Likelihood-based Optimization in Strong-motion Seismology

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Strong Ground Shaking

0

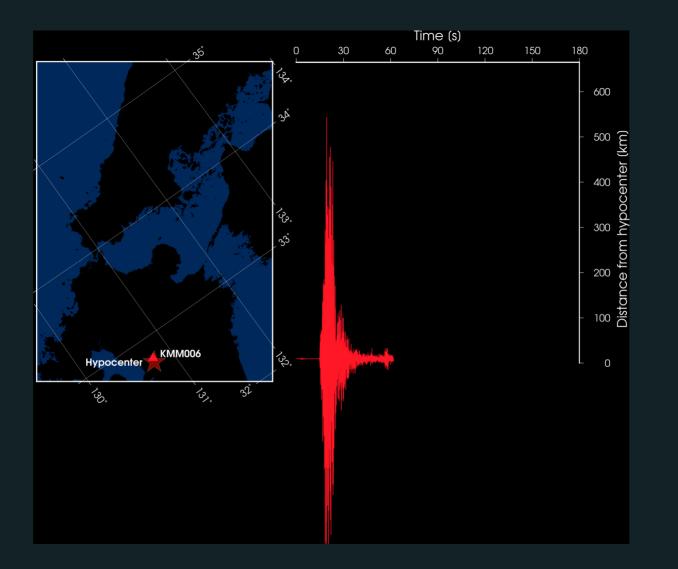
Strong Ground

Shaking

Hiroshi Fukuoka, The Japan Landslide Society

What is strong ground shaking?

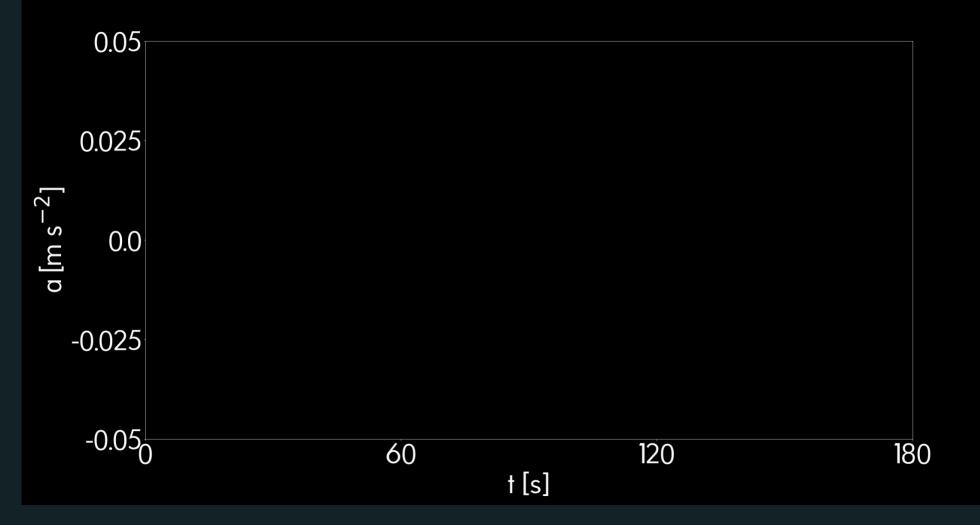
Shaking close to epicenter



What is strong ground shaking?

Shaking close to epicenter

with amplitudes capable to cause structural damage



What is strong ground shaking?

- Shaking close to epicenter
- with amplitudes capable to cause structural damage
- \rightarrow This is of engineering concern!
- Trigger for more disasters (e.g. landslides)
- → <u>Multihazard</u>

The Goal:

Modeling ground-shaking with data-driven methods to reduce expert elicitation and to tackle the increase of data with automation

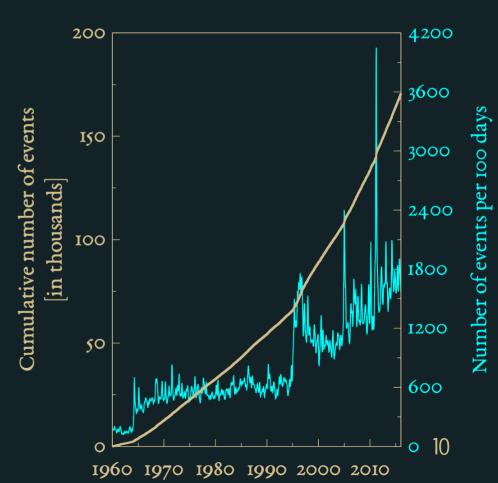
Motivation

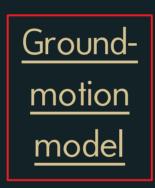
Accumulation of data with time

increasing rate of data accumulation

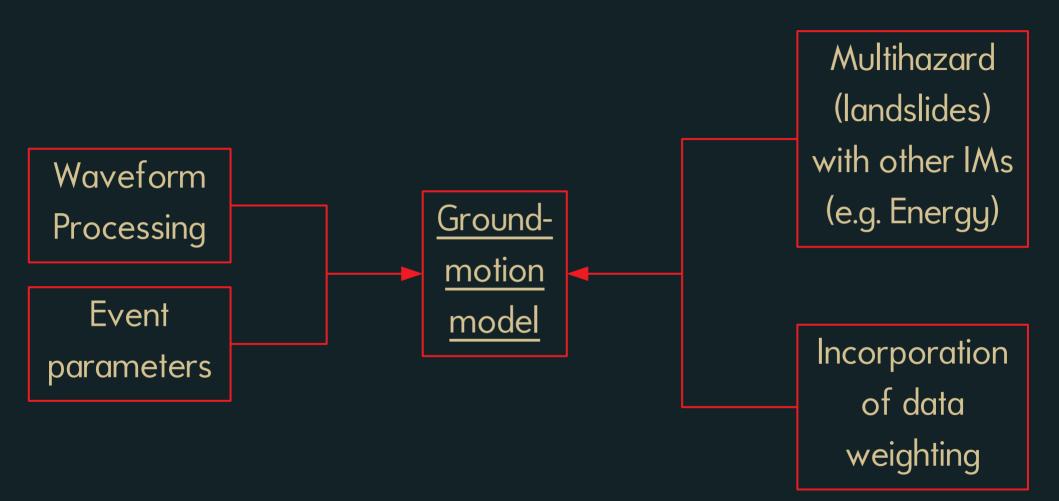
limit of manual processing

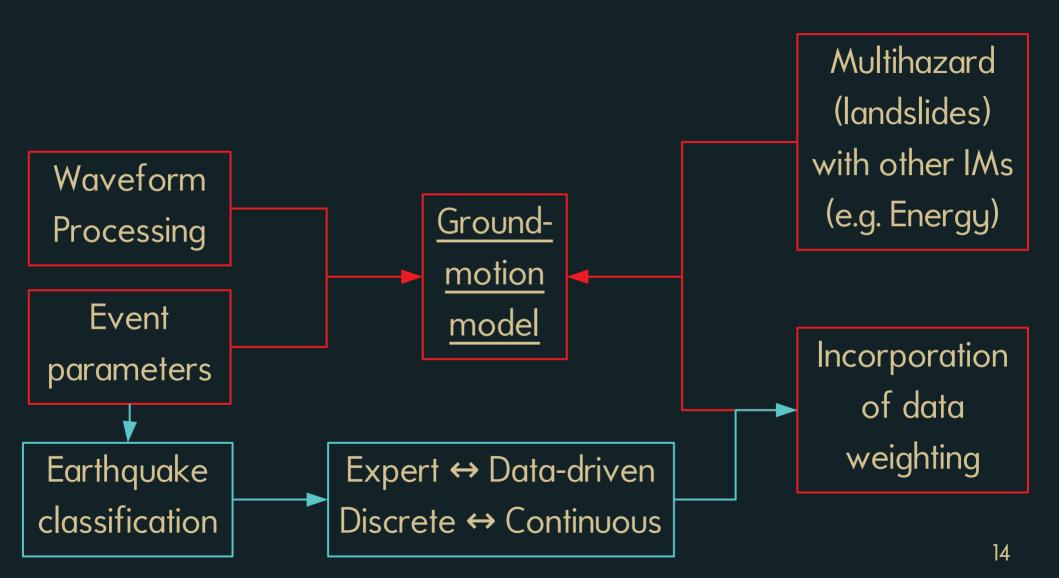
 \rightarrow Automation

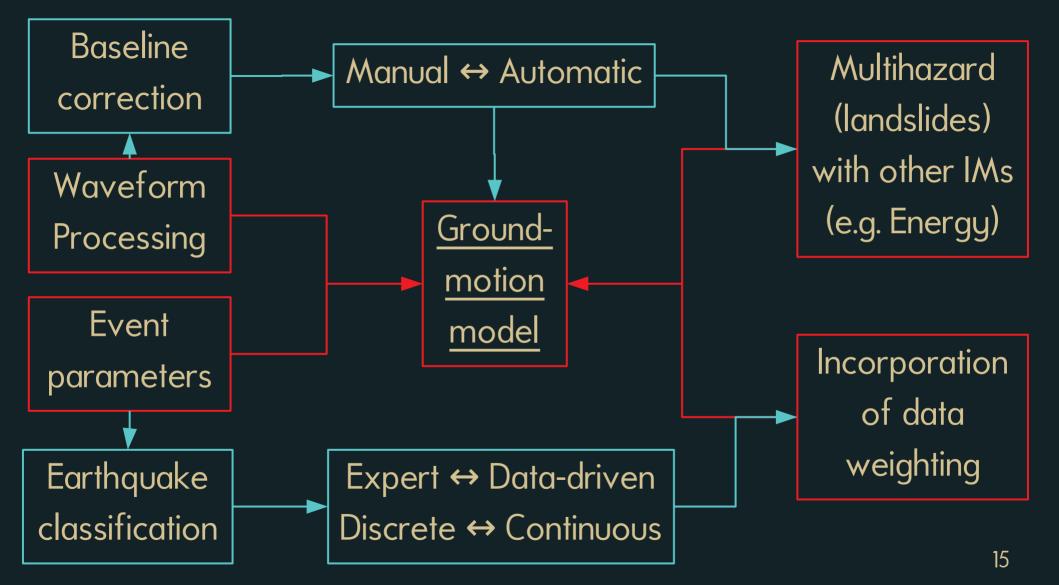






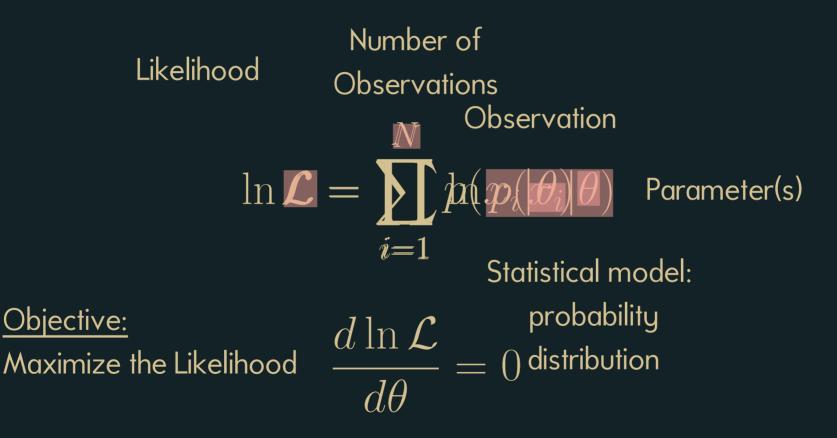


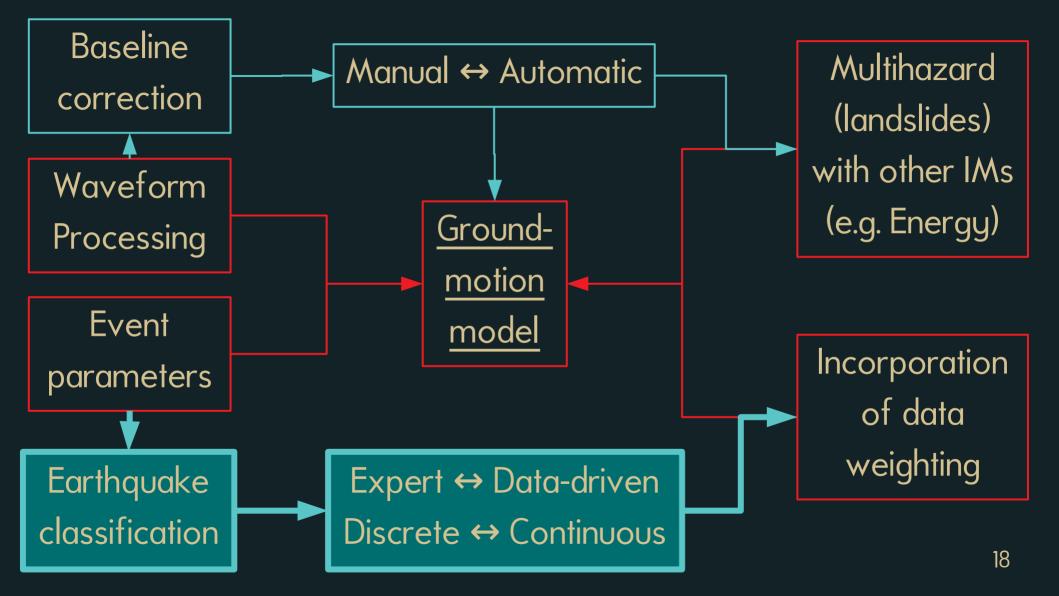




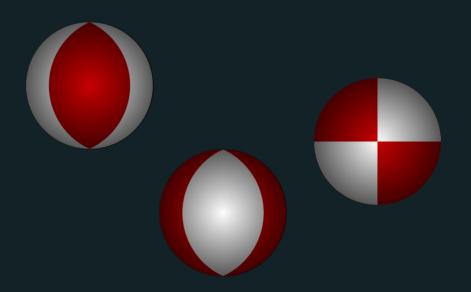
I addressed these questions by a common method

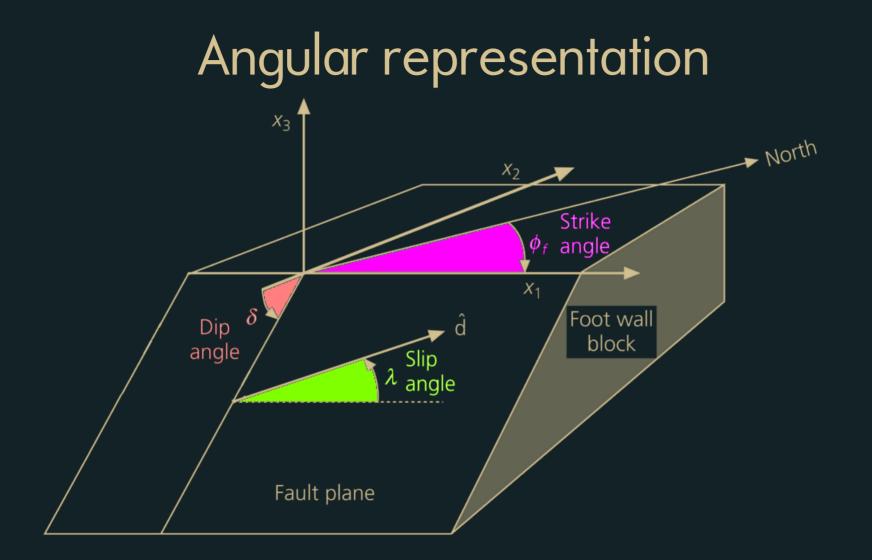
The likelihood



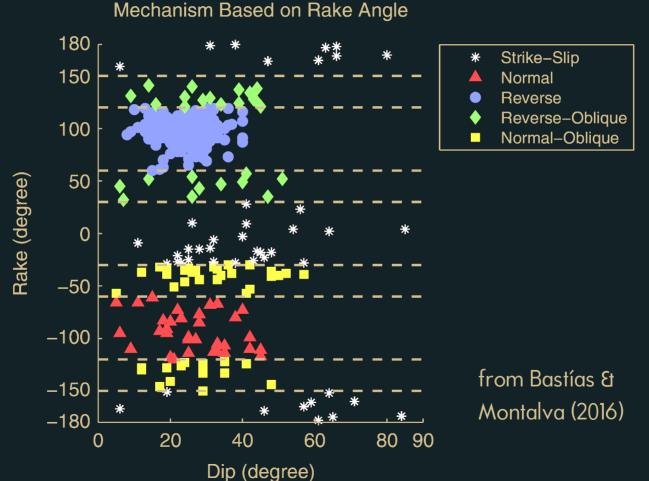


Angular Clusterization with Expectation-Maximization





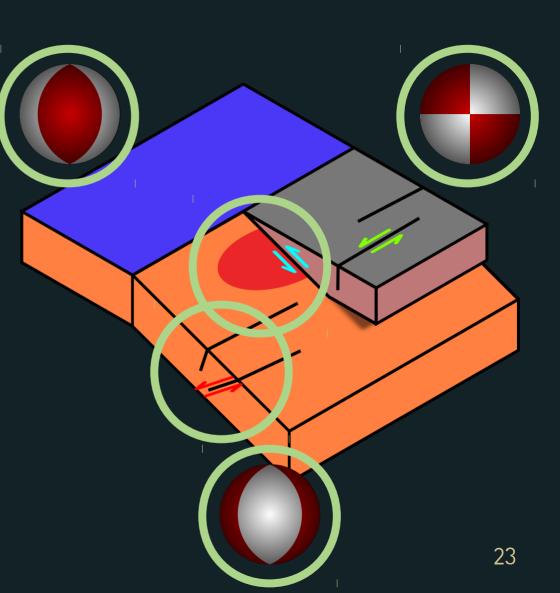
The classical classification

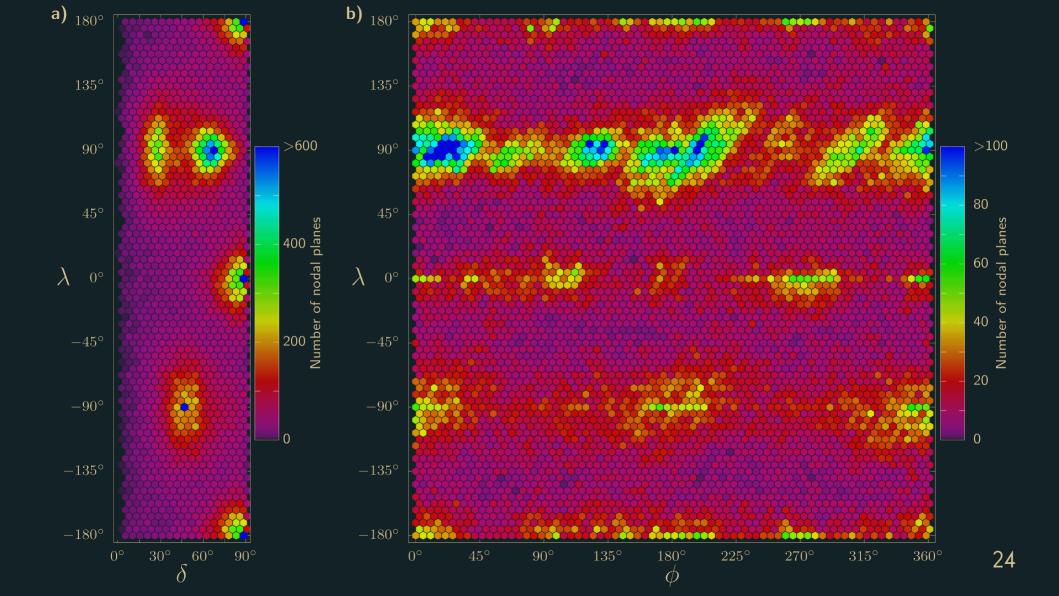


Expectation-Maximization

Mixture of probability distributions

Parameterization of models of latent variables Latent Variables in ACE: Style-of-Faulting Reverse Normal Strike-Slip Tectonic setting Interface Intraslab

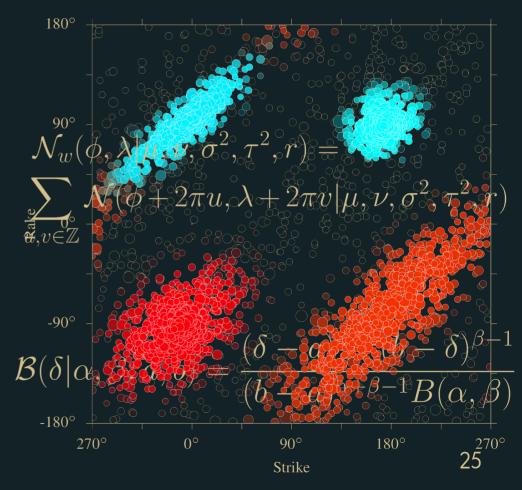




Objective

Identify clusters of same SoF / tectonic setting Model: joint probability **Bivariate** wrapped normal distribution

Beta distribution



discrete VS continuous -

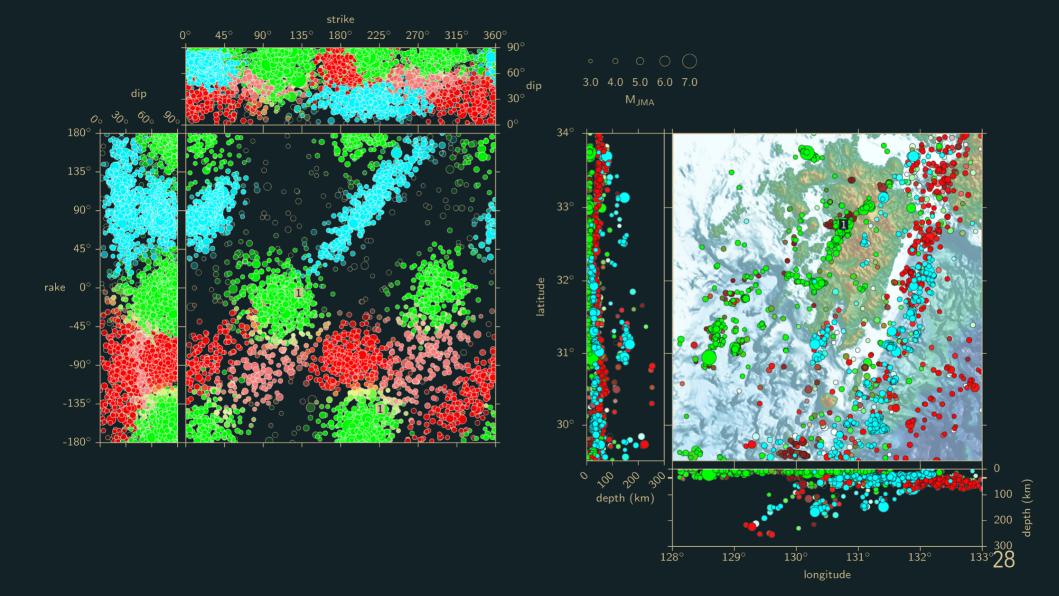
3. Choose your noodles: WHEAT NOODLES 被絕 regular-sized noodles also used for Japanese ramen EGG NOODLES 鍵蛋麵 thin wheat-and-egg noodles with the vibrant flavor of egg 4. Choose your burn: IST DEORES **BURN** 第一熱度 BURN BURN

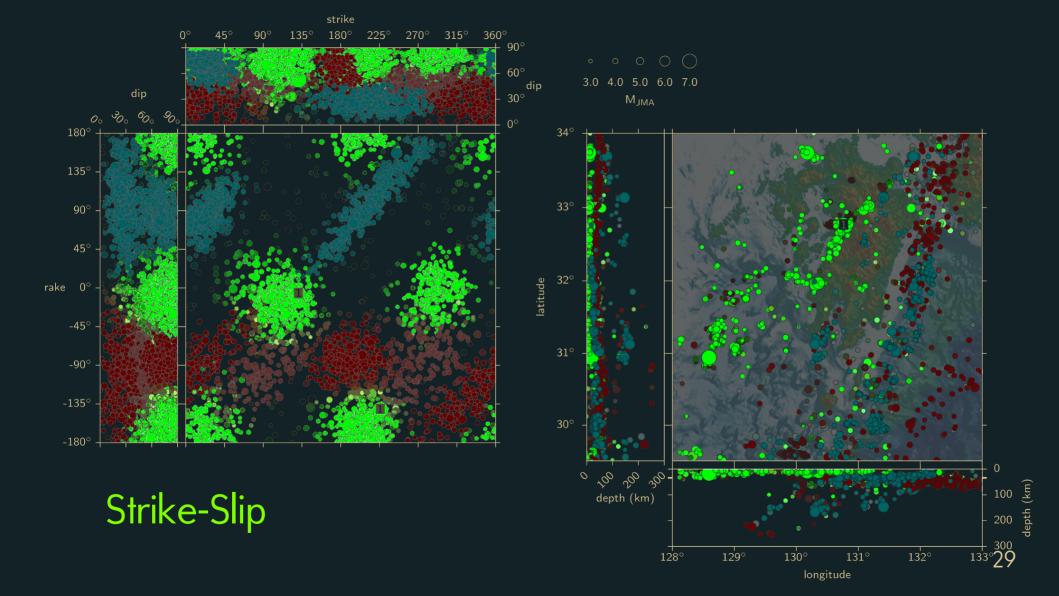
BURN

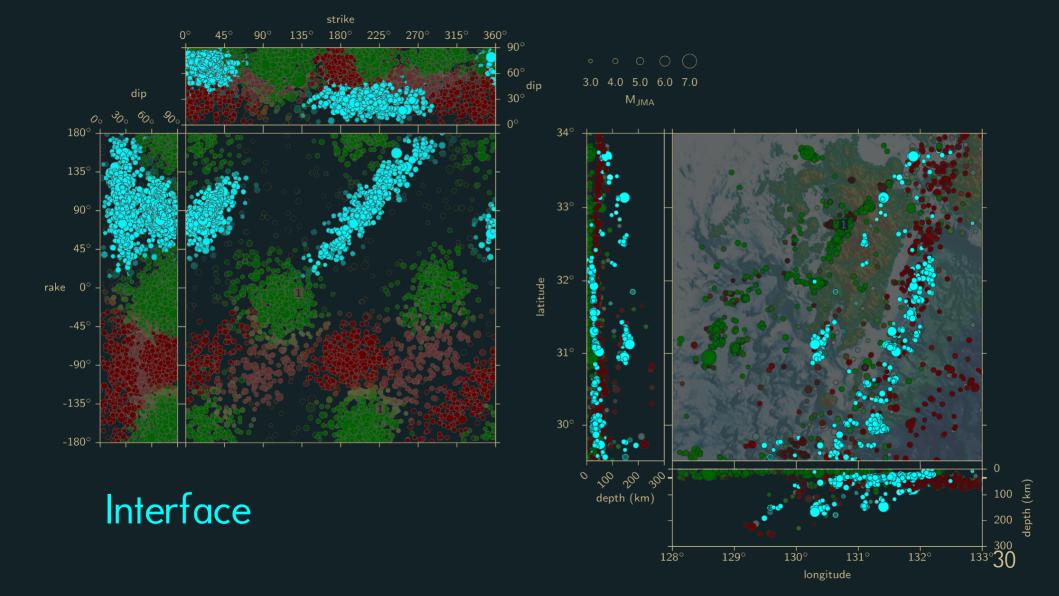
26

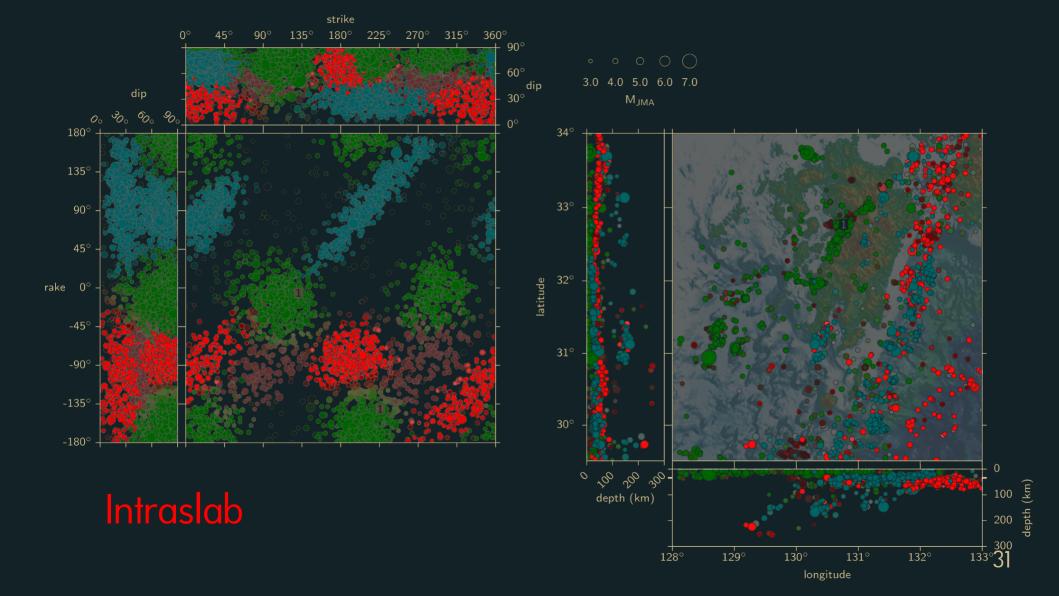


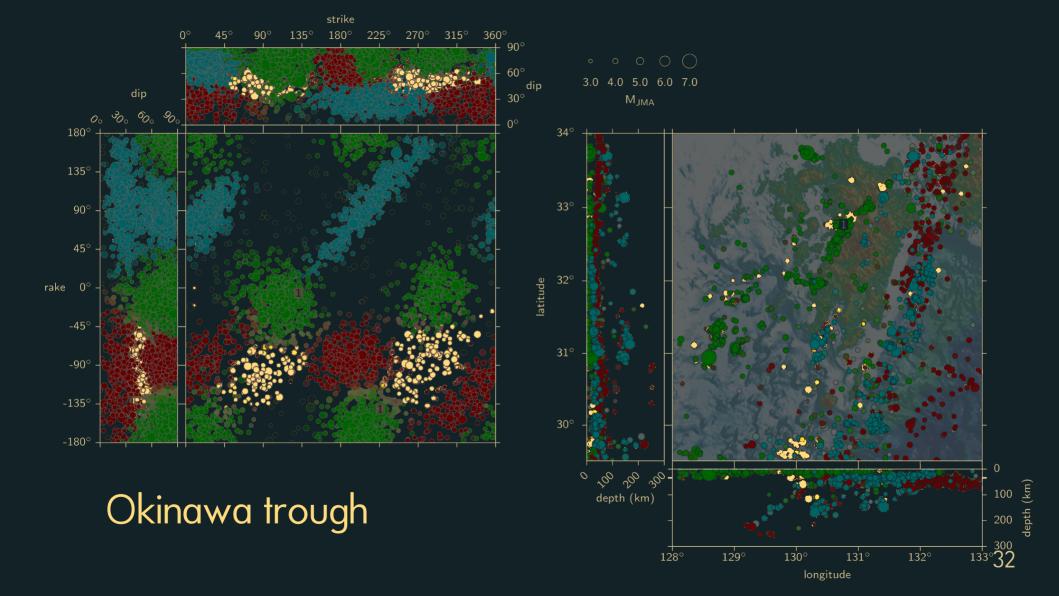
九州(日本) Kyushu (Japan)

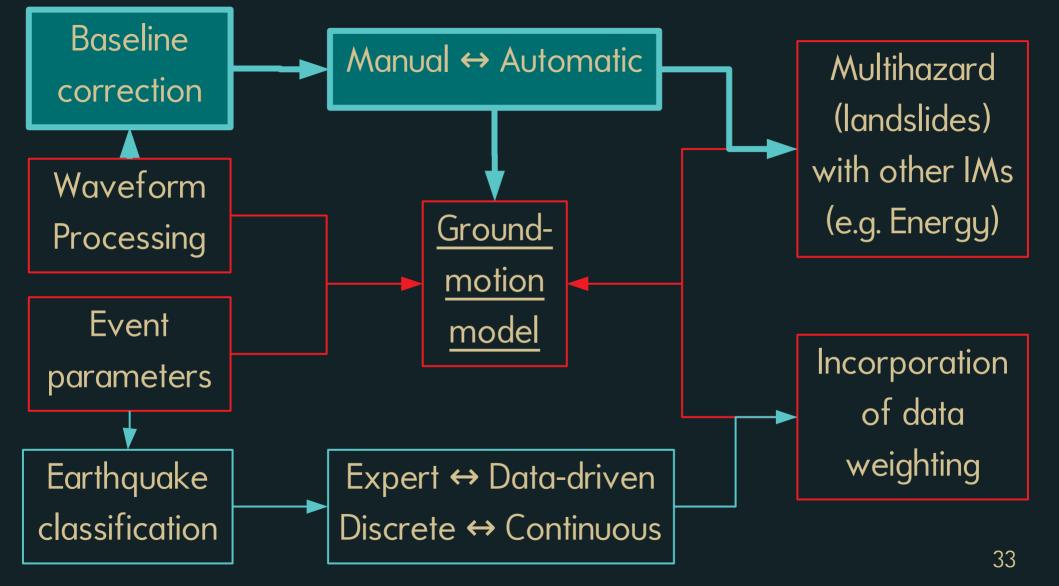












Integrated Combined Baseline Modification

nationalinterestorg

Why correcting?

Spurious discontinuities in accelerograms

Causes:

tilt, station movement

instrument noise

Effects:

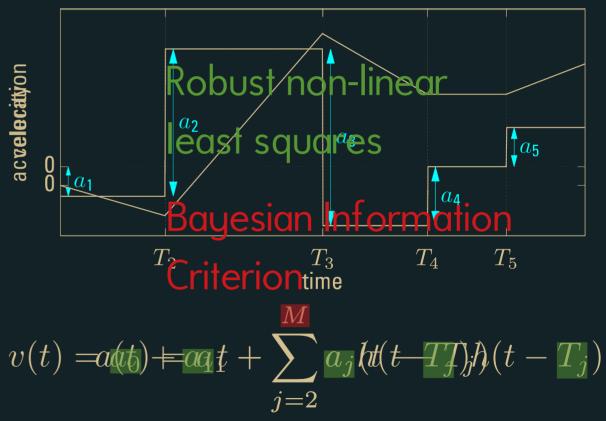
drift in velocity & displacement traces

bias in seismic energy, high period response spectra, PGV / PGD

Objective

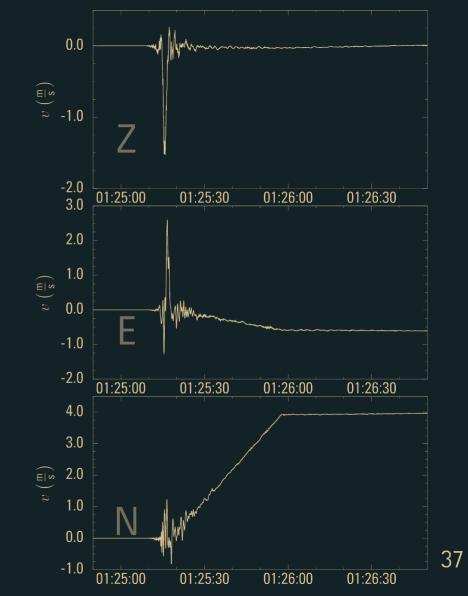
Identify number, amplitudes and times of jumps

use integrated traces (velocity)



Simultaneous jumps on up to all three instrument components

→ reduction of free parameters

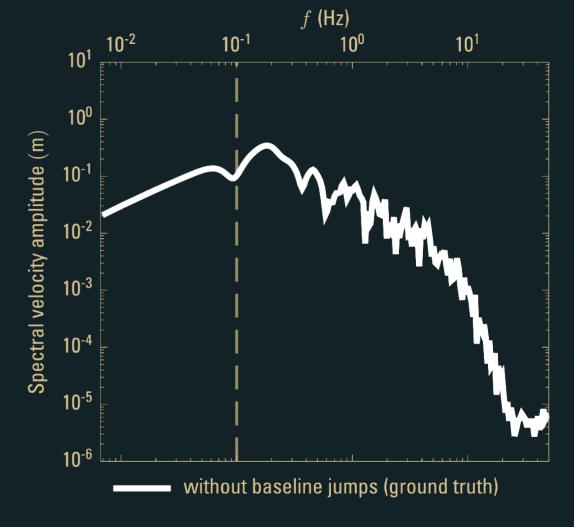


Avoid spectral contamination

jump segments related to corner frequency / signal duration

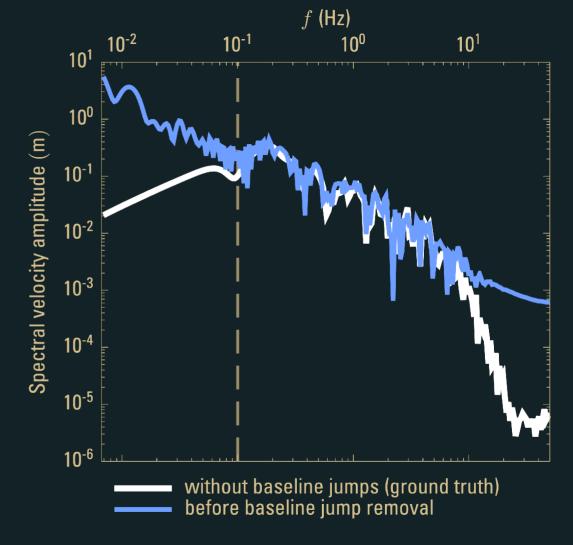
Example:

Synthetic spectrum (after Boore 2003)



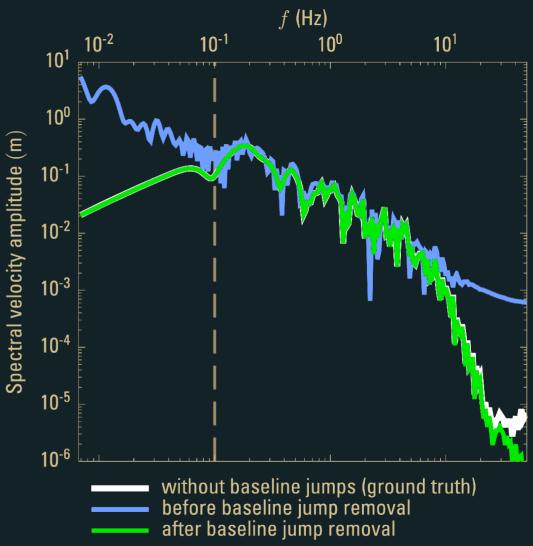
Example:

add baseline jumps (on acceleration)

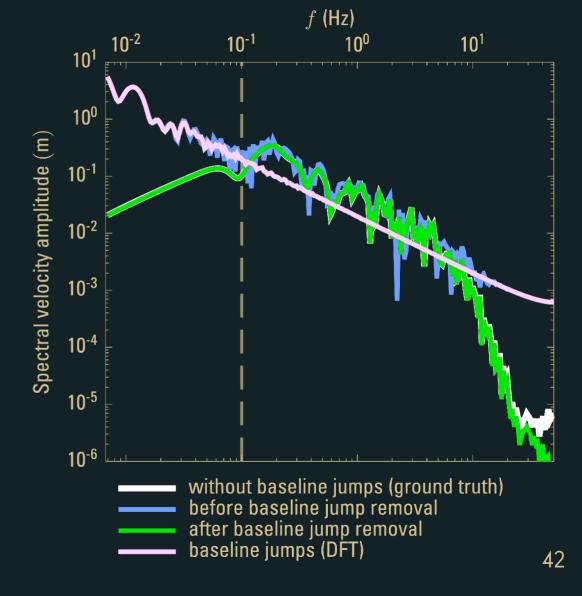


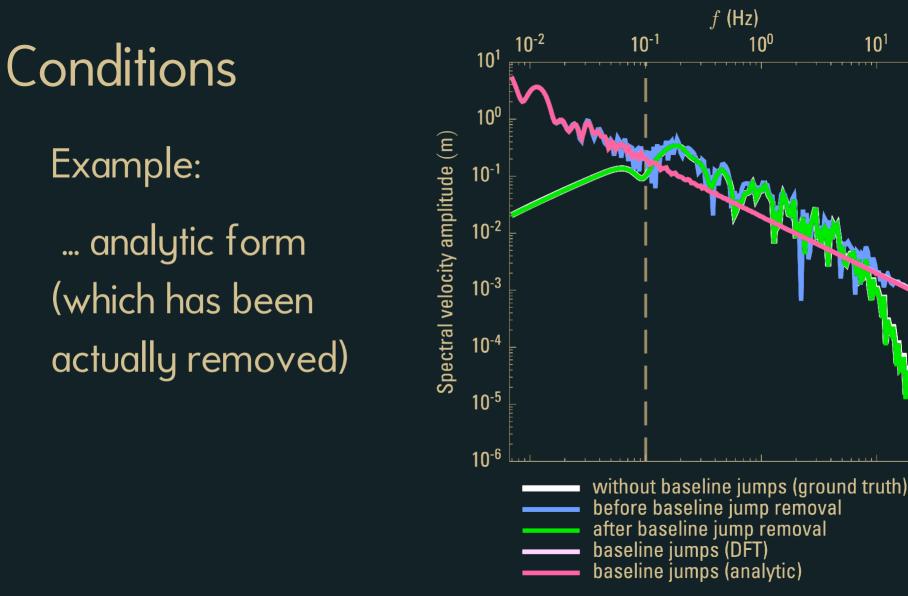
Example:

invert for baseline jumps and remove them



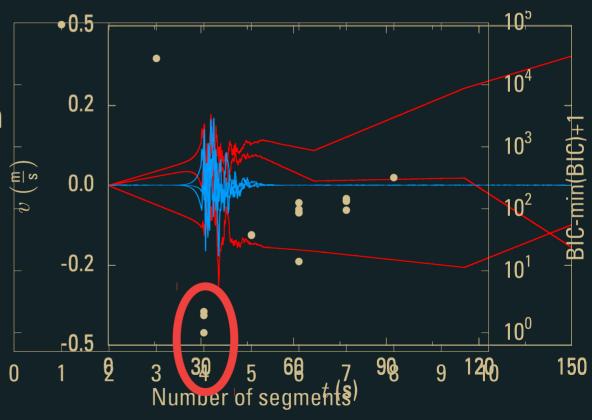
Example: actual baseline jump spectrum versus ...





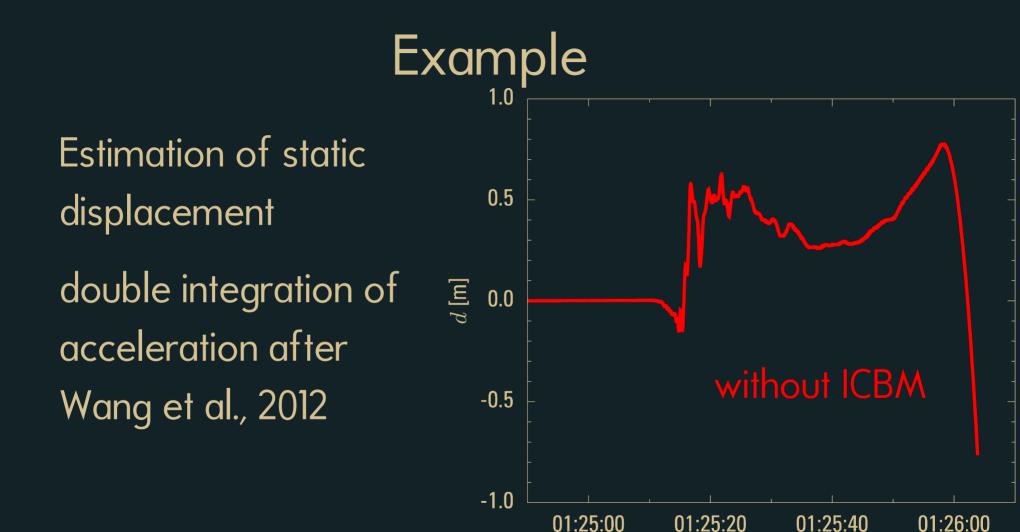
Number of segments

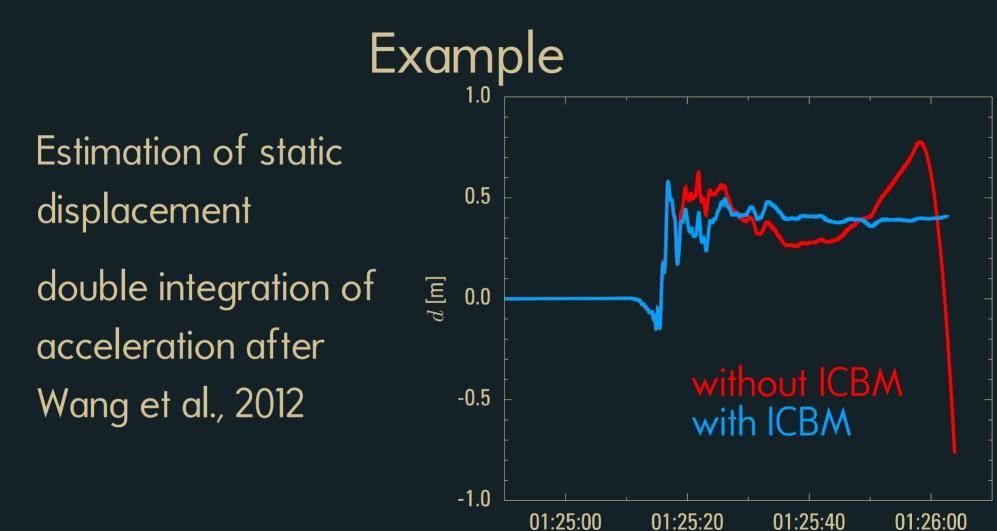
parsimonious selection with Bayesian Information Criterion (BIC)

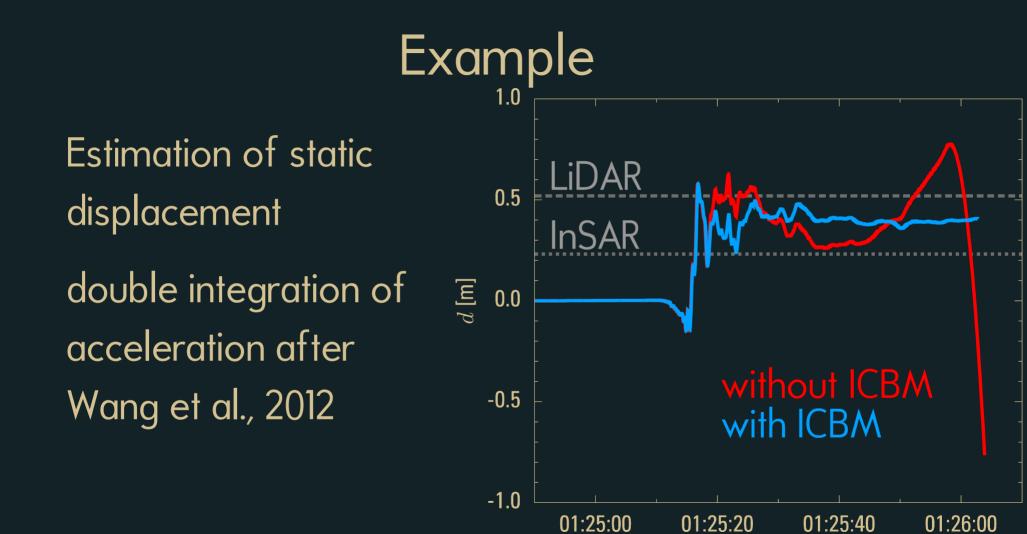


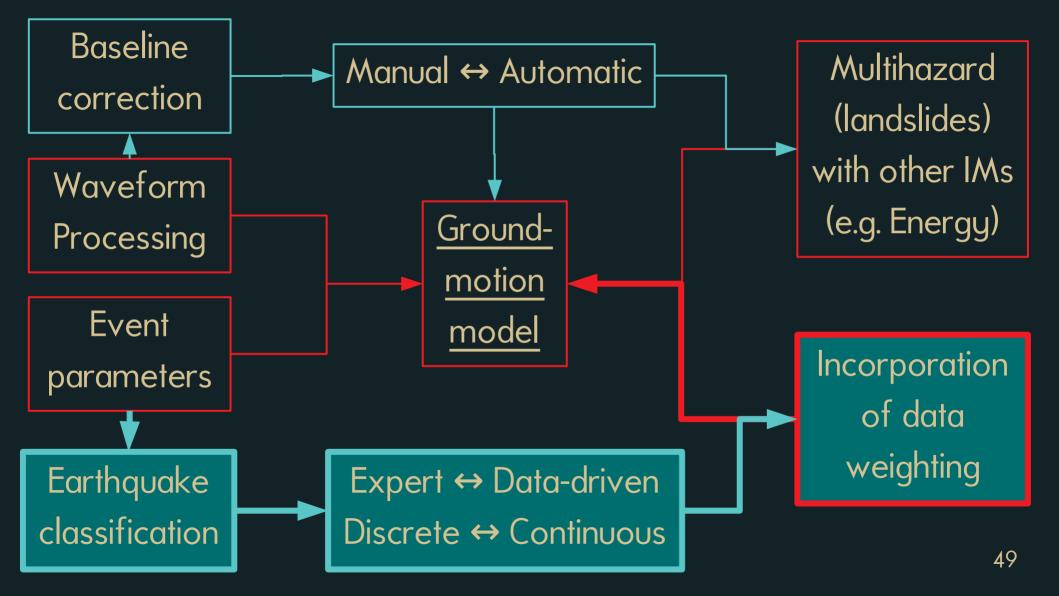
Example

- Estimation of static
- displacement
- double integration of acceleration after Wang et al., 2012









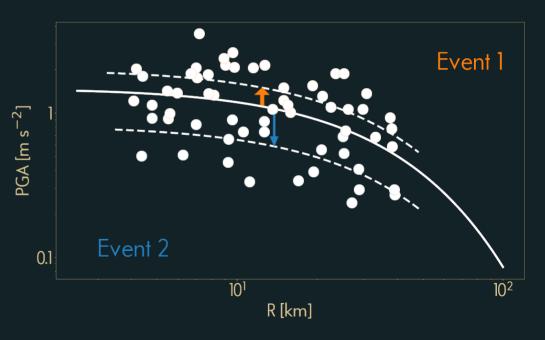
Mixed effect regression

Widely used model

 $\mathbf{y} = \mathbf{A}\mathbf{p} + \mathbf{B}\mathbf{q} + \boldsymbol{\epsilon}$

consists of two parts:

fixed effects model random effects model



Mixed effect regression

Different approaches to handle random effects

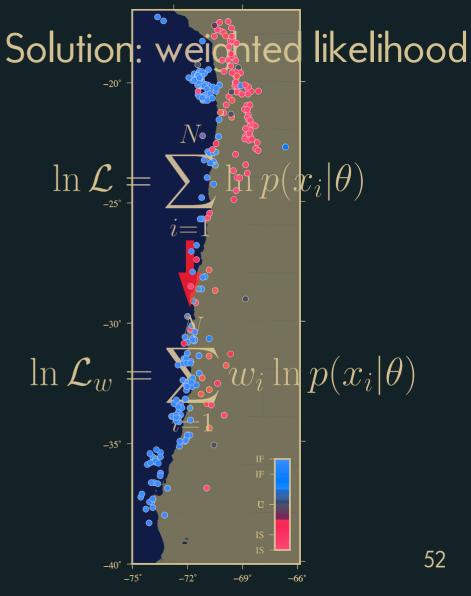
some hybrid models inconsistent with theory possibly biased results harder to reproduce How to incorporate data classification, e.g. SoF?

Discrete:

easy \rightarrow leave out

Continuous:

→ bias in variance estimates



One form to fit them all

Standardized approach:

implementation of any function type

→ meta-model

Example:

 $y = aM_W + br_{rup} - (c + dM_W) \ln r_{rup}$ $+e \begin{cases} z & \text{if } z \leq 125 \text{ km} \\ 125 & \text{otherwise} \end{cases}$ $+\begin{cases} w_i(q_i(M_W - 6.3)^2 + s_i) \\ w_s(q_s(M_W - 6.5)^2 + s_s + s_{sl} \ln r_{rup}) \end{cases}$ $+ x \ln v_{S30}$ from Händel 53 et al. (2015)

One form to combine them

clear work frame for any

random effect structure

$$\hat{\mathbf{p}} = \left(\mathbf{A}^{\mathrm{T}}\mathbf{V}\mathbf{S}^{-1}\mathbf{V}\mathbf{A}
ight)^{-1}\mathbf{A}^{\mathrm{T}}\mathbf{V}\mathbf{S}^{-1}\mathbf{V}\mathbf{y}$$

continuous data classification

and

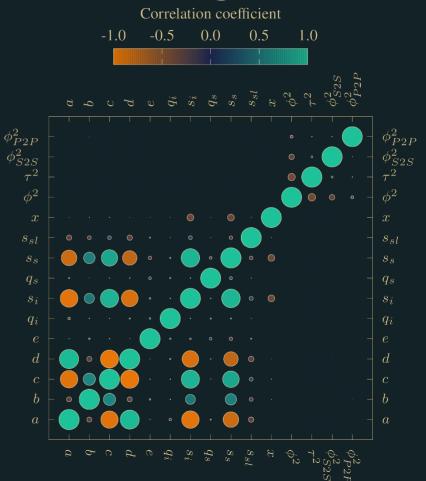
One form to bring them all...

```
... and in the
                                                 \mathbf{C} = \sigma^2 \mathbf{I}
                                                 Within-event covariance
covariance bind them
                                                \mathbf{B} = egin{pmatrix} \mathbf{B}_1 & \mathbf{B}_2 & \cdots & \mathbf{B}_K \end{pmatrix}
                                                     random effects matrix
                                                     (between events, \cdots
                                                     betweerOsiteT, etc.) · · ·
\mathbf{S} = \mathbf{C} + \mathbf{B} \mathbf{D} \mathbf{B}^{\mathrm{T}}
                                                                random effects covariance
                     factor matrix
```

Random effects splitting

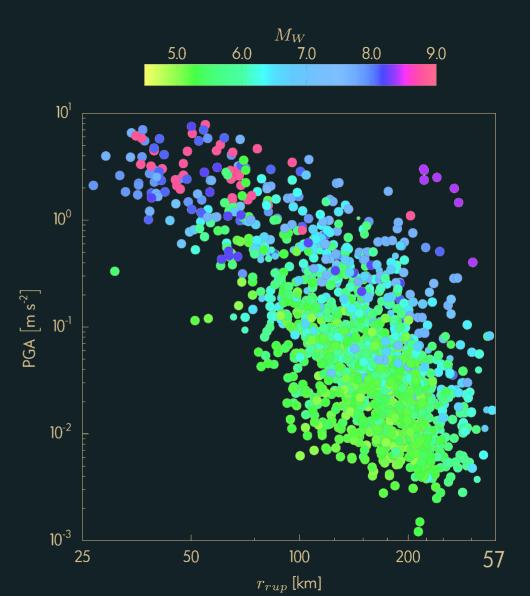
variable residual splitting possible

→ Fisher information shows: random effect variances independent from fixed effects parameters



Strong-motion flatfile (Bastias & Montalva, 2016)

Focal mechanisms for ACE from Global CMT



Functional form:

$$y = aM_W + br_{rup} - (c + dM_W) \ln r_{rup} + e \begin{cases} z & \text{if } z \le 125 \text{ km} \\ 125 & \text{otherwise} \end{cases} + \begin{cases} w_i(q_i(M_W - 6.3)^2 + s_i) \\ w_s(q_s(M_W - 6.5)^2 + s_s + s_{sl} \ln r_{rup}) \\ + x \ln v_{S30} \end{cases} v^2$$

 M_W

6.0

5.0

10¹

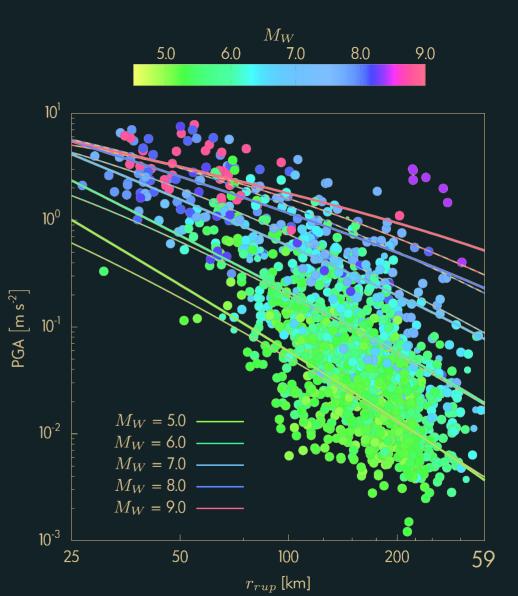
10⁰

7.0

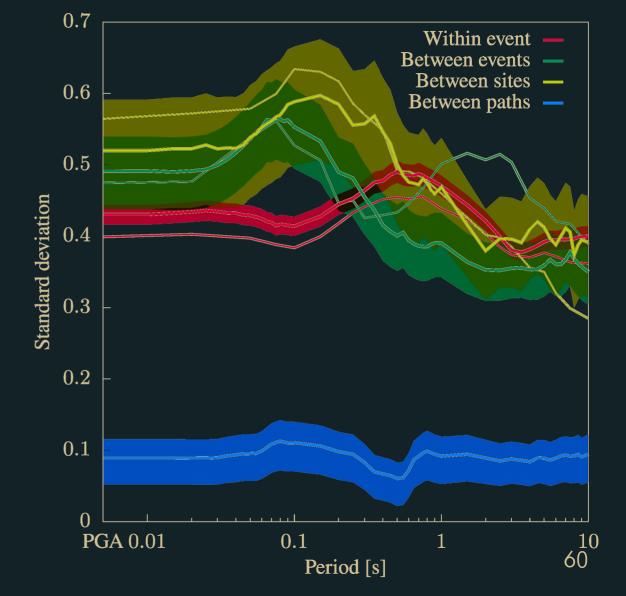
8.0

9.0

Similar performance in fixed effects and ...



... random effect variances



Conclusion & Outlook

- Event classification
 - apply globally

combination with seismicity models (e.g. Bayona et al., 2019) use in stress tensor inversion (von Specht et al., 2018) time dependent parameters

Conclusion & Outlook

Baseline correction

implement in standard processing for strong motion flat files (e.g. NGAwest)

preprocessing step for radiated seismic energy estimation (e.g. von Specht et al., 2019)

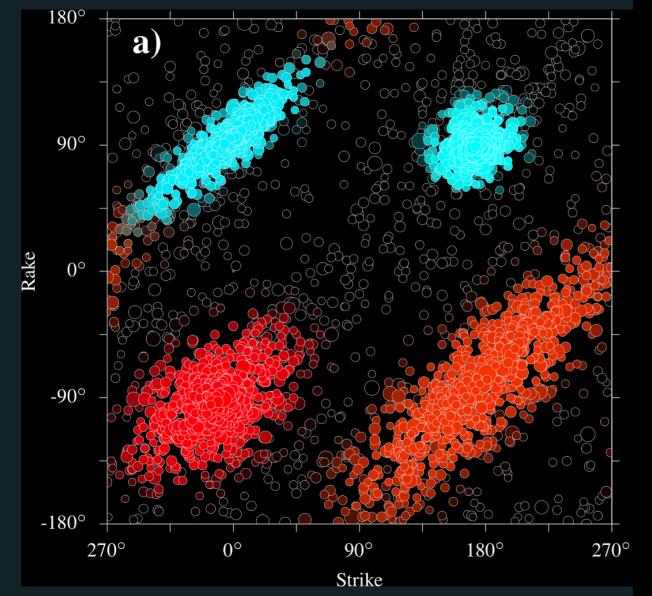
Conclusion & Outlook

- Ground motion model
 - development of purely data-driven ground motion models (from database to final model)
 - random effect terms as function of magnitude, distance etc.
 - hazard cascade specific models (seismically triggered landslides)

Thank you!

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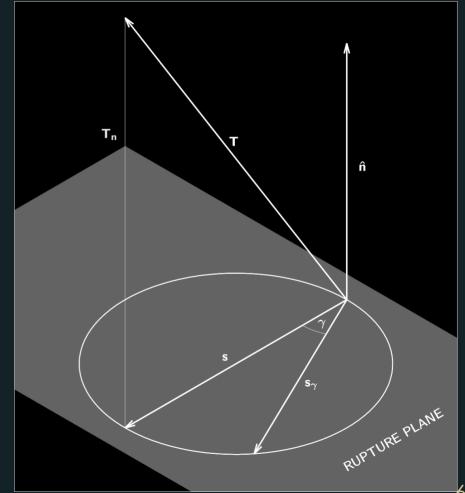
地震の桜

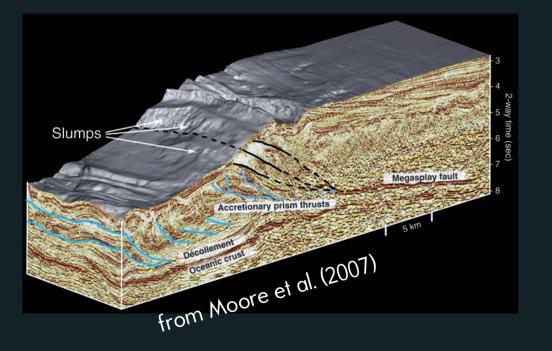


All models are wrong – some are useful

Fault slip oriented in maximum resolved shear stress (Wallace-Bott hypothesis)

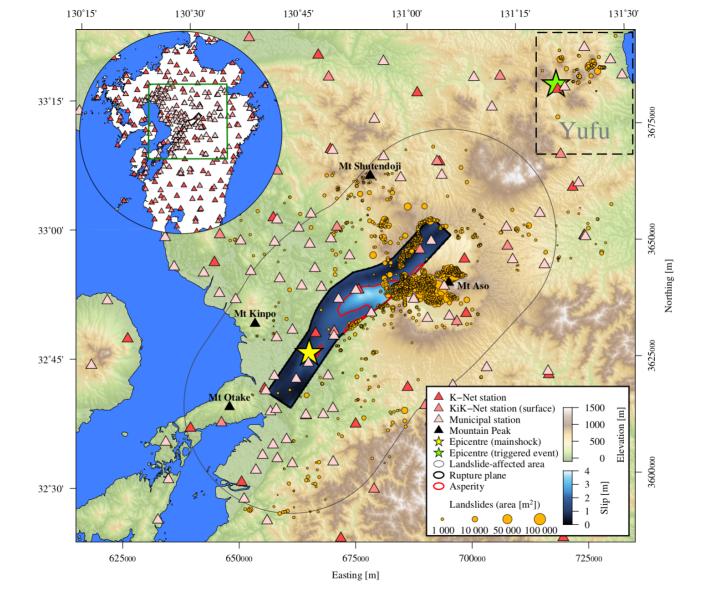
→ large deviation possible but less likely

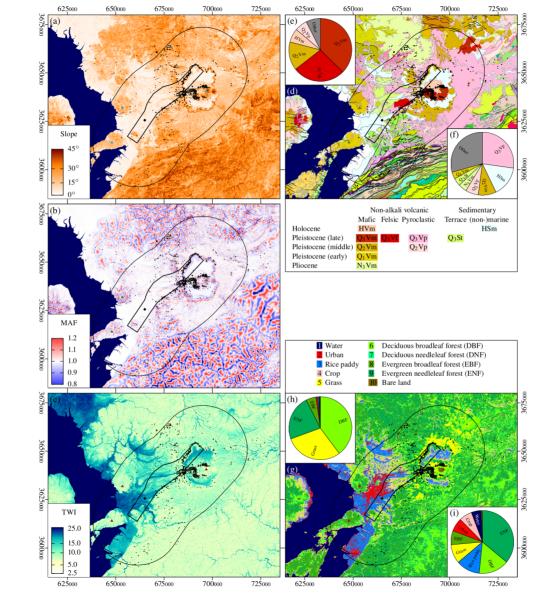


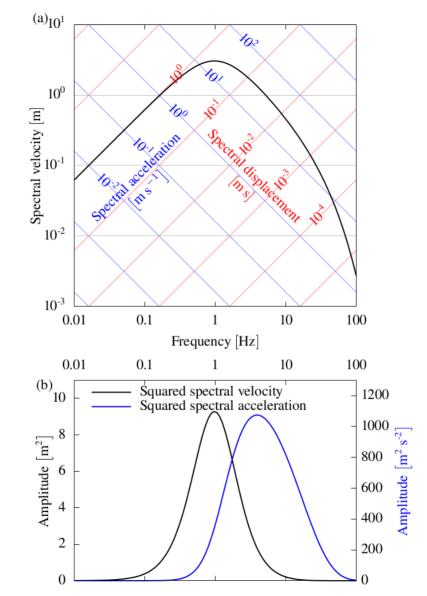


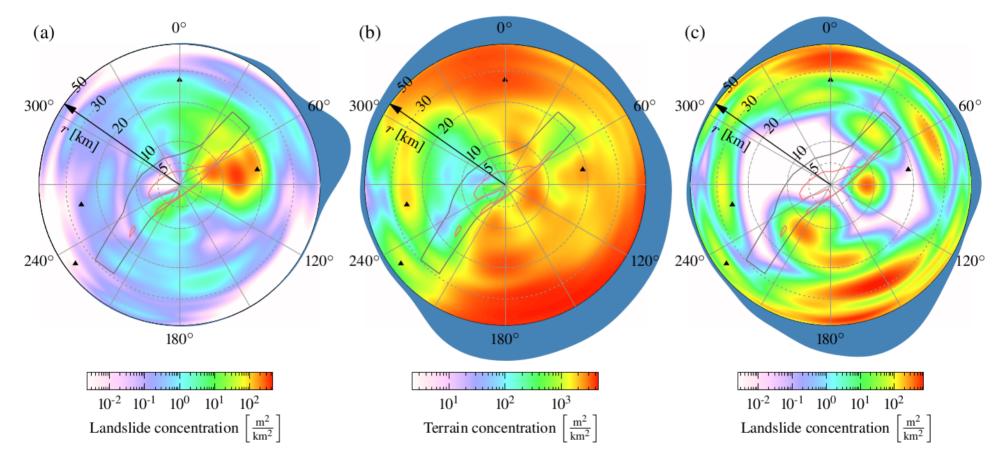
Thrust but not interface: splay faults in accretionary prism dip steeper than subduction interface

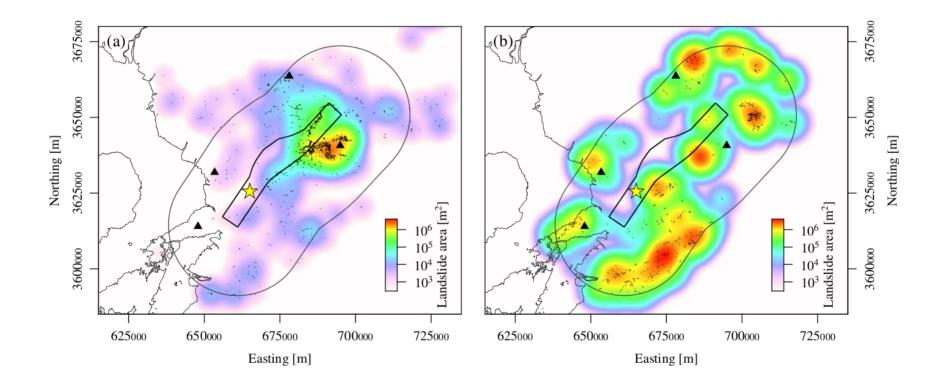
 \rightarrow less seismicity

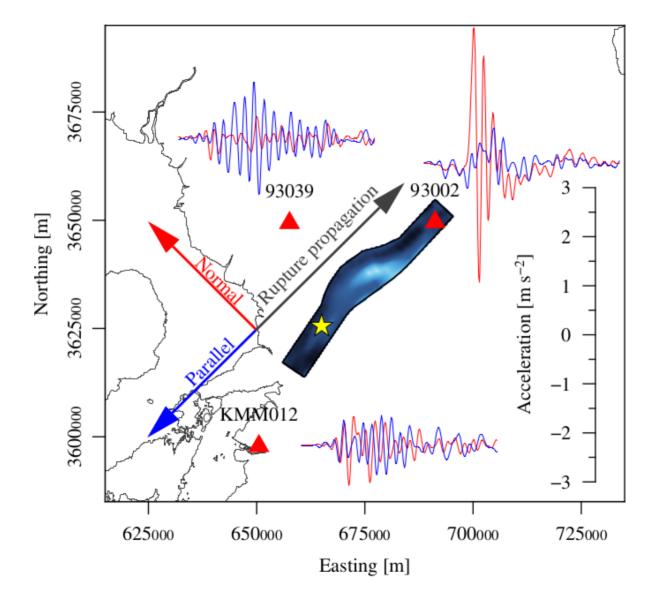


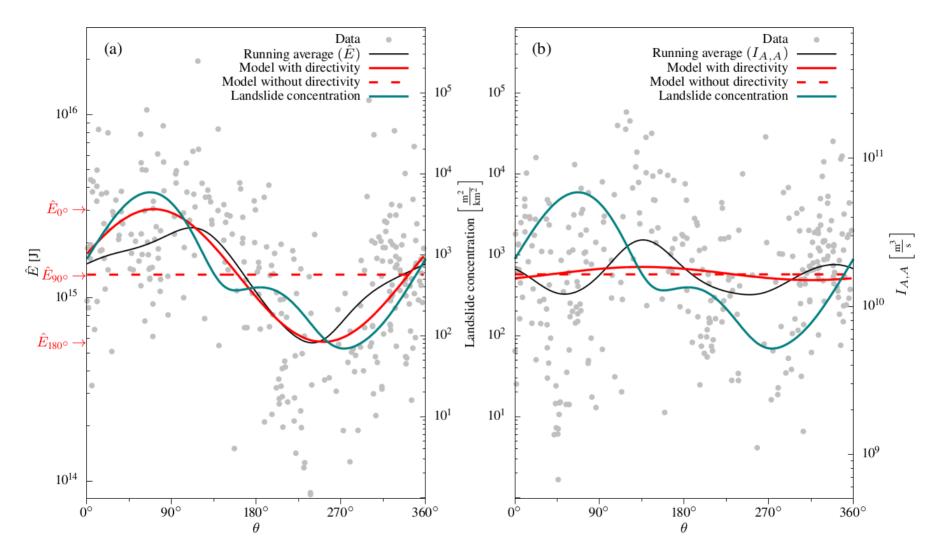


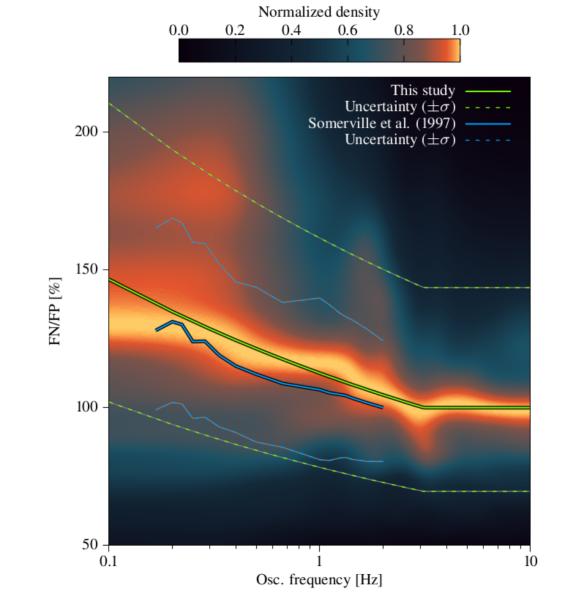




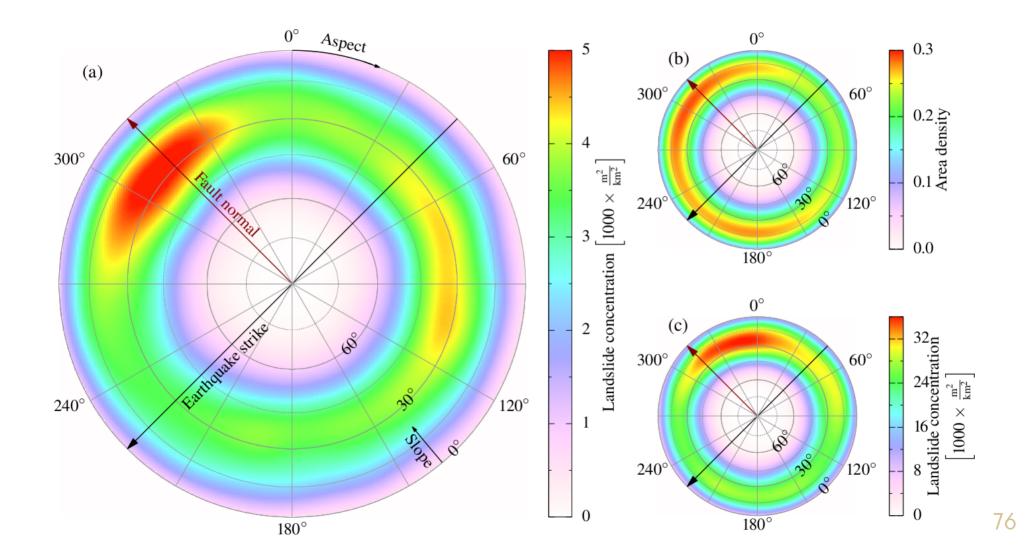


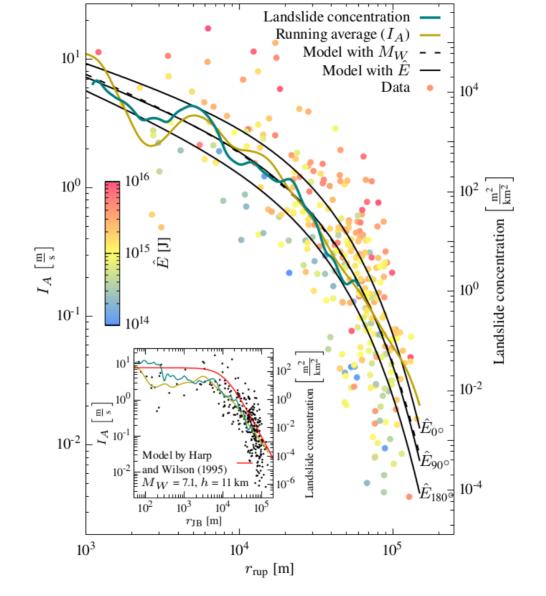


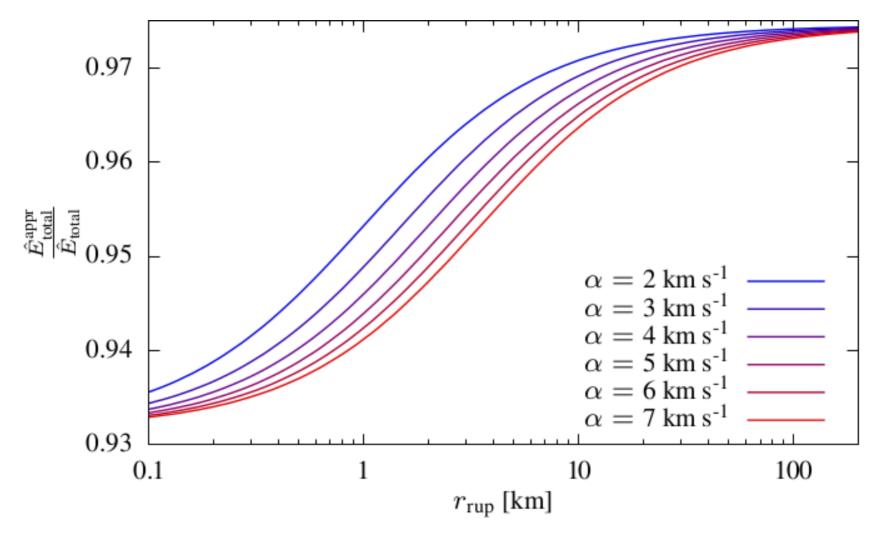


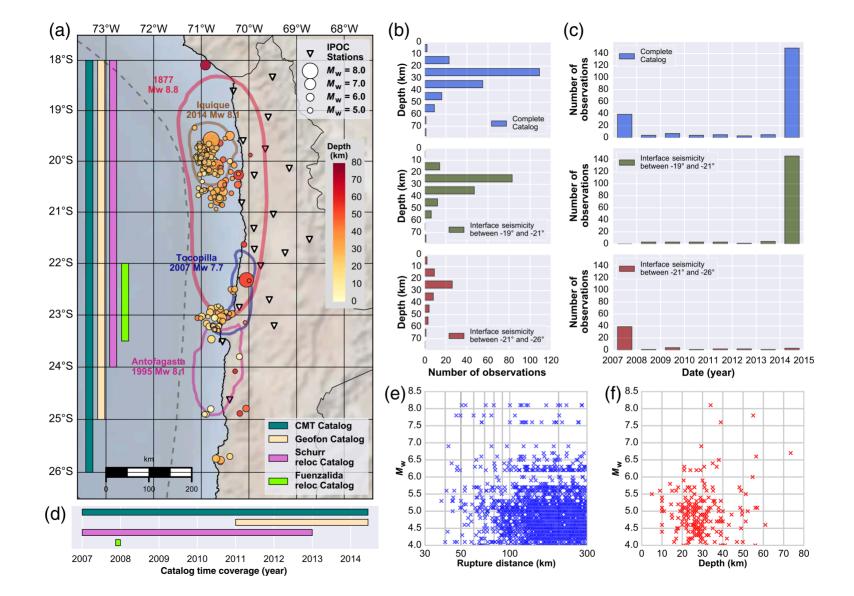


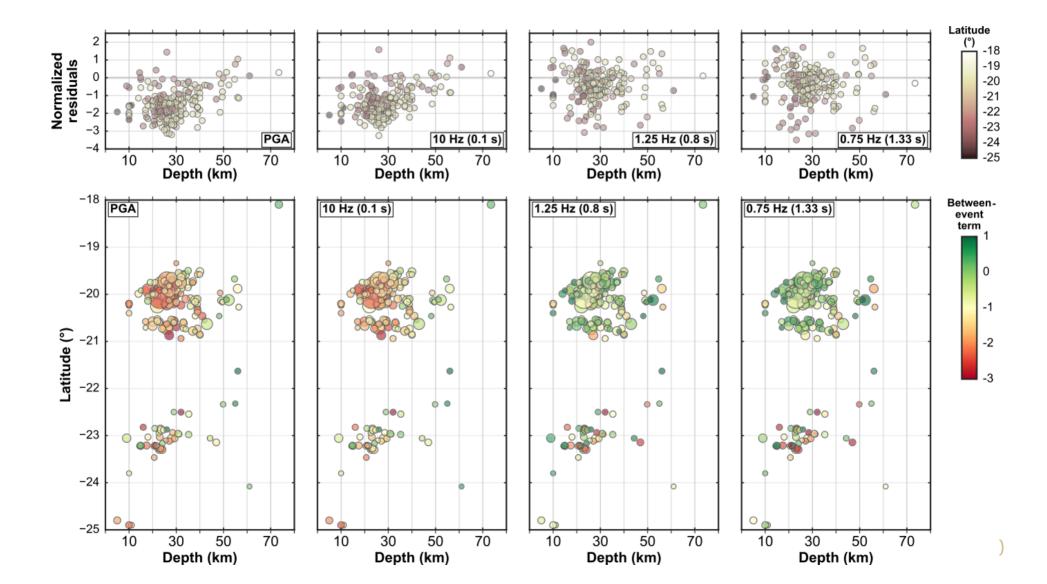


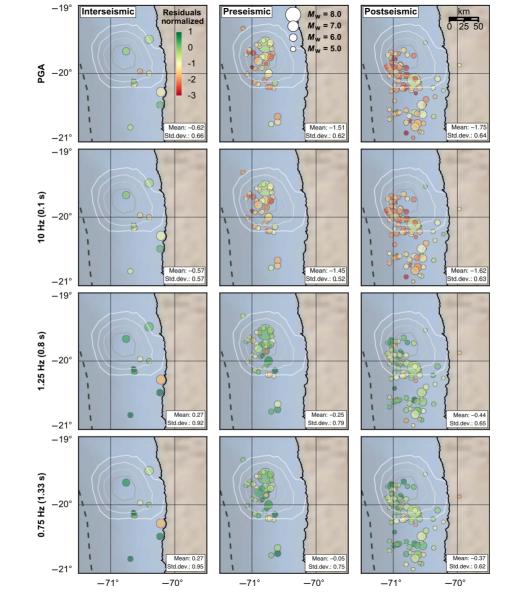


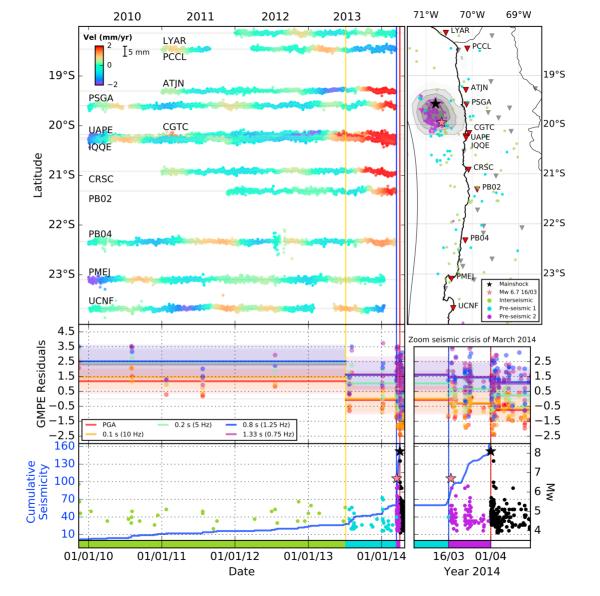


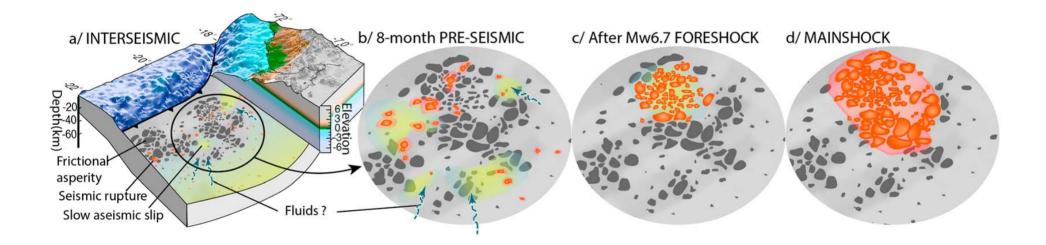


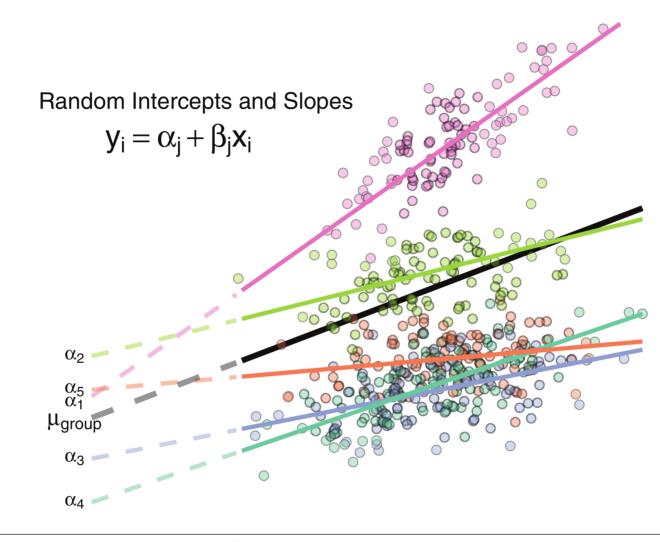












Dependent Variable y

Predictor Variable x

