

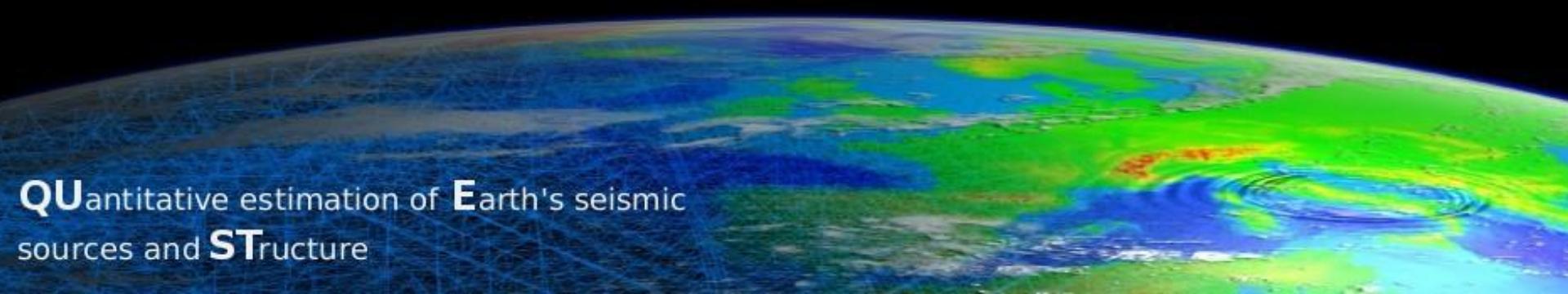
Resolution analysis in full waveform inversion

by

Andreas Fichtner & Jeannot Trampert



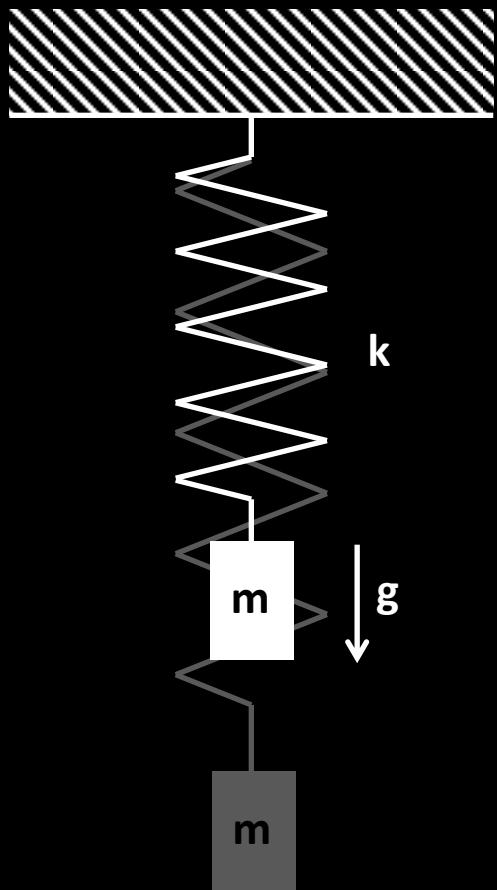
Universiteit Utrecht
Department of Earth Sciences



QUantitative estimation of **E**arth's seismic
sources and **ST**ructure



Welcome to the beginner's practicals !

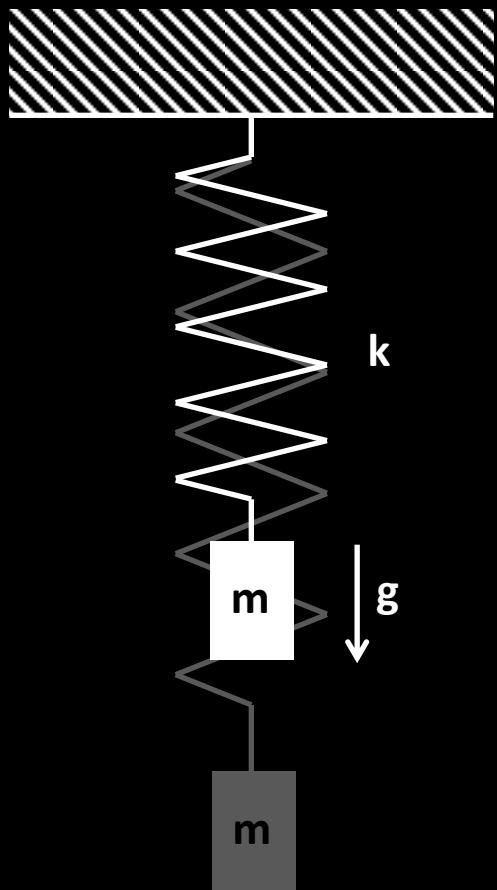


Initial-value problem

$$mg = kx + m\ddot{x}, \quad x(0) = \dot{x}(0) = 0$$

Oscillatory solution

$$x(t) = \frac{1}{2}x_{\max} [1 - \cos(\omega t)], \quad k = \omega^2 m$$



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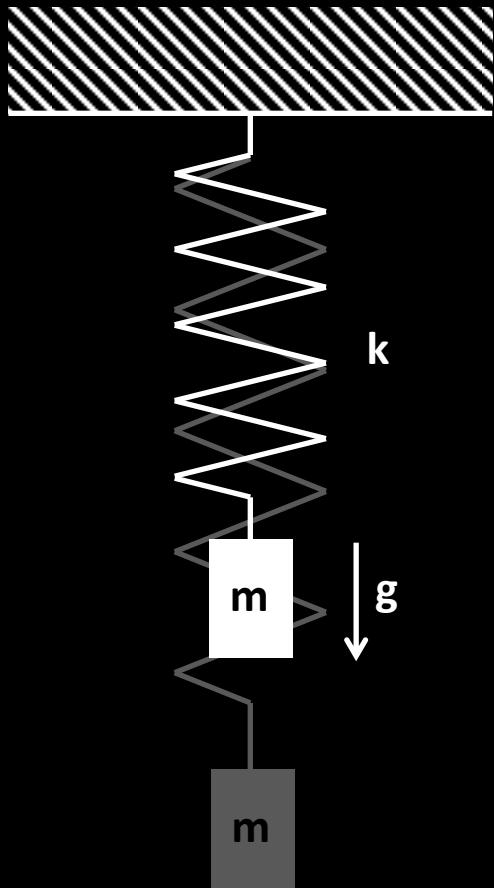
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Simplistic error propagation

$$\Delta k = 2\omega m \Delta \omega + \omega^2 \Delta m$$



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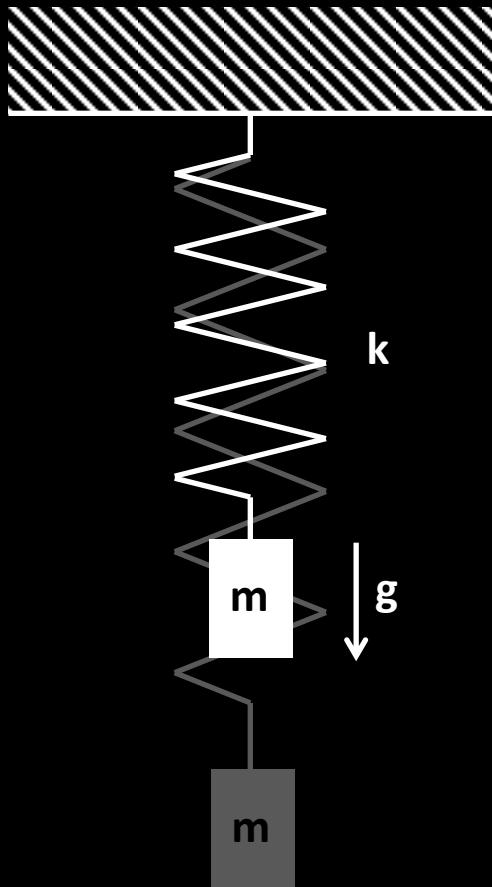
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- Extent to which a quantity can be constrained by observations.
→ Relevance of results, experimental design.



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Classical linear inverse theory

- Backus & Gilbert, 1968. **The resolving power of gross Earth data.** Geophys. J.
- Kennett & Nolet, 1978. **Resolution analysis for discrete systems.** Geophys. J.
- Deal & Nolet, 1996. **Nullspace shuttles.** Geophys. J. Int.
- Nolet, Montelli & Virieux, 1999. **Explicit approximate expressions for the resolution and a posteriori covariance of massive tomographic systems.** Geophys. J. Int.
- Boschi, 2003. **Measures of resolution in global body wave tomography.** Geophys. Res. Lett.

Non-linear inverse problems

- Duane & Kennedy, 1987. **Hybrid Monte Carlo.** Phys. Lett. B
- Sambridge, 1999. **Geophysical inversion with a neighborhood algorithm.** Geophys. J. Int.
- Devillee, Curtis & Roy-Chowdhury, 1999. **An efficient, probabilistic neural network approach to solving inverse problems.** J. Geophys. Res.
- Sambridge & Mosegaard, 2002. **Monte Carlo methods in geophysical inverse problems.** Rev. Geophys.
- Tarantola, 2005. **Inverse problem theory and methods for model parameter estimation.**

„classical“ tomography:

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Nature of synthetic inversions

- Very easy to perform
- Optimistic (show what you can recover, hide what you cannot see)

Complexity of alternatives

- Require large efforts
- Difficult to interpret (posterior covariance, probabilities)

Cultural

- Lack of (self-) criticism
- Seductive power of colourful pictures

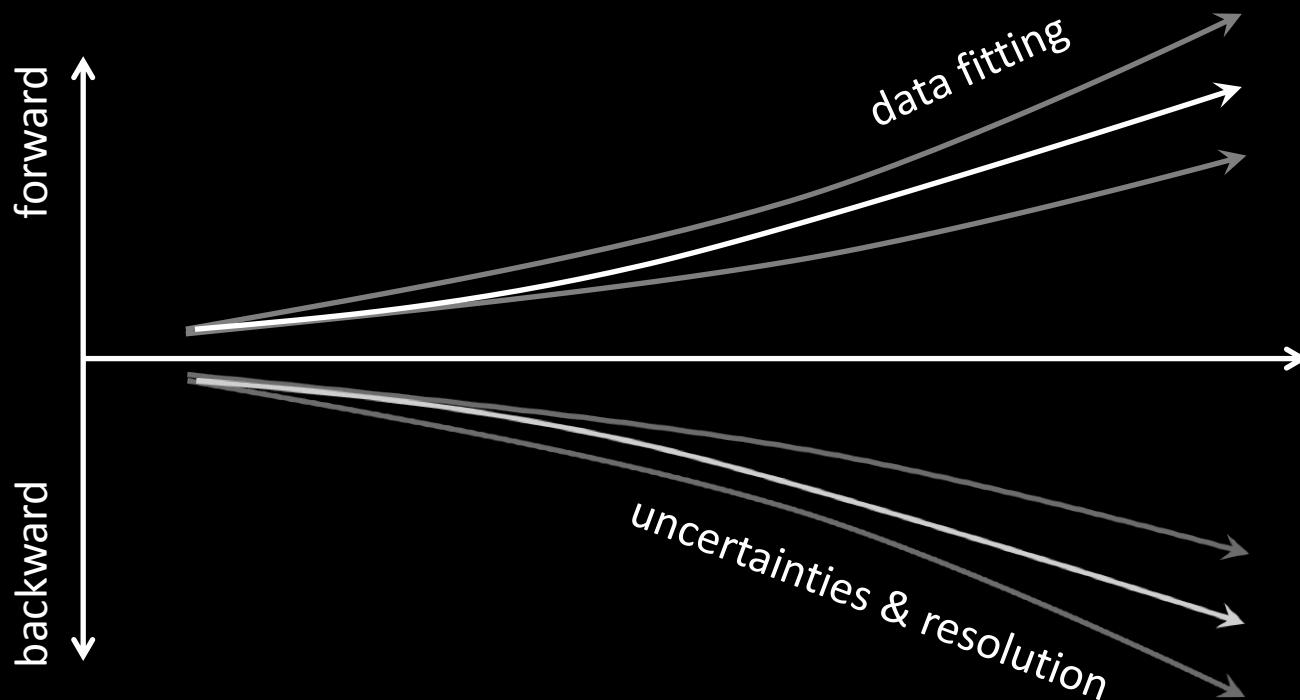
„classical“ tomography:

- tendency to ignore these developments
- restriction to synthetic inversions (chequer boards, hardly informative and misleading)

full waveform inversion (numerical modelling + adjoint or scattering integral methods):

- no resolution analysis
- often without synthetic inversions
- visual analysis and data fit

INVERSION = DATA FITTING + UNCERTAINTY ANALYSIS + ...



„classical“ linear &
smaller-scale
tomographic problems

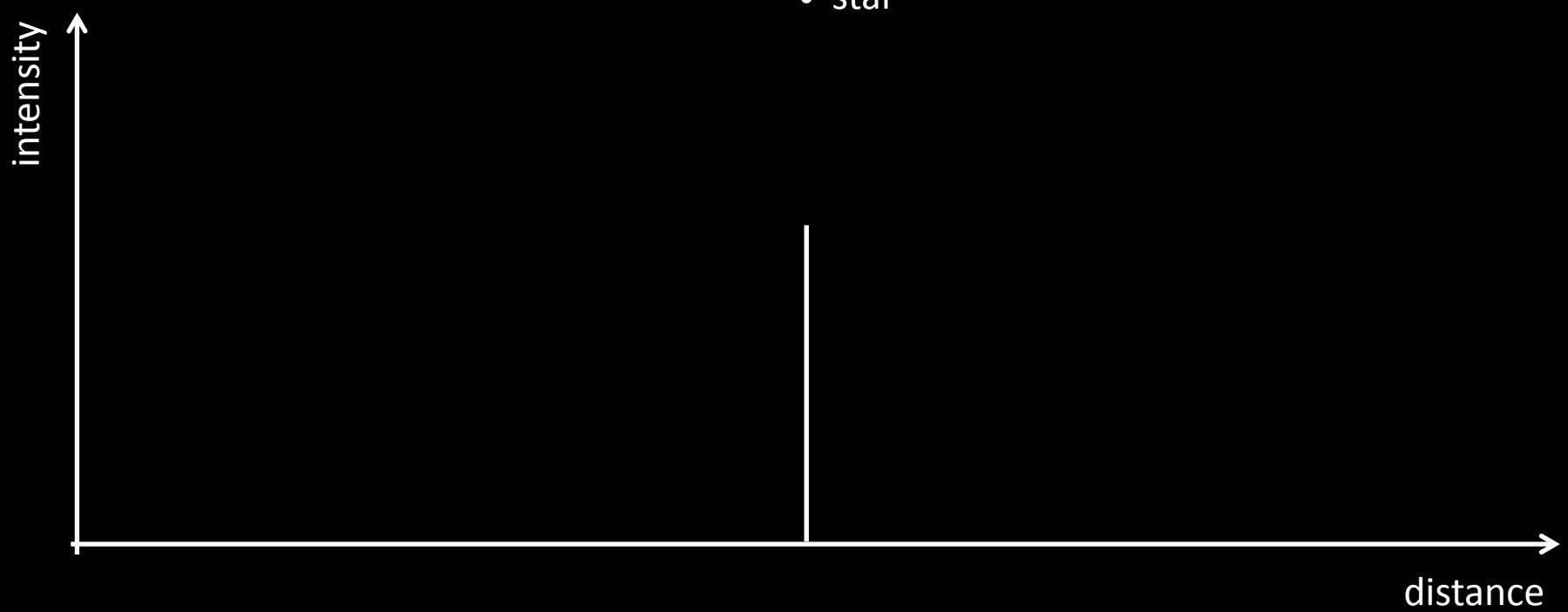
full waveform
and adjoints,
large-scale

Have we really made progress
in solving inverse rather than just data fitting
problems ?

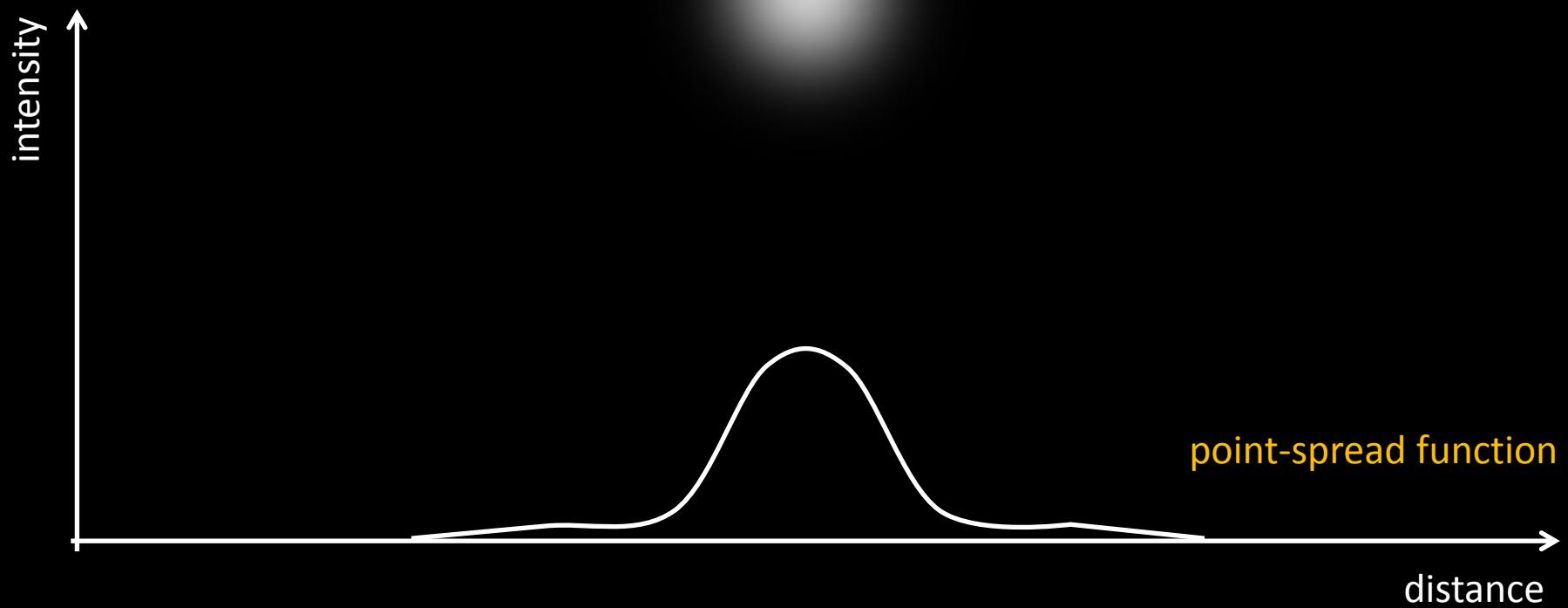
A wide-angle photograph of the night sky, dominated by the bright, glowing band of the Milky Way galaxy. The galaxy's disk is visible, showing a mix of blue, white, and reddish-brown colors from different star populations and interstellar dust. Numerous small stars are scattered across the dark, black void of space. A single, very bright star is visible in the lower-left quadrant.

Astronomical detour

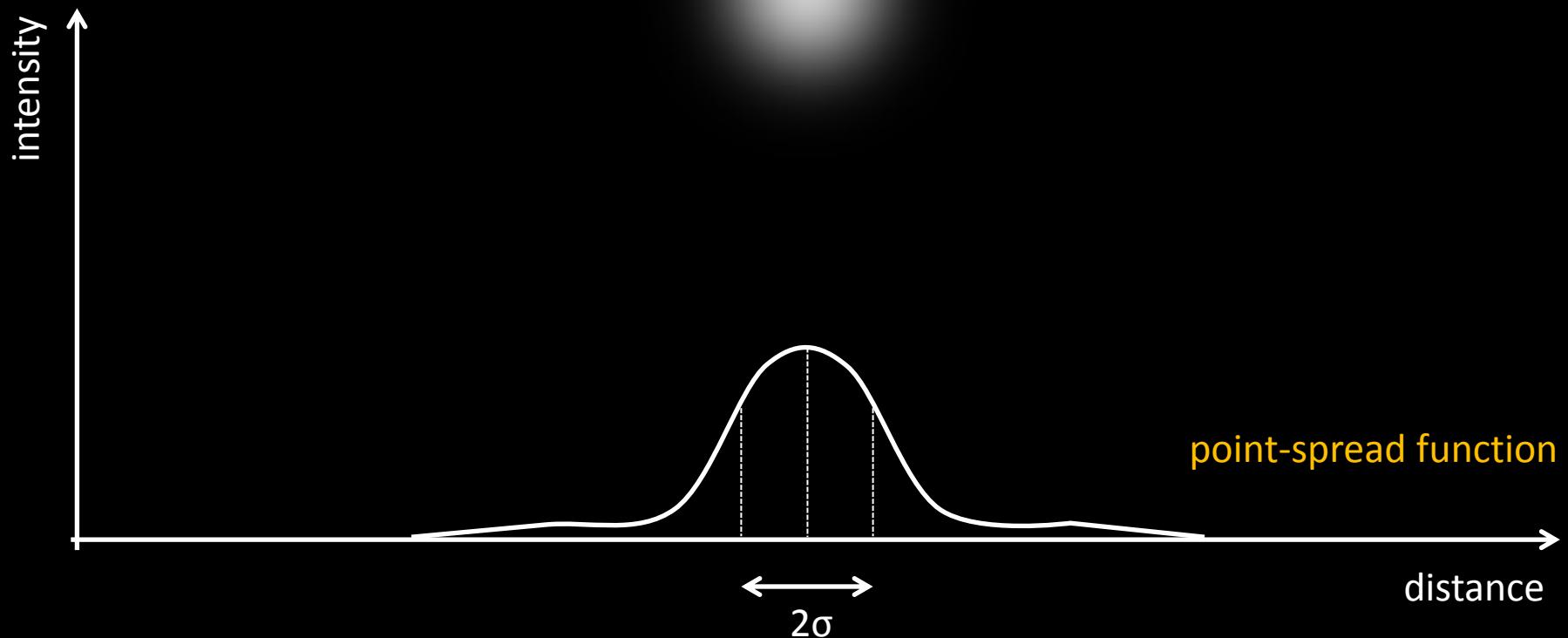
perfect (non-existent) telescope

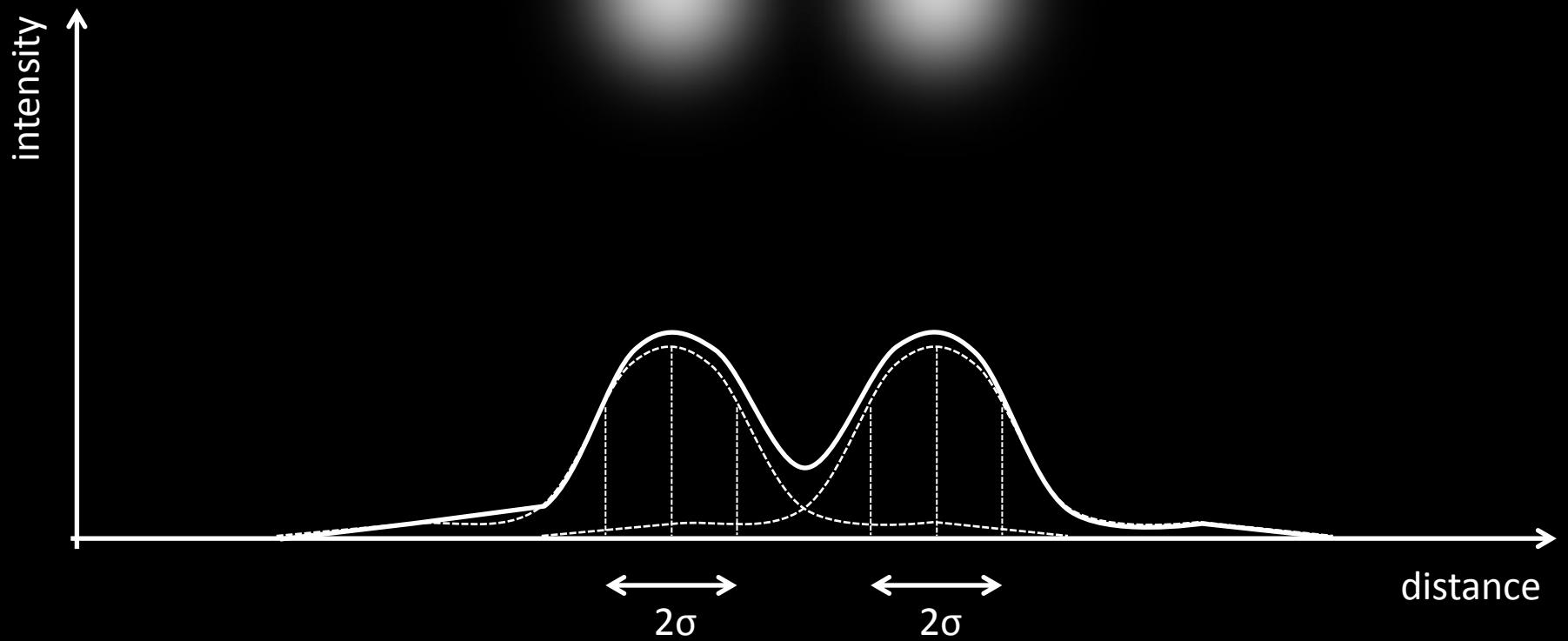


imperfect (real) telescope

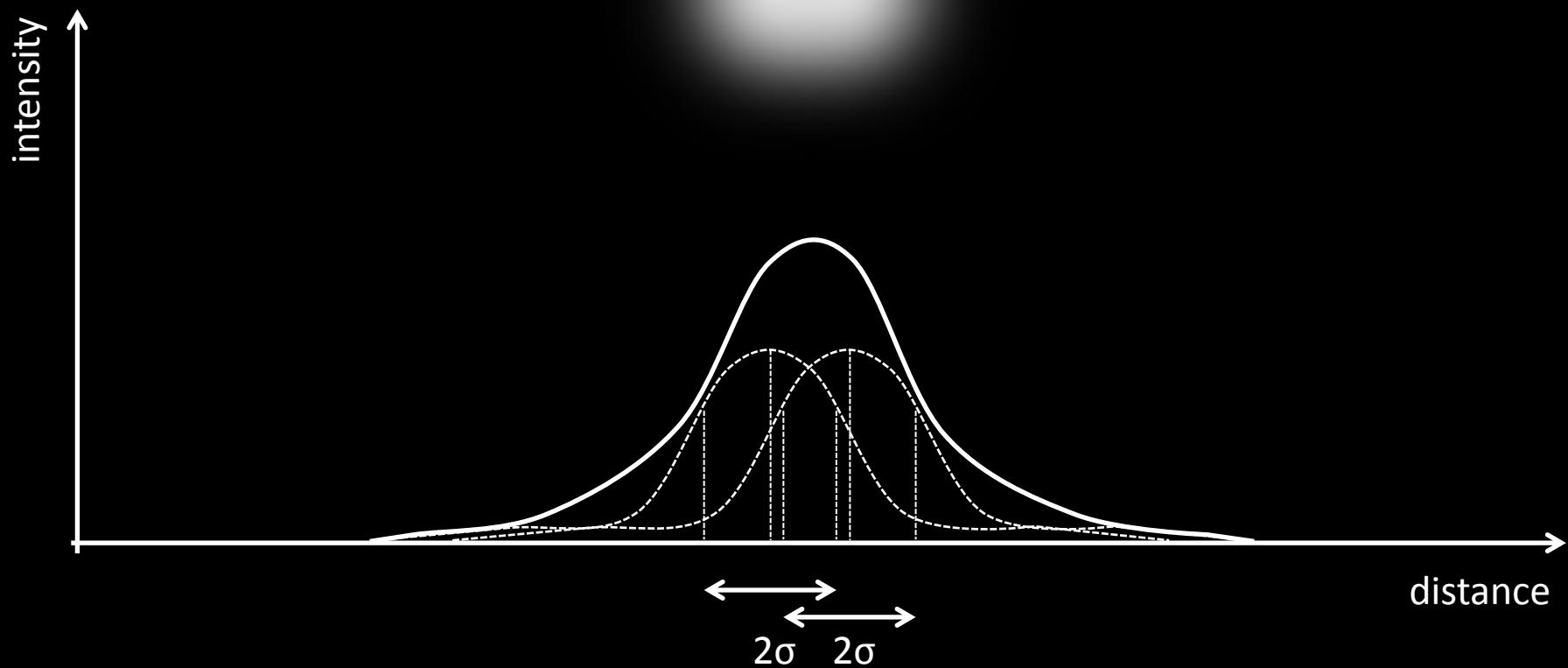


imperfect (real) telescope





Definition: resolution length = 2σ





Back to Earth

Waveform tomography

- Misfit for model \mathbf{m} : $\chi(\mathbf{m})$
 - frequency-dependent traveltimes, time-frequency misfits, amplitudes, ...
- Iterative minimisation of χ to reach optimal model: \mathbf{m}_{opt}
 - steepest descent, conjugate gradients, Newton-like methods, ...

Waveform tomography

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- Approximate tomographic equivalent of the point-spread function:

structural heterogeneity

seen through the tomographic telescope

- x_0

Hessian of χ at m_{opt} : $H(\vec{x}, \vec{x}_0)$

Waveform tomography

- Misfit for model m : $\chi(m)$
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- Approximate tomographic equivalent of the point-spread function:
- Interpretations of the Hessian:
 - Point-spread function
 - Inverse posterior covariance
 - Extremal bounds analysis

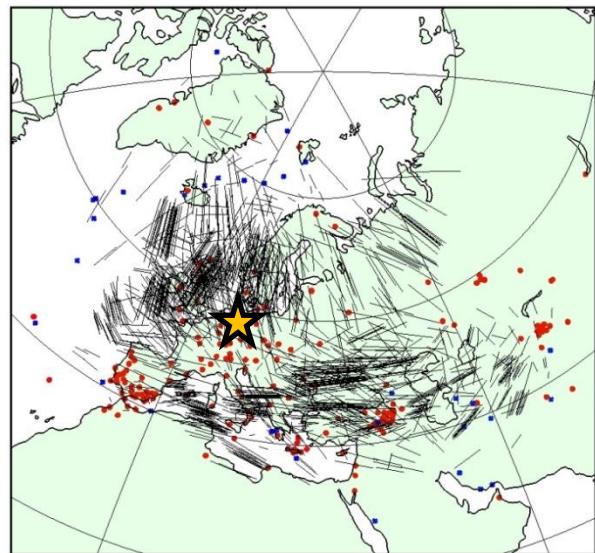
→ To make a first step towards a quantitative resolution analysis, we have to get efficient access to the Hessian!

- Computation of $H \cdot \delta m$ via an extension of the adjoint method
 - (1) 2 forward simulations
 - (2) 2 adjoint simulations (time-reversed)

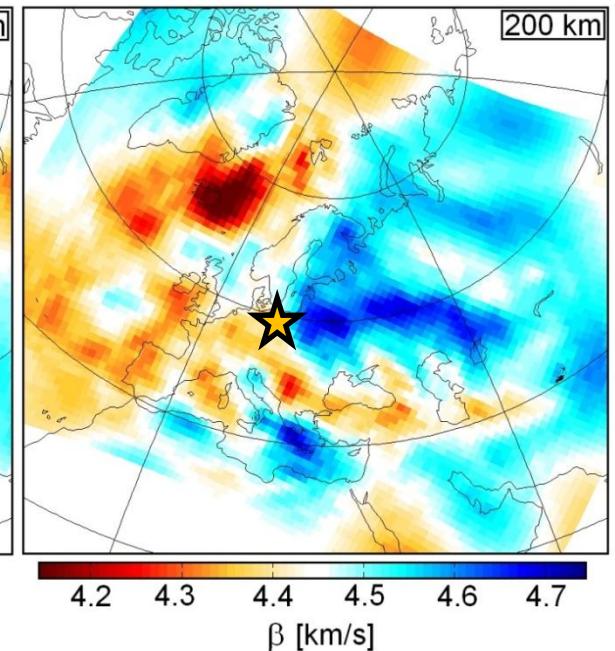
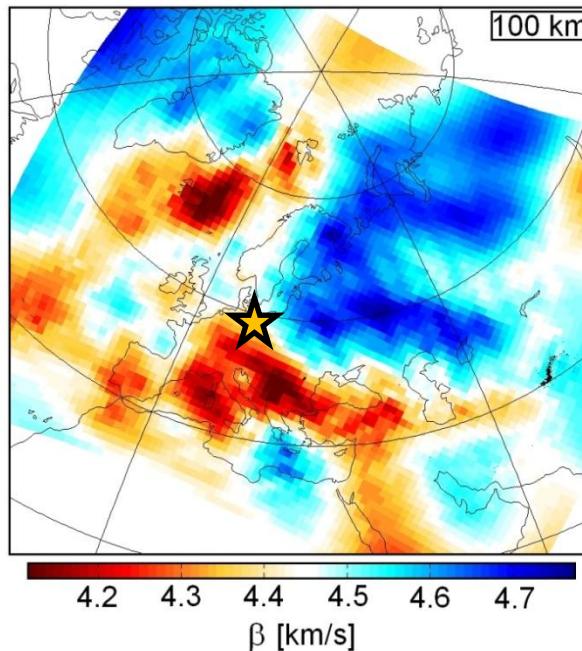
Fichtner & Trampert, 2011. **Hessian kernels of seismic data functionals based upon adjoint techniques.** Geophys. J. Int.

- Efficient computation of $\mathbf{H} \cdot \delta\mathbf{m}$ via an extension of the adjoint method
- Example:** $\delta\mathbf{m}$ = point-localised S velocity perturbation at ★

central ray segments

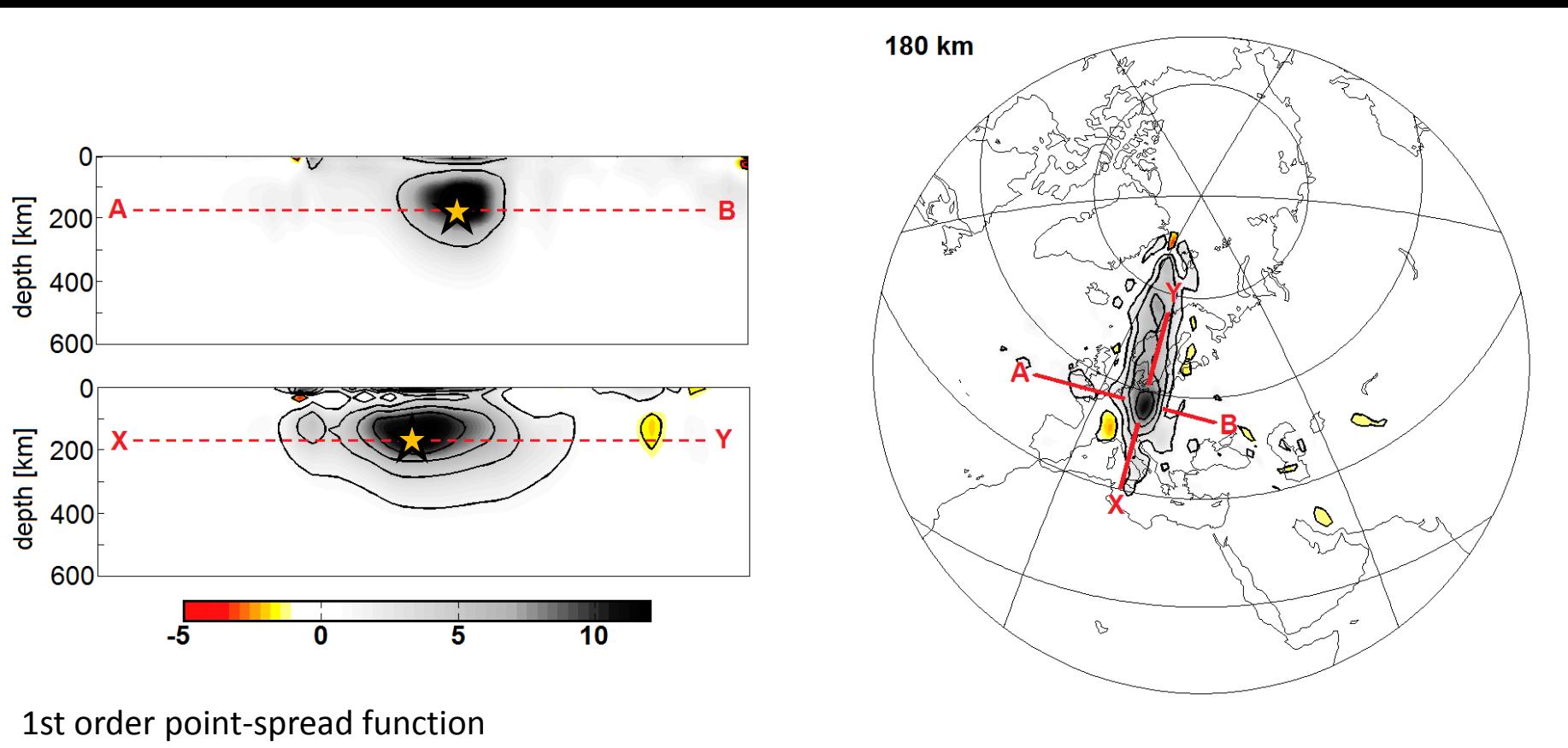


S velocity

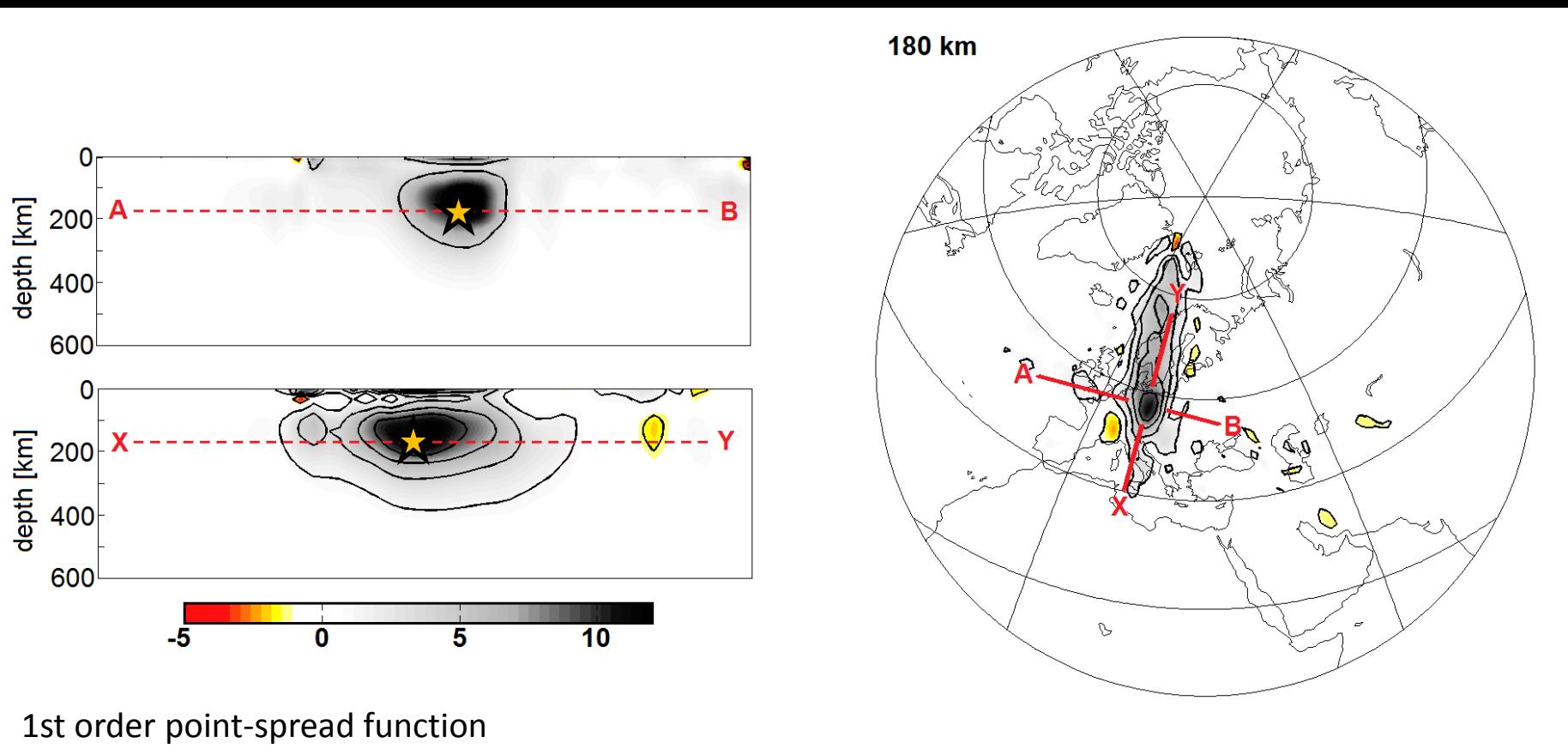


dominant period: $T = 100$ s

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- Example:** $\delta\mathbf{m}$ = point-localised S velocity perturbation at ★

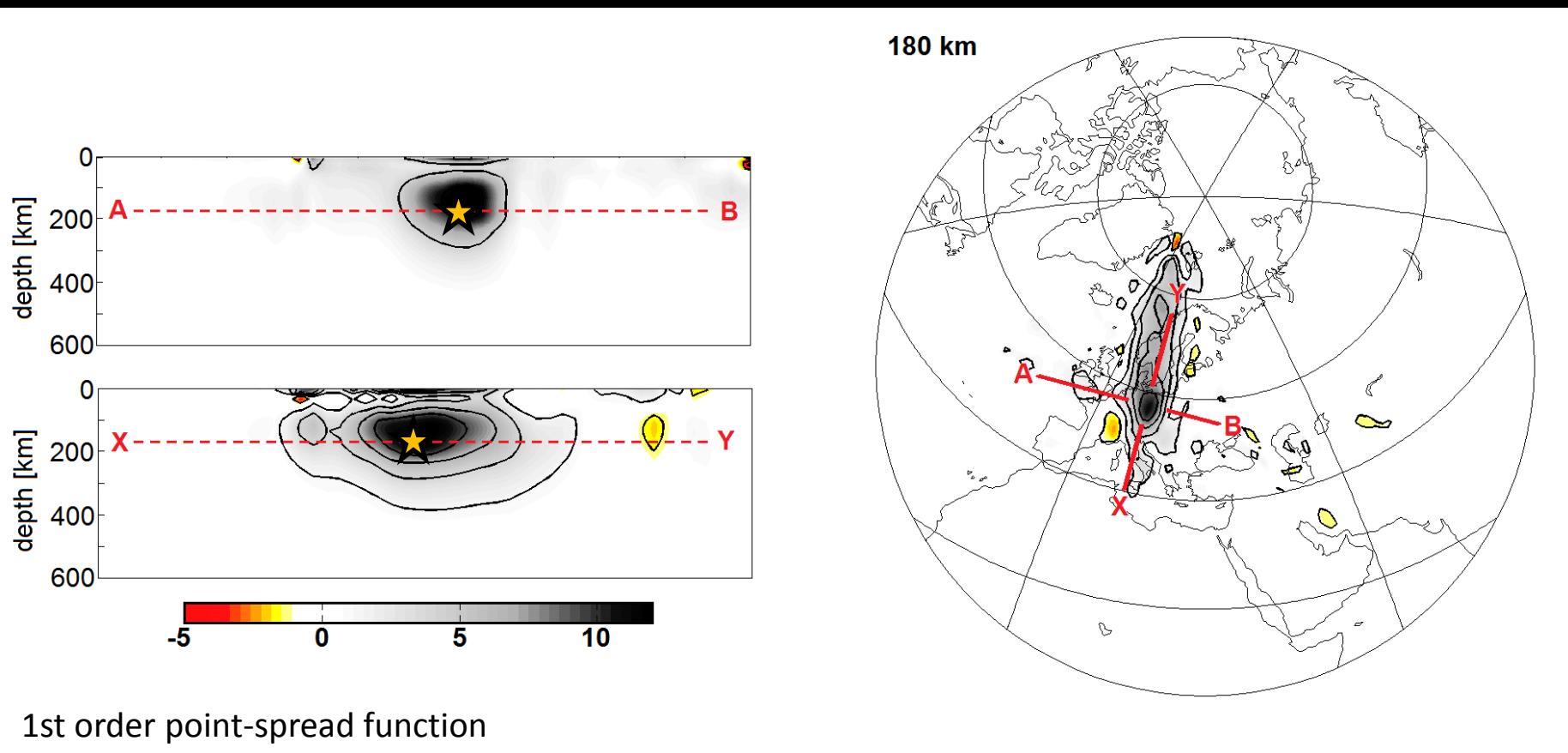


- Efficient computation of $H \cdot \delta m$ via an extension of the adjoint method
- Example:** δm = point-localised S velocity perturbation at ★



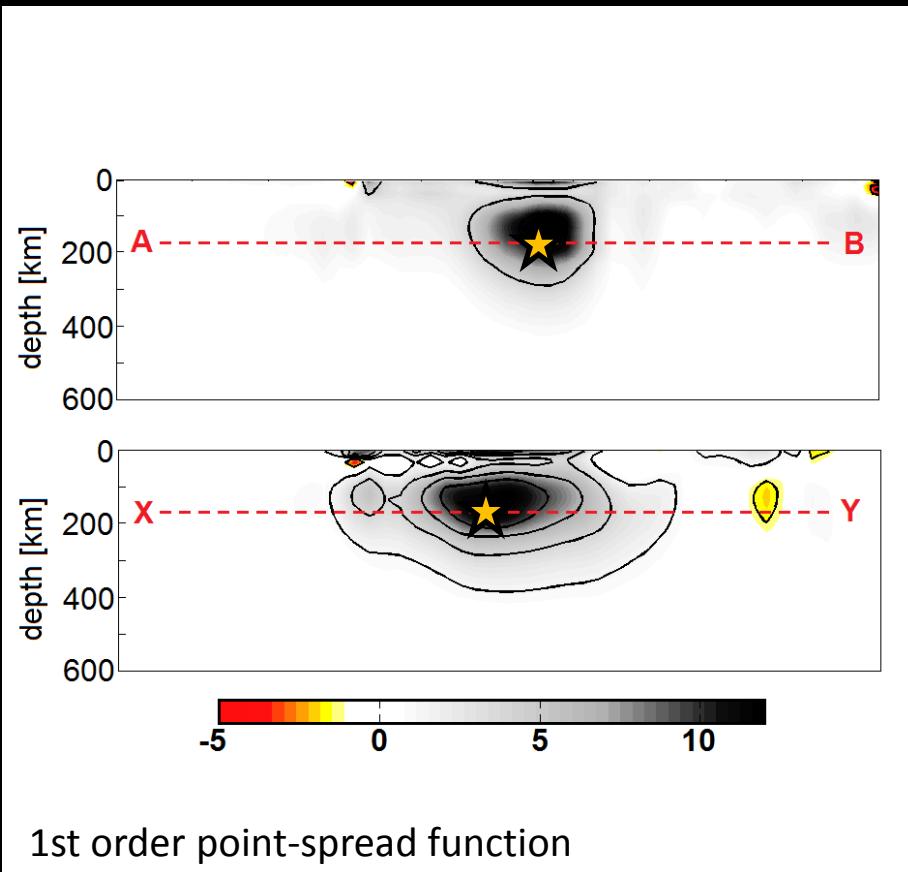
- Tomographic point-spread functions are **not** generally
 - symmetric
 - positive definite

- Efficient computation of $\mathbf{H} \cdot \delta\mathbf{m}$ via an extension of the adjoint method
- Example:** $\delta\mathbf{m}$ = point-localised S velocity perturbation at ★



- Ideally for every point in the model volume
- Prohibitively expensive

- Parameterise the Hessian by a position-dependent Gaussian:



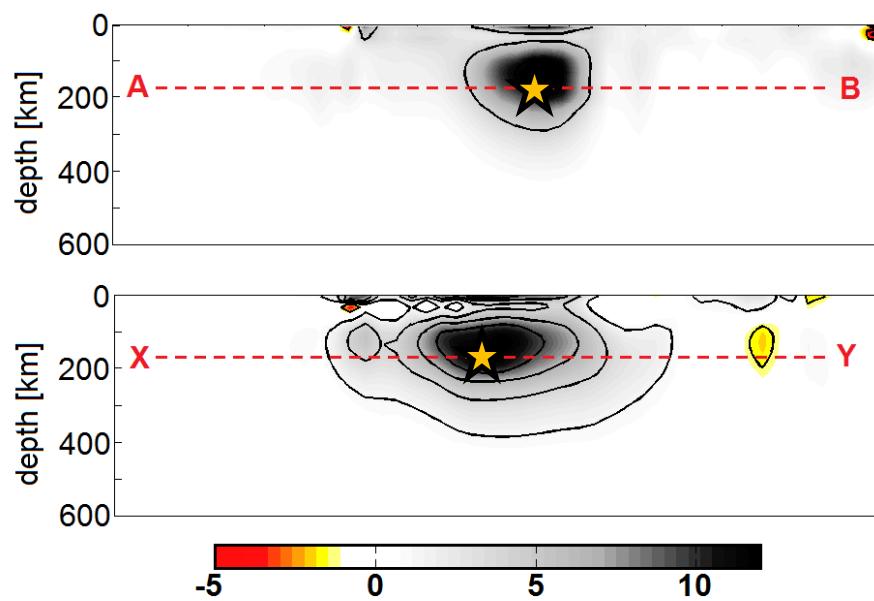
$$H(x, y) \approx A(x) e^{-(x-y)^T C(x) (x-y)}$$

A(x): position-dependent **amplitude**
C(x): position-dependent **width**

- **Generalisation:**
 - Gram-Charlier expansion of H
 - sum over a parent function and its successive derivatives

- **How can we compute the position-dependent parameters of the approximation?**

- Parameterise the Hessian by a position-dependent Gaussian:
- Determine the parameters using Fourier transforms of the Hessian



1st order point-spread function

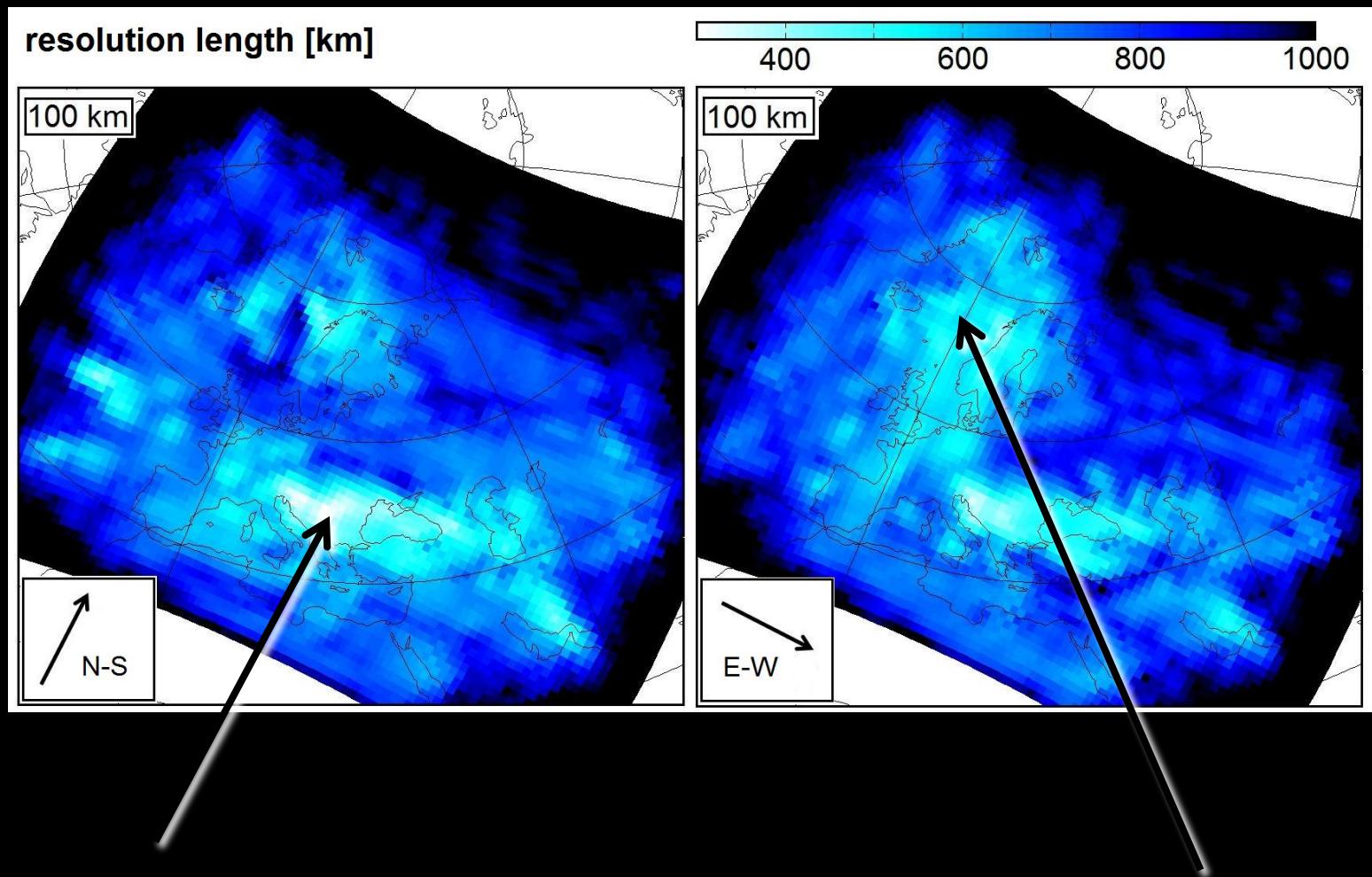
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$$\tilde{H}(x, k) = \underbrace{\int H(x, y) e^{-ik^T y} dy}_{\text{Hessian applied to a sinusoidal model perturbation, } H \cdot \delta m} \propto A(x) e^{i k^T C^{-1}(x) k}$$

Hessian applied to a sinusoidal model perturbation, $H \cdot \delta m$ - easy to compute

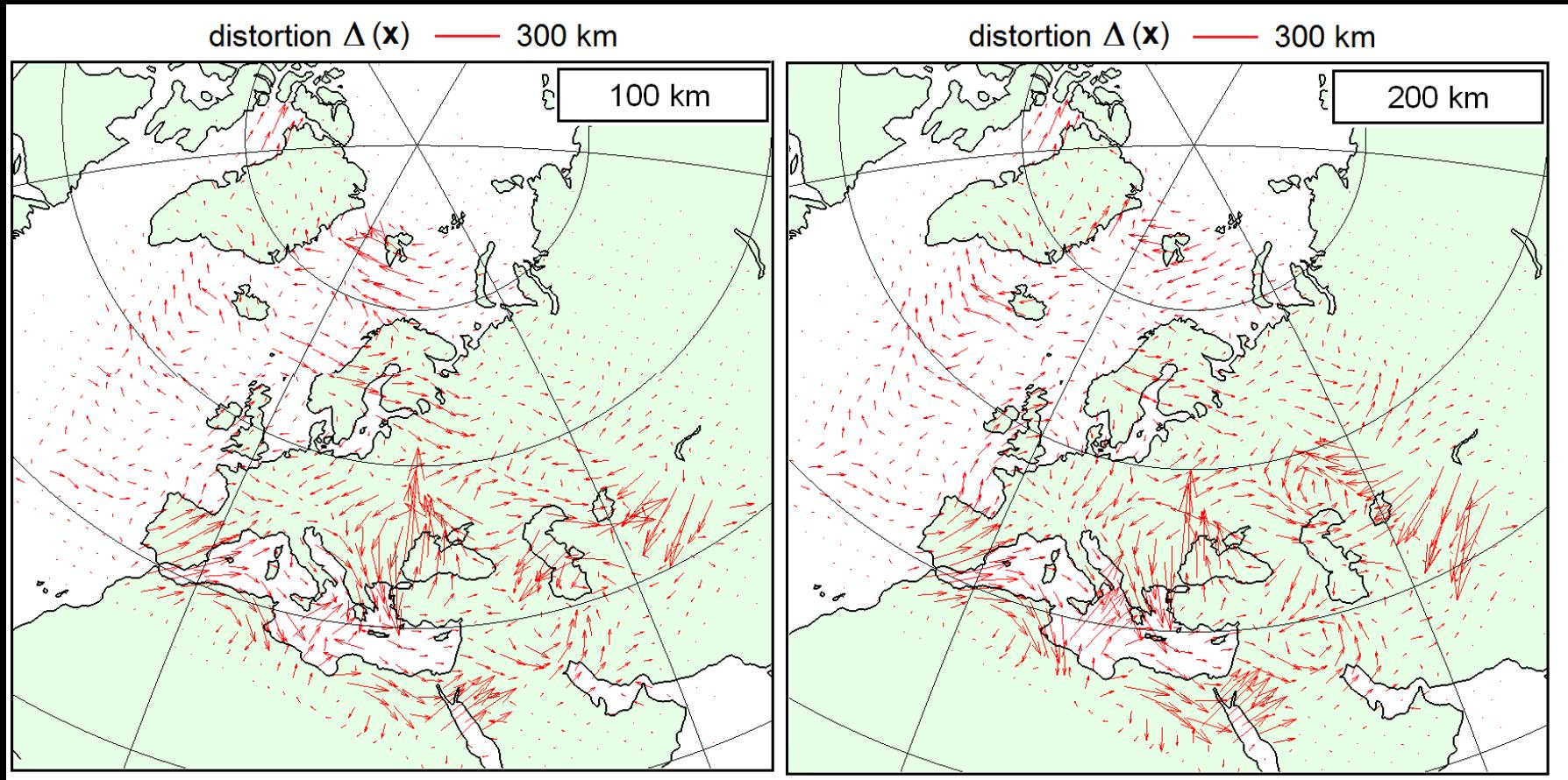
- **Resolution length:** The direction-dependent width of the point-spread function



high resolution in N-S direction
information propagating E-W

high resolution in E-W direction
information propagating N-S

- Point-perturbations are displaced through the tomographic imaging.
- Distortion = [position of point perturbation] – [centre of mass of its blurred image]



→ What you see may actually be somewhere else!

Conclusions

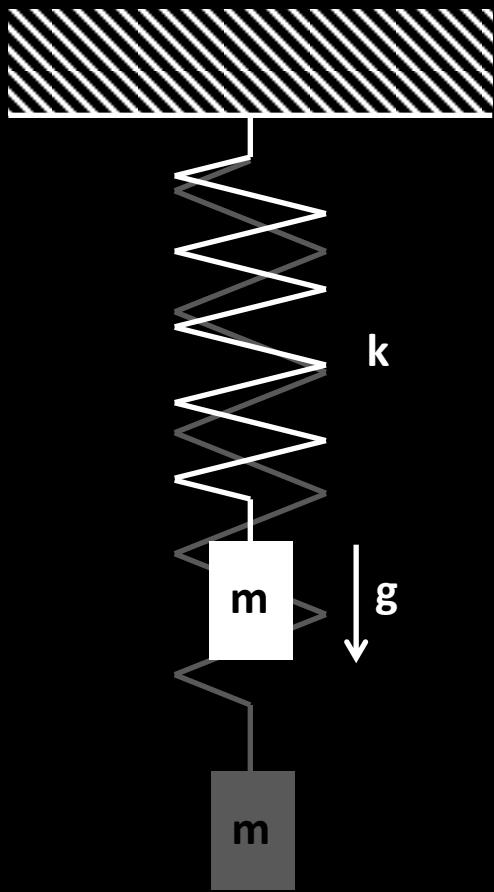
- Resolution analysis based on the Hessian of the misfit functional.
- Quantitative measures of resolution independent from misleading synthetic inversions.
- Method built on
 - Parametrisation of the Hessian (borrowed from astronomy)
 - Computation of the parameters via Fourier transforms
- 3D distributions of **resolution length** and **distortion**

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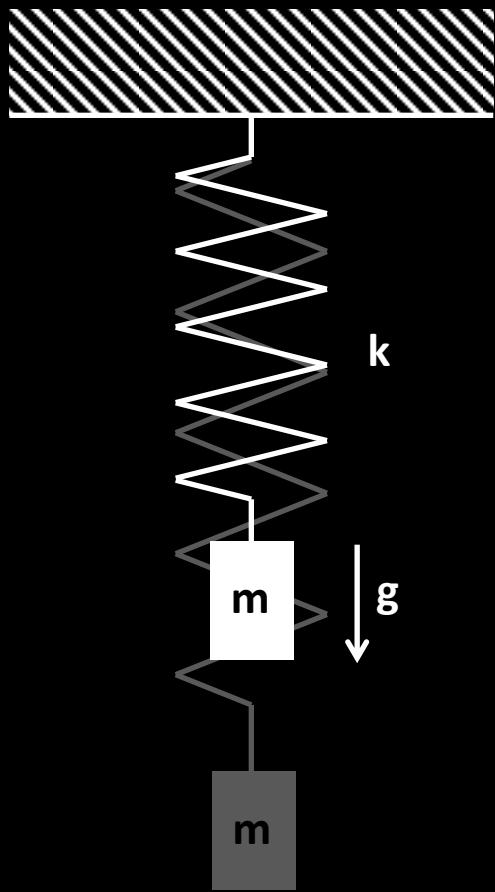
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Corollaries:

- New family of Newton-like methods with an explicitly computed approximate Hessian
- Efficient pre-conditioner for conjugate-gradient algorithms
- Approach to adaptive parametrisation independent from ray theory



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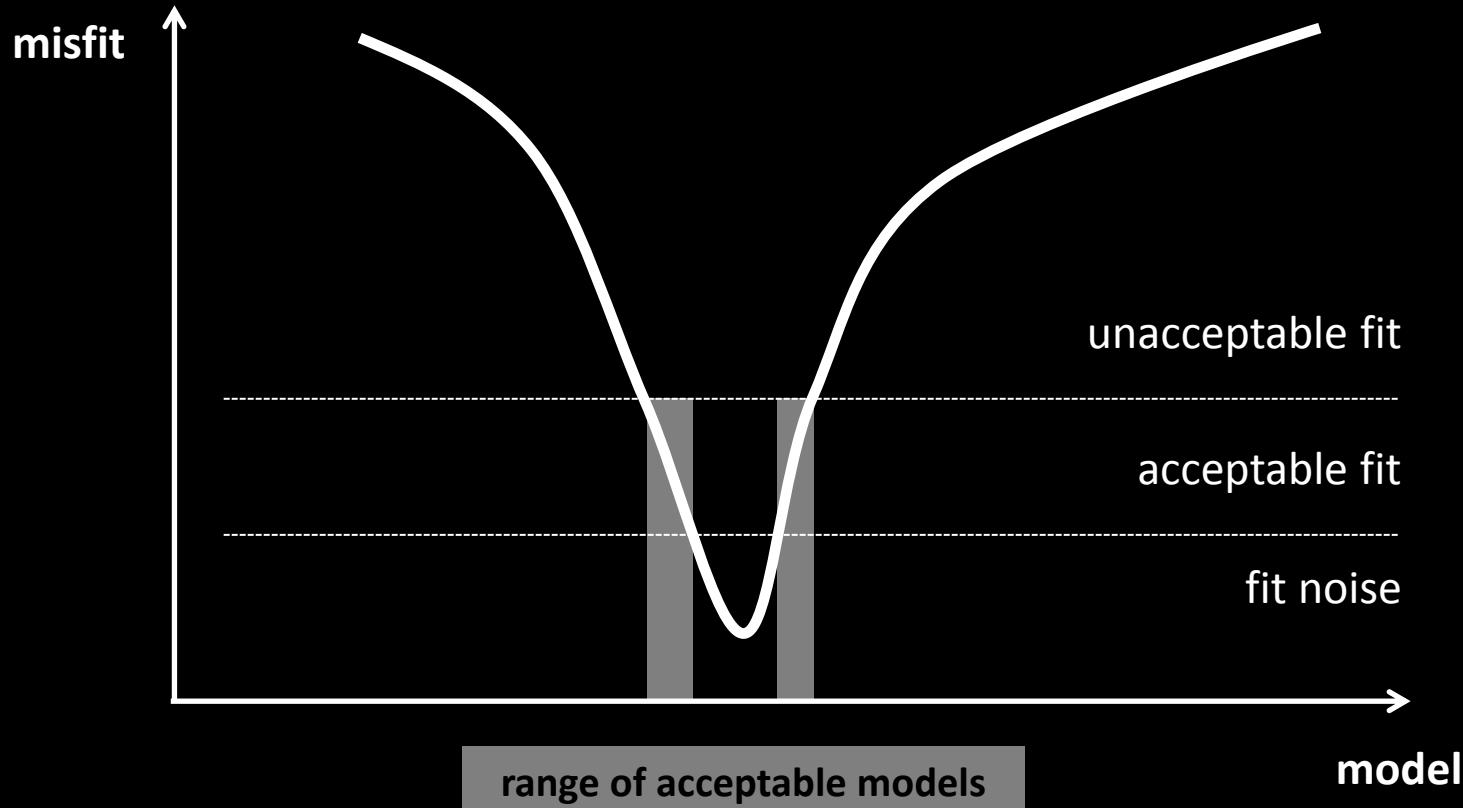


A. Tarantola

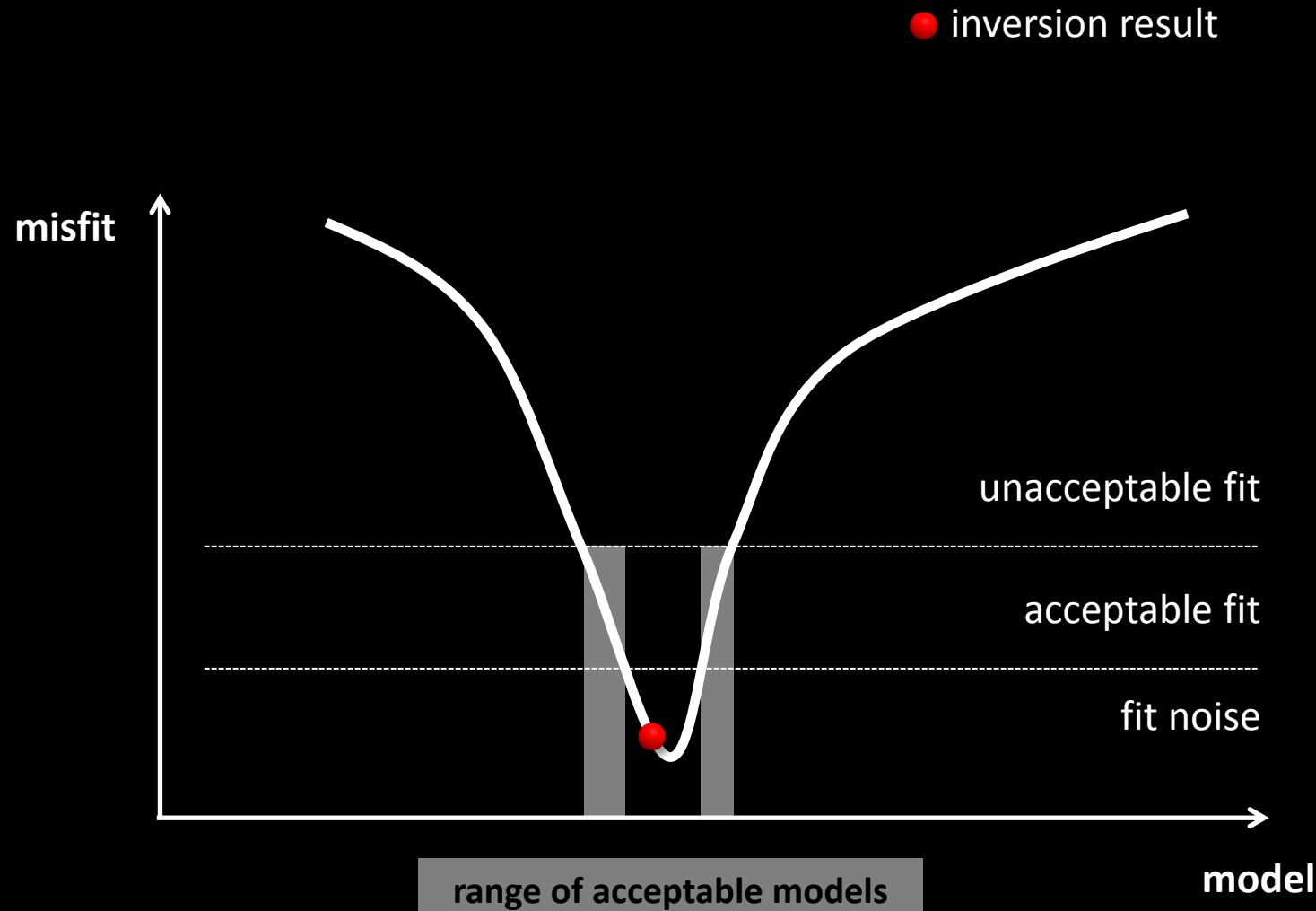
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Thank you very much for your attention !

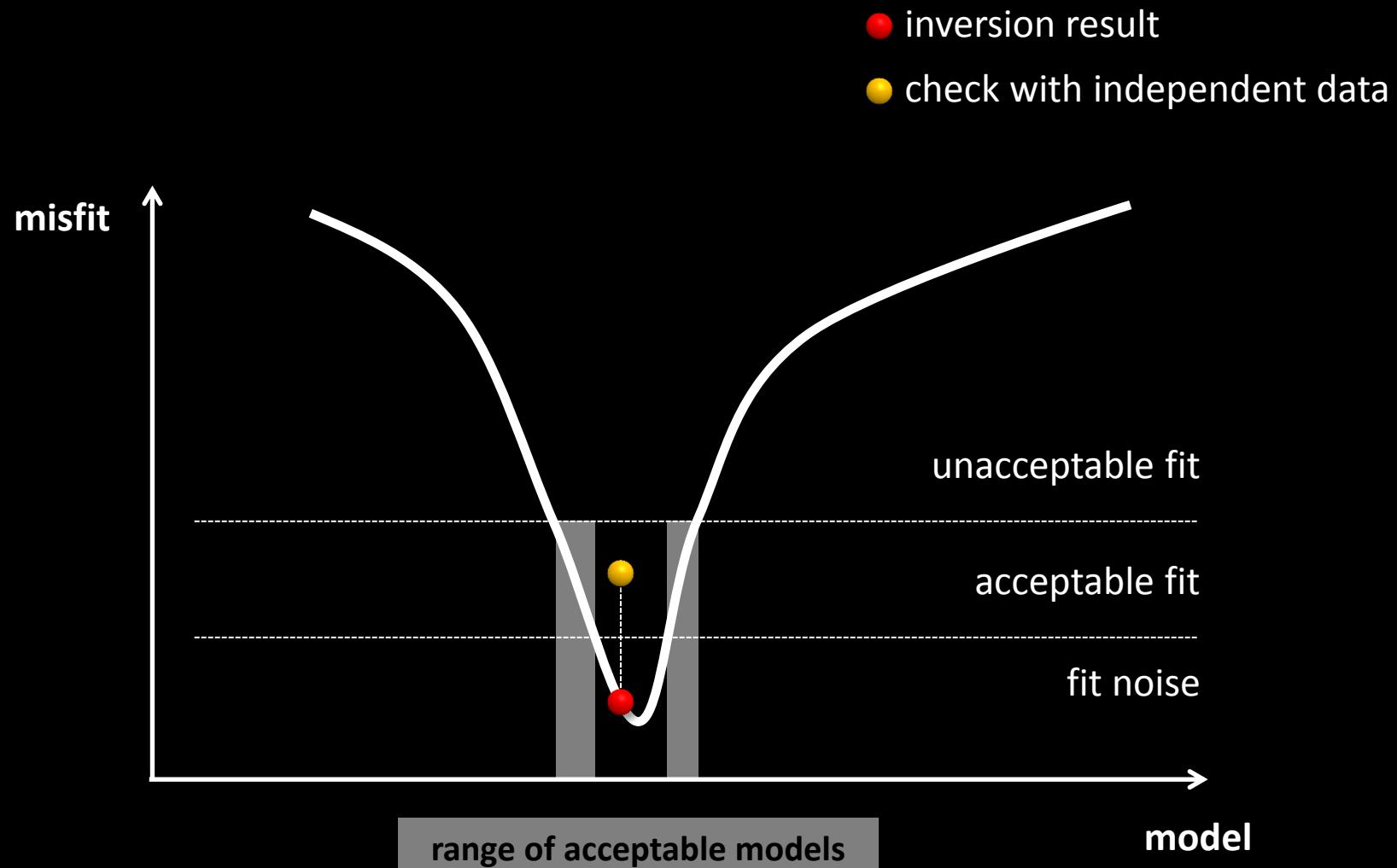
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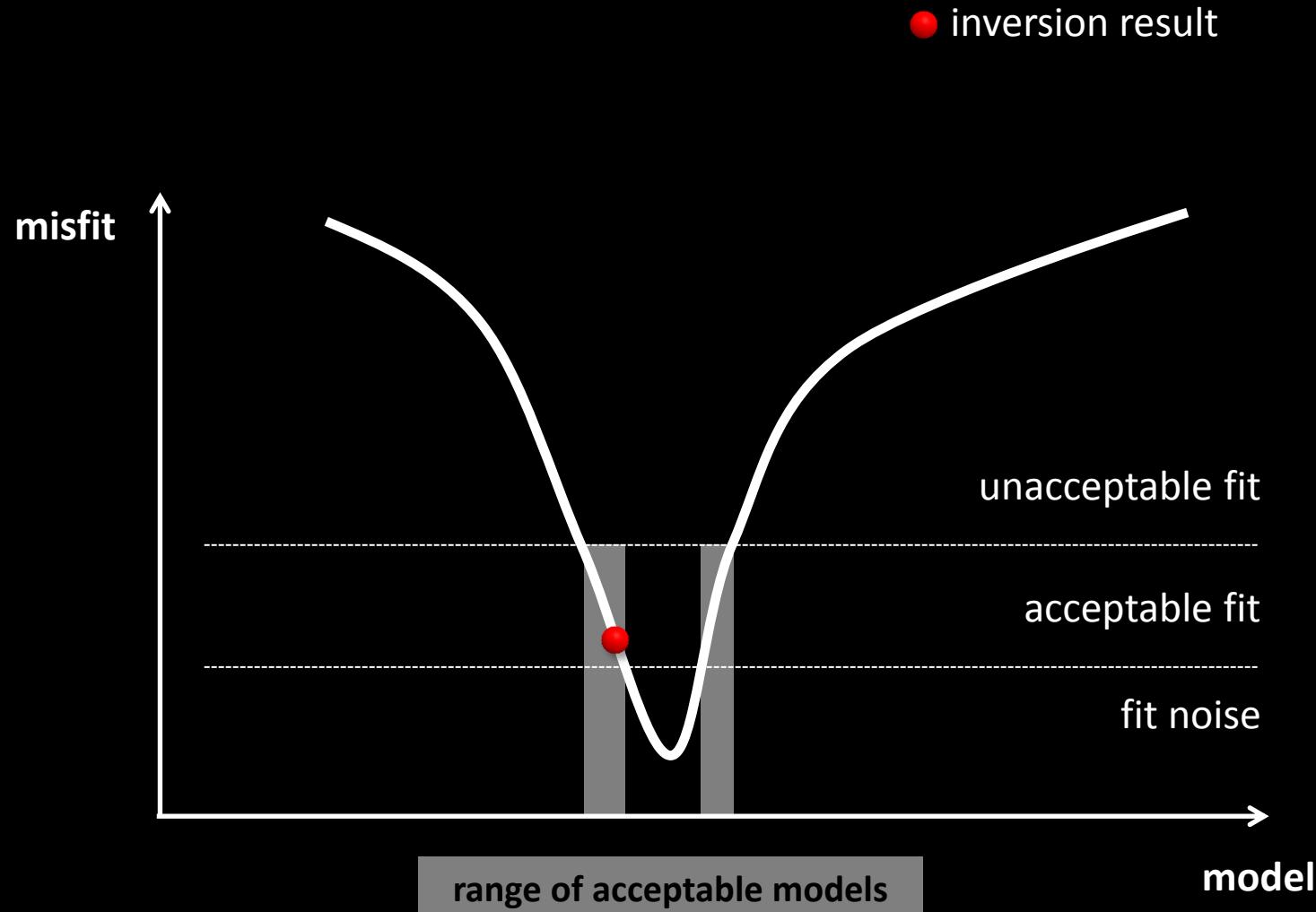
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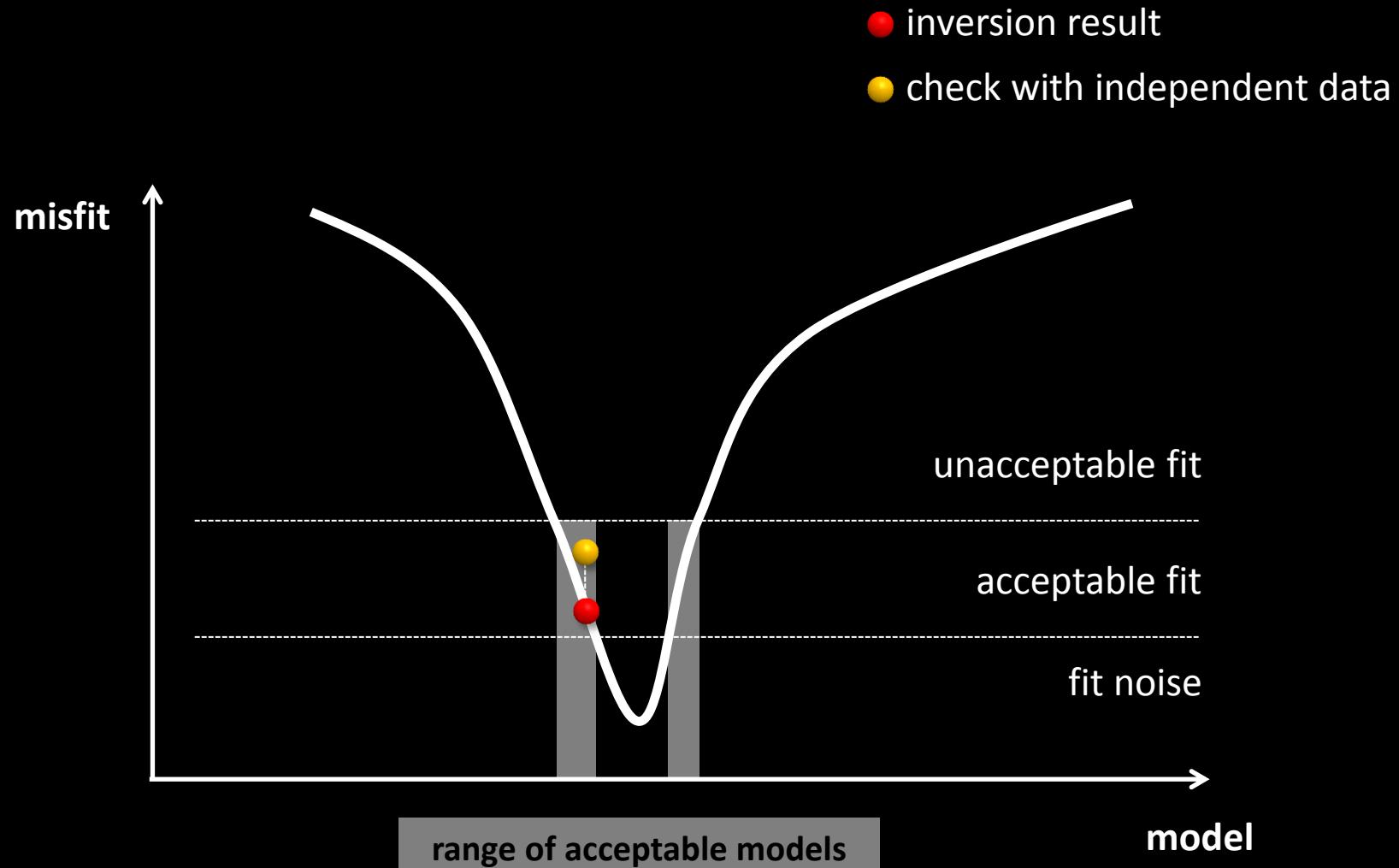
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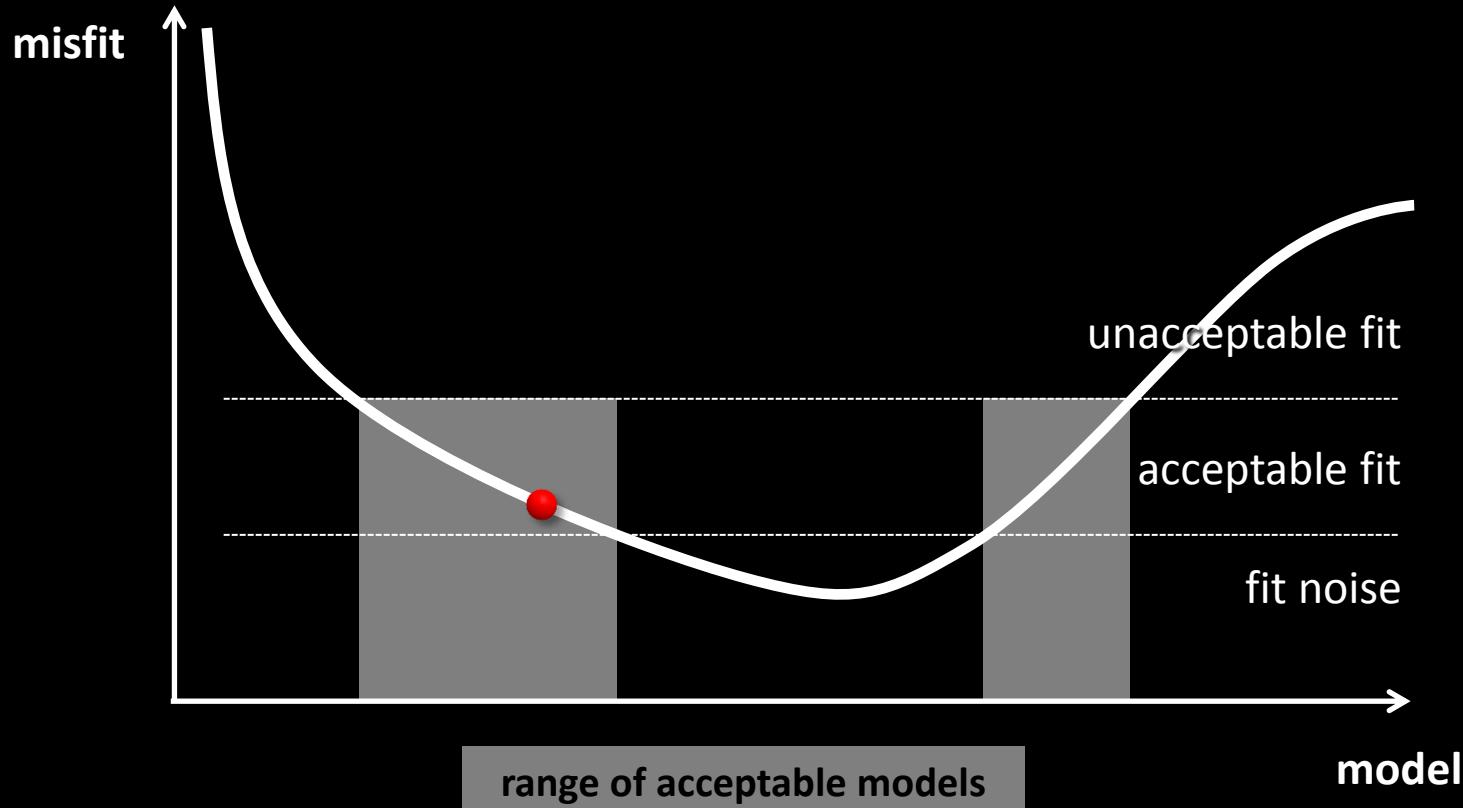
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Reproduction of the errors statisitcs with independent data

- necessary to avoid over-fitting of data and over-structuring of models
- **not sufficient** to ensure the model is well-constrained

