

1 **Response of dust emissions in southwestern North America to 21st**
2 **century trends in climate, CO₂ fertilization, and land use:**
3 **Implications for air quality**

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9

10 **Abstract.** Climate models predict a shift toward warmer and drier environments in southwestern
11 North America over the 21st century. The consequences of climate change for dust mobilization
12 and concentrations are unknown, but could have large implications for human health, given
13 connections between dust inhalation and disease. Here we drive a dynamic vegetation model (LPJ-
14 LMfire) with future scenarios of climate and land use, and link the results to a chemical transport
15 model (GEOS-Chem) to assess the impacts of land cover on dust mobilization and fine dust
16 concentrations (defined as dust particles less than 2.5 microns in diameter) on surface air quality.
17 In the most extreme warming scenario (RCP8.5), we find that surface temperatures in southwestern
18 North America during the season of greatest dust emissions (March, April, and May) warm by 3.3
19 K and precipitation decreases by nearly 40% by 2100. These conditions lead to vegetation dieback
20 and an increase in dust-producing bare ground. Enhanced CO₂ fertilization, however, offsets the
21 modeled effects of warming temperatures and rainfall deficit on vegetation in some areas of the
22 southwestern United States. Considering all three factors in RCP8.5 scenario, dust concentrations

23 decrease over Arizona and New Mexico in spring by the late 21st century due to greater CO₂
24 fertilization and a more densely vegetated environment, which inhibits dust mobilization. Along
25 Mexico's northern border, dust concentrations increase as a result of land use intensification. In
26 contrast, when CO₂ fertilization is not considered in the RCP8.5 scenario, vegetation cover
27 declines significantly across most of the domain by 2100, leading to widespread increases in fine
28 dust concentrations, especially in southeastern New Mexico (up to ~2.0 $\mu\text{g m}^{-3}$ relative to the
29 present day) and along the border between New Mexico and Mexico (up to ~2.5 $\mu\text{g m}^{-3}$). Our
30 results have implications for human health, especially for the health of the indigenous people who
31 make up a large percentage of the population in this region.

32 **1 Introduction**

33 The arid and semi-arid regions of the southwestern United States and northwestern Mexico
34 are characterized by large concentrations of soil-derived dust particles in the lower atmosphere,
35 especially in spring (Hand et al., 2016). By causing respiratory and cardiovascular diseases, fine
36 dust particles – i.e., those particles with diameter less than 2.5 microns – can have negative effects
37 on human health (Tong et al., 2017; Meng and Lu, 2007; Gorris et al., 2018). A key question is to
38 what extent climate change and other factors will influence future dust concentrations in this region,
39 which we define here as southwestern North America. In this study, we use a suite of models to
40 predict the future influence of three factors – climate change, increasing atmospheric CO₂
41 concentrations, and land use change – on land cover in this region, and assess the consequences
42 for dust mobilization and dust concentrations.

43 Wind speed and vegetation cover are two key factors that determine soil erodibility and
44 dust emissions. Wind gusts mobilize dust particles from the earth's surface, while vegetation
45 constrains dust emissions by reducing the extent of bare land and preserving soil moisture (Zender
46 et al., 2003). The high temperatures and reduced soil moisture characteristic of drought play an
47 important role in dust mobilization, since loss of vegetative cover during drought increases soil
48 erosion (Archer and Predick, 2008; Bestelmeyer et al., 2018).

49 Southwestern North America is covered by desert grassland, perennial grassland, savanna,
50 desert scrub, and grassy shrublands or woodlands (McClaran and Van Devender, 1997). In recent
51 decades, a gradual transition from grasslands to shrubland has been observed across much of this
52 region, with increased aridity, atmospheric CO₂ enrichment, and livestock grazing all possibly
53 playing a role in this trend (Bestelmeyer et al., 2018). Future climate change may further prolong
54 this transition, especially since shrubs fare better than grasses under a climate regime characterized

55 by large fluctuations in annual precipitation (Bestelmeyer et al., 2018; Edwards et al., 2019).
56 Climate models predict a warmer and drier environment in southwestern North America through
57 the 21st century, with more frequent and severe drought (Seager and Vecchi, 2010; MacDonald,
58 2010; Stahle, 2020; Prein et al., 2016; Williams et al., 2020). Such conditions would decrease
59 vegetative cover and allow for greater dust mobilization. On the other hand, elevated CO₂
60 concentrations in the future atmosphere could increase photosynthesis and decrease transpiration
61 of some vegetation species, allowing for more efficient water use and enhancing growth (Poorter
62 and Perez-Soba, 2002; Polley et al., 2013). Land use practices, e.g., farming and ranching,
63 industrial activities including mining, and urban sprawl, have changed dramatically over the
64 southwestern North America in recent decades, with Arizona and New Mexico showing decreasing
65 cropland area and northern Mexico experiencing increasing pasture area (Figure S1). Future land
66 use practices could also influence the propensity for dust mobilization by disturbing crustal
67 biomass (e.g., Belnap and Gillette, 1998).

68 Previous studies have investigated the relative importance of climate, CO₂ fertilization,
69 and/or land use in present-day and future dust emissions and concentrations, sometimes with
70 contradictory results. For example, Woodward et al., 2005 predicted a tripling of the global dust
71 burden by 2100 relative to the present day, while other studies suggested a decrease in the global
72 dust burden (e.g., Harrison et al., 2001, Mahowald and Luo, 2003 and Mahowald et al., 2006).
73 These estimates of future dust emissions depended in large part on the choice of model applied, as
74 demonstrated by Tegen et al., 2004.

75 In southwestern North America, a few recent studies examined statistical relationships
76 between observed present-day dust concentrations and meteorological conditions or leaf area index
77 (LAI). Hand et al., 2016 found that fine dust concentrations in spring in this region correlated with

78 the Pacific Decadal Oscillation (PDO), indicating the importance of large-scale climate patterns in
79 the mobilization and transport of regional fine dust. Tong et al., 2017 further determined that the
80 observed 240% increase in the frequency of windblown dust storms from 1990s to 2000s in the
81 southwestern United States was likely associated with the PDO. Similarly, Achakulwisut et al.,
82 2017 found that the 2002–2015 increase in average March fine dust concentrations in this region
83 was driven by a combination of positive PDO conditions and phase of the El Nino-Southern
84 Oscillation. More recently, Achakulwisut et al., 2018 identified the Standardized Precipitation-
85 Evapotranspiration Index as a useful indicator of present-day dust variability. Applying that metric
86 to an ensemble of future climate projections, these authors predicted increases of 26-46% in fine
87 dust concentrations over the U.S. Southwest in spring by 2100. In contrast, Pu and Ginoux, 2017
88 found that the frequency of extreme dust days decreases slightly in spring in this region due to
89 reduced extent of bare ground under 21st century climate change.

90 These regional studies relied mainly on statistical models that relate local and/or large scale
91 meteorological conditions to dust emissions in southwestern North America. Pu and Ginoux, 2017
92 also considered changing LAI in their model, but these dust-LAI relationships were derived from
93 a relatively sparse dataset, casting some uncertainty on the results (Achakulwisut et al., 2018). In
94 this study, we investigate the effects of climate change, increasing CO₂ fertilization, and future
95 land use practices on vegetation in southwestern North America, and we examine the response of
96 dust mobilization due to these changes in vegetation. With regard to climate, we examine whether
97 a shift to warmer, drier conditions by 2100 enhances dust mobilization in this region by reducing
98 vegetation cover and exposing bare land. To that end, we couple the LPJ-LMfire dynamic
99 vegetation model to the chemical transport model GEOS-Chem to study vegetation dynamics and
100 dust mobilization under different conditions and climate scenarios, allowing consideration of

101 several factors driving future dust mobilization in the southwestern North America. We focus on
102 fine dust particles in springtime (March, April, and May), because it is the season of highest dust
103 concentrations in the southwestern U.S. (Hand et al., 2017). Given the deleterious impacts of
104 airborne dust on human health, our dust projections under different climate scenarios have value
105 for understanding the full array of potential consequences of anthropogenic climate change.

106

107 **2 Methods**

108 We examine dust mobilization in southwestern North America, here defined as 25°N –
109 37°N, 100°W – 115°W (Figure 1), during the late-21st century under scenarios of future climate
110 and land use based on two Representative Concentration Pathways (RCPs). RCP4.5 and RCP8.5
111 capture two possible climate trajectories over the 21st century, beginning in 2006. RCP4.5
112 represents a scenario of moderate future climate change with gradual reduction in greenhouse gas
113 (GHG) emissions after 2050 and a radiative forcing at 2100 relative to pre-industrial values of +4.5
114 W m⁻², while RCP8.5 represents a more extreme scenario with continued increases in GHGs
115 throughout the 21st century and a radiative forcing of +8.5 W m⁻² at 2100. For each RCP, we
116 investigate the changes in vegetation for three cases: 1) an all-factor case that includes changes in
117 climate, land use, and CO₂ fertilization; 2) a fixed-CO₂ case that includes changes in only climate
118 and land use; and 3) a fixed-land use case that includes changes in only climate and CO₂
119 fertilization.

120 We use LPJ-LMfire, a dynamic global vegetation model, to estimate changes in vegetation
121 under future conditions (Pfeiffer et al., 2013). Meteorology to drive LPJ-LMfire is taken from the
122 Goddard Institute for Space Studies (GISS) climate model (Nazarenko et al., 2015). Using the
123 GEOS-Chem emission component (HEMCO), we then calculate dust emissions based on the LPJ-

124 generated vegetation area index (VAI) for all scenarios. We apply the resulting dust emissions to
125 the global chemical transport model GEOS-Chem to simulate the distribution of fine dust across
126 the southwestern North America.

127

128 **2.1 GISS Model E**

129 Present-day and future meteorological fields for RCP4.5 and RCP8.5 are simulated by the
130 GISS Model E climate model (Nazarenko et al., 2015), configured for Phase 5 of the Coupled
131 Model Intercomparison Project (CMIP5; <https://esgf-node.llnl.gov/search/cmip5/>, last accessed on
132 17 July 2020). The simulations cover the years 1801 to 2100 at a spatial resolution of 2° latitude x
133 2.5° longitude. Changes in climate in the GISS model are driven by increasing greenhouse gases.
134 In RCP4.5, CO₂ concentrations increase to 550 ppm by 2100; in RCP8.5 the CO₂ increases to 1960
135 ppm ((Meinshausen et al., 2011).

136 Under RCP4.5, the GISS model predicts a slight increase of 0.45 K in springtime mean
137 surface temperatures and an increase in mean precipitation by ~17% over the southwestern North
138 America by the 2100 time slice (2095-2099), relative to the present day (2011-2015). In contrast,
139 under RCP8.5, the 5-year mean springtime temperature increases significantly by 3.29 K by 2100
140 and mean precipitation decreases by ~39%. The spatial distributions of the changes in temperature
141 and precipitation by 2100 under RCP8.5 are presented in the Supplement (Figure S2). In addition,
142 lightning strike densities decrease by ~0.006 strikes km⁻² d⁻¹ over Arizona in RCP4.5, but increase
143 by the same magnitude in this region in RCP8.5 (Li et al., 2020). Lightning strikes play a major
144 role for wildfire ignition in this region, while wildfires may influence landscape succession (e.g.,
145 Bodner and Robles, 2017). Finally, future surface wind speeds do not change significantly under
146 RCP4.5, but increase slightly by ~4% across southwestern North America under RCP8.5 by 2100

147 (not shown). The increasing winds in RCP8.5 will influence the spread of fires in our study, but
148 will not affect the simulated dust fluxes directly, as described in more detail below. Compared to
149 those from other climate models, the GISS projections of climate change in southwestern North
150 America are conservative (Ahlström et al., 2012; Sheffield et al., 2013), implying that our
151 predictions of the impact of climate change on dust mobilization may also be conservative.

152 In our study, we do not specifically track drought frequency under future climate, as the
153 definition of drought is elusive (Andreadis et al., 2005; Van Loon et al., 2016). Nonetheless, the
154 meteorological conditions predicted in the RCP8.5 scenario for 2100 align with previous studies
155 projecting increased risk of drought in this region (e.g., Williams et al., 2020), and as we shall see,
156 such conditions, in the absence of CO₂ fertilization, result in decreased vegetation and greater dust
157 mobilization.

158 **2.2 LPJ-LMfire**

159 LPJ-LMfire is a fork of the LPJ dynamic vegetation model (Sitch et al., 2003) that includes
160 a process-based representation of fire (Pfeiffer et al., 2013). Input to LPJ-LMfire includes
161 meteorological variables, soil characteristics, land use, and atmospheric CO₂ concentrations, and
162 the model then simulates the corresponding vegetation structure, biogeochemical cycling, and
163 wildfire at a spatial resolution of 0.5° latitude x 0.5° longitude. Here “vegetation structure” refers
164 to vegetation types and the spatial patterns in landscapes.

165 More specifically, LPJ-LMfire simulates the impacts of photosynthesis, evapotranspiration,
166 and soil water dynamics on vegetation structure and the population densities of different plants
167 functional types (PFTs). The model considers the coupling of different ecosystem processes, such
168 as the interactions between CO₂ fertilization, evapotranspiration, and temperature, as well as the
169 competition among different PFTs for water resources (e.g., precipitation, surface runoff, and

170 drainage). The different PFTs in LPJ-LMfire respond differently to changing CO₂, with CO₂
171 enrichment preferentially stimulating photosynthesis in woody vegetation and C₃ grasses
172 compared to C₄ grasses (Polley et al., 2013). Wildfire in LPJ-LMfire depends on lightning ignition,
173 and the simulation considers multiday burning, coalescence of fires, and the spread rates of
174 different vegetation types. The effects of changing fire activity on vegetation cover are then taken
175 into account (Pfeiffer et al., 2013; Sitch et al., 2003; Chaste et al., 2019). Li et al., 2020 predicted
176 a ~50% increase in fire-season area burned by 2100 under scenarios of both moderate and intense
177 future climate change over the western United States. However, the effects of changing fire on
178 vegetation cover are insignificant in the grass and bare ground-dominated ecosystems of the desert
179 Southwest, where the low biomass fuels cannot support extensive spread of fires.

180 For this study we follow Li et al., 2020, in linking meteorology from GISS-E2-R to LPJ-
181 LMfire in order to capture the effects of climate change on vegetation. Meteorological fields from
182 the GISS model include monthly mean surface temperature, diurnal temperature range, total
183 monthly precipitation, number of days in the month with precipitation greater than 0.1 mm,
184 monthly mean total cloud cover fraction, and monthly mean surface wind speed. Monthly mean
185 lightning strike density, calculated using the GISS convective mass flux and the empirical
186 parameterization of Magi, 2015, is also applied to LPJ-LMfire. To downscale the 2° x 2.5° GISS
187 meteorology to finer resolution for LPJ-LMfire, we calculate the 2010-2100 monthly anomalies
188 relative to the average over the 1961-1990 period, and then add these anomalies to an
189 observationally based climatology (Pfeiffer et al., 2013). LPJ-LMfire then simulates the response
190 of natural vegetation to the 21st century trends in these meteorological fields and to increasing CO₂.
191 We apply the same changes in CO₂ concentrations as those applied to the GISS model.

192 We overlay the changes in natural land cover with future land use scenarios from CMIP5

193 (LUH; Hurtt et al., 2011; <http://tntcat.iiasa.ac.at/RcpDb/>, last accessed on 17 July 2020). These
194 scenarios include land used for crops, ranching (rangeland), and urban areas, all of which result in
195 reduction in aboveground biomass, an increase in herbaceous relative to woody plants, and an
196 increase in the extent of bare ground. The present-day land use in the LUH dataset is taken from
197 the HYDE database v3.1 (Goldewijk, 2001; Goldewijk et al., 2010), which in turn is based on
198 array of sources, including satellite observations and government statistics. In RCP8.5, the extent
199 of crop- and rangeland cover increases by ~30% in Mexico but decreases by 10-20% over areas
200 along Mexico's northern border in the U.S. (Hurtt et al., 2011). Only minor changes in land use
201 practices by 2100 are predicted under RCP4.5 (Hurtt et al., 2011).

202 We perform global simulations with LPJ-LMfire on a $0.5^\circ \times 0.5^\circ$ grid for the two RCPs
203 from 2006-2100, and analyze results over southwestern North America, where dust emissions are
204 especially high. For each RCP we consider the effects of changing climate on land cover, as well
205 as the influence of land use change and CO₂ fertilization. The LPJ-LMfire simulations yield
206 monthly timeseries of the leaf area indices (LAI) and fractional vegetation cover (σ_v) for nine plant
207 functional types (PFTs): tropical broadleaf evergreen, tropical broadleaf raingreen, temperate
208 needleleaf evergreen, temperate broadleaf evergreen, temperate broadleaf summergreen, boreal
209 needleleaf evergreen, and boreal summergreen trees, as well as C₃ and C₄ grasses. We further
210 discuss the LPJ-LMfire present-day land cover in the Supplement.

211 **2.3 VAI calculation**

212 Vegetation constrains dust emissions in two ways: 1) by competing with bare ground as a
213 sink for atmospheric momentum, which results in less drag on erodible soil (Nicholson et al., 1998;
214 Raupach, 1994); and 2) by enhancing soil moisture through plant shade and root systems (Hillel,
215 1982). Here we implement the dust entrainment and deposition (DEAD) scheme of Zender et al.,

216 2003 to compute a size-segregated dust flux, which includes entrainment thresholds for saltation,
217 moisture inhibition, drag partitioning, and saltation feedback. The scheme assumes that vegetation
218 suppresses dust mobilization by linearly reducing the fraction of bare soil exposed in each grid
219 cell:

$$220 \quad A_m = (1 - A_l - A_w)(1 - A_s)(1 - A_v) \quad (1),$$

221 where A_l is the fraction of land covered by lakes, A_w is the fraction covered by wetlands, A_s is the
222 fraction covered by snow, and A_v is the fraction covered by vegetation.

223 For this study, we use VAI as a metric to represent vegetation because it includes not only
224 leaves but also stems and branches, all of which constrain dust emission. VAI is used to calculate
225 A_v in equation (1) through

$$226 \quad A_v = \min [1.0, \min(VAI, VAI_t) / VAI_t] \quad (2),$$

227 where VAI_t is the threshold for complete suppression of dust emissions, set here to $0.3 \text{ m}^2 \text{ m}^{-2}$
228 (Zender et al., 2003; Mahowald et al., 1999).

229 To compute the dust fluxes, we need to convert LAI from LPJ-LMfire to VAI. VAI is
230 generally defined as the sum of LAI plus stem area index (SAI). Assuming immediate removal of
231 all dead leaves, the fractional vegetation cover, σ_v , can be used to represent SAI for the different
232 PFTs (Zeng et al., 2002). Given that the threshold VAI_t for no dust emission is relatively low (0.3
233 $\text{m}^2 \text{ m}^{-2}$), leaf area dominates stem area in the suppression of dust mobilization in the model. In
234 areas where LAI is greater than SAI, we therefore assume that SAI does not play a role in
235 controlling dust emissions, and we set LAI equivalent to VAI. We also assume that C_3 and C_4
236 grasses have zero stem area to avoid overestimating VAI during the winter and early spring when
237 such grasses are dead. Based on the method of Zeng et al., 2002, with modifications, we calculate
238 VAI in each grid cell as

239
$$VAI = \max (\sum_{PFT=1}^9 LAI, \sum_{PFT=1}^7 \sigma_v) \quad (3)$$

240 where LAI is for the nine PFTs from LPJ-LMfire, and σ_v is for just seven PFTs, with σ_v for C₃
241 and C₄ grasses not considered. Of the nine PFTs, temperate needleleaf evergreen, temperate
242 broadleaf evergreen, temperate broadleaf summergreen, and C₃ grasses dominate the region, with
243 temperate needleleaf evergreen having the highest LAI in spring. This mix of vegetation type is
244 consistent with observations (e.g., McClaran and Van Devender, 1997).

245 **2.4 Calculation of dust emissions**

246 Dust emissions are calculated offline in the DEAD dust mobilization module within the
247 Harvard-NASA Emissions Component (HEMCO). We feed into the DEAD module both the VAI
248 generated by LPJ-LMfire and meteorological fields from the Modern-Era Retrospective analysis
249 for Research and Applications (MERRA-2) at a spatial resolution of 0.5° latitude x 0.625°
250 longitude (Gelaro et al., 2017). Dust emission is nonlinear with surface windspeed. Following
251 Ridley et al., 2013, we characterize subgrid-scale surface winds as a Weibull probability
252 distribution, which allows saltation even when the grid-scale wind conditions are below some
253 specified threshold speed. The scheme assumes that the vertical flux of dust is proportional to the
254 horizontal saltation flux, which in turn depends on surface friction velocity and the aerodynamic
255 roughness length Z_0 . As recommended by Zender et al., 2003, and consistent with Fairlie et al.,
256 2007 and Ridley et al., 2013, we uniformly set Z_0 to 100 μm across all dust candidate grid cells.

257 With this model setup, we calculate hourly dust emissions for two five-year time slices for
258 each RCP and condition, covering the present day (2011-2015) and the late-21st century (2095-
259 2099). Dust emissions are generated for four size bins with radii of 0.1 – 1.0 μm , 1.0 – 1.8 μm , 1.8
260 – 3.0 μm , 3.0 – 6.0 μm . These dust emissions are then applied to GEOS-Chem. Calculated present-
261 day VAI and fine dust emissions are shown in Figure S3, and we compare modeled VAI with that

262 observed in Figures S4 and S5.

263

264 **2.5 GEOS-Chem**

265 We use the aerosol-only version of the GEOS-Chem chemical transport model (version
266 12.0.1; <http://acmg.seas.harvard.edu/geos/>). For computational efficiency, we apply monthly mean
267 oxidants archived from a full-chemistry simulation (Park et al., 2004). To isolate the effect of
268 changing dust mobilization on air quality over the southwestern North America, we use present-
269 day MERRA-2 reanalysis meteorology from NASA/GMAO (Gelaro et al., 2017) for both the
270 present-day and future GEOS-Chem simulations. In other words, we neglect the direct effects of
271 future changes in wind speeds on dust mobilization, allowing us to focus instead on the indirect
272 effects of changing vegetation on dust. For each time slice, we first carry out a global GEOS-Chem
273 simulation at 4° latitude x 5° longitude spatial resolution, and then downscale to 0.5° x 0.625° via
274 grid nesting over the North America domain. In this study, we focus only on dust particles in the
275 finest size bin (i.e., with radii of 0.1 – 1.0 μm), as these are most deleterious to human health. We
276 compare modeled fine dust concentrations over southwestern North America for the present-day
277 against observations from the IMPROVE network in Figures S6-S7.

278

279 **3 Results**

280 **3.1 Spatial shifts in springtime vegetation area index**

281 Figure 1 shows large changes in the spatial distribution of modeled springtime VAI in the
282 southwestern North America for the three cases under both RCPs by 2100. In RCP4.5, the
283 distributions of changes in VAI are similar for the all-factor and fixed-land use cases. Strong
284 enhancements (up to ~2.5 m² m⁻²) extend across much of Arizona, especially in the northwestern

285 corner. The model exhibits moderate VAI increases in most of New Mexico and in the forest
286 regions along the coast of northwestern Mexico. We find decreases in modeled VAI (up to ~ -1.6
287 $\text{m}^2 \text{m}^{-2}$) in the southwestern corner of New Mexico, to the east of the coastal forests in Mexico and
288 in the forest regions near the Mexican border connecting with southern Texas. The similarity
289 between the all-factor and fixed land use cases indicates the relatively trivial influence of land use
290 change on vegetation cover in RCP4.5, compared to the effects of climate change and CO_2
291 fertilization. For the fixed- CO_2 case, western New Mexico and northern Mexico show greater
292 decreases in VAI, indicating how CO_2 fertilization in the other two cases offsets the effects of the
293 warmer and drier climate on vegetation in this region. Figure S8 further illustrates the strong
294 positive impacts that CO_2 fertilization has on VAI.

295 Compared to RCP4.5, the RCP8.5 scenario shows larger changes in climate, CO_2
296 concentrations, and land use by 2100 (Figure 1). The net effects of these changes on vegetation
297 are complex. As in RCP4.5, Arizona experiences a strong increase in VAI in the all-factor and
298 fixed-land use cases, but now this increase extends to New Mexico. In contrast to RCP4.5, modeled
299 VAI decreases in the northern Sierra Madre Occidental (Mexico) in the all-factor case for RCP8.5.
300 In the fixed-land use case, however, the VAI decrease in northern Mexico is nearly erased,
301 indicating the role of vegetation/forest degradation caused by land use practices in this area (Figure
302 S9). For the fixed- CO_2 case for RCP8.5, VAI decreases in nearly all of southwestern North
303 America, except the northeastern corner of Arizona and the northwestern corner of New Mexico.

304 To better understand the changes in VAI, we examine changes in LAI, which represents
305 the major portion of VAI, for the four dominant plant functional types (PFTs) in this region. For
306 example, decreases in LAI in the fixed- CO_2 case under RCP8.5 are dominated by the loss of
307 temperate broadleaf evergreen (TeBE) and temperate broadleaf summergreen (TeBS) (Figure S10).

308 Temperate needleleaf evergreen (TeNE) shows areas of increase in the northern part and south of
309 Texas in this scenario, while both TeBE and TeBS show increases in northern Arizona and New
310 Mexico. In other areas, TeBS reveals strong decreases, especially in southern Arizona and Mexico.
311 As predicted by previous studies (Bestelmeyer et al., 2018; Edwards et al., 2019), C₃ perennial
312 grasses (C₃gr) in this case decrease across a large swath extending from Arizona through Mexico,
313 showing the impacts of warmer temperatures and reduced precipitation, as well as (for Mexico)
314 land use change. Increased fire activity also likely plays a role in the simulated decreases of forest
315 cover and C₃ grasses for RCP8.5 in southern Arizona, where fires together with drought may have
316 affected landscape succession (Williams et al., 2013; Bodner and Robles, 2017). We also
317 investigate trends in LAI for different months in spring from the present day to 2100. We find that
318 the greatest percentage decreases in TeBS and C₃ grasses occur in May, consistent with the largest
319 decreases in precipitation in that month (not shown).

320 In summary, we find that the warmer and drier conditions of the future climate strongly
321 reduce vegetation cover by 2100, especially in RCP8.5. In addition, CO₂ fertilization and land use
322 practices further modify future vegetation, but in opposite ways, as illustrated by Figure S8. Under
323 a warmer climate, higher CO₂ concentrations facilitate vegetation growth everywhere in the
324 southwestern North America, with larger VAI increases occurring over Arizona and New Mexico.
325 Combined changes in land use are greater under RCP8.5 than RCP4.5, with large increases in
326 RCP8.5 across Mexico but only modest changes in Arizona, New Mexico, and Texas (Figure S9).
327 The increases in Mexico result in the fragmentation of forested landscapes and decrease VAI,
328 especially in coastal forest regions and along the border with the United States.

329 **3.2 Spatial variations in spring fine dust emissions**

330 Unlike the widespread changes in VAI, future changes in fine dust emissions are

331 concentrated in a few arid areas, including: 1) the border regions connecting Arizona, New Mexico,
332 and northern Mexico (ANM border), 2) eastern New Mexico, and 3) western Texas (Figure 2). In
333 RCP4.5, slight increases in fine dust emission (up to $\sim 0.3 \text{ kg m}^{-2} \text{ mon}^{-1}$) are simulated in the ANM
334 border in all the three cases. In contrast, fine dust emissions decrease by up to $\sim -1.0 \text{ kg m}^{-2} \text{ mon}^{-1}$
335 in eastern New Mexico and western Texas in RCP4.5 due to warmer temperatures and increasing
336 VAI. Consistent with the modest changes in VAI (Figure 1), the three cases in RCP4.5 do not
337 exhibit large differences, with only the fixed- CO_2 case showing slightly greater increases in dust
338 emissions along the ANM border and in western Texas. In RCP8.5 in the all-factor case, spring
339 fine dust emissions increase slightly by up to $\sim 0.4 \text{ kg m}^{-2} \text{ mon}^{-1}$ along the ANM border, but
340 decrease more strongly in western Texas by up to $\sim -1.4 \text{ kg m}^{-2} \text{ mon}^{-1}$ (Figure 2). In contrast, with
341 fixed CO_2 the sign of the change in dust emissions reverses, with significant emissions increases
342 along the ANM border and in New Mexico. The area with decreasing emissions in western Texas
343 also shrinks in the fixed CO_2 case. These trends occur due to the climate stresses, e.g., warmer
344 temperatures and decreased precipitation, that impair the growth of temperature broadleaf trees
345 and C_3 grasses. In this case, such stresses are not offset by CO_2 fertilization (Figure S10).

346 Figure 3 shows more vividly the opposing roles of CO_2 fertilization and projected land use
347 change in southwestern North America. In RCP8.5, changing CO_2 fertilization alone promotes
348 vegetation growth and dramatically reduces dust mobilization by up to $\sim -1.2 \text{ kg m}^{-2} \text{ mon}^{-1}$. Figure
349 3 also reveals that land use trends are a major driver of increased dust emissions along the ANM
350 border and western Texas in RCP8.5, as crop- and rangelands expand in this region and
351 temperature broadleaf trees decline (Hurtt et al., 2011). Similarly, the expansion of rangelands in
352 northern Mexico in RCP8.5 reduces natural vegetation cover there (Hurtt et al., 2011), contributing
353 to the increase of fine dust emissions by up to $\sim 0.7 \text{ kg m}^{-2} \text{ mon}^{-1}$.

354 **3.3 Spring fine dust concentrations under the high emission scenario**

355 Our simulations suggest that fine dust emissions will increase across arid areas in
356 southwestern North America under RCP8.5, but only if CO₂ fertilization is of minimal importance
357 (Figure 2). To place an upper bound on future concentrations of fine dust in this region, we apply
358 only the fixed-CO₂ emissions to GEOS-Chem at the horizontal resolution of 0.5° x 0.625°. Given
359 the large uncertainty in the sensitivity of vegetation to changing atmospheric CO₂ concentrations
360 (Smith et al., 2016), we argue that this approach is justified.

361 Results from GEOS-Chem in the fixed-CO₂ case for RCP8.5 show that the concentrations
362 of spring fine dust are significantly enhanced in the southeastern half of New Mexico and along
363 the ANM border, with increases up to ~2.5 μg m⁻³ (Figure 4). The model also yields elevated dust
364 concentrations over nearly the entire extent of our study region by 2100. As Figure 3 implies, land
365 use along the ANM border contributes to the increased dust emissions in that area, by up to ~0.7
366 kg m⁻² mon⁻¹. Climate change impacts on natural vegetation, however, account for the bulk of the
367 modeled increases in dust emissions in this scenario, by as much as ~1.2 kg m⁻² mon⁻¹ (Figure 2).
368 The modeled wind fields, which are the same in all scenarios, transport the dust from source
369 regions, leading to the enhanced concentrations across much of the domain, as seen in Figure 4.
370 We find that dust concentrations decrease only in a limited area in western Texas due to decreased
371 pasture (Figures 3 and S9).

372

373 **4 Discussion**

374 We apply a coupled modeling approach to investigate the impact of future changes in
375 climate, CO₂ fertilization, and land use on dust mobilization and fine dust concentration in
376 southwestern North America by the end of the 21st century. Table 1 summarizes our findings for

377 the two RCP scenarios and three conditions – all-factor, fixed CO₂, and fixed land use – in spring,
378 when dust concentrations are greatest. We find that in the RCP8.5 fixed-CO₂ scenario, in which
379 the effects of CO₂ fertilization are neglected, VAI decreases by 26% across the region due mainly
380 to warmer temperatures and drier conditions, yielding an increase of 58% in fine dust emission
381 averaged over the southwestern North America. In addition, we find that the increase in fine dust
382 emission in northern Mexico is mainly driven by the increases in the extent of cropland and pasture
383 cover in this area, signifying the crucial role of land use practices in modifying dust mobilization.

384 Our findings of decreasing VAI with future climate change are consistent with observed
385 trends in vegetation during recent droughts in this region. For example, Breshears et al., 2005
386 documented large-scale die-off of overstory trees across southwestern North America in 2002-
387 2003 in response to short-term drought accompanied by bark beetle infestations. Similarly, during
388 a multi-year (2004-2014) drought in southern Arizona, Bodner and Robles, 2017 found that the
389 spatial extent of both C₄ grass cover and shrub cover decreased in the southeastern part of that
390 state.

391 The 58% increase predicted in this study in fixed-CO₂ RCP8.5 scenario is larger than the
392 26-46% future increases in fine dust for this region predicted by the statistical model of
393 Achakulwisut et al., 2018. That study relied solely on predictions of future regional-scale
394 meteorology and did not take into account the change in vegetation, as we do here. In contrast, the
395 statistical model of Pu and Ginoux, 2017 estimated a 2% decrease in the springtime frequency of
396 extreme dust events in the Southwest U.S., driven mainly by reductions in bare ground fraction
397 and wind speed. Like Pu and Ginoux, 2017, we also find that dust emissions decrease across a
398 broad region of the Southwest when CO₂ fertilization is taken into account, as shown in Figure 2.
399 Pu and Ginoux, 2017 relied on limited data for capturing the sensitivity of dust event frequency to

400 land cover in this region, and neither that study nor Achakulwisut et al., 2018 considered changes
401 in land use, as do here. The direct effects of changing wind speed on dust mobilization, however,
402 are not included in our study, but could be tested in future work.

403 We further find that consideration of CO₂ fertilization can mitigate the effects of changing
404 climate and land use on dust concentrations in southwestern North America. The all-factor and
405 fixed-land use simulations both yield decreases of ~20% in mean dust emissions compared to the
406 early 21st century. In the IPCC projections, CO₂ reaches ~550 ppm by 2100 under RCP4.5 and
407 ~1960 ppm under RCP8.5 (Meinshausen et al., 2011). Correspondingly, in the RCP4.5 scenario
408 for 2100, CO₂ fertilization enhances VAI by 30% in the all-factor case compared to the fixed-CO₂
409 case (1.07 m²m⁻² vs. 0.79 m²m⁻²); in RCP 8.5, the 2100 enhancement is 64% (1.11 m²m⁻² vs. 0.55
410 m²m⁻²), as shown in Table 1. These enhancements further decrease fine dust emissions by 21%
411 under RCP4.5 and 78% under RCP8.5, compared to the present day. Except along the ANM border
412 and a few other areas, trends in land use have only minor impacts on dust mobilization under the
413 two RCPs in southwestern North America.

414 In summary, we find that as atmospheric CO₂ levels rise vegetation growth is enhanced
415 and dust mobilization decreases, offsetting the impacts of warmer temperatures and reduced
416 rainfall, at least in some areas. These results are consistent with evidence that CO₂ fertilization is
417 already occurring in arid or semiarid environments like southwestern North America (Donohue et
418 al., 2013; Haverd et al., 2020). In such environments, water availability is the dominant constraint
419 on vegetation growth, and the recent increases in atmospheric CO₂ may have reduced stomatal
420 conductance and limited evaporative water loss. The effects of CO₂ fertilization on vegetation
421 growth are uncertain, however, and may be attenuated by the limited supply of nitrogen and

422 phosphorus in soil (Wieder et al., 2015). These nutritional constraints vary greatly among different
423 PFTs (Shaw et al., 2002; Nadelhoffer et al., 1999).

424 Understanding the drivers in historic dust trends has sometimes been challenging
425 (Mahowald and Luo, 2003; Mahowald et al., 2002), making it difficult to validate dust
426 mobilization models. A further drawback of our approach is that the LPJ-LMfire model is driven
427 by meteorological fields from just one climate model, GISS-E2-R. Given that the GISS model
428 yields a conservative prediction of climate change in the southwestern North America compared
429 to other models (Ahlström et al., 2012; Sheffield et al., 2013), our predictions of the impact of
430 climate change on dust mobilization may also be conservative. Other uncertainties in our study
431 can be traced to the dust simulation. The different vegetation types in our model are quantified as
432 fractions of gridcells, which have relatively large spatial dimensions of $\sim 50 \text{ km} \times 60 \text{ km}$. This
433 means the model cannot capture the spatial heterogeneity of land cover, and the aerodynamic
434 sheltering effects of vegetation on wind erosion are neglected, as they are in most 3-D global model
435 studies. Such sheltering could play a large role in dust mobilization (e.g., Liu et al., 1990). New
436 methods involving satellite observations of surface albedo promise to improve understanding of
437 the effects of aerodynamic sheltering on dust mobilization, at least for the present-day (Chappell
438 and Webb, 2016; Webb and Pierre, 2018). Implementation of aerodynamic sheltering in
439 simulations of future climate regimes would need to account for fine-scale spatial distributions of
440 vegetation. In addition, as recommended by Zender et al., 2003, we apply a globally uniform
441 surface roughness Z_0 in the model, which means that the impact of changing vegetation conditions
442 on friction velocity is not taken into account. Future work could address this weakness by varying
443 friction velocity according to vegetation type. Finally, our study focuses only on the effect of
444 changing vegetation on dust mobilization and does not take into account how changing windspeeds

445 or drier soils in the future atmosphere may more directly influence dust. Given the slight increase
446 in monthly mean winds in RCP8.5 by 2100, future dust emissions in this scenario could be
447 underestimated.

448 Within these limitations, our study quantifies the potential impacts of changing land cover
449 and land use practices on dust mobilization and fine dust concentration over the coming century
450 in southwestern North America. Our work builds on previous studies focused on future dust in this
451 region by (1) more accurately capturing the transport of dust from source regions with a dynamical
452 3-D model, (2) considering results with and without CO₂ enhancement, and (3) including the
453 impact of land use trends. Given the many uncertainties, it is challenging to gauge which of the
454 three factors investigated here – climate impacts on vegetation, CO₂ fertilization, or land use
455 change – will play the dominant role in driving future changes in dust emissions and concentrations.
456 This study thus brackets a range of possible dust scenarios for the southwestern North America,
457 with the simulation without CO₂ fertilization placing an upper bound on dust emissions. In the
458 absence of increased CO₂ fertilization, our work suggests that vegetated cover will contract in
459 response to the warmer, drier climate, exposing bared ground and significantly increasing dust
460 concentrations by 2100. In this way, dust enhancement could impose a potentially large climate
461 penalty on PM_{2.5} air quality, with consequences for human health across much of southwestern
462 North America.

463 Our finding of the potential for an increased dust burden in the future atmosphere has
464 special relevance for environmental justice in this region, where much of the current population is
465 of Native American and/or Latino descent. For example, in New Mexico, 10% of the population
466 is Native American and 50% identifies as either Hispanic or Latino. By some measures, New

467 Mexico has also one of highest poverty rates of the United States (<https://www.census.gov>
468 /quickfacts/NM, last accessed on August 20, 2020).

469

470 **Code and data availability**

471 GEOS-Chem model codes can be obtained at <http://acmg.seas.harvard.edu/geos>. LPJ-LMfire
472 model codes can be obtained at <https://github.com/ARVE-Research/LPJ-LMfire>. IMPROVE
473 datasets are available online at <http://vista.cira.colostate.edu/improve>. Any additional information
474 related to this paper may be requested from the authors.

475

476 **Author contributions**

477 Y.L. conceived and designed the study, performed the GEOS-Chem simulations, analyzed the data,
478 and wrote the manuscript, with contributions from all coauthors. J.O.K. performed the LPJ-LMfire
479 simulations.

480

481 **Competing interests**

482 The authors declare that they have no competing interest.

483

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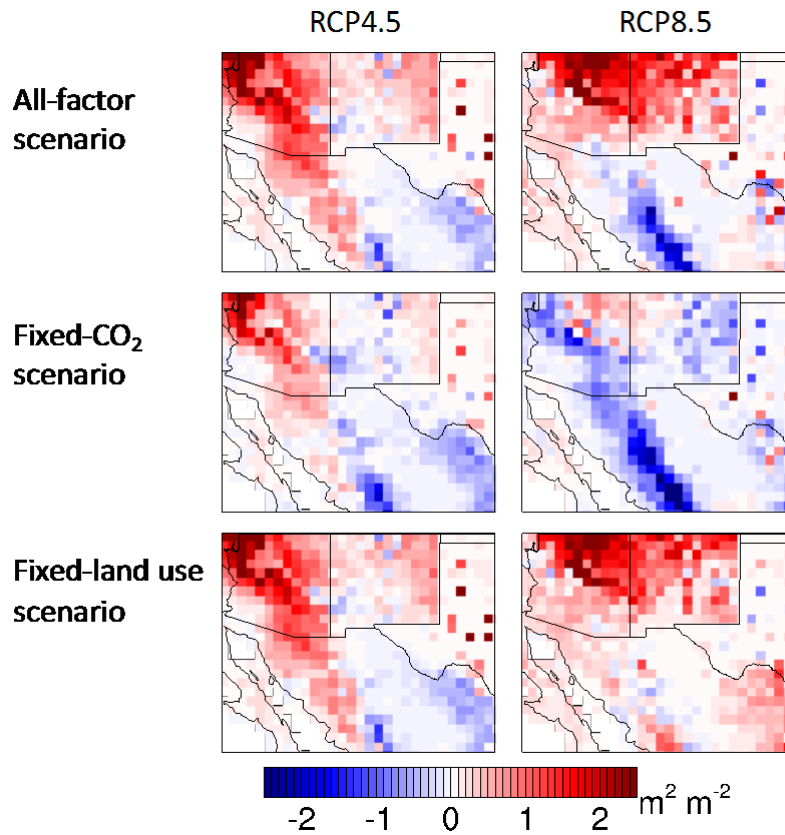
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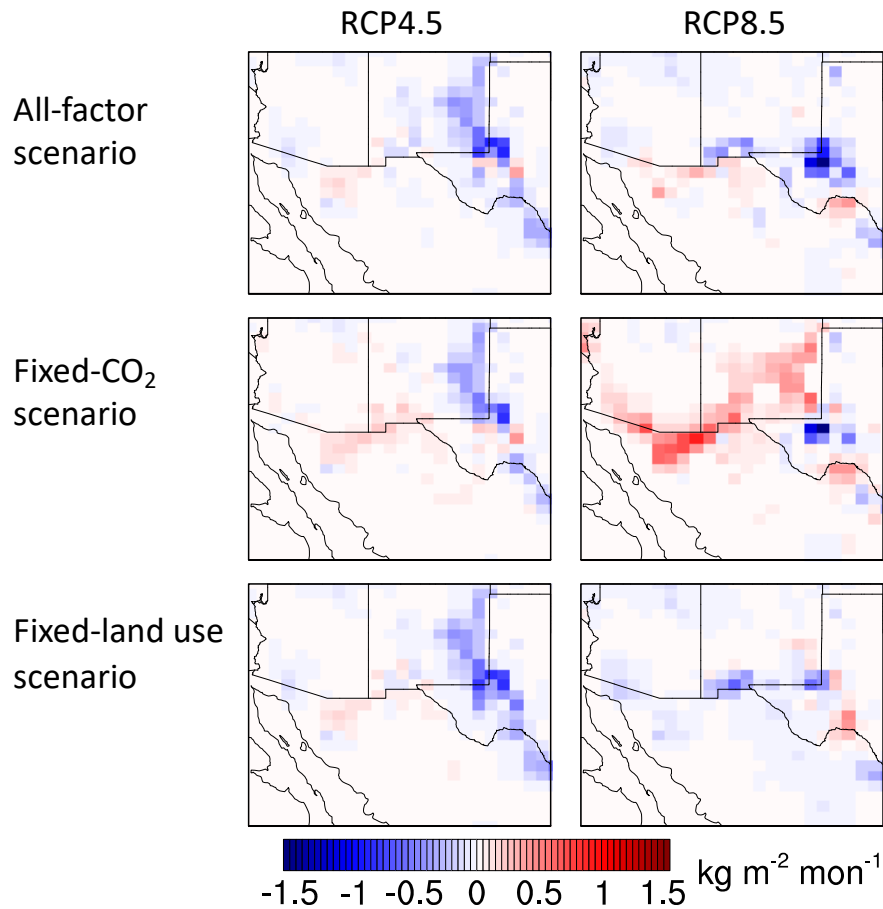
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686 model to the NCAR community climate model, *Journal of Climate*, 15, 1832-1854, 2002.



689

690 **Figure 1.** Simulated changes in spring averaged monthly mean vegetation area index (VAI) in
 691 southwestern North America under the three conditions for RCP4.5 and RCP8.5. Changes are
 692 between the present day and 2100, with five years representing each time period. The All-factor
 693 case (top row) includes the effects of climate, CO₂ fertilization, and land use on vegetation. Only
 694 climate and land use are considered in the Fixed-CO₂ case (middle), and only climate and CO₂
 695 fertilization are considered in the Fixed-land use case (bottom). Results are from LPJ-LMfire.

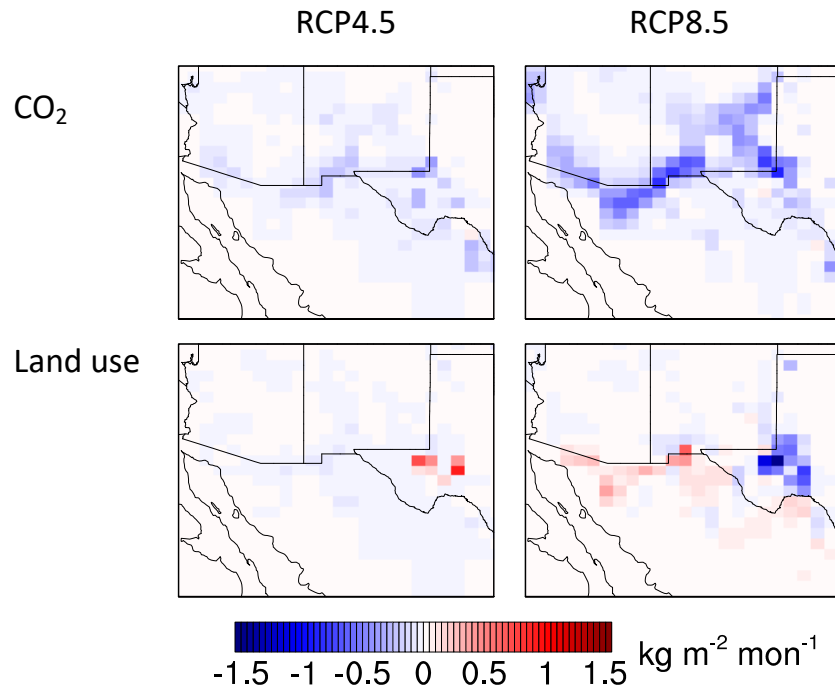
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697

698 **Figure 2.** Simulated changes in spring averaged monthly mean dust emission in southwestern
 699 North America under the three conditions for RCP4.5 and RCP8.5. Changes are between the
 700 present day and 2100, with five years representing each time period. The top row shows results for
 701 the all-factor condition, the middle row is for the fixed-CO₂ condition, and the bottom row is for
 702 the fixed-land use condition. Cases are as described in Figure 1. Results are generated offline using
 703 the GEOS-Chem emission component (HEMCO).

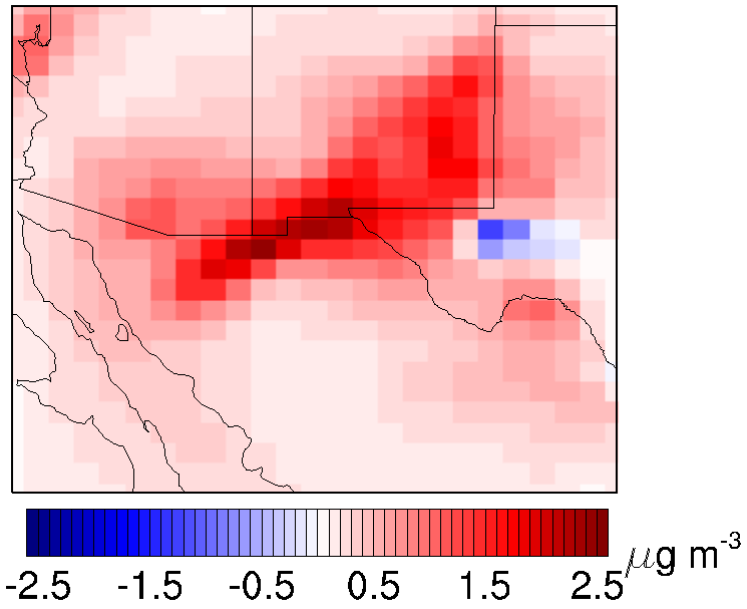
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705

706 **Figure 3.** Contributions of CO₂ fertilization and land use change to changing dust emissions in
 707 spring in southwestern North America for RCP4.5 and RCP8.5. Changes are between the present
 708 day and 2100, with five years representing each time period. The top row shows the response of
 709 dust emission to only CO₂ fertilization and the bottom row shows the response to only trends in
 710 land use. Results are generated offline using the GEOS-Chem emission component (HEMCO).

711



712

713 **Figure 4.** Simulated changes in springtime mean concentrations of fine dust over southwestern
714 North America for the RCP8.5 fixed-CO₂ case, in which the effects of CO₂ fertilization are
715 neglected. Changes are between the present day and 2100, with five years representing each time
716 period. Results are from GEOS-Chem simulations at 0.5° x 0.625° resolution.

717

718 **Table 1.** Averaged spring vegetation area index (VAI) and fine dust emission in southwestern
719 North America for the present-day and future for two scenarios (RCP4.5 and RCP8.5) and three
720 cases. The all-factor case includes changes in climate, land use, and CO₂ fertilization; the fixed-
721 CO₂ case includes changes in only climate and land use; and the fixed-land use case includes
722 changes in only climate and CO₂. The rows labeled “2100-2010, %” give the percentage changes
723 in VAI and fine dust emissions between the present day and future, with positive values denoting
724 increases in the future.

		VAI ^b , m ² m ⁻²			Fine dust emission ^b , kg m ⁻² mon ⁻¹		
		All-factor	Fixed CO ₂	Fixed land use	All-factor	Fixed CO ₂	Fixed land use
RCP4.5	2010^a	0.75±0.26	0.71±0.24	0.75±0.26	0.10±0.07	0.11±0.08	0.10±0.07
	2100^a	1.07±0.48	0.79±0.34	1.07±0.48	0.08±0.04	0.10±0.05	0.08±0.04
2100-2010, %		42	12	42	-25	-4	-26
RCP8.5	2010^a	0.80±0.27	0.75±0.24	0.75±0.24	0.09±0.04	0.09±0.05	0.09±0.04
	2100^a	1.11±0.71	0.55±0.33	0.55±0.33	0.07±0.04	0.14±0.09	0.07±0.06
2100-2010, %		38	-26	52	-20	58	-16

725 ^aEach time slice represents 5 years (i.e., 2011-2015 represents the 2010 time slice and 2095-2099 represents the 2100
726 time slice); ^bValues are spring (MAM) averages over southwestern North America.