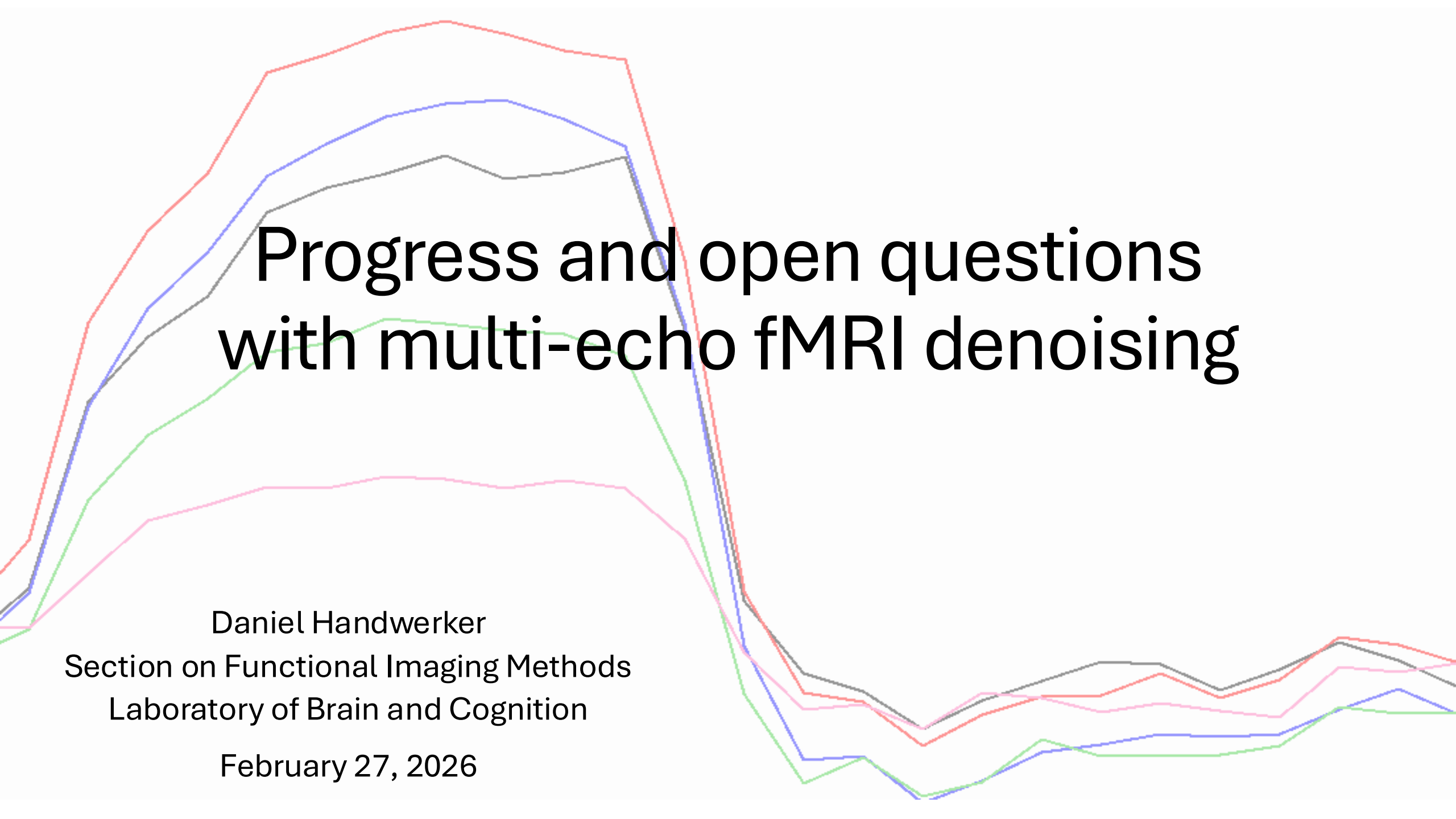




Progress and open questions with multi-echo fMRI denoising

Daniel Handwerker
Section on Functional Imaging Methods
Laboratory of Brain and Cognition

February 27, 2026

About me

I am the grandchild of child refugees
& the child of an immigrant

Grew up: Long Island, NY

Johns Hopkins

Biomedical Engineering, Computer Science
Psych minor

- **Michael Steinmetz:** Spatial encoding, primate Electrophysiology (and fMRI!) (John Port, Christos Constantinidis)
- **Steven Yantis:** Attention, human psychophysics and fMRI (Susan Courtney, Trisha Van Zandt)

NIMH Summer student with Karen Berman
Room 4C101

- MRI voxel-based morphometry & schizophrenic patients (precollected data)
- Piloting PET study with Williams Syndrome patients

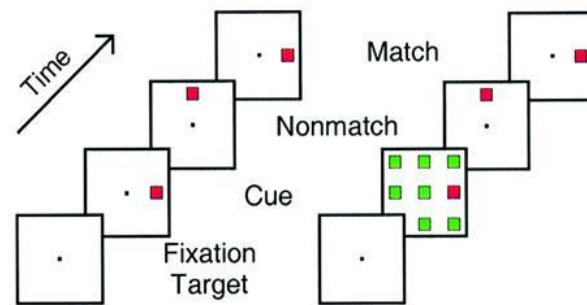
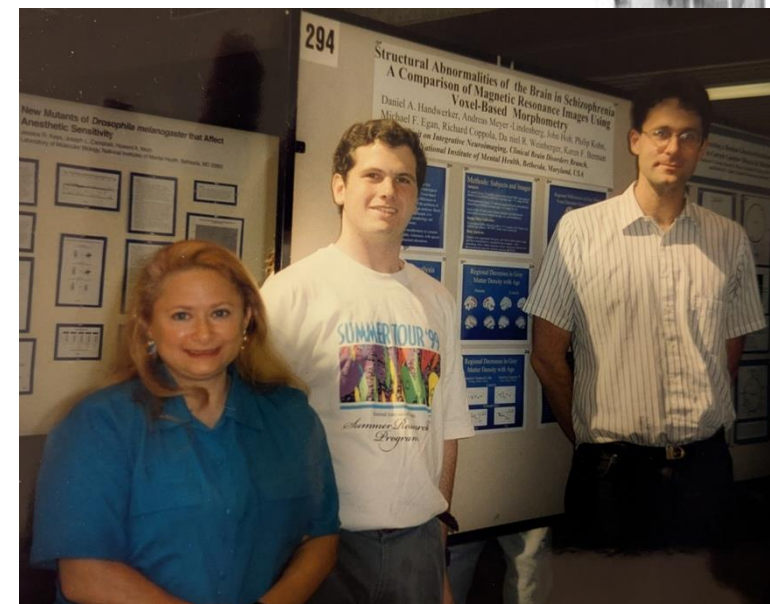
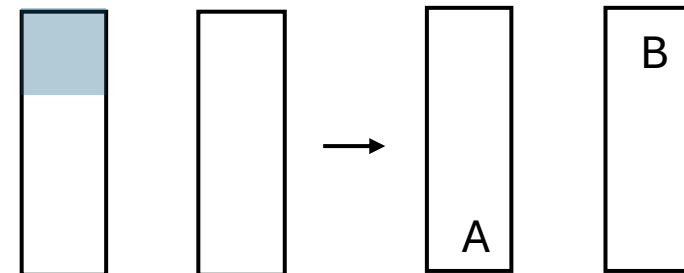
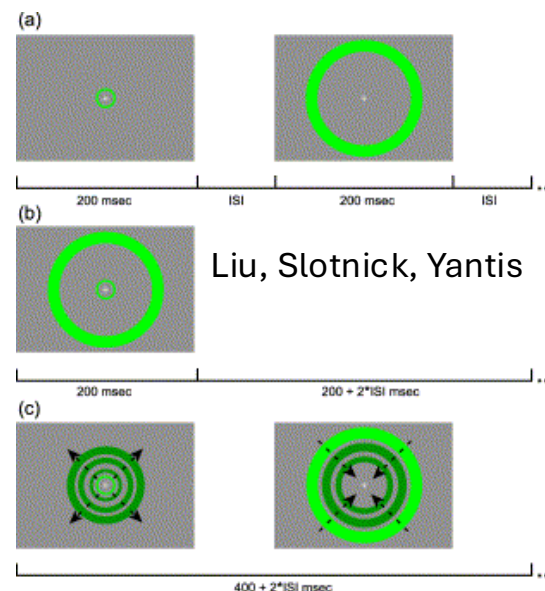


Figure 2
Constantinidis & Steinmetz



About me

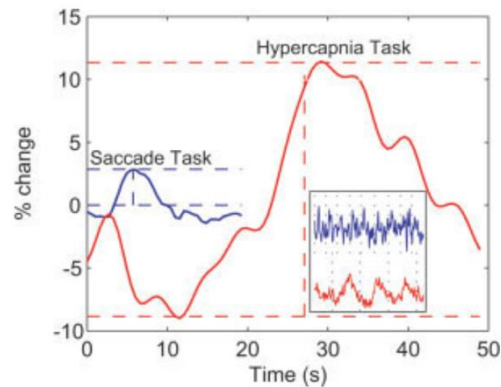
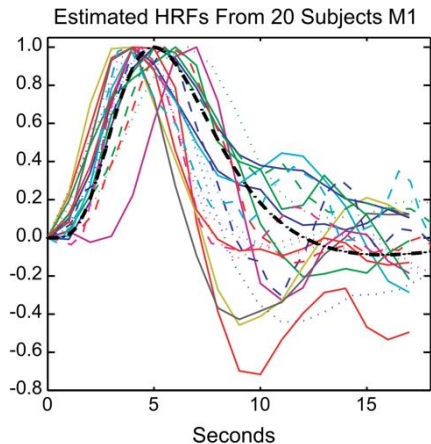
Grad school application:

Many researchers who study the brain are neuroscientists or psychologists who have developed engineering skills, or are engineers interested the brain. The field has a growing number of people who received their primary education in multiple disciplines. I hope to continue my studies **in both fields**.

My current interests are in the uses of fMRI to study the brain. I believe fMRI has **barely begun to reach its full potential** and I hope that in my future research, I can both use and improve on existing technology.

UC Berkeley, UCSF Joint grad group in Bioengineering

PhD mentor Mark D'Esposito



UCSF Postdoc

Roland Henry
DTI + fMRI +
ECoG

Personal
photo
removed

Personal
photo
removed

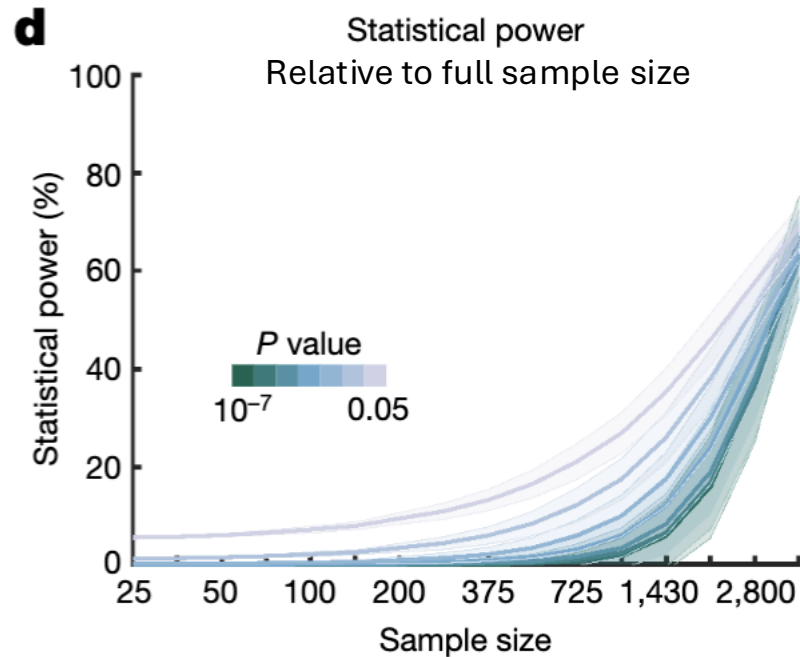
NIMH Postdoc then Staff Scientist

Personal
photo
removed

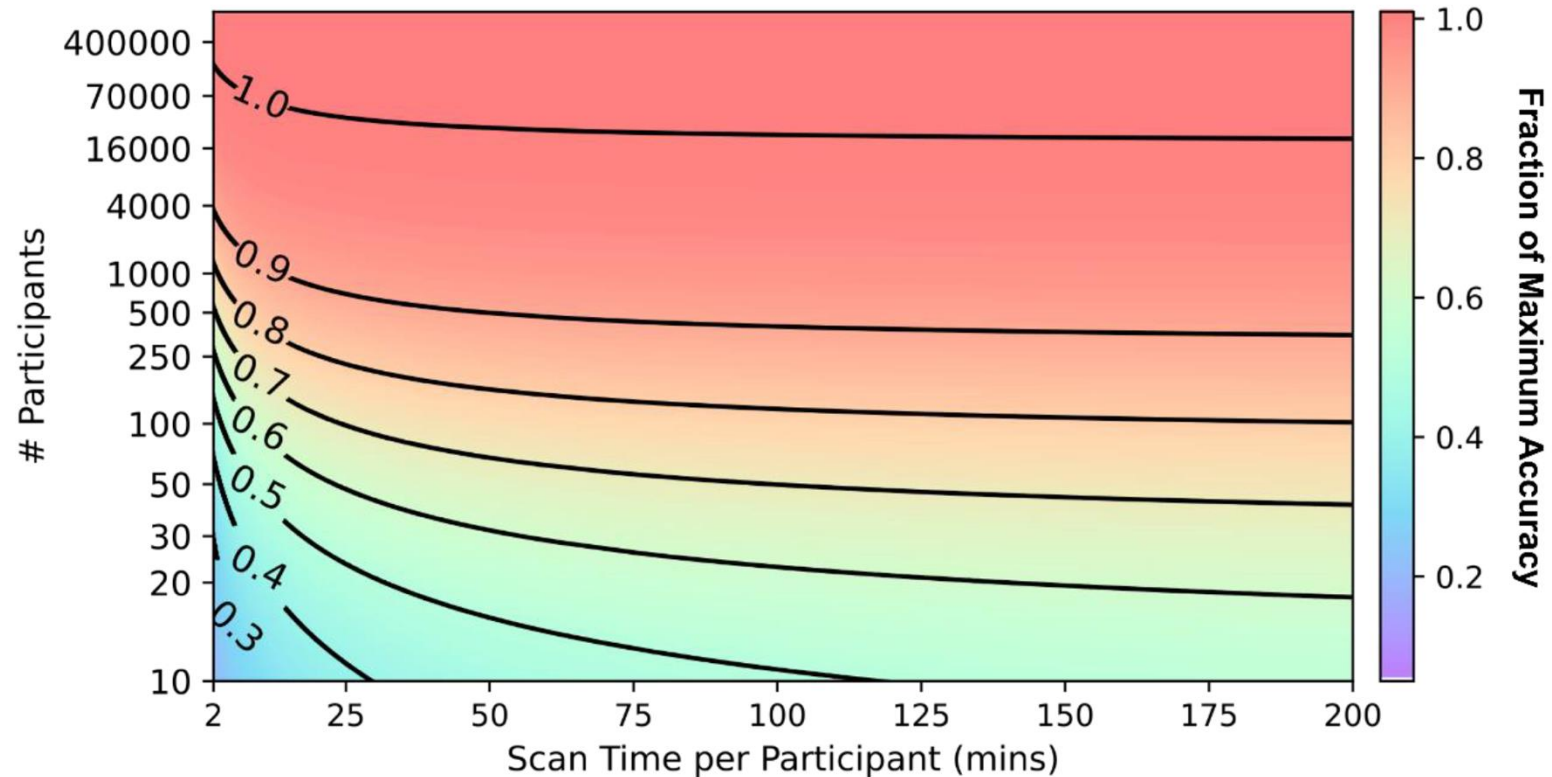
Personal
photos
removed

Reaching fMRI's full potential

Many participants?



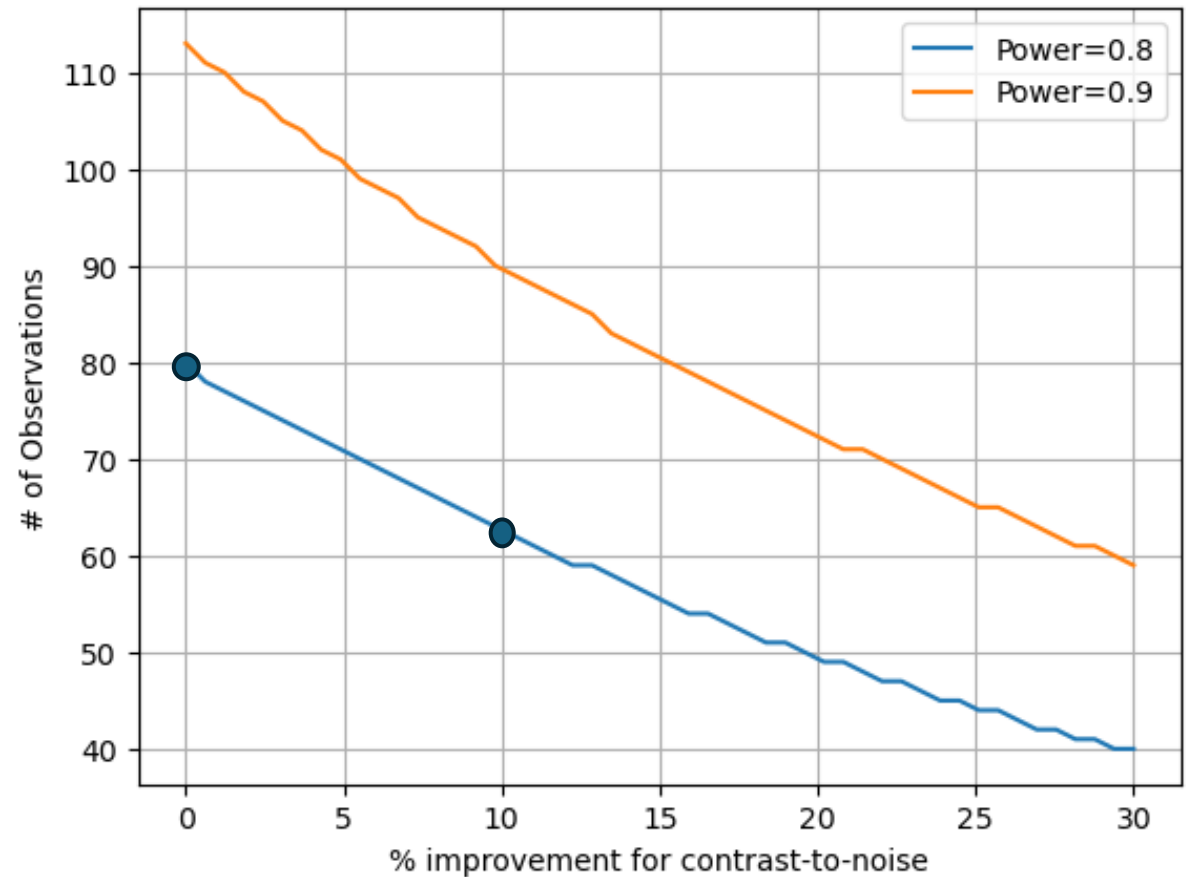
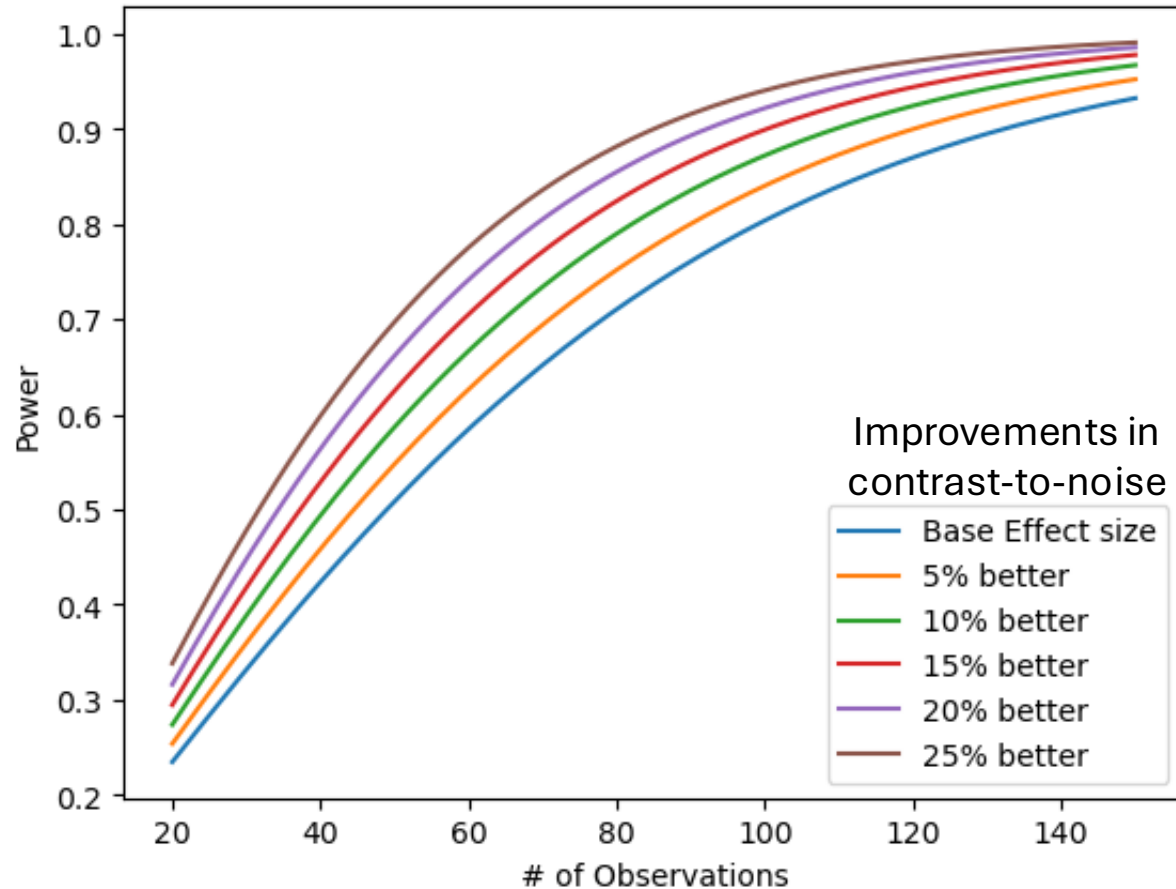
Scan time per participant?



Marek, Tervo-Clemmens, et. al. 2022
Reproducible brain-wide association studies require thousands of individuals <https://doi.org/10.15154/1503209>

Ooi, Orban, et. al. 2024
MRI economics: Balancing sample size & scan duration in BWAS
<https://www.biorxiv.org/content/10.1101/2024.02.16.580448v2>

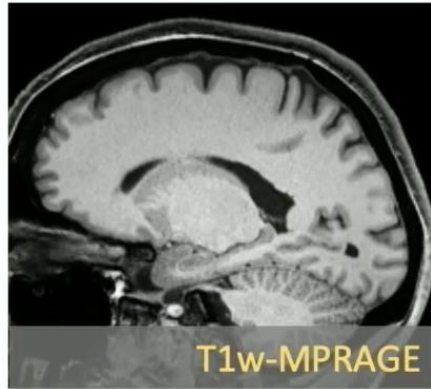
Data quality also matters!



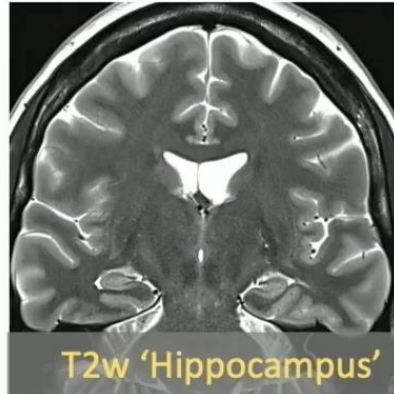
A 10% improvement in contrast-to-noise means a statistical power of 0.8 is possible with 63 vs 80 subjects

Example: Australian Epilepsy Study

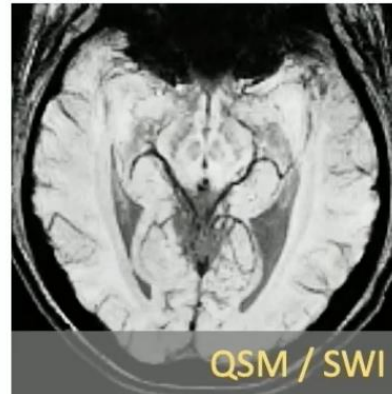
One part of larger imaging protocol



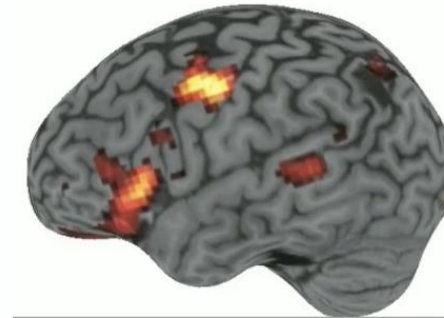
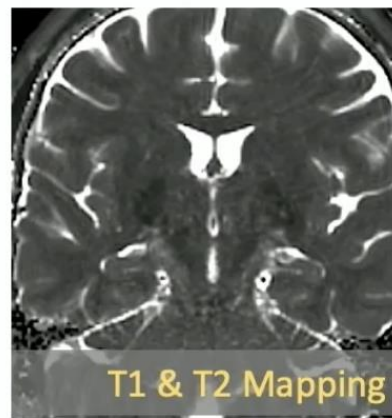
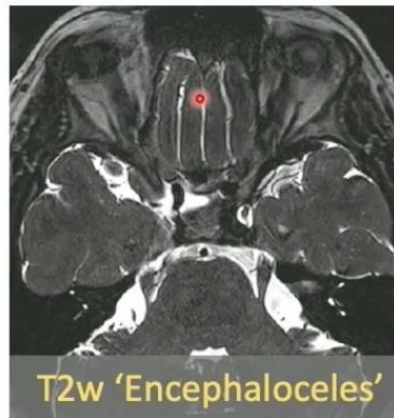
3D whole brain



Targeted FOV



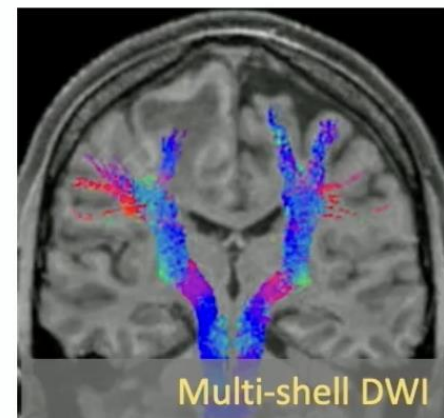
Quantitative MRI



Functional MRI

- a) Language
- b) Working memory
- c) Movie watching

Connectomes



Patient MRI time is very limited.

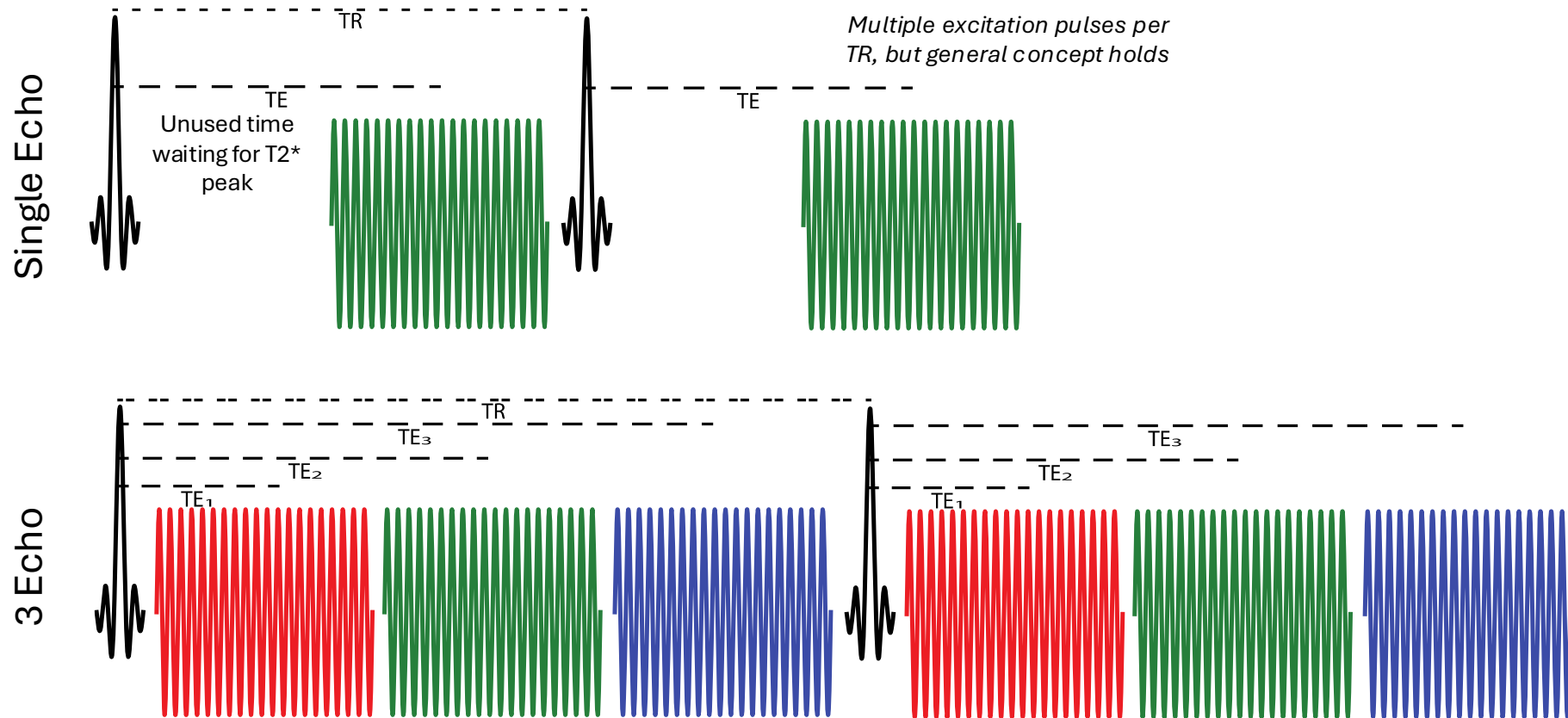
Protocol has 3 min for fMRI with the goal of identifying language lateralization

TR=0.9s, MB=4
Multi-echo=3
3mm³ voxels

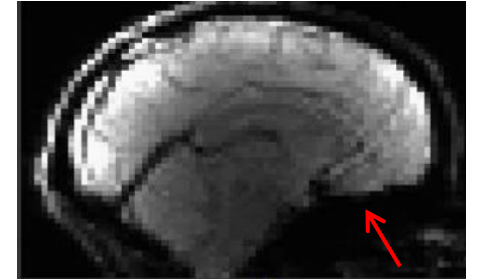
Uses tedana denoising

Of 271 patients scanned only 2% of fMRI scans were unsuccessful

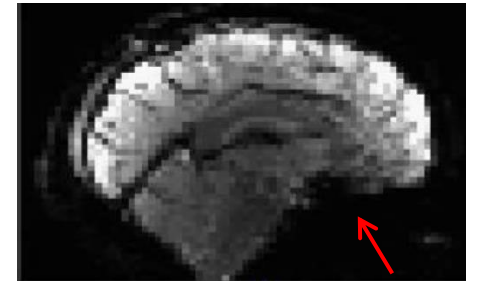
What is multi-echo fMRI?



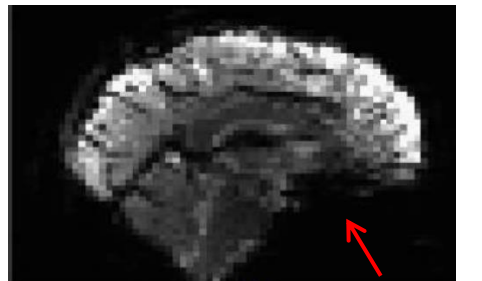
Echo 1



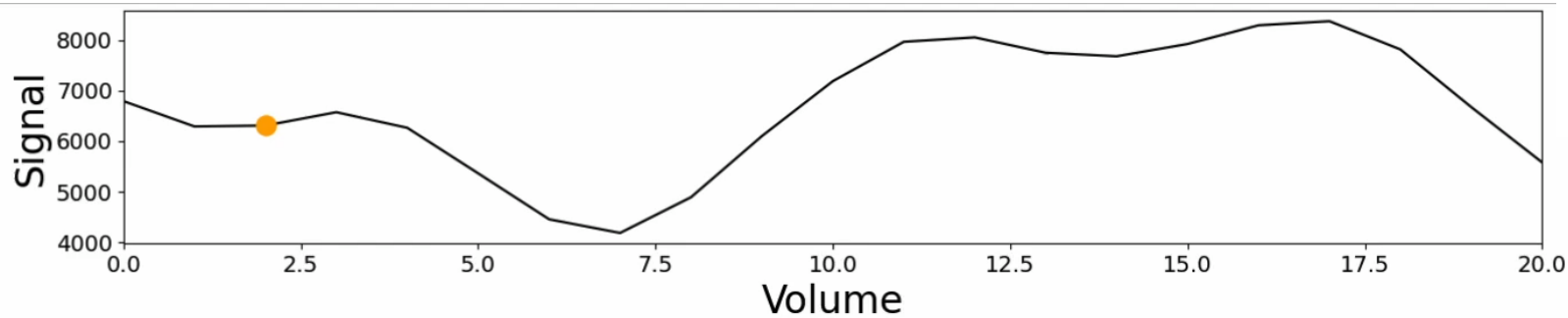
Echo 2



Echo 3

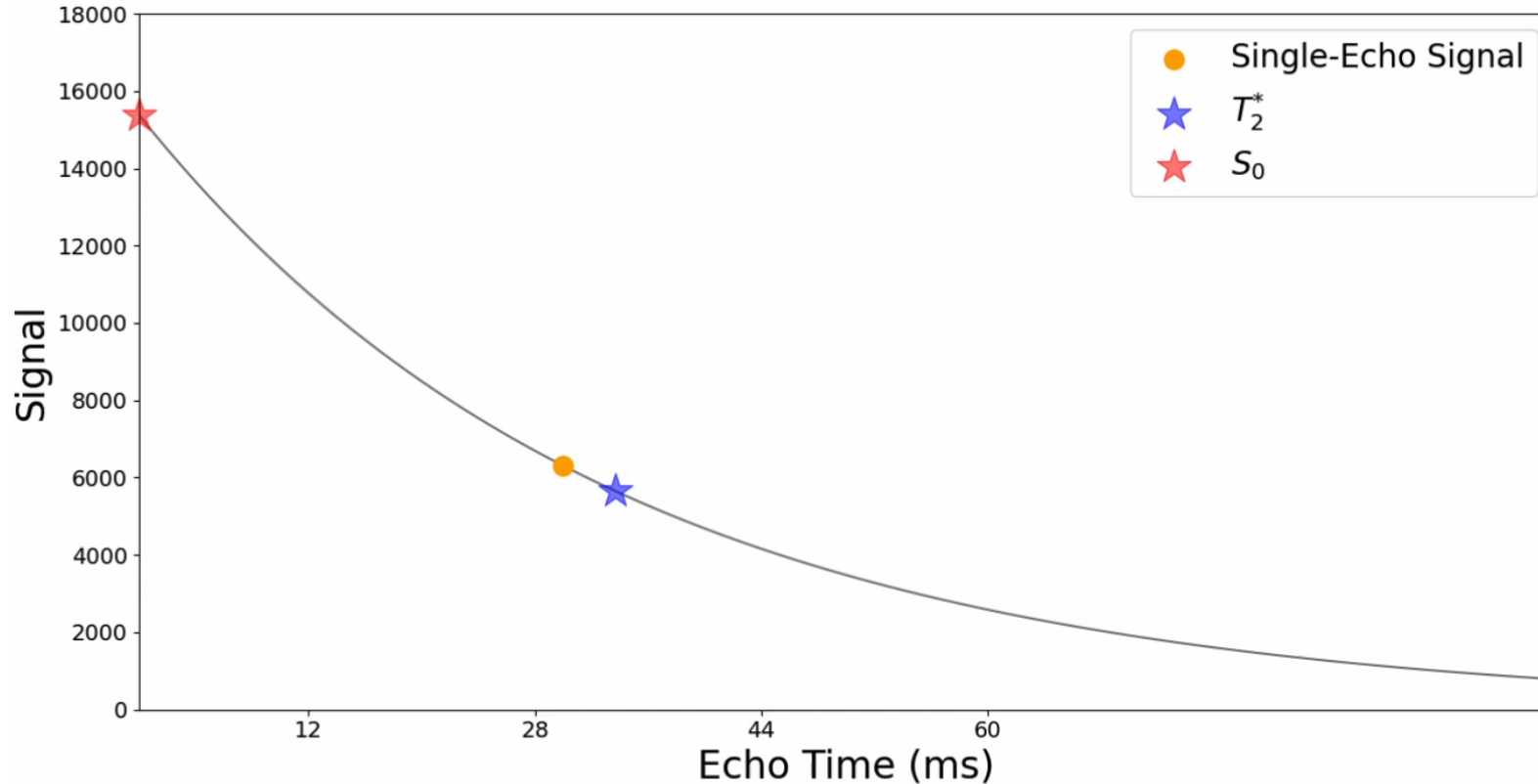


Using Multi-echo information



$$S(t, TE) = S_0(t) \times e^{\frac{-TE}{T_2^*(t)}} + \text{noise}$$

The signal decays with echo time



S_0

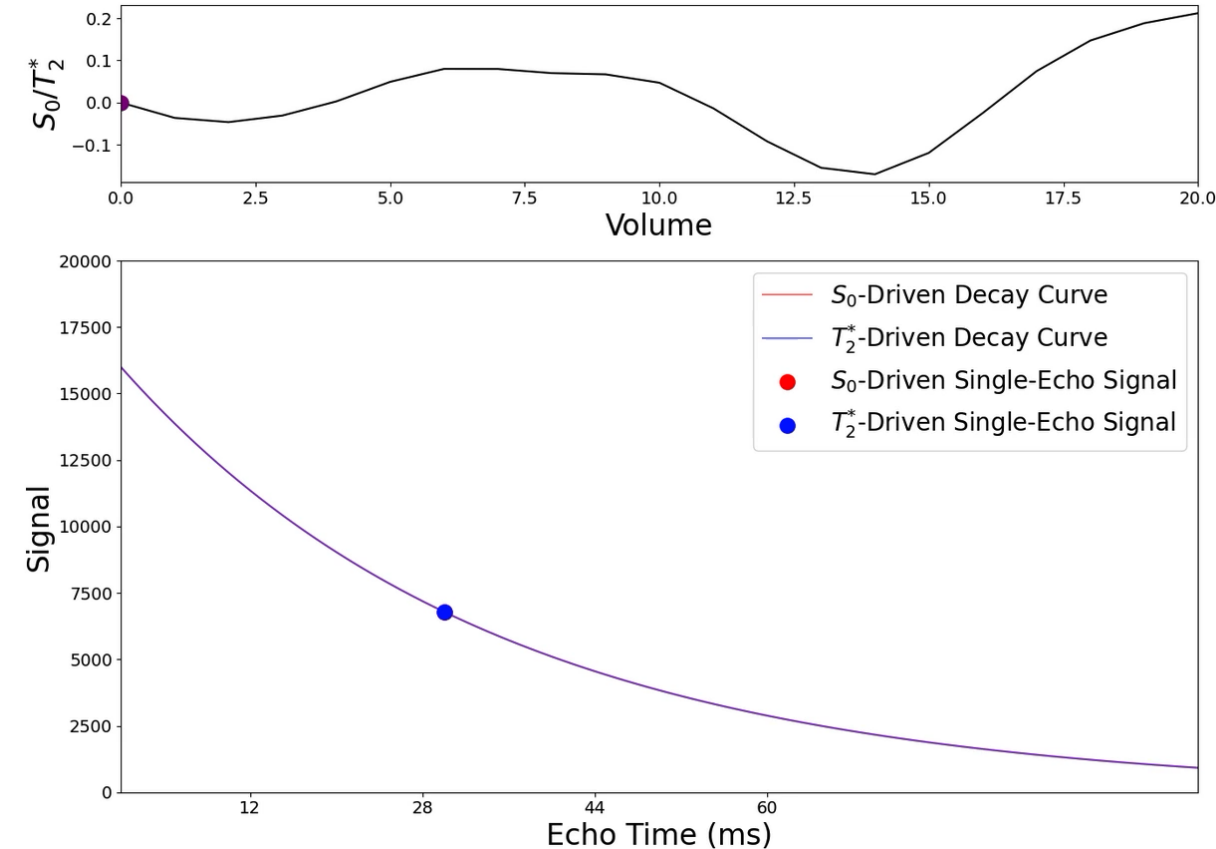
- The y-intercept at TE=0
- Non-neural signals including head motion & scanner artifacts

T_2^*

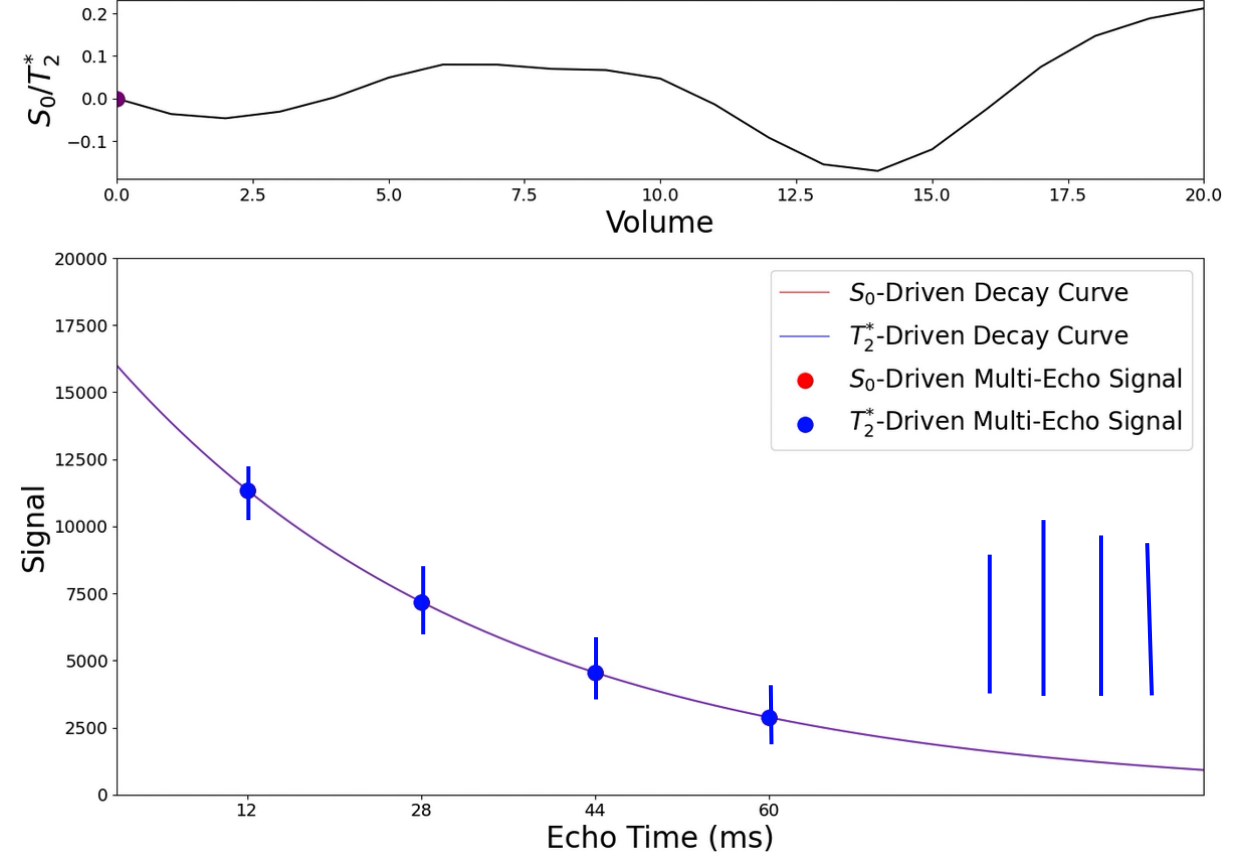
- The rate of decay
- Altered by several factors including changes in blood oxygenation

Multi-echo gives us info to distinguish T_2^* and S_0

What we measure with single-echo fMRI

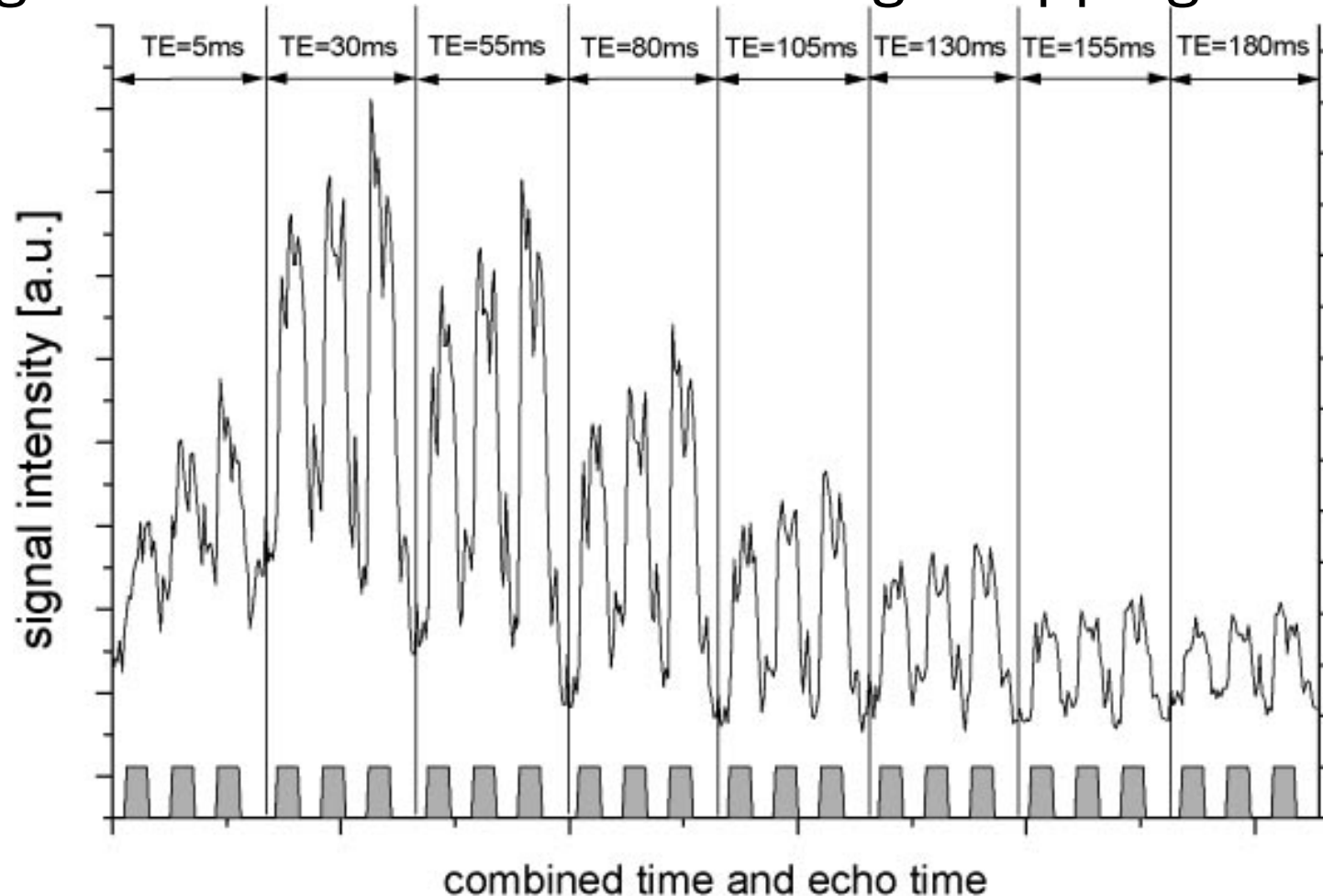


What we measure with multi-echo fMRI



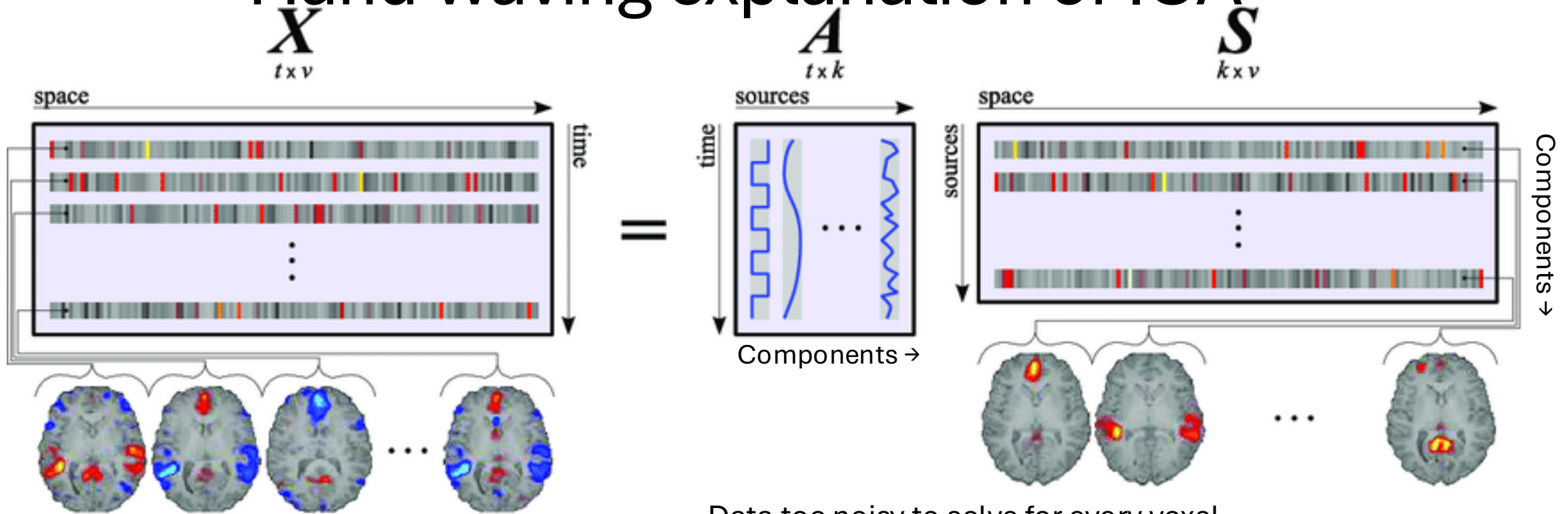
fMRI response magnitudes by echo time

Average across active voxels in a finger tapping task at 3T



MEICA or denoising

Hand waving explanation of ICA



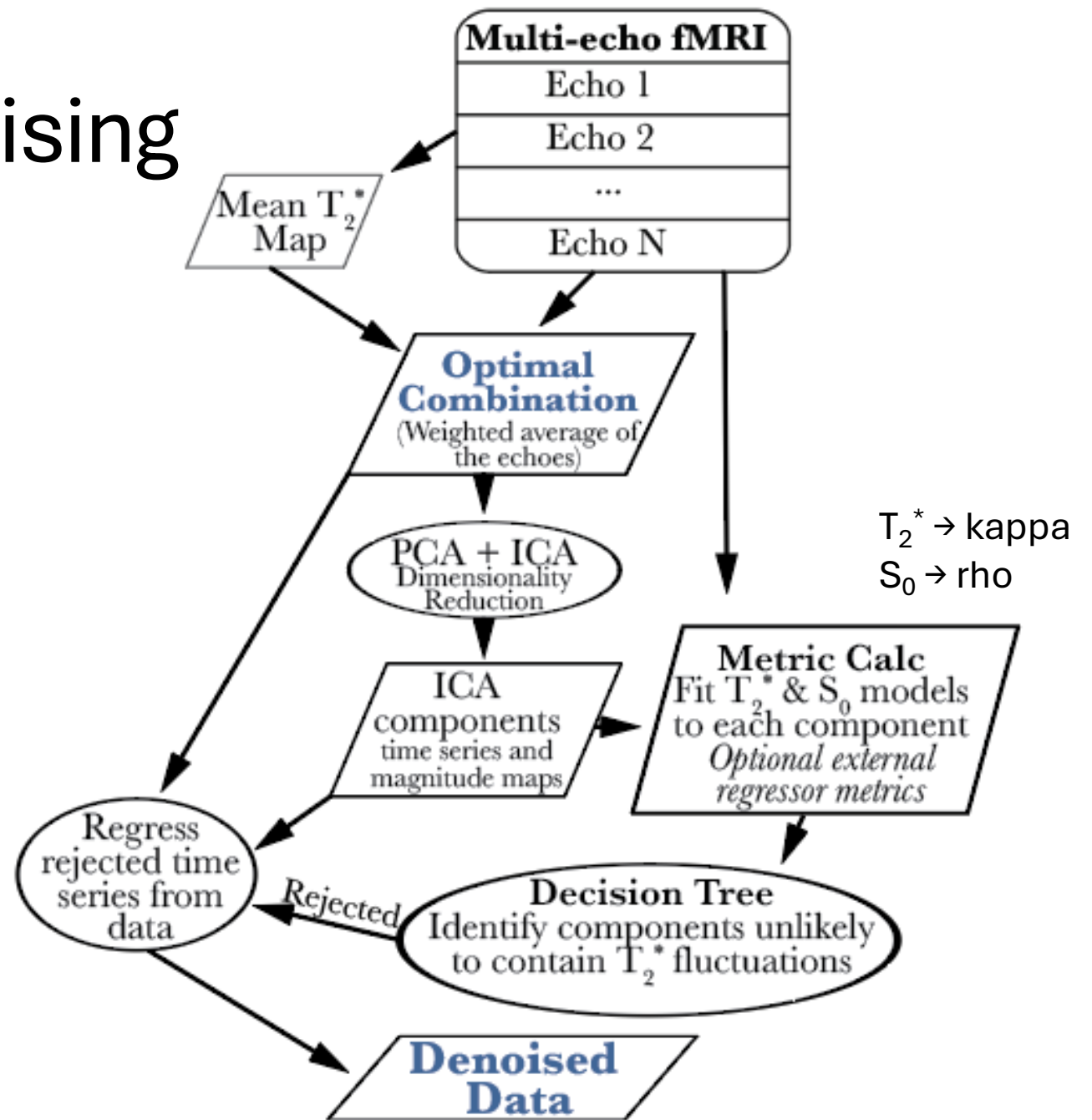
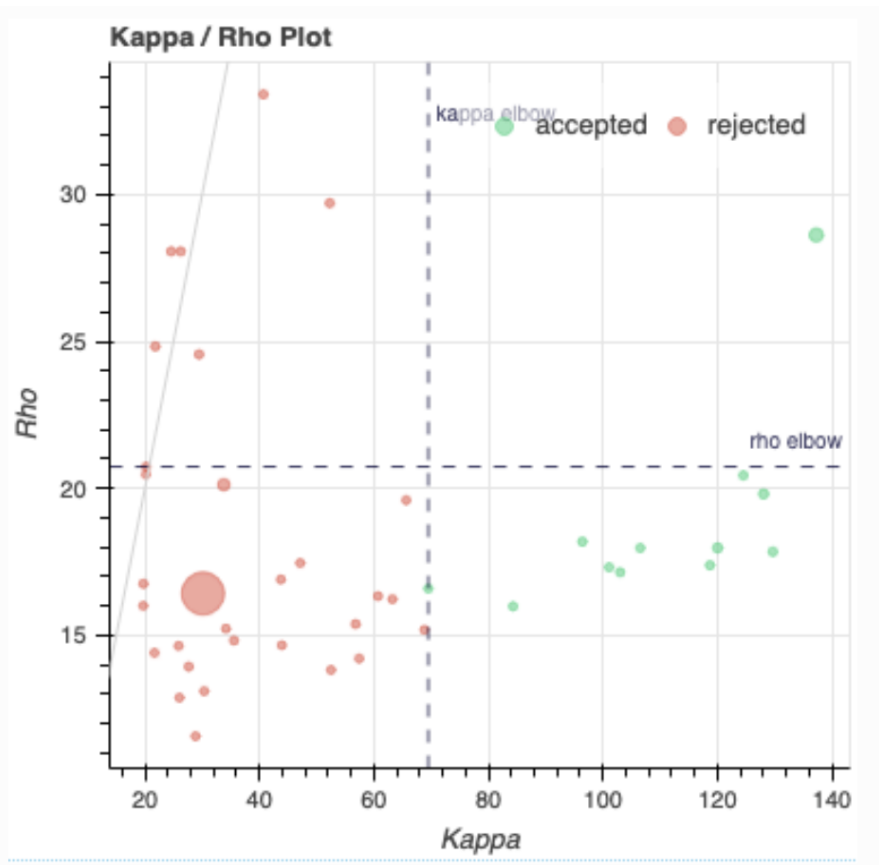
Data too noisy to solve for every voxel and time point separately

$$S(t, TE) = S_0(t) \times e^{\frac{-TE}{T_2^*(t)}} + noise$$

Solving for 2 variables
 ≥ 3 measurements

ICA is “intelligent”
 signal averaging

MEICA or tedana denoising



MEICA (Kundu 2012 & 2013)

tedana (DuPre, Salo, et al, JOSS, 2011)

tedana.readthedocs.io

What's new in

- **Version 26.0.0 released 2 days ago!**
- 11 people added to the code in the past year
- More contributed content and ideas
- Multi-echo users' meetings:
<https://www.youtube.com/@tedana-devs>
- New contributors in the past year



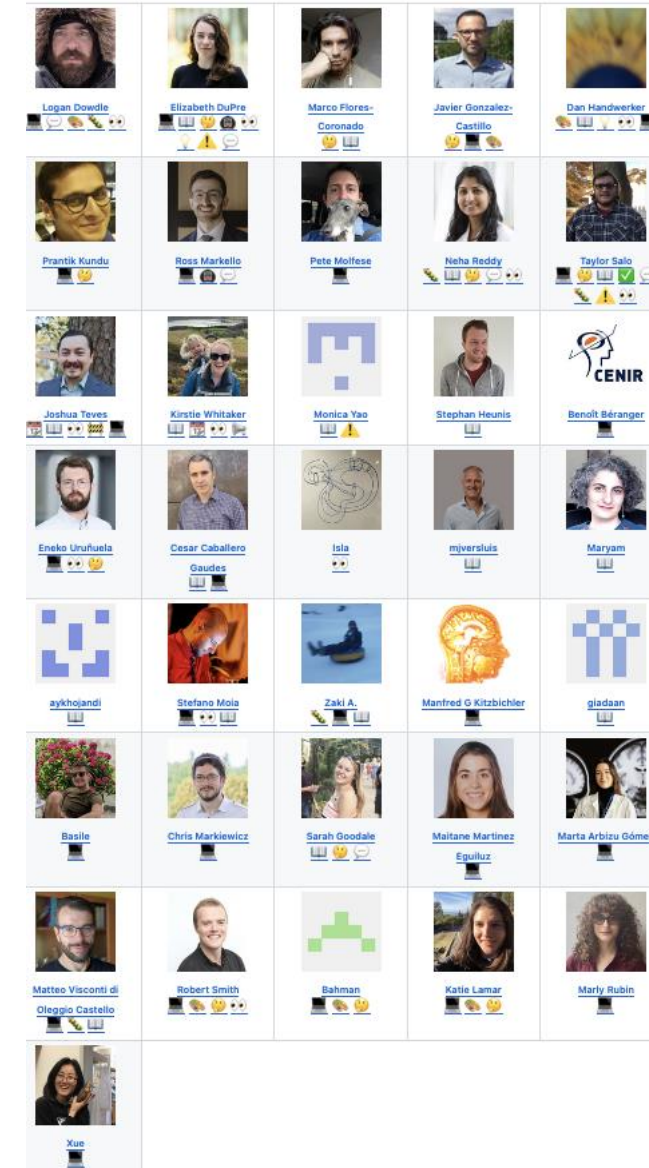
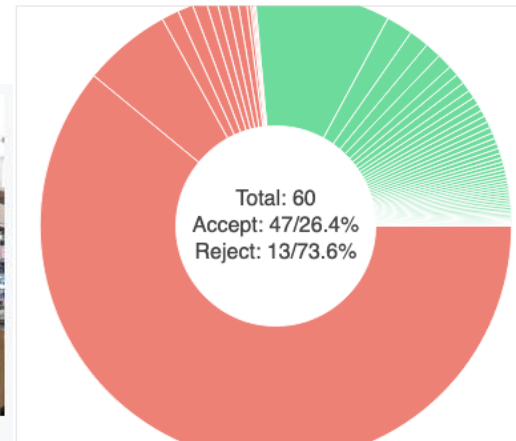
[Katie Lamar](#)



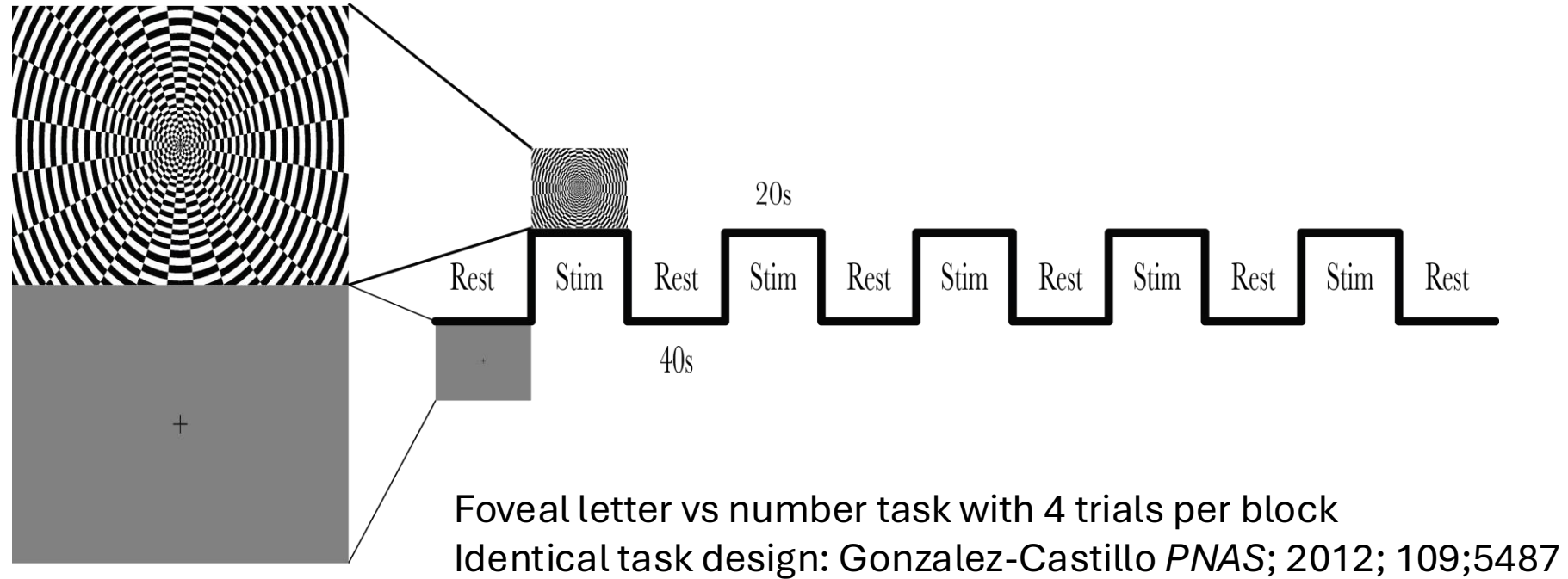
[Marly Rubin](#)



[Xue](#)

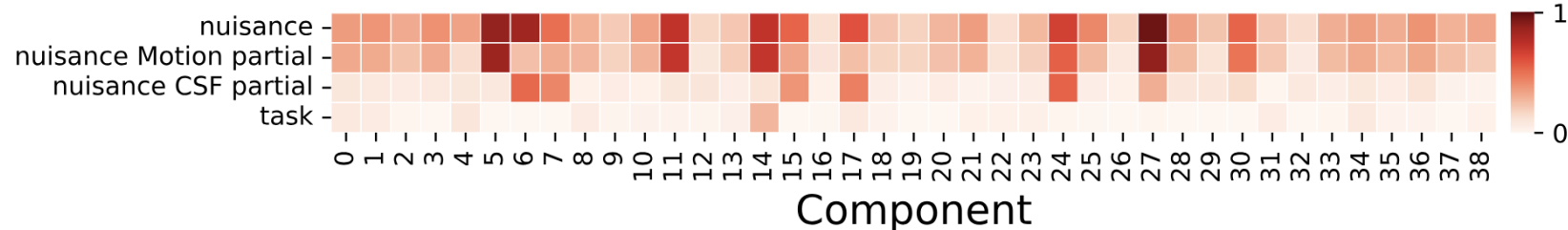
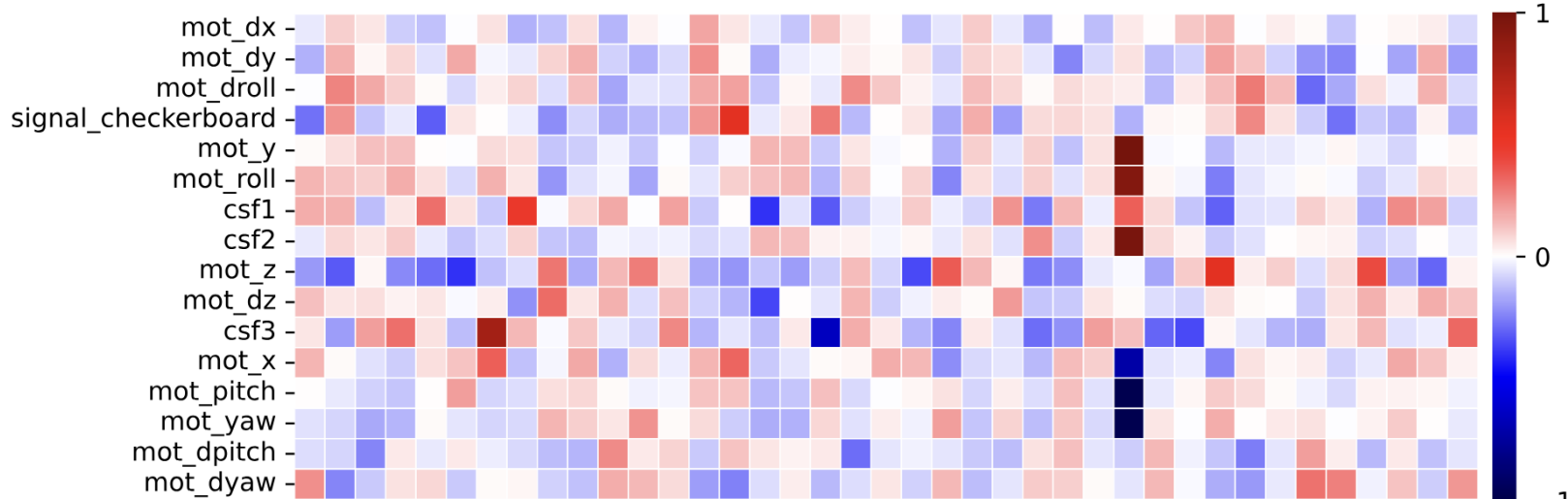


Block design checkboard + attention task

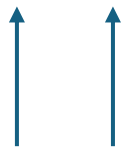


- GE MR750, 3T, 32 channel coil
- EPI: 3.5mm³, **5 echoes**, **TE=15.4, 29.7, 44.0, 58.3, 72.6ms**
FA=75°, TR = 2s,
- 5.5 minutes, 161 volumes

External Regressors

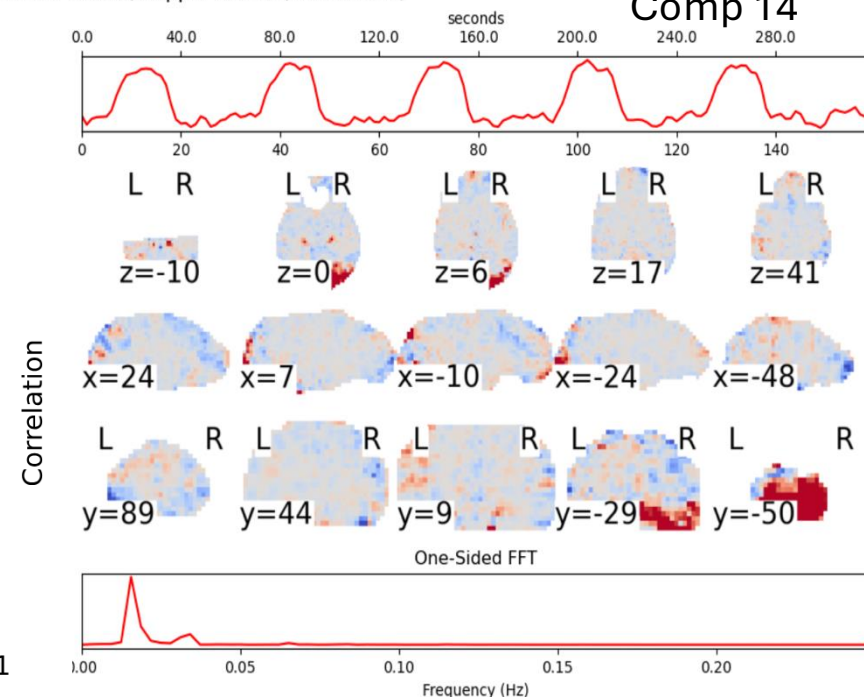


Component



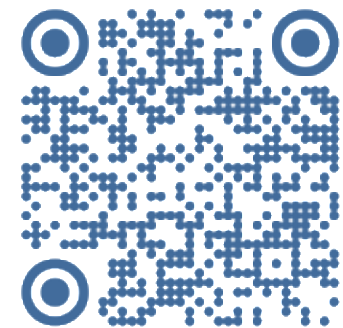
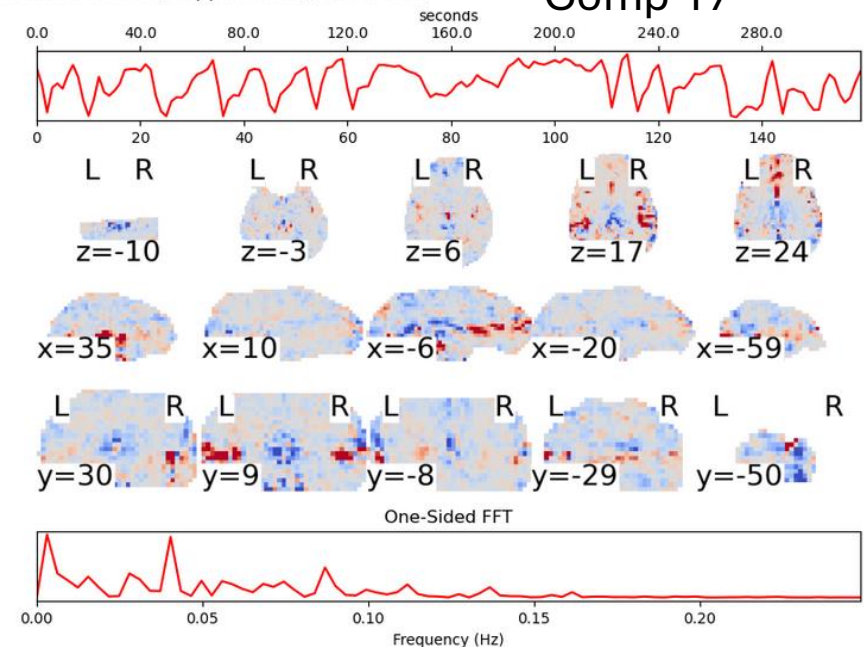
variance: 9.46%, kappa: 182.42, rho: 28.70,

Comp 14



variance: 0.81%, kappa: 25.62, rho: 30.96,

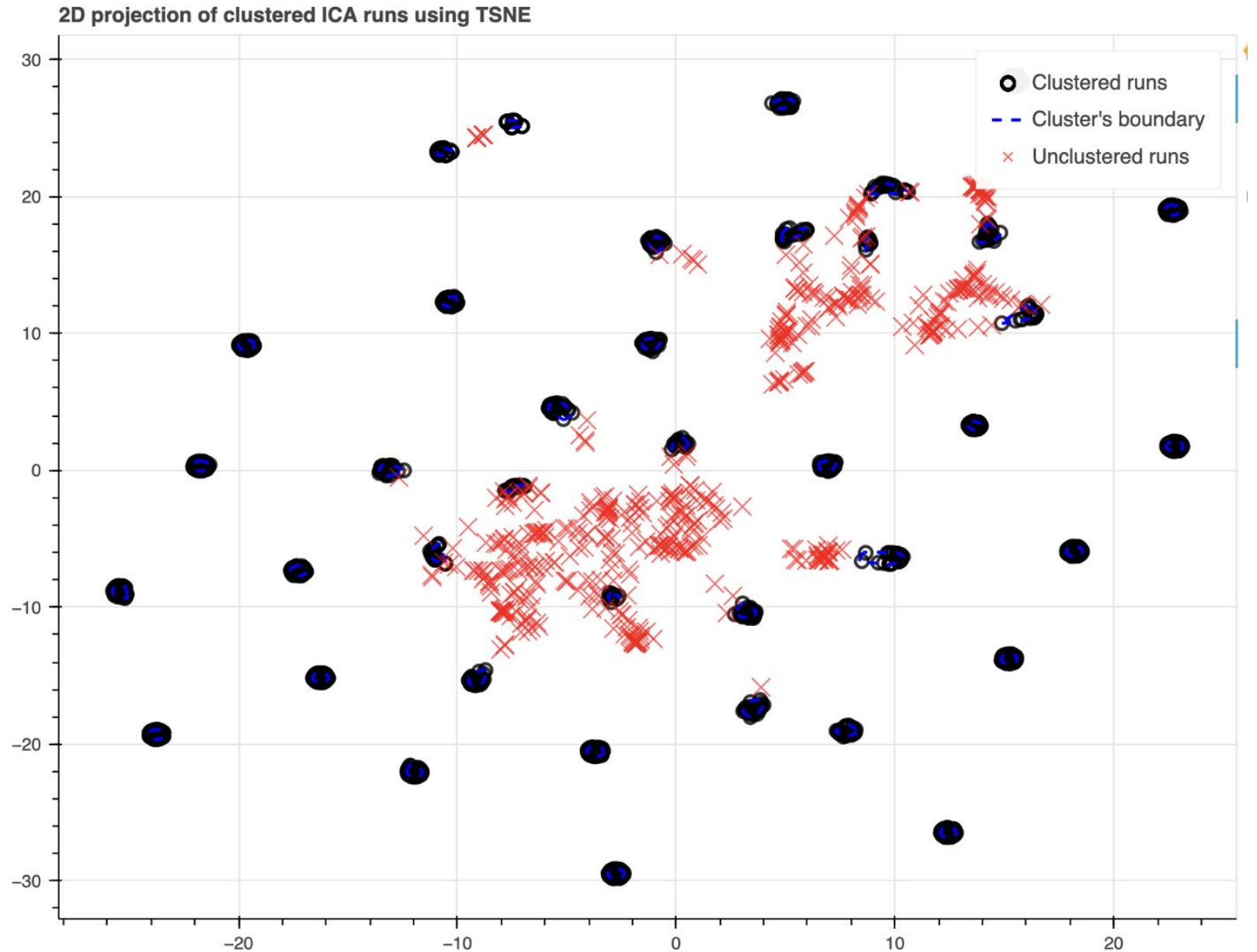
Comp 17



https://me-ica.github.io/ohbm-2025-multiecho/tedana_external_regress_processed/tedana_report.html

Robust ICA

Added by Australian
Epilepsy Project team



Better access to RICA interactive visualization

Info

ICA

Carpets

QC

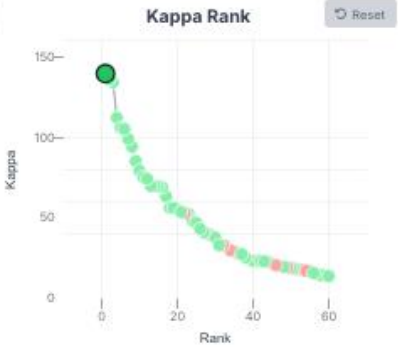
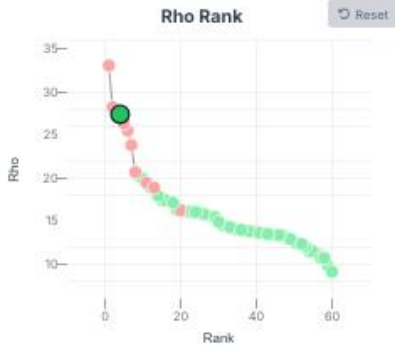
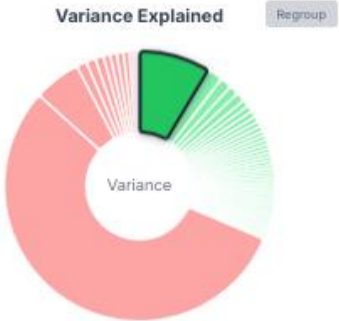
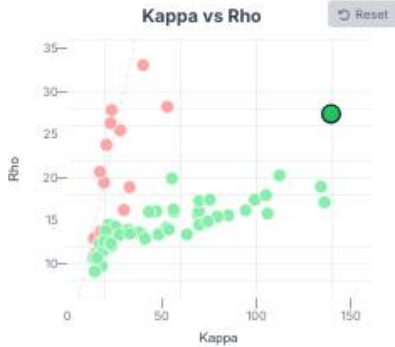
Accepted

Rejected

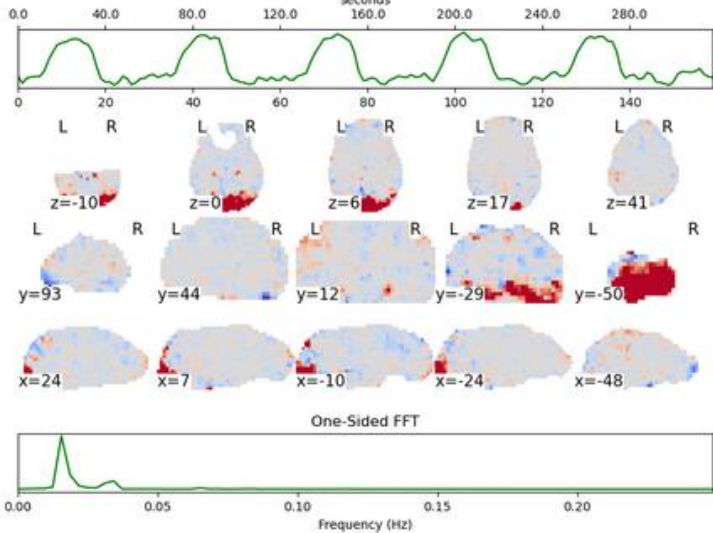
Reset

Save

Click to select. Scroll to zoom. Double-click to zoom in. Use A/R keys to classify.



Comp. 30: variance: 9.34%, kappa: 139.49, rho: 27.45, accepted reason(s): Likely BOLD



Hide

Component Metrics (60 components)

Component	Kappa	Rho	Variance %	κ Rank	ρ Rank	Dice FT2	Dice FS0	S/N t	Classification	Tags
ICA 28	21.80	14.54	0.18	45.00	32.00	0.57	0.30	10.17	accepted	Likely BOLD
ICA 29	69.84	14.55	0.19	14.00	31.00	0.63	0.52	16.48	accepted	Likely BOLD
ICA 30	139.49	27.45	9.34	1.00	4.00	0.72	0.63	6.59	accepted	Likely BOLD
ICA 31	69.86	17.32	0.42	13.00	17.00	0.76	0.66	19.88	accepted	Likely BOLD

More tedana additions

- Better memory management: Only storing voxels within masks
- Better parallelization for some processing steps
- Added more metrics for understanding each component
- Can designate dummy scans within tedana
- Input user-specified degrees of freedom for EPITl data
- Custom decision trees can be updated to figshare and automatically integrated into tedana
- Revived work on interactive book:
<https://me-ica.github.io/multi-echo-data-analysis>
- Tracking openly shared multi-echo data
<https://me-ica.github.io/open-multi-echo-data>
- Everything: <https://github.com/ME-ICA/tedana/releases>

How many components for ICA?

(Dimensionality estimation)



Marlene Smith

- Tedana has several well-established options for estimating the number of components
- They sometimes fail
 - If you have 200 timepoints, 10 or 199 components are implausible
 - 199 isn't ICA anymore and your brain is doing more than 10 things at once
- Want to understand why and ideally find a method to estimate the “right” number of components
- ICA is supposed to model the number of non-Gaussian statistically independent components + 1
- Many methods assume data are independent and identically distributed (i.i.d.). FMRI data are not, but there are existing methods that are supposed to deal with this... Easy to solve?

Audio vs Visual + Word vs NonWord Task



bright

crowd

bolt

bolt

4 sec trials



bright

crowd

bolt

bolt

Press button if a stim repeats in a trial

Jittered event related design

25 minutes of data over 3 runs

340-345 volumes/run

3T Siemens Prisma

TE=13.44, 31.7, & 49.96ms

TR=1.5s, SMS=2 iPAT=2

3.0 mm³ isotropic voxels.

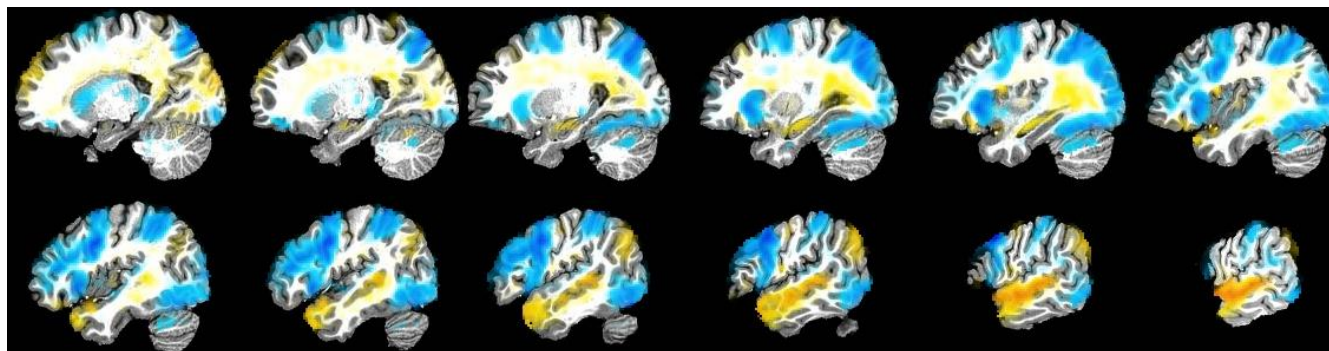
25 volunteers



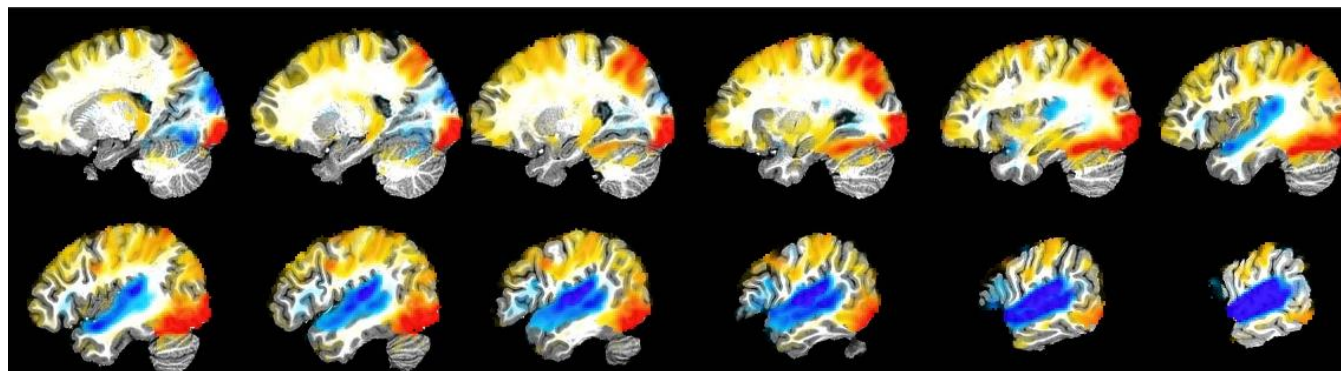
Vocoded words

Group contrasts

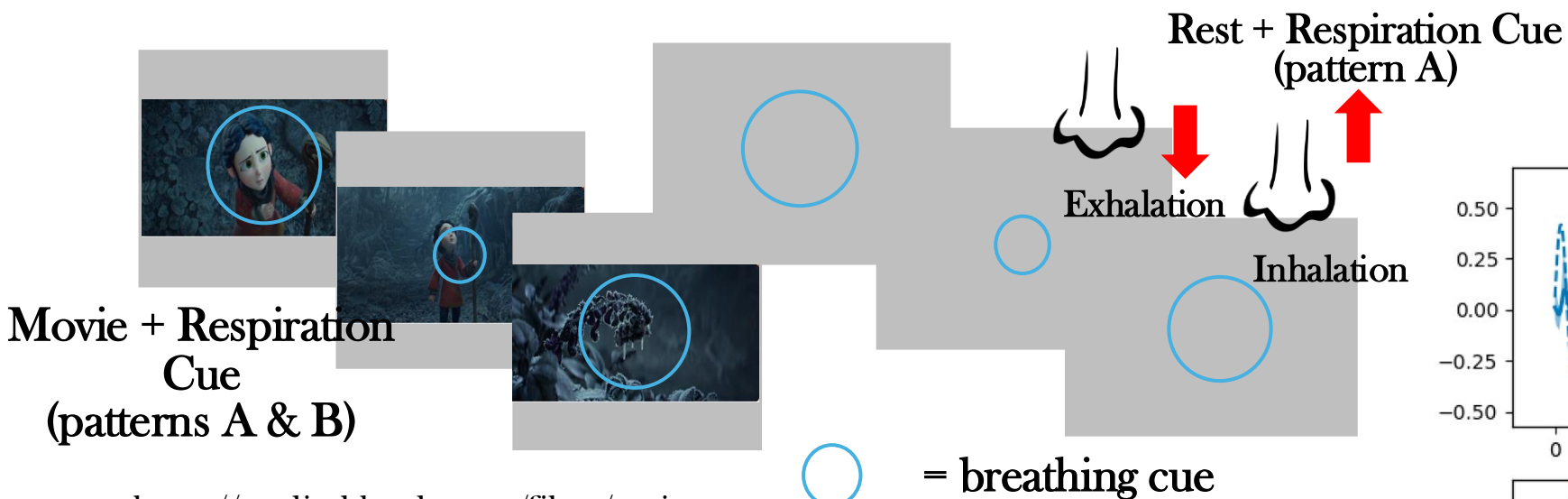
Word - Nonword



Visual - Audio



Paced Breathing and Movie Task



<https://studio.blender.org/films/spring>

3T Siemens Prisma

TR=1.5s

TE=13.44, 31.7, & 49.96ms

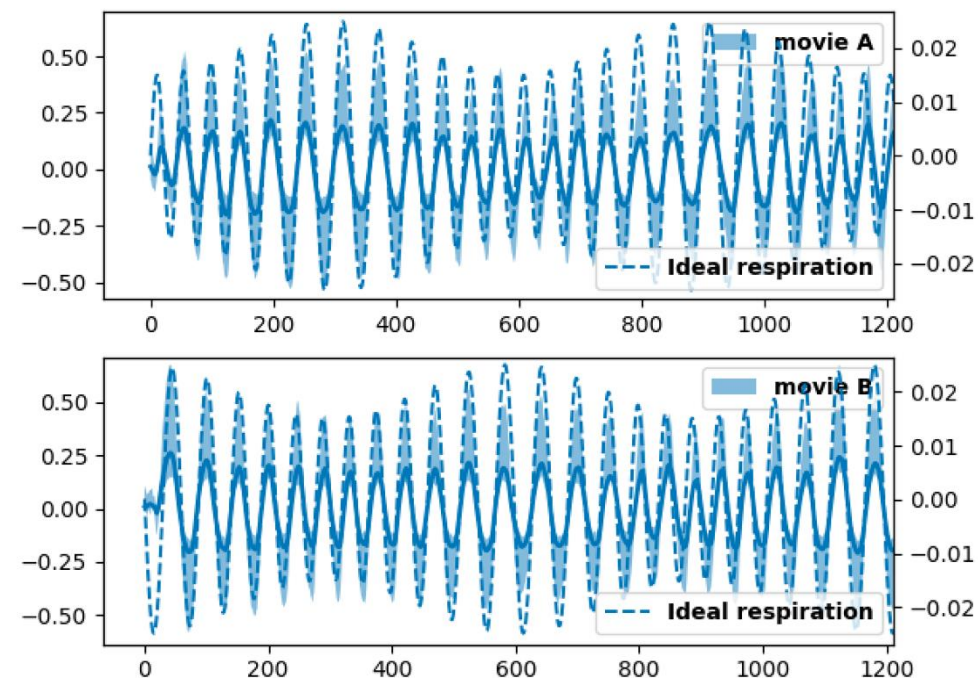
SMS=2 iPAT=2

3.0 mm³ isotropic voxels.

25 volunteers

299 volumes per run

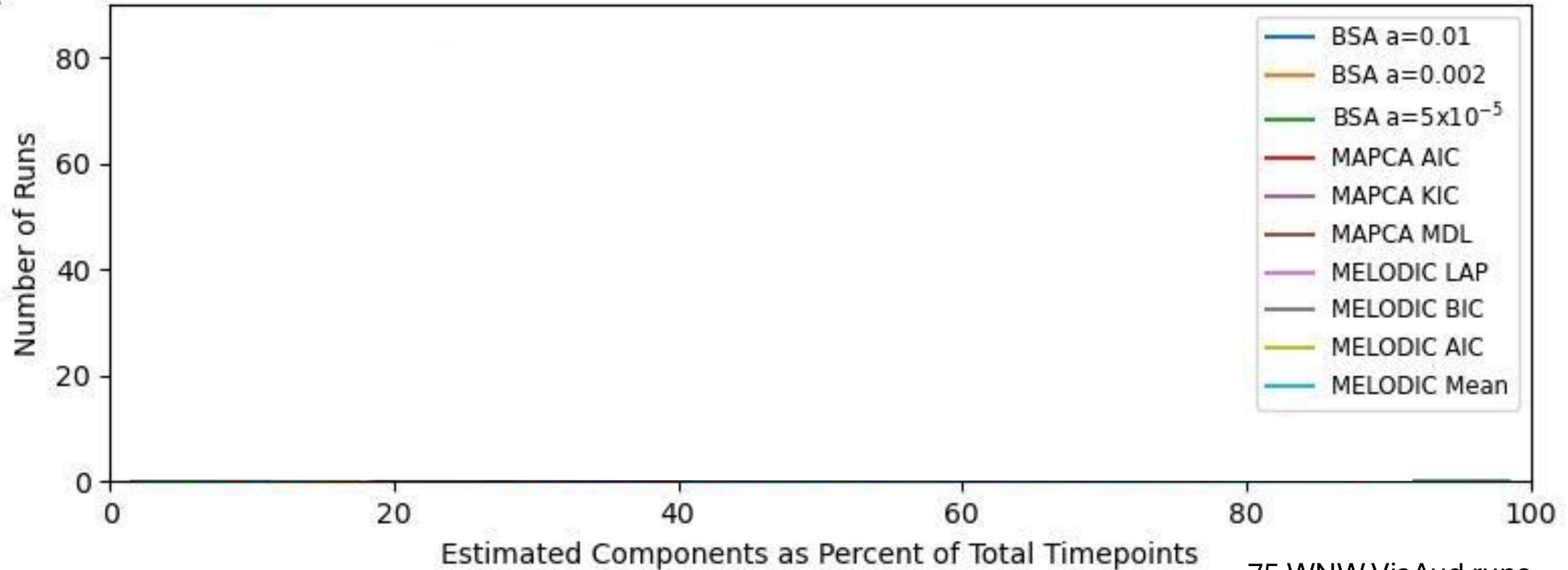
Breathing depth & rate



Comparing Methods for Component Number Estimation

A

Distributions of Component Estimate Methods



75 WNW VisAud runs

99 Breathing / Movie runs

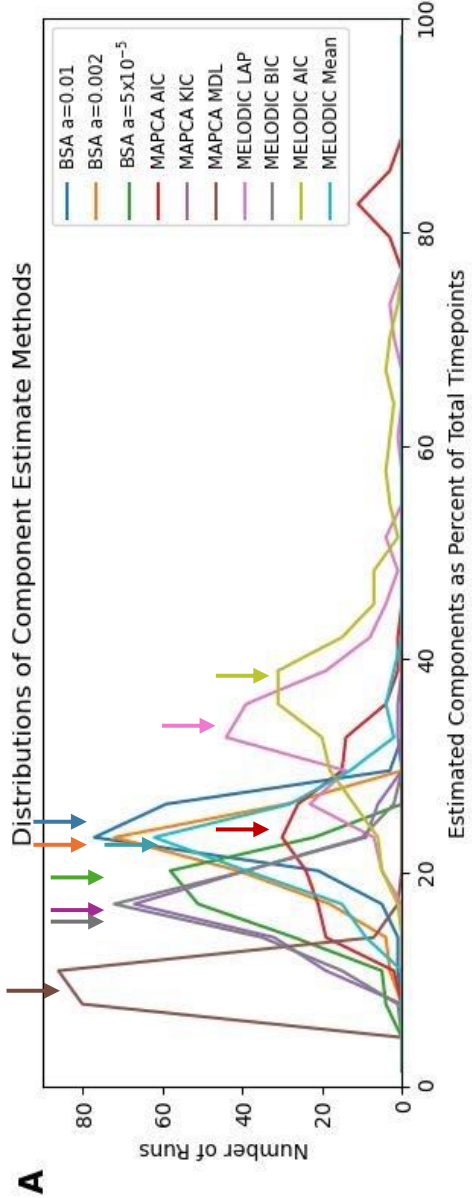
MAPCA = Used by tedana and a port of the methods in GIFT

MELODIC = Used by MELODIC in FSL. Some similar, some use Bayesian tools

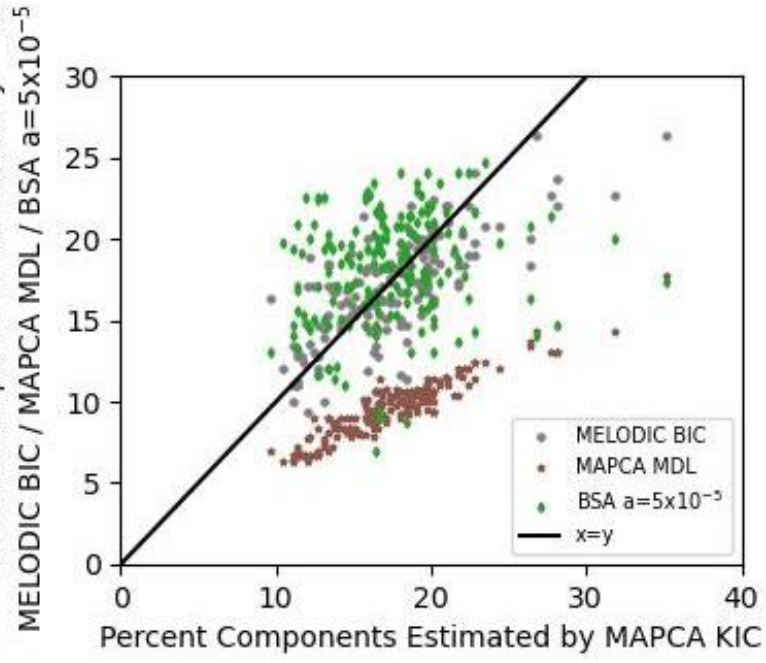
BSA = Bootstrap Analysis (Majeed & Avison, PLoS ONE 2014) + coded & enhanced by Ka Chun Lam

Run-by-run comparisons of dimensionality estimates

Comparing Methods for Component Number Estimation



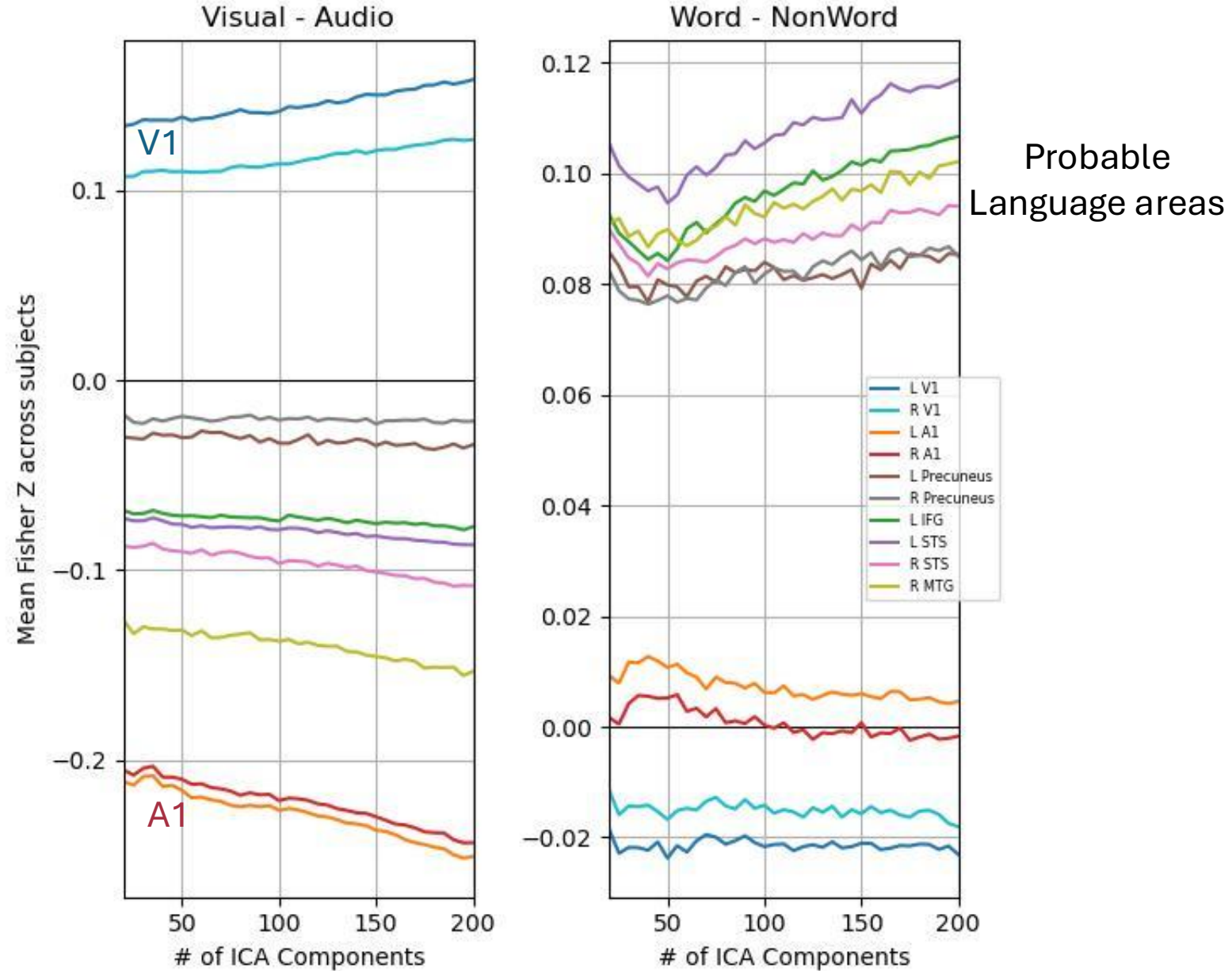
B Percent Components Estimated by



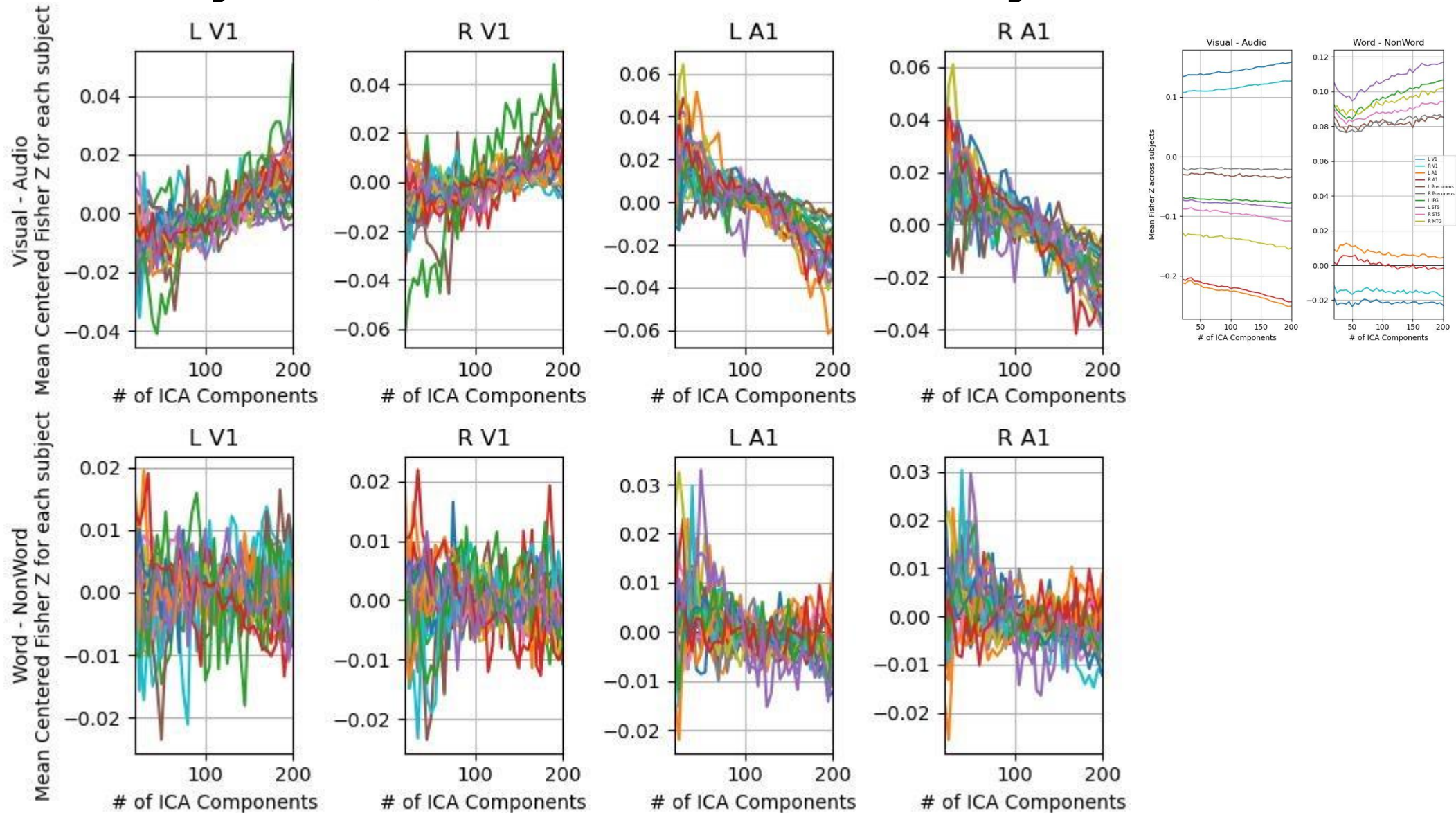
Many dimensionality estimations disagree

- We do not have a consensus estimate across methods
- Is one of these methods more accurate than others?
- What is accurate?
- Proxy measure: A good dimensionality estimate should result in better characterization of the data and better denoising
 - Using Word-Nonword, Vis-Aud task (340 volumes/run)
 - Run tedana with 20-200 ICA components pre-specified
 - Calculate $R^2 \rightarrow r \rightarrow \text{Fisher } Z$
 - Subject-specific anatomically+functionally localized ROIs for each contrast
 - Contrast-to-Noise measure: Average Fisher Z values for each subject and ROI
- What is the correct number of ICA components?

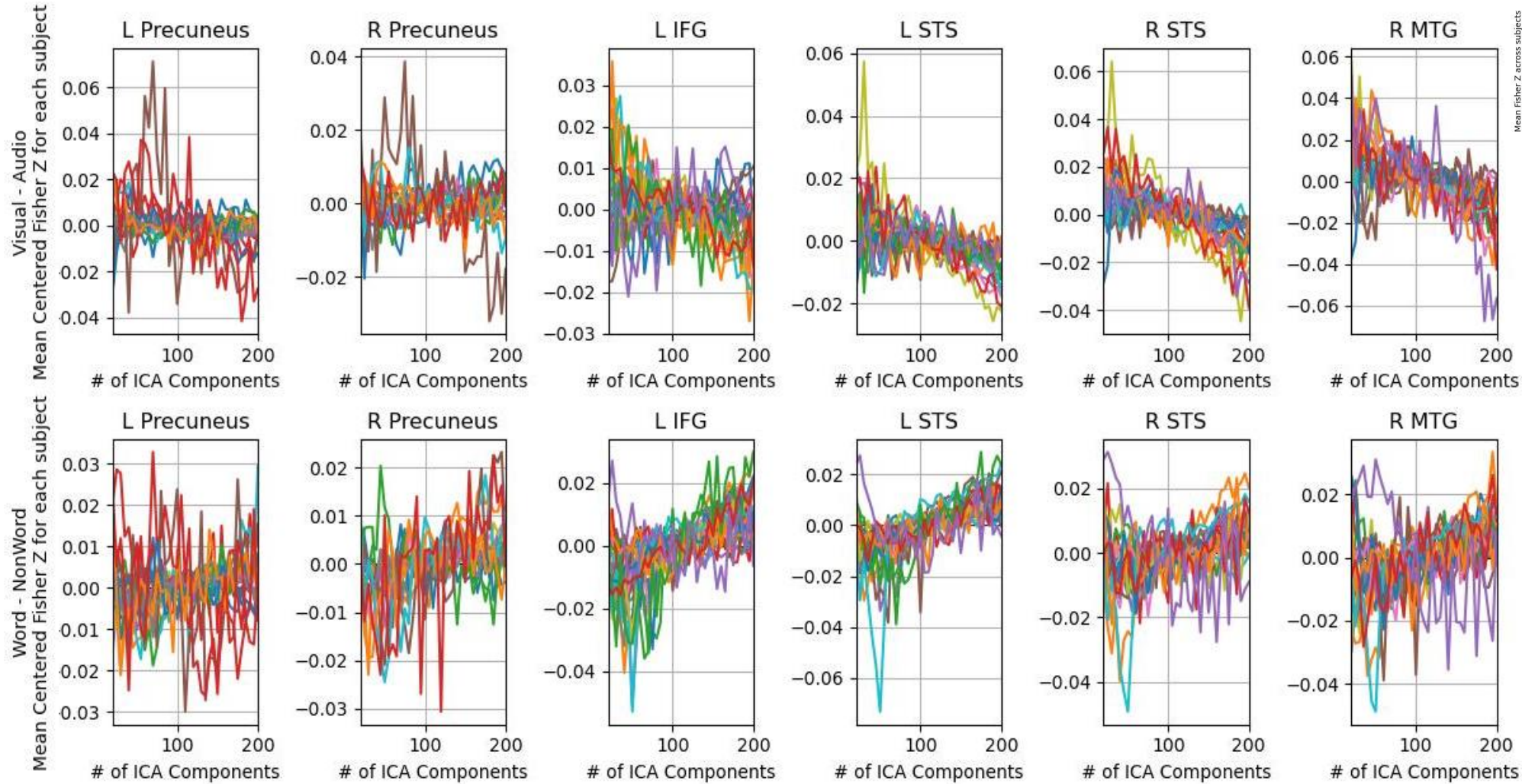
Contrast to noise vs # of ICA components



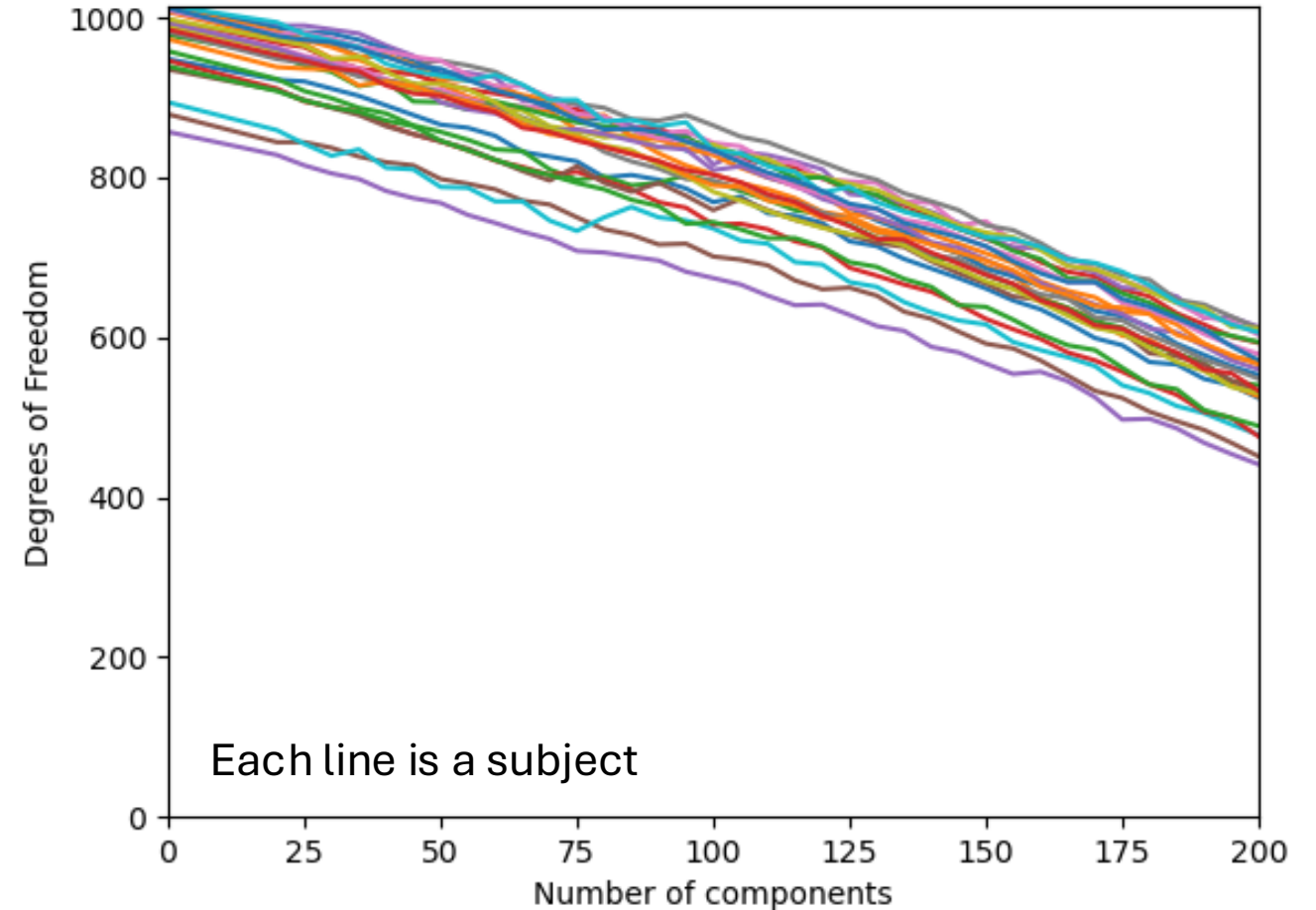
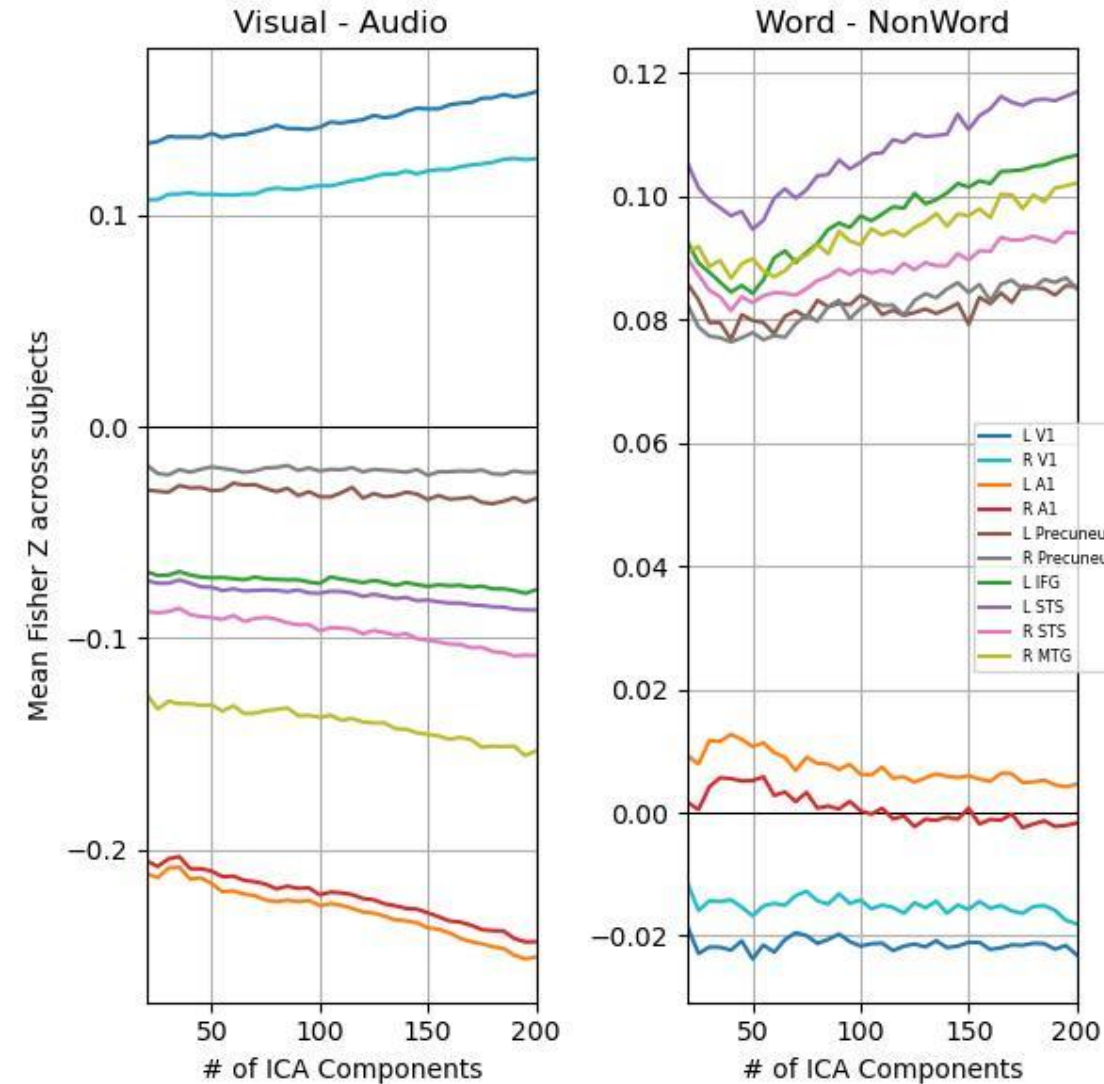
All subjects, Visual and Auditory ROIs



All subjects, language ROIs



Degrees of freedom decrease with more components



Current rec for selecting # of ICA components

- This choice will affect results, but don't stress whether 50, 60, or 70, is more correct
- Pick something that makes sense across your dataset and runs in a sane amount of time
- ICA is “intelligent” signal averaging
- What is better than ICA?

LBC Questions

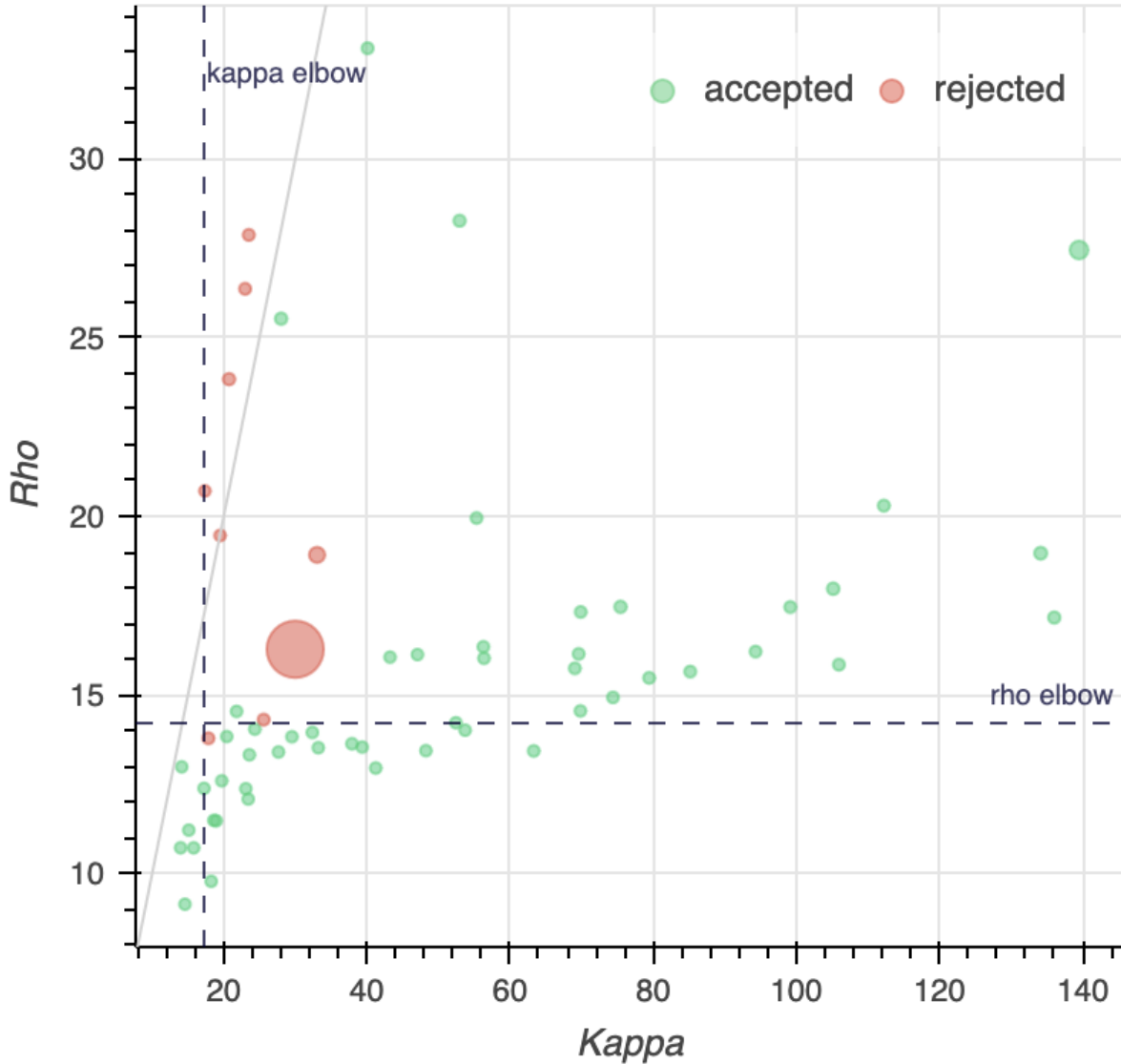


Where to “cut” for good vs bad?

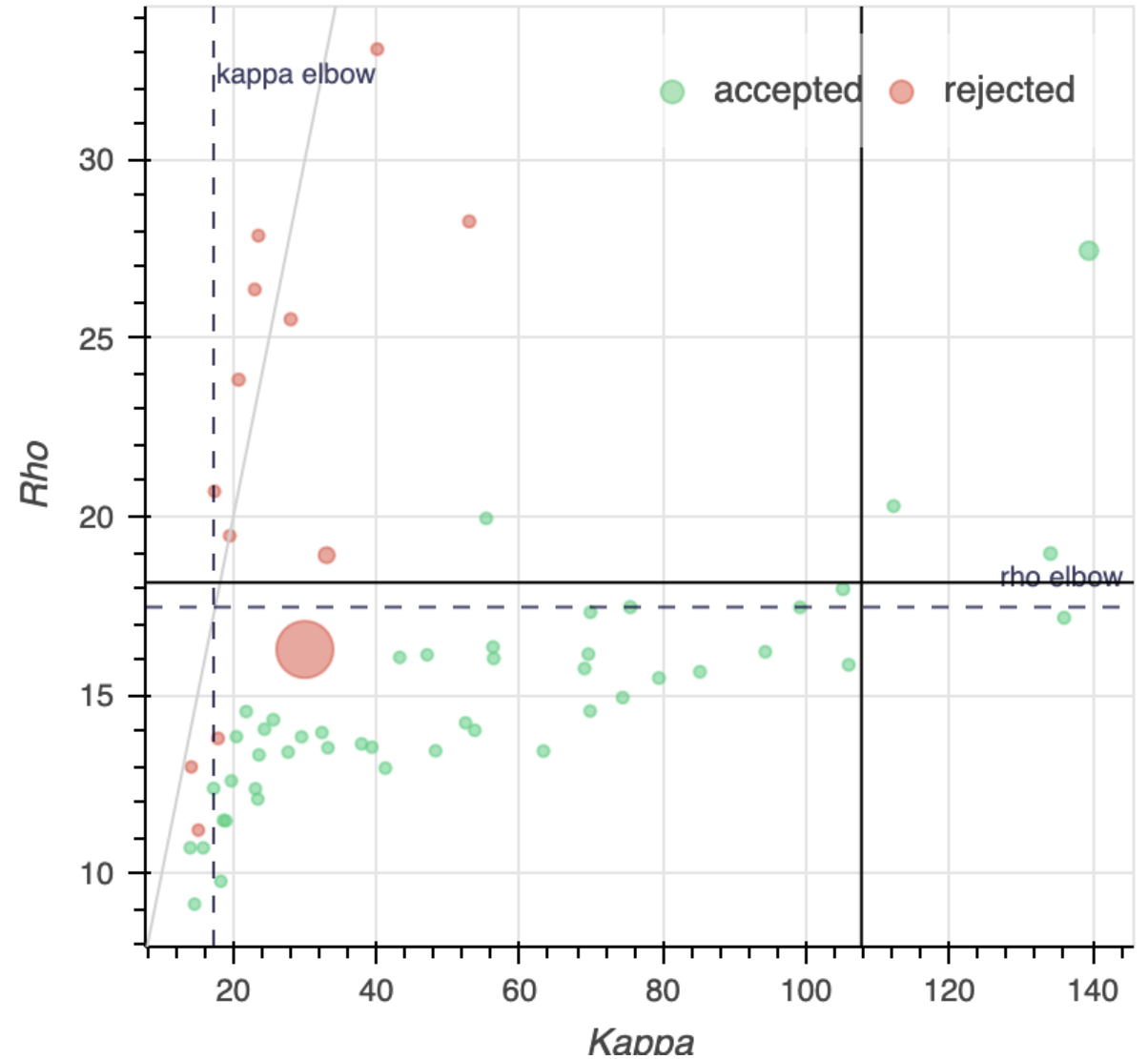
Original MEICA “decision tree”

Tedana “decision tree”

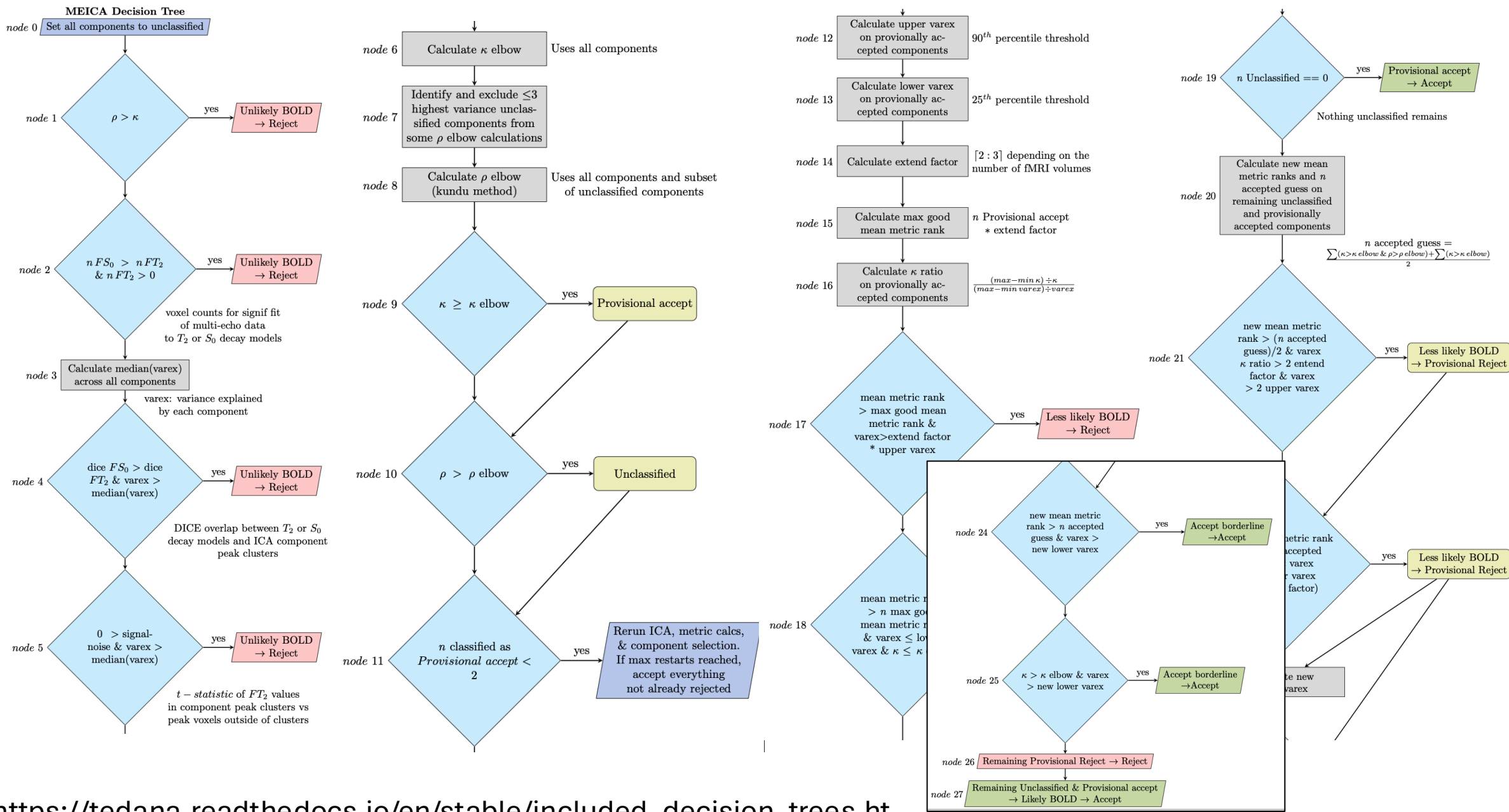
Kappa / Rho Plot



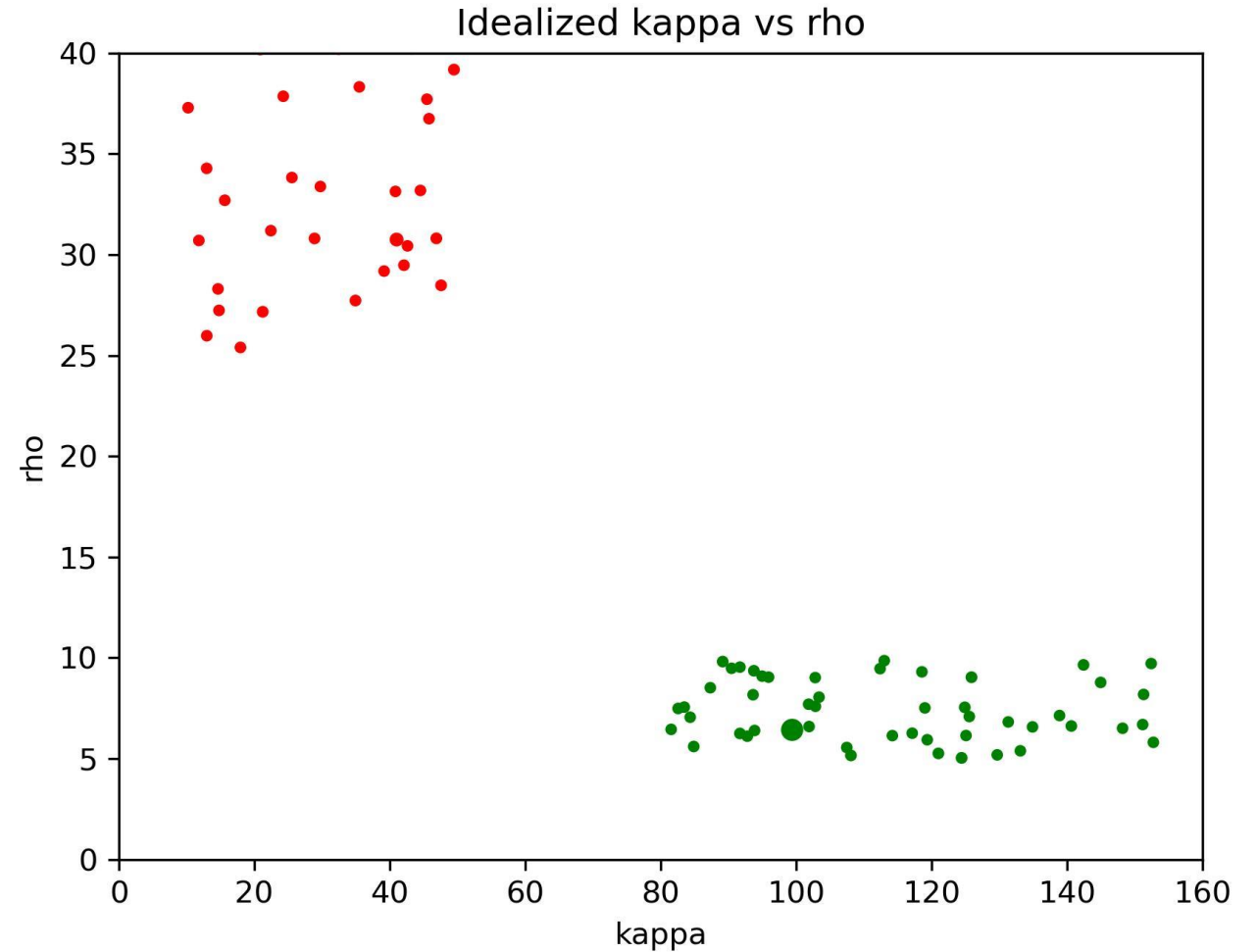
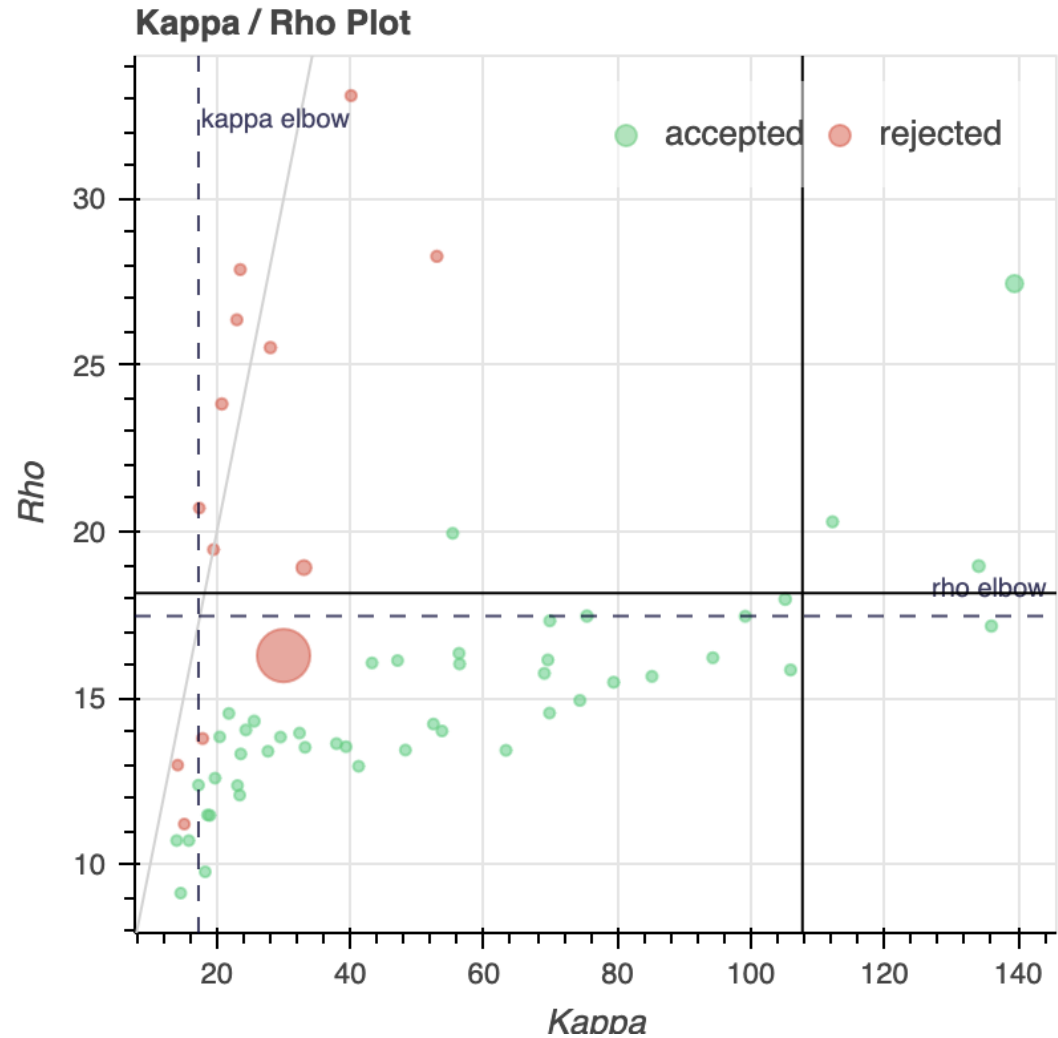
Kappa / Rho Plot



The meica decision process is... quirky

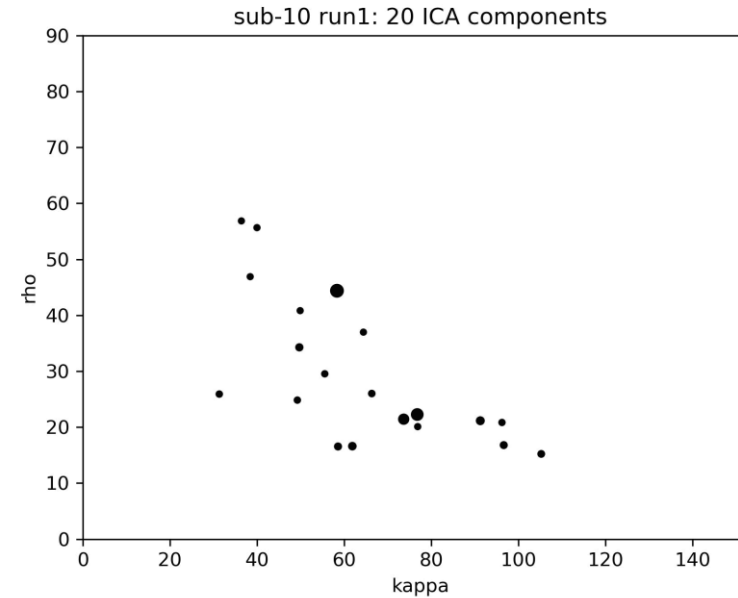
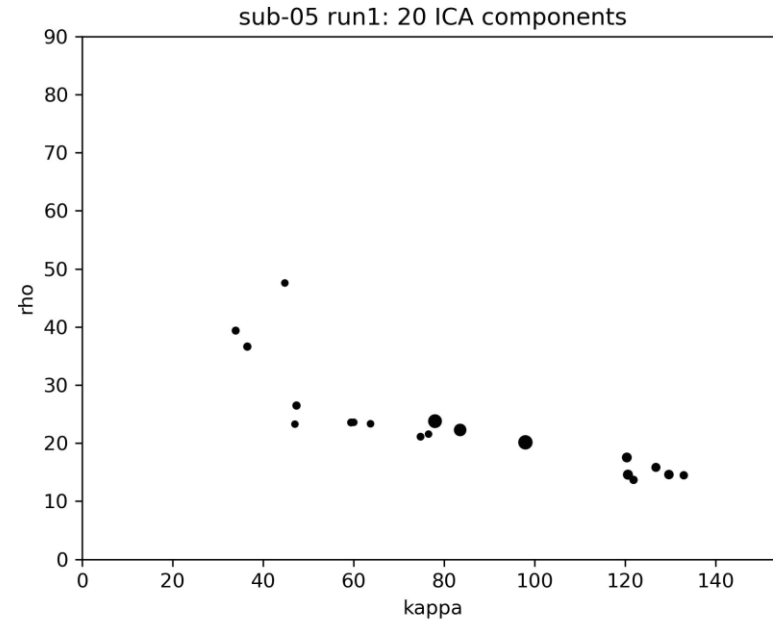


Why is the decision process unclear?

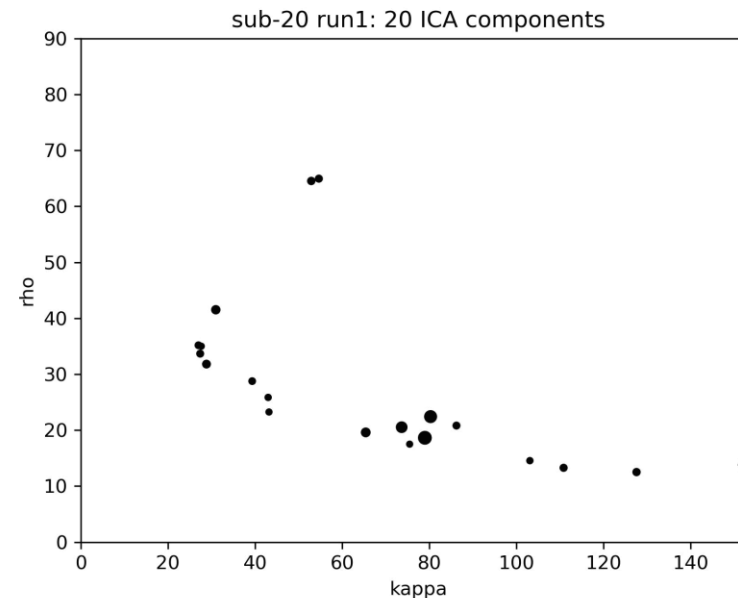
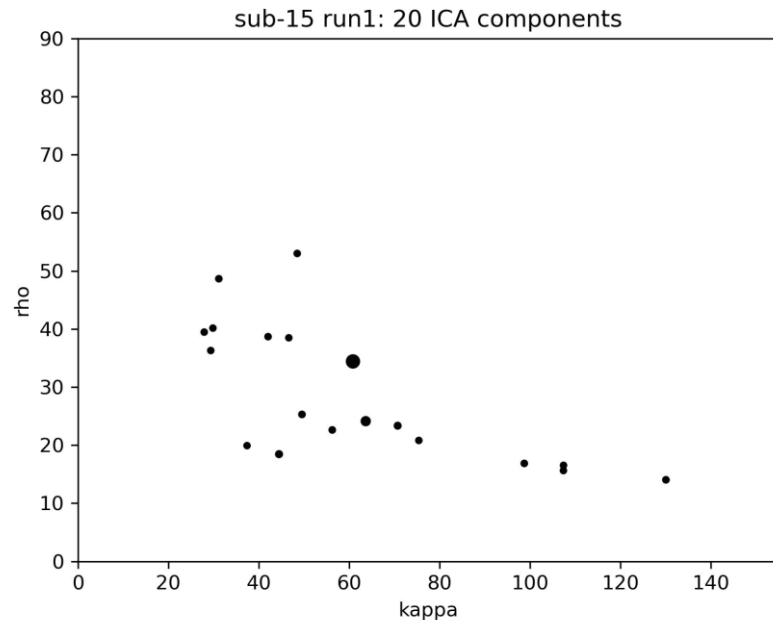


Everything far from the extremes is not only imprecise to classify,
but is inherently a mix of signal we want and don't want

ICA is signal averaging, but are there better options?



Making the # of components study a bit more interesting



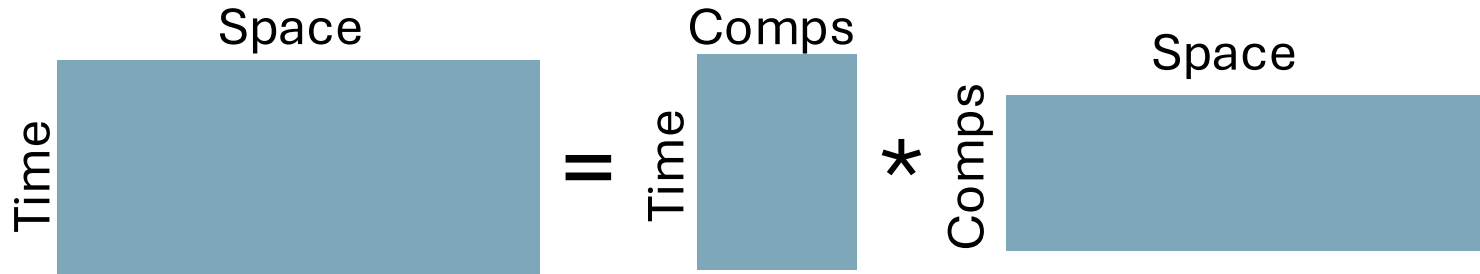
More components does bring more to extremes, but also more to closer to (0,0)

Other methods

The **VERY** preliminary & exploratory part of this talk

- Overall goal: A method that instead or in addition to optimizing for spatial statistical independence, optimizes for T2* vs S0 (kappa vs rho) separation
- **Tensorial ICA (tICA)**
- **Dynamic Mode Decomposition (DMD)**
- Iterative ICA to fit solutions with better separation
- Using T2* and S0 time series
- Machine Learning based approaches

Standard ICA

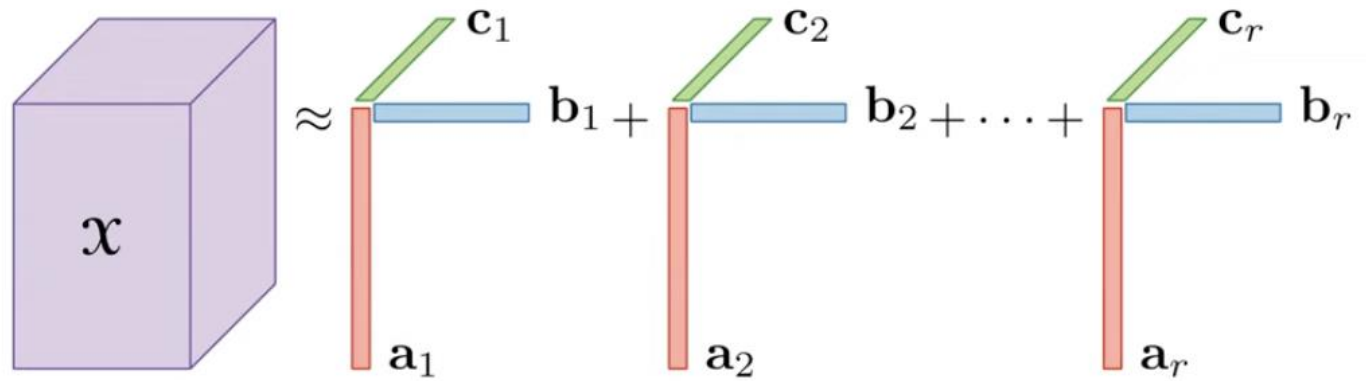


Tensorial ICA

(Generalized)

Canonical Polyadic (CP)
Decomposition

a.k.a.
CANDECOMP or
PARAFAC



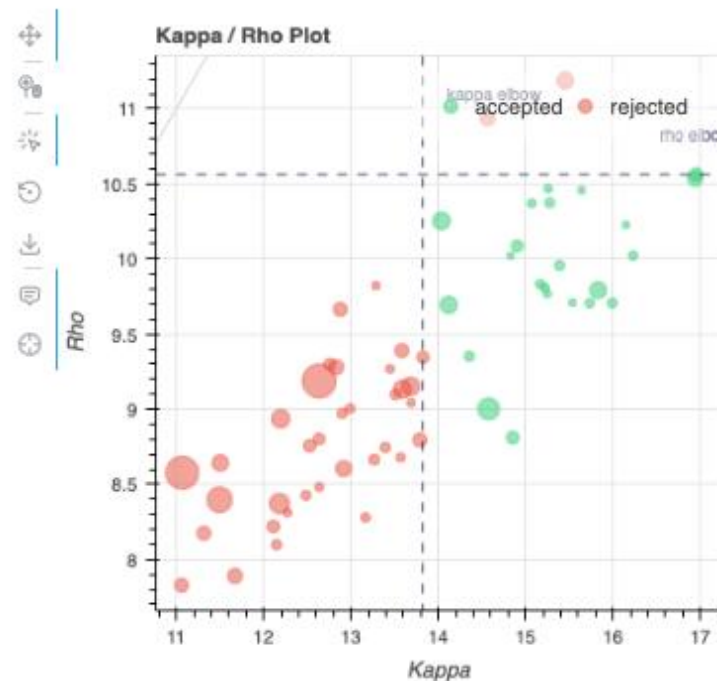
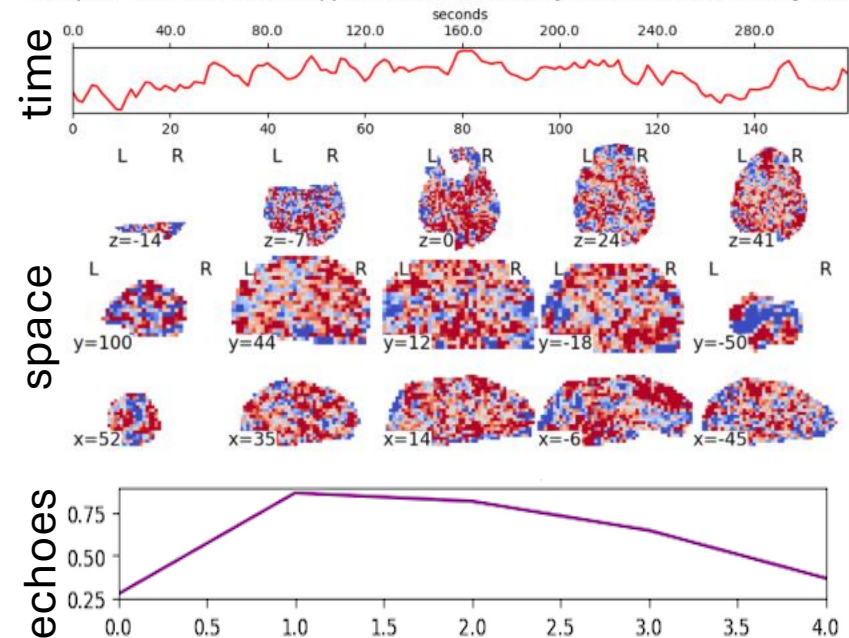
Potentially useful across echoes

Tensorial ICA

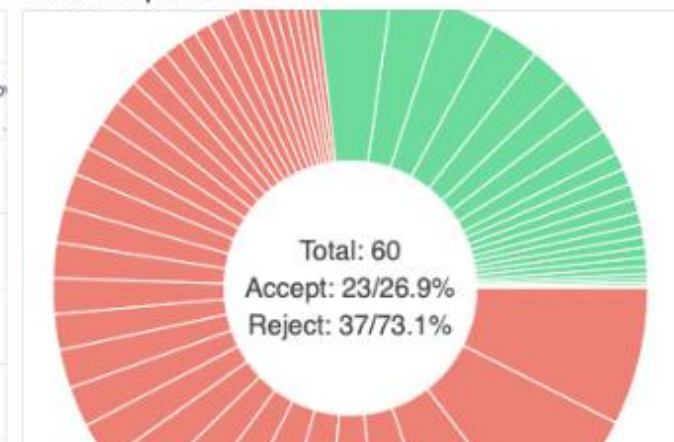
ICA components

One mode (component)

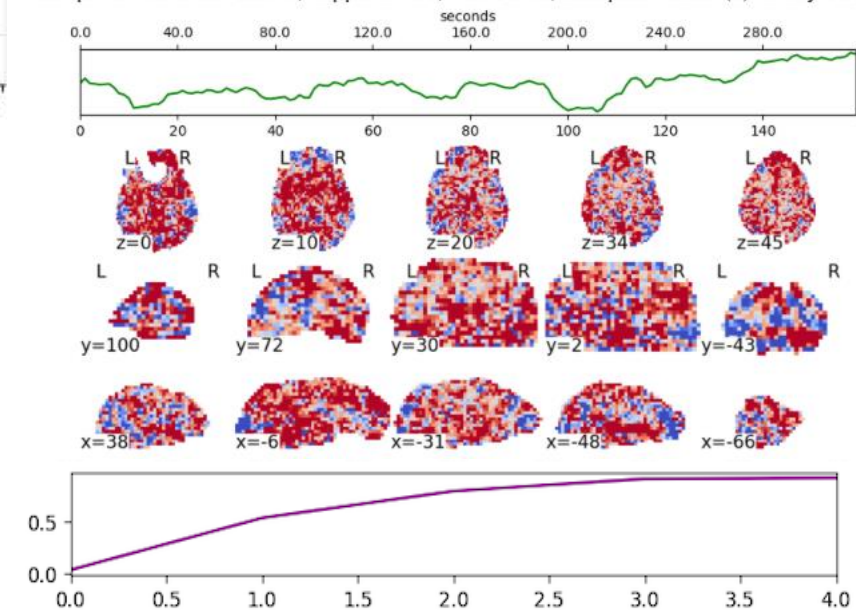
Comp. 0: variance: 0.88%, kappa: 13.40, rho: 8.75, rejected reason(s): Unlikely BOLD



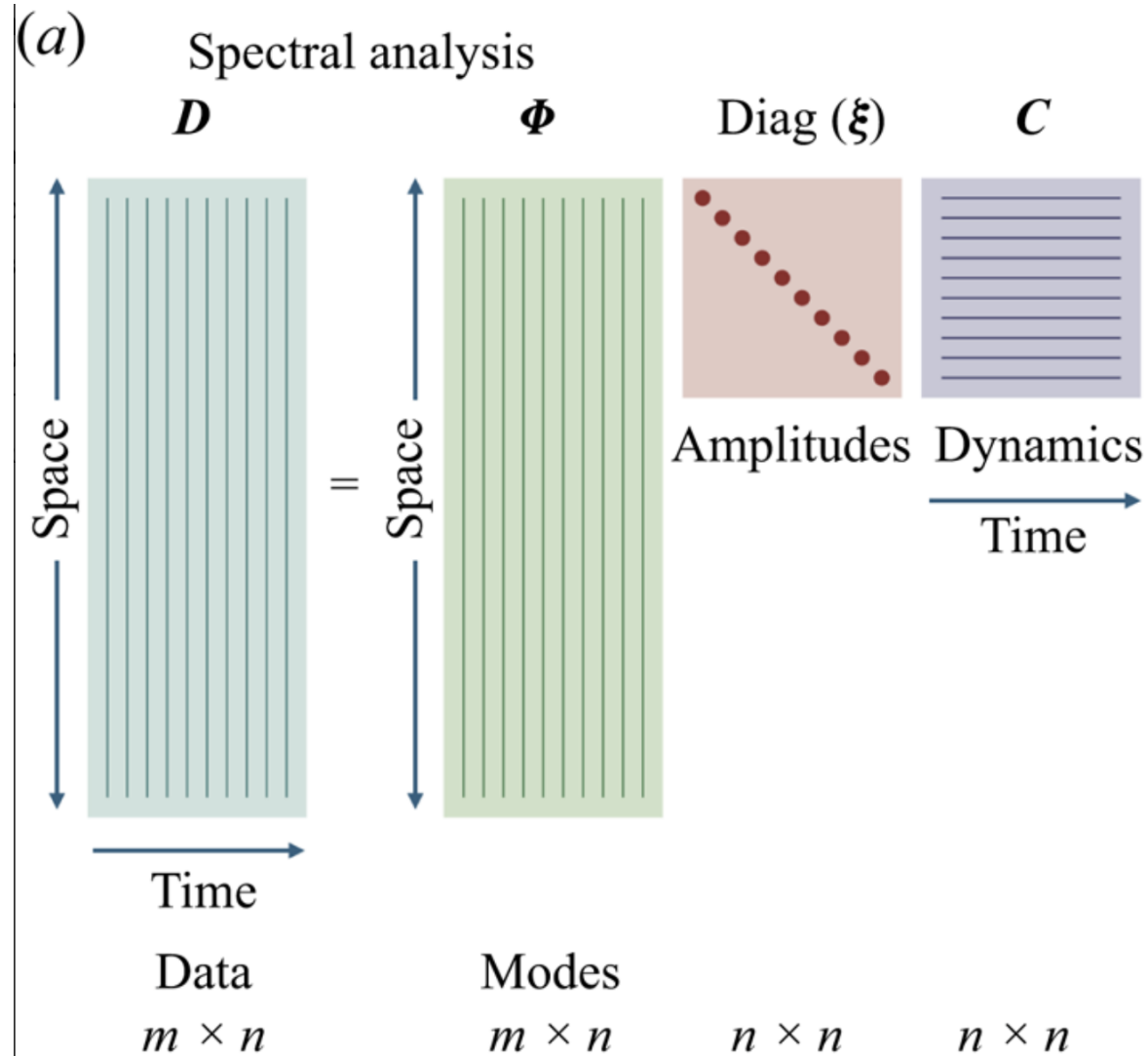
Variance Explained View



Comp. 45: variance: 1.39%, kappa: 14.90, rho: 10.09, accepted reason(s): Likely BOLD



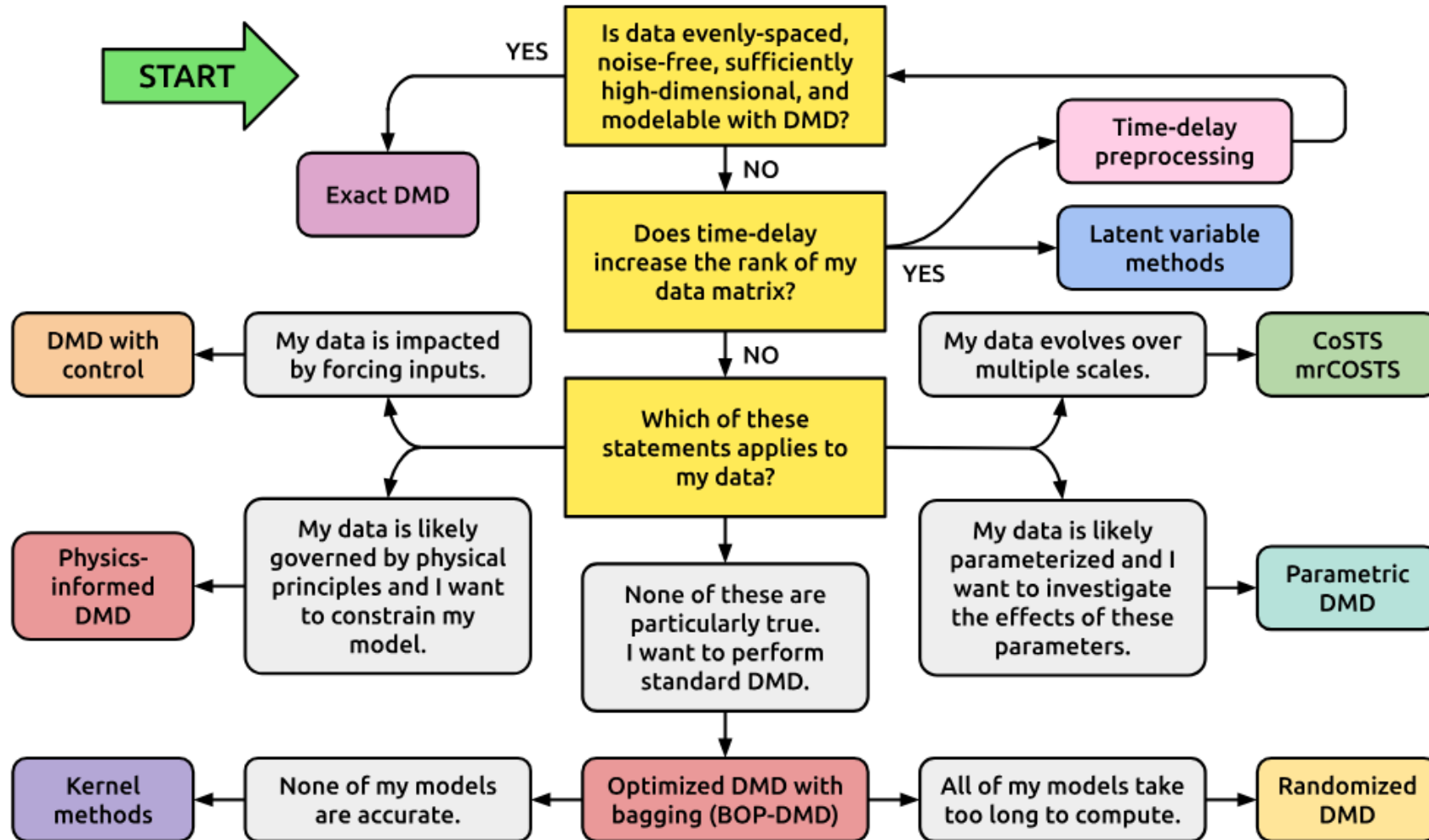
Dynamic Mode Decomposition



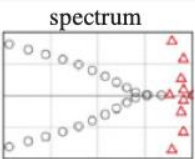
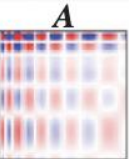
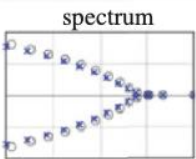
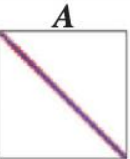
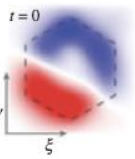
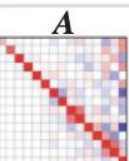
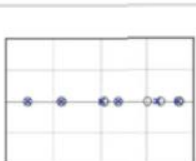
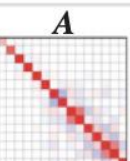
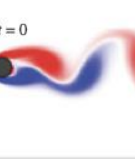
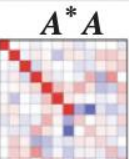
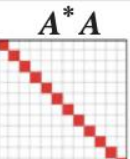
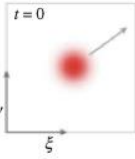
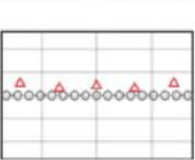
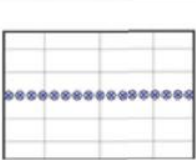
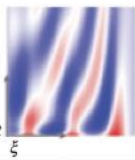
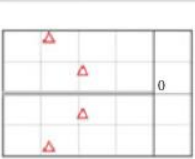
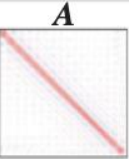
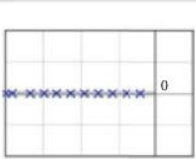
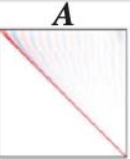
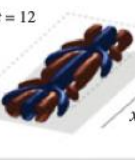
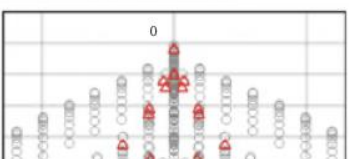
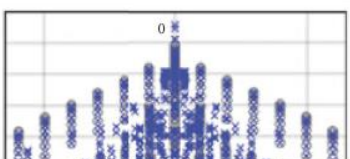
Dynamics is key, but how to define each can vary by data and goals

Also uses complex values which model phase and time delays

Many DMD options, but need to really understand application



Physics informed DMD

system	physics (matrix structure)	data	standard DMD	physics-informed DMD
advection-diffusion $\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial \xi^2} - v(\xi) \frac{\partial u}{\partial \xi}$	local (tri-diagonal)		spectrum  	spectrum  
Schrödinger equation $i\hbar \frac{d}{dt} \psi(t)\rangle = \hat{H} \psi(t)\rangle$	self-adjoint (Hermitian)		 	 
cylinder flow $\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} = \frac{1}{Re} \nabla^2 \mathbf{u} - \nabla p$	conservative (unitary)		 	 
two-dimensional advection $\frac{\partial u}{\partial t} = -c \cdot \nabla u$	shift-equivariant (block circulant)		 	 
Volterra integro-differential equation $\frac{\partial u}{\partial t} = \int_{-1}^{\xi} K(\xi, v) u(v, t) dv$	causal (upper triangular)		 	 
channel flow $\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} = \frac{1}{Re} \nabla^2 \mathbf{u} - \nabla p$	shift-equivariant (block circulant)			

Sets up data in and set up equations to conform in particular ways.

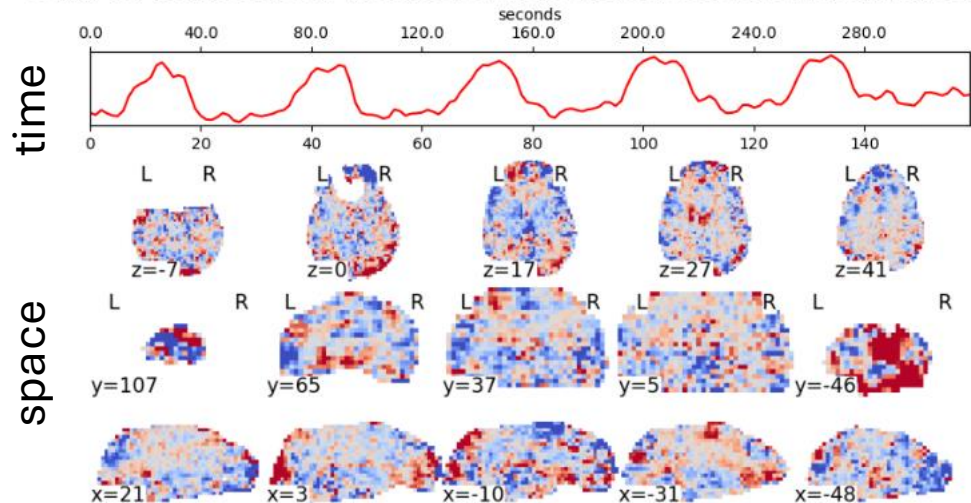
Can we set up the dynamics to be modeled on known relationships across echoes?

A test with Exact DMD

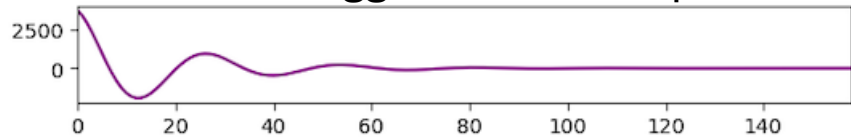
Modeling optimally combined data to same data one timepoint in the future

One mode

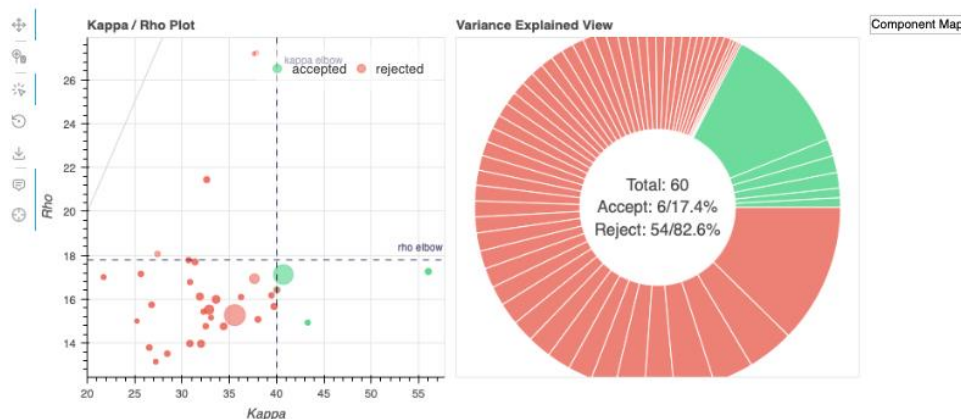
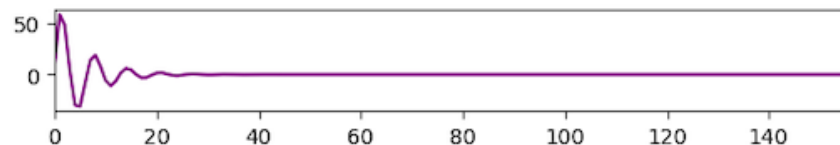
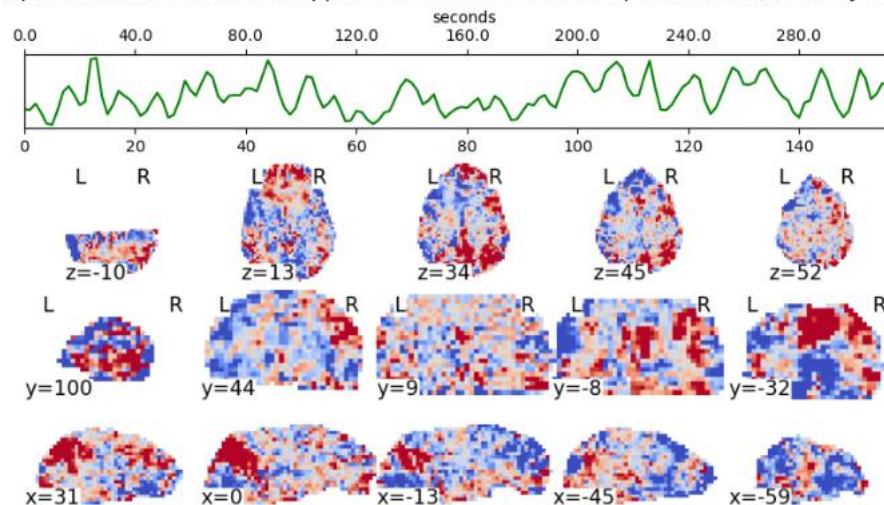
Comp. 45: variance: 0.82%, kappa: 30.85, rho: 16.78, rejected reason(s): Unlikely BOLD



Time lagged relationship



Comp. 17: variance: 1.49%, kappa: 56.03, rho: 17.26, accepted reason(s): Likely BOLD



Models some neurally related aspects of data, but currently worse kappa vs rho separation than ICA

Next steps

- Continuing to compile potential options for better methods
 - Try to understand the methods
 - Conduct tests of current and potential efficacy
- The properties of ICA give us nice pictures, but are statistically independent spatial maps really the best way to separate T2* and S0 signals.
- Potential for very different separation or clustering approaches?

Acknowledgements

- Marlene Smith
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 - Neha Reddy
 - Bahman Tahorini
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 - Ben Gutierrez

LBC Questions

