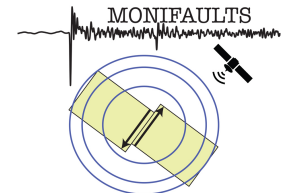


Direct fault states assessment from wavefield properties: application to the 2009 L'Aquila earthquake

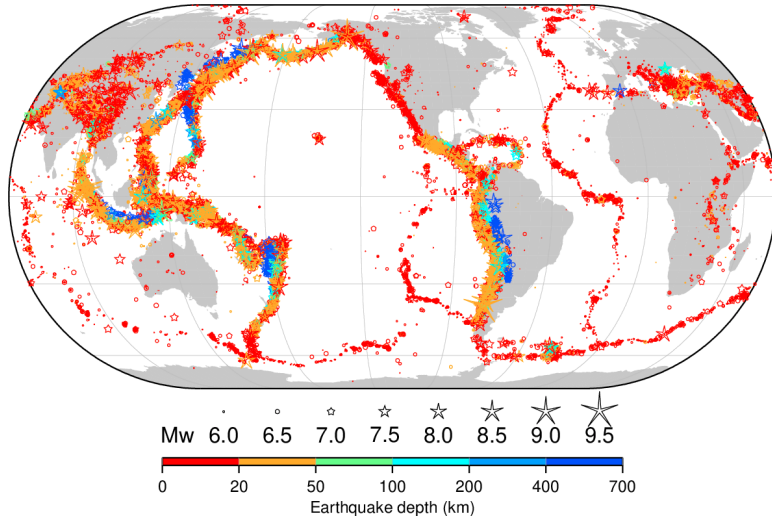
Peidong Shi^{1*}, Leonard Seydoux¹, Piero Poli¹

¹ Institut des Sciences de la Terre, Université Grenoble Alpes

* peidong.shi@univ-grenoble-alpes.fr



Backgrounds Go beyond the catalog

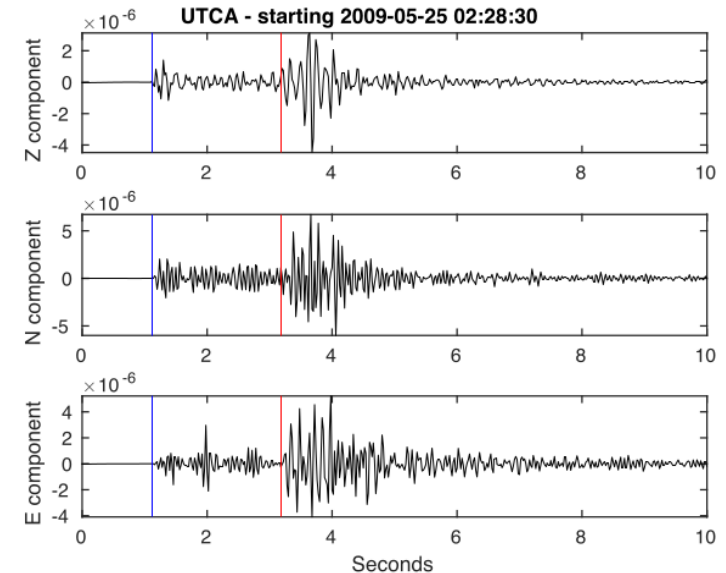


ISC-GEM Catalogue

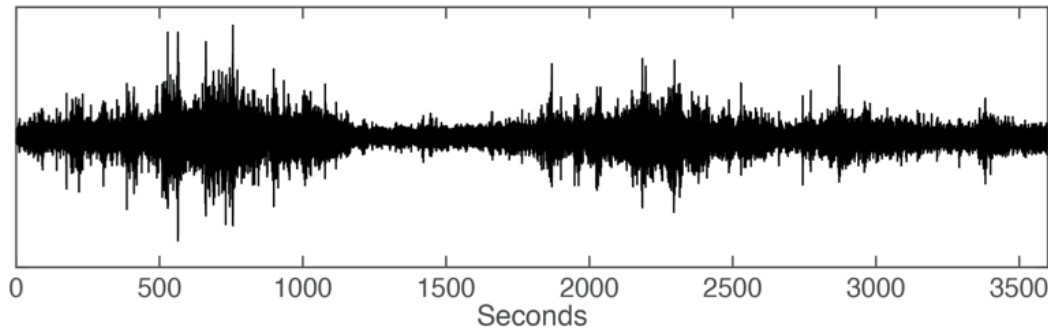
Seismic catalog is pervasively used for studying earthquake and fault physics.

Seismic catalogs are currently the main way of labeling seismic data.

In continuous seismic data, there may be signals that cannot be easily included into a seismic catalog.



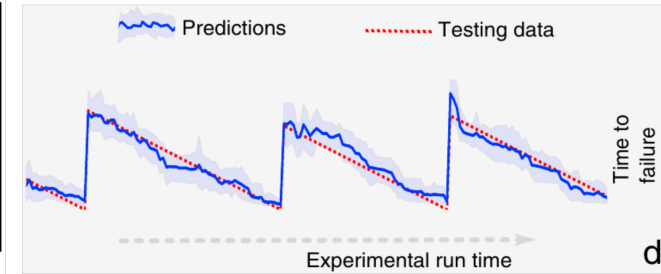
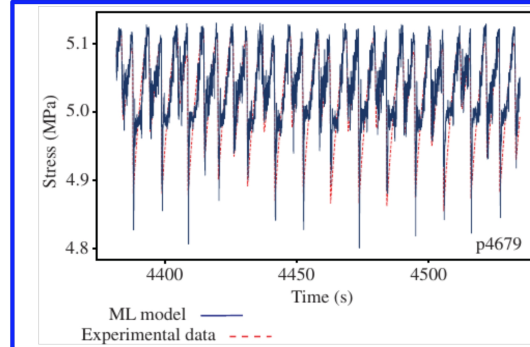
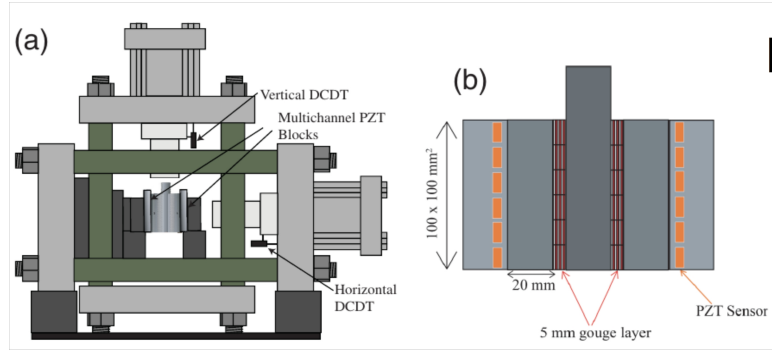
Tremors, slow-earthquakes, anomalous signals, ...



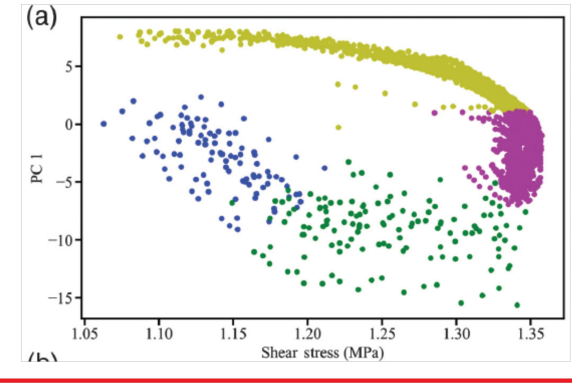
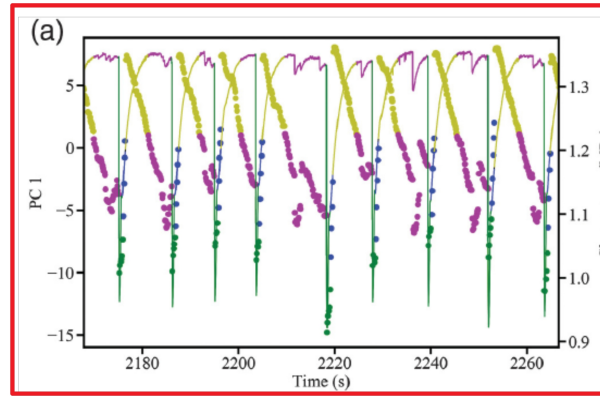
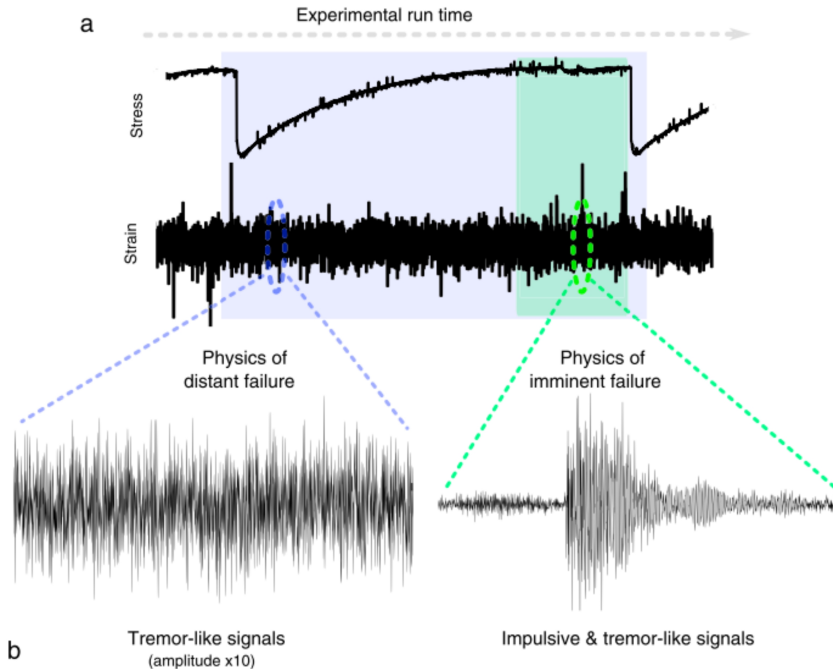
Continuous seismic data contain much more information about fault physics than what is actually employed.

Introduction: utilize continuous wavefield (laboratory experiments)

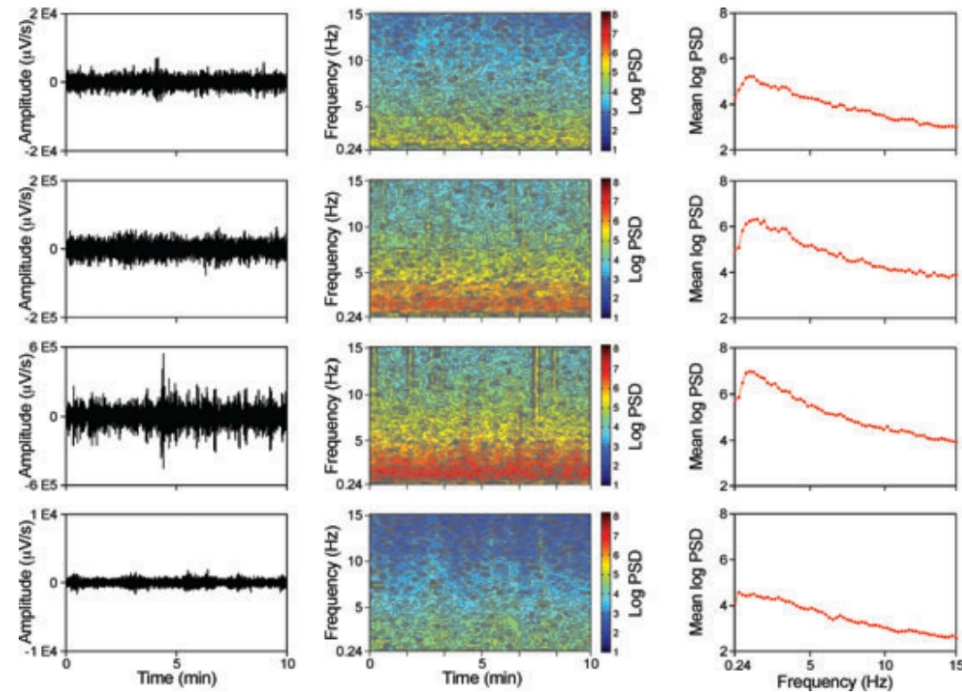
Laboratory experiments to simulate earthquakes in the Lab.



Supervised machine learning to predict failure times



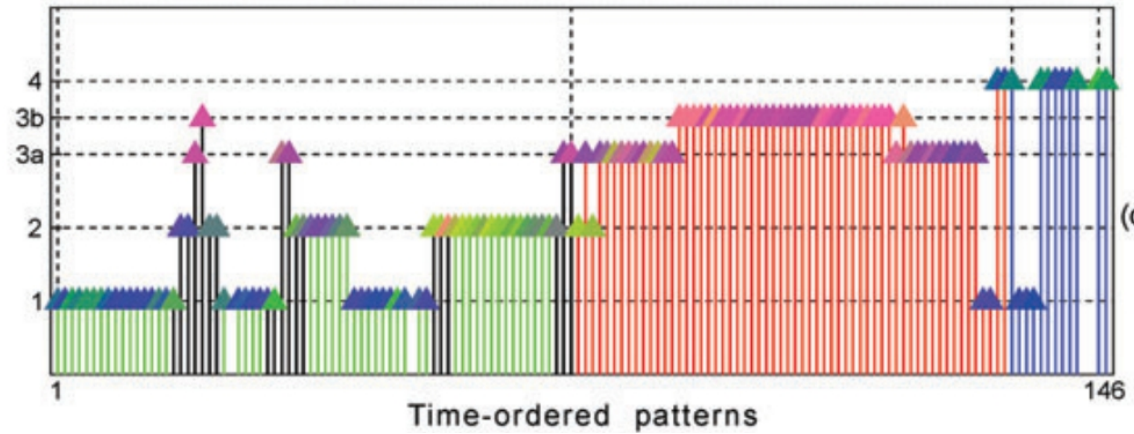
Unsupervised machine learning to identify patterns in AEs



time-series

spectrograms

feature vector



unsupervised pattern classification

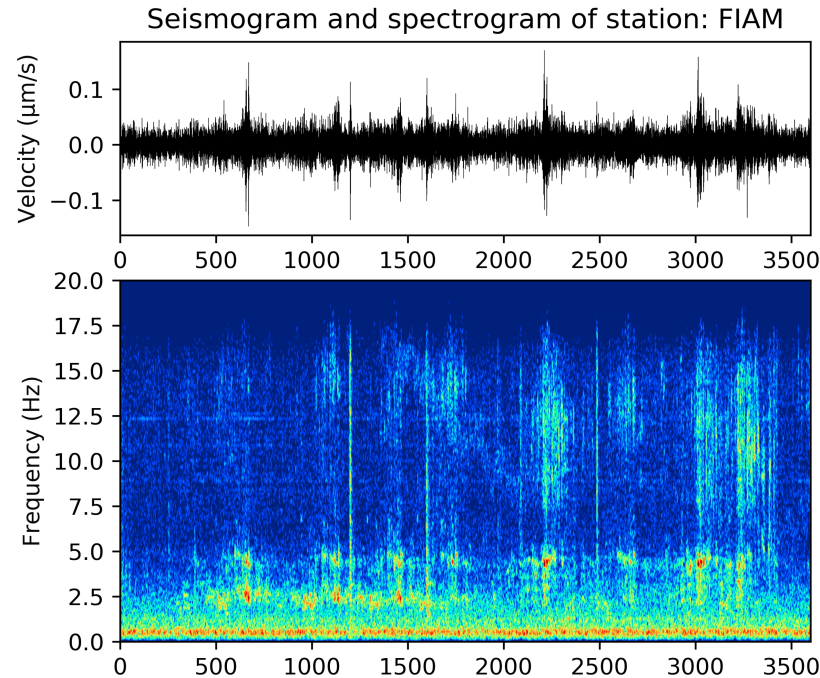
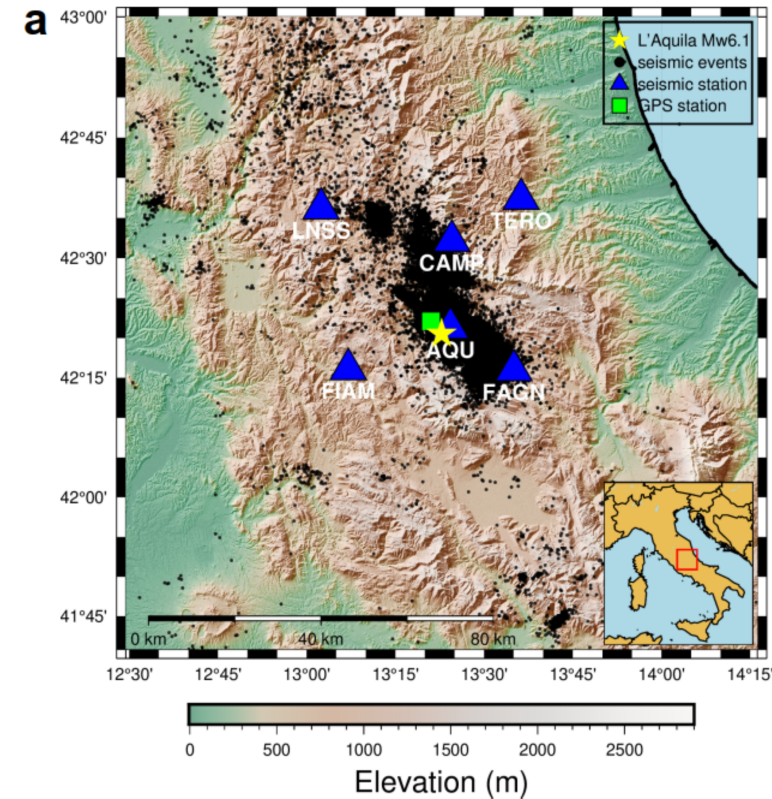
Through analyzing continuous data, one is able to identify the regimes of volcanic activity in an automatic way, which is of importance to volcano monitoring.

Different volcanic activities, e.g. **pre-eruption**, **lava fountains**, **eruption** and **post-eruption**, show seismic wavefield of distinct characteristics (especially in frequency content).

How about **earthquakes and faults**?

Introduction: utilize continuous wavefield (faults and earthquakes)

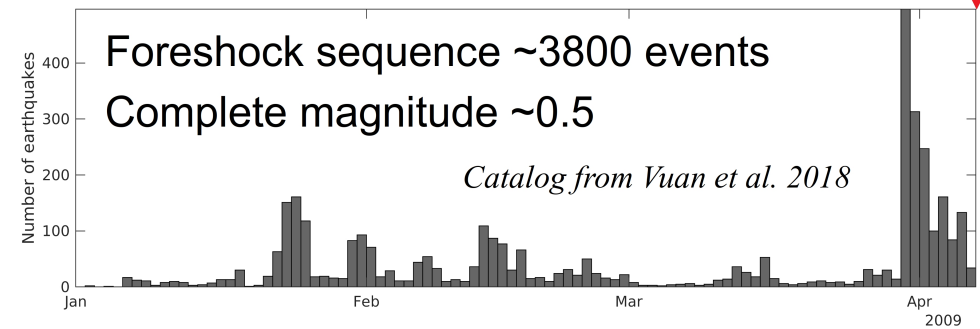
5/13



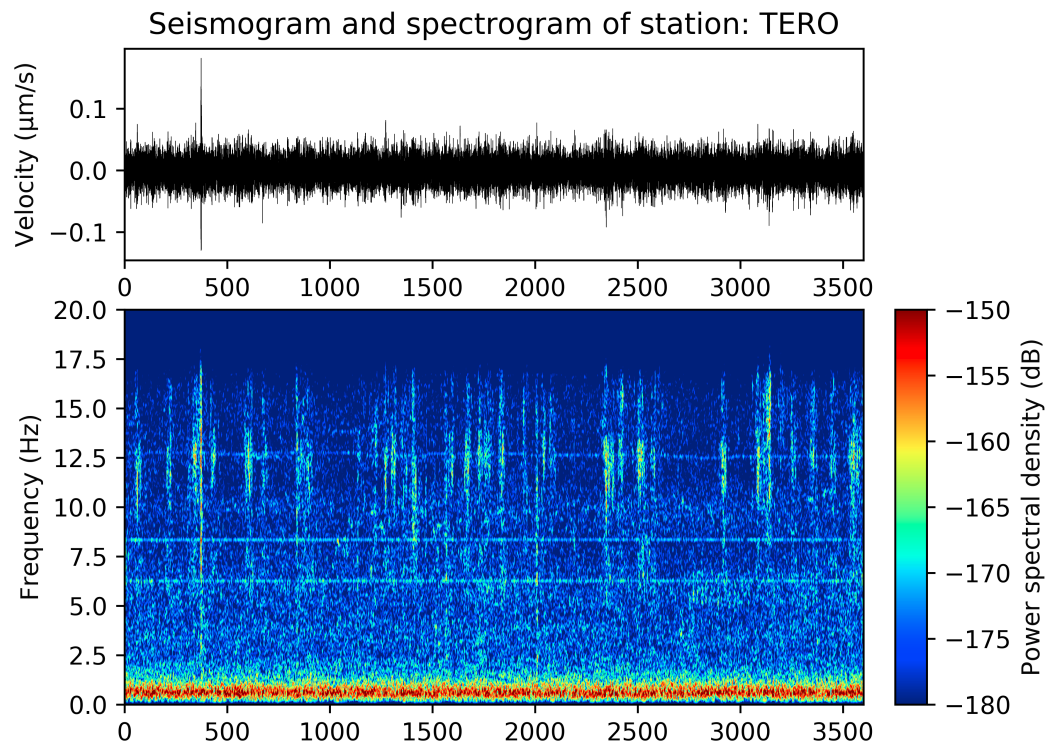
Local earthquakes;
Anomalous signals

Mw 6.1

L'Aquila region: well instrumented area
+ complex faulting processes

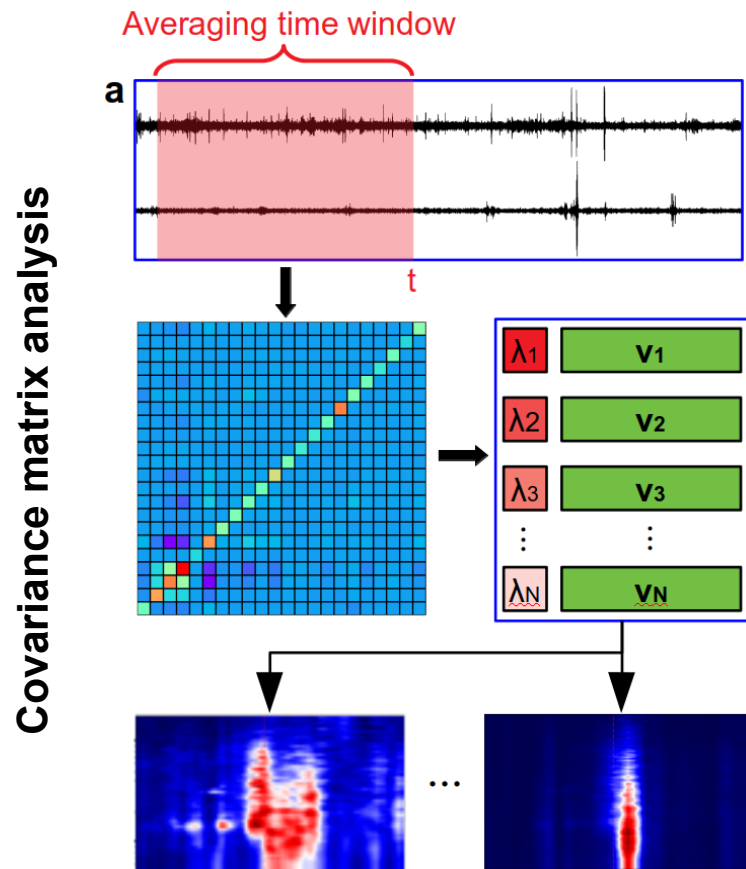


Method: extract wavefield properties/features



Signals are easier to recognize in the frequency domain.

Seydoux et. al. 2016



- Features extracted based on seismic arrays;
- Analyzed in frequency domain;
- Factorization separates independent sources;

Method: extract wavefield properties/features

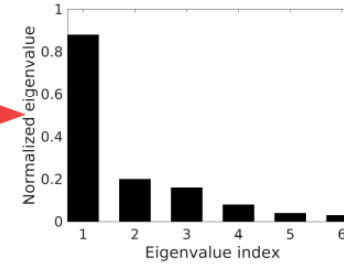
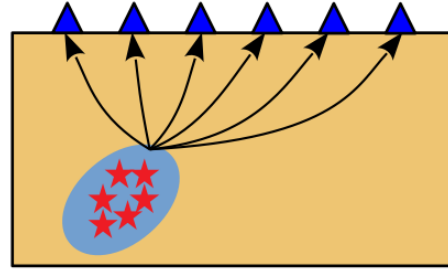
Wavefield features are extracted in a long-term averaging window of 60 days.

Wavefield coherence

Coherency

Entropy

Spectral-width



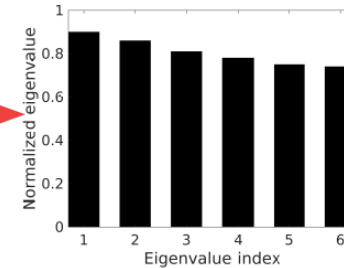
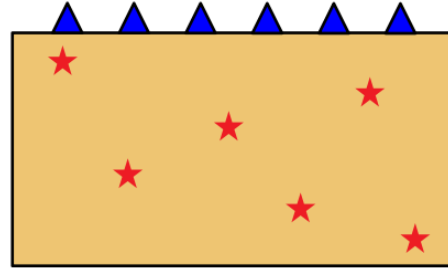
Localized sources

High coherence

High energy

Wavefield energy

First-eigenvalue



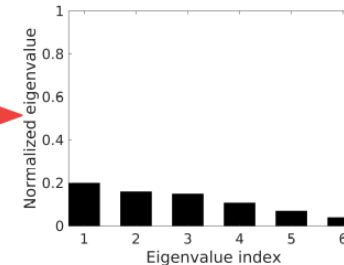
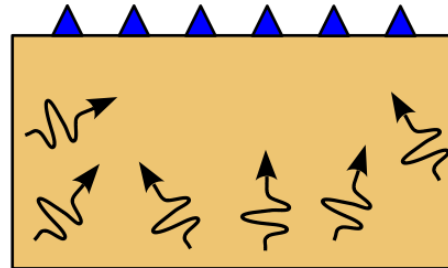
Scattered sources

Low coherence

High energy

Coherence & energy

Source variance



Noise sources

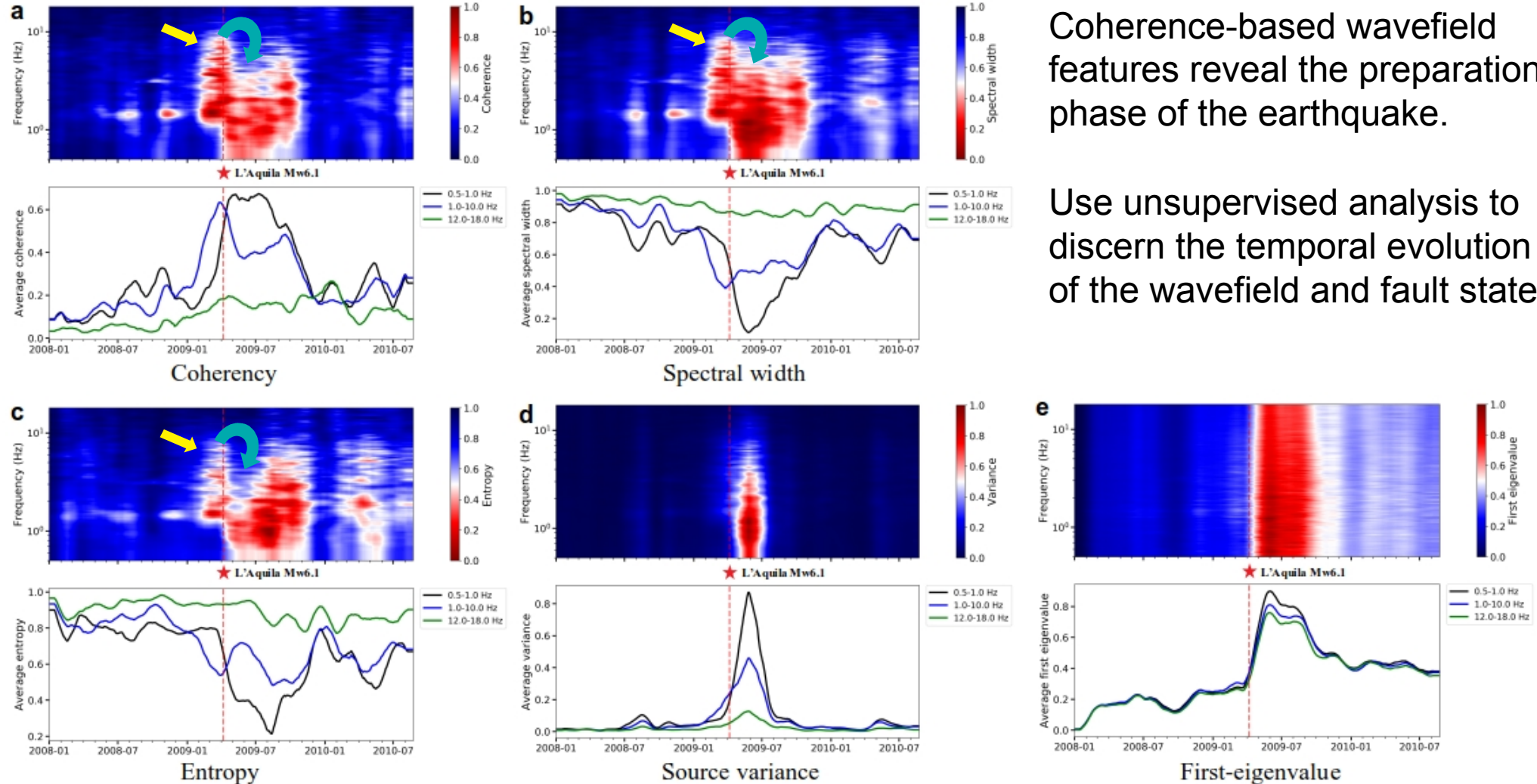
Low coherence

Low energy

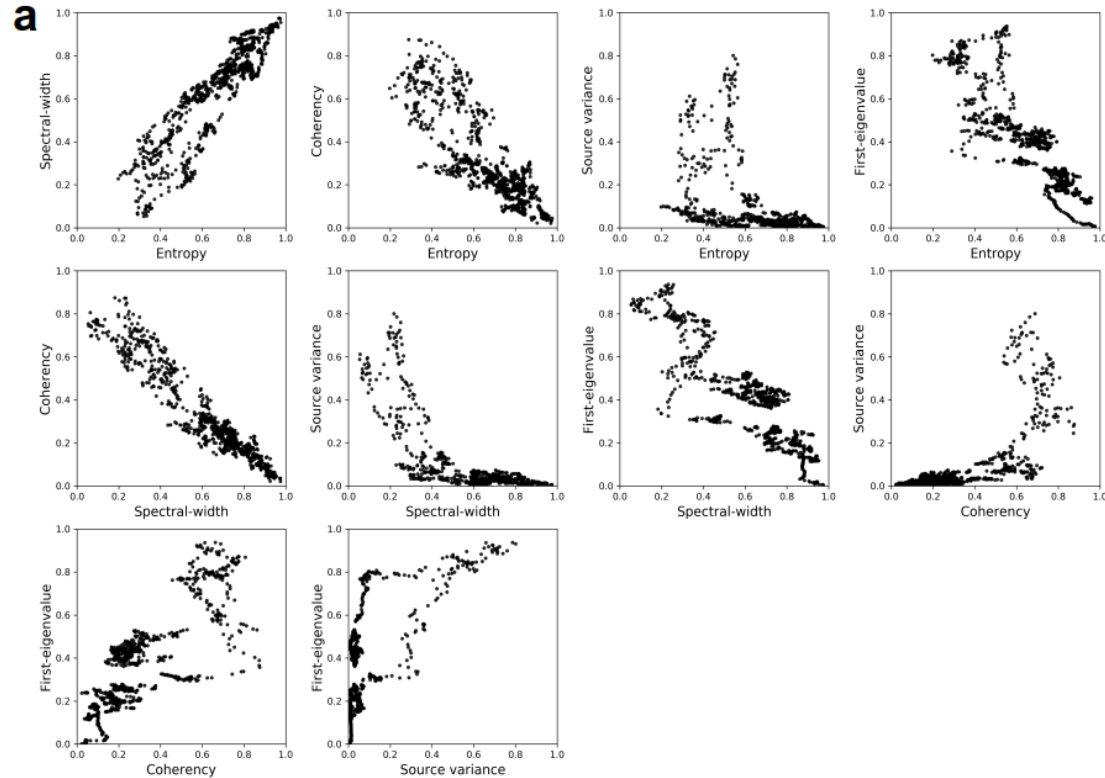
Results: wavefield properties/features

Coherence-based wavefield features reveal the preparation phase of the earthquake.

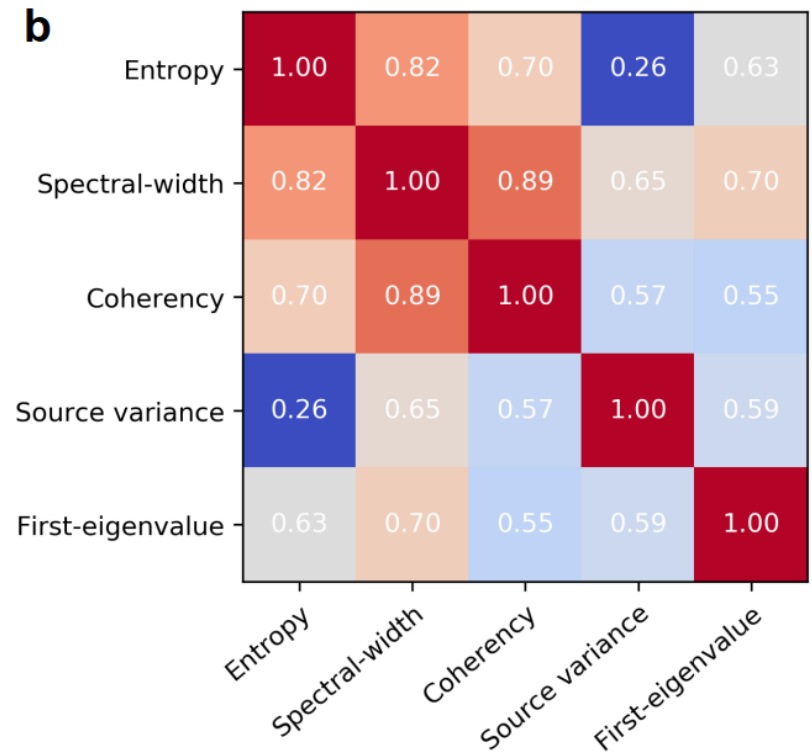
Use unsupervised analysis to discern the temporal evolution of the wavefield and fault states.



Results: correlation analysis between features

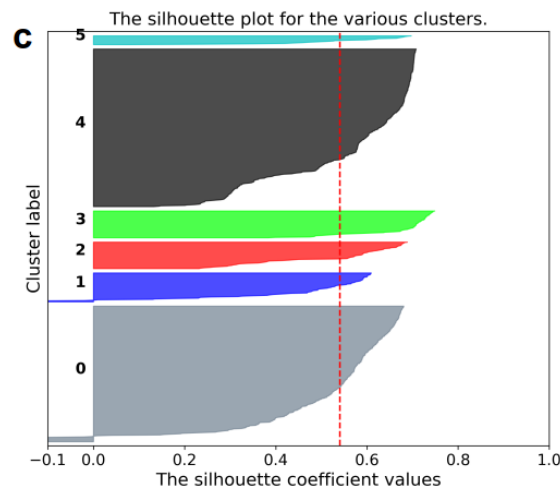
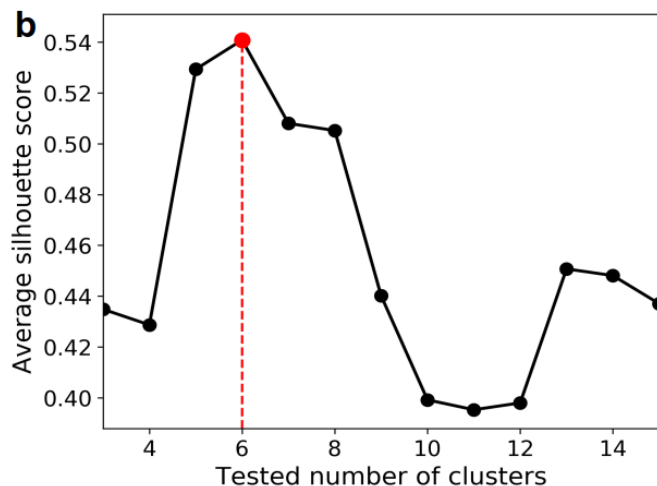
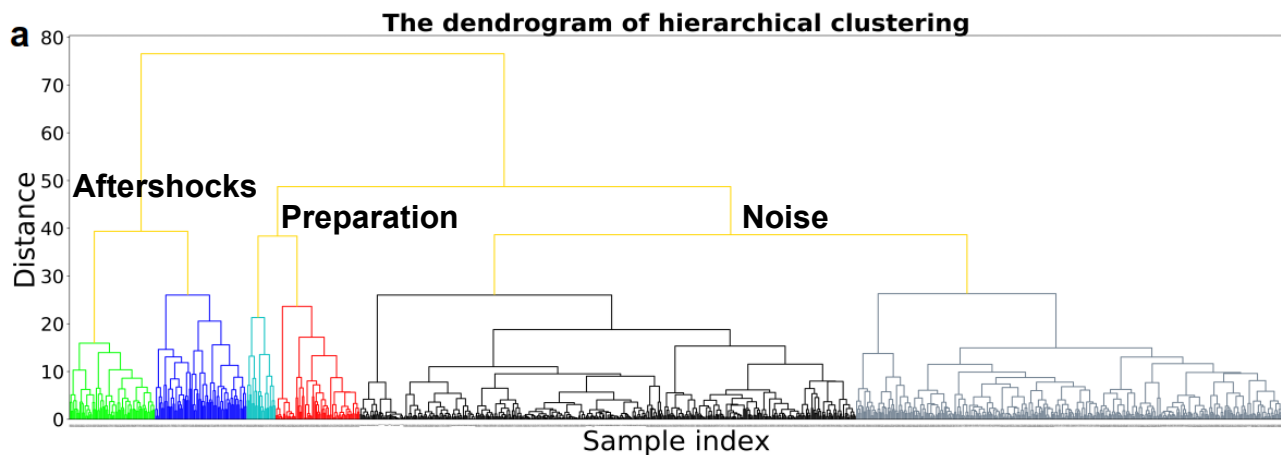


Cross-plot of different features (2-2.1 Hz)



Average cross-correlation coefficient between different features

Results: unsupervised analysis



Dimension of feature dataset:
285-D in the feature space;
966 time samples;

Different features are
linearly normalized to 0-1
before clustering.

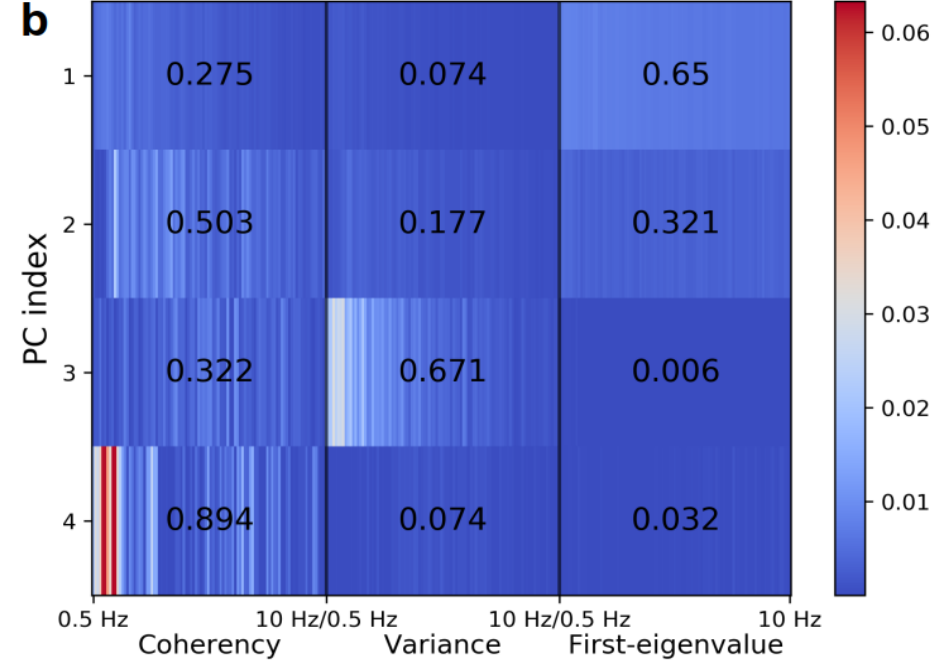
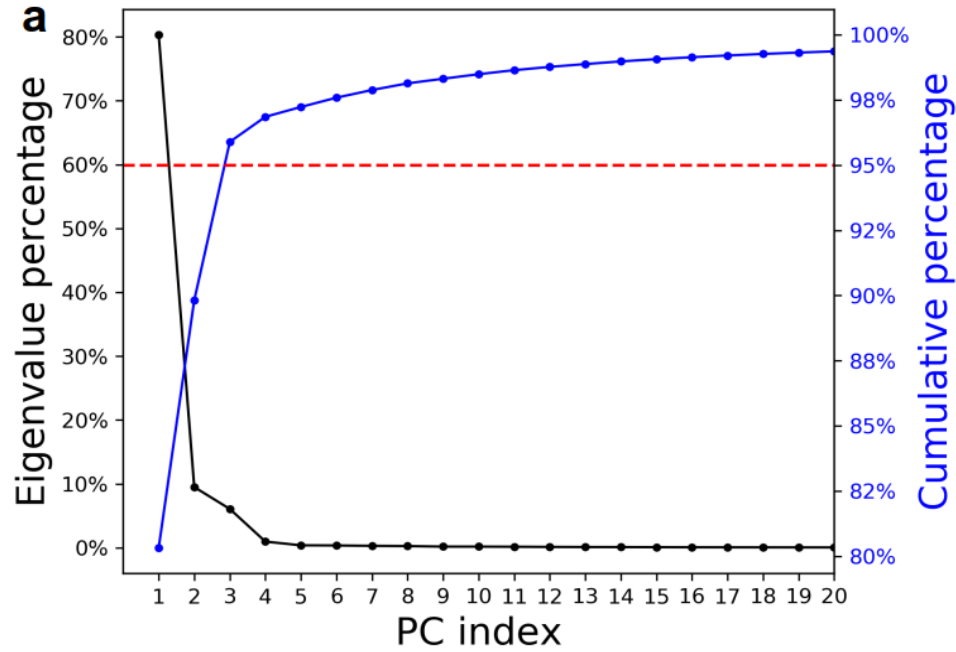
Hierarchical clustering:

- uneven cluster size;
- Non-flat geometry;
- Non-Euclidean distances

Determine the number of
clusters using **Silhouette
analysis**.

Number of clusters: 6

Results: unsupervised analysis



PCA of all the input features to (1) find suitable domain to **visualize** the clustering analysis;
 (2) identify the most **relevant** features and frequency ranges respect to different PCs.

PC1: Source energy

PC2: wavefield coherence and source localization

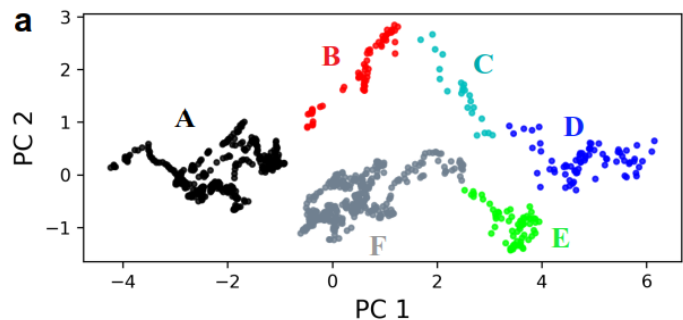
PC3: Source variance: energy + coherence

Results: unsupervised analysis

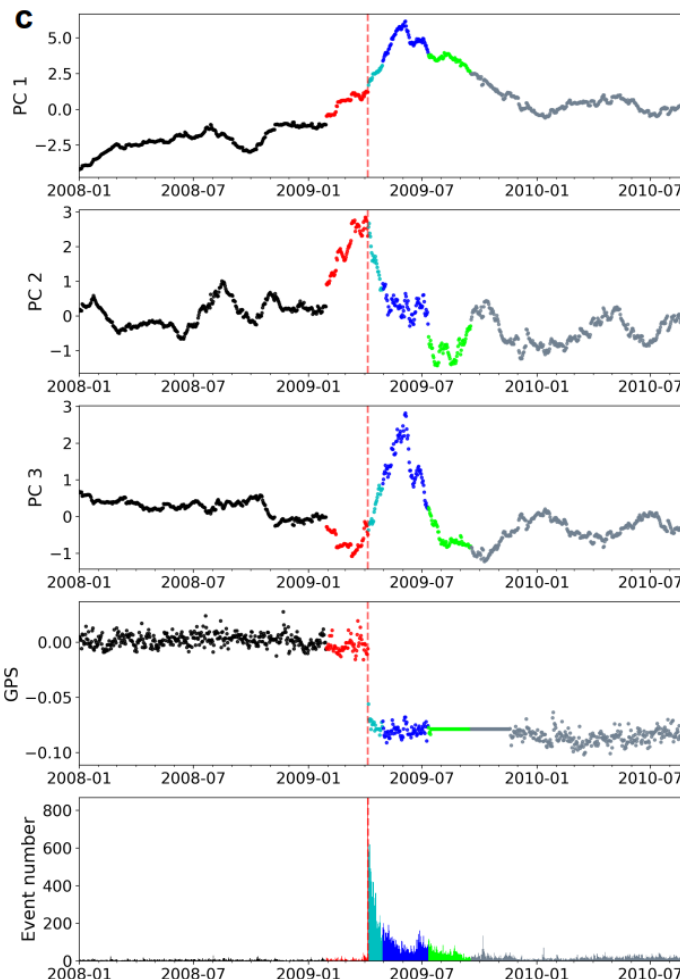
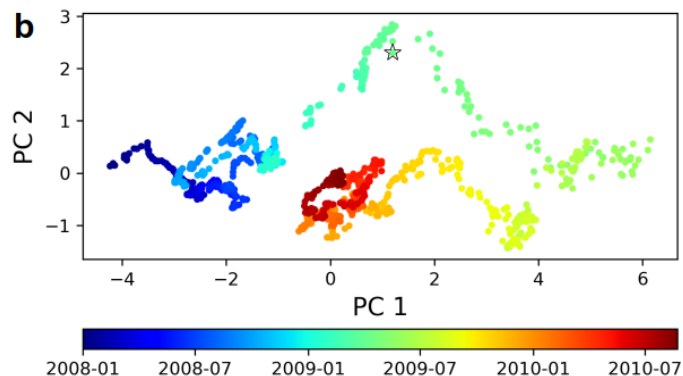
PC1: Source energy

PC2: source localization

12/13



Cluster A Cluster B Cluster C
Cluster D Cluster E Cluster F



- Clustering results in PC space clearly show there are patterns and systematic transitions in continuous wavefield.
- Temporal evolution of clusters highlights the transitional stages from a regime of fault activity to another one.

Cluster A: noise period

Cluster B: preparation phase

Cluster C: aftershock 1

Cluster D: aftershock 2

Cluster E: aftershock 3

Cluster F: recover to noise

- We have observed patterns and systematic transitions relating to the fault states from continuous seismic wavefield.
- The proposed array wavefield properties and unsupervised analysis enable us to identify the different regimes of fault activity, which can be important for hazard monitoring and fault physics studies.
- The proposed analysis method can be implemented as a powerful and complementary tool (in addition to the seismic catalog) to directly assess the fault state and track its temporal evolution in a blind way.