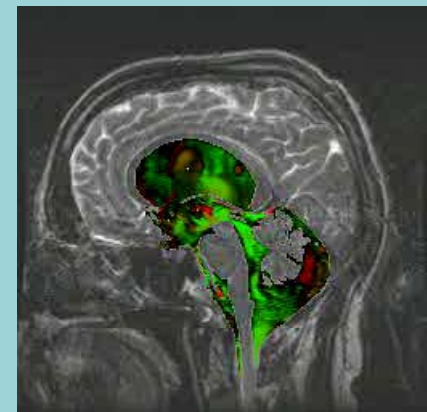
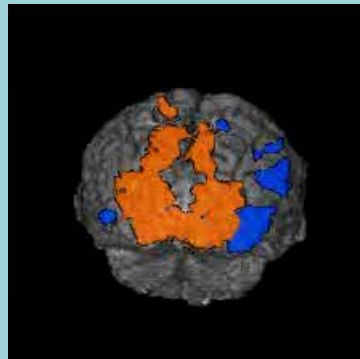
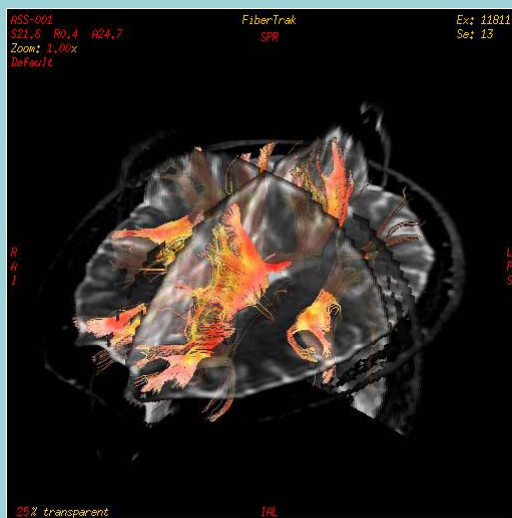
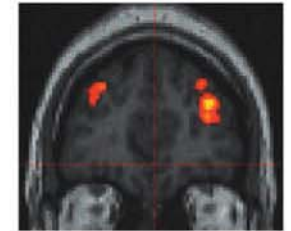
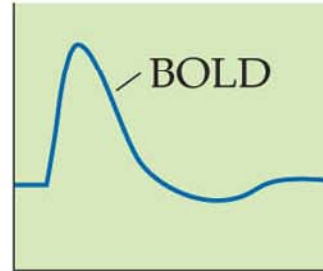
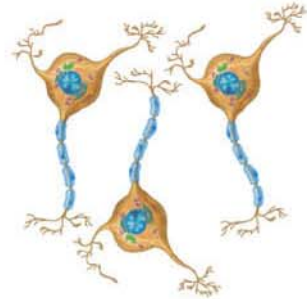


fMRI Experimental Design -- a Linear System Perspective

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Departments of Radiology and Psychology



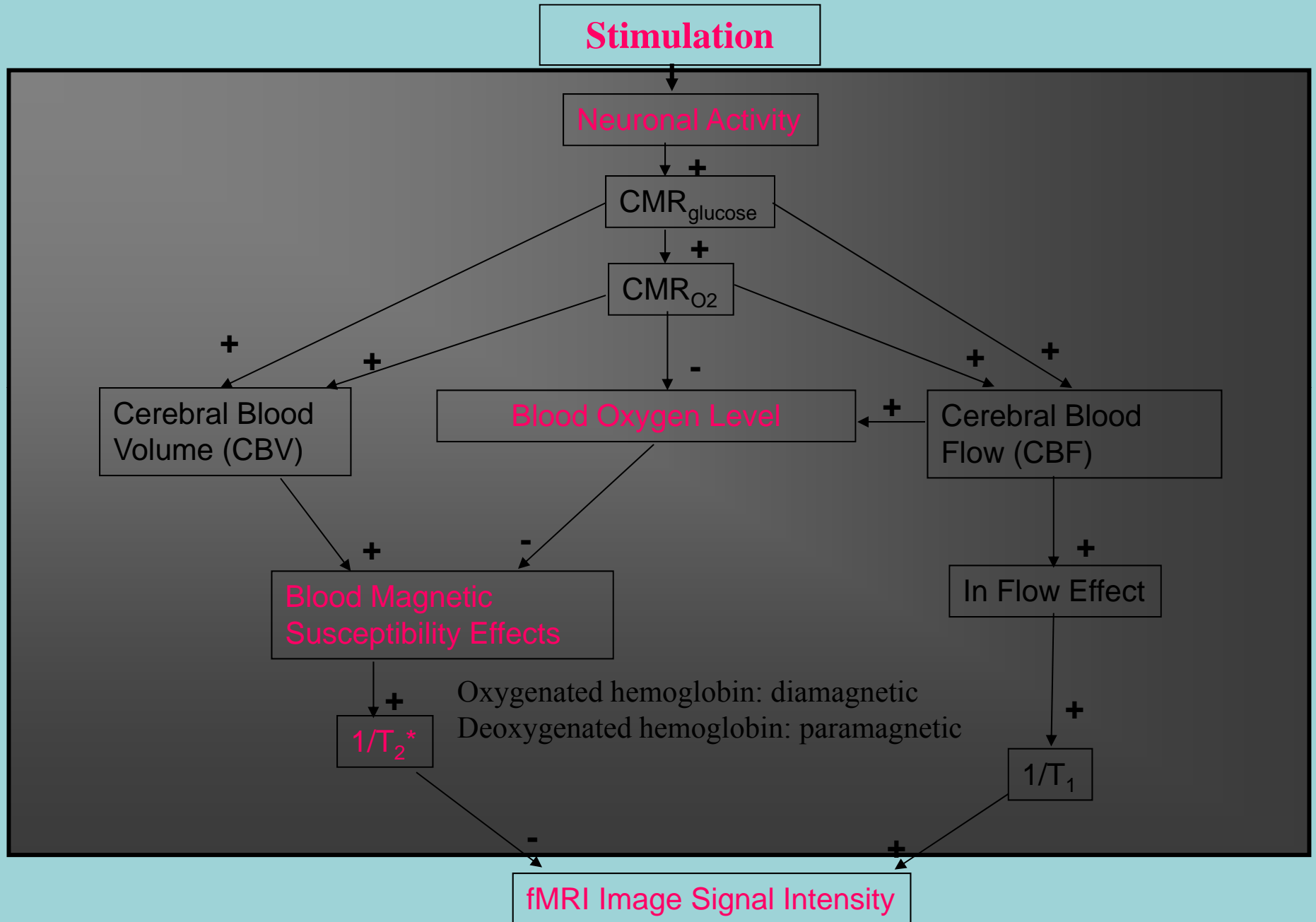


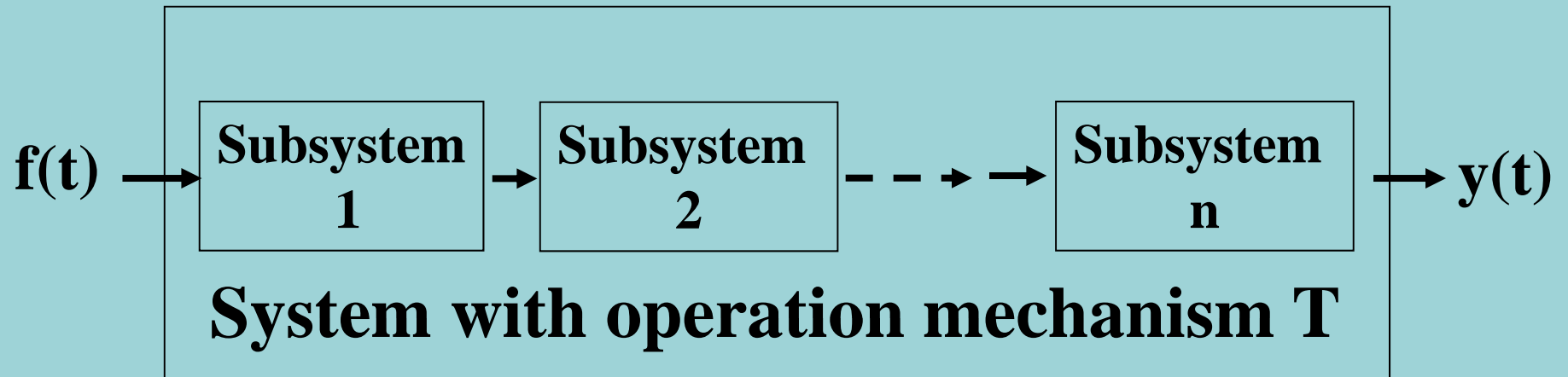
Hemodynamic hypotheses

Neuronal hypotheses

Psychological hypotheses

Mechanisms of BOLD (Blood Oxygen Level-Dependent) fMRI





$$y(t) = T\{f(t)\}$$

- ⇒
- (1) Find T ⇒ Event-related design
 - (2) Assume $T' = T$ based on some model
Find expected $y'(t) = T'\{f(t)\}$
Compare $y(t)$ and $y'(t)$
⇒ Block related design

Linear System

$f(t) \longrightarrow$ System $\longrightarrow y(t)$

$g(t) \longrightarrow$ System $\longrightarrow z(t)$

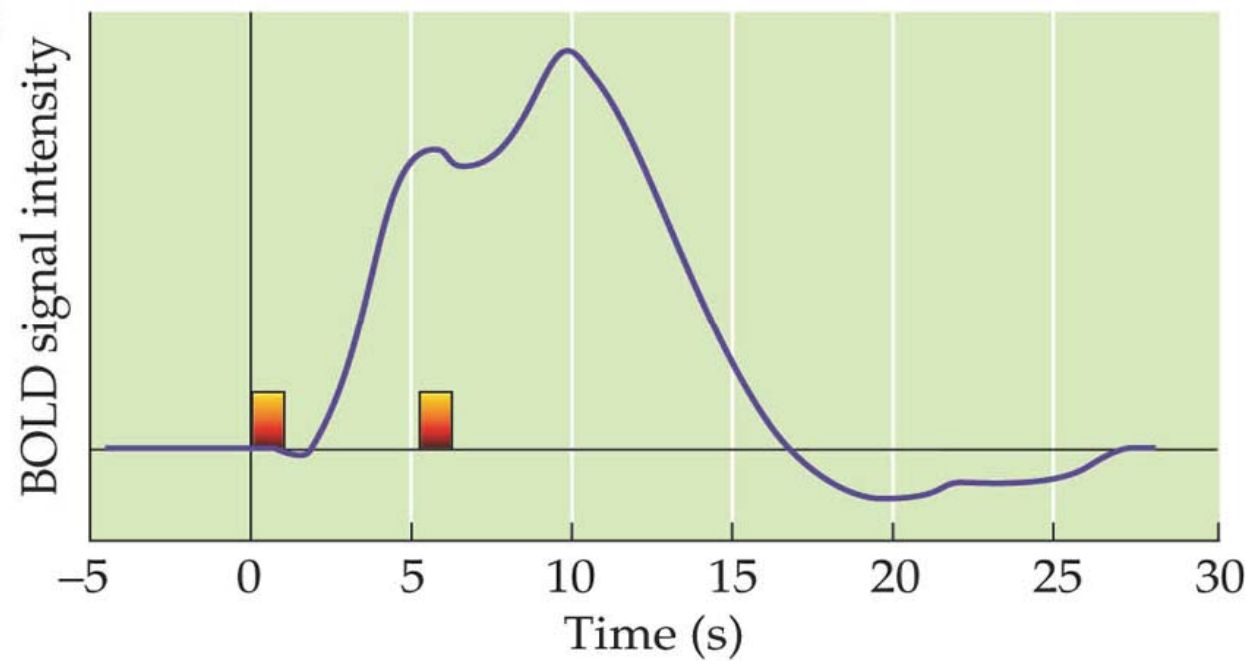
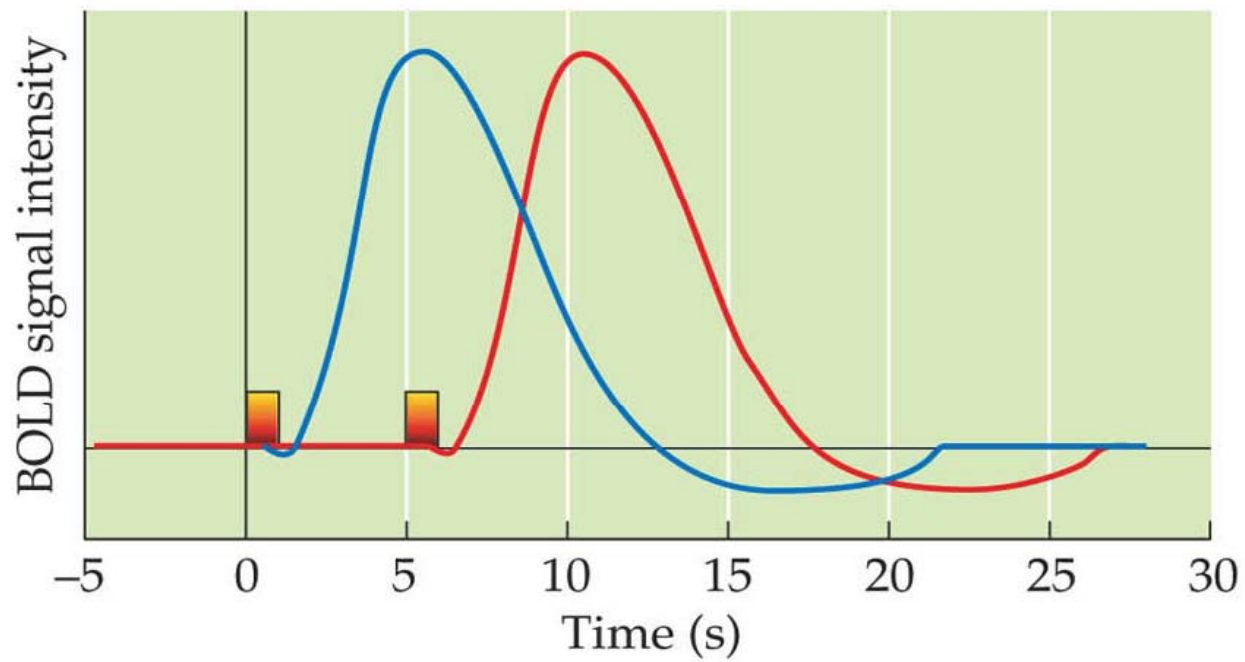
$af(t) + bg(t) \longrightarrow$ System $\longrightarrow ay(t) + bz(t)$



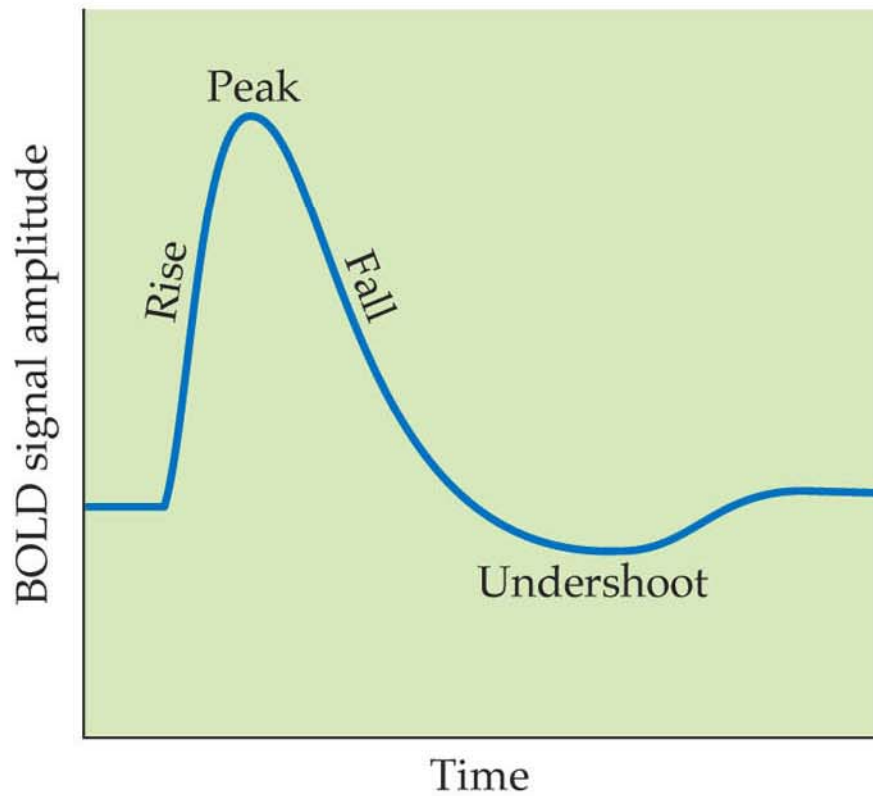
$f(t) \longrightarrow$ System $\longrightarrow y(t)$

$f(t-t_0) \longrightarrow$ System $\longrightarrow y(t-t_0)$

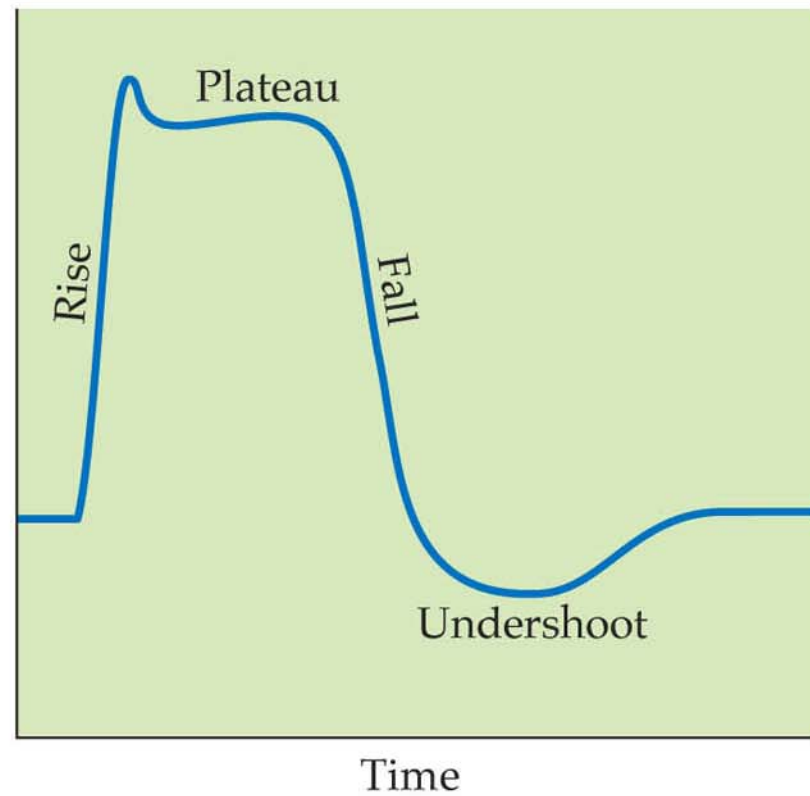
Linear Combination



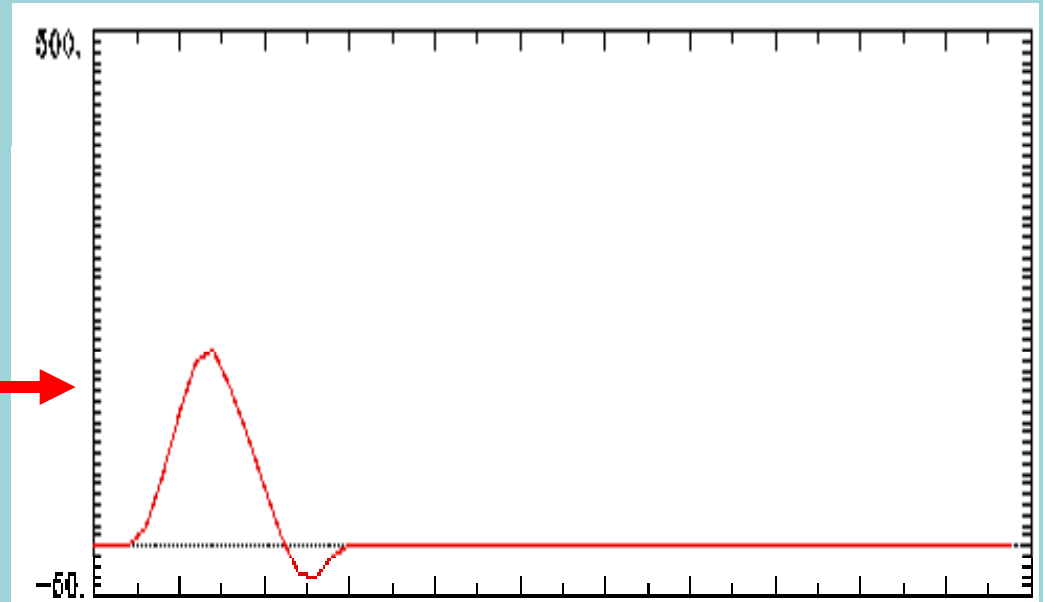
(A) A single short-duration event



(B) A block of multiple consecutive events



Present for 2 seconds



impulse



impulse response

$\delta(t)$

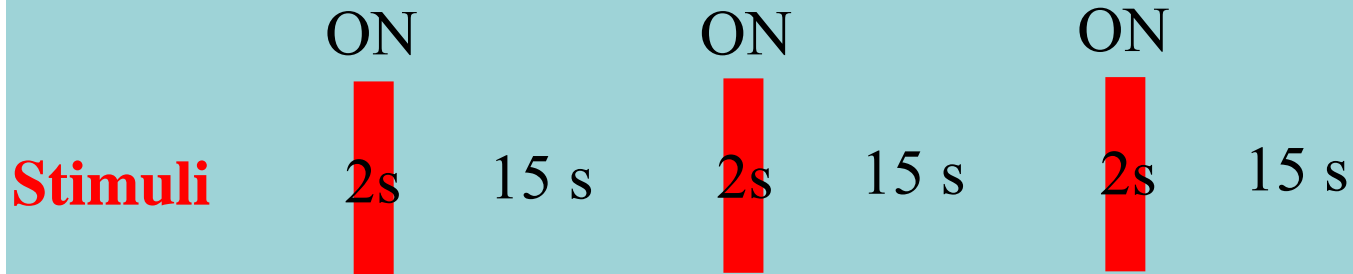


System

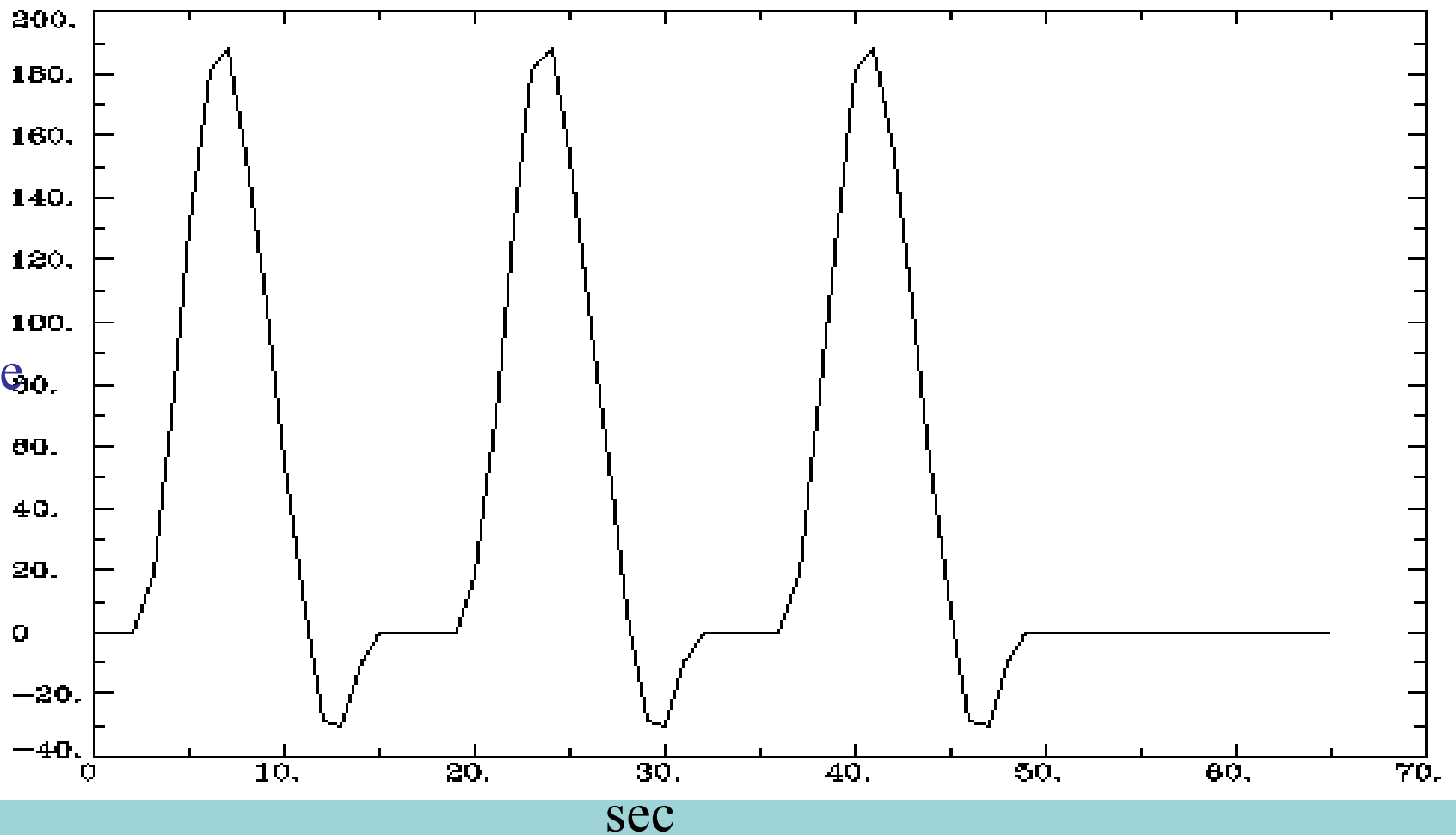


$h(t)$

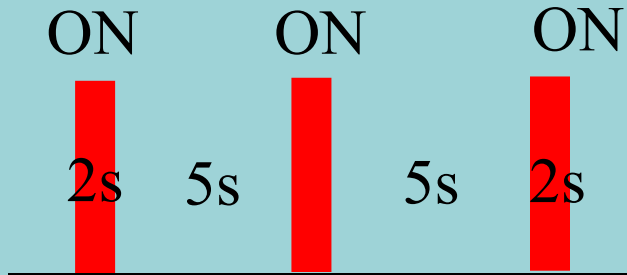
Traditional (Slow) Event Related Design



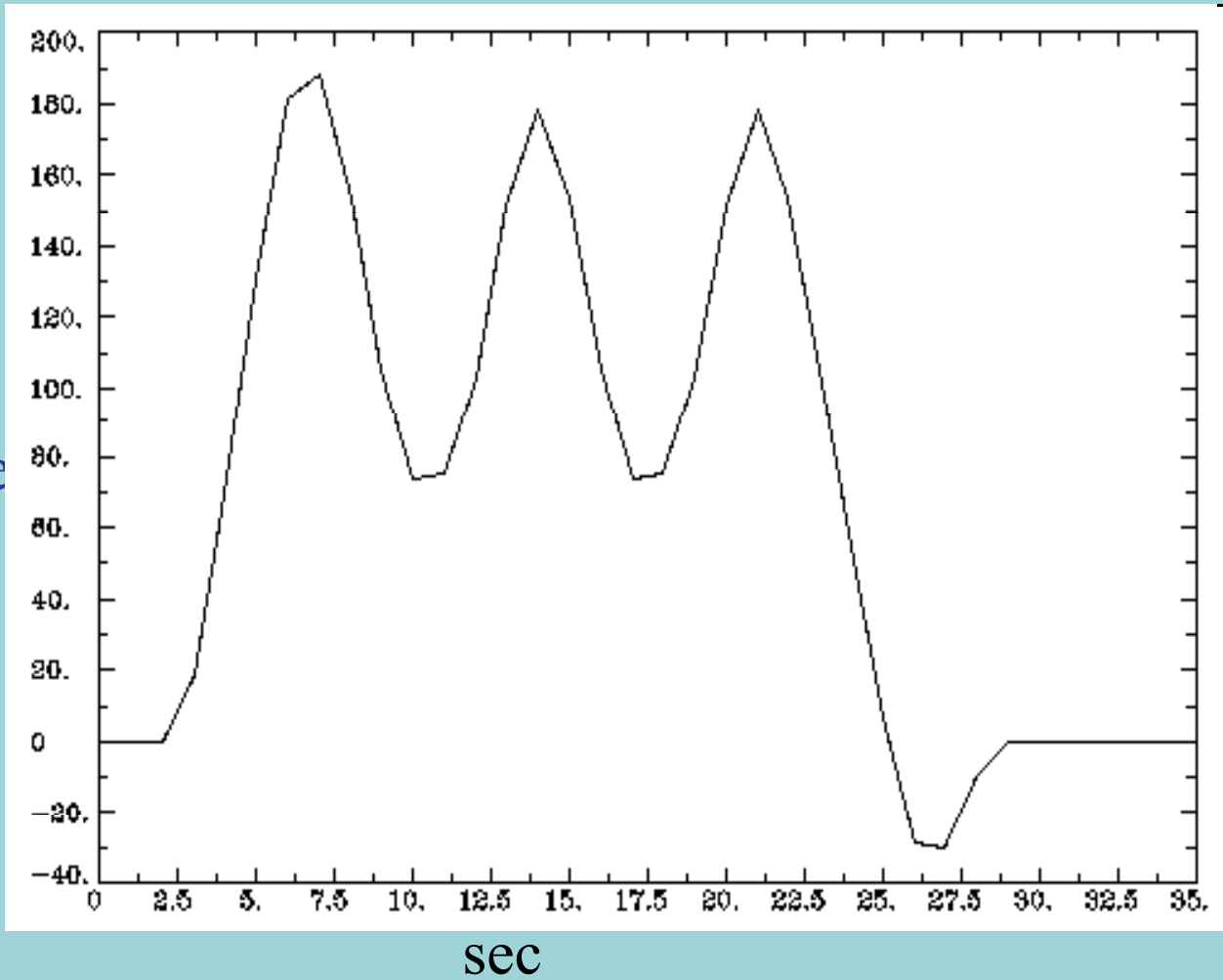
Response



Stimuli



Response

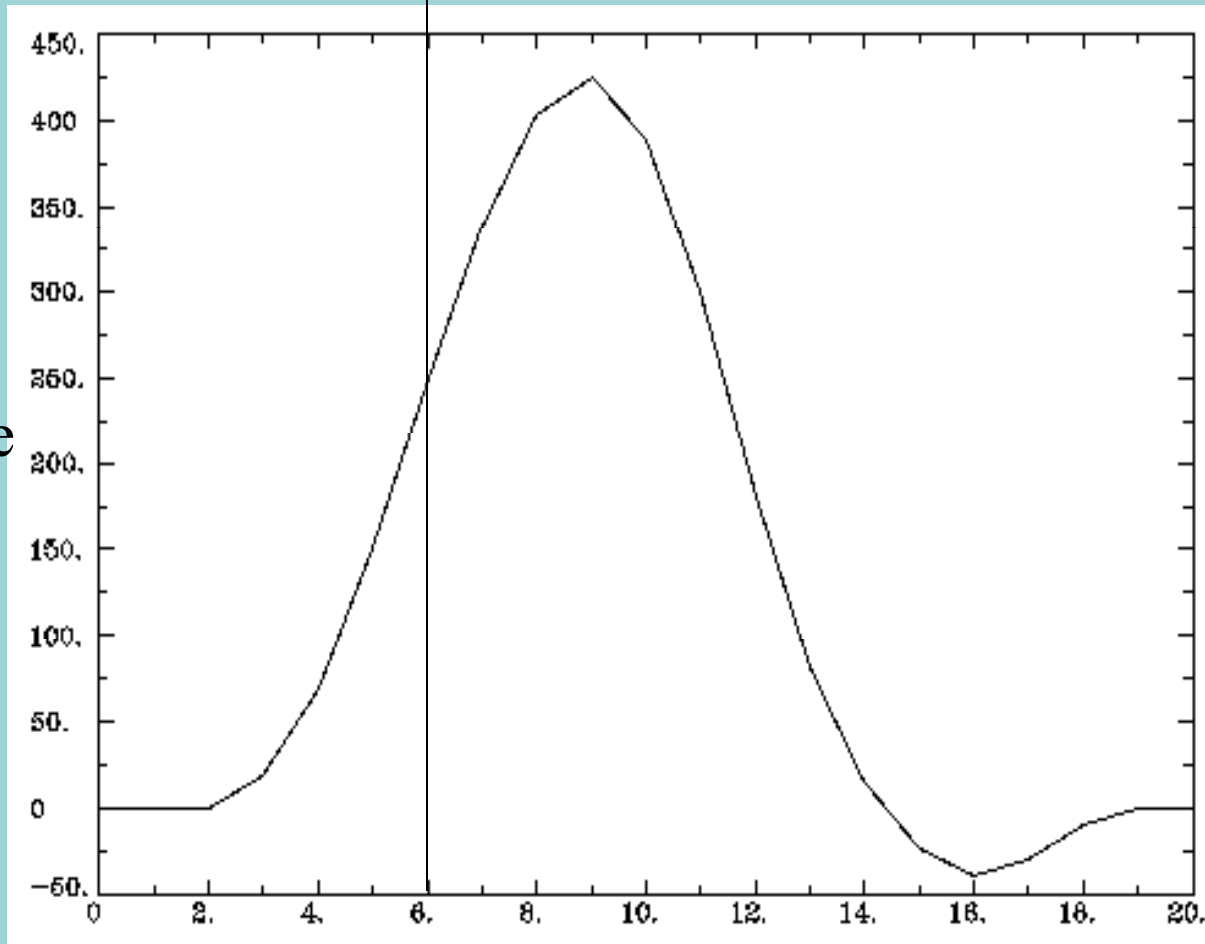


Stimuli

ON



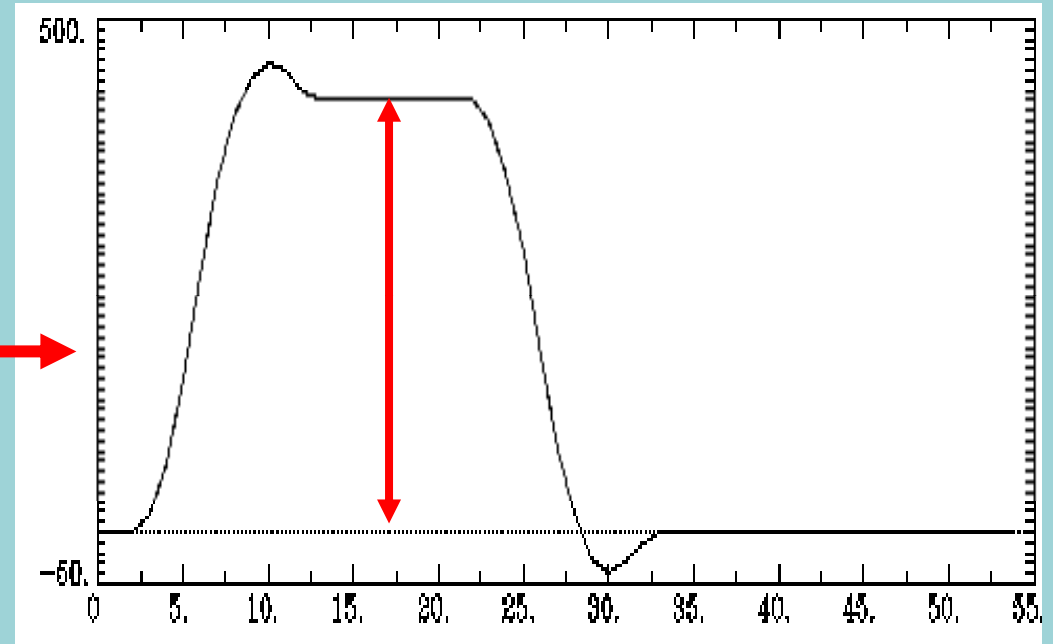
Response

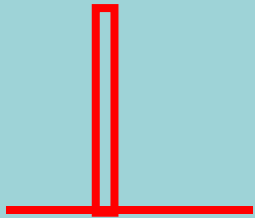


Time

Block Design

Present 10 pictures
With 2 seconds each



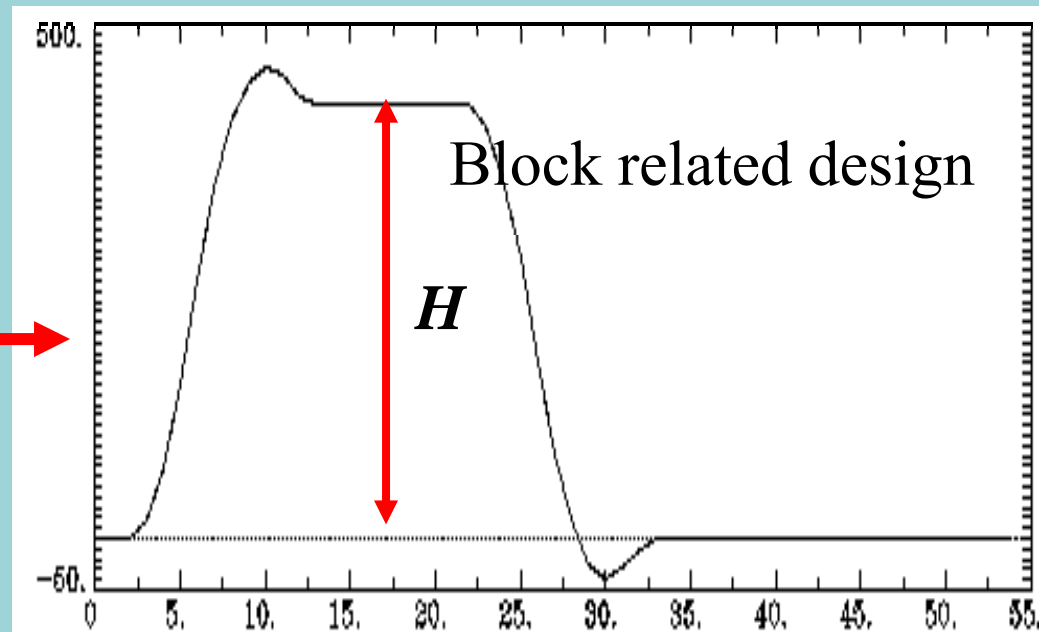
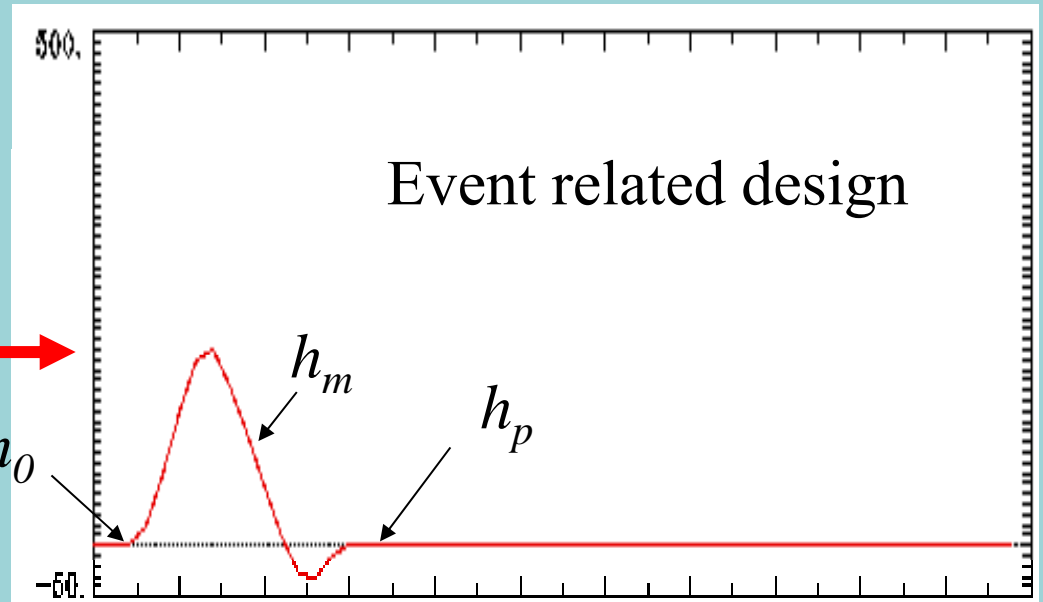


Present for 2 seconds



$$H \approx \sum_{m=0}^p h_m$$

Present 10 pictures
With 2 seconds each



Rapid Event Related Design

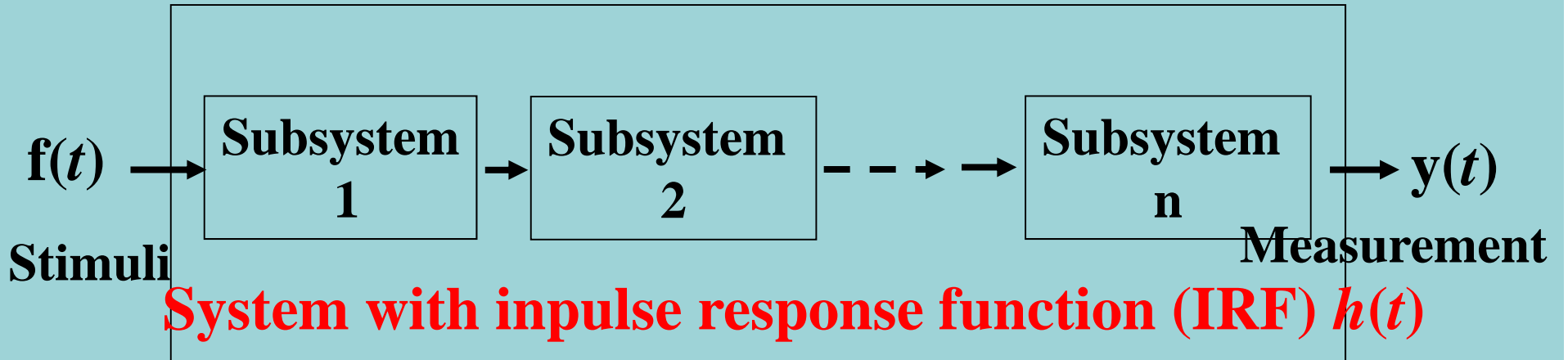
Randomizing the stimuli -> More real life

More stimuli per unit time comparing to traditional event related design -> time efficiency

Design flexibility -> more fun

Glover GH. Deconvolution of impulse response in event-related BOLD fMRI. Neuroimage. 1999;9:416-29.

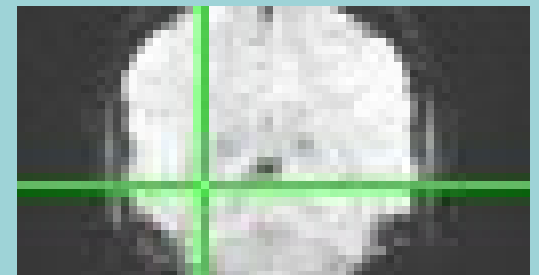
Modeling of fMRI



deconvolution

$h(t)$

$$y(t) = f(t) \otimes h(t) + \text{noise}(t)$$
$$= \int_0^t f(\tau) h(t - \tau) d\tau$$



Continuous: $y(t) = f(t) \otimes h(t) = \int_0^t f(\tau)h(t - \tau)d\tau$

Discrete times: $y(n\Delta t) = \sum_{m=0}^n f(m\Delta t)h(n\Delta t - m\Delta t)\Delta t$

In short hand,

$$\begin{aligned}y_n &= \sum_{m=0}^n f_m h_{n-m} \\ &= \sum_{m=0}^n f_{n-m} h_m\end{aligned}$$

Assume $h_m = 0$ for $n \geq p$, then

$$y_n = \sum_{m=0}^p f_{n-m} h_m$$

$$\begin{array}{ccc}
 & \mathbf{Z}_n = \sum_{m=0}^p f_{n-m} h_m + \boldsymbol{\varepsilon}_n & \\
 \nearrow & \underbrace{\hspace{10em}}_{y_n} & \nwarrow \\
 \text{measurement} & & \text{Error}
 \end{array}$$

Including constant baseline + linear trend, the MR signal measured

$$\begin{aligned}
 \mathbf{Z}_n &= y_n + \beta_0 + \beta_1 n + \boldsymbol{\varepsilon}_n \\
 &= \beta_0 + \beta_1 n + h_0 f_n + h_1 f_{n-1} + \cdots + h_p f_{n-p} + \boldsymbol{\varepsilon}_n
 \end{aligned}$$

For $n = p, p+1, \dots, N-1$

Using the matrix notation,

$$\mathbf{Z} = \begin{bmatrix} Z_p \\ Z_{p+1} \\ \cdot \\ \cdot \\ \cdot \\ Z_{N-1} \end{bmatrix}, \quad \mathbf{X} = \begin{bmatrix} 1 & p & f_p & \cdots & f_0 \\ 1 & p+1 & f_{p+1} & \cdots & f_1 \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ 1 & N-1 & f_{N-1} & \ddots & f_{N-p-1} \end{bmatrix}$$

$$\boldsymbol{\beta} = \begin{bmatrix} \beta_0 \\ \beta_1 \\ h_0 \\ \cdot \\ \cdot \\ \cdot \\ h_p \end{bmatrix} \quad \boldsymbol{\varepsilon} = \begin{bmatrix} \varepsilon_p \\ \varepsilon_{p+1} \\ \cdot \\ \cdot \\ \cdot \\ \varepsilon_{N-1} \end{bmatrix}$$

$$\mathbf{Z} = \mathbf{X} \boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

Error
term

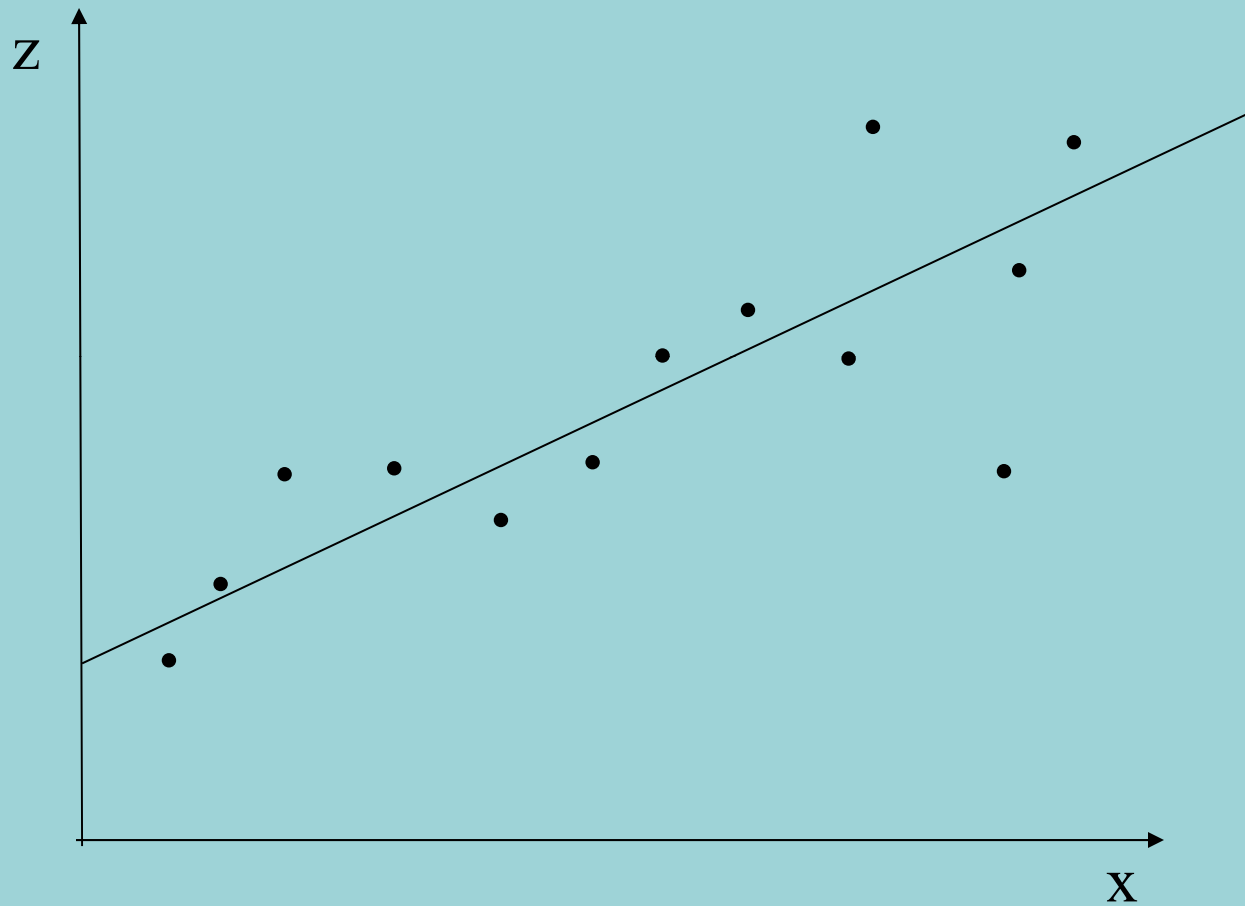
The MR signal
intensity at a voxel
from a 7-min run

Baseline signal +
linear trend + IRF

The design
matrix (when
the stimulus
ON and OFF)

$$\hat{\boldsymbol{\beta}} = \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{h}_0 \\ \cdot \\ \cdot \\ \cdot \\ \hat{h}_p \end{bmatrix} = (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{Z}$$

$$Z = \beta_0 + k X + \text{Noise}$$



Example

$$\begin{cases} a_1x_1 + b_1x_2 + c_1x_3 = 1 \\ a_2x_1 + b_2x_2 + c_2x_3 = 2 \\ a_3x_1 + b_3x_2 + c_3x_3 = 3 \end{cases}$$



$$\begin{bmatrix} a_1x_1 + b_1x_2 + c_1x_3 \\ a_2x_1 + b_2x_2 + c_2x_3 \\ a_3x_1 + b_3x_2 + c_3x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$



$$\begin{bmatrix} a_1x_1 + b_1x_2 + c_1x_3 \\ a_2x_1 + b_2x_2 + c_2x_3 \\ a_3x_1 + b_3x_2 + c_3x_3 \end{bmatrix} = \begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \Rightarrow \begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

$$\mathbf{Z} = \mathbf{X} \boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

$$\hat{\boldsymbol{\beta}} = \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{h}_0 \\ \cdot \\ \cdot \\ \cdot \\ \hat{h}_p \end{bmatrix} = \underbrace{(\mathbf{X}^t \mathbf{X})^{-1}} \mathbf{X}^t \mathbf{Z}$$

Design Evaluation:

(1) Multicollinearity issue

(2) Minimize error

Ward, B. (2002). *Deconvolution analysis of fMRI time series data*. Milwaukee, WI: Biophysics Research Institute, Medical College of Wisconsin.

Analogy of Multicollinearity Issue

$$\left\{ \begin{array}{l} X + Y + Z = 1 \\ 2X + 2Y + 2Z = 2 \\ 3X + 3Y + 3Z = 3 \end{array} \right. \longrightarrow \left\{ \begin{array}{l} X + Y + Z = 1 \\ X + Y + Z = 1 \\ X + Y + Z = 1 \end{array} \right. \longrightarrow \text{Cannot Resolve X, Y and Z}$$

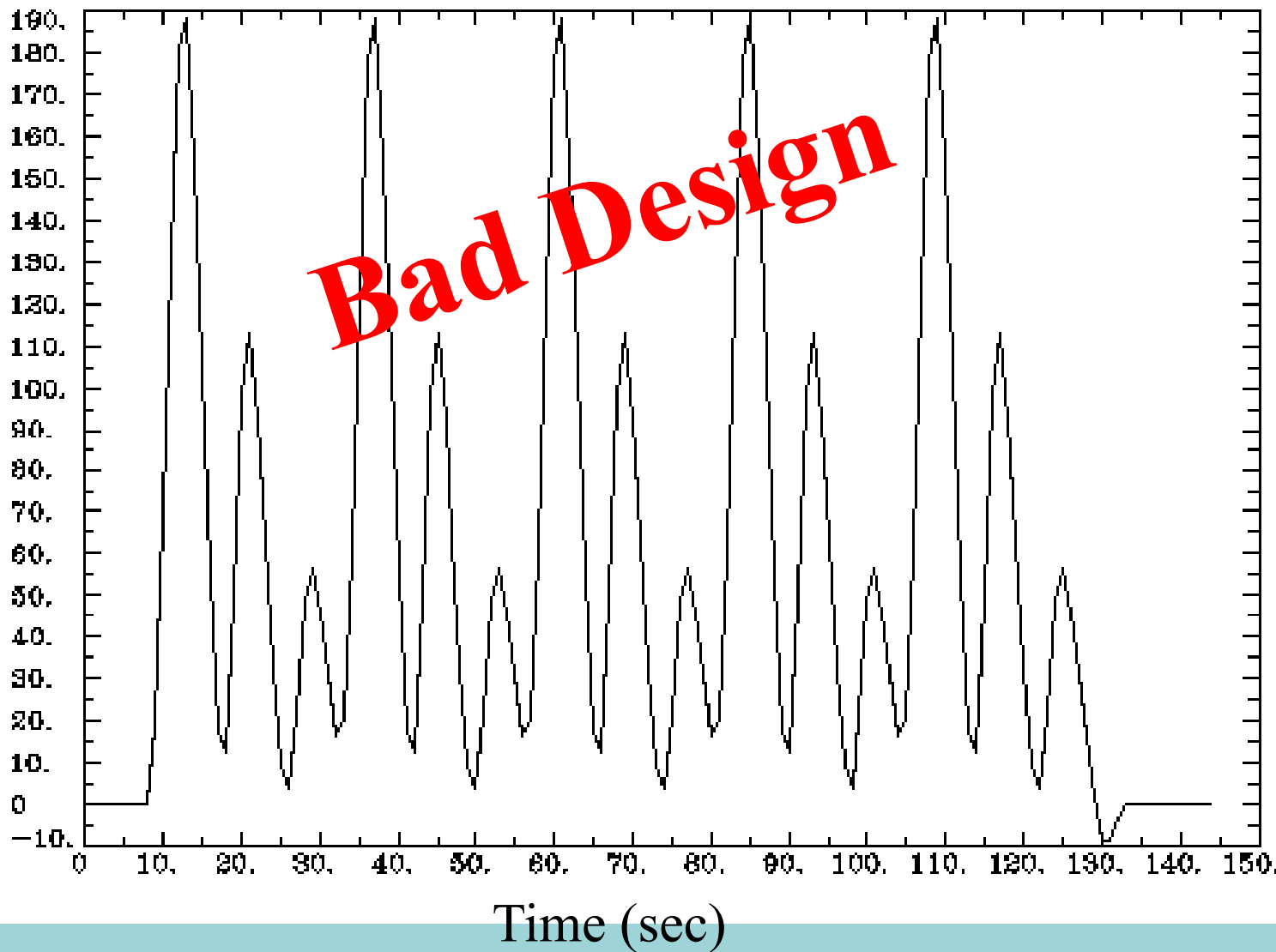
$$\left\{ \begin{array}{l} X + Y + Z = 1 \\ X + 3Y + Z = 2 \\ X + Y + 3Z = 3 \end{array} \right. \longrightarrow \text{Can resolve X, Y and Z}$$

2 sec for
each
stimulus

Repeat

--- 6sec-**A**-6sec-**B**-6sec-**C**-6sec ---

Signal

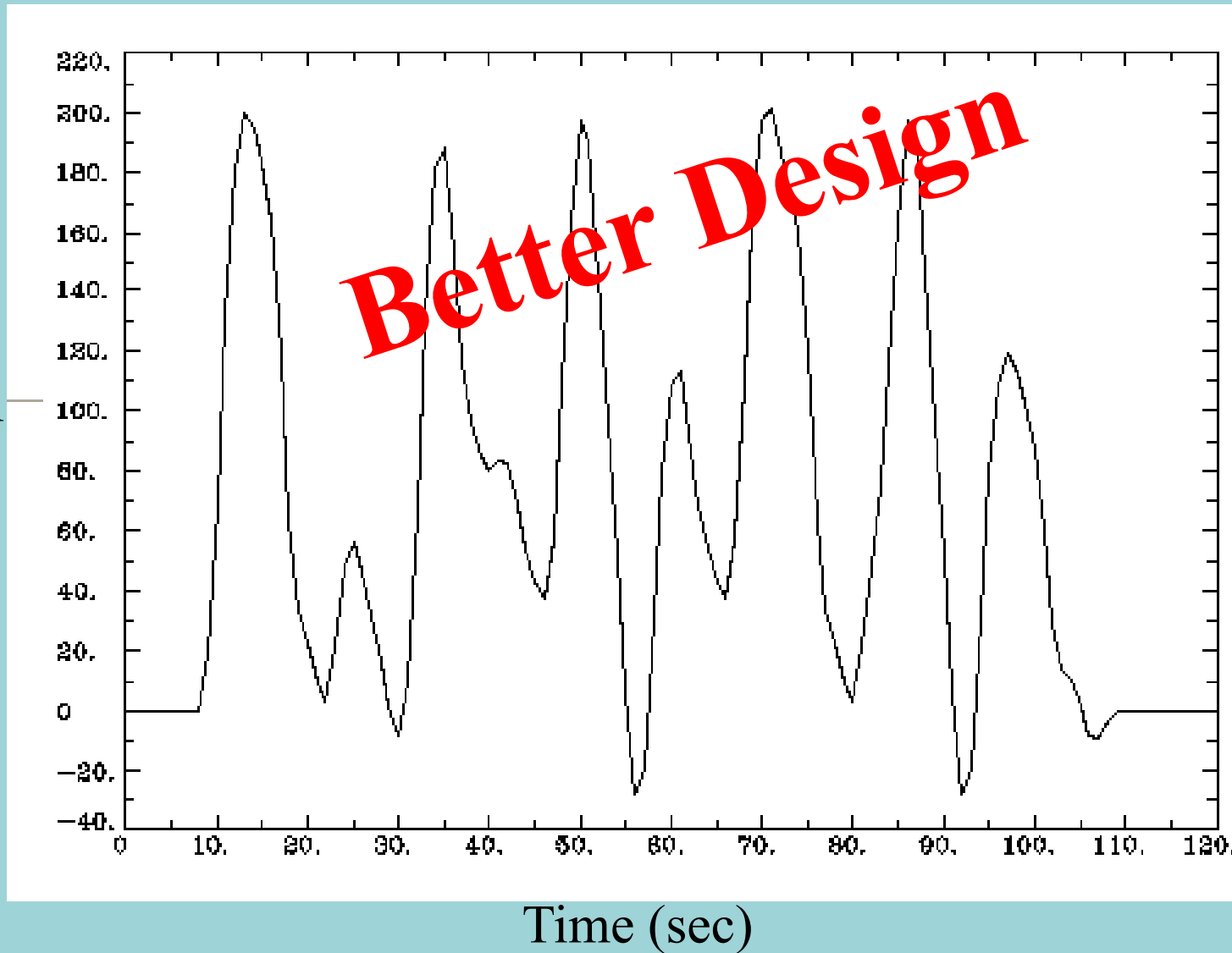


2 sec for
each
stimulus

Repeat with Various ISIs

--- 6sec-**A**-2sec-**B**-6sec-**C**-8sec-**A**-4sec-**B**-4sec-**C**-

Signal

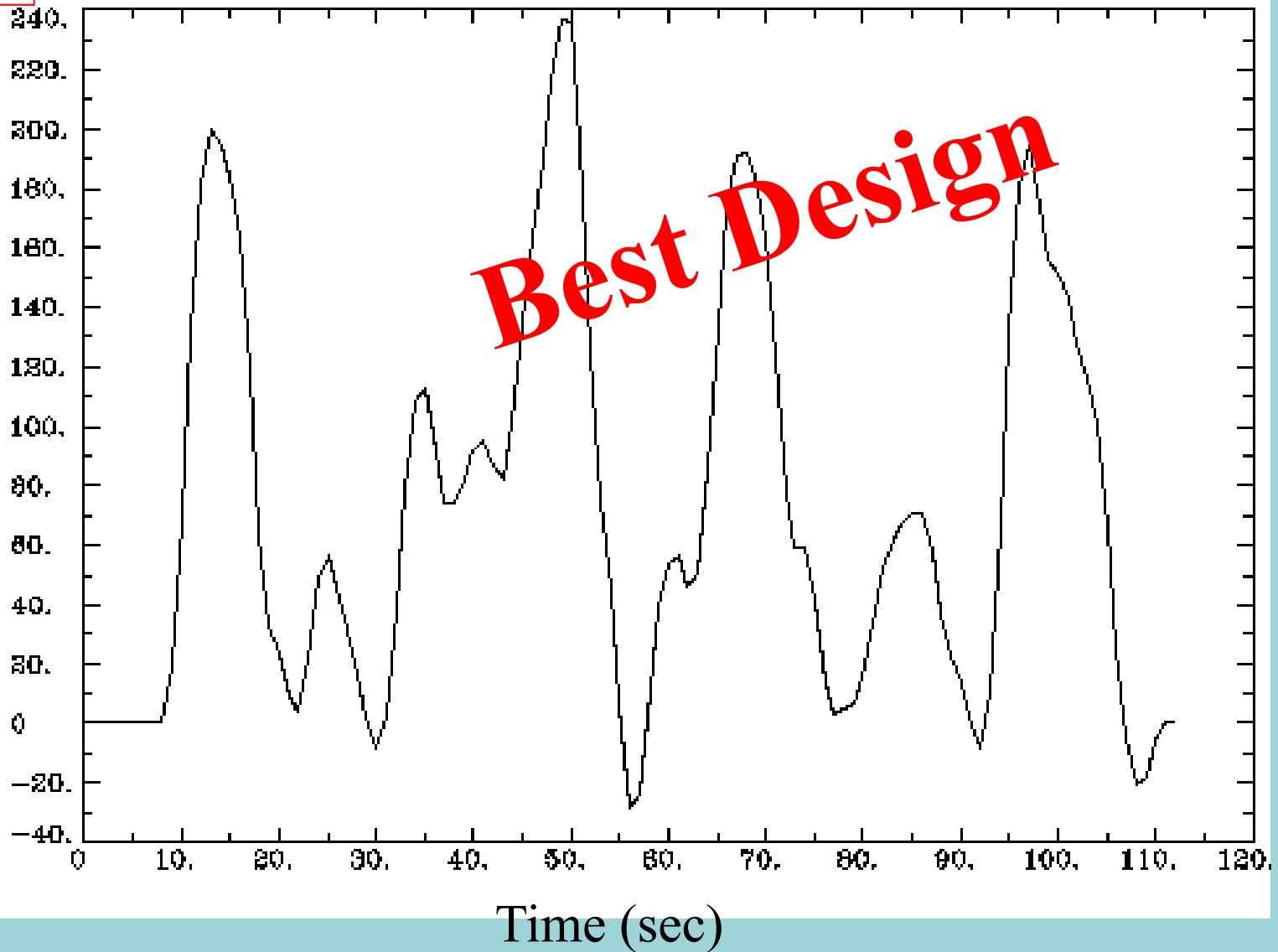


2 sec for
each
stimulus

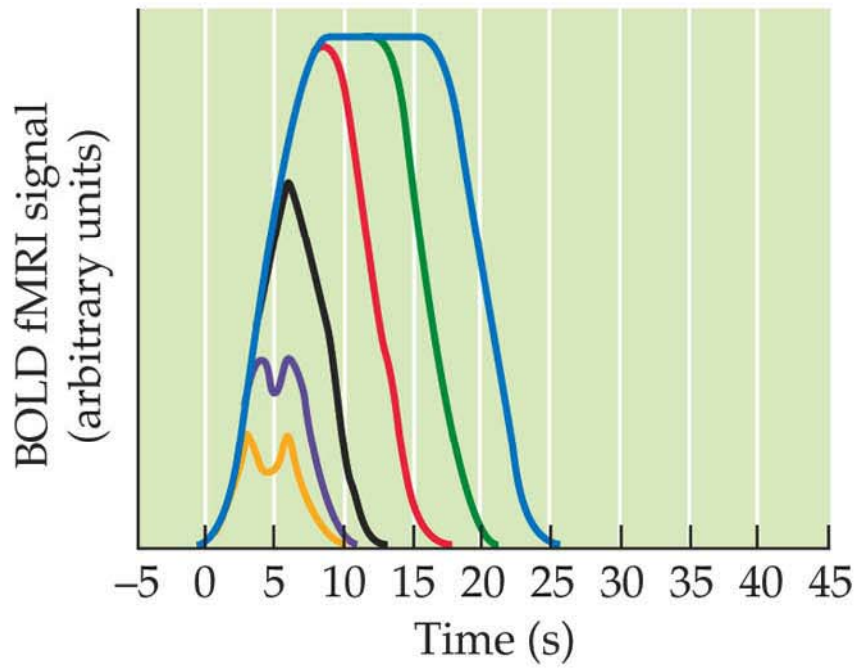
Repeat with Various ISIs and Random Order of Stimuli

---6sec-**A**-2sec-**B**-6sec-**C**-8sec-**B**-4sec-**B**-4sec-**A**-2sec-**A**---

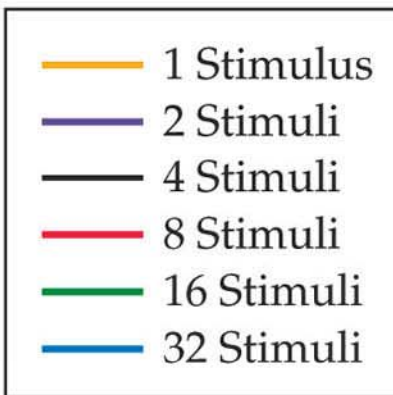
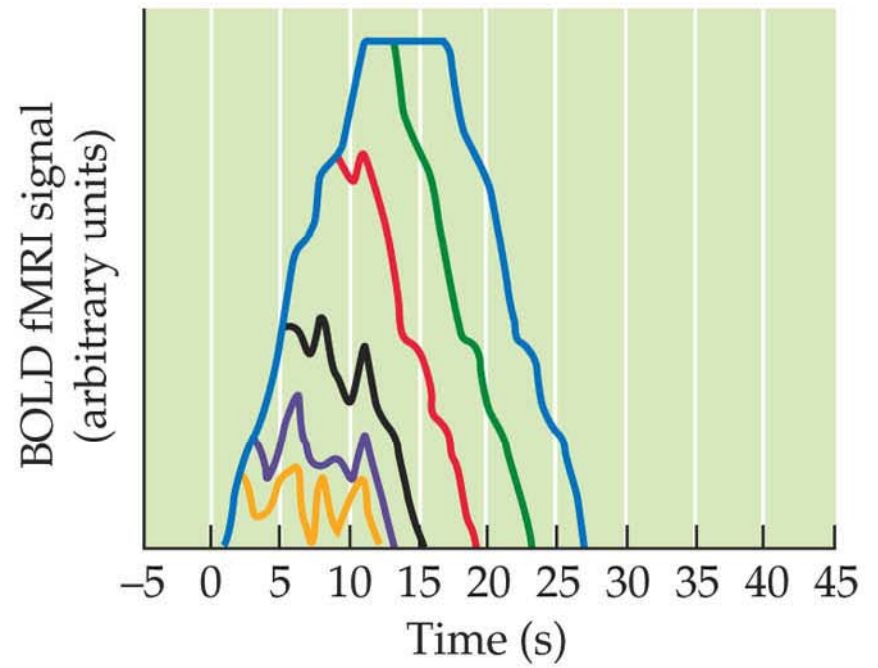
Signal



(C)



(D)



Summary

Block Design

Advantage: Best for detection
Easy to implement paradigm
Relative immune to timing error

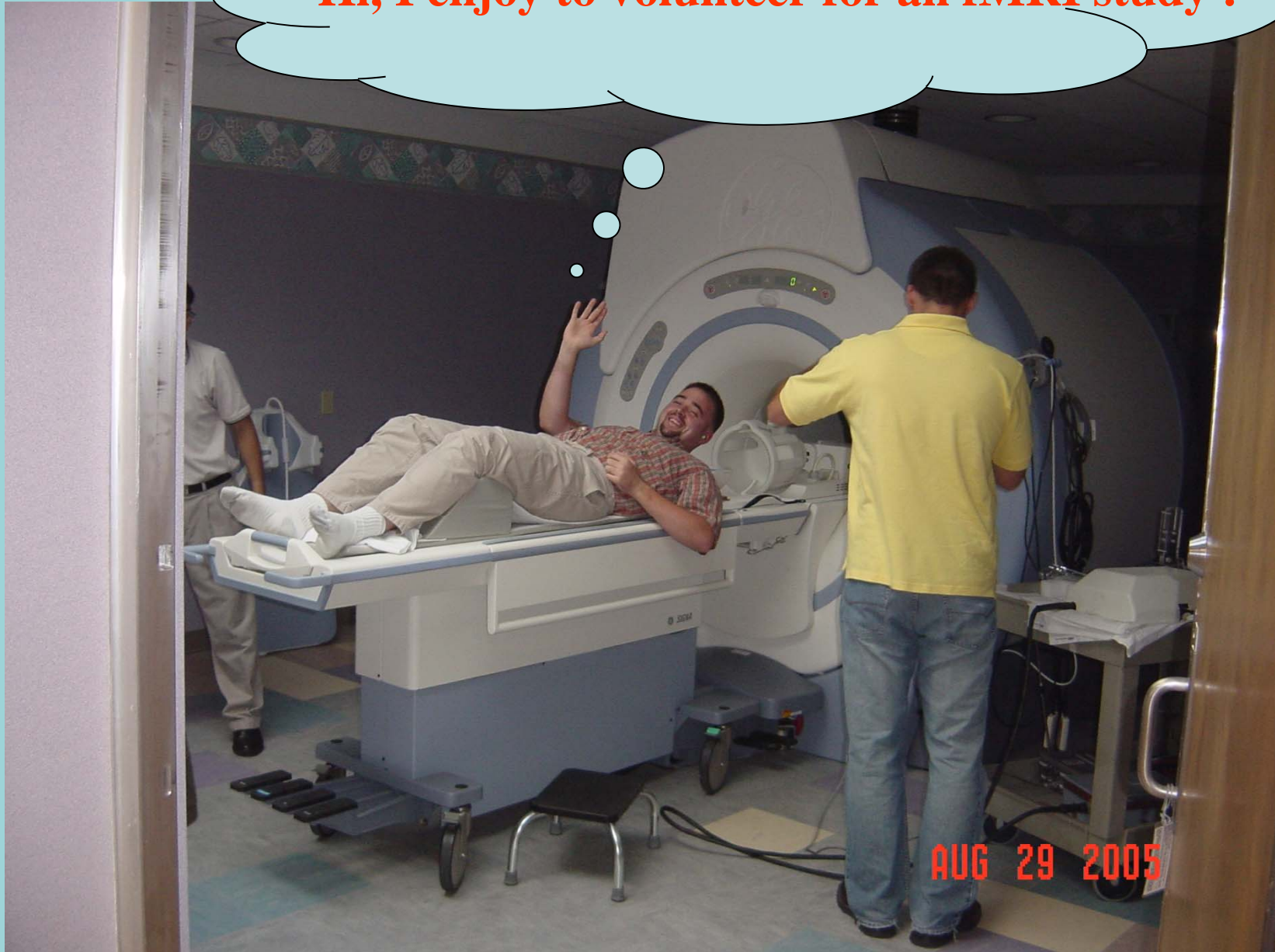
Traditional (Slow) Event Related Design:

Advantage: No assumption of linear system
Able to characterize IRF

Rapid Event Related Design:

Advantage: More life like
More time efficient
Flexible design
Able to characterize IRF

Hi, I enjoy to volunteer for an fMRI study !



Common Neuroimaging Data Analysis and Display Software

AFNI

FSL

FreeSurfer

SPM