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Burning in southwestern Brazilian Amazonia, 2016-2019

Abstract

Fire is one of the most powerful modifiers of the Amazonian landscape and knowledge about its drivers is needed for planning control and suppression. A plethora of factors may play a role in the annual dynamics of fire frequency, spanning the biophysical, climatic, socioeconomic and institutional dimensions. To uncover the main forces currently at play, we investigated the area burned in both forested and deforested areas in the outstanding case of Brazil's state of Acre, in southwestern Amazonia. We mapped burn scars in already-deforested areas and intact forest based on satellite images from the Landsat series analysed between 2016 and 2019. The mapped burnings in already-deforested areas totalled 550,251 ha. In addition, we mapped three forest fires totalling 34,084 ha. Fire and deforestation were highly correlated, and the latter occurred mainly in federal government lands, with protected areas showing unprecedented forest fire levels in 2019. These results indicate that Acre state is under increased fire risk even during average rainfall years. The record fires of 2019 may continue if Brazil's ongoing softening of environmental regulations and enforcement is maintained. Acre and other Amazonian states must act quickly to avoid an upsurge of social and economic losses in the coming years.

Keywords: Amazon; fires; deforestation; droughts, Acre, El Niño

1. Introduction

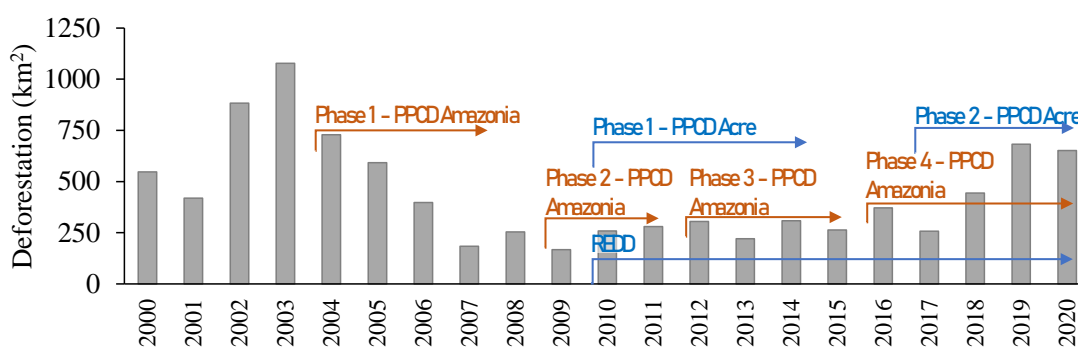
Amazon fires are associated almost exclusively with human activities (Barlow et al., 2019). These fires vary across space and time with changes in land use and cover. These changes are driven by complex interactions among factors such as governance (or lack thereof), international trade, the domestic land market and local climate (Barlow et al., 2019; Tasker and Arima, 2016). During unusually dry and hot years, accidental and illegal fires tend to escape from agricultural fields into standing old-growth, secondary and degraded forests. Although fires affect large areas of the Amazon, there is high variability in fire activity across the Basin (Aragão et al., 2007). This is partially explained by regional heterogeneity in the economic and biophysical factors that drive fires and in the regulatory measures that constrain fires. Understanding how fire activity has changed spatially and temporally across the Amazon is useful for improving fire policy effectiveness, including both prevention and suppression of forest fires.

Although periods of high precipitation seasonally dampen fire activity, droughts are becoming common in the region (Jiménez-Muñoz et al., 2016). Forest fires burn larger areas during droughts, especially when deforestation rates are high (Aragão et al., 2008). One study estimated that if deforestation continues to claim Amazon forests, up to 16% of southern Amazonian forests may burn in the near future (Brando et al., 2020). Projections for the end of the century in a land-use scenario with high forest fragmentation indicate that increase by up to 73.2%, mainly in the southern portion of the Amazon (Fonseca et al., 2019).

Fires are used as a tool for eliminating the felled trees in recently deforested areas, in clearing secondary forest or in renewing pasture (Barlow et al., 2019; Dias Filho, 2011). When these fires escape from control in years of extreme drought, such as 2005, 2007, 2010 and 2015/2016, they can cause large-scale forest fires (Alencar et al., 2015; Anderson et al., 2015; Morton et al., 2013; Silva et al., 2018). Even when there were

51 reductions in deforestation fires, there was still enough burning activity to generate
 52 large-scale forest fires during drought years (Aragão et al., 2018). Recent informal
 53 statements by politicians at the federal and state levels attest to the reduction of
 54 enforcement investment, which appears to have led to a significant increase in
 55 deforestation and fires in the Amazon (Thomaz et al., 2020). This process culminated
 56 with the 2019 Amazon fire crisis (Barlow et al., 2019), leading to a presidential decree
 57 prohibiting fires and allowing the use of the army for law enforcement (Brazil, 2019).
 58 However, this did not reduce burning and contributed to further weakening of IBAMA,
 59 the federal environmental agency (Ferrante and Fearnside, 2020; OC, 2020). Weakening
 60 environmental regulations and agencies leads to an increase in the area burned in
 61 association with the return of high rates of deforestation in the Amazon, as was
 62 observed during the first 6-months of 2020 (INPE, 2020a).

63 Acre, which is in the 5th position in the deforestation ranking of Brazil's nine
 64 Amazonian states, has a solid history of forest conservation, for which it was granted
 65 the first jurisdictional REDD+ program in the world (Acre, 2013). This important
 66 leadership is being threatened by a substantial increase in deforestation and fires since
 67 2007 (INPE, 2020b), where the Action Plan for Prevention and Control of Deforestation
 68 in Amazonia has not prevented the resumption of deforestation in Acre in recent years
 69 (Figure 1). The state is located in the southwestern Brazilian Amazon, and more than
 70 84% of its ~ 164,000-km² area is under old-growth forests (INPE, 2020c), with 46% of
 71 the forest area protected by conservation units (Acre, 2010). In recent years, the advance
 72 of the agricultural frontier in the “arc of deforestation” (Fearnside, 2005), makes Acre a
 73 focus for land speculation, contributing to a significant increase in deforestation. Acre
 74 was the epicenter of two recent mega-droughts, in 2005 and 2010 (Lewis et al., 2011).
 75 The state is among the ten poorest of Brazil, with approximately 40% of its citizens
 76 below the poverty threshold (IBGE, 2019), but much of the deforestation is done by
 77 wealthy ranchers. This scenario of deforestation, droughts and the weakening of public
 78 policies, contributes to inefficient environmental management that leads to socio-
 79 economic and environmental conflicts with exacerbation of inequality and increases
 80 burning by rural actors.



81
 82 **Fig. 1** Annual deforestation rates PRODES / INPE in the State of Acre indicating
 83 Action Plan for Prevention and Control of Deforestation in Amazonia (MMA, 2016)
 84 and State of Acre and jurisdictional REDD + program (Acre, 2018) .
 85

86 To understand the dynamics of fire, it is essential to analyze its spatial and
 87 temporal distributions and also to disentangle forest fires from burning in already-
 88 deforested areas. The main type of near real-time satellite data available for this purpose
 89 is the so called “hot pixels,” which indicate the location at which fires occur but do not

90 allow estimation of the areal extent of the burns. Global remote-sensing products for
91 burned areas are also available, but these underestimate fire-affected areas in dense
92 tropical forests (Anderson et al., 2017; Pessôa et al., 2020), and detailed maps have only
93 been produced for restricted spatial domains and/or time periods (Alencar et al., 2015;
94 Anderson et al., 2015). Other estimates have spatial resolution that does not allow
95 detecting the dynamics of small fires (INPE, 2020d; Morton et al., 2013).

96 In order to provide novel information on fire dynamics in the southwestern
97 Amazon, which is a region that has recently been impacted by severe droughts and
98 where there is a paucity of information on fire use, we investigated the interconnection
99 between deforestation, agricultural burning and forest fires. We also explored the
100 relationships of these phenomena with climate.

102 **2. Methods**

103 2.1. Study area

104 Acre State has an area of 16,423,979 ha and is located in the southwestern part of
105 the Brazilian Legal Amazon (Figure 2a). According to the Köppen classification
106 system, local climate is Af (without dry season) and Am (monsoon), with average
107 annual temperatures between 22 °C and 26 °C and annual precipitation between 2200
108 mm and 2500 mm. By 2019, the state had 2,259,990 ha (14%) of its territory deforested
109 (INPE, 2020c). Data from the TerraClass project show that deforested areas in Acre are
110 normally occupied by cattle pasture (67%) and secondary forests (areas abandoned after
111 use for agriculture or pasture) (EMBRAPA, 2017). Acre experienced extreme-drought
112 events in 1998, 2005, 2010 and 2016, with maximum cumulative water deficits of up to
113 300 mm (Aragão et al., 2007; Silva et al., 2018).

115 2.2. Mapping of burning in already-deforested areas

116 In this study, burning in already-deforested areas was defined as fire scars in areas
117 without native forest that are covered by pasture, agriculture or bare ground in areas of
118 recently deforested native or secondary vegetation. The burned areas mapping were
119 based on supervised classification of Landsat 8 Operational Land Imager (OLI) satellite
120 images from 2016 to 2019. Three images per year were used for the 14 scenes covering
121 the state of Acre to encompass the entire burning season, from July to September
122 (Supplementary Material, Table S1). The choice of several images to represent the year
123 is due to the rapid disappearance of the scars from the fires, which occurs between three
124 to four weeks after the fire event.

125 We used the supervised minimum-distance classification method with cloud
126 processing on the Google Earth Engine based on reflectance information from the
127 Landsat 2, 3, 4, 5 and 6 spectral bands. This classifier calculates the spectral distance
128 between the measurement vector for the candidate pixel and the average for each class
129 signature. The classifier compares the Euclidean distance between the value for each
130 pixel and the average for each cluster. Four classes were used: intact forest, water,
131 deforestation and burn scar, with at least 20 samples per class.

132 After the supervised classification, the minimum mapped area was defined as 0.5
133 ha, representing five contiguous pixels. Areas smaller than this size were excluded from
134 the analysis because they have less reliability due to the spatial resolution of the sensor.
135 The burn-scar mapping was audited by manual adjustment or elimination of area that
136 presented confusion with other targets, such as water bodies or deforestation. The audit
137 was carried out by a team of four people, with the last stage being carried out by a
138 specialist in the fire dynamics of Amazonian landscapes and in remote sensing.

139 The validation of the mapping of burn scars was based on field points and random
140 points. Twenty nine field points were collected between August 2nd and 28th 2019 along
141 federal highway BR 364. The random points were distributed between unburned and
142 burned areas, totaling 1000 and 1500 points, respectively. These points were verified by
143 experienced interpreters (Figure 2b). Assessment of the overall accuracy of the
144 classification and estimation of errors of omission and commission were performed
145 using an error matrix as proposed by Anderson et al. (2017).
146

147 2.3. Mapping of forest fires

148 Forest fires were defined in this study as those in which the crowns of the trees
149 were directly or indirectly affected by fire to the point that they cause a detectable
150 impact on the optical satellite images, representing the scar left by the fire (Silva et al.,
151 2018). These mapping procedures represent a continuity of the study performed by
152 Silva et al. (2018) and are based on image processing of the Landsat series using the
153 mixing model produced by CLASlite software. This software uses a spectral-mixing
154 model associated with a robust spectral library to generate the following fractions:
155 photosynthetically active vegetation, non-photosynthetic vegetation and soil. The dates
156 of the images used for processing are from September to December (Supplementary
157 Material, Table S2).
158

159 2.4. Analysis of the spatio-temporal patterns of burning in already-deforested areas

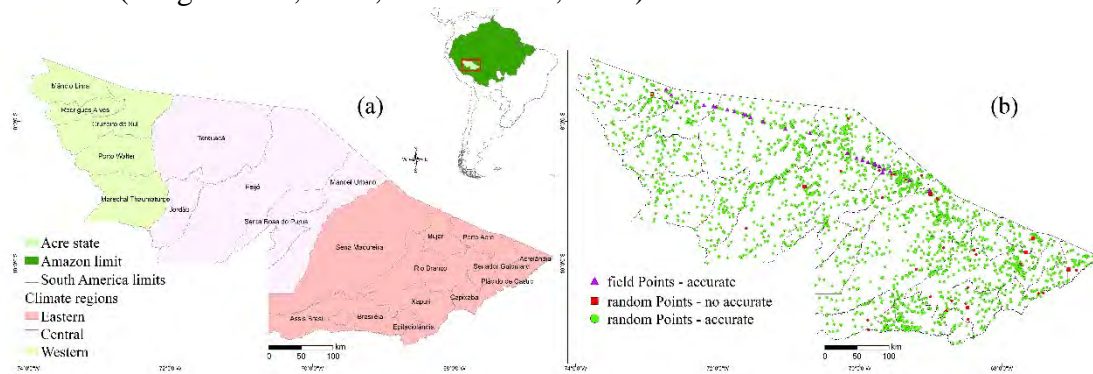
160 For the study period (2016 to 2019), the total burning in already-deforested areas
161 was quantified by year and by recurrence. For each year, we quantified the size of the
162 mapped areas of fire using seven classes: (0.5 - 2, 2 - 5, 5 - 10, 10-25, 25-50, 50-100,
163 ≥ 100 ha). These analyses allow us to understand the magnitude and patterns of burning
164 in already-deforested areas.

165 To uncover factors correlated with burnings, we categorized data according to
166 land-tenure categories such as settlement projects, undesignated public land, private
167 properties, conservation units and indigenous land) (Acre, 2010). We applied analysis
168 of variance with Levene's test and a post-hoc Tukey's test to evaluate the null-difference
169 hypothesis between the means for burning in already-deforested areas and in new
170 deforestation. This analysis helps clarify the use of fire to advance deforestation in areas
171 in different land-tenure categories.

172 The relationship between deforestation and burning in already-deforested areas
173 was assessed using data from the PRODES Project (INPE, 2020c). The "PRODES
174 year" used for deforestation estimates refers to the period from August 1st of the
175 previous year to July 31st of the nominal year (i.e., "2019" refers to August 1 2018 to
176 July 31, 2019). Based on these data, we performed three analyses: (I) quantification of
177 the proportion of the total annual area of fires that came from the new annual
178 deforestation, for example, we account for the burned area that occurred in the 2019
179 PRODES year that was not detected as deforestation by PRODES 2018/2019, (II)
180 quantification of the proportion of the total annual area of fires that came from the
181 management of deforested areas consolidated in previous years, for example, burned in
182 2019 that had been detected as deforestation by PRODES 2018/2019, (III) analysis of
183 the correlation between annual burning in already-deforested areas and annual
184 deforestation for the same PRODES year using the municipal boundaries (IBGE, 2016)
185 as the sample unit. The Spearman correlation test was used for this analysis.

186 The relationship between droughts and burning in already-deforested areas was
187 assessed using monthly precipitation estimates based on satellite data from TRMM
188 (Tropical Rainfall Measuring Mission v7, 3B43). For this study three climatic regions

189 were defined based on the mean values of maximum cumulative water deficit (MCWD):
 190 Aragão et al. (2007) for the 1998–2005 period: the eastern, central and western regions
 191 (Figure 2a). Drought intensity was measured as the MCWD between the months of June
 192 and September. We applied the Spearman test for correlation significance. We tested the
 193 hypothesis of a null correlation between water deficit and fire extent in order to evaluate
 194 the evidence that drought acted as an influential factor on fires. This test addresses the
 195 fact that anthropogenic forces were not the exclusive source of the fires detected, and
 196 the significance of weather influence remains an open question for the period in the
 197 literature (Aragão et al., 2018; Barlow et al., 2019).



198
 199 **Fig. 2** Location of the study area showing the three climatic regions in Acre (a) and the
 200 distribution of burned mapping validation points (b).
 201

202 2.5. Analysis of the spatio-temporal patterns of forest fire

203 The definition used for forest-fire burn scars in this study was based on Silva et al.
 204 (2018), where trees were detected that were directly or indirectly affected by fire to the
 205 point that they cause an impact visible on the optical satellite images. We quantified the
 206 total area of forest fire per year and its recurrence. We used violin plots to analyze the
 207 distribution patterns of areas sizes, including the median, maximum and minimum for
 208 each year. We quantified forest fires by categorizing the data according to land-tenure
 209 categories such as federal government land, settlement projects, private properties,
 210 conservation units and Indigenous Lands (Acre, 2010).

211 We analyzed the relationship between droughts and forest fire based on the
 212 MCWD for the three climatic regions, as described in Section 2.4 and Figure 2a. We
 213 calculated the Spearman correlation coefficient to test the relationship of the different
 214 predictor variables to fire occurrence.
 215

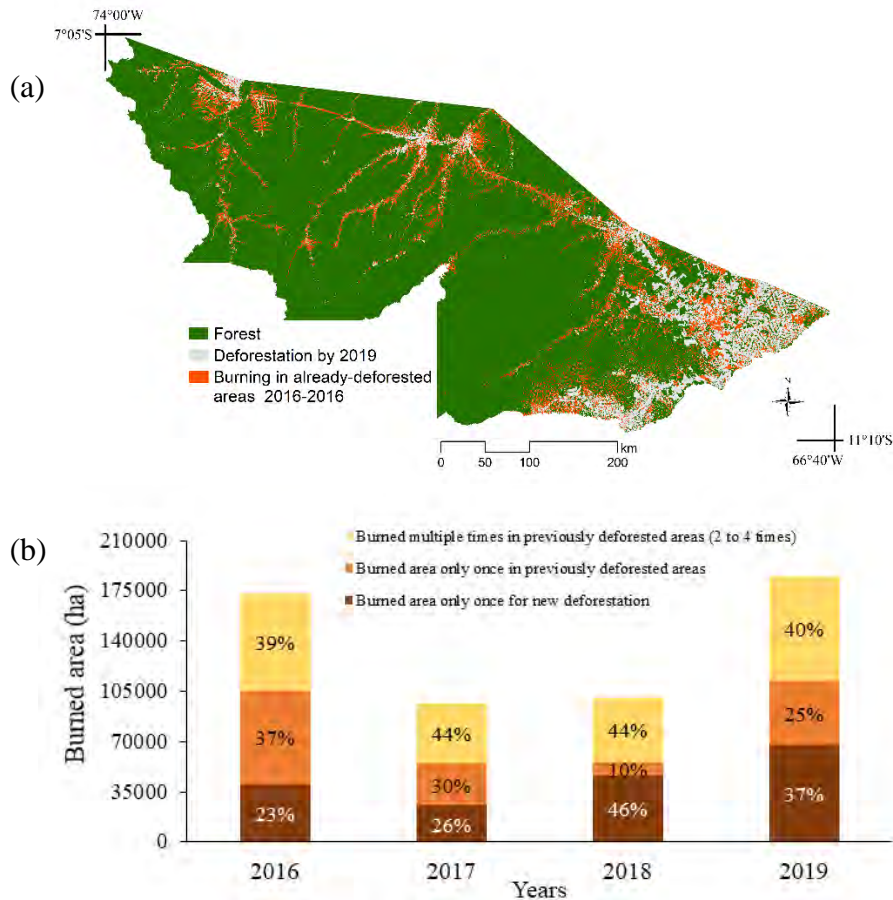
216 3. Results

217 3.1. Spatio-temporal distribution of burning in already-deforested areas

218 We mapped 550,251 ha of burning in already-deforested areas in the state of Acre
 219 in four years (2016–2019), corresponding to 64% of the total area mapped over the
 220 period occurred in 2016 and 2019 (Figure 3a). The overall accuracy of estimates of
 221 burned area was 98.5% (97.9% to 99.0%).

222 The year 2019 had the largest amount of burning in already-deforested areas
 223 among all of the years analyzed: 44% more than 2018, 46% more than 2017 and 4%
 224 more than the 2016 El Niño year. On the other hand, in 2017 and 2018, the area burned
 225 in already-deforested areas was 44% smaller than in 2016 and 2019. In the whole period
 226 analyzed, 67% (371,207 ha) of the burning occurred at least once in previously
 227 deforested areas and 33% (179,895 ha) occurred at least once in newly deforested areas.
 228 Of the total burning in already-deforested areas, 76% (323.284 ha) burned once, 19%
 229 (78,549 ha) burned twice, 4% (18,413 ha) burned three times and 1% (3868 ha) burned

230 four times (Figure 3b). Annually, 23% to 46% of the burned area was associated with
 231 newly deforested areas.



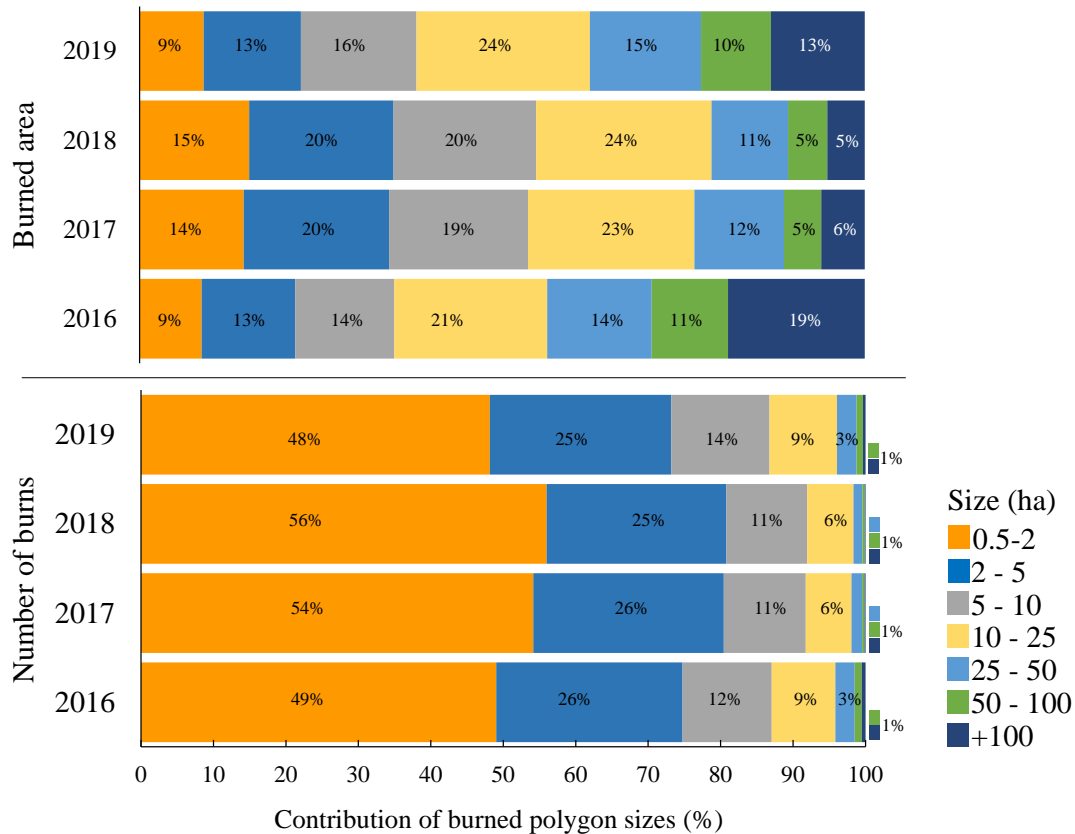
232

233
 234

235 **Fig. 3** Total burning in already-deforested areas in the state of Acre from 2016 to 2019
 236 (a). Shown in brown are the burnings that occurred in freshly deforested areas in each
 237 year, in orange, the burnings that occurred only once in grid cells deforested in previous
 238 years (before the period analyzed), and in yellow, the grid cells in which burnings were
 239 detected multiple times during the period (b).

240

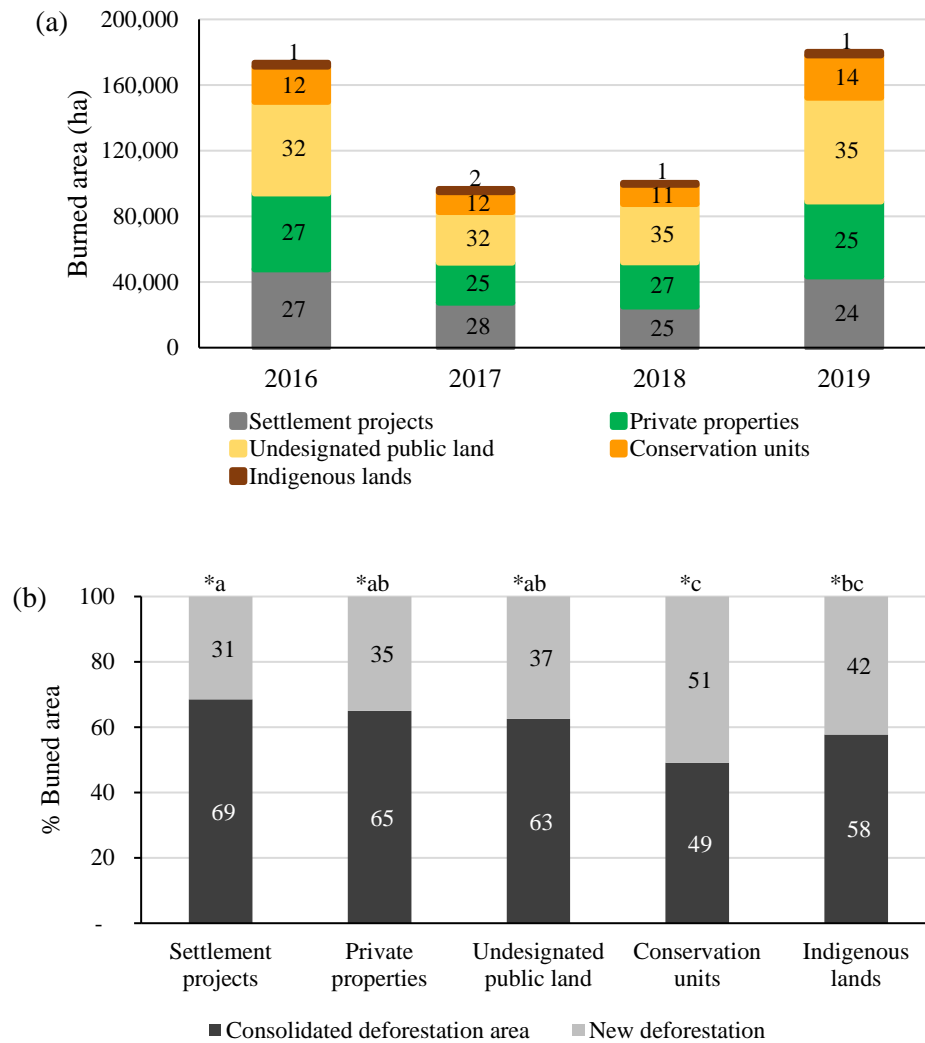
241 The distribution pattern of areas revealed that between 52,000 and 69,000 ha were
 242 burned every year (Figure 4, upper panel). These areas represent, respectively, 3 and 5%
 243 of the cumulative deforested area in Acre by the end of 2019 and are 17 to 55% greater
 244 than the 44,460 ha average area deforested annually in Acre over the 2016-2019 period
 245 (INPE, 2020c). The years 2016 and 2019 had the highest numbers of fires in already-
 246 deforested areas with more than 10 ha. These years also account for the largest
 247 percentages of the total burning in already-deforested areas (65 and 62%, respectively)
 248 (Figure 4, lower panel). This difference becomes larger in areas > 50 ha, since in the
 249 years 2019 and 2016 the burning in already-deforested areas within this class was four
 250 times larger than in 2017 and 2018. Areas with up to 5 ha represented 22% (2019 and
 251 2016) to 35% (2017 and 2018) of the total burning in already-deforested areas. These
 252 small areas represented 74% (2019 and 2016) to 81% (2017 and 2018) of the total
 253 number of areas affected by fires.



254
255 **Fig. 4** Distribution of areas by size class for burning in already-deforested areas by
256 burned area (upper panel) and number of burns (lower panel).
257

258 In regards to land-tenure categories, the years 2016 and 2019 had the largest
259 amounts of burning in already-deforested areas when compared to 2017 and 2018 in all
260 land-tenure categories (Figure 5). Undesignated public land had the largest contribution
261 to the total burned area ($34\% \pm 1.9\%$), an average of 46,000 ha. On the other hand,
262 Indigenous Lands had the smallest area burned in all years ($1\% \pm 0.4\%$), an average of
263 $1,989 \text{ ha year}^{-1}$, with only 1% of the total area burned. Together, all land-tenure
264 categories except Indigenous Lands, totalized an area burned in 2019 and 2016 (349.134
265 ha) 80% larger than the area observed between 2017 and 2018 (193.870 ha).

266 In fact, incidence of fire in conservation units and federal government land was
267 larger in 2019 than in all of the years analyzed (burned areas in conservation units were
268 17% larger than in 2016, 53% larger than in 2017 and 55% larger than in 2018; burned
269 areas in federal-government land were 12% larger than in 2016, 51% larger than in
270 2017 and 44% larger than in 2018). The percentage of burned areas in new deforestation
271 (31% to 42%) and in the already-deforested areas (58% to 69%) were equivalent among
272 all land-tenure categories with the exception of conservation units (ANOVA Levene's
273 test and Tukey HSD test, $p < 0.001$), with more burning in new deforestation (51%)
274 than in the already-deforested areas (49%), indicating the advance of new frontiers of
275 deforestation (Figure 5b).
276



277
 278 **Fig. 5** Burning in already-deforested areas by land-tenure category in the 2016-2019
 279 period: (a) area burned per year by land-tenure category, and (b) percentages of area
 280 burned in new deforestation and in the burning in already-deforested area. The values in
 281 the bars indicate the percentage contribution of each class. Different letters indicate
 282 significantly different means (ANOVA and Tukey HSD test, $p < 0.001$)
 283

284 Undesignated public lands (*terras devolutas*), settlement projects and private
 285 properties, represent 93% to 95% of the area larger than 50 ha of burning in already-
 286 deforested areas (Table 1). In 2016 and 2018, private properties represented the largest
 287 contributor to large areas. In 2018 and 2019, undesignated public land was the main
 288 contributor to large burned areas, reaching 46% of the total.
 289

290 Table 1. Area occupied by the class of areas greater than 50 ha of burning in already-
 291 deforested areas by land-tenure category and year.

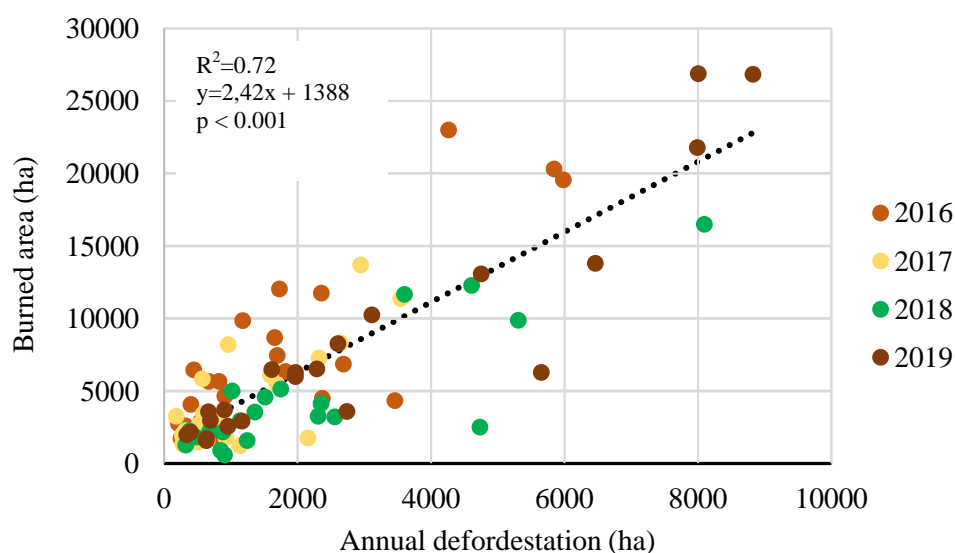
	2016		2017		2018		2019	
	ha	%	ha	%	ha	%	ha	%
Undesignated public land	16,341	32	3,595	33	4,306	41	18,892	46
Settlement projects	13,295	26	2,588	24	1,977	19	8,074	20
Private properties	18,975	37	4,191	39	3,573	34	11,775	29
Indigenous lands	167	0	117	1	72	1	0	0
Conservation units	2,126	4	274	3	672	6	1,945	5
Total	50,904	100	10,765	100	10,600	100	40,686	100

292

293 3.2. Deforestation and burning in already-deforested areas

294 At the municipal level, burning in already-deforested areas is significantly related
 295 to the annual rate of deforestation recorded by INPE ($p < 0.001$, $r = 0.74$, Spearman;
 296 Figure 6). The year 2019 had the largest area affected by fire (180,000 ha) and the
 297 highest annual deforestation rate in the last 14 years (68,800 ha) ($p < 0.001$, $r = 0.74$)
 298 and the lowest correlation was for the year 2017 ($p = 0.023$, $r = 0.49$). Burning in
 299 already-deforested areas was more intense in 2019, representing 41% of the mapped
 300 area (34% - 56.783 ha - in 2016, 37% - 35.387 ha - in 2017, 39% - 39.312 ha - in 2018
 301 and 41% - 71.344 ha - in 2019).

302



303

304 **Fig. 6** Relationship between annual increase in deforestation and burning in already-
 305 deforested areas for the state of Acre from 2016 to 2019.

306

307

308

309

310

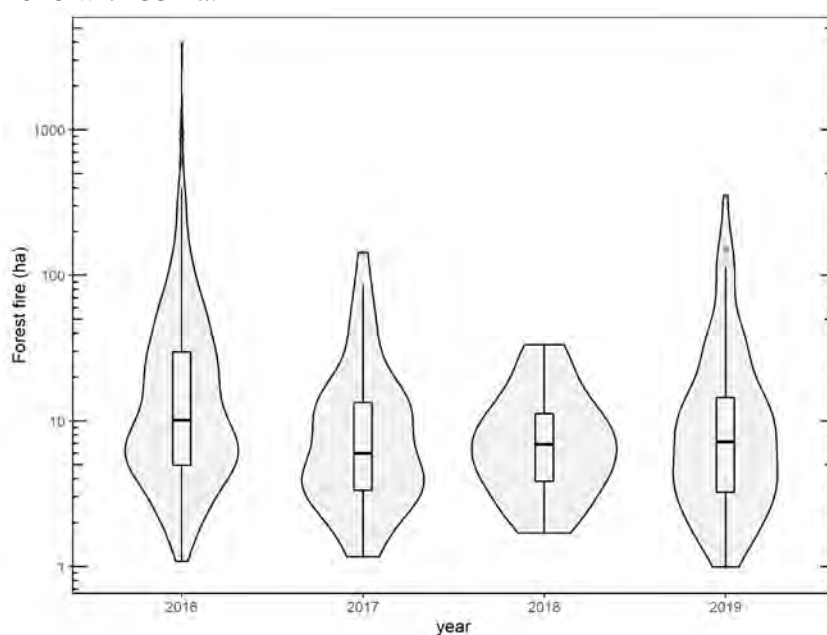
311

The total burning in already-deforested areas represented 4% to 8% of the deforested area for the entire state of Acre (Supplementary Material, Table S2). In the central region of Acre, the municipality (county) of Santa Rosa do Purus burned 30% of the cumulative deforested area detected by PRODES, Manoel Urbano, Feijó, Manoel Urbano, Jordão, Sena Madureira and Porto Walter burned from 10 to 22%. The

312 municipality of Assis Brasil, in the Alto Acre region, is an extreme case where, in 2016
 313 and 2019, 20% to 24% of the deforested area was affected by fire, respectively. These
 314 processes confirm the fact that fire is used not only to burn areas that are being
 315 deforested, but also in previously deforested land such as pasture or secondary
 316 vegetation (“*capoeiras*”), with the objective of managing these already-deforested areas.
 317 It should be remembered that many recently deforested areas are not burned in the same
 318 year, a detail that is not captured by this analysis.

320 3.3. Spatio-temporal distribution of forest fires

321 Forest fires in the period analyzed reached 34,084 ha, with the year 2016
 322 accounting for 91.3% of the total fire-affected area (31.117 ha), followed by 2019, with
 323 5.6% of the total (1920 ha) (Figure 7). The average overall accuracy of forest-fire
 324 identification by this method is 98.8% (98.1% to 99.4%). In 2016, there were areas of
 325 up to 3,927 ha, followed by 2019 with a maximum area of 350 ha, 2017 with 143 ha and
 326 2018 with 33 ha.



327
 328 **Fig. 7** Areas of forest fire (inner boxplot) and the empirical distribution of area by sizes
 329 (outer curve).

331 Forest fires were clustered in the eastern region of the state of Acre in all years:
 332 97% in 2016, 84% in 2017, 94% in 2018 and 100% in 2019. The land-tenure categories
 333 that contributed most to forest fires were settlement projects and private properties in all
 334 of the years analyzed, with the exception of 2019, where 78% of the fire occurred in
 335 conservation units, Indigenous Lands and undesignated public land in the extreme
 336 southeastern portion of Acre (Table 2; Supplementary Material, Figure S2).
 337

338 Table 2. Contribution of land-tenure categories to the area of forest fires in the state of
 339 Acre between 2016 and 2019.

Land-tenure category	2016		2017		2018		2019	
	ha	%	ha	%	ha	%	ha	%
Settlement projects	9,543	31	487	65	108	36	228	12
Private properties	12,525	40	148	20	85	29	195	10
Undesignated public land	7,364	24	114	15	69	23	466	24
Conservation units	1,657	5	0	-	36	12	476	25
Indigenous Lands	28	0	0	-	0	-	555	29
Total	31,117	100	749	100	298	100	1,920	100

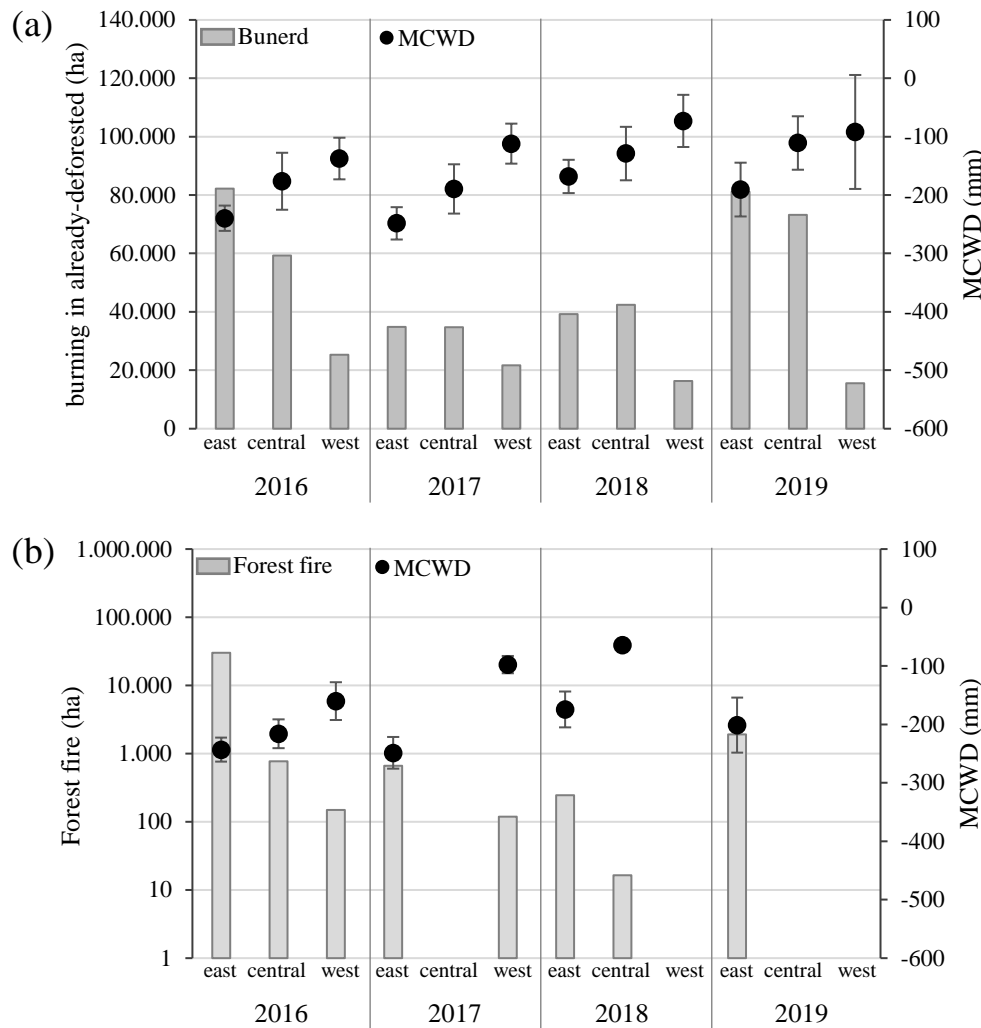
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341

342 3.4. Droughts, burnings in already-deforested areas and forest fires

343 We identified a gradient in the maximum cumulative water deficit (MCWD) in
 344 the three climatic regions of Acre in the period analyzed. The eastern region had the
 345 greatest water deficit during the dry season every year, followed by the central and
 346 western regions (Supplementary Material, Figure S2). Furthermore, the correlation
 347 between MCWD and occurrence of burnings in already-deforested areas was positive
 348 for all years ($p = 0.03$, $r = -0.62$, Spearman; Figure 8a), except for 2019. Similarly,
 349 MCWD had a negative correlation with occurrence of fires in the climatic regions of
 350 Acre, with the eastern and central regions exhibiting the strongest relation, especially
 351 for 2016, when there was an El Niño event ($p = 0.0016$, $r = -0.80$, Spearman; Figure
 352 8b). The burning in already-deforested areas was a record in 2019, the year in which
 353 there was weak El Niño, with water deficits equivalent to the years 2017 and 2018
 354 (visible in Figure S3 in the Supplementary Material).

355



356

357 **Fig. 8** Temporal relationship between (a) maximum cumulative water deficit (MCWD)
 358 and burning in already-deforested areas and (b) MCWD and forest-fire area in the state
 359 of Acre from 2016 to 2019 in the three regions analyzed (east, central and west regions).
 360

361 4. Discussion

362 4.1. Impact of burning in already-deforested areas

363 This study retrieved georeferenced and time-varying data on burning in already-
 364 deforested areas. This provides an important complement to the widely used PRODES
 365 deforested-area product from the Brazil's National Institute for Space Research (INPE)
 366 (INPE, 2020c). The main addition to the already-available fire products from INPE,
 367 namely active-fire point detections ("hot pixels") and the new burned-area product
 368 (INPE, 2020d), is the greater accuracy of our data due their higher spatial resolution,
 369 with which the extent of burned areas is measured (70 m in the new product *versus* 1
 370 km in the INPE products). The information presented by this study allows visualization
 371 of burned areas as small as 0.5 ha in area, allowing assessment of the annual burning
 372 rate for 2016-2019, in which fire frequency and extent achieved record breaking levels
 373 in the Amazon. The data presented here provide a valuable expansion of the information
 374 available for policymakers.

375 Our results indicate that 2019 had the largest area of burning in already-deforested
 376 areas in Acre, 80% greater than 2018 and even 4% greater than the area burned during
 377 the El Niño in 2016. The proportion of fires in recently deforested land was higher in

378 2018 and 2019. In 2019, the contribution was 41%, following the trend of increasing
379 deforestation in Acre (INPE, 2020c).

380 The contribution of areas larger than 10 ha to the total area of burning in already-
381 deforested areas was greater in 2019 than in 2017 and 2018, representing 32% of the
382 total area mapped. Fire in areas larger than 50 ha represented an average of 18% of the
383 total burned area we mapped over the whole period, with peaks reaching 26% in the
384 2016 and 2019. These areas are usually associated with extensive cattle ranching
385 (EMBRAPA, 2017); they amount to 66% of the deforested area in Acre. The areas are
386 in medium and large landholdings and represent the main source of ignition for forest
387 fires in Amazonia (Cano-Crespo et al., 2015; Dias Filho, 2011).

388 Deforestation rates have been increasing throughout Brazilian Amazonia since
389 2012 (INPE, 2020c) and our results show that the increasing trend in fires associated
390 with this was maintained and that it accelerated with an upward surge in the 2019. A
391 likely future trend of increased deforestation would be associated with still more fires.
392 Aragão et al. (2018) divided Brazilian Amazonia into one-degree grid cells and showed
393 that, although fire increases in grid cells where deforestation increases, there is also a
394 very large amount of forest burned in grid cells without increased deforestation, and that
395 in the 2015 drought fires burned large areas of forest throughout the region independent
396 of the amount of deforestation.

397

398 4.2. Impact of forest fire

399 The fact that major forest fires occurred in Acre in 2019 despite this not being a
400 year of extreme drought, when all other years without extreme droughts had almost no
401 forest fires, reflects the virulently anti-environmental rhetoric and policies currently in
402 place. Among the years analyzed, it was only in 2016 that an extreme drought event was
403 recorded, when the El Niño was very strong (NOAA, 2020).

404 The size of the largest mapped areas of forest fires reflect the magnitude of the
405 fires, showing the size of the spread of the fire. In 2016, we had a maximum area with
406 size of 3900 ha (mean = 38 ha, median = 10 ha). In 2019, we mapped a maximum areas
407 of 350 ha (mean = 17 ha, median = 5 ha), which was greater than in any year without
408 extreme droughts recorded by Silva et al. (2018).

409 The relationship between droughts and forest fires is different from the
410 relationship with burned areas, where the occurrence of fires observed in this study
411 coincided with rainfall deficit below -180 mm. Burning after deforestation was shown
412 to occur throughout Acre. The concentration of forest fires in the analyzed period
413 (2016-2019) was in the eastern region of Acre, which is a historically drier region
414 compared to the other regions of Acre (Aragão et al., 2007). In 2019, forest fires
415 occurred only in this region, where the largest areas were located near the triple national
416 boundary where Brazil, Peru and Bolivia meet and where a drought was recorded with
417 MCWD of -240 mm (Supplementary Material Figure S3). The concentration of large
418 fires in specific regions may be a reflection of the fact that the climate in the
419 southwestern portion of Amazonia is getting progressively drier, as reported by Aragão
420 et al. (2008), Fu et al. (2013) and Staal et al. (2018).

421 Projections for the future indicate that Acre is among the areas with the greatest
422 risk of prolonged drought periods and major forest fires (Faria et al., 2017; Fu et al.,
423 2013). These scenarios cause ecological concerns regarding the degradation of forests
424 and loss of biodiversity (Barlow et al., 2016). This also has implications for the state's
425 economy (Campanharo et al., 2019; Mendonça et al., 2004) and public health
426 (Machado-Silva et al., 2020; Morello et al., 2019). Policy concerns include loss of the
427 benefits from REDD (reducing emissions from deforestation and forest degradation),

428 which is under implementation in Acre through the Acre-California agreement (Acre,
429 2013).

430

431 4.3. The prominence of the 2019 fires may hide a worrying trend

432 Aside from the anomalous 2016 El Niño year, 2019 was outstanding in terms of
433 total burned area and also in terms of forest area burned. Since this was also true for the
434 annual deforestation rate, which increased to a level over 150% of the average for the
435 2016-2019 it may be the case that a new trend of greater forest suppression and
436 degradation was started in 2019. Additional evidence comes from the significant
437 negative correlation of 62% between burned area and water deficit. This shows that
438 climate was not behind the record fires in 2019 (Barlow et al., 2019), suggesting these
439 fires were intentional and were not unintended accidental fires (Stabile et al., 2020).

440 Institutional change favoring deforestation and fires is another new trend. Barlow
441 et al. (2019) argued that the deforestation policy under the presidential administration
442 that began in January 2019 deviated from the successful approach of 2004-2012, a
443 reference to the Action Plan for Prevention and Control of Amazon Deforestation
444 (PPCDAM), which, together with macroeconomic factors, helped reduce the annual
445 deforestation in Brazilian Amazonia by 84% in the decline that ended in 2012 (West et
446 al., 2019; West and Fearnside, 2021). The current federal administration has weakened
447 institutional enforcement capacity, resorting to an emergency approach to
448 environmental policy (Ferrante and Fearnside, 2019; Pereira et al., 2019). The
449 administration's discourse has led deforesters to believe that violations of
450 environmental laws will be forgiven and that regulations will be further relaxed
451 (Klingler and Mack, 2020). The combination of concrete institutional changes and anti-
452 environmental discourse encourages both deforestation and burning, even when the
453 government attempts to reverse this effect.

454 The government decreed a moratorium on fires for 60 days in August 2019, and a
455 120-day ban was decreed in 2020 (Brazil, 2019). Military enforcement of these bans did
456 not prevent large amounts of deforestation and burning (Finer et al., 2020; Moutinho et
457 al., 2020; OC, 2020), and the federal environmental agencies that the presidential
458 administration has largely dismantled have not had their surveillance and enforcement
459 capacities restored. Another controversial case was that Brazil's Amazonian fire crisis is
460 caused by indigenous people and subsistence farming by traditional communities
461 (Ferrante et al., 2020). However, our results show that in all years evaluated in Acre
462 only 1 to 2% of the total burning was in Indigenous Lands and 11 to 14% was in
463 conservation units (which include extractive reserves inhabited by traditional
464 communities) (Figure 5a); in contrast, 32 to 35% was in undesignated public land,
465 which is the primary target of large land grabbers (*grileiros*).

466

467 4.4. Protected areas are under increased pressure

468 Despite 2019 not having the largest area of forest burned, the year stood out in the
469 share of the burning that was in conservation units and Indigenous Lands (hereafter,
470 "protected areas"). The conjecture that at least part of the burning envisaged illegal
471 occupation of protected areas is supported by recent studies (Keles et al., 2020). These
472 use remote sensing to show the routine transgression of the legal constraints on land use
473 that involve deforestation and fires, in addition to land-grabbing (*grilagem*). They also
474 demonstrate that forest suppression and degradation are used as a strategy for pressuring
475 institutions to dismember protected areas or withdraw their protected status.

476 Among the protected areas, the Chico Mendes Extractive Reserve is under the
477 most social, political and economic pressure (Hoelle, 2011; Mascarenhas et al., 2018;

478 Vadjunec et al., 2009). This protected area represented 43 to 66% of the total burning in
479 already-deforested portions of protected areas. Between the 2018 and 2019, burning in
480 already-deforested areas in this extractive reserve increased by 340%. According to
481 Fearnside et al. (2018), the Chico Mendes Extractive Reserve had the fourth largest loss
482 of forest by 2014 of the 73 extractive reserves of the Brazilian Amazon. This protected
483 area is under strong pressure from deforestation, driven by the appreciation of livestock,
484 invasion of land and devaluation of forest productive chains such as rubber and Brazil
485 nuts (Hoelle, 2011; Vadjunec et al., 2009). Even though conservation units are
486 identified as a barriers to deforestation and burning (Pfaff et al., 2014), the results
487 presented here must be taken as an important warning sign with regard to their
488 preservation.

490 4.5. Stricter land and environmental policies could bring great gains

491 The fact that the greatest share of the area burned (65%) was in areas owned by
492 government demonstrates that the bulk of Brazil's Amazon burning is due to weak
493 enforcement, as has also been the case for deforestation (Araujo et al., 2009). In
494 addition, the share of burned areas above 50 ha in area, which peaked in 2019, suggests
495 that medium to large landholders play a relevant role in fire-assisted land-cover change,
496 which is another parallel with deforestation (Cano-Crespo et al., 2015; Dias Filho, 2011;
497 Godar et al., 2014). Therefore, there are two reasons why more rigorous policy would
498 bring great gains. First, avoiding deforestation in government-owned land is saving
499 economically valuable resources for current and future generations of Brazilians, which
500 is a duty of the government (Stabile et al., 2020). Second, large areas and landholdings
501 are detected with less error than smaller ones, and targeting them is more cost-effective
502 (Godar et al., 2014). This means that there is a clear opportunity for more rigorous
503 policy to deliver significant outcomes.

505 **5. Conclusions**

506 A novel high-resolution measurement of the areal extent of fires was developed as
507 part of this study and was applied to the case of Brazil's state of Acre, whose leadership
508 in sustainability has been challenged by rising amounts of fire and deforestation. Two
509 classes of fire were investigated: burning in already-deforested areas and forest fires.
510 The spatial and temporal patterns showed the prominence of 2019 fires in both classes
511 and the propensity of a given location to burn more than once. Significant correlations
512 that were positive with deforestation and negative with water deficit were also found, as
513 well as the dominance of federal lands, including protected areas, among the land
514 classes with large areas burned. The importance of burns above 50 ha in area shows the
515 role of large actors. Importantly, it was shown that climate was not a driver of the 2019
516 fire season. This adds evidence to attributing the upsurge to the discourse and policies
517 of Brazil's presidential administration that began in January 2019.

518 Our arguments arrive at a moment when the needed changes to preserve Acre's
519 regional leadership in sustainability are still possible. Authorities should undertake
520 strong action and target budgetary resources for surveillance and enforcement of
521 environmental restrictions. Authorities must also alter their discourse to emit signals
522 consistent with sustainability.

523 One important limitation of the analysis, and also a task for future study, is the
524 lack of precise investigation of which, among relevant biophysical, climatic,
525 socioeconomic and institutional factors, are the main predictors of burned area. This
526 would improve the usefulness of the data generated here for policy planning, including
527 the positioning of fire brigades.

528

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533

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Supplementary Material

Burning in southwestern Brazilian Amazonia, 2016-2019

Table S1. Data for the Landsat images used in this study for the mapping of fires in the state of Acre

Table S2. Deforestation and burning in already-deforested areas by municipality in the state of Acre from 2016 to 2019.

Figure S1. Parts of Landsat OLI 002/067 R5 G4 B3 images before and after the impact of fire. (a) fire scar a in pasture area. (b) burn scars in a pasture and agriculture area. (c) burn scars in a recently deforested area.

Figure S2. Forest fire in 2019 in the eastern region of the state of Acre.

Figure S3. MCWD for the state of Acre for (a) 2016, (b) 2017, (c) 2018 and(d) 2019.

Table S1. Data for the Landsat images used in this study for the mapping of fires in the state of Acre

Year	1/67	2/66	2/67	2/68	3/66	3/67	3/68	3/69	4/65	4/66	4/67	5/65	5/66
2016	16/Jul.	23/Jul.	23/Jul.	23/Jul.	30/Jul.	30/Jul.	14/Jul.	14/Jul.	07/Jul.	21/Jul.	21/Jul.	12/Jul.	12/Jul.
	17/Aug.	24/Aug.	24/Aug.	24/Aug.	15/Aug.	15/Aug.	15/Aug.	15/Aug.	06/Aug.	06/Aug.	08/Aug.	13/Aug.	13/Aug.
	18/Sept.	09/Sept.	09/Sept.	09/Sept.	16/Sept.	cloud	16/Sept.	16/Sept.	09/Sept.	09/Sept.	09/Sept.	14/Sept.	14/Sept.
2017	19/Jul.	26/Jul.	26/Jul.	26/Jul.	cloud	17/Jul.	17/Jul.	cloud	24/Jul.	24/Jul.	24/Jul.	15/Jul.	15/Jul.
	04/Aug.	27/Aug		11/Aug	02/Aug	02/Aug	02/Aug.	cloud	25/Aug.	25/Aug.	25/Aug.	16/Aug.	16/Aug.
	21/Sept.	12/Sep		cloud	19/Sep	3/Sep	19/Sep	19/Sep	10/Sept.	10/Sept.	cloud	01/Sept.	01/Sept.
2018	06/Jul.	29/Jul.	29/Jul.	29/Jul.	04/Jul.	04/Jul.	20/Jul.	cloud	cloud	27/Jul.	27/Jul.	02/Jul.	02/Jul.
	23/Aug.	14/Aug.	14/Aug.	14/Aug.	cloud	cloud	cloud	cloud	28/Aug.	28/Aug.	28/Aug.	19/Aug.	19/Aug.
	08/Sept.	17/out	cloud	cloud	06/Sept.	06/Sept.	06/Sept.	06/Sept.	13/Sept.	13/Sept.	13/Sept.	04/Sept.	cloud
2019	09/Jul.	16-Jul.	16/Jul.	01/Aug.	08/Aug.	23/Jul.	23/Jul.	cloud	30/Jul.	30/Jul.	30/Jul.	21/Jul.	21/Jul.
	10/Aug.	01/Aug.	01/Aug.	17/Aug.	24/Aug.	08/Aug.	24/Aug.	24/Aug.	15/Aug.	15/Aug.	31/Aug.	06/Aug.	06/Aug.
	11/Sept.	18/Sept.	18/Sept.	18/Sept.	25/Sept.	25/Sept.	25/Sept.	25/Sept.	16/Sept.	16/Sept.	16/Sept.	07/Sept.	07/Sept.

Table S2. Deforestation and burning in already-deforested areas by municipalities of the state of Acre from 2016 to 2019.

	2016			2017			2018			2019		
	Cumulative deforestation	Burned area	%	Cumulative deforestation	Burned area	%	Cumulative deforestation	Burned area	%	Cumulative deforestation	Burned area	%
Acrelândia	114,724	4,537	4	115,313	3,493	3	117,624	3,313	3	119,592	6,319	5
Assis Brasil	27,369	6,511	24	28,015	1,668	6	28,924	651	2	30,891	6,052	20
Brasileia	137,385	11,801	9	139,708	7,329	5	142,059	4,191	3	146,813	13,107	9
Bujari	123,287	6,388	5	124,079	2,890	2	125,592	4,635	4	128,201	8,313	6
Capixaba	88,676	8,724	10	89,810	1,282	1	90,950	2,996	3	93,691	3,651	4
Cruzeiro do Sul	101,939	9,889	10	102,901	8,256	8	104,652	5,183	5	106,269	6,523	6
Epitaciolândia	85,976	1,572	2	86,549	1,873	2	87,398	964	1	88,032	1,634	2
Feijó	151,124	23,018	15	154,072	13,719	9	162,171	16,519	10	170,995	26,855	16
Jordão	16,041	1,769	11	16,320	2,054	13	17,004	2,309	14	17,963	2,626	15
Mâncio Lima	40,974	4,119	10	41,158	3,307	8	41,541	2,292	6	42,177	1,767	4
Manoel Urbano	37,384	6,909	18	39,076	5,512	14	43,809	2,557	6	46,922	10,298	22
Marechal Thaumaturgo	25,150	2,654	11	25,427	1,925	8	25,834	1,886	7	26,175	2,039	8
Plácido de Castro	143,757	5,707	4	144,264	1,527	1	145,512	1,637	1	146,419	3,745	3
Porto Acre	128,813	7,488	6	129,740	1,843	1	131,103	3,592	3	133,393	6,566	5
Porto Walter	20,377	2,954	14	20,697	2,319	11	21,172	1,883	9	21,570	2,219	10
Rio Branco	268,033	20,324	8	270,677	8,386	3	274,281	11,702	4	282,275	21,817	8
Rodrigues Alves	49,256	5,722	12	49,831	5,899	12	50,850	5,045	10	52,022	2,989	6
Santa Rosa do Purus	9,304	2,808	30	9,581	1,390	15	9,906	1,324	13	10,600	3,022	29
Sena Madureira	189,544	19,585	10	193,089	11,420	6	198,396	9,918	5	206,400	26,904	13
Senador Guiomard	167,461	4,706	3	167,989	2,766	2	168,870	2,235	1	169,534	3,613	2
Tarauacá	151,822	12,068	8	153,416	6,126	4	158,027	12,310	8	164,488	13,828	8
Xapuri	136,816	4,398	3	138,975	1,816	1	141,533	3,272	2	147,188	6,338	4
Total	2,215,211	173,653	8	2,240,688	96,798	4	2,287,209	100,414	4	2,351,610	180,222	8

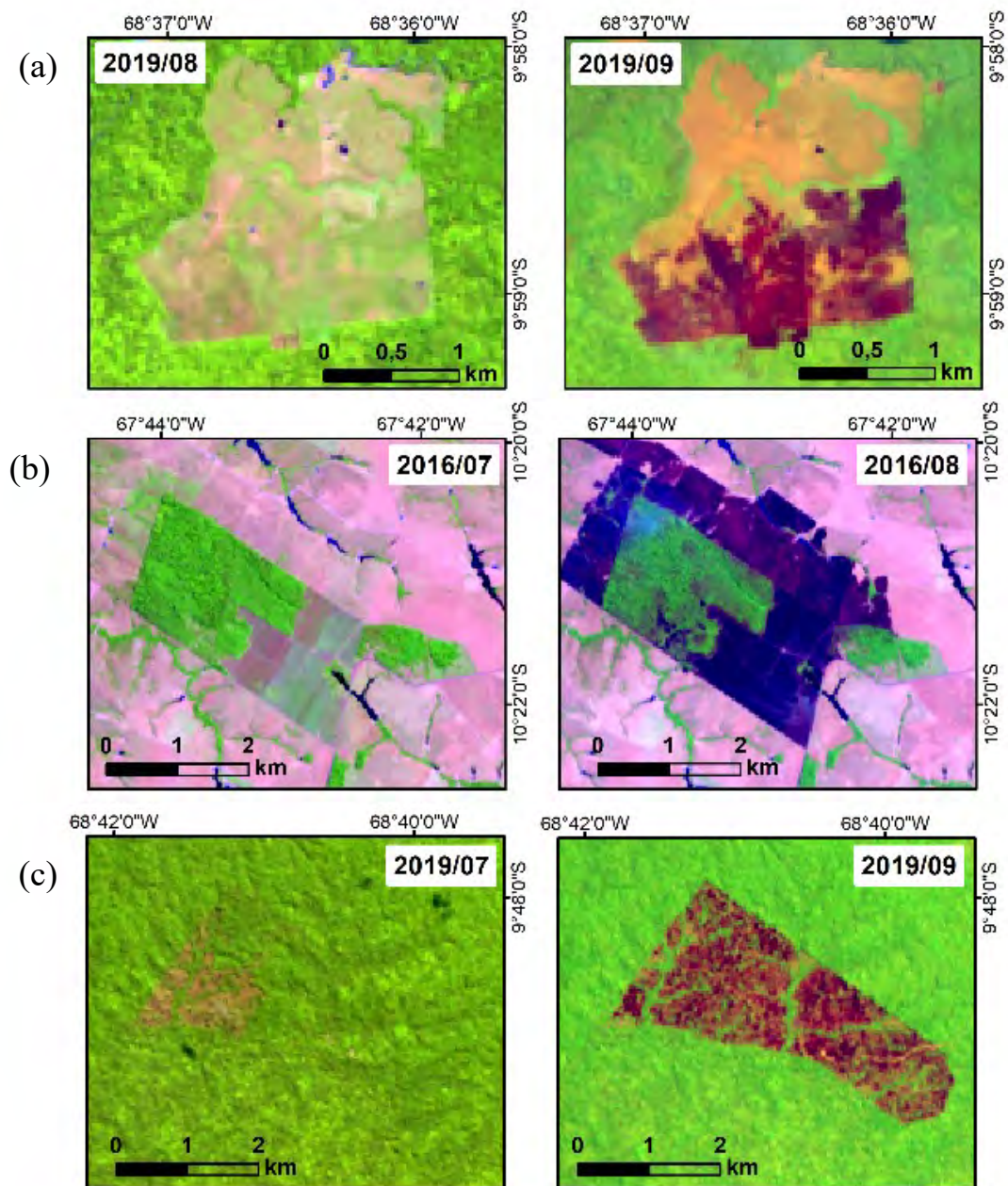


Figure S1. Parts of Landsat OLI 002/067 R5 G4 B3 images before and after the impact of fire. (a) fire scar in a pasture area. (b) burn scar in a pasture and agriculture area. (c) burn scars in a recently deforested area.

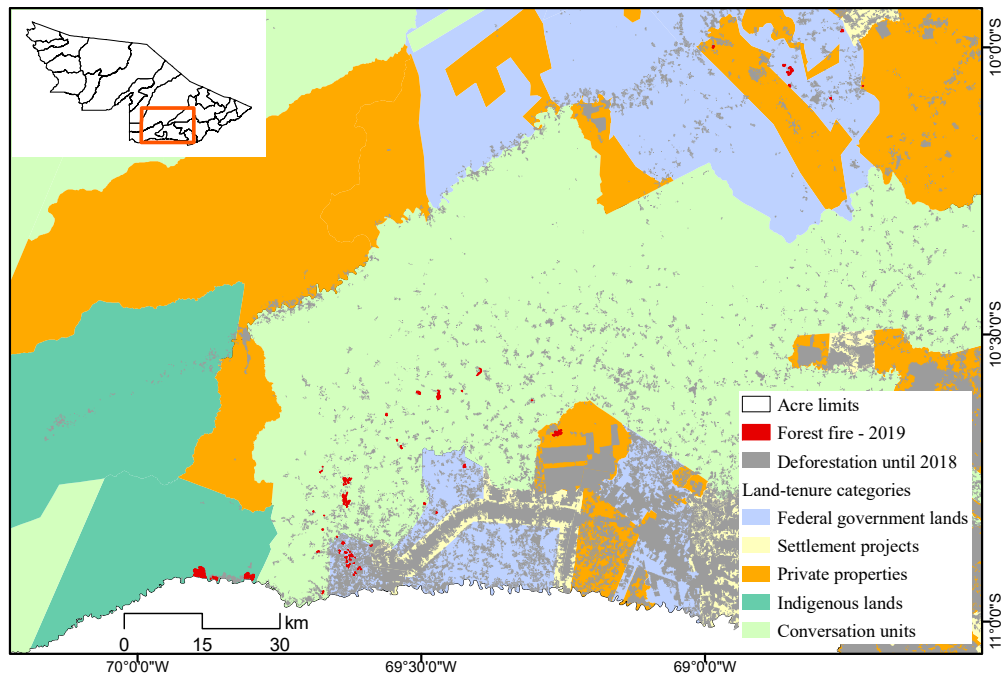


Figure S2. Forest fire in 2019 in the eastern region of the state of Acre.

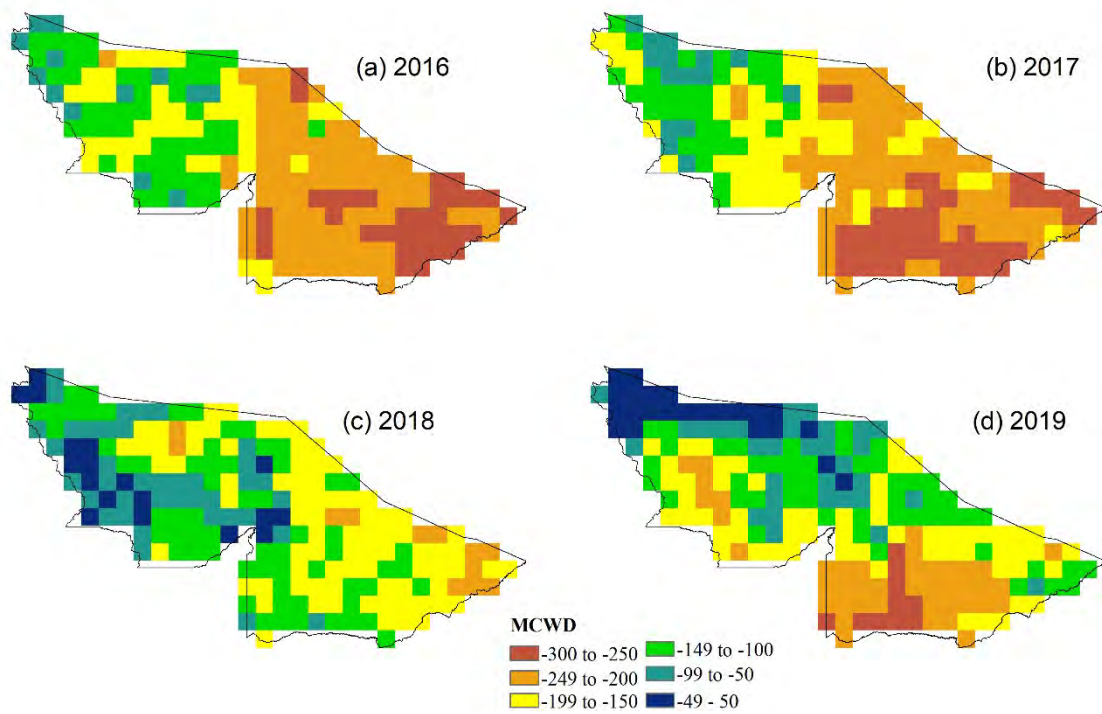


Figure S3. MCWD for the state of Acre for (a) 2016, (b) 2017, (c) 2018 and (d) 2019.