#### Development and evaluation of ice-ocean reanalyses using the S(T) assimilation system

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Boning

#### Dalhousie, February 24th, 2009











## Overview

- Ocean Reanalysis
  - Uses of reanalysis
  - Challenges of historical datasets
- Assimilation on isotherms
  - Depth versus temperature level assimilation
  - Impact of Argo assimilation
- New global ice-ocean reanalyses
  - Mean biases
  - Water mass properties
  - Transports

#### Uses of ocean reanalysis

- Initialization of seasonal forecasts:
  - Seasonal: ECMWF (Vidard et al. MWR, 2007)
  - Decadal : UK Metoffice DePreSys (Smith et al., Science, 2007)

#### • Climate signals:

- Sea level rise:
  - Wunsch et al. (J. Clim., 2007)
- Ocean heat content:
  - Carton and Santorelli (Submitted to J. Clim.)
  - Kohl and Stammer (JPO, 2007)
- Meridional overturning circulation:
  - Wunsch and Heimbach (JPO, 2006)
  - Balmaseda et al. (GRL, 2007)

#### CLIVAR Global Synthesis and Observations Panel (GSOP) Intercomparison

#### • Main goals:

- Evaluate quality and skill of existing global synthesis products (reanalyses) for climate applications
- Determine common strengths and weaknesses and their usefulness for various climate applications

#### Reanalyses included:

- ECCO-GODAE, ECCO-JPL, GECCO, ECMWF, SODA, CERFACS, MERCATOR, INGV, MOVE, GFDL, Reading, UKDP, UKOI
- Includes model resolutions from 2 to ¼ degree
- Range of assimilation methods (e.g. OI, KF, 4DVAR)
- Some span last several decades, although most only cover recent period (1992 onwards)

www.clivar.org/organization/gsop/projects.php

#### Global ocean heat content variability

Carton and Santorelli, submitted to J. Clim.



## Challenges posed by historical ocean datasets

- Satellite:
  - SST, Sea level
- Buoy, tide gauge, drifters
- XBT
  - Bias problems with fall rate (Wijffels et al. 2008)
  - Only near-surface (top 300-500m)
  - Poor spatial distribution (localized to ship tracks)
- CTD casts and moorings
- Argo:
  - Autonomous profiling floats
  - Near-global coverage of T,S over upper 2000m
  - Radical improvement in subsurface ocean sampling beginning around 2002







#### Challenges posed by historical ocean datasets

June, 1975





120°E

60°E

120°E

180°W

30°S

60°S

180°W

#### Argo radically improves:

- spatial sampling
- salinity observations

#### Questions:

- 1. How can we best make use of observations prior to Argo?
- 2. Can Argo help us with this?

# How can we best make use of the available observations?

# Two types of variability: dynamic and that due to water mass changes



- Dynamic: high frequency short correlation scales
- Water mass: low frequency long correlation scales

# Implications for data assimilation:

- collocalization
- error covariances
  - (i.e. length scales )

## Collocalization

## RMS S(z)

- Screenshot of OceanDIVA output visualised in GoogleEarth
- Pins are coloured by RMS misfit of model – observed salinity
- Comparison is for January 2004 of 47yr 1 degree model control run

RMS > 0.4 psu RMS < 0.1 psu



## Collocalization

## RMS S(T)

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#### Longer covariance length scales along isotherms

One-point correlation maps in HadCEM



Haines et al. (MWR, 2006)

#### Assimilation of 1 observation (near Gulf Stream)

$$T(z), S(z) \rightarrow z(\rho), \pi(\rho) \rightarrow Assimilate \rightarrow T(z), S(z)$$

## Density level depth $z(\rho)$ before and after assimilation



50 km length scale

Spiciness increment  $\pi$  ( $\rho$ )



300 km length scale

## Data Assimilation

Kalman Filter Equation:

$$\mathbf{x}_{a} = \mathbf{x}_{b} + \mathbf{K} (\mathbf{y}_{o} - \mathbf{H}\mathbf{x}_{b})$$

where  $\mathbf{x}_{a}$  is the model analysis state vector  $\mathbf{x}_{b}$  is the model background state  $\mathbf{y}_{o}$  is the observation vector  $\mathbf{K}$  is the gain matrix

vector(first guess)

## Data Assimilation

Kalman Filter Equation:

$$\mathbf{x}_{a} = \mathbf{x}_{b} + \mathbf{K} \left( \mathbf{y}_{o} - \mathbf{H}\mathbf{x}_{b} \right)$$

where vector(first guess)	$\mathbf{x}_{a}$ is the model analysis state vector $\mathbf{x}_{b}$ is the model background state
	$\mathbf{y}_{\mathrm{o}}$ is the observation vector
	$\mathbf{K}$ is the gain matrix:
<b>R</b> ) <sup>-1</sup>	$\mathbf{K} = \mathbf{B}\mathbf{H}^{\mathrm{T}} (\mathbf{H}\mathbf{B}\mathbf{H}^{\mathrm{T}} +$
observation operator	where $\mathbf{H}$ is the
(interpolation to observation spac	e)
the observation error covariance	<b>R</b> is
the background error covariance	<b>B</b> is

#### Model background error covariance

- Commonly specified using:
  - Covariance length scales (e.g. SODA, Carton et al. 2000)
  - Model EOFs (Mercator (SEEK); Brasseur and Verron, 2006)
  - Model forecast error covariances (Bluelink; Oke et al. 2006)
- Our approach:
  - "Flow dependent" error covariance
    - Assimilation along isotherms or isopycnals

#### S(T) Assimilation method

#### Standard method:

<u>S(T) algorithm:</u>

$$\begin{split} T_{a}(z) &= T_{b}(z) + K_{z} \left[ T_{o}(z) - H \ T_{b}(z) \right] \\ S_{a}(z) &= S_{b}(z) + K_{z} \left[ S_{o}(z) - H \ S_{b}(z) \right] \end{split}$$

> K<sub>T</sub> allows spreading over much greater distances than K<sub>z</sub> due to increased covariance length scales on isotherms.

Also, second salinity increment is independent of the 1st!

 $\begin{array}{l} T_{a}(z) = T_{b}(z) + K_{z} \left[ T_{o}(z) - H T_{b}(z) \right] \\ S'_{a}(z) = S_{b}(z) + \Delta S_{bal}, \text{ such that} \\ \Delta S_{bal} \text{ ensures } S'_{a}(T_{a}) = S_{b}(T_{a}) \\ S_{a}(T_{a}) = S'_{a}(T_{a}) + K_{T} \left[ S_{o}(T_{a}) - H S_{b}(T_{a}) \right] \end{array} \right\} \text{ from an S obs}$ 

## Model/forcing details

- NEMO (v2.3) modelling framework:
  - OPA9 ocean model
    - 46 z-levels, free surface, partial steps, energy-enstrophy conserving momentum advection, TKE vertical mixing
    - Namelist settings and keys as in DRAKKAR 'G70' series
    - No 3D relaxation to climatology (apart from small regions used in G70)
  - LIM2 ice model
  - Tripolar grid :
    - ORCA1: Global 1° resolution, 1/3° tropical enhancement
    - ORCA025: Global 1/4° resolution
- Bulk forcing (DFS3) from :
  - T,Q,U,V: ERA40/ECMWF Operational Analyses
  - Qlw, Qsw, Precip, Snow : CORE (ISCCP), with reduction applied to precipitation at high latitudes.
  - 60 day / 10m SSS relaxation with 5X under-ice relaxation
  - Forcing details identical to G70

## Assimilation scheme and forcing details

- Assimilate in situ temperature and salinity data only
- Observations from quality-controlled ENSEMBLES data set (Ingleby and Huddleston, 2007) from UK MetOffice (EN3\_v1c)
  - includes WOD05 and Argo
- Uses NEMOVAR online observation operator (FGAT)
- Analysis Correction Method (Lorenc, 1991) for z and T levels implemented within NEMO code (<10% increase in computation cost)</li>
- Spatially-varying length scales (Carton et al., 2000):
  - ZONAL: 450km tropical to 375km mid-latitudes
  - MERID: 250km tropical to 375km mid-latitudes
- T-level increments only used between 40N-40S, and below 100m depth. Outside this region *z*-level increments are used
- 5 day assimilation cycle with 1 day IAU.

## Ocean Reanalysis Experiments

## • <u>Illustrate:</u>

- Impact of Argo
- Difference between assimilation on Z and T levels
- <u>1 degree model ( ORCA1 ):</u>
  - 3-year experiments (Jan. 1, 2002 Dec. 31, 2004), initialized from a 44-year control run.
    - ALL: Assimilate all in situ observations from ENSEMBLES data (i.e. XBT, CTD, moorings and Argo)
    - NOARGO : Withold Argo data from above
  - ALL and NOARGO but using standard Z-level assimilation

Smith and Haines, QJRMS (in press)

#### RMS Temperature Misfits : 2002-2004



## RMS Temperature Misfits : 2002-2004



## RMS Salinity Misfits : 2002-2004



## RMS Salinity Misfits : 2002-2004



## Global reanalysis using S(T) assimilation

#### Sea surface temperature

- <sup>1</sup>/<sub>4</sub> degree reanalysis
  - Eddy-permitting
  - 1987-2007

- 1 degree reanalysis
  - 1/3° Eq enhancement
  - 1958-2007

Both reanalyses available at: BODC, Godiva2 and OceanDIVA



## Differences with climatology

#### Average 300-1000m



- Annual mean for 2004
- Large bias in Subtropical North Atlantic in control
- Biases corrected in reanalysis

Smith and Haines, QJRMS (in press)

## Differences with climatology

#### Average 300-1000m



Smith and Haines, QJRMS (in press)

#### Biases in the mean state

#### Average 300-1000m



## Biases in the mean state

#### Average 300-1000m



## RMS Temperature Misfits : 1988-2004



## RMS Salinity Misfits : 1988-2004



## RMS Temperature Misfits : 1988-2004



## RMS Salinity Misfits : 1988-2004





\*Created using OceanDIVA: www.resc.reading.ac.uk

#### North Atlantic S(T) : CLIVAR GSOP water mass intercomparison



Gemmell et al. (2008)



- Reanalysis shows much tighter distribution than control run for all temperature classes.
- In particular, note the correction of the positive salinity bias (for T=2-10C) in the control
- Calculation made using OceanDIVA online web service

#### North Pacific S(T) across syntheses



Gemmell et al. (2008)

#### Bias v Standard Deviation North Pacific – S(T) – over T range 5-17 °C



## Global ocean heat content variability



Reading reanalyses similar to other products
Assimilation corrects drifts in control runs

## **Tropical Pacific SST**



## Global mean quantities

#### ORCA025-G70 : 1/4° Control ORCA025-R07 : 1/4° Reanalysis ORCA1-R07 : 1° Reanalysis



## **Transport through Denmark Strait**

- Figure shows transport as a function of density class, with warm colours indicating southward flow

- Dense overflow in control weakens and freshens over time

- Reanalysis maintains strong southward dense flow and shows increased interannual variability

#### 1/4° Control



1/4° Reanalysis

#### Improvements to the Arctic Ocean

#### 1/4° Control Run (no assimilation)

1/4 °Reanalysis

Mugford et al. (in prep.)



#### **Arctic Freshwater Fluxes**





Mugford et al. (in prep.)





#### <sup>1</sup>/<sub>4</sub> ° control run

<sup>1</sup>/4 ° reanalysis

## Meridional overturning

#### ORCA025-G70 : 1/4° Control ORCA025-R07 : 1/4° Reanalysis ORCA1-R07 : 1° Reanalysis







## Summary

- The S(T) algorithm has been implemented into the NEMO global ice-ocean model
- Two reanalyses have been made: a 50-year reanalysis at 1° resolution, and a 21-year reanalysis at 1/4° resolution.
- Overall, the assimilation is able to prevent drifts in many ocean metrics, and brings the model in better agreement with accepted values.
- An evaluation of water mass properties in various ocean syntheses performed as part of the CLIVAR-GSOP intercomparison, shows that the S(T) reanalyses provide excellent agreement with in situ observations.
- Results suggest that assimilation of salinity data along isotherms should provide better recovery of historical water mass properties than using depth level method
- Studies underway to use reanalysis for:
  - Heat and salt content variability
  - Arctic freshwater budget
  - Impact of assimilation on ecosystem models
  - Force a global coastal ocean modelling system (GCOM)
- Future Work:
  - Still need to determine most appropriate S(T) lengthscales (requires front detection)
  - Density/spice assimilation
  - Investigate sensitivity to DFS4 and ERA-Interim forcing
  - Implement altimetry assimilation

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