

2019

Forced and unforced variations in ocean oxygen

Matthew C. Long

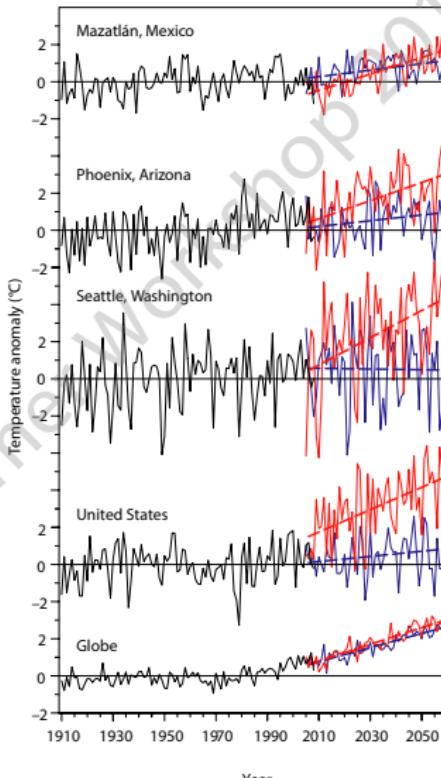
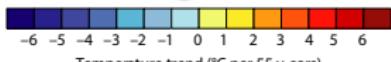
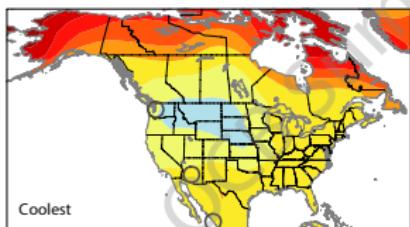
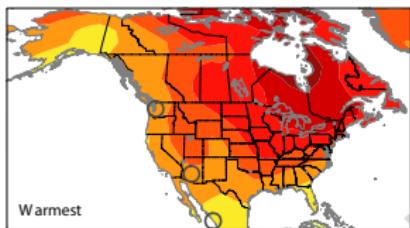
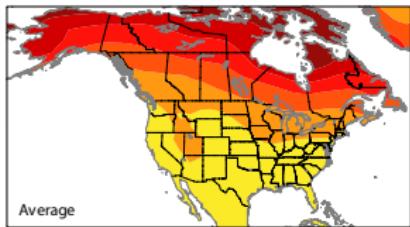
Climate & Global Dynamics Laboratory

National Center for Atmospheric Research

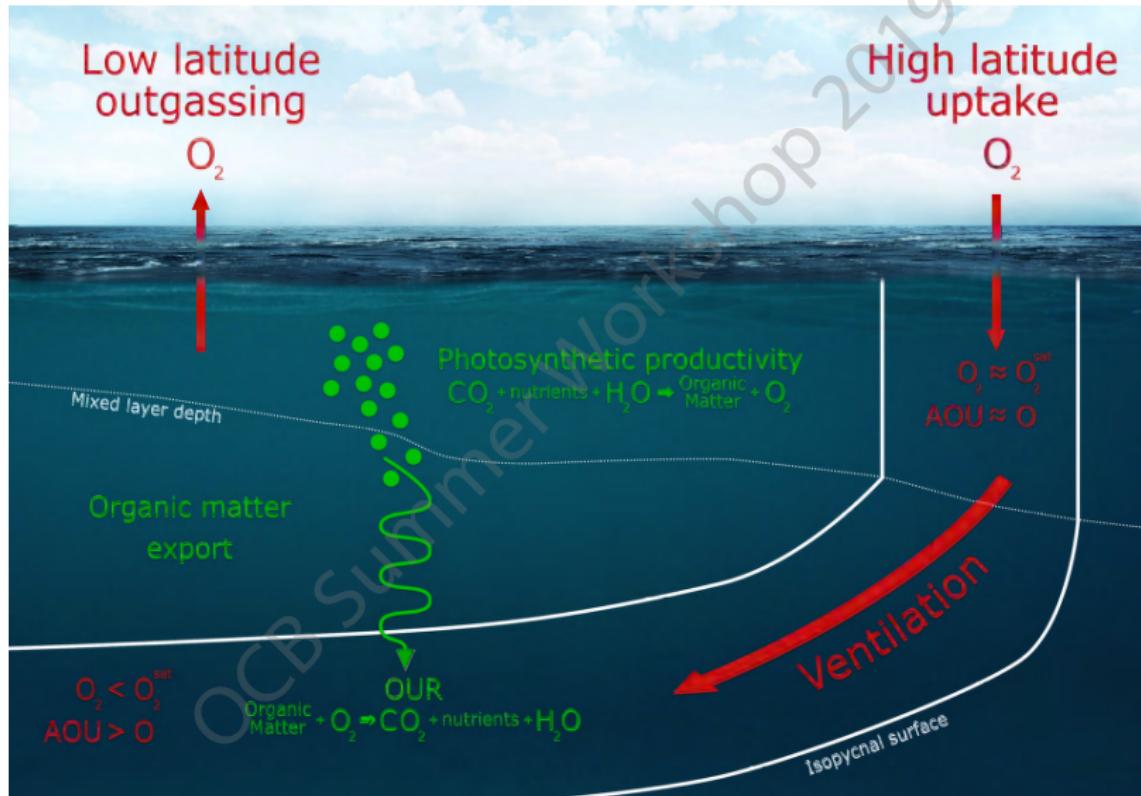


Earth system models: internally generated variability

Range of future climate outcomes: DJF temperature trends (CCSM3, SRES-A1B)



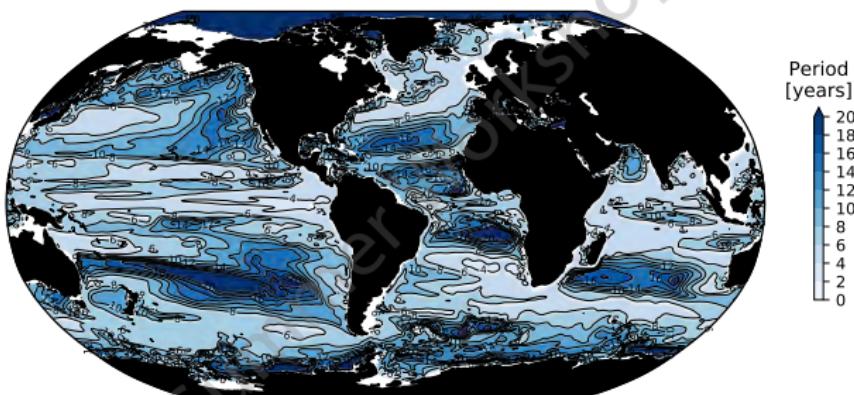
Physical & biological controls on interior oxygen



Graphic credit: M. Long and R. Johnson (NCAR)

Timescales of natural variability in thermocline O₂

Variance-weighted mean period in CESM 1850-control (400–600 m O₂)

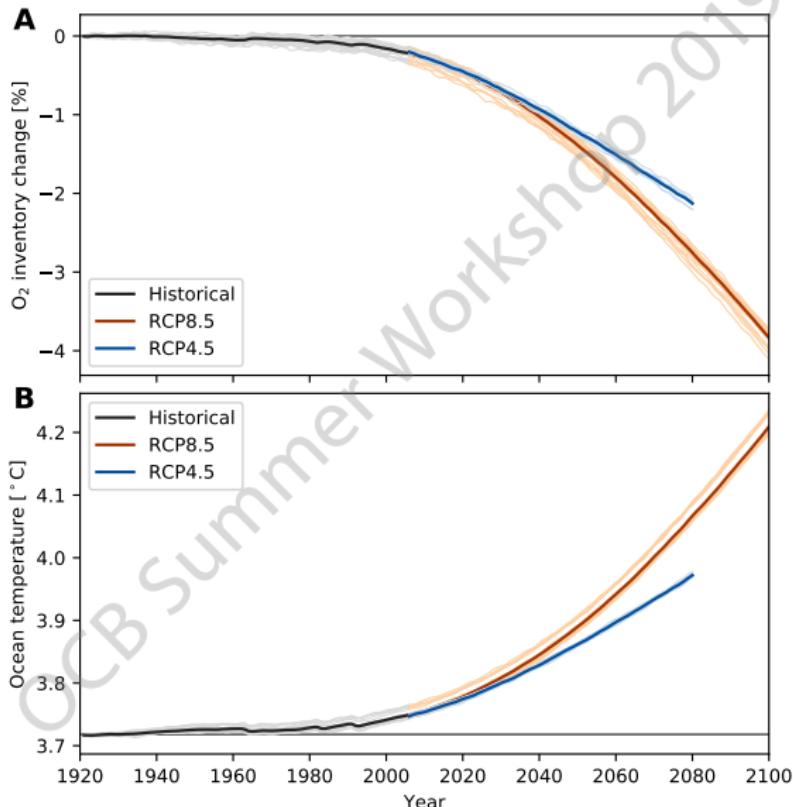


$$T_x = \sum_k V(f_k, x) / \sum_k f_k V(f_k, x)$$

Oxygen decline projected to accelerate

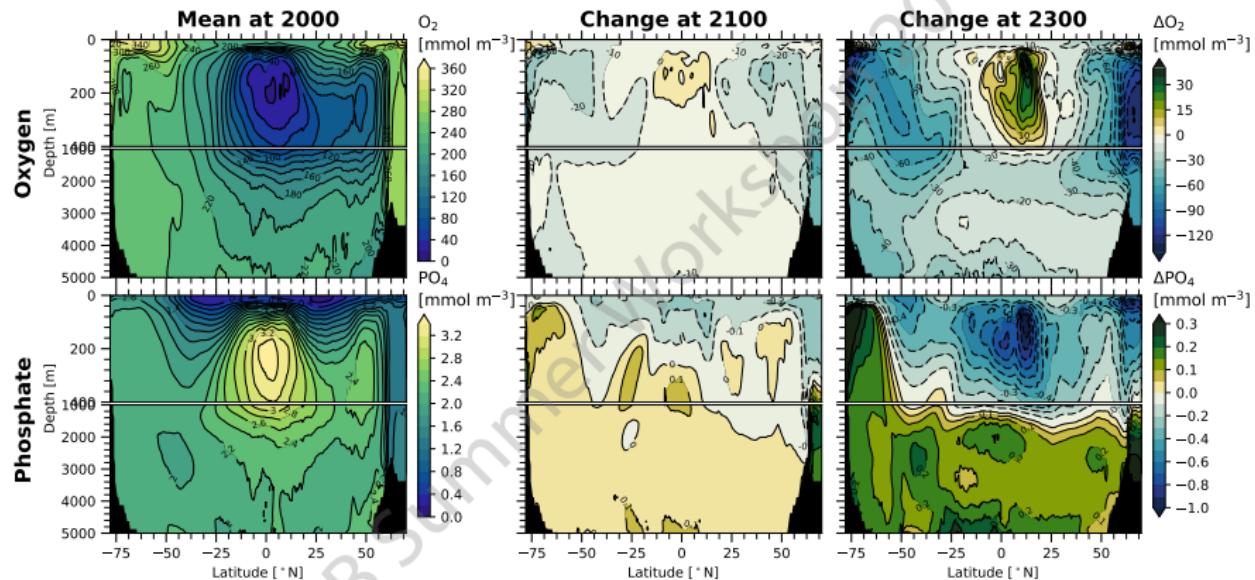
CESM-LE*: projected global change in O₂ & heat

* Community Earth System Model, Large Ensemble



Deep future looks to be dark: Southern Ocean nutrient trapping

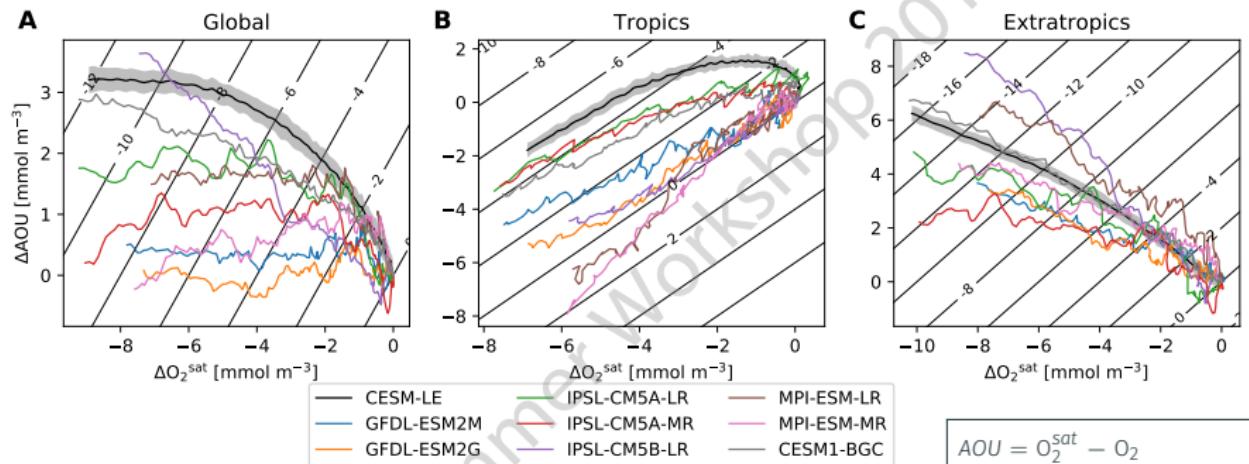
Global zonal-mean properties: RCP8.5 & ECP8.5



Long et al., 2019; Moore et al. 2018

CMIP5 projection: global and regional drivers of deoxygenation (2100)

AOU v. O_2^{sat} phase space: Simulated $\langle O_2 \rangle$ change ($z > -1$ km)

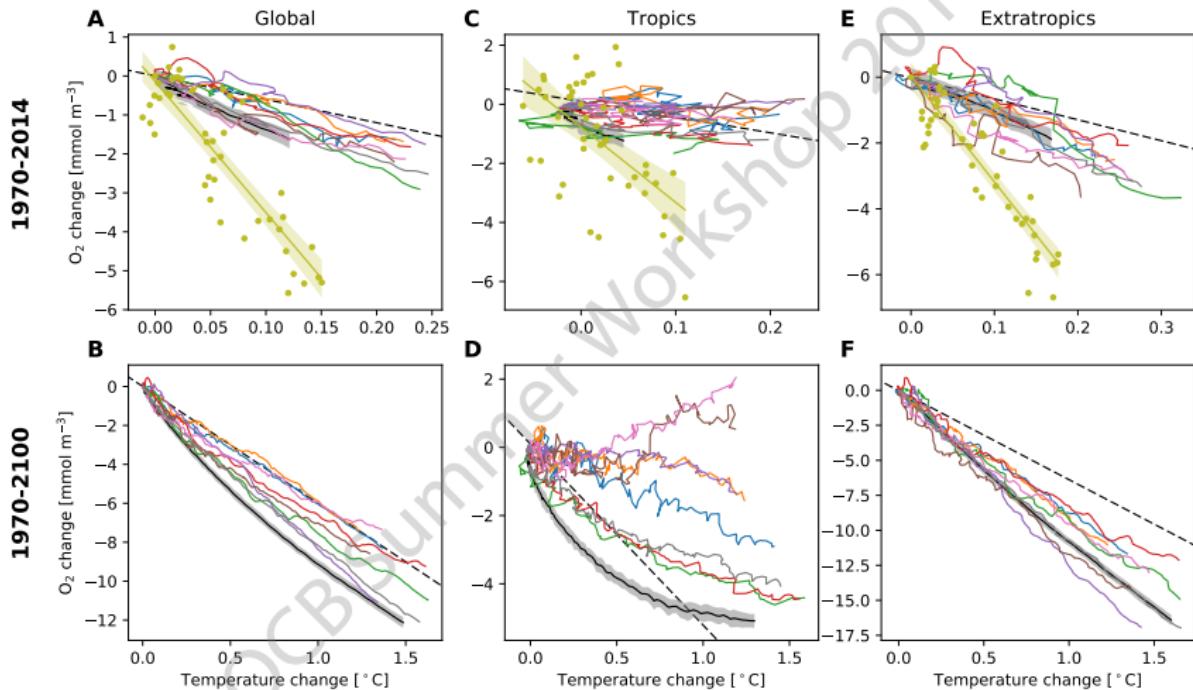


$$AOU = O_2^{sat} - O_2 \\ = f(\text{Age}, O_2\text{-utilization})$$

- Tropics: warming compensated by reduced AOU;
- Extra-tropics: reinforcing AOU and solubility change;
- Cancelation between tropical & extra-tropical ΔAOU :
global O_2 decline dominated by solubility effect.

Oxygen declines related to ocean heat content anomaly

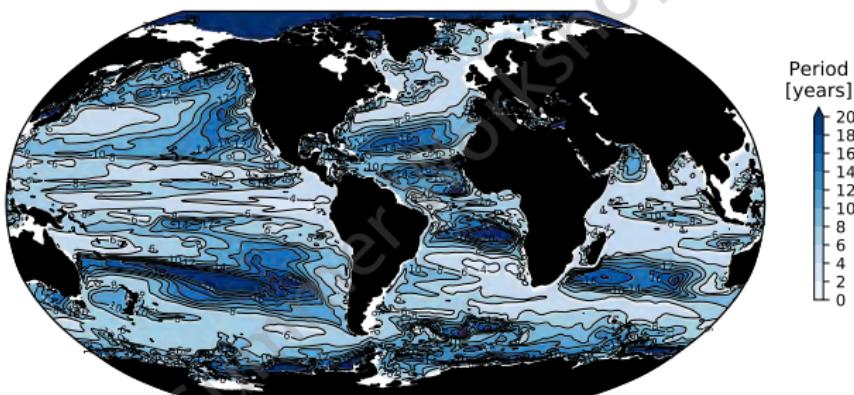
O_2^{sat} v. heat: Simulated $\langle O_2 \rangle$ change ($z > -1$ km)



| | | |
|--------------|--------------|------------|
| CESM-LE | IPSL-CM5A-MR | MPI-ESM-MR |
| GFDL-ESM2M | IPSL-CM5B-LR | CESM1-BGC |
| GFDL-ESM2G | MPI-ESM-LR | |
| IPSL-CM5A-LR | | --- |

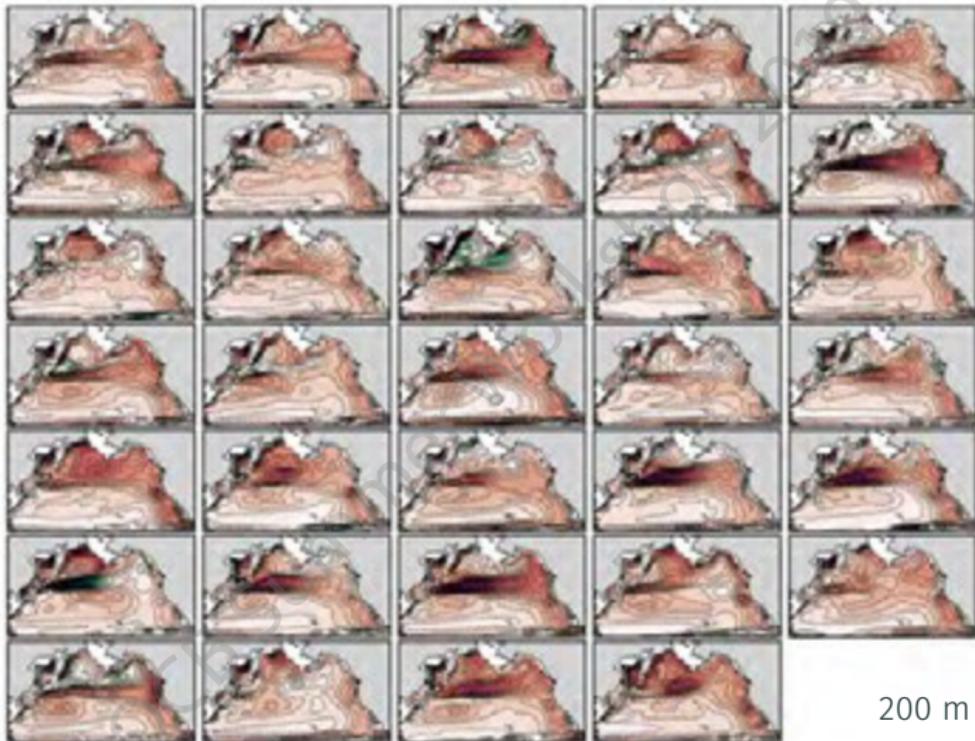
Timescales of natural variability in thermocline O₂

Variance-weighted mean period in CESM 1850-control (400–600 m O₂)

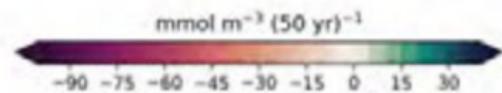


$$T_x = \sum_k V(f_k, x) / \sum_k f_k V(f_k, x)$$

Superposition of forced signal and internal variability: Total



200 m O₂ trends
2006–2055



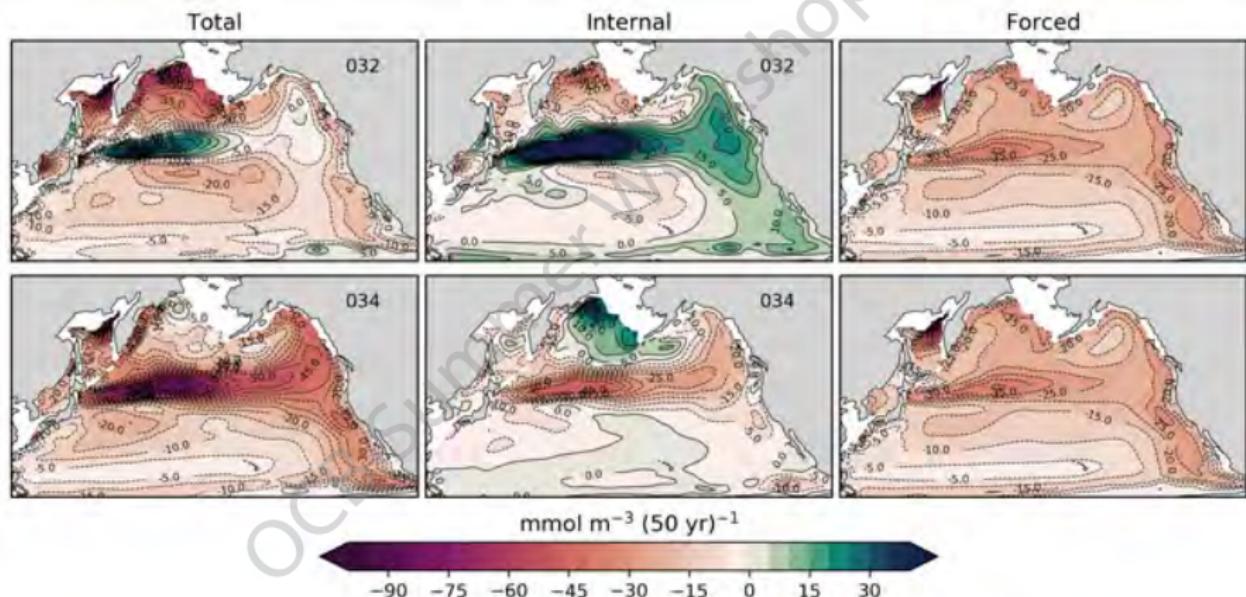
Superposition of forced signal and internal variability

CESM Large Ensemble: Linear trends in 200 m dissolved oxygen (2006–2055)

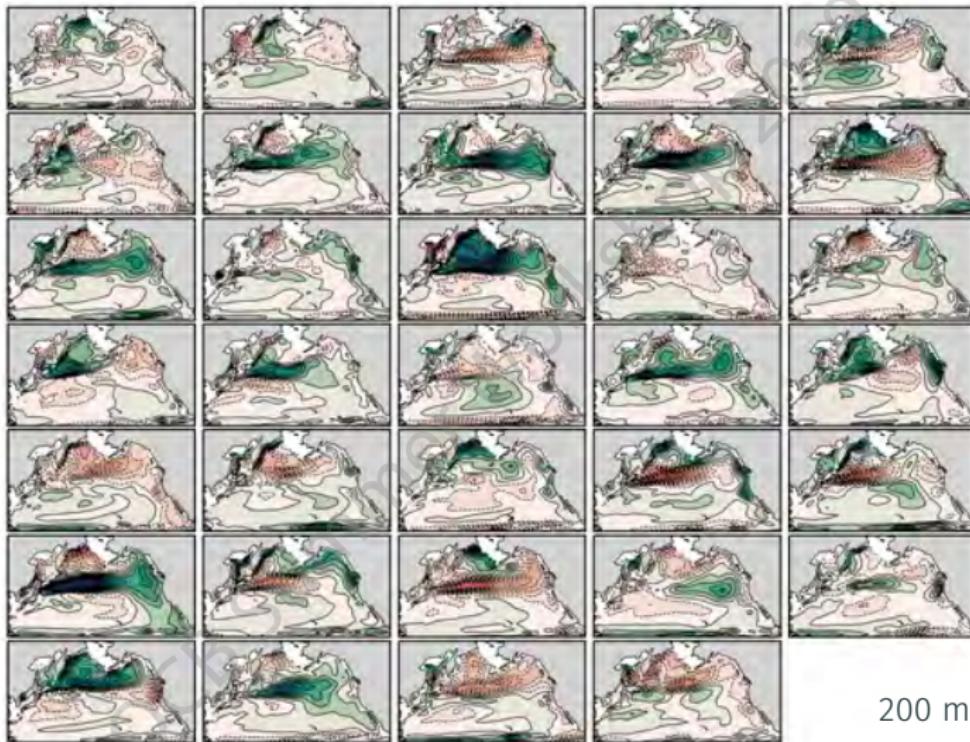
$$\psi_i(t, x) = \tilde{\psi}_i(t, x) + \psi^s(t, x)$$

$$\tilde{\psi}_i(t, x)$$

$$\psi^s(t, x) = \frac{1}{m} \sum_{i=1}^m \psi_i(t, x)$$



Superposition of forced signal and internal variability: Internal variability



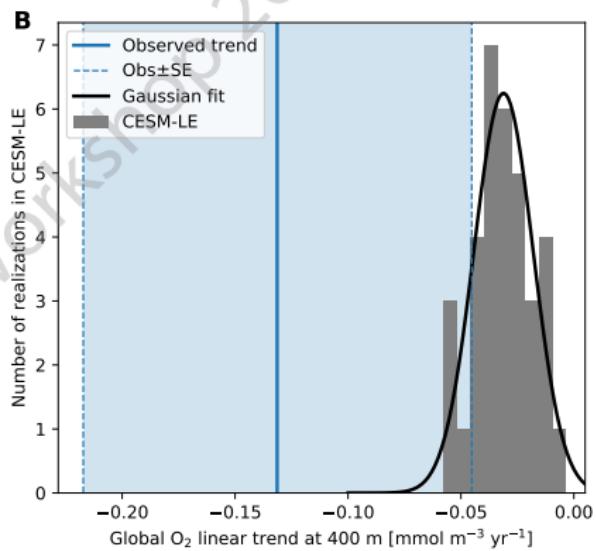
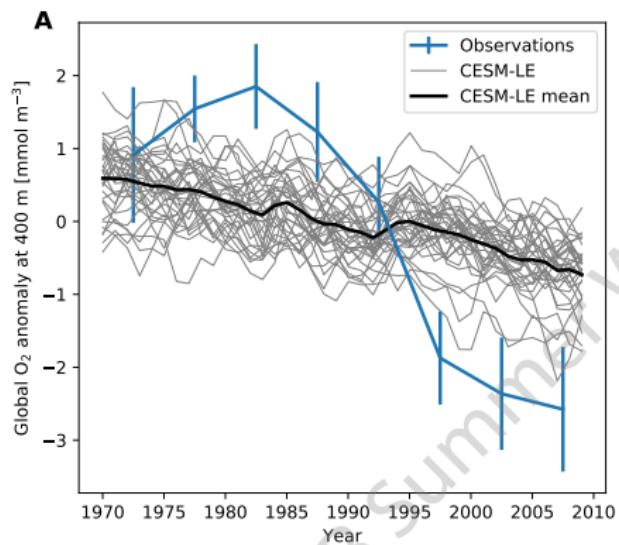
200 m O₂ trends
2006–2055

mmol m⁻³ (50 yr)⁻¹

-90 -75 -60 -45 -30 -15 0 15 30

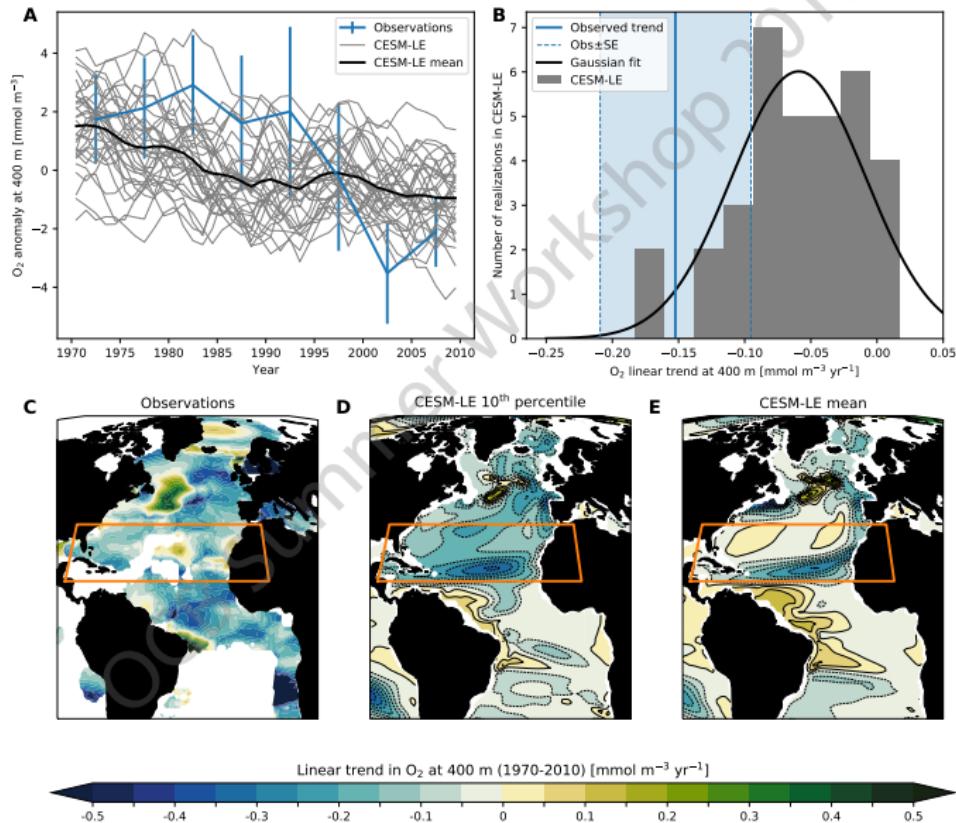
Can internal variability account for model-observations discrepancy?

Observed and simulated trends: global



Can internal variability account for model-observations discrepancy?

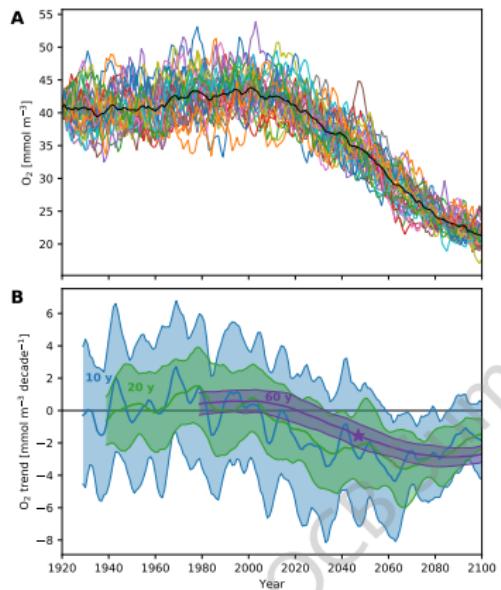
Observed and simulated trends: Subtropical North Atlantic



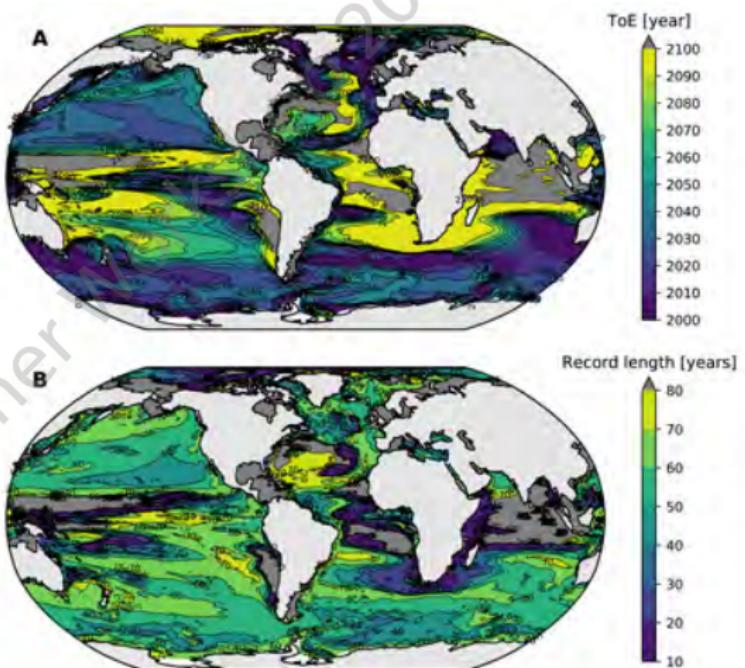
Long et al., 2019

Natural variability challenges detection and attribution

Thermocline O₂ California Current



Time of emergence



Quantifying spatial structure of $[O_2]$ variability and change

Time evolving spatial pattern

$$\psi_i(t, x) = \tilde{\psi}_i(t, x) + \psi^s(t, x)$$

where $t = 1, \dots, n$ and $x = 1, \dots, p$ denote discrete time (annual means) and space (model grid points).

The forced signal

$$\psi^s(t, x) = \frac{1}{m} \sum_{i=1}^m \psi_i(t, x)$$

Climate anomalies

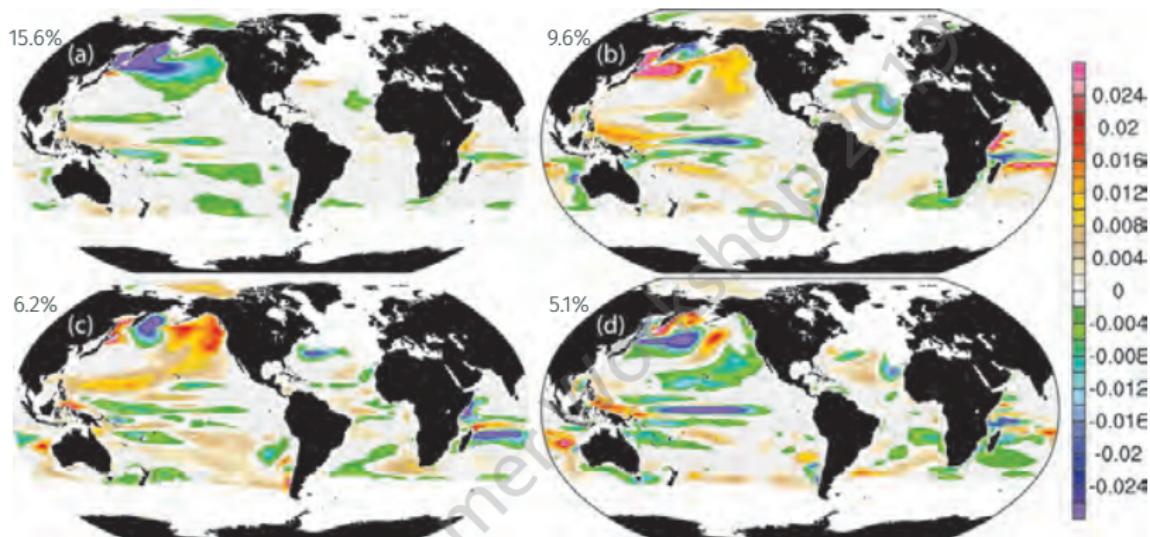
$$\psi'_i(t, x) = \psi_i(t, x) - \bar{c}(x)$$

EOF decomposition of anomalies

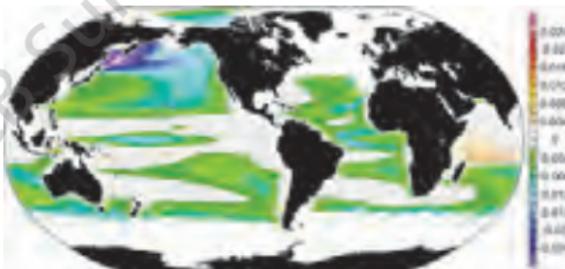
$$\psi'_i(t, x) = \sum_{j=1}^q \alpha_j^i(t) e_j^i(x)$$

Leading EOFs: Large-scale spatial patterns of variability

Natural variability



Forced signal



Quantifying spatial structure of [O₂] variability and change

EOF decomposition of anomalies

$$\psi'_i(t, x) = \sum_{j=1}^q \alpha_j^i(t) e_j^i(x)$$

Projection of each ensemble member ($\psi'_i(t, x)$) onto EOF of forced signal

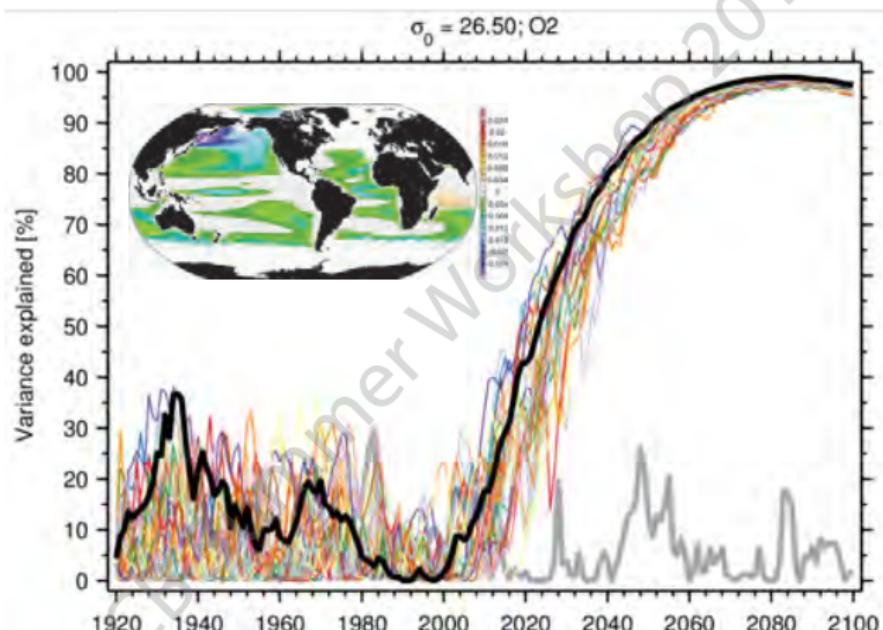
$$\alpha^{is}(t) = \sum_{x=1}^p \psi'_i(t, x) e^s(x)$$

Contribution of forced pattern to overall spatial variance

$$V_i^{is}(t) = \alpha^{is}(t)^2 \left/ \sum_{x=1}^p \psi'_i(t, x)^2 \right.$$

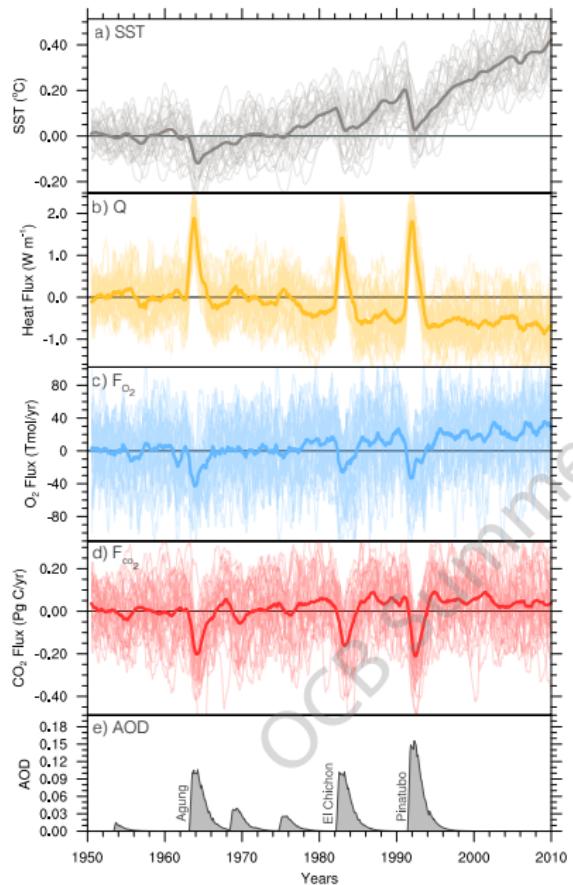
Rising dominance of the forced signal

Percentage of total spatial variance attributable to the forced signal

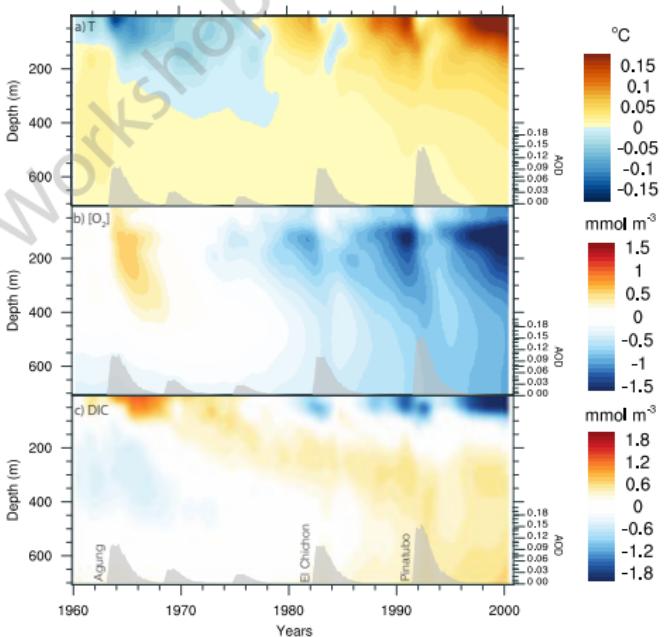


Long et al., 2016

Another forced signal: volcanic eruptions

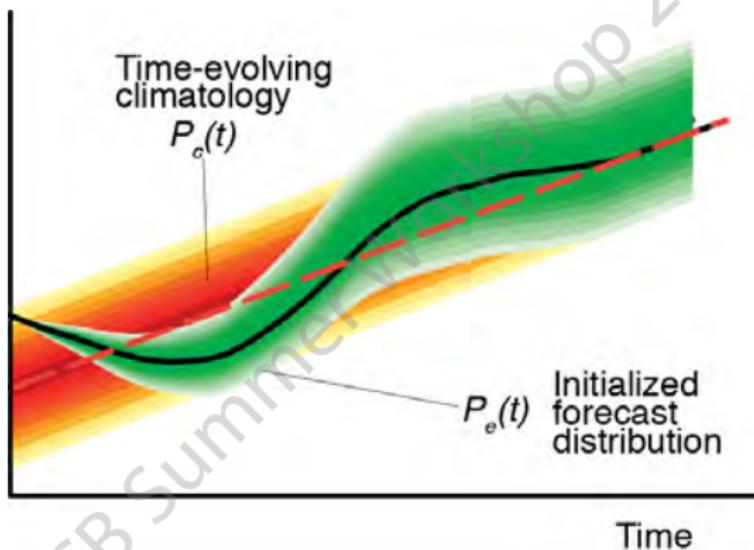


Global-mean ocean properties
CESM Large Ensemble



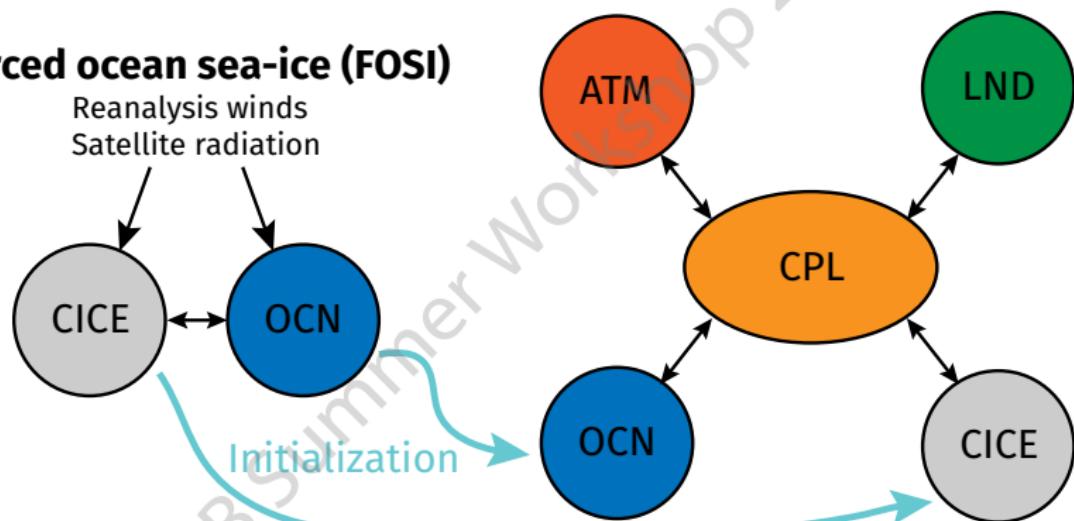
Initial-Value “Decadal Predictability”

Time-evolving distributions under changing external forcing



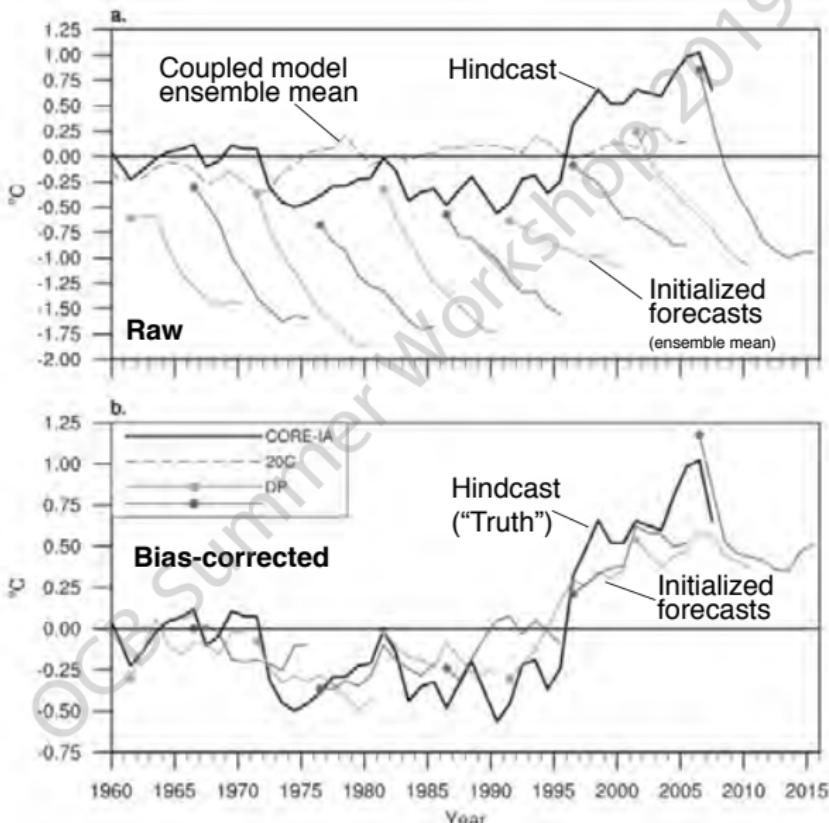
Branstator & Teng et al. 2010

Forced ocean sea-ice (FOSI) Fully-coupled CESM



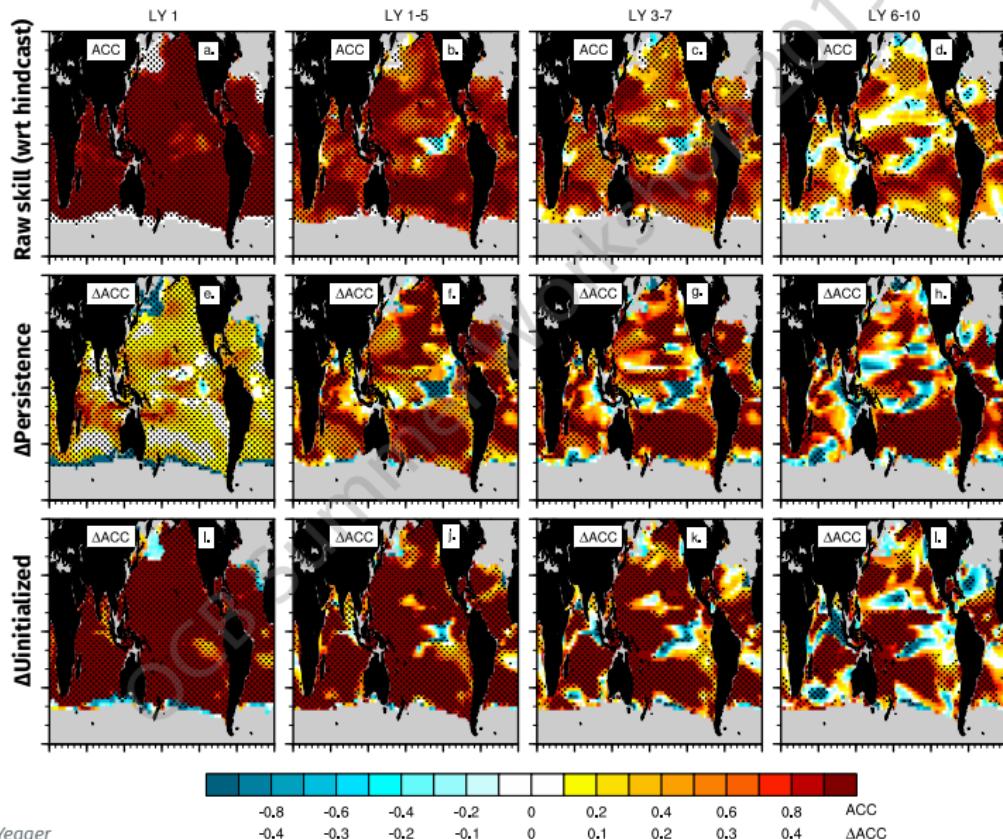
Skillful forecasts of upper ocean heat content on decadal timescale

Heat content anomaly, N. Atlantic Subpolar gyre ($z > -275$ m)



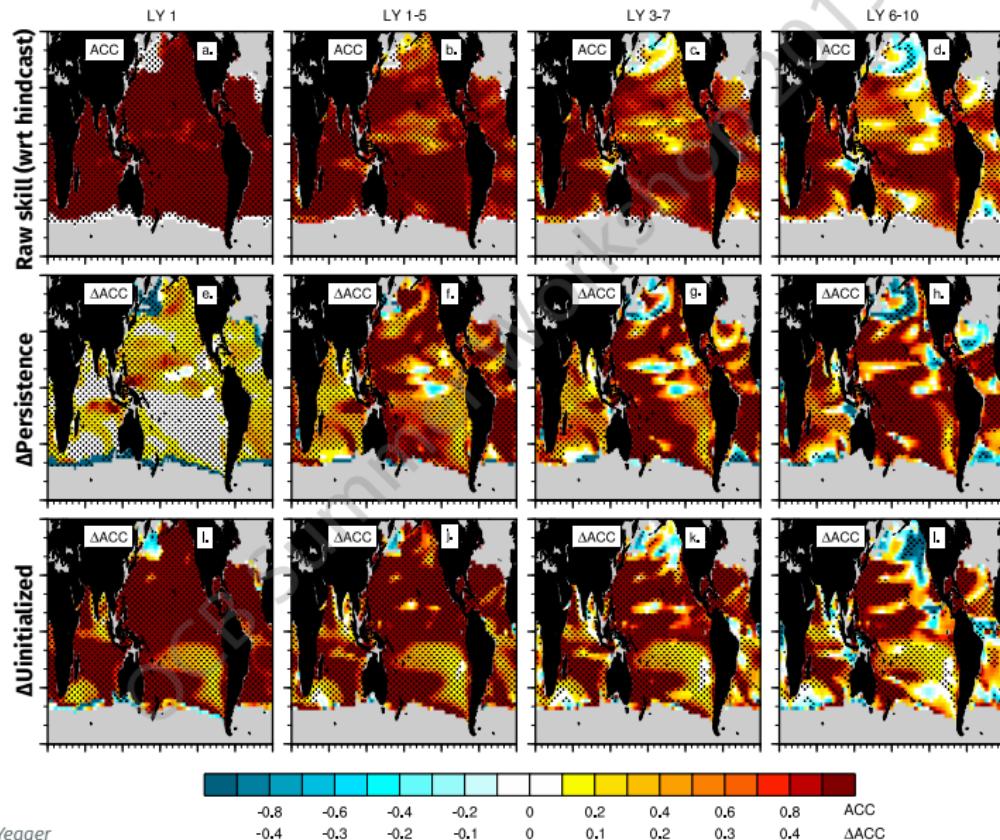
Thermocline oxygen concentrations look to be highly predictable

Anomaly correlation coefficient: O_2 on $\sigma_\theta = 26.5$

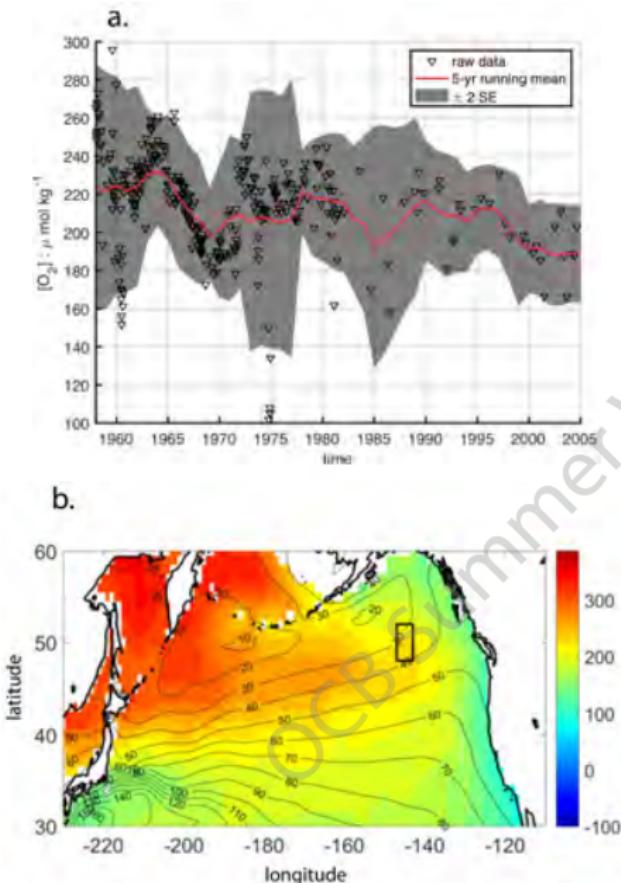


Thermocline oxygen concentrations look to be highly predictable

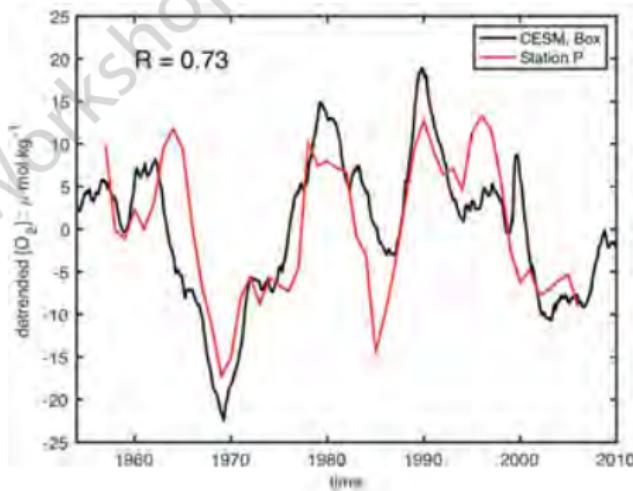
Anomaly correlation coefficient: Salinity on $\sigma_\theta = 26.5$



Oxygen variability at Ocean Station Papa ($\sigma_\theta = 26.5$)



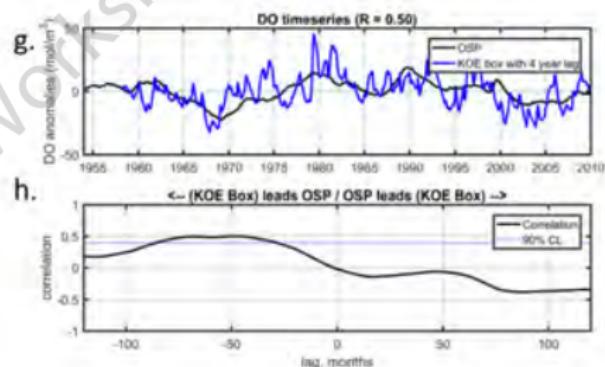
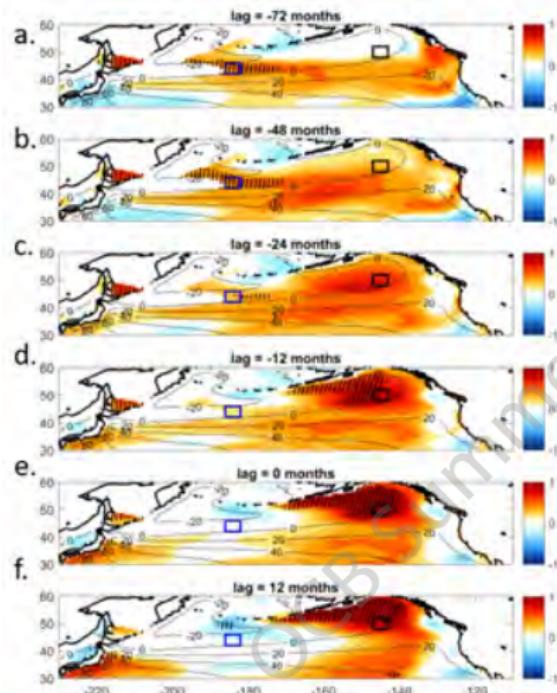
Detrended annual O_2 anomalies



Sun & Ito, in prep

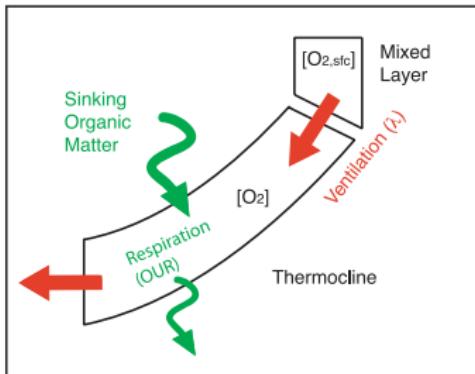
Oxygen variability at Ocean Station Papa ($\sigma_\theta = 26.5$)

Simulated lag correlation between OSP O₂ and full field



Sun & Ito, in prep

Oxygen variability at Ocean Station Papa ($\sigma_\theta = 26.5$)



Ito & Deutsch, 2010

Local autoregressive (AR1) model

$$\frac{dO_2}{dt} = -\lambda O_2 + f'$$

Linear Inverse Model (LIM)

$$\frac{dO_2}{dt} = LO_2 + f'$$

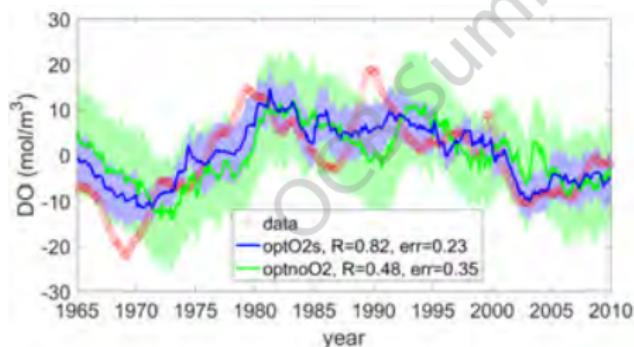
$$C(\tau) = G(\tau)C(0)$$

$$G(\tau) = \exp(L \cdot \tau)$$

$$O_2(t + \tau) = G(\tau)O_2(t)$$

$$L = \tau_0^{-1} \ln [C(\tau_0)C(0)^{-1}]$$

LIM: 3-month OSP forecast



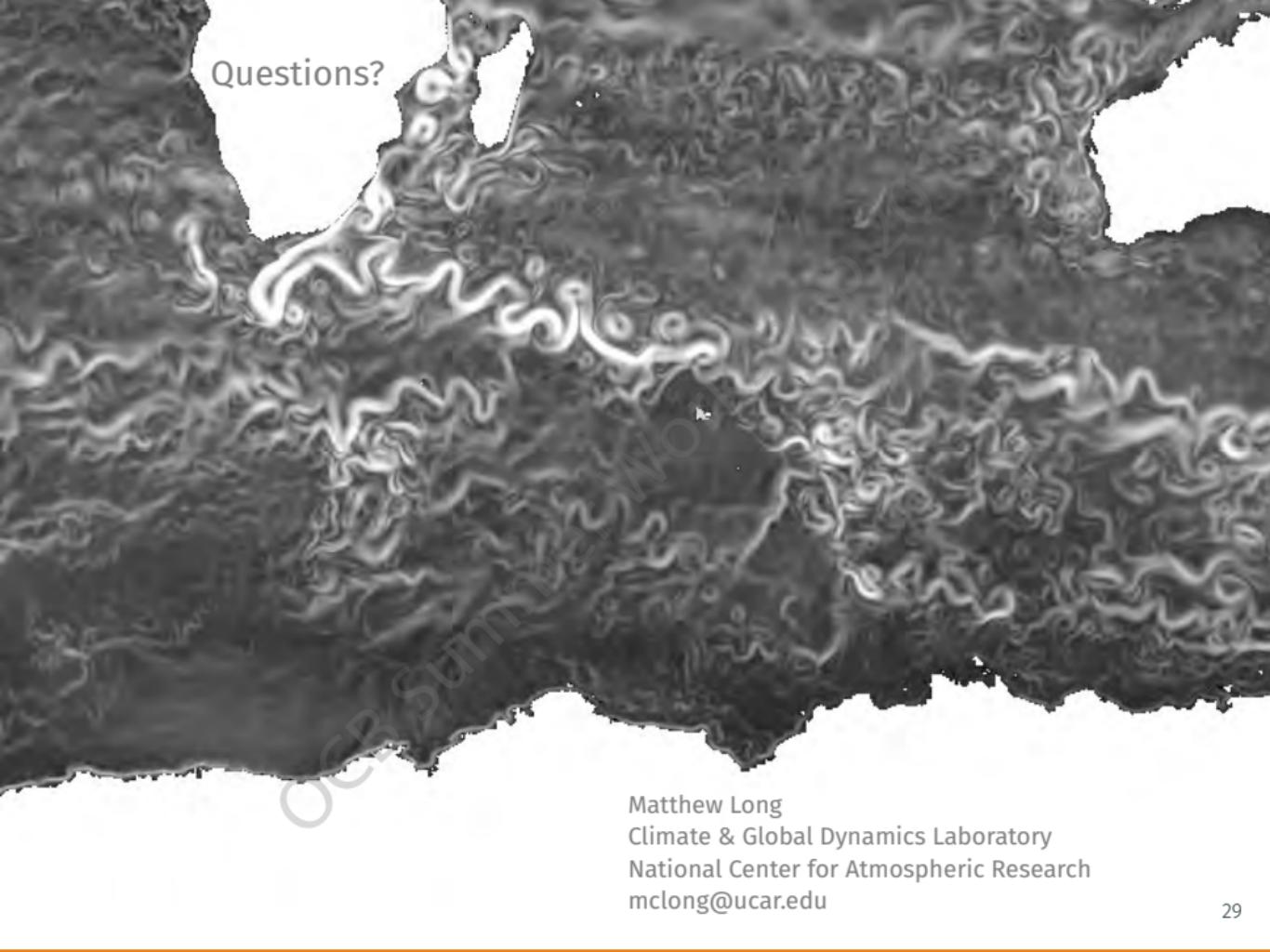
Sun & Ito, in prep

Summary

- Forced changes in O_2 result from combined effects of solubility and AOU; models show reasonable agreement on changes at high-latitudes, but weak consensus in the tropics.
- Models underestimate AOU-driven deoxygenation in the upper ocean; it is unlikely that internal variability can explain this discrepancy.
- Natural variability challenges definitive attribution of O_2 trends.
- Capabilities are emerging that enable predicting O_2 on seasonal to interannual timescales.

Acknowledgements

- Eddebar, Y. A., K. B. Rodgers, M. C. Long, A. C. Subramanian, S.-P. Xie, and R. F. Keeling (2019), El Niño–Like Physical and Biogeochemical Ocean Response to Tropical Eruptions, *J. Climate*, 32(9), 2627–2649, 10.1175/jcli-d-18-0458.1.
- Eddebar, Y. A., M. C. Long, L. Resplandy, C. Rödenbeck, K. B. Rodgers, M. Manizza, and R. F. Keeling (2017), Impacts of enso on air-sea oxygen exchange: Observations and mechanisms, *Global Biogeochemical Cycles*, 31(5), 901–921, 10.1002/2017gb005630.
- Ito, T., S. Minobe, M. C. Long, and C. Deutsch (2017), Upper Ocean O₂ trends: 1958–2015, *Geophys. Res. Lett.*, 10.1002/2017GL073613.
- Long, M. C., C. A. Deutsch, and T. Ito (2016), Finding forced trends in oceanic oxygen, *Global Biogeochem. Cycles*, 30, 10.1002/2015GB005310.
- Long, M. C., T. Ito, and C. Deutsch (2019), Oxygen projections for the future, in *Ocean deoxygenation: everyone's problem. Causes, impacts, consequences and solutions.*, edited by D. Laffoley and J. Baxter, IUCN, Gland, Switzerland, in press.
- Moore, J. K., W. Fu, F. Primeau, G. L. Britten, K. Lindsay, M. C. Long, S. C. Doney, N. Mahowald, F. Hoffman, and J. T. Randerson (2018), Sustained climate warming drives declining marine biological productivity, *Science*, 359(6380), 1139–1143, 10.1126/science.aa06379.
- Yeager, S. G., G. Danabasoglu, N. Rosenbloom, W. Strand, S. Bates, G. Meehl, A. Karspeck, K. Lindsay, M. C. Long, H. Teng, and et al. (2018), Predicting near-term changes in the earth system: A large ensemble of initialized decadal prediction simulations using the Community Earth System Model, *Bull. Amer. Meteor. Soc.*, 10.1175/bams-d-17-0098.1.

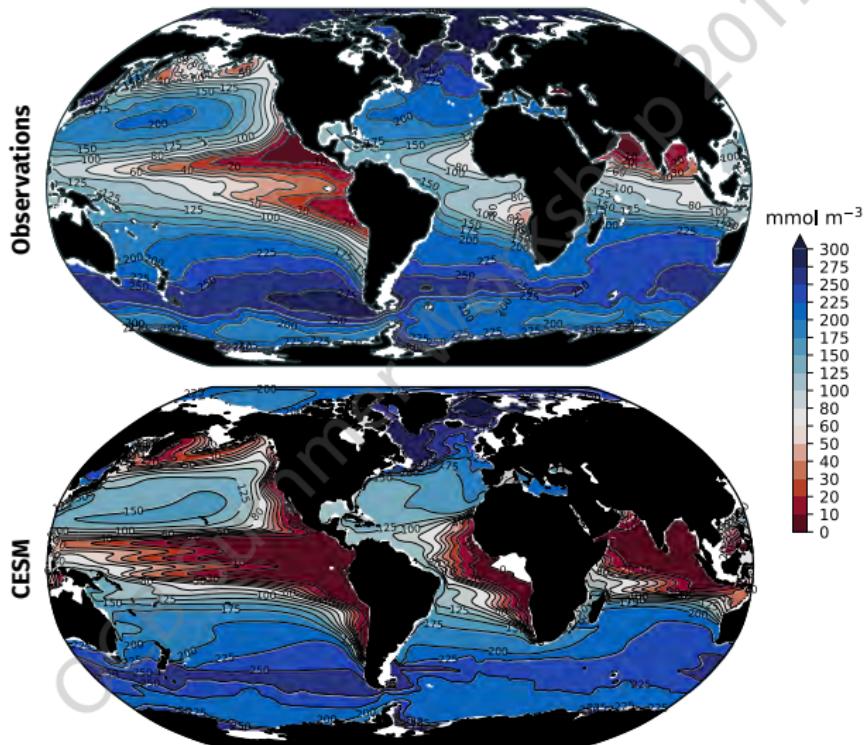


Questions?

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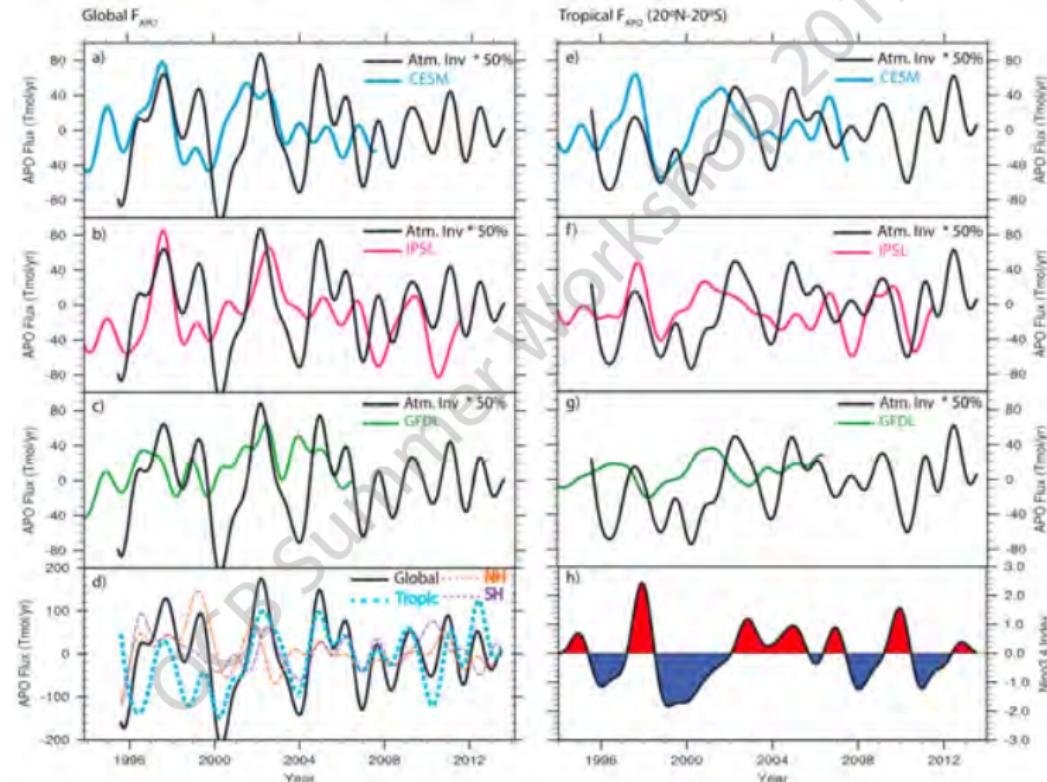
A persistent bias in Earth system models: Extensive OMZs

Thermocline (400–600 m) O₂ distributions



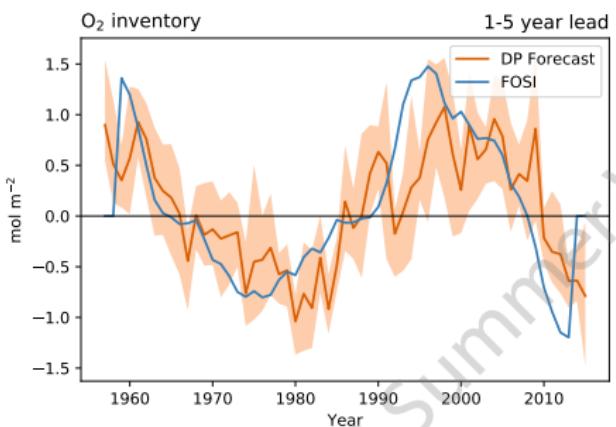
Simulated variability is weak

Simulated APO (mostly O₂) fluxes versus atmospheric inversion estimates

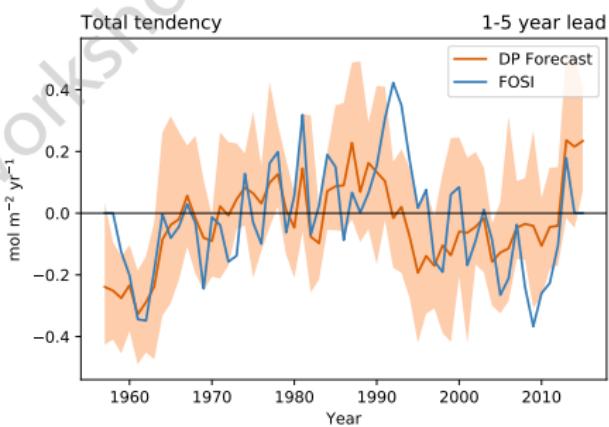


CalCOFI dissolved oxygen is skillfully predicted

Thermocline O₂ in CalCOFI region

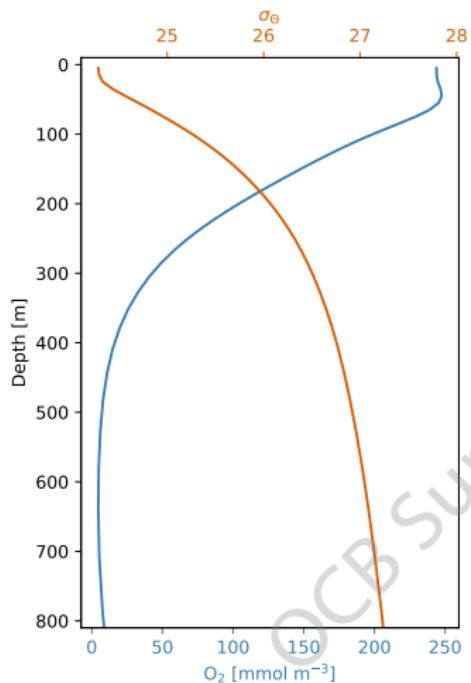


Thermocline O₂ tendency



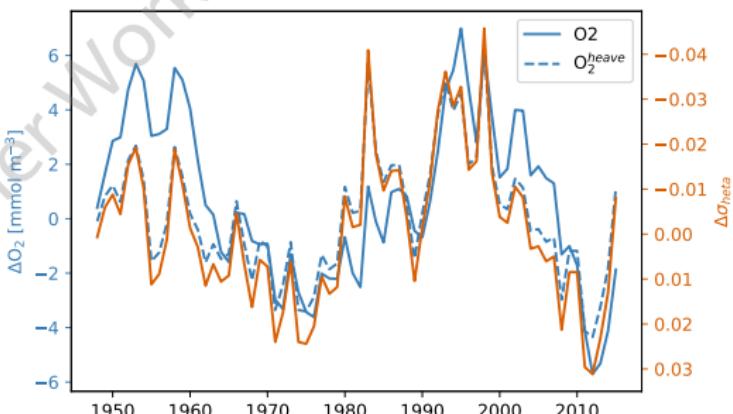
What mechanisms provide predictability for O₂?

Mean vertical gradients



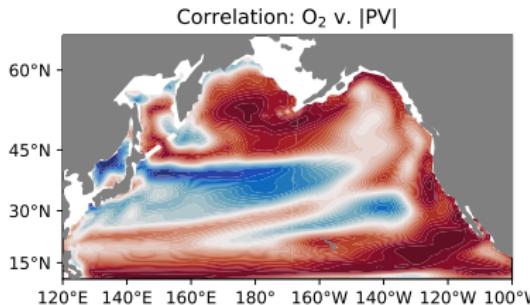
“Heave”

$$O_2^{heave} = \left(\frac{\partial \overline{O_2}}{\partial z} \right) \left(\frac{\partial \overline{\rho_\theta}}{\partial z} \right)^{-1} \rho'_\theta$$

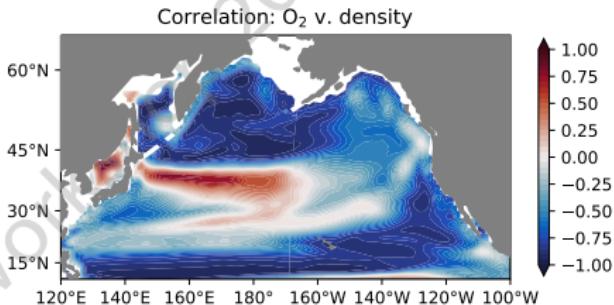


East-west difference in anomaly generation mechanism

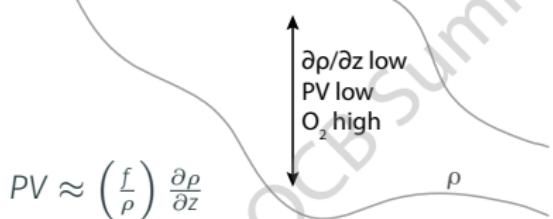
“Ventilation regime”



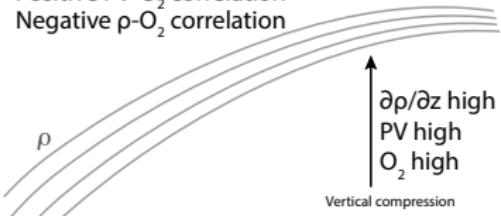
“Heave regime”



Negative PV-O₂ correlation

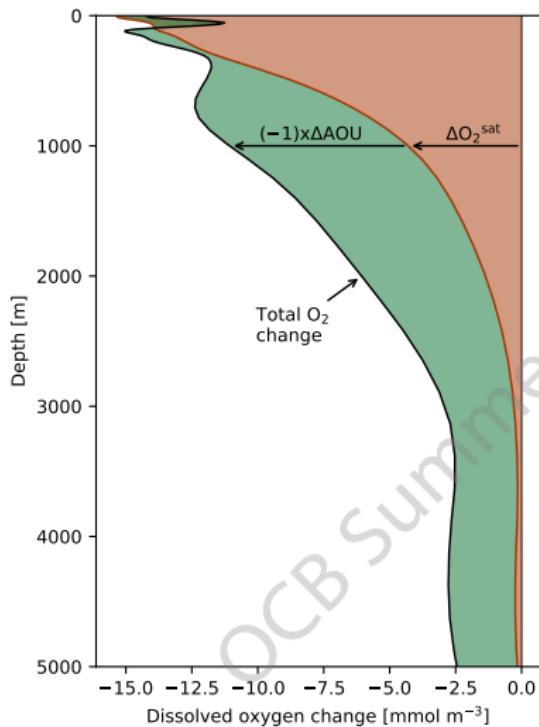


Positive PV-O₂ correlation
Negative p-O₂ correlation



CESM-LE projection: global-mean drivers of deoxygenation

O₂ change at 2100

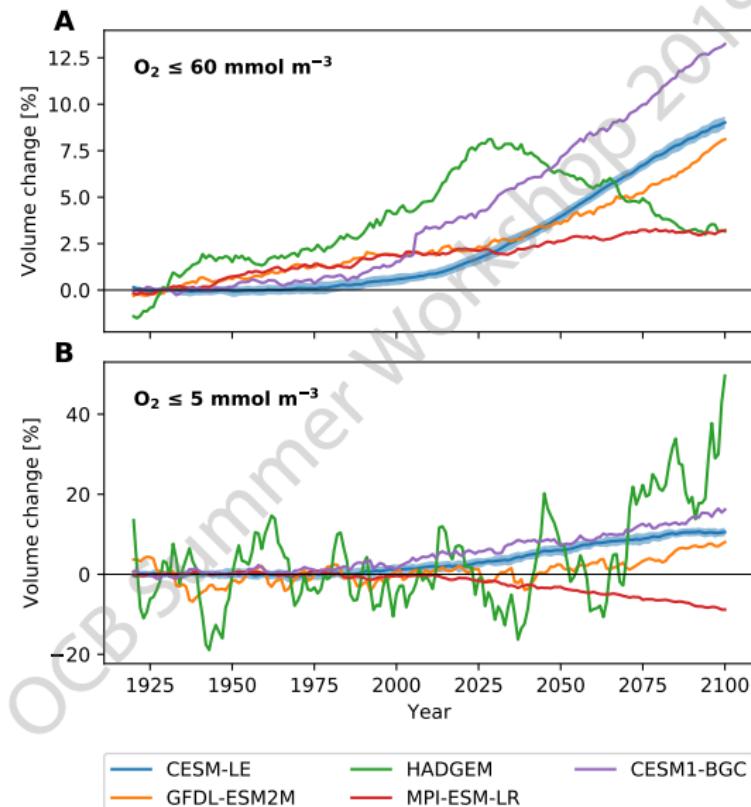


$$\Delta \text{O}_2 = \Delta \text{O}_2^{\text{sat}} - \Delta \text{AOU}$$

- Warming declines with depth;
- Surface AO₂ reduction: closer to equilibrium;
- Deep deoxygenation is AO₂-dominated.

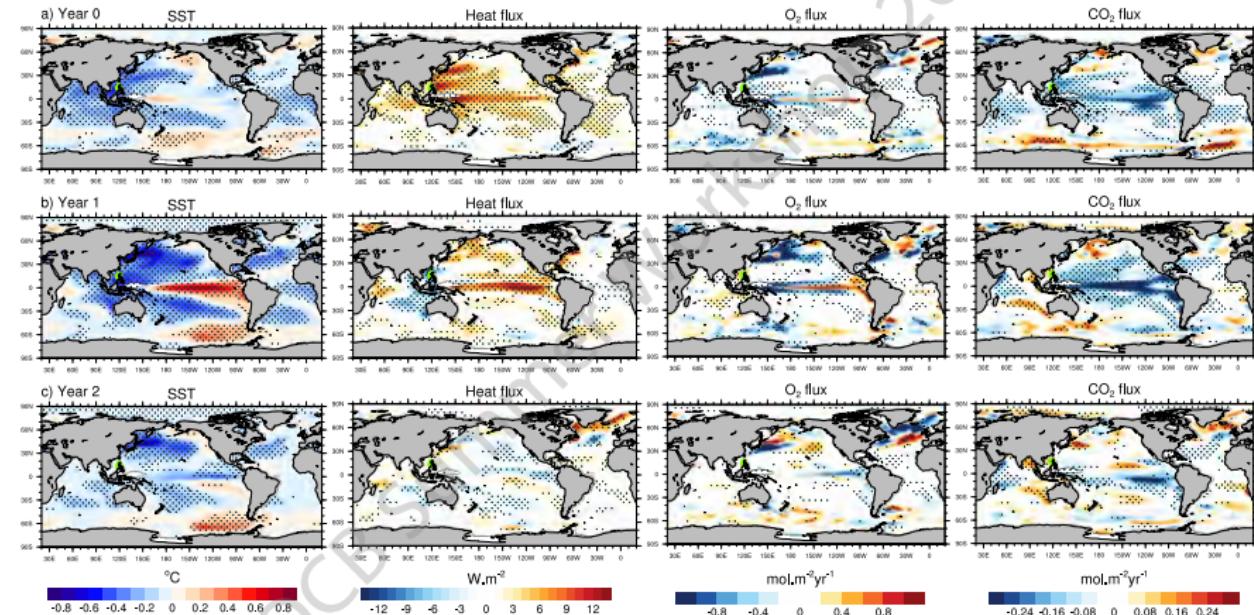
Suboxic and hypoxic volumes projected to increase

Simulated change in oxygen deficient zones



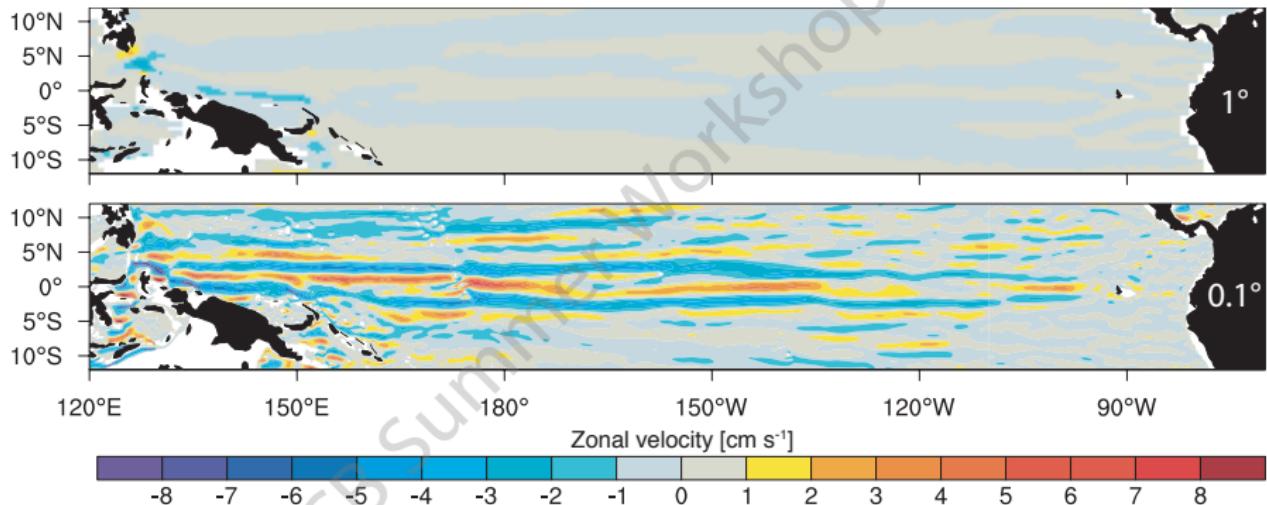
Imprint of volcanic eruptions on ocean biogeochemistry

Response to Pinatubo (CESM-LE)



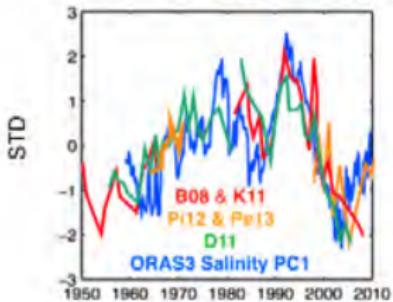
Model resolution determines ventilation dynamics

Simulated zonal velocity at 1000 m

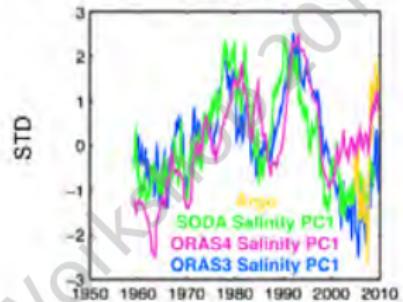


O₂ anomalies in the California Current

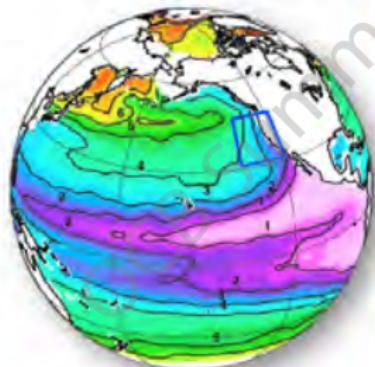
a Observations of oxygen in the CCS



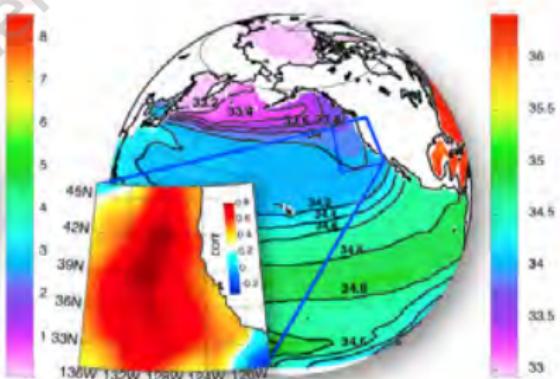
b Salinity PC1



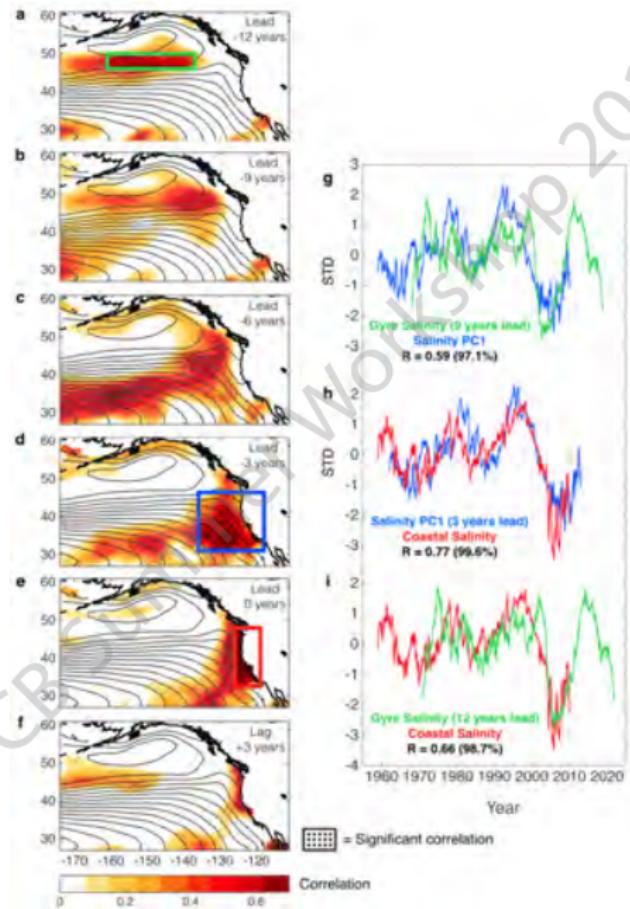
c Oxygen (ml/l) at $\sigma_\theta = 26.5$



d Salinity at $\sigma_\theta = 26.5$



O₂ anomalies in the California Current



Reinforcing drivers at high-latitudes; compensation in tropics

Simulated change in zonal-mean O₂

