

ACOUSTICAL METHODS FOR THE MONITORING OF THE MARINE ENVIRONMENT

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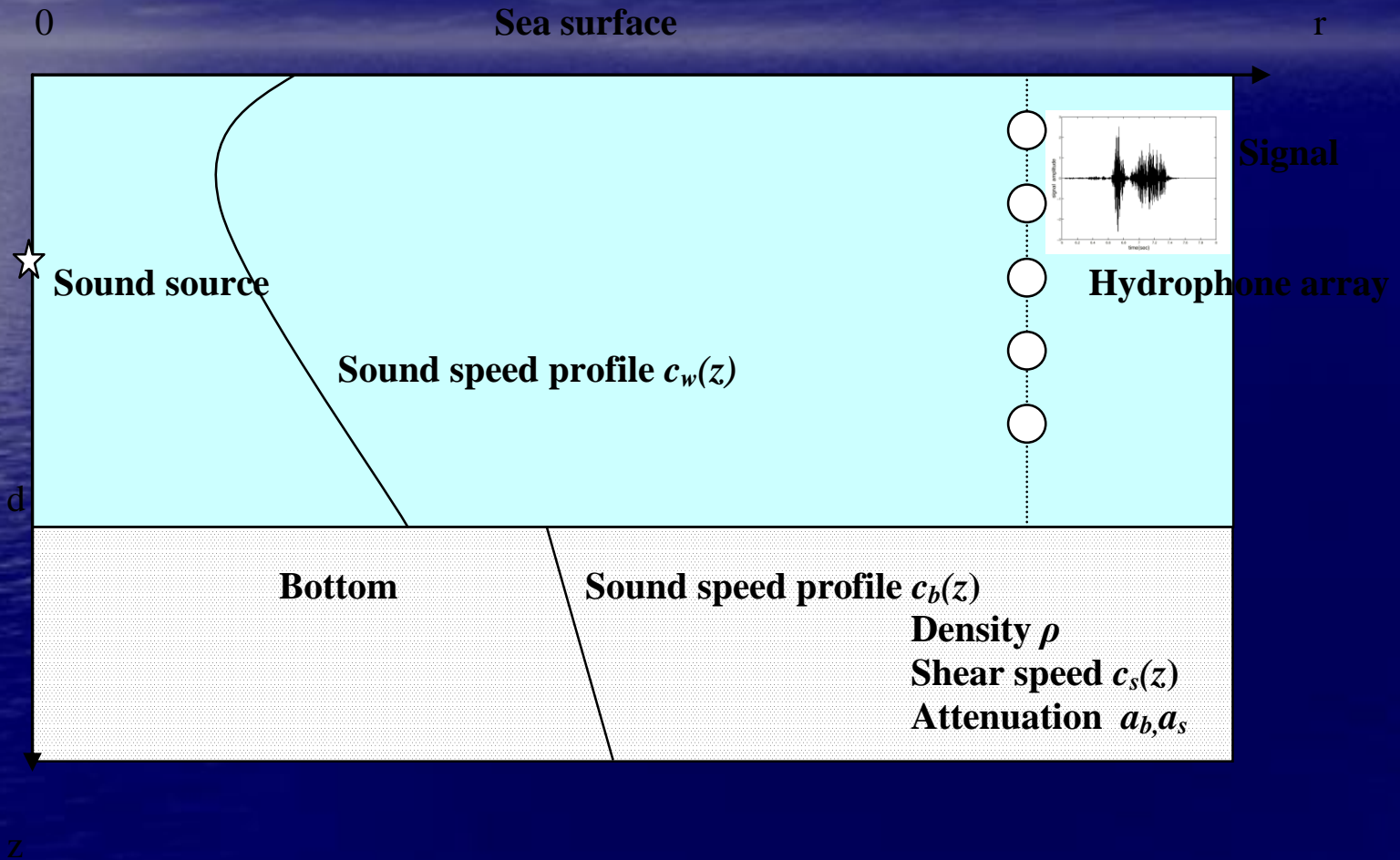
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The geometry at a vertical slice



By representation theorem the acoustic pressure can be expanded in terms of the eigenfunctions of the Depth Problem

$$p(r, z; \omega) = \sum_{n=1}^N A_n(r; \omega) u_n(z; \omega)$$

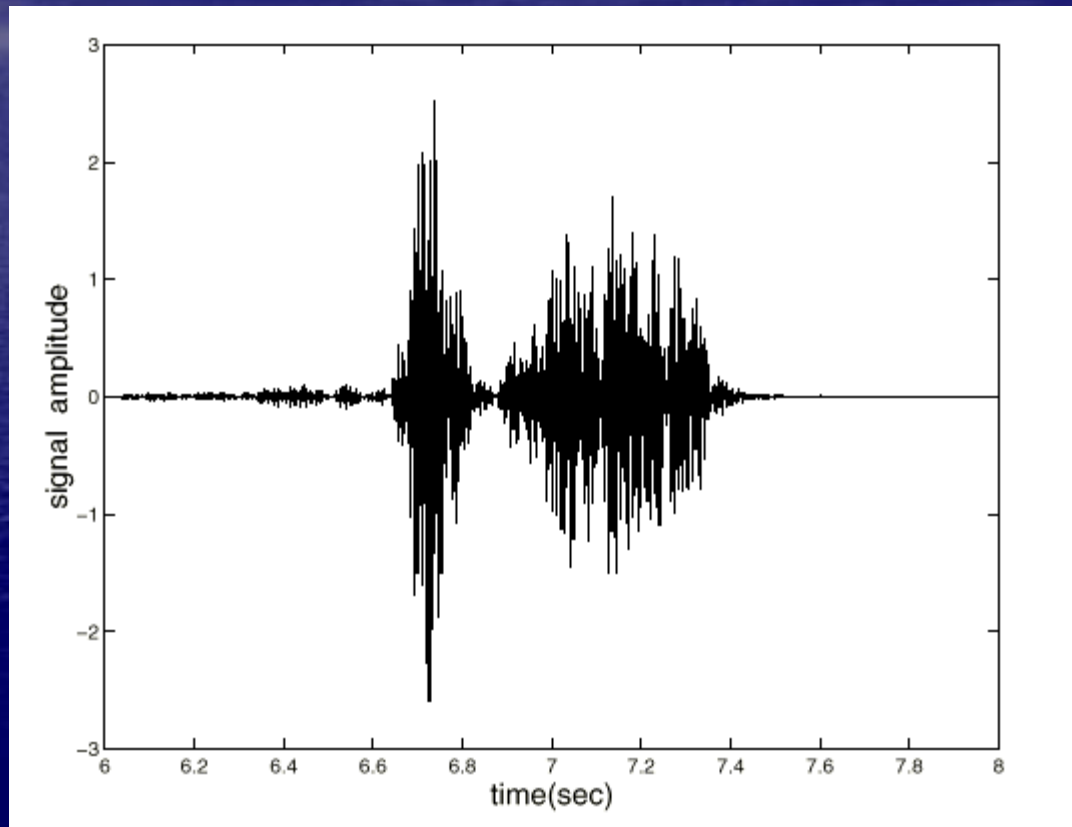
$$p(r, z; \omega) = \frac{i}{4\rho_1} \sum_{n=1}^N H_0^{(1)}(\kappa_n r) u_n(z_0; \omega) u_n(z; \omega)$$

Broad-band propagation

$$p'(r, z, \omega) = p(r, z; \omega)S(\omega)$$

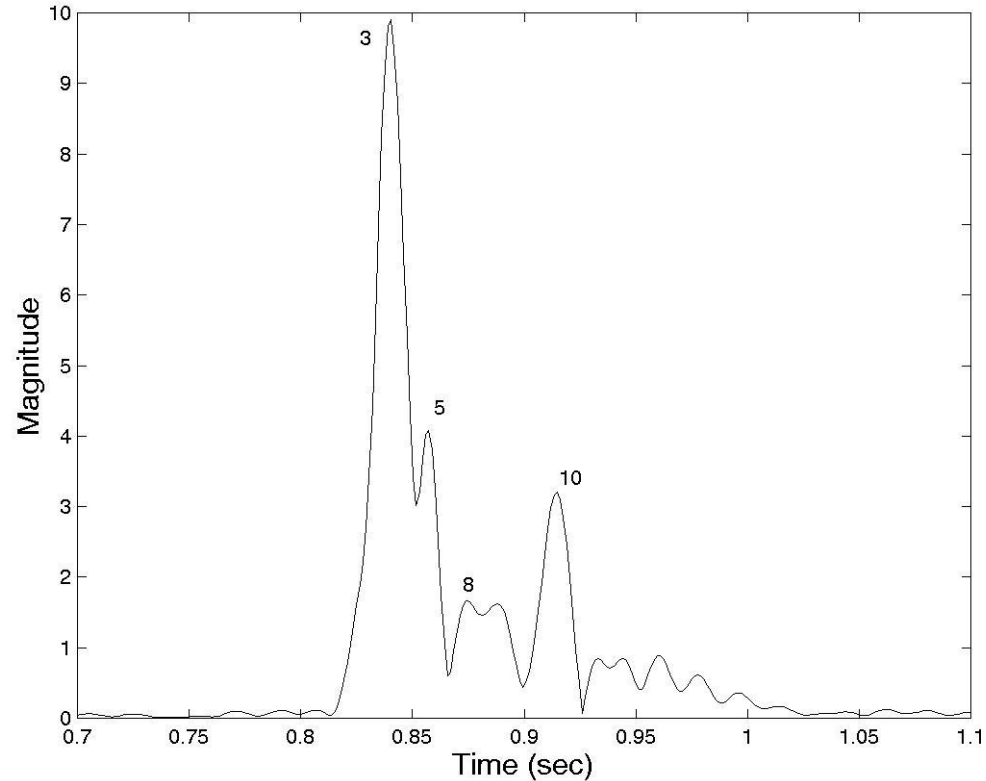
The pressure in the time domain

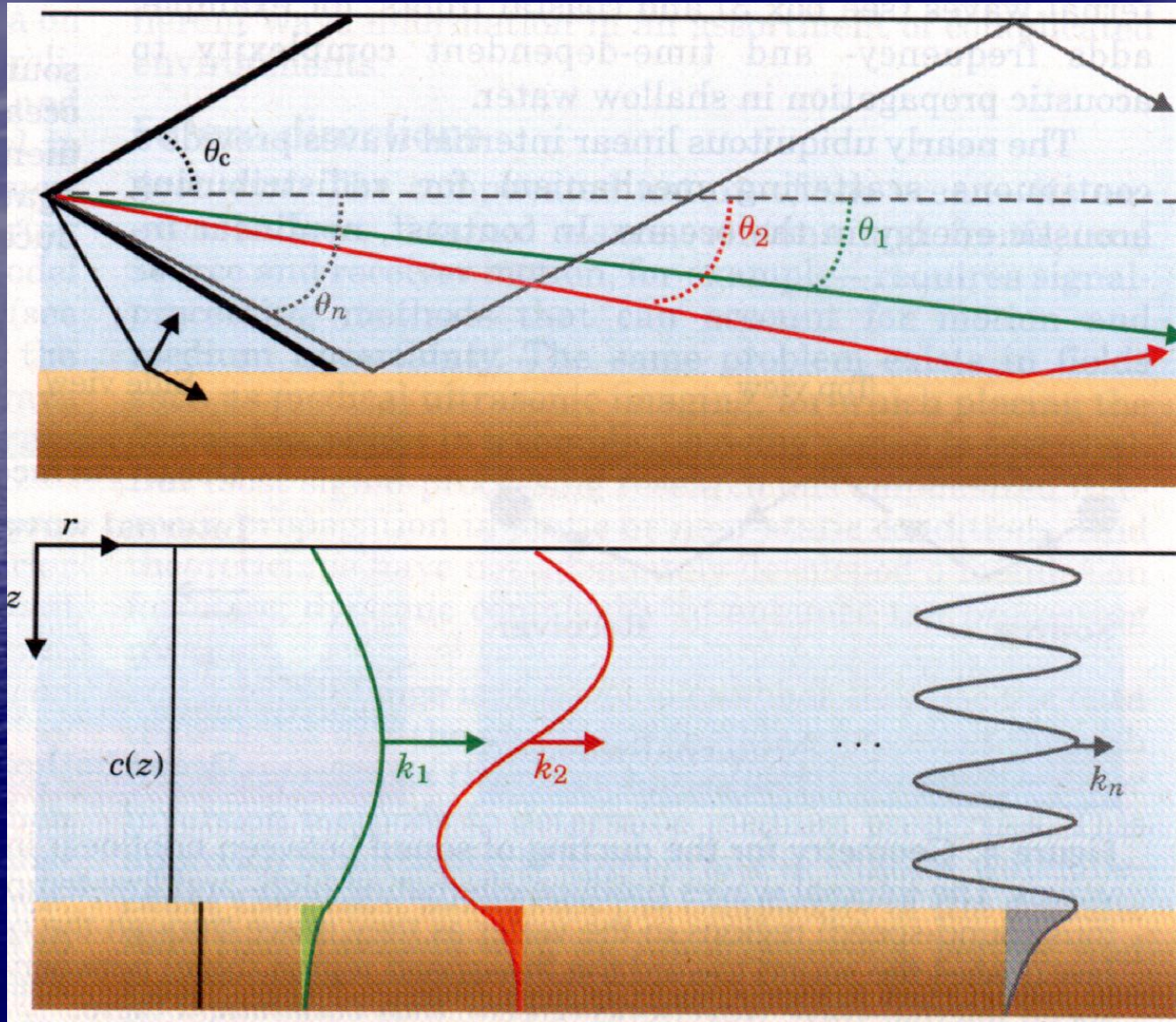
$$p(r, z; t) = \mathfrak{F}^{-1} [p'(r, z, \omega); \omega \rightarrow t]$$



Each mode propagates at a different group velocity

$$v_{gn} = \left. \frac{\partial \omega}{\partial \kappa_n} \right|_{\omega_0}$$





$$v_{gn} = \left. \frac{\partial \omega}{\partial \kappa_n} \right|_{\omega_0}$$

$$\tau_n = \left. \frac{\partial \bar{\kappa}_n}{\partial \omega} \right|_{\omega_0} r$$

Methods of Ocean Acoustic Tomography

Linear methods

Ray Inversions

Linearizing with respect to a reference environment

$$c(\vec{x}) = c_0(\vec{x}) + \delta c(\vec{x})$$

$$\delta\tau_n = \int_{\Gamma_n} \frac{\delta c(\vec{x})}{c_0^2(\vec{x})} ds$$

$$d_n = \int_{\Gamma_n} \frac{v(\vec{x})}{c_0^2(\vec{x})} ds$$

Ray Inversions

$$c(\vec{x}) = c_0(\vec{x}) + \delta c(\vec{x}) \qquad \delta\tau_n = \int_{\Gamma_n} \frac{\delta c(\vec{x})}{c_0^2(\vec{x})} ds$$

Model ?

Determination of the eigen-ray path Γ_n

Parameters ?

$$\delta c(\vec{x})$$

Measurements ?

$$\delta\tau_n \longrightarrow d_i$$

Definition of an inverse problem of the form:

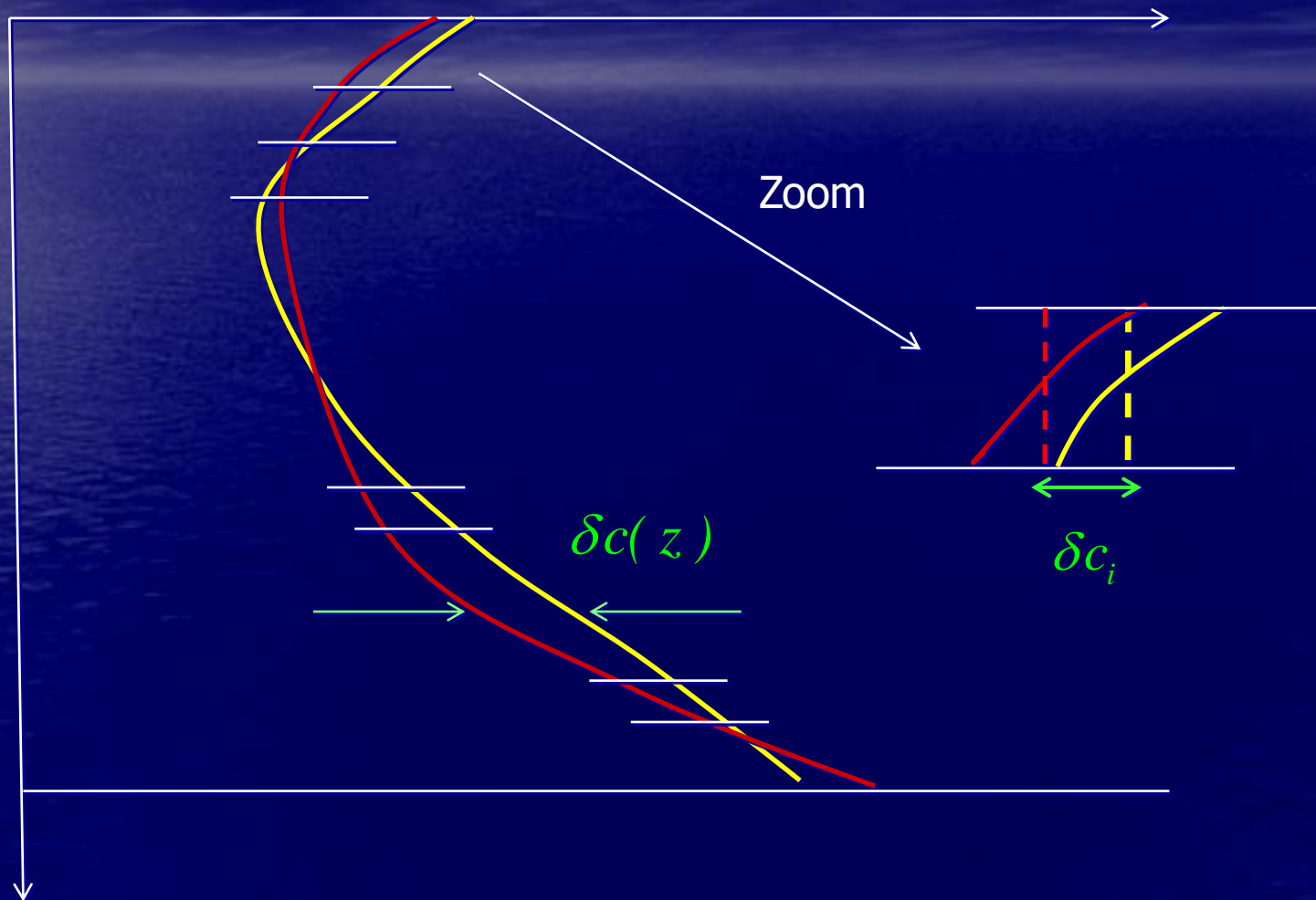
$$d_i = \int G_i(\vec{x}) m(\vec{x}) d\vec{x}$$

We need the theory to predict $G_i(\vec{x})$

After suitable discretization

$$d_i = \sum_{j=1}^N G_{ij} m_j$$

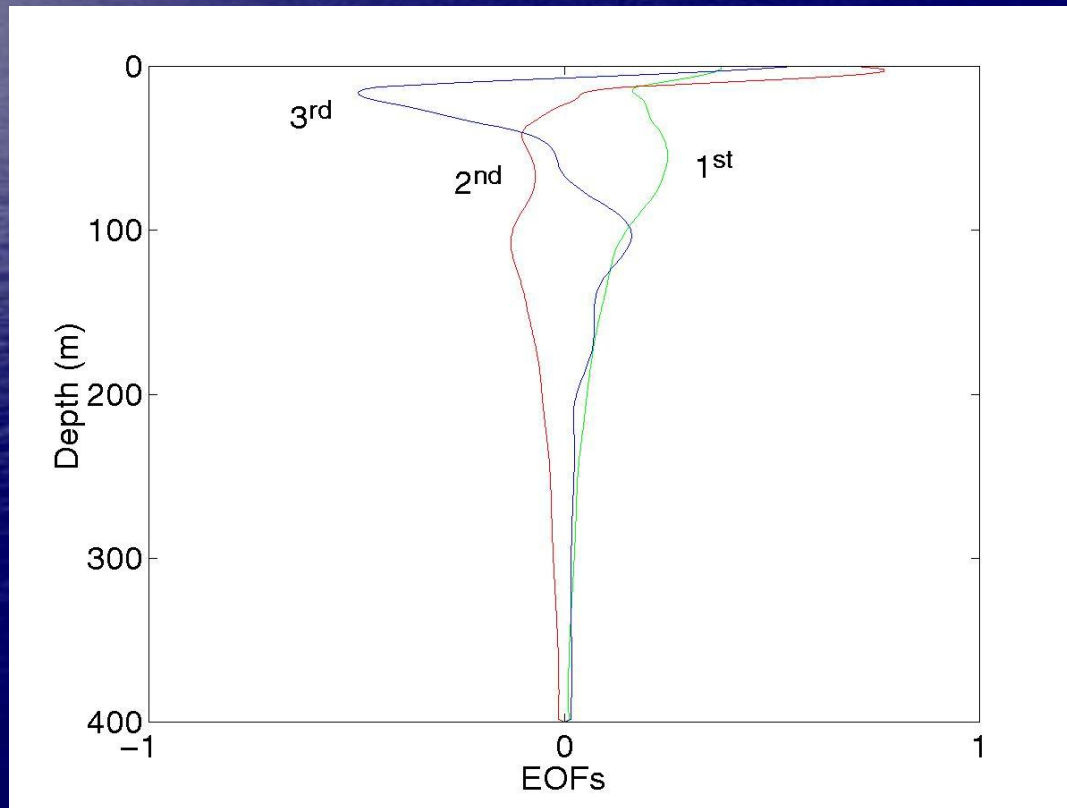
Discrete Linear Inverse Problem



Using Empirical Orthogonal Functions

$$\delta c(z) = \sum_{\ell} \theta_{\ell} f_{\ell}(z)$$

$$\delta c(r, z) = \sum_{\ell} \theta_{\ell}(r) f_{\ell}(z)$$



Additional a-priori
information !

Necessary Assumption :

Ray arrivals can be resolved and identified !

If not

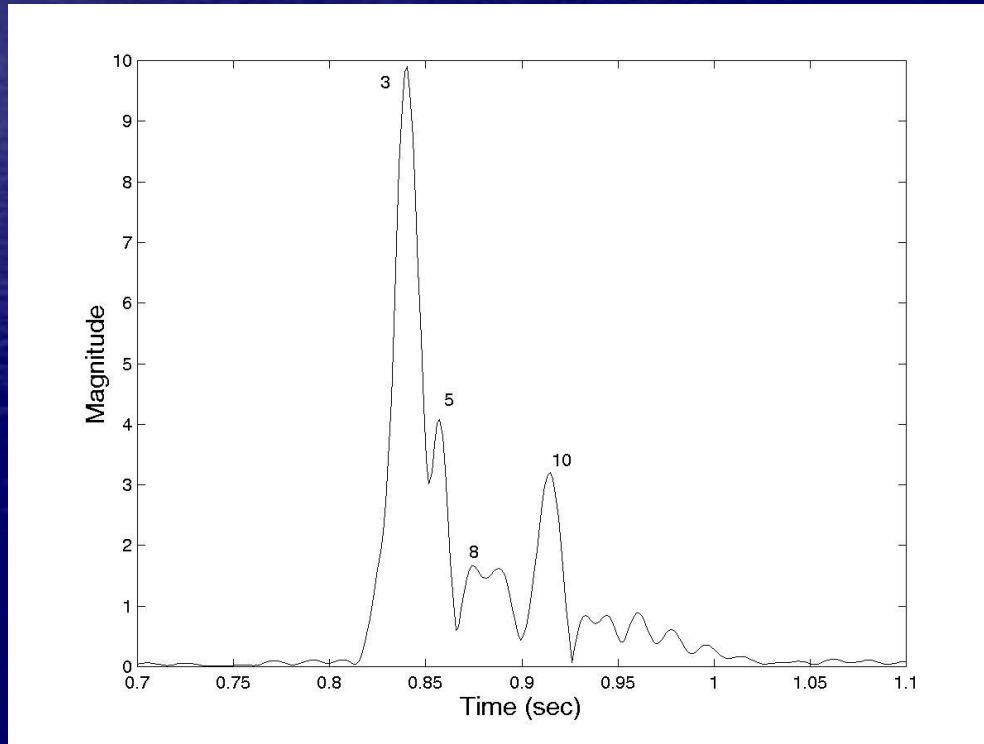
Check for modal arrivals !!

Modal Travel-Time Inversions

If modal arrivals can be resolved and identified !!

Modal packets travel with their respective group velocity

$$v_{gn} = \left. \frac{\partial \omega}{\partial K_n} \right|_{\omega_0}$$



Linearizing with respect to a reference environment we get an expression of the travel time differences in terms of sound speed differences

$$\tau_n = \left. \frac{\partial \bar{\kappa}_n}{\partial \omega} \right|_{\omega_0} r \quad \delta \tau_n(r) = \left. \frac{\partial \delta \bar{\kappa}_n r}{\partial \omega} \right|_{\omega_0}$$

From theory and assuming low order perturbations

$$\delta \bar{\kappa}_n = \frac{1}{\kappa_n^0} \int_D \frac{1}{\rho(\vec{x})} \left| u_n^0(\vec{x}) \right|^2 \frac{k^{02} \delta c(\vec{x})}{c_0(\vec{x})} d\vec{x}$$

By suitable discretization or expression of the sound speed difference in terms of EOFs we define a linear system of the form

$$\delta\tau_n(r) = \left. \frac{\partial \delta\bar{\kappa}_n r}{\partial \omega} \right|_{\omega_0} = \sum_{i=1}^I \sum_{j=1}^J \left(\left. \frac{\partial Q_{n,i,j}^0}{\partial \omega} \right|_{\omega_0} \right) \delta c_{i,j} \Delta z \Delta r$$

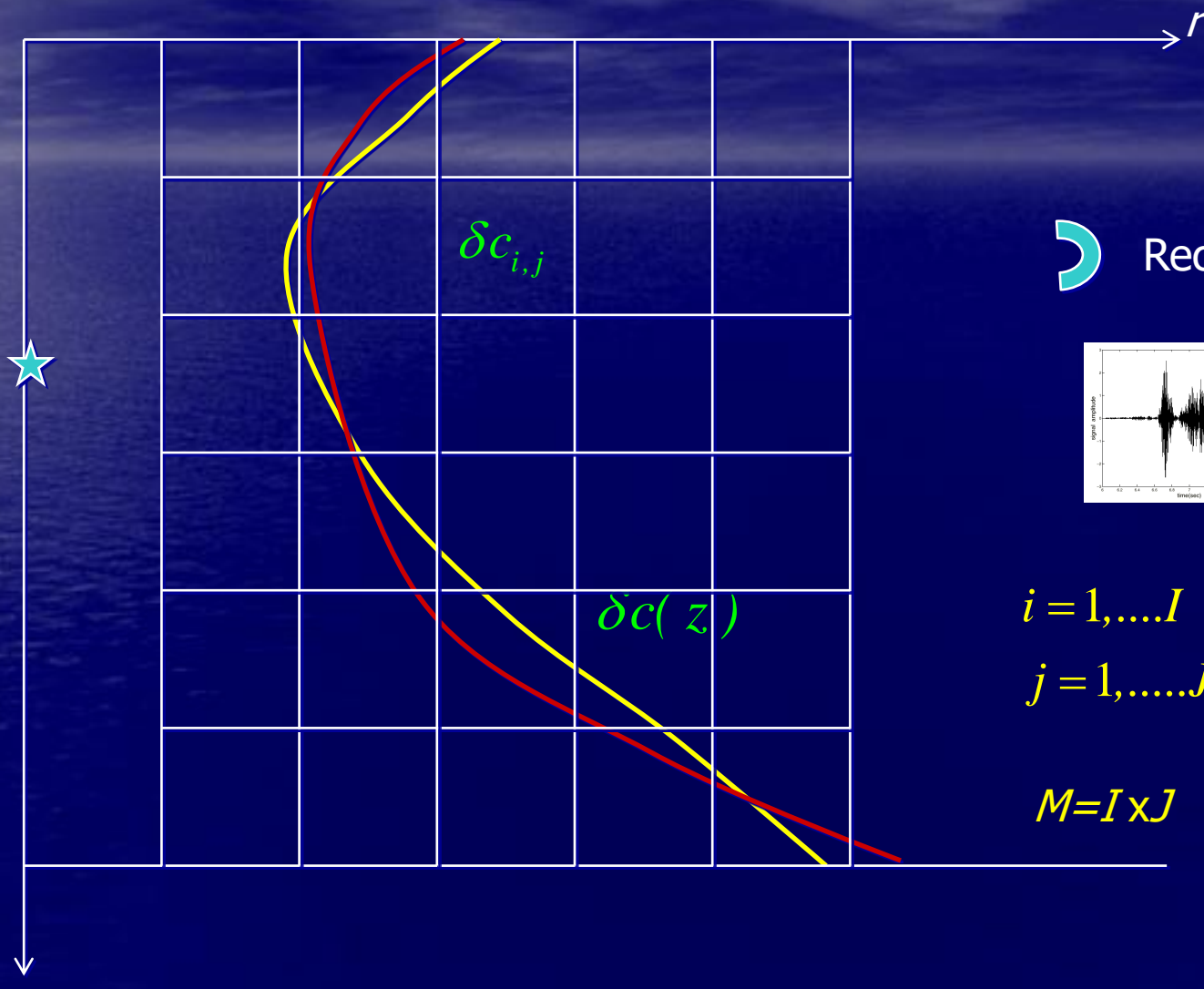
$$d_n \rightarrow \delta\tau_n$$

$$n=1, \dots, N$$

$$d_n = \sum_{k=1}^M G_{nk} m_k$$

N : number of identifiable modal arrivals

Source



$$i = 1, \dots, I$$

$$j = 1, \dots, J$$

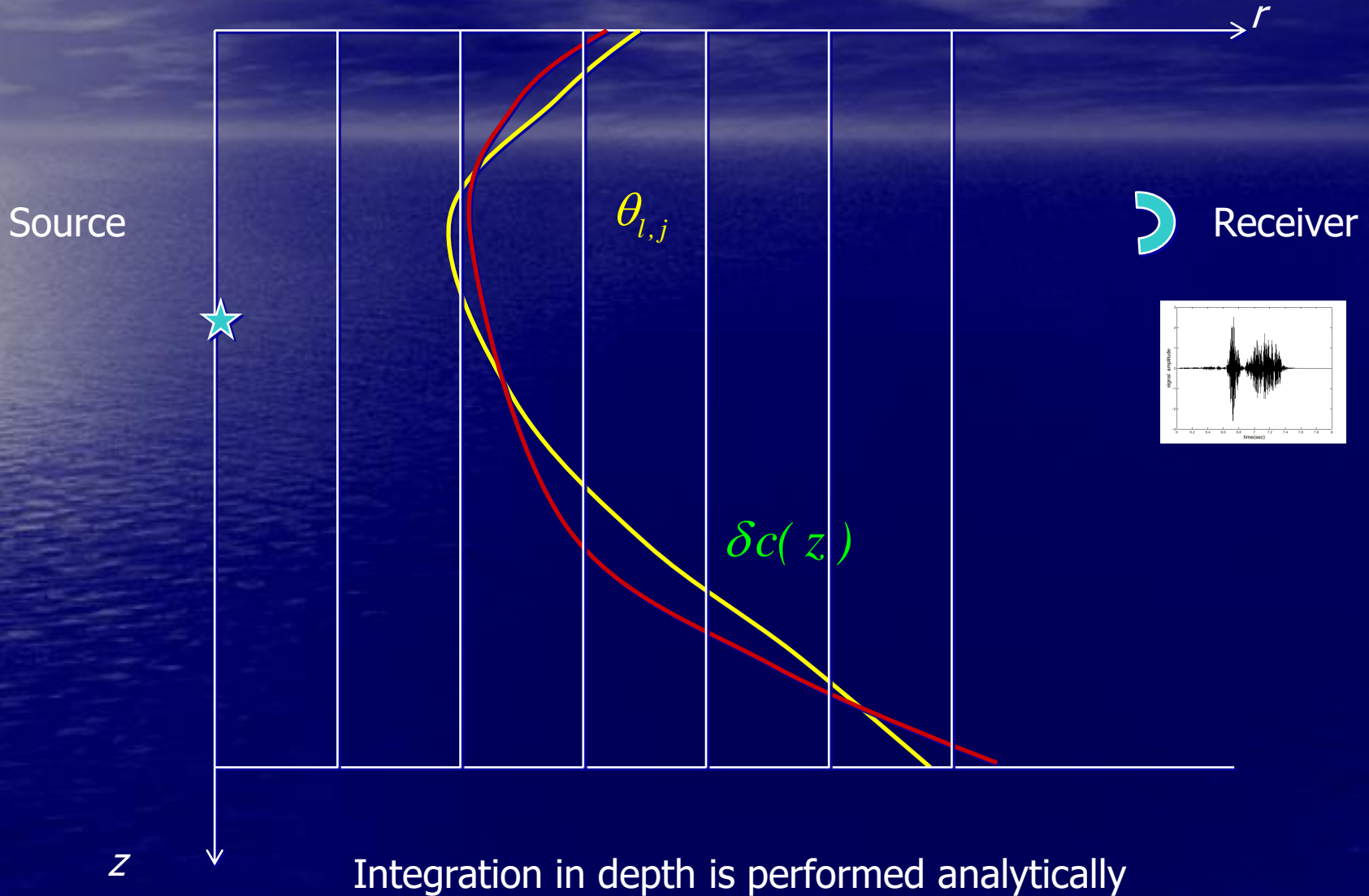
$$M = I \times J$$

z

r

Using EOFs

$$\delta c(r, z) = \sum_{\ell} \theta_{\ell}(r) f_{\ell}(z) \Rightarrow \delta c(r_j, z) = \sum_{\ell} \theta_{\ell, j} f_{\ell}(z)$$



Advantages of linear methods :

- Easy to implement
- Computationally efficient and fast

Disadvantages of linear methods :

- Require good a-priori knowledge of the environment
- The kernel matrix sometimes is characterized by bad condition

Non linear methods

Measurements (Data)

- We need to define the “observables”
- They have to be easily measured or inferred after some post-processing of the measured data.
- They should be sensible to changes of the environmental parameters.

Measurements (Data)

- Measurements are always the acoustic field either at a single hydrophone or on an array of hydrophones.
- The data to be used in the inverse problem can be

Measurements (Data)

- The field at a specific frequency or at a frequency band (acoustic pressure)
- The ray or modal arrivals (time of arrival)
- The dispersion curves (characteristic of modal propagation)
- The modal “phase” for each one of the propagating modes
- Statistical or probabilistic features of the signal

Matched Field Inversions

If a vertical array of hydrophones is available, the full-field can be used for inversions

Matching at a single frequency :

Perform Discrete Fourier Transform to obtain the field in the frequency domain at each one of the N hydrophones

$$F_n(\vec{x}; \omega) = \sum_{k=0}^K p_n(\vec{x}; t_k) w(t_k) e^{-i\omega t_k}$$

- Collect measurements of the acoustic field in $N \geq 1$ hydrophones
- Select a propagation model and use it to compute the acoustic field for candidate model parameters
- Use a suitable processor to cross-correlate the measured with modelled/predicted replica fields and search for the *highest* correlation

Optimization Process !!

Measured in N
hydrophones

$$\mathbf{F} = (F_1, F_2, \dots, F_N)^T$$

Modeled for N
hydrophones

$$\hat{\mathbf{F}} = (\hat{F}_1, \hat{F}_2, \dots, \hat{F}_N)^T$$

$$C = \langle \mathbf{F}\mathbf{F}^+ \rangle$$

Compare

$$w = \langle \hat{\mathbf{F}}\hat{\mathbf{F}}^+ \rangle^{-1} \hat{\mathbf{F}}$$

$$L(\mathbf{m}) = \mathbf{w}^+ C \mathbf{w}$$

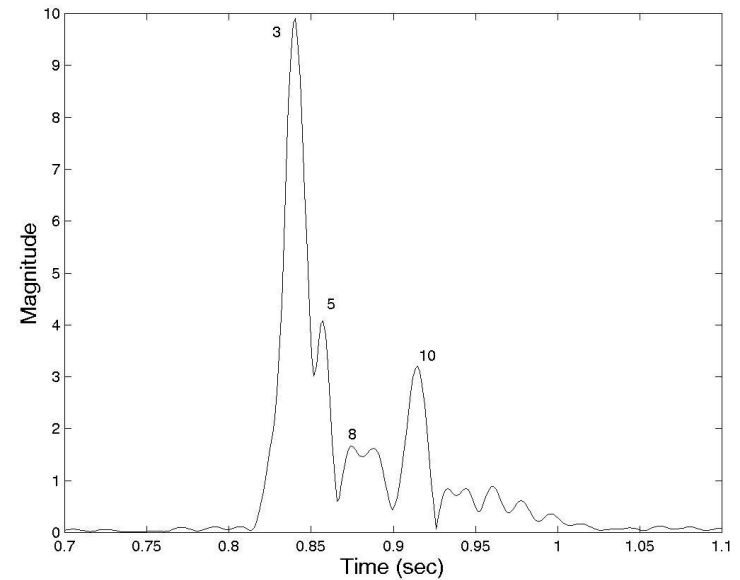
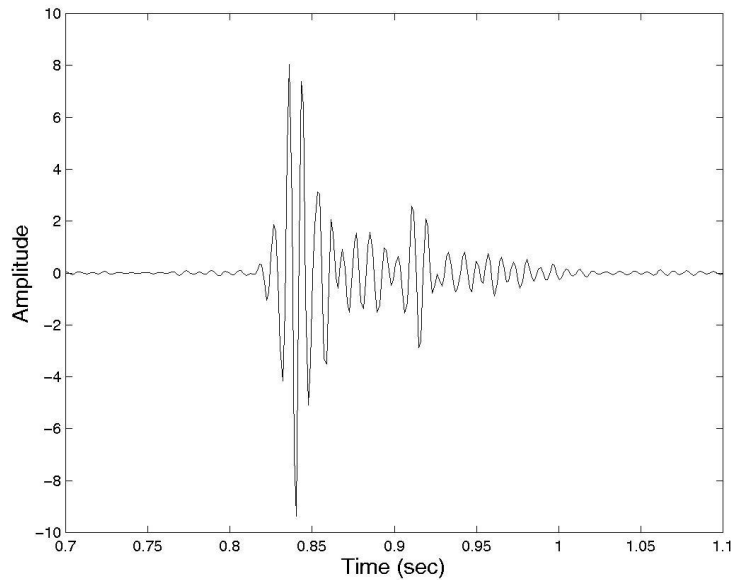
Bartlett processor

\mathbf{m}^{est} will be the vector maximizing the Processor

For normalized data, maximum of the $L(\mathbf{m})$ is 1

Inversions based on travel time

Necessary Condition : Identification of the type of signal peaks:



Not always easy !

Inversions based on modal travel time

Hypothesis : Identification of N modal arrivals

Replica fields are produced based on the pre-specified search space of the model parameters.

Use of the group velocities calculated for the modes of the replica fields to identify the modeled arrivals of the propagating modes ,

$$\hat{t}_n, n = 1, \dots, N$$

Define the actual modal arrivals t_n and compare the corresponding times with the arrival times of the replica signals.

$$\delta t_n = t_n - \hat{t}_n$$

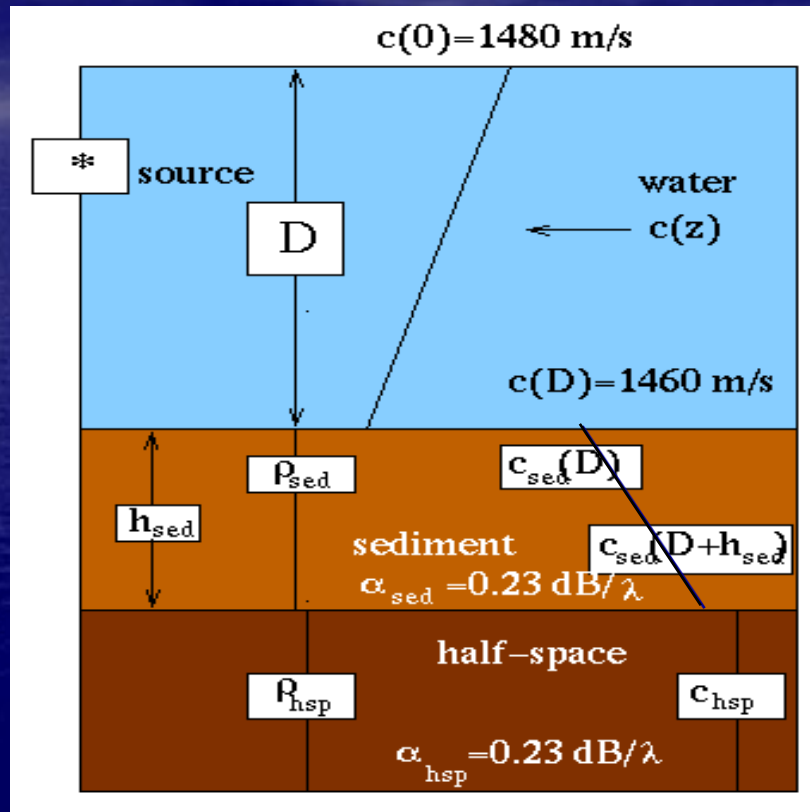
The inverse problem is formulated as an implicit non-linear problem

$$\delta t_n = g(\mathbf{m})$$

Define an appropriate processor P to solve the inverse problem as an optimization process

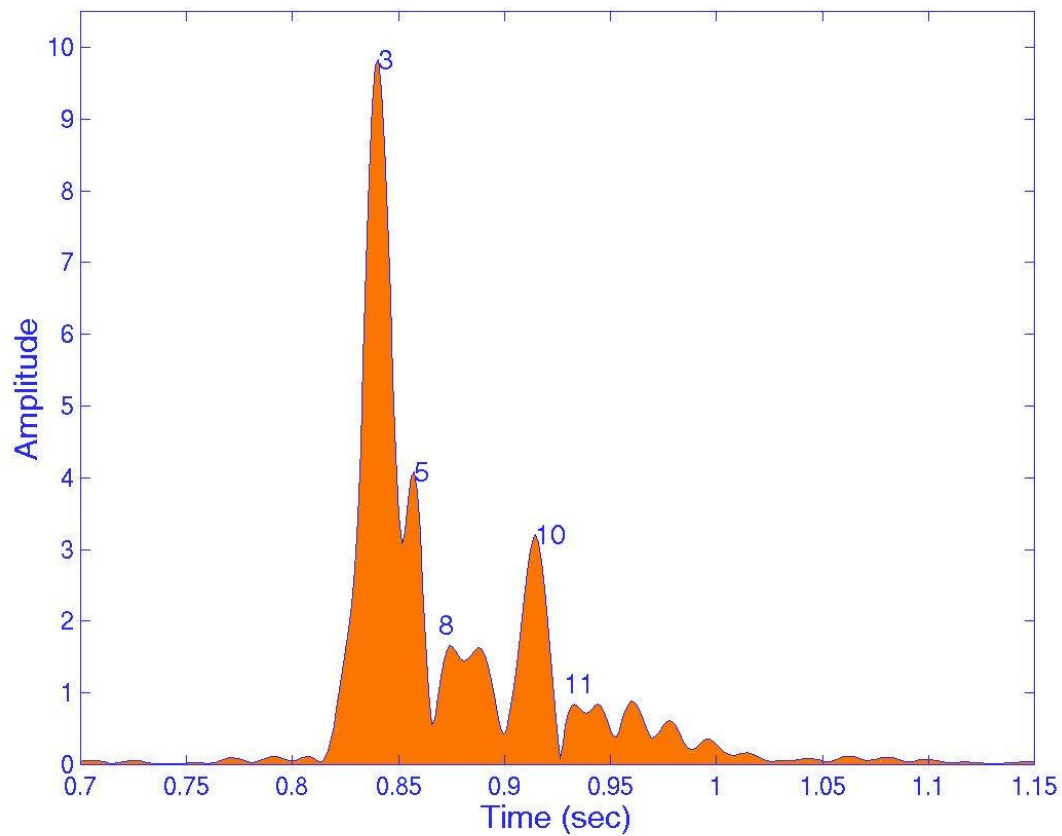
$$P(\delta t_n) = \sqrt{\frac{1}{N} \sum_{n=1}^N \delta t_n^2}$$

Test case

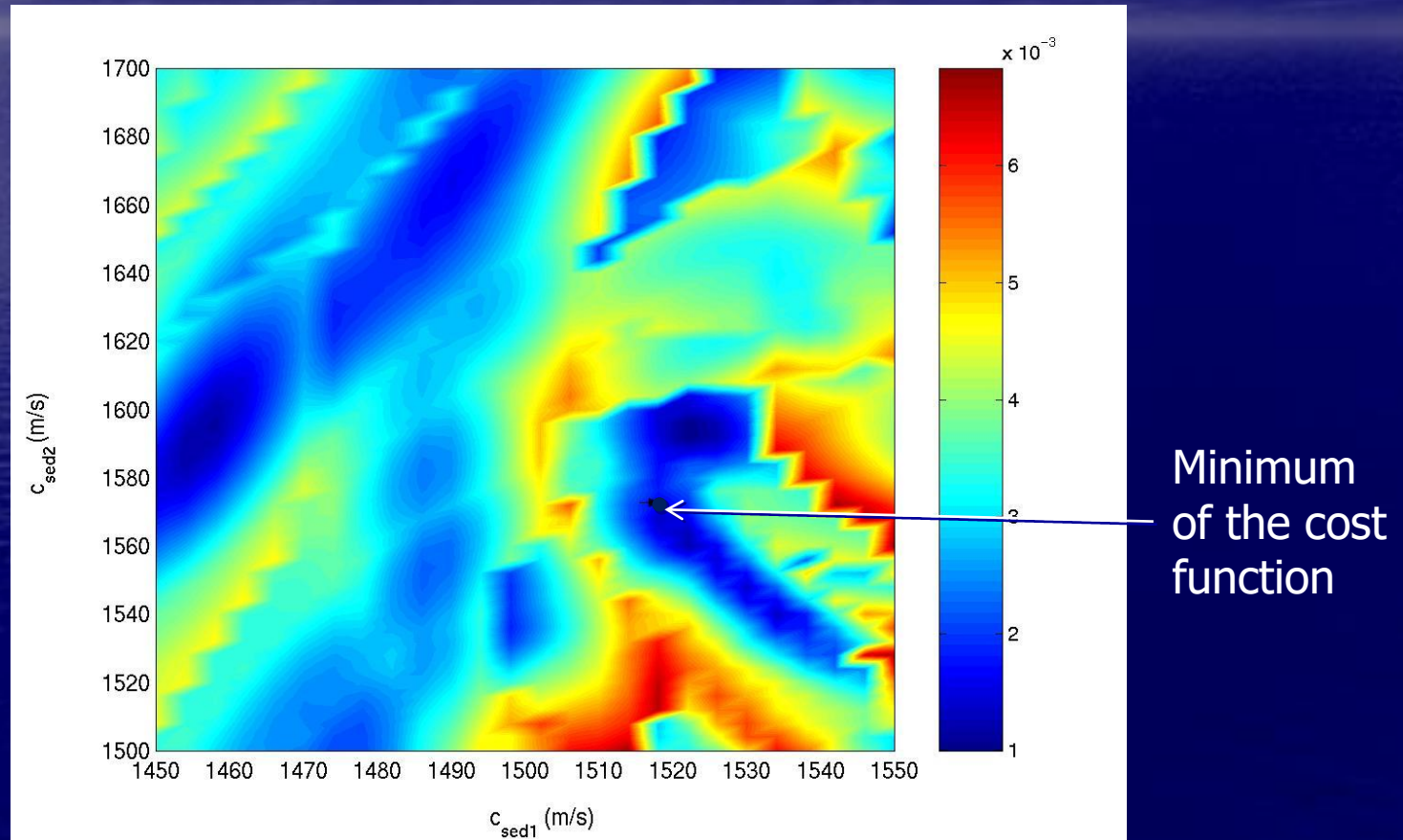


A description of the test environment based on the benchmark case WAa of the Vancouver 97 Workshop

Arrival pattern ($f_0 = 112$ Hz)



The cost function over the whole search space for the sediment properties



Inversion results

Parameter	Actual	Reference	Recovered
$C_{sed}(D)$	1516,2	1506,1	1517,44
$C_{sed}(D+h)$	1573,2	1539,5	1569,84

A systematic search over the search space has been performed

No specific optimization algorithms have been applied