

# Weighted RANdom sampling in Seismic Event Detection/Location (WRASED): Applications to Local, Regional and Global Seismic Networks

Lijun Zhu<sup>1</sup>, Zefeng Li<sup>2</sup>, Zhigang Peng<sup>2</sup>, Entao Liu<sup>1</sup>, and James McClellan<sup>1</sup>

<sup>1</sup>Center of Energy and Geo Processing at Georgia Institute of Technology,

<sup>2</sup>School of Earth and Atmospheric Sciences at Georgia Institute of Technology

April 19, 2017



Georgia Institute  
of Technology

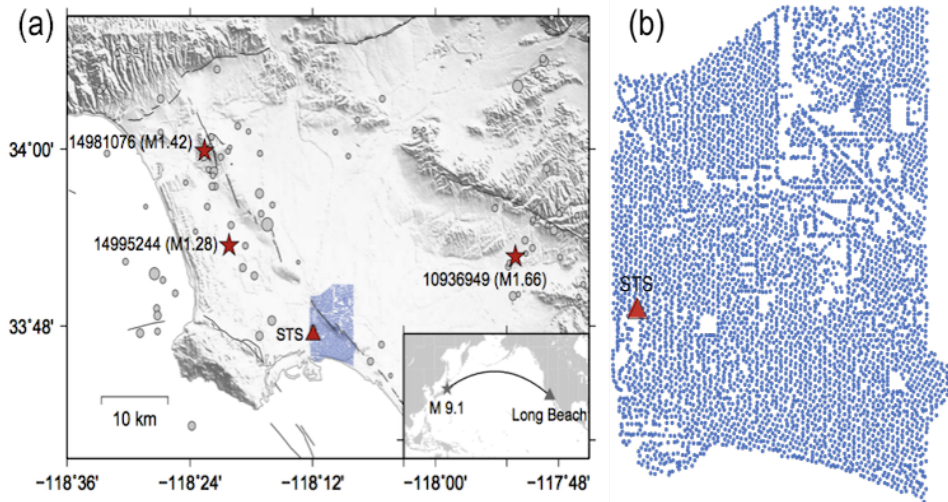
Center for Energy & Geo Processing



- 1 Introduction
- 2 RANdom SAmpling Consensus (RANSAC)
- 3 Local Coherence based Weighting
- 4 Future Work

- 1 Introduction
- 2 RANdom SAmpling Consensus (RANSAC)
- 3 Local Coherence based Weighting
- 4 Future Work

# Continuous Seismic Recording on a 5200-element Long Beach Nodal Array [Inbal et al., 2016, Li et al., 2017]





# Recent Study of Event Detection

- Stacking of *cross-correlation* between adjacent stations results in local coherence measure [Li et al., 2017]
- Event detection is significantly improved by local coherence
- Rearranging traces according to receiver locations results in *picks* forming a **moveout surface**

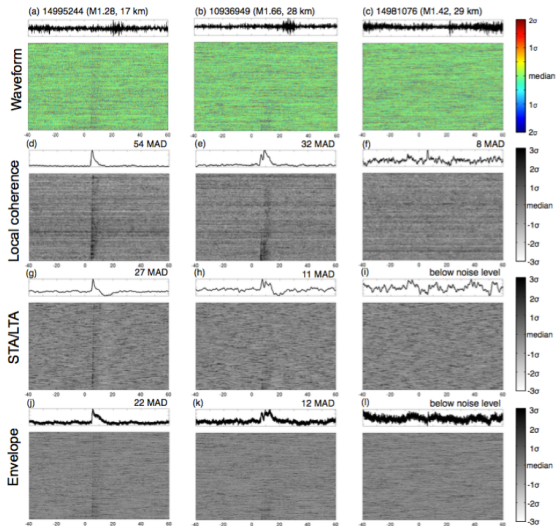
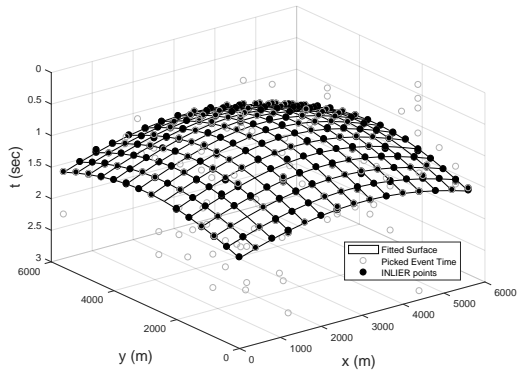


Figure 2: Results from the study in 2016

# Challenges for Processing Dense Array Data



**Figure 3:** Picks from an event form a hyperbolic surface.

# Challenges for Processing Dense Array Data

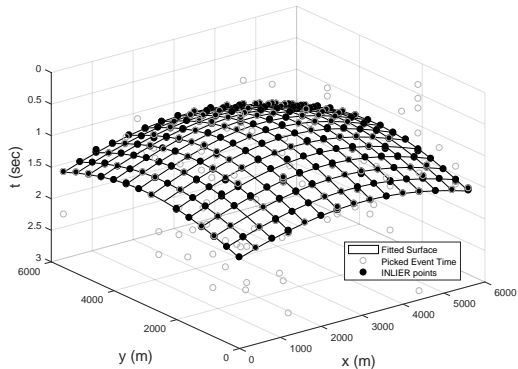


Figure 3: Picks from an event form a hyperbolic surface.

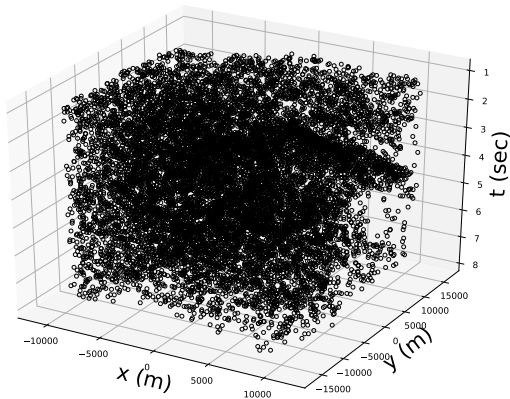


Figure 4: Picks from noisy field data.

# Challenges for Processing Dense Array Data

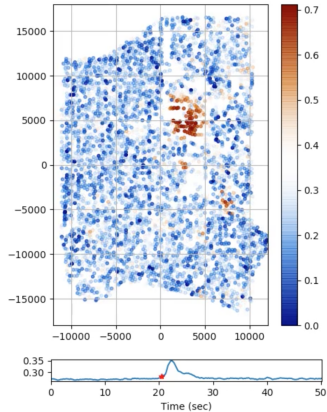


Figure 5: Wave propagation received on surface array.

# Challenges for Processing Dense Array Data

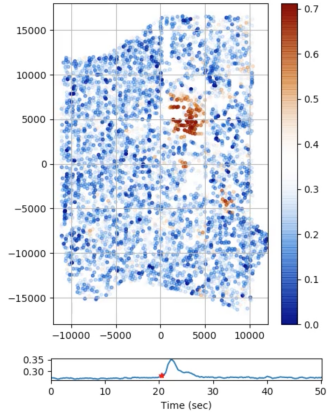


Figure 5: Wave propagation received on surface array.

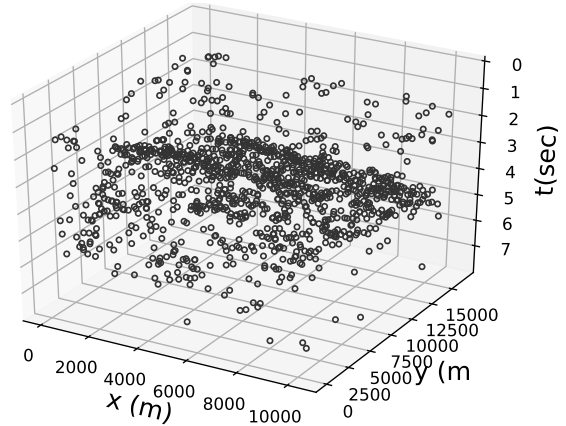


Figure 6: Picked arrival times on northeast quadrant of the surface array.

## Two Challenges in Processing Dense Array

- 1 Associate picks from the same event and eliminate false picks
- 2 Isolate receivers that are event dominant (Good SNR)

# Outline

- 1 Introduction
- 2 RANdom SAmpling Consensus (RANSAC)
- 3 Local Coherence based Weighting
- 4 Future Work

# Eliminate False Picks by RANSAC-based Curve Fitting

- Fit moveout curve to time picks
- **Robust** in the presence of many outliers [Fischler and Bolles, 1981]
- Hypothesize-and-test strategy [Zhu et al., 2016]
- Computationally efficient
  - two parameters for line
  - five for hyperbola
  - nine for hyperbolic surface

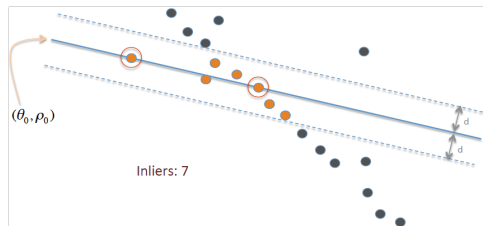


Figure 7: Illustration of RANSAC for line fitting (downloaded from Wikipedia).



# Synthetic Examples

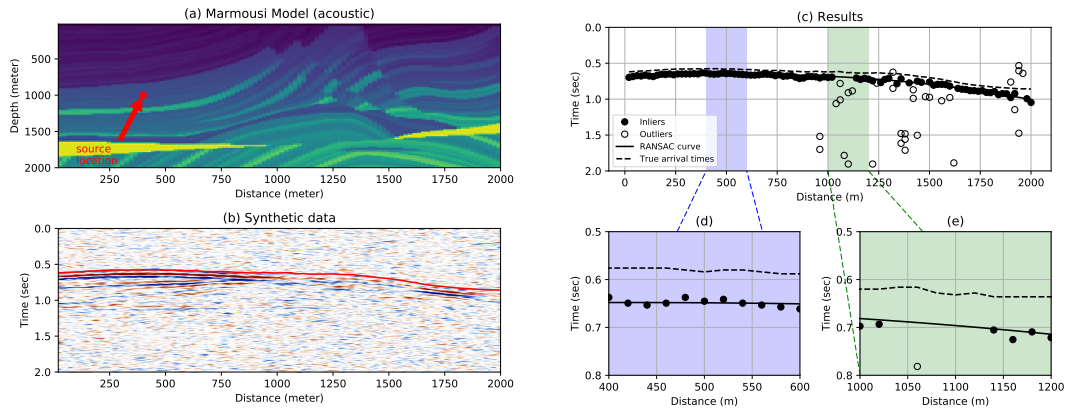


Figure 8: Synthetic example for non-layered medium.

## RANSAC Fitting Results on Northeast Quadrant

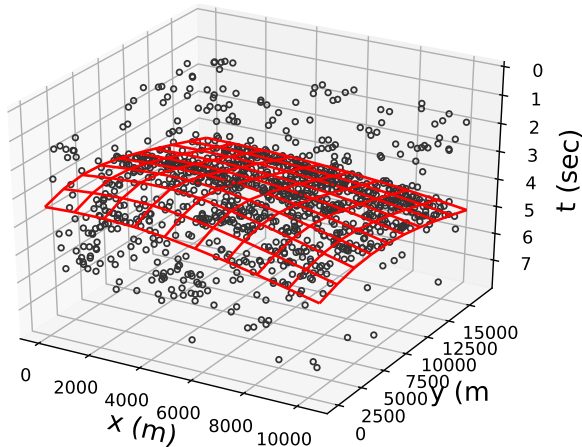


Figure 9: 3-D view of the picks ( $\circ$ ) from the 2-D sensor array with fitted hyperboloid surface in red.

# Outline

- 1 Introduction
- 2 RANdom SAmpling Consensus (RANSAC)
- 3 Local Coherence based Weighting**
- 4 Future Work

# Map-views of Max Value on Traces

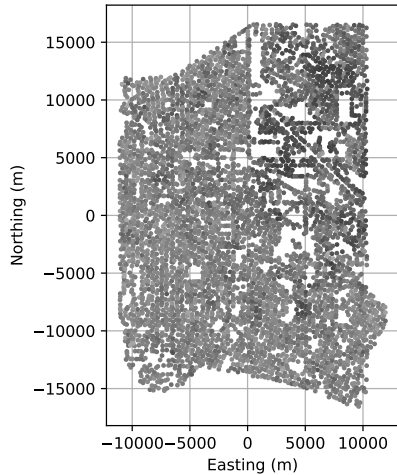


Figure 10: Max coherence value on traces.

# Map-views of Max Value on Traces

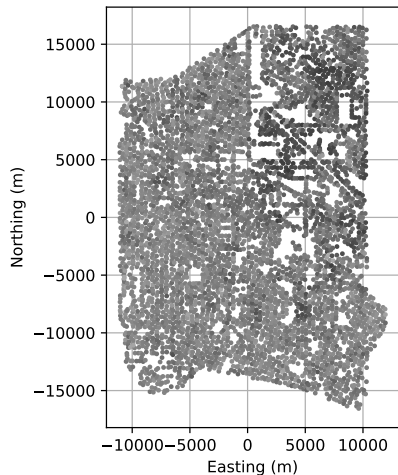


Figure 10: Max coherence value on traces.

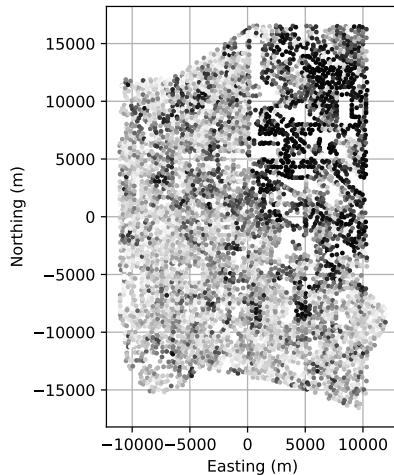


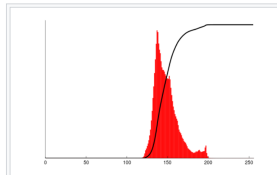
Figure 11: Scaled max coherence value on traces.

# Scale Local Coherence as Weighting Function

- Flatten local coherence by histogram equalization
- Soft-thresholding by logistic function



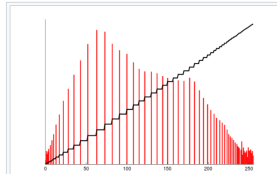
An unequalized image



Corresponding histogram (red) and cumulative histogram (black)



The same image after histogram equalization



Corresponding histogram (red) and cumulative histogram (black)

**Figure 12:** Example demonstrating histogram equalization for a natural image.

# Scale Local Coherence as Weighting Function

- Flatten local coherence by histogram equalization
- Soft-thresholding by logistic function

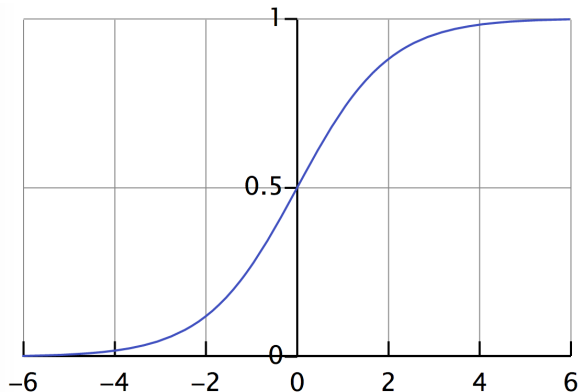


Figure 13: Logistic function centered at zero.

# Weighting on Time Picks

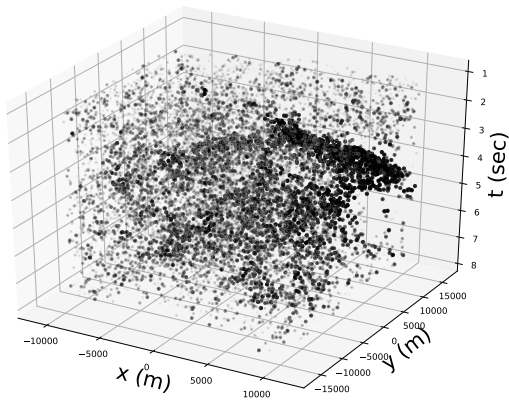


Figure 14: Weighted time picks



# Weighting on Time Picks

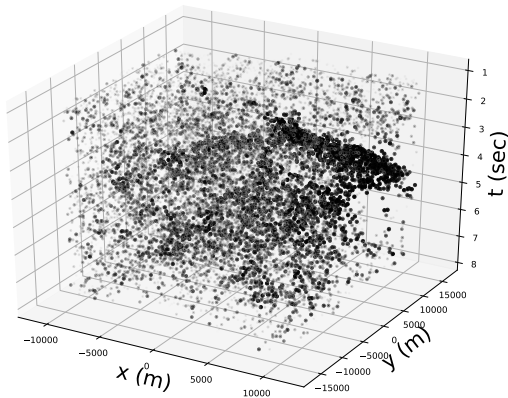


Figure 14: Weighted time picks

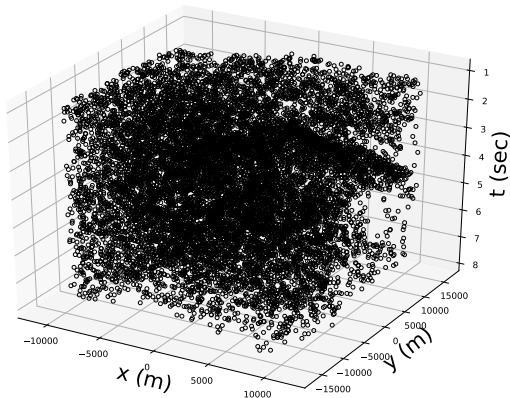


Figure 15: Original time picks

# Event Location Results

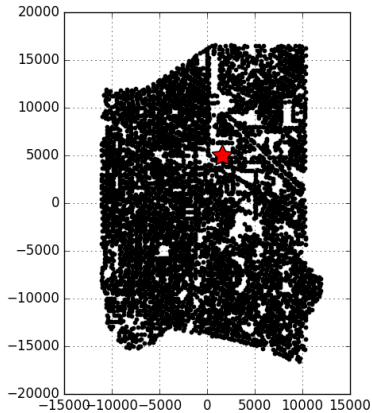


Figure 16: Event location estimated from weighted time picks.

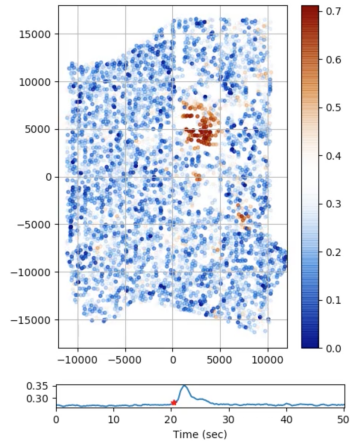


Figure 17: Wave propagation received.





# Outline

- 1 Introduction
- 2 RANdom SAmpling Consensus (RANSAC)
- 3 Local Coherence based Weighting
- 4 Future Work

- Fit P- and S-wave travel time difference using parabola to improve event depth estimation
  - Useful strategy for borehole arrays
  - Fewer parameters (three) for faster computation

- Fit P- and S-wave travel time difference using parabola to improve event depth estimation
  - Useful strategy for borehole arrays
  - Fewer parameters (three) for faster computation
- Extend the current scheme to regional network using the Earth-flattening transformation

# References

-  Fischler, M. A., and R. C. Bolles, 1981, Random sample consensus: a paradigm for model fitting with applications to image analysis and automated cartography: Communications of the ACM, **24**, 381–395.
-  Inbal, A., J. P. Ampuero, and R. W. Clayton, 2016, Localized seismic deformation in the upper mantle revealed by dense seismic arrays: Science, **354**, 88–92.
-  Li, Z., Z. Peng, and D. Hollis, 2017, High-resolution seismic event detection using local coherence for large-n arrays: Scientific Report., in revision.
-  Zhu, L., E. Liu, and J. H. McClellan, 2016, *in* An Automatic Arrival Time Picking Method Based on RANSAC Curve Fitting: EAGE Annual meeting 2016.