Development of ROMS/EnKF forecasting system for GBR

Pavel Sakov, Paul Sandery, Frank Colberg, Gary Brassington

Bureau of Meteorology

ROMS workshop 17–20 October 2016, Hobart

Outline

The EnKF

$\mathsf{ROMS}/\mathsf{EnKF}$ system for GBR

System settings Biases Fields Single observation experiments SST bias correction

The EnKF

DA: the minimisation problem

$$\begin{split} \{\mathbf{x}_{i}^{a}\}_{i=1}^{k} &= \arg\min_{\{\mathbf{x}_{i}\}_{i=1}^{k}} \mathcal{L}(\mathbf{x}_{1}, \dots, \mathbf{x}_{k}), \\ \mathcal{L}(\mathbf{x}_{1}, \dots, \mathbf{x}_{k}) &= (\mathbf{x}_{1} - \mathbf{x}_{1}^{f})^{\mathrm{T}}(\mathbf{P}_{1}^{f})^{-1}(\mathbf{x}_{1} - \mathbf{x}_{1}^{f}) \\ &+ \sum_{i=1}^{k} \left[\mathbf{y}_{i} - \mathcal{H}_{i}(\mathbf{x}_{i})\right]^{\mathrm{T}}(\mathbf{R}_{i})^{-1} \left[\mathbf{y}_{i} - \mathcal{H}_{i}(\mathbf{x}_{i})\right] \\ &+ \sum_{i=2}^{k} \left[\mathbf{x}_{i} - \mathcal{M}_{i}(\mathbf{x}_{i-1})\right]^{\mathrm{T}}(\mathbf{Q}_{i})^{-1} \left[\mathbf{x}_{i} - \mathcal{M}_{i}(\mathbf{x}_{i-1})\right]. \end{split}$$

(i = cycle number)

The EnKF

- Based on a recursive solution in the linear case ("Kalman Filter")
- State of the DA system $SDAS_i = {x_i, P_i}$
- ► The EnKF carries the SDAS by an ensemble of model states: SDAS_i = {E_i}

The EnKF (2)

 $\mathsf{Cost}(\mathsf{EnKF}) = \mathsf{Cost}(\mathsf{model\ propagation}) \, \times \, \mathsf{ensemble\ size}$

Main features:

- simple
- for propagation needs forward model only

 long-term behaviour needs to be managed

The EnKF(2)

 $\mathsf{Cost}(\mathsf{EnKF}) = \mathsf{Cost}(\mathsf{model\ propagation}) \, \times \, \mathsf{ensemble\ size}$

Main features:

- simple
- for propagation needs forward model only

 long-term behaviour needs to be managed

Example of dynamic covariance



GBR ROMS/EnKF system

- Our first ocean forecasting experience with ROMS
- One of the most challenging regions around Australia for the EnKF
- \blacktriangleright ... and for a model with vertical σ coordinates

GBR ROMS/EnKF system: the domain

Some domain characteristics

- Includes both shelf and deep ocean
- Has long ocean boundary
- Has quite a bit of internal variability: not as chaotic as the EAC region, and not as forcing-driven as the Gulf of Carpentaria
- Enough of internal variability to generate ensemble spread and propagate it on shelf
- Rugged terrain, steep shelf break

Implications for the system design Compared to the Tasman Sea system:

- Good atmospheric and boundary forcing is critical
- Forcing perturbation is necessary for the best performance



System settings

- EnKF with 96 members (EnKF-C code)
- \blacktriangleright 3-day cycle ($\sim40\,m/1,1152$ CPUs, $541\times197\times30$ grid, 600 s time step)
- Assimilating SLA, SST, T, S (SSS) obs.
- Assimilating SLA and SST asynchronously with 1-day binning
- Using night time SST observations only
- Using instantaneous fields for T/S
- Using de-tided daily average zeta for SLA (propagating extra 1/2 day)
- \blacktriangleright DEnKF scheme, moderated with ALPHA = 0.8

$$\mathbf{A}^{f} = \left(\mathbf{I} - rac{oldsymbol{lpha}}{2} \, \mathbf{K} \mathbf{H}
ight) \mathbf{A}^{a}$$

- Inflation 5%, capped
- $R_{loc} = 150 \,\mathrm{km}$ for SLA/SST, 400 km for T/S
- R-factor = 1.5 for SLA, 16 for SST, 7 for T/S, 3 for SSS
- SST bias correction: RANDOMISE 0.96 0.3
- Forcing perturbation: TAIR 0.03; UV 0.05 SWRAD 0.05 RIVER 0.2

Performance

Innovation statistics for July 2013 - December 2014

	SLA	SST	Т	S	SSS
EnKF	5.83	0.279	0.575	0.0896	0.213
EnOI	5.96	0.321	0.510	0.0885	0.207
BRAN2015*	4.18	0.274	0.438	0.0805	-

 $^*0.1^\circ$ model, ERA-interim forcing, stats for domain 135E-165E, 25S-5S



Biases



9 / 24

A bit more stats

Innovation statistics for July 2013 - December 2014

	SLA	SST	Т	S	SSS
MAD	5.83	0.279	0.575	0.0896	0.215
Bias	0.82	0.020	-0.203	-0.0132	0.0806
Persistence MAD	5.87	0.307	0.569	0.0891	0.219

- SST looks good
- Biases for all fields but SST are quite substantial
- The model is consistently too warm
- The model is too fresh at surface

SST on 14-07-2007

ensemble mean SST







SSS on 14-07-2007

ensemble mean SSS







Surface velocity on 14-07-2007

member 1 SV ensemble mean SV SV ensemble spread □0.5 m 1 0.9 0.45 0.8 0.4 0.7 0.35 0.6 0.3 0.5 0.25 0.4 0.2 0.3 0.15 0.2 0.1 0.1 0 0.05 0

Single observation experiments

-10

.12

-14

-16

-18

-20

.22

-24

-28



temp increment from SLA observation at (102,346)

Single observation experiments (2)

-10

-12

-14

-16

-18

-20

-23

-28

section j = const 7 = 0section i = const 0.1 45 45 175 175 0.05 425 425 Single SST observation 675 675 run enkf-27 on 14-07-2007 925 925 -0.05 innovation = 1° 1350 1350 error STD = 0.5° 3250 3250 at (i, j) = (102, 346)salt increment from SST observation at (102,346) z = 0 section i = const section i = const Model bathymetry 0.01 45 45 175 175 0.005 4500 425 425 4000 675 675 3500 925 925 -0.005 3000 1350 1350 2500 -0.01 3250 3250 2000 i. 1500 v increment from SST observation at (102.346) 1000 500 z = 0 section i = const section i = const 45 0.01 152 154 175 148 150 156 158 160 175 146 0.005 425 425 675 675 -0.005 925 925 1350 1350 -0.01 3250 3250

temp increment from SST observation at (102,346)

15/24

Single observation experiments (3)

-10

.12

-14

-16

-18

-20

.22

-24

-28

section j = const 7 = 0section i = const 45 45 0.1 175 425 425 Single SLA observation 675 675 run enkf-27 on 14-07-2007 925 925 -0.1 innovation = 5 cm1350 1350 error STD = 3 cm3250 3250 at (i, j) = (67, 117)salt increment from SLA observation at (67.117) z = 0 section i = const section i = const Model bathymetry 0.02 45 175 0.01 4500 425 425 4000 675 675 3500 925 925 -0.01 3000 1350 1350 -0.02 2500 3250 3250 1500 1000 v increment from SLA observation at (67,117) 500 z = 0 section j = const section i = const 45 4 0.02 142 144 146 148 150 152 154 156 158 160 175 175 425 425 675 675 925 925 -0.02 1350 1350 3250 3250 16/24

temp increment from SLA observation at (67,117)

Single observation experiments (4)

-10

-12

-14

-16

-18

-20

-22

-24

.24

-28

section i = const section j = const 7 = 045 45 0.1 175 175 425 425 Single SLA observation 675 675 run enkf-27 on 14-07-2007 925 925 -0.1 innovation = 5 cm1350 1350 error STD = 3 cm3250 3250 at (i, j) = (76, 73)salt increment from SLA observation at (76,73) z = 0 section i = const section i = const Model bathymetry 45 45 0.02 175 175 4500 0.01 425 425 4000 675 675 3500 925 925 -0.01 3000 1350 1350 -0.02 2500 3250 3250 1500 1000 v increment from SLA observation at (76,73) 500 section i = const section j = const z = 0 0.04 45 45 142 144 146 148 150 152 154 156 158 160 175 175 0.02 425 425 675 675 925 925 -0.02 1350 1350 -0.04 3250 3250

temp increment from SLA observation at (76,73)

17/24

Single observation experiments: EnKF vs. EnOI

 $\mathsf{SLA}\to\mathsf{T}$



Single observation experiments: EnKF vs. EnOI (2)

 $\mathsf{SLA}\to\mathsf{S}$



salt increment from SLA observation at (102.346)

salt increment from SLA observation at (102,346)





EnOI

FnKF

Single observation experiments: EnKF vs. EnOI (3)

 $\mathsf{SLA}\to\mathsf{V}$

z = 0



v increment from SLA observation at (102,346)





EnKF



Comparison of run enkf-15 (w/o SSTB correction) and enkf-20 (with SSTB correction); Jan 2006 – Feb 2009

	SLA	SST	Т	S
enkf-15 MAD	5.78	0.303	0.672	0.145
enkf-20 MAD	5.64	0.291	0.538	0.131
enkf-15 bias	-0.001	-0.088	0.148	-0.055
enkf-20 bias	-0.006	-0.006	-0.134	0.038

Effect of SST bias correction (2)

Mean T increments for runs enkf-15 and enkf-20



Effect of SST bias correction (3)

Mean S increments for runs enkf-15 and enkf-20



Conclusions

- EnKF forecasting system for GBR region has been developed based on a 4-km ROMS model with CFSR forcing and BRAN BC, and EnKF-C code (https://github.com/sakov/enkf-c)
- The system involves SST bias correction and (simplistic) forcing perturbation
- The system is quite robust and relatively inexpensive to run, doing a 3-day cycle in about 40 minutes on 1,152 CPUs
- The performance is rather good on SST, and OK on SLA, subsurface T and S (loosing on those to BRAN2015)

On ROMS

- ROMS internal de-tiding works well
- Looses on SSH (zeta) to MOM

Conclusions

- EnKF forecasting system for GBR region has been developed based on a 4-km ROMS model with CFSR forcing and BRAN BC, and EnKF-C code (https://github.com/sakov/enkf-c)
- The system involves SST bias correction and (simplistic) forcing perturbation
- The system is quite robust and relatively inexpensive to run, doing a 3-day cycle in about 40 minutes on 1,152 CPUs
- The performance is rather good on SST, and OK on SLA, subsurface T and S (loosing on those to BRAN2015)

On ROMS

- ROMS internal de-tiding works well
- Looses on SSH (zeta) to MOM