



DAS and DTS Monitoring at MiDAS Taiwan 井下光纖於米崙斷層的觀測研究及未來應用

馬國鳳 (Kuo-Fong Ma^{1, 2}) , and MiDAS team

¹Institute of Earth Science, Academia Sinica

²E-DREaM Center, National Central University, TAIWAN



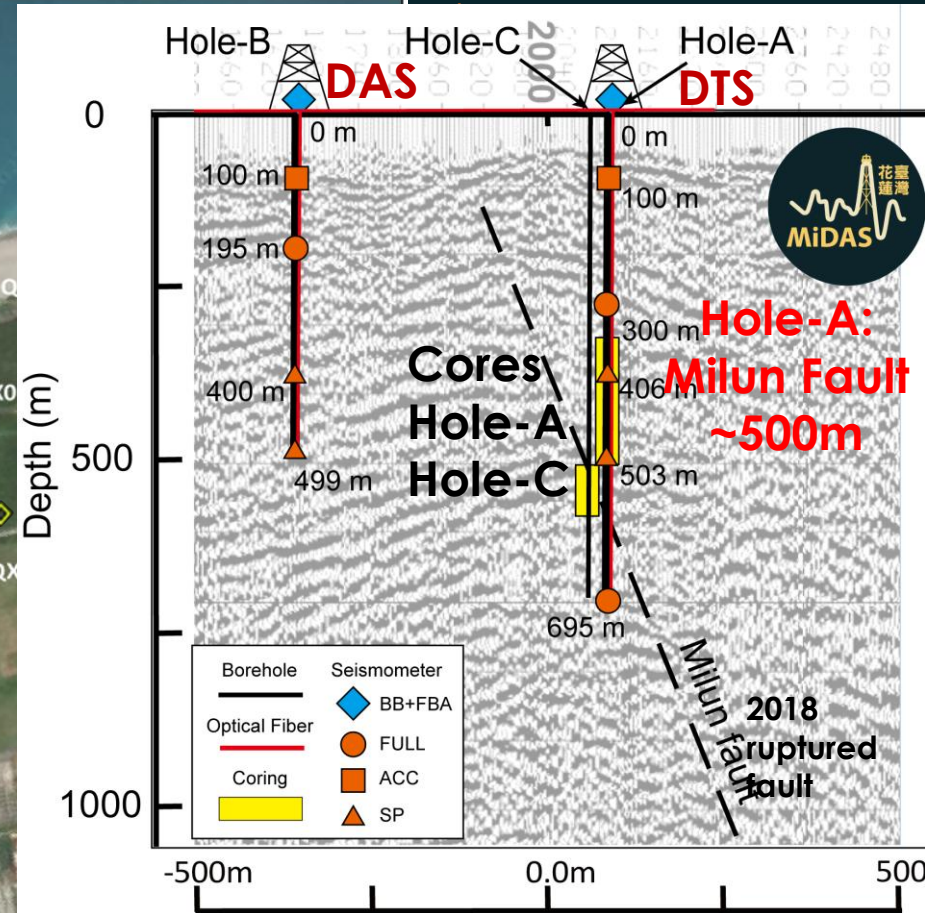
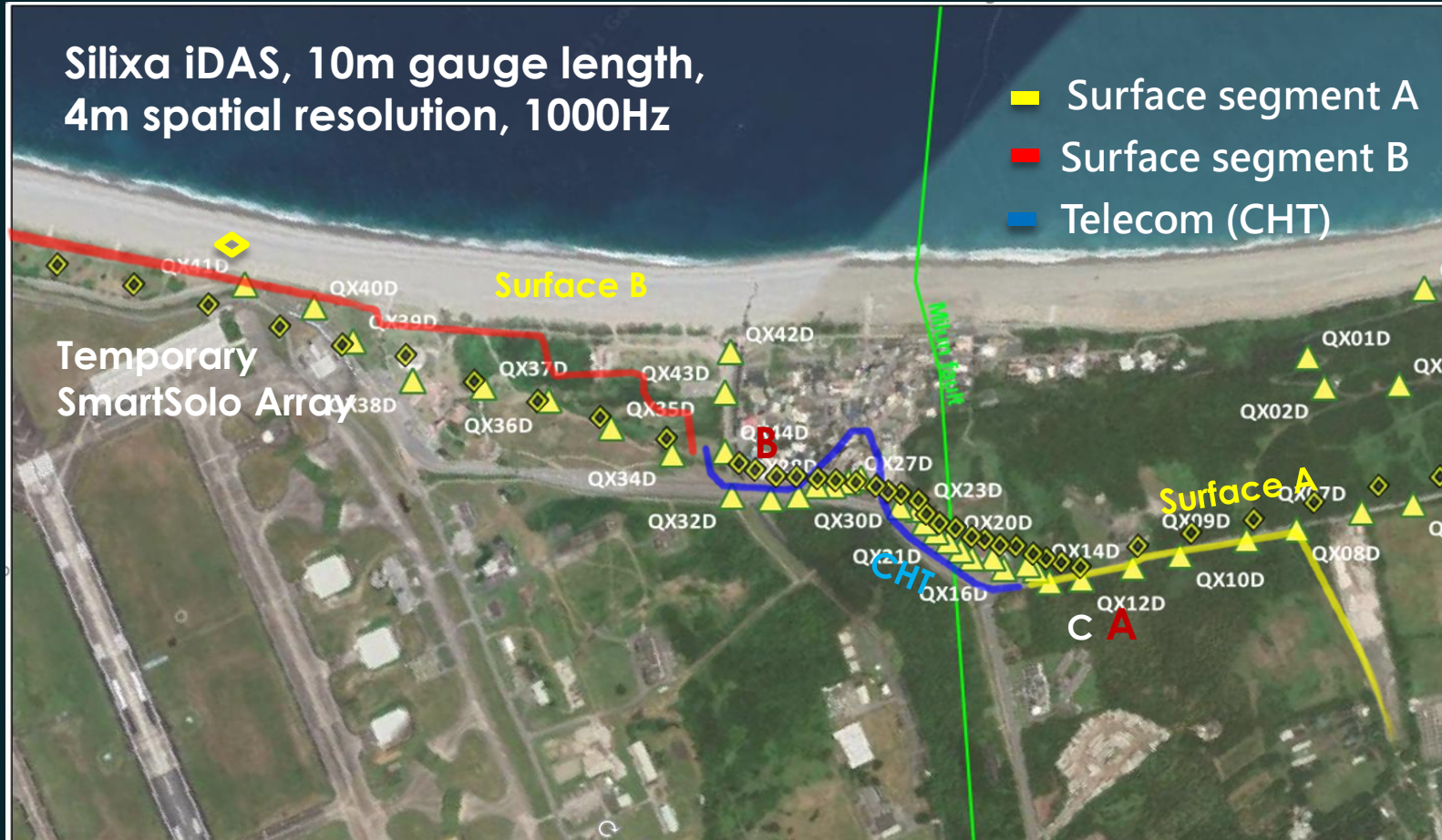
Milun fault Drilling and All-inclusive Sensing (MiDAS)

crossing fault holes: 700m Hole-A, footwall 500m Hole-B

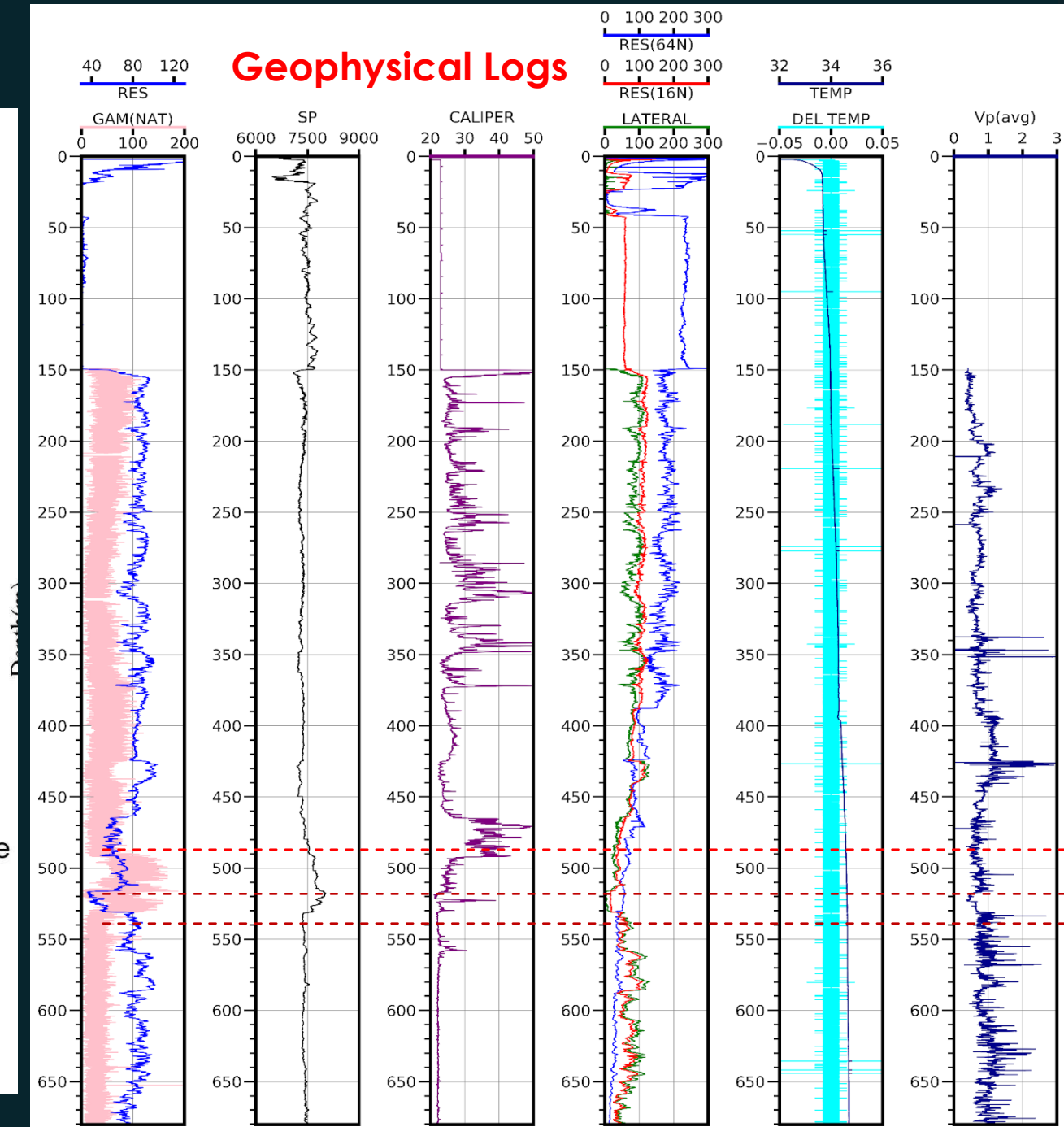
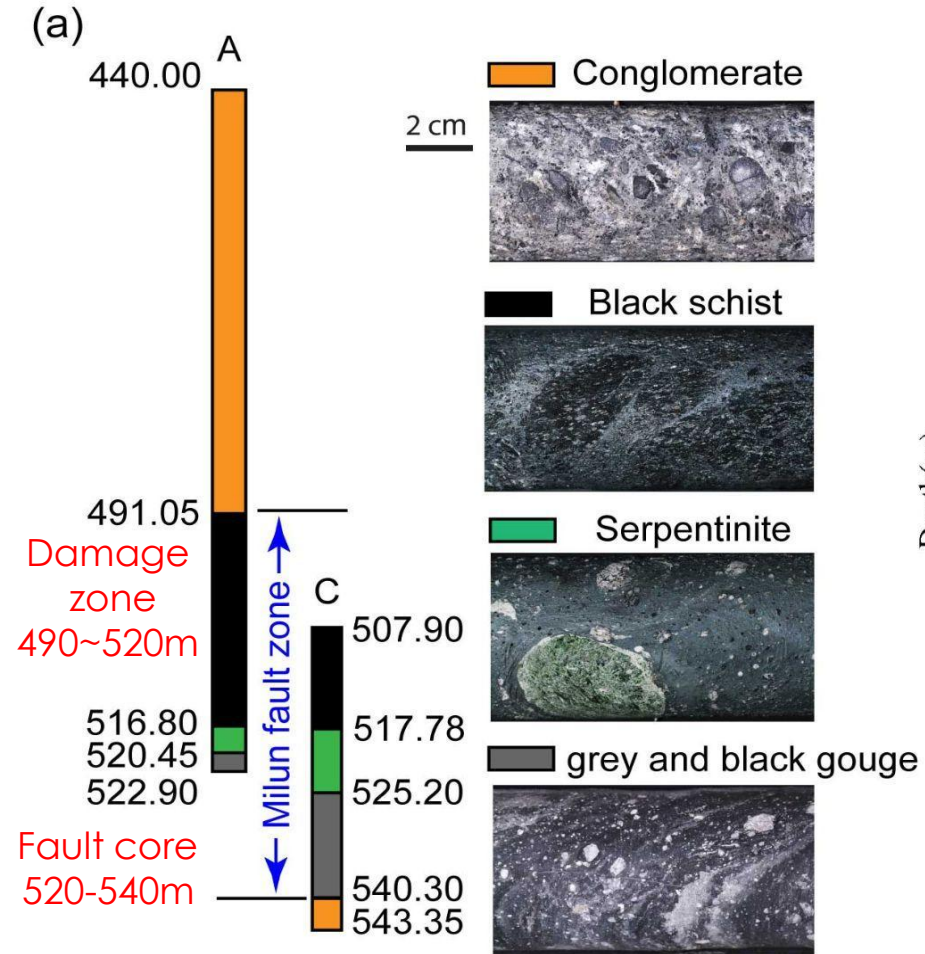
Hole-C: Fluid and gas monitoring



Crossing fault three-dimensional optical fiber configuration (HoleA, 0-700m-0, Surface A (forth-345m-back), Commercial Telecom, CHT (1052m), HoleB, 0-500m-0, Surface B (forth-1128m-back))



Retrieved Cores



conglomerate

Damage zone
fault core

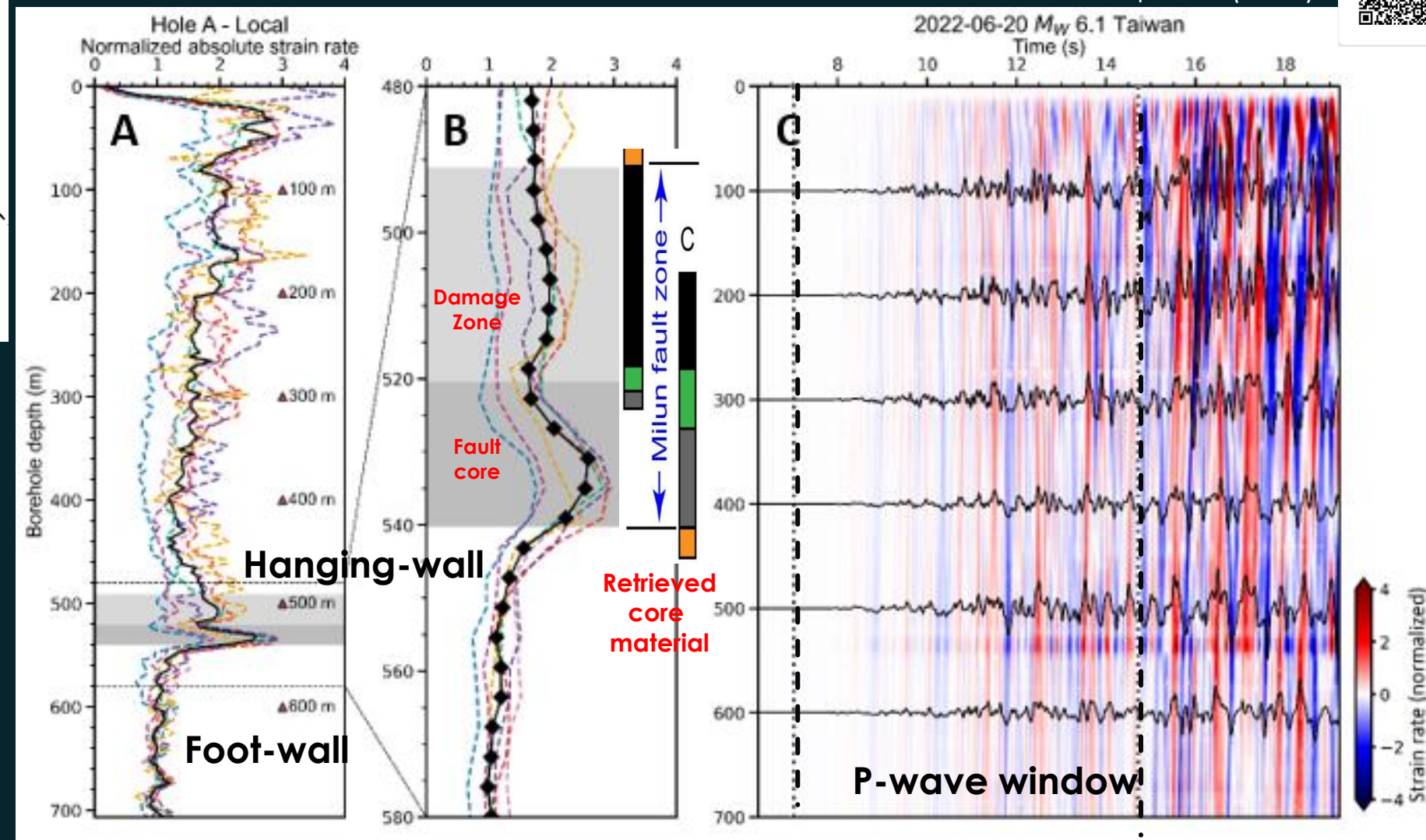
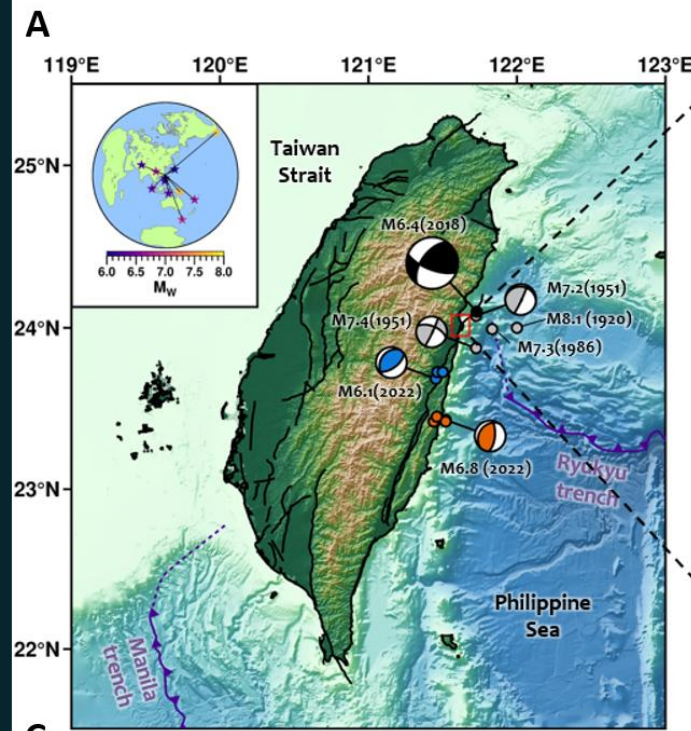
solidated
conglomerate



Downhole Crossing fault Fiber Sensing at depth Observation from Local Earthquakes

Broad-band strain amplification ~ 2.5 in the fault zone due to low rigidity

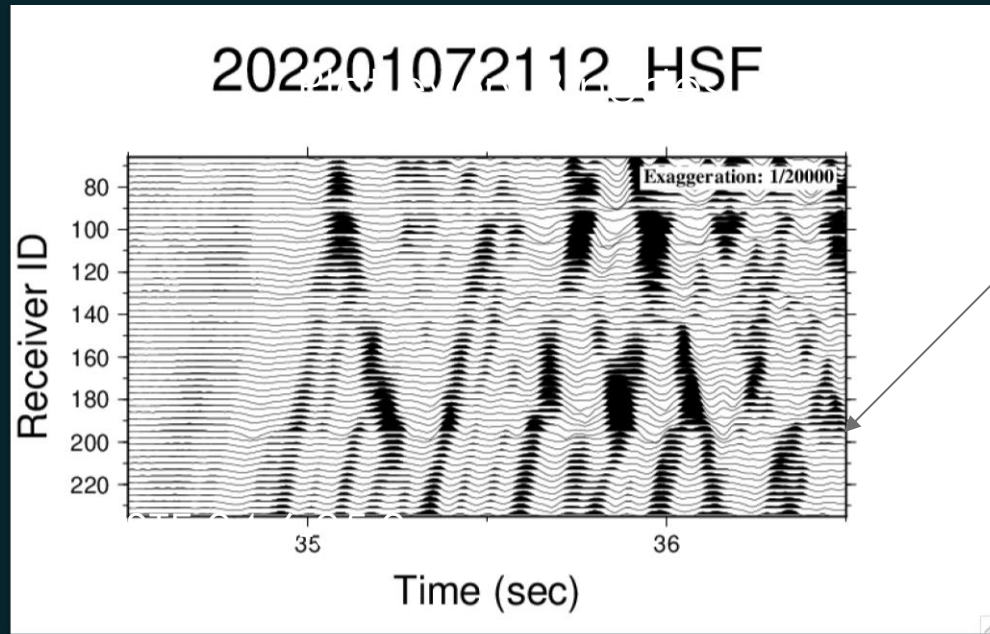
Ma and von Specht (2024)



DAS (Distributed Acoustic Sensing) strain observations are highly sensitive to changes in **rigidity**, which are influenced by **fluid dynamics** and **temperature variations**.

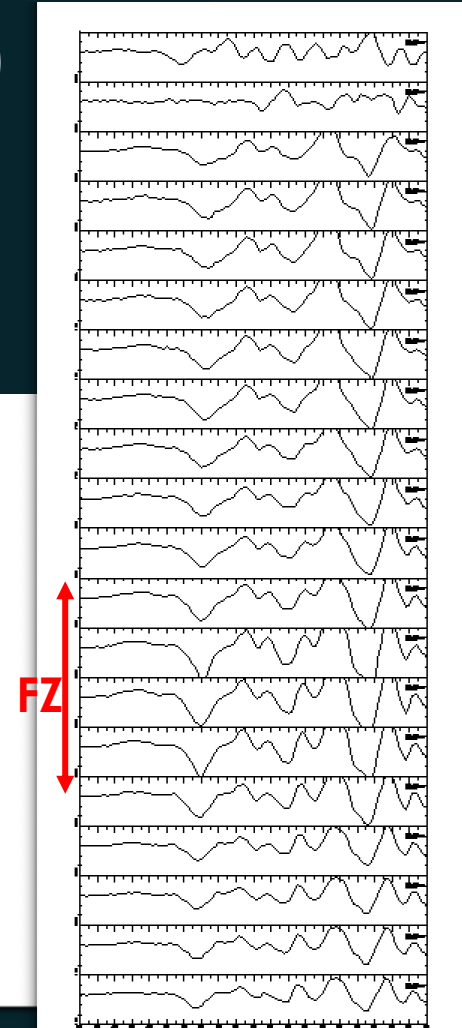
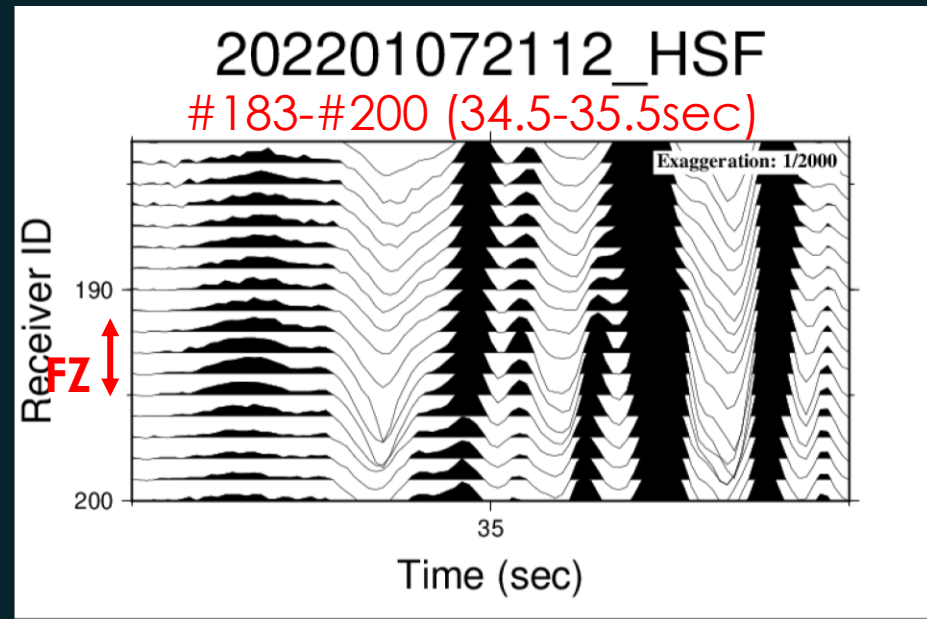
These parameters are critical for effective CCS (Carbon Capture and Storage) monitoring.

How does strain (rate) vary across different lithological layers? (Identifying the lithological layers, for CCS monitoring, using event based and ambient noise)



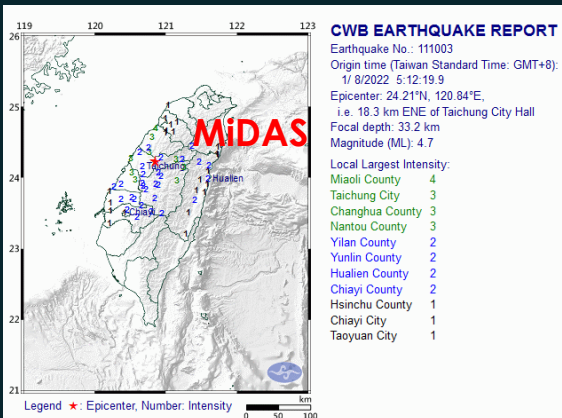
Sediment thickness
Disturbance in waves (Fracture Zone?)

Major Fault Zone LVZ to amplify the phases
More consistent features in the footwall



FZ

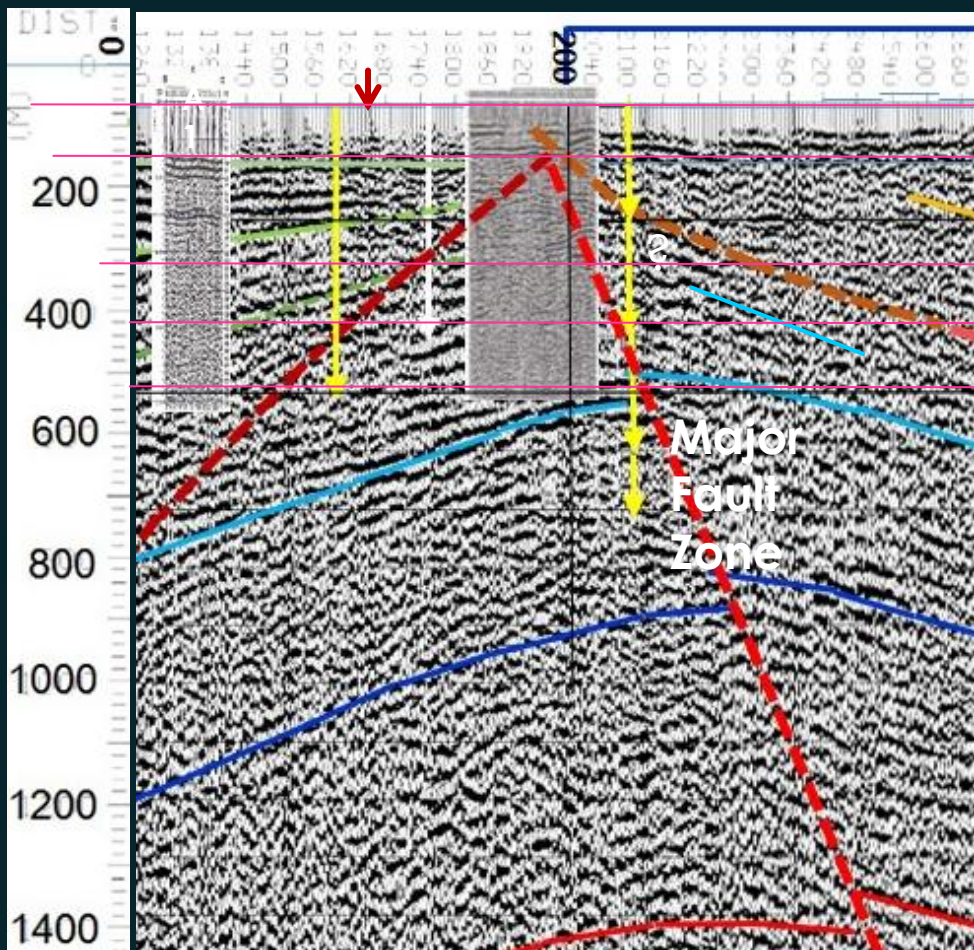
Direct Observation by DAS at Hole-A
20220107 M4.6 D~33km



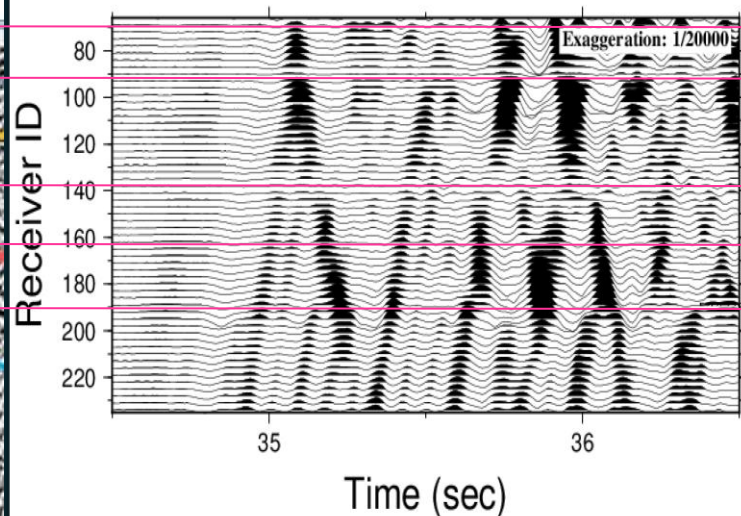
FZ amplification, and change in period
Fault Zone, 520-540m~ 20m

Fault Zone Features on Optical Fiber (strain-rate)

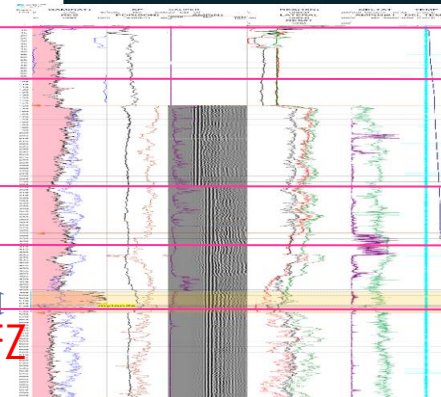
Seismic Reflection Profile



202201072112_HSF



Geophysical Logs (gamma, res, etc.)



Casing

Lithology I

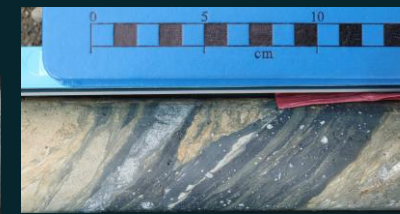
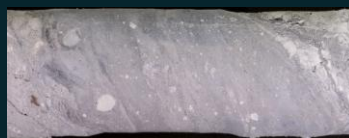
Lithology II

Lithology II

Fracture Zone within Lithology II
Major Fault Zone along the lithological boundary.

Fault Zone

Retrieved core



LW Kuo
WJ Wu

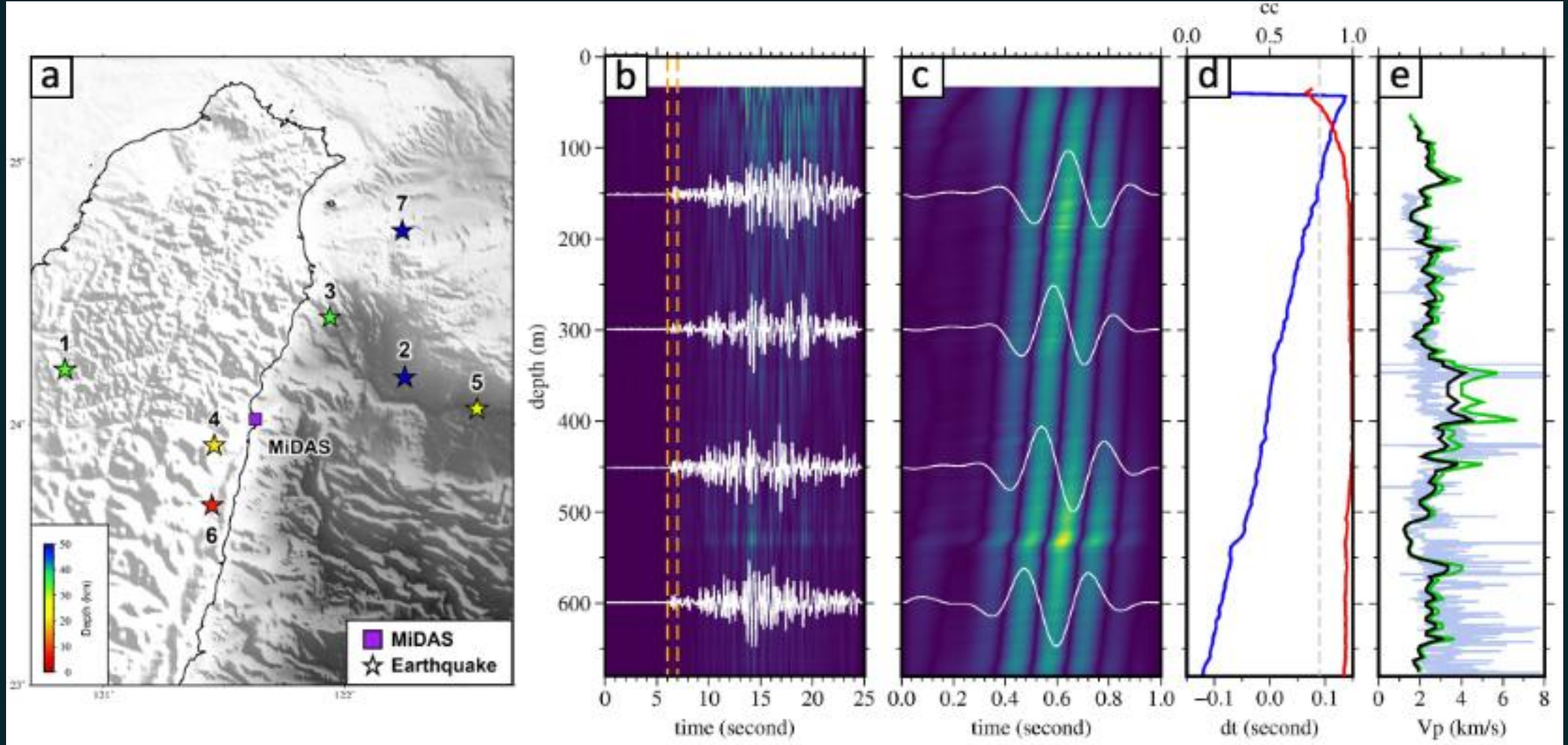
Fault core ~#160
@426m

Fault core #183
@491.05 – 491.15

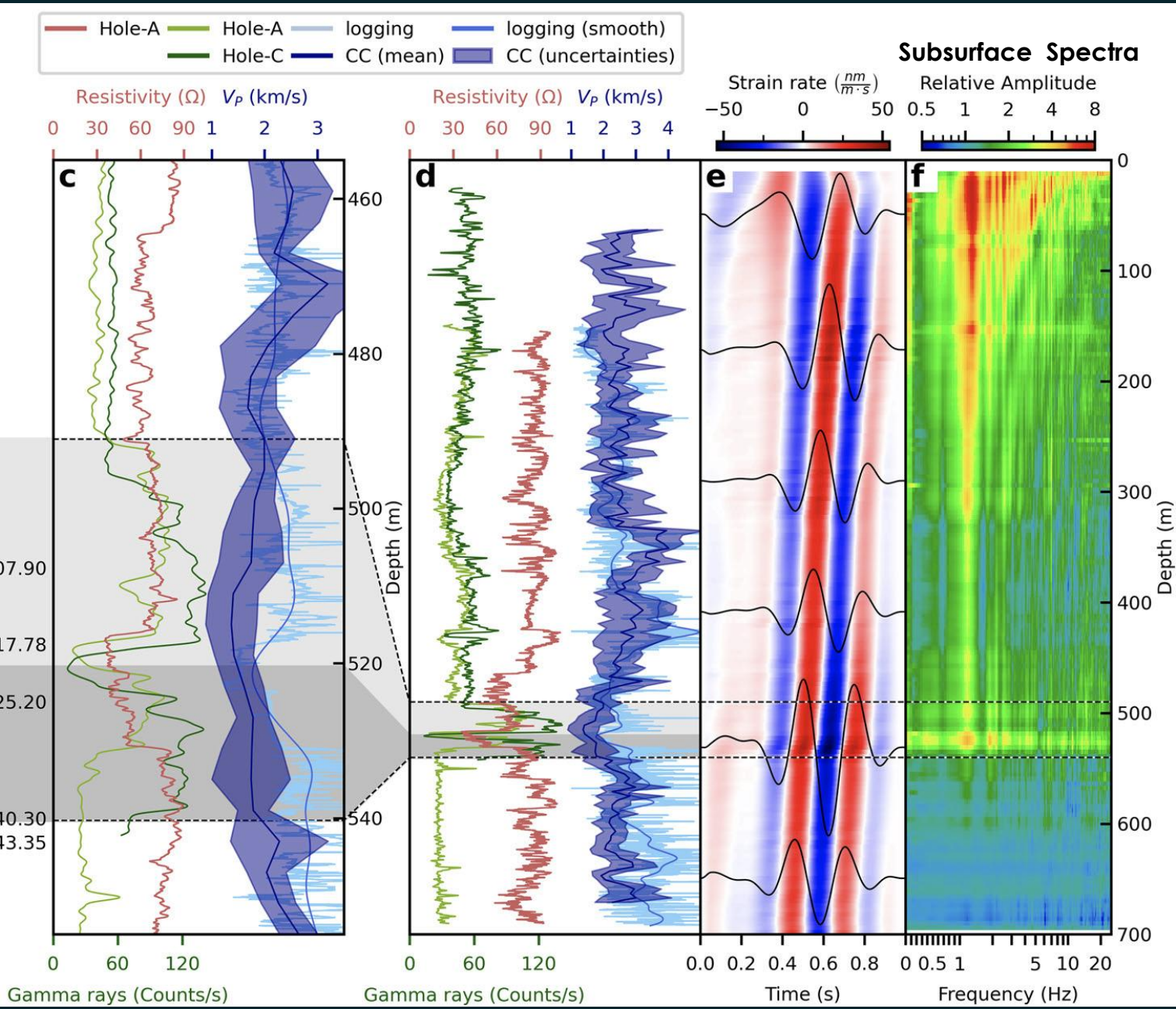
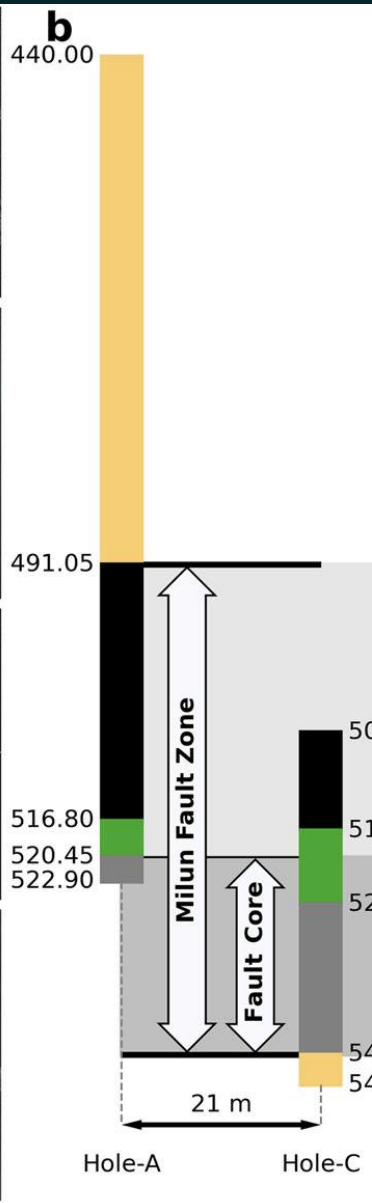
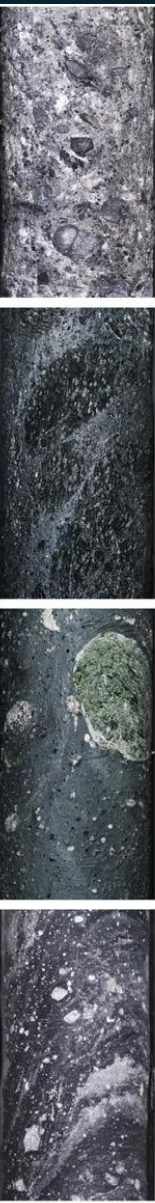
#190
@516.70-516.85

@522.45 – 522.55
~#192

Sensing fault zone velocity structure using cross-correlation through downhole optical fiber

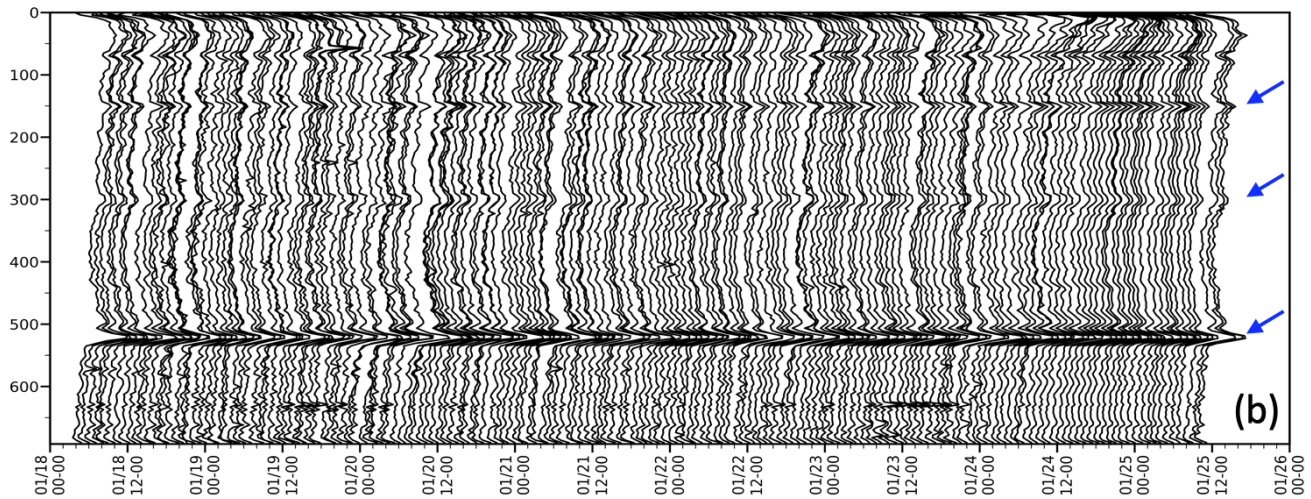
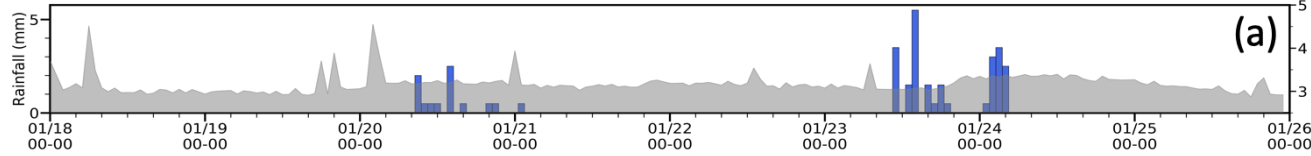
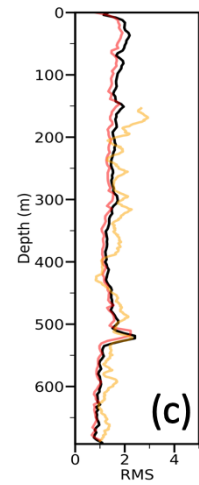


Conglomerate
Spotted Schist
Serpentinite
Fault Gouge



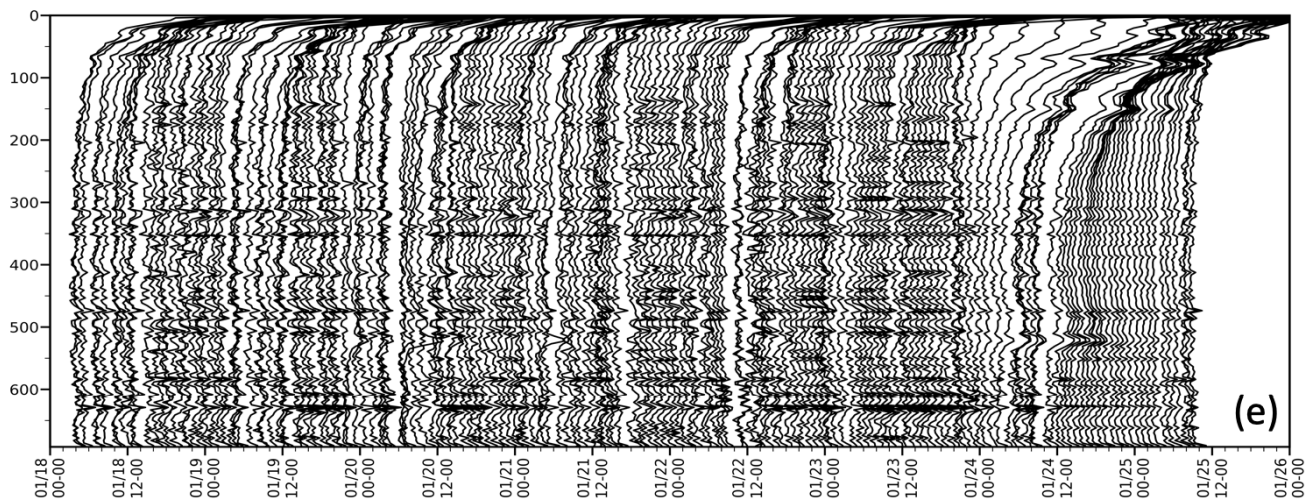
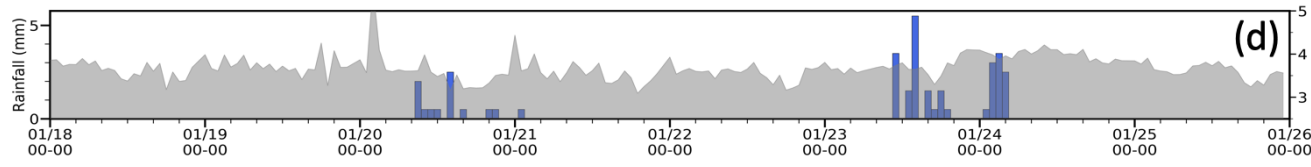
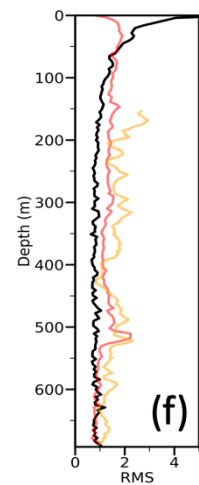
0.1-1 Hz

— Mean RMS
— EVT RMS
— Normalized logging slowness



1-30 Hz

— Mean RMS
— EVT RMS
— Normalized logging slowness



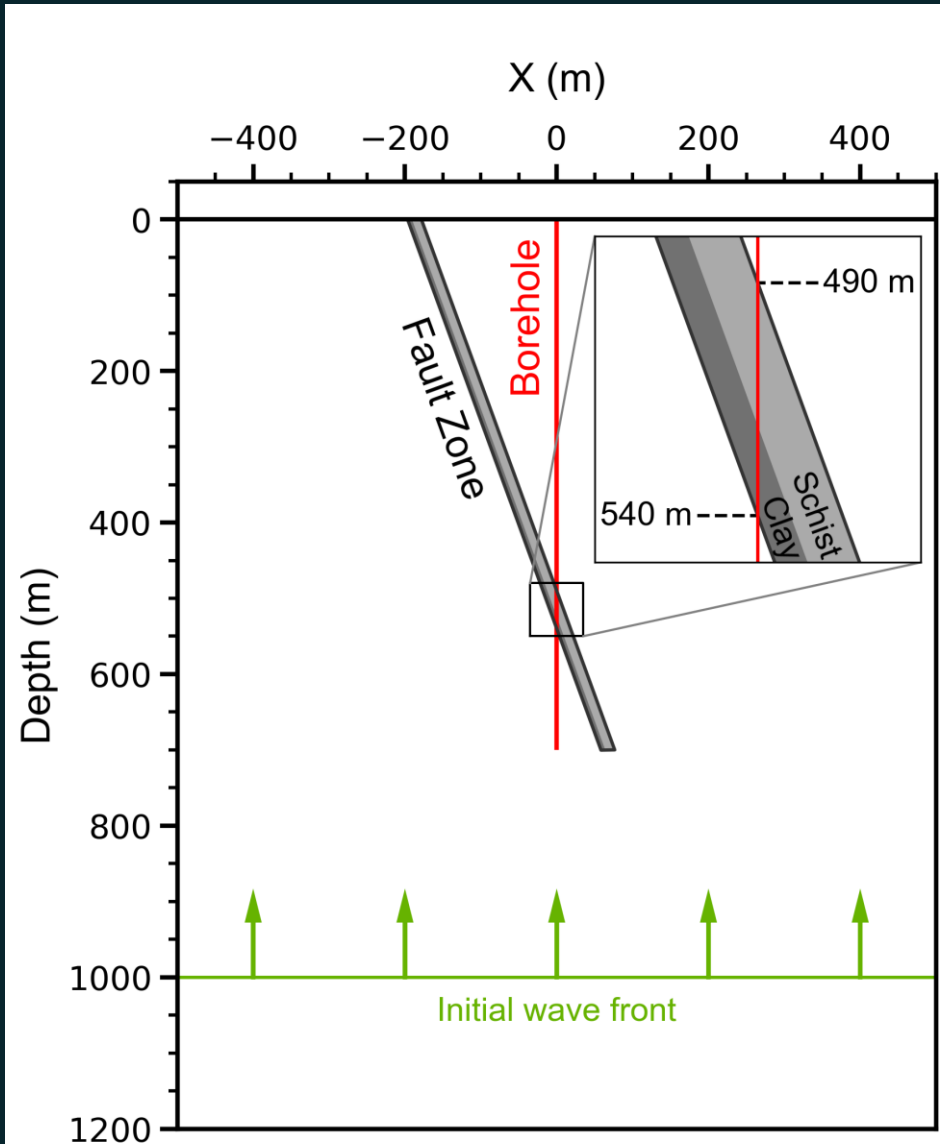
Ambient Noise Long-term temporal RMS amplitude ratio profiles at Hole A from 18-25 January 2023 in (a-c) 0.1-1 Hz and (d-f) 1-30 Hz frequency bands. (a) and (d) show the hourly precipitation in blue bars and the hourly peak ground motions in gray. (b) and (e) show the hourly moving-window calculation of RMS amplitude ratios in corresponding frequency bands. (c) and (f) show the comparison between the mean RMS amplitude ratio profile (black), mean RMS amplitude ratio profile derived from earthquakes, and logging P-wave slowness (1/velocity) profile (orange). **Blue arrows indicate the low velocity zones or rigidity contrasts.**

Huang, Ma et al. (AGU books, Distributed Acoustic Sensing in Borehole Geophysics, 2024)

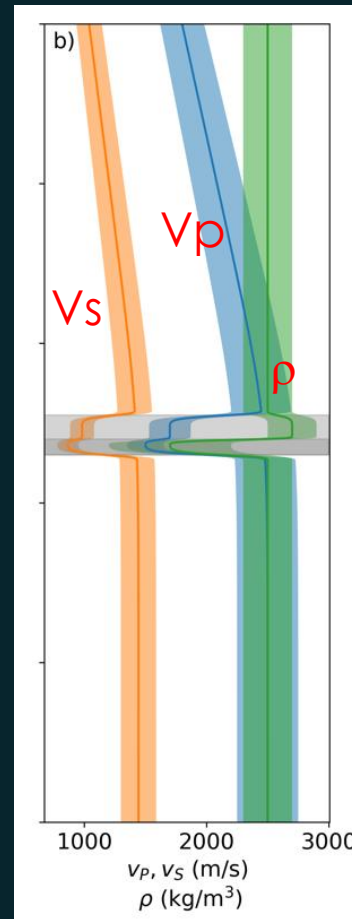
2D finite-difference simulation, heterogenous medium in elastic modulus,

$dx=dy=2m$; $dt=0.00015$ sec, Source Time Function, Gaussian derivative

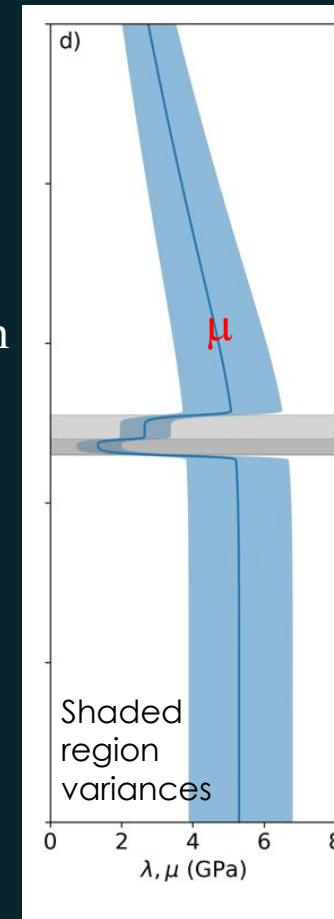
Velocity models with variances, Logging and OFS through CC



Velocity models, and elastic constant (rigidity) profile
(Assuming Poisson material)



Schoen
(2015)

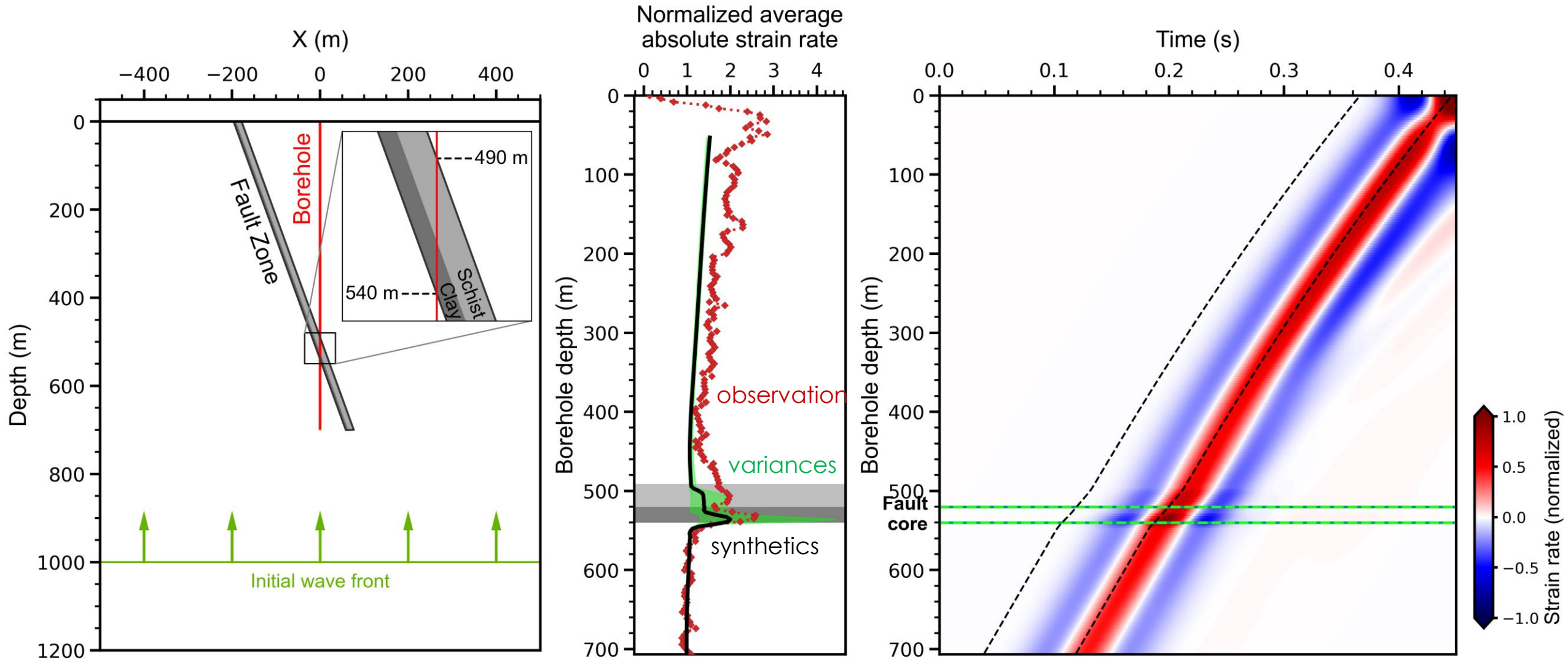


Reduction
within fault zone
Damage zone
Fault core

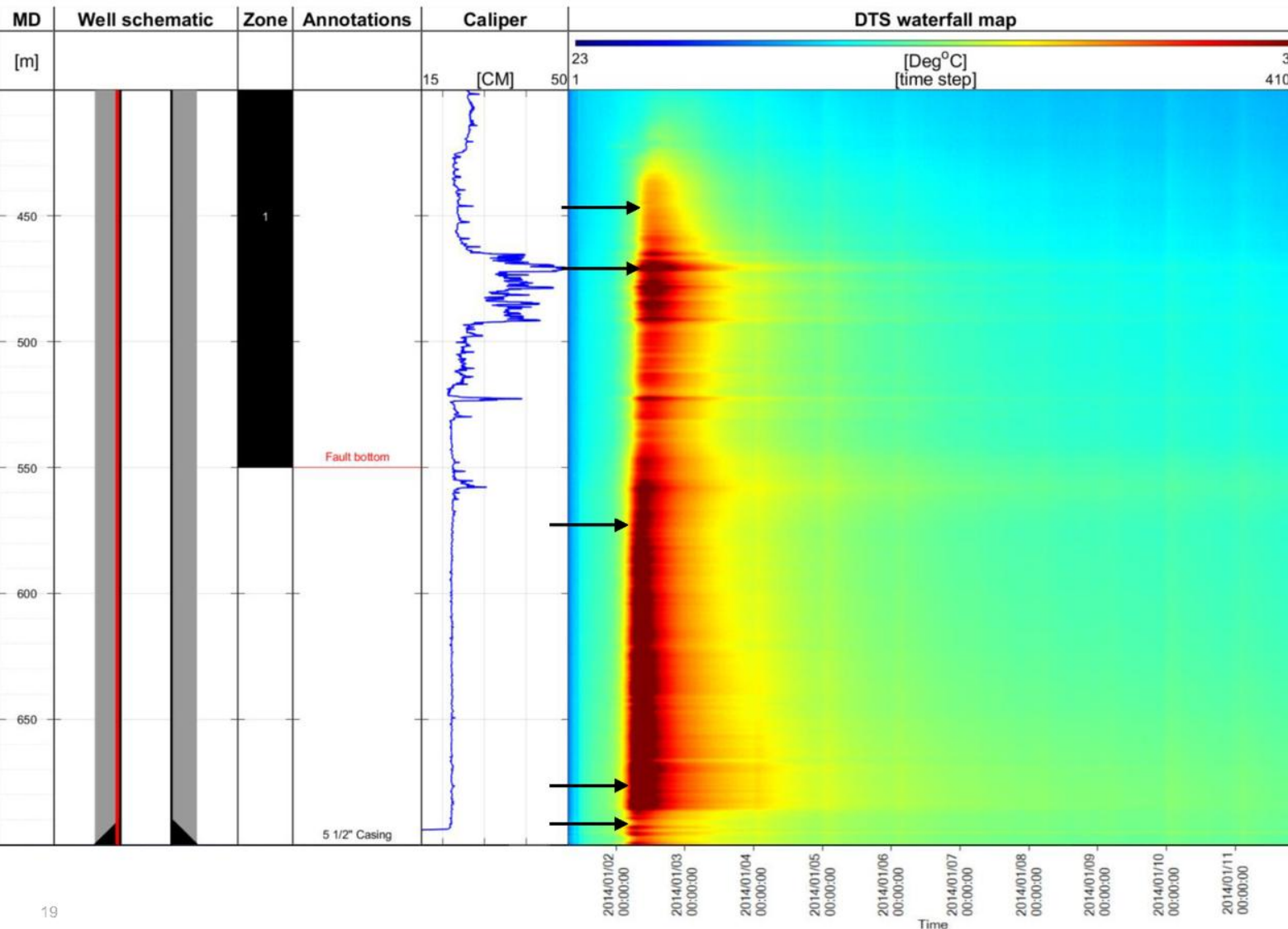
- Strain is sensitive to rigidity changes

=> Optical fiber sensing for monitoring temporal rigidity changes (fluid, velocity changes, environment changes)

Simulation results
full displacement, taking
temporal and spatial derivatives to strain-rate)



From Sebastian von Specht

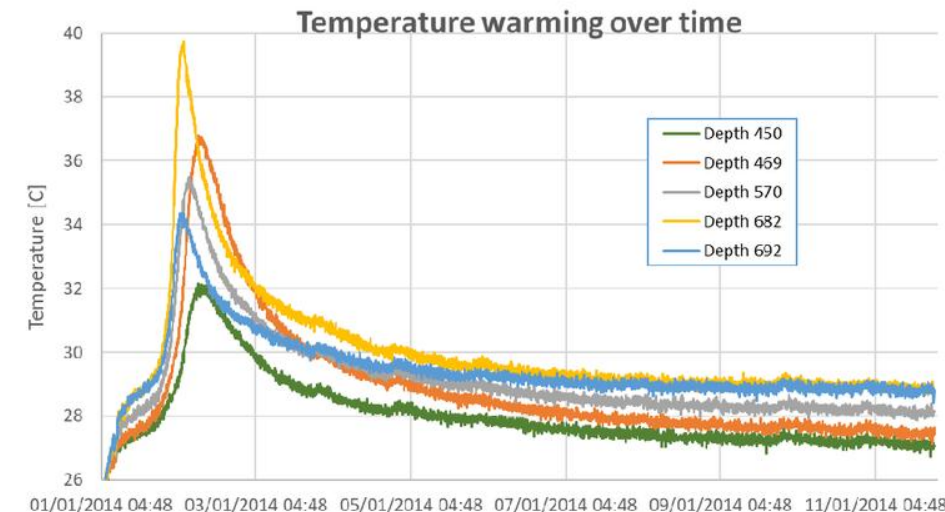


The following depths were selected to create a temperature warming over time comparison.

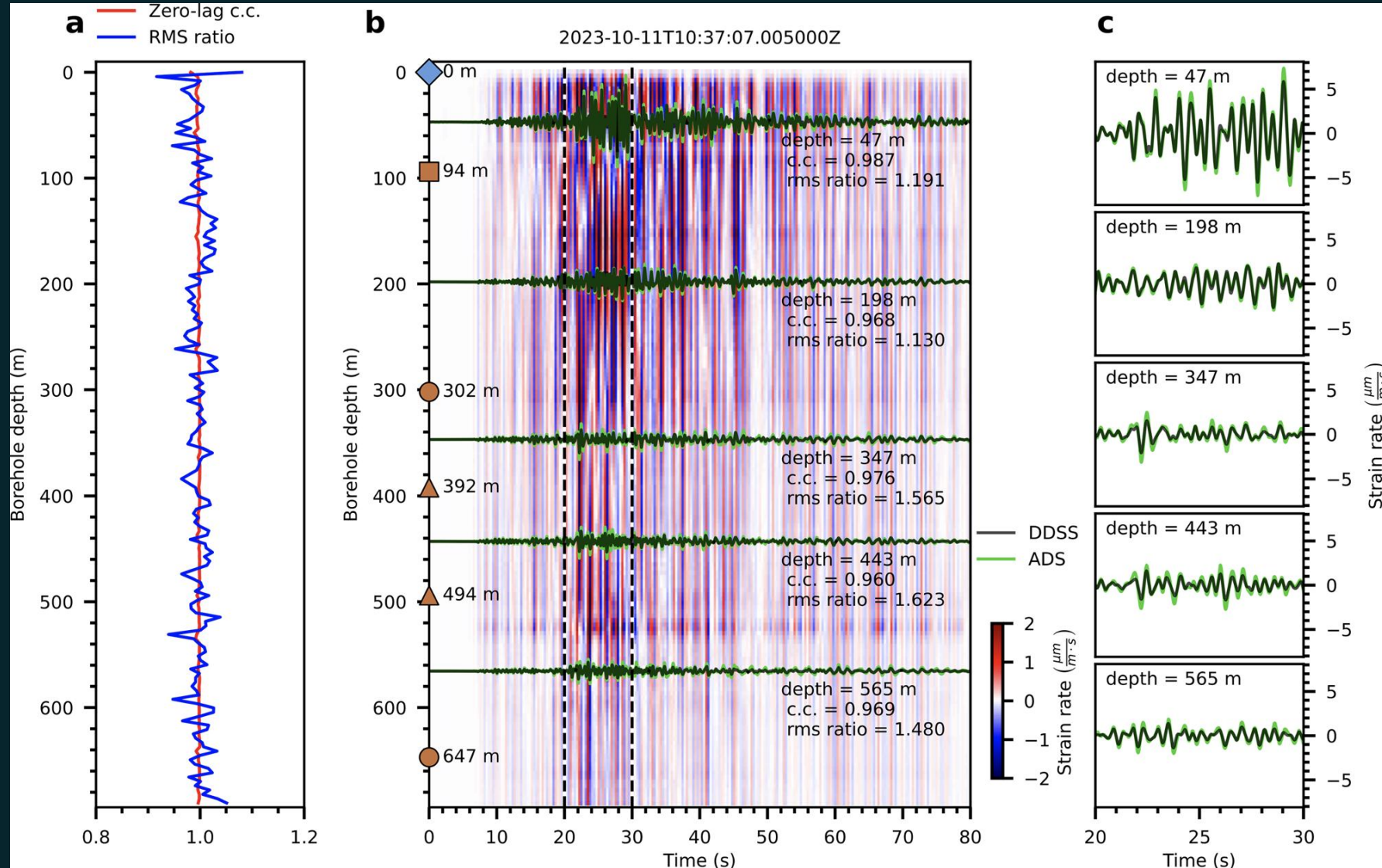
It can be noticed how after the cool-back [post cement curing warming] the temperatures go back to higher temperatures in the deepest points selected, which would agree with a vertical well temperature gradient.

It can be appreciated that even though 469m is a shallower depth, this one reaches higher temperature warming during cement curing than 470m and 692m.

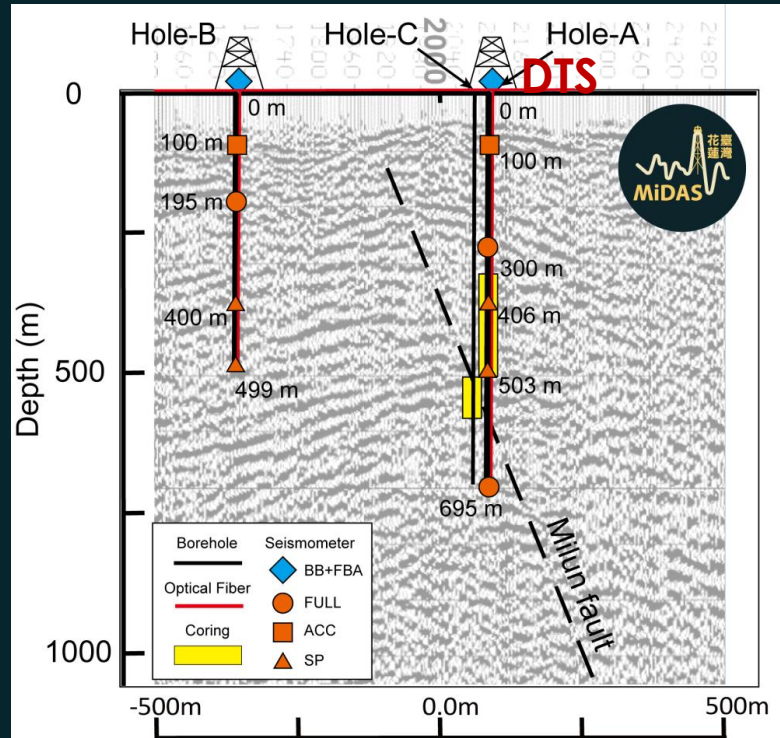
The depth that reaches the highest temperature can be seen at 682m.



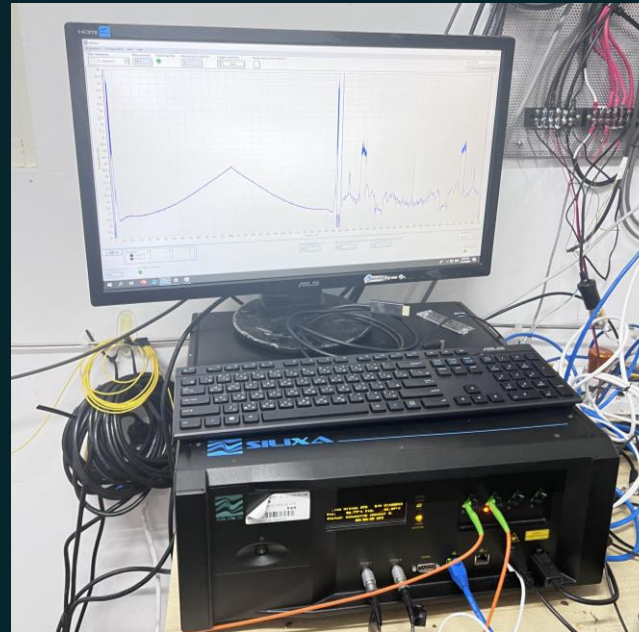
Reliability of the borehole optical fiber from DAS and seismic array derived strain: ensuring good quality of cementing and fading effects



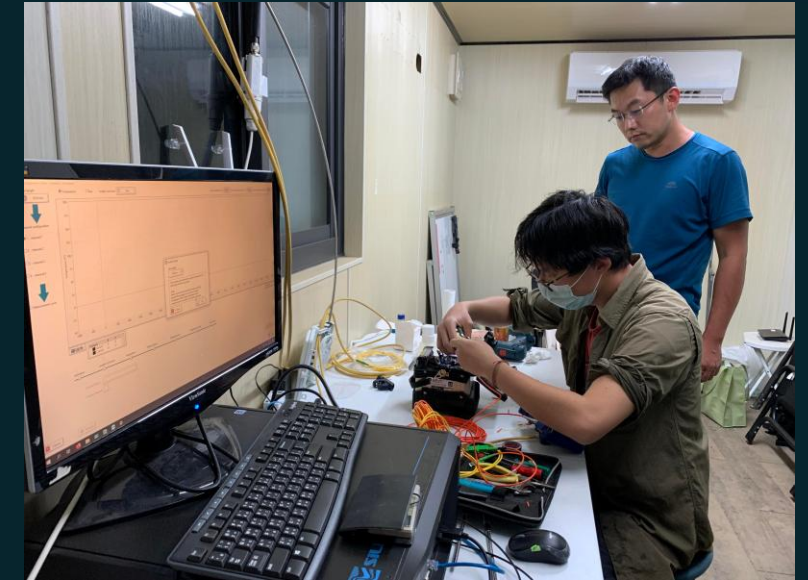
DAS/DTS observatory



DTS @Hole-A (0.254 m SR, 20sec SR)

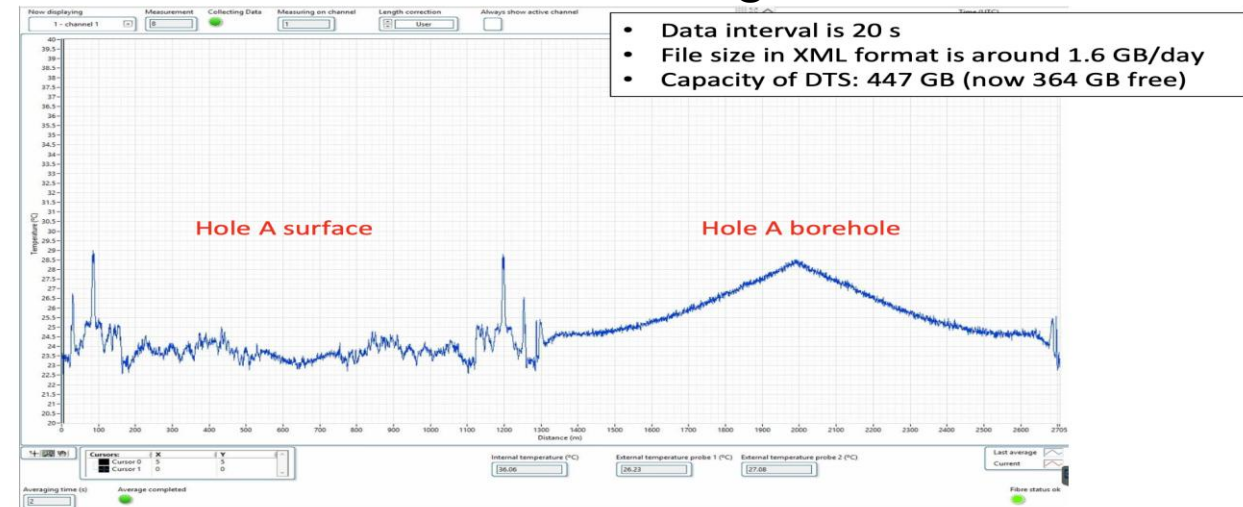


Final splicing for DAS loops

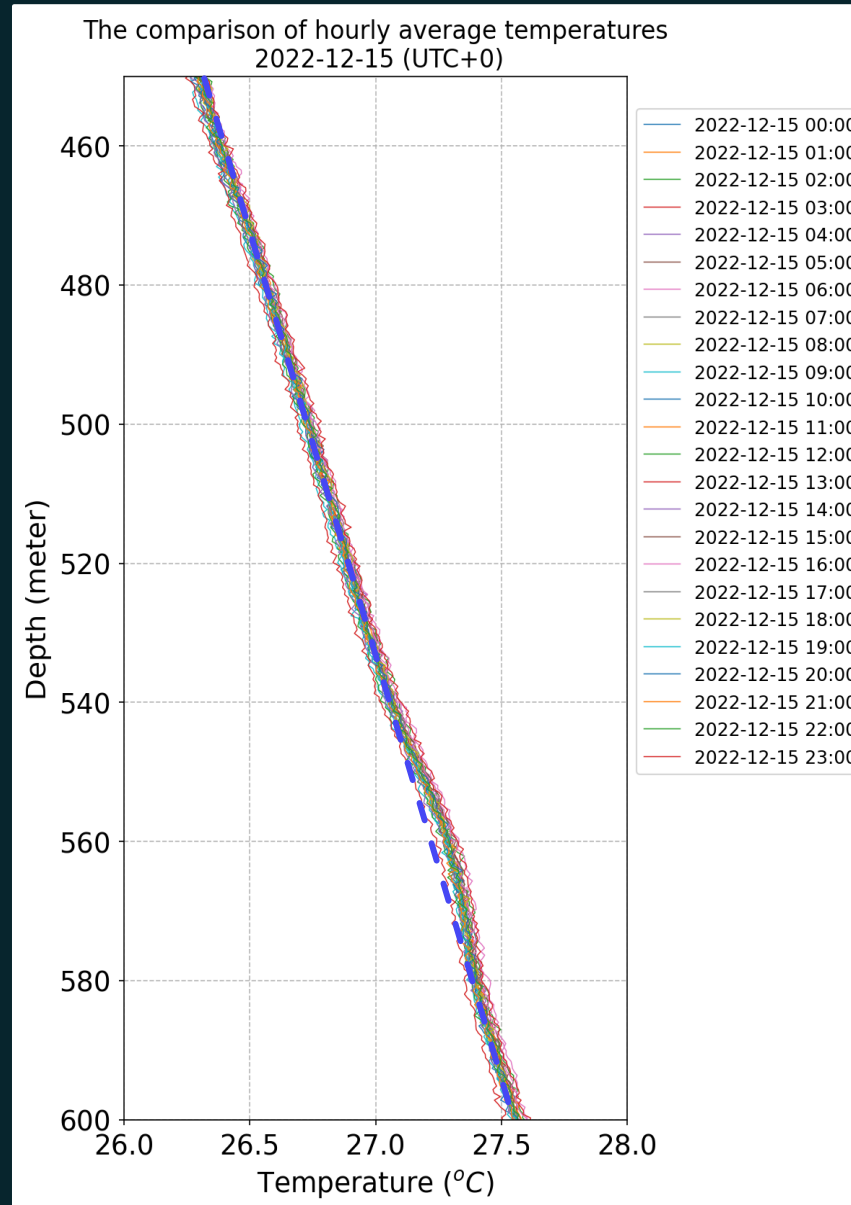
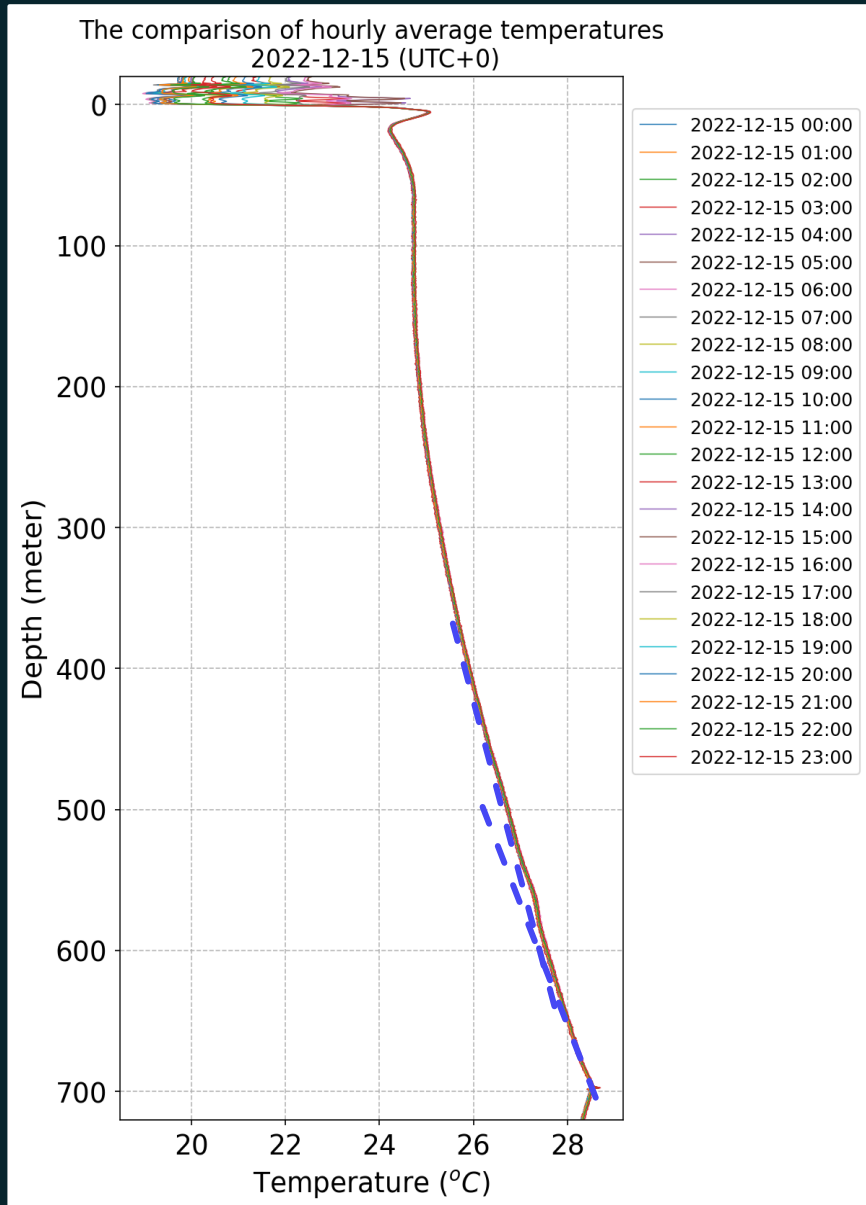


2022/11/15

DTS monitoring



DTS hourly average temperature profile



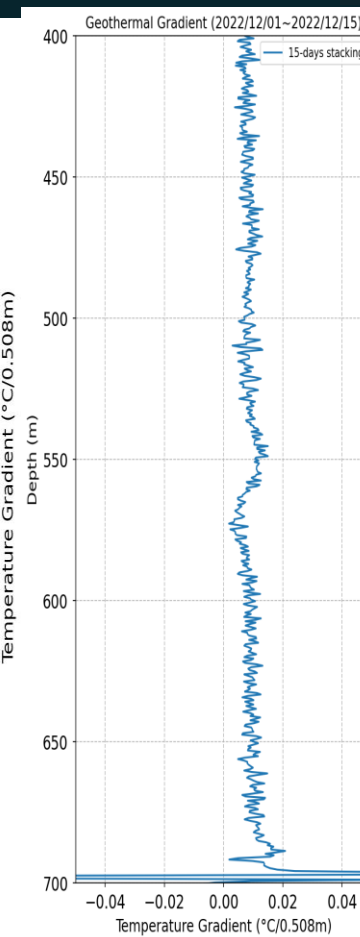
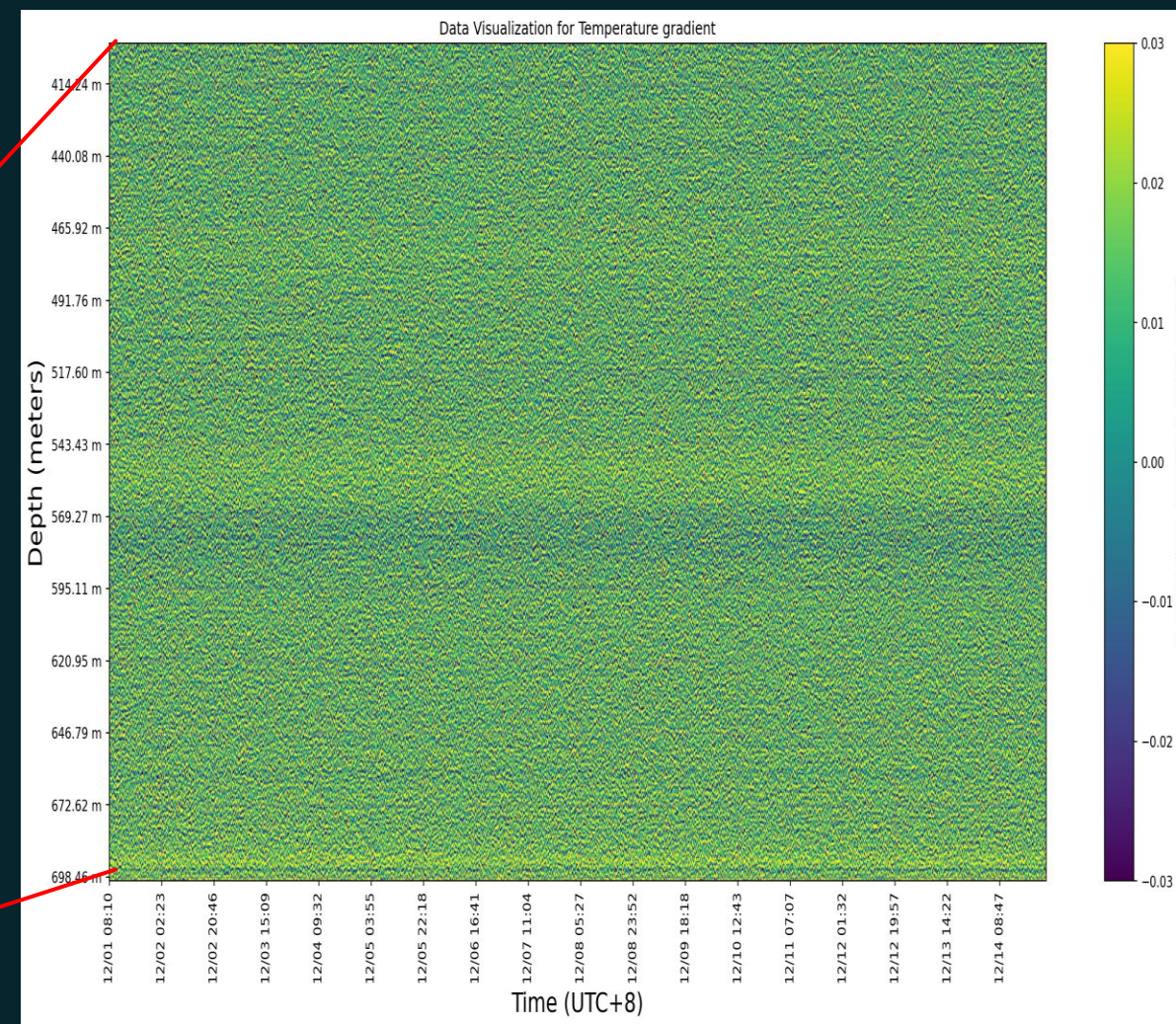
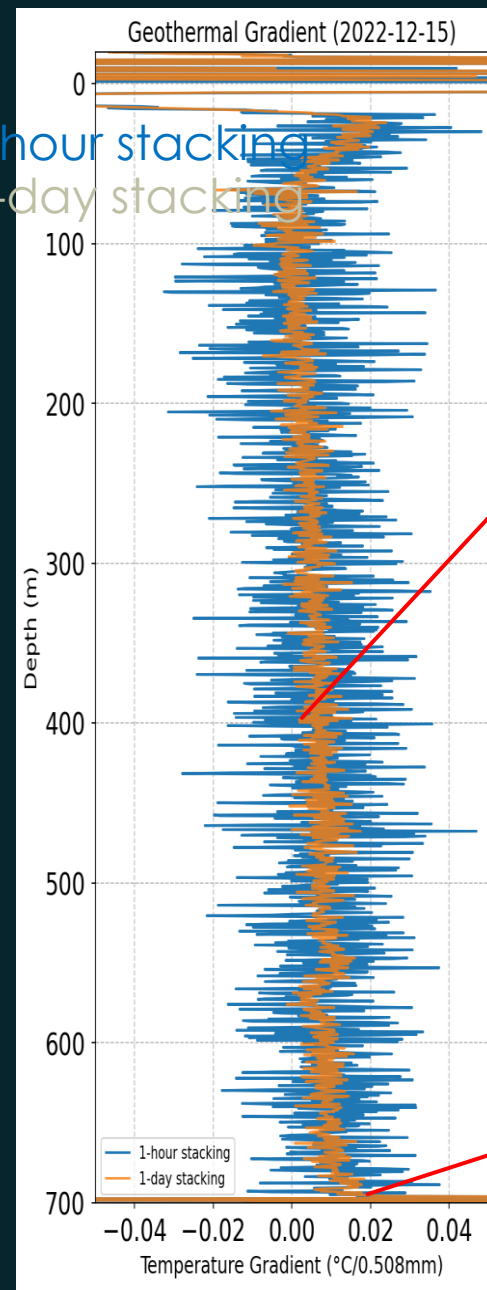
Background geothermal gradient

- Top 150m, environmental influence
- Disturbance deeper than 150m, might be crustal related features
- Nonlinear geothermal gradient
- The average temperature profile is rather stable regardless the variations with and without controlled head temperature
- Subtle temperature anomaly within the range of 550-580m
- Subtle geothermal gradient changes above and below the fault gouge zone

The changes in geothermal

Geothermal gradient map in the depth of 400-700 m (2022/12/01~12/14)

12/01 -----> 12/14



15-days sta

Geothermal Gradient changes below MFZ within the range of 540-580m

Borehole temperature profile in Hole-A

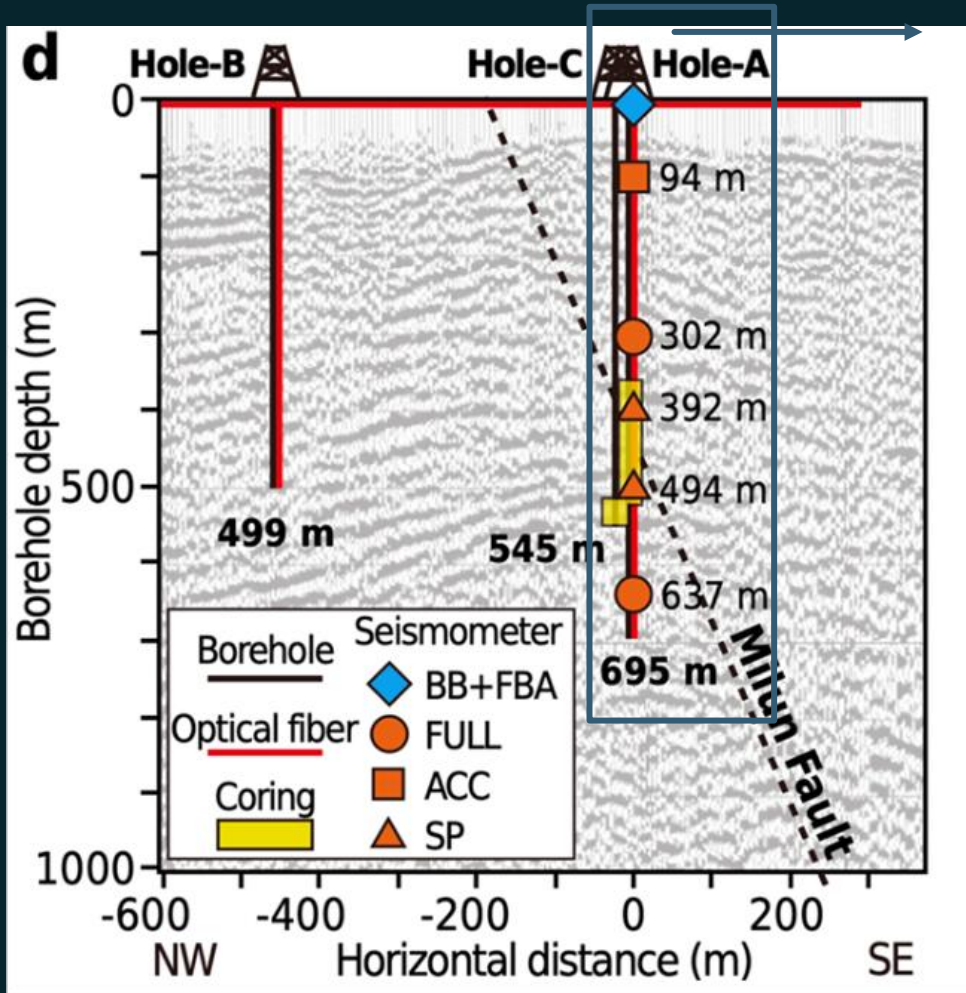
Distributed Temperature Sensing (DTS)

- Depth resolution : 0.245 m
- Temp. resolution : 0.01 °C

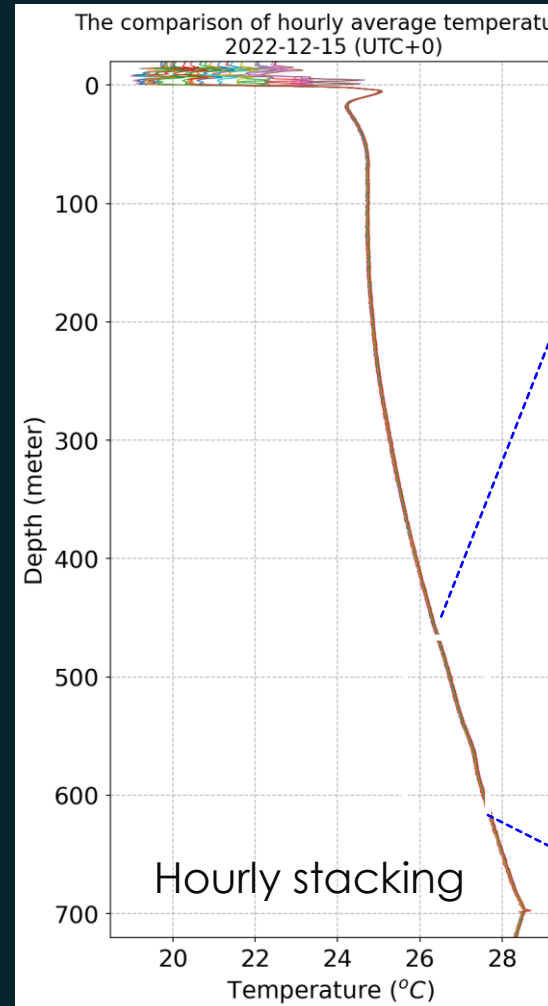


Since 2022 till now

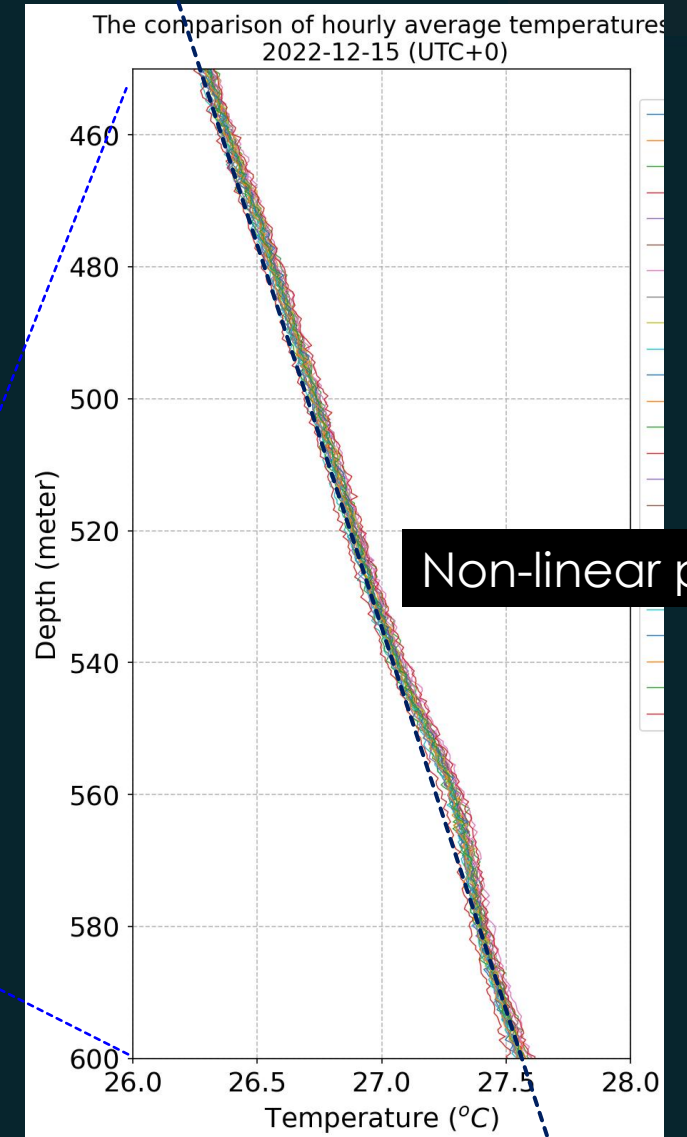
Surface (0m)



(Ma et al, 2024)

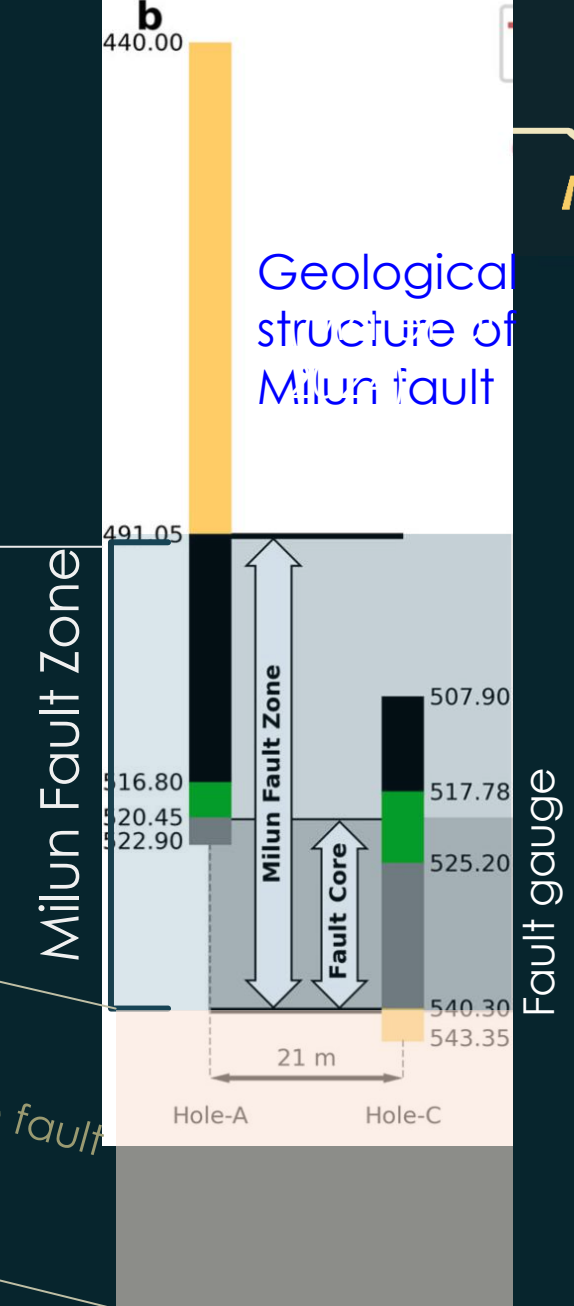
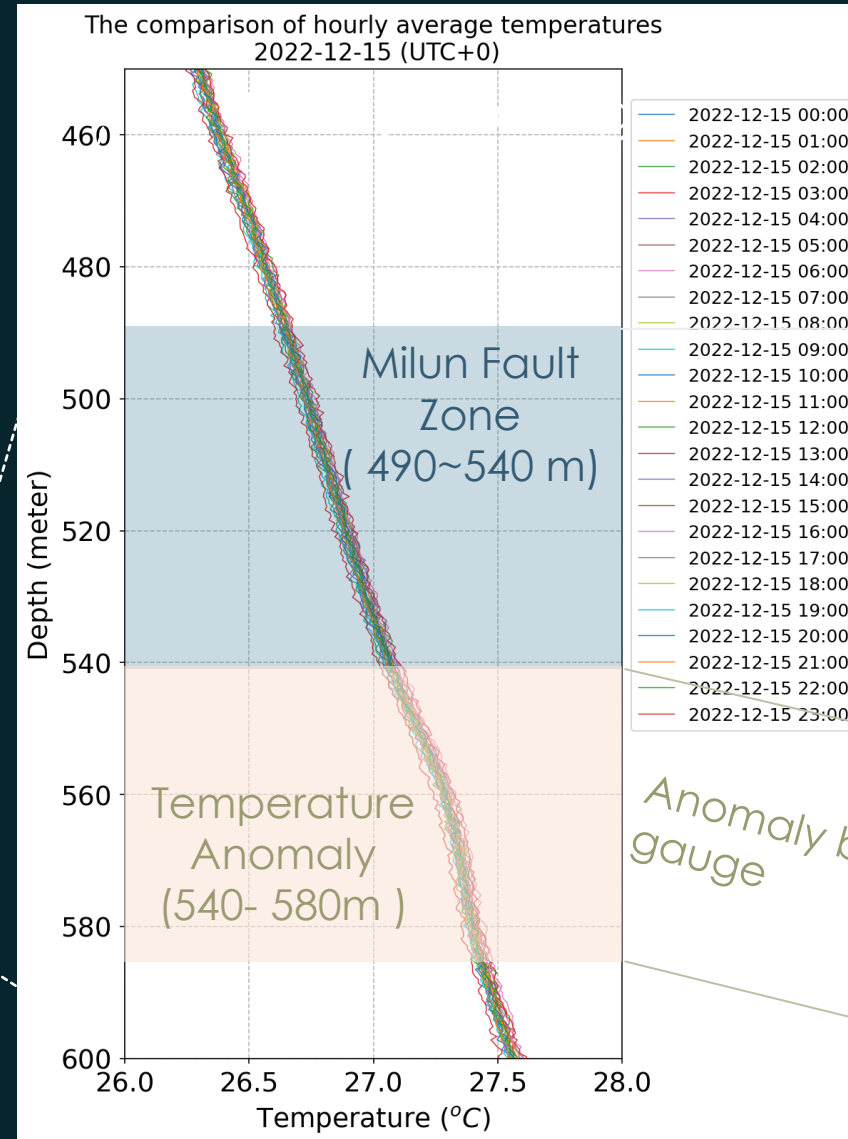
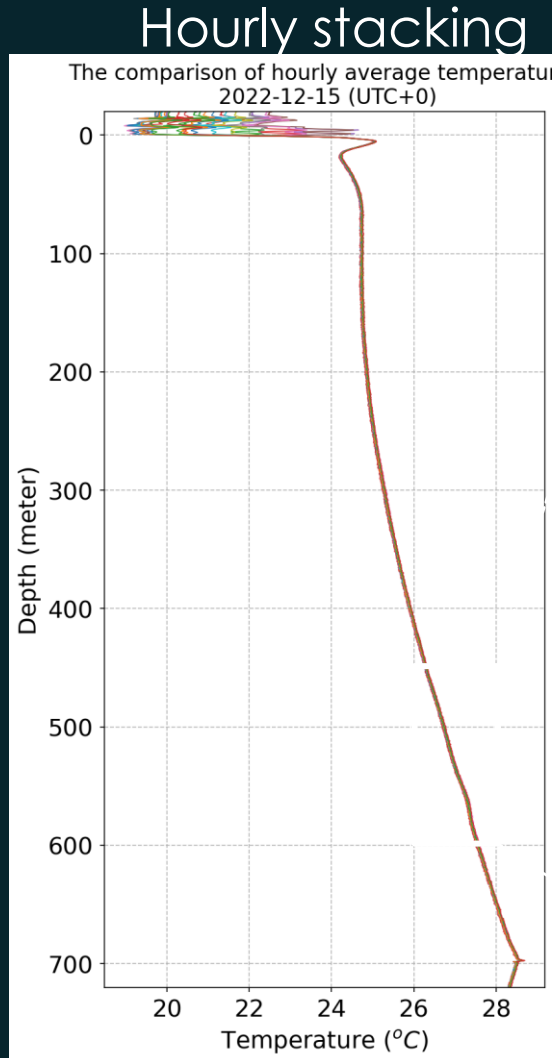


Bottom (695m)



Hsian-Ting Lai TGA/ 2025.06.17

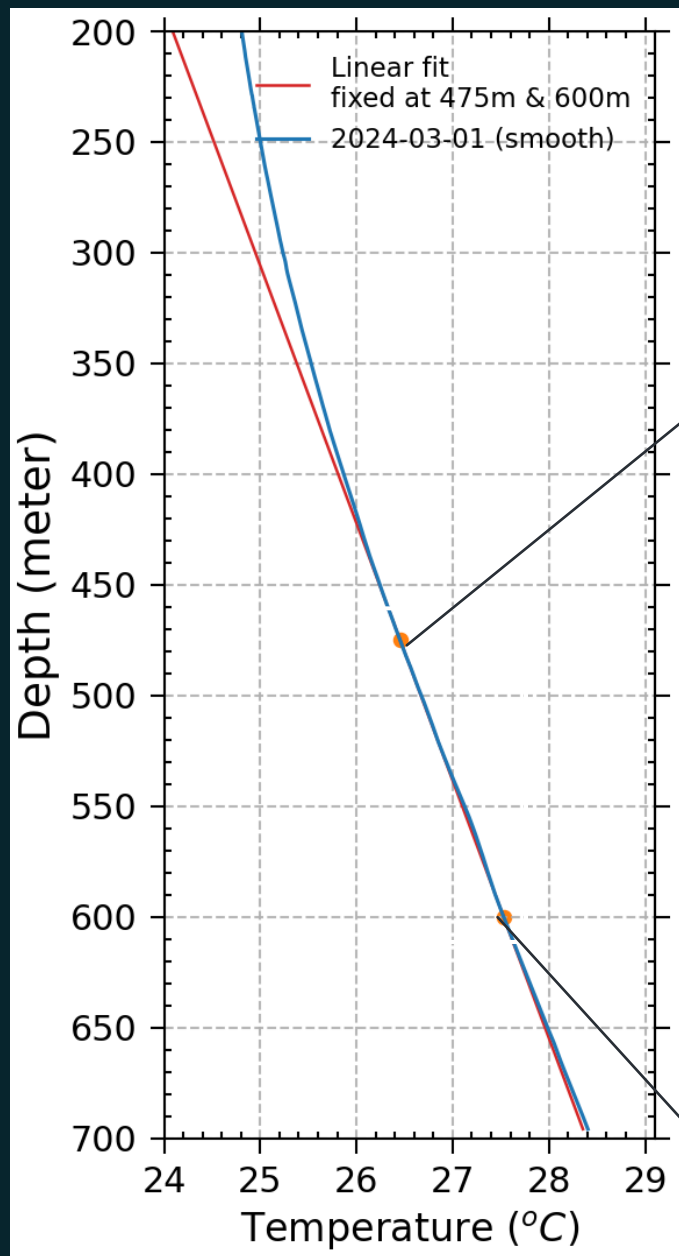
DTS hourly average temperature profile



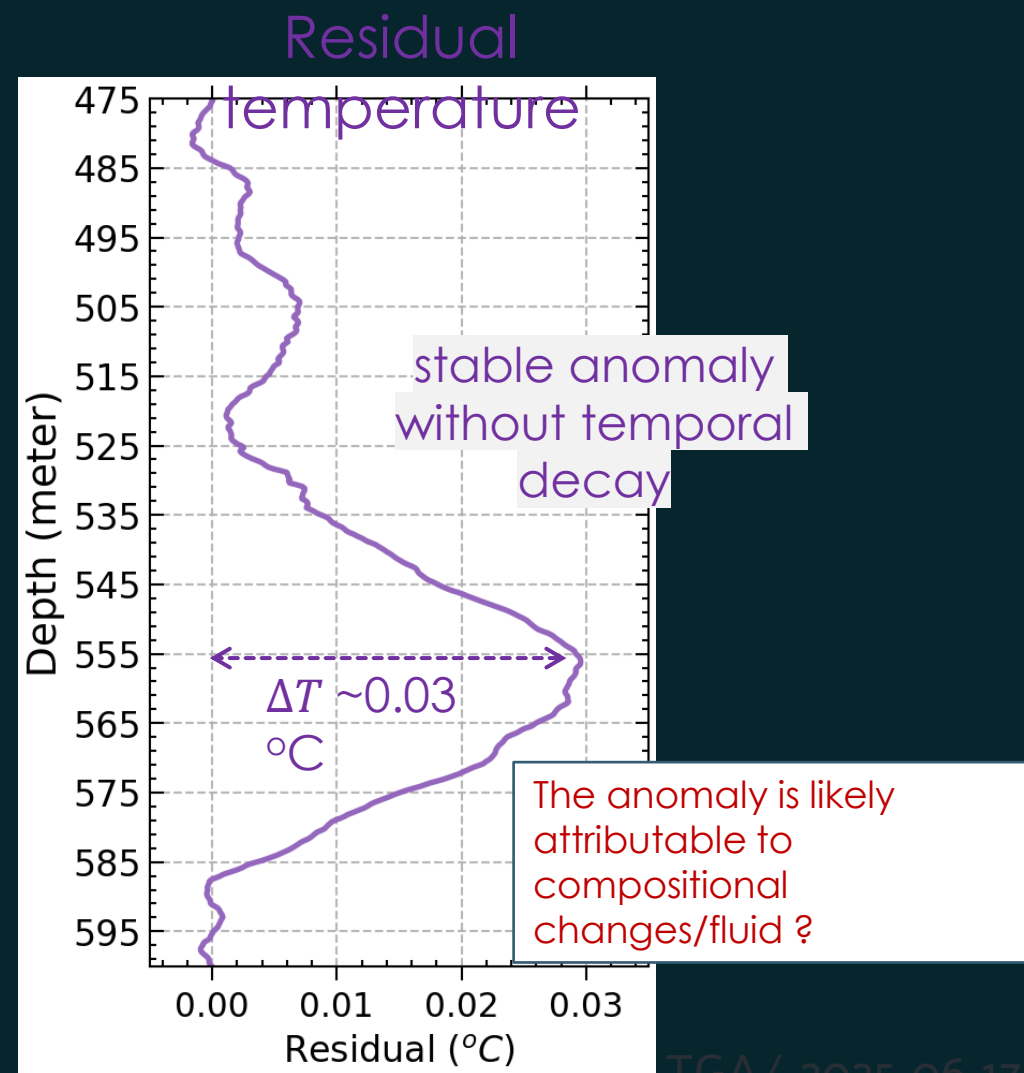
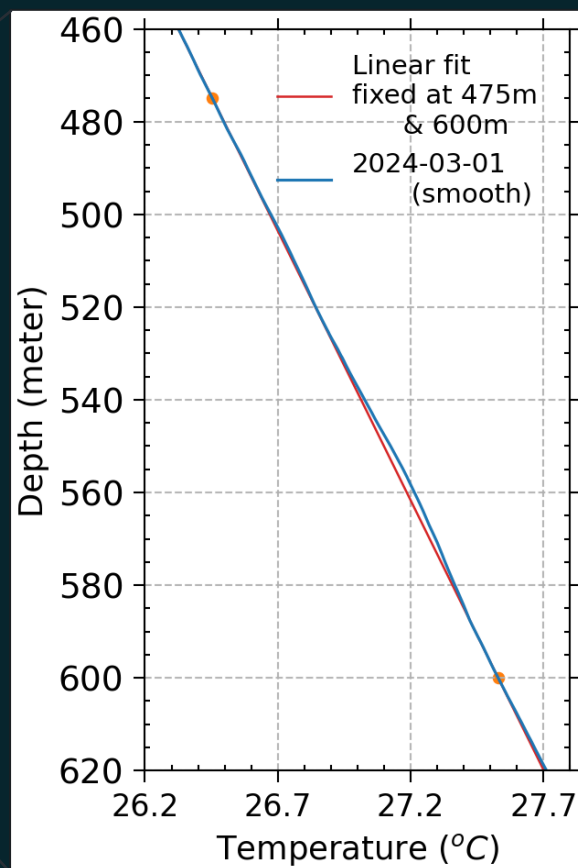
Persistent Thermal Anomaly

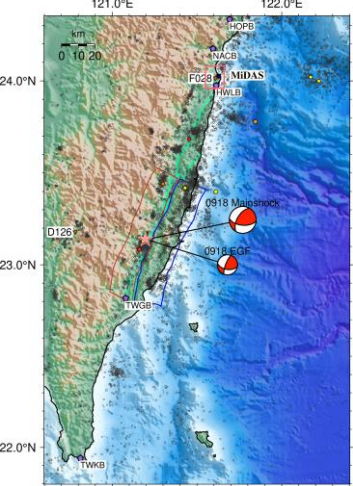


- Daily temperature stacking
- Depth resolution : 0.254 m
- Temperature resolution $\sim 0.01^\circ\text{C}$
- Moving average = 4.826 m (pts=20)



2024/03/01
Daily stacking



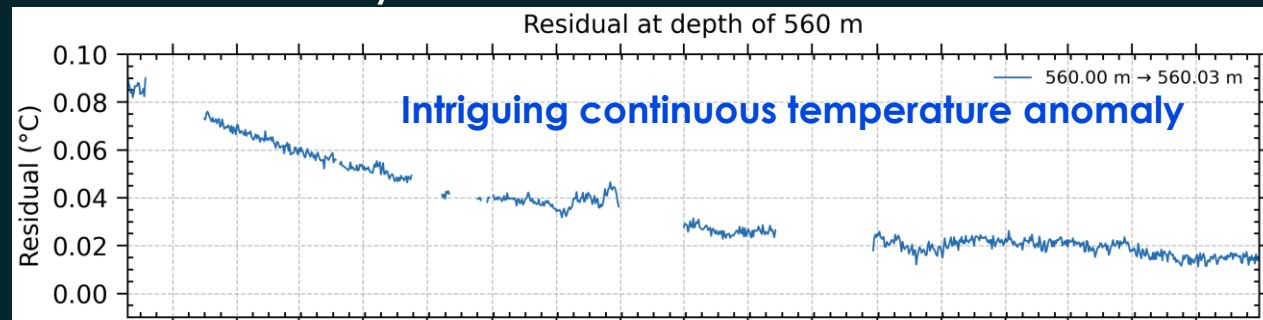


2022/09/1 M7.0 Chishang Earthquake, ~120km from MiDAS: Warm Fluid flow from the depth?

DTS analysis from 2022/09 to 2025/08



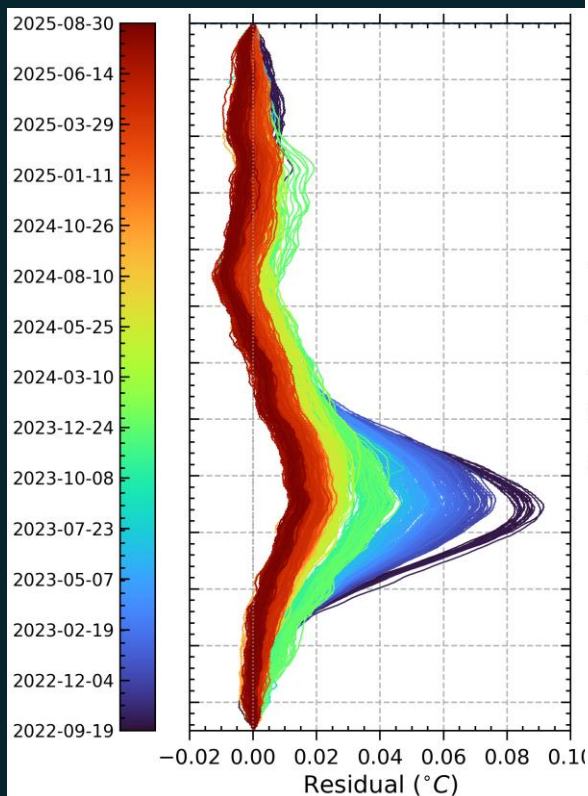
- **Daily temperature stacking**
- Depth : 475 ~ 600 m
- Date : 2022/09/19 ~ 2025/08/10



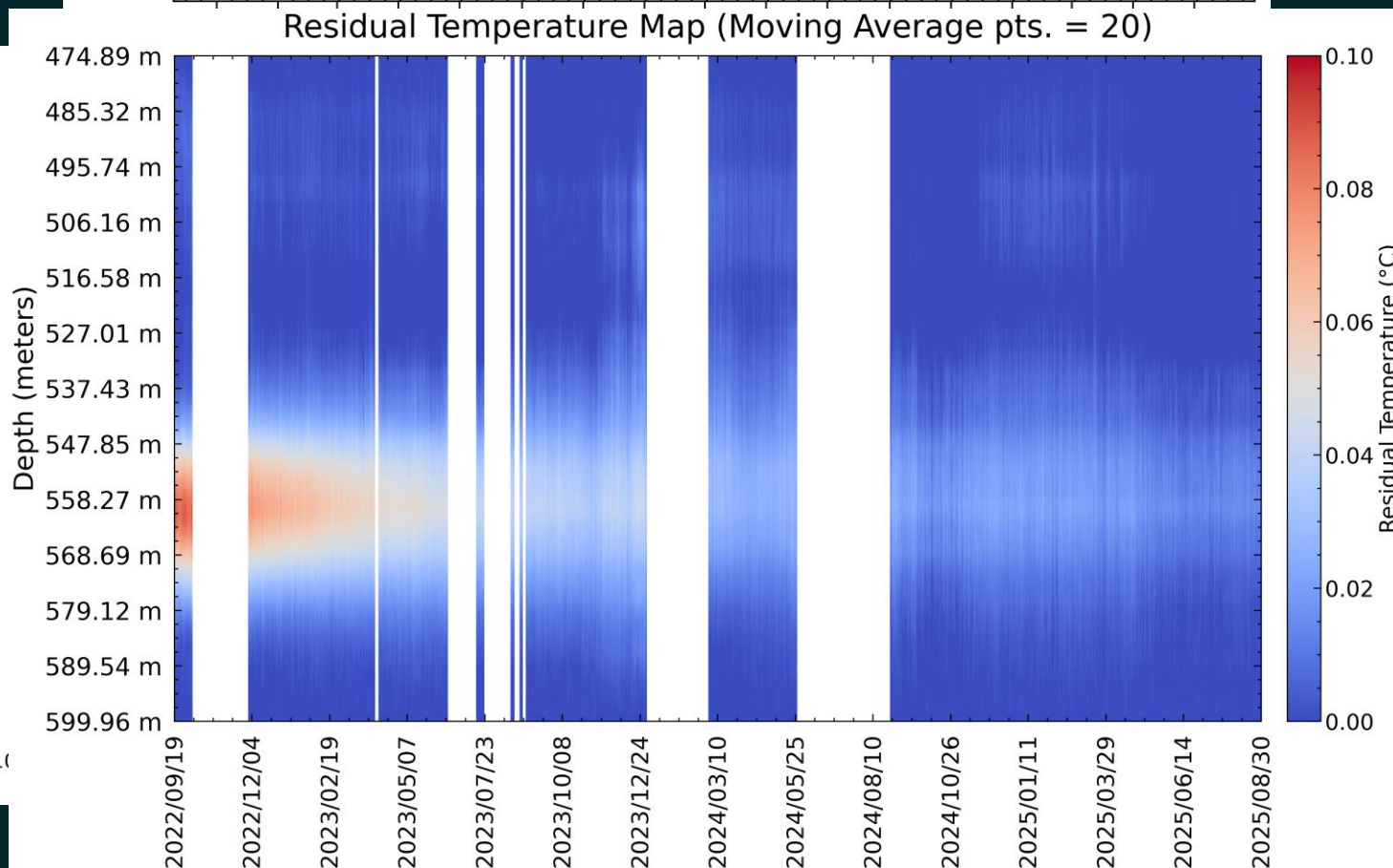
2025/08/30



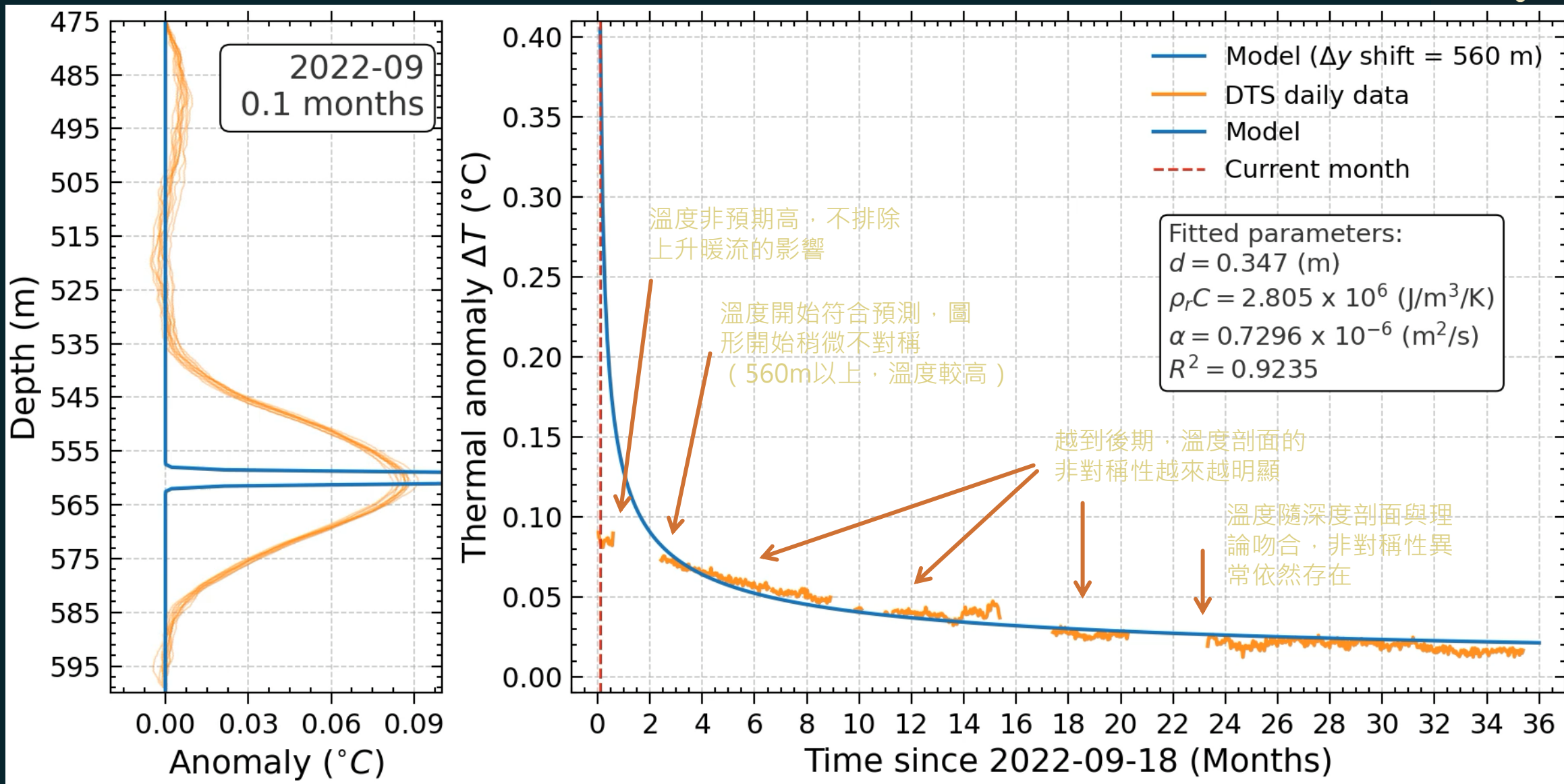
2022/09/19



From HT Lai



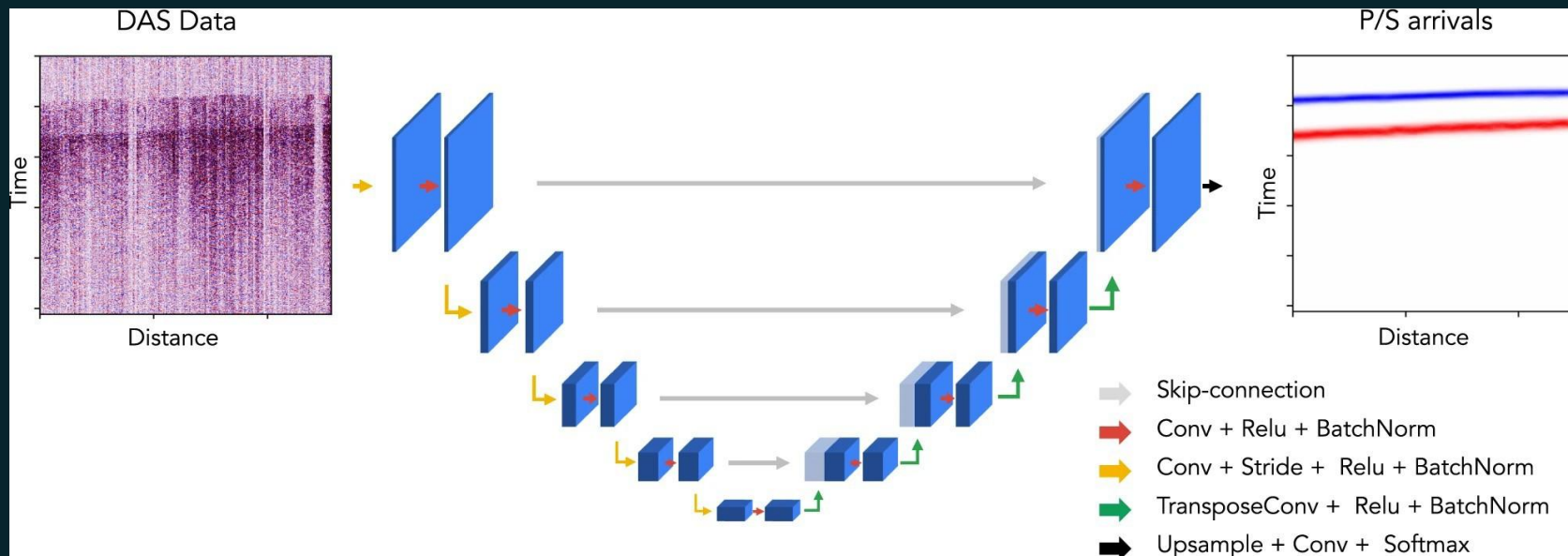
With hydrostatic pressure (Wet condition)



DAS BIG DATA=> AI APPLICATION FOR SEISMICITY INVESTIGATION



- * Exploring the application of Phasenet-DAS to continuous optical fiber DAS recordings of MiDAS
 - * Convolutional neural network designed to pick P and S wave arrivals in DAS data
 - Uses data from the MiDAS project. February 2023; Boreholes A and B (separately)
- Notes: DAS data is noisier than seismometer data (especially surface DAS data)
Typical methods don't leverage the extremely high spatial density of DAS



Trained on surface data
from fiber-optic cables
in Long Valley and
Ridgecrest, California

Zhu et al. (2023)

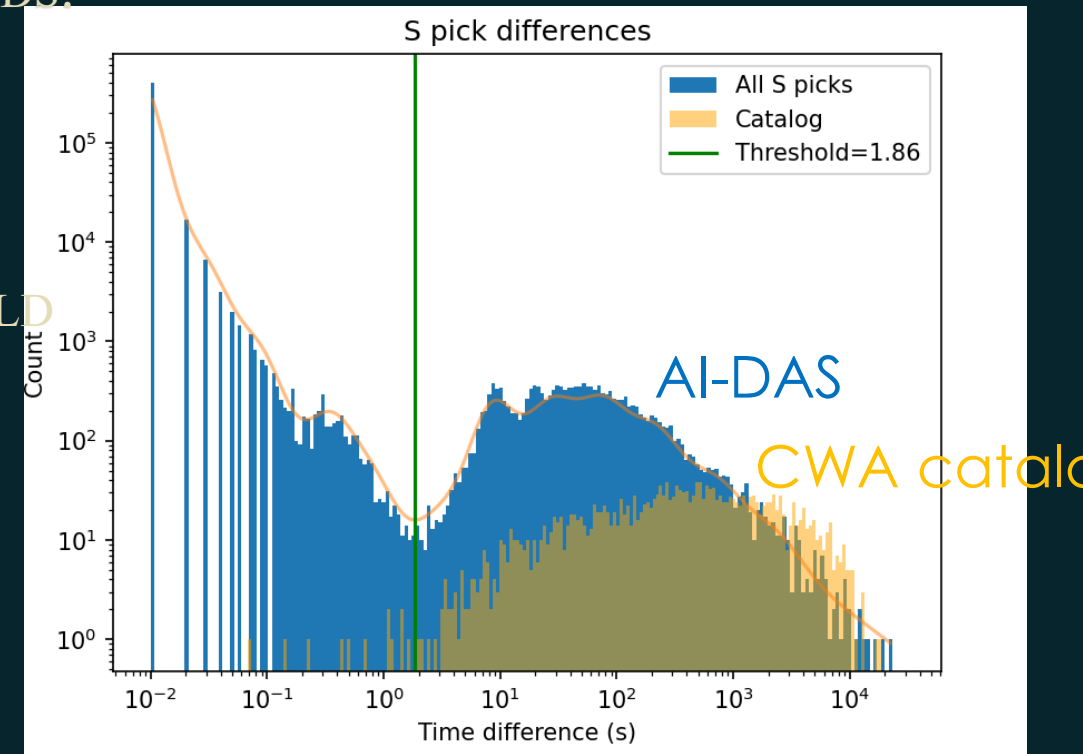
ONE MONTH CONTINUOUS DATA (~20TB)



- PARTITION INTO 30 MINUTE LONG SECTIONS (NO OVERLAP)
- BANDPASS INTO ONE OF 15 DIFFERENT FREQUENCY BANDS:
- LOWER FREQUENCY IN 0.01, 0.1, 1 HZ
- UPPER FREQUENCY IN 10, 20, 30, 40, 50 HZ
- PHASENET-DAS WITH DEFAULT PROBABILITY THRESHOLD (0.5)
- RAN FOR ABOUT A WEEK ON A SINGLE SERVER

=> Originally split picks into events with at least 1.86 second time difference (only differentiating boreholes)

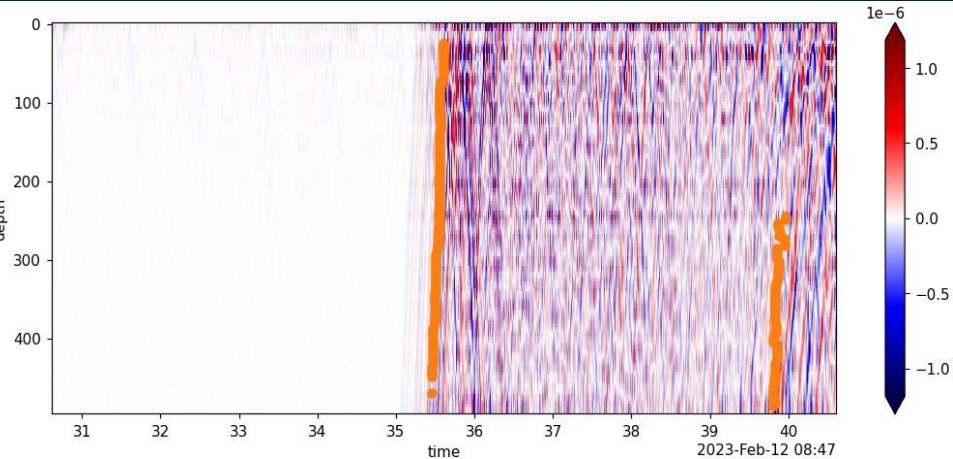
=> Threshold picked based on histogram of time differences



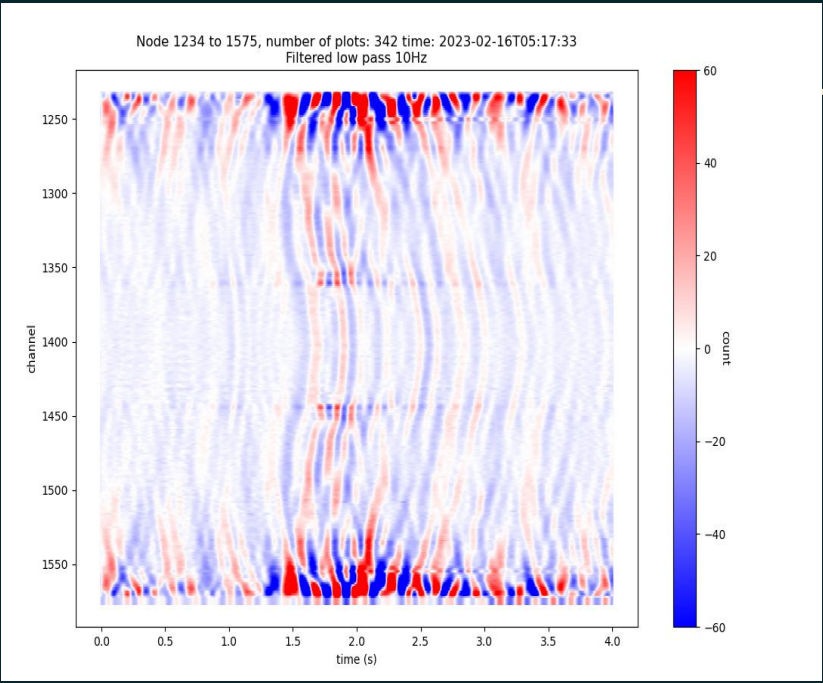
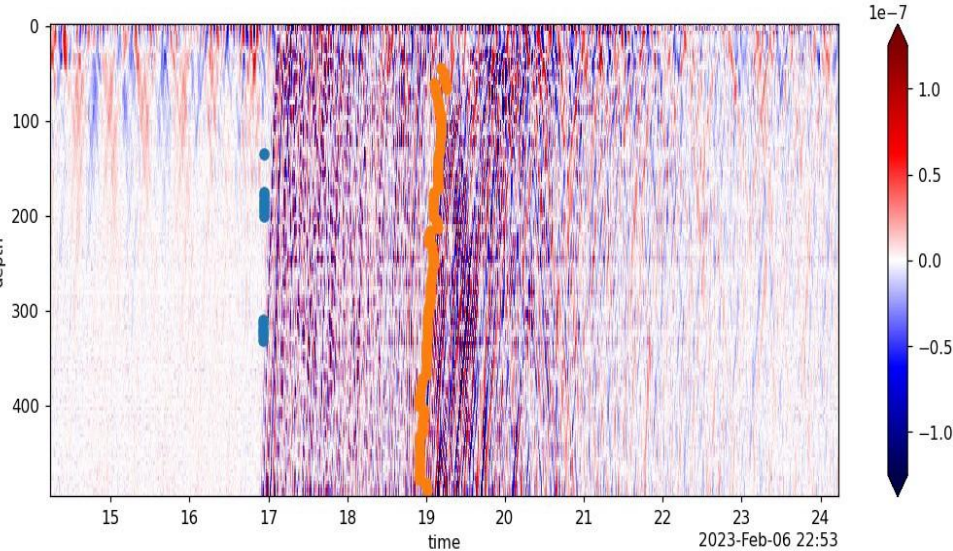
* Log-Log histogram of time between adjacent picks

* It is also possible using slope as an event identifier.

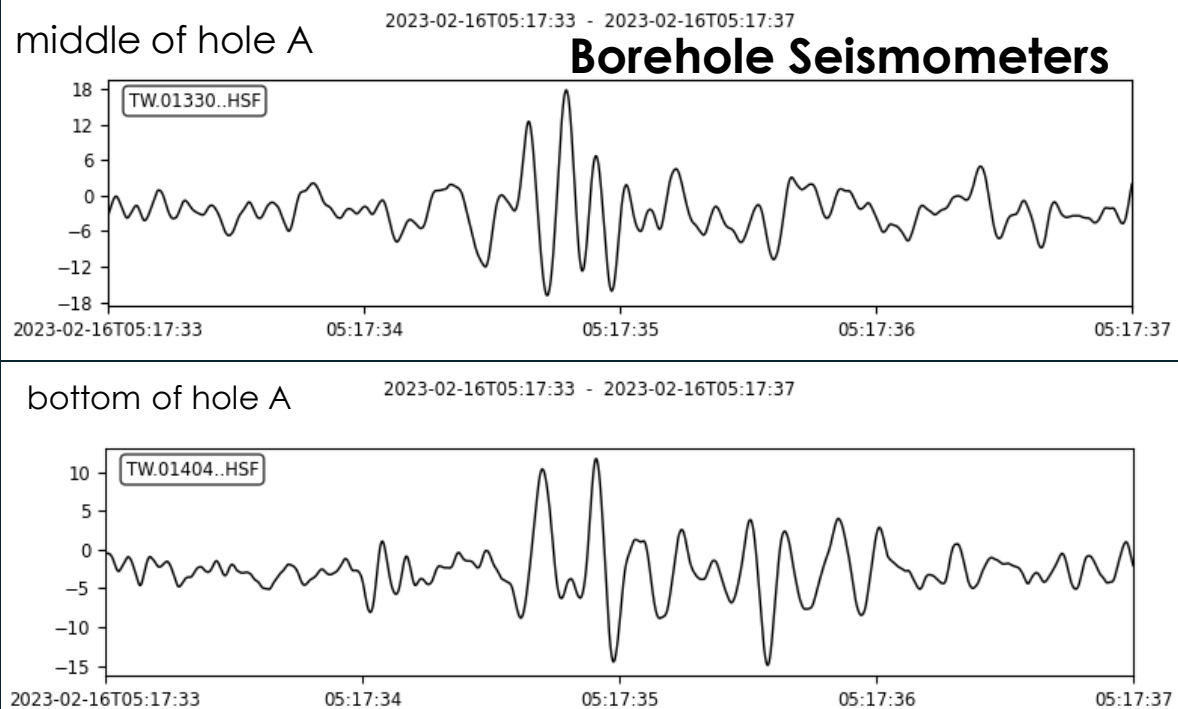
5 KM AWAY M_L 3.44 (CATALOG EVENT)



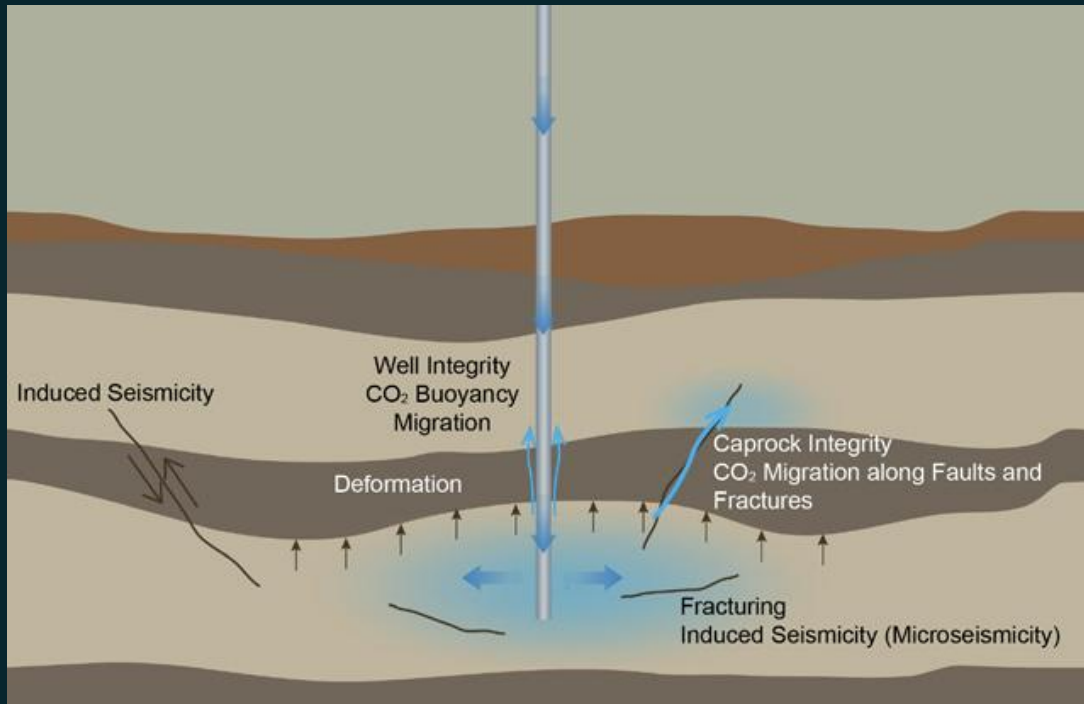
Event not in the catalog (20230206~M0)



Very close-by shallow
possibly hydraulic event
($M \sim -1$)



By
Yen-Yu Lin



- Well integrity: leakage of CO₂ upward to shallow aquifers or the surface.
- Migration of CO₂ along faults and fractures
- Migration of CO₂ plume outside of the storage reservoir
- Induced seismicity:

Silixa, Carbon Capture and Storage Monitoring with distributed fiber optic sensing, (March, 2022)

How Borehole DAS/DTS Observations at the MiDAS Site Can Help:

- Borehole DAS (Distributed Acoustic Sensing) and DTS (Distributed Temperature Sensing) technologies offer high-resolution, real-time data for enhanced monitoring.
- These methods improve understanding of subsurface dynamics before, during, and after the injection period.

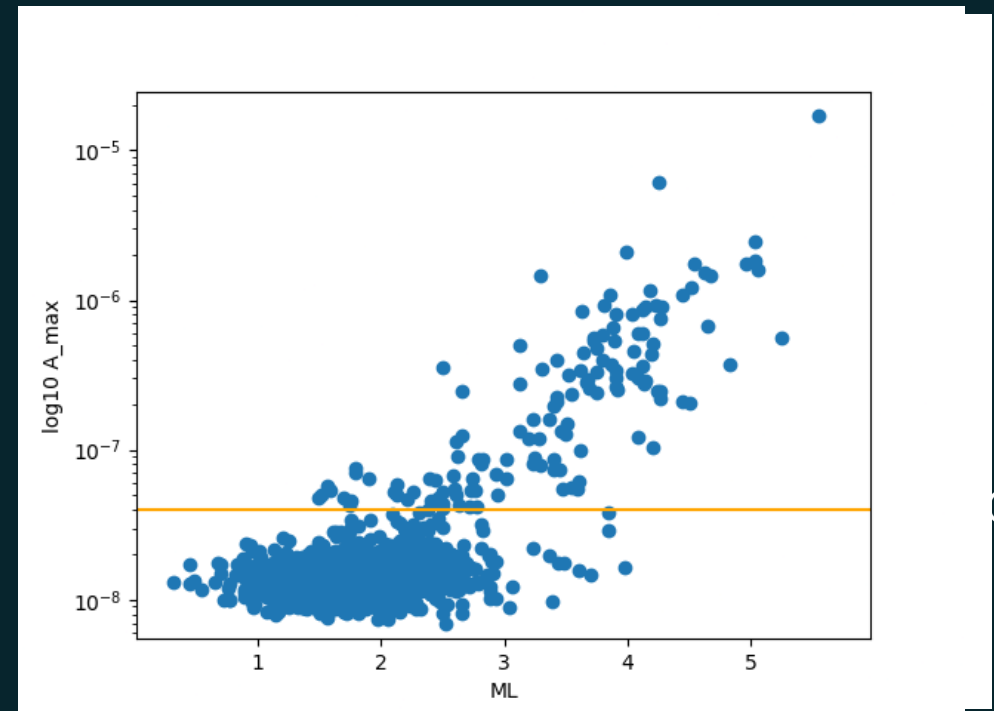
Key Insights for Future Applications:

- Fiber optic distributed sensing significantly enhances spatial and temporal data resolution.
- Lessons learned from the MiDAS site can guide the development of advanced monitoring techniques for injection-related processes, ensuring more effective characterization and management of subsurface systems.

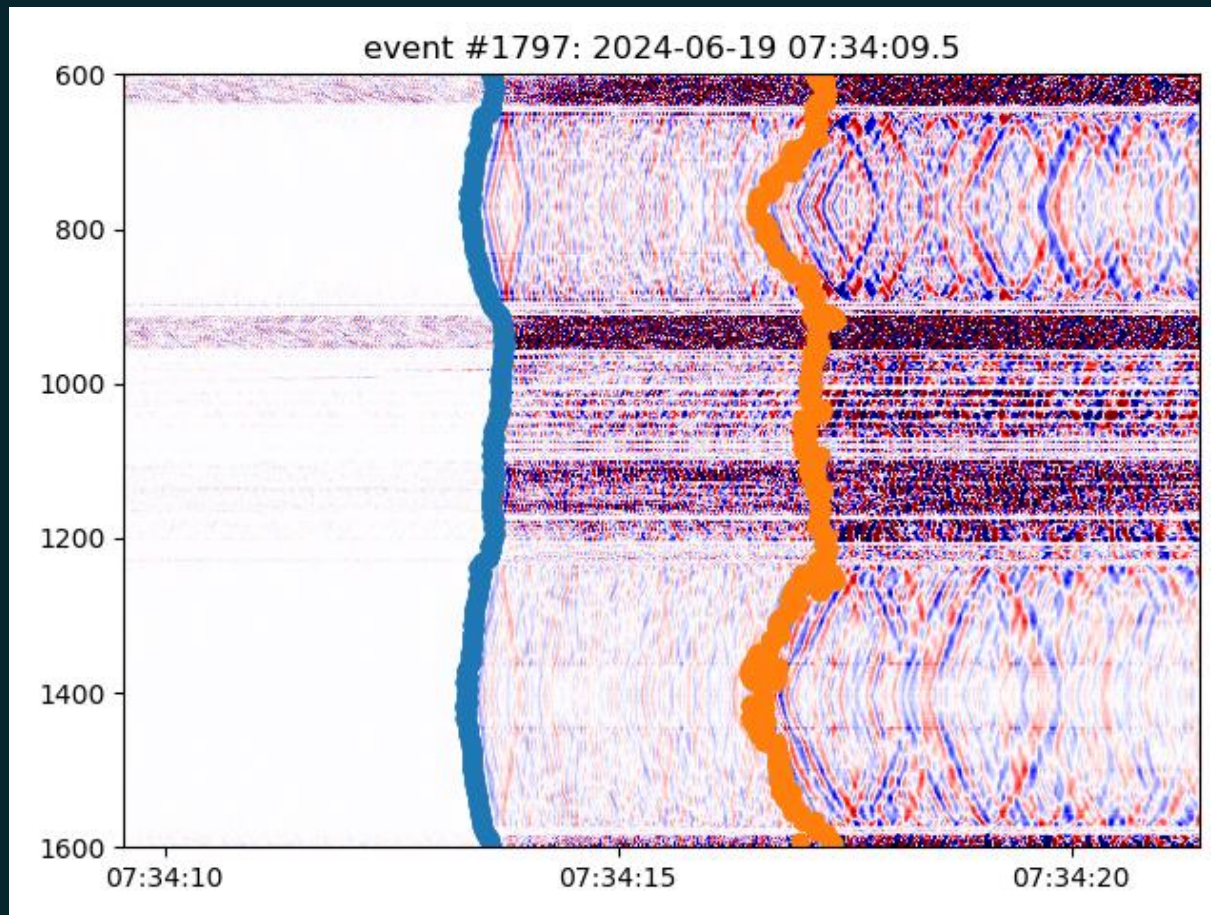
FIBER DATA FOR MAGNITUDE DETERMINATION

DATA USED

- CWA CATALOG FOR MONTH OF JUNE 2024 AS WELL AS FIRST WEEK OF MAY 2025 WERE USED
- FILTERED OUT EVENTS WITHIN 3 MINUTES OF EACH OTHER
- PLOT OF ML VS A_MAX REVEALED NOISE WAS AROUND $7\text{E-}9$ TO $3\text{E-}8$ M/M/S SO FILTERED OUT EVENTS WITH $A_{\text{MAX}} < 4\text{E-}8$
- EVENTS REMAINING WERE PROCESSED USING PHASENET-DAS AND MANUALLY FILTERED (SEE NEXT SLIDE)
- EVERYTHING FILTERED 0.1-20 Hz

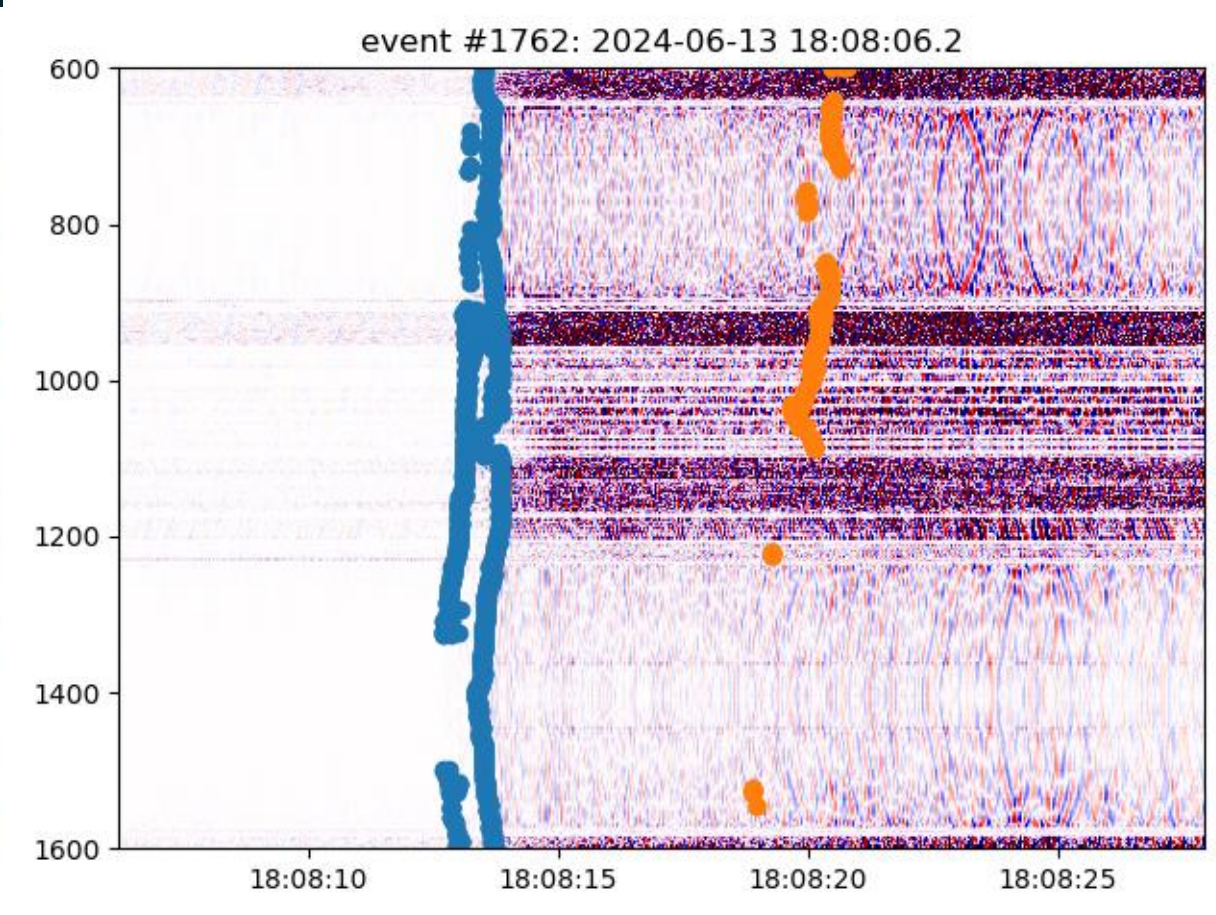


GOOD AND BAD RESULTS FROM PHASENET-DAS



Good

Phasenet-DAS run on entire fiber for 1 minute before to 2 minutes after the earthquake origin time. Results were then cut to only picks between the origin and twice the expected S-arrival time.



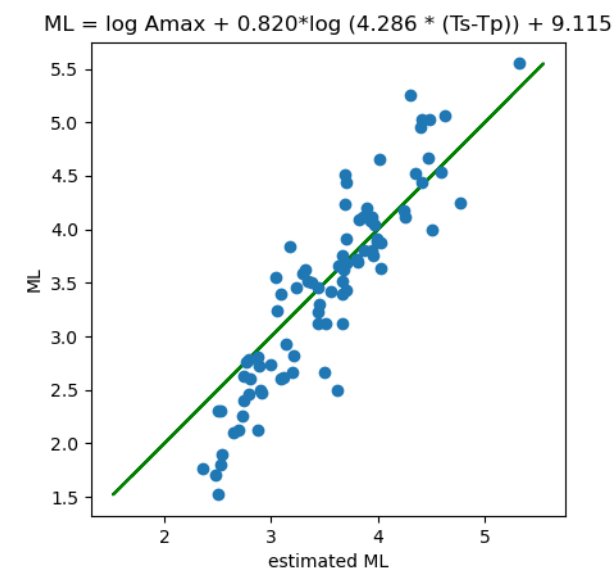
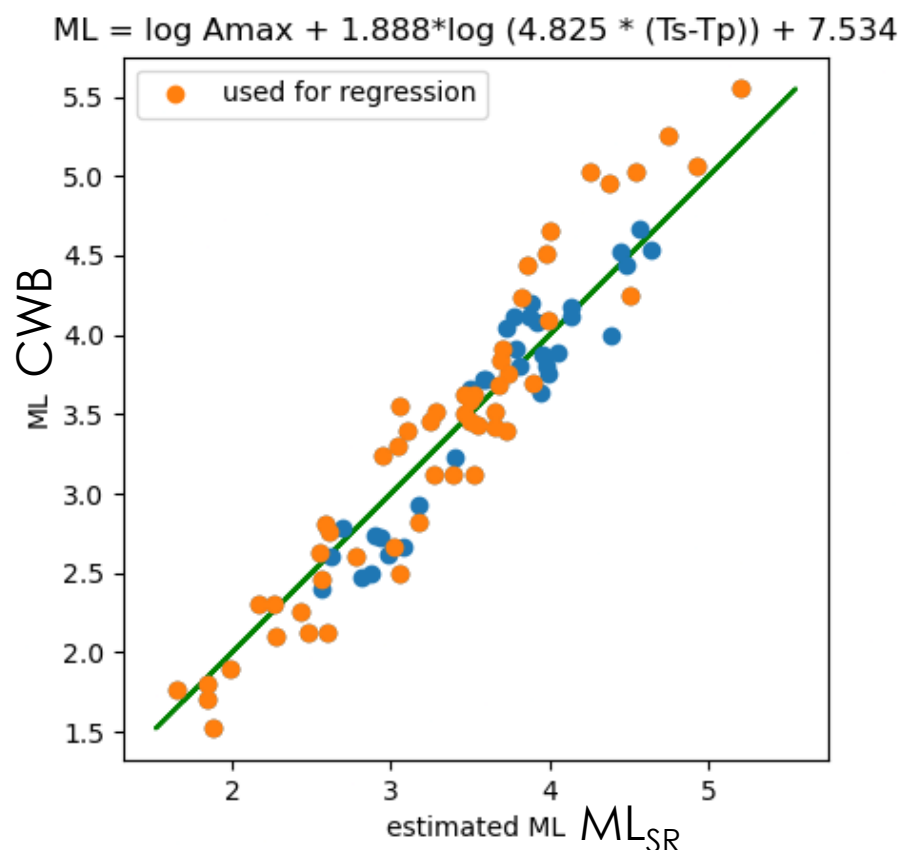
Bad

ESTIMATING ML USING AMPLITUDE AND TS-TP TIME

$$ML_{SR} = \log A_{max} + 1.888 \log(4.825 * (ts - tp)) + 7.534.$$

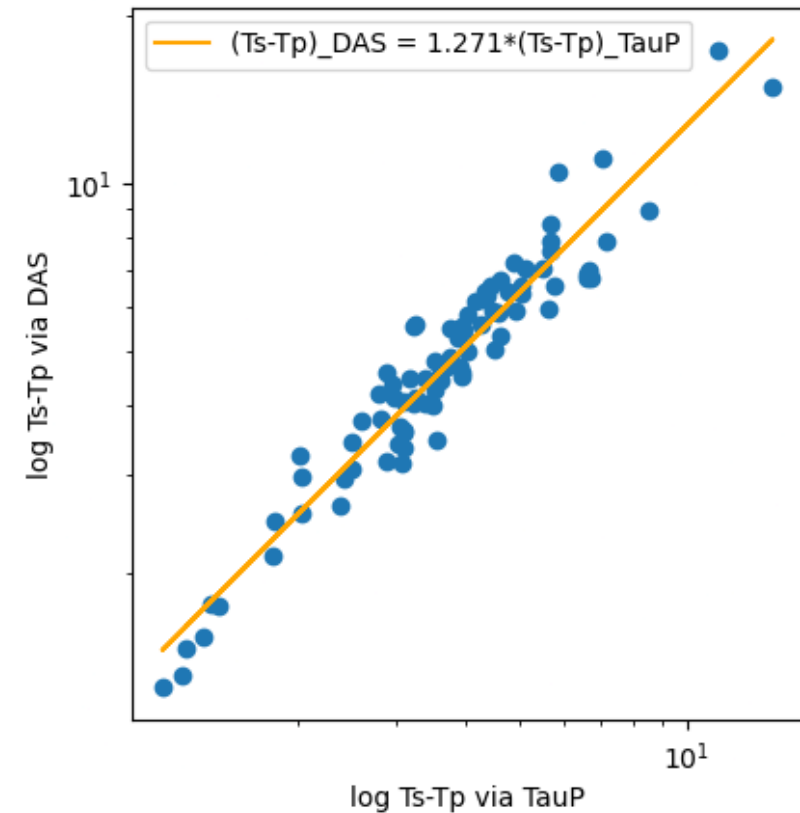
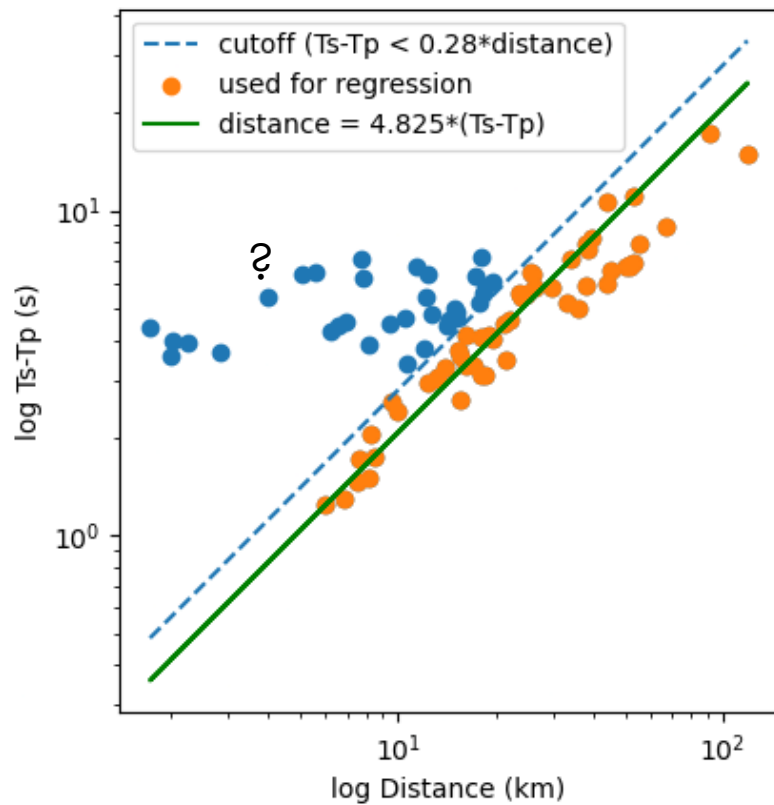
(A_{max}: strain rate, m/m.sec; ts, tp: sec)

Use ts-tp time to estimate distance, and then use that distance and the max amplitude to estimate magnitude



If using regressions based on all points

ESTIMATE DISTANCE FROM TS-TP TIME (HOLE-A, JUNE CATALOG EVENTS)

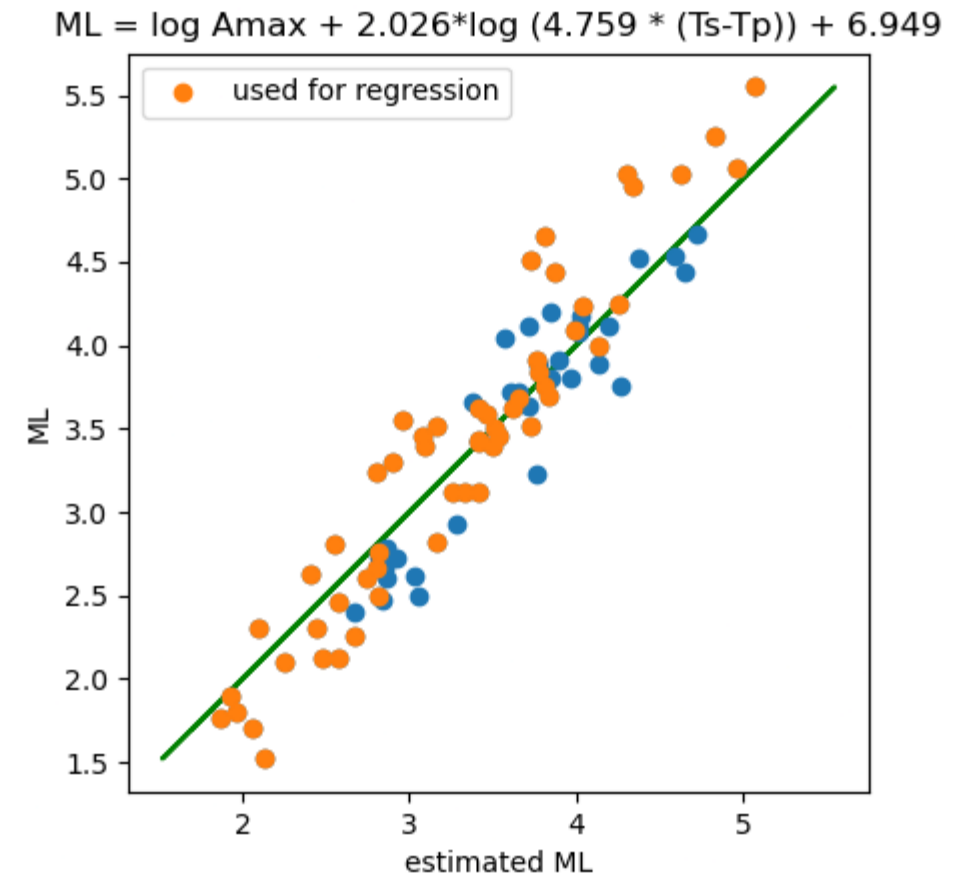
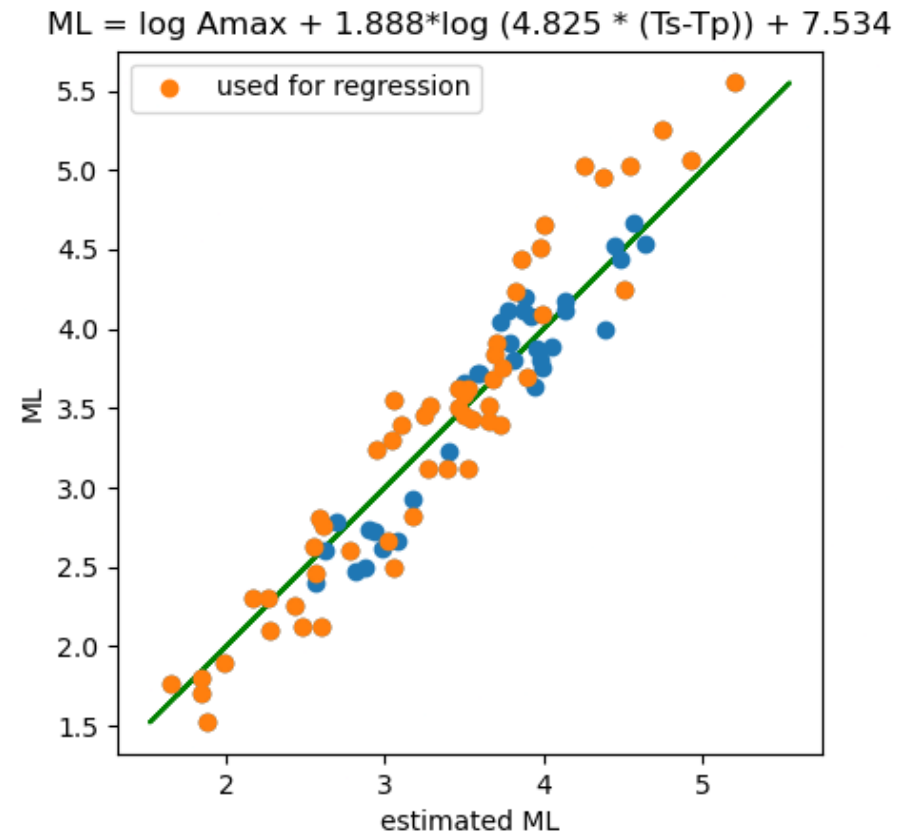


Points in orange are used for all other regressions

COMBINED REGRESSION HOLE A VS HOLE B

Hole A

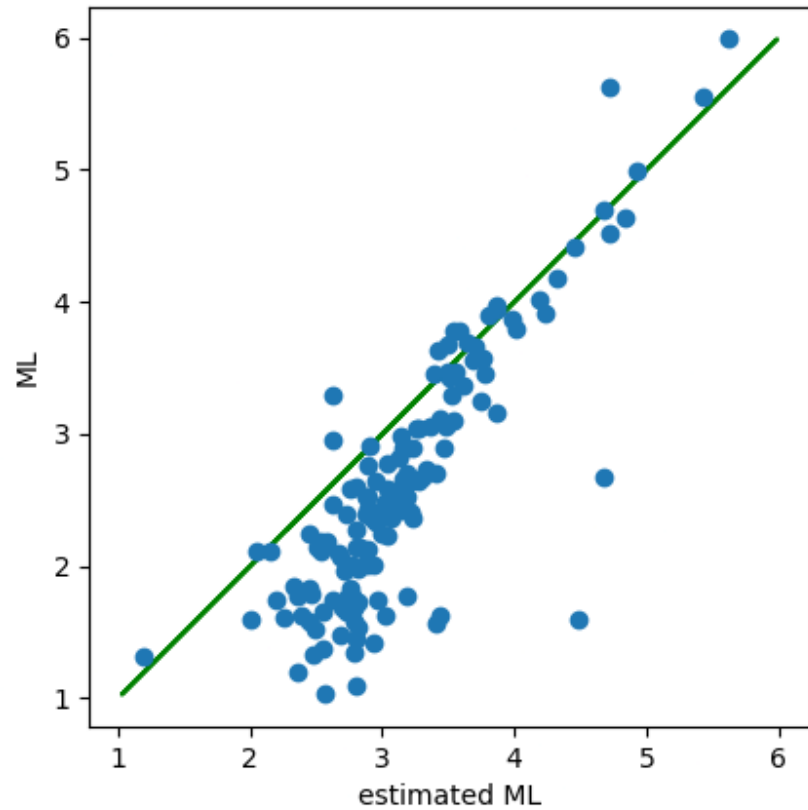
Hole B



TESTING REGRESSION ON MAY, 2024

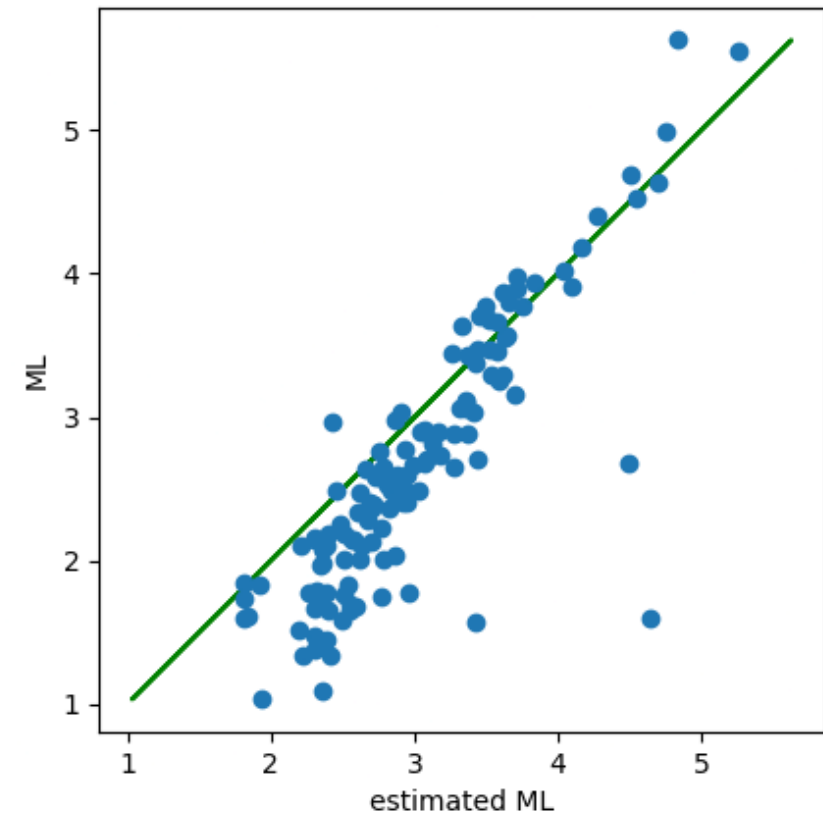
Hole A

$$ML = \log A_{max} + 1.888 * \log (4.825 * (Ts - Tp)) + 7.534$$



Hole B

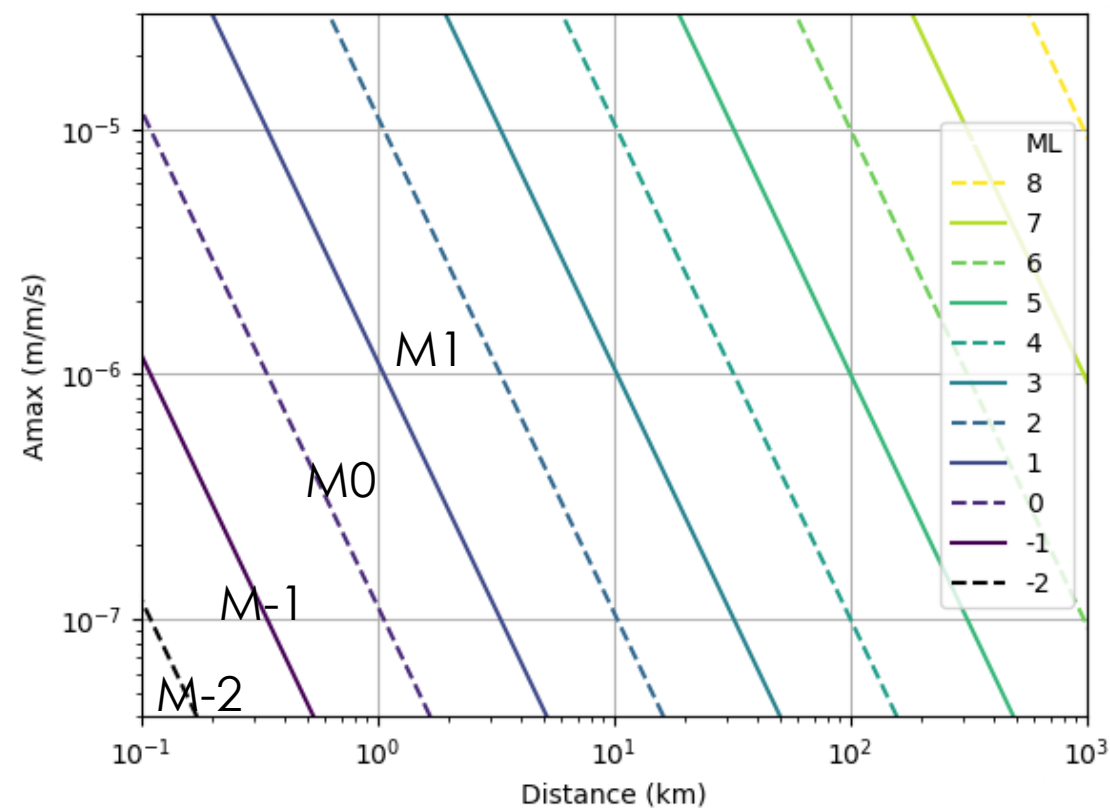
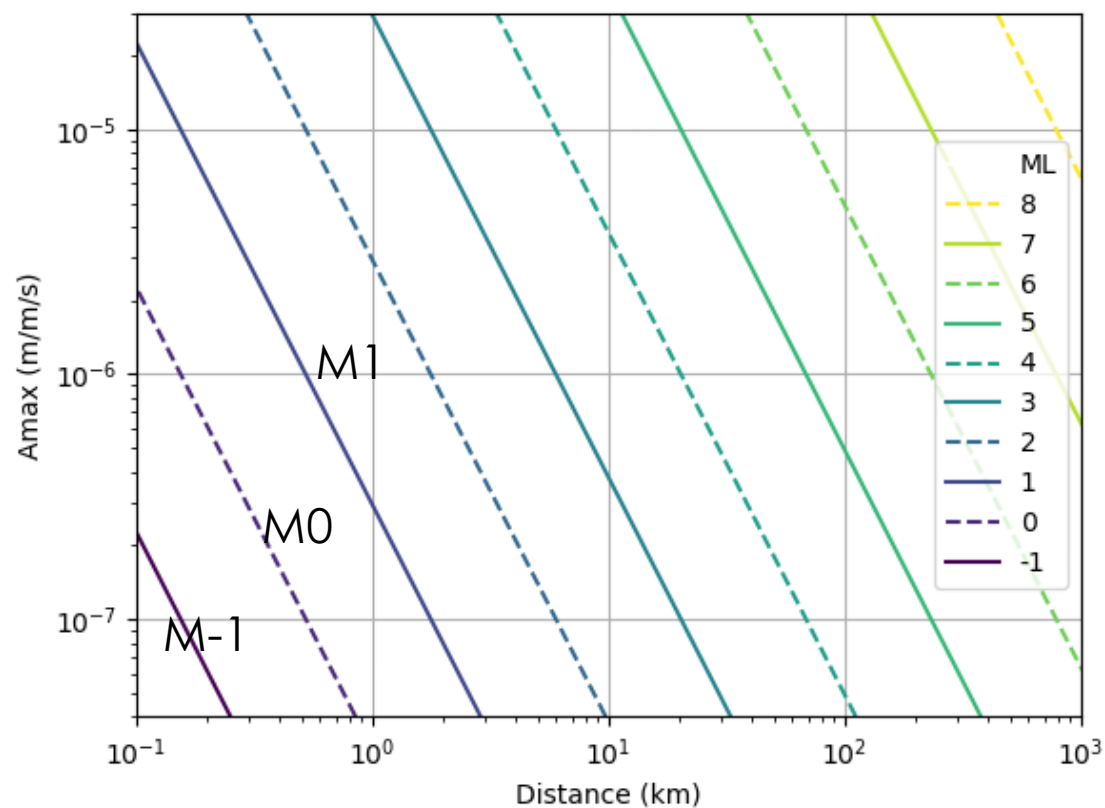
$$ML = \log A_{max} + 2.026 * \log (4.759 * (Ts - Tp)) + 6.949$$



CAPABILITY HOLE A VS HOLE B

Amplitude limits are noise floor ($\sim 4e-8$) and saturation point ($\sim 3e-5$)

Saturated distance and magnitude

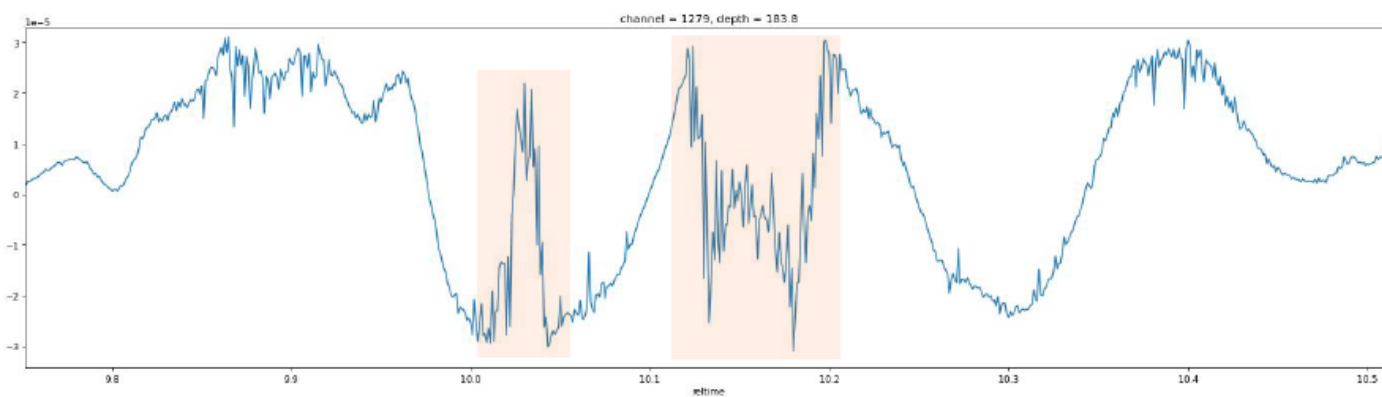


Capability of magnitude and distance threshold

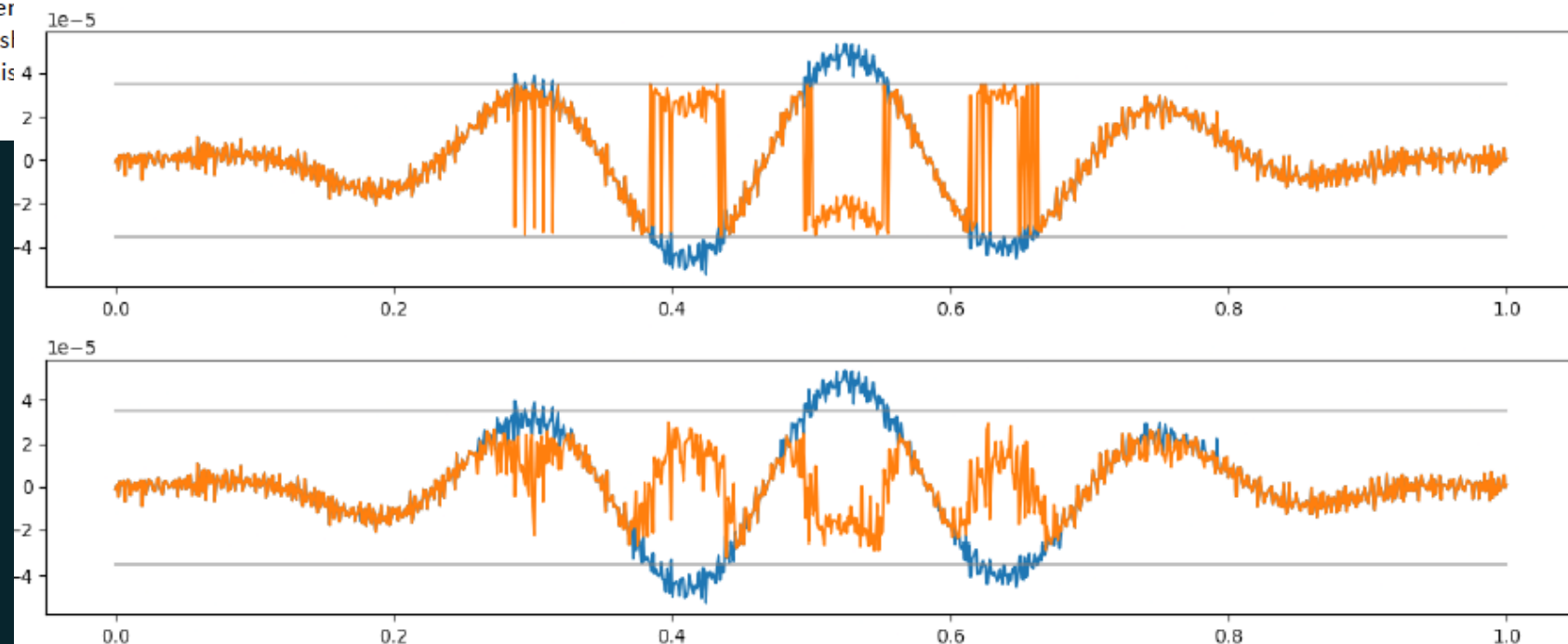
What's the issue with wrapping?

Phase shift is by definition measured in a limited range from $-\pi$ to π , but strain rate is not so limited.

Normally this wouldn't be an issue (methods dealing with e.g. InSAR data would work), however in our case the instrument uses some proprietary denoising algorithm on the raw data, which has the side effect of mostly eliminating the large jumps that usually characterize wrapped phase data. (The shaded boxes below indicate wrapping sections)



That said, the instrument samples at a 25cm spatial sampling frequency. An optimal way to downsample this to 2m (by first correcting the time shift using a weighted mean). When the raw phase shift is wrapping however, this is problematic (try stacking first then wrapping, vs wrapping first then stacking)



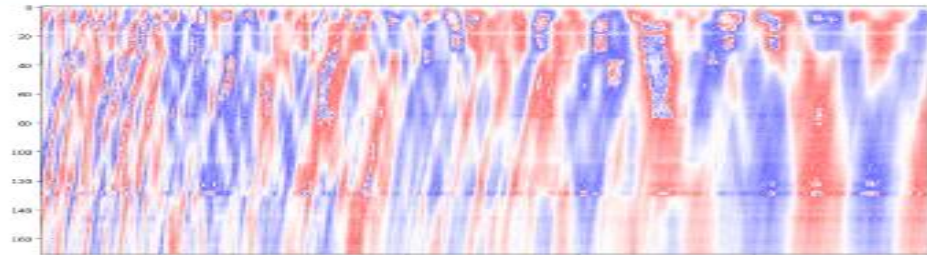
“Saturation” Issue for Strong Motion

Wrapping/Unwrapping*

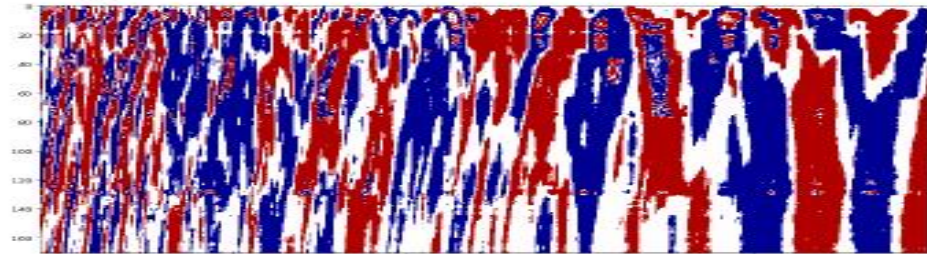
By Alexander Ristich

Method

Note that wrapping occurs in more or less contiguous regions. We should be able to unwrap the data by simply adding or subtracting a constant value to those regions. So the problem becomes discriminating the wrapping regions. This is done by using the following steps:



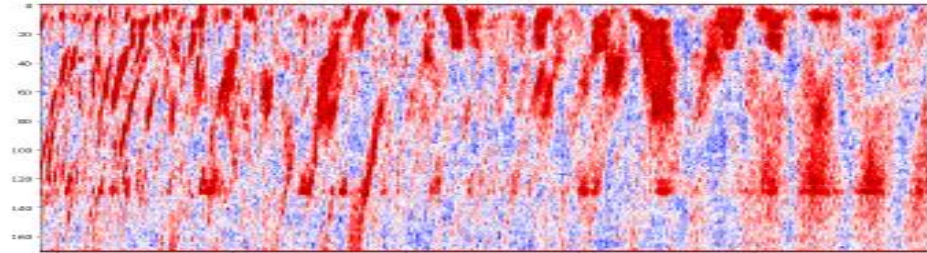
Original data, sliced to only clipping parts and scaled to the range $(-1, 1)$



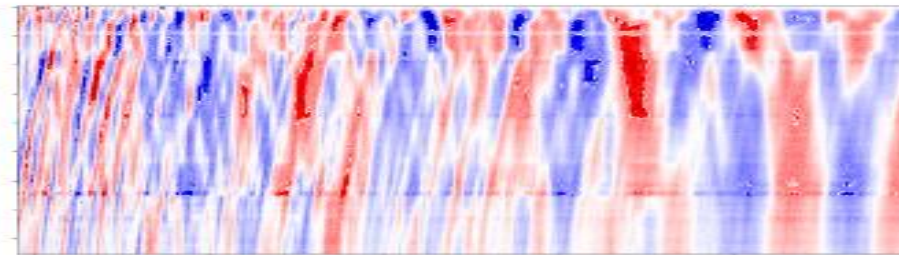
Mark points with absolute value > 0.2



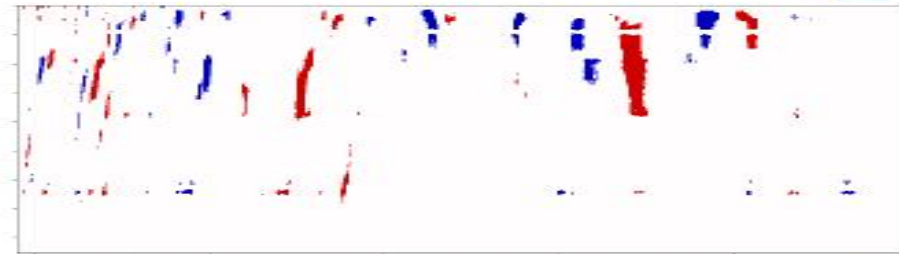
Use random walker image segmentation algorithm to divide data into positive and negative regions



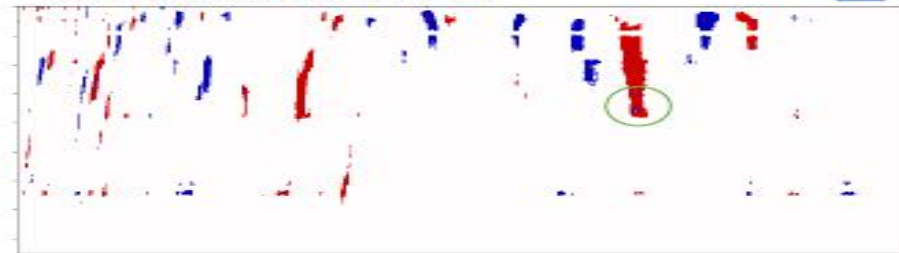
Calculate local noisiness of original data. Here we do that by finding a best fit paraboloid to a pixel and its 8 neighbors, and output the squared sum of differences



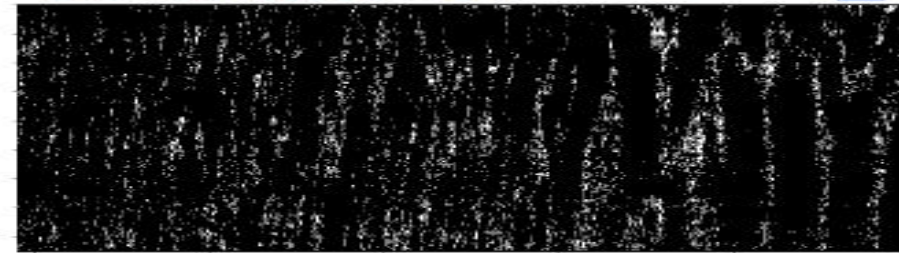
Add difference to original data



Use median filter to remove speckles of incorrect signs (see in green circle below. Repeat this step as necessary)



Zero out unwrapped regions based on whether any of the regions points are already known to be unwrapped



Threshold the noisiness in order to find points that are not wrapped

Verification

In order to verify our results we can compare the unwrapped data to seismometer data. The way we do that is the integral of strain rate from 100m to 695m in hole A should be equivalent to the difference in vertical velocities at those points.

While in theory these should be equal, in practice we scale the seismometer data by about 0.7 (parameters found by least squares fit with a small earthquake).

We'll be looking at the results on four earthquakes. From least to most saturation:

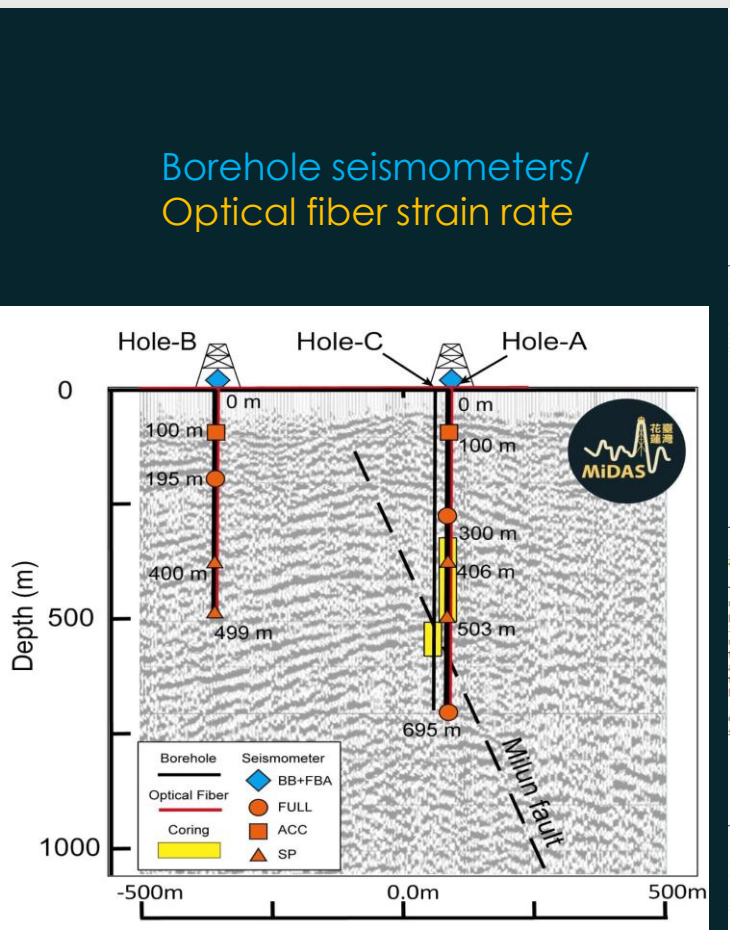
	UTC Date/time	Magnitude	Latitude	Longitude	Depth (KM)	Distance (KM)
1	2024-04-26 18:21:24	6.3	24.18° N	121.67° E	35.5	39.7
2	2024-04-22 18:32:48	6.2	23.85° N	121.52° E	7.7	23.5
3	2024-04-03 00:11:26	6.5	24.13° N	121.65° E	13.4	18.0
4	2024-04-02 23:58:09	7.1	23.88° N	121.57° E	19.7	26.0

Comparison of the DAS integrated strain rate (in blue) and the acceleration derived strain rate (in orange)

Recovering of the wrapping signals

On-going test with borehole seismometers

(Alexander Ristich)



#1 Minimal Clipping

Looking at the waterfall plot, it appears to work, although there is very little difference in the seismometer comparison (a handful of changes between 44000ms and 46000ms).

#2 Some clipping

Again, the waterfall plot doesn't reveal any issues, and this time there is a much more noticeable improvement in the comparison (especially around 40000)

#3 Strong clipping

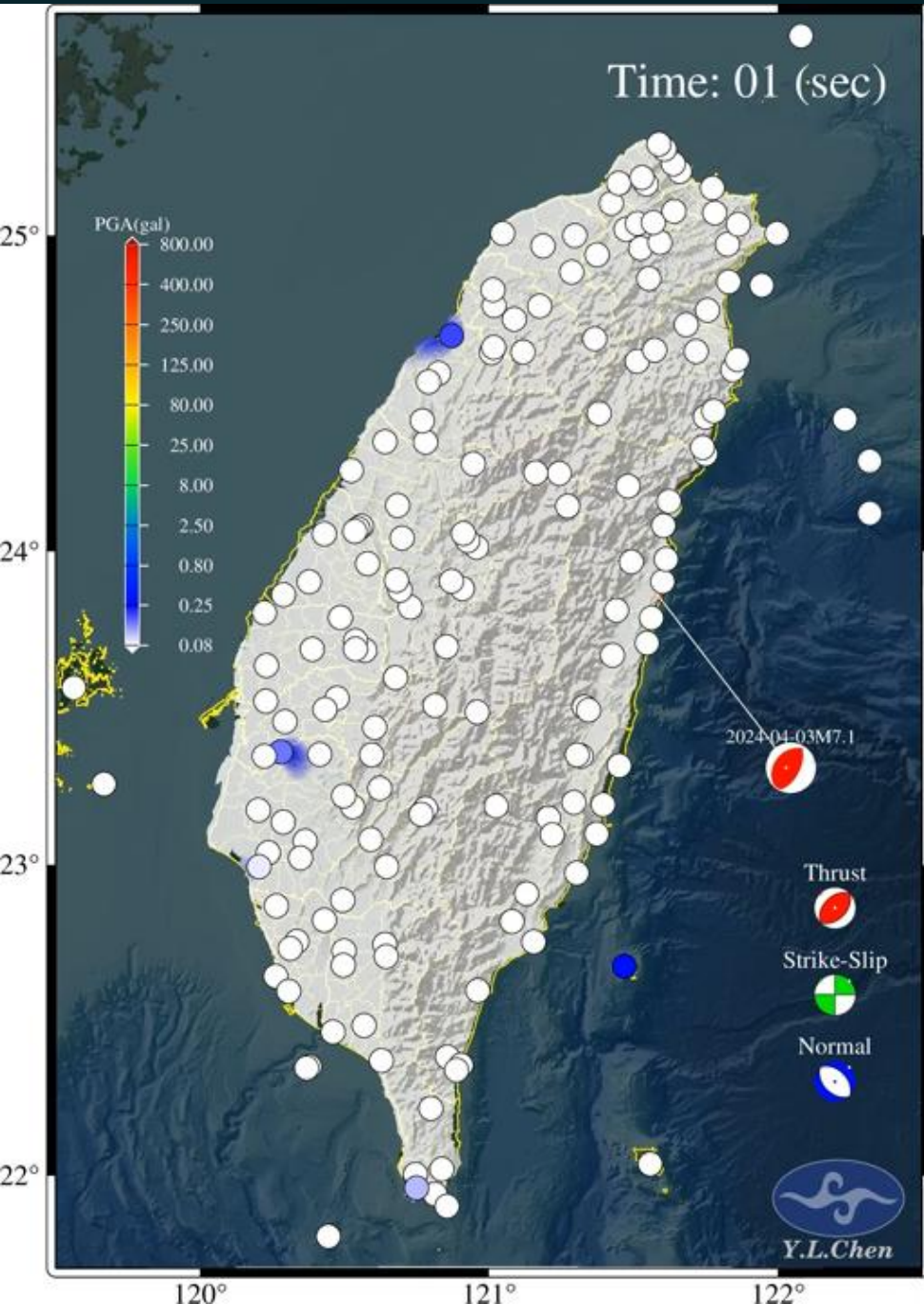
This time there are a number of noticeable issues from the waterfall plot. For example, between 3200 and 3400, one region is completely uncorrected, and a region near the top is partially uncorrected. In addition some regions are unwrapped that weren't actually wrapped to begin with.

On the other hand, the comparison shows a pretty striking improvement.

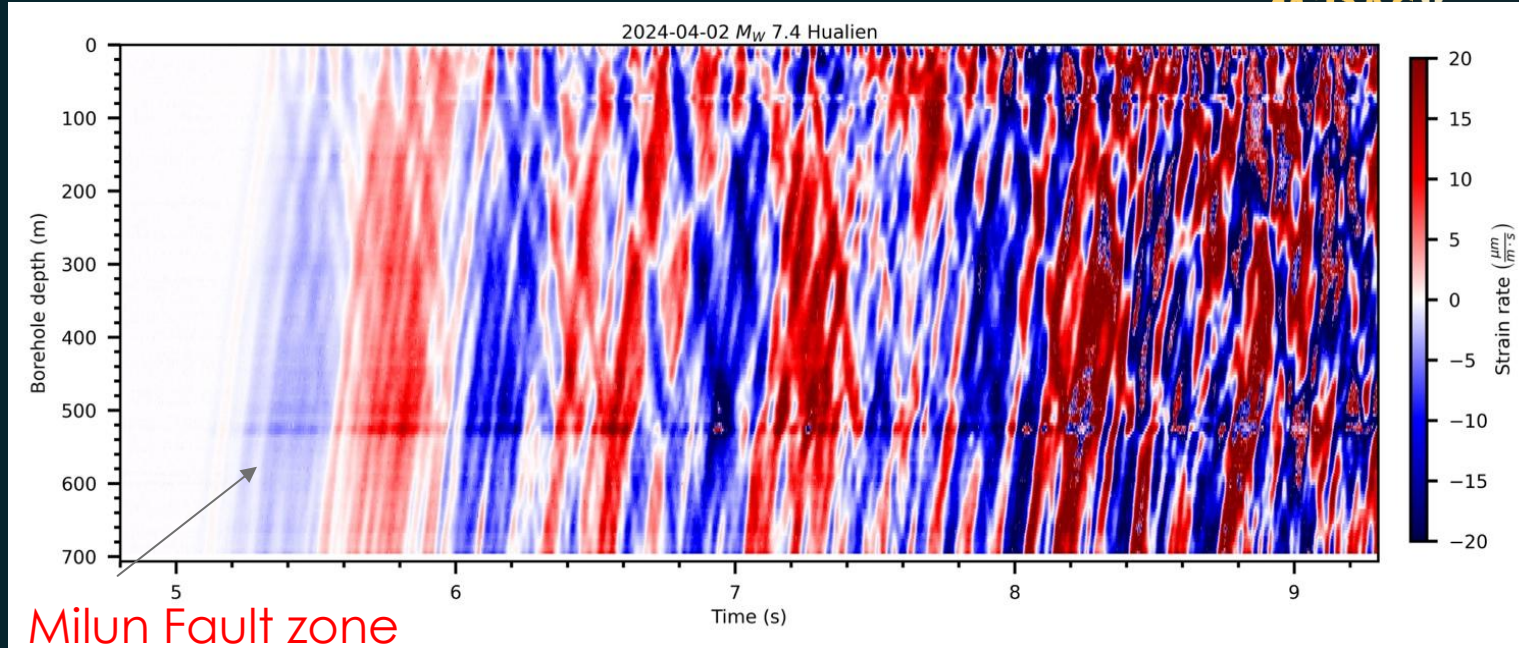
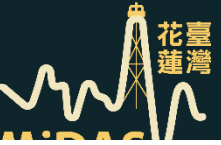
#4 Very Strong clipping

Regarding the main shock, the waterfall plot looks pretty bad.

The comparison also indicates the around 3x range increase achievable.



20240403 M7.2 earthquake (Hole-A) @MiDAS (20 km away from MiDAS site)



The time it
shakes @MiDAS
(PGA ~250 gal,
permanent
uplift, ~30cm)

Summary



- Well integrity: leakage of CO₂ upward to shallow aquifers or the surface.
- Migration of CO₂ along faults and fractures
- Migration of CO₂ plume outside of the storage reservoir
- Induced seismicity:

- DAS/DTS Emerging technology for ultra-fine structure spatial and temporal mapping
- Broadband strain amplification directly related to the reduction in elastic modulus, namely the rigidity (μ).
- Good indicator for fluid flow monitoring (or gaseous and liquid carbon dioxide).
- Continuous spatial resolution of DAS data provide real-time seismicity mapping using AI. Together with geophone and borehole seismic array for locations.

Key Insights for Future Applications:

- Fiber optic distributed sensing significantly enhances spatial and temporal data resolution.
- Lessons learned from the MiDAS site can guide the development of advanced monitoring techniques for injection-related processes (geothermal, CCS...) , ensuring more effective characterization and management of subsurface systems.

Milun fault Drilling and All-inclusive Sensing MiDAS observatory at beautiful Chishintan (七星潭)



Looking forward to
Future more
applications

THANK YOU



Autonomous Real-Time and On-Demand Monitoring System

