

1 **Unveiling Sulfate Aerosol Persistence as the Dominant Control of the**
2 **Systematic Cooling Bias in CMIP6 Models: Quantification and**
3 **Corrective Strategies**

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16

17 **Abstract**

18 Including sophisticated aerosol schemes in the models of the sixth Coupled Model
19 Inter-comparison Project (CMIP6) has not improved historical climate simulations. In
20 particular, the models underestimate the surface air temperature anomaly (SATa) when
21 anthropogenic sulfur emissions increased in 1960~1990, making the reliability of the
22 CMIP6 projections questionable. Biases in cooling among the models are correlated
23 with sulfate burden and the deposition of sulfur is the process responsible. Accordingly,
24 we define a diagnostic tool, named Sulfur Assessment Metric for Earth system models
25 (SAME), for model evaluation and improvement. We show that the SAME index
26 determines the cooling biases. Reducing the biases to within the observational
27 uncertainty is consistent with a physically plausible SAME of around 1.35 days, which
28 is overestimated by almost all the CMIP6 models. Based on targeting a reduction of
29 SAME, post-CMIP6 improvements to two models are shown to greatly improve SATa
30 reproduction.

33 **1. Introduction**

34 Atmospheric aerosols have rapidly increased since the Industrial Revolution. Over
35 this time period, the total aerosol effective radiative forcing (ERF) was dominated by
36 the sulfate cooling effect, and offset a substantial portion of global-mean forcing from
37 well-mixed greenhouse gases (IPCC, 2023). Without this historical aerosol ERF, the
38 Paris Agreement's target of limiting global warming to 1.5°C above pre-industrial
39 levels would have already been missed in 2015 (Hienola et al., 2018). Similarly,
40 stopping all present-day anthropogenic aerosol emissions is estimated to induce a
41 global-mean surface heating of 0.5~1.1°C (Samset et al., 2018). The year 2024 has been
42 confirmed as the hottest year in human history, and was the first year to breach the
43 1.5°C warming limit (Bevacqua et al., 2025). Moreover, recent years have seen
44 temperature trends accelerate, which may be due to reductions in atmospheric aerosols,
45 especially aerosols produced by commercial shipping (Hansen et al., 2025). Hence, it
46 has been suggested that even small emissions in relatively pristine air have substantial
47 effects, and better constraining the ability of global-climate models to predict aerosol
48 effects may be crucial to obtaining reliable projections.

49 The observed temporal evolution of historical surface air temperature (SAT) is one
50 of the major metrics used for evaluating the performance of climate models. However,
51 the SAT anomalies in the CMIP6 models are systematically lower than was observed
52 for the 1960~1990 period, whereas the CMIP5 models on average track the
53 instrumental record quite well (e.g., Flynn and Mauritsen, 2020). The 1960~1990
54 period is referred to as the “pothole cooling period” (PHC) in our previous study (Zhang
55 et al., 2021a), due to the ‘pot-hole’ shaped dip in SAT at that time, and in this study
56 hereafter. The PHC period is coincident with the so-called Great Acceleration period,
57 in which the human enterprise was boosted remarkably and led to global-scale impacts
58 on Earth System functioning (Steffen et al., 2007). Recent studies hypothesized that
59 aerosol forcing in CMIP6 is stronger than in CMIP5 and is responsible for the
60 suppressed late 20th-century warming (Dittus et al., 2020; Smith and Forster, 2021).

61 The anomalous cooling points towards a problem with the sulfur cycle in recent

62 ESMs or the emissions data (Hardacre et al., 2021; Wang et al., 2021). Considering the
63 importance of the sulfur cycle in historical aerosol ERF, we examine the sulfur related
64 processes in eleven CMIP6 models with aerosol schemes in this study. All the models
65 are forced with CMIP6 historical anthropogenic aerosol emissions (Hoesly et al., 2018),
66 and therefore differences in their sulfate burdens are mainly due to different
67 representations of the sulfur cycle in the models.

68 We will identify the key processes that determine sulfate-burden in these models,
69 and introduce a simple index for measuring the level of activity in the sulfur cycle in
70 the models on the global scale. This index (an effective diagnostic tool for global
71 cycling of atmospheric sulfate) can be easily calculated from time series of global
72 means only, without the need for complex diagnostics of the sulfur-cycle processes. We
73 show that the index is strongly correlated with sulfate burden and anomalous cooling
74 and has a clear physical interpretation that allows each model's sulfur cycle to be
75 calibrated using historical temperature biases.

76

77 **2. Model, data, and method**

78 **2.1 CMIP6 models and data**

79 **Table 1.** Information of the eleven CMIP6 models with aerosol schemes.

Model	Country	Interactive Chemistry	Members	Reference
BCC-ESM1	China	Yes	3	Wu et al. (2020); Zhang et al. (2021b)
CESM2	US	No	11	Danabasoglu et al. (2020)
CESM2-FV2	US	No	3	Danabasoglu et al. (2020)
EC-Earth3-AerChem	European consortium	Yes	2	Döscher et al. (2021)
GFDL-ESM4	US	Yes	3	Dunne et al. (2020)
MIROC6	Japan	No	50	Tatebeet et al. (2019)
MIROC-ES2L	Japan	No	30	Hajima et al.

					(2020)
MPI-ESM-1-2-HAM	Germany	Yes	3	Mauritsen et al. (2019)	
MRI-ESM2-0	Japan	Yes	10	Yukimoto et al. (2019)	
NorESM2-LM	Norway	Yes	3	Seland et al. (2020)	
UKESM1-0-LL	UK	Yes	19	Sellar et al. (2019)	

80

81 Eleven CMIP6 climate models with interactive aerosol schemes are utilized in this
 82 study, including seven models with interactive chemistry and four without (Table 1).
 83 The outputs from two CMIP6 experiments are used: (1) the historical experiment of
 84 climate change over the period 1850~2014, forced by time-varying external forcings
 85 that are based on observations of natural processes (e.g., solar activity, volcanic
 86 eruptions) and human-induced changes (e.g., greenhouse gas, aerosol emissions, land-
 87 use changes). All the available realizations for each model were used to minimize the
 88 uncertainty from internal variability in the climate system; (2) the 1pctCO₂ simulations,
 89 in which CO₂ is gradually increased at a rate of 1% per year. The 1pctCO₂ experiment
 90 is designed for studying model responses to CO₂ and is somewhat more realistic than
 91 rapidly increasing CO₂ such as in the abrupt-4×CO₂ experiment.

92 Model outputs used in this study comprise surface air temperature (SAT) and five
 93 key sulfur-cycle variables: sulfate burden (loadSO₄), sulfate wet deposition and sulfate
 94 dry deposition, sulfur-dioxide (SO₂) wet deposition and SO₂ dry deposition. For these
 95 sulfur-cycle variables, the inter-member variability within the historical experiment is
 96 substantially smaller than that of SAT. The standard deviation of loadSO₄ in PHC
 97 across the 11 CESM2 members is only 4% of its interannual variability, compared to
 98 approximately 21% for SAT. Similar results are also evident in the 19 UKESM1
 99 members, where the standard deviation of loadSO₄ is 3% of its interannual variability,
 100 versus 32% for SAT. Therefore, given the relatively small inter-member variability in
 101 sulfur-cycle variables compared to their interannual fluctuations and to SAT variability,
 102 we utilize the first realization of the historical simulations and neglect inter-member

103 differences for these sulfur-related quantities.

104 The monthly mean SAT from the Met Office Hadley Centre/Climatic Research
105 Unit global surface temperature (HadCRUT) data version 5 from 1850 to 2014 is used
106 for model evaluations (Morice et al., 2021). Considering the lack of long-term reliable
107 observations in polar regions, we focus on SAT changes between 60°S to 65°N and the
108 ‘global’ mean is calculated as the area-weighted mean in this latitudinal belt.

109

110 **2.2 The Sulfur Assessment Metric for ESMs (SAME) index**

111 Atmospheric sulfate concentrations are determined by the emission and oxidation
112 of sulfate precursors, as well as deposition processes. Together these processes make
113 up the atmospheric part of the Earth’s sulfur cycle. Anthropogenic SO₂ emissions are
114 the major source of sulfate aerosol over land in polluted regions. Given that the same
115 anthropogenic SO₂ emissions are used in all the CMIP6 models, most of the differences
116 in simulated atmospheric sulfate concentrations occur due to oxidation of SO₂ and
117 sulfate deposition processes. Since much of the loss of SO₂ occurs locally to its
118 emission source by oxidation and deposition, faster SO₂ deposition is associated with
119 weaker SO₂ oxidation. For sulfate itself, the faster the sulfate deposition rate, the less
120 the sulfate will reside in the atmosphere. That is, both the SO₂ deposition and sulfate
121 deposition are important for the sulfate concentrations in the atmosphere, directly or
122 indirectly.

123 Here we define the **Sulfur Assessment Metric for ESMs (SAME) index**. The
124 SAME index is calculated as the ratio of the sulfate burden anomaly and sulfur
125 deposition anomaly in PHC, relative to preindustrial period, to mitigate the influence
126 of differing model climatology. Sulfur deposition encompasses the deposition fluxes of
127 sulfate aerosol and its major precursor SO₂:

128 $\text{SAME} = \text{loadSO4a} / (\text{DSO4a} + \text{DSO2a})$

129 where:

130 • loadSO4a is the total sulfate loading anomaly in the atmosphere,
131 • DSO4a denotes the total (wet plus dry) sulfate deposition anomaly, and
132 • DSO2a denotes the total (wet plus dry) SO₂ deposition anomaly during the
133 PHC period.

134

135 **2.3 The transient Climate Response (TCR) index**

136 The transient Climate Response (TCR) index is calculated as the mean SAT
137 anomaly of a 1pctCO₂ simulation in a 20-year period centered on year-number 70, by
138 which a doubling CO₂ concentration has occurred. It is an important metric representing
139 CO₂-related historical warming and has been widely used for model evaluations and
140 comparisons (e.g., Bevacqua et al., 2025; O'neill et al., 2016).

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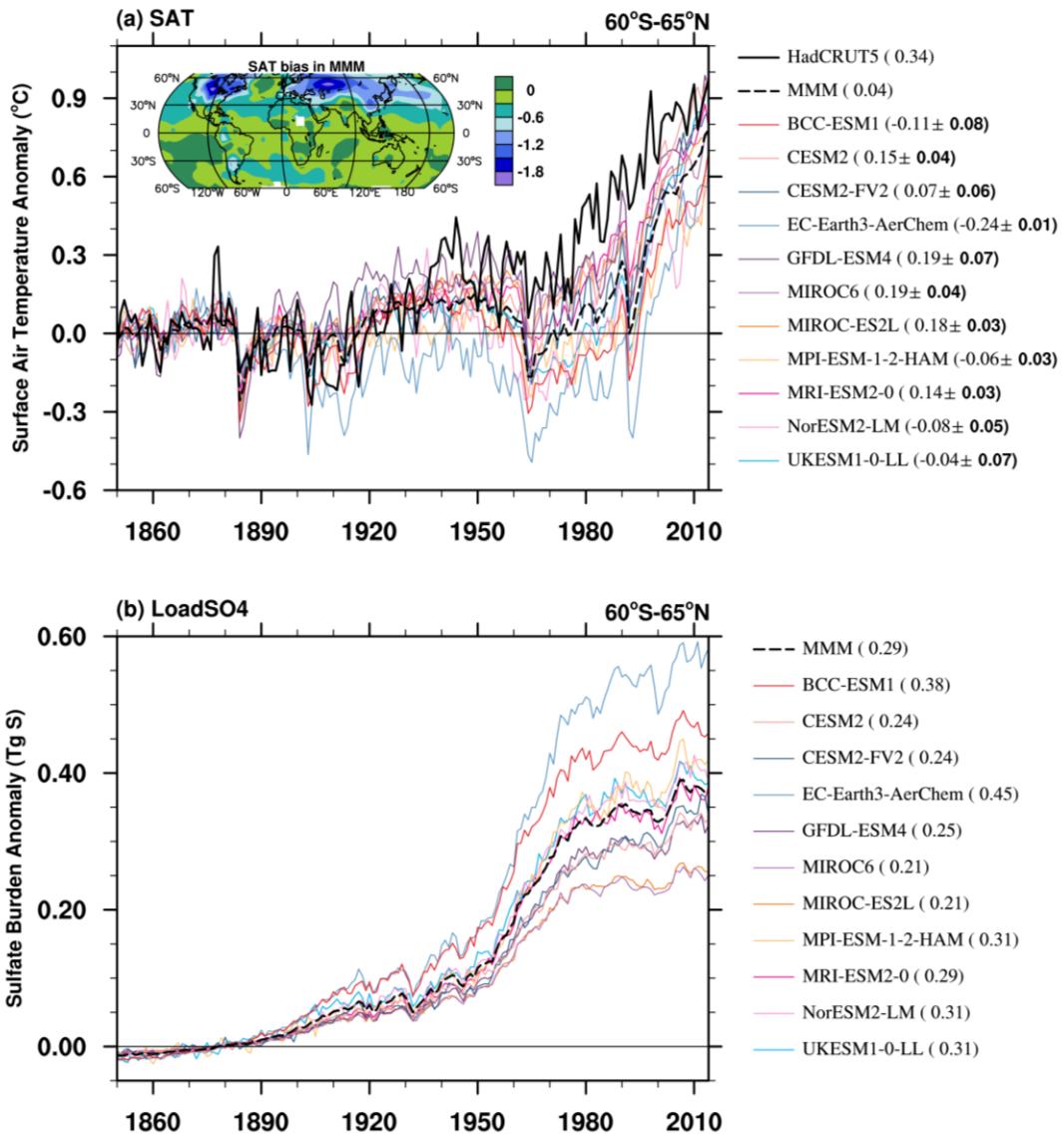
142 **3. Results**

143 **3.1 SAT and sulfate burden**

144 The historical evolutions of near-global mean (60°S to 65°N) SATa in the eleven
145 CMIP6 models with interactive aerosol schemes are shown in Fig. 1a. All the models
146 tend to underestimate SATa since the 1930s. The anomalous cooling in CMIP6 model
147 marked a notable departure from earlier model generations, which can effectively
148 capture the instrumental SAT record with observation falling well within model spread
149 (e.g., Flynn and Mauritsen, 2020; Hegerl, et al., 2007).

150 The cooling bias is most pronounced from 1960 to 1990, i.e., the PHC period. The
151 SATa in PHC is about 0.34°C in the observations. However, the multi-model mean
152 (MMM) SATa in the models is about 0.3°C lower with a large model spread. The SATa
153 ranges from -0.24°C in EC-Earth3-AerChem to 0.19°C in GFDL-ESM4 and MIROC6.
154 Significant cooling is evident across the mid- and high-latitudes of the Northern
155 Hemisphere, as illustrated by the SATa map in Fig. 1a. The sudden drop in SATa in
156 the early 1960s and 1990s may be due to the stronger model responses to large volcanic

157 eruptions, Mount Agung in 1963 and Mount Pinatubo in 1991, than in the observations
 158 (Chylek et al., 2020). The anomalous cooling biases gets smaller later in the
 159 simulations, which is related to the generally high sensitivity of the models to GHG
 160 forcing (Smith and Forster, 2021).



161
 162 **Figure 1.** (a) Historical surface air temperature anomalies (SATa) relative to 1850~1900 mean for
 163 HadCRUT5 (thick black line), the ensemble mean for each CMIP6 model (solid color lines), and
 164 multi-model mean (MMM, dashed black line). The numbers in brackets are the mean results during
 165 the pothole period (1960~1990) together with the inter-member spread for each model. Units: °C.
 166 Panel (b) is the same as panel (a) but for sulfate burden anomalies for the first realization from each
 167 CMIP6 model (solid color lines) and MMM (dashed black line). Units: Tg S yr⁻¹.

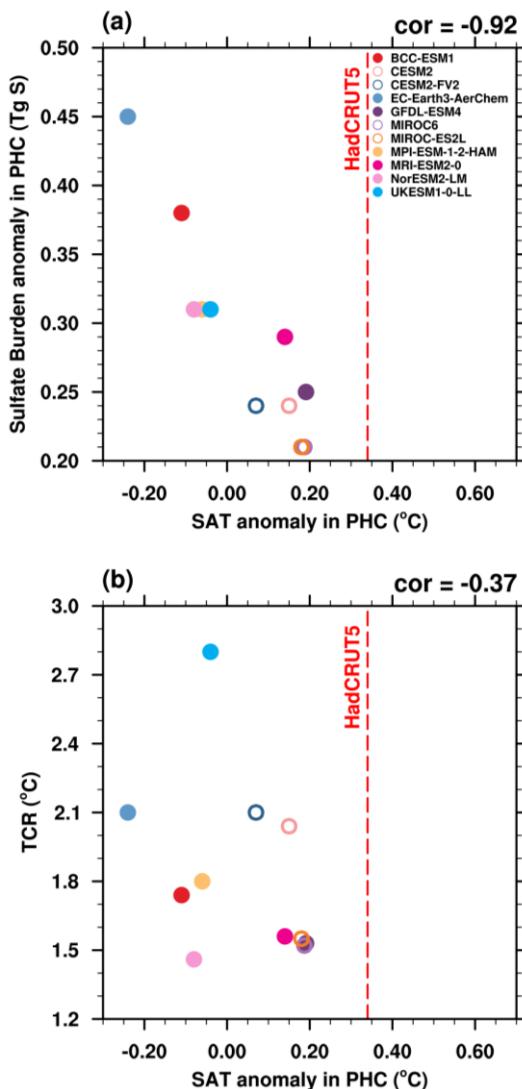
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169 The PHC coincides with increased anthropogenic emissions, particularly of sulfate

170 precursors such as SO₂ (Zhang et al., 2021a). Global emissions of SO₂ grew steadily
171 after the 1950s and peaked in the 1970s at 180Tg yr⁻¹, which is about 3.6 times the
172 1950s' emissions (Hoesly et al., 2018). As the main precursor of sulfate, the growing
173 emission of SO₂ led to the accumulation of sulfate in the atmosphere, which interrupted
174 a decades-long warming trend via the cooling effects of sulfate aerosols on climate,
175 even though carbon-dioxide emissions continued to rise (Wilcox et al., 2013). Because
176 of the emission control policies in Europe and North America (Hand et al., 2012;
177 Vestreng et al., 2007), such as the Gothenburg Protocol (Eb, 1999) and the 1990 Clean
178 Air Act Amendments in the U.S. (Likens et al., 2001), global anthropogenic SO₂
179 emissions were suppressed after the 1980s and SAT started to increase rapidly in the
180 observations (Aas et al., 2019). However, anthropogenic SO₂ emission continued to
181 increase over Asia due to industrial developments, although they have also decreased
182 since 2006 in East Asia (Wang et al., 2021). Some of this decrease in SO₂ emissions at
183 the beginning of the 21st century is not well represented in the CMIP6 emission
184 inventory. But it is outside of the PHC period and the impact on SAT reproduction is
185 beyond the scope of this paper.

186 In the 11 CMIP6 models, sulfate concentrations increased rapidly during the PHC
187 period (Fig.1b). The intensified emission of anthropogenic SO₂ mainly comes from
188 industries and the energy-transformation sector (e.g., Ohara et al., 2007; Vestreng et al.,
189 2007). The SAT anomalies simulated by CMIP6 models are systematically lower than
190 observations during the PHC period, indicating an excessively strong sulfate-induced
191 cooling effect in CMIP6 models. The sulfate burden is the lowest in MIROC models
192 (0.21 Tg S) with the smallest cooling bias (0.15°C lower than HadCRUT5), and is
193 doubled in EC-Earth3-AerChem (0.45 Tg S) with the largest cooling (0.58°C lower
194 than HadCRUT5). Generally, the models with higher sulfate burdens anomalies also
195 tend to underestimate SAT anomalies the most. As shown in Fig. 2a, the correlation
196 coefficient between anomalous sulfate burden and SAT during the PHC is -0.92,
197 significant at the 1% level using a Student's *t*-test. Interactive chemistry may have an
198 impact on sulfate formation and affect the sulfate aerosol burdens in the atmosphere

199 (Mulcahy et al., 2020). As shown in Fig.2a, models with interactive chemistry (color
 200 dots) seem to have higher sulfate burden anomaly and lower SATa than models without
 201 (color circles). However, the relationship between sulfate burden anomaly and SATa is
 202 consistent among models with and without interactive chemistry. That is, there is no
 203 obvious difference in relationship between sulfate burden anomaly and SATa for
 204 models with and without interactive chemistry.



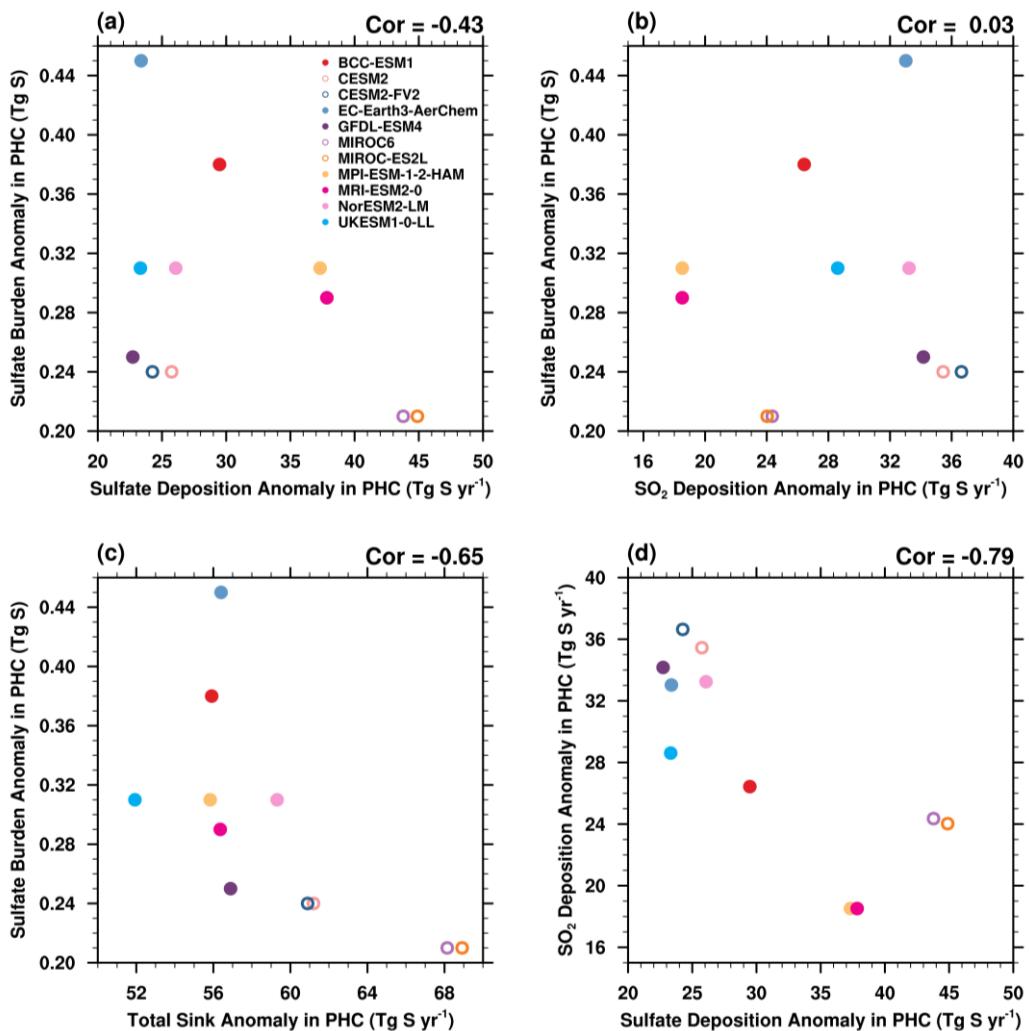
205
 206 **Figure 2.** Scatter plots of SATa in PHC period (x-axis, °C) versus (a) sulfate burden anomaly in
 207 PHC period (y-axis, Tg S) in historical experiments, and (b) the transient climate response (TCR,
 208 °C) for each model calculated by 1pctCO₂ experiments. The corresponding correlation coefficient
 209 (cor) is shown at the top-right corner of each panel. The anomalies are relative to 1850~1900 mean.
 210 Models with and without interactive chemistry are marked by color dots and color circles,
 211 respectively.

212

213 Greenhouse gases (GHGs) also show a rapidly increasing trend in the PHC period.
 214 However, TCR, which can generally indicate the impact of GHGs, is insignificantly
 215 correlated with SAT anomalies in CMIP6 models and the correlation coefficient is even
 216 negative (Fig.2b). Therefore, the biases of atmospheric sulfate burden and the
 217 associated sulfate aerosol cooling effects play an essential role in the anomalous-
 218 cooling biases in the CMIP6 models.

219

220 3.2 Sulfur deposition and a metric for the global sulfur cycle diagnostic (SAME)

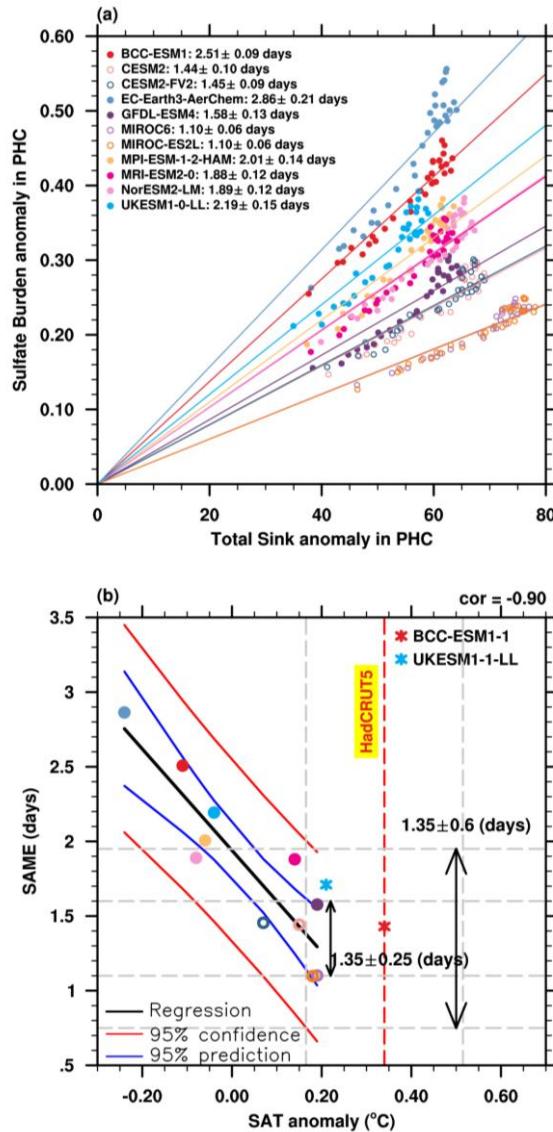


221

222 **Figure 3.** (a) Sulfate deposition anomaly, (b) SO₂ deposition anomaly, and (c) total sulfur sink
 223 (sulfate and SO₂ deposition) anomaly versus sulfate burden anomaly in PHC period (Tg S, y-axis).
 224 (d) Sulfate deposition anomaly (x-axis) versus SO₂ deposition anomaly (y-axis) in PHC period. Unit
 225 for deposition anomaly is Tg S yr⁻¹.

226

227 Fig. 3 shows comparisons of the global mean total sulfate burdens versus sinks of
228 sulfur as anomalies in the PHC period relative to a baseline period of 1850~1900. As
229 shown in Fig.3a, the sulfate burden anomaly is negatively correlated with sulfate
230 deposition anomaly. However, the correlation is not statistically significant, partly
231 attributable to a subset of five models characterized by low sulfate deposition and low
232 sulfate burden. These models prevent the robustness of a robust linear fit derived from
233 the remaining models. There is no clear statistical relationship between sulfate burden
234 anomaly and SO_2 deposition anomaly (Fig. 3b). However, the correlation between
235 sulfate burden anomaly and total sulfur sink (deposition of sulfate and SO_2) anomaly
236 increases to -0.65, significant at the 5% level using a Student's t-test (Fig.3c). Notably,
237 within the subset of five models exhibiting both low sulfate deposition and low sulfate
238 burden, most display higher SO_2 deposition in relative to the ensemble mean. The high
239 SO_2 deposition compensates for their low sulfate deposition, making the total sulfur
240 deposition magnitude sufficiently to sustain a significant correlation with sulfate
241 burden. We can hypothesize that in these 5 models' oxidation of precursors to sulfate
242 proceeds slower than in the other models. This is reflected by their larger SO_2
243 deposition rates, and leads to less sulfate in the atmosphere. That is, both the sulfate
244 deposition and the SO_2 deposition (via its relationship with oxidation rates) are
245 responsible for the sulfate burden anomalies, although the relative ratio of both
246 deposition processes is different among the models. Further examination indicates that
247 the anomalous SO_2 deposition rate among the models is highly negatively correlated
248 with the anomalous sulfate deposition rate with correlation coefficient of -0.79 (Fig.3d).
249 The total sulfur sink is examined and discussed hereafter.



250

251 **Figure 4.** (a) Scatter plots of yearly total sulfur sink anomaly (x-axis, $Tg\ S\ yr^{-1}$) versus sulfate burden
252 anomaly (y-axis, $Tg\ S$) in PHC period in relative to 1850~1900 mean. Number in legend shows the
253 mean and standard deviation of ratio between sulfate burden anomaly and total sulfur sink anomaly
254 in PHC period, defined as SAME, units: days. (b) The mean SATa ($^{\circ}C$, x-axis) versus SAME (days,
255 y-axis) in PHC period for each model. The black solid line is the linear fitting. The blue and red
256 solid lines are the 95% confidence interval (CI) and 95% prediction interval (PI), respectively. SAT
257 anomaly in HadCRUT5 and its $0.175^{\circ}C$ boundaries are shown by the red dashed line and parallel
258 gray dashed lines, respectively. The red and blue asterisks are the results in the two post-CMIP6
259 models BCC-ESM1-1 and UKESM1-1-LL, respectively.

260

261 Considering the importance of anomalous total sulfur sink to sulfate burden,
262 Fig.4a examines their relationship during the PHC period in each model. Generally, the
263 anomalous sulfate burden and total sink are positively correlated and co-vary almost

264 linearly in all the models. The ratio between anomalous sulfate burden and total sulfur
265 sink is defined as the SAME index in Section 2.2. The mean SAME in PHC ranges
266 from 1.1 days in MIROC models to 2.86 days in EC-Earth3-AerChem. The SAME is
267 generally longer in models with interactive chemistry (color dot) than without (color
268 circle).

269 The standard deviation of SAME for each model in PHC ranges from 0.03 to 0.12
270 days, about 3.0% of the mean SAME. That is, although the sulfate burden increased
271 significantly in the PHC period, the SAME hardly changed. This is an important sign
272 that SAME is a robust index for evaluating the sulfur cycle in model development. Our
273 finding that a single value of SAME is capable of characterizing the anomalous cooling
274 for each model, makes it a convenient target for model tuning. Focusing on a single,
275 representative parameter can make tuning more efficient and help to reduce the
276 computation cost, especially when the model resolutions become relatively high.
277 Moreover, because SAME has a clear physical interpretation as a globally defined
278 efficiency factor for sulfur removal processes, tuning based on SAME can give
279 confidence that SAT biases are reduced for a ‘right’ (i.e., physically sensible) reason.

280

281 **3.3 The recommended SAME value**

282 Tuning based on SAME requires an empirical best-estimate SAME value to aim
283 for. Therefore, a further question is how to estimate the reasonable values for SAME.
284 Here we try to constrain the SAME using the SATa in observations. Fig. 4b shows the
285 SAME and SATa in PHC in each model. The SATa is highly correlated with SAME
286 with a correlation coefficient of -0.90. The SAT anomaly in PHC is 0.34°C, shown by
287 the vertical red dash line (HadCRUT5). Considering the internal variability in the
288 climate system and the uncertainty in observation, the observed uncertainty is suggested
289 to be 0.175°C (the vertical gray dash line parallel with the red dash line). The observed
290 uncertainty is estimated as the standard deviation of observed annual mean globally
291 averaged SAT in HadCRUT5 from 1850 to 2014 after removing the least squares linear

292 trend. We calculate the linear fitting between SATa and SAME (black line in Fig. 4b),
293 the 95% confidence interval (CI, blue curves), and the 95% prediction interval (PI, red
294 curves), respectively. SATa in seven CMIP6 models falls beyond the observational
295 range, which in the remaining four models, SATa closely approaches the lower bound
296 of observation, giving a range of SAME between 1.1 to 1.58 days. That is, the SO₂
297 oxidation or deposition terms in CMIP6 models may need to be modified.

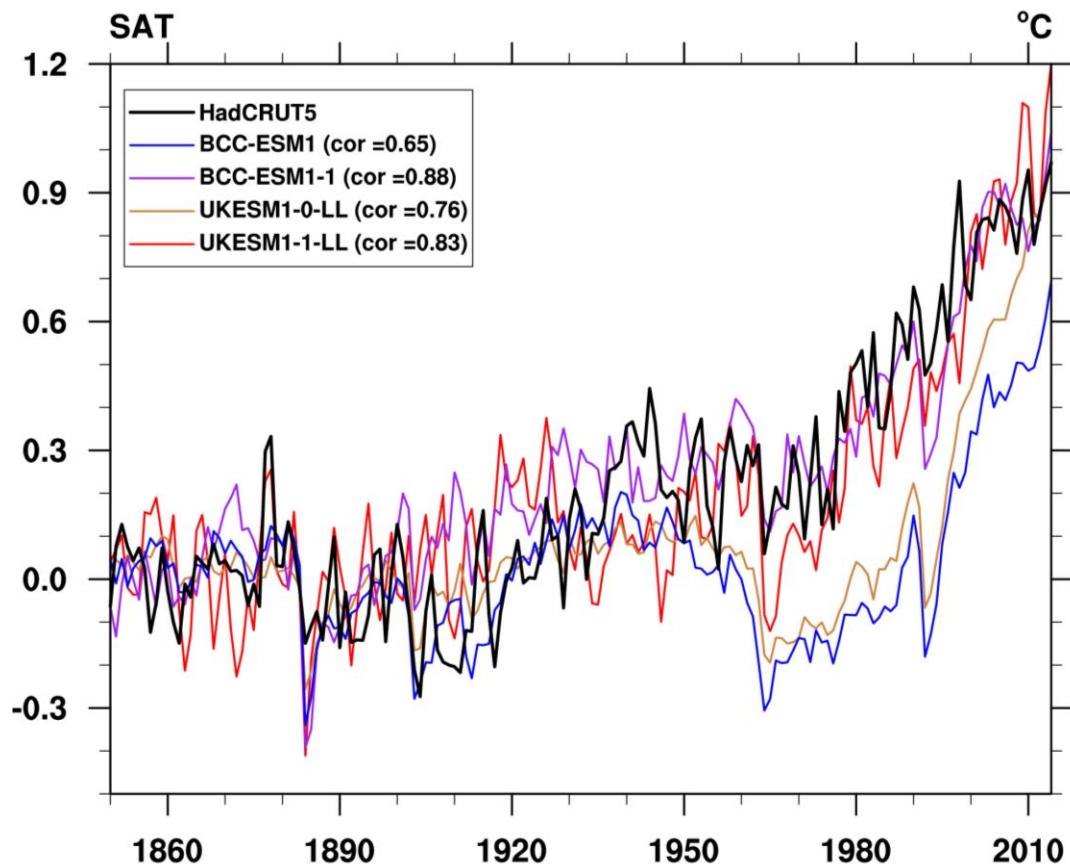
298 Here we use this metric to modify the sulfur cycle in BCC-ESM1, more
299 specifically we quadruple the SO₂ dry deposition over land and multiply the SO₂ dry
300 deposition over the ocean by 1.5. This effect is similar to that in UKESM1-0-LL by
301 modifying SO₂ dry deposition parameterization (Hardacre et al., 2021; Mulcahy et al.,
302 2023). The impact of changes to the SO₂ dry deposition parameterization in UKESM1-
303 0-LL is an increase of SO₂ dry deposition by a factor of 2 to 4. As shown by the red
304 asterisk in Fig.4b, the SAME reduced from 2.51 days to 1.43 days in updated BCC-
305 ESM1 (BCC-ESM1-1), falling right within the PI constraint. The new SAME index is
306 57% of its previous values. Accordingly, the SATa in PHC is 0.34°C, falling within the
307 observational range from 0.165°C to 0.515°C. We also examine the SAME in
308 UKESM1-1-LL with modified SO₂ dry deposition parameterization. The SAME is
309 shortened from 2.19 days to 1.71 days, falling within the CI constraint. Accordingly,
310 the SATa in PHC period increases by about 0.25°C.

311 Given that most models underestimate SATa relative to observations,
312 extrapolating SAME values for SATa exceeding the observation (0.34°C) becomes
313 highly uncertain. Result from BCC-ESM1-1 suggests that the rate of decrease in SAME
314 predicted by the regression line may not hold for SATa values above the observed lower
315 bound (0.165 °C). Therefore, we recommend a central SAME estimate of 1.35 days by
316 the linear fitting at the observed lower bound. Critically, this value carries inherent
317 uncertainties that must be quantified:

318 - The 95% confidence interval (CI) of ± 0.25 days (i.e., 1.10–1.60 days).
319 - The wider 95% prediction interval (PI) of ± 0.6 days (i.e., 0.75–1.95 days).

320 The substantial difference between the CI and PI ranges underscores the challenge
321 in precisely constraining SAME. We advise using the PI for applications requiring
322 robustness against individual model deviations.

323



324

325 **Figure 5.** Evolutions of SAT anomalies relative to 1850~1900 mean for HadCRUT5, BCC-ESM
326 models, and UKESM models. The numbers in legend are the corresponding correlation coefficients
327 with HadCRUT5.

328

329 As demonstrated by the global mean SATa in BCC-ESM and UKESM models
330 (Fig.5), both models on average tracked the instrumental record quite well with
331 statistically higher correlation coefficients with observation (HadCRUT5). That is,
332 improvements in sulfur deposition parameterizations, which reduced the SAME index,
333 improved the representation of historical surface temperature variation.

334

335 **4. Discussion: Sulfate lifetime in CMIP6 models and the two post-CMIP6 models**

336 **Table 2** Sulfate burden, sulfate depositions, and sulfate lifetime in CMIP6 models,
 337 BCC-ESM1-1 and UKESM1-1-LL in PHC period.

model name	Sulfate (Tg S)	burden	Sulfate Deposition (Tg S yr-1)		Sulfate lifetime (days)
			DrySO4	WetSO4	
BCC-ESM1	0.59	2.07	43.78		4.70
CESM2	0.40	6.14	35.13		3.54
CESM-FV2	0.43	5.92	32.60		4.07
EC-Earth3-AerChem	0.75	1.29	40.39		6.57
GFDL-ESM4	0.46	7.58	31.68		4.28
MIROC6	0.33	7.50	61.29		1.75
MIROC-ES2L	0.33	5.67	67.42		1.65
MPI-ESM-1-2-HAM	0.74	2.41	69.41		3.76
MRI-ESM2-0	0.53	0.75	56.96		3.35
NorESM2-LM	0.52	6.39	40.33		4.06
UKESM1-0-LL	0.63	7.00	34.79		5.50
BCC-ESM1-1	0.48	1.34	19.2		8.53
UKESM1-1-LL	0.52	5.57	27.34		5.77

338

339 Generally, the SAME metric is used to facilitate model tuning of the sulfate
 340 burden, ensuring that models do not overestimating the sulfate cooling effect over the
 341 historical period, as was the case in CMIP6 and is a current concern for model
 342 performance in the upcoming CMIP7 experiments. SAME is relative to sulfate
 343 lifetime but it is calculated by the anomalous changes and also considers SO₂
 344 deposition.

345 Sulfate lifetime is critical for validating the model's physical realism of sulfate
 346 cycle. Therefore, sulfate lifetime should be further examined to ensure model
 347 credibility. Here we calculated sulfate lifetime in PHC as the ratio of sulfate burden to
 348 total sulfate deposition (wet plus dry). As shown in Table 2, sulfate lifetime in CMIP6
 349 models ranges from 1.65 days in MIROC-ES2L to 6.57 days in EC-Earth3-
 350 AerChem. The mean sulfate lifetime is 3.93 days, consistent with previous literatures,
 351 particularly the mean value of 4.12 days in AeroCom models with standard deviation

352 of 18% (Textor et al., 2006). The wide sulfate lifetime range in CMIP6 models is
353 attributed to variations in both sulfate burden (0.33 to 0.75 Tg S) and deposition rates
354 (0.75 to 7.58 Tg S yr⁻¹ for dry deposition, and 31.68 to 69.41 Tg S yr⁻¹ for wet
355 deposition).

356 Sulfate lifetimes in the two post-CMIP6 models, 8.53 days in BCC-ESM1-1 and
357 5.77 days in UKESM1-1-LL, are generally longer than those of their CMIP6 versions.
358 The longer sulfate lifetimes in the two post-CMIP6 models may be due to lower SO₂ in
359 these revised models but also could be due to physical climate changes (e.g.,
360 temperatures, clouds, rainfall). Compared to prior lifetime measures reported in the
361 literature and considering the range of lifetimes found in recent models, the sulfate
362 lifetimes in BCC-ESM1-1 and UKESM1-1-LL also appear reasonable (e.g., Charlson
363 et al, 1992; Kristiansen et al. 2012; Textor et al., 2006).

364

365 **5. Conclusions**

366 Aerosol cooling effect is considered to be the second most important
367 anthropogenic forcing over the 20th Century. Our study, based on the 11 CMIP6 models
368 with aerosol schemes, demonstrates that the anomalous cooling bias in the PHC period
369 is closely related to the sulfate burden changes in the atmosphere. Sulfate burden in the
370 models, and hence the strength of the anomalous cooling, is determined by sulfur
371 deposition. We introduce a metric, called the SAME index, which incorporates the
372 effects of sulfur removal processes on sulfate concentration. The index is highly
373 correlated with cooling, and can be used to constrain sulfur removal processes in
374 models, on a global scale.

375 A constraint on SAME by observed SATA, is used to inform a choice of tunable
376 parameters for model depositions. Modifying sulfur deposition properties leads to an
377 improved SAME in BCC-ESM1-1 and UKESM1-1-LL, as well as SATA reproductions.
378 The optimal target value of SAME is 1.35 days with uncertainty range of ± 0.25 days
379 by 95% CI and ± 0.6 days by 95% PI. Sulfate lifetime is critical for validating the

380 model's physical realism and should be further examined to ensure model credibility.

381 Given that CMIP6 models overestimate the cooling effects of sulfate during the
382 PHC period, when emissions were rising, it is reasonable to assume that they will
383 underestimate the rate of warming during periods in climate projections. This has
384 potential implications for the use of CMIP6 in scenarios that incorporate clean-air
385 measures to inform the Paris Agreement goals of limiting warming to below 2 or 1.5°C,
386 i.e., SSP1-2.6 and SSP1-1.9 in CMIP6 (O'neill et al., 2016). To improve the reliability
387 of projections, sulfur cycle processes in models should be improved. The SAME metric
388 introduced in this paper provides a physically meaningful measure of the activity of the
389 sulfur cycle, at a global scale, which we have shown can be used to improve modeled
390 sulfur processes.

391

392 **Code availability**

393 All data processing codes are available if a request is sent to the corresponding authors.

394

395 **Data availability**

396 The HadCRUT5 dataset is accessible through Met Office Hadley Centre observations
397 database (<https://www.metoffice.gov.uk/hadobs/hadcrut5/>). All the model data can be
398 freely downloaded from the Earth System Grid Federation (ESGF) nodes
399 (<https://aims2.llnl.gov/search/cmip6/>).

400

401

402 **Author contributions**

403 The main ideas were formulated by J.Z. and K.F. J.Z. wrote the original draft. The
404 results were supervised by K.F. and S.T.T. All the authors discussed the results and
405 contributed to the final manuscript.

406

407 **Competing interests**

408 The authors declare no competing financial and/or non-financial interests.

409

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413

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