



### DECLARATION OF CONFLICT OF INTEREST OR RELATIONSHIP

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I have no conflicts of interest to disclose with regard to the subject matter of this presentation.



# Artifacts in fMRI

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#### What I can tell you about:

PET imaging
 Radionuclide production
 PET data analysis
 fMRI data analysis
 Coregistration
 Morphometry measures



Not so much: <u>MRI physic</u>s, pulse sequences, data acquisition

## What is an Artifact?

#### To neuroscientists, this brain looks funny.



#### To Wisconsin Badger fans, the brain obscures the logo.

An artifact is anything that prevents you from seeing what you want to see.

# What is the Signal?



# What is the goal?



change

# **Big-picture goal**



## **Artifact Sources**

= effect variance

#### Physiological

cardiac, respiratory motion 🧶 subject movement poor intersubject alignment B-field susceptibility (dropout) foreign objects Scanner **SNR** B-field inhomogeneity detector coil sensitivity reconstruction tradeoffs slice-timing (2D) effects

## A Scientific Experiment



#### Before

- measure twice, cut once
- During
  - look at your data. with your eyes.
- After
  - pre-processing: modeling and removing variancedata analysis: modeling and discovering signal

### The single biggest artifact source











### **Being Clever Ahead of TIme**

#### Before

🔵 Learn a package provides an orderly introduction to analysis. Software programming skills File format conversion Experimental paradigm event-related paradigms can isolate stimulus, HR Acquisition sequence Collect enough data multispectral anatomicals (T1, T2, PD, ...) field map physiological (cardiac, respiratory, skin conductance, eye tracking) estimator for HRF (e.g. motor response)

# **Post-hoc Corrections**

Some folks call it image processing...
spatial registration
field map correction
slice time correction
motion correction
smoothing- spatial, temporal

These corrections are ignorant of the experimental model.

### EPI Dropout (susceptibility) artifact



phase dispersion => signal loss

## EPI susceptibility artifact

clever ahead of time: optimize acquisition parameters



TE = 30 ms 4mm thick slices



TE = 20 ms 2mm thick slices

courtesy of Andy Alexander





### EPI registration: indirect approach

![](_page_16_Picture_1.jpeg)

![](_page_17_Figure_0.jpeg)

## Off to a bad start

🖤 Spamalize Window 1

![](_page_18_Picture_2.jpeg)

frame # (TR)

# **Field Map Correction**

The magnetic field is not uniform, leading to misplaced signal, since recording/reconstruction assumes a uniform gradient.
 3T => worse

 Nonuniformities are object-dependent.
 Largest source of B<sub>0</sub> inhomogeneity: air-tissue interfaces => susceptibility differences

Different acquisition sequences may have different affect.
 EPI data (Gradient Echo) most affected (fMRI, DTI)
 Spin Echo affected little

Luckily, this can be modeled and corrected...

But, a seperate measurement is required.

2 scans with different TEs (e.g. 8, 11 msec), minimal distortion.

## **B-Field Distortion**

#### The Cause: EPI Phase Error Accumulation

![](_page_20_Figure_2.jpeg)

For a linear field offset, phase error accumulates linearly.Cummulative phase errors cause a shift in position

### **Distortion (Fieldmap) Correction**

![](_page_21_Picture_1.jpeg)

![](_page_21_Picture_2.jpeg)

TE=10ms

TE=7ms

Peter Jezzard and Stuart Clare, "Sources of distortion in functional MRI data", *Human Brain Mapping*, 8:80-85, 1995.

Strategy: Collect two gradient

Phase-difference proportional

to the pixel-offset along the

Resample the EPI image to

echo images at short but

different echo times.

Unwrap the phase.

phase-encode axis.

correct for the offset.

courtesy of John Ollinger

### **Field Map Correction**

![](_page_22_Picture_1.jpeg)

# **Field Map Correction**

Magnitude of correction depends on the magnitude of the susceptibility artifact, which depends on the size of the sinuses.

- Larger in men than women.
- Larger in adults than children

Although the magnitude of the correction is often small, it can reduce confounds between gender, age and susceptibility.

It cannot recover missing data from the dropout region.

## **Inhomogeneity** Correction

original

![](_page_24_Picture_2.jpeg)

inhomoger eity corrected

![](_page_24_Picture_4.jpeg)

![](_page_24_Picture_5.jpeg)

![](_page_24_Picture_6.jpeg)

![](_page_24_Picture_7.jpeg)

joint histogram

Difference (~ bias field)

# Inhomogeneity Correction

Increase gradient magnetic field strength.
 Decrease echo time.
 Smaller pixels (better resolution).
 Phase encoding.
 Postprocessing.

![](_page_25_Picture_2.jpeg)

![](_page_25_Picture_3.jpeg)

![](_page_25_Picture_4.jpeg)

![](_page_25_Picture_5.jpeg)

# **Slice-time Correction**

![](_page_26_Picture_1.jpeg)

Make all slices appear to have been acquired at the same time.

A "reference" HRF should look the same in all slices.

![](_page_26_Figure_4.jpeg)

More important for longer TR. Usually best prior to motion correction.

# **Motion Artifacts**

#### Sources

- Subject motion
- Peripheral movement (changes B-field)
- Respiration, cardiac
- Magnitude / Importance
  - a significant fraction of the fMRI signal!
- Model
  - framewise 3D volumetric
    rigid body (translation, rotation)
    - ignores intraframe motion
  - assumes only small movement
- Correlation between motion, stimulus
  - correcting motion can attenuate activation signal
  - motion estimates can be incorporated into GLM

### **Motion Correction: Magnitude**

![](_page_28_Figure_1.jpeg)

## Motion Correction

![](_page_29_Picture_1.jpeg)

#### Image realignment

144

cut

move

resize

text

edit

 1
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0000.img

 2
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0002.img

 3
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0002.img

 4
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0003.img

 5
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0004.img

 6
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0005.img

 7
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0005.img

 8
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0006.img

 9
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0008.img

 10
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0009.img

 11
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_0010.img

 12
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_001.img

 12
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_001.img

 12
 /scratch/1MRI\_data/data/1MRI\_vis\_stim\_001.img

Clear

5PM99 (oakes): 11:50:56 - 27/02/2002

ColorMap

=

Effects

Print

![](_page_29_Figure_4.jpeg)

![](_page_29_Figure_5.jpeg)

### **Motion Correction: Subject Motion**

![](_page_30_Picture_1.jpeg)

# Combining scan runs

![](_page_31_Figure_1.jpeg)

### Motion Artifact: Typical pattern

![](_page_32_Figure_1.jpeg)

## **Motion Correlated with Stimulus**

![](_page_33_Figure_1.jpeg)

![](_page_33_Figure_2.jpeg)

### **Motion Parameters:**

#### How can we use this information for Good?

Apply motion correction: reslice each 3D volume
 "standard" aproach
 loss of sensitivity if motion correlated with activation

#### Use parameters as covariates in GLM

- 🧔 may increase sensitivity
- 🧔 more flexible data analysis
- small loss of degrees of freedom in GLM

### **Motion Parameters as GLM Covariates**

#### Block design:

if motion correlated with stimulus, "standard" reslice best
Event-related design:

so motion parameters in GLM usually work well.

![](_page_35_Figure_4.jpeg)

T Johnstone et al., "Motion Correction and the Use of Motion Covariates in Multiple-Subject fMRI Analysis", Hum Brain Mapp 27:779 -788, 2006.

## **Motion Correction at Work**

![](_page_36_Figure_1.jpeg)

![](_page_36_Picture_2.jpeg)

![](_page_36_Picture_3.jpeg)

# Smoothing - spatial

Increase S:N
 Objects similar in size to smoothing kernal emphasized
 matched filter theorem
 Manage imperfect registration
 Fulfill "Gaussian random field" assumption

![](_page_37_Picture_2.jpeg)

![](_page_37_Picture_3.jpeg)

![](_page_38_Figure_0.jpeg)

# Smoothing – when?

#### EPI time series

- Spatial smoothing removes small (uninteresting) clusters
- Mild temporal smoothing beneficial:
  - 🌌 trend removal
  - 🌌 high-pass filter
  - most versatile if left until GLM stage
- Parameter estimate (cluster) maps
  - more versatile for analysis
  - changes fitted results

## Group-wise artifacts

Multi-subject or 2cnd-level analysis

- subject-to-subject variance dominates
- most analysis assume:
  - similar within-subject variance
  - similar data acquisition, analysis

![](_page_40_Figure_6.jpeg)

from work by Pradeep Chilakamarri

# **Software Recommendations**

#### fMRI - specific

- AFNI, FSL, SPM have largest market share, similar results
- Two-step procedure for selecting software:
  - Find the smartest person in your lab.
  - Use what they use.

#### Scripting

bash

fast, easy, universal

idiosyncratic, hard to distribute

Python

versatile, extensible, good for distribution

mostly universal, moderate learning curve

Matlab, IDL

sed & awk? You don't need my advice.

Other handy tools: 

http://www.r-project.org/

🤵 fmristat

NiPy

🔘 R

http://www.math.mcgill.ca/keith/fmristat/

http://neuroimaging.scipy.org/