



1     **SEIA: a scale-selective eddy identification algorithm for the**  
2                                   **global ocean**

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## 22    **Abstract**

23        Automatic eddy identification algorithms are crucial for global eddy research. This  
24    study   presents   a   scale-selective   eddy   identification   algorithm   (SEIA;  
25    <https://github.com/Yk-Yang/SEIA>) for the global ocean based on closed sea level  
26    anomalies (SLAs) that features two improvements in the detection and tracking  
27    processes. First, the scale-selective scheme replaces the previously used threshold for  
28    defining the eddy boundary and restricts the numbers of upper and lower grid points  
29    based on the data resolution and eddy spatial scale. Under such conditions, the eddy  
30    boundary is as large as possible, while the eddy region is not overestimated.  
31    Furthermore, a novel and effective overlap scheme is used to track eddies by calculating  
32    the intersecting ratio of time-step-successive eddies. SEIA generates 278,630  
33    anticyclonic eddies and 274,351 cyclonic eddies from AVISO's SLA dataset over a  
34    five-year period (2015-2019; <https://doi.org/10.11922/sciencedb.o00035.00004>; Yang  
35    et al., 2022). The global distribution of eddies, propagation speed, and eddy path  
36    characteristics in the Southern Ocean verify the validity of SEIA.

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## 41 1. Introduction

42 Mesoscale eddies generally refer to swirling signals with space scales of  $50^1$ - $50^2$   
43 km and time scales of  $10^1$ - $10^2$  days in the ocean (Morrow et al., 2012; Faghmous et al.,  
44 2015). Eddies tend to trap fluid inside (Chelton et al., 2011), causing matter and energy  
45 transport (Wyrli et al., 1976). Many studies have been conducted to explore the effects  
46 of eddies on atmospheric phenomena (Chelton et al., 2004; Frenger et al., 2013) and  
47 the transport and re-distribution of heat, salt, water masses, chlorophyll and other  
48 chemical variables (Jayne et al., 2002; Volkov, 2008; Gordon et al., 2013; Dong et al.,  
49 2014; Zhang et al., 2014; Dufois et al., 2016; Yang et al., 2021).

50 The study of mesoscale eddies is highly reliant on the physical characteristics of  
51 the real ocean based on in situ observations. Eddy identification algorithms are  
52 constantly emerging as a result of coarse in situ observation data and the development  
53 of satellite altimeter data. The output of the algorithms allows researchers to assess the  
54 generation, development, and extinction, as well as hydrodynamic conditions and other  
55 dynamic processes, of eddies. A complete eddy identification algorithm contains the  
56 Eulerian procedure, i.e., the detection of snapshots, and the Lagrangian procedure,  
57 namely, the tracking of eddy movement. Nencioli et al. (2016) further categorized the  
58 detection into three types: physical parameter based, flow geometry based, and hybrid.  
59 The effect of eddies can be detected in the change in sea surface variables, so in this  
60 study, we divide the variable-based algorithms into three types: 1) ecology, 2)  
61 thermodynamics and 3) dynamics.



62       The ecological method detects eddies mainly based on chlorophyll concentration  
63       (Rodriguez et al., 2003; Dong et al., 2009; Nencioli et al., 2010; He et al., 2019). Yu et  
64       al. (2011) applied such a scheme based on MODIS chlorophyll concentration products,  
65       but it had an unsatisfactory effect on anticyclonic eddies (AEs). Sea surface temperature  
66       (SST) is the key indicator for the thermodynamics method. A flexible quasi-contour  
67       tracing and clustering method was used in Zhang et al. (2015), imposing no fixed  
68       restrictions or limitations during the course of “suspected” eddy detection. Using a  
69       high-resolution SST dataset, Moschos et al. (2020) introduced the deep learning  
70       technique to eddy detection. The issue with this type of method is that the outermost  
71       threshold of temperature may be arbitrary. Sea surface salinity (SSS) is also an effective  
72       way to detect oceanic eddies, but there are few precedents due to limited datasets.

73       Following the third type, many algorithms use sea surface height (SSH; Chelton et  
74       al., 2011; Mason and Pascual, 2014), SLA (Faghmous et al., 2015; Li et al., 2014, 2016;  
75       Liu et al., 2016) field, streamlines derived from the flow field (Sadarjoen and Post 2000;  
76       Nencioli et al., 2010; Le Vu et al., 2018), and physical parameters (Okubo, 1970; Weiss,  
77       1991; Doglioli et al., 2007; Souza et al., 2011; Williams et al., 2011; Petersen et al.,  
78       2013) calculated based on geometrical properties to detect and track oceanic eddies.  
79       These methods have been widely used in related studies (we cite some; Chelton et al.,  
80       2007; Chaigneau et al., 2008, 2009; Chen et al., 2011; Liu et al., 2012; Frenger et al.,  
81       2015; Li et al., 2015; Yang et al., 2019; Yang et al., 2021). Nevertheless, these  
82       approaches suffer from a variety of deficiencies and defects. Nencioli et al. (2010) noted  
83       that the smoothing algorithm used in the Okubo-Weiss method (O-W; Okubo, 1970;



84 Weiss, 1991) might remove physical information. Lian et al. (2019) found that the O-  
85 W method tends to perform poorly in areas with a high shear current. Furthermore, a  
86 relatively low detection rate of the O-W method was shown in Xing and Yang (2020).  
87 Similar to the SST-based method, the SSH- and SLA-based methods are typically  
88 restricted by a given threshold (Li et al., 2014; Chelton et al., 2011; Faghmous et al.,  
89 2015), which might constrain the eddy shape and, more importantly, indiscriminately  
90 remove both spurious and real features (Faghmous et al., 2015). The flow-field-based  
91 winding-angle (W-A; Sadarjoen and Post, 2000) method overestimates the eddy  
92 boundary (Tao and Yang, 2020). Nencioli et al. (2010) presented a flow-field-based  
93 method with two important parameters allowing the algorithm to be applied to data with  
94 different resolutions. Indeed, preset parameters make the method flexible but subjective  
95 and not user-friendly.

96 No ideal eddy detection and tracking method exists. In this study, we aim to present  
97 a scale-selective eddy identification algorithm (hereafter, SEIA) that quickly outputs  
98 results. To avoid the noted parameter issue, we enhance the initial scale selection based  
99 on the space scale definition of mesoscale eddies and the resolution of the input dataset.  
100 The upper and lower limits of grid points are set to remove false eddy-like structures  
101 and to solve the given threshold problem. The dataset and methodology of SEIA are  
102 detailed in section 2. On the basis of the output from SEIA, statistical analysis is  
103 presented in section 3 to state its reliability. Finally, a summary is provided in section  
104 4.



## 105    **2. Data and methodology**

### 106    **2.1 Satellite and topography data**

107        The multi-mission merged daily SLA with  $1/4^\circ$  resolution in both latitude and  
 108        longitude from the French Archiving, Validation, and Interpretation of Satellite  
 109        Oceanographic Data (AVISO) project from 2015 to 2019 is utilized to detect and track  
 110        eddies. In the nearshore region, the satellite data contain signal from tides and internal  
 111        waves (Yuan et al., 2006), thus the dataset of land topography and ocean bathymetry  
 112        ETOPO1 from NOAA is used to mask such area.

### 113    **2.2 Eddy detection principles: searching, restricting and clustering**

114        Hausmann and Czaja (2012) reported that there are near-isotropic rotating cores  
 115        within eddy composites with a typical amplitude of approximately 15 cm in the region  
 116        of the Gulf Stream and the Antarctic Circumpolar Current. On the basis of this  
 117        acknowledgement, with SLA from AVISO, the basic principle of eddy detection is to  
 118        find appropriate closed SLA contours with a single core. The streamlines of eddies  
 119        approximately correspond to closed contours of SLA, so it is possible to detect eddies  
 120        based on the outermost closed contour of SLA (Chelton et al., 2011). Following Krantz  
 121        et al. (1999), Zhang et al. (2015) separated the contours into four types: simple open,  
 122        simple closed, concentric closed and intersecting closed (Figs. 1a-d). Various  
 123        algorithms consider the multi-core structure, i.e., concentric closed contour (Fig. 1c),  
 124        to study the splitting and merging of eddies (Cui et al., 2019, 2020). Here, we consider



125 only the simple contour condition with only one core (Fig. 1b): the concentric (Fig. 1c)  
126 and intersecting (Fig. 1d) closed types degrade to this type. The red lines in Figs. 1a-d  
127 indicate the possible detection boundaries of SEIA, which we call the “mononuclear-  
128 closed” principle.

129 For the determination of the outermost boundary of an eddy, a scale-selective  
130 scheme is utilized instead of giving a fixed threshold of SLA (detailed in (iii) below).  
131 A preset arbitrary threshold will undoubtedly result in failed or poor detection of some  
132 eddies (Fig. 1e), as eddies evolve all the time. Taking the assessment of SEIA in the  
133 northwestern Pacific as an example (Fig. 2), the specified detection processes are as  
134 follows:

135 (i) Searching for SLA peaks

136 Initially, for a snapshot, peaks (maxima or minima) of SLA are searched in a  
137 moving 3×3 grid window (Fig. 2a). A peak occurs when the value of a grid is smaller  
138 or larger than that of the 8 neighbouring grids. Faghmous et al. (2015) implemented a  
139 5×5 neighbourhood, suggesting that a 3×3 neighbourhood might result in too many  
140 peaks by random chance. However, the peaks detected in this step are seen only as  
141 potential or suspected cores, and unqualified peaks are deleted in the following steps.  
142 Notably, due to the nearshore issue of satellite data mentioned above, SLA shallower  
143 than certain depth (50 m for SEIA) will be masked.

144 (ii) Searching for closed-SLA contours



145 Due to the AVISO SLA error of approximately 0.03 m (Ponte et al., 2007), the  
 146 extreme outermost value is set to  $\pm 0.025$  m (either increasing or decreasing with a step  
 147 of 0.02 m) to search the closed contours. SEIA is run in MATLAB, where contours are  
 148 interpolated from the original data. Therefore, a self-defined function is used to trace  
 149 back to the original grids of input data under the principle of the closest values. Hence,  
 150 quasi-contours are actually used in this study (Fig. 2b).

151 (iii) Restricting the contour grid points

152 After step (ii), several closed contours will exist for one eddy-like structure.  
 153 Therefore, how can we define the “edge” of eddies without a given threshold? A scale-  
 154 selective scheme of two restrictions is applied to solve this problem. The first is the  
 155 one-core policy, which allows concentric closed contours to contract and shrink into  
 156 simple closed contours (Figs. 1b-d). Furthermore, initial spatial scale filtering is the  
 157 next vital limitation. The size of eddies is strictly limited to the space scale of  $50^1$ - $50^2$   
 158 km with an influence radius  $L$  ranging from 25 to 125 km. Viewing eddies as quasi-  
 159 circles, the approximate grid points of the influence region (Fig. 2c) can be estimated  
 160 as:

$$161 \quad P = 2\pi L / R_s + c, \quad c = 1 \quad (1)$$

162 where  $2\pi L$  is the perimeter of the eddy boundary;  $R_s$  is the resolution of the input  
 163 dataset; and index  $c$  is an error-compensating correction. With  $R_s = 0.25^\circ$  (taking  
 164  $1^\circ \approx 100$  km in both longitude and latitude) of the SLA from AVISO, the lower and





165 upper grid points are  $P_{\min} \approx 7.2$  and  $P_{\max} \approx 32.4$ , respectively.

166 Xing and Yang (2020) proposed the concept of a below-standard eddy whose radius  
 167 fell below the resolution of the data used. With  $P_{\min} \approx 7.2$ , eddy boundaries that occupy  
 168 only 7 grid points can easily be deduced as below-standard. For the  $3 \times 3$  grid window  
 169 used in searching for the SLA peaks, a  $P_{\min}$  of 8 is much more reasonable. Additionally,  
 170 assessments have been conducted in the northwestern Pacific, showing good  
 171 performance with  $P_{\min}=10$ . In other words,  $P_{\min}$  can be preset from 8 to 10, where 8  
 172 is ideal.  $P_{\max}$  can be assigned the value calculated from formula (1), for example,  
 173  $P_{\max}=32$  in this study. Then, contours with  $P$  smaller than  $P_{\min}$  or greater than  $P_{\max}$   
 174 will be deleted.

175 Indeed, there is no universal definition of an eddy boundary. However, pre-setting  
 176 a given threshold of SLA is not the best choice. Here, we innovatively propose a new  
 177 scheme of restricting the pixels of boundaries accompanied by a one-core policy based  
 178 on the scale definition. No arbitrary values are set, and the largest influence regions as  
 179 possible are detected.

180 (iv) Clustering the closed contours and the peaks

181 After the restrictions of closed contours, the qualified peaks can be screened. Those  
 182 peaks without closed contours will be removed, and the rest will be “assigned” probable  
 183 closed contours ( $P_{\min} \leq P \leq P_{\max}$ ); thus, other eddy information, such as radius, can be  
 184 obtained. Finally, all eddies are detected in a Euler manner for one snapshot (Fig. 2d).



## 185 2.3 Eddy tracking scheme

186 The tracking process is a complicated task as eddies deform constantly. The  
 187 methodologies in previous studies vary based on the connectivity algorithm (Henson  
 188 and Thomas, 2008), centre-comparing based on the background current velocity  
 189 (Doglioli et al., 2007; Chaigneau et al., 2008; Nencioli et al., 2010), nondimensional  
 190 parameters (Penven et al., 2005; Chen et al., 2011), parameter W from the O-W method  
 191 (Chelton et al., 2011) and so on. In this study, a quick and effective methodology is  
 192 proposed for the first time: the overlap scheme.

193 For a detected eddy, the overlap scheme searches the several nearest eddies in  
 194 successive time steps and calculates the ratio of intersecting area with the larger eddy  
 195 region, as shown in the following formula:

$$196 \quad E_s \ni \left( \frac{e_0 \cap \{e_i, i=1 \dots n\}}{\max\{e_0, e_i, i=1 \dots n\}} \geq R_t \right) \quad (2)$$

197 where  $E_s$  represents a successful tracked eddy in a Lagrangian manner;  $e_0$  is the  
 198 initial eddy region and the series of  $e_i$  denotes the regions of nearest eddies;  $D_t$  is the  
 199 largest searching distance of the nearest eddies; and  $R_t$  is the overlapping ratio. Eddies  
 200 are determined to be the same when the ratio is larger than a certain value. Clearly, the  
 201 search distance  $D_t$  and the overlapping ratio  $R_t$  are two key factors.

202 The determination of  $D_t$  is highly related to the nearly due west eddy propagation  
 203 speed, which is essentially equal to the phase speed of long nondispersive baroclinic



204 Rossby waves (Chelton et al., 2007; Faghmous et al., 2015). The largest phase speed of  
 205 long nondispersive baroclinic Rossby waves is approximately  $0.16 \text{ m}\cdot\text{s}^{-1}$  outside the  
 206 equator ( $\pm 10^\circ$ ), as reported in Chelton et al. (2007). For the daily SLA dataset, an eddy  
 207 theoretically propagates approximately 13.8 km or approximately  $0.14^\circ$  at  $30^\circ\text{N}$  with  
 208 an Earth radius of 6371 km. The data resolution used in this study is  $0.25^\circ$ , which means  
 209 that movement of  $0.14^\circ$  cannot be captured. Therefore, the search distance  $D_t$  will be  
 210 too conservative if it is set to  $0.14^\circ$ . However, this finding further confirms the  
 211 reasonability of the overlap scheme since eddies propagate slowly. The propagation  
 212 speed is usually obtained by dividing the distance of two eddy centres (either the peak  
 213 or geometric centre of the boundary) by the time step. Supposing an eddy remains  
 214 relatively still in two successive time steps, the change in the eddy centre is limited.  
 215 The largest possible change is a shift of half of the eddy region, equal to the length of  
 216 the influence radius  $L$ . In summary, the largest search distance  $D_t$  is universally set as  
 217 125 km or  $1.25^\circ$ . A recent study also suggested that dipole eddy pairs can travel at more  
 218 than 10 times the speed of Rossby waves (Hughes and Miller, 2017), which validated  
 219 the set parameter  $D_t$ .

220 For  $R_t$ , the derivation is much easier. Assuming a circle moves half of its diameter  
 221 (Fig. 3a), the ratio of the intersecting area is approximately 0.39. However, taking the  
 222 development of eddy boundaries and shapes (Chen et al., 2019) into consideration, 0.39  
 223 is too ideal for application in a real scenario. Good performance is achieved in the  
 224 assessment in the northwestern Pacific with  $R_t=0.25$ , which is the recommended value  
 225 on a global scale.



226 In other words, the quick overlap scheme calculates the intersection area of two  
 227 successive eddy bounds within a search distance of  $1.25^\circ$  to assess how relevant they  
 228 are to each other. If the ratio is larger than 0.25 (including 0.25), then the eddies are  
 229 considered the same eddy; otherwise, the eddies do not match and are marked as  
 230 missing for one time step (Chelton et al., 2011; Liu et al., 2016).

231 The complete tracking logic of SEIA is illustrated in Fig. 3b. Notably, when an  
 232 eddy has a state of “-tracked-missing-” in two continuous time steps, all the missing  
 233 information is temporally replaced with the former tracked information and seen as a  
 234 complement state, allowing the tracking procedure to continue. The eddy is considered  
 235 dead with the following state of “-tracked-missing-missing-”, and information will only  
 236 be stored until the last “tracked” time step. For “-tracked-missing-tracked-”, it means  
 237 the eddy is successfully re-tracked, and the missing information, such as the bounds  
 238 and centre, is interpolated from the time-neighbouring information.

### 239 **3. Datasets and code availability**

240 Based on the SLA dataset, five years (2015-2019) of SEIA’s output (Yang et al.,  
 241 2022) are provided without any further processing and available on the repository  
 242 (<https://doi.org/10.11922/sciencedb.o00035.00004>). With input dataset in pre-set  
 243 format, SEIA will output series of eddies as mat (MATLAB) files with incremental IDs,  
 244 for example, AE\_1: the first output AE marked as No.1. Table 1 displays all of the  
 245 information contained in a single eddy. Specifically, the *alive* index denotes the  
 246 detection state of an eddy during the tracking process outlined in Fig. 3b, which will be



247 eliminated from the final output mat file.

248       The SEIA code is open-source and runs in MATLAB, allowing anyone to detect  
249 and track global eddies (<https://github.com/Yk-Yang/SEIA>). Regional- and global-  
250 oriented programs are both provided to address the eddy geographical boundary-  
251 crossing issue. For regional eddy identification with limited geographic range, an eddy  
252 will be marked as emerging (vanished) once it enters (exits) the regional boundary. The  
253 equator is a natural boundary since eddy generation is highly connected to the Coriolis  
254 force, over which eddy cannot cross. The land barrier makes  $0^\circ$  longitude a proper  
255 boundary in the north hemisphere, and the  $\sim 67^\circ$  W longitude for the south hemisphere  
256 since it has the shortest distance across the sea (from Cape Horn to the Antarctic  
257 continent). That is to say, under statistical level, eddies detected by global-oriented  
258 SEIA cannot cross the  $\sim 67^\circ$  W longitude in the south hemisphere. However, this setup  
259 simplifies and reduces data operation and manipulation. As a result, for global-oriented  
260 eddy identification, SLA dataset are separated into north and south hemispheres.  
261 Furthermore, regional-oriented identification is recommended for those eddy-related  
262 studies off the Cape Horn.

#### 263 **4. Application and validation**

264       Given the input dataset of SLA, SEIA outputs a series of eddies with universal  
265 information of the existing time, boundary, centre and radius. As detailed in Fig. 1, all  
266 eddies of SEIA are mononuclear. However, multinuclear eddies are a key consideration  
267 in related studies (Yi et al., 2014; Li et al., 2014; Cui et al., 2019, 2020), so the multi-



core structure (McS) index is added to the output of SEIA. The McS index indicates whether an eddy belongs to the same closed contour as other eddies (Table 1), which ensures that researchers can study the splitting and merging of eddies.

Executing the code on five years (from 2015 to 2019) of SLA data generates 278,630 AEs and 274,351 CEs, with a nearly 1:1 ratio. The distribution of the eddy ( $\geq 10$  days outside the tropics) interior occupation in each  $1^\circ \times 1^\circ$  square is shown in Fig. 4a. Most eddies occur in the mid-latitude and western boundary current regions, while relatively few occur at lower and higher latitudes, which is consistent with previous studies highlighting active eddy regions, such as the Gulf Stream, Kuroshio and Southern Ocean (Chelton et al., 2007; Chelton et al., 2011). Fifty percent of eddies fall in the duration span ranging from 10 to  $\sim 34$  days and radius span ranging from  $\sim 22$  to  $\sim 76$  km (Fig. 4b). The radius gradually increases with duration for eddies globally, among which the eddies in the Pacific exhibit this relation the most, and those in the Atlantic and Indian feature larger increasing amplitudes (Fig. 5). Notably, the radius of global eddies tends to decrease when the duration exceeds approximately 220 days.

Global eddy trajectories are characteristic of a strong tendency toward westward propagation at speeds similar to the westward phase speeds of long dispersive Rossby waves (Chelton et al., 2007). Figure 4c illustrates the similar connection between the eddy westward propagation speed calculated from SEIA and the nondispersive long Rossby wave speed. Furthermore, the analysis in Ni et al. (2020) suggested that the speed of westward eddy propagation equals the speed of baroclinic Rossby waves at



approximately 25°N, which separates the global ocean into a low-latitude anisotropic wavelike regime and a high-latitude isotropic turbulence regime. The results in Figure 4c indicate a turning latitude of approximately 30°N, confirming the above theory to some extent. On the other hand, eastward propagation occurs in the Kuroshio, Gulf Stream and Southern Ocean, where eddies are advected eastward by strong depth-mean flow (Chelton et al., 2011; Klocker and Marshall 2014; Ni et al., 2020). Taking the Southern Ocean as an example, nearly 18% of eddies exhibit eastward movement, unlike the dominant propagation trend in the global ocean, while only approximately 12% of eddies exhibit westward movement (Fig. 6). More long-lived eddies are active in the southern Indian part of the Southern Ocean, as detailed in Frenger et al. (2015), which again verifies the reliability and reasonability of the output from SEIA.

## 5. Summary

This study aims to construct a scale-selective eddy identification algorithm for the global ocean named SEIA, based on closed SLA contours, whose basic idea is that isotropic sea surface changes are regulated by eddies. SEIA includes two components, i.e., the detection and tracking process, with two improvements compared to other algorithms.

In the detection process, the standard of the determination of the eddy boundary varies in previous studies utilizing the closed contour idea. The given thresholds for the outermost contour are arbitrary and subjective. One of the two improvements of SEIA is that the given threshold is replaced by a scale-selective scheme, which provides the



310 upper and lower limits of grid points based on data resolution and eddy spatial scale.  
311 Its essence is that the radius of the eddy cannot be smaller than the data resolution or  
312 larger than half of the spatial scale of the mesoscale under the assumption of a quasi-  
313 circle. Based on the mononuclear framework, a boundary of the eddy that is as large as  
314 possible in the span will be obtained, which is consistent with the fact that eddies  
315 continue evolving constantly. Scale-selective conditions can capture more information  
316 without overestimating the eddy region.

317 The second improvement of SEIA is a novel, simple but effective tracking scheme.  
318 Chelton et al. (2007) found that the propagation speed of eddies was related to the speed  
319 of long nondispersive Rossby waves, whose maximum value is approximately 0.16 m/s  
320 outside the tropics. Thus, the propagation distance of eddies is limited to daily datasets.  
321 Therefore, an overlap scheme is applied to cluster eddies in two successive time steps  
322 into the same eddy by calculating the intersecting ratio of eddy regions. Assessment of  
323 the scheme in the northwestern Pacific indicates good performance, with successful  
324 adaptation to various complex dynamic environments. Five years of eddy product and  
325 source code of SEIA are available and accessible for anyone without restriction.

326 The analysis of the output from SEIA is consistent with previous studies on the  
327 global distribution of eddies, propagation speed and characteristics of eddy paths in the  
328 Southern Ocean, which confirms SEIA's reasonability and reliability. Among all the  
329 eddy identification algorithms, none can claim to be the best. SEIA is constructed to  
330 make small progress to better assist the related study of mesoscale eddies. Currently,





331 deep machine learning is gradually being applied to the eddy identification process, and  
332 we expect to follow this approach in future research.

333

### 334 **Author contribution**

335 YY designed the formulation of overarching research goals and aims, developed  
336 the methodology, programmed the codes, generated the SEIA datasets and wrote the  
337 initial draft. LZ and QW verified the research protocol and supervised the research  
338 activity planning and execution. QW assisted with the analysis of the results and LZ  
339 revised the draft.

### 340 **Competing interests**

341 The authors declare that they have no conflict of interest.

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508 Table 1. Information contained in a single eddy detected by the SEIA.

Term	Illustration
<i>time</i>	Existing dates of an eddy
<i>bound</i>	Scale-selective SLA contours that seen as eddy boundary
<i>center</i>	Peaks of SLA: maximum grid points for AE and minimum grid points for CE
<i>radius</i>	Mean distance between <i>center</i> and all the <i>bound</i> grid points
<i>McS</i>	Index indicating the state of multi-core structure: 1 means the existence of such state and 0 the opposite
<i>alive</i>	Temporary index indicating the state of an eddy (Exists only during program operation): 1 means successful detection of an eddy, 0 means death (ready to be output), and -1 means a successful detected eddy ‘missing’ in the latest time step

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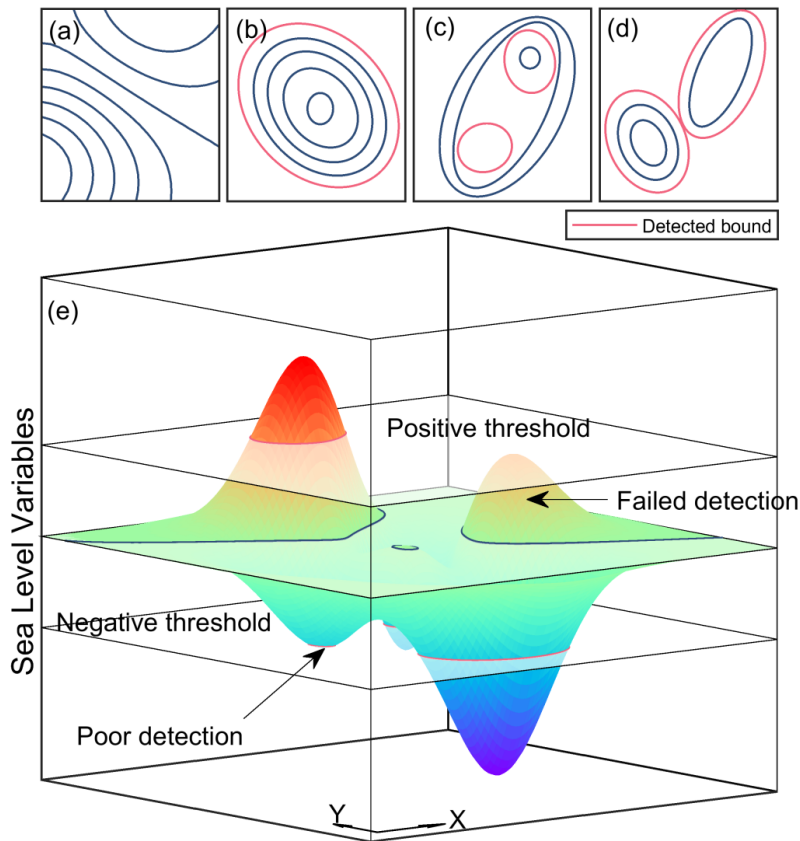
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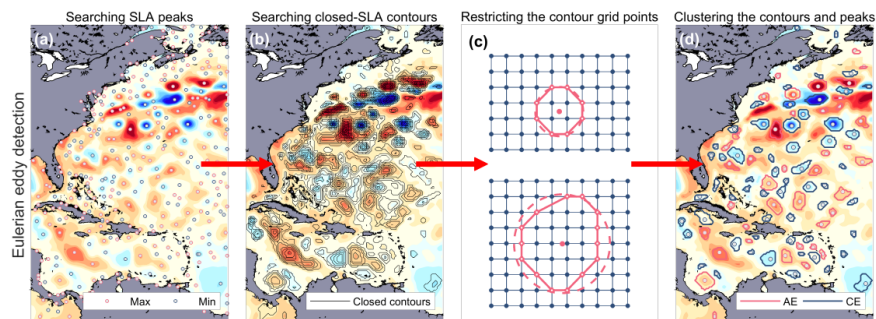
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**Figure 1.** Contour types of (a) simple open, (b) simple closed, (c) concentric closed and (d) intersecting closed contours in Krantz et al. (1999). Red lines indicate the detected boundaries of eddies under each type; notably, the intersecting closed contours in (d) are detected as two separate eddies. (e) Diagram of eddy identification algorithms using a given threshold to determine the outermost contour or boundary of an eddy. The value of the schematic sea level variable is modified from the built-in function Peak from MATLAB 2018a.

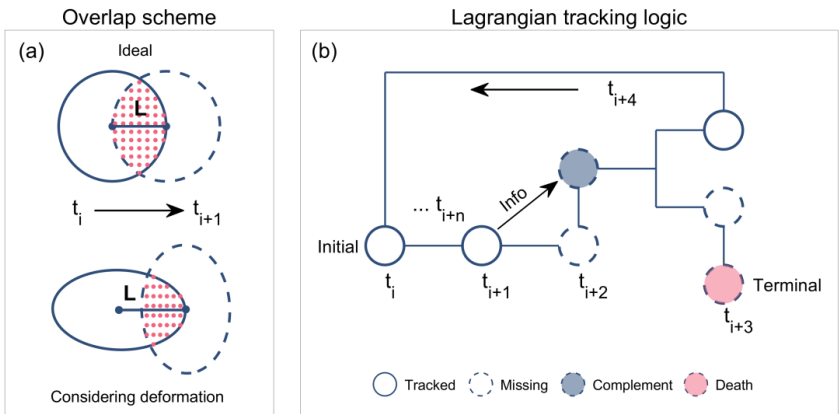
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**Figure 2.** Four steps of Eulerian eddy detection in the northwestern Atlantic. (a) Searching for SLA peaks in a moving  $3 \times 3$  grid window: red and blue dots are the maxima and the minima, respectively. (b) Searching for closed SLA contours. (c) Restricting the contour grid points based on the one-core policy and the initial spatial filtering result. (d) Clustering the closed contours with peaks into eddies.



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546 **Figure 3.** (a) Overlap scheme used to judge whether eddies are identical in two  
547 successive time steps. (b) Diagram of Lagrangian tracking logic: the black solid circle,  
548 black dotted circle, blue dotted circle and pink dotted circle represent successfully  
549 tracked, missing, complemental and dead eddies, respectively. The oblique arrow  
550 indicates that the missing eddy in  $t_{i+2}$  inherits eddy information from eddy in  $t_{i+1}$  to  
551 ensure the continuity of tracking, while the horizontal arrow represents a subsequent  
552 time step.

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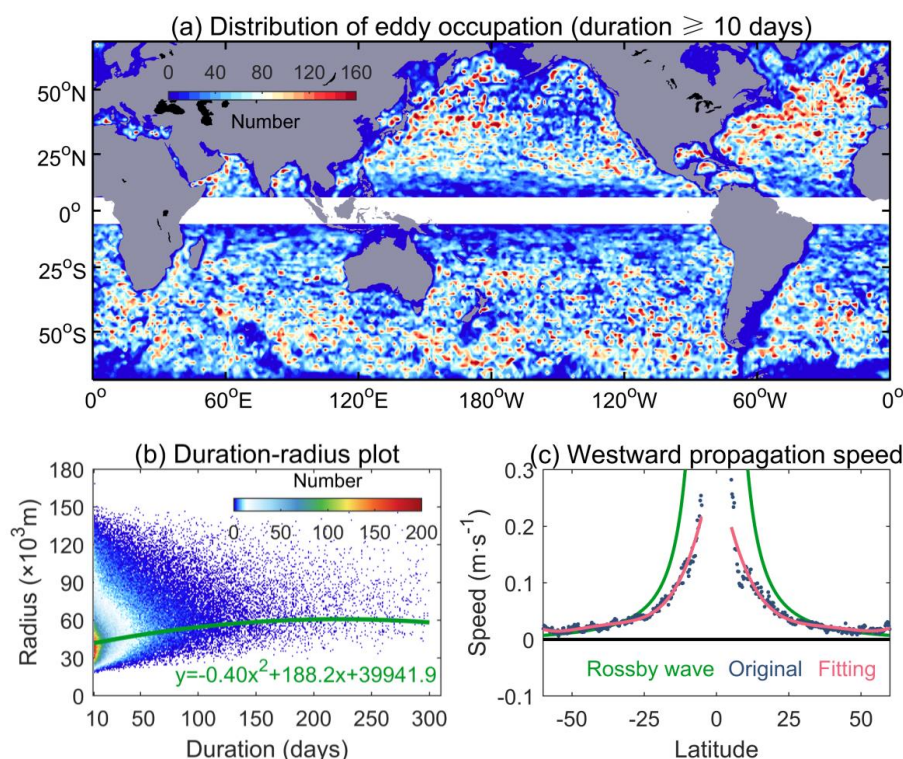
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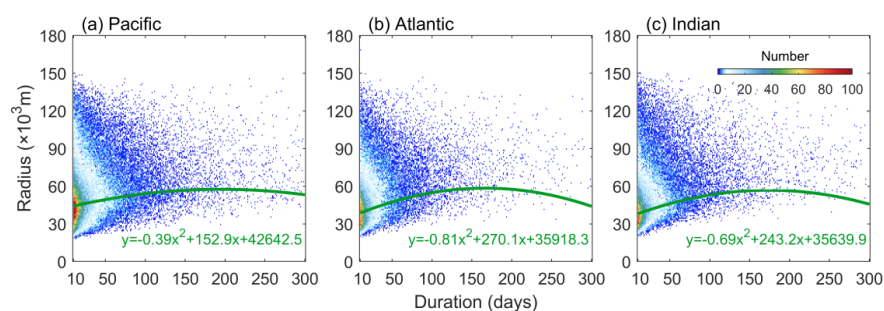
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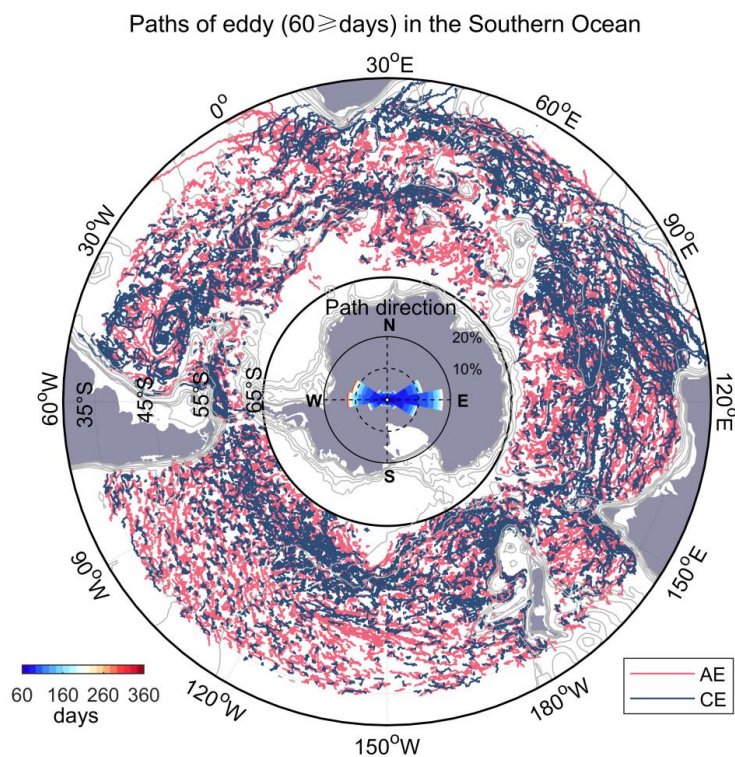


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561 **Figure 4.** (a) Distribution of the eddy interior occupation in each  $1^\circ \times 1^\circ$  square from  
 562 2015 to 2019. (b) Duration-radius plot of the results from SEIA in the global ocean.  
 563 Colours represent certain ratios between the duration and radius; increasingly blue or  
 564 red colours indicate lower or higher ratios, respectively. The green line is a quadratic  
 565 fitting curve. (c) Comparison between the global eddy westward propagation speed  
 566 (blue dots;  $\text{m/s}$ ) and the nondispersive long Rossby wave speed (green line;  $\text{m/s}$ )  
 567 evolving by latitude. The red line is the fitted curve of the original eddy westward  
 568 propagation speed. Notably, only those eddies propagating westward are taken into  
 569 consideration.



**Figure 5.** Duration-radius plots of the results from SEIA in the (a) Pacific, (b) Atlantic and (c) Indian Oceans. Colours represent certain ratios between the duration and radius; increasingly blue or red colours indicate lower or higher ratios, respectively. The green line is a quadratic fitting curve.



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581 **Figure 6.** Paths of eddy ( $\geq 60$  days) in the Southern Ocean: AEs: red lines, CEs: blue  
 582 lines. The grey lines are the 3500-, 2500-, 1500-, 500-, 300-, 200- and 100- m isobaths.  
 583 The eddy rose map shows the ratio of the propagation direction in 16 azimuths. The  
 584 magnitude of shading indicates the lifetime of the eddy; increasingly red or blue colours  
 585 represent longer or shorter lifetimes, respectively.

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