

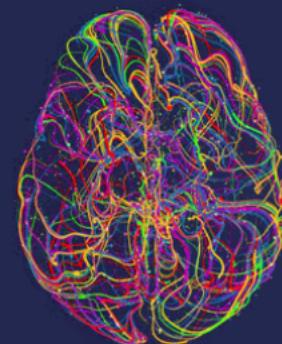
**Two is better than one (and many are better): Multi-echo fMRI methods and applications**

# How to decide if multi-echo fMRI can improve your study?

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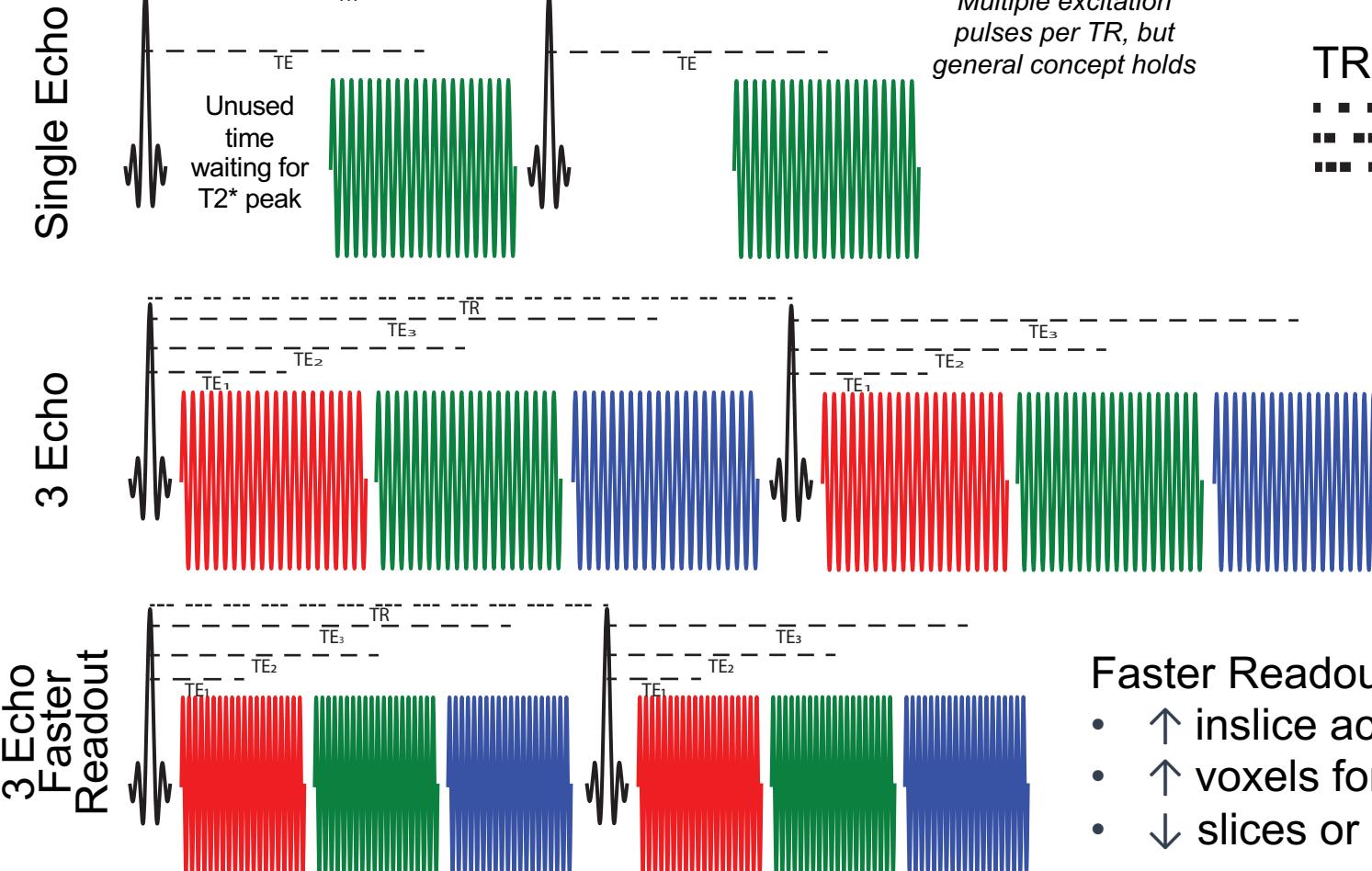


**OHBM 2020**  
A Virtual Experience for  
Engaging Minds & Empowering Brain Science

- ⌚ Multi-echo fMRI: One of many acquisition choices
- ⌚ On example of how multi-echo fMRI can help
- ⌚ Evaluating multi-echo fMRI with open eyes

# Multi-echo fMRI: One of many acquisition choices

- ⌚ Common question: Multi-echo fMRI or a short TR?
- ⌚ Better question: How many echoes for how short a TR?



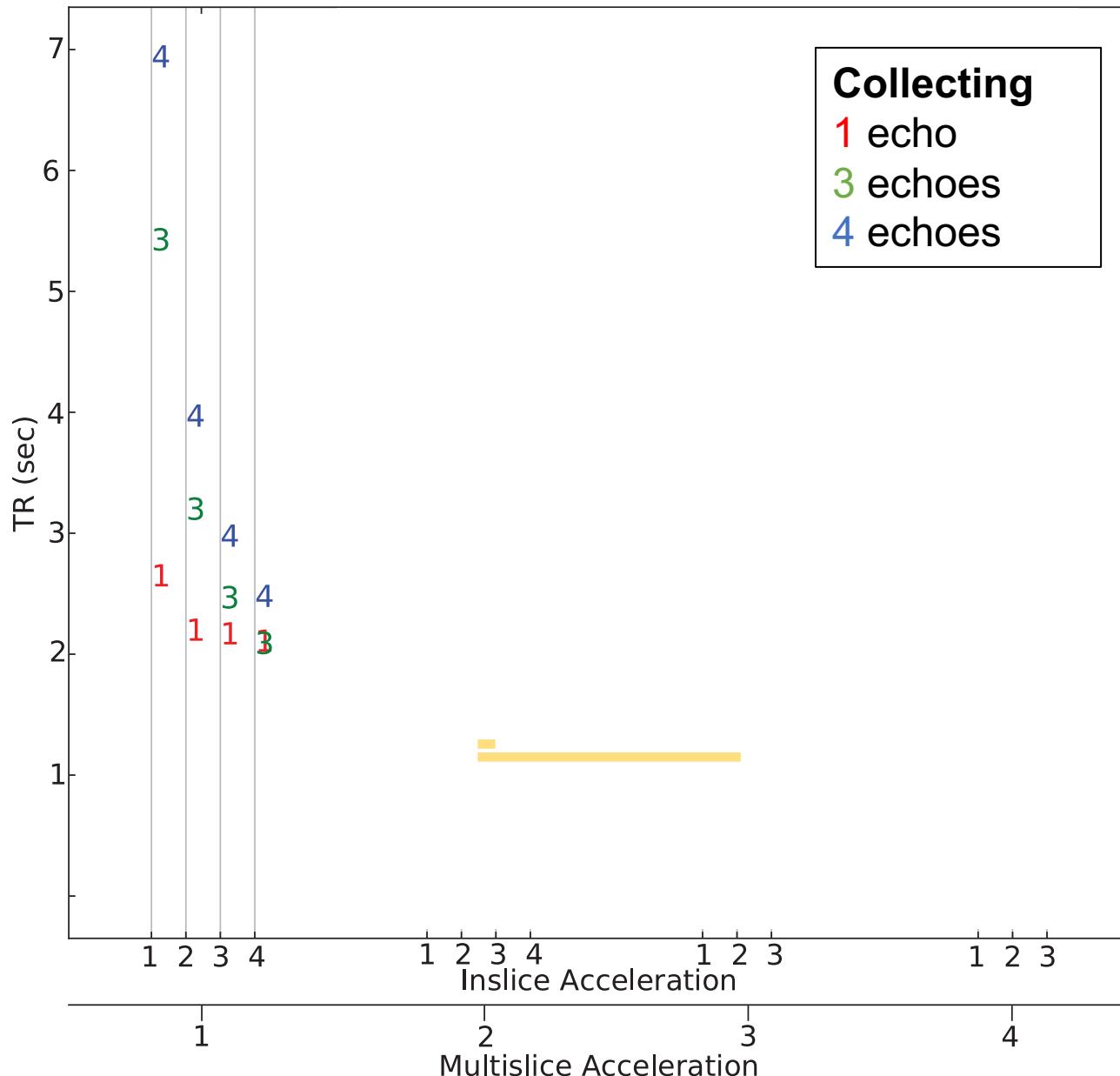
**Dual-echo methods**  
Helpful for motion correction & physiological artifact removal  
*Bright NeuroImage 2013; 62:526*  
*Buur NMR Biomed 2009; 22:551*

**Optimize many acquisition parameters based on study goals, just like any other fMRI study**

## Faster Readout

- ↑ inslice acceleration (GRAPPA, ASSET, etc)
- ↑ voxels for same field-of-view
- ↓ slices or ↑ multislice or multiband acceleration

# TR variation for multi-echo vs single-echo



Collected on a Siemens Prisma 3T MRI  
Siemens OS VD11

CMRR Multiband pulse sequence  
64 channel head coil

3mm<sup>3</sup> voxels 42-44 slices  
cortex & cerebellum coverage

3 vs 1 echo for:

Inslice Acceleration = 2: 40-50% TR cost

Inslice Acceleration = 3: ~14% TR cost

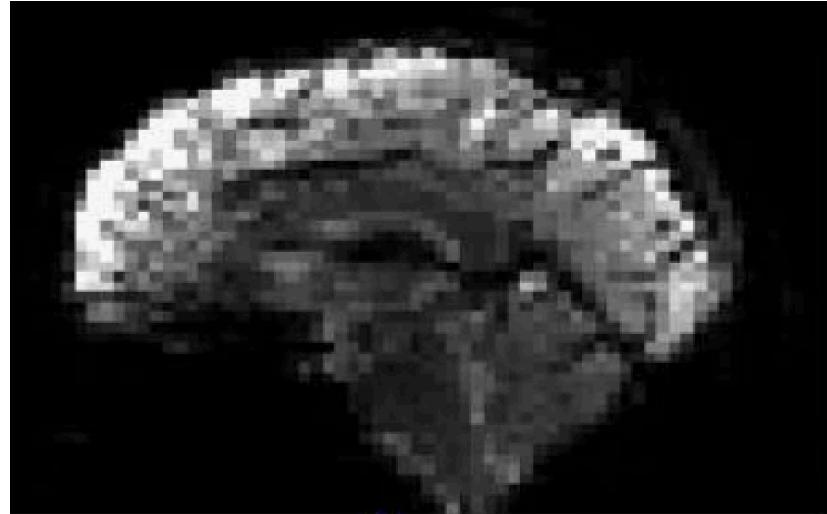
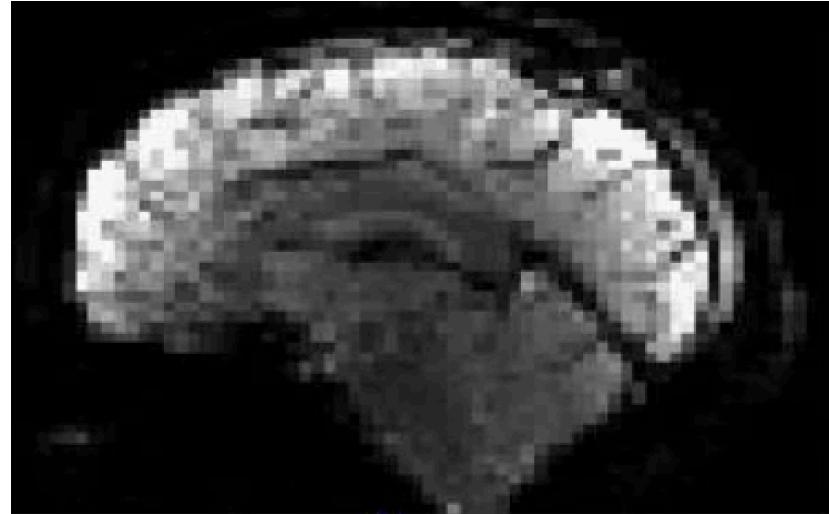
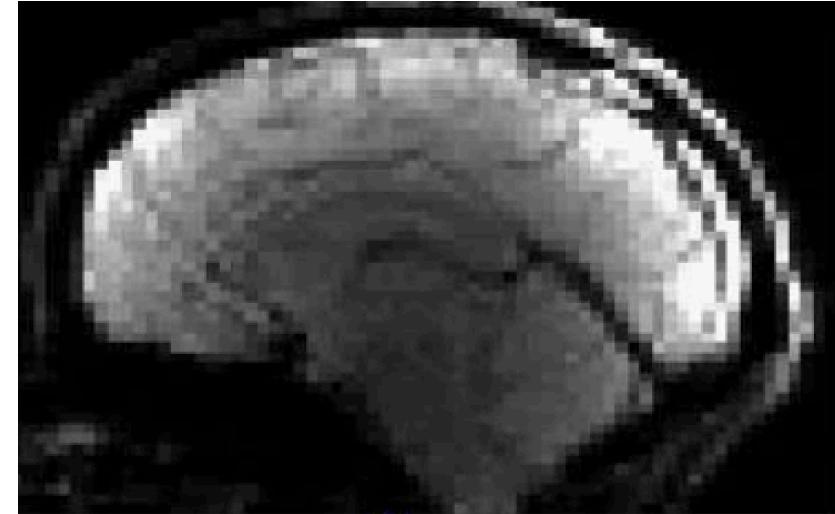
Inslice Acceleration = 4: ~-1% TR cost

# Multi-echo fMRI data with a 1 sec TR

Echo 1

Echo 2

Echo 3



- ❖ Collected on a Siemens Prisma 3T MRI, 64 channel head coil, Siemens OS VD11
- ❖ CMRR Multiband pulse sequence, Multi-slice acceleration=4, in-slice acceleration=2, 3mm<sup>3</sup> voxels
- ❖ TEs=13.6, 31.86, 50.12ms
- ❖ Data from an ongoing study led by Emily Finn

# How multi-echo fMRI can help

- ⌚ "Optimal Combination" of echoes: Weighted average (Posse 1999, Poser 2006)
  - ⌚ Weighted average of echoes based on voxel-wise T2\* estimates
  - ⌚ Straightforward math
  - ⌚ Automatically calculated in AFNI, fMRIprep, and tedana
- ⌚ Denoising
  - ⌚ Removal of non T2\* weighted signal based on models of signal change across echoes (Kundu 2012 & 2013)
  - ⌚ Uses empirical models of noise to get away from assumptions of what "looks like noise"
  - ⌚ Potential for bigger gains, such as separating slow drift of fMRI signal from slow neural changes (Evans 2015)

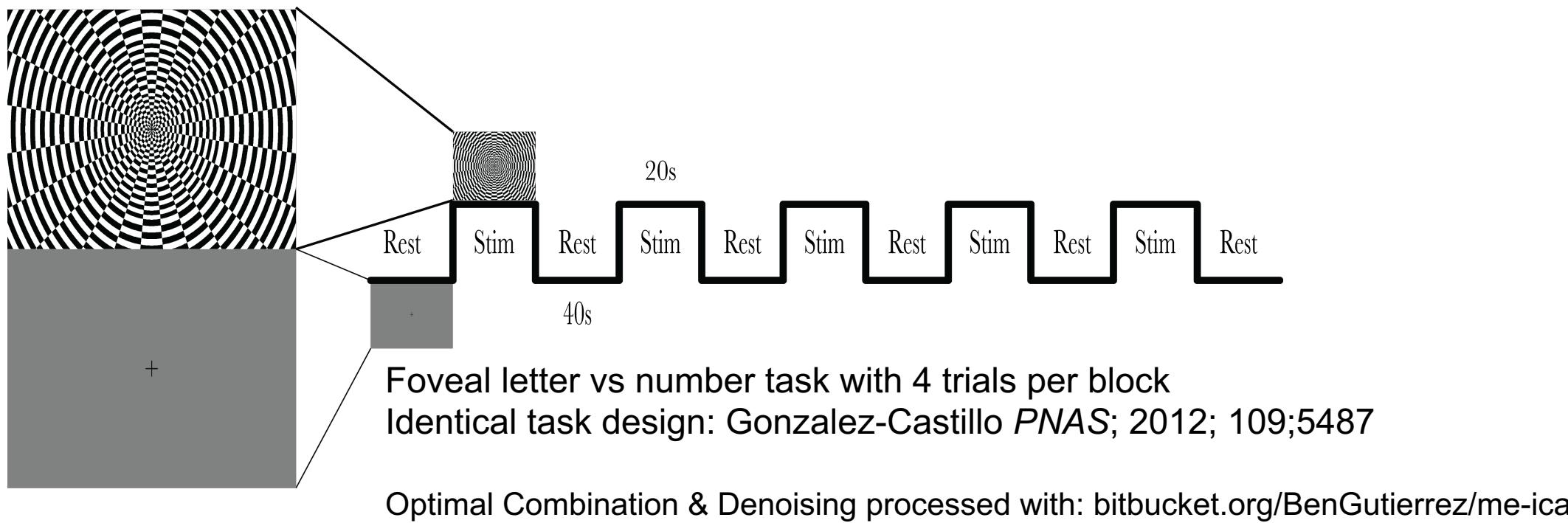
**More details at**

<https://tedana.readthedocs.io/en/latest/resources.html>  
<https://tedana.readthedocs.io/en/latest/approach.html>

Evans *NeuroImage* 2015; 105:189; Kundu *NeuroImage* 2012; 60:1759; Kundu *PNAS* 2013; 110:16187  
Poser, *MRM* 2006; 55:1227; Posse *MRM* 1999; 42:87

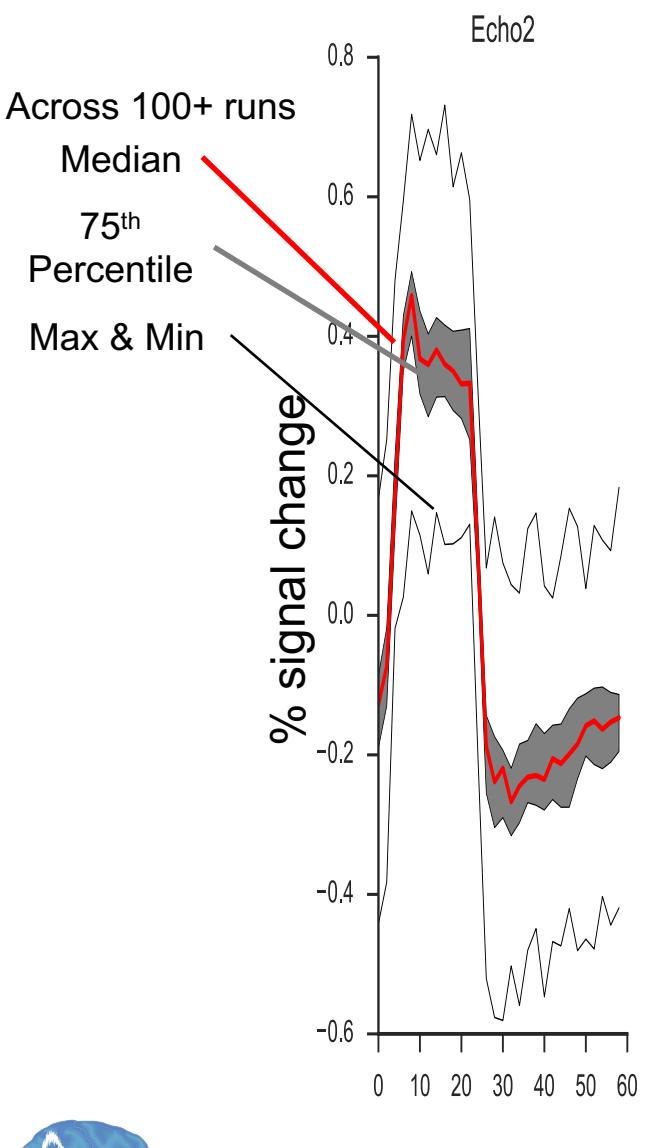
## Experimental Design

- 2 Volunteers, 9 sessions, 103 runs each, 9 hours of data per subject
- GE MR750, 3T, 32 channel coil
- EPI:  $3.5\text{mm}^3$ , **3 echoes, TE=15.4, 29.7, & 44.0ms** FA=75°, TR = 2s, 33 slices
- 5.5 minutes, 161 volumes (150 volumes used in each run)

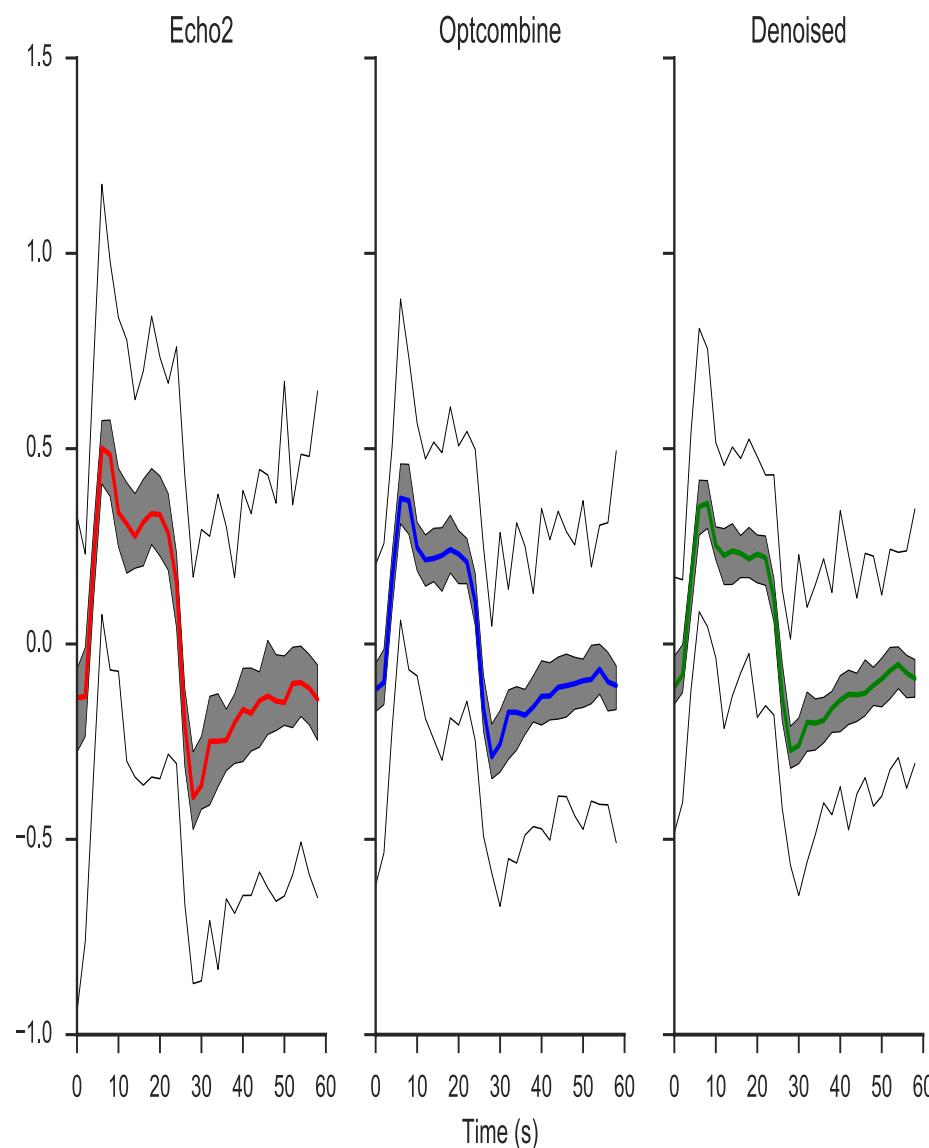


# Lateral Geniculate Nucleus (LGN) Responses

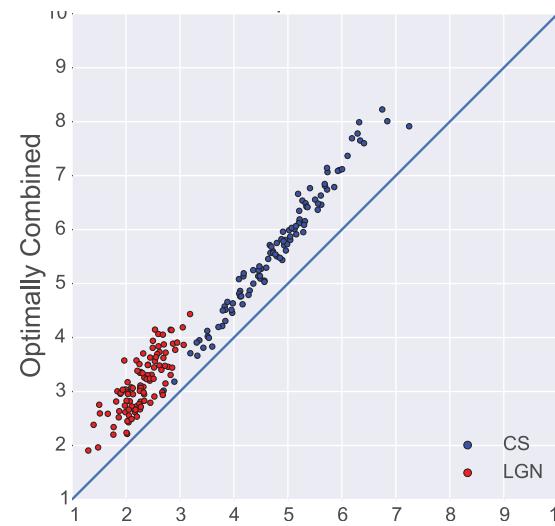
Volunteer 1



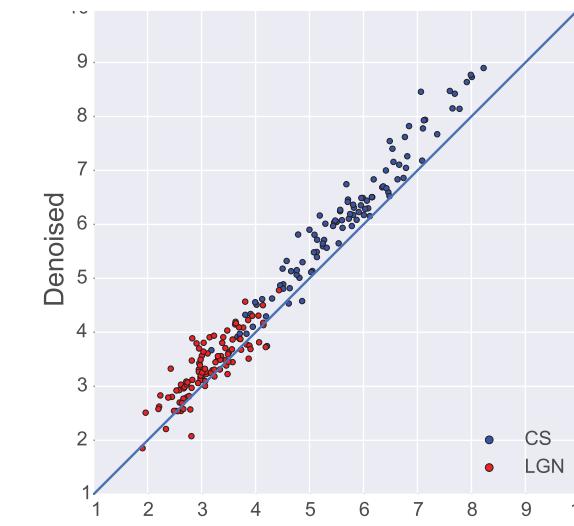
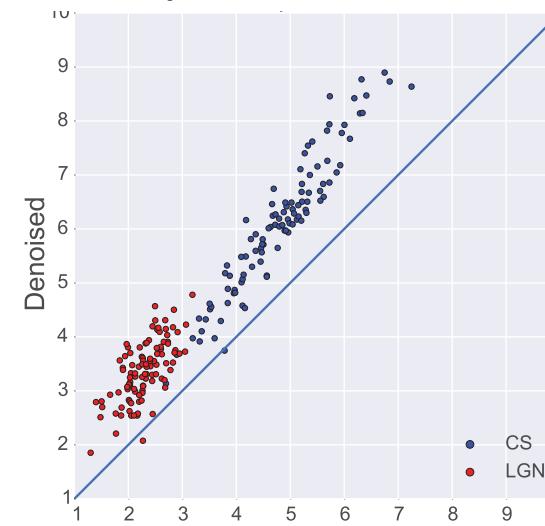
Volunteer 2



# Contrast-to-Noise By Run



CNR Comparison for Volunteer 1



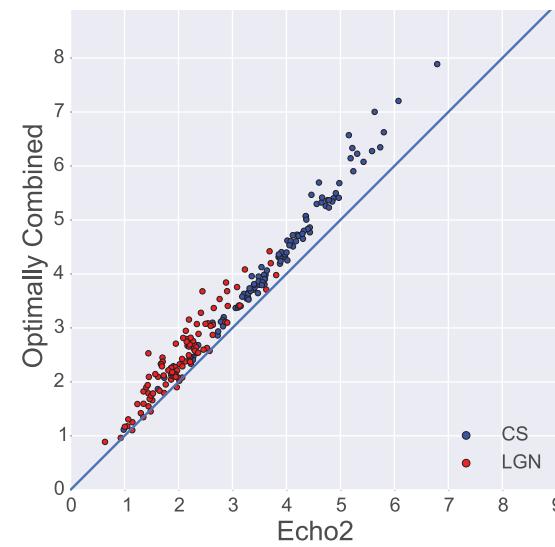
Regions of Interest

**CS**

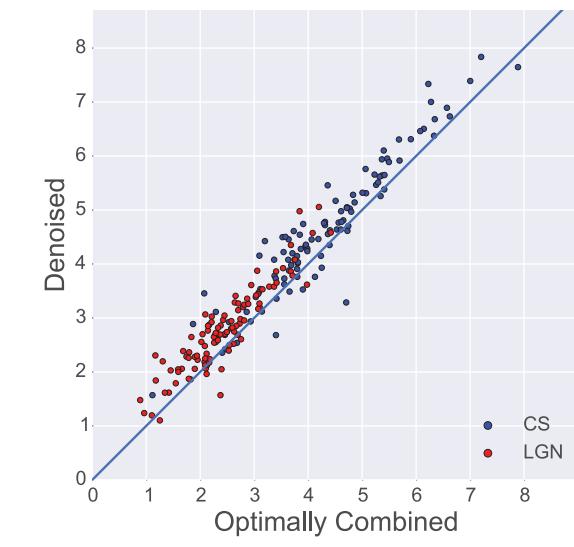
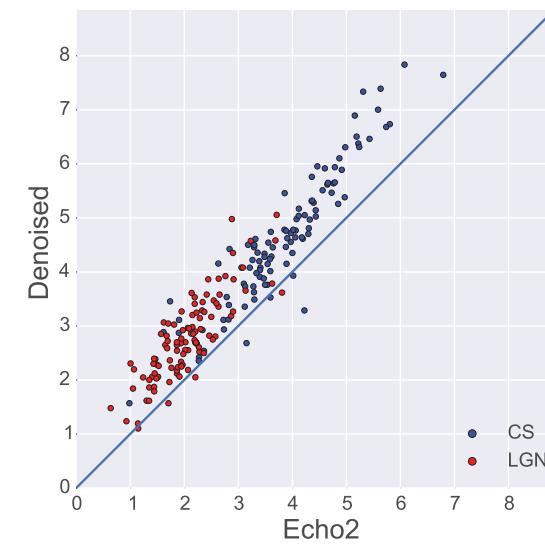
Calcarine  
Sulcus

**LGN**

Lateral  
Geniculate  
Nucleus

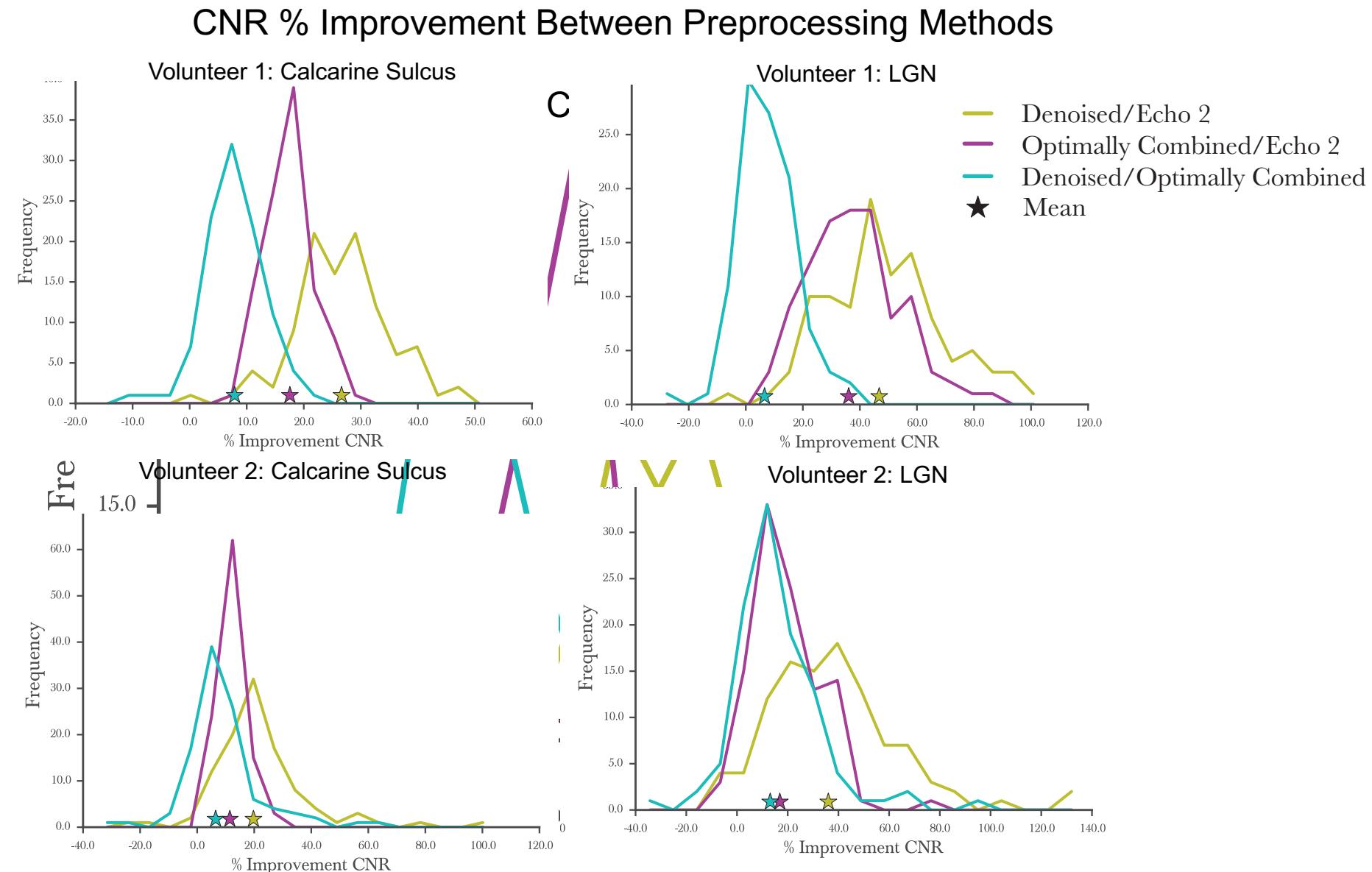
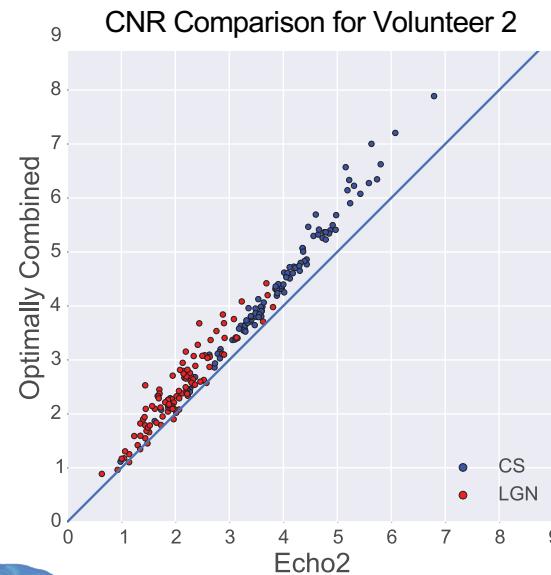
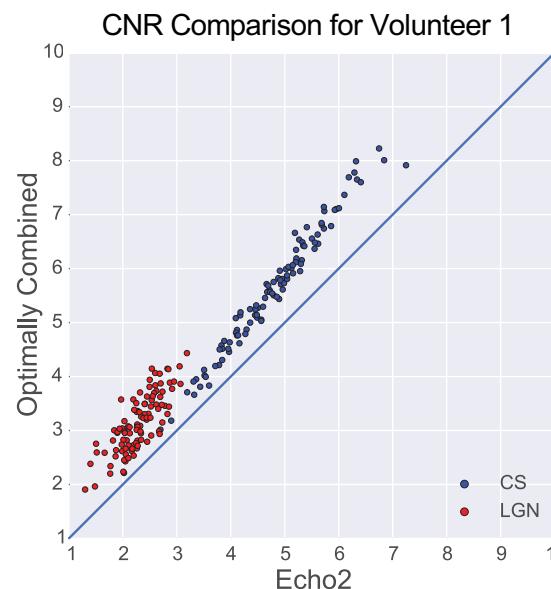


CNR Comparison for Volunteer 2



Each dot is  
the CNR  
values from  
1 run

# Contrast-to-Noise By Run



- ⌚ Optimal Combination reliably improves CNR over single echo
- ⌚ Denoising can be similar, much better, or worse than the optimal combination
  - ⌚ More to understand & improve on methods of defining noise to remove
  - ⌚ Use denoising, but don't assume everything worked perfectly
- ⌚ Limits of presented data
  - ⌚ Awesome volunteers: <1.5mm max head motion in all but 2/206 runs
  - ⌚ Single, stable scanner with a regular Quality Assurance testing
  - ⌚ Benefits of denoising may be greater with more noise to potentially remove

# Summary and Recommendations

- ⌚ Acquiring multi-echo can be balanced with a combination of acceleration↑, slices↓, & TR ↑
- ⌚ More CNR from “optimal combination” of echoes **should** balance lower SNR from acceleration↑
- ⌚ With multi-echo data, you can empirically identify and remove non-T2\* weighted noise
  - ⌚ Algorithms still under development & should not be blindly used
  - ⌚ These methods will get better: [tedana.readthedocs.io](https://tedana.readthedocs.io) & Elizabeth DuPre's talk
- ⌚ **Recommendations**
  - ⌚ Planning to go from acquisition to publication in a couple of years: Consider multi-echo
    - ⌚ You may see modest benefits with optimal combination
  - ⌚ Planning a longer-term project with goals of data re-use: Strongly consider multi-echo
    - ⌚ Immediate, modest benefits, and larger longer-term benefits are likely
    - ⌚ Development of additional ways to use multi-echo fMRI is likely (see César Caballero-Gaudes' talk)

## Introduction to Multi-Echo



**Stefano Moia**

Multi-echo is a simple concept that opens up many possibilities

## *tedana* software and community



**Elizabeth Dupre**

A community based, open source software development to improve method and application together

## Multi-Echo beyond preprocessing



**César Caballero-Gaudes**

With Multi-Echo we can estimate the activity-inducing neural signal better

## Multi-Echo fMRI in practice



**Angela Laird**

Setting up a Multi-echo study for the first time will be challenging, but worth the effort

# Acknowledgements

- ⌚ 100-runs multi-echo study:
  - ⌚ The volunteers!
  - ⌚ Peter Bandettini
  - ⌚ Javier Gonzalez-Castillo
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  - ⌚ Laura Buchanan
  - ⌚ Colin Hoy
- ⌚ NIH Biowulf computing cluster: hpc.nih.gov

- ⌚  developers including:
  - ⌚ Elizabeth DuPre
  - ⌚ Logan Dowdle
  - ⌚ César Caballero-Gaudes
  - ⌚ Javier Gonzalez-Castillo
  - ⌚ Ross Markello
  - ⌚ Stefano Moia
  - ⌚ Taylor Salo
  - ⌚ Joshua Teves
  - ⌚ Eneko Uruñuela
  - ⌚ Kirstie Whitaker
  - ⌚ You?