Atmosphere-ocean boundary layers and fluxes

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Brown et al., 2014

IPCC AR5, 2013

Presence of observable variability In practice, it is easier to observe the integrated ocean effects (ocean heat content (OHC), salinity) rather than the fluxes themselves.

However, problematic prediction and attributionthis is where modeling helps!

What does hydrography show? OHCs and fluxes are not fixed! 90% anomalous (anthropogenic?) warming ends up in oceans. Hansen et al. (2011).



Fig. 10. (a) Estimated contributions to planetary energy imbalance in 1993–2008, and (b) in 2005–2010. Except for heat gain in the abyssal ocean and Southern Ocean, ocean heat change beneath the upper ocean (top 700 m for period 1993–2008, top 2000 m in period 2005–2010) is assumed to be small and is not included. Data sources are the same as for Figs. 8 and 9. Vertical whisker in (a) is not an error bar, but rather shows the range between the Lyman et al. (2010) and Levitus et al. (2009) estimates. Error bar in (b) combines estimated errors of von Schuckmann and Le Traon (2011) and Purkey and Johnson (2010).

BUDGET is for Heat Content

GMST vs. SST vs. MLT vs. OHC

Atmosphere Recent Warm: 0.15K/decade 3.4m Ocean: 0.15K/decade =34m Ocean: 0.15K/century 0.01% this seasonality



Global climate models do pretty well at matching heat fluxes and watermasses.

Models get better every generation due to improved resolution and parameterizations

What do we usually do to make these improvements?

Changes to model physics, clouds, resolution, numerics, etc. Updates of the flux laws (but not recently)



FIG. 1. (top) CORE (left) total air-sea heat flux and (right) total freshwater flux (air-sea + runoff) into the ocean. Also shown are biases in the present-day mean of these fluxes from the (middle) CCSM3 and (bottom) CCSM4 20C ensemble means. Units: W m⁻². The increment in latitude is 15°.

S. C. Bates, BFK, S. R. Jayne, W. G. Large, S. Stevenson, and S. G. Yeager. Mean biases, variability, and trends in airsea fluxes and SST in the CCSM4. Journal of Climate, 25(22):7781-7801, November 2012. Often agreement in time mean fluxes

Often disagreement in annual band flux variability

Region	Q_E	Р	Q_H	Q_L	Q_S
1	0.85	1.00	0.97	0.71	0.99
2	1.00	0.73	0.94	0.01	0.06
3	0.78	0.00	0.76	0.32	0.87
4	0.11	0.35	1.00	0.98	0.01
5	1.00	0.92	0.04	1.00	0.21
6	0.93	1.00	1.00	0.99	1.00
7	1.00	1.00	0.79	1.00	0.97
8	0.86	0.68	0.26	0.80	1.00
9	1.00	0.80	0.25	1.00	1.00
10	0.84	1.00	0.19	1.00	1.00
11	0.99	0.19	0.05	1.00	1.00
12	0.23	0.17	0.83	1.00	0.09
13	1.00	0.53	1.00	0.75	0.53
14	0.99	0.22	0.92	1.00	0.97
15	0.75	0.99	0.42	1.00	1.00
16	1.00	0.99	0.16	0.23	0.13
17	1.00	0.89	0.55	1.00	0.19
18	1.00	1.00	0.82	1.00	1.00
19	0.33	0.74	0.32	0.23	0.51
20	0.90	0.15	0.42	0.05	1.00
21	0.91	0.78	0.50	1.00	0.86



S. C. Bates, BFK, S. R. Jayne, W. G. Large, S. Stevenson, and S. G. Yeager. Mean biases, variability, and trends in airsea fluxes and SST in the CCSM4. Journal of Climate, 25(22):7781-7801, November 2012.

Control: Isotropic

temp bias - bass - a01e



S. Reckinger, BFK, S. Bachman, F. O. Bryan, G. Danabasoglu. Anisotropic shear dispersion parameterization for mesoscale eddy transport. Ocean Modelling, In prep, 2015.

Anisotropic

temp bias - flow - a01e



Along transects



Map for a16n₂003a



Mesoscale anisotropy often reduces mean biases: pCFC by up to 24% Mesoscale Eddies have a profound effect on QBML Temp by up to 48% Even small changes affect Salinity by up to 63% surface warming budget



By comparing resolved mesoscale eddies to parameterized ones (with same 50km atmosphere), Griffies et al show global differences of O(0.7 W/m²) or O(0.14 K/century)



Stephen M. Griffies, Michael Winton, Whit G. Anderson, Rusty Benson, Thomas L. Delworth, Carolina O. Dufour, John P. Dunne, Paul Goddard, Adele K. Morrison, Anthony Rosati, Andrew T. Wittenberg, Jianjun Yin, and Rong Zhang, 2015: Impacts on Ocean Heat from Transient Mesoscale Eddies in a Hierarchy of Climate Models. J. Climate, 28, 952–977. doi: <u>http://dx.doi.org/10.1175/JCLI-D-14-00353.1</u>

Estimating the Circulation & Climate of the Ocean LLC4320 Model



Scale-Aware (Leith) Viscosity: BFK, S. Bachman, B. Pearson, and S. Reckinger. Principles and advances in subgrid modeling for eddy-rich simulations. CLIVAR Exchanges, 19(2):42-46, July 2014.

Estimating the Circulation & Climate of the Ocean LLC4320 Model





Movie: Z. Jing

Brown Visitor from S. China Sea Institute of Ocean.

Local Analysis: Z. Jing, Y. Qi, BFK, Y. Du, and S. Lian. Seasonal thermal fronts and their associations with monsoon forcing on the continental shelf of northern South China Sea: Satellite measurements and three repeated field surveys in winter, spring and summer. Journal of Geophysical Research-Oceans, August 2015. Submitted.

LES as big as we can?

Perform large eddy simulations (LES) of Langmuir turbulence with a submesoscale temperature front

Use NCAR LES model to solve Wave-Averaged Eqtns.

2 Versions: 1 With Waves & Winds 1 With only Winds

Computational parameters: Domain size: 20km x 20km x -160m Grid points: 4096 x 4096 x 128 Resolution: 5m x 5m x -1.25m 1000x more gridpoints than CESM Movie: P. Hamlington



P. E. Hamlington, L. P. Van Roekel, BFK, K. Julien, and G. P. Chini. Langmuir-submesoscale interactions: Descriptive analysis of multiscale frontal s simulations. Journal of Physical Oceanography, 44(9):2249-2272, September 2014.



P. E. Hamlington, L. P. Van Roekel, BFK, K. Julien, and G. P. Chini. Langmuir-submesoscale interactions: Descriptive analysis of multiscale frontal spin-down simulations. Journal of Physical Oceanography, 44(9):2249-2272, September 2014.

Near Future of Modeling

- LES 5m, 20km x 20km, weeks. Atmosphere & Ocean separate —not coupled.
- NCOM 3-4 km, Global, Forecasts < Annual, ocean-only</p>
- JPL ECCO MITgcm LLC4320, 2km, Global, Months, ocean-only
- CESM 10km, 100km, Global, Centuries, coupled

 GFDL 10km, 25km, 100km, Global, Centuries, coupled
For foreseeable future, air-sea flux & boundary layer turbulence will be
parameterized except on very small domains on both climate & weather timescales. Modeling of decadal variability First-Principle Process & GCM Modeling: Predictions and Biases Quantify process uncertainty, how much do Langmuir mixing or anisotropy of mesoscale eddies affect OHC?



Roughly 1 W/m² each as estimated by integrated T difference from control run. Model versions differ in net air-sea fluxes by 1-6 W/m² in mean and rms. This is 2-10x the observed trend! Retuning, parameterizations, resolution.

Q. Li, A. Webb, BFK, A. Craig, G. Danabasoglu, W. G. Large, and M. Vertenstein. Langmuir mixing effects on global climate: WAVEWATCH III in CESM. Ocean Modelling, August 2015. in press.

S. C. Bates, BFK, S. R. Jayne, W. G. Large, S. Stevenson, and S. G. Yeager. Mean biases, variability, and trends in air-sea fluxes and SST in the CCSM4. Journal of Climate, 25(22):7781-7801, November 2012.

 Modeling of decadal variability
Stochastic (unpredictable beyond persistence) Model: Frankignoul & Hasselmann (77)



One difficulty is getting the reservoir in communication with the atmospheric variability right. Another is getting predictable variability right!

These factors are affected by mixed layer depth.

Langmuir Mixing in CESM: Reduces MLD Errors

Table 3: Root mean square difference (m) of summer and winter mean mixed layer depth in comparison with observation (de Boyer Montégut et al. (2004), updated to include the ARGO data to 2012).^a

Case	Summer			Winter		
	Global	South of $30^\circ {\rm S}$	$30^\circ S-30^\circ N$	Global	South of 30°S	$30^{\circ}S-30^{\circ}N$
CTRL	10.62 (13.40)	17.24 (21.73)	5.38 (6.71)	43.85 (45.50)	57.19 (56.53)	12.57 (16.16)
MS2K	15.37	15.47	17.03	119.91	171.92	40.31
SS02	36.79	63.83	7.54	99.32	164.34	17.39
VR12-AL	9.06	13.47	6.49	40.45	50.33	14.52
VR12-MA	8.73 (11.83)	12.65 (18.13)	6.61 (7.52)	40.99 (42.02)	51.78 (50.78)	14.23 (15.67)
VR12-EN	8.95	10.52	8.91	41.94	52.98	19.58
	Case CTRL MS2K SS02 VR12-AL VR12-MA VR12-EN	Case Global CTRL 10.62 (13.40) MS2K 15.37 SS02 36.79 VR12-AL 9.06 VR12-MA 8.73 (11.83) VR12-EN 8.95	Case Summer Global South of 30°S CTRL 10.62 (13.40) 17.24 (21.73) MS2K 15.37 15.47 SS02 36.79 63.83 VR12-AL 9.06 13.47 VR12-MA 8.73 (11.83) 12.65 (18.13) VR12-EN 8.95 10.52	Case Summer Global South of 30°S 30°S-30°N CTRL 10.62 (13.40) 17.24 (21.73) 5.38 (6.71) MS2K 15.37 15.47 17.03 SS02 36.79 63.83 7.54 VR12-AL 9.06 13.47 6.49 VR12-MA 8.73 (11.83) 12.65 (18.13) 6.61 (7.52) VR12-EN 8.95 10.52 8.91	Case Summer Global South of 30°S 30°S-30°N Global CTRL 10.62 (13.40) 17.24 (21.73) 5.38 (6.71) 43.85 (45.50) MS2K 15.37 15.47 17.03 119.91 SS02 36.79 63.83 7.54 99.32 VR12-AL 9.06 13.47 6.49 40.45 VR12-MA 8.73 (11.83) 12.65 (18.13) 6.61 (7.52) 40.99 (42.02) VR12-EN 8.95 10.52 8.91 41.94	Case Summer Winter Global South of 30°S 30°S-30°N Global South of 30°S CTRL 10.62 (13.40) 17.24 (21.73) 5.38 (6.71) 43.85 (45.50) 57.19 (56.53) MS2K 15.37 15.47 17.03 119.91 171.92 SS02 36.79 63.83 7.54 99.32 164.34 VR12-AL 9.06 13.47 6.49 40.45 50.33 VR12-MA 8.73 (11.83) 12.65 (18.13) 6.61 (7.52) 40.99 (42.02) 51.78 (50.78) VR12-EN 8.95 10.52 8.91 41.94 52.98

^a Numbers shown in the parentheses are for the fully coupled experiments.



L. P. Van Roekel, BFK, P. P. Sullivan, P. E. Hamlington, and S. R. Haney. The form and orientation of Langmuir cells for misaligned winds and waves. Journal of Geophysical Research-Oceans, 117:C05001, 22pp, May 2012.

Q. Li, A. Webb, BFK, A. Craig, G. Danabasoglu, W. G. Large, and M. Vertenstein. Langmuir mixing effects on global climate: WAVEWATCH III in CESM. Ocean Modelling, August 2015. in press.

Mesoscale Anisotropy & Mixed Layer Depth



Anisotropy deepens MLD in Southern Ocean, shallows MLD in North Atlantic, and reduces winter mean rms bias by 15% (annual by 18%)

S. Reckinger, BFK, S. Bachman, F. O. Bryan, G. Danabasoglu. Anisotropic shear dispersion parameterization for mesoscale eddy transport. Ocean Modelling, In prep, 2015.

Twenty years ago, bulk flux schemes were considered to be uncertain by about 30%; the authors find COARE 3.0 to be accurate within 5% for wind speeds of 0–10 m/s and 10% for wind speeds of between 10 and 20 m/s." (Fairall et al. 2003).

Since then, COARE has been updated to v3.5 (Edson et al. 2013). Other observation-based schemes exist as well.

$$\frac{dT}{dt} = \frac{f_1'}{h} - \lambda T$$

GFDL uses a version of Beljaars (1994)

CESM uses a version of Bryan et al. (1996).
This factor is affected by flux laws.

Conclusions

- Improvements to mesoscale, fluxes, boundary layer schemes are similar in bias change magnitude to introducing new physics (submesoscale, Langmuir).
- Mesoscale resolution will soon fix many problems—some difficulty to parameterization (e.g., mesoscale air-sea coupling)
- Scale-aware subgrid models needed for mesoscale resolution
- Climate model air-sea flux schemes have not been refreshed in 20 years, progress has been made in obs, process, theory since then.
- Entrainment, subduction, seasonality are critical to determining the reservoir of OHC and its timescale—which relate to variability, persistence, predictability. They depend on getting many things right—some easy (Ekman pumping), some hard (turbulent entrainment under diverse forcing)

Equivalent Depths of Watermasses by Source (Gebbie & Huybers, 2011)

Consider 1D Oceans: one per watermass

Ekman flushing gives upper limit to λ^{-1} damping timescale





In Press.