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Future Changes in Wet and Dry Season Characteristics in CMIP5 and CMIP6 simulations

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ABSTRACT

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

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.

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

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

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

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

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

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

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

2. Data and Methodology

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 RCP4.5/RCP8.5 (CMIP5) or SSP245/SSP585 (CMIP6) simulations for periods at the end of the 20th and 21st centuries. For the present day climate, the historical simulation was used over the period 1 January 1985 - 31 December 2014, chosen to be in line with IPCC AR6. For the CMIP5 simulations the historical simulation is only available until 31 December 2005; CMIP5 RCP4.5 simulations were used for 1 January 2006 - 31 December 2014 (taken from the same model) and appended to the historical simulations for 1 January 1985 - 31 December 2005. Recent analysis suggests that CO_2 emissions have been above RCP4.5 levels over the recent period, so this gives a conservative estimate (Schwalm et al. 2020). Projections from RCP4.5 and RCP8.5 do not differ significantly over this 2005-2014 period, therefore the choice of scenario was assumed to have a minimal impact. Furthermore, the primary focus is on the newer CMIP6 models; results from the CMIP5 models are mainly presented for traceability with previous studies. The period 1 January 2070 - 31 December 2099 was used for the end of the 21st century period. Only the first ensemble members were used (r1i1p1/r1i1p1f1) – for two CMIP6 models r1i1p1f2 was used due to r1i1p1f1 being unavailable (see Table S2). Other studies have taken a similar approach. Daily maximum temperature data were also used for the same set of models and time periods, however, they were unavailable for a couple of the models at the time when the analysis was conducted; this is indicated in Table S2.

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

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*

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

1) SEASONALITY CLASSIFICATION

The seasonality of precipitation varies across the tropics from monsoonal regions with one summer wet season per year, to dry desert regions with little rainfall, to rainforest regions that are wet year round. East Africa is known to experience two wet seasons per year (Yang et al. 2015). To distinguish between wet and dry seasons, it is first important to determine the seasonal regime.

A number of different methodologies have been utilised for this purpose; Herrmann and Mohr (2011) used monthly precipitation and temperature to define rainfall seasonality regimes across Africa. Pascale et al. (2016) and Feng et al. (2013) used the Relative Entropy and Dimensionless Seasonality Index to define and examine changes in the 'Global Monsoon Domain'. Liebmann et al. (2012) and Dunning et al. (2016) use harmonic analysis to separate regions with one or two wet seasons per year across Africa. A combination of these metrics is used here.

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:

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.

2. Grid points with a relative entropy (a measure of how uniform or variable the seasonal sequence of monthly rainfall is, see below) less than 0.3 were defined as wet year round with 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

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

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

211 classification is similar to that produced by Herrmann and Mohr (2011). This classification gives
212 an overall picture – in regions of complex rainfall regimes, with multiple rainfall peaks separated
213 by a relatively drier period lasting only one to two months (e.g. Central America; Small et al.
214 2007) it may not capture this complex seasonal regime, and defines the region as having one wet
215 season per year.

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

228 2) WET AND DRY SEASON DISTINCTION

229 The method of anomalous accumulation was used to determine start and end dates of the
230 climatological wet and dry seasons. This method is suitable for application at a global scale as
231 it does not use a set precipitation threshold, and therefore can be applied to regions with very
232 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.

3) WET AND DRY SPELL METRICS

A range of metrics were calculated for wet and dry seasons. The wet season was defined to be the period after the onset and before the cessation date, while the dry season was defined to be the period between the cessation date and the onset date. For the metrics calculated here, a threshold of 1 mm day^{-1} was used to define a wet/dry day. The well-established Climdex indices also use a threshold of 1 mm day^{-1} to define wet/dry days for the calculation of CWD and CDD, used in 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. 2019; Giorgi et al. 2019), and this is confirmed by analysis of cumulative frequency of daily rainfall (Figure S7).

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.

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

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

281 4) REGIONS

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

285 5) TIMESERIES ANALYSIS

286 Timeseries were produced for certain regions to examine the changes over the whole period
287 (1985-2099). The regions for timeseries were selected over areas of particular interest, where
288 significant changes in wet/dry spell length had been identified. For the timeseries it is assumed
289 that all grid points have an annual regime; for the three regions used (NES, SEAF and WAF) this
290 is a reasonable assumption (Figure 1).

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

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.

a. Seasonal Timing

The multi-model median change in onset and cessation of the annual wet season in CMIP6 under the SSP585 scenario, shown in Figure 3, shows that onset is projected to get later across South America, Southern Africa, West Africa and the Sahel, with changes of up to two weeks. Over South America this continues the recent trend of delayed onset found by Correa et al. (2020). Over East Asia Figure 3 shows the onset is projected to be earlier, with advances of over a week in some locations. Changes in cessation are smaller than changes in onset; Figure 3b shows later cessation by up to a week over the Sahel, and parts of Asia, with the largest changes of around two weeks over central America. These results show good agreement with previous studies of changing onset over Africa (Dunning et al. 2018), and over Asia, with Kitoh et al. (2013) reporting a later retreat over South Asia and Ha et al. (2020) reporting an earlier onset over East Asia. The results also show good agreement with Pascale et al. (2016), who analysed changes in the timing of the rainfall centroid, and found a delay over Africa, Central and South America and parts of South Asia, with an advance over East Asia. More broadly, a number of studies have found a global phase delay of the seasonal cycle of precipitation (Biasutti and Sobel 2009; Dwyer et al. 2014; Marvel et al. 2017) and a redistribution of rainfall from early to late in the monsoon season (Seth et al. 2013), in agreement with the generally later onset/cessation shown in Figure 3. A number of mechanisms 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.

b. Seasonal Rainfall Characteristics

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

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

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

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

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.

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

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

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

417 east South America, Southern Africa and Australia in all three scenarios and CMIP eras, but the
418 magnitude and statistical significance was reduced, particularly under SSP245 where only Southern
419 Africa and small parts of South America showed a statistically significant increase. For the mean
420 length of wet spells in the wet season, the signal was similar in CMIP5 RCP8.5 to CMIP6 SSP585,
421 but much reduced in CMIP6 SSP245, with only the decrease over northern South America present.

422 Figure 3 showed changes in the length of wet and dry seasons under future climate change, which
423 may affect the length of wet and dry spells. Therefore the analysis was repeated, calculating wet and
424 dry spell lengths as a percentage of the wet/dry season length; results were similar demonstrating
425 that the changes found here are not a consequence of changing season lengths.

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

434 *d. Regional Changes*

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

1) NORTHERN SOUTH AMERICA

Decreases in precipitation over northern South America under future climate change have been widely reported (Scheff and Frierson 2012; Wang et al. 2020), and linked to changes in the tropical Pacific (Parsons 2020) and the response of vegetation to elevated CO₂ (Richardson et al. 2018). Recent declines in dry season precipitation over tropical South America can be attributed to elevated greenhouse gases and land use change (Barkhordarian et al. 2018), while recent increased frequency of dry days over the southern Amazon has been linked to tropical North Atlantic warming and enhanced subsidence of the southern branch of the Hadley cell (Espinoza et al. 2019). Three of the IPCC AR6 regions are considered; South American Monsoon (SAM), north South America (NSA) and north-east South America (NES). Figure 6b-c shows there is generally good model agreement in the projection of longer dry spells in the dry season across all three regions; for all scenarios the 25th percentile (across the models) is above 0. Under the high emissions scenarios for CMIP5 and CMIP6 the median increase in the maximum length of dry spells in the dry season ranges between 9 and 26 days, with median increases in the mean length of dry spells in the dry season between 1.4 and 9.4 days. In their analysis of changes in CWD and CDD over Brazil in the CMIP5 models, Avila-Diaz et al. (2020) also found longer dry spells under future climate change, with the largest increases in the north-east. This is a continuation of the recent trend of longer dry spells in north-east Brazil over 1980-2016 (Avila-Diaz et al. 2020).

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

between 0.6 and 3.7 days. It is interesting to note that the largest changes in dry spell lengths are over NES, while the largest changes in wet spell lengths are over NSA; this is consistent with Avila-Diaz et al. (2020) who found the largest decreases in CWD over the central northern part of 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 wet season for the NES region (Figure 7) show the same patterns of longer dry spells in the dry season and shorter wet spells in the wet season. The dots at the top of each panel indicate when the range of values in that year is statistically significantly different from the range over 1985-2014 and indicates the emergence of the signal from the noise. This suggests that the increase in the mean length of dry spells in the dry season becomes statistically significantly different between 2020 and 2040 (depending on the emissions scenario) whereas the decrease in the mean length of wet spells in the wet season is not statistically significant until 2060-2080 (timings are similar for maximum spell lengths). These times of emergence are indicative, but may not be fully quantified, as we have only used one ensemble member per model. For more robust times of emergence the internal variability should be calculated using a large ensemble. For wet spells in the wet season the change is less significant under the SSP245 scenario and diverges from the SSP585 scenario in the 2060s (Figure 7b). The apparent jump in dry spell length around 2065 is not a model artefact, but reflects the combination of internal variability and the levelling off of warming in the SSP245 scenario. Figure 7 also shows the large interannual variability in these metrics and the large model spread. Two methodologies were used for calculating the timeseries; one using varying season

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

495 2) AFRICA

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

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

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

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

found for NES (Figure 9a). For dry spells in the dry season the scenarios diverge in the 2060s, while for wet spells in the wet season the scenarios diverge in the 2040s.

3) ASIA

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

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

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

579 changes in relative humidity, lead to large increases in the vapor pressure deficit and widespread
580 increases in PET (Dai et al. 2018). Zhao and Dai (2015) found that rising temperatures and vapor
581 deficits explain most of the projected ubiquitous global PET increase. On the other hand, elevated
582 CO₂ concentrations modify how vegetation responds to changing water availability through a
583 complex and uncertain combination of increased plant growth and increased water use efficiency
584 (Mankin et al. 2019; Peters et al. 2018; Kooperman et al. 2018). Furthermore, rainfall changes exert
585 a substantial influence on temperature changes since less cloud and rainfall increases absorbed solar
586 radiation while drier soils increase the sensible heat fraction. Therefore temperature responses in
587 wet and dry seasons are not expected to be uniform and are thus of considerable additional interest.

588 Figure 11 shows increasing temperatures across the tropics in both wet and dry seasons, as
589 projected under future climate change. Over northern South America, Southern Africa, and parts
590 of Asia, the increase in dry season maximum temperature is up to 3°C greater than the increase
591 in wet season maximum temperature (Figure 11a-b); Zhou et al. (2015) also found the largest
592 temperature increases over drier regions. In both northern South America and Southern Africa
593 this increase in dry season maximum temperature is combined with a decrease in dry season
594 rainfall (Figure 4b), which may lead to a substantial drying of soils during the dry season, affecting
595 perennial crops. Dai et al. (2018) found large increases in PET and decreases in soil moisture
596 over northern South America and southern Africa. Decreases in soil moisture across northern
597 South America and southern Africa were also reported by Collins et al. (2013) and Zhao and Dai
598 (2015). Furthermore, Zhao and Dai (2015) found decreases in soil moisture (hence increased
599 surface drying) in regions projected to receive higher precipitation amounts, due to fewer wet days
600 and longer dry spells and high temperatures, highlighting the importance of changes in frequency
601 in precipitation, in addition to changes in amount.

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

4. Conclusions

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

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).

- 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.

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

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

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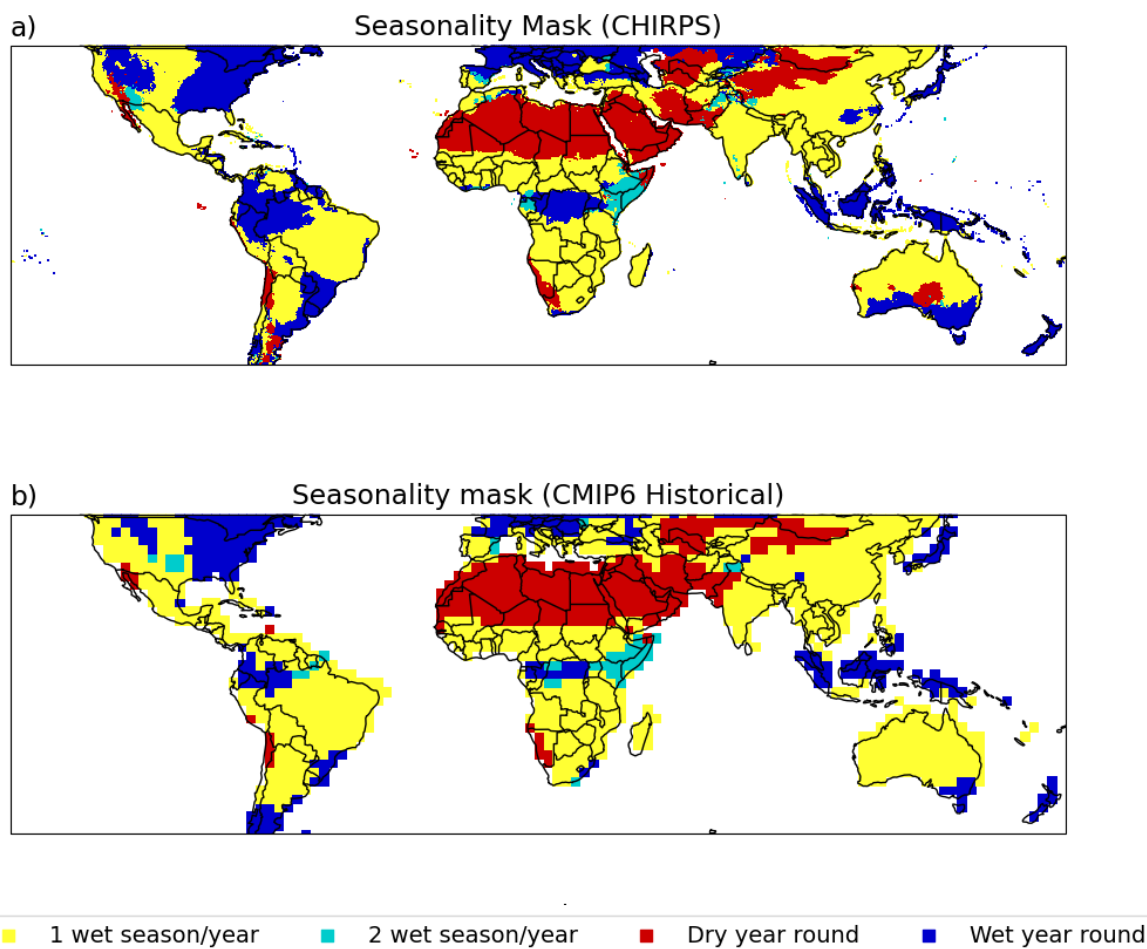


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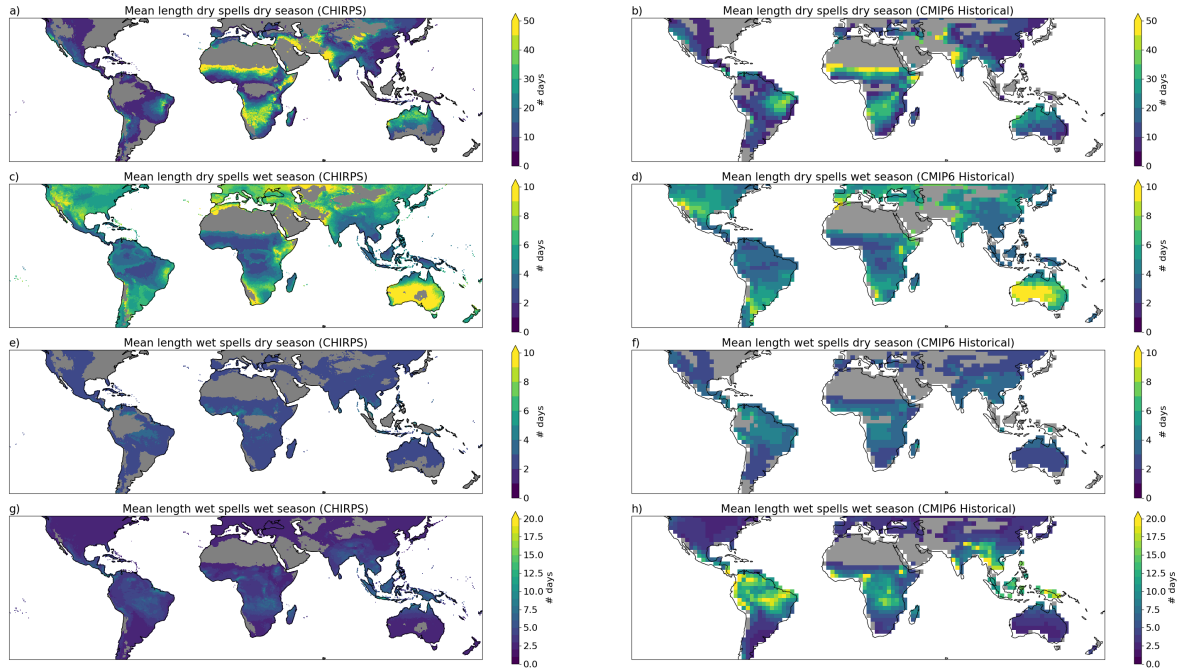


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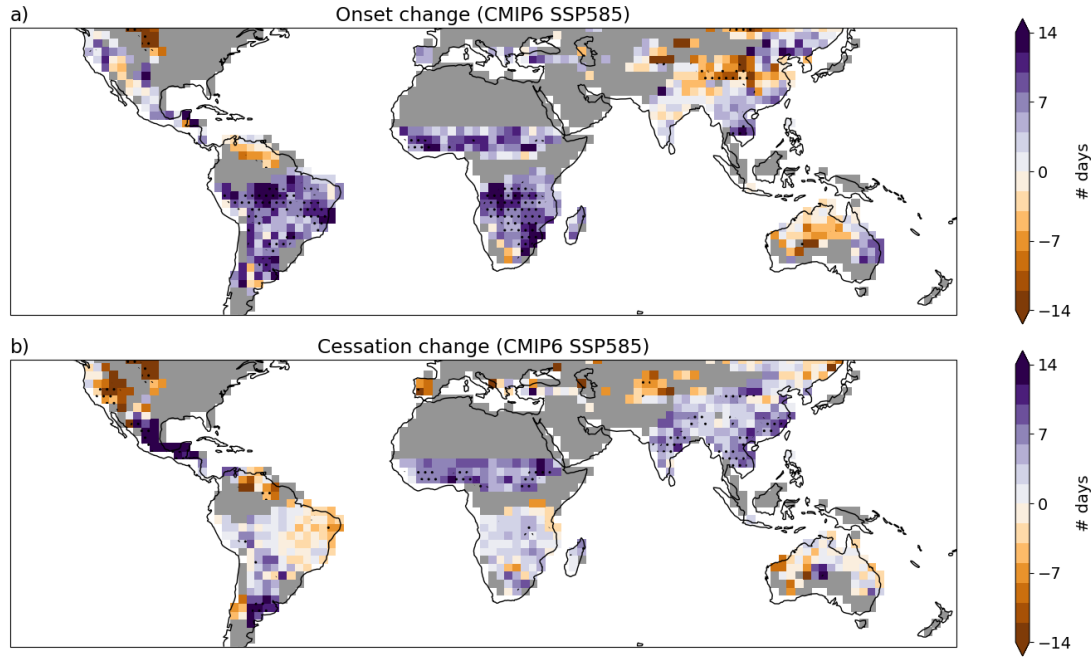


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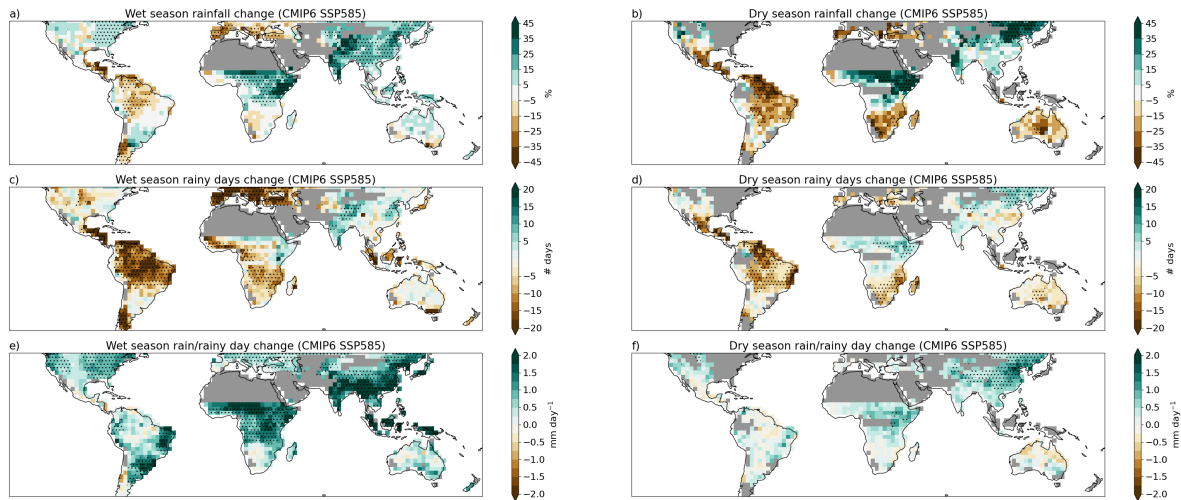


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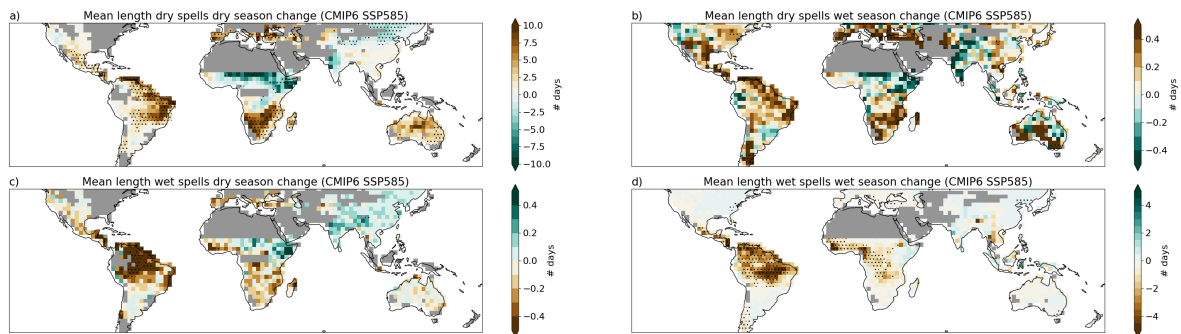


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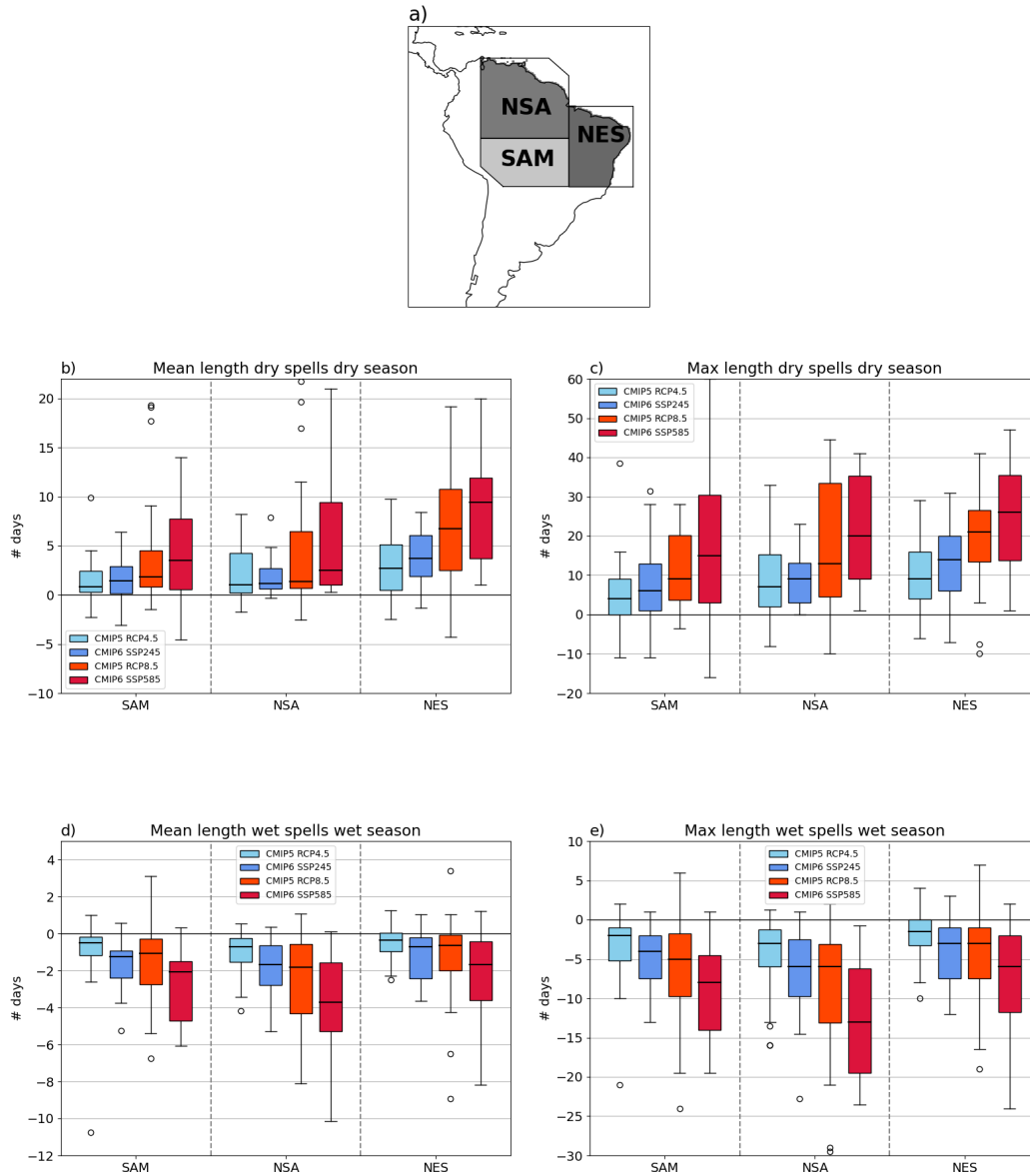


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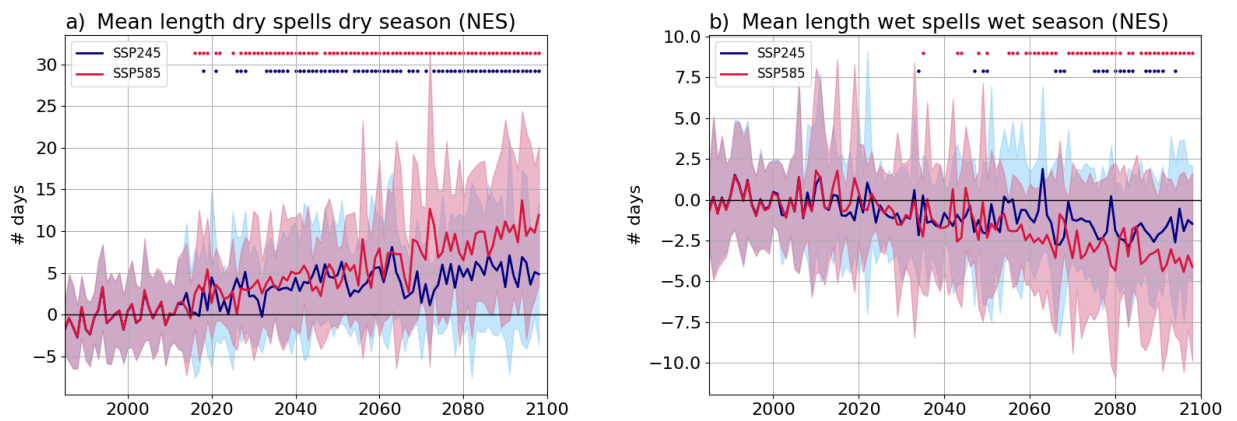


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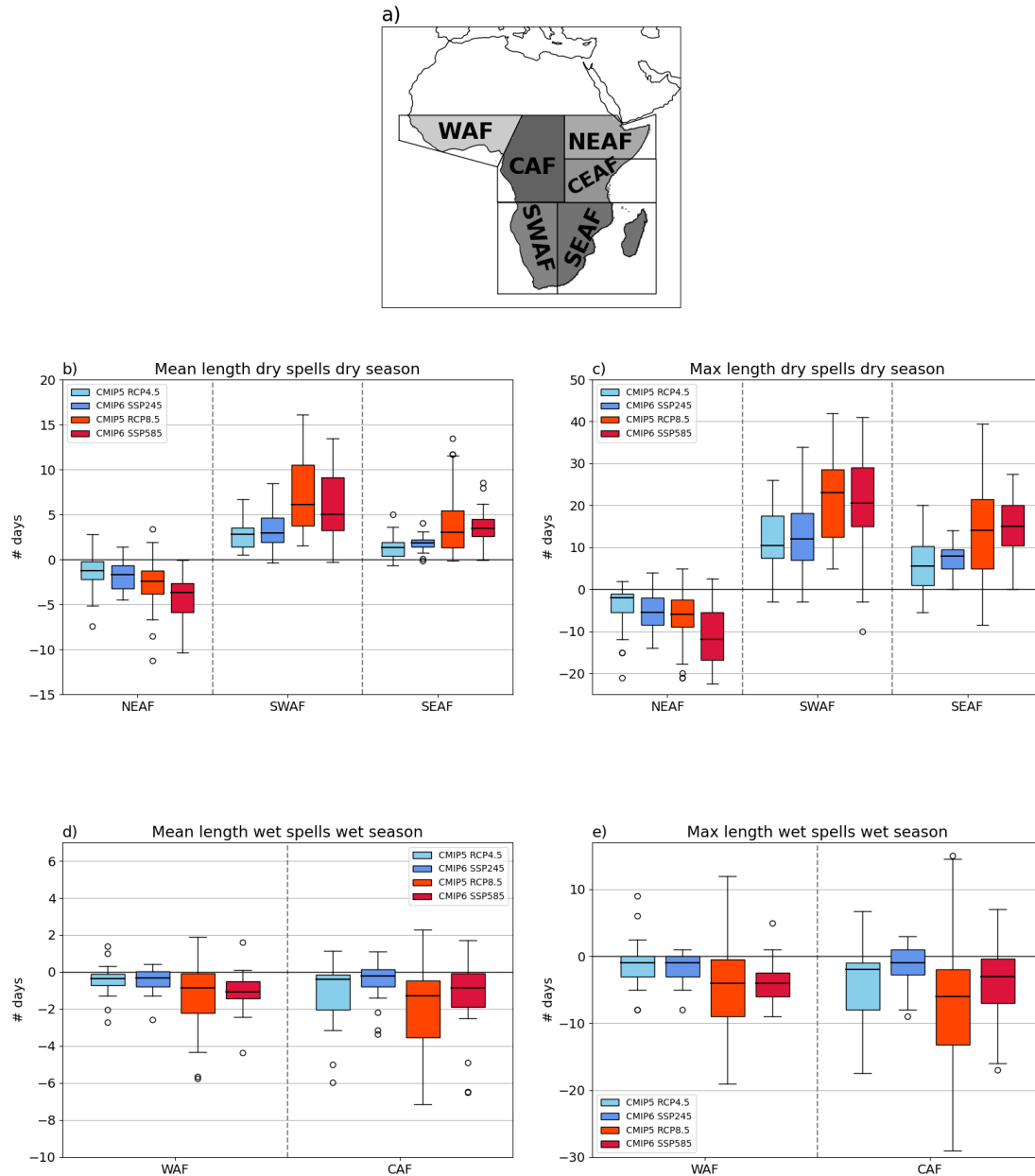


FIG. 8. a) Region map showing the location of West Africa (WAF), Central Africa (CAF), North East Africa (NEAF), Central East Africa (CEAF), South West Africa (SWAF), and South East Africa (SEAF). b-c) Boxplots showing the change in mean/max length of dry spells in the dry season over NEAF, SWAF and SEAF. d-e) Boxplots showing mean/max length of wet spells in the wet season over WAF and CAF. 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. Results for all regions are shown in Figure S17.

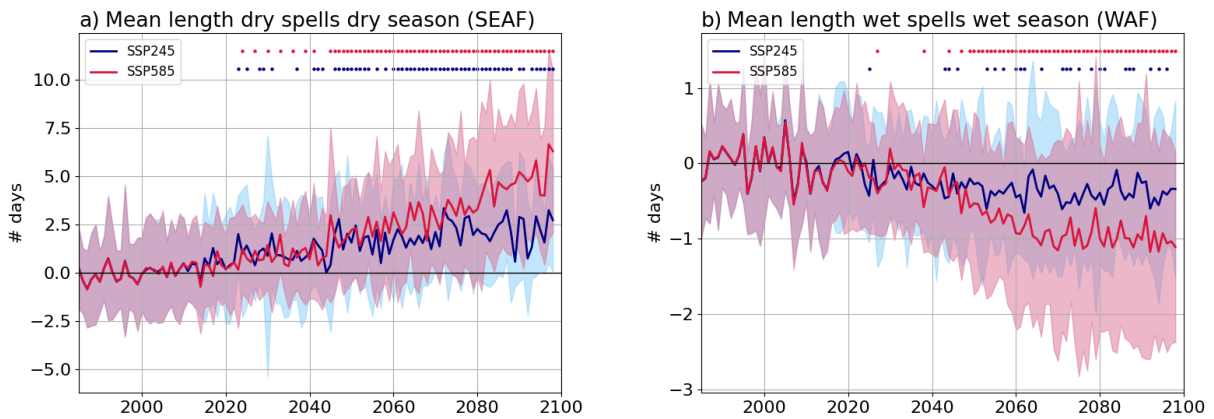


FIG. 9. Timeseries of a) mean length of dry spells in the dry season over south-east Africa (SEAF) and b) mean length of wet spells in the wet season over West Africa (WAF). The blue/red line shows the multi-model mean over the CMIP6 models for the SSP245/SSP585 scenario; the shading indicates one standard deviation. The dots indicate when the range of values from 18 models for that year is significantly different from the range for 1985–2014 at the 5% level, using a Mann–Whitney U and t test.

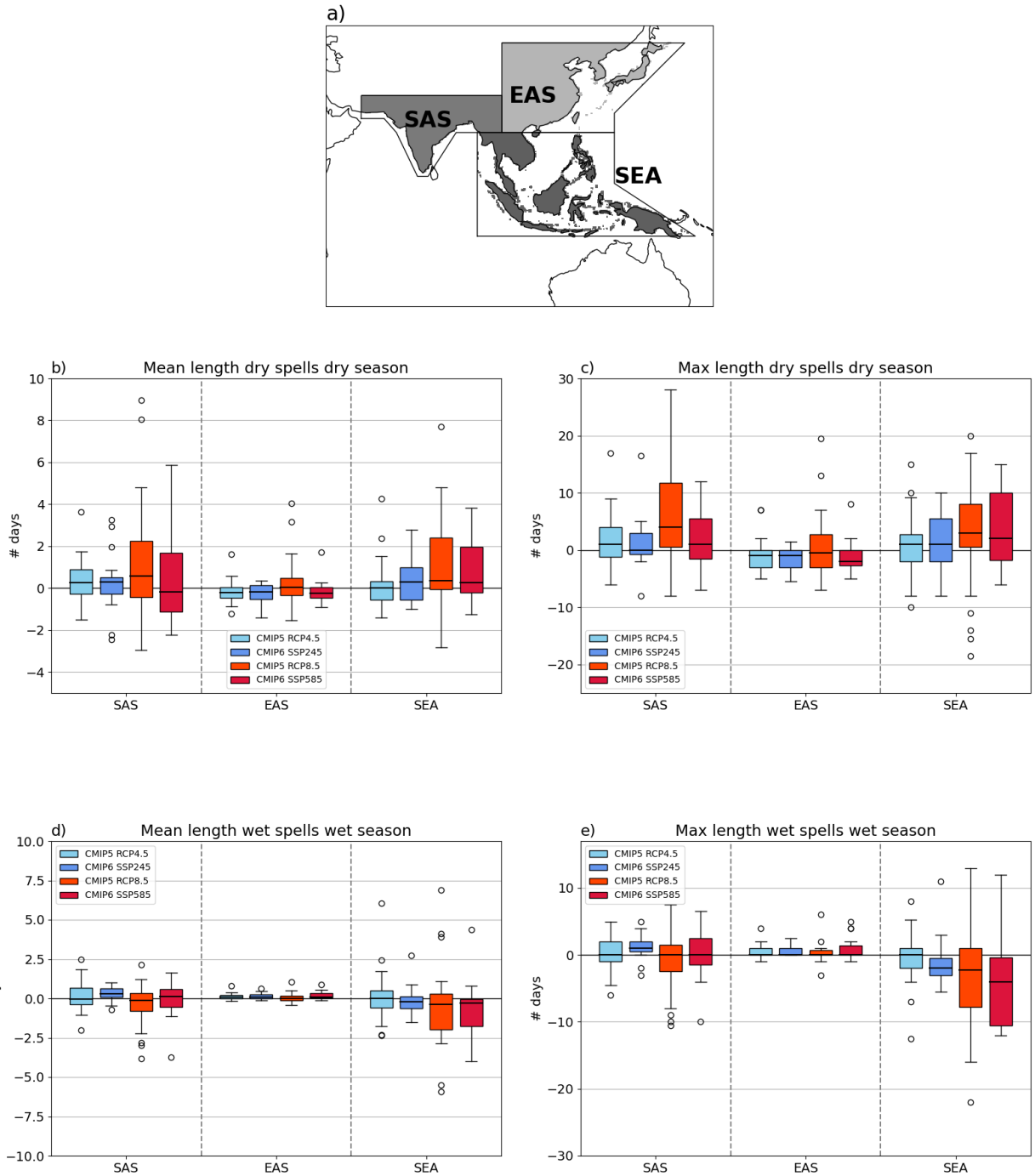


FIG. 10. a) Region map showing the location of South Asia (SAS), East Asia (EAS), and South East Asia (SEA). b-c) Boxplots showing the change in mean/max length of dry spells in the dry season. d-e) Boxplots showing the change in mean/max length of wet spells in the wet season. The different colours 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.

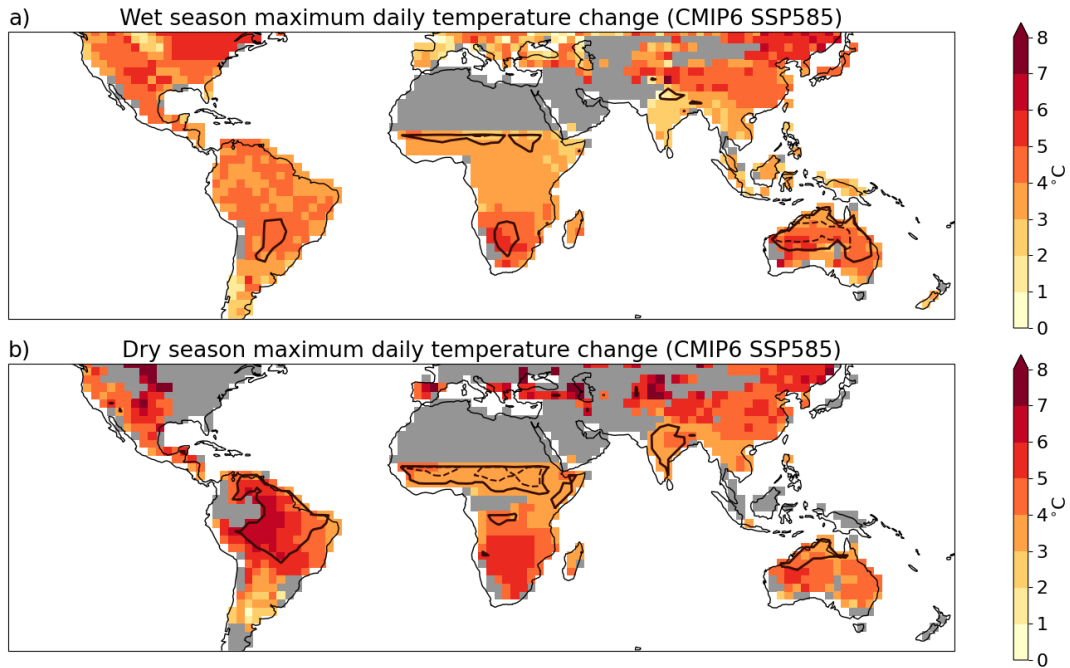
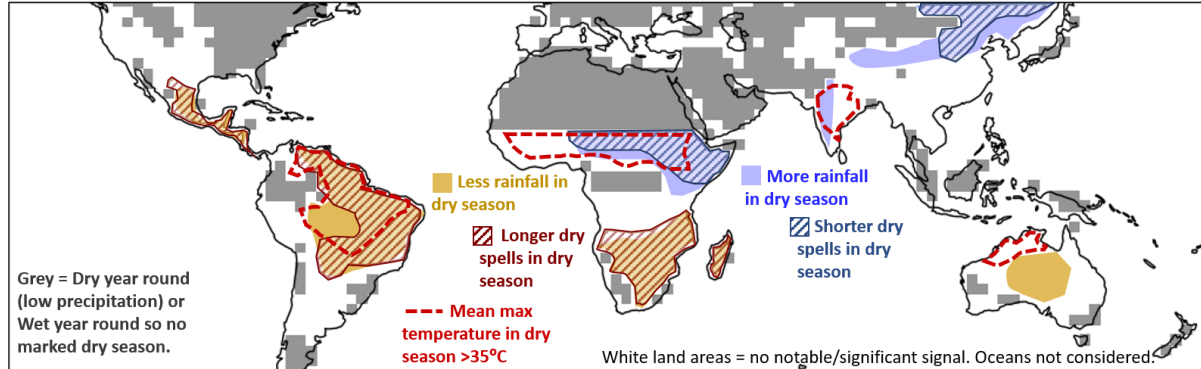


FIG. 11. CMIP6 multi-model median change in a) mean wet season maximum daily temperature and b) mean dry season maximum daily temperature for the SSP585 scenario over 2070-2099 compared with the historical simulation over 1985-2014. The contours indicate where the CMIP6 multi-model median wet and dry season mean daily maximum temperature is greater than 35°C for the historical simulation over 1985-2014 (dashed line) and the SSP585 simulation over 2070-2099 (solid line). In all panels grey regions indicate where at least 50% of the models do not give a value (dry year round or wet year round in the dry season plots).

DRY SEASON



WET SEASON

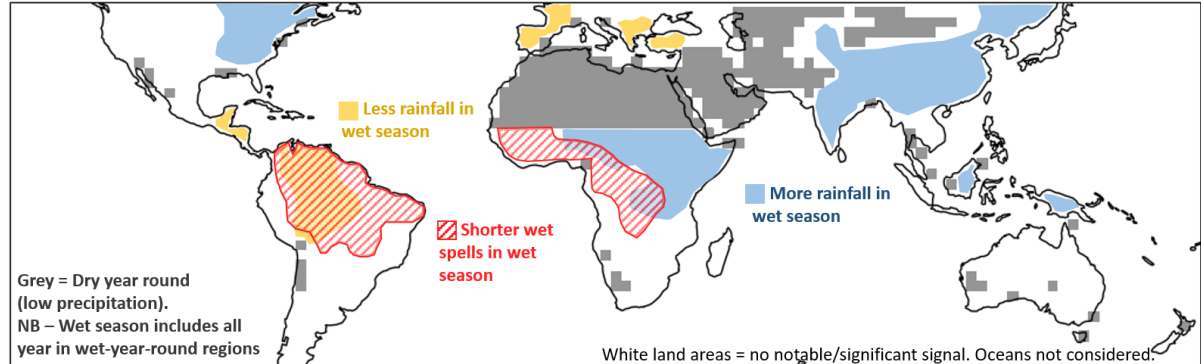


FIG. 12. Schematic summarising the changes in wet/dry season rainfall and wet/dry spell lengths in wet/dry seasons found here; the top panel is for dry seasons and the bottom row is for wet seasons (including regions that are wet year round).

(Top) Longer dry spells and lower rainfall during the dry season are found over Central and South America and Southern Africa. Shorter dry spells and more rainfall during the dry season are found over East Africa and parts of Asia and the Sahel.

(Bottom) More rainfall in the wet season is found over East Africa and Asia. Less rainfall in the wet season is found over northern South America. Reductions in the length of wet spells in the wet season are found over South America and West and Central Africa.