

1 **The response of stomatal conductance to vapor pressure deficit over global wheat**
2 **areas in CMIP6 warming projections**

3
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12 **Key Points:**

- 13 • CMIP6 projections show a sustained increase in vapor pressure deficit (VPD) associated
14 with global warming over global wheat areas
- 15 • The global response of wheat stomatal conductance (g_s) shows a decreasing trend from
16 around 2040 in the three CMIP6 scenarios evaluated
- 17 • The higher sensitivity of g_s to VPD than to air temperature suggests that the areas of
18 decreasing g_s could increase in warmer climates

19

Abstract

20 Climate change is expected to alter the conditions in which plants develop. The mechanisms by
21 which plants can adapt to changing conditions must be studied in terms of the magnitude of their
22 response and the implications for productivity. Recognizing wheat as a main crop sustaining
23 global livelihoods as well as the need for long-term adaptation strategies to climate change, this
24 work assesses the response of wheat stomatal conductance (g_s) to changes in vapor pressure
25 deficit (VPD) associated with global warming in three future scenarios from the Coupled Model
26 Inter-comparison Project phase 6 (CMIP6). We used multiple datasets and a modeling approach
27 to estimate g_s as a function of atmospheric variables only over global wheat areas. The results
28 show a sustained increase in both temperature and VPD in the historical period and future
29 CMIP6 scenarios, and a generalized decrease in wheat g_s that becomes clear from around the
30 year 2040. The pattern of change is spatially divergent, with areas that present increases in
31 relation to the historical average, which mostly concentrate towards high latitudes. Negative
32 trends in g_s area mainly observed in North America, Southern Europe, North Africa and Asia.
33 Notwithstanding, the proportion of areas of positive change in g_s tend to decrease in warmer
34 climate scenarios. g_s is more sensitive to changes in VPD than to global warming, which implies
35 that the areas of negative trends in g_s could expand to higher latitudes. These results may assist
36 the regional focus of long-term wheat adaptation programs to climate change.

37

38 Plain language summary

39 Climate change projections are of great relevance for crops, on the one hand due to the increase
40 in temperatures, but also due to the changes in vapor pressure deficit (VPD), an indicator of air
41 dryness. One of the plant responses to changes in temperature and VPD is stomatal closure,
42 which affects transpiration, photosynthesis, and consequently, productivity. This is of great
43 relevance for agricultural crops such as wheat given its importance for global food security. The
44 present study assesses the global response of wheat stomatal conductance (g_s), an indicator of the
45 magnitude of leaf-air gas exchanges, to the changes in VPD under future CMIP6 global warming
46 scenarios. The results show a generalized increase in VPD that is consistent with projected
47 warming over global wheat areas. However, a divergent response in g_s is projected. In general,
48 high latitude areas of both the Northern and Southern hemisphere show an increase in g_s , while
49 lower latitude areas in North America, Asia, and Europe show a decrease in g_s . The results also
50 revealed that the changes in g_s are more sensitive to changes in VPD than to temperature.

51

52 1 Introduction

53

54 The water losses and CO₂ photosynthetic uptake by plant leaves are controlled by
55 stomatal behavior, which is in turns controlled by biophysical processes counteracting water
56 availability and atmospheric demand (Buckley, 2005). The water flow through the soil-plant-
57 atmosphere continuum is driven by the differences in water potential between the soil and the air
58 surrounding the leaves, locally controlled by leaf transpiration (Passioura, 1982). A higher
59 atmospheric water demand reduces leaf water status, so that the stomata reduce their opening to
60 prevent the decline in water status (Buckley, 2005), playing a key role in the adaptation of plants
61 to changing environmental conditions (Damour et al., 2010). Given the strong relationship

62 between transpired water, crop productivity, and water use efficiency (Miner et al., 2017), the
63 stomatal conductance (g_s), an estimator of the magnitude and dynamics of leaf-air gas exchanges,
64 has been widely used as an indicator of plant water status and as a predictor of yields in crop
65 species such as wheat (Fisher et al., 1998). In this regard, higher yielding wheat varieties have
66 been monitored and described as having higher g_s , consequently lower canopy temperature and
67 heat resistance, thus making g_s useful as a selection metric for high yields under optimal water
68 supply conditions (Lu et al., 1998), and drought tolerance (Bota et al., 2004).

69 The global observed and projected atmospheric warming leads to an increase in the
70 drying power of air (Broz et al., 2021). This drying effect is driven by the vapor pressure deficit
71 (VPD), the difference between the actual water vapor content (actual vapor pressure) and the
72 maximum amount that the air can hold in the gas phase at a specific temperature (saturated vapor
73 pressure), whose magnitude depends on multiple factors such as land surface evaporation and the
74 large-scale atmospheric environment (Ficklin & Novick, 2017). Increasing attention has been
75 paid to VPD in recent years as its observed increase in the last decades has been associated with
76 a global reduction in vegetation growth (Yuan et al., 2019), offsetting the CO₂ fertilization effect,
77 and with an enhancement of heat stress on crops given its association with land-surface water
78 loss (Lobell et al., 2013). Moreover, anomalously high seasonal values of VPD have been
79 directly associated with the occurrence of droughts (Sanginés de Cárcer et al., 2018). The
80 immediate response of plant to increases in VPD is the stomatal closure, consequently reducing
81 g_s , with impacts on multiple biophysical processes such as evapotranspiration (Grossiord et al.,
82 2020). Plants close their stomata after a certain level of soil water depletion to avoid water loss
83 (Fang et al., 2021) and modern crop varieties are no exception. However, it has been observed
84 that VPD can limit stomatal conductance in a greater extent than soil water content, explaining
85 up to 90% of the variability in g_s in relation to soil moisture in maize and soybean (Kimm et al.,
86 2020), and up to 70% in global croplands and natural ecosystems (Novick et al., 2016).
87 Similarly, VPD has been quantified as the main factor triggering reductions in terrestrial gross
88 primary production (Fu et al., 2022; Lu et al., 2022), and a main driver of plant stress compared
89 with soil water content and available energy (Tong et al., 2019). These studies show that the
90 expected increase in VPD due to climate change could exacerbate stress conditions for crops and
91 terrestrial ecosystems, in addition to the relative importance of atmospheric water demand that
92 could be accentuated under climate change scenarios (Novick et al., 2016).

93 The current context of climate change poses challenges to global agriculture, especially
94 to field crops such as wheat, given its importance for food security and livelihoods, and the wide
95 range of environments in which wheat is grown (Gbegbelegbe et al., 2017). However, despite the
96 observed increase and robust projections in air temperature and evaporative demand (Byrne &
97 O’Gorman, 2018; Zhu & Troy, 2018; Fang et al., 2022), there is still much uncertainty regarding
98 the projections of rainfall (Sheffield et al., 2012; Shiogama et al., 2022) and therefore on
99 integrating processes such as droughts (Cook et al., 2014) and associated plant responses (Kath
100 et al., 2022). Although VPD is projected to increase in the future as a consequence of the
101 combination of global warming and saturated vapor pressure, changes in actual vapor pressure
102 and relative humidity vary across regions (Byrne & O’Gorman, 2013). The latter results from
103 complex feedback mechanisms that can modify the surface energy partitioning and evaporative
104 fraction (Barkhordarian et al., 2019). Notwithstanding, increases in VPD can exacerbate the
105 water demand and stomatal closure (Damour et al., 2010; Medlyn et al., 2011). Consequently,
106 assessing projected climate conditions acting as drivers of agriculturally relevant hydroclimatic
107 events is necessary to plan long-term adaptation strategies for more water-efficient crop varieties

108 and cropping systems (Damour et al., 2010; Mondal et al., 2020).

109 Considering the relationship between VPD, stomatal closure and g_s , its consequences on
110 water status of plants, and the role of increasing temperatures, the main objective of this work is
111 to assess the implications that projected changes in VPD associated with global warming would
112 have in terms of climatological g_s trends over global wheat cultivation areas. Although much
113 attention has been paid to the impacts of changes of air temperature and rainfall on wheat and
114 other crops, however, little has been done with respect to the influence of changes in evaporative
115 demand on processes relevant to crop water consumption and productivity. The latter is an
116 important issue for a major staple food for many countries since drought and heat stress are
117 among the most severe factors that reduce wheat productivity. We used outputs from a suite of
118 twenty-first century General Circulation Models (GCMs) simulations issued from the Coupled
119 Model Intercomparison Project phase 6 (CMIP6; Eyring et al., 2016). We calculated g_s with the
120 widely-used Jarvis-type (Jarvis, 1976) multiplicative model, to address the questions on (1) what
121 are the expected changes in VPD associated with global warming over global wheat-producing
122 areas, (2) what might be the expected responses in terms of g_s , and (3) what is the relative
123 importance of changes in temperature and in VPD on the projected g_s . The results from this work
124 represent an assessment of background climate and biophysical conditions affecting wheat
125 physiology that accompany other effects associated with atmospheric composition such as CO₂
126 fertilization, and can contribute to the identify climate pathways to develop adaptation strategies
127 and provide guidance for wheat breeding.

128

129 **2 Materials and Methods**

130

131 *2.1 CMIP6 and historical climate datasets*

132

133 As a primary dataset, we used daily climate data of mean air temperature, downwelling
134 shortwave radiation and relative humidity, from five bias-corrected and downscaled CMIP6
135 (Eyring et al., 2016) GCMs (Table 1) and three Shared Socioeconomic Pathways (SSP) for the
136 21st century (SSP126, SSP370 and SP585) at $0.5^\circ \times 0.5^\circ$ spatial resolution. These CMIP6 model
137 outputs belong to the Inter-Sectoral Impact Model Intercomparison Project (ISIMIP,
138 <https://www.isimip.org/>; Lange and Büchner, 2022). The historical period from 1971 through
139 2014 was considered, and two future periods: midcentury, from 2041 through 2070, and end of
140 century, from 2071 through 2100.

141

142

143 Table 1. List of CMIP6 GCMs from ISIMIP used in this study.

Model	Institution (country)
GFDL-ESM4	Geophysical Fluid Dynamics Laboratory (United States)
IPSL-CM6A-LR	Institut Pierre-Simon Laplace (France)
MPI-ESM1-2-HR	Max Planck Institute for Meteorology (Germany)
MRI-ESM2-0	Meteorological Research Institute (Japan)
UKESM1-0-LL	Met Office Hadley Centre (United Kingdom)

144

145 In addition to CMIP6 projections, we used data from European Centre for Medium-
 146 Range Weather Forecasts (ECMWF) AgERA5 product (Copernicus Climate Change Service,
 147 C3S, 2019) as an observational reference. AgERA5 corresponds to a statistically downscaled
 148 ($0.1^\circ \times 0.1^\circ$) daily time-step version of ERA5 reanalysis. In this case, minimum and maximum
 149 air temperature, shortwave radiation and dewpoint temperature data were used, for the period
 150 1981 through 2014. AgERA5 data were resampled to the $0.5^\circ \times 0.5^\circ$ CMIP6 grid resolution using
 151 bilinear interpolation.

152

153 *2.2 Computing vapor pressure deficit*

154

155 VPD was calculated using the widely used equation of Allen et al. (1998) for water vapor
 156 pressure estimation. However, the procedure differs for CMIP6 and AgERA5 data. For CMIP6,
 157 saturated vapor pressure (e_s) is computed from air temperature (T) as:

158

$$159 \quad e_s = 0.611 \times \exp\left(\frac{17.27 \times T}{237.3 + T}\right), \quad (1)$$

160

161 and actual vapor pressure (e_a) is calculated from relative humidity (RH):

162

$$163 \quad e_a = RH \times e_s, \quad (2)$$

164

165 with e_s and e_a expressed in kPa, T in $^\circ\text{C}$, and RH in percentage. VPD is simply calculated as the
 166 difference between e_s and e_a :

167

$$168 \quad VPD = e_s - e_a. \quad (3)$$

169

170 For AgERA5, since dewpoint temperature (T_d ; $^\circ\text{C}$) is available, e_a is calculated as:

171

$$e_a = 0.611 \times \exp\left(\frac{17.27 \times T_d}{237.3 + T_d}\right), \quad (4)$$

173

174 and VPD is obtained by equation (3).

175

176 *2.3 Estimating wheat stomatal conductance*

177

178 We performed a simulation analysis over global wheat areas in order to assess the
 179 response of wheat stomatal conductance to global warming and VPD. We calculated stomatal
 180 conductance following the widely-used semi-empirical Jarvis analytical formulation, which
 181 allows estimating g_s ($\text{mol m}^{-2} \text{s}^{-1}$) as the multiplicative effect of environmental factors controlling
 182 the relative stomatal closure:

183

$$g_s = g_{smax} f_1(\text{PAR}) f_2(T) f_3(\text{VPD}), \quad (5)$$

185

186 where g_{smax} is the maximum leaf stomatal conductance when environmental factors are not
 187 limiting, in this study set as $0.833 \text{ mol m}^{-2} \text{ s}^{-1}$ (Houshmandfar et al., 2015), f_1 , f_2 , and f_3 are stress
 188 functions representing the influence of independent environmental factor on g_s , which vary
 189 between 0 and 1. Following previous studies parameterizing g_s for wheat, the following stress
 190 functions were considered:

191

$$f_1(\text{PAR}) = 1 - \exp(-0.0075 \times \text{PAR}), \quad (6)$$

193

$$f_2(T) = 1 - (T - T_{opt})^2 / (T_{opt} + T_{min})^2, \quad (7)$$

195

$$f_3(\text{VPD}) = \frac{1}{1 + (\text{VPD}/3.75)}, \quad (8)$$

197

198 where PAR in Eq. (6) is the incoming photosynthetically active radiation, T in Eq. (7) is the daily
 199 mean air temperature, T_{opt} and T_{min} in Eq. (7) are the optimal and minimum temperatures for
 200 growth, respectively taken as 23 °C and 8 °C. The empirical parameters in the above equations
 201 (6) to (8) are taken from previous studies based on field experiments carried out to parameterize
 202 the Jarvis model on wheat (Danielsson et al., 2013; Houshmandfar et al., 2015). Recognizing that
 203 volumetric soil water content is a relevant variable that is usually incorporated as a stress
 204 function in Eq. (5), in this work we assumed well-watered conditions in order to focus only on
 205 the impact of atmospheric forcing on g_s . Furthermore, although multiple models exist to calculate
 206 stomatal conductance with varying complexity and forcing data requirements (Ball et al., 1987;
 207 Leuning, 1990; Tuzet et al., 2003), the Jarvis multiplicative formula corresponds to a
 208 meteorological approach that has been implemented in multiple land surface schemes coupled to

209 atmospheric models (Chen & Dudhia, 2001; Xiu & Pleim, 2001; Godfrey & Stensrud, 2010). In
210 the present work, we focused on quantifying the sensitivity of wheat g_s to VPD and warming, so
211 the Jarvis model, which represent the response of g_s to environmental factors, appears to be
212 appropriate.

213

214 *2.4 Representing wheat area and harvest dates*

215

216 We obtained global wheat area data from the Spatial Production Allocation Model
217 (SPAM) global crop production product (Yu et al., 2020). This gridded product provides
218 information on harvested area and production for the year 2005 at a spatial resolution of 5-min
219 ($0.083^\circ \times 0.083^\circ$) for multiple crops and is generated by combining information sources about
220 land use and crop production statistics. In this work, SPAM wheat growing area data were
221 initially resampled to the $0.5^\circ \times 0.5^\circ$ CMIP6 grid resolution using bilinear interpolation. In a
222 second step, the resampled data were converted into a binary mask indicating global wheat crop
223 area. SPAM grid cells with a wheat area of less than 5 % were removed. Additionally, global
224 wheat harvest dates for the year 2000 were extracted from the global crop calendar dataset for
225 specific crops of Sacks et al. (2010).

226

227 *2.5 Analyzes*

228

229 All the analyzes were performed for the period corresponding to the last 30 days of the
230 wheat growing cycle (as a proxy of the anthesis to maturity period), tanking the harvest dates
231 from Sacks et al. (2010). The mean changes for the 30-year average for the midcentury and end
232 of century periods in relation to the historical average 1971-2014 (CMIP6) and 1981-2014
233 (AgERA5) were analyzed. Also, the multi-model mean was considered. The first part of the
234 results shows an analysis of the projected changes in VPD in relation to global warming over
235 wheat growing areas. The second part of the results corresponds to the assessment of the
236 implications of changes in VPD and temperature on wheat g_s , which was calculated using the
237 complete historical and future series of VPD, temperature and solar radiation. To quantify the
238 contribution of changes in VPD and temperature on changes in g_s , the influence of VPD and
239 temperature were isolated by calculating g_s for the twentieth century using the detrended VPD
240 (denoted by g_s-tas) and temperature (denoted by g_s-vpd) of the future periods, and setting the
241 mean of the corresponding time series to be equal to the historical period, as a way to quantify
242 the impact of projected trends of both VPD and temperatures on g_s separately. We excluded solar
243 radiation from this sensitivity analysis to focus on the influence of global warming and VPD
244 trends. This approach has been used in previous studies on the relative contribution of VPD,
245 temperature and radiation to changes in drought indices (Cook et al., 2014; Noguera et al., 2022),

246

247 **3 Results**

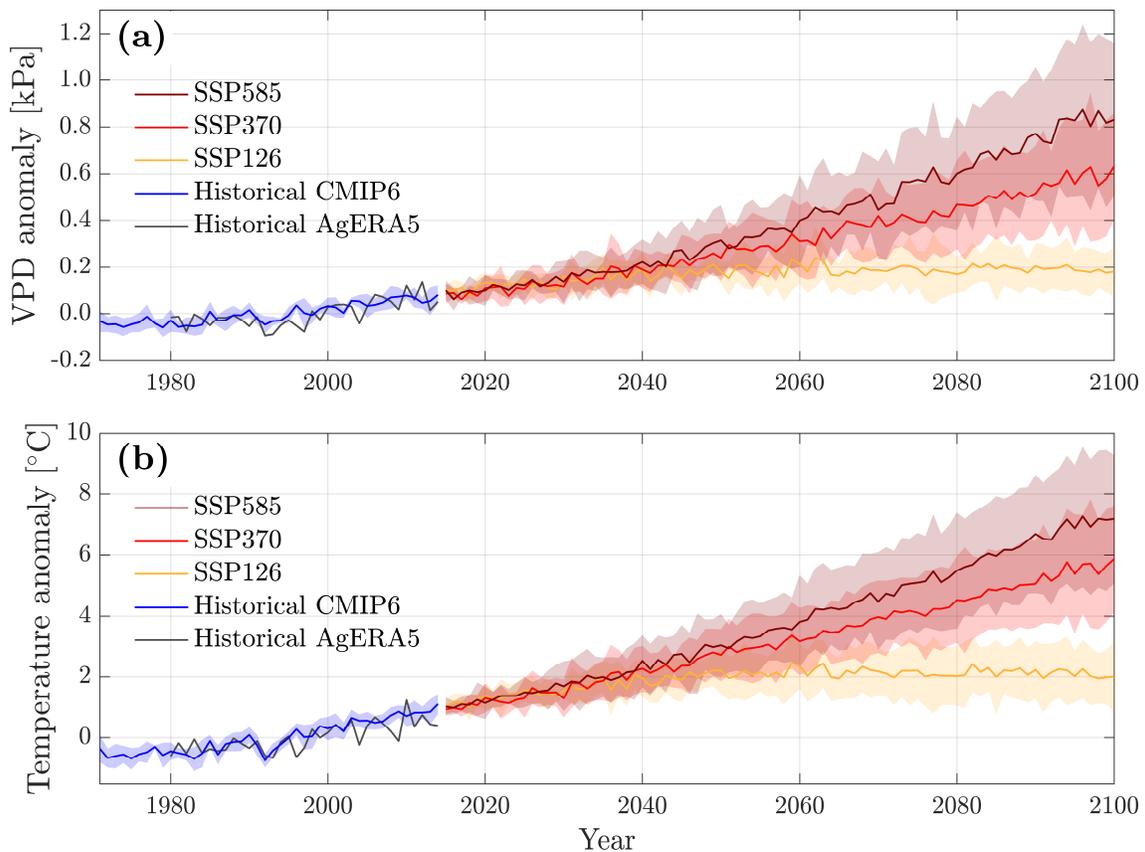
248

249 *3.1 Historical and projected VPD and temperature changes over global wheat areas*

250

251 Figure 1 shows the historical and projected changes in VPD and air temperature for
 252 CMIP6 and AgERA5. A relatively good agreement between the mean CMIP6 changes and
 253 AgERA5 is observed, with a Pearson correlation coefficient of 0.64 and 0.72, and a Root Mean
 254 Square Error of 0.04 kPa and 0.39 °C respectively for VPD and temperature. VPD anomalies
 255 (obtained by removing the average 1971-2014 and 1981-2014 value for CMIP6 and AgERA5,
 256 respectively) over global wheat areas show increasing values that progressively shift to positive
 257 from the 1990s in both AgERA5 and CMIP6 (Figure 1a), trends that have been previously
 258 highlighted globally (e.g., Yuan et al., 2019; Fang et al., 2022). The latter roughly coincides with
 259 the shift from negative to positive anomalies presented by temperature in observations and
 260 CMIP6 (Figure 1b). Future projections show a very similar mean evolution until around the year
 261 2040, from which the three future scenarios begin to diverge at an anomaly value of ~0.2 kPa
 262 and ~2 °C, respectively for VPD and temperature (Figure 1a and 1b). Regarding future scenarios,
 263 both SSP370 and SSP585 show sustained positive trends until the end of the 21st century,
 264 although they start to diverge from around 2040. On the other hand, the SSP126 scenario rather
 265 shows an increase until the 2040s to then stabilize at a value close to 0.2 kPa and 2 °C in VPD
 266 and temperature, respectively. Trends in actual values are presented in Figure S1 in Supporting
 267 Information S1.

268



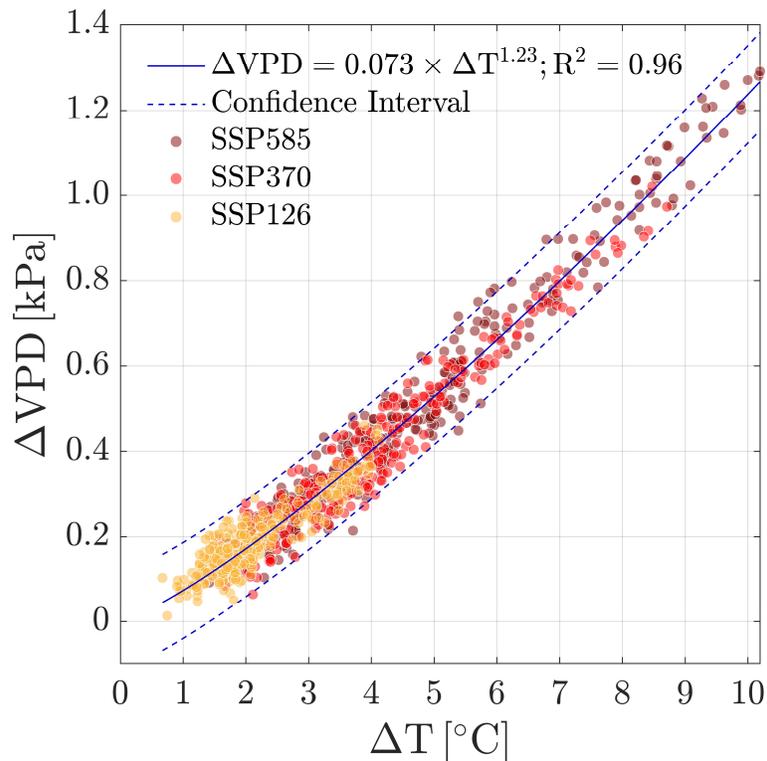
269

270 *Figure 1. Time series of historical and projected (a) VPD and (b) air temperature anomalies*
 271 *over global wheat areas. Solid lines are the multi-model mean, and the shaded area represent*
 272 *the spatial standard deviation. Calculations are made for the 30 days wheat heading periods.*

273

274 The scatter plot of the relationship between the change in VPD and global warming over
 275 global wheat areas is presented in Figure 2, which shows the global mean values for each CMIP6
 276 scenarios and GCMs. According to these results, global warming is expected to increase VPD in
 277 a slightly exponential relationship that is well captured by an exponential factor of 1.23 (Figure
 278 2). The low-end SSP126 scenario, which was generated to analyze the response of the climate
 279 system to declining greenhouse gasses emissions, shows an average increase in VPD of up to 0.4
 280 kPa in warmer climates. The magnitude of these changes is comparable to what has been
 281 observed historically in both CMIP6 and observations (Figure 1), suggesting that agricultural and
 282 natural systems could respond in a similar way to what has been observed in recent decades
 283 (Yuan et al., 2019). The SSP370 and SSP585 scenarios are associated with higher levels of
 284 warming, with values typically above 2 °C in warmer climates (Figure 2). Considering reference
 285 global warming values of 2 °C and 4 °C, Figure 2 shows an increase in VPD between 0.2 and 0.4
 286 kPa on average, which is relatively high in relation to the historical average of 1.1 kPa over
 287 wheat areas (Figure S2 in Supporting Information S2).

288



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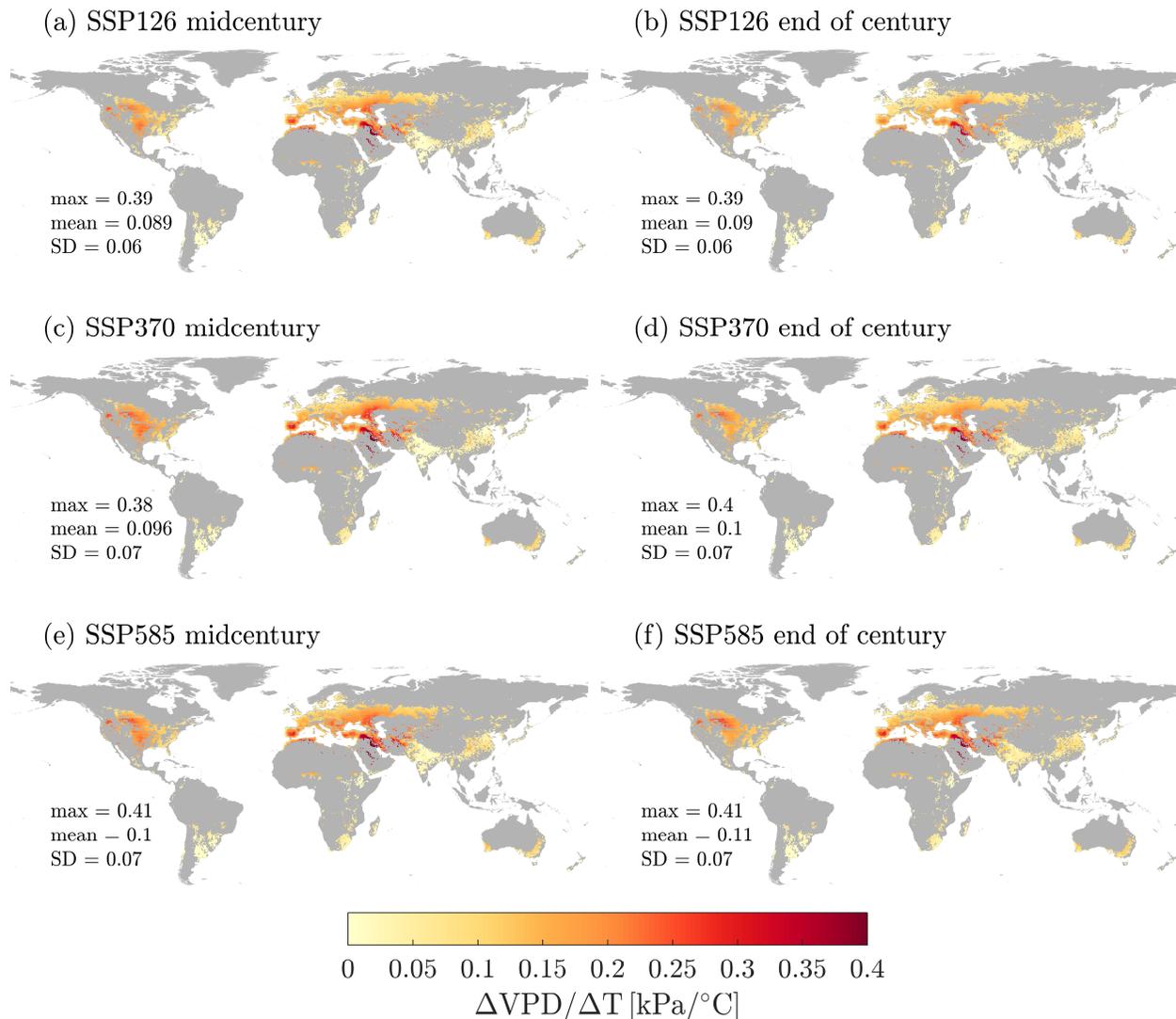
290 *Figure 2. Scatter plot between global annual change (Δ) in VPD and annual mean warming (ΔT)*
 291 *over global wheat areas for the three CMIP6 scenarios and periods (2041-2100). Each dot*
 292 *represents the mean value from the five GCMs used ($n = 900$). The continuous line corresponds*
 293 *to the exponential fit, and dashed lines bound the 95% confidence interval. The equation of the*
 294 *exponential fit is also displayed.*

295

296 The climatology of projected changes in VPD in relation to global mean warming over

297 wheat areas for multi-model mean, CMIP6 scenarios and future periods are presented in Figure
 298 3. The spatial patterns of VPD change look similar for the three CMIP6 scenarios and future time
 299 windows (midcentury and end of century). Maximum VPD increases range between 0.32 kPa/°C
 300 and 0.36 kPa/°C, with higher values mostly concentrated towards lower latitudes in Southern
 301 Europe, Asia and North America, areas in which pronounced trends in VPD are projected
 302 (Figure S3 in Supporting Information S3). Lower values are observed over areas such as South
 303 America, South Asia and Southern Africa. Clearly, the mean values close to 0.1 kPa/°C found
 304 across scenarios is associated with the similar rate of increase of both VPD and temperature
 305 (Figure 1). Thus, based on these results, it can be inferred that the sensitivity of the change in
 306 VPD is about 0.1 kPa per degree Celsius of change in temperature.

307



308

309 *Figure 3. Maps of change in VPD per degree of warming ($\Delta\text{VPD}/\Delta\text{T}$) over global wheat areas*
 310 *for multi-model midcentury (2041-2070) and end of century (2071-2100) means and for the three*
 311 *CMIP6 scenarios. The maximum, mean and standard deviation (SD) are also displayed.*

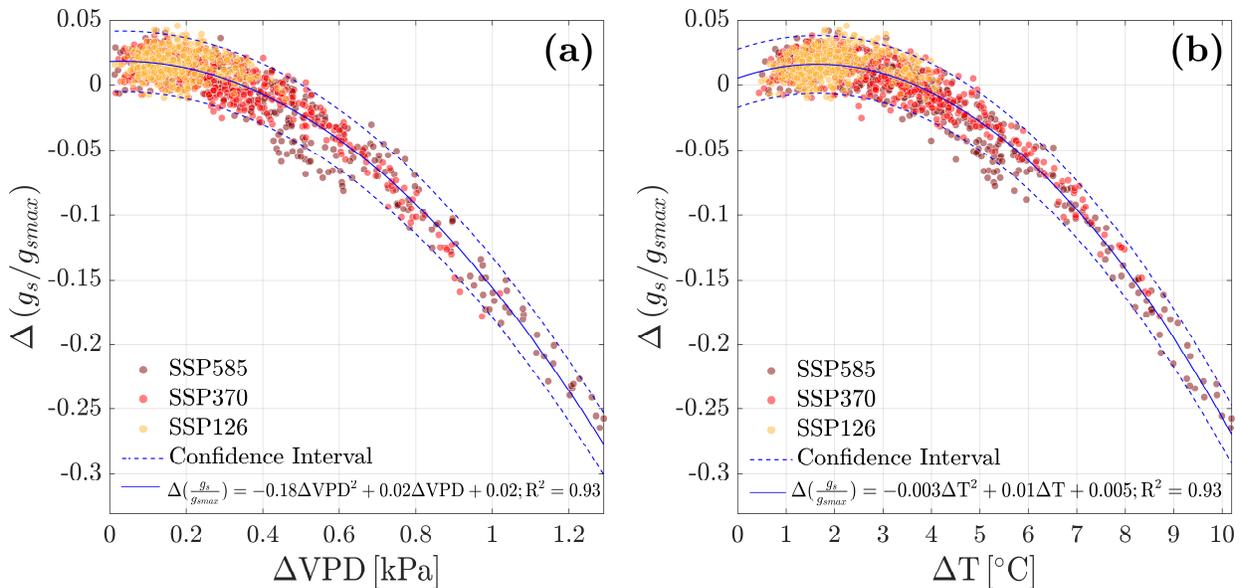
312

313 3.2 Wheat stomatal conductance and global warming

314

315 The response of g_s to changes in VPD and temperature are presented in this section. The
 316 main patterns in projections are summarized in Figure 4, which shows the scatter plot of the
 317 relationship between future g_s , expressed as relative to the maximum value (g_s/g_{smax}), and the
 318 change in VPD and temperature. Similar to Figure 2, the global mean values for each CMIP6
 319 scenario and GCM are displayed. Figure 4a shows a sharp decrease at increasing rates of g_s as a
 320 function of VPD in future climates, suggesting a continuous decrease in g_s as global warming
 321 drives higher rates of changes in VPD. Similarly, the relationship between g_s and temperature
 322 shows a strong drop in g_s as global scenarios become warmer, similar to VPD (Figure 3a).
 323 However, in the case of temperature, there seems to be a global warming level that generates a
 324 positive response in g_s , concentrated in the SSP126 scenarios, which represents a stabilization in
 325 the global temperature at a level close to 2 °C from around the year 2040 in relation to the
 326 historical average 1971-2014 (Figure 1). On the contrary, the scenarios of a sustained increase in
 327 temperature at a rate similar to recent decades for the midcentury period are characterized by a
 328 sustained decrease g_s , up to values close to 25% (i.e., $g_s/g_{smax} = -0.25$).

329



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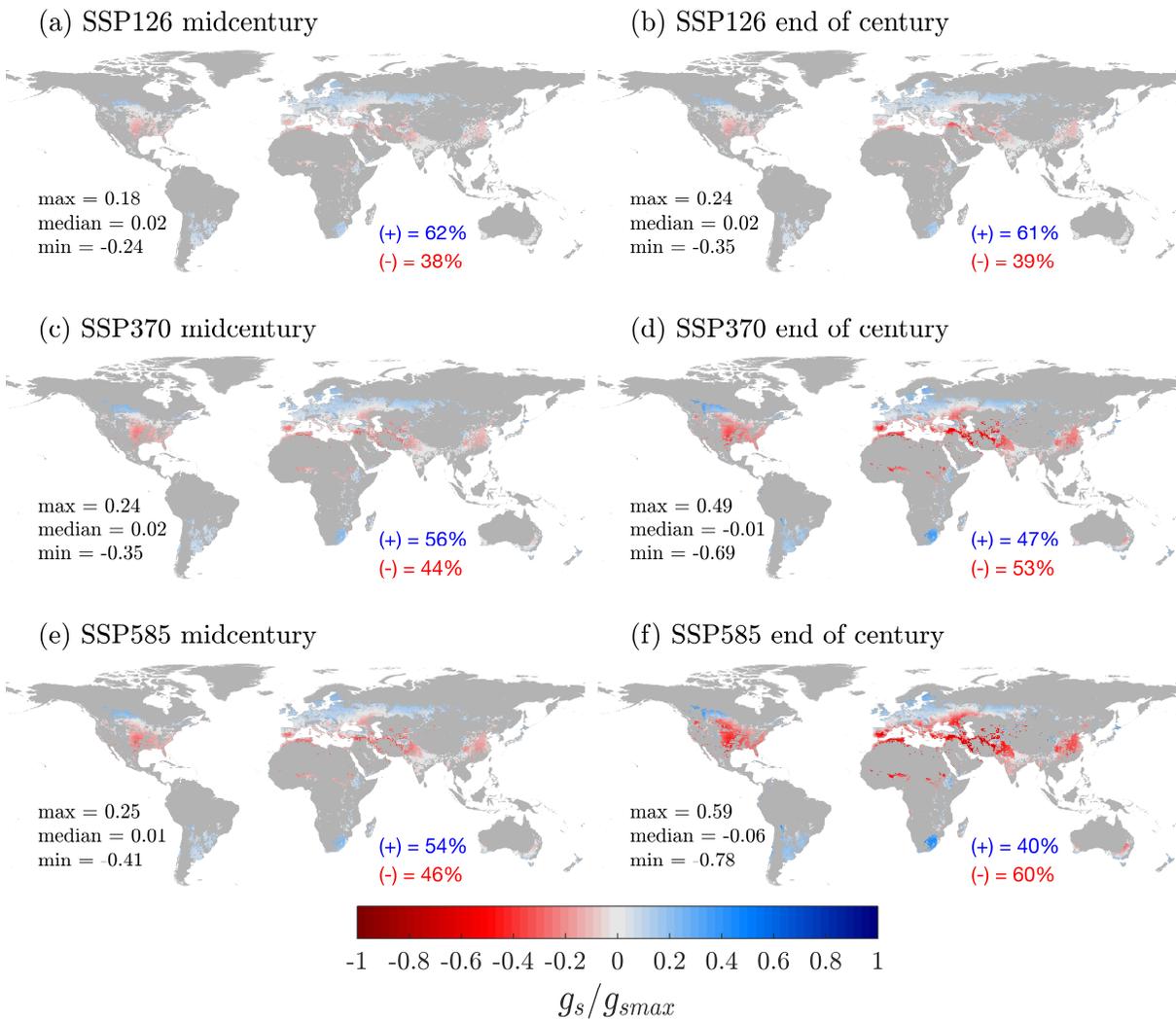
331 *Figure 4. Scatter plot between global annual change (Δ) in g_s relative to the maximum value*
 332 *(g_s/g_{smax}) and (a) mean VPD and (b) temperature (T) over global wheat areas for the three*
 333 *CMIP6 scenarios and periods (2041-2100). Each dot represents the mean value from the five*
 334 *GCMs used ($n = 900$). The continuous line corresponds to the polynomial fit, and dashed lines*
 335 *bound the 95% confidence interval. The equation of the second-degree polynomial fit is also*
 336 *displayed.*

337

338 The future climatology of g_s for midcentury and end of century periods and scenarios
 339 show a spatial distribution of positive and negative values of g_s/g_{smax} (Figure 5). For the three
 340 scenarios, median values remain close to zero from midcentury to the end of the century, but
 341 higher positive and negative changes are observed. In general, increases in g_s are observed at

342 high latitudes of the Northern and Southern Hemisphere, and a reduction in g_s over lower
 343 latitudes areas of North America, Europe, and Asia, which is coincident with the areas of greater
 344 increase in both VPD (Figure 3) and temperature (Figure S4 in Supporting Information S4).
 345 Australia appears as the only country in the Southern Hemisphere that shows dominant
 346 decreasing g_s . In addition, the maps of Figure 5 show an increasing/decreasing relative area of
 347 negative/positive changes in g_s as climate becomes warmer. In this way, while the SSP126
 348 scenario presents a very slight temporary change from midcentury or end of century (Figure 5a
 349 and 5b), both SSP370 and SSP585 scenarios represent a dominant area of negative changes in g_s
 350 that increases in time (Figure 5c-5f). The latter is evidenced, for example, in the expansion of
 351 negative changes in g_s/g_{smax} values over Southern Europe and in South Asia.

352



353

354 *Figure 5. Multi-model mean g_s relative to the maximum value (g_s/g_{smax}) over global wheat areas*
 355 *for midcentury (2041-2070) and end of century (2071-2100) and for the three CMIP6 scenarios.*
 356 *The maximum, median and minimum (SD) are displayed, as well as the percentage area with*
 357 *positive and negative changes.*

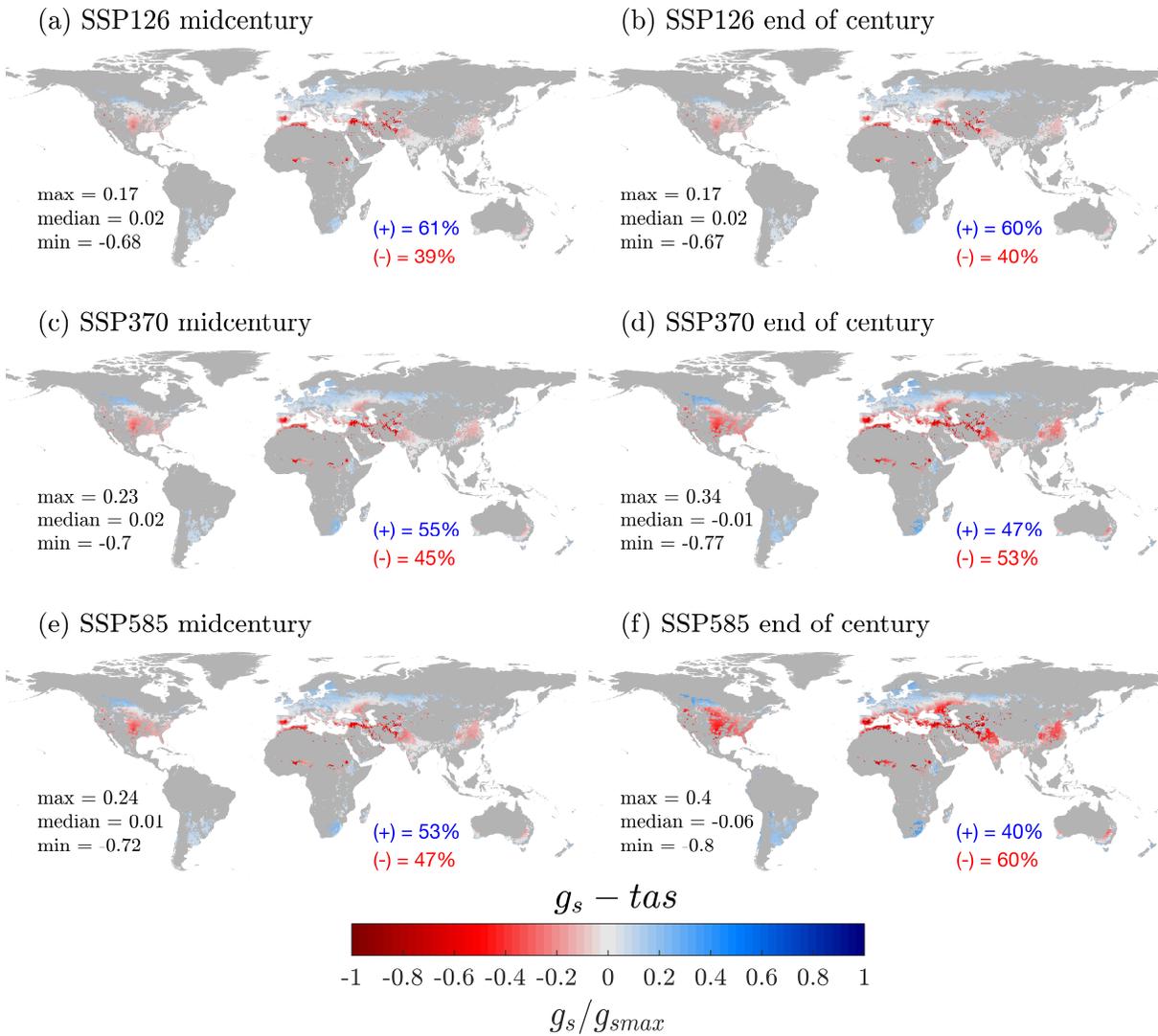
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359 *3.3 Sensitivity of wheat stomatal conductance to vapor pressure deficit and temperature*

360

361 The maps of projected changes in g_s after detrending VPD and setting the mean value to
362 the historical 30-day wheat grain filling period (g_s-tas) are displayed in Figure 6, where a
363 spatially contrasting response in terms of positive and negative changes in g_s/g_{smax} is observed.
364 This diverging pattern is similar to that obtained by including both VPD and temperature (Figure
365 5), with an increase in g_s at high latitudes in the Northern and Southern hemispheres, and a
366 negative response at low latitudes that concentrate in the Northern Hemisphere. In terms of
367 values, differences between scenarios and future period (midcentury and end of century) are low,
368 which range between -0.8 and 0.4 in the most extreme means. Nevertheless, a decrease/increase
369 in the proportion of positive/negative changes appears again as a clear pattern that differentiates
370 future scenarios. Additionally, the maps of Figure 6 and Figure 5 show a similar spatial
371 distribution of g_s/g_{smax} , although the higher magnitudes in negative changes. On the other hand,
372 sensitivity of g_s to changes VPD isolating the effect of temperature (g_s-vpd), presented in Figure
373 7, show a remarkable expansion of the areas with negative changes in g_s , with the exception of a
374 small area in high latitudes of the Northern and Southern hemispheres. A particular feature of g_s-
375 vpd is the spatially uniform and similarity of the median and extreme values for the three future
376 scenarios and two climatological periods, along with the proportion of positive and negative
377 changes.

378

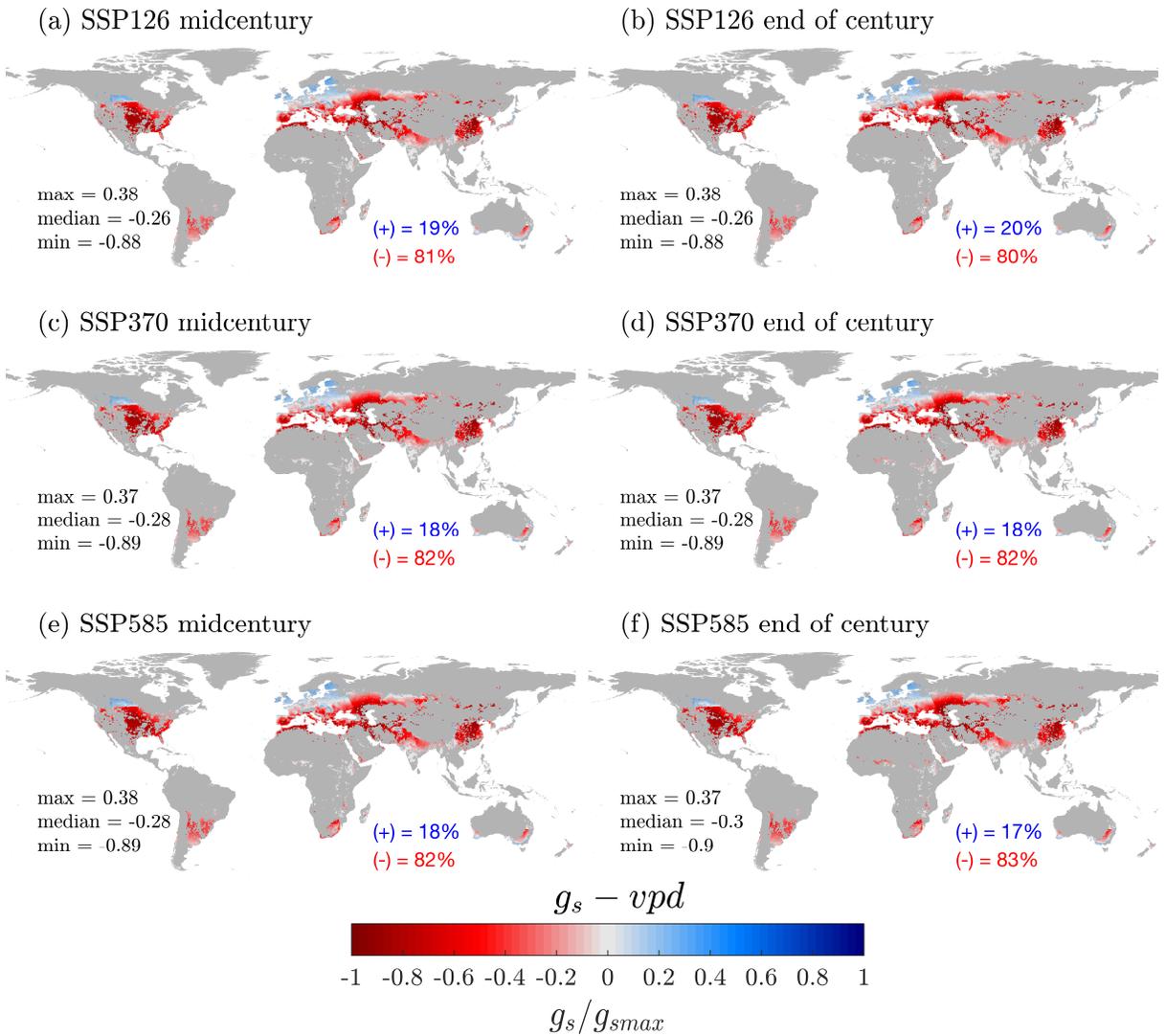


379

380 *Figure 6. Multi-model mean $g_s - tas$ relative to the maximum value (g_s/g_{smax}) over global wheat*
 381 *areas for midcentury (2041-2070) and end of century (2071-2100) and for the three CMIP6*
 382 *scenarios. The maximum, median and minimum (SD) are displayed, as well as the percentage*
 383 *area with positive and negative anomalies.*

384

385



386

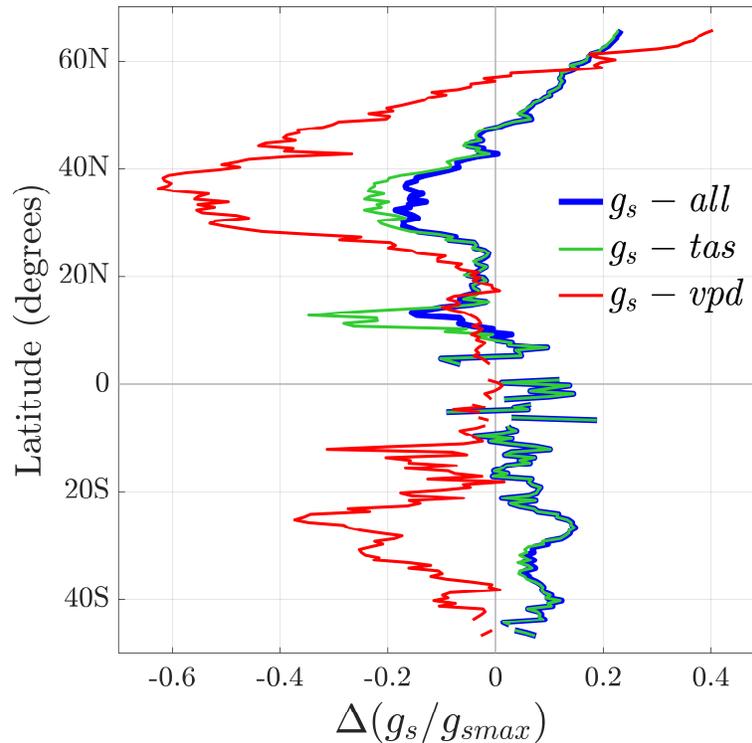
387 *Figure 7. Multi-model mean g_s -vpd relative to the maximum value (g_s/g_{smax}) over global wheat*
 388 *areas for midcentury (2041-2070) and end of century (2071-2100) and for the three CMIP6*
 389 *scenarios. The maximum, median and minimum (SD) are displayed, as well as the percentage*
 390 *area with positive and negative anomalies.*

391

392 The above-presented results are summarized in Figure 8, where the mean g_s/g_{smax}
 393 obtained from the three CMIP6 scenarios are presented as zonal averages of the difference
 394 between the future (2041-2100) and historical (1971-2014) periods. First, the difference between
 395 the g_s response between the Northern and Southern hemispheres is clear, associated with the
 396 difference in wheat area. In general, in the case of g_s calculated using both VPD and temperature
 397 (g_s -all in Figure 8), g_s decreases from northern high latitudes towards more tropical latitudes,
 398 shifting from positive to negative changes from around 50 °N to then become positive generally
 399 from around 10 °N and in the Southern hemisphere, evidencing a contrasting response. The curve
 400 of g_s removing the effect of VPD (g_s -tas) looks similar to that of g_s -all, except for a band of large
 401 drop in g_s near 30 °N-40 °N, which more or less corresponds to North America, Southern Europe,
 402 North Africa, Central Asia, as well as a region close to 10 °N-15 °N, represented mainly by the

403 North of the Sub-Saharan Africa (Figure 6). In contrast, the curve of differences for the case of
 404 sensitivity to VPD ($g_s\text{-vpd}$) is more pronounced in terms of the decrease with respect to $g_s\text{-all}$. In
 405 this case, the change from positive to negative anomalies in the Northern Hemisphere is more
 406 abrupt, remaining negative throughout all latitudes, except for a small latitudinal band between
 407 about 10 °N and 20 °N. In the Southern Hemisphere, while both $g_s\text{-all}$ and $g_s\text{-tas}$ represent an
 408 increase in g_s , $g_s\text{-vpd}$ remains negative.

409



410

411 *Figure 8. Zonally averaged multi-model and scenario mean g_s/g_{smax} (2041–2100). $g_s\text{-all}$ (Figure*
 412 *5), $g_s\text{-tas}$ (Figure 6) and $g_s\text{-vpd}$ (Figure 7).*

413

414 **4 Discussion**

415

416 *4.1 Global warming, VPD, and wheat stomatal conductance*

417

418 By focusing on the climatological aspects related with global warming and the associated
 419 increasing in atmospheric water demand, which has become evident in recent decades, the main
 420 goal of this work is the assessment of the possible future trajectories in VPD and global warming
 421 over global wheat areas, and their implications for g_s . The results represent an overview of the
 422 potential background conditions associated with climate change to which wheat cultivation could
 423 be subjected in the coming decades. Although trends in the drying power of the air, represented
 424 by VPD, has been characterized (Fang et al., 2022) and found as controlling water cycle and
 425 limiting plant growth globally (Zhang et al., 2015; Yuan et al., 2019), no specific studies has

426 focused on staple crops such as wheat. The data- and modeling-based results suggest a steady
427 increase in historical and projected VPD over global wheat areas that is very consistent with
428 temperature projections, except for SSP126 which represents a stabilization in both VPD and
429 temperature from around 2040. The reasons for this increase can be attributed to the increase in
430 temperature and consequently in the saturated vapor pressure, and a decrease in actual vapor
431 pressure, for example, over areas with negative trends in rainfall, such as North America (Seager
432 et al., 2015) or South America (Barkhordarian et al., 2019), or globally as a result of the global
433 depletion in soil moisture (Deng et al., 2020). However, this increase in VPD is more
434 pronounced over areas such as Southern Europe, Northern Africa and Central Asia, which could
435 be considered as hotspots of increasing evaporative demand (Ficklin & Novick, 2017) and
436 associated consequences on stress conditions (Grossiord et al., 2020) enhanced by higher
437 temperatures (Hatfield & Prueger, 2015). Furthermore, taking VPD 3 kPa as a global reference
438 value above which transpiration rate is affected (Tamang et al., 2022), CMIP6 projections show
439 a significant increase in the number of days exceeding 3 kPa during the last 30 days of the wheat
440 growing cycle (Figure S5 in Supporting Information S5).

441 The global response of g_s to CMIP6 projections in VPD and warming shows an
442 exponential decay in g_s (Figure 4). It is worthwhile commenting that this corresponds to a
443 climatological relationship that differs from the inverse and nonlinear relationship often found in
444 field measurements of g_s and VPD, where a drop in g_s with an asymptotic convergence to a
445 certain minimum value of g_s is observed (e.g., Grossiord et al., 2020). Notwithstanding, the
446 previously-presented results in terms of VPD and temperatures show a diverging spatial response
447 of g_s (Figure 5), which suggests that some regions of the global wheat production area would
448 benefit from climate change in terms of the response in g_s . The foregoing provided that water
449 supply is assured, which is very unlikely to happen in areas of projected decrease in rainfall,
450 higher drought conditions and water demand (Cook et al., 2014). However, the results show a
451 dominant increasing trend in negative changes in g_s , which represents adverse conditions for
452 wheat cultivation given the limitations to transpiration and photosynthetic rate of lower g_s ,
453 (Sperry et al., 2016), as well as the soil moisture depletion and drought effects of high VPD
454 (Farahmand et al., 2021). In this sense, irrigated areas could need greater access to irrigation
455 infrastructure, and rainfed areas would most likely suffer from water stress and lower yields
456 (Zhou et al., 2019). On the other hand, cross-year effects could occur due to low g_s in drier
457 regions, over which soil moisture could remain in the soil for the next wheat growing season. In
458 this way, advances in water conservation technologies (e.g., irrigation, conservation tillage) in
459 addition to cultivars with different sensitivity to VPD could help to increase adaptation
460 capacities. Although multiple studies have quantified the fertilization effect of the higher
461 atmospheric CO₂ concentrations (e.g., Degener, 2015), which could help to offset the effect of
462 the lower g_s , its effect on wheat could be very low or even null under water stress conditions
463 partly due to changes in stomatal traits (Zheng et al., 2020).

464 The sensitivity of g_s to changes in VPD and temperature reveals the relevance that
465 increasing atmospheric water demand may have in the future, as it has been highlighted in recent
466 years (Yuan et al., 2019). The latter is evidenced in the proportion of the global wheat area that
467 responds positively to changes in temperature (Figure 6), versus the negative changes in g_s when
468 considering trends in VPD only (Figure 7), as well as the sustained decrease in g_s in global
469 averages (Figure S6 and S7 in Supporting Information S6 and S7, respectively). These results
470 highlight the importance of considering suitable parameterizations of g_s and water stress effects

471 in terms of their response to changes in VPD in current crop and terrestrial biosphere models in
472 order to have realistic estimates of the possible trajectories in crop response to changing climate.
473 However, these findings should in the future be supported by regional observations on the
474 relationship between VPD and wheat physiology, as it has been done previously for the case of
475 corn, where minor effects of increasing VPD and potential evapotranspiration on yields have
476 been projected in North America (Basso et al., 2021; Riha & Melkonian, 2022;)

477

478 *4.2 Uncertainties and limitations*

479

480 The main data source for this study corresponds to the state-of-the-art CMIP6 climate
481 projections. However, we used a modeling approach and a combination of datasets that have
482 associated multiple sources of uncertainty. For instance, we used the Jarvis multiplicative model
483 of g_s , which allows to easily isolate the effect of changing VPD and in temperature on g_s , using
484 specific parameters for wheat. Although the incorporation of other processes such as carbon
485 uptake and photosynthesis or the effect of soil moisture, which can be carried out with the Jarvis
486 or others models, would help to further analyze the potential response of g_s to climate change.
487 However, this is out of the scope of this work since we focus only on the response to VPD in
488 relation to global warming.

489 We used secondary information as boundary conditions and assumptions in terms of
490 wheat cropping calendars and phenology, which represents an important source of uncertainty
491 and limitations since we have tried to project the ecophysiological response of wheat to future
492 climate scenarios based on wheat phenology, among other information. For instance, we used
493 fixed harvest dates from Sacks et al. (2010), which are naturally variable from year to year.
494 Similarly, the SPAM for wheat area representation corresponds to a static product that might
495 over- and under-estimate the wheat cultivation area (Montes et al., 2022). The latter becomes
496 delicate when assuming that, for example, wheat development and phenology should also be
497 affected by global warming (Jägermeyr et al., 2021). However, our approach seeks at providing
498 insights of the possible long-term future trajectories that wheat cultivation could be heading,
499 assuming the deep uncertainty existing when it comes to climate change. Fortunately, genetic
500 variation in response of g_s to VPD has been observed in wheat (Tamang et al., 2022) and other
501 crops.

502

503 **5 Conclusions**

504

505 Global warming is expected to alter the functioning of agro-ecosystems in many ways.
506 One of them is in the magnitude of gas exchanges between plants and the atmosphere, which can
507 be quantified by stomatal conductance (g_s). The aim of this work was to assess the possible
508 future climate scenarios in terms of the magnitude in which g_s could be affected, representing a
509 challenge for wheat cultivation given the implications on water use and stress. Our results show a
510 significant increase in both temperatures and VPD over global wheat areas, which represents, *per*
511 *se*, a threat to wheat cultivation since VPD is a well-known driver of drought-related stresses on
512 crops. Regarding the projected response of g_s to both global warming and VPD, our results
513 suggest a differential response ranging from negative to positive climatological multi-model

514 anomalies for the three scenarios considered (SSP126, SSP370, SSP585) across wheat
515 cultivation areas. The contrasting response of g_s , with positive changes over high-latitude areas
516 versus negative changes in lower latitude areas of the Northern Hemisphere, may be indicative of
517 a relative benefit of higher temperatures over some areas, which was evidenced in the
518 temperature sensitivity analysis (g_s -*tas*; Figure 6). However, the progressive decrease of the area
519 with positive anomalies in g_s as the climate becomes warmer suggest that this effect could be
520 offset. In addition, the clear negative effect of VPD on g_s should alert about possible climate
521 change scenarios for wheat, and in turn could assist in the design and development of long-term
522 adaptation strategies. In this way, efforts could focus on the use of soil moisture conservation
523 techniques and development of new wheat varieties adapted to conditions of high atmospheric
524 water demand and drought over those areas with higher changes in g_s such as in Asia and
525 Northern Africa, where drought-related stress conditions (Cook et al., 2014) could be exacerbated
526 by atmospheric factors limiting g_s . The latter is supported by our results highlighting the trends
527 in VPD and its importance for wheat g_s . Accordingly, future efforts in understanding the
528 response of g_s to VPD in historical observational data and modeling would allow for a better
529 interpretation on how g_s could respond to future climate scenarios, and in this way to incorporate
530 more biophysical processes in the analyzes. Such results can be used to refine breeding targets so
531 that, for example, a cultivar's response of g_s to VPD suits a particular water-deficit profile
532 considering VPD and other environmental factors.

533

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538

539 **Data Availability Statement**

540 All the GCMs and boundary conditions data sets used in this study are publicly available and can
541 be accessed from the original references provided in the text. The archiving process of the
542 historical and projected VPD and stomatal conductance data sets for global wheat areas is
543 underway.

544

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