Intraseasonal variability of mixed layer depth in the tropical Indian Ocean

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Abstract :

In this paper, we use an observational dataset built from Argo in situ profiles to describe the main largescale patterns of intraseasonal mixed layer depth (MLD) variations in the Indian Ocean. An eddy permitting (0.25A degrees) regional ocean model that generally agrees well with those observed estimates is then used to investigate the mechanisms that drive MLD intraseasonal variations and to assess their potential impact on the related SST response. During summer, intraseasonal MLD variations in the Bay of Bengal and eastern equatorial Indian Ocean primarily respond to active/break convective phases of the summer monsoon. In the southern Arabian Sea, summer MLD variations are largely driven by seemingly-independent intraseasonal fluctuations of the Findlater jet intensity. During winter, the Madden-Julian Oscillation drives most of the intraseasonal MLD variability in the eastern equatorial Indian Ocean. Large winter MLD signals in northern Arabian Sea can, on the other hand, be related to advection of continental temperature anomalies from the northern end of the basin. In all the aforementioned regions, peak-to-peak MLD variations usually reach 10 m, but can exceed 20 m for the largest events. Buoyancy flux and wind stirring contribute to intraseasonal MLD fluctuations in roughly equal proportions, except for the Northern Arabian Sea in winter, where buoyancy fluxes dominate. A simple slab ocean analysis finally suggests that the impact of these MLD fluctuations on intraseasonal sea surface temperature variability is probably rather weak, because of the compensating effects of thermal capacity and sunlight penetration: a thin mixed-layer is more efficiently warmed at the surface by heat fluxes but loses more solar flux through its lower base.

<u>1. Introduction</u>

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The mixed-layer, i.e., the quasi-homogeneous upper ocean layer with a fairly uniform 50 density profile, is critical to the ocean variability as it acts as the interface between the 51 atmosphere and ocean interior. This layer is also an essential parameter for air-sea 52 interactions: a shallow Mixed Layer Depth (MLD) exhibits a reduced thermal capacity and 53 can hence promote large sea surface temperature (SST) anomalies (e.g. Shinoda and Hendon, 54 1998). This is particularly critical in the Indian Ocean (IO), 40% of which is covered by 55 waters warmer than 28.5°C. At those high surface temperatures, small SST perturbations can 56 57 induce strong variations of the deep atmospheric convection (Gadgil et al. 1984) with both 58 local and remote consequences on the atmospheric circulation (see review by Schott et al. 2009). In addition to air-sea interaction, MLD also affects the primary productivity and the 59 timing of phytoplankton blooms through controlling the availability of nutrients and light, as 60 well as the dilution of grazers (Sverdrup, 1953; Behrenfeld and Michael 2010). Because of 61 62 those impacts, it is important to describe MLD variability in the IO and the main processes that control it. While the MLD seasonal variability has been abundantly described, as detailed 63 64 below, there are fewer studies that investigate its variability at other timescales, and in particular at the omnipresent intraseasonal timescale in the IO (e.g. Goswami, 2005; Zhang, 65 66 2005). The objective of this study is thus to describe intraseasonal MLD variability in the IO, 67 as well as the associated climate modes and driving processes, using a combination of observations and model experiments. 68

Numerous studies have already investigated the patterns and mechanisms of seasonal 69 70 MLD variations in the IO (Rao et al. 1989; McCreary and Kundu 1989; Rao and Sivakumar 2003; Prasad 2004; de Boyer Montegut et al. 2007; Sreenivas et al. 2008) and discussed their 71 72 biological impact (e.g. Wiggert et al. 2005; Levy et al. 2007; Kone et al. 2009). The amplitude of the seasonal MLD variations is particularly large in the Arabian Sea, reaching up to 30 m 73 74 (Figure 1a). The mechanisms driving these variations differ during the southwest and northeast monsoons. The summer monsoon is characterized by a strong southwesterly 75 76 Findlater jet (Findlater, 1969) along the western coast of the Arabian Sea, which markedly 77 deepens the MLD in the southern central Arabian Sea owing to Ekman convergence. The winter monsoon is characterized by a negative heat flux at the air-sea interface that plays a 78 dominant role in the convective deepening and cooling of the MLD (Rao et al. 1989; 79 Prasanna Kumar and Narvekar, 2005; de Boyer Montegut et al. 2007). As shown on Figure 80

1a, the Bay of Bengal (BoB) exhibits weaker seasonal MLD variations of ~10 m (Gopalakrishna et al. 1988; Rao et al. 1989; Shenoi et al. 2002; Prasad, 2004; Babu et al. 2004; Narvekar and Prasanna Kumar, 2006), especially in its northern part where strong salinity stratification prevents convective cooling. Finally, the southwestern tropical IO also exhibits large seasonal MLD variations (~20-30 m; Figure 1a), which have been related to the annual cycle of the wind, through both its stirring effect and impact on buoyancy fluxes, and to thermocline depth (Foltz et al. 2010).

As depicted by Keerthi et al. (2013) using both an ocean model and in-situ 88 89 observations, the IO is also home to very significant MLD fluctuations at interannual timescales. They showed that the MLD interannual variability is typically of ~10 m (Figure 90 91 1b), about two to four times smaller than the seasonal cycle (Figure 1a), except in the eastern 92 equatorial IO and along the Sumatra and Java coast. Aside from coastal and subtropical 93 regions where eddy-related small-scale structures dominate interannual MLD variations, Keerthi et al. (2013) found that a large fraction of IO MLD interannual variations could be 94 95 related to large-scale climate modes. The Indian Ocean Dipole, a mode of interannual variability intrinsic to the tropical IO and arising from a positive feedback between the ocean 96 97 and atmosphere (e.g. Saji et al. 1999), explains most of the MLD interannual variability close to the equator. The Subtropical Indian Ocean Dipole, a large-scale mode of SST variability 98 that apparently arises from atmospheric forcing in the subtropical southern IO (Behera and 99 Yamagata 2001), largely controls large-scale winter MLD variations in the southern IO. In 100 contrast, while El Niños in the neighbouring Pacific Ocean drive SST fluctuations in the IO 101 through zonal shifts in the Walker circulation (e.g. Klein et al. 1999), MLD variations related 102 to El Niño and interannual variations of the summer monsoon appear to be rather weak in the 103 IO. Buoyancy fluxes appear to dominate interannual MLD fluctuations in most of these 104 regions, with wind stirring and Ekman pumping only playing a role in a few regions. 105

106 The tropical IO is also home to very clear atmospheric intraseasonal variability, which arises from the interaction between atmospheric large-scale dynamics and deep atmospheric 107 convection. Fluctuations in deep atmospheric convection, rainfall, surface winds and air-sea 108 109 fluxes within the 30-90 days frequency band develop as the result. This variability is dominated by the northward propagating active and break phases of the monsoon in summer 110 (e.g. Goswami, 2005), and by the eastward propagating Madden Julian Oscillation (MJO, e.g. 111 Zhang, 2005) in winter. Those phenomena have strong regional consequences, for instance 112 impacting agriculture (e.g. Gadgil, 2003; Ingram et al. 2002), modulating the occurrence of 113 tropical weather systems and cyclones (e.g. Webster and Hoyos, 2004; Bessafi and Wheeler, 114

2006) or influencing the IO chlorophyll variability (e.g. Resplandy et al. 2009; Jin et al.2012).

Many studies have identified strong intraseasonal SST fluctuations in response to the 117 aforementioned atmospheric signals in the IO for both winter (Harrison and Vecchi, 2001; 118 Duvel et al 2004, Vialard et al. 2008; Vialard et al. 2013) and summer (Sengupta et al 2001; 119 Vecchi and Harrison 2002; Duvel and Vialard, 2007; Vialard et al. 2012). As an illustration, 120 an index of intraseasonal monsoon activity proposed by Goswami (2005) and detailed in 121 Section 2.4 reveals two strong intraseasonal convective perturbations from July to September 122 123 2000 (Figure 2a): these perturbations are associated with large SST variations of ~0.8°C in the BoB (Figure 2b). Similarly, a strong MJO event in January-February 1999 as depicted by 124 125 the Wheeler and Hendon (2004) index detailed in Section 2.4 (Figure 2d) was found to force a peak-to-peak intraseasonal SST perturbation of ~1°C south of the equator (Figure 2e; 126 127 Harrison and Vecchi, 2001; Duvel and Vialard, 2007). It is important to understand the processes responsible for these SST variations because they appear to feed back onto the 128 129 atmospheric intraseasonal variability (e.g., Maloney and Sobel, 2004; Matthews, 2004; Bellon et al. 2008; Bellenger and Duvel, 2009). 130

While SST intraseasonal variability has been extensively studied in the IO, there have 131 132 only been a handful of studies discussing MLD intraseasonal variations. Yet, those MLD intraseasonal variations are far from negligible. An ocean general circulation model 133 simulation (to be described in detail in the next section) for example suggests that the 134 intraseasonal SST signals in Figure 2b, e are associated with MLD intraseasonal variations 135 ranging from 10 to 20 m in these regions (Figure 2c, f). Historically, intraseasonal MLD 136 variations have been difficult to estimate from observations due to the scarcity of *in situ* data. 137 The advent of the ARGO program considerably increased the number of available in situ 138 profiles over the recent decade. Compiling these data, Drushka et al. (2012, 2014) estimated a 139 MLD fluctuation of more than 15 m peak-to-peak in the eastern equatorial IO in response to 140 winter MJO forcing. In contrast to seasonal and interannual timescales, there is however to 141 142 date no exhaustive description and understanding of the main patterns of intraseasonal MLD variations at the basin scale of the IO. Modelling results from Keerthi et al. (2013), however, 143 suggest that interannual and intraseasonal MLD variations have roughly the same magnitude 144 (around 10 m), except along the equator and northern BoB where intraseasonal fluctuations 145 are about twice as large (Figure 1b, c). Intraseasonal MLD variations along the equator are 146 even larger than their seasonal counterpart (Figure 1a, c). 147

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Atmospheric heat flux forcing appears to be the dominant process driving

intraseasonal SST variability in the BoB in response to active/break phases of the summer 149 monsoon (Waliser et al. 2004; Bellon et al. 2008; Duncan and Han, 2009; Vialard et al. 2012) 150 and south of the equator in response to winter MJO forcing (Duvel and Vialard, 2007; 151 Javakumar et al. 2011). Much debate however remains about the possible impact of 152 intraseasonal MLD variations on these intraseasonal SST variations. Several studies indeed 153 suggest that a slab ocean model using climatological MLD estimates can reasonably well 154 capture observed intraseasonal SST signals in the southwestern tropical IO (Jayakumar et al. 155 2011) and northwestern Australian Basin (Vialard et al 2013) in winter, as well as in the 156 157 northern BoB in summer (Vialard et al. 2012). This suggests that MLD intraseasonal variability does not strongly contribute to SST intraseasonal variability in these regions. 158 159 Drushka et al. (2012) have shown that intraseasonal MLD variations in the southwestern tropical IO and northwestern Australian Basin regions do not affect intraseasonal SST, but 160 161 that not accounting for intraseasonal MLD variations in the eastern equatorial IO could result in an overestimation of intraseasonal SST signals by up to 40% there. 162

163 Our aim in the present paper is to investigate intraseasonal MLD fluctuations. To that end, an observational dataset built from Argo data and outputs from an eddy permitting 164 (0.25°) regional ocean general circulation model will be used to describe the main large-scale 165 patterns of intraseasonal MLD variations in the IO. We will also show that these MLD 166 fluctuations are linked with well-known modes of atmospheric intraseasonal variability in 167 most regions (the MJO in winter and active/break phases of the Indian monsoon during 168 summer), except in the Arabian Sea where more local atmospheric fluctuations related to 169 intraseasonal fluctuations of the Findlater jet in summer and intraseasonal air temperature 170 perturbations in winter explain the large intraseasonal MLD variations there. Our modelling 171 approach will finally allow us to understand the main mechanisms responsible for these MLD 172 variations (buoyancy fluxes vs. wind stirring), and to assess their potential impact on the 173 related SST response. The paper is organized as follows. Section 2 describes the numerical 174 experiments, the observed MLD validation product, as well as the statistical methods used to 175 176 extract the intraseasonal signals. In Section 3, we describe the patterns of intraseasonal MLD variability in the model and observations for both summer (Section 3.1) and winter (Section 177 3.2), and relate them to atmospheric variability. Model analysis and sensitivity experiments 178 are further used to discuss the respective control of wind-driven mixing and buoyancy fluxes 179 on MLD intraseasonal fluctuations (Section 4.1) and their impact on intraseasonal SST 180 signals (Section 4.2). The last section provides a summary and discussion of our results. 181

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183 **<u>2. Data and Methods</u>**

This section describes the observational and modelling tools used in the present study. Section 2.1 describes the model configuration and reference experiment along with the sensitivity experiment used to disentangle the respective influence of buoyancy fluxes and wind stirring on these MLD fluctuations. Section 2.2 describes the processing used to infer intraseasonal MLD variations from Argo data and the datasets used to describe the associated atmospheric variability. The filtering and composite analysis methods are described in Sections 2.3 and 2.4.

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192 2. 1. Modelling tools

The model configuration used here is based on the NEMO ocean general circulation 193 modelling system (Madec 2008) and is an IO sub-domain from the global 0.25° resolution 194 (i.e. cell size ~ 25 km) coupled ocean/sea-ice configuration described by Barnier et al. (2006). 195 The African continent closes the western boundary of the domain. The oceanic portions of the 196 eastern, northern and southern boundaries use radiative open boundaries (Treguier et al. 197 2001), constrained with a 150-day timescale relaxation to 5-day-average velocities, 198 199 temperature and salinity from an interannual global 0.25° simulation (Dussin et al. 2009), using a similar atmospheric forcing as the regional simulation detailed below. This simulation 200 is a product of the DRAKKAR hierarchy of global configurations (Drakkar Group 2007) and 201 has been extensively validated in the tropical Indo-Pacific region (Lengaigne et al. 2012; 202 203 Keerthi et al. 2013; Nidheesh et al. 2013).

204 The model starts from World Ocean Atlas temperature and salinity climatologies (Locarnini et al. 2010) at rest and is forced from 1990 to 2007 with the Drakkar Forcing Set 205 206 #4 (DFS4, Brodeau et al. 2010) which consists of a modified version of the CORE dataset (Large and Yeager, 2004). In this forcing dataset, ERA40 reanalysis (Uppala et al. 2005) and 207 European Centre for Medium-Range Weather Forecasts (ECMWF) analysis after 2002 are 208 used to compute latent and sensible heat fluxes. Radiative fluxes are based on corrected 209 International Satellite Cloud Climatology Project-Flux Dataset (ISCCP-FD) surface radiations 210 (Zhang et al. 2004) and precipitation forcing consists of a blending of several products 211 proposed by Large and Yeager (2004), including two of the most widely used datasets: the 212 global precipitation climatology project (GPCP, Huffman et al. 1997) and the Climate 213

Prediction Center Merged Analysis of Precipitation (CMAP, Xie and Arkin, 1997). All atmospheric fields are corrected to avoid temporal discontinuities and remove known biases (see Brodeau et al. 2010 for details). This experiment successfully reproduces the observed boreal summer intraseasonal SST variations along the coasts of India (Nisha et al. 2013) and boreal winter intraseasonal SST variations in the thermocline ridge region and Northwest Australian basin (Vialard et al. 2013). A more detailed description of the reference experiment can be found in Nisha et al. (2013) and Vialard et al. (2013).

The MLD is controlled by air-sea fluxes of both momentum and buoyancy. 221 Momentum fluxes drive vertically sheared currents, thereby inducing upper ocean mixing and 222 modulating the MLD, while buoyancy fluxes across the air-sea interface modulate the MLD 223 through their stabilizing or destabilizing effect (e.g., Weller and Price, 1988; McWilliams et 224 al., 1997). We hence also perform a sensitivity experiment to evaluate the respective influence 225 of wind stresses and atmospheric buoyancy fluxes in forcing intraseasonal MLD signals in the 226 model. After storing the wind stress computed by the model in the reference simulation, the 227 sensitivity experiment (hereafter NOWIND) is forced by smoothed wind stress that filtered 228 229 out the intraseasonal component, keeping the buoyancy flux forcing identical to the reference simulation. This sensitivity experiment is run over the same 1990–2007 period from the same 230 initial condition as in the reference experiment. 231

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2. 2. Observed datasets

As in Drushka et al. (2012), observed MLD signals are derived from Argo profiles 234 235 downloaded from the Global Ocean Data Assimilation Experiment (GODAE) database. To avoid erroneous MLD estimates, profiles with less than five measurements within the top 236 237 100m as well as without measurements in the top 6 m were discarded. As in Drushka et al. 238 (2012), MLD for each profile was calculated as the depth at which density exceeds density at a 6 m reference depth by 0.05 kg/m³. Modelled MLD is calculated online using a 0.01 kg/m³ 239 criterion, lower than the one used in observations because of the absence of a proper diurnal 240 variability in the model (de Boyer Montégut et al. 2004). We also validate the model MLD 241 climatology to the climatology derived from observations by de Boyer Montégut et al. (2004, 242 dBM04). 243

Typical intraseasonal perturbations of convection, wind, air and SST associated with intraseasonal MLD perturbations will be described using daily data from the National Oceanic and Atmospheric Administration 2.5° resolution gridded Outgoing Longwave Radiation
(OLR) product (Liebmann and Smith, 1996), 10 m winds and 2 m surface air temperature
from ERA-Interim reanalysis data (Dee et al. 2011), windstress data from QuikSCAT
scatterometer produced at Centre ERS d'Archivage et de Traitement (CERSAT, Bentamy et
al. 2003) and optimally interpolated Tropical Rainfall Measuring Mission Microwave
Instrument 0.25° resolution SST data produced by Remote Sensing Systems (available at
www.remss.com).

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2.3. Filtering method

Intraseasonal signals are isolated using 20 to 110-day filtering based on Fourier 255 transform for all datasets (except MLD Argo-based estimates; see below). Using different 256 257 filtering methods and different bandpass windows (e.g. 30 to 60 days and 30 to 90 days windows) does not significantly affect our results. In contrast with other data sources in this 258 study, Argo data are unevenly distributed in space and time, and Fourier filtering can 259 therefore not be applied. Intraseasonal signals from Argo data are thus estimated as the 260 261 difference between the raw signal and a background signal representative of the seasonal and interannual components. This background MLD signal is estimated from the monthly gridded 262 density dataset produced by Roemmich and Gilson (2009), which is based exclusively on 263 Argo profiles and is available from 2004 onward. MLD from this product was then low-pass 264 filtered with a 110-day cutoff and projected onto the exact time and position of each Argo 265 profile using linear interpolation to provide the expected background MLD component for 266 each profile. 267

In our eddy-permitting simulation, there is a significant amount of meso-scale MLD variability associated with oceanic eddies or other small-scale features. Since we are interested in large-scale MLD variations, the model MLD is filtered in space to retain only large spatial scales (> 250 km). We do this by applying the iterative application of the heat diffusion equation described in Weaver and Courtier (2001), which is well suited to conduct spatial filtering in domains with complex boundaries, like the ocean.

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275 <u>2.4. Composite analysis</u>

The sparse and irregular temporal and spatial distribution of Argo profiles does not easily allow mapping MLD variations for individual intraseasonal events. We therefore

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compute composites by averaging all measurements made at a given grid point during a given 278 phase of, for example, the MJO or monsoon active/break cycle, defined using one of the 279 indices described below. The floats provide patchy spatial coverage in some regions, so we 280 281 restrict our analysis to grid boxes where more than twenty Argo profiles were available for a given phase. Regions where the magnitude of the composite average is smaller than the 282 standard error are masked to highlight the significant patterns of variability. Shaded areas 283 indicate a signal that is coherent across various events, and not merely noise. As shown in the 284 following, both summer and winter intraseasonal MLD variations derived from these 285 286 composites are of order of 10 m peak-to-peak but it must be kept in mind that individual MLD events can reach amplitude of up to 30 to 40 m in all regions discussed below. 287

We use two well-known indices to define the phases of the main modes of 288 intraseasonal variability in the IO, namely the MJO in winter and the active/break phase of the 289 monsoon in summer. The temporal evolution of the MJO is based on the real-time multi-290 variate MJO indices (RMM1 and RMM2) proposed by Wheeler and Hendon (2004). They 291 correspond to the principal components of a pair of empirical orthogonal functions of the 292 293 combined fields of near-equatorially averaged 850-hPa zonal wind, 200-hPa zonal wind, and satellite-based outgoing longwave radiation data (see Wheeler and Hendon, 2004 for details). 294 295 These indices can be used to separate the MJO evolution into eight discrete phases that represent the location of the active MJO as it moves eastward over the IO and through the 296 Pacific Ocean. The RMM1 evolution for the 1999 winter season is shown on Figure 2d as an 297 illustration, with the 8 phases indicated. Composites based on these eight phases (referred to 298 299 MJO phases in the following) will be used to describe the MLD signals associated to the MJO forcing. 300

301 The Wheeler and Hendon (2004) index, however, fails to capture the northward 302 propagation of the monsoon intraseasonal oscillation in boreal summer (Kikuchi et al. 2012). We hence also use a simple index of monsoon active and break phases proposed by Goswami 303 (2005; hereafter Monsoon index) and constructed as the difference between BoB (70°E–95°E, 304 10°N-20°N) and equatorial IO (70°E-95°E, 5°S-5°N) intraseasonal-filtered outgoing 305 longwave radiation (a proxy for atmospheric convection). As illustrated on Figure 2a for the 306 2000 summer monsoon, we divide this index into six discrete phases that depict the northward 307 308 propagation of the intraseasonal monsoon spells. Positive and negative parts of the index are each divided into 3 phases of equal duration. As we will show in the following, Phase 2 309 corresponds to the index maximum and captures the monsoon break phase, while Phases 3 310

and 4 correspond to the transition phase from a break to an active phase. Similarly, the index
mimimum of Phase 5 captures the monsoon active phase, with Phases 6 and 1 corresponding
to the subsequent transition to the break phase.

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315 **<u>3. MLD intraseasonal variability in the IO and its mechanisms</u>**

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3.1. Summer intraseasonal MLD variations

We will first provide a general overview of the seasonal and intraseasonal MLD variations. We will show that intraseasonal MLD variations in the BoB and eastern equatorial Indian Ocean (EEIO) are related to active/break convective phases of summer monsoon while those in the southern Arabian Sea (SAS) are largely driven by seemingly independent intraseasonal fluctuations of the intensity of the Findlater jet.

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3.1.1. General overview

The observed June to September (JJAS) climatological MLD and wind stress in the 324 325 Tropical IO are shown on Figure 3a. The northern IO, BoB and Arabian Sea exhibit contrasted MLD patterns. The mixed layer is deeper in the Arabian Sea (up to 50 m), because 326 327 of the intense Findlater jet that causes both mixing and downwelling to the east of the jet axis (de Boyer Montegut et al. 2007). In contrast, the northern BoB displays a shallower MLD in 328 329 response to the stabilizing effect of the intense freshwater flux received by this basin during the summer monsoon (Shenoi et al. 2002). South of the equator, the MLD is largely driven by 330 331 the intensity of climatological winds, with a deeper MLD south of 10°S where easterlies are 332 strongest. The model reproduces these large-scale MLD structures reasonably well (Figure 333 3b). However, the model simulates a shallower MLD than the one inferred from the observational dataset in the coastal regions, which is likely to arise because a lack of Argo 334 profiles in these regions leads to uncertainities in the MLD estimates (Nisha et al. 2013). The 335 model MLD is also shallower along the equator, which cannot be explained by observational 336 coverage but is probably related to a deficiency in either atmospheric forcing or the vertical 337 mixing scheme. 338

The strongest summer intraseasonal MLD fluctuations are found in the EEIO, where typical MLD variations exceed 15 m (Figure 3c). Variations of the order of 10 m are also evident in the BoB, SAS and south of 10°S. In the BoB and EEIO, these intraseasonal MLD variations occur over regions of relatively shallow climatological MLD (20-40 m) and could

therefore influence the mixed layer heat budget at intraseasonal timescales. The MLD 343 variability depicted on Figure 3c can be either the result of large-scale intraseasonal 344 atmospheric forcing or the intraseasonal signature of oceanic meso-scale variations. Contours 345 on Figure 3c show the standard deviation of the model intraseasonal MLD variations after 346 applying a 250-km low-pass filter (discussed in section 2.3). This analysis illustrates that 347 intraseasonal MLD variability in the southwestern Arabian Sea is largely an intraseasonal 348 signature of small-scale variations, most likely related to energetic meso-scale eddies 349 occurring in the Somalia and Oman upwellings (Brandt et al. 2003), with larger-scale 350 351 variations occurring further east. We will thus focus on the three regions framed on Figure 3c (the boxes' boundaries are provided in Table 1), where large-scale MLD maxima are found: 352 353 the BoB, the EEIO and the SAS.

During summer, monsoon active/break phases are the most prominent mode of 354 355 intraseasonal atmospheric variability. Figure 4a, d maps the percentage of variance of largescale intraseasonal modelled MLD and OLR explained by the Monsoon index: this figure 356 357 illustrates to which extent this mode drives the MLD fluctuations on Figure 3c and where this mode is related to large atmospheric convective perturbations at intraseasonal timescales. As 358 359 expected, this OLR-based index explains a large fraction of atmospheric convection 360 intraseasonal variance in the EEIO and BoB boxes (Figure 4d). Figure 4a further reveals that this index is also able to explain a large fraction of the MLD variance in these two regions. 361 Box-averaged intraseasonal MLD variations in the BoB (resp. EEIO) box indeed display a 362 maximum correlation with the Monsoon index of -0.74 (0.72) at 5-day lag. This illustrates 363 that the EEIO and BoB MLD vary out of phase at intraseasonal timescales, under the 364 influence of active/break phases of the summer monsoon as will be discussed in section 3.1.2. 365

In contrast, the Monsoon index is unable to explain the MLD variations in the SAS 366 box (Figure 4a). Joseph and Sijikumar (2004) report that intraseasonal modulation of the 367 Findlater jet induces strong low-level wind perturbations over the Arabian Sea in summer. 368 We therefore constructed an index based on averaged intraseasonal zonal wind over the SAS 369 box, referred to as the "Jet index" in the following. In contrast to the monsoon index, the Jet 370 index is able to explain a large part of the MLD variance in the SAS (Figure 4b), with a 0.8 371 correlation between intraseasonal MLD and zonal wind fluctuations averaged over this 372 373 region. This result illustrates that summer SAS intraseasonal MLD variations are largely driven by intraseasonal wind fluctuations associated with the Findlater jet. The Jet index is 374 not strongly related with the Monsoon index (maximum lag correlation of 0.3), suggesting 375

that intraseasonal Findlater jet fluctuations in the Arabian Sea are quite independent from 376 convective perturbations. The Jet index indeed only explains up to 30% of OLR variance 377 along the western coast of India (Figure 4e), and ~15-25% in the BoB where strongest 378 convective perturbations are found in summer. Collectively, the Monsoon and Jet indices 379 explain a large part of MLD variations (50-70%) in the 3 regions of strongest variability 380 (SAS, BoB and EEIO, Figure 4c, f). In the following, we will use the Monsoon index (i.e. 381 active/break monsoon phase) to describe intraseasonal MLD fluctuations in the BoB and 382 EEIO regions and the Jet index (i.e. enhanced / reduced Findlater Jet) for MLD fluctuations in 383 the SAS region. 384

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3.1.2. MLD response to active/break phases of the monsoon

Figure 5 displays the composite patterns of intraseasonal OLR and wind along with 387 modelled and observed intraseasonal MLD for the Monsoon index phases 1 to 4. The wind 388 389 and convection patterns (Figure 5a-d) are typical of the evolution from a break to an active phase of the summer monsoon (e.g. Goswami 2005). Phase 1 characterizes the onset of a 390 391 break phase with weakly supressed convection over the Indian subcontinent and weakly enhanced convection in the equatorial region. Phase 2 is typical of the peak of the break 392 phase, with increased convection (with typical OLR signals of -20W.m⁻²) south of the equator 393 while a tilted band of suppressed convection occupies the northern IO (Figure 5b). The 394 395 associated wind stress anomaly displayed as contours on Figure 5f indeed shows a decreased monsoonal wind flow across the SAS and BoB, and increased eastward flow at the equator, as 396 a consequence of the monsoon jet deflection around the southern tip of India. The bands of 397 excess and suppressed convection progress northward during the transition between Phases 3 398 and 4 (Figure 5c-d), with enhanced convection over the southern part of the BoB and southern 399 India during Phase 4 (Figure 5d). Phases 5 and 6 are almost exactly the opposite of Phases 2 400 and 3 and are therefore not shown: they characterize a monsoon active phase with an 401 increased monsoon flow and deep atmospheric convection across the Indian subcontinent and 402 northern part of the BoB. 403

The model MLD response to those monsoon active/break phases is shown on the middle panels of Figure 5, with largest and out-of-phase MLD variations in the EEIO and BoB regions. The main patterns of MLD changes generally agree well with wind stress intensity changes (coutours on the middle panels of Figure 5). MLD anomalies are largest during Phase 3 (and Phase 6, not shown). During Phase 3, increased westerly winds and 409 convection in the EEIO result in a MLD deepening of up to 10m while reduced monsoonal 410 south-westerly winds and convection in the BoB act to shoal the MLD by up to 7 m there. In 411 contrast with convective signals, MLD anomalies do not exhibit any clear northward 412 propagation from BoB to EEIO but rather appear as a standing oscillation. In contrast, the 413 Arabian Sea exhibits a weak MLD shoaling signal (up to 2 m) that appears to propagate 414 northward from its southern (Phase 1; Figure 5e) to its northern boundary (Phase 4; Figure 415 5h).

416 These modelled MLD composites generally agree well with the observed estimates from Argo both in terms of structure and amplitude, except for Phase 2 (Figure 5, lower 417 418 panels), for which observations do not exhibit the significant signal in the BoB that is seen in the model. Figure 6a, b provides more quantitative comparison of the box-averaged 419 420 composites in regions of largest MLD variations. Consistent with Figure 5, MLD anomalies are out of phase between the EEIO and BoB boxes, with maximum deepening during Phase 3 421 in the BoB and Phase 6 in the EEIO. Peak-to-peak amplitude of composite MLD signals 422 derived from Argo data reaches 8 m for the BoB and EEIO regions. In both regions, the 423 model MLD evolution closely matches the observed one, despite a slight amplitude 424 overestimation in the EEIO region. It must however be noted that the amplitude of individual 425 events can largely exceed those derived from the composite analysis: for example, peak-to-426 peak MLD variations during summer 2000 reached 30 m (20 m) in the EEIO (BoB) (Figure 427 2). 428

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3.1.3. MLD response to intraseasonal Findlater jet variations

As discussed above, summer MLD intraseasonal variability also exhibits a clear 431 maximum in the SAS region (Figure 3), associated with intraseasonal modulation of the 432 Findlater jet over the Arabian Sea. Composite patterns of OLR, wind and MLD intraseasonal 433 anomalies associated with the Findlater Jet index are displayed on Figure 7. The upper panels 434 on Figure 7 illustrate the onset (Phase 1; Figure 7a), mature (Phase 2; Figure 7b), decay 435 (Phase 3; Figure 7c) and termination phases (Phase 4; Figure 7d) of intraseasonal pulses of 436 this jet, which are evident in the composite wind field over the SAS. As expected from the 437 weak maximum lag-correlation between Findlater Jet and Monsoon indices (0.3), 438 intraseasonal fluctuations of the Findlater jet are only related to modest convective 439 perturbation over the BoB (up to 6 W.m⁻² to be compared with the 20 W.m⁻² perturbations 440 related to monsoon active/break phases). Largest convective perturbations (up to 12 W.m⁻²) 441

442 are found southwest of India during the mature phase of the intensification of this jet (Phase 443 2; Figure 7b). As already noted by previous authors (e.g. Murtugudde et al. 2007), there are 444 large Ekman pumping signals of opposite phases on both sides of the jet, associated with 445 fluctuations in the jet intensity (most clearly during phases 2-3, see Figure 7fg). We will 446 discuss the role of those Ekman pumping perturbations on the mixed layer depth in section 4.

447 Composite patterns of modelled MLD anomalies related to these phases are provided 448 on Figure 7e-h. Strongest MLD fluctuations occur during phases 2 and 3 in the SAS region, 449 when the jet is most intense (Figure 7b, c), with MLD deepening of up to 7 m associated with 450 increased winds (Figure 7f, g). MLD signals are weak during the transition phases 1 and 4 451 (Figure 7e and 7h). The spatial pattern and amplitude of this modelled MLD composite is in broad agreement with the one derived from the observation (Figure 7i-l), where the deepest 452 453 MLD signals also occur in the SAS region during phases 2 and 3 (Figure 7i, k). Figure 6c provides a more quantitative comparison of the modelled and observed MLD variations in 454 this region. The peak-to-peak amplitude of composite MLD signals derived from Argo data 455 (green line) reaches 12 m. The modelled MLD phase and amplitude reasonably matches the 456 457 observed one, despite a slight tendency for the model to lead the observed signal.

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3.2. Winter intraseasonal MLD variations

We will now describe winter intraseasonal MLD variations, and show they mostly occur in the southeastern equatorial Indian Ocean (where they are primarily driven by the Madden Julian Oscillation) and in the northern Arabian Sea (in response to advection of continental air temperature anomalies).

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465 <u>3.2.1. General overview</u>

466 In boreal winter, the Eurasian continent cools and a high-pressure region develops on 467 the Tibetan plateau with resulting north/northeasterly winds over the Arabian Sea (Smith and 468 Madhupratap, 2005; DileepKumar, 2006; Figure 8a, vectors). Though the winds are not as strong as during summer (Figure 3a), they are cold and dry, leading to strong evaporative 469 470 cooling (Dickey et al. 1998). This buoyancy forcing at the air-sea interface leads to convective mixing and ocean mixed layer deepening in the northern Arabian Sea (Lee et al. 471 2000; de Boyer Montegut et al. 2007). The MLD remains shallow along the equator and 472 southern IO, due to relatively weak winter winds there. Figure 8a, b shows the model is able 473

to capture these observed seasonal MLD patterns, despite a slight overestimation in deepMLD regions and underestimation in shallow MLD regions.

476 Strongest winter intraseasonal MLD variations are found in the southeastern Equatorial IO (SEEIO) and in the northern Arabian Sea (NAS) regions, where the typical 477 MLD amplitude exceeds 10 m (Figure 8c). MLD fluctuations of about 8m also occur in the 478 479 BoB. Contours on Figure 8c display the large-scale model MLD variations and illustrate that most of the signal in NAS and SEEIO regions is large-scale, while it is largely mesoscale in 480 the BoB, consistent with previous observational results (Drushka et al., 2014). We will 481 therefore focus our analysis on the NAS and SEEIO regions framed on Figure 8c (see boxes 482 details in Table 1). 483

The most prominent mode of atmospheric winter intraseasonal variability is the MJO. 484 Figure 9a, d maps the variance percentage of large-scale modelled intraseasonal MLD and 485 OLR fluctuations that can be explained by the MJO index. This index explains 30 to 60% of 486 OLR variations in the central and eastern IO and 20 to 30% of intraseasonal MLD variations 487 south of the equator and in the SEEIO box. SEEIO box-averaged intraseasonal MLD has a 488 489 maximum correlation of 0.5 at lag 0 with the Wheeler and Hendon (2004) MJO index. However, the MJO index explains a weaker percentage of variance than the monsoon index in 490 491 summer for the BoB and EEIO boxes. A more local wind index (average intraseasonal zonal wind over the SEEIO box) enhances the correlation with MLD variations in the SEEIO region 492 493 from 0.5 to 0.75. This local index results in similar MLD patterns to those obtained with the Wheeler and Hendon (2004) MJO index and we therefore decided to illustrate the MLD 494 495 variations in the EEIO box using this widely used Wheeler and Hendon (2004) MJO index.

The MJO index is unable to explain the strong MLD fluctuations in the NAS region 496 497 (Figure 9a). At seasonal timescales, strong evaporative cooling associated with cooler and drier air drives the MLD deepening is the northern Arabian Sea (Prasanna Kumar and 498 Narvekar 2005). Hypothesizing that this mechanism also operates at intraseasonal timescales, 499 500 we constructed an index based on intraseasonal air temperature fluctuations averaged over the NAS box (NAS index), which represents intraseasonal fluctuations of the outbreaks of cold 501 502 air over the northern Arabian Sea. This index explains a large part of MLD variance in the NAS region (Figure 9b), with a 0.8 correlation between average intraseasonal MLD and air 503 temperature fluctuations over the NAS box. This illustrates that, as for seasonal timescales, 504 intraseasonal winter MLD fluctuations in the northern Arabian Sea are driven by intraseasonal 505

air temperature fluctuations. The NAS index is uncorrelated with atmospheric convection anywhere in the IO and is only weakly correlated with the MJO indices (maximum lagcorrelation of 0.2), indicating that those air temperature fluctuations are not related to intraseasonal atmospheric convective variations. Drivers of the intraseasonal NAS air temperature variations are discussed below.

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<u>3.2.2. MLD response to the MJO</u>

Figure 10 (top panel) shows the winter MJO typical OLR and wind evolution for 513 suppressed convection over the Indian Ocean. The MJO is associated with eastward 514 propagation of suppressed convective signals in the equatorial IO, with positive OLR and 515 easterly wind anomalies propagating from the western part (Phase 1, Figure 10a) to the 516 eastern part of the IO (Phase 4, Figure 10d). Suppressed MJO conditions are strongest in 517 phases 3 and 4 in the eastern Indian Ocean slightly south of the equator, with positive OLR 518 519 anomalies of ~15 W.m-2. Maximum wind anomalies are found south of the equator, where climatological winds are westerly (Figure 8b). Westerly wind anomalies during Phase 1 520 521 therefore correspond to an intensification of these climatological westerly winds while easterly wind anomalies during Phase 3 and 4 correspond to a reduction of the climatological 522 523 westerlies (contours on Figure 10e-f). Phases 5 to 8 are almost exactly the opposite of Phases 1 to 4 and are therefore not shown: they characterize an MJO active phase over the IO with an 524 increased deep atmospheric convection and westerly anomalies south of the equator. 525

The model MLD response to MJO forcing is largest in the SEEIO region, where MLD 526 shoals during phases 3 and 4 (colors on Figure 10e-f) in response to reduced climatological 527 westerlies (contours on Figure 10e-f) and suppressed convection (Figure 10a-b). Model MLD 528 patterns (Figure 10e-h) generally agree with Argo observations (Figure 10i-l), with a 529 maximum (minimum) MLD in the SEEIO box during Phase 1 (Phase 4). Figure 11a provides 530 a quantitative comparison of modelled and observed MLD intraseasonal variations in this 531 region. The peak-to-peak amplitude of composite MLD signals derived from Argo data (green 532 533 line) reaches 5 m. The model MLD generally agrees with the observed MLD within the uncertainties, with a slight tendency for the model to lag the observed signal. 534

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<u>3.2.3. MLD response in the NAS</u>

537 We saw earlier that winter intraseasonal MLD variations in the NAS region are 538 strongly related to local air temperature fluctuations. During winter, snowfall and weaker

solar radiation cools the Asian continent and continental high pressure builds up. This results 539 in northerly winds advecting dry and cold air masses from the continent towards the equator 540 over the IO (de Laat and Lelieveld 2002, Figure 12a). Figure 12b shows a latitude-time 541 542 section of lag-regressed air temperature over the Arabian Sea and continent to the north of it to air temperature in the NAS box. This figure reveals that air-temperature fluctuations over 543 the NAS region are related to large intraseasonal air temperature fluctuations over northwest 544 India and south Pakistan and propagate southward at ~5 degrees latitude per day. Mean winds 545 over the continent are relatively weak (Figure 12a). Figure 13 further shows composite maps 546 of temperature and wind anomalies associated with intraseasonal variations of surface air 547 temperature over the NAS. These clearly illustrate that Phases 2-3 correspond to anomalously 548 549 warm air over the continent and the NAS region and associated southerly wind anomalies. Phases 5-6 (not shown) are associated with anomalies opposite in sign to those in Phases 2-3 550 551 and are related to anomalously cold air and northerly wind anomalies in these regions.

These intraseasonal cold air intrusions and related wind fluctuations over the northern 552 part of the Arabian Sea result in large-scale MLD variations. These variations are associated 553 554 with a positive latent heat flux anomaly (not shown) that shoals the model MLD, with the strongest signals occurring during Phase 2 and 3 (Figure 13f-g). The model MLD signal is 555 consistent with the one derived from observations for Phase 3 (Figure 13k, g) but the 556 insufficient number of observed Argo profiles prevents a proper validation of the signal 557 during Phase 2. Time evolution of the MLD signal in the NAS box agrees well between 558 model and observations, although the model displays a somewhat larger amplitude (Figure 559 560 11b).

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563 <u>4. Related mechanisms and SST impact</u>

In this section, we will explore the mechanisms driving intraseasonal tropical Indian Ocean MLD variations (Section 4.1) and discuss their potential impact on intraseasonal SST variations (Section 4.2).

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4.1. Mechanisms driving intraseasonal MLD fluctuations

569 Our objective in this subsection is to better quantify the processes that control 570 intraseasonal MLD fluctuations in the regions of largest variability in summer (BoB, EEIO and SAS) and winter (SEEIO and NAS). The influence of buoyancy fluxes and wind stirring influences can be respectively estimated by calculating surface buoyancy fluxes and the cube of the friction velocity, which are roughly proportional to the amount of energy transferred from the atmosphere to the mixed layer (Niiler and Kraus, 1977). We will use these two parameters to qualitatively infer their respective contribution onto the modelled MLD variations. The net surface buoyancy flux B_0 is computed as follows:

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$$B_o = \frac{\alpha Q_{net}}{c_p} + \beta (P - E) S_o \tag{1}$$

where the first and second terms on the right hand side are respectively the buoyancy fluxes due to heat and fresh water fluxes. α and β the coefficients of thermal and haline expansion, Q_{net} is the net heat flux at the air-sea interface, C_p the specific heat capacity of seawater, P-E the net surface fresh water flux and S_o the surface salinity (Gill 1982). The friction velocity u^{*} is calculated as:

583
$$u^* = \sqrt{\frac{\tau}{\rho}}$$
(2)

where τ is the surface wind stress and ρ the density of seawater. Previous studies (e.g. 584 Murtugudde et al. 2007) indicate that a third mechanism can control MLD in the Arabian Sea 585 in summer. Strong Ekman pumping variations on the southern flank of the Findlater jet (see 586 Figure 7fg and 14c) can indeed also influence the mixed layer depth by making the 587 thermocline shallower or deeper, and hence stabilizing or destabilizing the ocean column near 588 the bottom of the mixed layer. We verified that significant intraseasonal Ekman pumping 589 variations only occur in the SAS box (i.e. the only box close to the Findlatter jet in summer) 590 591 and will hence only show Ekman pumping variations for that region (Figure 14c).

Figure 14f-j demonstrates that MLD deepening is associated with both reduced 592 593 buoyancy fluxes and increased frictional velocity in all regions, except in the NAS where friction velocity variations are negligible. These two forcing mechanisms therefore combine 594 595 to produce intraseasonal MLD fluctuations in most regions. MLD deepening (resp. shoaling) for all regions except NAS are indeed associated with a wind intensification (resp. reduction) 596 597 (see contours on middle panels of Figures 5, 7 and 10 and Figure 14, upper panels), which both increases (resp. reduces) the frictional velocity and reduces (resp. increases) the 598 599 buoyancy fluxes through a modulation of the amplitude of the evaporative cooling (Figure 14,

middle panels). MLD deepening (resp. shoaling) for all regions except NAS are also 600 associated with an OLR reduction (resp. increase) (see colors on top panels of Figures 5, 7 601 602 and 10 and Figure 14, upper panels), which contribute to reduce (resp. increase) the buoyancy fluxes through a modulation of the amplitude of incoming shortwave flux (Figure 14, middle 603 604 panels). In the NAS region, there is no strong wind variation (see contours on middle panel of Figure 13 and Figure 14e), and hence no wind stirring, but the changes in air temperature 605 drive evaporation and hence buoyancy changes. The non-solar heat flux component 606 (dominated by latent heat flux variations; not shown) significantly contributes to buoyancy 607 608 flux fluctuations in all regions (Figure 14a-e). Latent heat fluxes dominate buoyancy fluxes fluctuations in the Arabian Sea for winter (NAS box, Figure 14j), due to the modest deep 609 610 atmospheric convection and surface solar heat flux perturbations associated with MLD 611 variations there (see Figure 9e and Figure 14e). In contrast, summer MLD fluctuations in the 612 BoB and EEIO regions and winter MLD fluctuations in the SEEIO are related to phenomena (MJO and monsoon active/breaks phases) that involve a clear modulation of atmospheric 613 614 convection (Figure 14a, b, d) and related surface solar flux. As a result, solar heat flux also contributes to buoyancy flux in these regions (Figure 14f, g, and i). Decreased atmospheric 615 616 convection is generally associated with reduced winds (Figure 14a-e), explaining the in-phase 617 relationship of solar and non-solar heat fluxes contribution to buoyancy fluxes. In the SAS region, the intensification of the Findlater jet is associated with a negative wind stress curl, 618 i.e. wind-driven downwelling (Fig14c) that slightly lags the maximum MLD deepening. 619 These wind stress curl variations probably contribute to the MLD deepening in the SAS 620 region, in addition to the wind stirring and buoyancy effects. This is confirmed by the spatial 621 pattern of the deepening (Figure 7fg) that is collocated with the maximum Ekman pumping 622 rather than with the largest wind stress anomalies. 623

It is difficult to quantify the respective influences of buoyancy fluxes and wind stirring 624 on MLD fluctuations from the above analysis, as these two terms strongly co-vary (see 625 626 correlations between buoyancy fluxes and frictional velocity on Figure 14f-j). We therefore use the NOWIND sensitivity experiment described in section 2.1, forced by intraseasonal-627 628 filtered wind stress and identical buoyancy flux forcing to the reference simulation. Comparing REF (green line) and NOWIND experiment (dashed green line) on Figure 14k-o 629 630 therefore allows to quantitatively assessing the respective role of buoyancy fluxes and wind stirring (plus Ekman pumping in the SAS box) on intraseasonal MLD variations. The only 631 632 region where buoyancy fluxes almost entirely control (~90%) intraseasonal MLD fluctuations

is the NAS box in winter. In the other regions (BoB, EEIO and SAS in summer and SEEIO in 633 winter), the buoyancy fluxes and the wind stirring have a similar contribution, with the 634 contribution from buoyancy fluxes ranging from 45% in the SAS box in summer to 65% in 635 the SEEIO box in winter. In these four boxes, MLD fluctuations are associated with similar 636 amplitude buoyancy fluxes signals but frictional velocity fluctuations are comparatively 637 weaker in the equatorial regions (EEIO and SEEIO). This suggests that in these regions, a 638 relatively small wind stress perturbation produces a MLD response that is quite similar to 639 other regions, due to the increased responsiveness of currents (and hence shear and 640 641 turbulence) to wind in the equatorial waveguide. Finally, it should be noted that for the SAS region (where the largest ~55% effect of wind stress intraseasonal variations is found), this 642 effect probably results from a combination of wind stirring and Ekman pumping during 643 Findlater jet intraseasonal fluctuations. 644

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4.2. Impact of intraseasonal MLD fluctuations on SST

647 As shown on the bottom panels of Figure 14, the maximum SST warming generally lags the maximum MLD shoaling by ~5 days. This suggests that MLD variations could 648 649 influence the SST variations although this influence would be larger if the SST and MLD were in quadrature. Several previous studies (e.g. Jayakumar et al. 2011; Vialard et al. 2012, 650 651 2013) have demonstrated the ability of a simple slab ocean model (i.e. fixed MLD) to reproduce intraseasonal SST fluctuations in the IO. We hence assess the impact of 652 intraseasonal MLD variations on intraseasonal SST fluctuations using such a slab ocean 653 described as follows: 654

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$$\partial_t T = \left[\frac{Q_s(1-f(-h))+Q^*}{\rho C_p h}\right]^{\prime}$$
(3)

where T is SST; Q_S the surface shortwave flux; Q* the sum of longwave, latent and sensible fluxes; the f (z) function describes the fraction of shortwave that penetrates down to the depth z following the double exponential rule corresponding to type I water in the Jerlov (1968) classification; h is the mixed-layer depth; and ' denotes intraseasonal filtering. We will quantify the importance of intraseasonal MLD fluctuations on SST by applying Eq. (3) for climatological (i.e. without intraseasonal variations) and time-varying MLD.

Figure 15a-e first allows a rough validation of the modelled intraseasonal SST signals

and the relevance of the slab ocean for modelling it. For all boxes, the model reproduces the 663 observed SST intraseasonal fluctuations well, despite a tendency for the model to 664 underestimate those fluctuations in the EEIO, SAS and NAS regions. The simple slab ocean 665 approach is generally in agreement with the REF experiment (Figure 15a-e), suggesting that 666 heat flux forcing dominates SST intraseasonal variations in these regions, consistent with past 667 studies (e.g. Jayakumar et al. 2011; Vialard et al. 2012, 2013). However, the slab ocean model 668 SST amplitude is larger than in REF experiment in the BoB. In this region, Nisha et al. (2013) 669 indicate that oceanic processes tend to damp SST intraseasonal fluctuations: mixed layer 670 671 cooling decreases the temperature vertical gradient, hence resulting in reduced cooling by vertical mixing. Despite this negative feedback from oceanic processes in the BoB, heat flux 672 673 forcing is the first order mechanism that drives SST fluctuations in all the considered regions, 674 justifying our slab ocean approach.

The impact of intraseasonal MLD fluctuations is illustrated on Figure 15f-j by 675 comparing slab ocean model SST computed using the actual (blue) and intraseasonal filtered 676 (green) model MLD. Neglecting intraseasonal MLD fluctuations has a minor influence on 677 678 SST fluctuations in all regions, suggesting that MLD intraseasonal variability does not significantly modulate SST intraseasonal variability (compare green and blue curves on 679 Figure 15f-j). The primary mechanism by which intraseasonal MLD variations can affect 680 intraseasonal SST variations is the modulation of the mixed layer thermal capacity (hereafter 681 "scaling effect"). However, in regions of shallow mixed layer such as the tropics, a 682 significant part of the incoming solar heat flux penetrates below the mixed layer and therefore 683 does not contribute to mixed layer heating. A small variation of the mixed layer depth can 684 change the amount of heat flux that is "lost" beneath the mixed layer quite significantly 685 (hereafter the "penetrative effect"), because of the exponential nature of shortwave 686 penetration into the ocean. Red curves on Figure 15f-j exhibit SST fluctuations derived from 687 the slab ocean model when not accounting for intraseasonal MLD variations when calculating 688 689 the solar penetration in (3): comparing red and blue curves on these panels allows quantifying the impact of the "penetrative effect" on the amplitude of intraseasonal SST fluctuations. 690 691 Similarly, brown curves on Figure 15f-j exhibit SST fluctuations derived from the slab ocean 692 model when not considering intraseasonal MLD fluctuations in the denominator of Eq. (3), 693 comparing brown and blue curves on these panels allows the impact of the "scaling effect" to be quantified. Not accounting for the penetrative effect results in overestimated SST 694 695 intraseasonal amplitude (from 20% in the BoB to 70% in the EEIO) for all regions. This is

because less incoming solar heat is trapped in the mixed layer when the MLD is shallower 696 than normal during the warming phase (and vice versa during the cooling phase), so the 697 penetrative effect damps the SST fluctuations. The impact of the "scaling effect" is more 698 subtle, as it depends on both the sign of the heat flux forcing and the amplitude of the mixed 699 layer depth in each phase. As noted by Shinoda and Hendon (1998) and Drushka et al. (2012), 700 negative net heat fluxes are associated with deep mixed layers (i.e., reduced cooling) and 701 positive heat fluxes with shallow mixed layers (i.e., enhanced warming), so that the scaling 702 effect nearly always induces a relative warming. Because the mixed layer is thinner during the 703 704 phase with positive heat fluxes, this warming effect is stronger compared to the phase with negative heat fluxes. As a result of this asymmetry, the "scaling effect" results in a SST 705 anomaly with a larger mean (which is filtered out when the intraseasonal variations are 706 707 extracted) and a larger amplitude. The overall impact of the scaling effect is therefore to 708 amplify SST fluctuations (blue against brown curves on Fig. 15f-j). The compensation between the "scaling" and "penetrative" effects at intraseasonal timescales therefore seems to 709 710 result in an overall weak impact of intraseasonal MLD fluctuations on intraseasonal SST variations. The slab model approach above is however very simple and its limitations will be 711 712 discussed in section 5.

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715 <u>5. Summary and Discussions</u>

716 <u>5.1. Summary</u>

The winter MJO and the active and break phases of the summer monsoon are the dominant modes of atmospheric intraseasonal variability in the IO. To date, there was no exhaustive study describing the intraseasonal MLD response to atmospheric intraseasonal variability over the IO. This paper hence aims at a better description of large-scale intraseasonal variability of MLD over the IO. Our study relies on the joint analysis of a dataset built from 2002–2013 Argo data and an eddy permitting (0.25°) regional ocean model, which reproduce observed intraseasonal MLD variations reasonably well.

During the summer monsoon, largest intraseasonal MLD signals are found in eastern equatorial Indian Ocean, Bay of Bengal and southern Arabian Sea. Active and break phases of the summer monsoon drive most of the MLD fluctuations in the eastern equatorial Indian Ocean and Bay of Bengal. During the break phase, enhanced convection south of India is associated with a MLD deepening in the eastern equatorial basin, while suppressed convection over the Bay of Bengal results in shallow MLD there. Intraseasonal MLD fluctuations in the southern Arabian Sea are relatively independent from MLD and atmospheric convection variability in the two previous regions. Intraseasonal MLD variations in the southern Arabian Sea are driven by fluctuations of the Findlater jet intensity.

During winter, strongest large-scale MLD variations occur in southeastern equatorial Indian Ocean and in the northern Arabian Sea, while MLD perturbations in Bay of Bengal are mostly small-scale and related to eddy variability. The MLD intraseasonal variability in the southeastern equatorial Indian Ocean is related to MJO forcing, with suppressed convection and light winds associated with shallow MLDs. The southward advection of continental air temperature anomalies induces intraseasonal air temperature fluctuations over the northern Arabian Sea, which drive intraseasonal convective MLD variations.

Buoyancy fluxes and friction velocity both contribute significantly to intraseasonal 740 MLD fluctuations in all regions, except in the northern Arabian Sea in winter, where 741 742 buoyancy flux forcing dominates and in the southern Arabian Sea in summer, where Ekman 743 pumping on the southern flank of the Findlatter jet also contributes. A slab ocean model analysis suggests that these intraseasonal MLD fluctuations have a weak impact on 744 745 intraseasonal SST signals in any of the regions (less than 10% of the amplitude). This weak response is largely explained by the compensation between the "scaling" (i.e. modulation of 746 747 mixed layer thermal capacity by MLD fluctuations that acts to enhance SST variations) and "penetrative" effects (i.e. modulation of the amount of incoming solar heat flux lost through 748 the base of the mixed layer that has the opposite impact) in our simple framework. 749

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4.2. Discussion and perspectives

To our knowledge, Drushka et al (2012, 2014) are to date the only observational 752 studies that have described intraseasonal MLD variations in the Indian Ocean, focussing on 753 754 the MLD response to the MJO in winter in the central and eastern equatorial part of the basin. For this particular region and season, our results echo the observational analysis of Drushka et 755 756 al (2012, 2014), with more than 10 m peak-to-peak fluctuations in this region. The present study expands this description of the intraseasonal MLD variability for the entire Indian 757 Ocean and for both winter and summer seasons, complementing the analysis of in-situ data 758 with an oceanic simulation. Although model and observationally-derived MLD intraseasonal 759

composites exhibit consistent patterns in all the regions of strong intraseasonal variability, the limited density of Argo data did not allow providing a complete mapping of these intraseasonal anomalies. In addition, the use of a composite analysis to extract meaningful MLD variations from observations does not allow monitoring the large event-to-event variability. Future studies with other models and a longer Argo dataset will be needed to ascertain the MLD patterns and amplitudes presented here.

The main goal of this study was to explore MLD variations and their causes in the 766 tropical Indian Ocean. Our results however raise two interesting questions regarding 767 atmospheric variability in the Indian Ocean. Intraseasonal variations in most regions are 768 linked with well-known modes of atmospheric intraseasonal variability: the MJO in winter 769 and active/break phases of the Indian monsoon during summer. On the other hand, MLD 770 intraseasonal variations in the Arabian Sea cannot be clearly connected with a known mode of 771 intraseasonal atmospheric variability. In summer, MLD variations in the southern Arabian 772 773 Sea are driven by intraseasonal fluctuations of the Findlater jet intensity. Joseph and Sijikumar (2004) already noted changes in the monsoon jet position and intensity linked to 774 775 active and break phases of the monsoon, which can be seen on Figure 5a-d. In contrast, the wind variations that drive intraseasonal MLD variations in the southern Arabian Sea are 776 upstream (Figure 7a-d), and are independent from monsoon active/break phases and 777 convection over the BoB and appear to be more driven by convective variability southwest of 778 779 India (Figure 7b and 7d). A more thorough study is needed to assess if this corresponds to a different "flavour" of active/break phase with main convective perturbations over the AS 780 781 rather than over the BoB, and how the dynamics of this intraseasonal "mode" compare with the more standard active/break phases. Similarly, the occurrence of intraseasonal temperature 782 perturbations over the northern Arabian Sea has to be investigated in more detail. Our results 783 suggest that they are associated with southward advection of continental temperature 784 anomalies by northerly winds. However, the exact nature, process and dominant timescale (if 785 786 any) of this phenomenon has yet to be understood.

Regarding the potential impact of MLD intraseasonal fluctuations, our slab ocean model results are in line with those of Jayakumar et al. (2011) and Vialard et al. (2012, 2013), which suggested a rather weak influence of intraseasonal MLD fluctuations on intraseasonal SST variations. Our analysis suggests that the "scaling" and "penetrative" effects tend to cancel each other, explaining the overall weak effect of MLD variations on SST. This result apparently contradict those of Drushka et al (2012), which suggests that intraseasonal MLD

fluctuations may reduce the amplitude of the SST signal during the MJO active phase in the 793 southeastern equatorial Indian Ocean. These differences may well be explained by the very 794 795 localised SST impact discussed in Drushka et al. (2012), which may be wiped out when averaging over a large region as in the present paper. Alternatively, our results could also be 796 797 hampered by methodological caveats. First, our assessment using a simple slab ocean model does not account for oceanic processes (lateral advection, entrainment, upwelling). In 798 addition, modelled intraseasonal SST and MLD estimates may suffer from errors inherent to 799 800 the forcing dataset.

In addition to these methodological caveats, other processes that we did not consider 801 802 may also influence intraseasonal SST fluctuations. We did not attempt for instance to isolate the contribution of internal oceanic instabilities (e.g. eddies) on large-scale intraseasonal 803 fluctuations. This internally driven variability may indeed constructively/destructively 804 805 interacts with the intraseasonal variability forced by the large-scale climate modes such as the MJO or the monsoon active/break phases. Jochum and Murtugudde (2005) indeed showed 806 that these small-scale features could significantly contribute to the large-scale SST variations 807 808 in specific regions of the Indian Ocean. The very intense eddy variability off the Somalia upwelling region may in particular contribute to large-scale upper ocean variations there, as 809 demonstrated by Jochum and Murtuggude (2005). Intraseasonal chlorophyll fluctuations may 810 also alter the vertical profile of solar penetration and hencefore the SST variability. A crude 811 812 estimate of this effect by including surface chlorophyll variations derived from the satellite data and their impact on the SST of our slab ocean model however suggests that the average 813 814 impact is very small, although it can be significant for peculiar events (not shown). Finally, scale interaction mechanisms such as the potential rectification of intraseasonal variability 815 onto lower frequency suggested by Waliser et al. (2003, 2004) or the potential influence of 816 diurnal cycle onto longer timescales as suggested by Wiggert et al. (2002) may also operate. 817 Addressing such issues would require a more idealized model setup similar to the one used in 818 819 Waliser et al. (2003, 2004) and a properly resolved diurnal cycle, which is lacking in the present model configuration. 820

Mixed layer depth variability is not only crucial for air-sea interactions and climate but also from a biogeochemical perspective. Mixed layer entrainment and thickness are important determinants of the nutrient flux into the euphotic zone and average light intensity experienced by phytoplankton (McCreary et al. 2001). In the Bay of Bengal, there is a strong coupling between the seasonal cycle of mixed layer depth and the processes that affect upper

ocean chlorophyll pigment concentrations (Narvekar, 2013). The mixed layer in the central 826 Arabian Sea deepens considerably during both monsoons seasons (McCreary et al. 2001; 827 Wiggert et al. 2005). This gives rise to competing mechanisms that can either lead to a 828 phytoplankton biomass increase or decrease. On the one hand, nutrient concentration 829 increases due to entrainment and grazing-pressure decreases because of a vertically wider 830 habitat, but on the other hand light-limitation increases because of less time spent in the 831 euphotic layer (Levy et al. 2007). Although there have been significant advances in our ability 832 to describe and model the oceanic biogeochemistry in the Indian Ocean, the biogeochemical 833 834 impact of MLD variations in response to climate variability at intraseasonal timescales in the IO remain largely unknown. Only a handful of studies have examined the ocean ecosystem 835 response to the MJO (e.g. Waliser et al. 2005; Resplandy et al. 2009; Jin et al. 2012). While 836 the MJO drives large intraseasonal chlorophyll signals in the southern IO and in the Bay of 837 838 Bengal, there are strong intraseasonal chlorophyll fluctuations in the Arabian Sea that only seem to be marginally related to the MJO and whose driving processes remain unclear (Jin et 839 840 al. 2012). In future, we will use a combination of observations and modelling to investigate the impacts of MLD variations on chlorophyll and primary production in the Arabian Sea. 841

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852 *References:*

- Babu KN, Sharma R, Agarwal N, Agarwal VK, Weller RA (2004) Study of the mixed layer
 depth variations within the north Indian Ocean using a 1-D model, J Geophys Res,
 109:1-9.
- Barnier B, Madec G, Penduff T, Moilnes JM, Treguier AM, Sommer Le, Beckmann A,
 Biastoch A, Boning C, Deng J, Derval C, Durand E, Gulev S, Remy E, Talandier C,
 Theetten S, Maltrud M, McClean J, De Cuevas B (2006) Impact of partial steps and
 momentum advection schemes in a global ocean circulation model at eddy permitting
 resolution, Ocean Dyn, 56:543–567.
- Behera SK, Yamagata T (2001) Subtropical SST dipole events in the southern Indian Ocean,
 Geophys Res Lett, 28(2): 327-330,doi: 10.1029/2000GL011451.
- Behrenfeld, Michael J (2010) Abandoning Sverdrup's critical depth hypothesis on
 phytoplankton blooms, Ecology, 91(4), 977-989.
- Bellenger H, J Duvel (2009) An analysis of tropical ocean diurnal warm layers, J Climate, 22,
 3629–3646.
- Bellon G, AH Sobel, J Vialard (2008) Ocean-atmosphere coupling in the monsoon
 intraseasonal oscillation: a simple model study, J Climate, 21, 5254-5270.
- Bentamy A, Katsaros KB, Alberto M, Drennan WM, Forde EB, Roquest H (2003) Satellite
 estimates of wind speed and latent heat flux over global oceans, J climate, 16:637-656
- Bessafi, Miloud, MC. Wheeler (2006) Modulation of South Indian Ocean tropical cyclones
 by the Madden-Julian Oscillation and convectively coupled equatorial waves. Monthly
 Weather Review, 134.2, 638-656.
- Brandt P, Dengler M, Rubino A, Quadfasel D, Schott F (2003) Intraseasonal variability in the
 southwestern Arabian Sea and its relation to the seasonal circulation. Deep Sea Res II
 50:2129–2142
- Brodeau L, Barnier B, Treguier AM, Penduff T, Gulev S (2010) An ERA 40-based
 atmospheric forcing for global ocean circulation models. Science Direct, 31:88-104,
 doi: 10.1016/j.ocemod.2009.10.005.

- de Boyer Montegut C, Madec G, Fischer AS, Lazar A, Iudicone D (2004) Mixed layer depth
 over the global ocean : an examination of profile data and a profile-based
 climatology. J Geophys Res, 109, C12003, doi: 10.1029/2004JC002378
- de Boyer Montegut C, Vialard J, Shenoi SSC, Shankar D, Durand F, Ethé C, Madec G
 (2007) Simulated seasonal and interannual variability of mixed layer heat budget in
 the northern Indian Ocean, J Climate, 20:3249–3268.
- Dee DP, Uppala SM, Simmons AJ, Berrisford P, Poli P, Kobayashi S, Andrae et al (2011)
 The ERA-Interim reanalysis: Configuration and performance of the data assimilation
 system. Quarterly Journal of the Royal Meteorological Society, 137(656), 553-597.
- de Laat ATJ, J Lelieveld (2002) Interannual variability of the Indian winter monsoon
 circulation and consequences for pollution levels, J. Geophys. Res., 107(D24), 4739,
 doi: 10.1029/2001JD001483.
- Dickey T, J Marra, DE Sigurdson, RA Weller, CS Kinkade, SE Zedler, JD Wiggert, C
 Langdon (1998) Seasonal variability of bio-optical and physical properties in the
 Arabian Sea: October 1994–October 1995, Deep Sea Research Part II: Topical Studies
 in Oceanography, 45.10 (1998), 2001-2025.
- Biogeochemistry of the North Indian Ocean; IGBP-WCRP-SCOPE
 Rep. Ser.1, Indian Nat.Sci. Acad., New Delhi, India.
- B98 Drakkar Group (2007) Eddy-permitting Ocean circulation hindcasts of past decades. Clivar
 B99 Exchanges 12(3): 8–10, No 42.
- Drushka K, Sprintall J, Gill ST (2012) In situ observations of Madden-Julian Oscillation
 mixed layer dynamics in the Indian and western Pacific Oceans, J Climate, 25.7, 23062328.
- Drushka, K., J. Sprintall, and S. T. Gille (2014). Subseasonal variations in salinity and
 barrier-layer thickness in the eastern equatorial Indian Ocean. J. Geophys. Res. 119,
 805–823, doi: 10.1002/2013JC009422.
- Duncan B, W Han (2009) Indian Ocean intraseasonal sea surface temperature variability
 during boreal summer: Madden– Julian Oscillation versus submonthly forcing and
 processes. J. Geophys. Res., 114, C05002, doi: 10.1029/2008JC004958.

29

- Dussin R, Treguier A-M, Molines JM, Barnier B, Penduff T, Brodeau L, Madec G (2009)
 Definition of the interannual experiment ORCA025-B83, 1958–2007, LPO Report
 902.
- Duvel J-P, Roca R, Vialard J (2004) Ocean mixed layer temperature variations induced by
 intraseasonal convective perturbations over the Indian Ocean, J. Atm. Sciences
 61:1004–1023.
- Duvel J, J Vialard (2007) Indo-Pacific sea surface temperature perturbations associated with
 intraseasonal oscillations of tropical convection. J. Climate, 20, 3056–3082.
- Findlater J (1969) A major low level air current over the Indian Ocean during the northern
 summer, Quart J Roy Meteor Soc, 95:362–380.
- Foltz GR, Vialard J, Praveen Kumar B, McPhaden MJ (2010) Seasonal mixed layer heat
 balance of the southwestern tropical Indian Ocean, J. Climate, 23: 947-965.
- Gadgil S, PV Joseph, NV Joshi (1984) Ocean atmosphere coupling over monsoon regions,
 Nature, 312, 141145.
- Gadgil S (2003) The Indian Monsoon and its variability, Annu Rev Earth Planet Sci, 31:429–
 467.
- Gill AE (1982) Atmosphere-Ocean Dynamics, Academic Press, 662 pp.
- Gopalakrishna VV, Sadhuram Y, Ramesh Babu V (1988) Variability of mixed layer depth in
 the northern Indian Ocean during 1977 and 1979 summer monsoon seasons, IJMS, 17:
 258-264..
- Goswami BN (2005) South Asian Monsoon. In: Lau WKM, Waliser DE (eds) Intraseasonal
 variability in the atmosphere-ocean climate system. Praxis Springer, Berlin, pp 19–55.
- Harrison DE, Vecchi GA (2001) January 1999 Indian Ocean cooling event. Geophys Res Lett
 28: 3717–3720.
- Huffman GJ, and co-authors (1997) The Global Precipitation Climatology Project (GPCP)
 combined data set. Bull. Amer. Meteor. Soc., 78, 5-20.
- Ingram KT, Roncoli MC, Kirshen PH (2002) Opportunities and constraints for farmers of
 west Africa to use seasonal precipitation forecasts with Burkina Faso as a case study,

- 937 Agr Syst 74, 331–349.
- Jayakumar A, Vialard J, Lengaigne M, Gnanaseelan C, McCreary JP, Praveen Kumar B
 (2011) Processes controlling the surface temperature signature of the Madden–Julian
 Oscillation in the thermocline ridge of the Indian Ocean, Clim Dyn, 37, 2217-2234.
- 941 Jerlov NG (1968) Marine Optics, Elsevier Oceanography Series, 5.
- Jin D, DE Waliser, C Jones, R Murtugudde (2012) Modulation of tropical ocean surface
 chlorophyll by the Madden–Julian Oscillation, Clim Dyn, 40.1-2, 39-58.
- Jochum M, Murtugudde R (2005) Internal variability of Indian ocean SST, J.Climate, 18(18),
 3726-3738.
- Joseph PV, Sijikumar S (2004) Intraseasonal variability of the low level jet stream of the
 Asian summer monsoon, J.Climate, 17, 1449–1458.
- Keerthi MG, Lengaigne M, Vialard J, de Boyer MC, Muraleedharan PM (2013) Interannual
 variability of the Tropical Indian Ocean mixed layer depth, Clim Dyn, 40, 743–759.
- Kikuchi K, B Wang, Y Kajikawa (2012) Bimodal representation of the tropical intraseasonal
 oscillation, Clim Dyn, 38.9-10, 1989-2000.
- Klein SA, BJ Soden, NC Lau (1999) Remote sea surface temperature variations during
 ENSO: Evidence for a tropical atmospheric bridge, J Climate, 12: 917–932.
- Koné V, O Aumont, M Lévy, L Resplandy (2009), Physical and biogeochemical controls of
 the phytoplankton seasonal cycle in the Indian Ocean: A modeling study, in Indian
 Ocean Biogeochemical Processes and Ecological Variability, Geophys. Monogr. Ser.,
 185, edited by J. D. Wiggert et al., pp. 147–166, AGU, Washington, D. C.
- Large WG, Yeager SG (2004) Diurnal to decadal global forcing for ocean and sea-ice models:
 the datasets and flux climatologies, Techical report TN-460+STR, NCAR, 105 pp.
- Lee CM, BH Jones, KH Brink, AS Fischer (2000), The upper ocean response to monsoonal
 forcing in the Arabian Sea: Seasonal and spatial variability, Deep Sea Res., Part II, 47,
 1177–1226.
- Liebmann B, Smith CA (1996) Description of a complete (interpolated) outgoing longwave
 radiation dataset, Bull. Amer. Meteor. Soc. 77:1275-1277.

- Lengaigne M, Hausmann U, Madec G, Menkes C, Vialard J, Molines JM (2012) Mechanisms
 controlling warm water volume interannual variations in the equatorial Pacific:
 diabatic versus adiabatic processes. Clim Dyn 38:1031–1046.
- Levy M, D Shankar, JM Andre, SSC Shenoi, F Durand, C de Boyer Montegut (2007)
 Basinwide seasonal evolution of the Indian Ocean's phytoplankton blooms, J.
 Geophys. Res., 112, C12014.
- P71 Locarnini RA, Mishonov AV, Antonov JI, Boyer TP, Garcia HE, Baranova OK, Zweng MM,
 P72 Johnson DR (2010) World Ocean Atlas 2009, vol 1: Temperature. S. Levitus (ed)
 P73 NOAA Atlas NESDIS 68. U.S. Government Printing Office, Washington, D.C.
- Madec G (2008) NEMO, the Ocean Engine. Tech. Rep, Notes de l'IPSL (27), ISSN 12881619, Université P. et M. Curie, B102 T15-E5, 4 Place Jussieu, Paris Cedex 5, p. 193
- Maloney E, A Sobel (2004) Surface fluxes and ocean coupling in the tropical intraseasonal
 oscillation. J. Climate, 17, 3717–3720.
- Matthews AJ (2004) The atmospheric response to observed intraseasonal tropical sea surface
 temperature anomalies, Geophys Res Lett, 31, L14107
- McCreary JP, Kohler KE, Hood RR, Smith S, Kindle J, Fischer AS, Weller RA (2001)
 Influences of diurnal and intraseasonal forcing on mixed layer and biological
 variability in the central Arabian Sea, J. Geophys. Res Oceans, 106(C4): 71397155.
- McCreary JP, Kundu PK (1989) A numerical investigation of sea surface temperature
 variability in the Arabian Sea, J Geophys Res, 94:16 097–16 114.
- McWilliams JC, Sullivan PP, Moeng CH (1997) Langmuir turbulence in the ocean, J. Fluid
 Mech. 334, 1–30
- Murtugudde R, R Seager, P Thoppil (2007) Arabian Sea response to monsoon variations,
 Paleoceanography, 22, PA4217, doi:10.1029/2007PA001467.
- Narvekar J, Prasanna Kumar S (2006) Seasonal variability of the mixed layer in the central
 Bay of Bengal and associated changes in nutrients and chlorophyll, Deep-Sea Res I,
 53: 820- 835.

- Narvekar J, Prasanna Kumar S (2013) Mixed layer variability and chlorophyll a biomass in
 the Bay of Bengal, Biogeosciences Discuss, 10, 16405-16452.
- Nidheesh AG, Lengaigne M, Vialard J, Unnikrishnan AS, Dayan H (2013) Decadal and longterm sea level variability in the tropical Indo-Pacific Ocean, Clim Dyn, 41(2), 381402.
- Niiler PP, EB Kraus (1977) One-dimensional models of the upper ocean. Modelling and
 Prediction of the Upper Layer of the Ocean, E. B. Kraus, Ed. Pergamon press, Oxford,
 143-172.
- Nisha K, M Lengaigne, VV Gopalakrishna, J Vialard, S Pous, A-C Peter, F Durand, S Naik
 (2013) Processes of summer intraseasonal sea surface temperature variability along
 the coasts of India, Ocean Dyn, 63, 329-346.
- Prasad TG (2004) A comparison of mixed-layer dynamics between the Arabian Sea and Bay
 of Bengal: one-dimensional model results, J Geophys Res, 109, C03035.
- Prasanna Kumar S, J Narvekar (2005) Seasonal variability of mixed layer in the central
 Arabian Sea and its implication to nutrients and primary productivity, Deep-Sea Res.
 II, 52, 1848-1861.
- Rao RR, Molinari RL, Festa JF (1989) Evolution of the climatological near-surface thermal
 structure of the tropical Indian Ocean, J Geophys Res, 94: 1081- 10815.
- 1010 Rao RR, Sivakumar R (2003) Seasonal variability of sea surface salinity and salt budget of
 1011 the mixed layer of the north Indian Ocean, J Geophys Res, 108, 3009.
- 1012 Resplandy L, J Vialard, M Lévy, O Aumont, Y Dandonneau (2009) Seasonal and
 1013 intraseasonal biogeochemical variability in the thermocline ridge of the southern
 1014 tropical Indian Ocean, J. Geophys. Res., 114, C07024.
- Roemmich D, J Gilson (2009) The 2004-2008 mean and annual cycle of temperature, salinity,
 and steric height in the global ocean from the Argo program. Prog. Oceanogr., 82 (2),
 81–100.
- Saji NH, Goswami BN, Vinaychandran PN, Yamagata T (1999) A dipole mode in the tropical
 Indian Ocean, Nature, 401, 360-363.

- Schott FA, Xie SP, McCreary Jr JP (2009) Indian Ocean circulation and climate variability,
 Rev of Geophys, 47, RG1002/2009.
- Sengupta D, Goswami BN, Senan R (2001) Coherent intraseasonal oscillations of ocean and
 atmosphere during the Asian summer monsoon, Geophys Res Lett, 28, 4127–4130.
- Shenoi SSC, Shankar D, Shetye SR (2002) Differences in heat budgets of the near-surface
 Arabian Sea and Bay of Bengal: Implications for the summer monsoon, J Geophys
 Res, 107, doi 10.1029/2000 JC000679.
- Shinoda T, Hendon HH (1998) Mixed layer modelling of intraseasonal variability in the
 tropical western Pacific and Indian Oceans, J Climate, 11, 2668–2685.
- Smith S, M Madhupratap (2005) Mesozooplankton of the Arabian Sea: patterns influenced by
 seasons, upwelling, and oxygen concentrations, Progress in Oceanography, 65, 2(4),
 214-239.
- Sreenivas P, Patnaik KVKRK, Prasad KVSR (2008) Monthly Variability of Mixed Layer
 over Arabian Sea Using ARGO Data, Marine Geodesy, 31, Issue 1, doi:
 1034 1080/1490410701812311.
- Sverdrup HU (1953) On conditions for the vernal blooming of phytoplankton, J. Cons. Cons.
 Int. Explor. Mer, 18, 287-295.
- Treguier A-M, Barnier B, De Miranda AP, Molines JM, Grima N, Imbard M, Madec G,
 Messager C, Reynaud T, Michel S (2001) An eddy-permitting model of the Atlantic
 circulation: evaluating open boundary conditions. J Geophys Res 106:22,115–22,129.
- 1040 Uppala SM et al (2005) The ERA-40 re-analysis, Q J R Meteorol Soc, 131:2961–3012.
- 1041 Vecchi GA, Harrison DE (2002) Monsoon breaks and subseasonal sea surface temperature
 1042 variability in the Bay of Bengal. J Climate, 15, 1485–1493.
- 1043 Vialard J, Foltz G, McPhaden M, Duvel J-P, de Boyer Montegut C (2008) Strong Indian
 1044 Ocean sea surface temperature signals associated with the Madden–Julian Oscillation
 1045 in late 2007 and early 2008, Geophys Res Lett, 35:L19608.
- 1046 Vialard J, Jayakumar A, Gnanaseelan C, Lengaigne M, Sengupta D, Goswami BN (2012)
 1047 Processes of 30–90 day sea surface temperature variability in the Northern Indian

- 1048 Ocean during boreal summer, Clim Dyn, 38, 1901–1916.
- Vialard J, Drushka K, Bellenger H, Lengaigne M, Pous S, Duvel JP (2013) Understanding
 Madden-Julian induced sea surface temperature variations in the North Western
 Australian basin, Clim Dyn, 41(11-12), 3203-3218.
- Waliser DE, Murtugudde R, Lucas LE (2004) Indo-Pacific Ocean response to atmospheric
 intraseasonal variability: 2. Boreal summer and the intraseasonal oscillation. J
 Geophys Res, 109, C03030.
- Waliser DE, Murtugudde R, Strutton P, Li JL (2005) Subseasonal organization of ocean
 chlorophyll: Prospects for prediction based on the Madden-Julian oscillation, Geophys
 Res Lett, 32(23).
- Waliser D, R Murtugudde, L Lucas (2003) Indo-Pacific ocean response to atmospheric
 intraseasonal variability: 1. Austral summer and the Madden-Julian Oscillation. J.
 Geo-phys. Res., 108, 3160, doi:10.1029/2002JC001620.
- Weaver A, Courtier P (2001) Correlation modelling on the sphere using a generalized
 diffusion equation, Quart J Roy Meteor Soc, 127:1815-1846.
- 1063 Webster PJ, C Hoyos (2004) Prediction of Monsoon Rainfall and River Discharge on 15-30
 1064 day Time Scales, Bull. Amer. Met. Soc., 85 (11), 1745-1765.
- Weller RA, JF Price (1988) Langmuir circulation within the oceanic mixed layer, Deep-Sea
 Res., 35, 711–747.
- Wheeler M, H Hendon (2004) An all-season real-time multivariate MJO index: Development
 of an index for monitoring and prediction, Mon. Wea. Rev., 132, 1917–1932.
- Wiggert JD, R Murtugudde, CR McClain (2002) Processes controlling interannual variations
 in wintertime (northeast monsoon) primary productivity in the central Arabian Sea,
 Deep Sea Res., Part II, 47, 2319–2343
- 1072 Wiggert JD, RR Hood, K Banse, JC Kindle (2005), Monsoon driven biogeochemical
 1073 processes in the Arabian Sea, Prog. Oceanogr., 65, 176–213.
- 1074 Xie P, PA Arkin (1997) Global precipitation: a 17-year monthly analysis based on gauge 1075 observations, satellite estimates, and numerical model outputs. Bull. Amer. Meteor.

Soc., 78, 2539-2558.

1077	Zhang	С	(2005)	Madden-Julian	Oscillation,	Rev	Geophys,	43,	RG2003,	doi:
1078		10.1	029/2004	RG000158						

Zhang Y, Rossow WB, Lacis AA, Oinas V, Mishchenko MI (2004) Calculation of radiative
fluxes from the surface to top of atmosphere based on ISCCP and other global data
sets: refinments of the radiative trasfer model and the input data, J Geophys Res
1082 109:D19105. doi: 10.1029/2003JD00445

1100 Figure Captions:

Figure 1: Standard deviation of MLD variations (in meters) in the Indian Ocean at (a)
seasonal, (b) interannual and (c) intraseasonal timescales from a ¹/₄° simulation
provided by the DRAKKAR project detailed in Keerthi et al. (2013). Contours on
panel b (resp. panel c) show the ratio of interannual (resp. intraseasonal) against
seasonal MLD standard deviation. This figure is adapted from Keerthi et al. (2013).

1106

Figure 2: May to September 2000 time series of (a) intraseasonal summer monsoon 1107 active/break index (Goswami et al. 2005) detailed in section 2.4, (b) averaged 1108 intraseasonal TMI SST (in °C) and (c) averaged intraseasonal modelled MLD (in m) 1109 over the Bay of Bengal (10°N-20°N; 80°E-100°E). December 1998 to April 1999 time 1110 series of (d) intraseasonal MJO index (RMM1 from Wheeler and Hendon 2004) 1111 1112 detailed in section 2.4, (e) averaged intraseasonal TMI SST (in °C) and (f) averaged intraseasonal modelled MLD (in m) over the thermocline ridge of the Indian Ocean 1113 (2°S-10°S; 60°E-90°E). The climatological seasonal SST and MLD depth are 1114 indicated on the bottom left corner of the corresponding panels. Details on the model 1115 simulation from which intraseasonal MLD are calculated are provided in Section 2.1 1116 while intraseasonal monsoon and MJO indices and TMI satellite SST data are detailed 1117 in Section 2.2. An illustration of the phase definition for monsoon and MJO indices is 1118 also provided on the top panels. 1119

1120

Figure 3: (a) Observed summer climatological MLD (color) from de Boyer Montegut et al
2004 climatology and wind stress (arrows) from Tropflux and (b) Modeled summer
(JJAS) climatological MLD (color) and wind stress (arrows). (c) Summer standard
deviation of MLD (color) and large-scale MLD (contour) intraseasonal variations. The
black boxes indicate regions of maximum large-scale MLD variability, whose
boundaries are provided in table 1.

1127

Figure 4: Percentage of summer intraseasonal modelled MLD variance explained by (a) the
Monsoon index (Goswami 2005), (b) the JET index (zonal wind averaged over the

[55°E-75°E; 2.5°N-12.5°N] box) and (c) the two previous indices, collectively. (d-f)
Same but for intraseasonal OLR variance. Contours on panel c and f display the
standard deviation of summer intraseasonal large-scale MLD and OLR, respectively.
Black boxes indicate the three regions of largest intraseasonal MLD variations, whose
boundaries are provided in table 1.

1135

Figure 5: Composites of the phases 1 to 4 of the intraseasonal summer monsoon index from Goswami (2005) for (top) OLR (color) and winds (arrow), (Middle) large-scale model MLD (color) overlaid with large-scale model wind stress intensity anomalies (contours in 10⁻² N.m⁻²), (Bottom) Argo MLD. Regions where composite values are less than the standard error are displayed in white. Phases 5 and 6 are almost exactly the opposite of Phases 2 and 3 and are therefore not shown.

1142

Figure 6: Box-averaged composite evolution of model (black) and Argo (green) intraseasonal
MLD anomalies for the six phases of the intraseasonal summer monsoon index in the
(a) BoB and (b) EEIO boxes. (c) Same but for MLD composites anomalies based on
the Findlater Jet index in the SAS box. The error bars represent the standard error.

1147

Figure 7: Composites of phases 1-4 of the intraseasonal Findlater "jet index" of (top) large-scale OLR (color) and winds (arrow), (middle) large-scale model MLD (color) overlaid with large-scale model wind stress intensity anomalies (contours in 10⁻² N.m⁻¹) and large scale model wind stress curl (blue contours in 10⁻⁷ N.m⁻³), (bottom) Argo MLD (color). Regions where composite values are less than the standard error are displayed in white. Phases 5 and 6 are almost exactly the opposite of Phases 2 and 3 and are therefore not shown.

1155

Figure 8: (a) Observed summer climatological MLD (color) from de Boyer Montegut et al
 2004 climatology and wind stress (arrows) from Tropflux and (b) Modeled winter
 (DJFM) climatological MLD (color) and wind stress (arrows). (c) Winter standard
 deviation of MLD (color) and large-scale MLD (contour) intraseasonal variations. The

1160 1161 black boxes indicate regions of winter maximum large-scale MLD variability, whose boundaries are provided in table 1

1162

Figure 9: Percentage of winter intraseasonal modelled MLD variance explained by (a) the MJO index (b) the NAS temperature index (2-m air temperature averaged over the [55°E-75°E; 15°N-25°N] box) and (c) the two previous indices. (e-f) Same but for intraseasonal OLR variance. Contours on panel (c) and (f) display the standard deviation of winter intraseasonal large-scale MLD and OLR respectively. Black boxes indicate the boxes used for calculating the winter indices.

1169

Figure 10: Composites of phases 1-4 of the Wheeler and Hendon (2004) MJO index for (top)
 large-scale OLR (color) and winds (arrow), (Middle) large-scale model MLD (color)
 overlaid with large-scale model wind stress intensity anomalies (contours in 10⁻² N.m⁻
 (Bottom) Argo MLD (color). Regions where composite values are less than the
 standard error are displayed in white. Phases 5-8 are almost exactly the opposite of
 phases 1-4 and are therefore not shown.

1176

Figure 11: Box-averaged composite evolution of model (black) and Argo (green)
intraseasonal MLD anomalies for (a) the eight phases of the Wheeler and Hendon
(2004) index in the SEEIO box and (b) the six phases of the NAS temperature index in
the NAS box. The error bars represent the standard error.

1181

Figure 12: (a) Climatological map of DJFM 2m air temperature (color) and wind (arrows).
(b) Lag-regression of intraseasonal air temperature anomalies zonally averaged
between 55°E and 75°E onto NAS air-temperature index. This box is marked on panel
(a). The land-sea limit is marked as thick black line in panel (a).

1186

Figure 13: Composites of phases 1-4 of the intraseasonal NAS winter air temperature index
for (top) large-scale near surface air temperature (color) and winds (arrow), (Middle)

1189 large-scale model MLD (color) overlaid with large-scale model wind stress intensity 1190 anomalies (contours in 10^{-2} N.m⁻²), **(Bottom)** Argo MLD (color). Regions where 1191 composite values are less than the standard error are displayed in white. Phases 5-8 are 1192 almost exactly the opposite of phases 1-4 and are therefore not shown.

1193

1194 Figure 14: Lag regression onto the relevent local climate modes of intraseasonal variations of (top) OLR, wind stress module and wind stress curl (only for panel c), (middle) 1195 frictional velocity and buoyancy fluxes including the solar and non-solar heat flux 1196 component from REF experiment and (bottom) MLD and SST from REF experiment 1197 and MLD from NOWIND experiments in the (a, f, k) BoB (Monsoon index), (b, g, l) 1198 EEIO (Monsoon index), (c, h, m) SAS (Jet index), (d, i, n) SEEIO (MJO index) and 1199 (e, j, o) NAS (NAS index). The regression coefficient of the NOWIND on to the REF 1200 is indicated on the upper right corner of each panel in (k-o). The correlation between 1201 frictional velocity and buoyancy fluxes intraseasonal variations is indicated on the 1202 upper right corner of each panel in (f-j). 1203

1204

Figure 15: Lag regression of SST onto the relevent local climate modes: (a, f) BoB
(Monsoon index), (b, g) EEIO (Monsoon index), (c, h) SAS (Jet index), (d, i) SEEIO
(MJO index) and (e, j) NAS (NAS index). Top panels show SST from TMI (black),
REF (purple) and slab ocean model (blue). Bottom panels are for slab ocean model
SST (blue) and slab ocean model SSTrecalculated neglecting the impact of
intraseasonal MLD variations on the "scaling" effect (brown), the "penetrative" effect
(red) or both (green). See text for details.

1212

1213 **Table Captions:**

Table 1 : Regions of strong large-scale MLD intraseasonal signals in the Indian Ocean.

1215

1216

1217









ACTIVE MONSOON ONSET COMPOSITE





FINDLATER JET INTENSIFICATION COMPOSITE







MJO PHASE COMPOSITE







NAS INTRASEASONAL EVENT COMPOSITE



-7 -5 -3 -1 1 3 5 7





Acronym	Name	Season	Boundaries
BoB	Bay of Bengal	Summer	[10°N-20°N; 80°E-100°E]
EEIO	eastern equatorial Indian Ocean	Summer	[5°S-5°N; 70°N-95°N]
SAS	southern Arabian Sea	Summer	[55°E-75°E; 2.5°N-12.5°N]
SEEIO	southeastern equatorial Indian Ocean	Winter	[8°S-2°N ; 85°E-105°E]
NAS	northern Arabian Sea	Winter	[15°N-25°N ; 55°E-75°E]