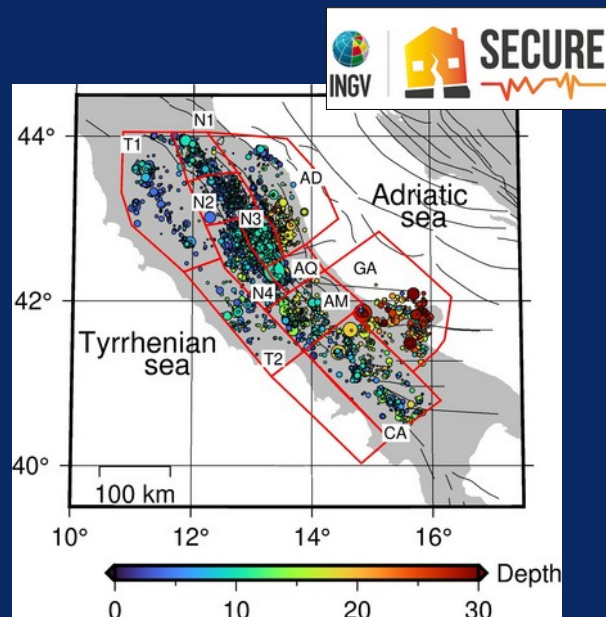




Bindi et al., 2024 (BSSA)

Bridging Spectral Decomposition and Ground-Motion Models: Results from **SECURE** and Perspectives

Dino Bindi & WP2, WP3

GFZ-Potsdam

Rome, 12 March 2026

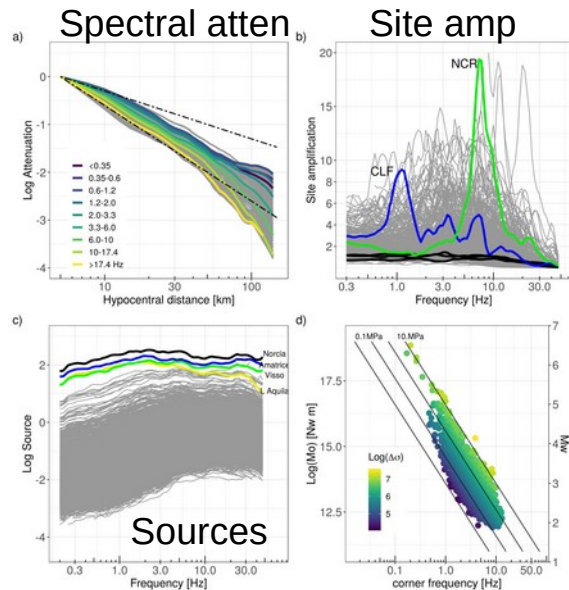


regional-Scale Earthquake ground motion prediCtions throUgh pyhsics-based and empiRical approachEs: a case study Central Italy

- **WP1** Enhancements of Existing Crustal Velocity Models and Seismic Datasets
- **WP2** Characterization of Earthquake Source and Seismic Wave Propagation in the Central Apennines
- **WP3** Ground Motion Predictions using Innovative Approaches
- **WP4** Testing the Performance of Ground Motion Models and Empirical Relationships

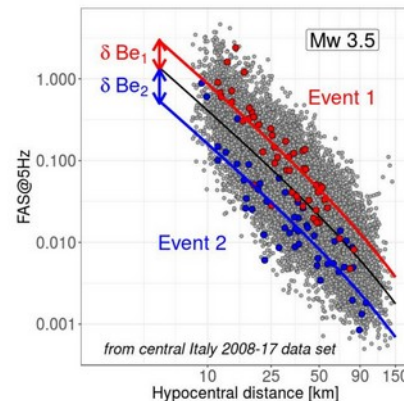
In the context of SECURE

WP2 – Physical ingredients

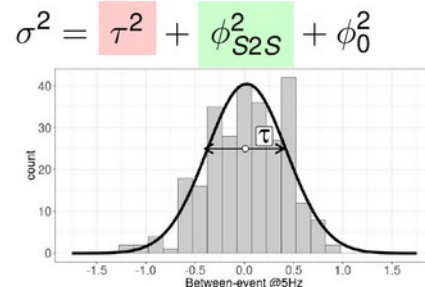


WP3 – Ground motion models

$$\text{Log}[FAS(f)] = \text{Scaling}(R) + \text{Scaling}(M) + \delta S2S + \delta Be + \epsilon$$



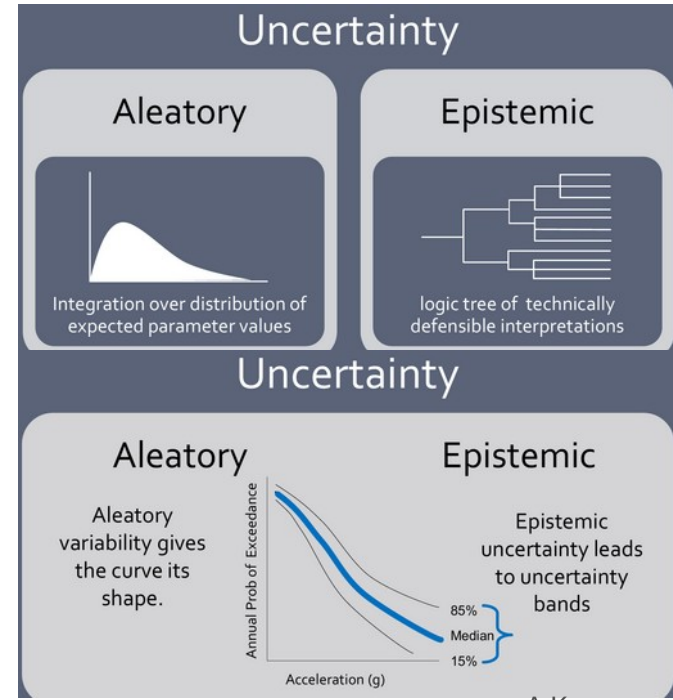
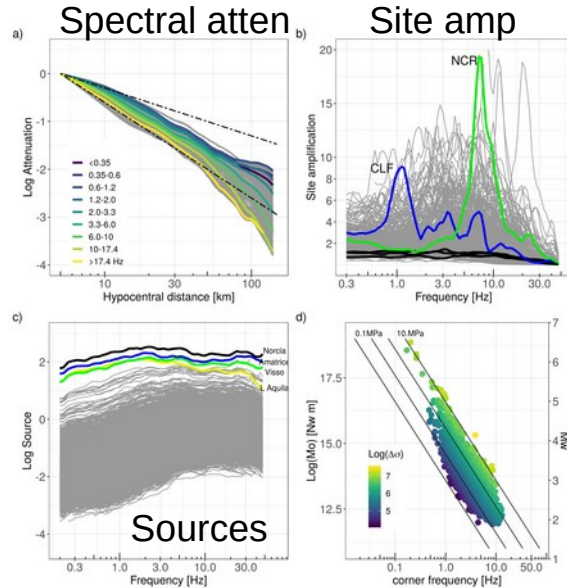
Station-term δBe_1 δBe_2 Event-term



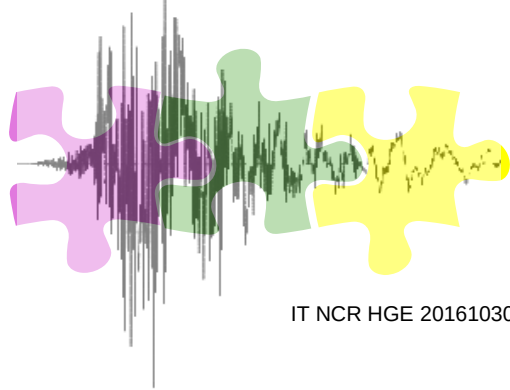
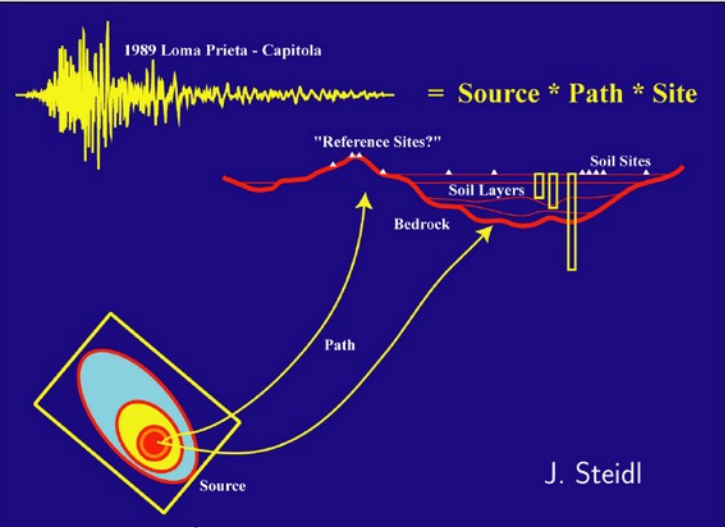
Why we need to decompose

WP2 – Physical ingredients

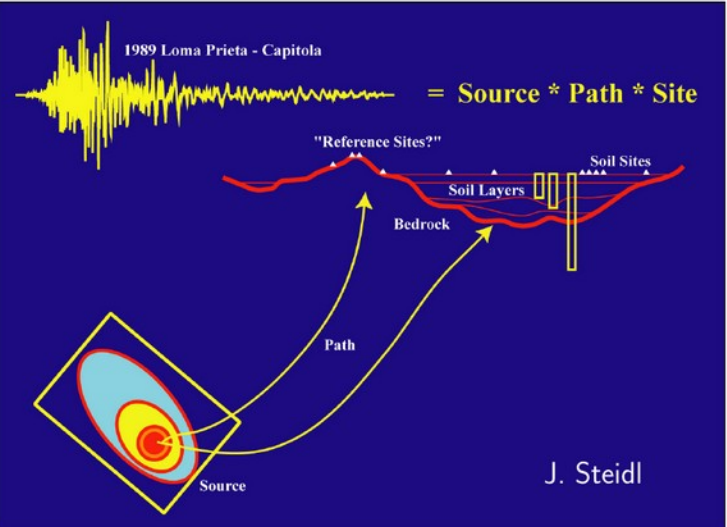
$$\text{Log}[FAS(f)] = \text{Scaling}(R) + \text{Scaling}(M) + \delta S2S + \delta Be + \epsilon$$



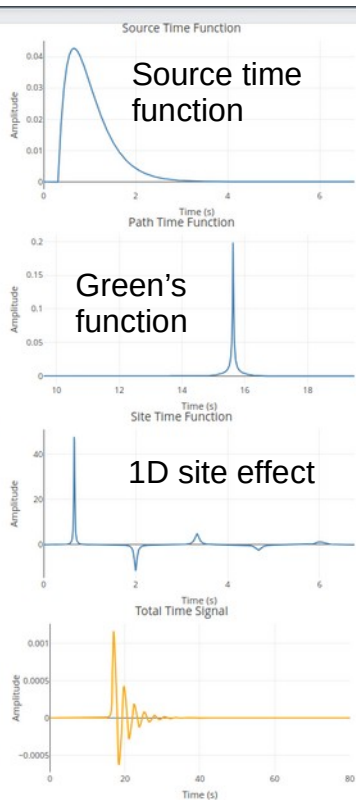
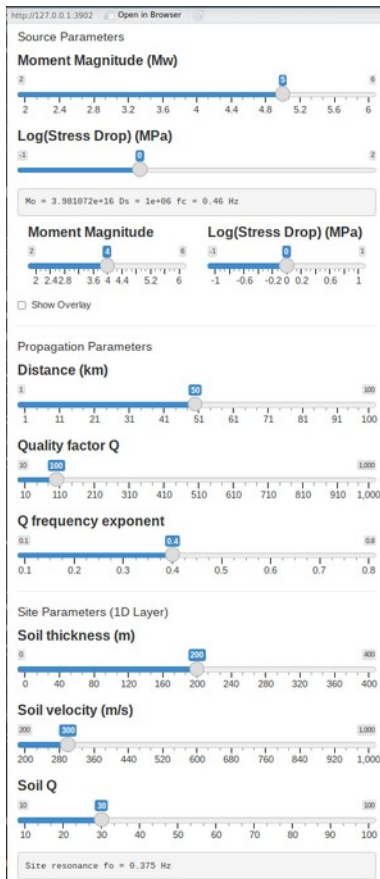
A.Kammerer



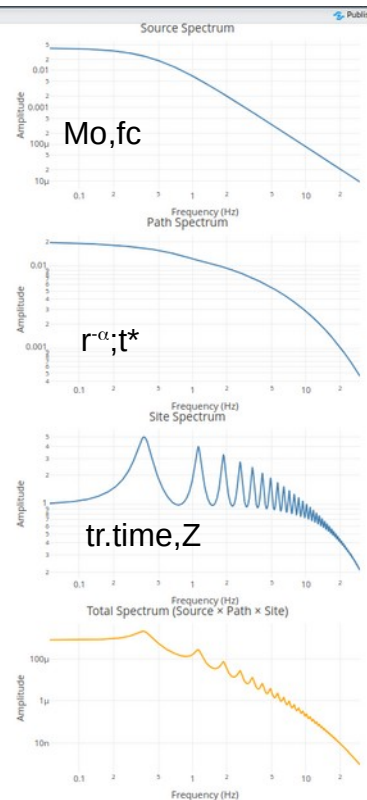
IT NCR HGE 20161030 esm-db.eu



IT NCR HGE 20161030 esm-db.eu

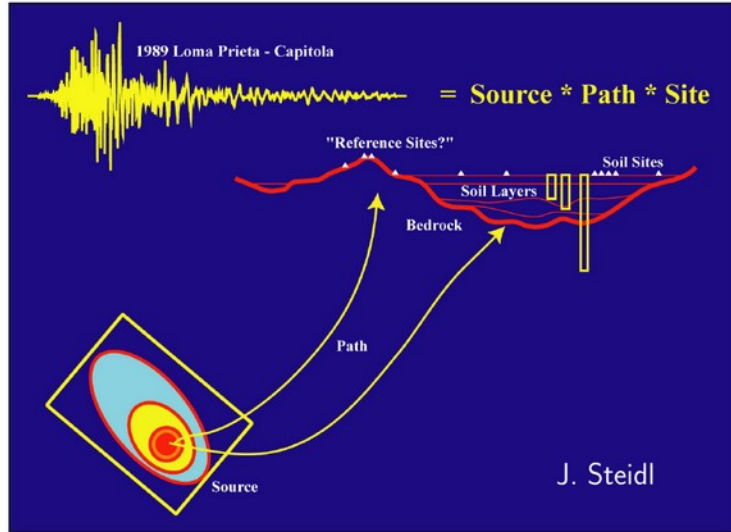


Convolution in time domain



Convolution in spectral domain

Spectral decomposition GIT & SECURE WP2



$$FAS(f) = S(f) P(f) G(f)$$

Source Site
Propagation

$$\log_{10} FAS_{ij}(f) = \sum_{k=1}^{Nev} \delta_{ik} \log_{10} S_k(f) +$$

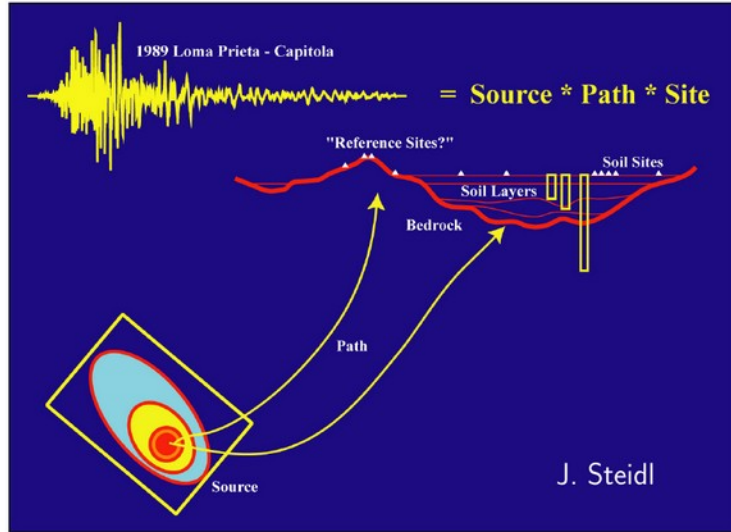
$$+ a_n \log_{10} P_n(f) + a_{n+1} \log_{10} P_{n+1}(f) +$$

$$+ \sum_{k=1}^{Nst} \delta_{jk} \log_{10} Z_k(f)$$

where $a_n = (R_{n+1} - R_{ij}) / (R_{n+1} - R_n)$

$$a_{n+1} = 1 - a_n \quad R_n \leq R_{ij} < R_{n+1}$$

Spectral decomposition GIT & SECURE WP2



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Source Site
Propagation

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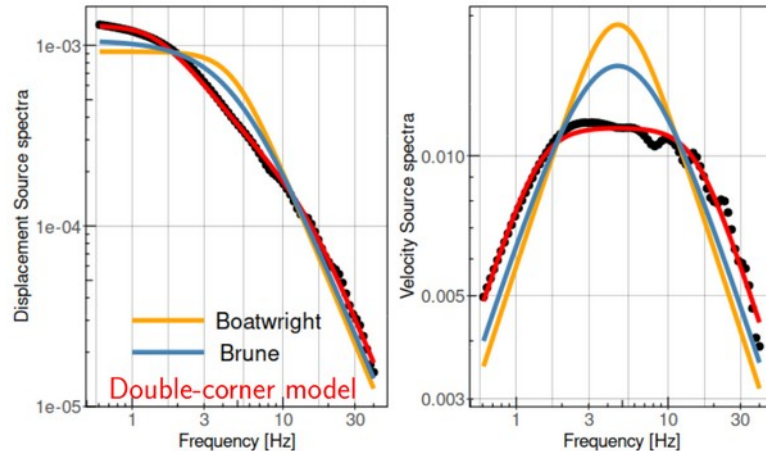
15:10 WP2 Calderoni G. *Multi-Method Stress Drop Estimation and Uncertainty Analysis for the 2016–2017 Amatrice–Visso–Norcia Seismic Sequence.*

Spectral decomposition GIT & SECURE

$$S_k(f) = [s_1, s_2, \dots, s_{N_{fre}}]$$

$$FAS(f) = S(f) P(f) G(f)$$

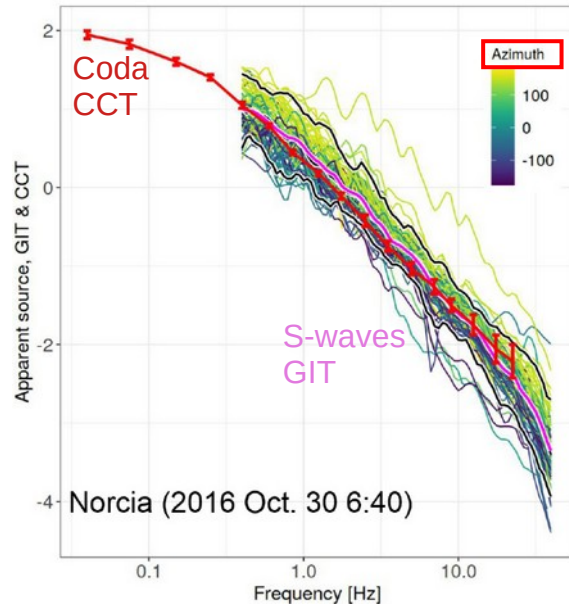
Source Site
Propagation



$$\begin{aligned} \log_{10} FAS_{ij}(f) = & \sum_{k=1}^{N_{ev}} \delta_{ik} \log_{10} S_k(f) + \\ & + a_n \log_{10} P_n(f) + a_{n+1} \log_{10} P_{n+1}(f) + \\ & + \sum_{k=1}^{N_{st}} \delta_{jk} \log_{10} Z_k(f) \end{aligned}$$

Spectral decomposition GIT & SECURE

$$S_{kj} = \{ [s_{1j}, \dots, s_{N_{\text{fre}},j}]_{j=1, N_{\text{sta}}} \}$$



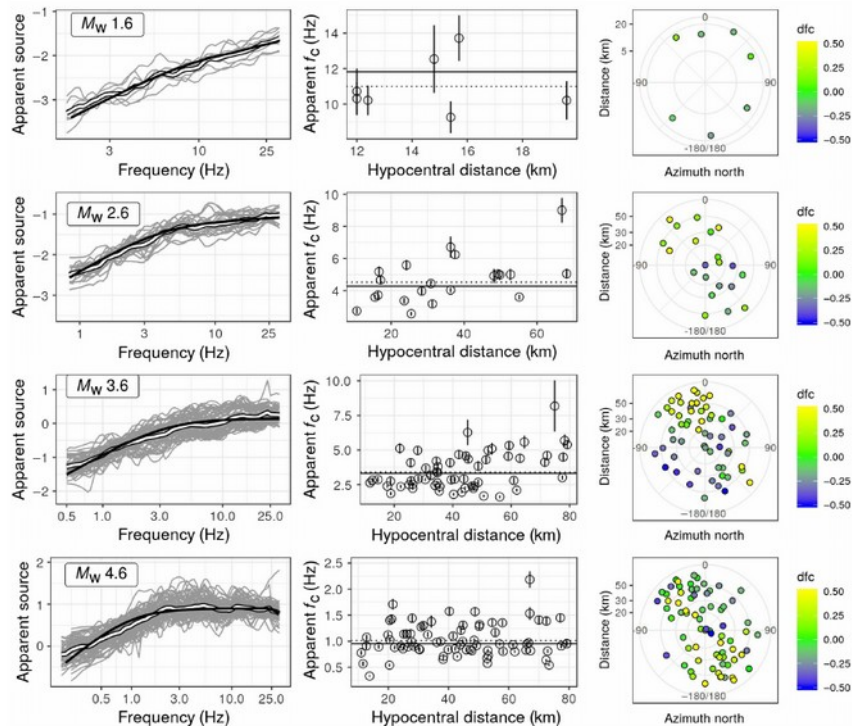
$$FAS(f) = S(f) P(f) G(f)$$

Source
Propagation
Site

$$\begin{aligned} \log_{10} FAS_{ij}(f) = & \sum_{k=1}^{N_{\text{ev}}} \delta_{ik} \log_{10} S_k(f) + \\ & + a_n \log_{10} P_n(f) + a_{n+1} \log_{10} P_{n+1}(f) + \\ & + \sum_{k=1}^{N_{\text{st}}} \delta_{jk} \log_{10} Z_k(f) \end{aligned}$$

Morasca et al. 2022 GJI

Spectral decomposition GIT & SECURE



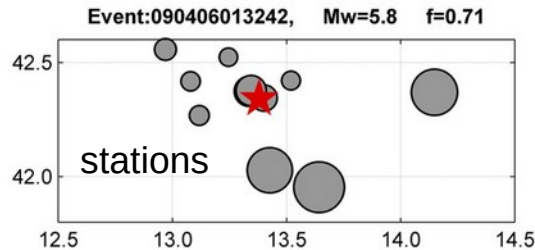
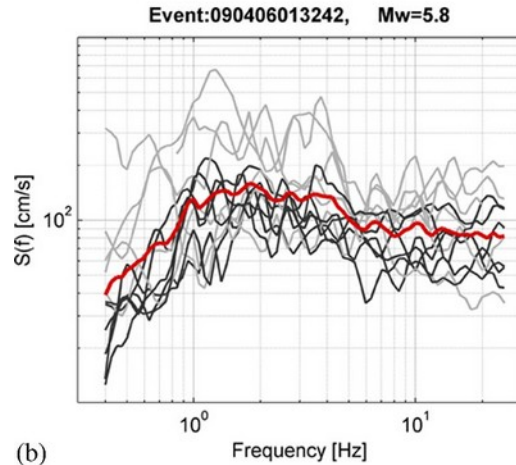
$$FAS(f) = S(f) P(f) G(f)$$

Source Site
Propagation

$$\begin{aligned} \log_{10} FAS_{ij}(f) = & \sum_{k=1}^{N_{ev}} \delta_{ik} \log_{10} S_k(f) + \\ & + a_n \log_{10} P_n(f) + a_{n+1} \log_{10} P_{n+1}(f) + \\ & + \sum_{k=1}^{N_{st}} \delta_{jk} \log_{10} Z_k(f) \end{aligned}$$

Bindi et al. 2020, BSSA

Spectral decomposition GIT & SECURE



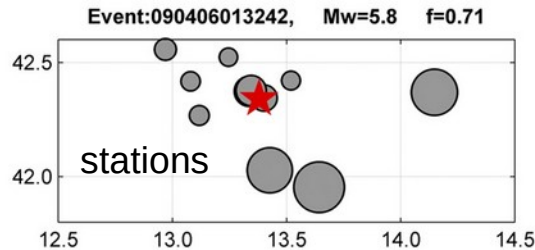
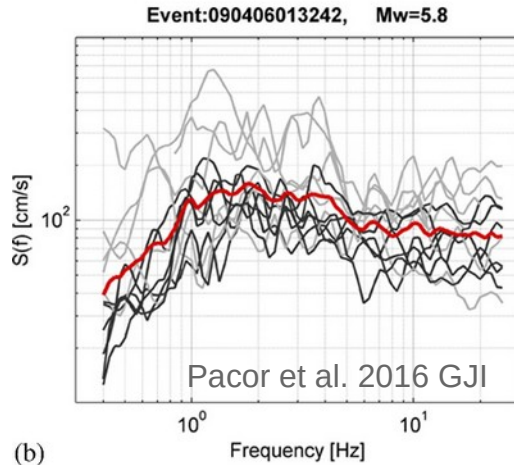
$$FAS(f) = S(f) P(f) G(f)$$

Source
Propagation Site

$$\begin{aligned} \log_{10} FAS_{ij}(f) = & \sum_{k=1}^{N_{ev}} \delta_{ik} \log_{10} S_k(f) + \\ & + a_n \log_{10} P_n(f) + a_{n+1} \log_{10} P_{n+1}(f) + \\ & + \sum_{k=1}^{N_{st}} \delta_{jk} \log_{10} Z_k(f) \end{aligned}$$

Pacor et al. 2016 GJI

Spectral decomposition GIT & SECURE



$$FAS(f) = S(f) P(f) G(f)$$

Source Site
Propagation

$$\log_{10} FAS_{ij}(f) = \sum_{k=1}^{N_{ev}} \delta_{ik} \log_{10} S_k(f) +$$

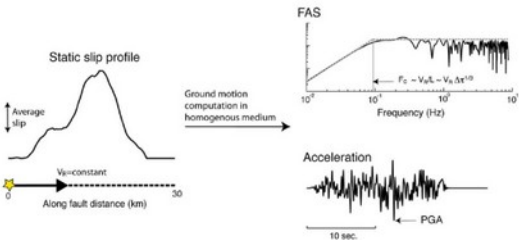
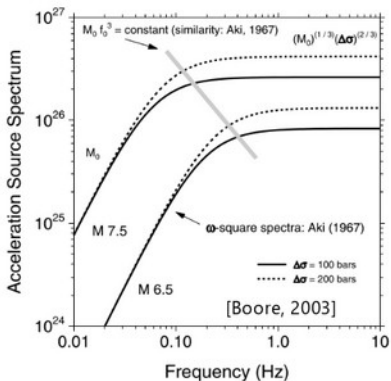
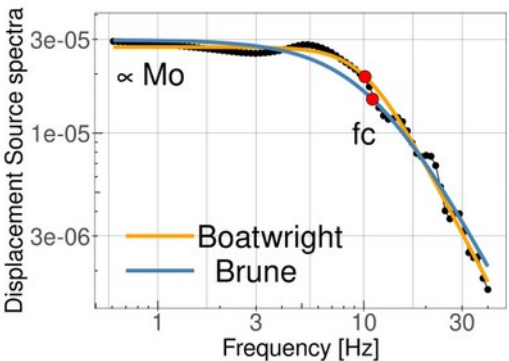
$$+ a_n \log_{10} P_n(f) + a_{n+1} \log_{10} P_{n+1}(f) +$$

15:25 WP2 Cirella, A., *Linking Stress Drop and Slip Heterogeneity to Assess Source Rupture Directivity for Earthquakes in Central Italy.*

15:55 WP2 Colavitti, L. *Data-Driven Empirical Modeling of Source Directivity Effects in a Ground Motion Model for Central Italy*

10:30 – Keynote Lecture, Gallovič, F., (Charles University in Prague, Czech Republic), *Constraining kinematic and dynamic rupture models for broadband strong-motion simulations.*

GMMs are based on self-similar source scaling



[Causse & Song, 2015]



Models

$$FAS_D(f) = \Gamma \frac{M_o}{\left[1 + \left(\frac{f}{f_c}\right)^{\gamma n}\right]^{\frac{1}{\gamma}}}$$

n=2

γ=1 Brune [1970]

γ=2 Boatwright [1980]



Definition of stress drop Δσ

$$\Delta \sigma = \frac{7}{16} \left(\frac{\bar{f}_c}{k \beta} \right)^3 M_o$$

- Circular crack model with uniform Δσ

[Eshelby, 1957]

- κ depends on the rupture model



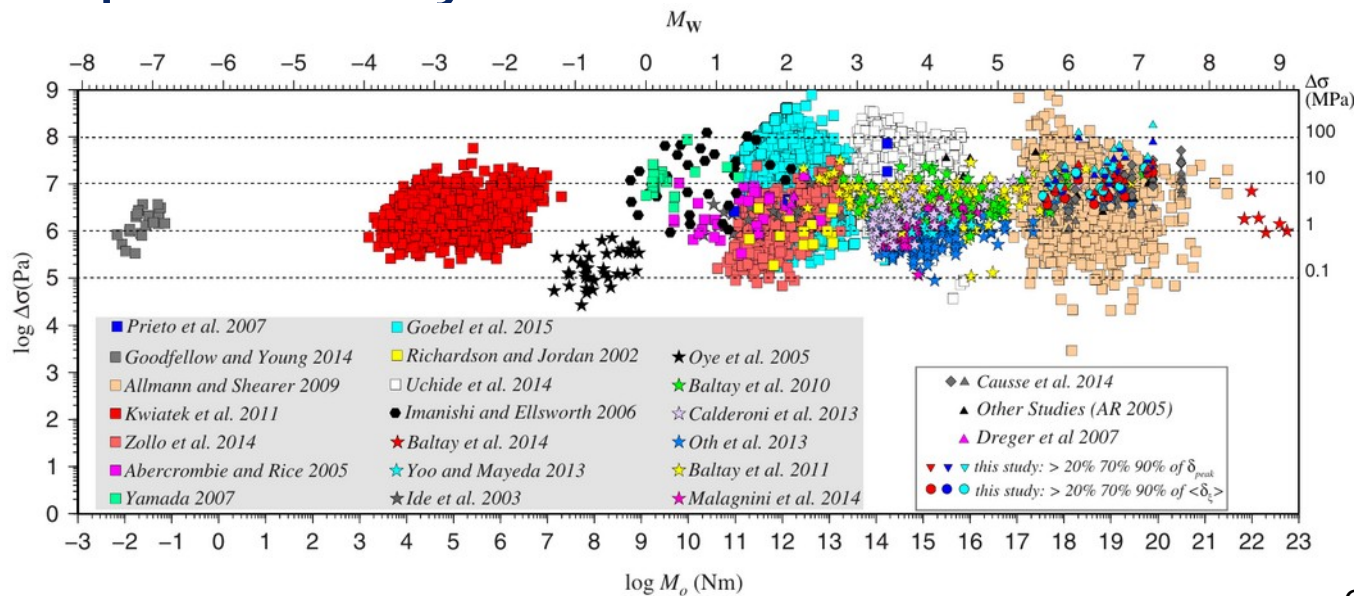
Stochastic simulations

$$\log(PGA) \sim 0.31M + 0.8 \log(\Delta \sigma)$$

[Hanks & McGuire, 1981; Boore, 2003]

Stress drop variability

Aggregation bias;
Between-groups confounding



Cocco et al. 2016

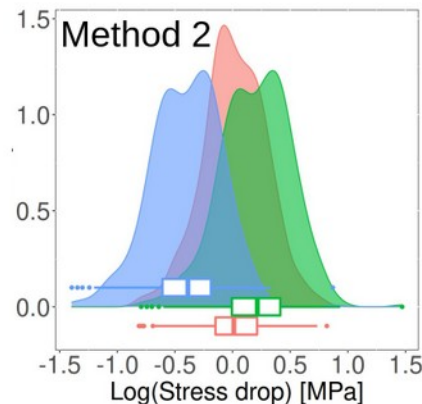
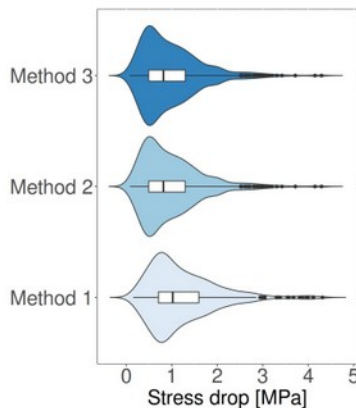
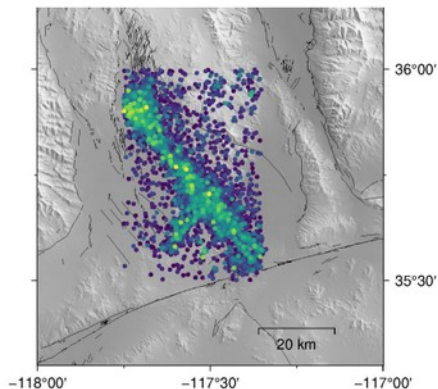
Pooled data set of results
(between-studies)

- Large $\Delta\sigma$ variability (~ 3 orders)
- No clear trend

Individual data set
(within-study)

- Large $\Delta\sigma$ variability (> 2 orders)
- Some data sets show a trend

Stress drop variability SCEC stress drop benchmark [Ridgecrest]



↑ Data set: Ridgecrest 2019
From Trugman, 2020

$N_{ev} = 556$
 $N_{st}=67$
 $N_{rec}=$
 $M \text{ in } [2.5, 7.1]$
 $R: 5-112 \text{ km}$

↑ Between-Methods variability

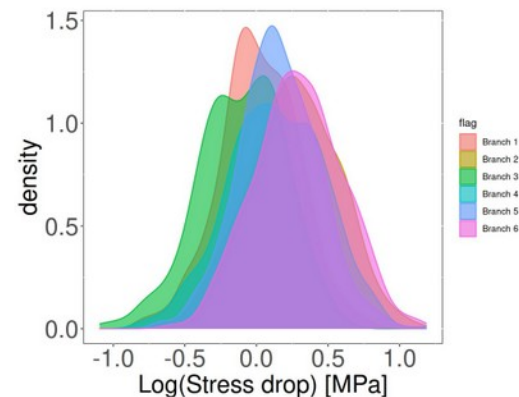
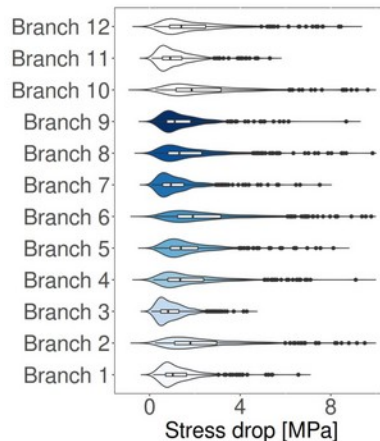
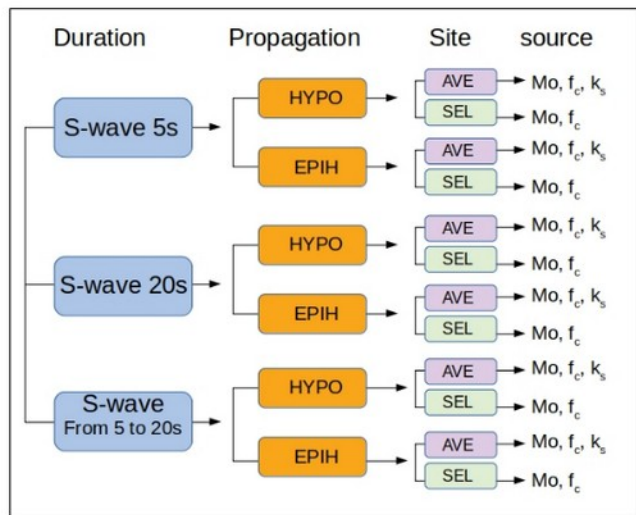
The participants applied different methods such as spectral decomposition, Spectral ratios, coda-analysis, time deconvolution, finite-fault kinematic approaches, etc

↑ Within-method variability

Considered method: GIT

Investigated alternative choices for different model components

Stress drop variability SCEC stress drop benchmark [Ridgecrest]

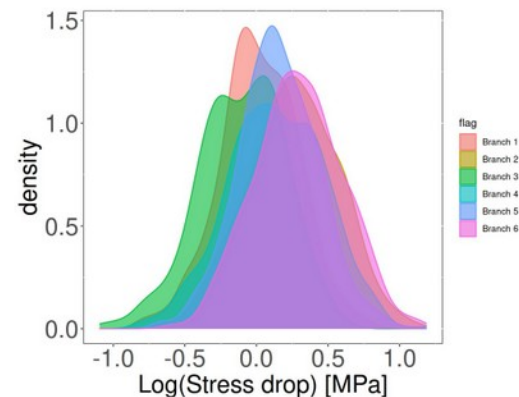
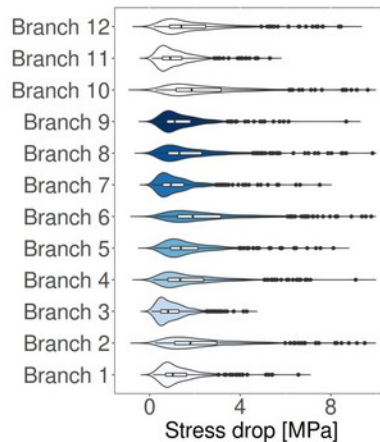
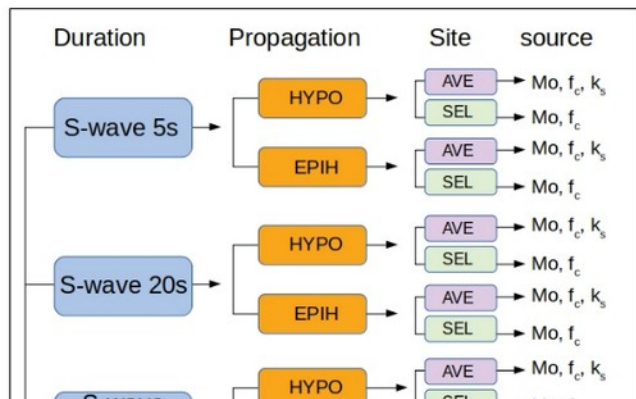


Statistics for the whole population

```

> quantile((df$drop_vs_3D), probs = c(0.025, 0.5, 0.975))
      2.5%      50%      97.5%
0.2806694 1.2339084 5.1477031 MPa
> quantile(log10(df$drop_vs_3D), probs = c(0.025, 0.5, 0.975))
      2.5%      50%      97.5%
-0.55180653 0.09128291 0.71161279
  
```

Stress drop variability SCEC stress drop benchmark [Ridgecrest]



```
> sort(result$p50)
      50%      50%      50%      50%      50%      50%      50%      50%      50%      50%
-0.08681238 -0.04017378 -0.02551702  0.01075241  0.04658297  0.11905079  0.13125700  0.14099140  0.14693779  0.25583089
      50%      50%
  0.26902009  0.28135963
      log10Δσ
> 0.28+0.087
[1] 0.367      (2.3)
```

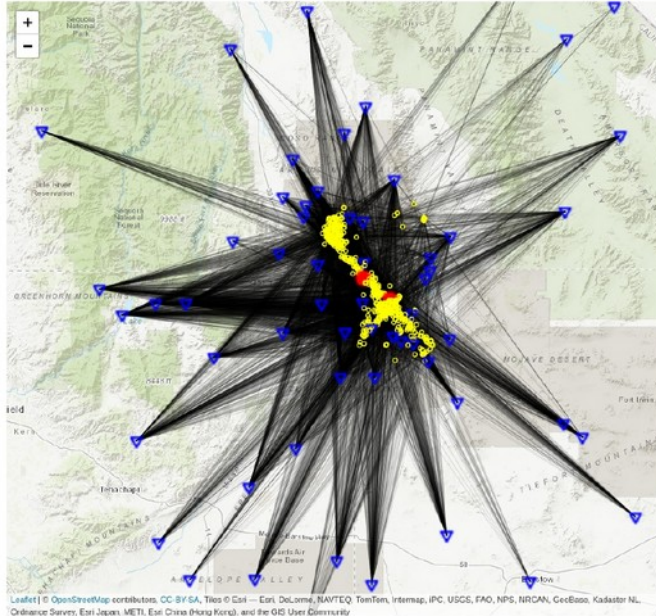
Our companions along the GIT journey

$$FAS(f) = S(f) P(f) G(f)$$

- **Non-Uniqueness of the solution**
- Attenuation-source interplay
 - Regional attenuation: polygons, cell-grid
 - Depth dependency

Spectral decomposition GIT – non uniqueness

Ridgecrest M7.1, 2016



Singular value decomposition

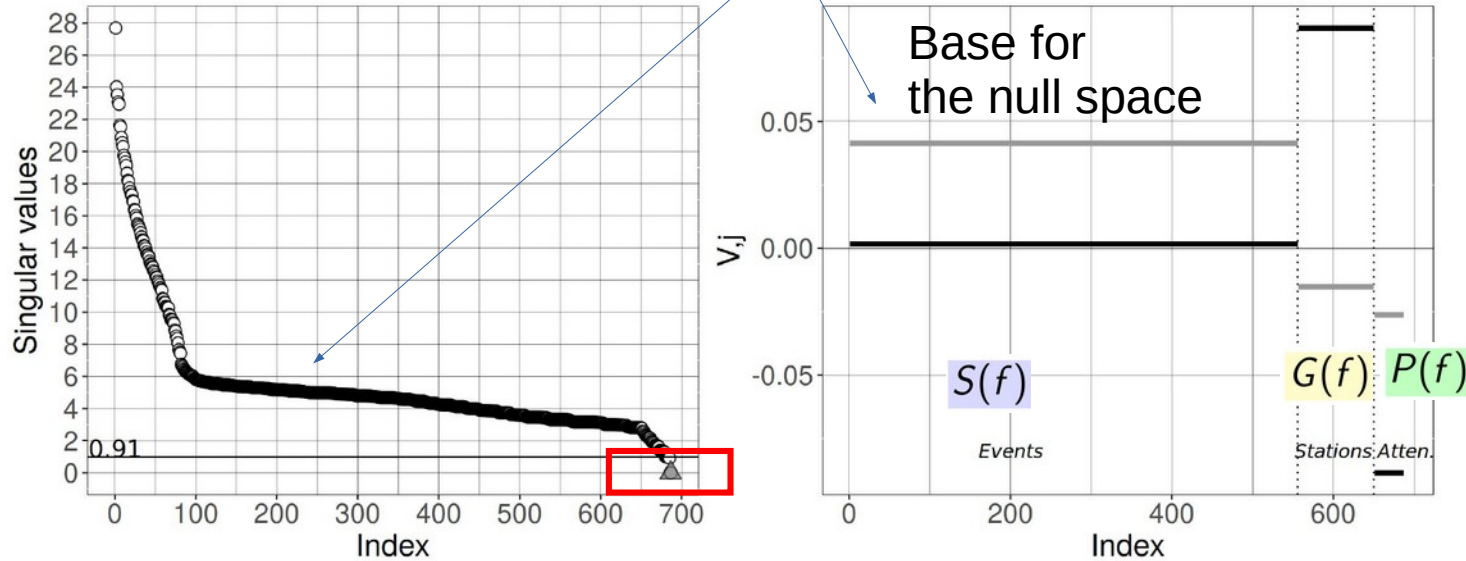
$$\text{SVD: } G = U \Lambda V^T$$

- Λ singular value matrix
- columns of V associated to $\lambda = 0$

$$FAS(f) = S(f) P(f) G(f)$$

Spectral decomposition GIT – non uniqueness

$$\text{SVD: } G = U \Lambda V^T$$

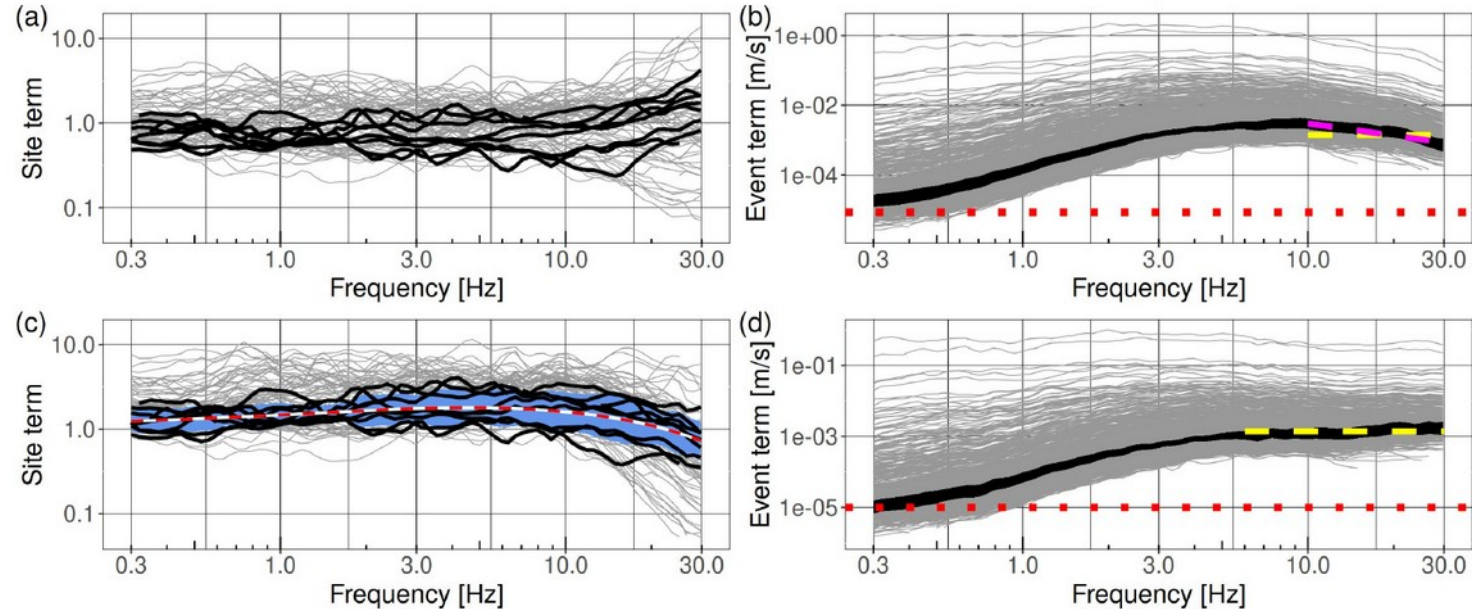


$$FAS(f) = S(f) P(f) G(f)$$

Bindi et al. 2023, SRL

Spectral decomposition GIT – non uniqueness

Impact of site constraint on source spectra



Bindi et al. 2023, SRL

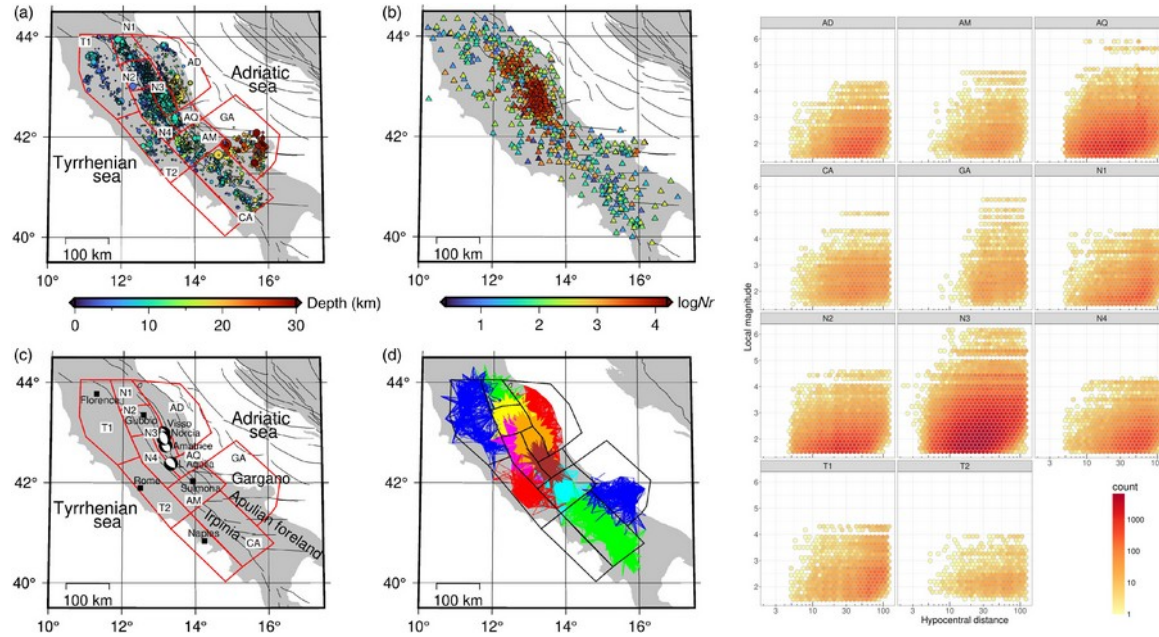
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 - **Regional attenuation: polygons, cell-grid**
 - Depth dependency

15:40 WP2 Gabrielli, S., *Spectral Corrections for Propagation and Site Effects: Implications for Stress Drop Estimation in the 2016 Mw 6.0 Amatrice Earthquake.*

Example regional attenuation

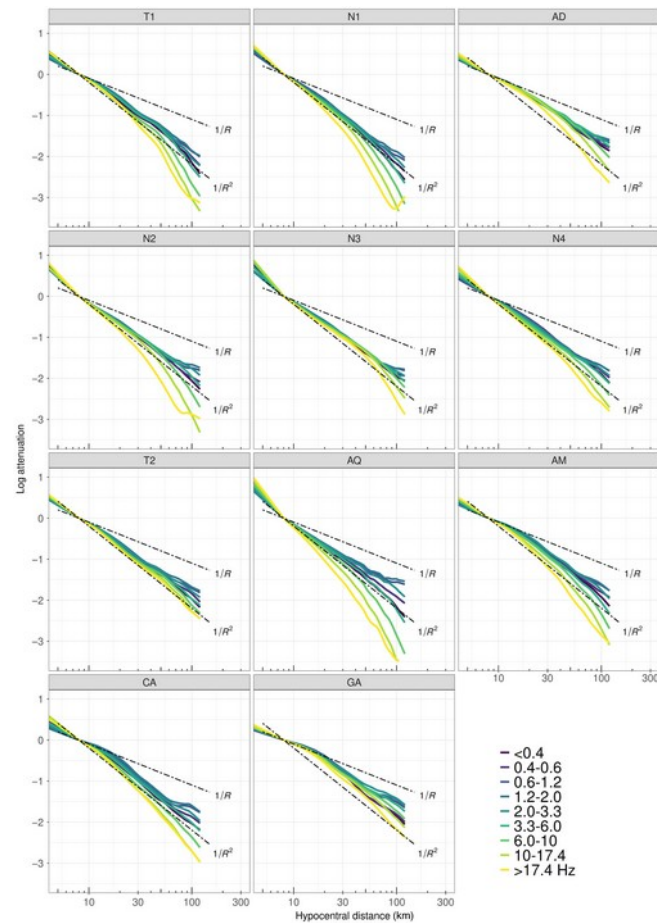
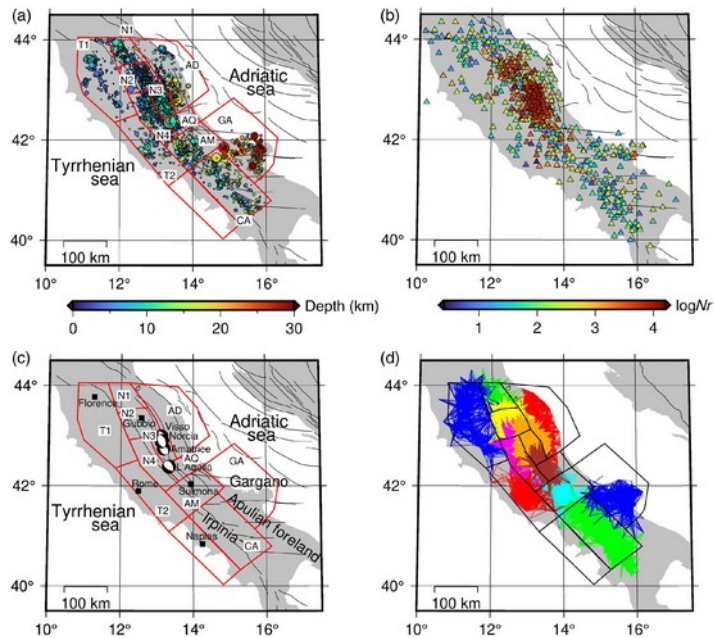


Bindi et al. 2024, BSSA
SECURE

Source regional models are useful for:

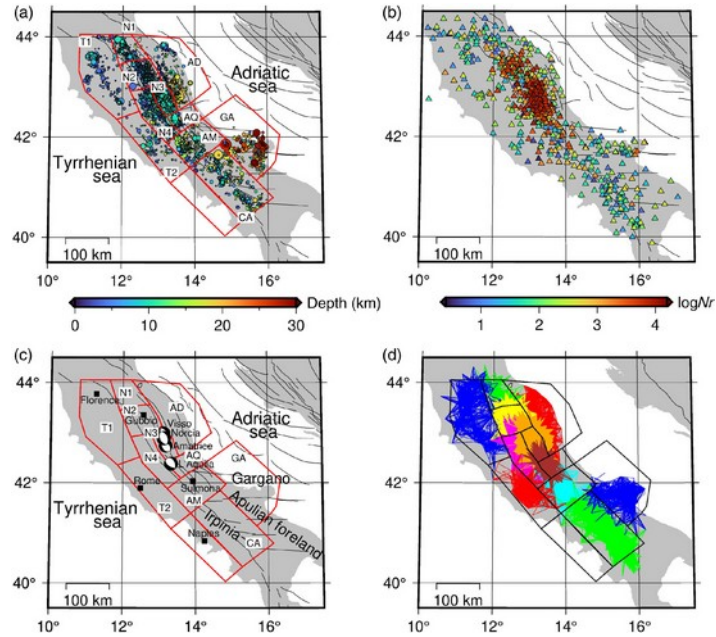
- capturing fault-system related differences in source behavior
- analysis of $\delta L2L/\delta Be$ residuals
- support development of magnitude conversions (e.g., T Allen, ongoing effort in Australia)
- prepare the set-up for numerical simulations

Example regional attenuation



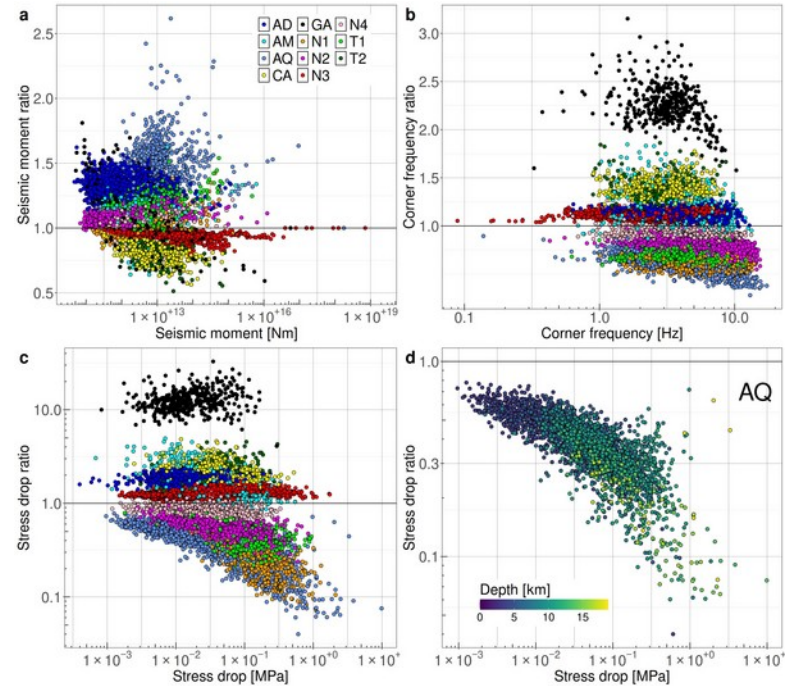
Bindi et al. 2024, BSSA
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Impact of regional attenuation



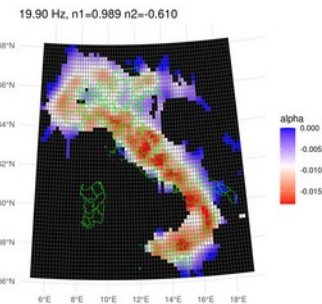
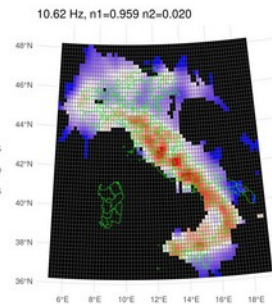
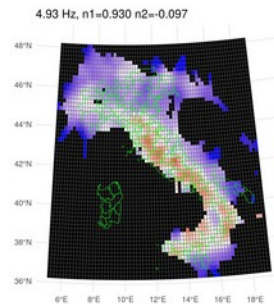
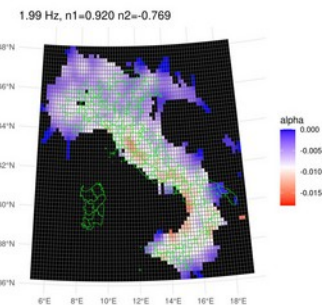
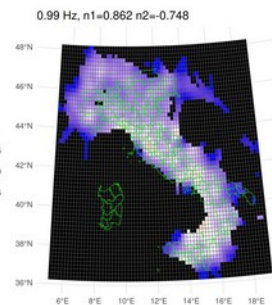
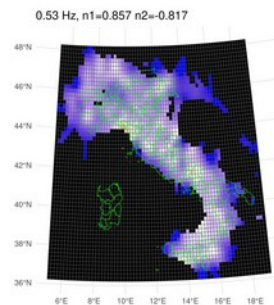
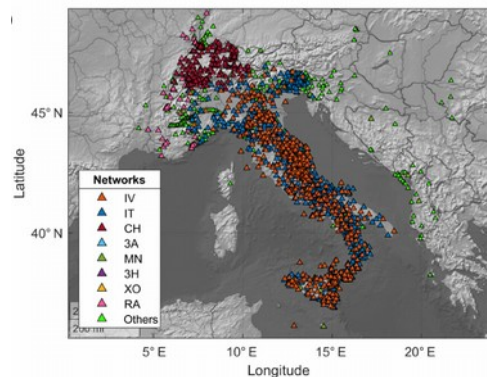
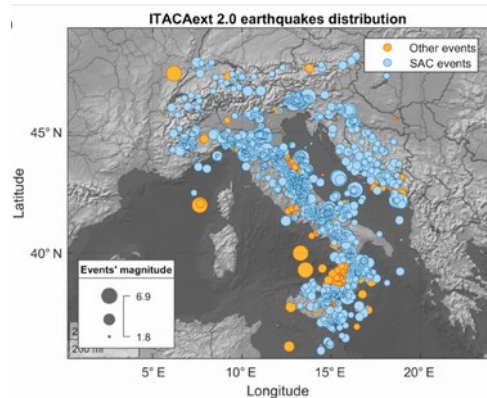
Bindi et al. 2024, BSSA
SECURE

RATIO single model/multiple models



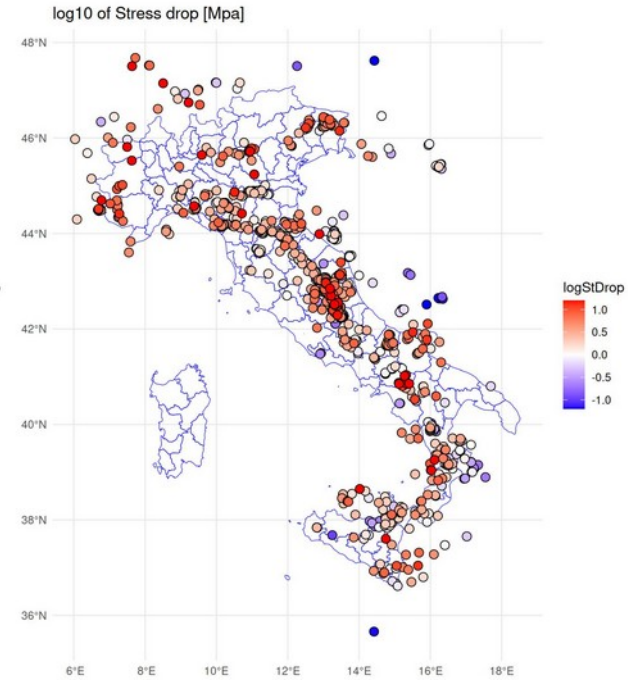
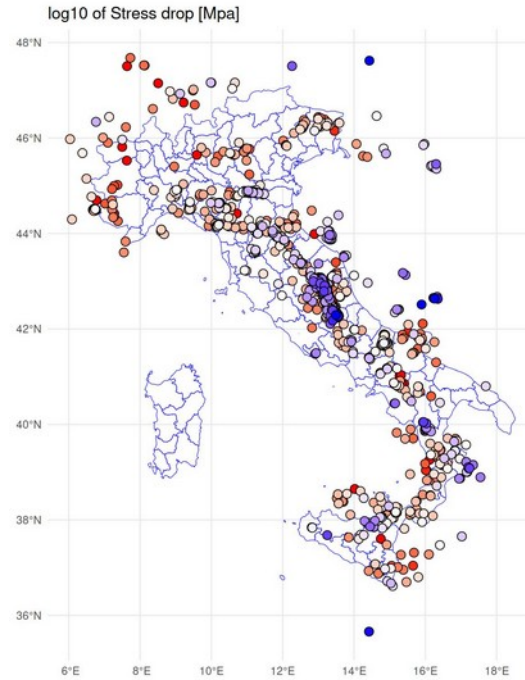
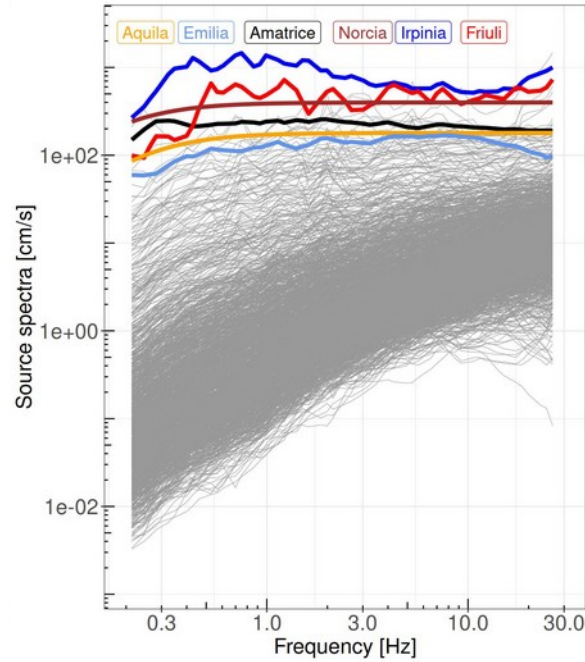
SECURE ITACA-extended

$$\log(\text{Path}_{ij}(f)) = \log G(R_{ij}; f) + \sum_k^{N_c} \lambda_{ijk} \alpha_k(f) \quad \begin{bmatrix} \mathbf{A} \\ \lambda \mathbf{L} \end{bmatrix} \mathbf{m} = \begin{bmatrix} \mathbf{d} \\ \mathbf{0} \end{bmatrix}$$



SECURE ITACA-extended

$$\log(\text{Path}_{ij}(f)) = \log G(R_{ij}; f) + \sum_k^{N_c} \lambda_{ijk} \alpha_k(f) \quad \begin{bmatrix} \mathbf{A} \\ \lambda \mathbf{L} \end{bmatrix} \mathbf{m} = \begin{bmatrix} \mathbf{d} \\ \mathbf{0} \end{bmatrix}$$

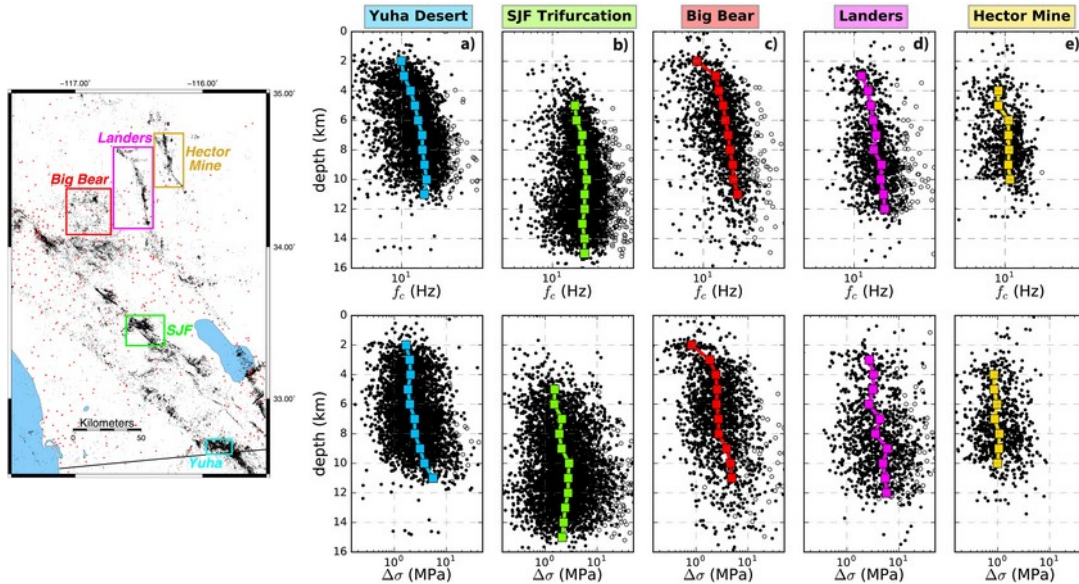


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- Non-Uniqueness of the solution
- Attenuation-source interplay
 - Regional attenuation: polygons, cell-grid
 - **Depth dependency**

The depth-dependency issue

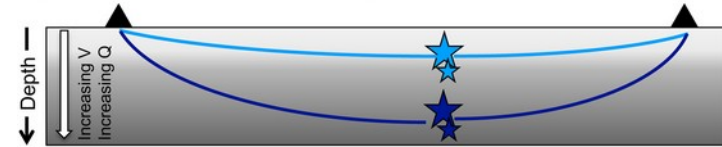


Trugman and Shearer 2017
<http://dx.doi.org/10.1002/2017JB013971>

(a) Same small EGF events for deep and shallow main events



(b) Separate small EGF events for deep and shallow main events



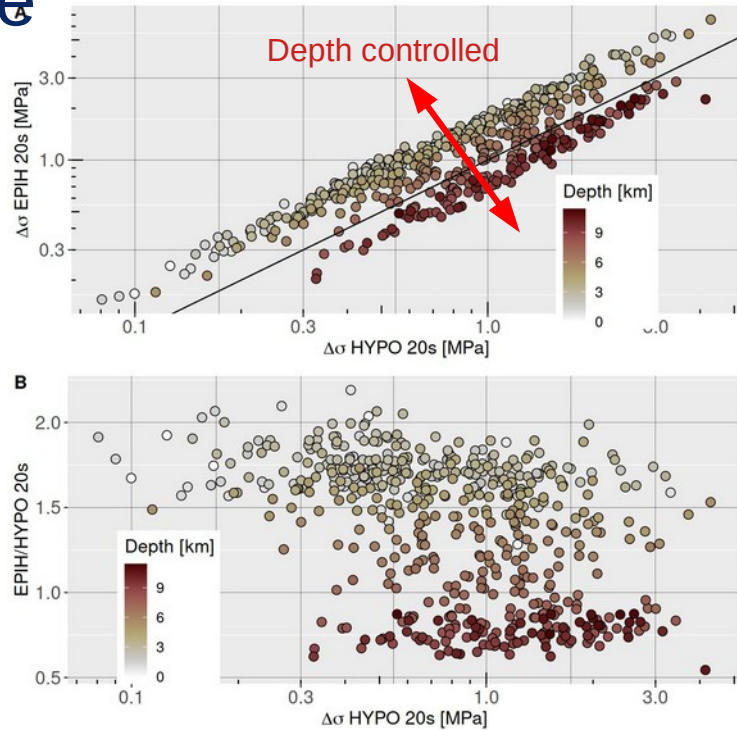
Does Earthquake Stress Drop Increase With Depth in the Crust?

Rachel E. Abercrombie, Daniel T. Trugman, Peter M. Shearer, Xiaowei Chen, Jiewen Zhang, Colin N. Pennington, Jeanne L. Hardebeck, Thomas H. W. Goebel, Christine J. Ruhl

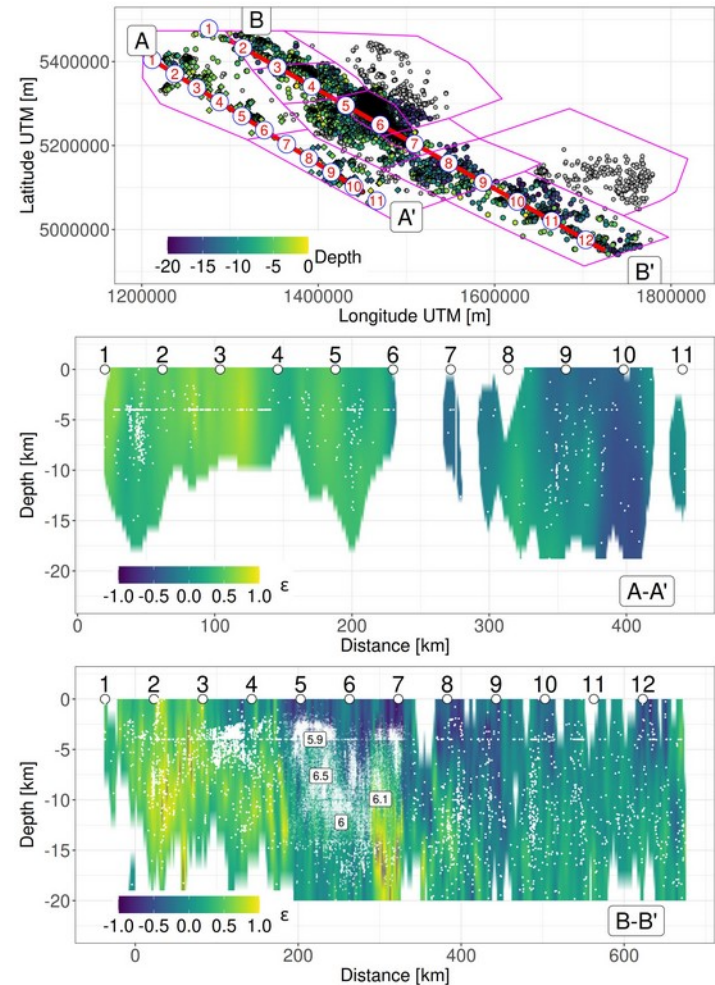
First published: 16 September 2021 | <https://doi.org/10.1029/2021JB023214>

Depth issue

Depth dependent /
Depth independent



Bindi et al. 2023, SRL



Bindi et al. 2024, BSSA SECURE

SECURE - Farewell

Ground-motion models describe variability statistically. Spectral decomposition allows us to give that variability a physical meaning.

In **SECURE**, we applied this idea at regional scale in Central Italy.

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In SECURE, we applied this idea at regional scale in Central Italy.

Thank you, SECURE community, for the nice journey we took together.

Alla prossima !!! Dino