FMRI based on Blood Oxygenation Level Dependence (BOLD) contrast

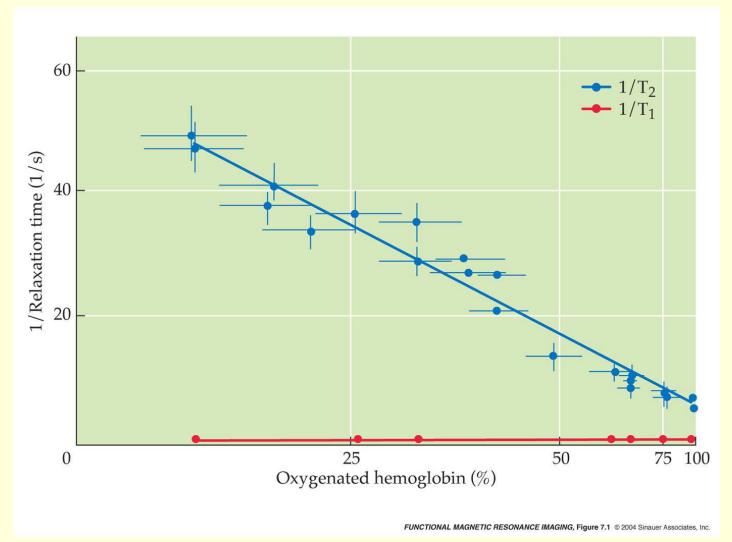
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Contrast Agents in MRI

- Definition: Substances that alter magnetic susceptibility of tissue or blood, leading to changes in MR signal
 - Affects local magnetic homogeneity: decrease in T₂*
- Two types
 - <u>Exogenous</u>: Externally applied, non-biological compounds (e.g., Gd-DTPA)
 - <u>Endogenous</u>: Internally generated biological compound (e.g., deoxyhemoglobin, dHb)

Blood Deoxygenation affects T₂^{*} Decay

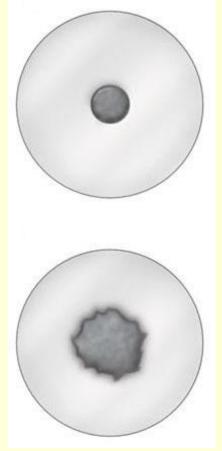


Thulborn et al., 1982

BOLD Endogenous Contrast

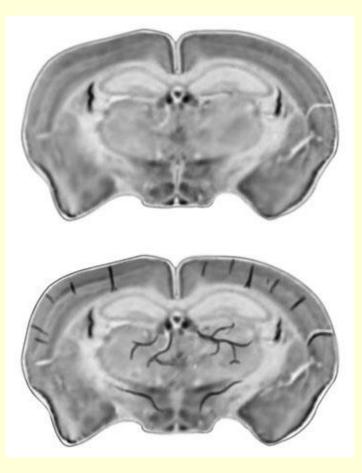
- <u>Blood Oxyenation Level Dependent Contrast</u>
 - Deoxyhemoglobin is paramagnetic
 - Magnetic susceptibility of blood increases linearly with increasing Deoxygenation
- Oxygen is extracted during passage through capillary bed
 - Brain arteries are fully oxygenated
 - During activation Venous (and capillary) blood has increased proportion of Doxyhemoglobin
 - Then oxygen is compensated in veins
 - Difference between oxy and deoxy states becomes greater for veins \rightarrow BOLD sensitive to venous changes

Deoxygenated Blood \rightarrow Signal Loss



Oxygenated blood? No signal loss...

Deoxygenated blood? Signal loss!!!



Images from Huettel, Song & McCarthy, 2004, Functional Magnetic Resonance Imaging

Susceptibility Artifacts

T1-weighted image

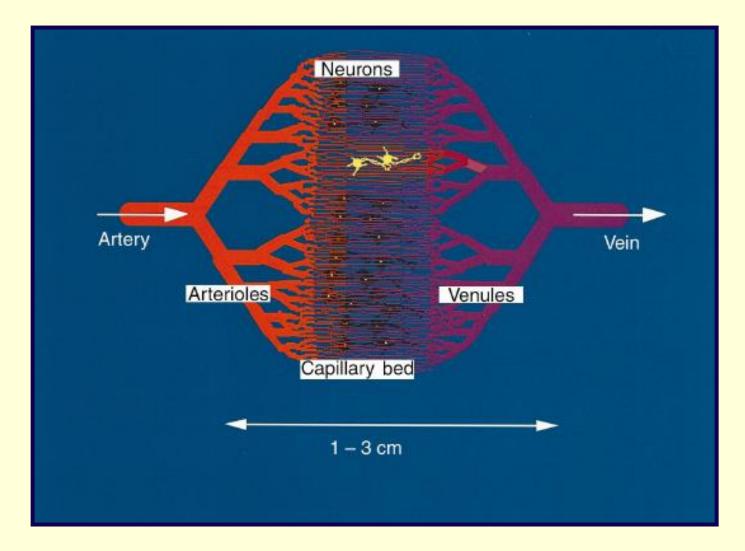
-In T2* images, artifacts occur near junctions between air (sinuses, ear canals) and tissue

Measuring Deoxyhemoglobin

 fMRI measurements are of amount of oxyhemoglobin per voxels in Venus pool

 We assume that amount of <u>deoxygenated</u> <u>hemoglobin</u> in vein (and oxyhemoglobin in later stage) is predictive of neuronal activity

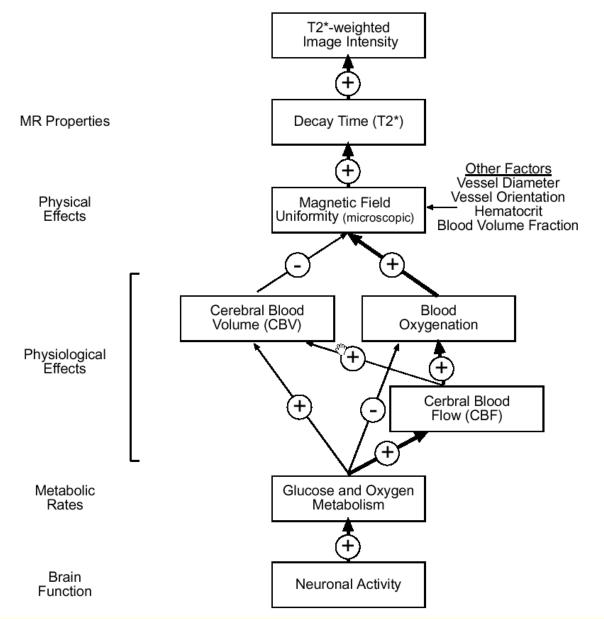
Vasculature



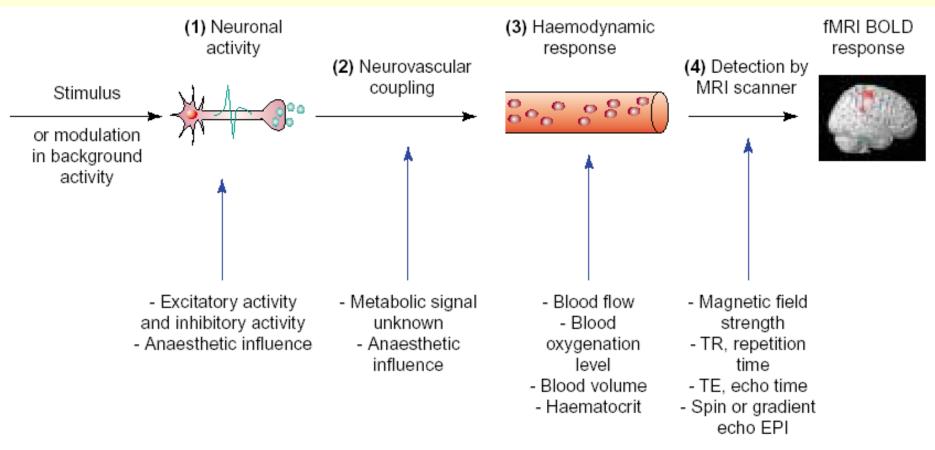
Source: Menon & Kim, TICS

Scott Huettel, Duke University

BOLD signal



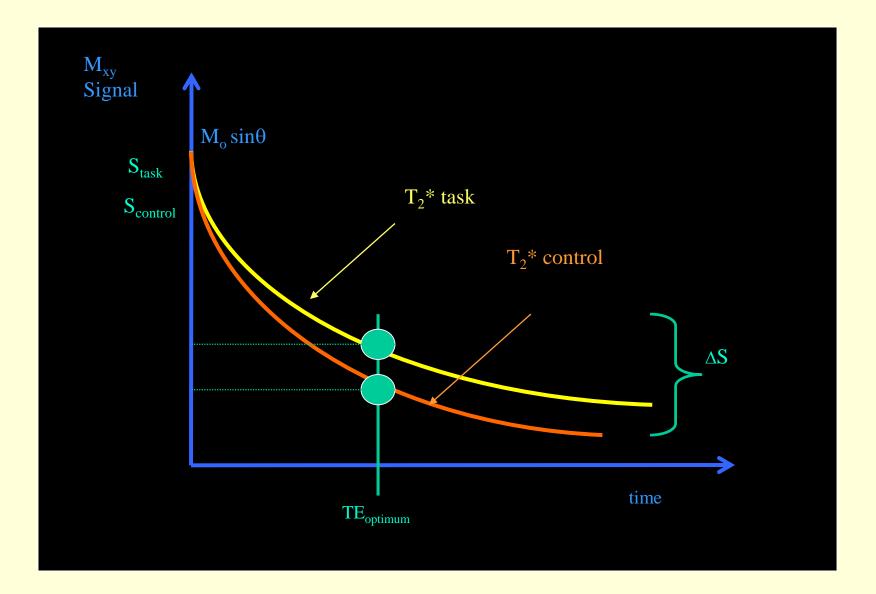
Stimulus to BOLD



TRENDS in Neurosciences

Source: Arthurs & Boniface, 2002, *Trends in Neurosciences*

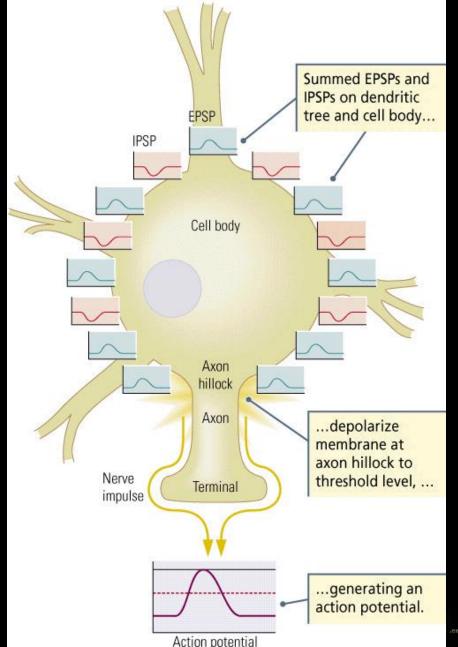
BOLD signal



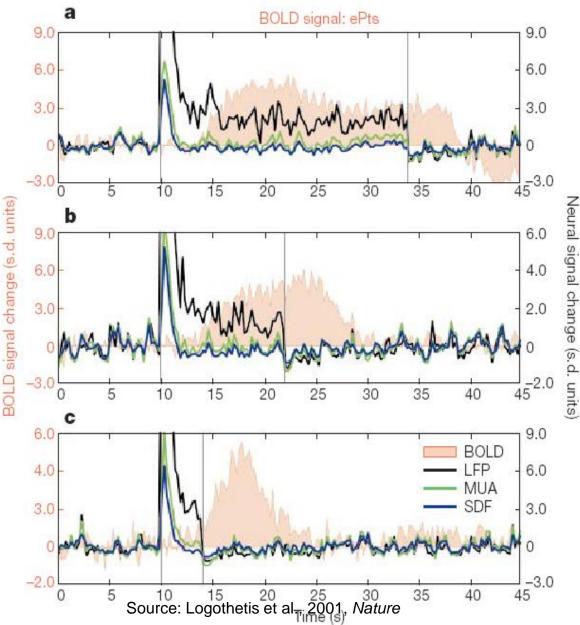
Physiology of BOLD Response (The Hemodynamic Response)

Post-Synaptic Potentials

- The inputs to a neuron (post-synaptic potentials) increase (excitatory PSPs) or decrease (inhibitory PSPs) the membrane voltage
- If the summed PSPs at the axon hillock push the voltage above the threshold, the neuron will fire an action potential



BOLD Correlations



Local Field Potentials (LFP)

- reflect post-synaptic potentials
- similar to what EEG (ERPs) and MEG measure

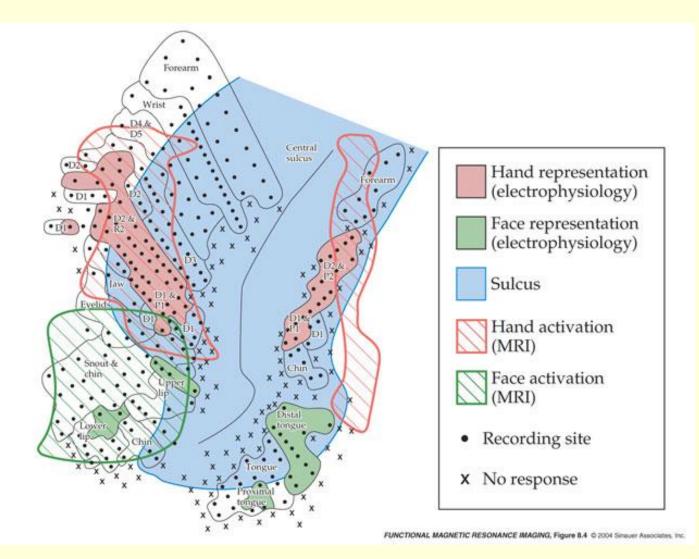
Multi-Unit Activity (MUA)

- reflects action potentials
- similar to what most electrophysiology measures

BOLD activity

- is more closely related to LFPs than MUA
- **SDF=** spike-density function

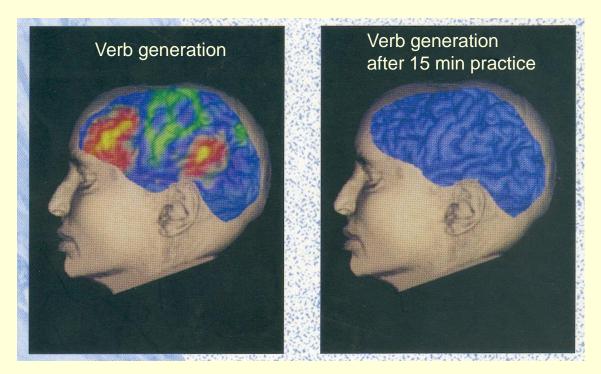
Comparing Electrophysiolgy and BOLD



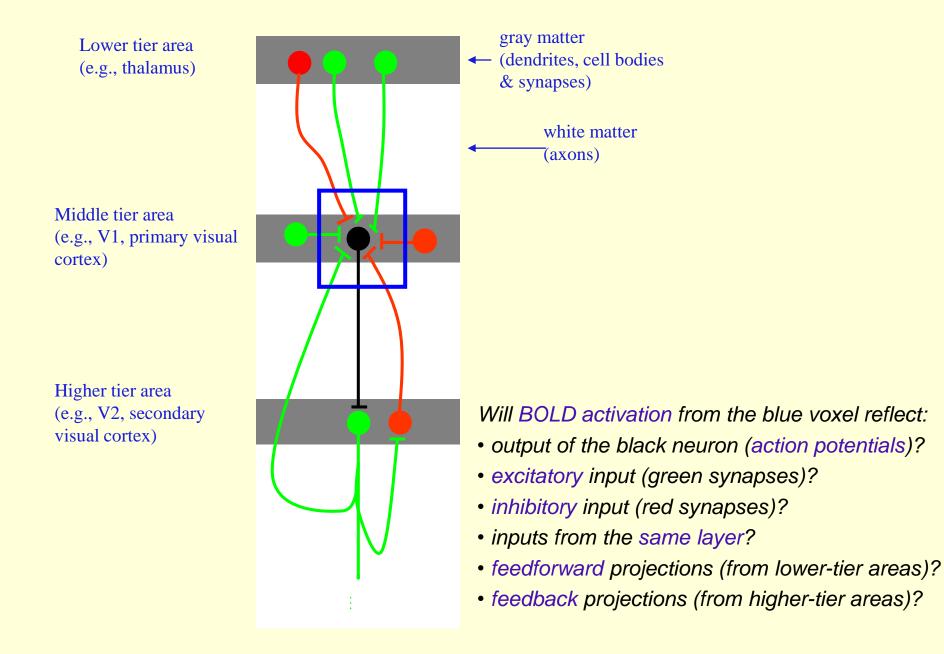
Data Source: Disbrow et al., 2000, *PNAS* Figure Source, Huettel, Song & McCarthy, *Functional Magnetic Resonance Imaging*

fMRI Measures the Population Activity

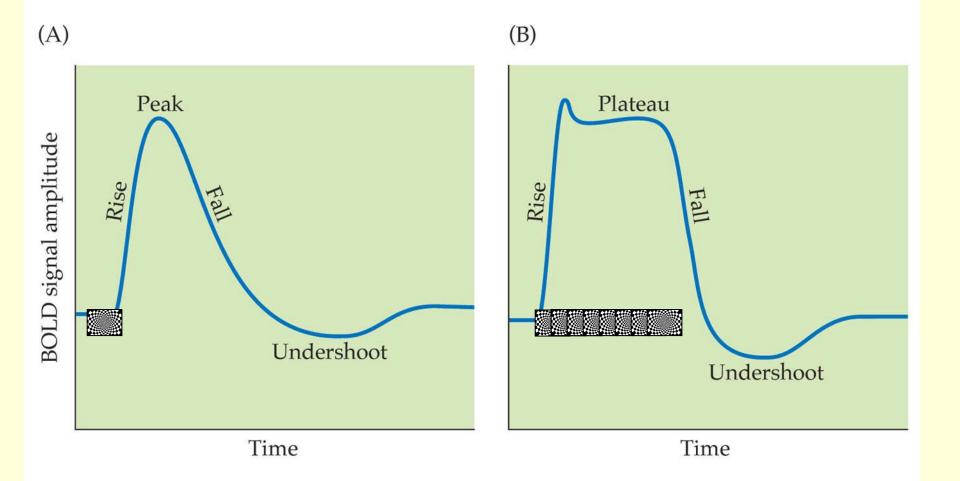
- fMRI may not match single neuron physiology results
- population activity depends on
 - how active the neurons are
 - how many neurons are active

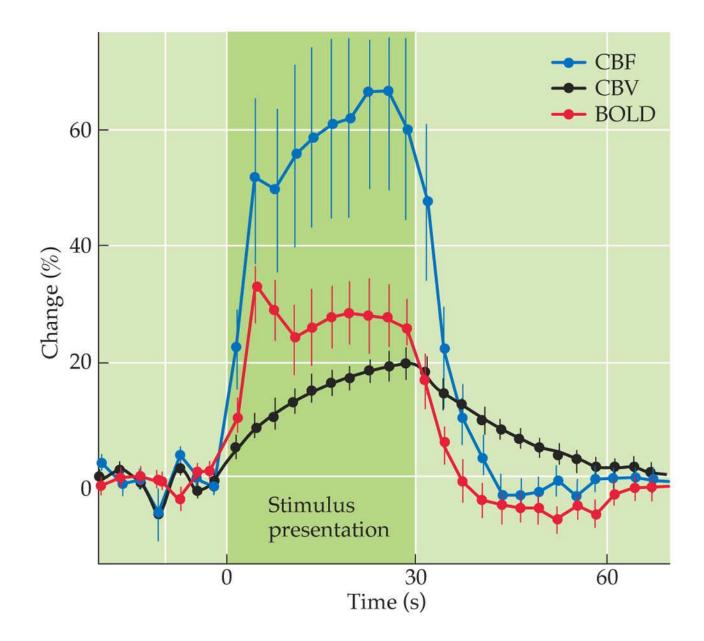


Ideas from: Scannell & Young, 1999, Proc Biol Sci Raichle & Posner, Images of Mind cover image

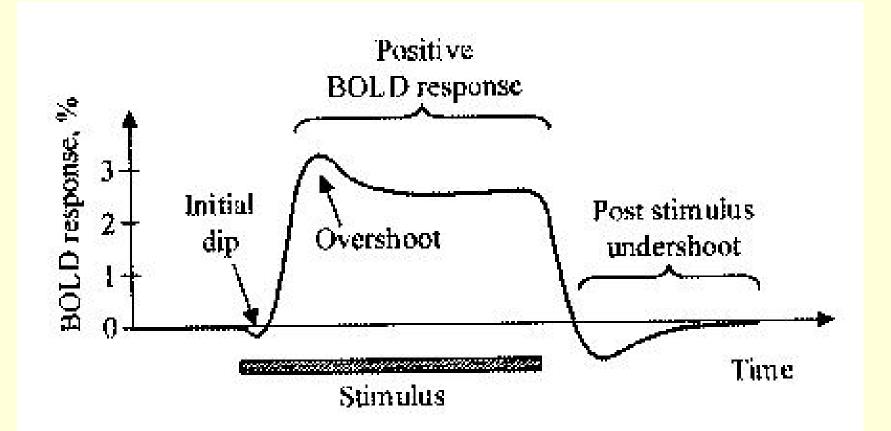


Basic Form of Hemodynamic Response





BOLD Time Course



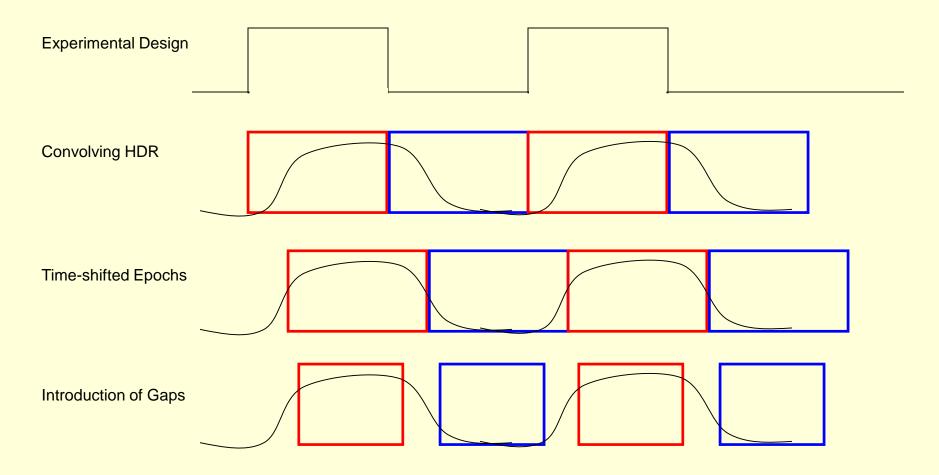
Amplitude of the HDR

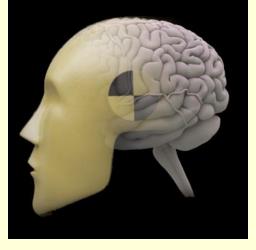
- Peak signal change dependent on:
 - Brain region
 - Task parameters
 - Voxel size
 - Field Strength

Why does the hemodynamic response matter?

- Delay in the hemodynamic response (HDR)
 - Hemodynamic activity lags neuronal activity
- Amplitude of the HDR
- Variability in the HDR
- Linearity of the HDR
- HDR as a relative measure

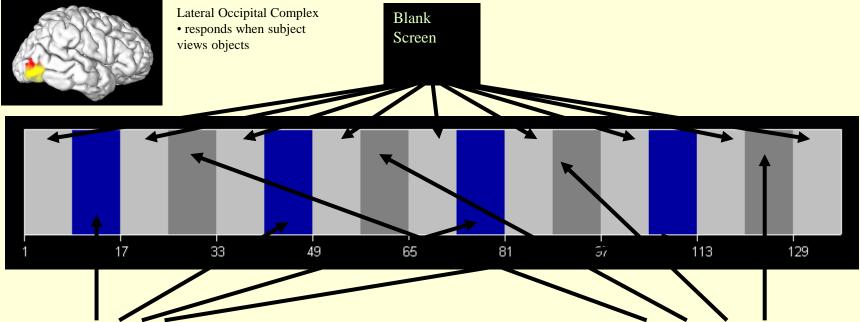
The Hemodynamic Response Lags Neural Activity

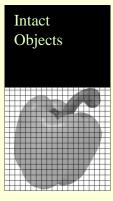




fMRI Analysis with emphasis on the general linear model

A Simple Experiment





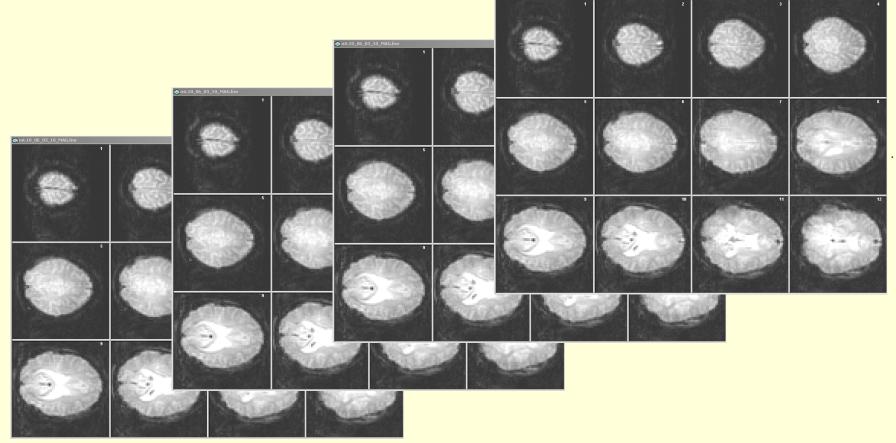


Condition changes every 16 seconds (8 volumes per Block), 17 block One volume (12 slices) every 2 seconds

for 272 seconds (4 minutes, 32 seconds)



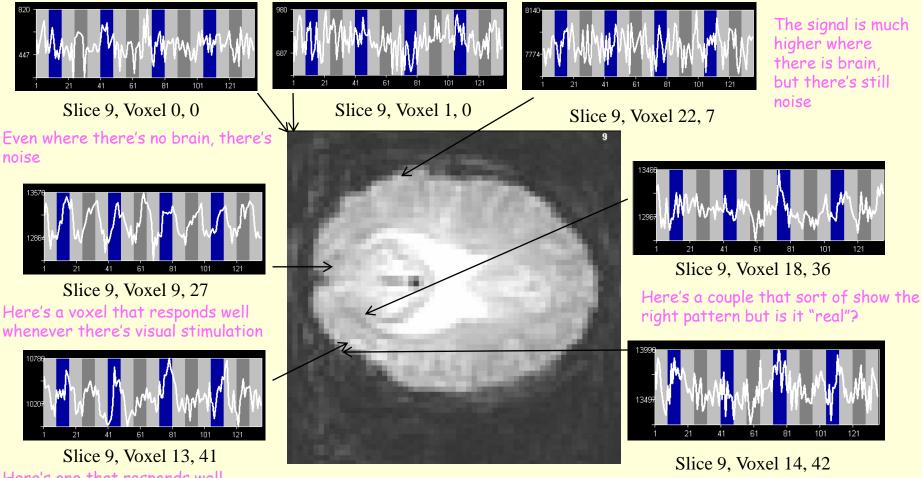
What data do we start with



- 12 slices * 64 voxels x 64 voxels = 49,152 voxels
- Each voxel has 136 time points
- Therefore, for each run, we have 6.7 million data points
- We often have several runs for each experiment

Why do we need stats?

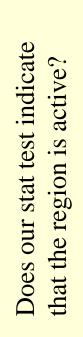
 We could, in principle, analyze data by voxel surfing: move the cursor over different areas and see if any of the time courses look interesting

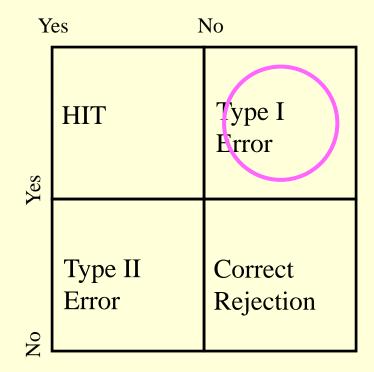


Here's one that responds well whenever there's intact objects

Types of Errors

Is the region truly active?





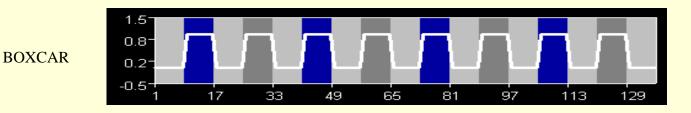
p value: probability of a Type I error

e.g., p <.05

"There is less than a 5% probability that a voxel our stats have declared as "active" is in reality NOT active

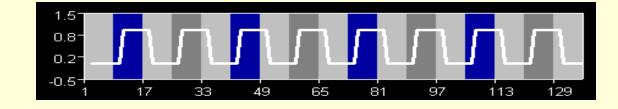
Modeling the Predicted activation

- It takes about 5 sec for the blood to catch up with the brain, therefore we can model the predicted activation in one of two ways:
 - 1. shift the boxcar by approximately 5 seconds (2 images x 2 seconds/image = 4 sec, close enough)
 - 2. convolve the boxcar with the hemodynamic response to model the shape of the true function as well as the delay

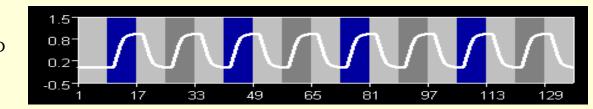


PREDICTED ACTIVATION IN VISUAL AREA

SHIFTED



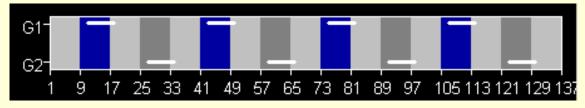




Statistical Approaches

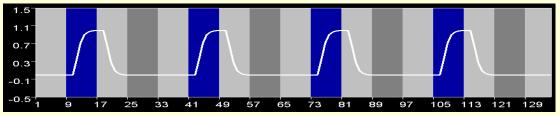
t-tests

compare activation levels between two conditions (eg. Activation and Rest)



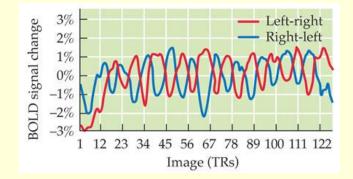
correlations

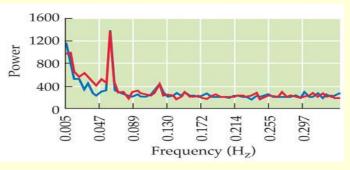
• model activation and see whether any areas show a similar pattern



Fourier analysis

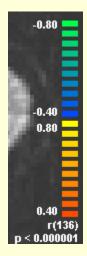
• Do a Fourier analysis to see if there is energy at your paradigm frequency





Fourier analysis images

Effect of Thresholds



r =

49

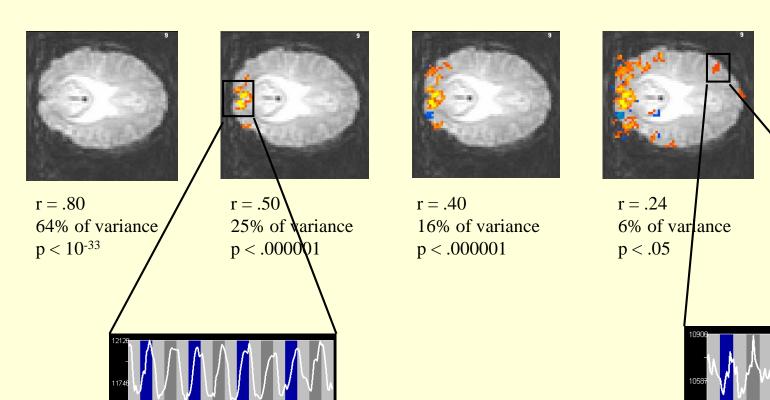
65

p < 1

0% of variance

97

113



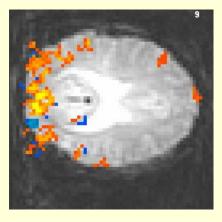
49

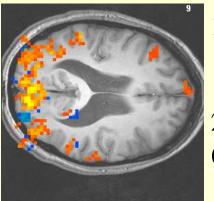
65

Complications

• There are all sorts of statistical problems:

What's wrong with these data?



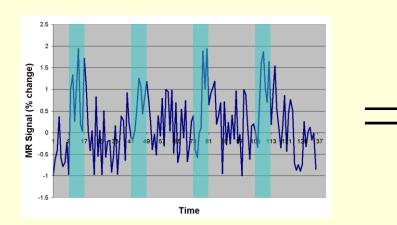


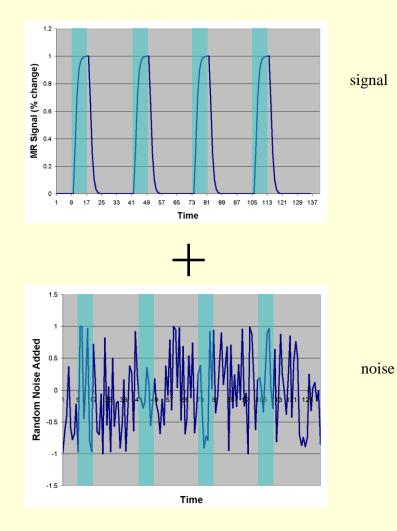
r = .24 6% of variance p < .05 data may be contaminated by artifacts (e.g., head motion, breathing artifacts)

2. "significant" voxels by chance alone.(P=05) * 49,152 = 2457 voxels

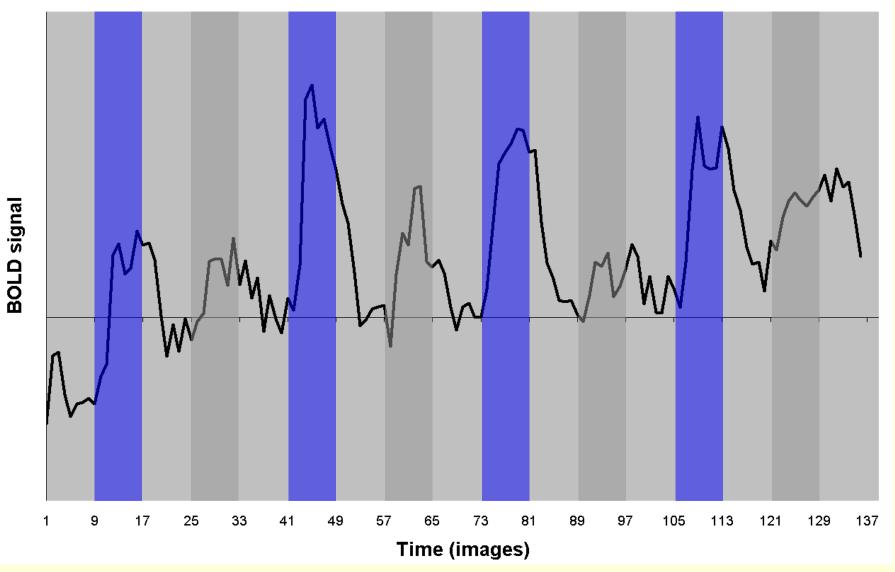
3. many assumptions of statistics are false.
(e.g. adjacent voxels uncorrelated with each other; adjacent time points uncorrelated with one another)

What's real?



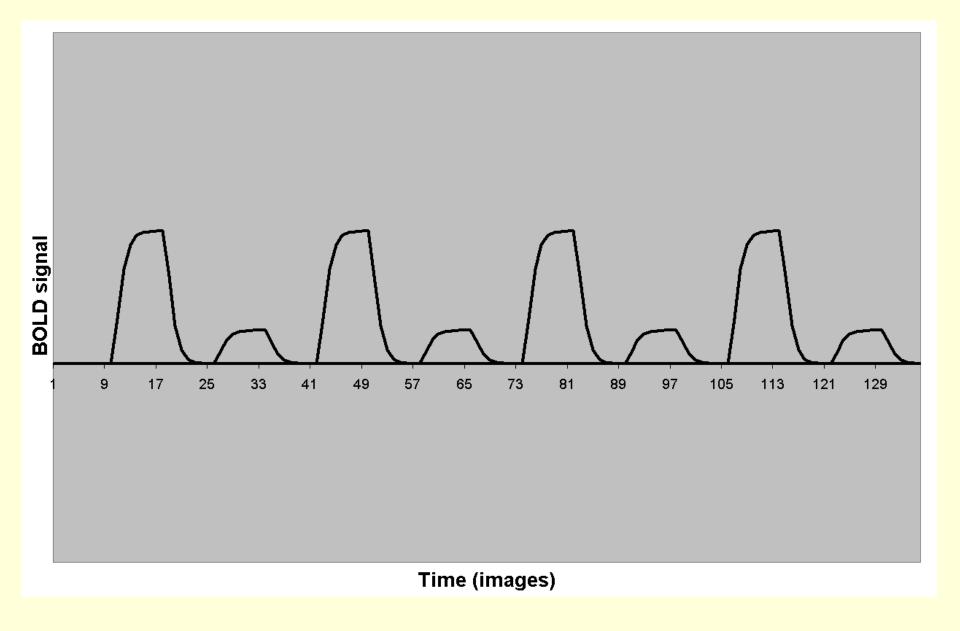


Let's create a time course for one voxel Intact Objects is greater than Scrambled

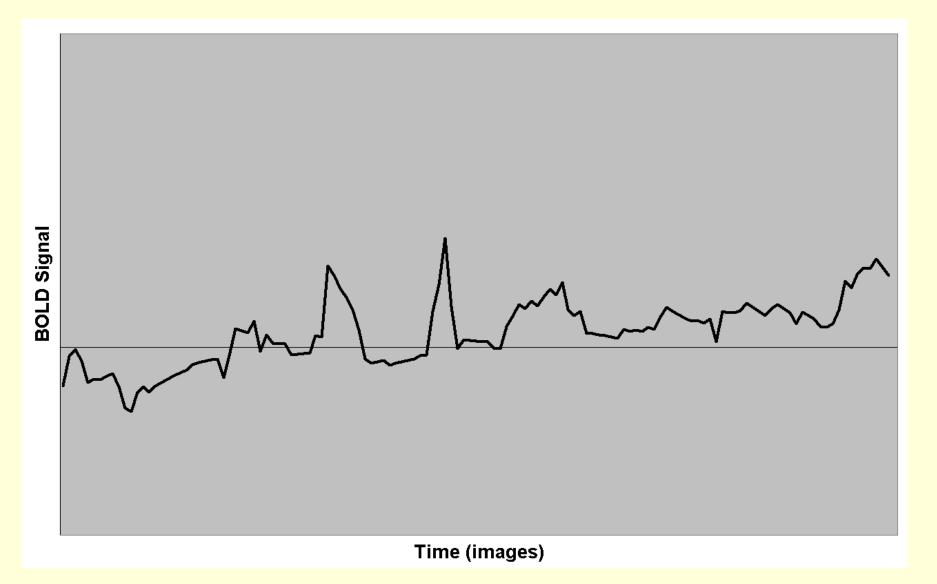


How this signal is build up?

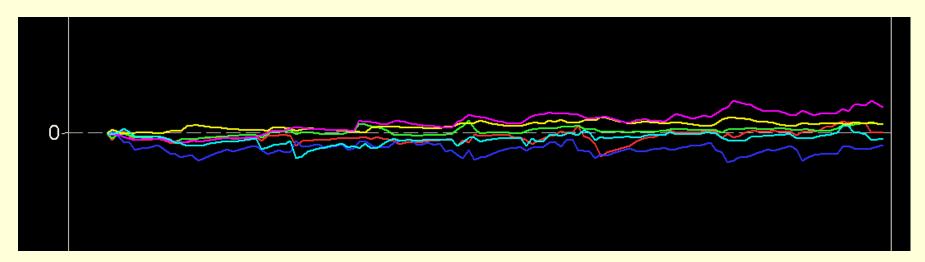
Response to Intact Objects which is 4X greater than Scrambled Objects



Now let's add some variability due to head motion

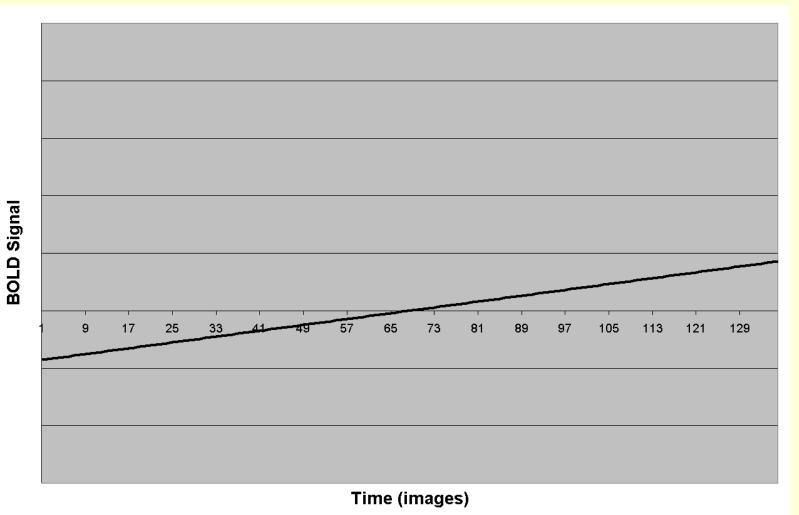


...though really motion is more complex

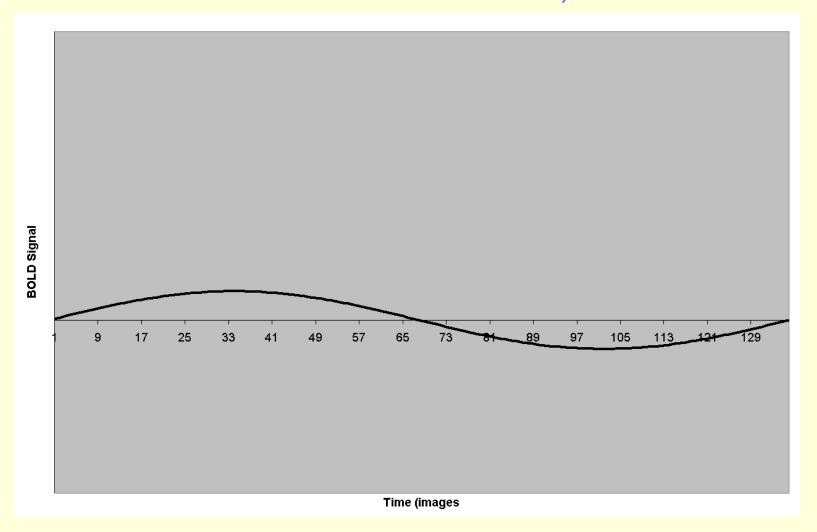


- Head motion can be quantified with 6 parameters given in any motion correction algorithm
 - x translation
 - y translation
 - z translation
 - xy rotation
 - xz rotation
 - yz rotation
- For simplicity, I've only included parameter one in our model
- Head motion can lead to other problems not predictable by these parameters

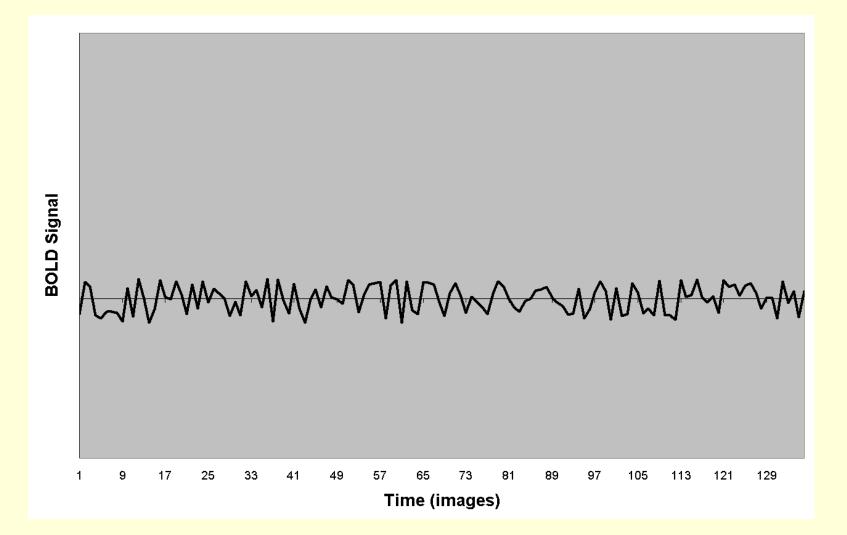
Adding linear drift from magnet noise (e.g., parts warm up) or physiological noise (e.g., subject's head sinks



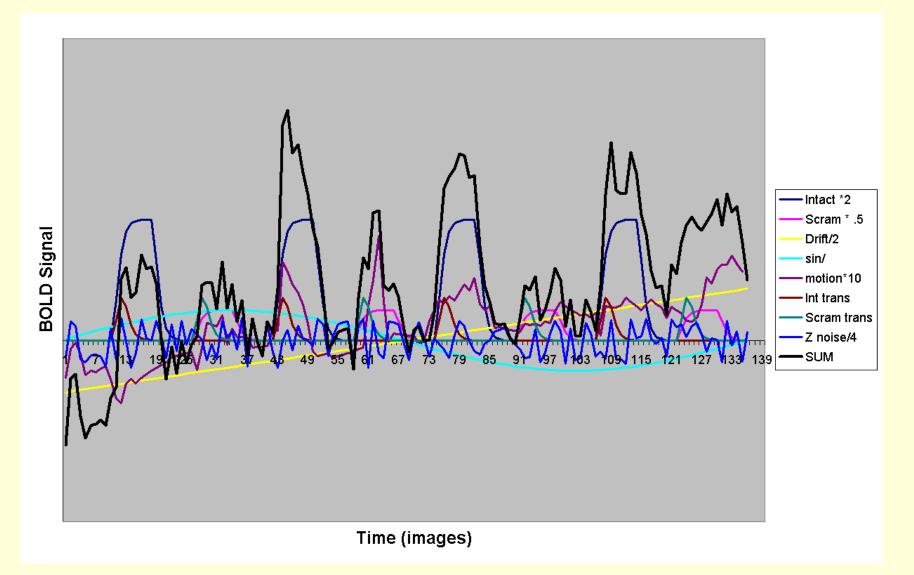
Add a dash of low frequency noise from magnet noise or physiological noise (e.g., subject's cycles of alertness/drowsiness)



Adding some high frequency noise from magnet noise or physiological noise (e.g., subject's breathing rate and heartrate

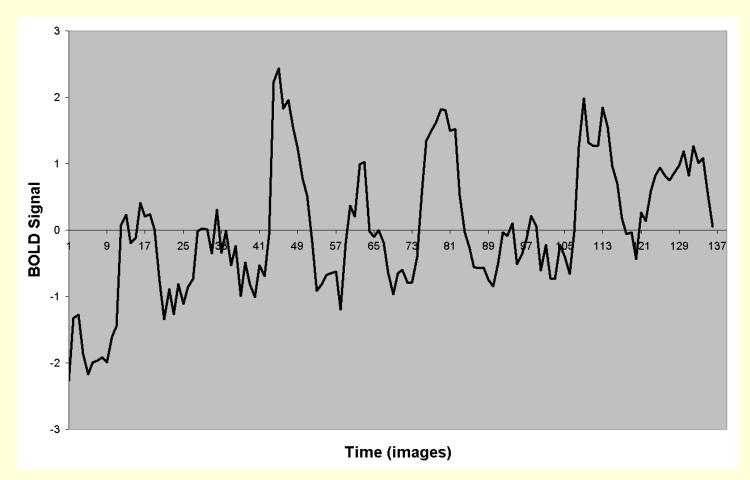


When we add these all together, we get a realistic time course



Now let's be the experimenter

- First, we take our time course and normalize it using z scores
- z = (x mean)/SD
- normalization leads to data where: mean = zero SD = 1



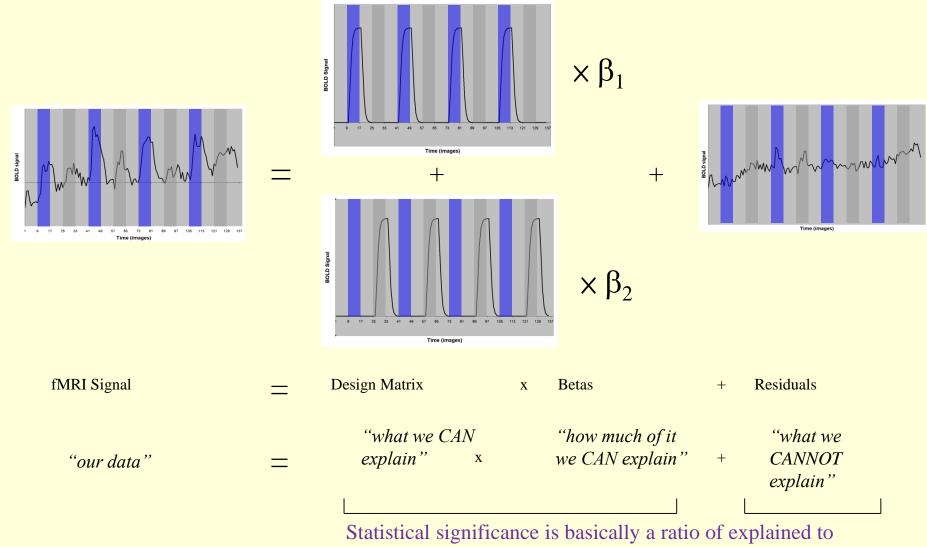
Using General Linear Model

- T-tests, correlations and Fourier analysis work for simple designs.
- The General Linear Model (GLM) can be used

Why is the GLM so great?

- Any combination of contrasts can be used (e.g., intact scrambled, scrambled - baseline) with one GLM rather than multiple correlations
- the GLM allows for combining data within subjects and between subjects
- the GLM allows you to model things that may account for variability in the data (e.g., head motion)
- GLM allows using more complex designs (e.g., factorial designs)

We create a GLM with 2 predictors



unexplained variance

