

# Multiplatform Autonomous Adaptive Sampling

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Special Thanks to Michel Rixen, Pierre Lermusiaux, Ed Gough

# View from the Top

Transfer and reduction of uncertainty from ocean sampling –  
through acoustic propagation

- “Optimal is in the eye of the beholder”

The Naval uncertainty problem is:

- Multi-Disciplinary (not just inter..)

- Ocean Dynamics (Physics)
- Acoustics (Physics)
- Signal Processing (EE)
- Military Application

- Nonlinear (T/S-P-TL-DCLT)

- Complex

- Fronts/Eddies/Bathymetry/ducts/waves

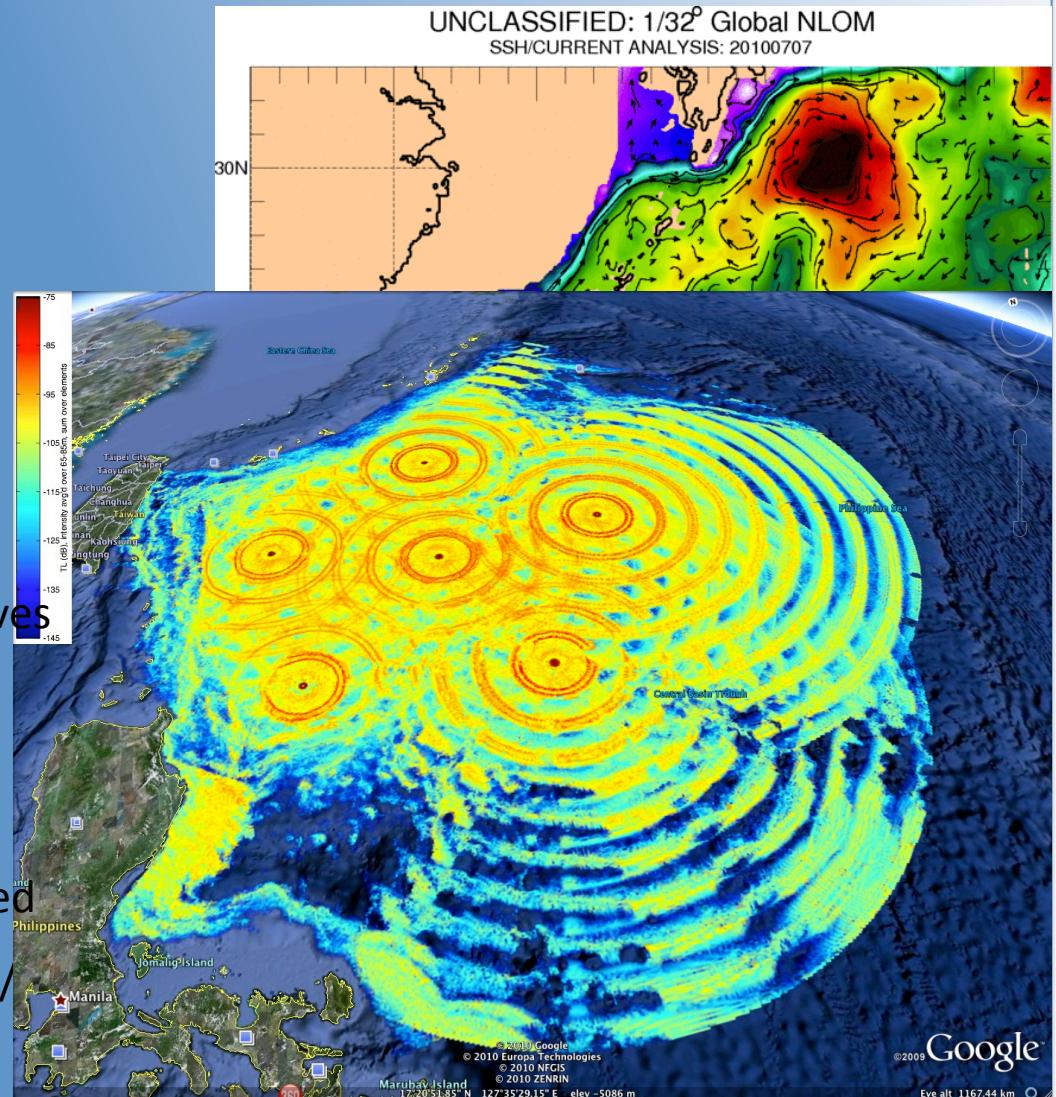
- Multiple Scales

- Large scale circulation

- Small scale variability – not well understood or surveyed

- Perpetually undersampled and understaffed

Inherently involves motion (target/receiver/ocean) and 3D environment



# The EMPath Algorithm

## (Environmental Measurements Path planner)

### The EMPATH Algorithm History

2004 SBIR out of SPAWAR for OASIS (MATLAB, GA)

Experiment: SW06

2007 Transition to NAVOCEANO/NRL via LBSFI (C, GALib)

Experiment: Valiant Shield 07

Experiment: Rimpac 08

Experiment: ANNUALEX 09

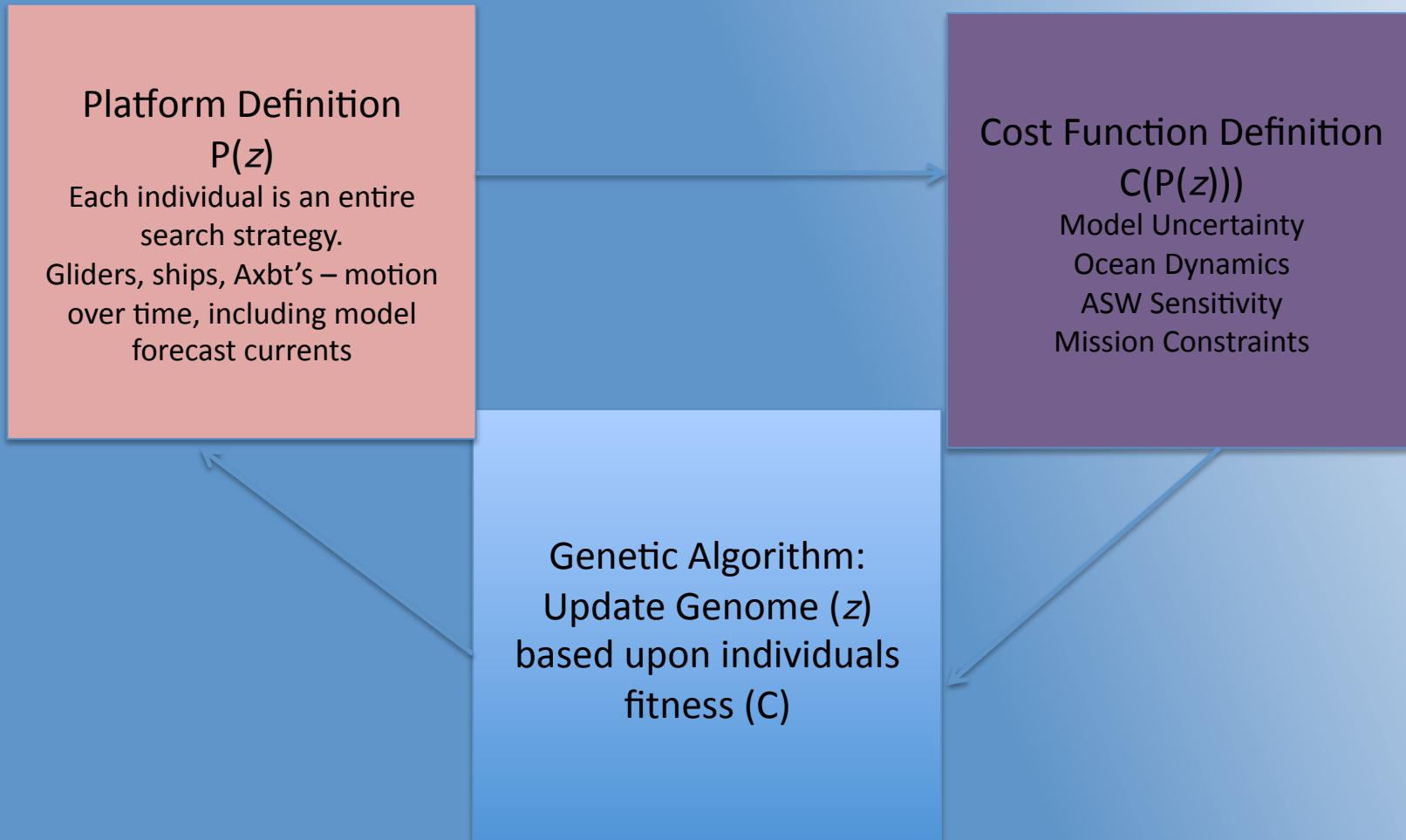
2010 OSSE and VTR with NRL (C, internal GA, GoogleEarth)

ONR QPE Experiment: Taiwan 2008, 2009

Experiment: Rep10

VTR for operational use at NAVOCEANO 2011.

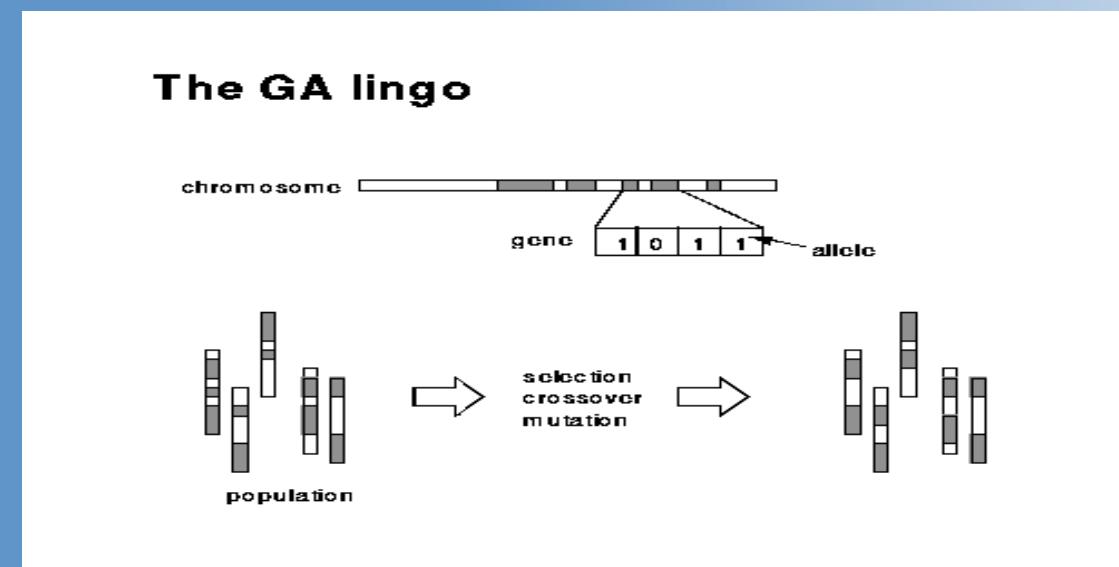
# EMPath (Environmental Measurements Path planner)



[Heaney, Lermusiaux and Duda, JFR-2007 and OM-2010]

# Genetic Algorithm Solutions

- Solution method to non-linear multi-dimensional scalar space.
- Based upon evolutionary biology –
  - Population of individuals (search planning algorithms)
  - Each search parameter is coded into a gene, the individual is the aggregated gene (chromosome)
  - Each individual's fitness is evaluated (Cost function evaluation)
  - Next generation is reproduced from fittest parents
    - Selection (of fit individuals)
    - Cross-over (of parents genes)
    - Mutation – random perturbation to escape local minima



# Initial Cost Functions

- Spatial Variability:

$$T_{spatial} = \frac{1}{N} \sum_i \int d\vec{r}_i \left( T(\vec{r}) - \bar{T}(\vec{r}) \right)^2$$

- Temporal/Ensemble Variance:

$$T_{temporal} = \frac{1}{N} \sum_i \int d\vec{r}_i \sigma_t^2 \left( T(\vec{r}, t) \right)$$

# Constituent Cost Functions

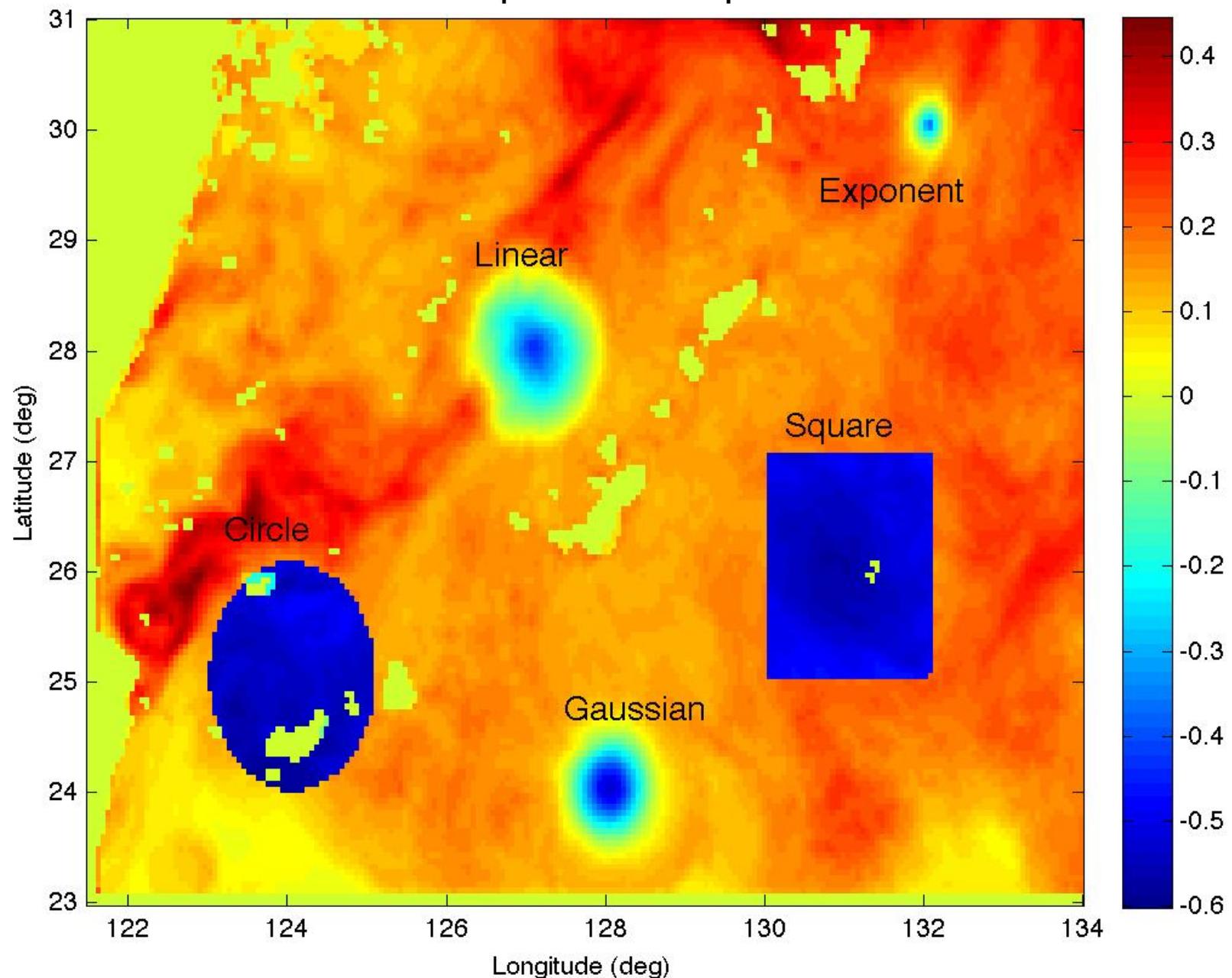
- Central Forecast (Single forecast run)
  - Temporal Variability (dynamics)
  - Spatial Variability (Gradients/fronts)
  - Acoustic Sensitivity
- Ensemble Forecasts:
  - Ensemble Spread (T/S @ depth)
  - Acoustic Computation Sensitivity
    - Area Coverage Variance
    - Below Layer Gradient
    - Mixed Layer Depth
    - ...
  - Impact of measurements on Target Area (Coelho TOFU).
- Operational Constraints
  - Operational Safety
    - Bathymetry
    - Currents
  - Collection of Vehicles
  - Waterspace Constraints... Aussie subs... ;)

# Normalized Weighted Linear Sum

$$E(\bar{r}) = \sum_i \frac{W_i C_i(\bar{r})}{\sigma(C_i)} + W_b C_b(\bar{r}) + W_{dp} C_{dp}(\bar{r})$$

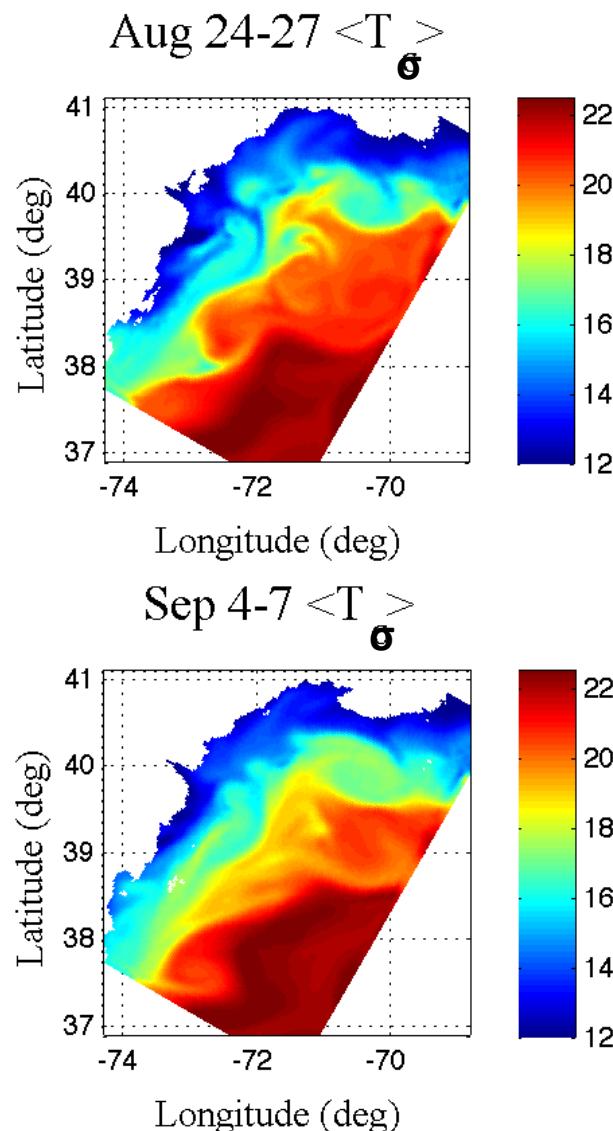
$$C_{dp} = \left\{ \left( \frac{\hat{r}}{\Delta r + \alpha \hat{r}} \right)^2 + 1 \right\} \left\{ \left( \frac{\hat{t}}{\Delta t + \alpha \hat{t}} \right)^2 + 1 \right\} - 1$$

## WaterSpace Examples

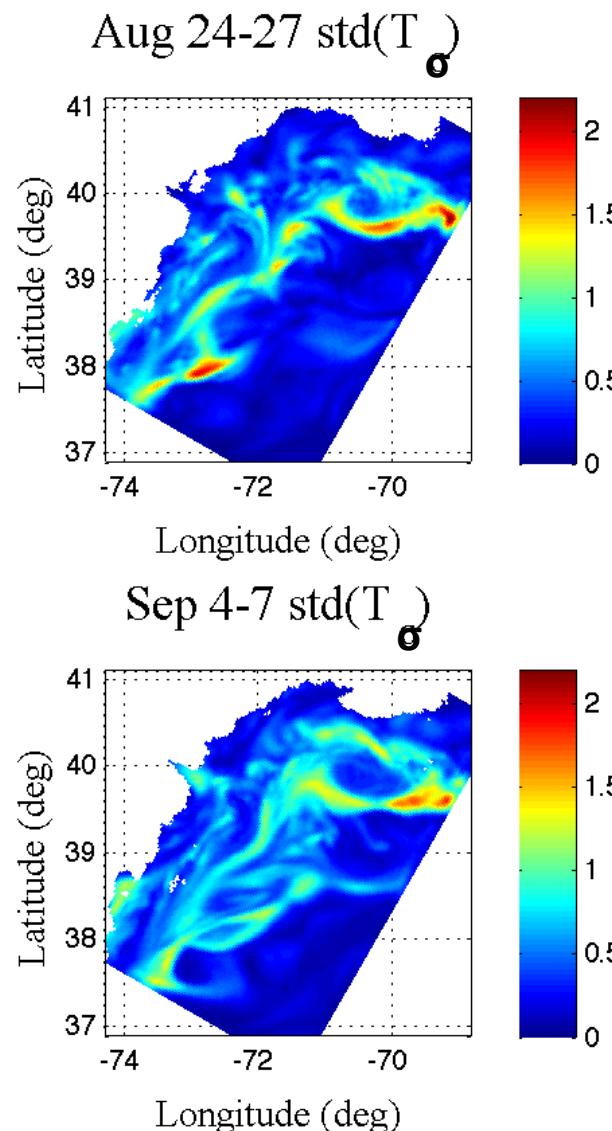


## HOPS Model Hindcasts for Shallow Water 2006 – NJ SHeLF

Before  
Tropical  
Storm  
Ernesto



After  
Tropical  
Storm  
Ernesto

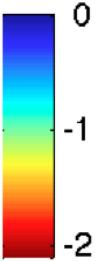
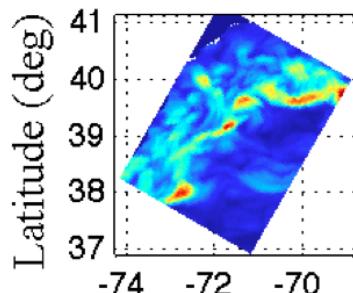


**Time-Averaged Temperature**  
( $T_s = T$  along the  $24.7 \text{ g/cm}^3$  isopycnal)

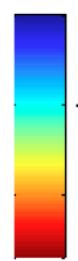
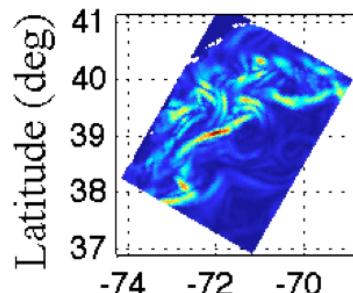
**Temperature Uncertainty**  
(Ensemble standard deviation)

## Cost Function Morphology

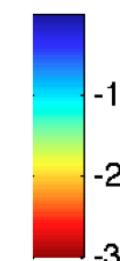
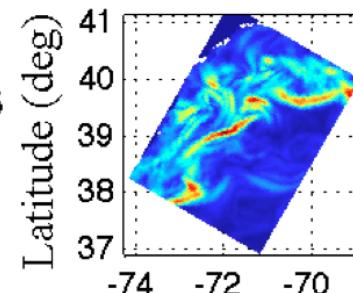
Aug 24-27 Tens



Aug 24-27 Tsig



Aug 24-27 Comb

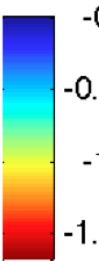
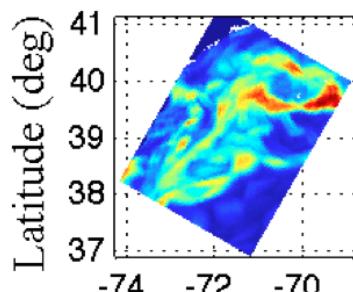


Longitude (deg)

Longitude (deg)

Longitude (deg)

Sep 4-7 Tens

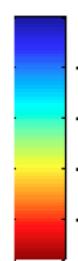
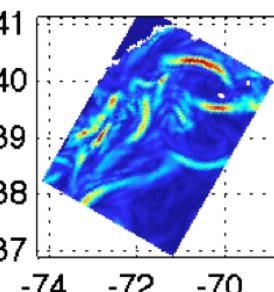


Longitude (deg)

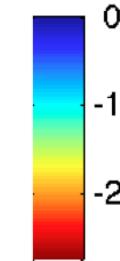
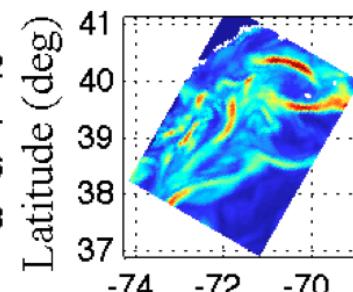
Longitude (deg)

Longitude (deg)

Sep 4-7 Tsig



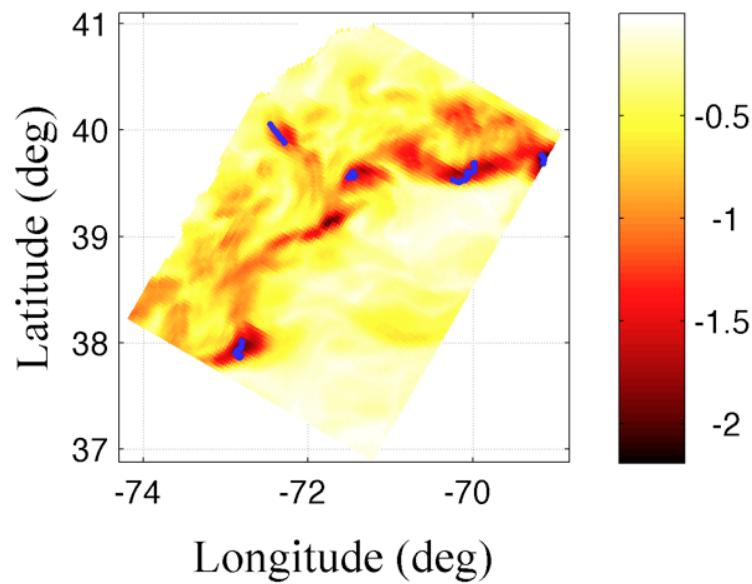
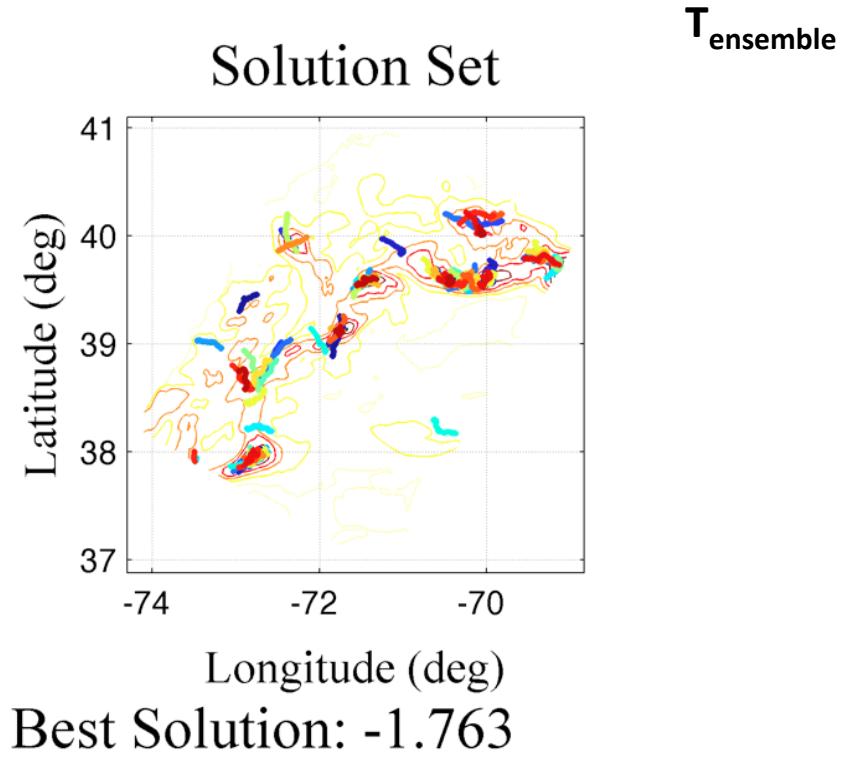
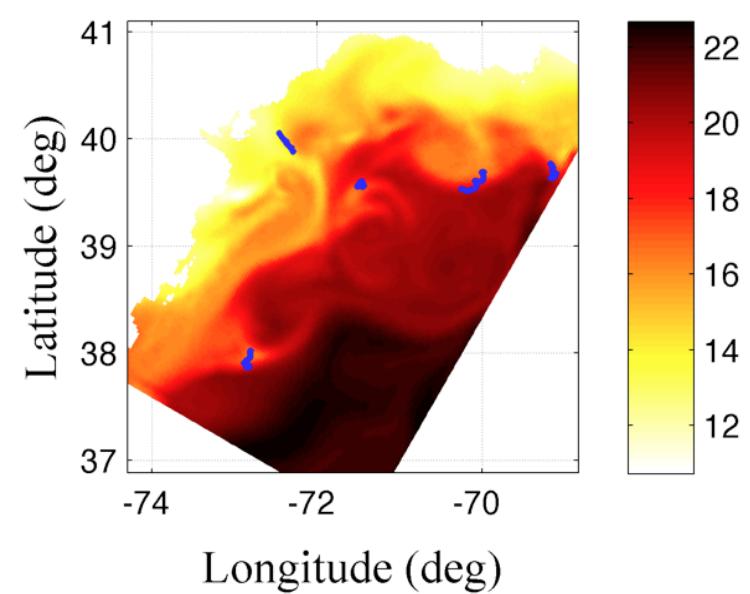
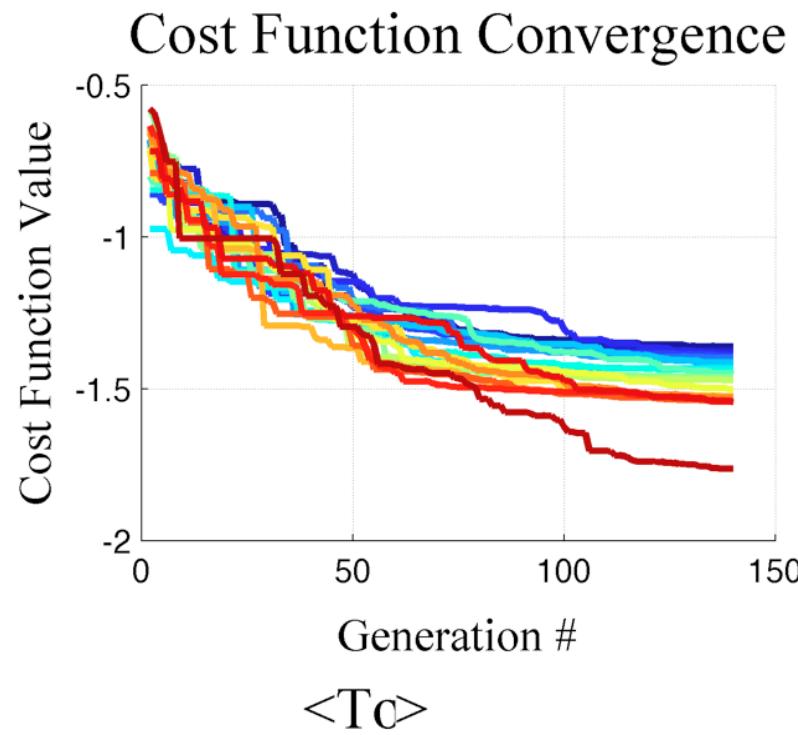
Sep 4-7 Comb



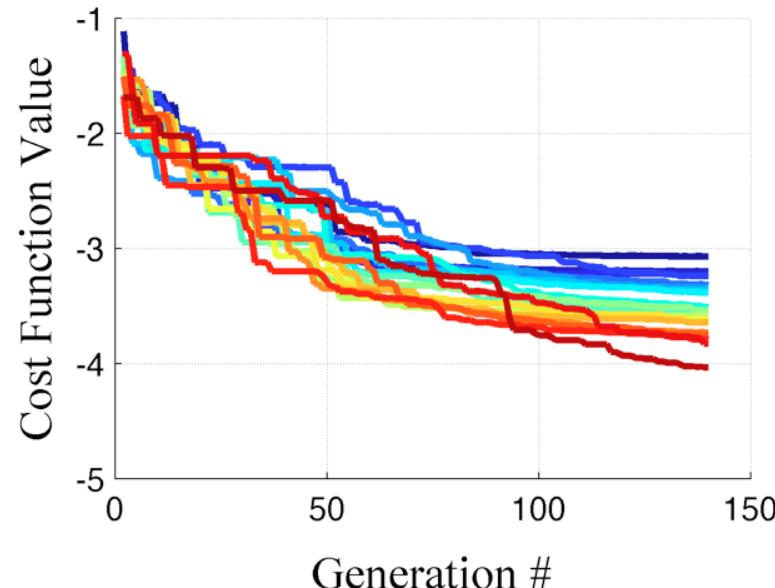
Longitude (deg)

Longitude (deg)

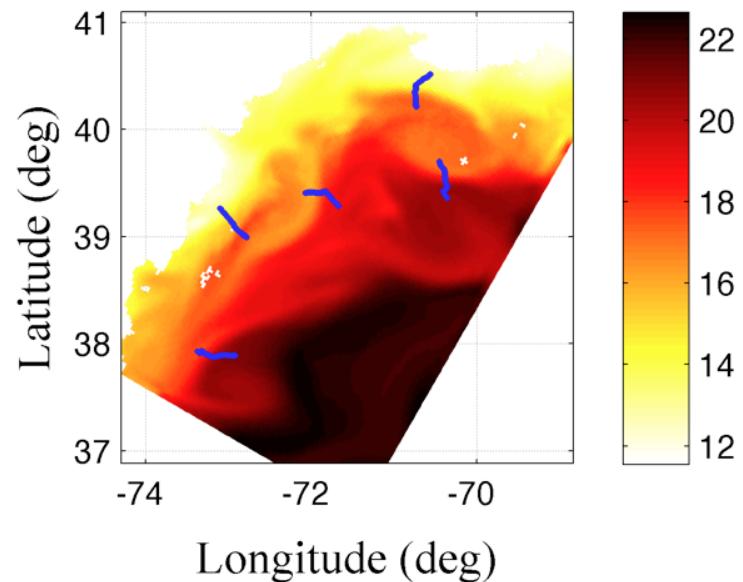
Longitude (deg)



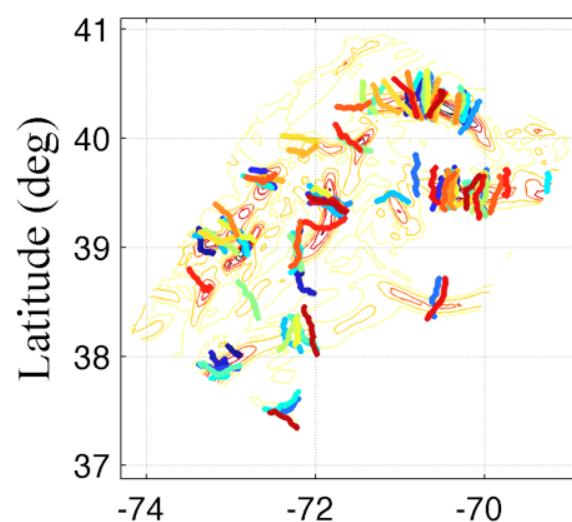
## Cost Function Convergence



$\langle T_c \rangle$

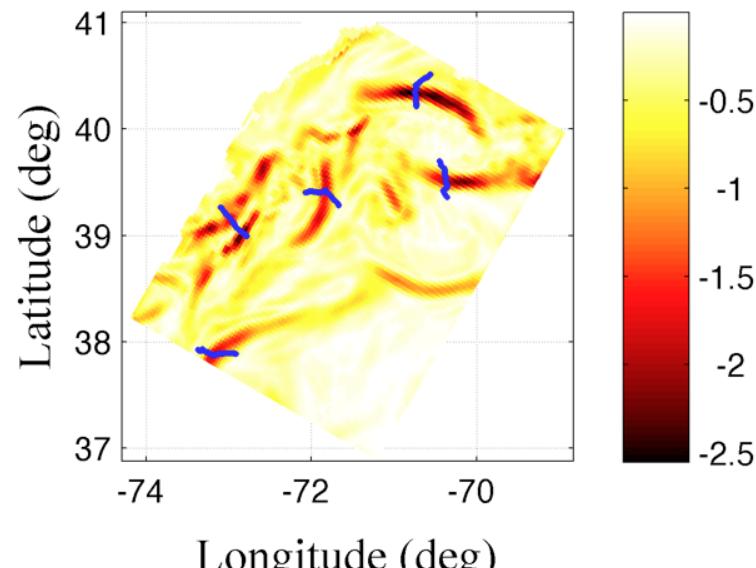


## Solution Set

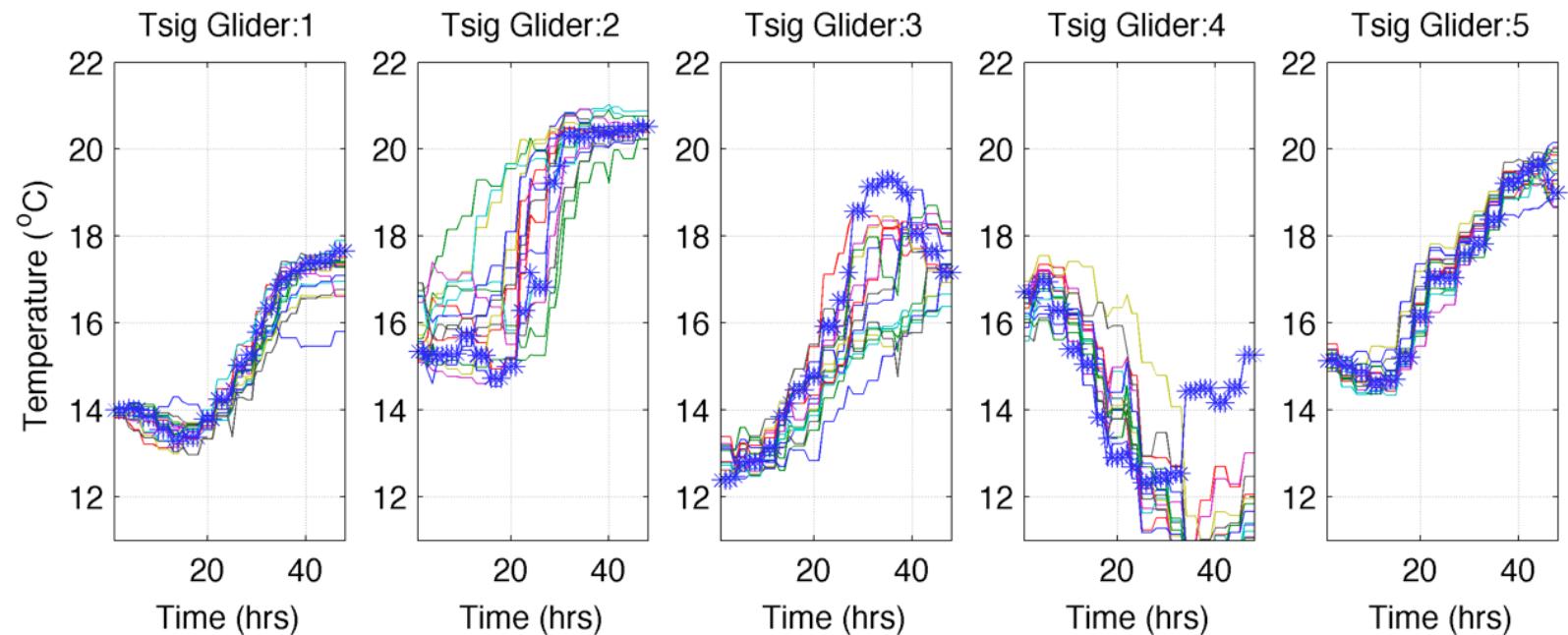
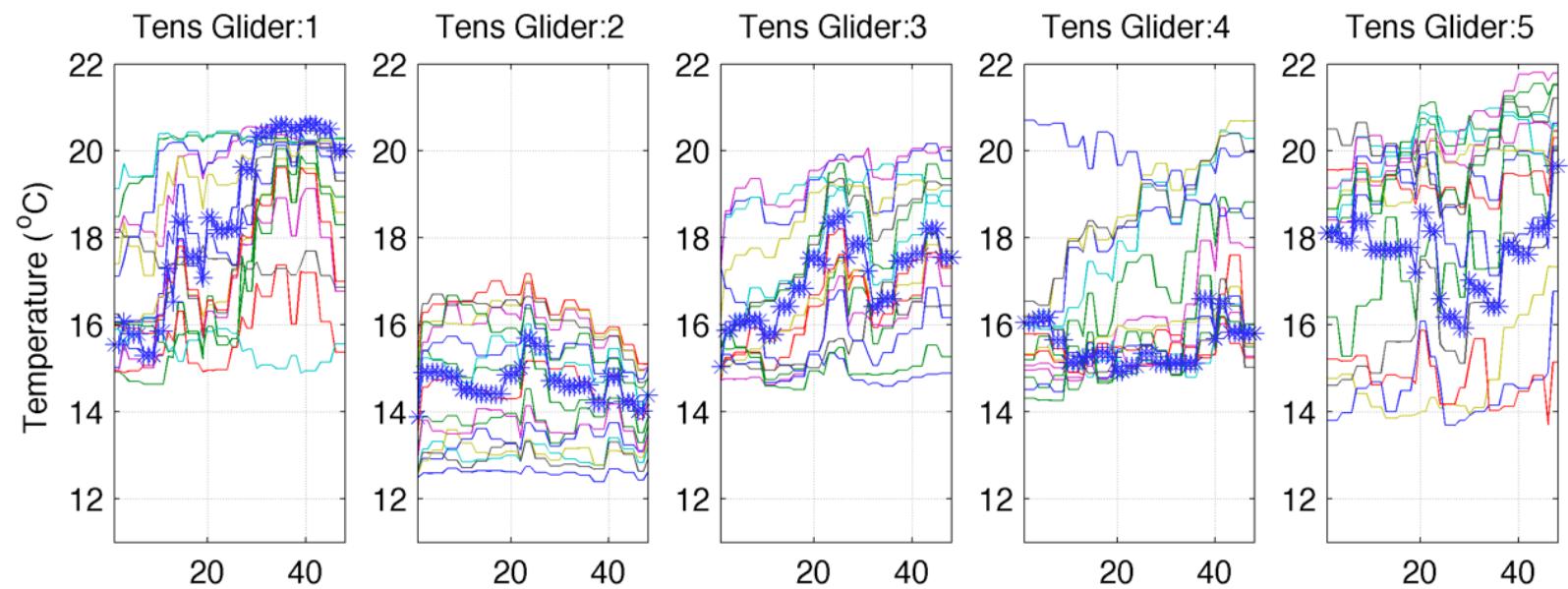


Longitude (deg)

Best Solution: -4.033



$T_{\text{spatial}}$

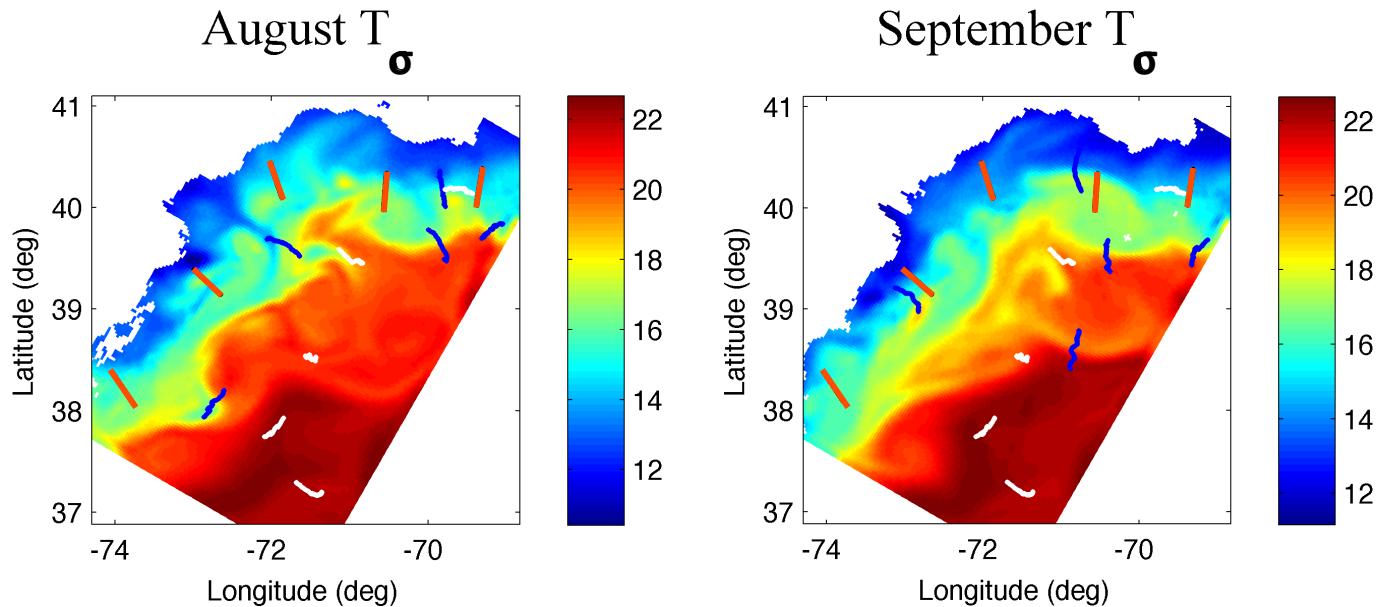


### c) GA Scheme Evaluation: Comparing Strategies using Data-Assimilation

#### Three Types of Strategies

[Heaney, Lermusiaux, Haley and Duda, 2009]

Two dynamic situations:  
before (Aug)  
and after (Sep)  
Ernesto



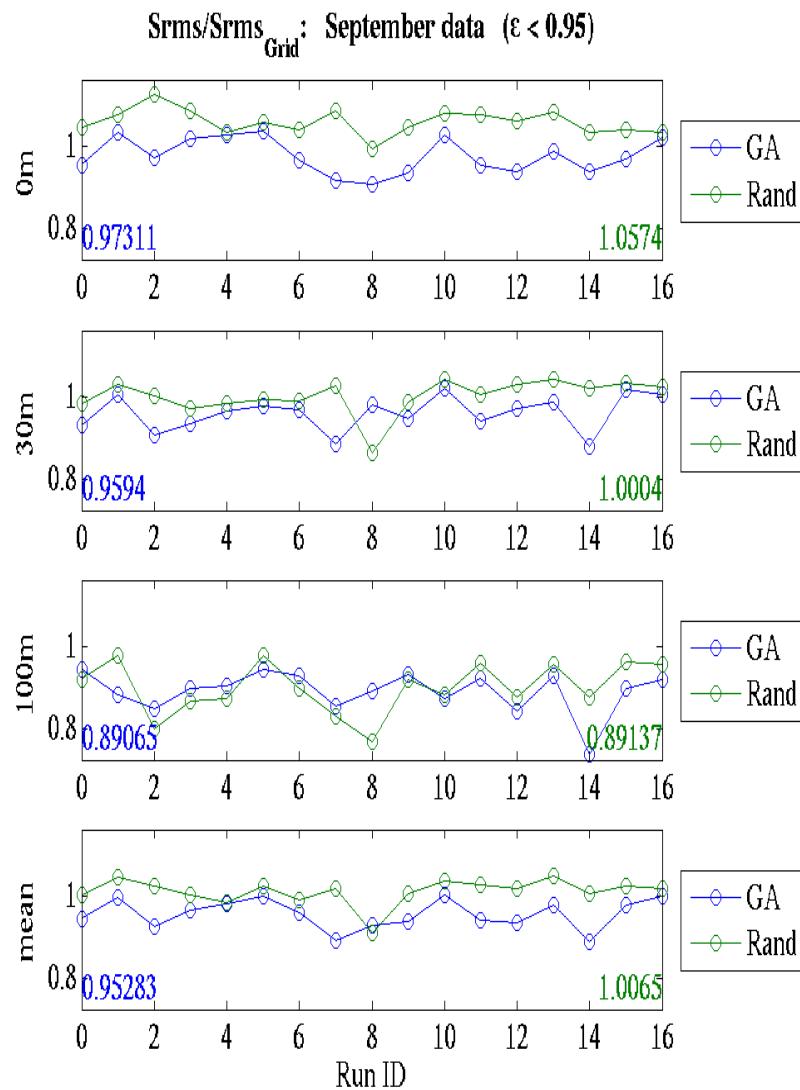
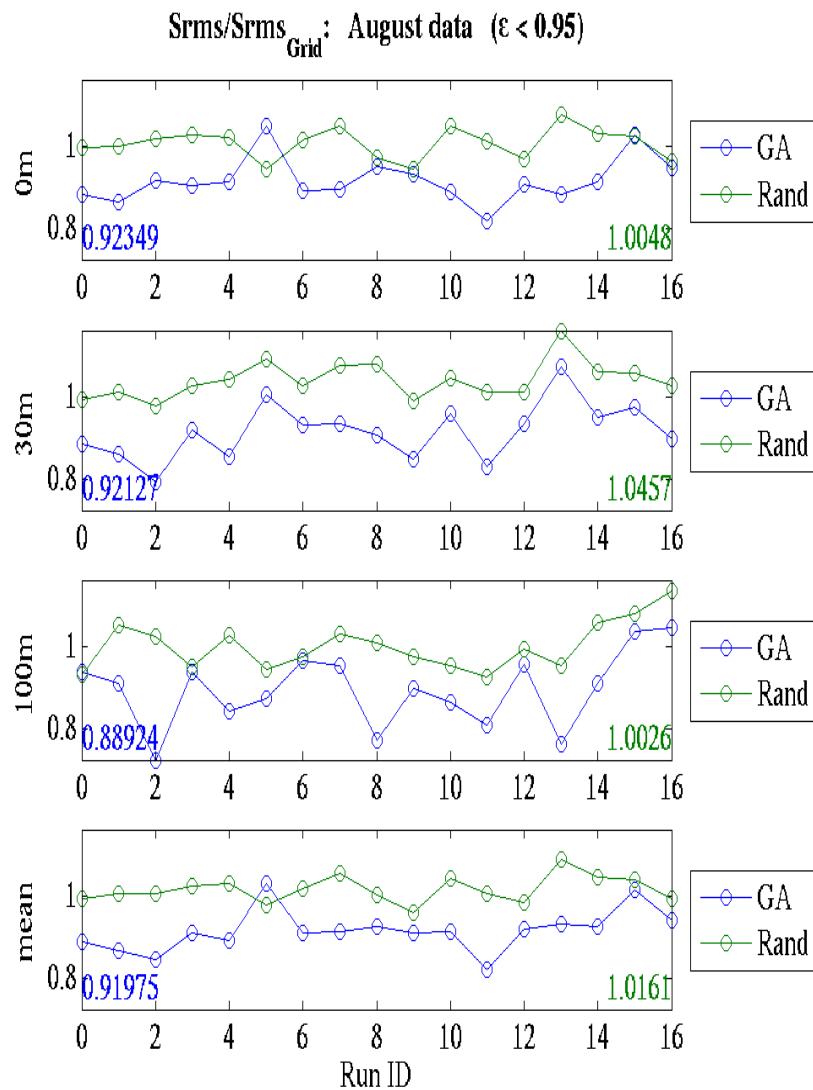
Five gliders to be optimized over 48 hours of sampling within large domain

#### Three Types of Sampling Paths

- ❖ *Grid* (red): the “smart” oceanographer
- ❖ *Random* (white)
- ❖ *Genetic Algorithm* (blue): GA paths computed to minimize chosen cost function

Paths overlaid on  $T_s$  surface for August 24-27 (left) and September 4-7 (right)

## 5-10% Gain over Random

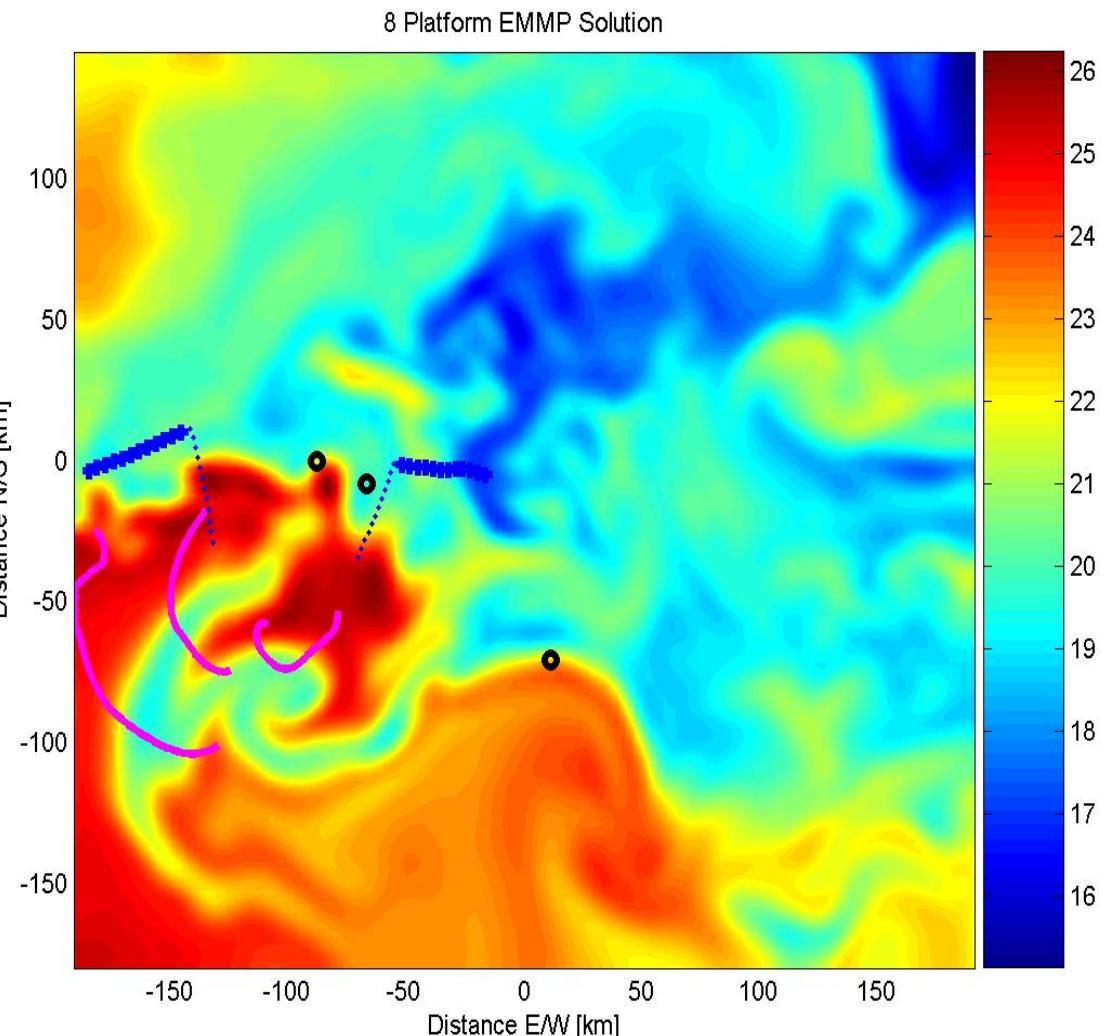


# Multi-Platform Dynamic

Time Dependent Optimization

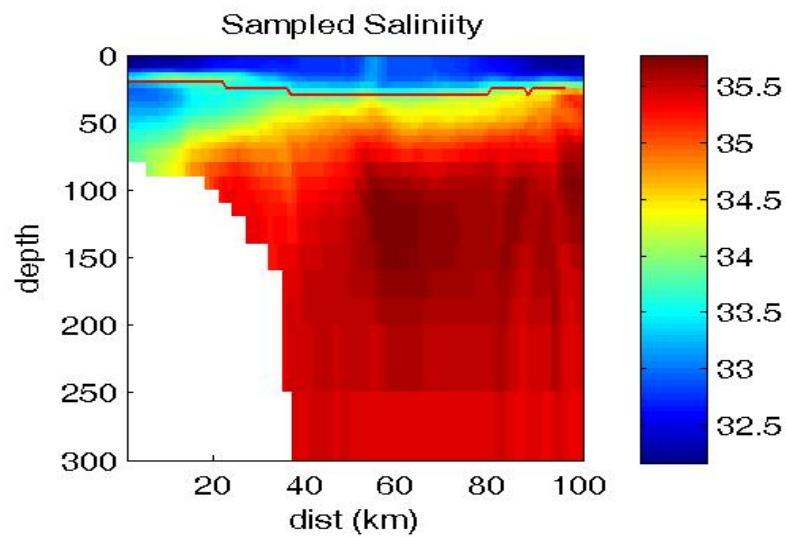
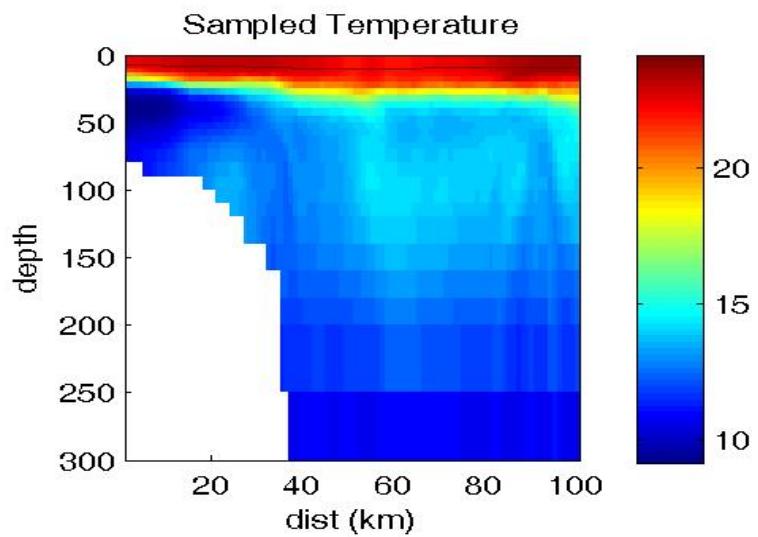
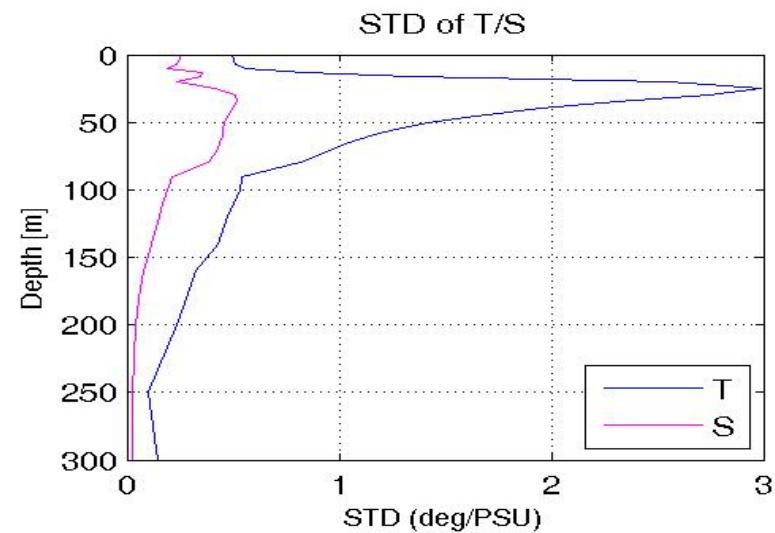
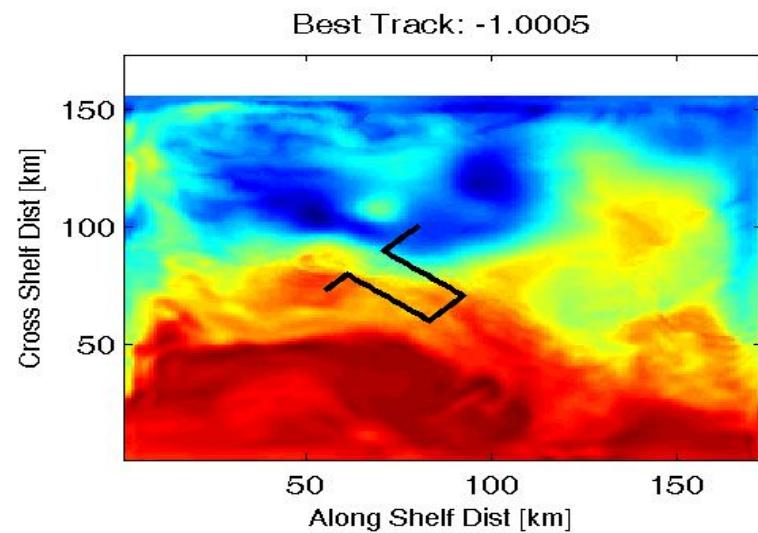
With:

- 3 – moorings (1 week)
- 3 – gliders (1 week)
- 2 – powered AUVs (24 hours)

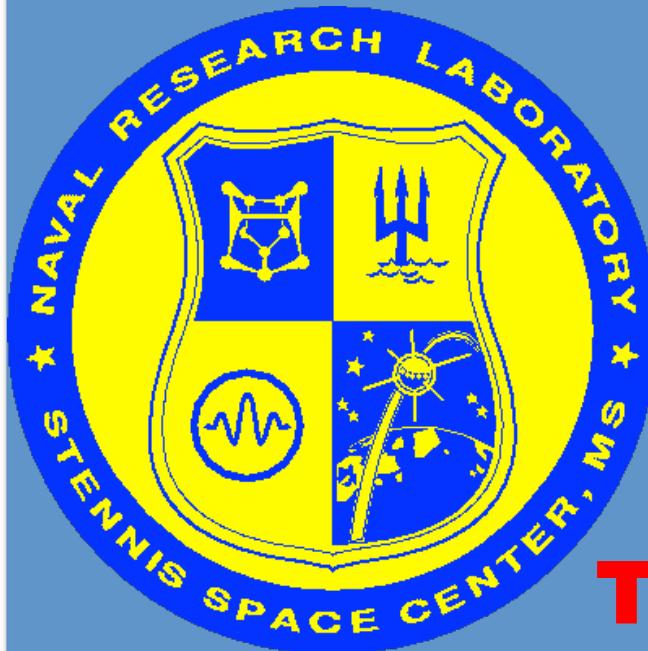


# Movie...(hopefully)

# ScanFish Survey – SW06



**20 Oct 2010**



# **FORECAST SYSTEM AND ADAPTIVE SAMPLING IN SUPPORT OF REP\_10**

**The NRLSSC Team:**

**Germana Peggion, E. Coelho, C. Rowley,  
L. Smedstad, K. Heaney, P. Spence, and D. Sitton.**

**The wind team: M. Phelps, P. Posey**

**The computer team: B. Maloy, D. Goolsby**

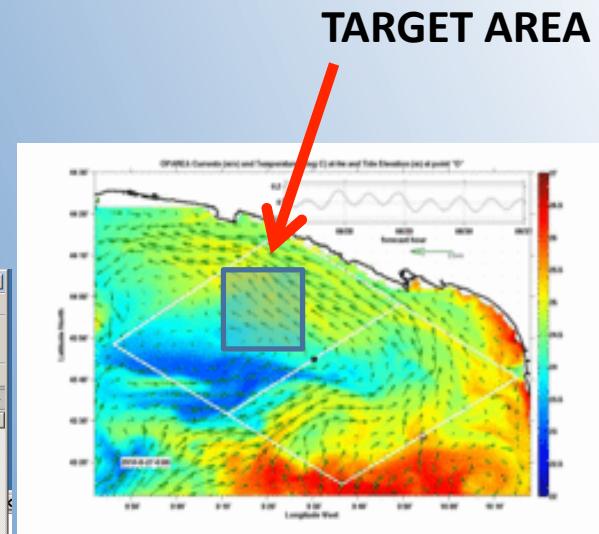
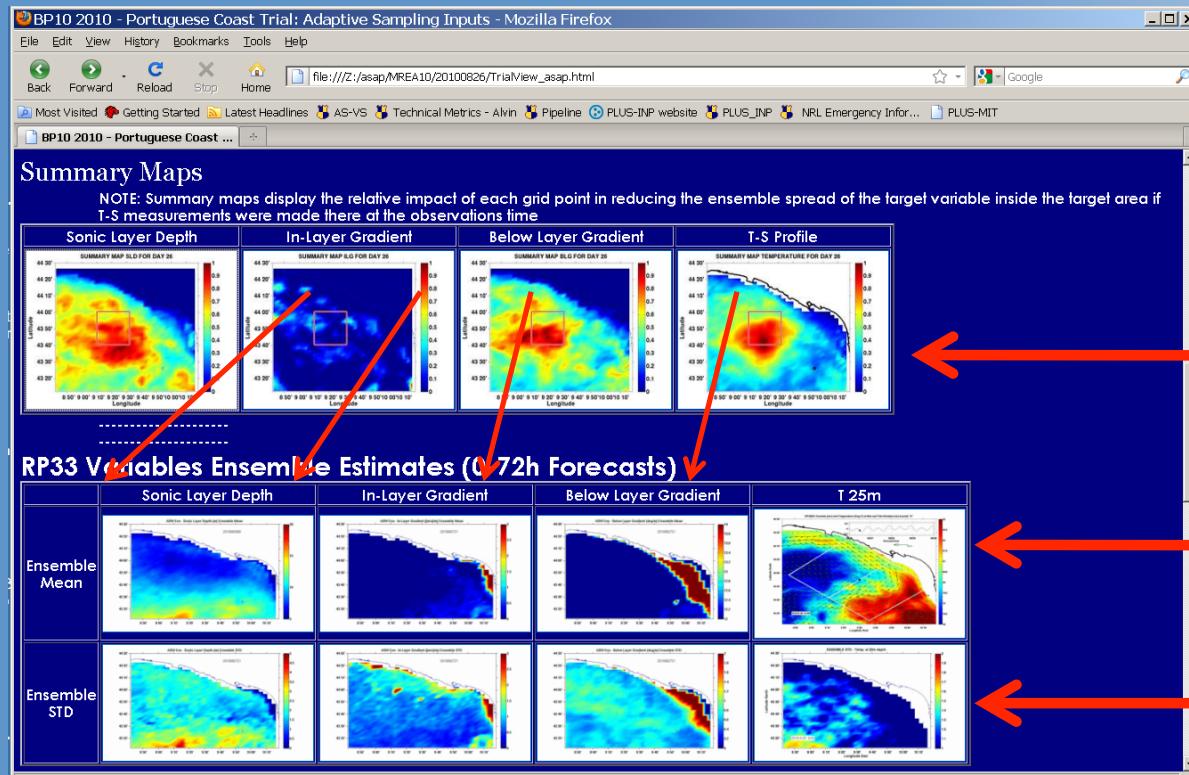


**[gpeggion@uno.edu](mailto:gpeggion@uno.edu)**

# Sampling Strategies

## MREA10 Test Case

### TOFU PROCESSING ENSEMBLE AND COST FUNCTIONS FOR KEY VARIABLES SLD – ILD – BLG – Temp/Sal Prof



Sensitivity (summary) Maps  
for target area

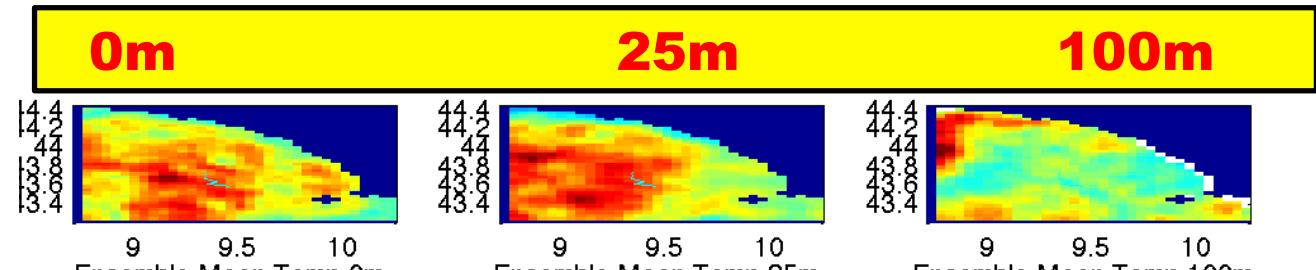
Space-Time Variability

Ensemble Spread (uncertainty)

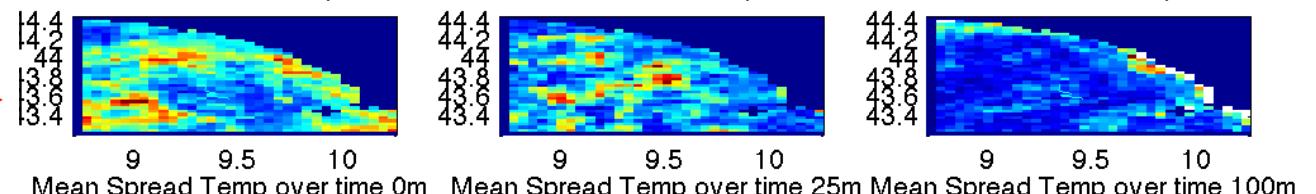
# Cost Function Morphologies

**Central Forecast**

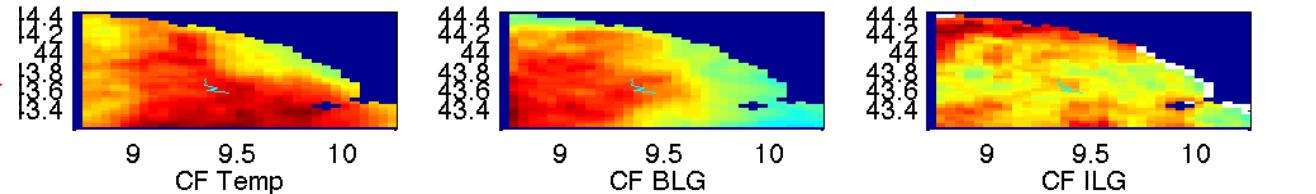
**Ens\_Spread\_T**



**Ens\_Mean\_T**



**Mean\_Spread\_T\_over\_time**



**TOFU**

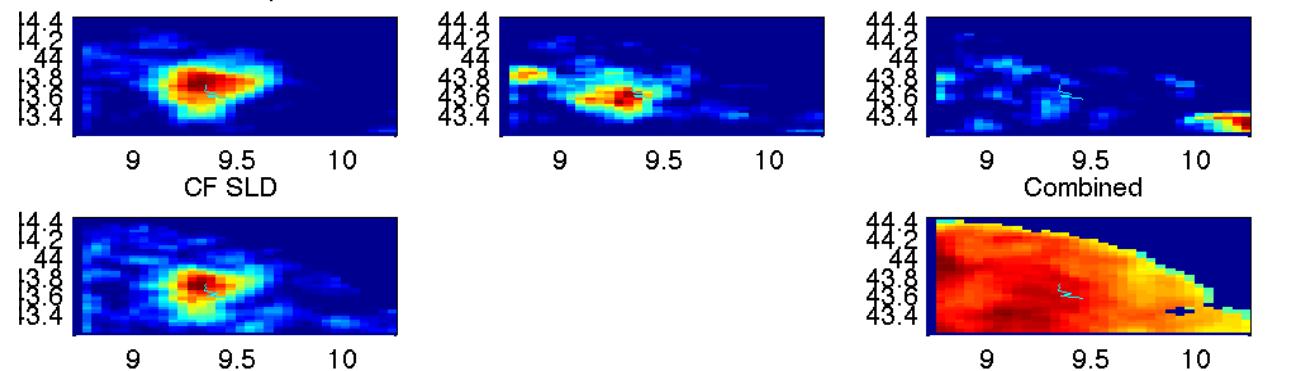
**CF Temp**

**CF BLG**

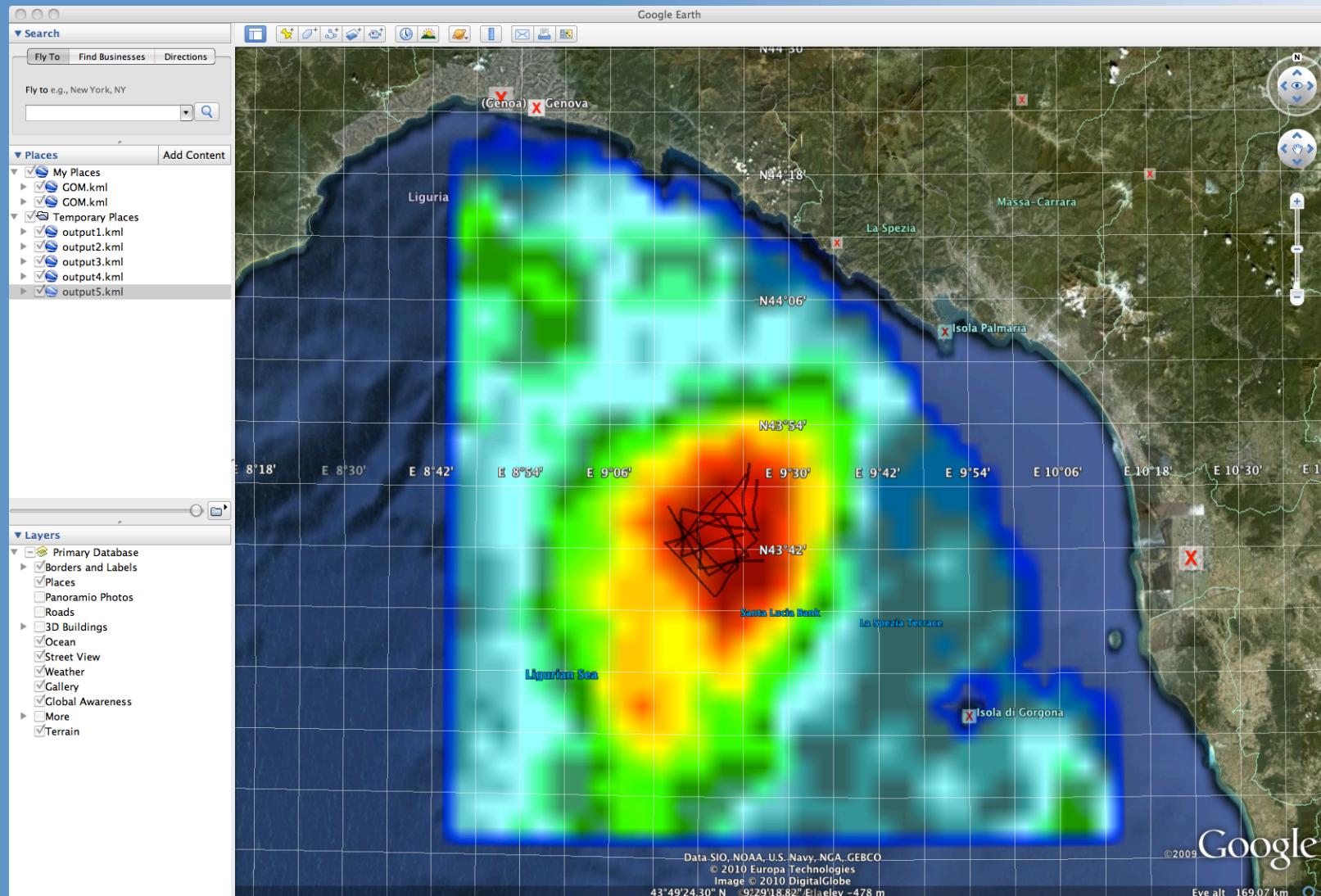
**CF ILG**

**CF SLD**

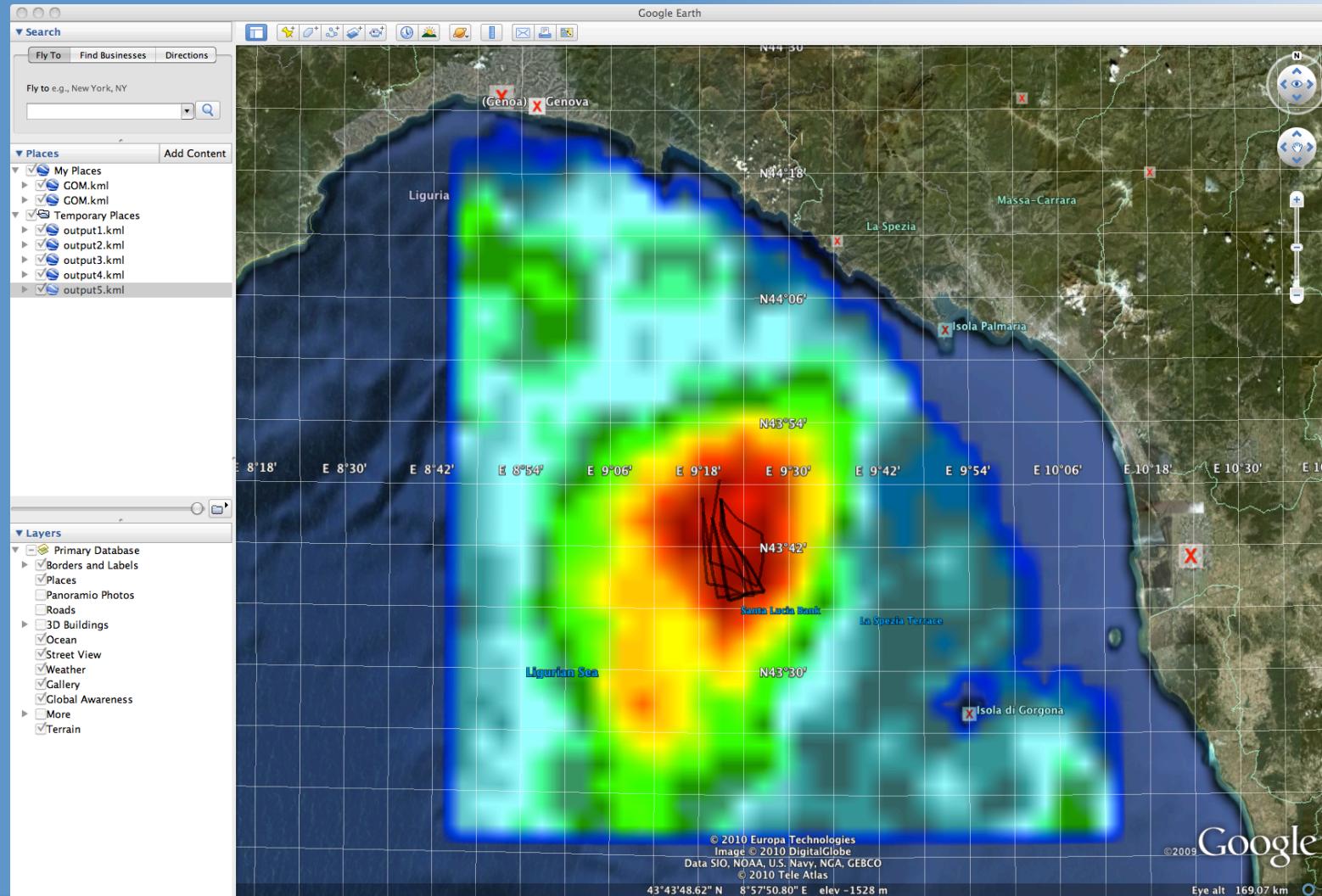
**Combined**



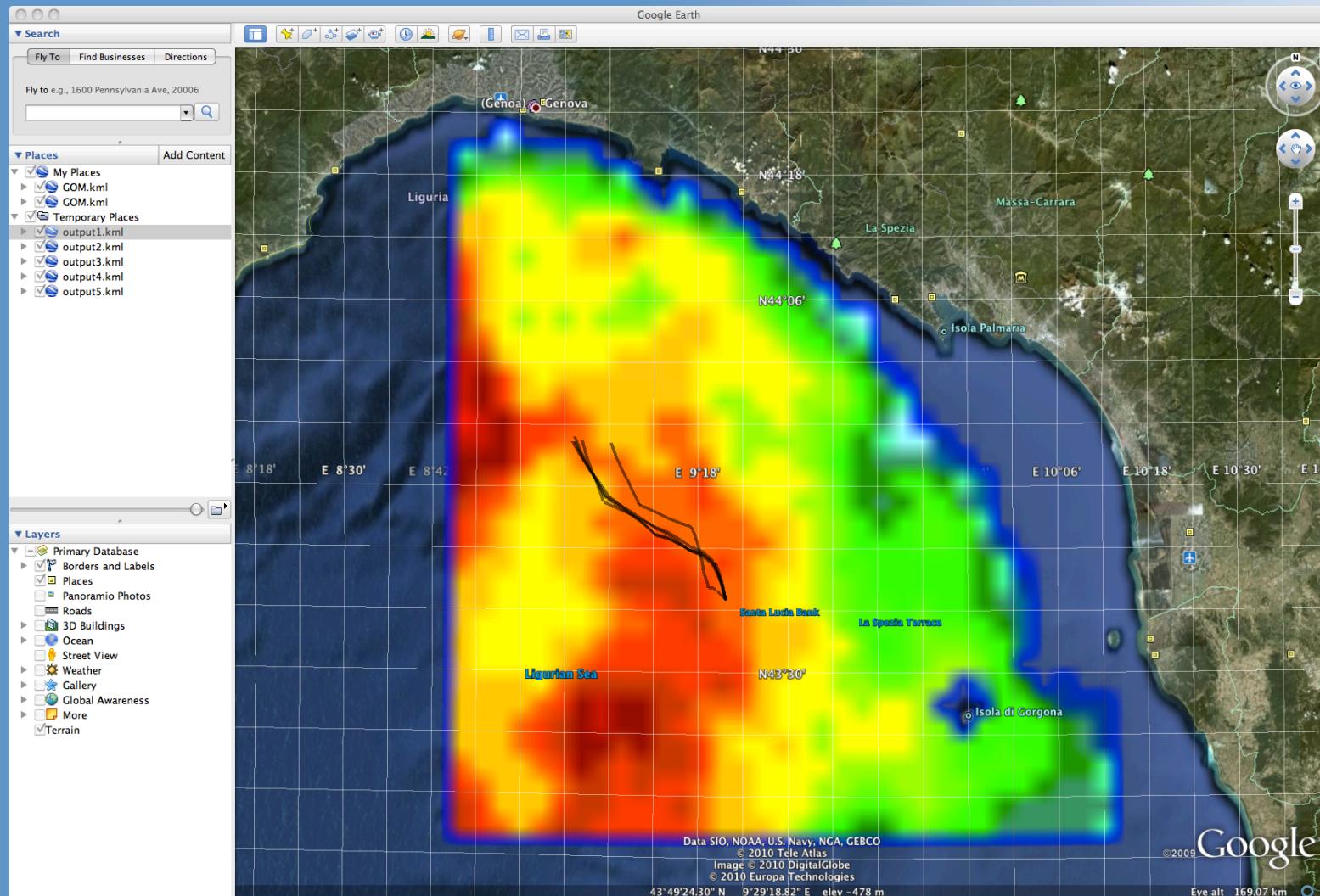
# 5 Best single glider searches

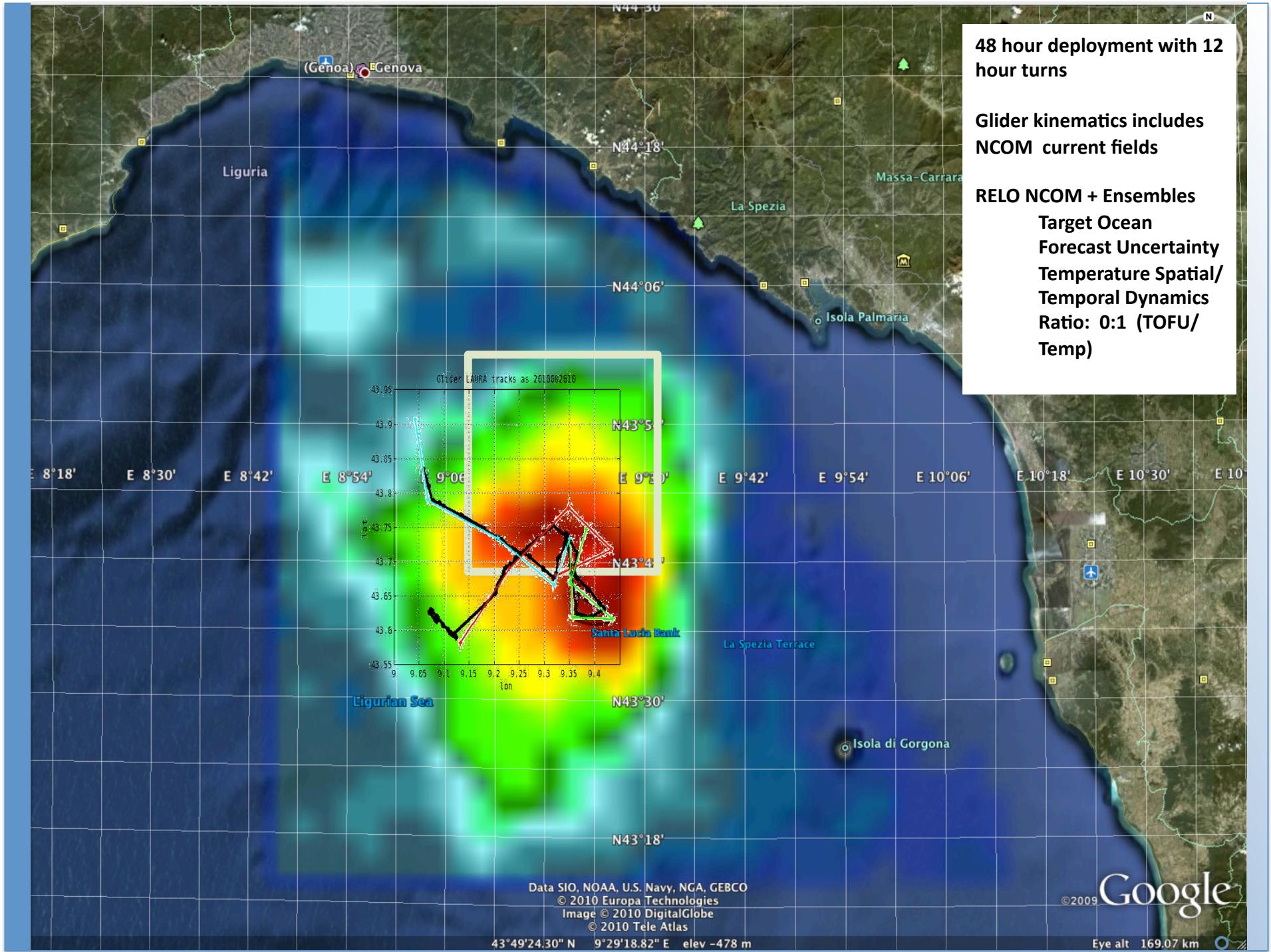


# 5 Best Solutions – REP 10 Day 1



# 5 Best Solutions – REP10 Day 6





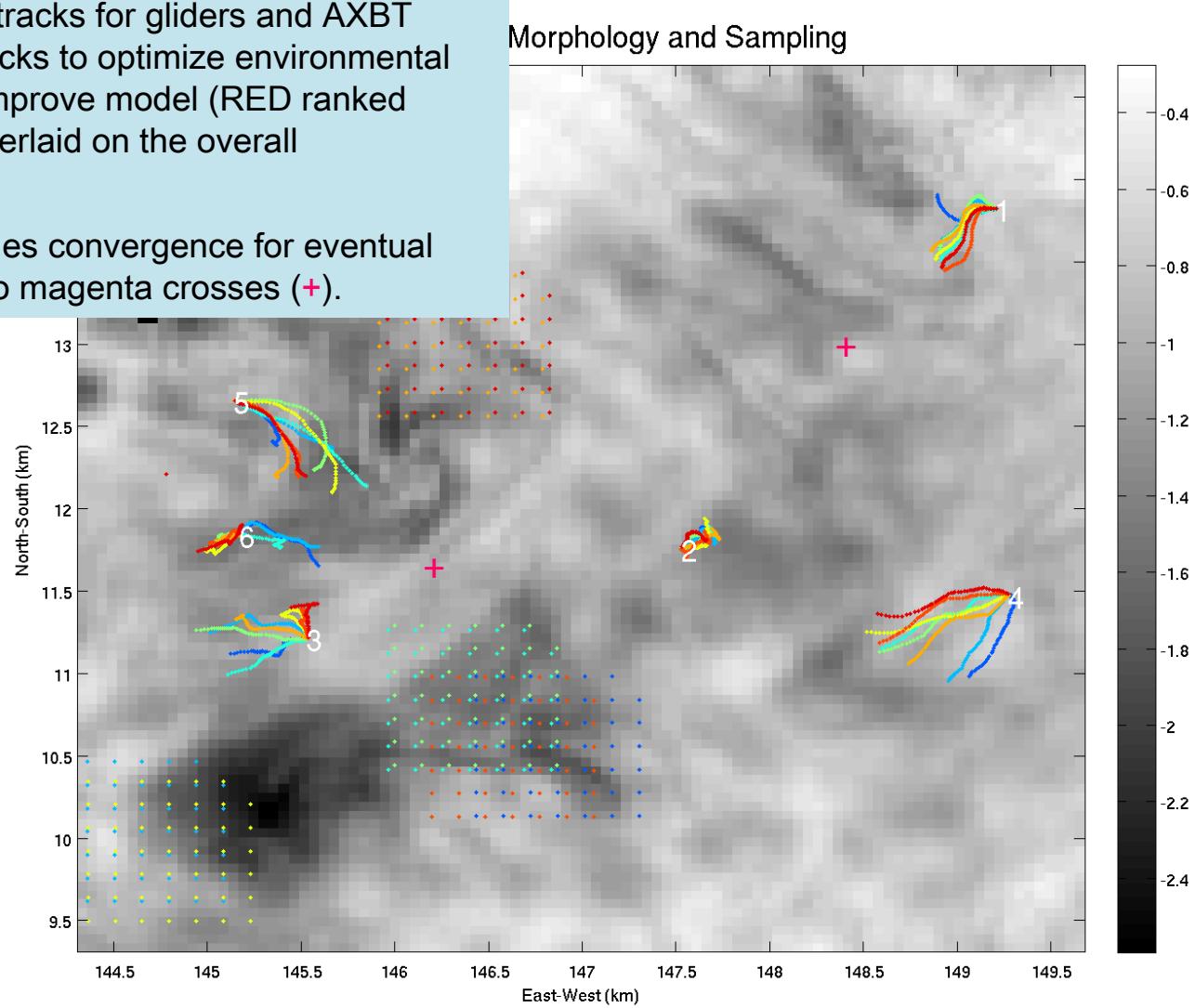
# The Main Challenges

- “I could have done that...”
- Communicating results to the operator
  - Did the GA converge? (convergence)
  - Are you sure? (uniqueness/sensitivity)
  - What physics is driving the selection of the path plans? (why?)

# 2007 Operational Display for Preliminary Op-Test

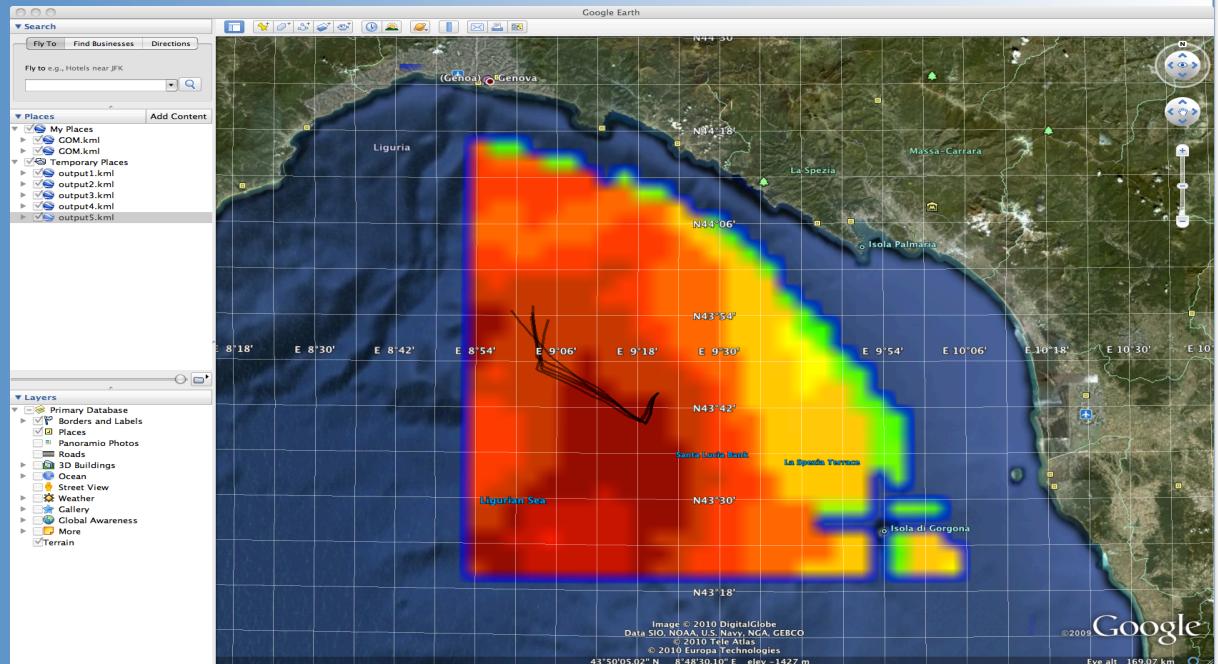
5 possible 48-hour tracks for gliders and AXBT sampling array. Tracks to optimize environmental observations and improve model (RED ranked best). These are overlaid on the overall morphology.

Cost function includes convergence for eventual pick up near the two magenta crosses (+).

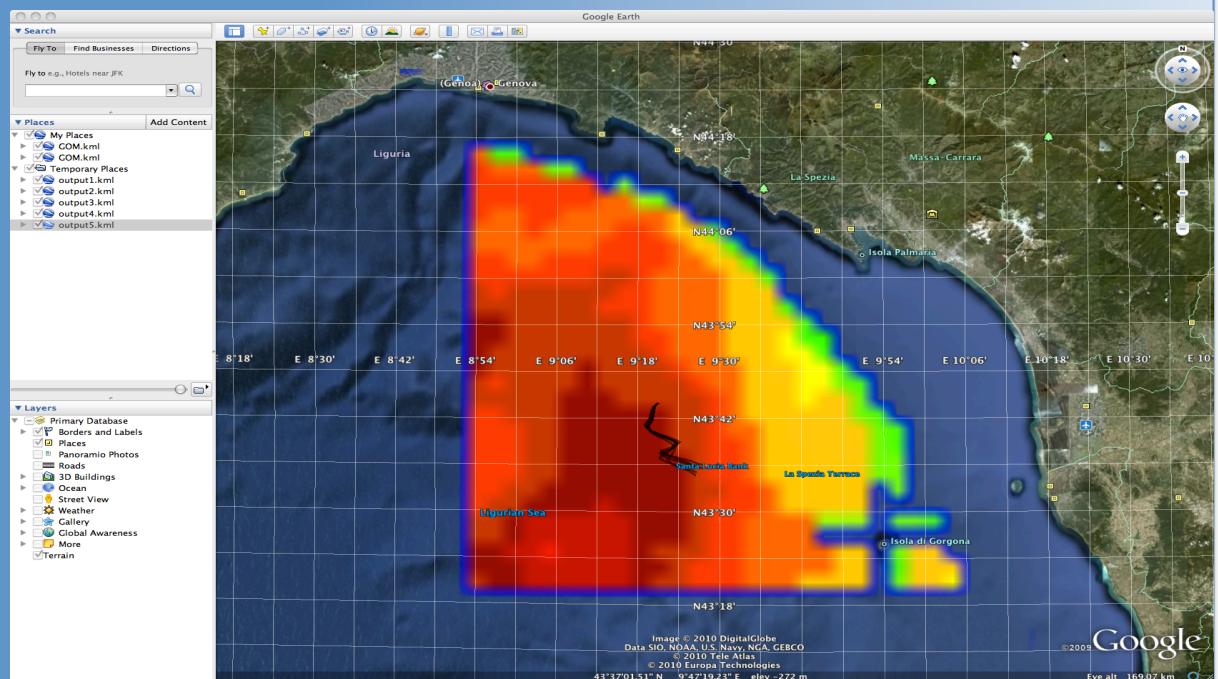


# Sensitivity to Glider Velocity

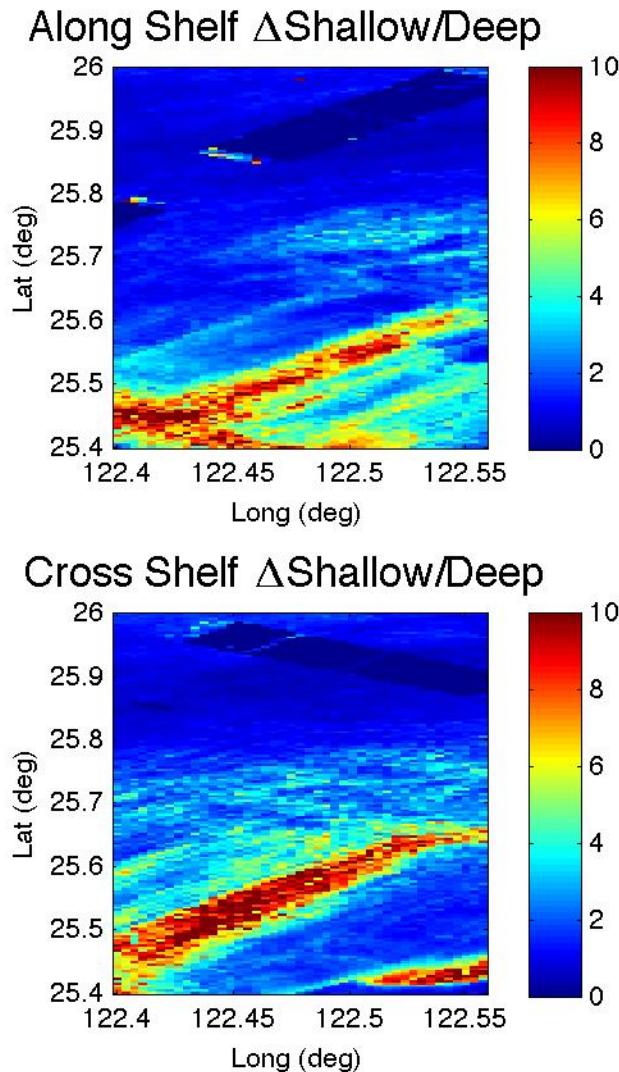
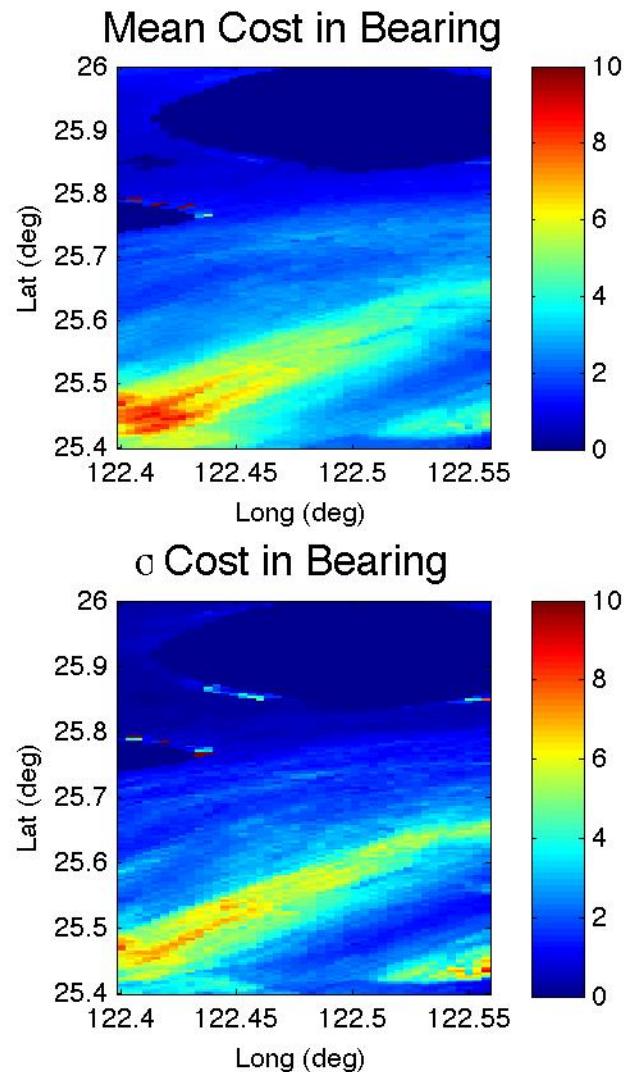
5 knot glider



4 knot glider



# Acoustic Adaptive Sampling



# Conclusions

EMPath (Cost function, platform model, GA) provides a flexible, robust approach to non-linear automatic optimization of multiple platforms.

The hard part:

- Defining what the user means by “optimal”
- Integrating downstream “costs” into the search plan
- Communicating effectively with end-users
- Demonstrating quantitative value added
- Fielding and directing more than 4 gliders for an extended period of time in the open ocean.

The way forward:

- Incorporation of measurements on future cost function
- Sensor planning within Lagrangian framework (LCS)