

