

# Site Classification Derived From Spectral Clustering of Empirical Site Amplification Functions

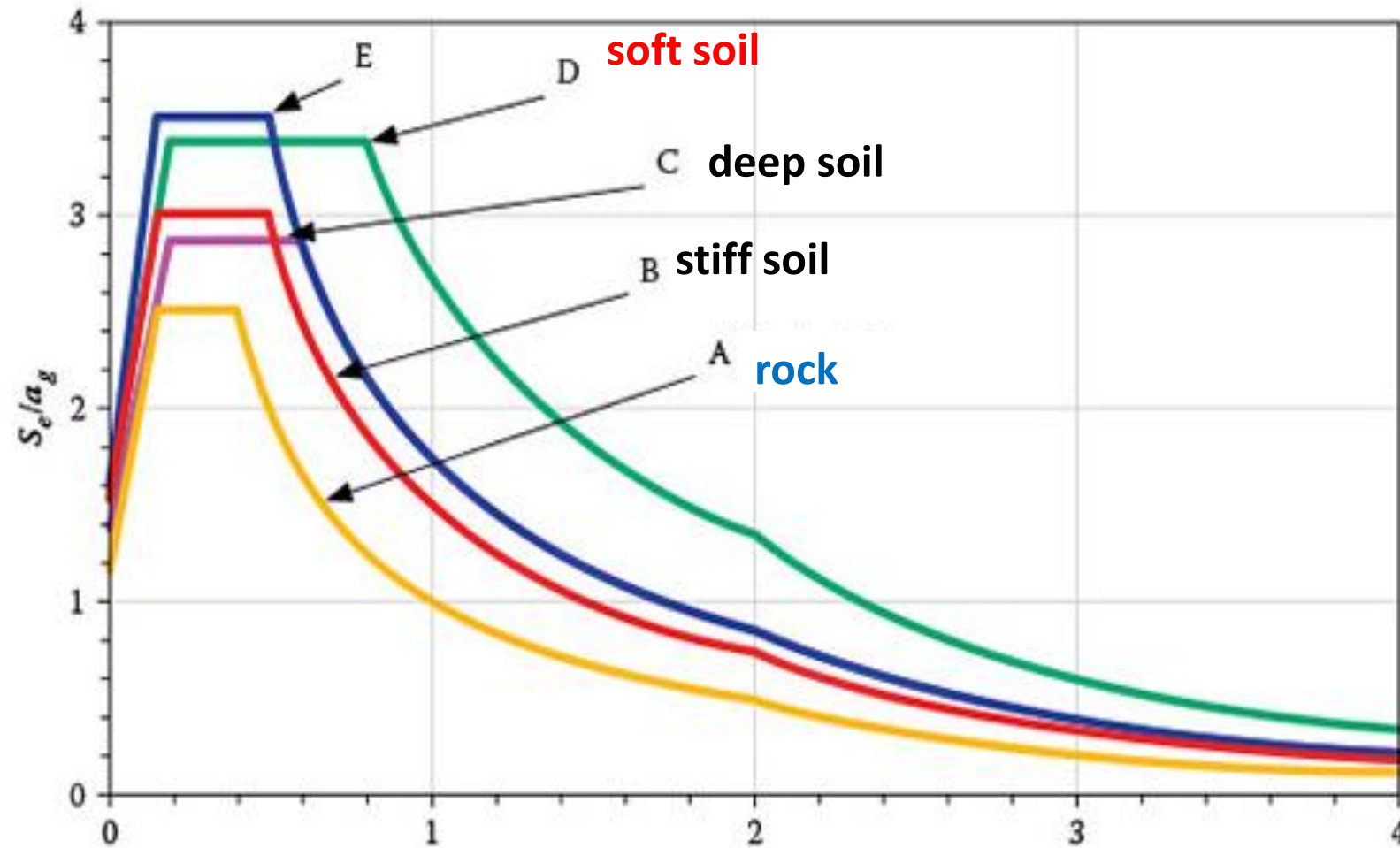
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with Fabrice Cotton & Dino Bindi

Section 2.6: Seismic Hazard and Stress Field

GFZ Potsdam

Site effects: “The effect of ‘local site conditions’ on ‘ground motion’”



WORKU, A. *Soil-structure-interaction provisions: A potential tool to consider for economical seismic design of buildings?*. J. S. Afr. Inst. Civ. Eng. [online]. 2014, vol.56, n.1 , pp.54-62.

## What are the issues?

1. Traditionally, site classes are defined a priori:  $V_{s30}$ , SPT, PI ranges etc.
2. Within each site class, the site-to-site variability of amplification is large

## What is our plan?

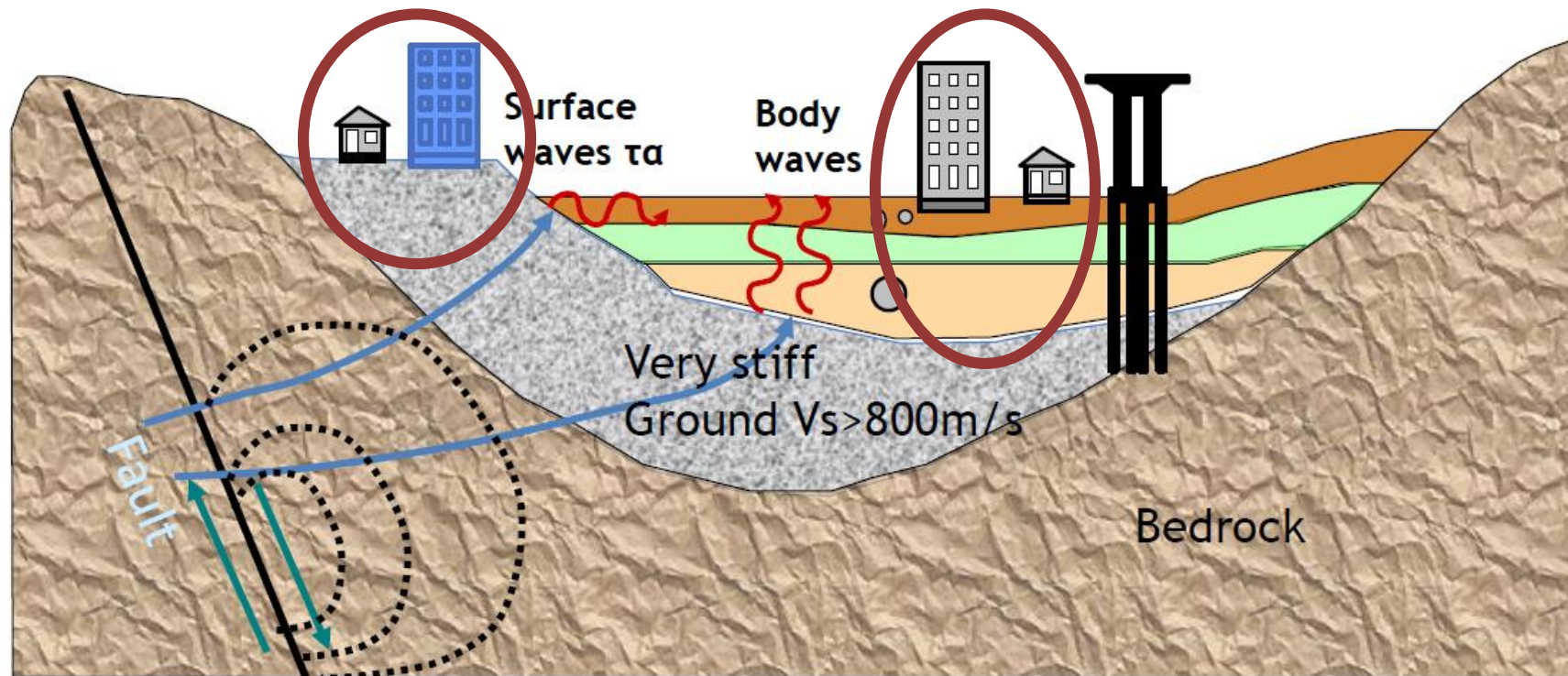
1. Use a rich strong motion dataset
2. Derive empirical site amplification functions for well-recorded sites
3. Use machine learning techniques to identify and cluster similar sites
4. Evaluate site response proxies that explain the new site classes

# Empirical Site Amplification Functions : $\delta S_2 S_s(T)$

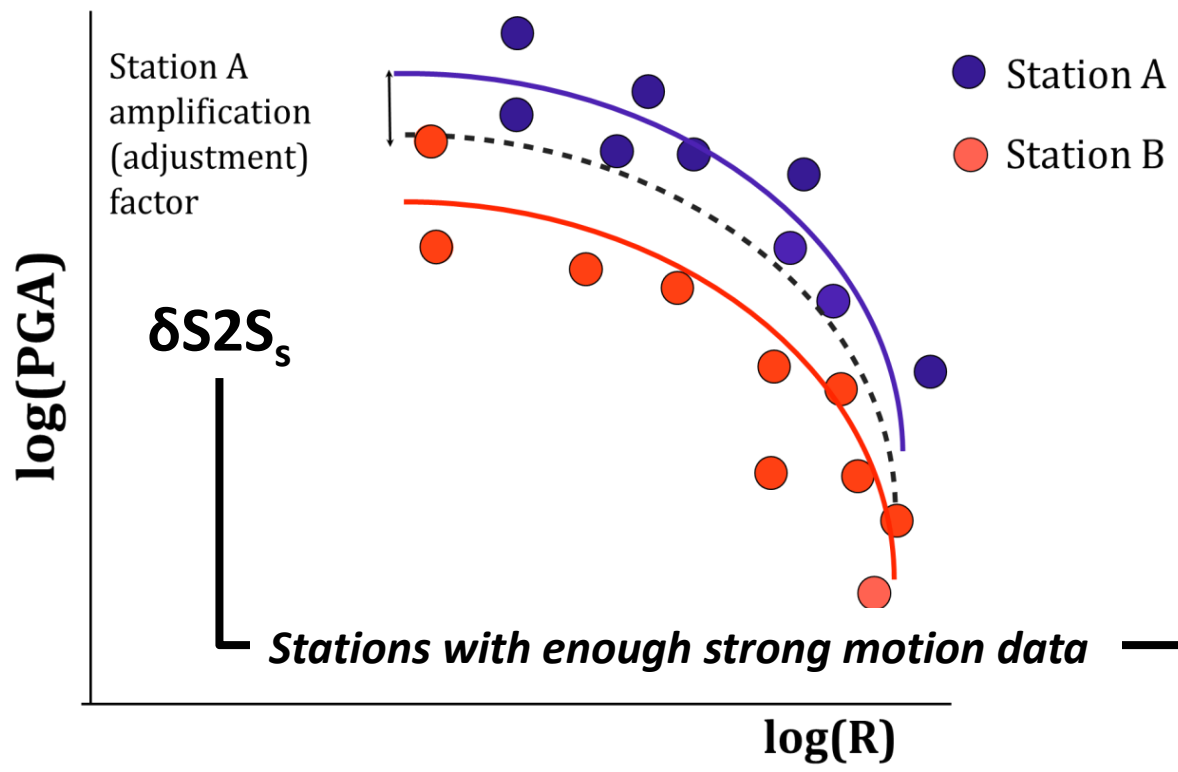
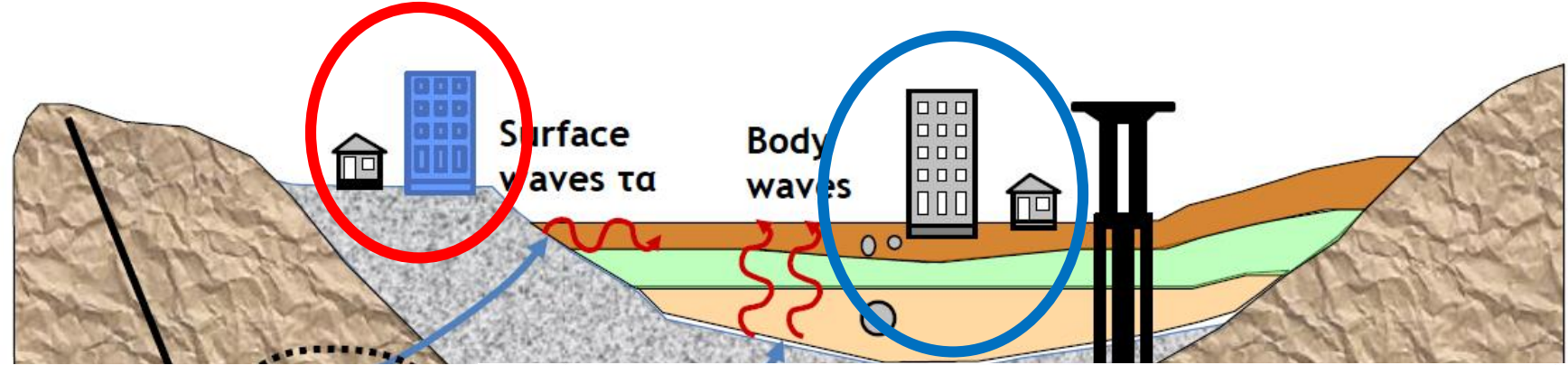
## Ground Motion Prediction Equations (GMPE)

$$\ln(GM_{e,s}) = F_M(M_e) + F_D(R_{e,s}, M_e) + F_S(\theta_s) + \delta B_e + \delta S_2 S_s + \delta W S_{e,s}$$

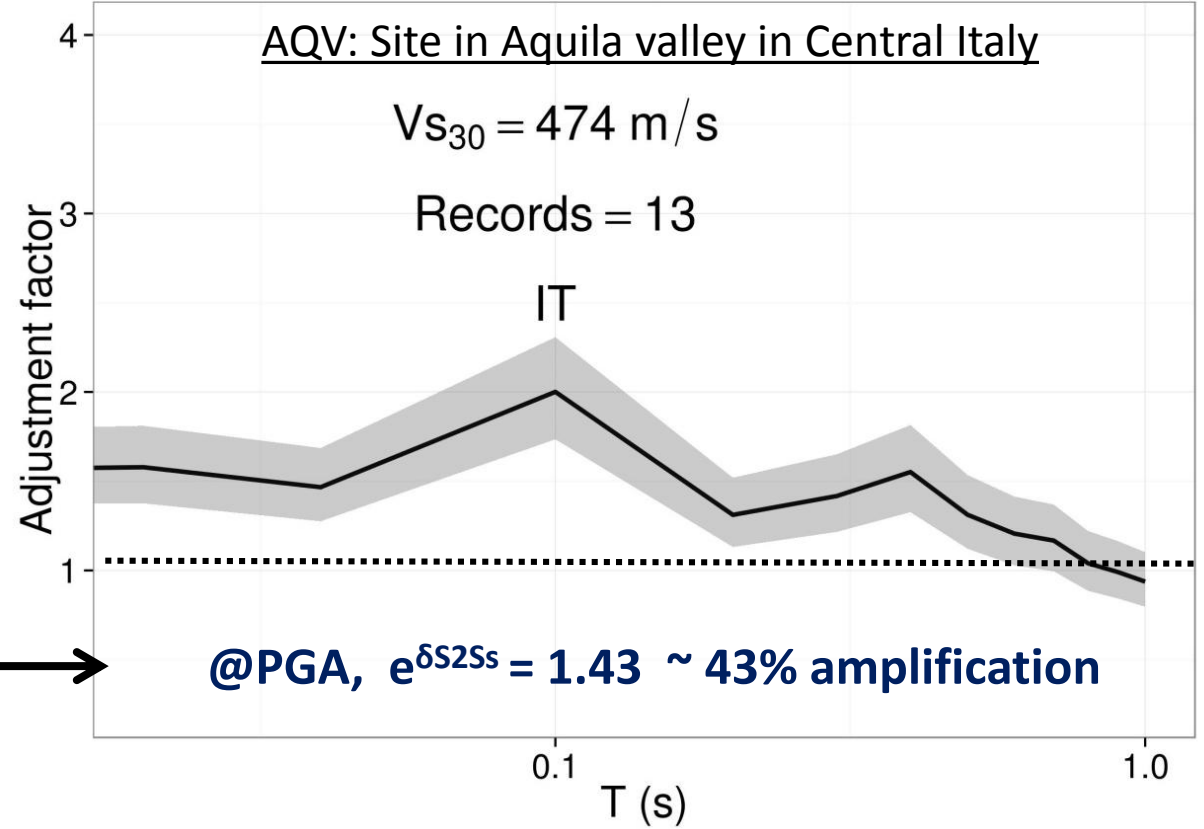
Observations                      Predictor functions                      Random effects                      Residuals



# Empirical Site Amplification Functions : $\delta S2S_s(T)$



*Stations with enough strong motion data* →





# Strong Motion Dataset: KiK-Net

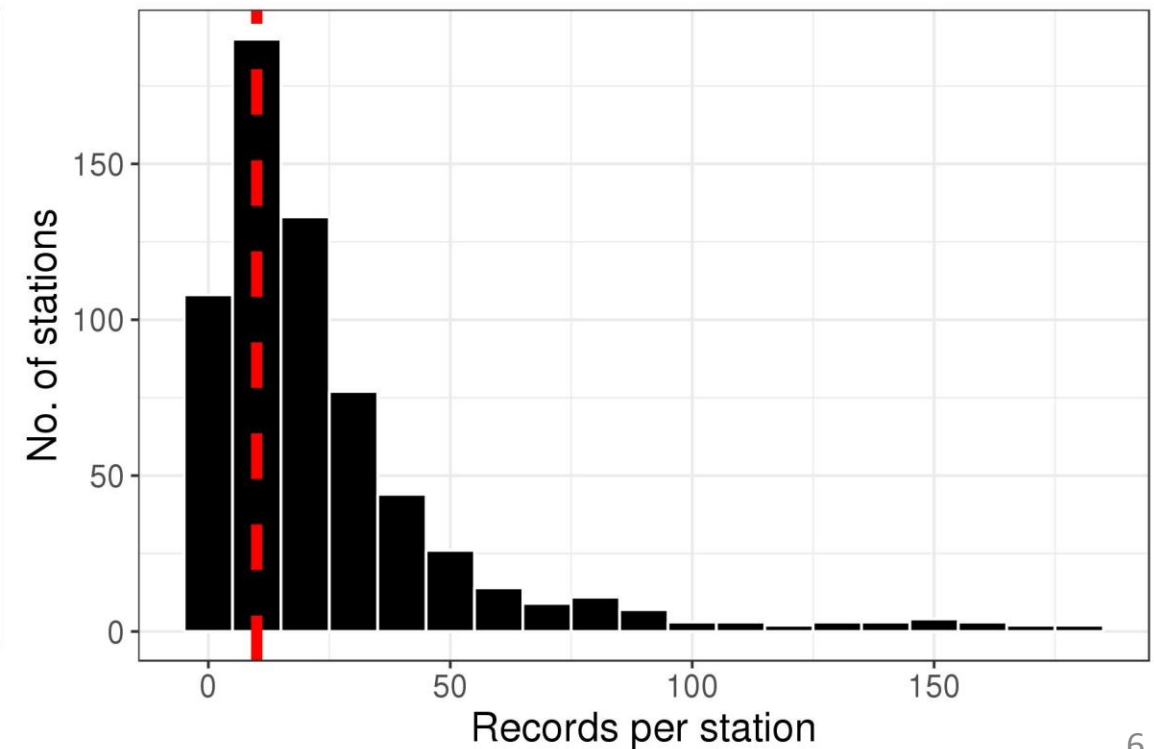
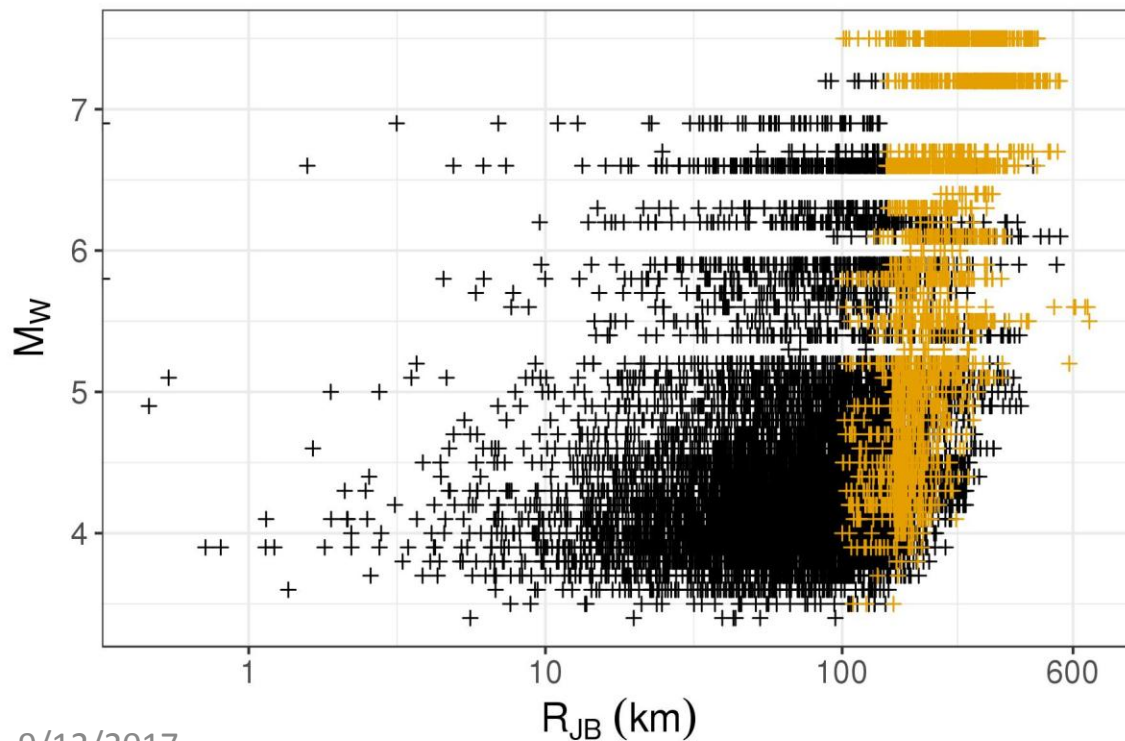
GMPE for GM of H - components of Response Spectra for Shallow Crustal events

~16000 records :  $M_{3.4}-M_{7.5}$ ,  $0\text{km} < R_{JB} < 600\text{km}$ ,  $T = 0.01\text{s} - 7\text{s}$

~ 500 sites with more than 10 records

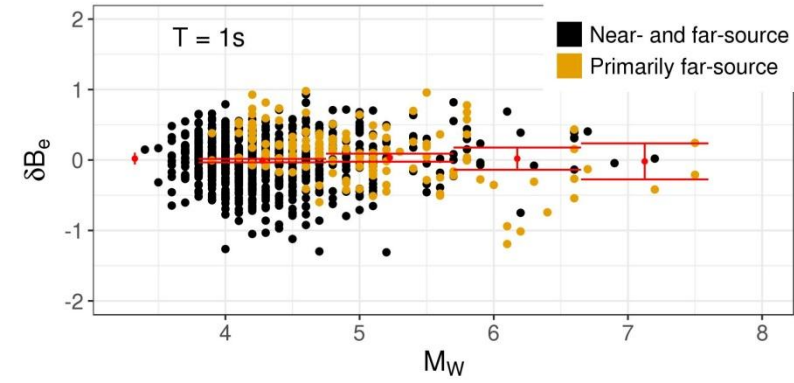
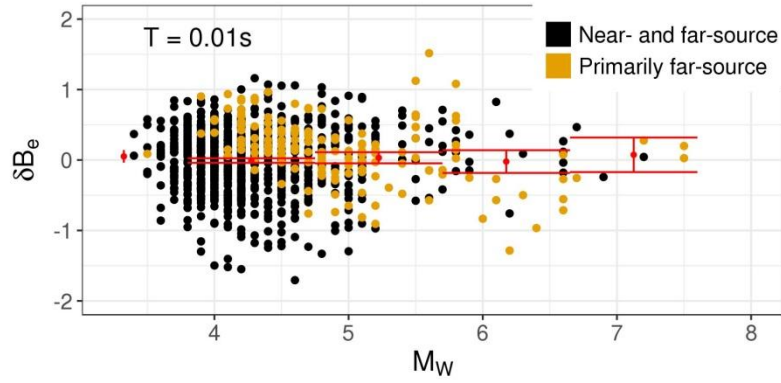
(Dawood, Rodriguez-Marek et al. 2016)

## Data distribution

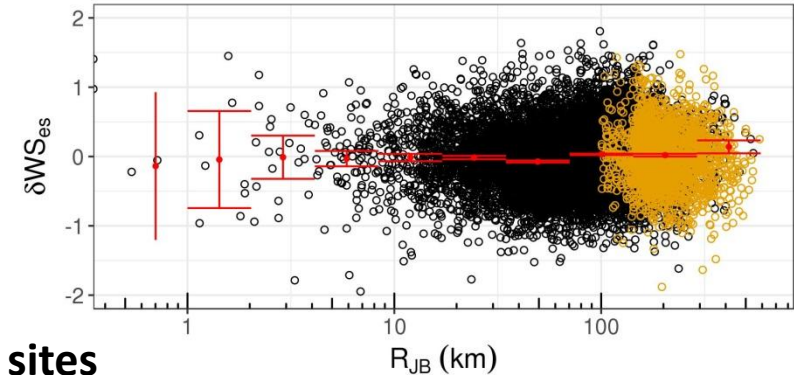
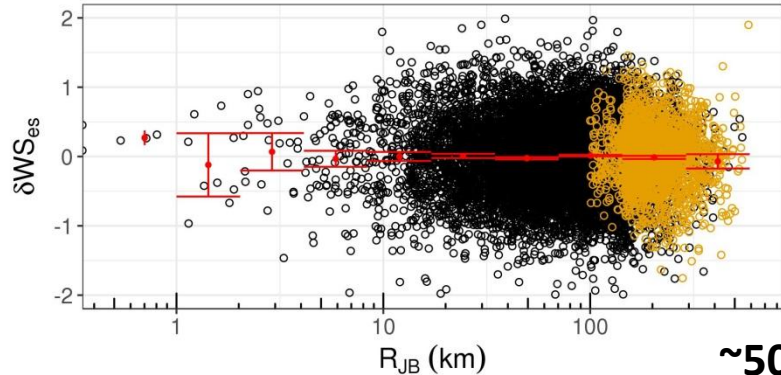


High frequency  $\delta S2S_s$  shows a weak trend with  $V_{s30}$

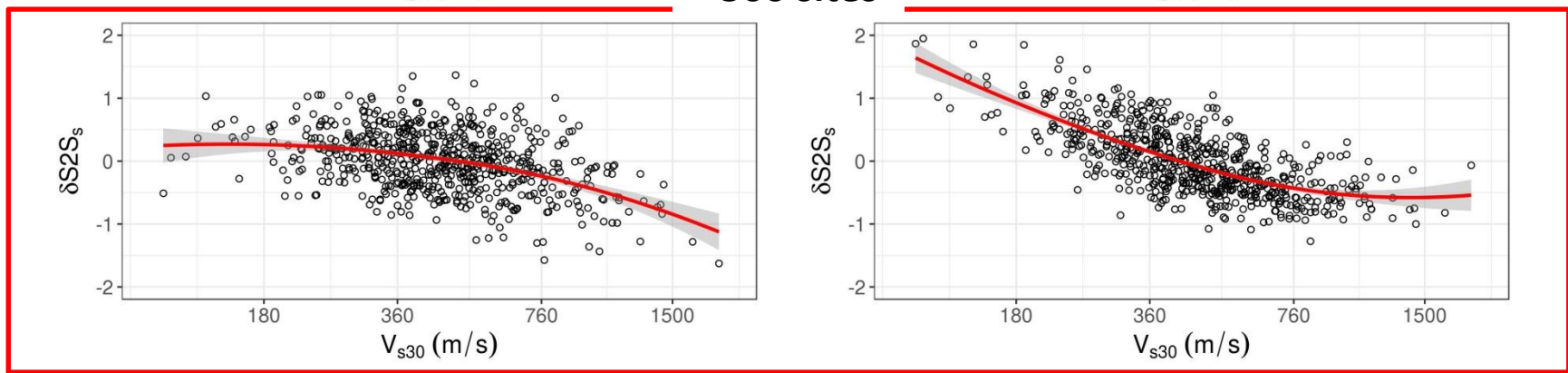
**T = 0.01s**



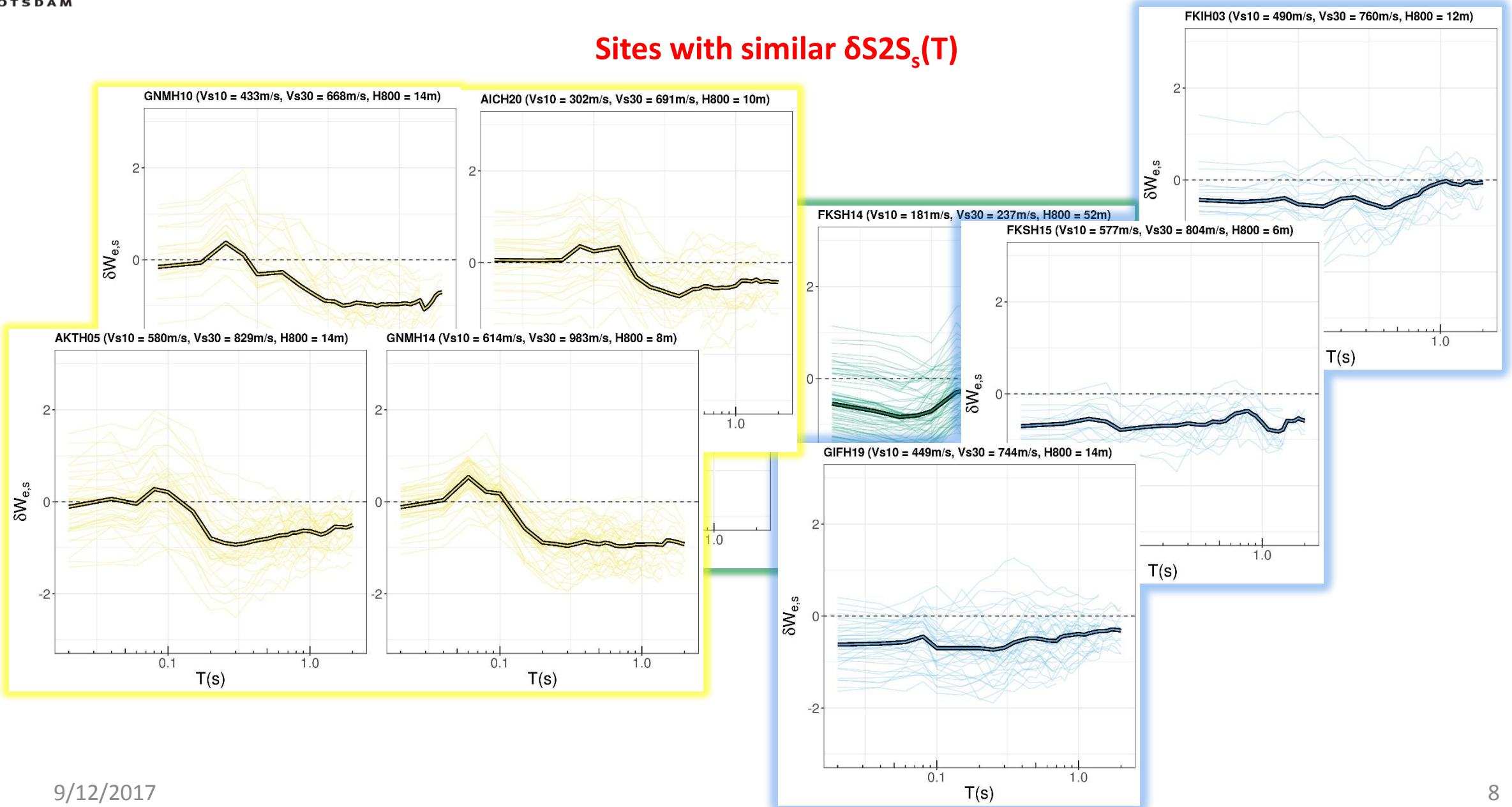
**T = 1s**



~500 sites



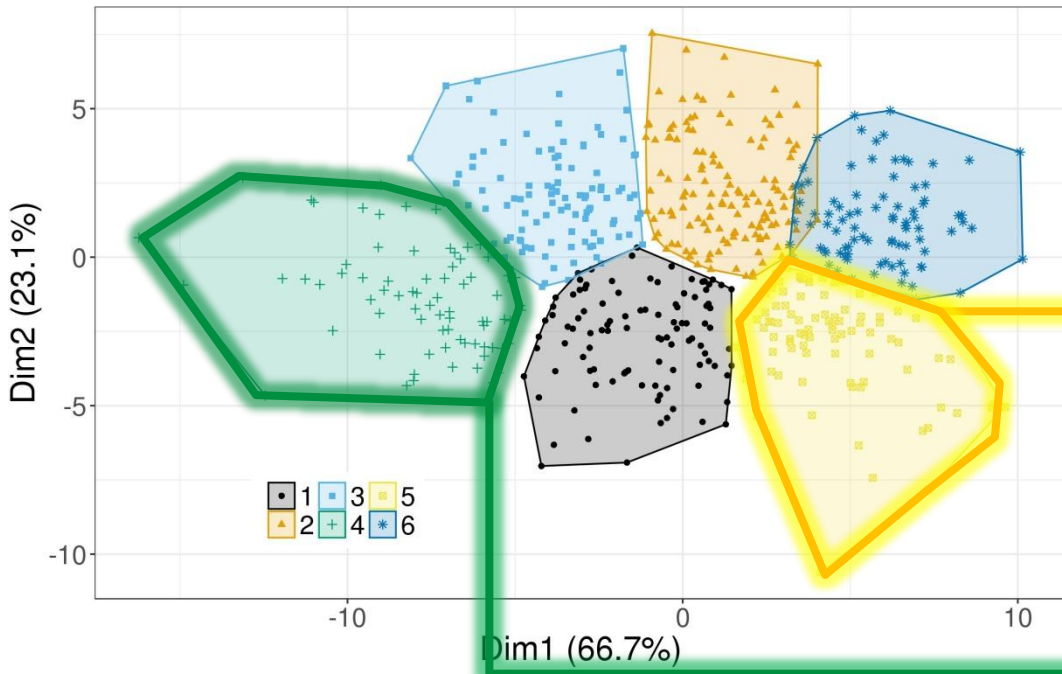
## Sites with similar $\delta S2S_s(T)$



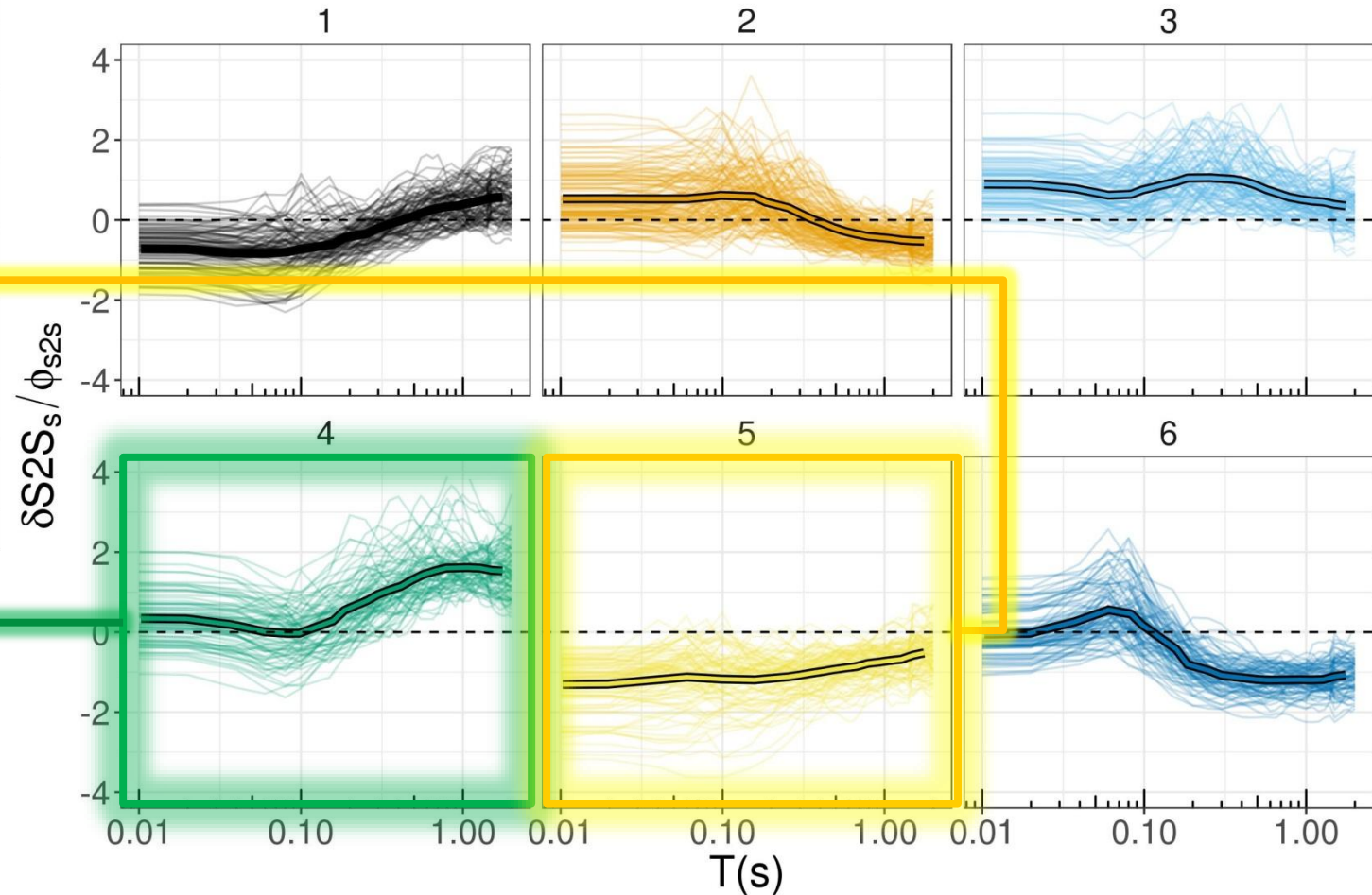


## K-mean clustering of sites with similar response $\delta S2S_s(T)$

**K-mean clusters**

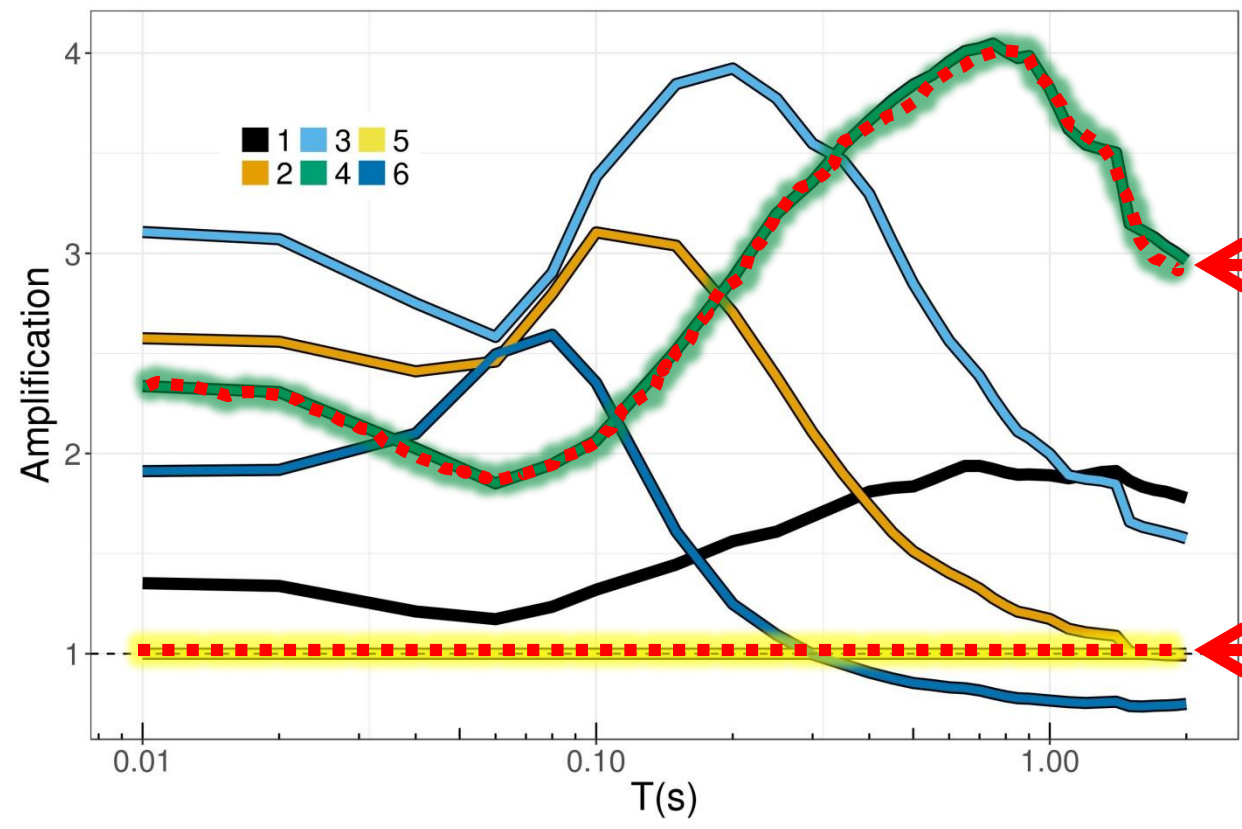


**Clustered  $\delta S2S_s(T)$**

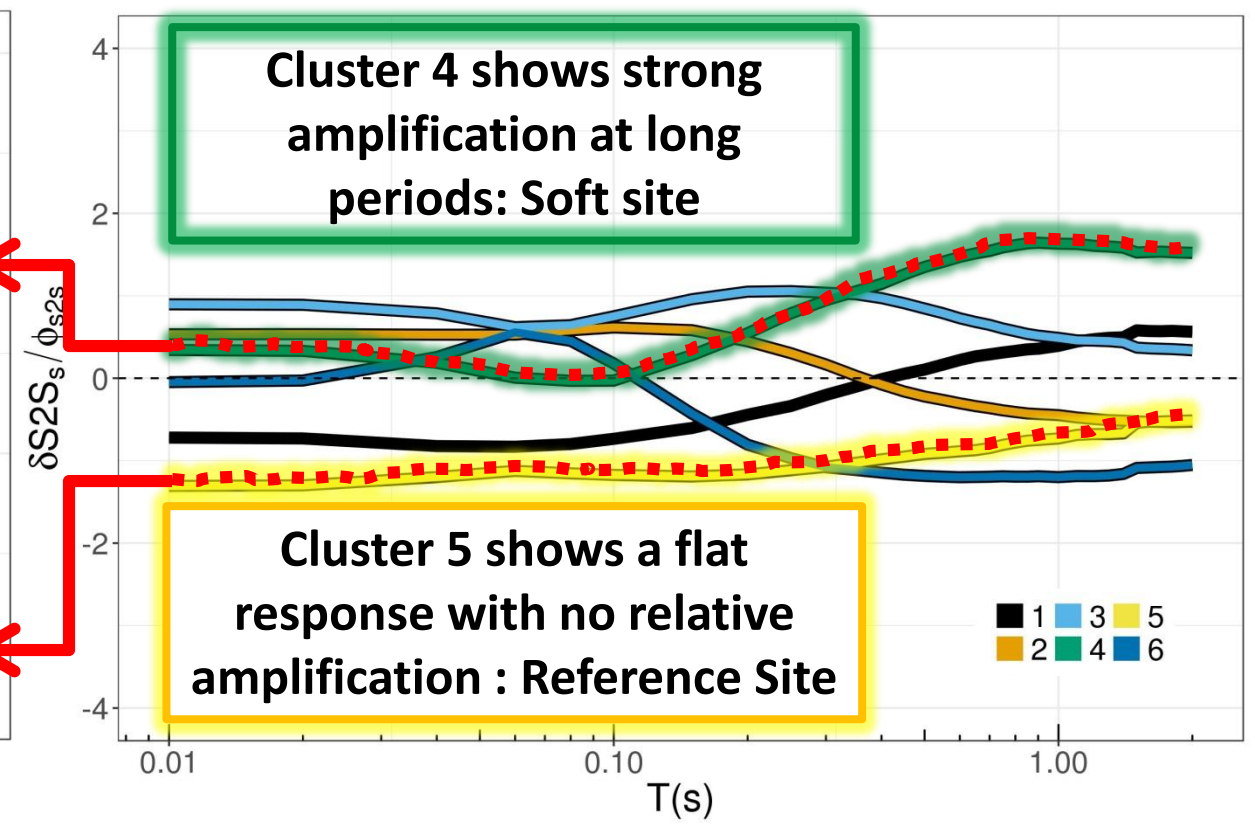


Scale w.r.t reference site  $\delta S2S_s(T)$ , and then  $e^{\delta S2S_s(T)}$

**Amplification functions**

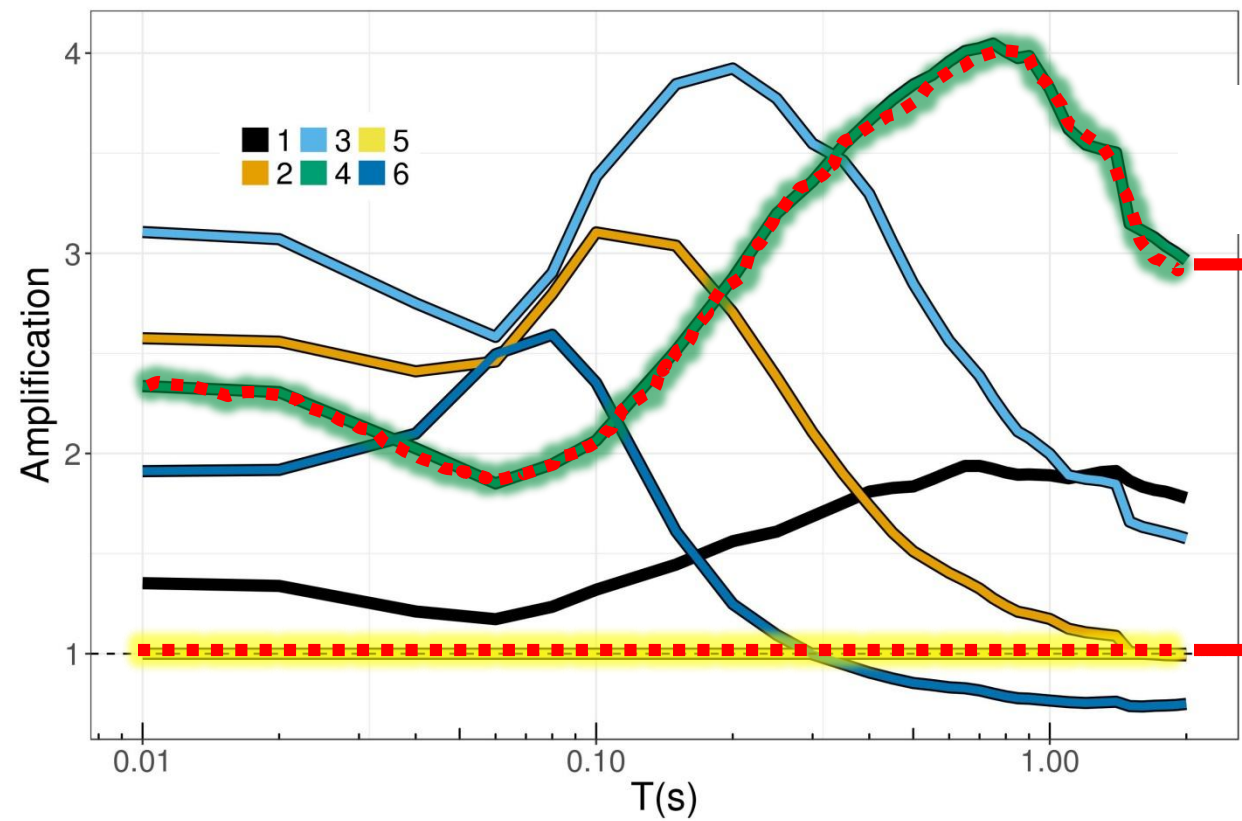


**Means of clustered  $\delta S2S_s(T)$**

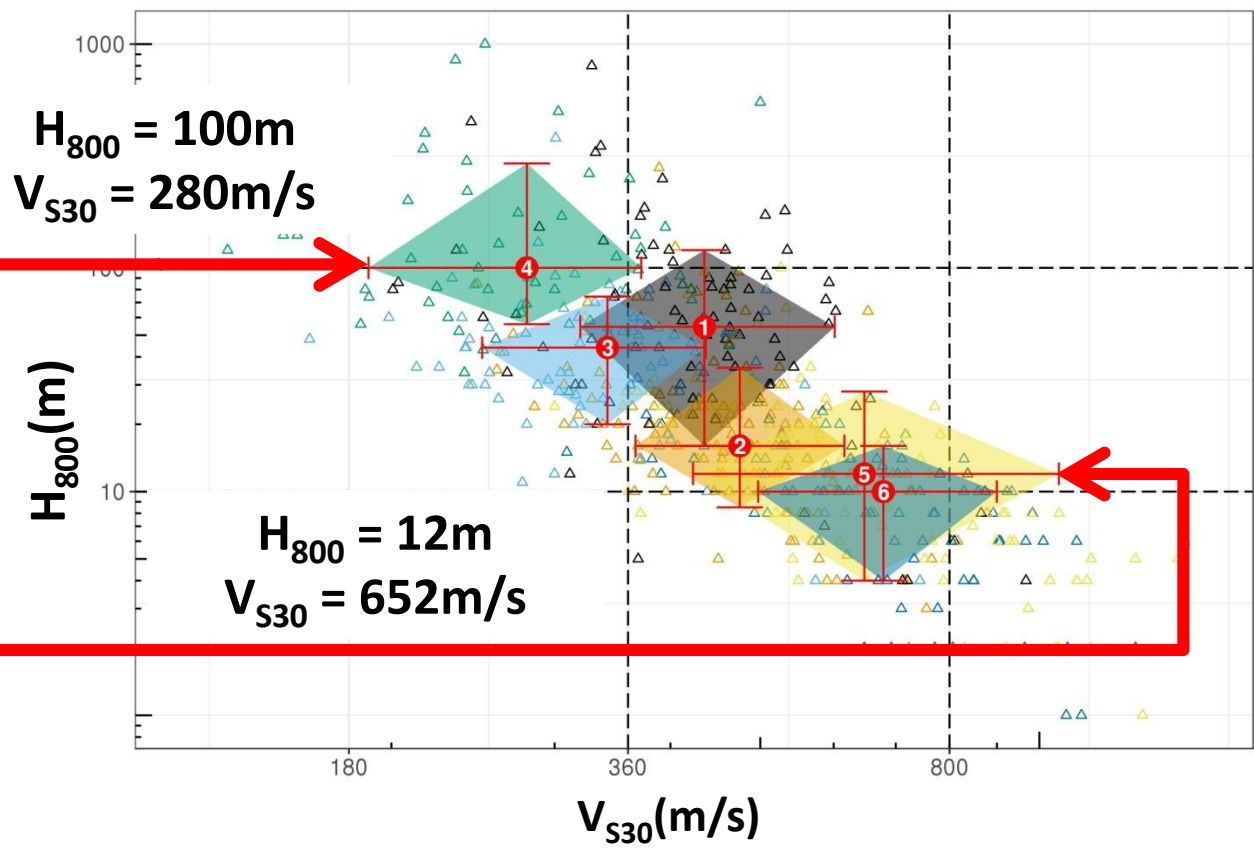


## Physical meaning of cluster specific $\delta S_{2S_s}(T)$

Amplification functions



Site conditions

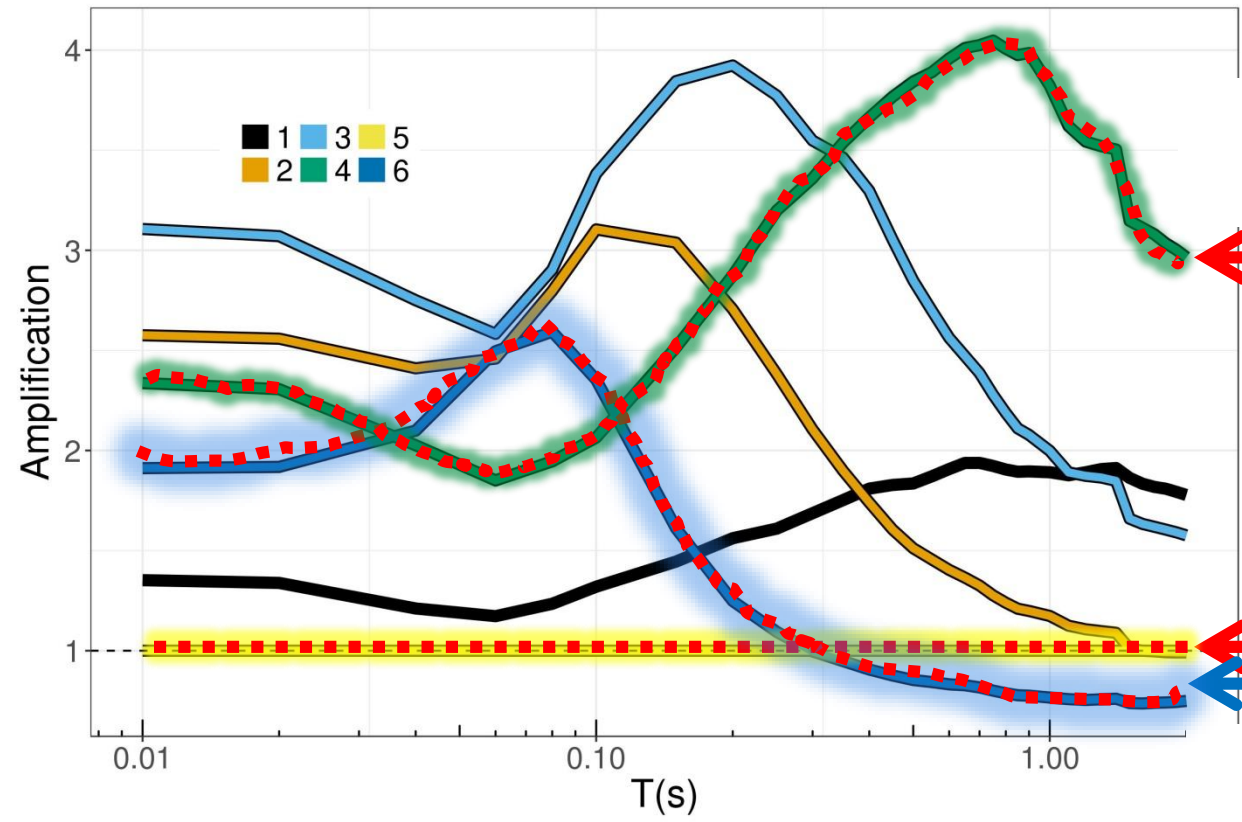


$H_{800} = 100\text{m}$   
 $V_{S30} = 280\text{m/s}$

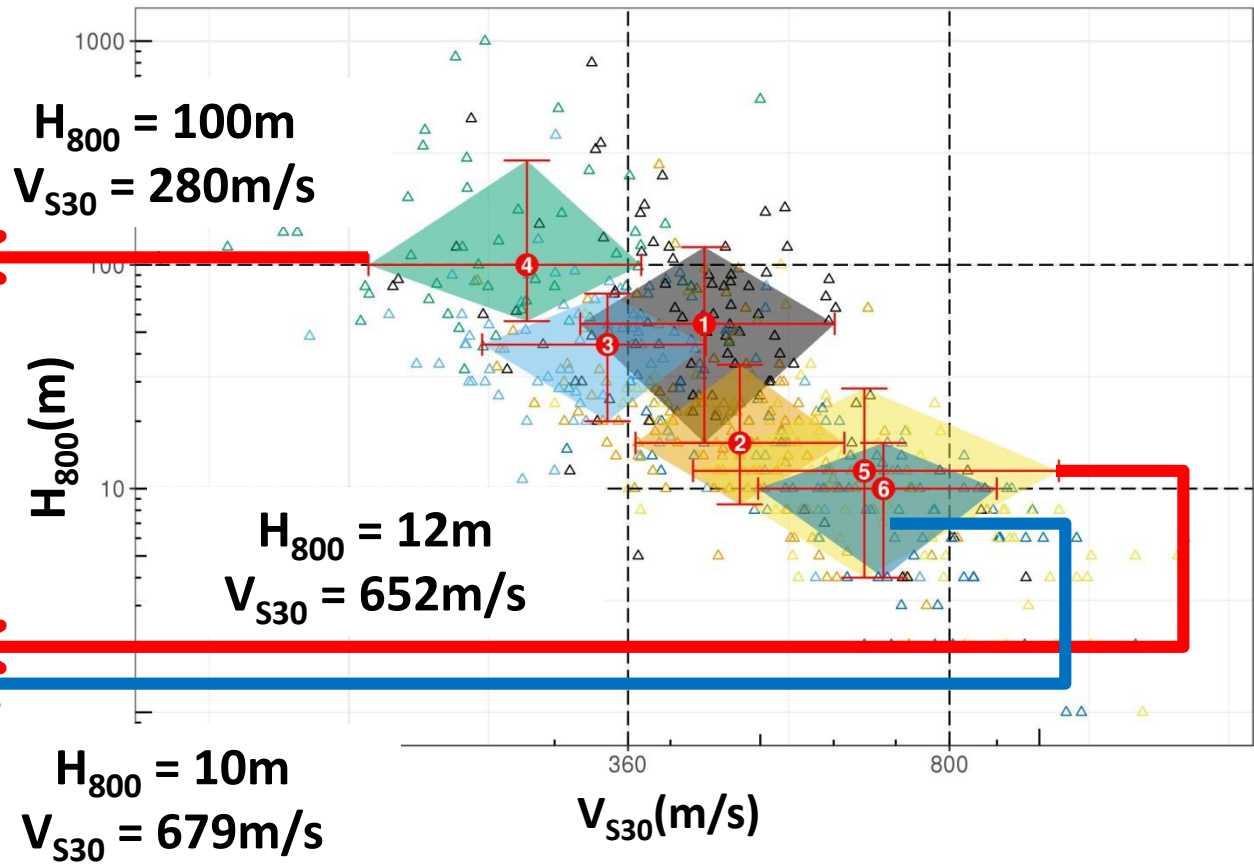
$H_{800} = 12\text{m}$   
 $V_{S30} = 652\text{m/s}$

## Reference 'rock' site conditions?

**Amplification functions**

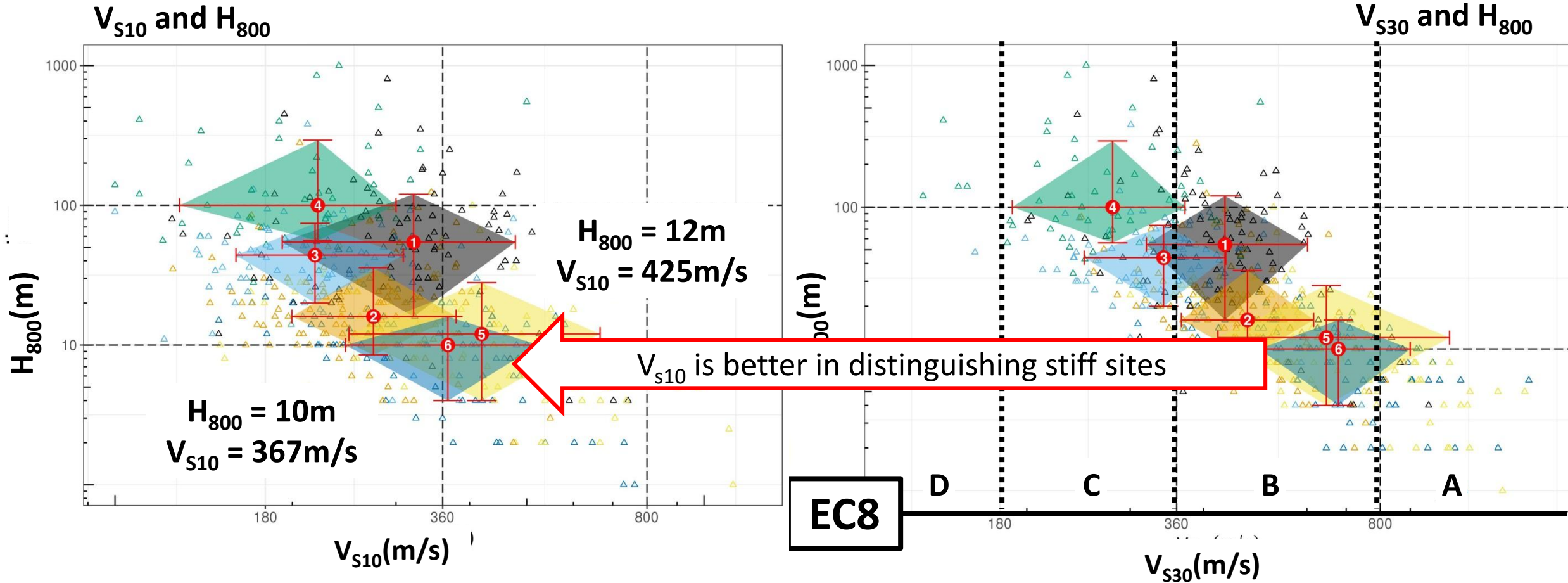


**Site conditions**



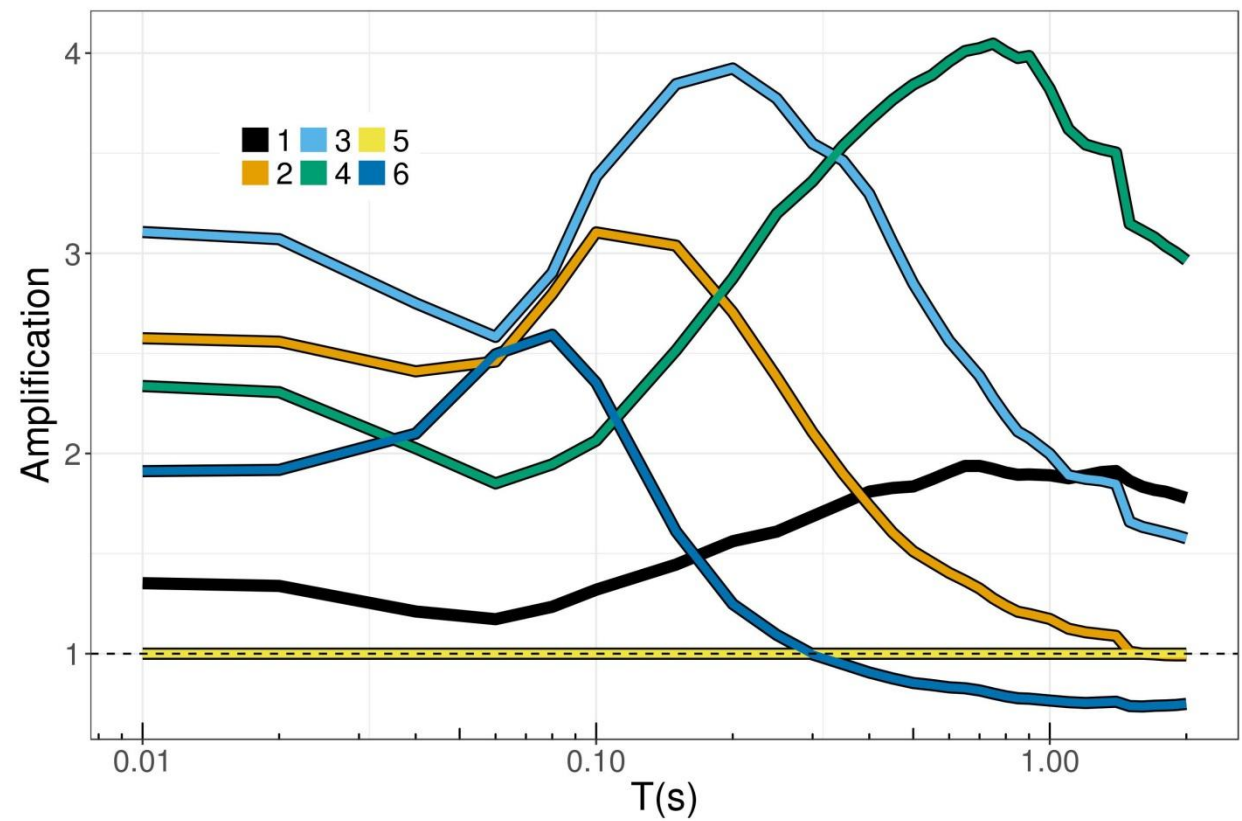


**$V_{S30}$  based classification may not be efficient**

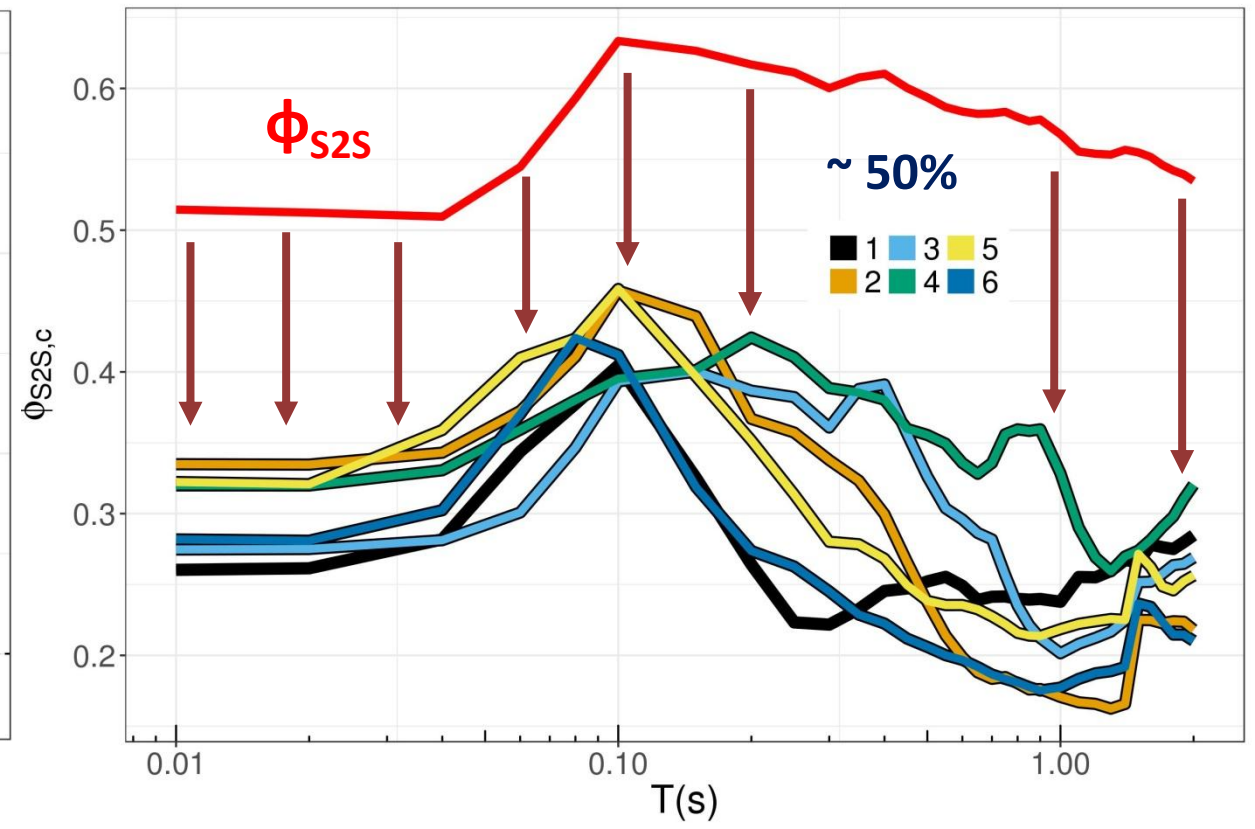


**Within cluster site-to-site response variability ~ 50% smaller**

**Amplification functions**



**Site-to-site variability**



## What were the key issues?

1. Pre-defined sites classes based on  $V_{S30}$  may not be efficient
2. Large site-to-site variability within  $V_{S30}$  based classes

## What we tried?

1. Site-specific random effects  $\delta S2S_s(T)$  as empirical site AFs
2. Unsupervised machine learning techniques to cluster sites with similar response

## What we found?

1.  $V_{S10} - H_{800}$  is an optimal proxy to classify 6 site clusters
2.  $\sim 50\%$  smaller within-cluster site-to-site variability

**...Thank you... review?**

## What next?

1. The tools are open-source and easy to use... more sophistication is needed?
2. With a pan-European dataset, we may expect very different results!!!