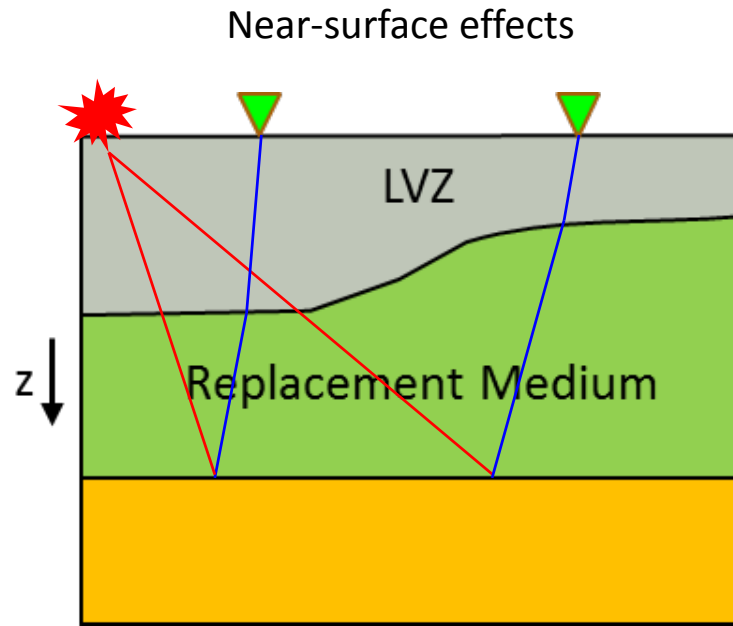


# PS-wave traveltimes difference inversion for near-surface characterization in the tau-p domain

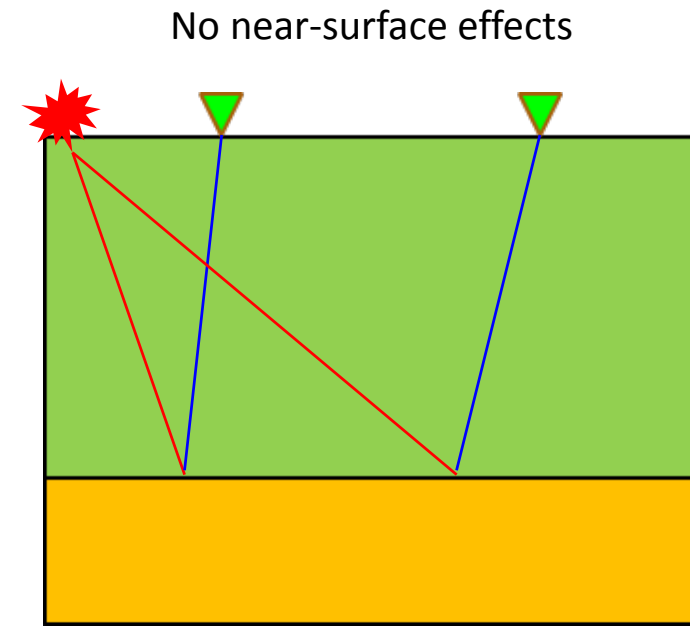
Prepared by

Raul Cova and Kris Innanen

# Near-surface Effects



LVZ: Low Velocity Zone



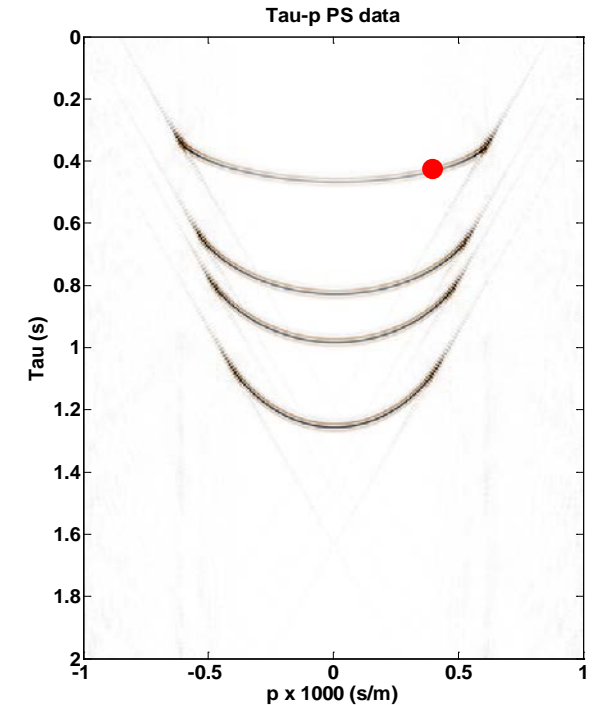
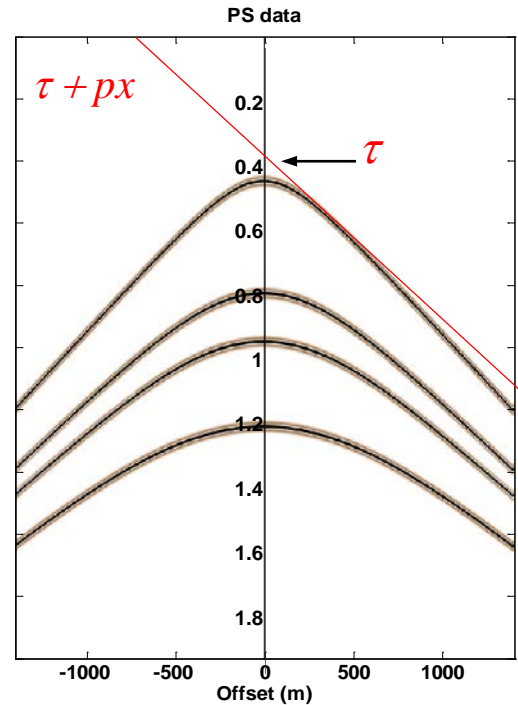
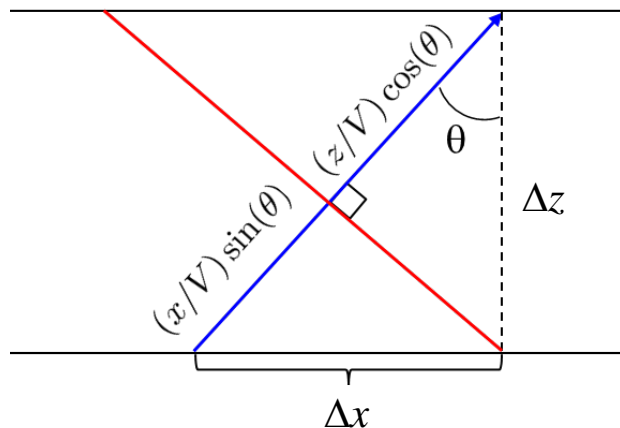
How to compute a near-surface velocity model for S-wave “static” corrections using PS data?

# Tau-p coordinates

Tau-P Transform

$$U(\tau, p) = \int_{-\infty}^{\infty} u(\tau + px, x) dx$$

Sensitive to the emerging angle of the wavefield at the surface



$$\Delta t = \frac{\sin(\theta)}{V} \Delta x + \frac{\cos(\theta)}{V} \Delta z \quad p^2 + q^2 = \frac{1}{V^2}$$

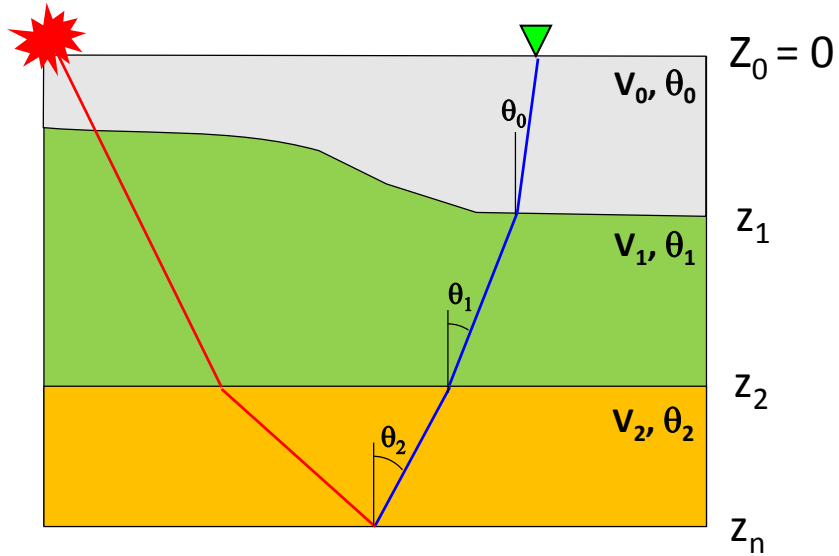
$$= p\Delta x + q\Delta z$$

$$= p\Delta x + \Delta\tau$$

$p$ : Horizontal slowness

$q$ : Vertical slowness

# $\tau$ and the near surface



$\tau$  contribution from  $z_2$  to the surface with velocity  $v_1$  and raypath angle  $\theta_1$

$$\tau^u = \sum_{i=2}^{n-1} \Delta z_i q_i^u + z_2 q_1^u + z_1 (q_0^u - q_1^u)$$

$\tau$  contribution from the reflector to the base of the replacement medium  $z_2$

Near-surface effect

Near-surface correction

Subtract  $\tau$  contribution from  $z_1$  to the surface with velocity  $v_0$  and raypath angle  $\theta_0$

$$\Delta \tau_{NS}^u = z_1 (q_1^u - q_0^u)$$

Add  $\tau$  contribution from  $z_1$  to the surface with velocity  $v_1$  and raypath angle  $\theta_1$

# Dipping near-surface base

In this case  $p$  is not constant at all layers, therefore we introduce,

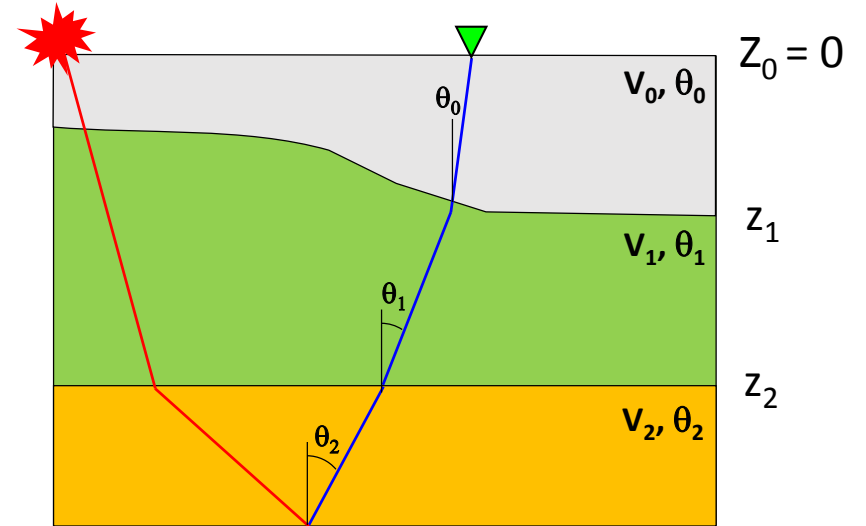
$$\Delta\tau_{NS}^u = z_1 (q_1 - q_0(\phi))$$

where,

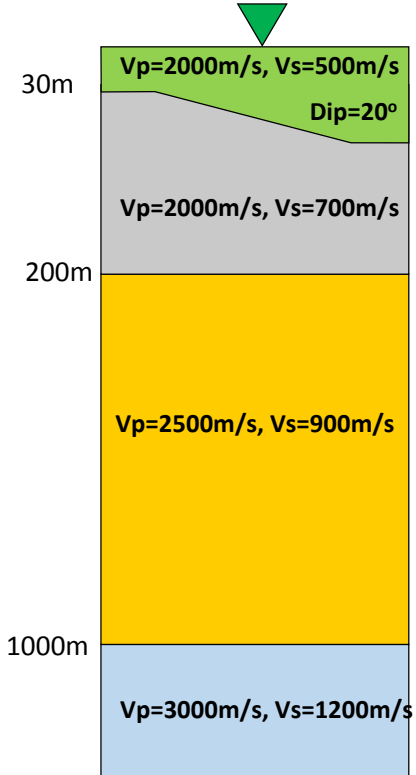
$$q_0(\phi) = q_a \cos(\phi) - p_a(\phi) \sin(\phi)$$

with,

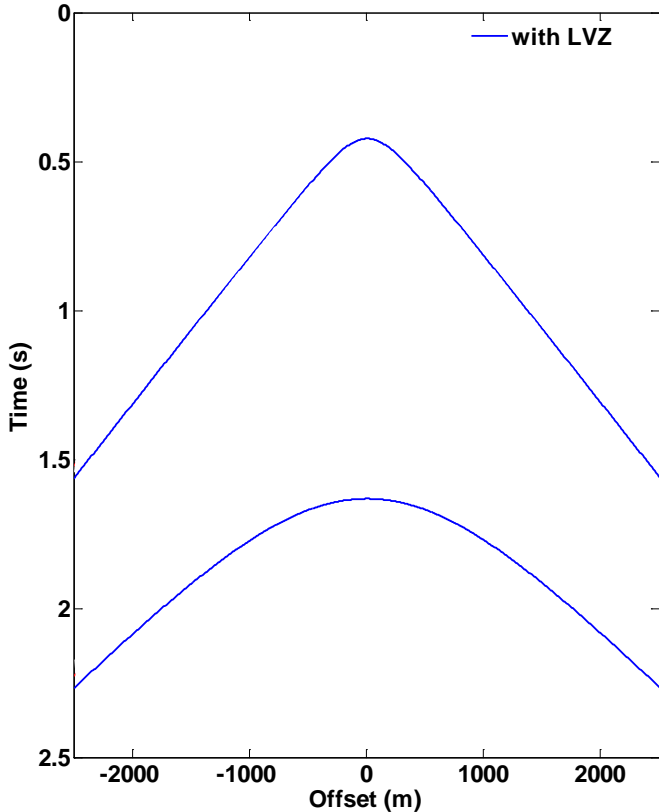
$$q_a = \sqrt{\frac{1}{v_0^2} - p_a^2} \quad \text{and} \quad p_a(\phi) = p \cos(\phi) - q_1 \sin(\phi)$$



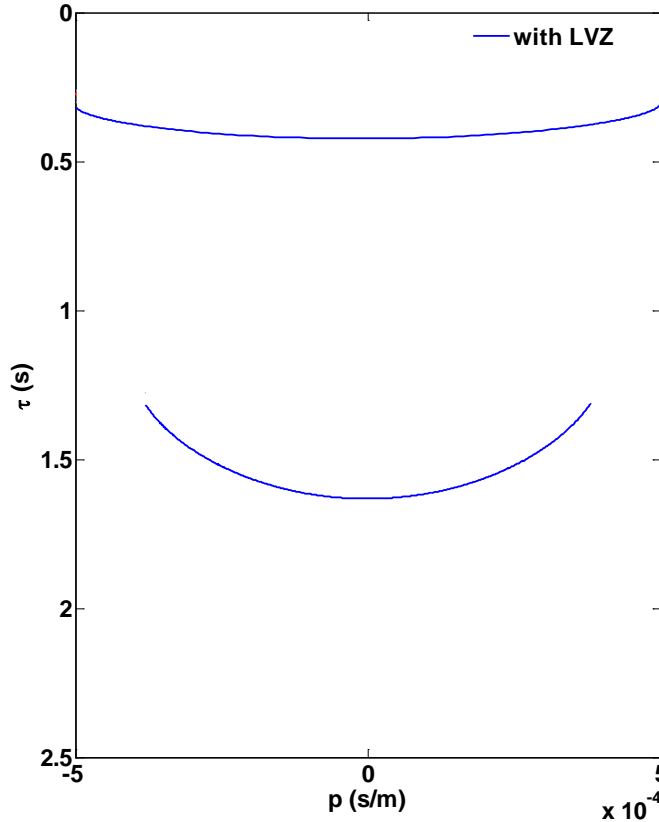
# Travel times analysis



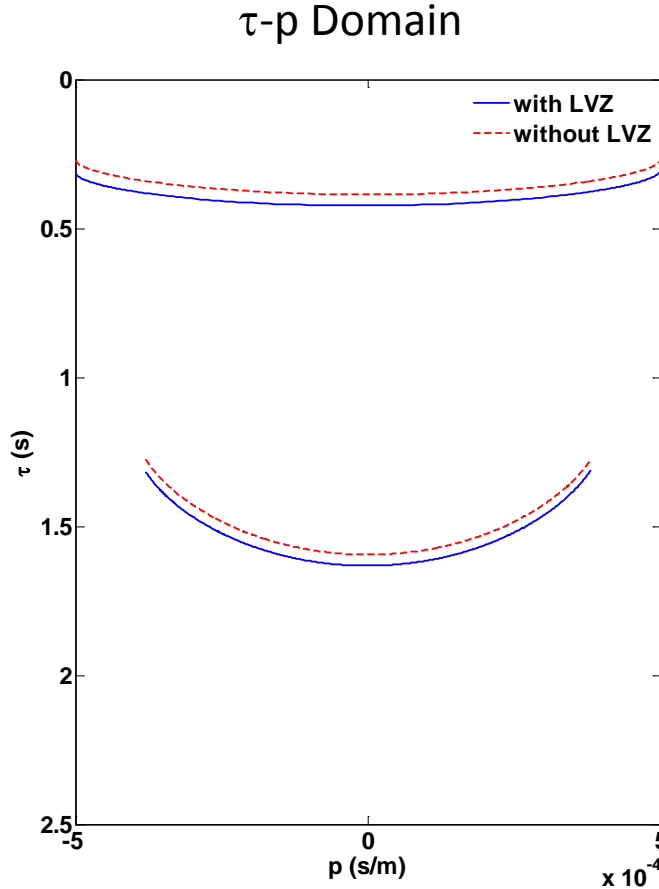
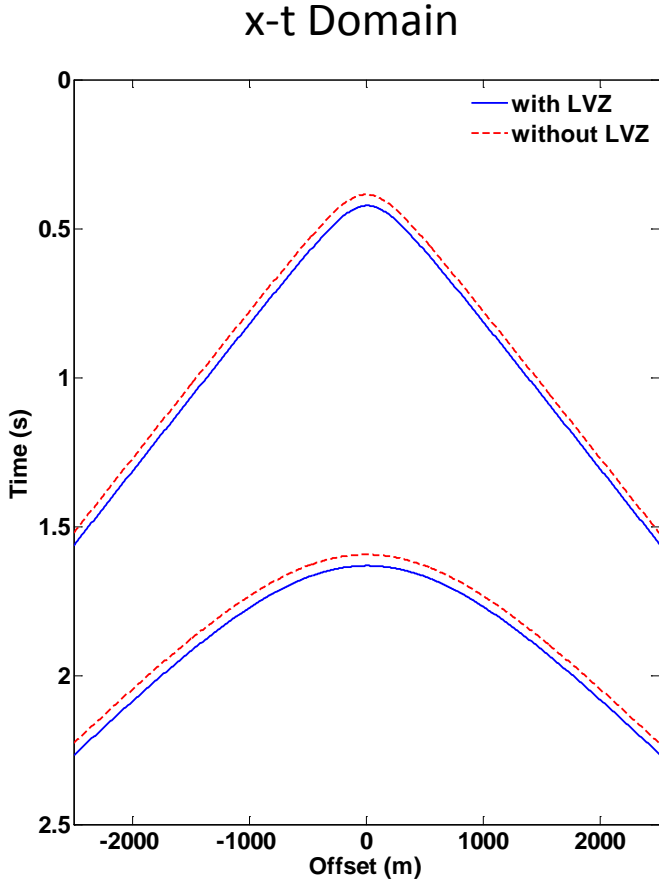
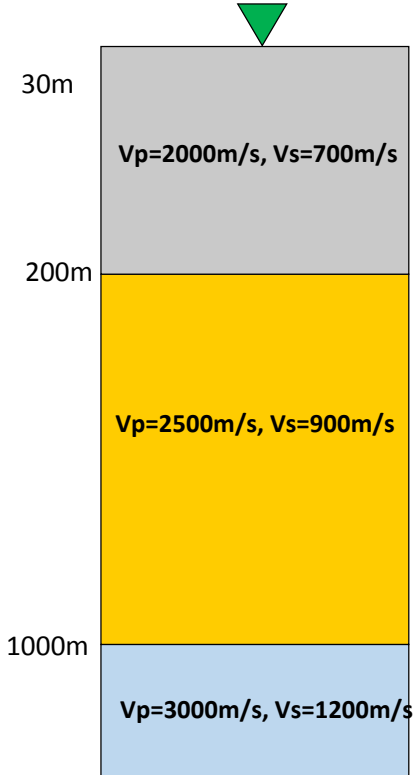
x-t Domain



$\tau$ -p Domain

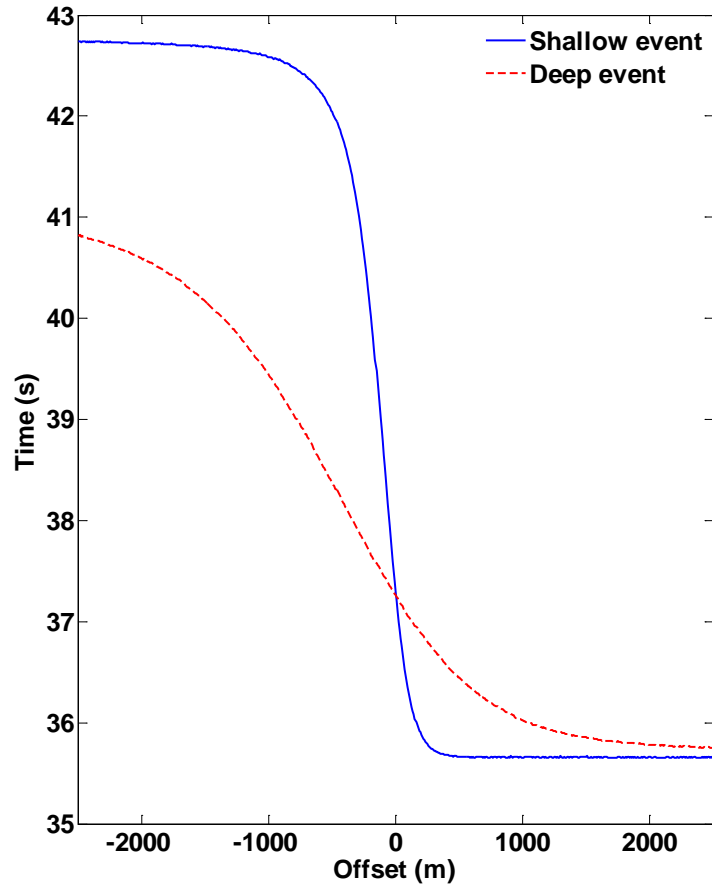


# Travel times analysis



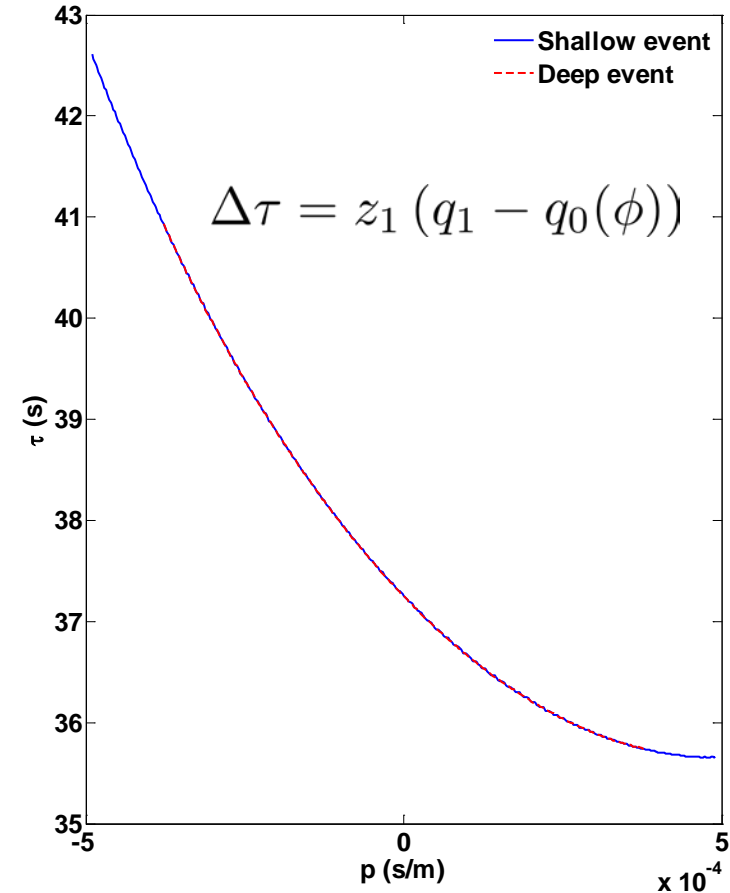
# Dipping LVZ Model

x-t Domain



Non-stationary

$\tau$ -p Domain



Stationary



# Quasi-Newton Inversion

Objective Function

$$\Phi(m) = \|\delta d\|^2 = \|g(m) - d_{\text{obs}}\|^2,$$

Diagram labels: Data Residuals (points to  $\delta d$ ), Forward modelled data (points to  $g(m)$ ), Observed data (points to  $d_{\text{obs}}$ )

$$m_i = m_{i-1} + \delta m_i, \leftarrow \text{Model Update}$$

$$\delta m = [J(m)^\dagger J(m) + \mu I]^{-1} J(m)^\dagger \delta d.$$

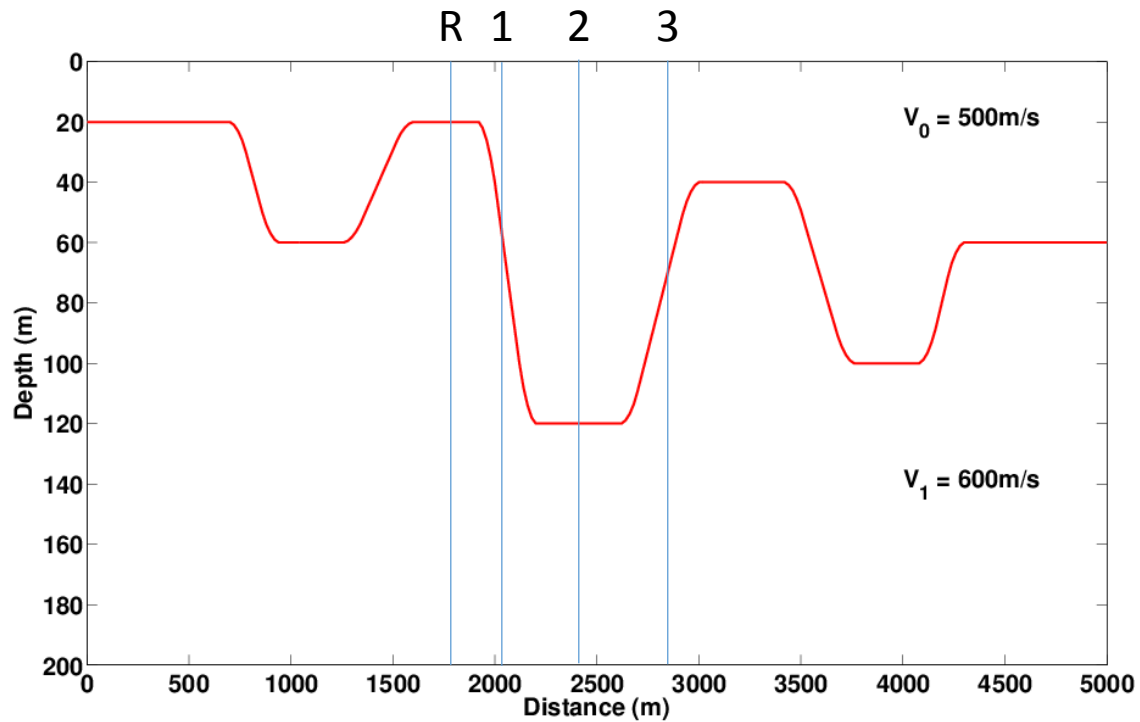
$I$  : Identity matrix  
 $\mu$  : Regularization weight

Jacobian

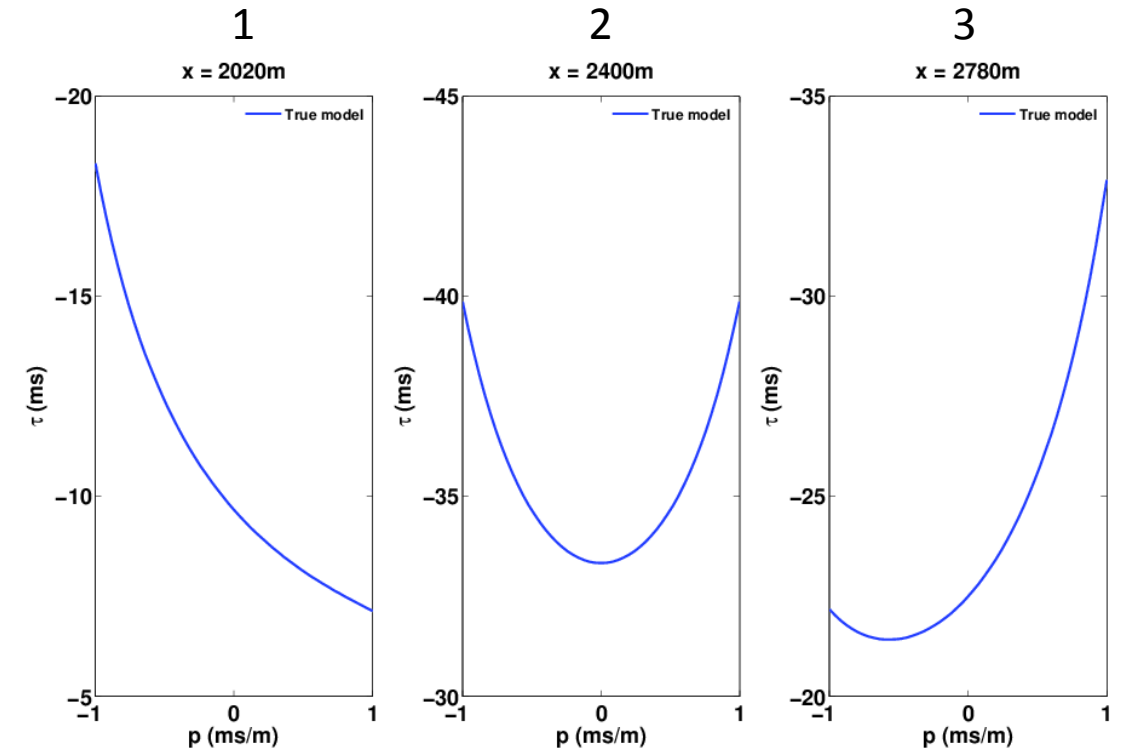
$$J(m) = \left[ \frac{\partial g(m)}{\partial m} \right] = \left[ \frac{\partial g(m)}{\partial V_0}, \frac{\partial g(m)}{\partial V_1}, \frac{\partial g(m)}{\partial z}, \frac{\partial g(m)}{\partial \phi} \right].$$

# Modelled differences at three locations

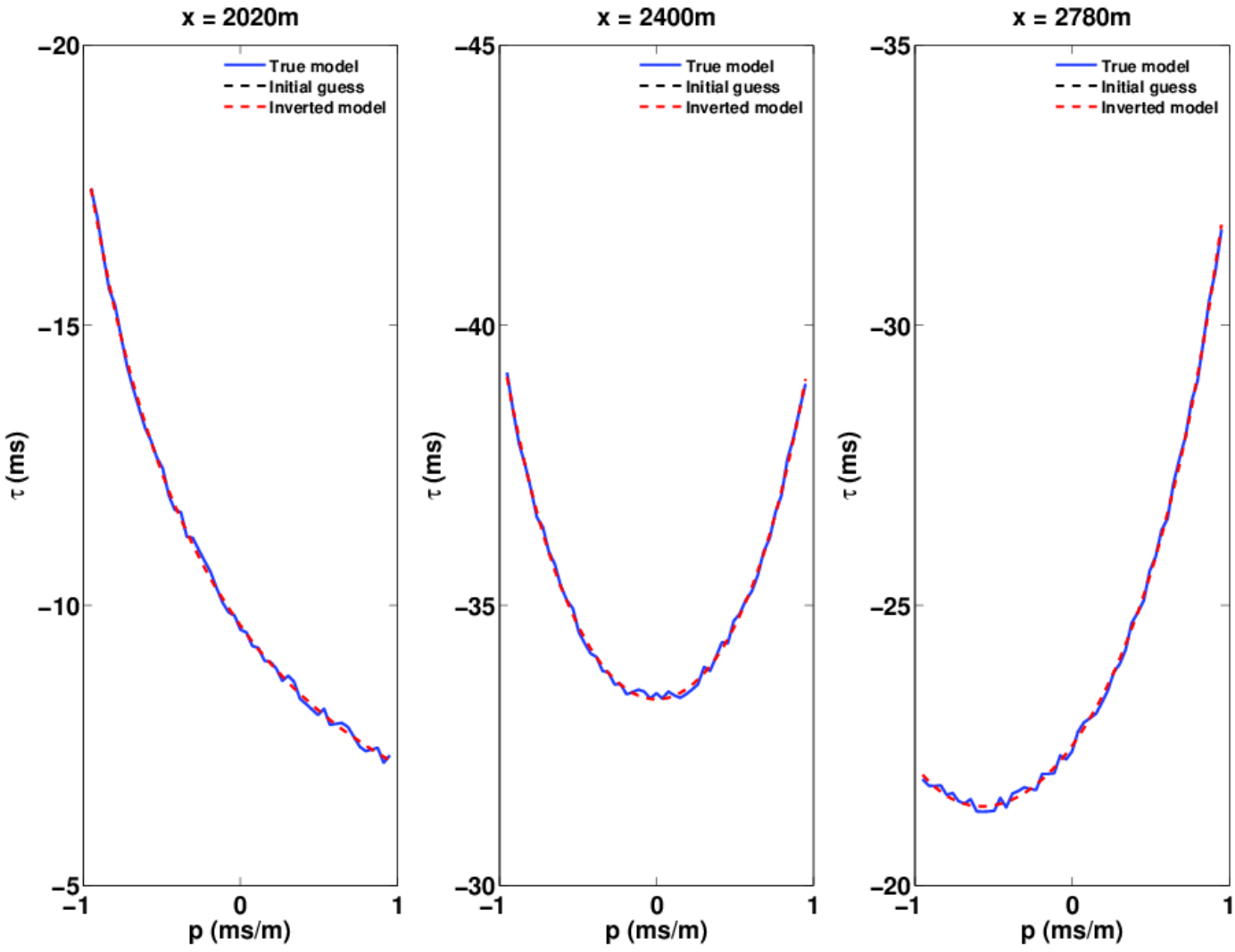
## Velocity Model



## Data plots

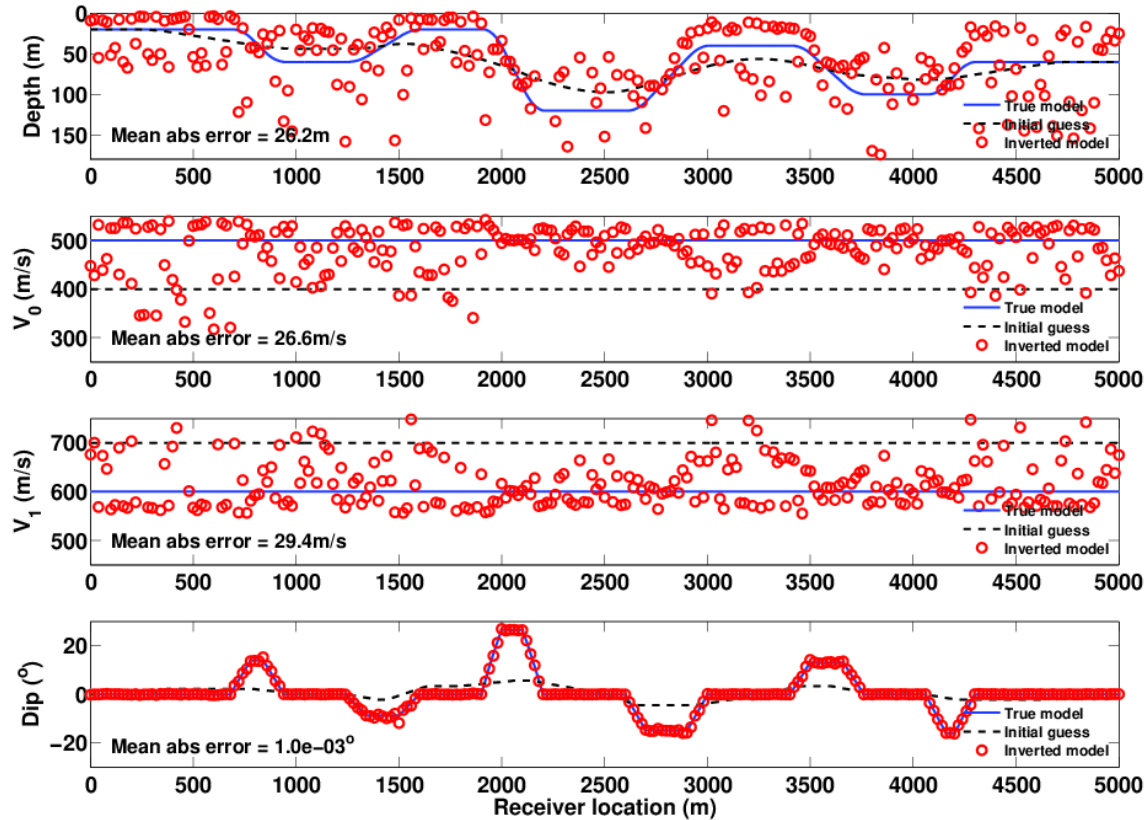


# Adding random noise [-0.1, 0.1]ms

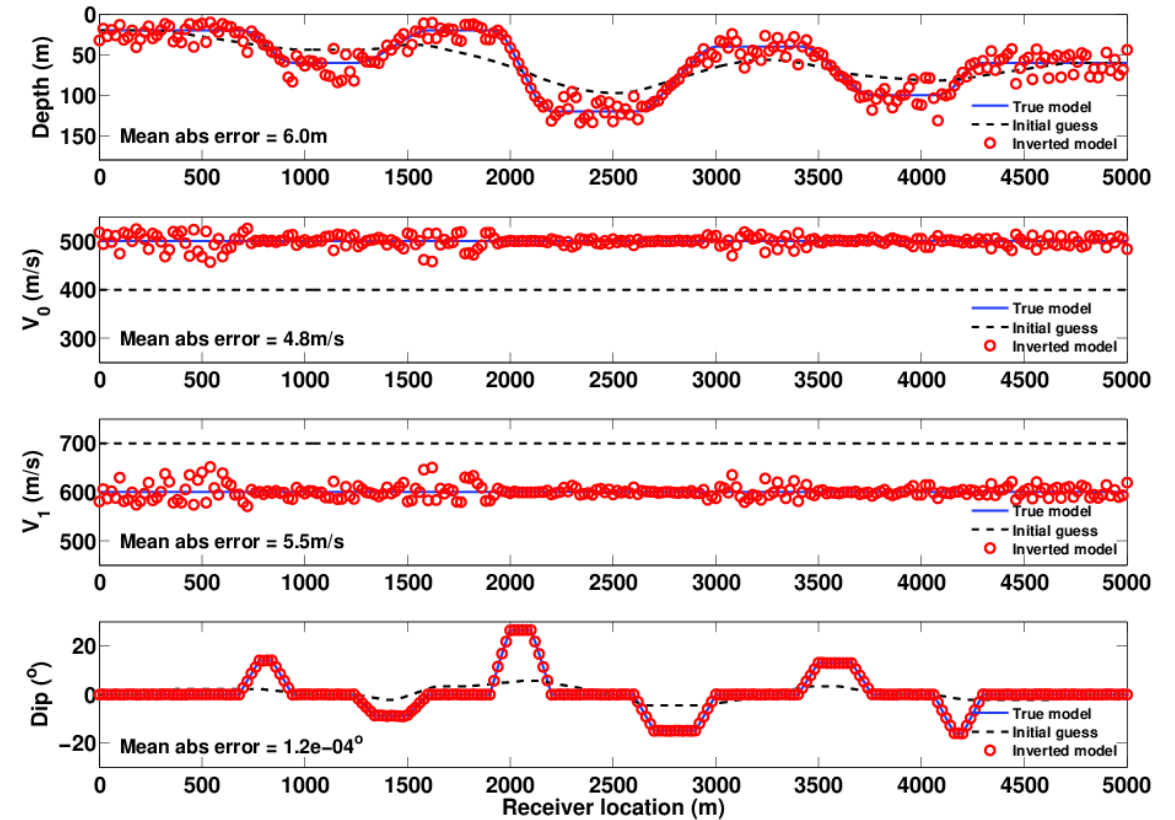


# Random noise effect

Random noise [-0.1, 0.1] ms



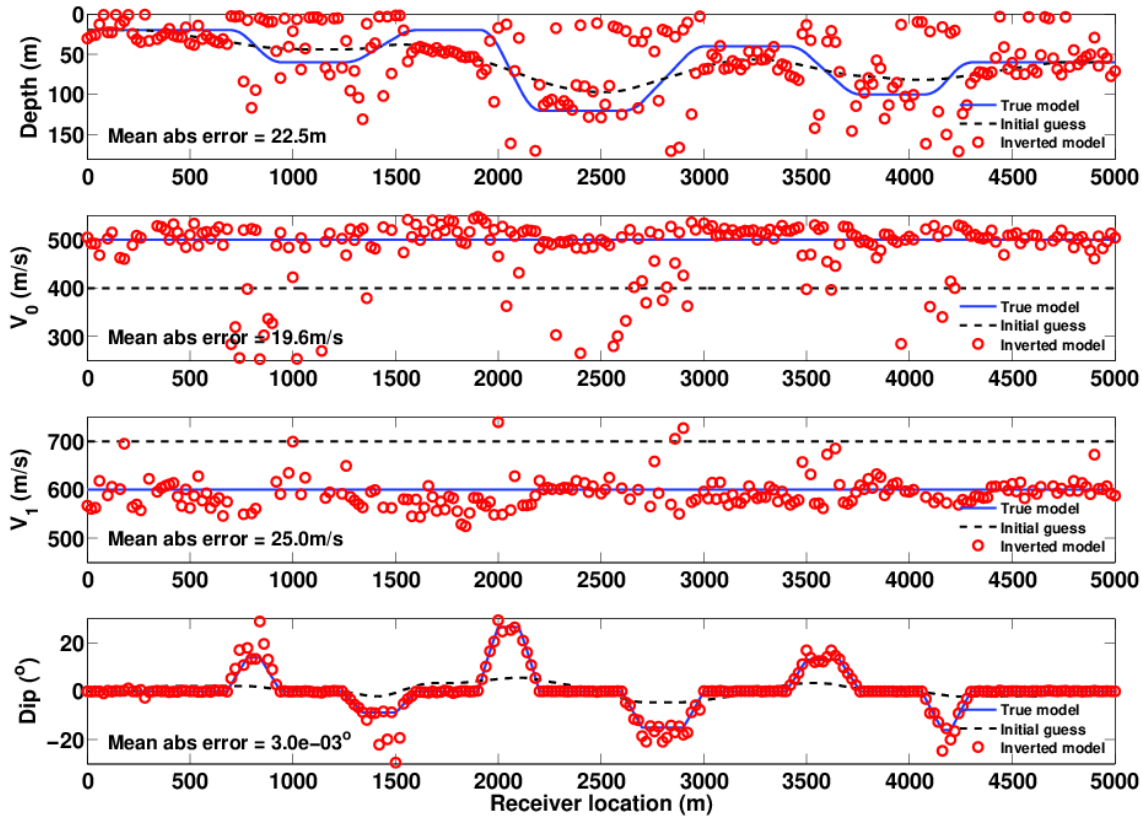
Random noise [-0.01, 0.01] ms



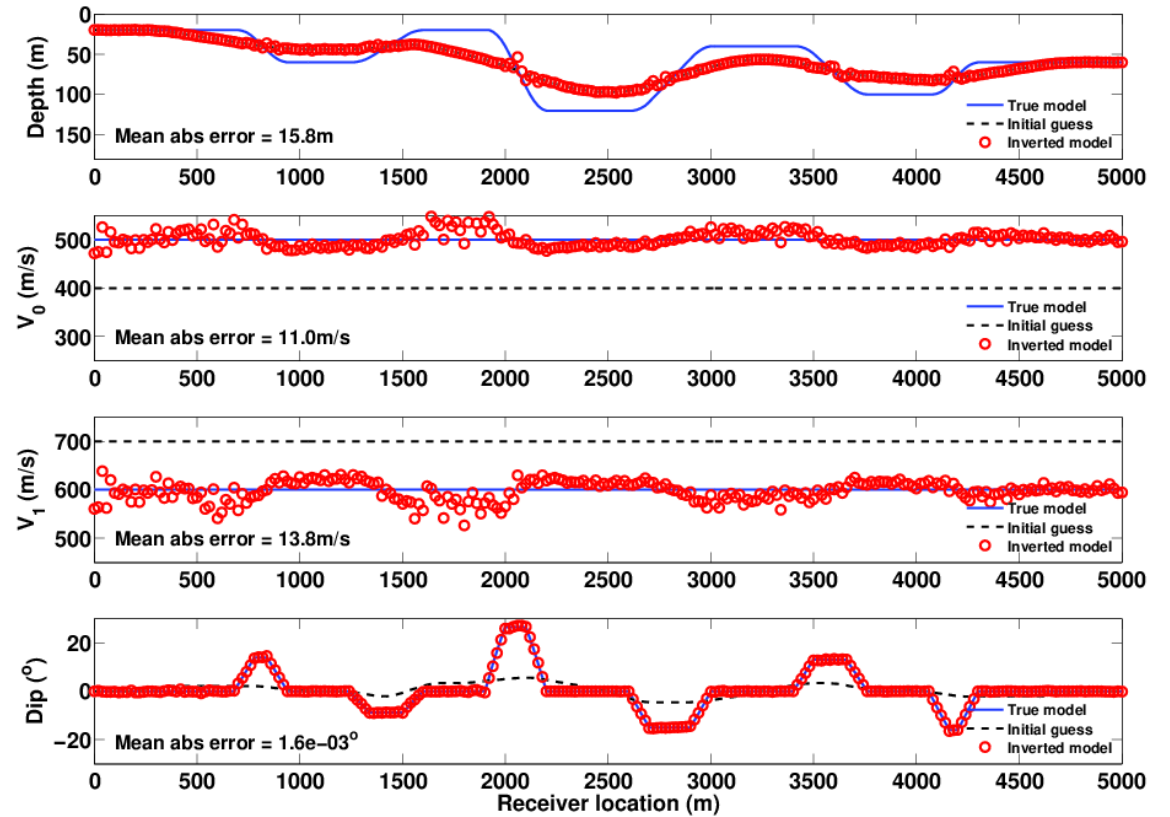
- Inversion seems to be very sensitive to noise in traveltimes differences.
- Dip estimations display stable results

# Random Noise [-0.1 0.1]ms, $\rho=[-0.5, 0.5]$ ms/m

$u=0.01$



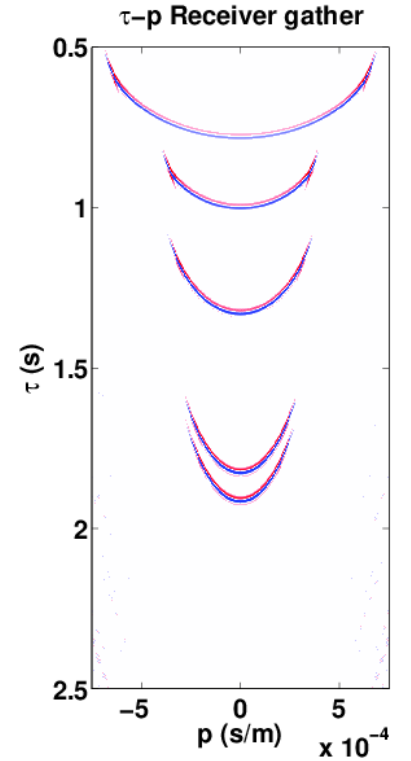
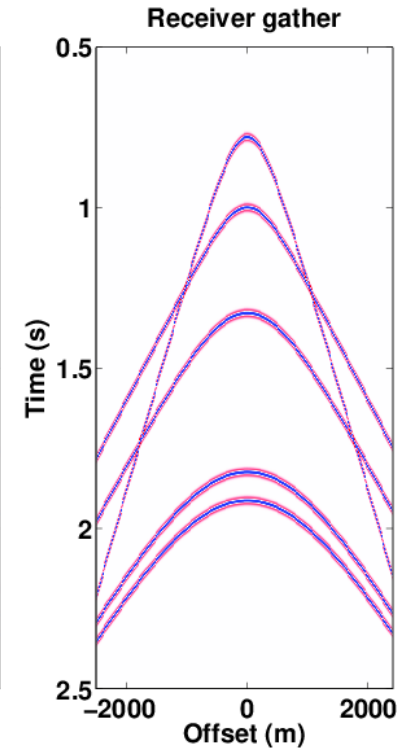
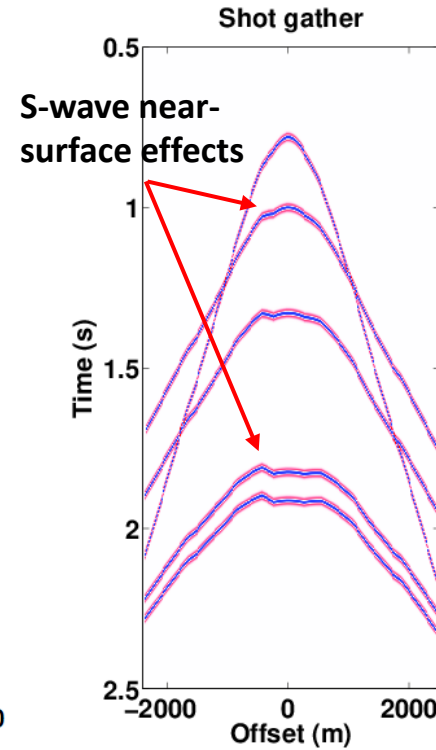
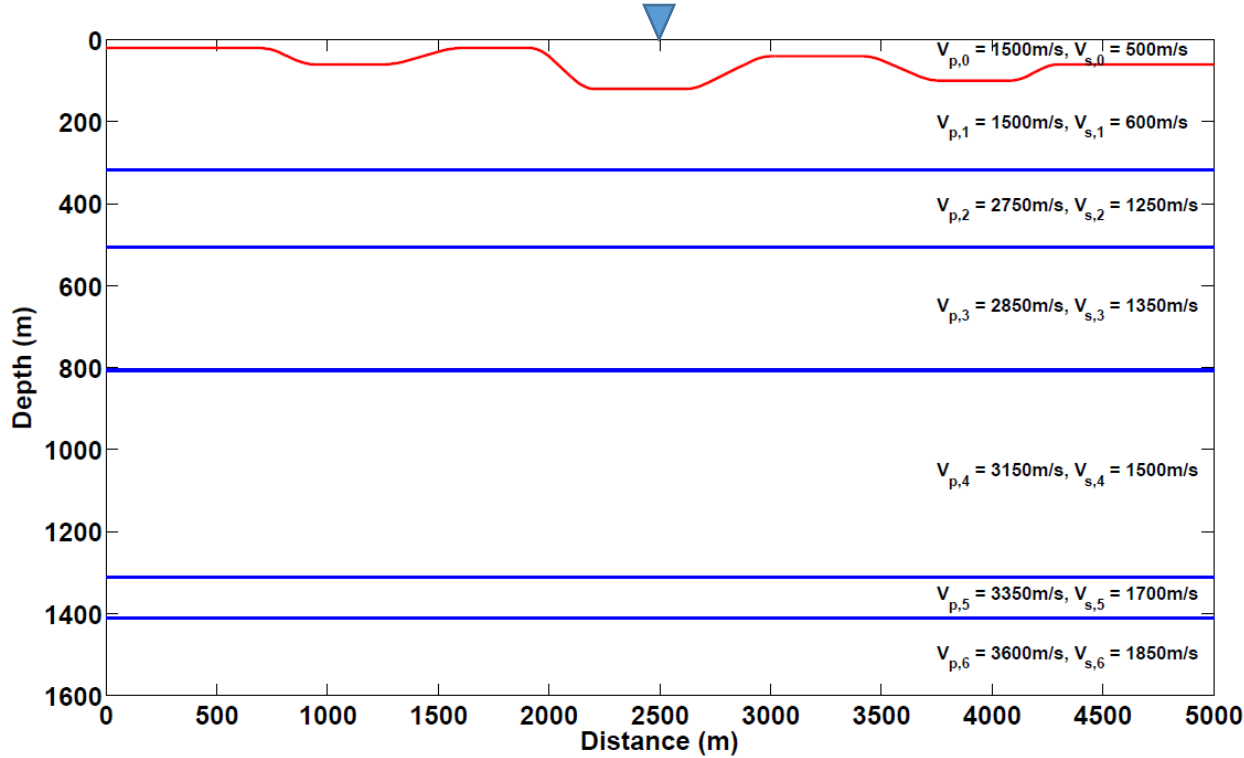
$u=0.1$



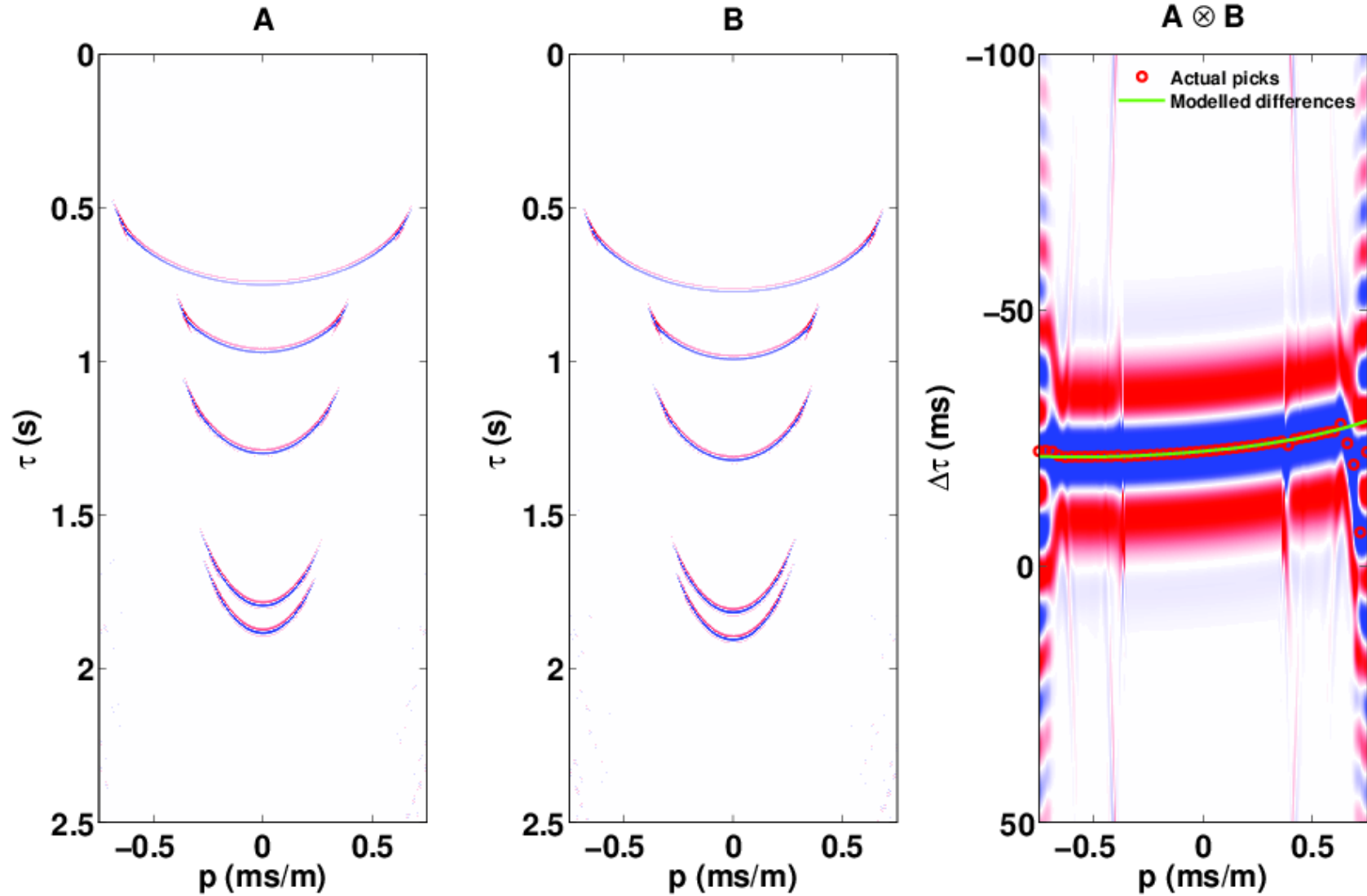
- Increasing regularization weight from 0.01 to 0.1, stabilizes the inversion.
- Depth estimation is now largely constrained by the initial depth model
- Dip inversion is very well behaved

# Raytrace Modelling

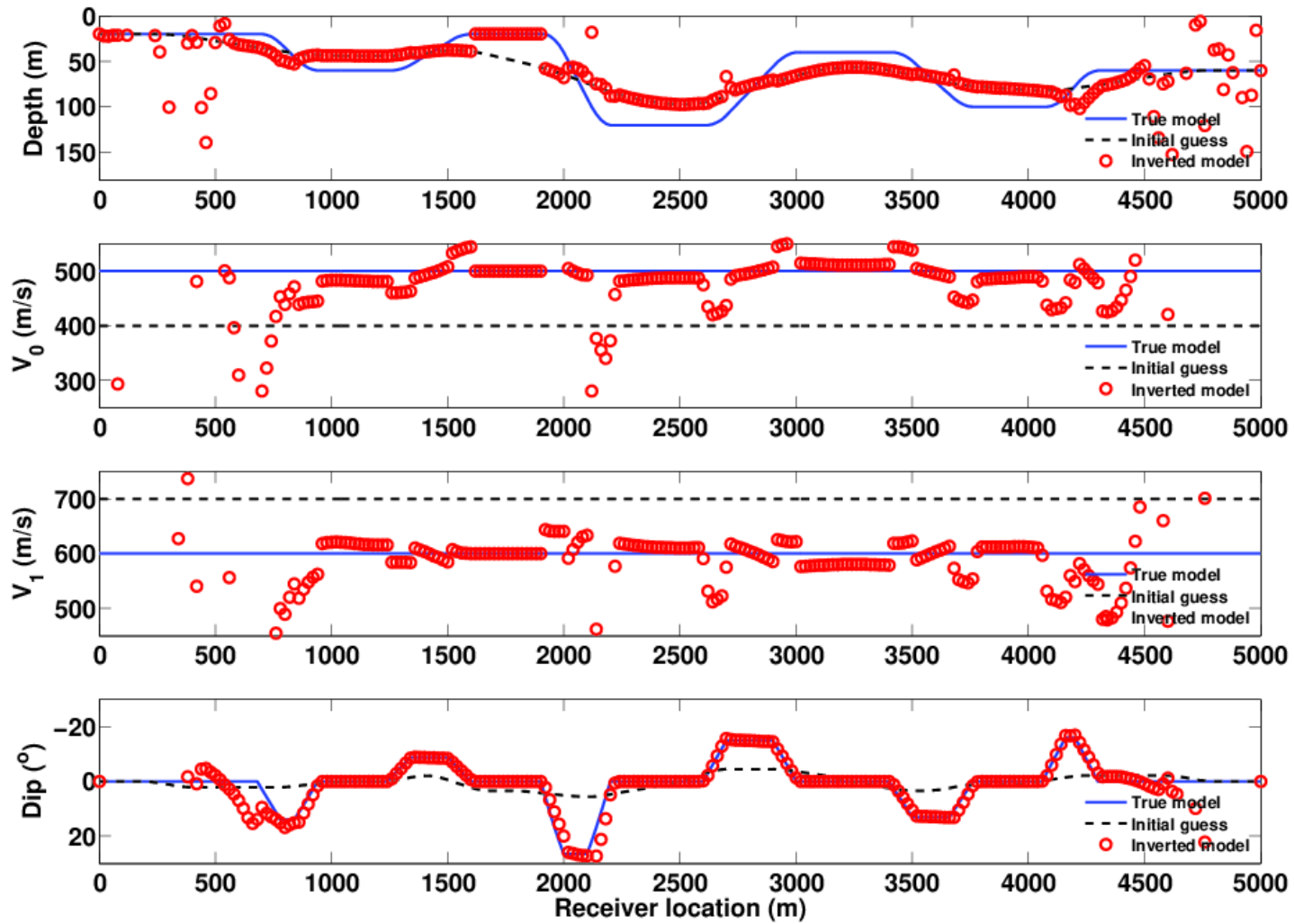
Velocity model



# Data Crosscorrelation



# Ray Trace inversion $u=0.1$





# Conclusions

- Since our approach requires the near-surface parameters to be known at one reference location, any result provided by the inversion will depend of the accuracy of this information.
- The presence of noise in the picks or the lack of a wide range of p-values had an important effect on the stability of the inversion of the depth values. The inverted velocities were also affected by these limitations although to a lesser degree.
- The inverted dips displayed by far the most stable results. Since this parameter controls the shape of the data, it is less sensitive to errors in the individual picks.
- Different parameterizations and inversion methods should be explored to improve the results for this study. Application of this method on real datasets remains to be explored.

# Acknowledgements

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- CREWES faculty, staff and students.

**Thanks!!!**