

Future changes in wet and dry season characteristics in CMIP5 and CMIP6 simulations

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Future Changes in Wet and Dry Season Characteristics in CMIP5 and
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ABSTRACT

Climate change will result in more dry days and longer dry spells, however, the resulting impacts on 13 crop growth depend on the timing of these longer dry spells in the annual cycle. Using an ensemble 14 of Coupled Model Intercomparison Project Phase 5 and Phase 6 (CMIP5 and CMIP6) simulations, 15 and a range of emission scenarios, here we examine changes in wet and dry spell characteristics 16 under future climate change across the extended tropics in wet and dry seasons separately. Delays 17 in the wet seasons by up to two weeks are projected by 2070-2099 across South America, Southern 18 Africa, West Africa and the Sahel. An increase in both mean and maximum dry spell length during 19 the dry season is found across Central and South America, Southern Africa and Australia, with 20 a reduction in dry season rainfall also found in these regions. Mean dry season dry spell lengths 21 increase by 5-10 days over north-east South America and south-west Africa. However, changes 22 in dry spell length during the wet season are much smaller across the tropics with limited model 23 consensus. Mean dry season maximum temperature increases are found to be up to 3°C higher 24 than mean wet season maximum temperature increases over South America, Southern Africa and 25 parts of Asia. Longer dry spells, fewer wet days, and higher temperatures during the dry season 26 may lead to increasing dry season aridity, and have detrimental consequences for perennial crops. 27

28 1. Introduction

²⁹ Climate change will result in more dry days and longer dry spells; this has the potential to lead ³⁰ to negative impacts on crop yields and food security, as water stress limits crop growth. Across the ³¹ tropics most crops are grown solely during the annual wet season, with few crops grown during ³² the dry season. Therefore the timing of these longer dry spells in the annual cycle is of crucial ³³ importance in determining their impact.

As the climate warms, the contrasting constraints of energy budgets at global scales and moisture 34 budgets at regional scales drive a general increase in precipitation intensity and decrease in fre-35 quency that are further altered as atmospheric circulation patterns shift in location (Trenberth 2011; 36 Held and Soden 2006; Funk et al. 2019; Giorgi et al. 2019; Allan et al. 2020). A larger increase in 37 precipitation intensity compared to mean precipitation is balanced by a decrease in the number of 38 wet days, resulting in longer dry spells and shorter wet spells (Giorgi et al. 2011; Seneviratne et al. 39 2012; Sillmann et al. 2013b). These longer dry spells may have negative impacts on crop yields 40 and food production (Rockström et al. 2010), as reductions in water availability limit crop growth. 41 Heavier rainfall events punctuating longer dry spells can also affect hydrological characteristics; 42 for example, more rainfall can be taken up by the soil rather than producing runoff yet on the other 43 hand, more intense rainfall falling on encrusted ground can lead to higher runoff (Eekhout et al. 44 2018; Yin et al. 2018). 45

Many studies have reported an increase in the number of dry days and dry spell lengths under future climate change. Giorgi et al. (2019) found an increase in the number of dry days and dry spell lengths and a decrease in wet spell lengths aggregated over land areas in the tropics and the extra-tropics over the 21st century in 10 Coupled Model Intercomparison Project Phase 5 (CMIP5) model projections. Giorgi et al. (2014) report that the response in wet and dry spell lengths is more

pronounced and spatially consistent over the tropics than the extratropics. Polade et al. (2014) 51 also identified an increasing dry day frequency, in particular over northern South America, Central 52 America, Southern Africa and the Mediterranean; regions also identified by Pascale et al. (2016) 53 as showing increases in the number of dry days. While Giorgi et al. (2014) and Giorgi et al. (2019) 54 examined the mean wet and dry spell lengths over large land areas, Sillmann et al. (2013b) looked 55 at the maximum wet and dry spell lengths over a number of regions. They find significant increases 56 in maximum dry spell length over Central America, parts of South America, the Mediterranean, 57 Southern Africa, and south and south-east Asia, (consistent with the regions where Polade et al. 58 (2014) found more dry days) with decreases in maximum dry spell length over the high northern 59 latitudes, north-east Asia and East Africa. Lau et al. (2013) also suggest increasing dry spell 60 lengths over Southern Africa, the Mediterranean and parts of South America, although they use a 61 threshold of <0.024 mm day⁻¹ to look at the changes in the frequency of dry months. Overall, the 62 Intergovernmental Panel on Climate Change (IPCC) Special Report on Extremes concluded that 63 an increase in dry spell length and frequency was very likely over the Mediterranean and southern 64 Australia, and likely over most subtropical regions (Seneviratne et al. 2012). 65

Most of these previous studies have looked at changes in wet and dry spells across the entire 66 calendar year (Sillmann et al. 2013b). In the tropics, however, most crops are grown during the 67 annual wet season. Therefore the impact of longer or shorter wet and dry spells may vary throughout 68 the year. Longer dry spells during the main wet season may have significant detrimental impacts 69 on agriculture, whereas longer dry spells within the period of the year that is climatologically 70 dry may have less impact. Conversely, longer wet spells within the dry season may be beneficial 71 e.g. for recharging water supplies. Therefore, understanding the timing of changes in wet and 72 dry spell lengths is of crucial importance. Orlowsky and Seneviratne (2012) looked at changes 73 in consecutive dry days (CDD) in the four meteorological seasons and found some differences in 74

the projected changes in different seasons; for example, over Asia there is a decrease in CDD in 75 December-February (DJF), but an increase in June-August (JJA). The magnitude of changes was 76 also different in different seasons; for example, over Southern Africa the increase in CDD was 77 greater in JJA than in DJF. However, these 3-month meteorological seasons may not map directly 78 onto wet and dry seasons and there may be changes in wet and dry season length in the future 79 (Dunning et al. 2018). Other studies looking at changes in wet and dry spell lengths in wet seasons 80 have focused over certain regions, for example, Klutse et al. (2018) looked at changing wet and dry 81 spell lengths in June-September over West Africa, and Mba et al. (2018) examined changing wet 82 and dry spell lengths in March-May and September-November over Central Africa. Furthermore, 83 very few studies have also considered the impact of temperature changes; increases in temperature 84 combined with longer dry spells may lead to reductions in soil moisture and increased water stress 85 for crops. Changes in extreme temperatures may differ in wet and dry seasons, hence it is important 86 to separately examine the wet and dry seasons, so as not to exaggerate the impact of temperature 87 changes on agriculture. 88

Although the seasonal timing of changes in wet and dry spells is of significant societal importance, 89 it has not been investigated consistently in previous studies. Many studies have found a trend of 90 wet seasons getting wetter and dry seasons getting drier under future climate change (Chou et al. 91 2013; Kumar et al. 2015; Schurer et al. 2020), suggesting there may be notable seasonal differences 92 in wet and dry spell projections. Thus, the aim of the present study is to answer the question 93 "Do changes in the length of wet and dry spells differ in climatologically wet and dry seasons?". 94 Unlike previous dry spell studies, a recently developed methodology is used to determine the 95 timing of the wet and dry season at each location (Liebmann et al. 2012; Dunning et al. 2016), 96 thus enabling investigation of changes in the wet and dry season using a location-specific, not 97 month-based, seasonal definition. Furthermore, we utilise projections from the latest version of 98

coupled global climate models, which have recently been released (Coupled Model Intercomparison 99 Project Phase 6, CMIP6, Eyring et al. 2016); these are of higher resolution with more complex 100 models than the previous generation of coupled global climate models (CMIP5, Taylor et al. 2012), 101 and thus represent a new opportunity for exploring changes in daily quantities. Here, results 102 from both CMIP5 and CMIP6 are compared, using a range of emissions scenarios. Many previous 103 studies have used the well-established Climdex (https://www.climdex.org/) indices to explore 104 changes in wet and dry spells (Zhang et al. 2011; Seneviratne et al. 2012; Sillmann et al. 2013b). 105 These define CDD (Consecutive Dry Days; maximum length of dry spell) and CWD (Consecutive 106 Wet Days; maximum length of wet spell) as "the maximum number of consecutive days with daily 107 precipitation less than/greater than 1mm day⁻¹" (for CDD/CWD respectively). This therefore only 108 looks at the longest wet/dry spell in a year. In this study we aim to look at a more complete set of 109 metrics, including number of wet days and mean length of wet/dry spells, to gain a more complete 110 picture of how wet and dry spells are changing under future climate change. Along with changes 111 in wet and dry spells, changes in maximum temperatures are also considered, as large temperature 112 increases may exacerbate the impacts of long dry spells. 113

114 2. Data and Methodology

115 a. Model Data

Daily precipitation data were used from 31 CMIP5 (Coupled Model Intercomparison Project Phase 5, Taylor et al. 2012) models and 19 CMIP6 (Coupled Model Intercomparison Project Phase 6, Eyring et al. 2016) models under a medium (Representative Concentration Pathway (RCP) 4.5/ Shared Socioeconomic Pathway (SSP) 245) and high (RCP8.5/SSP585) emissions scenario; a full list of models used is included in Tables S1 and S2. SSP585 is the highest emissions scenario and represents an update to the RCP8.5 emissions scenario (the highest emission scenario for CMIP5); both produce a radiative forcing of 8.5 W m⁻² in 2100 (Van Vuuren et al. 2011; O'Neill et al. 2016). While some have argued that this pathway is unrealistic, it should be noted that current emissions are in line with the RCP8.5 scenario (Schwalm et al. 2020). The SSP245 scenario represents a medium emissions scenario and updates the RCP4.5 pathway; both produce a radiative forcing of 4.5 W m⁻² in 2100 (Van Vuuren et al. 2011; O'Neill et al. 2016).

Models were selected based on the availability of daily precipitation data for historical and 127 RCP4.5/RCP8.5 (CMIP5) or SSP245/SSP585 (CMIP6) simulations for periods at the end of the 128 20th and 21st centuries. For the present day climate, the historical simulation was used over the 129 period 1 January 1985 - 31 December 2014, chosen to be in line with IPCC AR6. For the CMIP5 130 simulations the historical simulation is only available until 31 December 2005; CMIP5 RCP4.5 131 simulations were used for 1 January 2006 - 31 December 2014 (taken from the same model) and 132 appended to the historical simulations for 1 January 1985 - 31 December 2005. Recent analysis 133 suggests that CO₂ emissions have been above RCP4.5 levels over the recent period, so this gives a 134 conservative estimate (Schwalm et al. 2020). Projections from RCP4.5 and RCP8.5 do not differ 135 significantly over this 2005-2014 period, therefore the choice of scenario was assumed to have a 136 minimal impact. Furthermore, the primary focus is on the newer CMIP6 models; results from the 137 CMIP5 models are mainly presented for traceability with previous studies. The period 1 January 138 2070 - 31 December 2099 was used for the end of the 21st century period. Only the first ensemble 139 members were used (r1i1p1/r1i1p1f1) – for two CMIP6 models r1i1p1f2 was used due to r1i1p1f1 140 being unavailable (see Table S2). Other studies have taken a similar approach. Daily maximum 141 temperature data were also used for the same set of models and time periods, however, they were 142 unavailable for a couple of the models at the time when the analysis was conducted; this is indicated 143 in Table S2. 144

Data were considered over the region 50° S- 50° N. The focus here is on differences in wet and 145 dry seasons so the decision was taken to focus on this tropics and sub-tropics region that has a 146 well defined wet season (equatorward parts of mid-latitudes are also included in 50° S- 50° N); high 147 latitude regions tend to have less well-defined wet and dry seasons and thus are not included. All 148 analysis was done on the models at native resolution. For the multi-model median maps the metrics 149 were first computed at the native resolution and were then regridded onto the lowest resolution grid 150 using nearest-neighbouring remapping. For the regional analysis the regions were calculated for 151 each model grid and the model's native resolution grid was used. 152

153 b. Observational Data

¹⁵⁴ CHIRPS (Climate Hazards group Infrared Precipitation with Stations) precipitation data were ¹⁵⁵ used over 50°S-50°N for 1 January 1985 – 31 December 2014 (to be comparable to historical ¹⁵⁶ CMIP6 simulations). CHIRPS precipitation estimates use thermal infrared imagery and gauge ¹⁵⁷ data in addition to a monthly precipitation climatology, CHPClim (Climate Hazards Center's ¹⁵⁸ Precipitation Climatology) and CFS (Coupled Forecast System) version 2 reanalysis fields to ¹⁵⁹ produce rainfall estimates (Funk et al. 2015).

160 *c. Methodology*

To calculate metrics, including mean wet and dry spell lengths, for wet and dry seasons separately, each grid point is first classified into one of four seasonality categories (section 1), then season start and end dates are computed using a methodology for objectively calculating wet season onset and cessation dates (section 2). Metrics are then calculated for wet and dry seasons separately (section 3).

166 1) SEASONALITY CLASSIFICATION

The seasonality of precipitation varies across the tropics from monsoonal regions with one 167 summer wet season per year, to dry desert regions with little rainfall, to rainforest regions that are 168 wet year round. East Africa is known to experience two wet seasons per year (Yang et al. 2015). 169 To distinguish between wet and dry seasons, it is first important to determine the seasonal regime. 170 A number of different methodologies have been utilised for this purpose; Herrmann and Mohr 171 (2011) used monthly precipitation and temperature to define rainfall seasonality regimes across 172 Africa. Pascale et al. (2016) and Feng et al. (2013) used the Relative Entropy and Dimensionless 173 Seasonality Index to define and examine changes in the 'Global Monsoon Domain'. Liebmann 174 et al. (2012) and Dunning et al. (2016) use harmonic analysis to separate regions with one or two 175 wet seasons per year across Africa. A combination of these metrics is used here. 176

It was decided that four categories need to be distinguished: regions that are dry year-round, regions that are wet year-round, regions with one wet season per year (annual regime), and regions with two wet seasons per year (biannual regime). While Herrmann and Mohr (2011) defined additional regimes, here just four are used for simplicity. The methodology used has three steps:

- 181 1. The mean annual rainfall was computed at each land grid point (between 50° S and 50° N). The 20% of driest land grid points were excluded as dry regions with no marked wet season.
- ¹⁸³ 2. Grid points with a relative entropy (a measure of how uniform or variable the seasonal
 ¹⁸⁴ sequence of monthly rainfall is, see below) less than 0.3 were defined as wet year round with
 ¹⁸⁵ no marked wet/dry seasons.

3. The remaining grid points were defined as having one/two wet seasons per year based on the ratio of the amplitude of the second harmonic to the amplitude of the first harmonic. If the

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ratio is greater than 1/less than 1 then the grid point experiences a biannual/annual regime with two/one wet season(s) per year.

The relative entropy quantifies how different the observed sequence of monthly rainfall is from a uniform monthly sequence with unchanging monthly rainfall. The relative entropy attains the maximum value when the annual rainfall is concentrated in one single month, and equal to 0 for a uniform precipitation sequence (Feng et al. 2013; Pascale et al. 2016). Details of the calculation of relative entropy are in the Supplementary Information.

Regions that are wet year round have a low value of relative entropy. Pascale et al. (2016) defined 195 the 'Global Monsoon Domain' to be where $D_k > 0.4$ (and where the Dimensionless Seasonality 196 Index is >0.05, but this is mainly used to remove dry regions), but when examining changes in 197 the timing of the rainfall centroid Pascale et al. (2016) used a relative entropy threshold of 0.3 to 198 "exclude regions without a pronounced dry season". Here a threshold of 0.3 is used; a range of 199 thresholds were considered and it was decided that a D_k threshold of 0.3 classified regions such as 200 the Maritime Continent, Central Africa and eastern North America, known to be wet year round, 201 correctly, while also classifying most of South America as having an annual regime. The aim was 202 to just exclude those points that are definitely wet year round (without a pronounced dry season), 203 as opposed to just including those with a well-defined monsoon, hence the threshold of 0.3 was 204 used. 205

The seasonality classification for CHIRPS (Figure 1) shows that the dry desert regions (including the Sahara and Gobi Deserts, and the Arabian Peninsula) are correctly identified as dry-year-round, Central Equatorial Africa and the Maritime Continent are correctly identified as wet year-round, and the Horn of Africa is categorised as biannual. Monsoon regions, such as India, East Asia and South America are correctly identified as experiencing an annual regime. Over Africa, the classification is similar to that produced by Herrmann and Mohr (2011). This classification gives
an overall picture – in regions of complex rainfall regimes, with multiple rainfall peaks separated
by a relatively drier period lasting only one to two months (e.g. Central America; Small et al.
2007) it may not capture this complex seasonal regime, and defines the region as having one wet
season per year.

The seasonality classification was calculated for each model and scenario; it was calculated 216 separately for the present and future periods. Figure 1 shows the seasonality mask for CHIRPS, 217 and the modal seasonality classification for CMIP6 historical (for CMIP5 and all scenarios see 218 Figure S1). For the most part the model seasonality classification shows good agreement with 219 CHIRPS with a few minor differences; southern Australia and eastern China are shown as having 220 one wet season in models whereas they are defined as wet year round in observations. In terms 221 of the future projections there are not large changes in the overall pattern of seasonality (<10% of 222 grid points change regime), and most changes are located on regime boundaries (Figure S1). The 223 notable changes are over southern Europe/middle East where we see changes from wet year round 224 to an annual regime, and over central Africa CMIP6 shows some grid points change from wet year 225 round to two wet seasons per year; presumably due to larger rainfall increases in the wetter seasons 226 than the drier seasons. 227

228 2) WET AND DRY SEASON DISTINCTION

The method of anomalous accumulation was used to determine start and end dates of the climatological wet and dry seasons. This method is suitable for application at a global scale as it does not use a set precipitation threshold, and therefore can be applied to regions with very different rainfall amounts (Liebmann et al. 2012; Dunning et al. 2016). It is applicable to regions with both one and two wet seasons per year (Dunning et al. 2016). Full details of the calculation
can be found in the Supplementary Information.

Onset and cessation dates were calculated for each model and scenario; thus accounting for changing seasonal timing under future climate change. Historical patterns of onset and cessation in the CMIP5 and CMIP6 models were similar to those produced using observations, with a wet season during the boreal summer in the northern hemisphere and a wet season during the austral summer in the southern hemisphere (see Figure S2-6 in the Supplementary Information); Dunning et al. (2017) showed that CMIP5 models generally captured the correct patterns of onset and cessation across Africa, when assessed using this methodology.

242 3) WET AND DRY SPELL METRICS

A range of metrics were calculated for wet and dry seasons. The wet season was defined to be 243 the period after the onset and before the cessation date, while the dry season was defined to be the 244 period between the cessation date and the onset date. For the metrics calculated here, a threshold 245 of 1 mm day⁻¹ was used to define a wet/dry day. The well-established Climdex indices also use 246 a threshold of 1 mm day⁻¹ to define wet/dry days for the calculation of CWD and CDD, used in 247 Seneviratne et al. (2012) and Sillmann et al. (2013b). Other studies looking at wet and dry days and spells also use a threshold of 1 mm day⁻¹ (Polade et al. 2014; Giorgi et al. 2014; Funk et al. 249 2019; Giorgi et al. 2019), and this is confirmed by analysis of cumulative frequency of daily rainfall 250 (Figure S7). 251

²⁵² The following metrics were calculated:

• Wet/Dry Season Rainfall: Sum of daily rainfall over the wet/dry season.

• Wet/Dry Season Rainy Days: The number of days with rainfall greater than 1 mm day⁻¹ in the wet/dry season.

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• Wet/Dry Season Rain per Rainy Day: The mean rainfall over all days with rainfall greater than 1 mm day⁻¹ in the wet/dry season.

• Maximum wet/dry spell length in wet/dry seasons: The length (in days) of the longest wet/dry spell in each wet/dry season.

• Mean wet/dry spell length in wet/dry seasons: The mean length (in days) of all wet/dry spells (of length two days or more) in the wet/dry season.

• Wet/Dry Season Mean Maximum Temperature: The mean of the daily maximum temperature over the wet/dry season.

Metrics were computed for one wet season for the annual regime region and for two wet seasons for the biannual regime regions. For the wet-year-round regions the metrics were calculated for each calendar year (wet season was assumed to last all year); these points have a wet season but no dry season, so are included in wet season plots but not dry season plots. The median was taken over all seasons and the historical and future periods compared. For the maps, the change was regridded onto the lowest resolution model using nearest-neighbour remapping.

Figure 2 and Figure S8 show that the historical simulations from the CMIP5 and CMIP6 models 270 capture the observed distribution of wet and dry spells in wet and dry seasons as found in CHIRPS. 271 The main difference is the mean length of wet spells in the wet season where the CMIP6 models 272 contain much longer wet spells than the observations. Similarly, Sillmann et al. (2013a) found good 273 model representation of CDD in the CMIP5 models, but found that CWD was overestimated in 274 both models and reanalyses. Results from a second observational dataset (Multi-Source Weighted-275 Ensemble Precipitation (MSWEP, Beck et al. 2019); Figure S9) are very similar to those shown in 276 Figure 2 for CHIRPS. The mean length of wet spells in the wet season is longer in MSWEP over 277 parts of South America, Ethiopia and the Himalayas in MSWEP, presumably due to the inclusion 278

of reanalysis data, but the mean lengths in CMIP6 still exhibit a positive bias over these regions
when compared with MSWEP.

281 4) REGIONS

To explore changes over certain regions a regional classification was required. The regions used here are those defined for the 6th IPCC report (Iturbide et al. 2020). The locations and regions used are shown in Figures 6a, 8a and 10a.

285 5) TIMESERIES ANALYSIS

Timeseries were produced for certain regions to examine the changes over the whole period (1985-2099). The regions for timeseries were selected over areas of particular interest, where significant changes in wet/dry spell length had been identified. For the timeseries it is assumed that all grid points have an annual regime; for the three regions used (NES, SEAF and WAF) this is a reasonable assumption (Figure 1).

When calculating the timeseries of wet/dry spell lengths over the IPCC regions two methodologies 291 were employed. The first methodology defined a climatological onset and cessation date using the 292 full 115 years of the timeseries and then used the same onset and cessation date for each year, thus 293 not accounting for changes in seasonal timing. The wet and dry spell lengths were then calculated 294 for each year, in the same manner as above. The second methodology used a moving 30-year 295 period to calculate the onset and cessation dates, with the first 15 and last 15 years of the full 296 record set to the first-30-years value and the last-30-years value. Thus this second methodology 297 takes into account changes in seasonal timing. In section 3, timeseries produced using the second 298 methodology were presented. For the most part the differences between the two methodologies 299 were similar.

301 3. Results and Discussion

Results are presented here for wet and dry seasons; section 3a on seasonal timing presents changes in timing of the onset and cessation of the wet season, while sections 3b-3e present changes in rainfall and temperature metrics for wet and dry seasons separately. Figure S2 in the Supplementary Information indicates the timing of the wet season(s) in the calendar year for reference.

306 a. Seasonal Timing

The multi-model median change in onset and cessation of the annual wet season in CMIP6 under 307 the SSP585 scenario, shown in Figure 3, shows that onset is projected to get later across South 308 America, Southern Africa, West Africa and the Sahel, with changes of up to two weeks. Over 309 South America this continues the recent trend of delayed onset found by Correa et al. (2020). Over 310 East Asia Figure 3 shows the onset is projected to be earlier, with advances of over a week in some 311 locations. Changes in cessation are smaller than changes in onset; Figure 3b shows later cessation 312 by up to a week over the Sahel, and parts of Asia, with the largest changes of around two weeks 313 over central America. These results show good agreement with previous studies of changing onset 314 over Africa (Dunning et al. 2018), and over Asia, with Kitoh et al. (2013) reporting a later retreat 315 over South Asia and Ha et al. (2020) reporting an earlier onset over East Asia. The results also 316 show good agreement with Pascale et al. (2016), who analysed changes in the timing of the rainfall 317 centroid, and found a delay over Africa, Central and South America and parts of South Asia, with 318 an advance over East Asia. More broadly, a number of studies have found a global phase delay 319 of the seasonal cycle of precipitation (Biasutti and Sobel 2009; Dwyer et al. 2014; Marvel et al. 320 2017) and a redistribution of rainfall from early to late in the monsoon season (Seth et al. 2013), in 321 agreement with the generally later onset/cessation shown in Figure 3. A number of mechanisms 322 have been proposed for this; Dwyer et al. (2014) linked the delay to changes in the seasonality of the circulation, Seth et al. (2013) linked the delay to an enhanced spring convective barrier, while Song et al. (2018) linked the delay to changes in cross-equatorial energy transports and change in the strength of the subtropical highs.

327 b. Seasonal Rainfall Characteristics

Firstly, changes in seasonal rainfall totals, seasonal number of rainy days and seasonal rain per 328 rainy day were explored; Figure 4 shows the changes from CMIP6 under SSP585. South and East 329 Asia, Eastern Africa and Eastern North America all show projected increases in wet season rainfall 330 (Figure 4a); due to a combination of increasing rainfall intensity in the wet season (Figure 4e) and 331 either limited change or a slight increase in the number of wet days in the wet season (Figure 4c). 332 Wang et al. (2021), Wang et al. (2020) and Collins et al. (2013) also show an increase in boreal 333 summer rainfall over South and East Asia; Wang et al. (2020) show increases of around 1 mm 334 day⁻¹ in JJAS over South and East Asia. The Mediterranean, Central America and northern South 335 America exhibit significant decreases in wet season rainfall (of 5-25%, Figure 4a); these regions 336 also experience a significant decrease in the number of rainy days in the wet season (decrease of 337 over 15 days, Figure 4c), and smaller increases in rain per rainy day (Figure 4e). The asymmetry 338 between the American and Asian-African monsoon systems, with the North American monsoon 339 getting drier and the Asian-African monsoon getting wetter, identified in other studies (e.g. He 340 et al. 2020), is apparent in Figure 4. Other regions, including Southern Africa, the Maritime 341 Continent, and parts of southern Europe, also exhibit a decrease in the number of wet season 342 rainy days (Figure 4c). However, a significant decrease in rainfall is not found in these regions 343 (Figure 4a) as the decrease in the number of rainy days is compensated by an increase in rain per 344 rainy day (Figure 4e), broadly consistent with expectations from joint thermodynamic and energetic 345 constraints (Allen and Ingram 2002; Trenberth 2011).

Considering the dry season, Central America, South America, Southern Africa and Australia 347 all show decreases in dry season rainfall (Figure 4b), decreases in the number of wet days in the 348 dry season (Figure 4d) and little change in dry season rain per rainy day (Figure 4f). A decrease 349 in boreal summer rainfall over Southern Africa is also found by Wang et al. (2021), Wang et al. 350 (2020), and Collins et al. (2013), and a number of studies have identified an increasing contrast 351 between wet and dry seasons, with wet seasons getting wetter and dry seasons getting drier (Chou 352 et al. 2013; Kumar et al. 2015; Polson and Hegerl 2017; Lan et al. 2019; Deng et al. 2020; Schurer 353 et al. 2020). Eastern Africa, the Eastern Sahel and East Asia all show increases in dry season 354 rainfall (Figure 4b), increases in the number of dry season rainy days (Figure 4d) and increases in 355 dry season rain per rainy day (Figure 4f). Rainfall increases over East Asia are in good agreement 356 with Wang et al. (2021), Wang et al. (2020), and Collins et al. (2013) who identified increased 357 boreal winter rainfall in this region. In general, the increases in dry season rain per rainy day are 358 less than the increases in wet season rain per rainy day. Changes in seasonal rainfall are shown as 359 a percentage of the wet/dry season rainfall; thus while changes in dry season rainfall may appear 360 larger, in absolute terms the changes in wet season rainfall are larger. Central and northern South 361 America show rainfall decreases in both seasons, in agreement with Wang et al. (2020) who also 362 show precipitation decreases over the region in boreal summer and winter, although in winter they 363 find increases over the east and west coast, although it should be noted that they use a month-based 364 seasonal definition, and the increases are very close to the coast, which may have been removed in 365 the land-sea mask applied here. 366

³⁶⁷ Changes from CMIP6 under SSP245 and from CMIP5 under RCP8.5 (Figure S10-11) were ³⁶⁸ generally of the same sign of those shown in Figure 4, but of lower magnitude, possibly related to ³⁶⁹ the higher climate sensitivity in CMIP6 than in CMIP5 (Meehl et al. 2020). Over northern South ³⁷⁰ America decreases in dry season rainfall were ubiquitous, but decreases in wet season rainfall were

only found in CMIP6 under SSP585. Although they all show a decrease in the number of wet 371 season rainy days over northern South America, CMIP5 RCP8.5 shows a larger increase in wet 372 season rain per rainy day, and CMIP6 SSP245 has a smaller decrease in the number of rainy days, 373 but similar change in rain per rainy day to CMIP6 SSP585. Increases in wet season rainfall over 374 Eastern Africa and South and East Asia, and increases in dry season rainfall over East Africa and 375 East Asia, were found to be robust across the three scenarios and CMIP eras. Decreases in dry 376 season rainfall over Southern Africa were present in all three scenarios and CMIP eras, but were 377 only significant under SSP585 and RCP8.5. 378

379 c. Wet and Dry Spells

Figure 5 shows the projected changes in the mean length of wet and dry spells in wet and dry seasons from CMIP6 under SSP585; the changes in maximum spell length are generally of the same sign and of larger magnitude (see Figure S14).

During the dry seasons, projections indicate longer dry spells over Central America, north-east 383 South America, Southern Africa and Australia, with increases of up to 10 days (Figure 5a). This 384 is consistent with Figure 4 which showed reductions in dry season rainfall and fewer dry season 385 rainy days over these regions. Shorter dry spells in the dry season are found over East Africa and 386 parts of the Sahel, and East Asia (Figure 5a); these regions showed increases in dry season rainfall 387 and more rainy days in the dry season (Figure 4). Orlowsky and Seneviratne (2012) also show 388 increases in CDD over Southern Africa in JJA (local dry season). Over northern South America, 389 Orlowsky and Seneviratne (2012) show increases in CDD in different regions in all four seasons, 390 with the smallest increases in DJF (the peak of the wet season), thus agreeing on the projection 391 of longer dry spells in the dry season, but also demonstrating the impact of different definitions 392

³⁹³ of seasonality. Sillmann et al. (2013b) also show large increases in CDD over north-east South ³⁹⁴ America, Southern Africa, Australia and central America.

For the mean length of dry spells in the wet season (Figure 5b) the changes are generally 395 very small (multi-model median is generally less than half a day) and not statistically significant. 396 Only Southern Europe and the northern Mediterranean shows a statistically significant increase in 397 dry spell length in the wet season; the signal is also seen over the same region for dry seasons 398 (Figure 5a). Parts of the Mediterranean are classified as having one wet season per year, while 399 other parts are wet year round, thus Figure 5a-b indicates a general increase in dry spell length 400 in this region. Other studies, including Orlowsky and Seneviratne (2012) and Seneviratne et al. 401 (2012) show increases in CDD in the Mediterranean; and Brogli et al. (2019) show year-round 402 decreases in Mediterranean rainfall. 403

The changes in the mean length of wet spells are generally of smaller magnitude than the changes 404 in the mean length of dry spells (Figure 5c-d). The only statistically significant change in the mean 405 length of wet spells in the dry season is over northern South America, which shows a decrease 406 in the mean length of wet spells in the dry season; but the magnitude of the change is small 407 (Figure 5c). Only two regions show significant changes in the mean length of wet spells in the 408 wet season (Figure 5d); Central and northern South America and West/Central Africa both show 409 decreases of less than five days. These are the regions with the longest wet spells during the wet 410 season in current climate (Figure 2), and therefore are likely to exhibit the largest changes. These 411 regions also show fewer wet season rainy days under future climate change (Figure 4c), consistent 412 with shorter wet spells. 413

Again, changes from CMIP6 under SSP245 and from CMIP5 under RCP8.5 (Figure S12-13,S15-16) were generally of the same sign of those shown in Figure 5, but of lower magnitude. For the mean length of dry spells in the dry season, increases were present across Central America, north-

east South America, Southern Africa and Australia in all three scenarios and CMIP eras, but the 417 magnitude and statistical significance was reduced, particularly under SSP245 where only Southern 418 Africa and small parts of South America showed a statistically significant increase. For the mean 419 length of wet spells in the wet season, the signal was similar in CMIP5 RCP8.5 to CMIP6 SSP585, 420 but much reduced in CMIP6 SSP245, with only the decrease over northern South America present. 421 Figure 3 showed changes in the length of wet and dry seasons under future climate change, which 422 may affect the length of wet and dry spells. Therefore the analysis was repeated, calculating wet and 423 dry spell lengths as a percentage of the wet/dry season length; results were similar demonstrating 424 that the changes found here are not a consequence of changing season lengths. 425

Overall, Figure 5a-b indicates that the significant increases in dry spell length expected under 426 future climate change are projected to occur during the dry season not the wet season. Only parts 427 of the Mediterranean and southern Europe, where some locations experience rainfall year round, 428 show a significant increase in dry spell length in the wet season. Therefore agriculture grown solely 429 during the wet season may be less affected by longer dry spells. However, Figures 4-5 indicate 430 longer dry spells, fewer rainy days, and less rainfall during the dry season in certain regions, which 431 may make the dry seasons more intense, negatively impacting perennial crops and crops grown 432 during the dry season. 433

434 d. Regional Changes

Figures 4-5 have presented changes across the tropics; three regions are now considered in more detail. Some of the largest changes in wet and dry spell lengths identified in Figures 4-5 were for northern South America and Africa; hence these regions are explored further. South and East Asia is a region of high population density, highly dependent on agriculture, so this region has also been considered in more detail.

440 1) NORTHERN SOUTH AMERICA

Decreases in precipitation over northern South America under future climate change have been 441 widely reported (Scheff and Frierson 2012; Wang et al. 2020), and linked to changes in the tropical 442 Pacific (Parsons 2020) and the response of vegetation to elevated CO_2 (Richardson et al. 2018). 443 Recent declines in dry season precipitation over tropical South America can be attributed to 444 elevated greenhouse gases and land use change (Barkhordarian et al. 2018), while recent increased 445 frequency of dry days over the southern Amazon has been linked to tropical North Atlantic warming 446 and enhanced subsidence of the southern branch of the Hadley cell (Espinoza et al. 2019). Three 447 of the IPCC AR6 regions are considered; South American Monsoon (SAM), north South America 448 (NSA) and north-east South America (NES). Figure 6b-c shows there is generally good model 449 agreement in the projection of longer dry spells in the dry season across all three regions; for all 450 scenarios the 25th percentile (across the models) is above 0. Under the high emissions scenarios 451 for CMIP5 and CMIP6 the median increase in the maximum length of dry spells in the dry season 452 ranges between 9 and 26 days, with median increases in the mean length of dry spells in the dry 453 season between 1.4 and 9.4 days. In their analysis of changes in CWD and CDD over Brazil in the 454 CMIP5 models, Avila-Diaz et al. (2020) also found longer dry spells under future climate change, 455 with the largest increases in the north-east. This is a continuation of the recent trend of longer dry 456 spells in north-east Brazil over 1980-2016 (Avila-Diaz et al. 2020). 457

For the mean and maximum length of wet spells in the wet season the 25th percentile is below 0 for all regions and scenarios indicating strong model agreement on shorter wet spells in the wet season under future climate change (Figure 6d-e). Under the high emissions scenarios for CMIP5 and CMIP6 median decreases in the maximum length of wet spells in the wet season range between 3 and 13 days, while median decreases in the mean length of wet spells in the wet season range

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⁴⁶³ between 0.6 and 3.7 days. It is interesting to note that the largest changes in dry spell lengths
⁴⁶⁴ are over NES, while the largest changes in wet spell lengths are over NSA; this is consistent with
⁴⁶⁵ Avila-Diaz et al. (2020) who found the largest decreases in CWD over the central northern part of
⁴⁶⁶ Brazil.

For both variables and all the regions the changes are larger for the CMIP6 high emissions scenario than the CMIP5 high emissions scenario; this was also noted in the previous section and is thought to be related to the higher climate sensitivity in CMIP6 than in CMIP5 (Meehl et al. 2020).

Timeseries of mean length of dry spells in the dry season and mean length of wet spells in the 471 wet season for the NES region (Figure 7) show the same patterns of longer dry spells in the dry 472 season and shorter wet spells in the wet season. The dots at the top of each panel indicate when 473 the range of values in that year is statistically significantly different from the range over 1985-2014 474 and indicates the emergence of the signal from the noise. This suggests that the increase in the 475 mean length of dry spells in the dry season becomes statistically significantly different between 476 2020 and 2040 (depending on the emissions scenario) whereas the decrease in the mean length of 477 wet spells in the wet season is not statistically significant until 2060-2080 (timings are similar for 478 maximum spell lengths). These times of emergence are indicative, but may not be fully quantified, 479 as we have only used one ensemble member per model. For more robust times of emergence the 480 internal variability should be calculated using a large ensemble. For wet spells in the wet season 481 the change is less significant under the SSP245 scenario and diverges from the SSP585 scenario in 482 the 2060s (Figure 7b). The apparent jump in dry spell length around 2065 is not a model artefact, 483 but reflects the combination of internal variability and the levelling off of warming in the SSP245 484 scenario. Figure 7 also shows the large interannual variability in these metrics and the large model 485 spread. Two methodologies were used for calculating the timeseries; one using varying season 486

start and end dates and one using fixed seasons start and end dates (see methodology section 5). 487 The results shown in Figure 7 were the same for both methodologies, indicating the result is robust 488 and insensitive to changing season length. However, for the mean length of dry spells in the wet 489 season and mean length of wet spells in the dry season the signal (increase/decrease respectively) 490 becomes weaker and less statistically significant when varying onset is used (not shown). This 491 indicates that erroneously including parts of the other season when the time-invariant dates are 492 used can influence the calculation of these metrics, and indicates changes occurring in the margins 493 between the seasons. 494

495 2) AFRICA

Projected changes in the mean length of wet and dry spells are not consistent across the African 496 continent (Figure 8, Figure S17). While south-west Africa (SWAF) and south-east Africa (SEAF) 497 show longer dry spells in the dry season (increase in mean and maximum length), with median 498 increases in the maximum length greater than 20 days over SWAF, north-east Africa (NEAF) shows 499 shorter dry spells (median change in the maximum length of dry spells is a decrease of 12 days 500 under CMIP6 SSP585, Figure 8b-c). There is good model agreement in shorter dry season dry 501 spells over NEAF, with the 25th percentile below 0 for all scenarios. Pinto et al. (2016) also found 502 increases in CDD over Southern Africa, with increases of greater than 25 days over Zimbabwe and 503 Botswana, with larger increases under RCP8.5 than under RCP4.5, in agreement with Orlowsky 504 and Seneviratne (2012) and Sillmann et al. (2013b). Over East Africa previous studies report both 505 increases and decreases in dry spell lengths. Over southern Ethiopia, Tegegne et al. (2020) found a 506 decrease in CDD during the 2080s under RCP8.5, in agreement with the findings in Sillmann et al. 507 (2013b). However, Gudoshava et al. (2020) show only small changes (generally less than one day) 508 in CDD in March-May, June-September and October-December over the Horn of Africa; Ogega 509

et al. (2020) also found no large changes in CDD over Kenya and Uganda. Conversely, Osima 510 et al. (2018) find increasing CDD over the Horn of Africa. Our results here are consistent with 511 the studies of Sillmann et al. (2013b), Orlowsky and Seneviratne (2012) and Tegegne et al. (2020); 512 further detailed studies on this region should explore the range of projections from different studies, 513 taking into account the large variability in model projections over East Africa (Rowell et al. 2015). 514 West Africa (WAF) and Central Africa (CAF) show shorter dry spells in the future (Figure S17), 515 although the changes are smaller than those over NEAF and model agreement is weaker; Sillmann 516 et al. (2013b) show that changes in CDD are not significant over these regions. 517

Only two regions show notable changes in the length of wet spells in the wet season; WAF and 518 CAF show median decreases of 3-6 days in the maximum length of wet spells (under high emissions 519 scenarios, Figure 8d-e). Changes in the mean length also show decreases. While there is generally 520 good model agreement on shorter wet spells in the wet season over WAF and CAF (for the most part 521 the 25th percentile is below 0), there is a large range across the models, particularly under CMIP5 522 and over CAF. This may be related to poor model representation of the mean climate over this 523 region in CMIP5 models (Creese and Washington 2018) and to the inclusion of a wet-year-round 524 region in equatorial Africa. Klutse et al. (2018) also found a reduction in CWD in June-September 525 over West Africa, with reductions of up to four days over Nigeria and Guinea under 2°C of global 526 warming. Over central Africa Mba et al. (2018) found decreases in CWD in both March-May and 527 September-November. 528

Figure 9b shows that the decrease in the length of wet spells in the wet season over WAF becomes statistically significantly different from 1985-2014 at around 2050 under SSP585. However, under SSP245 the change does not become consistently statistically significantly different in the period of this timeseries (up to 2100). For dry spells in the dry season over SEAF the increase in length becomes statistically significantly different from 1985-2014 around 2040-2060, later than was ⁵³⁴ found for NES (Figure 9a). For dry spells in the dry season the scenarios diverge in the 2060s, ⁵³⁵ while for wet spells in the wet season the scenarios diverge in the 2040s.

536 3) ASIA

Generally, changes in the length of wet and dry spells are smaller and less significant, or exhibit 537 limited model consensus, over Asia. Figure 5 showed significant decreases in the mean length of 538 dry spells in the dry season over north-east China (also found by Han et al. 2018; Dong et al. 539 2020), but no other significant changes over this region. Figure 10b-c confirms that changes in 540 the length of dry spells in the dry season are generally small. The largest change is an increase of 541 four days in the maximum length of dry season dry spells over South Asia (SAS) under CMIP5 542 RCP8.5; this change is larger in CMIP5 than in CMIP6 (median in CMIP6 SSP585 is an increase 543 of one day), consistent with the slight increase in dry season rainfall over India found in CMIP6 544 SSP585, but little change in dry season rainfall found in CMIP5 RCP8.5. While there is an increase 545 in the maximum length of dry spells, the change in the mean length of dry spells is small (less 546 than 0.6 days), indicating that for the most part dry season dry spells are not getting longer. An 547 increase in the maximum, but not the mean also suggests an increase in the variability. This 548 highlights the importance of considering the mean as well as the maximum spell length. There is 549 limited model agreement in these changes. Han et al. (2018) also found little change in CDD over 550 South Asia and South East Asia in their analysis of CMIP5 models. Aadhar and Mishra (2020) 551 concluded that projections based on CMIP6 multi-model ensembles are not reliable over South 552 Asia as projections are different when the whole ensemble is used compared with just those models 553 that correctly represent current climate. Further work is required to investigate this region in more 554 detail. 555

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Figure 10b-c also shows a slight increase in the maximum length of dry season dry spells over south-east Asia (SEA). It should be noted that much of this region experiences a wet-year-round rainfall regime, therefore only a small region (over Myanmar and Thailand) will be contributing to this signal; this is supported by the findings of Ge et al. (2019), who show the largest increases in CDD in the SEA region are over the Myanmar/Thailand peninsula. Figure 10b-c shows the median changes are small (two and three days), and again, only seen in the maximum not the mean, with limited model agreement.

Figure 10d-e shows changes in the length of wet spells in the wet season, with median decreases 563 in the maximum length of two and four days over SEA and little change in other regions. This 564 is consistent with a decrease in the number of wet season rainy days over this region. Again, the 565 change in the mean is small (less than 0.4 days), indicating only the longest wet spells are getting 566 shorter, not all wet spells, and also indicating a change in the variability. While Ge et al. (2019) 567 show increases in CWD in the CORDEX models over SEA, they also find that the increase is 568 larger under 1.5°C of global warming than under 2°C of global warming, and under 2°C of global 569 warming some regions show decreases (e.g. Borneo). 2°C of global warming is reached at around 570 2040 under SSP585 (O'Neill et al. 2016); therefore if the increased warming continues to decrease 571 the change in CWD, this is consistent with our finding of shorter wet spells. Over East Asia (EAS) 572 the changes are small, consistent with the results in Figure 5. 573

574 e. Seasonal Maximum Temperature

Interpreting agricultural implications of precipitation changes cannot be considered in isolation from temperature, as the aridity or dryness over a region is determined by the difference between supply (precipitation) and demand (measured by potential evapotranspiration (PET)) (Dai et al. 2018). Elevated greenhouse gas levels lead to increasing temperatures, which combined with small

changes in relative humidity, lead to large increases in the vapor pressure deficit and widespread 579 increases in PET (Dai et al. 2018). Zhao and Dai (2015) found that rising temperatures and vapor 580 deficits explain most of the projected ubiquitous global PET increase. On the other hand, elevated 581 CO_2 concentrations modify how vegetation responds to changing water availability through a 582 complex and uncertain combination of increased plant growth and increased water use efficiency 583 (Mankin et al. 2019; Peters et al. 2018; Kooperman et al. 2018). Furthermore, rainfall changes exert 584 a substantial influence on temperature changes since less cloud and rainfall increases absorbed solar 585 radiation while drier soils increase the sensible heat fraction. Therefore temperature responses in 586 wet and dry seasons are not expected to be uniform and are thus of considerable additional interest. 587 Figure 11 shows increasing temperatures across the tropics in both wet and dry seasons, as 588 projected under future climate change. Over northern South America, Southern Africa, and parts 589 of Asia, the increase in dry season maximum temperature is up to 3° C greater than the increase 590 in wet season maximum temperature (Figure 11a-b); Zhou et al. (2015) also found the largest 591 temperature increases over drier regions. In both northern South America and Southern Africa 592 this increase in dry season maximum temperature is combined with a decrease in dry season 593 rainfall (Figure 4b), which may lead to a substantial drying of soils during the dry season, affecting 594 perennial crops. Dai et al. (2018) found large increases in PET and decreases in soil moisture 595 over northern South America and southern Africa. Decreases in soil moisture across northern 596 South America and southern Africa were also reported by Collins et al. (2013) and Zhao and Dai 597 (2015). Furthermore, Zhao and Dai (2015) found decreases in soil moisture (hence increased 598 surface drying) in regions projected to receive higher precipitation amounts, due to fewer wet days 599 and longer dry spells and high temperatures, highlighting the importance of changes in frequency 600 in precipitation, in addition to changes in amount. 601

The contours on Figure 11 indicate that under the historical simulation over 1985-2014 only 602 northern Australia and the Sahel have mean seasonal maximum temperatures greater than 35°C 603 (dashed contour), whereas by the end of the 21st century under SSP585 parts of India, the Sahel, 604 northern South America and parts of northern Australia may all experience mean dry season 605 maximum temperatures of greater than 35°C (solid contour). This may have negative impacts on 606 crops including sorghum (FAO 2020). However, during the wet season temperatures are generally 607 less than 35°C, except over northern Australia and parts of the Sahel. This may mean that crops 608 grown during the wet season are less impacted by these elevated temperatures. 609

4. Conclusions

⁶¹¹ Changes in the mean and maximum lengths of wet and dry spells (and mean maximum tempera-⁶¹² ture) have been calculated separately for the climatological wet and dry seasons using an ensemble ⁶¹³ of CMIP5 and CMIP6 models across the global tropics and subtropics. An objective methodology ⁶¹⁴ is used to classify the seasonality at each location and to determine the timing of the climatological ⁶¹⁵ wet and dry season(s); metrics including total seasonal rainfall, number of rainy days, rain per ⁶¹⁶ rainy day and the mean and maximum length of wet and dry spells are then calculated for the wet ⁶¹⁷ and dry seasons separately.

⁶¹⁸ Our main results are (also summarised in Figure 12):

Onset of the annual wet season is projected to get later across South America, Southern Africa,
 West Africa and the Sahel, with changes of up to two weeks. Over East Asia the onset is
 projected to be earlier. Cessation of the annual wet season is projected to be later over the
 Sahel, parts of Asia, and central America. These changes are consistent with the projected
 global phase delay of the seasonal cycle of precipitation under future climate change (Dwyer
 et al. 2014; Song et al. 2018).

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Changes in the mean and maximum length of dry spells are largest during the dry season.
 Increasing dry season dry spell lengths were found over central America, north-east South
 America, Southern Africa and Australia. Shorter dry spells in the dry season are found over
 East Africa and parts of the Sahel, and East Asia. Changes in the length of dry spells in the
 wet season are generally small, with only the Mediterranean and southern Europe showing
 significant increases.

- Statistically significant decreases in the length of wet spells in the wet season are found over Central and northern South America and West/Central Africa.
- For the most part, changes were of the same sign in CMIP5 and CMIP6, but of larger magnitude under CMIP6, possibly related to the higher climatic sensitivity (Meehl et al. 2020).

The time at which the change becomes statistically significantly different from the historical period varies with metric and region; while the increase in length of dry spells in the dry season over north-east South America is statistically significant around 2020-2040, the decrease in the length of wet spells in the wet season over West Africa is not statistically significant until around 2050. Times of emergence here are indicative; more robust times of emergence could be calculated using a larger ensemble.

Mean maximum temperature increases are greater for the dry season than for the wet season,
 with mean maximum dry season temperatures increasing by up to 7°C, particularly over South
 America, Southern Africa and parts of Asia. While only northern Australia is projected to
 have large areas with mean maximum wet season temperatures >35°C at the end of the 21st
 century under SSP585, northern South America, the Sahel, and India are projected to have
 mean maximum dry season temperatures >35°C at the end of the 21st century under SSP585.

This paper aimed to address the question "Do changes in the length of wet and dry spells differ 647 in climatologically wet and dry seasons?". Here it has been decisively shown, considering both 648 the mean and maximum wet/dry spell lengths (thus considering all wet/dry spells, not just the 649 longest) that changes in the length of wet and dry spells are different in wet and dry seasons, and 650 that the increases in dry spell lengths expected under future climate change mostly occur during 651 the annual dry seasons. This has important implications for agriculture across the tropics, and 652 suggests that crops grown solely during the wet season may be less affected by longer dry spells 653 and increased water stress. However, for crops grown during the dry season and perennial crops, 654 longer dry spells in the dry season, combined with lower dry season rainfall and fewer dry season 655 rainy days may lead to elevated dry season water stress. This elevated dry season water stress 656 will be further exacerbated by increased evaporative demand due to higher temperatures; Scheff 657 and Frierson (2015) found aridification generally dominates over humidification in the tropics and 658 subtropics due to temperature-driven increases in potential evapotranspiration. Regions such as 659 central America, north-east South America, Southern Africa and Australia will be most impacted 660 by this. Further work, using detailed crop models, is required to determine the impacts on individual 661 crops. 662

Regional variations in the response of wet and dry spells to elevated greenhouse gas concen-663 trations have been identified; the aim of this paper is to assess the changes in wet/dry season 664 precipitation characteristics and not to understand the regional changes in terms of circulation 665 change. Regional changes in rainfall across the tropics are driven by a range of factors including 666 changes in the Intertropical Convergence Zone (contraction, intensification and timing; Dwyer 667 et al. 2014; Su et al. 2017; Song et al. 2018) and subtropical drying, which is determined by 668 complex interplay between Hadley Cell expansion (Seidel et al. 2008; Scheff and Frierson 2012), 669 land ocean warming contrasts (Karnauskas and Ummenhofer 2014; He and Soden 2017), and 670

vegetation responses to elevated CO₂ (Mankin et al. 2018). Further work is required to understand
 the changing regional/local circulations and energy and water fluxes that lead to these changes over
 different regions.

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⁶⁹¹ *Data availability statement*. CHIRPS rainfall data are available from https://data.chc. ⁶⁹² ucsb.edu/products/CHIRPS-2.0/. CMIP5 and CMIP6 data were obtained from the British ⁶⁹³ Atmospheric Data Centre (BADC) hosted at the Centre for Environmental Data Analysis (CEDA).

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FIG. 10. a) Region map showing the location of South Asia (SAS), East Asia (EAS), and South East Asia (SEA). b-c) Boxplots showing the change in mean/max length of dry spells in the dry season. d-e) Boxplots showing the change in mean/max length of wet spells in the wet season. The different colours indicate the CMIP era (CMIP5 or CMIP6) and scenario (SSP 245, SSP 585, RCP 4.5, RCP 8.5). For all boxplots the change is computed from the historical simulation for 1985-2014 to the select scenario for 2070-2099.



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FIG. 12. Schematic summarising the changes in wet/dry season rainfall and wet/dry spell lengths in wet/dry seasons found here; the top panel is for dry seasons and the bottom row is for wet seasons (including regions that are wet year round).

(Top) Longer dry spells and lower rainfall during the dry season are found over Central and South America and
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