

Mapping the Ocean Floor

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Introduction

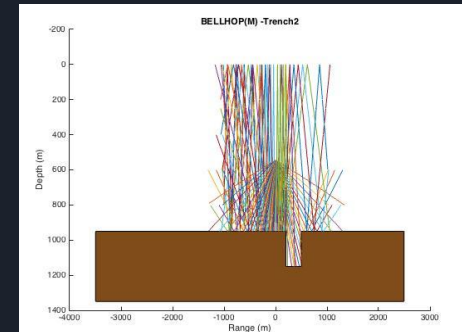
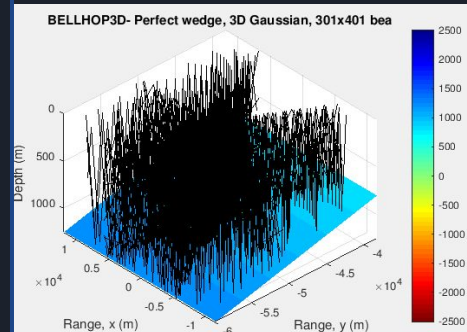
- Competing in the Shell Ocean Discovery XPRIZE. Goal to map 500 km² of the ocean floor autonomously with high efficiency and high resolution (5 meters in horizontal resolution and 0.5 meters vertically).

- Autonomous drone and pod will be deployed into the mapping area.
- Drone with hybrid gas-electric engine ensures the long flight time.
- Pod with self-designed sonar system will gather ocean data.



Simulation Software

- Simulation software built from Bellhop, an open-source program distributed by HLS Research
- 2D/3D ray traces and arrival calculation for various simulated ocean bathymetries
 - Specification of source and receiver positions
 - Gaussian and geometric beam simulation
 - Arrival data - amplitude and time delay



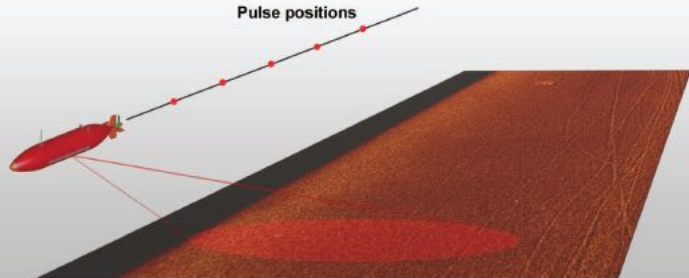
Approach A - Synthetic Aperture Sonar (SAS)

Overview of SAS

- Creates a synthetic aperture, equivalent to an extremely long receiver antenna
- Sonar source and receiver array move along a near linear path, emitting pulses at selected intervals. Typically horizontal path and 2D images.
- SAS was first developed based on existing methods for Synthetic Aperture Radar (SAR)
- Generally uses a horizontal path to produce a two-dimensional image of the sea floor
- With precise enough location information, can produce extremely high-resolution images

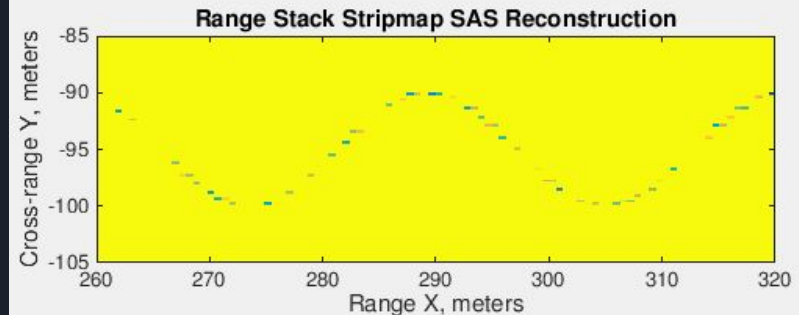
Methods and Goals

- Relies on Fourier transform of reflected signal $s(t,u)$, where t is time and u is synthetic aperture position, to frequency domain $S(\omega,k)$. Currently based on SAR code by Mehrdad Soumekh.
- Pod design uses a vertical SAS path, imaging ocean floor points as 2D, vertical slices.
- Requires directional information for sonar echos. Can be obtained using spatial frequency information or image coregistration between two receivers. Plans to limit source directionality also.



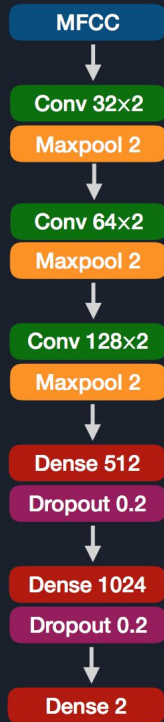
< Typical SAS
Imaging Path


Sine Wave >
Bathymetry
Reconstructed
By Simulation



Approach B - Neural Nets and Error Reduction

Sonar data classification



- Convolutional Neural Networks (CNN) have been used for classifying different bathymetry types.


The image shows two side-by-side bathymetry maps. The left map is labeled 'Bump' and shows a dark brown silhouette of a bump on a lighter brown background. The right map is labeled 'Valley' and shows a dark brown silhouette of a valley on a lighter brown background.
- Mel Frequency Cepstrum Coefficient (MFCC) was used at the beginning of the project to extract useful features from the sonar sound files (input).
- Obtained overall accuracy of 96.7%.

Error reduction method

- In this approach, we add random coherent noise to a low-resolution map to form several high-resolution maps. The map that is most consistent with the underwater sonar data is chosen as the final map.
- Low resolution maps of the ocean floor are available online.
- The detailed processes are:

