

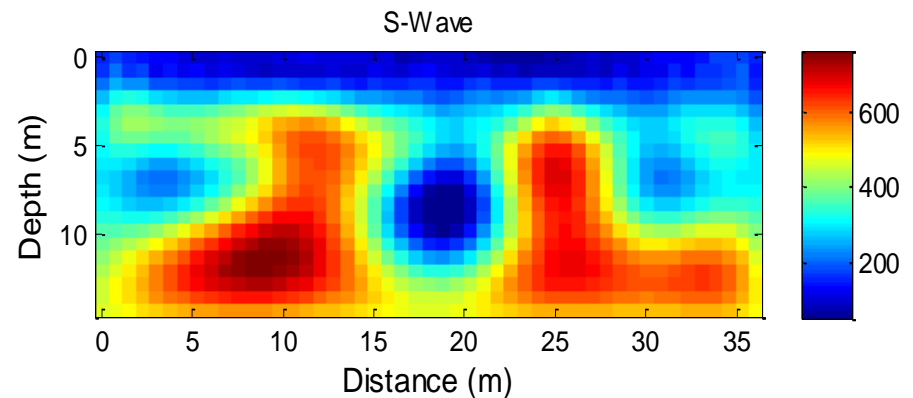
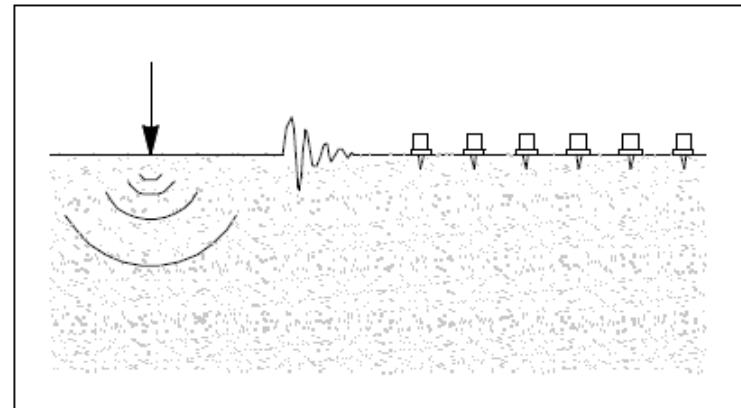
# Site Characterization with 3D Full Seismic Waveform Tomography

Stiffness-based ground deformation  
predictions workshop  
NHERI@UTexas  
Seattle 2018

by

Khiem Tran, Ph.D.

Department of Civil and Coastal  
Engineering



# Outline of presentation

- Need for site investigation
- FWI motivation
- FWI challenges at geotechnical scales
- Overview of FWI
- 3D waveform tomography methods
  - 3D FWI using Adjoint gradient
  - 3D FWI using Gauss-Newton
  - Synthetic data applications
  - Field data applications
- Conclusion

# Need of site investigation

## Problem

- Structural collapses that lead to significant property damage and even fatalities

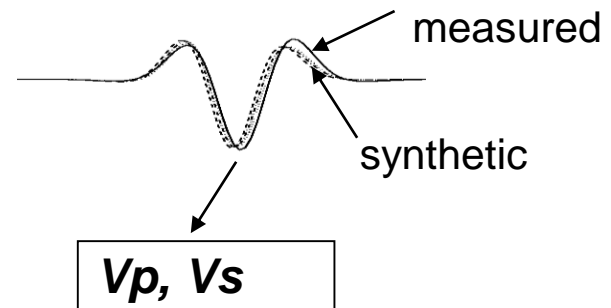
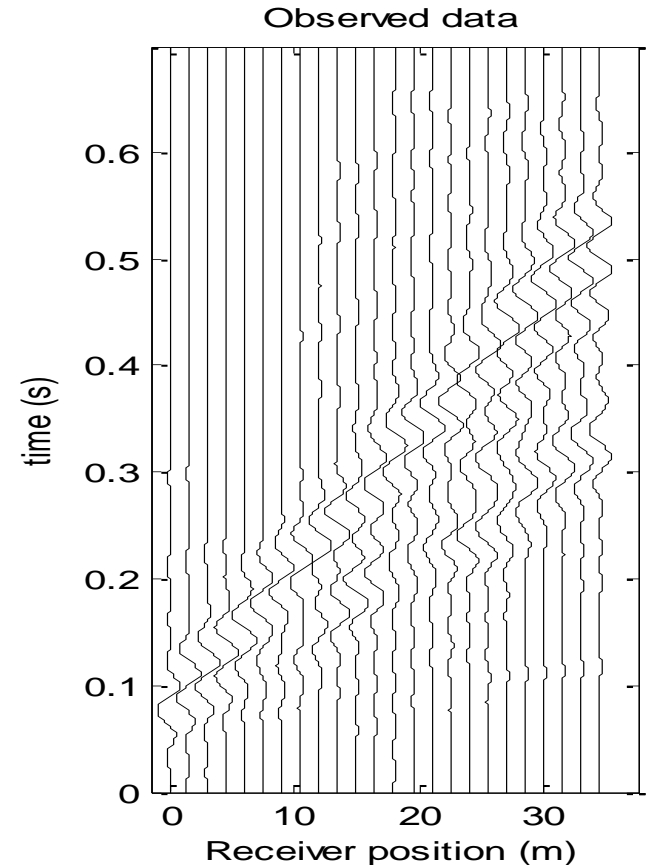
## Site investigation

- Typical invasive testing SPT, CPT – tests  $< .1\%$  of material
- Seismic methods can test over large volume of materials
- Soil/rock property and stratigraphy, and embedded voids/anomalies



# FWI Motivation

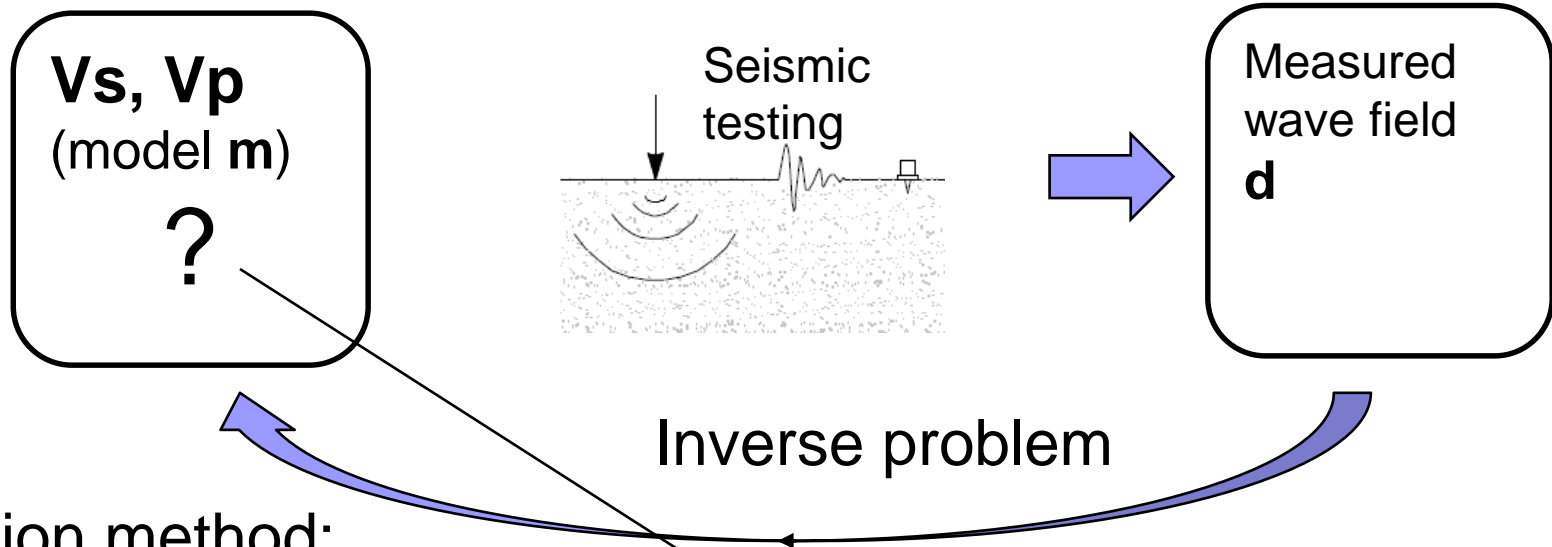
- Conventional seismic methods analyse travel times of certain wave types
  - inversion of P-wave first arrival travel time
  - inversion of surface wave dispersion
  - migration
  - use only phase, not magnitude
- FWI is wave-equation based and has the potential to
  - use full information content (waveforms), both phase and magnitude
  - consider all measured wave types (P-, S-, Rayleigh waves)
  - **characterize both  $V_p$  and  $V_s$  at high resolution (meter pixel)**



# **FWI challenges at geotechnical scales**

- inconsistent wave excitation, unknown source signatures (inversion artifacts near source locations)
- strong variability of near surface soil/rock, poor priori information in the initial model (shallow inversion artifacts, local minimum)
- dominant Rayleigh waves, small body waves with strong attenuation (large model updates at shallow depths, poorly resolved deeper structures)

# Overview of full waveform inversion



Inversion method:

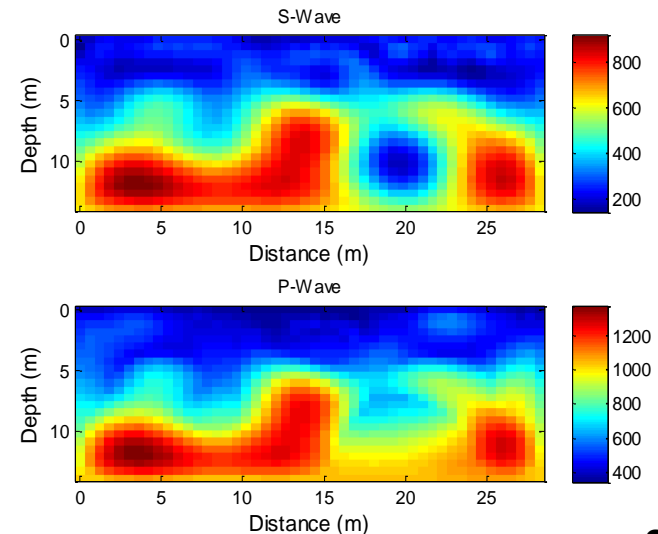
➤ Forward modeling  $\mathbf{d} = f(\mathbf{m})$

- 3D elastic wave equations

$$\rightarrow \mathbf{d}_{\text{est}} = f(\mathbf{m}_{\text{est}})$$

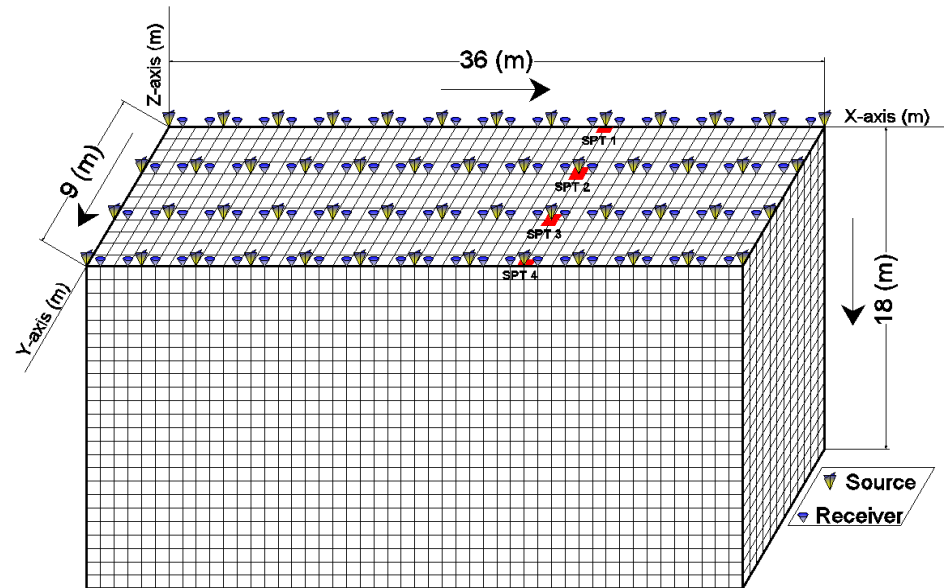
➤ Model updating to match  $\mathbf{d}_{\text{est}} \approx \mathbf{d}$

- Global optimization: simulated annealing, genetic algorithm, ANN
- Deterministic optimization: Gradient, Newton, Gauss-Newton methods

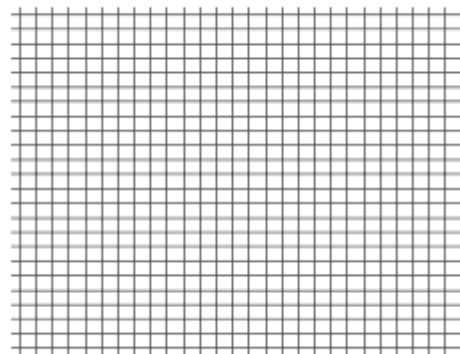
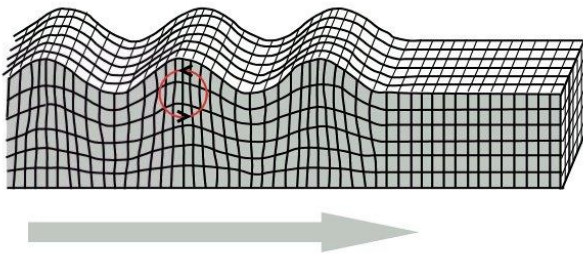


# Data Acquisition

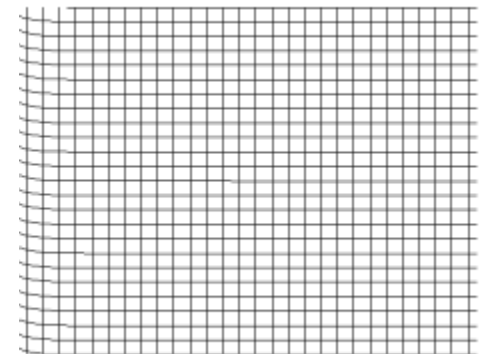
- 2D grids of sources & receivers at 1 to 3 m spacing
- Propelled energy generator or Shaker (5-80 Hz signals)
- P-, S-, and Rayleigh waves are all recorded and used for analysis



Rayleigh Wave



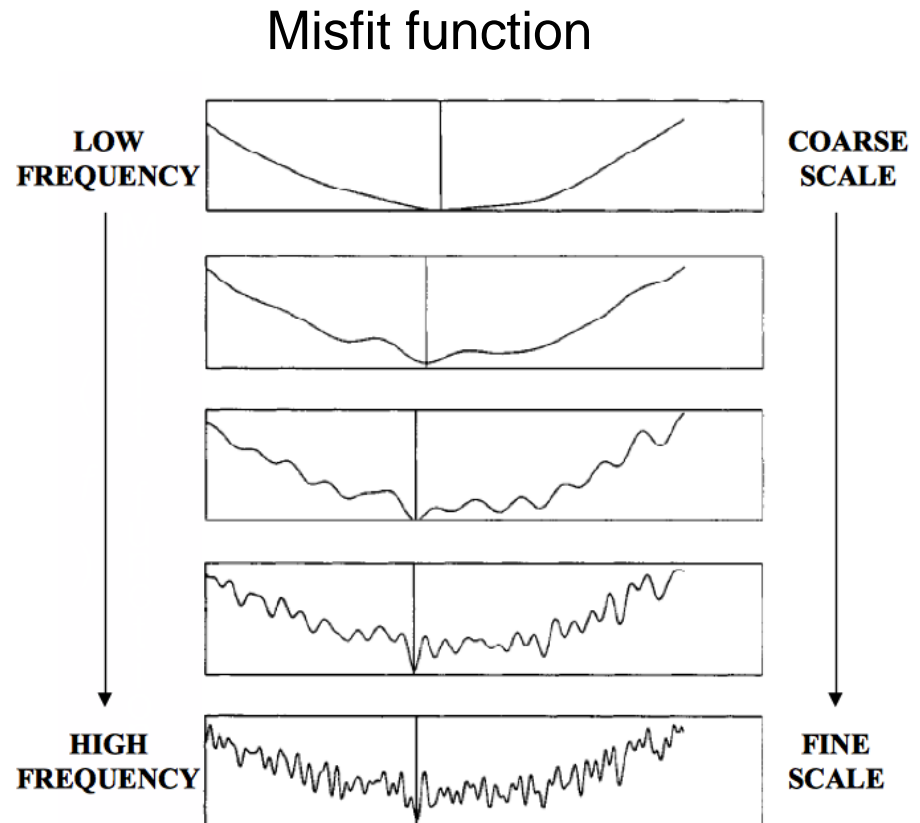
P-wave



S-wave

# Data Analysis

- Start analysis at lowest frequencies and move up
- Low frequencies (large wavelengths) require less detailed information of initial model
- Adding high frequency data gradually helps to resolve detailed features and improve resolution



*Bunks et al. (1995)*



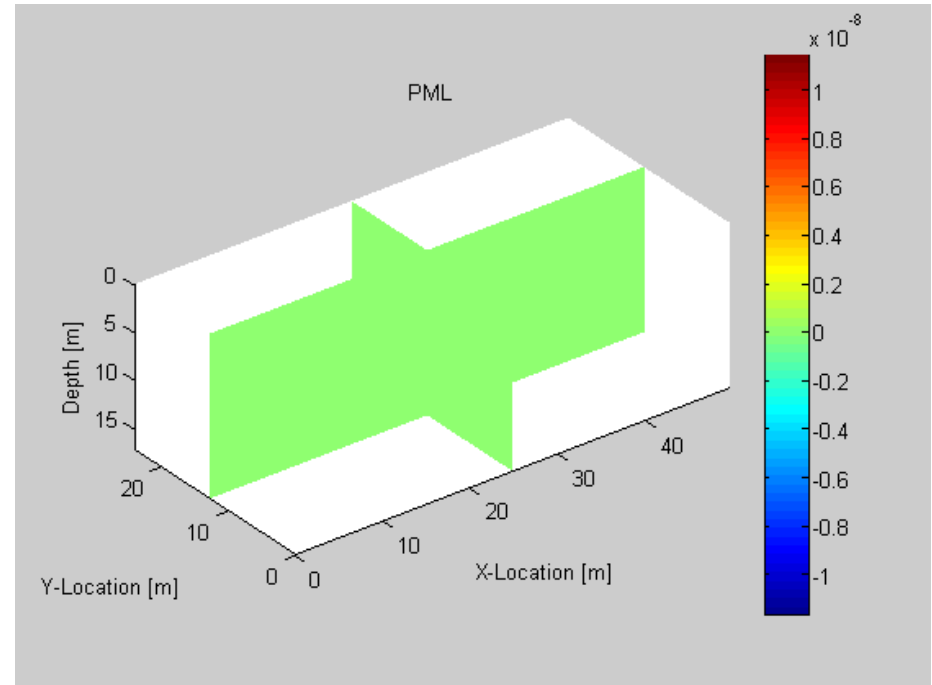
# 3D FWI

## ➤ Forward modeling by 3D wave equations

$$\rho \frac{\partial v_i}{\partial t} = \frac{\partial \sigma_{ij}}{\partial x_j} + f_i \quad \text{where } i, j = 1, 2, 3$$

$$\frac{\partial \sigma_{ij}}{\partial t} = \lambda \frac{\partial v_k}{\partial x_k} + 2\mu \frac{\partial v_i}{\partial x_j} \quad \text{if } i \equiv j$$

$$\frac{\partial \sigma_{ij}}{\partial t} = \mu \left( \frac{\partial v_i}{\partial x_j} + \frac{\partial v_j}{\partial x_i} \right) \quad \text{if } i \neq j$$



Perfectly Matched Layer (PML) is used  
at bottom and 4 vertical boundaries.

# 3D FWI

## ➤ Model updating by **Adjoint Gradient**

Displacement residual: 
$$\Delta u_{i,j}(t) = \int_0^t F_{i,j}(\mathbf{m}, \tau) d\tau - \int_0^t d_{i,j}(\tau) d\tau$$

Misfit function:

$$E(\mathbf{m}) = \frac{1}{2} \Delta \mathbf{u}^t \Delta \mathbf{u}, \text{ where } \Delta \mathbf{u} = \{\Delta u_{i,j}, i = 1, \dots, NS, j = 1, \dots, NR\}$$

Gradients for Lamé's parameters:

$$\delta\lambda = - \sum_{i=1}^{NS} \int_0^T dt \left[ \left( \frac{\partial u_x}{\partial x} + \frac{\partial u_y}{\partial y} \right) \left( \frac{\partial \psi_x}{\partial x} + \frac{\partial \psi_y}{\partial y} \right) + \left( \frac{\partial u_x}{\partial x} + \frac{\partial u_z}{\partial z} \right) \left( \frac{\partial \psi_x}{\partial x} + \frac{\partial \psi_z}{\partial z} \right) + \left( \frac{\partial u_y}{\partial y} + \frac{\partial u_z}{\partial z} \right) \left( \frac{\partial \psi_y}{\partial y} + \frac{\partial \psi_z}{\partial z} \right) \right]$$
$$\delta\mu = - \sum_{i=1}^{NS} \int_0^T dt \left[ \left( \frac{\partial u_x}{\partial y} + \frac{\partial u_y}{\partial x} \right) \left( \frac{\partial \psi_x}{\partial y} + \frac{\partial \psi_y}{\partial x} \right) + \left( \frac{\partial u_x}{\partial z} + \frac{\partial u_z}{\partial x} \right) \left( \frac{\partial \psi_x}{\partial z} + \frac{\partial \psi_z}{\partial x} \right) + \left( \frac{\partial u_y}{\partial z} + \frac{\partial u_z}{\partial y} \right) \left( \frac{\partial \psi_y}{\partial z} + \frac{\partial \psi_z}{\partial y} \right) \right. \\ \left. + 2 \left( \frac{\partial u_x}{\partial x} \frac{\partial \psi_x}{\partial x} + \frac{\partial u_y}{\partial y} \frac{\partial \psi_y}{\partial y} + \frac{\partial u_z}{\partial z} \frac{\partial \psi_z}{\partial z} \right) \right]$$

# 3D FWI

## ➤ Model updating by **Adjoint Gradient**

Gradients for  $V_s$ ,  $V_p$ :

$$\delta V_P = 2\rho V_P \delta \lambda$$
$$\delta V_S = -4\rho V_S \delta \lambda + 2\rho V_S \delta \mu$$

Conditioning gradients:

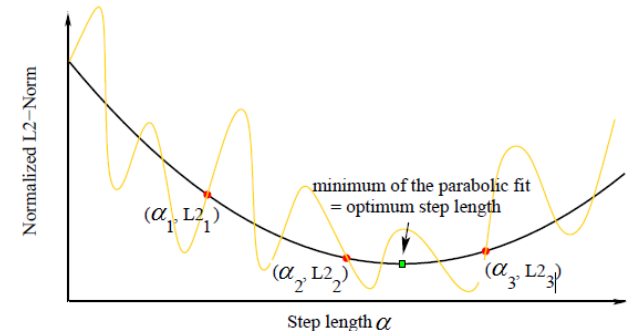
- tapering to suppress large gradient values near source and receiver locations
- tapering to linearly increase the gradient scales with depth to better resolve deeper structures

Regularization:  $\delta^* V_P = R_{V_P} (LV_P) + \delta V_P$

$$\delta^* V_S = R_{V_S} (LV_S) + \delta V_S$$

Model update:

$$V_P^{n+1} = V_P^n - \alpha_P \delta^* V_P$$
$$V_S^{n+1} = V_S^n - \alpha_S \delta^* V_S$$

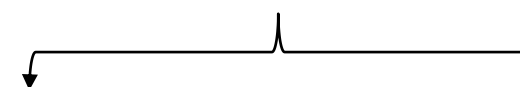


# 3D FWI

## ➤ Model updating by Gauss-Newton

- Velocity residual:  $\Delta \mathbf{d}_{i,j} = \mathbf{F}_{i,j}(\mathbf{m}) - \mathbf{d}_{i,j}$
- Misfit function:  $E(\mathbf{m}) = \frac{1}{2} \Delta \mathbf{d}^t \Delta \mathbf{d}$
- Model updating:  $\mathbf{m}^{n+1} = \mathbf{m}^n - \alpha^n [\mathbf{J}^t \mathbf{J} + \lambda_1 \mathbf{P}^t \mathbf{P} + \lambda_2 \mathbf{I}^t \mathbf{I}]^{-1} \mathbf{J}^t \Delta \mathbf{d},$   

Filter, focus,  
balance gradient  
vector


- Jacobian matrix:  $\mathbf{J}_{i,j} = \frac{\partial \mathbf{F}_{i,j}(\mathbf{m})}{\partial m_p}$
- Gauss-Newton inversion is done in frequency domain to reduce RAM

$$\tilde{u}(\mathbf{x}, \omega) = \sum_{l=1}^{nt} \exp(\sqrt{-1} \omega l \Delta t) u(\mathbf{x}, l \Delta t) \Delta t$$

# Derivative wave-field (Jacobian Matrix J)

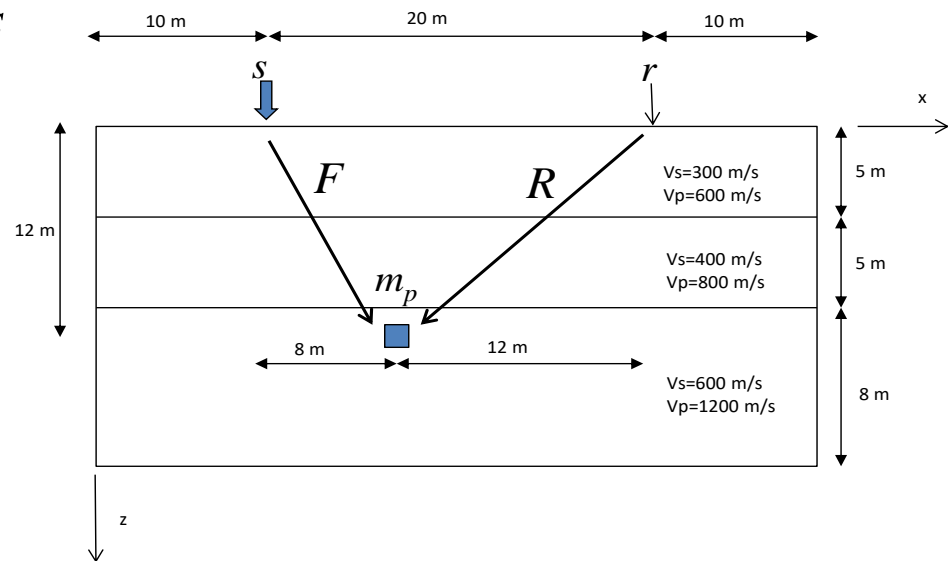
## ➤ Explicit

- two forward simulations with and without the model perturbation for each unknown
- Required number of forward simulations = *number of shots*  $\times$  (*number of unknowns* + 1)

$$J_{i,j}^p = \frac{\partial \mathbf{F}_{i,j}(\mathbf{m})}{\partial m_p} = \frac{\mathbf{F}_{i,j}(\mathbf{m} + \Delta m_p) - \mathbf{F}_{i,j}(\mathbf{m})}{\Delta m_p}$$

## ➤ Implicit

- Virtual source (F) and reciprocal wave-fields (R)
- Required number of forward simulations = (*number of shots* + *number of receivers*)



$$J_{i,j}^p = F_x * R_x + F_y * R_y + F_z * R_z$$

# Virtual source for $V_s$

Wave equations are differentiated with respect a parameter  $V_{s_n}$ :

$$\rho \frac{\partial v_i}{\partial V_{s_n}} = \frac{\partial \sigma_{ij,j}}{\partial V_{s_n}}$$

Virtual source

$$\rho \frac{\partial \sigma_{ij}}{\partial V_{s_n}} = \lambda \frac{\partial v_{k,k}}{\partial V_{s_n}} + 2\mu \frac{\partial v_{i,j}}{\partial V_{s_n}} \left( -4\rho V_s \frac{\partial V_s}{\partial V_{s_n}} (v_{k,k} - v_{j,i}) \right) \quad \text{if } i = j$$

$$\rho \frac{\partial \sigma_{ij}}{\partial V_{s_n}} = \mu \left( \frac{\partial v_{i,j}}{\partial V_{s_n}} + \frac{\partial v_{j,i}}{\partial V_{s_n}} \right) \left( +2\rho V_s \frac{\partial V_s}{\partial V_{s_n}} (v_{i,j} + v_{j,i}) \right) \quad \text{if } i \neq j$$

# Virtual source for $V_p$

Wave equations are differentiated with respect a parameter  $V_{p_n}$ :

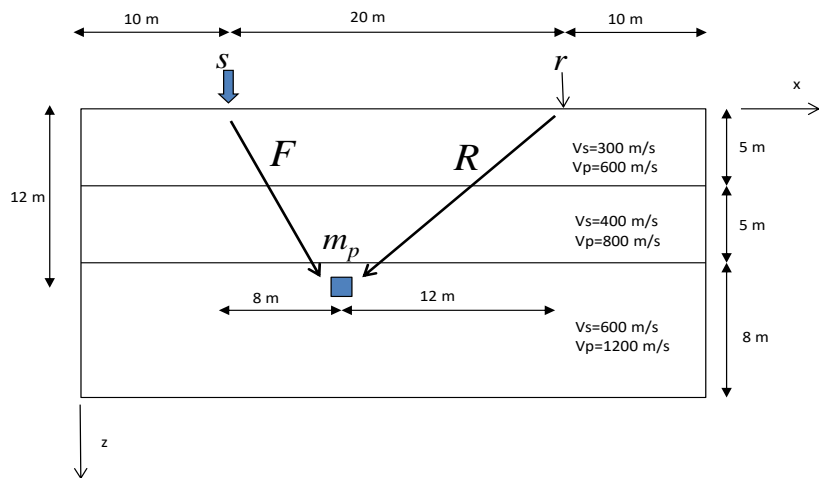
$$\rho \frac{\partial v_i}{\partial V_{p_n}} = \frac{\partial \sigma_{ij,j}}{\partial V_{p_n}}$$

Virtual source

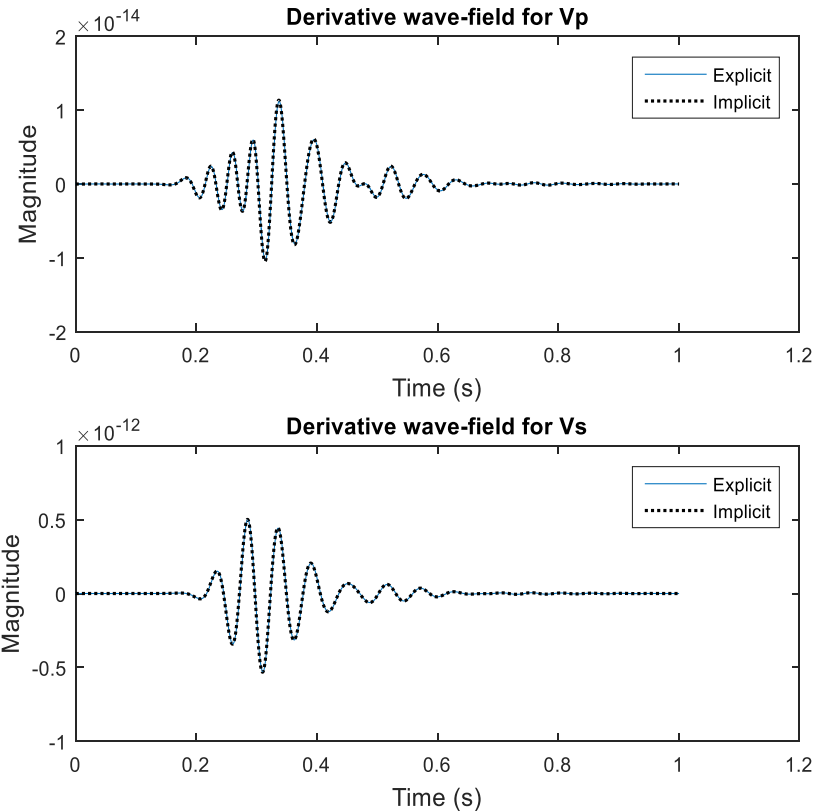
$$\rho \frac{\partial \sigma_{ij}}{\partial V_{p_n}} = \lambda \frac{\partial v_{k,k}}{\partial V_{p_n}} + 2\mu \frac{\partial v_{i,j}}{\partial V_{p_n}} + 2\rho V_p \frac{\partial V_p}{\partial V_{p_n}} v_{k,k} \quad \text{if } i = j$$

$$\rho \frac{\partial \sigma_{ij}}{\partial V_{p_n}} = \mu \left( \frac{\partial v_{i,j}}{\partial V_{p_n}} + \frac{\partial v_{j,i}}{\partial V_{p_n}} \right) \quad \text{if } i \neq j$$

# Comparison of derivative wave-field



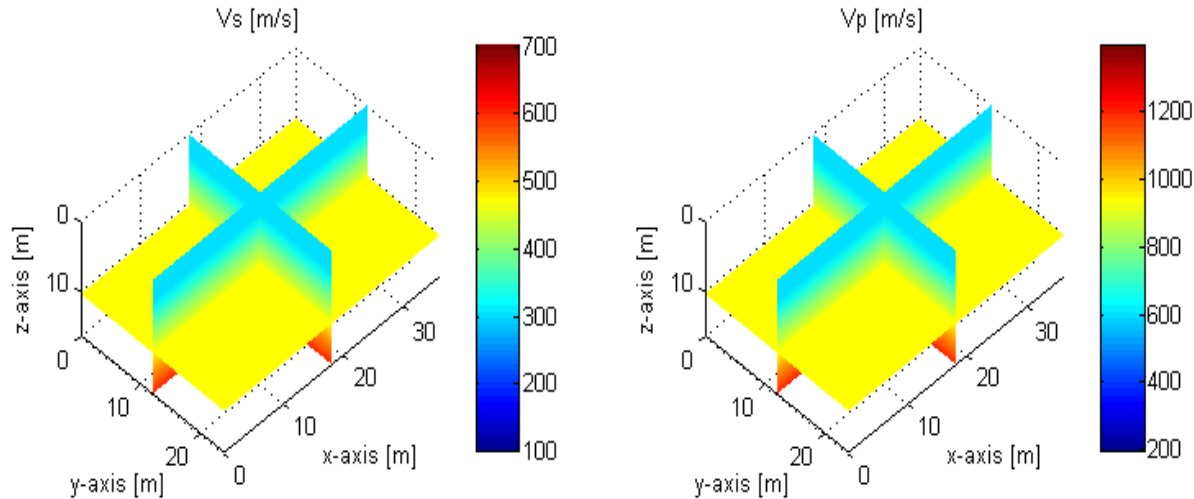
40 x 40 x 18 m  
model of 3 layers



Explicit and Implicit are identical



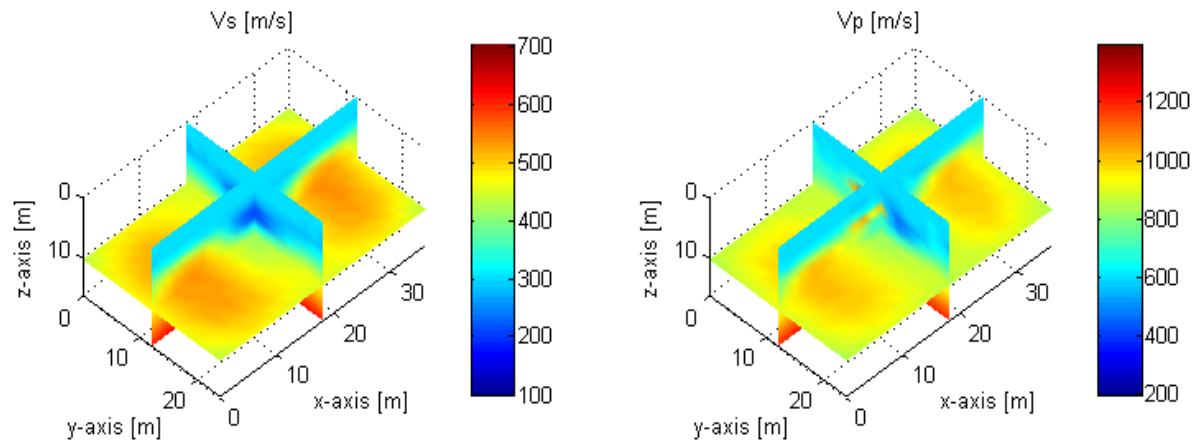
# 3D FWI: Synthetic test



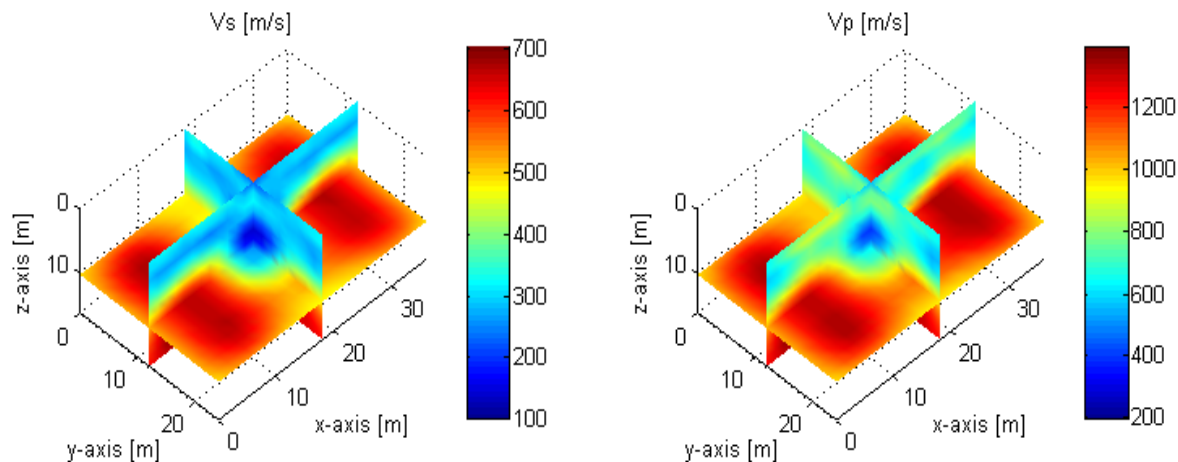
Initial model used for both  
Adjoint and GN inversion

- 2 inversion runs at 15 and 25 Hz central frequencies
- about 40 hours for both Adjoint gradient and Gauss-Newton inversions on a desktop computer (32 cores of 3.46 GHz each and 256 GB of memory)

# 3D FWI: Synthetic test results



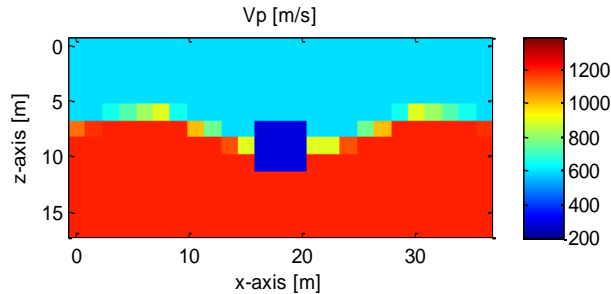
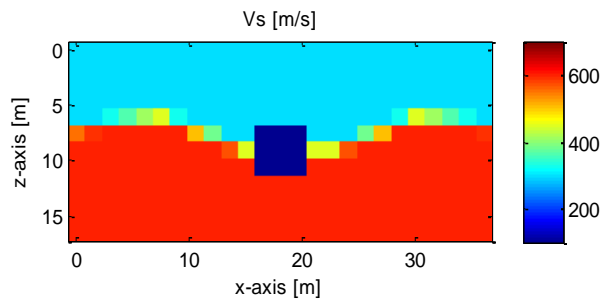
Adjoint gradient



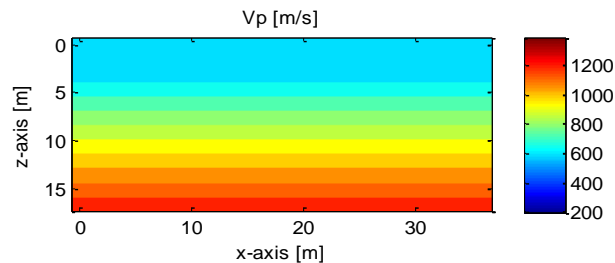
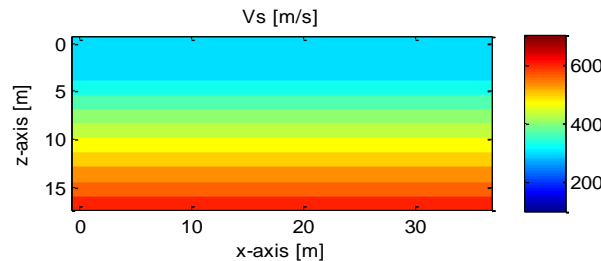
Gauss-Newton

# 3D FWI: plane comparison at void center

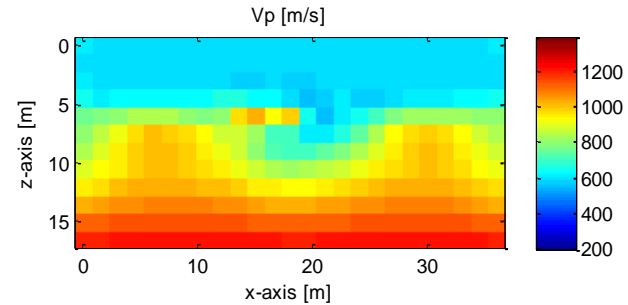
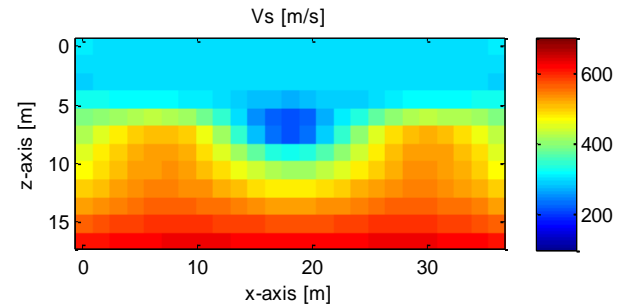
True  
model



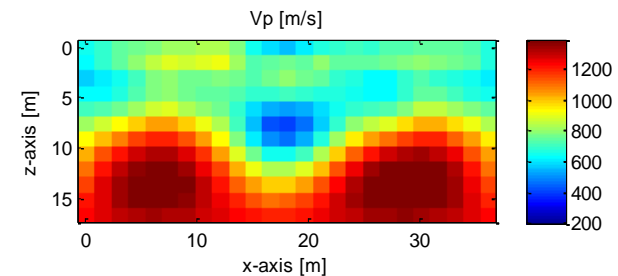
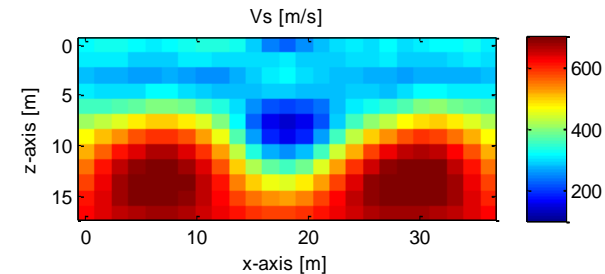
Initial  
model



Adjoint  
gradient

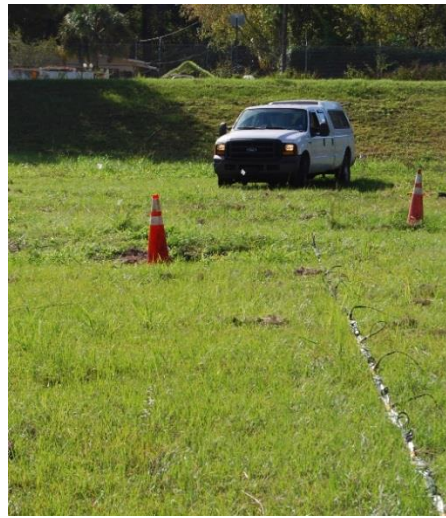


Gauss-  
Newton

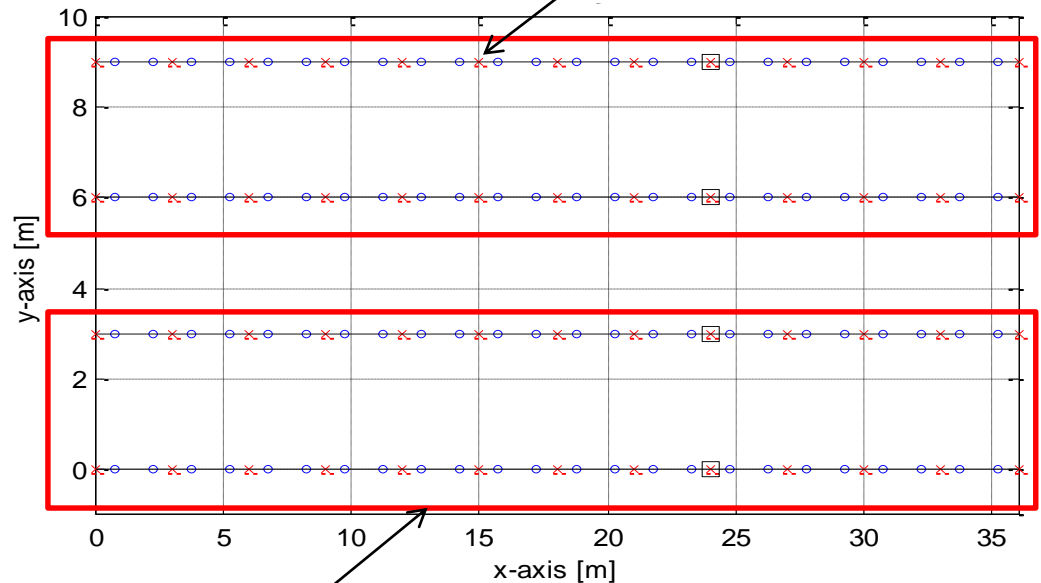


# Gainesville site

- dry retention pond in Gainesville, FL
- test area of 36 x 9 m
- 96 receivers located in 24 x 4 grid
- 52 shots located in 13 x 4 grid
- 48 geophones twice
- PEG active source



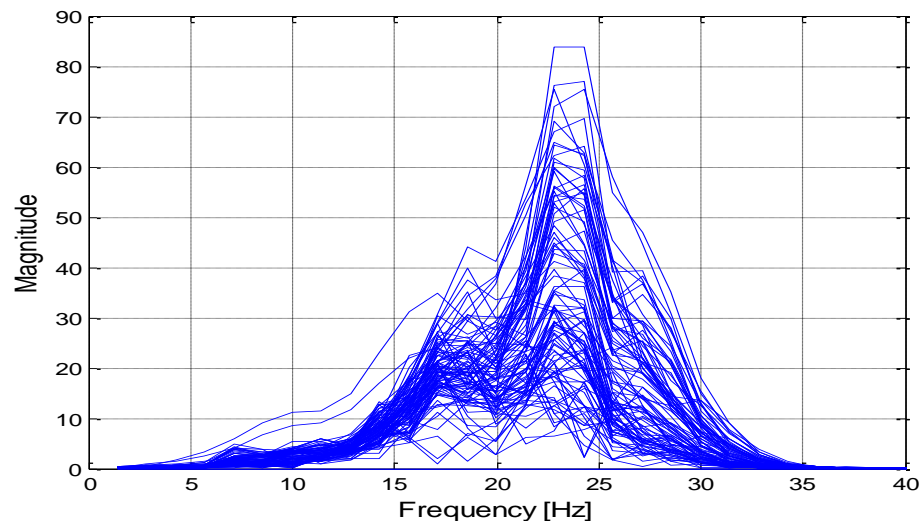
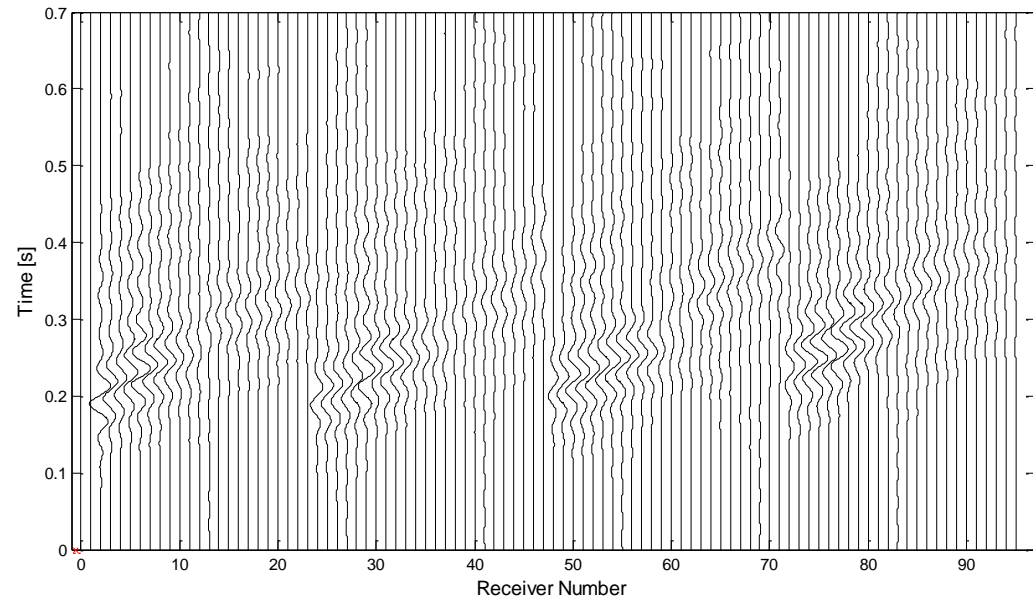
Stage 1



Stage 2

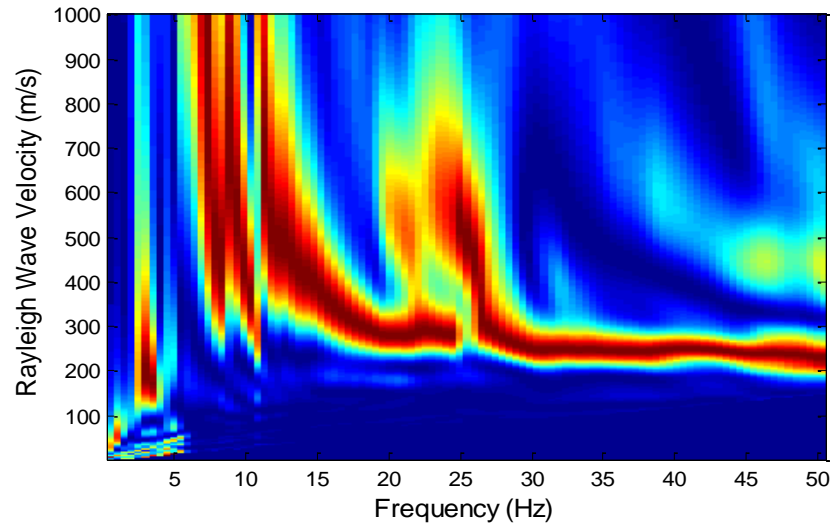
# Gainesville site: sample field data

- measured data combined from the two stages for 96-channel shot gather
- consistent wave magnitudes and propagation pattern

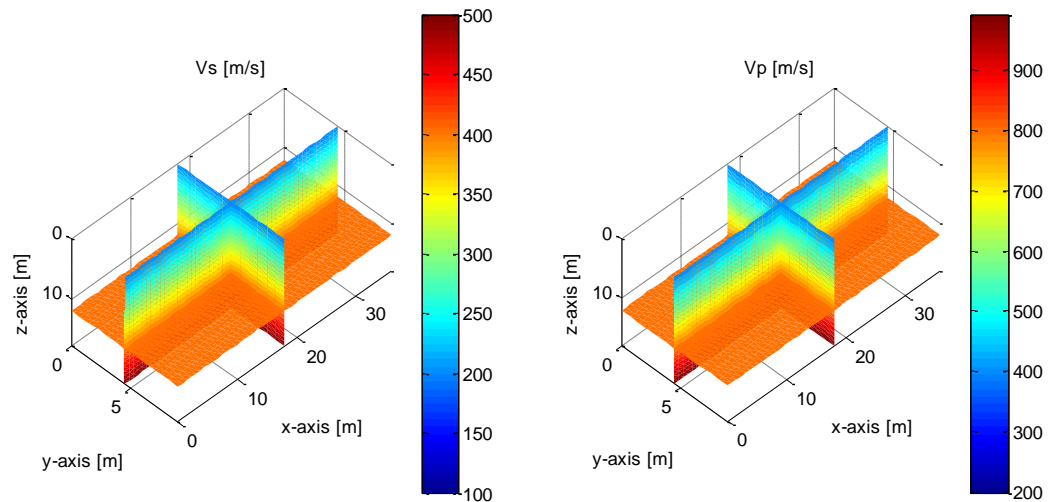


# Gainesville site: data analysis

- 2 inversion runs at 12 and 22 Hz central frequencies
- Medium divided into about 14,000 cells of 0.75 m
- About 30 hours for both Adjoint gradient and Gauss-Newton methods

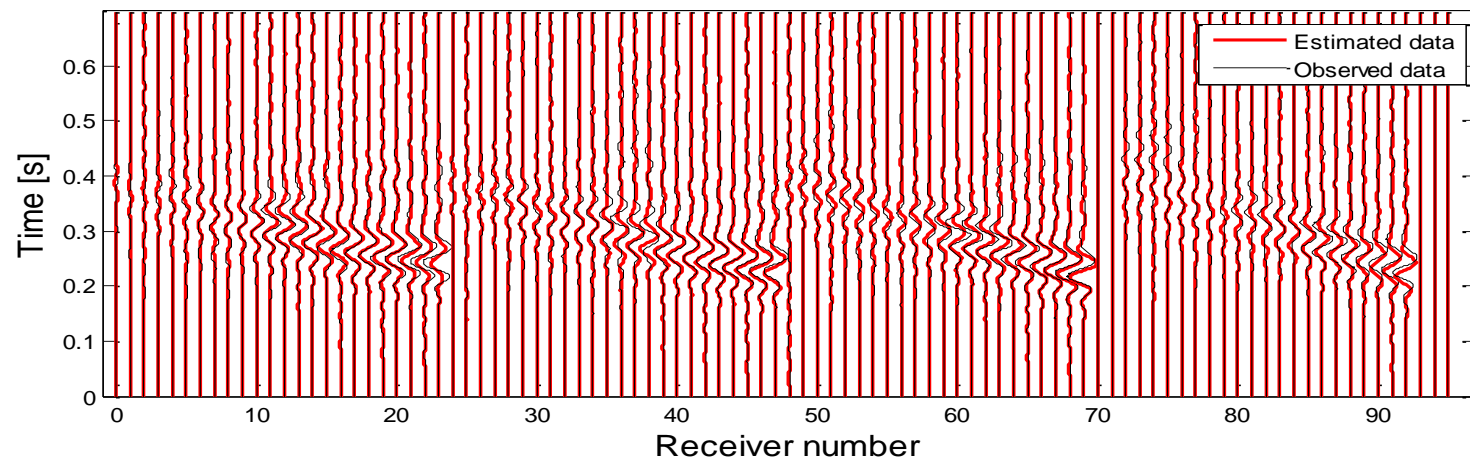
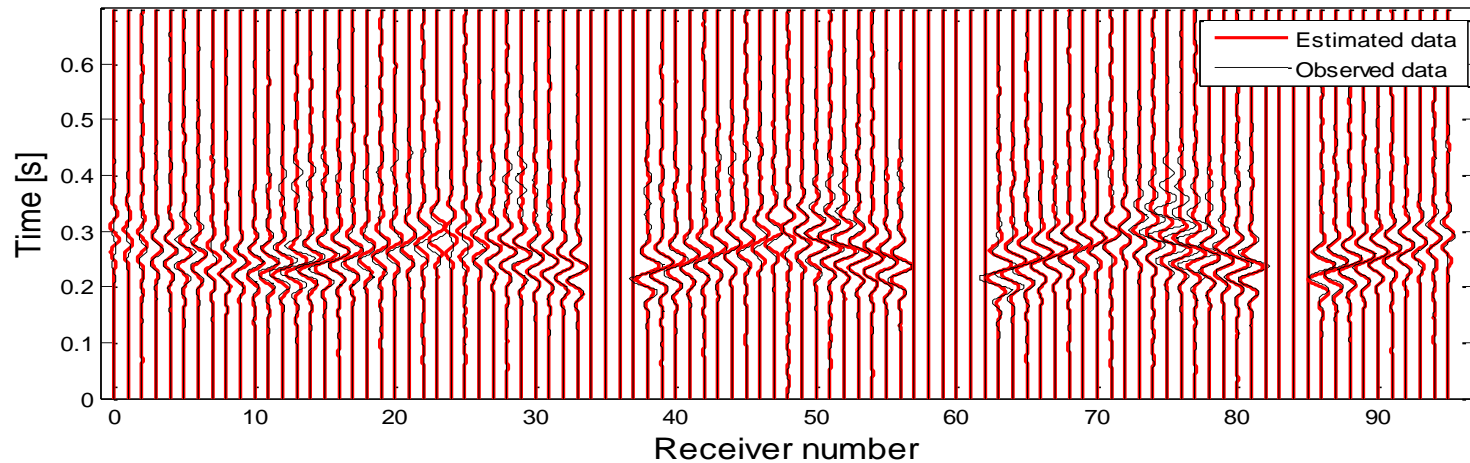


## ■ Power spectrum



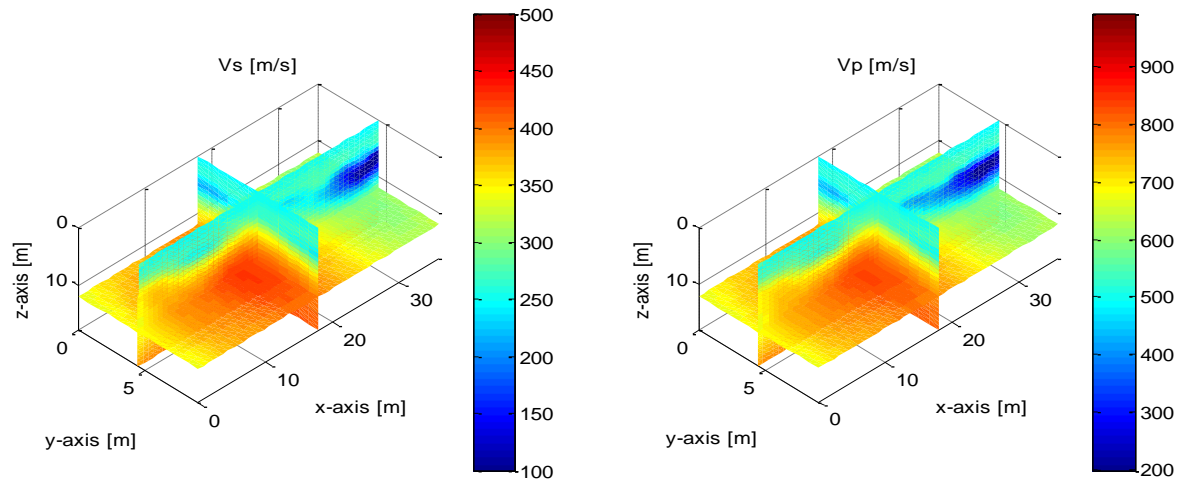
## ■ Initial model

# Gainesville site: data analysis

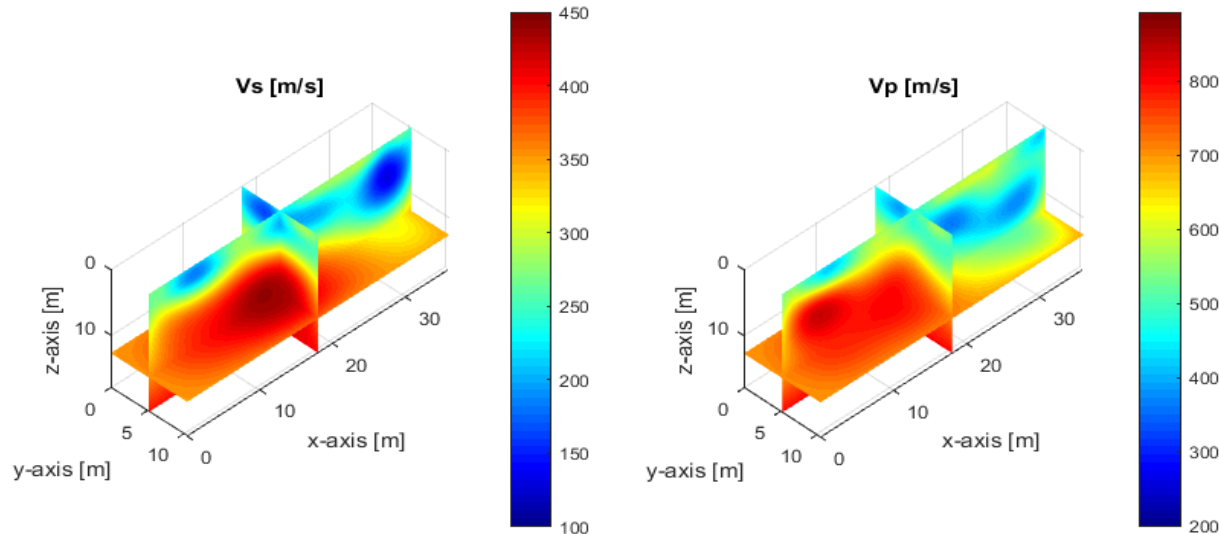


Waveform comparison for 2 sample shots

# Gainesville site: 3D FWI results



Adjoint gradient

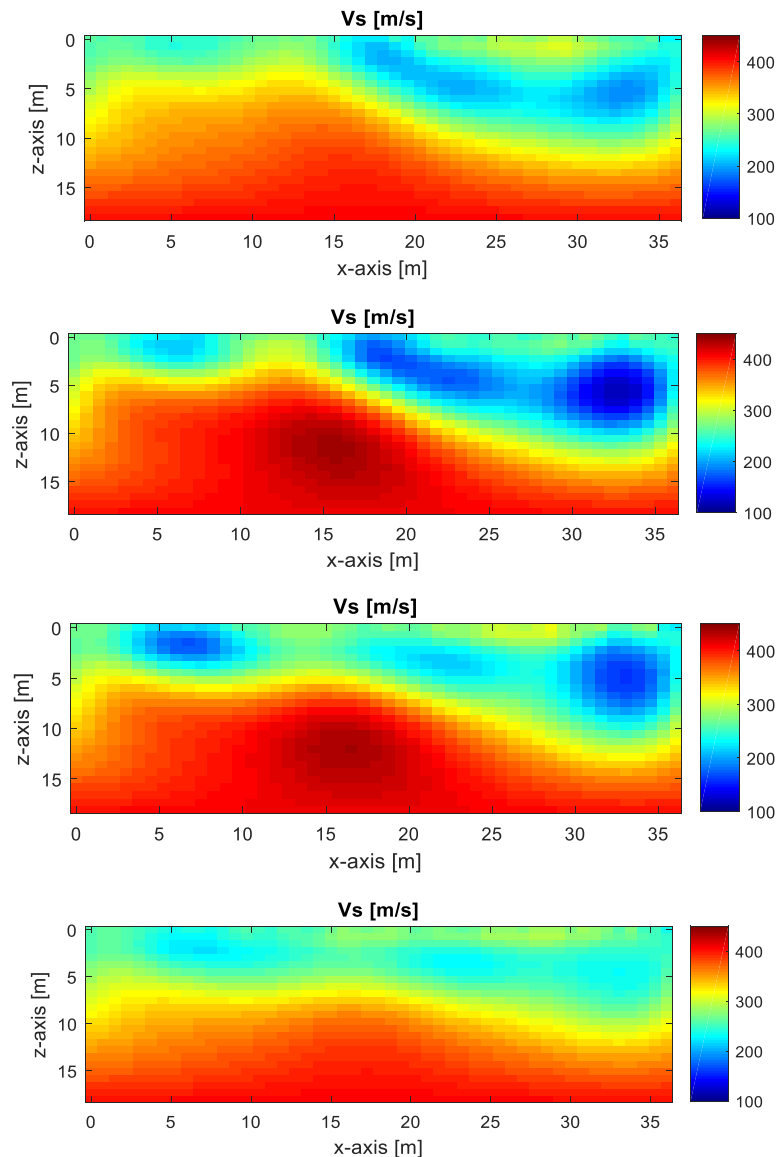


Gauss-Newton

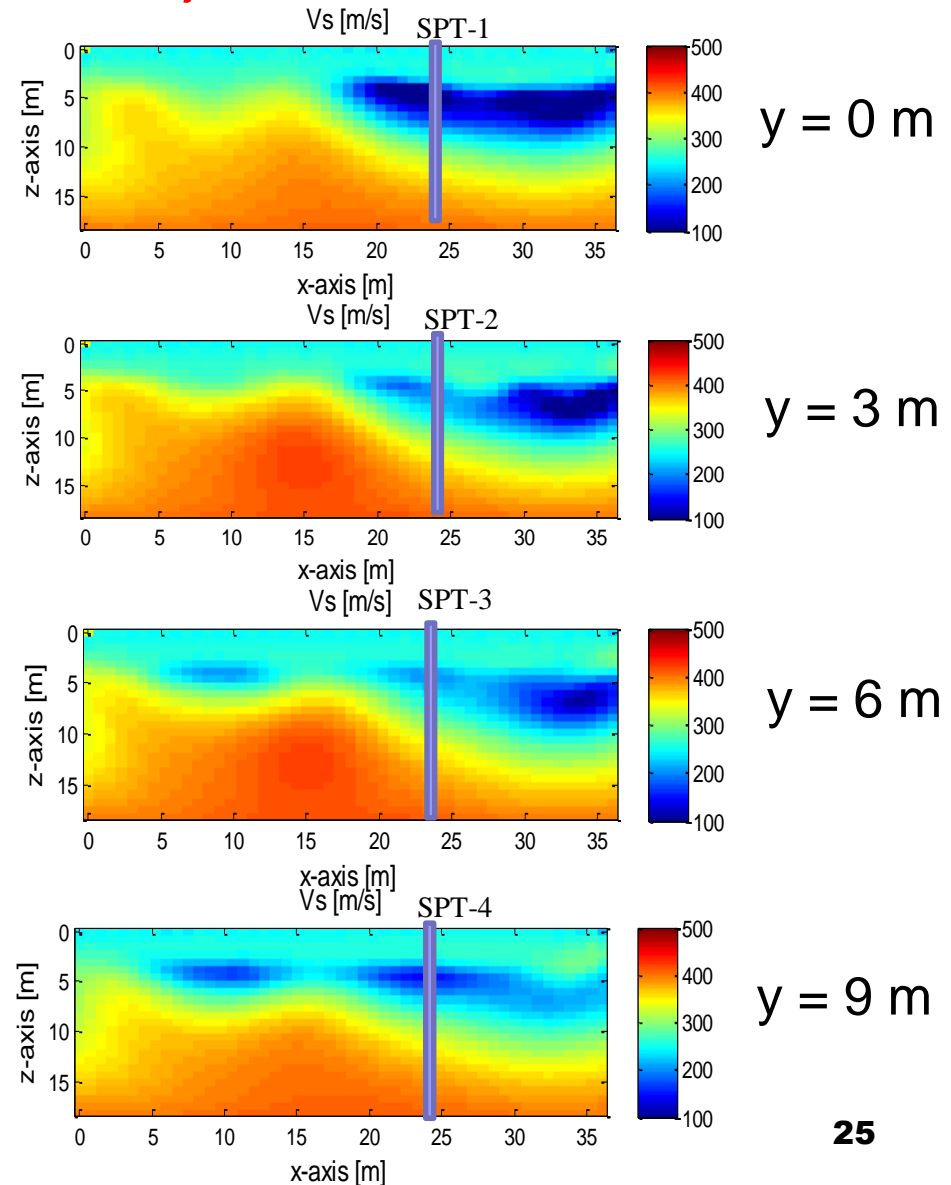


# Gainesville site: 3D FWI results at planes

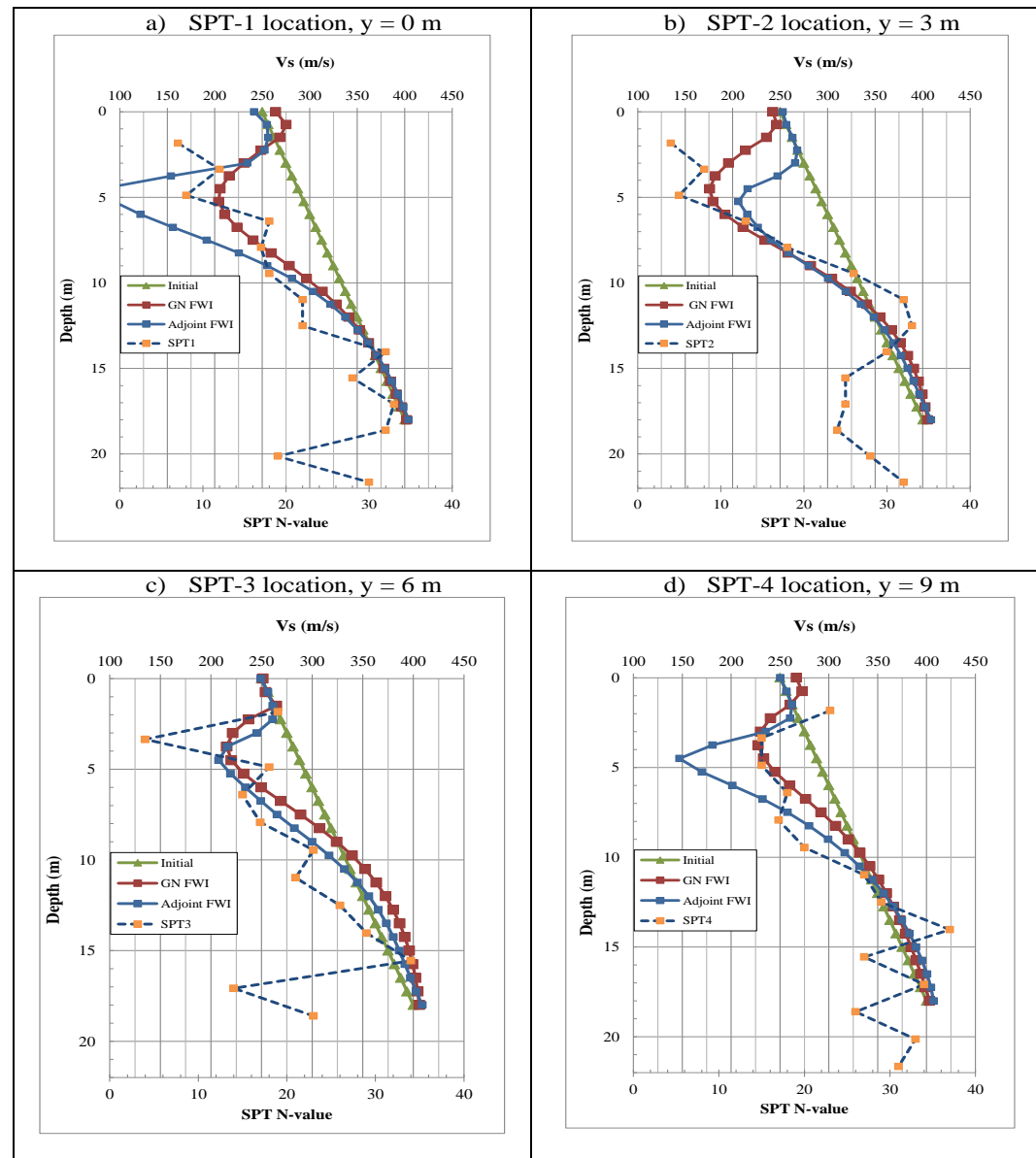
## Gauss-Newton



## Adjoint Gradient

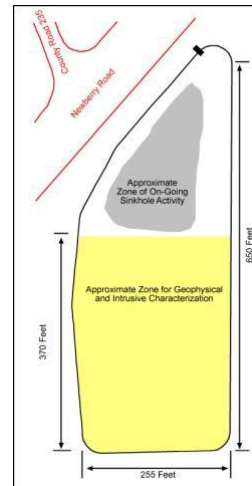
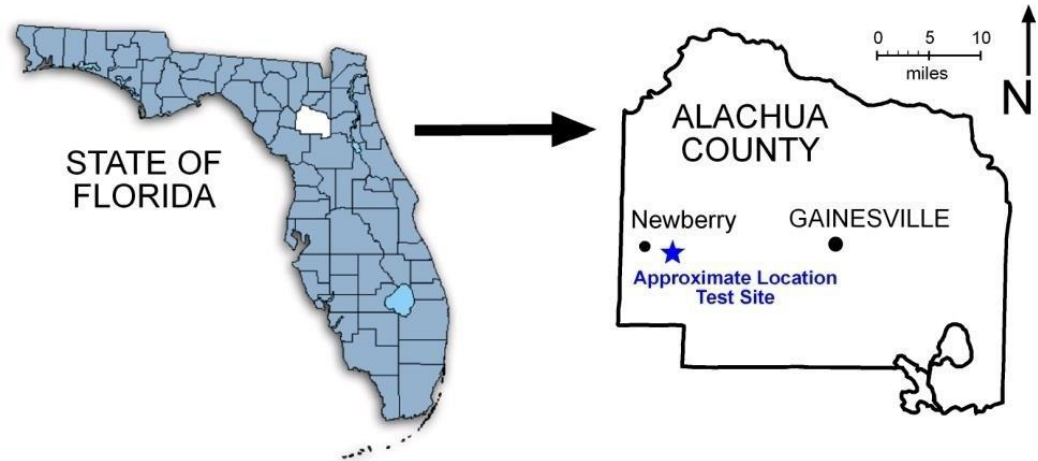


# 3D FWI vs. SPT results



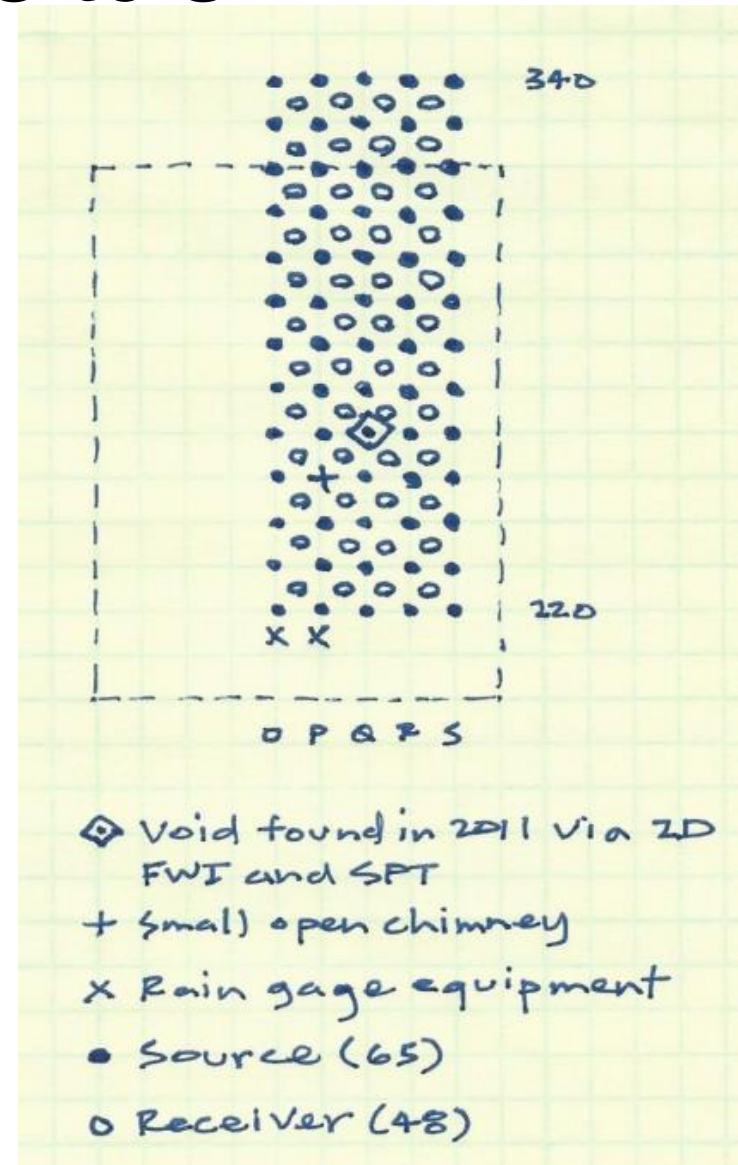
# Newberry site

- Dry retention pond in Newberry, FL
- Mix sand and clay over lime stone bedrock
- Site was marked by 25 lines (A to Y) at 3 m spacing
- Data were collected by NHERI @UTexas team using 48 4.5 Hz vertical geophones and Thumper shaking source.



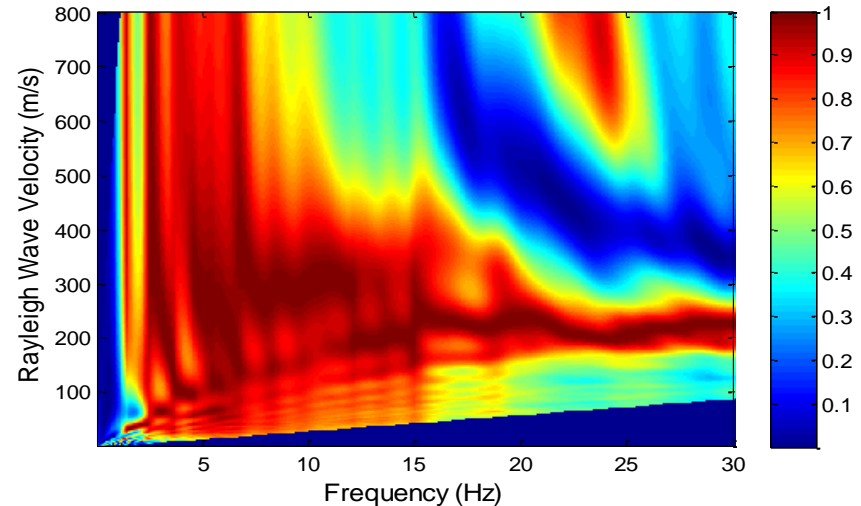
# Seismic Survey at Lines O to S

- Test area of 36 x 12 m
- 48 geophones located in 12 x 4 grid
- 65 shots located in 13 x 5 grid
- Thumper shaker source

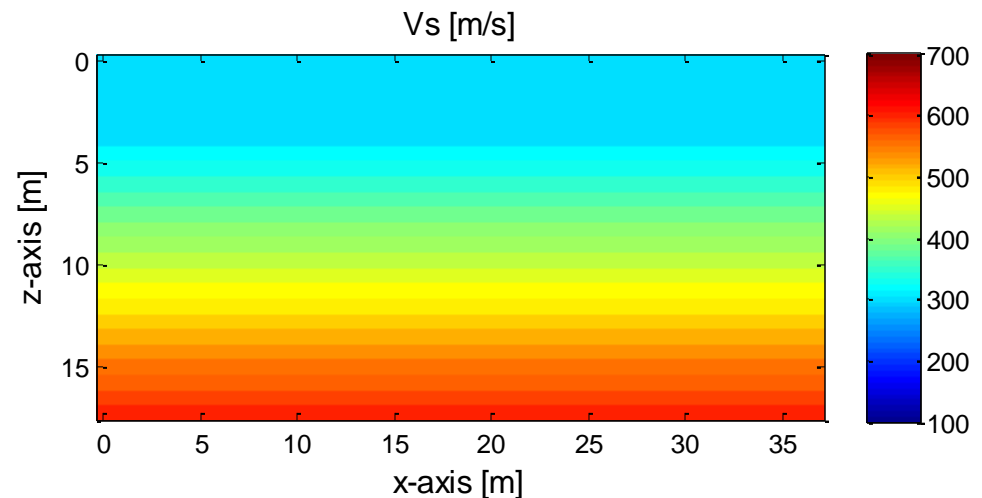


# Newberry data analysis

- Analyzed by Gauss-Newton 3D FWI
- 2 inversion runs at 10 and 25 Hz central frequencies
- 60 hours on a desktop computer (32 cores of 3.46 GHz each and 256 GB of memory)



■ Power spectrum

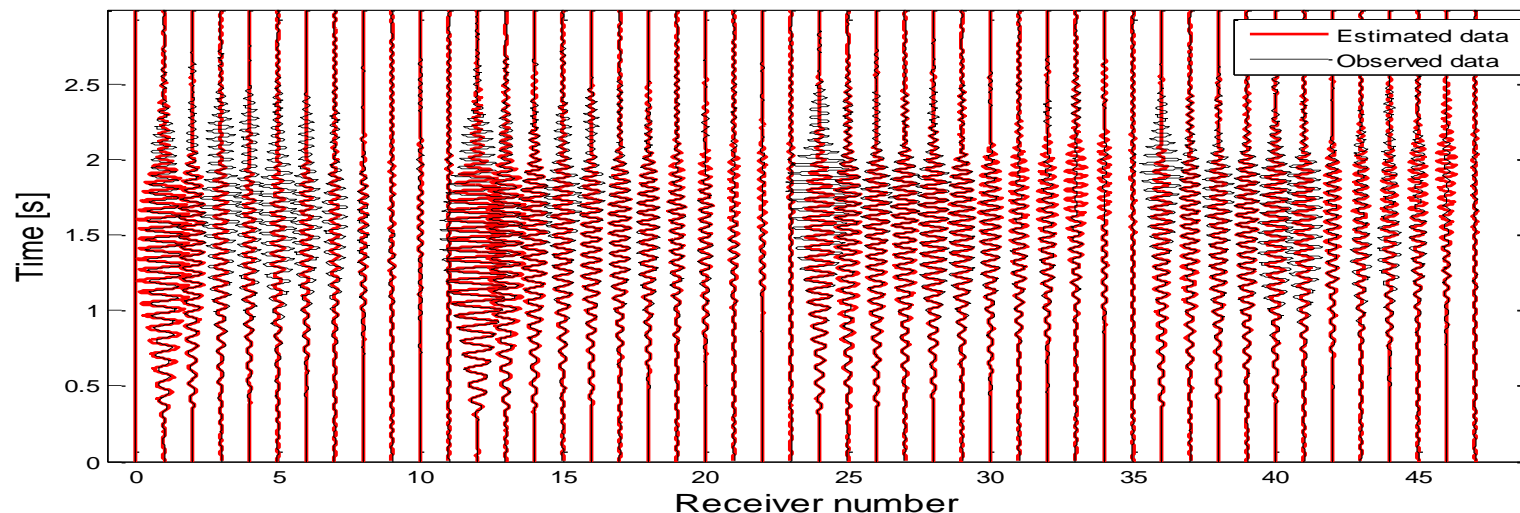


■ Initial model

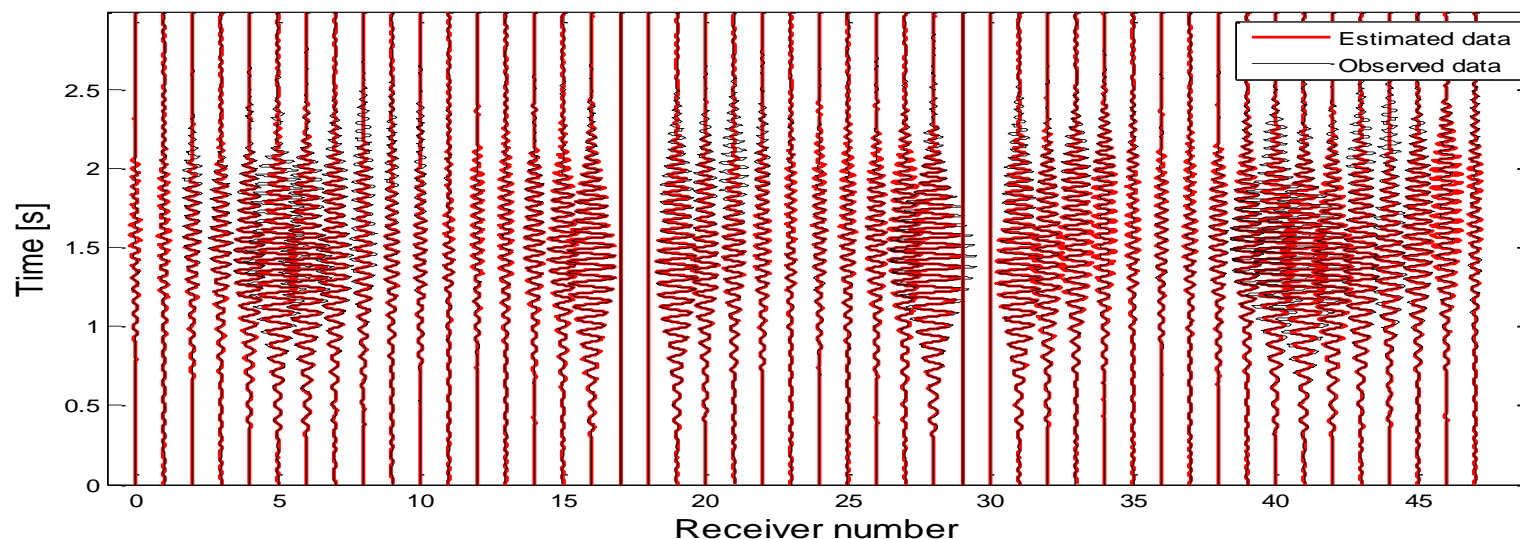


# Newberry: data analysis

Shot 1  
Line O  
x= 0m  
y= 0m



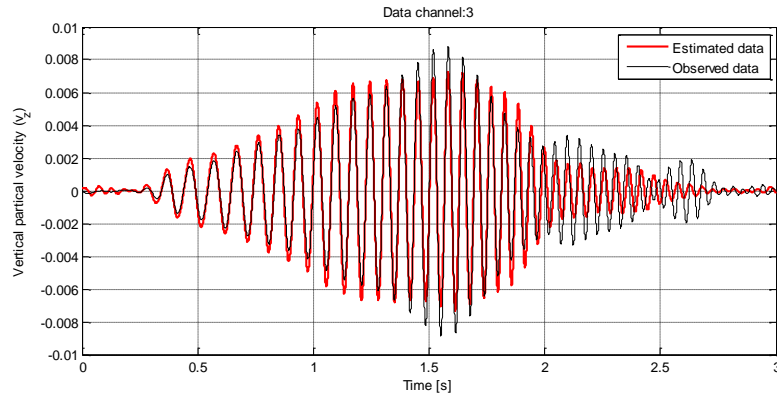
Shot 33  
Line Q  
x= 18m  
y= 6m



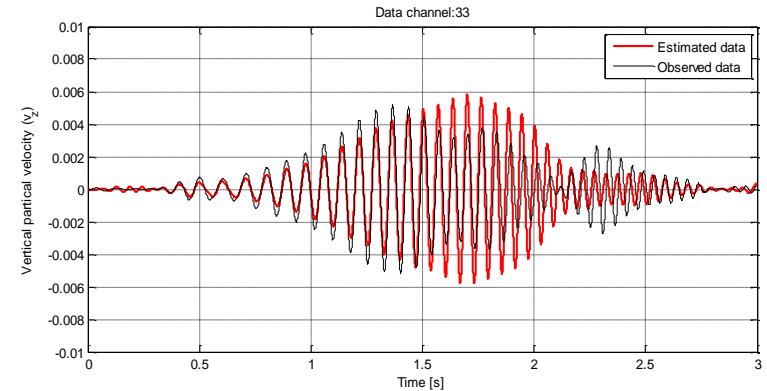
Waveform comparison for 2 sample shots

# Newberry: data analysis

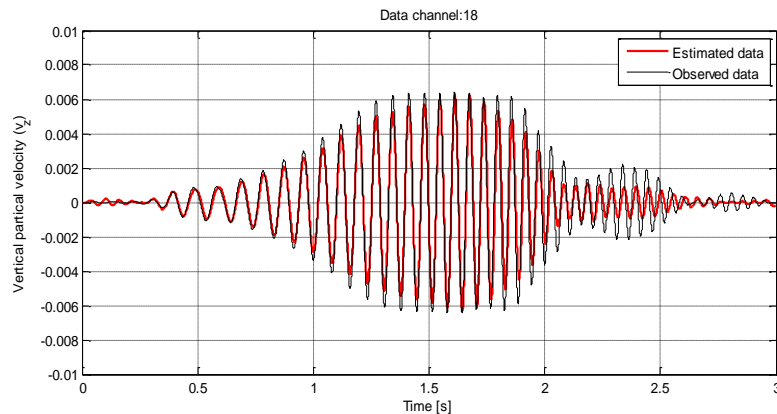
Channel 3, Line OP, x= 7.5m, y= 1.5m



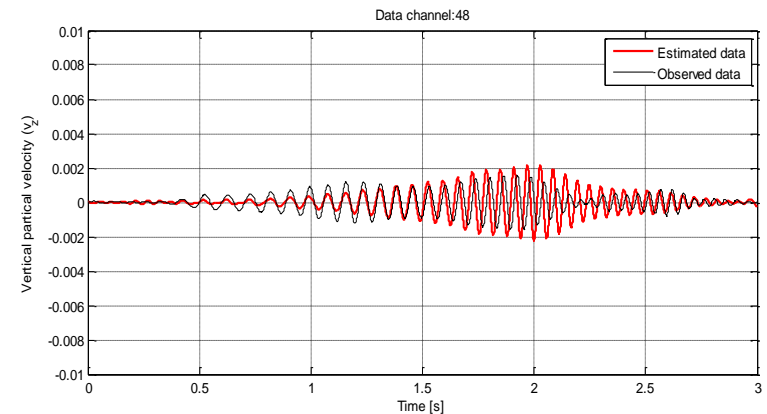
Channel 33, Line QR, x= 25.5m, y= 7.5m



Channel 18, Line PQ, x= 16.5m, y= 4.5m

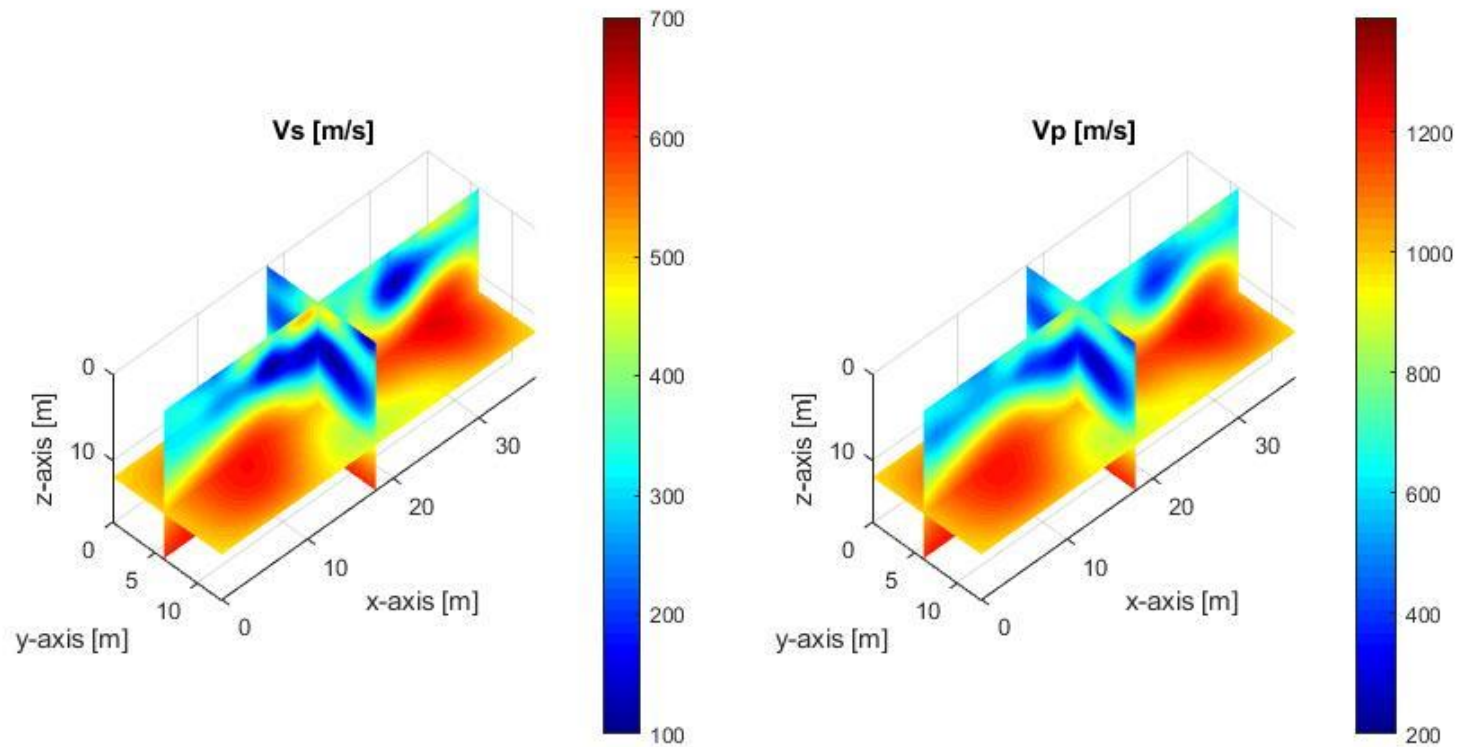


Channel 48, Line RS, x= 34.5m, y= 10.5m



Waveform comparison for 4 sample channels for shot 1 at x=0m, y=0m

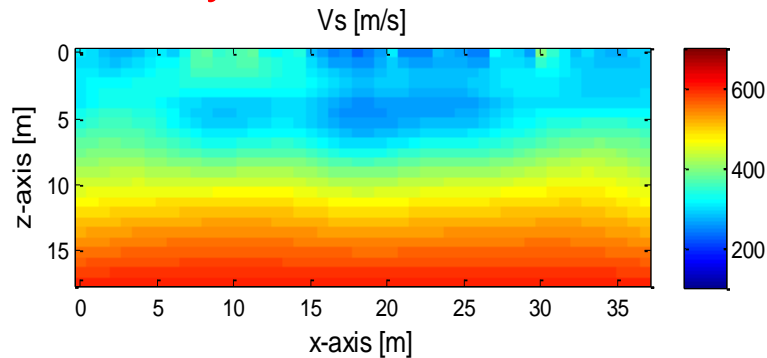
# Newberry: 3D FWI Results



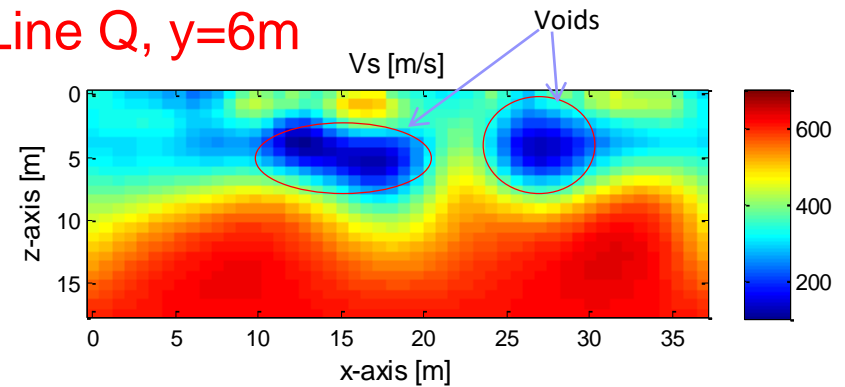


# Newbery: 3D FWI Results at planes

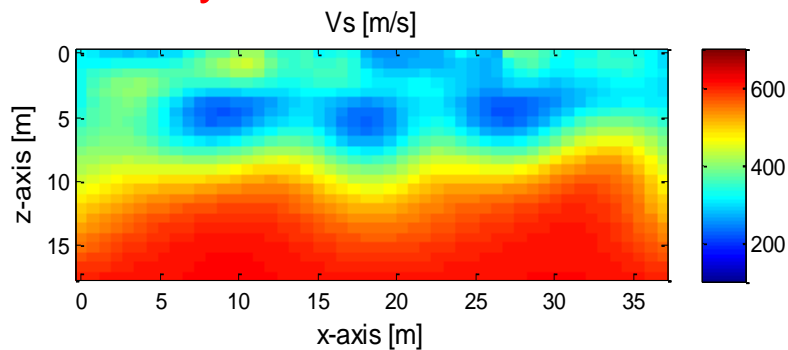
Line O,  $y=0\text{m}$



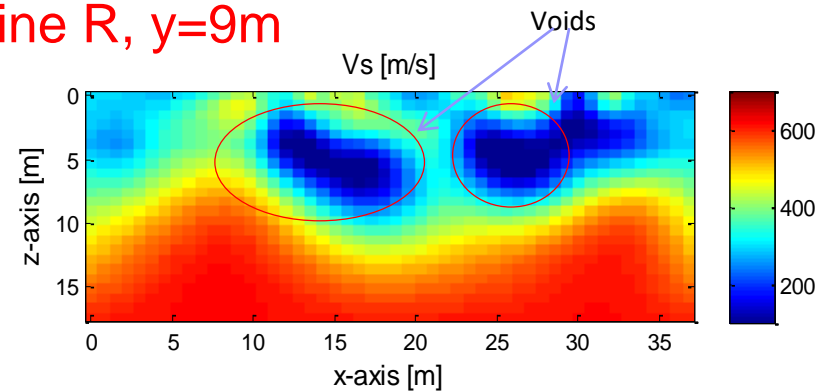
Line Q,  $y=6\text{m}$



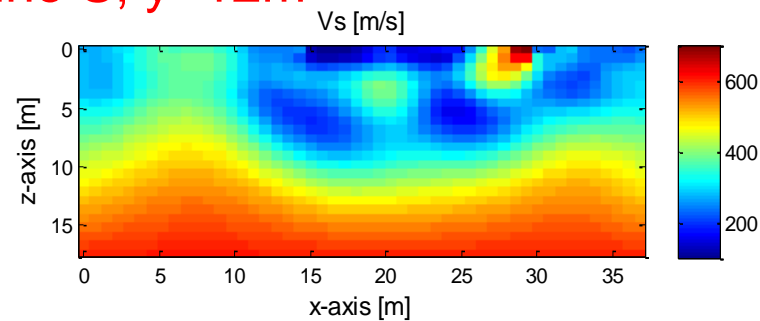
Line P,  $y=3\text{m}$



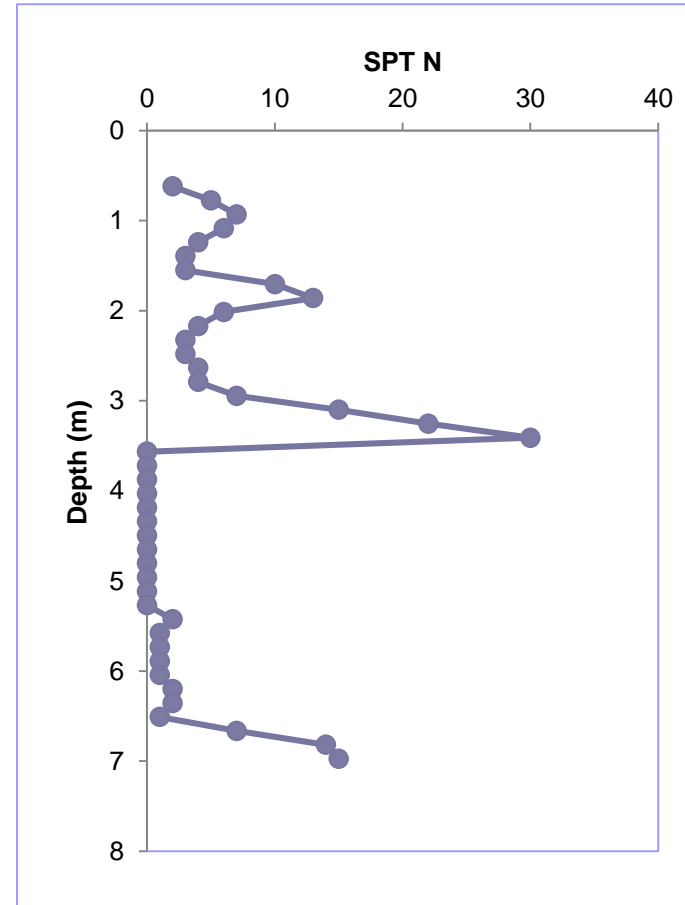
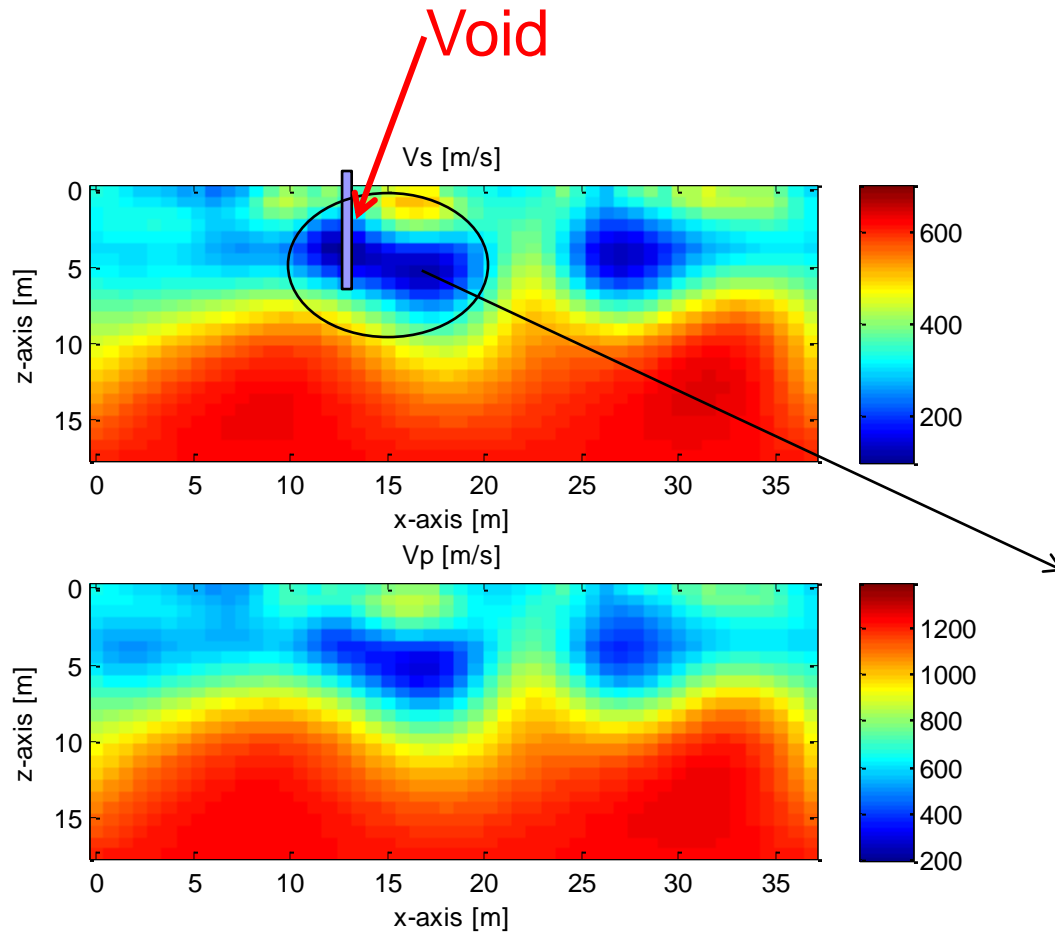
Line R,  $y=9\text{m}$



Line S,  $y=12\text{m}$



# Newbery: Void at Line Q



# Conclusion

- Both  $V_s$  and  $V_p$  can be characterized at high resolution (meter pixel) to 20 m in depth by 3-D FWI methods
- Buried void can be identified to a depth of about 3 void diameters with surface measurement
- Gauss-Newton provides better results than Adjoint gradient inversion, particularly for void imaging

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# References

- Tran K.T, Mirzanejad M. McVay M. and Horhota D., “3D Time-Domain Gauss-Newton Full Waveform Inversion for Near-Surface Site Characterization”, ***Geophysical Journal International***, under 2<sup>nd</sup> review.
- Nguyen D.T. and Tran K.T. (2018), “Site Characterization with 3-D Elastic Full Waveform Tomography”, ***Geophysics***, Vol. 83 (5), pp. R389–R400
- Tran K.T. and Sperry J. (2018), “Application of 2-D Full Waveform Tomography on Land-streamer Data for Assessment of Roadway Subsidence”, ***Geophysics***, SEG, Vol. 83, (3), pp. EN1–EN11.
- Tran K.T., McVay M., Horhota D., and Faraone M. (2013), “Sinkhole Detection Using 2D Full Seismic Waveform Tomography”, ***Geophysics***, Vol. 78 (5), pp. R175–R183.

# *Thank You!*

