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Clouds and temperature drive dynamic changes in tropical flower production

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Authors

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30 Tropical forests are incredibly dynamic, showing rapid and longer-term changes in growth,
31 mortality, and net primary productivity (NPP)^{1,2,3}. Tropical species may be highly sensitive to
32 temperature increases associated with climate change because of their narrow thermal tolerances.
33 However, at the ecosystem scale the competing effects of temperature, light, and precipitation on
34 tropical forest productivity have been difficult to assess. Here we quantify cloudiness over the
35 past several decades to investigate how clouds, together with temperature and precipitation,
36 affect flower production in two contrasting tropical forests. Our results show that temperature,
37 rather than clouds, is critically important to tropical forest flower production. Warmer
38 temperatures increased flower production over seasonal, interannual, and longer timescales,
39 contrary to recent evidence that some tropical forests are already near their temperature threshold
40^{4,5}. Clouds were primarily important seasonally and limited production in a seasonally dry forest
41 but enhanced production in an ever-wet forest. A long-term increase in flower production at the
42 seasonally dry forest is not driven by clouds and instead may be tied to increasing temperatures.
43 These relationships show that tropical forest productivity, which is not widely thought to be
44 controlled by temperature, is indeed sensitive to only small temperature changes (1-4° C) across
45 multiple timescales.

46 Tropical forests play a large role in the global carbon budget, accounting for about 35% of
47 terrestrial productivity⁶. This productivity in turn has important cascading effects on numerous
48 species with about half of the world's species residing in the tropics⁷. Temperature has long
49 been recognized as a fundamental constraint on many biological processes across a wide range of
50 temporal and spatial scales⁸. It has been hypothesized that tropical species are highly sensitive
51 to climate change and temperature increases because of the narrow temperature range that they
52 occupy⁹, and because they may already exist near their upper thermal limits⁸. Some however,

53 have suggested that the ecological impacts of increasing temperature in the tropics will be less
54 pronounced than in higher latitudes because tropical regions experience smaller variability in
55 temperature^{10,11}. Tropical forests are warm year-round, usually receive ample precipitation, and
56 thus may instead be more strongly limited by light¹. Solar radiation has been shown to be
57 highly dynamic over tropical regions and there is some evidence that tropical cloudiness has
58 been decreasing over the past several decades¹². Numerous experimental and observational
59 studies have shown that cloud cover can limit tropical forest productivity because clouds reduce
60 light availability¹³⁻¹⁶, thus a decrease in cloudiness should result in increased productivity.

61 One of the challenges to understanding drivers of change in tropical forests is considering
62 the simultaneous effects of temperature, clouds, and precipitation, in part because long-term
63 changes in cloudiness have been difficult to quantify^{3,17}. Here we advance research in this area
64 by using a new globally gridded satellite dataset, NOAA NCDC GridSat, which provides visible
65 and infrared data to directly quantify cloudiness over the past several decades. Our empirical
66 approach is based on rare long-term flower production records from two contrasting tropical
67 forest sites— one seasonally dry, Barro Colorado Island, Panama (BCI) and one ever wet,
68 Luquillo, Puerto Rico (Figure S1 and S2). We focus on flower production because it is an
69 important measure of reproductive activity as well as an indicator of primary productivity¹⁸.
70 Using regressions of monthly data we examine what the relative effects of clouds, temperature,
71 and precipitation are on seasonal (intra-annual) and year-to-year (interannual) patterns of flower
72 production. We also examine whether there are long-term trends in flower production and
73 associated climate variables.

74 On a seasonal timescale higher temperatures were associated with greater flower production
75 at both sites, but cloudiness affected flower production in site-specific ways (Table 1, Figure 1).

76 At the seasonally dry site, flower production decreased with more clouds. In contrast, at the wet
77 site, flower production increased with increased cloudiness, which may be due to the effects of
78 diffuse light. The light-use efficiency of forests can be higher under cloudy or partly cloudy
79 conditions (low irradiance conditions) because diffuse light is scattered more uniformly
80 throughout the canopy and understory, whereas on clear-sky days sunlight comes from a single
81 direction and many leaves remain in shadow^{19,20}. Unlike the seasonally dry site, the wet site is
82 located outside the thick deep convective clouds of the Intertropical Convergence Zone (ITCZ)
83 and cloudiness values were on average lower (Figure 1 and 3), indicating fewer and/or optically
84 thin clouds that enhance diffuse radiation²¹. Results suggest that clouds limit light availability at
85 the seasonally dry site by blocking solar radiation but enhance light availability at the wet site by
86 increasing the diffuse fraction of radiation. Even though clouds had divergent effects on flower
87 production, both sites appear to respond positively to increases in light availability.

88 We further examined the role of direct radiation on flower production by estimating direct
89 light availability. To do this we reduced top of the atmosphere (TOA) insolation proportional to
90 % cloudy values (i.e., incoming minus reflected; see Supplementary Information). We found
91 that flower production increased with direct light availability at the seasonally dry site (Table
92 S1), supporting the role of direct radiation. At the wet site, however, there was no significant
93 relationship between flower production and estimated direct light availability, again suggesting
94 that the primary effect of clouds on flower production may be in altering diffuse radiation (see
95 further discussion in Supplementary Information).

96 Relationships with climate differed at seasonal (Table 1) and interannual time scales (Table
97 2). At the seasonally dry site, the positive effect of temperature was still evident at interannual
98 time scales in addition to the positive effect of precipitation (Table S3). Although irrigation

99 experiments have shown that water addition does not affect leaf litter, wood, or fine root
100 production at the seasonally dry site^{22,23}, a reduction in water availability associated with
101 droughts may lead to water stress, increased tree mortality and decreased flower production^{24,25}.
102 At the wet site, the relationship between flower production and clouds became negative while
103 controlling for the effect of temperature (Figure S4; Table S3). This negative relationship
104 suggests that interannual climate variability results in changes in cloud cover that block solar
105 radiation, possibly associated with large storms or hurricanes as opposed to more predictable
106 seasonal variability.

107 The seasonally dry site exhibited significant long-term increases in flower production at an
108 average rate of 3% more flowers per year ($R^2 = 0.84$, $df = 21$, $p < 0.001$, Figure 2; see
109 Supplementary Information). At the wet site there was no long-term trend in flower production
110 after accounting for the effects of Hurricane Hugo in 1989 (Figure S2). Decreases in cloudiness
111 and concomitant increases in solar radiation reaching tropical forests have been hypothesized to
112 drive long-term changes in tropical forests³. Contrary to these hypotheses^{3,12} our results show
113 that the lack of significant long-term trends in cloudiness at either site indicates that clouds and
114 light availability are likely not contributing to long-term directional changes at these tropical
115 forest sites (Figure 3). Instead, the trend at the seasonally dry site may be attributed to increasing
116 maximum temperature ($0.03^\circ \text{C yr}^{-1}$ or about 1% of mean monthly maximum temperatures; $R^2 =$
117 0.20 , $df = 21$, $p < 0.05$; Figure 2) or precipitation (0.17 mm yr^{-1} or 0.2% of yearly total; $R^2 = 0.23$,
118 $df = 21$, $p < 0.05$), which is also seen in our interannual analyses of flower production (Table 2,
119 Figure S4). The $\sim 0.03^\circ \text{C yr}^{-1}$ temperature increase is similar to changes across the entire
120 tropical forest biome ($\sim 0.024^\circ \text{C yr}^{-1}$)³, whereas changes in cloudiness are regionally variable.

121 Our results demonstrate that clouds and light availability affect flower production on a
122 seasonal basis, whereas temperature is a major driver of flower production across several
123 timescales (seasonal, interannual, and the long-term trend at the seasonally dry site). Although
124 the most recent IPCC projections show that temperature increases in the tropics will be smaller
125 in magnitude compared to higher latitude regions²⁶, our results demonstrate that the productivity
126 and reproductive activity of both seasonally dry and wet tropical forests are sensitive to
127 temperature changes of just 1-2°C in interannual analyses and 2-4°C in seasonal analyses
128 (Figures 1, 2, and S1) .

129 Other studies have suggested that tropical forests are already near their temperature
130 threshold^{4,5}, however our results based on flower productivity do not show evidence of this—
131 relationships with temperature were always positive at both sites (seasonal, interannual, and
132 long-term trends). Increasing temperatures may enhance flower productivity through both a
133 direct effect on photosynthesis or indirect effects such as increasing rates of litter decomposition
134 and nutrient cycling²⁷. Nonetheless, if temperatures continue to increase, it is likely that the
135 productivity of these forests will decline because of increasing rates of respiration or direct cell
136 damage^{17,28}.

137 We focused on the role of climate to help answer how tropical forests may respond to
138 climate change in the future but other factors are known to affect forest dynamics. Patterns of
139 flowering phenology in the tropics are affected by biotic interactions such as competition for
140 resources and the timing of peak pollinator activity²⁹. Disturbance events such as hurricanes and
141 El Niño-Southern Oscillation are also important to understanding long-term forest dynamics,
142 threshold responses and resilience to disturbance. Additionally, rising atmospheric CO₂
143 concentrations have been suggested as a potential driver of directional change in tropical forests.

144 However the wet site shows no directional trend, which would be the expected response to rising
145 CO₂ (although CO₂ does not vary substantially across regions, it is possible that disturbances
146 have masked this effect). Results from other tropical forest sites are also inconsistent with a
147 pantropical increase in productivity associated with rising CO₂ concentrations³⁰.

148 Our work addressing climatic drivers of tropical forests highlights the need for further
149 efforts to tease apart the relative importance of the direct and indirect effects of temperature and
150 clouds. We used an unparalleled set of satellite data and ground observations, but completely
151 disentangling competing effects will require physiologically-based experiments to complement
152 our results. Additionally, while measures of diffuse radiation are lacking, our results show that
153 these measurements are crucial for the accurate prediction of tropical forest response to global
154 change. In order to advance our understanding and improve predictions of climate change
155 impacts on tropical forests, efforts to combine and synthesize results from process-based
156 experiments and measurements with longer-term ecosystem-scale observations will be critical.

157 **Methods**

158 We examined the relative effects of clouds, temperature, and precipitation on flower production
159 using regression analyses and monthly averaged data. Because observations in time series data
160 are often serially correlated and therefore non-independent, we fit appropriate error correlation
161 structures (autoregressive and moving average components). Flower production was log-
162 transformed to reduce heteroscedasticity. We used Akaike's Information Criterion (AIC) to
163 assess model fit in a full model, which included each predictor and all two-way interactions
164 (model A in Table 1), compared to reduced models (models B-E in Table 1). Models explained
165 primarily seasonal variation in flower production, therefore to address interannual variation we
166 performed regressions using de-seasonalized monthly data (year-to-year anomalies from the

167 monthly mean for each month). Finally we examined whether there were long-term trends in
168 flower production and associated climate variables using year to predict mean values each year.
169 See Supplementary Information for further details about study sites, remote sensing of clouds,
170 and statistical analyses.

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248

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264 **Author contributions**

265 S.P., B.I.C., E.M.W., and S.J.W. developed and designed the primary analyses. S.P. and J.R.
266 analyzed satellite data and developed cloud detection algorithms. J.K.Z., C.J.N., and S.J.W.
267 curated and assisted with interpretation of the flower production and meteorological data. S.P.
268 performed all analyses and wrote the first draft of the manuscript. All authors discussed the
269 results and helped edit the manuscript.

270 **Figure Legends**

271 **Figure 1.** Empirical relationships between flower production with clouds and temperature at
272 BCI, a seasonally dry tropical forest (a) and Luquillo, a wet tropical forest (b). Data were
273 averaged to monthly values across several decades. Solid points are the observed data with
274 colors indicating differences in temperature (increasing with darker shade). The plane is fitted
275 using a multiple regression with autocorrelated errors, which explained primarily seasonal
276 variation in flower production (all $p < 0.05$; see Table 1 and Supplementary Information).
277 Precipitation was never significant except in its interaction with clouds at the seasonally dry site
278 (Figure S3).

279 **Figure 2.** Long-term trends in flower production (solid line) and maximum temperature (dotted
280 line) at BCI from 1987-2009. Flower production was calculated as the number of flower
281 presences for each species in each trap-census combination and averaged to monthly values.
282 Daily temperature was recorded at a meteorological tower above the forest canopy. Both flower
283 production and temperature were averaged to yearly values and regressed against 'year' to
284 examine long-term trends (flower production: $R^2 = 0.84$, $df = 21$, $p < 0.001$; temperature: $R^2 =$
285 0.20 , $df = 21$, $p < 0.05$).

286
287 **Figure 3.** Seasonal and interannual variation in cloudiness at BCI, a seasonally dry tropical
288 forest (a) and Luquillo, a wet tropical forest (b). Data are from NOAA NCDC GridSat-B1,
289 which provides visible ($6\mu\text{m}$) and thermal infrared ($11\mu\text{m}$) data every 3 hours since 1980 at 8-km
290 grid cell spatial resolution. Daily data were used to calculate '% of the day cloudy' and then
291 averaged to monthly values.

Table 1. Models of seasonal flower production at a seasonally dry (BCI) and wet (Luquillo) tropical forest using monthly values. The full model (model A; includes clouds, temperature, and precipitation and their two-way interactions) was compared to reduced models (B-E) based on Akaike’s information criterion (AIC). Shading indicates equivalent best-fit models. Clouds and temperature were always included in the best-fit models. The main effect of precipitation was never significant in any of the models at either site ($p > 0.05$; see Table S2). These models explained primarily seasonal (not interannual) variation in flower production (see Supplementary Information). To account for autocorrelated errors in monthly data, all models included an autoregressive parameter and models for Luquillo included an additional moving average parameter (see Supplementary Information).

BCI flower production	Parameters	AIC	psuedo R²	k
A) full model	cloud1 + temp + precip + cloud1*temp + cloud1*precip + temp*precip	150.93	0.42	9
B) without interactions	cloud1 + temp + precip	151.47	0.43	6
C) clouds and temperature	cloud1 + temp	149.65	0.43	5
D) clouds and precipitation	cloud1 + precip	165.72	0.30	5
E) temperature and precipitation	temp + precip	193.29	0.28	5
Luquillo flower production				
A) full model	cloud1 + temp1 + precip + cloud1*temp1 + cloud1*precip + temp1*precip	157.45	0.12	10
B) without interactions	cloud1 + temp1 + precip	154.29	0.10	7
C) clouds and temperature	cloud1 + temp1	155.89	0.08	6
D) clouds and precipitation	cloud1 + precip	185.28	0.04	6
E) temperature and precipitation	temp1 + precip	157.90	0.09	6

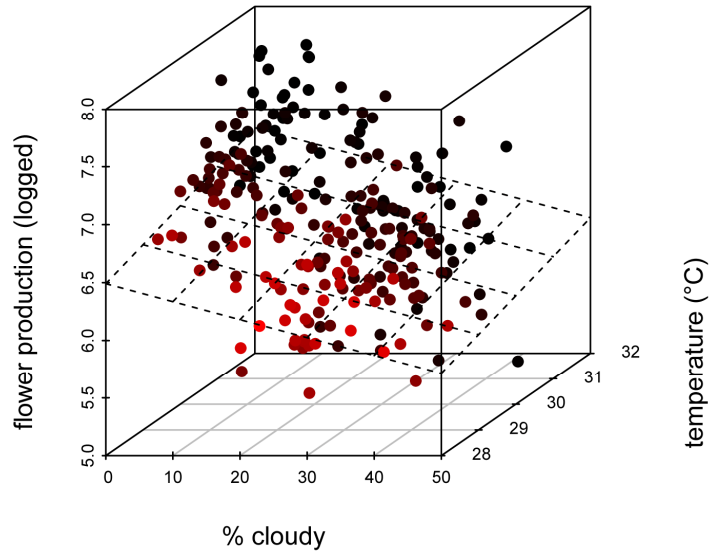
“cloud” = % of the day cloudy; “temp” = temperature; “precip” = precipitation; “1” indicates the preceding month’s value (all data were aggregated to monthly values); “pseudo R²” = the squared correlation coefficient of observed against predicted values for flower production; “k” = number of model parameters.

Table 2. Models of interannual flower production at a seasonally dry (BCI) and wet (Luquillo) tropical forest using de-seasonalized monthly values (Figure S4). The full model (model A; includes clouds, temperature, and precipitation and their two-way interactions) was compared to reduced models (B-E) based on Akaike's information criterion (AIC). Shading indicates equivalent best-fit models. Clouds and their interactions were never significant in any models at either site when using all available data for both sites (Table S3). For post-1994 Luquillo data (excluding the effect of Hurricane Hugo), clouds were significant ($p < 0.05$; see Table S3) with the additive effect of temperature (model C) and no other variables were significant in any of the other models. De-seasonalized monthly values were calculated as observed monthly values minus the mean value over all years for the appropriate month.

BCI flower production	Parameters	AIC	adj R2	k
A) full model	cloud+ temp + precip + cloud*temp + cloud*precip + temp*precip	3812	0.07	8
B) without interactions	cloud + temp + precip	3806	0.07	5
C) clouds and temperature	cloud + temp	3817	0.04	4
D) clouds and precipitation	cloud + precip	3820	0.03	4
E) temperature and precipitation	temp + precip	3805	0.08	4
Luquillo flower production - all data				
A) full model	cloud+ temp + precip + cloud*temp + cloud*precip + temp*precip	2363	0.11	8
B) without interactions	cloud + temp + precip	2362	0.10	5
C) clouds and temperature	cloud + temp	2372	0.05	4
D) clouds and precipitation	cloud + precip	2371	0.05	4
E) temperature and precipitation	temp + precip	2361	0.11	4
Luquillo flower production - post-hurricane				
A) full model	cloud+ temp + precip + cloud*temp + cloud*precip + temp*precip	1787	0.01	8
B) without interactions	cloud + temp + precip	1781	0.03	5
C) clouds and temperature	cloud + temp	1779	0.03	4
D) clouds and precipitation	cloud + precip	1781	0.02	4
E) temperature and precipitation	temp + precip	1782	0.02	4

Figure 1.

a) Seas. Dry (BCI)



b) Wet (Luquillo)

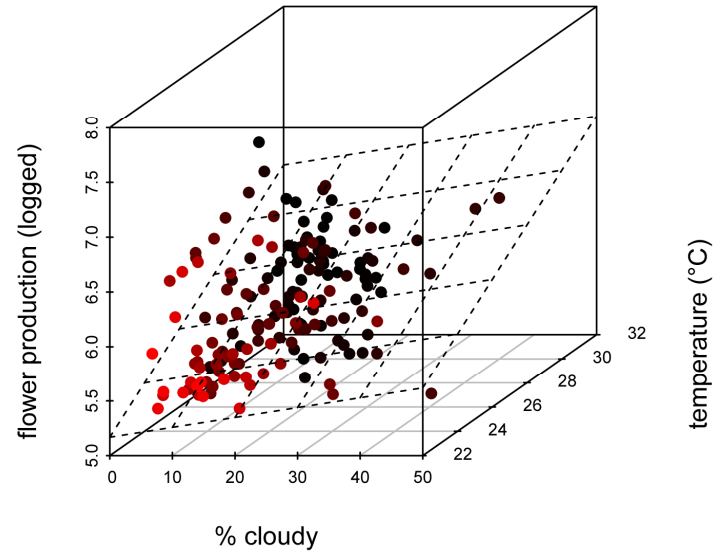


Figure 2.

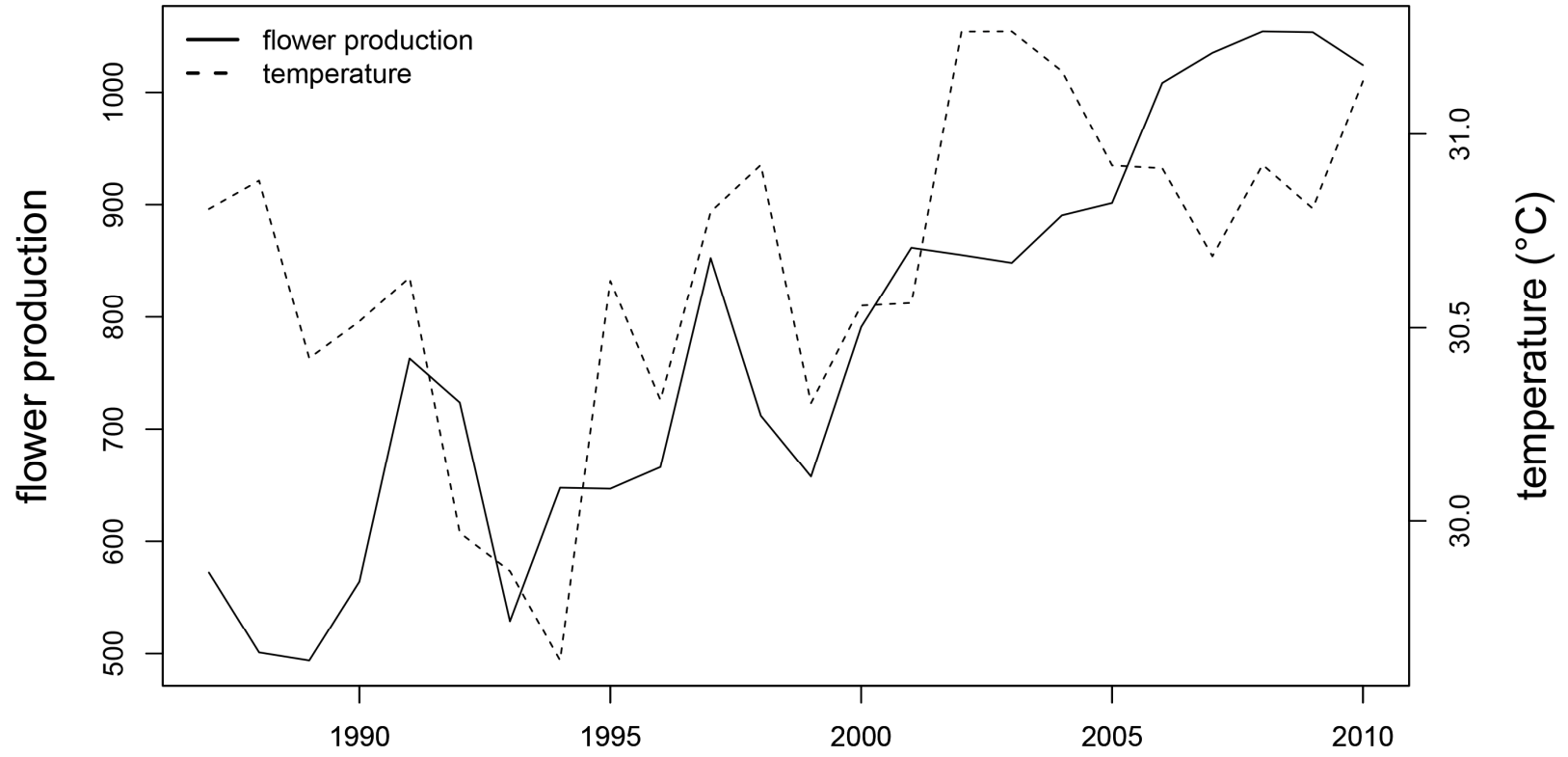


Figure 3.

