#### Acoustic multipath arrival time estimation via blind channel estimation

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SCHOOL OF OCEAN AND EARTH SCIENCE AND TECHNOLOGY

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### **Objective and Background**

- Objective: Estimate multipath arrival times by blindly calculating underwater acoustic impulse responses using marine mammal vocalizations
  - Blind?
    - No knowledge of source signal required
    - No assumptions on environment (e.g., no range-independent assumption); non-parametric
  - Potential uses include:
    - Localization of both impulsive and non-impulsive marine mammal vocalizations
    - Ocean tomography via unknown sources

### What is an impulse response (IR)?

- The output (or response) of a system when the input is a unit impulse
- Given the system IR, the system output given any input signal is:  $x(t) = h(t) \otimes s(t)$

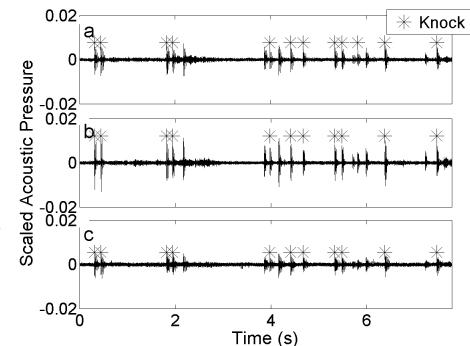
x(t) = system output h(t) = impulse response s(t) = system input

### Impulse Responses and Marine Mammals

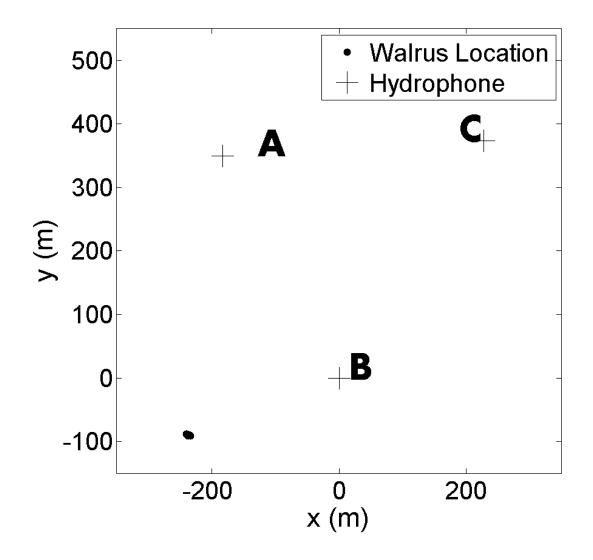
- Consider the case of a walrus (or sperm whale) producing impulsive vocalizations
- Effectively, the received signals from these animals is an approximation of the impulse responses for the acoustic channels between the animal and the receiver
- One impulse response for each source:receiver pair

#### Walrus Vocalizations

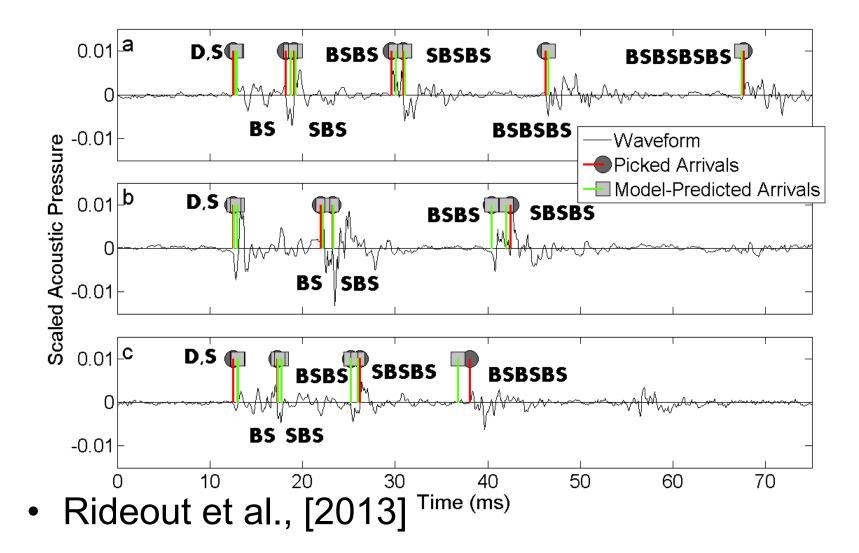
- Walrus vocalizations include knocks, grunts, and bells
- Knocks are impulsive vocalizations made underwater, primarily by male walruses



#### Impulse Response Information Content



# Impulse Responses and Marine Mammals, cont'd.



### **Theory: Problem Formulation**

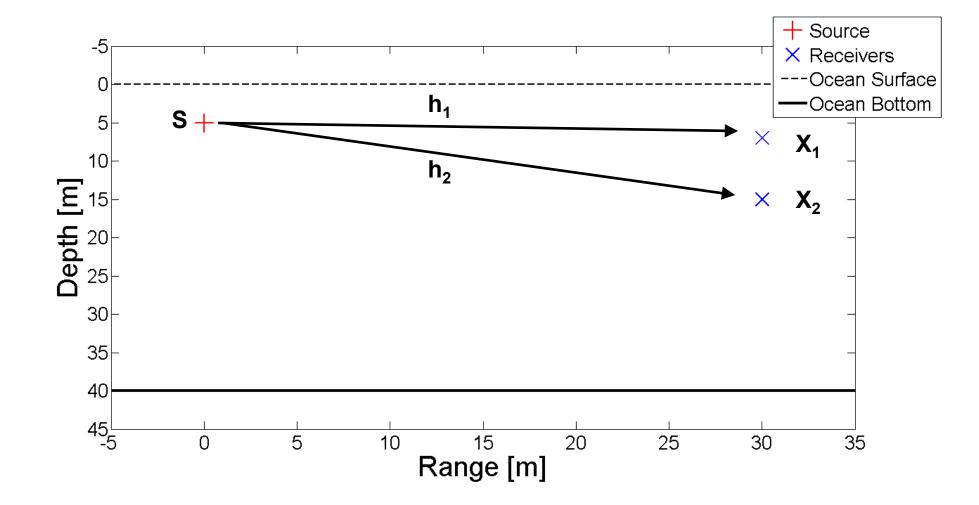
• Received data:

$$x_1(k) = h_1(k) \otimes s(k)$$
$$x_2(k) = h_2(k) \otimes s(k)$$

• If the channels share a common source:

$$h_{2}(k) \otimes x_{1}(k) = h_{2}(k) \otimes [h_{1}(k) \otimes s(k)]$$
$$= h_{1}(k) \otimes [h_{2}(k) \otimes s(k)]$$
$$= h_{1}(k) \otimes x_{2}(k)$$
$$h_{2}(k) \otimes x_{1}(k) - h_{1}(k) \otimes x_{2}(k) = 0$$

## Theory: Problem Formulation, cont'd.



# Theory: Problem Formulation, cont'd.

 Convolution can be expressed as a system of equations (L = # samples in the channels, N = # data):

$$\begin{bmatrix} X_1(L) \vdots - X_2(L) \end{bmatrix} \begin{bmatrix} h_2 \\ h_1 \end{bmatrix} = 0 \qquad \qquad X(L)h = 0$$

• Where

$$h_m = [h_m(L), \dots, h_m(1)] \qquad X_m(L) = \begin{bmatrix} x_m(1) & x_m(2) & \dots & x_m(L) \\ x_m(2) & x_m(3) & \dots & x_m(L+1) \\ \vdots & \vdots & \ddots & \vdots \\ x_m(N-L) & x_m(N-L+1) & \dots & x_m(N) \end{bmatrix}$$

- Solve the homogeneous system of equations for the IR vector, h
- Generalizable to any number of receivers

## **Theory: Optimization Algorithm**

- To solve the over-determined system of equations X(L)h = 0 (i.e., estimate the IRs), an iterative, L1 optimization routine called NESTA (Becker et al., 2011) is used.
  - Designed to efficiently acquire compressible signals, and adapted to estimate IRs
  - Doesn't require precise knowledge of channel length, nor knowledge of number of significant non-zero IR samples

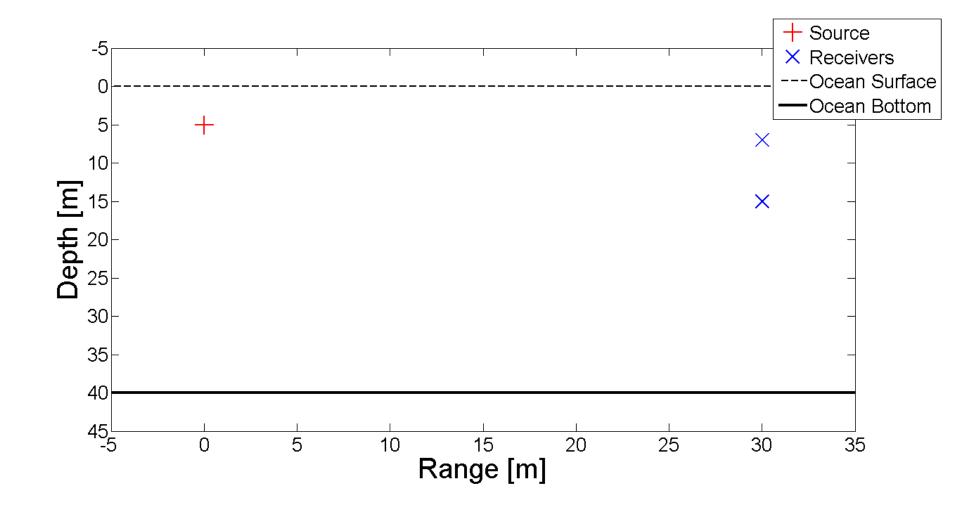
# Theory: Optimization Algorithm, cont'd.

- NESTA (Nesterov's Algorithm)
  - Application of compressed sensing which builds upon classical Basis Pursuit techniques
    - Compressed Sensing
      - Identification of smallest number of signal components containing the most information
    - Basis Pursuit
      - Decomposition of a signal (e.g., IR) into an 'optimal' (i.e., minimum I1 norm) combination of components
  - Estimated IRs balance data fit with solution size using tradeoff parameter

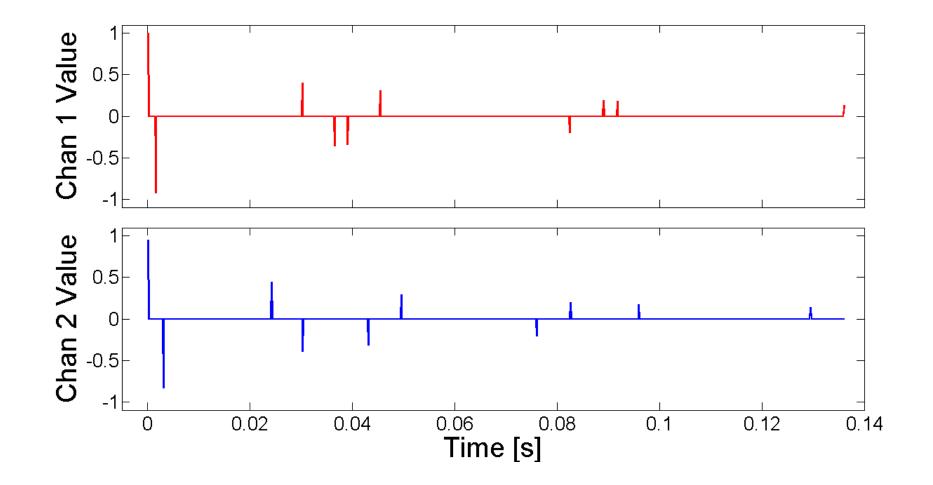
• Cost Function = 
$$\lambda \|x\|_{\ell_1} + \frac{1}{2} \|b - Ax\|_{\ell_2}^2$$

- Tradeoff parameter proportional to noise power
  - As noise increases, smaller solutions preferred

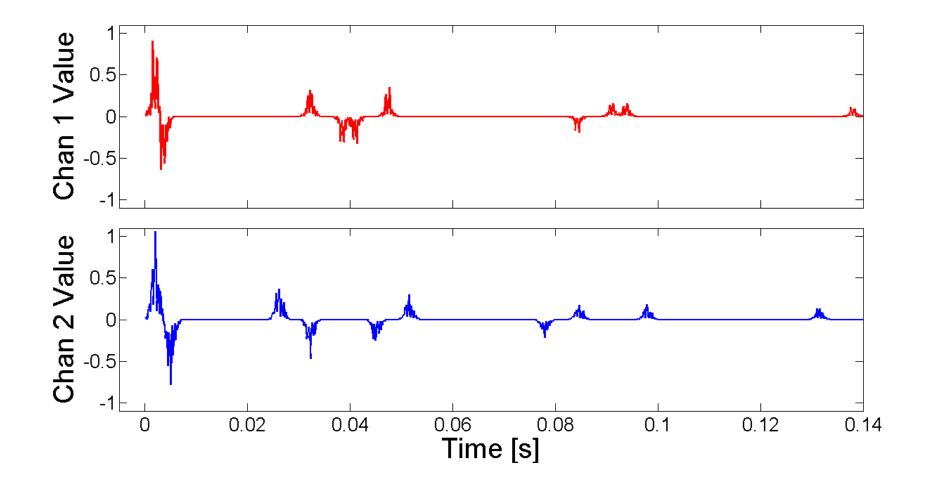
### Simulation: Environment



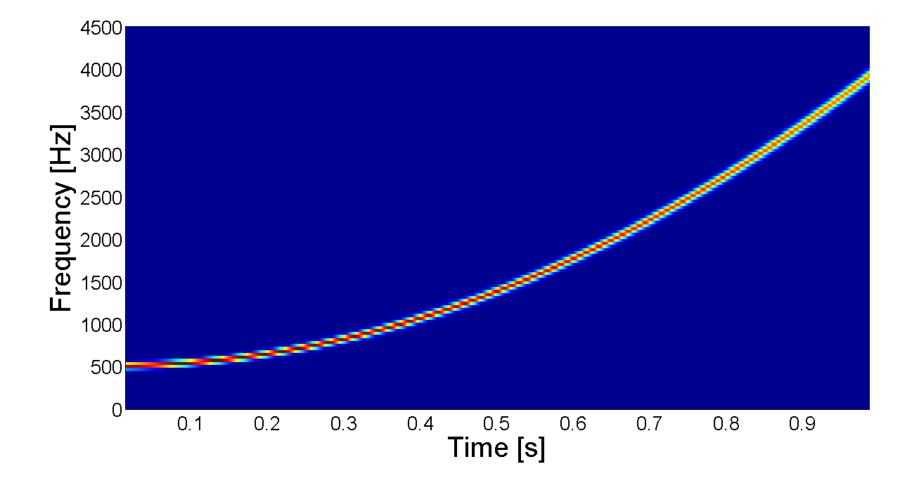
#### Simulation: Calculated IR

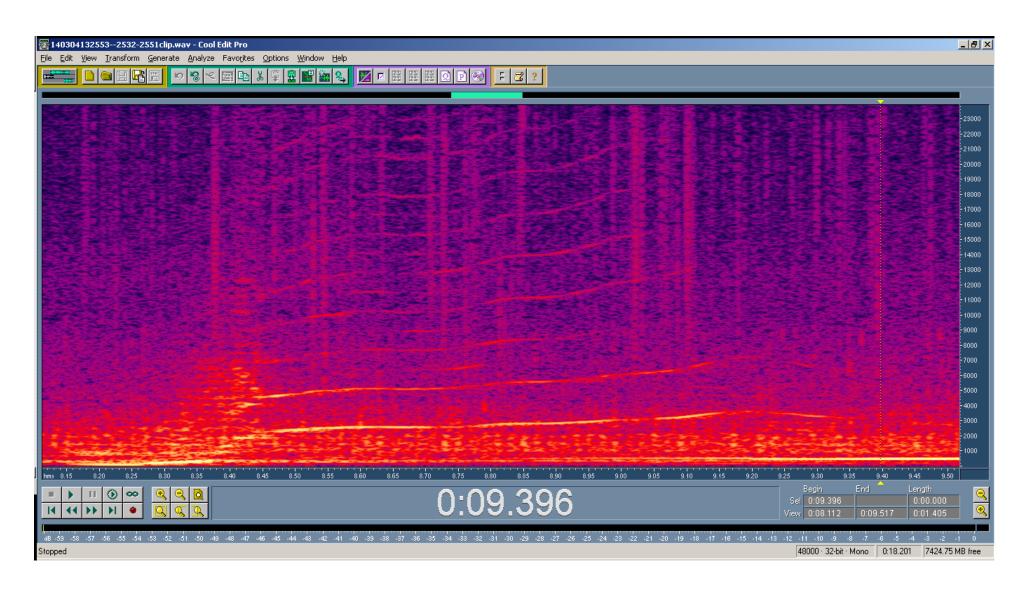


#### Simulation: IR with Scattering Model



#### Simulation: Source Waveform

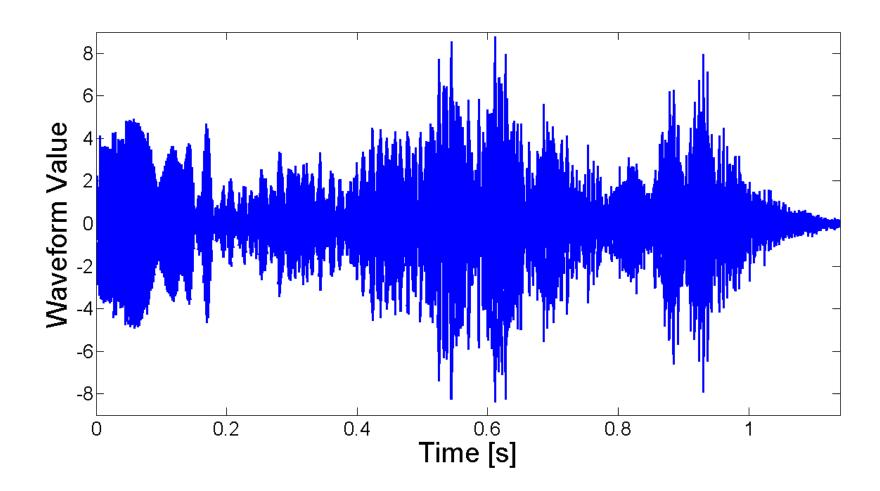




• Chen et al., [unpublished]

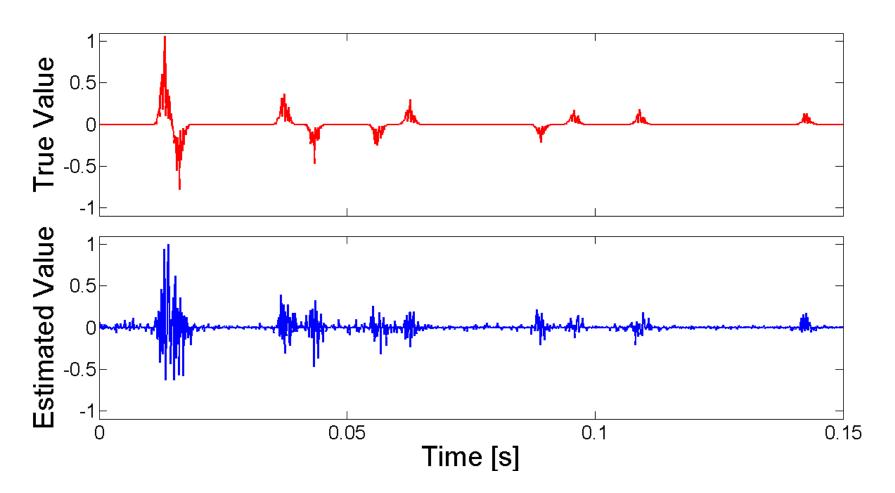
#### Simulation: Received Waveform

• SNR= = ~44dB



#### Simulation: True & Estimated Channels

Channel 2

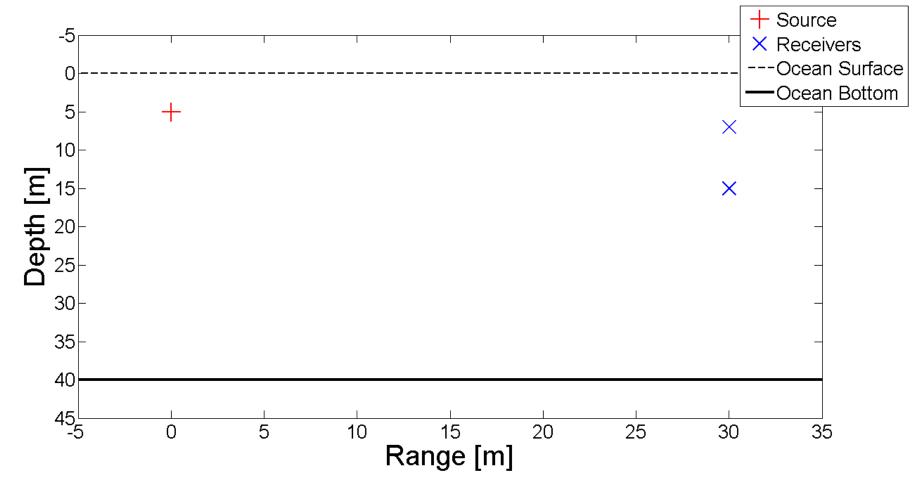


## Summary and Next Steps

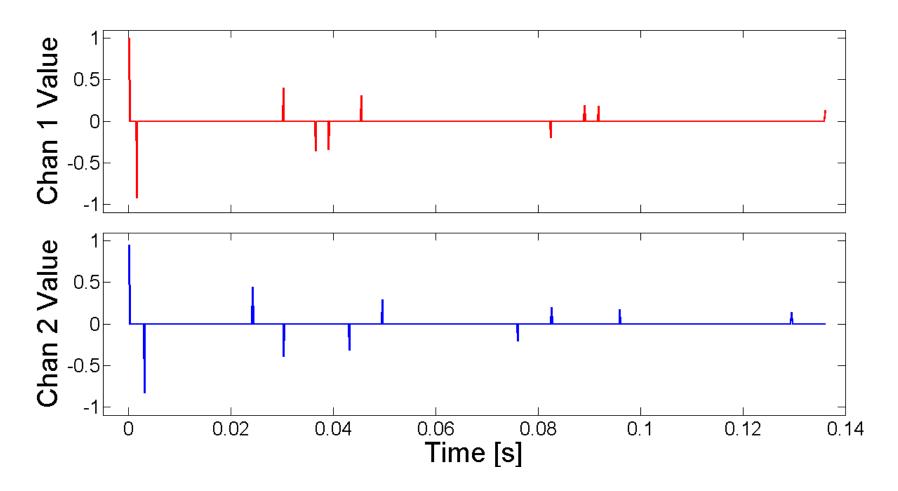
- Adapted a blind channel estimation algorithm to estimate underwater acoustic channel impulse responses using bioacoustic signals
- Simulations show promising results
  - Identification of multipath arrival times possible in estimated IR
- Currently working to process measured data

### Questions?

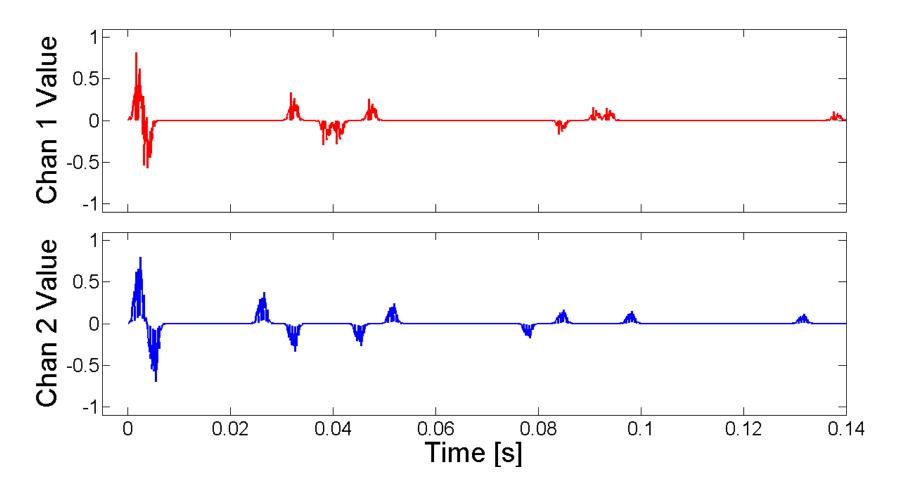
• 1. Define Environment



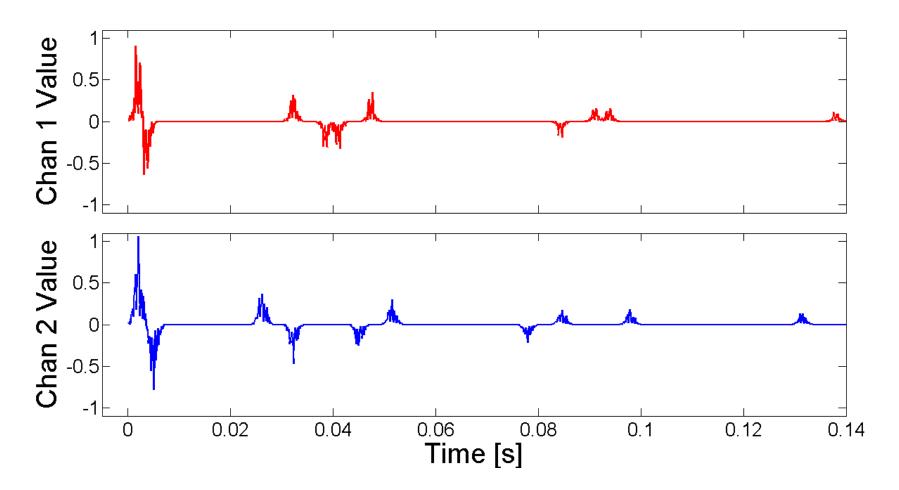
• 2. Calculate arrival times and TL



3. Random Gaussian Pulse Convolution



• 4. Uniform Random Multiplication

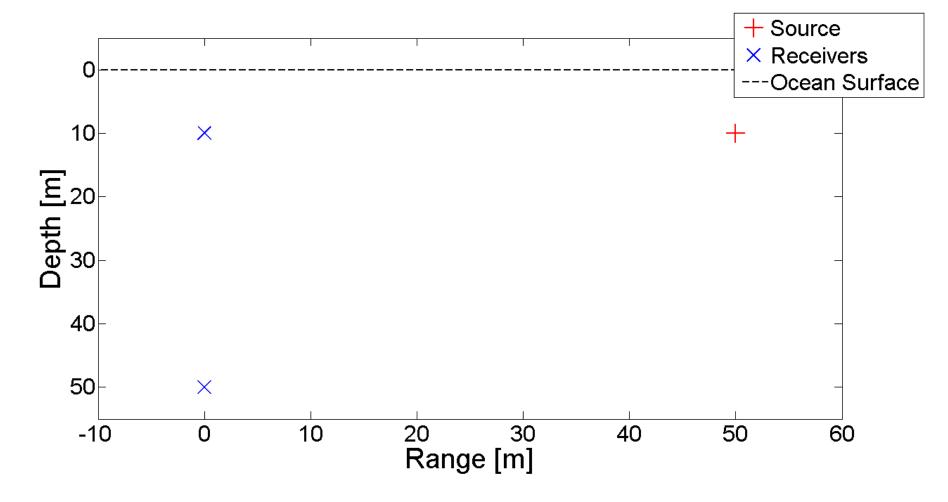


## **Processing Steps**

- Recorded Waveform
- Estimate Impulse Responses
- Estimate Multipath Arrival Times

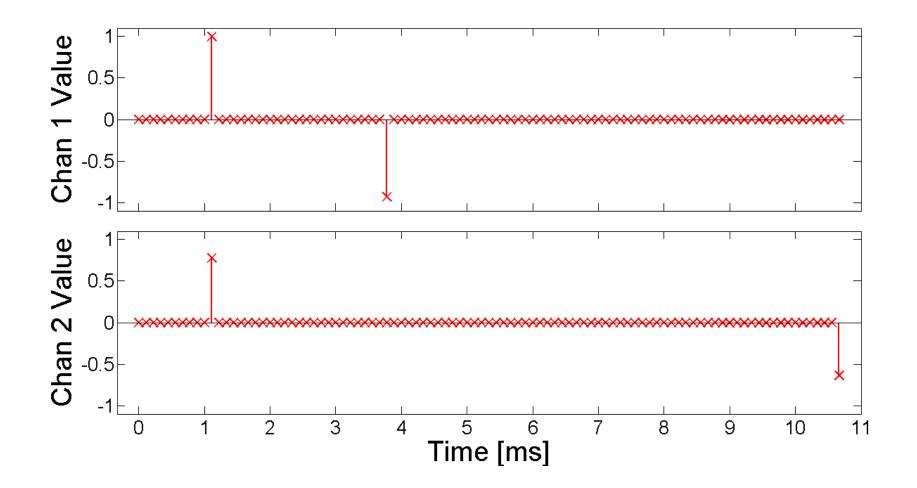
## **Theory: Cost Function Depiction**

Simulated Environment



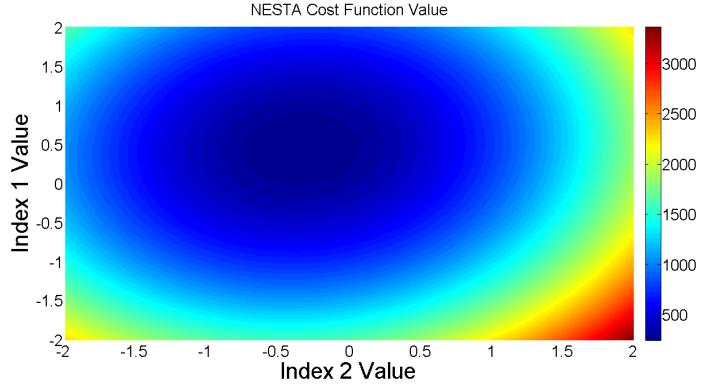
## Theory: Cost Function Depiction, cont'd.

• True Impulse Responses (time shifted)



# Theory: Cost Function Depiction, cont'd.

• Cost Function =  $\lambda \|x\|_{\ell_1} + \frac{1}{2} \|b - Ax\|_{\ell_2}^2$ SNR = -1dB True = [0.78, -0.64] Est = [0.37, -0.34]



### Theory: Performance Considerations

- For quality IR recovery,
  - Sparse channels (e.g., minimal scattering)
  - Wide bandwidth vocalizations
  - High signal-to-noise ratio

## Impulse Responses and Marine Mammals, cont'd.

