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## Supplemental Material

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### **Interior Water-Mass Variability in the Southern Hemisphere Oceans during the Last Decade**

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# 1 **Supplementary Material**

2

## 3 **1.Uncertainty computation and MLD smoothing**

4

5 Data-based studies, mainly those with gridded fields, contain several sources of uncertainty:  
6 sampling and interpolation error and uncertainties due to the residual of estimations are among the  
7 most common ones. The error derived from sampling bias is difficult to estimate, however, the map  
8 of percentage of variance (Figure 2) works as a good proxy of this error, showing the latitudinal  
9 gradient in the data availability. The uncertainty associated to the volume trend computation is  
10 given by the significance level. In our discussion we only took in account the  $\sigma$ - $\tau$  classes with trends  
11 significant at 95%, which are represented by the dots in Figure 4b, c and d.

12

13 After performing several preliminary tests, we have confirmed that uncertainty in the subduction  
14 rate is the most important source of error among all the computations performed in this study.  
15 Moreover, of the three components of subduction, it is the lateral induction due to the horizontal  
16 gradient of the MLD which accounts for most of the error. We verified, by computing the MLD  
17 with different criteria, that the detection method of the MLD used in this study was the more  
18 accurate for our data set and region. Afterwards, we identified the two processes that are  
19 introducing the main variability to the MLD computation: the MLD interpolation and the  
20 smoothing.

21

22 To account for the MLD variability due to the interpolation process, we computed the mapping  
23 error as the standard deviation between three different interpolation methods as done in Pellichero  
24 et al. (2018): from the previously computed MLD of each single profile we carried out objective  
25 mapping following (i) the method of Kirill Pankratov,  
26 (<http://globec.whoi.edu/software/saga/objmap.m>) and (ii) objective analysis based on Barnes

27 technique (Barnes 1964). In addition (iii) we obtained the MLD from previously gridded fields of  
28 temperature and salinity obtained from ISAS. The error in MLD mapping was then propagated to  
29 the total subduction computation and the resulting error bars are shown in Figure 4a of the main  
30 manuscript.

31

32 The second important source of variability of the subduction is the smoothing of the MLD.  
33 Different methods as Gaussian filter and convolution were tested in this study. The choice of the  
34 running mean was based on the best compromise between reliability and physical meaning that this  
35 method provided. While the results obtained by different methods were somehow comparable, one  
36 of the key factors to chose the running mean was that the other possibilities propagated the blank  
37 area given by the land mask (NaN in the MATLAB language), what lead to an important loss of  
38 information.

39

40 In order to set the adequate moving window of the running mean, we performed several tests. The  
41 lateral induction (the term of subduction which is the most sensitive to the MLD gradient) was  
42 computed using ISAS and ECCOv4 without and with only MLD smoothing with different moving  
43 windows for the running-mean. The result is shown in Figure S1.

44

45 As expected, and in contrast with ISAS, the main lateral induction structures and magnitude from  
46 ECCOv4 does not show an important change with or without MLD smoothing. ISAS MLD has to  
47 be considerably smoothed in order to remove all the fine-scale structures. These structures comprise  
48 the unresolved scales (by the Argo network) and the lack of consistency in ISAS between the MLD  
49 gradient and the velocity field. On the contrary, in ECCOv4, the MLD gradient is consistent with  
50 the velocity field, and only a short moving window is needed to remove the small-scale features  
51 observed around the Drake Passage and other specific locations, mainly at the southern border of  
52 the domain.

53

54 Figure S1 suggests that the magnitude and dominant structures of the induction as computed from  
55 ECCOV4 can be a reference to validate the moving window chosen for the running mean applied on  
56 the MLD from ISAS. Panel c) shows similar structures as those found in ECCO (panels e and f)  
57 while shorter moving windows in ISAS result in a too noisy induction also characterized by too  
58 high magnitude.

59

60 It is worth noticing that in both of the modeling studies of Langlais et al. (2017) and Downes et al.  
61 (2017) subduction appears to be a patchy field due to chaotic behaviors of the ACC and stationary  
62 Rossby waves. With the Argo dataset, the resolution ( $3^{\circ} \times 3^{\circ}$ ) is not enough to resolve these small-  
63 scale structures and therefore in this study we focus only in the medium-large scale processes.

64

65 The effect of the moving-window size on the resulting subduction magnitude from ISAS along a  
66 zonal section at  $40^{\circ}\text{S}$ , can be seen in Figure S2. As observed, the subduction magnitude converges  
67 to an asymptotic value, this suggests that after this value, a greater moving window is not necessary  
68 linked to a smoother field. Given this results, and the comparison between ECCOV4 and ISAS  
69 showed in Figure S1, the moving window of the running mean for the MLD was set to  $10^{\circ}$ .

70

71 Smoothing is a very common procedure when working with data, however it is rarely reported in  
72 detail in the literature. Also, smoothing is subjective, as it is not possible to quantify the suitability  
73 of a specific method over all the possible ones, rather, the criterion is often chosen as a trade off  
74 between the preservation of the main characteristics of the given field and the possibility to obtain  
75 reliable, robust and physically consistent conclusions.

76

77 Here we have learned that the degree of smoothing, even if it does not change significantly the  
78 spatial patterns of the subduction terms, has a pivotal effect in their magnitude and therefore it is

79 indispensable to carefully describe it. In addition, it has been shown how in the case of this study,  
80 this smoothing is necessary to find physically consistent patterns, otherwise undetectable.

81

82 Finally, the uncertainty in transformation is mainly correlated to that of subduction. The remaining  
83 source of error in the transformation computation is given by the residual of the linear regression  
84 from the linear equations system which is indicated in the text to be of the order  $10^{-11}$  with  
85 maximum at  $10^{-5}$ .

86

## 87 **2. Comparison ISAS vs ECCOv4: Lateral Induction computation**

88

89 As previously stated, subduction is a critical computation with a high uncertainty. Since  
90 transformation relies on volume trend and subduction, it is important to ensure the robustness of our  
91 results regarding the subduction field. To this end, in addition to the Figure S1, which shows  
92 important aspects of the smoothing process, we have computed induction with ECCOv4 following  
93 two different methods: (I) by obtaining the induction field for every single month over the period  
94 2006-2015 and time-averaging the final induction (fine temporal resolution). (ii) Following the  
95 same procedure as with ISAS, that is, computing induction for every climatological month and then  
96 time-averaging the final induction (coarse temporal resolution). The deepest MLD over the entire  
97 period was chosen in both cases. The resulting induction fields together with the lateral induction  
98 computed from ISAS climatological monthly fields are shown in Figure S3. In all the cases we used  
99 the same smoothing scale ( $10^\circ$  of moving window).

100

101 Despite the differences between the subduction from ISAS (panel a) and ECCOv4 (panel b and c),  
102 the main patterns of subduction/obduction persist in both cases. When computed with ECCOv4, the  
103 two induction fields (panel b and c) show very similar structures and magnitude; the finely-resolved

104 induction (panel c) presents slightly more spatial variability than the climatologically computed  
105 induction (panel b).

106

107 Of particular interest are the dipolar, or rather, multi-polar structures, between the East of Australia  
108 and New Zealand, and in the central South Pacific. These regions of strong subduction/obduction  
109 are present in all three panels of Figure S3 which increases the confidence in our computations.  
110 Their location coincides with the narrowing of the main ACC flow (blue lines) and its fronts (black  
111 lines) which suggests that it could be a topographically-induced feature due to the standing  
112 meanders of the ACC rather than MLD-gradient derived. In fact, these dipolar structures are  
113 somehow compensated by the MLD gradient. When computed with ECCOv4, the lateral induction  
114 obtained without smoothing shows a much weaker dipoles (Figure S1d). The dipoles become more  
115 evident with the increase of the MLD smoothing level (Figure S1e, f), while the velocity field is  
116 always the same.

117

### 118 **3. Components of the transformation term:**

119

120 Figure S4 shows a more detailed view of the relative contribution of the isopycnal and diapycnal  
121 transformation components, as well as the exchange flow in  $\sigma$ - $\tau$  coordinates. We can see that  
122 isopycnal transformation dominates over all the AAIW density layers (highlighted with red squares)  
123 while diapycnal transformations are key for the total formation within the SAMW range (marked  
124 with black squares). The exchange flow becomes important for the spicier varieties of the SAMW  
125 and for the less spice IW.

126

127 Figures:

128

129 Figure S1

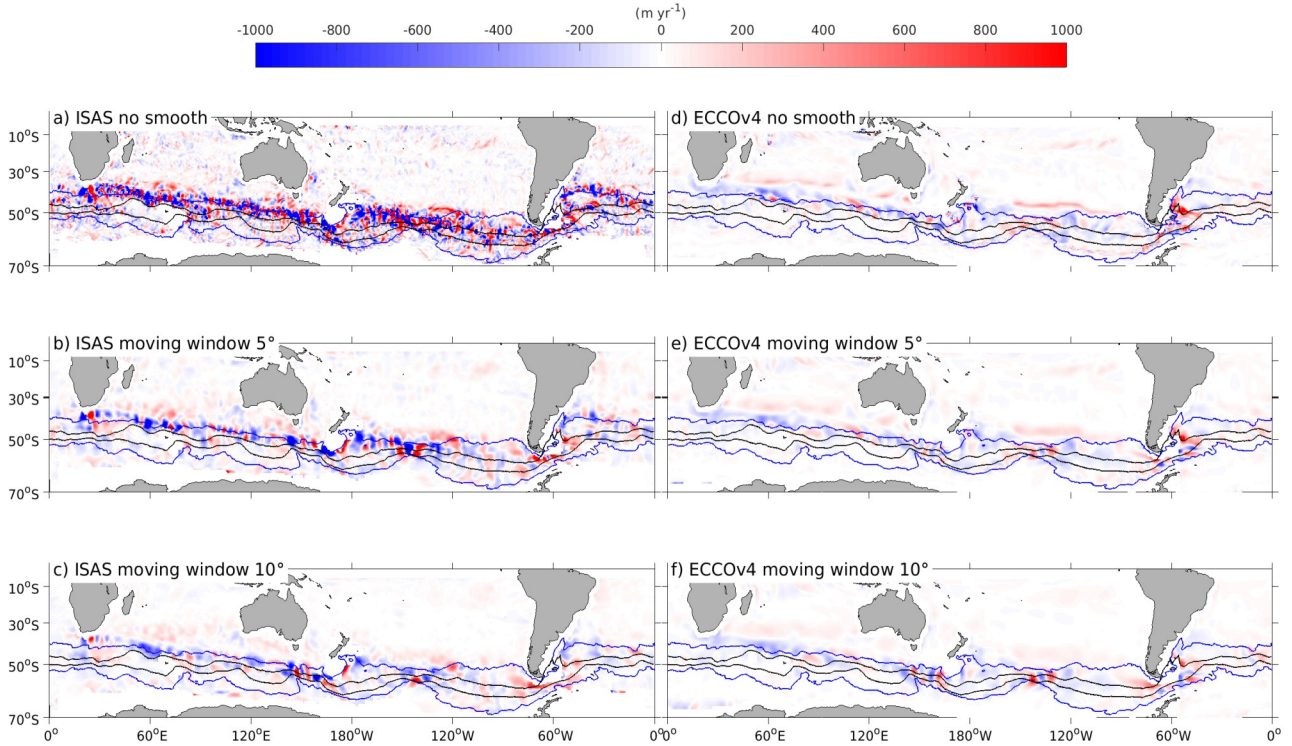


Figure S1. Lateral induction computed from (a-c) ISAS and (d-f) ECCOV4 by time-averaging the monthly climatological data. Different smoothing was applied in each row. A,d) No smooth was applied to the MLD field. (b,e) the running mean smoothing method with a moving window of  $5^\circ$  was applied to the MLD. (c,f) the running mean smoothing method with a moving window of  $10^\circ$  was applied to the MLD. In all the panels blue contours represent the northern and southern boundaries of the ACC computed as the outermost close contours of sea surface height through the Drake passage. The black lines inside are the Subantarctic and Polar fronts from North to South respectively as computed by (Sallée et al. 2008).

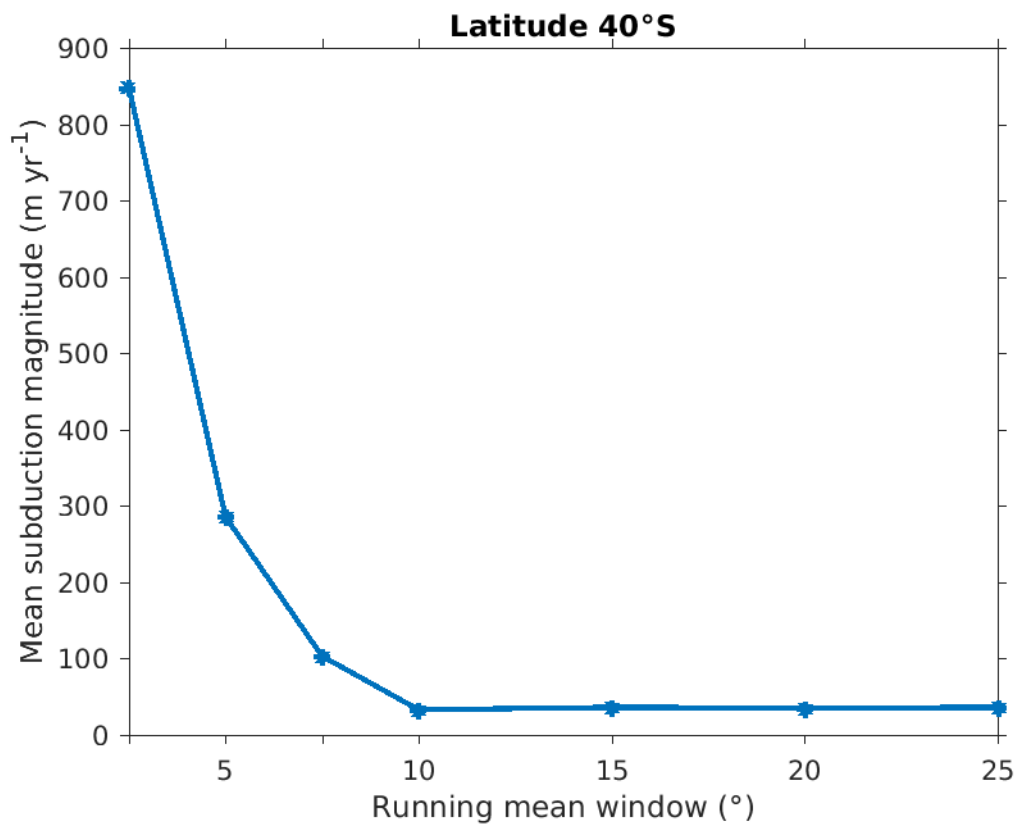
131

132 **Figure S2**

133

134

135



*Figure S2. Average relation between the degree of the smoothing (given by the window of the running mean) and the magnitude of subduction along a zonal section at 40°S*



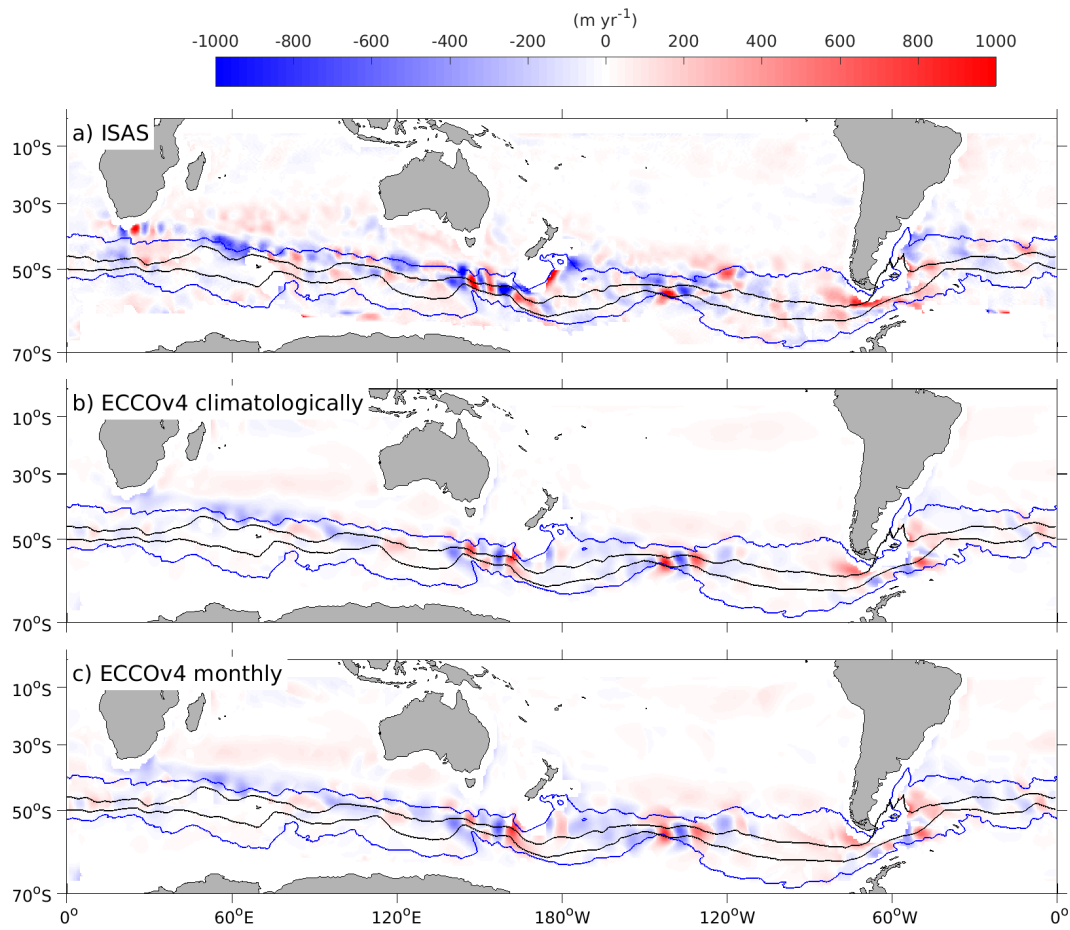


Figure S3: Lateral induction term of subduction as computed with a) ISAS by averaging the climatological induction, b) ECCOv4 in the same way as in panel a and c) ECCOv4 by time-averaging the computed monthly induction over the 10-years period between 2006 and 2015

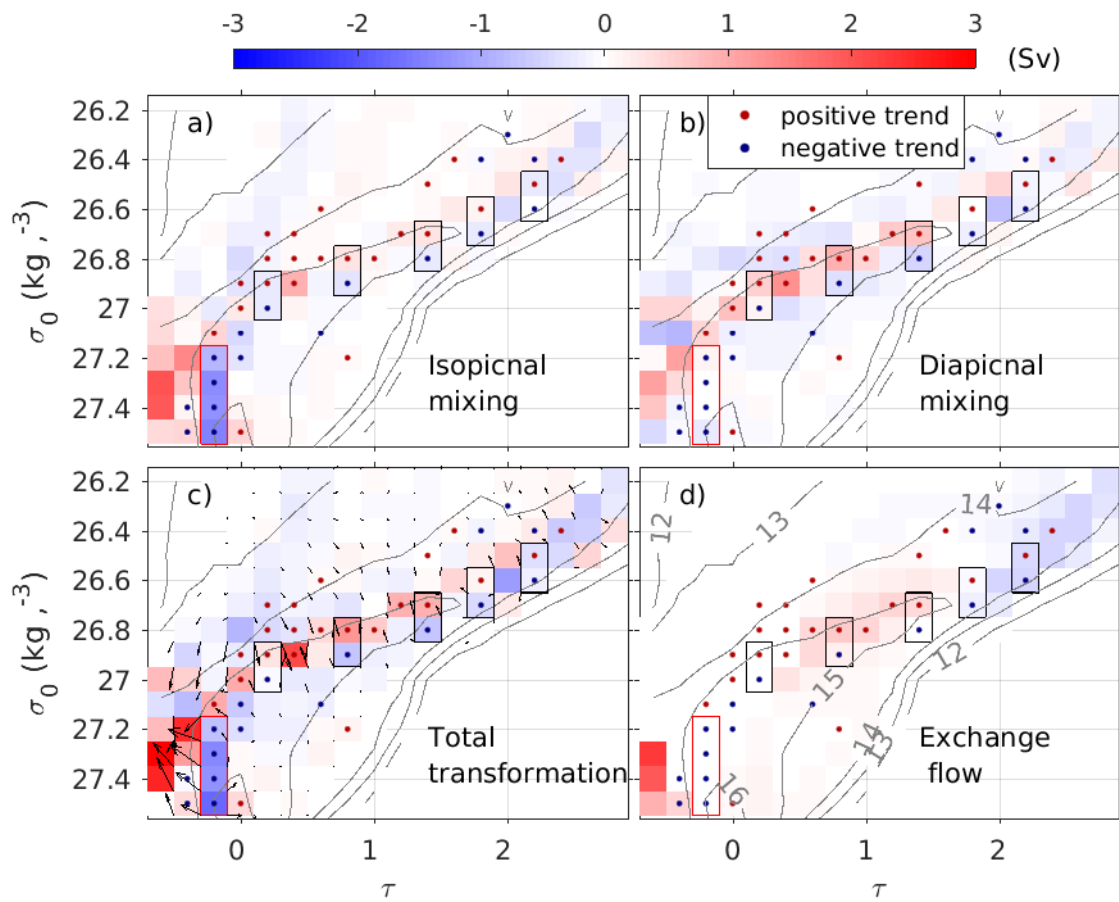


Figure S4. a) Isopycnal, b) diapycnal transformation (arrows) and formation (color coded) and c) the sum of both components. d) Exchange flow across the domain's limits where red indicates water into the domain and blue means flow out from the domain. The diagrams are shown in  $\sigma$ - $\tau$  coordinates.