

Clustering Independent Components of EEG Data



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S. Makeig 2016

Why cluster independent components across subjects or sessions?

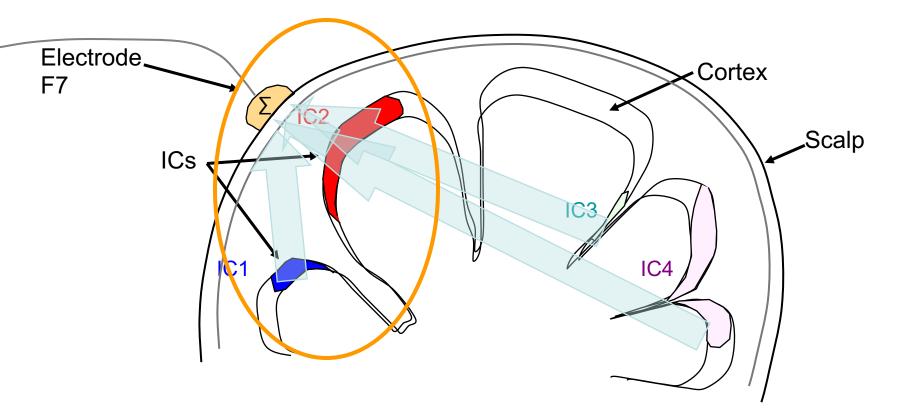


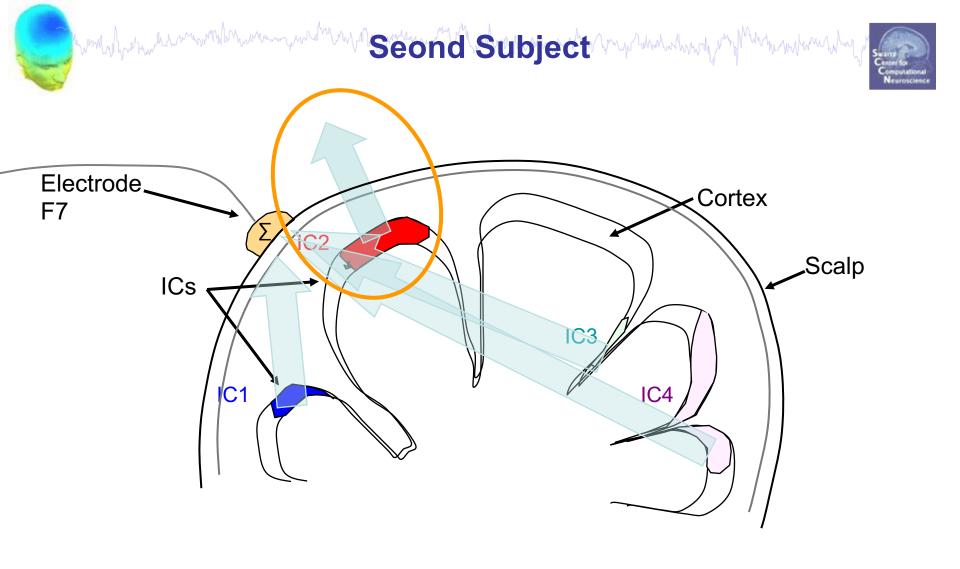
- ICA transforms the data from a channel basis
 - (activity recorded at each channel)
 - to a component basis (activity computed at each IC).
- Normally, EEG researchers assume that, for example, electrode channel F7 == F7 == F7 ... in each subject – and then 'cluster' their data assuming channel equivalence.
- This amounts to the simple assumption

"Your Cz is My Cz!"

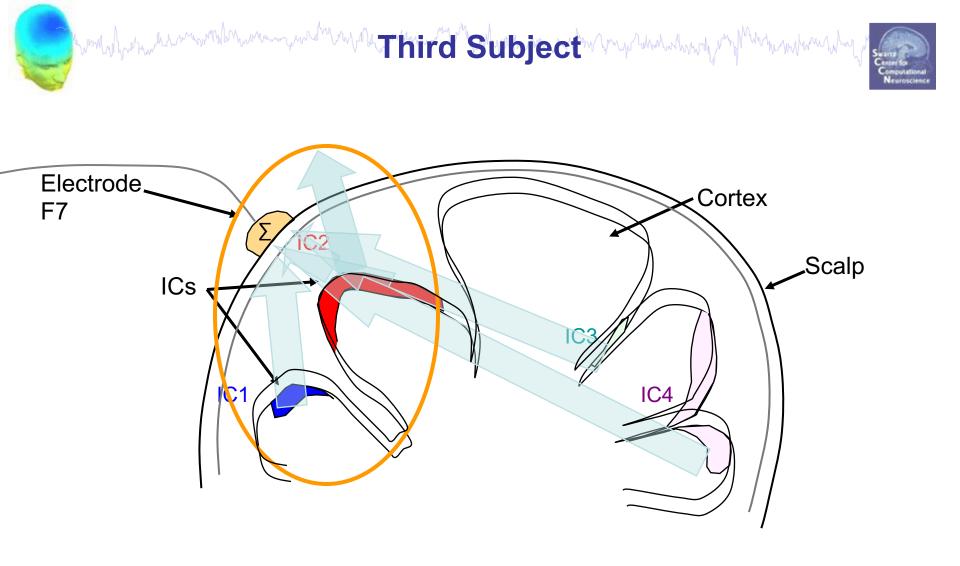
But this is only *roughly* correct !



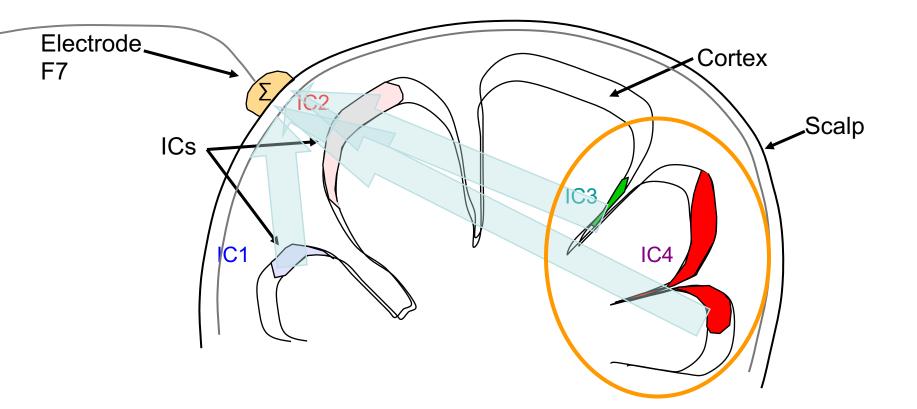




Makeig, 2005

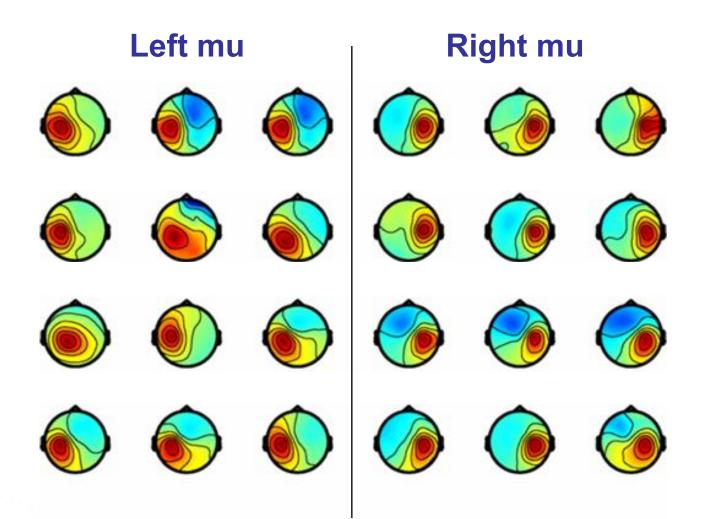






Clustering ICA components by eye





Makeig et al., ~2000 unpublished

Man So how to cluster components?



The same problems hold for clustering independent components

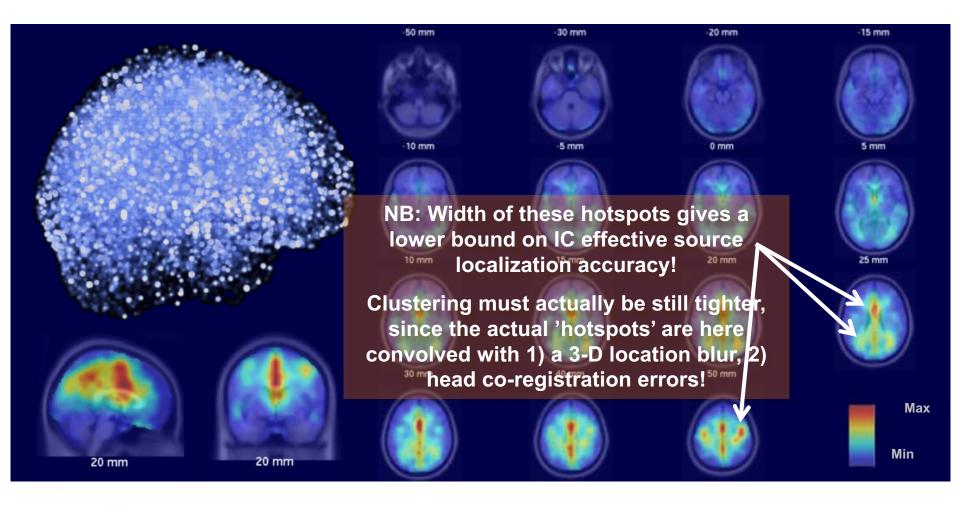
Across Ss, components don't even have "the same" scalp maps! \rightarrow Are "the same" components found across subjects?

- What should define "*the same*" (i.e., "*component equivalence*")?
 - Similar scalp maps?
 - Similar cortical or 3-D equivalent dipole locations?
 - Similar activity power spectra?
 - Similar ERPs?
 - Similar ERSPs?
 - Similar ITCs?
 - Or similar *combinations* of the above?? ...

EEGIC Source Locations



(135,794 IC equivalent dipoles!)



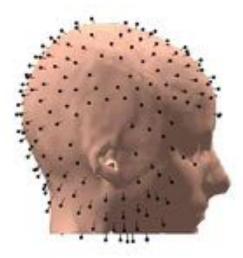
... Some caveats

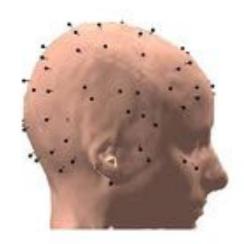
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In this *dipoledensity()* assay ...

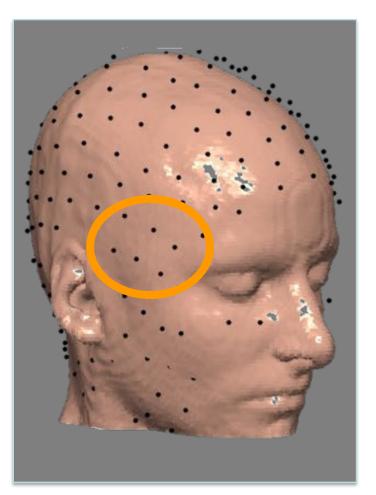
- MR head images were not available \rightarrow brain co-registration crude.
- Single versus dual-dipole model selection was subjective.
- Different electrode montages \rightarrow mis-localization effects.
- Electrode locations were not all digitized some 'guestimated' !
- Brain geometries differ!





Co-Registration of Electrodes with MR Image



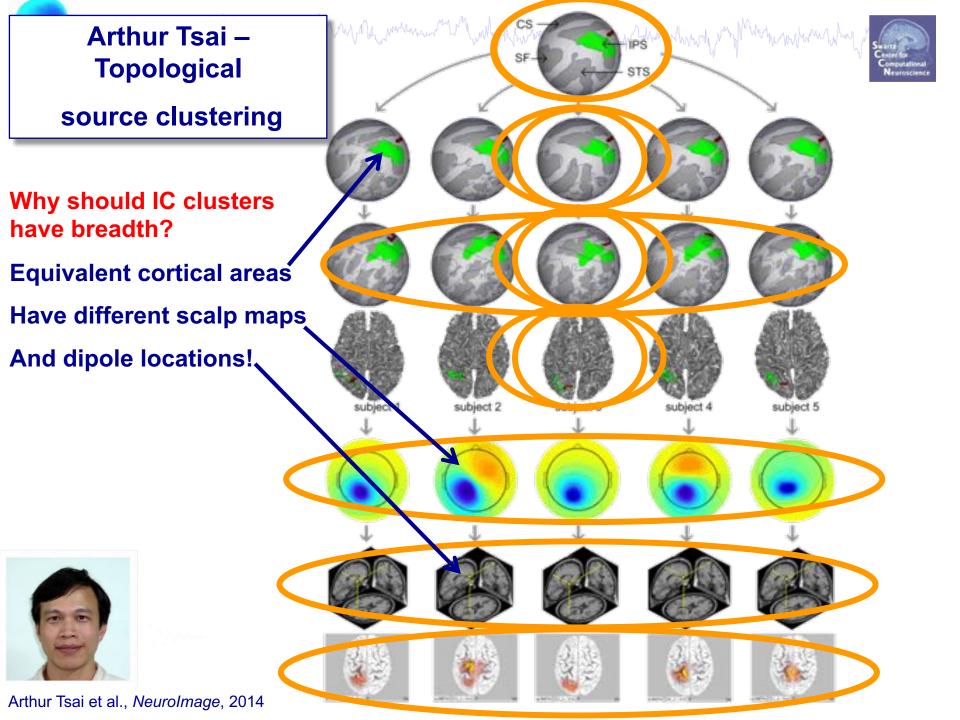




MR + EEG

EEG

Onton & Makeig, 2004





Does the spatial distribution of IC equivalent dipole source locations depend on the task the subject performs?

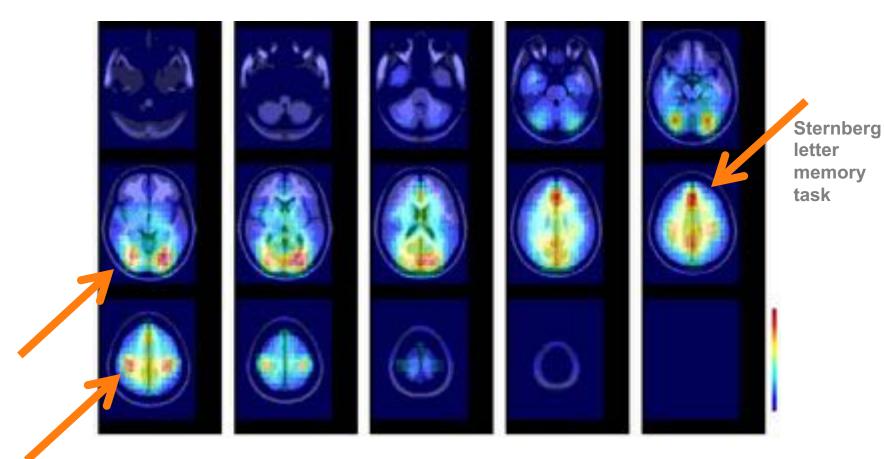
i.e.

Do "the same" ICs (and IC clusters) appear for every task?

Makeig, 2007

www.www.weity.org/www.www.w



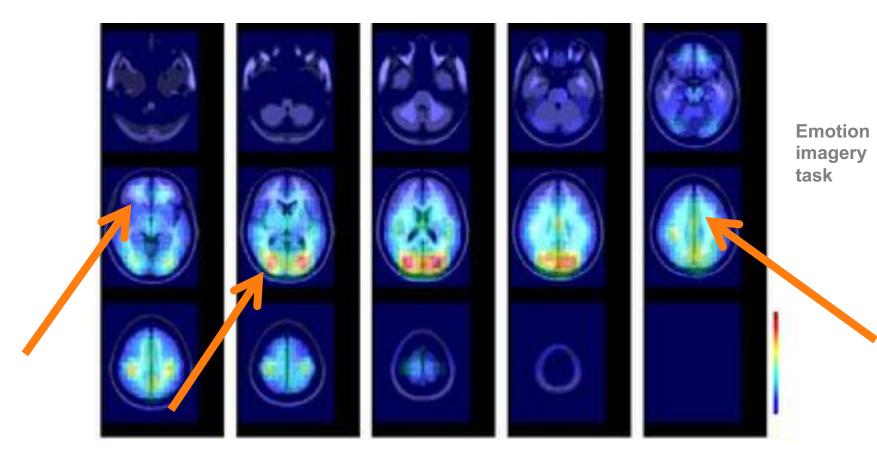


>> dipoledensity()

Onton et al., 2

Onton et al., '05





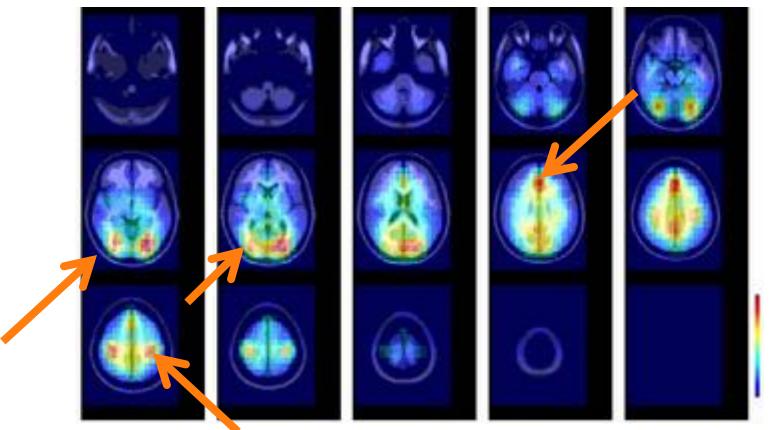
>> dipoledensity()

Onton et al., 2

Onton et al., '05

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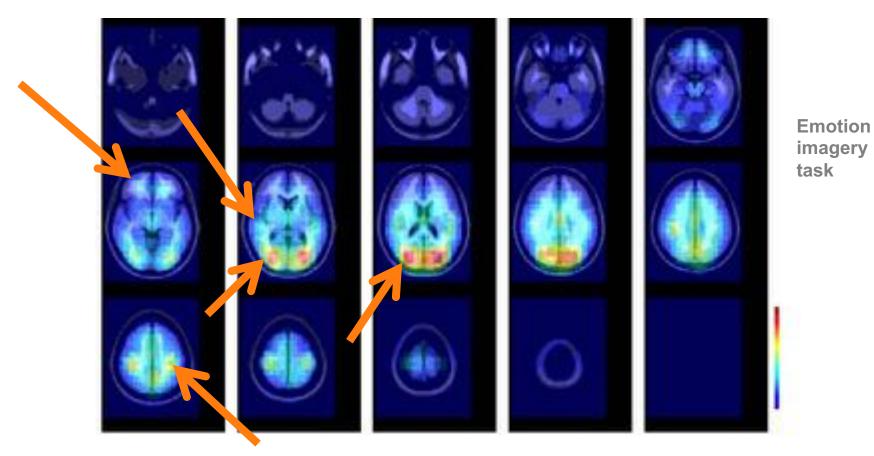




Sternberg letter memory task

>> dipoledensity()





>> dipoledensity()

Onton et al., 2

Onton et al., '05

M. M. So how to cluster components?



The same problems hold for clustering independent components

Across Ss, components don't even have "the same" scalp maps! → Are "the same" components found across subjects?

- What should define "the same" (i.e., "component equivalence")?
 - Similar scalp maps?
 - Similar cortical or 3-D equivalent dipole locations?
 - Similar activity power spectra?
 - Similar ERPs?
 - Similar ERSPs?
 - Similar ITCs?
 - Or similar *combinations* of the above?? ...
 - EEGLAB clustering supports all these possibilities.



Study IC Clustering: Assumptions

• Assumes there are *functionally* equivalent ICs across most subjects.

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- Assumes these ICs have similar responses to experimental conditions across a set of measures (ERP, ERSP, ITC...)
- Creates *non-overlapping IC partitions* making each IC belong to only one cluster.



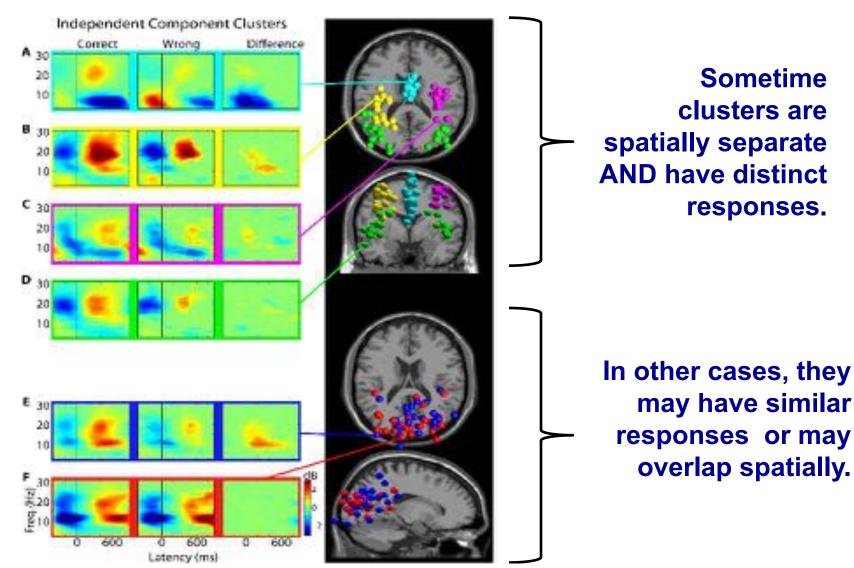
EEGLAB Study Clustering strategy

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- 1. Cluster on **multiple measures** (**dipole locations**, scalp maps, spectra, ERPs, ITCs, ERSPs, ...) in one or more conditions.
- 2. Reduce the dimension of each measure to a principal component subspace.
- 3. Compose a PCA-reduced **position vector** for each component.
- 4. Cluster the composed component vectors using k-means or other.
- Use the computed component measures (not PCA-reduced) to visualize the activities and spatial properties of the clustered components.
- 6. Compute and visualize the **cluster-mean measures**.
- 7. Use **clustered Study set data** as input into **'std_yyy'** functions.

Study IC Clustering





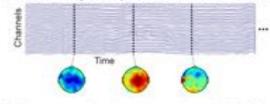
Onton & Makeig, 2007

EEGLAB Study Clustering procedure

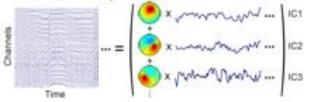


- 1. Identify a set of datasets as an EEGLAB **Study**.
- 2. Specify the **subject** code, subject **group**, **condition** and/or **session** for each dataset in the Study.
- 3. Identify components to cluster in each Study dataset.
- 4. Decide on **component measures** to use in clustering the Study and/or to evaluate the obtained component clusters.
- 5. Compute the component measures for each Study dataset.
- 6. Cluster the components on these component measures.
- 7. Review the obtained clusters (e.g., their scalp maps, dipoles, and activity measures).
- 8. Edit the clusters (manually remove/shift components, make subclusters, merge clusters, re-cluster).
- 9. Statistically test differences within or between selected clusters.

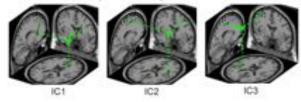




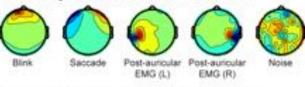
2. Decompose single-subject data with AMICA



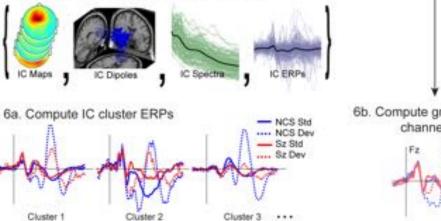
3. Estimate IC equivalent dipole locations



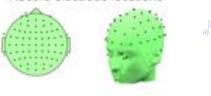
4a. Identify & remove non-brain artifact ICs

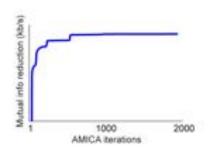


5. Cluster brain ICs across subjects based on

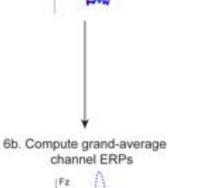


Record electrode locations

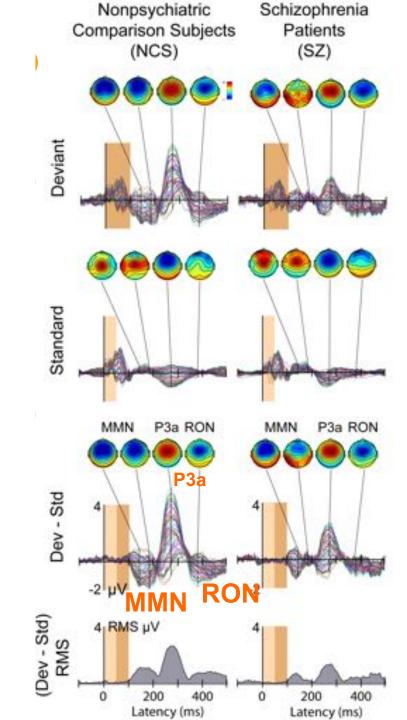




4b. Compute artifact-removed channel ERPs

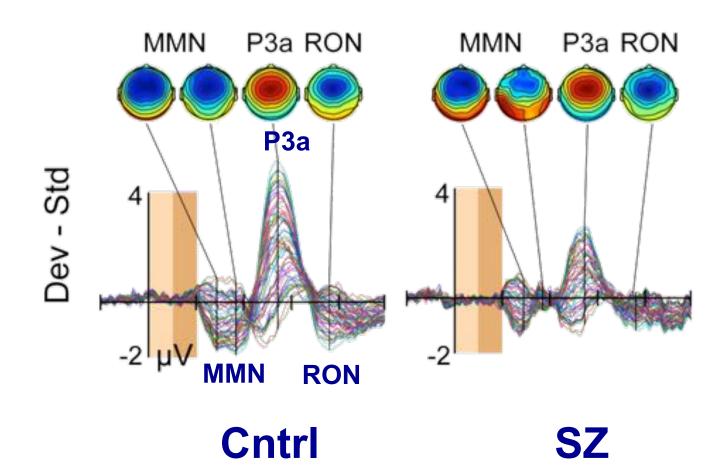


After removal

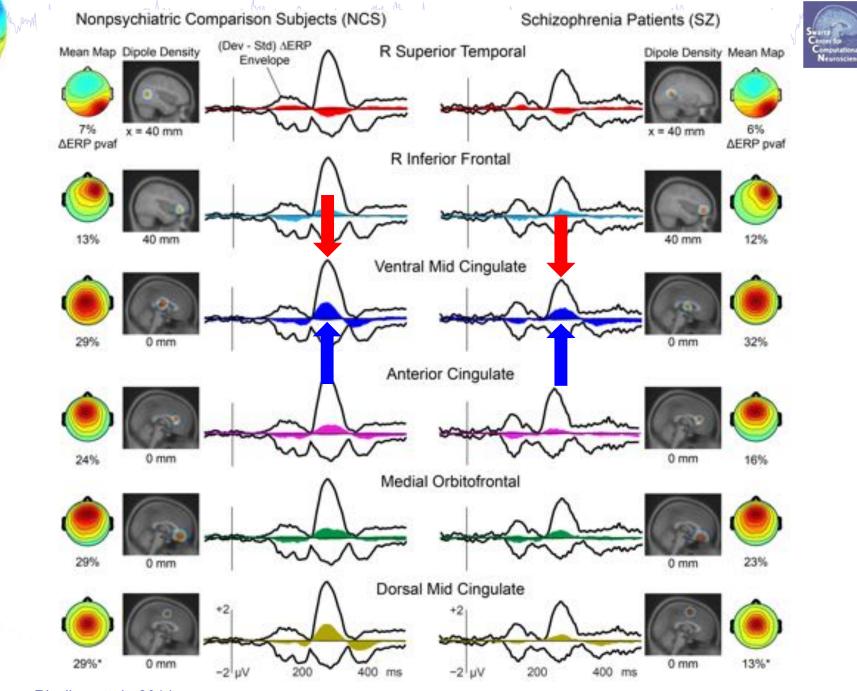


Man Auditory Deviance Response



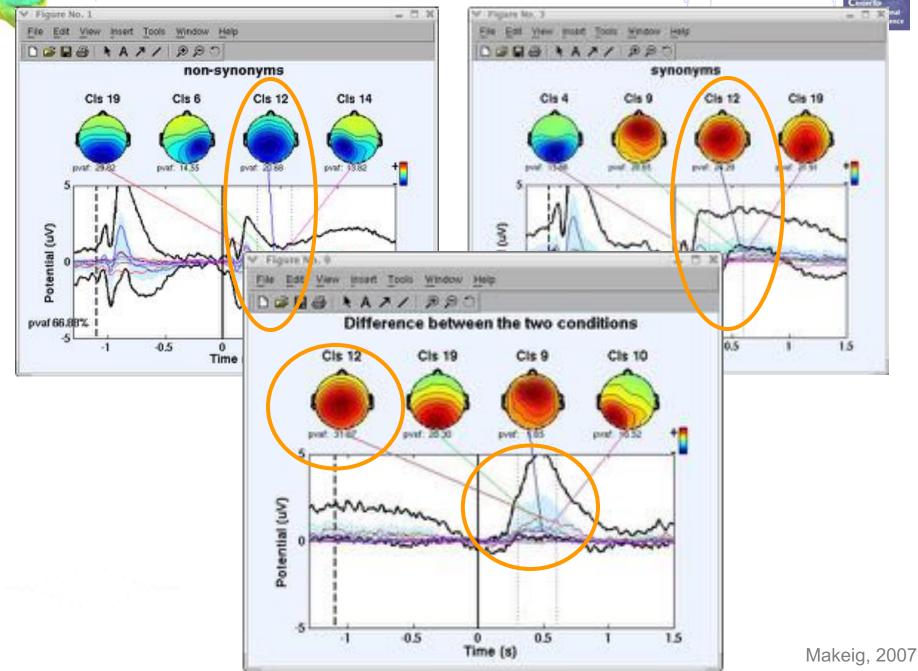


The deepest mental trap in science lies in the word "THE" !!!



Rissling et al., 2014

Cluster ERP contributions - std_envtopo()



PEAK AMPLITUDES	MALAN ERP	mun m 2 M	Myrkharma	vinder when a provide	With Man And March	M My My My My My My My
Scalp Electrode (Fz)						DR
Verbal IQ (WRAT)	P3a	0.11				
Functional Capacity (UPS	RON	0.12		MMN	P3a RON	MMN P3a R0
K Superior Tomporal						
Working Memory (LNS Reorder)	RON	0.15				
Verbal IQ (WRAT)	RON	0.15				
Immediate Verbal Memory (CVLT)	RON	0.28				
Delayed Verbal Memory (CVLT)	RON	0.26	Std		A	4
Functional Capacity (UPSA)	MMN	0.48	S	4		
Functional Capacity (UPSA)	RON	0.26	5			
R Inferior Frontel			Dev	à		
Negative Symptoms (SANS)	RON	0.36		Harris Contraction		Marrie Marriel
Psychosocial Functioning (501)	KUN	0.24				A S at
Auditory Attention (LNS Forward)	MMN	0.38		-2'µV	4	-2
Working Memory (LNS Reorder)	MMN	0.30				~7
Verbal IQ (WRAT)	MMN	0.46		C	ntrl	SZ
Ventral Mild Emgulate Positive Symptoms (SAPS)	RON	0.29	\sim			
Negative Symptoms (SAIS)	P3a	0.36	2			
Immediate Verbal Momory (CVIT)	DOU	0.30				
Delayed Verbal Memory (CVLT)	RON	0.24				
Verbal IQ (WRAT)	RON	0.29				
Executive Functioning (WCST)	RON	0.24				
Anterior Cingulate						
Functional Status (GAF)	MMN	0.18				
Functional Status (GAF)	RON	0.17				
Immediate Verbal Memory (CVLT)	RON	0.25				
Delayed Verbal Memory (CVLT)	RON	0.17				
Medial Cripitorrontal						
Positive Symptoms (SAPS)	P3a	0.40				
Negative Symptoms (SANS)	P3a	0.54	/			
Psychosocial Functioning (SOE)		0.37				
Functional Capacity (UPSA)	P3a	0.32				
Dorsal Mid Cingulate						
Verbal IQ (WRAT)	P3a	0.15				

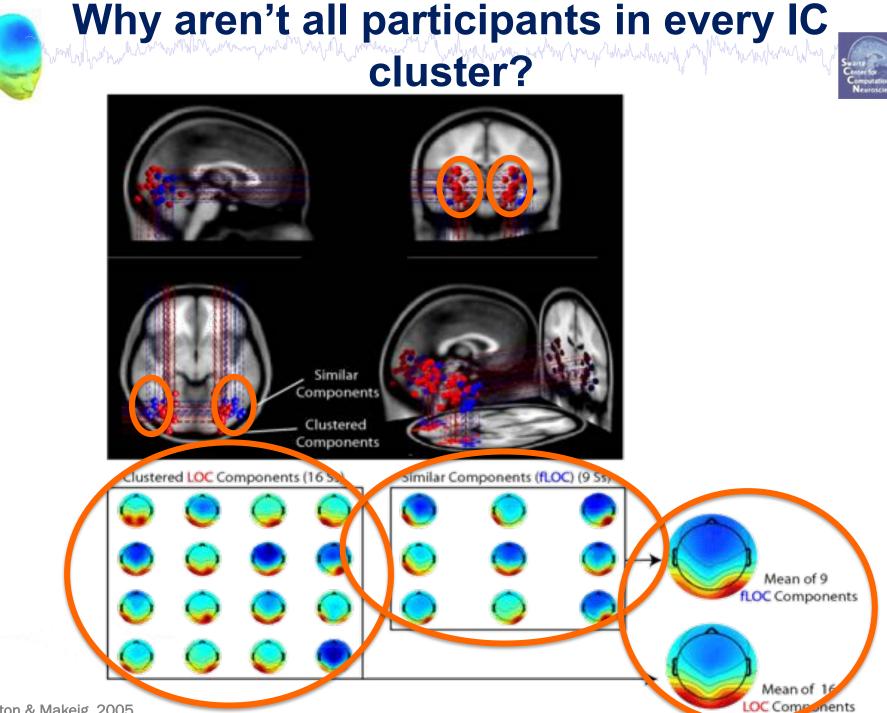
PEAK LATENCIES	M. M. ERP. M	1. Marty	Kharman Mary Mary	Amy Water And	Mar Mar Mar Star
Scalp Electrode (Fz)					C.N.N
n/a			M	MN P3a RON	MMN P3a RON
K Superior Temporal			IVII.		MMN P3a RON
Functional capacity (UPSA)	MMN	0.25			
Delayed Verbal Memory (CVLT)	MMN	0.17	\sim		\sim \sim \sim \sim \sim \sim
R Inferior Frontal			~ \		
Negative Symptoms (SANS)	RON	0.51	4 Std		4
Psychosocial Eunctioning (SOE)	RON	0.25	i l		
Executive Functioning (WCST)	MMN	0.30			
Executive Functioning (WCST)	P3a	0.28	- marker	A BUILDE	Min Ali Ant
Ventral IVII Cingulate					A. 6. 444
Negative Symptoms (SANS)	РЗа	0.33	-2'µ		-2
Negative Symptoms (SANS)	NON	0.33			07
Psychosocial Functioning (SOF)	P3a	0.31		Cntrl	SZ
Verbal IQ (WRAT)	MMN	0.25			
Executive Functioning (WCST)	P3a	0.30			
Anterio: Cingulate					
Functional Capacity (UPSA)	RON	0.17			
Verbal IQ (WRAT)	MMN	0.24			
Auditory Attention (LNS-Forward)	MMN	0.17			
Medial Orbitofrontal					
Negative Symptoms (SANS)	RON	0.41			
Positive Symptoms (CAPC)	PON	0.40			
Auditory Attention (LNS-Forward)	MMN	0.29			
Executive Functioning (WCST)	P3a	0.32			
Dorsal Mid Cingulate					
Negative Symptoms (SANS)	MMN	0.20			
Negative Symptoms (SANS)	P3a	0.17			
Global Functioning (GAF)	RON	0.24			
Functional Capacity (UPSA)	P3a	0.13			



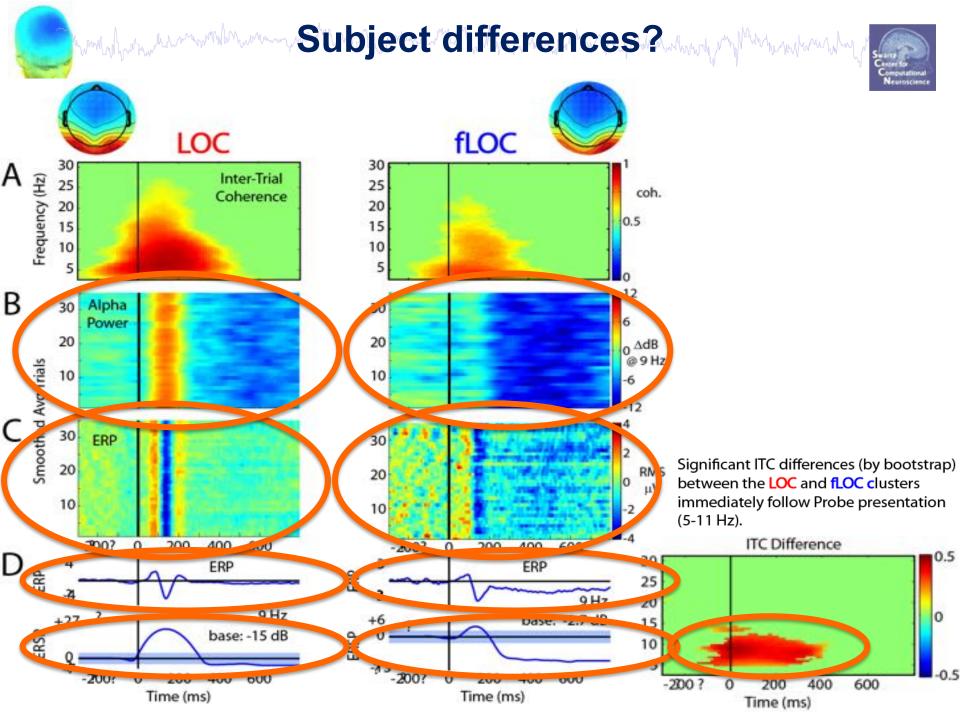
Not all subjects contribute components to each cluster.

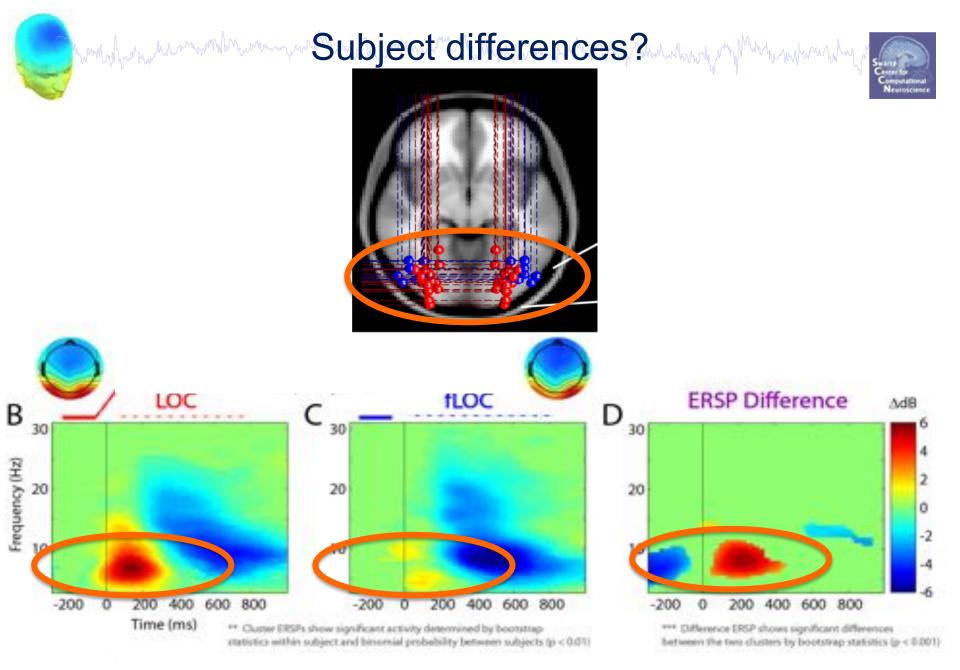
Why not?

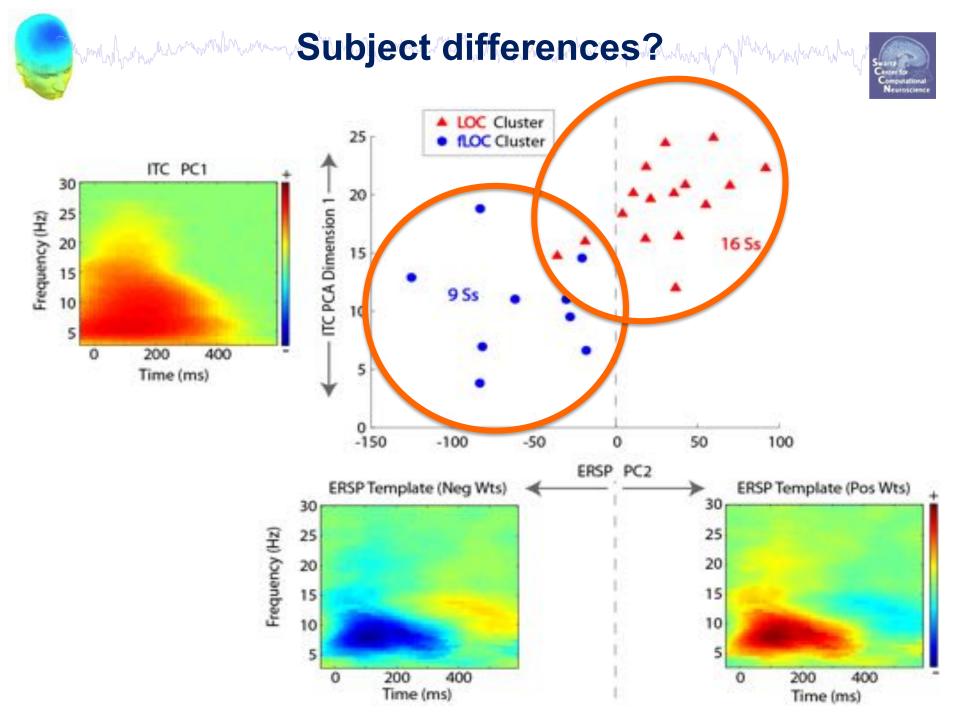
- Different numbers of artifact components
- Subject differences!?
- Does my subject group really exhibit a Gaussian cloud of individual differences around 'a mean subject' in 'subject space' ??



Onton & Makeig, 2005





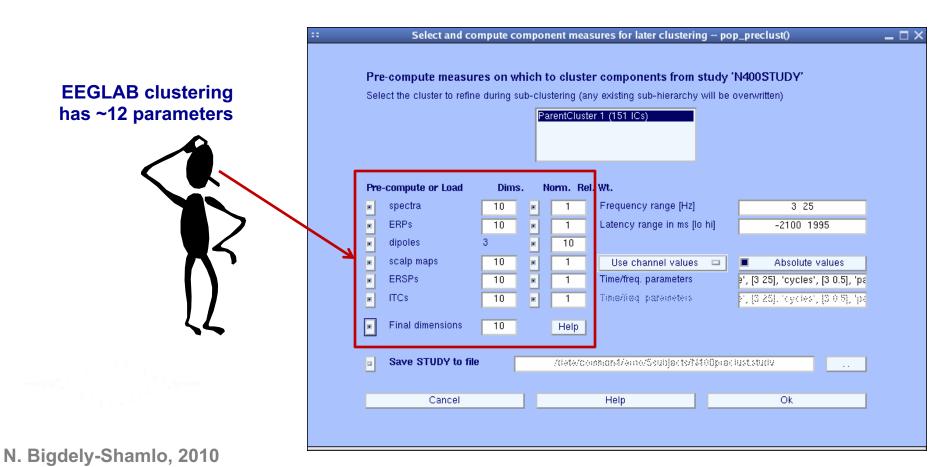


would have been and the second of the second



STUDY IC Clustering: Practical Problems

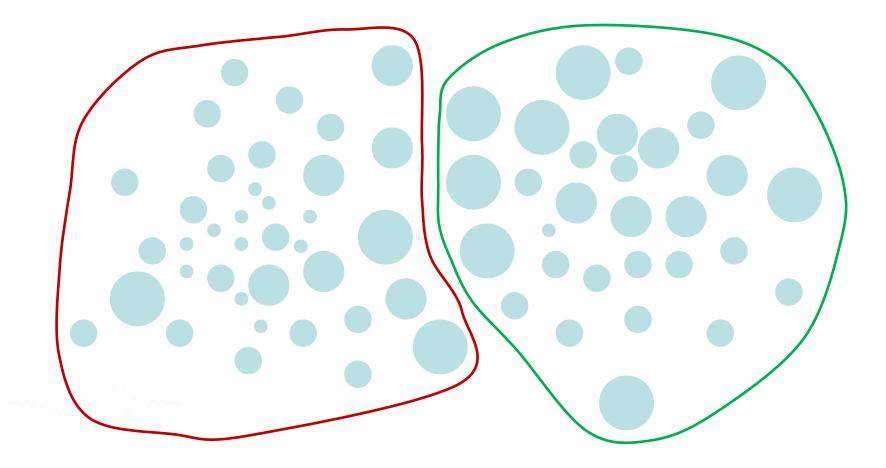
Large parameter space problem: many different clustering solutions can be produced by changing parameters and measure subsets. Which one should we choose?



Problems with multi-measure clustering

What are the clusters according to location?

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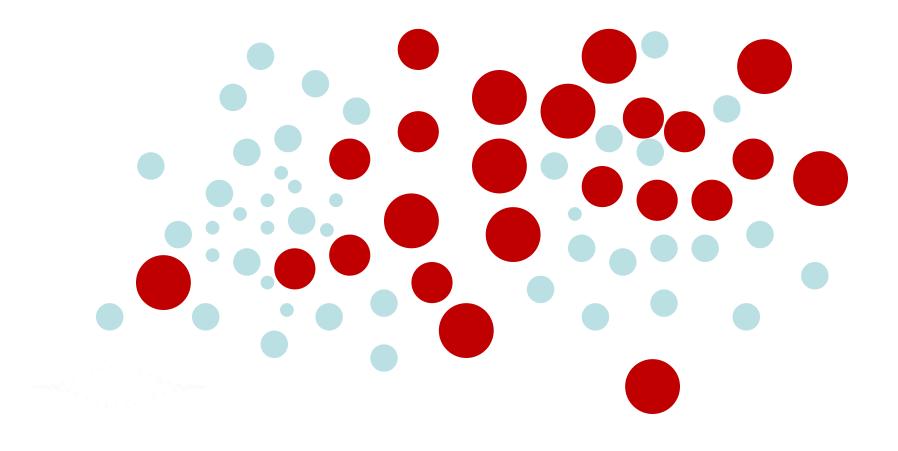




Problems with multi-measure clustering

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What are the clusters according to size ?



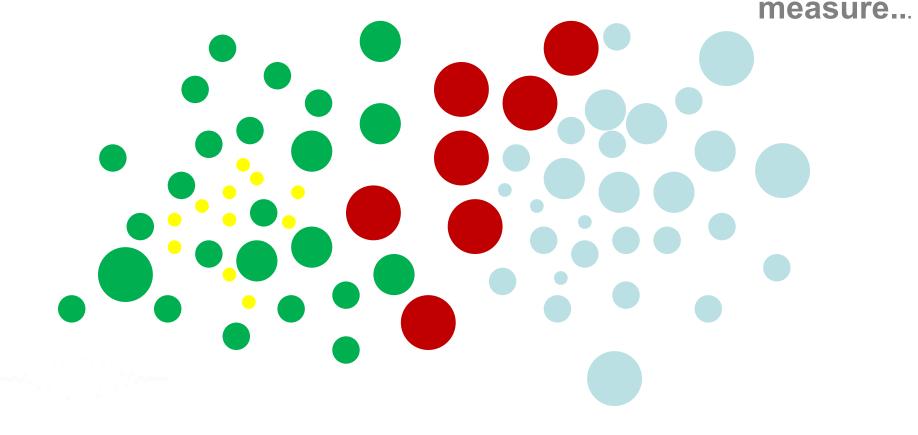
Problems with multi-measure clustering

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What are the clusters according to location and size?

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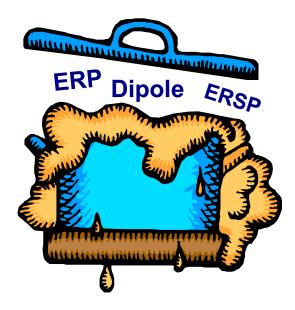
Well... it depends on how much weight we give each



would be when any approximation of the manual of the property of the property



- With either clustering method, we basically mix together distances for a subset of EEG measures (ERP, ERSP, ITC, mean spectrum, dipole location).
- This may make clustering distance less interpretable.



Instead of IC Clustering, use Measure Projection



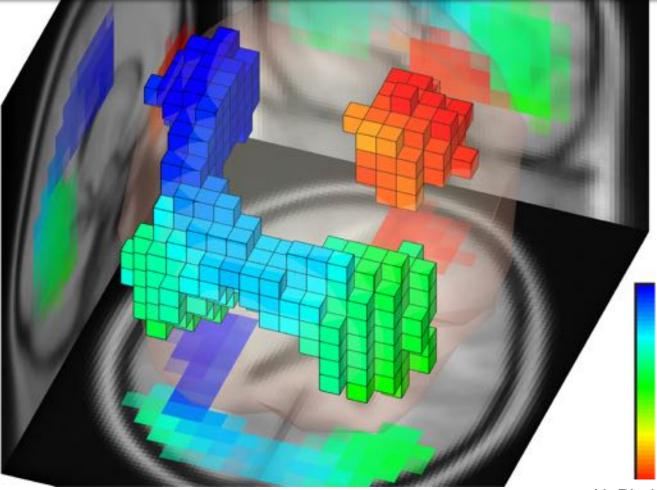
- Instead, we can directly work on pair-wise similarity matrices and prevent ICs with similarities less than certain threshold (e.g., ERSP corr. < 0.5) to be clustered together.
- The most important measure is **equivalent dipole location**.
- Assuming a certain variability estimate for dipole location (due to error in localization and subject variability), one can also estimate an optimum number of clusters.

Measure Projection asks:

- 1. Where in 'template brain space' does our data have evidence that our measure of interest is consistent across nearby ICs?
- 2. Which such brain space voxel *domains* show consistent differences?

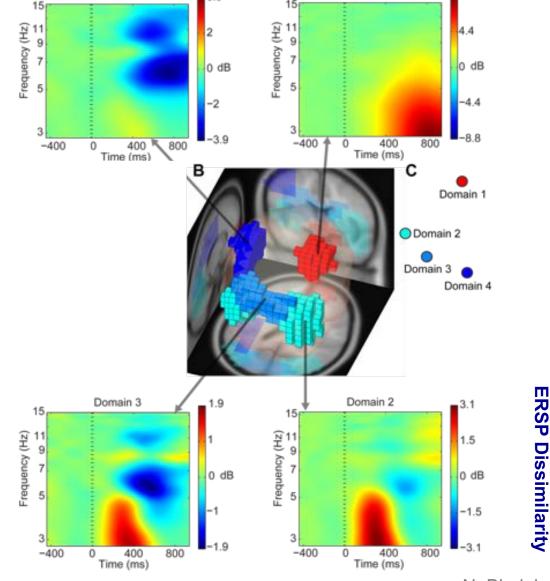


Other nice graphics modes in the Measure Projection Toolbox



ERSP Dissimilarity

Measure Projection Toolbox Measure Projection: Measure Exemplars



N. Bigdely-Shamlo, 2011