

Stochastic Dynamic Causal Modelling

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Overview

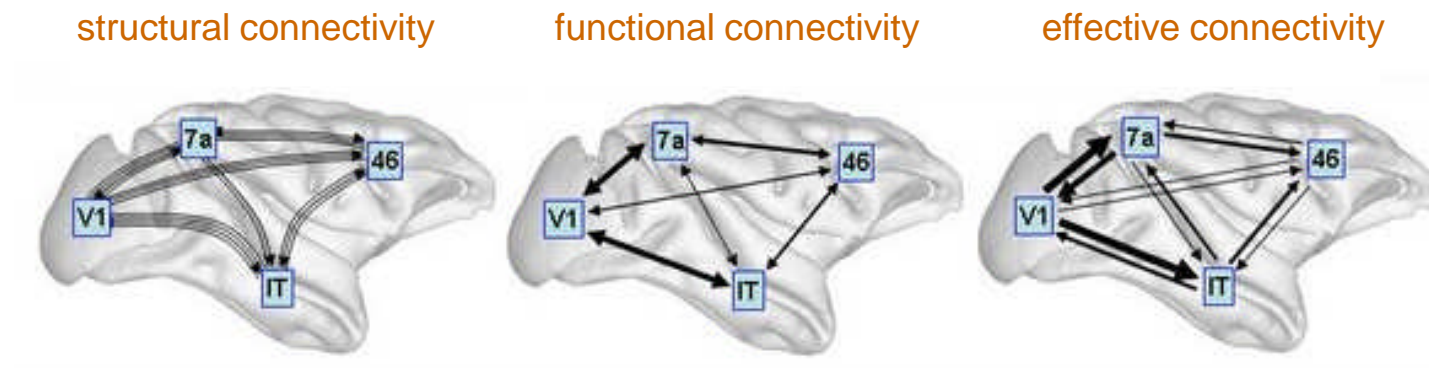
- 1 Dynamic Causal Modelling (DCM)
- 2 System's states dynamics
- 3 Stochastic DCM: a variational Bayesian approach
- 4 VB evaluation
- 5 Stochastic DCM: example
- 6 Conclusion

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Dynamic Causal Modelling

structural, functional and effective connectivity



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

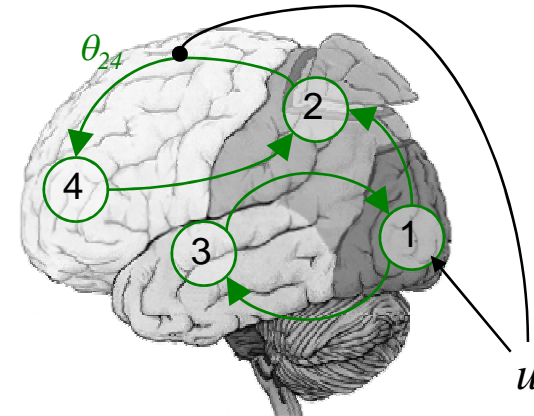
! connections are recruited in a *context-dependent* fashion

Dynamic Causal Modelling

parametric statistical approach

- DCM: model structure

$$\begin{cases} y = g(x, \varphi) + \varepsilon \\ \dot{x} = f(x, u, \theta) \end{cases} \quad \text{likelihood} \Rightarrow p(y|\theta, \varphi, m)$$



- DCM: Bayesian inference

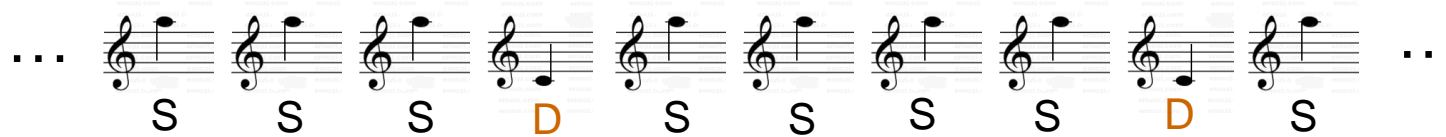
marginal posterior density: $p(\theta|y, m) \propto \int p(y|\theta, \varphi, m) p(\theta|m) p(\varphi|m) d\varphi$ priors on parameters

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

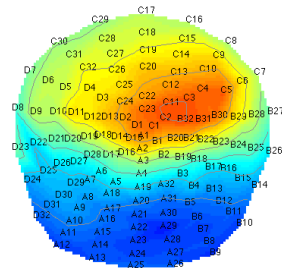
Dynamic Causal Modelling

DCM for EEG-MEG: auditory mismatch negativity

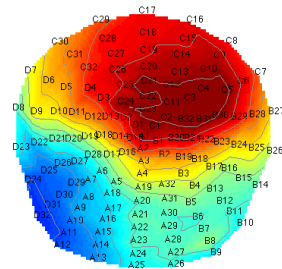
sequence of auditory stimuli



standard condition (S)

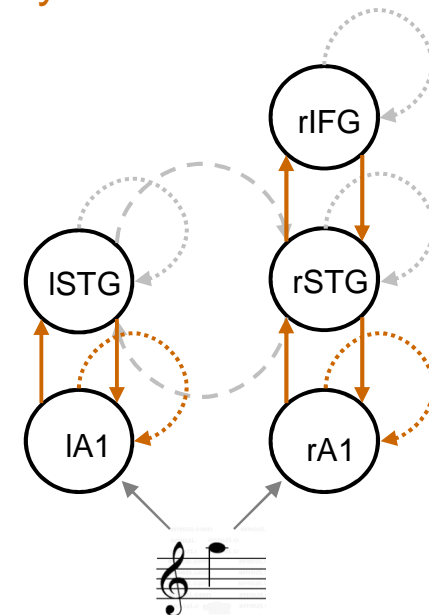
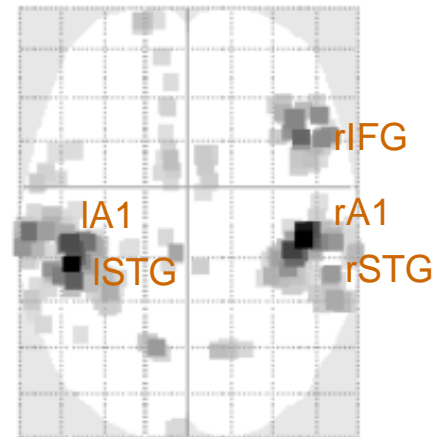


deviant condition (D)



$t \sim 200$ ms

S-D: reorganisation
of the connectivity structure



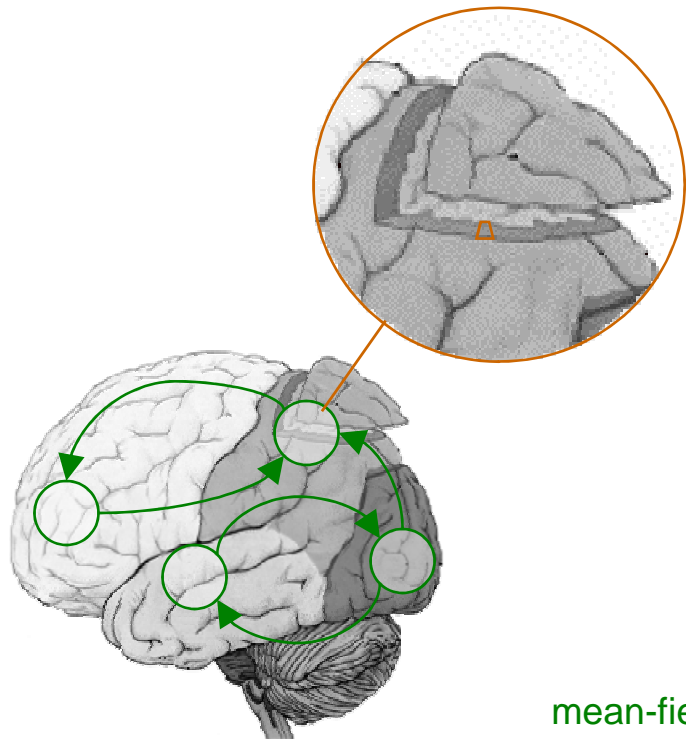
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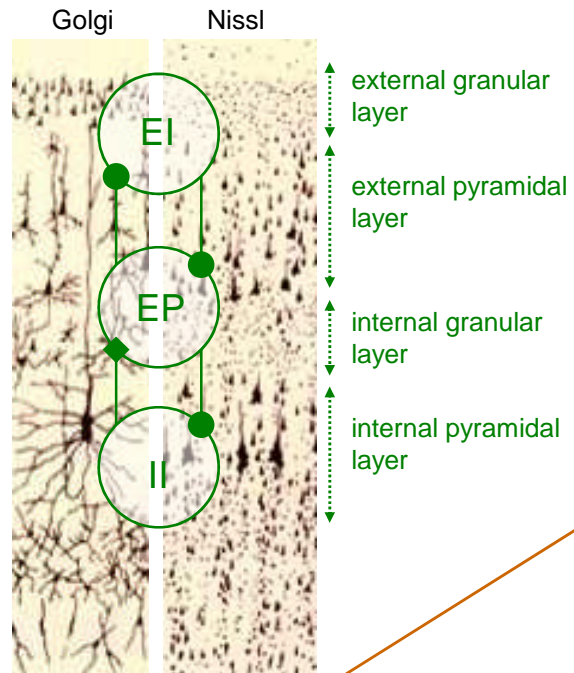
System's states dynamics

DCM for EEG/MEG: neural ensembles dynamics

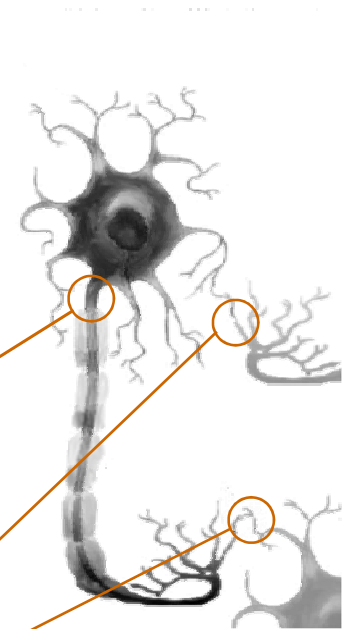
macro-scale



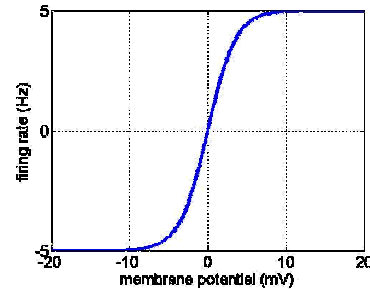
meso-scale



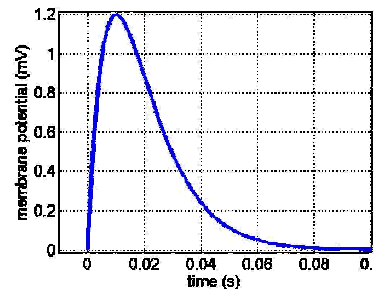
micro-scale



mean-field firing rate

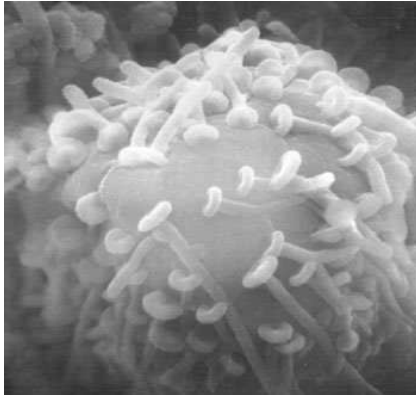


synaptic dynamics



System's states dynamics

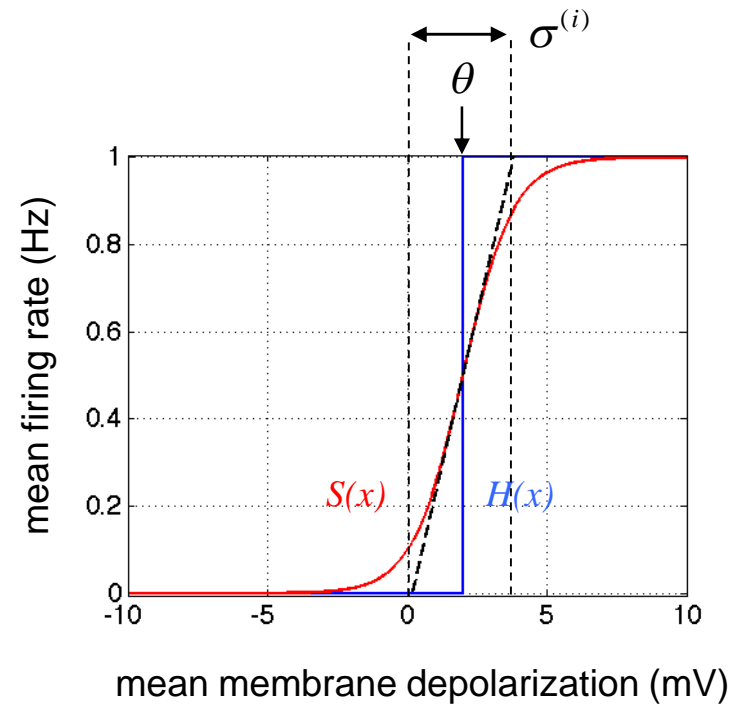
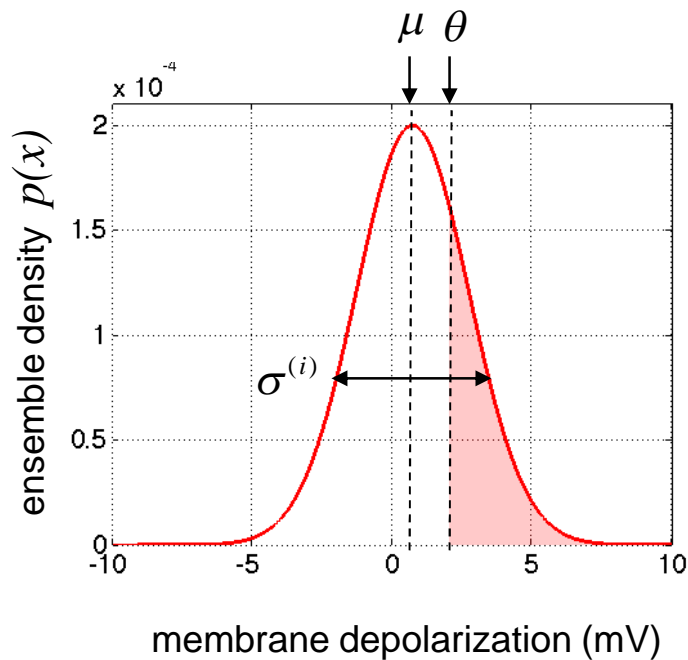
DCM for EEG/MEG: from micro- to meso-scale



$x_j(t)$: post-synaptic potential of j^{th} neuron within its ensemble

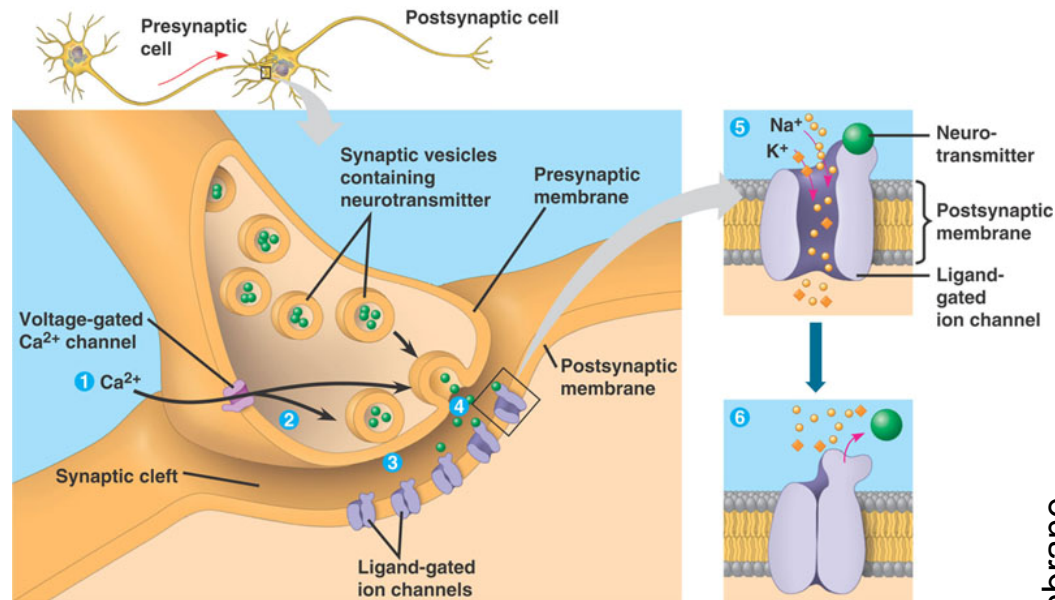
$$\frac{1}{N-1} \sum_{j' \neq j} H(x_{j'}^{(i)}(t) - \theta) \xrightarrow{N \rightarrow \infty} \int H(x^{(i)}(t) - \theta) p(x^{(i)}(t)) dx^{(i)}$$

$\approx S(\mu^{(i)})$ **mean-field firing rate**

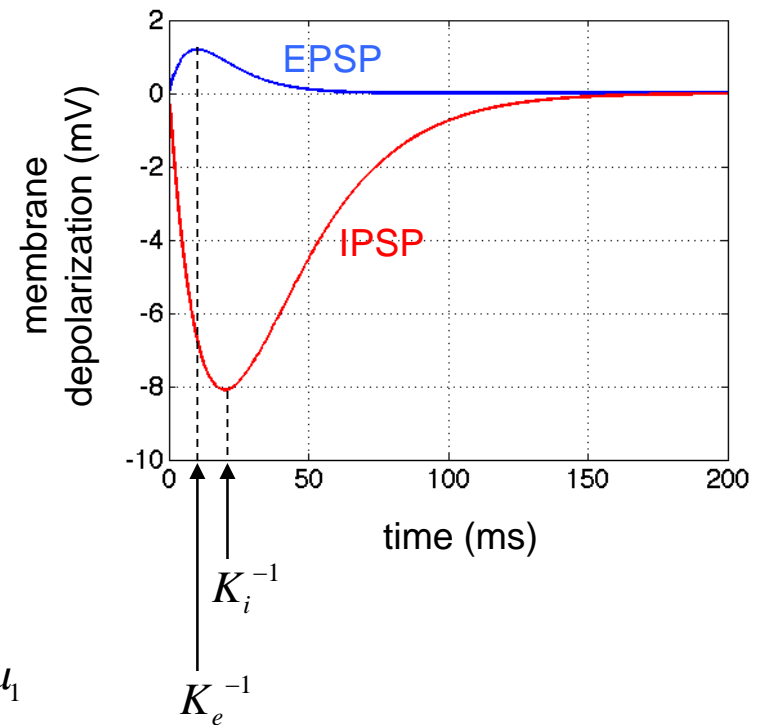


System's states dynamics

DCM for EEG/MEG: synaptic dynamics



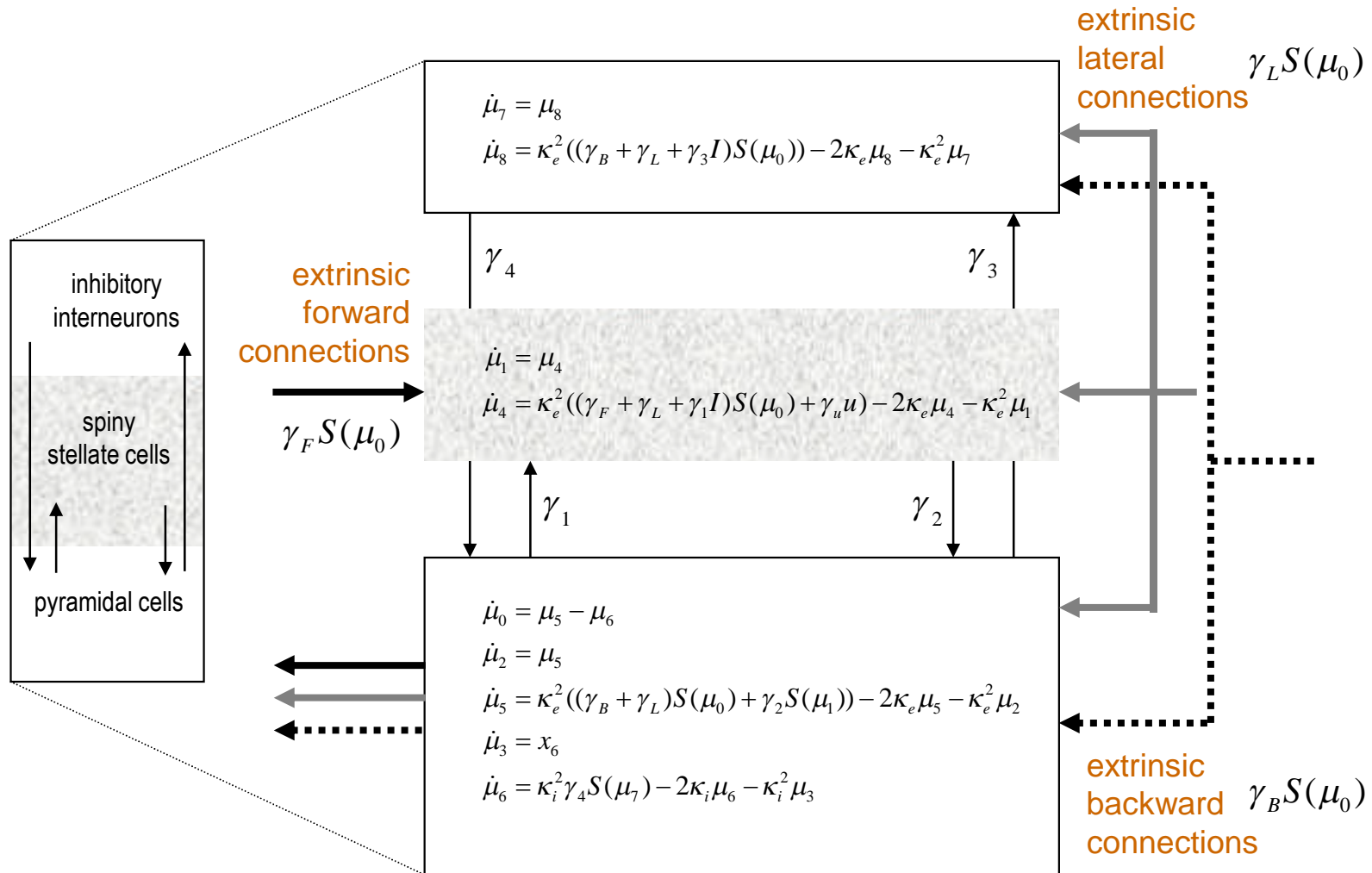
post-synaptic potential



$$\begin{cases} \dot{\mu}_1 = \mu_2 \\ \dot{\mu}_2 = \kappa_{i/e}^2 S(\cdot) - 2\kappa_{i/e} \mu_2 - \kappa_{i/e}^2 \mu_1 \end{cases}$$

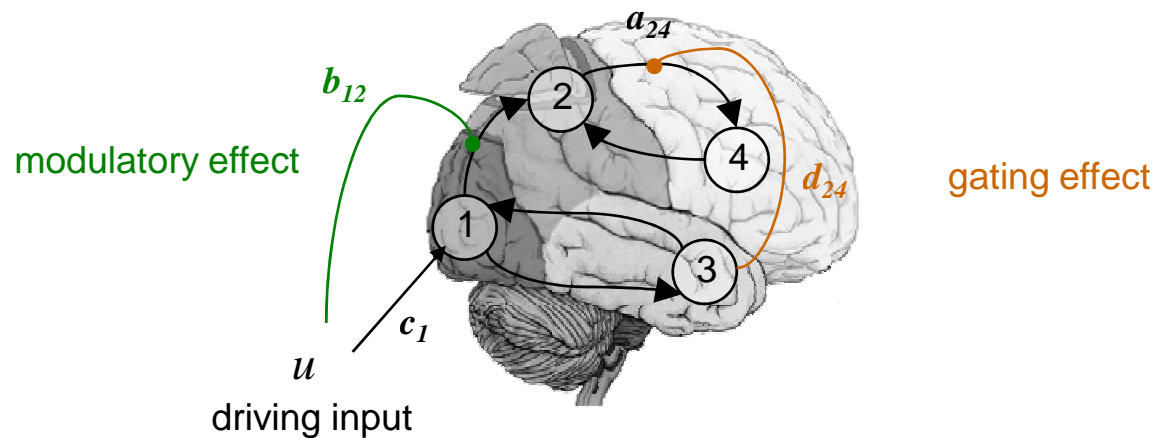
System's states dynamics

DCM for EEG/MEG: extrinsic connections



System's states dynamics

DCM for fMRI: neuronal states dynamics



$$\dot{x} = f(x, u) \approx f(x_0, 0) + \frac{\partial f}{\partial x} x + \frac{\partial f}{\partial u} u + \frac{\partial^2 f}{\partial x \partial u} ux + \frac{\partial^2 f}{\partial x^2} \frac{x^2}{2} + \dots$$

bilinear state equation:

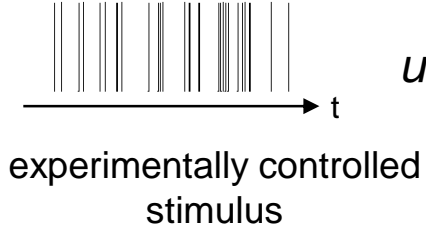
$$\frac{dx}{dt} = \left(A + \sum_{i=1}^m u_i B^{(i)} \right) x + Cu$$

nonlinear state equation:

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

System's states dynamics

DCM for fMRI: the neuro-vascular coupling

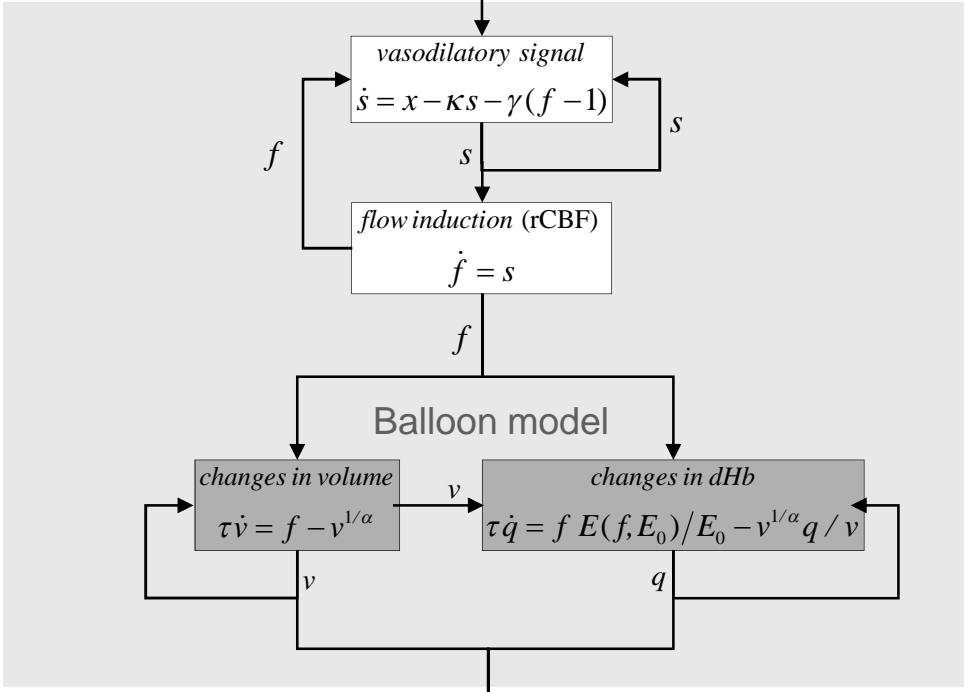


$$u \rightarrow \dot{x} = \left(A + \sum_{j=1}^m u_j B^{(j)} + \sum_{j=1}^n x_j D^{(j)} \right) x + Cu$$

neural states dynamics

$$\theta^h = \{\kappa, \gamma, \tau, \alpha, E_0, \varepsilon\}$$

$$\theta^n = \{A, B^{(i)}, C, D^{(j)}\}$$



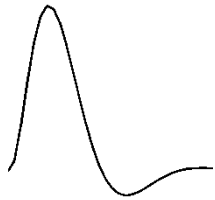
$$\lambda(q, v) = \frac{\Delta S}{S_0} \approx V_0 \left[k_1(1-q) + k_2 \left(1 - \frac{q}{v} \right) + k_3(1-v) \right]$$

$$k_1 = 4.39_0 E_0 TE$$

$$k_2 = \varepsilon r_0 E_0 TE$$

$$k_3 = 1 - \varepsilon$$

BOLD signal change observation



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stochastic DCM

stochastic processes: basics

stochastic differential equation:

$$dx_t = \underbrace{f(x_t, \theta, u_t)}_{\text{drift}} dt + \underbrace{b(x_t, \theta, u_t)}_{\text{diffusion}} d\omega_t$$

→ limit of a discrete Markov process:

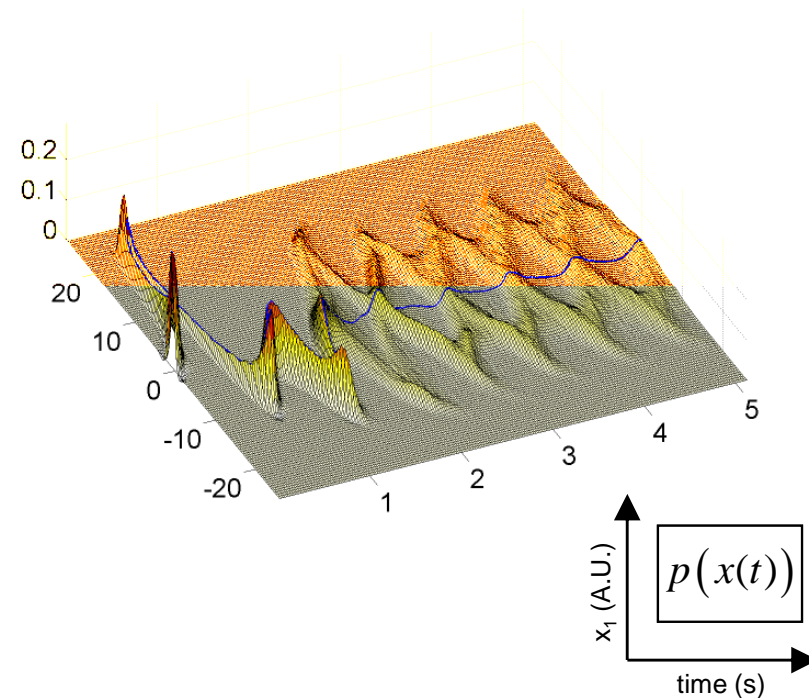
$$x_{t+1} \approx \tilde{f}(x_t) + b\sqrt{\Delta t} \eta_t$$

$$\eta_t \sim N(0, I)$$

$$\tilde{f}(x_t) = J^{-1}(\exp[J\Delta t] - I_n)a(x_t)$$

$$\xrightarrow{\Delta t \rightarrow 0} x_t + \Delta t f(x_t),$$

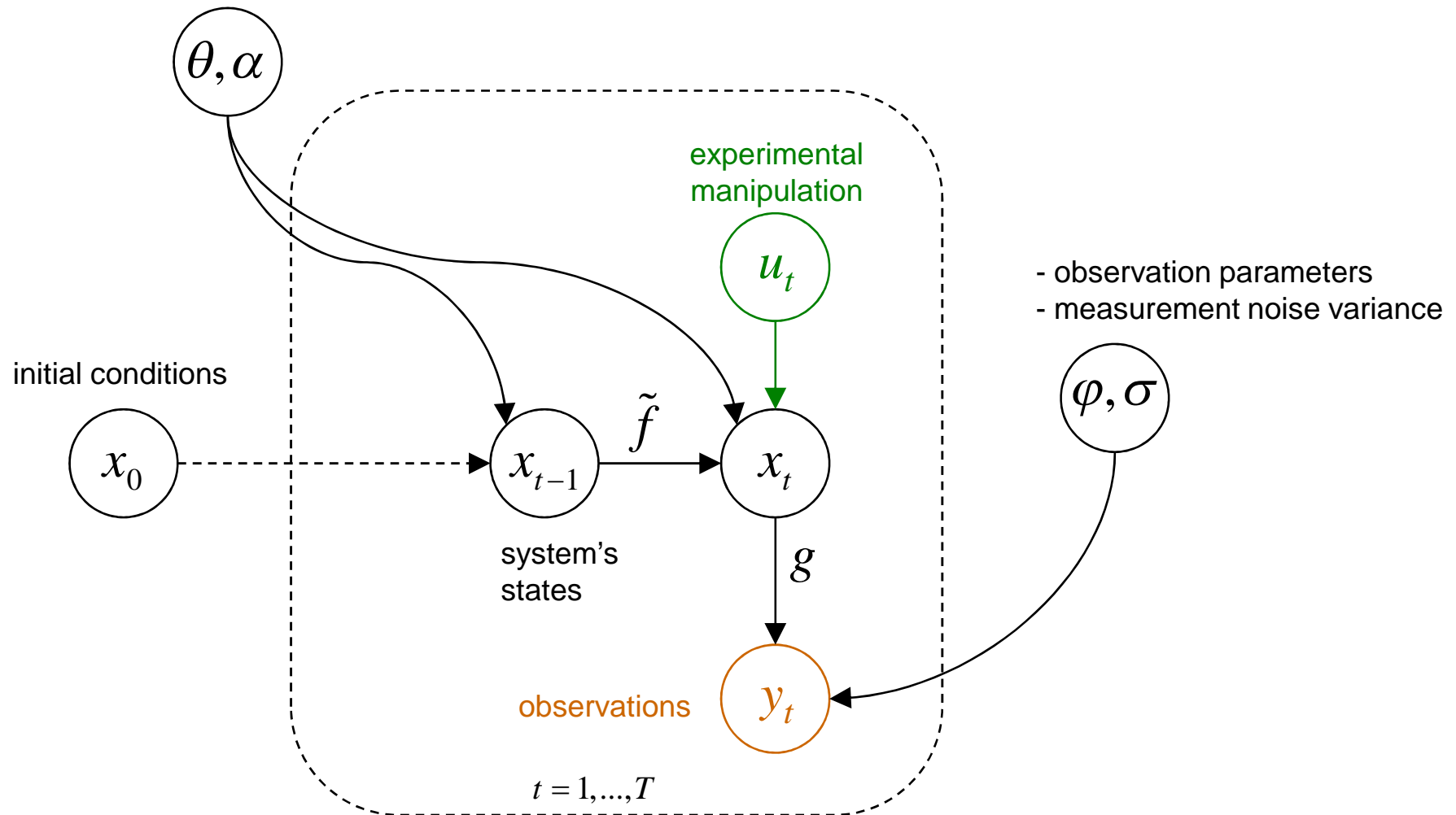
DCM for fMRI with gating (nonlinear) effects:



stochastic DCM

the generative model

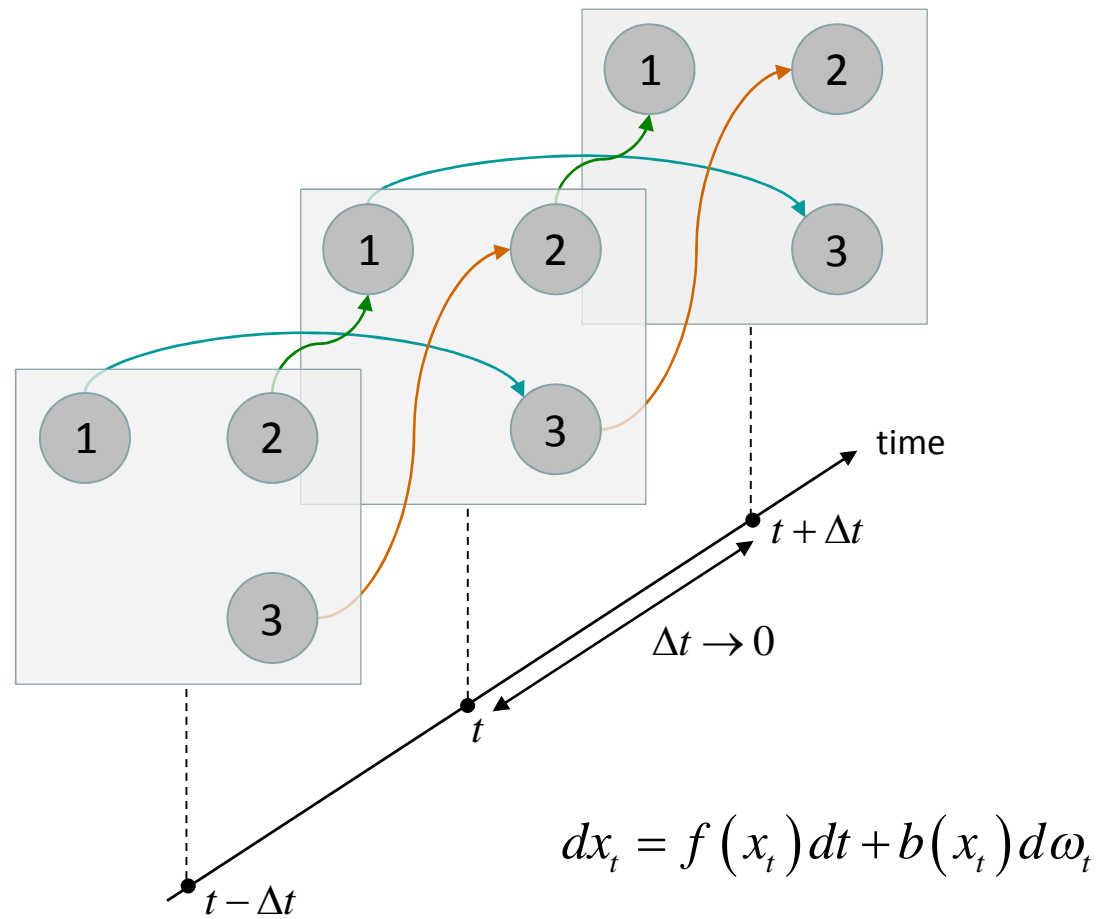
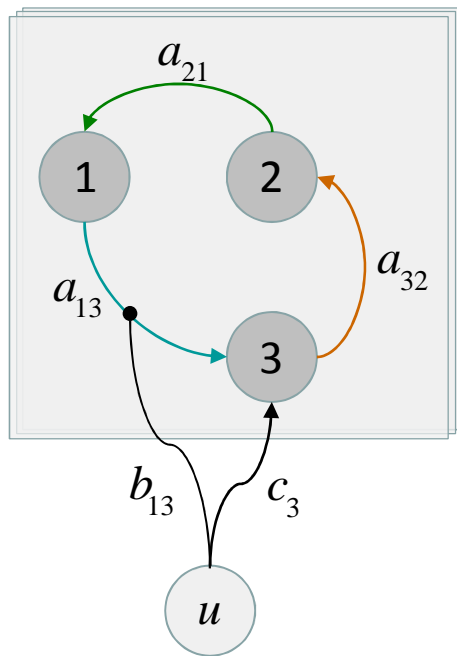
- evolution parameters
- stochastic innovations variance



stochastic DCM

a note on causality

$$u \xrightarrow{\theta} x \xrightarrow{\phi} y$$

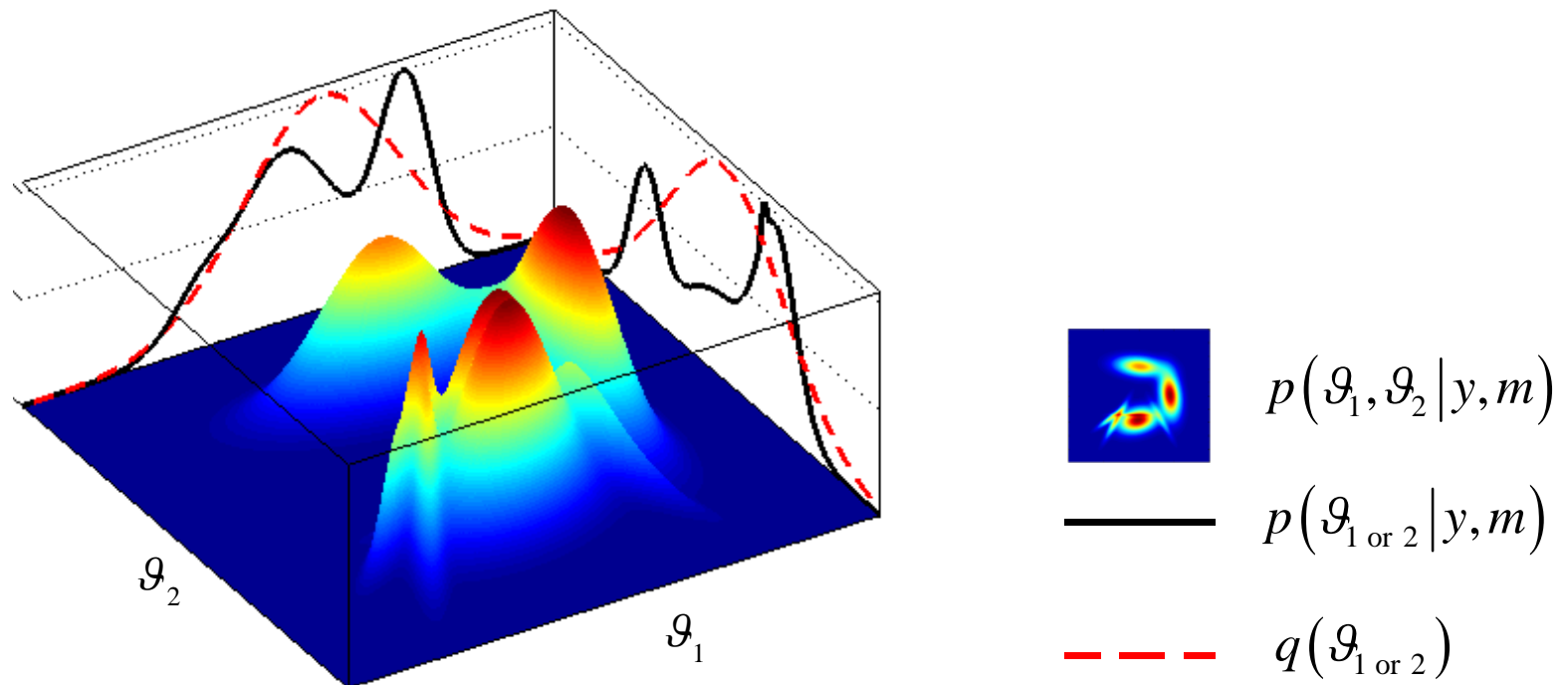


stochastic DCM

VB: mean field - Laplace approximation

$$\ln p(y|m) = \underbrace{\left\langle \ln p(\mathcal{Z}, y|m) \right\rangle_q + S(q) + D_{KL}(q(\mathcal{Z}); p(\mathcal{Z}|y, m))}_{\text{free energy : functional of } q}$$

approximate (marginal) posterior probability density functions: $\{q(\mathcal{Z}_1), q(\mathcal{Z}_2)\}$

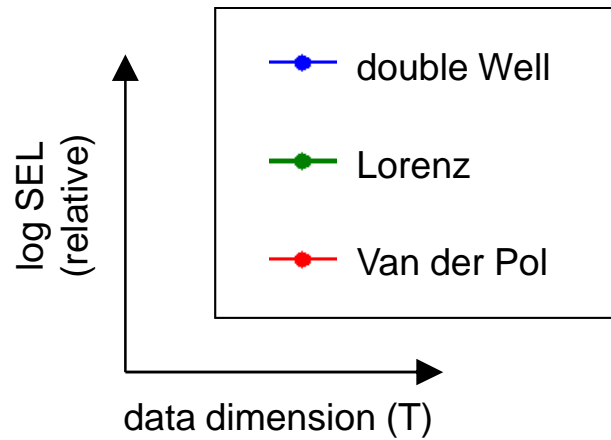
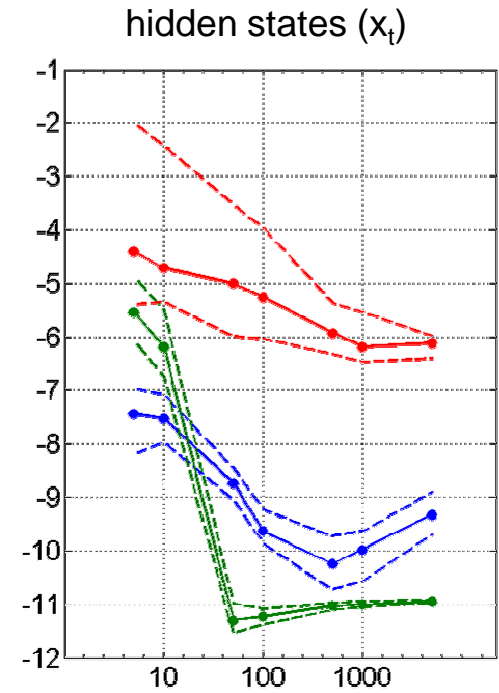
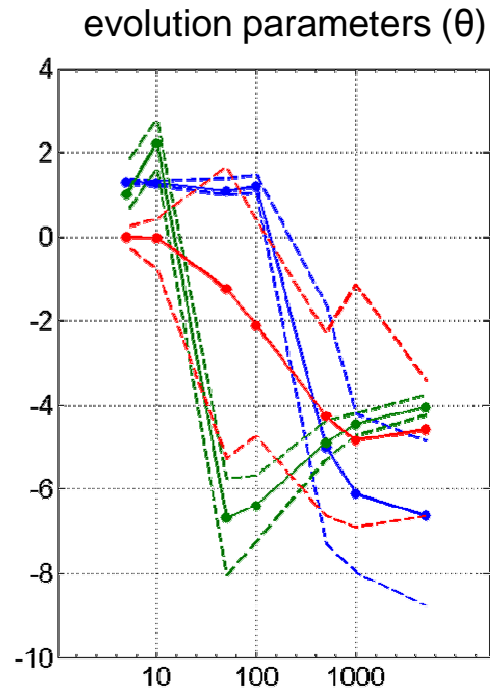
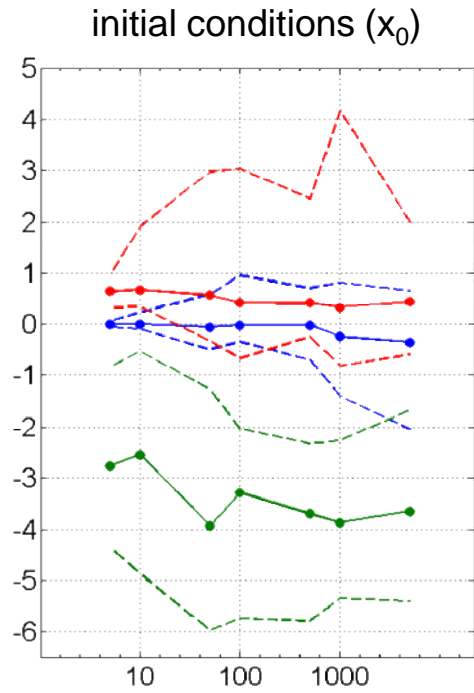


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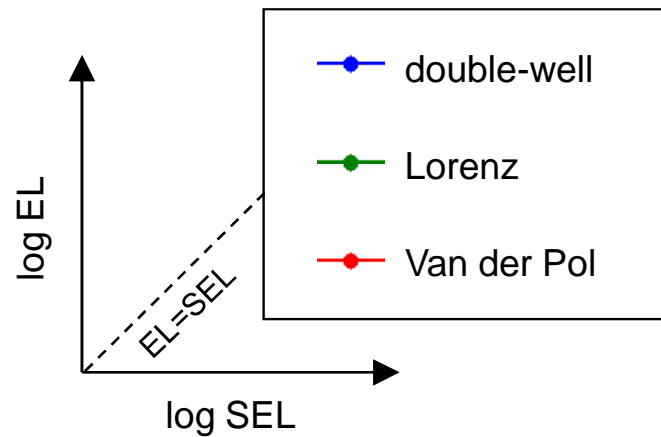
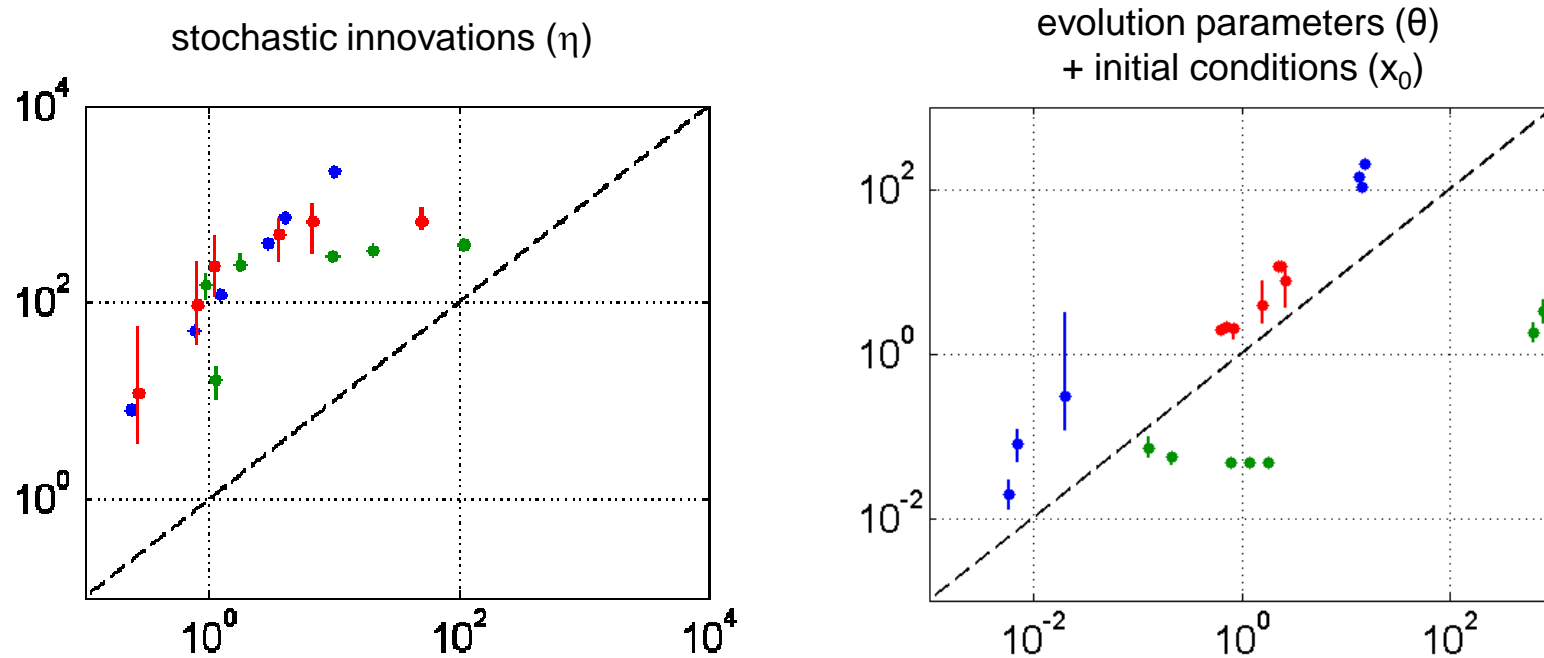
VB evaluation

asymptotic efficiency



VB evaluation

posterior confidence intervals

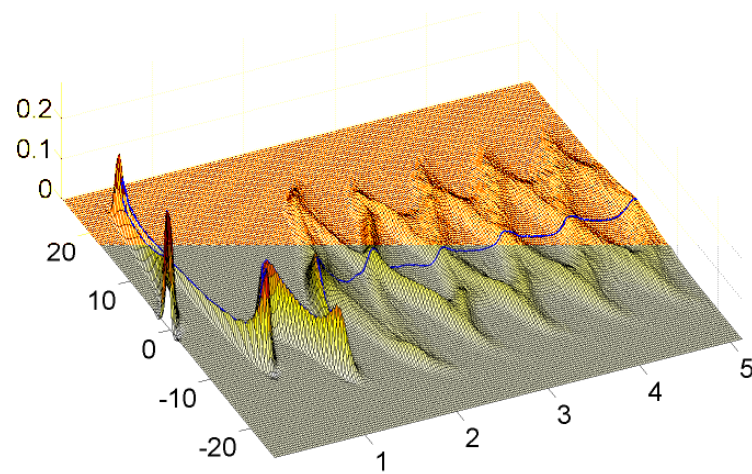


VB evaluation

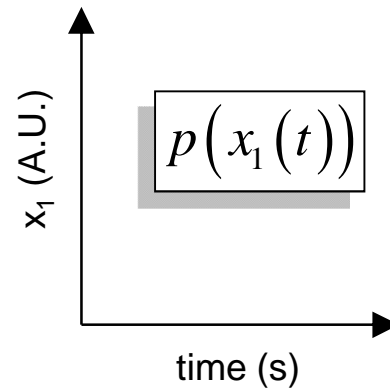
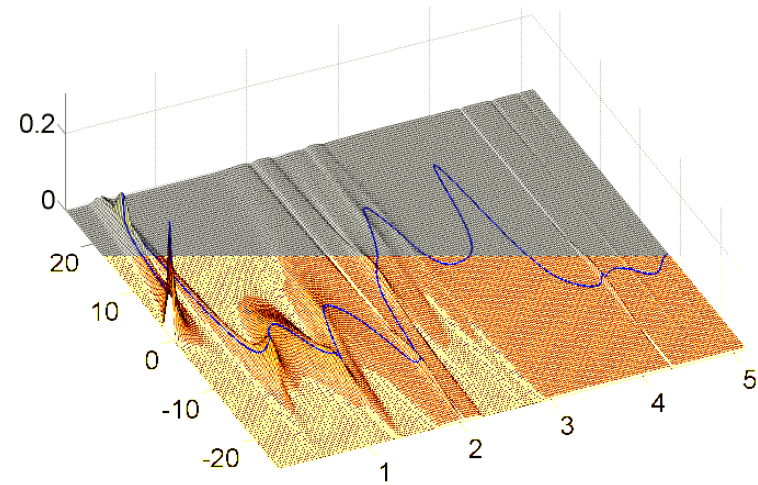
prediction ability

Lorenz system: predictive density (x_1)

MCMC

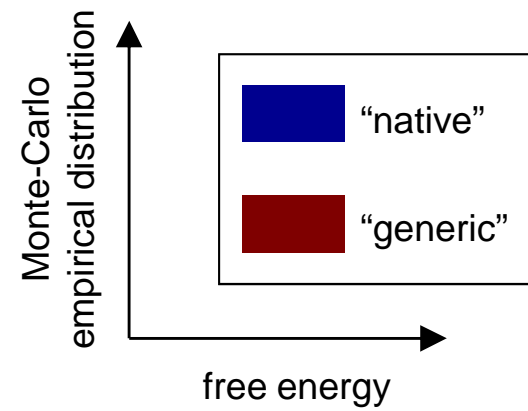
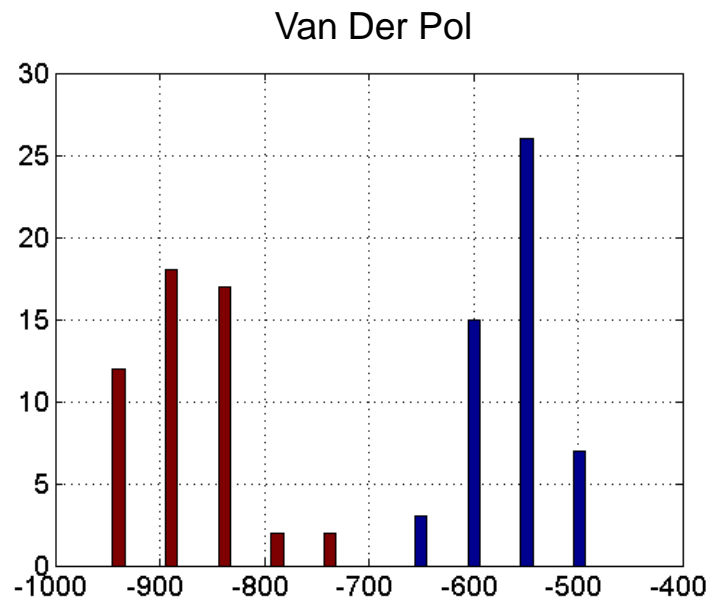
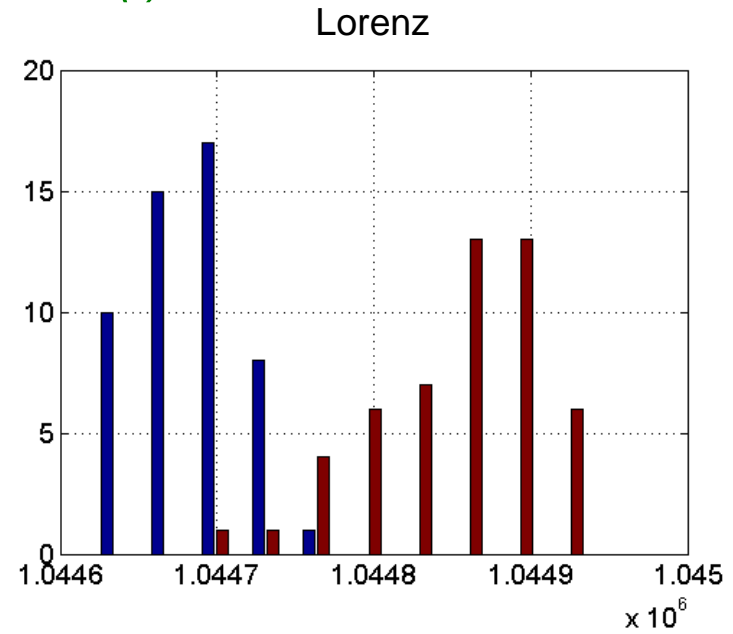
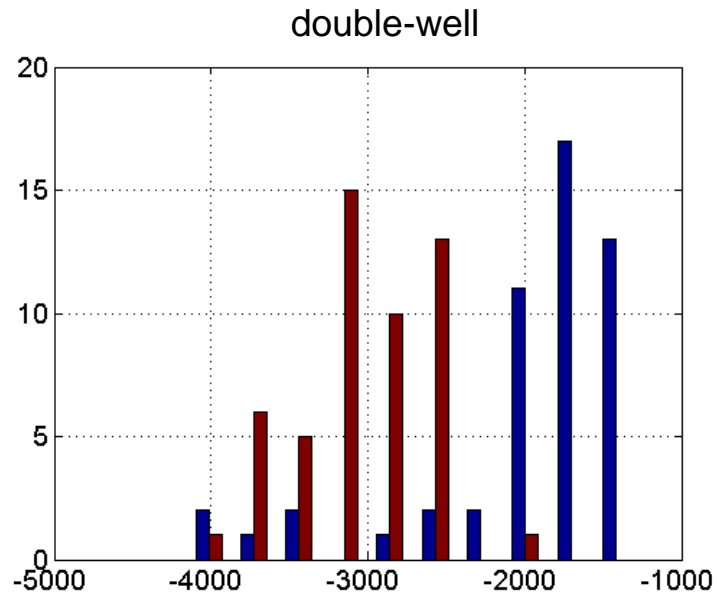


Laplace approximation



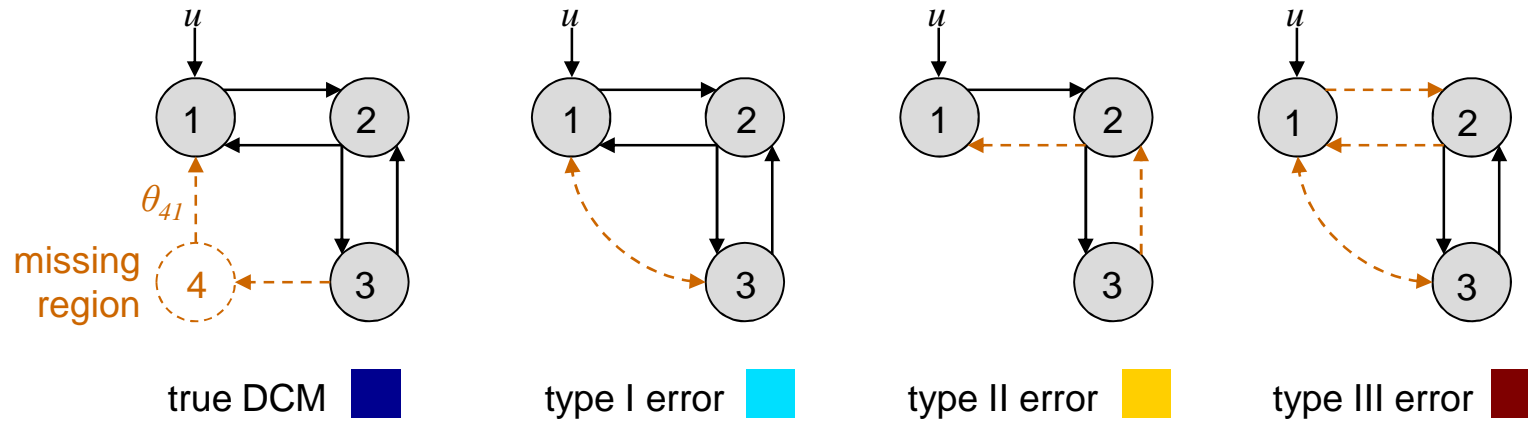
VB evaluation

model comparison (I)

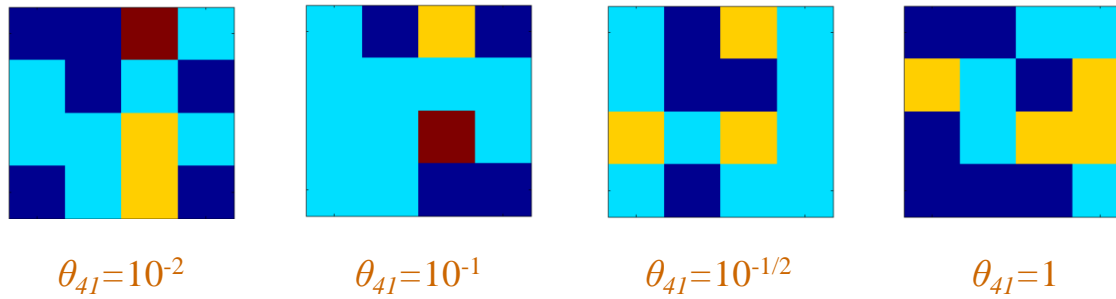


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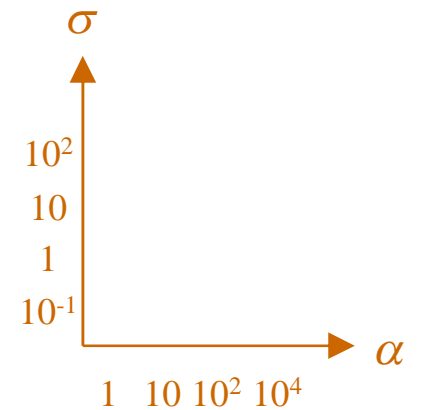
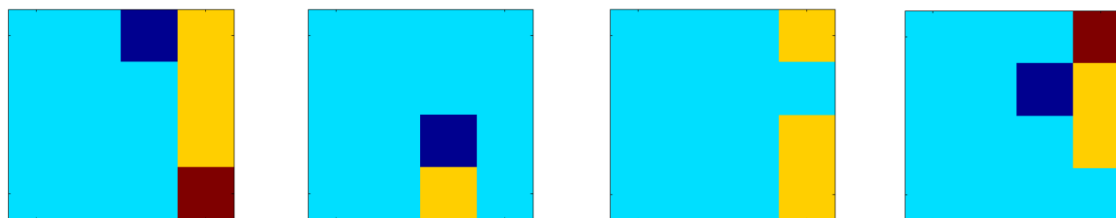
model comparison (II): robustness to missing regions



VB-Laplace for stochastic DCM



VB-Laplace for DCM



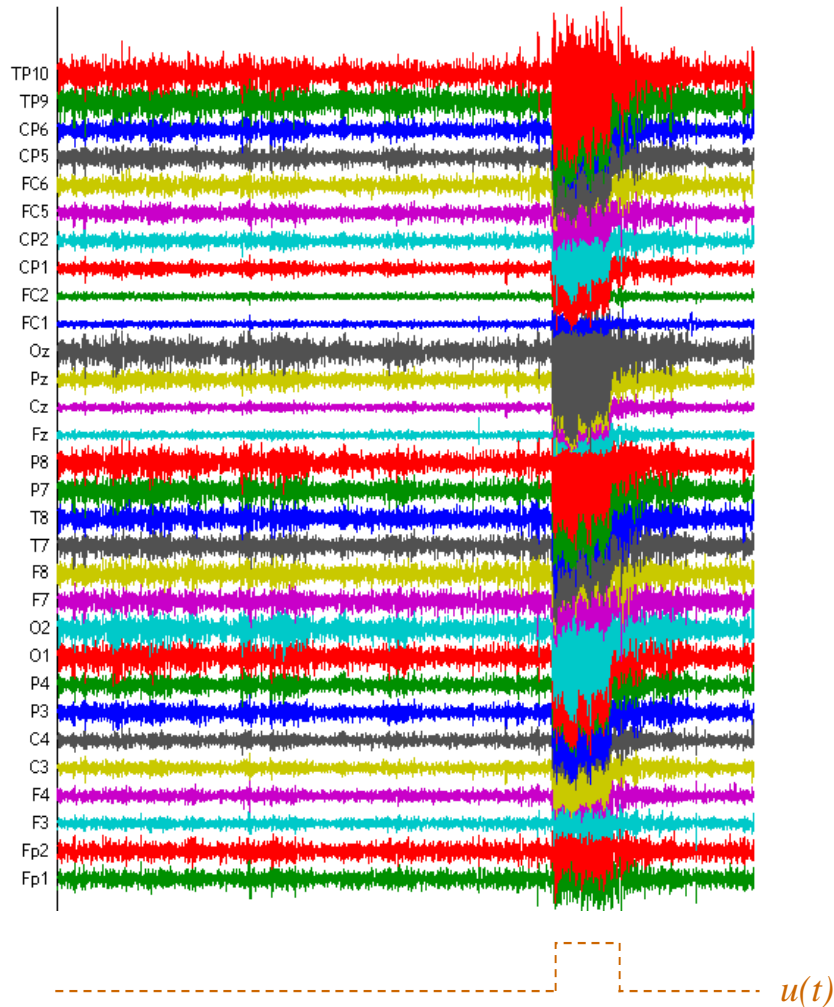
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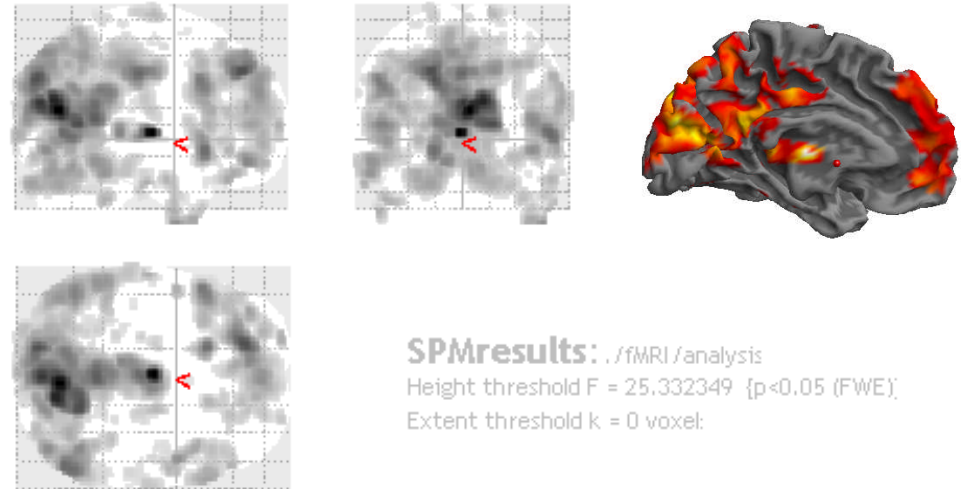
sDCM: example

epileptogenic network mapping

EEG: ictal activity



fMRI: GLM activation results

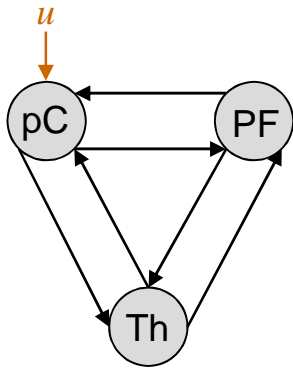


group-level consistent ROIs

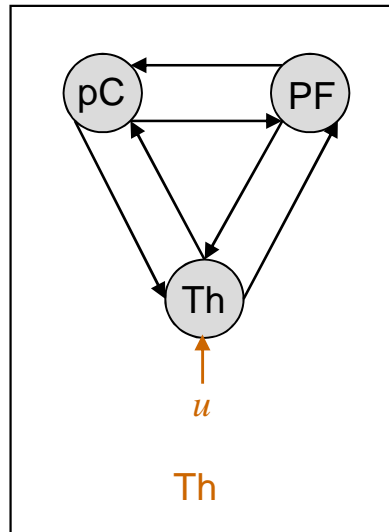


sDCM: example

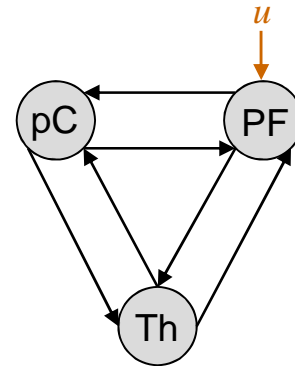
Bayesian model comparison



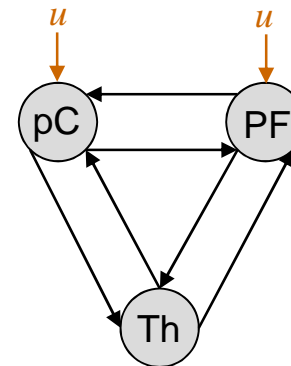
PC



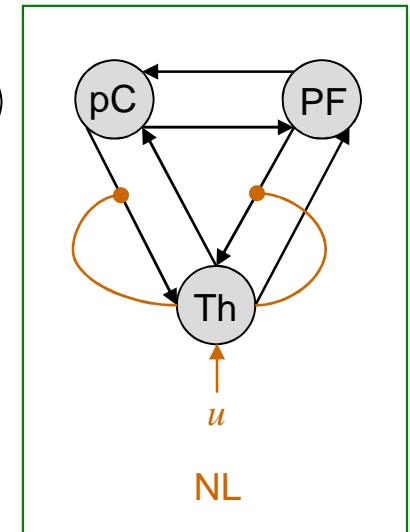
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PF

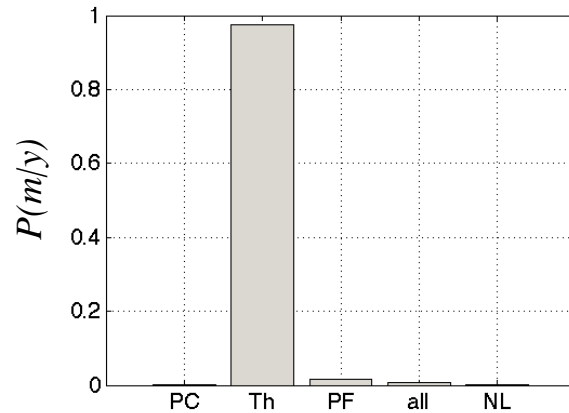


all

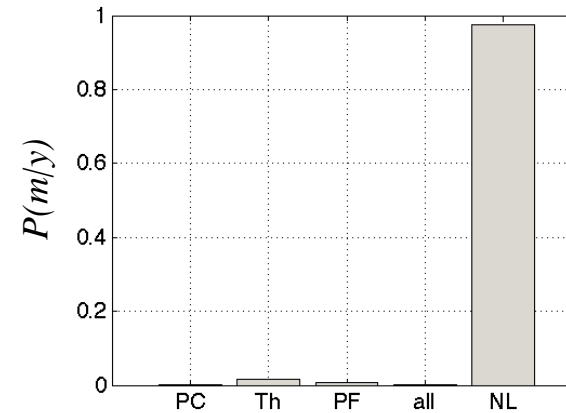


NL

deterministic DCMs



stochastic DCMs

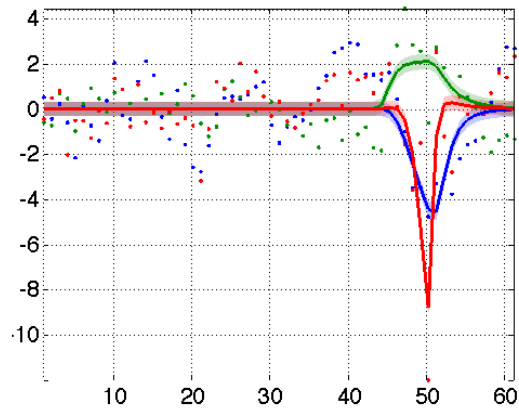


sDCM: example

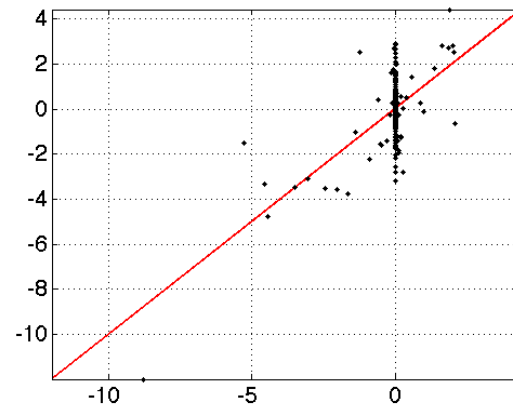
stochastic versus deterministic (most likely) DCM

posterior predictive density
 $p(g(x)/y, m)$ as a function of time

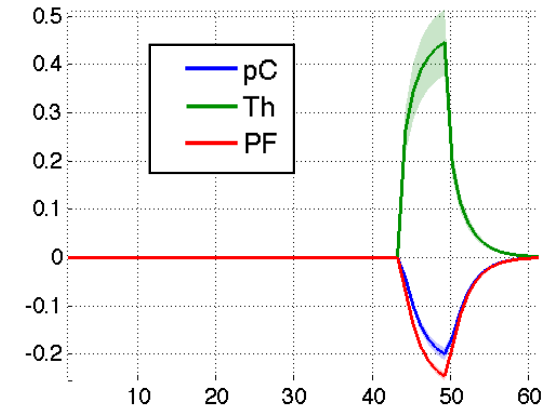
DCM



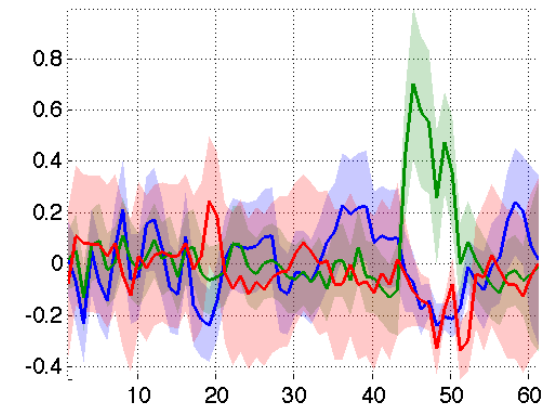
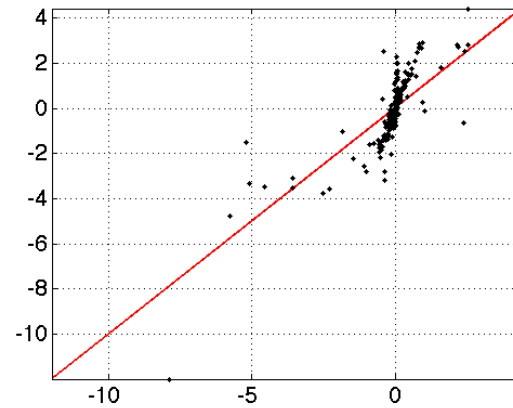
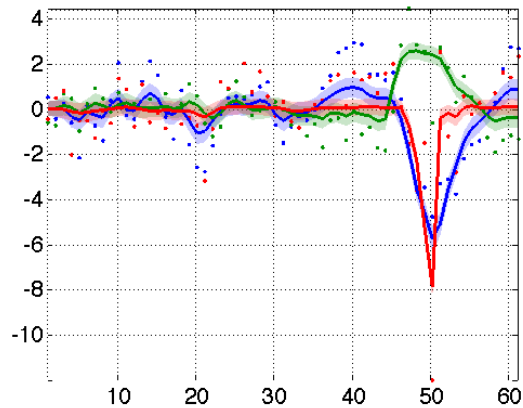
model fit:
 $\langle g(x) \rangle$ versus y



neural states dynamics:
 $\langle x \rangle$ as a function of time



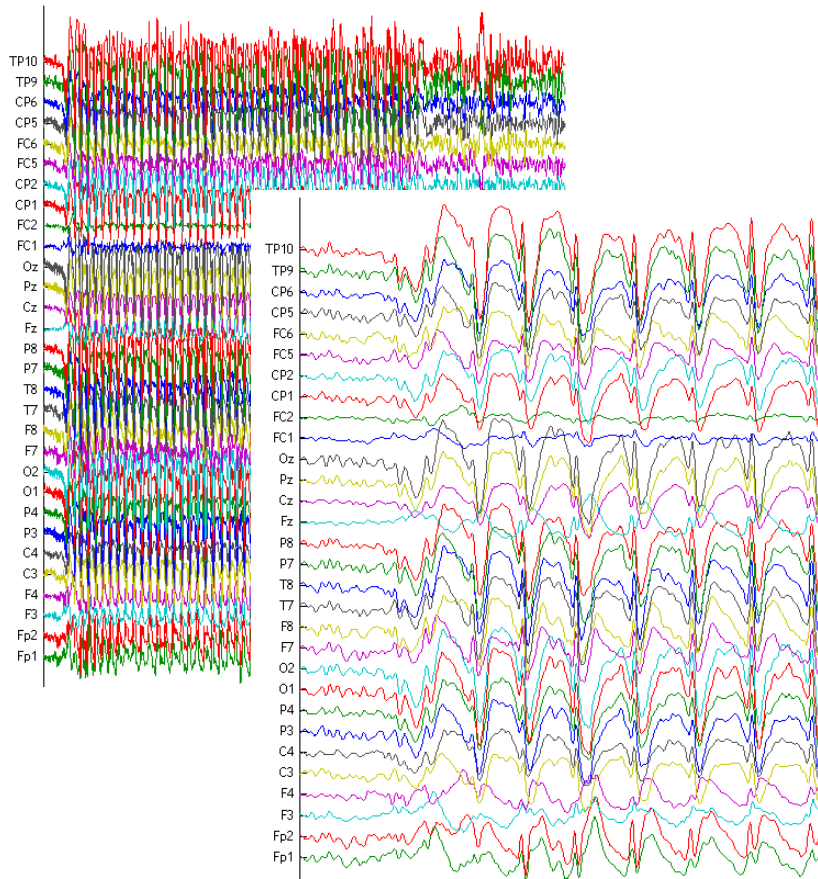
sDCM



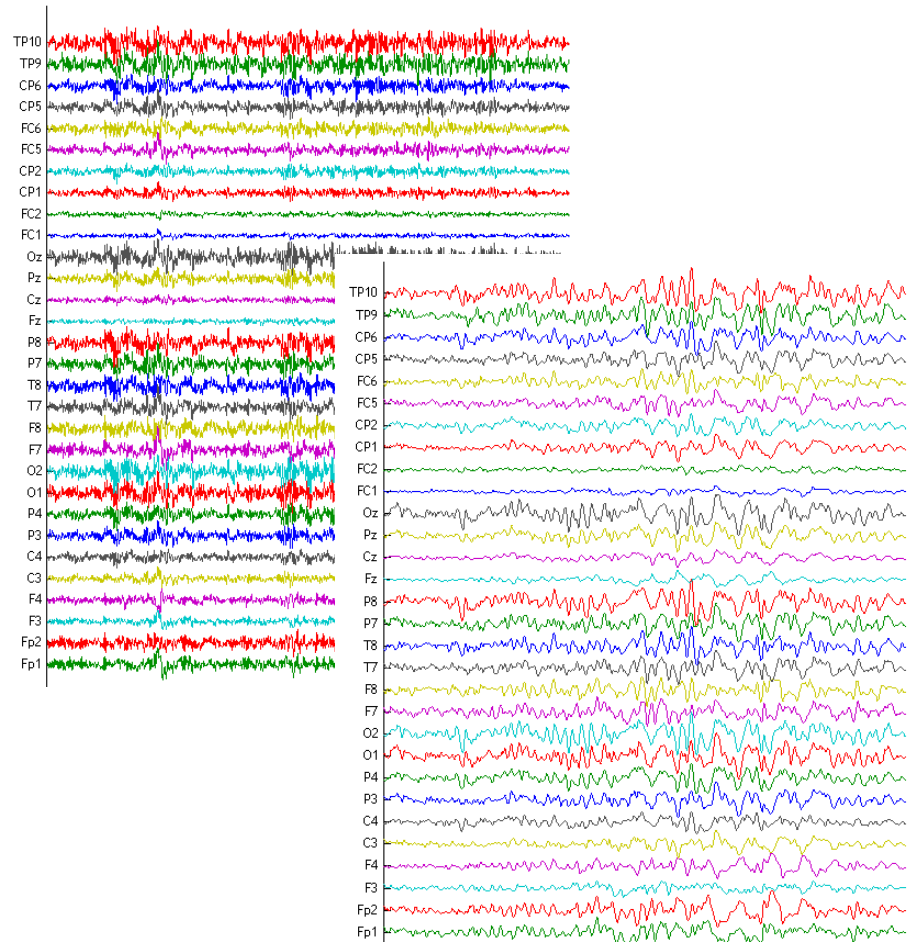
sDCM: exemple

back to the EEG

ictal activity (absence seizure)



interictal activity (no clinical sign)



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Conclusion

→ Motivations:

- model imperfections
- endogenous non-controlled activity
- robustness to the “missing region” problem

→ VB: balanced computational load/accuracy (curse of dimensionality):

- asymptotic efficiency
- posterior confidence intervals
- model comparison

→ Limitations:

- prediction: lack of robustness to violation of *quasideterminism*
- non-stationarity? (volatile models)
- estimation of “neural noise”: robustness of the backward pass?
- potential conflicts between deterministic and stochastic effects

→ What about *continuous* stochastic systems?

- augmented Kushner-Stratonovitch-Pardoux equations?
- **Dynamic Expectation-Maximization** (Friston K.; Neuroimage 2008)

many thanks to:

Karl J. Friston (London, UK)

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Klaas E. Stephan (Zurich, Switzerland)

Louis Lemieux (Chalfon, UK)