

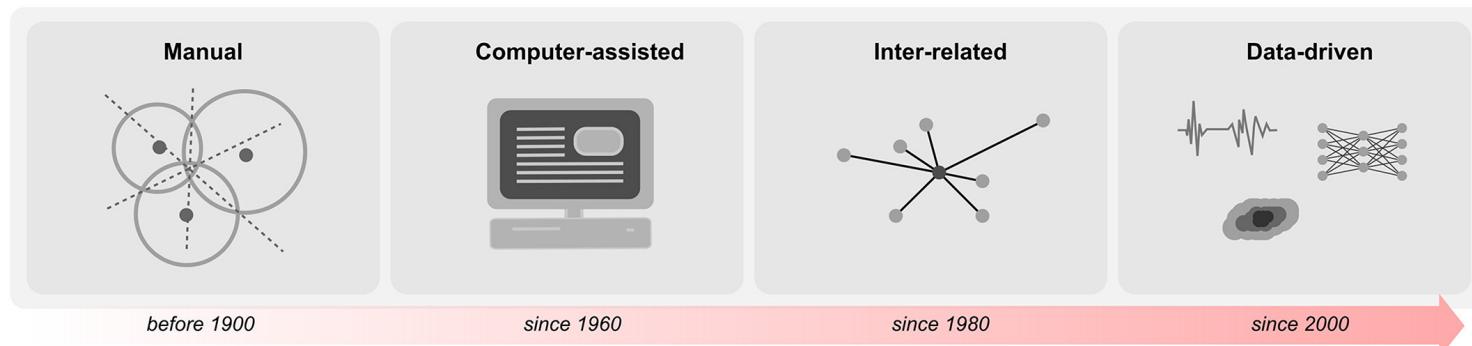
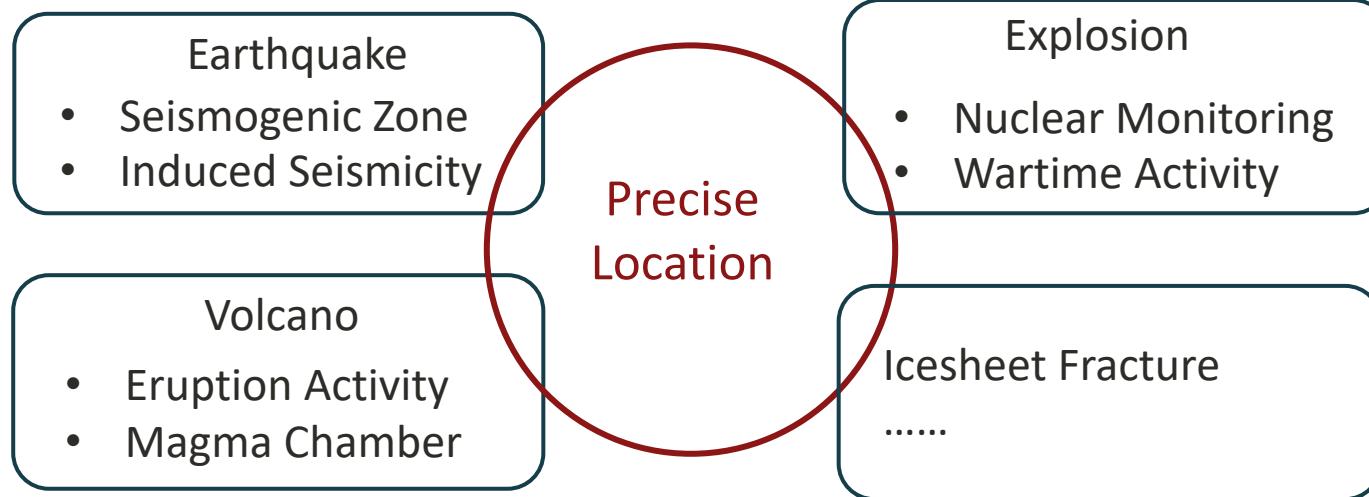
Accuracy and Precision of Earthquake Locators: Insights from a Synthetic 2019 Ridgecrest Sequence Experiment

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Jan. 17 2024

Introduction – Earthquake Location

Fundamental but Challenging Topic in Seismology

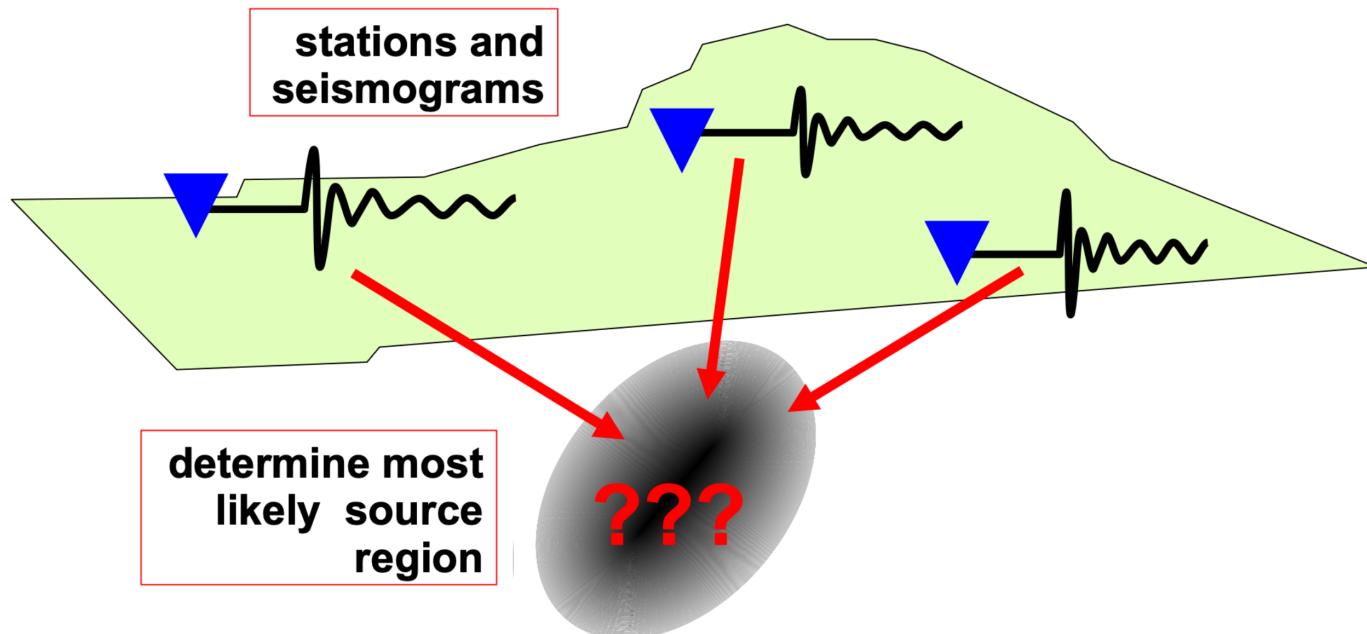


Li et al., 2020, Reviews of Geophysics

Introduction – Earthquake Location

A Well Defined Nonlinear Inverse Problem

$$t_{arr} = t_0 + \int_r \frac{1}{v(s)} ds \quad d_{obs} = (t_{arr}^1, \dots, t_{arr}^n) = G(m) = G([x, y, z, t_0])$$



From: NonLinLoc

Introduction – Hypocenter Locators

Hypoinverse

An iterative linearized inversion scheme

$$R = G * dM$$

$$dM = (G^T G)^{-1} G^T R$$

VELEST

$$t_{res} = H * h + M * m + e = Gm + e$$

Constant Error

Hypocenter Velocity Model

Non_Lin_Loc

Nonlinear inversion: Grid Search

EDT (Equal Differential Time)

$$pdf(x) \propto \sum_{obs_a, obs_b} e^{-\frac{[(T_{obs_a}(x) - T_{obs_b}(x)) - (TT_{calc_a}(x) - TT_{calc_b}(x))]^2}{\sigma^2}}$$

“satisfy **the most pairs** of observations”
• no origin time estimate required

HypoSVI

Machine Learning model!
Grid search with ML travelttime calculator

Introduction – Hypocenter Locators

HypoDD

$$dr_k^{ij} = (t_k^i - t_k^j)^{obs} - (t_k^i - t_k^j)^{cal}$$

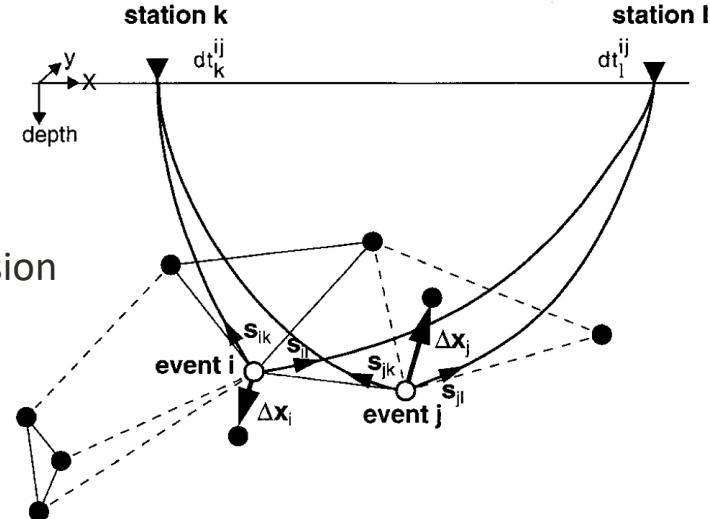
$$\frac{\partial t_k^i}{\partial \mathbf{m}} \Delta \mathbf{m}^i - \frac{\partial t_k^i}{\partial \mathbf{m}} \Delta \mathbf{m}^j = dr_k^{ij}$$

Growclust

L1 Norm, more resilient to outliers
Traveltime table search instead of matrix inversion

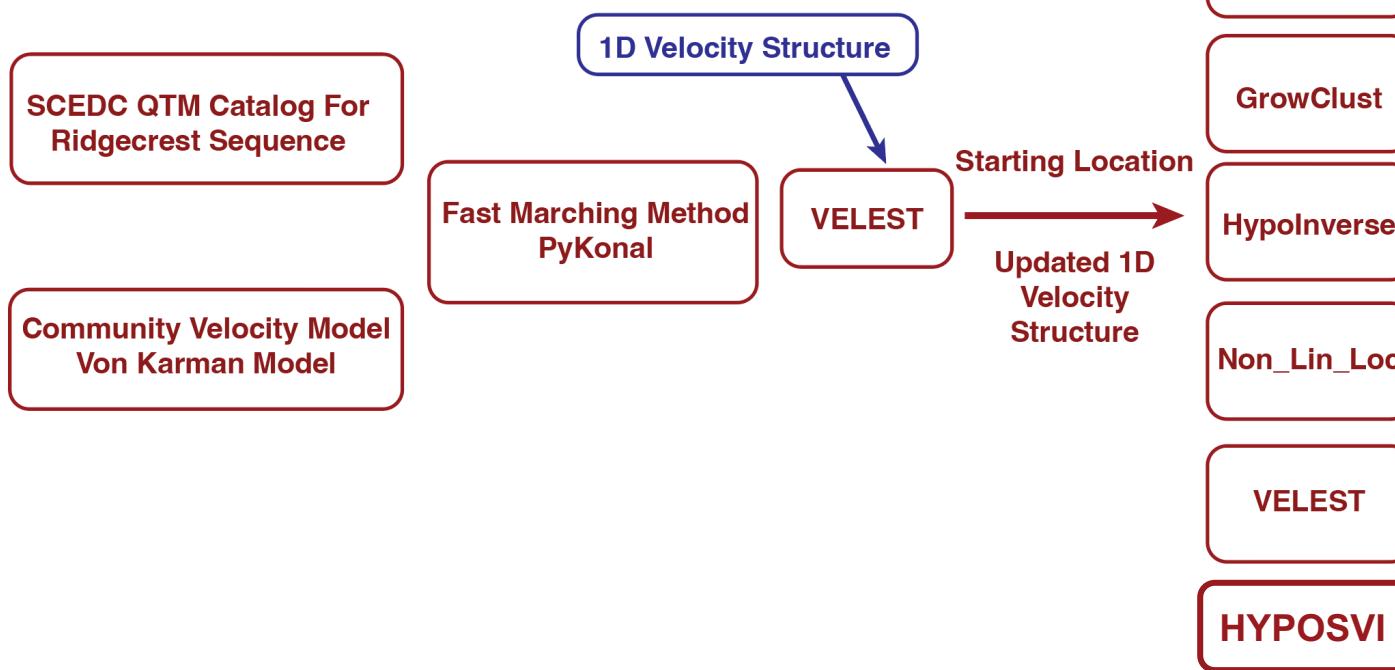
XCORLOC

Similar to Growclust
But with Source Station Specific Term



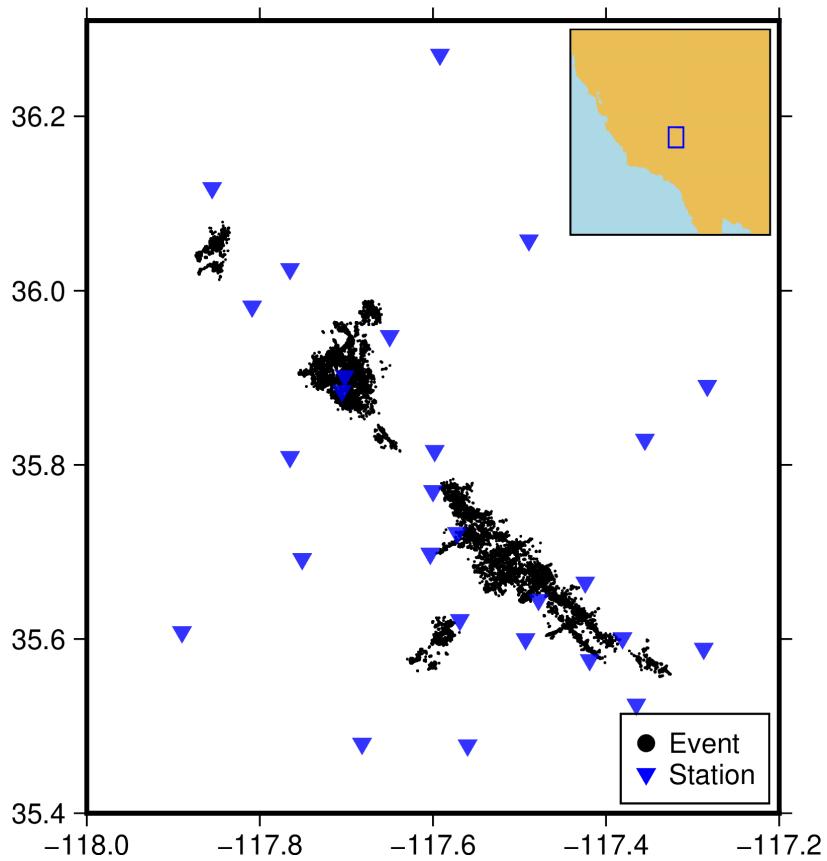
How to compare their performance?

Compare the seven location methods on a realistic synthetic dataset



Setup – Earthquake Location

2019 Ridgecrest Earthquake Sequence – Data Mine



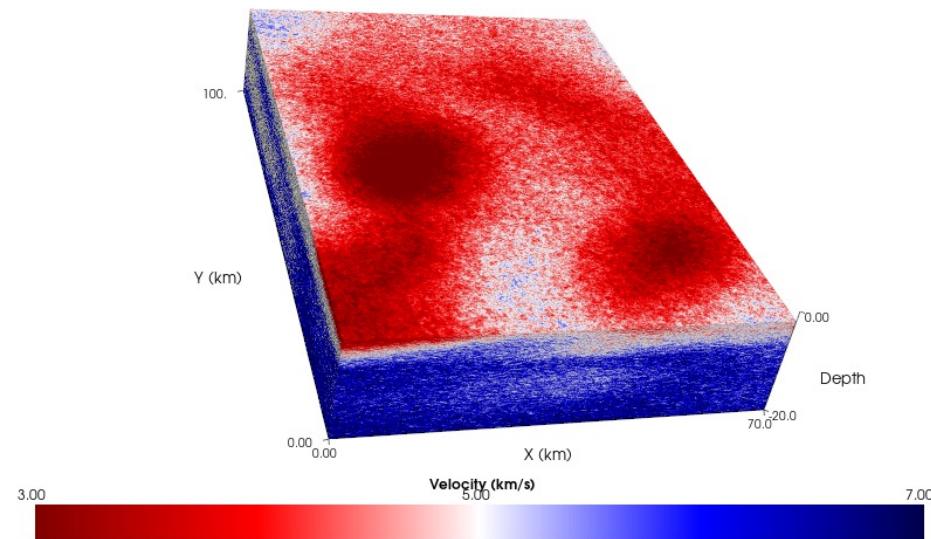
SCEDC QTM Catalog For Ridgecrest

We randomly draw 1000 sources out of the catalog (cluster > 400) as true locations

Real stations distribution from SCEDC

Setup – Velocity Structure

3D, Heterogeneous



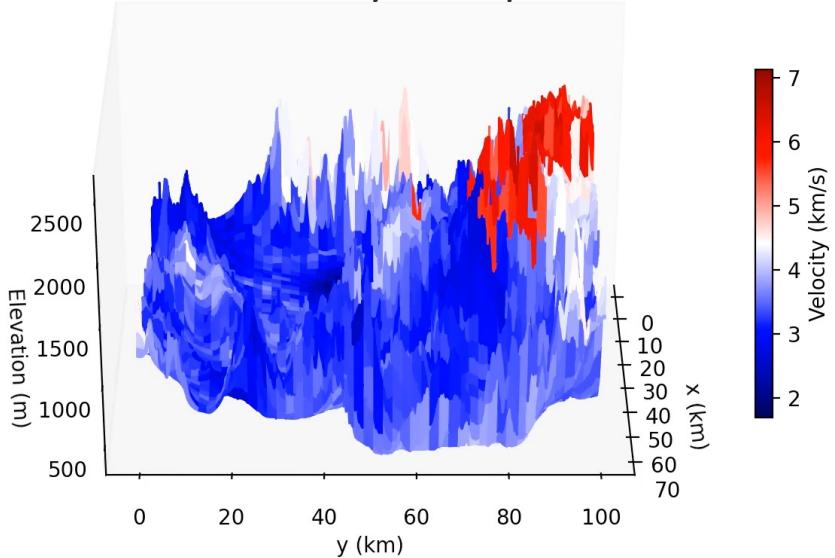
- SCEC Community Velocity
- Von Karman Perturbation
($a_z=100\text{m}$, $a_h=500\text{m}$, $\epsilon = 0.05$)
- $70\text{km} \times 100\text{km} \times 20\text{km}$ depth +
(max. 2km elevation)
($\text{dx}=\text{dy}=100\text{m}$, $\text{dz}=50\text{m}$)

Setup – Elevation Effect

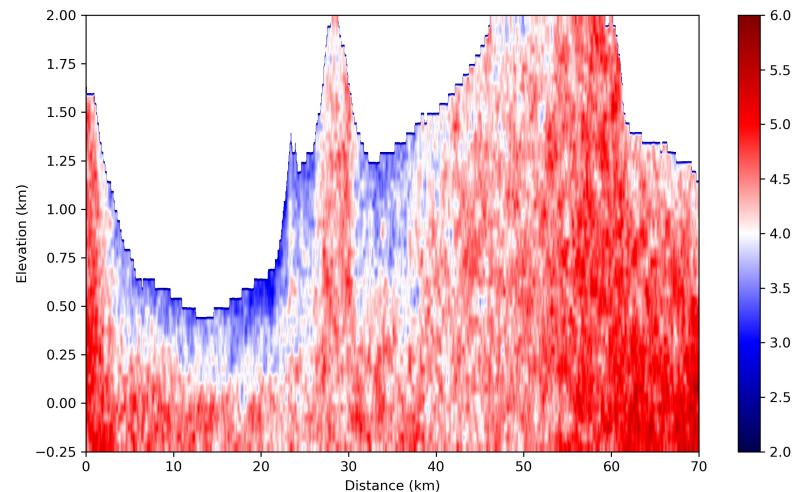
Mountains and basins could impact the location!

1. Classify as Granite or Sediments
2. Velocity decrease with elevation in different slope based on empirical relations from experiment (Brocher, 2008)

Elevation Velocity example



Velocity profile example



Setup – Travelttime Calculator

Eikonal Function Based
Fast Marching Method: **PyKonal**

RESEARCH ARTICLE | JUNE 03, 2020

PyKonal: A Python Package for Solving the Eikonal Equation in Spherical and Cartesian Coordinates Using the Fast Marching Method 

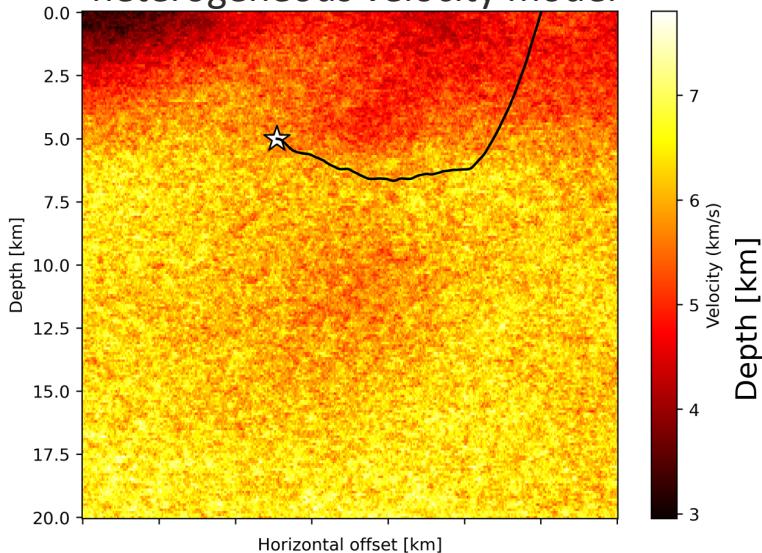
Malcolm C. A. White ; Hongjian Fang; Nori Nakata; Yehuda Ben-Zion

+ Author and Article Information

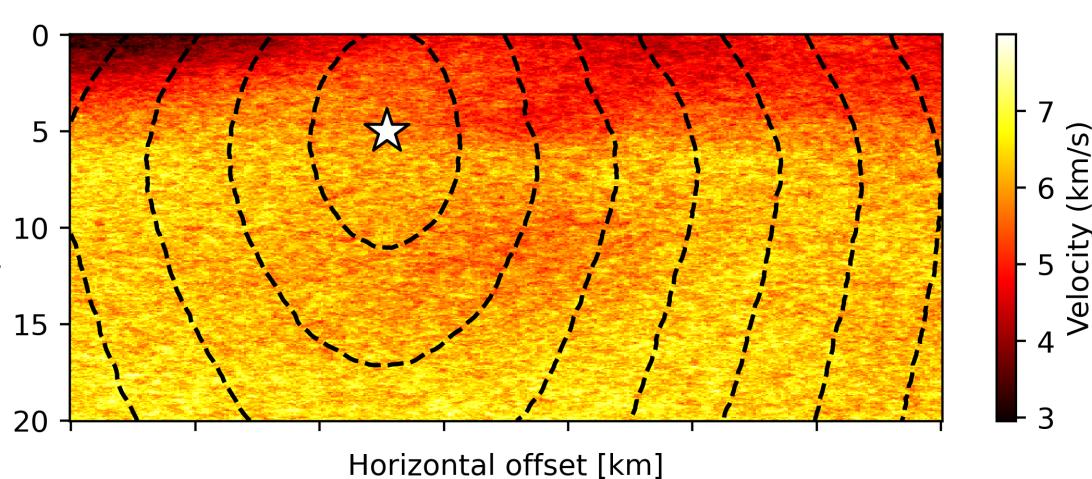
Seismological Research Letters (2020) 91 (4): 2378–2389.

<https://doi.org/10.1785/0220190318> | Article history 

Ray tracing verification on heterogeneous velocity model

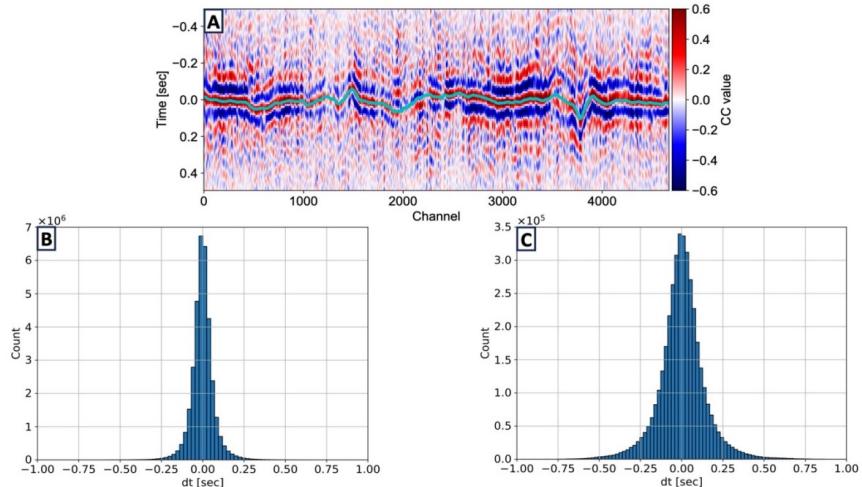


Travelttime contour verification



Setup – Realistic Error

The observation is not perfect.

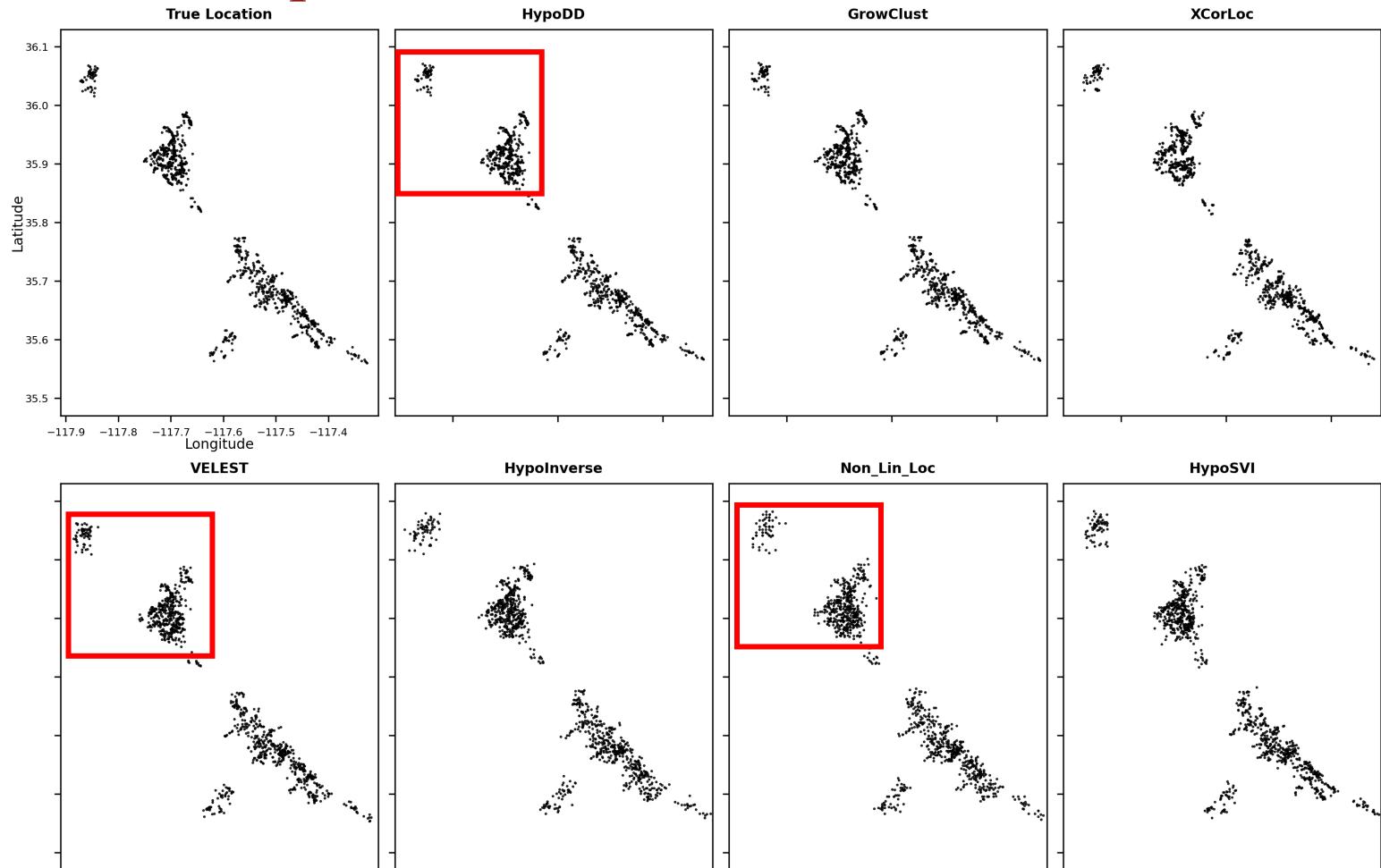


- Realistic traveltimes reading error in Laplacian distribution
- Phase availability for each stations in probability: 0.67 for P, 0.5 for S
- 1% P phase and 4 % S phase outliers exist

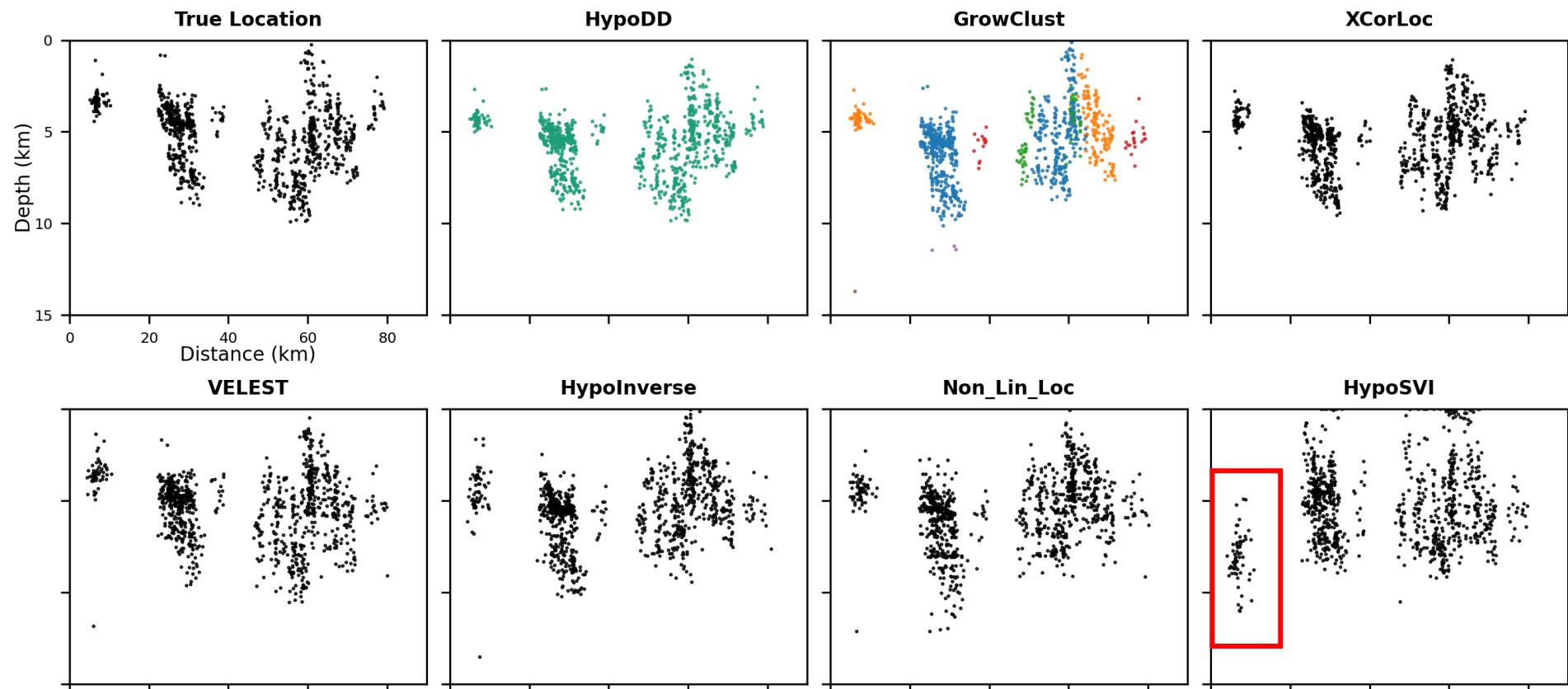
Biondi et al., 2023

While both cross correlation and direct arrival phases error can be considered in Laplacian distributions, CC is more accurate than direct. And P phase is more accurate than S.

Results – Mapview



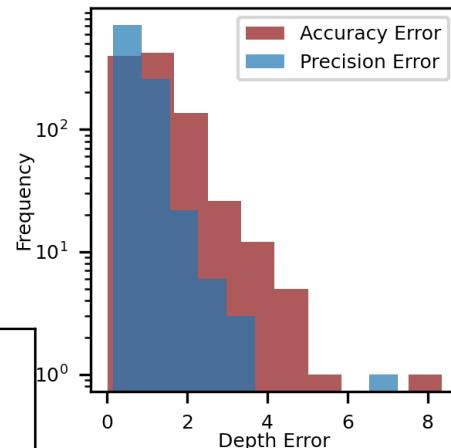
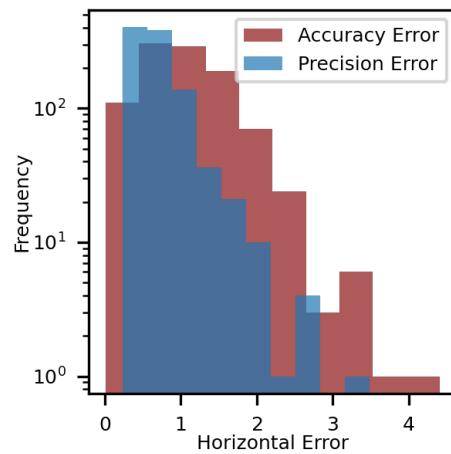
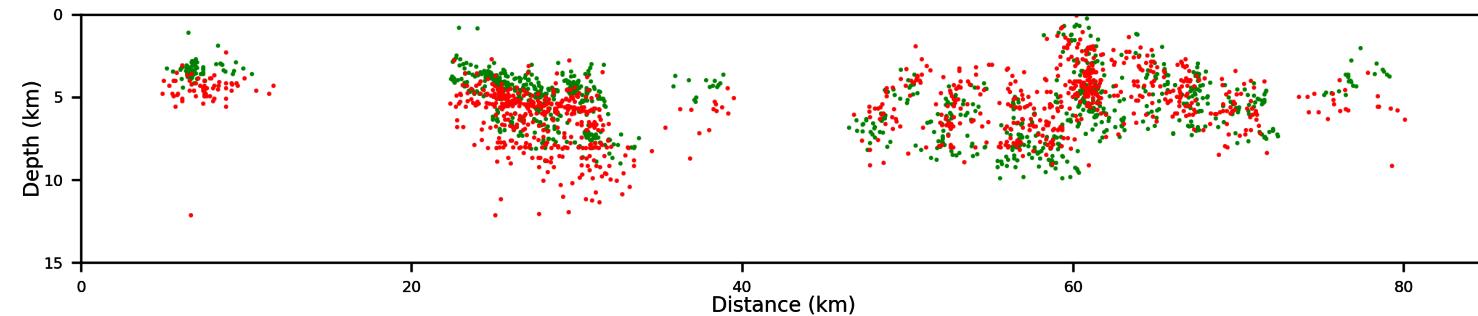
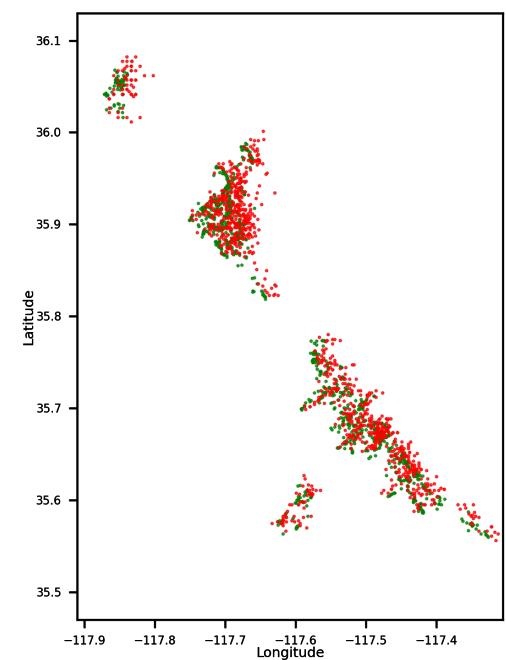
Results - Profile



Results - NonLinLoc

Green: true location
Red: recovered location

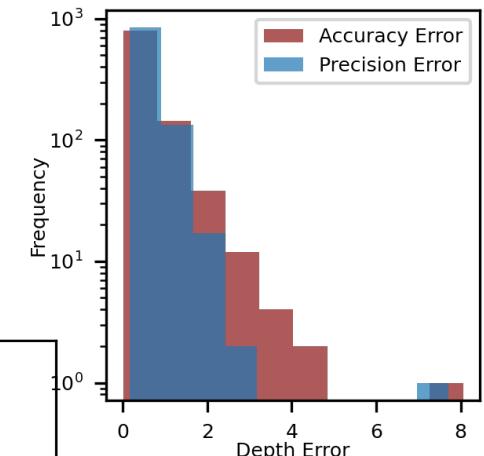
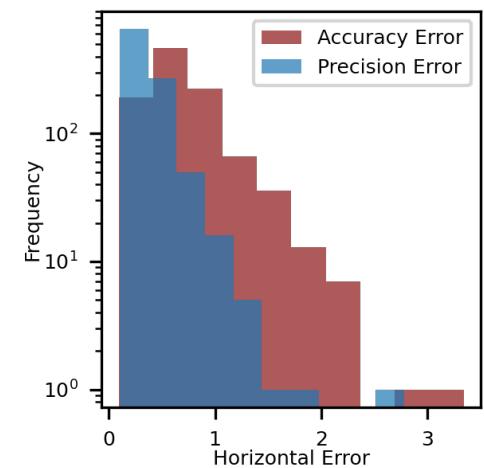
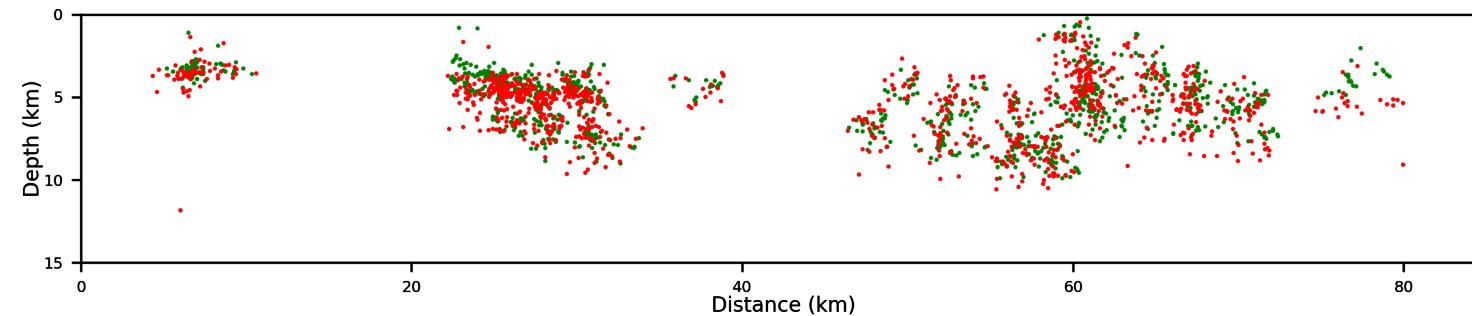
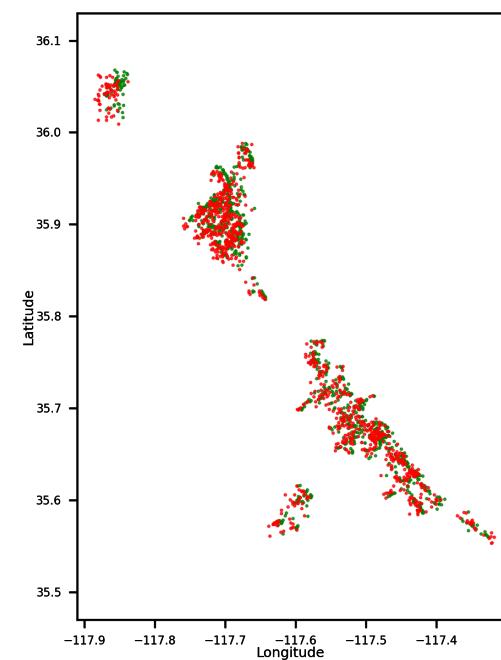
- Capture the general positions
- The depth profile is diffusive
- Overall error is in big range



Results - VELEST

Green: true location
Red: recovered location

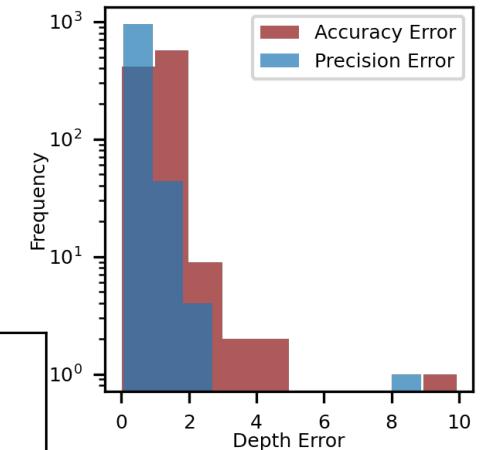
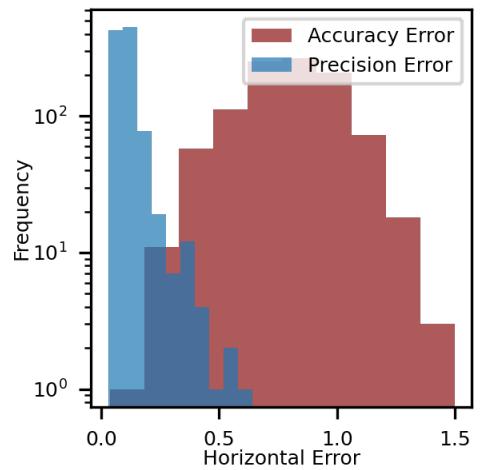
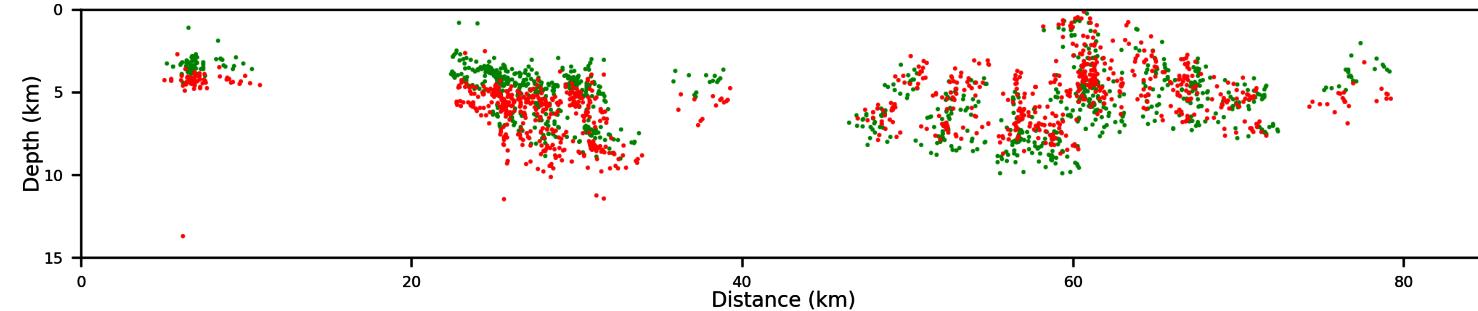
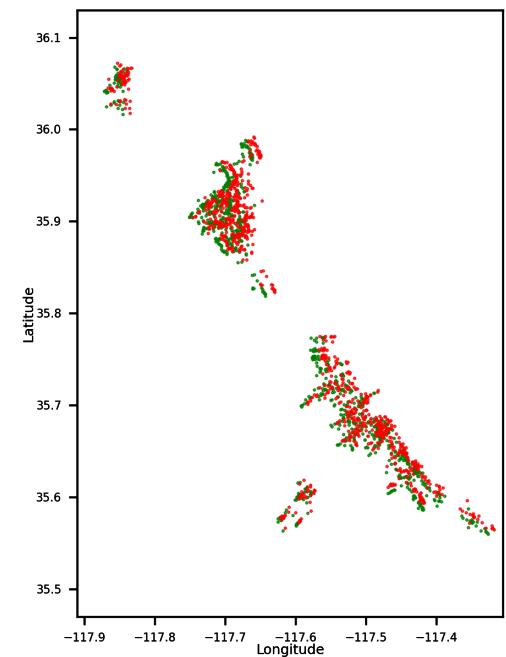
- Well recovered positions with small systematic shifts
- Well recovered depth
- Overall error is in small range



Results - Growclust

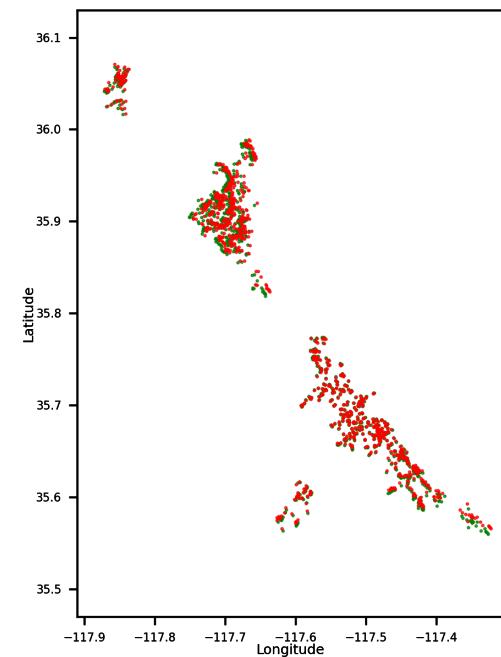
Green: true location
Red: recovered location

- The relative locations are well recovered
- Good relative structure in systematic shift
- Much smaller error in horizontal

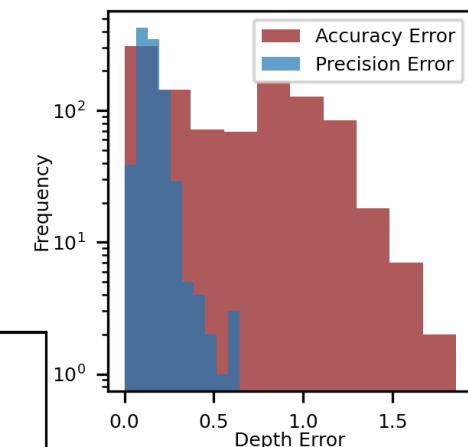
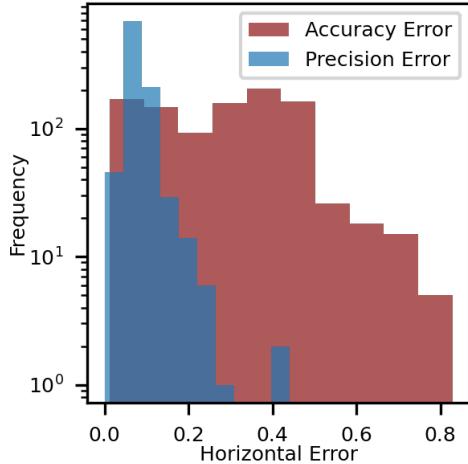
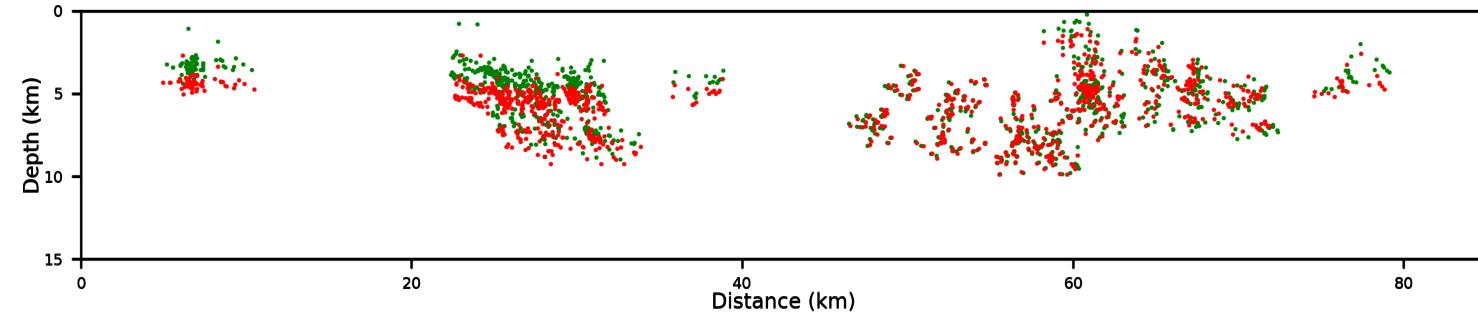


Results - HypoDD

Green: true location
Red: recovered location



- The relative locations both in mapview and depth profile are well recovered and in small shift
- Much smaller error



Results – Overall Performance

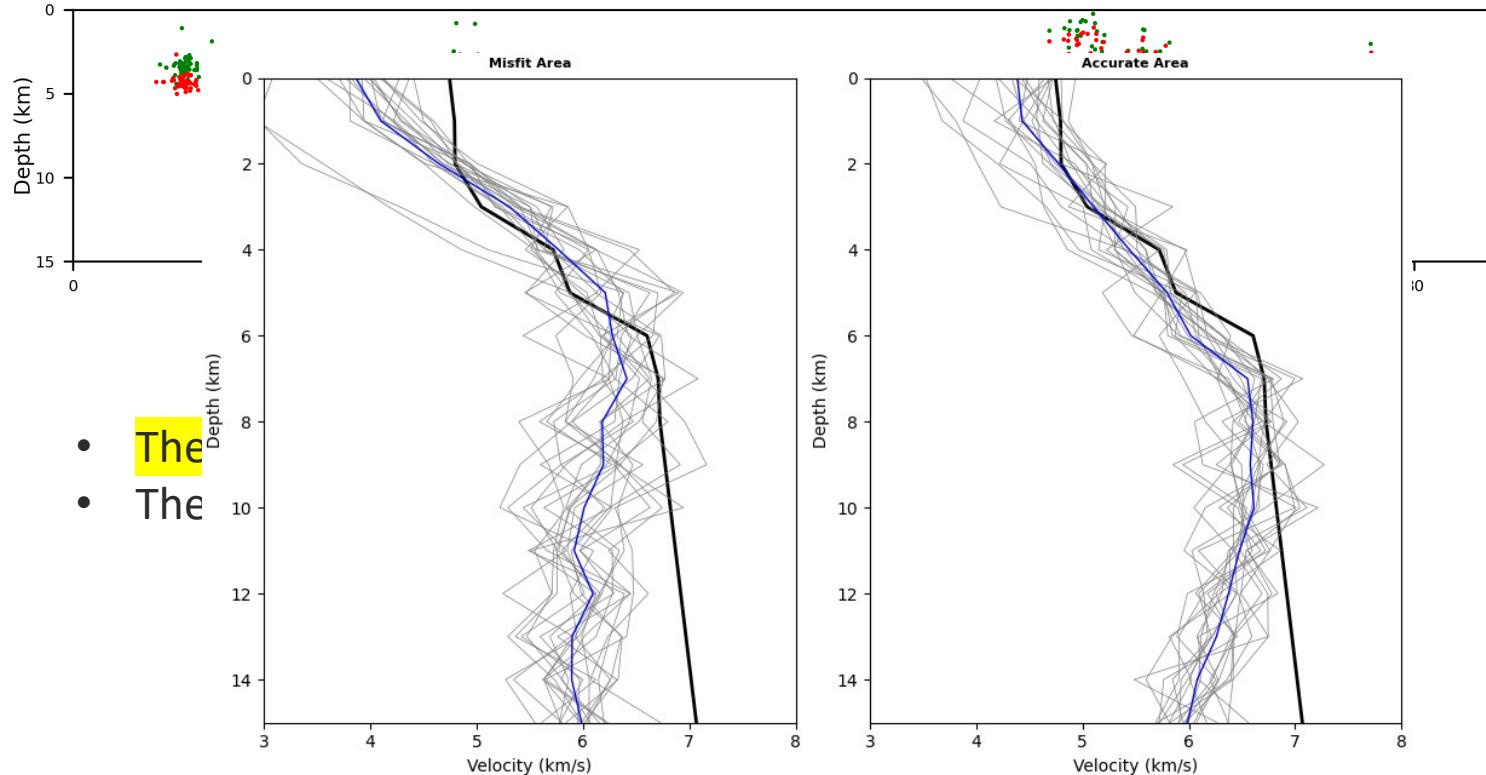
Chamfer distance:

$$d_{CD}(S_1, S_2) = \sum_{x \in S_1} \min_{y \in S_2} \|x - y\|_2^2 + \sum_{y \in S_2} \min_{x \in S_1} \|x - y\|_2^2$$

Method	Mean Accuracy Error (km)		Mean Precision Error (km)		Chamfer Distance	Execution Time (s)
	Horizontal	Depth	Horizontal	Depth		
VELEST	0.696	0.559	0.372	0.644	1.170	60
HypoDD	0.289	1.124	0.079	0.147	0.946	17
Growclust	0.802	1.071	0.110	0.297	1.702	140
HypoInverse	0.824	1.118	0.560	0.671	1.617	2
Non_Lin_Loc	1.070	1.124	0.705	0.799	1.750	2000
XCorLoc	2.376	1.486	1.233	0.669	1.629	120
HypoSVI	0.911	1.152	0.454	0.951	1.854	2000

Discussion – Limits of 1D Velocity Structure

What causes the depth misfit variation across the region?

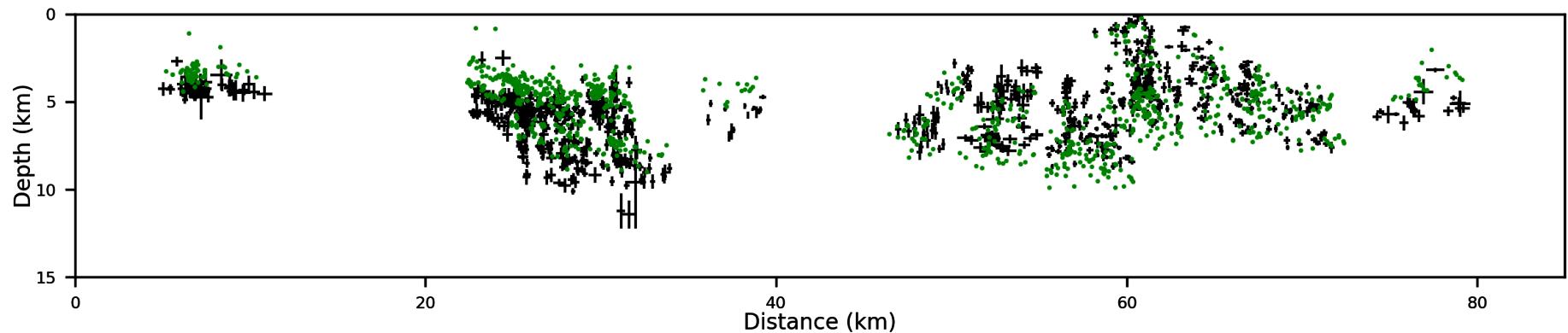


Discussion – Uncertainties Analysis

1. Bootstrapping: Growclust, XCORLOC
2. Estimating from likelihood function: NonLinLoc, HypoSVI
3. Covariance Matrix: Hypoinverse

They all underestimate the error range

Example from Growclust's bootstrapping error analysis



Discussion – Near Source S Phase

Gomberg et al. (1990) argues that the S phase within distance of $1.4 * \text{source depth}$ is important for constraining the depth

Randomly take down one near source S phase V.S. take down one far S phase

VELEST	Mean Accuracy Error (km)		Mean Precision Error (km)	
	Horizontal	Depth	Horizontal	Depth
Without near S	0.64	0.61	0.39	0.70
Without far S	0.58	0.51	0.35	0.62

Conclusion

- This controlled experiment proves the effectiveness of the relative location methods over absolute one, while VELEST and HypoDD stands out.
- 1D velocity structure is limited to represent the realistic situation.
- Our findings should motivate revisiting earthquake location uncertainty assessment.

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Thank You