Model-data fusion Workshop: Ocean Data Assimilation Summary

Peter Oke et al. CAWCR - CSIRO

Talk Outline

- Status of ocean data assimilation internationally
- Current capabilities in Australia
- > Aspirations for the Australian community
- Stakeholder needs and requirements
- Gaps in capability and/or capacity
- Potential avenues for addressing gaps
- Known unknowns
- Unknown unknowns
- Workshop goals

Operational ocean data assimilation capabilities

GODAE short-range prediction

System Name	Country	Data Assimilation Method	Reference
BODAS	Australia	Ensemble Optimal Interpolation	Oke et al., 2008
ECCO-JPL	USA	Kalman filter and smoother	Fukumori, 2002
FOAM	UK	Analysis Correction	Martin et al., 2007
Mercator	France	Static SEEK filter	Brasseur et al., 2005
MOVE/MRI.COM	Japan	Multivariate 3DVAR	Fujii and Kamachi, 2003
NCODA	USA	Multivariate Optimal Interpolation	Cummings, 2005
NEMOVAR	European Union	Multivariate Incremental 3DVAR	Weaver et al., 2005
TOPAZ	Norway	Ensemble Kalman filter	Evensen, 2006

Table 1. Data assimilation methods used by GODAE systems

Cummings et al. 2009: Ocean Data Assimilation systems for GODAE, Oceanography, 22, 80-97.

	MRI-JMA http://ds.data.jma.go.jp/tcc/tcc/products/elnino/index.html			
Operational oc	 Operational MOVE/MRI.COM-G (1x1 +0.3eq) .50 vertical levels Multi-variate 3DVAR JRA-25(1979-2003)+JMA ops fluxes 	 COBE-SST gridded product Subsurface T&S (GTS+α) SLA along track from AVISO MDT analyzed with historical T, S 	 Weighted mean of clim. (1%) and model fields is used for FG From 1979 onwards Both Delayed (34-38 days) and NRT (3-7 days) 	
	• 10 days assimilation cycle+IAU		Lagged ensemble	
GODAE short-ra	ORA-S3 (ECMWF) http://ds.data.jma.go.jp/tcc/tcc/products/elnino/ index.html ECMWF System 3 http://www.ecmwf.int/products/forecasts/d/charts/seasonal/			
	Operational	HadISST+ Reynolds OI v2 SST	Bias correction (T,S and	
Seasonal predict		 Subsurface T&S (ENACT/ENSEMBLES+GTS) SLA maps from AVISO MDT by assimilating T,S 	 bias correction (1,5 and Pressure) 10-year relaxation to clim From 1959 onwards Both Delayed (12 days) and RT 5-analyses ensemble 	
	POAMA – PEODAS (CAWCR, Melbourne)			
	 In transition to Operations ACOM-2 Ocean Model Based on MOM2. ERA-40 (1980-2001)+NCEP2 (2002 onwards) fluxes 3 Day assimilation cycle Multi-variate ensemble OI 	 ENACT observational data set T&S SST from NCEP re-analysis Time evolving covariances from ensemble perturbed about main analysis using forcing perturbations 	 T&S 3D relaxation to Levitus with 2year time scale Ensemble initial conditions from ensemble of states used for covariance calculation 	
	GODAS (NCEP) http://www.cpc.ncep.noaa.gov/products/GODAS/			
	 Operational MOMv3 (1x1 + 1/3 at eq). 40 vertical levels, 3DVar NCEP/DOE R2 atmospheric fluxes 	 Temperature from NODC,GTSPP and GTS. Along Track SLA from Jason-1 SST from Reynolds OIv2. 	 From 1979 onwards Both 14 days and 1 day lags relative to RT No relaxation to climate 	
	• 12 hour assimilation cycle+IAU	MDT by assimilation of T,S		
	MERCATOR (Meteo France) http://bulletin.mercator-ocean.fr/html/welcome_en.jsp			
	 Operational OPA8.2 ORCA2 (2cos\u03c6 x 2° + 0.5° eq.) reduced order Kalman Filter (SEEK) ERA-40 (1979-2001) + ops fluxes 7-day assimilation cycle 	 Reynolds OI v2 SST Subsurface T&S (ENSEMBLES+CORIOLIS) SLA along track from AVISO Model MDT 	 10-year relaxation to clim From 1979 onwards 7 days behind RT 	
Balmaseda et al. 2010,	MO (MetOffice) http://www.metoffice.gov.	uk/research/seasonal/		
Ocean Obs Proceedings	 Operational HadOM3/OI (1x1+0.3 eq) ERA40 (1985-2002)+ ECMWFops fluxes 7 days assim cycle+IAU 	 HadISST+ Reynolds OI v2 SST Subsurface T&S (ENACT/ENSEMBLES+GTS) 	 Bias correction (Pressure) From 1985 onwards 3 days behind RT 5-analyses ensemble 	
	GMAO ODAS-1 http://gmao.gsfc.nasa.gov/research/oceanassim/ODA_vis.php GMAO Seasonal Forecasts: http://gmao.gsfc.nasa.gov/cgi-bin/products/climateforecasts/index.cgi			
	Operational	Reynolds OI v2 SST	Bias correction (SLA)	
	 Operational Poseidon/OI and EnKF (5/8 x 1/3) NCEP R1+ SSMI winds GPCP precip 5 day assim cycle+IAU 	 Keynolds OF V2 551 Subsurface T&S (NCEP GTS + TAO + GDAC Argo) SLA along-track from JPL (EnKF only) Online bias correction for MDT 	 From 1993 onwards Delayed (about 7 days) 3-analyses ensemble 	
	Table 1: Summary of different ocean assin	(EnKF only)	n of operational and	

Table1: Summary of different ocean assimilation systems used in the initialization of operational and quasi-operational seasonal forecasts.

Research ocean data assimilation capabilities

- > Existing community systems:
 - Regional Ocean Modelling System (ROMS) 4d-Var:
 - Community modelling system used at Rutgers, UCLA, Scripps, UNSW, …
 - Independently developed system used at OSU
 - ECCO:
 - German effort (Stammar et al.)
 - US effort (Lee et al. JPL; Cornelle et al. SCRIPPS)
 - ???

Current capabilities in Australia

Research and operational:

- Mesoscale Open Ocean → Bluelink Ensemble OI @ CSIRO/BoM
- Seasonal → Seasonal ensemble prediction @ CSIRO/BoM
- Coastal \rightarrow CSIRO, UNSW, ...
- > Research:
 - ROMS 4d-Var @ UNSW

> EnOI is based on the Kalman Filter equations:

$$\mathbf{w}^{\mathrm{a}} = \mathbf{w}^{\mathrm{b}} + \mathbf{K}(\mathbf{w}^{\mathrm{o}} - \mathbf{H}\mathbf{w}^{\mathrm{b}}),$$

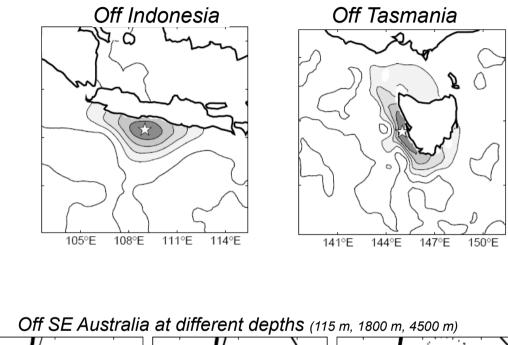
$$\mathbf{K} = (\rho \circ \mathbf{P})\mathbf{H}^*(\mathbf{H}(\rho \circ \mathbf{P})\mathbf{H}^* + \mathbf{R})^{-1}$$

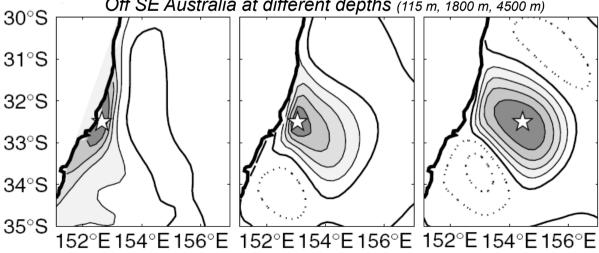
> where the model state is

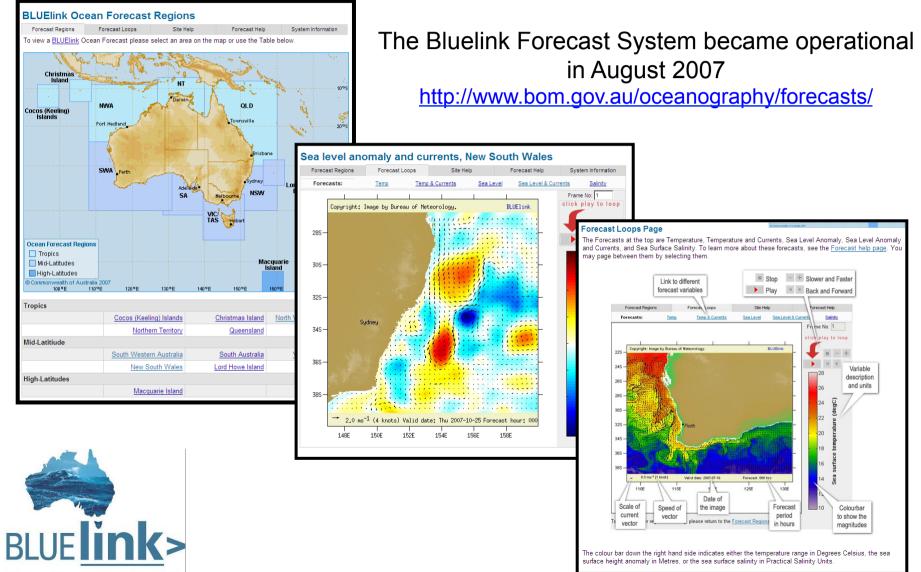
$$\mathbf{w} = \begin{bmatrix} \eta & T & S & u & v \end{bmatrix}^*$$
$$\mathbf{P} = \mathbf{A}\mathbf{A}^{\mathrm{T}}/(n-1)$$
$$\mathbf{A} = \alpha \begin{bmatrix} \mathbf{w}_1' & \mathbf{w}_2' & \cdots & \mathbf{w}_n' \end{bmatrix}$$

Localisation ρ, is required when the model sub-space >>> ensemble size – and is here implemented as a distancedependent reduction of P.

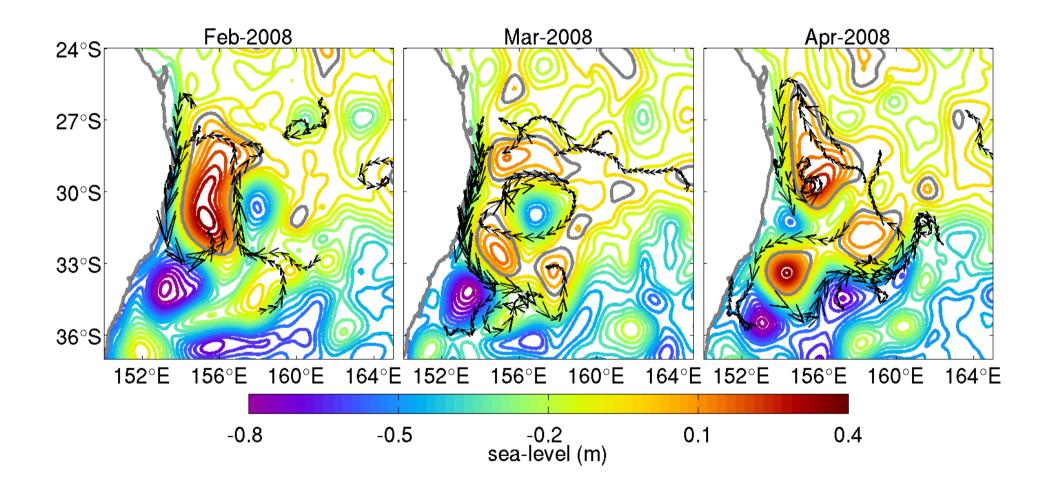
- Ensemble-based covariances summarise the linear dynamics of the circulation
 - inhomogeneous
 - anisotropic
 - concisely represent lengthscales and model variance
 - can easily be seasonally varying (but we haven't exploited this yet)

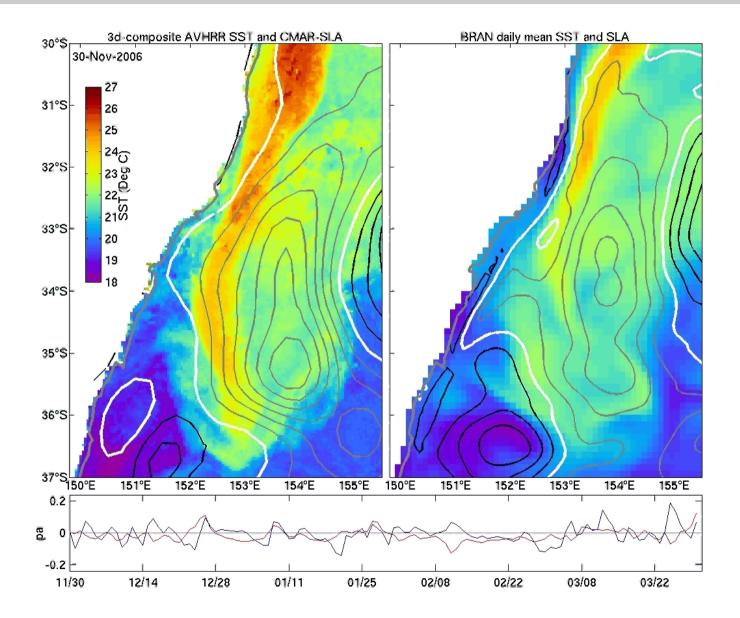


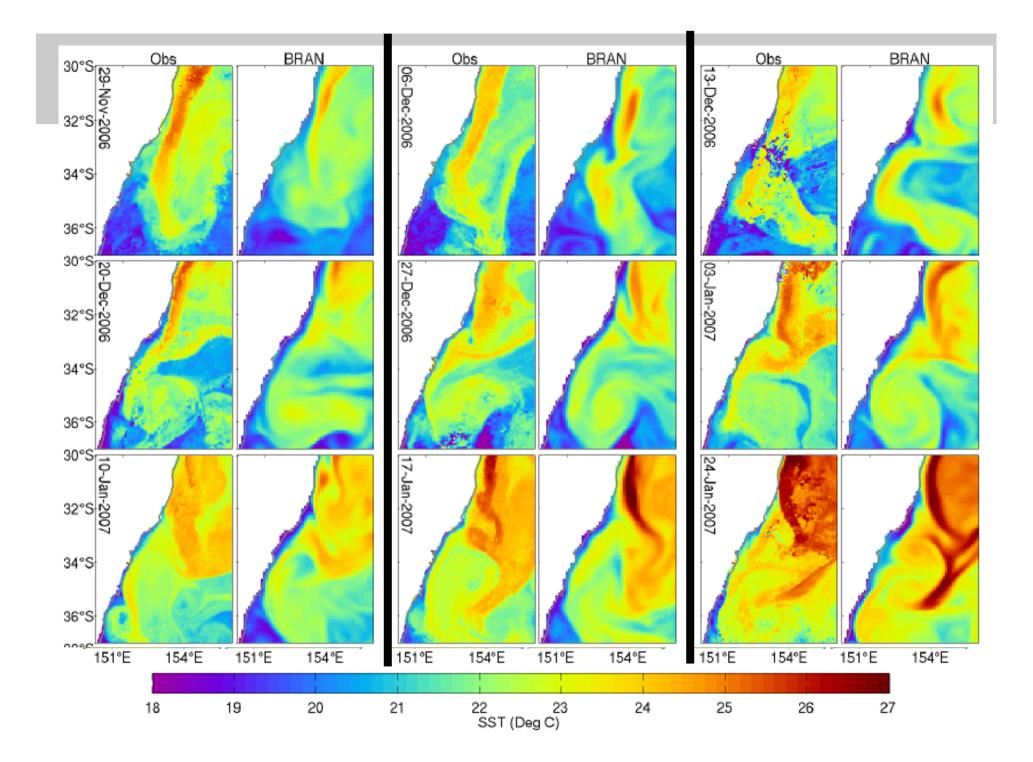


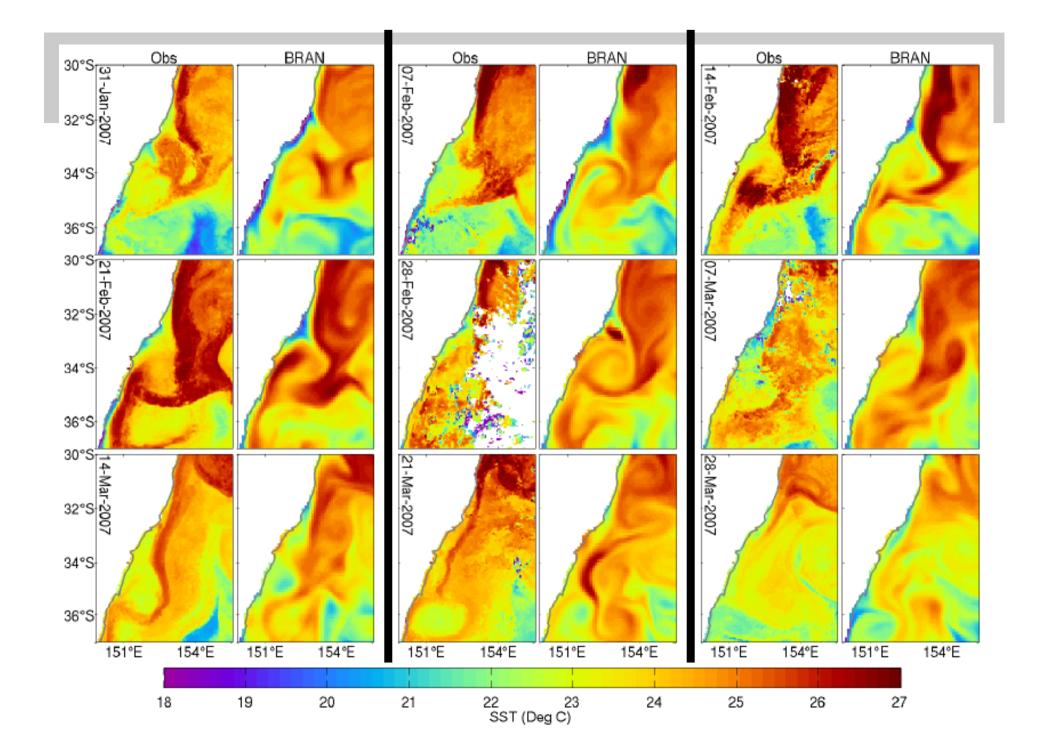


Ocean Forecasting Australia

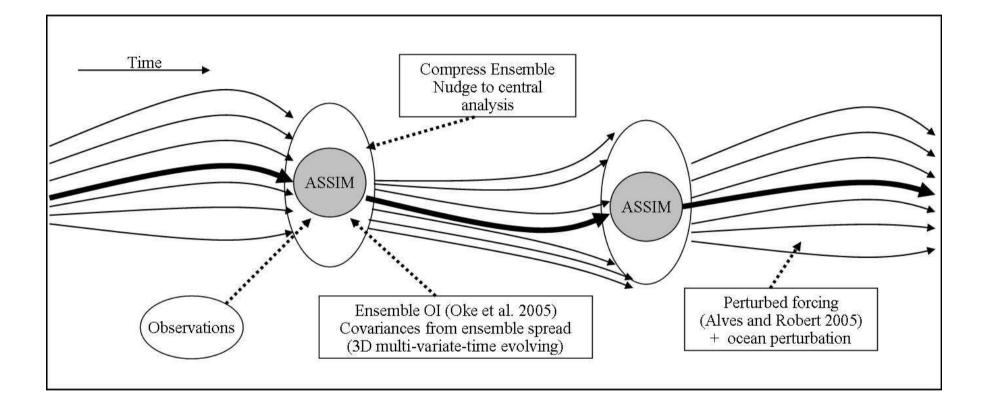








Seasonal: POAMA

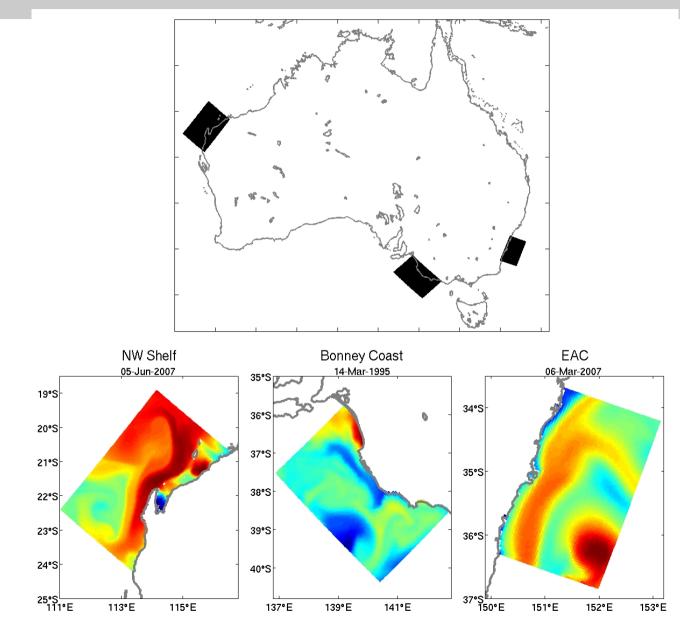


Seasonal

LONGITUDE : 185E LONGITUDE : 185E LATITUDE : D LATITUDE : D TIME : 01-NOV-1988 D6:44 to 31-JAN-2005 17:15 TIME : 31-JAN-1986 06:42 to 31-JAN-2005 17:17 40 40 (m) 80 HLH 120 DEPTH (m) 80 120 160 160 200 200 0.90 0.05 0.15 0.25 0,35 D.45 0.55 0.10 0.30 0.50 0,70 1,10 1.30 Anomalous Sal rmse Anomalous Temp rmse Figure 2. RMS error of interannual 40 anomalies of (a) temperature, (b) salinity and 80 (c) zonal current. Shown are the PEODAS DEPTH (m) reanalysis (black), the old POAMA re-120 analysis (green), the ECMWF ORA-S3 (red) and the Control simulation (blue). The 160 verifying observations are from the TAO 200 mooring at location 165°E. 2400,15 0,25 0.05 D.35 0.45 Anomalous U rmse PEODAS ECMNF POAMA1

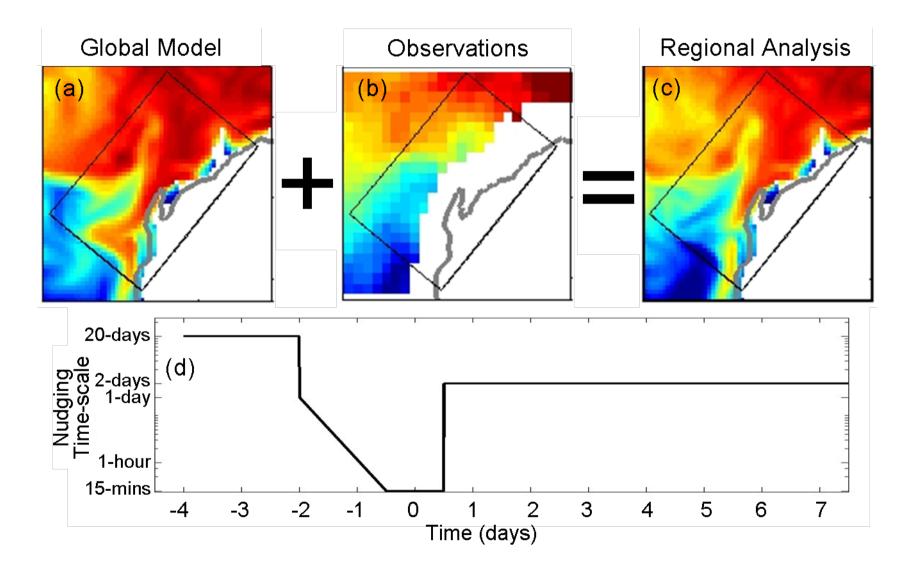
Control

Coastal Ocean DA: EnOI

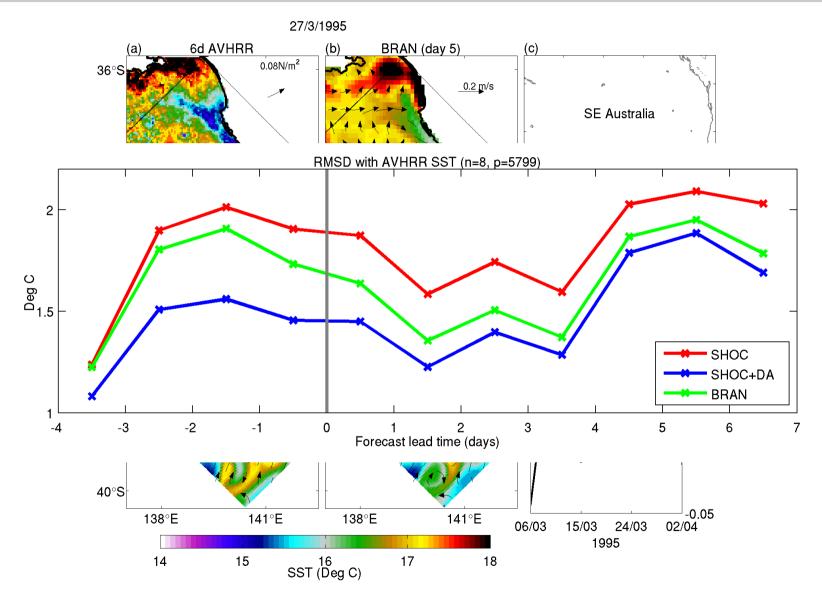


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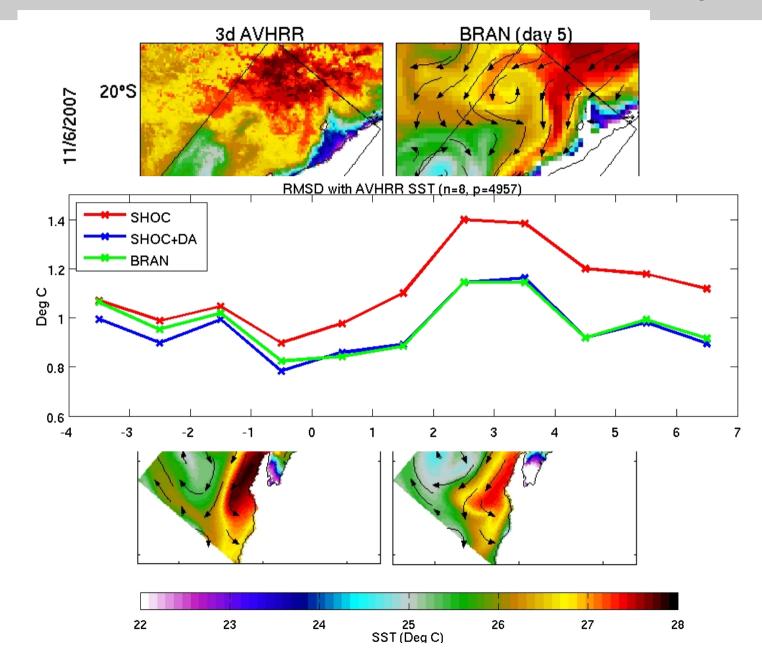
Coastal Ocean DA: EnOI



Coastal Ocean DA: EnOI -- Bonney Coast



Coastal Ocean DA: EnOI -- NW Cape



1

Coastal / Shelf scale Ocean DA: 4d-Var

➤ UNSW – MacDonald et al. to advise

Sea-ice DA

Forecast system exists, but no DA

Ocean BGC DA

- ➤ See Parslow et al.
- Plans to develop Ocean BGC DA capability under Bluelink-3 (Lenton & Matear)

Aspirations for the Australian community

- CSIRO/BoM aspirations include:
 - 4d Ensemble DA extension
 - Ensemble-Var development
- > University aspirations:
 - ???
- > Community aspirations:
 - Method development?
 - Do we want to be research leaders?
 - Or bask in the shadow of other international efforts?

Stakeholder needs and requirements

- Stakeholders:
 - Defence (D)
 - Meteorology (e.g., TC, fog forecasting ...) (M)
 - Search and Rescue (S&R)
 - Oil and Gas (O&G)
 - Environmental stakeholders (e.g., EPA, State & Local Authorities) (E)
 - Research (R)
 - Agriculture (A)
- > Variables required:
 - Open Ocean
 - Coastal
 - T/S \rightarrow D, EPA, R, M
 - Littoral
 - $U/V \rightarrow D, R$

• T/S \rightarrow D, EPA, R, M, A U/V \rightarrow D, S&R, O&G, EPA, R

- $U/V \rightarrow D$, S&R, EPA, R
- Waves \rightarrow D, R

Stakeholder needs and requirements

- Stakeholders:
 - Defence (D)
 - Meteorology (e.g., TC, fog forecasting ...) (M)
 - Search and Rescue (S&R)
 - Oil and Gas (O&G)
 - Environmental stakeholders (e.g., EPA, State & Local Authorities) (E)
 - Research (R)
 - Agriculture (A)
- Derived properties:
 - Prediction of trajectories/dispersion (D, S&R, O&G, E, R)
 - Underwater acoustics (D)
 - Uncertainty (all)

Gaps in capability and/or capacity

See known unknowns …

Areas where rapid progress is possible

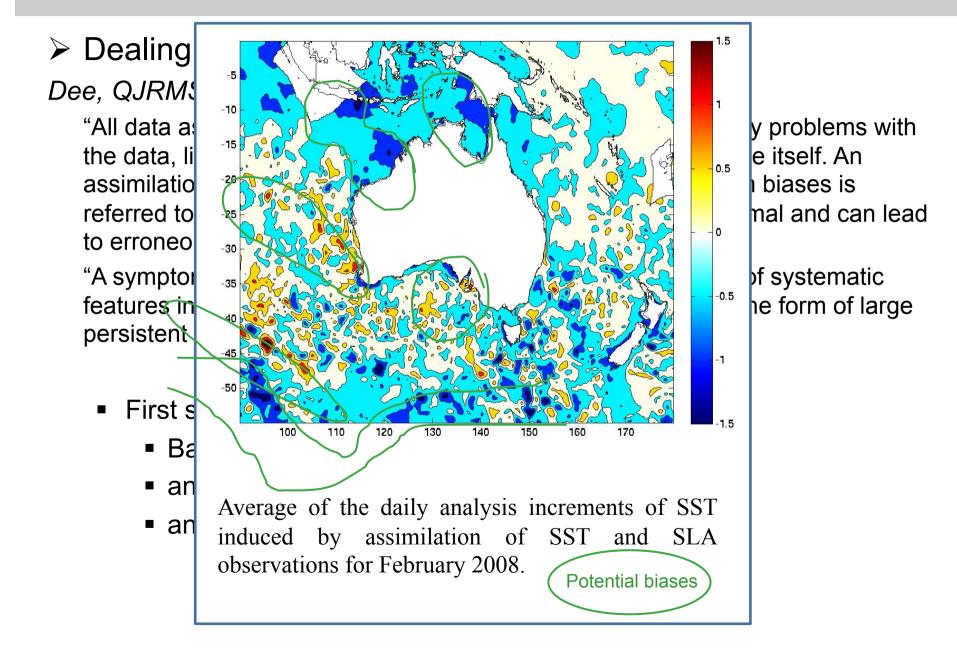
- Assimilation of all data types ... e.g., from IMOS
- Combining ensemble and Var DA

- Important areas where progress is likely to be limited
 - Oceans are data-sparse (physics, but esp BGC)
 - Extension of systems to BGC ...
 - Recruiting researchers with DA experience is difficult

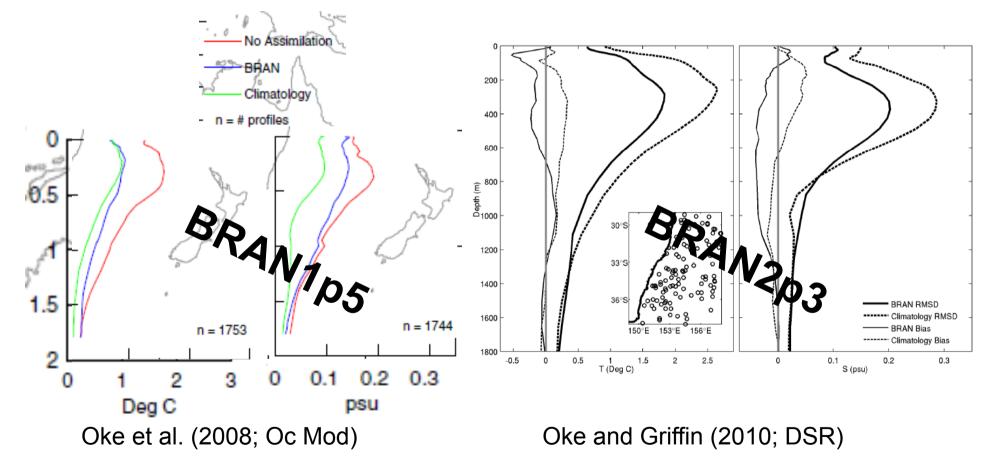
Potential avenues for addressing gaps

- > Development of community DA tools:
- Training / proactive at attracting PhD students
 - How ... ? Who ... ?
- > Systems:
 - Implementation on GPU clusters / and on parallel platforms

... In no particular order



- Dealing with bias
- Forecast skill ... beating persistence (getting there)



BRAN1p5 vs BRAN2p2: main difference is the initialisation: Nudging vs IAU

- Dealing with bias
- Forecast skill ... beating persistence (getting there)
- Mean sea-level ... for assimilation of altimetry
 - Several international programs offer "solutions" to this problem ??
 - Some ocean DA systems use obs-products ... some still use model estimates

- Dealing with bias
- Forecast skill ... beating persistence (getting there)
- Mean sea-level ... for assimilation of altimetry
- Applications to sea-ice
 - CSIRO to work on solution with UTAS
 - Special challenge → sea-ice extent changes, so ensemble approach requires novel treatment

- Dealing with bias
- Forecast skill ... beating persistence (getting there)
- Mean sea-level ... for assimilation of altimetry
- Applications to sea-ice
- > Applications to BGF ... Parslow et al. to offer solution
 - What obs to assimilate?
 - How to relate ocean colour to modelled fields (e.g., NPZ)?
 - ???

- Dealing with bias
- Forecast skill ... beating persistence (getting there)
- Mean sea-level ... for assimilation of altimetry
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- Model error
 - DA requires explicit estimates of model error
 - Several studies show that model error is poorly represented (e.g., seasonal prediction)

QUARTERLY JOURNAL OF THE ROYAL METEOROLOGICAL SOCIETY Q. J. R. Meteorol. Soc. 135: 1538–1559 (2009) Published online 31 July 2009 in Wiley InterScience (www.interscience.wiley.com) DOI: 10.1002/qj.464

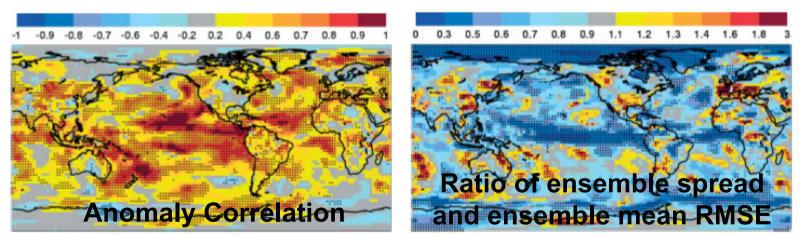


Addressing model uncertainty in seasonal and annual dynamical ensemble forecasts

F. J. Doblas-Reyes,^a* A. Weisheimer,^a M. Déqué,^b N. Keenlyside,^c M. McVean,^{d†} J. M. Murphy,^d P. Rogel,^e D. Smith^d and T. N. Palmer^a

All systems suggest

that spread cannot be considered a useful predictor of skill



unat spread cannot be considered a userul predictor of skill. Annual-mean predictions showed lower forecast quarty than seasonal predictions. Only small differences between the systems were found. The full multi-model ensemble has improved quality with respect to all other systems, mainly from the larger ensemble size for lead times longer than four months and annual predictions. Copyright © 2009 Royal Meteorological Society and Crown Copyright

- Dealing with bias
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- Mean sea-level ... for assimilation of altimetry
- Applications to sea-ice
- > Applications to BGF ... Parslow et al. to offer solution
- Model error
- How to combine ensemble and variational DA
 - ???

Known unknowns

- Dealing with bias
- Forecast skill ... beating persistence (getting there)
- Mean sea-level ... for assimilation of altimetry
- Applications to sea-ice
- > Applications to BGF ... Parslow et al. to offer solution
- Model error
- How to combine ensemble and variational DA
- Coupled ocean/atmosphere DA
 - Different time-scales ... lagged coupling
 - BoM committed to 4d-Var, Bluelink/seasonal invested in ensemble
 - BoM exploring for TC applications under Bluelink
 - BoM/CSIRO exploring for seasonal applications under GRDC

Known unknowns

- Dealing with bias
- Forecast skill ... beating persistence (getting there)
- Mean sea-level ... for assimilation of altimetry
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- > Applications to BGF ... Parslow et al. to offer solution
- Model error
- How to combine ensemble and variational DA
- Coupled ocean/atmosphere DA

Unknown unknowns

Workshop Goals

- Obtain an accurate picture of the status of ocean DA
- Identify how we can improve our capability



Breakout session

> Methods:

- Community tools/data
 - What, how, who
- Coupled DA
 - Ocean / atmosphere
 - Ocean / BGC
 - Ocean / sea-ice
 - Sea-ice / atmosphere

Current capabilities in Australia (updated)

- > Ocean data assimilation
 - Bluelink: Andreu-Burillo, Brassington, Oke, ...
 - Status: BODAS complete and tested; further developments planned (4d, additional obs types, ...)
 - Coastal (e.g., INFORMD): Jones
 - Status: developing an EnOI for a coastal model; main challenges are the lack of observations and dealing with tides.
 - ROAM: Oke, Sun Status: Preliminary system developed based on Bluelink system
- Coupled ocean / atmosphere
 - Regional: Andreu-Burillo, Brassington, Sandery, …
 Status: Preliminary system using MOM4 ocean, TC-LAPS atmosphere, BODAS, and OASIS-3 coupler. Next step to use UM. No coupled DA yet.
 - Seasonal: Alves, Hudsen, O'Kane, Oke, Okely, Webb, Yin Status: Operational system running, and a new PEODAS system being transferred to operations. Started work on coupled DA. 50-year reanalysis completed.
- Biogeochemical data assimilation
 - Systems to be developed under BL-3. Status: Tested various techniques for BGC for toy models. Developed a "good framework" for 4d application.
- > Uncertainty
 - Ensemble prediction methods using CLAM to estimate forecast errors and adaptive sampling: O'Kane, Oke
 - Status: Breeding system developed for MOM4; application to WBC complete

Aspirations for the Australian community (updated) – page 1

Need to include FTEs ...

- Perfect forecasts (Nugzar)
- Use assimilation to understand model error (Alves)
- Treatment of (model or obs) bias rigorously (Andreau-Burillo)
- Dynamical downscaling of climate/seasonal forecasts to meet needs of eg aquaculture (Parslow; Sun)
- > Ocean data assimilation
 - Development of advanced methods (Oke, O'Kane, Sun). Plans to develop an ensemble-Var ocean DA system – funding pending from ONR. Initial applications will be to coastal ocean applications.
 - Developments of methods to optimise BCs, surface forcing, parameters (Jones, Oke, Parslow, Sun, ...) for coastal ocean and climate
 - Ocean state estimation capability; decadal prediction (Hirst, Lenton)
 - Model error (O'Kane): apply systematic sub-grid-scale pareterisation theory to realistic applications
 - Optimisation of model configuration using DA tools (Parslow)
 - Implementation of ROMS 4d-Var to UNSW ROMA configuration (MacDonald)
 - Dimension reduction and emulation (explore different options to with uncertainty other than really crass estimates) of fine-scale DA models (Nugzar)

Aspirations for the Australian community (updated) – page 2

Coupled ocean/atmos

- World leaders in coupled ensemble DA (Brassington, Sandery, ...)
- Develop the assimilating global ocean model to couple to a global NWP (Brassington)
- POAMA- coupled DA → Including new observing systems (e.g., altim) → Ensemble generation strategy (Alves, Hudsen, O'Kane, Oke, Okely, Webb, Yin)
- Consider including waves coupled to POAMA (Brassington)
- Extend skill of POAMA to multi-week forecasting (Hudson; Alves); to match NWPocean forecast systems.
- ➢ BGC DA
 - Capability to run scenarios and understand uncertainty of those scenarios (Parslow)
 - Short-term BGC forecasts from 100m 10 km scales to predict eg harmful algal blooms)
 - Use BGC DA to constrain BGC <u>and</u> physics involves algorithm development (Jones, Parslow)
 - Quantify error propagation from uncertainty in physics to BGC (Nugzar)
- > Uncertainty
 - Capability to quantify uncertainty of all ocean forecast systems: O'Kane

Stakeholder needs and requirements (updated)

- Stakeholder needs and requirements
 - Stakeholders want to better understand uncertainty of climate forecasts due to parameter choices

Gaps in capability and/or capacity and avenues for addressing gaps (updated)

- Forcing data sets
 - Systems are configured and tuned for certain NWP product that is not always available
- > Community tool box:
 - Toy models and toy DA systems
 - Web site with links to DA tools
 - Links to data sets (international sources, eMII, AODN, ...)
 - List of papers published and data used
- > Outreach:
 - Annual DA summer school (Jeff Walker to organise @ Monash)
 - Lecturer from ocean, atmos, BGC, terrestrial, ...
 - CSIRO and maybe Australian mathematics society to fund
 - E.g., Day 1-3 focus on methods (theory and exercises using abovementioned toolbox); day 4-5 focus on applications
 - Other coutries have run successfully (eg Oxford Uni; MSRI Berkley)



Bluelink open ocean DA

- Applications to OFAM and CLAM
 - EnOI+GOOS provides useful background state for surface T,S,u,v,eta
 - Not sufficient for high precision surface currents
 - Not sufficient for high precision vertical state and currents
 - Achieving higher precision state estimation?
 - More optimal DA methods and initialisation
 - Progress observing system
 - More optimal yet practical options:
 - Hybrid EnKF, Bred vectors, Time-lagged ensemble (FGAT 4dEnOI),
 ...
 - Fully exploit current GOOS:
 - Remote sensing methods to estimate u,v
 - Remote sensing SSS, SMOS and Aquarius
 - Motivate / contribute to enhancements to GOOS:
 - e.g., Adaptive intensive observations

How should we represent tides? Explicit or parameterised?

- Motivation: Baroclinic conversion, T/S mixing, Transport, residual flux, non-linear sea level
- Risk: tide error > mesoscale errors; explicit tides could degrade lower frequency variability
- Possible strategy:
 - Multi-scale analysis (separation of time-scales, ...)
 - BLUElink(global) non tidal => BLUELink(regional) nested + tidal b.c.'s
- Resolution
 - Horizontal resolution resolved vs constrain variability
 - Vertical resolution sensitivity, limited validation data

Closure of flux errors (momentum, heat and mass)

- tune parameterisations
- surface currents, waves, SST, mixed layer depth (sound ducts)
- ABL winds, heat and moisture
- Coupled projection of innovations
 - wave errors => atmos. winds and ocean currents
 - SST errors => atmos. winds, clouds, ocean state and currents
- Opportunity of ACCESS / UM + BLUElink
 - Ocean prediction and GOOS has matured
 - ACCESS supporting coupled infrastructure

Seasonal DA - POAMA

Ocean DA

Coupled DA

Coastal DA

≻ ROAM DA

- Currently assimilated SST
- Will soon assimilate altim, T/S

> INFORMD

➢ ROMS / UNSW