

## INTRODUCTION

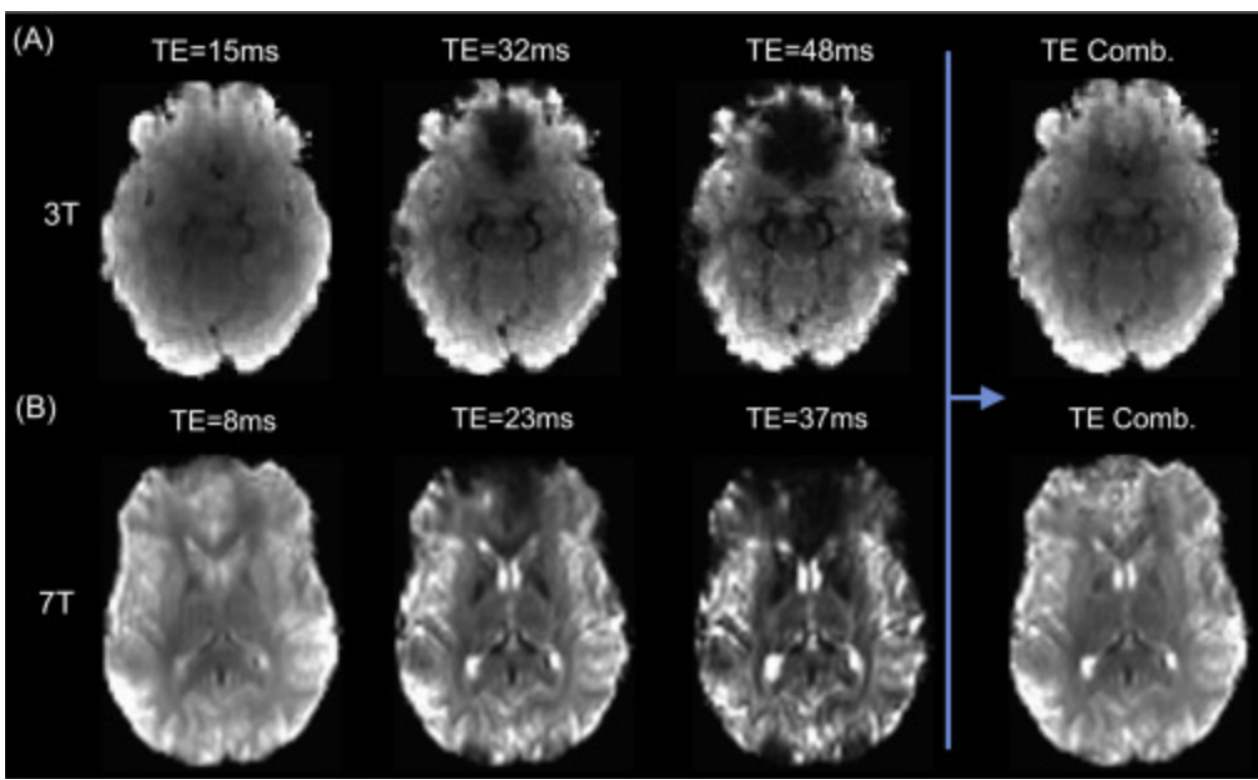
To access an audio recording and transcript describing this poster, [click here](https://github.com/ramyav97/multi-echo-fMRI-OHBM-2020.git):  
<https://github.com/ramyav97/multi-echo-fMRI-OHBM-2020.git>

- Validation of multi-echo data analysis methods benefits from a dataset with predictable signal changes across multiple brain regions and acquisition options.
- We are collecting such a dataset to help validate and compare existing and new changes to processing methods.

## METHODS

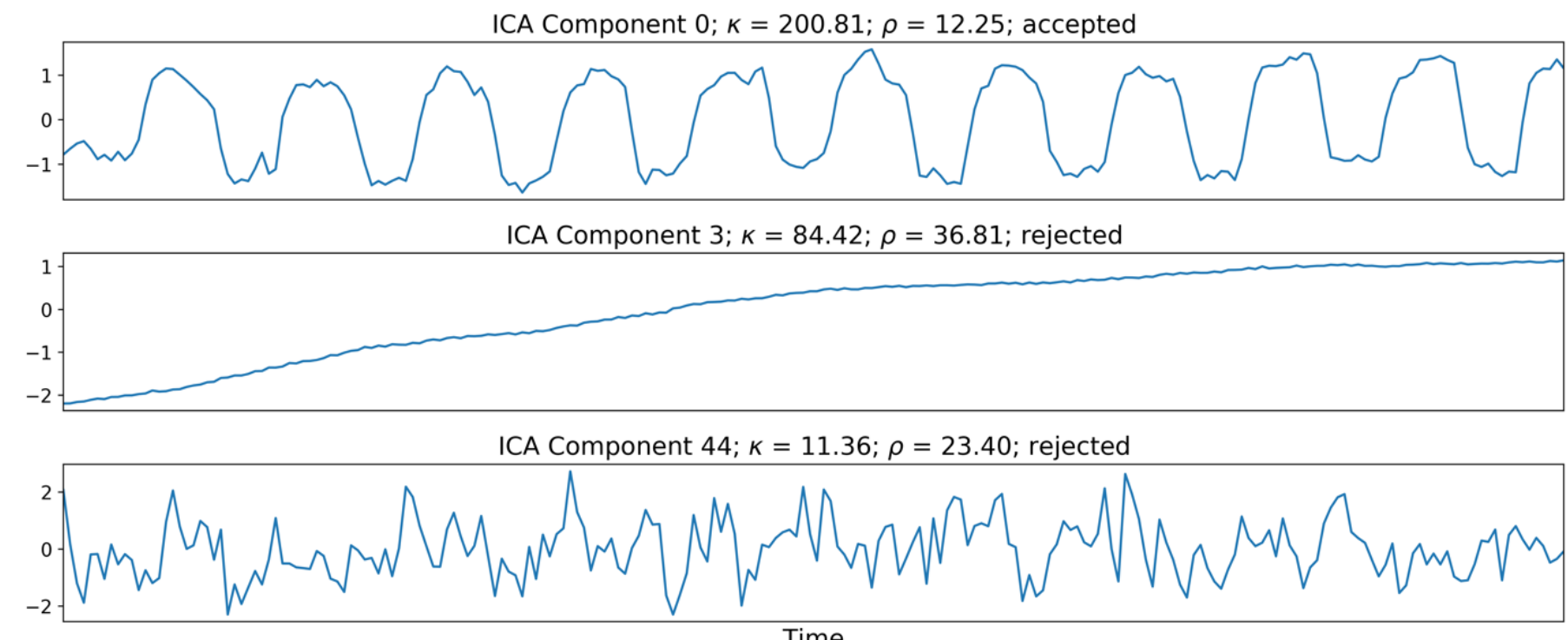
In order to check that our dataset was able to display robust contrasts in multiple subjects and preprocessing conditions, we processed our data in the following three ways:

- Single echo:** The second echo was used to look at brain activation in response to the task.
- Optimally combined:** A weighted average of the echoes was used to optimize T2\* weighting. This has been shown to give a reliable boost in data quality.



**Figure 1.** Generating an optimally combined image with a combination of echo images from a multi-echo experiment. (Figure from Kundu 2017)

- Denoised:** The data were denoised with the standard tedana.readthedocs.io pipeline. In this approach, ICA is performed to regress out components classified as noise.



**Figure 2.** Applying TE-dependent ICA to separate data into components and then remove non-BOLD noise components. (Figure from tedana.readthedocs.io)

Data were processed in AFNI before denoising. Finally, we compared brain activation and task response across these three preprocessing streams in multiple subjects using a **group analysis**. We looked at significant ROIs for the two contrasts present in the task.

### REFERENCES

- Kundu P, Voon V, Balchandani P, Lombardo MV, Poser BA, Bandettini PA. Multi-echo fMRI: A review of applications in fMRI denoising and analysis of BOLD signals. *Neuroimage*. 2017;154:59-80. doi:10.1016/j.neuroimage.2017.03.033
- "The Tedana Pipeline". *Tedana*, 2017, tedana.readthedocs.io/en/latest/approach.html#multi-echo-data.
- Malins JG, Gumkowski N, Buis B, et al. Dough, tough, cough, rough: A "fast" fMRI localizer of component processes in reading. *Neuropsychologia*. 2016;91:394-406. doi:10.1016/j.neuropsychologia.2016.08.027

### ACKNOWLEDGMENTS

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## TASK DESIGN

13 healthy volunteers (3 males and 10 females, aged 22-48) performed a language localizer task.<sup>3</sup> They viewed 4-second groupings of 4 audio or visual stimuli and pressed a button if the same word or non-word appeared twice in each grouping. The visual stimuli were words or false font words. The audio stimuli were words or vocoded sounds derived from words. This task allows for a word-nonword contrast as well as a visual-audio contrast. These two contrasts mean a spatially diffuse number of brain regions will show measurable activation, and evaluation and validation of multi-echo methods will include a range of response magnitudes and drop out regions.

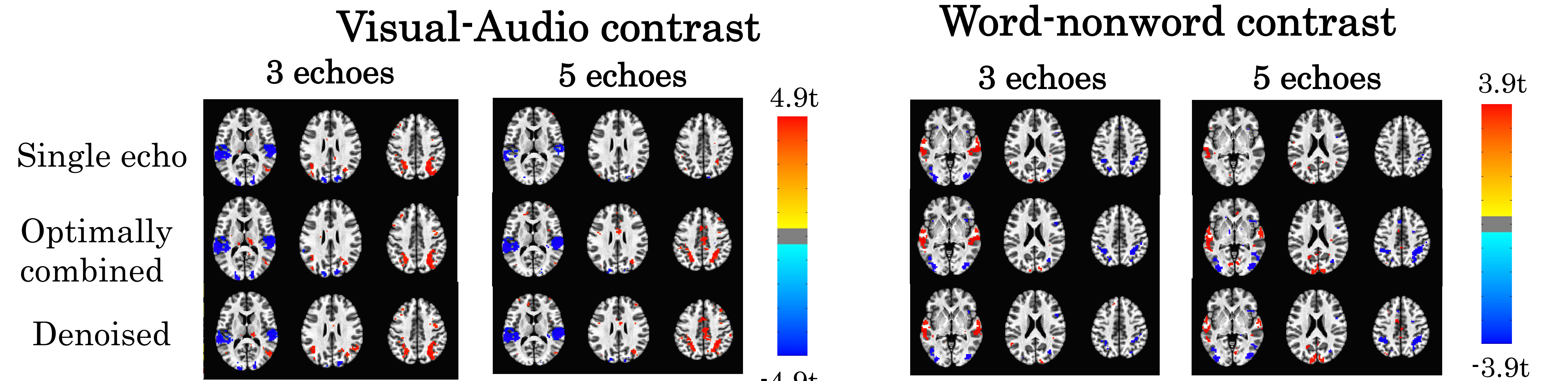


**Figure 3.** An example of the visual task conditions (*left*: visual-word and *right*: visual-nonword).

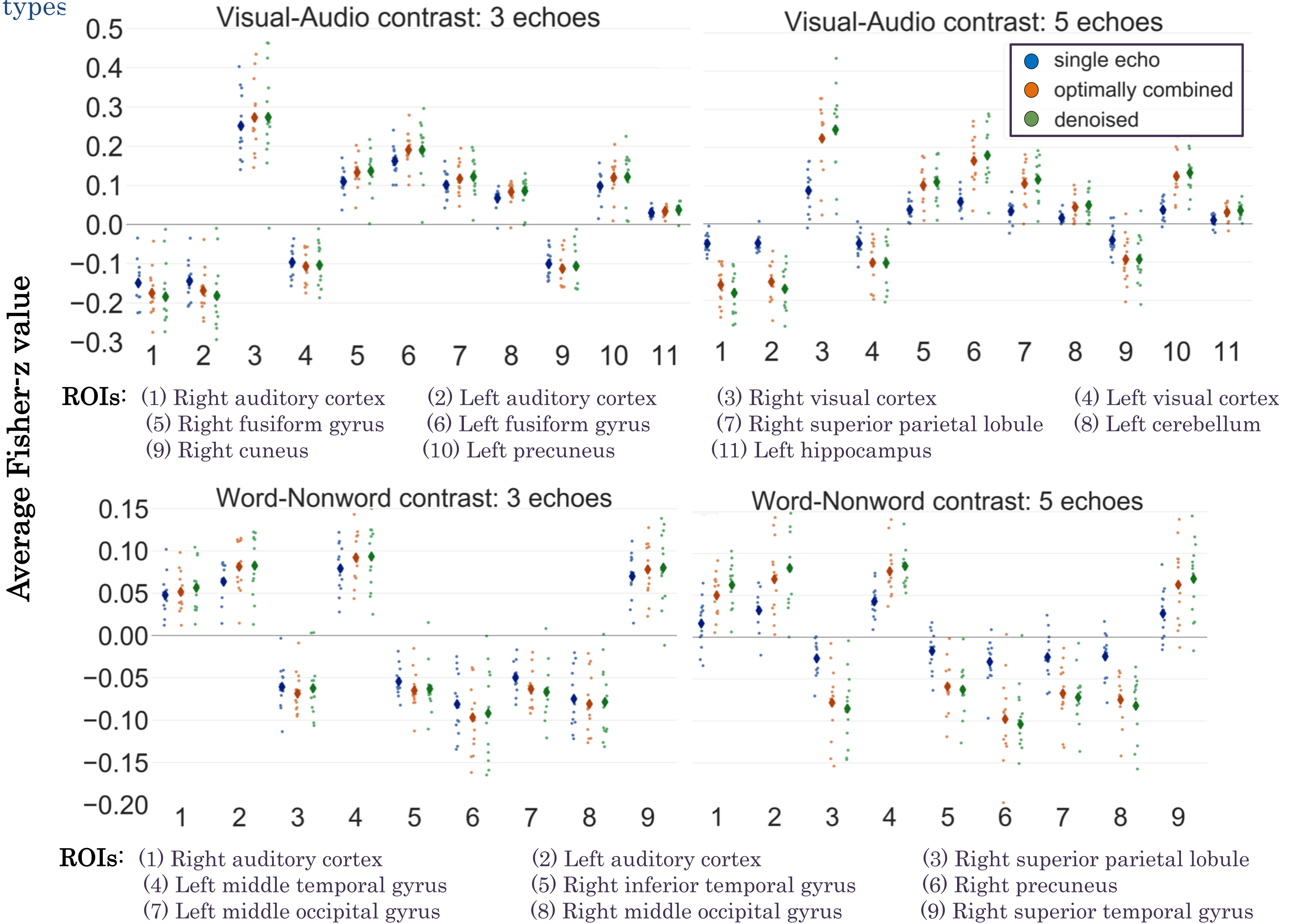
Each of the four conditions appeared 34 times over 22 minutes in a jittered event-related design. Data were acquired on a GE MR750 3T MRI with a 2 sec TR  
 Acquisitions include:

- 3 echoes, in-slice accel=2, 3.3 mm<sup>3</sup> voxels, TE=10, 25, 40 ms
- 5 echoes, in-slice accel=3, 3.3 mm<sup>3</sup> voxels, TE=7.8, 18.4, 29, 39.6, 50.2 ms
- 3 echoes, in-slice accel=3, 2.75 x 2.75 x 3 mm<sup>3</sup> voxels, TE=9, 22, 35 ms (fewer volunteers & not presented here)

## RESULTS



**Figure 4.** Group maps of visual-audio and word-nonword conditions visualized across preprocessing types



**Figure 5.** These dot plots show average Fisher-Z values in significant ROIs (>10 voxels) in the group maps. Effect magnitudes vary across ROIs, and consistent effect directions with all preprocessing choices. As expected, values tend to be slightly larger after optimal combination and denoising.

## CONCLUSIONS

- The task generates robust and consistent contrasts with distinct acquisition parameters
- These can be used to validate and compare different preprocessing streams, like those being developed in tedana

## FUTURE STEPS

- Continue evaluations of current analysis pipelines using these data
- Fully document and share these data
- Make these an option in the tedana integration and testing process