

Global seismic monitoring as probabilistic inference Nimar Arora, Stuart Russell, Paul Kidwell, Erik Sudderth University of California Berkeley, Lawrence Livermore National Lab, Brown University.

Introduction

• The Comprehensive Nuclear-Test-Ban Treaty (CTBT) bans all nuclear explosions on Earth whether for military or for peaceful purposes.



Locations of all known nuclear explosions.

•. A sparse global network of seismic, hydroacoustic, infrasound, and radionuclide stations monitors the earth for potential violations.

• We focus here on seismic events.



Blue dots and triangles are

primary seismic stations.

The Problem

- Roughly 10000 detections a day of which 90% are spurious, i.e. small local events like passing trains, falling trees, etc.
- Many real events (i.e. magnitude 2 or higher) are not detected at all.
- Data Association Problem: What were the true events given the observed





• The current automated system (SEL3) detects 69% of real events and creates twice as many spurious (nonexistent) events.

• 16 human analysts find more events, correct existing ones, throw out spurious

- events, generate **LEB** ("ground truth")
- Unreliable below magnitude 4 (about 1 kiloton).



Yellow stars – LEB. Red circles – SEL3. Results for 1 week.



- Event magnitudes are distributed as per the Guttenberg Richter distribution (exponential distribution with rate log(10)). •Event detection probabilities depend on the station, the seismic wave type (phase), event magnitude, and distance from the event to the station.
- distributions.



- Noise detections are generated by a station-specific time-homogenous Poisson process.
- All parameters are estimated from historical training data.

Our Approach

• Event parameters – arrival time, azimuth, amplitude, etc. – have station-specific

Inference

- Given the set of detections at all the stations, we need to infer the most probable **explanation (MPE)** – a sequence of events and the association of events to detections.
- Inference works by modifying the current world through a sequence of moves which mainly focus on events and detections in the current window.



• The birth move adds new events by probabilistically "inverting" detections



• The re-associate move shuffles detections among the events.



- $D_1 \quad D_2 \quad D_3$ $P_4 D_5 D_6 D_7 D_8 D_9 D_{10} D_{11}$
- The death move kills unlikely events.



• The window moves forward, new detections are added and old events are output





Results

SEL3: Current automated seismic bulletin

- LEB: Analyst bulletin starting from SEL3.
- NET-VISA: MPE with generative model.

m _b range	SEL3			NET-VISA	
	Recall	Error (km)	Recall	Error (km)	
0 – 2	64.9	101	89.2	87	
2 – 3	50.0	186	80.6	134	
3 – 4	66.5	104	85.8	106	
> 4	86.6	70	93.9	70	

• Precision and recall computed using max-cardinality bipartite matching with LEB (assumed to be ground truth).

• Average error is the average distance between matched events.



•SEL3 extrapolation is based on scores from an SVM trained on true and false SEL3 events (Mackey, Kleiner, and Jordan . AGU 2009)

• Results are based on a 3 month dataset of which 1 week was used for validation

• LEB is not perfect, the table below shows the performance of LEB and NET-VISA for prediction on the continental United States with the USGS as ground truth.

LE	B	NET-VISA		
Precision	Recall	Precision	Recall	
4/4	4/33	9/51	9/33	

Conclusions

- NET-VISA is twice as effective as SEL3.
- Under consideration for deployment by CTBTO.
- Next step: SIG-VISA extends generative model down to signal level.

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