

8. Traveltime inversion

M. Ravasi

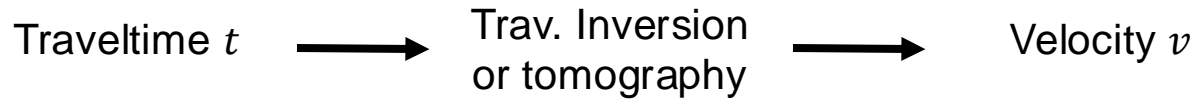
ERSE 210 Seismology

Raytracing vs. travelttime inversion

Forward problem:



Inverse problem:

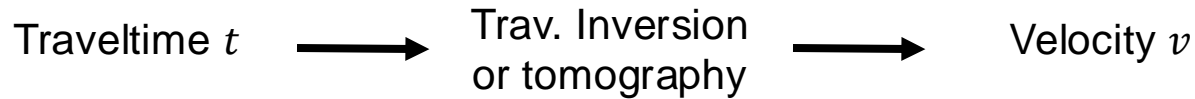


Raytracing vs. travelttime inversion

Forward problem:



Inverse problem:



Traveltime inversion

2 steps approach:

- Find a simple (possibly 1D parametric, e.g. gradient) velocity field to be used as **background model**

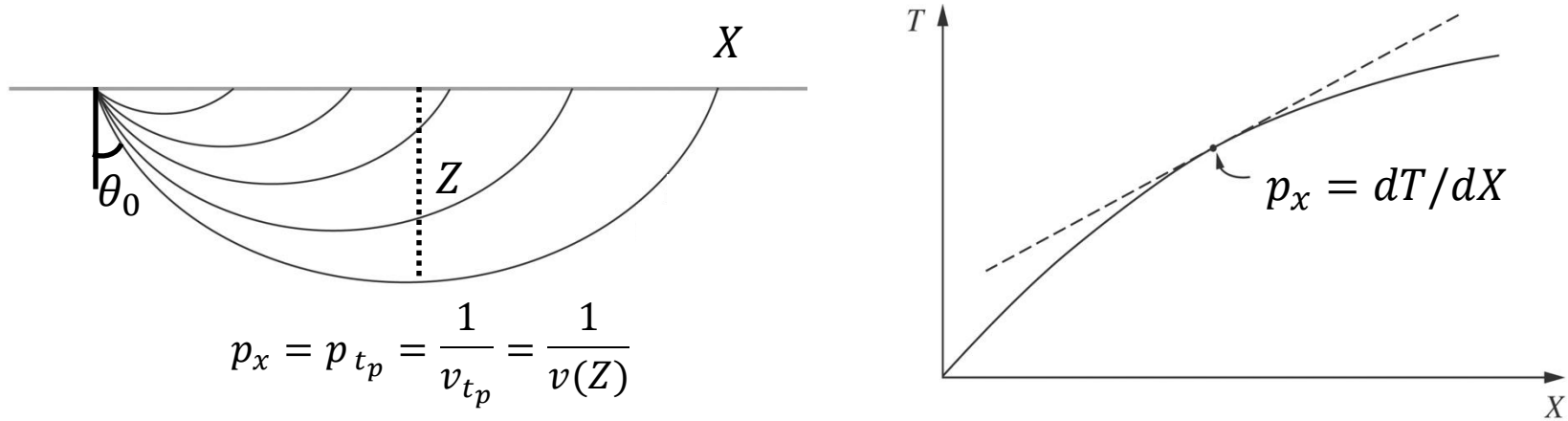
$$t \rightarrow v_0: t = f(v_0) \quad e.g. \quad v_0(x, z) = a(x)z + b(x)$$

- Refine the background velocity by inverting the **residual traveltimes**

$$\Delta t = t^{obs} - t_0 \rightarrow \Delta v \quad v = v_0 + \Delta v$$

1D Velocity inversion

Let's start from the traveltimes curve of a turning wave



→ Given a traveltime curve, if we estimate the slope at a given offset ($p_x = dT/dX$), we know the velocity at the turning point ($v(Z)$). If we start from the top, we can bootstrap a 1D velocity model.

1D Velocity inversion

We first compute velocities for a set of offsets where receivers are located (X_0, X_1, X_2, X_n):

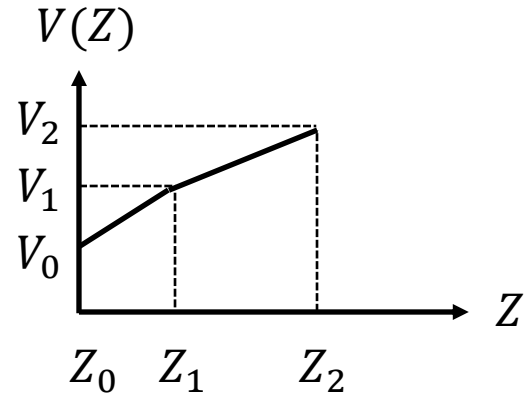
$$v_0 = v(Z_0 = 0) = \frac{1}{dT/dX|_{X_0=0}}$$

$$v_1 = v(Z_1) = \frac{1}{dT/dX|_{X_1}}$$

...

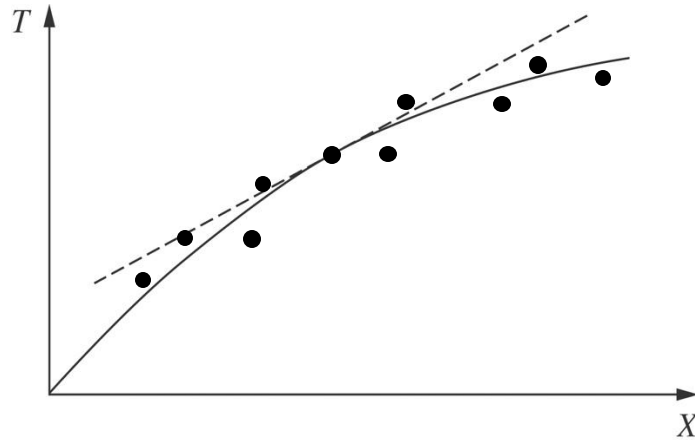
Then given a measure of $T(X_1)$, we find Z_1 such that the profile $V = \{v(Z_0), v(Z_1)\}$ matches the observed time. Repeat for $T(X_2)$ and so on.

$$\text{Using: } X(p_x) = 2p_x \int_0^{Z_1} \frac{dz}{\sqrt{p^2(z) - p_x^2}} \quad T(p_x) = 2 \int_0^{Z_1} \frac{p^2(z) dz}{\sqrt{p^2(z) - p_x^2}}$$



1D Velocity inversion

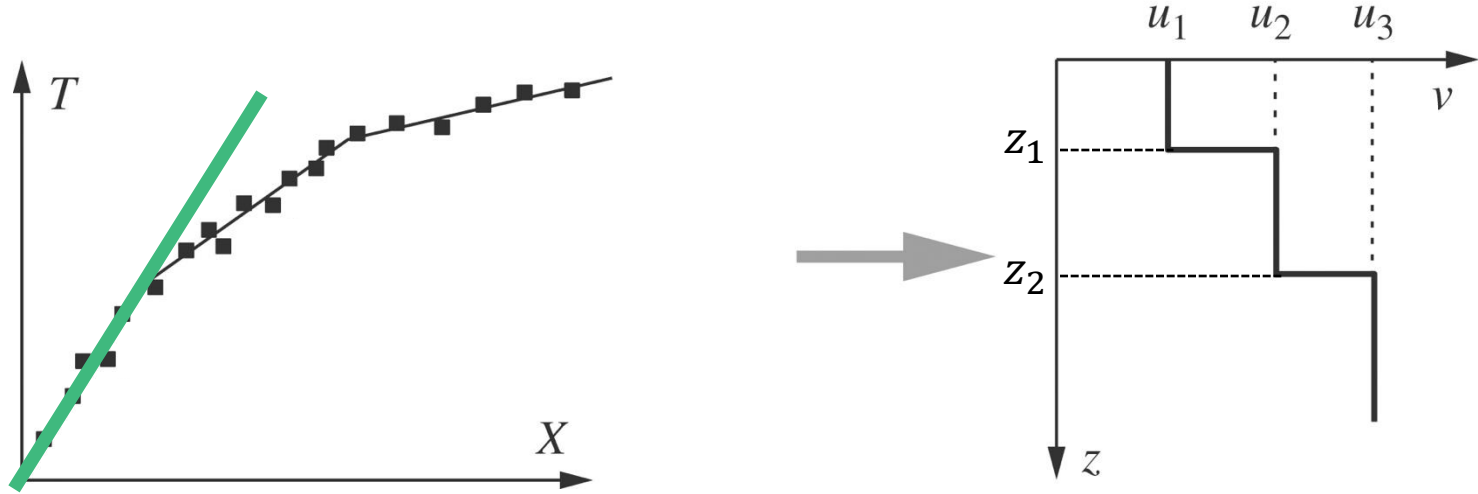
In practice we do not have exact traveltimes



- Noisy measurements

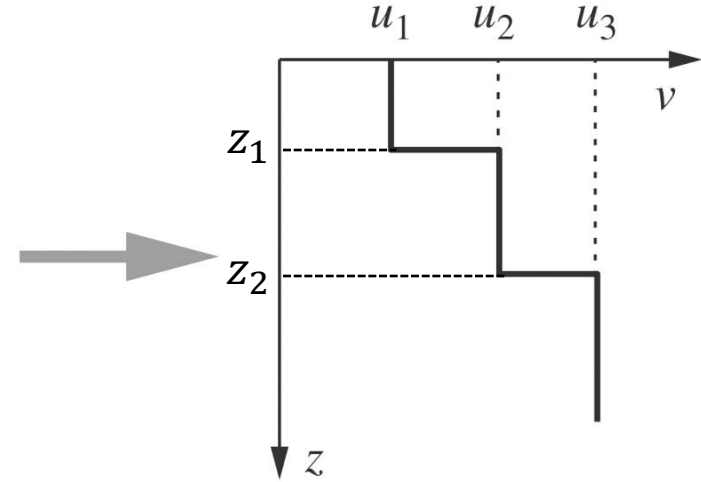
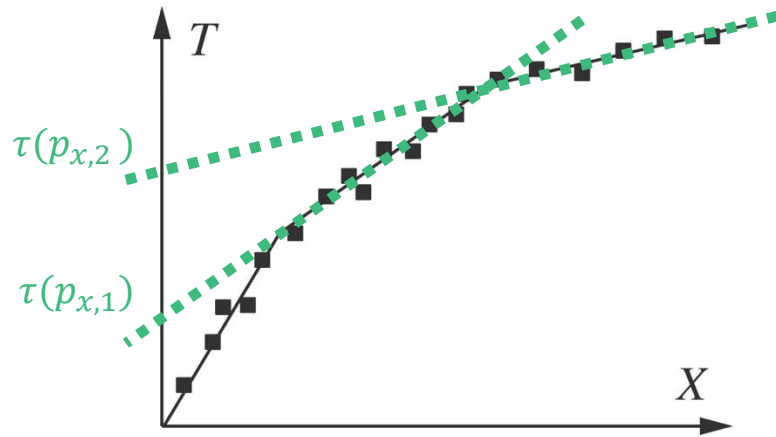
The previous approach can be somehow unstable!

1D Velocity inversion – straight line approach



For the first layer: $t = X/V_0 \rightarrow$ Find slope of the straight line

1D Velocity inversion – straight line approach



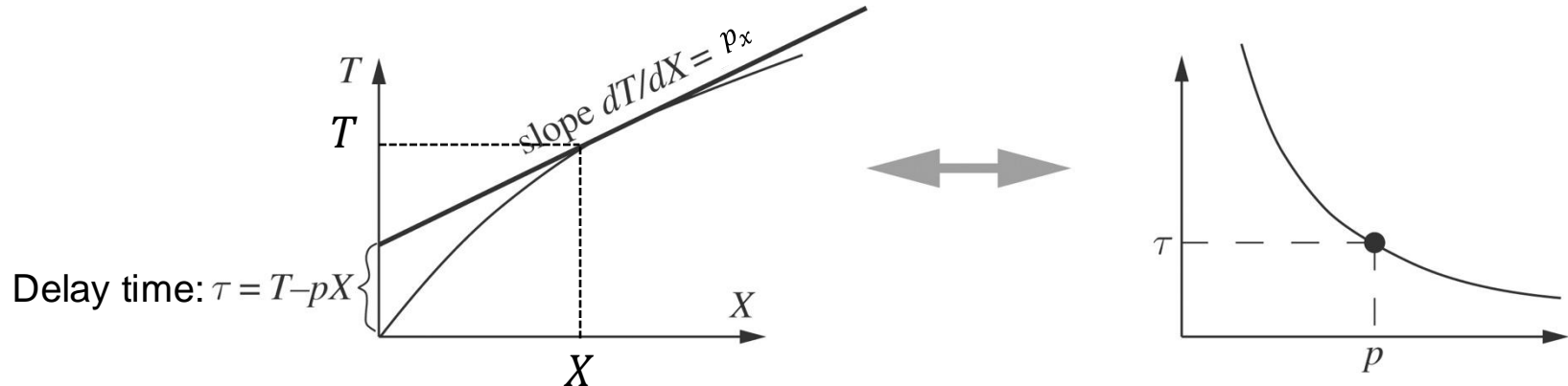
For all other layers, the slope is proportional to $1/V_i$

Tau-P function

The curve $T(X)$ is not always well behaved (i.e., caustics). However, $X(p)$ is well behaved in that there is only one value of X for any value of p .

Let's define the $\tau(p_x)$ curve:

$$\tau(p_x) = T(p_x) - p_x X(p_x)$$



Tau-P function

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$$\tau(p_x) = T(p_x) - p_x X(p_x)$$

This can be derived from the Taylor expansion of $t(x)$ at $x = X$:

$$t = T + p(x - X) + \dots$$

$$\text{at } x = 0: t(x = 0) = \tau = T - pX$$

Tau-P function

Since we know the expressions for $X(p_x)$ and $T(p_x)$, we can write:

$$\begin{aligned}\tau(p_x) &= 2 \int_0^{Z_p} \left[\frac{p^2(z)}{\sqrt{p^2(z) - p_x^2}} - \frac{p_x^2}{\sqrt{p^2(z) - p_x^2}} \right] dz \\ &= 2 \int_0^{Z_p} \sqrt{p^2(z) - p_x^2} dz\end{aligned}$$

Which in a 'discretized' Earth:

$$\tau(p_{x,i}) = 2 \sum_j \sqrt{p_j^2 - p_{x,i}^2} \Delta z_j, \quad p_j > p_{x,i}$$

1D Inversion with Tau-P function

This leads to a simple bootstrapping method:

$$v_0 = \frac{1}{p_{x,0}}$$

$$\Delta z_0 = \frac{\tau_1}{2\sqrt{p_0^2 - p_{x,1}^2}}$$

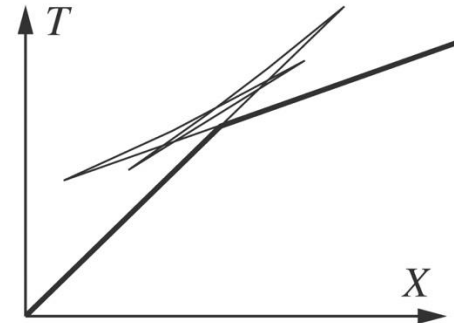
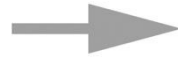
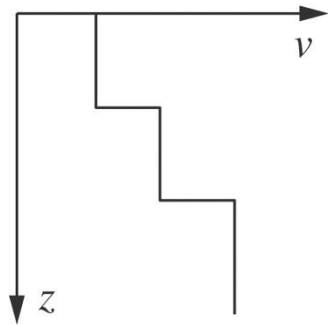
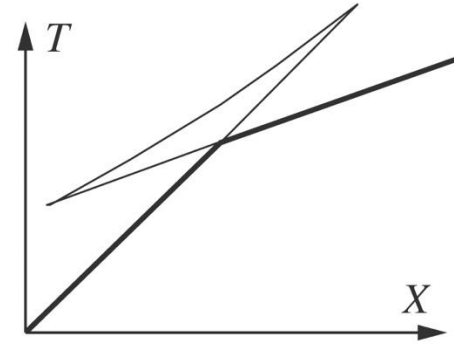
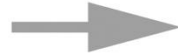
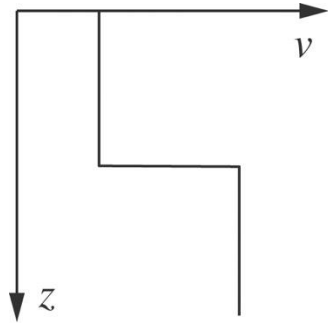
$$v_1 = \frac{1}{p_{x,1}}$$

$$\Delta z_1 = \frac{\tau_2 - 2\sqrt{p_0^2 - p_{x,2}^2}\Delta z_0}{2\sqrt{p_1^2 - p_{x,2}^2}}$$

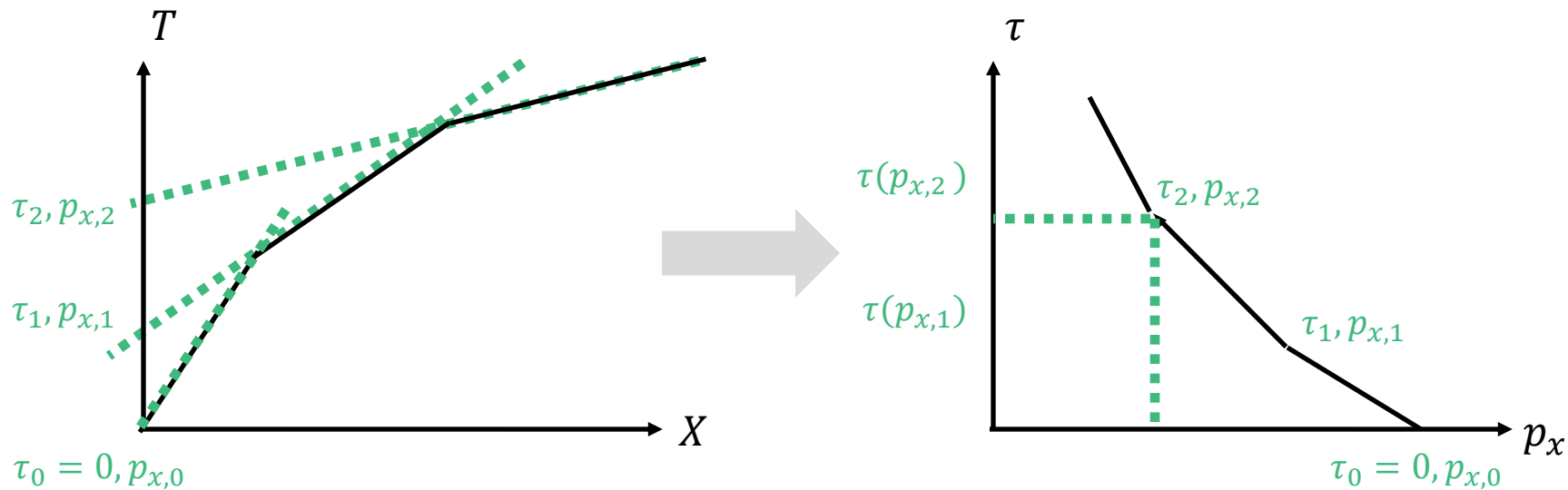
$$v_i = \frac{1}{p_{x,i}}$$

$$\Delta z_i = \frac{\tau_{i+1} - 2\sum_{j=1}^i \sqrt{p_j^2 - p_{x,i+1}^2}\Delta z_0}{2\sqrt{p_i^2 - p_{x,i+1}^2}}$$

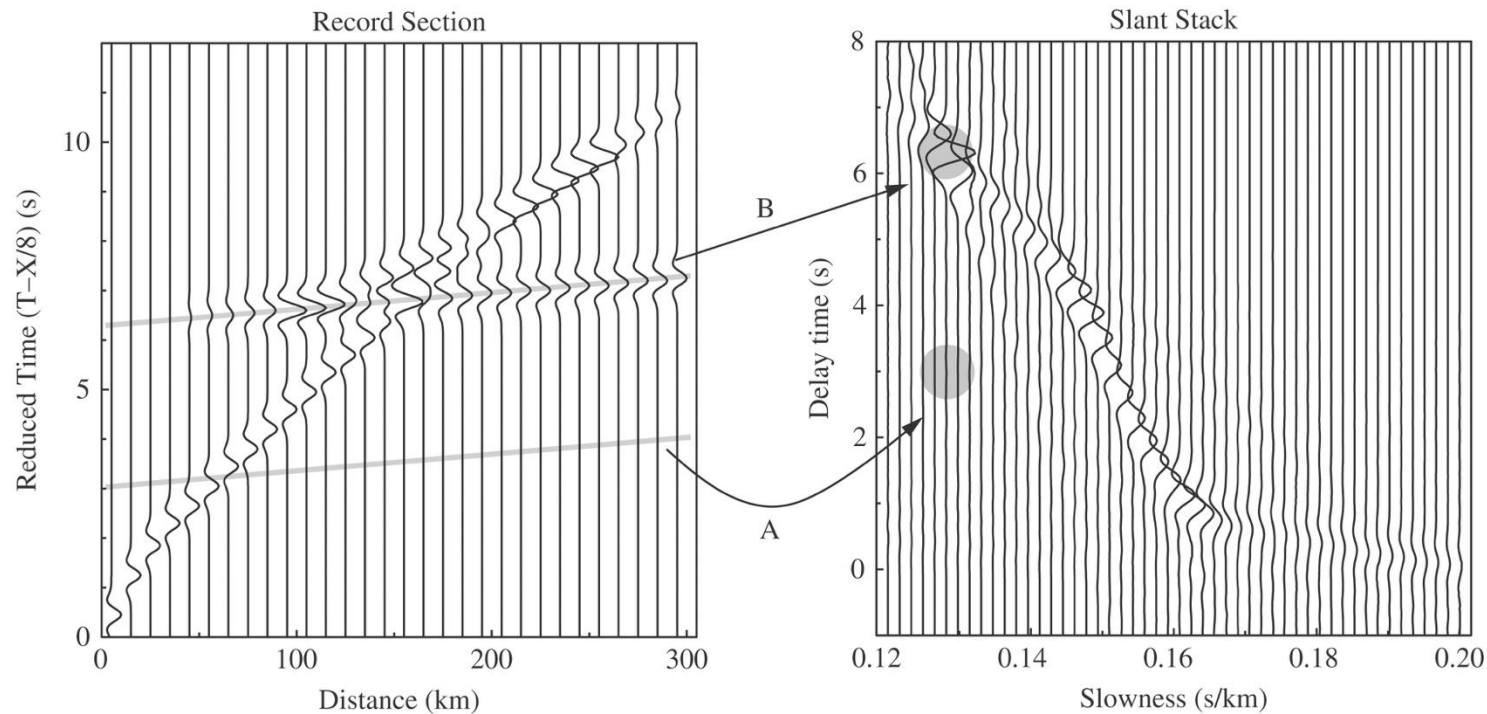
1D Non-uniqueness



Slant stack / Radon transform



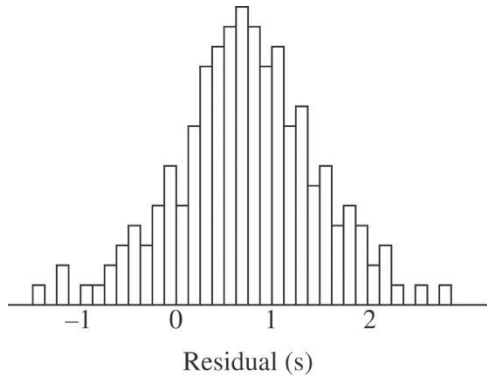
Slant stack / Radon transform



From 1D to 3D inversion

Once a simple 1D velocity model $v(z)$ is available, one can compute

$$\Delta t = t^{obs} - t(v(z))$$



$\Delta t < 0$ slower than correct prediction

$\Delta t > 0$ faster than correct prediction

$\sigma(\Delta t)$ is due to noise in data and should not be overfitted

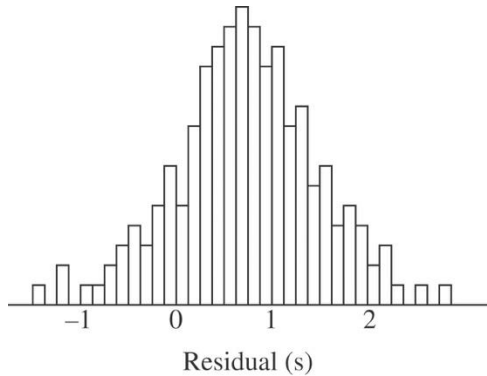
$E[\Delta t] \neq 0 \rightarrow$ systematic error, may require lateral changes to explain it

Δt as input for a residual 3D inversion (aka **seismic tomography**).

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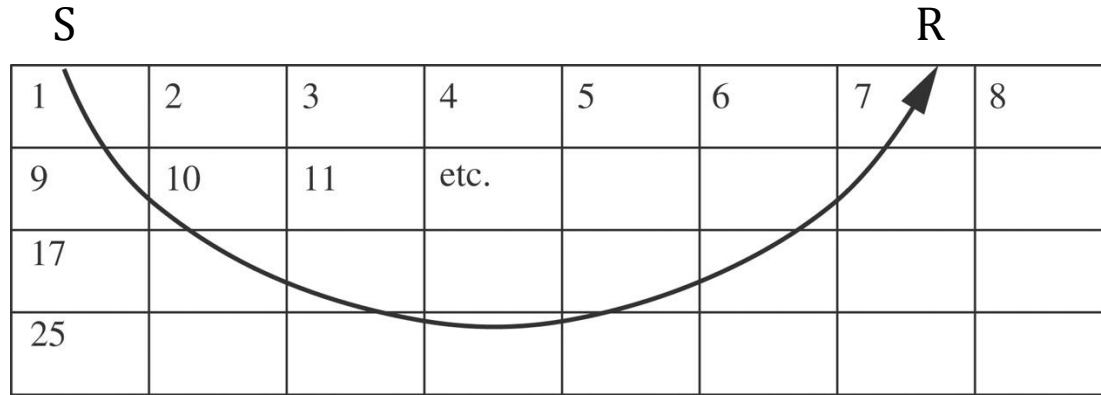
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Seismic tomography

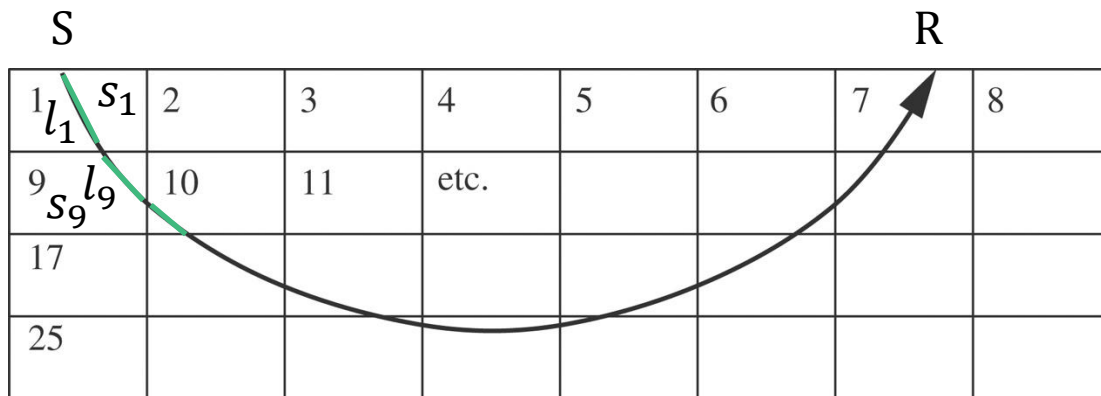
For each source-receiver pair:



$$t_{SR} = \int_{l_{SR}} s \, dl$$

Seismic tomography

For each source-receiver pair:



$$t_{SR} = \int_{l_{SR}} s \, dl \rightarrow t_{SR} = \sum_i s_i l_i$$

Seismic tomography

In practice, the first problem is to find a ray that goes from S to R:

- **Ray shooting:** shoot a fan of rays from the source with different take-off angles and choose the one that reaches the surface closer to R
- **Ray bending:** S and R are fixed and the ray is moved to fit Fermat's principle

Seismic tomography

The second problem is to discretize the rays (as shown in previous slides) and set up a linear system of eqs:

$$\begin{bmatrix} t_{S_1 R_1} \\ t_{S_1 R_2} \\ \dots \\ t_{S_1 R_{N_R}} \\ t_{S_2 R_1} \\ \dots \\ t_{S_{N_S} R_{N_R}} \end{bmatrix} = \begin{bmatrix} l_{1-S_1 R_1} & l_{2-S_1 R_1} & \dots & \dots & l_{N_x N_z - S_1 R_1} \\ l_{1-S_1 R_2} & l_{2-S_1 R_2} & \dots & \dots & l_{N_x N_z - S_1 R_2} \\ \dots & \dots & \dots & \dots & \dots \\ l_{1-S_{N_S} R_{N_R}} & l_{2-S_{N_S} R_{N_R}} & \dots & \dots & l_{N_x N_z - S_{N_S} R_{N_R}} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \dots \\ \dots \\ \dots \\ \dots \\ s_{N_x N_z} \end{bmatrix}$$

$$\downarrow$$

$$\mathbf{t} = \mathbf{L}\mathbf{s}$$

After removing the
traveltime of the
background model

$$\downarrow$$

$$\Delta \mathbf{t} = \mathbf{L} \Delta \mathbf{s}$$

Note: \mathbf{L} is usually a very sparse matrix

Seismic tomography

The tomographic problem is usually over-determined ($N_S N_R > N_x N_z$)

$$\widehat{\Delta \mathbf{s}} = (\mathbf{L}^T \mathbf{L})^{-1} \mathbf{L}^T \Delta \mathbf{t}$$

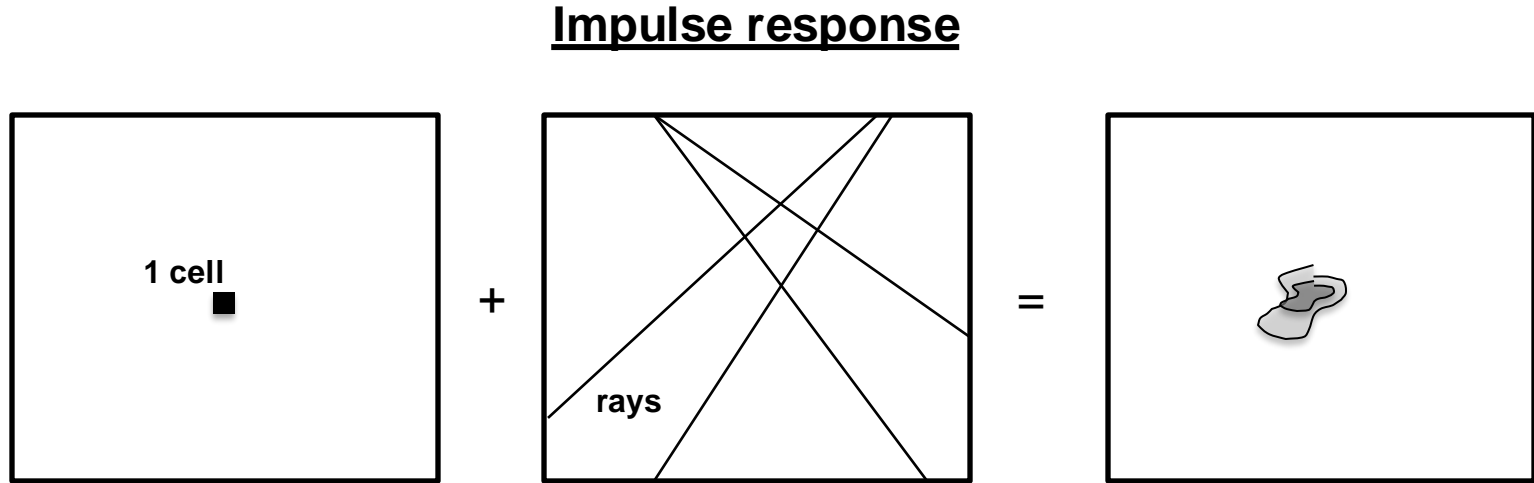
However, $\mathbf{L}^T \mathbf{L}$ is hard to invert since:

- multiple rays can pass through a single cell and give contrasting info (for noisy data)
- some cells may be undersampled (or not sampled at all)

Solution: add prior information (aka regularization)

Seismic tomography

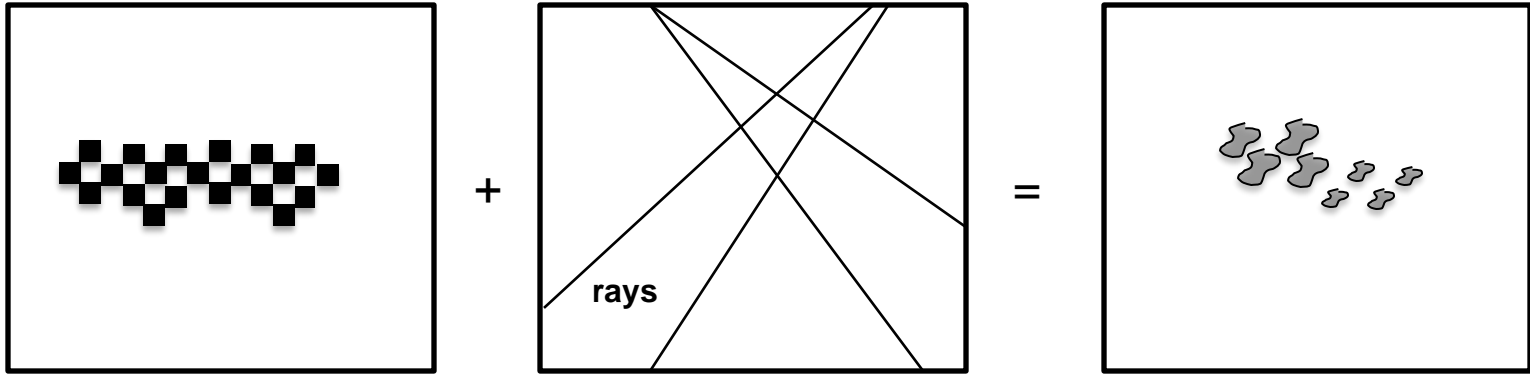
An initial assessment of the 'expected' quality of the inversion (purely based on the acquisition geometry and background velocity model):



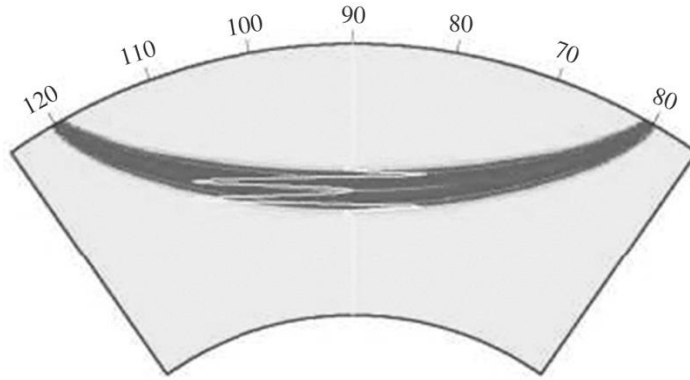
Seismic tomography

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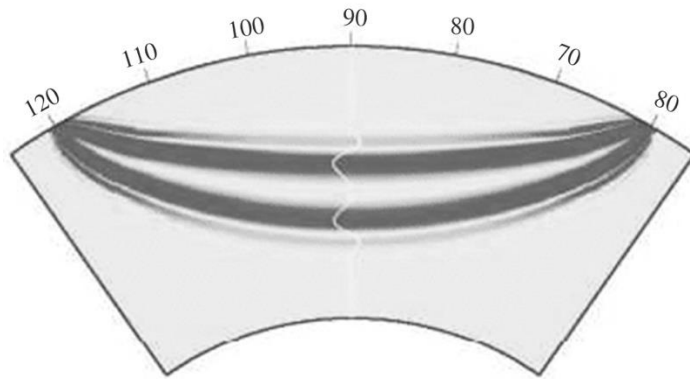
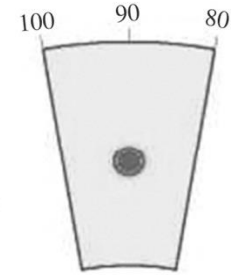
Checkerboard test



Finite-frequency tomography



period = 2 s



period = 20 s

