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## Climate change projection for the western tropical Pacific Ocean using a high-resolution ocean model: implications for tuna fisheries

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#### Abstract

The Western Pacific Warm Pool is a region of high tuna catch, and how future climate change might impact the tuna fisheries is an important regional issue. By using a high-resolution ocean model forced by the simulated climate of the 2060s, we investigate whether enhanced spatial resolution and bias correction of the mean state could alter the climate change projection for the western tropical Pacific and examine the consequences this might have for tropical tuna distributions.

For most of the physical environmental variables, enhanced resolution and bias correction had only a minor impact on the projected changes. The climate projections showed a maximum surface warming east of the Warm Pool, a shoaling of the thermocline in the Warm Pool, and an eastward expansion of the Warm Pool. In the Warm Pool, the shoaling of the thermocline raises the nutricline into the photic zone and increases phytoplankton and primary productivity, a feature that is most evident in the high-resolution model projection but also weakly present in the coarse-resolution projection.

The phytoplankton and primary productivity response to climate change was where ocean model resolution produced a clear difference. With enhanced resolution, the simulation had stronger and better-defined zonal currents, which were more consistent with observations. Along the equator, the high-resolution model enabled vertical current shear mixing to generate

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a sub-surface phytoplankton maximum both inside and outside the Warm Pool, which is an observed phenomenon. With climate change, the enhancedresolution model projected enhanced vertical shear mixing, increased vertical supply of nutrients to the photic zone, and increased sub-surface phytoplankton concentrations. The increase in sub-surface phytoplankton concentrations helps to offset the decline in surface phytoplankton concentrations and results in a projection of almost no change in the western tropical Pacific primary productivity. In contrast, the low-resolution model projected a substantial reduction in phytoplankton concentrations and primary productivity; such a response is typical of climate change projections for the region. Importantly, enhanced resolution dramatically altered the projected response of phytoplankton and primary productivity to climate change. Using the enhanced-resolution model, the projected increase in the Warm Pool with little change in primary productivity and in suitable habitat for skipjack tuna suggest that by the 2060s climate change will not have a large impact on skipjack tuna fisheries.

*Keywords:* climate change, western equatorial Pacific, primary productivity, tuna

#### 1 1. Introduction

The upper waters of the equatorial Pacific Ocean are divided into two 2 regions, which have distinct physical, biogeochemical and ecosystem charac-3 teristics. In the central and eastern Pacific, there is an equatorial upwelling 4 system with relatively cold, salty, macronutrient-rich water, where primary 5 production is iron-limited (Christian et al., 2002). In the western tropical 6 Pacific, the water is warm, fresh and oligotrophic, and encompasses a prominent oceanographic region called the Western Pacific Warm Pool (Le Borgne 8 et al., 2002). The Warm Pool has some of the warmest surface water in 9 the ocean (McClain et al., 1999), and this warm water is fundamental to 10 the large-scale deep atmospheric convection in the western Pacific region, 11 the circulation and stratification of the upper ocean, and El Niño Southern 12 Oscillation (ENSO) variability (Maes et al., 2010). 13

The zonal movement of the eastern edge of the Warm Pool appears to be important for the onset of the ENSO phases (Picaut et al., 1996), with the eastern edge moving westward during La Niñas and eastward during El Niños (Maes, 2008; Bosc et al., 2009; Maes et al., 2010). The location

of the Warm Pool's eastern edge also seems to modulate the distribution 18 of tuna in the equatorial Pacific (Lehodey et al., 2011). For example, the 19 skipjack tuna catch appears to move with the large zonal displacement in 20 the Warm Pool that occurs during ENSO events (Lehodey et al., 2011). 21 Tuna fisheries contribute significantly to the livelihoods and economies of 22 many Pacific Island Countries and Territories (Bell et al., 2013), so the way 23 in which future climate change might impact tuna populations is a critical 24 issue for this region. 25

Under the influence of climate change, the mean climate of the western 26 tropical Pacific will probably undergo significant changes, with potentially 27 important consequences for ENSO variability (Collins et al., 2010) and for 28 tuna distributions (Lehodey et al., 2011). Coupled global circulation models 29 (CGCMs) have common spatial biases in the western tropical Pacific, such 30 as a Warm Pool eastern edge that is too far west (Brown et al., 2013a), 31 which can potentially affect their future climate projections for the tropical 32 Pacific (Brown et al., 2013b). To investigate the impact of climate change on 33 the western tropical Pacific, we use simulations from a high-resolution ocean 34 model (HOM) that gives a good representation of the present-day western 35 tropical Pacific ocean state to make a climate projection for the 2060s (Cham-36 berlain et al., 2012). The simulations are configured to determine the change 37 in the mean ocean state. They also include the lower levels of the food 38 web (i.e. phytoplankton and zooplankton). A previous study used the same 30 simulations to predict future climate change in the Western Boundary Cur-40 rent region of the Southwest Pacific (Matear et al., 2013); the study showed 41 that by resolving mesoscale features (e.g. the East Australian Current and 42 its eddies), the oligotrophic water of the Tasman Sea is projected to have 43 increased primary productivity, because of increased eddy activity. By com-44 paring our climate projections with previously generated CGCM projections 45 (e.g. Ganachaud et al., 2013), we investigate whether climate projections of 46 the ocean state will be modified by a less-biased ocean state with enhanced 47 model resolution. For this study, we focus on the western tropical Pacific be-48 cause of its importance for tuna. In particular, we are interested in whether 49 enhanced resolution can significantly alter the projection of primary produc-50 tivity and suitable thermal habitat for skipjack tuna. 51

The paper is structured as follows. First, we briefly discuss the key oceanic features of the western tropical Pacific in § 2. Then, in § 3 we summarize how the future climate change projections are performed with our HOM. In § 4 we present results of the HOM simulation of the present-day

ocean state and compare them with observational data and with the low-56 resolution model that we used to produce the climate change projection. 57 Next, we describe in  $\S5$  the climate change projection for the 2060s and 58 compare our simulated projections from the high- and low-resolution mod-59 els. This section also includes a comparison of the projected changes with 60 previous results, discussion of the implications of our projected changes for 61 tuna distributions in the western tropical Pacific, and remarks on the robust-62 ness of the projections. Finally, in  $\S 6$ , we present a short summary of the 63 limitations of our modelling approach and discuss the direction of our future 64 work. 65

#### 66 2. Oceanography of the Western Pacific Warm Pool

The Western Pacific Warm Pool has warm surface water, with a shallow 67 mixed layer (at 30–40 m depth) separated from the thermocline (deeper than 68 65 m) by a high-salinity-gradient barrier layer (Lukas and Lindstrom, 1991). 69 In the Warm Pool, the phytoplankton are macronutrient-limited, and a deep 70 chlorophyll maximum occurs below the mixed layer (Barber and Chavez, 71 1991), where most of the primary productivity occurs (Le Borgne et al., 72 2011). Surface-nutrient depletion in the Warm Pool reflects the lack of up-73 welling and a deep thermocline, which under average climatic conditions is 74 located near the lower limit (approximately 80 m) at which there is sufficient 75 light for phytoplankton growth (Le Borgne et al., 2011). In addition to the 76 large horizontal movement of the eastern edge of the Warm Pool with ENSO. 77 the vertical structure within the Warm Pool also changes with ENSO phases. 78 During an El Niño, the thermocline can shoal to 40 m, which raises macronu-79 trients into the photic zone and increases primary productivity (Le Borgne 80 et al., 2011). 81

The tuna fisheries of the tropical Pacific Ocean mostly consist of skipjack 82 (Katsuwonus pelamis), yellowfin (Thunnus albacares), bigeye (T. obesus) and 83 albacore (T. alalunga) (Lehodey et al., 2011). In 2009, catches from the 84 western Pacific represented around 60% of the global tuna catch, of which 85 about 70% comprises skipjack (Lehodey et al., 2013). Skipjack are found 86 throughout the equatorial and subtropical Pacific, but catches are highest in 87 the Warm Pool (Lehodey et al., 1997). Sustaining benefits from the tuna 88 resources is a challenge for the Pacific Island Countries and Territories, as 89 the quantity and distribution of the fish catch display large variability from 90 year to year (Lehodey et al., 1997), and a changing ocean (e.g. Durack et al.,

<sup>92</sup> 2012) will make it even more difficult to maintain catch levels (Bell et al.,
<sup>93</sup> 2013).

#### 94 3. Methods

The climate model used in this study is the CSIRO Mk3.5 model of 95 Rotstayn et al. (2010), hereafter referred to as CSIRO35. The CSIRO35 pro-96 jection of the SRES (Special Report on Emissions Scenarios) A1B scenario 97 (Nakicenovic et al., 2000) for the decade of the 2060s is used to force an 98 HOM (Chamberlain et al., 2012). The selected SRES scenario describes a 99 future world of very rapid economic growth, with a global population peak-100 ing in the middle of the century and declining thereafter, and where from 101 mid-century there is also rapid introduction of new and more efficient tech-102 nologies balanced across fossil and non-fossil energy sources (Nakicenovic 103 et al., 2000). The HOM used in this study is the Ocean Forecasting Aus-104 tralia Model (Brassington et al., 2007; Oke et al., 2008), which is a near-global 105 model (covering latitudes of 70°S to 70°N). The HOM has 47 vertical levels, 106 with 10 m resolution in the upper 200 m, while the horizontal grid is vari-107 able: eddy-resolving around Australia (with  $0.1^{\circ}$  resolution between  $90^{\circ}E$ 108 and 180°E and between 20°N and 70°S) and increasing to a maximum of 109  $2^{\circ}$  in the north Atlantic. The HOM also has a simple ocean biogeochemi-110 cal formulation, namely the Whole Ocean Model with Biogeochemistry And 111 Trophic-dynamics (WOMBAT). WOMBAT is based on Kidston et al. (2011) 112 and has been implemented in the 3D ocean model 'Modular Ocean Model 113 version 4' (Dietze et al., 2009); details of WOMBAT are given in Matear 114 et al. (2013). 115

The HOM simulations used in this study are briefly summarised below, and Chamberlain et al. (2012) provides a detailed explanation of how the CSIRO35 climate change projection was used to simulate future climate change in the HOM.

To prepare the HOM, an initial spin-up of the ocean physics was per-120 formed, where the model was initialised with observed climatological fields 121 (Chamberlain et al., 2012) and forced by atmospheric reanalysis products (i.e. 122 windstresses, heat and freshwater fluxes) from 1991 to 2004 (ERA-40, Up-123 pala et al., 2005), while the surface layer was relaxed to the observed surface 124 temperatures (Reynolds and Smith, 1994) and salinities (Levitus, 2001) on a 125 30-day time-scale. HOM was then run for a second loop of atmospheric forc-126 ings in the same manner as the original spin-up for the period 1991–1994 but 127

with WOMBAT activated. The ocean state at the end of this spin-up period 128 was used as the initial state for the HOM present-day simulation. From the 129 HOM spin-up, the windstresses and the heat and freshwater fluxes from the 130 years 1993–2001 were averaged to produce a monthly climatology. To these 131 monthly climatologies we added diurnal variability in the atmospheric forcing 132 fields, which was obtained from the difference between the 1995 fields and 133 the corresponding monthly climatology computed for 1995. The year 1995 134 was chosen because it was a moderate year, with none of the major climate 135 indices (North Atlantic Oscillation, Antarctic Oscillation, North Pacific Os-136 cillation and ENSO) at an extreme (Large and Yeager, 2004). High-frequency 137 forcing can be important to the mixed-layer depth evolution (Large and Yea-138 ger, 2004), so we wanted to retain it in the forcing fields. The combined 139 monthly climatologies with diurnal variability gave the present-day atmo-140 spheric forcing fields used to force the HOM present-day simulation. With 141 these atmospheric forcing fields, the HOM present-day simulation was run 142 for 10 years, and we present results from the last five years of this simulation. 143 Analysis of the HOM simulations showed that after five years the simulations 144 were stable (Chamberlain et al., 2012). Longer HOM simulations of just the 145 physical system (Sun et al., 2012) revealed no decadal trend to the simulated 146 climate change, justifying the use of a shorter simulation period to investigate 147 the impact of climate change on phytoplankton. 148

For the HOM future climate change projection, we added to the present-149 day fields the changes in ocean state and changes in atmospheric forcings 150 from the CSIRO35 simulation to obtain the initial ocean state and atmo-151 spheric forcing fields for the HOM future simulation. From the CSIRO35 152 simulation, we compute the change in atmospheric forcing and ocean state 153 as the difference between the results of the 2060s and 1990s (i.e. 2060s state 154 minus 1990s state). WOMBAT was incorporated into the CSIRO35 simu-155 lation to allow us to compare the simulated phytoplankton change resulting 156 from the two models. With the future forcing fields, the HOM simulation 157 was run for 10 years and the averaged results over the last five years of the 158 simulation are reported here. 159

To investigate the impact of atmospheric changes over two decades (the 1990s and 2060s), the HOM simulations were performed as ocean forced simulations with atmospheric forcings that remove interannual variability by averaging a decade of atmospheric fields. Therefore, an important dynamical process of the western tropical Pacific, ENSO variability, is not represented in the projection and hence the simulations do not provide information on

how the character of ENSO might change with climate change. What the
projections do provide is information on how the mean ocean state may
change with climate change.

To explore ocean-atmosphere coupling, we run an additional HOM sim-169 ulation, where the 2060s winds are modified to assess how the future ocean 170 warming pattern in the HOM could alter the atmospheric circulation and 171 how this might affect future ocean dynamics (Chamberlain et al., 2012). To 172 modify the winds, an atmospheric model is driven by the projected sea sur-173 face temperatures from the 2060s HOM simulation and the 2060s CGCM 174 simulation (Chamberlain et al., 2012). These atmospheric-only simulations 175 allow us to quantify how changes in the future ocean warming pattern be-176 tween the HOM and CSIRO35 simulations alter the atmospheric circulation. 177 The difference in the winds from these two atmospheric simulations are then 178 added to the winds used to force the 2060s HOM simulation (called the wind-170 stress feedback) to investigate the potential interaction between the ocean 180 and atmosphere in the future climate change projection of the ocean state. 181

#### <sup>182</sup> 4. Present-Day Simulation

Before describing the projected changes in the western tropical Pacific with climate change, we present an initial assessment of the present-day HOM simulation and compare its results with both the observed fields and the fields simulated by CSIRO35. The key features included in this assessment are the sea surface temperature (SST), sea surface salinity (SSS), mixed-layer depth (MLD), zonal flows, ocean properties along the equator, and chlorophyll *a* (Chla) concentrations.

### 190 4.1. Sea Surface Temperature, Salinity and Mixed-Layer Depth

To assess the simulated SST in the 1990s, we compare the annual mean 191 SST pattern generated by the HOM with the observed climatological field 192 from Reynolds and Smith (1994) (see Figure 1a,b). In the western tropical 193 Pacific, the annual mean SST pattern is reproduced by the HOM, with r =194 0.93 and a root mean square (RMS) temperature difference of  $0.4^{\circ}$ C. The 195 model captures the observed east-west gradients in SST along the equator 196  $(3 \pm 0.5^{\circ}C/70$  degrees from the simulation versus an observed gradient of 197  $2.5 \pm 0.5^{\circ}$ C/70 degrees) but tends to underestimate the temperature by 1°C 198 north of 5°N. The simulated extent of the Warm Pool, using the dynamic 199 Warm Pool edge definition of Brown et al. (2013b) (i.e. the isotherm where 200

the salinity gradient along the equator is maximal), was 29.5°C, in good agreement with the observed Warm Pool extent given by the 29.2°C isotherm (Maes et al., 2010) (compare Figure 1a,b). According to the dynamic Warm Pool edge definition (Brown et al., 2013b), the edge of the Warm Pool at the equator in the HOM 1990s simulation was located around 170°E, compared to an observed location of 165–170°E (Maes et al., 2010).

For the 1990s, the CSIRO35 simulation gives a much warmer western 207 tropical Pacific (Figure 1c) and, using the dynamic Warm Pool edge defini-208 tion, the 29.7°C isotherm defines the extent of the Warm Pool, which at the 209 equator places the edge at about 160°E, slightly west of the observed edge at 210 165–170°E. For comparison with the HOM simulation, the CSIRO35 simu-211 lated SST correlation with the observations was similar (r=0.9) but the RMS 212 temperature difference was much greater  $(0.9^{\circ}C)$ . Along the equator in the 213 eastern equatorial Pacific, CSIRO35 displays a cold tongue bias, with sur-214 face water several degrees colder than the observations (Figure 1a.c): this is 215 a common feature of many global climate models. Off the equator, CSIRO35 216 yields a much more extensive Warm Pool than the observations. 217

For the 1990s, the HOM-simulated annual mean SSS field shows good 218 agreement with the observed climatological field from the 2009 CSIRO Atlas 219 of Regional Seas (CARS2009; this is an updated dataset that uses the same 220 methodology as Dunn and Ridgway (2002) and Ridgway et al. (2002) but 221 includes more recent data; it is available at www.cmar.csiro.au/cars), with 222 r = 0.92 and a RMS difference of 0.18 practical salinity units (Figure 2a,b). 223 In comparison, the CSIRO35 1990s surface salinity has a large fresh bias 224 (RMS difference of 1.3 practical salinity units) and a poor correlation with 225 the observations (r=0.3) (Figure 2c). 226

The MLD controls the exchange of heat between the atmosphere and the 227 ocean, as well as the light environment of the upper ocean, which affects 228 phytoplankton growth (Ryan et al., 2002). To assess the HOM 1990s sim-229 ulation, we compare the simulated monthly mean MLD with the observed 230 mean value from CARS2009. For the CARS2009 dataset, the MLD is de-231 fined as the minimum depth at which the temperature is  $0.4^{\circ}$ C less than the 232 value at  $10 \,\mathrm{m}$  and the salinity is 0.03 greater than the value at  $10 \,\mathrm{m}$  (Condie 233 and Dunn, 2006); we use the same definition to compute the MLD in our 234 model simulations. This definition of the MLD eliminates the possibility of 235 density-compensating temperature and salinity gradients being interpreted 236 as a well-mixed layer (Condie and Dunn, 2006). The spatial variability in 237 the HOM-simulated 1990s MLDs is consistent with observations. The Warm 238

Pool has the shallowest MLDs (less than 40 m), and the HOM simulations 239 gave MLDs about 10 m shallower than the observations (Figure 3a,b). To the 240 east of the Warm Pool, the HOM-simulated MLDs vary between 50 m and 241 100 m, which is consistent with the observations. The exception occurs just 242 north of the equator, where the simulated MLDs are slightly greater (20 m)243 than observed. The CSIRO35-simulated MLDs in the 1990s are generally 244 too shallow (Figure 3c), particularly in the Warm Pool region, where the 245 simulated MLD is only 20 m. 246

#### 247 4.2. Temperature and Salinity Along the Equator

Along the equator, the observations from CARS2009 show a strong zonal 248 gradient, with the warmest and freshest water found on the western side of 249 the section (Figure 4). In the west, the isotherms shoal, and a sub-surface 250 salinity maximum develops between 100 m and 200 m. In the HOM 1990s 251 simulation, the equatorial temperature and salinity sections (Figures 5a and 252 6a) display the same features as evident in the observations: the simulation 253 captures the magnitude of the zonal temperature and salinity gradients and 254 exhibits a sub-surface salinity maximum on the western side of the section 255 at the correct depth. The CSIRO35 1990s simulation also displays strong 256 zonal temperature and salinity gradients along the equator (Figures 5c and 257 6c). However, the zonal temperature gradient is much greater than observed 258 while the salinity is much lower than observed. The CSIRO35 simulation 259 displays a sub-surface salinity maximum on the west side of the section, but 260 it is about 30 m shallower than that in the observations. 261

#### 262 4.3. Zonal Flow

As described by Kessler et al. (2003), the western equatorial region has 263 an alternating pattern of upper ocean zonal currents, and both the HOM 264 and the CSIRO35 simulations show this behaviour (Figure 7). Around the 265 equator at about 3°N and 3°S, broad surface currents transport water west-266 ward; this South Equatorial Current (SEC) is present in the HOM simulation 267 but has more north-south structure than in the CSIRO35 simulation (Fig-268 ure 7a.c). Beneath the atmospheric convergence zones on both sides of the 260 equator, eastward flowing currents appear near the surface, known as the 270 North Equatorial Countercurrent (NECC) and the South Equatorial Coun-271 tercurrent (SECC). Both of these currents are present in the HOM and in 272 the CSIRO35 simulations, but the NECC is much stronger in the HOM sim-273 ulation than in the CSIRO35 simulation (Figure 7a,c). 274

At the equator and beneath the SEC, a strong sub-surface Equatorial 275 Undercurrent (EUC) carries water to the east. The core of the EUC is 276 nearly 200 m deep in the Warm Pool, and it shoals as the EUC flows eastward 277 (Figure 8a). Across the equator, east of the Warm Pool at 180°E, the EUC 278 flows east for depths between  $100 \,\mathrm{m}$  and  $300 \,\mathrm{m}$ , and this is evident in both 279 the HOM and the CSIRO35 simulations (Figure 9a,c). Above the EUC, in 280 the HOM simulation the SEC shoals at the equator, with deeper branches on 281 either side of the equator, whereas in the CSIRO35 simulation the SEC has 282 only one branch with a maximum just south of the equator (Figure 9a,c). 283

Comparing the simulated zonal currents of the HOM and CSIRO35, one 284 can see that the currents in the HOM simulation are generally stronger and 285 have a more prominent north-south gradient. In particular, the EUC is much 286 stronger in the HOM simulation. To assess the equatorial flow, we compare 287 the simulations with the average zonal current data from the TAO/TRITON 288 mooring at 165°E and 0°S (Figure 10). The figure clearly shows that the 289 magnitude and vertical structure of the HOM simulation are much more con-290 sistent with the mooring observations than the CSIRO35 simulation. While 291 the HOM simulation does a good job of representing the current profile and 292 magnitude of the EUC in general, the HOM-simulated EUC is about 20 m 293 shallower than that in the mooring data. This may reflect either a bias in 294 the model or a mismatch in time, as the mooring data comes from the 2000s 295 rather than the 1990s period simulated by the HOM. That the HOM sim-296 ulation has stronger equatorial currents, with more defined structure, is a 297 clear difference between it and the CSIRO35 present-day simulation, and is 298 an aspect where it is in better agreement with the observations. 299

#### 300 4.4. Phytoplankton

To assess the phytoplankton field produced by the simulations, we com-301 pare the simulated annual mean phytoplankton concentrations with chloro-302 phyll a concentrations estimated from the Sea-viewing Wide Field-of-view 303 Sensor (SeaWiFS) 1997–2008 mean climatology of eight-day, 9 km compos-304 ites generated by the NASA Goddard Space Flight Center. To perform the 305 comparison, we first convert the simulated phytoplankton concentrations to 306 chlorophyll a concentrations (expressed in nitrogen units) by using a con-307 version factor of 1 mmol N/1.59 mg Chla (Matear, 1995). Within the Warm 308 Pool, the HOM-simulated mean chlorophyll a concentrations are low (less 309 than  $0.15 \,\mathrm{mg}$  Chla/m<sup>3</sup>) but slightly greater than the observed values, which 310 are less than  $0.10 \,\mathrm{mg}$  Chla/m<sup>3</sup> (Figure 11a,b). To the east of the Warm Pool, 311

<sup>312</sup> both the HOM-simulated and the observed chlorophyll *a* concentrations show their highest values. However, the HOM-simulated values  $(0.35 \text{ mg Chla/m}^3)$ are slightly greater than the observed values  $(0.30 \text{ mg Chla/m}^3)$ . The CSIRO-<sup>315</sup> 35-simulated concentrations, on the other hand, are more than double the observed values at 140°W (Figure 11a,c). Within the Indonesian Seas, both the simulated and observed fields show mean chlorophyll *a* concentrations that are generally higher than 0.3–0.4 mg Chla/m<sup>3</sup>.

In general, both models overestimate the chlorophyll a concentrations 319 (Figure 11), but the CSIRO35 simulation has a much greater chlorophyll a320 concentration gradient along the equator, with too-high concentrations in 321 the eastern part of the region. The HOM-simulated spatial distribution of 322 chlorophyll a is consistent with the data from SeaWiFS (r = 0.75), showing 323 a similar magnitude of variability and small difference from the observa-324 tions (RMS difference of  $0.06 \text{ mg Chla/m}^3$ ), which gives us some confidence 325 that the HOM provides a realistic representation of the processes control-326 ling phytoplankton variability in the western tropical Pacific. In contrast, 327 the CSIRO35 simulation shows lower correlation with the data, with greater 328 spatial variability and a greater difference from the observations (r = 0.7)329 and RMS difference of  $0.15 \text{ mg Chla/m}^3$ ). We emphasise, however, that 330 the chlorophyll a comparison can be problematic for several reasons. First, 331 the conversion of modelled phytoplankton concentration (in nitrogen units, 332 mmol  $N/m^3$ ) to chlorophyll a concentration (mg Chla/m<sup>3</sup>) assumes a fixed 333 ratio, but the actual ratio is expected to vary (Taylor et al., 1997). Sec-334 ond, satellite-derived chlorophyll a concentrations are based on estimates of 335 water-leaving radiances, which are sensitive to the effects of poorly deter-336 mined corrections of the atmosphere on these radiances. Third, satellite-337 derived calculations tend to overestimate chlorophyll a concentrations near 338 the coast, because of the influence of dissolved organic matter and sediment 339 resuspension (Moore et al., 2007). Fourth, the nominal uncertainty in the 340 SeaWiFS estimates of chlorophyll a concentrations in the open ocean water is 341  $\pm 25-35\%$  (Behrenfeld et al., 2006). Because of these uncertainties in the ob-342 servations, in our assessment of the HOM simulation we focus on the spatial 343 pattern rather than the magnitude of simulated chlorophyll a concentrations. 344 Along the equator, both the HOM- and CSIRO35-simulated phytoplank-345 ton concentrations for the 1990s show high surface values in the eastern part 346 of the section, with a deep phytoplankton maximum existing beneath the 347 Warm Pool (Figure 12a,c). To produce a deep phytoplankton maximum, 348 sufficient light and nutrients are needed to sustain the phytoplankton. The 349

exponential decline in light levels with depth and the presence of the sub-350 surface phytoplankton maximum combine to reduce light levels below the 351 phytoplankton maximum and help confine the deep phytoplankton maxi-352 mum to a thin layer, thus preventing the occurrence of phytoplankton below 353 a depth of 110 m. While not allowing the simulated phytoplankton to adapt 354 to the low-light conditions beneath the deep phytoplankton maximum helps 355 to limit the extent of this layer in the model, it is a real feature of the Warm 356 Pool region and is observed to be only tens of metres thick (Maes et al., 357 2010). 358

The east-west gradient in the simulated phytoplankton concentrations and the existence of a deep chlorophyll maximum in the Warm Pool are both consistent with observations (Le Borgne et al., 2002). The presence in the HOM simulations of a deep phytoplankton maximum to the east of the Warm Pool (east of  $170^{\circ}$ E) is also observed in chlorophyll data. In this region, the observed deep chlorophyll maximum exceeds 0.3 mg Chla/m<sup>3</sup> (Maes et al., 2010), which is similar to the HOM-simulated values.

While both the HOM and the CSIRO35 simulations produce a deep phy-366 toplankton maximum along the equator, this feature is more extensive in the 367 HOM simulation, which is more consistent with observations. Further, the 368 phytoplankton concentrations in the HOM simulation have smaller magni-369 tude, which is also more consistent with the observations than the CSIRO35 370 simulation. The CSIRO35 simulation, with its cold tongue bias, has too 371 much upwelling of high-nutrient water in the eastern equatorial Pacific, and 372 this has the effect of maintaining much higher phytoplankton concentrations 373 in the eastern half of the section than what is observed. In the CSIRO35 374 simulation, upwelling of nutrients and their westward transport supply the 375 nutrients for phytoplankton growth along the equator. The HOM simulation 376 does have upwelling, but the presence of a deep phytoplankton maximum 377 along the equatorial section implies that nutrient supply to the photic zone 378 from below is occurring along the entire section. In the HOM simulation, 379 the existence of a strong vertical zonal current shear along the equator (Fig-380 ure 8) provides a mixing mechanism for supplying nutrients to the photic 381 zone, which lies below the mixed layer, and hence producing a sub-surface 382 phytoplankton maximum. 383

#### <sup>384</sup> 5. Climate Change Results and Discussion

In the following discussion, we define the climate change projected by 385 the HOM or by the CSIRO35 as the difference between the simulated ocean 386 states of the 2060s and the 1990s (i.e. 2060s state minus 1990s state). For 387 the HOM, we use monthly averages of the last five years of each 10-year 388 period of simulation; for CSIRO35, we use decadal averages for both periods. 389 We shall compare the HOM and CSIRO35 simulations with each other and 390 with recent analyses of global climate model projections (e.g. Brown et al., 391 2013a; Ganachaud et al., 2013) as well as observed multi-decadal trends in 392 the region. We will also use the HOM simulation with windstress feedback 393 to assess how interactions between the ocean and atmosphere could modify 394 future climate change projections. 395

#### <sup>396</sup> 5.1. Changes in Temperature, Salinity and Mixed-Layer Depth

The HOM climate change projection for the western tropical Pacific shows 397 considerable surface warming, with the greatest warming occurring along 398 the equator in the east Pacific and the least warming in the west (Fig-399 ure 13a). DiNezio et al. (2009) analysed multiple climate model projections 400 and also found the global warming maximum to occur along the equator east 401 of 150°W. Using the dynamic Warm Pool edge definition of Brown et al. 402 (2013b), the Warm Pool regions of the two models for the 1990s and 2060s 403 are shown in Figure 13. In the HOM projection, there is an eastward migra-404 tion of the Warm Pool with climate change, and the greatest warming occurs 405 along the equator in the expanded Warm Pool region (Figure 13a). The HOM 406 projects less than 2°C surface warming in the Warm Pool, but the warming 407 is in excess of 3°C just east of the 1990s Warm Pool edge. CSIRO35 projects 408 a similar magnitude and pattern of warming, with maximum warming tak-409 ing place along the equator east of the 1990s Warm Pool edge (Figure 13b). 410 CSIRO35 also projects an eastward movement of the Warm Pool along the 411 equator (Figure 13b). 412

Along the equator, both the HOM and CSIRO35 project the greatest warming to occur in the upper 100 m, while in the Warm Pool region the projections show prominent sub-surface cooling (up to 1°C) in the thermocline, revealing an uplift of the thermocline (Figure 5). CSIRO35 projects a similar magnitude of warming to the HOM (Figure 5b,d), but the maximum warming and maximum sub-surface cooling occur further west than in the HOM projection, which is consistent with the model having a 1990s Warm

Pool edge which was further west than that of the HOM (160°E compared 420 to 170°E). Ganachaud et al. (2013) showed that the projected multi-model 421 mean (MMM) warming of the CMIP3 climate models was generally restricted 422 to the upper ocean, with warming of  $2.5^{\circ}$ C at 50 m and 1°C at 100 m be-423 tween the 1990s and the 2100s. The projected MMM warming lacked the 424 sub-surface cooling at the equator beneath the Warm Pool, but this may 425 reflect the longer time period that Ganachaud et al. (2013) used to compute 426 the warming (1990s to 2100s, compared with 1990s to 2060s in our study). 427

The surface waters of the HOM and CSIRO35 projections show the great-428 est freshening in the western tropical Pacific; freshening declines to nearly 429 zero east of the Warm Pool (Figure 14). The freshening of the surface has a 430 very similar pattern in the two simulations and is consistent with observed 431 historical trends, which reveal a multi-decadal decline in salinity in the Warm 432 Pool (Cravatte et al., 2009; Durack et al., 2012). In the Warm Pool at the 433 equator, both models project that the freshening will extend down to 200 m. 434 while east of the Warm Pool the salinity decline is projected to be small. In 435 the HOM projection the freshening in the Warm Pool is deeper, and there is 436 also greater freshening outside the Warm Pool than in the CSIRO35 projec-437 tion (Figure 6). 438

Both climate change projections show a general shoaling of the MLD, with the maximum decline being less than 30 m (Figure 15). In the HOM projection, the greatest shoaling occurs near the eastern edge of the Warm Pool (maximum decline of less than 30 m). The CSIRO35 simulation also projected the largest declines in MLD to occur around the edge of the Warm Pool, but these changes appear to be greatest just off the equator (20 m decline), and along the equator the change in MLD is nearly zero.

Observational data from 1950 to 2008 showed that the maximum warming 446 of the western tropical Pacific occurred near the eastern edge of the Warm 447 Pool (Johnson and Wijffels, 2011), so both climate change projections are 448 consistent with this observation. In the Warm Pool, observed water tem-449 peratures have decreased by up to  $2^{\circ}$ C in the thermocline (100–150 m) over 450 58 years (Johnson and Wijffels, 2011; Durack and Wijffels, 2010); a similar 451 pattern of cooling is produced by the HOM and CSIRO35 climate change 452 projections, but the projected magnitude of cooling is less, at only 1°C. Such 453 cooling seems to be related to a weakening of the easterly equatorial winds, 454 which causes an adiabatic lifting of the thermocline (Han et al., 2006). Weak-455 ening of the easterly equatorial winds is a robust feature of future climate 456 change projections (Collins et al., 2010), and it is present in the climate 457

change projection used to force the HOM. The HOM-projected uplift of the
thermocline in the Warm Pool is consistent with the observed trend over the
past 50 years and with the expected response due to climate change.

#### 461 5.2. Change in Equatorial Currents

With climate change, along the equator both the HOM and CSIRO35 462 projected a weakening of the SEC, with the emergence of a weak eastward 463 flow in the Warm Pool (Figure 7b,d). However, in the CSIRO35 projection 464 these changes in the zonal flow are smaller and more diffuse than in the 465 HOM projection. Below the surface, the EUC is still prominent in the 2060s 466 in both the HOM and the CSIRO35 projections (Figure 8b,d). The HOM 467 predicts a less than 5% weakening in its EUC strength, but the core of the 468 EUC is projected to shoal by about 30 m east of the Warm Pool. CSIRO35 469 predicts similar changes in the EUC strength and position. 470

In the tropical Pacific, the upper ocean currents are expected to change in 471 the future as a result of weakened equatorial and northeasterly trade winds 472 together with strengthened southeasterly trade winds (Sen Gupta et al., 473 2012). Ganachaud et al. (2013) showed that according to the MMM, climate 474 change will decrease the velocity of the westward-flowing SEC, from 30 cm/s 475 in the 2000s to 20 cm/s in 2100, which is about double the decline predicted 476 by the HOM climate projections. Ganachaud et al. (2013) also showed that 477 from the MMM, by 2100 the EUC is expected to increase substantially, with 478 an approximately 20 m shoaling of the EUC core. With climate change, the 479 HOM predicted EUC shoaling but little change in its strength. Overall, the 480 HOM projects less change in the EUC and SEC than does the MMM, but 481 this discrepancy may reflect the difference in time periods covered by the 482 studies (the HOM used the 2060s to compute the change, while the MMM 483 used 2100). 484

#### 485 5.3. Changes in Phytoplankton

In the western tropical Pacific, the HOM projects a decline in surface 486 phytoplankton concentrations with climate change, except in the Indonesian 487 Seas, where there is a small increase (Figure 16a). The decline in phyto-488 plankton concentration is greatest along the equator near the Warm Pool 489 edge. CSIRO35 also projects a decline in surface equatorial phytoplankton 490 (Figure 16b), but the decline it predicts is much greater than that of the 491 HOM projection. At 100 m depth, the two climate change projections start 492 to differ (Figure 17). CSIRO35 projects a general decline in phytoplankton 493

concentrations, while the HOM projects a band of increased phytoplankton 494 concentrations along the equator across the whole region. Within the Warm 495 Pool, the increase in phytoplankton reflects the shoaling of the thermocline, 496 which raises the nutricline into the photic zone and thus increases phyto-497 plankton concentrations. This feature is most evident in the HOM projection 498 at 150°E; it is weakly present in the CSIRO35 projection but is shifted to 499 the west, at 140°E (Figure 12b,d). Le Borgne et al. (2011) hypothesised that 500 the shoaling of the thermocline with climate change, similar to what occurs 501 during an El Niño (Le Borgne et al., 2011), could increase nutrient supply 502 to the photic zone in the Warm Pool; however, the one climate projection 503 they looked at did not actually produce such a response. Along the equator 504 outside of the Warm Pool, CSIRO35 projected a decline in phytoplankton 505 concentration across the region while HOM projected an increase at 95 m 506 depth across the whole region (Figure 12b,d). The HOM projection suggests 507 that nutrient supply in the western tropical Pacific can increase with climate 508 change. 509

The HOM-projected increase in phytoplankton concentrations at 100 m 510 approximately cancels the decrease at the surface, and results in primary pro-511 ductivity in the equatorial Pacific remaining nearly unchanged in the HOM 512 projection (Figure 18b). In contrast, primary productivity declines substan-513 tially in the CSIRO35 projection (Figure 18d). Like the CSIRO35 projection, 514 other climate models generally project declines in the western tropical Pa-515 cific primary productivity with climate change (Steinacher et al., 2010). The 516 discrepancy in the responses of primary productivity to climate change is a 517 significant difference between the two projections. To help understand this 518 difference, let us look at the simulated behaviour along the equator. 519

Vertical shear mixing along the equator can supply nutrients to the photic 520 zone (Ryan et al., 2002), and in the HOM simulation this occurs in both 521 the 1990s and the 2060s, as demonstrated by the presence of a sub-surface 522 phytoplankton maximum along the equator (Figure 12). Since both the 523 zonal current strengths and the vertical current shears are much stronger 524 in the HOM than in the CSIRO35 simulation (Figure 10), this mechanism 525 is only apparent in the HOM simulation. Without the enhanced vertical 526 and horizontal resolution at the equator, CSIRO35 has much weaker zonal 527 currents with much less vertical current shear, and in this model the eastern 528 equatorial Pacific upwelling of nutrients and their subsequent transport west 529 is the main process supplying nutrients to the equatorial phytoplankton. This 530 is the dominant mechanism of nutrient supply in climate models (Steinacher 531

et al., 2010). As the ocean warms and stratifies and the upwelling declines, CSIRO35 projects a significant decline in phytoplankton concentrations and primary productivity in the western tropical Pacific, consistent with other climate model projections (Steinacher et al., 2010).

In both the HOM and the CSIRO35 projections, the EUC shoals with 536 little change in its strength (Figure 8). In the HOM projection, the shoaling 537 of the EUC increases the vertical current shear and increases vertical shear 538 mixing. The increase in vertical shear mixing increases the nutrient supply to 539 the photic zone and thus increases sub-surface phytoplankton concentrations 540 (Figure 12). Hence, this nutrient supply mechanism can counter the reduc-541 tion in nutrient upwelling in the eastern equatorial Pacific to yield a small 542 increase in primary productivity. The high resolution in our model is neces-543 sary to enable representation of the vertical shear mixing and the subsequent 544 primary productivity response to climate change. The increase in sub-surface 545 phytoplankton in the HOM climate change projection, accompanied by little 546 change in primary productivity, represents an important modification to ex-547 isting climate change projections and potentially has significant consequences 548 for the marine ecosystem. 549

#### 550 5.4. Impact of Projected Climate Change on Tuna Distribution

The warming and changes in primary productivity projected by the model 551 simulations could influence tuna distributions both directly, through changes 552 in preferred thermal environment, and indirectly, through changes in prev 553 abundance. To explore the impact of climate change on tuna distribution, 554 we consider how the projected changes might affect skipjack tuna habi-555 tat. By optimising the Spatial Ecosystem And Population Dynamics Model 556 (SEAPODYM), Lehodey et al. (2013) estimated that the preferred thermal 557 range was 26.5–32.5°C for spawning skipjack tuna and 16–25°C for adult 558 skipjack. Defining skipjack habitat as the water column thickness of the 559 tuna's preferred thermal range, we compare the 1990s skipjack habitat from 560 the HOM simulation with observations and with the projections for the 2060s 561 (Figures 19 and 20). 562

The simulated thickness of the spawning skipjack habitat in the 1990s compares well with the thickness calculated from the CARS2009 mean temperature data (Figure 19a,b). The thickest layer of habitat is found just south of the equator, where it exceeds 140 m. With climate change, the HOM projects that the maximum thickness of the spawning habitat will migrate southward to become centred around 10°S. By bias-adjusting the CSIRO35

climate change projection, we get a preferred spawning habitat which is very 569 similar to that in the HOM projection (Figure 19c,d). The increased resolu-570 tion of the HOM did not significantly change the projected thickness of the 571 spawning habitat. For comparison, the simulations of Lehodev et al. (2013), 572 using SEAPODYM and a bias-corrected climate change projection, showed 573 that by 2100 the favourable skipjack spawning ground would shift to higher 574 latitudes but also into the central and eastern Pacific. Neither of the climate 575 projections studied here display a large eastward shift, but perhaps this is a 576 feature that emerges only after the 2060s. 577

The HOM-simulated thickness of the adult skipjack thermal habitat in 578 the 1990s also shows good agreement with observations (Figure 20a,b). The 579 HOM simulation captures the slightly thicker habitat along the equator that 580 is apparent in the observations. With climate change, HOM projects little 581 change in the thickness of the adult thermal habitat  $(\pm 10 \text{ m})$  (Figure 20c). 582 CSIRO35 projected a similar thickness of adult habitat to the HOM but 583 with a slightly increased thickness (20 m) in the western equatorial Pacific 584 (Figure 20c,d). What the CSIRO35 projection misses is the narrow band of 585 increased thickness along the equator evident in the HOM projection. The 586 equatorial band of increased adult habitat is associated with the increased 587 vertical shear mixing that occurs in the HOM projection, which led to the 588 increased primary productivity. Lehodey et al. (2013), using the IPSLc cli-589 mate model projection, predicted that the biomass of adult tuna will shift its 590 core habitat from the western to the central equatorial region by 2100. Our 591 climate projections do not suggest such a shift, but it is possible that this 592 shift may develop only after the 2060s. For the 2060s, our model predicts 593 that across the entire western tropical Pacific, suitable adult skipjack habitat 594 will remain greater than 50 m, which is comparable to the thickness of the 595 1990s habitat in the Western Pacific Warm Pool (Figure 20a,c). 596

From the MMM, Ganachaud et al. (2013) concluded that with climate 597 change, skipjack are likely to move substantially eastward and poleward by 598 2100. While an eastward and poleward extension of the population may oc-590 cur, we emphasise that in the western tropical Pacific, a region for which 600 little change in the future habitat is projected, one would still expect to find 601 skipjack tuna in the 2060s. Further, given the increase in sub-surface phy-602 toplankton concentrations in a narrow band along the equator, little change 603 in primary productivity, and the continued presence of suitable habitat for 604 adult and spawning skipjack, conditions are such that one would expect to 605 see little change in the 2060s skipjack population. Perhaps, as projected by 606

Lehodey et al. (2013), future declines in skipjack habitat and biomass will occur only after the 2060s.

The HOM simulation that incorporated windstress feedback predicted a 609 slightly more positive situation for skipjack; that is, in the western equa-610 torial Pacific the adult and spawning thermal habitats were projected to be 611 slightly greater (about  $10 \,\mathrm{m}$  thicker) and the primary productivity about 10%612 higher along the equator than in the original HOM projection. This modi-613 fied HOM projection has slightly greater vertical shear mixing caused by the 614 EUC, which increased by about 10% with climate change. While the HOM 615 projected slightly positive conditions for skipjack, other species of tuna with 616 a preference for the Warm Pool could also benefit from the projected changes 617 for the 2060s. 618

#### <sup>619</sup> 5.5. Robustness of the Climate Projection

To compute the climate projection with the HOM, we defined climate 620 change as the difference in the CSIRO35-simulated climate state between 621 the 1990s and the 2060s. Given the potential for multi-decadal variability 622 in the western tropical Pacific, it is possible that our simulations are biased 623 because we derived climate change from the difference between the climate 624 states of two decades. To assess decadal bias, we compare the change in zonal 625 windstress between the 1990s and the 2060s with the difference derived from 626 a three-decade average centred on the periods of interest; that is, we use the 627 difference between the years 1980–2009 and the years 2050–2079 to compute 628 climate change (Figure 21a,b). The magnitude and pattern of change are 629 very similar for the two calculations. A similar comparison was also made 630 for the windstress curl (Figure 21c,d), and we again found good agreement 631 in the pattern and magnitude for the two calculations. The similarity of the 632 windstress changes demonstrates that our climate change estimate is unlikely 633 to have been biased by multi-decadal variability. 634

In the HOM projection, a key driver of the increase in the deep phy-635 toplankton maximum along the equator and the weak response of primary 636 productivity to climate change was the shoaling of the EUC. This change in-637 creased the vertical current shear and increased vertical shear mixing. Anal-638 ysis of multiple climate projections suggests both shoaling of and an increase 639 in the EUC with climate change (Sen Gupta et al., 2012). Sen Gupta et al. 640 (2012) showed that the strengthening of the EUC is purely a wind-driven 641 response to the projected intensification of southeasterly trade winds and an 642 associated off-equatorial windstress curl change in the southern hemisphere. 643

CSIRO35 projects a small reduction in the EUC strength, as opposed to 644 most other models, which have shown that the EUC increases with climate 645 change (Sen Gupta et al., 2012); in particular, the MMM exhibits a sub-646 stantial strengthening of the EUC with climate change (Ganachaud et al., 647 2013). A greater increase in EUC strength should further increase vertical 648 shear mixing and hence the supply of nutrients to the photic zone along 649 the equator. Therefore, because CSIRO35 did not project a substantial in-650 crease in EUC strength, the HOM projection here may be underestimating 651 the vertical shear mixing and the potential for climate change to increase 652 phytoplankton concentrations and primary productivity in the western trop-653 ical Pacific. The HOM climate change projection with windstress feedback 654 has an EUC that increased by about 10%, and it did lead to slightly greater 655 primary productivity. 656

#### 657 6. Summary

For most of the physical environmental variables, the 2060s HOM climate 658 projection was similar to the CSIRO35 projection. Enhanced resolution and 659 bias correction appear to have only a minor impact on the climate change 660 projection of the physical ocean state. Both low- and high-resolution climate 661 projections showed a maximum surface warming east of the Warm Pool, a 662 shoaling of the thermocline in the Warm Pool, and eastward expansion of 663 the Warm Pool. In the Warm Pool, the shoaling of the thermocline raises 664 the nutricline into the photic zone and increases phytoplankton and primary 665 productivity, a feature that is most evident in the HOM projection but is 666 also weakly present in the CSIRO35 projection. 667

Where the HOM projection displayed a clear difference from the CSIRO35 668 projection was in the impact of climate change on phytoplankton concentra-669 tions. For phytoplankton, the enhanced resolution of the HOM had an impor-670 tant effect. The HOM simulation had stronger and better-defined zonal cur-671 rents than the CSIRO35 simulation, and this enabled vertical current shear 672 mixing to play an important role in generating a phytoplankton sub-surface 673 maximum along the equator in the Western Pacific. The HOM projected 674 a shoaling of the EUC with climate change, which enhanced the vertical 675 shear mixing and increased the vertical supply of nutrients to the photic 676 zone, resulting in an increase of sub-surface phytoplankton concentrations. 677 The increase in sub-surface phytoplankton concentrations helped to offset the 678 decline in surface phytoplankton and to produce simulation results showing 679

almost no change in primary productivity in the western tropical Pacific with 680 climate change. In contrast, CSIRO35 projected a substantial reduction in 681 phytoplankton concentrations and primary productivity, a response that is 682 typical of climate change projections for the region (Steinacher et al., 2010). 683 The projected expansion of the Warm Pool along with little projected 684 change in primary productivity and in suitable habitat for skipjack tuna 685 suggest that by the 2060s, climate change will not have had an adverse im-686 pact on skipjack tuna populations. Beyond the 2060s the situation may be 687

different, as suggested by Lehodey et al. (2013), but there is a need to repeat their study using a high-resolution model that can resolve current shear mixing and its response to climate change.

An important limitation of the HOM is that the simulations were not 691 dynamically coupled to the atmosphere. Chamberlain et al. (2012) used sim-692 ulations with varying heat, freshwater and windstress coupling of the HOM 693 with the atmosphere to assess the robustness of the HOM climate change 694 projection. These sensitivity experiments showed that the pattern of phyto-695 plankton change was robust (with a spatial correlation of 0.8 between pro-696 jections), and there was a slight amplification of the response (10% increase)697 when the feedback of the ocean state on the windstresses was included in the 698 simulations. While Chamberlain et al. (2012) in their sensitivity experiments 699 probed the impact of changes in ocean warming on atmospheric dynamics, 700 they did not investigate the response of the coupled system. Given the impor-701 tance of atmosphere-ocean coupling in the western tropical Pacific, there is 702 a need to undertake global climate model simulations with a high-resolution 703 ocean model. Such simulations should also investigate how ENSO variability 704 is affected by climate change. 705

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719 7. Figures

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Figure 1: Annual mean sea surface temperature (°C) from a) the observations of Reynolds and Smith (1994), b) the HOM simulation for the 1990s, and c) the CSIRO35 simulation for the 1990s. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) for the three datasets. The thick white line in a) shows the dynamic Warm Pool edge in the 1990s HOM simulation.



Figure 2: Annual mean sea surface salinity from a) the observations based on CARS2009, b) the HOM simulation for the 1990s, and c) the CSIRO35 simulation for the 1990s. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) for the three datasets.



Figure 3: Monthly averaged mixed-layer depth (m) from a) the observations based on CARS2009, b) the HOM simulation for the 1990s, and c) the CSIRO35 simulation for the 1990s. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) for the three datasets.

#### a) Observed Temperature (°C) 0 40 30 29 80 28 27 26 25 24 23 22 21 (m) 120 -Ht 160 -200 -240 280 130°E 150°E 170°E Longitude 170°W 150°W 110°E b) Observed Salinity 0. 40 35.5 80 (ju 120 - 1 160 - 1 160 - 1 200 - 1 35 34.5 34 240 33.5 280 33 170°E Longitude 130°E 150°E 170°W 150°W 110°E

Figure 4: Observations along the equator (between  $3^{\circ}N$  and  $3^{\circ}S$ ) from the CARS2009 climatology: a) annual mean temperature (°C); b) annual mean salinity.



Figure 5: Simulated averaged mean temperature (°C) along the equator (between 3°N and 3°S): a) temperatures from the 1990s HOM simulation; b) difference between temperatures of the 2060s and the 1990s obtained from the HOM projection; c) temperatures from the 1990s CSIRO35 simulation; d) difference between temperatures of the 2060s and the 1990s obtained from the CSIRO35 projection.



Figure 6: Simulated annual averaged salinity along the equator (between  $3^{\circ}N$  and  $3^{\circ}S$ ): a) salinities from the 1990s HOM simulation; b) difference between salinities of the 2060s and the 1990s obtained from the HOM projection; c) salinities from the 1990s CSIRO35 simulation; d) difference between salinities of the 2060s and the 1990s obtained from the CSIRO35 projection.



Figure 7: Annual averaged upper ocean (0-50 m) mean zonal flow obtained from a) HOM simulation for the 1990s, b) HOM simulation for the 2060s, c) CSIRO35 simulation for the 1990s, and d) CSIRO35 simulation for the 2060s. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s in the respective simulations.



Figure 8: Simulated eastward velocity (cm/s) along the equator (between 3°N and 3°S), obtained from a) HOM simulation for the 1990s, b) HOM simulation for the 2060s, c) CSIRO35 simulation for the 1990s, and d) CSIRO35 simulation for the 2060s.



Figure 9: Simulated eastward velocity (cm/s) at  $180^{\circ}$ E, obtained from a) HOM simulation for the 1990s, b) HOM simulation for the 2060s, c) CSIRO35 simulation for the 1990s, and d) CSIRO35 simulation for the 2060s.



Figure 10: Observed zonal averaged current profile (cm/s) at the TAO/TRITON current mooring site ( $165^{\circ}E$  and  $0^{\circ}S$ ), along with simulated values from the HOM and CSIRO35 for the 1990s and 2060s.



Figure 11: Annual mean surface chlorophyll *a* concentration (mg Chla/m<sup>3</sup>): a) computed from 1997–2008 eight-day, 9 km composites of SeaWiFS; b) obtained from the 1990s HOM simulation; c) obtained from the 1990s CSIRO35 simulation. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s from the observations and the respective simulations.



Figure 12: Simulated annual average phytoplankton concentration (mmol N/m<sup>3</sup>) along the equator (between 3°N and 3°S): a) concentrations from the 1990s HOM simulation; b) difference between concentrations of the 2060s and the 1990s obtained from the HOM projection; c) concentrations from the 1990s CSIRO35 simulation; d) difference between concentrations of the 2060s and the 1990s obtained from the CSIRO35 projection.



Figure 13: Projected annual averaged change in sea surface temperature between the 1990s and the 2060s, obtained from a) the HOM simulation and b) the CSIRO35 simulation. The thick black (white) lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s (2060s) in the respective simulations.



Figure 14: Projected annual averaged change in sea surface salinity between the 1990s and the 2060s, obtained from a) the HOM simulation and b) the CSIRO35 simulation. The thick black (white) lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s (2060s) in the respective simulations.



Figure 15: Projected change in the annual mean mixed-layer depth (m) between the 1990s and the 2060s, obtained from a) the HOM simulation and b) the CSIRO35 simulation. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s in the respective simulations.



Figure 16: Projected change in annual mean surface phytoplankton concentration (mmol  $N/m^3$ ) between the 1990s and the 2060s, obtained from a) the HOM simulation and b) the CSIRO35 simulation. The thick black (white) lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s (2060s) in the respective simulations.



Figure 17: Projected change in annual mean phytoplankton concentration (mmol  $N/m^3$ ) at a depth of 100 m between the 1990s and the 2060s, obtained from a) the HOM simulation and b) the CSIRO35 simulation. The thick black (white) lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s (2060s) in the respective simulations.



Figure 18: Simulated annual mean primary productivity (mol  $C/m^2/y$ ): a) primary productivity from the 1990s HOM simulation; b) change in primary productivity between the 1990s and the 2060s obtained from the HOM projection; c) primary productivity from the 1990s CSIRO35 simulation; d) change in primary productivity between the 1990s and the 2060s obtained from the CSIRO35 projection. The thick black (white) lines represent the dynamic Warm Pool edge (Brown et al., 2013b) of the 1990s (2060s) in the respective simulations.



Figure 19: Suitable thermal habitat for spawning skipjack tuna, defined as the thickness of the water column with a temperature between 25°C and 32°C, from a) the observations based on CARS2009, b) the HOM simulation for the 1990s, c) the HOM simulation for the 2060s, and d) the bias-corrected CSIRO35 simulation for the 2060s. The thick white lines represent the dynamic Warm Pool edge (Brown et al., 2013b) in the 1990s HOM simulation.



Figure 20: Suitable thermal habitat for adult skipjack tuna, defined as the thickness of the water column with a temperature between 20°C and 26°C, from a) the observations based on CARS2009, b) the HOM simulation for the 1990s, c) the HOM simulation for the 2060s, and d) the bias-corrected CSIRO35 simulation for the 2060s. The thick white lines represent the dynamic Warm Pool edge (Brown et al., 2013b) in the 1990s HOM simulation.



Figure 21: CSIRO35-simulated change in annual mean zonal windstress: a) between the 1990s and the 2060s; b) between 1980–2009 and 2050–2079. CSIRO35-simulated change in the annual mean windstress curl: c) between the 1990s and the 2060s; d) between 1980–2009 and 2050–2079. The thick black lines represent the dynamic Warm Pool edge (Brown et al., 2013b) in the 1990s HOM simulation.

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