Dynamic causal modelling

of brain-behaviour relationships

J. Daunizeau

Brain and Spine Institute, Paris, France Wellcome Trust Centre for Neuroimaging, London, UK

Overview

- ✓ DCM: introduction
- ✓ Augmenting DCM with behavioural outputs
- ✓ Proof of concept: inhibitory control

Overview

✓ DCM: introduction

✓ Augmenting DCM with behavioural outputs

✓ Proof of concept: inhibitory control

Brain connectivities



O. Sporns 2007, Scholarpedia

- structural connectivity
 - = presence of axonal connections
- functional connectivity
 - = statistical dependencies between regional time series
- effective connectivity
 - = causal (directed) influences between neuronal populations

! connections are recruited in a *context-dependent* fashion

Functional segregation / integration

localizing brain activity: functional segregation

effective connectivity analysis: functional integration



 $u_1 X u_2$



« Where, in the brain, did my experimental manipulation have an effect? »

« How did my experimental manipulation propagate through the network? »

DCM for fMRI: example



Dynamical systems theory



 Δx $\xrightarrow{\Delta t \to 0} \dot{x}$ Δt

, time

2

3

Evolution and observation mappings



System identification: agnostic neural dynamics



nonlinear state equation:

$$\dot{x} = \left(A + \sum_{i=1}^{m} u_i B^{(i)} + \sum_{j=1}^{n} x_j D^{(j)}\right) x + Cu$$

The neuro-vascular coupling



Parametric statistical approach

DCM: model structure

$$\begin{cases} y = g(x, \varphi) + \varepsilon \\ \dot{x} = f(x, u, \theta) \end{cases}$$

likelihood $\Rightarrow p(y|\theta, \varphi, m)$



• DCM: Bayesian inference

parameter estimates:
$$\hat{\theta} = \int \theta p(y|\theta, \varphi, m) p(\theta|m) p(\varphi|m) d\theta d\varphi$$

model evidence:

$$p(y|m) = \int p(y|\theta,\varphi,m) p(\theta|m) p(\varphi|m) d\varphi d\theta$$

The variational Bayesian approach

$$\ln p(y|m) = \underbrace{\left\langle \ln p(\vartheta, y|m) \right\rangle_q + S(q) + D_{KL}(q(\vartheta); p(\vartheta|y,m)) \right\rangle}_{q}$$

free energy : functional of q

mean-field: approximate marginal posterior distributions: $\{q(\theta_1), q(\theta_2)\}$





✓ DCM: introduction

✓ Augmenting DCM with behavioural outputs

✓ Proof of concept: inhibitory control

Identifying the brain-behaviour mapping



✓ modelling the brain input-output transform (through the network)

✓ decomposing the relative contribution of brain regions and their interactions to the behavioural response

Identifying the brain-behaviour mapping



$$p(o|x) = s(r)^{o} (1 - s(r))^{r o}$$
$$r(t) = \int_{-\infty}^{t} h(x(\tau), u(\tau)) e^{-\alpha \tau} d\tau \implies \dot{r} = h(x, u) - \alpha r$$

$$h(x,u) \approx h(0,0) + \frac{\partial h}{\partial x}x + \frac{\partial h}{\partial u}u + \frac{\partial^2 h}{\partial x \partial u}ux + \frac{\partial^2 h}{\partial x^2}\frac{x^2}{2} + \dots$$

bDCM: face validity



bDCM: face validity



bDCM: behavioural susceptibility analysis



bDCM: predicting the effect of lesions





- ✓ DCM: introduction
- ✓ Augmenting DCM with behavioural outputs
- ✓ Proof of concept: inhibitory control

Go/noGo: paradigm and fMRI results





Go/noGo: model comparison set



Go/noGo: Bayesian model selection



Go/noGo: behavioural fit







Go/noGo: behavioural susceptibility analysis



stop (NoGo)

execution (Go)

preparation











left response

right response

Go/noGo: lesion-induced behavioural deficits



Overview

- ✓ DCM: introduction
- ✓ Augmenting DCM with behavioural outputs
- ✓ Proof of concept: inhibitory control

Many thanks to Lionel Rigoux