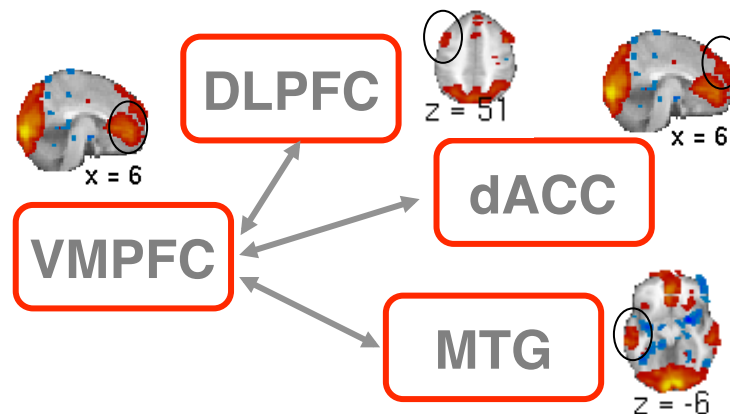


# Module 23:

## Functional Connectivity

# Brain Connectivity

- Functional Connectivity
  - Undirected association between two or more fMRI time series and/or performance and physiological variables.
  - Makes statements about the structure of relationships among brain regions.
  - Usually makes no assumptions about the underlying biology.



# Functional Connectivity

- Methods include:
  - Seed analysis
  - Inverse covariance methods
  - Multivariate decomposition methods
    - Principle Components Analysis
    - Independent Components Analysis
    - Partial Least Squares

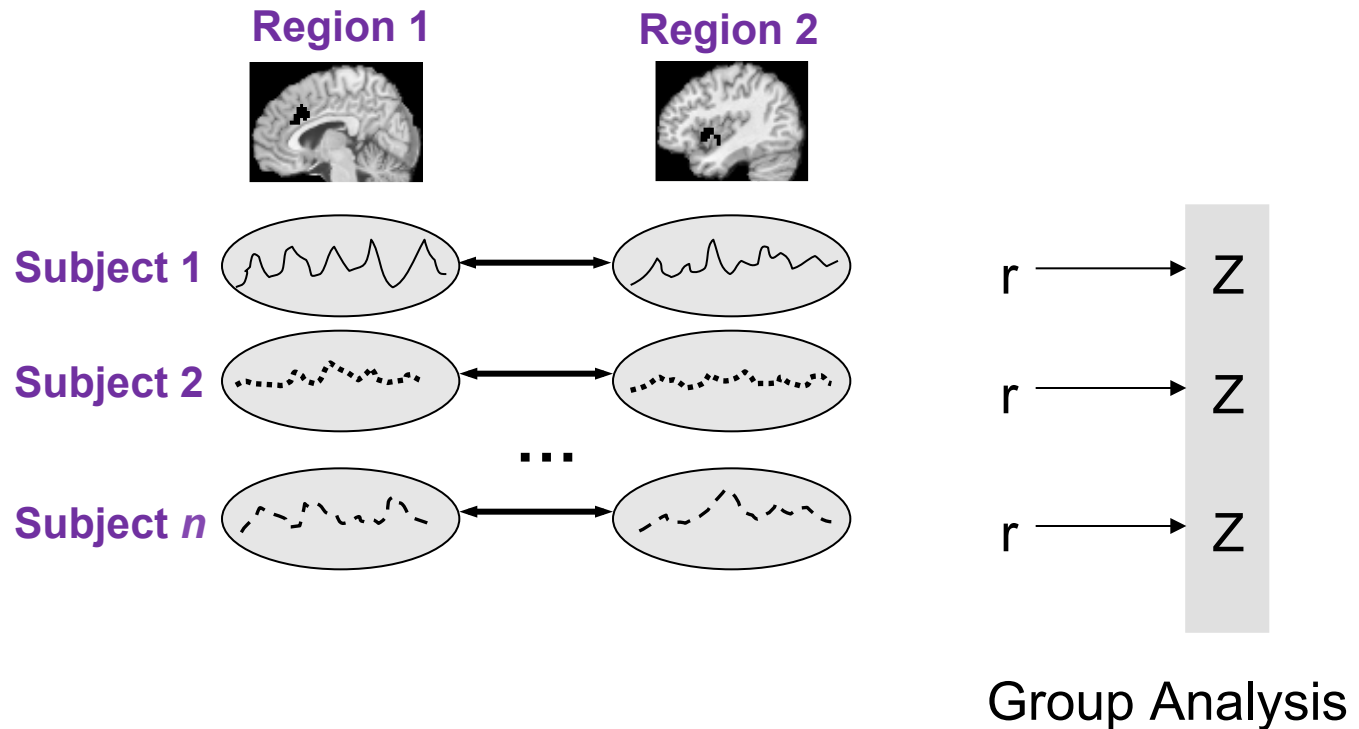
# Bivariate Connectivity



- Simple functional connectivity
  - Region A is related to Region B.
  - Provides information about relationships among regions.
  - Can be performed on time series data within a subject, or individual differences (contrast maps, one per subject).

# Time Series Connectivity

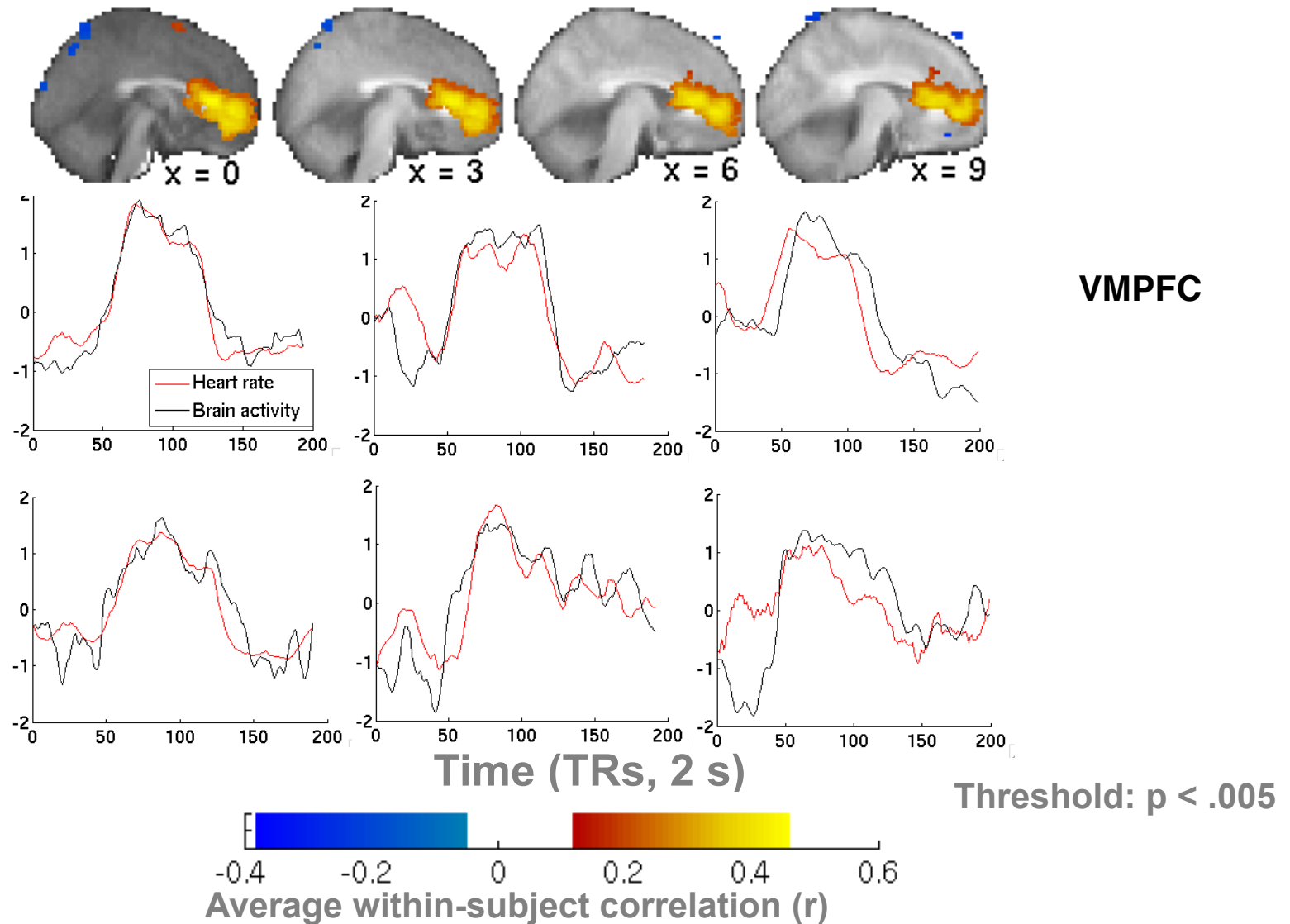
- Calculate the cross-correlation between time series from two separate brain regions.



# Seed Analysis

- In **seed analysis** the cross-correlation is computed between the time course from a predetermined region (**seed region**) and all other voxels.
- This allows researchers to find regions correlated with the activity in the seed region.
- The seed time course can also be a performance or physiological variable

# Correlations between brain activity and heart-rate



# Issues

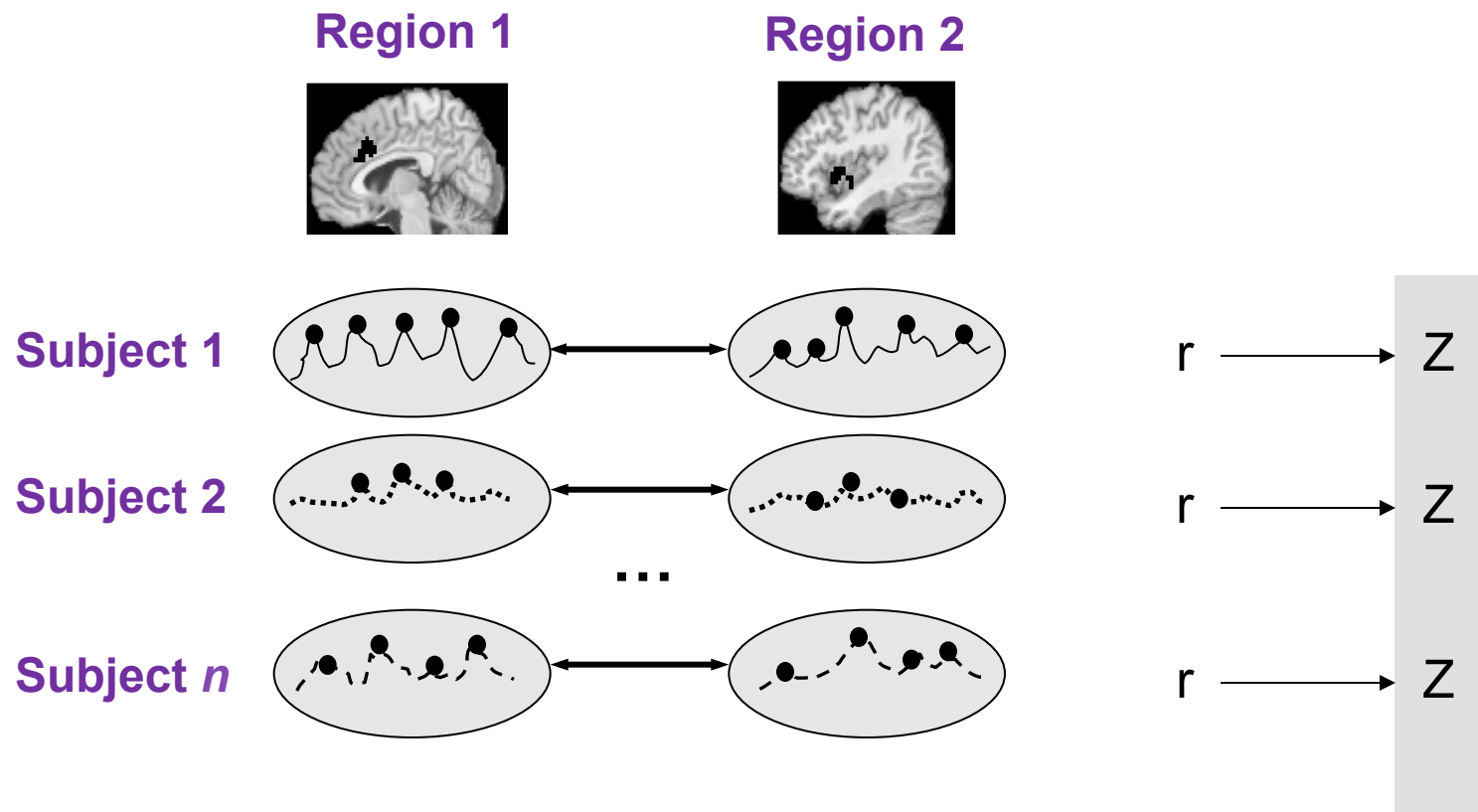
- One of the main problems with time series connectivity is the fact that there may be different hemodynamic lags in different regions:
  - Time series from different regions may not match up, even if neural activity patterns match up.
  - If lags are estimated from data, temporal order may be caused by vascular (uninteresting) or neural (interesting) response.



# Beta Series

- The **beta series** approach can be used to minimize issues of inter-region neurovascular coupling.
- Procedure:
  - Fit a GLM to obtain separate parameter estimates for each individual trial.
  - Compute the correlation between these estimates across voxels.

# Beta Series



Group Analysis

# Individual Differences

- One way of studying differences between groups of subjects is to use some trait as a covariate in between-subject analysis.
- These types of analysis of individual differences allow researchers to study relationships between brain and behavior.

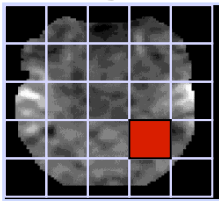
# Individual Differences

Subject

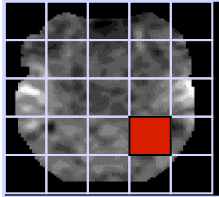
Contrast  
Image

Seed Value    Group Results

1

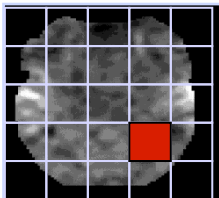


2



⋮

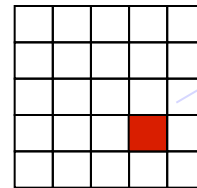
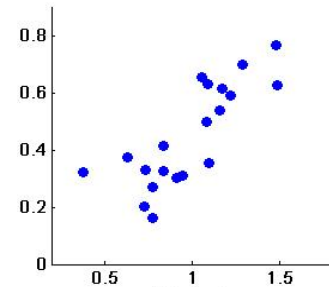
N



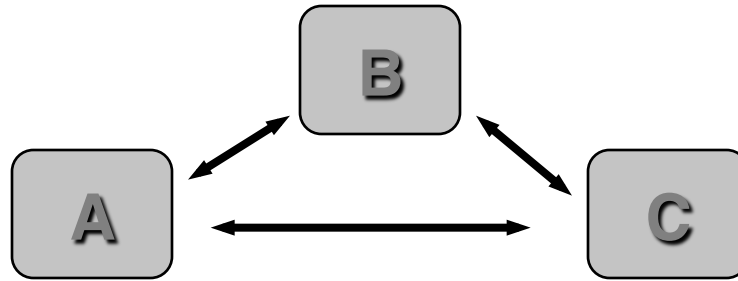
$x_1$

$x_2$

$x_N$



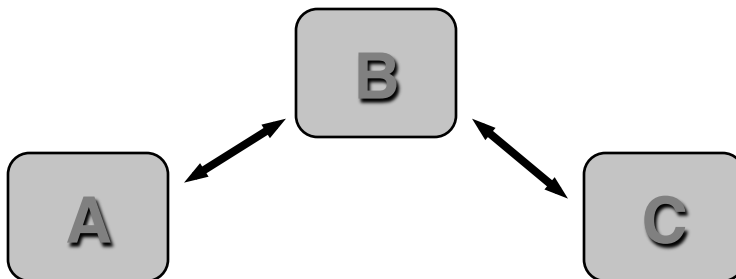
# Partial Correlation



- Partial Correlation
  - Correlation between two regions, after the effect of all other regions have been removed.
  - Helps protect against ‘illusory’ correlations between regions (e.g., A and C uncorrelated after controlling for B).

# Inverse Covariance Methods

- For multivariate normal data there exists a duality between the inverse covariance (**precision**) matrix and the graph representing relationships between regions.
  - Conditional independence between variables (regions) corresponds to zero entries in the precision matrix.
  - Graphical lasso (GLASSO) can be used to estimate sparse precision matrices and graphs.



$$\Sigma^{-1} = \begin{bmatrix} & & 0 \\ & & \\ 0 & & \end{bmatrix}$$

# End of Module



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