NRL Multivariate Ocean Data Assimilation System Overview

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DA Technique Based on Optimal Estimation Theory

- oceanographic version of MVOI method widely used in NWP systems (Daley, 1991)
- simultaneous analysis of five ocean variables: temperature, salinity, geopotential, and u-v velocity components (T, S, Φ , u, v)
- multivariate in mass and velocity

Observation Space Formulation

$$x_a = x_b + P_b H^T (HP_b H^T + R)^{-1} [y - H(x_b)]$$

where $\mathbf{x}_{a} = analysis$, $\mathbf{x}_{b} = background$ $\mathbf{P}_{b} = background error covariance$, $\mathbf{R} = observation error covariance$ $\mathbf{H} = forward operator (spatial interpolation in 3 dimensions)$ $(\mathbf{x}_{a} - \mathbf{x}_{b}) = analyzed increment$ $[\mathbf{y}-\mathbf{H}(\mathbf{x}_{b})] = innovation vector (synoptic T, S, u, v observations)$



NRL Multivariate Ocean Data Assimilation

Flexible System

- global or regional applications
- supports re-locatable, multi-scale analyses on nested, successively higher resolution grids
- used to initialize/update ocean forecast model or run stand-alone
- interfaced to multiple ocean forecast models
 - Hybrid Coordinate Ocean Model (HYCOM)
 - Navy Coastal Ocean Model (NCOM)
 - Wavewatch III wave forecast model (WW3)

Designed as Complete End-to-End Analysis System

- fully automated ocean data quality control
- multivariate analysis
- performance diagnostics package

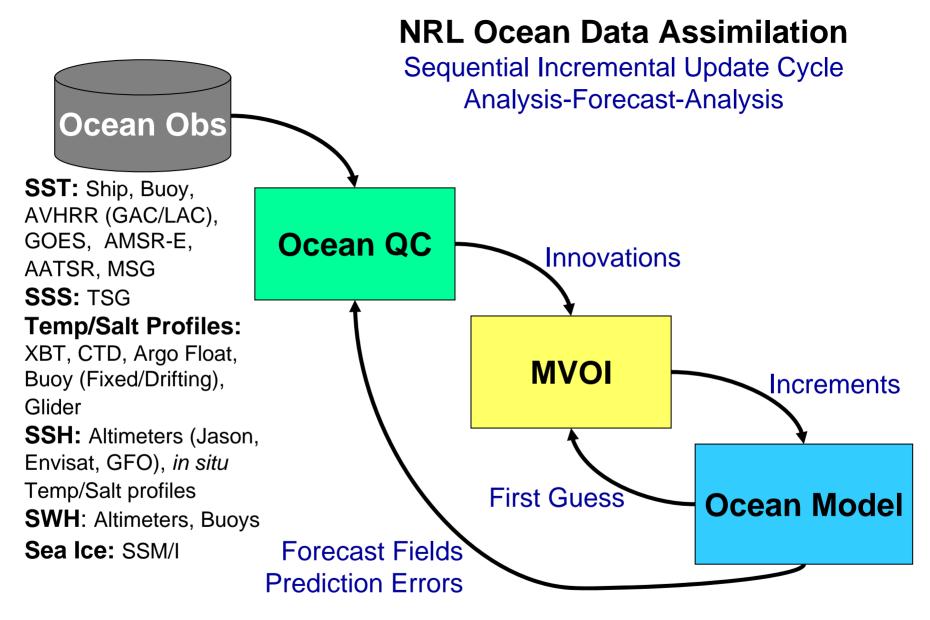


Operational Applications

- SST and sea ice analyses as LBCs for COAMPS and NOGAPS NWP models at FNMOC
- global and regional 3D analysis-only runs at FNMOC and NAVOCEANO

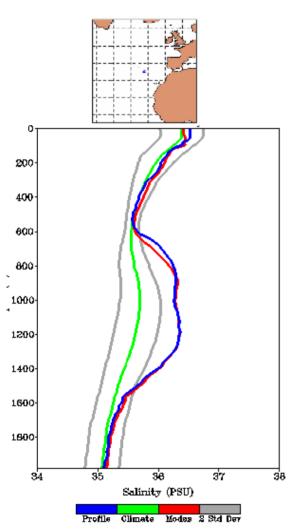
Pre-Operational Applications

- MVOI cycling with global HYCOM (NOPP/GODAE project)
- MVOI cycling with regional HYCOM and NCOM (various NRL and NOPP/CODAE projects)
- ensemble-based adaptive sampling project with ocean gliders
- assimilation altimeter SWH in global Wavewatch III at FNMOC



Model forecast fields and prediction errors are being evaluated for use in the QC of newly received ocean observations





observed salinity (Argo, CTD) is directly assimilated
salinity is derived for temperature-only profiles using MODAS historical TS relationships (bi-monthly, global)

$$S_{i,k}(T) = \overline{S}_{i,k} + \alpha_{i,k}(T - \overline{T}_{i,k})$$

$$\alpha_{i,k} = \sum b_{i,j}(T_{j,k} - \overline{T}_{i,k})(S_{j,k} - \overline{S}_{i,k}) / \sum b_{i,j}(T_{j,k} - \overline{T}_{i,k})^{2}$$

$$b_{i,j} = \exp(-((x_{j} - x_{i}) / L_{x})^{2} - ((y_{j} - y_{i}) / L_{y})^{2} - ((t_{j} - t_{i}) / L_{t})^{2})$$

• derived salinity values are put through the automated QC procedures in same way as an observed salinity

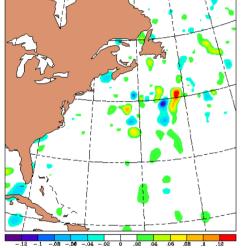
MODAS salt profile derived from Argo float temperature in MEDDY (red) - verifying Argo salt profile in blue – MODAS does a good job predicting MEDDY salinities



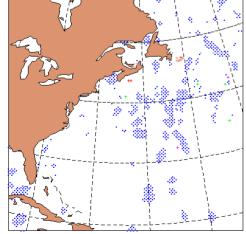
Altimeter SSH Assimilation

Synthetic T/S profiles generated using one of two approaches: Synthetics are assimilated in the MVOI as an observing system





Synthetic Sampling Locations



- 1. Direct method (modified form of Cooper Haines)
 - adjusts model density profile to be consistent with measured change in model forecast SSH (creates T/S innovations)
- observation errors set to forecast error variance plus residual error from iterative fit of density adjustment
- 2. MODAS synthetic BT method
 - computes temperature at depth from SSHA using stored regressions of climate anomalies of temperature and dynamic height
- salinity is then computed from synthetic temperatures
- observation errors are set to stored regression residuals

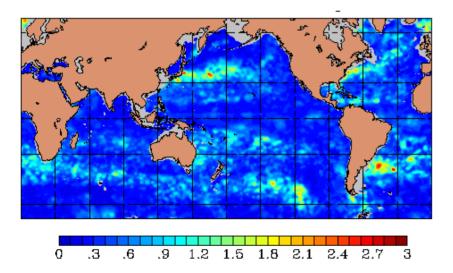
synthetic profiles are generated where analyzed change in SSH exceeds altimeter measurement errors (~2 cm)



Separable formulation: product of a variance and a correlation

Error Variances

- vary by position, depth and analysis variable
- evolve with time, updated continuously using analyzed increment fields
- error growth parameterization used in long term absence of observations
 - function of age of data on grid and innovation temporal autocorrelations
 - evolves to climate or free running model errors in limit of no data

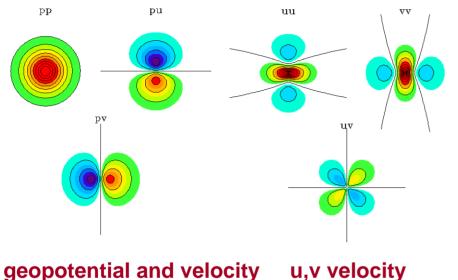


400 M temperature background error standard deviation (°C) from FNMOC global analysis

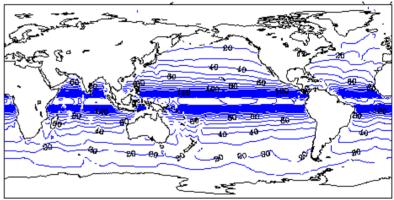


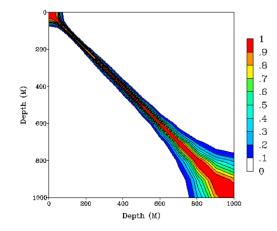
Background Error Correlations

Separable formulation: product of a horizontal and a vertical correlation



Horizontal correlations are multivariate in velocity and geopotential - length scales depend on location (default is Rossby radius)





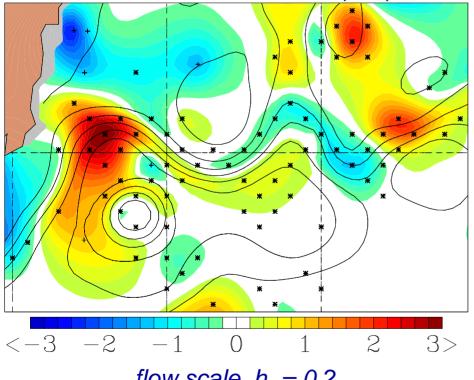
Rossby Radius Deformation (km)

Vertical correlation length scales are computed from background vertical density gradients - correlations evolve with time but do not vary with location

scales are large (small) when stratification is weak (strong)



Temperature Increments at 100 M (color) Forecast Surface Height (lines) Observation Locations (*, +)



flow scale, $h_s = 0.2$

MVOI correlations modified to incorporate flow dependence from forecast surface height.

 $\mathbf{C} = \mathbf{C}_{h}\mathbf{C}_{v}\mathbf{C}_{f}$

 C_h – horizontal correlation C_v – vertical correlation C_f – flow dependent correlation

Small (large) values of flow dependence scalar (h_s) produces strong (weak) flow dependence

Note increments stretched along Kuroshio front and decorrelation in cold eddy south of the Kuroshio



- sum of measurement error (ϵ_i), representation error (ϵ_r), and data age error (ϵ_{τ})
- measurement errors are from data provider or table look-up
 - satellite SST measurement errors computed from buoy match-ups
 - in situ measurement errors function of instrument type
- satellite representation error function of resolution of model (r_g), resolution of observing system (r_o), and background horizontal gradient (∇_b)

 $\varepsilon_{\rm r} = \nabla_{\rm b} \cdot ({\rm r_o} / {\rm r_g})$

 $r_o > r_g$

• profile representation error function of mesoscale variance (σ_t) and uncertainty associated with internal wave activity (vertical gradient used as proxy)

 $\varepsilon_r = \kappa_t \sigma_t + \lambda_t \cdot (\mathsf{dT} \cdot \mathsf{d} z^{-1})$

• κ_t and λ_t are determined empirically for temperature and salinity

• data age error a function of observation time (τ_o), time correlation scale (τ_c), and observation depth (z_o) relative to maximum depth of time correlation (z_c)

 $\begin{aligned} \varepsilon_{\tau} &= \varepsilon_{i} \cdot (\tau_{o} \cdot \delta) / \tau_{c} \\ \delta &= 1 - z_{o} / (z_{o} + z_{c}) \end{aligned}$

 \bullet observation error variances tuned based on $J_{\mbox{\scriptsize min}}$ statistics

 $J_{min} \ll 1$ variances too large, $J_{min} \gg 1$ variances too small (or bad data)

Projects Underway and Present Developments

Projects Underway – Present Developments

- Conversion to 3D-Var assimilation system based on NAVDAS
 - natural development pathway from MVOI
 - global solution, no data selection
 - direct assimilation of observations with (weak) nonlinearities in the observation operator
 - non-separable covariances (horizontal and vertical)
 - vertical flow dependence (isopycnal coordinates)
 - new balance operators (based on Anthony Weaver's work)
- Development of 4D-Var assimilation based on NAVDAS-AR
 - focus on Navy Coastal Ocean Model (NCOM) in limited domains
- Adaptive Sampling
 - incorporation of ensemble covariances in assimilation (so called hybrid approach)

Projects Underway – Present Developments

- Assimilation of new observing systems
 - glider T/S data and METOP, MSG, AMSR-E and MTSAT SST
- Diagnostics and automated tuning of error covariances
- Wave Model assimilation of altimeter SWH
 - Global and regional Wavewatch III assimilation at FNMOC
 - evaluation of alternative wave model updating strategies
- Global and regional HYCOM/NCODA NOPP/GODAE transitions
 - real-time demonstration later this year (2007)
- Development and implementation of adaptive QC methods
- Improvement in pre-processing functions
 - adaptive data thinning to handle large amounts of satellite observations
 - bias detection/correction

END