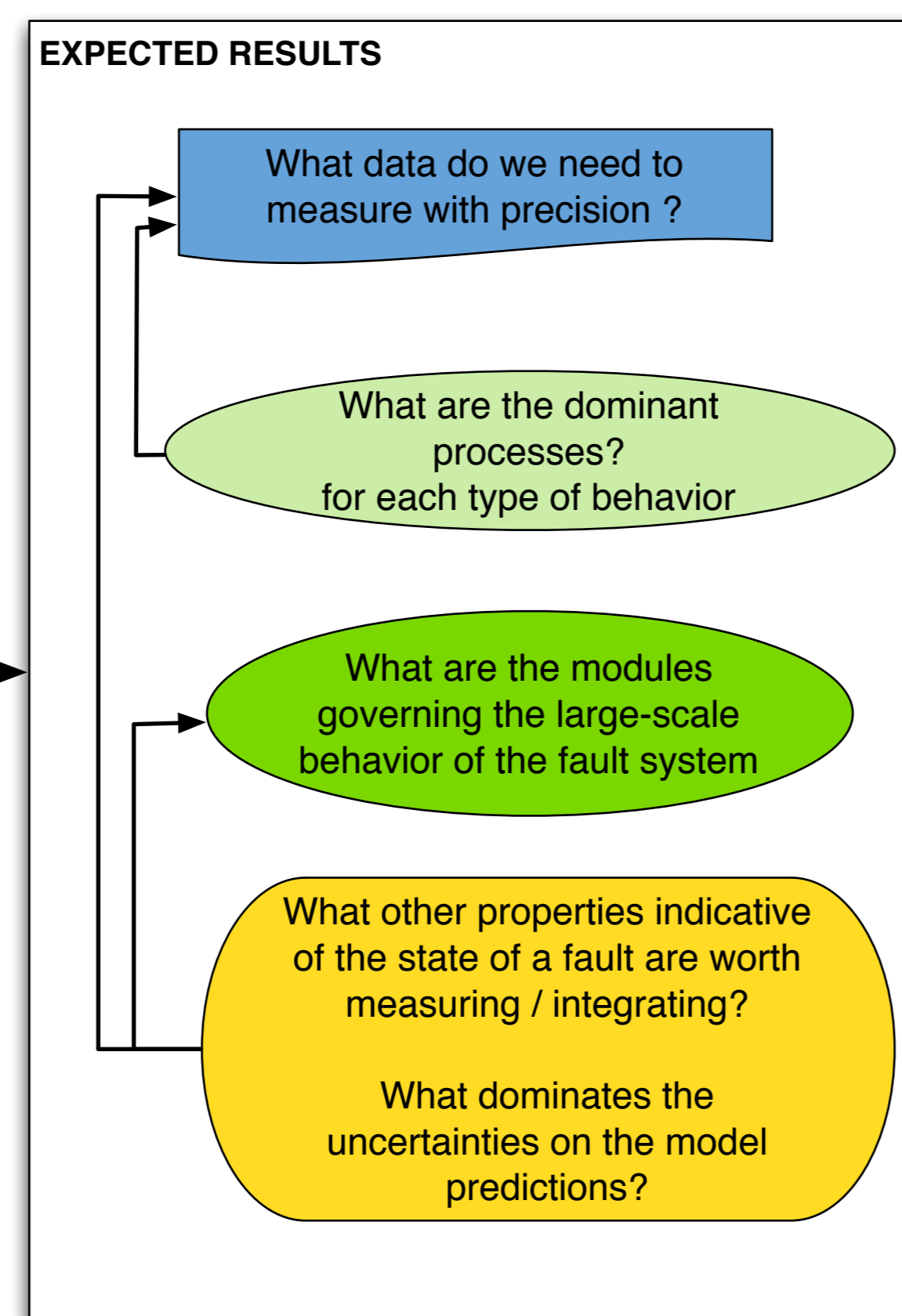
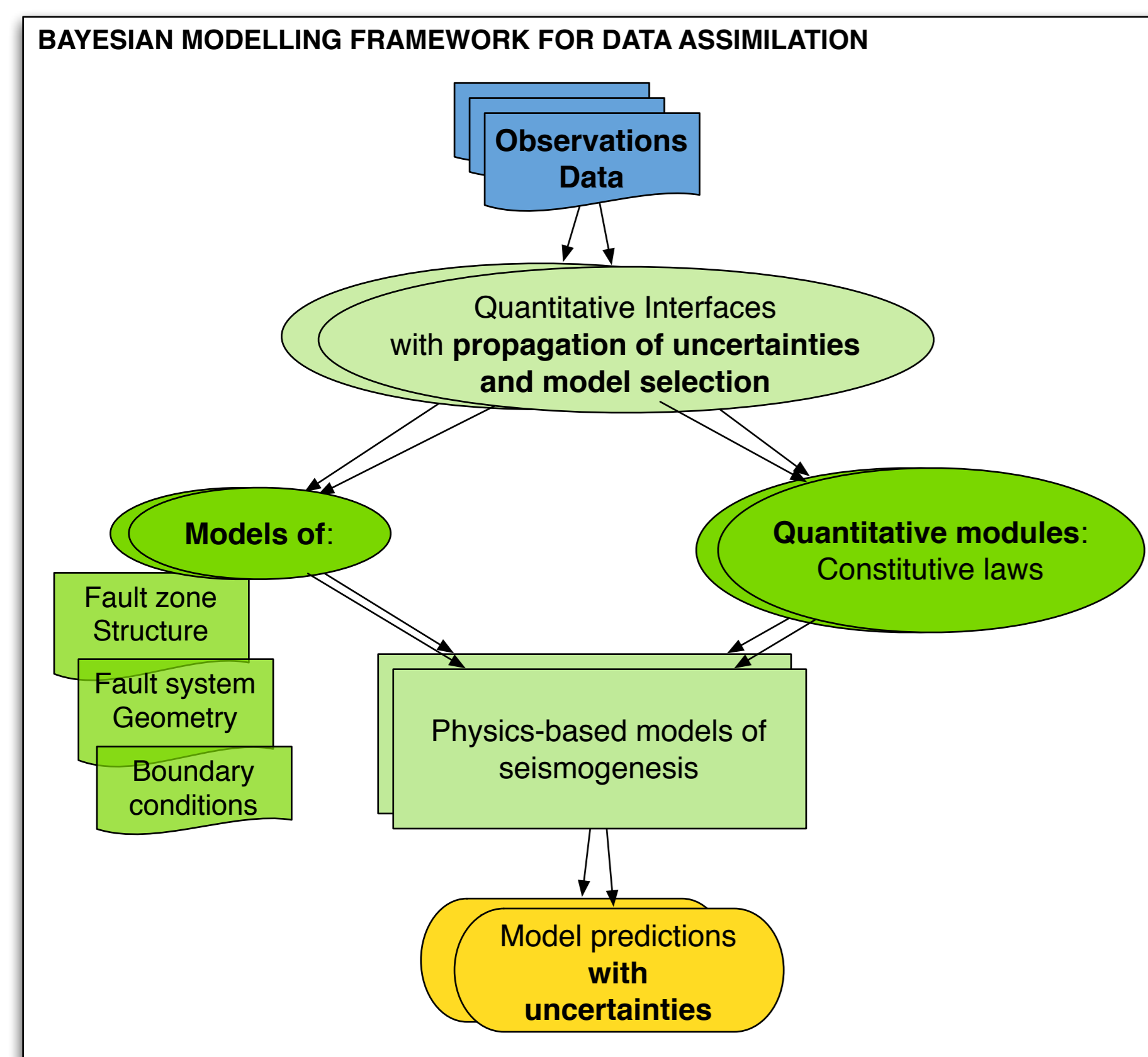


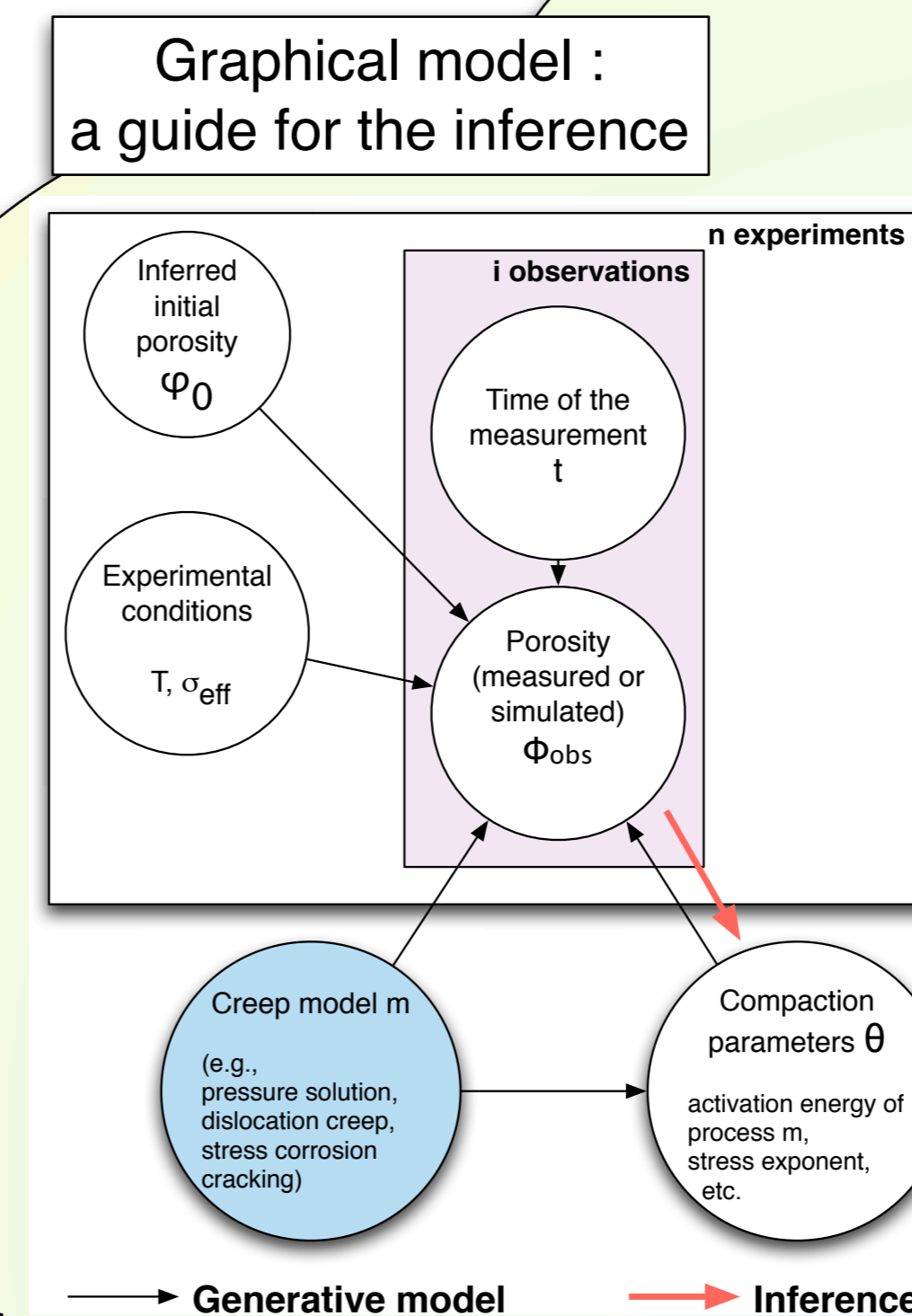
Probabilistic modeling of earthquake occurrence: first examples of data integration within a Bayesian framework.

Delphine D. Fitzenz (EOST Strasbourg, France), Steve H. Hickman (USGS Menlo Park, CA), Andre Jalobeanu (LSIIT Strasbourg, France), Chris Spiers (HPT Lab Utrecht University, The Netherlands)

Building upon our recent experience in the analysis and integration of isotropic creep experiments into numerical models of interseismic fault processes, we study creep under deviatoric stresses, and then perform time-forward simulations of interseismic fault behaviour. Given a shear loading rate and a rupture criterion, our model provides probability density functions for the time to failure and fault zone physical properties at the onset of failure. The first step in the forward modeling is the point-source model, in which we evaluate the robustness of the modeling results in response to uncertainties in the input parameters and alternative models for the creep law. Our modeling framework addresses two big issues in seismic hazard assessment: the evaluation of the aleatory uncertainties and the reduction of the epistemic uncertainties (via model selection). Current efforts also include extending the approach to study the relative influence of more complex simulations (with 2D to 3D faults), to provide a modular probabilistic "synthetic earthquake simulator". This will allow us to test the impact of different sources of heterogeneity in fault zone physical properties and loading conditions on the statistics of time to failure.



Data analysis tool



The five steps of the inference

- A:** Choose the parametrization making the problem as linear as possible
- B:** When a factorization appears in the new Bayesian network, adopt a hierarchical inference scheme
- C:** Calculate the joint probability density function according to the graph structure (the joint pdf is proportional to the posterior pdf).
- D:** Eliminate the nuisance parameters (marginalization step)
- E:** Revert from the new parametrization to the original parameters θ_i *

Prerequisite

- Integration of the creep law: we want to use the data, not a subproduct of it
- Reparametrization such that the problem is as linear as possible (work with Gaussians)

$$f(\varphi_0, t) = -\mu \log(t e^{-\lambda'/\mu} + e^{-\varphi_0/\mu})$$

$$\Theta = F(\theta) = \{\log(-\theta_0)/\theta_3, \theta_1/\theta_3, \theta_2/\theta_3, 1/\theta_3\}$$

$$\lambda' = h(\theta) = -\theta_0 - k_1 \theta_1 + k_2 \theta_2 + \theta_3 \log \theta_3$$

$$\mu \equiv \theta_3$$

Advantages of integrating

- no need to define explicitly the notion of state
- no need to compute strain or compaction rates
- we can **analyse all data simultaneously**

Observations
Experimental or simulated porosity time series

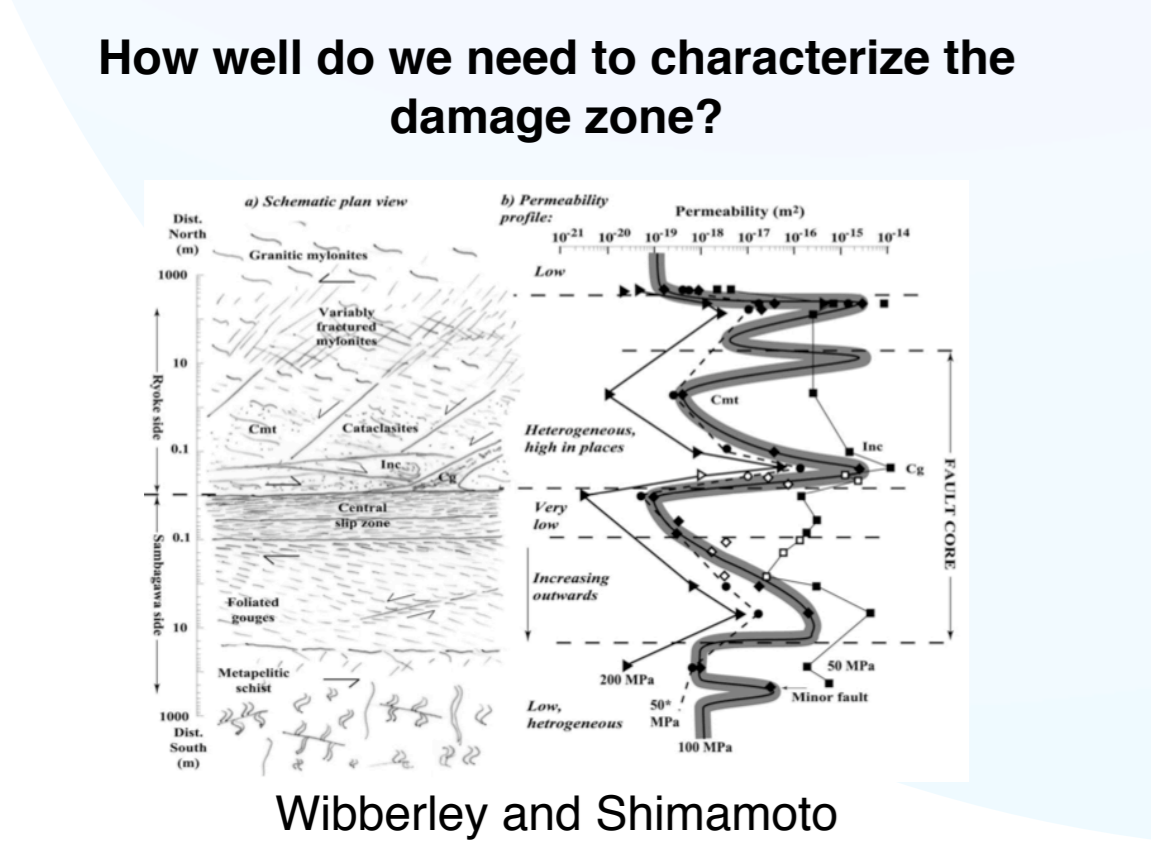
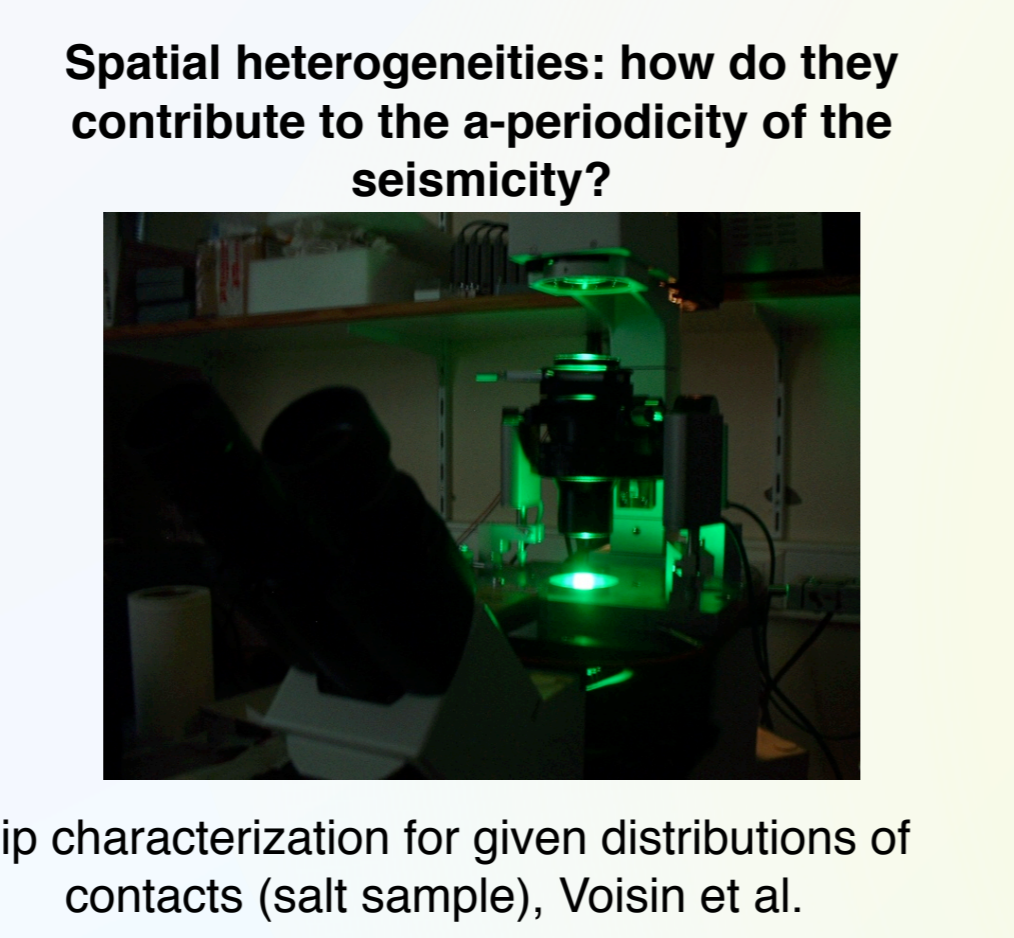
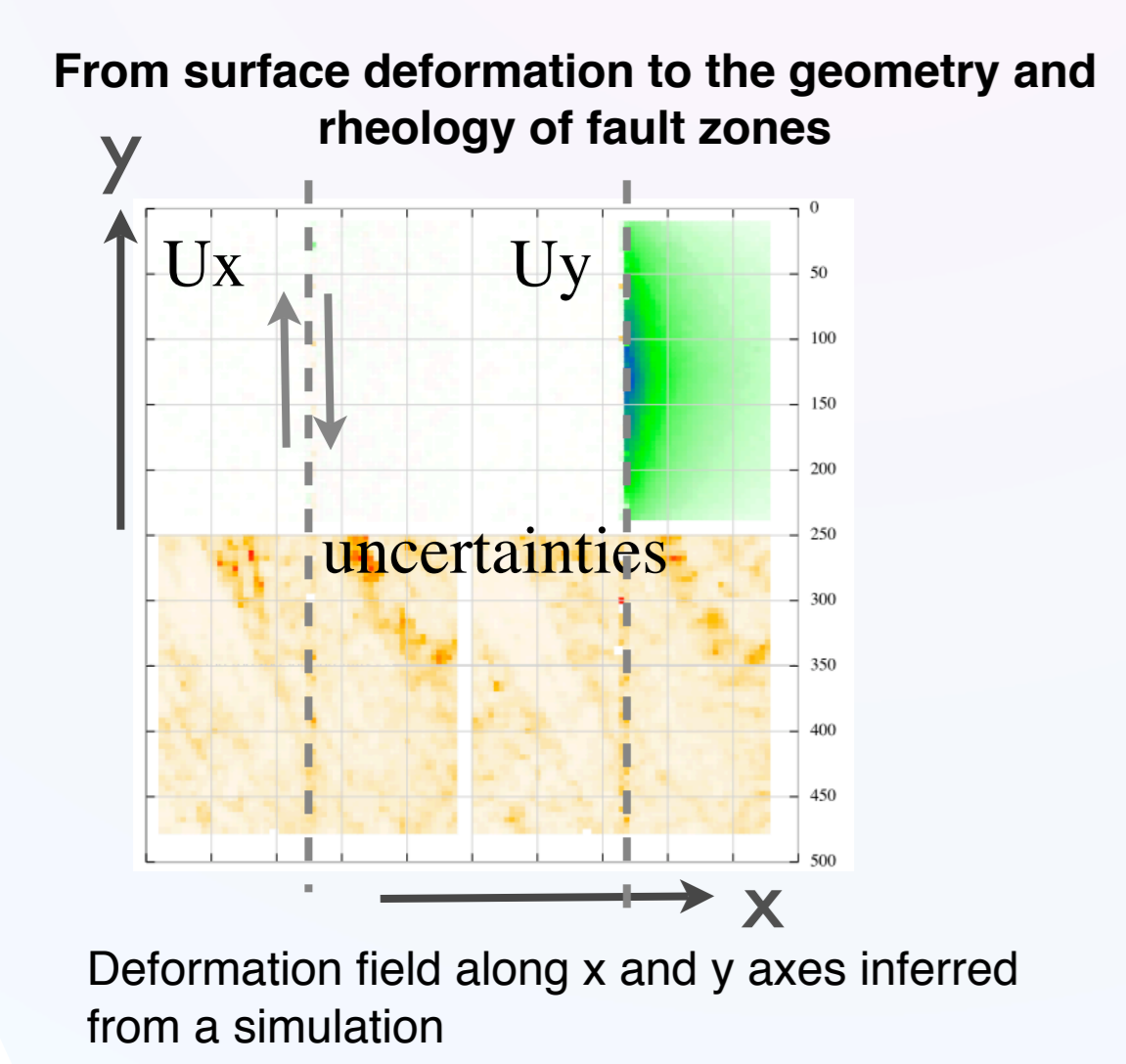
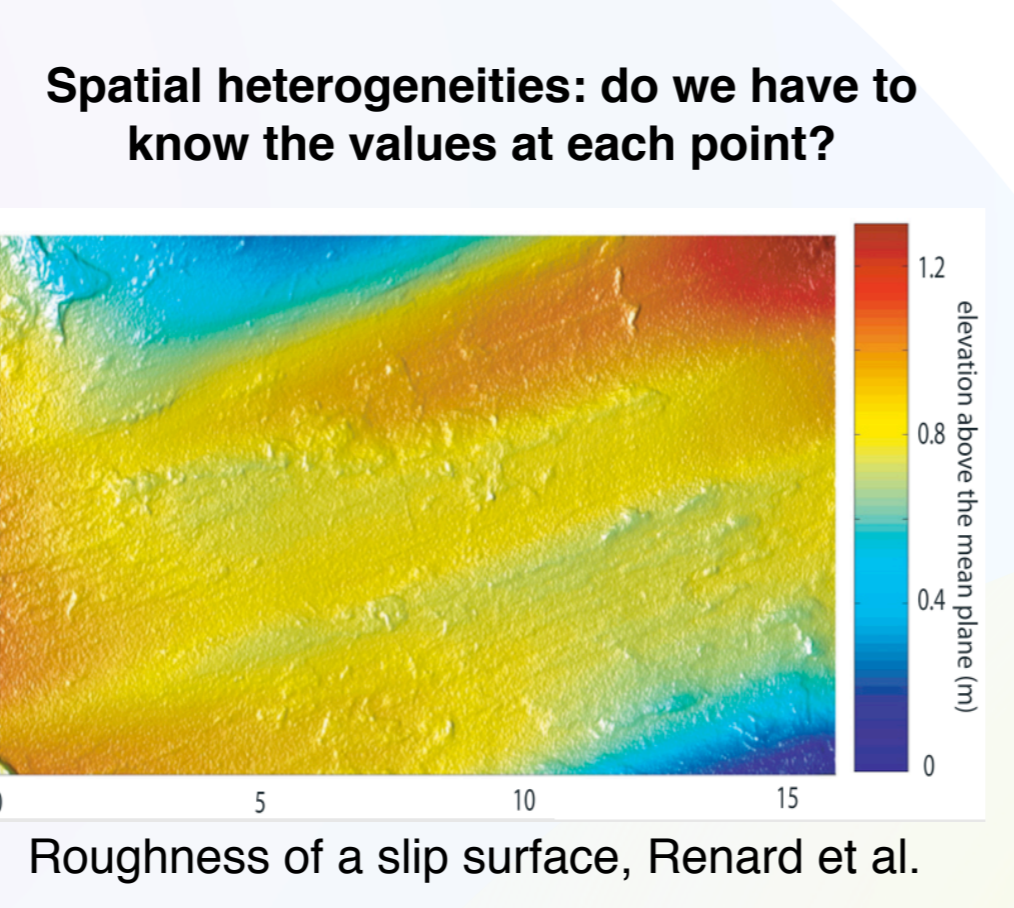
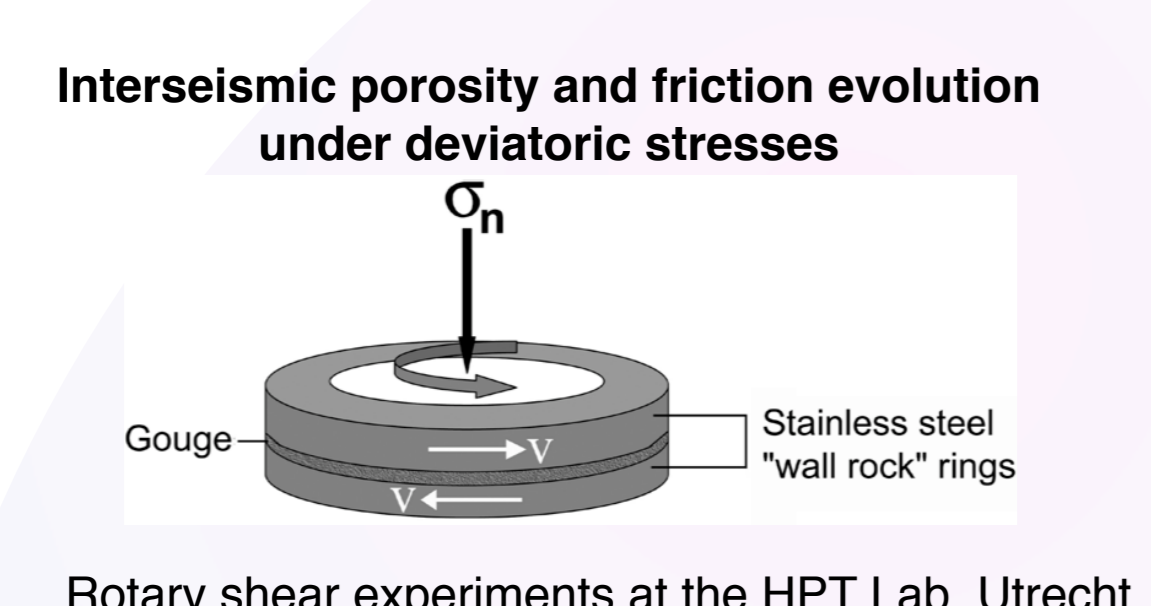
Choice of the model m *

When the grain-size effects can be neglected, and we make no assumption on the mechanisms, we can choose a creep law of the type:

$$\frac{\partial \varphi}{\partial t} = \theta_0 \times \sigma_{\text{eff}}^{\theta_1} \times \exp(-\theta_2/(RT)) \times \exp(\theta_3 \varphi)$$

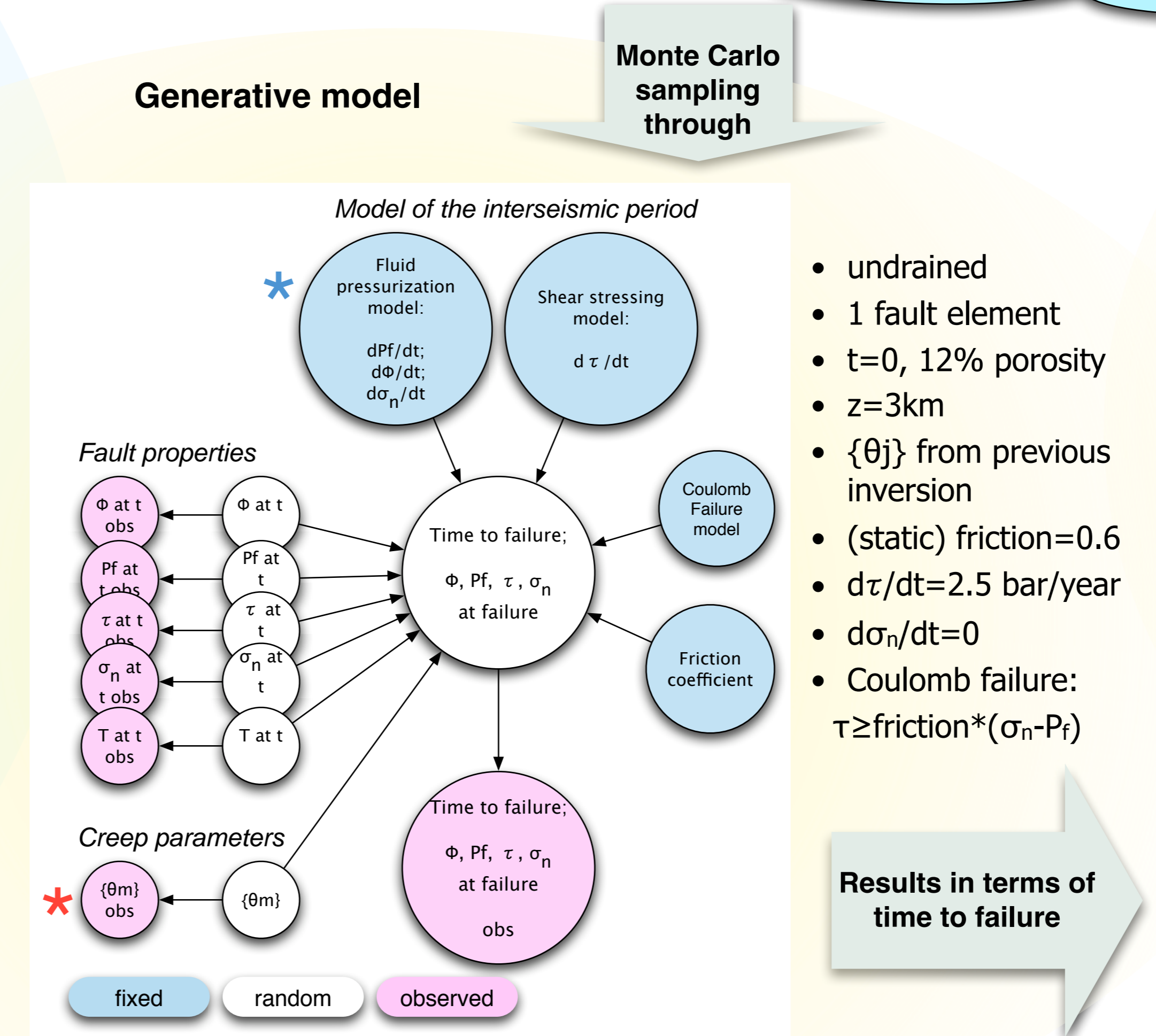
Stress exponent Apparent activation energy

Future directions



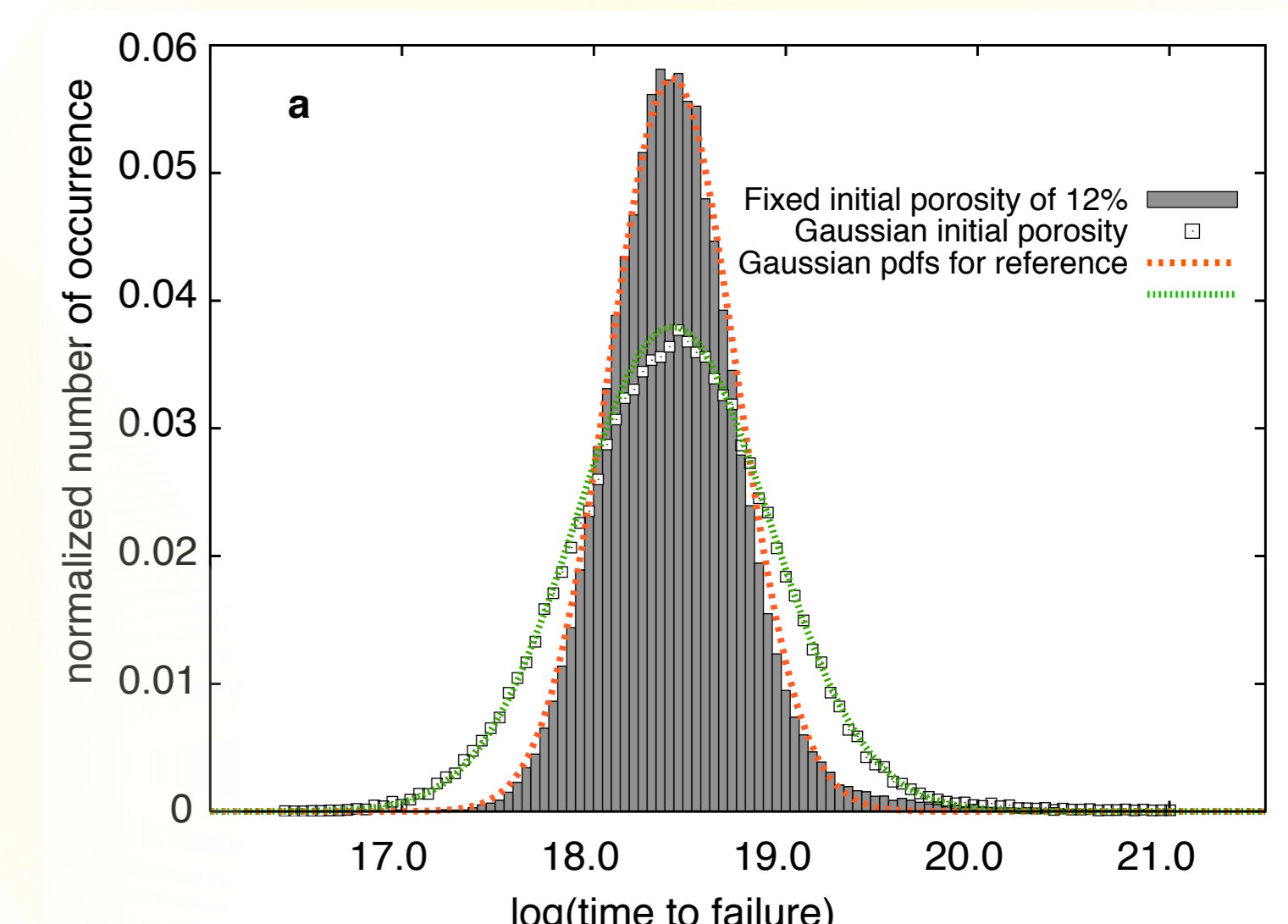
Suggestions are welcome!!

Data integration tool



- undrained
- 1 fault element
- t=0, 12% porosity
- z=3km
- {theta_j} from previous inversion
- (static) friction=0.6
- dtau/dt=2.5 bar/year
- dsigma_n/dt=0
- Coulomb failure: tau >= friction*(sigma_n - Pf)

Results in terms of time to failure



- Uncorrelated stress exponent and apparent activation energy lead to lognormal time to failure close to the optimum value.
- Random variations in post EQ porosity leads to increased variability in time to failure, and stable mean.