# Lagrangian versus Eulerian spectral estimates of surface kinetic energy over the global ocean

Zhang Xinwen<sup>1</sup>, Xiaolong Yu<sup>1</sup>, Aurélien L Ponte<sup>2</sup>, Zoé Caspar-Cohen<sup>2,3</sup>, Sylvie Le Gentil<sup>2</sup>, Lu Wang<sup>2</sup>, and Wenping Gong<sup>1</sup>

<sup>1</sup>School of Marine Sciences, Sun Yat-sen University, Southern Marine Science and Engineering Guangdong Laboratory (Zhuhai)
<sup>2</sup>Ifremer, Laboratoire d'Océanographie Physique et Spatiale, Université de Brest, CNRS, IUEM
<sup>3</sup>Scripps Institution of Oceanography, University of California

April 11, 2024

#### Abstract

In this study, we carried out a novel massive Lagrangian simulation experiment derived from a global 1/48° tide-resolving numerical simulation of the ocean circulation. This first-time twin experiment enables a comparison between Eulerian (fixed-point) and Lagrangian (along-flow) estimates of kinetic energy (KE), and the quantification of systematic differences between both types of estimations. This comparison represents an important step forward for the mapping of upper ocean high-frequency variability from drifter database. Eulerian KE rotary frequency spectra and band-integrated energy levels (e.g., tidal and near-inertial) are considered as references, and compared to Lagrangian estimates. Our analysis reveals that, apart from the near-inertial band, Lagrangian spectra are systematically smoother, e.g., with wider and lower spectral peaks compared to Eulerian counterparts. Consequently, Lagrangian KE levels obtained from spectra band integrations tend to underestimate Eulerian levels on average at low-frequency and tidal bands. This underestimation is more significant in regions characterized by large low-frequency KE. In contrast, Lagrangian and Eulerian near-inertial spectra and energy levels are comparable. Further, better agreements between Lagrangian and Eulerian KE levels are generally found in regions of convergent surface circulation, where Lagrangian particles tend to accumulate. Our results demonstrate that Lagrangian estimates may provide a distorted view of high-frequency variance. To accurately map near-surface velocity climatology at high frequencies (e.g., tidal and near-inertial) from Lagrangian observations of the Global Drifter Program, conversion methods accounting for the Lagrangian bias need to be developed.

# Lagrangian versus Eulerian spectral estimates of surface kinetic energy over the global ocean

# Xinwen Zhang<sup>1</sup>, Xiaolong Yu<sup>1</sup>, Aurélien L. Ponte<sup>2</sup>, Zoé Caspar-Cohen<sup>2,3</sup>, Sylvie Le Gentil<sup>2</sup>, Lu Wang<sup>2</sup>, Wenping Gong<sup>1</sup>

5	<sup>1</sup> School of Marine Sciences, Sun Yat-sen University, and Southern Marine Science and Engineering
6	Guangdong Laboratory (Zhuhai), Zhuhai, China
7	<sup>2</sup> Ifremer, Université de Brest, CNRS, IRD, Laboratoire d'Océanographie Physique et Spatiale, IUEM,
8	Brest, France
9	<sup>3</sup> Scripps Institution of Oceanography, University of California, San Diego, La Jolla, California, USA

# <sup>10</sup> Key Points:

3

11	•	The accuracy of Lagrangian high-frequency kinetic energy estimates is evaluated with
12		a novel twin global numerical simulation experiment.
13	•	Lagrangian velocity spectra are smoother than Eulerian counterparts at tidal fre-
14		quency peaks, but not at near-inertial ones.
15	•	Lagrangian and Eulerian tidal energies agree better in regions characterized by weak
16		low-frequency kinetic energy and high drifter density.

Corresponding author: Xiaolong Yu, yuxlong5@mail.sysu.edu.cn

#### 17 Abstract

In this study, we carried out a novel massive Lagrangian simulation experiment derived 18 from a global  $1/48^{\circ}$  tide-resolving numerical simulation of the ocean circulation. This first-19 time twin experiment enables a comparison between Eulerian (fixed-point) and Lagrangian 20 (along-flow) estimates of kinetic energy (KE), and the quantification of systematic differences 21 between both types of estimations. This comparison represents an important step forward 22 for the mapping of upper ocean high-frequency variability from drifter database. Eulerian 23 KE rotary frequency spectra and band-integrated energy levels (e.g., tidal and near-inertial) 24 are considered as references, and compared to Lagrangian estimates. Our analysis reveals 25 that, apart from the near-inertial band, Lagrangian spectra are systematically smoother, 26 e.g., with wider and lower spectral peaks compared to Eulerian counterparts. Consequently, 27 Lagrangian KE levels obtained from spectra band integrations tend to underestimate Eu-28 lerian levels on average at low-frequency and tidal bands. This underestimation is more 29 significant in regions characterized by large low-frequency KE. In contrast, Lagrangian and 30 Eulerian near-inertial spectra and energy levels are comparable. Further, better agreements 31 between Lagrangian and Eulerian KE levels are generally found in regions of convergent sur-32 face circulation, where Lagrangian particles tend to accumulate. Our results demonstrate 33 that Lagrangian estimates may provide a distorted view of high-frequency variance. To 34 accurately map near-surface velocity climatology at high frequencies (e.g., tidal and near-35 36 inertial) from Lagrangian observations of the Global Drifter Program, conversion methods accounting for the Lagrangian bias need to be developed. 37

### <sup>38</sup> Plain Language Summary

Ocean surface currents play a pivotal role in transporting heat and energy across the 39 global ocean, and thus affect global climate patterns and marine ecosystems. Yet, despite 40 ocean currents' significant role in the Earth system, much of the rapid (high frequency) 41 ocean variability is not known accurately at the moment. In this study, we show that the 42 information derived from the movements of surface drifters, which track ocean currents, 43 may help fill this gap. This is demonstrated with global ocean numerical models, which are 44 now able to represent high-frequency variability associated with tides, winds and eddies, 45 and are therefore powerful tools to evaluate ocean multi-scale variability. We compare here 46 fixed-point (i.e., "Eulerian") and along-flow (i.e., "Lagrangian" or drifter) kinetic energy es-47 timates. Our results show that the Lagrangian frame of reference can induce distortions of 48 rapid motion signals when compared to Eulerian frame of reference, particularly in regions 49 of large kinetic energy and low drifter density. Nevertheless, these two different perspec-50 tives can be reconciled in the estimation of energy levels, as long as adequate frequency 51 bandwidths are chosen. This work highlights the potential of drifter-based observations in 52 enhancing our understanding of high-frequency ocean variability. 53

#### 54 **1** Introduction

The ocean circulation controls the transport and distribution of physical properties 55 and biochemical tracers across the global ocean. Ocean motions at horizontal scales smaller 56 than several hundreds of kilometers and temporal scales shorter than months account for a 57 dominant fraction of kinetic energy (KE; Ferrari & Wunsch, 2009). Its two main contributors 58 are quasi-geostrophic balanced motions, which include mesoscale eddies (horizontal scales of 59 20-300 km, periods of weeks to months) and submesoscale motions (horizontal scales of 0.2-60 20 km, periods of hours to days), and unbalanced internal waves (horizontal scales < 300 km61 62 and periods <1 day). Mesoscale eddies account for most of the global ocean KE and play a key role in the physical equilibrium and biogeochemical functioning of the ocean at climatic 63 scales (McWilliams, 2008; McGillicuddy et al., 2007; Treguier et al., 2014). Submesoscale 64 motions induce, on the other hand, a vigorous vertical circulation and determine the vertical 65 exchanges of heat, carbon, and nutrients (McWilliams, 2016; Lévy et al., 2018; Taylor & 66 Thompson, 2023). Internal waves are a major driver for turbulent mixing in the ocean, 67 which is of fundamental importance to the global overturning circulation (Whalen et al., 68 2020). Internal waves are commonly organized around frequency, and are observed to have 69 energy peaks at tidal and near-inertial frequencies, and a continuous energy distribution 70 across higher frequencies, commonly known as the internal wave continuum. 71

Provided sufficient information is available along spatial and temporal dimensions, one 72 way of characterizing quasi-geostrophic balanced motions and unbalanced internal waves is 73 to estimate the distribution of surface KE as function of spatial and temporal scales. Torres 74 et al. (2018) examined for instance the distribution of surface KE in wavenumber-frequency 75 space from a high-resolution numerical simulation, and showed that lower-frequency motions 76 emanate from larger scales and spread to finer spatial and temporal scales. The emergence 77 of wide-swath altimetry and surface current measuring satellite missions has fostered efforts 78 aiming at improving our understanding of oceanic variability down to O(10 km) and of its 79 manifestation on satellite and in situ observations (Morrow et al., 2019; Du et al., 2021). 80

An emerging dataset to proceed with in situ observational descriptions across scales is 81 that of the Global Drifter Program (GDP; Elipot et al., 2016). With the development of 82 satellite tracking system, the GDP dataset provides global velocity measurements at hourly 83 resolution, and thus enables studies of ocean variability at high frequencies. Yu et al. (2019) 84 compared frequency spectra estimated from GDP drifter data (i.e., Lagrangian) and out-85 put from a high-resolution Massachusetts Institute of Technology general circulation model 86 (MITgcm) simulation (i.e., Eulerian), which enable to point towards inaccurate represen-87 tations of tidal and near-inertial variability in the numerical model. Arbic et al. (2022) 88 performed a similar yet more detailed comparison based on Yu et al. (2019) datasets and 89 an additional global tide-resolving simulation of the HYbrid Coordinate Ocean Model (HY-90 COM). In global maps and zonal averages, numerical models captured the low-frequency 91 and high-frequency variance qualitatively. HYCOM simulation, because of its more fre-92 quently updated wind-forcing and a more finely tuned implementation of tidal variability, 93 was found closer to GDP drifter values compared to MITgcm simulation. However, both 94 studies questioned the equivalence between Eulerian and Lagrangian estimates, which has 95 not been demonstrated yet. 96

Further, GDP drifter observations have been extensively used to achieve global and 97 regional climatology of time-mean and mesoscale oceanic flows (Lumpkin & Johnson, 2013; 98 Lumpkin, 2016), while such mapping at high frequencies remains relatively understudied 99 (e.g., Liu et al., 2019). Understanding and quantifying the differences caused by Lagrangian 100 inherent sampling nature with respect to Eulerian is a key step to ensure the rationality 101 102 of the mapping of high-frequency variance using the GDP data. At semidiurnal frequencies, Caspar-Cohen et al. (2022) recently demonstrated that the displacement of surface 103 drifters may distort low-mode internal tide signals which translated to wider spectral peaks, 104 a mechanism coined "apparent incoherence". 105

In this work, we compare Lagrangian and Eulerian spectral decompositions of surface 106 KE at global scale, with the aid of output from a high-resolution ocean numerical model 107 (LLC4320 simulation; Yu et al., 2019). A central question addressed here is whether high-108 frequency Eulerian KE levels can be accurately estimated from Lagrangian drifters. The 109 paper is organized as follows. Section 2 describes the LLC4320 simulation, the Lagrangian 110 numerical simulation experiments, and methods of spectral analysis and energy level esti-111 mates. Comparisons between Eulerian and Lagrangian KE fields are described in Section 112 3. Discussions and conclusions are given in Sections 4 and 5, respectively. 113

#### <sup>114</sup> 2 Materials and Methods

#### 115 2.1 LLC4320 simulation

The LLC4320 simulation was performed using MITgcm (Marshall et al., 1997) on a 116 global Latitude-Longitude-polar Cap (LLC) grid (Forget et al., 2015) for a period of 14 117 months between 10 September 2011 and 15 November 2012. The model has a horizontal grid 118 spacing of  $1/48^{\circ}$  (approximately 2.3 km at the equator and 0.75 km in the Southern Ocean), 119 and thereby resolves mesoscale eddies and permits submesoscale variability. The model time 120 step was 25 seconds, and model variables were stored at hourly intervals. The model was 121 forced by 6-hourly surface flux fields (including 10-m wind velocity, 2-m air temperature 122 and humidity, downwelling long- and short-wave radiation, and atmospheric pressure load) 123 from the European Centre for Medium-Range Weather Forecasts (ECMWF) operational 124 reanalysis, and included the full lunisolar tidal constituents that are applied as additional 125 external forcing. The LLC4320 uses a flux-limited monotonicity-preserving (seventh order) 126 advection scheme, and the modified Leith scheme of Fox-Kemper and Menemenlis (2008) for 127 horizontal viscosity. The K-profile parameterization (Large et al., 1994) is used for vertical 128 viscosity and diffusivity. In this study, we use a yearlong record of the instantaneous surface 129 fields at every hour, starting on 15 November 2011. 130

131

#### 2.2 Lagrangian experiments

Lagrangian simulations are performed with LLC4320 hourly surface velocity outputs 132 using the 'Parcels' Python package (Lange & Van Sebille, 2017; Delandmeter & Van Se-133 bille, 2019). Surface virtual drifters are initially released every 50 grid points of LLC4320 134 grid (about 50 to 100 km spacing), and drifter positions and velocity fields are stored at 135 hourly rate. The Lagrangian simulation is about one year long (from 15 November 2011 136 to 9 November 2012). Virtual drifters are released every 10 days at initial release positions 137 if no virtual drifter is present within a radius equal to the distance to closest neighbor at 138 initial release. This continuous seeding enables to maintain a continuous coverage through-139 out the Lagrangian simulation. The number of virtual drifters is of about 60,000 at the 140 start, and reaches about 95,000 drifters at the end of the simulation. The Lagrangian par-141 ticles in the LLC4320 simulation (i.e., virtual drifters) are scattered throughout the open 142 ocean worldwide (Figure 1: also see Movie S1 in Supporting Information), and their spatial 143 distribution is broadly in line with that of GDP drifters over the global ocean albeit with 144 an instantaneous drifter density larger by about two orders of magnitudes (Elipot et al., 145 2016; Yu et al., 2019). Heavily sampled regions concentrate in flow convergence zones (e.g., 146 the interior of subtropical gyres). In contrast, areas of flow divergence (e.g., the equatorial 147 region, upwelling areas) as well as polar and coastal regions are generally less sampled. As 148 a result, the number of 60-day particle trajectory segments at midlatitudes (30°-60°N and 149 S) is at least a factor of 2 larger than that in the equatorial region  $(10^{\circ}\text{S}-10^{\circ}\text{N})$ . 150

151 152

# 2.3 Frequency rotary spectrum and bandwidth selection for energy integration

For Eulerian estimates, hourly surface horizontal velocity time series are used to compute rotary spectra of horizontal velocity at each model grid point. For Lagrangian esti-

mates, rotary spectra are computed from horizontal velocities along particle trajectories. 155 For both datasets, we first divide velocity time series into segments of 60 days overlapping 156 by 50% and linearly detrend over each segment, and then compute the 1D discrete Fourier 157 transform of complex-valued fields (u + iv), where u and v are zonal and meridional velocity, 158 respectively) multiplied by a Hanning window. Spectra are formed by multiplying Fourier 159 amplitudes by their complex conjugates and averaged over time for Eulerian estimates and 160 according to segment mean drifter's latitudes and longitudes for Lagrangian estimates (Fig-161 ure 1b). Given the geographical distribution of particle trajectories, velocity data in polar 162 regions with latitude higher than  $60^{\circ}$  and in coastal waters with depth shallower than 500 163 m are not considered in the calculation for both datasets. 164

Rotary frequency spectral densities are integrated over four frequency bands to compute 165 KE components of interest, including high-frequency (>0.5 cpd, absolute values here and 166 hereinafter), semidiurnal, near-inertial, diurnal bands. Total KE is estimated from temporal 167 averages of instantaneous velocity fields, and low-frequency KE is computed as total KE 168 minus high-frequency KE. We examine the sensitivity of the regression coefficient and root 169 mean square error between Eulerian and Lagrangian semidiurnal, near-inertial and diurnal 170 KE levels to different frequency bandwidths of integration (Figure 2). For semidiurnal 171 band, the closest match between Eulerian and Lagrangian energy levels is achieved for the 172  $\pm 0.3$  cpd bandwidth with a regression coefficient value closest to unity and a root mean 173 square error plateauing at approximately  $10^{-3}$  m<sup>2</sup> s<sup>-2</sup> (equivalent to 15.6% of the averaged 174 Eulerian semidiurnal energy level). In contrast, for diurnal and near-inertial bands, the 175 narrowest bandwidth (i.e.,  $\pm 0.1$  cpd) yields the best comparison based on the two metrics. 176 Consequently, the semidiurnal, near-inertial, diurnal bands are respectively defined as 1.7-177 2.3 cpd, 0.9-1.1f and 0.9-1.1 cpd, where f is the Coriolis frequency. 178

To achieve a balance between drifter density and spatial variability of bin-averaged 179 diagnostics, a bin size of 1° latitude is employed to compare Eulerian and Lagrangian zonally 180 averaged rotary spectra and associated band integrals. For global maps, the band-integrated 181 KE estimates are averaged in  $1^{\circ} \times 1^{\circ}$  spatial bins. Finally, following Arbic et al. (2022), we 182 compute the ratio of Lagrangian KE divided by the sum of Lagrangian and Eulerian KE. 183 Note that a ratio of 0.5 indicates equality between Lagrangian KE and Eulerian KE, a ratio 184 exceeding 0.5 indicates Lagrangian KE overestimates Eulerian KE, and a ratio below 0.5185 indicates Lagrangian KE underestimates Eulerian KE. 186

#### 187 **3 Results**

#### 3.1 Zonally-averaged spectrum and KE

Lagrangian and Eulerian zonally averaged spectra both show expected peaks at low, 189 near-inertial and tidal frequencies (Figures 3a and 3b). Along with Figures 3c and 3d, 190 Lagrangian spectral peaks appear to be systematically broader and weaker than Eulerian 191 ones, indicating a spreading of energy in the Lagrangian perspective. This spreading is 192 clear around main tidal peaks and increases with frequency such that Lagrangian higher 193 frequency tidal constituents (e.g. 3 cpd, 4 cpd, ...) are hardly noticeable unlike Eulerian 194 ones. The ratio of Lagrangian to Eulerian spectra consistently indicates that Lagrangian 195 peak values at tidal frequencies are lower but wider than Eulerian ones (Figure 3c). This is 196 in line with the findings of Zaron and Elipot (2021), which noted that the drifter tidal peaks 197 do not stand out above the background spectrum as strongly as in tide model predictions. 198 Caspar-Cohen et al. (2022) consistently demonstrated that the distortion of tidal internal 199 waves induced by surface drifter motions, a process coined as apparent incoherence, leads 200 to broader tidal peaks. 201

At subinertial frequencies, a similar mechanism may be invoked to explain the smoothing of the low-frequency energy peak in Lagrangian diagnostics: Lagrangian particles sample both spatial and temporal variability, which leads to shorter velocity decorrelation timescales

<sup>188</sup> 

and broader spectra (Middleton, 1985; Davis, 1983; Lumpkin et al., 2002; LaCasce, 2008). 205 At near-inertial frequencies, the smoothing of the peak is not visible in latitude dependent 206 spectra (Figures 3a and 3b), and the ratio of Lagrangian to Eulerian spectra indicates values 207 close to unity (Figure 3c). This suggests that drifter displacements do not distort the signa-208 ture of near-inertial waves similarly to internal tides, in line with findings from Shakespeare 209 et al. (2021). Another obvious contrast is that energy levels at the anticyclonic frequen-210 cies are substantially higher than those at the cyclonic frequencies, particularly below the 211 semidiurnal frequency band (Figure 3d). This conforms expectations from the natural po-212 larization of internal gravity waves which leads to a ratio between anticyclonic and cyclonic 213 kinetic energies that scales as  $(\omega + f)^2/(\omega - f)^2$  (Gill, 1982; van Haren, 2003), where  $\omega$  is 214 the frequency, and is consistent with the observational findings of elevated near-inertial KE 215 in the anticyclonic domain (Elipot et al., 2010; Vic et al., 2021; Yu et al., 2022). 216

Zonally averaged low-frequency, semidiurnal, near-inertial, diurnal KE estimated from 217 Eulerian velocity field and Lagrangian particle trajectories with ("2D binned") and without 218 ("raw") spatial binning are displayed in Figure 4. The overall trends of the Eulerian and 219 Lagrangian (with and without binning) estimates show good visual similarities. Raw La-220 grangian estimates tend to underestimate KE compared to 2D binned Lagrangian estimates, 221 particularly at low-frequency energy peaks. Low-frequency energy peaks near the equator, 222 at 35°N and 55°S at the locations of Northern Hemisphere western boundary currents and 223 the Antarctic Circumpolar Current respectively (Figure 4a). At 35°N, Lagrangian ener-224 gies underestimate Eulerian ones, which will be argued to partly result from the unequal 225 sampling of high vs low energy regions (Davis, 1985) and may be mitigated with alterna-226 tive geographical binning (see Discussion section). Lagrangian zonally averaged semidiurnal 227 KE is slightly lower than Eulerian KE at almost all low and mid-latitudes (Figure 4b). 228 In contrast, Lagrangian zonally averaged near-inertial KE follow Eulerian estimates rela-229 tively well over most latitudes, except for a clear underestimation near 30°S, where the local 230 inertial frequency coincides with diurnal frequencies (Figure 4c). For the diurnal KE, dis-231 crepancies are relatively larger, with a 15.4% difference in average (Figure 4d), compared 232 to low-frequency (7%), semidiurnal (11.1%), and near-inertial (6.7%) bands. There are two 233 substantial mismatches between Eulerian and Lagrangian diurnal KE, one is in 20°N, which 234 may be associated with the less Lagrangian particles in Luzon strait, and one is in 30°S, in 235 line with the underestimation in near-inertial KE. 236

#### 237

#### 3.2 Low-frequency KE maps

Global maps of low-frequency KE highlight prominent large-scale currents and energetic 238 areas, including equatorial and western boundary currents such as the Gulf Stream and the 239 Antarctic Circumpolar Current (Figures 5a and 5b). Low-frequency KE dominates total 240 energy and therefore mimic total KE variations (cf. Figures 1a and 5a). Lagrangian and 241 Eulerian low-frequency KE are generally in good agreement, with a mean value and standard 242 deviation of the ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) about 0.49 and 243 0.07, respectively (Figures 5c and 5d). A noticeable difference is that Lagrangian estimates 244 appear smoother than Eulerian ones around large-scale current features, which presumably 245 results from the spatial advection of Lagrangian particle over the temporal window of energy 246 integration. 247

We next examine the dependence of the ratio of Lagrangian KE/(Lagrangian KE+Eulerian 248 KE) on two factors, the number of Lagrangian particles per bin (i.e., drifter density) and 249 the intensity of low-frequency KE. As the number of Lagrangian particles increases, the 250 mean ratio gradually approaches 0.5, and the ratio range is more concentrated (Figure 5e). 251 This suggests that Lagrangian KE tends to align more closely with Eulerian KE in regions 252 where Lagrangian particles accumulate, indicating convergence of the flow. Further, the ra-253 tio exhibits a more pronounced dependence on the intensity of low-frequency KE. In regions 254 of strong low-frequency flow, the ratio is significantly reduced, with a mean value smaller 255 than 0.4, indicating a substantial underestimation of low-frequency KE from a Lagrangian 256

perspective (Figure 5f). This underestimation results from a larger projection of spatial
variability onto temporal one along particle trajectories in energetic regions (LaCasce, 2008;
Caspar-Cohen et al., 2022). Indeed, nearby energetic current features, Lagrangian energy
thus tends to underestimate Eulerian energy maxima in the core of these features (ratio
below 0.5) and overestimate Eulerian energy on the surroundings (ratio above 0.5; Figure
5c).

#### 3.3 Semidiurnal KE maps

263

286

Semidiurnal, diurnal and near-inertial KE are all an order of magnitude smaller than 264 low-frequency KE. The comparison between Lagrangian and Eulerian semidiurnal KE es-265 timates show significant similarities globally (Figures 6a and 6b). Both clearly display 266 hotspots of internal tide generation, e.g., near Hawaii islands, the French Polynesian is-267 lands, the Aleutians island chain, 40°S and 40°N in the Atlantic as well as the western 268 Pacific. The discrepancies at semidiurnal frequencies are relatively larger compared to low 269 frequencies, with a mean energy ratio of 0.47 and standard deviation of 0.06. Figures 6c and 270 6d show that Lagrangian semidiurnal KE systematically underestimate Eulerian semidiur-271 nal KE, particularly in the ocean's major current regions of high kinetic energy. Noticeable 272 differences between Lagrangian and Eulerian estimates also occur near coastal areas, where 273 274 the Lagrangian field may exhibit semidiurnal KE levels considerably larger than the Eulerian field. This overestimate is likely to result from Lagrangian particles crossing continental 275 shelves, where tidal currents are faster, over the 60-day window of energy integration. 276

Interestingly, for semidiurnal tides, the dependence of the ratio on the drifter density is 277 relatively weak, although there is a slight increase in the ratio towards 0.5 with increasing 278 particle counts (Figure 6e). Instead, the ratio shows a clear decreasing and scattering trend 279 as the low-frequency KE intensity increases (Figure 6f). That is, the ratio ranges from about 280 0.23 to 0.67 with a mean value somewhat smaller than 0.5 in areas of strong low-frequency 281 KE, while the range of the ratio reduces to 0.34 and 0.61 with a mean value closer to 0.5282 in areas of weak low-frequency KE. This indicates that the bandwidth tuned based on a 283 global criterion is likely not sufficient in areas of strong low-frequency KE to account for the 284 smearing of the semidiurnal tidal spectral peak (Figure 3d). 285

#### 3.4 Diurnal KE maps

The global map of diurnal Lagrangian KE also closely reproduces the Eulerian KE 287 visually (Figures 7a and 7b). The most prominent feature of diurnal KE is the enhancement 288 around  $\pm 30^{\circ}$  latitudes, where the diurnal wind-forcing (sea breeze) aligns with the local 289 inertial frequency. Similar to the semidiurnal band, the Lagrangian field shows larger values 290 than the Eulerian field in several coastal regions (Figure 7c). These coastal regions are 291 mostly located outside 30°S and 30°N, where diurnal internal tides are not expected to freely 292 propagate, indicating that their differences may be caused by barotropic tides or trapped 293 baroclinic tides. Over the global ocean, Lagrangian diurnal KE slightly underestimate 294 Eulerian one (Figure 7d). 295

Similar to low-frequency motions, diurnal tides show a clear dependence on the drifter 296 density and the intensity low-frequency KE (Figures 7e and 7f). For diurnal tides, the 297 ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) gradually inclines towards 0.5 with 298 the increase in Lagrangian particle counts, and becomes more focused on 0.5 with the 299 decrease in low-frequency KE. In other words, Lagrangian diurnal KE matches Eulerian KE 300 better in regions of weak and convergent flows, and tends to underestimate in regions of 301 strong and divergent flows. This may explain the underestimation of Lagrangian diurnal 302 KE in the Luzon Strait (approximately 20°N; Figure 4d), where the background flow is 303 strong and divergent. Lastly, apparent incoherence is in general expected to be weaker at 304 diurnal frequencies than at semidiurnal frequencies, due to the larger horizontal wavelength 305 of diurnal tides (Caspar-Cohen et al., 2022). 306

#### 307 3.5 Near-inertial KE maps

Lagrangian and Eulerian estimates of near-inertial KE show particularly similar spatial 308 patterns across the global ocean (Figures 8a and 8b). Intensified near-inertial KE generally 309 occurs at mid latitudes, with largest values concentrated in the North Pacific. This is 310 broadly in line with storm-track regions and spatial distribution of wind work (Alford, 2003). 311 Expected enhancements also occur at  $\pm 30^{\circ}$  latitudes where the local inertial frequency 312 coincides with diurnal frequencies. Nearly meridionally oriented beams appear in the low 313 to mid latitudes and are particularly evident in the Eulerian field, likely associated with 314 individual tropical cyclones and storms in the model forcing fields. The mean value and 315 standard deviation of the energy ratio are 0.49 and 0.07, respectively (Figure 8c). Differences 316 between Eulerian and Lagrangian near-inertial KE are modest in global maps compared to 317 those within semidiurnal and diurnal bands. And this also embodies in Figure 8d, in which 318 Eulerian and Lagrangian near-inertial KE closely follow a 1-to-1 relationship. Nonetheless, 319 Lagrangian energy slightly underestimates near-inertial KE over open ocean regions in the 320 Southern Hemisphere, around 30°S. It is probably related to the influence of diurnal bands 321 and a change in the nature of motions (larger contribution from internal tides for instance). 322

Similar to low-frequency and tidal motions, the ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) in the near-inertial band also shows a dependence on the drifter density (Figure 8e). The mean ratio approaches 0.5 and the entire range becomes more focused on 0.5 with the increasing number of Lagrangian particles. However, there is no apparent association between the ratio and the intensity of low-frequency KE (Figure 8f).

#### 328 4 Discussion

In previous studies, ocean surface KE in high-resolution global simulations has been 329 compared with KE from GDP surface drifters (Yu et al., 2019; Arbic et al., 2022). Model-330 drifter comparisons showed good qualitative agreement over a wide range of frequency bands 331 but systematic discrepancies were also observed, and the pending question is whether these 332 discrepancies could be attributed to Lagrangian/Eulerian biases. We find that the difference 333 between Lagrangian and Eulerian KE levels in LLC4320 is significantly lower than the 334 one observed in model-drifter analysis. As in the comparison between LLC4320 and GDP 335 drifters, a deficit of low-frequency energy within the equatorial region was observed, with 336 energy peak values reaching  $0.15 \text{ m}^2 \text{ s}^{-2}$  for the model and  $0.34 \text{ m}^2 \text{ s}^{-2}$  for GDP drifters 337 (Yu et al., 2019). This difference (approximately  $0.2 \text{ m}^2 \text{ s}^{-2}$ ) is one order of magnitude 338 larger than the Lagrangian-Eulerian difference near the equator, which is of order 0.01339  $m^2 s^{-2}$  as shown in Figure 4a. Moreover, Yu et al. (2019) reported that the LLC4320 340 simulation exhibits KE four times higher in the semidiurnal band and three times lower in 341 the near-inertial band compared with GDP drifter data. However, the global mean ratios 342 of Lagrangian KE to Lagrangian KE+Eulerian KE obtained in this study are  $0.47\pm0.06$ 343 for the semidiurnal band (Figure 6c) and  $0.49\pm0.07$  for the near-inertial band (Figure 8c), 344 both of which are close to 0.5. This means that, on average, Lagrangian KE is nearly equal 345 to Eulerian KE for the semidiurnal and near-inertial bands in the LLC4320 simulation. 346 Therefore, our results suggest that Lagrangian/Eulerian biases are very likely not the main 347 cause of the model-drifter discrepancies. 348

Arbic et al. (2022) identified the sensitivity of the Lagrangian semidiurnal energy esti-349 mate to the bandwidth of integration, which the present study corroborates. This sensitivity 350 arises from apparent incoherence which leads to a widening of the semidiurnal tidal peak 351 (Caspar-Cohen et al., 2022). A single common value (i.e., 1.7-2.3 cpd) for the bandwidth 352 of integration that produced the best match between Lagrangian and Eulerian energy es-353 timates was chosen in the present study. However, the bandwidth of integration may not 354 be the same for Eulerian and Lagrangian energy diagnostics. Limiting bandwidth may be 355 desirable in order to mitigate contamination from the background energy spectrum. Given 356 the sharper shape of Eulerian semidiurnal peaks, smaller bandwidths may be afforded for 357

Eulerian diagnostics. The width of Eulerian semidiurnal peaks is related to internal tide 358 incoherent timescales, whose geographical variations may lead to geographically varying 359 choices for the bandwidth of integration of Eulerian energy. Caspar-Cohen et al. (2022) 360 theoretically predict that the intensity of apparent incoherence and thus the associated 361 widening of the Lagrangian spectrum depends on parameters that may vary geographically 362 such as the low-frequency energy level and decorrelation timescale or internal tide properties 363 (wavenumber, incoherent timescale). The bandwidth of integration of Lagrangian estimates 364 may thus be modulated geographically in order again to mitigate contamination from the 365 background energy spectrum. Such more advanced choices for bandwidth of energy inte-366 gration would be good material for future studies, even though the present study indicates 367 this would be mostly relevant for the semidiurnal band which exhibits most sensitivity to 368 integration bandwidth. 369

Lastly, Lagrangian low-frequency KE estimates considerably underestimate Eulerian 370 ones in energetic regions (Figure 4a and 5f), likely due to the preferential sampling of 371 weak-energy regions by Lagrangian particles (Freeland et al., 1975). The deficit caused 372 by the inhomogeneous sampling of Lagrangian particles, in KE levels of up to 20%, can be 373 compensated by averaging Lagrangian diagnostics into longitude/latitudes bins. Our recom-374 mendation is to geographically bin energy estimates prior to integration over larger domains 375 (e.g., zonally, globally) to mitigate such sampling biases. However, it should be noted that 376 along a similar line but at the bin level, the combination of spatio-temporal inhomogeneities 377 and energy variability within individual bins may also lead to systematic differences between 378 Eulerian and Lagrangian estimates (Davis, 1991), such as those observed nearby large cur-379 rent systems. The role of such sampling bias in explaining observed differences has not 380 been investigated here but could constitute an interesting follow up study. Further, we have 381 chosen here a 60-day time window for spectral decompositions and energy estimates, and 382 this choices induces spatial smoothing compared to Eulerian estimates. Above-mentioned 383 study may be useful in order to identify whether statistical techniques enabling more local 384 (temporally and therefore spatially for drifters) estimates of high-frequency energy levels 385 should be devised. 386

#### 387 5 Summary

In this study, we quantify the relationship between Eulerian and Lagrangian KE spectral content and its geographical variability based on a novel twin global numerical simulation experiment. A practical objective is to assess the extent to which Lagrangian particles can estimate Eulerian KE levels, especially at high frequencies. To achieve this, we have compared the surface KE estimated using the LLC4320 global ocean model, and the Lagrangian simulations performed by the LLC4320 hourly surface velocity output. Our main findings are summarized as follows:

1) Eulerian and Lagrangian KE exhibit broad qualitative similarities, with the domi-395 nance of low-frequency motions and the presence of distinct spectral peaks at semidiurnal, 396 near-inertial, and diurnal frequencies. A common feature among all dominant frequency 397 bands is that Lagrangian spectra appear smoothed compared to Eulerian ones. This smooth-398 ing is least pronounced in the near-inertial band and most pronounced in the semidiurnal 300 and low-frequency bands. At low frequencies, this smoothing is attributed to the simultane-400 ous sampling of spatial and temporal variability by particles and the associated decrease in 401 the decorrelation timescale (Middleton, 1985; LaCasce, 2008). The widening of the semidi-402 urnal peak is consistent with the mechanism of apparent incoherence (Caspar-Cohen et al., 403 2022). The relatively minor difference between the Lagrangian and Eulerian near-inertial 404 frequency peaks remains to be explained. 405

2) With a tuned choice of bandwidth of integration, good agreements for semidiurnal (1.7-2.3 cpd), near-inertial (0.9-1.1f) and diurnal (0.9-1.1 cpd) KE can be achieved between Eulerian and Lagrangian simulations. This implies that Lagrangian particles advected by Eulerian filed can qualitatively reproduce the original Eulerian high-frequency variance.
Compared to Eulerian estimates, Lagrangian estimates are more sensitive to bandwidth,
as expected from their character of broadened spectral peaks. Particularly, Lagrangian
semidiurnal tides are featured with a wider bandwidth than other high-frequency motions.
We have identified avenues to refine further this choice of bandwidth of integration.

3) The intensity of low-frequency motions affects Lagrangian KE estimates at low and
tidal frequencies. Lowest Lagrangian to Eulerian energy ratio is observed in energetic and
turbulent areas. Conversely, Lagrangian KE in near-inertial band has no clear connection
with low-frequency KE. For all bands, Lagrangian and Eulerian KE levels have a better
agreement in regions of convergent flows, where Lagrangian particles accumulate, than in
regions of divergent flows, where Lagrangian particles scatter.

Our findings confirm that the drifter data may provide an estimate of high-frequency variance, such as tidal and near-inertial motions. Drifter and model differences, as shown in Yu et al. (2019) and Arbic et al. (2022), are not mainly caused by Lagrangian vs Eulerian sampling nature. This work may motivate future studies on particular aspects of the modelobservation and model-model discrepancies, and is a substantial step towards the production of high frequency KE climatologies.

# 426 Acknowledgments

This work was funded by grants from the National Natural Science Foundation of China (42206002, 42361144844), and Guangdong Basic and Applied Basic Research Foundation (2023A1515010654). This work was also supported by ANR project 17-CE01-0006-01 entitled EQUINOx (Disentangling Quasi-geostrophic Motions and Internal Waves in High Resolution Satellite Observations of the Ocean).

# 432 Open Research

433

The LLC4320 simulation output is available at https://data.nas.nasa.gov/ecco/data.php.

# 434 References

- Alford, M. H. (2003). Redistribution of energy available for ocean mixing by long-range propagation of internal waves. *Nature*, 423(6936), 159-162.
- Arbic, B. K., Elipot, S., Brasch, J. M., Menemenlis, D., Ponte, A. L., Shriver, J. F., ...
   Nelson, A. D. (2022). Near-surface oceanic kinetic energy distributions from drifter observations and numerical models. *Journal of Geophysical Research: Oceans*, 127(10), e2022JC018551.
- Caspar-Cohen, Z., Ponte, A., Lahaye, N., Carton, X., Yu, X., & Gentil, S. L. (2022).
   Characterization of internal tide incoherence: Eulerian versus Lagrangian perspectives.
   Journal of Physical Oceanography, 52(6), 1245-1259.
- Davis, R. E. (1983). Oceanic property transport, Lagrangian particle statistics, and their prediction. *Journal of Marine Research*, 41, 163-194.
- Davis, R. E. (1985). Drifter observations of coastal surface currents during CODE: The
   statistical and dynamical views. Journal of Geophysical Research: Oceans, 90(C3),
   447
   448
   4756-4772.
- <sup>449</sup> Davis, R. E. (1991). Observing the general circulation with floats. *Deep Sea Research Part* <sup>450</sup> A. Oceanographic Research Papers, 38, S531-S571.
- <sup>451</sup> Delandmeter, P., & Van Sebille, E. (2019). The Parcels v2.0 Lagrangian framework: new <sup>452</sup> field interpolation schemes. *Geoscientific Model Development*, 12(8), 3571-3584.
- <sup>453</sup> Du, Y., Dong, X., Jiang, X., Zhang, Y., Zhu, D., Sun, Q., ... Peng, S. (2021). Ocean
  <sup>454</sup> surface current multiscale observation mission (OSCOM): Simultaneous measurement
  <sup>455</sup> of ocean surface current, vector wind, and temperature. *Progress in Oceanography*,
  <sup>456</sup> 193, 102531.

- Elipot, S., Lumpkin, R., Perez, R. C., Lilly, J. M., Early, J. J., & Sykulski, A. M. (2016). A
   global surface drifter data set at hourly resolution. *Journal of Geophysical Research: Oceans*, 121(5), 2937-2966.
- Elipot, S., Lumpkin, R., & Prieto, G. (2010). Modification of inertial oscillations by the mesoscale eddy field. *Journal of Geophysical Research: Oceans*, 115(C9).
- Ferrari, R., & Wunsch, C. (2009). Ocean circulation kinetic energy: Reservoirs, sources, and sinks. Annual Review of Fluid Mechanics, 41, 253-282.
- Forget, G., Campin, J. M., Heimbach, P., Hill, C. N., Ponte, R. M., & Wunsch, C. (2015).
   ECCO version 4: an integrated framework for non-linear inverse modeling and global ocean state estimation. *Geoscientific Model Development*, 8(10), 3071-3104.
- Fox-Kemper, B., & Menemenlis, D. (2008). Can large eddy simulation techniques improve
   mesoscale rich ocean models? In Ocean modeling in an eddying regime (p. 319-337).
   American Geophysical Union (AGU).
- Freeland, H. J., Rhines, P. B., & Rossby, T. (1975). Statistical observations of the trajectories of neutrally buoyant floats in the north atlantic. *Journal of Marine Research*, 33.
- 473 Gill, A. E. (1982). Atmosphere-ocean dynamics. Academic press.

482

483

485

486

487

488

489

492

493

494

495

- LaCasce, J. H. (2008). Lagrangian statistics from oceanic and atmospheric observations.
   In J. B. Weiss & A. Provenzale (Eds.), *Transport and mixing in geophysical flows: Creators of modern physics* (p. 165-218). Springer Berlin Heidelberg.
- Lange, M., & Van Sebille, E. (2017). Parcels v0.9: Prototyping a Lagrangian ocean analysis framework for the petascale age. *Geoscientific Model Development*, 10(11), 4175-4186.
- Large, W. G., Mcwilliams, J. C., & Doney, S. C. (1994). Oceanic vertical mixing a review and a model with a nonlocal boundary-layer parameterization. *Reviews of Geophysics*, 32(4), 363-403.
  - Liu, Y., Jing, Z., & Wu, L. (2019). Wind power on oceanic near-inertial oscillations in the global ocean estimated from surface drifters. *Geophysical Research Letters*, 46(5), 2647–2653.
  - Lumpkin, R. (2016). Global characteristics of coherent vortices from surface drifter trajectories. Journal of Geophysical Research: Oceans, 121(2), 1306-1321.
  - Lumpkin, R., & Johnson, G. C. (2013). Global ocean surface velocities from drifters: Mean, variance, El Nino-Southern Oscillation response, and seasonal cycle. *Journal* of Geophysical Research: Oceans, 118(6), 2992-3006.
- Lumpkin, R., Treguier, A.-M., & Speer, K. (2002). Lagrangian eddy scales in the northern atlantic ocean. *Journal of Physical Oceanography*, 32(9), 2425-2440.
  - Lévy, M., Franks, P. J. S., & Smith, K. S. (2018). The role of submesoscale currents in structuring marine ecosystems. *Nature Communications*, 9.
  - Marshall, J., Adcroft, A., Hill, C., Perelman, L., & Heisey, C. (1997). A finite-volume, incompressible navier stokes model for studies of the ocean on parallel computers. *Journal of Geophysical Research: Oceans*, 102(C3), 5753-5766.
- McGillicuddy, D. J., Anderson, L. A., Bates, N. R., Bibby, T., Buesseler, K. O., Carlson,
   C. A., ... Steinberg, D. K. (2007). Eddy/wind interactions stimulate extraordinary
   mid-ocean plankton blooms. *Science*, *316*(5827), 1021-1026.
- McWilliams, J. C. (2008). The nature and consequences of oceanic eddies. In Ocean modeling in an eddying regime (p. 5-15). American Geophysical Union (AGU).
- McWilliams, J. C. (2016). Submesoscale currents in the ocean. Proceedings of the Royal Society A: Mathematical Physical and Engineering Sciences, 472(2189).
- Middleton, J. F. (1985). Drifter spectra and diffusivities. Journal of Marine Research, 43, 37-55.
- Morrow, R., Fu, L. L., Ardhuin, F., Benkiran, M., Chapron, B., Cosme, E., ... Zaron, E. D.
   (2019). Global observations of fine-scale ocean surface topography with the surface water and ocean topography (SWOT) mission. *Frontiers in Marine Science*, 6, 232.
- Shakespeare, C. J., Gibson, A. H., Hogg, A. M., Bachman, S. D., Keating, S. R., & Velze boer, N. (2021). A new open source implementation of Lagrangian filtering: A method

to identify internal waves in high-resolution simulations. Journal of Advances in Modeling Earth Systems, 13(10), e2021MS002616.

Taylor, J. R., & Thompson, A. F. (2023). Submesoscale dynamics in the upper ocean. Annual Review of Fluid Mechanics, 55, 103-127.

511

512

524

525

526

527

528

- Torres, H. S., Klein, P., Menemenlis, D., Qiu, B., Su, Z., Wang, J. B., ... Fu, L. L.
   (2018). Partitioning ocean motions into balanced motions and internal gravity waves: A modeling study in anticipation of future space missions. *Journal of Geophysical Research: Oceans*, 123(11), 8084-8105.
- Treguier, A. M., Deshayes, J., Le Sommer, J., Lique, C., Madec, G., Penduff, T., ... Talandier, C. (2014). Meridional transport of salt in the global ocean from an eddyresolving model. *Ocean Science*, 10(2), 243-255.
- van Haren, H. (2003). On the polarization of oscillatory currents in the bay of biscay.
   Journal of Geophysical Research: Oceans, 108(C9).
  - Vic, C., Ferron, B., Thierry, V., Mercier, H., & Lherminier, P. (2021). Tidal and nearinertial internal waves over the Reykjanes Ridge. *Journal of Physical Oceanography*, 51(2), 419–437.
  - Whalen, C. B., de Lavergne, C., Garabato, A. C. N., Klymak, J. M., MacKinnon, J. A., & Sheen, K. L. (2020). Internal wave-driven mixing: governing processes and consequences for climate. *Nature Reviews Earth & Environment*, 1, 606-621.
- Yu, X., Naveira Garabato, A. C., Vic, C., Gula, J., Savage, A. C., Wang, J., ... MacKinnon, J. A. (2022). Observed equatorward propagation and chimney effect of near-inertial waves in the midlatitude ocean. *Geophysical Research Letters*, 49(13), e2022GL098522.
- Yu, X., Ponte, A. L., Elipot, S., Menemenlis, D., Zaron, E. D., & Abernathey, R. (2019).
   Surface kinetic energy distributions in the global oceans from a high-resolution numerical model and surface drifter observations. *Geophysical Research Letters*, 46(16), 9757-9766.
- Zaron, E. D., & Elipot, S. (2021). An assessment of global ocean barotropic tide models using
   geodetic mission altimetry and surface drifters. *Journal of Physical Oceanography*,
   51(1), 63-82.



Figure 1. (a) Global map of Eulerian total KE at the surface layer in  $1^{\circ} \times 1^{\circ}$  bins. (b) Distribution of the number of 60-day Lagrangian trajectory segments over the global ocean in  $1^{\circ} \times 1^{\circ}$  bins.



Figure 2. The (a) regression coefficient and (b) root mean square error  $(m^2 s^{-2})$  between Lagrangian and Eulerian semidiurnal, near-inertial and diurnal KE as a function of bandwidths of  $\pm 0.1$  cpd,  $\pm 0.2$  cpd,  $\pm 0.3$  cpd,  $\pm 0.4$  cpd.



Figure 3. Zonally averaged rotary frequency spectra in 1° latitude bins from (a) Lagrangian and (b) Eulerian horizontal velocity fields at the surface layer and (c) their ratio, with positive (negative) frequencies corresponding to counterclockwise (clockwise) rotating motions, which are cyclonic (anticyclonic) in the Northern Hemisphere. The cyclonic inertial frequency  $(f/2\pi \text{ cpd})$  is indicated by the gray dashed line and the anticyclonic inertial frequency  $(-f/2\pi \text{ cpd})$  is indicated by the black dashed line. (d) Globally averaged anticyclonic (at negative frequencies) and cyclonic (at positive frequencies) spectra of the Eulerian (blue) and Lagrangian (orange) horizontal velocity fields.



**Figure 4.** Zonally averaged (a) low-frequency, (b) semidiurnal, (c) near-inertial and (d) diurnal KE in 1° latitude bins estimated from Eulerian velocity field (blue) and Lagrangian particle trajectories with (orange) and without (green) binning.



Figure 5. (a-c) Global maps of Lagrangian and Eulerian low-frequency KE at the surface layer and the ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) in  $1^{\circ} \times 1^{\circ}$  bins. Mean value and one standard deviation of the ratio are given in the black box in (c). (d) Joint plot of the comparison between Lagrangian and Eulerian low-frequency KE levels. (e) Box plot of the ratio under different ranges of counts of Lagrangian particles. (f) Box plot of the ratio under different ranges of lowfrequency KE. The dashed red lines in (e) and (f) indicate the conditional means of the ratio.



Figure 6. (a-c) Global maps of Lagrangian and Eulerian semidiurnal KE at the surface layer and the ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) in  $1^{\circ} \times 1^{\circ}$  bins. Mean value and one standard deviation of the ratio are given in the black box in (c). (d) Joint plot of the comparison between Lagrangian and Eulerian semidiurnal KE levels. (e) Box plot of the ratio under different ranges of counts of Lagrangian particles. (f) Box plot of the ratio under different ranges of lowfrequency KE. The dashed red lines in (e) and (f) indicate the conditional means of the ratio.



Figure 7. (a-c) Global maps of Lagrangian and Eulerian diurnal KE at the surface layer and the ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) in  $1^{\circ} \times 1^{\circ}$  bins. Mean value and one standard deviation of the ratio are given in the black box in (c). (d) Joint plot of the comparison between Lagrangian and Eulerian diurnal KE levels. (e) Box plot of the ratio under different ranges of counts of Lagrangian particles. (f) Box plot of the ratio under different ranges of low-frequency KE. The dashed red lines in (e) and (f) indicate the conditional means of the ratio.



Figure 8. (a-c) Global maps of Lagrangian and Eulerian near-inertial KE at the surface layer and the ratio of Lagrangian KE/(Lagrangian KE+Eulerian KE) in  $1^{\circ} \times 1^{\circ}$  bins. Mean value and one standard deviation of the ratio are given in the black box in (c). (d) Joint plot of the comparison between Lagrangian and Eulerian near-inertial KE levels. (e) Box plot of the ratio under different ranges of counts of Lagrangian particles. (f) Box plot of the ratio under different ranges of lowfrequency KE. The dashed red lines in (e) and (f) indicate the conditional means of the ratio.