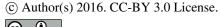
Atmospheric Chemistry and Physics



- Title:
- 2 Climate and demographic impacts on wildfire air pollution hazards
- 3 during the 21st century
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- 14 Abstract:
- 15 Wildfires pose a significant risk to human livelihoods and are a substantial health
- hazard due to emissions of toxic smoke. It is widely believed that climate change,
- through increasing the frequency of hot weather conditions, will also lead to an
- increase in wildfire activity. More recently, however, new research has shown that
- 19 trends in population growth and urbanisation can be as important for fire prediction as
- 20 changes in climate and atmospheric CO₂, and that under certain scenarios, fire activity
- 21 may continue to decline through most of the 21st century. The present study re-
- 22 examines these results from the perspective of air pollution risk, focusing on
- 23 emissions of airborne particulate matter (PM2.5). We combine an existing ensemble
- of simulations using a coupled fire-dynamic vegetation model with current
- observation-based estimates of wildfire emissions to predict future trends. Currently,

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26 wildfires PM2.5 emissions exceed those from anthropogenic sources in large parts of

the world, while emissions from deforestation or peat fires constitute minor sources.

We find that for Sub-Saharan Africa and southern China predictions of wildfire

29 pollution risks depend almost entirely on population dynamics, whereas for North

30 Australia and South America, it is mainly determined by climate change, with

31 Southeast Asia lying somewhere in-between. Under a scenario of current legislation

of anthropogenic emissions, global high population growth and slow urbanisation,

wildfires may seize to be the dominant source in large parts of Sub-Saharan Africa.

However, if anthropogenic emissions are strongly reduced, wildfires may both

become the dominant source and lie above critical levels for health impacts in large

36 parts of Australia, Africa, Latin America and Russia, and parts of southern China and

southern Europe. This implies that controlling anthropogenic emissions will not

suffice for attaining the World Health Organization air quality targets.

1 Introduction

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Wildfires are a major natural hazard (Bowman et al. 2009) and an important source of

air pollutants (Langmann 2009). Of these, emissions of fine aerosol particles, i.e.

42 particulate matter up to a size of 2.5 microns (PM2.5), are of particular health

concern, with no known safe levels of PM2.5 concentration in air, as noted by the

World Health Organization (WHO 2005). While globally, most PM2.5 emissions

45 come from human activities, wildfires can be an important source in large, more

remote areas (Granier et al. 2011, van der Werf et al. 2010). There is an expectation

that such emissions will become more important in the future (Kloster et al. 2010,

48 Knorr et al. 2016a), because of a widely held view among both the general public and

49 members of the research community that wildfire occurrence and severity have been

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increasing in recent decades, and will continue to increase due to climate change 50 (Doerr and Santin 2016) and efforts to reduce anthropogenic emissions (EEA 2014). 51 52 Climate warming has already led to frequent hot and dry weather around the globe, increasing the probability of wildfires (Flannigan et al. 2012), and this is 53 expected to continue into the future. Studies based on predicted fire severity indices 54 55 from climate argue for large increases in burned area as a result of climate warming (Flannigan et al. 2005, Amatulli et al. 2013). However, a long-term increase in the 56 length of the fire season or in weather conditions conducive of wildfires does not 57 58 necessarily lead to increases in burned area (Doerr and Santin 2016). This is because at longer time scales, vegetation also responds to climate change, as well as directly to 59 60 rising atmospheric CO₂ levels (Buitenwerf et al. 2012, Donohue et al. 2013). While CO₂ fertilization will lead to increased fuel load, enhancing emissions, it also leads to 61 an increase in woody as opposed to herbaceous vegetation, with lower emissions due 62 to decreased fire spread in shrublands (Kelley et al. 2014, Knorr et al. 2016b). Indeed, 63 simulations with coupled fire-vegetation or statistical climate-envelope models 64 generally show less increase in fire activity until 2100 when accounting not only for 65 climate, but also for these vegetation factors (Krwachuk et al. 2009, Kloster et al. 66 2010, Knorr et al. 2016c). 67 Another factor that has so far received less attention are changes in human 68 69 population density. Contrary to common perception, higher population density tends to be associated with lower burned area (Archibald et al. 2009, 2010, Lehsten et al. 70 2010, Knorr et al. 2014, Bistinas et al. 2014), even though more humans tend to lead 71 to more, but smaller fires (Archibald et al. 2009, 2010). This can be explained by the 72 concept of the ignition-saturated fire regime, which is reached at very low levels of 73 population density. Above this level, human impact is less manifested as enhancing 74

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burned area by providing ignitions, but more by creating barriers to and suppressing fire spread, thus reducing area burned (Guyette et al. 2002). Indeed, coupled 76 77 vegetation-fire models that include human effects suggest a reduced rate of increase of fire activity during the 21st century compared to simulations not accounting for 78 demographic changes (Kloster et al. 2010), or even a decline in burned area (Knorr et 79 80 al. 2016c) or emissions (Knorr et al. 2016b) for moderate levels of climate change combined with slow urbanisation and fast population growth. It was found that 81 differences between demographic scenarios can be more important than differences 82 83 between climate scenarios or climate models. There is also observational evidence for a long-term declining trend in past fire activity or emissions from wildfires (Marlon et 84 85 al. 2008, Wang et al. 2010, van der Werf et al. 2013), and more recent negative trends in Africa have been related to the expansion of cropland, that is itself a result of 86 increasing population density (Andela and van der Werf 2014). 87 The question is therefore not only how climate and vegetation change in the 88 future will impact on wildfire hazards, but also what the role of total population 89 growth and changes in spatial population distribution is for those predictions. 90 Following a similar study for Europe (Knorr et al. 2016a), we will use PM2.5 91 emissions from wildfires as an example fire hazard to illustrate the relative effects of 92 climate, vegetation and demographics, and base our projections on observation-based 93 wildfire emissions, using vegetation-fire model simulations to project relative 94 changes. The results are meant to be indicative of the importance of demographic and 95 climatic changes for the expected future development of wildfire hazards. All this, 96 however, needs to be seen against a background of considerable uncertainties 97 surrounding future projections of wildfire emissions (Knorr et al. 2016a, Doerr and 98 99 Santin 2016).

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2 Methods

101 2.1 Models and driving data We use the LPJ-GUESS global dynamic vegetation model (Smith et al. 2001, 102 Ahlström et al. 2012) coupled to the global semi-empirical fire model SIMFIRE 103 (Knorr et al. 2014). A detailed description of the coupling between SIMFIRE and 104 LPJ-GUESS and of methods used to compute wildfire emissions in terms of biomass 105 106 can be found in Knorr et al. (2016b). LPJ-GUESS is a patch-scale dynamic vegetation 107 model that represents age cohorts and computes vegetation establishment and growth, 108 allocation of carbon pools in living plants, and turnover of carbon in plant litter and soils. SIMFIRE provides burned area to LPJ-GUESS on an annual basis, which then 109 evokes plant mortality according to a plant-functional-type (PFT) dependent 110 probability. Specified fractions of plant litter and live leaf biomass are burnt and 111 emitted into the air in a fire, while the remaining biomass of the killed vegetation is 112 transferred to the litter pool (see Knorr et al. 2012). Population data needed to drive 113 SIMFIRE are based on gridded data from HYDE 3.1 (Klein-Goldewijk et al. 2010) up 114 to 2005, and then re-scaled using per-country relative growth in population and 115 urbanisation rates, retaining the urban masks of the HYDE data. Grid cells with more 116 than 50% past or future cropland area (in either the RCP6.0 or 4.5 land use scenarios 117 of Hurtt et al. 2011) were also excluded (see Knorr et al. 2016b, c for details). 118 In order to compute emissions of chemical species, we use the emission factors of 119 the Global Fire Emissions Database version 4 (GFED 4, Van der Werf et al. 2010, 120 based mainly on Akagi et al. 2011, see http://www.falw.vu/~gwerf/GFED/GFED4), 121 which are fixed ratios between emission rates of various pollutant species and rates of 122 combustion of dry biomass differentiated between fires in (1) savannas and 123 grasslands, (2) tropical, (3) boreal and (4) temperate forests. We assign a grid cell to 124

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development) is a grass, to (2) if it is a tropical, to (3) a boreal and to (4) a temperate 126 127 woody plant (see Knorr et al. 2012 for list of PFTs). 2.2 Simulations and scenarios 128 Simulations are driven by output from an ensemble of eight global climate models 129 from the Climate Model Intercomparison Project 5 (CMIP5, Taylor et al. 2012) for 130 two RCP (Representative Concentration Pathway, van Vuuren et al. 2011) climate 131 scenarios: 4.5 (moderate) and 8.5 (high degree of climate change). Simulations for 132 1901 to 2100 are carried out on a global equal-area grid with one by one degree 133 spatial resolution at the equator, but constant east-west spacing of the grid cells when 134 moving towards the poles in order to keep the grid cell area constant (Knorr et al. 135 136 2016b). Population projections follow the Shared Socioeconomic Pathways (SSPs, Jiang 137 2014). The SSPs are based on qualitative narratives following five different 138 development pathways which have been translated into quantitative projections of a 139 range of socio-economic and biophysical factors. Globally, SSP2 reflects an 140 intermediate case (medium population and economic growth and a central 141 142 urbanisation case), SSP3 high population growth and slow urbanisation with slow economic development, and SSP5 rapid but fossil-fuel driven economic growth with 143 slow population growth and fast urbanisation. However, there are regional variations 144 in demographic trends under each SSP. In contrast to developing countries and the 145 world as a whole, high-income countries have low population growth for SSP3 but 146 high population growth for SSP5. We did not consider the SSP1 scenario because its 147 148 sustainability assumptions lead to low emissions and the scenario is therefore not 149 compatible with the RCP8.5 climate scenario, nor did we use SSP4, since it is similar

(1) if the dominant PFT (the one with the largest leaf area index at full leaf

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to SSP2 in its population projections. The matrix of three SSPs and two RCP 150 scenarios represents a wide range of future climate and socio-economic conditions. 151 In addition to the emission fields simulated by LPJ-GUESS-SIMFIRE, we also use 152 the GFED4.1s observation-based emissions fields for wildfires (van der Werf et al. 153 2010, updated using Randerson et al. 2012 and Giglio et al. 2013) aggregated to 0.5 154 by 0.5 degrees resolution and then re-scaled in time by country or groups of countries 155 in some cases (following the methodology of Knorr et al. 2016a). For larger countries, 156 scaling is done by sub-national regions, which were chosen in such a way as to isolate 157 158 major fire areas found in GFED4.1s. For a list of regions/countries, see Table A1 in the Appendix. The use of countries accounts for the high degree of policy or cultural 159 impact on fire regime (Bowman et al. 2011). In order to account for demographic 160 effects at the grid-cell scale, we combine a scalar accounting for climate and 161 vegetation effects, f_{cv} , which is uniform in space across each region, with a scalar 162 accounting for demographic effects, f_p , which is applied at each grid cell separately: 163 164 $E(x,t) = f_{cv}(R(x),t) * f_p(p(x,t)) * E_{GFED}(x)$ (1) with E the re-scaled emissions, x the geographic location on the 0.5 by 0.5 degree grid 165 used for the analysis, t time, R the region/country found at location x, E_{GFED} the 166 167 annual emissions climatology from GFED 4.1s (average for 1997 to 2014). The 168 population effect, f_p , is equal to the population multiplier of SIMFIRE (Knorr et al. 2016b): 169 $f_p(p) = \exp(-0.0168*p'(p)).$ 170 (2) p' here is the minimum of population density p and 100 inhabitants per km², i.e. the 171 function is constant for values of p above 100 inhabitants per km² (Fig. A1). We have 172 173 introduced this cap, which is only used for scaling observation-based inventories by

LPJ-GUESS-SIMFIRE output but not by SIMFIRE itself, in order to prevent large

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relative increases in emissions during the scaling procedure, when population density
decreases from high values. We thus consider areas with higher population density
than 100 per km² to be essentially wildfire free. The combined climate and vegetation
effect is defined as

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$$f_{cv}(R,t) = \sum_{R} E(x',t) / \left[\sum_{R} E(x',t_0) * \sum_{R} f_{p}(p(x',t)) \right].$$
 (3)

Here, the sums are over all grid cells x' of the LPJ-GUESS 1-degree equal-area grid that belong to region/country R. For countries where 90% or more of the grid cells of the LPJ-GUESS grid have been excluded because they have a current or future cropland fractions of 50% or higher (highly agricultural regions: Moldavia and Bangladesh), or for which LPJ-GUESS simulates zero current emissions in at least one simulation (Greenland), we set $f_{cv}(R,t)$ =1. Gridded population data is based on HYDE 3.1 (Klein-Goldewijk et al. 2010), and future population patterns are re-scaled from 2005 population data using per-country population increases and changes in urbanization level, retaining the urban masks of the HYDE data (see Knorr et al. 2016c for details).

2.3 Analytical Framework

In our analysis, we focus on four time windows: current, 2030, 2050 and 2090. For current, we use 2010 population fields and annual anthropogenic emission data, as well as the mean annual emissions of GFED4.1s, which span the period 1997-2014. For the future time windows, we again use population fields and anthropogenic emissions from that year, but average emissions simulated by LPJ-GUESS-SIMFIRE spanning a period of 21 years centered on each of these years (i.e. 2020 to 2040 for the 2030 time window, etc.). While LPJ-GUESS-SIMFIRE simulations are carried out on a 1-degree equal-area grid, all spatially explicit analysis is carried out on a global 0.5 by 0.5 degree grid.

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To assess the relevance of PM2.5 emission rates, we both compare them to those 200 from anthropogenic sources, and consider an approximate threshold above which they 201 202 can be considered relevant for human health and air quality policy. The World Health Organization has adopted an air quality policy target of 10 µg/m³ on an annual 203 average, pointing out that there is no established safe upper limit and that the target 204 was set considering background concentrations of 3–5 µg m⁻³ in North America and 205 Western Europe. We follow here Knorr et al. (2016a) and assume a typical boundary 206 207 height of 1000 m and a life time of 1/50 years (about 7 days). Pollutants from 208 wildfires are inject into the atmosphere from large plumes, which have a global average height of around 1400 m, but only about 4-5% of wildfire emissions are 209 210 emitted into the free troposphere, the rest into the boundary layer (Veira et al. 2015). Here, we assume that after about one week, most of PM2.5 is either deposited or 211 effectively mixed into the free troposphere. We also neglect horizontal transport 212 between 0.5 by 0.5 degree grid cells and compute an annual budget based on annual 213 mean emissions and pollutant life time. Using these idealized conditions, which are 214 meant as a first guidance, we arrive at a threshold of 0.5 g m⁻² yr⁻¹ for PM2.5 215 emissions corresponding to a mean annual concentration of 10 µg m⁻³. In this 216 analysis, we use 0.2, 0.5 and 1 g m⁻² yr⁻¹ as alternative thresholds spanning a critical 217 range for health and air-quality policy purposes. 218 219 For anthropogenic emissions, we use the GAINS 4a data (Amann et al., 2011) developed as part of the ECLIPSE project (Granier et al. 2011, Klimont et al. 2013, 220 Stohl et al., 2015) for the years 2010 (for current conditions), 2030 and 2050. There 221 are two future scenarios: current legislation (CLE), and maximum feasible reductions 222 (MFR). MFR corresponds to a policy driven abatement scenario with the aim, among 223 others, to lower PM2.5 emissions to a level to minimize health impacts (Amann et al. 224

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constant per capita emissions after 2050 but changing population according to the 226 227 SSP3 scenario. For MFR, 2090 emissions are assumed half of the corresponding 2050 levels. As the CLE and MFR scenarios do not account for CO2 emissions, both are in 228 229 principal compatible with the CO₂ equivalent greenhouse gas emissions of both RCP 230 scenarios, even though the non-CO₂ greenhouse gas emissions may differ. 231 For a regional analysis, we use a global map of nine major world regions to facilitate a global-scale analysis of our results (see Fig. A2). Of these, three belong to 232 233 the high-income group of countries of the SSP scenarios (see Jiang and O'Neill 2015): High-income Europe, Australia & New Zealand, and North America. Countries of 234 235 Europe belonging to the middle-incoming group were assigned Eastern Europe and Central Asia, which also includes Russia. Countries of the Middle East (Israel, oil-236 rich states of the Persian Golf) or East Asia (Japan, South Korea) belonging to the 237 high-income group were excluded, which only account for a very small fraction of 238 wildfire emissions in their respective group. 239 3 Results 240 241 3.1 Current patterns of wildfire pollutant emissions The analysis presented in this sub-section concerns exclusively observation-based 242 emission inventories. Currently 14 million km² of land area are affected by wildfire 243 PM2.5 emissions that exceed 0.5 g m⁻² yr⁻¹, used as an indicative threshold for serious 244 245 health impacts, mainly in Sub-Saharan Africa, North America, South Australia, Southeast Asia, and the boreal zone (numbers are for the 0.5 by 0.5 degree grid: 23 246 million km² for a threshold of 0.2 and 8 million km² for a threshold of 1 g m⁻² yr⁻¹) 247

2011). Following Knorr et al. (2016a), we estimate 2090 CLE emissions assuming

According to the GFED4.1s and ECLIPSE inventories (Fig. 1), PM2.5 emissions over

large parts of the globe are dominated by wildfires, in particular the boreal zone and

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the semi-arid tropics. Even in the humid tropics, such as the Amazon basin or 250 Southeast Asia, wildfires are prevalent and deforestation fires still play a 251 252 comparatively minor role. Only Indonesia is clearly dominated by deforestation and peat fire emissions. Of the nine world regions, four (Sub-Saharan Africa, Latin 253 America & Caribbean, Eastern Europe-Russia-Central Asia and Australia & New 254 255 Zealand) have higher wildfire than anthropogenic emissions of PM2.5 on an annual basis (Table 1). 256 There are large, remote areas where, despite low emission rates, wildfires are the 257 258 highest emissions source because anthropogenic emissions are even lower. This applies for example to large parts of the boreal zone, central parts of Australia, much 259 260 of the western US, or the northern part of the Amazon. However, large regions also have emissions above the upper critical range from 0.5 to 1 g m⁻² yr⁻¹, mainly in the 261 boreal-forest areas (Alaska, Canada, Russia), the semi-arid tropics (the African 262 savannas, the areas south of the Amazon basin, Southeast Asia from Myanmar to 263 Cambodia and Northern Australia), and southeastern Australia in the temperate zone. 264 Other pollutants show a similar pattern of dominance (Fig. 2), but with some 265 important differences: for CO and NO_x, anthropogenic sources are more important, 266 and only Sub-Saharan Africa and Australia & New Zealand have emissions of these 267 gases from wildfires close to or surpassing those from anthropogenic sources (Table 268 269 1). For NO_x, deforestation fires in the Amazon are also of minor importance compared to wildfires (Fig. 2b). For BC, Sub-Saharan Africa has very high but 270 Australia & New Zealand very low anthropogenic sources, which is reflected in the 271 dominance pattern shown in Fig. 2c. BC from wildfires is also important in temperate 272 North America and central Asia. 273

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A breakdown of PM2.5 emission patterns by population density is shown in Fig. 274 3. It shows current emissions per area averaged over all grid cells of a given region 275 276 that fall into a certain population-density range. Fig. 3 reveals the expected trend of increasing anthropogenic emissions where more people live. By contrast, for Sub-277 Saharan Africa, Latin America & Caribbean, Eastern Europe-Russia-Central Asia and 278 279 South & Southeast Asia, wildfires show peak values with maximum emissions in regions of intermediate population density: Sub-Saharan Africa and Latin America & 280 Caribbean at 1 to 10, Eastern Europe-Russia-Central Asia at 0.1 to 1, and South & 281 Southeast Asia at 10 to 100 people km⁻². Deforestation fires are of minor importance 282 for air pollutant emissions, except for Latin America & Caribbean, where they occur 283 284 mainly in sparsely populated area, and for South & Southeast Asia, where they are as important as wildfires and most significant in areas of high population density. It is 285 important to note that within South & Southeast Asia (Fig. A2), wildfires occur 286 mainly in South-East Asia proper, but deforestation fires in Indonesia (Fig. 1a). 287 Indonesia is also the only region where emissions from peat fires are relatively 288 important for air pollution. In High-Income Europe and Developing Middle East & 289 North Africa, wildfires show a similar increase with population density as 290 anthropogenic emissions with the consequence that they are much smaller than 291 anthropogenic emissions in all areas (for Developing Middle East & North Africa 292 293 their magnitude is also very low *per se*). For North America, wildfires have the reverse trend compared to anthropogenic emissions, and for Australia & New Zealand 294 and Developing East Asia, wildfire emissions happen at a similar average rate across 295 all population density categories. These different trends between wildfires and 296 anthropogenic sources lead to a situation where the former become the dominant 297 298 source below a certain value of population density, which is 10 for Australia & New

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also has the highest per-area anthropogenic emissions, which makes wildfire 300 301 emissions appear to be generally of minor importance compared to current anthropogenic-emission levels. 302 3.2 Simulated changes in emissions 303 In this sub-section we present results from simulation with LPJ-GUESS-SIMFIRE. 304 These differ from those presented in the following sub-sections, where the relative 305 temporal changes from the LPJ-GUESS-SIMFIRE simulations are used to scale 306 307 current observed emissions (Equ. 1). The ensemble mean PM2.5 emission shows a continuous declining trend for Sub-Saharan Africa across both the 20th and the 21st 308 centuries for the moderate climate change scenario (RCP4.5, Fig. 4). Only the two 309 ensembles with fast urbanisation show a slight increasing trend starting after 2050 for 310 the medium population projection (SSP2 with fast urbanisation), or around 2030 for 311 the low population growth scenario (SSP5). The range of predictions across climate 312 313 models is about as large as the range of predictions across demographic scenarios. The result is surprisingly similar for the high climate change scenario (RCP8.5), 314 where the central SSP2 demographic scenario still shows no clear increase in 315 emissions even towards the end of the 21st century (Fig. 5). For Sub-Saharan Africa, 316 317 demographic trends are by far the dominant driver of changes in fire regime, while differences between climate scenarios are minor. For the two scenarios with fast 318 319 urbanisation (with either SSP2 or SSP5 population trends), this region shows a continuing decline in PM2.5 emissions from wildfires. The effect of changing only 320 321 the urbanisation scenario on emissions is approximately half compared to changing both the urbanisation and population scenarios (i.e. changing SSP2 to SSP2 with fast 322 urbanisation vs. changing SSP2 to SSP5; or changing SSP2 to SSP2 with slow 323

Zealand, and 1 inhabitant per km² for the other world regions. Developing East Asia

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urbanisation vs. changing SSP2 to SSP3). This region clearly stands out not only 324 because it has by far the largest share of global wildfire pollutant emissions (Table 1), 325 but also because of its large decline in fire activity driven by population trends. 326 The other two tropical regions, Latin America & Caribbean and South & 327 Southeast Asia, differ from Sub-Saharan Africa in that the simulated historical decline 328 is less steep, and that the future scenarios show an upward trend that is only slight for 329 RCP4.5, but steep for RCP8.5. The range of predictions for only the central SSP2 330 scenario (ensemble ranges for the medium population and central urbanisation 331 332 scenario, dark grey area in Fig. 4 and 5) is almost as large as that for the entire ensemble, indicating a reduced role for demographic change, with climate change as 333 334 the main driver. For Latin America & Caribbean, the effect of urbanisation is negligible, whereas urbanisation plays an important role for South & Southeast Asia. 335 However, in the case of moderate climate change (RCP4.5) combined with SSP3 high 336 population growth and slow urbanisation, even in these regions wildfire activity may 337 not increase during the 21st century. The arid Developing Middle East & North Africa 338 region has a similar declining trend, with a reversal around the middle of the 21st 339 century that is strongly dependent of climate scenario. 340 Two northern regions that belong to the middle-income group and therefore have 341 lower population growth with SSP5 than with SSP3 (as have the tropical regions 342 343 discussed so far) are Eastern Europe-Russia-Central Asia and Developing East Asia. For both, LPJ-GUESS-SIMFIRE simulates no trend during the historical period, 344 except for an increasing upward trend beginning late in the 20th century. The climate 345 scenario has a strong impact on both regions' predictions, and demographics a small, 346 albeit still discernable one. In both regions, urbanisation plays an important role as 347

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349 urbanisation. 350 North America and High-Income Europe have a similar temporal profile as 351 Eastern Europe-Russia-Central Asia, but with the order of demographic scenarios reversed. For North America, demographics are predicted to play only a minor role 352 353 compared to climate change, but for High-Income Europe, there is still a marked influence even though changing the demographic scenario does not change the 354 general trend in predictions. Here, differences between urbanisation scenarios are 355 356 unimportant, but SSP5, which here has the highest population growth, leads to markedly higher predictions compared to SSP3 with low population growth. Another 357 358 region that belongs to the high-income group but where most wildfire emissions are from the tropics (Australia & New Zealand, cf. Fig. 1) stands out as showing almost 359 no change in fire activity across both centuries. Only for RCP8.5 there is a very slight 360 increase, and difference between demographic scenarios have almost no impact on the 361 results. 362 We also note that simulations with LPJ-GUESS-SIMFIRE sometimes differ 363 substantially from GFED4.1s (Fig. 4, 5, Table 1), and show in particular higher 364 emissions in the boreal zone (e.g. higher for Eastern Europe-Russia-Central Asia than 365 for Latin America & Caribbean, and also high emissions for North America). We 366 367 attribute this to differences in the assumed litter load and combustion completeness between GFED (van der Werf et al. 2010) and SIMFIRE (Knorr et al. 2012). These 368 quantities are generally not well constrained, as noted by Knorr et al. (2012). We 369 expect, however, that the relative change in emissions, which we compute by country 370 (Table A1), is much less affected by those differences. In Fig. 6, we show this relative 371 change from current conditions to 2090 by country/region for the two scenarios that 372

seen in the difference between SSP2 with central and SSP2 with slow/fast

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lead to the lowest (SSP3 with RCP4.5) or highest (SSP5 with RCP8.5) end-century global emissions. 374 375 For SSP3/RCP4.5, there is again a very pronounced decline in fire activity in most but not all countries of Africa. South Africa, Namibia and in particular 376 Botswana show a relative increase in emissions of up to 50%. The strongest decline is 377 378 found in parts of West Africa, in particular Nigeria, where wildfires in this scenario 379 almost completely disappear. By contrast, central Europe, which has a pronounced population decline under SSP3, shows a strong increase, albeit from a low base (Fig. 380 381 1). An arch of countries spanning from Turkey to Southeast Asia also show either a strong decline or only a small increase in fire activity. Western boreal North America 382 383 (ALK, CAN-W, Table A1) and eastern boreal Russia (RUS-NW) show moderate to strong increases, which are much higher for SSP5/RCP8.5, driven mainly by climate 384 change given the low population density in these regions. A very large increase with 385 SSP5/RCP8.5 by around 150% is found for eastern and southern China (CHN-E), 386 driven by reduced population size, fast urbanization, and climate change. A 387 pronounced increase is found mainly for the northern part of the Amazon basin 388 389 (BRA-N). For Australia, we find a decline in the north (AUS-N) for both climate scenarios, but a slight to pronounced increase for the remaining areas (demographics 390 391 in this region play almost no role). 392 3.3 Predicted changes in emissions by population density 393 The strongest change in the distribution of wildfire emissions against population density (cf. Fig. A3) is found for Sub-Saharan Africa under the SSP3 and RCP4.5 394 scenarios, where the 0.1 to 1 and 1 to 10 people per km² categories see a decline by 395 around a factor of 10 between 2010 and 2090 (Fig. 7). As the area extent of these 396 categories hardly changes (dotted lines) and the decline is absent for SSP5/RCP8.5, 397

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the decrease is mainly because population density within a given category increases, 398 which leads to more fire suppression. (Figs. 1, A2). Woody encroachment, which also 399 leads to a decline in fire activity, would respond strongly to the higher CO₂ levels of 400 RCP8.5 (Buitenwerf et al. 2012, Knorr et al. 2016b, c). Conversely, areas with more 401 than 100 people per km² see an increase in both extent and emissions per area, as 402 403 more people move into fire-prone area in this slow-urbanisation scenario. For the CLE anthropogenic-emissions scenario, we expect most changes in the relative 404 dominance of wildfire vs. anthropogenic emissions to happen in the 10 to 100 people 405 per km² category. 406 This pattern of increasing emissions in the most densely populated areas is seen 407 408 for all middle to lower-income regions (all but Australia & New Zealand, North America and High-Income Europe). Of these, Latin America & Caribbean, Eastern 409 Europe-Russia-Central Asia and Developing Middle East & North Africa show 410 relatively small changes in emissions, while for South & Southeast Asia and 411 Developing East Asia, a decline in emissions in sparsely populated regions is 412 accompanied by a similar decline in anthropogenic emissions, so that no significant 413 changes in the relative importance of the two emission sources are expected for this 414 particular scenario. For High-Income Europe, wildfire emissions are projected to 415 remain well below anthropogenic emissions in all categories, while for Australia & 416 417 New Zealand, a continuing decrease in emissions in the most densely populated category will make wildfire emissions increasingly relevant in such areas. For North 418 America simulated changes in wildfire are also minor and wildfires will continue to 419 be the dominant source mainly in remote areas. 420 The situation of relative importance changes drastically if we consider the MFR 421 422 anthropogenic scenario (Fig. 8). According to this scenario combined with RCP8.5

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climate and SSP5 demographic change (rapid urbanisation, low population growth in low to middle income countries), wildfires could become the dominant emission 424 425 source in Sub-Saharan Africa and Australia & New Zealand in all population-density categories as early as 2030, and be at least of comparable magnitude as anthropogenic 426 emissions for Latin America & Caribbean and to a lesser extent South & Southeast 427 Asia and Eastern Europe-Russia-Central Asia. High-Income Europe and Developing 428 Middle East & North Africa, who both have the same increasing relationship between 429 emissions and population density for both sources, wildfires will continue to be minor 430 431 in all categories despite strong reductions in anthropogenic emissions. For Developing East Asia, there is an approximately fourfold increase predicted for wildfire emissions 432 in the 10 to 100 inhabitants per km² category, with the result that they might become 433 comparable to anthropogenic emissions in areas that comprise a rather large 434 population. 435 3.4 Future patterns of pollutant exposure 436 437 The previous analysis only compared wildfire and anthropogenic emissions, but in some areas, both might be so low that they do not constitute a relevant health hazard. 438 A further analysis therefore considers if wildfire emissions exceed a threshold of 0.5 g 439 m⁻² yr⁻¹ (Fig. 9; see Fig. A4 and A5 for thresholds of 0.2 and 1 g m⁻² yr⁻¹). Large areas 440 441 in the boreal zone, South America, Central Asia and Australia where wildfires dominate do not reach this level. However, the analysis also reveals the demographic 442 443 scenario as the main driver of change for Africa. For SSP3, with high population growth and slow urbanisation, many areas drop below this threshold in the future, 444 445 independent of climate scenario, but not for SSP5 (low population growth and fast urbanisation). For northern Australia, the result is independent of demographic 446 scenario, while RCP4.5 sees a small contraction of high-emission areas, but RCP8.5 a 447

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much larger one with an additional zone with high emissions emerging further south. 448 Changes in the boreal zone and South America are slight but new high-emissions 449 450 areas appear for RCP8.5 (light and dark blue). Larger new high-emission areas are found in southern China, mainly driven by demographic and to a lesser extent climate 451 change. The same result is found for Portugal, but with the opposite demographic 452 scenario as this is a high-income region where SSP5 has low population growth and 453 leads to the extension of high-emission areas. Other temperate areas in North America 454 and Australia see little change in any of the scenarios. 455 456 If we consider the number of people living in areas exceeding a certain wildfireemissions threshold, we find an almost universal increase in both the absolute 457 458 numbers, and the percentage of global population independent of climate and demographic scenario (Table 2). Currently, between 7.9 and 1.8% of world 459 population is affected, depending on where the threshold is set (3.6% for the 0.5 g m⁻² 460 yr⁻¹ threshold used with Fig. 10). Only for the 1 g m⁻² yr⁻¹ threshold and SSP5/RCP4.5 461 scenario, this percentage will very slightly decrease to 1.7% by 2030, but the absolute 462 number still increase from 126 to 146 Million people affected by dangerously high 463 levels of wildfire air pollution. For all other scenarios and for SSP5/RCP4.5 from 464 2050 there will be an increase both in the absolute and relative numbers of affected 465 population. The demographic scenario is also more important than the degree of 466 climate change. For the 0.5 g m⁻² yr⁻¹ threshold, changing the RCP scenario changes 467 the percentage by between 0.1 for 2030 and 0.4 for 2090, but changing from SSP5 468 (low growth, fast urbanisation) to SSP3 (high growth, slow urbanisation) changes the 469 percentage of affected population by between 0.4 for 2030 and 1.2 for 2090. This 470 difference is even more pronounced for the 0.2 g m⁻² yr⁻¹ threshold, while for the 1 g 471

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m⁻² yr⁻¹ threshold, both climate and demographic change have about the same importance. 473 Absolute numbers of people affected by the intermediate 0.5 g m⁻² yr⁻¹ threshold 474 can reach as high as 973 Million for SSP3 with high population growth and slow 475 urbanisation combined with RCP8.5 climate change, from currently 256 Million, i.e. 476 almost four times the current estimate. Finally, the spatial exceedance patterns are 477 predicted to change very little for 0.2 g m⁻² yr⁻¹ (Fig. A4), but are similar between the 478 0.2 and 0.5 g m⁻² yr⁻¹ cases (Fig. A5). This stands in contrast to the percentage of 479 480 affected population (Table 2), and can be explained by a change in the geographic location of areas that are either added or subtracted (cf. China in Fig. A4 vs. Africa in 481 482 Fig. A5). Using the idealized conditions that have led us to the 0.5 g m⁻² yr⁻¹ threshold, i.e. 483 1000 m boundary layer height and 1/50 yr life time of PM2.5, we can calculate a 484 mean annual concentration value for every grid cell and from there a population-485 weighted average for each of the nine world regions. For six of those regions, the 486 resulting concentration is below 1 µg m⁻³ for all years and scenarios and therefore not 487 considered relevant. For the other three, results are shown in Table 3. For Latin 488 America & Caribbean, we only find a small increase, mainly for RCP8.5. For 489 Australia & New Zealand, the increase is more substantial and much higher for 490 491 RCP8.5, and the results for both regions are almost the same across all demographic scenarios. Both findings are in accordance with Fig. 10. For Sub-Saharan Africa, 492 however, we find a universal decrease in exposure estimates from currently 7.7 to as 493 low as 5.3 µg m⁻² by 2090 for SSP3 with high population growth, slow urbanisation 494 and RCP8.5. Differences between RCP scenarios amount to about 0.1 µg m⁻², but 495 496 between demographic scenarios to between 0.3 for RCP4.5 by 2030, and 1.3 for

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affected by the 0.5 g m⁻² yr⁻¹ threshold, which we assume is dominated by Sub-498 499 Saharan Africa (Table 1). Here, the idealized exposure in Sub-Saharan Africa is 500 higher for the fast-urbanisation/low population growth scenario (SSP5) than for those with slow-urbanisation/high population growth (SSP3), but the reverse is true for both 501 502 relative and absolute population in high-emission areas. For both measure of human 503 risk from PM2.5 exposure (Tables 1 and 2), changing only the urbanisation scenarios (with the same SSP2 population scenario) has a markedly smaller impact than 504 505 changing only the population scenario (and keeping either fast or slow urbanisation). 4 Discussion 506 An important question is whether past climate change has already led to increases in 507 508 wildfire activity and related pollutant emissions. The many uncertainties associated with modelling wildfire emissions are discussed in detail by Knorr et al. (2012, 509 2016a). This study simulates an increase from around 1980 to 1990 for Russia and 510 North America, which seems to agree with the observation of a climate-driven 511 increase in fire activity in the western U.S. based on data from 1982 to 2012 512 (Westerling, 2016). However, our simulated relative increase is only very slight for 513 the western U.S. region (Fig. A6), and only reaches about 20% by 2090 for RCP8.5 514 (Fig. 6). Doerr and Santin (2016) argue that this increase may be regional and highly 515 policy dependent. Ironically, there is the possibility that the increase has been driven 516 by increased fire suppression, which has led to fewer but more intense fires and more 517 area burned. While past climate-driven increases in fire activity remain debatable, this 518 study shows a general picture of climate-driven increases that may be overridden by 519 demographic changes only in Sub-Saharan Africa. 520

RCP4.5 by 2090. There is also a reversal of the order compared to the population

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A particularly large relative increase in wildfire emissions is projected for China 521 and the boreal areas of North America and Russia for the RCP8.5 climate scenario. 522 523 Southern China is also identified as a new area of high risk from wildfire air pollution. While forest fires in China may have received less attention, they can still be 524 substantial, with 670,000 ha area burnt annually between 1950 and 1999 (Shu et al. 525 2003). 526 The present study broadly re-confirms the results of a previous analysis that was 527 based on burned area as a measure of wildfire risk (Knorr et al. 2016c). Sub-Saharan 528 529 Africa, currently by far the most fire-prone region, is projected to see a demographically driven decline in fire activity, except for a scenario of low 530 531 population growth and rapid urbanisation. This decline is in agreement with observations of declining burned area linked to demographic trends for northern part 532 of Sub-Saharan Africa (Andela and van der Werf, 2014). At the same time, wildfire 533 risk to humans will broadly increase for almost all scenarios considered. For 534 developing countries, differences between the high and low population growth 535 scenarios tend to be more important for the projected changes in fire hazard than 536 differences between the two climate scenarios. For the developed world and northern 537 wildfire regions, climate and vegetation change appear to be the main drivers. 538 There are some noteworthy differences between the approaches. Fire risk in 539 540 Knorr et al. (2016c) is based on the probability of a point in space to be affected by wildfire, while the present study focuses on emissions-related hazards and uses PM2.5 541 emissions as an indicator, taking into account the amount of fuel burnt and its 542 efficiency at producing PM2.5. Both aspects are important for how human societies 543 are impacted by fire but represent different types of hazard, as the former relates 544 545 mainly to potential loss of property, whereas the latter affects human health. While

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the burned-area study uses output from LPJ-GUESS-SIMFIRE directly, the present 546 one uses observation-based estimated and re-scales them based on smaller regions 547 according to LPJ-GUESS-SIMFIRE projections. This approach has been used before 548 by Knorr et al. (2016a) and accounts for the importance of policy on wildfire 549 occurrence by scaling emissions mostly by country. The importance of policy is 550 551 evident for example when comparing observed burned area in Scandinavia to adjacent 552 areas of Russia (Giglio et al. 2013). There is also an underlying assumption in both studies that the impact of population density on area burnt within each country is 553 554 invariant over time. The present study also confirms the important role of PM2.5 emissions for wildfire air pollution risks, as other pollutants tend to have a larger 555 556 relative contribution from anthropogenic sources (Knorr et al. 2016a). LPJ-GUESS-SIMFIRE only simulates wildfires. The predictions presented in this 557 study therefore leave out the possibility of significant increases in deforestation or 558 559 peat fire sources. Therefore, peat and deforestation fires have been excluded from the predictive part of the present study. Peat fires can be associated with considerable 560 emissions (Page et al. 2002, Kajii et al. 2002), and forest conversion is often 561 accompanied by burning (van der Werf et al. 2010). The comparative analysis shown 562 here, however, shows that globally both are of minor importance except for Southeast 563 Asia. The south-east Asian deforestation and peat fires occur mainly in Indonesia 564 565 (Field et al. 2009), where they are the dominant pollution source and occur even in more densely populated areas. In other world regions, including Russia, peat fires are 566 of minor importance. 567 Whether or not future land-use change will lead to an increase or a decrease in 568 deforestation, is unknown. Based on four integrated-assessment model realisations of 569 570 the four RCPs, Hurtt et al. (2011) projected little increase, and if any, then in future

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crop and pasture areas. However, in studies that examined land-use change from a broader perspective, a much larger range of crop and pasture changes emerged (Eitelberg et al. 2015, Prestele et al. 2016), which makes the relative change of deforestation and wildfires highly uncertain. In the present analysis, declining wildfire emissions are only predicted for Sub-Saharan Africa, where it appears to be related to conversion of savanna to cropland (Andela and van der Werf 2014). Interestingly, increased fire activity is predicted for southern Africa for both climate scenarios, in accordance with the result of Andela and van der Werf, who found a recent increase for that region driven by declining precipitation. We therefore believe that the results of the present study are broadly representative of possible future changes in wildfire risk, even though one needs to take into account that in certain areas, deforestation may remain the main driver of air pollution for a while. Demographic trends will be an important and often the main factor driving changes in wildfire hazards. One factor is that higher population density in rural areas means lower burned area and emissions, but also more people exposed, and vice versa. In the analysis of burned area patterns by Knorr et al. (2016a), there was a large impact of urbanisation (using the same SSP2 per-country population scenario), with more people living in fire prone areas at slow than at fast urbanisation, but with a relatively minor affect due to overall population change. The average fractional surface area burned in densely populated areas was also higher for slower urbanisation. This is because the suppression of fire by higher population density was over-compensated by a higher number of people living in rural, fire-prone areas. In the present analysis, we find a much smaller impact of urbanisation on the number of people living in areas with high wildfire emissions, but a large impact of total population change. Even though more people tend to suppress fire, the percentage of

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people living in high-emission areas increases when the overall population is higher. 596 597 This is the opposite of what we find for the average pollutant concentration levels 598 experienced by the population: For Sub-Saharan Africa, fast urbanisation or high population growth lead to higher emissions in rural areas and overall higher exposure 599 to wildfire pollutants even though people move away from areas with high wildfire 600 601 activity. While any additional emission source of PM2.5 poses a health risk (WHO 2005), 602 in practice wildfires are likely to be ignored by air quality policy if they emit 603 604 considerably less than anthropogenic sources, in particular as their occurrence tends to be sporadic and of short-term nature. One factor is that wildfire emissions are much 605 606 more difficult to legislate given the sometimes unexpected results of fire suppression policies (Donovan and Brown 2007). However, we find that in large parts of the 607 world, wildfires are the main air pollutant source. While in many of those regions, 608 wildfires dominate by absence of large anthropogenic sources, Sub-Saharan Africa, 609 610 Brazil, northern Australia, Southeast Asia and the boreal zone are regions where they not only emit more PM2.5 than anthropogenic sources, but emissions are higher than 611 some approximate threshold of health relevance in the region of 0.5 to 1 g m⁻² yr⁻¹ 612 This implies that even controlling all anthropogenic sources, the WHO air quality 613 goals can not be attained. 614 615 It will therefore of critical importance whether future air quality policy objectives in the various regions will converge to the current WHO guidelines, in which case in 616 these regions fire management will become increasingly important. At current 617 legislation, wildfires will seize to be important for large parts of Africa and 618 considerable parts of South America. If, however, anthropogenic emissions are 619 aggressively curtailed (MFR scenario), wildfires in both regions are predicted to 620

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areas with wildfire emission levels relevant for air quality policy. Such reductions in 622 anthropogenic emissions would bring those down to levels similar to those of 623 wildfires even in the most densely populated areas, making wildfires the most 624 important pollution source in many regions (Sub-Saharan Africa, Latin America & 625 Caribbean, and to a lesser degree Australia & New Zealand, Eastern Europe-Russia-626 Central Asia). Because past efforts aimed at a lasting reduction in wildfire activity 627 have largely failed despite high costs (Doerr and Santin 2016), it is questionable 628 629 whether it is even possible to devise policy measure aimed at bringing down wildfire emissions to meet WHO guidelines. Because wildfires are an essential part of many 630 631 ecosystems (Bowman et al. 2009), it may therefore better to discount for wildfire emissions as a natural phenomenon and rather adapt urban and suburban planning 632 accordingly (Moritz et al. 2014). 633 This study has some important limitations. It does not consider atmospheric 634 transport or injection height (Gonzi et al. 2015, Sofiev et al. 2012), nor horizontal 635 advection of pollutants, and predictions are based on a single fire and vegetation 636 model. Demographic scenarios do not currently account for changes in the urban 637 mask. It only considers climatological annual emissions during specified time 638 windows, even though wildfires impacts on air quality can have large interannual 639 640 (Jaffe et al. 2008) and intra-seasonal variations, caused in part by long-range transport (Niemi et al. 2005), which is also not accounted for. It also does not account for 641 relevant secondary emission products, such as ozone from wildfires, which can reach 642 policy relevant levels (Jaffe and Wigder 2012). This contrasts with previous studies 643 on the possible impact of climate change on wildfire-related air pollution hazards 644 have concentrated on changes in meteorological conditions (Jacob and Winner 2008, 645

decline less than anthropogenic sources, and climate change will even lead to new

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Tai et al. 2010) instead of emissions. The study by Kaiser et al. (2012) focuses on 646 current conditions and includes atmospheric transport, using satellite-observed fire 647 radiative energy (Wooster et al. 2005) as well as satellite-derived aerosol optical 648 depth data to constrain wildfire emissions (Kaiser et al. 2012), as opposed to satellite-649 derived burned area as used by GFED (van der Werf et al. 2010, Giglio et al. 2013). 650 651 There is also recent progress in the incorporation of injection height (Sofiev et al. 2012) into chemistry-enabled atmospheric general circulation models (Veira et al. 652 653 2015). 654 By contrast, the present study focuses on changes in emissions, and is the first 655 global-scale study to consider changes both climate and demographic drivers of air 656 pollutant emissions from wildfires. Future work should aim at using general circulation models with realistic plume heights for a series of dedicated present and 657 future time slices at combining observed plume height information, fire radiative 658 energy data (for their finer temporal resolution), satellite-derived burned area (for 659 better spatial coverage), projected emission changes from coupled dynamic 660 vegetation-fire models (as the present study), and improved demographic scenarios 661 accounting for changes in urban population density. Such studies would then not only 662 account for long-range transport pollutants and secondary products such as ozone, but 663 also simulate the temporal statistics of pollution events on a daily time scale. Such 664 665 results could then be used, for example, to assess for how many days the WHO 24hour PM2.5 limit (WHO 2005) is a exceeded as a result of wildfire emissions. 666

5 Summary and conclusions

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 So far, there does not seem to be compelling evidence for a long-term trend towards increased pollutant emissions from wildfires due to climate warming.
 While in the Western U.S. burned area from wildfires seems to have increased

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- and the increase may be linked to climate, the present study simulates only very small relative increase for the region. Most of the predicted increase for North America concerns the boreal forest zone.
 - Demographic changes appear to be the main driver for the expected changes in wildfire emissions in Sub-Saharan Africa. For a scenario of high population growth and slow urbanisation, there will be large decreases in emissions in many parts of the continent, often dropping below thresholds that make them relevant for air quality policy. The decrease will be much smaller or turn into an increase for a scenario of low population growth and fast urbanisation.
 - Exposure of humans to PM2.5 in Sub-Saharan Africa is expected to drop for all demographic scenarios, but mostly for high population growth and slow urbanisation. Stronger fire suppression by higher rural population outweighs the effect of larger populations in rural areas.
 - Globally, both the number of people and the percentage of world population exposed to dangerously high PM2.5 emissions from wildfires is expected to increase in all scenarios considered. Both relative and absolute increase are highest for high population growth, while the degree of urbanisation plays only a minor role. This is opposite to the average fractional burned area in densely populated regions – a measure of fire risk to properties and lives – where the projected increase was earlier found to depend mostly on the degree of urbanisation.
 - The goal of reducing PM2.5 emissions globally such that the WHO guidelines for PM2.5 concentrations are met everywhere may not be attainable because in many regions wildfire emissions will remain above critical thresholds. So far, there is no generally accepted method for wildfire management that has been

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696	shown to lead to lasting reductions in fire activity or emissions. The still
697	widely used approach of aggressive fire suppression is not only costly, but
698	may even have led to increased overall fire activity. It may therefore be
699	prudent to accept the existence of wildfires as a natural phenomenon with
700	important ecosystem function and adapt urban planning accordingly.
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703	climate interaction Study, PEGASOS) and grant 603542 (Land-use change: assessing
704	the net climate forcing, and options for climate change mitigation and adaptation,
705	LUC4C).
706	Author contributions: WK conceived of the study, carried out the analysis and wrote
707	the first draft of the manuscript. All authors contributed to discussions and writing.
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Table 1: Emissions [Gg/yr] by world region from various sources

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Region	Wildfire ¹ I	Deforestation ¹ Peat fire ¹	Peat fire ¹	$Anthropogenic^2$	$Wildfire^{1}$	Deforestation ¹ Peat fire ¹	Peat fire ¹	Anthropogenic ²	
		[P]	PM2.5			Z	NOx		
Saharan Africa	14,973	538	0	5,864	8,141	151	0	2,169	
Latin America & Caribbean	3,138	1,886	0	2,534	1,655	528	0	4,956	
Easter Europe-Russia-Central Asia	2,832	0	18	2,490	265	0	6	2,967	
South & Southeast Asia	1,593	1,499	598	10,392	791	420	297	6,155	
Australia & New Zealand	1,536	22	0	186	747	9	0	834	
North America	1,349	0	30	1,462	126	0	15	13,654	
Developing East Asia	364	17	0	13,324	73	S	0	11,587	
High-income Europe	31	0	0	1,630	13	0	0	7,216	
Developing Middle East & North Africa	7	0	0	1,561	4	0	0	4,236	
Globe	25,842	3,968	646	40,370	11,822	1,112	321	59,652	
			93			I	BC		
Sub-Saharan Africa	131,545	5,500	0	69,581	773	31	0	801	
Latin America & Caribbean	27,315	19,270	0	41,113	160	108	0	344	
Easter Europe-Russia-Central Asia	23,461	0	286	27,338	76	0	2	338	
South & Southeast Asia	13,623	15,319	9,739	98,011	80	98	72	1,217	
Australia & New Zealand	13,056	221	0	4,503	92	1	0	33	
North America	10,976	0	496	64,754	46	0	4	324	
Developing East Asia	2,886	173	3	146,050	14	1	0	1,559	
High-income Europe	250	0	0	27,150	1	0	0	357	
Developing Middle East & North Africa	59	0	0	23,898	0	0	0	248	
Globe	223,331	40,557	10,524	518,576	1,249	227	77	5,397	
GFED4.1s									
² ECLIPSE-GAINS 4a									





7	Table 2: Global	Table 2: Global population affected by wildfire PM2.5 emissions above given limit	ed by wild	fire Pl	<i>M</i> 2.5 emis	sions a	bove giv	en limit.		٠							
	Population	Urbanization			RCP4.5						RCP8.5						
	Limit [g m ⁻² yr ⁻¹]] 0.2		-													
			Current	% ₁₎	2030	%	2050	%	2090	%	2030	%	2050	%	2090	%	
	SSP5	fast			721	9.8	892	10.0	1054	12.5	721	8.6	888	6.6	1080	12.8	
	SSP2	fast			771	6.8	1032	10.6	1498	14.5	775	8.9	1034	10.6	1531	14.9	
	SSP2	central	268	7.9	780	0.6	1042	10.7	1466	14.2	782	0.6	1037	10.7	1537	14.9	
	SSP2	slow			787	9.1	1045	10.8	1468	14.3	785	0.6	1042	10.7	1560	15.1	
	SSP3	slow			844	9.3	1302	12.1	2221	16.0	842	9.3	1273	11.9	2299	16.6	
	Limit [g m ⁻² vr ⁻¹]	1 0.5															
	SSP5	fast			313	3.7	384	4.3	456	5.4	317	3.8	391	4.4	489	5.8	
	SSP2	fast			333	3.8	441	4.5	646	6.3	337	3.9	451	4.6	681	9.9	
	SSP2	central	256	3.6	340	3.9	446	4.6	989	6.2	341	3.9	452	4.7	682	9.9	
	SSP2	slow			349	4.0	455	4.7	629	6.1	348	4.0	454	4.7	889	6.7	
	SSP3	slow			372	4.1	531	4.9	911	9.9	374	4.1	528	4.9	973	7.0	
	Limit $\lceil g m^{-2} vr^{-1} \rceil$	1															
	SSP5	fast			146	1.7	176	2.0	204	2.4	148	1.8	171	1.9	219	2.6	
	SSP2	fast			154	1.8	193	2.0	280	2.7	155	1.8	188	1.9	302	2.9	
	SSP2	central	126	1.8	156	1.8	187	1.9	267	2.6	158	1.8	187	1.9	291	2.8	
	SSP2	slow			162	1.9	194	2.0	247	2.4	162	1.9	192	2.0	266	2.6	
	SSP3	slow			173	1.9	211	2.0	330	2.4	172	1.9	209	2.0	351	2.5	
	1) Per cent of global population	bal population															





Table 3: Average annual exposure of population to PM2.5 [μ g m] under idealized conditions.	population to PM2.	5 [µg m] 5	under ideal	ized condit	ions.			
Population	Urbanization	Н	RCP4.5		H	RCP8.5		
•		Current	2030	2050	2090	2030	2050	. 1
Sub-Saharan Africa								
SSP5	fast		9.9	6.5	6.4	9.9	6.3	
SSP2	fast		6.5	6.2	0.9	6.5	0.9	
SSP2	central	7.7	6.5	0.9	5.7	6.5	5.9	
SSP2	slow		6.4	5.8	5.4	6.4	5.7	
SSP3	slow		6.3	5.6	5.1	6.2	5.5	
Australia and New Zeeland								
SSP5	fast		3.2	3.5	3.6	3.2	3.5	
SSP2	fast		3.2	3.5	3.6	3.2	3.5	
SSP2	central	3.1	3.2	3.5	3.6	3.2	3.5	
SSP2	slow		3.2	3.5	3.6	3.2	3.5	
SSP3	slow		3.2	3.5	3.7	3.2	3.5	
Latin America and Caribbean								
SSP5	fast		1.5	1.6	1.7	1.5	1.6	
SSP2	fast		1.5	1.5	1.6	1.5	1.6	
SSP2	central	1.5	1.5	1.5	1.6	1.5	1.6	
SSP2	slow		1.5	1.5	1.6	1.5	1.6	
SSP3	slow		1.5	1.5	1.5	1.5	1.5	





906 Figures

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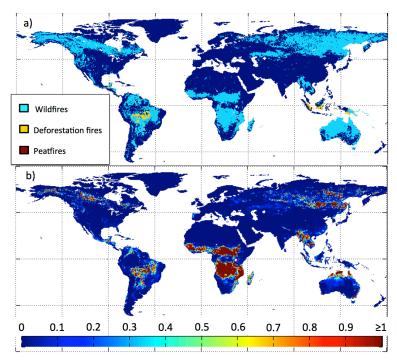


Figure 1: a) Largest current source of PM2.5 emissions, including anthropogenic sources (dark blue areas); b) wildfires emissions in g PM2.5 m⁻² yr⁻¹. Average annual PM2.5 emissions 1997 to 2014 are from to GFED4.1s, or ECLIPSE GAINS 4a for 2010 (anthropogenic).





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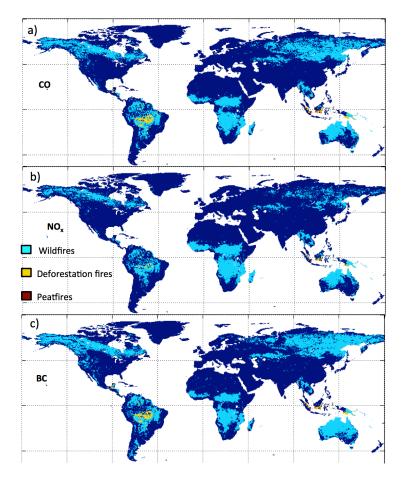


Figure 2: Largest current source of annual air pollutant emissions. Dark blue areas:

dominant source anthropogenic, or zero emissions. a) CO, b) NO_x, c) black carbon.





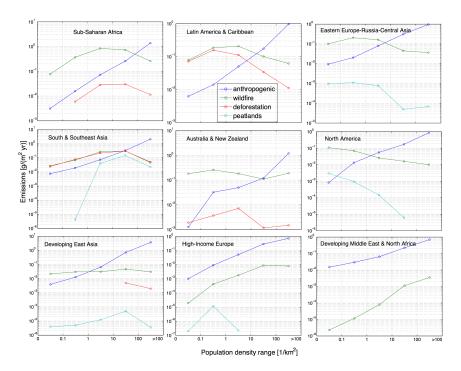


Figure 3: Comparison of current anthropogenic, wildfire, deforestation fire and peat fire PM2.5 emissions by region and range population density, constructed by relating emission rates to the population density found for the same grid box on a 0.5 by 0.5 degree grid. Emissions are from ECLIPSE GAINS 4a (anthropogenic source for 2010) and GFED4.1s (average of 1997-2014).





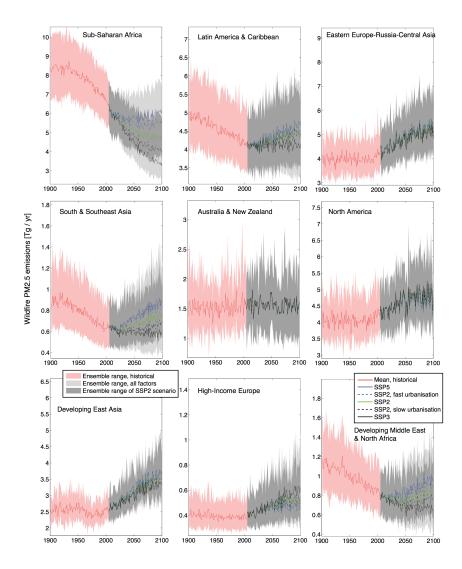


Figure 4. Annual PM2.5 emissions from ensemble of LPJ-GUESS-SIMFIRE simulations for nine world regions 1901 to 2100. Ensemble ranges and impact of population and urbanisation scenarios. Climate scenario: RCP4.5.





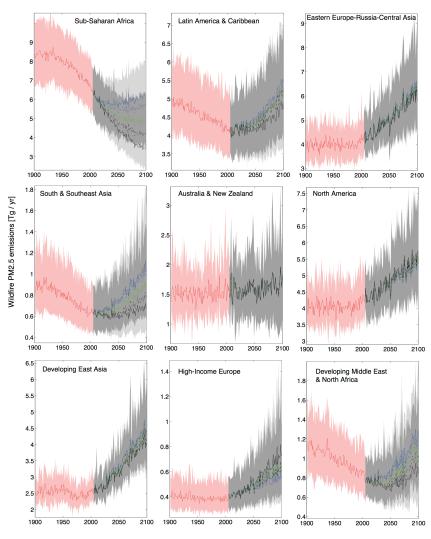


Figure 5. Same as previous figure, but for climate scenario RCP8.5.

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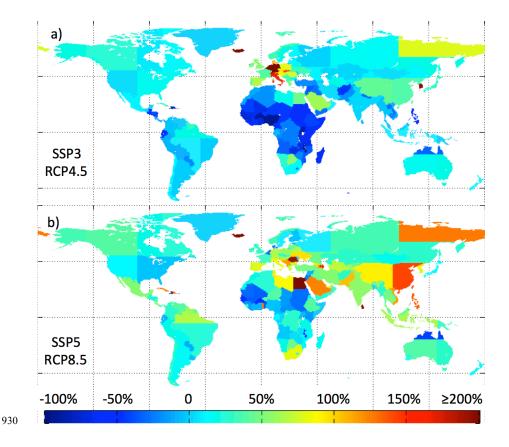


Figure 6: Relative change in annual PM2.5 emissions from current (1997-2014 mean) to 2090 (2080 to 2100 mean) by country/region. a) SSP3 globally high population growth (high-income countries: low population growth) with slow urbanisation and RCP4.5 climate scenario, b) SSP5 globally low population growth (high-income countries: high population growth) with slow urbanisation, RCP8.5 climate scenario.





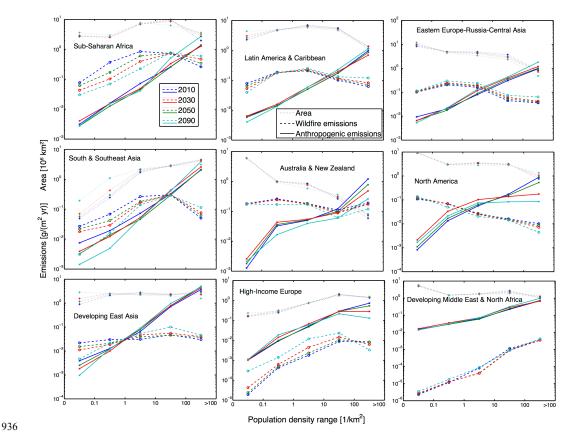


Figure 7: Predicted changes of annual PM2.5 emissions against ranges of population density from wildfires and anthropogenic sources, as well as changes in area extent of the population-density categories, for the nine world regions, based on re-scaled GFED4.1s wildfire emissions. SSP3 demographic scenario, RCP4.5 climate change and Current-Legislation (CLE) anthropogenic emissions scenario.

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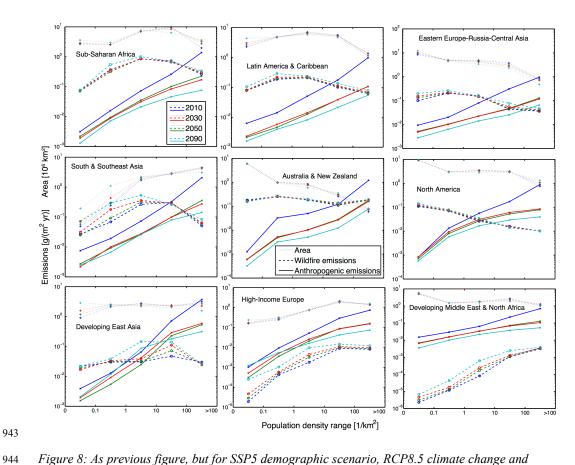


Figure 8: As previous figure, but for SSP5 demographic scenario, RCP8.5 climate change and

Maximum Feasible Reduction (MFR) anthropogenic-emissions scenario. 945





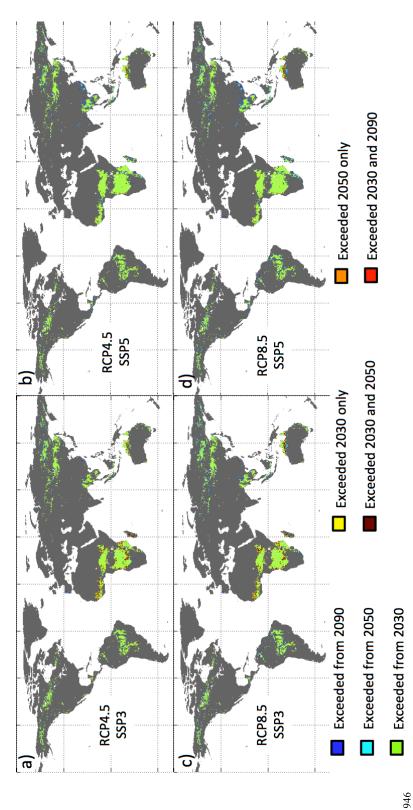


Figure 9: Timing of when annual wildfire emissions exceed 0.5 g PM2.5 m⁻² yr⁻¹ for the time windows 2030, 2050 and 2090a, b) RCP4.5 and c,

d) RCP8.5 climate scenario; a, c) SSP3 and b, d) SSP5 demographic scenario.

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Appendix 949

Table A1: Countries/regions used for scaling GFED4.1s wildfire emissions. 950

		_	grid cells on		_	•		grid cells			
		World	1-degree	non-crop			World	on 1-	non-crop	longitude	latitud
Code	Country name	region	grid	cells ¹	Code	Country name	region	degree grid	cells	range	range
AGO	Angola		100	100	ARM	Armenia		4	4		
BEN	Benin		10	6	AZE	Azerbaijan		7	6		
BWA	Botswana		46	46	GEO	Georgia		6	6		
BFA	Burkina Faso		23	21	KAZ	Kazakstan		213	206		
BDI	Burundi		3	3	KGZ	Kyrgyzstan		16	16		
CMR	Cameroon		37	35	TJK	Tajikistan		8	8		
CAF	Central African		49	46	TKM	Turkmenistan		35	35		
	Republic		100	100	UZB	Uzbekistan		35	32		
TCD	Chad				BLR	Belarus	Eastern	16	14		
COG	Congo		24	24	BGR	Bulgaria	Europe,	10	8		
ZAR	Congo, Dem.		176	176	Ш	Latvia and Lithunania	Russia and	10	5		
	Republic				ROM	Romania	Central Asia	19	10		
CIV	Cote d'Ivoire		25	24	RUS-SW			78		W of 60°E	S of 52
RI	Eritrea		12	12	RUS-NW			212		W of 55°E	N of 52
TH	Ethiopia		90	89	RUS-C	Russian Federation		553	506		ther RUS
SAB	Gabon		20	20	RUS-SE			232		E of 110°E	S of 60
HA	Ghana		18	13	RUS-NE			327		E of 110°E	N of 60
SIN	Guinea		20	20	UKR	Ukraine		47	5		
SNB	Guinea-Bissau	Sub-	1	1	YUA	Serbia, Montenegro,		13	11		
ŒN	Kenya	Saharan	41	40		Bosnia, Macedonia					
SO	Lesotho	Africa	1	1	CHN-W			348		W of 105°E	
.BR	Liberia	741166	5	5	CHN-E	China	Developing	283		E of 105°E	S of 43
ИDG	Madagascar		43	43	CHN-N		East Asia	122	103		N of 43
/WI	Malawi		10	10	PRK	North Korea	Lust Hald	10	8		
/LI	Mali		106	106	MNG	Mongolia		131	131		
/IRT	Mauritania		80	80	BTN	Bhutan		4	2		
1OZ	Mozambique		61	60	KHM	Cambodia		14	13		
IAM	Namibia		66	66	IND	India		256	50		
IER	Niger		94	80	IDN	Indonesia		125	116		
IGA	Nigeria		74	40	LAO	Laos		18	18		
EN	Senegal		17	16	MYS	Malaysia	e and and	23	23		
LE	Sierra Leone		5	5	MMR	Myanmar	South and South-East	44	36		
OM	Somalia		55	55	NPL	Nepal		12	8		
AF	South Africa		99	98	PAK	Pakistan	Asia	58	44		
DN	Sudan		207	199	PHL	Philippines		16	14		
GO	Togo		2	1	LKA	Sri Lanka		4	4		
GA	Uganda		16	13	THA	Thailand		42	30		
ZA	Tanzania		73	72	VNM	Viet Nam		27	22		
MB	Zambia		63	63	PNG	Papua New Guinea		31	31		
WE	Zimbabwe		30	30	AUS-SW			18	16	W of 120°E	S of 30
ZA	Algeria		189	184	AUS-E	A	Australia	200	178	E of 140°E	S of 18
GY	Egypt		77	76	AUS-C	Australia	and New	317	316	not in o	
BY	Libya		131	131	AUS-N		Zealand	76	76		N of 1
1AR	Morocco		56	49	NZL	New Zealand		22	22		
UN	Tunisia		14	11	CAN-W			385	341	W of 100°W	
FG	Afghanistan	Developing		52	CAN-C	Canada		192	185	10080°W	
RN	Iran	Middle East	134	129	CAN-E		North	176		E of 80°W	
RQ	Iraq	and North	37	31	USA-W		America	314		W of 100°W	
OR	Jordan	Africa	6	6	USA-E	United States of		372		E of 100°W	
AU	Saudi Arabia		154	154	ALK	America		116	116		N of 5
YR	Syria		15	9	CRI	Costa Rica		3	3		
UR	Turkey		57	44	CUB	Cuba		7	5		
EM	Yemen		31	31	DOM	Dominican Republic		4	4		
UT	Austria		7	7	GTM	Guatemala		15	15		
NL	Benelux		5	3	HTI	Haiti		2	2		
	Croatia and				HND	Honduras		9	9		
RS	Slovenia		3	2	MEX-W			120		W of 95°W	
			3								
7F				2	MEX-SE	Mexico					
	Czech Republic		5	3	MEX-SE NIC			19	19	E of 95°W	
NK	Czech Republic Denmark		5 6	3	NIC	Nicaragua		19 8	19 8		
NK ST	Czech Republic Denmark Estonia		5 6 4	3 4	NIC PAN	Nicaragua Panama		19 8 6	19 8 6		
NK ST IN	Czech Republic Denmark Estonia Finland		5 6 4 28	3 4 27	NIC PAN ARG	Nicaragua Panama Argentina	Latina	19 8 6 230	19 8 6 207		
INK ST IN RA	Czech Republic Denmark Estonia Finland France		5 6 4 28 41	3 4 27 24	NIC PAN ARG BOL	Nicaragua Panama	Latina America and	19 8 6 230 88	19 8 6 207 88	E of 95°W	s of r
INK ST IN RA EU	Czech Republic Denmark Estonia Finland France Germany		5 6 4 28 41 32	3 4 27 24 29	NIC PAN ARG BOL BRA-W	Nicaragua Panama Argentina Bolivia	America and	19 8 6 230 88 197	19 8 6 207 88 192	E of 95°W W of 49°W	S of 5
NK ST IN RA EU RC	Czech Republic Denmark Estonia Finland France Germany Greece	High-	5 6 4 28 41 32	3 4 27 24 29 9	NIC PAN ARG BOL BRA-W BRA-E	Nicaragua Panama Argentina		19 8 6 230 88 197 316	19 8 6 207 88 192 294	E of 95°W	
NK ST IN RA EU RC UN	Czech Republic Denmark Estonia Finland France Germany Greece Hungary		5 6 4 28 41 32 10 7	3 4 27 24 29 9	NIC PAN ARG BOL BRA-W BRA-E BRA-N	Nicaragua Panama Argentina Bolivia Brazil	America and	19 8 6 230 88 197 316 161	19 8 6 207 88 192 294 161	E of 95°W W of 49°W	
NK ST IN RA EU RC UN	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland	High-	5 6 4 28 41 32 10 7	3 4 27 24 29 9 2 7	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL	Nicaragua Panama Argentina Bolivia Brazil Chile	America and	19 8 6 230 88 197 316 161 61	19 8 6 207 88 192 294 161 60	E of 95°W W of 49°W	
NK ST N RA EU RC UN	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Ireland	High- income	5 6 4 28 41 32 10 7 7	3 4 27 24 29 9 2 7	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia	America and	19 8 6 230 88 197 316 161 61 88	19 8 6 207 88 192 294 161 60 88	E of 95°W W of 49°W	
NK ST IN RA EU RC UN SL SL	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Ireland Italy	High- income	5 6 4 28 41 32 10 7 7 5 5	3 4 27 24 29 9 2 7 4	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador	America and	19 8 6 230 88 197 316 161 61 88	19 8 6 207 88 192 294 161 60 88	E of 95°W W of 49°W	
NK ST N RA EU RC UN L IL IL CA	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Ireland Italy Norway	High- income	5 6 4 28 41 32 10 7 7 5 23	3 4 27 24 29 9 2 7 4 11 31	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana	America and	19 8 6 230 88 197 316 161 61 88 19 6	19 8 6 207 88 192 294 161 60 88 19 6	E of 95°W W of 49°W	
NK ST N RA EU RC UN L L IL CA OR	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Italy Norway Poland	High- income	5 6 4 28 41 32 10 7 7 5 23 31 25	3 4 27 24 29 9 2 7 4 11 31	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana	America and	19 8 6 2300 88 197 316 161 61 88 19 6	19 8 6 207 88 192 294 161 60 88 19 6	E of 95°W W of 49°W	
NK ST IN RA EU RC UN SL SL RA OR OR	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal	High- income	5 6 4 28 41 32 10 7 7 5 23 31 25 6	3 4 27 24 29 9 2 7 4 11 31 11	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Gulana Guyana Paraguay	America and	19 8 6 230 88 197 316 161 61 88 19 6	19 8 6 207 88 192 294 161 60 88 19 6	E of 95°W W of 49°W	
NK ST IN RA EU CC UN IL RL A OOR OOR	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Iraland Italy Norway Poland Portugal Slovakia	High- income	5 6 4 28 41 32 10 7 7 5 23 31 25 6	3 4 27 24 29 9 2 7 4 11 31 11 5	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL GUF GUY PRY PER	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guina Guyana Paraguay	America and	19 8 6 230 88 197 316 161 61 88 19 6 15 28	19 8 6 207 88 192 294 161 60 88 19 6	E of 95°W W of 49°W	
INK ST IN RA RA GIUN GL GR GO	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Italy Norway Poland Portugal Slovakia Spain	High- income	5 6 4 41 32 10 7 7 5 23 31 25 6 6	3 4 27 24 29 9 2 7 4 11 31 11 5 2	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru	America and	19 8 6 230 88 197 316 161 61 88 19 6 15 28	19 8 6 207 88 192 294 161 60 88 19 6 15 28	E of 95°W W of 49°W	
NK ST N RA EU RC UN IL RI A OR OL RT SSP WE	Czech Republic Denmark Estonia Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Syakia Sweden	High- income	5 6 4 4 28 411 32 100 7 7 7 5 23 311 25 6 6 6 400 39	3 4 27 24 29 9 2 7 4 4 11 31 11 5 2 2 24	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay	America and	19 8 6 230 88 197 316 161 61 188 19 6 15 28 100 11	19 8 6 207 88 192 294 161 60 88 19 6 15 28 100	E of 95°W W of 49°W	
NK ST N RA EU RC UN IL RI A OR OL RT SSP WE	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland	High- income	5 6 4 41 32 10 7 7 5 23 31 25 6 6	3 4 27 24 29 9 2 7 4 11 31 11 5 2	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay	America and	19 8 6 6 230 88 197 316 61 61 161 61 88 8 19 6 6 15 28 100 11 15 73	19 8 6 207 888 192 294 161 60 88 81 19 6 15 28 100 11 15 73	E of 95°W W of 49°W	
NK ST N RA EU RC UN IL L O O R O O N ST V K S P W E H E	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland United	High- income	5 6 4 4 288 41 32 2100 7 7 7 5 23 31 25 6 6 6 400 39 2 2	3 4 27 24 29 9 2 7 4 11 11 5 2 2 4 39	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela	America and	19 8 6 6 230 88 197 7 316 611 611 61 88 8 199 6 6 15 28 1000 111 15 73 4 4	19 8 6 6 207 88 81929 294 161 600 608 88 199 15 28 1000 111 15 73 4 4	E of 95°W W of 49°W	
NK ST N RA EU RC UN IL L O O R O O N ST V K S P W E H E	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland	High- income	5 6 4 4 28 411 32 100 7 7 7 5 23 311 25 6 6 6 400 39	3 4 27 24 29 9 2 7 4 4 11 31 11 5 2 2 24	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR JPN	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela Israel	America and	19 8 6 6 230 88 197 316 61 161 88 199 6 15 28 100 11 15 73 4 28 8	19 8 6 207 88 192 294 161 60 88 199 19 6 15 28 100 11 15 73 4 2 8 8	E of 95°W W of 49°W	
NK ST N N R E U R C U N L L A O R O L L S F E E E E E E E E E E E E E E E E E E	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland United	High- income	5 6 4 4 288 41 32 2100 7 7 7 5 23 31 25 6 6 6 400 39 2 2	3 4 27 24 29 9 2 7 4 11 11 5 2 2 4 39	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR JPN KOR	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela	America and	19 8 6 6 230 88 197 7 316 611 611 61 88 8 199 6 6 15 28 1000 111 15 73 4 4	19 8 6 6 207 88 81929 294 161 600 608 88 199 15 28 1000 111 15 73 4 4	E of 95°W W of 49°W	
NK ST N RA EU RC UN IL L O O R O O N ST V K S P W E H E	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland United	High- income	5 6 4 4 288 41 32 2100 7 7 7 5 23 31 25 6 6 6 400 39 2 2	3 4 27 24 29 9 2 7 4 11 11 5 2 2 4 39	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR JPN	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela Israel	America and	19 8 6 6 230 88 197 316 61 161 88 199 6 15 28 100 11 15 73 4 28 8	19 8 6 207 88 192 294 161 60 88 199 19 6 15 28 100 11 15 73 4 2 8 8	E of 95°W W of 49°W	
NK ST N N R E U R C U N L L A O R O L L S F E E E E E E E E E E E E E E E E E E	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland United	High- income	5 6 4 4 288 41 32 2100 7 7 7 5 23 31 25 6 6 6 400 39 2 2	3 4 27 24 29 9 2 7 4 11 11 5 2 2 4 39	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR JPN KOR	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela Israel Japan	America and	19 8 6 6 230 88 8197 316 161 661 88 19 6 15 28 100 11 15 73 4 4 28 8 6 6	19 8 6 6 2077 888 1922 2944 1611 600 888 199 155 28 1000 111 155 73 4 28 6 6	E of 95°W W of 49°W	
NK ST N N R E U R C U N L L A O R O L L S F E E E E E E E E E E E E E E E E E E	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland United	High- income	5 6 4 4 288 41 32 2100 7 7 7 5 23 31 25 6 6 6 400 39 2 2	3 4 27 24 29 9 2 7 4 11 11 5 2 2 4 39	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR JPN KOR OMN	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela Israel Japan South Korea	America and	19 8 6 6 230 88 88 197 316 61 61 161 1 15 28 100 11 15 73 4 4 28 6 6 25 6	19 8 8 6 207 88 88 192 294 161 161 160 175 28 100 111 15 73 4 28 6 6 26 6	E of 95°W W of 49°W	
ZE NK ST ST ST ST ST ST SE SE SE SE ST ST SE ST ST ST ST ST ST ST ST ST ST ST ST ST	Czech Republic Denmark Estonia Finland Finland France Germany Greece Hungary Iceland Ireland Italy Norway Poland Portugal Slovakia Spain Sweden Switzerland United	High- income	5 6 4 4 288 41 32 2100 7 7 7 5 23 31 25 6 6 6 400 39 2 2	3 4 27 24 29 9 2 7 4 11 11 5 2 2 4 39	NIC PAN ARG BOL BRA-W BRA-E BRA-N CHL COL ECU GUF GUY PRY PER SUR URY VEN ISR JPN KOR OMN ARE	Nicaragua Panama Argentina Bolivia Brazil Chile Colombia Ecuador French Guiana Guyana Paraguay Peru Suriname Uruguay Venezuela Israel Japan South Korea Oman	America and	19 8 8 6 230 88 8 197 316 161 161 182 8 8 199 6 6 155 28 100 111 155 73 4 4 28 8 6 6 26 8 8	19 8 6 207 288 192 294 161 16 60 888 199 6 6 155 28 1000 111 15 73 4 28 6 6 26 8 8	E of 95°W W of 49°W	S of 5

¹Cells with less than 50% cropland fraction in past or future scenarios

²Constant emissions assumed because dominated by croplands ³Constant emissions assumed because zero current wildfire emissions in some simulations





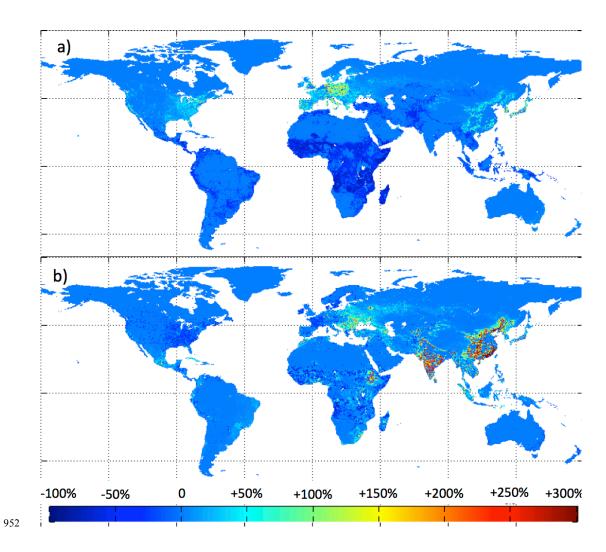


Figure A1: Relative change in wildfire emissions due to changes in population density from 2010 to

954 2090 according to Equ. 2. a) SSP3, b) SSP5 demographic scenario.

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Figure A2: World regions used in the analysis. Dark blue: not included





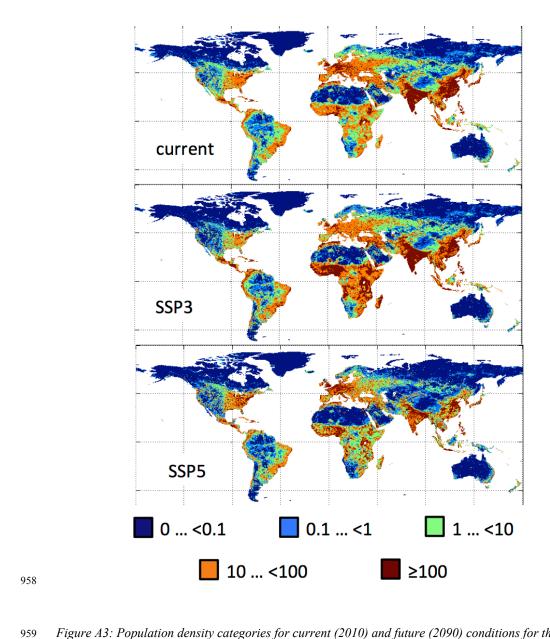


Figure A3: Population density categories for current (2010) and future (2090) conditions for the SSP3 and SSP3 demographic scenarios.





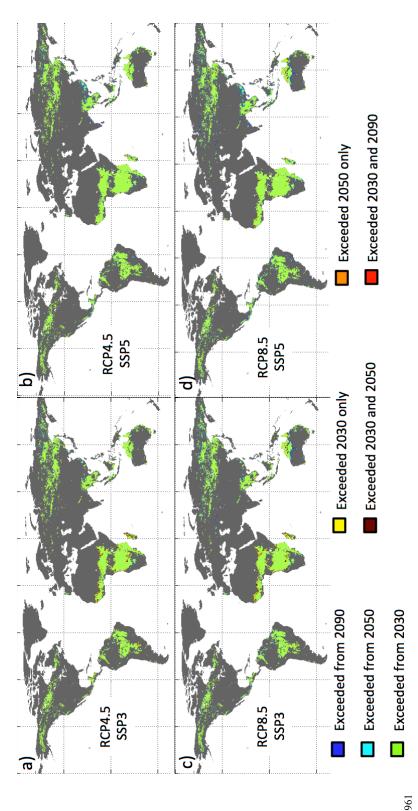


Figure A4: Timing of when annual wildfire emissions exceed 0.2 g PM2.5 m⁻² yr⁻¹ for the time windows 2030, 2050 and 2090a, b) RCP4.5 and c, d) RCP8.5 climate scenario; a, c) SSP3 and b, d) SSP5 demographic scenario. 963





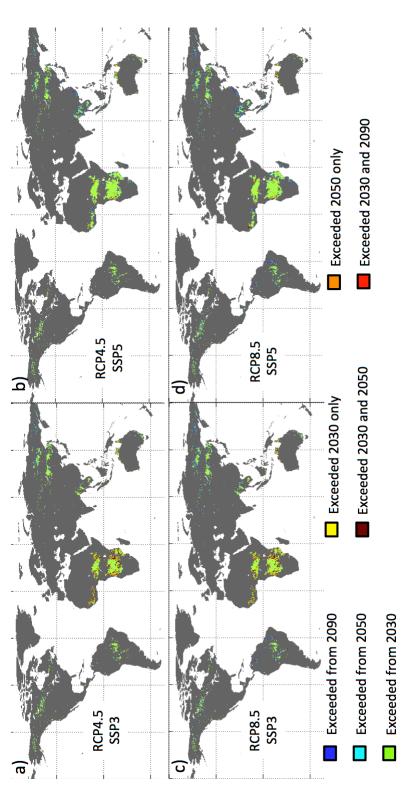


Figure A5: as previous figure, but for a threshold of 1 g PM2.5 m⁻² yr⁻¹.

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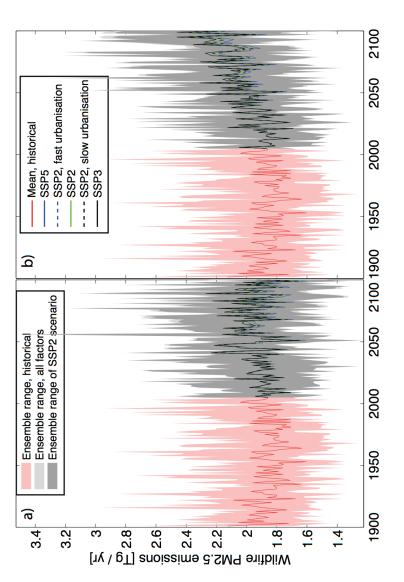


Figure 46: Simulated annual wildfire PM2.5 emissions by LPJ-GUESS-SIMFIRE for the Western U.S. (region USA-W, Table A1).

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