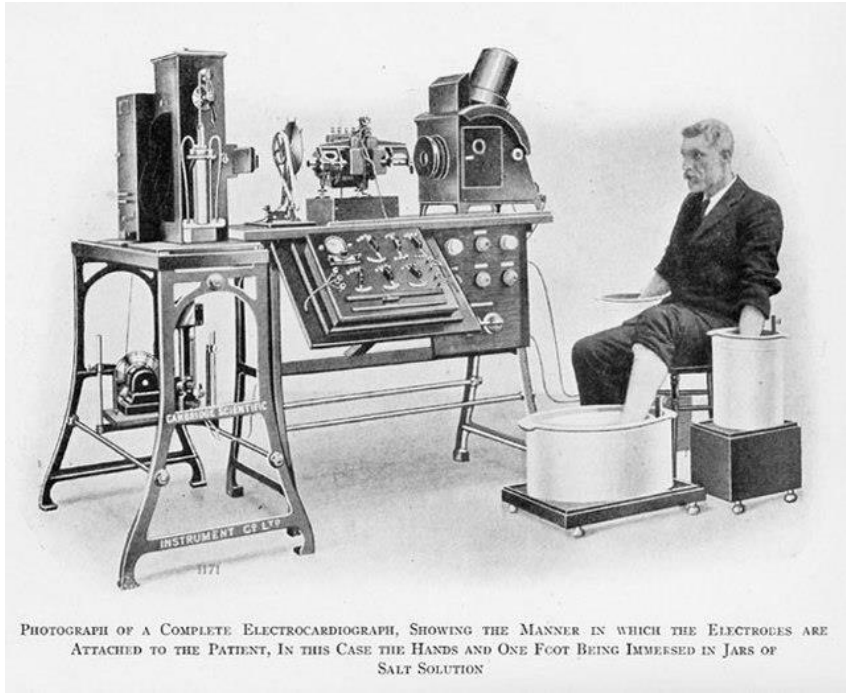


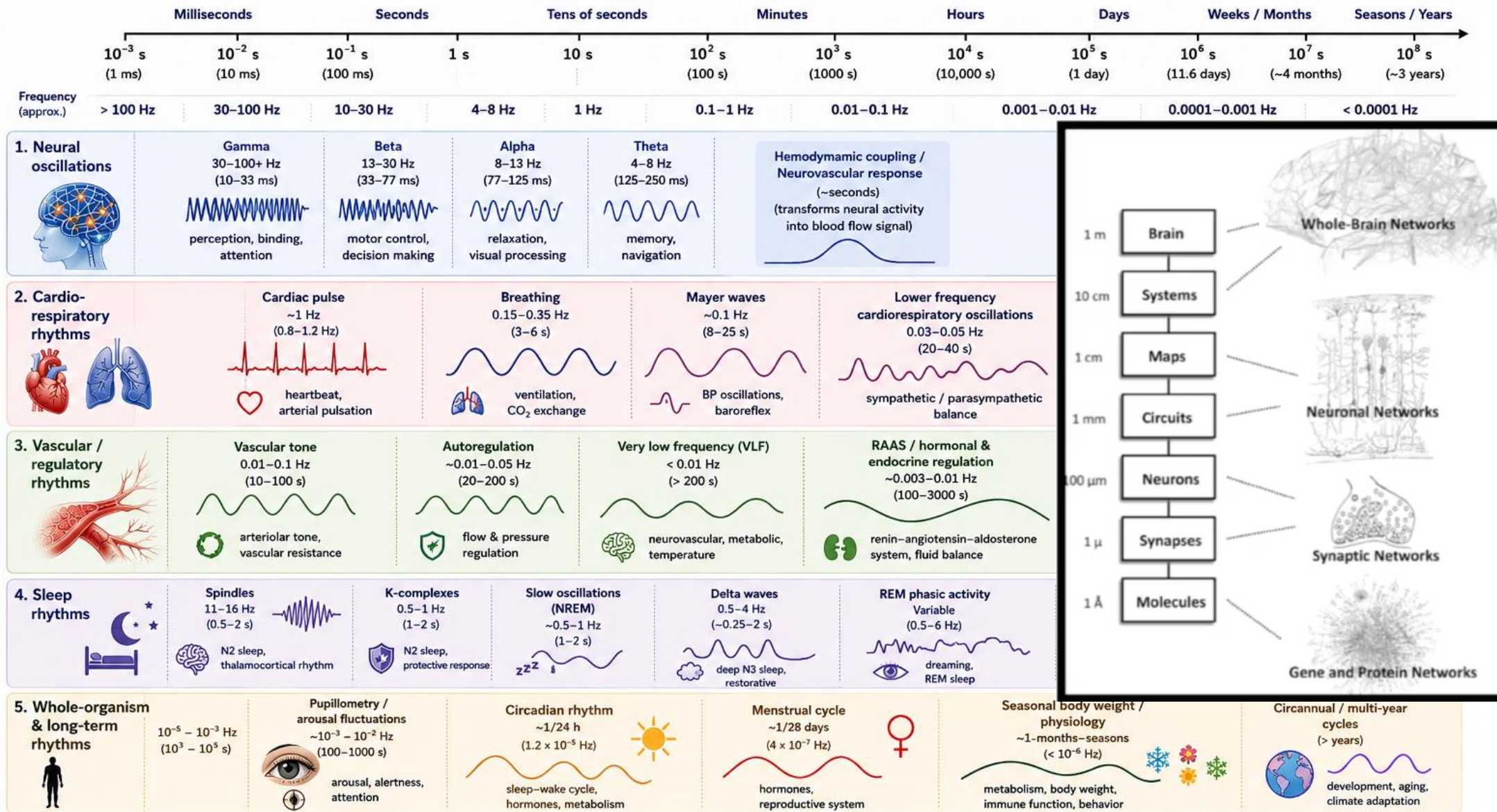


Physiology in time series signal

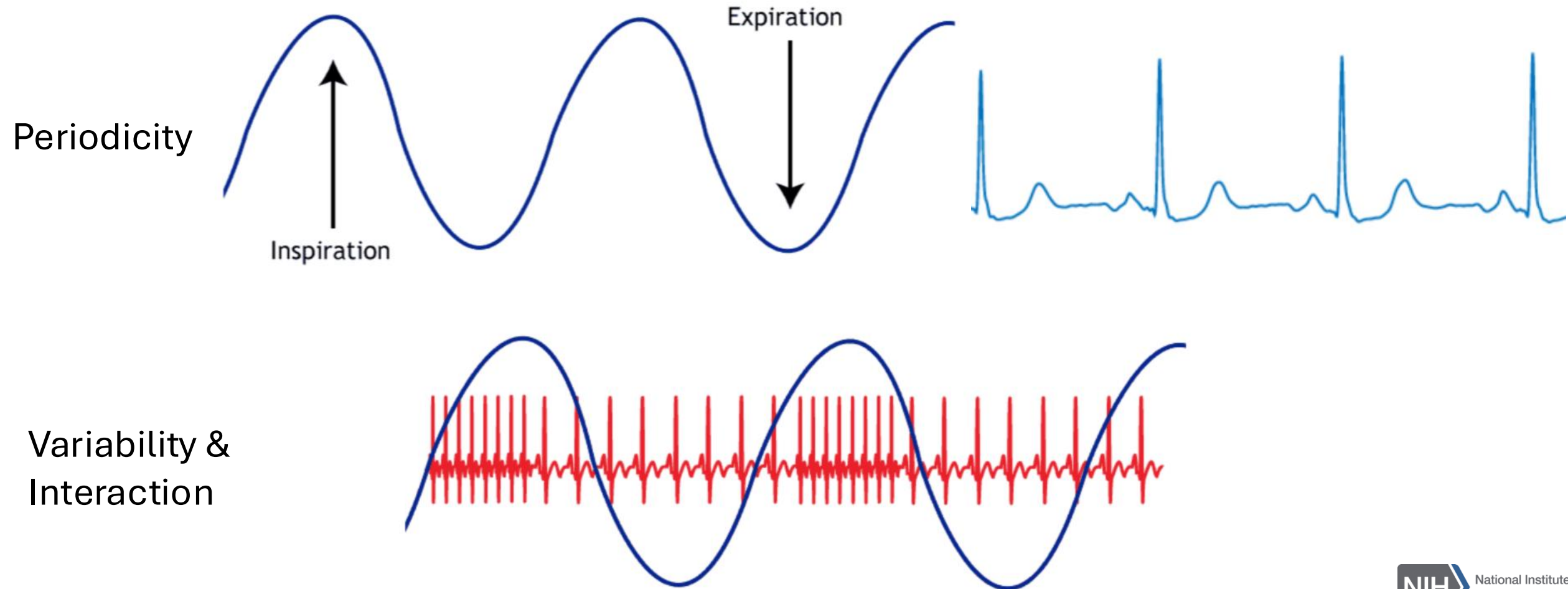
08/11/26 Burak Akin



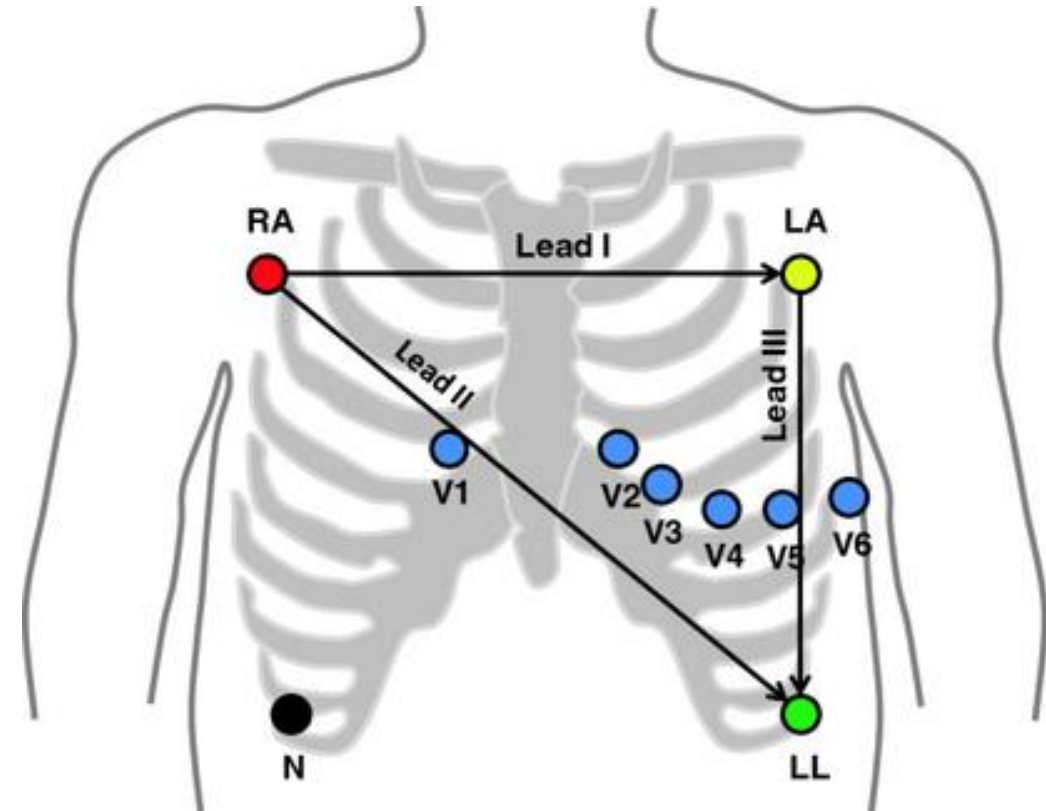
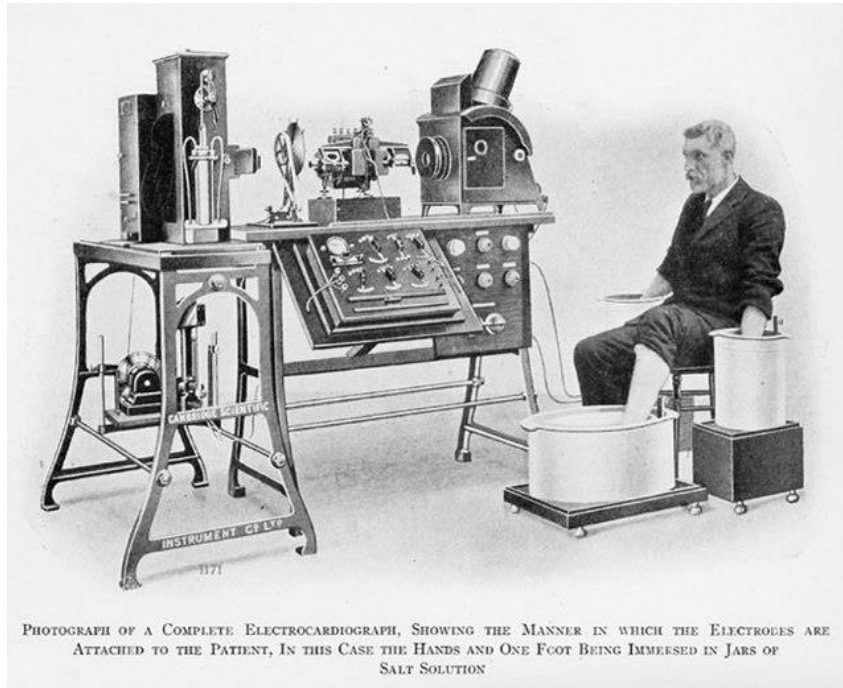
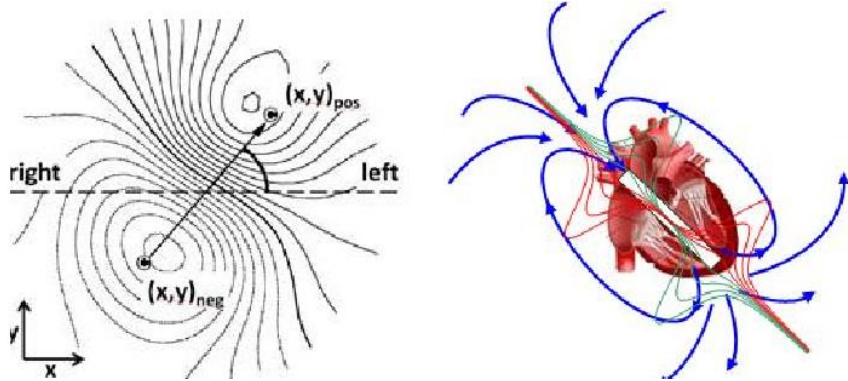
Biological rhythms



What are the characteristics of physiological signals?



How do we measure physiological signals? Cardiac activity

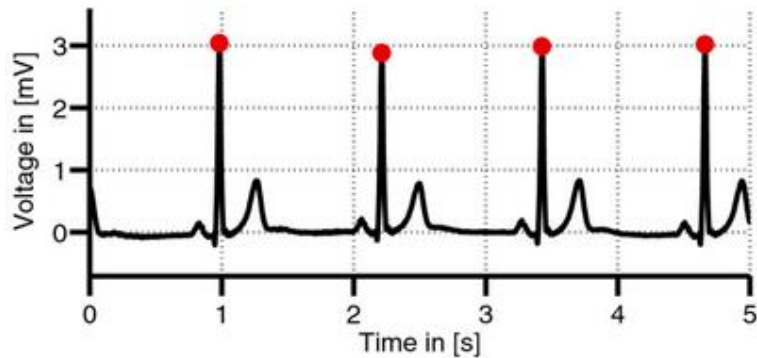


How do we measure physiological signals?

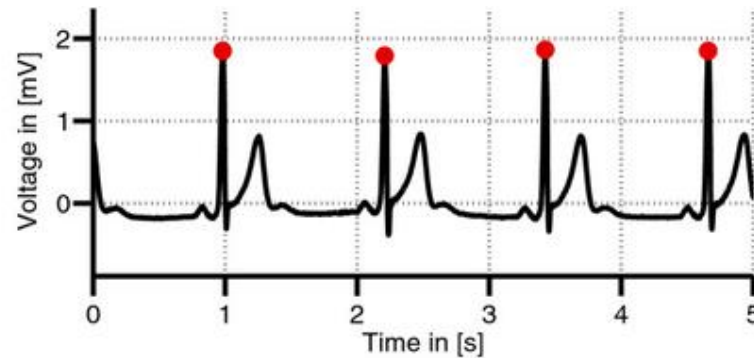


How do we measure physiological signals?

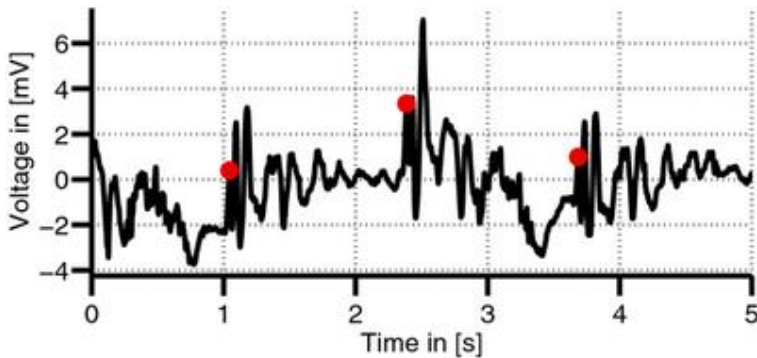
- ECG inside MRI
- The magneto-hydrodynamic effect : the motion of the blood



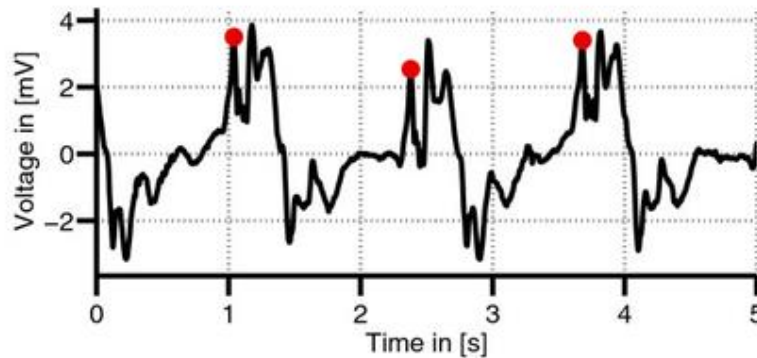
(a) Lead II, outside



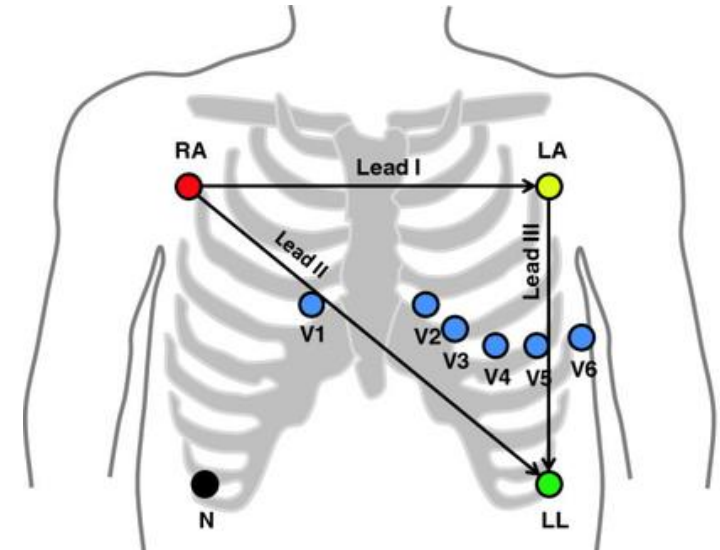
(b) Lead V3, outside



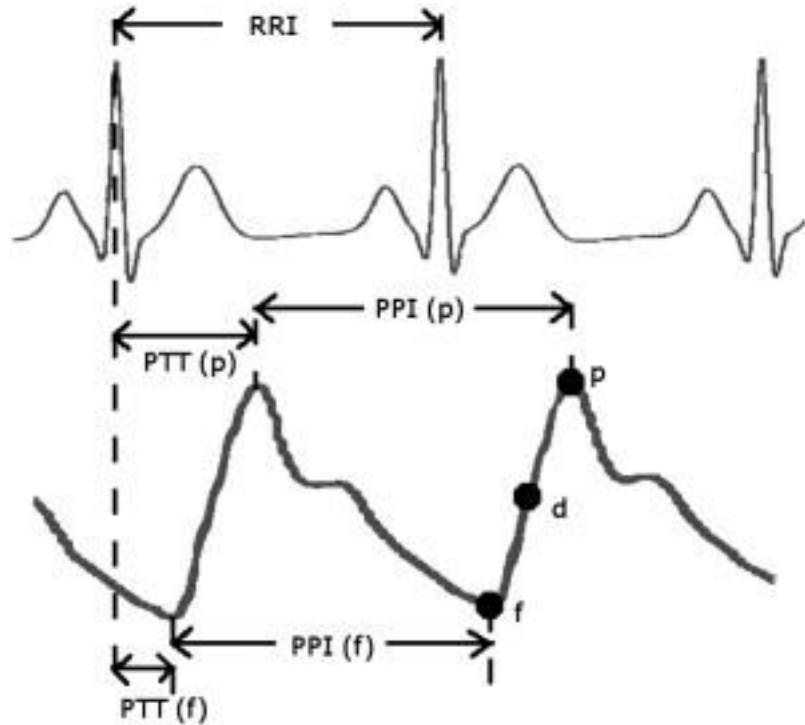
(c) Lead II, inside, Ff



(d) Lead V3, inside, Ff

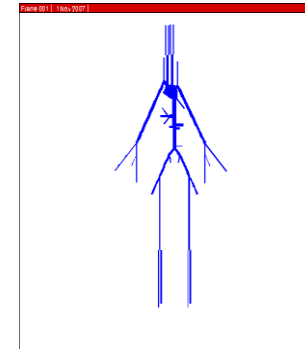


Pulse Oximeter (optical measure)

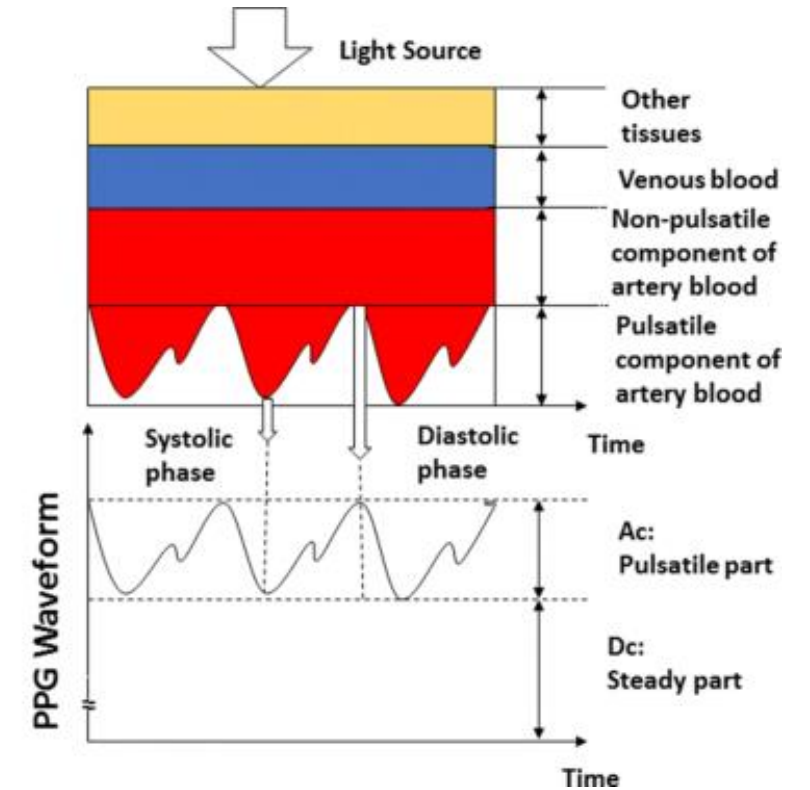


Schäfer et, al.2012

How accurate is pulse rate variability as an estimate of heart rate variability?



<http://haemod.uk/nektar>



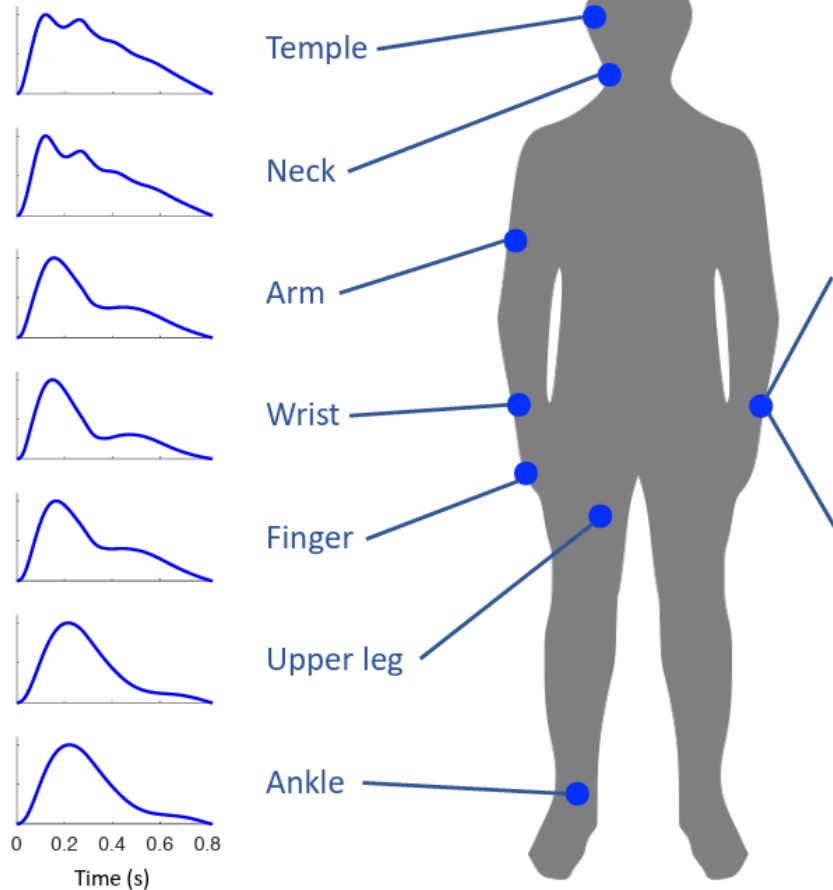
Pulse Oximeter (optical)

Photoplethysmography Signal Processing and Synthesis

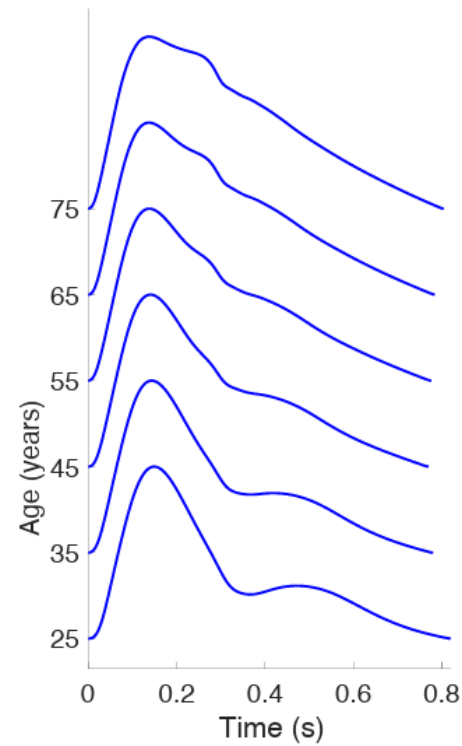
Elisa Mejía-Mejía, John Allen, Karthik Budidha, Chadi El-Hajj, Panayiotis A. Kyriacou and Peter H. Charlton*



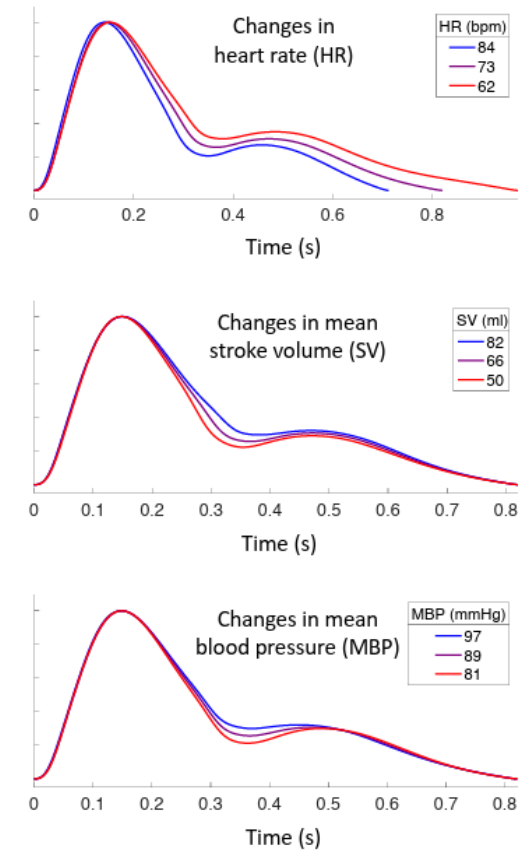
PPG pulse waves at different locations



PPG pulse waves at different ages

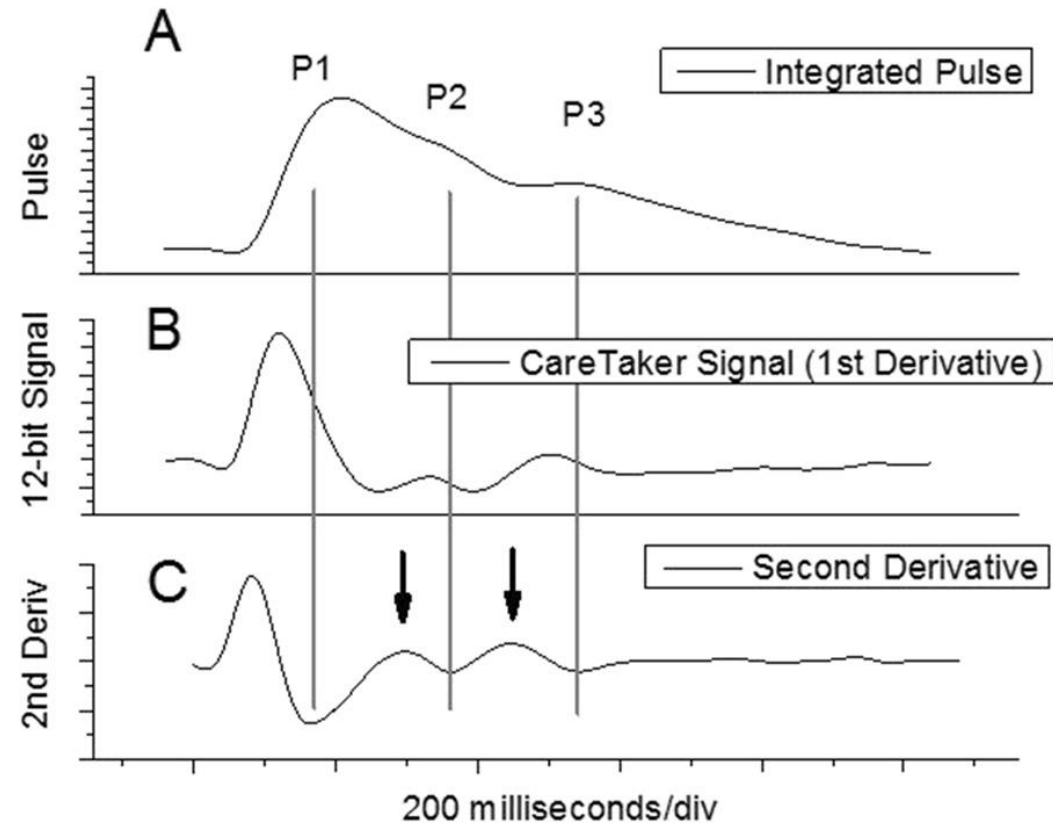


PPG pulse waves under different cardiovascular conditions

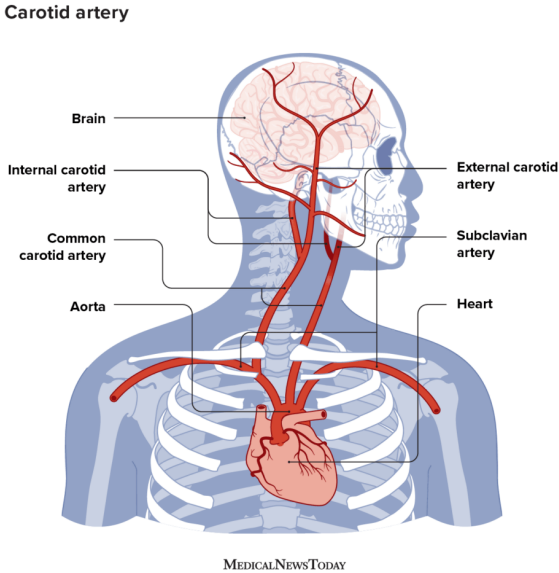
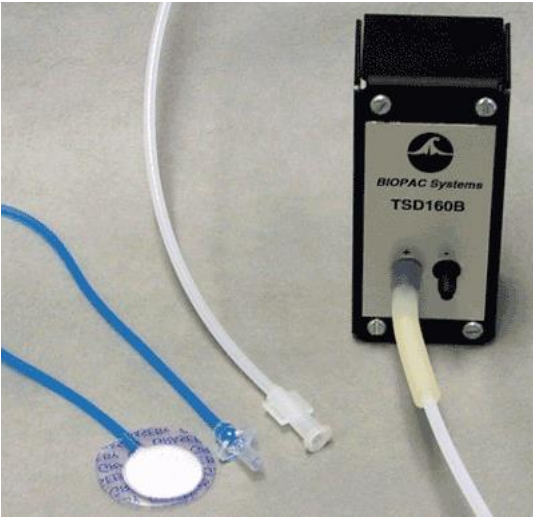


Measuring Blood Pressure (pneumatic)

”Continuous” Non-invasive blood pressure monitor(NIBP)

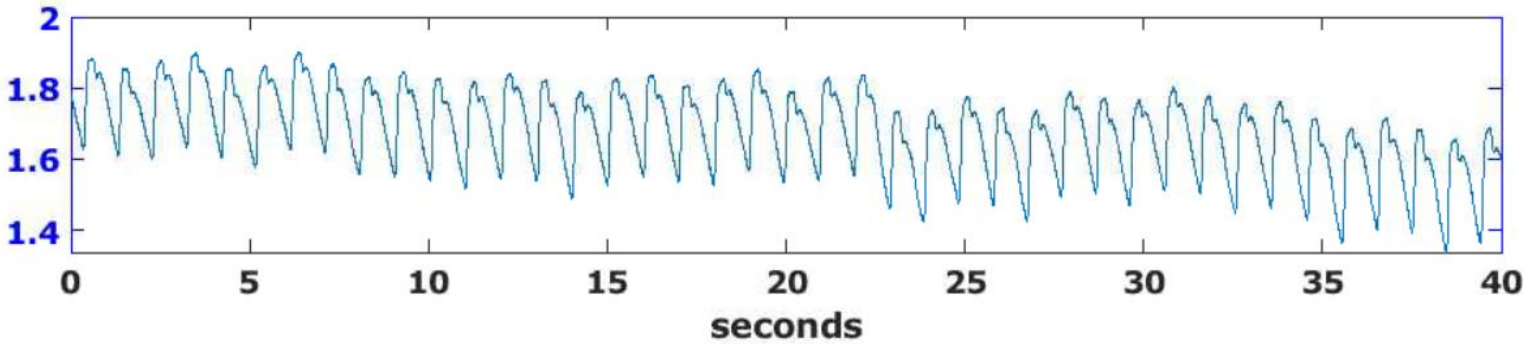


MRI compatible – Pressure pad (pneumatic)



Small pressure pad for surface pulsations

placed on carotid artery



Respiration (pneumatic-piezo)

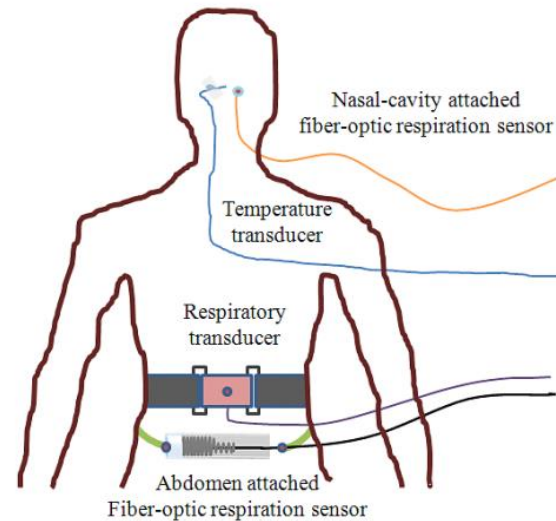
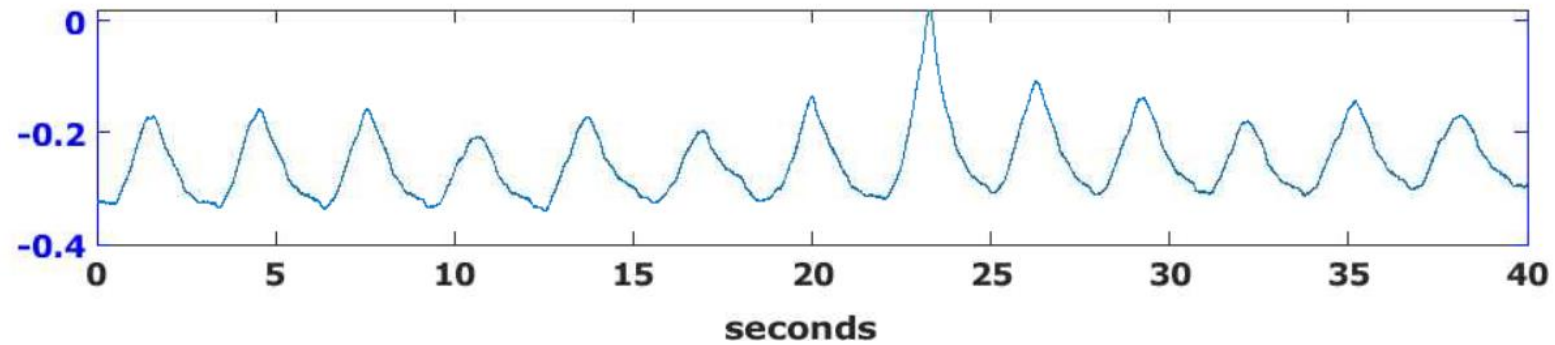


FIG. 4. Experimental setup using fiber-optic respiration sensors and a BIOPAC® system.

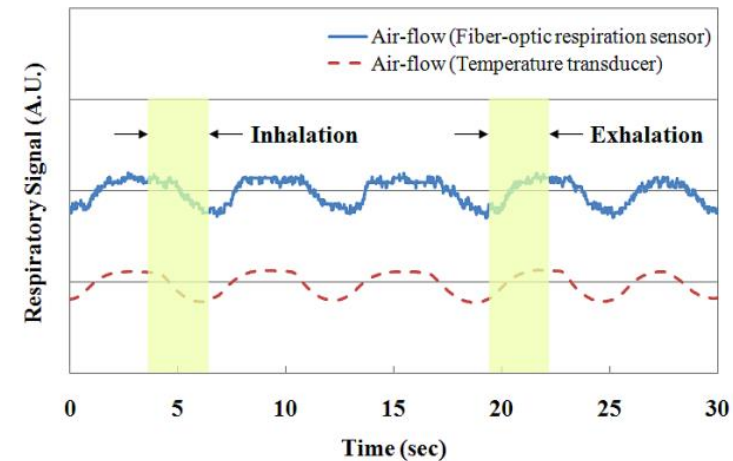


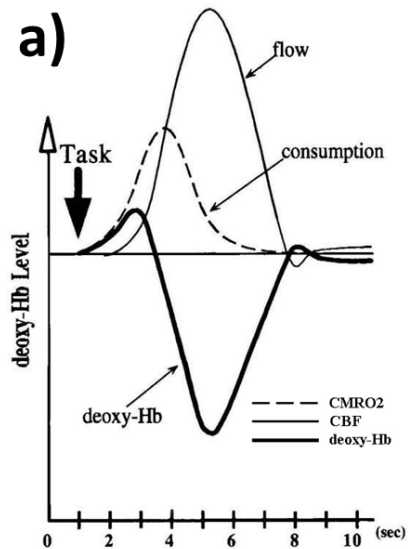
FIG. 5. Comparison of the respiratory signal of a nasal-cavity attached fiber-optic respiration sensor and a temperature transducer.

- Diaphragmatic
- Shallow breathing

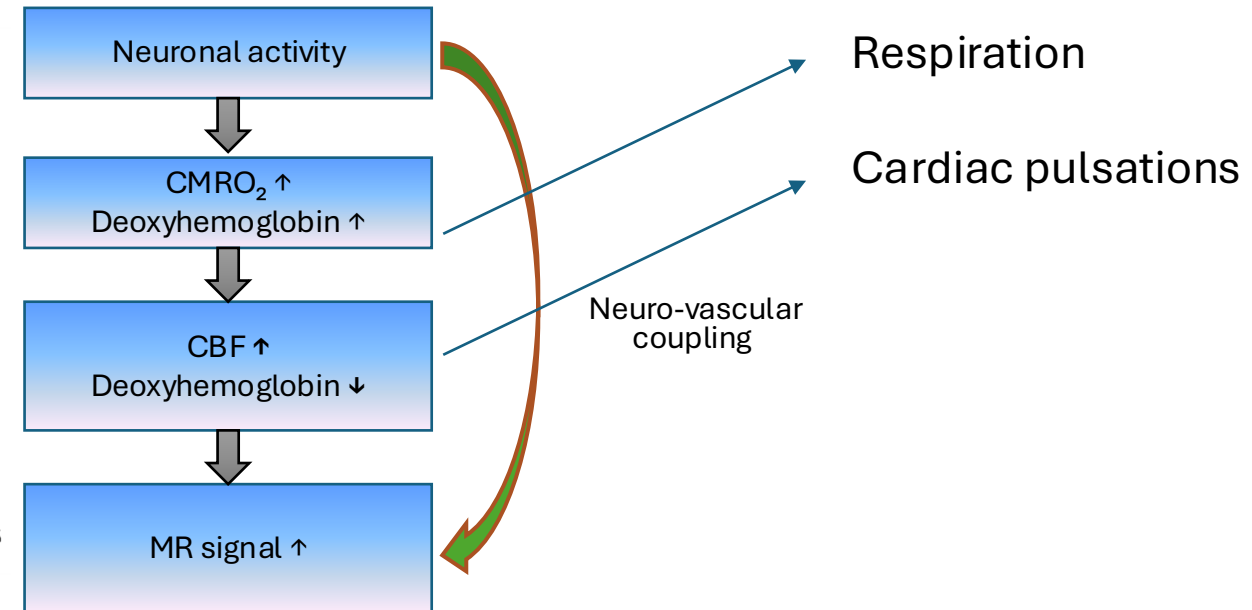
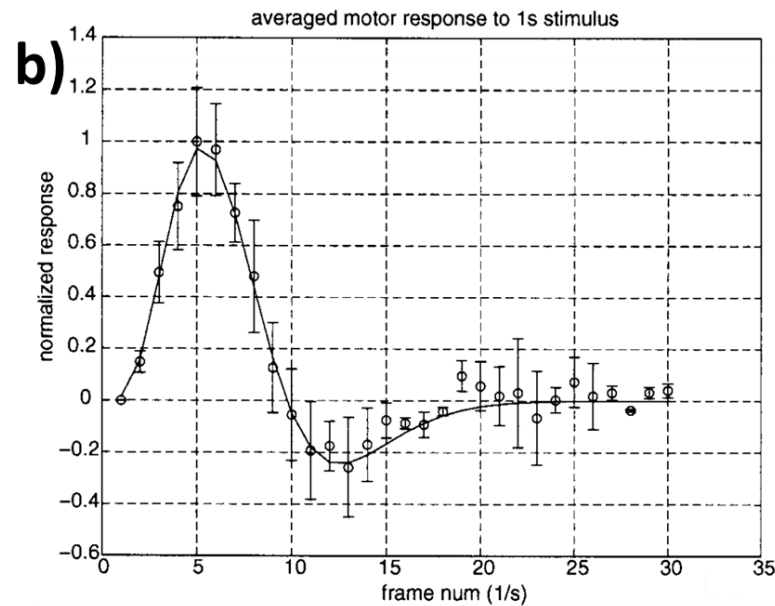


Functional MRI

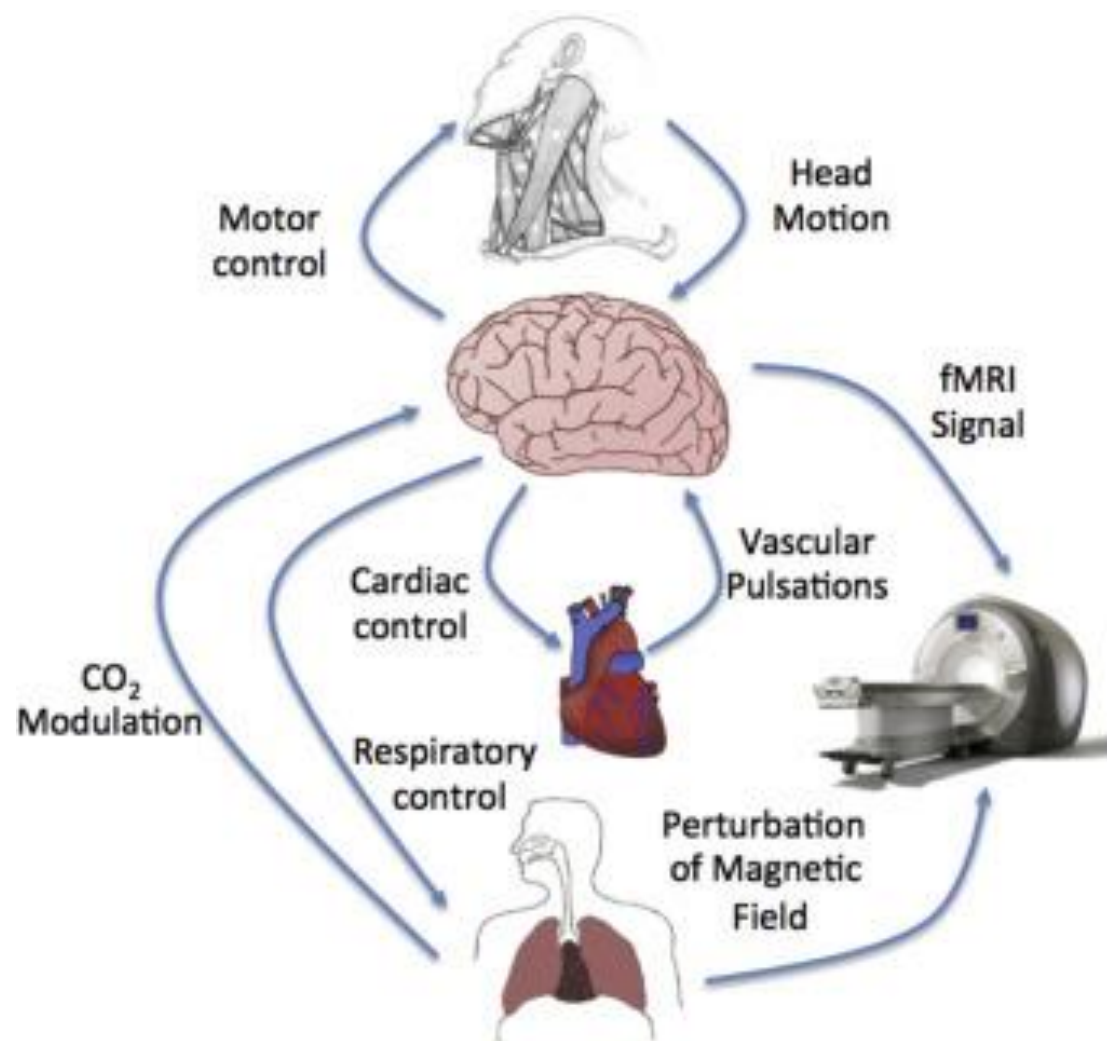
- T2*-weighted imaging sensitive to paramagnetic deoxyhaemoglobin concentration in the blood
- Blood-oxygenation level dependent (BOLD) effect



Glover et.al 1999

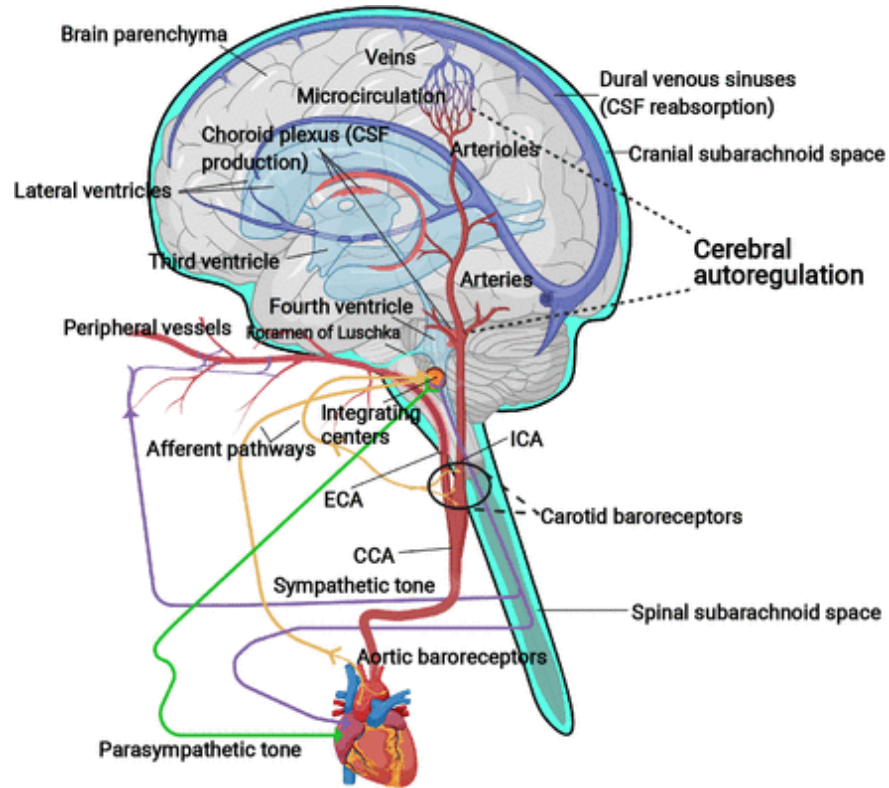


What modulates time series in fMRI

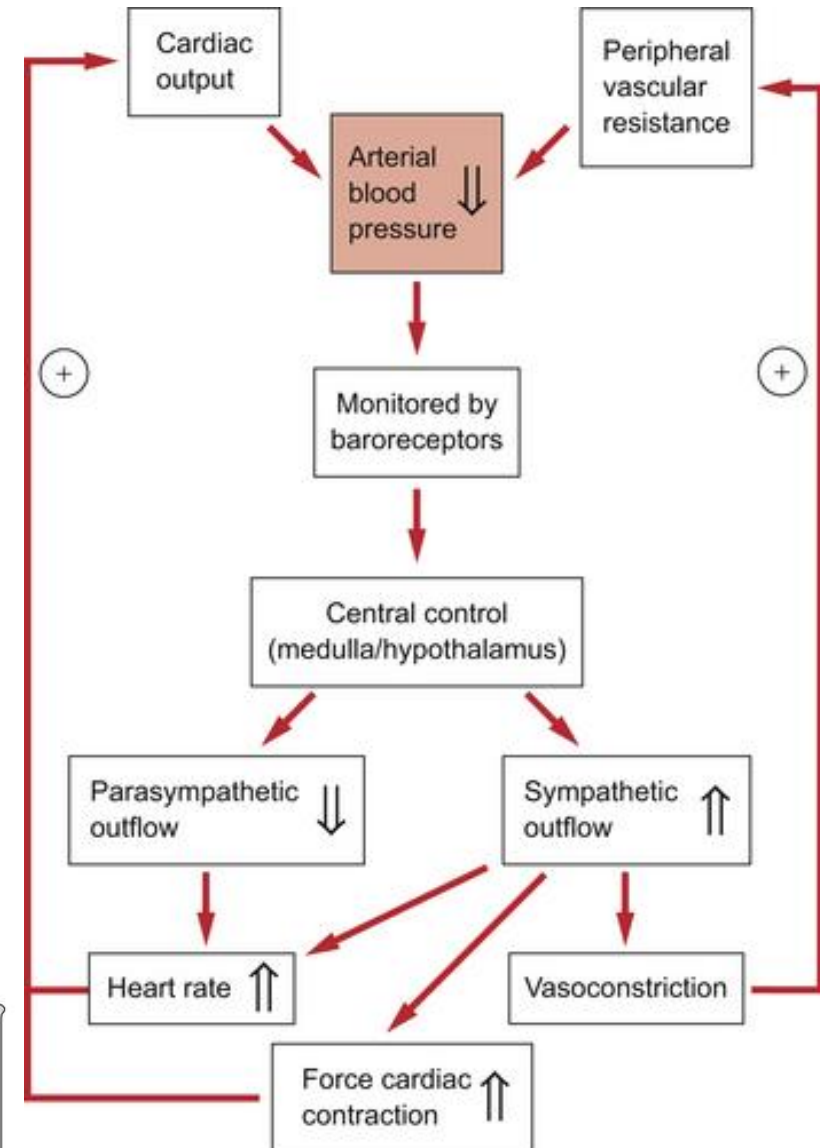
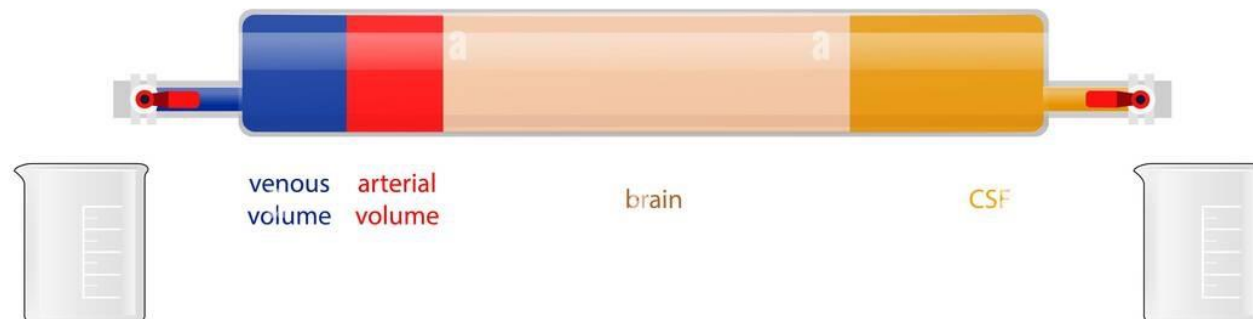


- Noise
 - Hardware related
 - Physiological
 - Motion
- Hardware related
 - Scanner drift
 - Thermal noise
- Motion
 - Respiration related
 - Cardiac related
 - Task related
 - Other head motion
- Physiological
 - Respiration
 - Time-locked
 - B₀-field
 - CO₂
 - Blood flow
 - Cardiac
 - Time-locked
 - Motion-related
 - Blood flow

What modulates time series in fMRI



Garcia et, al.2021

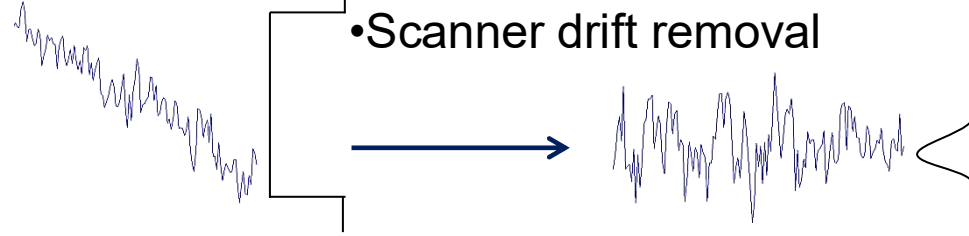


Functional MRI

Preprocessing

- fMRI images are not ready to do statistics on them.
- They need to be “conditioned”

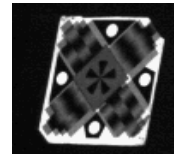
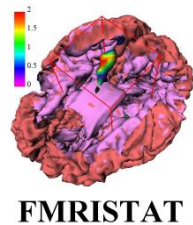
- Baseline wander correction
- Trend removal
- Scanner drift removal



- Physiological noise correction
- Skull stripping
- Spectral filtering
- Smoothing

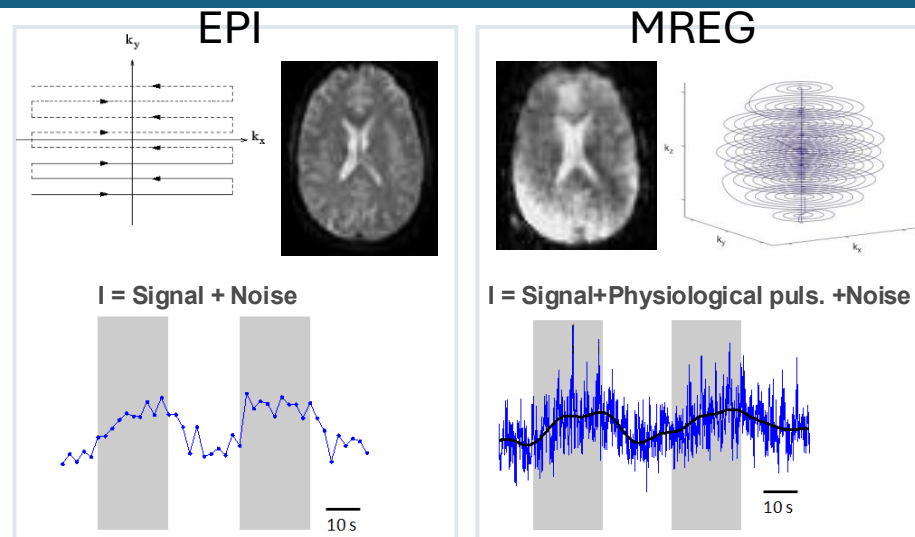


- Image alignment
- Registration
- Motion Correction

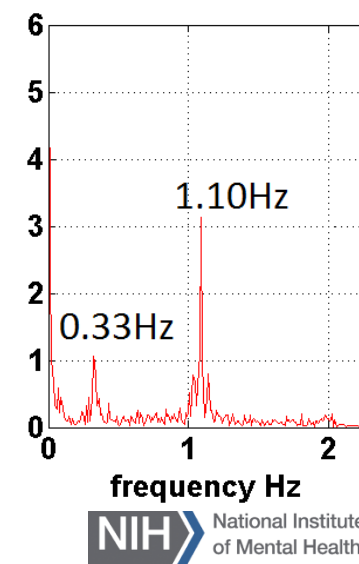
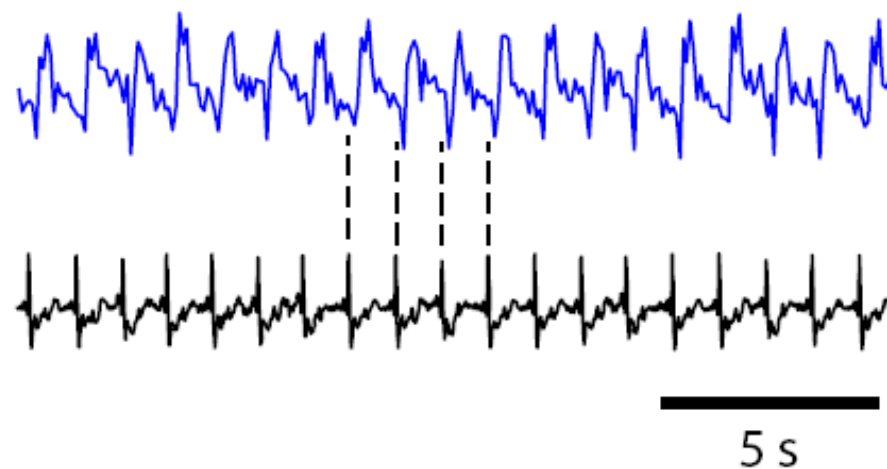
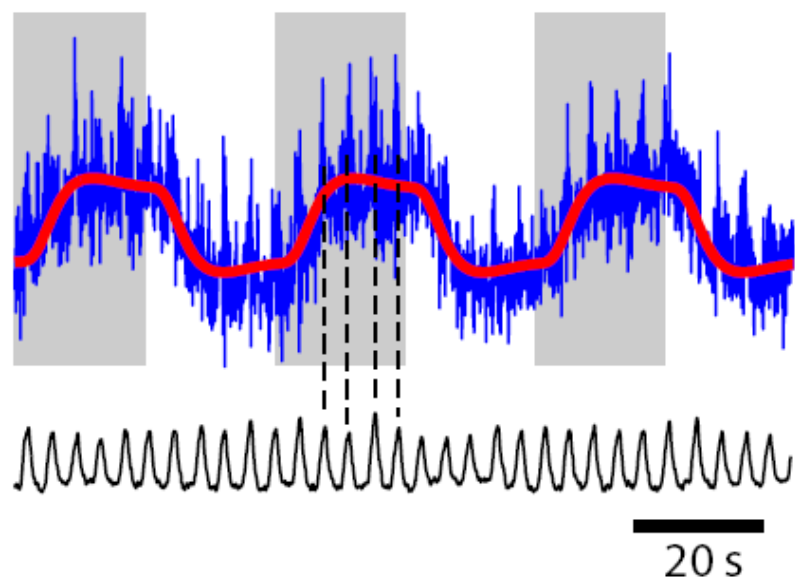


Improving temporal resolution

Fast fMRI

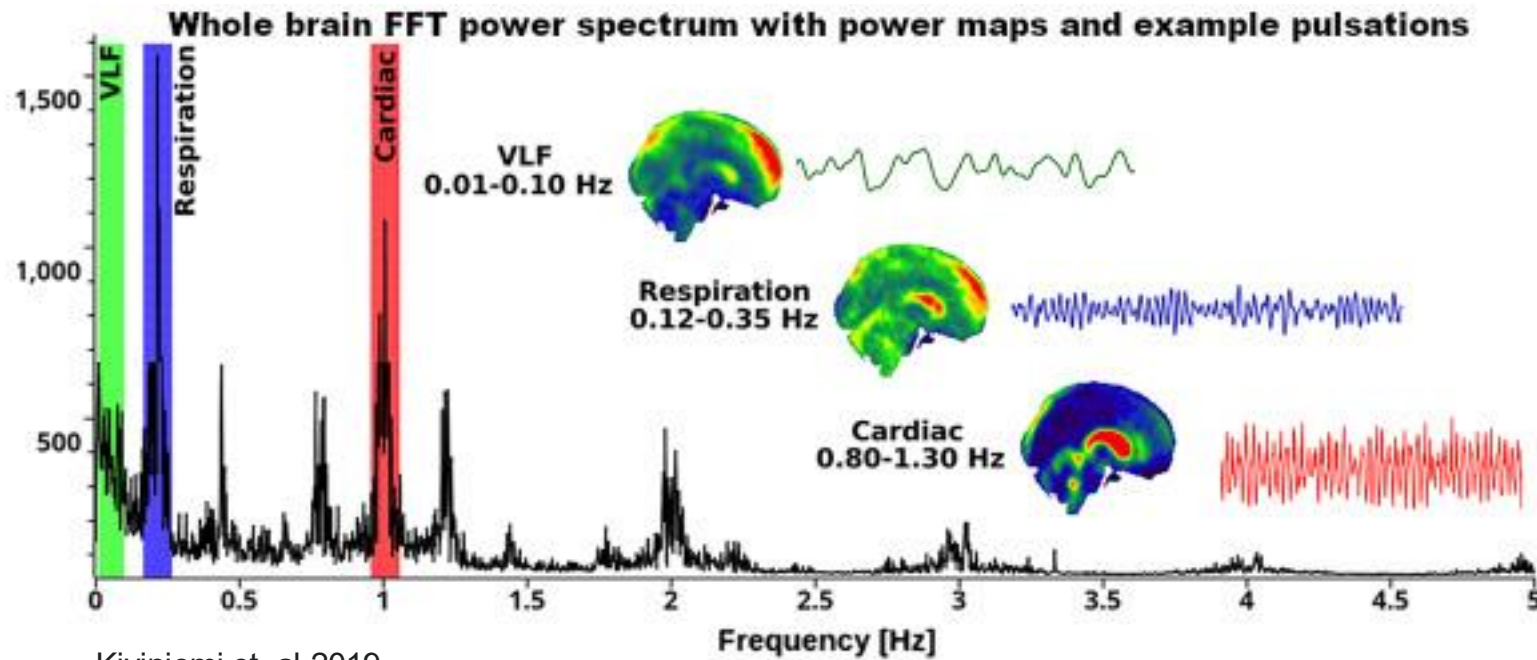


- Readout time: $\sim 65\text{ms}$
- 3D acquisition in $\sim 100\text{ ms}$ (in contrast to 2-3 seconds in conventional EPI)
- Iterative image reconstruction with off-resonance correction
- Result: whole-brain data with some blurring, effective spatial resolution $\approx 4\text{ mm}$)

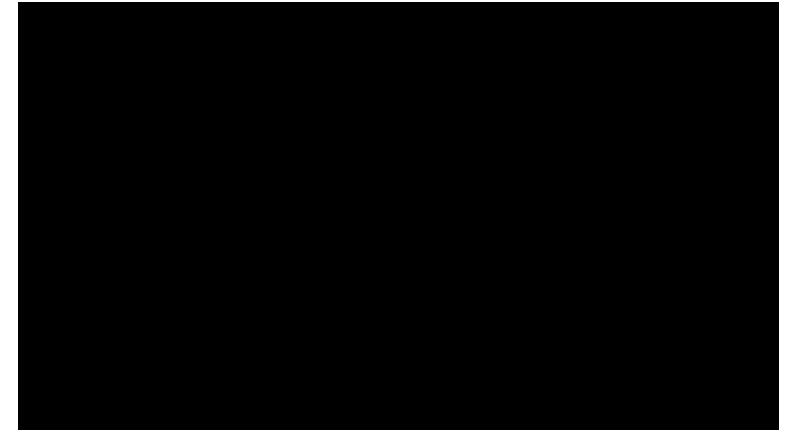


Improving temporal resolution

Fast fMRI

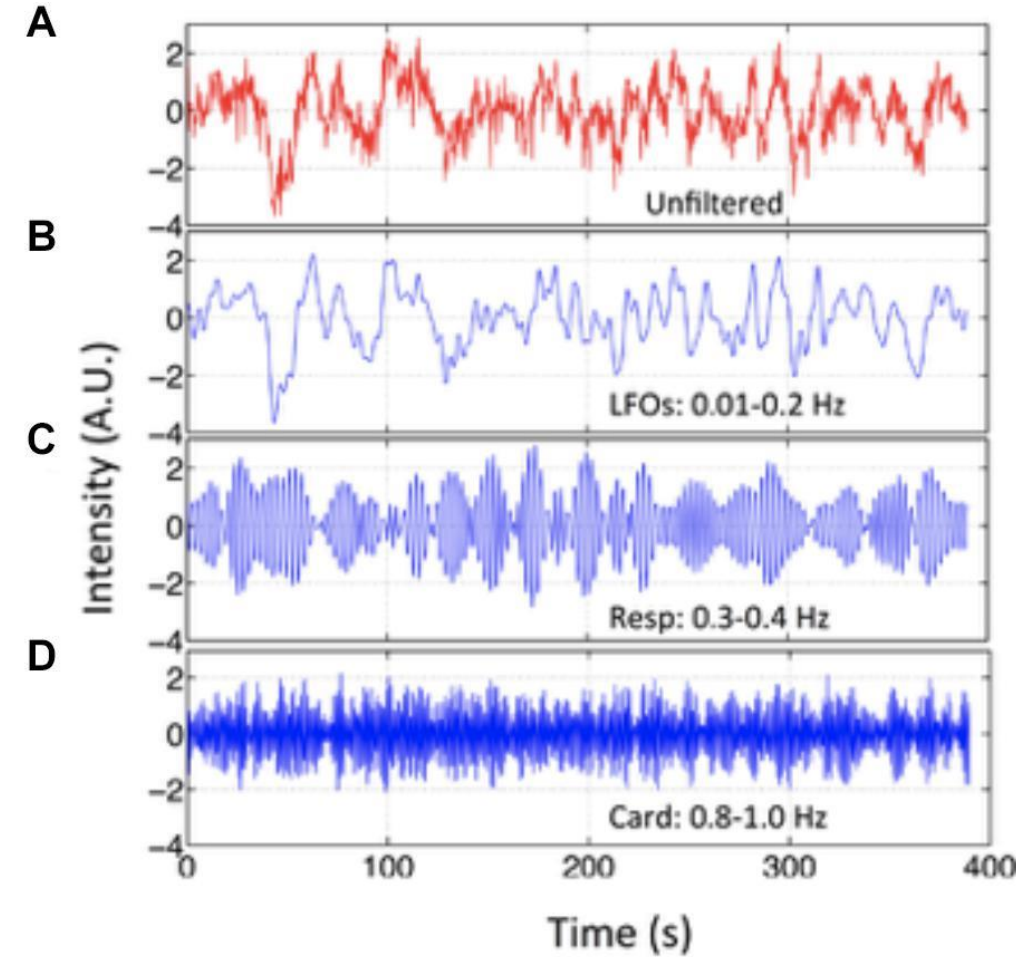
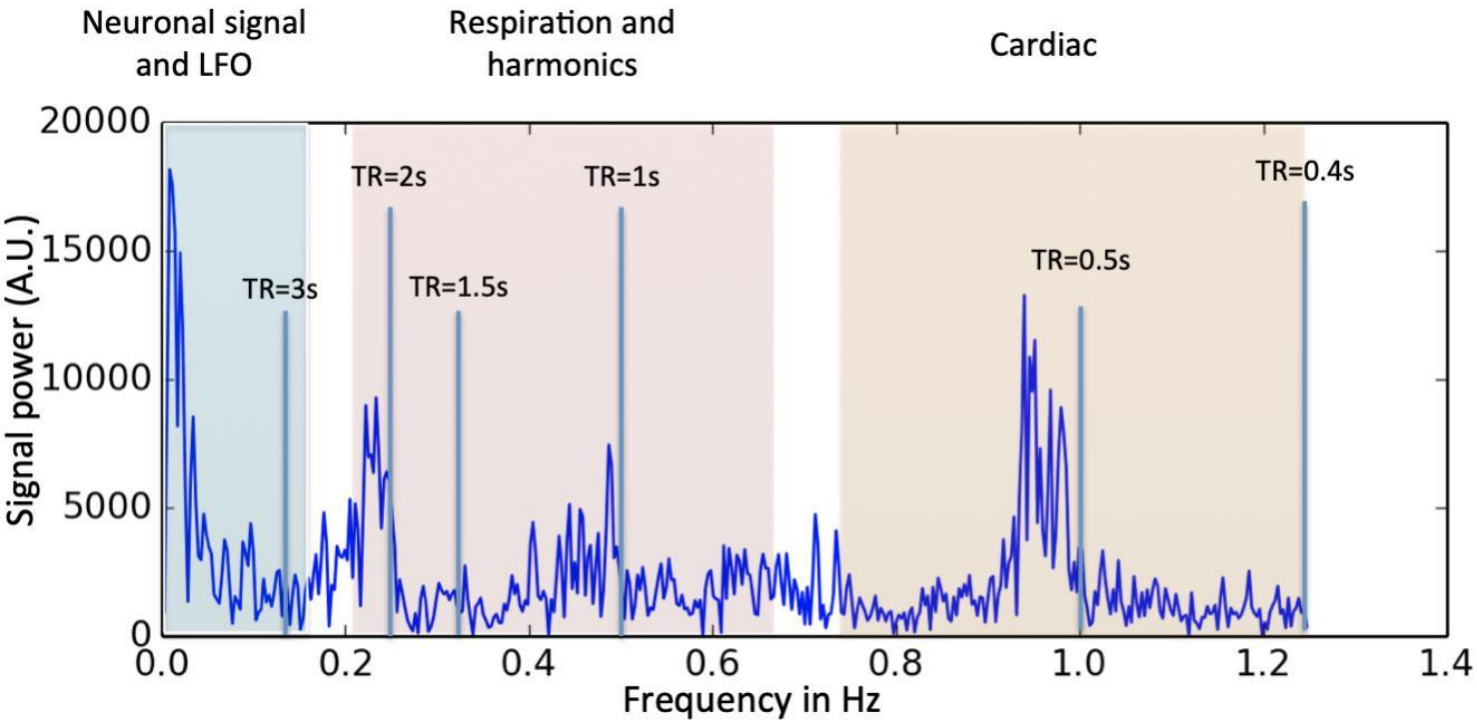


Kiviniemi et, al.2019



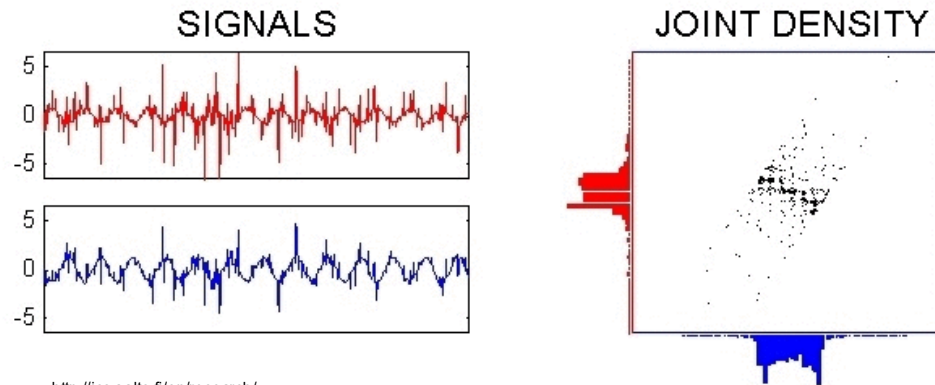
Improving temporal resolution

Fast fMRI

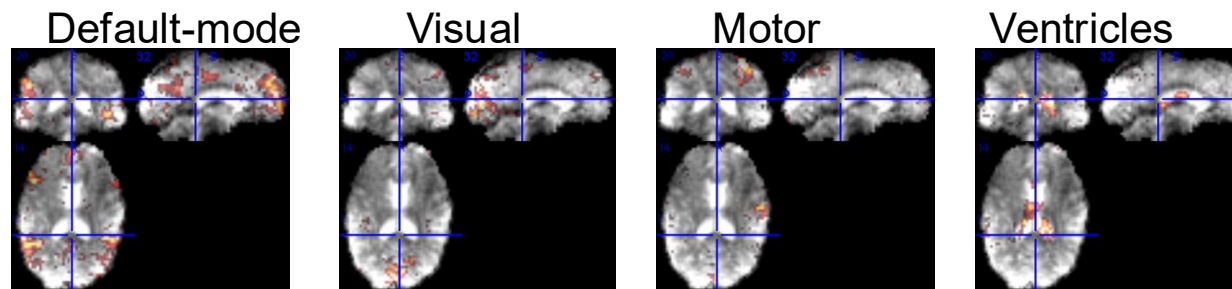
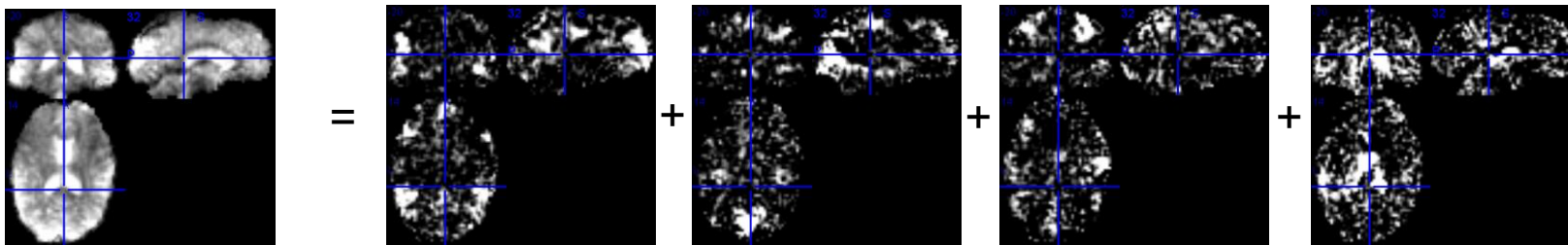


Independent Component Analysis

- ICA is the mathematical approach that performs this so-called “blind source separation”
- In fMRI: at each time point, we measure a mixture of unknown functional networks (as well as artifacts)
- Networks fluctuate and are thus mixed differently at each time point

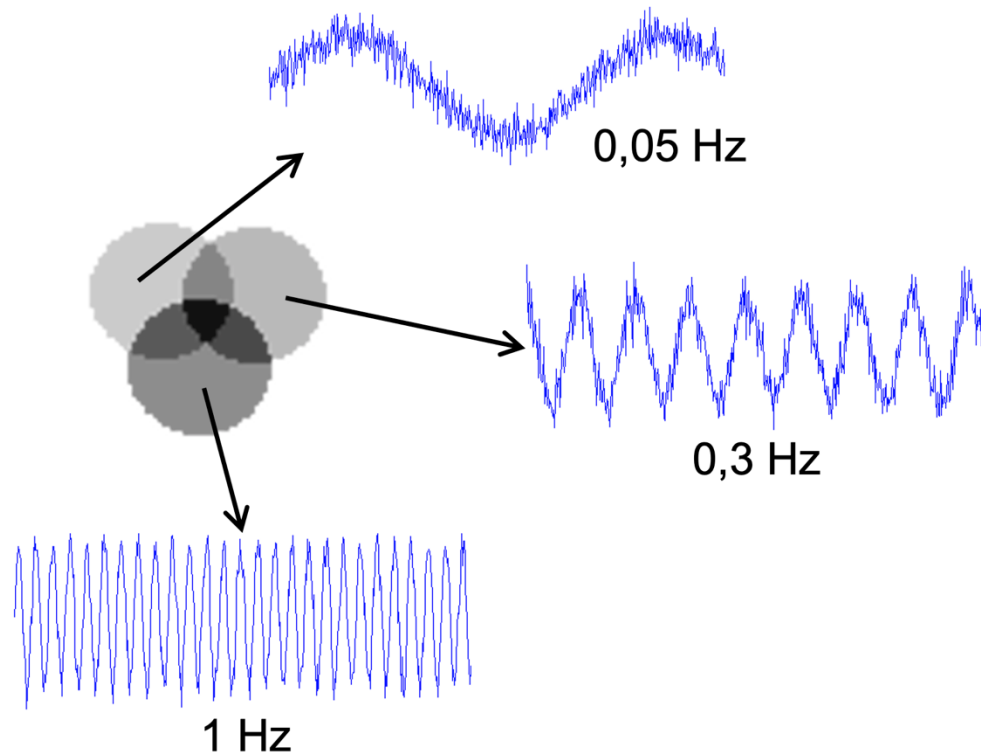


<http://ics.aalto.fi/en/research/>

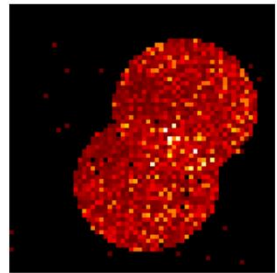
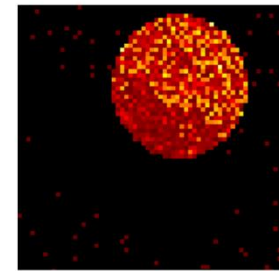
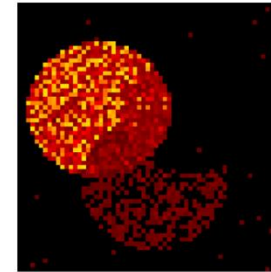


Improving temporal resolution

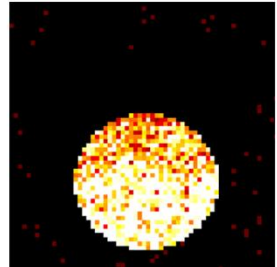
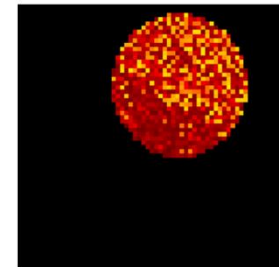
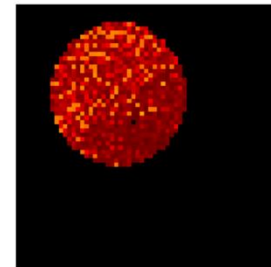
Fast fMRI



2 sec

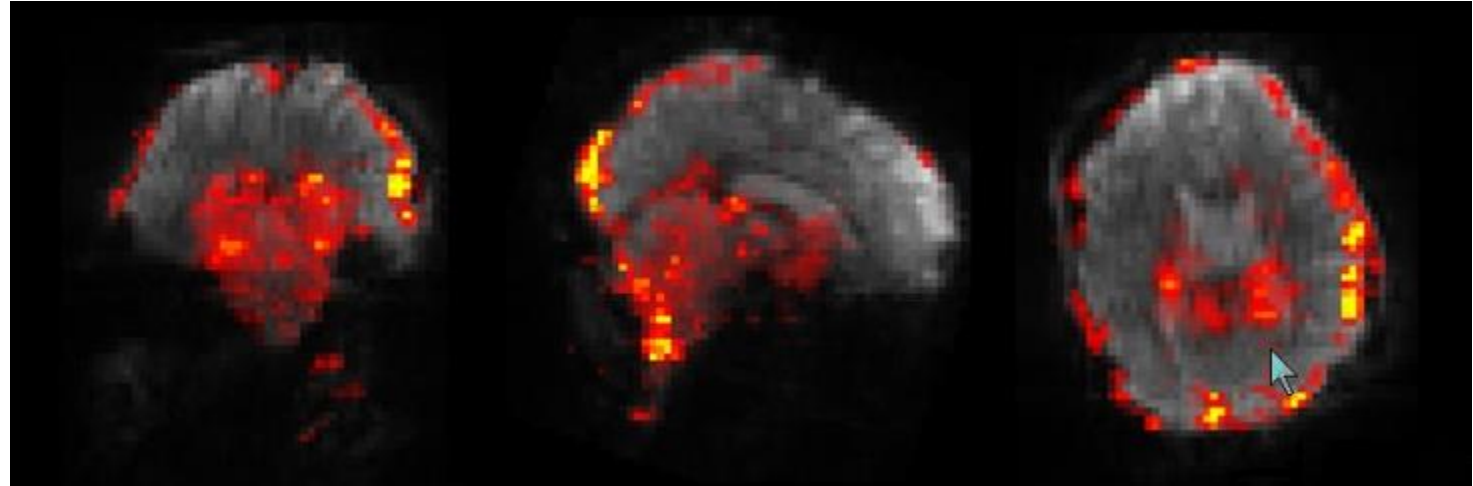


0.1 sec

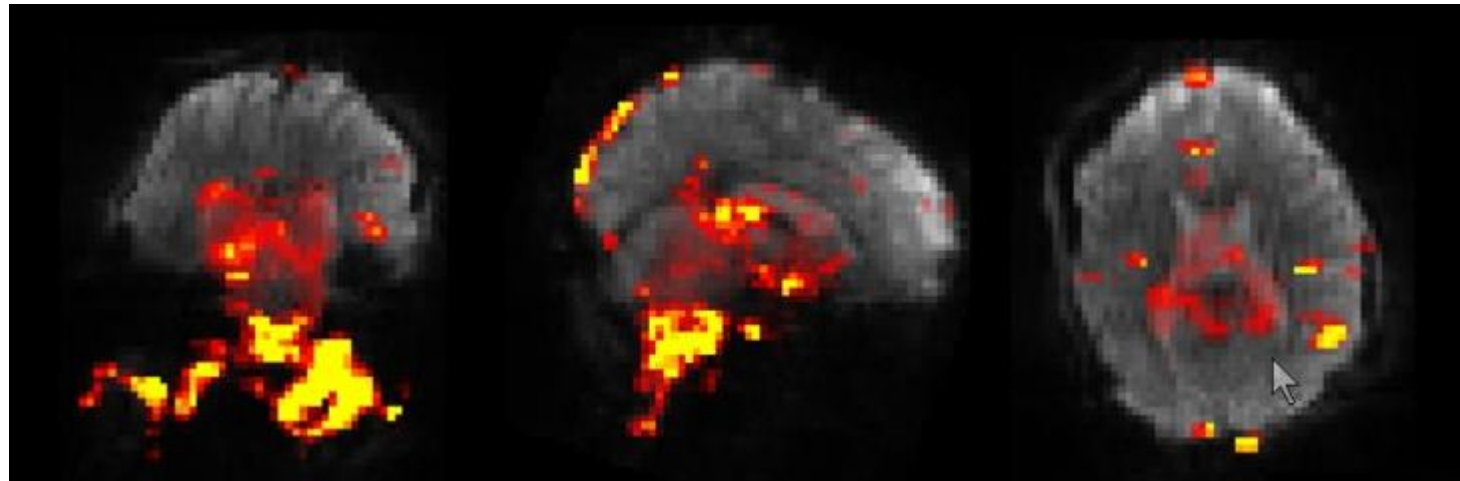


What are the characteristics of physiological signals ?

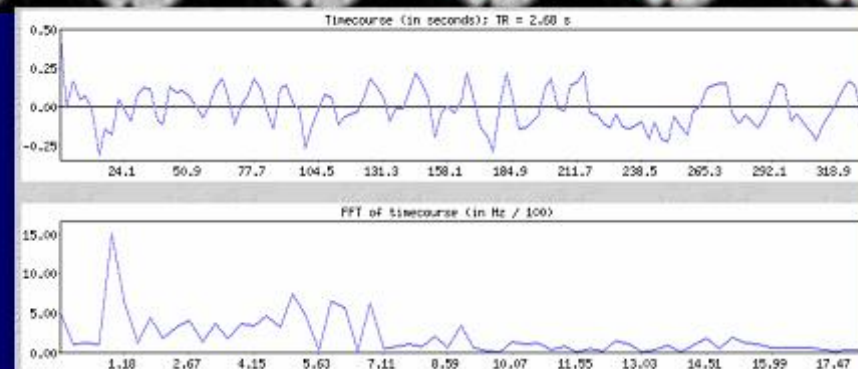
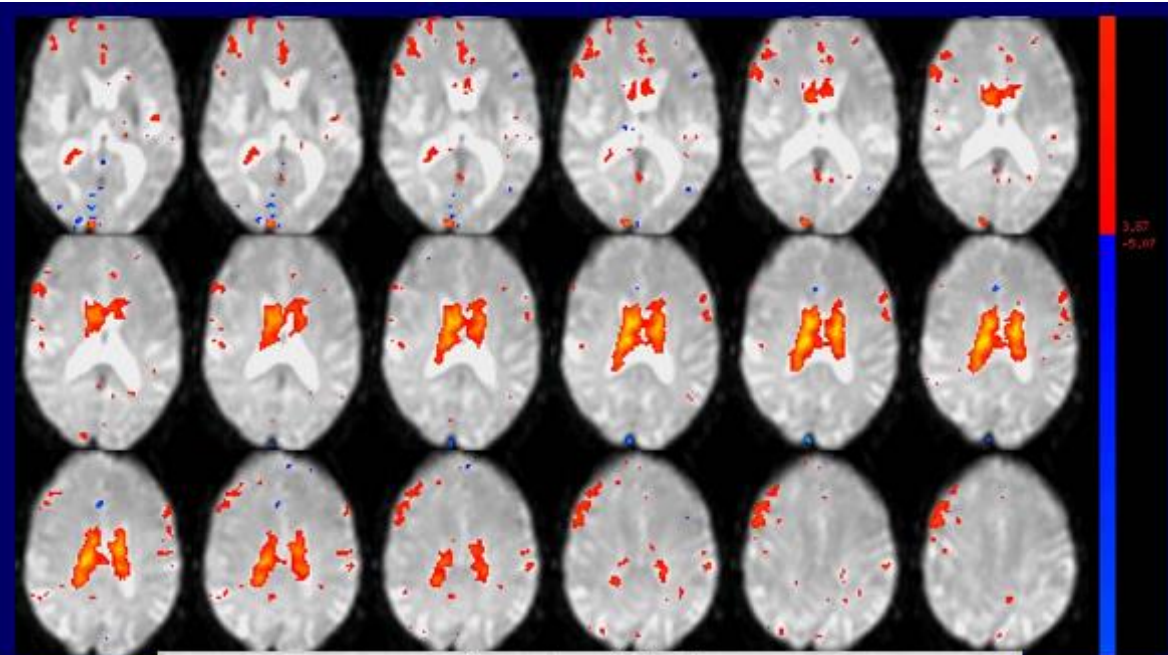
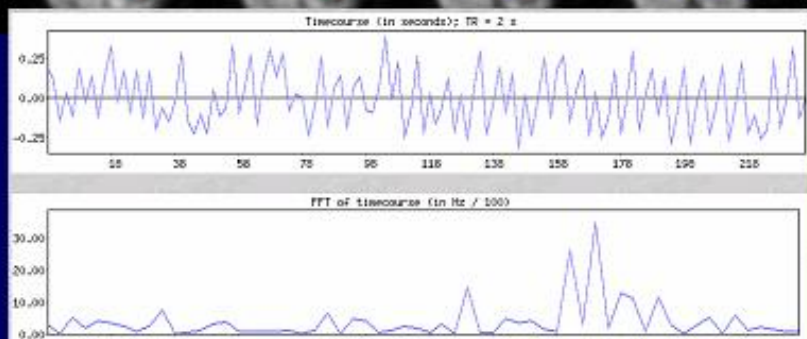
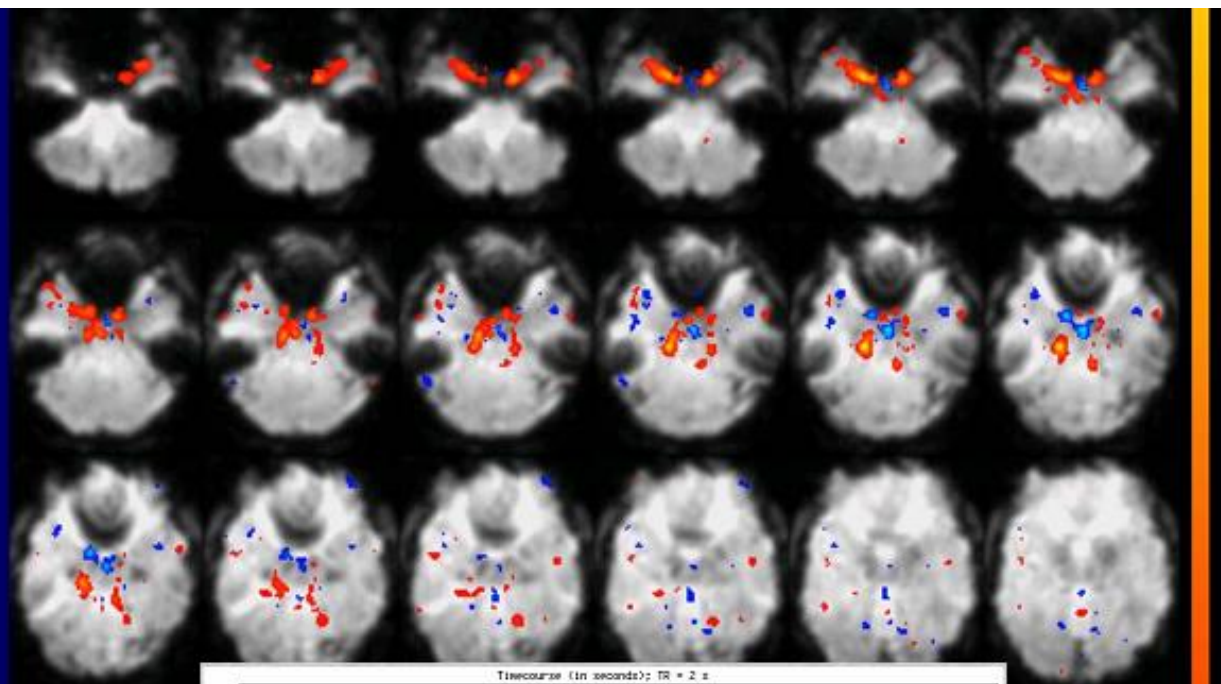
Respiratory 0.3 Hz



Cardiac 1.3 Hz

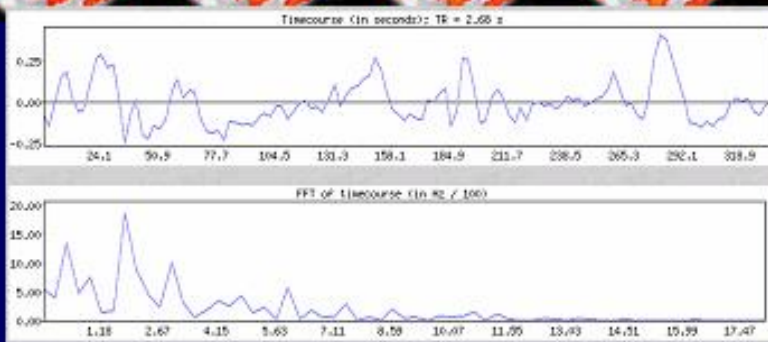
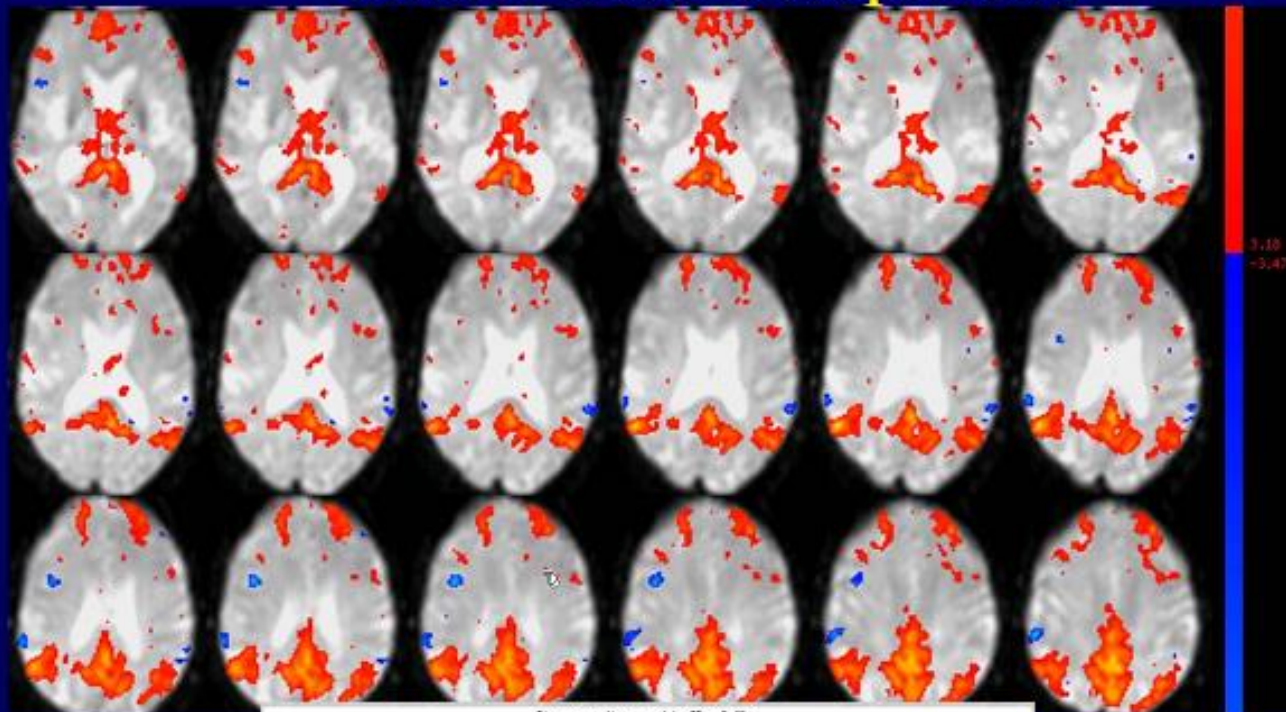


Bad components?



Good components?

“Good” ICA Component

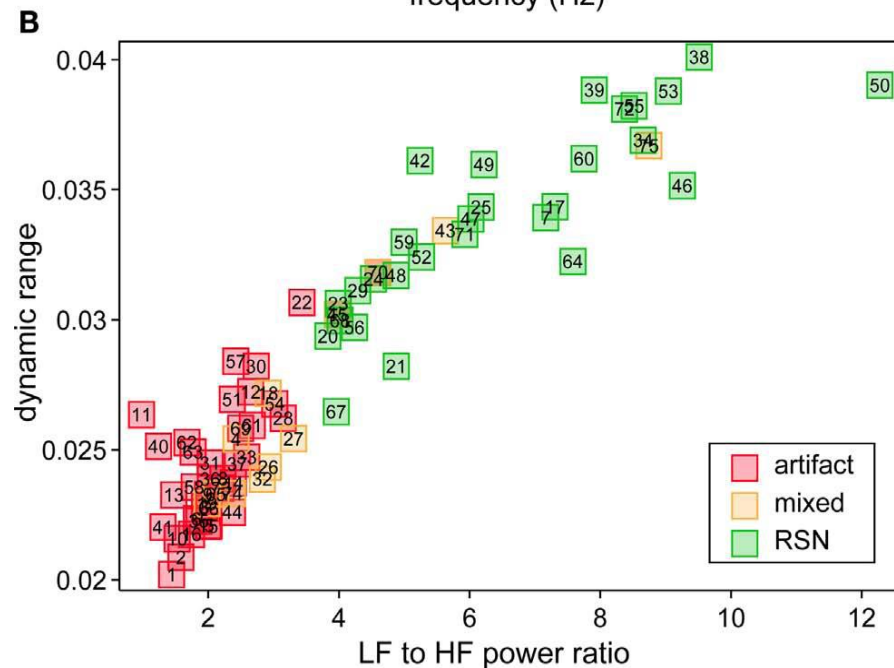
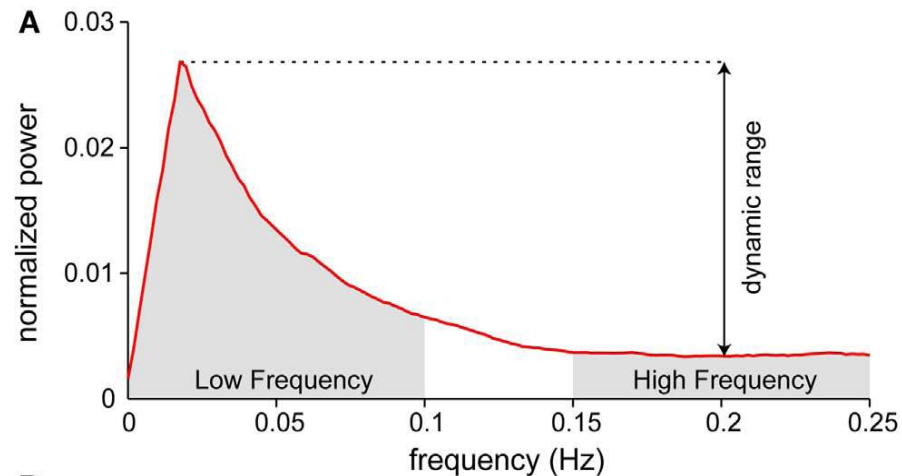


How to “denoise” fMRI data?

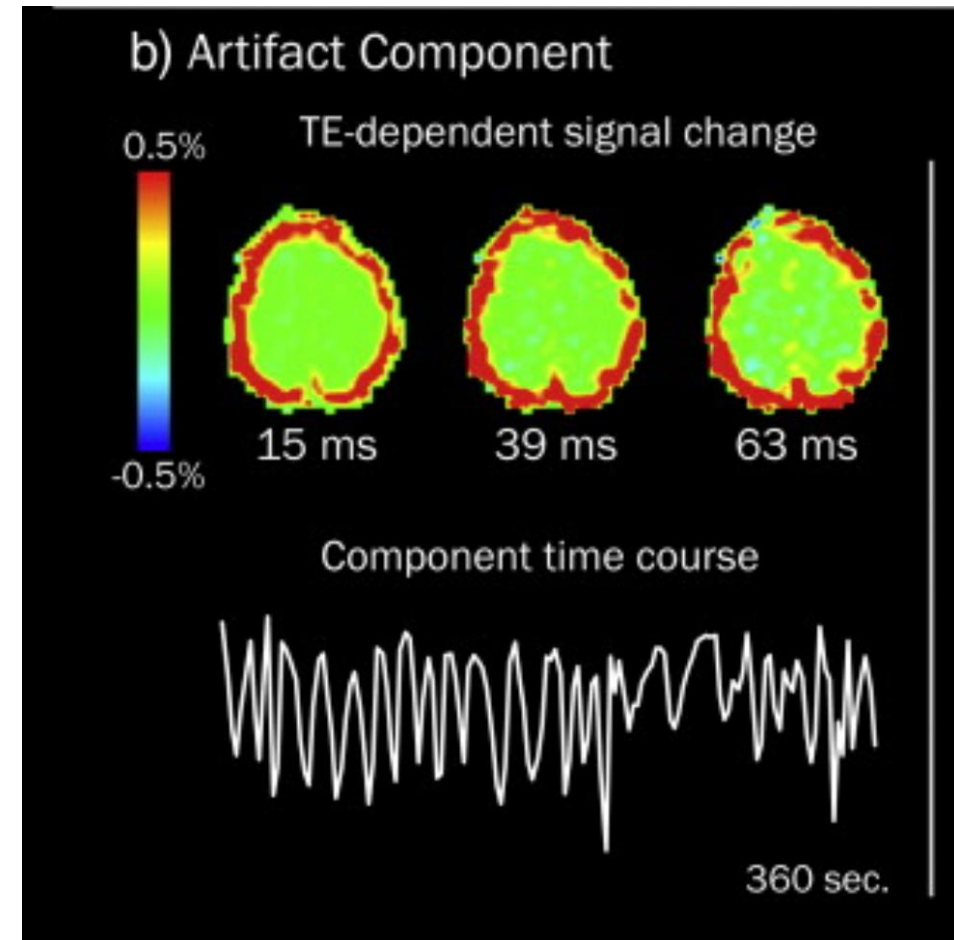
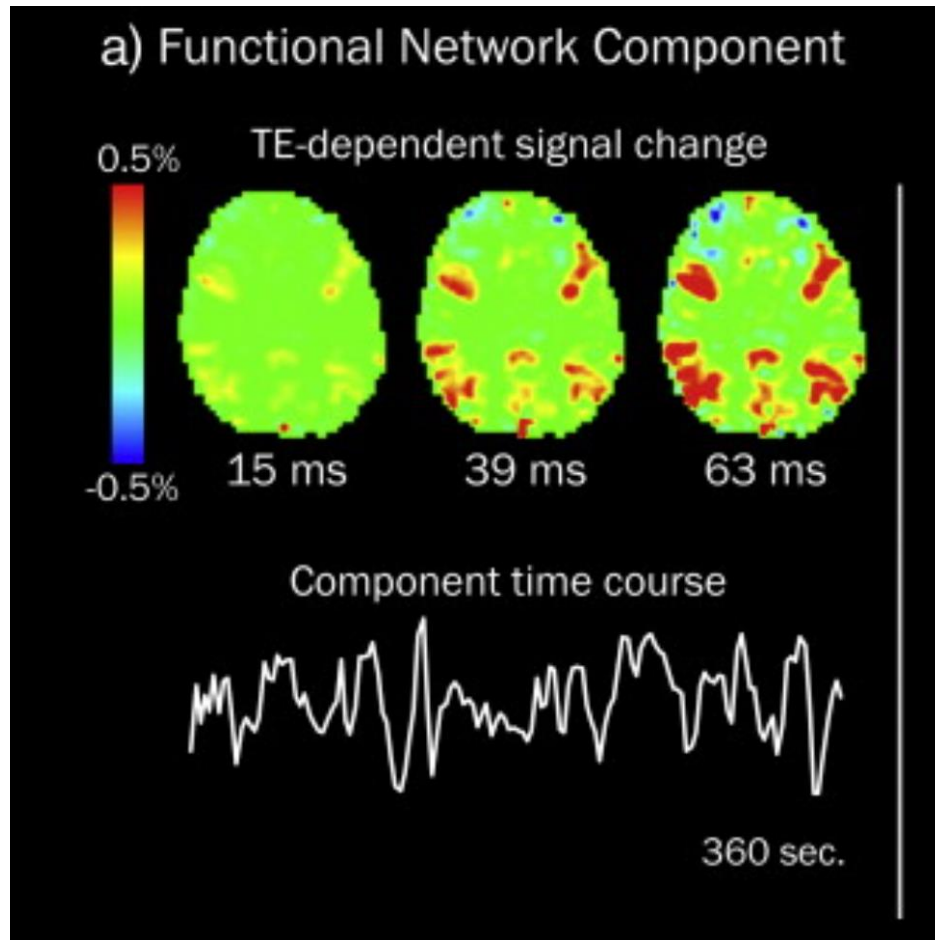
fALFF

fractional amplitude of
lowfrequency fluctuations

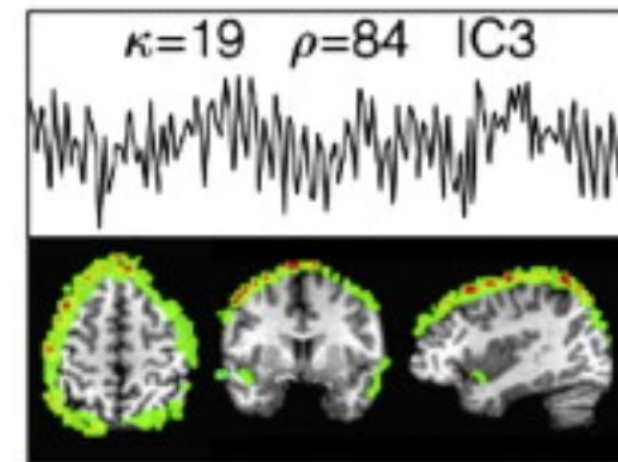
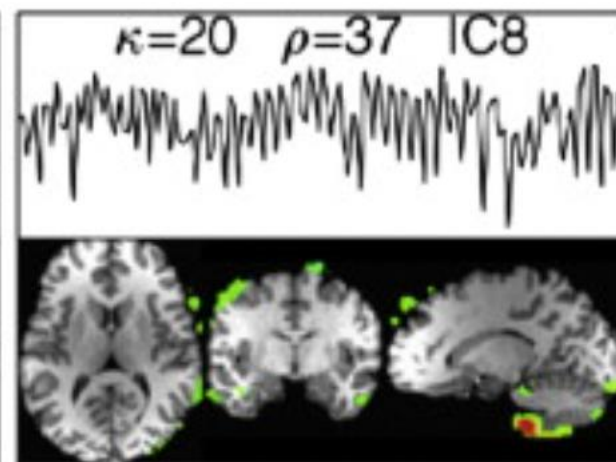
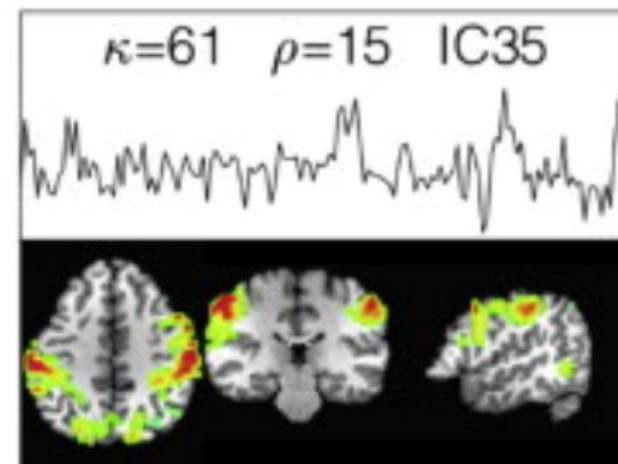
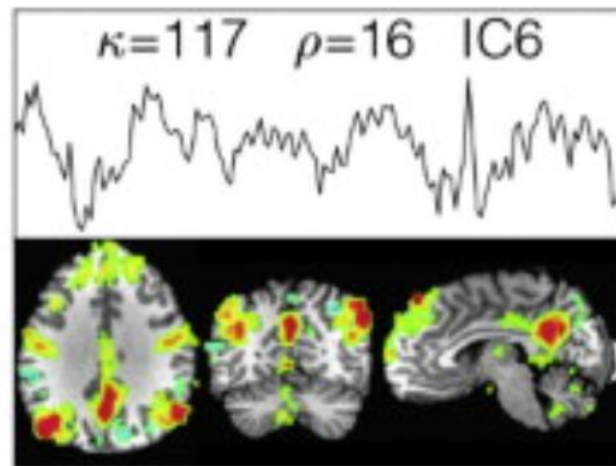
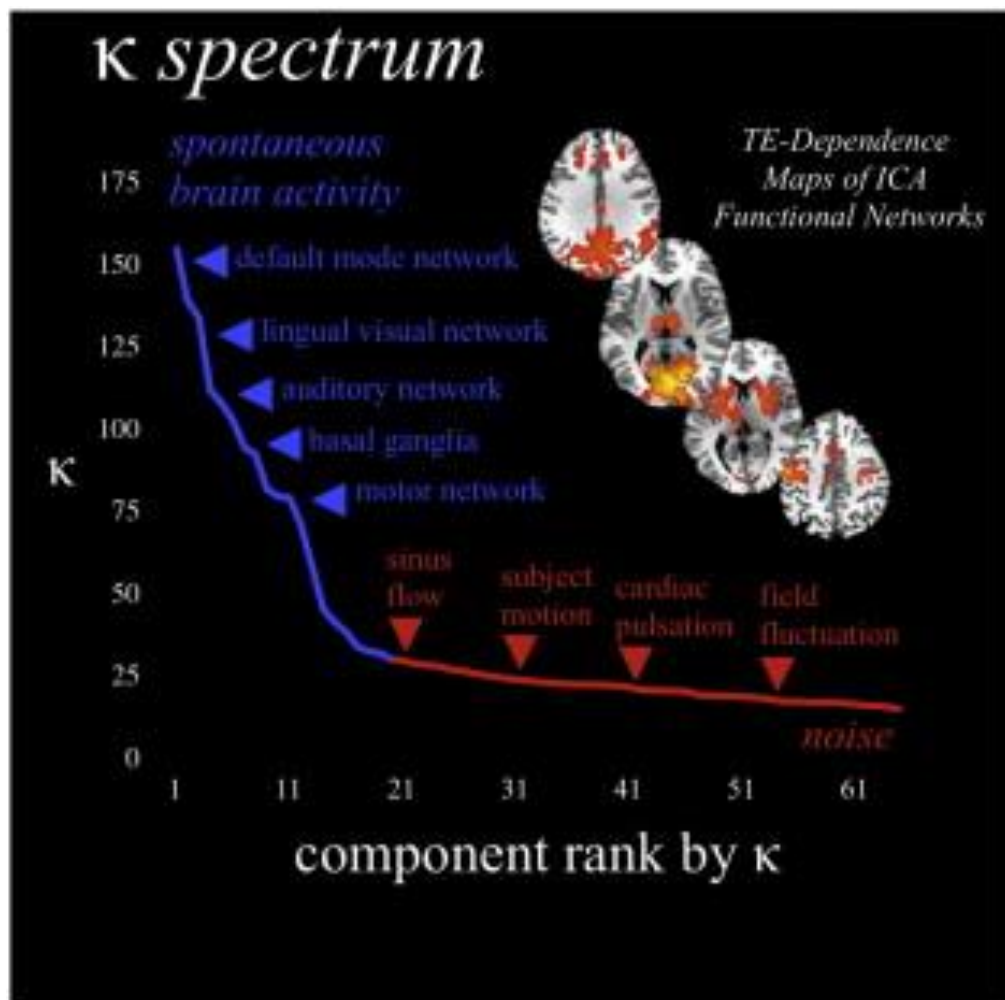
summing the oscillatory
amplitudes across the typical
0.01-0.08 Hz range, then
dividing by the amplitude sum
across 0-0.25 Hz



Multi-Echo EPI



Multi-Echo EPI



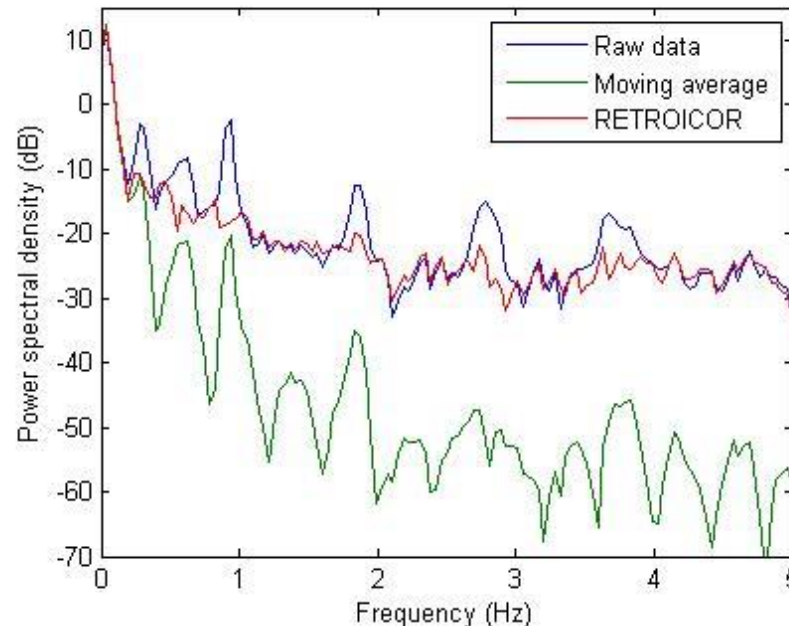
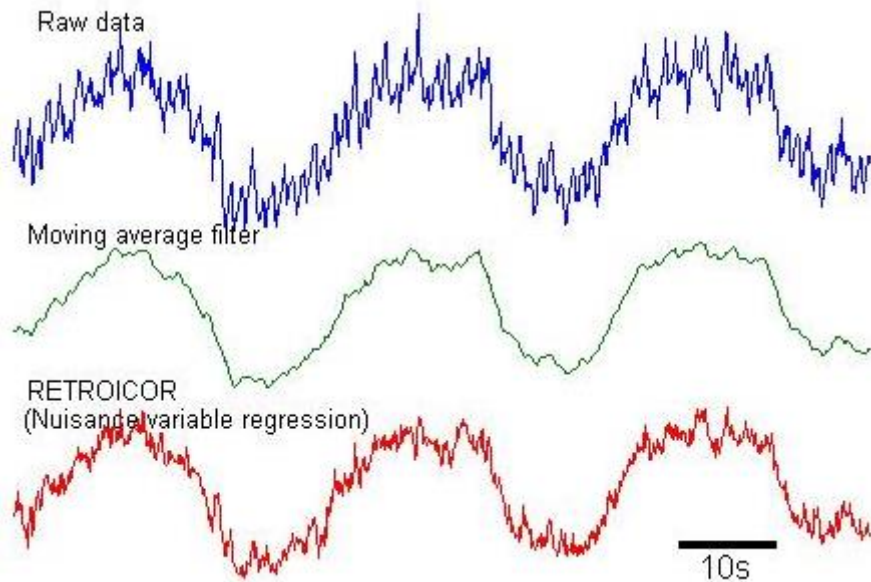
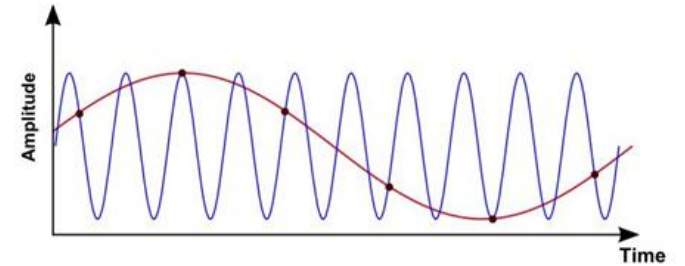
How to “remove” physiological pulsations

Moving average filter (Lin et al., 2011; Posse et al., 2012)

Removes high frequencies and affects noise properties

Artefact modeling (Glover et al., 2000; Lund et al., 2006)

Model noise as high-order autoregressive process



Regression

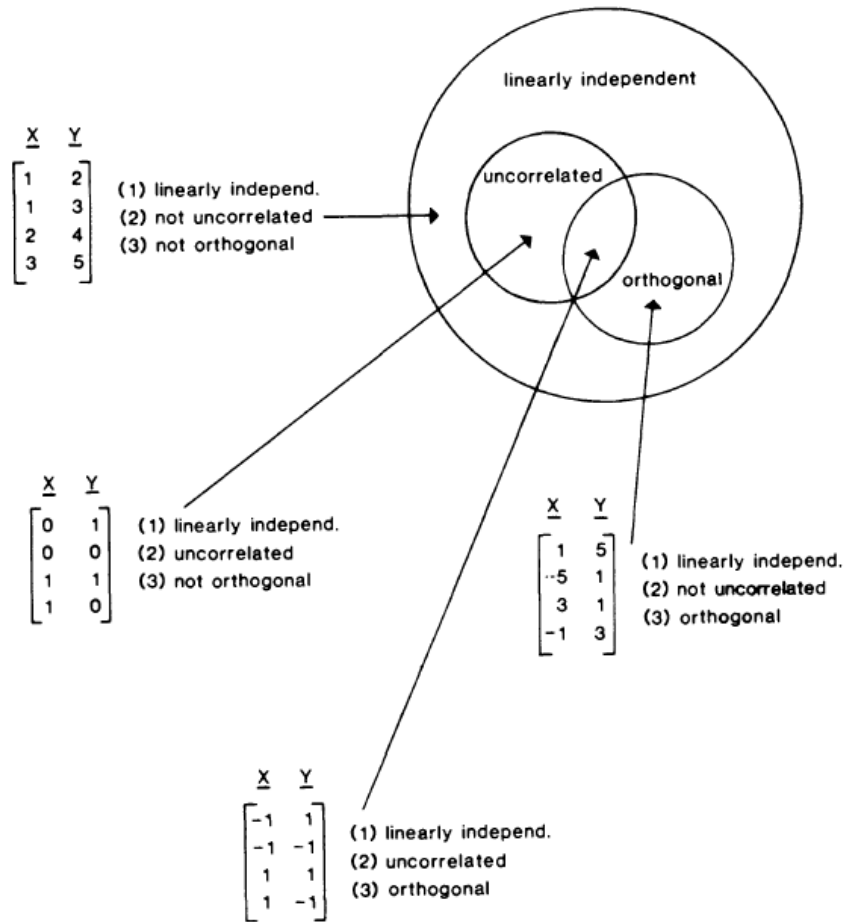
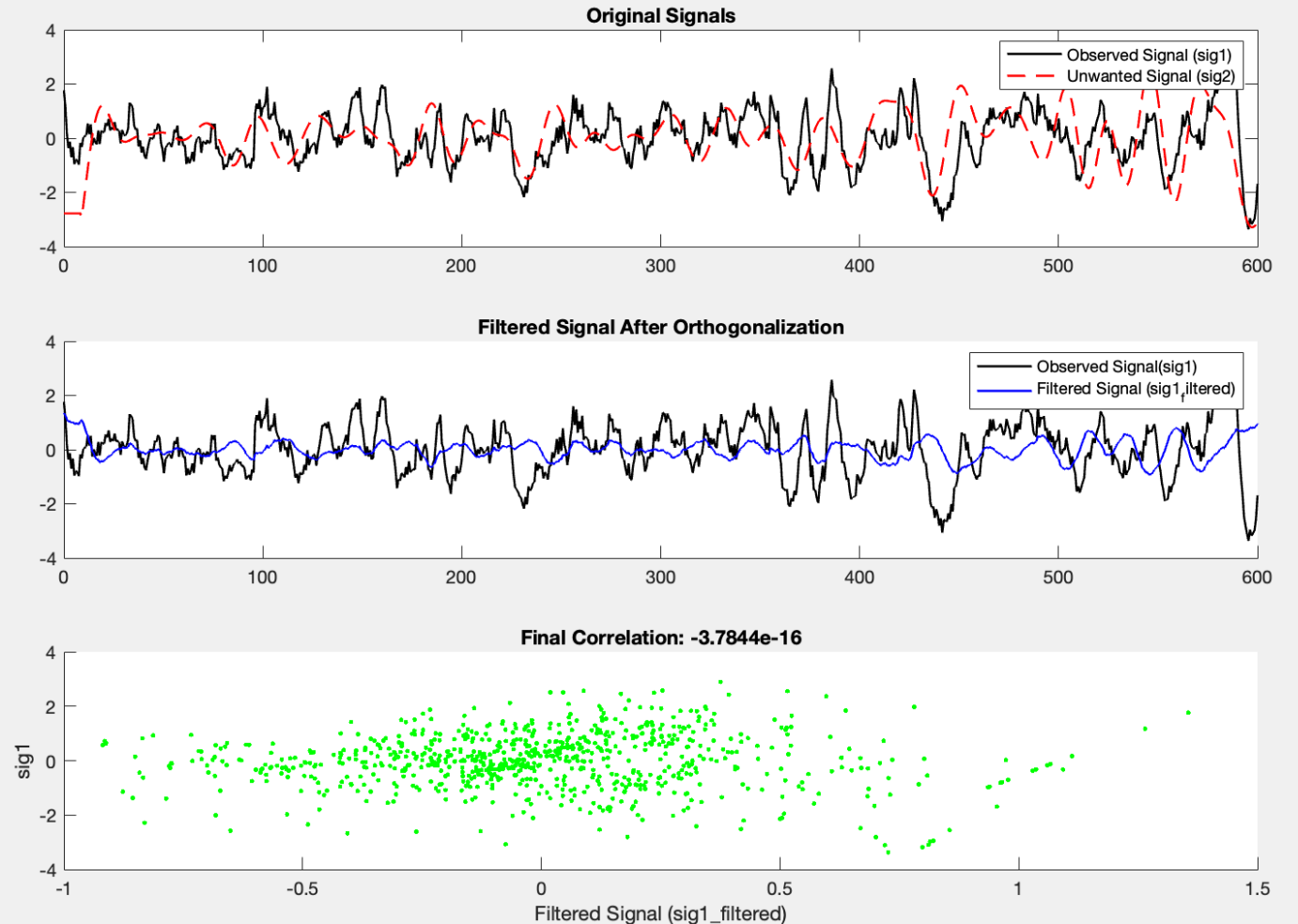
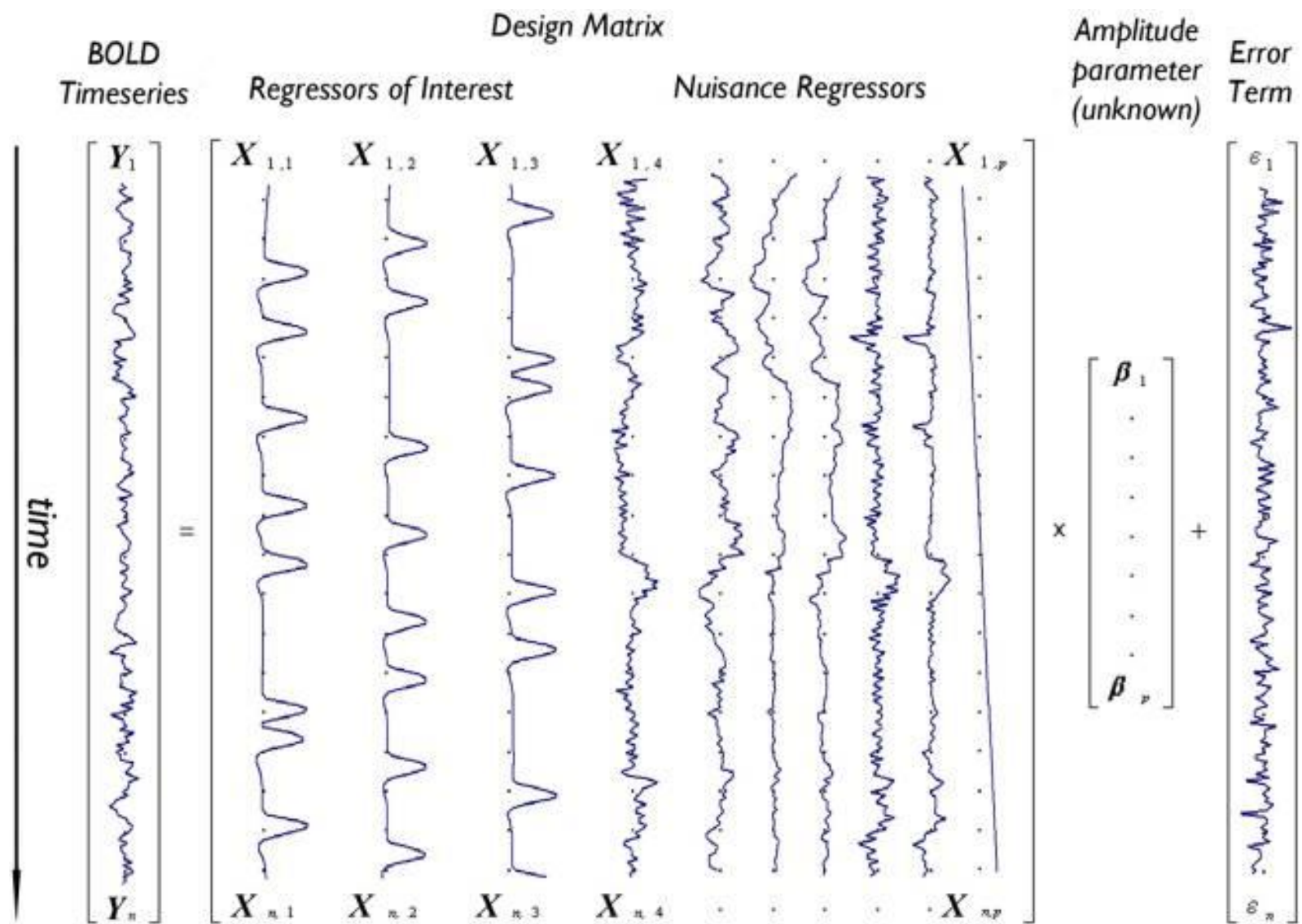


Figure 1. The relationship between linearly independent, orthogonal, and uncorrelated variables.



Regression



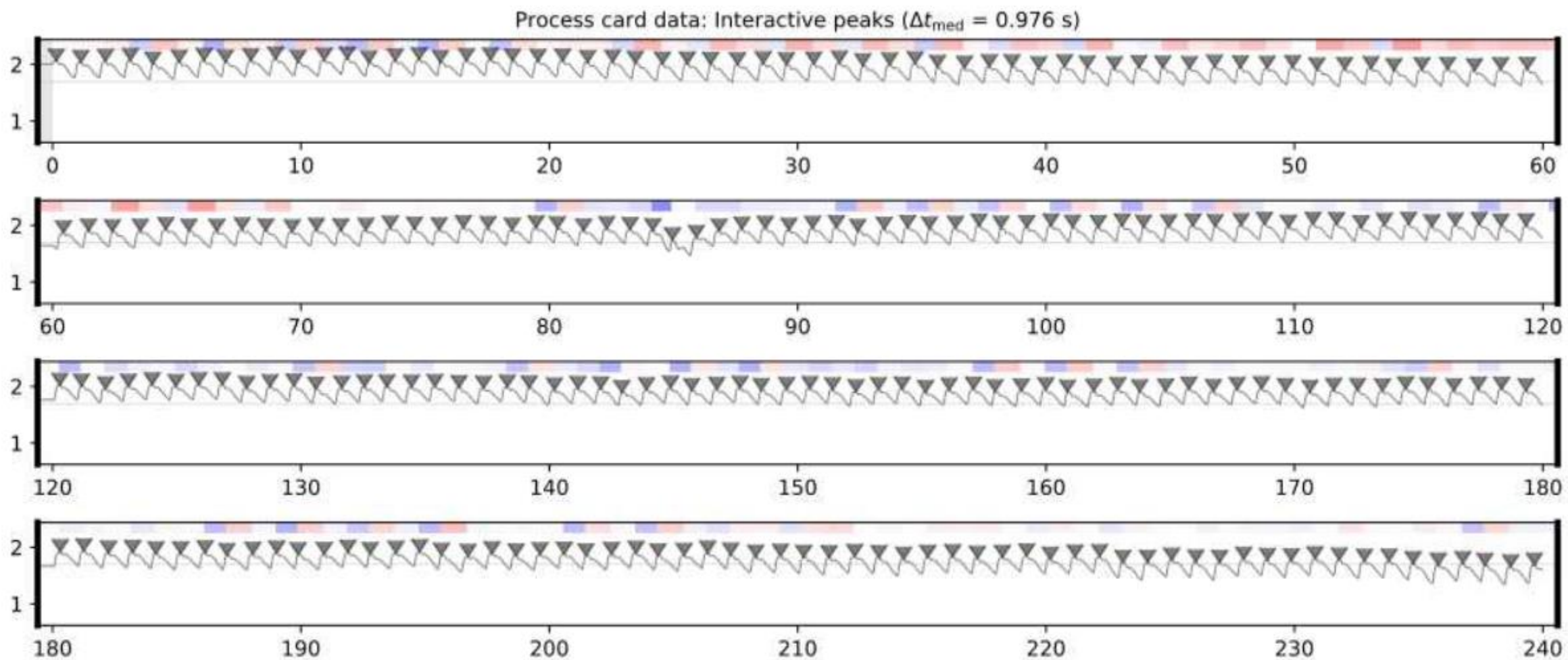
AFNI physio_calc.py



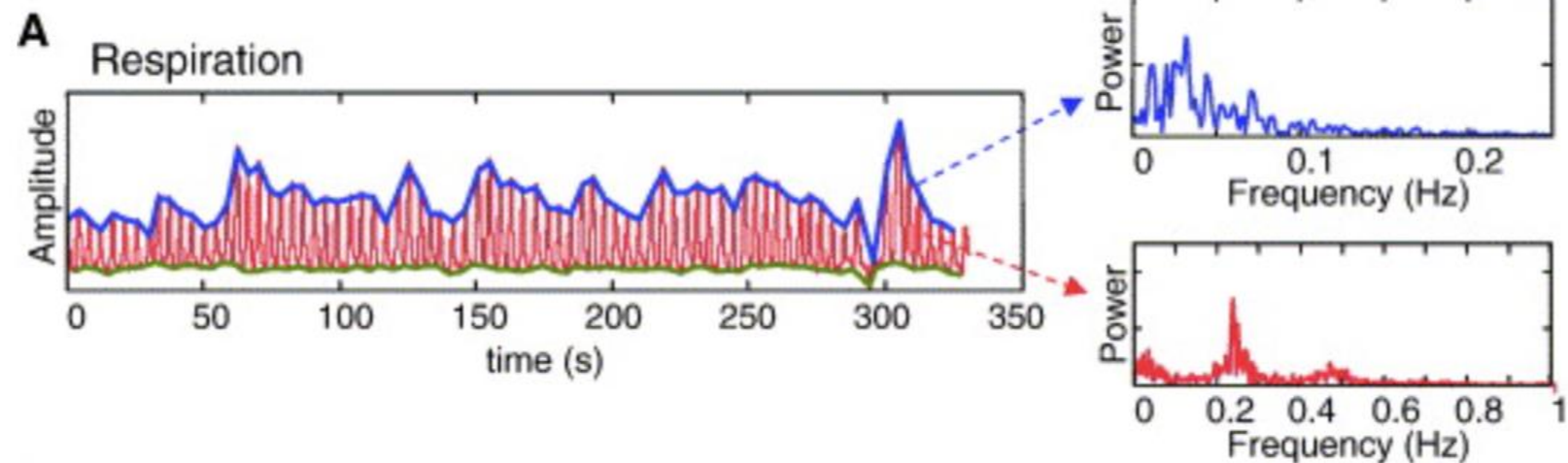
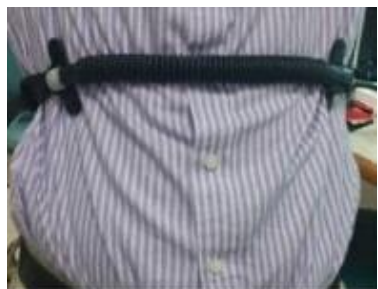
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physio_calc.py \
  -card_file      sub-001_ses-01_task-rest_run-1_physio-ECG.txt \
  -resp_file      sub-001_ses-01_task-rest_run-1_physio-Resp.txt \
  -freq           500 \
  -dset_tr        0.1 \
  -dset_nt        219 \
  -dset_nslice    33 \
```

```
afni_proc.py -subj_id bp10 \
  -blocks tcats despikes tshift align tlrc volreg mask scale regress \
  -radial_correlate_blocks tcats volreg \
  -copy_anat /data/akinb2/allbp/bp10/anat/anatSS.bp10.nii \
  -anat_has_skull no \
  -anat_follower_ROI aaseg anat /data/akinb2/allbp/fsurfs/bp10/SUMA/aparc.a2009s+aaseg.nii.gz \
  -anat_follower_ROI aaseg epi /data/akinb2/allbp/fsurfs/bp10/SUMA/aparc.a2009s+aaseg.nii.gz \
  -tcats_remove_first_trs 0 \
  -dsets /data/akinb2/allbp/bp10/rest.nii \
```

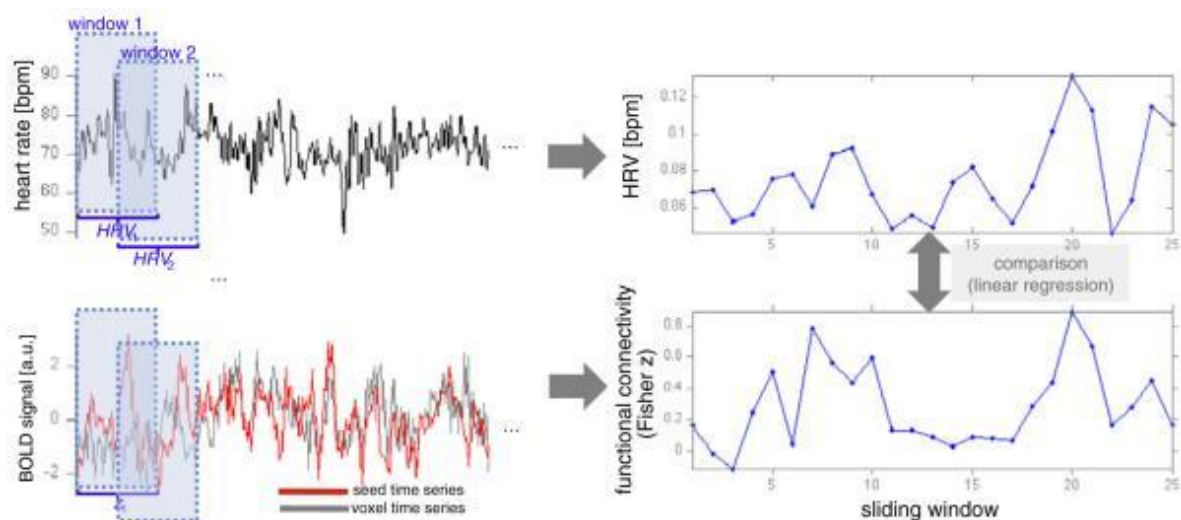
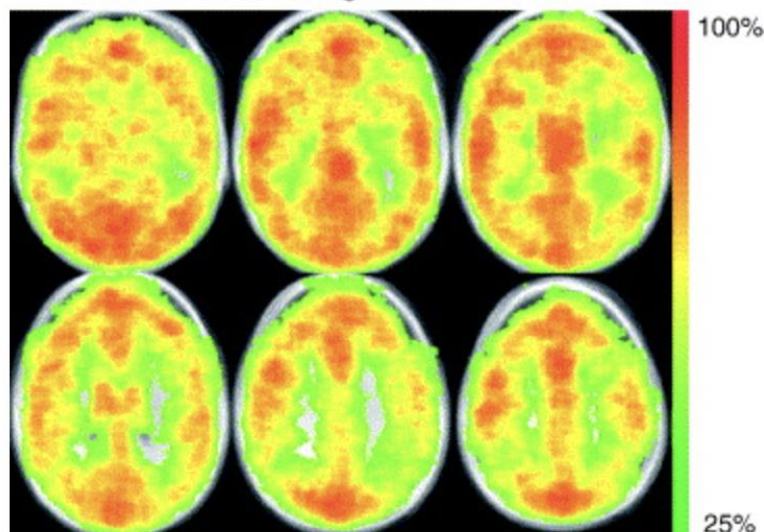

Finger pulsation- peak detection



Indirect modulation



% of time-series signif. corr. w/ RVT



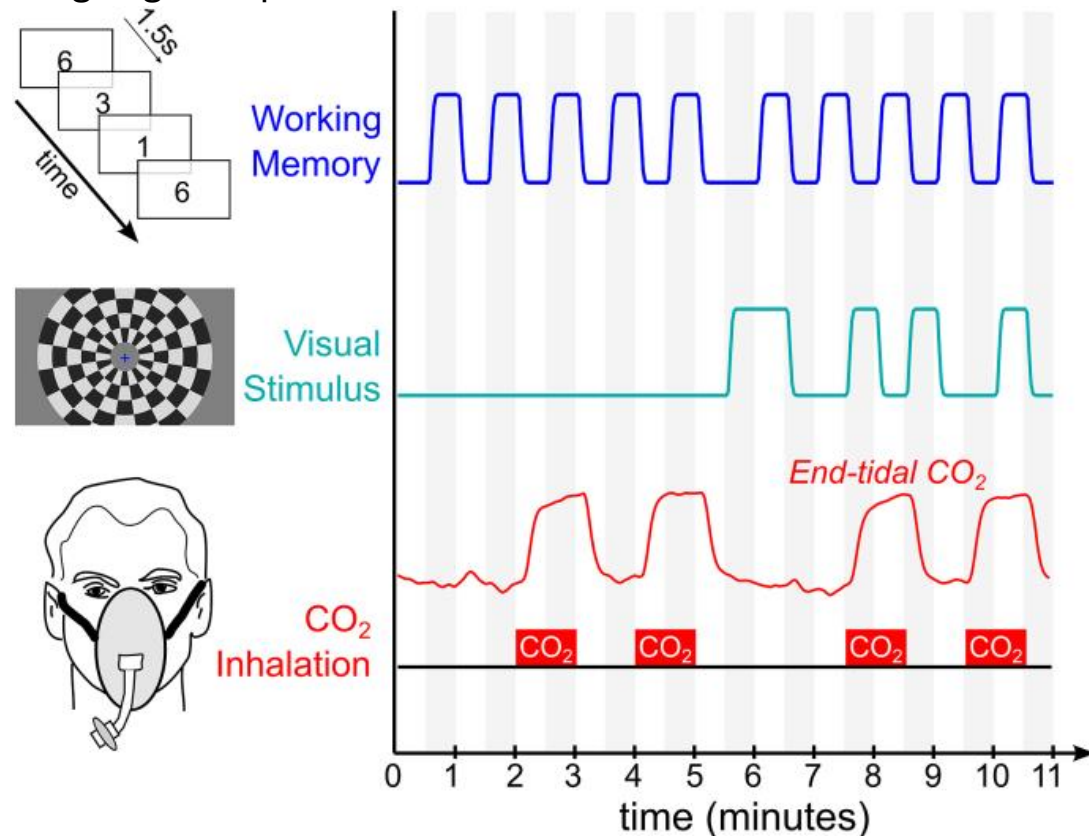
Vascular physiology drives functional brain networks

Molly G Bright ¹, Joseph R Whittaker ², Ian D Driver ³, Kevin Murphy ²

Affiliations + expand

PMID: 32387624 PMCID: [PMC7339138](#) DOI: [10.1016/j.neuroimage.2020.116907](#)

Designing an experiment with distinct tasks

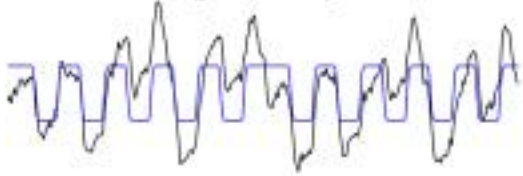


Designing an experiment with distinct tasks

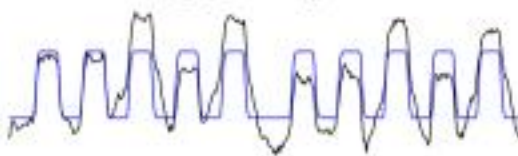
Step 1

Identify neuronal networks using temporal correlations

Maximum negative correlation with Working Memory task



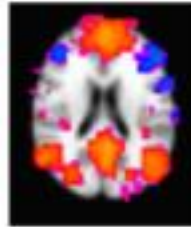
Maximum positive correlation with Working Memory task



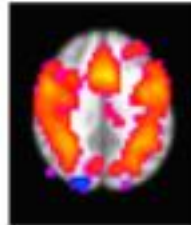
Maximum positive correlation with Visual stimulus



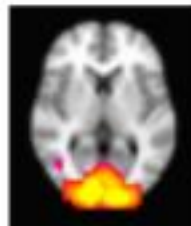
Default Mode Network



Task Positive Network



Visual Network



Step 2

Identify similar network using spatial overlap

overlap
= 0.34

overlap
= 0.48

overlap
= 0.29

Step 3

Analyze time-series of spatially similar networks

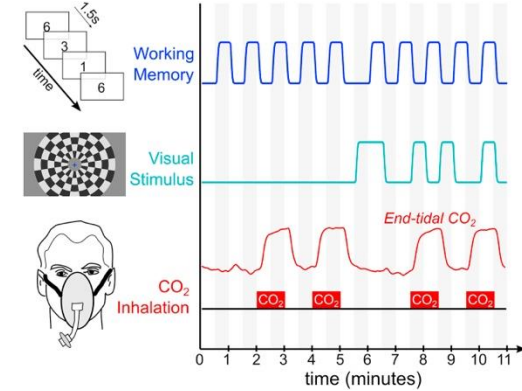
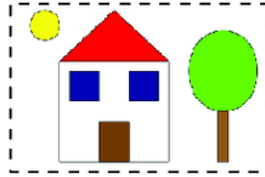


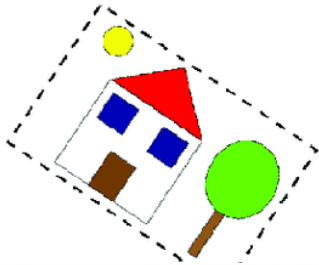
Image registration

Original Image

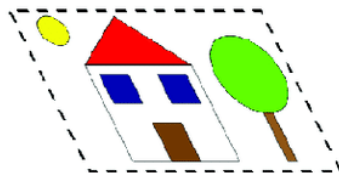


Transformation

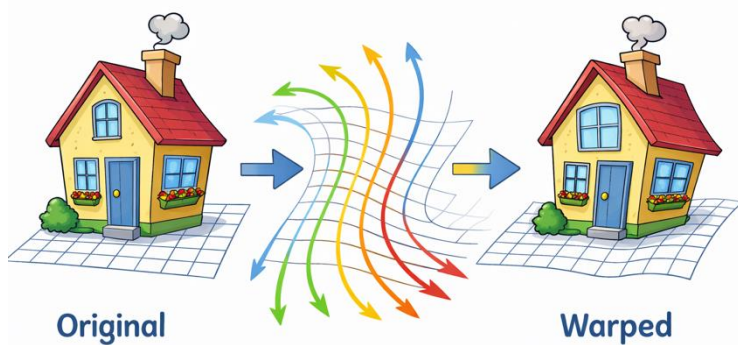
Rigid



Affine

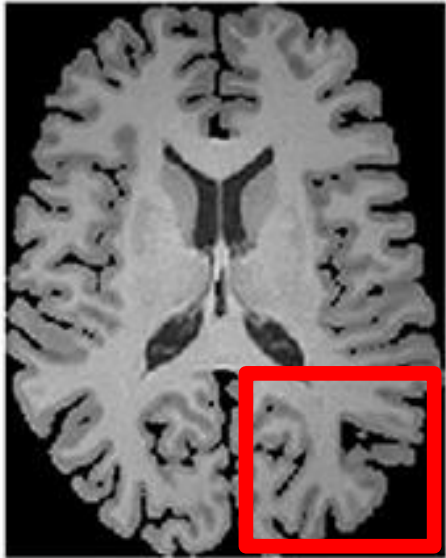


Deformable

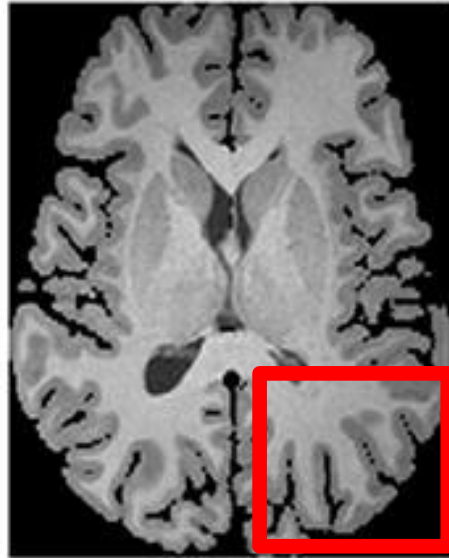


Time

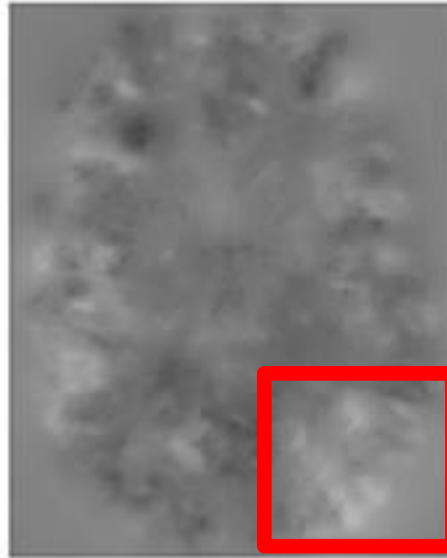
Non-linear registration



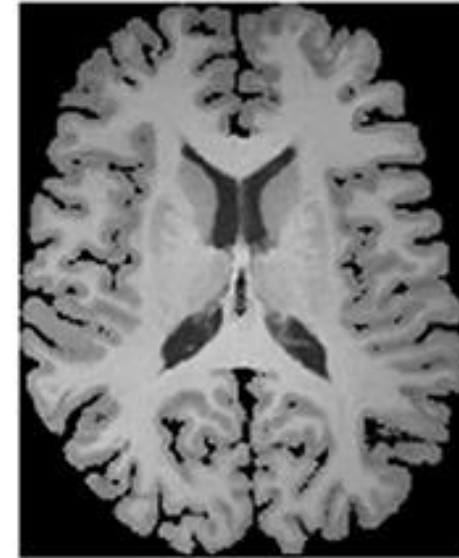
Fixed



Moving



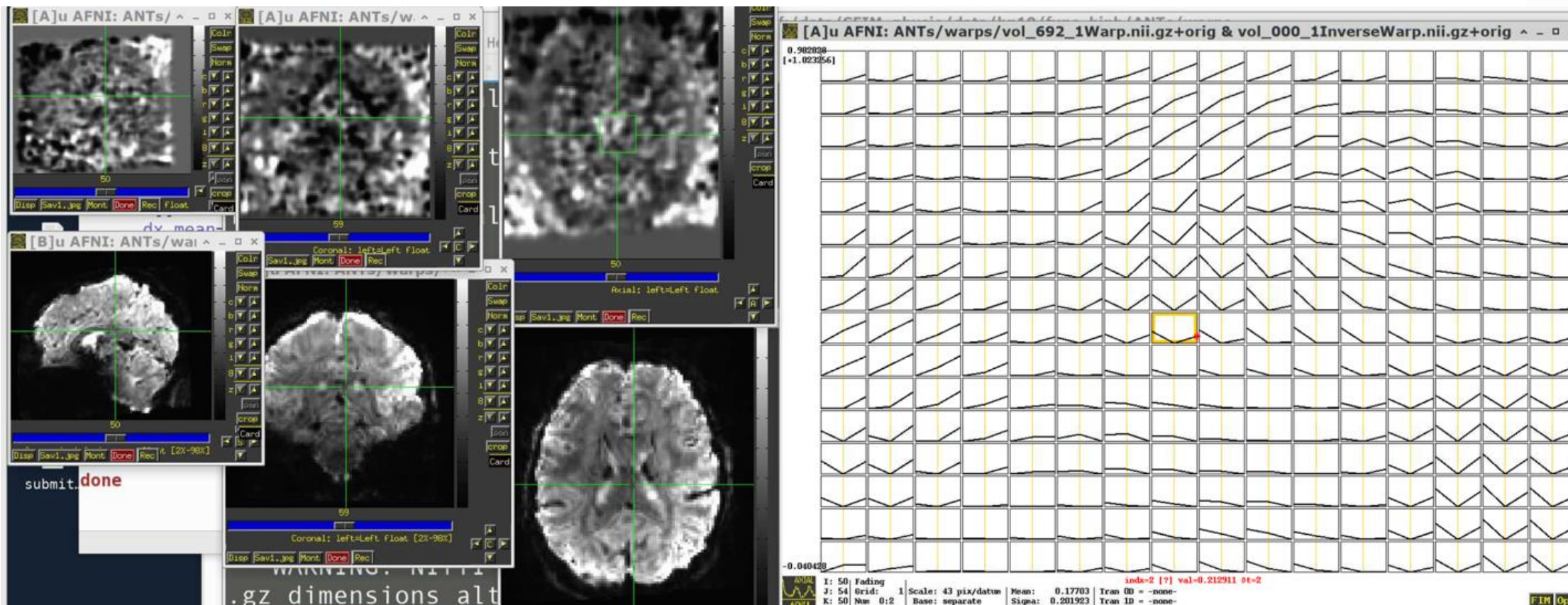
Deformation



Warped

```
antsRegistration \  
  --dimensionality 3 \  
  --float 1 \  
  --collapse-output-transforms 1 \  
  --output [ "${OUTDIR}/warps/vol_${v}_", "$OUTWARP", 0 ] \  
  --interpolation Linear \  
  --winsorize-image-intensities [0.01,0.99] \  
  --use-histogram-matching 0 \  
  --initial-moving-transform [ "$REF", "$MOV", 1 ] \  
  --transform SyN[0.03,2,0] \  
  --metric CC["$REF","$MOV",1,2] \  
  --convergence [10x5,1e-5,5] \  
  --shrink-factors 2x2 \  
  --smoothing-sigmas 1x1vox
```

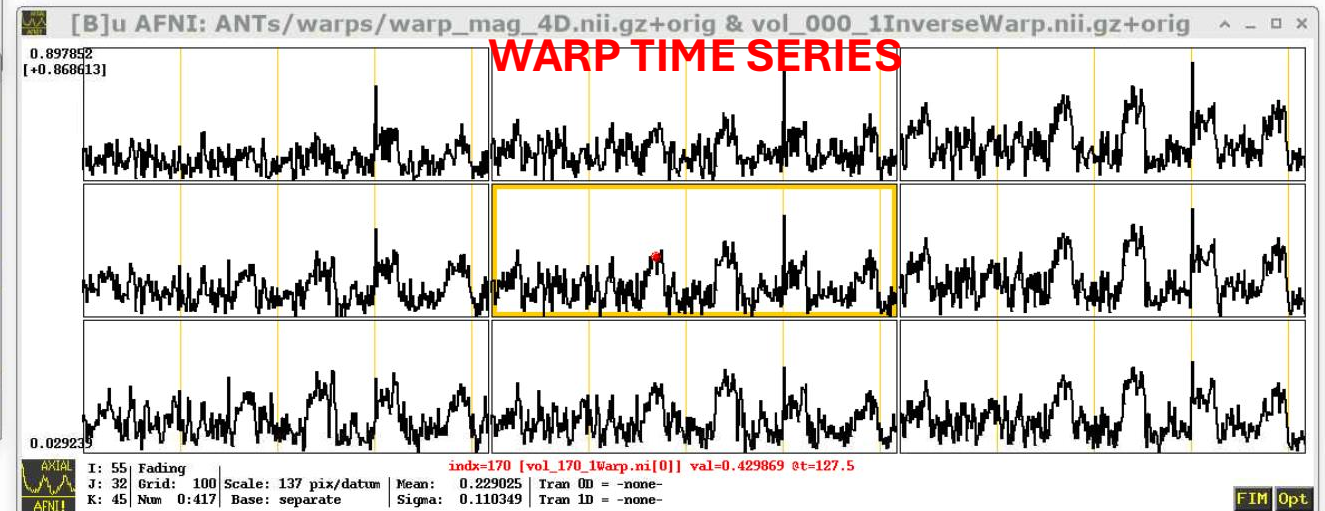
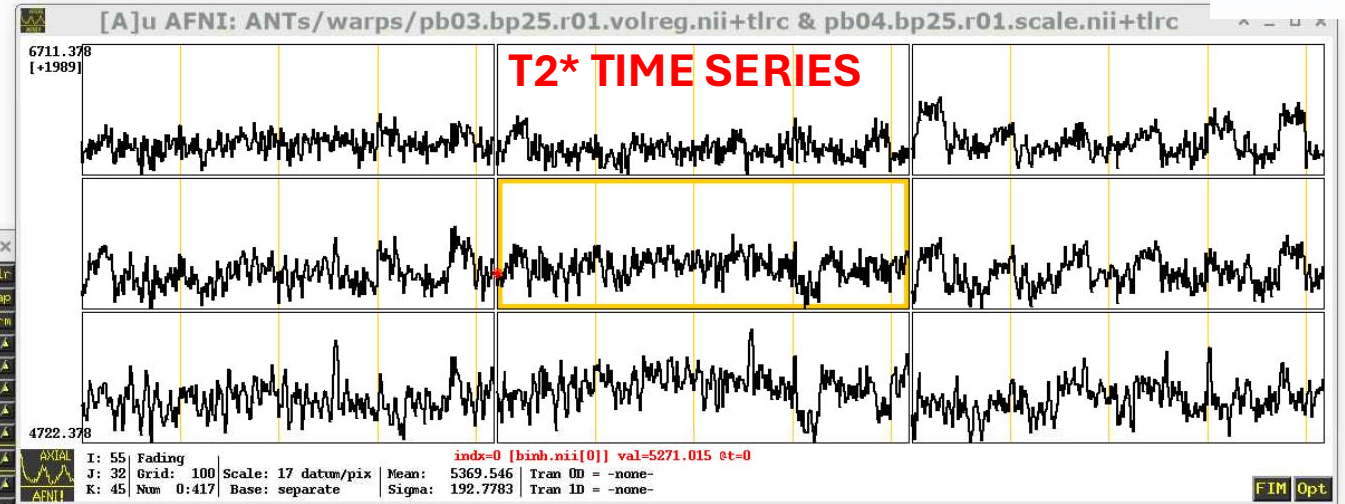
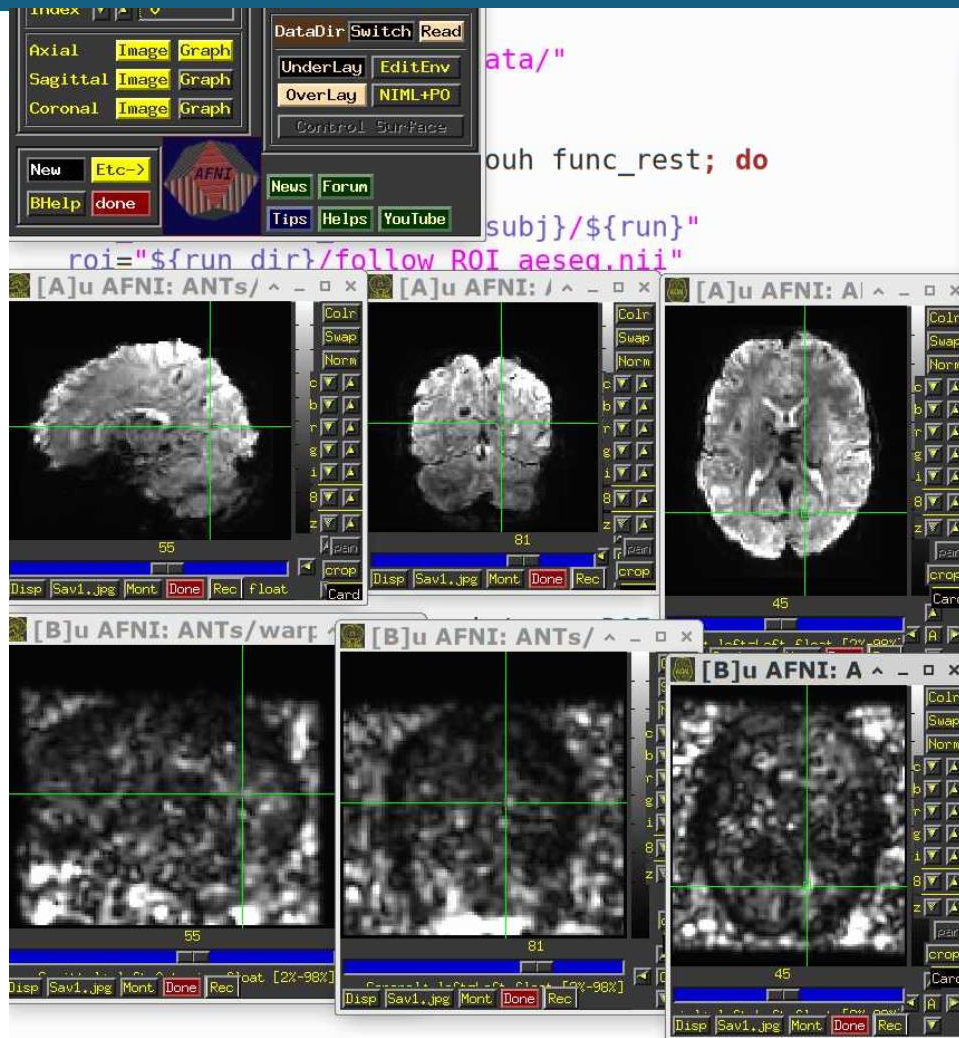

Image registration (T2* fMRI images)



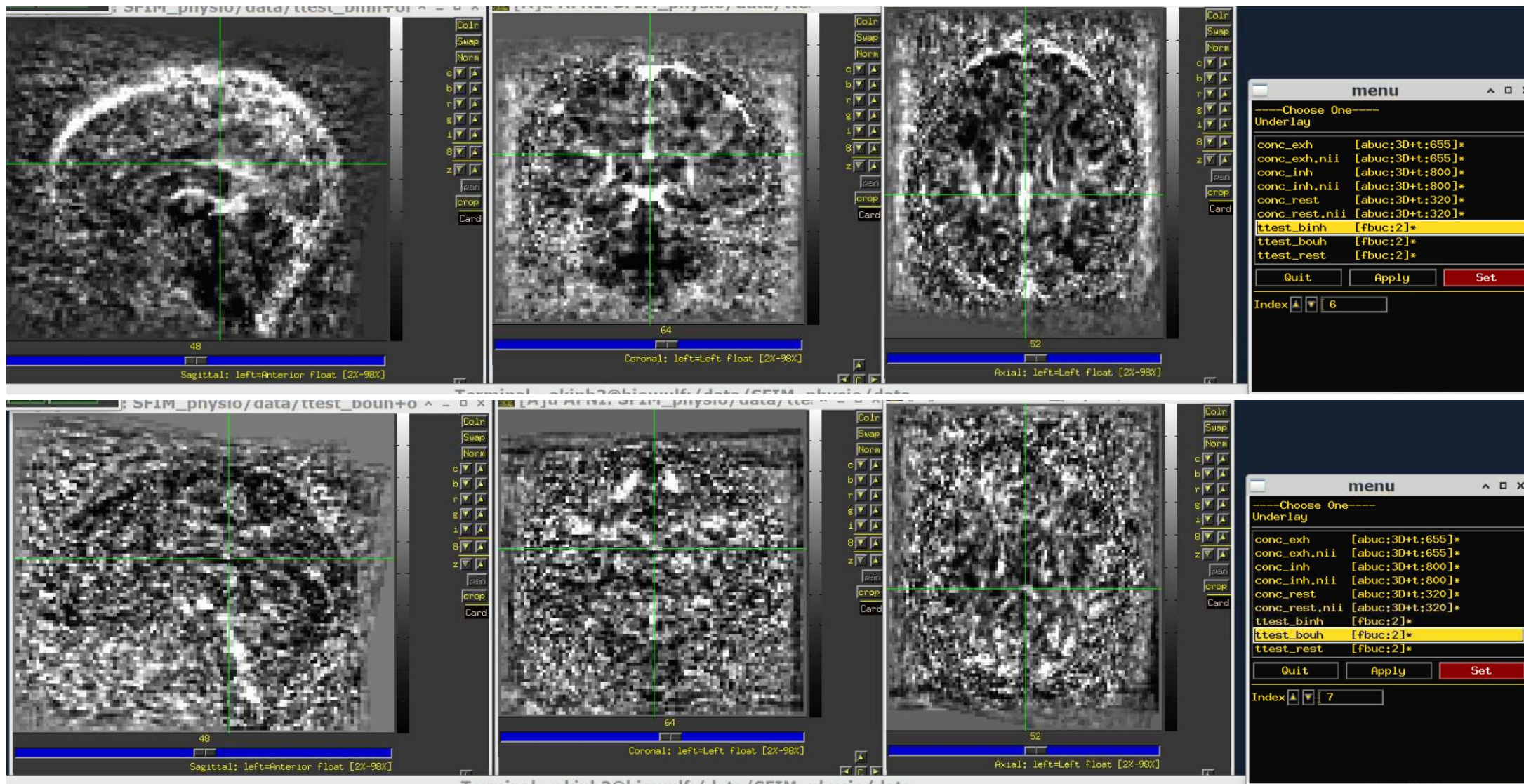
$$\text{Magnitude} = \sqrt{dx^2 + dy^2 + dz^2}$$

$$\theta = \arccos \left(\frac{dz}{\sqrt{dx^2 + dy^2 + dz^2}} \right)$$

Warp time series



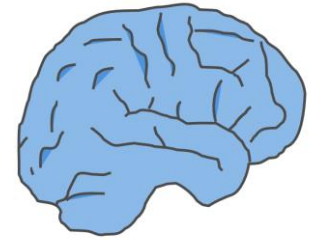
Group level concordance



Some take home messages

- Always look at your data
- Always acquire physiology
- Physiology might be both noise or signal
- Design a task paradigm that helps to disentangle sources of signals
- Design a processing pipeline that helps to delineate the task related activity

Thanks to colleagues...



**Any
questions ?**