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2 **2 Response to Reviewer #2's comments**

3 This paper aims to investigate the effect of ENSO on global dust emissions, concentration, and
4 deposition. A multivariate predictive model and the Ganger causality test were used to analyze
5 dust-relevant output from 12 CMIP6 models.

6 General Comments

7 A major and critical shortcoming of this paper, in current form, is the lack of detailed presentation
8 of methodology, results, and discussion. The paper seems to be written hastily.

9 **Response:** We thank the reviewer for your comments. We agree with the reviewer that some parts
10 of the manuscript need improvement. We modified the manuscript based on your suggestions as
11 below.

12 2.1 Methodology: Authors only provide a few lines about their approach, and refer readers to
13 their previous works and the supplementary document. However, a brief description of the
14 method should be presented in this section.

15 **Response:** We thank the reviewer for raising this point. We moved part of the supplementary to
16 Section 2.2 to further clarify the Methods used in this study as follows:

17 “We use the following multivariate predictive model (Mosedale et al., 2006; Stern and Kaufmann,
18 2013) to estimate the causal links between the ENSO and dust deposition:

$$19 \mathbf{X}_t = \sum_{i=1}^p \alpha_i \mathbf{X}_{t-i} + \sum_{i=1}^p \beta_i \mathbf{Y}_{t-i} + \sum_{j=1}^m \sum_{i=1}^p \delta_{j,i} \mathbf{Z}_{j,t-i} + \boldsymbol{\varepsilon}_t \quad (1)$$

20 where X_t is the annual mean (or seasonal mean) dust deposition for year t , Y_t is the ENSO index,
21 and $Z_{j,t}$ is the confounding factor j for year t . In the predictive model presented in equation 1, while
22 assessing the effect of Y on X (i.e., the contribution of the term $\sum_{i=1}^p \beta_i Y_{t-i}$ in predicting X), the
23 possible influence of past X events are considered by adding the term $\sum_{i=1}^p \alpha_i X_{t-i}$. Thus, the causal
24 influence of Y on X , if detected, is robust and the impact of past X events are accounted in the
25 analyses. Here, m is the number of confounding factors and $p \geq 1$ is the order of the multivariate
26 predictive model. The optimal order p is computed by minimizing the Schwarz criterion or the
27 Bayesian information criterion (Schwarz, 1978). The optimal orders may be different for each
28 model.

29 Here we take into account the impacts of confounding factors and therefore provide further
30 information of the real-world teleconnections. In the analyses, we use three different confounding
31 factors; hence, m is equal to 3. The noise residuals ε_t and the regression coefficients α_i , β_i and $\delta_{j,i}$
32 are computed by using the multiple linear regression analysis of the least squares method. We
33 detrend and normalize all the climate indices.”

34 2.2 Additionally, authors stated that they studied confounding effects of other climates modes,
35 namely NAO, SAM, and IOD. First, what is the basis for choosing these modes(?), and
36 second, no analysis or sensitivity test regarding this treatment was provided.

37 **Response:** We thank the reviewer for raising this point. We added the following sentences to
38 Section 2.2 to explain the selection of the three factors Indian Ocean Dipole, Southern Annular
39 Mode, and the North Atlantic Oscillation:

40 “The climate modes SAM, the IOD and the NAO are the important sources of global climate
41 variability (Hurrell et al., 2003; Luo et al., 2012; Roxy et al., 2015). For instance, the NAO is the
42 prominent mode of atmospheric circulation variability over the North Atlantic and surrounding
43 regions (Delworth et al., 2016; Hurrell et al., 2003) and variations in NAO are crucial for the
44 environment and society (Hurrell et al., 2003). The IOD affects climate extremes over the Indian
45 Ocean and surrounding areas (Abram et al., 2008; Kripalani et al., 2009; Kripalani and Kulkarni,
46 1997) and might cause severe economic consequences (Ummenhofer et al., 2009). The SAM is
47 the major mode of atmospheric circulation variability in the southern Hemisphere (Cai et al., 2011;
48 Raphael and Holland, 2006). In addition, changes in these modes may affect the variations of
49 ENSO (Abram et al., 2020; Cai et al., 2011, 2019; Le et al., 2020; Le and Bae, 2019). Nevertheless,
50 it is likely that these factors may alter the influences of ENSO on dust activities.”

51 2.3 Finally, what is the basis for choosing these 12 models as data crucial for these analyses are
52 missing in three of them (table S2)?

53 **Response:** We thank the reviewer for raising this point. We selected all the models with accessible
54 dust-related data. Considering the total models is somewhat low (i.e., 12 models), we kept all these
55 models. To our knowledge, dry and wet deposition of dust are key variables which directly affect
56 local environment, thus we selected all models having these data.

57 We add the following sentences to Section 2.1 to clarify this point:

58 “We limited our study to all the models having both dry dust and wet dust data (i.e., there is total
59 of 12 models with accessible dry dust and wet dust data as described in Table S2). Dust deposition
60 on land and ocean surface are important metrics to assess the impacts of dust activities on
61 ecosystems and environment (Bao et al., 2017; Fan et al., 2006; Jickells et al., 2005; Jiménez et
62 al., 2018; Kanakidou et al., 2018; Schulz et al., 2012). Additional data of od550dust and emidust
63 supplied by these 12 models provide further understanding of ENSO impacts on dust activities.”

64 2.4 Results and Discussion: Results were presented and discussed in a highly qualitative manner
65 without any in-depth analysis as required in a manuscript with an archival value. Authors
66 only reported the fraction of total “affected area” over ocean and land, but this number alone
67 is not useful in understanding the true impact of ENSO on dust activities in different regions
68 of the globe.

69 **Response:** We thank the reviewer for raising this point. The “affected area” provide important
70 information on the scale of ENSO impacts. The purpose of this work is to provide simple and
71 robust conclusions about the causal effects of ENSO on global dust activities. To our knowledge,
72 these analyses are lacking, and the results described here have not been shown before. We should
73 note that we also provided information on the impacts of ENSO on dust activities at regional scale,
74 as well as the consistency between models in simulating the connection between ENSO and dust
75 activities.

76 2.5 As expected, individual models show drastically different results (figures 4-6), but
77 conclusions of the paper were based only on the ensemble mean results with minimal
78 discussion about the difference between models. Note that the chosen models use different
79 dust emissions and deposition, as well as dust size partitioning schemes, so ensemble mean
80 results must be interpreted with caution.

81 **Response:** We thank the reviewer for raising this point. The use of ensemble mean is widely used
82 and is expected to reduce the uncertainty related to the connection between ENSO and dust
83 activities. In fact, despite using different aerosol model, there is high agreement (i.e., denoted by
84 stippling in Figures 1 and 2) between models for the results. Our conclusions are mainly based on
85 this high consistency.

86 We partly discussed the difference between models in Section 4 as below:

87 “Regarding the consistency across models, the response of dust emission to ENSO is much
88 stronger in the models INM-CM5-0, MIROC-ES2L, and UKESM1_0_LL compared to other
89 models (Figure 6). This difference might be due to the use of different dust schemes and soil
90 properties in this model which lead to higher dust emissions (Mulcahy et al., 2020; Zhao et al.,
91 2022). As models use different parameters to estimate dust emissions (Thornhill et al., 2020), this
92 discrepancy leads to low consensus across models in modeling the response of dust emissions to
93 ENSO (Figures 3 and 6).”

94 2.6 Finally, several conclusions of the work are not supported by the current results, for example,
95 “ENSO may initiate dust activities in (line 93)”, “dust deposited in the South Pacific and
96 the Southern Ocean might be originated from central Australia and southern South America
97 (line 117)”, “weak causal impacts of ENSO on regional dust emissions of major dust sources
98 (Figure 3) may indicate the important role of human influences in igniting local dust
99 activities... (line 132)”.

100 **Response:** We thank the reviewer for raising this point. We removed the lines 93 and 117 to avoid
101 confusing the readers.

102 We rewrote line 132 as this line serves as a discussion rather than a conclusion as below:

103 “Substantial influences of ENSO on dust emission over central Australia (Figure 3) suggest an
104 agreement with earlier work (Marx et al., 2009), while we observe weak causal impacts of ENSO
105 on regional dust emissions of major dust sources (Figure 3).”

106 “Previous studies indicate the important role of human influences in igniting local dust activities
107 (Duniway et al., 2019; Webb and Pierre, 2018).”

108 Specific Comments

109 2.7 Line 15 and all other places through the manuscript: Caution must be practiced with the term
110 “concentration” as the relationship between dust concentration and dust AOD depends on
111 the pre-defined and assumed dust particle size distribution, which is different in different
112 models.

113 **Response:** We thank the reviewer for raising this point. We removed the term “concentration” to
114 avoid confusing the readers.

115 2.8 Line 15 and all other places through the manuscript: Change “transportation” to “transport”

116 **Response:** We thank the reviewer for raising this point. We corrected as your suggestion.

117 2.9 Line 57: “Dry and wet deposition is related to different types of dust and aerosol.” Not clear
118 what authors mean here.

119 **Response:** We thank the reviewer for raising this point. We removed this sentence to avoid
120 confusing the readers.

121 2.10 Line 82: How are “areas affected by ENSO” defined? What are the criteria considered?

122 **Response:** We thank the reviewer for raising this point. We added the following sentence to clarify
123 the criteria for computing the areas affected by ENSO:
124 “In Figure 2b, the areas influenced by ENSO are computed as the areas limited by the cyan contour
125 line as shown in Figures 1 and 2a (i.e., p -value is lower than 0.33 or ENSO is unlikely to exhibit
126 no causal effects on dust activities over these regions).”

127 2.11 Line 116: Change “original” to “originated”

128 **Response:** We corrected as your suggestion.

129 2.12 Figure 1: What is the significance of studying dry and wet deposition separately in figure 1,
130 as paper provides no insightful comparison between the two?

131 **Response:** We thank the reviewer for raising this point. As we introduced in Section 2.1, dry and
132 wet deposition are related to different processes of dust deposition. Hence, direct comparison is
133 not necessary. We combined the results in 1 Figure as these dust deposition processes are
134 complementing for each other.

135 2.13 Figure 2(b) and S2: There two figures don’t provide any addition information beyond just
136 one number mentioned in the text, so they should be removed.

137 **Response:** We thank the reviewer for raising this point. We think these Figures provide quick
138 summary and illustration for the Text. Hence, we would like to keep these Figures.

139 2.14 Table S1 and S2: Should be merged into one table

140 **Response:** We thank the reviewer for this suggestion. We would like to keep these Tables
141 separately to avoid a too big Table. It would make the presentation easier.

142 2.15 The following recent publication might be of interest to the authors:
143 <https://acp.copernicus.org/articles/22/2095/2022/>

144 **Response:** We thank the reviewer for this suggestion. We include this publication as a reference.

145 **References**

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