















Underwater Computer Vision:

Two Decades of Traditional Approaches
and a New Paradigm in
3-D Reconstruction based on
Opti-Acoustic Stereo Imaging

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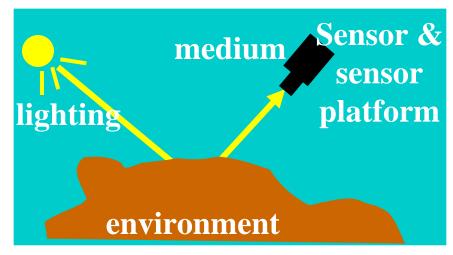


Overview

- Why underwater vision is difficult, and progress has been slow;
- What has been done in favorable undersea conditions;
- Some challenging problems and how they have been overcome;
- Integration of sonar and optical imaging as a new paradigm in 3-D target mapping.

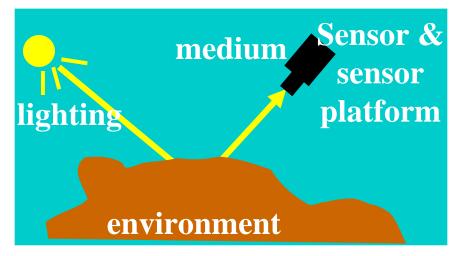
Why slow Progress?

- 1. More difficult problems;
- 2. Very small research community;
- 3. Need high levels of investment, resources (platform, sensors, ship time) and interdisciplinary expertise;
- 4. Science disciplines vary in needs (large effort goes into developing technology for specific application);
- 5. Fundamental problems highly depend on environmental conditions that vary with location, depth, time of day, season, etc. (e.g., sea floor –vs- water column vs- near sea surface);
- 6. Many marine science needs require development of tools, NOT always challenging problems that intrigue CV researchers;
- 7. Scaaaaaaaaaarcity of Funding sources
 - -- NSF Marine Sciences: Do not typically fund CV work
 - -- NSF CISE Program: Terrestrial domain priorities
 - -- ONR, NOAA, USCG: Very limited funding for basic research, only for specific applications (mine detection, instrumentation, data collection, HLS) and short timelines
 - -- Industrial support: Non-existent (small market, high develop. costs)
- 8. Very limited job opportunities: Discouraging to Graduate students

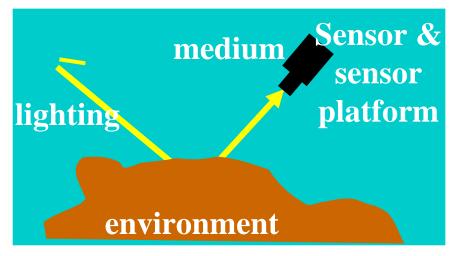


1. Medium

- Poor signal-to-noise ratio (Turbidity, absorption and scatter)
- Limited visibility/range: Can only see
 - short distance from camera
 - (small) parts of objects/scene in each image
- **Need to navigate and collect more data** to image the same region of the scene compared to terrestrial applications: more image registration, drift error, etc.
- Loss/ Distortion of color cues (wavelength-dependent attenuation)

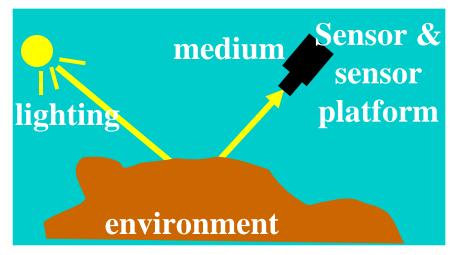


- 2. Environment: Complex & variable (unknown, unstructured, cluttered): Vision system works
- in one place but fails 50 meters further away;
- in the morning but not a few hours later;
- in deep but not shallow waters;
- one depth but not other (e.g., in the deep, but not shallow waters).



3. Illumination

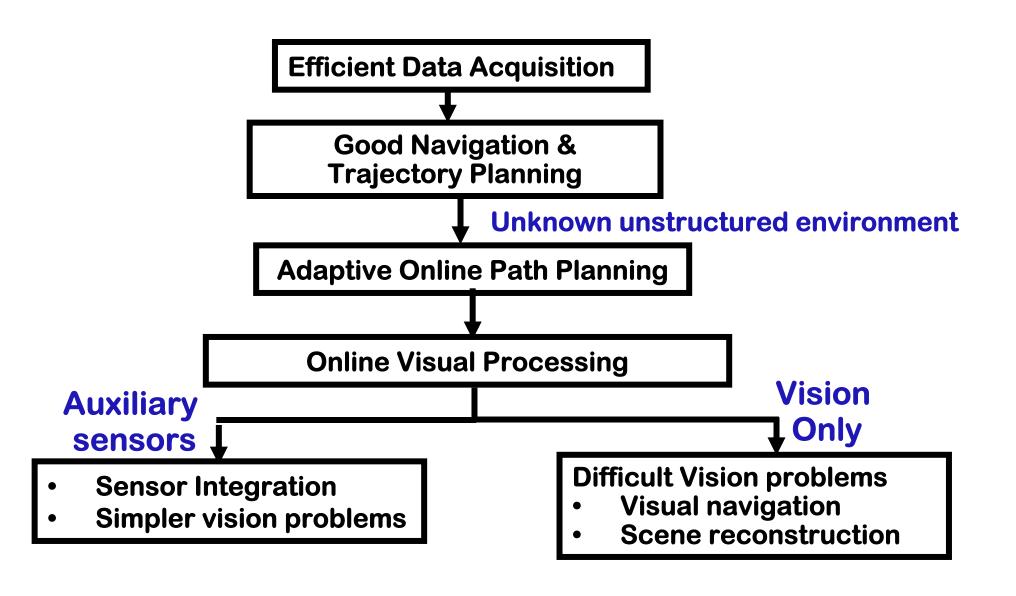
- Artificial lighting (deep sea)
- Complex cast shadows (deep and shallow waters)



4. Poor positioning:

- 6 d.o.f motion in response to wave and current disturbances
- No GPS; drift from various (inexpensive) sensors
- Easier to carry out target-based rather than absolute positioning
- Auxiliary positioning sensors are very useful, but integration with camera(s) requires accurate calibration
- Much easier to carry our offline tasks, e.g. mosaicing, object modeling (potentially with manual guidance)

Operating in Unstructured Scattering Environment With Short-Range Visibility



Real-time vision-based capabilities are critical for effective data acquisition, path planning & execution, target tracking & mapping:

- Optical Station Keeping (1989 1996)
 Maintaining position under current disturbances
- Video Mosaicing (1997)
 Navigating on a 2-D map
- Automatic Navigation (1997 present)
 Video-based ROV control
- 3D Reconstruction (2002 present)
 Benthic object mapping
 Target ID in search and inspection



Automatic Optical Station Keeping (1989-1996)



First ever ocean trial on commercial ROV Miami Beach, FL (June 1999)



Manual ROV Control

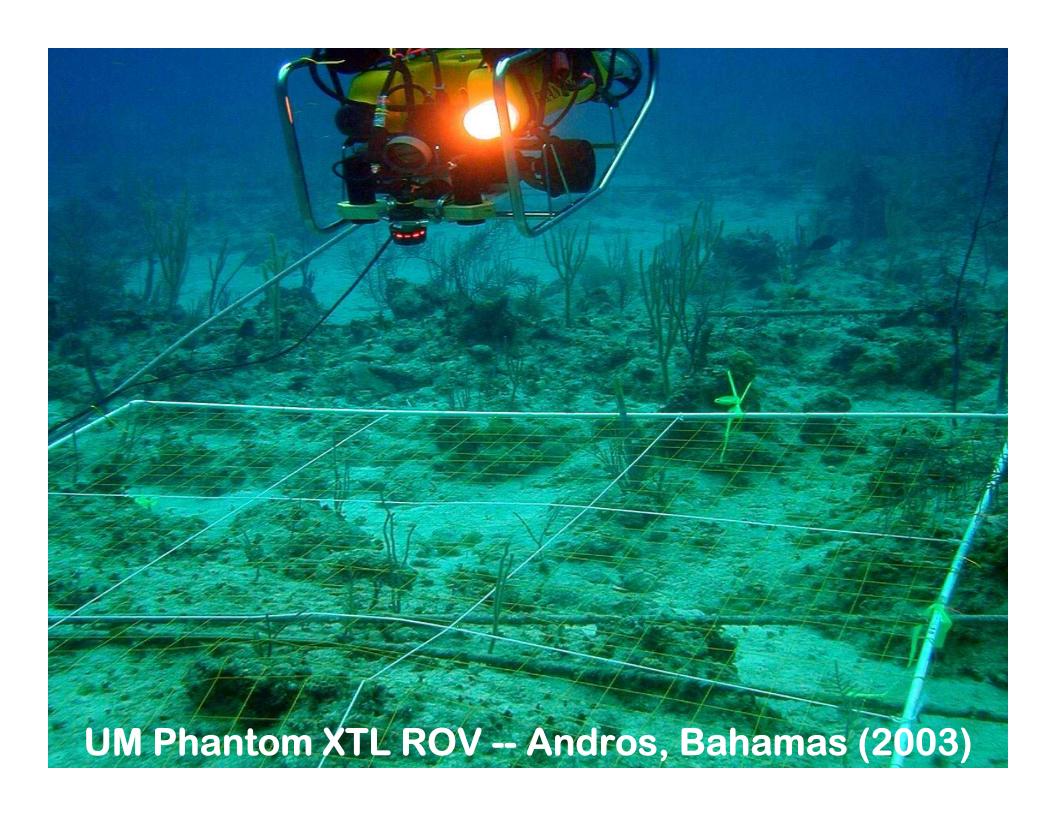
Automatic ROV Control





Example of Technological Developments under Favorable Visibility Conditions

Application of ROV Video Technology to complement
Coral Reef Resource Mapping and Monitoring

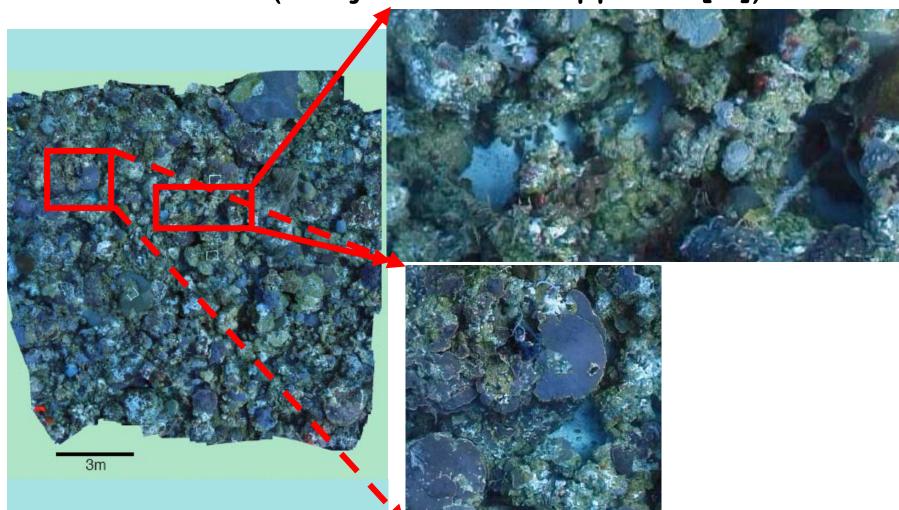




Application of 2D Mosaicing Deep Water Reef Community



Site: Dry Tortugas, FL (stony Montastrea spp at 30[m])



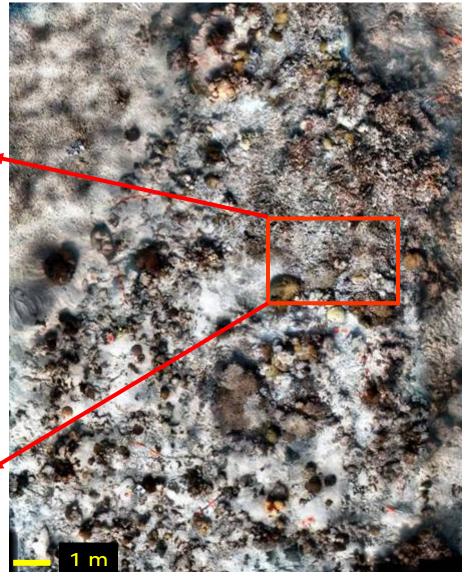


SERDP Application of 2D Mosaicing **Coral Bleaching**



Site: Andros, Bahamas December 2005)





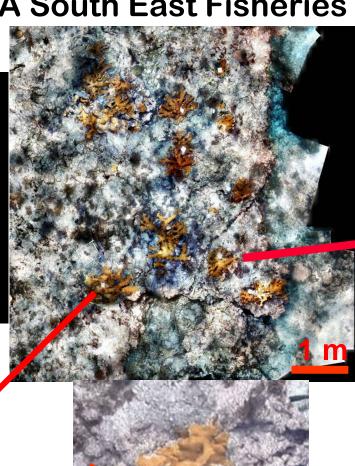


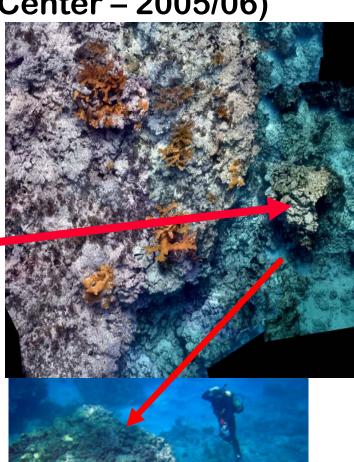
Application of 2D Mosaicing Hurricane Damage Assessment



(Site: NOAA South East Fisheries Center – 2005/06)

Left (May'05).
Right (Feb'06)
after Hurricanes
Dennis, Katrina,
Rita, and Wilma
passed over the
area.









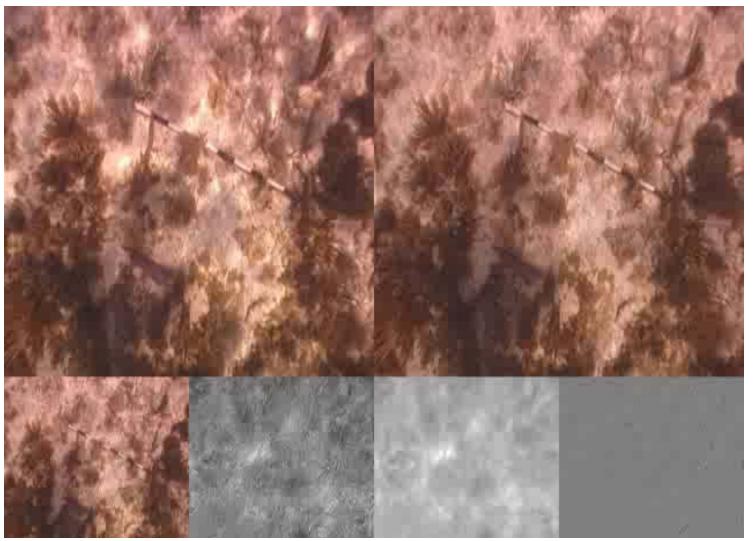




Strategic Environmental Research Application of 2D Mosaicing Ship Grounding Restoration



Site: Florida Keys



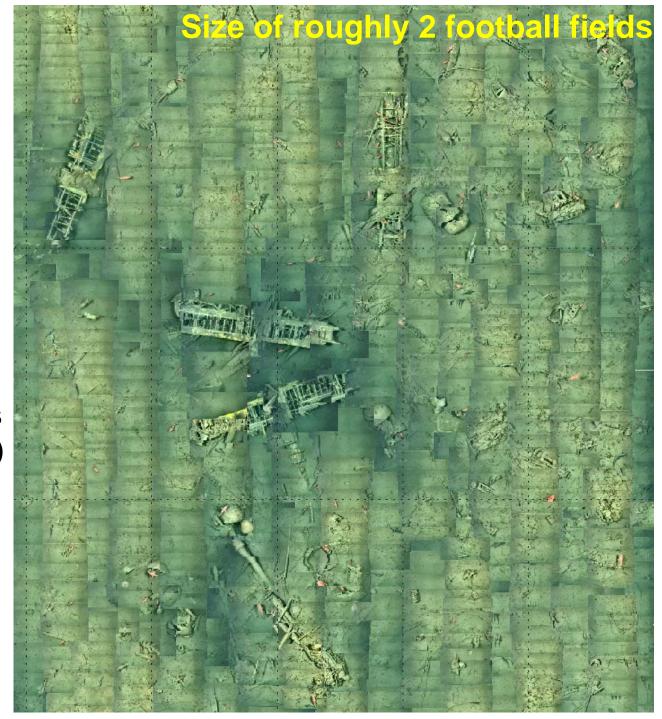


Archeology & wreck sites:
Navy ship USS
Macon
(Stanford/MBARI)

Positioning by

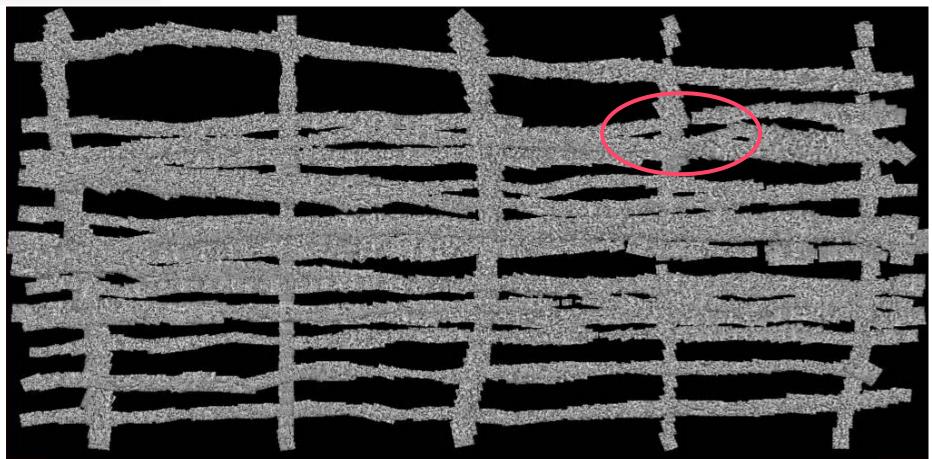
- Doppler Velocity Logger (DVL)
- Orientation sensors (pitch, roll, compass)
- Altitude sensor

Courtesy of Prof. Steve Rock, Stanford Univ.





Marine Geology



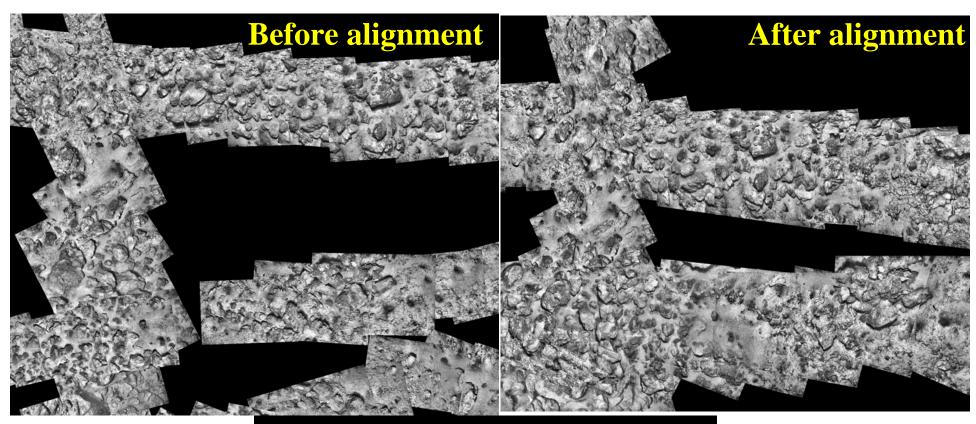
Mosaic based on LBL positioning and compass information

Positioning: Long Baseline Acoustic (LBL) and compass

Courtesy of: Prof. Rafael Garcia



Marine Geology



Alignment by image registration

Positioning: Long Baseline Acoustic (LBL) and compass

Courtesy of: Prof. Rafael Garcia





Issues:

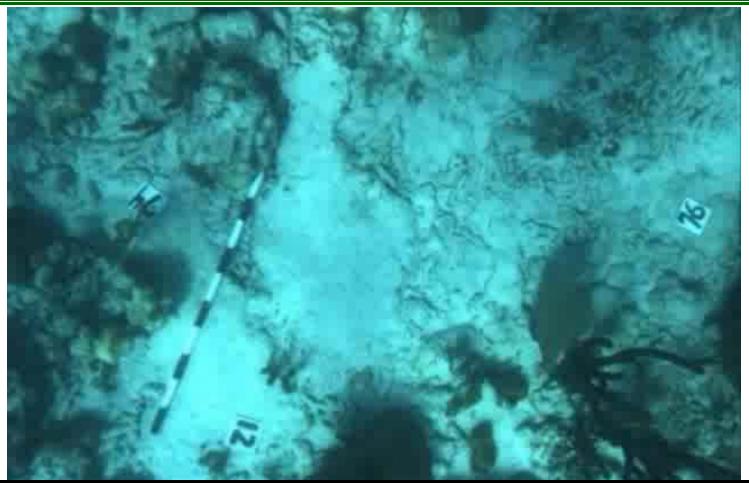
- 3-D mapping and navigation within relatively large high-relief areas;
- Processing large data effectively while dealing with data association problem (e.g., SLAM type formulations).

Approach

- Surface representation with minimum number of features while maintaining model accuracy;
- Identification of geometrically significance features that can be robustly tracked;
- Effective integration of optical-flow and feature-based approaches for dense surface reconstruction and camera trajectory estimation.







Data: Bahamas

500 frames @ 512x384 resolution with total of 400 features @ 29.5 fr lifespan used for surface reconstruction





Data: Bahamas

Reconstructed trajectory and piecewise planar terrain from 500 frames @ 512x384 resolution with total of 400 features @ 29.5 fr lifespan used for surface reconstruction









Data: Bahamas

Texture-mapped reconstructed terrain from 500 frames @ 512x384 resolution with total of 400 features @ 29.5 fr lifespan used for surface reconstruction



Sun flicker and cast shadow can be quite significant

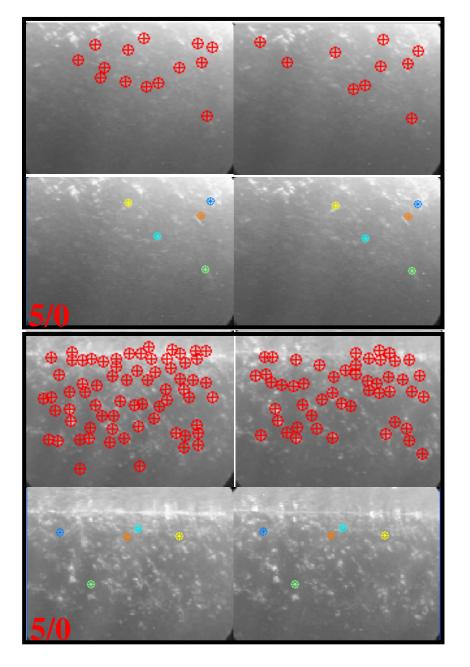


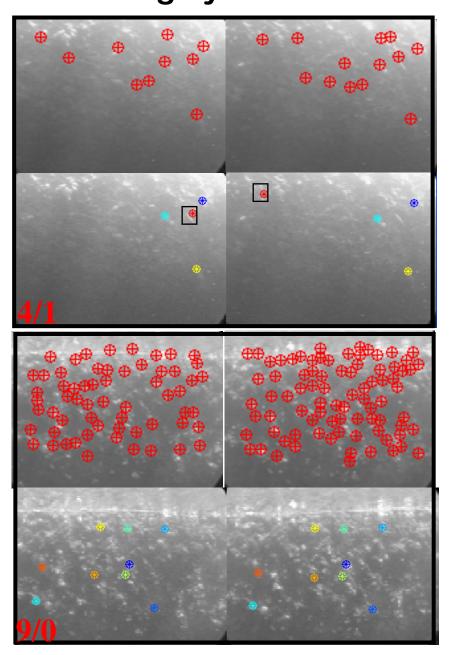
Shallow Water – Sea floor (Florida Keys)

Sea Surface – Port Structure (Panama City Naval Center)

Need a fundamental approach to deal with the problem, particularly for real-time operations

Sample Frames from Panama City DataFeature Detection and Matching by SIFT





Stereovision?

Cast Shadows: Become signal not noise source;

Improved accuracy: Redundant spatio-temporal data;

Instantaneous Target-Based Positioning

Improved robustness (for 6 d.o.f motion):

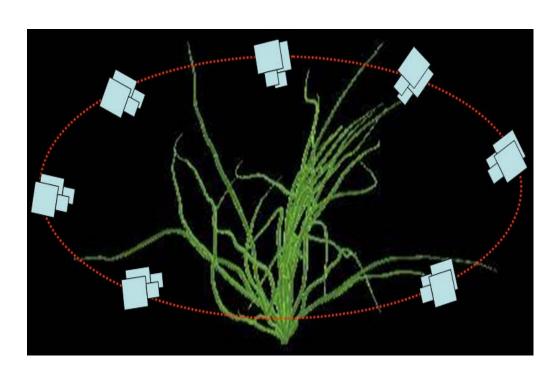
Require less features and less views; Avoid degenerate cases;

Improved efficiency: Closed-form solution of absolute orientation in place of methods based on fundamental matrix or trifocal tensor;

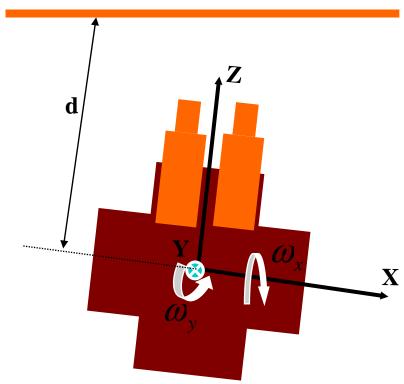
No scale drift (inherent in monocular motion vision);

More effective in dealing with dynamic feature sets (feature appearance/disappearance)

Stereovision: Has proven crucial in a range of underwater problems, handled by monocular vision in terrestrial domain

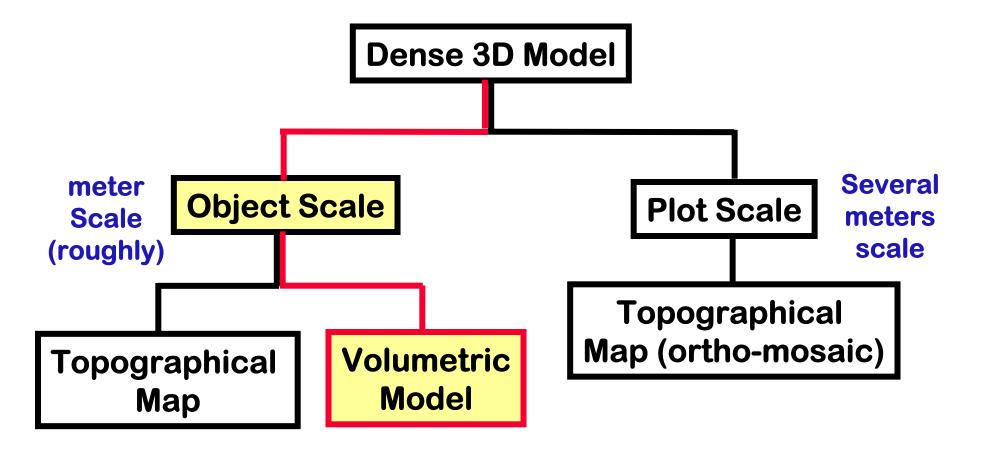


Example 1: Volumetric Modeling



Example 2: Visual Odometry

Example 1: Dense 3-D Scene Reconstruction



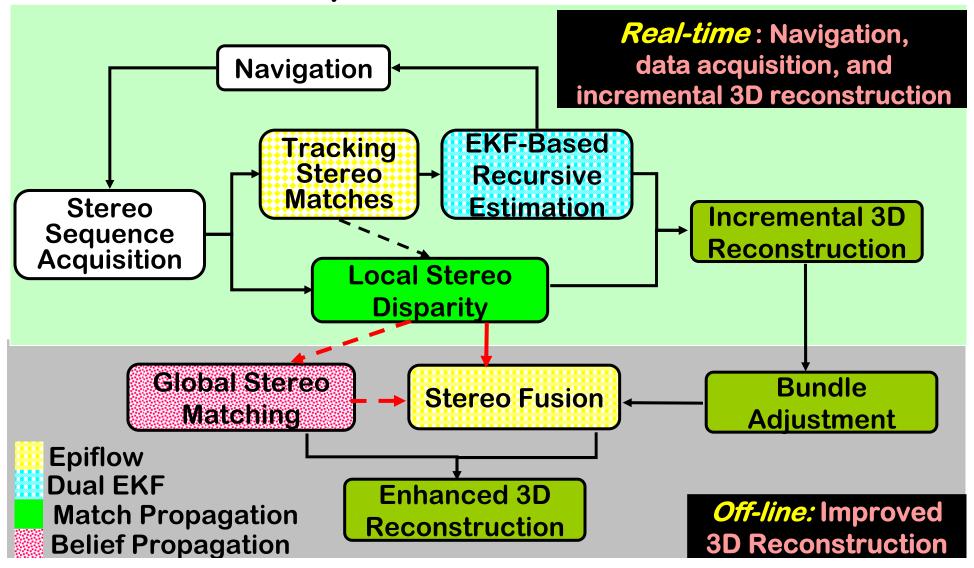


Example 1:

Miami

(Object Volumetric Modeling)

Computational Framework





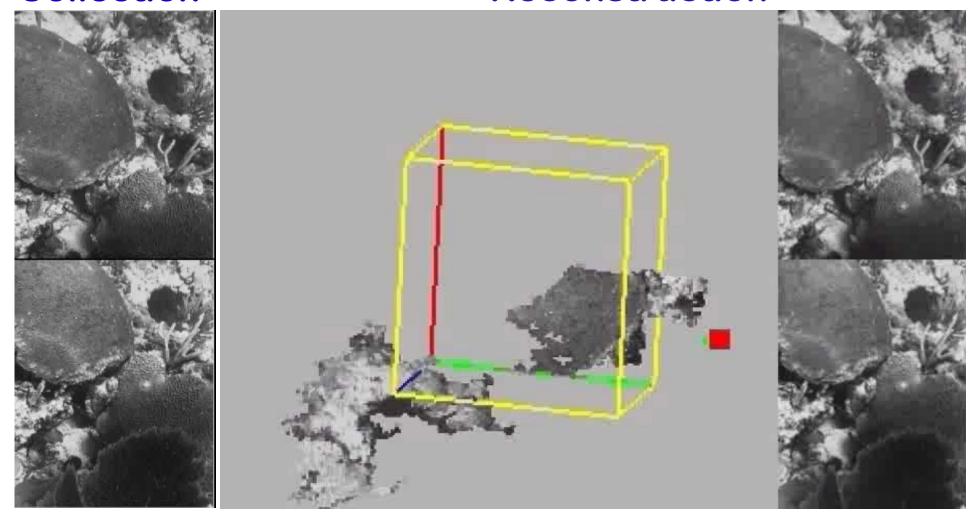
Example 1:

(Object Volumetric Modeling)



Data Collection

Trajectory & Object Reconstruction

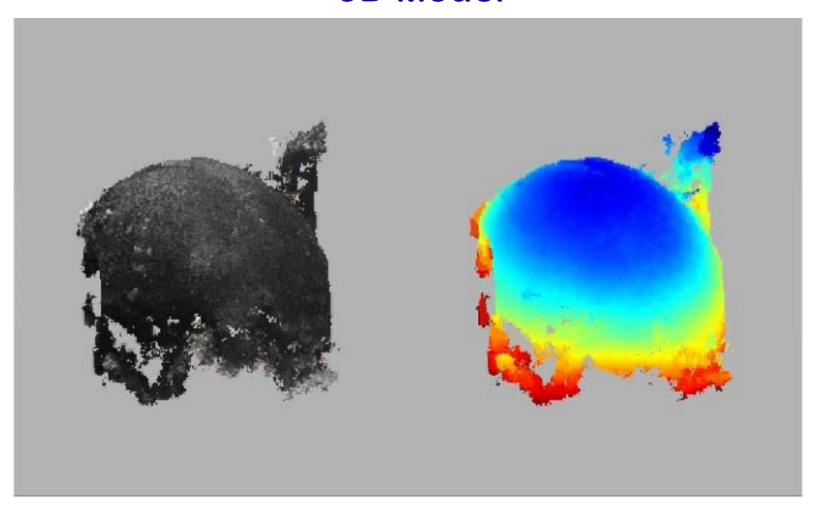






(Object Volumetric Modeling)

3D Model



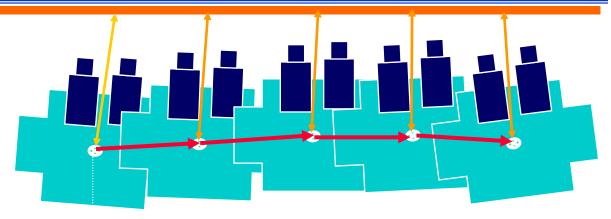




Example 2:

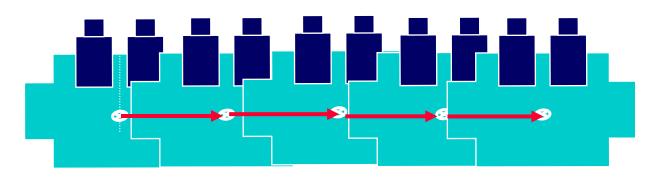


Shallow water Inspection and Target Mapping (Ship Hull Inspection 2003)



Actual vehicle motion

Video stabilization: Remove effects from roll motion, heading & scale change.



Computer generated video for mosaicing with 3 d.o.f in motion

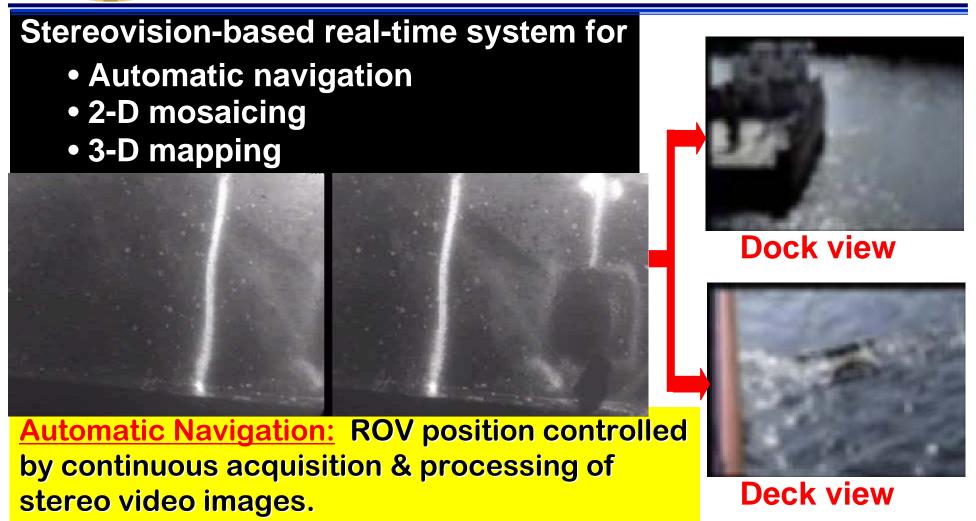








Ocean Demo; July 2003
FAU SeaTech Center, Dania Beach, FL





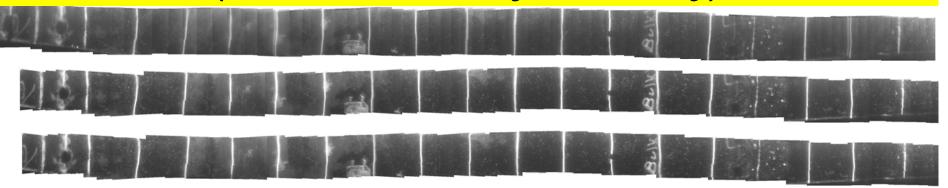
Shallow water Inspection and Target Mapping





Ocean Demo; July 2003 FAU SeaTech Center, Dania Beach, FL

Real-time Mosaics from left-camera video (3 different runs verify consistency)

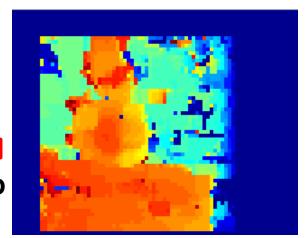




3-D Reconstruction

Sample left image and Color-coded 3-D map; (Red: close, Blue: far)

Hull, step & spherical ball placed on diving boat step detectable in 3-D map.





Application of Motion Video Processing to Sonar Imagery



- Unfortunately, visibility range limits optical imaging;
- Sonar penetrate through turbidity, and high-frequency sonar imaging systems produce images with high details
- Can we develop same vision-based capabilities with 2-D sonar cameras (DIDSON, BlueView)? How effectively?
 - -- Constructing 2-D map of targets (Photo-mosaicing);
 - -- Determining motion of sensor platform relative to target (Target-based positioning);
 - -- Determining 3-D position of points on target surfaces (3-D reconstruction from stereo imagery).

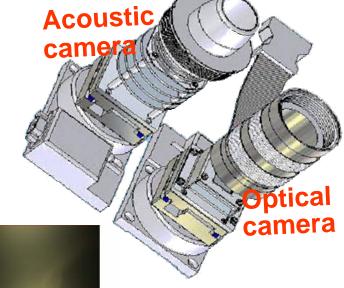
Answer: Need to integrate with auxiliary positioning systems



Opti-Acoustic Stereo Imaging; New Paradigm for Underwater 3-D Target Mapping



New Paradigm: Can we achieve improved performance by integrating information from optical and sonar cameras?



Good OpticalVisibility

Both cameras provide valuable information



Poor Optical Visibility

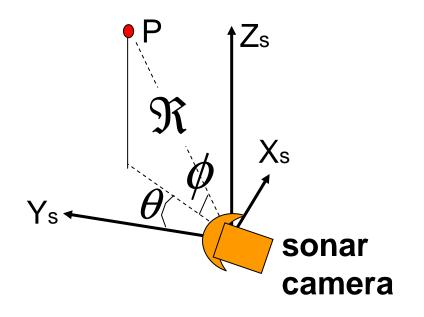
Rely solely on sonar camera

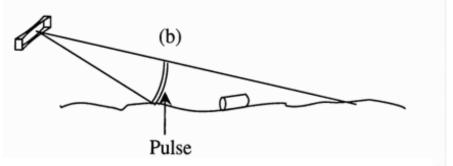
Currently exploring sensor fusion in the context of Opti-Acoustic Stereo Imaging



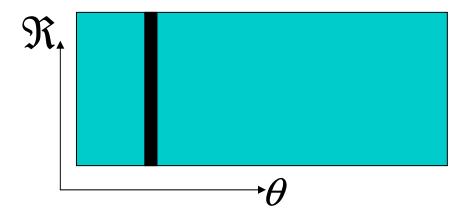


2-D Imaging Sonar





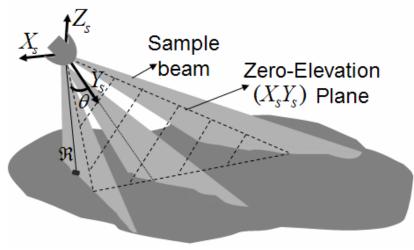
Scene "illumination" for particular (azimuth) direction



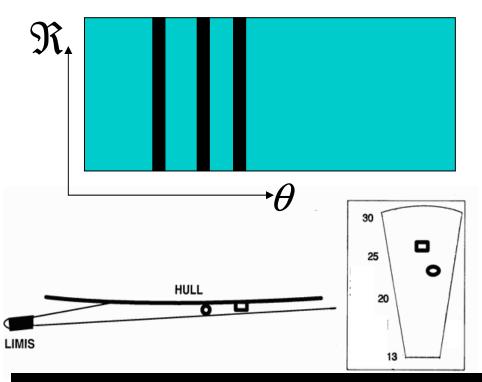




2-D Imaging Sonar



- Each array of sonar measurements correspond to target scene ranges \Re for a particular bearing (azimuth direction) θ ;
- 2-D Image $I(\Re\theta)$ is constructed from range measurements at different bearing directions.

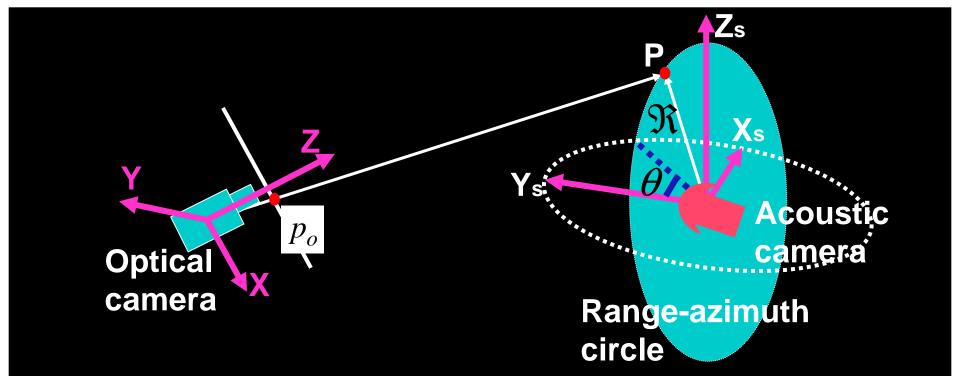


Target illumination and the image





Opti-Acoustic Stereo Imaging



Opti-acoustic stereo pair: Projections of a scene feature onto optical and sonar views in the form of opti-acoustic correspondences:

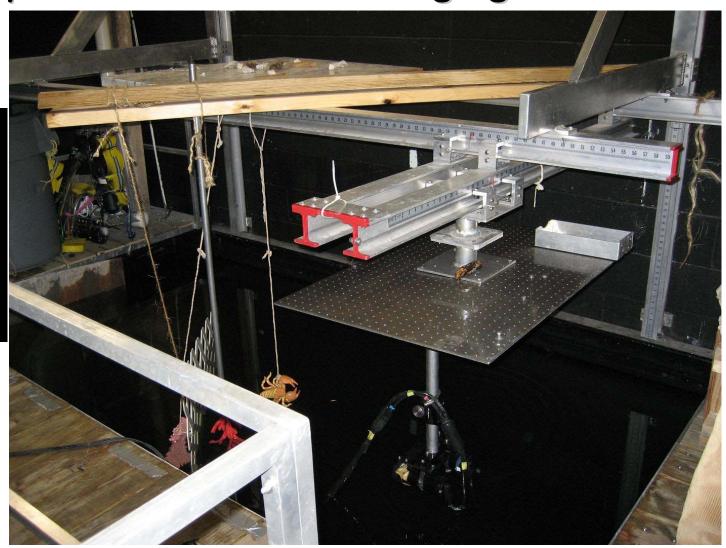
$$\begin{cases} p_o = (x, y) \\ p_s = (\Re, \theta) \end{cases} \rightarrow \begin{cases} p_o = (x, y) \\ p_s = (\Re \sin \theta, \Re \cos \theta) \end{cases}$$





Opti-Acoustic Stereo Imaging

Water tank set for controlled underwater experiments

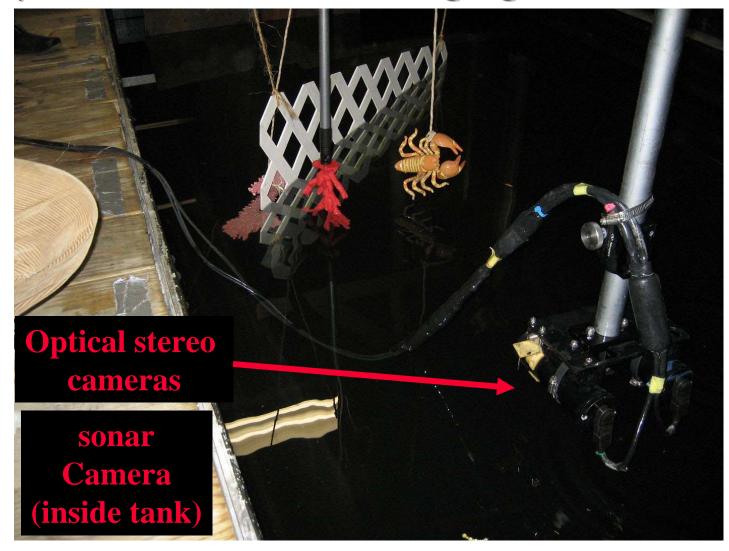






Opti-Acoustic Stereo Imaging

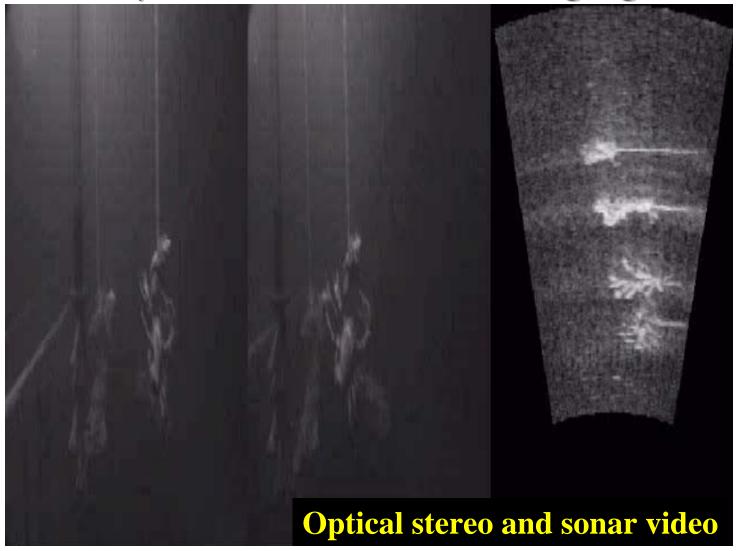
Water tank set up and targets







Opti-Acoustic Stereo Imaging







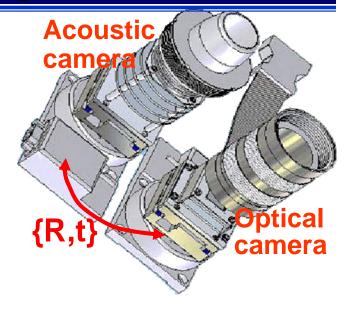
Epipolar Geometry of Opti-Acoustic Stereo Imaging

• For *an acoustic-image point s*, what is epipolar contour in optical image?

$$\mathbf{p}^{\mathrm{T}}U(s)\mathbf{p}=0;$$







$$U(\mathfrak{R}, \theta) = \left(\frac{(\tan \theta t_y - t_x)^2}{t \cdot t - \mathfrak{R}^2}\right) I + (r_1 - \tan \theta r_2)(r_1 - \tan \theta r_2)^T +$$

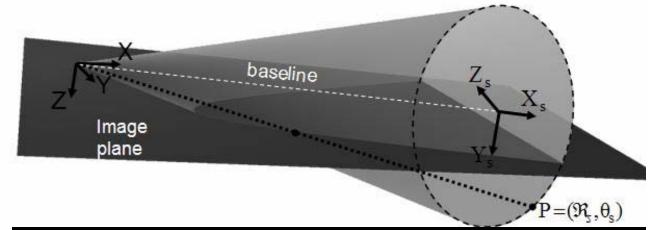
$$\left(\frac{\tan\theta\,t_y-t_x}{t.\,t-\Re^2}\right)\left((r_1-\tan\theta\,r_2)t^TR+R^Tt(r_1-\tan\theta\,r_2)^T\right)$$

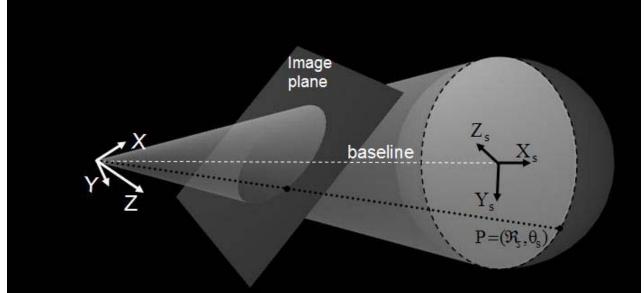




Opti-Acoustic Stereo Epipolar Geometry

Epipolar contours in optical image are quadrics: Either hyperbolas or ellipses.

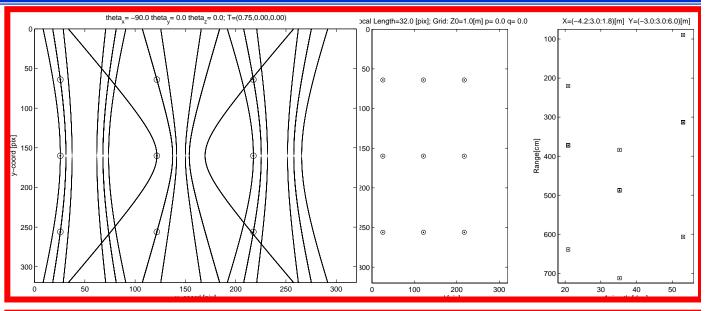


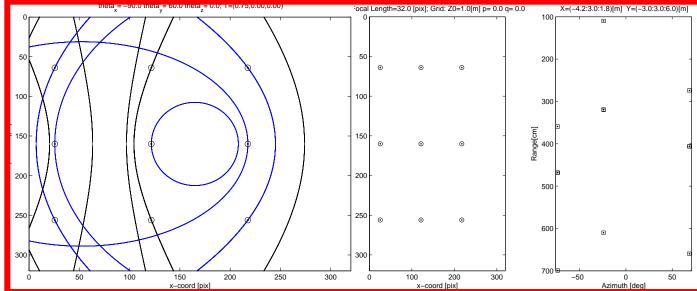






Epipolar Geometry of Opti-Acoustic Stereo Imaging



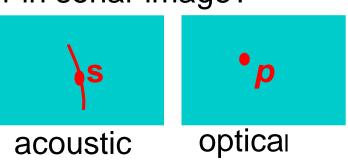


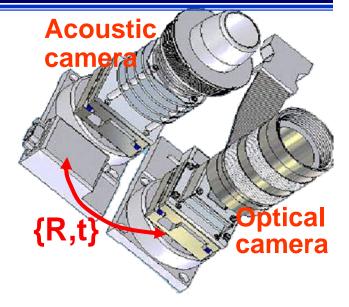




Epipolar Geometry of Opti-Acoustic Stereo Imaging

• For every **optical-image point p**, what is epipolar contour in sonar image?





$$\mathfrak{R} = \sqrt{\frac{N(\theta)}{D(\theta)}}$$

$$N(\theta) = (u_{31}\sigma_2 - u_{32}\sigma_1)^2 + ((u_{12}\sigma_1 - u_{11}\sigma_2)\sin\theta + (u_{22}\sigma_1 - u_{21}\sigma_2)\cos\theta)^2$$

$$D(\theta) = ((u_{31}u_{12} - u_{32}u_{11})\sin\theta + (u_{31}u_{22} - u_{32}u_{21})\cos\theta)^2$$

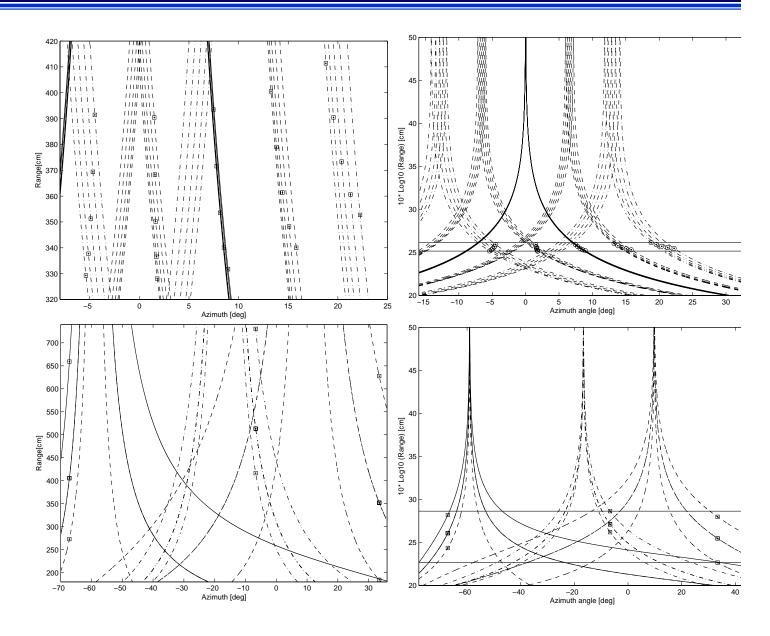
$$u_{k1} = r_{3k}y - r_{2k} \quad u_{k2} = r_{3k}x - r_{1k}$$

$$\sigma_i = t_x u_{1i} + t_y u_{2i} + t_z u_{3i}$$





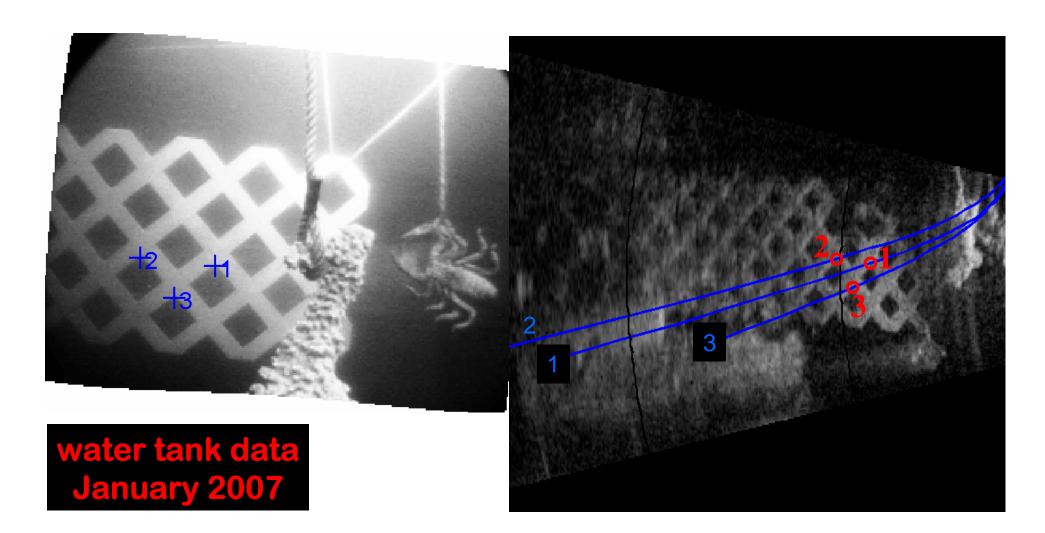
Epipolar
Geometry
of
OptiAcoustic
Stereo
Imaging







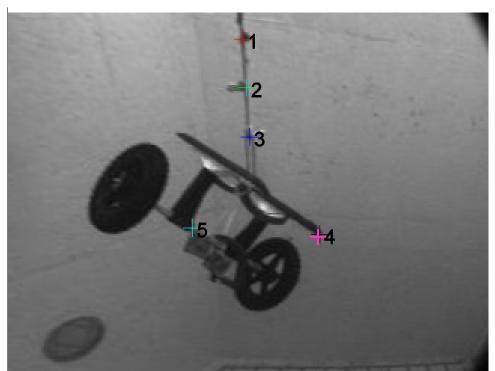
Opti-Acoustic Stereo Epipolar Geometry



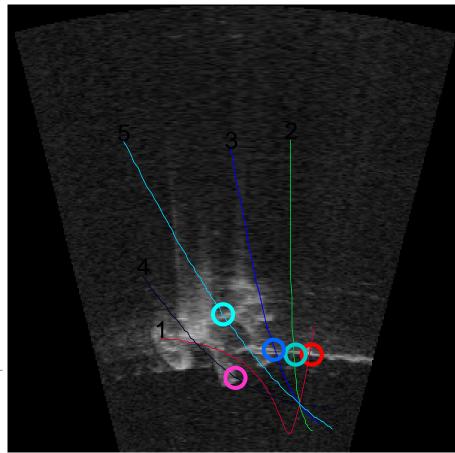




Opti-Acoustic Stereo Epipolar Geometry



Teledyne Benthos pool facility (N. Falmouth, MA) November'06







Epipolar Geometry of Opti-Acoustic Stereo Imaging

Visibility Constraint:

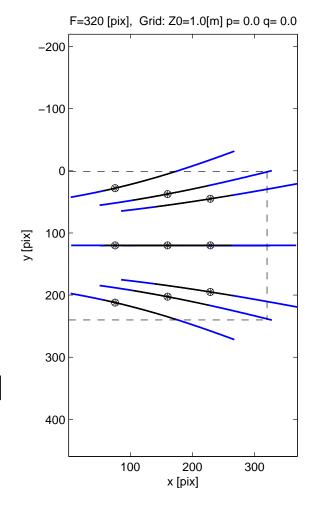
Each acoustic beam insonifies the scene within small sector in elevation direction:

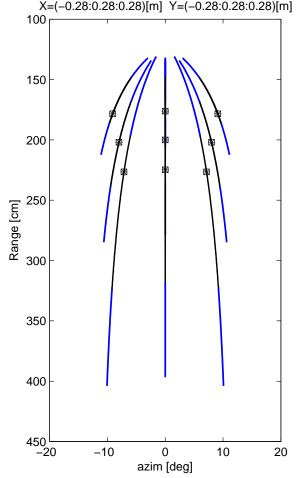
DIDSON &

BlueView: $\pm 7 [deg]$

Black: elevation $\pm 7 [deg]$

Blue: elevation ±12[deg]

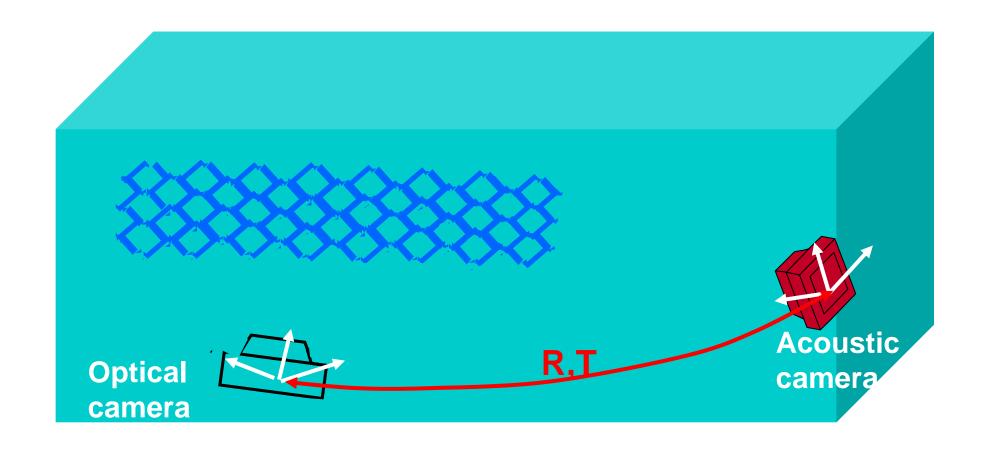








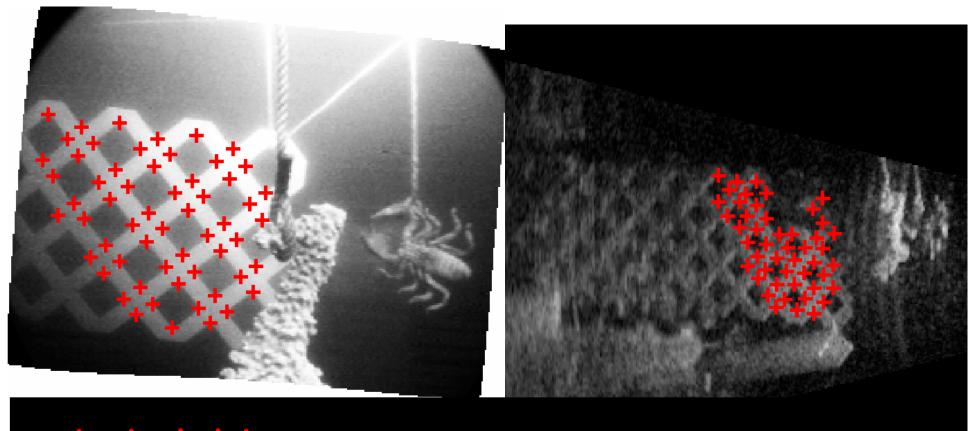
Stereo Calibration







Stereo Calibration

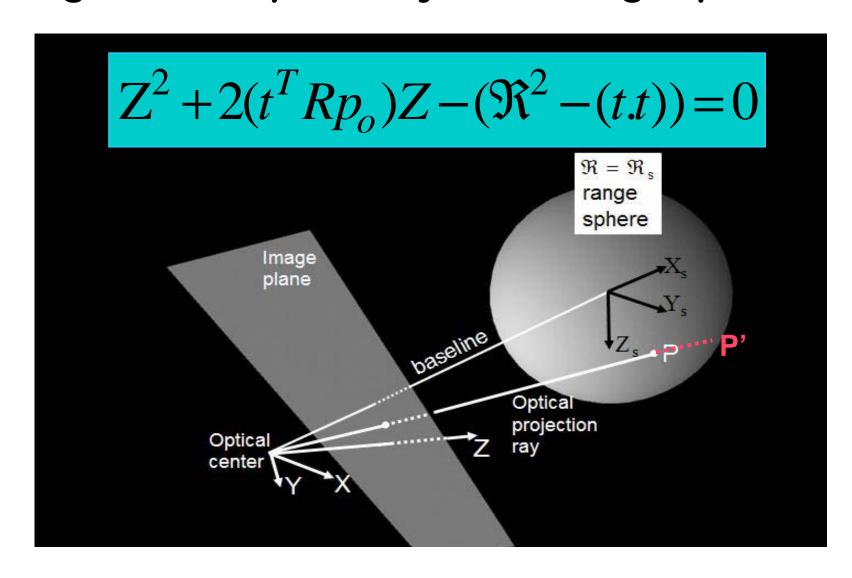


water tank data January 2007





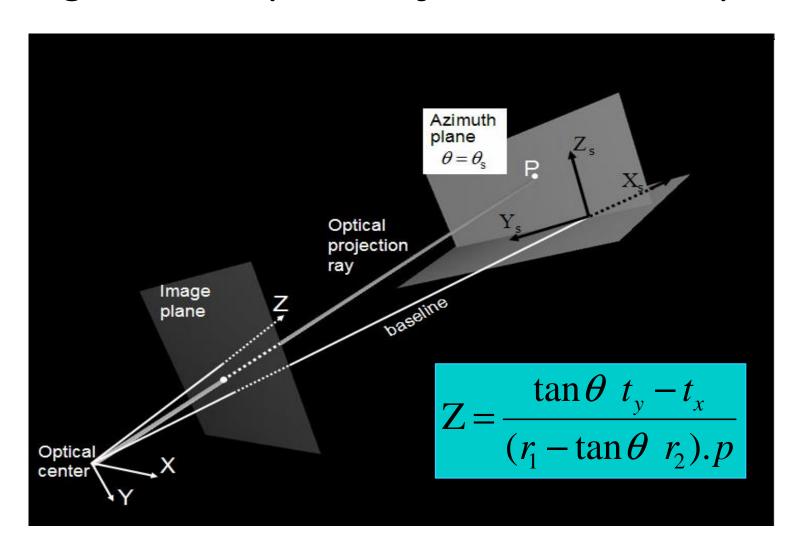
Triangulation: "optical ray" and "range sphere"







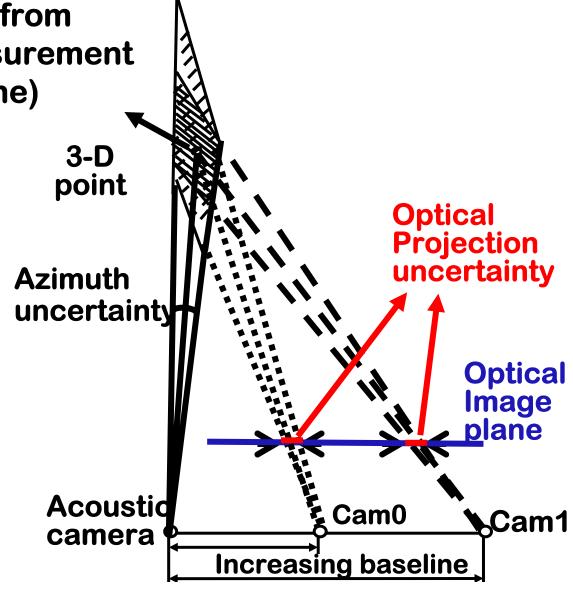
Triangulation: "optical ray" with "azimuth plane"







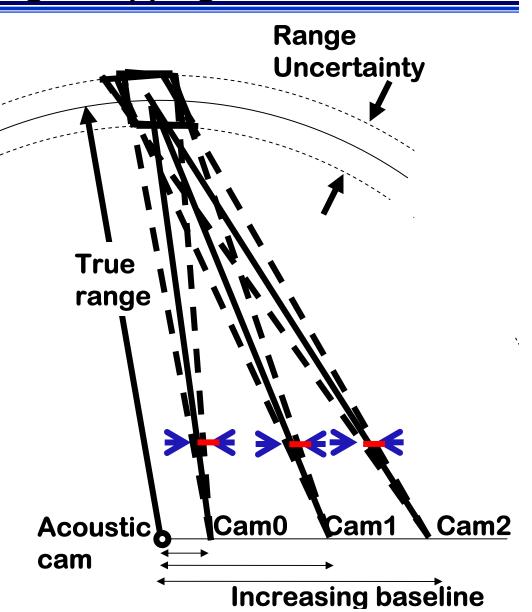
3-D reconstruction from bearing (azimuth) measurement (Impact of baseline)





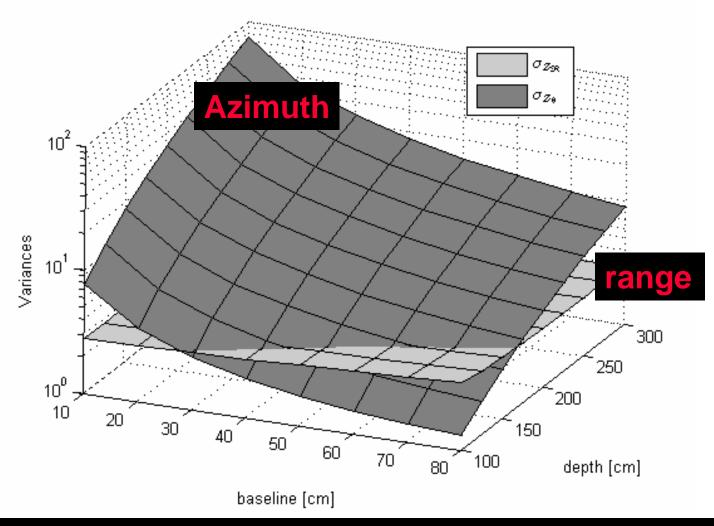


3-D reconstruction from range measurement (Impact of baseline)







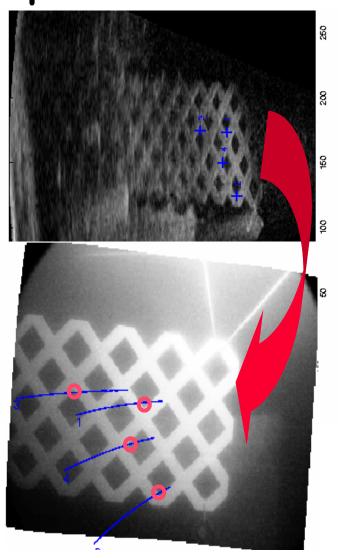


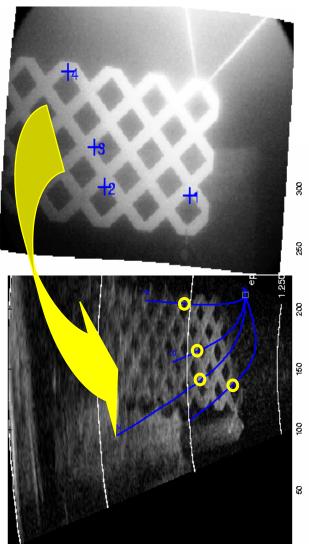
Azimuth and range measurements can be utilized more cleverly, accounting for target range and stereo baseline.





Opti-Acoustic Correspondence Problem?

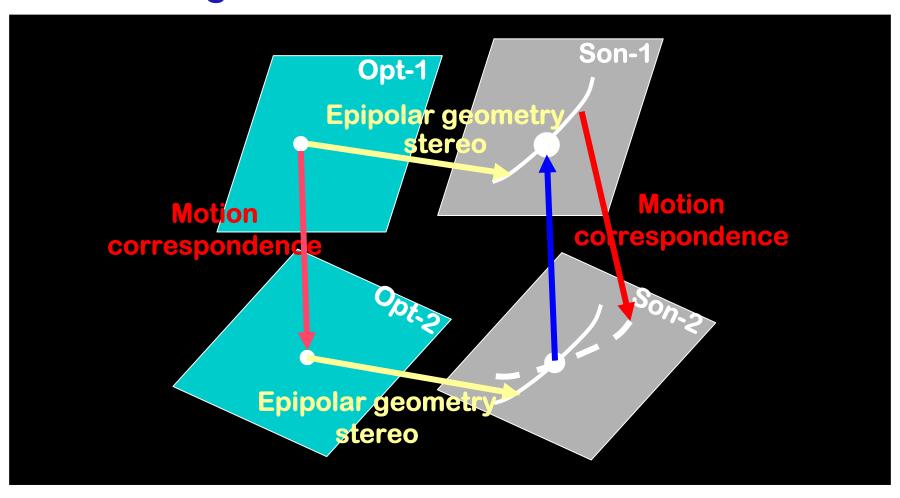








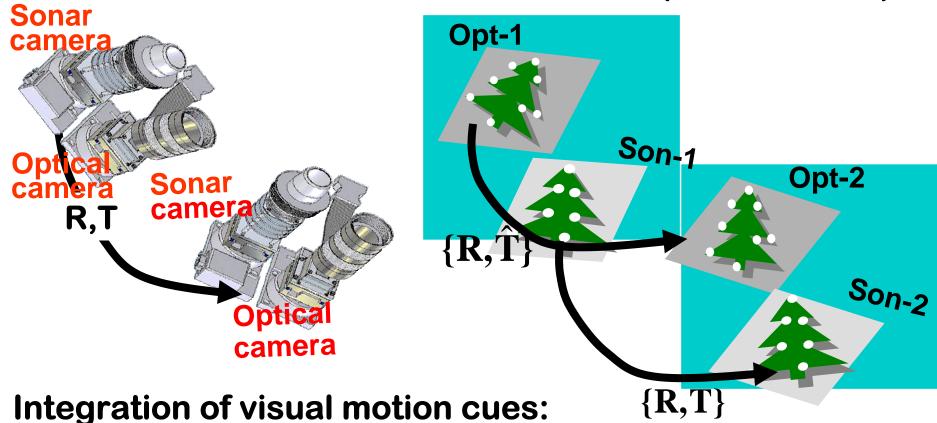
Current Approach:Integration of motion and stereo cues







3-D Camera Motion Estimation (BenCOS'07)



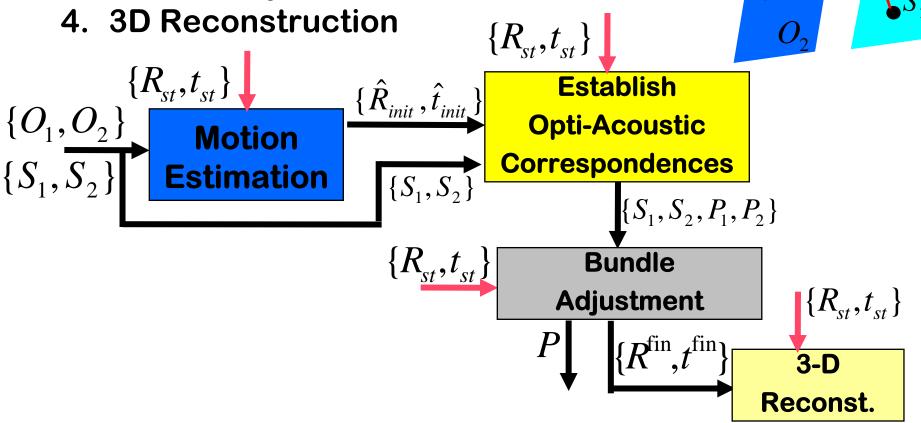
- Resolving scale factor ambiguity of monocular motion
- More robust & accurate estimation of sensor platform motion





Automated 3-D Reconstruction:

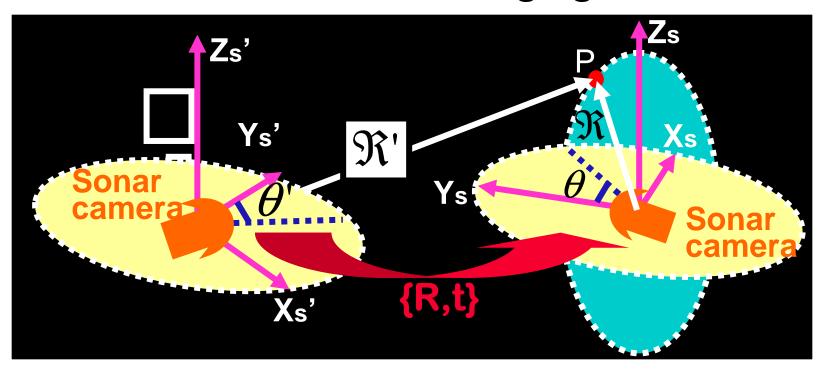
- 1. Motion Estimation
- 2. Opti-Acoustic Matching
- 3. Bundle Adjustment







Sonar Stereo Imaging

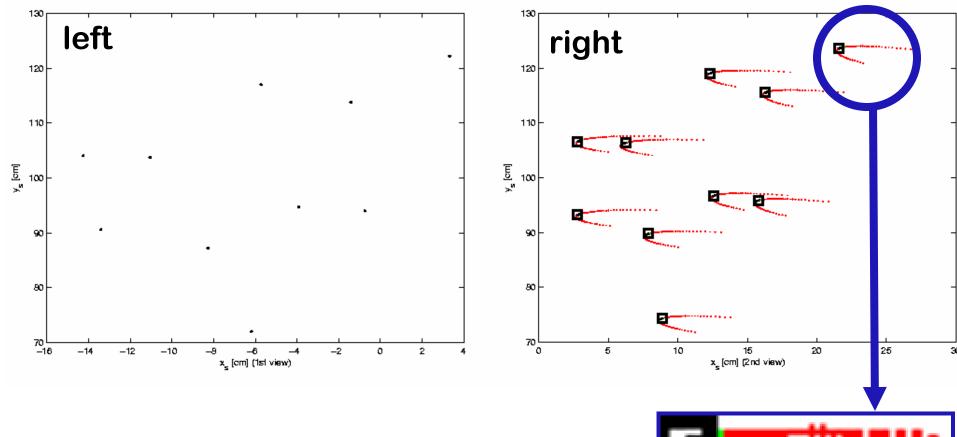


$$\begin{bmatrix} -1 & \tan \theta & 0 \\ r_{21} \tan \theta' - r_{11} & r_{22} \tan \theta' - r_{12} & r_{23} \tan \theta' - r_{13} \\ \mathbf{t} \cdot \mathbf{r_1} & \mathbf{t} \cdot \mathbf{r_2} & \mathbf{t} \cdot \mathbf{r_3} \end{bmatrix} P = \begin{bmatrix} 0 \\ t_x - t_y \tan \theta' \\ \Re'^2 - \Re^2 - \|\mathbf{t}\|^2 \end{bmatrix} \text{ subject to } \|P\| = \Re$$

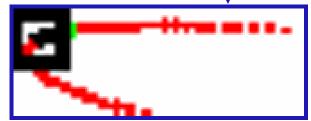




Sonar Stereo Imaging (Epipolar Geometry)



Black: Points visible in both views Green: Visible in first but not 2nd view







Concluding Remarks

- Underwater vision research is challenging, but also fun and rewarding (lot of diving in great locations);
- Theoretical developments of vision in scattering media will play a critical role in underwater vision;
- Underwater vision can truly benefit from contributions from a larger body of CV community;
- Perhaps, we can establish assessment criteria of 2-D and 3-D reconstruction methods, not merely for performance on existing data bases but also on underwater imagery;
- To promote this, UM has valuable resources and facilities for underwater research to be made available to researchers who are interested in collaborative activity.





Resources

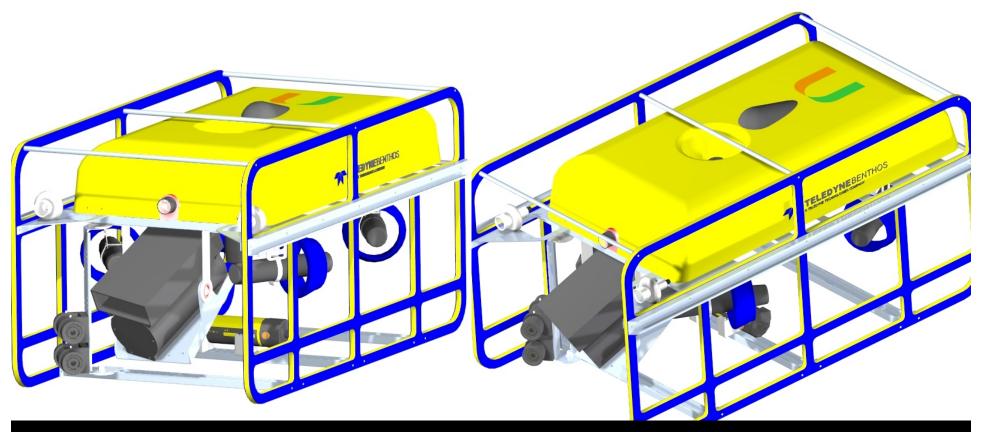
Phantom XTL with mega-pix digital stereo cameras, analog CCD camera, compass and pitch/roll sensors







Resources



Benthos Custom-designed Stingray ROV with stereo Camera sonar Camera and 6 d.o.f. acoustic positioning





Resources



Of course, large experimental tank in our backyard

The End