## Seismic elastic double-beam characterization of faults and fractures for CO<sub>2</sub> storage site selection

Project Number: DE-FE0032063

PI: Yingcai Zheng, University of Houston

Subcontract: LANL (Lianjie Huang) & Vecta Oil and Gas Ltd. (Bryan DeVault)

> Project start date: July 1, 2021; DOE share: \$799,932. Cost share: \$225,000

> > U.S. Department of Energy National Energy Technology Laboratory 2024 Project Review Meeting, Pittsburgh August 5, 2024

# Teams

- University of Houston
  - Yingcai Zheng (PI)
  - Jake Parsons (graduate student)
  - Joe McNease (graduate student)
  - Hao Hu (past postdoc)
- Los Alamos National Laboratory
  - Lianjie Huang (Co-PI)
  - David Li (graduate research assistant)
  - Neala Creasy (postdoc)
- Vecta Oil and Gas, Ltd.
  - Bryan DeVault (President/CEO; Co-PI)
  - Gulia Popov







#### FINANCIAL ASSISTANCE FUNDING OPPORTUNITY ANNOUNCEMENT



Department of Energy (DOE) Office of Fossil Energy (FE)

EMERGING CO<sub>2</sub> STORAGE TECHNOLOGIES: OPTIMIZING PERFORMANCE THROUGH MINIMIZATION OF SEISMICITY RISKS AND MONITORING CAPROCK INTEGRITY Funding Opportunity Announcement (FOA) Number: DE-FOA-0002401 Announcement Type: Amendment 1<sup>1</sup> CFDA Number: 81.089 EMERGING CO<sub>2</sub> STORAGE TECHNOLOGIES: OPTIMIZING PERFORMANCE THROUGH MINIMIZATION OF SEISMICITY RISKS AND MONITORING

#### **CAPROCK INTEGRITY**

Funding Opportunity Announcement (FOA) Number: DE-FOA-0002401

AOI 1a - Fault Detection, Characterization, and Hazard Assessment.

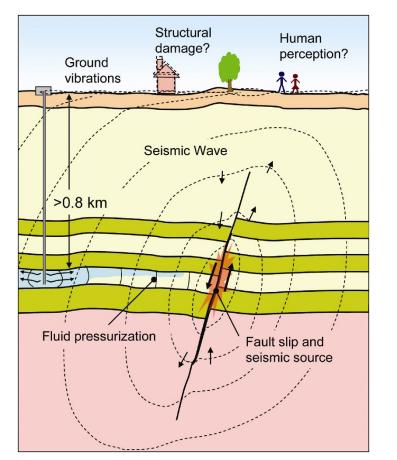
No field work. Novel method development. Final deliverable: a software package for subsurface analysis for Gigatonne storage scenarios.

# Motivation: induced seismicity is quite common in operating CCS sites

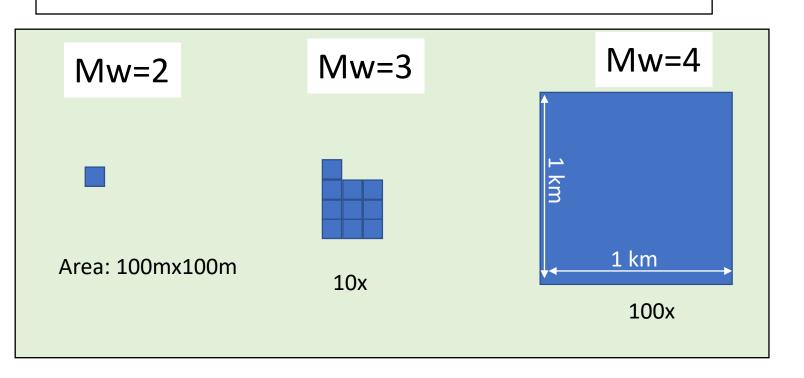
CCS Sites	Start Time	End Time	Injection Amount	Seismic Monitoring (Y/N)	Max Magnitude	Distance from injection	Rock Type	Injection Depth
Decatur, USA	2012	-	2.8Mt	Y	1.55	400m	Sandstone	2140m
In Salah, Algeria	2004	2011	3.8Mt	Y	1.6	1100m	Sandstone	1950m
Otway Australia	2007		30000t	Y	0.5	450m	Sandstone/Cal cite/Kaolinite	2000m
Weyburn, Canada	2000	2011	15Mt	Y	-0.5	300m	Sandstone	1420m
Sleipner, Norway	1996	-	15Mt (2016)	Ν	-	-	Sandstone	1000m
Tomakomai, Japan	2016	-	0.2Mt	Υ	0.5	1.5km	Sandstone	2400m
Fort Saskatchewan, Canada (Quest)	2015	-	5.7Mt		0.8	70m	Sandstone	2172m

(McNease et al. 2023)

#### Scientific rationale: ruptured fault size/area $\rightarrow$ seismic event magnitude

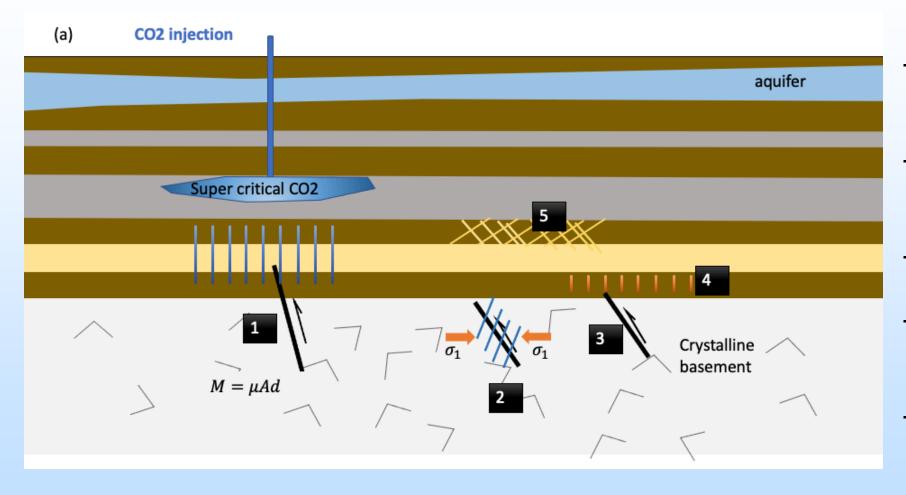


$$Mw = (2/3) Log_{10}(mu^*Area^*slip) - 6.0$$



The induced seismic event magnitude should be viewed with respect to the reservoir/seal thickness. Estimating fault/fracture size  $\rightarrow$  upper bound in event magnitude 6

### **Project Goals and Objectives for Gigatonne injection**

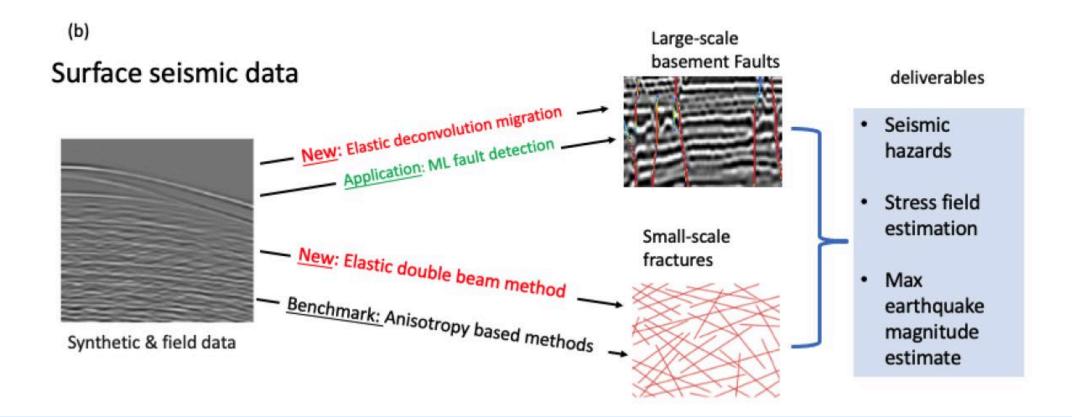


- Small-scale fractures & large-scale faults
- Sedimentary layers & basement
  - Stress field
- Potential seismic hazard; max Magnitude;
- Possible fluid pathways for CO<sub>2</sub> plume migration and leakage

## **Objectives budget periods (BP)**

- <u>BP 1.</u> Fault detection and fracture characterization in the basement using synthetic 9C surface seismic data (BP-1)
- <u>BP 2.</u> Fault detection and fracture characterization in the basement using <u>field</u> 9C surface seismic data (BP-2)
- **<u>BP 3.</u>** Determination of fault stress state and fault activation potential (BP-3)

## Technology Roadmap

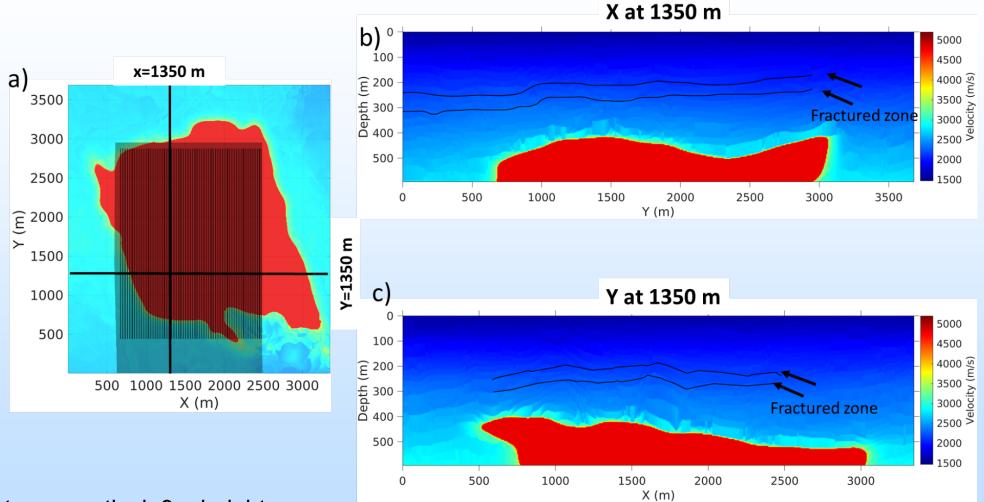




# Why another method for fracture characterization?

• Can seismic migration see the small-scale fractures?

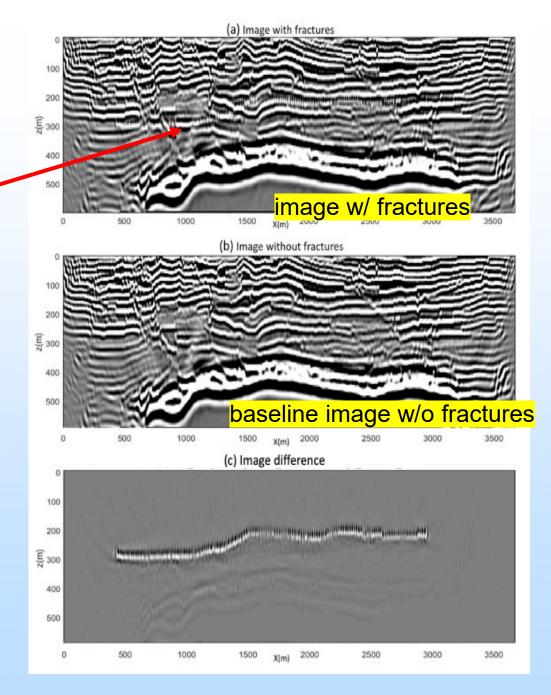
# Motivational example: fractures are hard to see



Fractures: vertical; 9m height; fracture compliance 1e-10 m/Pa

#### Finite difference modeling: Coates and Schoenberg (1995)

## Hard to see fractures in traditional seismic migrated images

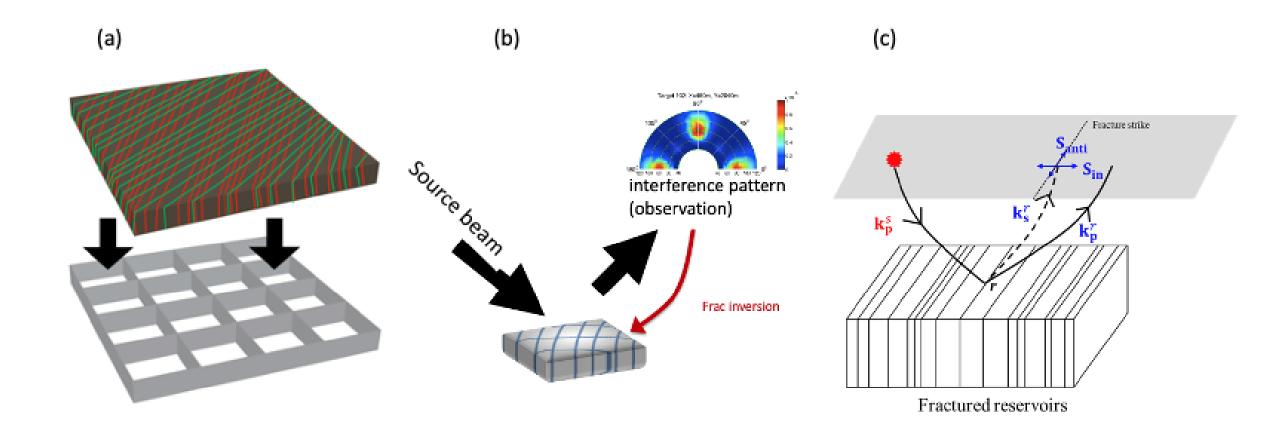


## Seismic double-beam method

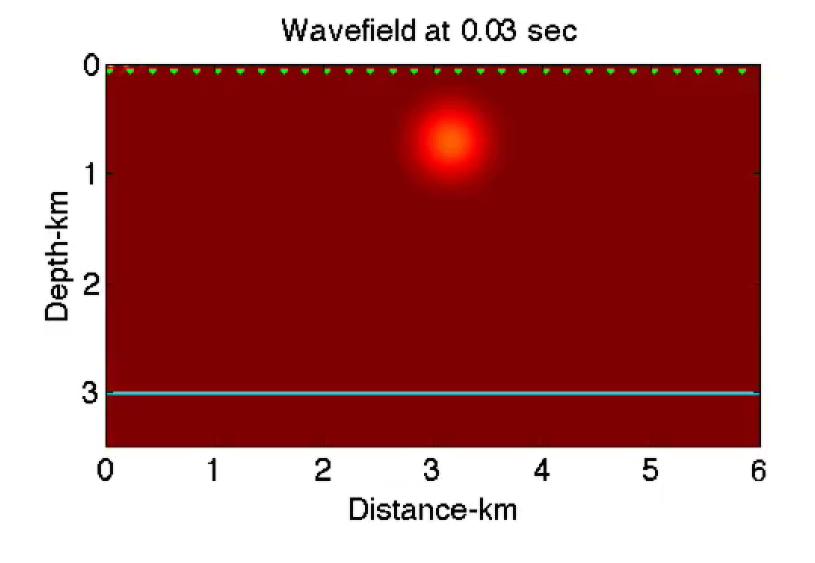
to characterize small-scale fractures

# Basic idea of double-beam method

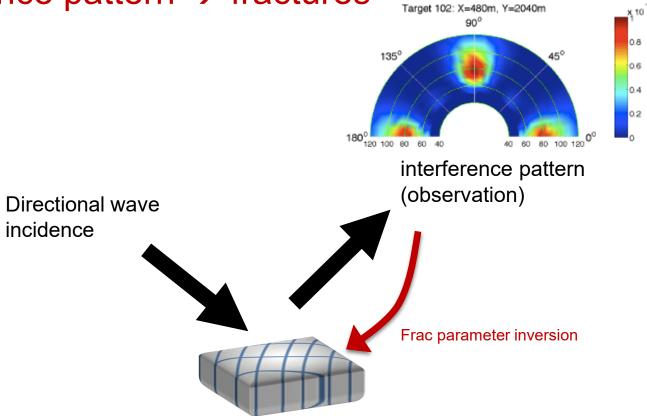
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## A directional seismic packet can selectively image



### Interference pattern $\rightarrow$ fractures

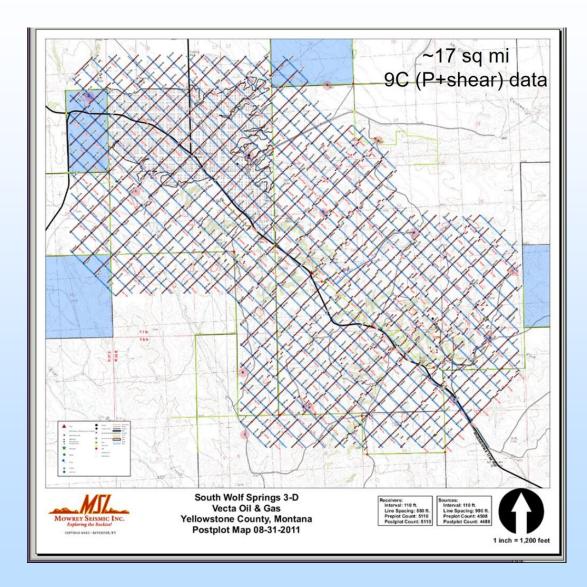


Fractured reservoir

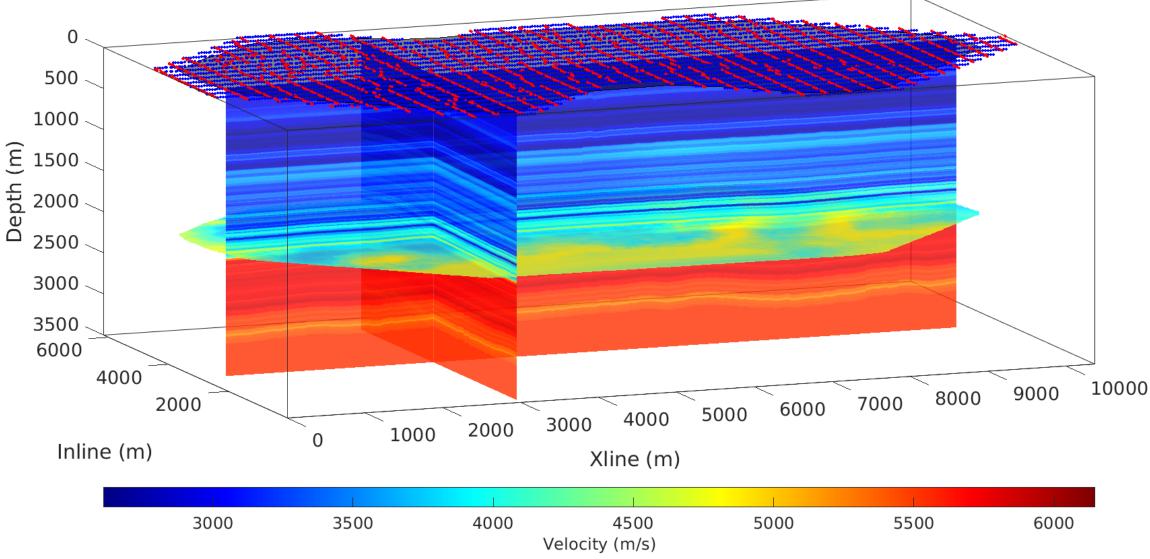
- Fracture orientation
- Density
- Compliance

# Build the synthetic elastic model from the field data

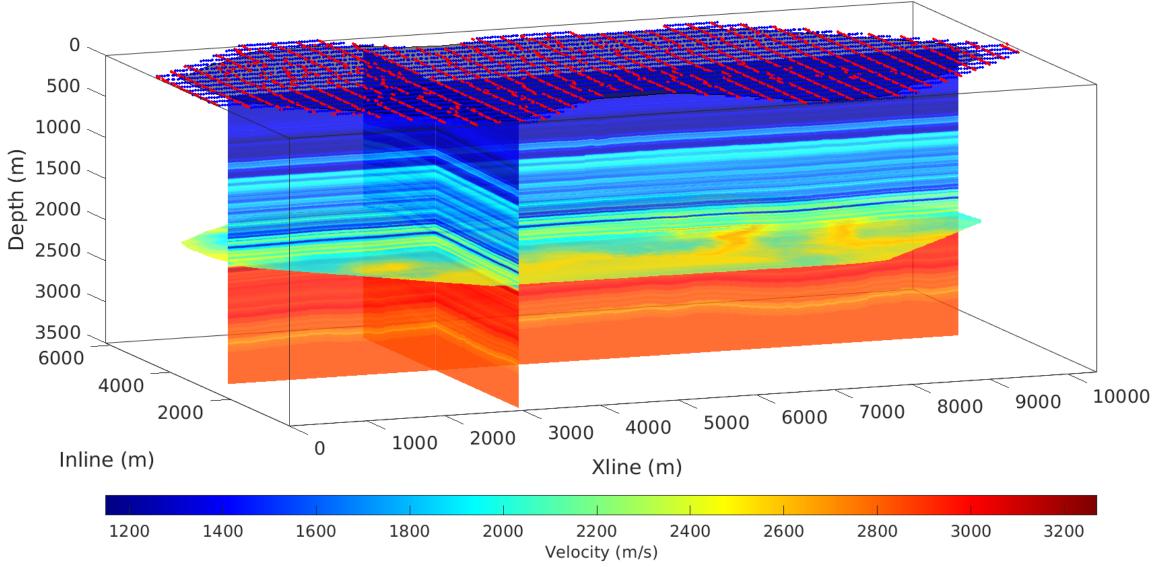
## Field seismic data (9C) in Montana



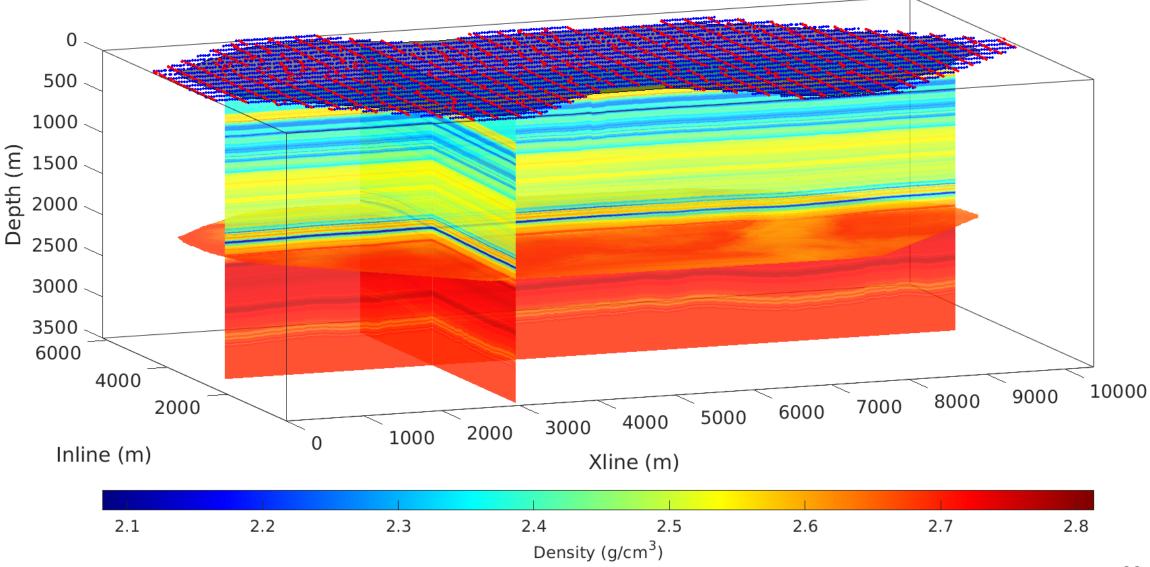
### P-wave velocity model from the field Vecta data



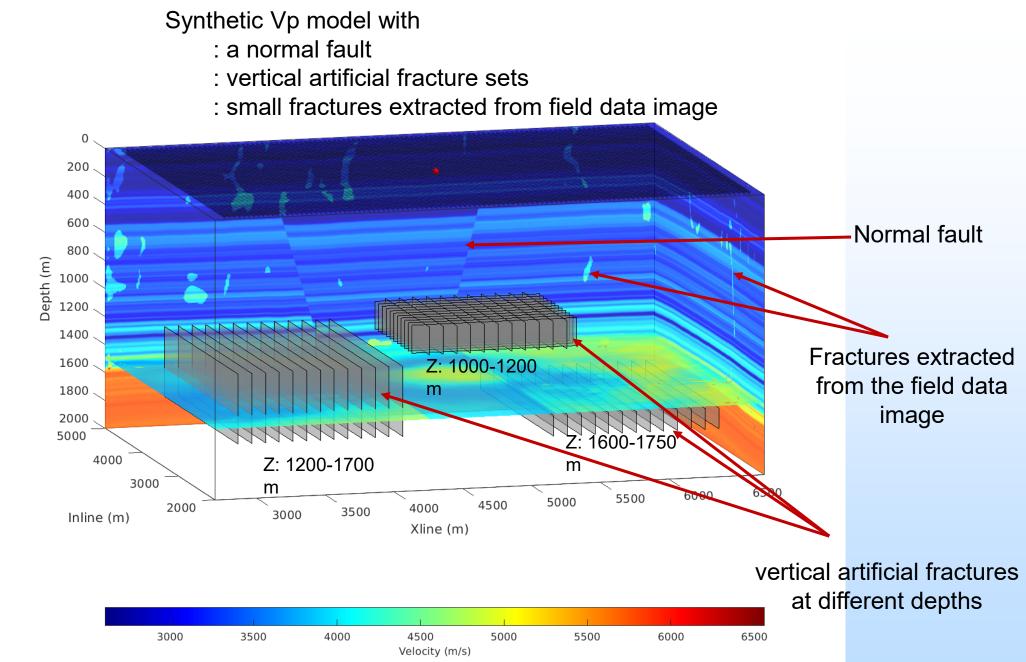
### Shear-wave velocity model from the field Vecta data

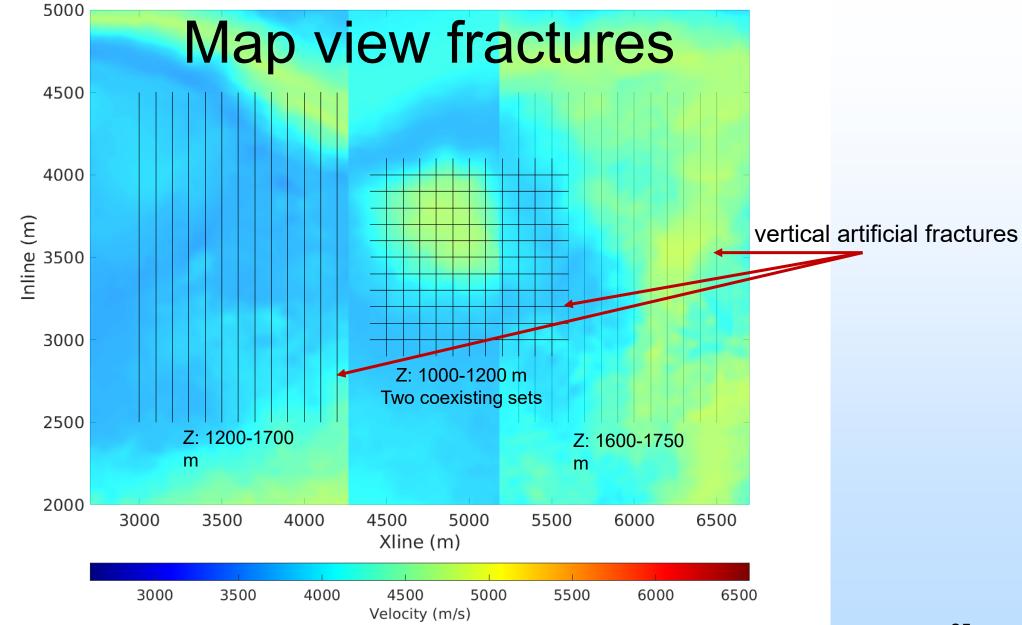


### **Density model from the field Vecta data**

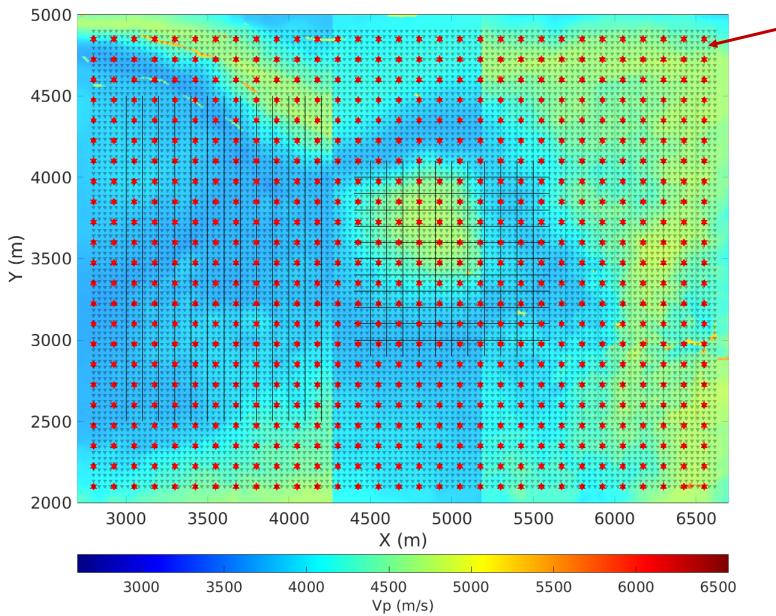


# **Detection Results: small-scale fractures**





### **Acquisition geometry**



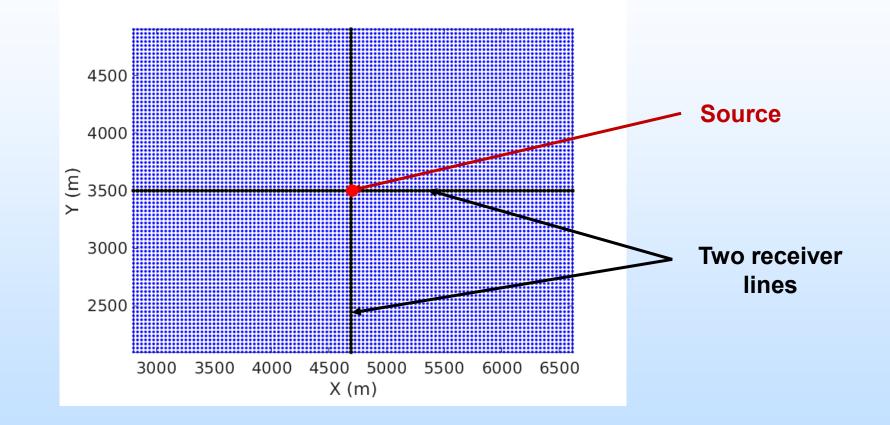
**★Source** ▼Receiver

Sources: X: 2800:125:6550 m Y: 2100:125:4850 m Total: 31\*23=**713** 

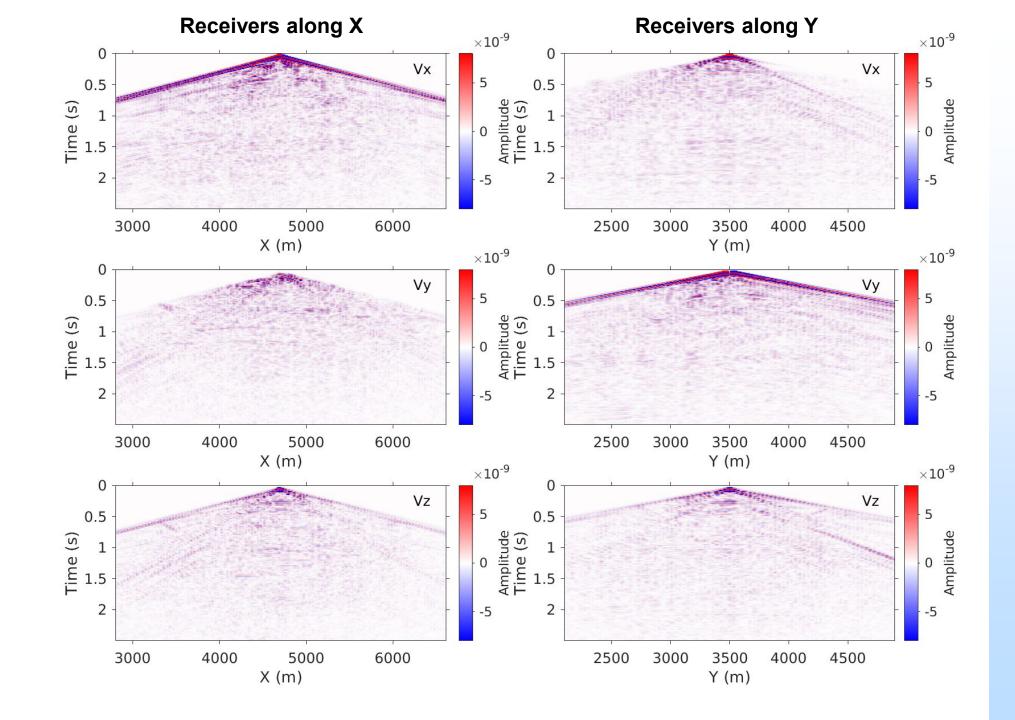
Receivers: X: 2800:35:6615 m Y: 2100:35:4900 m Total: 110\*81=**8910** 

Both Source and receivers are at surface

# Modeled common-shot gathers at one location with different types of source.

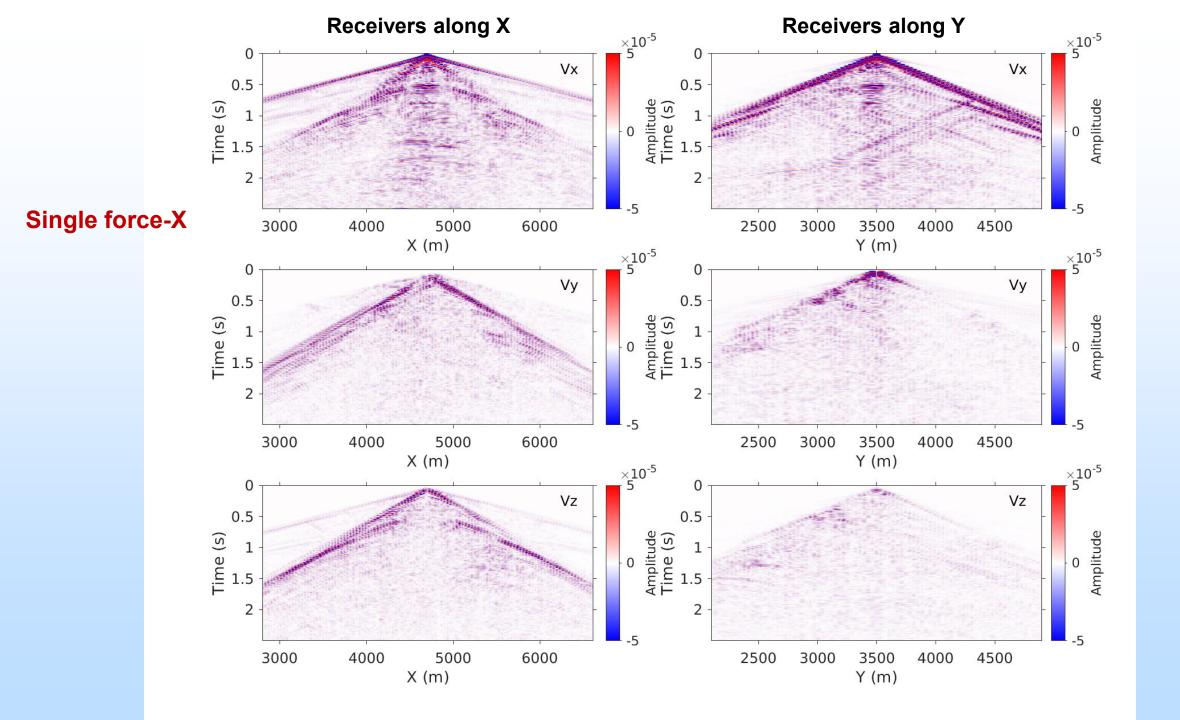


Source wavelet: 20 Hz Ricker 28



Explosive source

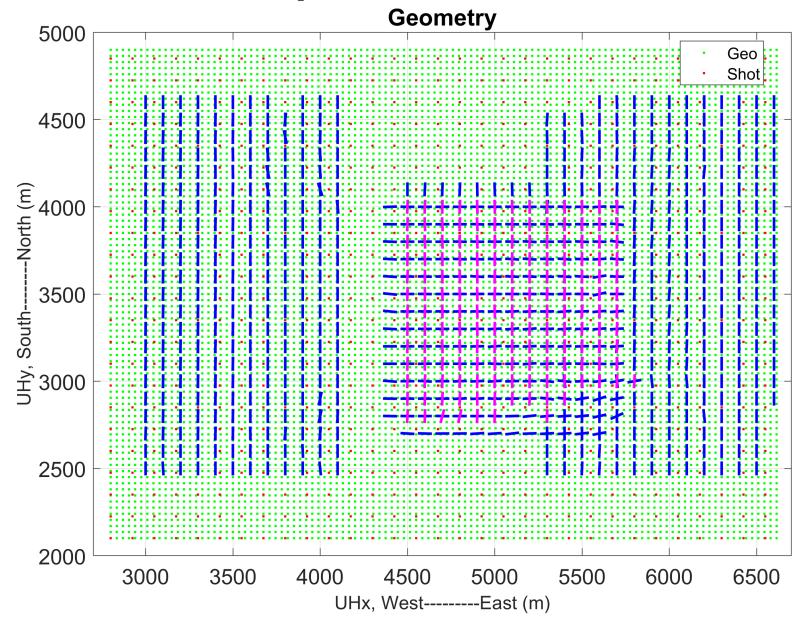
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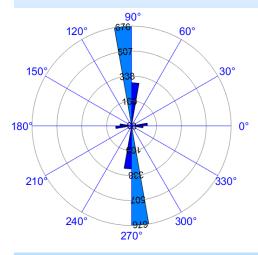
# Fracture detection results using Seismic Double-Beam method

(Zheng et al., 2023 SGW for methodology review)

### **Results: top view of double-beam detected fractures**

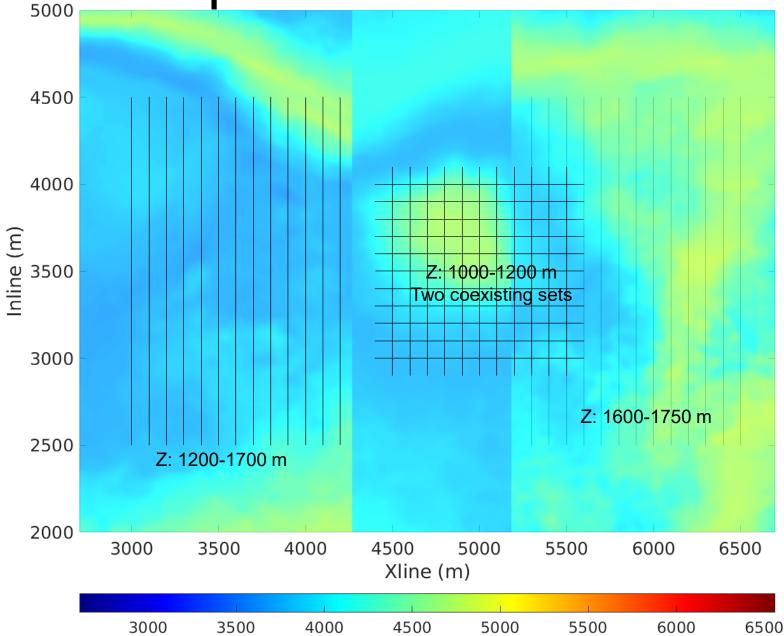


At three depths (1100 m, 1400 m and 1650 m) from frequencies 15 Hz, 20 Hz, 30 Hz and 40 Hz

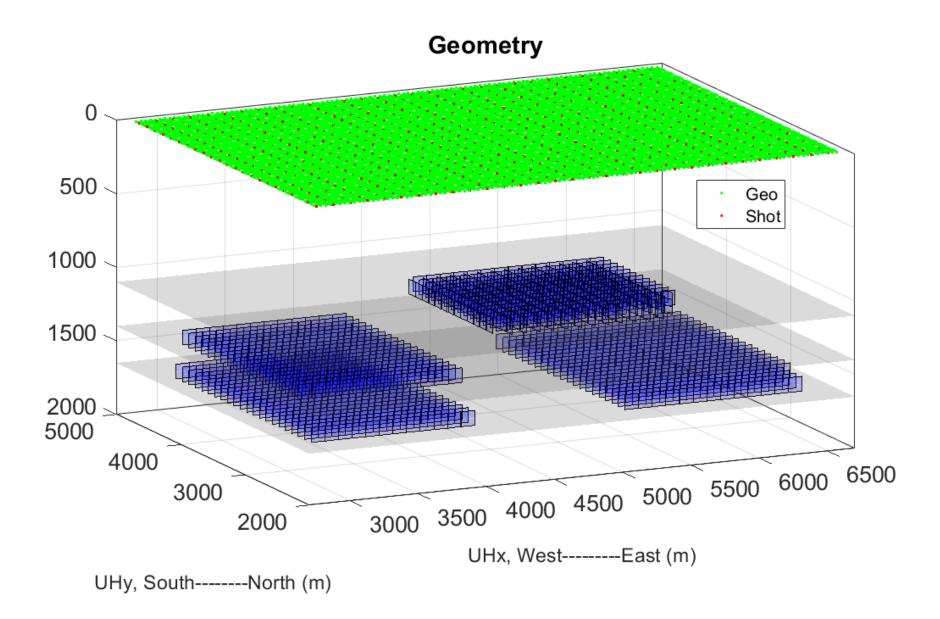


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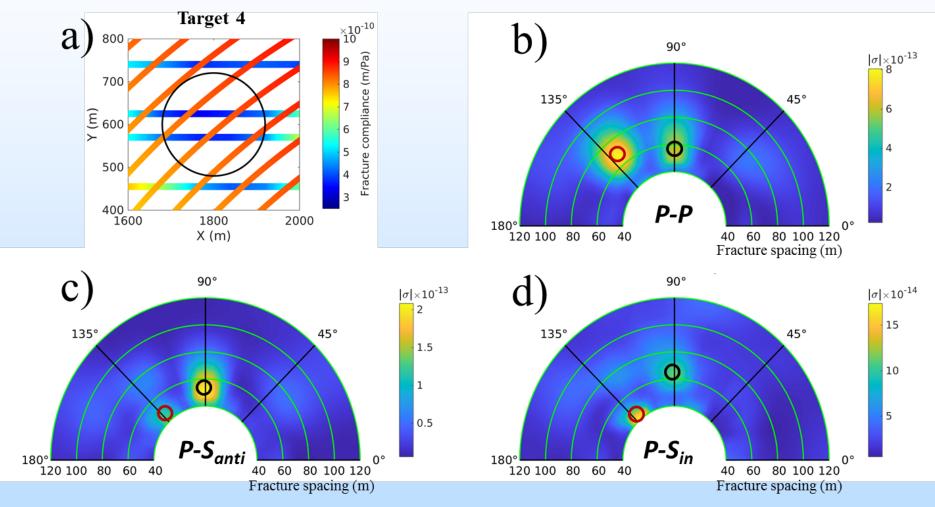
# Map view of true fractures



## **3D view of detected fractures**



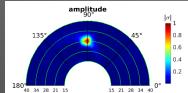
# Next steps: discrete fracture network using Machine learning & field data



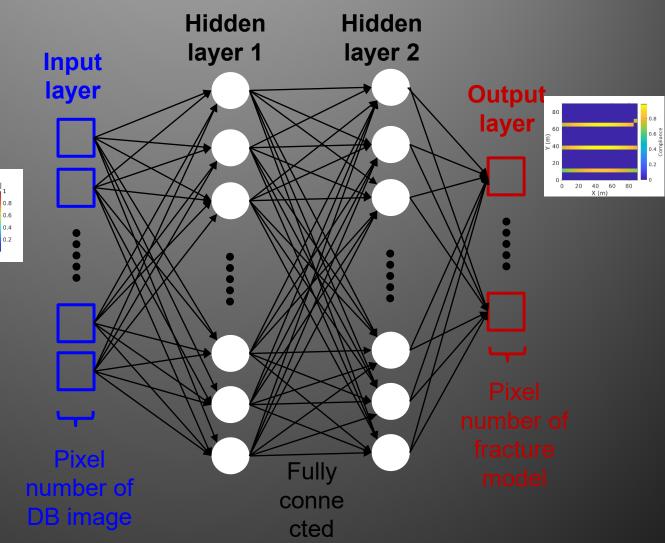
(Zheng et al., 2023 SGW)

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#### Elastic double-beam neural network (DBNN) machine learning

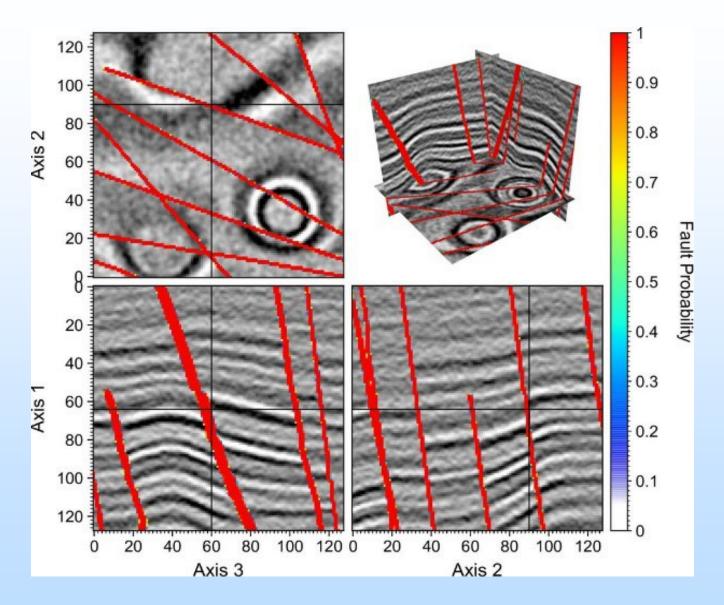


The architecture of our fully-connected neural network including two hidden layers.



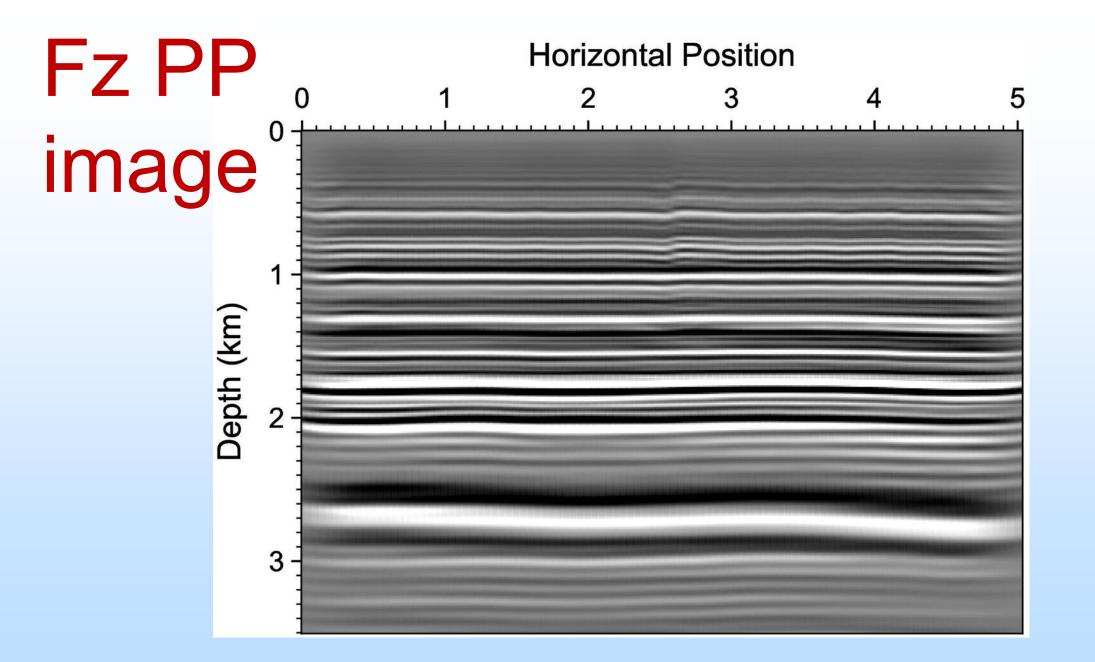
# Large-scale fault detection using machine learning (LANL)

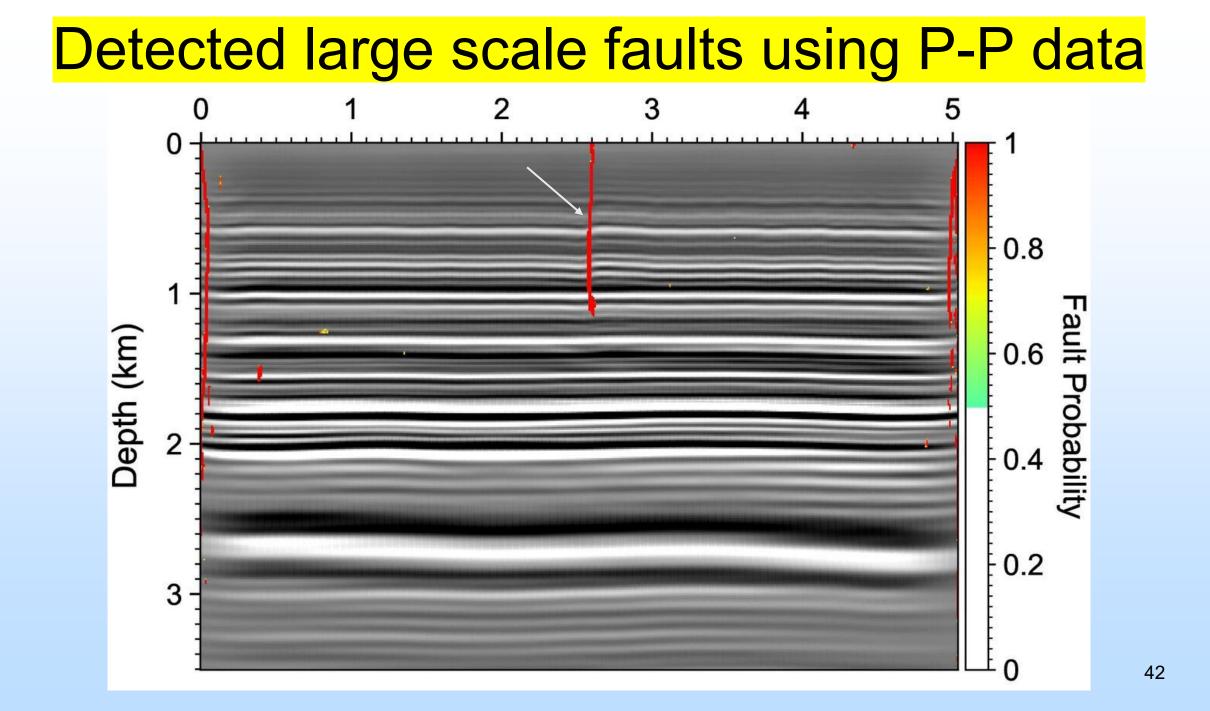
#### Large-scale faults detected using LANL's new NRU

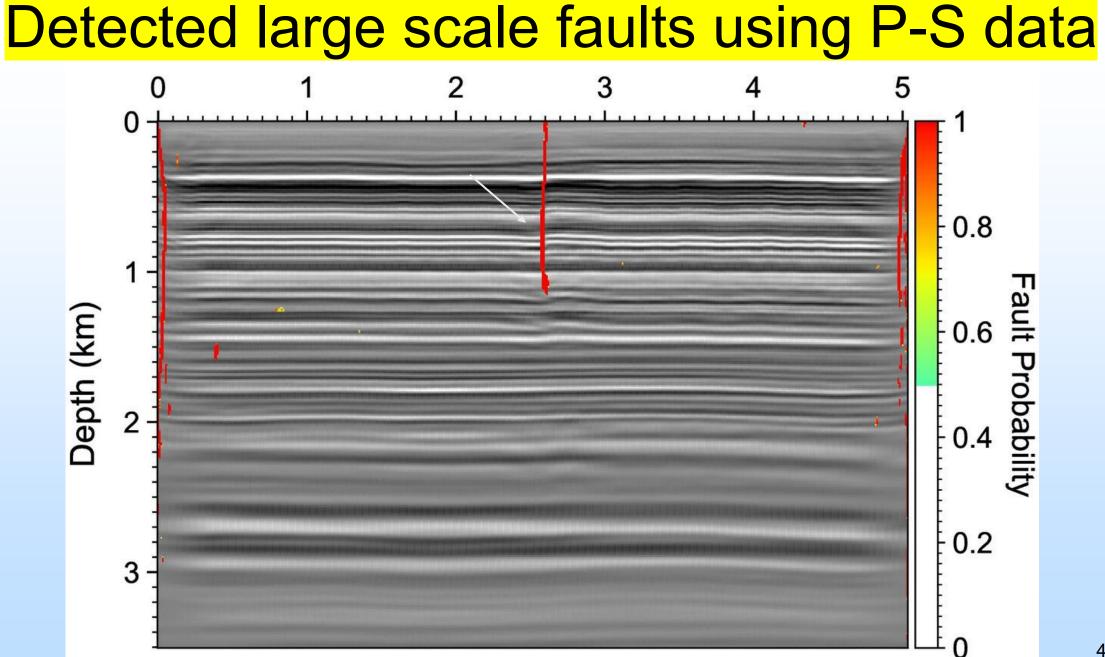


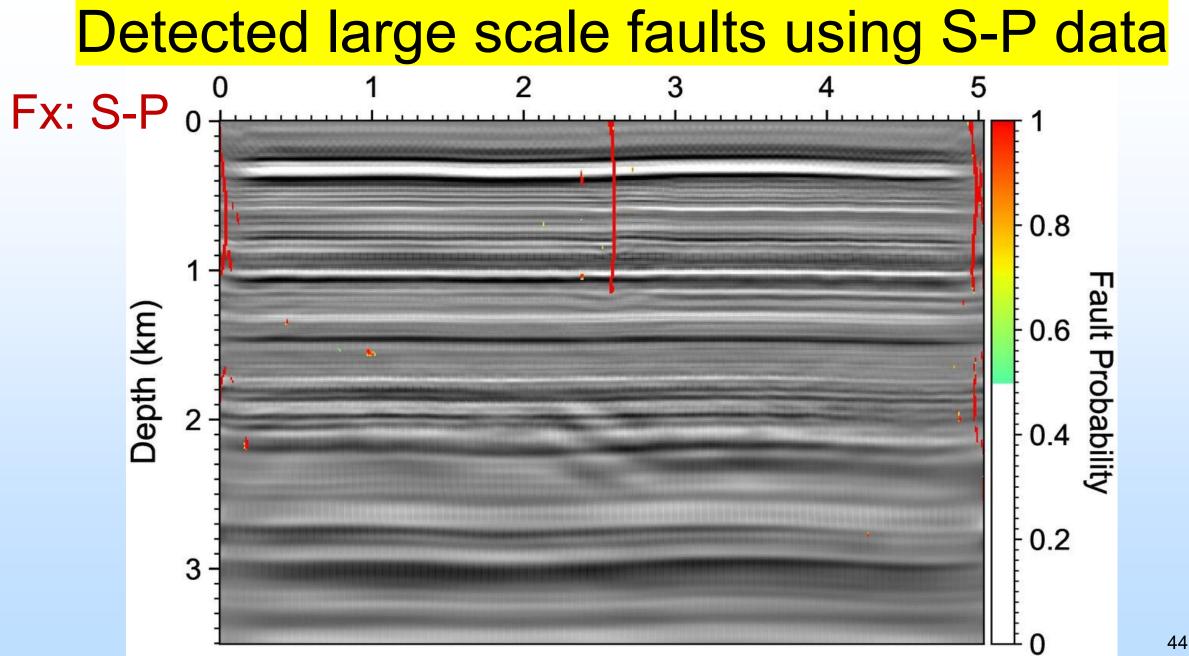
(Gao, Huang, Zheng 2022; **IEEE TGRS**) 40

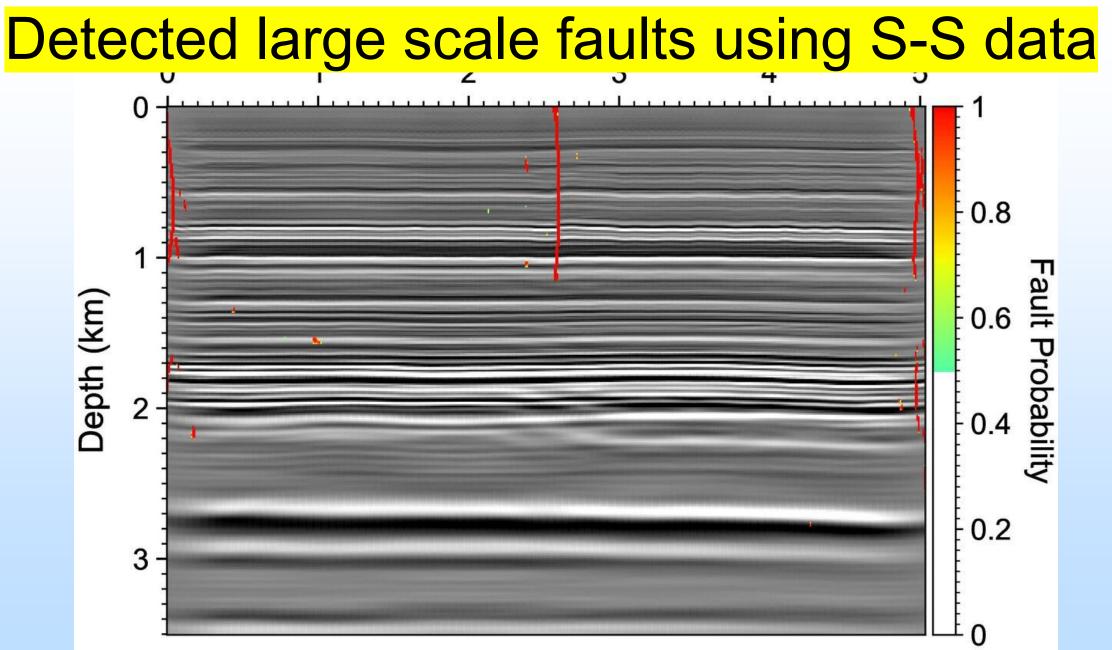
Nested Residual U-shaped convolutional neural network (NRU)









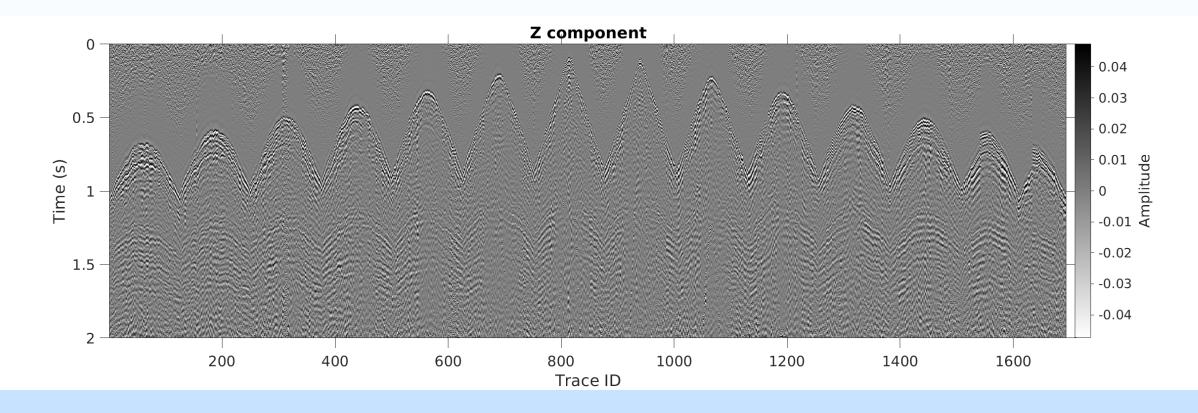


Lessons from: large-scale fault and smallscale fracture detection

 Different modes (P-P, P-S, S-P, S-S) give consistent results.

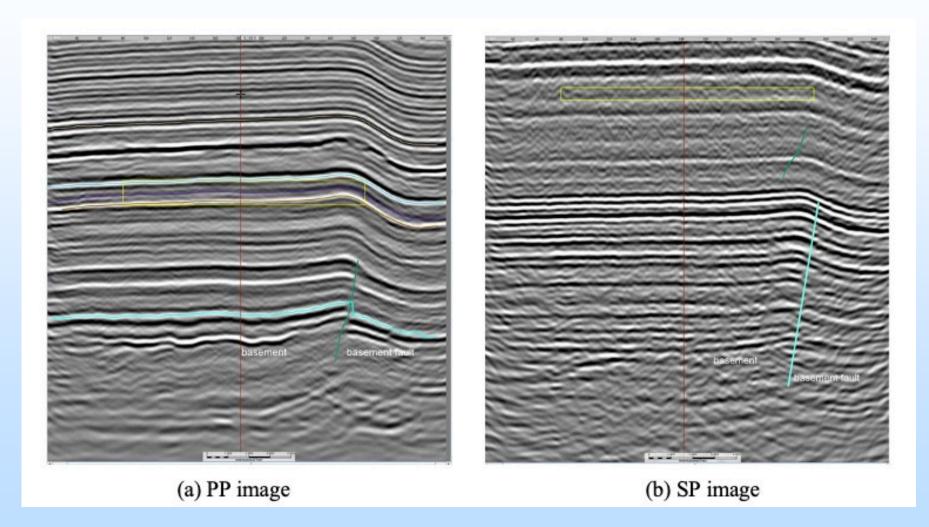
• The consistency reduces uncertainties & yields high confidence in the results.

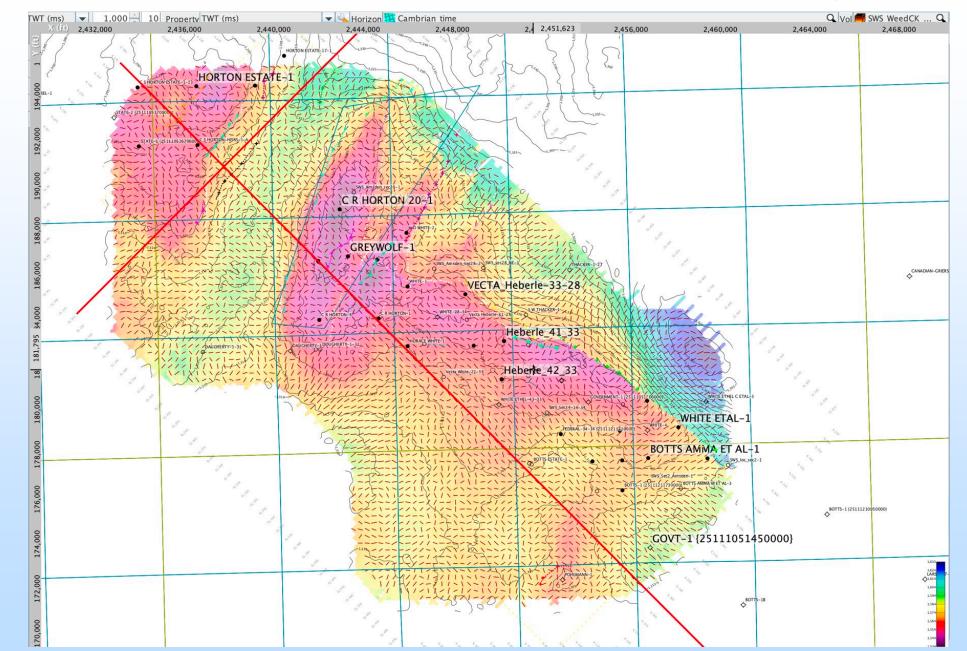
### Field data



Vertical vibreseis

## Field data: Fractures and Basement faults (Vecta processing)

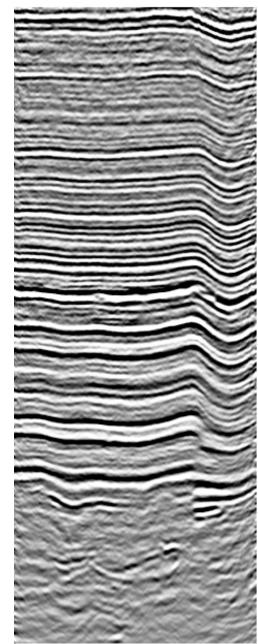


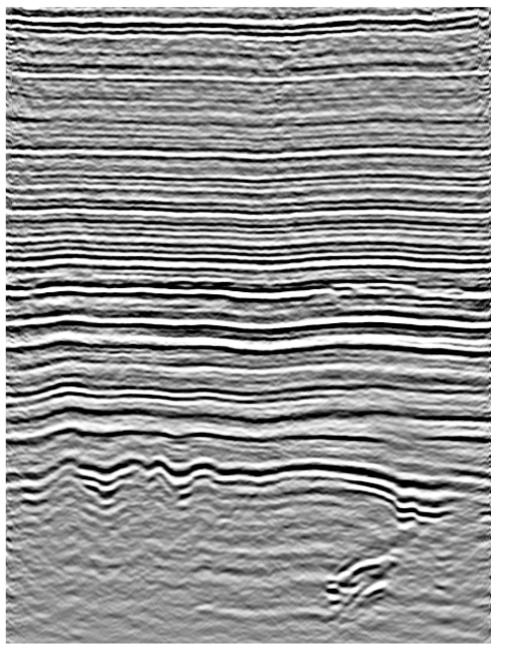


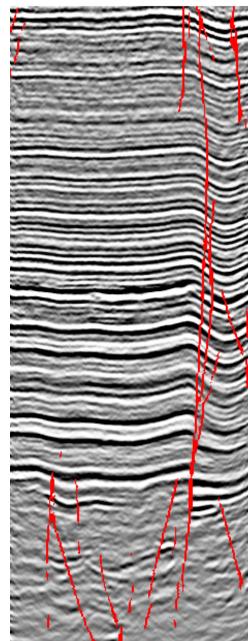
Fast azimuths from Bani corendered with Cambrian time structure: this will be compared with directions from splitting analysis in future

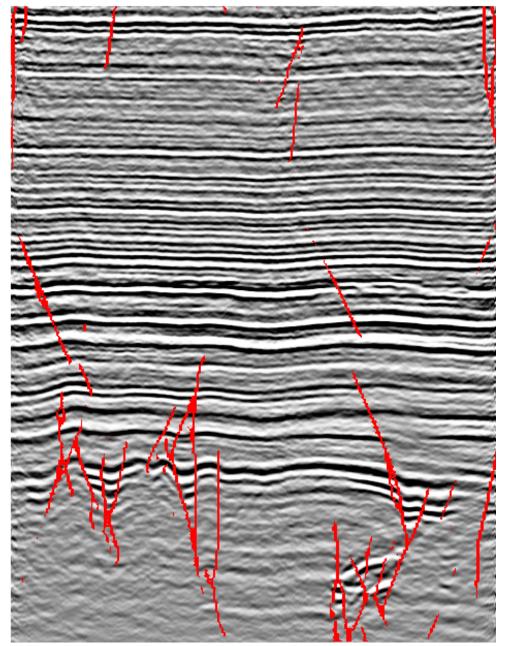
## **PP Images**

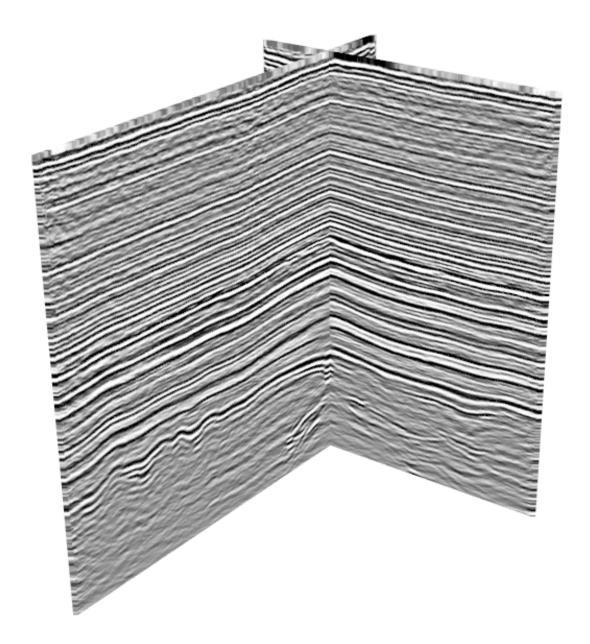
Xline: 111

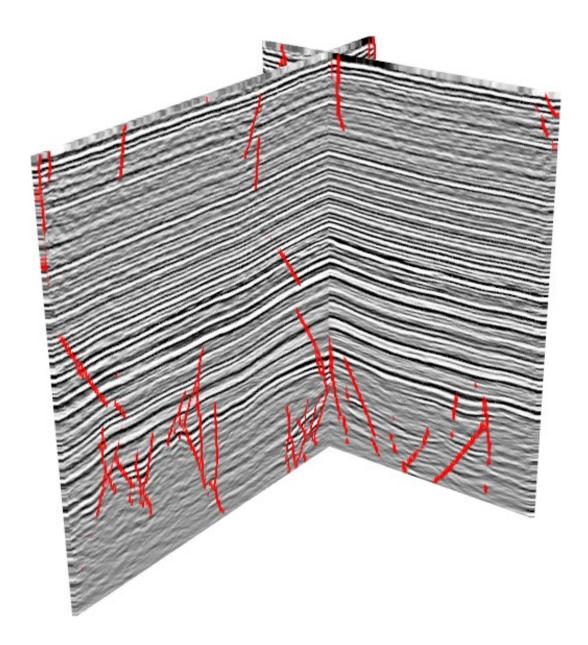


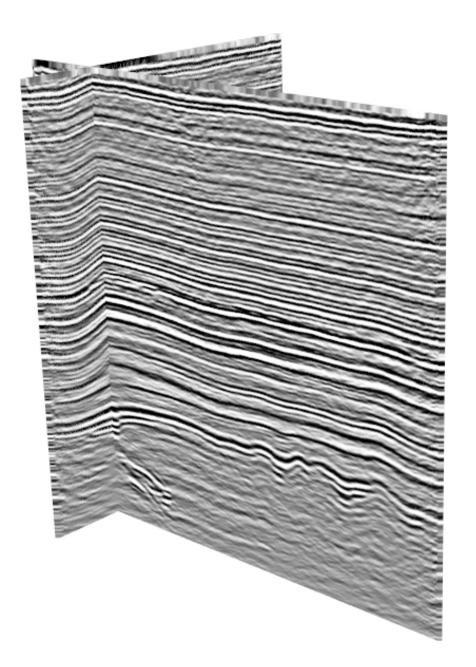


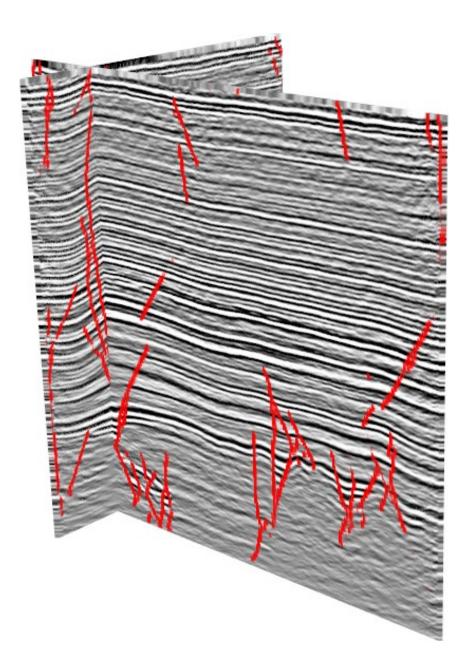




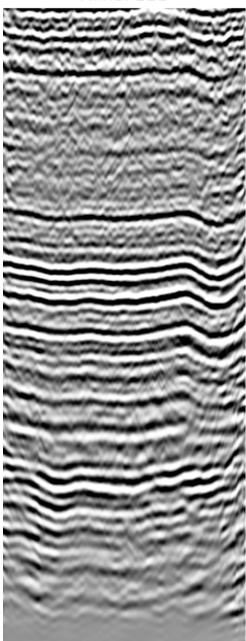


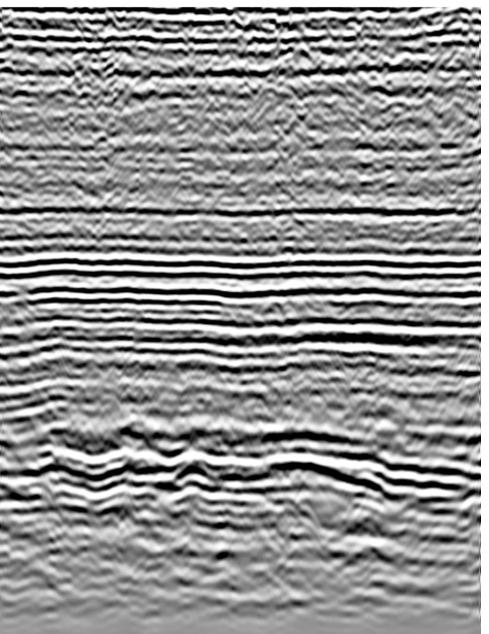


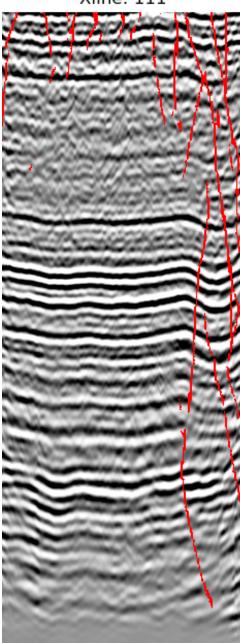


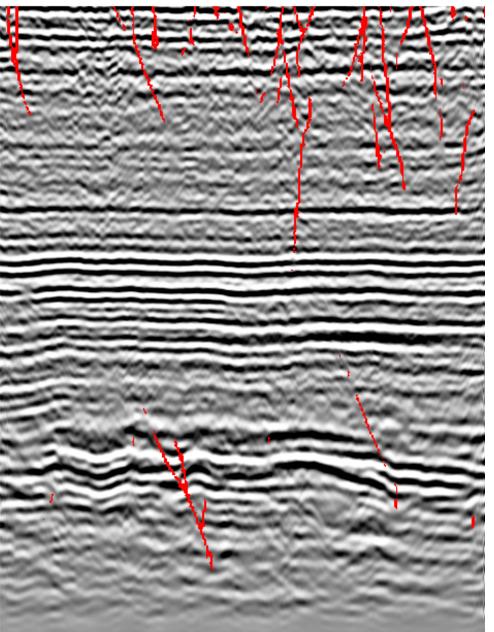


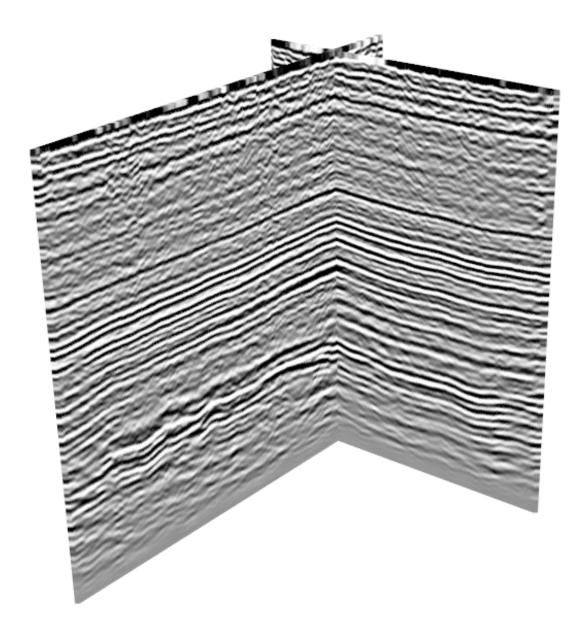


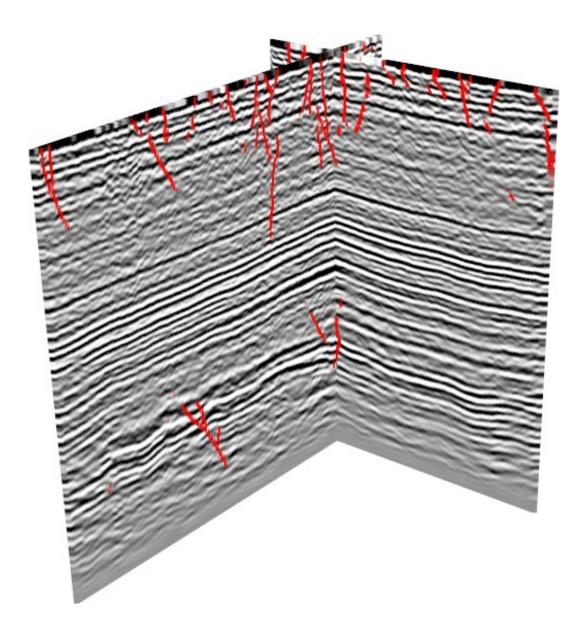


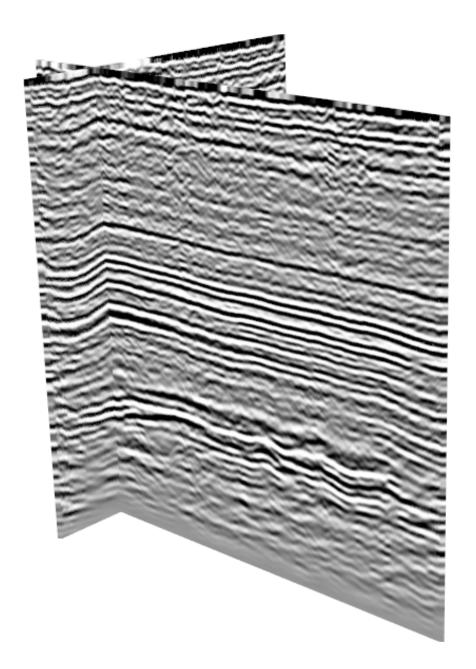


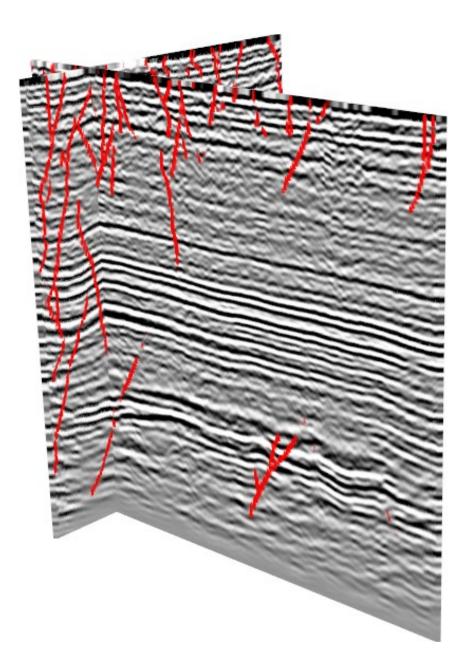




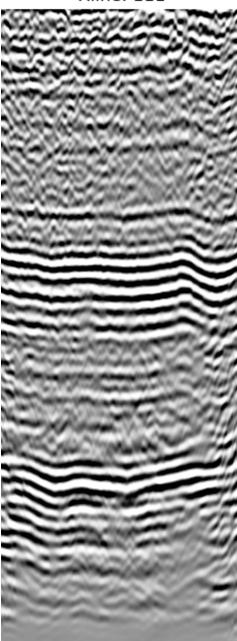


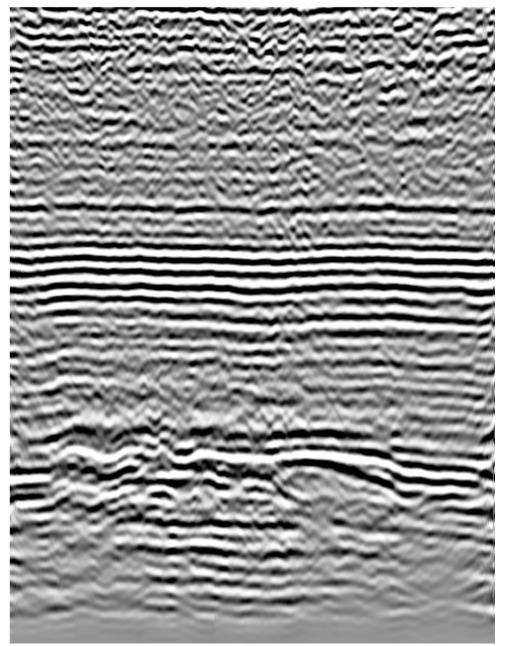


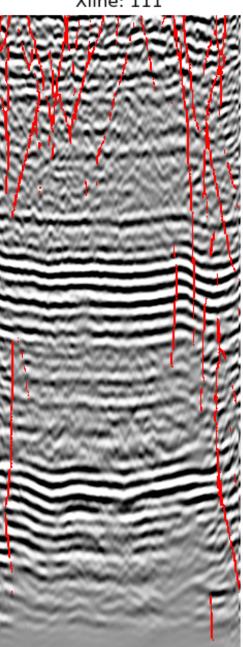


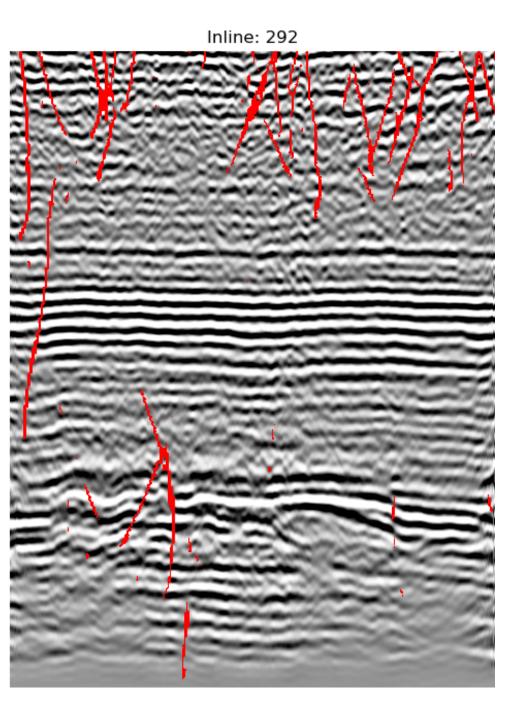


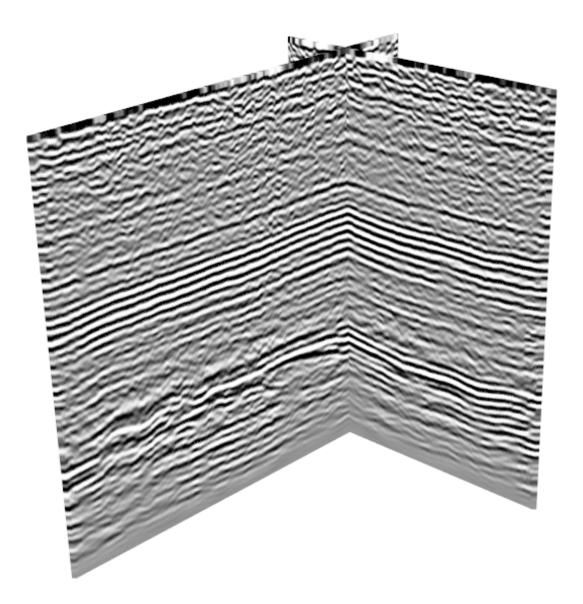
**RR Images** 

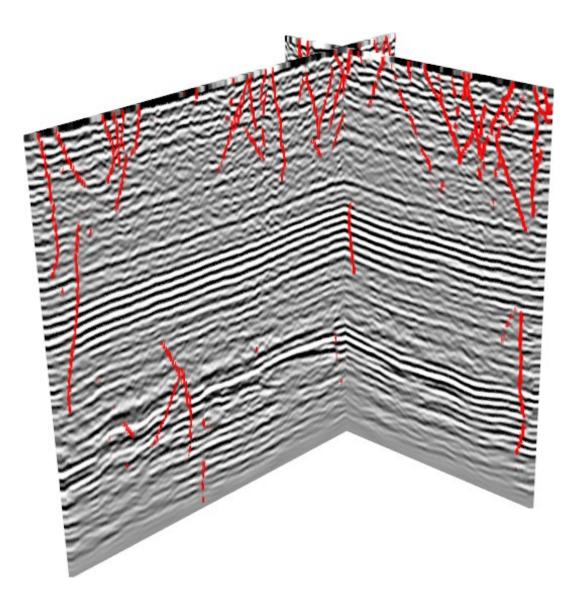


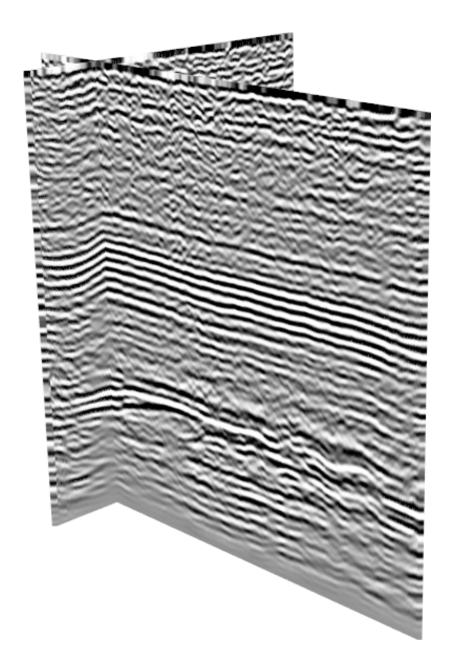


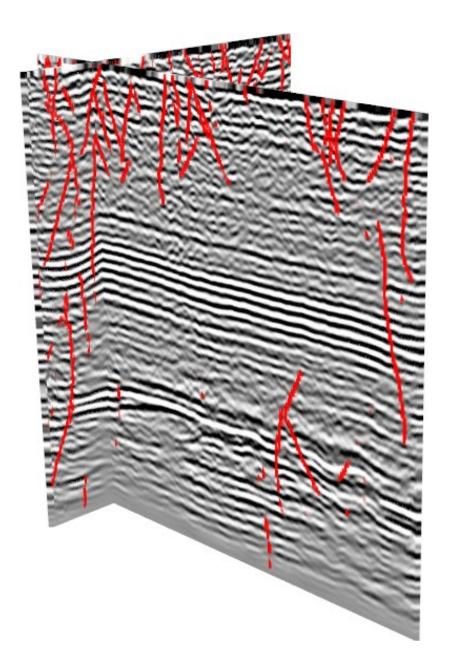




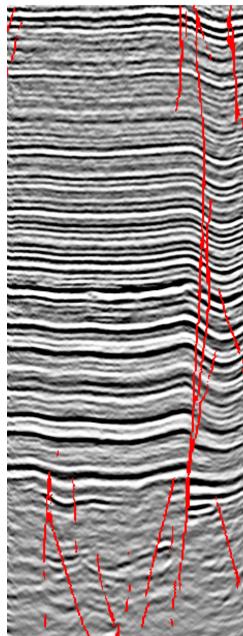


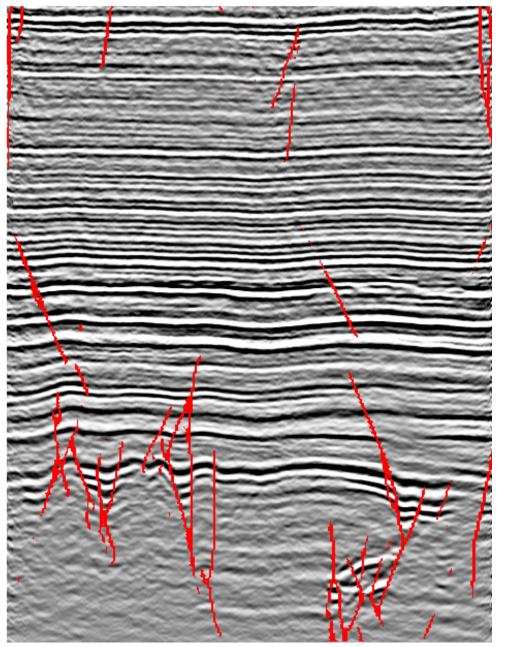




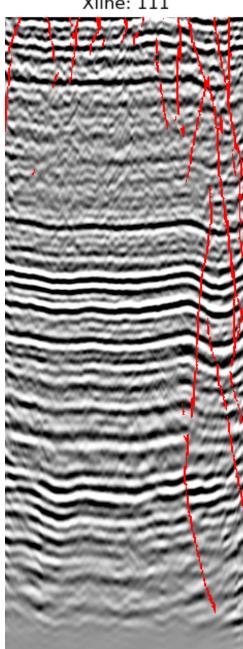


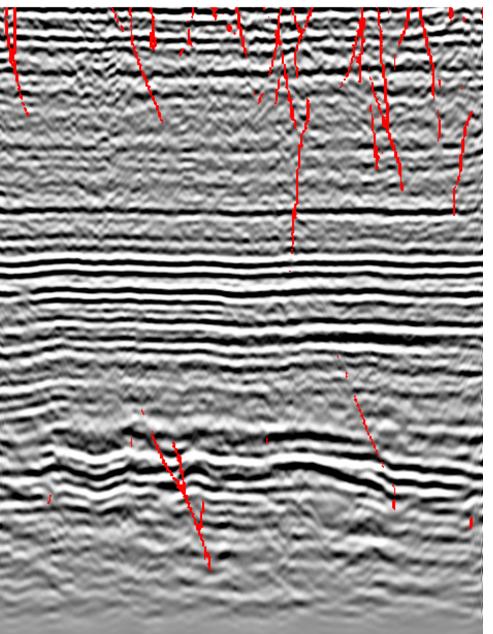




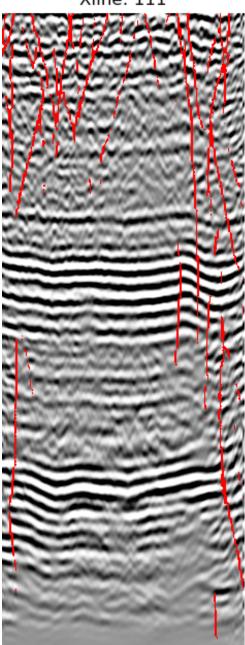


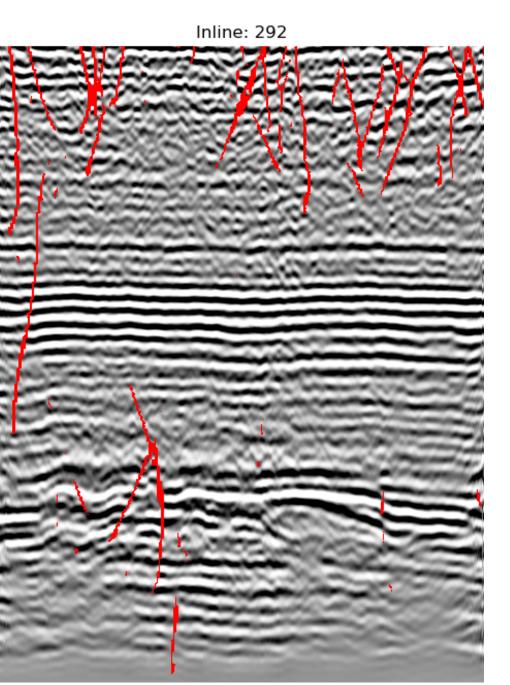
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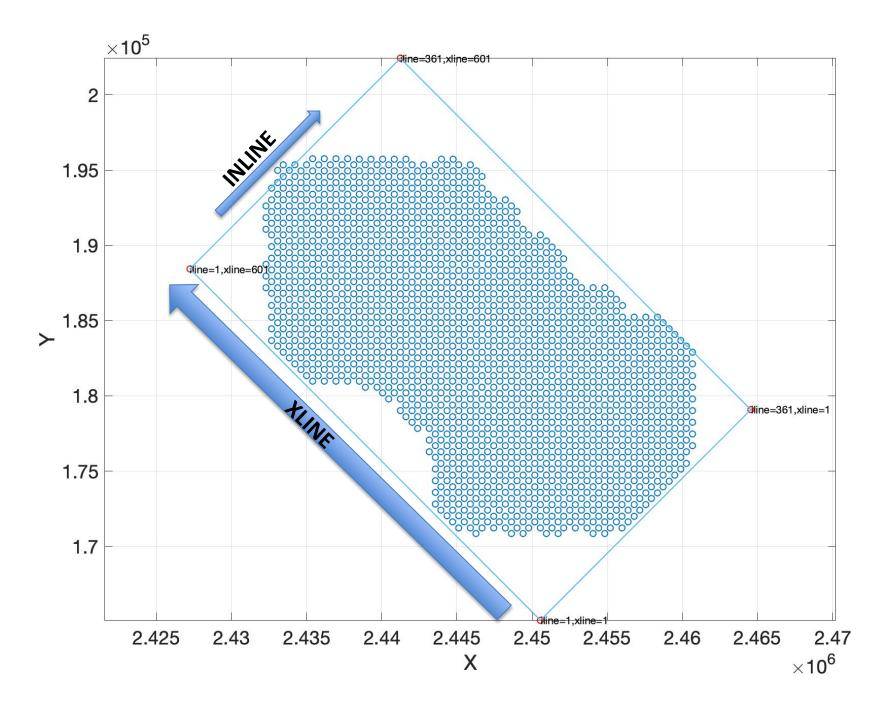


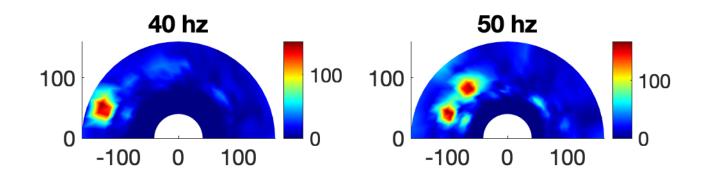


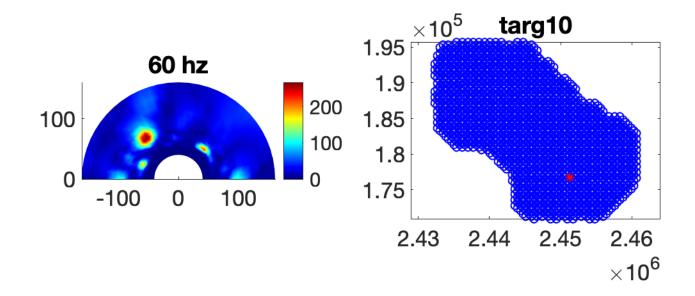


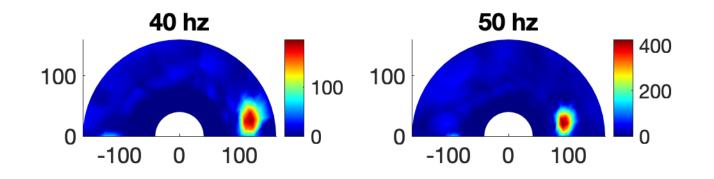


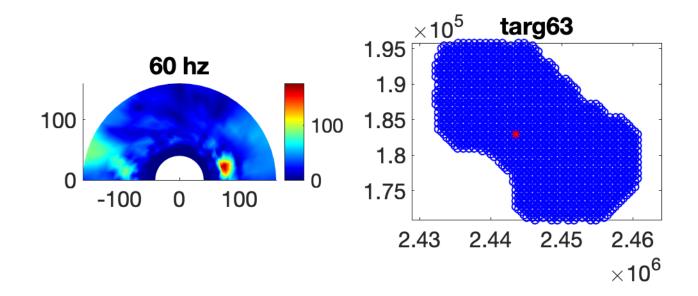
# Double-beam fracture detection on field data

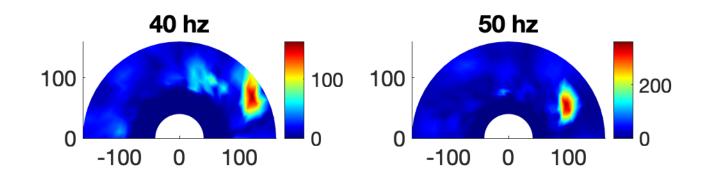


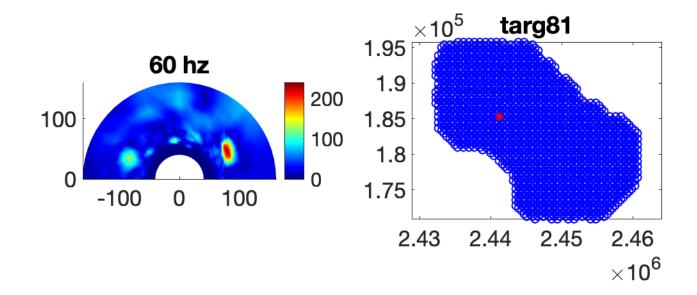


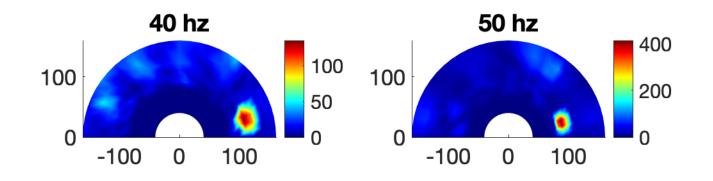


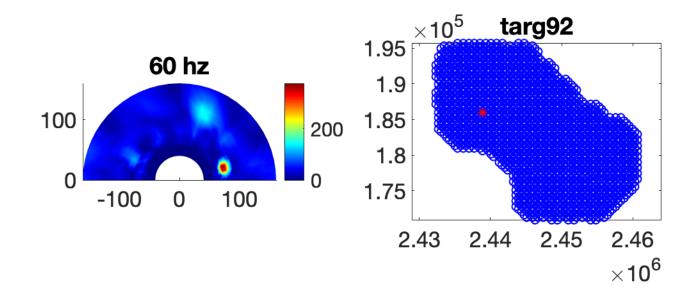


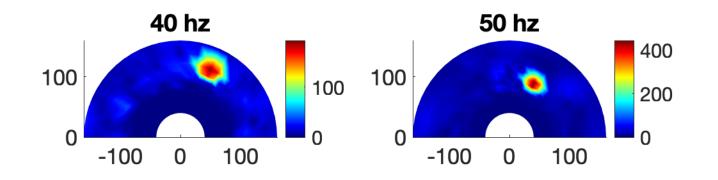


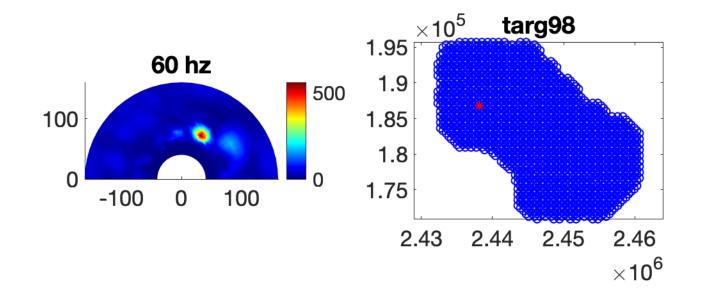












### Task/Subtask Breakdown

#### Task 2.0 - Fault detection and fracture characterization in the basement using synthetic 9C surface seismic data

This task will be in Year-1 which is the budget period 1. The work will focus on synthetic dataset based on a model with dimensions similar to the Wolf Springs field data.

- <u>Subtask 2.1</u> (UH/LANL/Vecta) Model building based on central Montana (M1-M3): Build a 3D elastic model using the Wolf Springs field geometry.
- <u>Subtask 2.2</u> (UH) Multicomponent synthetic seismic data modeling (M4-M6): Using the model and the locations of the sources and receivers in the field data, UH will run their elastic finite-difference code to generate the synthetic datasets. The computation will be done on PI's group cluster.
- <u>Subtask 2.3</u> (LANL) Migration imaging (M7-M9): LANL will conduct P-P, P-S, S-P, and S-S imaging on the synthetic dataset.
- <u>Subtask 2.4</u> (LANL) Machine learning fault detection (M10-M12): LANL will detect faults on P-P, P-S, S-P, and S-S images of the synthetic dataset.
- <u>Subtask 2.5</u> (UH) Fracture characterization using elastic double beam (M10-M12).

•Task 3.0 – Fault detection and fracture characterization in the basement using field 9C surface seismic data This task will be in the budget period 2. The work will focus on the field 9C seismic dataset at Wolf Springs, Yellowstone County, Montana.

•<u>Subtask 3.1</u> – (UH) Seismic data denoise such as the ground roll removal (M13-15).

Because the field data has ground coupling issues and different source strengths, UH will apply trace amplitude balance and ground roll removal to clean up the data.

•<u>Subtask 3.2</u> – (Vecta Oil and Gas) Quality control of the field data and geologic interpretation of faults and fractures (M13-M15).

Vecta Oil and Gas will perform quality control of the field data and analyze geological and well log information for fault/fracture interpretation.

•<u>Subtask 3.3</u> – (UH) eDB fracture characterization (M16-M24).

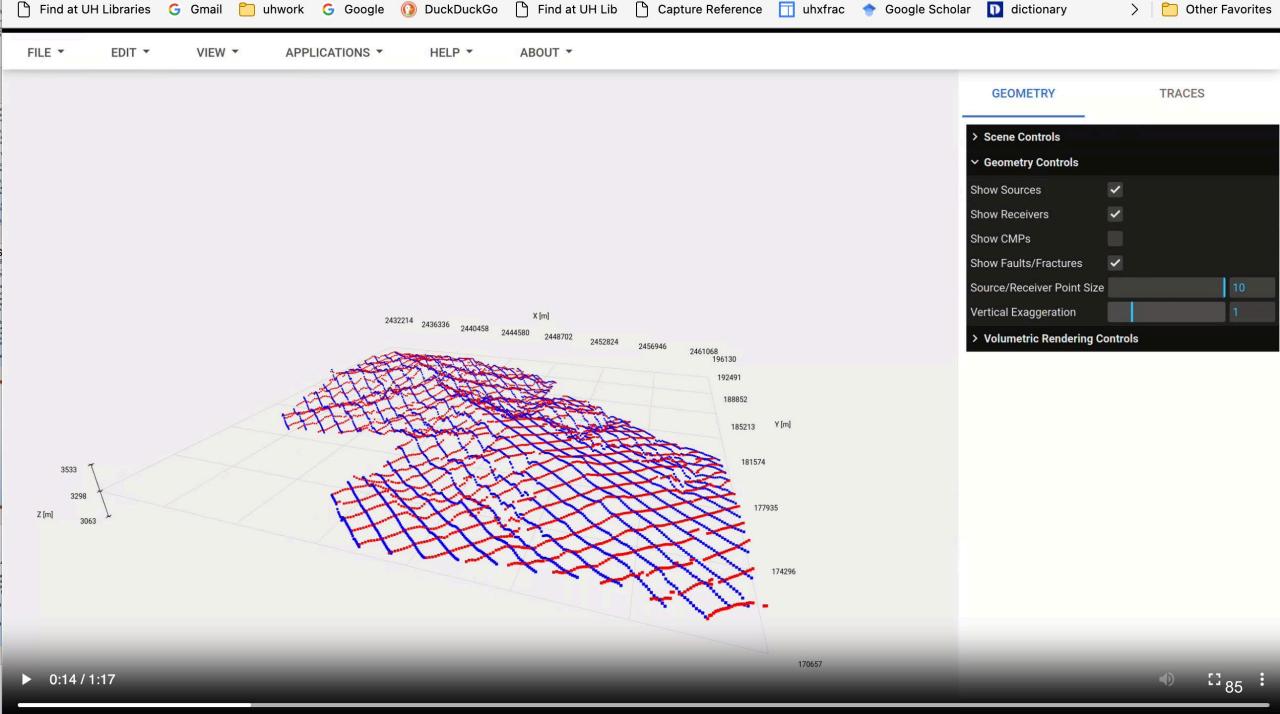
Perform eDB fracture characterization for P-P, P-S1, P-S2. Also use the neural network (DBNN) to obtain discrete fracture networks.

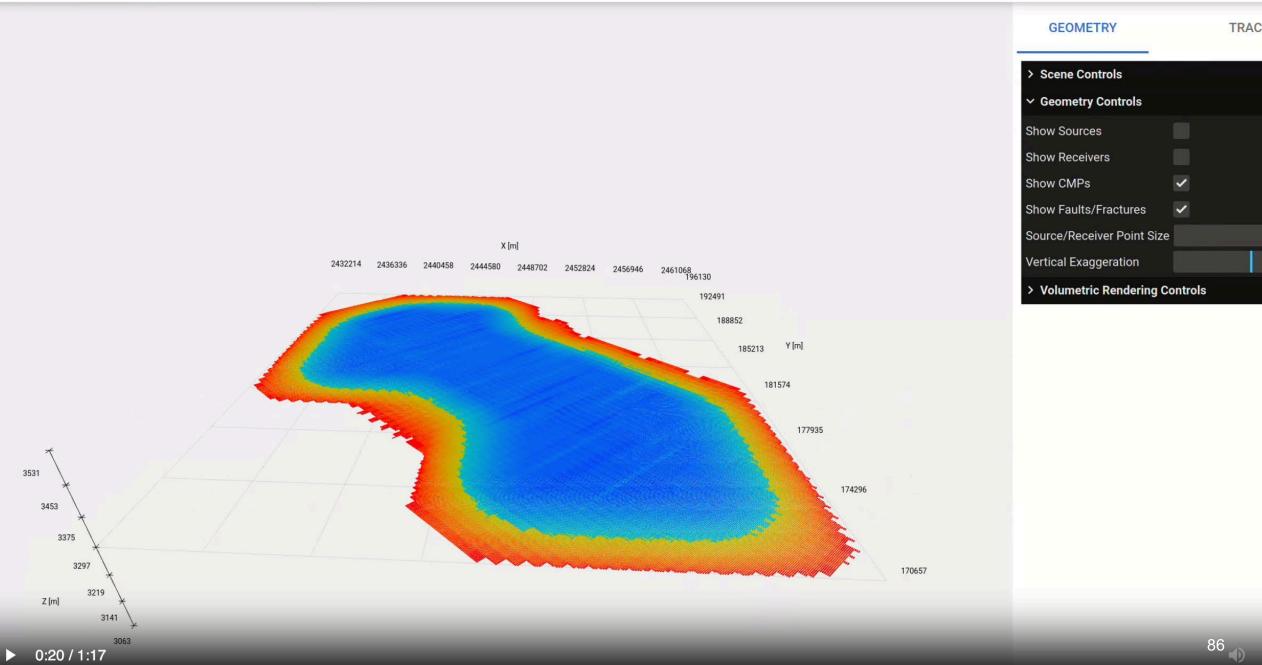
•<u>Subtask 3.4</u> – (LANL) Enhancing images in the basement using elastic deconvolution migration imaging (M16-M24) LANL will perform P-P, P-S, S-P, and S-S imaging on the field dataset.

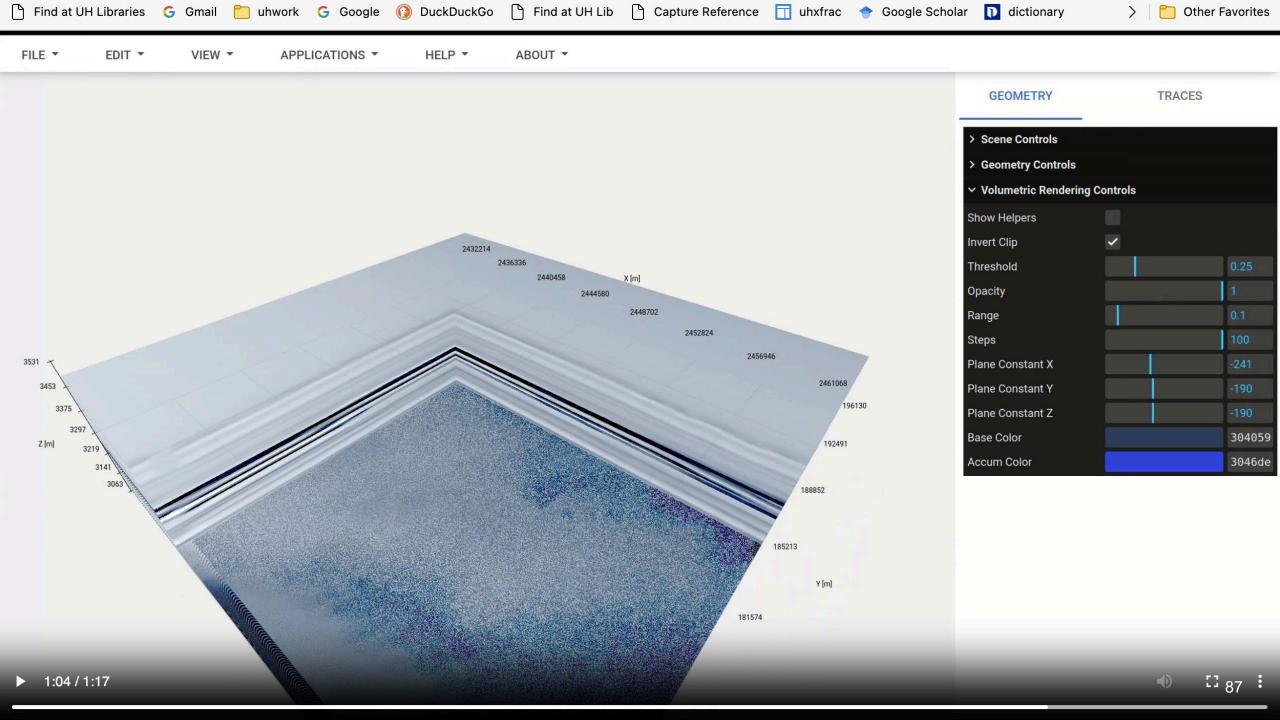
•<u>Subtask 3.5</u> – (LANL) Fault detection using machine learning (M16-M24).

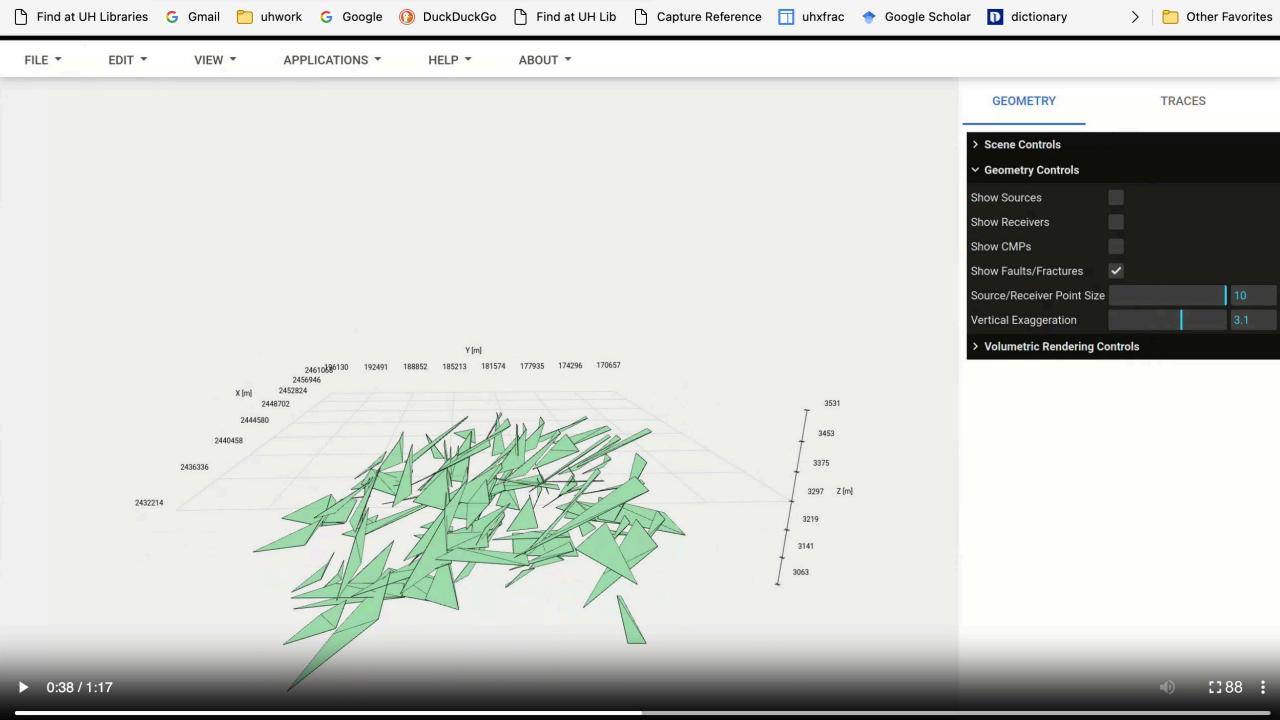
# Deliverable: Software development

• Web browser for seismic model/data visualization







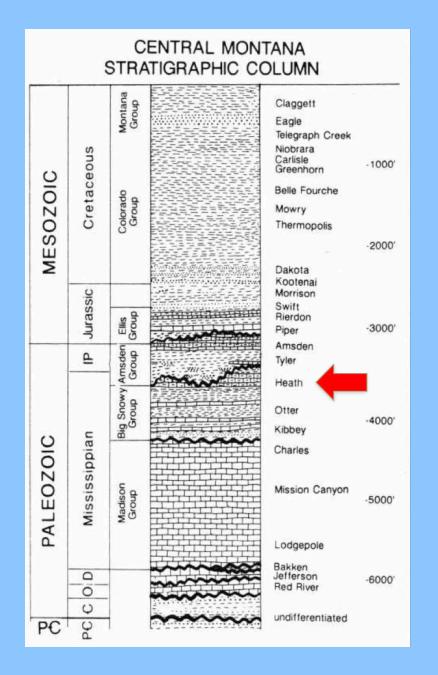


# Summary of Accomplishments

- Synthetic model and data tests; **DONE** 
  - UH, LANL, and Vecta Oil and Gas Itd. worked together and built a 3d seismic model: Vp, Vs, density and spatially varying fracture networks including conjugate fracture sets DONE
  - We modeled 3d 9-c shot gathers DONE
  - We applied the double-beam method on the modeled datasets DONE
    - If there are fractures, DB can invert for the true fractures
    - If there is on fracture in the model, DB reports 'no fracture'
    - Different frequencies give consistent results  $\rightarrow$  DB method is self-verifying
- BP-2: field data (Jan 1, 2023 December 31, 2023, revised to June 2024)
  - State-of-the-art oil/gas industry methods (shear wave splitting) DONE
  - Shear wave splitting **DONE**;
  - Double-beam neural network (DBNN); with preliminary results
  - Faults detection; with preliminary results; overall status
- BP-3: synthesis: fault stress state & activation potential

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- We also used computing facilities provided by the UHXfrac group at the University of Houston.
- Vecta Oil and Gas ltd. provided field data.



Heath Is Well-Documented as the Major Oil Source Rock in Central Montana