

1 'Eastern African Paradox' rainfall decline
2 due to shorter not less intense Long Rains

3

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13 **Abstract:**

14 An observed decline in the Eastern African Long Rains from the 1980s to late 2000s appears contrary
15 to the projected increase under future climate change. This “Eastern African climate paradox”
16 confounds use of climate projections for adaptation planning across Eastern Africa. Here we show
17 the decline corresponds to a later onset and earlier cessation of the long rains, with a similar
18 seasonal maximum in area-averaged daily rainfall. Previous studies have explored the role of remote
19 teleconnections, but those mechanisms do not sufficiently explain the decline or the newly
20 identified change in seasonality. Using a large ensemble of observations, reanalyses and atmospheric
21 simulations, we propose a regional mechanism that explains both the observed decline and the
22 recent partial recovery. A decrease in surface pressure over Arabia and warmer north Arabian Sea is
23 associated with enhanced southerlies and an earlier cessation of the long rains. This is supported by
24 a similar signal in surface pressure in many atmosphere-only models giving lower May rainfall and an
25 earlier cessation. Anomalously warm seas south of Eastern Africa delay the northward movement of
26 the tropical rain-band, giving a later onset. These results are key in understanding the paradox. It is
27 now a priority to establish the balance of mechanisms that have led to these trends, which are
28 partially captured in atmosphere-only simulations.

29 **Main Text:**

30 **Introduction:**

31 The March, April, and May (MAM) “Long Rains” of Eastern Africa have historically been the major
32 rainfall and primary agricultural season for much of the region^{1,2}. Since around 1985 the rains have
33 declined, with major consequences for livelihoods¹⁻⁶. In contrast, climate model projections show
34 increased long-rains rainfall: this has been termed the “Eastern African climate change paradox” that
35 has been used to question the reliability of the projections^{5,7} and so arguably restricts their utility for
36 informing suitable adaptation measures⁶.

37 Shortage of in-situ observations makes assessment of regional rainfall trends across Africa
38 challenging, yet the recent drying of the Eastern African long rains appears robust across
39 datasets^{4,6,8}, with a drying in March, April and May⁶. Coupled Model Intercomparison Project Phase 5
40 (CMIP5) simulations fail to consistently capture the MAM Eastern African drying with around half
41 simulating a drying, and half a wetting⁶ while most atmosphere-only (AMIP) simulations, driven by
42 observed sea surface temperature (SST) and realistic radiative forcings, correctly capture the
43 drying^{3,6}. The observed drying is thus partly explained by SSTs^{2,3,9}. Comparison of the observed long
44 rains trend with a range of trends from control runs of CMIP5 coupled models, in which
45 anthropogenic forcing (and external natural forcing) are absent, indicates that the observed rainfall
46 trend is unlikely to be entirely consistent with natural variability⁶. Mechanisms have been proposed
47 linking the decline in the long rains with decadal variability in the Pacific Ocean¹⁻⁵. Specifically,
48 changes in the zonal SST gradient in the Pacific¹⁰, warming of the Indo-Western Pacific SSTs (with
49 anomalous warm water transported from the tropical western Pacific to the Indian Ocean by
50 Indonesian throughflow¹¹) and enhanced convection over the Western Equatorial Pacific, are
51 associated with an anomalous Walker circulation over the Indian Ocean. Strengthening of the upper
52 level easterlies and the descending branch lead to increased subsidence over Eastern Africa and
53 reduced precipitation^{2,5,7}. However, there is limited evidence for the Indian Ocean Walker Cell
54 extending over the Horn of Africa¹, and an equatorial zonal circulation cell is not present in boreal
55 spring¹², although modulation of subsidence over Eastern Africa by upper level easterlies over the
56 Indian Ocean has also been proposed¹.

57 Projections of changes in the onset and cessation of wet seasons across Africa under future climate
58 change, produced using CMIP5 models, show the onset and cessation of the Eastern African
59 October-December (OND) “short rains” getting later, while the cessation of the long rains is
60 projected to get earlier¹³. Changes in the timing of onset and cessation across interior Africa (0°E-
61 35°E) have been linked to changes in the progression of the tropical rain-band and the Saharan Heat
62 Low (SHL)¹³. A stronger SHL (amplified by a water vapour - greenhouse gas warming feedback^{14,15}),

63 leads to a northward shift in the tropical rain-band during the boreal summer and a delayed retreat
64 of the rain-band southwards¹³. This delayed retreat may be associated with the later onset and
65 cessation of the short rains.

66 Here, specific characteristics of the long rains decline are investigated, and a regional mechanism is
67 proposed that explains both the observed decline, and the partial recovery since the late 2000s.

68 **Results:**

69 Figure 1a shows the decline in Eastern Africa MAM rainfall during 1998-2008 relative to 1986-1997,
70 and recent recovery from 2010 onwards, although with high year to year variability. The timeseries
71 is divided into 3 periods; a wetter period (1986-1997, P1), a drier period (1998-2008, P2) and the
72 recent partial recovery (2009-2018, P3). Analysis of the seasonal cycle (Figure 1b) reveals that the
73 decrease in rainfall during P2 is clearly associated with a delayed onset (Figure 1c, 4.9 days on
74 average) and earlier cessation (Figure 1d, 2.2 days on average), rather than a change in seasonal
75 maximum daily rainfall as often assumed. Onset in P3 recovers to P1 values (Figure 1e) while
76 cessation in P3 partially recovers but is still earlier than during P1 (Figure 1f). Supplementary Figure 2
77 shows the onset and cessation dates for individual years. These observed changes in seasonality are
78 not fully explained by the warming of the Indo-Western Pacific SSTs altering subsidence over Eastern
79 Africa since there is no decrease in seasonal maximum in area-averaged daily rainfall during P2 (Fig.
80 1b).

81 A delayed and then faster movement of the tropical rain-band northwards across Eastern Africa
82 during the boreal spring could explain later onset and earlier cessation (Supplementary Figure 3).
83 Previous studies, using multiple observational datasets, found wetting at the northern and southern
84 limits of the tropical rain belt extent during June - August (JJA) and December - February (DJF) over
85 1983-2008, with central Africa drying during MAM over the same period^{8,9}, consistent with a faster
86 progression of the rain-band.

87 A recent study highlighted that under future climate change the strengthened SHL delays the
88 southward progression of the rain-band across Africa, giving a later cessation and onset¹³. The recent
89 observed faster northward movement of the rain-band in MAM therefore similarly suggests the SHL
90 or a comparable mechanism may play a role in the earlier long-rains cessation revealed in Figure 1.
91 Following this study¹³, here changes in a different, proximal, heat low are investigated, to ascertain
92 whether this influences the recent changes in the timing of the long rains over Eastern Africa.

93 ***Early Cessation***

94 Focusing on the early cessation, the Arabian Heat Low is a dominant feature over the Arabian
95 Peninsula in spring and summer¹⁶, which interacts with the Somali Jet and Indian Monsoon¹⁷⁻¹⁹.
96 Figure 2 (and Supplementary Figure 6) reveal a strong decrease in geopotential height over Arabia
97 and the adjacent Arabian Sea in May in ERA-Interim from P1 to P2 (-8.46m over 50-75°E, 10-25°N).
98 This is also present in April (further north), is seen in other reanalyses (NCEP & MERRA) and is robust
99 to changing periods of the composite (for P3-P1 see Supplementary Figure 5 and Supplementary
100 Figure 6). The decrease in geopotential height is expected to draw the rain-band northwards faster
101 and further, consistent with the observed and analysed rainfall changes (Fig. 1-2, Supplementary
102 Figure 4). ERA-Interim captures the drying in Central and Eastern Africa in May, and the northward
103 shift in Indian Ocean and Western India rainfall seen in observations²⁰ (Supplementary Figure 4). The
104 long rains correspond to the seasonal northerly-to-southerly reversal of the Somali Jet off the Horn
105 of Africa, as the tropical rain-band crosses Eastern Africa. During P2 the southerlies are enhanced
106 across Eastern Africa in May (Figure 2), consistent with the earlier cessation of the faster
107 propagating rain-band. Temperatures over the Arabian Peninsula increased over 1979-2009²¹,
108 consistent with a pressure and geopotential height decrease. Furthermore, Arabian July - September
109 temperatures exhibited a somewhat step-wise increase in the mid-1990s²², coincident with the
110 Eastern African rainfall decline.

111 The decline in Eastern African May rainfall from 1998/9 onwards (coincident with the decrease in
112 MAM rainfall) followed the decrease in geopotential height over Arabia starting in 1997 (Figure 3).
113 The correlation between the geopotential height and May rainfall (CHIRPS) time-series is moderate
114 ($r=0.59$), but significant at the 1% significance level (Figure 3a-b). This correlation increases with
115 smoothing, which suggests that both are driven, at least to some extent, by a common relatively
116 persistent driver, with SSTs the most obvious candidate. This relationship might also address some of
117 the low predictability of MAM rainfall on a seasonal time-scale^{23,24}.

118 The AMIP multi-model mean captures the drying across Eastern Africa in March and May and the
119 increase in the strength of the Somali Jet in May (Figure 2). It also captures the decrease in
120 geopotential height in May over the north Arabian Sea (Figure 2f), although the decrease is further
121 southeast than in ERA-Interim, and restricted to over the ocean, not extending sufficiently over
122 Arabia (Figure 2 and Supplementary Figure 6). The decrease in precipitation and geopotential height
123 are of lower magnitude than the observed changes which may be linked to the lack of drying across
124 the CMIP5 simulations. The assimilation of observations into ERA-Interim helps capture the trends
125 compared with AMIP. AMIP simulations struggle to fully capture the response over land to SST and
126 other forcings^{25,26}. Summertime dust increases over Arabia from 1997 to 2009 have been shown to
127 deepen the Arabian Low in simulations²⁶, suggesting that the dust trend may influence Eastern
128 African MAM rainfall.

129 We correlate the change in May geopotential height southeast of Arabia (where the AMIP models
130 give the largest pressure decrease and ERA-Interim also shows a decrease) with both the change in
131 May rainfall and the change in long rains cessation over Eastern Africa across the AMIP simulations
132 (Figure 3 and Supplementary Figure 7). The correlation ($r=0.33$) shows that the spread in the
133 geopotential height decrease between models explains some of their range in the change in May
134 rainfall and cessation date of the long rains. The multi-model mean captures the interannual
135 correlation between the geopotential height over Arabia and the northern Arabian Sea and Eastern

136 African May rainfall (Supplementary Figure 8), with 25 out of 28 models capturing the positive
137 correlation and 11 out of 28 models exhibiting a statistically significant interannual correlation. The
138 AMIP simulated geopotential height changes are therefore important for explaining inter-model
139 variations in the change in long-rains cessation, May rainfall change and the total MAM rainfall.

140 From P1 to P2, May SST increased more over the very north Arabian Sea than further south in the
141 Indian Ocean (Figure 4b). Warm May SSTs in the Indian Ocean are associated with a later cessation
142 (Figure 4d), but this correlation is reversed over the very north Arabian Sea (north of 15-20°N). We
143 suggest that these changes in SST gradient across the Arabian Sea are associated with both the
144 rainfall variability and its trend, with warmer SSTs to the north and cooler SSTs to the south initiating
145 a pressure gradient across the Arabian Sea, drawing the rain-band north. Also, warmer SSTs to the
146 north will favour moist convection there. The withdrawal of rainfall from Eastern Africa at the end of
147 the long rains has previously been associated with the establishment of a low pressure zone and
148 active convection in the Arabian Sea²⁷. Two regions²⁸ were chosen to calculate the north-south SST
149 gradient across the Arabian Sea (Figure 4f) which correlates with the long-rains cessation at -0.61.
150 The correlation was tested for robustness using 10 SST realisations (see methods) and all
151 correlations were found to be significant, with a range of -0.45 to -0.70 across the 10 realisations.
152 This supports the hypothesis that the rain-band has been drawn north by the warmer SSTs and the
153 associated low-pressure, with an earlier cessation linked with relatively warmer SSTs to the north
154 from P1 to P2. Comparing P3 to P2 (Supplementary Figure 9), the SST has warmed more over the
155 very north Arabian Sea, but contrast in warming is less, contributing to the partial recovery in long
156 rains cessation, amongst other factors.

157 ***Late Onset***

158 Turning to the late onset, warming SSTs south of Madagascar (Figure 4a) are associated with
159 reduced March rainfall and later onsets (Figure 4c), as warmer SSTs to the south delay the
160 northward progression of the rain-band. Figure 4e shows a correlation of -0.41 (-0.29 to -0.48 across

161 10 realisations of HadISST) between long rains onset and the southern-hemisphere SST gradient.
162 From P2 to P3 March SST cooled south of Madagascar (Supplementary Figure 9), potentially
163 explaining the recent recovery in onset. Furthermore, during P2 the northerlies are enhanced in
164 March (Figure 2), also consistent with the delayed onset.

165 ***Future Climate Change***

166 Sub-tropical land is expected to warm faster than tropical oceans under climate change²⁹, which
167 together with water-vapour feedbacks can deepen sub-tropical heat lows, affecting rain-band
168 progression^{13-15,30,31}. CMIP5 projections for 2080-2100 suggest a greater and more extensive
169 deepening of the SHL compared to the Angola Low¹³. This leads to a strengthening north-south
170 asymmetry with the rain-band moving further north in boreal summer but not further south in
171 austral summer. CMIP5 projections also show earlier onset and earlier cessation for Eastern Africa in
172 MAM¹³. Considering the regional mechanisms identified here, projections under RCP 8.5 show a
173 greater increase in geopotential height over the Indian Ocean compared with Arabia in May and
174 strengthening of the Somali Jet (Supplementary Figure 10), consistent with projections of an earlier
175 cessation of the long rains¹³. In March, the surface temperature increases more in the western
176 Indian Ocean than south of Madagascar, which combined with the southerly wind anomaly is
177 consistent with projections of earlier onset¹³. Thus, CMIP5 projections of earlier onset and cessation
178 are consistent with the mechanism we have identified to describe the recent decline. The overall
179 increase in long-rains rainfall projected under future climate change³² (that is less than the projected
180 increase in the short rains)⁶, is therefore the consequence of little change in overall season length¹³
181 and an increase in the intensity of rainfall on individual days^{13,32,33}.

182 **Discussion:**

183 In summary, we find that the observed decline in Eastern African Long Rains is characterised by a
184 shortening of the rainy season (with later onset and earlier cessation) rather than by a decrease in

185 the peak daily rainfall. The cause of the shortening is a faster movement of the rain-band over
186 Eastern Africa during the boreal spring. This is a consequence of warmer SSTs to the north during
187 boreal summer and to the south during austral summer, which leads to an increased pressure
188 gradient and hence to more rapid travel of the rain-band during the whole January – August period.
189 The hypothesis that reduced pressure over Arabia and the adjacent ocean influences rain-band
190 progression is supported by the fact that this pressure change explains some of the variation in May
191 rain change across the AMIP ensemble. The results highlight the interhemispheric transitional nature
192 of the long rains, with the locations of March and May drivers in opposite hemispheres. Further
193 studies should quantify to what extent the mechanisms for rainfall decline revealed in this study are
194 driven by natural variability, amplified land-ocean pressure gradients driven by anthropogenic
195 carbon emissions and changing aerosol forcings; and the role of these mechanisms under future
196 climate change.

197 **Methods:**

198 *Precipitation Data.* The Climate Hazards Group InfraRed Precipitation with Stations (CHIRPS) daily
199 rainfall dataset³⁴ and Tropical Applications of Meteorology using Satellite data and ground-based
200 observations (TAMSAT)³⁵ daily rainfall dataset were used for 1985-2018. Both rainfall datasets use
201 thermal infrared imagery to calculate rainfall totals; CHIRPS also includes gauge data, a monthly
202 precipitation climatology, and atmospheric model rainfall fields from the NOAA Climate Forecast
203 System, version 2 (CFSv2)³⁴. TAMSAT, however, just uses thermal infrared imagery; gauge
204 observations are used for the time-invariant calibration but are not incorporated into the
205 estimates³⁵. The Global Precipitation Climatology Project (GPCP) monthly precipitation analysis was
206 also used over 1985-2018; it uses low-orbit satellite microwave data, geosynchronous-orbit satellite
207 infrared data, and surface rain gauge observations to calculate rainfall totals³⁶.

208 *Reanalysis data.* Geopotential height and wind data were taken from the ERA-Interim reanalysis for
209 1986-2018, produced by the European Centre for Medium-Range Weather Forecasts (ECMWF),
210 using the Integrated Forecast System combined with data assimilation³⁷.

211 *Sea Surface Temperature Data and Correlations.* Monthly HadISST (v1.1) observed SST data,
212 produced by the UK Met Office were used for 1986-2018³⁸. In order to compute the uncertainty on
213 the SST correlations, 10 different realisations of HadISST (v2.2) were used to compute a range on the
214 correlation values³⁹. These data are only available until 2015, and hence are used for 1986-2015.
215 Correlations were computed using the Pearson Correlation Coefficient, and data were detrended
216 first.

217 *Atmosphere-Only Model Simulations.* Atmosphere-only simulations were obtained from 28 models
218 from the CMIP5 generation of models⁴⁰. The atmosphere-only simulations are driven by historical
219 sea surface temperatures (SSTs), sea ice and radiative forcing agents. Daily and monthly
220 precipitation and monthly geopotential height and winds at 850hPa and 925hPa were obtained for
221 1986-2008. Only the first ensemble members (r1i1p1) are used. To enable the construction of multi-
222 model means, the data were re-gridded to a $1^{\circ}\times 1^{\circ}$ grid. The models used are listed in Table 1.

223 *Coupled Model Simulations.* Data from 29 coupled climate model simulations, used in CMIP5⁴⁰, were
224 used to assess changes in variables under future climate change. The coupled simulations include a
225 fully coupled ocean and are driven by historical radiative forcings for the observation period and use
226 radiative forcings from Representative Concentration Pathway (RCP) 8.5 for the future projections⁶¹;
227 this is a high emission scenario, with a radiative forcing of 8.5 Wm^{-2} at 2100. Only the first ensemble
228 members (r1i1p1) are used. Monthly geopotential height (850 hPa), near-surface air temperature
229 and eastward and northward winds (850 hPa) were obtained for 1980-1999 and 2080-2099. To
230 enable the construction of multi-model means, the data were re-gridded to a $1^{\circ}\times 1^{\circ}$ grid for
231 geopotential and temperature, and $3^{\circ}\times 3^{\circ}$ for winds. The models used are listed in Table 2.

232 *Onset/Cessation Methodology.* Onset and cessation dates were computed using the methodology of
233 anomalous accumulation, used in a number of studies of seasonality over Africa^{67,68}. This
234 methodology identifies wet seasons when the rainfall is persistent in occurrence, duration and
235 intensity⁶⁹. Full details of the methodology can be found in Dunning et al. (2016)⁶⁸; the two season
236 variant is used here. Firstly, the two periods of the year when the wet seasons occur, labelled the
237 climatological wet seasons, are determined by identifying local minima and maxima in the

238 climatological cumulative daily mean rainfall anomaly. To calculate the climatological cumulative
239 daily mean rainfall anomaly, the climatological mean rainfall for each day of the calendar year, R_i ,
240 and the long-term climatological daily mean rainfall, \bar{R} , are computed. The climatological cumulative
241 daily mean rainfall anomaly on day d , is:

$$C(d) = \sum_{i=1}^d R_i - \bar{R}$$

242 (1)

243 where i ranges from 1st January to the day for which the calculation applies. Local minima and
244 maxima in $C(d)$ determine the beginning and end of the climatological seasons; if the method
245 cannot identify two seasons then the point is excluded from this analysis, as the main interest is the
246 biannual regime over East Africa.

247 Onset and cessation dates are calculated for each season individually, by calculating the daily
248 cumulative rainfall anomaly for each climatological wet season (starting 20 days prior to the start of
249 the climatological wet season and ending 20 days after the end of the climatological wet season).
250 The minima in the daily cumulative rainfall anomaly is the onset date, and the maxima is the
251 cessation. Seasons of less than 14 days in length are excluded.

252 *Regions used in the analysis.* Harmonic analysis was used to determine the region of East Africa that
253 experiences a biannual regime. Using the CHIRPS daily rainfall data, the amplitude of the first
254 harmonic and second harmonic were computed at each grid point. The ratio was then calculated; if
255 the ratio was greater than 1, i.e. the amplitude of the second harmonic was greater than the
256 amplitude of the first harmonic, then the grid point was defined as biannual. The biannual region
257 was then smoothed using a Gaussian filter to give the region mask shown in Figure 1c-f. For other
258 datasets (e.g. AMIP data) the same mask was regridded.

259 For the correlations in Figure 3, the Arabian Heat Low was defined to be 50-75°E, 10-25°N; this
260 region captures the decline in geopotential in both ERA-Interim and the AMIP multi-model mean.

261 **Data Availability**

262 The TAMSAT data set is available from the TAMSAT website (<https://www.tamsat.org.uk/>). The
263 CHIRPS data set, produced by the Climate Hazards Group, is available
264 at <https://chc.ucsb.edu/data/chirps#>. GPCP data provided by the NOAA/OAR/ESRL PSD, Boulder,
265 Colorado, USA, from their web site at <https://www.esrl.noaa.gov/psd/>. ERA-I data were sourced
266 from CEDA (<ftp.ceda.ac.uk>).

267
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272 Organization for Earth System Science Portals. Model data were sourced from the CMIP5 data portal
273 (http://cmip-pcmdi.llnl.gov/cmip5/data_portal.html) and the British Atmospheric Data Centre
274 (<http://badc.nerc.ac.uk/>) via CEDA (<ftp.ceda.ac.uk>).

275

276 Monthly HadISST (v1.1) observed SST data are available from
277 <https://www.metoffice.gov.uk/hadobs/hadisst/data/download.html>. The first member of the 10
278 different realisations of HadISST (v2.2) can be found here:
279 <http://doi.org/10.22033/ESGF/input4MIPs.1221>; the rest are currently restricted until openly
280 published.

281 **Code Availability**

282 Code for computing onset and cessation dates can be obtained from Caroline Wainwright upon
283 request (c.wainwright@reading.ac.uk).

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301 **Competing interests:**

302 The authors declare no competing financial interests. The authors declare that there are no
303 competing interests.

304 **Author Contributions:**

305 JHM conceived the study, based on Dunning *et al.* 2018. CMW analysed the observed rainfall
306 datasets (CHIRPS and TAMSAT), including calculation of onset dates, and the AMIP and CMIP data.
307 CMW completed the SST correlation analysis. RJK analysed GPCP rainfall and ERA-Interim reanalysis
308 (Figures 2a,c,e and 3a-b). CMW constructed Figures 1, 3 and 4; RJK constructed Figure 2. CMW and
309 JHM led the writing of the paper. All authors interpreted and discussed the results and
310 commented on the figures and the manuscript.

311

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466 **Figure Legends:**

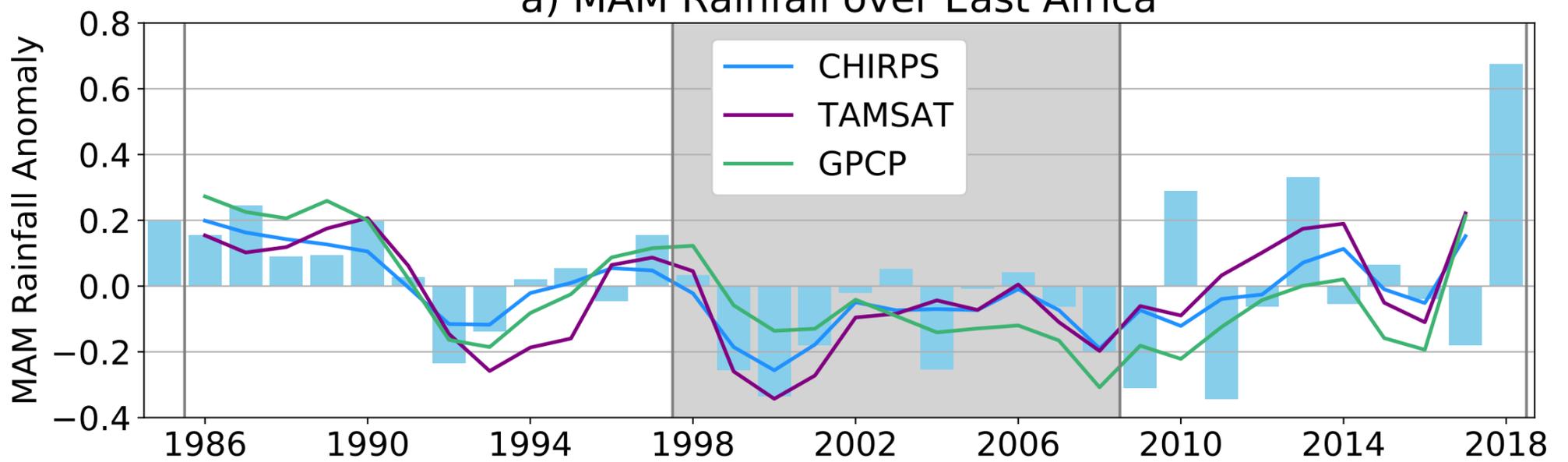
467 *Figure 1: Changes in seasonality of the long rains. a) MAM rainfall anomaly over Eastern Africa (see region in c-f) 1985-2018*
468 *from CHIRPS, TAMSAT and GPCP. The bars represent the anomaly for each year (CHIRPS); the lines are smoothed using a 3*
469 *year moving average. b) Mean seasonal cycle over Eastern Africa from CHIRPS over the 3 periods of study. c-d) Maps*
470 *showing the mean change in onset (c) and cessation (d) date of the long rains from 1986-1997 to 1998-2008. Onset and*
471 *cessation dates were calculated using CHIRPS data and the method of Dunning et al. (2016, see methods). The purple*
472 *contour demarcates the region used for Eastern Africa. e-f) As (c-d) but comparing 2009-2018 with 1986-1997.*

473 *Figure 2: Change in precipitation, geopotential height and 850hPa winds for P2 minus P1. Change in precipitation (coloured*
474 *shading), 850 hPa geopotential height (m, green contours: solid lines represent positive values and dashed lines represent*
475 *negative values) and 850 hPa winds (black arrows), for the period P2 minus period P1, for March (a,b), April (c,d) and May*
476 *(e,f). Data for (a,c,e) are from ERA-Interim (wind and geopotential height) and GPCP (rainfall) and for (b,d,f) are from the*
477 *AMIP multi-model mean (all variables).*

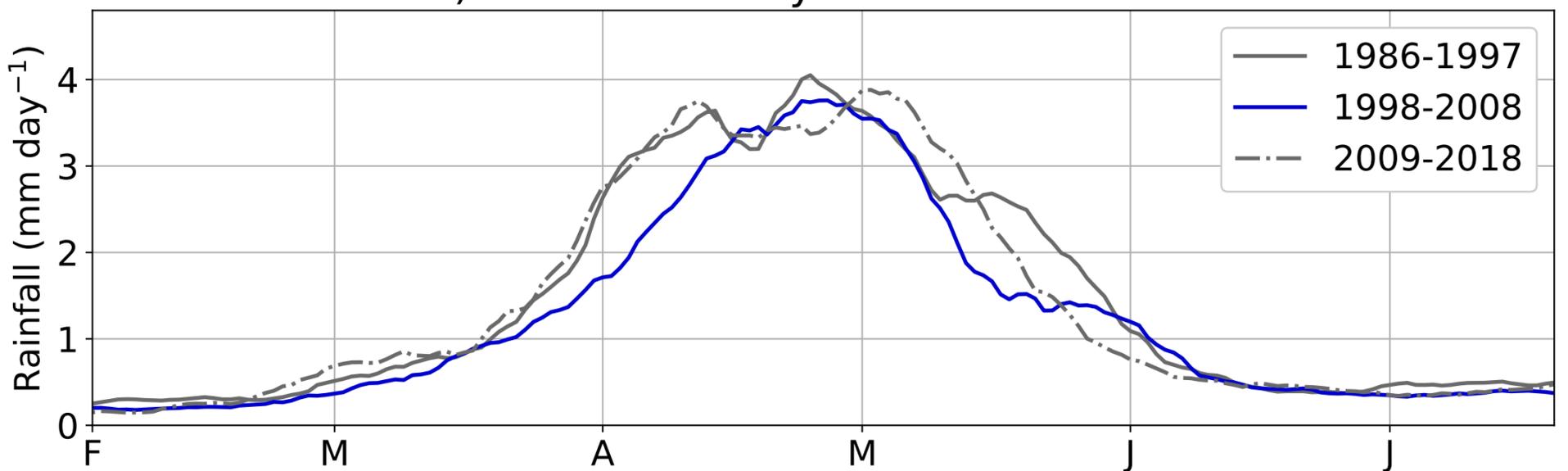
478 *Figure 3: Relationship between changes in geopotential over the north Arabian Sea and changes in East African May*
479 *rainfall. a-b) Mean May 850 hPa geopotential height over the North Arabian Sea from ERA-Interim and mean May*
480 *precipitation over East Africa from GPCP and CHIRPS over 1986-2018 smoothed with a 3 year moving average. The*
481 *corresponding scatter plot (CHIRPS) is shown in (b). Correlation was calculated on the unsmoothed timeseries. c) The*
482 *median change in cessation date of the long rains over East Africa from P1 to P2 in each AMIP model is plotted against the*
483 *mean change in 850 hPa geopotential height over the north Arabian Sea from P1 to P2. The numbers correspond to the*
484 *numbers for each AMIP model in Table 1 (methods); the cross shows the multi-model mean. The circle shows the*
485 *observations; the change in geopotential height was calculated from ERA-Interim, and the change in cessation from the*
486 *cessation dates computed using CHIRPS data.*

487 *Figure 4: Changes in March/May SST and relationship with Long Rains onset/cessation. a-b) Change in March (a) and May*
488 *(b) SST from P1 to P2 from HadISST. c-d) Correlation (Pearson correlation coefficient) of onset (c) and cessation (d) date for*
489 *the long rains across East Africa with March (c) and May (d) SST over 1986-2018. Onset and cessation dates were calculated*
490 *using CHIRPS data and the method of Dunning et al. (2016, see methods). Stippling indicates a p-value less than 0.1. e-f)*
491 *Scatter plot of the SST gradient in March/May (green box minus purple box) with the mean onset/cessation date from*
492 *CHIRPS across East Africa for each year. The values quoted in the title are the r and p values calculated using Pearson*
493 *correlation coefficient.*

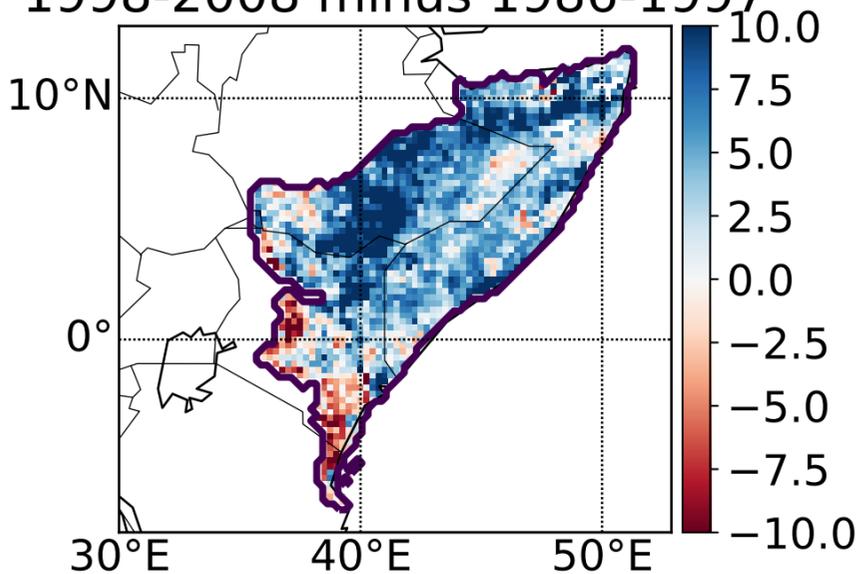
a) MAM Rainfall over East Africa



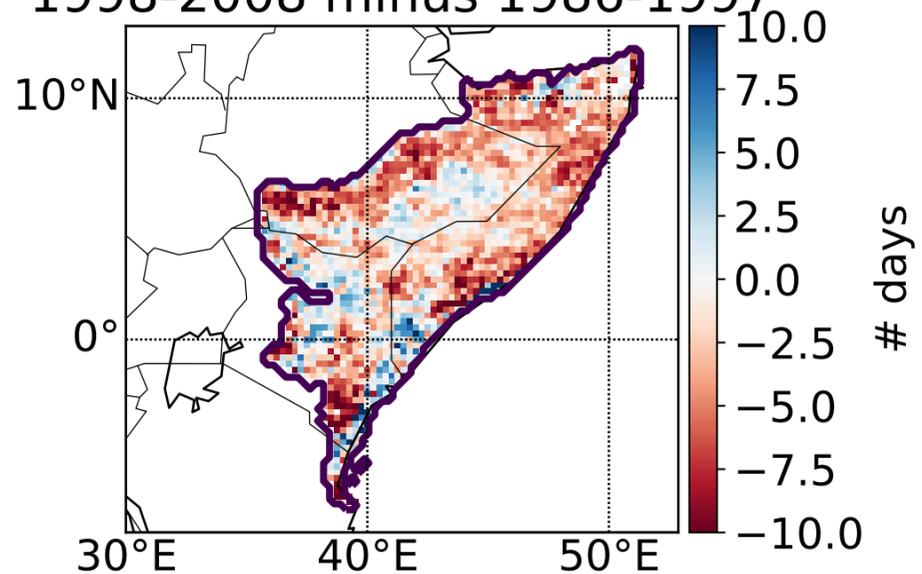
b) Mean annual cycle over East Africa



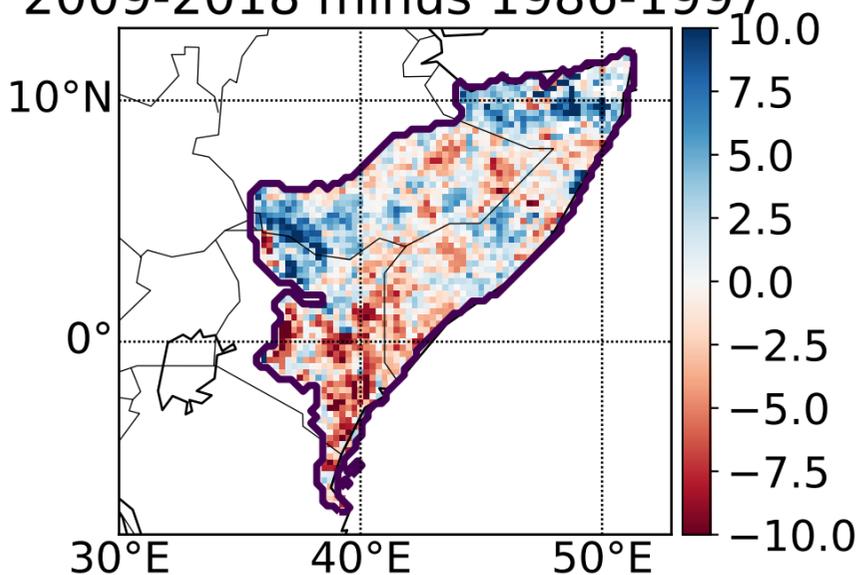
c) Long Rains Onset Change
1998-2008 minus 1986-1997



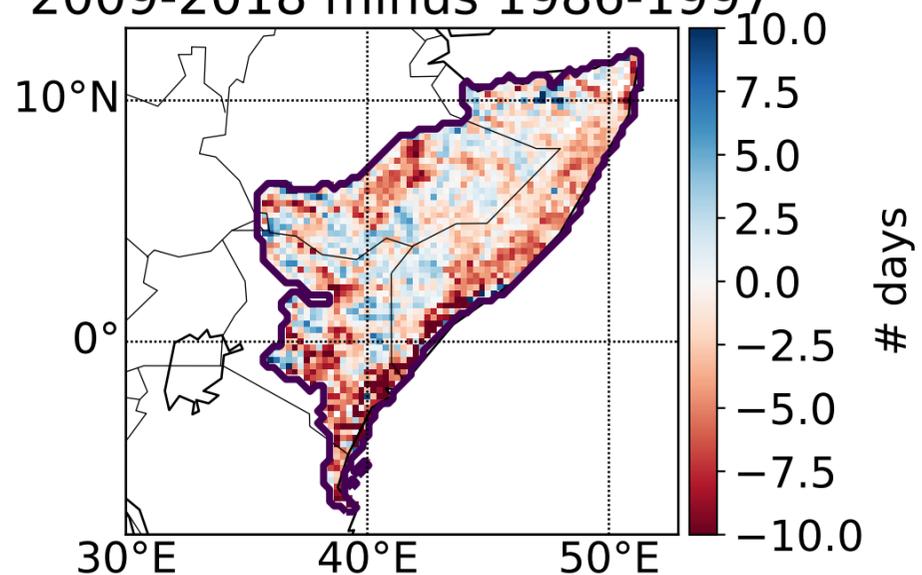
d) Long Rains Cessation Change
1998-2008 minus 1986-1997

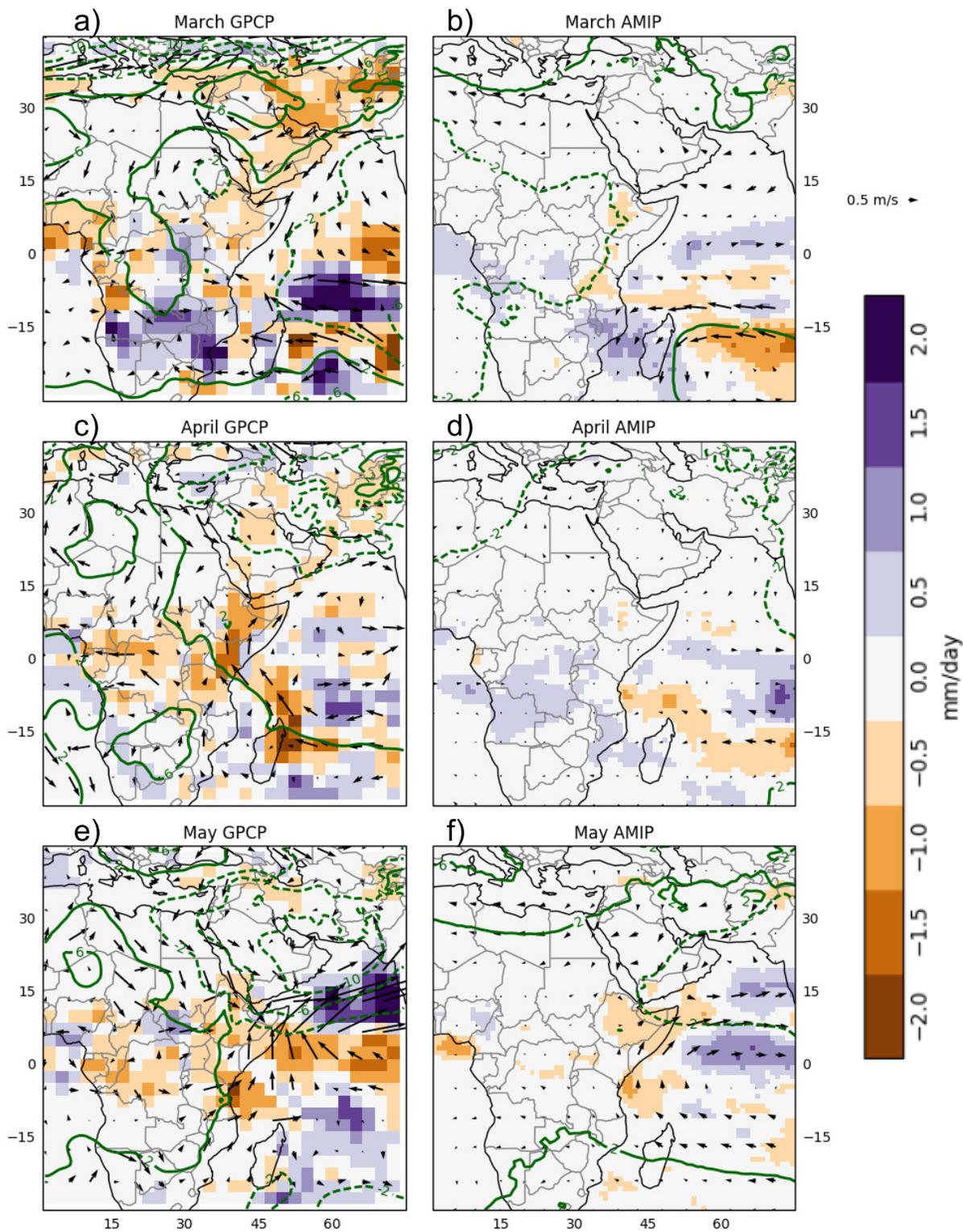


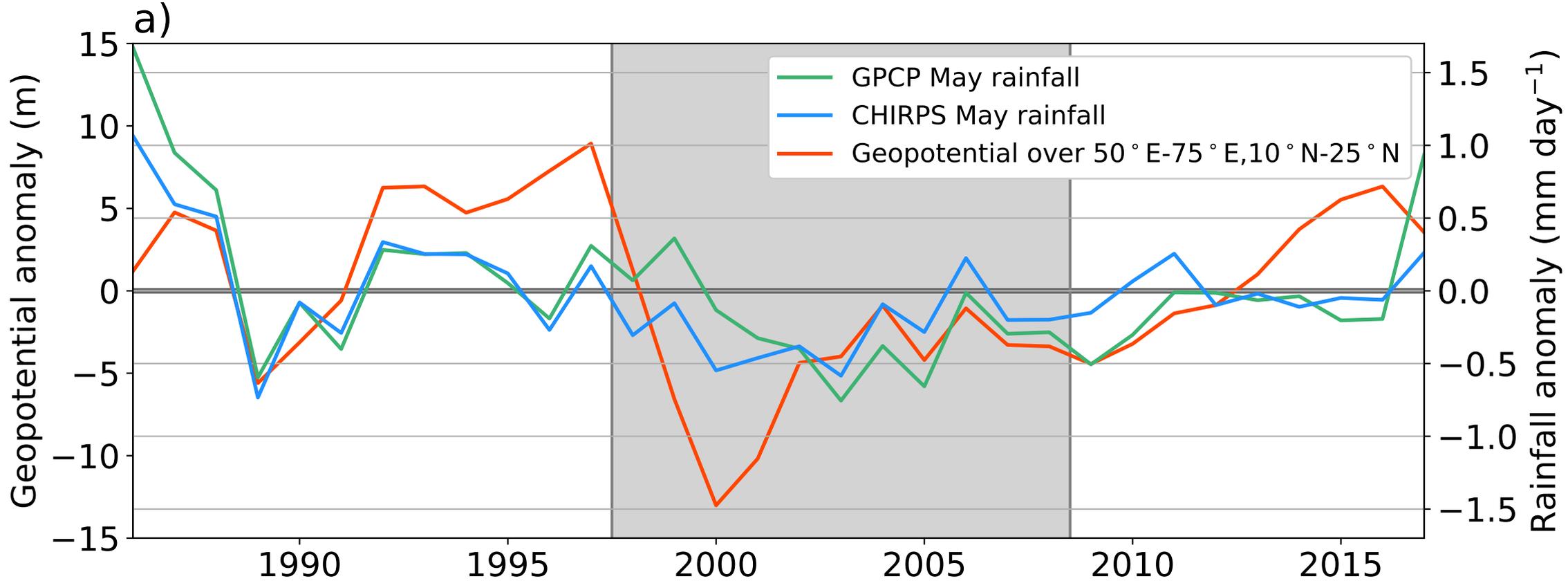
e) Long Rains Onset Change
2009-2018 minus 1986-1997



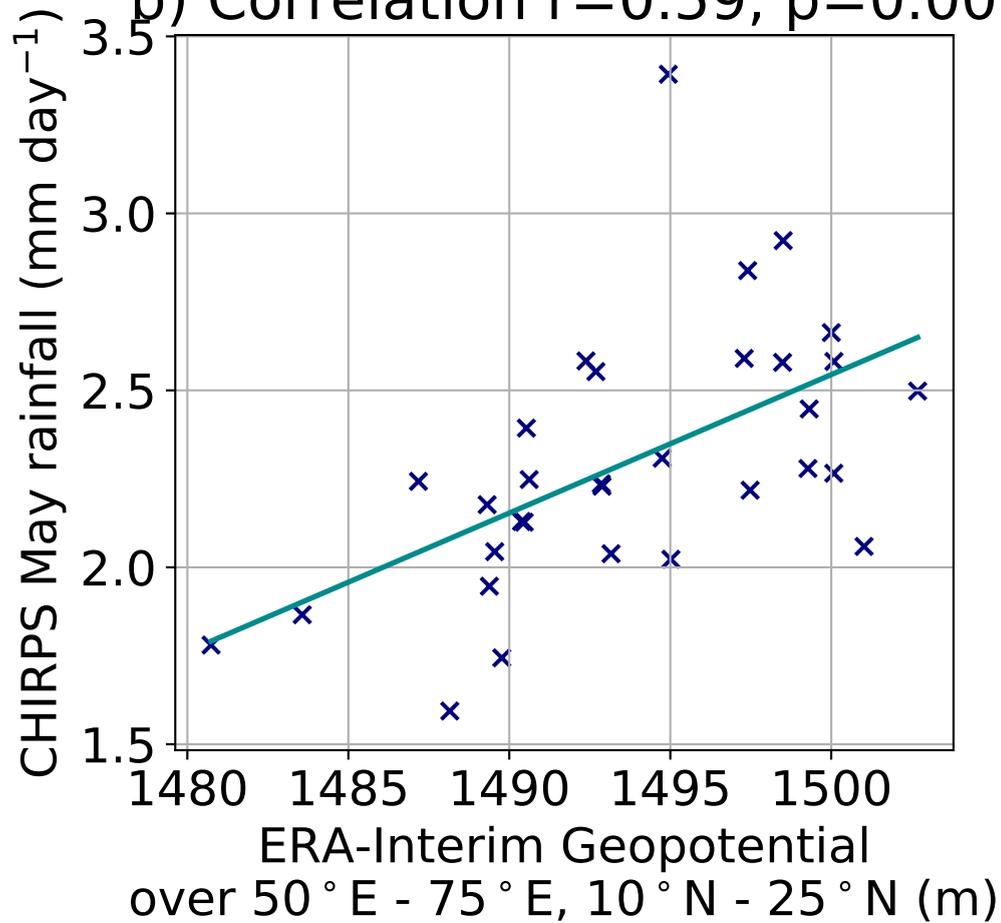
f) Long Rains Cessation Change
2009-2018 minus 1986-1997



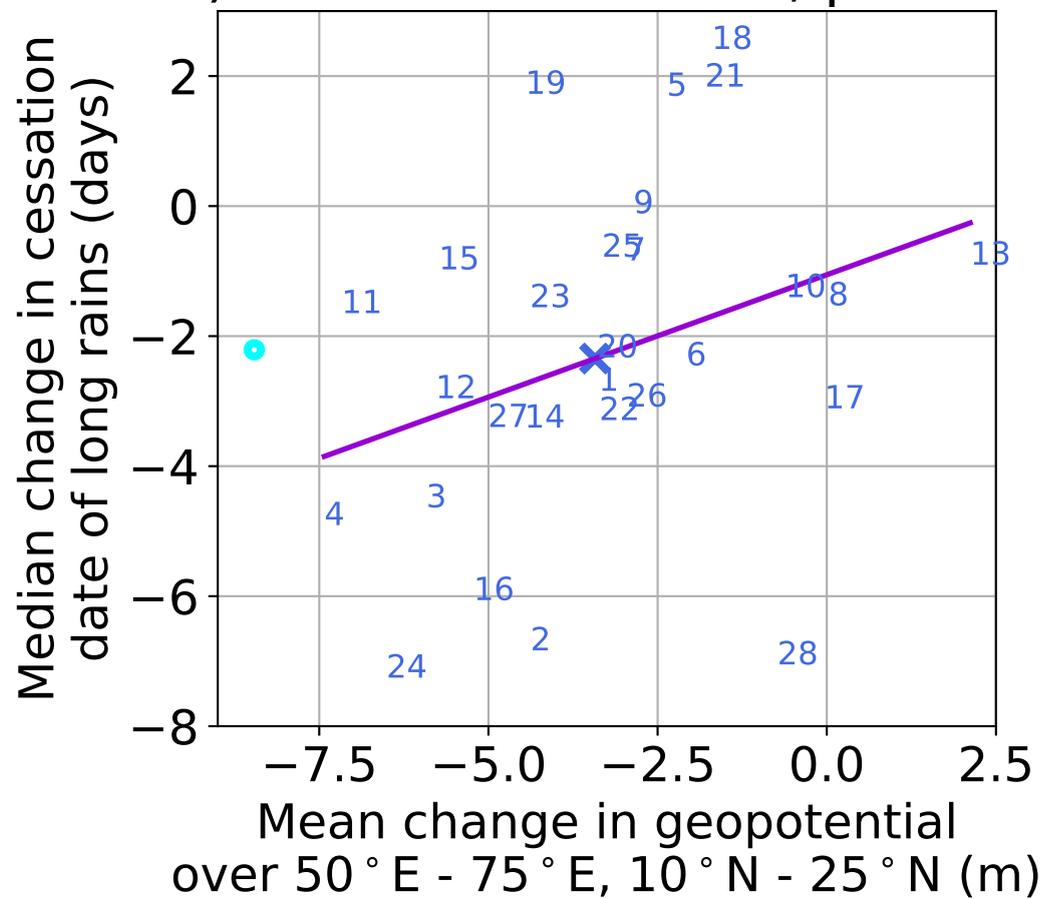




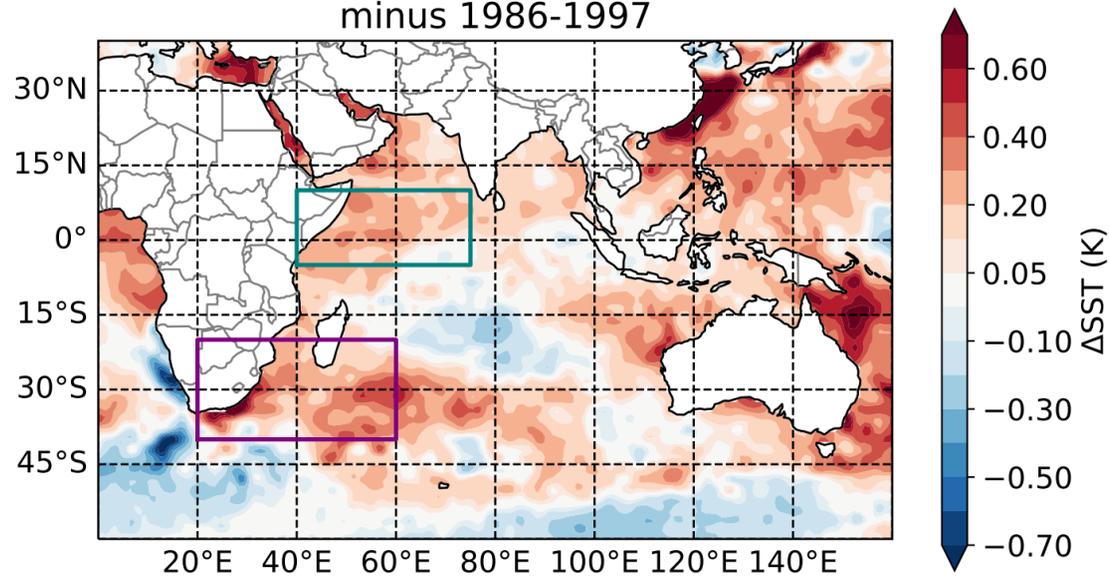
b) Correlation $r=0.59$, $p=0.00$



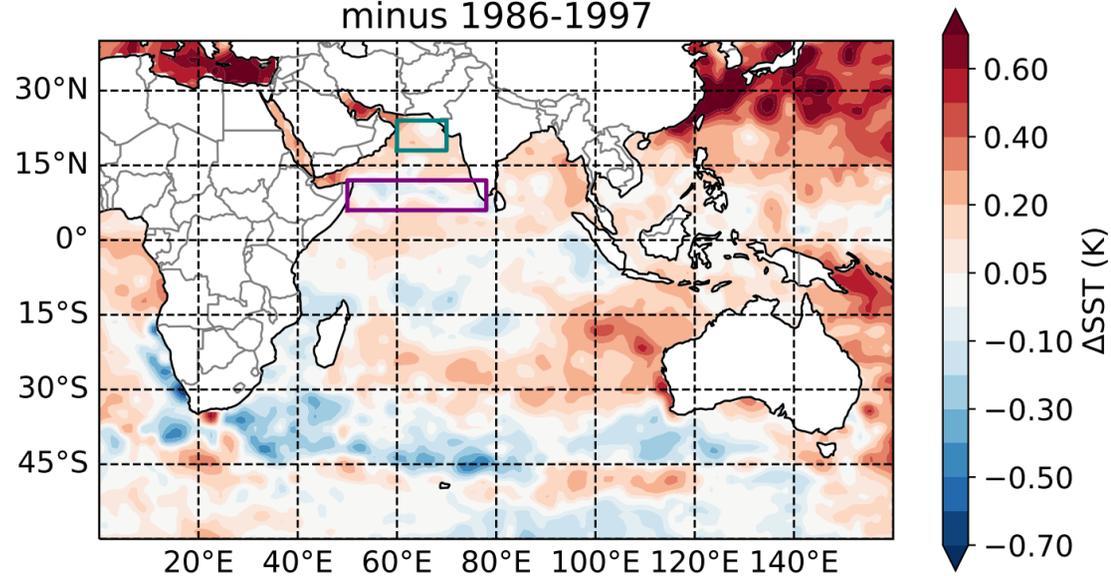
c) Correlation $r=0.33$, $p=0.08$



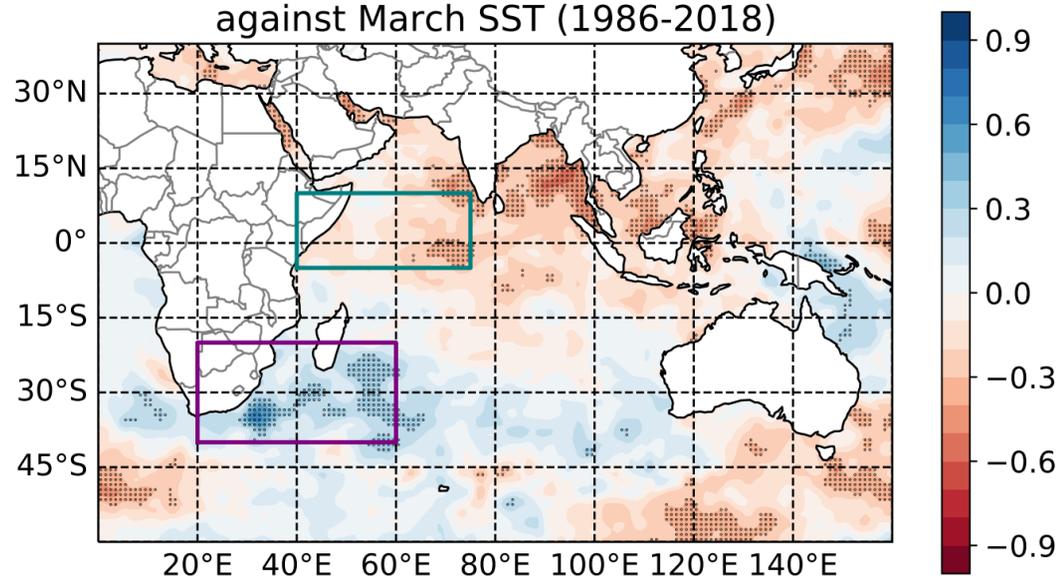
a) Change in March SST 1998-2008
minus 1986-1997



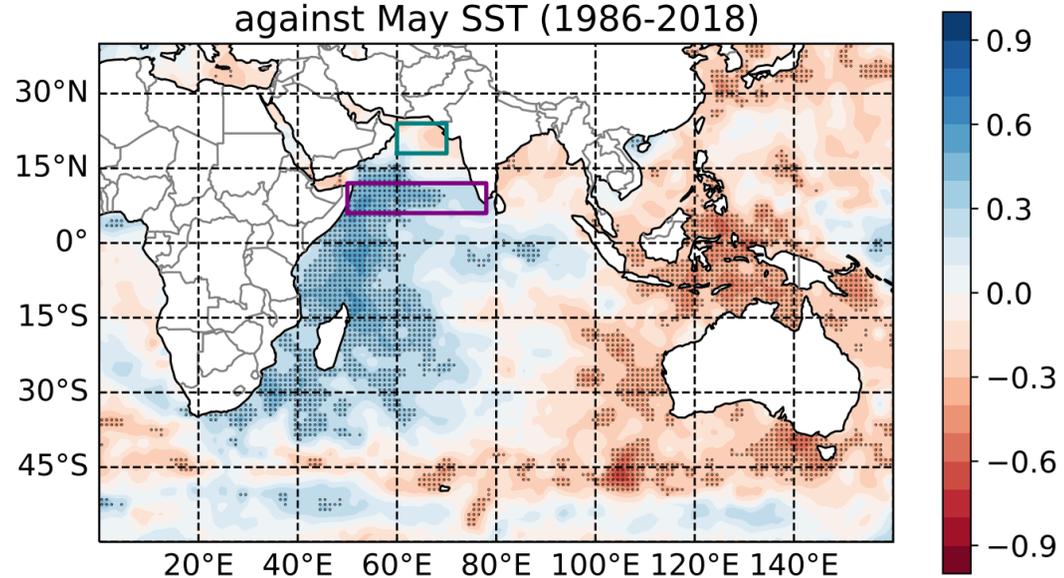
b) Change in May SST 1998-2008
minus 1986-1997



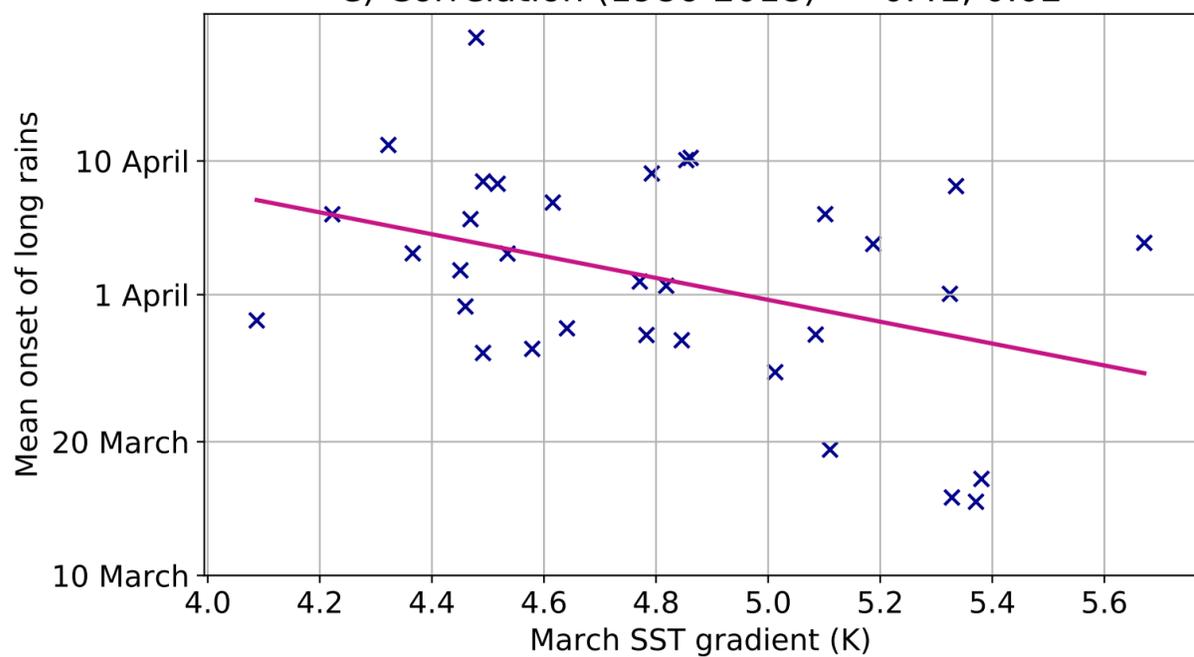
c) Correlation of Long Rains Onset over East Africa
against March SST (1986-2018)



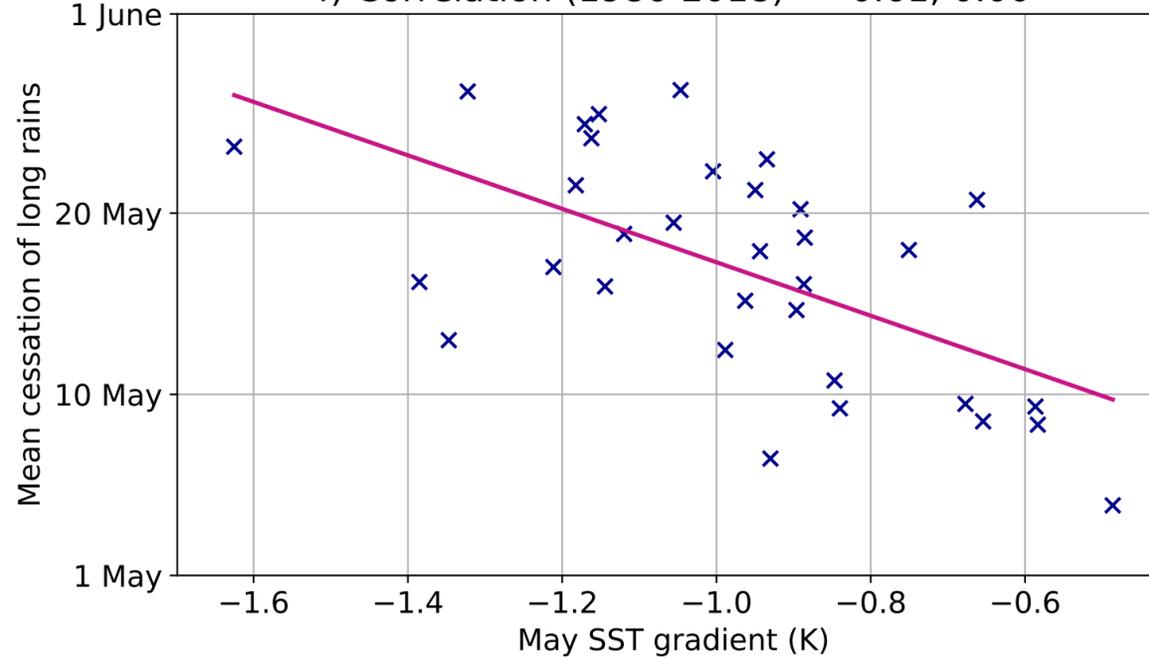
d) Correlation of Long Rains Cessation over East Africa
against May SST (1986-2018)



e) Correlation (1986-2018) = -0.41, 0.02



f) Correlation (1986-2018) = -0.61, 0.00



Model Number (Figure 3)	Institute	Model	Reference
1	CSIRO-BOM	ACCESS 1.0	41
2	CSIRO-BOM	ACCESS 1.3	41
3	BCC	bcc-csm1-1	42
4	BCC	bcc-csm1-1-m	42
5	BNU	BNU-ESM	43
6	CCCma	CanAM4	44
7	NCAR	CCSM4	45
8	CMCC	CMCC-CM	46
9	CNRM-CERFACS	CNRM-CM5	47
10	CSIRO-QCCCE	CSIRO-Mk3-6-0	48
11	ICHEC	EC-EARTH	49
12	LASG-CESS	FGOALS-g2	50
13	NOAA-GFDL	GFDL-CM3	51
14	NOAA-GFDL	GFDL-HIRAM-C180	51
15	NOAA-GFDL	GFDL-HIRAM-C360	51
16	NASA-GISS	GISS-E2-R	52
17	MOHC	HadGEM2-A	53
18	INM	inmcm4	54
19	IPSL	IPSL-CM5A-LR	55
20	IPSL	IPSL-CM5A-MR	55
21	IPSL	IPSL-CM5B-LR	55
22	MIROC	MIROC5	56
23	MPI-M	MPI-ESM-LR	57
24	MPI-M	MPI-ESM-MR	57
25	MRI	MRI-AGCM3-2H	58
26	MRI	MRI-AGCM3-2S	58
27	MRI	MRI-CGCM3	59
28	NCC	NorESM1-M	60

Table 1: List of AMIP models (and institutions) used in this study. The number in the first column is the symbol used for that model in Figure 3c.

Institute	Model	Reference
CSIRO-BOM	ACCESS 1.0	41
CSIRO-BOM	ACCESS 1.3	41
BCC	bcc-csm1-1-m	42
BNU	BNU-ESM	43
CCCma	CanESM2	44
NCAR	CCSM4	45
NSF-DOE-NCAR	CESM1-BGC	62
NSF-DOE-NCAR	CESM1-CAM5	63
CMCC	CMCC-CMS	46
CMCC	CMCC-CM	46
CNRM-CERFACS	CNRM-CM5	47
CSIRO-QCCCE	CSIRO-Mk3-6-0	48
ICHEC	EC-EARTH	49
LASG-CESS	FGOALS-g2	50
NOAA-GFDL	GFDL-ESM2G	64
NOAA-GFDL	GFDL-ESM2M	64
MOHC	HadGEM2-CC	65
MOHC	HadGEM2-ES	65
INM	inmcm4	54
IPSL	IPSL-CM5A-LR	55
IPSL	IPSL-CM5A-MR	55
IPSL	IPSL-CM5B-LR	55
MIROC	MIROC5	56
MIROC	MIROC-ESM-CHEM	66
MIROC	MIROC-ESM	66
MPI-M	MPI-ESM-LR	57
MPI-M	MPI-ESM-MR	57
MRI	MRI-CGCM3	59
NCC	NorESM1-M	60

Table 2: List of CMIP5 models and institutions that provided coupled model output used in this study.