



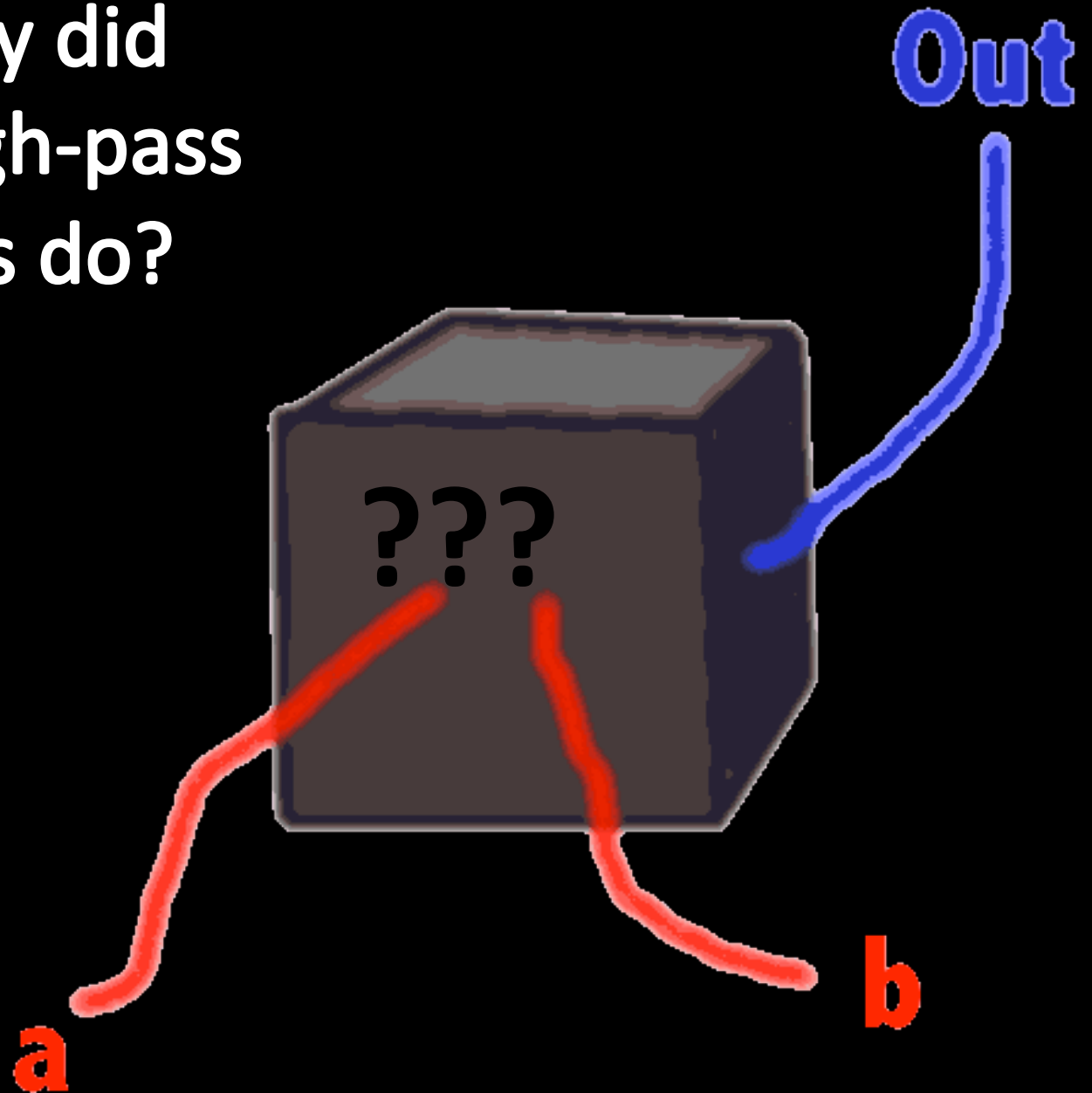
Preprocessing fMRI Data

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USC Neuroimaging Workshop

08.11.2011

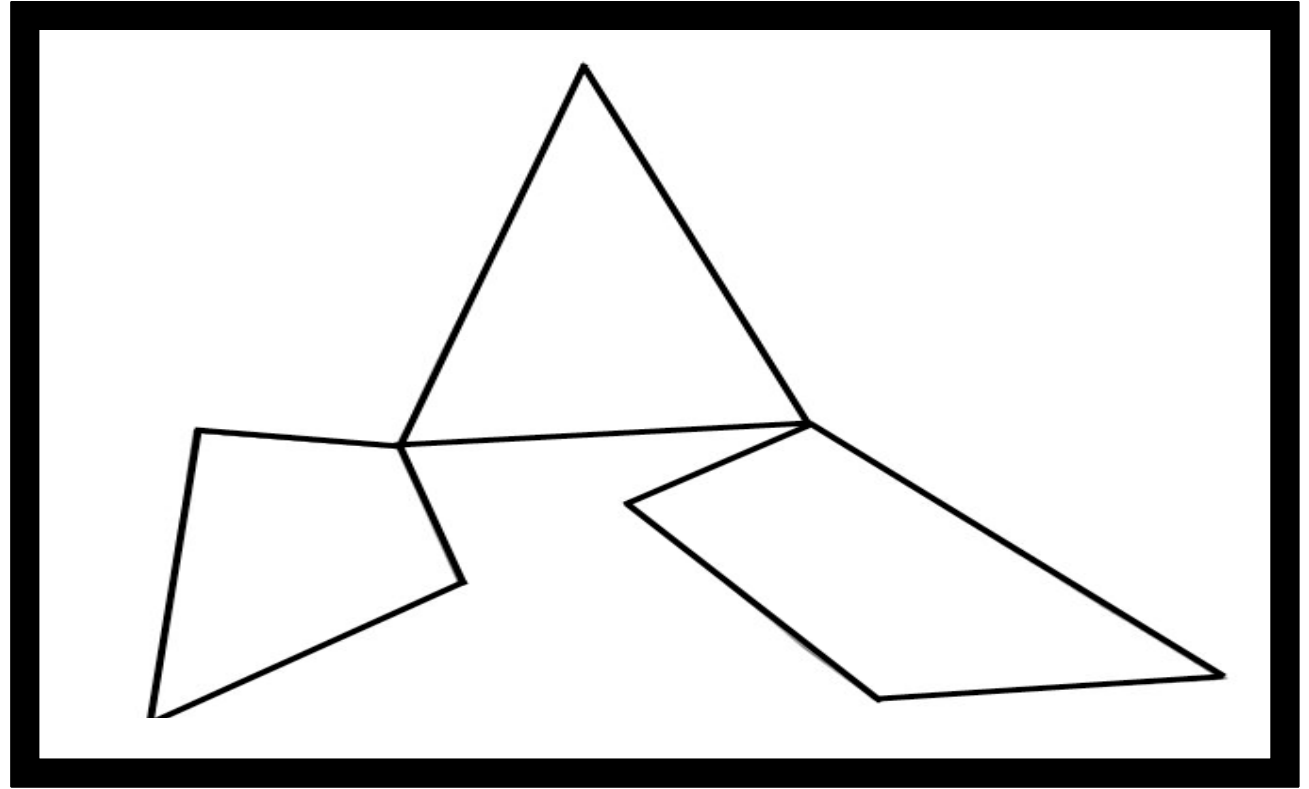
What exactly did
choosing a high-pass
filter of 100s do?



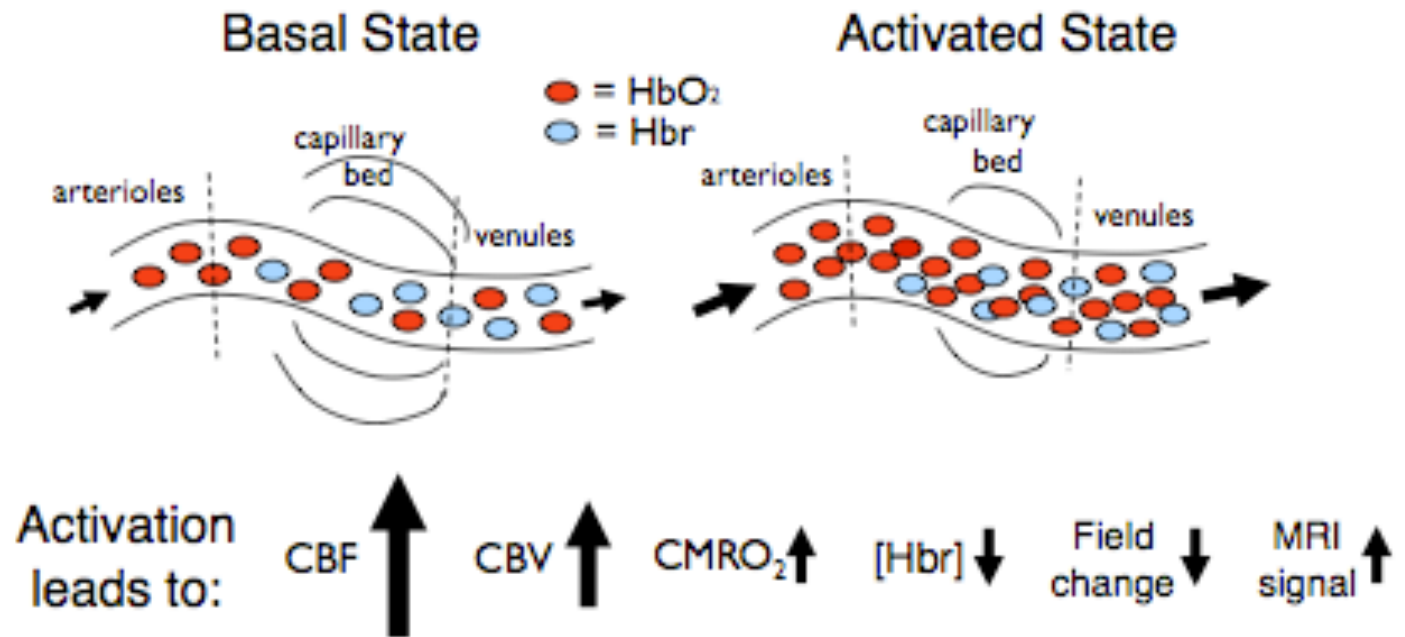
Typical Process:



Typical Process:



Typical Process:



- Field changes (perturbations) --> dephasing --> T₂^{*} effect
- BOLD-tuned MRI (T₂^{*}-weighted) is sensitive to this effect

Why Preprocess?

- Data is noisy!
- Subjects move!
- Things change over the time of your experiment (more if it's long and boring)!
- There a giant bulldozers and excavators 10 feet away from your scanner! Wait, that's not normal..?
- Etc etc etc...

Why Preprocess?

- Preprocessing attempts to increase Signal (e.g., BOLD contrast) to Noise (variance from movement, subject, scanner artifacts, other uncontrollables) ratio
- Helps you meet assumptions so you can do stats on your data

$$y = X \times \beta + \varepsilon$$

fMRI Signal = Design Matrix × Parameter + Residuals

"our data" = "what we CAN explain" × "how much of it we CAN explain" + "what we CANNOT explain"

Preprocessing Resources

- Full credit for all the info and most of the slides (and nice pictures) goes to:
- FSL tutorials-
 - <http://www.fmrib.ox.ac.uk/fslcourse/>
- UCLA NITP Summer Course (See Monti 7/12 Preprocessing)
 - <http://www.brainmapping.org/NITP/Summer2011.php>

Preprocessing Steps

- Pre-Preprocessing
 - DICOM transformation, Image reconstruction, BET
- Motion correction
- Slice-timing correction
- Spatial filtering
- Temporal filtering
- Global intensity normalization
- Registration/Normalizing (technically post-preprocessing)

Preprocessing Steps

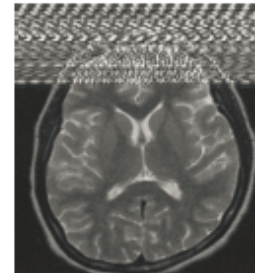
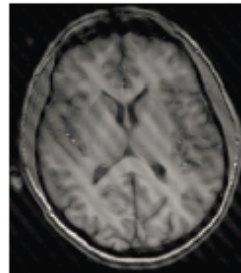
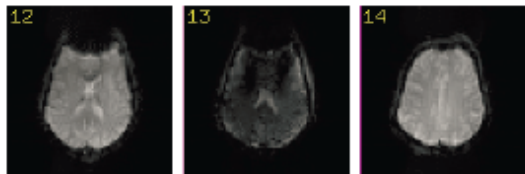
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Pre-Preprocessing

- Scanner output: DICOMS
- Analysis program input: Analyze (SPM), NiFTi (FSL, SPM), BrainVoyager
- SPM, BrainVoyager have built-in functions
- FSL – use MRICRON's dcm2nii convert
 - Can set up preferences on dcm2nii.ini file
 - Put all dcms in one folder
 - `dcm2nii -b /Applications/mricronmac/dcm2nii.ini firstfile.dcm`

Pre-Preprocessing

- Image registration
 - Look at your data! Use FSL video mode.
 - Osirix Viewer (or any other DICOM viewer):
 - <http://www.osirix-viewer.com/>



- Brain extraction
 - Mostly in FSL, use BET toolbox

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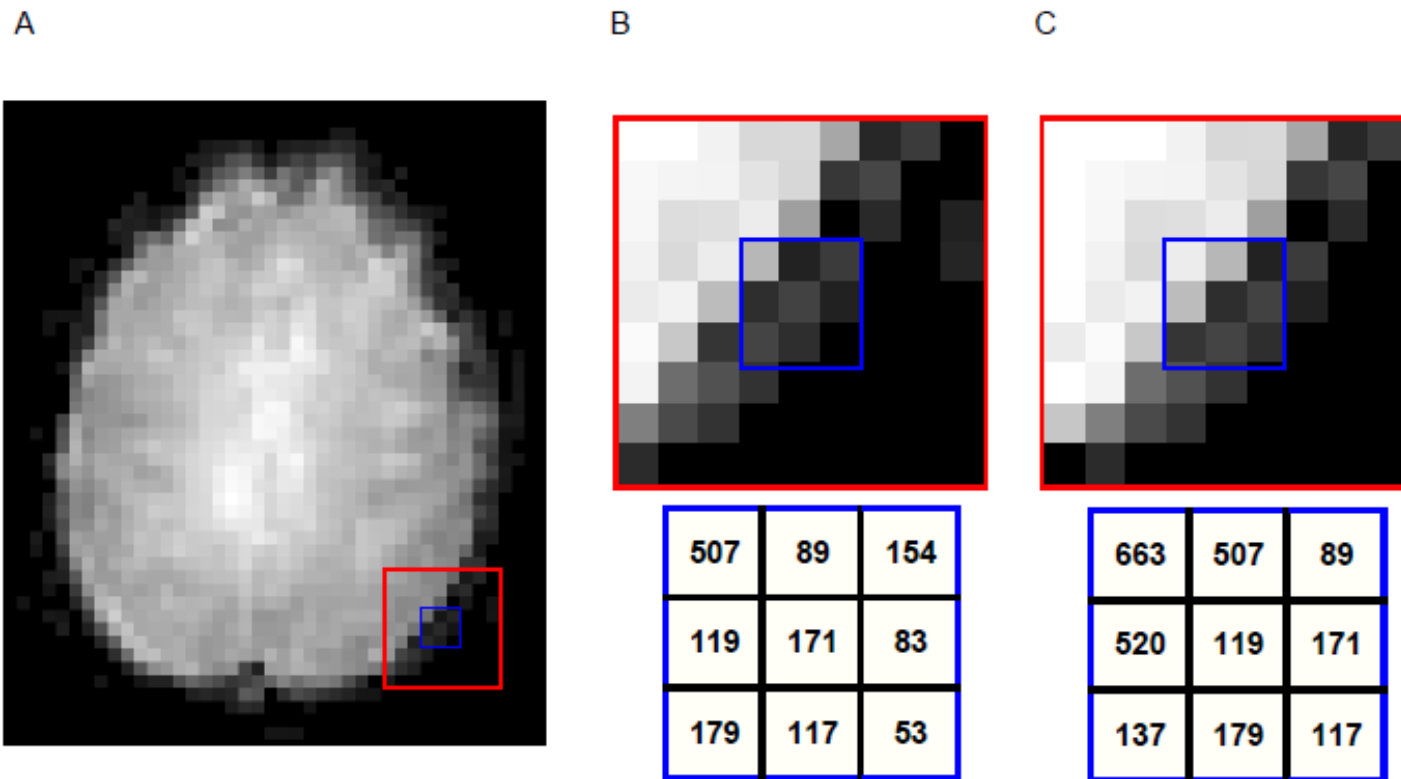
Motion Correction

- “Please be very, very, very still....”
- Movement that’s 1% of voxel size can induce a 1% signal change, which is sometimes greater than your actual BOLD signal! (say the FSL gods)
- Also, your voxel is no longer the same voxel...

Motion Correction

From Monti, UCLA NITP 2011

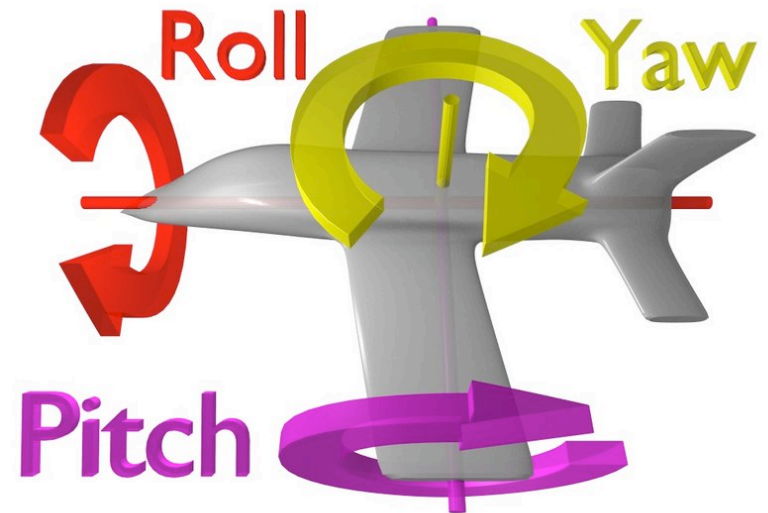
Subject Motion



Huettel et al. Functional Magnetic Resonance Imaging

Motion Correction

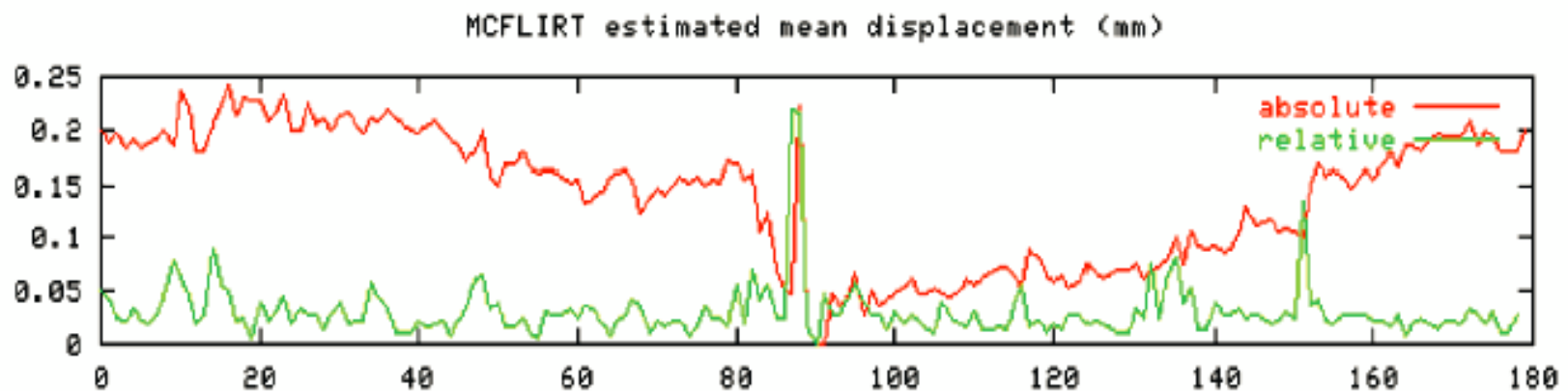
- Choose a volume to register all the other volumes to (e.g., first, middle, last, mean, standard space; in FSL its middle)
- Within subject: use 6 DOF rigid
- X, y, z, roll, pitch, yaw
- Realigns to reference to minimize variance



Motion Correction

- Output (from FSL):

Summary of total motion (**relative** and **absolute**)

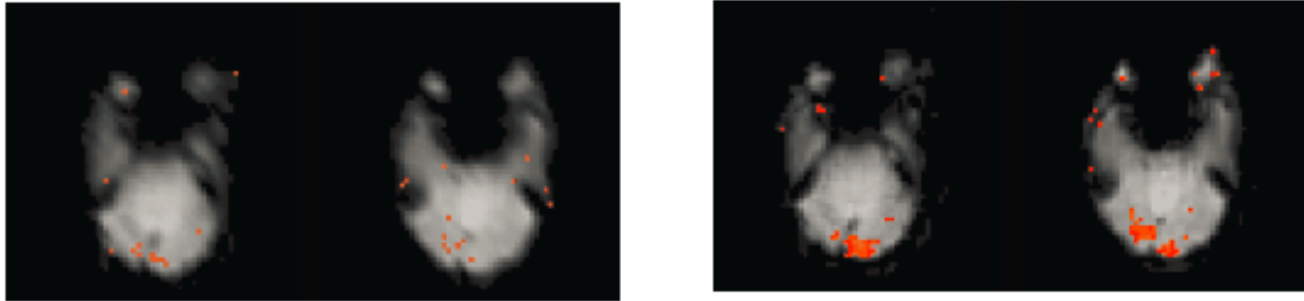


- Sudden spikes? Motion > 1 voxel size? Or .5?
- Model these 6 regressors!

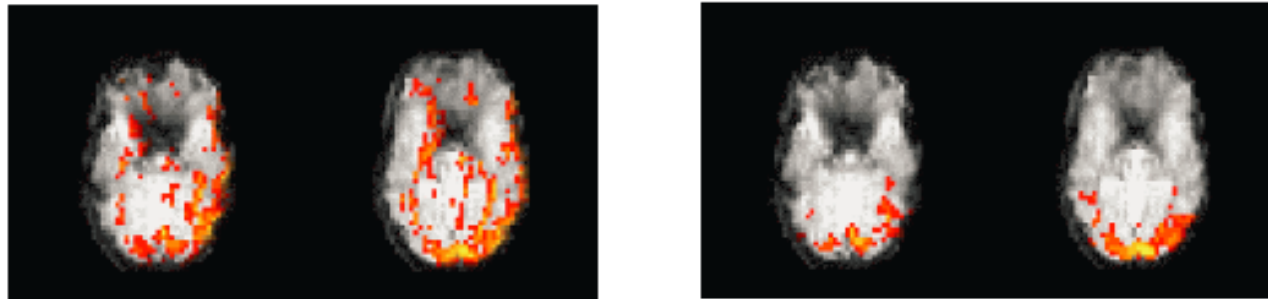
Motion Correction

- Results (from FSL):

Uncorrelated Motion



Stimulus Correlated Motion



Without MC

With MC

Motion Correction

- Other notes:
 - Can prevent with training, make subjects feel relaxed (saying “don’t move!” usually makes them more nervous!)
 - Give them a minute to get settled and comfortable
 - Make sure task isn’t correlated w/ motion
 - Too much motion? Throw out run/subject.. ☹️

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Slice-Timing Correction

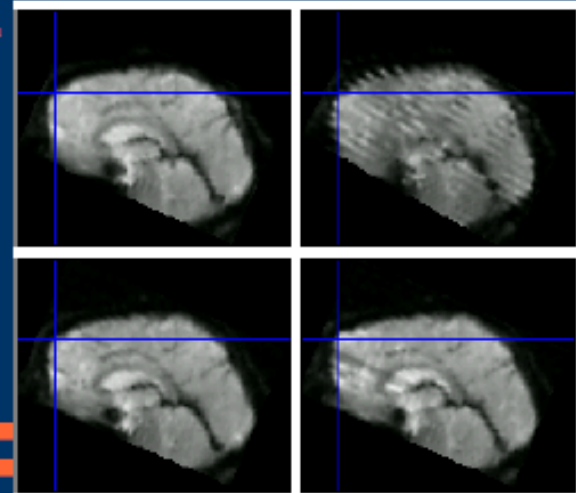
- Scanners (like our Siemens 3T) may acquire slices in an interleaved fashion (e.g., 0,2,4..1,3,5) to avoid contaminating neighboring slice
- Can correct and put slices back in order
- Current consensus is no need (FSL, Monti)
 - It's another interpolation (the less, the better)
 - It doesn't help that much, could make things worse
 - **Instead, add TEMPORAL DERIVATIVE to model**

Slice-Timing Correction

From Monti, UCLA NITP 2011

Slice timing correction

- Most people now suggest not to do it
 - i. Not all that helpful & requires interpolation
 - ii. It may worsen artefacts (e.g., smearing spikes)
 - iii. Interacts in unpredictable ways with motion correction
 - iv. We spatially smooth across proximal slices
 - v. Mismatching TR and task
 - vi. Include temporal derivative of HRF
- What order? Ascending, descending, contiguous, interleaved.

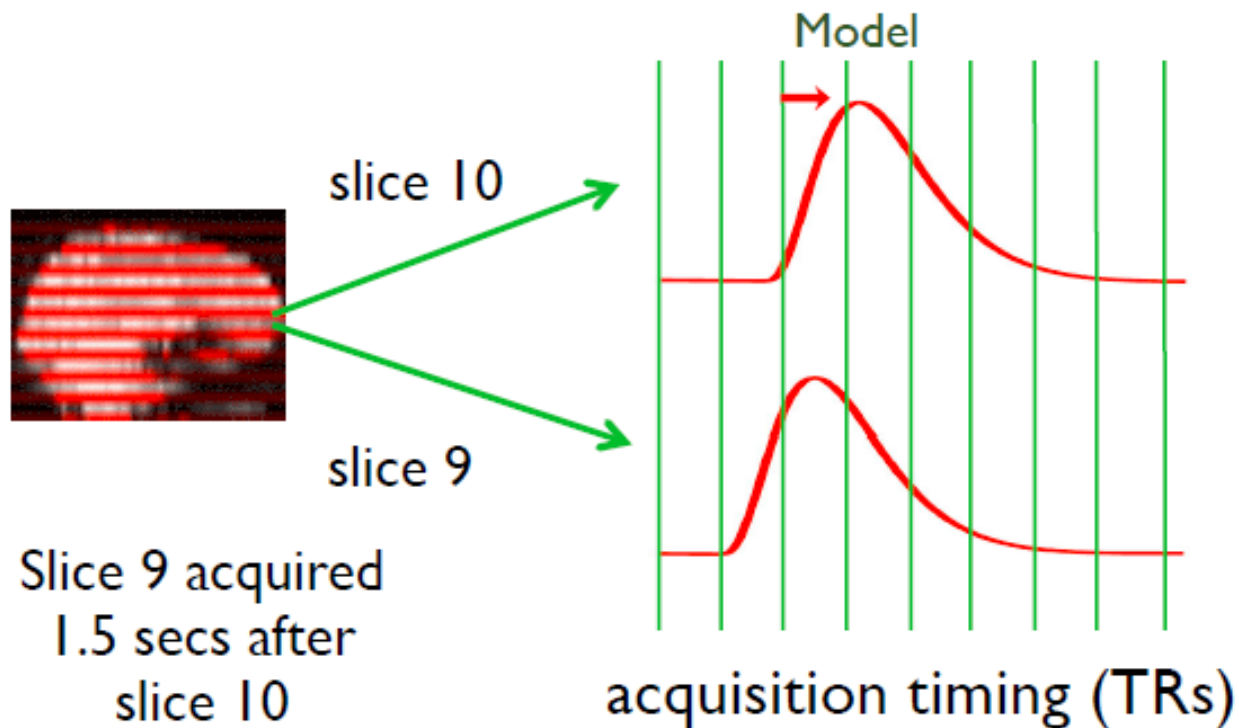


Slice-Timing Correction



Slice Timing

Alternatively, can get consistency by
shifting the **model**



Slice-Timing Correction



Slice Timing

One way to
shift the model
is to use the
**temporal
derivative in
the GLM**

Based on
Taylor approx:
 $m(t+a) = m(t) + a.m'(t)$

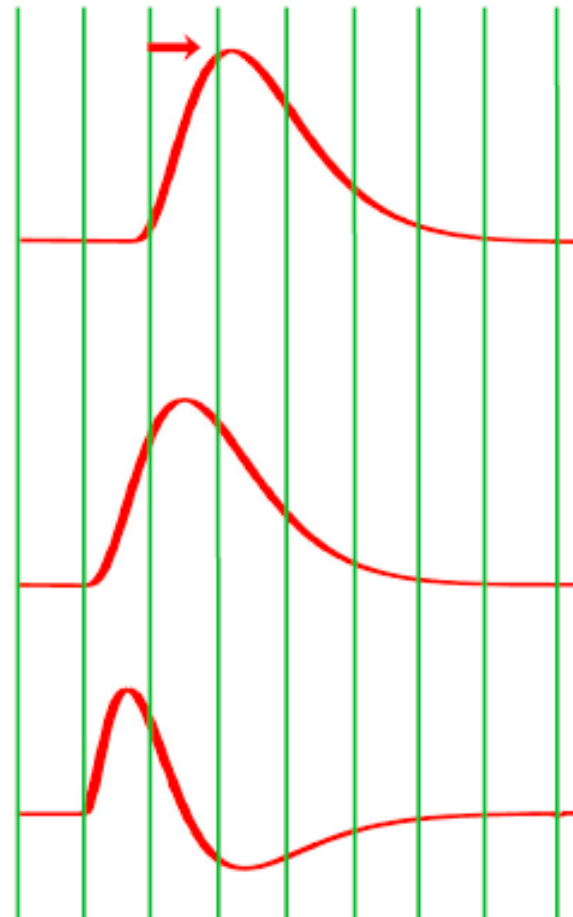
Shifted
Model

=

Original
Model

-

Temporal
Derivative



Slice-Timing Correction

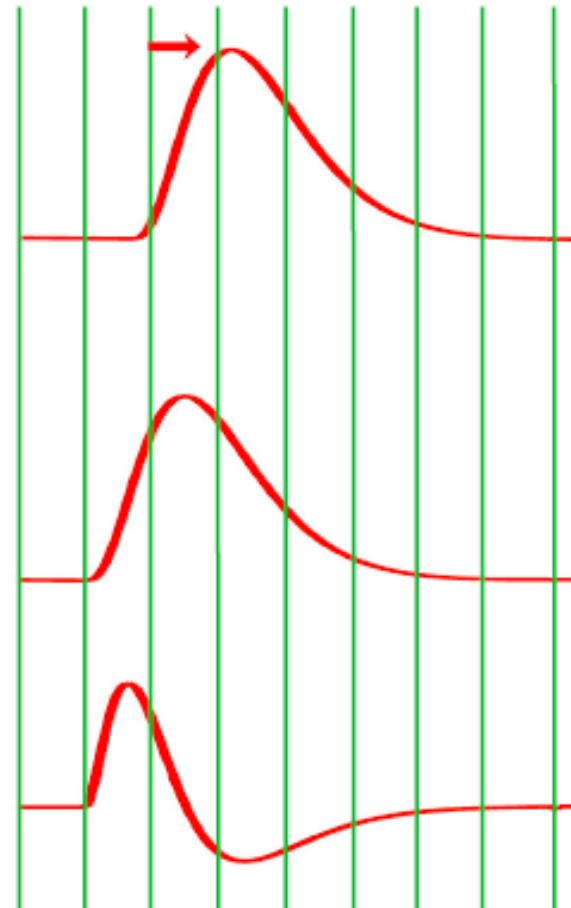


Slice Timing

Shifting the model also accounts for **variations in the HRF delay**

- as the HRF is known to vary between subjects, sessions, etc.

This is the recommended solution for slice timing



Preprocessing Steps

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- **Spatial filtering**
- Temporal filtering
- Global intensity normalization
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Spatial Filtering (Smoothing)

- Averages one voxel's values with its neighbors
- Minimum “smoothness” needed if using Gaussian random field theory (FSL)
- PROS: can increase SNR by decreasing variance
- CONS: may reduce signal if small activations
 - If you expect this, use a smaller kernel
 - Also reduces spatial resolution

Spatial Filtering (Smoothing)

- Gaussian Full Width Half Maximum (FWHM) kernel (from FSL tutorial)

Spatial filtering done by a 3D convolution with a Gaussian (cf. 1D convolution with HRF for model)

Each voxel intensity is replaced by a *weighted* average of neighbouring intensities

A Gaussian function in 3D specifies weightings and neighbourhood size

Weights

0.1	0.3	0.4	0.3	0.1
0.3	0.6	0.8	0.6	0.3
0.4	0.8	1.0	0.8	0.4
0.3	0.6	0.8	0.6	0.3
0.1	0.3	0.4	0.3	0.1

↔ FWHM

Specify amount by Full Width Half Maximum (FWHM)
= distance between 0.5 values

Spatial Filtering (Smoothing)

- On average, a kernel of 5 mm is adequate
 - Pick something about 2.5x FWHM
 - Can go up to 10-15 mm if expecting large activations
 - Or not use (other options for thresholding)

Preprocessing Steps

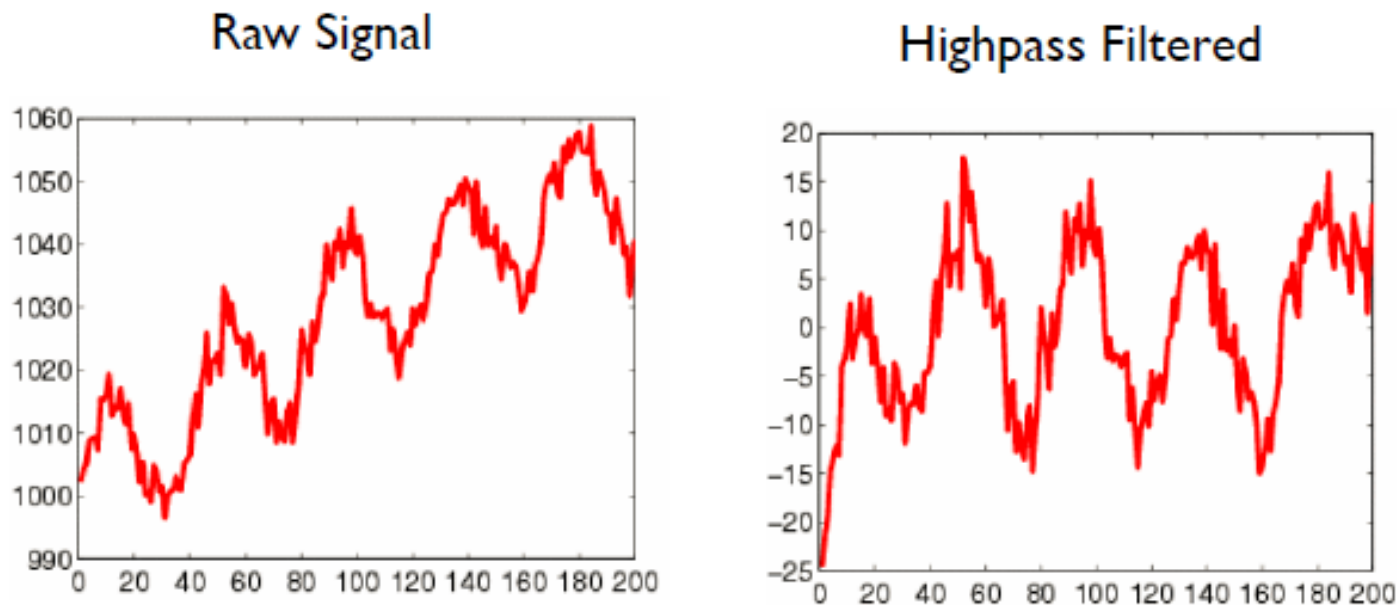
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Temporal Filtering

- Temporal noise:
 - Temporal drift from scanner; physiological cycles (cardiac, respiratory)
- These can mask your actual signal!
- High pass filter lets high frequencies through
- Low pass filter lets low frequencies through (removes high-frequency fluctuations) but can remove signal, esp if event-related.
- Usually just use high pass – FSL uses prewhitening to avoid low-pass

Temporal Filtering

- Use the high-pass filter to remove low-frequency (e.g., long, slow) noise (from FSL):



Removes low frequency signals, including linear trend

Temporal Filtering

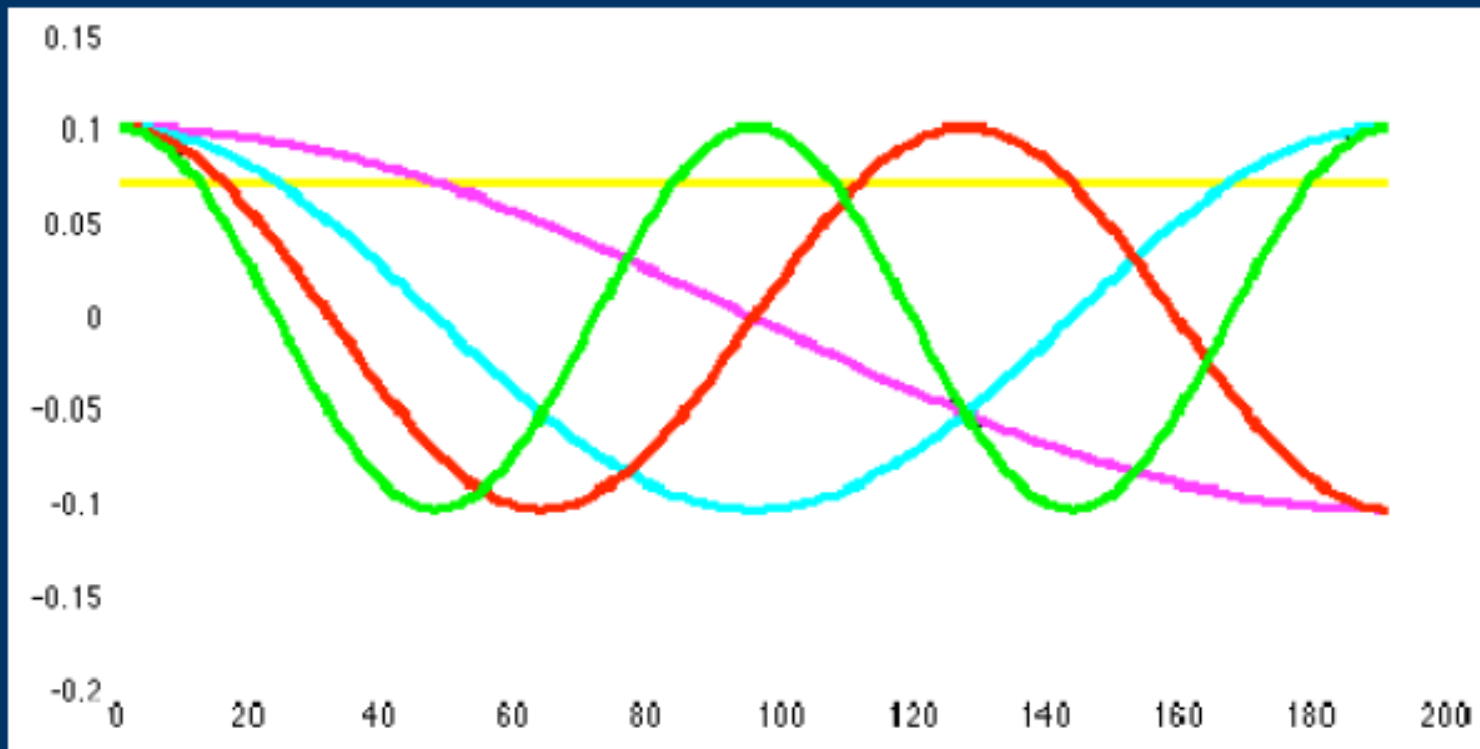
- Generally a HP Filter of 100 s is adequate
 - Change if your model is different (eg very long event)
- SPM: models low drifts to catch variance; uses cosine basis set
- FSL: removes low drifts and then convolves signal with gaussian-weighted running line

Temporal Filtering

From Monti, UCLA NITP 2011

HP Filtering Strategy I: SPM

- Model low drifts to “soak up” their variance (using a discrete cosine transform basis set).

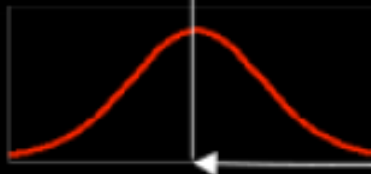
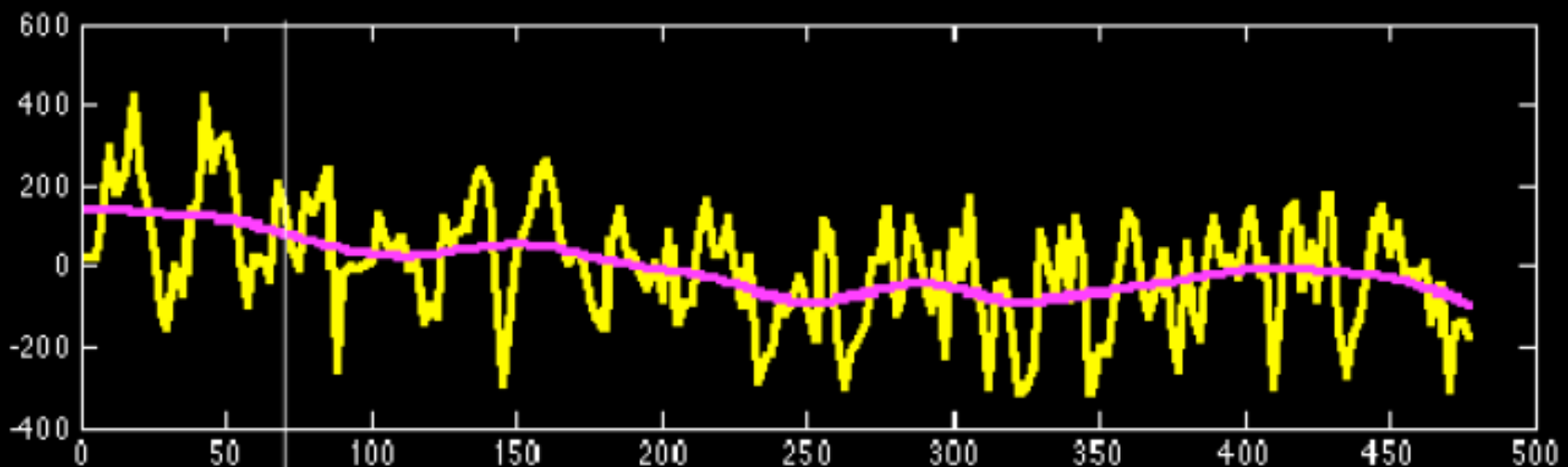


Temporal Filtering

From Monti, UCLA NITP 2011

HP Filtering Strategy II: FSL

- Remove low drifts from the signal:
 - i. Fit a Gaussian-weighted running line



Fit at time t is a weighted average of data around t

Preprocessing Steps

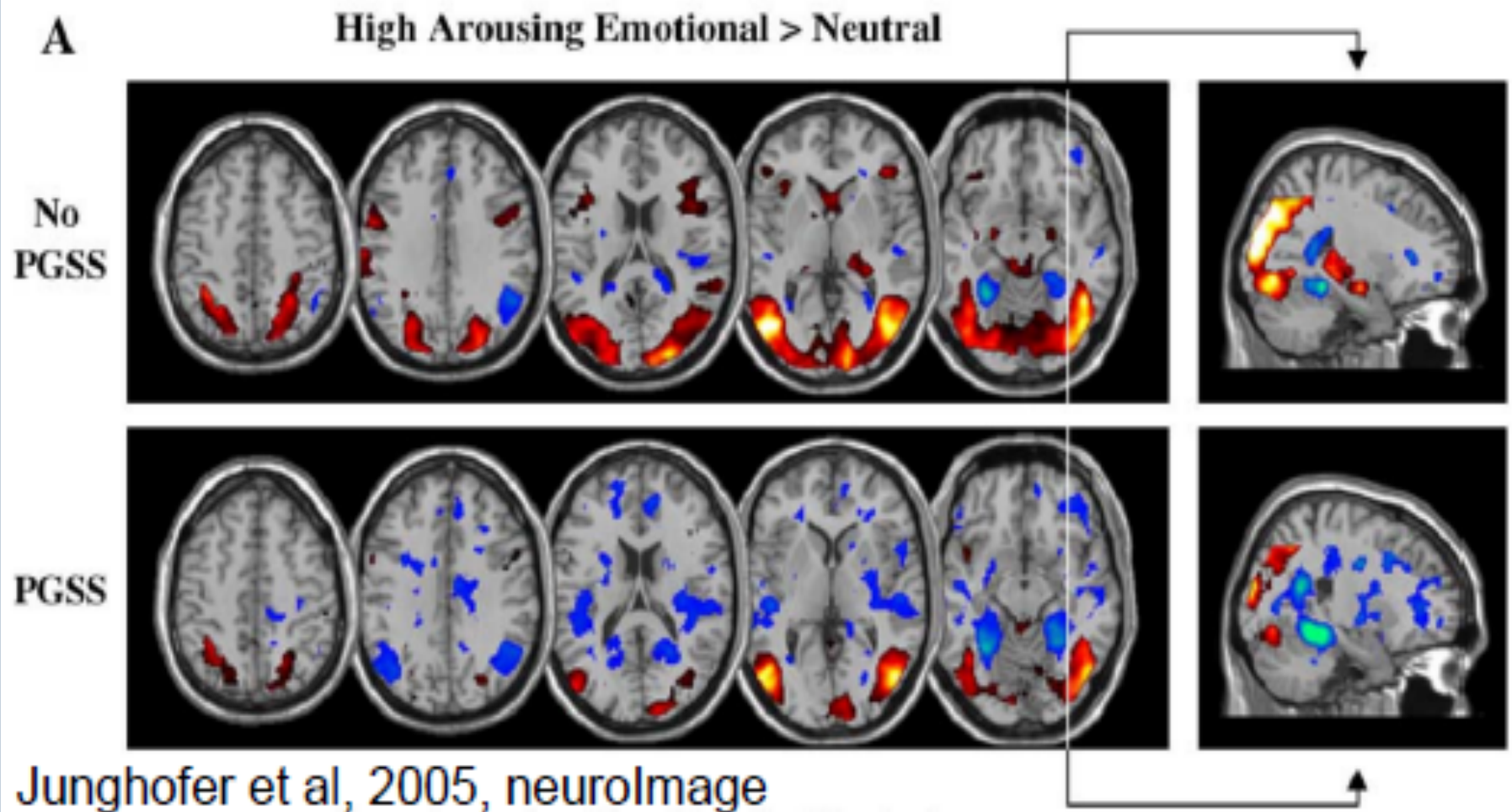
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- **Global intensity normalization**
- Registration/Normalizing (technically post-preprocessing)

Global Intensity Normalization

- From Monti 2011 slides:
- GOOD:
 - Between-sessions (grand mean scaling) so you are comparing all runs by a single factor (can be arbitrary) centered on the same mean
- BAD:
 - Within-session (global scaling), forces each VOLUME to have same mean intensity, that's silly

Global Intensity Normalization

- Don't do global scaling!



Preprocessing Steps

- Pre-Preprocessing
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- ****ICA Denoising**
- Registration/Normalizing (technically post-preprocessing)

ICA Denoising

- Manually remove noise components
- Feed raw data into MELODIC (FSL; Multivariate Exploratory Linear Optimised Decomposition into Independent Components)
- Model free – pulls out all the “components”
- Can manually go through components and identify noise! [demo]
- See Kelly 2010 JNeurosci Methods for details on what to remove
- Especially good if resting state data; or odd movements! Bad if noise relates to signal...

Registration/Normalization

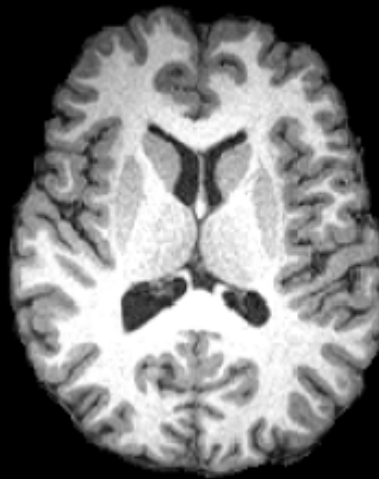
From Monti, UCLA NITP 2011

Registration

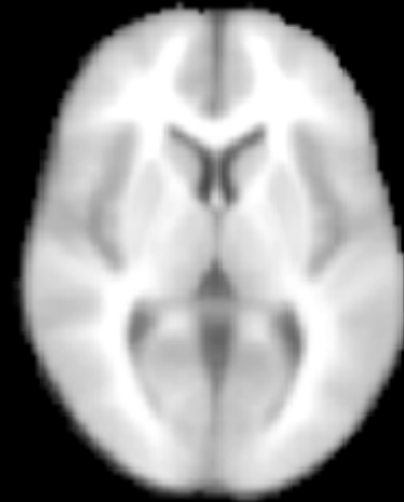
Functional



Anatomical



Template



Registration

- Allows comparing between individuals by mapping all to a template brain (from Monti 2011 slides)
- Rigid body:
 - 6 DOF – 3 rotations, 3 translations, typically within subj
 - 7 DOF – add global scaling, good for func to anatomical
- Affine:
 - 12 DOF – 3 rotations, 3 translations, 3 scalings, 3 skewings
 - Subject to template rotation
- Non-linear:
 - Represented by a deformation field (not a matrix)

Registration

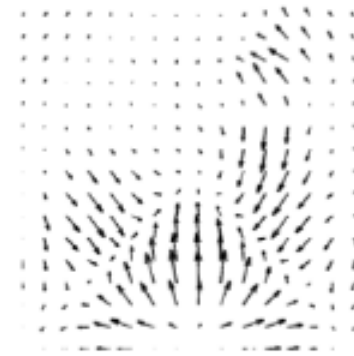
From Monti, UCLA NITP 2011

- Non-linear (> 12dof)
 - Can be local
 - Can be constrained (e.g., regularization, topology preservation)

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

An affine transformation is represented by these 12 numbers.

This matrix multiplies coordinate vectors to define the transformed coordinates.



A non-linear transformation is represented by a **deformation field**.

Registration

- Re: Non-linear transformations
- See FSL specifically on registration and Monti for more detailed information (FSL FLIRT, FNIRT)
- SPM DARTEL is also good (so I hear)
- Lots of options overall!

Registration

- When registering, you have to interpolate again – what to do with new empty spaces?
 - Nearest neighbor? Trilinear (4 neighbors)? Sinc (10)?
- FSL: analyzes in native space, then normalizes
- SPM: normalizes first, then GLM, decreases variance & SD
- Can account for some differences in results!

Questions/Discussion?

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