Improvements in Estimating Upper Ocean Heat Content in the North Atlantic Ocean with the NOAA Next-Generation Enterprise Ocean Heat Content Algorithm



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Motivation: Improve tropical storm prediction at NOAA — Particularly rapid intensification and weakening

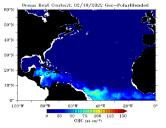


Overview

Motivation: Improve tropical storm prediction at NOAA — Particularly rapid intensification and weakening

Objective: Improve estimates of Ocean Heat Content (OHC). Current products generally climatologically based.





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Overview

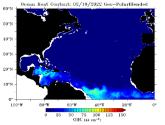
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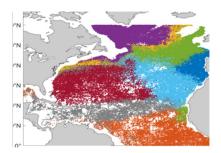
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Method: Use a variety of modern techniques including the Geostrophic Empirical Mode (GEM) and AI/ML to:

- 1) Quality control Θ & S profiles
- 2) Identify hydrographic regimes
- 3) Construct lookup tables







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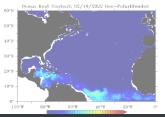
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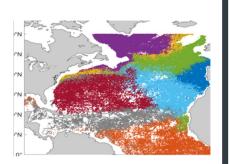
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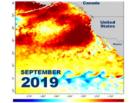
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Many applications for Ocean Heat Content across forecasting, monitoring, and planning domains



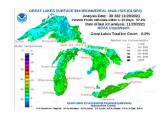
HURRICANE INTENSITY



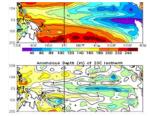
MARINE HEATWAVES



ECOSYSTEM MANAGEMENT (EBFM)



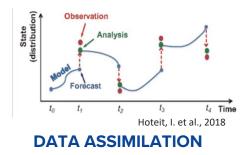
ICE EXTENT FORECASTING



METRICS (A SUBSTITUTE FOR REANALYSES)



PLANNING RESILIENT COASTAL INFRASTRUCTURE





INFORMING RISK ESTIMATES





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-Citation for "Data Assimilation" picture: Hoteit, I. et al., 2018: Data assimilation in oceanography: Current status and new directions. In "New Frontiers in Operation Oceanography", doi:10.17125/gov2018.ch17.

What is the Next Generation Enterprise Ocean Heat Content (NGE OHC) Product?

A statistically robust, observationally-based daily estimate of ocean thermohaline conditions $(\Theta(z), S(z))$ to ~1800 m, with a focus on high accuracy in the upper ocean.



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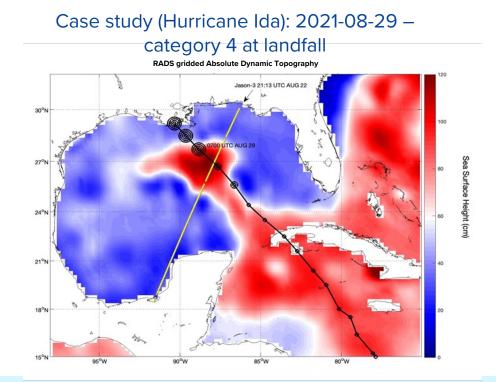
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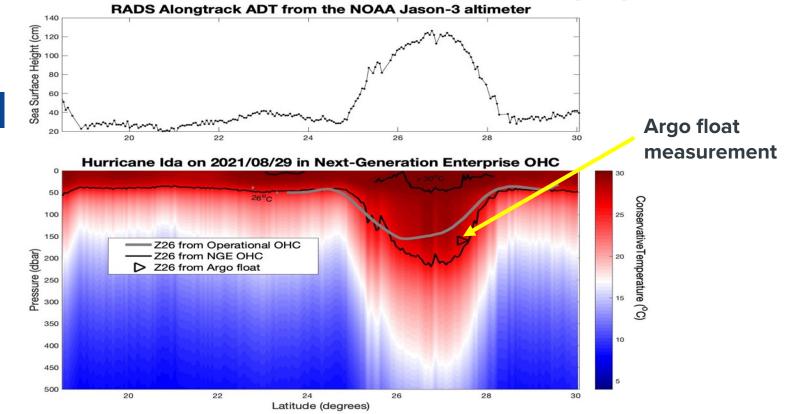
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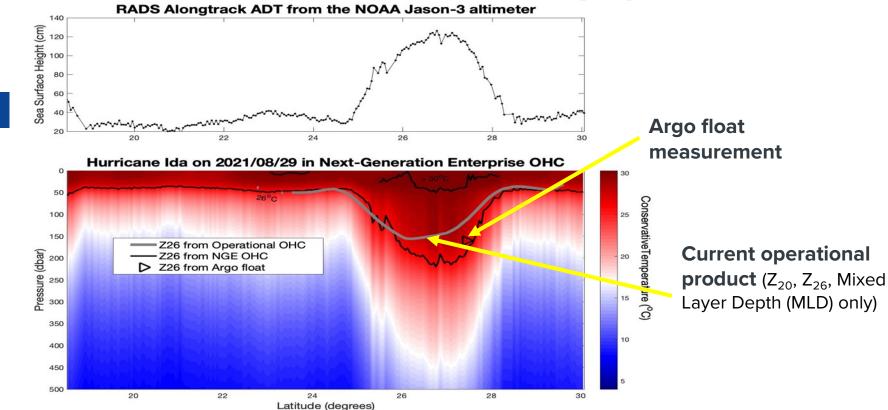
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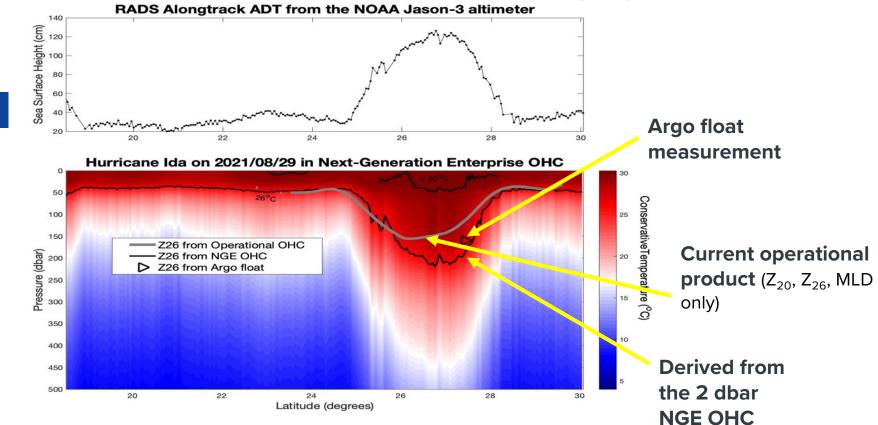
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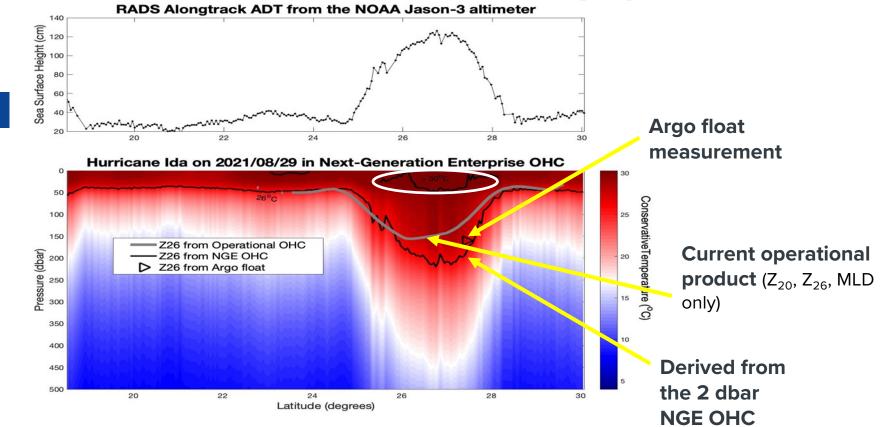
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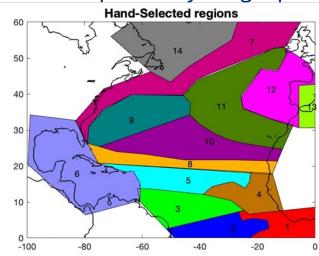
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How does the NGE OHC Work? Uses the BLT method (a meaty sandwich with trimmings)

Collect a large database of ocean profile data. Apply a method to divide it up into hydrographically (dynamically) similar regions:



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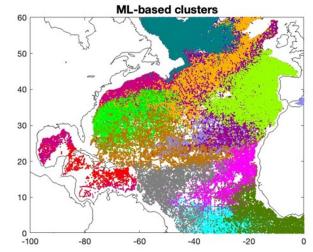
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How does the NGE OHC Work?

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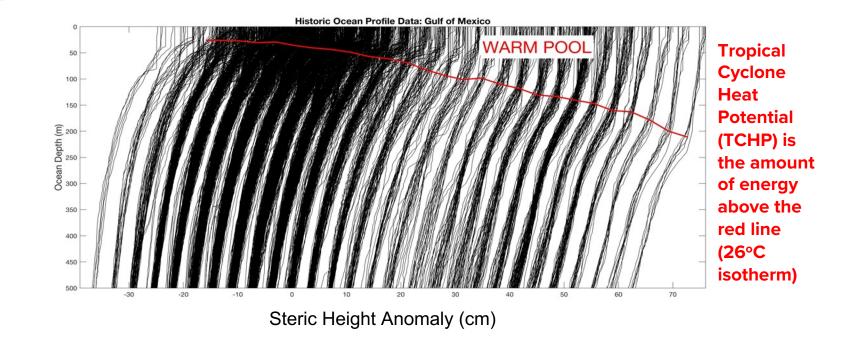
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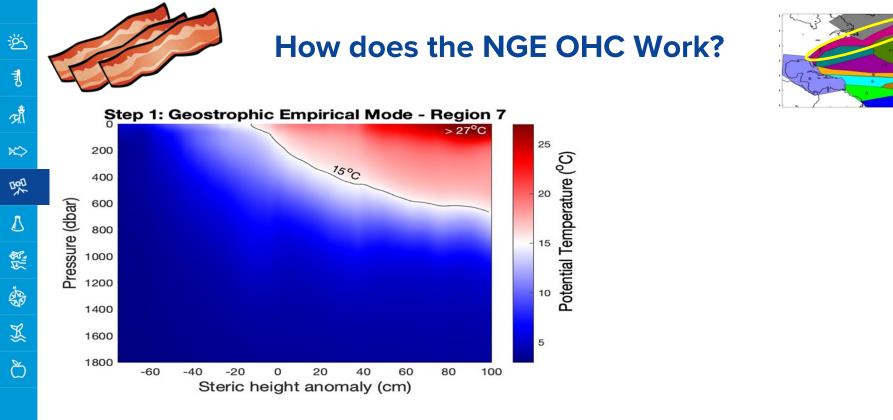
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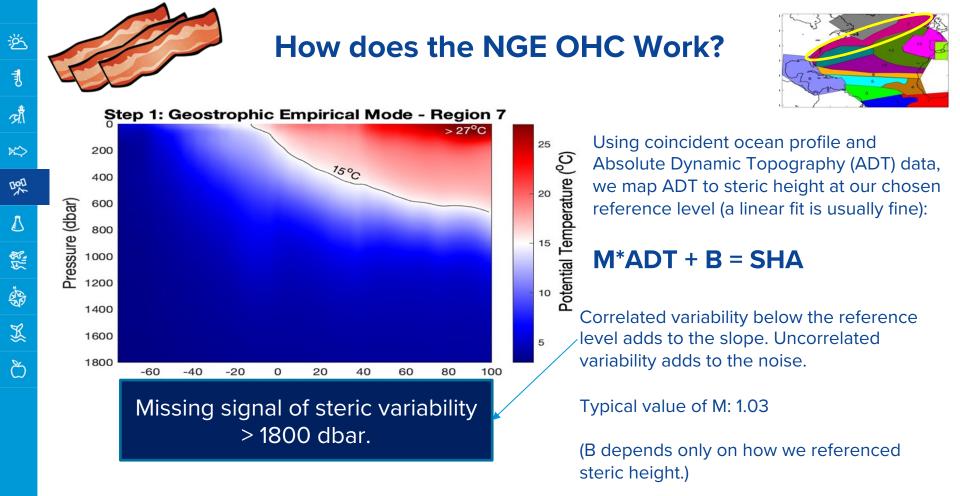


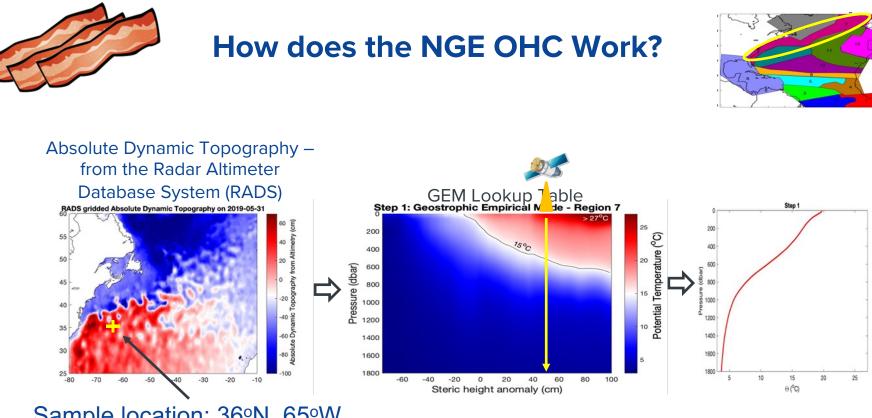
Collect a large database of ocean profile data. Organize it with respect to a parameter of interest, in our case **Steric Height Anomaly (SHA)**



For each region/cluster, create a lookup table based on ocean profile data







Sample location: 36°N, 65°W

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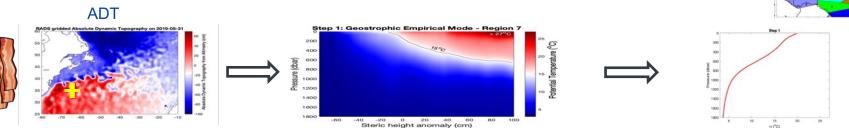
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GEM is the meat of the algorithm



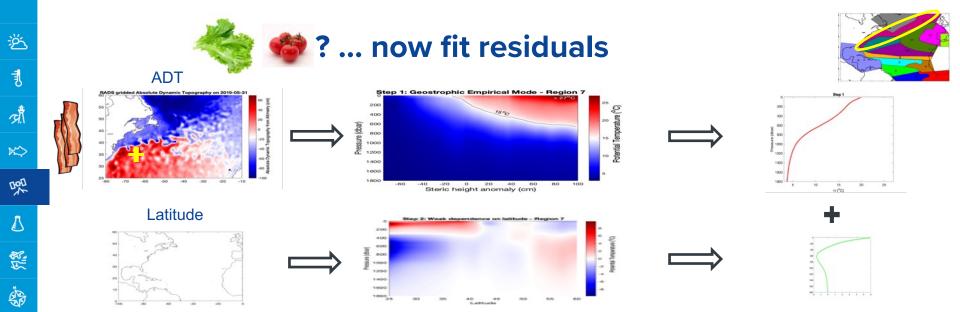
Goal: A set of lookup tables for each region

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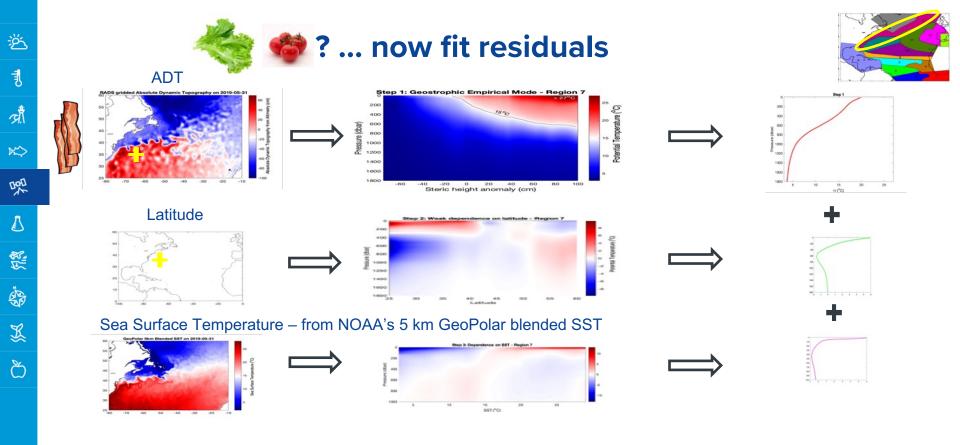


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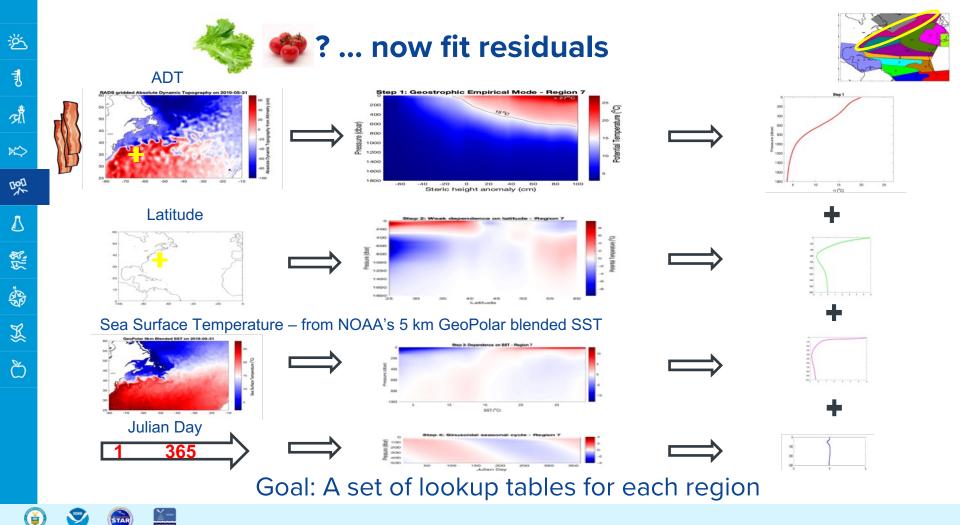
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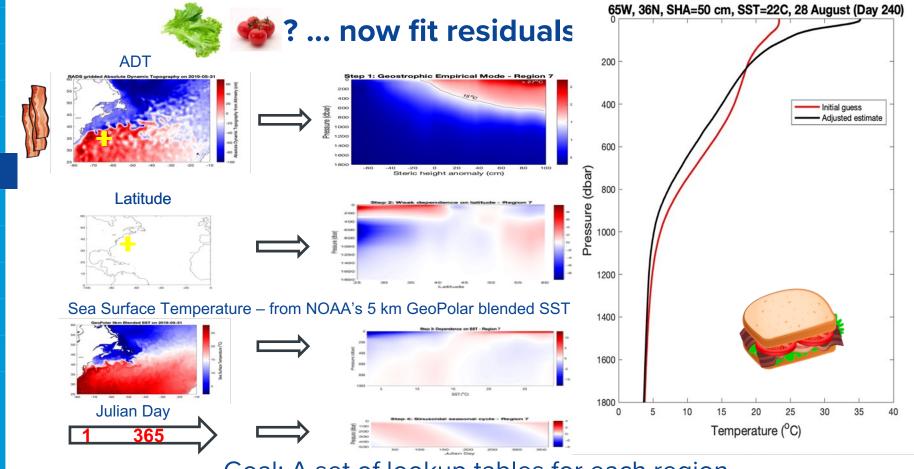
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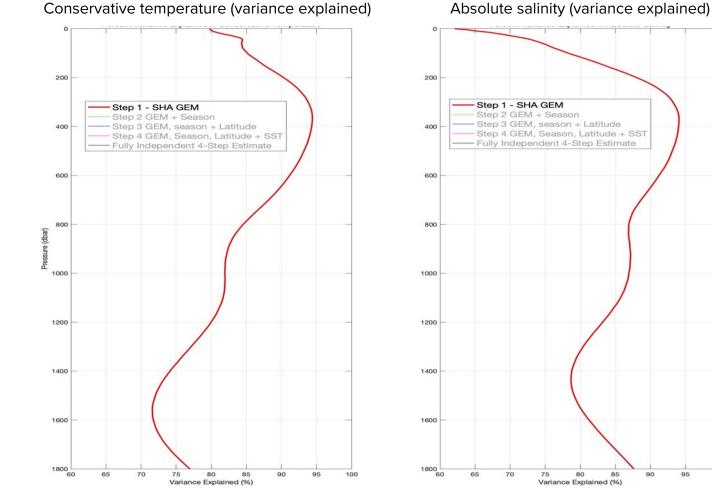
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Goal: A set of lookup tables for each region



How well do the NGE OHC profiles represent ocean conditions?

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Conservative temperature (variance explained)

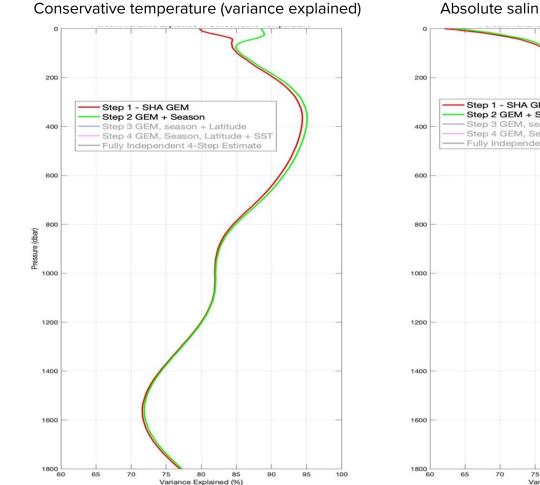
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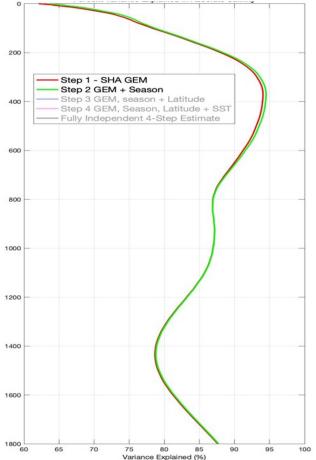
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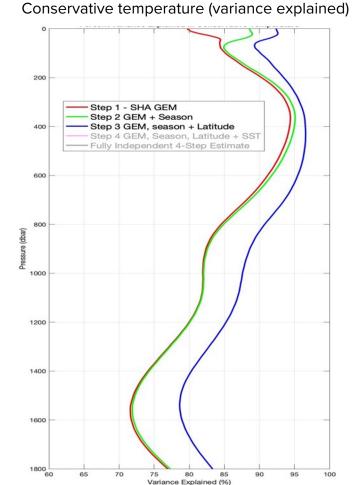
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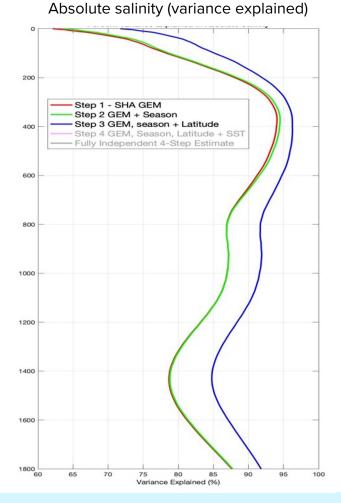
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Absolute salinity (variance explained)







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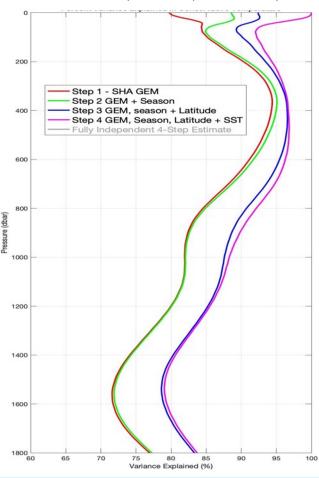
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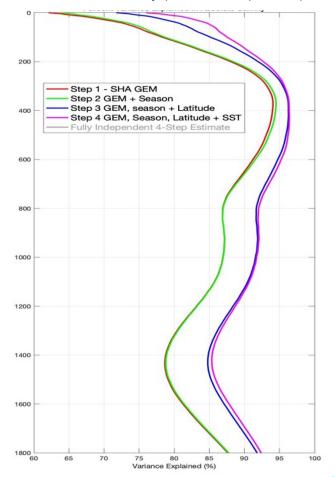
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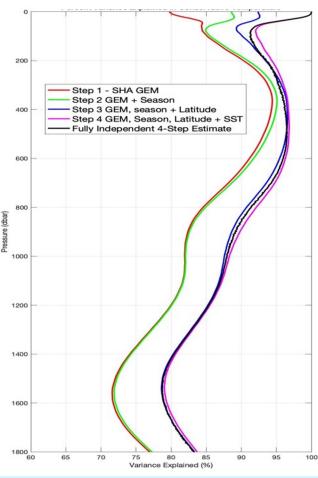
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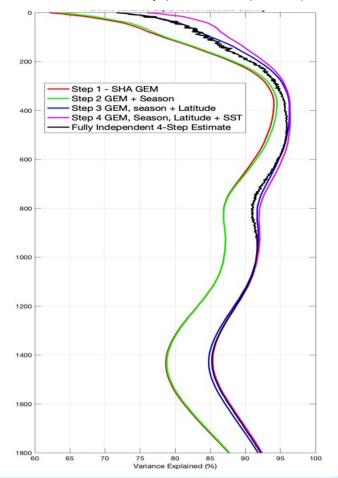


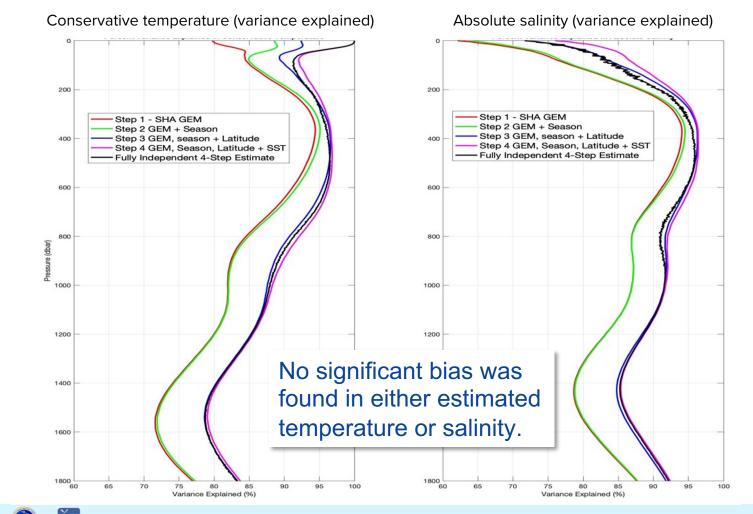


Conservative temperature (variance explained)



Absolute salinity (variance explained)





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Summary and Conclusions



Adaptable algorithms are needed for a warming world

- o Dynamically based NGE OHC better captures extremes
- Use of ML/AI allows for input of new profiles and annual updates to lookup tables
- o Based on satellite inputs: RADS ADT & NOAA GeoPolar blended SST
- Good depth resolution in T/S allows OHC (TCHP) & MLD estimates and other metrics to be derived

Coming in 2023

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- Finalize selection of (ML/AI) methods for clusters & residual lookup tables
- \circ Daily ¹/₄ degree gridded T(z), S(z) fields for the North Atlantic
- Daily 7-km along-track fields using RADS ADT: Jason-3, CryoSat-2, Sentinel-6 MF, Sentinel-3A,-3B, SARAL

Future Work

- o Expansions: Global Open Ocean; Coastal Zone (< 1800 dbar)
- o Error Estimates



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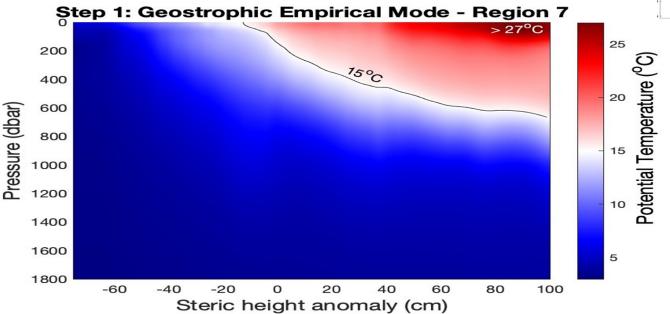
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How does the NGE OHC Work?





All variability that has a predictable sea surface height expression has been removed. **Anything left is density-compensated within the water column.**



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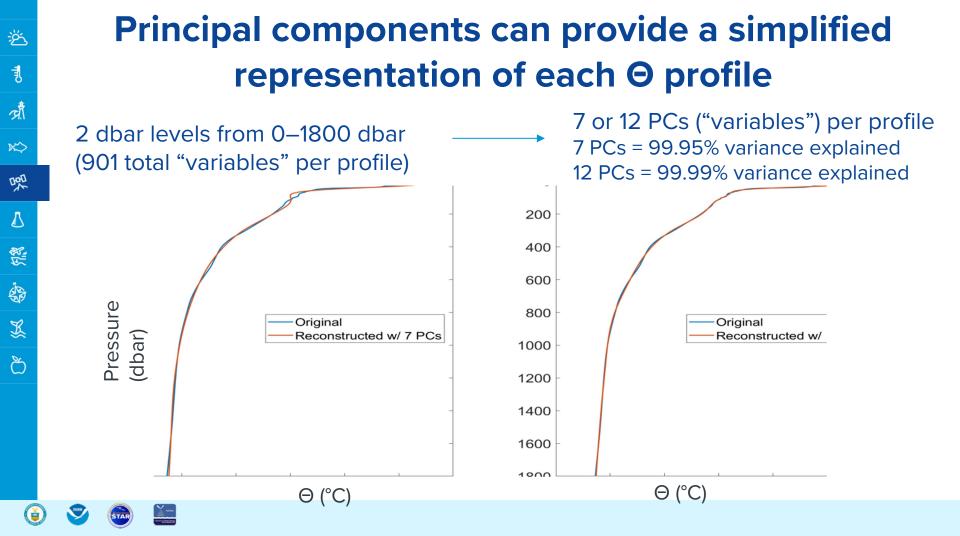
The selected reference level is a trade-off between including more profiles, and including more of the water column steric height variability. Number of Observations Variability in Steric Height Anomaly N_{700 dbar} = 332105 1000 1000 5 cm of σ_{SHA} added due to 11x bigger sample size N_{1800 dbar} = 166130 2000 2000 N_{2000 dbar} = 96411 N_{3000 dbar} = 15224 3000 3000 2.8 cm of $\sigma_{\rm SHA}$ unaccounted for due to variability between 1800 - 3000 dbar Pressure (dbar) 4000 $N_{5000 \text{ dbar}} = 2811$ 5000 5000 6000 6000 True variability in Steric Height Anomaly from variability in stratification and thermohaline 7000 7000 properties below 3000 dbar is difficult to estimate due to the extremely small sample size. However, it may be another 2 to 3 cm. 8000 8000 0 0.5 1.5 2.5 3 0 10 15 20 25 30 35 40 45 50 Number of Observations $\times 10^{5}$ Std. dev (o) Steric Height Anomaly (SHA)

Technical bits

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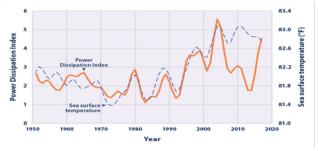


Focus on Hurricanes

- A tropical cyclone making landfall in the USA causes on average ~\$19.2 billion in damage.
- There have been more category 4 and 5 landfalls between 2017 – 2021 than from 1963 – 2016 and these storms collectively did \$353 billion damage to the USA through 2020.
- The over-ocean intensification of tropical storms can bring stronger winds and greater rainfall to land areas, making them even more dangerous and costly. This effect was seen in Hurricane Katrina, which went on to cause a record \$161 billion in damage.
- Of the nine strongest tropical storms to hit the continental USA, only one was hurricane strength more than 72 hours before landfall.
 - In 2021, Hurricane Ida brought a 3 m storm surge, left 1.5 million people without electricity, produced up to 432 mm of rain, and increased in strength by 105 kmh in its last 24 hours prior to landfall.
 - Hurricane Ian



New York Times, August 30, 2021



Emanuel, K.A. 2021 *update to data originally published in*: Emanuel, K.A. 2007. Environmental factors affecting tropical cyclone power dissipation. *J. Climate* **20**(22):5497– 5509.

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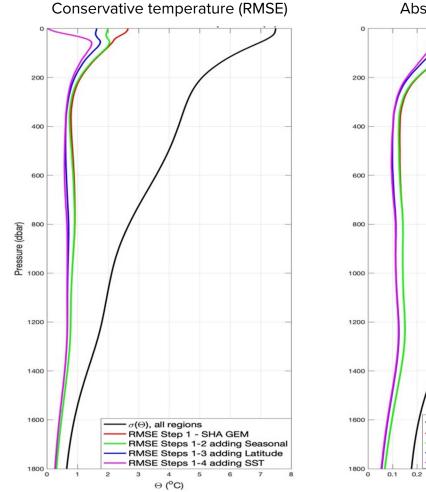
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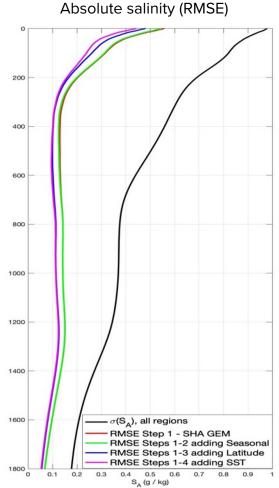
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Case Study: Hurricane Fiona

A collaboration between NESDIS/STAR/SOCD and OAR/AOML/HRD

HAFS_S, New NESDIS OHC, and ocean observations

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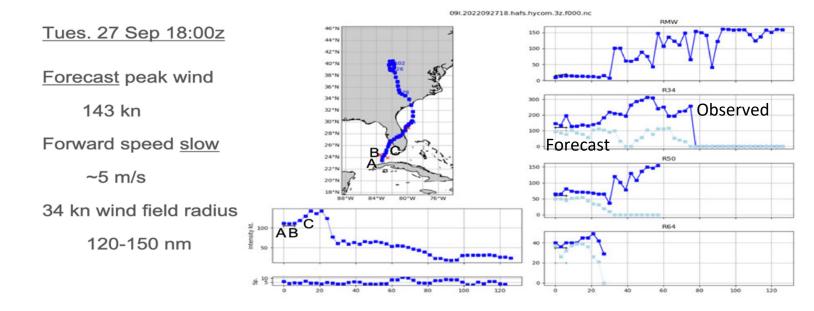
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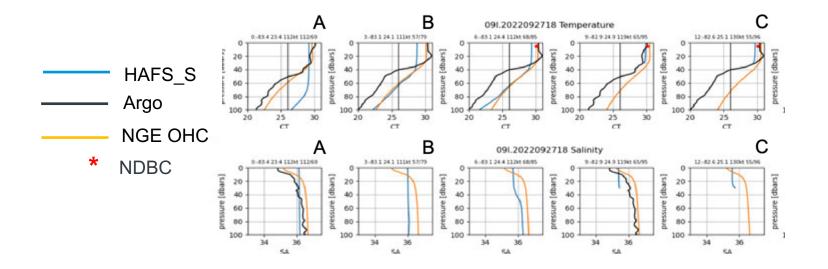
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Objectives of the NOAA Next Generation Enterprise Ocean Heat Content (NGE OHC) Algorithm

- Using primarily satellite altimetry as input, provide 4-Dimensional (spatial, depth, temporally resolved) estimates of ocean heat and salt
- Provide $\frac{1}{4}^{\circ}$ global coverage, and high resolution (7 10 km) where & when needed.
- Be able to use input from many (and new) satellites without requiring modification or recalibration.
- Exhibit accuracy and fidelity in extreme environmental conditions, e.g., anomalously high heat content.
- Adaptable be able to be updated in the face of a warming ocean. Updating uses advanced computational techniques (AI/ML) to be as routine and automated as possible.



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