

# Noise Models and Correction for fMRI

*- an Introduction to the PhysIO Toolbox*

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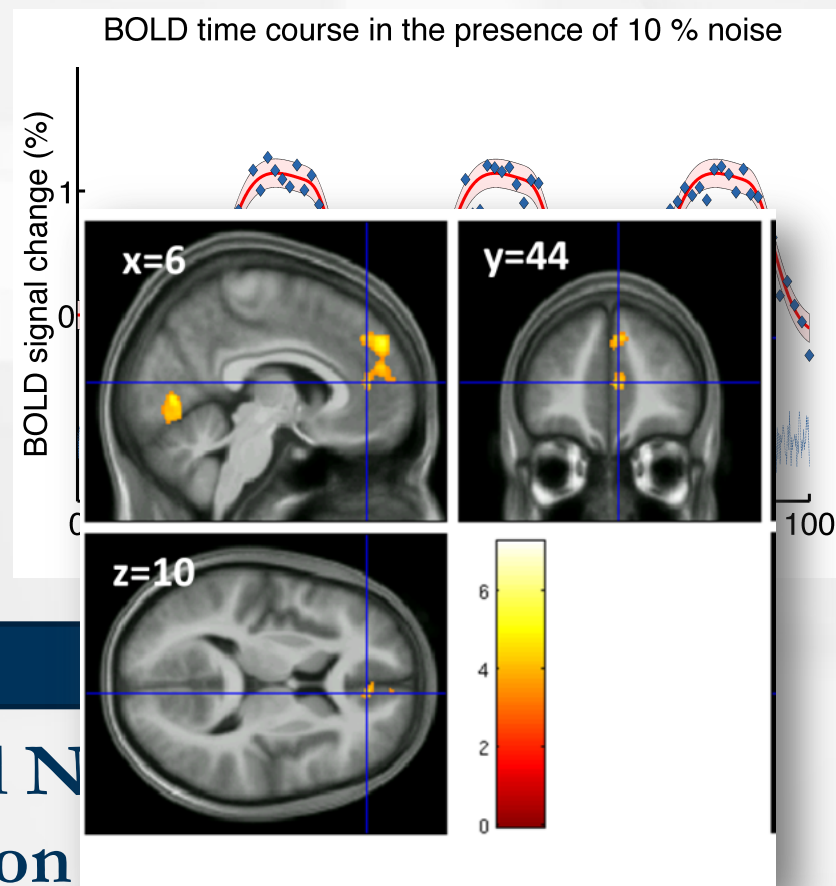
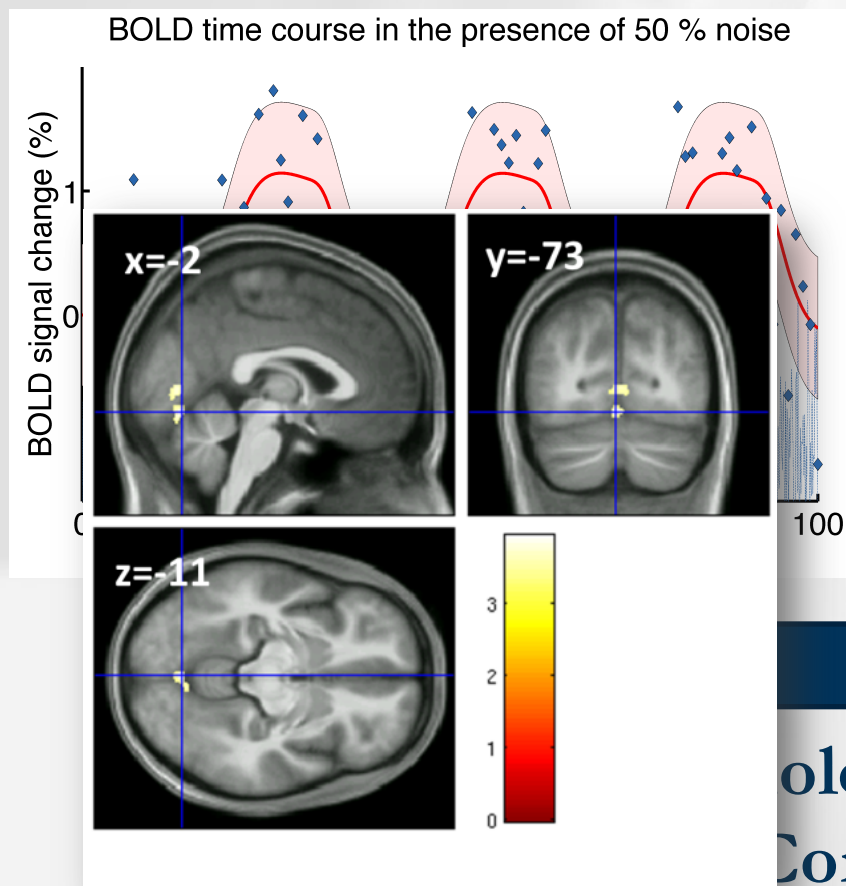


# The Goal of Noise Correction



Before

After



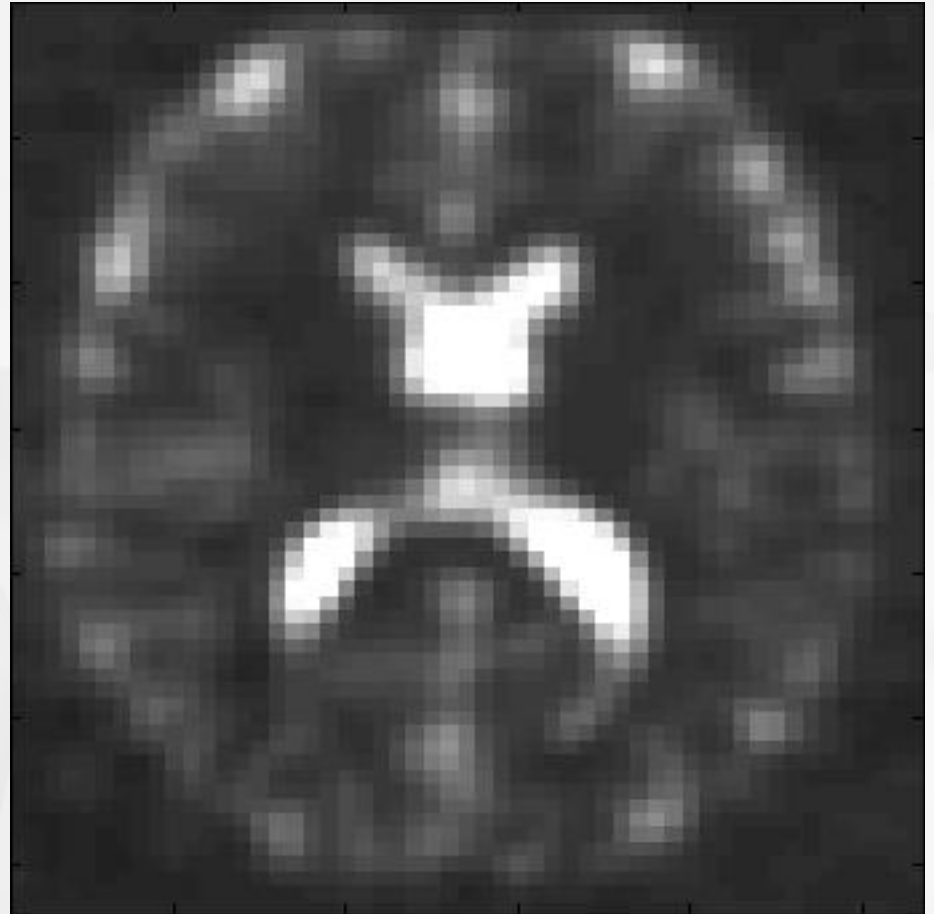
ological N  
Correction



# Previously...



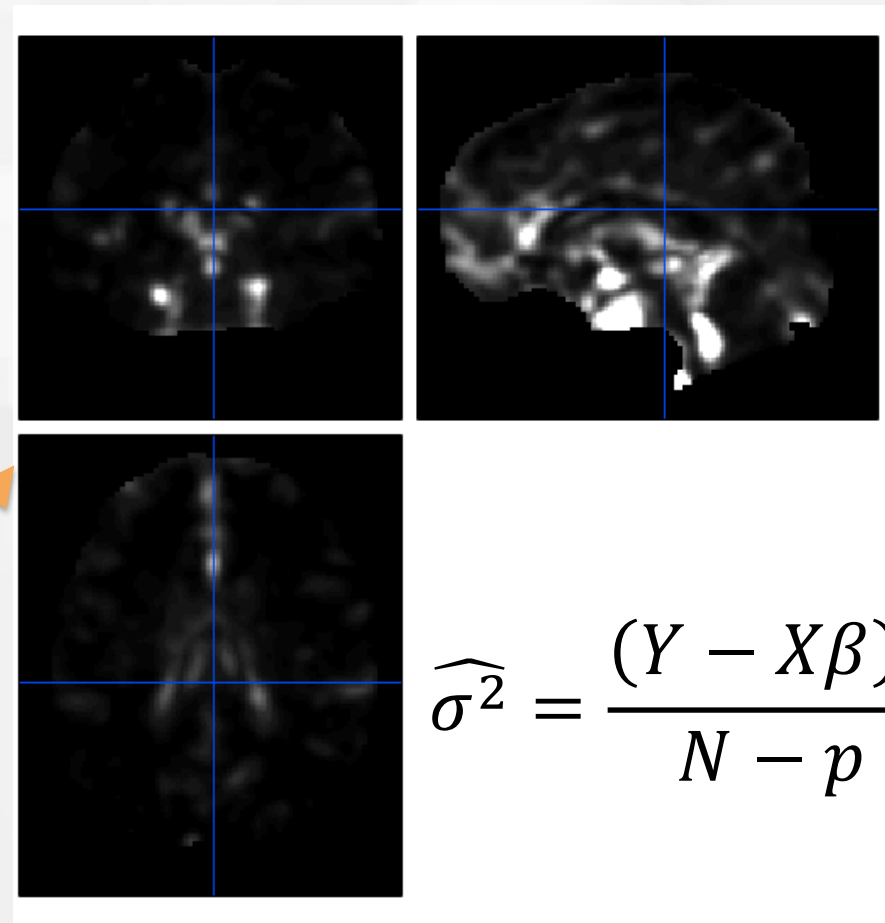
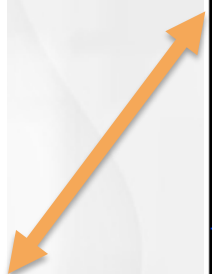
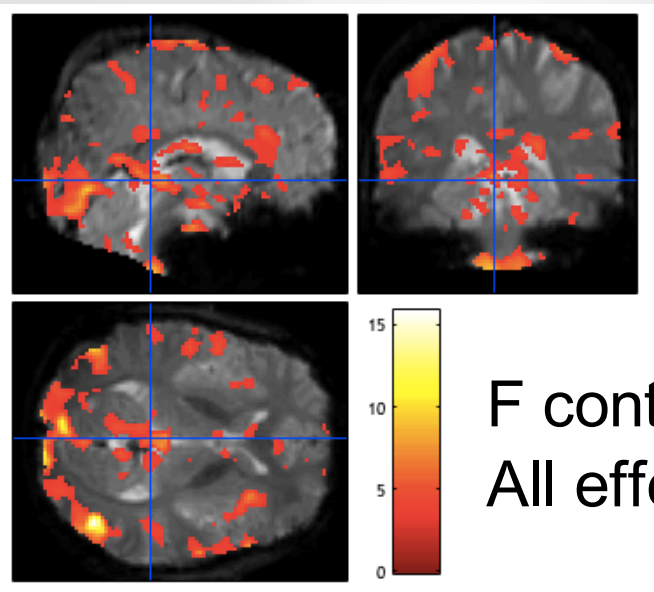
- How we ended 3 weeks ago
- After smoothing... still some fluctuation



# Previously....(continued)



- Last week: ResMS image
- Indicates where model incomplete...
- limits sensitivity...





# Sources of Noise in fMRI



- Acquisition Timing

Temporal Preproc

- Slice-Timing

- Subject Motion

Spatial Preproc

- Realignment

- Anatomical Identity

Spatial Preproc

- Co-registration

- Inter-subject variability

Spatial Preproc

- Segmentation

- Thermal Noise

Spatial Preproc

- Smoothing

- Physiological Noise

Noise Modelling

- PhysIO Toolbox



- MRI Time Series Recap and Noise Sources
  - Why de-noising? Noise pathways: Scanner, Cardiac/Respiratory/Motion
- Noise Correction Approaches
  - Correction Target: Drift, Motion, Cardiac/Breathing Cycle
  - Data Correction Point: Modelling VS Preprocessing
  - Noise Model Input: fMRI Data-driven VS Peripheral Measures
- Noise Correction Prospects
  - Effects of Physiological Noise on Group Statistics
- Noise Correction Limitations
  - Degrees of Freedom; Task-related “noise”; Interoception





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# fMRI = Acquiring Movies

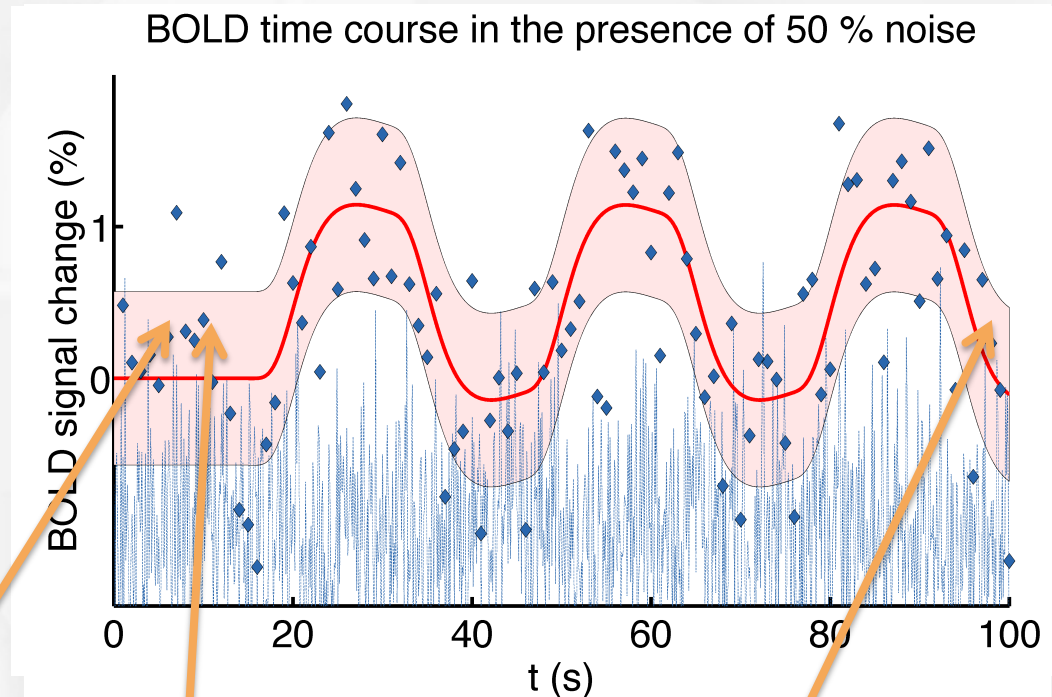


- The Localized Time-series is the Fundamental Information Unit of fMRI

**Signal:** Fluctuation through Blood oxygen level dependent (BOLD) contrast

**Noise:** All other fluctuations

- Run/Session: Time Series of Images



scan 1



...



scan N

time



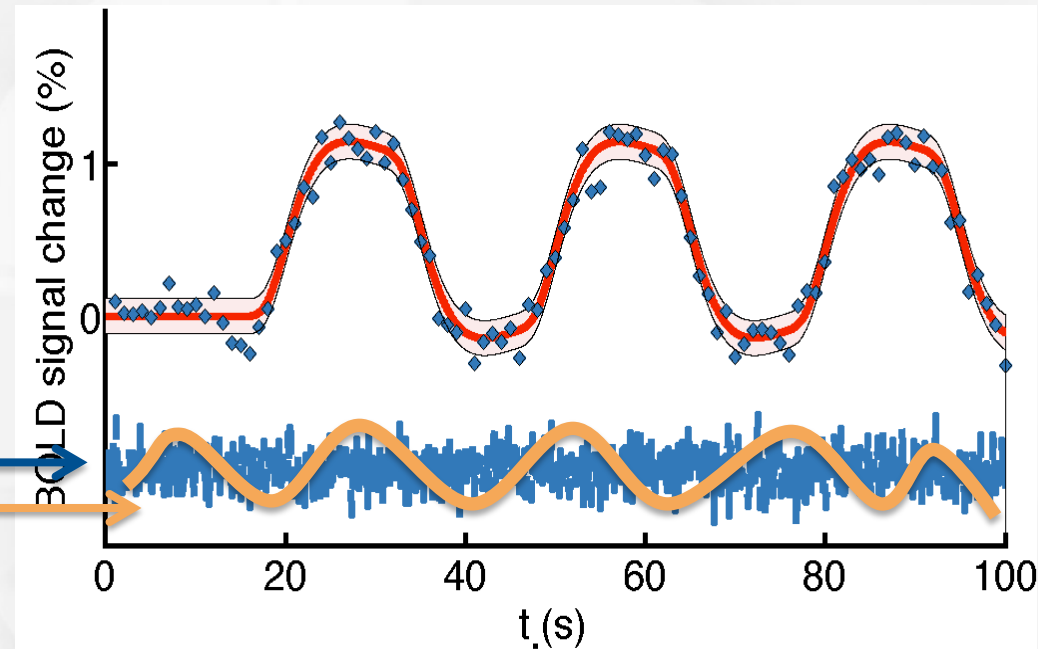
# Noise Categories & Reduction



- Thermal Noise
  - temporally uncorrelated
  - reduced SNR → risk of false negatives
  - Remedy: Spatial Smoothing

Noise: All other fluctuations

- “Structured” Noise
  - temporally correlated
  - reduced SNR → risk of false negatives
  - correlated with task → risk of false positives
  - Remedy: Noise Modelling (e.g. GLM)



Inference = Signal-To-Noise

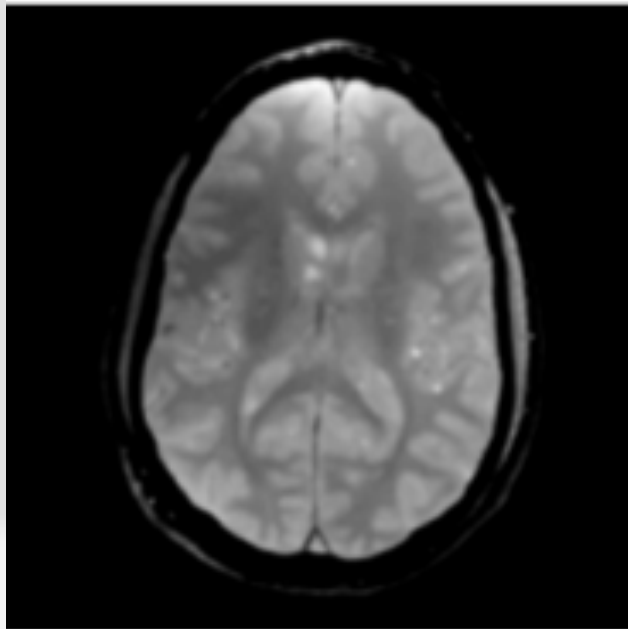
$$t = \frac{\beta}{\sqrt{\hat{\sigma}_\varepsilon^2 (X^T X)^{-1}}} = \frac{\beta \|\mathbf{x}\|}{\hat{\sigma}_\varepsilon}$$

$$F = \frac{N - M}{M_1} \cdot \frac{(\sigma_S^2 + \sigma_N^2) - \sigma_N^2}{\sigma_N^2}$$

# Recap: MR Image Encoding



Image Space ( $m$ )



$y$  ↑  
→  $x$

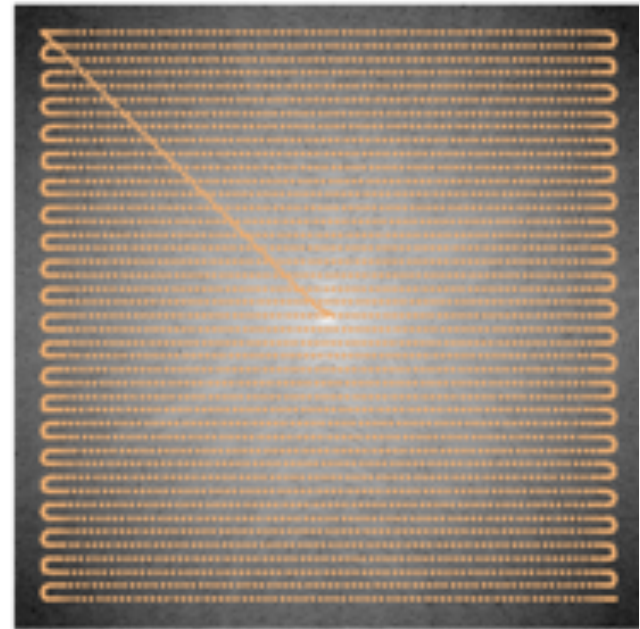
Fourier  
Transform  
(FT)



In general:  
Image  
Encoding

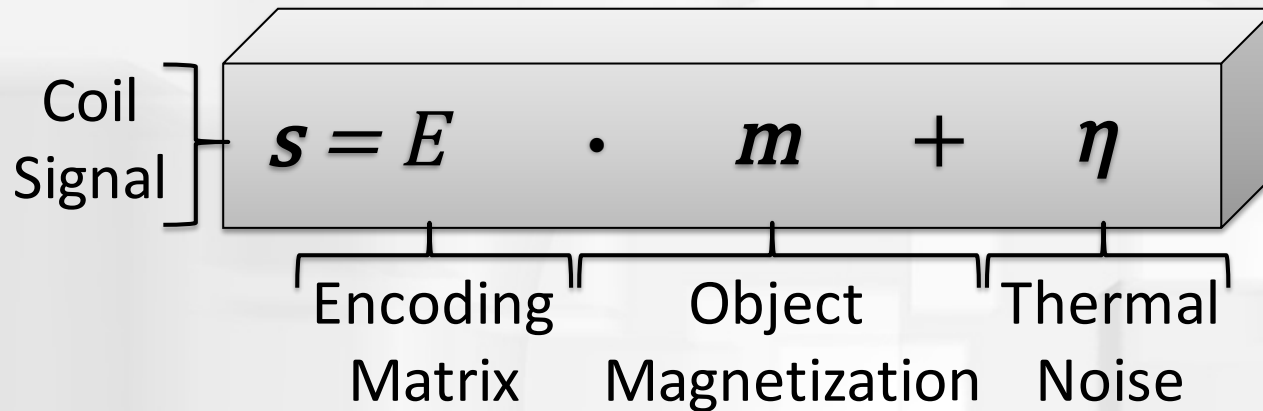
$E$

k-Space ( $s$ )



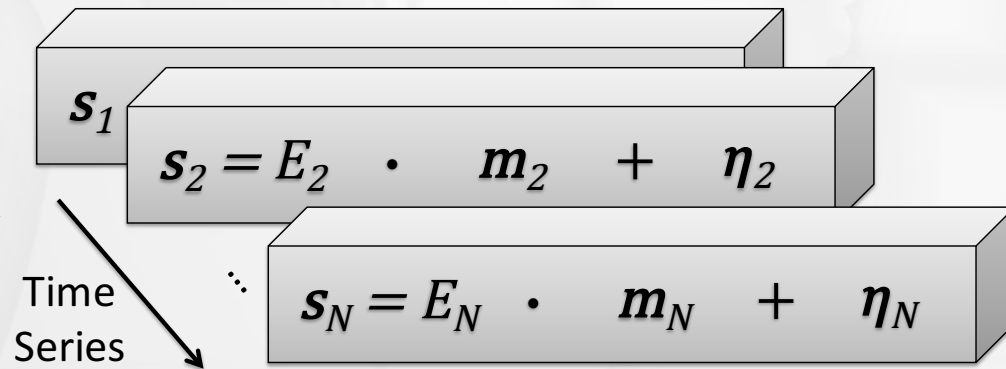
$k_y$  ↑  
→  $k_x$



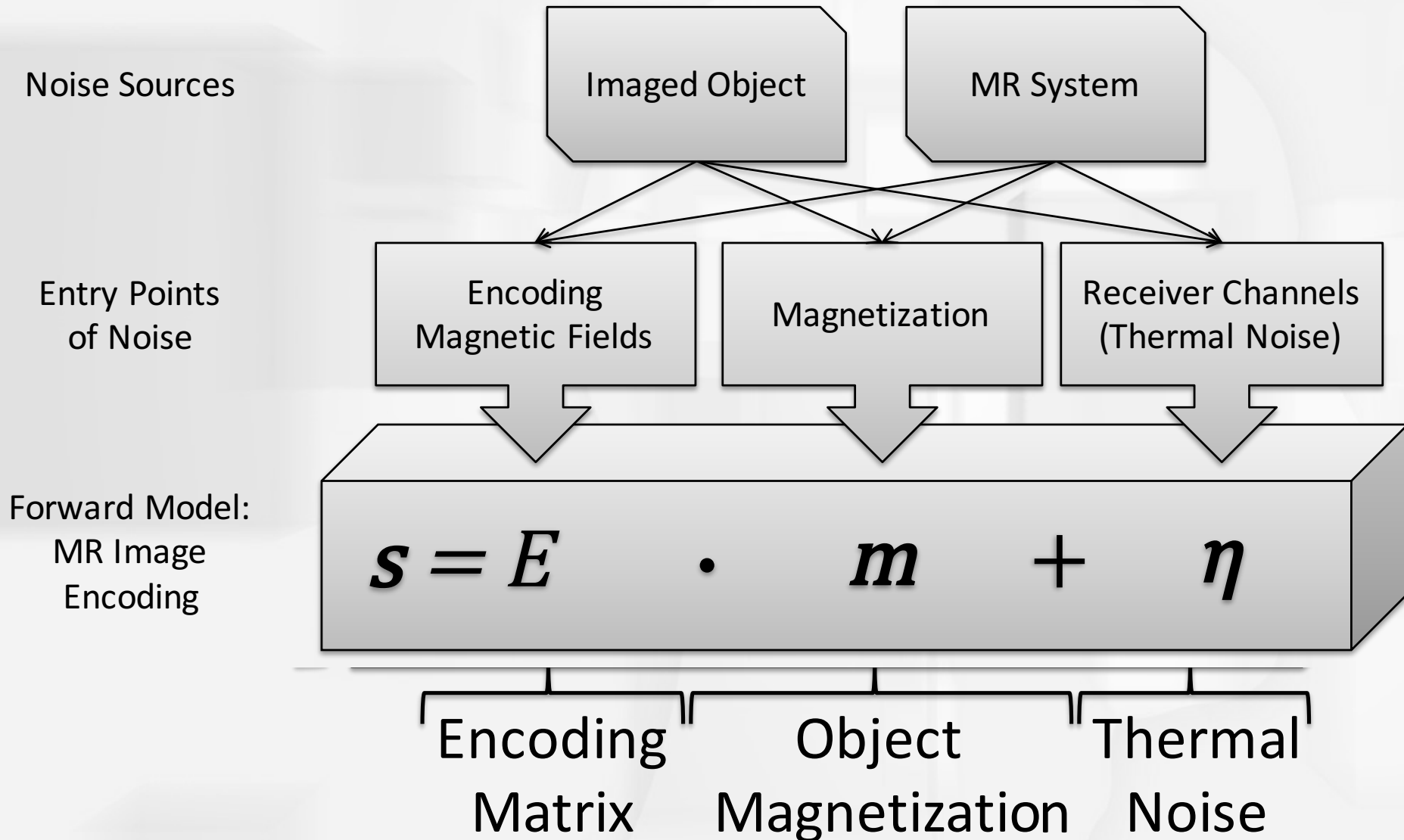


- Image reconstruction is also (often) a GLM, though a huge one,  $\sim 100,000$  rows
- Any changes between encodings (images) in encoding matrix (field), object magnetization and thermally induces image noise

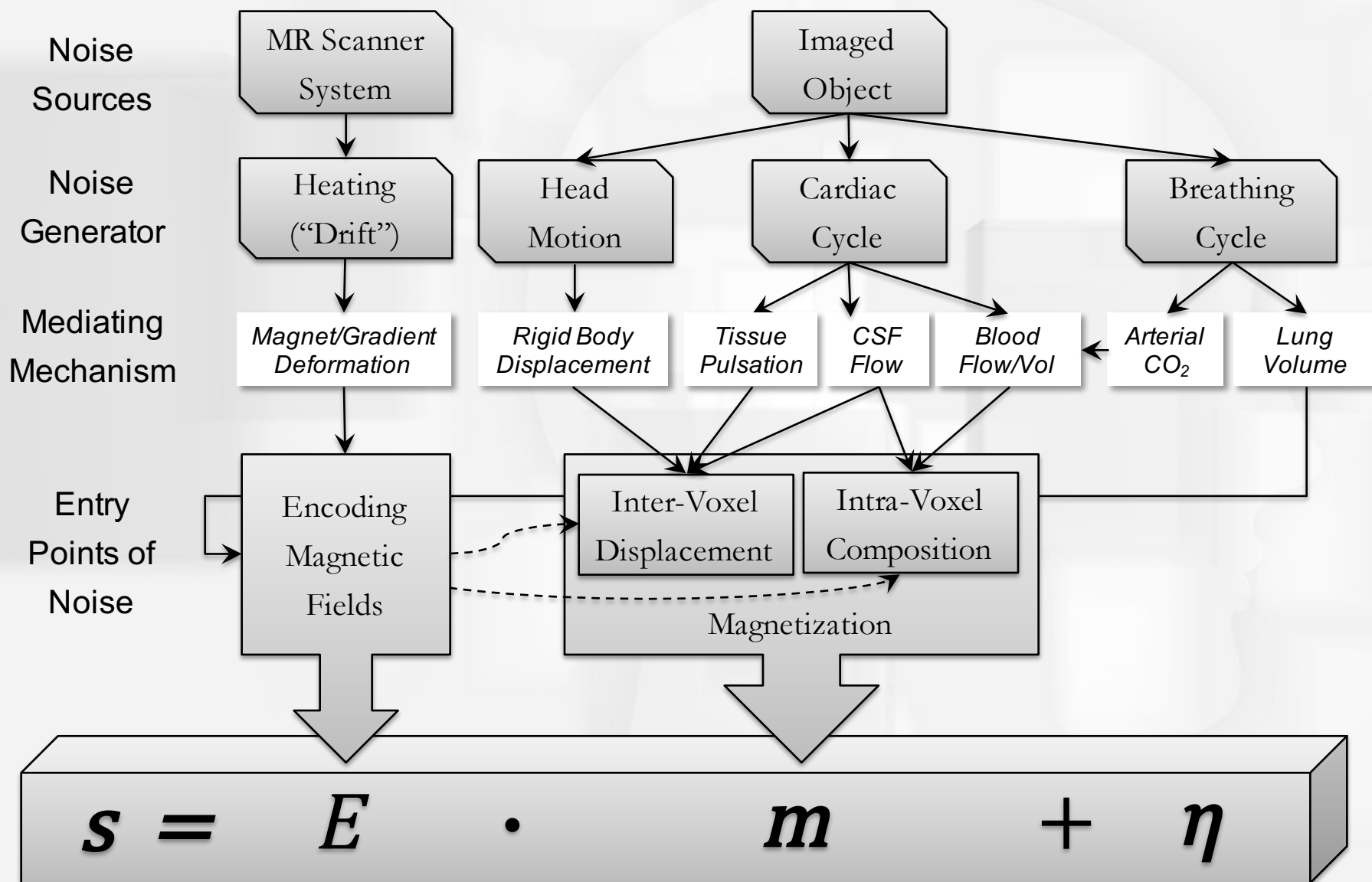
$$\hat{m} = (E^H E)^{-1} E^H s$$



# What fluctuates?



# Magnetization and Field Noise



# The Problem: Physiological Noise



- Cardiac effects

- Respiratory effects



# The Problem: Physiological Noise



## ■ Cardiac effects

- Systole:
  - Blood pumped into brain, vessel volume increases: pulsatile vessels
  - CSF pushed down: pulsatile CSF
- Diastole:
  - Vessel volume decreases
  - CSF flows back into “void” brain volume

## A Cardiac Cycle in the Brain

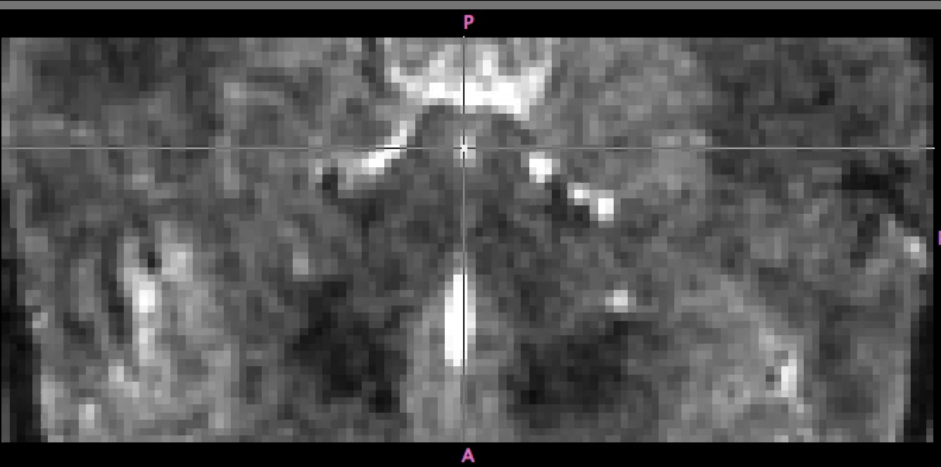
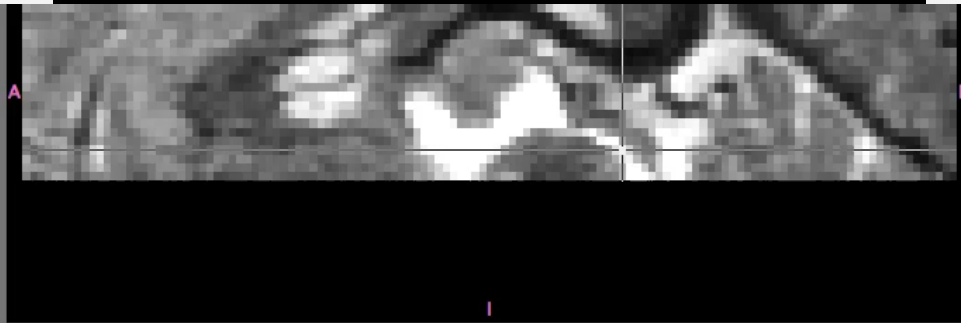
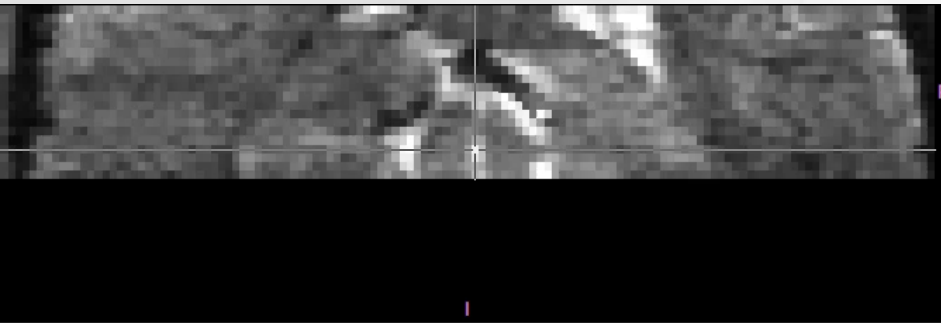


# The Problem: Physiological Noise



Triggered High-Resolution fMRI

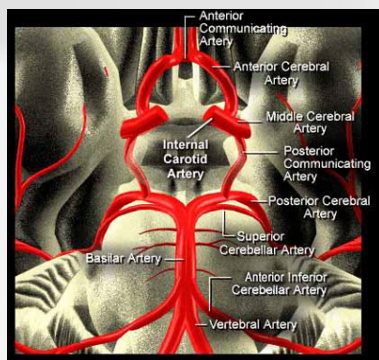
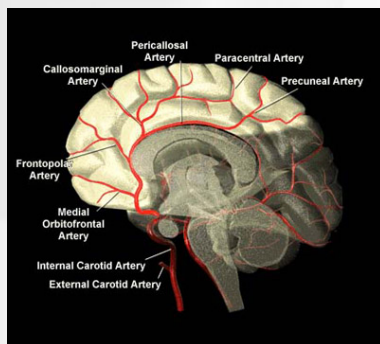
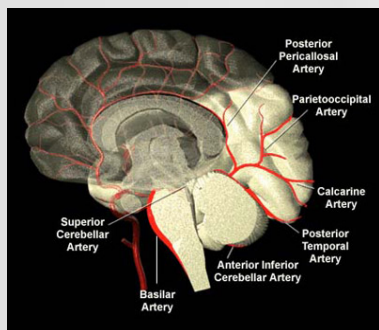
A Cardiac Cycle in the Brain



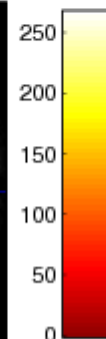
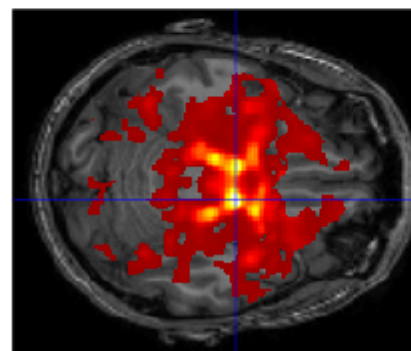
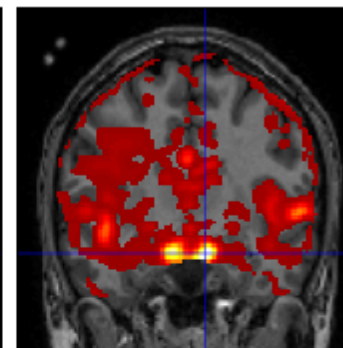
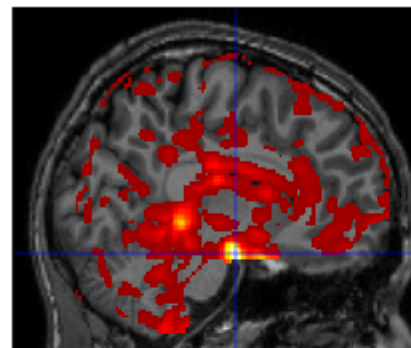
# The Problem: Physiological Noise



## ■ Cardiac effects

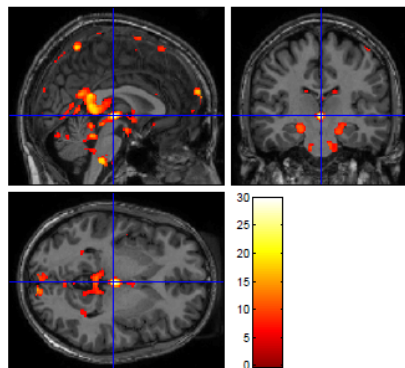
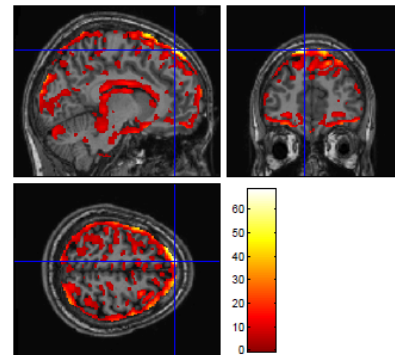
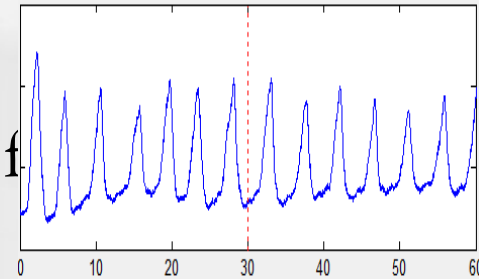


## Vessel Anatomy



## Locations of Fluctuations

# The Problem: Physiological Noise



## ■ Respiratory effects

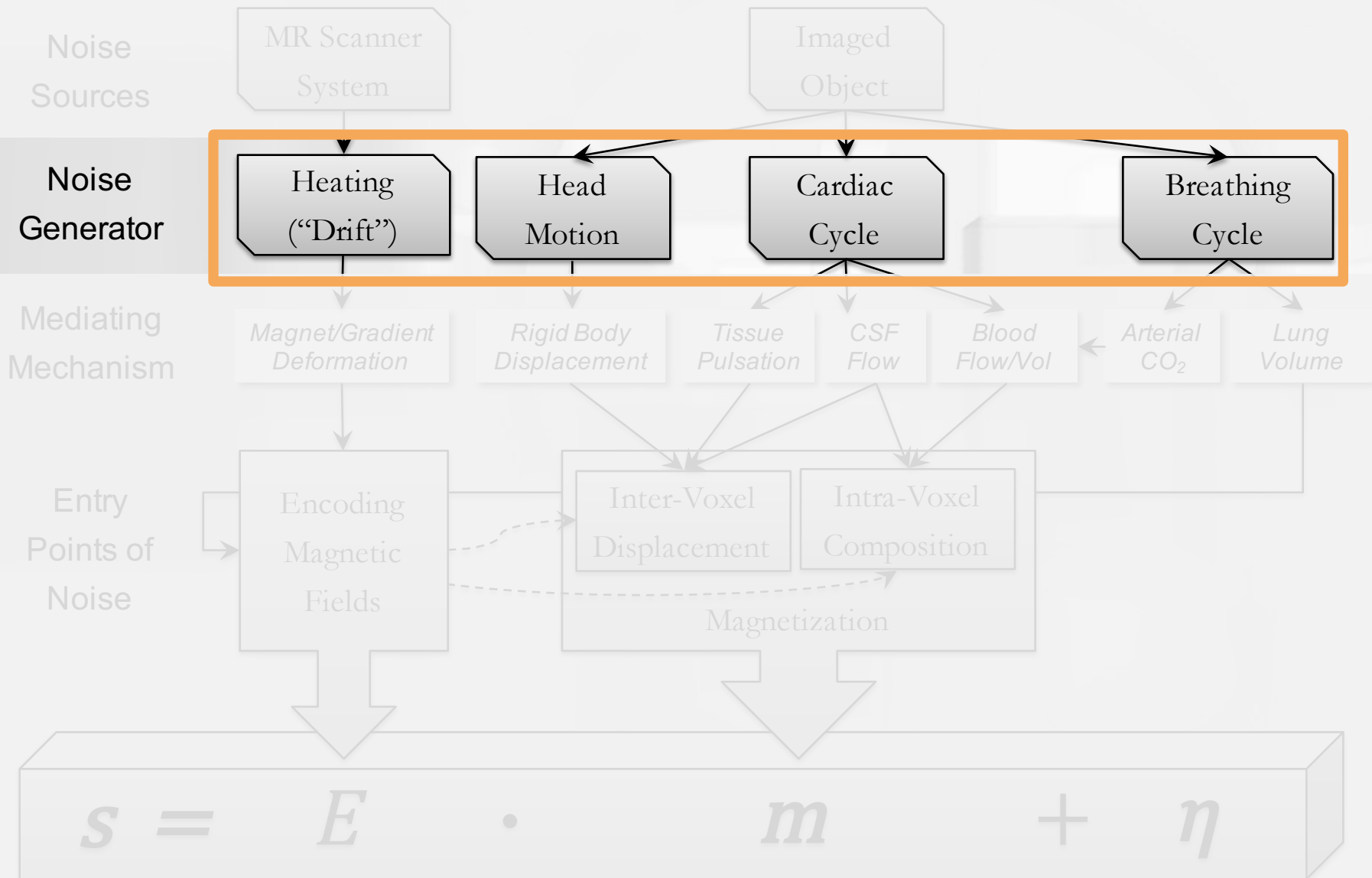
- Chest (& head) moves with respiratory cycle
- Changes in lung volume change encoding magnetic field for MR
  - Geometric distortion/scaling
- Respiratory-sinus arrhythmia
  - Heart beats faster during inhalation



- MRI Time Series Recap and Noise Sources
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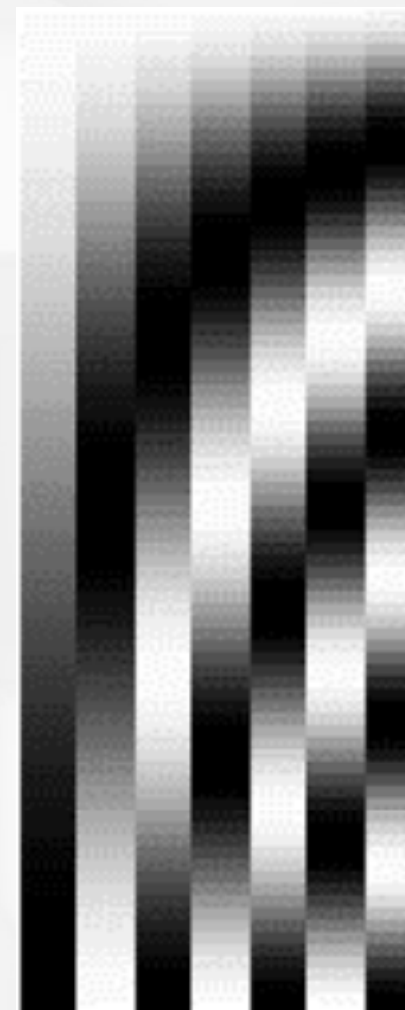
# Noise Correction Targets





- Discrete Cosine Model (last lecture) of slow oscillations (cycle  $\geq 128$  s)
- Was: Extra, non-task related columns in design matrix: **nuisance regressors**
- Now: Part of “hidden” preprocessing
  - Residual forming Matrix  $K = 1 - X_0$ 
    - With  $X_0$  being the design matrix modelling the confounds
  - In fact, GLM in SPM estimates

$$K \cdot \mathbf{y} = K \cdot X \cdot \boldsymbol{\beta} + K \cdot \boldsymbol{\varepsilon}$$





- Modelling:
  - Filters, projections (e.g. to independent components) etc. are all linear operations
  - could be in one design matrix, together with task
  - Simple test of correction efficacy: F-test on nuisance regressors
- Preprocessing:
  - The data  $y$  entering the GLM is altered  $\Rightarrow y' = X\beta + \varepsilon$
  - For non-linear changes of  $y$  or inter-voxel dependencies, alteration outside GLM necessary



- Problem: No inherent measure of efficacy (F-test in GLM), correlation with task regressors undetected
- “Advantage”: No loss of degrees of freedom (sensitivity of F-test)
  - But it is only a hidden loss, statistics for inference is biased, if performed modelling is not incorporated
- Modelling via GLM recommended, if possible
  - Motion Regr, Drifts, RETROICOR, HRV, RVT, aCompCor, (ICA)



- Correction for motion artifacts is actually a combination of Preprocessing and modelling
- Preprocessing:
  - Realignment
  - Motion “Scrubbing”
- Modelling (from estimated realignment parameters)
  - Retrospective Modelling: Motion Regressors
  - Motion Censoring





- Best: Avoid subject motion in the first place
- Better: Use Prospective Motion Correction
- Standard: Perform rigid-body realignment, use parameters as nuisance regressors
  - 6 parameters: translation+rotation
  - 12 parameters: include derivatives (for temporal shifts)
  - 24 parameters: include squared regressors
- 24-parameter model known as Volterra expansion

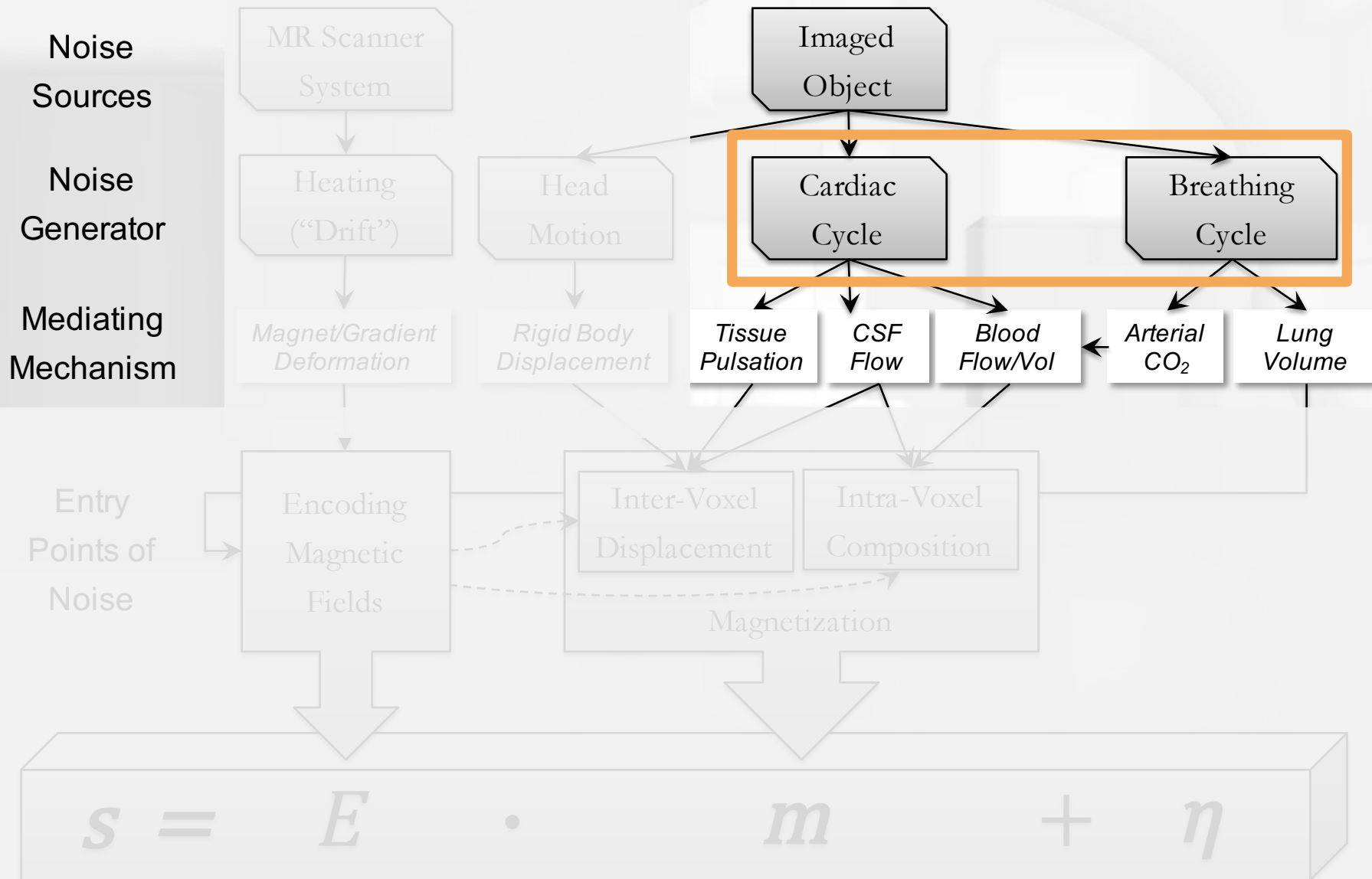
*Friston, MRM, 1996*

# Motion Censoring = “Scrubbing”

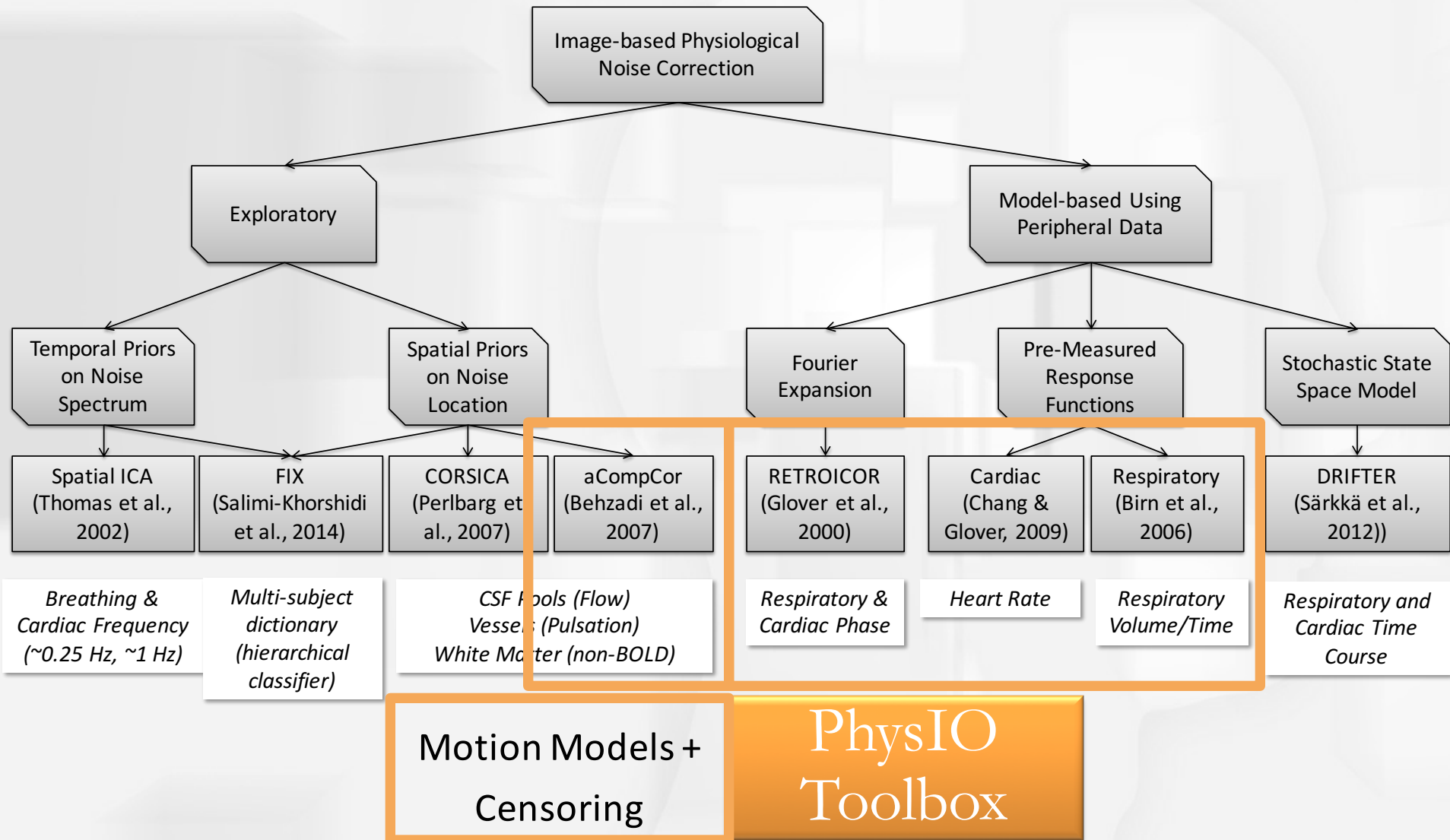


- Detect outlier volumes (strong movement, but also spikes, RF flip angle fluctuations)
- Inform the GLM of these bad volumes via stick regressors (zero everywhere else, 1 at volume)
  - Will absorb all variance of that volume
- Problem: Temporal filtering before GLM might create Gibbs ringing of outliers into neighbors
- Alternative: censoring during preprocessing
  - interpolate faulty volume by neighbors *Power, NeuroImage, 2012*

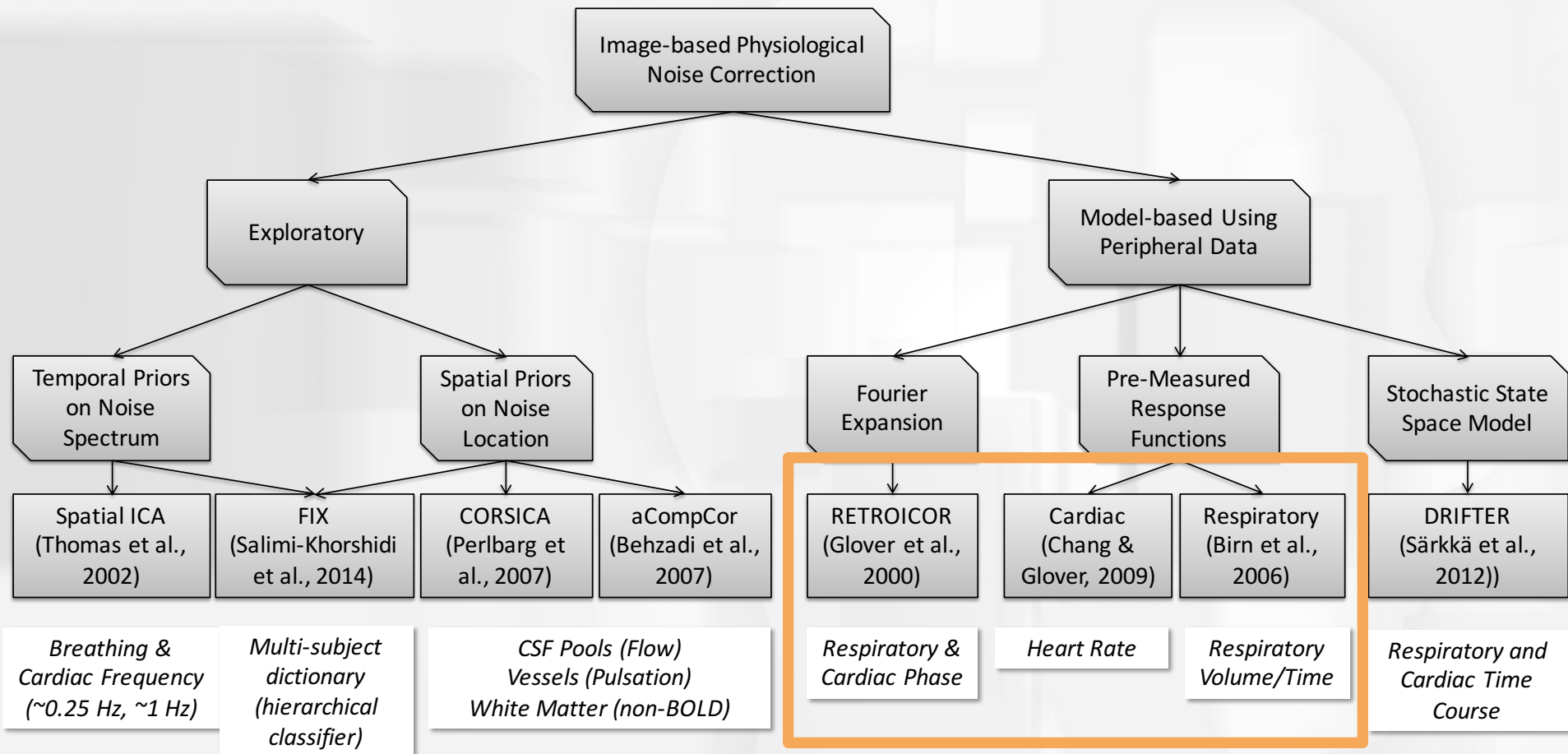
# Noise Correction Targets



# Image-based Noise Correction



# Model-based Phys Noise Correction







## RETROspective Image CORrection

## Cardiac Response Function

## Respiratory Response Function

- Cardiac/respiratory phase  $\varphi_c$   $\varphi_r$
- Fourier expansion (cosine/sine)
- Heart Rate
- convolved with CRF
- Resp. Volume per Time
- convolved with RRF
- evaluated at 1 time point (slice) per volume = regressor

# Noise Modelling

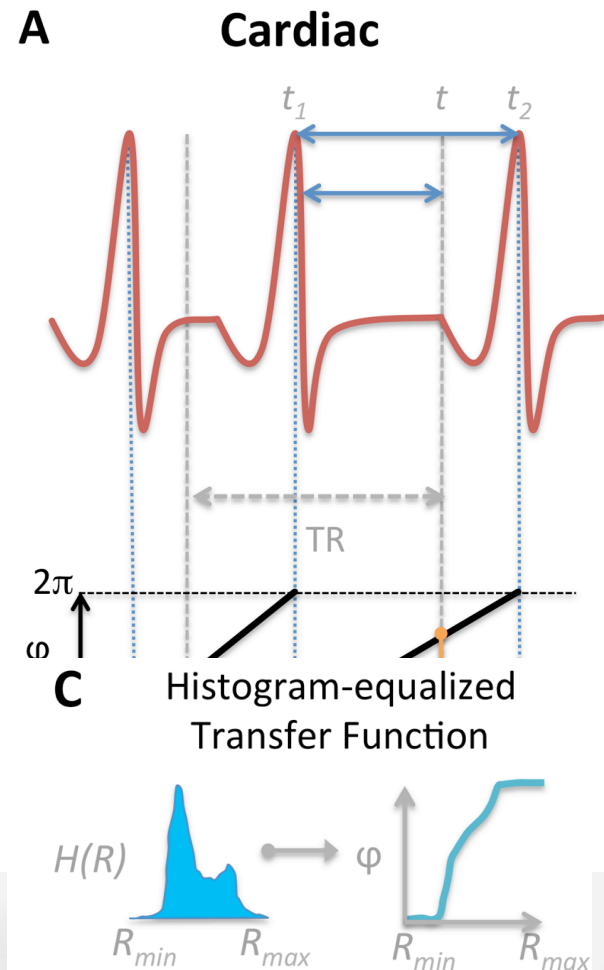


## RETROspective Image CORrection

- Cardiac/respiratory phase

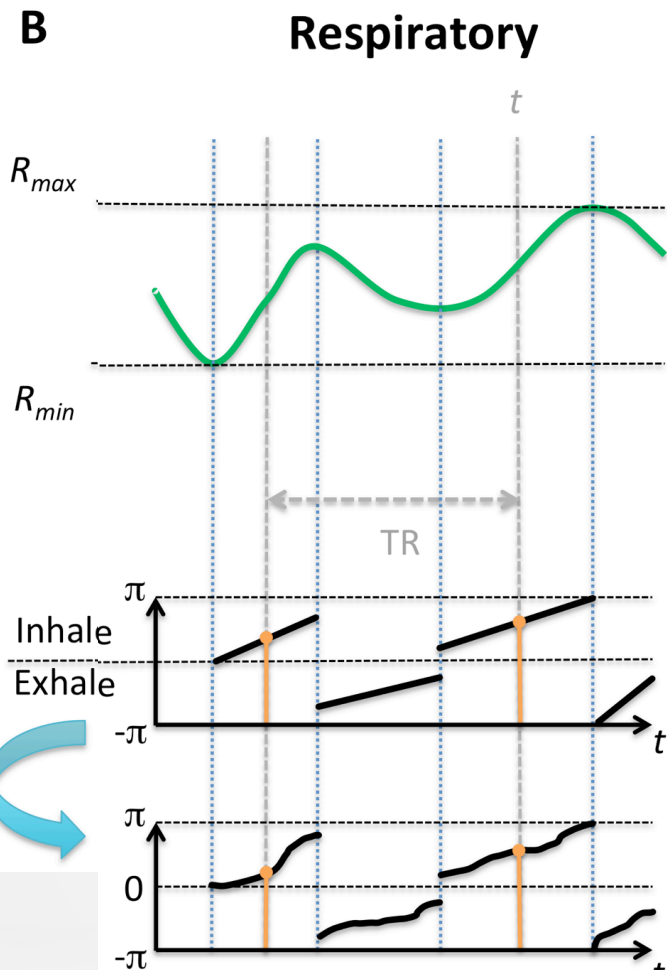
Cardiac Response Function  $\varphi_C$

Respiratory Response Function  $\varphi_r$



Peripheral Signal

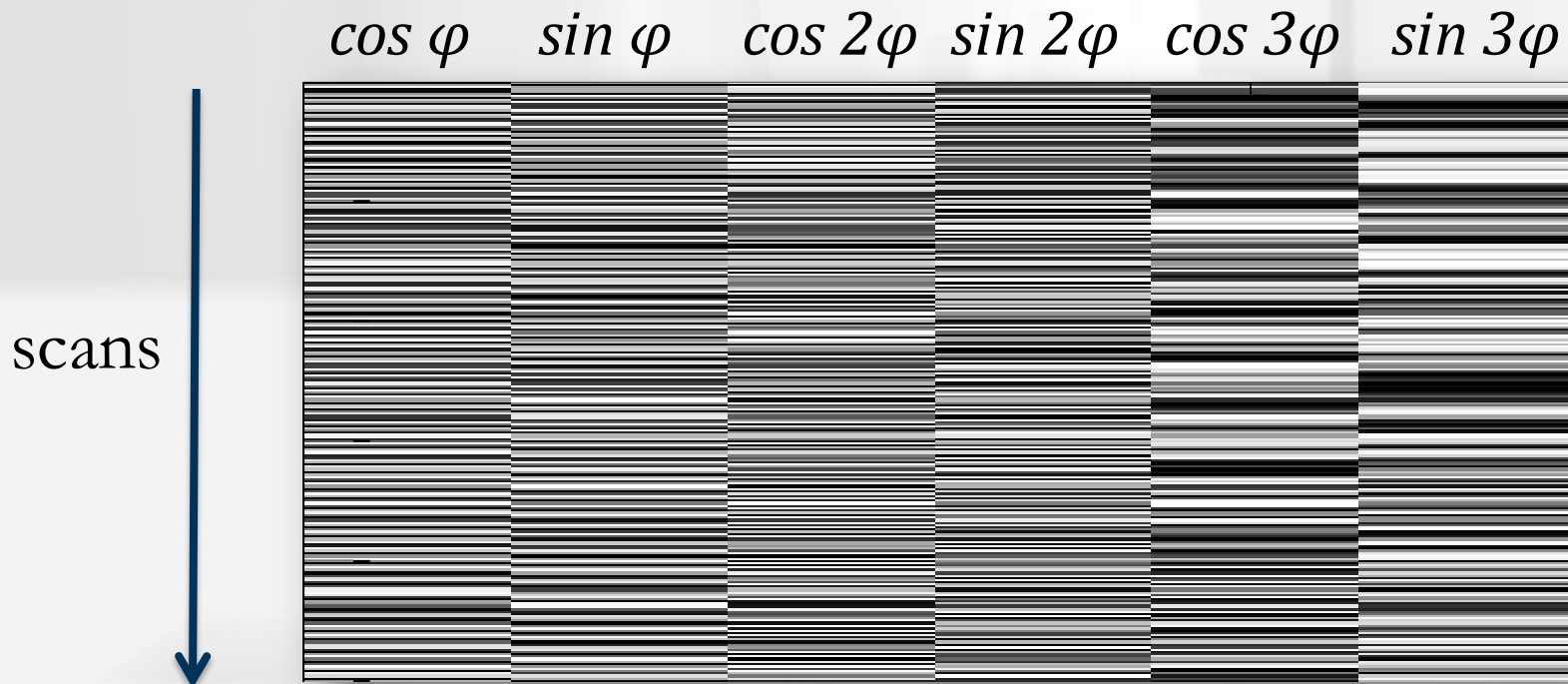
Estimated Phase



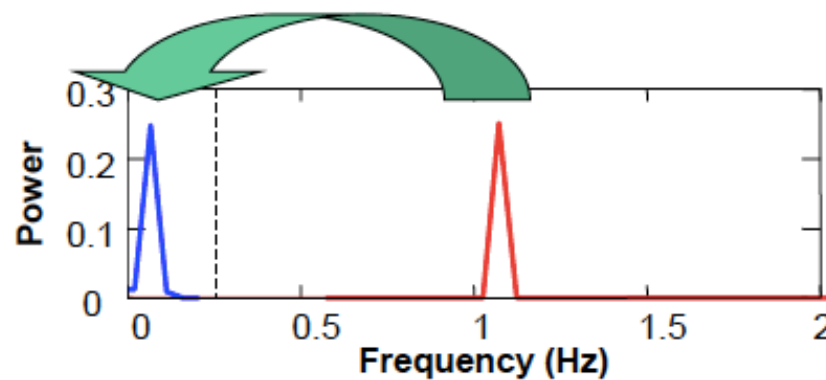
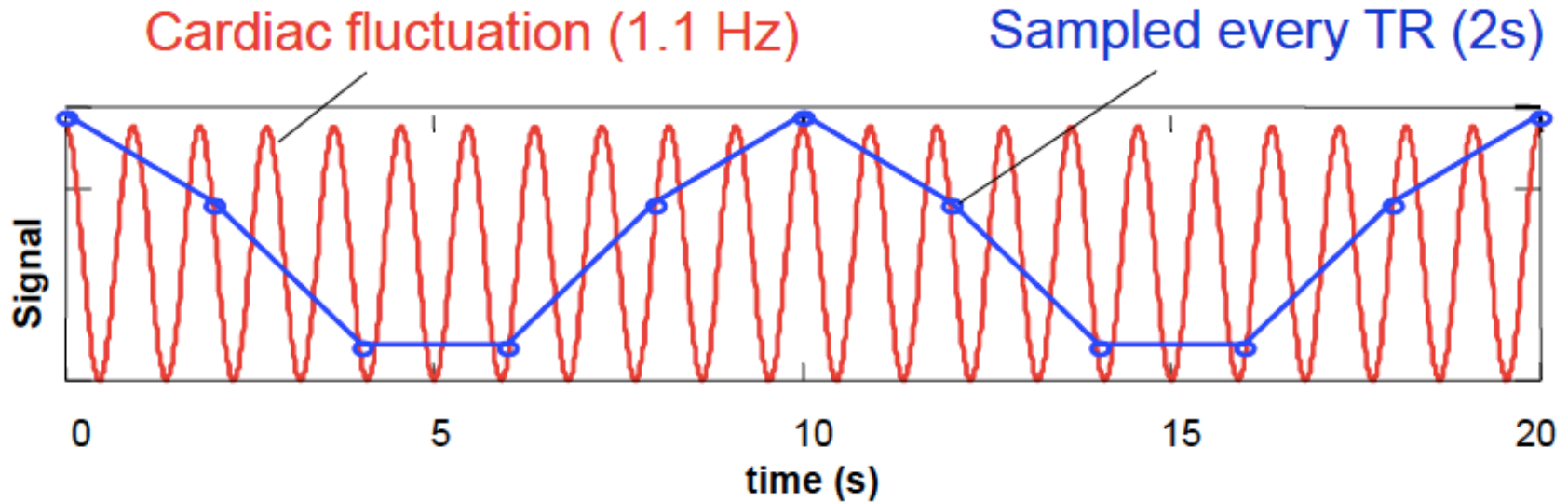
# Model: Fourier Phase Expansion



- Cosine and sine to allow for constant phase shifts per voxel
- Higher model orders to account for under-sampling of physiological frequencies with typical TR in fMRI



# Aliasing of Physiology



Courtesy: R. Birn, HBM 2015

# Noise Modelling



RETROspective  
Image CORrection

Cardiac Response  
Function

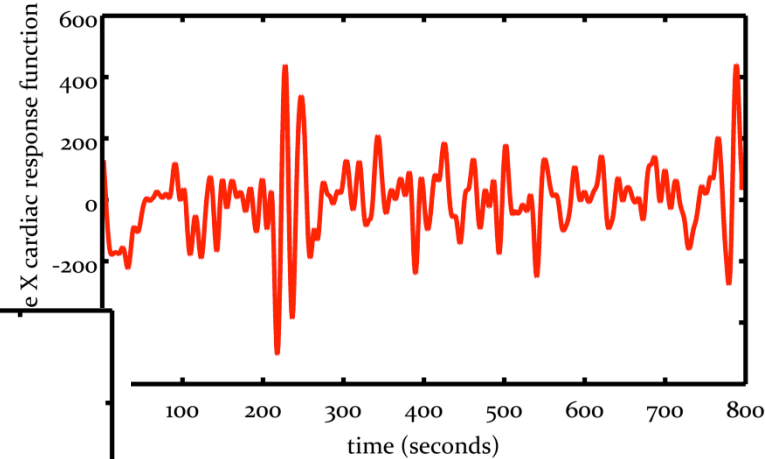
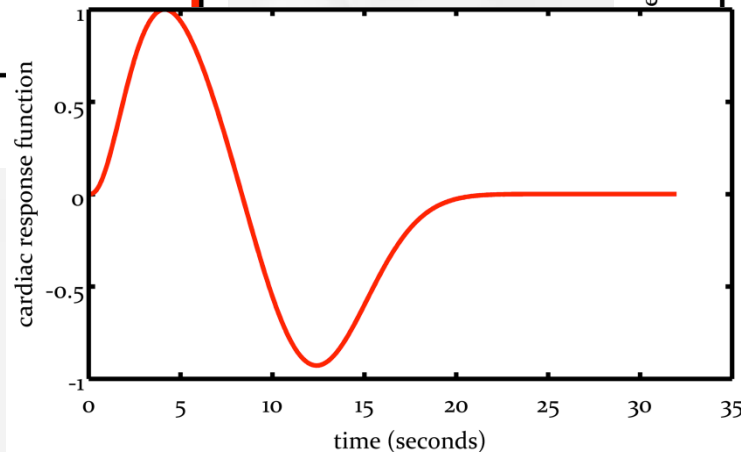
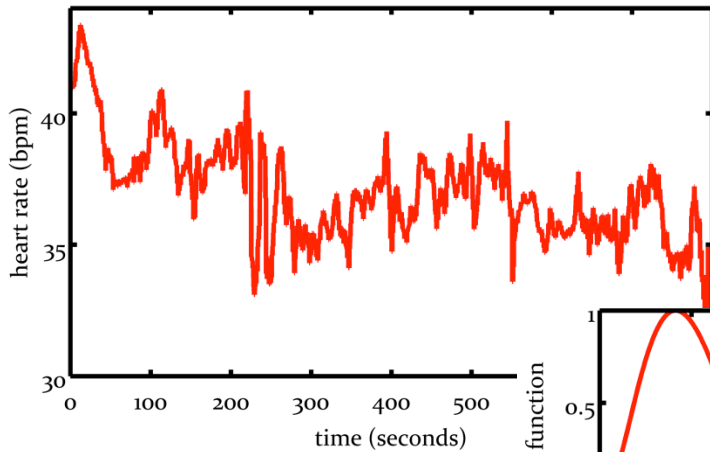
Respiratory  
Response Function

- Heart Rate

- Heart Rate Variability  
Response Regressor

- convolved with

CRF



# Noise Modelling



RETROspective  
Image CORrection

Cardiac Response  
Function

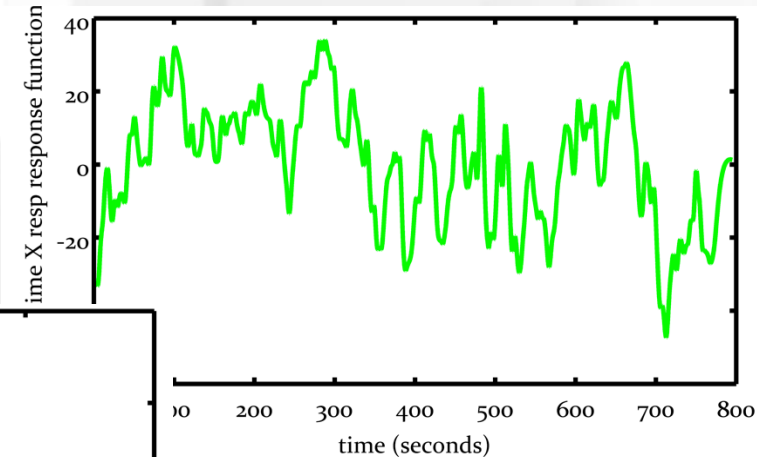
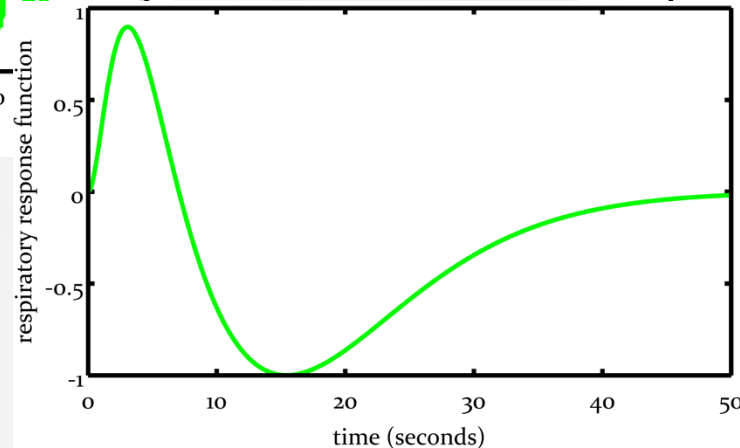
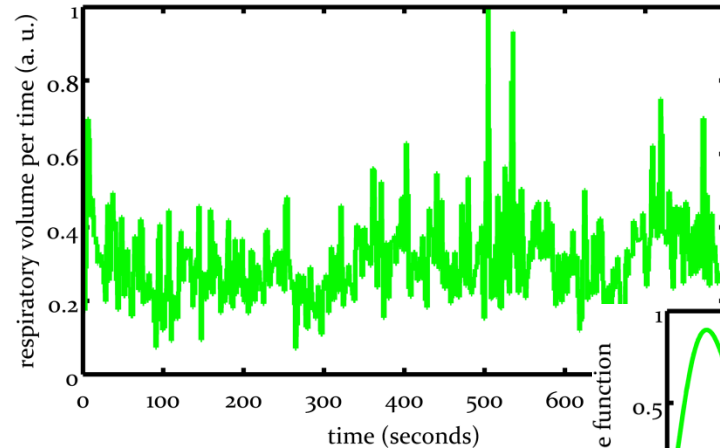
Respiratory  
Response Function

■ Respiratory  
Volume per Time

■ convolved with

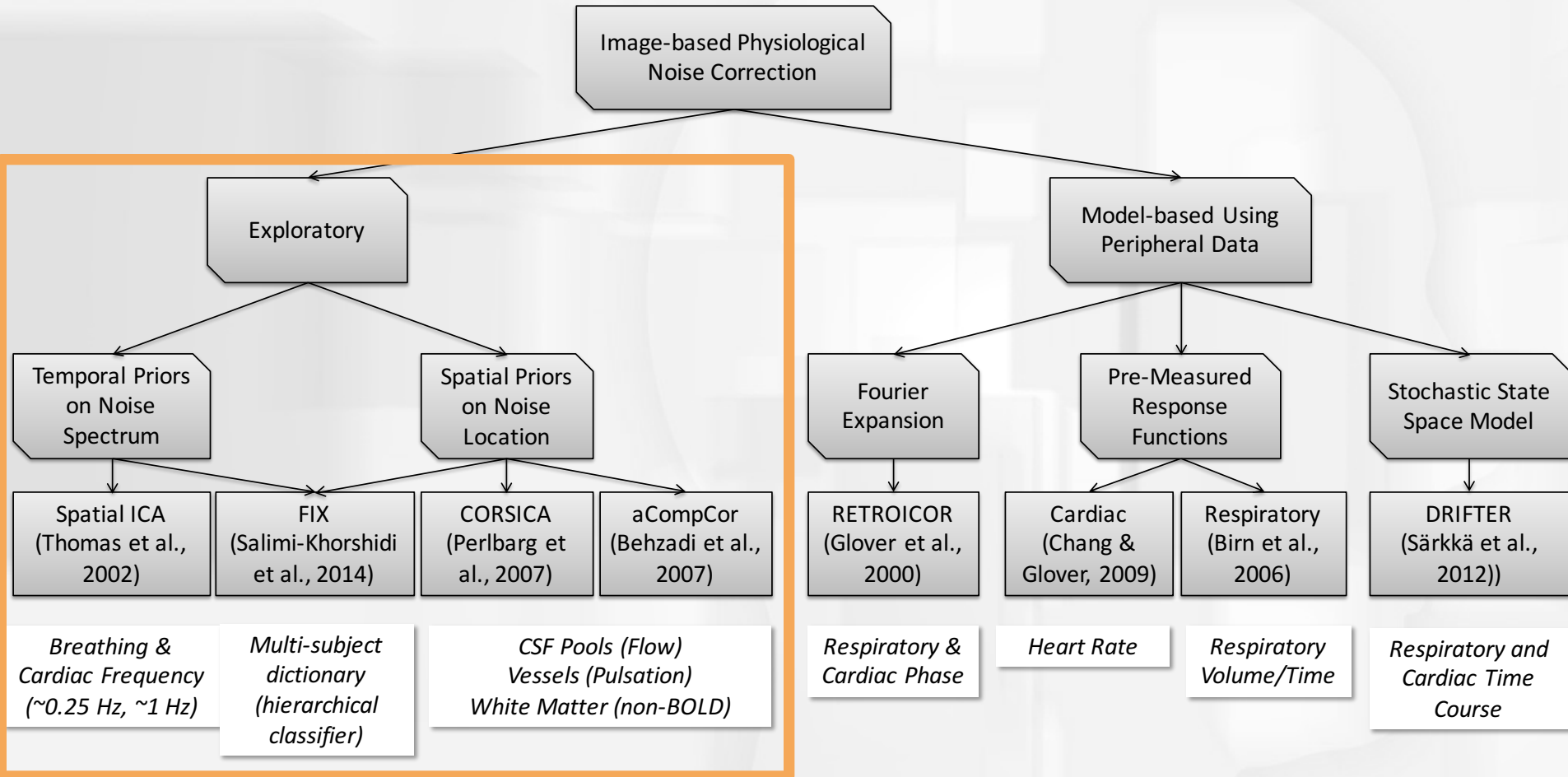
■ Respiratory Volume  
per Time Regressor

RRF





# Exploratory Phys Noise Correction



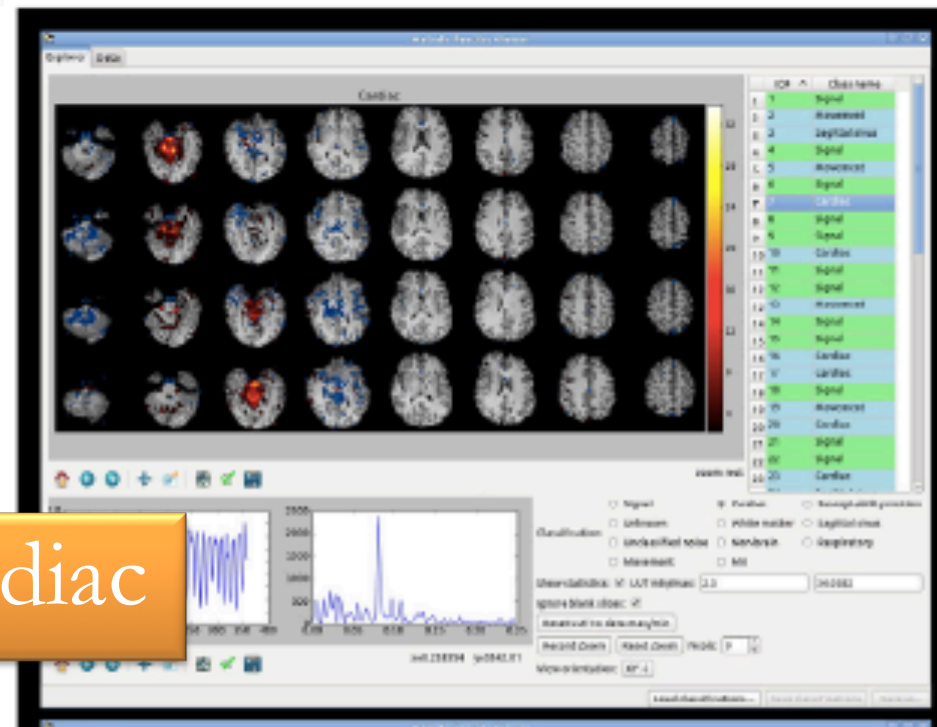
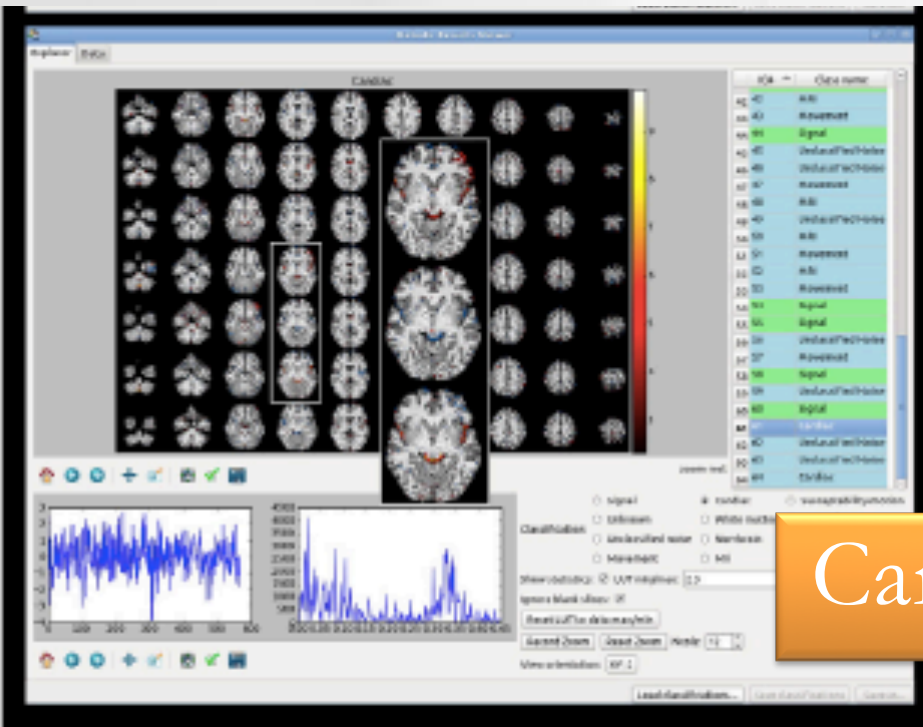


- Use priors about physiological noise to identify noise components (time series)
  - Spatial Priors: Mechanisms of physiological noise implicate physiological noise in CSF, blood vessels
  - Temporal Priors: Knowledge about typical physiological frequency contents (heart  $\sim 1$ Hz, breathing 0.2-0.4 Hz)
    - Note that simple filtering is impossible (cf. aliasing)
  - Population Priors: Use dictionary learning from manually labelled training set of subjects (FIX)



- Methods to extract components (i.e. summarize ROIs/spectra) differ:
  - Maximum variance representative time series: Principal Component Analysis (PCA) from region of interest (aCompCor, Behzadi 2007)
  - Maximally independent time courses/sites: spatial/temporal ICA, FSL MELODIC, FIX

# Preprocessing Techniques



Cardiac



- Non-linear models
  - DRIFTER: Kalman Filter, Bayesian, *Joint* Stochastic State-space model of peripheral physiology and BOLD
- Identify noise via task test-retest reproducibility
  - PHYCAA: e.g. via high-freq. autocorrelation, anatomy
  - GLMDnoise: PCA of noise regressors
- MEICA: Multi-Echo ICA
  - Use diff. TE-images to decompose proton density from  $T2^*$  changes

*Särkkä, NeuroImage, 2012*

*Churchill, NeuroImage, 2012/13*

*Kay, Front. Neurosc., 2013*

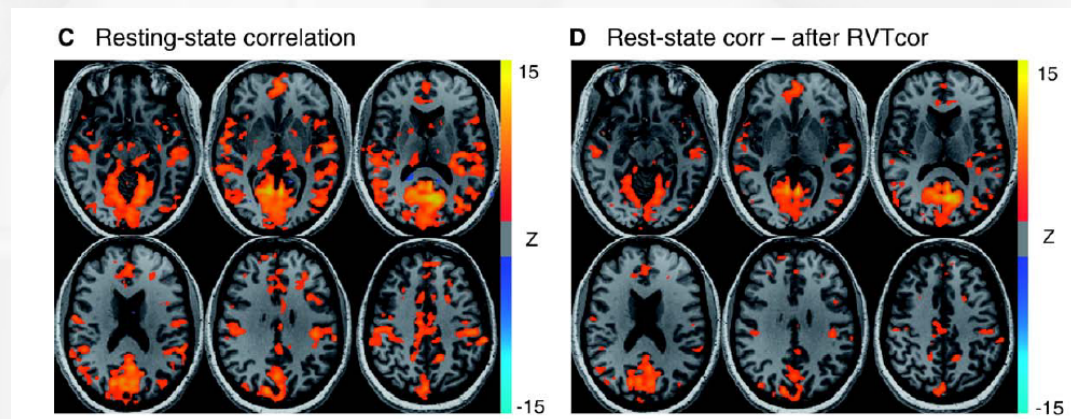
*Olafsson, NeuroImage, 2015*

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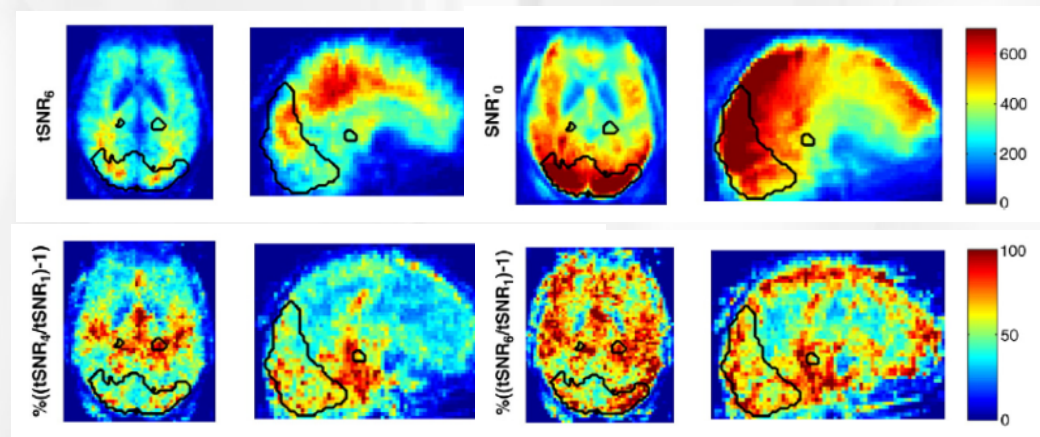
## ■ Resting-state:

- Birn, R. M. “The Role of Physiological Noise in Resting-state Functional Connectivity.” *NeuroImage* 62, 2012
- Birn, R. M., et al. “Separating Respiratory-variation-related Fluctuations from Neuronal-activity-related Fluctuations in fMRI.” *NeuroImage* 31, 2006



## ■ Task-based:

- Hutton, C., et al. “The Impact of Physiological Noise Correction on fMRI at 7 T.” *NeuroImage* 57, 2011:





- Physiological noise correction not a default pre-processing step in task-based fMRI
- Reasons
  - Impact on group level fMRI
    - no reports for non-trivial paradigms
  - Existing Toolboxes lack...
    - robust, automatic implementation
    - dealing with variable peripheral data quality



- Hierarchical learning of trustworthiness of advisor over time
- Contrasts: Prediction and Prediction Error about advice



recommendations of adviser were **veridical** (pre-recorded videos from behavioural study)

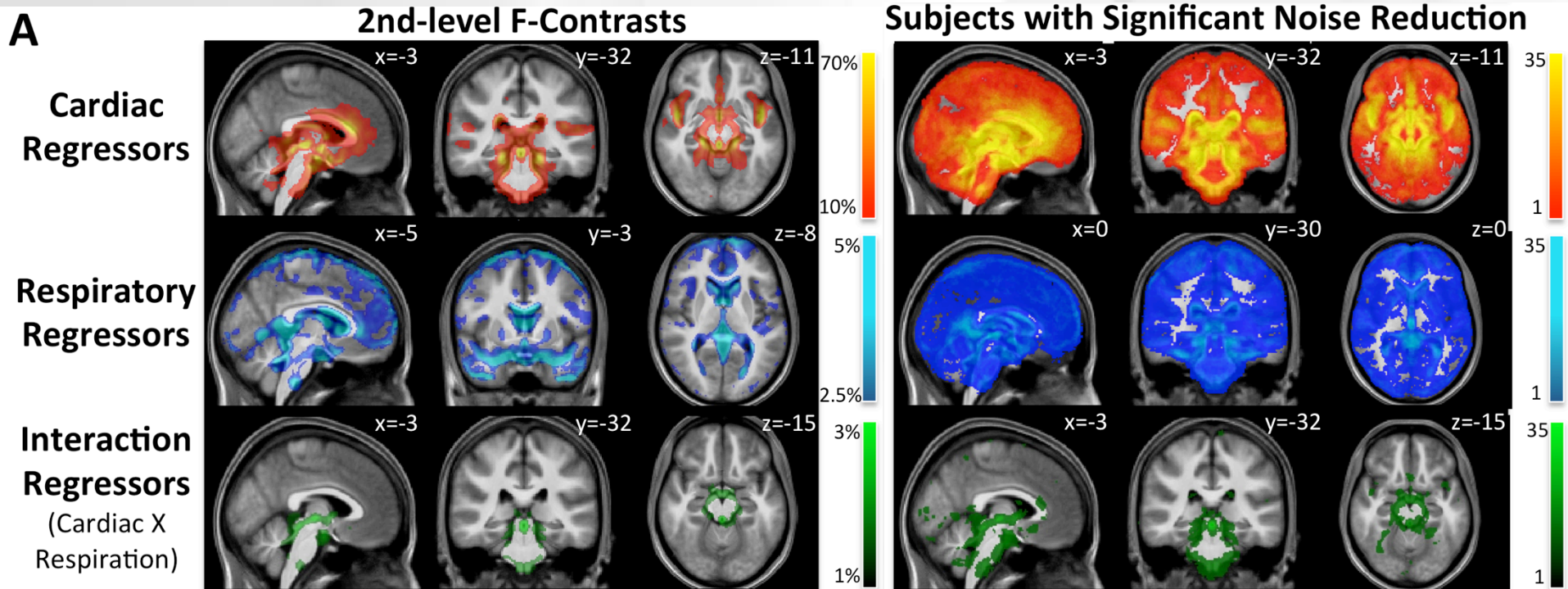
**volatility of advice** (changing intentions of adviser through incentive structure)

interactive, gender-matched (**40** male subjects)

**fMRI design:** Philips Achieva 3T  
TR/TE 2500/36ms, 2 x 2 x 3 mm<sup>3</sup>

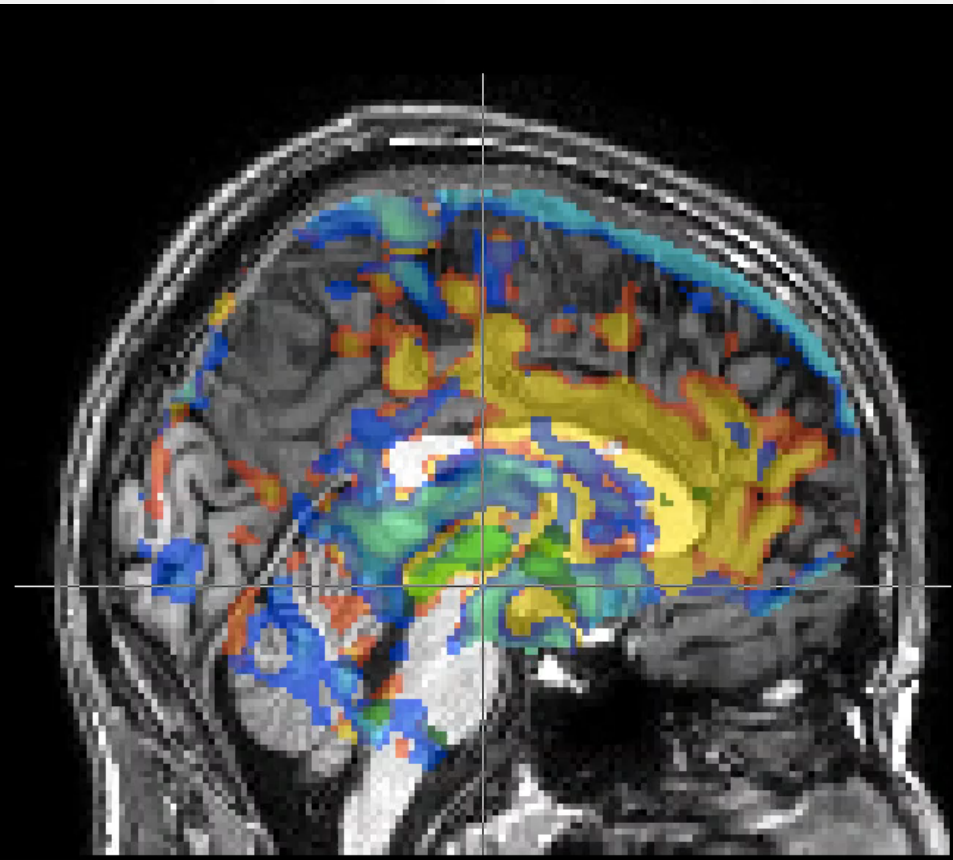
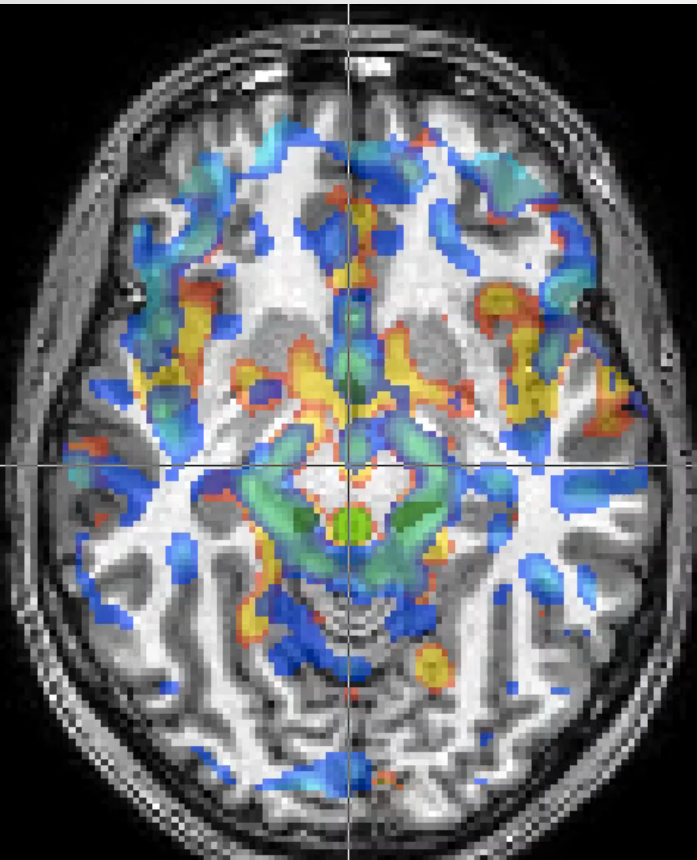
*Diaconescu et al, 2014, PLoS Comp. Biol.*

- Andreea Diaconescu (TNU): Social Learning Experiment 2012-2014, (N=35)
- F-contrast: Where does physiological noise model explain significant variance?



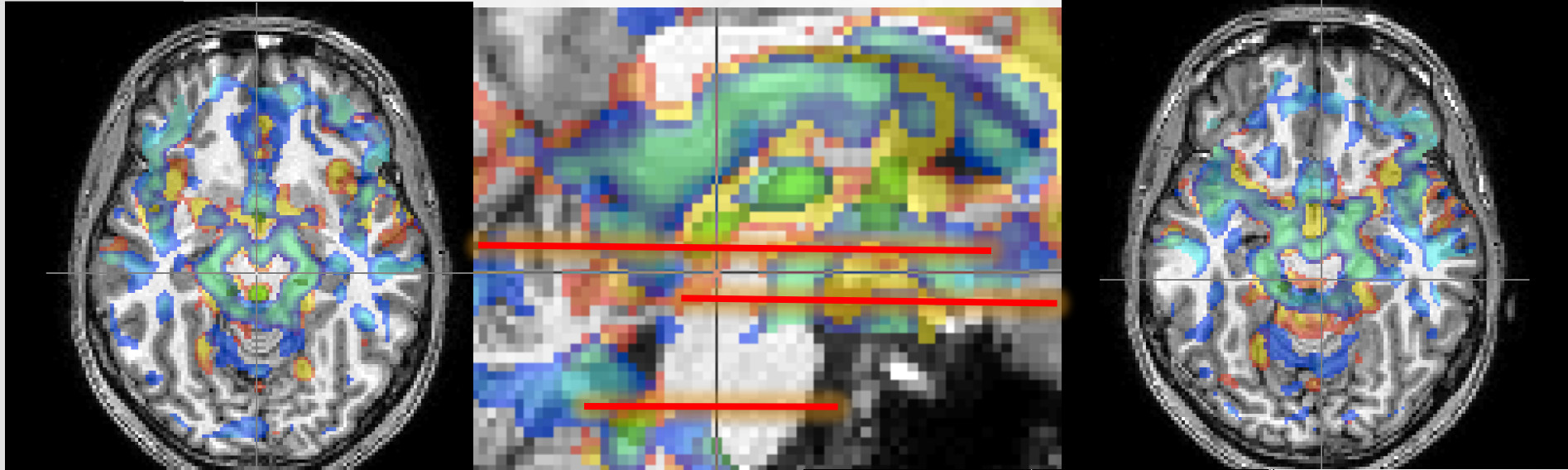


- cardiac (red), respiratory (blue), cardXresp (green)

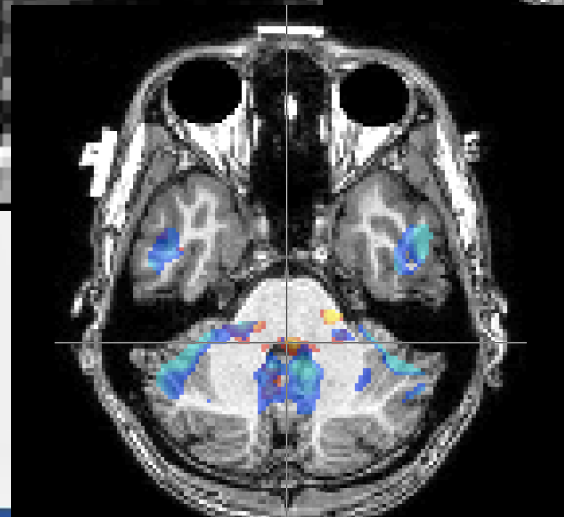


- VTA (DA)

- Raphe Nuclei (5-HT)

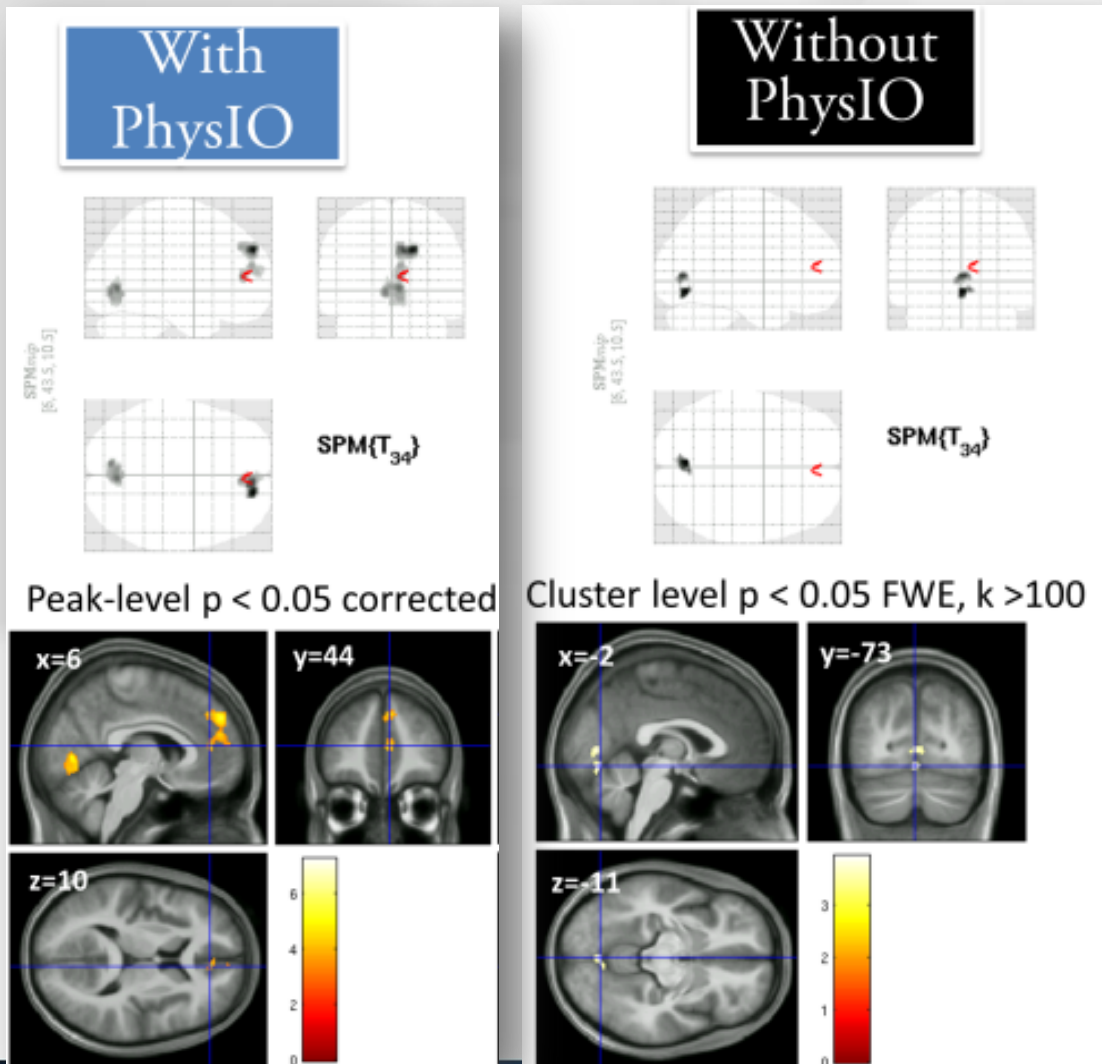


- Locus coeruleus (NA)

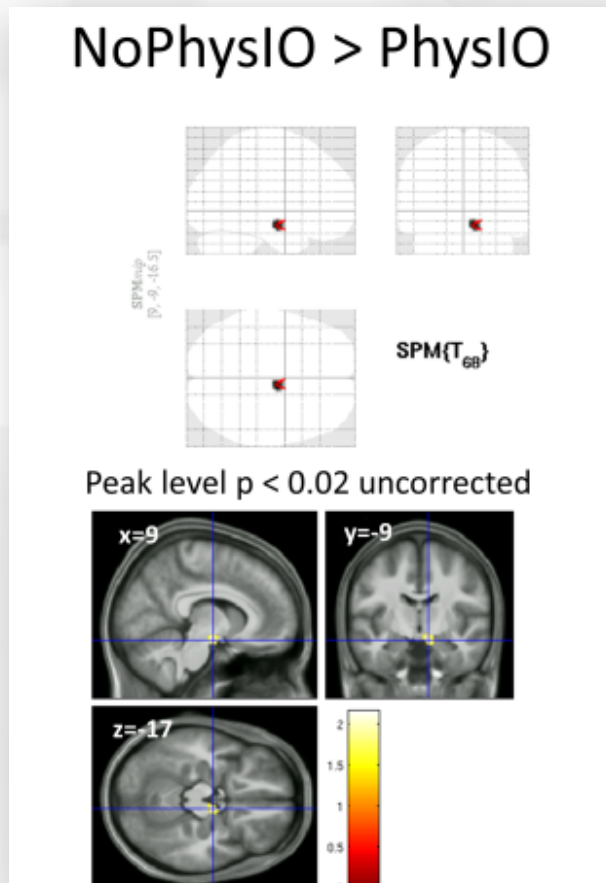




## Higher Sensitivity

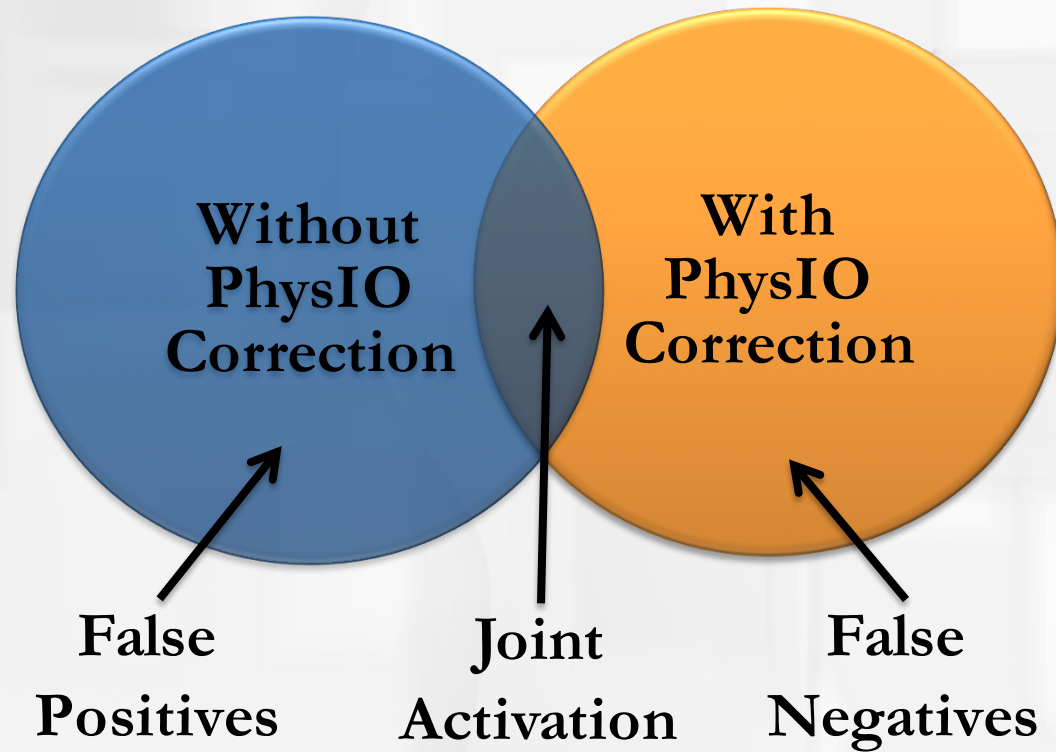


## False Positives



2<sup>nd</sup> level t-contrast  
Social Prediction Error

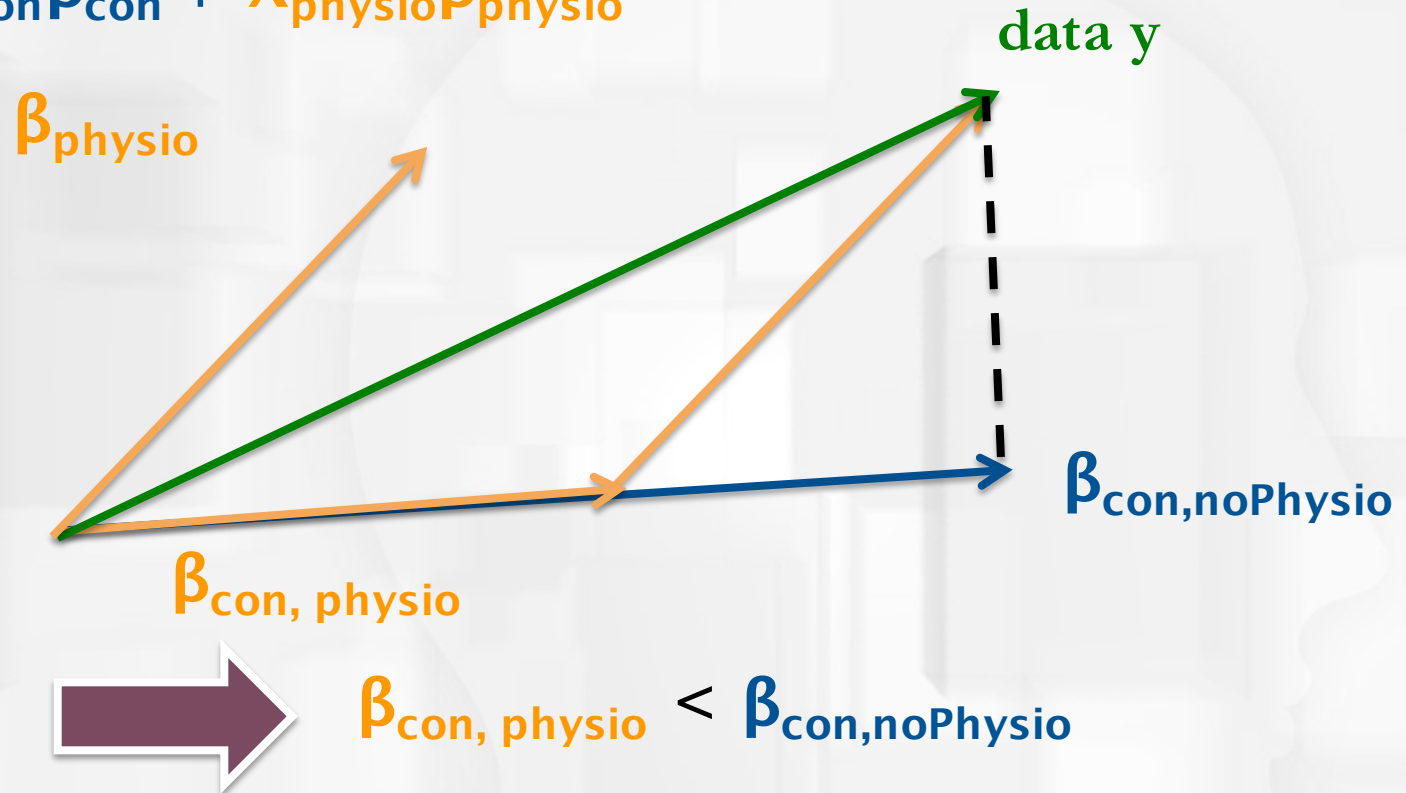
## Significant Clusters



# Noise modelling impact on task?



$$y = X_{\text{con}}\beta_{\text{con}} + X_{\text{physio}}\beta_{\text{physio}}$$



- Phys noise correction can change parameter estimates for regressors of interest (correlation!)
- Thereby change distribution of  $\beta \Rightarrow$  Mean? Variance?



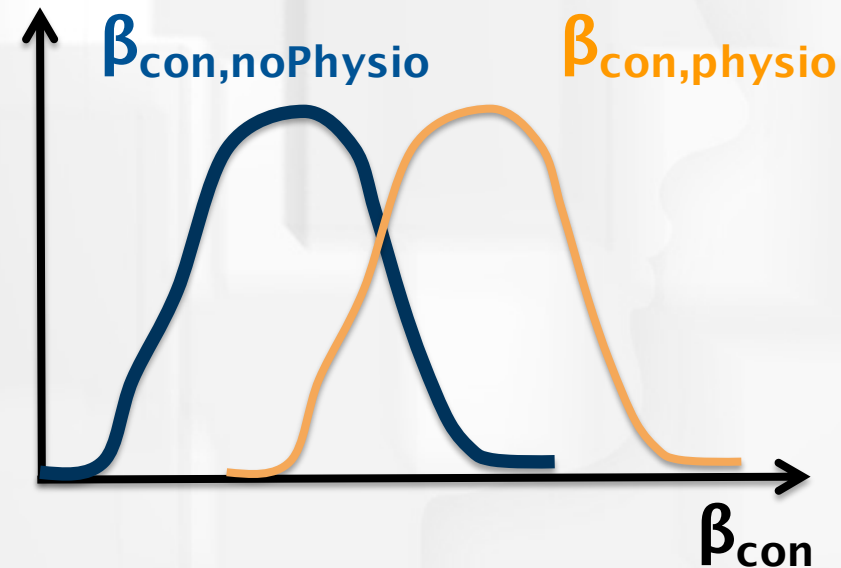
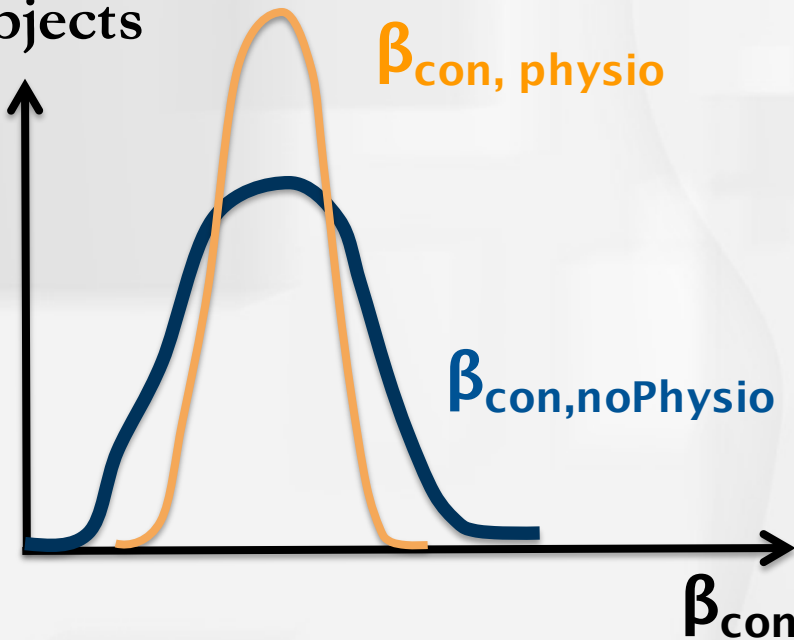
- Two mechanisms imaginable

Reduced inter-subject  
variance

Increased inter-subject  
mean estimates

frequency of  
subjects

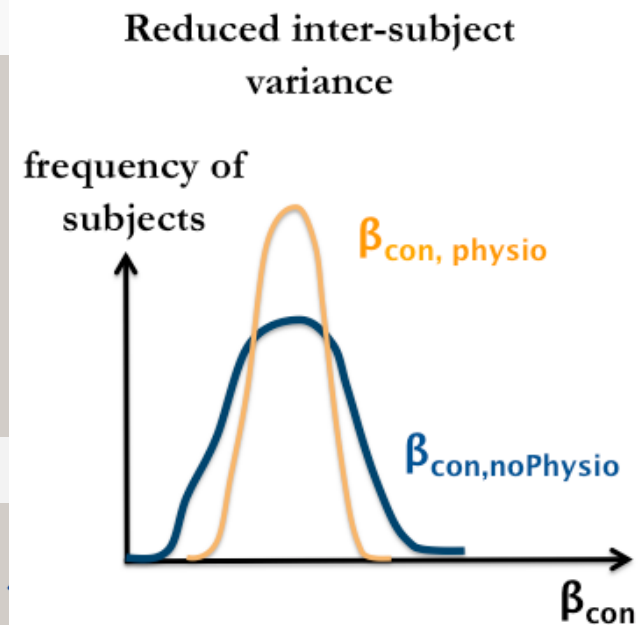
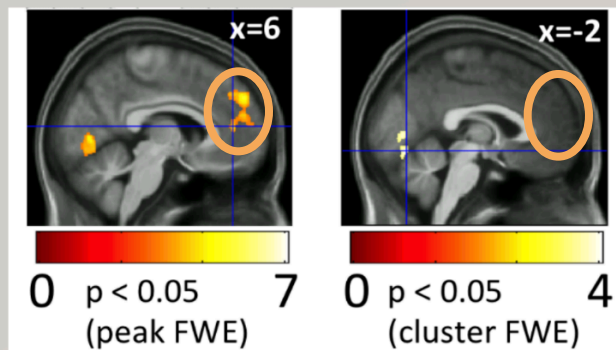
frequency of  
subjects



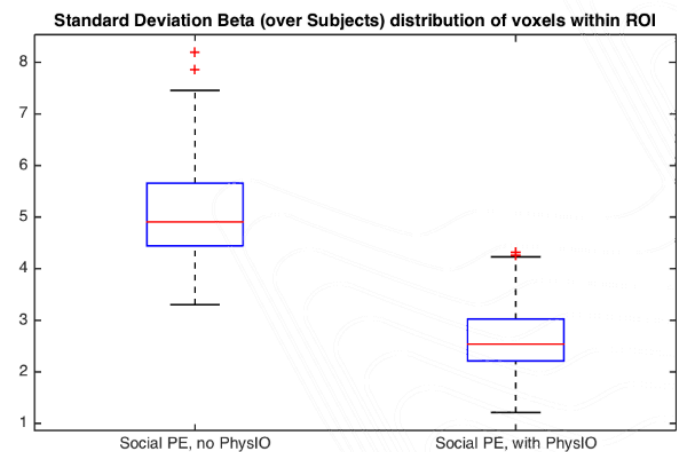
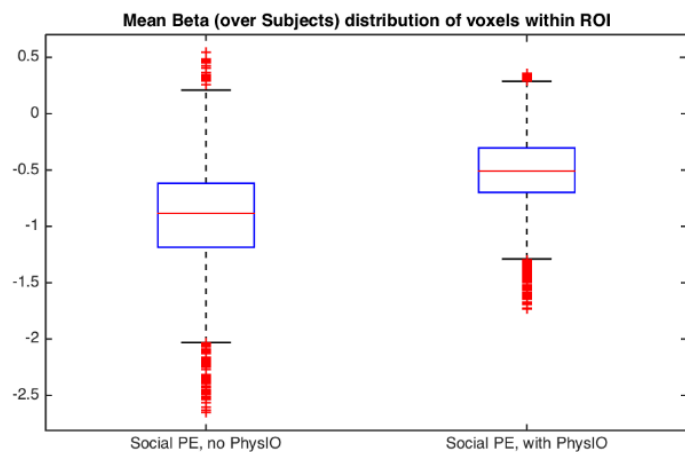
# The Corrective Mechanism



## Social Pred. Error



## Mean



- MRI Time Series Recap and Noise Sources
  - Why de-noising? Noise pathways: Scanner, Cardiac/Respiratory/Motion
- Noise Correction Approaches
  - Correction Target: Drift, Motion, Cardiac/Breathing Cycle
  - Data Correction Point: Modelling VS Preprocessing
  - Noise Model Input: fMRI Data-driven VS Peripheral Measures
- Noise Correction Prospects
  - Effects of Physiological Noise on Group Statistics
- **Noise Correction Limitations**
  - Degrees of Freedom; Task-related “noise”; Interoception



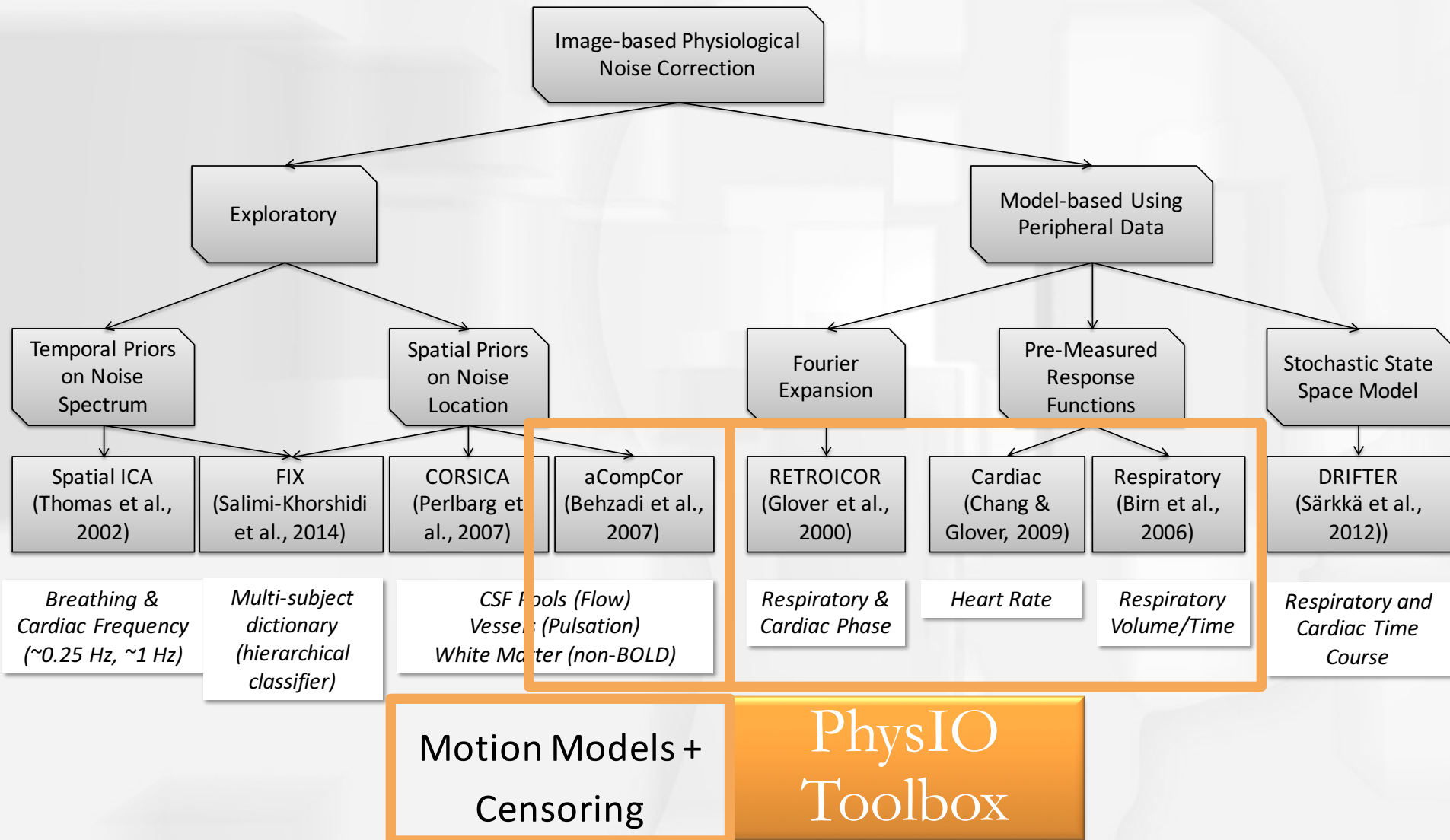


- Degrees of freedom, sensitivity reduced by too many ineffective regressors
  - F-test informative
- Intrinsic correlations of functional areas of interoception and peripheral physiology
  - E.g. Amygdala, Insula, ACC
  - Controversial reading: [fMRI of the Amygdala: All In Vein? – Neuroskeptic](#)
  - Alternative: Masking, Pure anatomical priors removing CSF, angiography (vessels)



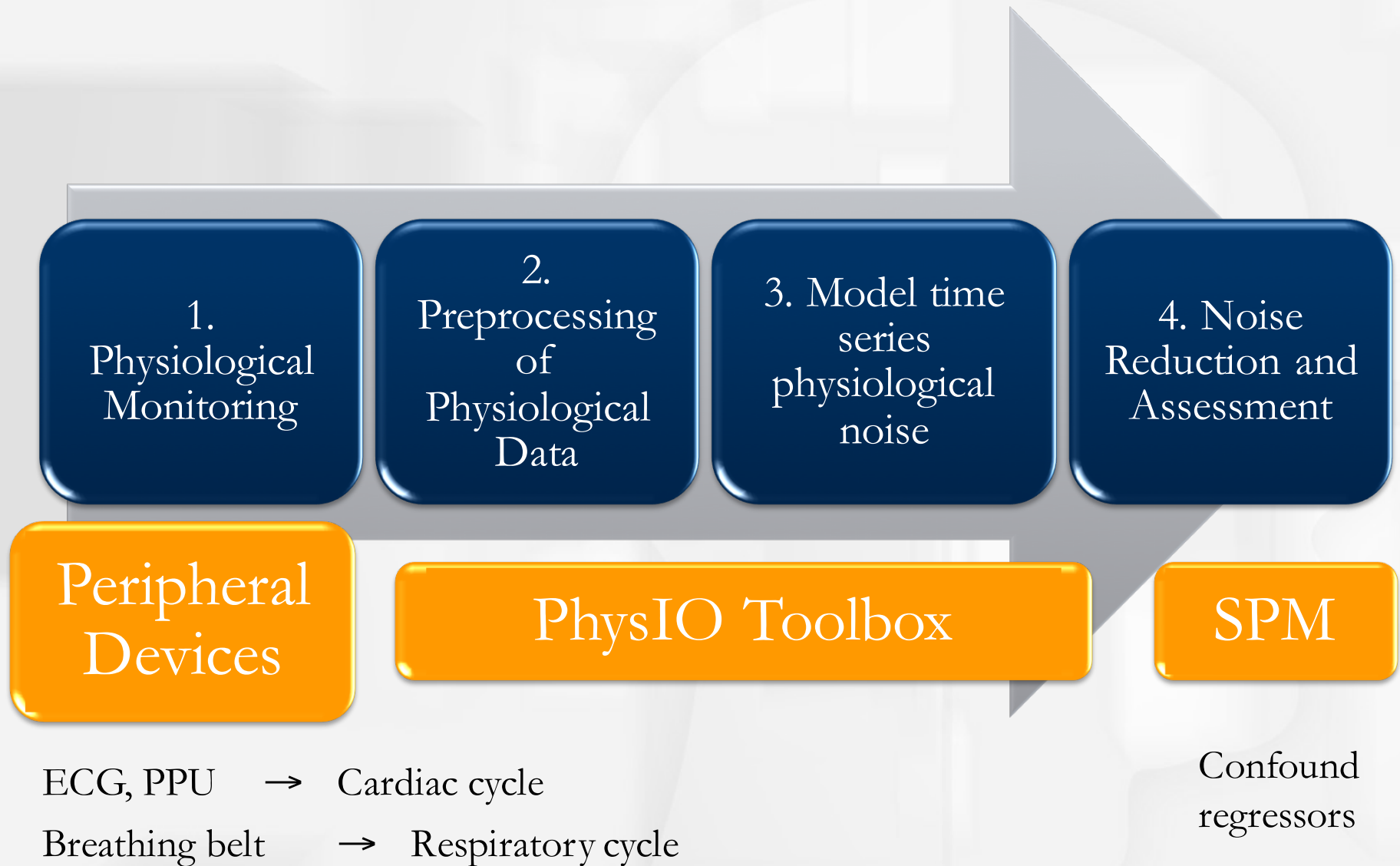
- MRI Time Series and Physiological Noise
- Noise Modeling
- Prospects: Group FX
- Image-Based Correction in the GLM
- The PhysIO Toolbox
- Structured noise through cardiac/resp cycle (70%)
- Increase group sensitivity (low inter-subject variability), fewer false positives
- Nuisance regressors from Fourier expansion or response functions
- Correction in SPM/Matlab practice => **NOW!**

# Image-based Noise Correction





- Developed at the Translational Neuromodeling Unit (TNU) since 2008
  - Lead programmer: Lars Kasper (TNU)
  - Contributors: Jakob Heinzle (TNU), Steffen Bollmann (KiSpi Zurich)
- Part of the TNU «TAPAS» software suite
- Used at the TNU, in Zurich and beyond by ~30 researchers (>500 subjects)
  - e.g., Iglesias 2013, Neuron; Kasper 2014, NeuroImage; Bollmann 2014, PhDThesis (ADHD in Children); Sulzer 2013, NeuroImage (NeuroFeedback)
- Current version:
  - <http://www.translationalneuromodeling.org/tapas/>
- Documentation & Example Data:
  - <http://www.translationalneuromodeling.org/software/documentation/>
  - <http://www.translationalneuromodeling.org/software/tapas-data/>



# Workflow PhysIO Toolbox



Read logfiles

Preprocess  
physiological data

Model time series  
physiological noise

Include confound  
regressors (GLM)

# Workflow PhysIO Toolbox



Read logfiles

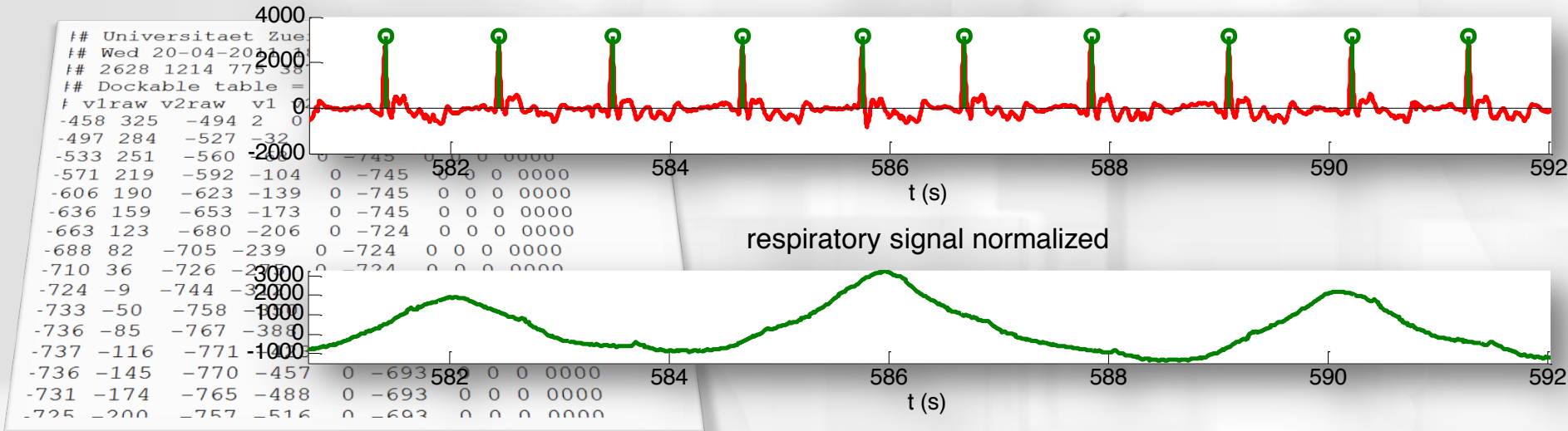
Preprocess  
physiological data

Model time series  
physiological noise

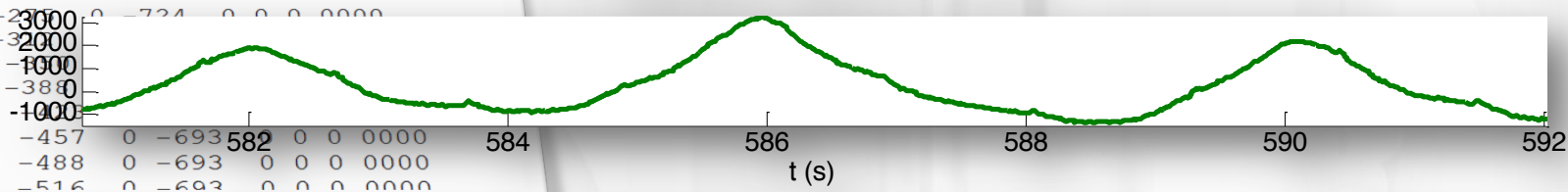
Include confound  
regressors (GLM)

SCANPHYSLOG.log

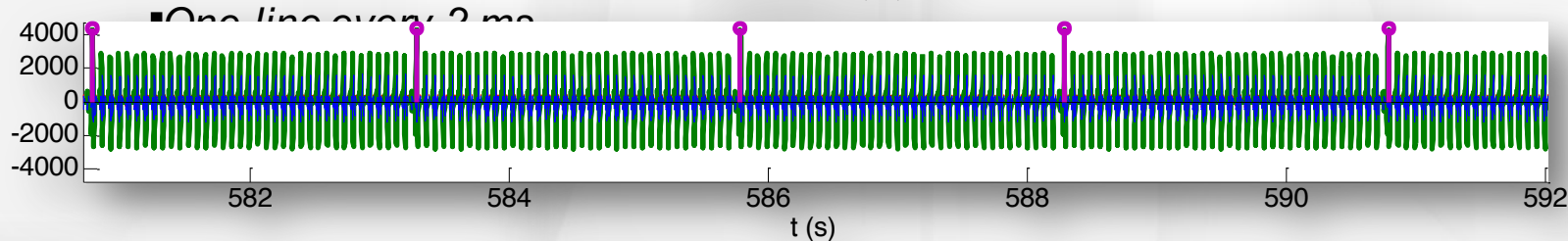
ECG normalized



respiratory signal normalized



Gradient X,Y,Z





# Workflow PhysIO Toolbox



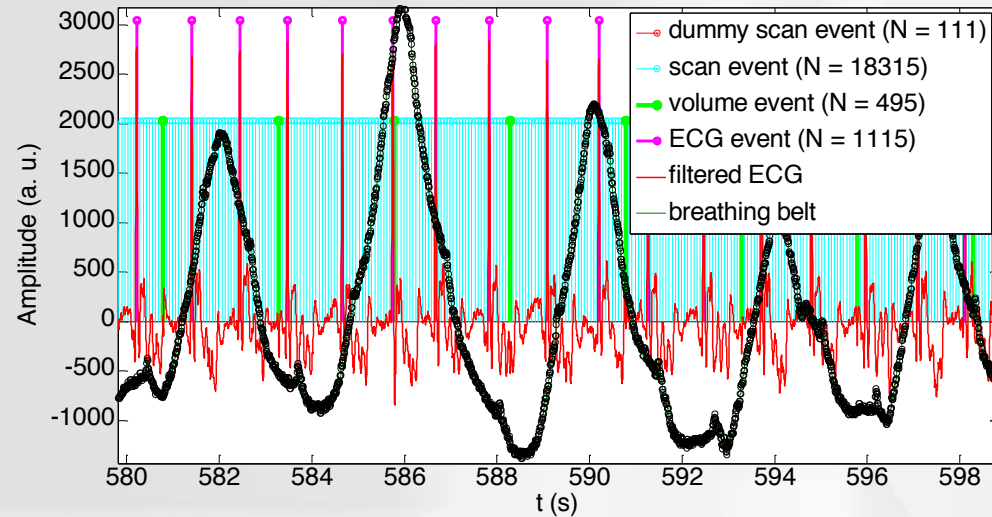
Read logfiles

Preprocess  
physiological data

Model time series  
physiological noise

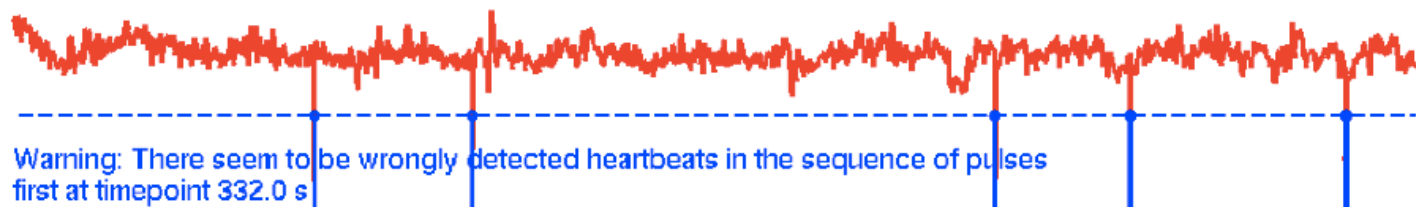
Include confound  
regressors (GLM)

Cutout region for RETROICOR



- Align scan timing to physiological time series

- Misdetected heartbeats





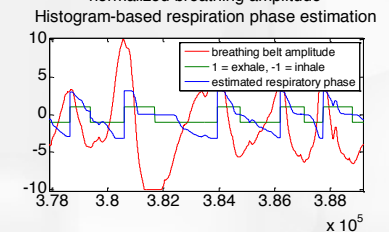
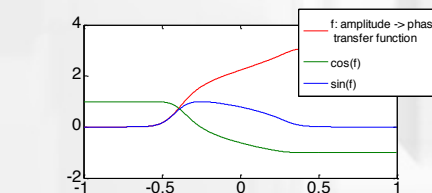
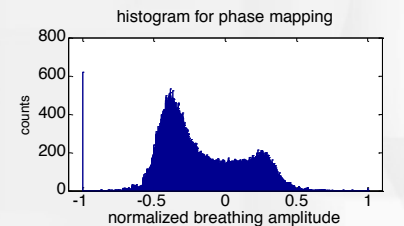
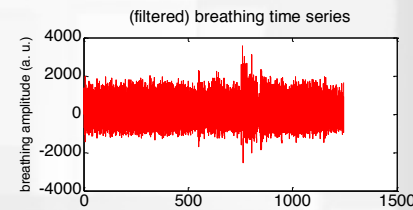
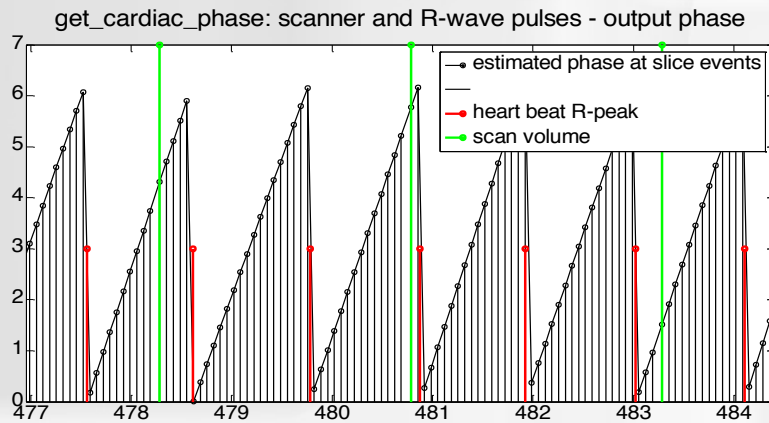
Read logfiles

Preprocess  
physiological data

Model time series  
physiological noise

Include confound  
regressors (GLM)

## Cardiac & respiratory phase estimation



## Regressors via Fourier expansion of phases:

RETROspective Image CORrection (RETROICOR)

# Workflow PhysIO Toolbox

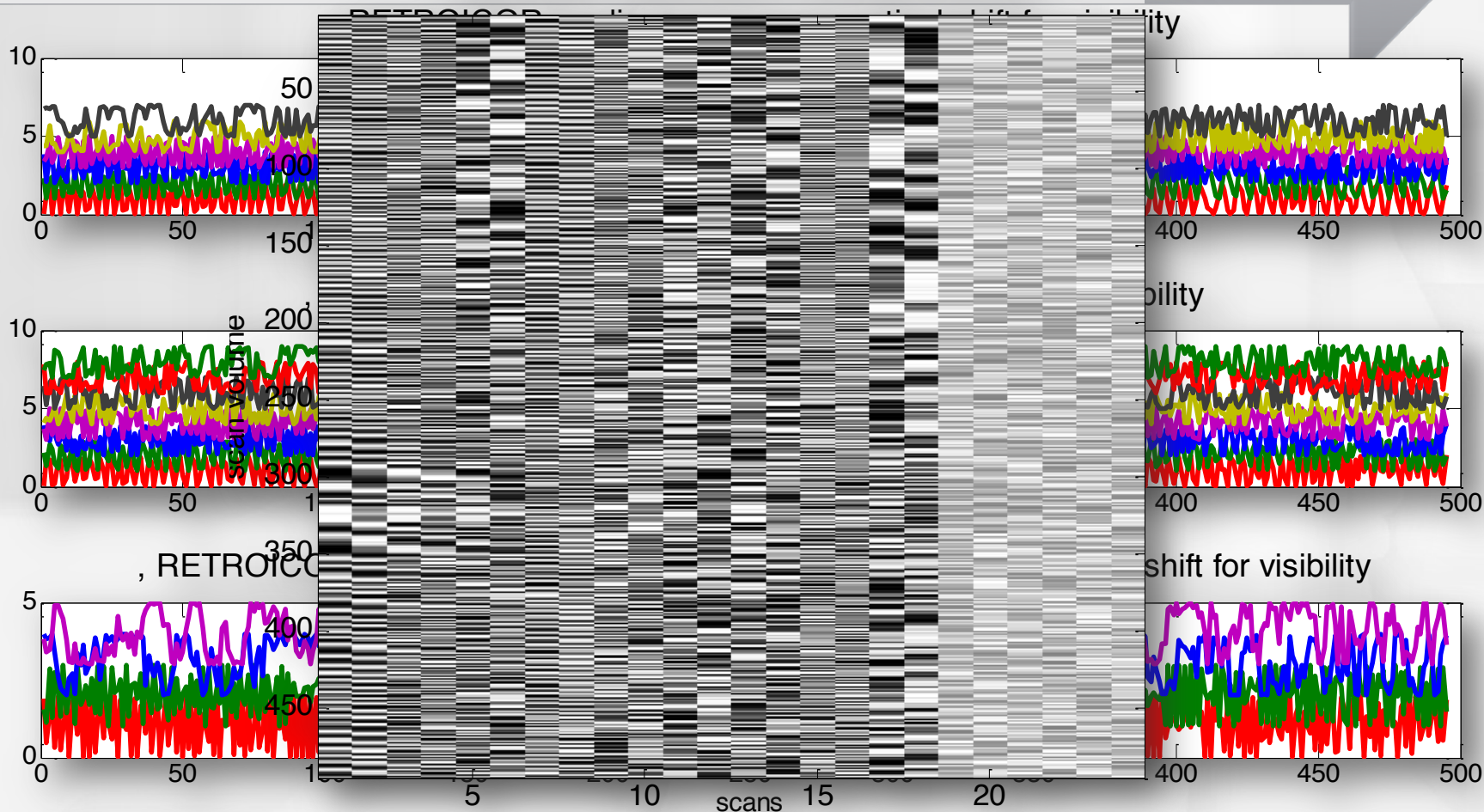


Read logfiles

Preprocess physiological data  
physiological regressors matrix for GLM  
- specified regressors orthogonalized -

Model time series physiological noise

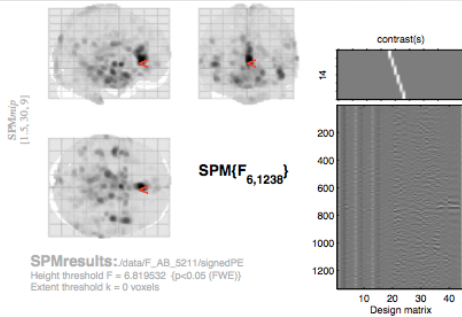
Include confound regressors (GLM)



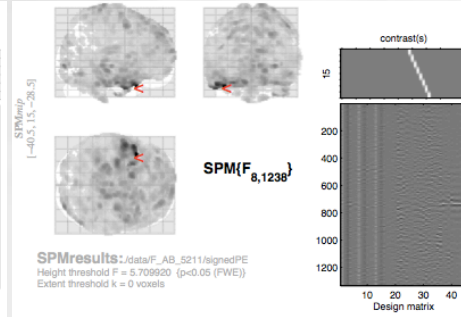
# Model Check: SPM F-contrasts



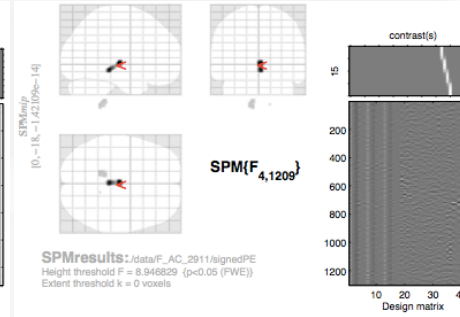
## Cardiac regressors



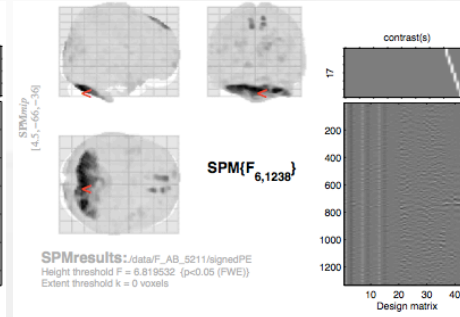
## Respiratory regressors



## Cardiac x Respiratory



## Movement regressors



Finally:

Check Influence of Physiological Noise (Correction) on Data

- SPM
- F-contrast on 1st and second level

## Read logfiles

- Of peripheral physiological data
- Vendor-specific

## Preprocess physiological data

- Filter noisy ECG & detect cardiac pulses
- Hand-pick missing pulses

## Model time series physiological noise

- RETROICOR
- Respiratory Volume
- Heart Rate

## Include confound regressors (GLM)

- Multiple\_regressors file for SPM

Finally:

## Check Influence of Physiological Noise (Correction) on Data

- SPM
- F-contrast on 1st and second level



# PhysIO: SPM Batch Interface



**A**

**B**

Module List

- Realign: Estimate & Res
- TAPAS PhysIO Toolbox**
- fMRI model specification

Current Module: TAPAS PhysIO Toolbox

Help on: TAPAS PhysIO Toolbox

```

save_dir
log_files
  .vendor Philips
  .log_cardiac <-X
  .log_respiration <-X
  .log_scan_timing <-X
  .sampling_interval []
  .relative_start_acquisition 0
sqpar (Sequence timing parameters)
  .Nsllices <-X
  .NsllicesPerBeat []
  .TR <-X
  .Ndummies <-X
  .Nscans <-X
  .onset_slice <-X
  .time_slice_to_slice []
  .Nprep []
thresh (Thresholding parameters for de-noising and timing)
  .Scan/Physlog Time Synchronization
  .nominal
cardiac
  .modality ECG
  .Initial Detection of Heartbeats
  .load_from_logfile
  .Post-hoc Selection of Cardiac Pulses
  .Off
model
  .type ... (RETRO)
  .order
  .cardiac 3
  .respiratory 4
  .cardiac X respiratory 1
  .orthogonalise none
  .input_other_multiple_regressors
  .output_multiple_regressors ...essors.txt
verbose
  .level 2
  .fig_output_file ""
  .use_tabs false
    
```

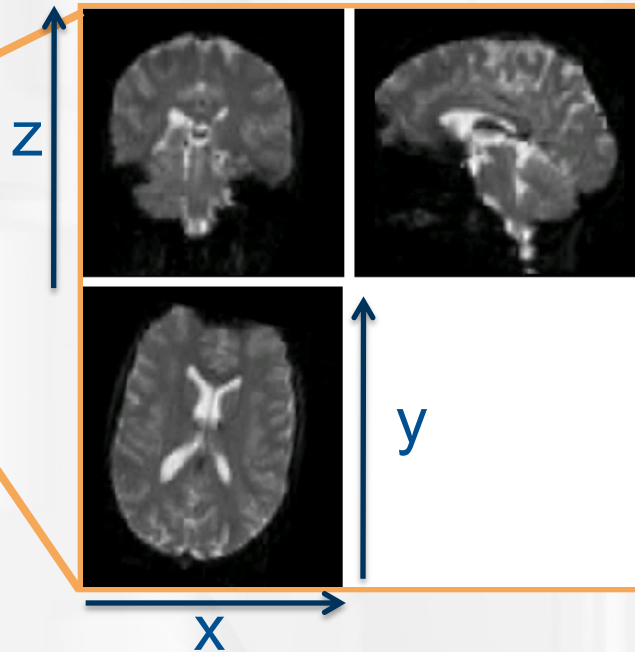
`tapas_physio_report_contrasts()`





- **Birn**, Rasmus M., Jason B. Diamond, Monica A. Smith, and Peter A. Bandettini. 2006. “Separating Respiratory-variation-related Fluctuations from Neuronal-activity-related Fluctuations in fMRI.” *NeuroImage* 31 (4) (July 15): 1536–1548. doi:10.1016/j.neuroimage.2006.02.048.
- **Glover**, G H, T Q Li, and D Ress. 2000. “Image-based Method for Retrospective Correction of Physiological Motion Effects in fMRI: RETROICOR.” *Magnetic Resonance in Medicine: Official Journal of the Society of Magnetic Resonance in Medicine / Society of Magnetic Resonance in Medicine* 44 (1) (July): 162–7.
- **Harvey**, Ann K., Kyle T.S. Pattinson, Jonathan C.W. Brooks, Stephen D. Mayhew, Mark Jenkinson, and Richard G. Wise. 2008. “Brainstem Functional Magnetic Resonance Imaging: Disentangling Signal from Physiological Noise.” *Journal of Magnetic Resonance Imaging* 28 (6): 1337–1344. doi:10.1002/jmri.21623.
- **Hutton**, C., O. Josephs, J. Stadler, E. Featherstone, A. Reid, O. Speck, J. Bernarding, and N. Weiskopf. 2011. “The Impact of Physiological Noise Correction on fMRI at 7 T.” *NeuroImage* 57 (1) (July 1): 101–112. doi:10.1016/j.neuroimage.2011.04.018.
- **Josephs**, O., Howseman, A.M., Friston, K., Turner, R., 1997. “Physiological noise modelling for multi-slice EPI fMRI using SPM.” *Proceedings of the 5th Annual Meeting of ISMRM, Vancouver, Canada*, p. 1682
- **Kasper**, Lars, Sarah Marti, S. Johanna Vannesjö, Chloe Hutton, Ray Dolan, Nikolaus Weiskopf, Klaas Enno Stephan, and Klaas Paul Prüssmann. 2009. “Cardiac Artefact Correction for Human Brainstem fMRI at 7 Tesla.” In *Proc. Org. Hum. Brain Mapping* 15, 395. San Francisco.

# fMRI = Acquiring Movies



- ...of three-dimensional Blood Oxygen-Level Dependent (BOLD) contrast images

- Run/Session: Time Series of Images

