

1    **1.    Introduction**

2            In Hawai‘i, floods contributed to the largest economic loss and the second largest cause  
3            of death from natural hazards between 1996 and 2018 (Data sources: hazards reported from the  
4            Storm Event Database (National Centers for Environmental Information (NCEI), 2019a).  
5            Flooding is associated with erosion that scours stream beds, abrades bedrock, and moves large  
6            amounts of sediment and debris downstream, and downcutting channels. Improving our  
7            understanding of changes in flooding will inform watershed planning and develop potential  
8            mitigation strategies, especially as climate warming continues to alter precipitation patterns and  
9            hydrological processes. In theory, increasing global temperature results in increased water vapor  
10          in the air with its greater water holding capacity that lead to larger and more frequent extreme  
11          precipitation events (Held et al., 2006; Knutson and Manabe, 1995). Extreme precipitation events  
12          have increased in magnitude and frequency with warming temperature in many regions  
13          (Alexander et al., 2006; Beck et al., 2015; Donat et al., 2013; Easterling et al., 2017; Groisman et  
14          al., 2005; Wentz et al., 2007; Westra et al., 2013). Nevertheless, an analysis using 390 watersheds  
15          across the U.S. found the 99<sup>th</sup> percentile precipitation does not totally contribute to 99<sup>th</sup>  
16          percentile streamflow, and the contributions varied with soil moisture condition (Ivancic and  
17          Shaw, 2015). Unlike rainfall, streamflow and flooding are scale-dependent based on local  
18          watershed characteristics (Wasko and Sharma, 2017). Therefore, extreme rainfall trends might  
19          not fully describe the observed patterns in extreme streamflow trends, and their linkage needs to  
20          be studied locally.

21

22            Previous studies have shown various trends in extreme streamflow across the world, such  
23          as in North America (Cunderlik and Ouarda, 2009; Groisman et al., 2001), Europe (Blöschl et

24 al., 2019; Hannaford and Marsh, 2008; Mediero et al., 2014; Renard et al., 2008), and Australia  
25 (Ishak et al., 2013; Zhang et al., 2016). The annual maximum peak flow in Canada decreased  
26 (Cunderlik and Ouarda, 2009), whereas the high discharge in the eastern U.S. increased  
27 (Groisman et al., 2001). In Europe, the annual 10-day streamflow and annual maximum peak  
28 flow between 1969 and 2003 in the U.K. increased (Hannaford and Marsh, 2008), but decreased  
29 in France between 1968 and 1998 (Renard, 2008). The magnitude and frequency of annual peak  
30 flow in Spain generally decreased from 1942 to 2009 (Mediero et al., 2014). In Australia, the  
31 annual peak flow decreased in the south but increased in the north (Ishak et al., 2013; Zhang et  
32 al., 2016). In addition to changes in the frequency and duration of extreme streamflow events, the  
33 timing of flooding has also shifted in Europe, including earlier spring snowmelt floods in  
34 northern areas, later winter floods along the Mediterranean coast, and earlier winter floods in  
35 western Europe (Blöschl et al., 2017). As trends in extreme streamflow are geographically  
36 heterogeneous, it is critical to study the trends of extreme streamflow locally. Most trend  
37 analyses of extreme streamflow events have been conducted in continental systems, and little has  
38 been discussed in tropical regions, which are sensitive to changes in large climate patterns.

39

40 Extreme events in the tropics, including Hawai‘i, are influenced by large-scale climate  
41 variability, including the El Niño-Southern Oscillation (ENSO), the Pacific Decadal Oscillation  
42 (PDO), and the Pacific North America teleconnection pattern (PNA) (Beck et al., 2015; Chu et  
43 al., 1993; Chu and Chen, 2005; Elison Timm et al., 2011; Lyons, 1982; Mantua et al., 1997).  
44 Recent evidence indicates that climate change has affected the ENSO’s behavior across the  
45 tropics (Diaz and Giambelluca, 2012; O’Connor et al., 2015; Trenberth and Hoar, 1997) and  
46 changed El Niño properties, including its earlier starting time (Wang et al., 2019). In Singapore,

47 both daily and hourly maximum rainfall were found to increase between 1981 and 2010 with  
48 ENSO-related changes in their intensity and timing (Beck et al., 2015). With global warming, the  
49 intensity and longevity of tropical cyclones (TCs) in the Northeast Pacific has slightly decreased  
50 between 1986 and 2005 (Klotzbach, 2006); however, the frequency of TCs around the Pacific are  
51 expected to occur more frequently in the future (Li et al., 2010; Murakami et al., 2013).

52

53 In Hawai‘i, there are two seasons, dry (May – Sep.) and wet (Oct. – Apr.) seasons. In the  
54 dry season, an extreme precipitation is often attributed to hurricanes or TCs (e.g., Nugent et al.,  
55 2020). The TC activity is more favorable under El Niño conditions (Chu and Wang, 1997; Jin et  
56 al., 2014) and associates with the Western North Pacific (WNP) biennial oscillations (Luo et al.,  
57 2020). In the wet season, winter storms contribute the most to extreme rainfall. A combination of  
58 PNA, ENSO, and PDO contributes to the rainfall variability in the wet season (Frazier et al.,  
59 2017). During El Niño conditions, the wet season months often become drier in Hawai‘i and  
60 wetter during La Niña conditions. In addition, negative periods of PDO that align with La Niña,  
61 typically strengthen the intensity, and lengthen the wetter period (Chen and Chu, 2014). Streams  
62 in Hawai‘i respond to rainfall rapidly (within hours), as majority of streams are first order  
63 streams with small drainage areas, steep slopes, and short time of concentration. The low-  
64 permeable volcanic geology of all islands combined with high intensity rainfall simultaneously  
65 produces runoff, generating flashy hydrographs (Oki, 2003). Besides, the coastal areas of  
66 Hawaiian watersheds are often urbanized with impervious surface and channelized streams,  
67 which contribute to the fast rising and falling hydrograph. In Hawai‘i, we expect extreme  
68 streamflow to highly correspond to its rainfall.

69

70        Although we anticipate a close relationship between extreme streamflow and rainfall in  
71    Hawai‘i, previous studies investigated their trends separately. Chu et al. (2010) and Chen and  
72    Chu (2014) showed that extreme precipitation between the 1950s and 2007 became more  
73    common on the largest island, Hawai‘i Island, but not on other islands, with the indices of the  
74    annual count of days when the precipitation over 25.4 mm, annual maximum consecutive 5-day  
75    precipitation, and the normalized 95<sup>th</sup> percentile precipitation. Bassiouni and Oki (2013)  
76    investigated streamflow trends in Hawai‘i and found that 16 out of 26 gauges of peak flow  
77    decreased between 1943 and 2008. Furthermore, Clilverd et al. (2019) assessed streamflow  
78    records for 23 unregulated streams from 1967 to 2016 across the Hawaiian Islands and found  
79    significant declines in baseflow and surface runoff. They also discovered significant declines in  
80    peak flow on Hawai‘i Island. However, in some cases, peak flow showed opposite trends from  
81    extreme precipitation (e.g., Chen and Chu, 2014; Chu et al., 2010; Clilverd et al., 2019). Because  
82    extreme rainfall and streamflow trends have been investigated separately, the linkages between  
83    these extreme events are not clear. To better understand their associations, we need a joined  
84    examination of extreme rainfall and peak flow trends.

85

86        This study aims to characterize local changes in the annual maximum daily rainfall  
87    ( $RF_{max}$ ) and the annual peak flow ( $PF_{max}$ ) by examining their spatial and temporal trends with  
88    their associations across five major Hawaiian Islands. We investigated the trends in  $RF_{max}$  and  
89     $PF_{max}$  to determine possible relationships between the meteorological and hydrologic response to  
90    climate change, by studying: 1) the spatial distribution of trends in  $RF_{max}$  and  $PF_{max}$  in Hawai‘i  
91    from 1970 to 2005; 2) the association between peak rainfall and peak flow by pairing  
92    representative rain gauges to crest gauges; and 3) the temporal shifts of  $RF_{max}$  and  $PF_{max}$ ,

93 respectively.

94

95 **2. Materials and methods**

96 **2.1 Study area**

97 Among the eight major Islands, we focus on the five largest islands, from west to east,  
98 Kaua‘i, O‘ahu, Moloka‘i, Maui, and Hawai‘i Island, spanning from 18.9°N, 154.8°W to  
99 22.24°N, 159.8°W, where have most and long-term rainfall and peak flow records. The climate  
100 in Hawai‘i is strongly affected by the Hadley cell atmospheric circulation patterns in the Pacific.  
101 These patterns generate the typical northeast trade winds in the northern hemisphere, which  
102 induces orographic rainfall when moist air encounters the steep island topography (Lyons, 1982).  
103 Thus, these islands’ windward facing sides experience more frequent rain and higher annual  
104 rainfall below trade wind inversion (TWI; approximate mean elevation at 2,000 m; Cao et al.  
105 2007), while the leeward sides are much drier. In addition, widespread and intense precipitation  
106 may be attributed to four other types of atmosphere conditions – (i) Kona storms, the low-  
107 pressure systems that usually develop on the west of the islands accompanied with southern  
108 winds, (ii) cold fronts, (iii) upper-level trough, or (iv) tropical cyclones (Caruso and Businger,  
109 2006; Kodama and Barnes, 1997). The wide range of terrain across the Hawaiian Islands with the  
110 large scale atmospheric systems results in highly heterogeneous climate patterns, and therefore  
111 steep rainfall gradients (mean annual rainfall from 200 mm to 10,000 mm; Giambelluca et al.,  
112 2013). Further, watersheds in Hawai‘i are typically characterized by young volcanic geology,  
113 small drainage size, steep topography, and limited channel storage (Craig, 2003). Thus, streams  
114 frequently experience flash flooding where water levels rise and fall rapidly within hours of  
115 locally intense rainfall events (Oki, 2003; Sahoo et al., 2006).

116

117 2.2 *Data*

118 The locations of rain and crest (annual peak flow) gauges across the Hawaiian Islands  
119 and their physiographical regions (windward vs. leeward) are shown in Figure 1. We sought crest  
120 gauges with the longest period of record and greatest overlap with rainfall records, which  
121 resulted in a study period of 1970 to 2005 (by water year). Daily rainfall data were obtained from  
122 the National Centers for Environmental Information (NCEI) and the U.S. Geological Survey  
123 (USGS). Eighty-four rain gauges (Kaua‘i, n=17; O‘ahu, n=25; Moloka‘i, n=3; Maui, n=20, and  
124 Hawai‘i Island, n=19), with data records longer than half of the study period, were used in this  
125 study (Figure 1). Most rain gauges were located at low elevations and only a few were above the  
126 average TWI ( $> 2,000$  m). The rain gauges were not evenly distributed across the islands, except  
127 for Kaua‘i. For O‘ahu and Moloka‘i, most of rain gauges were located on the leeward side,  
128 whereas for Maui and Hawaii island, rain gauges were located mostly on the windward side and  
129 along the coast. Annual records of peak flow were obtained for 111 long-term crest gauges  
130 (circles in Figure 1; Kaua‘i, n=18; O‘ahu, n=45; Moloka‘i, n=7; Maui, n=25, and Hawai‘i Island,  
131 n=16) from the water years 1970 to 2005, maintained by the USGS. Compared with the  
132 distribution of rain gauges, crest gauges have a more even spatial distribution, except on Hawai‘i  
133 Island, where crest gauges were predominantly located on the windward side. Peak flow values of  
134 each gauge were standardized by the watershed area and converted to a daily scale (mm/day) that  
135 is comparable to the daily rainfall values (mm/day).

136 We used two climate indices to examine shifts in the seasonality of peak events, the  
137 Oceanic Niño Index (ONI) and the Pacific Decadal Oscillation (PDO) index. The ONI is one of  
138 the primary indices for monitoring ENSO. It is calculated by averaging sea surface temperature

139 monthly anomalies of the east-central equatorial Pacific Ocean, Niño-3.4 region. We retrieved  
140 monthly ONI from the Climate Prediction Center (CPC) (2019). The PDO index is defined by  
141 ocean temperature anomalies in the northeast and the tropical Pacific Ocean. The PDO index is  
142 downloaded from NCEI (2019b).

143

144 2.3 *Trend analysis*

145 Trends in the magnitude of  $RF_{max}$  and  $PF_{max}$  were analyzed using the non-parametric  
146 Mann-Kendall test (Hirsch and Slack, 1984; Mann, 1945); a p-value  $< 0.05$  was used to indicate  
147 the significant trends. Changes of the magnitude were evaluated using Sen's slope (Sen, 1968).  
148 Then, the change of values per year were divided by the mean value of the study period to  
149 generate percent change in the value per year. These two trend analysis methods are

150 recommended for analyzing environmental time series data that are not normally distributed,  
151 with no data distribution assumptions required (Hirsch and Slack, 1984; Mann, 1945).

152 Additionally, evaluation with Sen's slope is not sensitive to outliers. These trend analyses have  
153 been widely applied to quantifying and testing the significance of streamflow trend (Bassiouni  
154 and Oki, 2013; Clilverd et al., 2019; Oki, 2004; Small et al., 2006) and rainfall (Chen and Chu,  
155 2014; Frazier and Giambelluca, 2016). The R package, 'trends' (Pohlert et al., 2018), was  
156 applied in this analysis.

157

158 2.4 *Analysis of paired rainfall gauges and stream crest gauges*

159 In addition to examining data from 84 rain gauges and 111 crest gauges, we paired crest  
160 gauges with rain gauges that most likely represent the rainfall received from upstream  
161 watersheds to better understand the associations between rainfall and peak flow. We paired them

162 geographically with following criteria, ranked in order of preference:

163 1) The rain gauge was located upstream of the crest gauge in the same watershed; if  
 164 multiple rain gauges existed in the same watershed, we chose the gauge with the highest  
 165 elevation but below the TWI.

166 2) The rain gauge was upstream from the crest gauge in the neighboring watershed.

167 3) The rain gauge was close to the crest gauge (within a 5 km radius).

168 4) The rain gauge best represented the rainfall of the upstream watershed of the crest gauge,  
 169 despite being in different watersheds. For example, three crest gauges on the windward  
 170 side of East Maui were paired with one rainfall gauge at Hana, Maui, that represents the  
 171 windward rainfall.

172

173 Subsequently, we extracted daily rainfall on the same day of the  $PF_{max}$  of these paired  
 174 gauges. Additionally, 2-day and 5-day accumulated rainfall were examined to consider  
 175 antecedent conditions and rainfall events that lasted more than one day.

176

177 2.5 *Examination of temporal shifts in annual maximum rainfall and peak flow*

178 We examined the temporal shifts of  $RF_{max}$  and  $PF_{max}$  with circular statistics (Zar, 1999)  
 179 by using the R package, “circular” (Lund et al., 2017). Circular statistics is powerful when  
 180 applying to the data with unit radius or degree on a circumference (Pewsey et al., 2013). We  
 181 applied a circular analysis to the occurrence time of  $RF_{max}$  and  $PF_{max}$  for the study period by  
 182 converting the occurrence date ( $d_y$ ) of  $RF_{max}$  and  $PF_{max}$  into angular values for each year,  $y$ :

$$\vartheta_y = 2\pi * \frac{d_y}{D_y} \quad 0 \leq \vartheta_y \leq 2\pi \quad \text{Eq. 1}$$

183 Where  $d_y = 1$  corresponds to January 1<sup>st</sup> and  $d_y = D_y$  to December 31<sup>st</sup> with  $D_y$  is the

184 number of days in that year (i.e., 365 or 366). Then, the Sen's slope,  $z$ , is adjusted for estimating  
 185 trends in the timing:

$$z = \text{median} \left( \frac{\vartheta_j - \vartheta_i + c}{j - i} \right) \quad \text{with } c = \begin{cases} -\pi & \text{if } \vartheta_j - \vartheta_i > \pi \\ \pi & \text{if } \vartheta_j - \vartheta_i < -\pi \\ 0 & \text{otherwise} \end{cases} \quad \text{Eq. 2}$$

186

187 Where  $i$  and  $j$  indicate the year covering all possible pairs of years within the study  
 188 period, and  $i < j$ . The adjustment factor,  $c$ , aims to represent the shifting trend (i.e., to earlier or  
 189 later) based on the occurrence time at the  $i$ th year, when the differences between  $\vartheta_i$  and  $\vartheta_j$  are  
 190 larger than  $\pi$ . Finally, we analyzed the temporal shifts by different physiographic zones (i.e.,  
 191 windward and leeward), to examine the effects of different rainfall forming mechanisms.

192

193 **3. Results**

194 **3.1 Changes in maximum rainfall magnitudes**

195 Out of 84 rain gauges, the majority (67%) of rain gauges exhibited decreasing  $RF_{\max}$   
 196 magnitude (Kaua'i,  $n=9$ ; O'ahu,  $n=19$ ; Moloka'i,  $n=1$ ; Maui,  $n=15$ , and Hawai'i Island,  $n=12$ ).  
 197 The magnitude of  $RF_{\max}$  significantly decreased at seven rain gauges: one on O'ahu ( $p = 0.021$ ;  
 198 changed  $-1.5\%$  per year; mean = 195.2 mm), five on Maui ( $p < 0.01$ ; changed from  $-2\%$  to  
 199  $-4\%$ ; mean: 22.5 mm to 70 mm), and one on Hawai'i Island ( $p = 0.046$ ; changed  $-1.6\%$ ; mean  
 200 = 171.6 mm) (Figure 2). The remaining 33% of rain gauges showed increasing  $RF_{\max}$  trends  
 201 (Kaua'i,  $n=8$ ; O'ahu,  $n=6$ ; Moloka'i,  $n=2$ ; Maui,  $n=5$ , and Hawai'i Island,  $n=7$ ) (Figure 2).  
 202 Significant increasing trends only occurred at one gauge on Hawai'i Island ( $p = 0.047$ ) and  
 203 changed  $2.7\%$  per year with a mean of 151.4 mm. On Kaua'i, increasing  $RF_{\max}$  trends were

204 primarily located on the windward side, whereas there were no discernable differences in RF<sub>max</sub>  
205 between leeward and windward on the other islands. Decreasing trends in RF<sub>max</sub> dominated on  
206 Oahu, Maui, and Hawai‘i Island, while no particular trend direction prevailed on Kauai and  
207 Moloka‘i. However, there are too few rain gauges on Moloka‘i to address the spatial distribution  
208 in RF<sub>max</sub> trends.

209

210 3.2 *Changes in peak flow magnitudes*

211 Decreasing magnitude in PF<sub>max</sub> occurred at 60% of the crest gauges (Kaua‘i, n=8; O‘ahu,  
212 n=29; Moloka‘i, n=3; Maui, n=16, and Hawai‘i, n=11), predominantly on O‘ahu, Maui, and  
213 Hawai‘i islands (Figure 3). Statistically significant decreasing trends occurred at a single gauge  
214 on O‘ahu ( $p < 0.002$ ; decreased 6.5%; mean = 195.8 mm/day) and two on Maui ( $p < 0.011$ ;  
215 decreased 4.2%; means = 224.8 mm/day and 107.8 mm/day, respectively). Although decreasing  
216 PF<sub>max</sub> trends dominated across O‘ahu, some crest gauges still showed increasing trends, with two  
217 statistically significant ( $p < 0.05$ ) peak flow trends that rose 5.8% and 3.4% with means of 302  
218 mm/day and 80 mm/day, respectively. There was a general pattern of increasing PF<sub>max</sub> on the  
219 windward sides of Kaua‘i, Moloka‘i, and East Maui (Figure 3). Compared with RF<sub>max</sub>, PF<sub>max</sub>  
220 shows a greater physiographic division (i.e., windward and leeward) in trend directions.

221

222 3.3 *Associations between peak flow, annual maximum rainfall and paired rainfall*

223 When pairing rainfall and crest gauges, there are 18 pairs identified following the first  
224 criterion, and there are nine, eight, and four pairs found in the subsequent criteria, respectively.  
225 This resulted in a total of 39 pairs across islands (2 on Kaua‘i, 15 on O‘ahu, 1 on Moloka‘i, 14  
226 on Maui, and 17 on Hawai‘i Island). Some rain gauges were paired with multiple crest gauges as

227 they were the most representative rainfall of the given watersheds of crest gauges. This applied at  
228 13 crest gauges on O‘ahu (5 at Poamoho, 2 at Waiāhole, 4 at Pauoa Flats, 2 at Palolo valley);  
229 seven crest gauges on Maui (4 at Haleakala and 3 at Hana); and at four crest gauges on Hawai‘i  
230 Island (2 at Waiakea and 2 at Mauna Kea) (Figure 4).

231 The timing of  $RF_{max}$  rarely coincided with the timing of  $PF_{max}$ . Out of the 39 pairs of  
232 rainfall and crest gauges, almost half (46%) of the pairs have different trend directions between  
233  $RF_{max}$  and  $PF_{max}$  (Figure 4). Within the study period, when we extracted the same date rainfall  
234 with  $PF_{max}$ , 67% of the paired rainfall and  $PF_{max}$  records had consistent trends (increasing or  
235 decreasing). However, when we extracted the 2-day and 5-day accumulated rainfall with  $PF_{max}$ ,  
236 the percent of consistency decrease (Table 1).

237 Among the pairs having consistent trends, 56% of them showed decreasing trends,  
238 whereas the remaining 44% exhibited increasing trends (Figure 4). There was no clear leeward  
239 versus windward patterns in trend direction across the islands. Paired decreasing trends occurred  
240 on both windward and leeward sides of O‘ahu and Maui, and on windward Hawai‘i Island.  
241 Consistent increasing trends occurred on the leeward side of O‘ahu, windward side of Maui, and  
242 north of Hawai‘i Island.

243

#### 244 3.4 *Changes in the timing of annual maximum rainfall and peak flow*

245  $RF_{max}$  and  $PF_{max}$  often occurred during the wet season (Oct. – Apr.) with the median  
246 occurrence time of  $RF_{max}$  in earlier January and the median occurrence time of  $PF_{max}$  in mid-  
247 January for all gauges during the study period (Figure 5a, b). From 1970 to 2005, the occurrence  
248 time of both  $RF_{max}$  and  $PF_{max}$  shifted to an earlier time in the wet season, from late January to  
249 late December (Figure 5c, d). Yet, the occurrence time oscillated between earlier and later

250 throughout the four sub-periods (Figure 5c, d). The Sen's slope statistics also supported the  
251 temporal shifts in the study period. The Sen's slope estimator for the timing of the RF<sub>max</sub> and the  
252 PF<sub>max</sub> were  $-1.34^\circ$  ( $\sim 1.5$  days earlier) and  $-1.24^\circ$  ( $\sim 1.5$  days earlier with p-value  $< 0.05$ ) in  
253 leeward regions, respectively;  $-0.404^\circ$  ( $\sim$  half day with p-value  $< 0.05$ ) and  $0.231^\circ$  ( $\sim 6$  hours  
254 later with p-value  $< 0.05$ ) in windward areas, respectively.

255

256 *3.5 The relationships of annual maximum rainfall and peak flow to large-scale climate  
257 variability*

258 The magnitude of RF<sub>max</sub> and PF<sub>max</sub> showed no clear relationships to either the strength of  
259 ENSO (**Error! Reference source not found.**) or PDO (**Error! Reference source not found.**)  
260 indices. On the other hand, the median occurrence times of both RF<sub>max</sub> and PF<sub>max</sub> shifted earlier  
261 with a two to seven years cycle, regardless in windward or leeward areas (Figure 6 and 7), which  
262 were correlated to the cycle of ENSO. Specifically, the median occurrence times of RF<sub>max</sub> and  
263 PF<sub>max</sub> shifted to earlier during warm phase of ENSO (shaded in red in Figures 6 and 7). The  
264 ENSO's impact on median occurrence times was more pronounced in leeward regions, i.e., with  
265 12(8) recognized El Niño(La Niña) events, 9(5) RF<sub>max</sub> and 9(6) PF<sub>max</sub> occurred earlier(later) in a  
266 water year (Figure 6). However, there is no correlation between the strength of ENSO (i.e.,  
267 higher values of index) and the amount of shifting in occurrence times of RF<sub>max</sub> and PF<sub>max</sub>, with  
268 a correlation coefficient -0.15. Besides, the magnitude and occurrence time of RF<sub>max</sub> and PF<sub>max</sub>  
269 were not directly correlated to PDO (Figure B2).

270

271

272 **4. Discussion**

273 **4.1 Overview**

274 Changes in the extreme events are driven by complex atmospheric conditions and  
275 antecedent hydrological conditions, and it is unclear whether observed changes in rainfall and  
276 streamflow are due to natural atmospheric variability or global warming. Climate models have  
277 projected more intense rainfall under a warming climate and led to more severe flooding  
278 (Trenberth, 2011). The dataset presented here extends our knowledge of changes in both  
279 magnitude and occurrence time of  $RF_{max}$  and  $PF_{max}$  and their relationships to physiography and  
280 atmospheric circulation across the Hawaiian Islands. We analyzed temporal and spatial trends in  
281  $RF_{max}$  and  $PF_{max}$ , examined the association between  $RF_{max}$  and  $PF_{max}$ , and inspected the temporal  
282 shifts of  $RF_{max}$  and  $PF_{max}$ . Five key findings of this study were: 1) decreasing  $RF_{max}$  trends in  
283 more than half of gauges, particularly on O‘ahu and Maui; 2) decreasing  $PF_{max}$  trends in most  
284 gauges on O‘ahu, Maui and Hawai‘i Island, and more physiographic patterns in  $PF_{max}$  trends  
285 than  $RF_{max}$  trends; 3) different timing between the  $PF_{max}$  and the  $RF_{max}$ ; 4) shifting in the  
286 occurrence times of  $RF_{max}$  and  $PF_{max}$  to earlier, except for  $PF_{max}$  in windward areas, and 5) an  
287 effect of ENSO on the occurrence time of  $RF_{max}$  and  $PF_{max}$ . Our results highlight that changes in  
288  $RF_{max}$  are not sufficient to explain the observed trends in  $PF_{max}$ . In addition, the observed shifts  
289 in the occurrence time of  $RF_{max}$  and  $PF_{max}$  are important for flood risk and environmental  
290 management.

291

292

293 **4.2 Comparing trends in the Hawaiian Islands to other studies**

294 The predominantly decreasing trends of  $RF_{max}$  and  $PF_{max}$  shown in this study are

295 consistent with previous findings in Hawai‘i (Bassiouni and Oki, 2013; Chu et al., 2010; Clilverd  
296 et al., 2019). However, we did not find increasing  $RF_{max}$  trends on Hawai‘i Island as suggested  
297 by Chu et al. (2010). This difference may be in part due to the differences in the examined period  
298 (1970-2005 vs. 1950-2007), and it highlights cautiously using fixed periods of record as  
299 statistical significances of long-term trends (Frazier and Giambelluca, 2016). The majority of  
300  $PF_{max}$  stations showed declining trends across the state of Hawai‘i during our study period,  
301 particularly on O‘ahu and Hawai‘i Island. Changes in the magnitude of  $PF_{max}$  range from  $-6.5\%$   
302 to  $5.8\%$  per year.

303 In general, the Hawaiian Islands are experiencing decreasing  $RF_{max}$  and  $PF_{max}$ , whereas  
304 extreme rainfall has increased in some continental regions globally (Changnon and Kunkep,  
305 1995; Groisman et al., 2004; Hannaford and Marsh, 2008; Lins and Slack, 1999; Petrow and  
306 Merz, 2009). Madsen et al. (2014) indicated the trends of extreme precipitation and floods in  
307 Europe are heterogeneous. Beside regional factor, trends of extreme hydrological events may  
308 vary due to different definitions or methodologies to detect the trend, e.g., the maximum value  
309 over a period, count of precipitation days over a threshold, the percentile values, frequency, etc.  
310 (Cunderlik and Ouarda, 2009; Do et al., 2017; Douglas et al., 2000; Hannaford and Marsh, 2008;  
311 Ishak et al., 2013; Lins and Slack, 1999; Madsen et al., 2014; Petrow and Merz, 2009). Further,  
312 variation in the observed peak flow trends may be due to different causes or mechanisms that  
313 generate peak flow, such as intense rainfall, saturated soil, or snow melt (Blöschl et al., 2017;  
314 Cunderlik and Ouarda, 2009). Studies in tropical regions (i.e., Singapore (Beck et al., 2015) and  
315 India (Pingale et al., 2014)) only showed an increase in annual maximum daily precipitation but  
316 little has been done in examining the changes of peak flow and in other tropical islands. In  
317 Hawai‘i, decreasing  $PF_{max}$  may be related to decreased  $RF_{max}$ ; however, little research has

318 examined higher temporal resolution (i.e., subdaily) rainfall and its changes. Subdaily rainfall  
319 could better associate with peak flow as streamflow responds to rainfall within hours in Hawai‘i  
320 streams. In addition, the currently increasing drought condition across the Hawaiian Islands  
321 (Frazier et al., 2019) may also contribute to decreasing  $PF_{max}$  (decreasing the opportunity of  
322 saturation excess overland flow). This study adds to the knowledge of peak flow changes in  
323 tropical regions, especially on those tropical islands that are often frontline to the impact of  
324 changing climate and are likely vulnerable to the change in the precipitation patterns.  
325 Nevertheless, further studies with long-term data recording of peak flows and high-resolution  
326 rainfall will assist future assessments of their relationships and the mechanisms driving the  
327 observed across the islands.

328

329 4.3 *The inconsistency between annual peak flow and paired rainfall trends*

330 We found that  $PF_{max}$  rarely occurred at the same time as or soon after  $RF_{max}$ , and only  
331 64% of paired gauges showed consistent trend directions between daily rainfall and the peak  
332 flow. The pairs with inconsistent trends may be an artifact of the limited availability of paired  
333 rain gauges, as pairs showing inconsistent trends were mainly located in different watersheds or  
334 sub-watersheds to each other. However, some of the inconsistency cannot be explained by the  
335 spatial mismatch, for example, the Poamoho rain gauge on O‘ahu best represented the rainfall on  
336 Ko‘olau Mountain for nearby watersheds, but interestingly three out of four nearby crest gauges  
337 were inconsistent with its rainfall trend. The inconsistency between  $PF_{max}$  and the paired rainfall  
338 may be due to that rainfall does not always lead to the peak flow, and additional factors in the  
339 runoff processes need further investigation. These watersheds and their associated hydrological  
340 characteristics may have local-scale patterns that vary the peak flow generation. Wasko and

341 Sharma (2017) suggested that disagreement between extreme rainfall and streamflow may be  
342 attributed to antecedent soil moisture conditions for larger watersheds. In contrast, in our  
343 analysis, we found only small differences when considering antecedent rainfall (as indication of  
344 antecedent soil moisture conditions) and annual maximum peak flow. In our study, it was  
345 challenging to demonstrate how sensitive the  $PF_{max}$  is to antecedent soil moisture in Hawai‘i by  
346 comparing with paired 1-day, 2-day, or 5-day accumulated rainfall. There are two possible  
347 reasons for this: 1) the  $PF_{max}$  is not sensitive to antecedent soil moisture likely due to the small  
348 watersheds with shorter concentration time in Hawai‘i, or 2) the 5-day accumulated rainfall is  
349 insufficient to indicate the soil moisture maximum. The pairing methods adopted in this study  
350 have improved our understanding and representation of rainfall events that cause peak flows.  
351 However, it remains unclear why some of the pairs exhibit negative trend relationships, and thus  
352 further investigation is required. The change in  $RF_{max}$  cannot be conclusively applied as a proxy  
353 for the change in  $PF_{max}$ , and we must be cautious in predicting changes in extreme streamflow  
354 based on changes in extreme rainfall alone.

355

356 4.4 *The shifts in seasonality of annual maximum rainfall and annual peak flow*

357 The occurrence time of  $RF_{max}$  and  $PF_{max}$  has receded from late January to December in  
358 Hawai‘i (Figure 5). In addition to the timing of winter storms, these temporal shifts in peak flow  
359 may be due to the soil moisture conditions. In continental or large watersheds ( $> 2000 \text{ km}^2$ ),  
360 changes in antecedent soil moisture play a notable part in temporal shifts of floods (Blöschl et  
361 al., 2017; Merz and Blöschl, 2003; Wasko and Sharma, 2017). Frazer and Gaimbelluca (2016)  
362 showed that Hawai‘i had experienced drying trends from 1920-2012. However, it is unclear how  
363 drying conditions have impacted the timing of maximum soil moisture and the occurrence time

364 of  $PF_{max}$ . Shifting climate patterns may also have altered atmospheric circulations and the timing  
365 of winter rainfall (Blöschl et al., 2017; Diaz et al., 2016, 2001), which would change the timing  
366 of  $RF_{max}$  and then  $PF_{max}$ . To our knowledge, no other studies have addressed temporal shifts of  
367  $RF_{max}$  and  $PF_{max}$  in Hawai‘i; hence this study provides an important contribution to our  
368 understanding of shifting hydrological regimes in Hawai‘i. It also adds further information and  
369 raises the awareness of temporal shifts in both  $RF_{max}$  and  $PF_{max}$ .

370 The occurrence time of  $RF_{max}$  in both windward and leeward regions exhibited 2 to 7  
371 years oscillations (Figures 5–7). These oscillations may be caused by ENSO or other large  
372 atmospheric cycles. Previous running trend analyses of the magnitude of monthly rainfall,  
373 extreme precipitation, and streamflow have also identified distinct associations with atmospheric  
374 oscillations (Chu et al., 2010; Clilverd et al., 2019; Frazier and Giambelluca, 2016). These  
375 oscillations are most likely due to natural coupled oceanic and atmospheric cycles, such as the  
376 ENSO. Chu et al. (2010) suggested that more extreme precipitation in the Hawaiian Islands  
377 occur during La Niña years, while fewer extreme precipitation events happen during El Niño  
378 years. However, we found no pronounced relationship between ENSO and the magnitude of  
379  $RF_{max}$  and  $PF_{max}$ , with possible bias due to different methods and study periods. Instead, we  
380 discovered that the occurrence time of both  $RF_{max}$  and  $PF_{max}$  likely depends on the ENSO phase  
381 regardless of its strength. It is unsurprising that ENSO impacted the occurrence time of both  
382  $RF_{max}$  and  $PF_{max}$ . Studies show that ENSO events delayed the starting time of monsoon in Asia  
383 (e.g., Joseph et al., 1994; Wang et al., 2013). Different El Niño properties might have led to  
384 different starting time of El Niño and thus on the timing of rainfall (Xin Wang et al., 2013; Bin  
385 Wang et al., 2019). Different mechanisms in occurrences of El Niño (e.g., the El Niño that had  
386 earlier starting time and occurred more frequent from 1901 to 2017 (Wang et al., 2019)) possibly

387 contribute to the shift of occurrence time of  $RF_{max}$  and  $PF_{max}$ . The shifts in occurrence time have  
388 important ecological implications – for instance, the timing of flash floods triggers the migration  
389 of native freshwater fish, o‘opu, from their coastal nursery grounds to core stream habitats where  
390 they mature and reproduce (Fitzsimons and Nishimoto, 1995; Radtke and Kinzie, 1996).  
391 Therefore, shifts in the timing of these migratory cues could have consequences for the dispersal  
392 and life cycles of native stream species across the Hawaiian Islands. We need further research on  
393 the sensitivity of peak flow to natural climate oscillations and climate change, and the influence  
394 on the magnitude and seasonality of  $RF_{max}$  and  $PF_{max}$ . This could be addressed with new stream  
395 gauges in watersheds, a more extensive rainfall gauge network, and support for long-term  
396 records, that would provide more effective streams and rainfall pairs needed to further our  
397 understanding of the association between peak flow and rainfall. These would improve our  
398 understanding of the drivers of observed changes in seasonal peak flow and improve our ability  
399 to predict future flooding events across the Hawaiian Islands, and the potential impacts on  
400 humans and the ecosystem.

401

402 *4.5 Conclusions*

403 We have gained an improved understanding of local changes in peak streamflow and  
404 responses to changes in rainfall. Despite limitations in the availability of long-term data, our  
405 study demonstrated that both annual maximum rainfall and annual peak flow have declined;  
406 local responses of both extreme rainfall and streamflow were varied, and changes in peak  
407 streamflow do not necessarily follow the changes in maximum rainfall. In addition, we conclude  
408 that spatial linkages between rainfall and streamflow gauges are important when studying the  
409 relationship between peak streamflow and maximum rainfall. Also, the temporal shifts in peak

410 streamflow and rainfall should be considered in understanding changes in extreme events.

411 Examining the inconsistency between the timing of peak streamflow and annual maximum

412 rainfall is much needed, and investigating the mechanisms that cause annual maximum rainfall

413 or heavy rainfall could shed light on the subsequent results in temporal shifts in peak streamflow.

414

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420

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644

645 Table 1 Trends of annual peak flow and trends of 1-day, 2-day, and 5-day accumulated rainfall of the same date as the annual peak flow for each pair (see figure 4  
 646 for the spatial distribution) between 1970 and 2005. Trends are shown in percent change per year, and the consistent trends (with the same sign) between peak  
 647 flow and paired rainfall are shaded in grey. (OA = O‘ahu, KA = Kaua‘i, MO = Moloka‘i, MA = Maui, HA = Hawai‘i, and ‘\*’: p-value < 0.05, ‘\*\*’: p-value <  
 648 0.01, ‘\*\*\*’: p-value < 0.00)

Island	Rain Gauge	Crest gauge	peak flow (%)	1-day rainfall (%)	2-day rainfall (%)	5-day rainfall (%)
KA	USC00515560	16060000	0.38	-4.48*	-5.87	27.01
KA	USC00514561	16097900	-0.3	-1.41	-4.21	-1.8
OA	X213215157552800	16200000	0	-1.73	-1.17	-1.17
OA	X213215157552800	16208000	-0.43	-2.55	-2.45	-2.45
OA	X213215157552800	16212700	1.28	-0.94	-1.12	-1.12
OA	USC00518964	16212800	0.6	1.29	1.14	1.14
OA	USC00517810	16232000	0.57	0.04	-0.23	-0.23
OA	USC00517810	16237500	-1.48	-1.66	-2.56	-2.56
OA	USC00517810	16238500	2.45	1.03	0.34	0.34
OA	USC00517810	16240500	0.12	0.15	-0.07	-0.07
OA	USC00517664	16244000	5.77*	2.57	1.79	1.79
OA	USC00517664	16247100	0.26	0.11	-0.34	-0.34
OA	USC00519523	16249000	-0.6	-0.57	-1.13	-1.13
OA	USC00513117	16274499	-0.91	-3.2	-3.07	-3.07
OA	USC00518964	16294900	-0.77	0.63	4.84	4.84
OA	X213215157552800	16296500	0.77	-1.65	-1.17	-1.17
OA	X213215157552800	16301050	0.67	-0.13	0	0
MO	USW00022534	16411800	-6.54	18.74	19.13	13.55
MA	USC00511125	16502400	0.4	0.22	0.62	-0.36
MA	USC00511125	16502800	1.41	-0.55	-0.5	0.08
MA	USC00511125	16502900	1.25	1.05	1.23	1
MA	USC00511004	16587000	0.59	-1.07	-0.44	-1.73
MA	USC00511004	16603700	-4.58*	-1.13	-0.69	-1.22
MA	USC00511004	16603800	-1.59	-1.54	-1.94	-1.69

MA	USC00518543	16603850	2.41	2.41	2.9	-2.36
MA	USC00515715	16630200	-0.21	-5.49*	-8.23	-11.49
MA	USC00515177	16638500	1.75	-2.67	-0.83	0
MA	USC00515408	16643300	-0.13	-3.78	-5.19	-4.89
MA	USC00517059	16646200	-0.28	-3.66	-2.24	-1.92
MA	USC00518060	16647500	-0.04	-0.56	-1.49	-0.55
MA	USC00511004	16658500	2.69	2.34	2.15	-3.11
MA	USC00514489	16659000	-0.85	6.02	8.83	7.88
HA	USC00519025	16701300	1.61	-0.83	1.18	-2.1
HA	USC00519025	16701400	-8.15	-6.45	-1.86	-8.75
HA	USC00511065	16704000	-1.08	-3.81	-0.22	0.63
HA	USC00519025	16717000	-0.46	1.51	1.45	1.06
HA	USC00517312	16717850	-33.11	-24.6	-9.9	-5.39
HA	USC00514680	16752600	9.16	1.26	11.95	17.78
HA	USC00513300	16770500	0.99	-3.74	3.54	2.03

650 **Figures**

651 Figure 1 Map of crest (circles) and rain (squares) gauges in different physiographic zones –  
652 windward (green), leeward (brown), and unclassified (transparent) superimposed on the  
653 digital elevation model (DEM) obtained from the U.S. Geological Survey across five major  
654 Hawaiian Islands; the brown line indicates the physiographic division of windward  
655 (northeast) and leeward (southwest) regions on the islands. .... 35

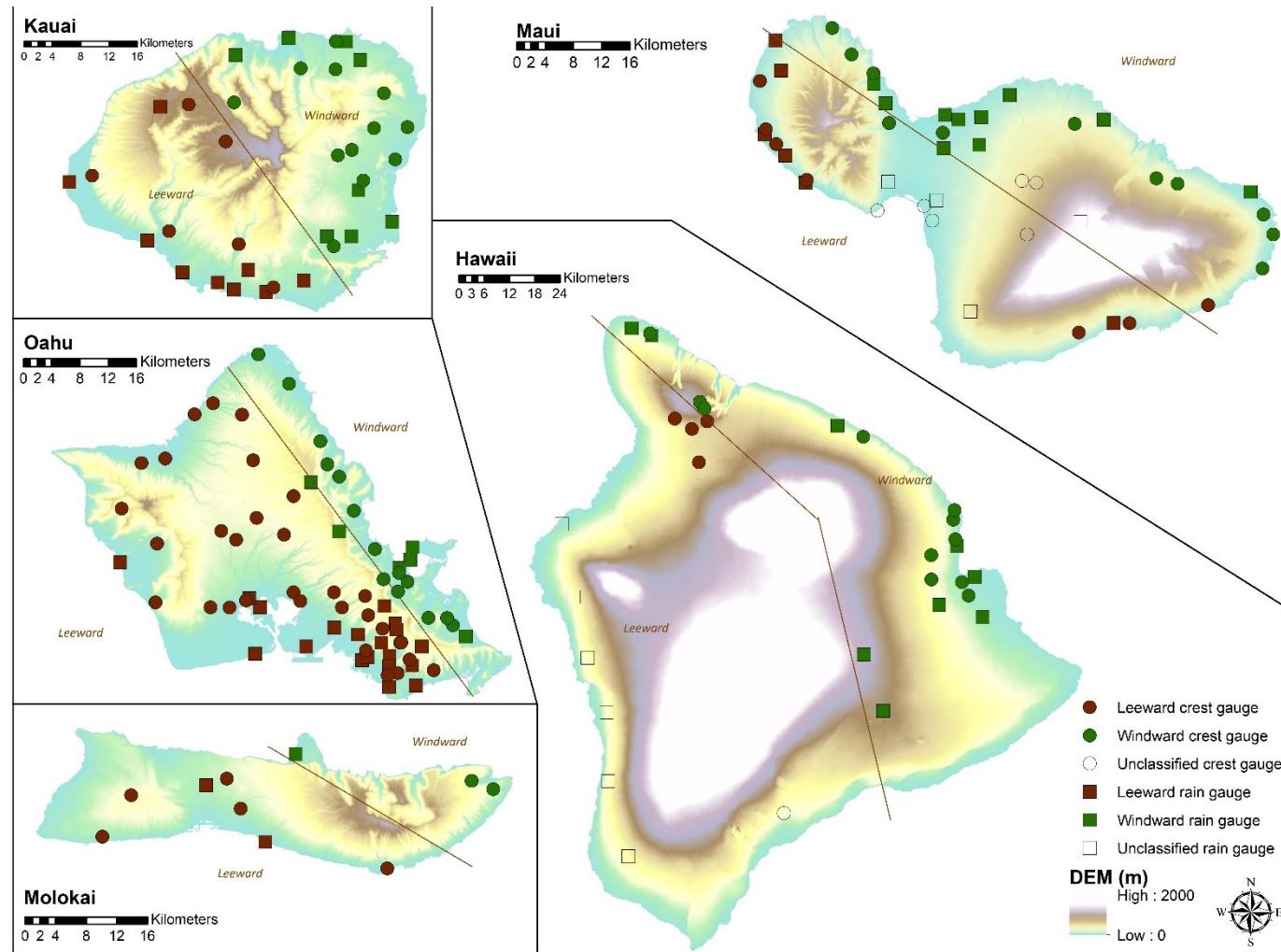
656 Figure 2 Annual maximum daily rainfall trends from water years 1970 to 2005 across five major  
657 Hawaiian Islands, superimposed on the DEM obtained from the U.S. Geological Survey.  
658 Blue inverted arrows show decreasing trends, red arrows show increasing trends, and black  
659 dots indicate significant trends ( $p < 0.05$ ). The trends denote percentage change per year. The  
660 brown line indicates the physiographic division of windward (northeast) and leeward  
661 (southwest) regions on the islands. .... 36

662 Figure 3 Annual peak flow trends from the water year 1970 to 2005 for the five main Hawaiian  
663 Islands. Blue inverted triangles show decreasing trends, the orange triangles show increasing  
664 trends, triangles with a black dot indicate significance ( $p < 0.05$ ). The trends indicate  
665 percentage change per year. The color shading is the elevation (m). The brown line indicates  
666 the physiographic division of windward (northeast) and leeward (southwest) regions on the  
667 islands. .... 37

668 Figure 4 Paired daily rainfall and annual peak flow trends that showed agreement in the direction  
669 of trends from 1970 to 2005 for five major Hawaiian Islands, superimposed on the DEM  
670 obtained from the U.S. Geological Survey. Brown lines indicate the watershed boundaries.  
671 The paired daily rainfall gauge (black square) and the crest gauge (triangles) exhibit the same  
672 trend (peak flow trend with agreement; blue triangle: decreasing peak flow trends and can be

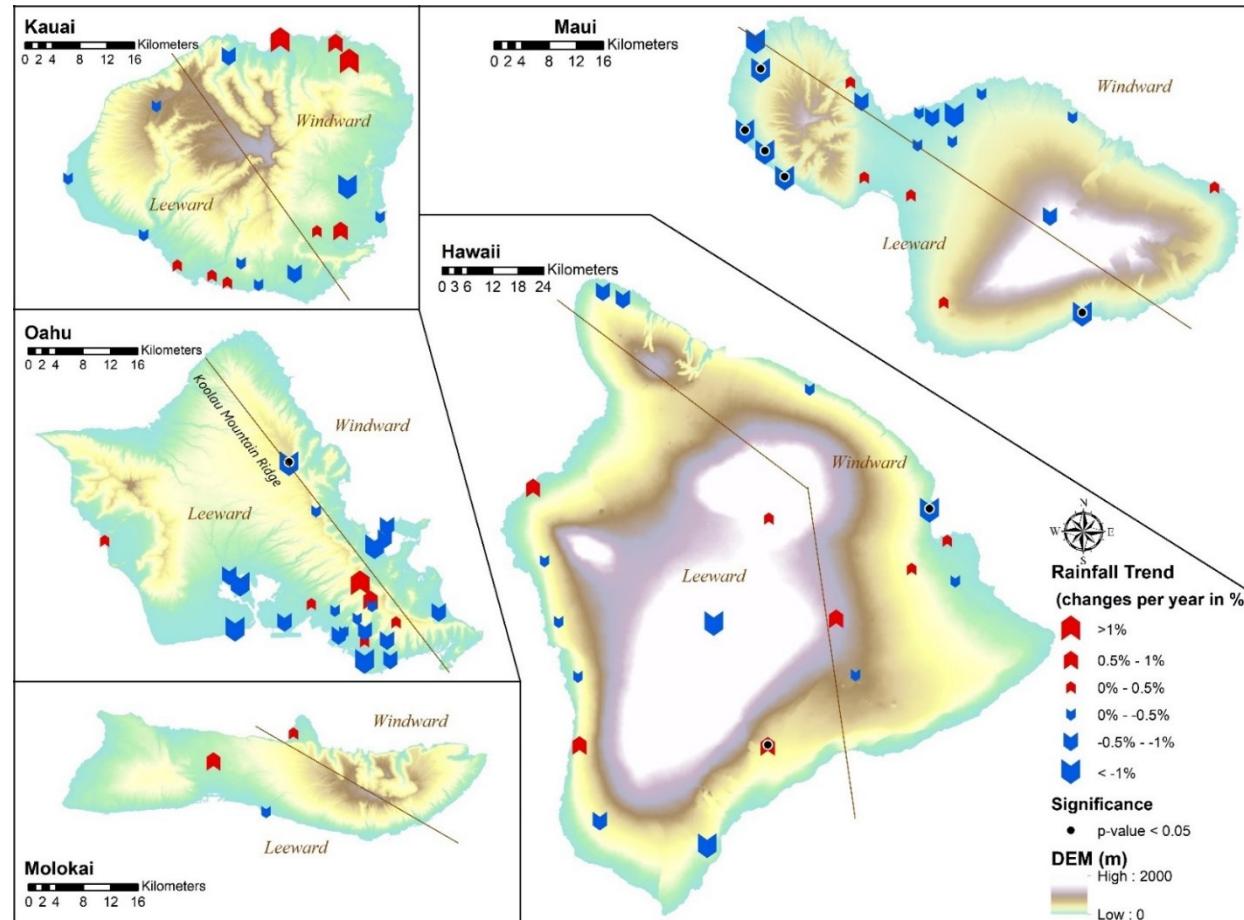
673	referred to paired daily rainfall; orange triangle: increasing trends of peak flow and can be	
674	referred to paired daily rainfall). Brown circles show the crest gauges that did not have	
675	consistent trends with the paired rainfall (peak flow trend w/o agreement).....	38
676	Figure 5 Comparison of the occurrence time of annual maximum daily rainfall (n=85 gauges) for	
677	(a) the analysis period, 1970-2005, and (c) every nine years from 1970-2005, with the	
678	occurrence time of annual peak flow (n=112 gauges) for (b) the analysis period, 1970-2005,	
679	and (d) for every nine years from 1910-2005, across the Hawaiian Islands. The dashed line is	
680	the estimated kernel density, and the arrow points to the median date of occurrence with	
681	mean resultant length. For (a) and (b), the gray dots indicate the frequency of the occurrence	
682	time. For (c) and (d), the colors reflect the time period, and the circles around the clock	
683	indicate the occurrence time. ....	39
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685	(green line; n=42) and annual peak flow (black line; n=54) on the leeward side with the	
686	ENSO phase shaded (red: warm phase; blue: cool phase, white: neutral). The y-axis indicates	
687	the months from June through May (bottom to top). Occurrences under ENSO warm phase	
688	are indicated by red squares for annual maximum daily rainfall, and a red circle for annual	
689	peak flow. The Oceanic Nino Index (ONI) on the top indicates the strength of the ENSO	
690	from 1970 to 2005.....	40
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692	(green line; n=34) and annual peak flow (black line; n=49) on the windward side with the	
693	ENSO phase shaded (red: warm phase; blue: cool phase, white: neutral). The y-axis indicates	
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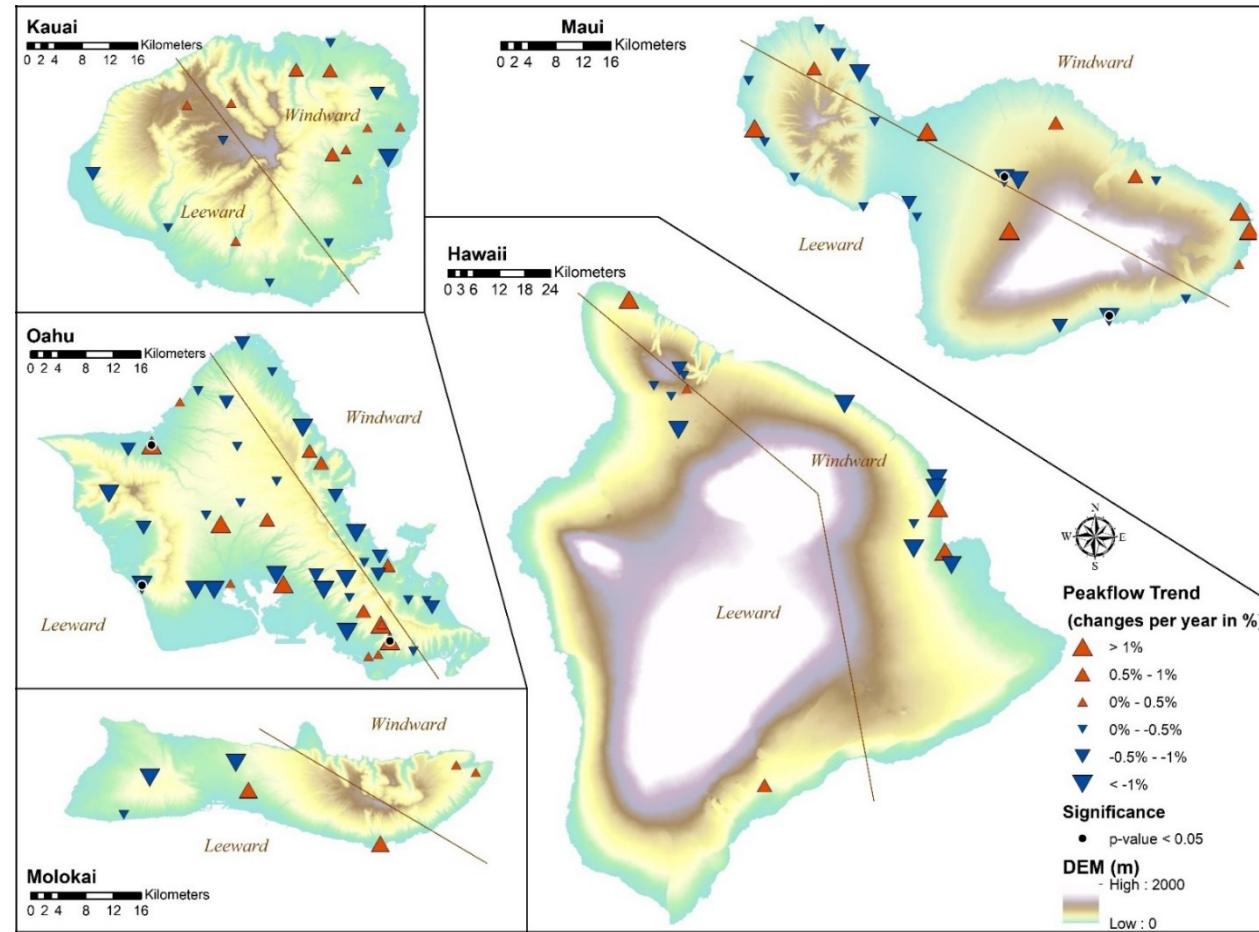


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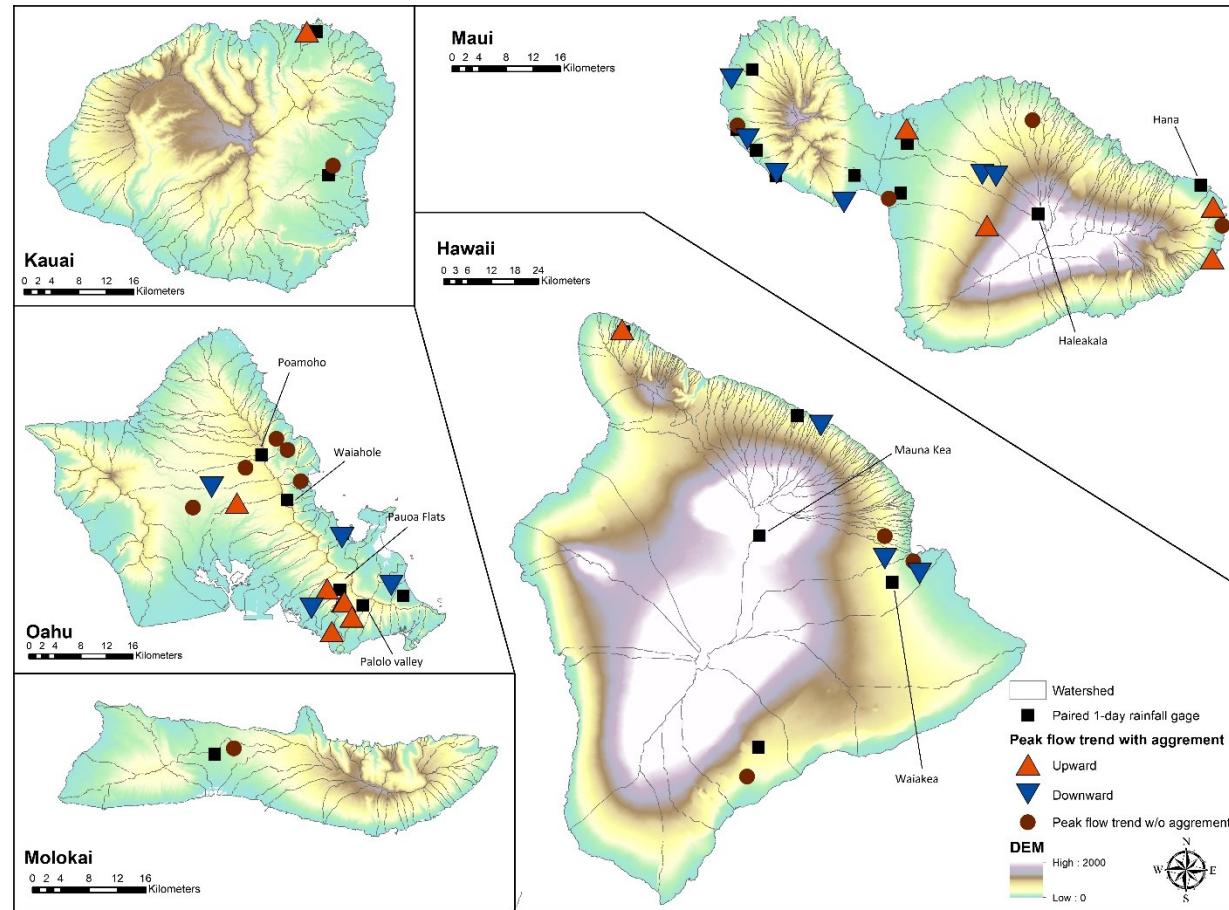


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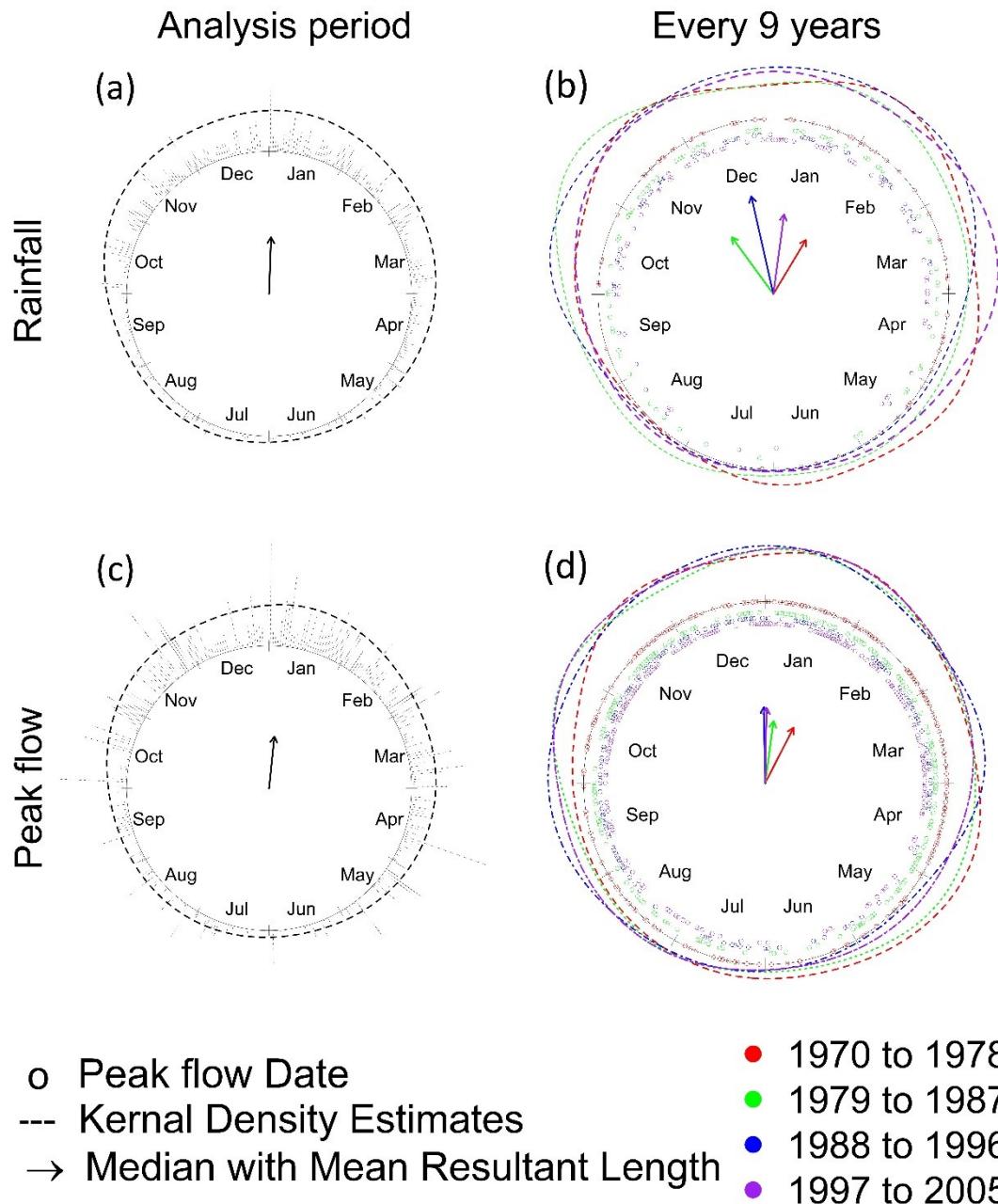
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709 Figure 3 Annual peak flow trends from the water year 1970 to 2005 for the five main Hawaiian Islands. Blue inverted triangles show decreasing trends, the  
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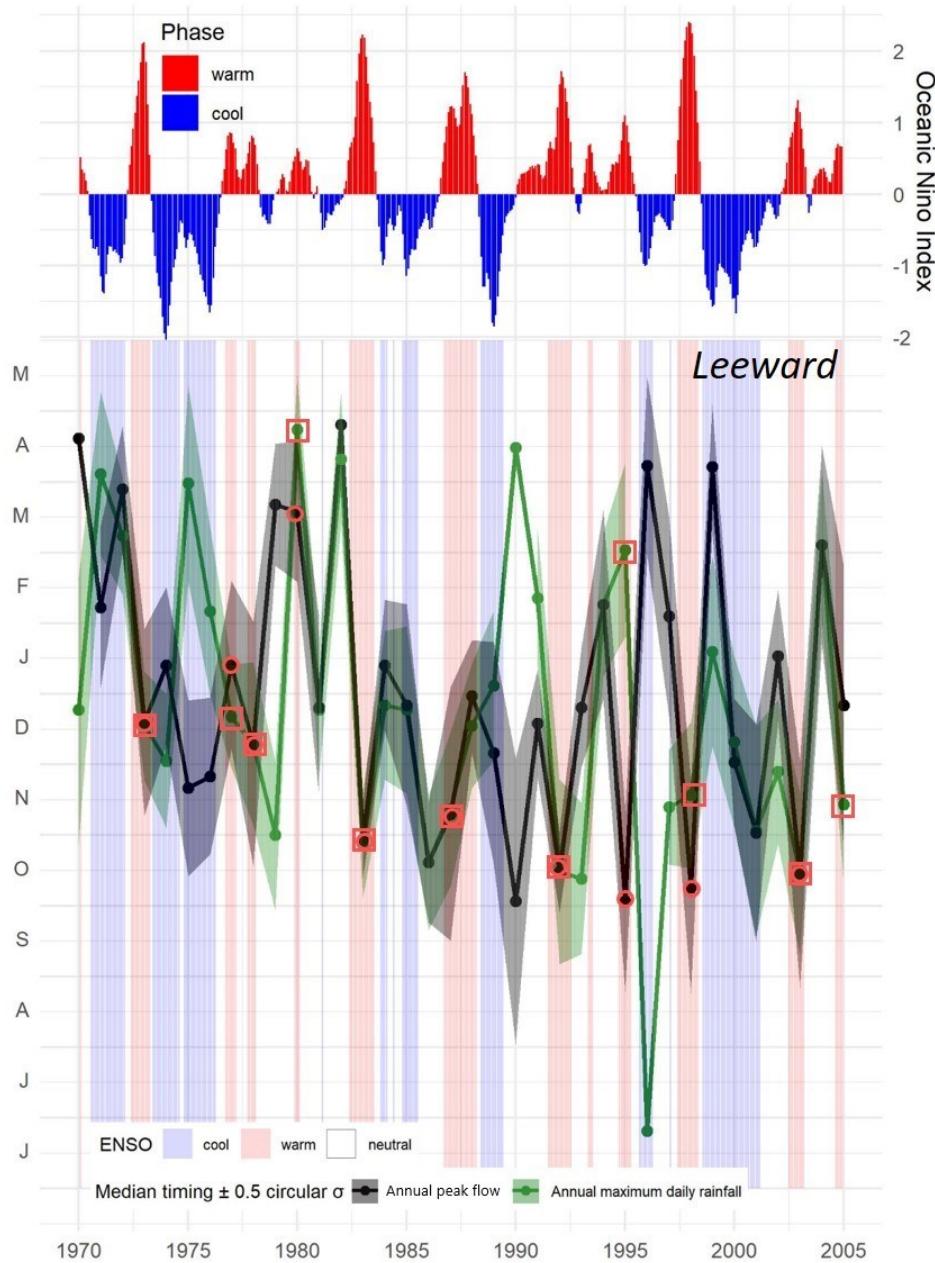
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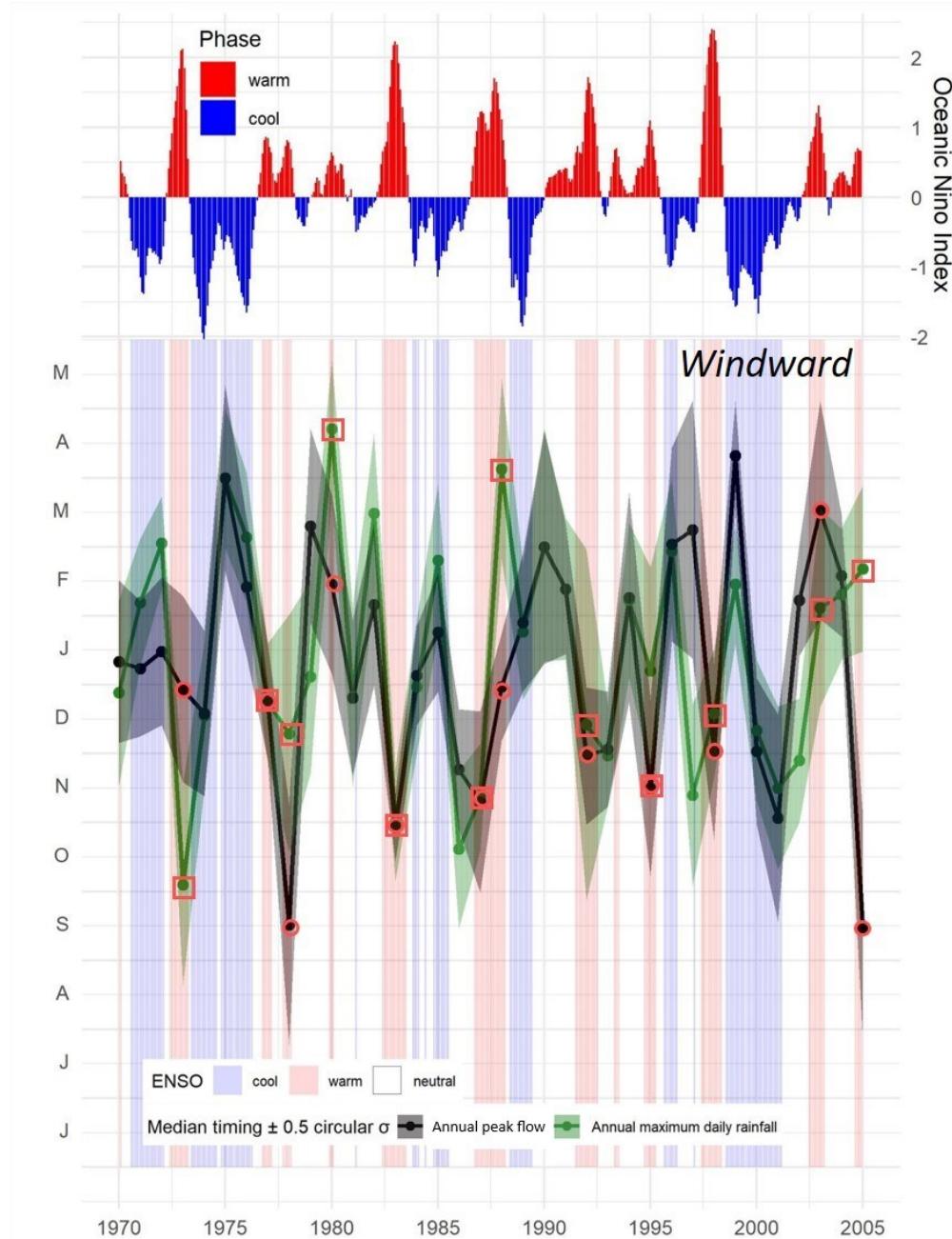
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725

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731

732 Figure 7 The occurrence time (median  $\pm 0.5$  circular deviation) of annual maximum daily rainfall (green line; n=34)  
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