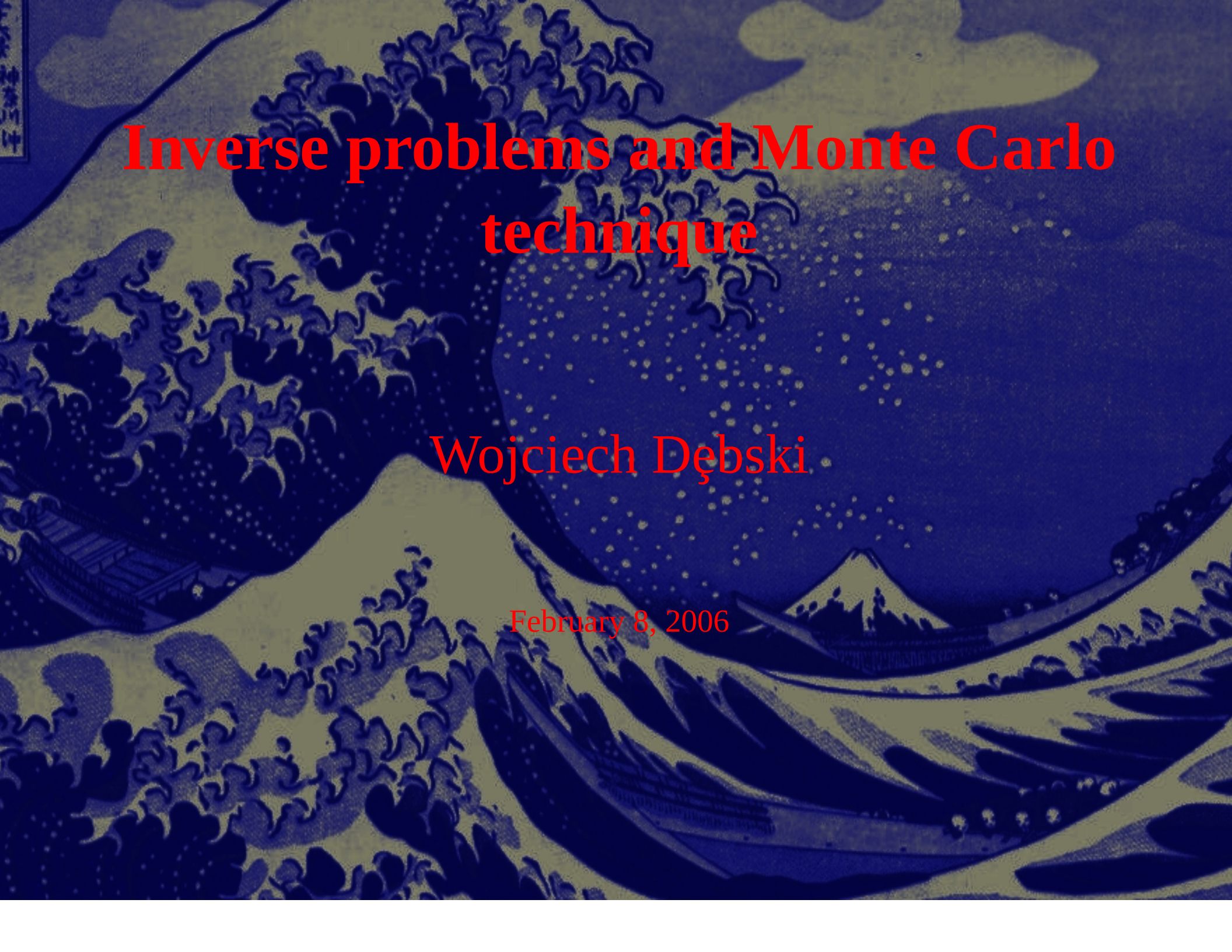




Inverse problems and Monte Carlo technique

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February 8, 2006

Plan of the talk

- ❖ Solutions to inverse problems
- ❖ Bayesian techniques - error estimation
- ❖ Examples
 - ★ Location of seismic events
 - ★ Imaging rupture process
 - ★ seismic tomography

Inverse problems

Inverse problems appears on the scen when:

1. we want to know parameters that cannot be measured directly (indirect measurements)
2. an inference from a given data (information) set is performed (non-parametric inversion)
3. given theory must be verified (!!!)

Inverse problem - Indirect Measurements

$$\mathbf{d}^{obs} \implies \mathbf{m}$$

Solution

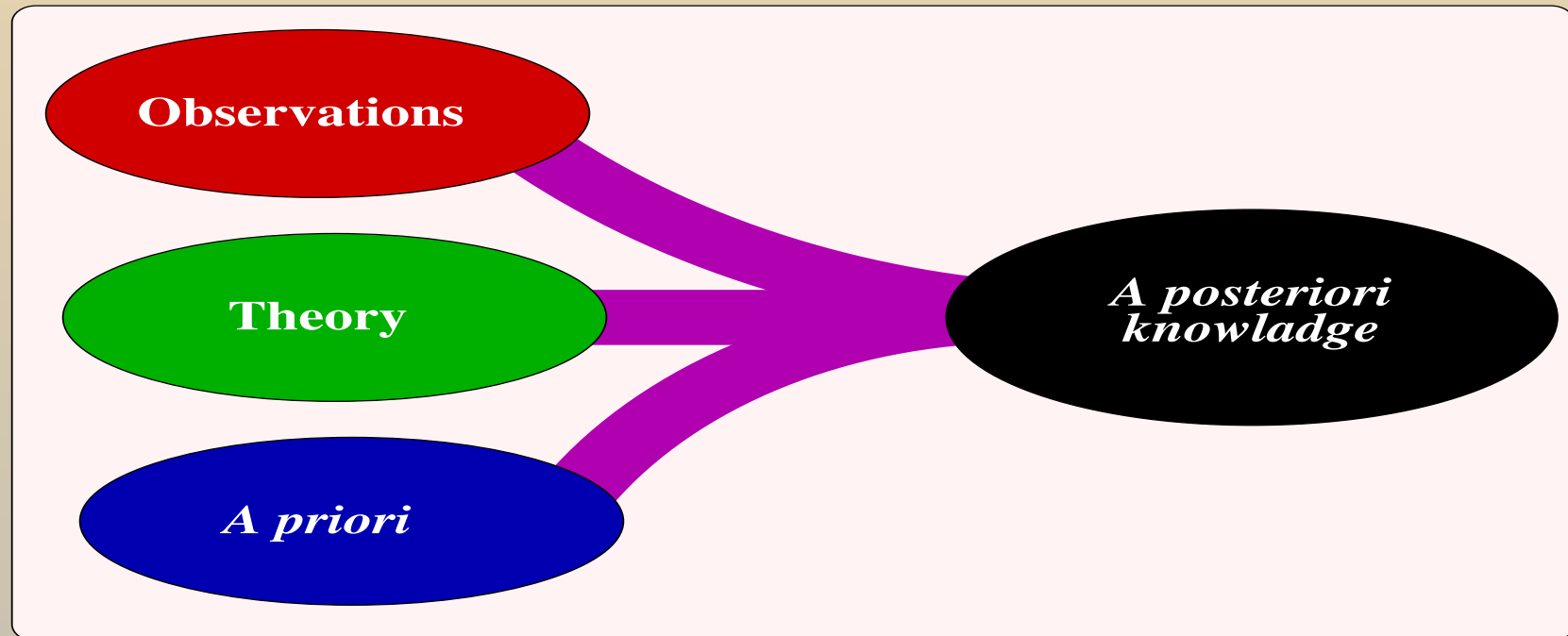
$$||\mathbf{d}^{obs} - \mathbf{d}^{th}(\mathbf{m})|| + \lambda ||\mathbf{m} - \mathbf{m}^{apr}|| = \min$$

Errors

$$\mathbf{m}^{true} = \mathbf{m}^{ml} + \epsilon_{\mathbf{m}}$$

$$\epsilon_{\mathbf{m}} = ???$$

Bayesian Inversion - Basic Ideas



$$\sigma_{post}(\mathbf{m}|\mathbf{d}^{obs}) \sim \sigma_{apr}(\mathbf{m}) \int_{\mathbf{D}} \sigma_{th}(\mathbf{d}|\mathbf{m}) \star \sigma_{obs}(\mathbf{d}|\mathbf{d}^{obs}|)$$

Bayesian Inversion

(parameter estimation)

A posteriori pdf

$$\sigma(\mathbf{m}) = f(\mathbf{m})L(\mathbf{m}, \mathbf{d})$$

- ♦ f - A priori pdf (prob. dens. funct.)
- ♦ L - Likelihood function

Errors independent of $\mathbf{m}$ and $\mathbf{d}$

$$\mathbf{d}^{th} = \mathbf{G}(\mathbf{m})$$

$$\sigma(\mathbf{m}) = f(\mathbf{m}) \exp \left(-||\mathbf{d}^{obs} - \mathbf{G}(\mathbf{m})|| \right)$$

Bayesian approach

A posteriori pdf $\sigma(m, d)$:

- ❖ always exists
- ❖ is unique
- ❖ describes all information
- ❖ is the **solution** of an inverse problem

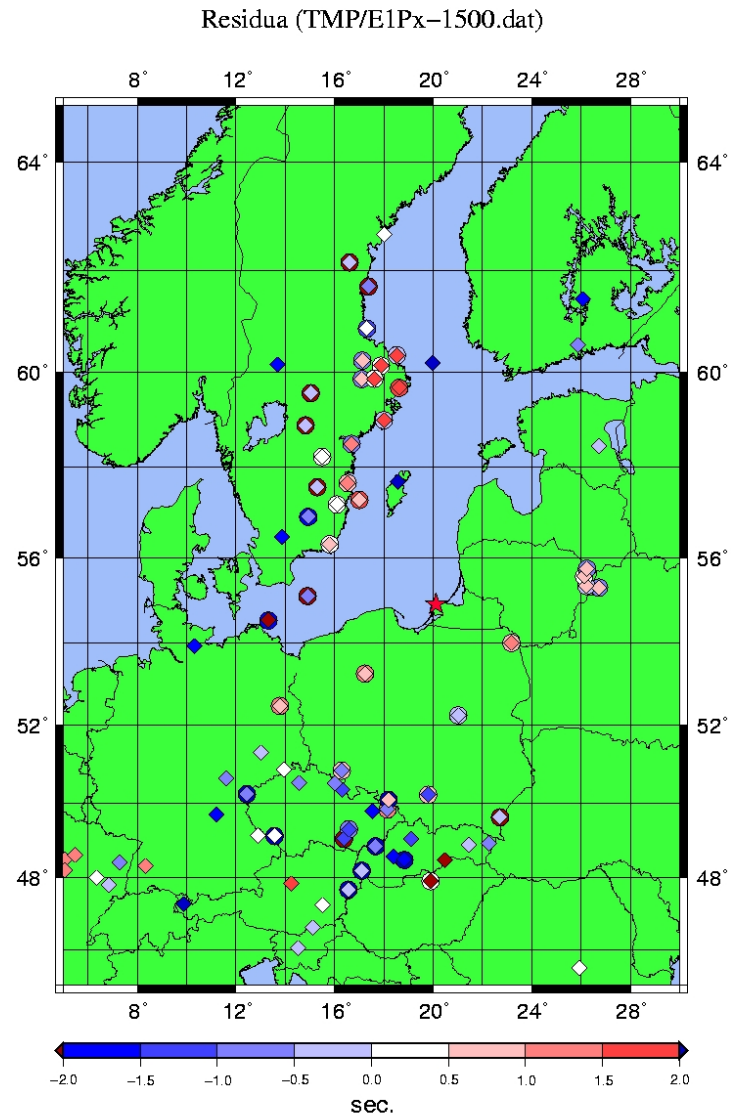
When and why we need to use this approach ???

ERROR ANALYSIS !!!

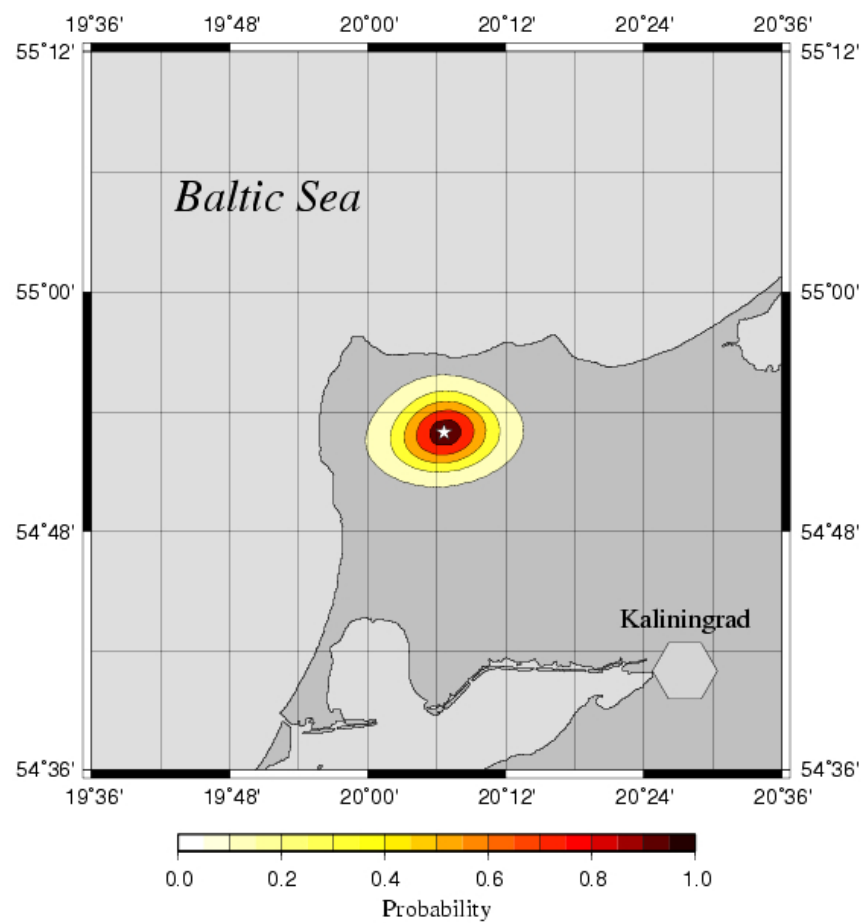
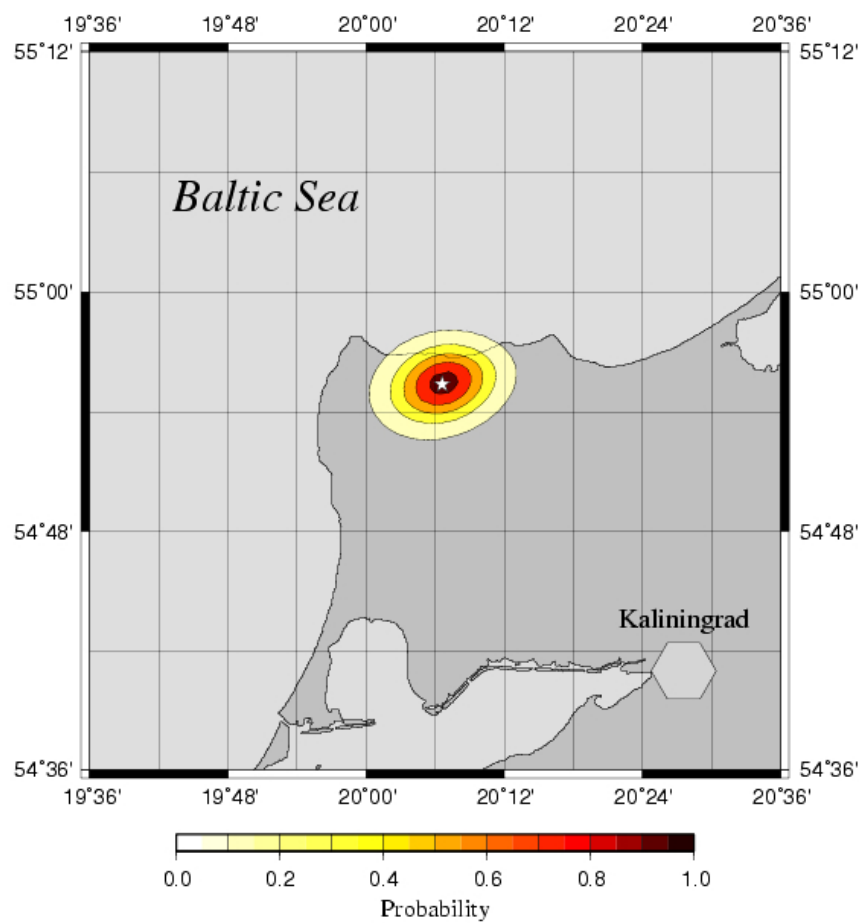
Examination of $\sigma(\mathbf{m})$

- ❖ Searching *Maximum Likelihood* solution
 - ★ Gradient methods
 - ★ Deterministic methods (e.g. simplex)
 - ★ Monte Carlo - Global Optimization
 - ➡ Simulated Annealing
 - ➡ Genetic Algorithm
 - ❖ Sampling over regular grid
 - ❖ Random (Monte Carlo) Sampling
-

2004 earthquakes in Kalinigrad area ($M_L \approx 5/5.2$)

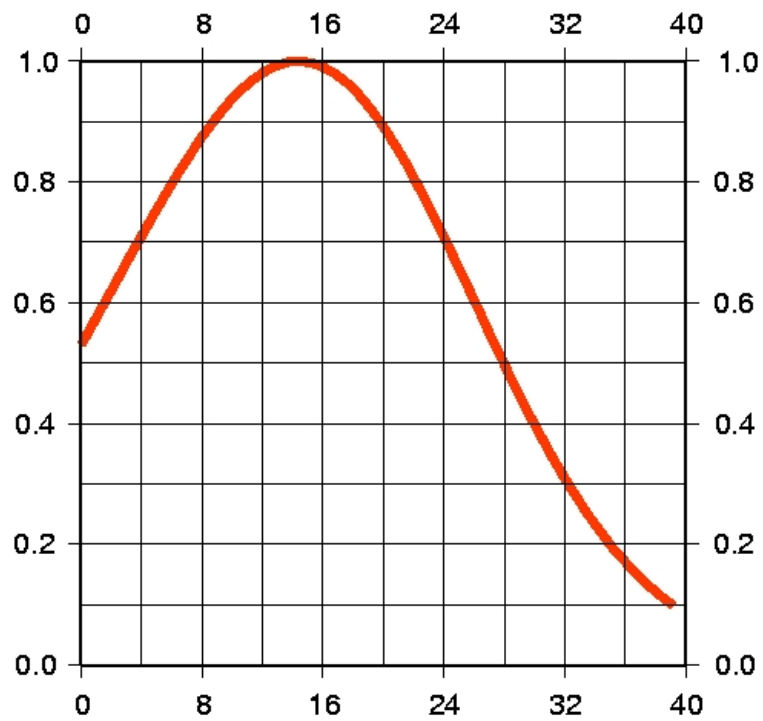


Epicenter location

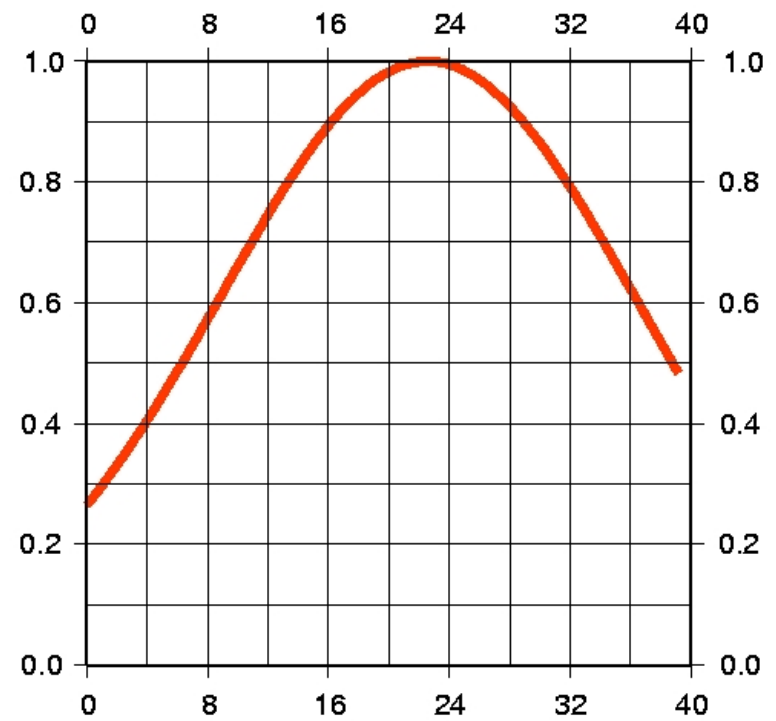


Depth

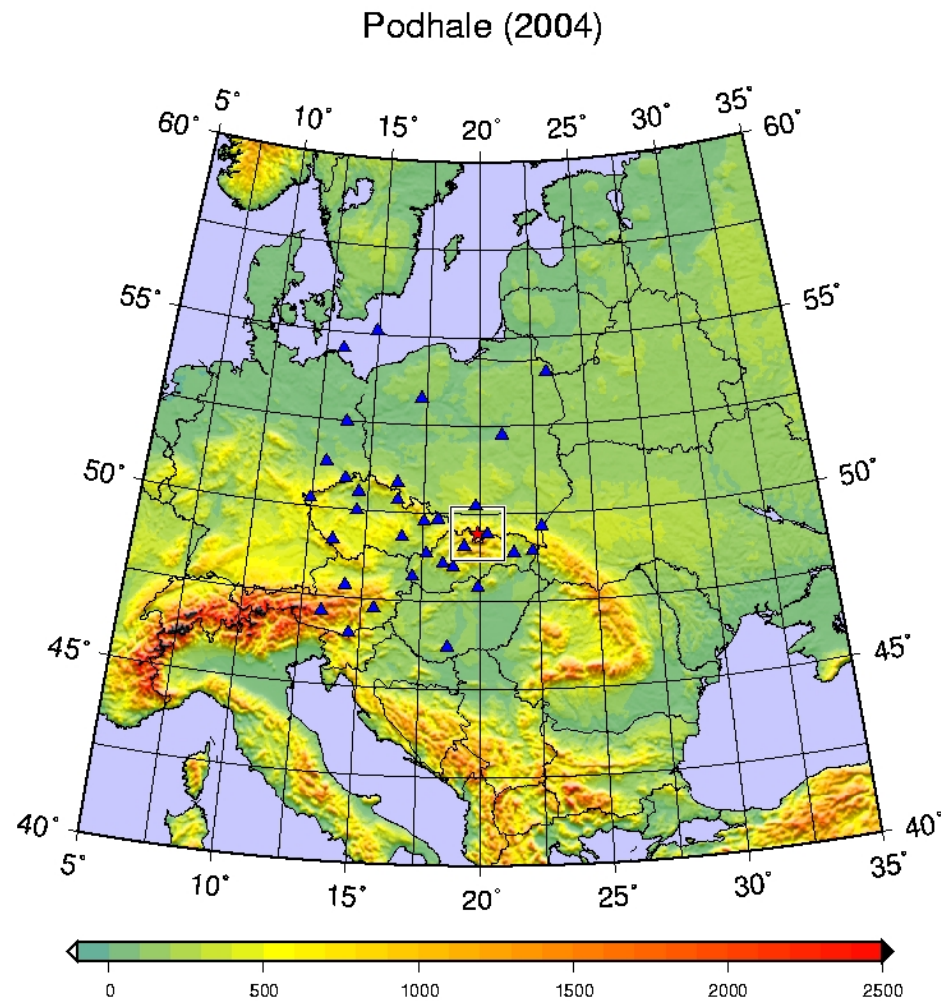
DEP



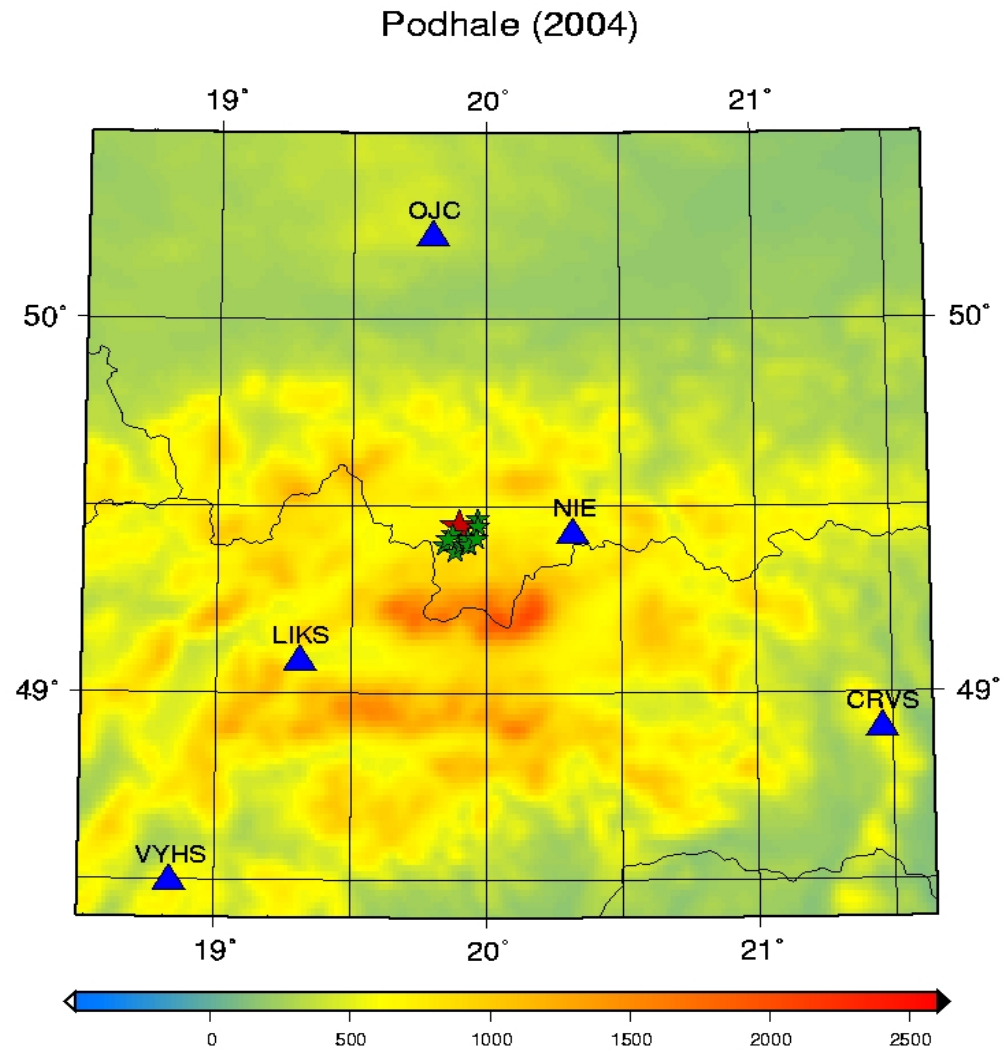
DEP



Natural seismicity in Poland (A)

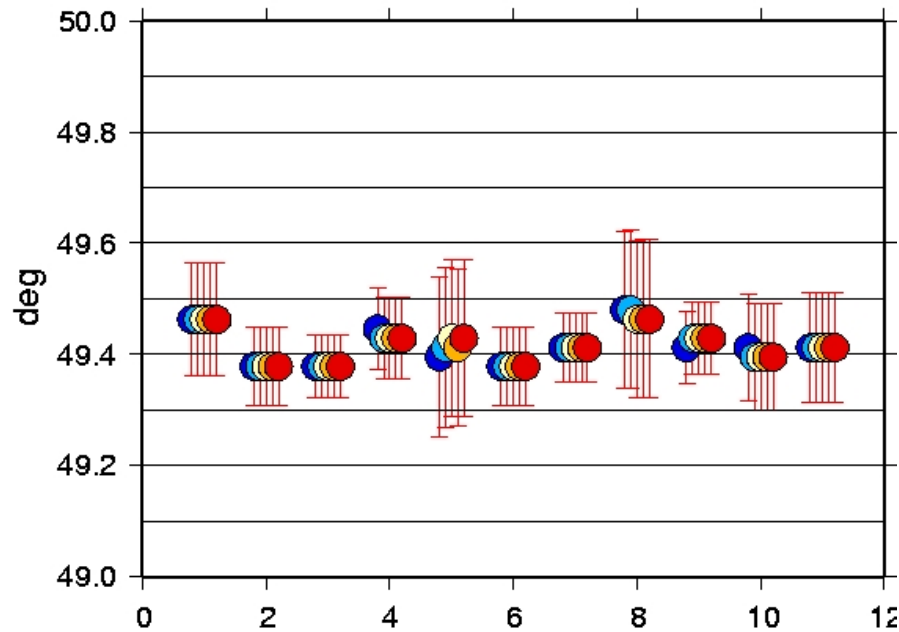


Natural seismicity in Poland (B)

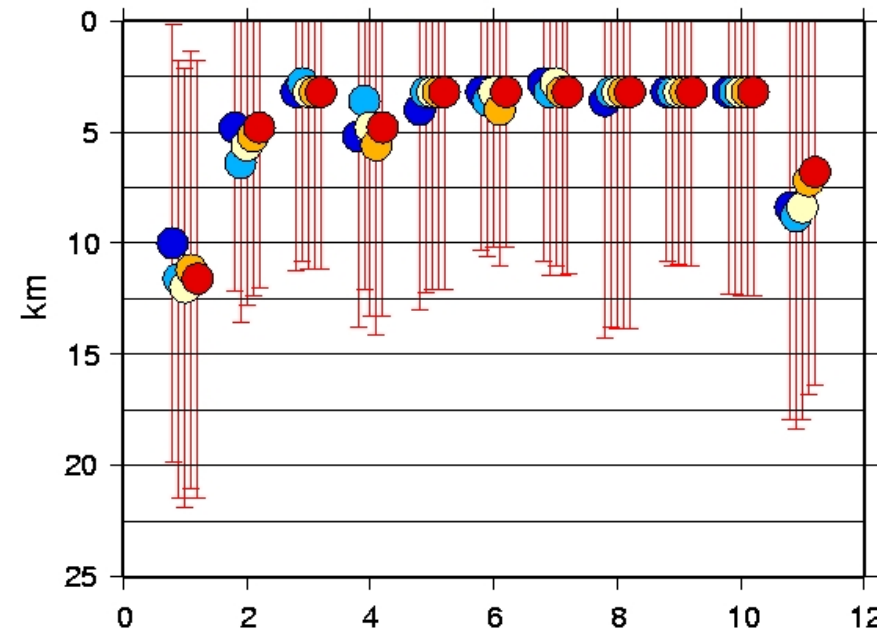


2004 Karpathy earthquakes (MC errors)

Latitude



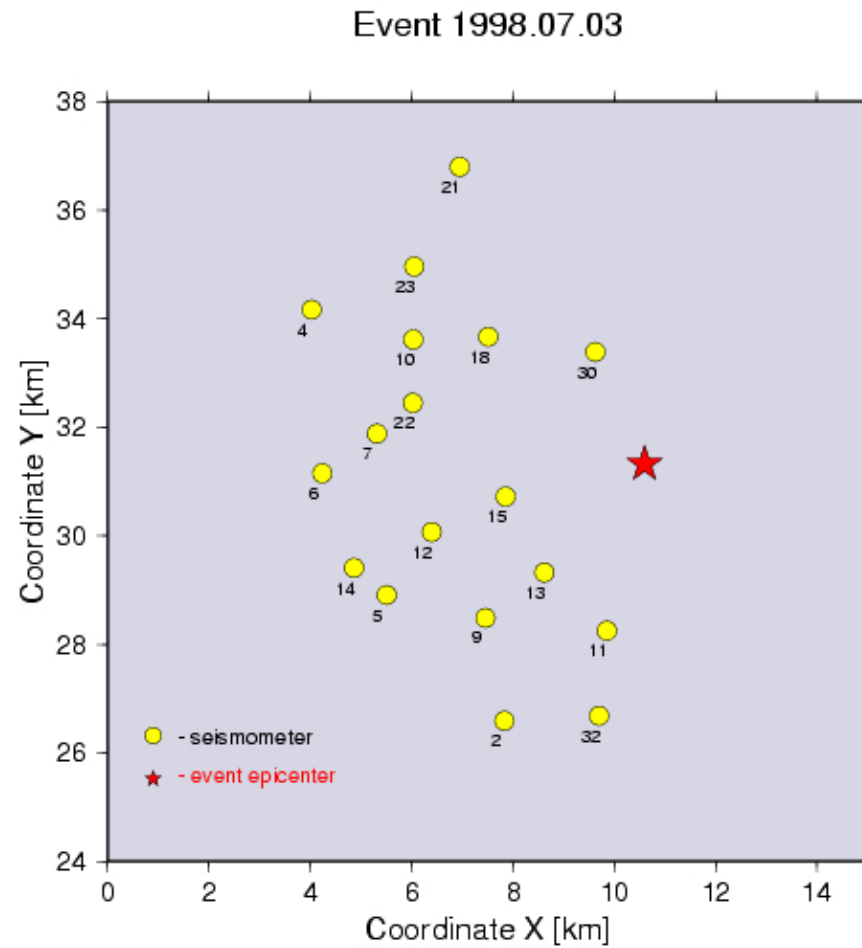
Depth [km]



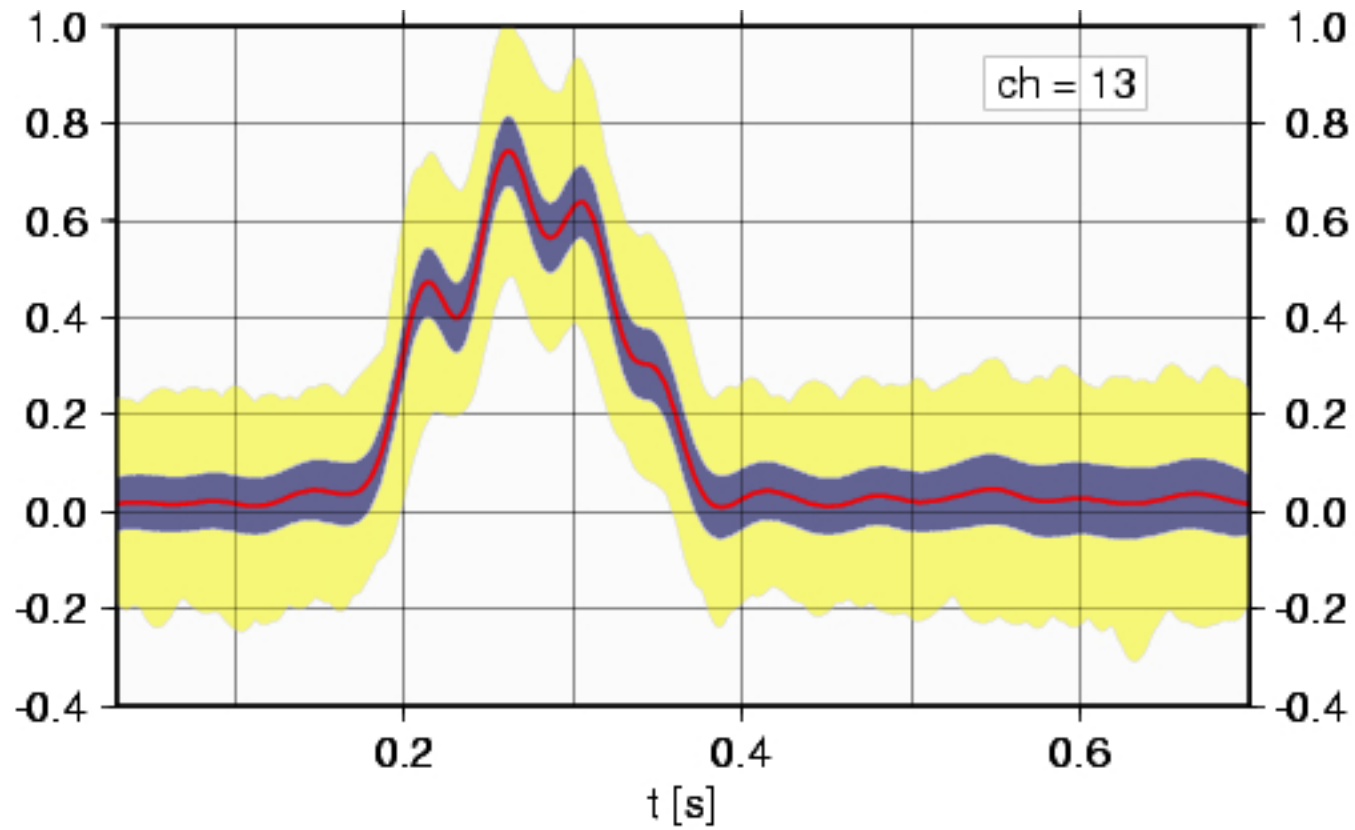
Imaging rupture process

$$u(x, t) = \int_{t, x'} \dot{M}(t) G(t - t', x - x')$$

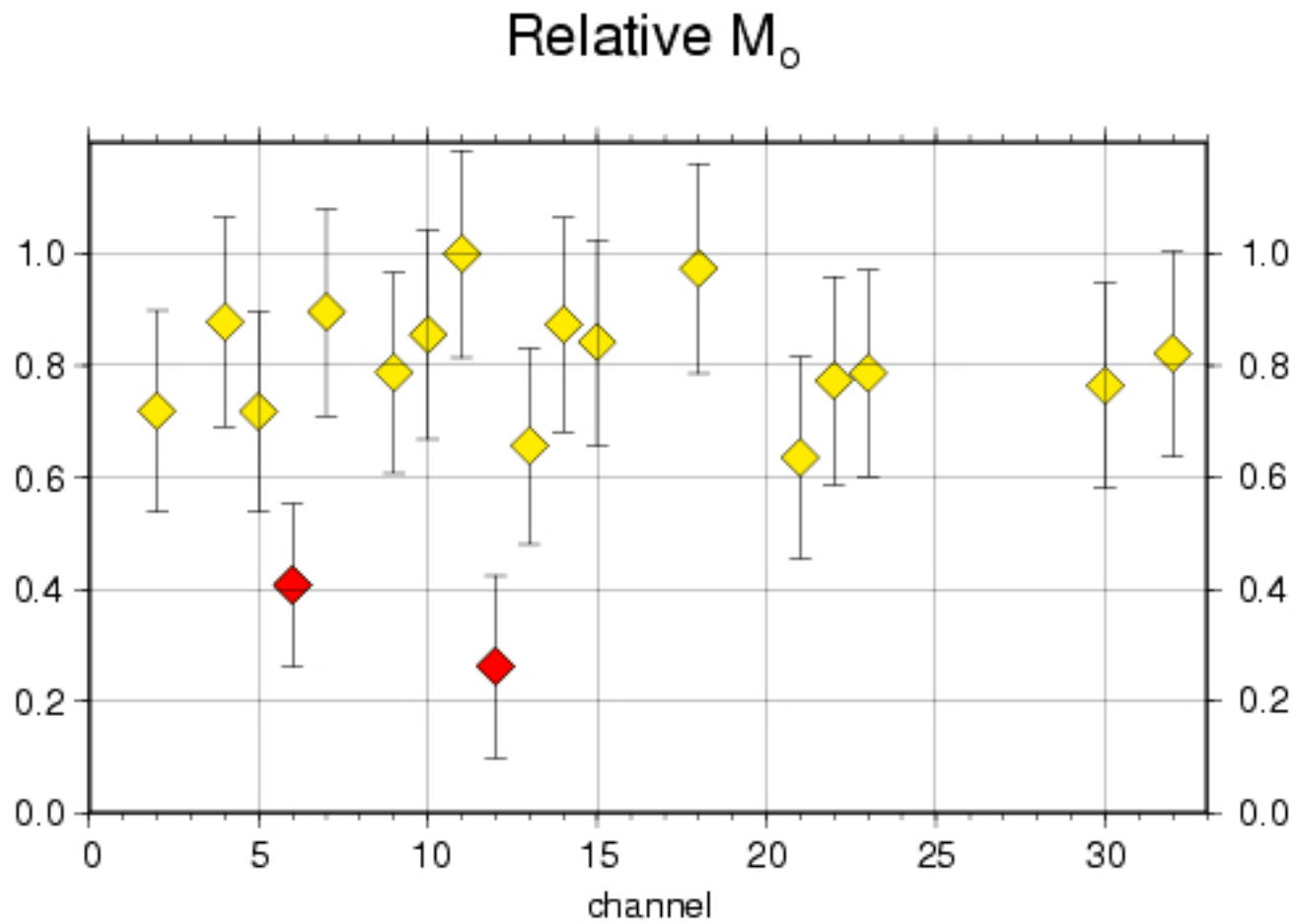
Copper mines seismic network



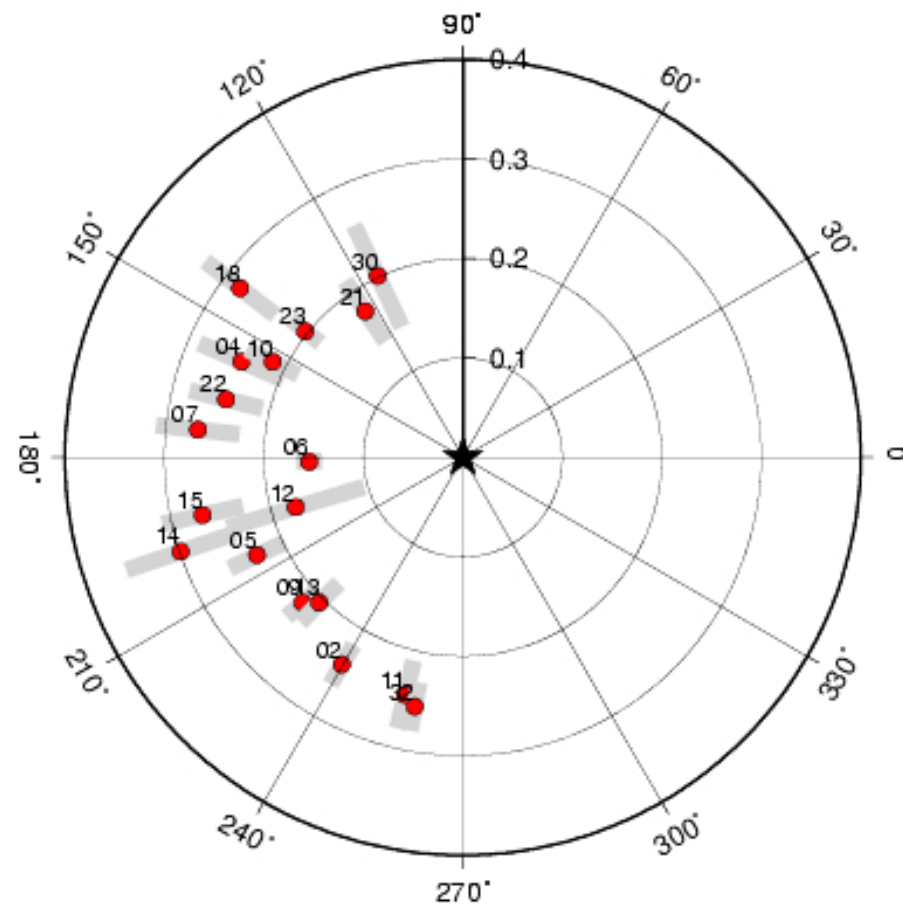
STF - channel 13



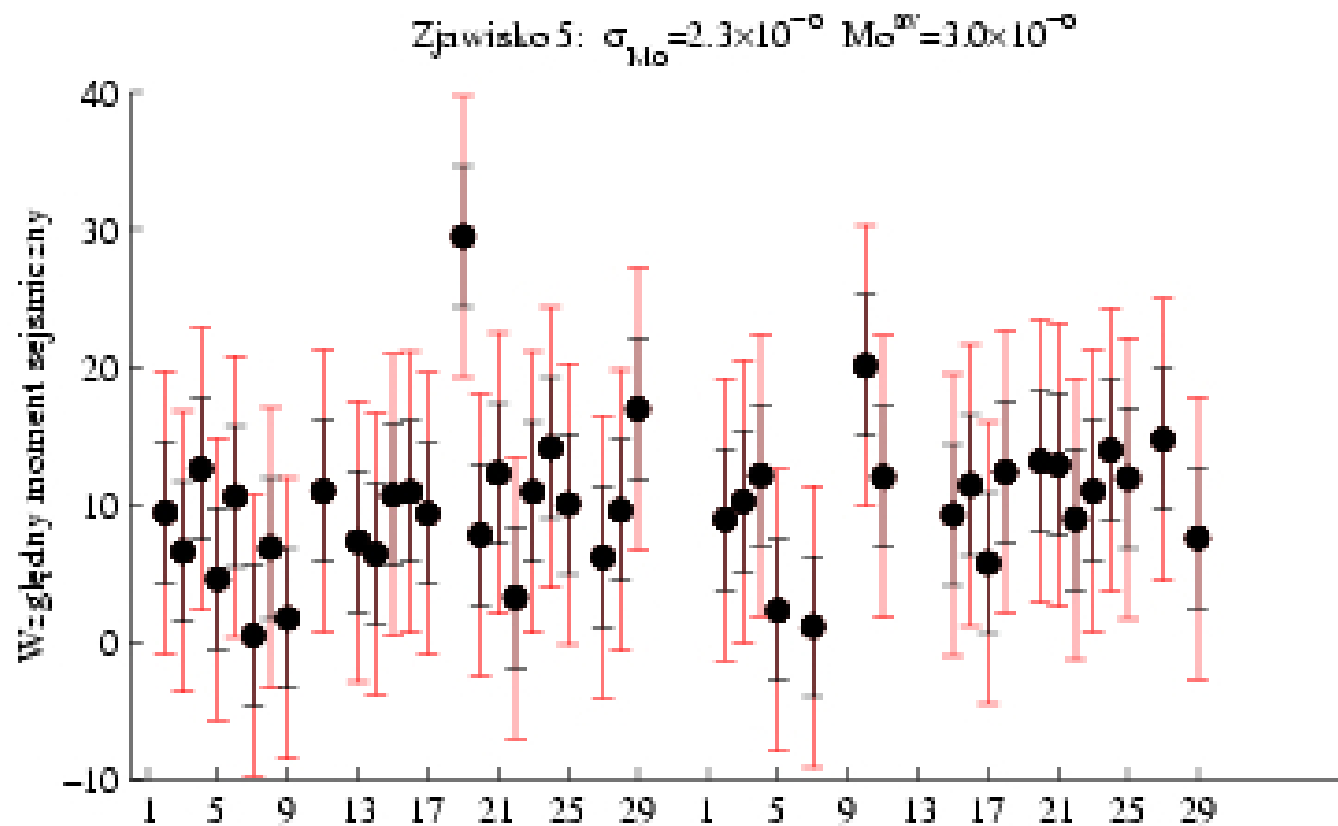
Seismic moment M_o



Radiation pattern



Another example



Tomography and expectations?

Tomography should provides:

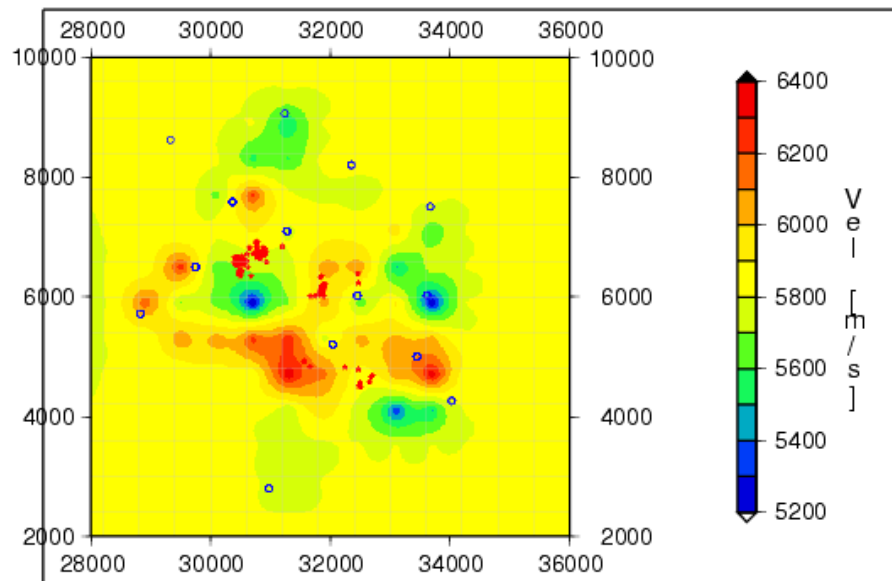
1. spatial heterogeneity distribution
2. quantify “strenght” of heterogeneities

PROBLEM:

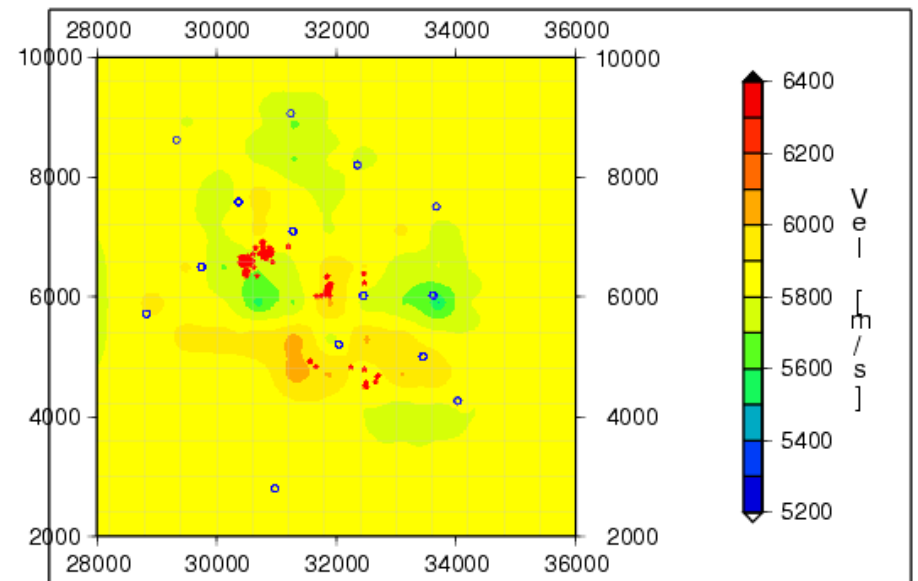
$$||\mathbf{d}^{obs} - \mathbf{d}^{th}(\mathbf{m})|| + \lambda ||\mathbf{m} - \mathbf{m}^{apr}|| = \min$$

Heterogeneities

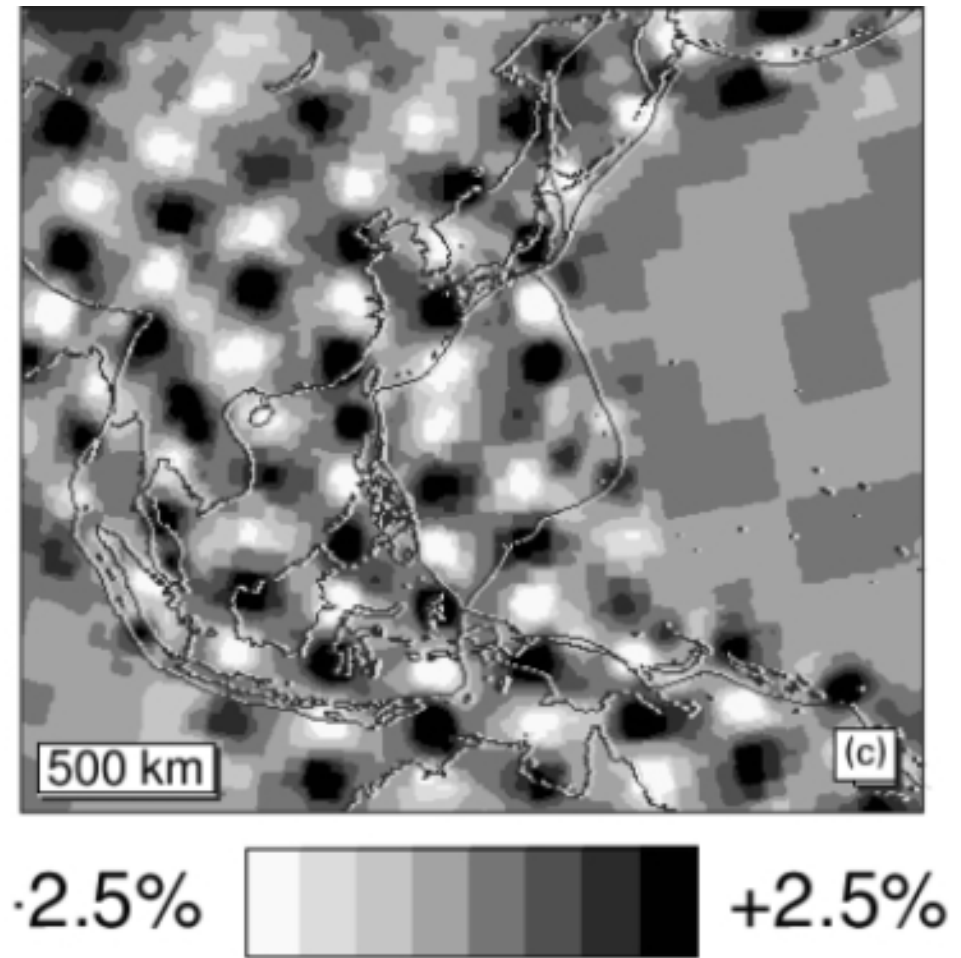
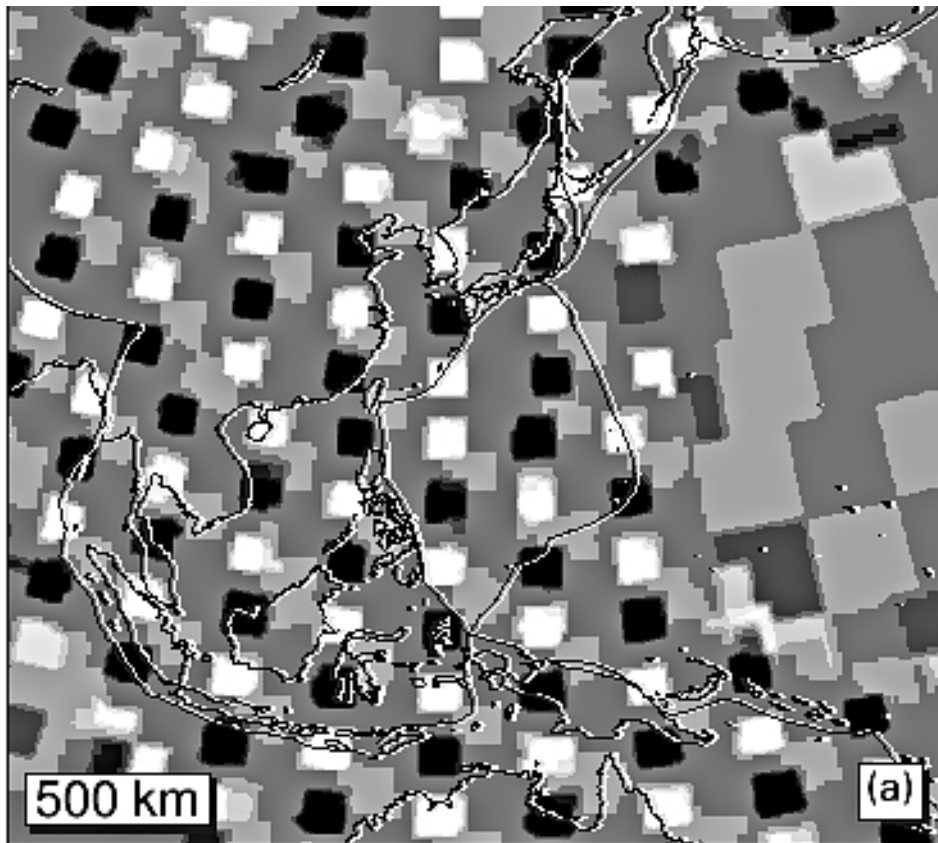
AVR-Cd=0.00001-Cm=200-Id=I-Im=I



AVR-Cd=0.005-Cm=200-Id=I-Im=I

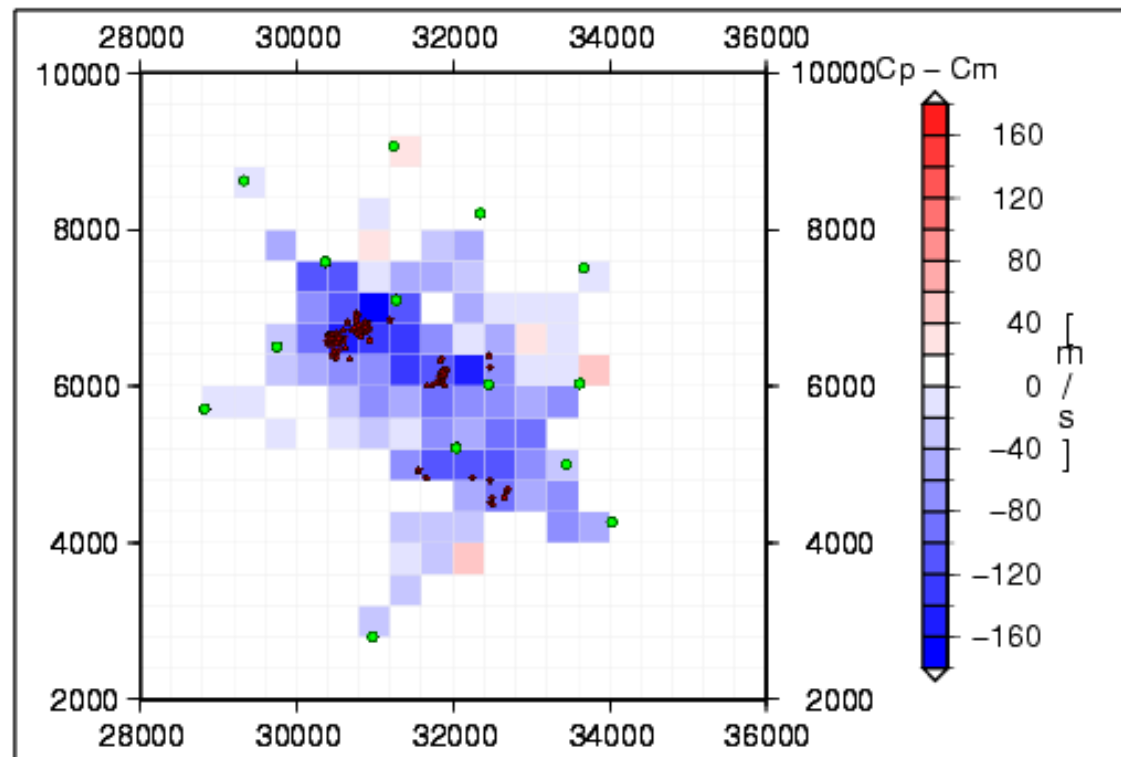


Classical approach



Example of an error estimation

$$\text{ERR}-C_d=0.002-C_m=200-l_d=l-l_m=l$$



Sampling to tomography solution

SEE MOVIE