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# 2 Internal gravity wave modeling on basin to global scales and its impacts on acoustic3 propagation

- 4 Martha C. Schönau<sup>1\*,7</sup>, Luna Hiron<sup>2</sup>, John Ragland<sup>3,10</sup>, Keshav J. Raja<sup>2</sup>, Joseph Skitka<sup>4,8</sup>, Miguel
- 5 S. Solano<sup>5,9</sup>, Xiaobiao Xu<sup>2</sup>, Brian K. Arbic<sup>4</sup>, Maarten C. Buijsman<sup>5</sup>, Eric P. Chassignet<sup>2</sup>,
- 6 Emanuel Coelho<sup>1</sup>, Robert W. Helber<sup>6</sup>, William Peria<sup>3</sup>, Jay F. Shriver<sup>6</sup>, Jason E. Summers<sup>3</sup>,
- 7 Kathryn L. Verlinden<sup>1</sup>, Alan J. Wallcraft<sup>2</sup>
- 8 <sup>1</sup>Applied Ocean Sciences (AOS), LLC
- 9 <sup>2</sup>Center for Ocean-Atmospheric Prediction Studies, Florida State University
- 10 <sup>3</sup>Applied Research in Acoustics LLC (ARiA)
- <sup>4</sup>Department of Earth and Environmental Sciences, University of Michigan
- 12 <sup>5</sup>School of Ocean Science and Engineering, The University of Southern Mississippi
- 13 <sup>6</sup>Naval Research Laboratory, Ocean Dynamics and Prediction
- 14 <sup>7</sup>Scripps Institution of Oceanography, University of California, San Diego
- 15 <sup>8</sup>Department of Physical Oceanography, Woods Hole Oceanographic Institution
- 16 <sup>9</sup>SOFAR Ocean Technologies
- <sup>10</sup>Department of Electrical and Computer Engineering, University of Washington
- 18 \*Corresponding author: <u>mschonau@ucsd.edu</u>.
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# 20 Abstract:

Underwater acoustic propagation depends on ocean temperature, salinity, pressure, and
 bathymetry. The realistic representation of the ocean state and its underlying dynamics within
 ocean models is essential to accurately model and predict underwater acoustic propagation.

24 Stratified, high-resolution global ocean models that include tidal forcing have only been 25 developed in the last two decades. Tidal forcing introduces internal tides and generates higher 26 frequency (supertidal) internal waves. The inclusion of tides in global ocean models better 27 captures the ocean state. Through the disciplines of internal wave modeling, acoustics and 28 machine learning, we examined the movement of internal wave energy through numerical 29 simulations, how the inclusion of tides in the models may impact acoustic propagation, and how 30 machine learning may be used to simplify the modeling of these impacts. The project used 31 global, basin-scale, and idealized HYbrid Coordinate Ocean Model (HYCOM) simulations as 32 well as regional Massachusetts Institute of Technology general circulation model (MITgcm) 33 simulations to examine the impacts of tidal inclusion on sea surface height variability, the 34 propagation and dissipation of internal-wave energy, and the sensitivity of internal wave 35 modeling to vertical and horizontal grid spacing. Sound speed, acoustic parameters, and acoustic 36 propagation were compared between simulations both with and without tidal forcing. Deep 37 learning algorithms were used to replicate some of the ocean structure variability at test locations 38 through generation of a "tidal" simulation output from global HYCOM without tidal forcing. The 39 cross-disciplinary effort between ocean modeling and acoustic assessment elucidated the impacts 40 of certain ocean modeling choices (e.g., vertical and horizontal grid spacing, the hydrostatic 41 assumption, and others) on sound speed and explored methods to reduce computational costs.

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#### 43 1. Introduction

44 Global ocean models are used by many stakeholders who are interested in (among others) 45 how the ocean state (e.g., velocities, temperature, and salinity) may impact navigation, climate, 46 the dispersal of biogeochemical tracers, biological productivity, and acoustic propagation, such 47 as passive acoustic soundscape monitoring and acoustic communications between vessels or 48 marine mammals. The propagation of acoustic waves (sound) in the ocean depends on the 49 thermohaline structure of the ocean—the temperature, salinity, and pressure—and the bathymetry (shape of the seafloor). Sound refraction is caused by changes in temperature and 50 51 salinity structure that impact density (relative pressure) and its gradients, such as fronts, eddies, 52 currents, vertical stratification, and mixing. Processes that impact the thermohaline structure 53 include energy input from wind, salinity changes at the ocean surface from evaporation and 54 precipitation, surface temperature evolution from ocean-atmosphere heat exchange, and spatial

and temporal variability of large-scale ocean currents that then spawn mesoscale eddies (with horizontal scales of about 100 km). A tidally forced ocean model, which simulates energy that can propagate then as internal tides and waves, will create sound-speed structure that may be very different from that of a non-tidally forced ocean model.

59 This paper presents new advancements in the modeling of internal tides and gravity 60 waves and their impact on the propagation of acoustic waves. Internal gravity waves (IGWs) are 61 waves with a restoring force of gravity, that exist as undulations along constant density ocean 62 layers called isopycnal surfaces. IGWs create a profile of depth-dependent velocities, alongside the displacement of isopycnals. They are often discussed in terms of their vertical structure, or 63 64 "modes" (Gill, 1982), which decomposes internal tides and waves to study the dynamics of IGW interactions and energy transfer. In this method, oceanic IGWs can be approximated as a linear 65 superposition of standing waves in the vertical direction and propagating waves in the horizontal 66 67 direction, which is a reasonable approximation in buoyancy driven flows where the horizontal 68 scale of the flow is much greater than that in the vertical. Each wave mode has a characteristic 69 length and phase speed (eigenvalue) and vertical structure (eigenfunction) that depends on the 70 frequency of the IGW, the Coriolis frequency (arising from the projection of Earth's rotation), 71 and vertical stratification (i.e., vertical density gradient or buoyancy frequency). The lowest-72 baroclinic mode has a singular, two-layer horizontal structure (i.e., the velocities are out of phase 73 above and below the thermocline), and higher modes have greater vertical structure.

74 Internal tides are a special case of internal gravity waves that exist at tidal frequencies 75 and are generated by tidal flow over bathymetric features of the seafloor (e.g., Bell, 1975; Arbic 76 et al., 2004; Buijsman et al, 2020). Such flows create vertical motions, including vertical 77 displacements along isopycnal surfaces. These are different from IGWs generated by high-78 frequency wind forcing, which causes near-inertial IGWs with a frequency that is near the 79 Coriolis frequency (Pollard and Millard, 1970; Alford, 2003; Simmons and Alford, 2012, Raja et 80 al., 2022). As the Coriolis frequency increases towards the poles, near-inertial IGW oscillation frequency also depends on latitude. Aside from internal tides and near-inertial waves, there exists 81 82 an internal gravity wave continuum spectrum (Garrett and Munk, 1975). The high-frequency 83 waves in the IGW continuum spectrum have supertidal frequencies (frequencies greater than 84 tidal frequencies) and are thought to arise from nonlinear interactions between internal tides and 85 near-inertial waves (Olbers, 1976; McComas and Bretherton, 1977; Müller et al., 1986).

86 Mesoscale eddies may also play a role in the formation of the spectrum (Barkan et al., 2017;

87 Skitka et al., 2024; Delpech et al., 2024). These nonlinear interactions can change the length and
88 time scales of IGWs, and we saw some evidence of them in this study.

IGW variability has not been well-captured by global ocean simulations, which may have 89 incomplete forcing (e.g., a lack of tidal forcing) or may parameterize, rather than resolve, 90 91 processes that happen at finer scales because the model horizontal and/or vertical discretization 92 is too coarse. Models of the large-scale barotropic tides (where there is uniform water movement 93 with depth) are often performed in a one-layer ocean of constant-density, that does not allow for 94 any flow fields with vertical structure or undulations of density surfaces. Barotropic tidal models 95 have been available since the 1970s (e.g., Hendershott, 1981). However, only in the last two 96 decades has the increase in computational power allowed accurate modeling of internal tides in a 97 stratified ocean. These have evolved from using uniform two-layer (Arbic et al., 2004) or 98 multilayer (Niwa and Hibiya, 2004; Simmons et al., 2004) stratification in the horizontal direction to the inclusion of realistic horizontally varying stratification by embedding tidal 99 100 forcing in ocean general circulation simulations (Arbic et al., 2010, 2012, 2018; Shriver et al., 101 2012; Buijsman et al., 2020; Arbic, 2022).

102 These models feature high-frequency undulations of density surfaces, including internal 103 tides and supertidal IGWs, that bring cold water up and push warm water down, and therefore 104 change the sound speed (Gill, 1982). The nonlinear interactions in the IGW continuum spectrum 105 eventually bring internal wave scales down to the order of 1 m or less, at which IGWs turn over 106 and break. Breaking IGWs play a dominant role in the mixing of the interior ocean, that is, the 107 ocean away from coastal boundaries, the air-sea interface, and the bottom boundary at the 108 seafloor (e.g., Munk and Wunsch, 1998; Alford, 2001; Waterhouse et al., 2014, MacKinnon et 109 al., 2017). The impacts on acoustic propagation will also likely be influenced by model grid 110 spacing, boundary conditions and parameterizations, as these all impact the IGW structure.

Internal waves and tides can cause significant impacts on sound speed variability and acoustic propagation (Flatté et al., 1979, Colosi et al. 2013). Acoustic tomography, an inverse method that uses long-range acoustic propagations to extrapolate the ocean structure, has been used to study the barotropic and baroclinic tides (e.g., Dushaw et al., 1995; Dushaw et al., 2011), and shown as a way to constrain ocean models themselves (Dushaw, 2019; Dushaw and Menemenlis, 2023). There have been many acoustical observational studies in strong tidal regions (e.g., Worcester et al., 2013) and efforts have been made to model these impacts using
both acoustic arrival times and acoustic mode amplitudes (Colosi and Flatté, 1996;

119 Chandrayadula et al., 2020). The advances in global models to include tidal forcing at high

120 resolutions provides an opportunity to address how acoustic propagation may responds to the

121 ocean model forcing and grid discretization.

122 This coordinated research project focuses on the modeling of IGWs and the impacts of 123 IGWs on underwater acoustic propagation. The project was funded under the Task Force Ocean 124 (TFO) Office of Naval Research (ONR) research initiative, and dubbed "TFO-HYCOM", after 125 the U.S. Navy's operational HYbrid Coordinate Ocean Model (HYCOM). This article describes 126 our progress in global internal wave modeling, our approach to assess the impacts that the 127 modeling choices may have on mid-range, upper-ocean acoustics, and a preliminary exploration 128 of how deep learning (DL) may be used to optimize the predictions of these impacts.

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#### 130 2. Internal wave modeling and their impacts on sound speed

131 The inclusion of tidal forcing in ocean models brings simulations closer to observations 132 (e.g., Müller et al. 2015), which is a central effort in ocean modeling. This is illustrated by 133 comparing high-resolution regional North Atlantic HYCOM simulations with tides and high-134 resolution bathymetry to satellite observations (Figure 1). HYCOM is the backbone of the 135 operational forecasting system of the U.S. Navy (Chassignet et al., 2003, 2009; Metzger et al., 136 2014). The Navy HYCOM simulations use a hybrid vertical-coordinate system: isopycnal 137 coordinates in the stratified ocean interior, and a dynamically transition to depth (z) coordinates 138 in the surface mixed layer and to terrain-following ( $\sigma$ ) coordinates (which follow the 139 bathymetry) in shallow shelf water. The simulations use realistic atmospheric forcing from the 140 Navy Global Environmental Model (NAVGEM) (Hogan et al., 2014), and can be run with or 141 without data assimilation and with or without tidal forcing. Sophisticated methods from the data-142 assimilation literature have also been applied to bring the tidal simulations closer to observations 143 (Ngodock et al., 2016). The HYCOM model was used in this study with a variety of vertical and 144 horizontal grid-spacing and conditions.

Sea-surface height (SSH), available from satellite altimetry, is a useful proxy for
mesoscale oceanic variability. The SSH wavenumber spectral slope is a measure of the relative
strength of oceanic flows as a function of length-scales within a given domain. Wavenumber is

148 defined as  $2\pi$  or one divided by wavelength, hence large wavenumbers are associated with 149 smaller spatial scales. For example, the wavelength of the first mode of the internal tide on the 150 Amazon Shelf is around  $\sim 120$  km and that of the second mode  $\sim 70$  km (Xu et al., 2022). 151 However, wavenumber spectral slopes vary greatly by location (Figure 1a, Xu and Fu, 2012; 152 Zhou et al., 2015; Dufau et al., 2016), with steep spectral slopes in western boundary current 153 regimes such as the Gulf Stream (which has strong currents and high eddy variability), flatter 154 slopes in the interior (such as the eastern North Atlantic), and much flatter slopes in the 155 equatorial region (Figure 4a). However, previously published results in high-resolution models 156 without tidal forcing could not replicate this spatial variability (e.g., Figures 1e-f; Chassignet and 157 Xu, 2017). Motivated by this apparent discrepancy, Xu et al. (2022) investigated the relative 158 impacts of the inclusion of tidal forcing, high-resolution bathymetry, and high-frequency 159 atmospheric forcing on the SSH wavenumber spectra using a series of 1/50° North and 160 equatorial Atlantic HYCOM simulations. At this resolution, the models can realistically capture 161 variability at scales of 10's to 100's of kilometers and both the barotropic and baroclinic tides. 162 The inclusion of internal tides was paramount to improving the agreement with observations 163 (Figures 1b-d). The surface signature of internal tides was greatest in the equatorial Atlantic and 164 the eastern subtropical North Atlantic where there are strong barotropic tides and a strong 165 stratification in the upper layer of the water column. The internal tides are the primary cause of 166 the observed large spatial variability of the spectral slope in the Atlantic. High-resolution 167 bathymetry and high-frequency wind variability had comparably minor impacts on the SSH 168 variability (Figure 1b-d). However, on a local scale, high-resolution bathymetry details, 169 especially the slope near the shelf break, played an important role in the generation of internal 170 tides, which could be important in locations with large internal tide energy.

Globally there is a distinct signature of tidal variability, as can be seen from the kinetic energy of tidally forced HYCOM model simulations band-passed at semidiurnal (lunar) tidal frequencies (Figure 2a) and at higher frequencies, called supertidal frequencies (Figure 2b). The distribution of internal wave energy in the ocean is uneven, with extremes near generation sites such as coastlines and steep bathymetry, such as the Luzon Strait, (Figure 2d), a strait between Taiwan and Luzon, Philippines, a region of well-known for its strong internal tidal generation (e.g., Alford et al., 2015).

178 To assess how tidally forced models may differ acoustically from those without tides in 179 locations such as these, we used a series of global HYCOM simulations: with or without tidal 180 forcing and with or without data assimilation (DA). These experiments are summarized as tidal 181 forcing and no DA (Exp. 19.0), no tidal forcing and no DA (Exp. 19.2), tidal forcing and DA 182 (Exp. 21.6), no tidal forcing and no DA (Exp. 19.1), where the numbers correspond to saved 183 experimental set-ups (e.g., https://hycom.org). Each was forced by wind and has a 1/25° (2-4 184 km) horizontal resolution and 41 layers. Hourly output was saved for each of these experiments 185 for one month from May to June 2019. For each model iteration temperature and salinity were 186 interpolated from the native grid to 101-depth layers and then used to compute sound-speed. 187 Using a 5000 km transect across the Luzon Strait (Figure 2d) as an example, the depth of a 188 single sound speed surface, 1510 m s<sup>-1</sup>, is shown in Figure 2e. At the ridge, located at 1000 km along the transect, internal tides propagate in both directions, which can be seen in the depth 189 190 change of this surface over time in the tidally-forced HYCOM simulation (Figure 2e). These are 191 absent in the simulation without tides (Figure 2f). Steep topography in the North Pacific causes 192 additional internal tide generation along this transect that is visible in the periodicity of the depth 193 of the sound-speed surface at additional ridge locations (e.g., at 4800 km). Again, these 194 undulations are absent in the simulation without tidal forcing.

195 Just as SSH spectral slope was compared to observations (i.e., Figure 1), it is also 196 important to compare the sound speed in the simulations to observations, such as by examining 197 sound speed variability at a single location. Here, the four simulations (with/without tides and 198 with/without data assimilation) were compared to Spray glider observations in the California 199 Current in the North Pacific (Figure 2c; Rudnick, 2016). The mean and standard deviation of the 200 sound speed were computed from each simulation at 3-hourly intervals over the region that the 201 glider was profiling. (The glider profiles from the surface to 500 m depth roughly every three 202 hours.) Although this is not a region of large tidal energy, the tidal forcing still increased the 203 sound speed variability. The simulation with both tidal forcing and data assimilation compared 204 the closest to the observations. However, even without data assimilation, the HYCOM 205 simulation with tidal inclusion lay closer to observations than the simulation without tidal 206 forcing (Figure 2c).

Although the data-assimilative simulations of sound speed variability were closer to
 observations, implausible jumps can occur in the sound-speed structure during an assimilation

209 window. Data-assimilation can abruptly alter the vertical temperature and salinity structure. It's 210 unclear if the increase of sound speed variability in the data assimilative models was due to 211 natural ocean variability or the assimilative 'shock' bringing the ocean closer to observations. 212 These 'shocks' bring the ocean out of geostrophic balance (i.e., the balance between horizontal 213 pressure gradients and rotation is disturbed). In the ensuing geostrophic adjustment process, in 214 which the ocean tries to restore this balance, spurious low-mode internal waves are generated 215 (Raja et al., 2024). These waves have frequencies that overlap with the tidal and inertial bands, 216 complicating the analysis of naturally occurring tidal and near-inertial waves. For this reason, 217 most of our HYCOM internal tide and IGW studies (e.g., Buijsman et al., 2020; Luecke et al., 218 2020; Raja et al., 2022; Arbic et al., 2022; Arbic 2022) have used HYCOM simulations without 219 data assimilation. The interaction of these spurious IGWs with other internal waves or eddies and their eventual dissipation alters the ocean energetics. Hence, for much of the research during this 220 221 project we used non-assimilative HYCOM simulations.

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## 223 3. Amazon Region: A case study on IGW energy transfer and acoustic impacts

224 In the open ocean, low-mode internal tides (e.g., first baroclinic mode) may decay via wave-wave interactions that cascade energy to higher frequencies and wavenumbers (Lamb 225 226 2004; Eden et al., 2020). The energy at these primary frequencies, such as the diurnal and 227 semidiurnal bands, dominate most of the internal tide spectrum, except along the path of large 228 amplitude internal tides near the equator. Most of the research on IGW-IGW interactions in the 229 open ocean has focused on subharmonic resonance, also called Parametric Subharmonic 230 Instabilities, which transfer energy from tidal to lower 'subharmonic' frequencies (e.g., Ansong 231 et al. 2018). In contrast, the decay of the low-mode internal tide due to superharmonic wave-232 wave interactions (interactions that transfer energy to higher than tidal frequencies), have not 233 been thoroughly studied in global simulations. For this project, Solano et al. (2023) evaluated the 234 energy transfers between tidal and supertidal frequencies and the interactions between low-mode 235 waves in global HYCOM simulations.

Globally, higher-harmonic (supertidal) kinetic energy (KE) accounts for a significant
portion of the total internal gravity wave energy, especially at low-latitudes. Band-pass filtering
the HYCOM KE at supertidal frequencies (Figure 2b) shows that supertidal KE energy is higher
at low-latitudes (<25°), propagating from 'hot spots' such as the Amazon Shelf. Supertidal</li>

energy accounts for about 5% of the total (tidal+supertidal) IGW energy, and internal tides
transfer energy to higher harmonics at a rate of about 45 GW, compared to the 500 GW of
energy available from barotropic tides (i.e., barotropic to baroclinic conversion). Locally, in
regions such as the Bay of Bengal, the Amazon Shelf, and the Mascarene Ridge, 25-50% of the
IGW KE is found at supertidal frequencies.

245 We focused on the two regions of high supertidal frequency: the Amazon Shelf and the 246 Mascarene Ridge (Figure 3). The nonlinear IGW energy transfer from primary to supertidal 247 frequencies revealed a banding pattern (Figure 3a-b) that agreed well with the horizontal 248 divergence of the supertidal energy flux (Figure 3c-d). Where the supertidal energy was high, a 249 regularly spaced banding pattern emerged where energy was transferred from primary to higherharmonic frequencies at a rate of about 10-50 mW m<sup>-2</sup> locally. The banding pattern in Figure 3 250 251 suggested a common mechanism for the nonlinear energy transfer between length scales. 252 Decomposing the energy into separate modes (Figure 3e-f) shows that the banding pattern 253 appeared when superimposing the lowest modes (1+2) but it was not present for individual 254 modes. This suggests that these regions of enhanced energy transfer to higher-harmonics can be 255 explained by the interactions between mode-1 and mode-2 internal tides, which interfere 256 constructively at the locations of the patches (e.g., their velocities are in phase, increasing the 257 tidal amplitude and steepening the internal tide). The location of these patches was modulated by 258 the slowly varying subtidal current and the spring-neap cycle, with higher amplitude spring tides 259 resulting in greater energy available to transfer to higher-harmonics.

260 The evidence of low to high mode IGW energy transfer make the Amazon basin and 261 Mascarene Ridge interesting regions to examine the impact of tidal forcing on acoustic 262 propagation. It was hypothesized that these active regions would strongly undulate the upper 263 ocean, impacting both sound speed and its gradients. Along the Amazon shelf we examined the 264 sound speed differences between tides and no-tides HYCOM simulations, and the impacts of 265 tides on midfrequency acoustic propagation in and above the thermocline. In Section 4, the 266 Mascarene Ridge is used to assess how model grid spacing and physics may cause differing 267 acoustic variability.

In the Amazon region, semidiurnal internal tides propagate in a northeast direction away from the coast as be the regions of energy transfer in Figure 2. Sound speed, acoustic parameters and acoustic transmission loss were examined along this line, in a 60° radial along the direction of tidal propagation, as shown by a line in Figure 3a. First, we took the mean and standard
deviation of sound speed in the upper ocean (Figure 3g, h) for both the tidal and non-tidally
forced HYCOM simulations for one week, May 20-28. Although the mean sound speed was
similar, there are notable differences in the thermocline (~150 m depth), where in the tidal case a
similar "banding" of sound speed variability is observed to that of internal wave energy transfer
(Figure 3a).

277 A virtual acoustic source was placed in the HYCOM simulations with and without tidal 278 forcing at 4.1°N and 44.8°W and 20 m depth (Figure 3a; yellow star). For the acoustic model, we 279 used a 1500 Hz source and a three-dimensional ray-tracing model, Bellhop 3D, available from 280 the Ocean Acoustics Laboratory (Porter, 2011). Running an acoustic propagation model provides 281 direct information of how sound at a given frequency and depth may propagate. We used 282 acoustic transmission loss (TL) to provide a relative estimate of acoustic pressure, a proxy of 283 how detectable a source may be at a certain range. A snapshot of the sound speed structure along 284 this 60° radial line shows undulations in the tidal case that were absent in the nontidal case 285 (Figure 4a, top panels). The internal tides oscillate the pycnocline thus changing both the vertical 286 and horizontal sound speed gradients (Figure 4a(i, ii)). At this time step, the tidal case showed 287 greater transmission loss in the surface layer, whereas strong transmission occurred in the 288 nontidal case (Figure 4a, bottom panels).

289 Acoustic parameters can be used to quantify the differences between the tidal and non-290 tidal ocean simulations. When examining the properties of potential surface duct propagation in 291 the upper ocean, as we are here, two acoustic parameters of interest are the sonic layer depth 292 (SLD) and below-layer gradient (BLG). The SLD is defined as the depth of the subsurface 293 sound-speed maximum (e.g., Helber et al., 2008). If the acoustic source is within the layer and 294 the frequency is supported by the layer, sound can potentially be trapped between the surface and 295 the SLD. The BLG, defined as the gradient in sound speed below the SLD, can influence the 296 potential of the layer to retain trapped energy. Comparing the tidal and nontidal snapshots, the 297 SLD was shallower and the BLG more variable in the HYCOM simulation forced with tides than 298 that without tidal forcing. The structure of the sound speed profiles (horizontally and vertically) 299 also guide the paths of the sound. A smooth profile will result in a greater concentration of 300 energy (less loss) and a profile with many small gradient variations will result in more refracted 301 energy (more loss) throughout the entirety of the waveguide.

302 A timeseries of TL, SLD and BLG showed the persistent differences between the 303 HYCOM simulations with and without tidal forcing. In the tidal case TL was greater, likely 304 because of a shallower SLD, particularly from 20 to 23 May. The transmission was sensitive to 305 this depth, with transmission occurring on a semidiurnal timescale (e.g., every 12 hours) with the 306 upper ocean undulation. In the nontidal case, there were no periodic fluctuations of TL, only 307 irregular fluctuations associated with submesoscale to mesoscale variability. The relationship 308 between SLD and TL thus appeared relatively straight forward, with poor transmission if the 309 source is positioned below the SLD. Less clear was the relationship of TL to BLG, which was 310 more variable in the tidal case (Figure 4b), or the relationship of TL to the horizontal sound 311 speed gradients introduced by tidal forcing.

312 The mesoscale differences between the tidal and non-tidal HYCOM simulations made a direct comparison between them problematic, and later impacted how DL parameterizations 313 314 were made (Section 5). Some of this variability was owing to how the tides interact with the 315 mesoscale field and atmospheric forcing. Correlation coefficients between wind and mixed layer 316 depths in the Amazon region were similar between the tides and no-tides simulations, but with 317 differences near the coast where currents and tidal variability were strong. This will be an avenue 318 of future research, but it made linking the relative amplitude of the tidal energy to acoustic 319 impacts difficult.

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#### 321 4. Horizontal and vertical grid spacing

322 Horizontal and vertical grid spacing, as well as the parameterizations commonly used to 323 represent unresolved processes in ocean models, significantly impact the ability of models to 324 represent internal tides and IGW processes (Nelson et al., 2020; Buijsman et al., 2020; Kelly et 325 al., 2021; Thakur et al., 2022). The grid spacing also affects temperature and salinity 326 distributions and thus acoustic propagation. During this project, horizontal grid spacings of 2-4 327 km were used for HYCOM (as seen in the previous section), which is a finer grid spacing than 328 1/12° grid spacing available in many global ocean models. There is a clear link between the 329 HYCOM grid spacing and the magnitude of the semidiurnal internal wave energy (Buijsman et 330 al., 2020). For example, decreasing the horizontal grid size from 8 to 4 km increases the internal 331 wave generation and energy density by about 50%. This is largely because of the number of

internal wave modes resolved, increasing from the first two modes in the 8-km simulation to thefirst 4 modes in the 4-km simulation.

334 The horizontal and vertical grid spacings determine how well vertical features of 335 bathymetry and, respectively, horizontal and vertical wavelengths of the modes are resolved. 336 Both the modal wavelengths and grid spacings are spatially variable in global ocean models. For 337 example, the horizontal grid spacing generally decreases poleward due to the sphericity of the 338 earth, while the wavelengths of tidally generated internal waves increase poleward due to the 339 increase of the Coriolis frequency. We applied existing and newly developed criteria to 340 determine what diurnal, semidiurnal, and supertidal vertical wave modes could be resolved 341 depending on both the horizontal and vertical resolution of the 4-km global HYCOM simulation 342 with 41 layers (Figure 5; Exp. 19.0). For the horizontal resolution we used the criterion that a 343 vertical mode is resolved if there were at least 6-8 horizontal grid spacings per wavelength 344 (Steward et al., 2017). A similar criterion was applied for the vertical resolution (vertical 345 criterion CZA). However, this criterion is designed for z-coordinate models, whereas HYCOM is 346 an isopycnal model below the mixed layer. Therefore, we applied an additional criterion called 347 CZB, which accounted for both the vertical and horizontal velocity structure. The goal was to see 348 how modes could be resolved globally using these criteria.

349 The zonal mean of the number of the resolved modes for seafloor depths >2000 m and 350 averaged over 10° latitude bins is presented in Figure 5a. Internal wave modes with lower 351 frequencies have longer wavelengths, and hence, they are better resolved by the horizontal 352 resolution. For example, for the lunisolar diurnal internal tide,  $K_1$ , around eight modes are 353 resolved at the equator and the number increases towards 20 modes near the K1 turning latitude 354 of about 30° due to the increase in wavelength. Poleward of this latitude, the tidal frequency is 355 smaller than the Coriolis frequency, and diurnal internal waves cannot exist. The shorter 356 wavelength principal lunar semidiurnal internal tide,  $M_2$ , had fewer modes resolved, with a 357 minimum number of about four modes resolved at the equator. Supertidal internal tidal waves 358 occur at higher harmonics of the primary tidal frequencies, as discussed in the previous section. 359 Globally, the supertidal waves with the quarter diurnal M<sub>4</sub> frequency have the most energy (see 360 Figure 1 of Solano et al., 2023). On average, only up to two M<sub>4</sub> modes were resolved. The 361 number of resolved modes was more sensitive to the vertical resolution criteria than to the 362 horizontal resolution. If CZA was applied to the hybrid vertical coordinates, mode-1 is barely

363 resolved at the low latitudes and not at all at higher latitudes (Figure 5a). In contrast, CZB
364 appeared to be a more appropriate criteria, and allowed for the resolution of more modes with a
365 maximum of 12 diurnal modes resolved at the equator. Having criteria for how many modes may
366 be resolved in a global model with a certain discretization will help interpret the meaning both of
367 internal tidal energy transfer and the sound speed variability.

368 Grid spacing does have a direct impact on sound speed variability, as smaller grid 369 spacing may resolve more processes and stronger temperature and salinity gradients. As an 370 example, we compared SLD and BLG at the Mascarene Ridge between a two-dimensional 371 nonhydrostatic simulation of the MITgcm, with a horizontal grid spacing of 100 m, to the 372 simulations of the tidally forced global HYCOM model (Exp. 19.0), which is hydrostatic and has 373 a horizontal grid spacing forty-times greater than that of the MITgcm simulation (Figure 5b, c). 374 The Mascarene Ridge is a location of nonlinear wave interactions, where solitons are generated 375 and propagate away from the Ridge (Figure 3b, d, f). Although a direct comparison of the sound-376 speed structure was not possible, as the HYCOM simulations were initialized with an offset in 377 temperature, a relative comparison of these acoustic parameters was insightful (Figure 5b, c). 378 The HYCOM simulation had semidiurnal fluctuations of the SLD and BLG, with each 379 oscillating twice a day (Figure 5b). The MITgcm simulation had a periodic signal as well, but it was disorganized, with propagation pathways appearing to overlap (Figure 5c). The closer grid 380 381 spacing of the MITgcm simulation likely allowed for nonlinear interactions to occur between 382 different internal waves which in turn impacted the sound speed structure. This structure is likely 383 more realistic when compared to observed ocean variability, but it would also make the sound 384 speed less predictable.

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386 With the increase in computational power, more submesoscale-resolving regional and 387 basin-scale models have been developed in the past years, such as the 1-km grid-spacing 388 HYCOM simulation for the Gulf of Mexico (Chassignet and Xu, 2017; Hiron et al., 2021, 2022; 389 Chassignet et al., 2023). Ongoing discussions within the oceanography community resolve 390 around performing global HYCOM simulations with tides with a finer grid-spacing on the order 391 of 1/50° (the most up-to-date global HYCOM has 1/25° grid-spacing). However, the optimal 392 number of vertical layers needed in submesoscale resolving models to resolve internal tides and 393 tidal energetics, and the consequent effect on acoustic propagation is still an open question. To

explore this question, we used an idealized HYCOM configuration with 1/100° horizontal gridspacing (1 km), forced only by the semidiurnal (M<sub>2</sub>) tides over a centrally spaced ridge, and
varied the number of layers between the simulations from 8- to 128- layers (Figure 6). The
idealized configuration allowed the problem to be isolated from contamination by ocean eddies
and currents while resolving all the physics allowed in HYCOM.

399 Each simulation was initialized with a climatological temperature profile averaged over 400 the Cape Verde area and a constant salinity structure. The domain size, approximately 8000 km 401 in the zonal direction, was large enough to prevent the reflection of internal tides at the 402 boundaries. The vertical grid-discretization of these layers was chosen based on characteristic 403 wavelengths of different IGW modes. To generate internal tides, a steep ridge with a Gaussian 404 shape was added in the center of the domain. The criticality of the slope, which is a measure of 405 the ridge steepness normalized by the ray slope of the internal waves, was larger than one, 406 allowing for the generation of nonlinear waves and wave beams (Garrett and Kunze, 2007; 407 Buijsman et al., 2010). Figure 6a is a snapshot of the vertical velocity for the 128-layer 408 simulation, highlighting the region near the ridge where the wave beams were the strongest. Figure 6b shows the depth-integrated, vertical KE ( $\frac{1}{2}\int w^2 dz$ ), where w is the vertical velocity, 409 for different vertical discretizations (8-, 16-, 32-, 48-, 64-, 96-, and 128-layer) and averaged over 410 411 one week. The vertical KE of the 8- and 16-layer simulations differed from the others in 412 amplitude and phase. As the number of layers increased, the simulations became more similar, 413 both in amplitude and phase. From the 48- to the 128-layer simulations, magnitude and phase are 414 similar across simulations. Strong vertical velocities led to greater vertical displacement of the 415 isopycnals which can then affect sound speed and, potentially, acoustic propagation. When 416 integrated over the whole domain (0 to 2000 km from the ridge), the tidal barotropic-to-417 baroclinic energy conversion, the vertical kinetic energy, and the turbulent dissipation were 418 greatest in the 128-layer simulation and decreased with coarser vertical grid spacing. The results 419 started to converge for the simulations with more than 48 layers, however, these results may 420 differ with a change in horizontal grid spacing.

The idealized model can be used to isolate the possible impact of vertical grid spacing on sound speed, without the confounding challenges of differences in mesoscale eddy fields or initialization states such as in the examples of the Amazon basin and Mascarene ridge. We interpolated the 8-, 16-, 32-, 48-, and 96-isopycnal layer simulations each to constant-depth 425 coordinates, then took the mean and standard deviation of sound speed over 72-hourly timestep 426 (Figure 6c, d). The lower vertical grid simulations (less than 32-layers) had greater mean sound 427 speeds and deeper SLDs (Figure 6c). These lower-layer simulations had some difficulty 428 resolving the locations of large sound speed variability (Figure 6d), which would likely impact 429 acoustic propagation. As the number of layers increased, the sound-speed profiles and their 430 variability converged with very little difference between the 48- and 96-layer simulations. 431 Hence, the number of layers affected the simulation of internal tides, with a minimum of 48 432 isopycnal layers (for a 1 km horizontal grid-spacing) necessary to ensure an accurate 433 representation of internal tide-induced isopycnal displacement. A similar impact was observed 434 for the mean sound speed and its variability, with 48-layers or greater needed to converge on 435 similar results.

436

## 437 5. A final word on energetics

438 The IGW spectrum covers both the transfer of energy between internal wave modes and 439 the transfer of oceanic kinetic energy from its injection at large scales in eddies, near-inertial-440 waves, and tides to the smallest scales. As such, properties of the IGW spectrum depend on the 441 ocean state, including the slowly varying background circulation and surface forcing that may 442 affect vertical stratification. An empirically determined form of the IGW spectrum, determined 443 by Garrett and Munk (1979), is applicable globally but uses free parameters to account for 444 regional and seasonal variations. Ongoing research focuses on what determines these parameters 445 and any deviation from this spectral form; nonlinear interactions involving IGWs, such as those 446 on display in the Amazon basin and near Mascarene Ridge, are thought to be of particular 447 importance.

448 Previous theoretical and idealized work on IGW-IGW interactions has identified some 449 important processes that move energy to smaller scales (McComas and Bretherton, 1977; 450 Dematteis et al., 2022). These studies did not analyze eddy-IGW interactions in the same manner 451 as IGW-IGW interactions and considered the latter to be the dominant processes. One IGW-IGW 452 mechanism involving the interaction of near inertial and tidal IGWs, called "induced diffusion", 453 is thought to be very important in transferring KE across length scales. Skitka et al. (2024) used a 454 unique framework to diagnose these IGW-IGW interactions along with analogous eddy-IGW 455 interactions from a regional MITgcm ocean simulation in the North Pacific (Nelson et al., 2020;

456 Pan et al., 2020, Thakur et al., 2022). They found that IGW-eddy interactions induce a 457 downscale KE flux in a manner analogous to IGW-IGW interactions. This "eddy-induced 458 diffusion" is the dominant mechanism of energy exchange within the IGW supertidal continuum 459 at 2-km horizontal grid spacing, which is typical grid spacing for the highest-resolution 460 global models that can presently be run, and is achievable in the HYCOM simulations used here. 461 This "eddy-induced diffusion" is comparable to the wave-induced diffusion at the higher 250-m 462 horizontal grid spacing that is achievable by regional models. Thus, the finer grid spacing in the 463 vertical and horizontal directions will make a large difference in the details of the IGW cascade, 464 including the rate of energy transfer, the dominant mechanisms that contribute to it, and the 465 mechanisms and patterns of dissipation. All of these have implications for the ability of global 466 models to resolve oceanic processes and thus sound speed and its variability.

467

#### 468 6. Developing awareness and creating solutions using Deep Learning (DL)

469 It is clear from the work thus far that the inclusion of tidal forcing in ocean simulations 470 increases the realism of the ocean state, as do finer grid spacing in both the vertical and 471 horizontal. However, tidally forced, global ocean models are computationally expensive, 472 especially those with fine grid spacing. DL was used to investigate the statistical differences 473 between HYCOM simulations with and without tidal forcing. The goal was to generate a tidally 474 forced ocean state from a non-tidally forced HYCOM simulation without numerically solving the 475 physical forcing equations, both to reduce the computational cost and to more rapidly predict 476 acoustic impacts.

477 A Generative Adversarial Network (GAN; Goodfellow et al., 2014; Creswell et al., 2018) 478 was proposed as a potential solution, using the simulations of global HYCOM with and without 479 tides (Exp. 19.0) as initialization states. General Adversarial Networks (GANs) are a DL 480 technique that tries to learn a transformation from one statistical distribution to another, instead 481 of trying to directly learn a specific distribution. The DL model is trained alongside an 482 adversarial discriminator, which is a classifier used to differentiate between actual data and 483 generated data (Goodfellow et al. 2014, Goodfellow et al. 2016). Given this formulation, GANs 484 are well suited to translate nontidal results to tidal results for vice versa. 485 To address this issue of the chaotic, turbulent nature of the ocean we considered the

486 HYCOM simulations with and without tidal forcing to be unpaired. This meant that we did not

487 compare a specific point in space and time between the two simulations. As we considered this 488 data unpaired, we trained the networks in a way that compares the simulations using cycle-489 consistency loss. For this method, we trained two separate generators and discriminators. One 490 generator,  $G_{NT \to T}(\cdot)$ , translates from the non-tidal to the tidal domain, and the other generator, 491  $G_{T \to NT}(\cdot)$ , translates from the tidal to non-tidal domain. For training, data was translated to the 492 opposite domain, and then translated back to the original domain. The cycle-consistency loss is 493 defined as the mean-squared difference between the original data sample and the doubly 494 translated data (Zhu et al., 2017). The cycle-consistency loss was then used in combination with 495 the traditional GAN losses to train the networks.

The Atlantic Ocean was used as a test-case region and split into 90% training data and 10% validation data. Figure 7 shows some examples of the results, with the original HYCOM model simulations with tides (Exp 19.0) and without tides (Exp. 19.2) and GAN outputs for selected samples in the validation set. The example subfigures are time series at each given location (Figure 7a), representing samples of how the GAN network performs in different oceanographic conditions.

502 The general structure of the temperature and salinity profiles was retained in the GAN translation. For example, south of New England in the Gulf Stream (Figure 7b), the tidally forced 503 504 observations had a fresher, cooler region extending to 150 m depth that dissipated after 505 approximately 90 hours. This was likely a mesoscale intrusion and was absent in the nontidal 506 results. Thus, it was poorly represented when comparing the original tidally forced HYCOM to 507 the GAN generated tidal fields. The GAN model captured the periodic structure consistent with 508 tidally forced simulations which varied depending on the sample location. The DL model 509 handled these differences consistently. For instance, in the mid-Atlantic (Figure 7c), there were 510 periodic signatures in temperature, salinity and velocity. Here, the periodicity of the outputs of 511  $G_{NT \rightarrow T}(\cdot)$  matched nicely with those from HYCOM with tides. It was more difficult to separate 512 the mesoscale variability from the tidal structure in the Gulf Stream (Figure 7d). The DL model 513 artificially increased the periodic structure (most visible in water velocity). This location, further 514 from the shore than that of 7b, likely was dominated by a strong mesoscale eddy field. This was 515 an instance where the DL model imposed a periodicity to make the sample like other tidally 516 forced results, even though the actual HYCOM sample did not contain this structure.

517 As the data we utilized was unpaired, it is difficult to evaluate the performance of the 518 model without visually inspecting the outputs. In future work, a discriminator could be trained to 519 differentiate between these results. Because the model output used to train the DL models were 520 sampled from a region of the globe during the same time of year, no two samples were 521 completely independent. This introduces the risk of overfitting. Using unpaired data made the 522 model more robust to overfitting but did not remove the risk entirely. Additionally, the sound 523 speed structure had a persistent offset of about 5 m/s greater in the GAN generated results than 524 the original HYCOM simulations (not shown). Thus, although this work provides a good starting 525 point, further work is necessary to improve the DL approach.

526

# 527 7. Conclusions

528 The TFO-HYCOM project was a cross-disciplinary investigation into the modeling of 529 internal tides and high-frequency IGWs, their sensitivity to model horizontal and vertical grid 530 spacing, the energy transfer and dissipation of IGWs, the impacts on sound speed structure and 531 acoustic propagation, and the ability to replicate tidal simulations using DL techniques. During 532 this project we examined the sensitivity of modeled IGWs to bathymetry and damping schemes 533 and compared them to observations.

534 The inclusion of tidal forcing in ocean models improves the representation of the ocean 535 state. Tidal forcing has a direct impact on acoustic propagation via changes that it induces to the 536 sound speed at scales from kilometers to hundreds of kilometers, and time scales on the order of 537 a few to several hours. HYCOM simulations run with and without tides showed clear differences 538 in mean sound-speed structure and sound-speed variance, that was evident both in mesoscale 539 features and at tidal frequencies. Simulations that included tidal forcing tended to have greater 540 sound-speed variance and were more consistent with observations. These impacts were sensitive 541 to vertical and horizontal discretization, as were the ability of the simulations to resolve IGW 542 interactions and energy transfer. In addition to the tilt of density surfaces caused by internal 543 waves, spice, temperature and salinity fluctuations along a constant density surface, can have a 544 similarly large impact on sound speed and its gradients (Dzieciuch et al., 2004; Colosi et al., 545 2012). Spiciness, is caused by ocean stirring by mesoscale eddies. Although we do not separate 546 out the effects of spice from isopycnal tilt here, it's likely spice would also differ between the 547 models as IGWs interact with the mesoscale field. Further investigations into acoustic impacts

could also be made by looking at different frequency ranges or acoustic arrival times, for
example performing long-range, low-frequency acoustic propagation and comparing model
results with observational studies (e.g., Dushaw et al. 2011).

As global operational models begin to include tidal forcing, along with a decrease in grid spacing, it is important to understand how well they represent physical processes and how energy cascades through the internal wave spectrum. As running these models at high resolution is computationally expensive, machine learning techniques may facilitate predictions of IGW impacts on ocean state using a model with coarser grid spacing or simulations without tidal forcing.

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558

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567

#### 568 Author information:

569 Martha C. Schönau was a Senior Scientist at Applied Ocean Sciences, LLC, and is now an

570 Assistant Project Scientist at the Scripps Institution of Oceanography at the University of

571 California San Diego.

- 573 Luna Hiron is a Postdoc at the Center for Ocean-Atmospheric Prediction Studies at Florida State574 University.
- 575 John Ragland was a summer intern at Applied Research in Acoustics, LLC (ARiA), and is a PhD
- 576 student in the Department of Electrical and Computer Engineering at the University of
- 577 Washington.

- 578 Keshav J. Raja is an Assistant Research Scientist at the Center for Ocean-Atmospheric
- 579 Prediction Studies at Florida State University.
- 580 Joseph Skitka was a Postdoc at the Department of Earth and Environmental Sciences at the
- 581 University of Michigan and is now a Postdoc in the Physical Oceanography department at
- 582 Woods Hole Oceanographic Institution.
- 583 Miguel S. Solano was a Postdoc at the School of Ocean Science and Engineering at the
- 584 University of Southern Mississippi and is now a Research Scientist at SOFAR Ocean585 Technologies.
- 586 Xiaobiao Xu is a Senior Research Scientist at the Center for Ocean-Atmospheric Prediction
- 587 Studies at Florida State University.
- 588 Brian K. Arbic is a Professor in the Department of Earth and Environmental Sciences at the589 University of Michigan.
- 590 Maarten C. Buijsman is an Associate Professor at the School of Ocean Science and Engineering591 at the University of Southern Mississippi.
- 592 Eric P. Chassignet is a Professor and Director of the Center for Ocean-Atmospheric Prediction593 Studies at Florida State University
- 594 Emanuel Coelho is Chief Scientist at Applied Ocean Sciences, LLC.
- 595 Robert W. Helber is an Oceanographer for the Naval Research Laboratory at the Stennis Space596 Center in Mississippi.
- 597

598 William Peria is a Senior Research and Development Scientist at Applied Research in Acoustics,599 LLC (ARiA).

- GO0 Jay F. Shriver is an Oceanographer for the Naval Research Laboratory at the Stennis Space
- 601 Center in Mississippi.
- Jason Summers is the Chief Scientist at Applied Research in Acoustics, LLC (ARiA).
- 603 Kathryn L. Verlinden is a Senior Scientist at Applied Ocean Sciences, LLC (AOS).

Alan J. Wallcraft is a Research Scientist at the Center for Ocean-Atmospheric Prediction Studiesat Florida State University.

606

# 607 Author Contributions

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609 This manuscript highlights the research efforts by postdocs and early career researchers on the 610 TFO-HYCOM project. The TFO-HYCOM team that focused on improving IGW modeling was 611 composed of researchers from the Naval Research Laboratory (NRL), Florida State University 612 (FSU), The University of Southern Mississippi (USM), and University of Michigan (U-M), each 613 with a complementary research focus for the project. The NRL team provided global HYCOM 614 simulations at 1/25° horizontal grid spacing. FSU researchers performed higher horizontal grid 615 spacing (1/50°) North Atlantic basin simulations and idealized simulations with finer horizontal 616 and vertical grid spacing (1 km and up to 128 vertical layers) to study the impact of vertical, 617 horizontal, and bathymetry grid spacing, tidal forcing, and data assimilation on IGWs in 618 HYCOM. USM researchers examined kinetic energy content and transfer between different IGW 619 modes and provided MITgcm simulations along the Mascarene Ridge, while U-M researchers 620 further examined the theory of IGW nonlinear energy transfer and dissipation in very-high-621 resolution regional simulations of the MITgcm. Researchers from NRL and Applied Ocean 622 Sciences (AOS) used the global HYCOM model simulations with and without tidal forcing to 623 assess the impact of tidal forcing on acoustic prediction parameters and acoustic transmission 624 loss. AOS also looked at acoustic propagation in the idealized models and the MITgcm. Finally, 625 researchers from Applied Research in Acoustics (ARiA) used deep learning (DL) algorithms to 626 derive a tidal ocean state from HYCOM model results without tidal forcing to reduce 627 computational costs.

628

629 Martha C. Schönau was the lead writer of this manuscript, project lead at her institution, and

630 performed the research contributing to Figures 3g,h, 4, 5b,c, and 7a. Figures and research were

also contributed by Maarten Buijsman (Figure 5a), Luna Hiron (Figure 6), John Ragland and

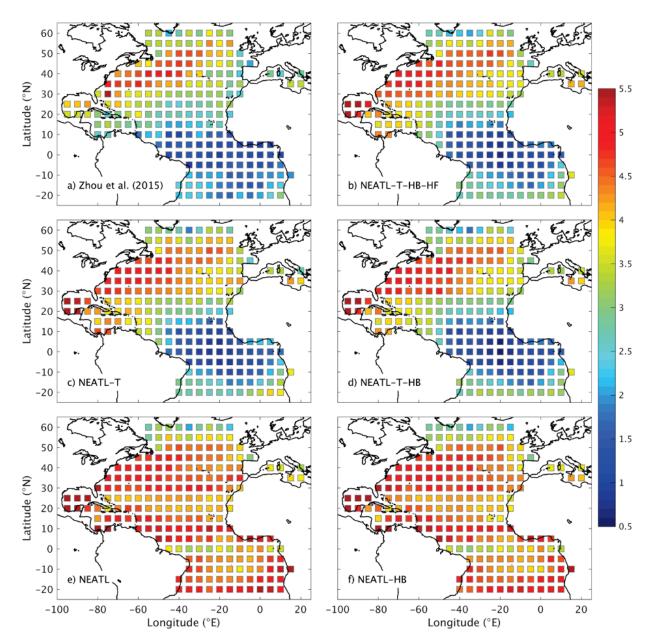
- 632 William Peria (Figure 7b-d), Keshav J. Raja (part of Section 2), Joseph Skitka (Section 5),
- 633 Miguel S. Solano (Figures 2a, b and 3), Xiaobiao Xu (Figure 1), and Jay F. Shriver and Robert

- 634 W. Helber (Figure 2c-f). Brian K. Arbic, Maarten C. Buijsman, Eric P. Chassignet, Emanuel
- 635 Coelho, Robert W. Helber, Jay F. Shriver, and Jason E. Summers served as co-PIs of this TFO-
- 636 HYCOM project at their respective institutions and oversaw the research efforts. Arbic
- 637 conceived the idea of a project on internal wave impacts on acoustics, co-led the proposal
- 638 writing, and organized regular online group meetings. Summers co-led the proposal writing for
- 639 the project and served as lead principal investigator. Kathryn L. Verlinden provided ocean-
- 640 atmospheric analysis in the Amazon region. Alan J. Wallcraft helped guide and run the Florida
- 641 State University basin-wide Atlantic HYCOM simulations.

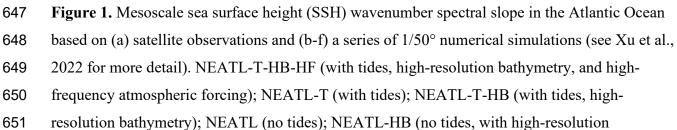


# 644 Figures:

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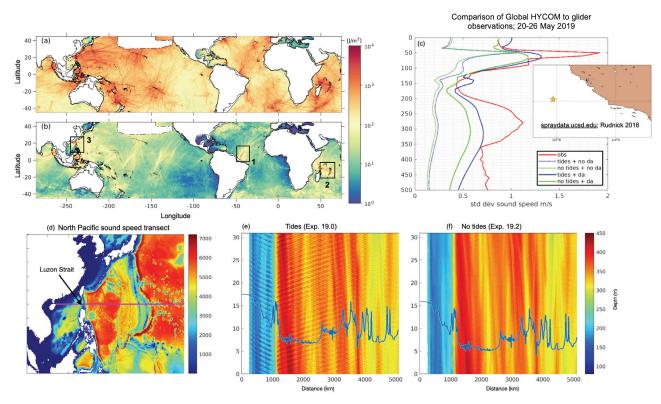






652 bathymetry). The result highlights that the large-scale spatial variability of the spectral slope is

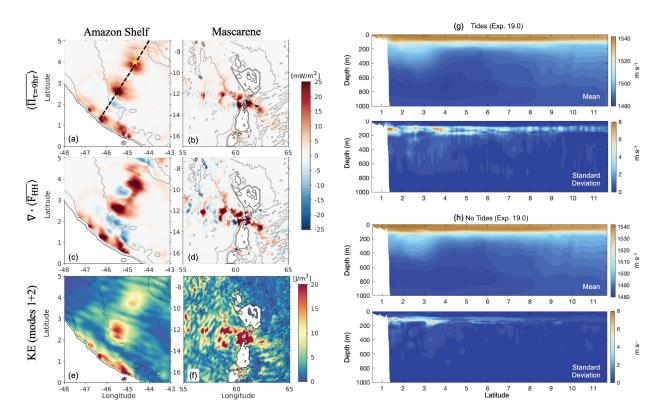
- 653 primarily due to the internal tides, whereas the impact of high-resolution bathymetry (difference
- between c and d) and high-frequency atmospheric forcing (difference between b and d) is
- relatively small. Taken from Xu et al. 2022, Figure 7. Printed with permission.
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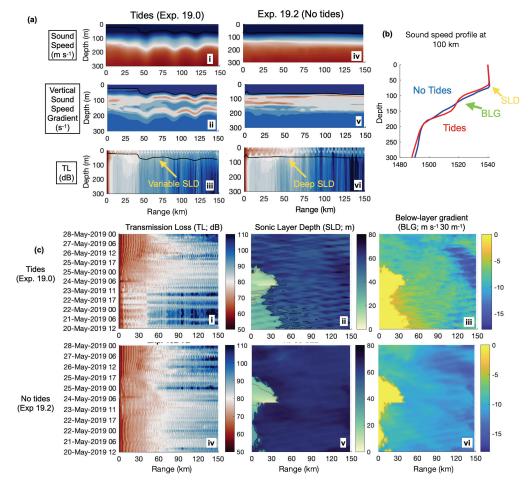
659 Figure 2: Time-mean and depth-integrated internal wave kinetic energy (J m<sup>-2</sup>) band-passed at 660 (a) semidiurnal and (b) supertidal frequency bands. Regions with relatively high supertidal 661 energy used as examples are indicated by the black rectangles: (1) the Amazon Shelf, (2) the 662 Mascarene Ridge and (3) the Luzon Strait. (c) Standard deviation of sound speed for 20-26 May 663 2019 from Global HYCOM simulations with and without tides and with and without data 664 assimilation at location indicated on map of the coast of California. Simulations are compared to 665 standard deviation computed from glider observations over the same week and location. (d) Map 666 of transect (magenta line) from the South China sea to the North Pacific, bisecting the Luzon 667 Strait. The depth of the 1510 m s<sup>-1</sup> sound speed surface is shown along the transect shown in (d) for the HYCOM simulations (e) with tidal forcing (Exp. 19.0) and (f) without tidal forcing (Exp. 668

- 669 19.0). Bathymetry is overlaid on each (e) and (f), with the Luzon Strait is located at 1000 km
- 670 distance.



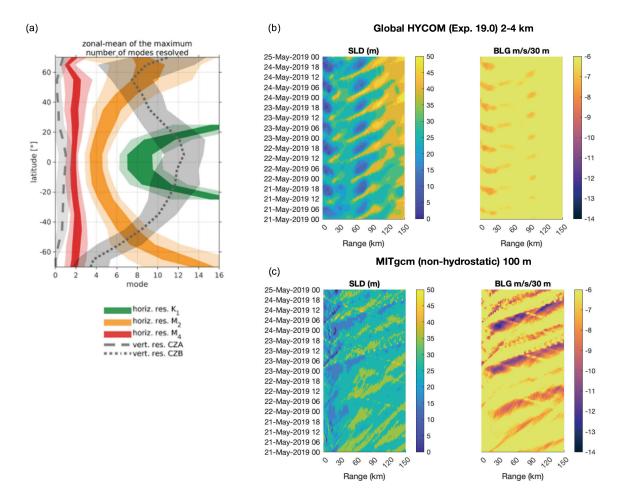


**Figure 3.** Time-mean and depth-integrated coarse-grained kinetic energy transfer ( $\langle \Pi_{(\tau=9hr)} \rangle$ ) at the (a) Amazon Shelf and (b) Mascarene Ridge. Time-mean, depth-integrated divergence of supertidal energy flux ( $\nabla$  ( $F_{HH}$ )) at the (c) Amazon Shelf and (d) Mascarene Ridge. Time-mean surface kinetic energy for the superposition of modes 1+2 at the surface at the (e) Amazon Shelf and (f) Mascarene Ridge. Panels (a)-(f) are modified from Solano et al. (2023). Sound speed mean and standard deviation for each the (g) tidal and (h) non-tidally forced HYCOM simulations from May 20-29, 2019, in the Amazon region, plotted by latitude along the dotted line shown in (a). In (a), the location of the acoustic source (star) and radial (dashed black line) used in Figure 4 are also noted. In (b), a short, dashed line indicates the transect used in Figures 5b and 5c. 



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688 Figure 4: Comparison of acoustic propagation and properties between non-assimilative global 689 HYCOM simulations with tidal forcing (Exp.19.0) and without tidal forcing (Exp.19.2). The 690 location of the source and radial are indicated in Figure 2a, with acoustic propagation shown at 60° counterclockwise from east. (a) A snapshot from 20 May 2019 18:00:00 of (i, iv) sound 691 speed (m s<sup>-1</sup>) (ii, v) vertical gradient of sound speed (s<sup>-1</sup>) and (iii,vi) TL (dB) for (i-iii) tidally 692 forced and (iv-vi) non-tidally forced simulations. (b) Comparison of a single velocity profile at 693 100 km distance from the source along the 60° radial. (c) Acoustic transmission loss (TL) at 20 694 695 m depth, sonic layer depth (SLD) and below-layer gradient (BLG) along a radial 60° counterclockwise from east for a 1500 Hz source at 20 m depth located in at 4.1°N and 44.8°W 696 697 for each (i-iii) a tidally (Exp. 19.0) and (iv-vi) a non-tidally (Exp. 19.2) forced HYCOM 698 simulation. 699

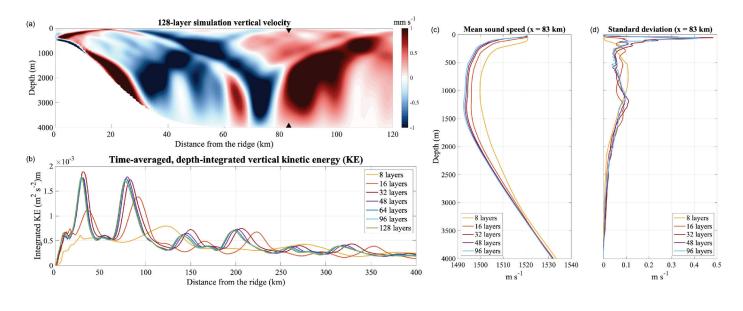


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Figure 5. (a) The predicted number of modes resolved for the horizontal resolution for various
 internal tide frequencies (colored polygons) and the vertical resolution (gray shaded polygons)
 area-averaged over longitude and 10° latitude bins. K<sub>1</sub>, M<sub>2</sub>, M<sub>4</sub> represent the dominant diurnal,
 semidiurnal, and supertidal constituents of internal tides with decreasing wavelengths,

respectively. For the horizontal resolution, the dark-colored polygons mark the range of the number of resolved modes bounded by 6 and 8 cells/wavelength and the light-colored polygons mark the extent of the zonal-mean  $\pm 1$  standard deviation. For the vertical resolution, the dashed lines mark the zonal-means and the gray-shaded polygons mark the extent of the zonal-mean  $\pm 1$ standard deviation. (b) SLD and BLG for global HYCOM simulation with tides (Exp 19.0) and (c) SLD and BLG for a nonhydrostatic regional MITgcm simulation near the Mascarene Ridge (see Figure 3b).

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**Figure 6.** (a) Snapshot of the vertical velocity for the 128-layer simulation, zooming in closer to

the ridge centered at 40°W (the domain extended from 60°W to 20°W). The black triangles

720 indicate the location where the sound speed profiles, shown in (c). (b) Time-averaged, depth-

integrated vertical kinetic energy  $(\frac{1}{2} \int w^2 dz)$ , where w is the vertical velocity, for different

vertical discretizations: 8-, 16-, 32-, 48-, 64-, 96-, and 128-layers. (c) Mean and standard

deviation of sound speed 83 km from the ridge for 8-, 16-, 32-, 48-, and 96-layers.

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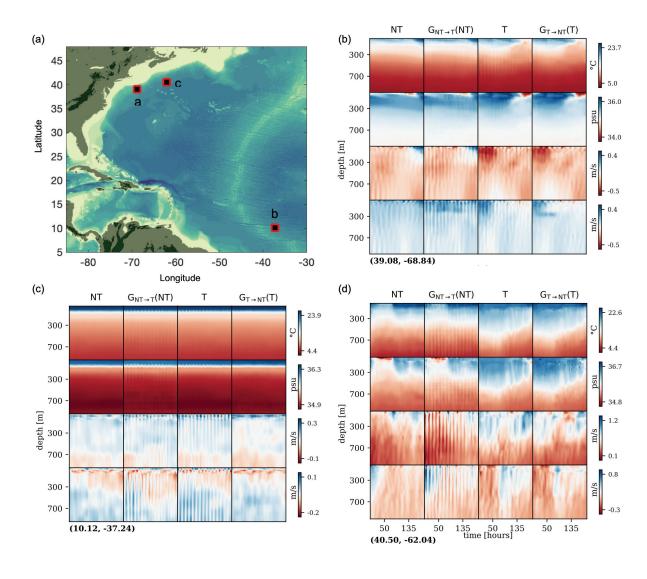




Figure 7. Temporal outputs of the trained DL model at the location noted in the bottom left of
each subfigure (b)-(c), mapped in (a) For each panel: the first column is the non-tidal (NT)
HYCOM results (Exp 19.2); the second column is the NT results translated into the tidal domain
using the DL model; the third column is the tidal (T) HYCOM results (Exp 19.0); the fourth
column is the T results translated into the NT domain using the DL model. Each row corresponds
to the oceanographic variables from top to bottom: water temperature, salinity, eastward
velocity, and northward velocity respectively.

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