# Granger Causality in fMRI connectivity analysis

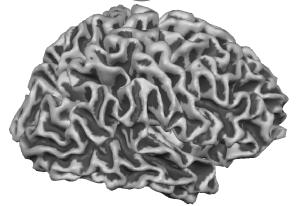
**Alard Roebroeck** 

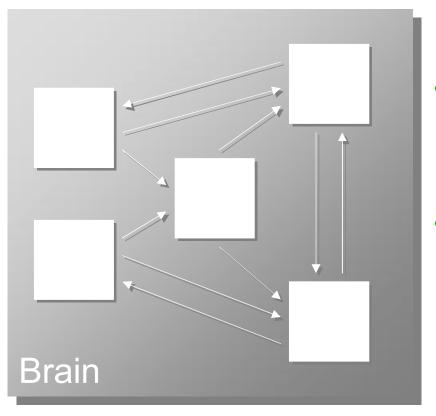
Maastricht Brain Imaging Center (MBIC) Faculty of Psychology & Neuroscience Maastricht University

#### Overview

- fMRI signal & connectivity
- Functional & Effective connectivity
- Structural model & Dynamical model
  - Identification & model selection
- Granger causality & fMRI
  - Granger causality and its variants
  - Granger causality mapping
- Issues with variable hemodynamics
  - Hemodynamic deconvolution

# Integration and connectivity

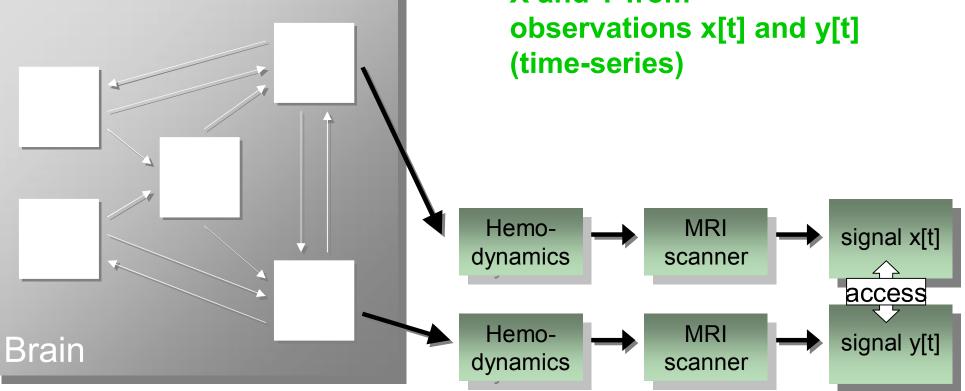




- Performance of complex tasks requires interaction of specialized brain systems (functional integration)
- Interaction of specialized areas requires connectivity
- Investigation of complex tasks requires connectivity analysis

#### A problem for fMRI connectivity

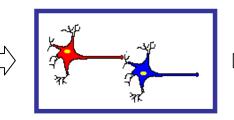
- In fMRI our access to the neural activity is *indirect*
- We want to infer interaction between Area X and Y from observations x[t] and y[t] (time-series)



# fMRI: The BOLD signal



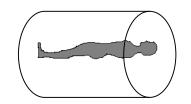
Stimulus



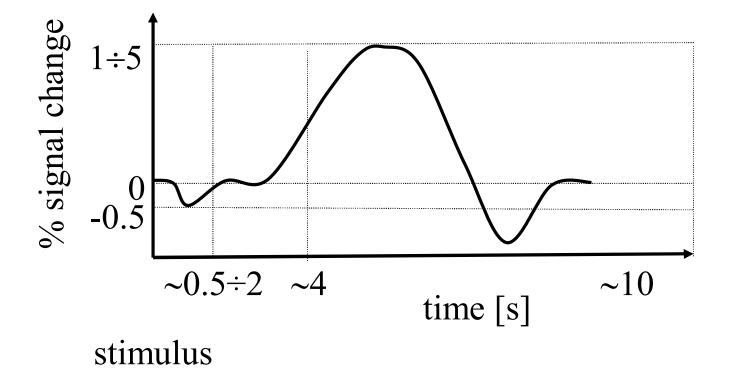
Neural pathway



Hemodynamics



MR scanner

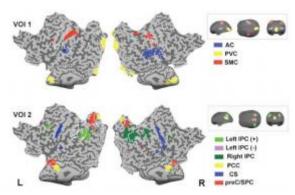


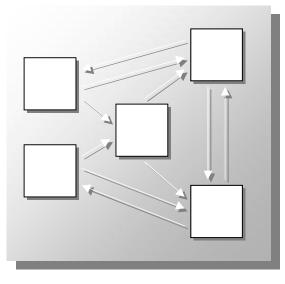
#### Overview

- fMRI signal & connectivity
- Functional & Effective connectivity
- Structural model & Dynamical model
  - Identification & model selection
- Granger causality & fMRI
  - Granger causality and its variants
  - Granger causality mapping
- Issues with variable hemodynamics
   Hemodynamic deconvolution

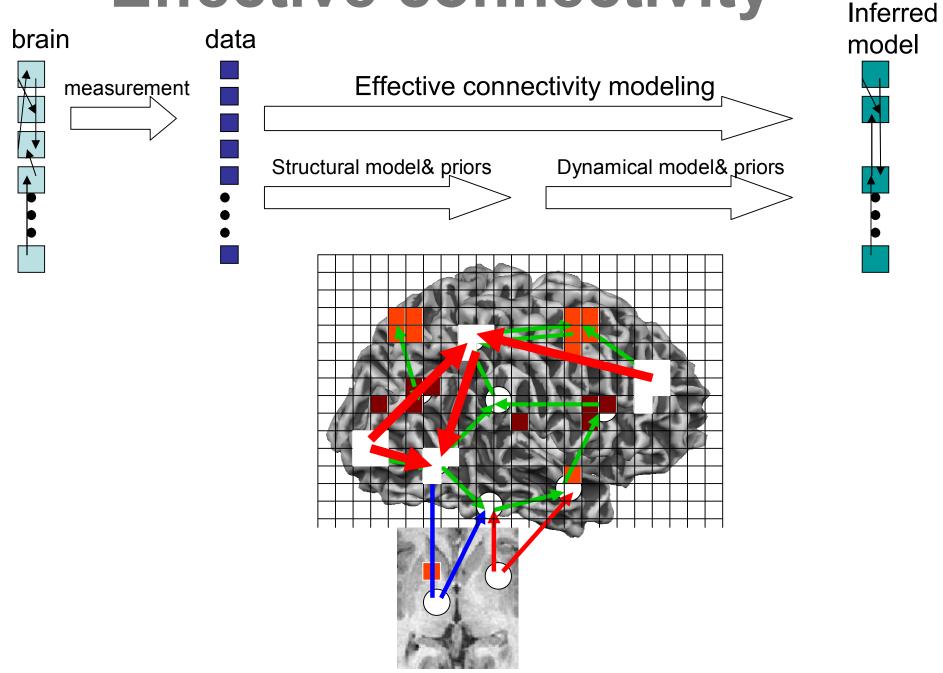
#### Functional & Effective Connectivity

- Functional connectivity
  - Association (mutual information)
  - Localization of whole networks
- Effective connectivity
  - Uncover network mechanisms (causal influence)
  - Directed vs. undirected
  - Direct vs. indirect





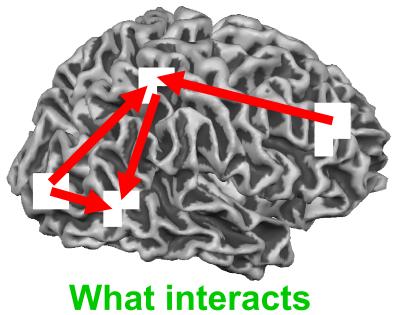
#### **Effective connectivity**



# **Effective connectivity**

# Structural model& priors

- ROI selection
- Graph selection



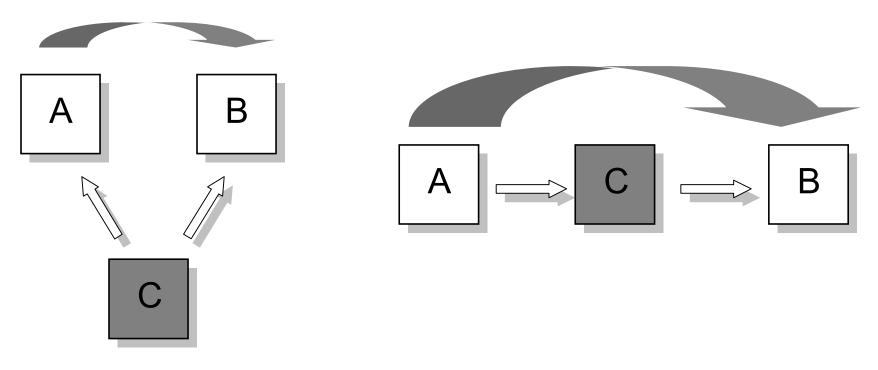
Dynamical model& priors

- Deterministic vs.
  stochastic models
- Linear vs. non-linear
- Forward observation models

 $\begin{pmatrix} x[t] \\ y[t] \end{pmatrix} = \sum_{i=1}^{p} \mathbf{A}_{i} \begin{pmatrix} x[t-i] \\ y[t-i] \end{pmatrix} + \begin{pmatrix} e_{x|y} \\ e_{y|x} \end{pmatrix} \quad \operatorname{cov} \begin{pmatrix} e_{x|y} \\ e_{y|x} \end{pmatrix} = \begin{pmatrix} \sigma_{x|y}^{2} & \sigma_{xy} \\ \sigma_{xy} & \sigma_{y|x}^{2} \end{pmatrix} = \Sigma$ 

How does it interact: signal model

#### **Problem: spurious influence**



- Danger of strong structural models:
- When important regions are 'left out' (of the anatomical model), ANY correct method will give 'wrong' answers

# Overview

- fMRI signal & connectivity
- Functional & Effective connectivity
- Structural model & Dynamical model
   Identification & model selection

#### Granger causality & fMRI

- Granger causality and its variants
- Granger causality mapping
- Issues with variable hemodynamics
   Hemodynamic deconvolution

 $\begin{aligned} & \mathbf{Granger \, causality} \\ & \mathbf{(G-causality)} \\ & \begin{pmatrix} x[t] \\ y[t] \end{pmatrix} = \sum_{i=1}^{p} \mathbf{A}_{i} \begin{pmatrix} x[t-i] \\ y[t-i] \end{pmatrix} + \begin{pmatrix} e_{x|y} \\ e_{y|x} \end{pmatrix} \quad \operatorname{cov} \begin{pmatrix} e_{x|y} \\ e_{y|x} \end{pmatrix} = \begin{pmatrix} \sigma_{x|y}^{2} & \sigma_{xy} \\ \sigma_{xy} & \sigma_{y|x}^{2} \end{pmatrix} = \Sigma \end{aligned}$ 

- Predictions are quantified with a linear multivariate autoregressive (AR) model
  - Though not necessarily: non-linear AR or nonparametric (e.g. Dhamala et al., NI, 2008)
- AR Transfer function form gives frequency distribution

#### Various normalizations

- Geweke's decomposition (Geweke, 1982; Roebroeck, NI, 2005)
- Directed transfer function (DTF; Blinowska, PhysRevE, 2004; Deshpande, NI, 2008)
- Partial directed coherence (PDC; Sameshima, JNeuSciMeth, 1999; Sato, HBM, 2009)

# **Sampling & Hemodynamics**

Influence ..... 0.8 Power 0.6 0.2 100 Delay(ms) 50 0.8 0.6 0.4 Sample intv.(s) 0.2 0 0 ? Х Sample Sample

Granger causality analysis

Roebroeck, NI 2005

#### Structural model for GC

ROI-based as in SEM, DCM

E.g. Stilla, 2007; Sridharan, 2008;
 Udaphay, 2008; Deshpande, 2008

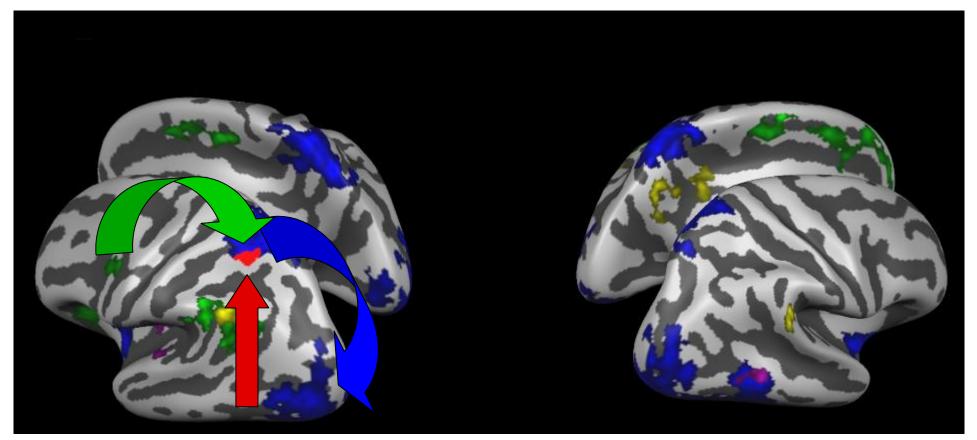
 Massively multivariate based on parcelation of the cortex

- Valdes Sosa, 2004, 2005

Granger causality mapping

 Massively bivariate without prior anatomical asumptions

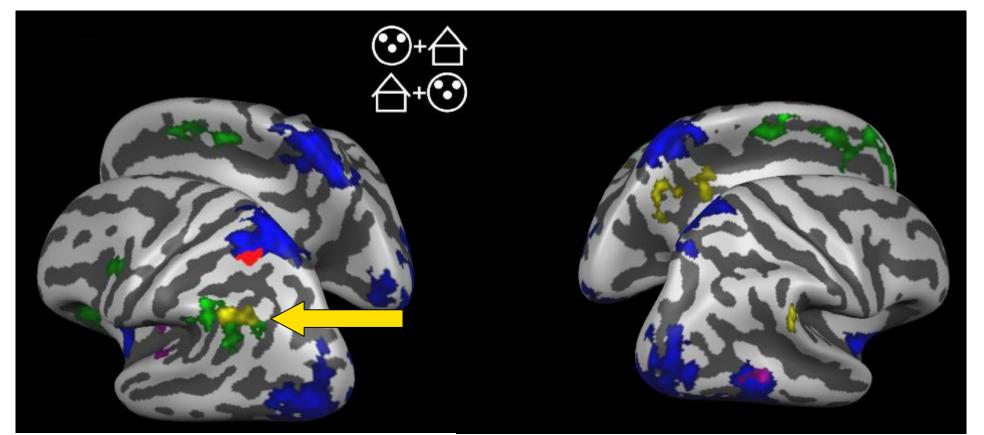
# Granger causality mapping (GCM)

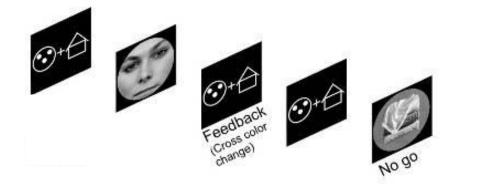


#### Random effects level GCMs

Roebroeck, NI 2005; Goebel, MRI 2004

# Granger causality mapping (GCM)





#### Experimental modulation:

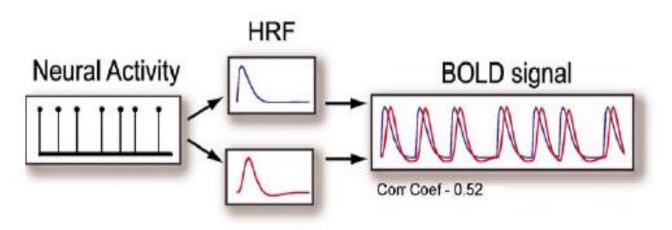
- Functional assignment
- Avoid HRF confound

Roebroeck, NI 2005; Goebel, MRI 2004

# Overview

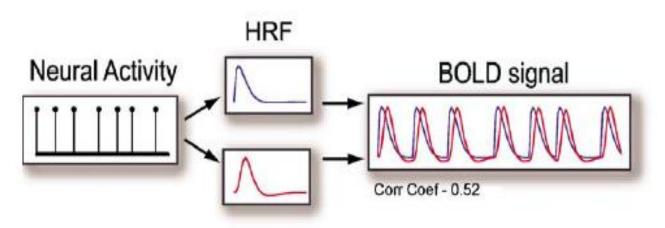
- fMRI signal & connectivity
- Functional & Effective connectivity
- Structural model & Dynamical model
   Identification & model selection
- Granger causality & fMRI
  - Granger causality and its variants
  - Granger causality mapping
- Issues with variable hemodynamics
  - Hemodynamic deconvolution

# Hemodynamics & GC



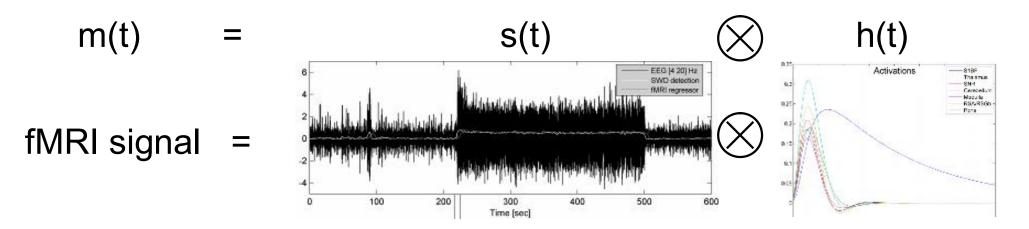
- GC could be due purely to differences in hemodynamic latencies in different parts of the brain
- Which are estimated to be in the order of 100's - 1000's ms (Aguirre, NI, 1998; Saad, HBM, 2001)

# Hemodynamics & GC



- Caution needed in applying and interpreting temporal precedence
- Tools:
  - Finding experimental modulation of GC
  - Studying temporally integrated signals for slow processes (e.g. fatigue; Deshpande, HBM, 2009)
  - Combining fMRI with EEG or MEG
  - Hemodynamic deconvolution

#### Hemodynamic deconvolution

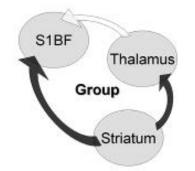


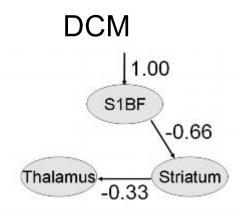
- Deconvolve neuronal source signal s(t) and hemodynamic response h(t) from fMRI signal
  - E.g. by wiener deconvolution (Glover, NI, 1999)
- Only possible if:
  - Strong constraints on s(t) are assumed (e.g. DCM: stimulus functions), or
  - An independent measure of s(t) is available (e.g. simultaneous EEG) and EEG/fMRI coupling can be assumed

#### Hemodynamic deconvolution

- Rat study of epilepsy
- Simultaneous fMRI/EEG
- Gold standard model =:

Granger without deconvolution





#### Granger using deconvolution

Striatum

0.04

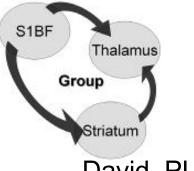
Thalamus

0.04

-0.04

0

Time [s]



p<0.002

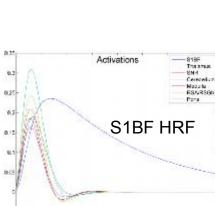
2 4 Time [s]

S1BF

p>0.3

0.04

-0.04





#### Summary

- G-causality and AR models are powerful tools in fMRI effective connectivity analysis
- GC is ideal for massive exploration
  of the structural model
- Caution is needed with GC in the face of variable hemodynamics