Greetings to all,

The Global Ocean Data Assimilation Experiment (GODAE) final symposium will be held in Nice in November 12-15 2008. This project has been a precursor to a worldwide experiment to demonstrate the feasibility of global ocean observing systems using state of the art assimilation techniques. Today, several teams are working on operational ocean systems to provide forecast and description of the ocean, using increasingly complex assimilation schemes and high resolution models. As we saw in the last newsletter, these systems have reached the coast and routinely provide real time ocean forecast. But they need input information for their boundaries and initialisation fields, from regional, basin wide or global configurations.

This month, the Newsletter is dedicated to global ocean systems resulting from the GODAE project.

In the first news feature, a review of the GODAE achievements in ocean observing systems is made by Le Traon et al. In a second introduction paper, Pierre Bahurel provides a “Global view on MyOcean” where he introduces the special ongoing efforts to improve products and services to users.

Four systems from three countries (U.S., France and Japan) are then presented, showing a variety of developments, model resolutions and assimilation schemes that are all facing the same challenges: to describe, understand and forecast the world ocean. The first contribution is from Chassignet et Hurlburt and is dedicated to the U.S. HYCOM 1/12° global configuration. Menemenlis et al. will then tell us how useful the ECCO2 system is in understanding and estimating ocean processes. Legalloudec et al. follow with the 1/12° Mercator global model and its ability to represent the mesoscale activity. Finally, Kamachi et al. will present the MRI global systems, including two nesting configurations dedicated to several applications from climate variability to boundary forcing or ocean weather.

The next newsletter will be published in January 2009 and dedicated to the Mediterranean Sea.

We wish you a pleasant reading.
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GODAE Oceanview: from an experiment towards a long-term Ocean Analysis and Forecasting International Program

By Pierre Yves Le Traon1, Mike Bell2, Eric Dombrowsky3, Andreas Schiller4, Kirsten Wilmer-Becker2

1 Ifremer Brest, Technopole Brest-Iroise, BP70, 29280 Plouzané cedex
2 Met Office, Exeter, UK
3 Mercator Ocean, 8-10 rue Hermès, Toulouse, France
4 CAWCR-CSIRO, Hobart, Tasmania, Australia

Over the past 10 years, GODAE through its International GODAE Steering Team (IGST) has coordinated and facilitated the development of global and regional ocean forecasting systems and has made excellent progress. It has been demonstrated that global ocean data assimilation is feasible and GODAE has made important contributions to the establishment of an effective and efficient infrastructure for global operational oceanography that includes the required observing systems, data assembly and processing centers, modeling and data assimilation centers and data and product servers. GODAE as an experiment will end in 2008. Its final symposium (Nice, November 12-15, 2008) will provide an opportunity to review the key achievements of the last 10 years. The symposium will also discuss the future of operational ocean analysis and forecasting and proposals for its international coordination. Main issues are summarized hereafter.

Although there are still major challenges to face, global operational oceanography now needs to transition from a demonstration to a permanent and sustained capability. Most GODAE groups have or are now transitioning towards operational or pre-operational status. GODAE systems are also evolving to satisfy new requirements (e.g. for coastal zone and ecosystem monitoring and forecasting, climate monitoring) and must benefit from scientific advances in ocean modeling and data assimilation.

In order to ensure the required long-term international collaboration and cooperation on these issues, it is thus proposed to set up an international program on ocean analysis and forecasting systems called GODAE OceanView. Through its science team, GODAE OceanView would provide international coordination and leadership in:

- The development and scientific testing of the next generation of ocean analysis and forecasting systems, covering biogeochemical and eco-systems as well as physical oceanography, and extending from the open ocean into the shelf sea and coastal waters.
- The exploitation of this capability in other applications (weather forecasting, seasonal and decadal prediction, climate change detection and its coastal impacts, etc).
- The assessment of the contribution of the various components of the observing system and scientific guidance for improved design and implementation of the ocean observing system.

GODAE OceanView science team will provide a forum where the main operational and research institutions involved in global ocean analysis and forecasting can develop collaborations and international coordination of their activities. It will include scientists from the main operational systems as well as scientific experts on specific fields (e.g. observation, modeling, data assimilation) and representatives of key observing systems. Some of the GODAE OceanView objectives will be pursued through a number of Task Teams (e.g. Intercomparison and Validation, Observing System Evaluation, Coastal Ocean and Shelf Seas, Marine Ecosystem Monitoring and Prediction). These teams will address specific topics of particular importance to GODAE OceanView usually in collaboration with international research programs (e.g. OOPC, CLIVAR, IMBER, WCRP). Operational aspects related to product harmonization and standardization and links with JCOMM will be carried out by the JCOMM ET-OOFS.

To summarize, operational oceanography faces many challenges with time scales ranging from weather to climate. It is inherently an international issue, requiring broad collaboration to span the global oceans. GODAE OceanView will promote the development of ocean modelling and assimilation in a consistent framework to optimize mutual progress and benefit. It will also promote the associated exploitation of improved ocean analyses and forecasts and provide a means to assess the relative contributions of and requirements for observing systems. GODAE OceanView detailed objectives and links with international research programs must now be discussed with the wider community. The GODAE final symposium will be a major opportunity for starting such an interaction.
A Global View on MyOcean

By Pierre Bahurel

Numerical global models have now joined satellites and in situ instruments to reinforce their ocean watch mission all around the planet. They form together a powerful brigade to monitor the ocean state, describe its real time situation anywhere, and forecast its short-term evolutions. They deliver a new and global view on the ocean.

“One planet, one ocean”, as it is declared these days on the frontpage of the International Oceanographic Commission (IOC) website. It sounds like an invitation to join efforts all around the world for a better depiction of the ocean. It sounds like an injunction to act collectively to increase our knowledge and respect of our ocean planet. During the past ten years, the international “Global Ocean Data Assimilation Experiment (GODAE)” conducted with IOC has been a first answer to this challenge, and lead to a major step forward in operational oceanography. It has lead to the emergence of a reliable, continuous, real-time, 3D and global capacity in ocean monitoring and forecasting. World-leading teams (amongst with the Japanese, US and French teams presenting their global ocean capacities in this newsletter) have joined effort and motivation to set up a new international network of operational oceanography centres.

Europe has undoubtedly taken an important role in the development of this modern operational oceanography. Born with successful national projects at the end of the 90’s (such as Mercator in France) now linked together and cross-fertilized through European projects from ESA or the Commission, the European capacity for ocean monitoring and forecasting has reached the point where the demonstration is over, and the service activated. This is what “MyOcean” is about.

MyOcean is the European service for ocean monitoring and forecasting, the marine component of the “Kopernikus” European program for a global monitoring of environment & security. The mission is straightforward: offer the best information available on the state of the global ocean and European regional seas for the benefit of any citizen, decision-maker, or downstream service provider requiring it. To serve a user community as wide as the marine application sectors, the MyOcean focus has been clearly set on the common denominator data requested for all users: a “core” information on the ocean provided by the European Marine “Core” Service.

Mercator Ocean is the coordinator of this new European service.

The MyOcean project will start in the first days of 2009, and the MyOcean service will open in the following months.

The FP7 project that provides for the 3 years coming (2009-2011) the European Union framework to set-up this new service gathers with Mercator Océan 60 partners, all the major operational oceanography centres, all maritime member states from UE, and the best skills in Europe for this challenge. This consortium is composed of old companions of the GODAE years, but half of it is formed by new teams from other countries or communities in Europe.

MyOcean is built indeed on the strong belief that sharing data and knowledge increases the value of the information service. That’s why connections with other initiatives in the world — on the model invented by GODAE — are priorities for the MyOcean European team to build this new view on our global ocean.
Introduction

During the past 5-10 years, a broad partnership of institutions has been collaborating in developing and demonstrating the performance and application of eddy-resolving, real-time global and basin-scale ocean prediction systems using the HYbrid Coordinate Ocean Model (HYCOM). These systems have been or are in the process of being transitioned for operational use by the U.S. Navy at the Naval Oceanographic Office (NAVOCEANO), Stennis Space Center, MS, and by the National Oceanic and Atmospheric Administration (NOAA) at the National Centers for Environmental Prediction (NCEP), Washington, D.C. The systems run efficiently on a variety of massively parallel computers and include sophisticated, but relatively inexpensive, data assimilation techniques for assimilation of satellite altimeter sea surface height and sea surface temperature as well as in-situ temperature, salinity, and float displacement. The partnership represents a broad spectrum of the oceanographic community, bringing together academia, federal agencies, and industry/commercial entities, spanning modeling, data assimilation, data management and serving, observational capabilities, and application of HYCOM prediction system outputs. All participating institutions were committed and the collaborative partnership provided an opportunity to leverage and accelerate the efforts of existing and planned projects, consequently producing a high quality product that should collectively better serve a wider range of users than would the individual projects.

The HYCOM partnership is a U.S. component of the international Global Ocean Data Assimilation Experiment (GODAE). GODAE is a coordinated international effort envisioning “a global system of observations, communications, modeling, and assimilation that will deliver regular, comprehensive information on the state of the oceans, in a way that will promote and engender wide utility and availability of this resource for maximum benefit to the community” (see Chassignet and Verron (2006) for a review). Navy applications, NOAA applications such as maritime safety, fisheries, the offshore industry, and management of shell/coastal areas are among the expected beneficiaries of the HYCOM ocean prediction systems (http://www.hycom.org). More specifically, the precise knowledge and prediction of ocean mesoscale features is used by Navy, NOAA, the oil industry, and fisheries for an optimal use of their resources. Examples are optimal ship and submarine routing, search and rescue, oil spill drift application, monitoring of the open ocean ecosystems, fisheries management, short range coupled atmosphere-ocean forecasts, forecast of the coastal and near-shore environment, etc…

Background

Numerical modeling studies over the past several decades have demonstrated progress in both model architecture and the availability of computational resources for the scientific community. Perhaps the most noticeable aspect of this progression has been the evolution from simulations on coarse-resolution horizontal/vertical grids outlining basins of simplified geometry and bathymetry and forced by idealized stresses, to fine-resolution simulations incorporating realistic coastal definition and bottom topography, forced by observational data on relatively short time scales. The choice of the vertical coordinate system in an ocean model however remains one of the most important aspects of its design. In practice, the representation and parameterization of the processes not resolved by the model grid are often directly linked to the vertical coordinate choice (Griffies et al., 2000). Oceanic general circulation models traditionally represent the vertical in a series of discrete intervals in either a depth, density, or terrain-following unit. Because none of the three main vertical coordinates (depth, density, and terrain-following) provide universal optimality, it is natural to envision a hybrid approach that combines the best features of each vertical coordinate. Isopycnic (density-tracking) layers work best for modeling the deep stratified ocean, levels at constant fixed depth or pressure are best to use to provide high vertical resolution near the surface within the mixed layer, and terrain-following levels are often the best choice for modeling shallow coastal regions. In HYCOM, the optimal vertical coordinate distribution of the three vertical coordinate types is chosen at every time step. The default configuration of HYCOM is isopycnic in the open stratified ocean, but makes a dynamically smooth transition to terrain-following coordinates in shallow coastal regions and to fixed pressure-level coordinates in the surface mixed layer and/or unstratified seas. In doing so, the model takes advantage of the different coordinate types in optimally simulating coastal and open-ocean circulation features (Chassignet et al., 2006, 2007). A user-chosen option allows specification of the vertical coordinate separation that controls the transition among the three coordinate systems. The assignment of additional coordinate surfaces to the oceanic mixed layer also allows the straightforward implementation of multiple vertical mixing turbulence closure schemes (Halliwell, 2004). The choice of the vertical mixing parameterization is also of importance in areas of strong entrainment, such as overflows.
Data assimilation is essential for ocean prediction because (a) many ocean phenomena are due to nonlinear processes (i.e., flow instabilities) and thus are not a deterministic response to atmospheric forcing, (b) errors exist in the atmospheric forcing, and (c) ocean models are imperfect, including limitations in numerical algorithms and in resolution. Most of the information about the ocean surface's space-time variability is obtained remotely from instruments aboard satellites [Sea Surface Height (SSH) and Sea Surface Temperature (SST)], but these observations are insufficient for specifying the subsurface variability. Vertical profiles from expendable bathythermographs (XBT), conductivity-temperature-depth (CTD) profilers, and profiling floats (e.g., Argo, which measures temperature and salinity in the upper 2000 m of the ocean) provide another substantial source of data. Even together, these data sets are insufficient to determine the state of the ocean completely, so it is necessary to exploit prior knowledge in the form of statistics determined from past observations as well as our present understanding of ocean dynamics. By combining all of these observations through data assimilation into an ocean model, it is possible to produce a dynamically consistent depiction of the ocean. However, in order to have any predictive capabilities, it is extremely important that the freely evolving ocean model (i.e., non-data-assimilative model) has skill in representing ocean features of interest.

To properly assimilate the SSH anomalies determined from satellite altimeter data, the oceanic mean SSH over the altimeter observation period must be provided. In this mean, it is essential that the mean current systems and associated SSH fronts be accurately represented (position, amplitude, and sharpness). Unfortunately, the earth’s geoid is not presently known with sufficient accuracy for this purpose, and coarse hydrographic climatologies (~0.5º-1º horizontal resolution) cannot provide the spatial resolution necessary when assimilating SSH in an eddy-resolving model (horizontal grid spacing of 1/10º or finer). At these scales of interest, it is essential to have the observed means of boundary currents and associated fronts sharply defined (Hurlburt et al., 2008). Figure 1 shows the climatological mean derived on a 0.5º grid using surface drifters by Maximenko and Niiler (2005) as well as the mean currently used in the Navy global HYCOM prediction system (see following section for details). The HYCOM mean was constructed as follows: a 5-year mean sea surface height field from a non-data assimilative 1/12º global HYCOM run was compared to available climatologies and a rubber-sheeting technique (Carnes et al., 1996) was used to modify the model mean in two regions, the Gulf Stream and the Kuroshio, where the western boundary extensions were not well represented and where an accurate frontal location is crucial for ocean prediction. Rubber-sheeting consists of a suite of computer programs specifically designed to operate on SSH fields, overlay contours from a reference field, and move masses of water in an elastic way (hence rubber-sheeting).
Figure 1
Mean SSH (in cm) derived from surface drifters (Maximenko and Niiler, 2005) (top panel) and from a non-data assimilative HYCOM run corrected in the Gulf Stream and Kuroshio regions using a rubber-sheeting technique (bottom panel). The RMS difference between the two fields is 9.2 cm.

The HYCOM Ocean Prediction Systems (http://www.hycom.org)

Two systems are in the process of being evaluated for operational use by the U.S. Navy at the Naval Oceanographic Office (NAVOCEANO), Stennis Space Center, MS, and by the National Oceanic and Atmospheric Administration (NOAA) at the National Centers for Environmental Prediction (NCEP), Washington, D.C.

The first system is the NOAA Real Time Ocean Forecast System for the Atlantic (RTOFS). The Atlantic domain spans 25°S to 76°N with a horizontal resolution varying from 4 km near the U.S. coastline to 20 km near the African coast. The system is run daily with one-day nowcasts and five-day forecasts. Prior to June 2007, only the sea surface temperature was assimilated. In June 2007, NOAA implemented the 3D-Var data assimilation of i) sea surface temperature and sea surface height (JASON and GFO), ii) temperature and salinity profile assimilation (ARGO, CTD, moorings, etc.), and iii) GOES data. The model outputs are available at http://polar.ncep.noaa.gov/ofsf/.

The second system is the pre-operational global U.S. Navy nowcast/forecast system using the 1/12° global HYCOM (6.5 km grid spacing on average, 3.5 km grid spacing at North Pole, and 32 vertical hybrid layers), which has been running in near real-time since December 2006 and in real-time since February 2007. The current ice model is thermodynamic (energy loan), but it will soon include more physics as it is upgraded to PIPS (based on the Los Alamos CICE ice model). The model is currently running daily on...
379 processors on the IBM Power 5+ at the Naval Oceanographic Office using a part of the operational allocation on the machine. The daily run (U.S. Navy requirement) consists of a 5 day hindcast and a 5 day forecast and takes about ~15 wall clock hours. The system assimilates sea surface height (Envisat, GFO, and Jason-1), ii) sea surface temperature (all available satellite and in-situ sources), iii) all available in-situ temperature and salinity profile (ARGO, CTD, moorings, etc.), and iv) SSM/I sea ice concentration. The three-dimensional multivariate optimum interpolation Navy Coupled Ocean Data Assimilation (NCODA) (Cummings, 2005) system is the assimilation technique. The NCODA horizontal correlations are multivariate in geopotential and velocity, thereby permitting adjustments (increments) to the mass fields to be correlated with adjustments to the flow fields. The velocity adjustments are in geostrophic balance with the geopotential increments, and the geopotential increments are in hydrostatic agreement with the temperature and salinity increments. Either the Cooper and Haines (1996) technique or synthetic temperature and salinity profiles (Fox et al., 2002) can be used for downward projection of SSH and SST. An example of forecast performance is shown in Figure 2.

![Figure 2](image-url)

**Figure 2**

Verification of 30-day ocean forecasts: median SSH anomaly correlation vs. forecast length in comparison with the verifying analysis for the global HYCOM over the world ocean and five subregions. The red curves verify forecasts using operational atmospheric forcing, which reverts toward climatology after five days. The green curves verify “forecasts” with analysis quality forcing for the duration and the blue curves verify forecasts of persistence (i.e. no change from the initial state). The plots show median statistics over twenty 30-day HYCOM forecasts initialized during January 2004 - December 2005, a period when data from three nadir-beam altimeters, Envisat, GFO and Jason-1, were assimilated.

The model outputs from the hindcast experiment are available through the HYCOM consortium web page, http://www.hycom.org. A validation of the results is underway using independent data with a focus on the large scale circulation features, sea surface height variability, eddy kinetic energy, mixed layer depth, vertical profiles of temperature and salinity, sea surface temperature and coastal sea levels. Figure 3 and 4 show some examples for the Gulf Stream region while Figure 5 documents the performance of HYCOM in representing the mixed layer depth. HYCOM is also an active participant in the international GODAE comparison of global ocean forecasting systems.
Ocean U.S. GODAE: Global Ocean Prediction with the Hybrid Coordinate Ocean Model

**Figure 3**
Surface (top panels) and 700 m (lower panels) eddy kinetic energy from observations (left panels) and HYCOM over the period 2004-2006 (right panels). The observed surface eddy kinetic energy (upper left panel) is from Fratantoni (2001) and the 700 m eddy kinetic energy (lower left panel) is from Schmitz (1996). The units are in cm$^2$/s$^2$.

**Figure 4**
Modeled analysis of the sea surface height field on September 8, 2008. The white line represents the independent frontal analysis of sea surface temperature observations performed by the Naval Oceanographic Office.
Ocean U.S. GODAE: Global Ocean Prediction with the Hybrid Coordinate Ocean Model

Figure 5

Median bias error (in meters) of mixed layer depth (MLD) calculated from simulated and approximately 66000 unassimilated observed profiles over the period June 2007-May 2008. Blue (red) indicates a simulated MLD shallower (deeper) than observed; 53% of the simulated MLDs are within 10 m of the observation and these are represented as gray. The basin-wide median bias error is -6.6 meters and the RMS error is 40 meters.

The HYCOM data distribution team developed and implemented a comprehensive data management and distribution strategy that allowed easy and efficient access to the global HYCOM-based ocean prediction system outputs to (a) coastal and regional modeling groups, (b) to the wider oceanographic and scientific community including climate and ecosystem researchers, and (c) the general public. The basic idea consists of the setup of a web server that acts as a gateway to backend data management, distribution and visualization applications (http://www.hycom.org/dataserver). These applications enable end users to obtain a broad range of services such as browsing of datasets, gif images, NetCDF files, FTP request of data etc. The 100 Terabytes HYCOM Data Sharing System is built upon two existing software components: the Open Project for a Network Data Access Protocol (OPeNDAP) and the Live Access Server (LAS). These tools and their use to distribute the data are described below. In the current setup, the OPeNDAP component provides the middleware necessary to access distributed data, while the LAS functions as a user interface and a product server. The abstraction offered by the OPeNDAP server also makes it possible to define a virtual data set that LAS will act upon, rather than physical files. An OPeNDAP “aggregation server” utilizes this approach to append model time steps from many separate files into virtual datasets. The HYCOM Data service has been in operation for the last four years and has seen a steady increase in the user base. In the last year, the service received approximately 20,000 hits per month. In addition to the numerous requests from educational institutes and researchers this service has been providing near real-time data products to several private companies in France, Portugal, and the USA.

An important attribute of the data assimilative HYCOM system is its capability to provide boundary conditions to even higher horizontal and vertical resolution regional and coastal models. The current horizontal and vertical resolution for the global forecasting system marginally resolves the coastal ocean [7 km at mid-latitudes, with up to 15 terrain-following (σ) coordinates over the shelf], but is an excellent starting point for even higher resolution coastal ocean prediction systems. Several partners within the HYCOM consortium evaluated the boundary conditions and demonstrated the value added of the global and basin HYCOM data assimilative system outputs for coastal ocean prediction models. The inner nested models may or may not be HYCOM, i.e. the nesting procedure can handle any vertical grid choice. Outer model fields are interpolated to the horizontal and vertical grid of the nested model throughout the entire time interval of the nested model simulation at a time interval specified by the user, typically once per day. The nested model is initialized from the first archive file and the entire set of archives provides boundary conditions during the nested run, insuring consistency between initial and boundary conditions. This procedure has proven to be very robust. Figure 6 shows an example of the ROMS West Florida Shelf domain embedded in the U.S. Navy HYCOM ocean prediction system (Barth et al., 2008).
Ocean U.S. GODAE: Global Ocean Prediction with the Hybrid Coordinate Ocean Model

Outlook

The long term goals of the HYCOM consortium for the global domain are to a) add 3-D and 4-D VAR data assimilation, b) increase the horizontal resolution of the global domain to 1/25°, c) implement zero depths coastlines with wetting and drying, and c) include tides. The scientific goals will include, but will not be limited to a) evaluation of the internal tides representation in support of field programs, b) evaluation of the model's ability to provide boundary conditions to very high resolution coastal models, c) interaction of the open ocean with ice, d) shelf-deep ocean interactions, e) upper ocean physics including mixed layer/sonic depth representation, and f) mixing processes. Other research activities will focus on coupled ocean-atmosphere prediction; bio–geo–chemical–optical and tracer/contaminant prediction; ecosystem analysis and prediction; and earth system prediction (i.e. coupled atmosphere-ocean-ice-land).

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References


ECCO2: High Resolution Global Ocean and Sea Ice Data Synthesis

By Dimitris Menemenlis¹, Jean-Michel Campin², Patrick Heimbach², Chris Hill², Tong Lee¹, An Nguyen⁰, Michael Schodlok¹, and Hong Zhang¹

¹ Jet Propulsion Laboratory, California Institute of Technology, Pasadena, USA
² Massachusetts Institute of Technology, Cambridge, USA

Abstract

The Estimating the Circulation and Climate of the Ocean (ECCO) project was established in 1998 as part of the World Ocean Circulation Experiment (WOCE) with the goal of combining a general circulation model (GCM) with diverse observations in order to produce a quantitative depiction of the time-evolving global ocean state. Such combinations, also known as data assimilation, are important because available remotely sensed and in-situ observations are sparse and incomplete compared to the scales and properties of ocean circulation. These combinations also provide rigorous consistency tests for models and for data. In contrast to numerical weather prediction that also combines models and data, ECCO estimates are physically consistent; in particular, ECCO estimates do not contain discontinuities when and where data are ingested. First generation ECCO solutions are available and widely used for numerous science applications but the coarse horizontal grid spacing and the lack of Arctic Ocean and of sea ice of these first-generation solutions limits their ability to describe the real ocean. To address these shortcomings, the follow-on ECCO, Phase II (ECCO2) project aims to produce a best-possible, global, time-evolving synthesis of most available ocean and sea-ice data at a resolution that admits ocean eddies. A first ECCO2 synthesis for the period 1992–2007 has been obtained using a Green's Function approach (Menemenlis, et al., 2005a) to estimate initial temperature and salinity conditions, surface boundary conditions, and several empirical ocean and sea ice model parameters. Data constraints include altimetry, gravity, drifter, hydrography, and observations of sea-ice. A large complement of high-frequency and high-resolution diagnostics has been saved; these diagnostics are made available to the scientific community via ftp and OPeNDAP servers at http://ecco2.org. This note provides a brief overview of this first ECCO2 synthesis and of some early science applications.

Introduction

Physically consistent estimates of ocean circulation constrained by in situ and remotely sensed observations, as produced by the ECCO project, have now become routinely available and are being applied to myriad scientific applications (Wunsch and Heimbach, 2007). The coarse horizontal grid spacing of current-generation ECCO solutions, however, is a severe limitation on their ability to describe the real ocean, for example, mesoscale eddies, flow over narrow sills, boundary currents, and regions of deep convection and of restratification. Despite the very great progress made toward parameterizing sub-grid scale processes, some problems remain intractable through this route. First, eddy parameterizations are not based upon completely fundamental principles and they fail to adequately account for known anisotropies in their fluxes. Everything we know about eddies suggests that their property fluxes can accumulate in the ocean, changing it measurably and importantly from what it would be if eddies were absent. That is, the long-wavelength, low-frequency features characterizing climate are controlled in part by eddy fluxes. Second, studies have shown that horizontal grid-spacing of order 2 km is required to resolve restratification processes, in which stratified fluid in the periphery of the convection patch is drawn over the surface, allowing the convected fluid to be "swallowed" by the ocean. If restratification is not represented adequately, then the water-mass properties of the modeled ocean deteriorate over time, there are model drifts, and the attendant air-sea fluxes become compromised and have to be "corrected". Restratiﬁcation of mixed layers is a ubiquitous feature of the ocean but is particularly important in strong frontal regions and in regions of deep-water formation. Third, scalar property transports (heat, fresh water, carbon, oxygen, etc.) are of central interest for climate studies and in the ocean, narrow western and eastern boundary currents make major contributions; these boundary currents are not parameterizable and, until they are resolved, there will always be doubts that the ocean model is carrying their property transports realistically. Ultimately, water mass properties in the ocean are important to climate and climate change. In the abyssal ocean, the inability to resolve major topographic features, e.g., fracture zones and sills, leads to systematic errors in the movement of deep-water masses with consequences, for example, on the accuracy of computation of carbon uptake.

Another limitation of current-generation ECCO solutions is that they exclude the Arctic Ocean and that they lack an interactive sea-ice model. This restricts the use of satellite data over ice-covered regions and the usefulness of current-generation solutions for describing and studying high-latitude processes. Coupled ocean and sea ice state estimation is an integral component of the ECCO2 project. The inclusion of an interactive sea-ice model provides for more realistic surface boundary conditions in Polar Regions and allows the model to be constrained by satellite observations over ice-covered oceans. The sea-ice model also provides the ability to estimate the time-evolving sea-ice thickness distribution and to quantify the role of sea-ice in the global ocean circulation. Improved representation of high-latitude processes will enhance hindcasting and forecasting...
ECCO2: High Resolution Global Ocean and Sea Ice Data Synthesis

Model description

ECCO2 data syntheses are obtained by least squares fit of a global full-depth-ocean and sea-ice configuration of the Massachusetts Institute of Technology general circulation model (MITgcm; Marshall et al., 1997) to the available satellite and in-situ data. The computational demands of rigorous ocean state estimation, aka data assimilation, are enormous. Depending on the method and on the approximations that are used, the computational cost of state estimation is several dozen to several thousand times more expensive than integrating a model without state estimation. This has limited the resolution of first-generation ECCO solutions to horizontal grid spacing of order 1 degree (except at the Equator where meridional grid spacing is 1/3-degree in one of the solutions). First-generation solutions also exclude the Arctic Ocean and lack an interactive sea-ice model, which restricts the utilization of satellite data over Polar Regions. Therefore, a necessary condition for a next generation synthesis is an efficient truly global model and significant computational resources.

A first, global ECCO2 solution was obtained on the MITgcm model configuration, which is described in Menemenlis et al. (2005b) and depicted on Figure 1. A cube-sphere grid projection is employed, which permits relatively even grid spacing throughout the domain and which avoids polar singularities (Adcroft et al., 2004). Each face of the cube comprises 510 by 510 grid cells for a mean horizontal grid spacing of 18 km; this is inadequate for fully resolving the processes discussed in the introduction but it is the best that can be achieved at the moment with available computational resources. The model has 50 vertical levels ranging in thickness from 10 m near the surface to approximately 450 m at a maximum model depth of 6150 m. Bathymetry is from the S2004 (W. Smith, unpublished) blend of Smith and Sandwell (1997) and of General Bathymetric Charts of the Oceans (GEBCO) one arc-minute bathymetric grid. The partial-cell formulation of Adcroft et al. (1997), which permits accurate representation of the bathymetry, is used. The model is integrated in a volume-conserving configuration using a finite volume discretization with C-grid staggering of the prognostic variables. The ocean model is coupled to a sea-ice model that computes ice thickness, ice concentration, and snow cover as per Zhang et al. (1998) and that simulates a viscous-plastic...
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rheology using an efficient parallel implementation of the Zhang and Hibler (1997) solver. Inclusion of sea-ice provides for more realistic surface boundary conditions in Polar Regions and allows the system to be constrained by polar satellite observations. The sea-ice model also permits estimation of the time-evolving sea-ice thickness distribution.

Estimation approaches

The first high-resolution global-ocean and sea-ice data synthesis was obtained for the period 1992-2007 by calibrating a small number of control variables using a Green's function approach (Menemenlis, et al., 2005a). The control parameters include initial temperature and salinity conditions, atmospheric surface boundary conditions, background vertical diffusivity, critical Richardson numbers for the Large et al. (1994) KPP scheme, air-ocean, ice-ocean, air-ice drag coefficients, ice/ocean/snow albedo coefficients, bottom drag, and vertical viscosity. Data constraints include sea level anomaly from altimeter data, time-mean sea level from Maximenko and Niiler (2005), sea surface temperature from GHRSSST-PP, temperature and salinity profiles from WOCE, TAO, ARGO, XBT, etc., sea ice concentration from passive microwave data, sea ice motion from radiometers, QuikSCAT, and RGPS, and sea ice thickness from ULS.

Figure 2
Sensitivity maps of sea-ice export through Fram Strait during December 1995 to changes in ice thickness at any point in the domain 1, 2, 3, and 4 years prior. The dominant pattern reflects the advective pathways of sea-ice through the Arctic (Heimbach, 2008)
In parallel with the Green’s function optimization, work is underway towards adjoint-method optimization on the same grid. One objective of the MITgcm ocean and sea-ice model development effort has been to provide the capability for automatic generation of the adjoint model from up-to-date versions of the MITgcm, which is invaluable for ocean state estimation. A major milestone was reached recently with the completion of an adjoint of the full-fledged dynamic/thermodynamic MITgcm sea-ice component. The coupled ocean/sea-ice adjoint now yields stable and physically meaningful adjoint sensitivities or Lagrange multipliers. By way of example, Figure 2 depicts sensitivity maps of sea-ice export through Fram Strait during December 1995 to changes in ice thickness at any point in the domain 1, 2, 3, and 4 years prior (Heimbach, 2008). The dominant pattern reflects the advective pathways of sea-ice through the Arctic. In general terms, points furthest away from Fram Strait are connected to Fram Strait export by faster advective time scales. Time-varying Lagrange multipliers for many other variables, including sea-ice (concentration, thickness, salinity, and velocity), snow (thickness and velocity), ocean (temperature, salinity, and velocity), atmospheric forcing (surface air temperature, specific humidity, precipitation, and wind velocity), and internal model parameters (vertical diffusivity, and bottom drag) are available, and are being analyzed to understand main causes of sea-ice variability.

Mazloff (2008) demonstrated the feasibility of high-resolution adjoint-based state estimation on a regional scale. At this point a preliminary solution (based on 22 iterations) of a 1/6-degree Southern Ocean State Estimate (SOSE) covering the years 2005 and 2006 is available and is being analyzed both in-house and by other research groups. Almost all of the data employed in the first-generation ECCO efforts have been utilized with major emphasis on the satellite altimetry, the Argo profiles, and satellite Sea Surface Temperature (SST). Given the success of this regional, eddy resolving, adjoint-based state estimation effort and of the sea-ice sensitivity experiments, we are currently attempting an adjoint-method optimization on the global, cubed-sphere ocean and sea ice model configuration as a way to increase the number of control variables relative to the existing Green’s function optimization.

In related work, the ECCO2 project is contributing to the development of an open-source Automatic Differentiation (AD) tool called OpenAD. Much effort over the last year has gone into generating efficient adjoint code for the MITgcm that can be applied for large-scale applications. After demonstrating that OpenAD can handle a simplified configuration of the MITgcm we are now in a position to adjoint a global coarse-resolution MITgcm configuration (including the GM/Redi eddy parameterization scheme), which has been a workhorse setup for various adjoint sensitivity studies. It is noteworthy to report that the availability of an adjoint model based on a different AD tool has enabled us to trace a bug in the current “production” AD tool TAF. This is just one aspect of the advantages of having an independent AD tool at our disposal. A major breakthrough is the implementation of a hierarchical checkpointing algorithm, without which adjoint integrations would have been limited to a few days. Using OpenAD, we have conducted a 100-year adjoint integration to compute transient sensitivities of Atlantic meridional heat transport.

**A first ECCO2 synthesis**

The specific objective of the first ECCO2 optimization was to reduce large-scale biases and drifts of the model relative to observations. Figure 3 displays some of the large-scale biases that were present in the baseline integration and the equivalent fields from the optimized solution, demonstrating significant improvements of the optimized solution relative to the baseline integration. Specifically, the top left panel compares the 1992-2002, 0-750-m time-mean temperature of the ECCO2 baseline and optimized solutions to the World Ocean Atlas 2005 (WOA05) climatology. The baseline simulation exhibits a global warm bias of up to 3°C, which is not present in the optimized solution. The top right panel shows baseline (top) and optimized (bottom) Antarctic sea ice distribution on September 1992 (left) and September 2002 (right). Simulated sea is shown in blue. The thin white line is passive radiometer observations of sea-ice extent (15% concentration). The problematic open-water winter polynyas in the Ross and Weddell Seas, which appear in the baseline integration, are no longer present in the optimized solution. The bottom left panel shows Weddell Sea T/S properties of the WGCH climatology (top), of the baseline integration (middle), and of the optimized solution (bottom). The optimized Weddell Sea T/S properties are considerably closer to the observations compared to the baseline integration. The bottom right panel shows Drake Passage transport in the baseline and optimized solutions, which can be compared to the estimate of Sloyan and Rintoul (2001). The Drake Passage transport in the baseline integration is 35 Sv too strong compared to the Sloyan and Rintoul (2001) estimate while that of the optimized solution is much more realistic due, in part, to a much-improved Southern Ocean hydrography.
Evaluation of the first ECCO2 optimization. Top left panel: comparison of the 1992-2002, 0-750-m time-mean temperature of the ECCO2 baseline and optimized solutions to the World Ocean Atlas 2005 (WOA05) climatology. Top right panel: baseline (top) and optimized (bottom) Antarctic sea ice distribution on September 1992 (left) and September 2002 (right). Simulated and optimized sea extent is shown in blue. The thin white line is passive radiometer observations of sea-ice extent (15% concentration). Bottom left panel: Weddell Sea T/S properties of WGCH climatology (top), baseline integration (middle), and optimized solution (bottom). Bottom right panel: Drake Passage transport in the baseline and optimized solution is compared to the estimate of Sloyan and Rintoul (2001)

Despite significant large-scale improvements in this preliminary ECCO2 solution relative to the baseline integration, there remain many problematic aspects, for example, the representation of Mode water formation processes, boundary current separation, Arctic sea ice distribution and hydrography, and ice-shelf-ocean interactions, which are the subject of ongoing work. These shortcomings are being addressed using improved model physics, for example, of deep water formation over continental shelves (Campin and Goose, 1999), of salt plumes rejected by sea ice during freezing (Nguyen et al., 2008), of ice shelf-ocean interactions (Losch, 2008), and of small-scale ice mechanics and thermodynamics (Kwok et al., 2008). As discussed in the previous Section, we are also seeking improved estimates of initial and boundary conditions using the adjoint method, which permits a much larger number of control variables than does the Green’s function approach.

Available ECCO2 products

ECCO2 modeling and estimation tools and results are freely available to scientific community. Model configurations and parameterizations are available at http://mitgcm.org. Automatic differentiation tools are available at http://www-unix.mcs.anl.gov/OpenAD. Finally, modeling and estimation results are available at http://ecco2.org. The ECCO2 results include 1992–2002 quasi-global simulations at 1/4, 1/8, and 1/16-degree horizontal grid spacing (Hill et al., 2007), a 2005–2006 adjoint-method Southern Ocean State Estimate at 1/6-degree horizontal grid spacing (Mazloff, 2008), and a 1992–present global-ocean and sea-ice Green’s function optimization on a cubed-sphere grid with 18 km horizontal grid spacing (Menemenlis et al., 2005b). For the global cubed-sphere grid configuration, in addition to the optimized solution, upwards of 80 forward model sensitivity experiments are available. These experiments, which were used for the Green’s function optimization, explore the model’s response to different surface boundary conditions, initial conditions, horizontal and vertical mixing parameters, sea-ice model parameters, and to the addition of various sub-grid scale parameterizations. There is also a longer, 1979–present, forward integration carried out using the optimized model parameters. For all of these solutions a large complement of high-frequency and high-resolution model diagnostics has been saved. The diagnostics include surface fluxes, sea surface height, bottom
pressure, mixed and mixing layer depths, sea-ice thickness, concentration, salinity, ocean temperature, density, velocity and eddy transports of mass, temperature, and salt. A large portion of these diagnostics (~100 TB) is readily available online via ftp, http, and OPeNDAP servers. The complete diagnostics are stored on tapes at the NASA Advanced Supercomputing (NAS) and are made available upon request.

**Early science applications**

This section lists some early science applications of the ECCO2 products. A first set of applications concerns improved error estimates and eddy parameterizations for coarser-resolution ocean simulations and estimations. Forget and Wunsch (2007) used hydrographic data and an early ECCO2 simulation to estimate global hydrographic variability and data weights in oceanic state estimates. Ponte et al. (2007) used altimeter data and an early ECCO2 simulation for spatial mapping of time-variable errors in Jason-1 and TOPEX/POSEIDON sea surface height measurements. ECCO2 high-resolution simulations have also been used to inform the model parameterization of sub-grid scale processes (Fox-Kemper and Menemenlis, 2008; Danabasoglu et al., 2008).

A second set of early science applications concerns the impact of mesoscale eddies on large-scale ocean circulation and its variability. Fu (2006) used correlation between successive maps of sea-surface height to estimate eddy propagation characteristics. For example, Figure 4 shows the zonally averaged zonal component of the eddy propagation velocity as a function of latitude. The excellent agreement between the model simulation and altimeter observations demonstrates the model’s skill in producing realistic eddy propagation characteristics. Similarities and differences between results from observed and simulated sea-surface height variability improve understanding of model and data errors and of the underlying physical processes. Volkov and Fu (2008) studied the dynamics of the Zapiola Anticyclone, which is situated in a highly energetic area of the ocean, by analyzing the vorticity balance of the anticyclone. This helped to understand the main physical mechanisms that drive the variability of the anticyclone. ECCO2 solutions are also being used to study the impact of eddies on mode water formation. For example, Maze et al. (2008) and Forget et al. (2008) investigated North Atlantic subtropical mode water formation while Davis (2008) studied the formation, evolution, and dispersal of Subtropical Mode Water (STMW) in the North Pacific Ocean.

![Figure 4](image)

Zonally averaged zonal component of the eddy propagation velocity from both model simulation and altimeter observation (Fu, 2006)
A third set of early science applications concerns the study of Polar Oceans. Condon et al. (2008) studied the response of the Arctic freshwater budget to extreme North Atlantic Oscillation (NAO) forcing. Kwok et al. (2006) used ECCO2 estimates of Arctic sea surface height variability to estimate contributions of the oceanographic circulation signal to Ice, Cloud, and land Elevation Satellite (ICESat) retrievals in order to help interpret ICESat altimetric and reflectivity profiles. Kwok et al. (2008) compared sea ice results from one of the ECCO2 solutions and from other coupled-ice-ocean models to observations obtained by the RADARSAT Geophysical Processor System (RGPS). Nguyen et al. (2008) investigated the reasons why coupled ocean and sea ice models in the Arctic tend to misrepresent the upper ocean stratification. Specifically, results from the Arctic Ocean Model Intercomparison Project (AOMIP) showed that participating ocean models consistently failed to either produce and/or maintain the cold halocline layer at the 50–200-m depth. Without a cold halocline, excess heat flux from the warm Atlantic water source at greater depths can inhibit production of realistic sea ice extent and thickness. To address this problem, a new sub-grid-scale parameterization of salt plumes was developed, resulting in a considerably more realistic representation of the cold halocline in the Arctic Ocean (Figure 5).

**Figure 5**

Vertical temperature and salinity profiles of the Canadian Basin in August 2003. CTD measurements are shown in light gray with the data mean shown in dashed heavy black. Blue lines are from a global ECCO2 solution. Red lines are from a regional optimization that includes a sub-grid-scale parameterization of salt plumes (Nguyen et al., 2008)

Finally, ECCO2 results are being used to supply boundary conditions for regional studies and to drive biogeochemical, geodetic, acoustic, and electromagnetic models. For example, Manizza et al. (2008) used ECCO2 results to examine the fate of riverine fluxes of Dissolved Organic Carbon (DOC) in the Arctic Basin. Dushaw et al. (2008) used ECCO2 results to design an acoustic array for observing gyre-scale acoustic variability in the North Atlantic. Glazman and Golubev (2005) used an early ECCO2 simulation to investigate the spatial and temporal variability of the Earth’s magnetic field component induced by ocean circulation.

**Conclusion**

The focus of ocean state estimation during the past ten years has been to demonstrate the feasibility and utility of physically-consistent, global, sustained estimates, with considerable success for upper ocean and for equatorial processes. But many pressing scientific questions, for example, quantifying and monitoring ocean sources and sinks in the global carbon cycle, understanding the recent evolution and variability of the Polar Oceans, and quantifying the time-evolving term balances within and between different components of the Earth System, require much improved accuracy in the estimation of water mass formation and transformation rates, eddy-mixed layer interactions, and high-latitude processes. The accurate monitoring of these processes in turn requires developing state estimation systems, of the sort we have described in this Note, that can fully capitalize on continuing advances in computational and observational technologies.
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References


Simulation of Meso-Scale Eddies in the Mercator Global Ocean High Resolution Model

By Olivier Le Galloudec1, Romain Bourdallé Badie2, Yann Drillet1, Corinne Derval2 and Clément Bricaud1

1 Mercator Ocean, 8-10 rue Hermes, Parc technologique du canal, 31520 Ramonville st Agne
2 CERFACS, 42 avenue Gustave Coriolis, 31057 Toulouse cedex 01

Abstract

The simulation of ocean eddies in the global high resolution ORCA12 model is compared to altimetric observations. At the global scale, the eddy kinetic energy (EKE) of the eddy resolving global ocean model is close to the eddy kinetic energy computed from the geostrophic velocity deduced from altimetric maps. Even if the model is generally overestimating the EKE, the main patterns corresponding to main meso-scale activity areas are well reproduced in term of intensity and geographical position. We have chosen to study particularly six regions relevant of the World Ocean : the Leeuwing Current and the Mozambique Channel for the Indian Ocean, the Alaska current and the Kuroshio for the Pacific Ocean and the Sargasso Sea and the Aghulas Current for Atlantic Ocean. In all these regions, the number of eddy simulated by the model is in good agreement with satellite data. The other result is the high significant correlation between the temporal evolution of the number of cyclonic (and anticyclonic) eddies for the model and observations. The higher correlations (0.8, and more) are found in the Leeuwing Current for cyclonic eddies, in the Kuroshio and in the Sargasso Sea for the both kind of eddies.

Introduction

Mercator Océan is developing a new global high resolution ocean forecasting system which will be the global component of the European MyOcean project. In this paper, study focuses on the validation and on the representation of ocean eddies in the first interannual simulation realized with the global high resolution ocean model. Results are compared to altimetry data which allow both a good representation of the ocean meso-scale activity and tracking of eddy structures like it is mentioned in Aviso web site (http://www.aviso.oceanobs.com/en/applications/ocean/meso-scale-circulation/altimetry-on-eddies-tracks/index.html). As it is the first time that a model allows us to follow eddies in all the world oceans, a brief review of the main ocean eddy formation areas is described by comparison between a “virtual” ocean simulated by the model and the “real” ocean observed by altimetric satellites. In a first part, the model configuration is described. In the second one, the eddy detection algorithm is presented and in the last section, results in 6 areas are commented.

Numerical model: description and validation

The eddy resolving Mercator Océan 1/12°OGCM (here after called ORCA12) is based on NEMO code [Madec, et al., 1998]. The global grid is a quasi isotropic tripolar ORCA grid [Madec and Imbard, 1996], with resolution from 9.3 km at equator to 1.8 km at high latitudes. The vertical coordinates are z-levels with partial cells parameterization [Barnier, et al., 2006]. The lateral diffusion on the tracers (125 m².s⁻¹) is ruled by an isopycnal laplacian operator and a horizontal bilaplacian is used for the lateral diffusion on momentum (-1.25e10 m².s⁻²). The global bathymetry is processed from a combination of ETOPO2v2 bathymetry and GEBCO for the Hudson Bay. Monthly climatological runoffs, from the Dai&Trenberth database, are prescribed [Dai and Trenberth, 2003; Bourdalle-Badie and Treguier, 2006]. The 99 major rivers are spread at mouth and others runoffs are applied as coastal, particularly, along the Antarctic [Jacobs, et al., 1992]. The model is initialised with the recent version of Levitus climatology [Boyer, et al., 2005]. This simulation is forced by the CLIO bulk formulae [Goosse, et al., 2001] using ECMWF analyses from 1999 to 2006. The last 4 simulated years (2002-2006) have been chosen as a significant period to realise statistics and to spin up the surface layer in the ocean. The ORCA12 simulation has been performed on Mercator-Océan SGI computer.

The data base used in this study to validate the ocean meso-scale activity simulated by ORCA12 model, is based on AVISO altimetry [Le Traon, et al., 1998] which contains weekly maps of the global sea level anomaly and the associated geostrophic velocity. The horizontal resolution of these maps is 1/3° which allows a representation of the main me-so-scale eddies, except the smaller one. This point will be discussed in the following parts.
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Figure 1

Eddy kinetic energy (cm$^2$/S$^2$) for the period 2003-2006. Top panel: EKE computed with the surface model velocity. Bottom panel: EKE computed with the geostrophic velocity deduced from the altimetric map of sea surface elevation.

To compare the meso-scale activity in the model and in the observation at the global scale, we have computed the Eddy Kinetic Energy (EKE) with the total velocity of the model surface layer and the EKE with the geostrophic velocities deduced from the altimetric maps. The global EKE (Figure 1) shows the area in the ocean where the meso-scale activity is the more intense. First, the main ocean currents are visible on the two maps with an intense activity in the Gulf Stream and the Kuroshio, in the tropical band and for the southern hemisphere in the Antarctic circumpolar current, all around the Australia, along the South African coasts and in the Argentina basin. In all these areas, the comparison between model and altimetric data shows very similar patterns. We can notice that generally, the model is more energetic than the observations. This is particularly true in the middle of the gyres for each basin. These differences are not studied in this paper which focuses more specifically on the number and size of ocean eddies. Nevertheless, several reasons can explain these differences:

- Considering the model, we used 3 days mean output of the total velocity and the mean surface EKE of the ocean is plotted (Figure 1). For the altimetric data, we used the weekly geostrophic velocity deduced from altimetry.
- EKE computed from the geostrophic velocity or surface velocity are different. In the area where the EKE is weak, like in the middle of the gyre, the geostrophic velocity under estimate the EKE. In eddy propagation area where the EKE is...
strong, the underestimation by the geostrophic velocity is less than 10% (in the Mozambic Channel, in the South east Indian Ocean, along the Alaska Peninsula and in the Aghulas current) but the difference is more important (around 20%) in the Gulf Stream and Kuroshio.

- The horizontal resolution of the altimetric data (1/3°) can’t capture the smallest meso-scale eddies but these eddies are represented in the model as it is explain in the following chapters.
- The model can be too energetic, several parameters can be tuned to correct such biases (like diffusion, viscosity or advection schemes), but at this time, the comparison with other data base (like the surface drifters for example) doesn’t substantiate this thesis.

**Eddies detection**

In this study, the Okubo Weiss criteria [Weiss, 1991] is used for the ocean model output and for the geostrophic velocity deduced from the SLA altimetric data.

The Okubo Weiss parameter is computed thanks to equation (1):

$$
\lambda = \left( \frac{\partial u}{\partial x} - \frac{\partial v}{\partial y} \right)^2 + \left( \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \right)^2 - \left( \frac{\partial v}{\partial x} - \frac{\partial u}{\partial y} \right)^2
$$

where $u$ is the zonal component of the surface current and $v$ the meridional part of the surface current. In this equation, the third term is the relative vorticity of the flow and the two first terms are the deformation of the flow. An ocean eddy is then characterized by this parameter with negative values in the centre of the eddy where the rotation dominates surrounded by positive values at the boundary of the eddy where the deformation dominates. An additional criteria on the sea level anomaly is added to the Okubo Weiss criteria to select only large amplitude eddies. The eddies with amplitude smaller than 15 cm can’t be follow in space and time. Moreover small structures which are not eddies could be detected, especially in the model output. A last criteria is based on the minimum number of pixel in the detected eddies. In the model the minimum number is fixed to 36 pixels, which represent around 3 grid points in an eddy radius, whereas it is only 4 pixels for the altimetric data. The same eddy detection method has been used in previous ocean studies [Henson and Thomas, 2008; Penven, et al., 2005].

**Ocean Eddies: characteristics and statistics**

The eddies have been detected with the Okubo Weiss criteria (previous paragraph) on each map for the period 2004-2006. For the model, a map is a 3-day mean and for the altimetric data, a map is a 7-day mean merging all available altimetric satellites. Six areas have been selected to perform the study of the meso-scale activity. Two of them are in the Indian Ocean (West coast of Australia, Mozambique channel), two in the Atlantic Ocean (West of south Africa and Sargasso Sea) and two in the Pacific Ocean (Alaska and Kuroshio regions). The mean number of eddies (Table 1) represents the mean number of eddies per map for all the period and over the selected domain described in each paragraphs. As the range of eddy scales detected in the model is wider than in the observations (resolution in the observations is coarser than in the model), we also computed the number of eddy with radius larger than 30 km (smallest scale detected in the observations.)To represent the spatial distribution of the eddy field in the model and in the observation, the probability of occurrence of an eddy in 1° x 1° boxes for the 4-year period boxes has been computed and presented in the figure 2 to the figure 7.

Informations about the size of the eddies are also provided (Table 1) with the percentage of eddies with a radius between 30 to 60 km which are the more common size of eddies in the study areas. Last, the proportion of anticyclonic eddies of the total number of eddies in the model and in the observation is compared.

For each studied area, the evolution of the eddy number (total, cyclonic and anticyclonic) have been compared and correlation between simulated and observed eddies using 21 days smoothed time series (Table 2) have been computed.

**Leeuwing Current**

The Leeuwing current is a warm and fresh ocean surface current which flows southward along the western Australian coast. The eddies are formed all along this current from north (around 22°S) to south (around 35°S) by barotropic and baroclinic instabilities. These anticyclonic and cyclonic eddies are advected in the Indian Ocean after the separation from the Leeuwing Current. The probability of eddy occurrence (Figure 2) illustrates the geographical repartition of the ocean eddies in the altimetry and in the model. Occurrence larger than 15% (and even larger than 20% in the model) represents the eddy formation place. In the eddy pathway in the Indian Ocean, the occurrence is larger than 10%. The eddies are formed in two places around 30°S and 37°S south westward from Australian coast. These eddies are thereafter advected in the Indian Ocean following a pathway
between 20°S and 30°S. In this area, the counting of eddy (Table 1) and the time correlation between model and observed eddies (Table 2) are realised on a box bounded from 71°E to 129°E and from 39°S to 20°S.

Figure 2
Spatial distribution of the probability of eddy occurrence computed by 1x1° boxes in Leeuwing current and southeastern Indian Ocean.

Morrow, et al., [2004] has described characteristics of these eddies detected by the altimetry data, the results of our study and the comparison with the ORCA12 simulation is in good agreement with this previous study. The mean number of eddies (around 40 per map) in the model is comparable to the altimetry (Table 1) with more than half with a radius smaller than 60 km. In the model as in the altimetry data, the number of anticyclonic eddies are higher than the cyclonic one (Table 1). The number of large anticyclonic eddies (radius larger than 60 km), in the model as in the altimetry, is larger than the number of cyclonic eddies. For the smallest structures (radius smaller than 60 km), the proportion of cyclonic and anticyclonic eddies are equivalent. A strong seasonal cycle, with a maximum value in spring (September to November in the southern hemisphere) is observed for the cyclonic eddies (not shown in this paper). For the anticyclonic one, the seasonal cycle is less obvious and is not in phase with the time series of the number of cyclonic eddies. Two maxima are observed during fall (May to June) and summer (January).

The number of anticyclonic (respectively cyclonic) eddy correlation between model and altimetry for the 4 years of study is 0.7 (respectively 0.8, table 2).

Mozambique Channel

The region around Madagascar Island is a region of strong meso-scale activity (Figure 1). It can be split in two domains: East of Madagascar and Mozambique Channel. These two regions feed the Agulhas current. Biastoch and Krauss, [1999] have estimated the transport in the Agulhas current at 65 Sv in the upper 1000m, 5 Sv coming from the Mozambique Channel and 20 Sv from the East of Madagascar. The observations show maxima of EKE are reached in these areas. The model reproduces very well this pattern. The major difference is the level of EKE in the north of the Mozambique Channel, which is more intense in the model. The box selected to perform eddy statistics is 30°E-60°E; 34°S-10°S. The number of eddies over the period is quite the same in the altimetric data and in ORCA12 (around 35, see table 1). Figure 3 shows a quite homogeneous 20% probability to find eddies in the model over the Madagascar area, whereas the proportion is more important in the Mozambique Channel (near 18%) than east of Madagascar (about 13%) in the altimetric data. In the model, there is a lack of eddies in the area around 40°E; 35°S (5%) compare to the altimetric data (10%). It appears clearly that preferential path are more localized in the current trajectory in the model output than in the observations, where eddies are widely spread. The number of anticyclonic eddies is more important in altimetric data (63%), on the contrary to the model (proportion of anticyclonic is 44%). A very strong seasonal cycle, both in observation and model, depending on the monsoon, is present, with a maximum in January and the correlation with the anticyclonic eddies is significant (0.68).
Aghulas Current

The Aghulas current is one of the more energetic currents in the global ocean. It takes source in the Indian Ocean and follows southward the south eastern African coast. Then, this current leaves the shelf, retroreflects and flows backward in the Indian Ocean. The retroreflection is located between 20°E and 15°E (Figure 1). Here warm eddies, called Aghulas rings are formed by loop occlusion. These anticyclonic and cyclonic eddies are advected in the South Atlantic Ocean over several thousand kilometres [Treguier, et al., 2003; Biastoch and Krauss, 1999].

The box selected to compute the statistics on the meso-scale activity in this area is 10°W-20°E; 42°S-20°S. The number of eddies during the studied period is of the same order than in the observations (25 eddies per map). In the data, more anticyclonic eddies are observed (73%) but in the model the proportion (52%) is quite the same (see table 1). The penetration of the anticyclonic (35°W) and cyclonic (20°W) eddies in the simulation is in good agreement with data. Figure 4 shows that the anticyclonic (cyclonic) eddies drift north-westward (south-westward) in the model as in the observation. A preferential path near 25°-30°S for anticyclonic eddies can be identified in the model, with occurrence of eddy between 5 and 10% along this pathway. In the altimetric data, anticyclonic eddies are observed between 25°S-35°S with a maximum at 33°S (not shown in this paper). The eddy number seasonal cycle is not well marked in both observations and model. The correlation is low (0.4) for the anticyclonic eddies because of a phase lag with a maximum in April for the model and in February for data.

Sargasso Sea

The Sargasso Sea is crossed by a south-westward current that flows between the Gulf Stream and the Bermuda. This near-surface flow drifts westward the Cold Core Rings (CCRs), which pinched form the Gulf Stream. We also can find in the Sargasso Sea others eddies eastward of the Gulf Stream, generated from baroclinic instabilities in the flow field. Using in situ measurements during the period 1996-2004, Luce and Rossby, [2008] found the CCRs with a typical radius of 57km +/- 16 km, in a band from 150 to 300 km of the Gulf Stream. They also found coherent vortices due to baroclinic instabilities with radius of
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64 +/- 18 km. In this study, statistics are realized in a box bounded from 81°W to 59°W and from 26°N to 37°N. The number of eddies in ORCA12 is the same than in altimetry data (17 eddies per map), but more eddies are created in the modelled meanders of Gulf Stream (as we will see below in the Kuroshio region) with less eddies in the south of Sargasso Sea in the simulation than in the data (figure 5). The correlation between the time series of number of eddies detected in altimetry and model is significant for both cyclonic and anticyclonic eddies (respectively 0.88 and 0.83, see Table 2).

**Figure 5**
Spatial distribution of the probability of eddy occurrence computed by 1x1° boxes in the Aghulas current.

**Alaska Current**

The circulation in the Gulf of Alaska is dominated by a wind-forced gyre in the ocean basin bounded southward by the North Pacific Current. It splits as it approaches the North American continent to form the equatorward California Current and the poleward Alaska Current. The Alaska Current turns south-westward at the head of the Gulf (56°N 145°W), and becomes a narrow, swift stream which closely follows the shelf break. A portion of the Alaskan Stream turns southward near the Aleutian Islands (165°W, 53°N) and recirculates as part of the North Pacific Current, closing the loop of the Alaska Gyre. A large part of eddies is generated on the path of the gyre, between the Queen Charlotte Islands (132°W, 53°N) and the eastern bound of the Gulf. The repartition of the ocean eddies (figure 6) in the altimetry and in the model confirms this point, with around 15% of occurrence of eddy for several 1°x1° boxes at this position. These eddies are thereafter advected along the Alaska Peninsula and the Aleutian Islands. In this area, the statistics are realized on a box bounded from 179°W to 114°W and from 47°N to 61°N. By analysing the altimetry maps, Henson and Thomas, [2008] has observed, a high proportion of anticyclone (about 85%) among the eddies. Even if the studied period is not the same, we obtain the same order of anticyclonic eddies with 78% in the altimetry data and 77% in the model. The seasonal cycle of anticyclones formation is marked, with maximum in summer as in Henson and Thomas, [2008]. The correlation between model and altimetry for the number of anticyclonic eddies during the 4 years of study is 0.7 (table 2).

**Figure 6**
Spatial distribution of the probability of eddy occurrence computed by 1x1° boxes in the Alaska current.
Kuroshio

In the Kuroshio extension, south of Japan, the meso-scale activity is important. Cyclonic and anticyclonic eddies are formed in the meander of the Kuroshio and interact with the current. Several studies have been realised in this area especially south of Japan for example in Ebuchi and Hanawa, [2000] or more southward in the China Sea [Chow, et al., 2008]. The studied area is bounded by 120°E to 160°E and 20°N to 36°N. It includes the starting point of the Kuroshio (north of the China Sea) to the Kuroshio extension in the North Pacific. The same number of eddies (Table 1) are detected in this area in the ORCA12 simulation (46 eddies per map) and in the altimetry data (43 eddies per map), Ebuchi and Hanawa, [2000] obtain the same result based on altimetry. But location of the eddies in this area are different. In the ORCA12 simulation, eddies are mainly situated in the Kuroshio with occurrence larger than 25% (figure 7). In the meander of this current, eddies are mainly anticyclonic (not shown) but they don’t systemically detached from it. These anticyclonic structures have a short lifetime (less than 1 month for most of eddies.) They are formed at the end of summer or in fall (from September to November) and they rapidly disappear in the mean flow of the Kuroshio. The minimum number of anticyclonic eddies is smaller in the model compare to the altimetry (around 10 for ORCA12 compare to 15 for the altimetry during winter, not shown) but the maximum of anticyclonic eddies is larger in the simulation (larger than 40 in ORCA12 and around 35 in altimetry, not shown).

In ORCA12 simulation, eddies are mainly smaller than 60 km against 90 km in the altimetry data (see table 1).

The correlation between simulation and data are both 0.8 for the anticyclonic and cyclonic eddies, that means that the seasonal cycle, which is the main signal on the temporal evolution, is correct. We can notice that the correlation for the total number of eddies is in this case 0, there is no seasonal cycle for the total number of eddies in the model and in the observation. This is explain by the seasonal cycle for the number of cyclonic and the anticyclonic eddies which is in opposition of phase.

Figure 7

Spatial distribution of the probability of eddy occurrence computed by 1x1° boxes in the Kuroshio region.
Simulation of Meso-Scale Eddies in the Mercator Global Ocean High Resolution Model

### Table 1

<table>
<thead>
<tr>
<th>Area</th>
<th>Number of eddies</th>
<th>% of eddies between 30 to 60 km for eddies &gt;30km.</th>
<th>% of anticyclonic eddies/total number of eddies</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Alti</td>
<td>Orca12 (&lt;\text{min alti})</td>
<td>Alti</td>
</tr>
<tr>
<td>Leeuwing</td>
<td>40.7</td>
<td>55.6</td>
<td>44</td>
</tr>
<tr>
<td>Mozambique</td>
<td>34.2</td>
<td>44.3</td>
<td>35.4</td>
</tr>
<tr>
<td>Alaska</td>
<td>12.2</td>
<td>22.7</td>
<td>17.6</td>
</tr>
<tr>
<td>Kuroshio</td>
<td>43.2</td>
<td>60.4</td>
<td>46.5</td>
</tr>
<tr>
<td>Sargasso</td>
<td>16.9</td>
<td>22.2</td>
<td>16.4</td>
</tr>
<tr>
<td>Aghulas</td>
<td>24.5</td>
<td>31.4</td>
<td>25.5</td>
</tr>
</tbody>
</table>

Eddy statistics in each area. The number of eddies is the mean number of eddies in the area per map. The column ORCA12>\text{min alti} is the number of eddies in ORCA12 when we omitted the eddies which are smaller than the smaller eddy in the altimetry.

### Table 2

<table>
<thead>
<tr>
<th></th>
<th>Leeuwing</th>
<th>Mozambique</th>
<th>Alaska</th>
<th>Kuroshio</th>
<th>Sargasso</th>
<th>Aghulas</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cyclonic eddies</td>
<td>0.83</td>
<td>0.44</td>
<td>0.6</td>
<td>0.8</td>
<td>0.88</td>
<td>0.57</td>
</tr>
<tr>
<td>Anticyclonic eddies</td>
<td>0.71</td>
<td>0.67</td>
<td>0.68</td>
<td>0.8</td>
<td>0.83</td>
<td>0.4</td>
</tr>
<tr>
<td>Total eddies</td>
<td>0.6</td>
<td>0.13</td>
<td>0.66</td>
<td>0.0</td>
<td>0.4</td>
<td>-0.37</td>
</tr>
</tbody>
</table>

Correlation coefficient between the time serie of the eddy number (cyclonic, anticyclonic and total) detected in the altimetry and in the model. The correlation is computed on the a time serie filtered at 21 days. The eddies in ORCA12 simulation smaller than the smallest eddy in the altimetry are removed from this statistic.

### Conclusion

The main conclusion of this study is the really good potentiality of ORCA12 model to simulate the meso-scale activity and particularly the ocean eddies. The number and the geographical distribution of eddies, in all the studied areas, are in good agreement with altimetric observations. The seasonal cycle of the number of anticyclonic and cyclonic eddies are also comparable to the altimetry. These two points are of great importance for the qualification of this simulation to provide realistic informations for the assimilation scheme used in Mercator-Océan forecast systems. This assimilation scheme based on the SEEK filter [Testut, et al., 2003; Tranchant, et al., 2008] needs 3D mode data base. These modes will be computed from the ORCA12 forced simulation. They have to represent the ocean meso-scale variability at time scale from one week to the seasonal cycle. But we can notice one of the biases in the model. In all the area, except in the Alaska current, the model seems to produce an equivalent number of cyclonic and anticyclonic eddies whereas the proportion is generally not equivalent in the altimetric data. In ongoing work, other diagnostics would be realized to characterize the ocean eddies in the model, particularly the 3D geometry of eddies and the associated heat and salt transport in the ocean.
Acknowledgements

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Ocean Data Assimilation and Prediction system in JM and MRI

By: Masafumi Kamachi, Yosuke Fujii, Norihisa Usui, Shiroh Ishizaki, Satoshi Matsumoto and Hiroyuki Tsujino

Meteorological Research Institute, Japan Meteorological Agency

Abstract

Ocean data assimilation/prediction systems MOVE/MRI.COM are introduced. The systems have been developed in Meteorological Research Institute, and been operated in Japan Meteorological Agency. The operation is for ocean state estimation and prediction of ocean weather (mesoscale) and of seasonal to interannual (e.g., ElNino) phenomenon. This report shows briefly the structure of ocean model, assimilation method and some examples of ocean state (velocity and water mass) and prediction of the Kuroshio path. Future perspective is also briefly stated.

Introduction

The Meteorological Research Institute (MRI) Multivariate Ocean Variational Estimation (MOVE/MRI.COM), Systems have been developed for research system to study mesoscale phenomenon, climate variability and for the operational ocean data assimilation system in Japan Meteorological Agency (JMA).

The MOVE/MRI.COM systems have three components: the global (MOVE/MRI.COM-G), North Pacific (MOVE/MRI.COM-NP) and Western North Pacific (MOVE/MRI.COM-WNP) systems (Usui et al., 2006a). The aims of MOVE/MRI.COM-G are the initialization of MRI coupled GCM for seasonal-interannual forecasting and analysis/reanalysis project, which is related to CLIVAR/GSOP project. The period of the analysis/reanalysis product is 1948 to 2007. The aims of MOVE/MRI.COM-NP and -WNP are the initialization of ocean forecast in the North pacific (especially around Japan) and analysis/reanalysis, which period is 1993-2007 and which is related to GODAE project. Information about the systems and recent publications are given in the MRI homepage http://www.mri-jma.go.jp/Dep/oc/oc.html. Here we introduce the system’s configurations and some demonstration results.

MOVE/MRI.COM

MOVE/MRI.COM systems are composed of OGCMs and a variational analysis scheme which synthesizes the observed information (i.e., temperature, salinity and sea surface height) together with the OGCMs.

Model

The numerical code for the OGCMs used in the MOVE/MRI.COM system is the MRI community ocean model (MRI.COM). MRI.COM has been developed in JMA/MRI and is independent of any other popular OGCM code (Ishikawa et al. 2005, Tsujino et al., 2006). It is a multilevel model code that solves the primitive equations under the hydrostatic and Boussinesq approximations. The vertical coordinate is a terrain following-depth (σ−z) hybrid, i.e., the levels near the surface follow the surface topography. It enables us to adopt a fine vertical resolution near the surface because it prevents the uppermost layer from vanishing during integration when the free surface variation is explicitly solved. For momentum advection, MRI.COM uses the generalized enstrophy-preserving scheme with the Takano–Oonishi scheme, which contains the concept of diagonally upward/downward mass momentum fluxes along a sloping bottom (Ishizaki and Motoi, 1999).

The OGCM used in MOVE/MRI.COM-G is a model with a global domain (model G) for climate variability. On the other hand, for ocean weather, MOVE/MRI.COM employs two models, namely the North Pacific and western North Pacific models (models NP and WNP). Model WNP is nested into model NP, i.e., the boundary conditions for the western, eastern, and southern boundaries in model WNP are passed from model NP (one-way nesting). Daily outputs from model NP are linearly interpolated both in time and space to replace boundary data of model WNP at every time step.

The domain of model G extends from 75°S to 75°N glo bally. The grid spacing in the zonal direction is 1° and that in the meridional direction is 0.3° within 5°S–5°N, 1° pol eward of 15°S and 15°N and gradually changed from 0.3° to 1° in the transition regions (i.e. 5°S-15°S, 5°N-15°N). There are 50 levels in vertical. The bottom topography is based on ETOPO5. The northern boundary is closed. The isopycnal diffusion (Redi, 1982) and the isopycnal thickness diffusion (Gent and McWilliams, 1990; hereafter GM), the background coefficients for vertical diffusion by Tsujino et al. (2000), the harmonic viscosity with the parameterization of Smagorinsky (1963, hereafter SMA63), and the level-2.5 turbulent closure scheme of Mellor and Yamada
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(1982) (hereafter MY2.5) are all used in model G. Figure 1 shows the geographical relation of the global, North Pacific and western North Pacific models.

Figure 1
Geographical relationship of Global (MOVE-G), North Pacific (MOVE-NP), and Western North Pacific (MOVE-WNP) models.

Monthly mean SST in December 2007 is shown in the panel of model G. Bottom topography is shown in the panel of model NP and the assimilation result (distribution of surface current speed in the 1st July 2004) is also shown in the panel of model WNP.

The model is driven by daily wind stress, heat flux, and fresh water flux fields calculated from NCEP–NCAR reanalysis (NCEP R-1; Kalnay et al., 1996). Latent and sensible heat flux fields are calculated with the bulk formula of Kara et al. (2000) (hereafter Kara00). The solar heat flux penetrates surface layers according to Paulson and Simpson (1977). The fresh water flux of NCEP R-1 is adjusted by adding time-independent (space-dependent) flux correction terms (Vialard et al. 2002).

The domain of model WNP extends from 15° N to 65° N, and 117° E to 160° W, with a grid spacing of 1/10° × 1/10° around Japan (15° N to 50° N, and 117° E to 160° E). Outside of the region, the grid spacing is a little coarse: zonal grid spacing from 160° E to 160° W is 1/6°, and meridional grid spacing from 50° N to 65° N is 1/6°. This model is nested into model NP, which model region extends from 15° S to 65° N, and 100° E to 75° W with a grid spacing of 1/2° × 1/2° (Figure 1). Model NP and WNP have a northern boundary located at the same latitude. The northern boundary of WNP is closed. All lateral boundaries of model NP are closed although they are planned to be nested in model G. Models NP and WNP (Figure 1) have the same vertical grid spacing (54 levels). Bottom topographies in models NP and WNP are based on Smith and Sandwell (1997). The configurations of models NP and WNP are the same as model G, except for a few differences. The biharmonic viscosity with a parameterization based on SMA63 (Griffies and Hallberg, 2000), instead of harmonic viscosity, and the turbulent closure scheme of Noh and Kim (1999), instead of MY2.5, are used. Biharmonic diffusion, instead of isopycnal diffusion in G model, is used in model WNP. The models are driven by the daily wind stress and flux fields calculated from the NCEP–DOE AMIP-II reanalysis (NCEP R-2; Kanamitsu et al., 2002). Models NP and WNP adopted the bulk formula of Kondo (1975), instead of Kara00, for latent and sensible heat fluxes. The water flux is corrected by restoring SSS to the climatology with a restoring time of 1 day (the time-independent correction term is not applied). A sea ice model with the elastic-viscous-plastic dynamics of Hunke and Dukowicz (1997) and the thermodynamics of Mellor and Kantha (1989) (EVP sea ice model) is also applied in models NP and WNP.
Assimilation system

The analysis fields for models G, NP and WNP are calculated separately. The analysis scheme adopted in the MOVE/MRI.COM system is a multivariate 3DVAR with vertical coupled T–S EOF decomposition of a background error covariance matrix. The scheme is based on Fuji and Kamachi, 2003a, 2003c and Fuji et al., 2005. The amplitudes of the coupled EOF modes are employed as control variables and the analyzed temperature and salinity fields are represented by the linear combination of the EOF in the scheme.

A preconditioned optimizing utility for large-dimension analysis (POpULar; Fujii and Kamachi, 2003b and Fujii, 2005) is developed and applied for minimizing the nonlinear cost function as the descent method. This scheme can minimize a cost function including a constraint of the background without inversion of the background error covariance matrix, even if the function is nonlinear. It is useful for handling the correlation among background errors.

The G, NP and WNP domains are divided into 40, 12 and 13 subregions, respectively. EOF modes are calculated in each subregion for each model from world ocean database 2001 (WOD2001), as well as the representativeness error covariance matrix, according to Fuji and Kamachi (2003b). We retained 12 dominant modes in each subregion. In fact, more than 85% of the total variance can be explained by the dominant 12 modes although this estimate will differ from one in different subregions. A Gaussian function is adopted as the horizontal correlation model applied in the background covariance matrix B. The e-folding scales along latitude and longitude lines are also different depending on the subregions and are computed from Kuragano and Kamachi (2000).

The model temperature and salinity fields are corrected by the analysis results through the Incremental Analysis Updates (IAU) technique (Bloom et al., 1996). The assimilation period is 1/3 month.

Temperature, salinity and along-track SSH observations are employed in the analysis. The temperature and salinity observations were collected from WOD2001 which is the original/raw data set of the climatological data set of WOA2001 and from the global temperature–salinity profile program (GTSPP) database. We also use the along track SSH anomaly data of TOPEX/Poseidon (T/P), Jason, ERS, ENVISAT after adding it to the mean SSH calculated from a preliminary analysis using temperature and salinity observations alone.

Experimental conditions of analysis/reanalysis and prediction

The assimilation experiment (analysis/reanalysis) was conducted from January 1948 to December 2007 for global and North Pacific systems, and from January 1993 to September 2007 for western North Pacific system.

Hereafter we introduce the western North Pacific version (MOVE/MRI.COM-WNP). The assimilation period is 1/3 month: the first and second assimilation periods in a month are 10 days and the third one varies from 8 to 11 days. Temperature and salinity profiles above 1500 m depth and SSHA (Sea Surface Height Anomaly) data are assimilated. Temperature and salinity data are collected from WOD2001 and Global Temperature-Salinity Profile Program (GTSPP) database. The SSHA data is the along track data from the TOPEX/Poseidon, Jason-1, ERS-1/2 and ENVISAT altimeters, which are extracted from the SSALTO/DUACS delayed time multimission altimeter products (CLS 2004). Latent and sensible heat fluxes are re-calculated in the model using model sea surface temperature (SST) and the bulk formula of Kondo (1975).

138 cases of prediction experiments for the Kuroshio path variability south of Japan were conducted from February 1993 to July 2004. Predictions start at the first day of every month and are integrated for 90 days. The wind-stress and heat fluxes used in the prediction experiments are NCEP R-2, the same as in the assimilation experiment. We should treat an external forcing as an unknown factor in the prediction. However, we treat it here as a known factor because our objective is to assess the predictive skill of the assimilation scheme and the dynamical model when a perfect external forcing is given. The predictive skill obtained from this protocol could be affected by the use of predictive forcing.

Analysis/Reanalysis and Prediction Results

Here we introduce some examples of the characteristics of the analysis/reanalysis and prediction experiments.

We give here one example of product validation concerning velocity field. Figure 2 shows a comparison of the velocity field from the assimilation results and independent JMA’s ADCP observation in January, 2001. Velocity field of the assimilation shows the correct Kuroshio path as well as positions and strengths of mesoscale eddies, whereas the positions and strength of the smaller scale eddies are rather shifted. The correlation coefficient of the zonal (meridional) velocity between the two datasets is 0.84 (0.47).
In the Western North Pacific, there are two typical shallow water masses: the warm and salty Kuroshio water in the subtropical gyre and the cold and fresher Oyashio water in the subpolar gyre. These two water masses merge and produce many mesoscale eddies at the boundary of the subtropical and subpolar gyres in the area east of Tohoku. Figure 3 shows temperature and salinity distributions in the meridional section along 144°E of the JMA’s line measurement. Figure 3 (a) and (c) show the temperature field. Symbol ST in the figures indicates the warm and saline Kuroshio water and its frontal structure south of 36°N and Kuroshio warm water eddy between 38°N and 39°N. The Kuroshio warm water eddy is surrounded by the cold Oyashio water (symbol SP). In figures 3 (b) and (d), salinity distributions highlight the Kuroshio warm eddy structure with salty water (ST) surrounded by Oyashio fresh water.
Comparison of temperature and salinity distributions along 144E of Japan Meteorological Agency repeated hydrography section. (a): temperature (assimilation), (b) salinity (assimilation), (c): temperature (independent observation), (d): salinity (independent observation). ST: subtropical water, SP: subpolar water.

Figure 3

Figure 4 shows salinity distributions in the meridional section along 165°E of the JMA’s observation line. It shows a typical water mass from surface to mid depth (600m to 800m.) This is the North Pacific Intermediate Water (NPIW), which is characterized by a minimum salinity core near 600m – 800m depth and is important to understand Pacific decadal oscillation. We can see that the assimilation result reproduces quite well the salinity distribution of NPIW. This is an example of the efficiency of the assimilation scheme (multivariate 3DVAR) to represent not only temperature but also salinity fields of the subtropical (Kuroshio) and subpolar (Oyashio) waters in the upper layers and NPIW distribution at mid depth (Usui et al. 2006a, 2008a,b).
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Mean water mass property (temperature and salinity) along JMA’s hydrographic observation lines around Japan are also examined. Figure 5 shows JMA’s observation lines of the repeated hydrography. Figure 6 shows the climatological mean water mass property calculated from the assimilation of the JMA’s observed data into MOVE/MRI.COM-WNP. Though the JMA’s observation data are assimilated (i.e., not independent data), the assimilation (analysis/reanalysis) results give a good representation of the mean water masses along almost all lines. Some sections have a larger bias. PH line (figure 5) in the subtropical-subpolar boundary region along 41.5°N has the largest bias. This is difficult to reproduce the ocean state/water mass into this area even if the water masses of the Kuroshio and Oyashio are well recovered (figure 3). PM line in the Japan Sea shows also large bias near the bottom. The structures, variations and trend (which is related to climate change) are examined with each water mass in the Pacific. It explains the causes of the variation, propagation path and changes of water property in formation regions and relation with the atmospheric conditions. It is an on-going study (see Matsumoto et al., 2008).

Figure 4
Salinity distribution along 165E (Japan Meteorological Agency repeated hydrography section). (a): assimilation, (b): JMA independent observation. The both panels (a) and (b) show composite distributions of salinity. Observation/assimilation period is in September (April), 2000 in each left (right)-half of the panels.
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Figure 5
JMA’s repeated hydrography sections

Water Mass Compared with Observation

Figure 6
Comparison of water property (temperature and salinity) on T-S diagram along JMA’s hydrographic sections (see Figure 3 for the section names and the positions)
Using the analysis/reanalysis datasets, prediction experiments are conducted. Results of the prediction are shown in figures 7 and 8. Figure 7 shows the comparison of the large meander Kuroshio path from prediction and assimilation experiments in 2004. The small meander, located in the south of the Kii Peninsula at the beginning of July 2004 (figure 7a), grows as it propagates eastward and develops into a large meander path (figures 7c and e) from the middle to the end of August 2004. After this period the Kuroshio sustains the large meander path. The developing process of the large meander path is successfully predicted. An ensemble of 138 cases of the prediction are compared with persistency and climatological variability of SSH field. Notice that persistency is the state of the initial condition continues in the prediction period. Figure 8 shows a time evolution of the ensemble mean SSH RMS difference between forecast and assimilation (blue bold line) and persistency and assimilation (red line) in the rectangular region (131°E-140°E, 30°N-35°N). The surrounding thin blue lines correspond to the standard deviation envelope from the 138 cases. We note that seasonal changes in the SSH field are expressed in the persistency because SSH anomaly is defined here as the anomaly from the 1/3 monthly mean SSH. Eliminating the seasonal variation in the assimilation field and using standard deviation of SSH of 15.3 cm depicted in figure 8 as a threshold for a predictive limit, we find that the predictive limit of our system is roughly 40-60 days, which is much longer than that of the persistency (see Usui et al. 2006b, 2008a,b for the dynamics of the meandering and variability of the Kuroshio).

**Figure 7**

Prediction of the 2004 Kuroshio large meander. Color bar shows the velocity intensity (cm s⁻¹). (a): initial condition (1 July, 2004), (b): assimilation (25 July), (c): assimilation (25 August), (d): 25 days prediction (25 July), (e): 55 days prediction (25 August)
Summary and Perspectives

We have introduced here the configuration of the MRI ocean data assimilation systems (MOVE/MRI.COMs). The systems are operated in JMA from March 2008. The system-G is used for El Nino forecast and the NP and WNP systems provide ocean mesoscale forecasting. Some examples of the comparison of assimilation and observation about velocity and water mass distributions are introduced. The prediction experiments shows 40-60 days predictability in the Kuroshio region.

Improvements and developments of the model and assimilation schemes in the MOVE/MRI.COM systems are ongoing work at MRI. On-going projects are OSE/OSSE type experiment, sensitivity experiment with adjoint version of MOVE/MRI.COM-WNP, ensemble initialization for El Nino prediction. JMA’s operational and MRI’s research groups also have planed of the developments of coastal/shelf sea applications and air-sea coupled assimilation. Project of coastal/shelf sea model/assimilation system development contains development of sea ice assimilation, study of wind wave-current interaction, model development with high resolution (e.g., 1/64°), two-way nesting method and influence of higher resolution information into lower resolution model, boundary condition sensitivity experiment of coastal current variations in a higher resolution model (such as Tokyo Bay model. This is affected by the Kuroshio path in the lower resolution assimilation result, see Kamachi et al., 2008).

References


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Contact:
Please send us your comments to the following e-mail address: webmaster@mercator-ocean.fr

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