Instantaneous tracking of earthquake growth using prompt elasto-gravity signals (PEGS) and deep learning

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How do we model PEGS ?



Schematic representation at a time between earthquake onset and first P-wave arrival (direct elastic waves are inside the grey area)

As soon as an earthquake occurs (and thus **before**) the arrival of seismic waves), a weak signal is expected to be recorded at a broadband seismometer, due to the combination of :

- **direct effect** : the gravity perturbation induced by the earthquake rupture and the elastic waves (Harms et al. 2015, Montagner et al. 2016)
- **induced effect** : the elastic relaxation of the Earth, itself affected by the gravity perturbation (Vallée et al. 2017, Juhel et al. 2018)









the 2011 M_w 9.1 Tohoku earthquake

- Prompt elastogravity signals (PEGS) depend on :
- the earthquake focal mechanism
- the earthquake magnitude



(Vallée et al. 2017, Juhel et al. 2018)

... within the duration of the rupture !





P-wave based vs. PEGS-based EEW



Time after origin (s)



<u>P-wave based earthquake early warning</u> :

- Magnitude estimates based on 3-4 seconds of P-waves sature for large earthquakes
- System latency due to P-wave speed

PEGS based earthquake early warning :

- No saturation for large earthquakes
- Information carried at the speed of light



Objective :

assessment of PEGS potential for early magnitude estimation

PEGSNet : the training database

Few real observations of PEGS are available : training must rely on synthetic data.



⁽Licciardi et al., submitted)

- Real noise added to synthetic PEGS
- 500k synthetic earthquake sources
- Location, dip and strike from Slab2.0 (*Hayes et al. 2018*)
- M_w follows uniform distribution U [5.5, 10.0]
- STF empirical model (*Meier et al. 2017*)
- P-wave travel times assumed known





PEGSNet : architecture and learning strategy



- T₁ is randomly chosen during training.
- The value of M_w at the end of the input window is used as label.
- The model learns patterns in the data as M_w evolves with time.
- The model is designed to track the evolving magnitude and not to forecast its value.



Results on test set : predictions accuracy

Successful prediction if the estimated $M_w(t)$ lies within \pm 0.4 magnitude units from the ground truth value.



- M_w > 8.6 : moment tracking with good accuracy and low error
- $8.2 < M_w < 8.6$: early tracking more difficult, final magnitude estimation achievable
- M_w < 8.2 : poorly constrained by data, $M_w 8.3$ lower limit of PEGSNet sensitivity









Results on test set : $M_w = 9.0 \pm 0.05$



• Time-dependent performance of M_w predictions for events with true final M_w of 9 ± 0.05

• Magnitude $M_w(t)$ estimation with zero delay once $M_w > 8.3$ (t > 40 seconds)

Real data : the 2011 M_w 9.1 Tohoku earthquake

2011/03/11 05:46:24 -- Tohoku-Oki STF JMA EEW FinDer2 BEFORES This study 100 150 200

- Retrospective analysis, compared with 'true' STF and other EEWS performances.
- t > 120 s : correct prediction, when rupture is almost over.

(Licciardi et al., submitted)

• 50 < t < 100 s : tracking with slight under-estimation, with a trend suggesting rupture is in progress.

Dealing with noise

<u>Synthetic PEGS + noise from different</u> pre-event recordings

- t < 55 s : high variability due to noise
- $t > 55 s (M_w > 8.3)$: similar predictions
- PEGSNet able to generalize well to real data

<u>Pre-event noise only, no PEGS</u>

- Predicted M_w is always below model sensitivity
- $M_w = 6.5$ is a baseline value for noise

Real data : $M_w < 8$ earthquakes

2008/07/19 02:39:28 -- Off East Coast of Honshu

Conclusions

• Instantaneous tracking of moment release (no saturation, zero time delay)

- Can be combined with other observables (seismic, GNSS) to increase performance in real time
- Tohoku-oki timeliness about 50 seconds, time scale for tsunami early warning
- Applicability to $M_w > 8.3$ Japanese subduction earthquakes
- Easy to scale to different focal mechanisms and tectonic settings

Thank you

How do we model PEGS ?

Rupture and seismic wave propagation :

Initial gravity field

transient redistribution of masses

-> remote, instantaneous recording

Seismic waves arrival

the 2011 M_w 9.1 Tohoku earthquake

- Bandpass filtering : 0.002 0.03 Hz
- Criterion to evaluate data quality : \pm 0.8 nm/s² in the 30 min-long interval preceding the event
- <u>Selected broadband stations</u> : networks : IC, IU, G, F-net
- - from 400 to 3000 km
 - good azimutal coverage

Time series truncated at P-wave arrival time

PEGS observations

• Single stations or array-based observations

(Vallée and Juhel, 2019)

• Observational limit : $M_w = 7.9$

Factors controlling PEGS detectability

For a given M_w and STF, **strike-slip** and ${\bullet}$ deep earthquakes generate larger PEGS than thrust earthquakes on shallow dipping interfaces

(Vallée and Juhel, 2019)

dashed : +/- 0.4 nm/s² | dotted : +/- 1.0 nm/s²

Direct relation between STF and gravity perturbations : a rapidly growing STF increases signal observability

Elementary moment tensors

Computation of the input synthetic database

Weighted sum of 4 elementary moment tensors :

(Aki and Richards, 2002)

pure strike-slip

dotted: +/- 0.2 nm/s² dashed: +/- 0.5 nm/s² solid: +/- 1.0 nm/s² $\mathbf{M} = \cos \delta \cos \lambda \mathbf{M}^{(1)} + \sin \delta \cos \lambda \mathbf{M}^{(2)} - \cos 2\delta \sin \lambda \mathbf{M}^{(3)} + \sin 2\delta \sin \lambda \mathbf{M}^{(4)}$

$$\mathbf{M}^{(1)} = M_0 \begin{pmatrix} 0 & 0 & -\cos \phi_{\rm s} \\ 0 & 0 & -\sin \phi_{\rm s} \\ -\cos \phi_{\rm s} & -\sin \phi_{\rm s} & 0 \end{pmatrix}, \\ \mathbf{M}^{(2)} = M_0 \begin{pmatrix} -\sin 2\phi_{\rm s} & \cos 2\phi_{\rm s} & 0 \\ \cos 2\phi_{\rm s} & \sin 2\phi_{\rm s} & 0 \\ 0 & 0 & 0 \end{pmatrix} \\ \mathbf{M}^{(3)} = M_0 \begin{pmatrix} 0 & 0 & \sin \phi_{\rm s} \\ 0 & 0 & -\cos \phi_{\rm s} \\ \sin \phi_{\rm s} & -\cos \phi_{\rm s} & 0 \end{pmatrix}, \\ \mathbf{M}^{(4)} = M_0 \begin{pmatrix} -\sin^2 \phi_{\rm s} & \frac{1}{2} \sin 2\phi_{\rm s} & 0 \\ \frac{1}{2} \sin 2\phi_{\rm s} & -\cos^2 \phi_{\rm s} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Elementary moment tensors

Direct computation using Global CMT moment tensor as input

Decomposition using the best double-couple solution :

- strike = 203°
- dip = 10°
- rake = 88°

Results on test set : low noise conditions (0.5 nm/s²)

Successful prediction if the estimated $M_w(t)$ lies within \pm 0.4 magnitude units from the ground truth value.

Average residuals

• Under favorable noise conditions :

 $\sigma_{\text{noise}} < 0.5 \text{ nm/s}^2$

• $7.9 < M_w < 8.3$: final M_w prediction with 70-80% accuracy, 150 seconds from origin time.

Results on test set : $M_w = 9.0 \pm 0.05$

- Magnitude $M_w(t)$ estimation with zero delay once $M_w > 8.3$
- Ability to recover the actual moment release sooner or later, depending on the source onset

Slow onset

Real data : the 2003 M_w 8.2 Hokkaido earthquake

2003/09/25 19:50:06 -- Hokkaido

- At the edge of PEGSNet's lower sensitivity limit
- Final magnitude estimation after around 2 minutes, with expected lower accuracy and higher errors