



1 **Methane gas emissions from savanna fires: What analysis of local**
2 **burning regimes in a working West African landscape tell us**

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12



13 **Abstract**

14 Savanna fires contribute significantly to greenhouse gas emissions. While it is recognized that these fires
15 play a critical role in the global methane cycle, there are too few accurate estimates of emissions from West Africa,
16 the continent's most active fire region. Most estimates of methane emissions contain high levels of uncertainty as
17 they are based on generalizations of diverse landscapes that are burned by complex fire regimes. To improve
18 estimates we used an approach grounded in the burning practices of people who set fires to working landscapes. We
19 conducted 97 experimental fires collecting data for savanna type, grass type, biomass composition and amount
20 consumed, scorch height, speed of fire front, fire type and ambient air conditions for two sites in Mali. We collected
21 smoke samples for 36 fires using a canister method. We report values for fire intensity, combustion completeness,
22 patchiness, modified combustion efficiency (MCE) and emission factor (EF). Our study finds that methane EFs
23 ranged from 3.71 g/kg in the early dry season (EDS) to 2.86 in the mid-dry season (MDS). We found head fires had
24 nearly double the CH₄ EF of backfires (4.89 g/kg to 2.92). Fires during the MDS have the lowest intensity values
25 and the lowest methane emissions 0.981 g/m² compared with 1.030 g/m² for EDS and 1.102 g/m² for the late dry
26 season (LDS). We conclude that policies aimed at shifting the burning regime earlier to reduce methane emissions
27 will not have the desired effects, especially if fire type is not considered. We recommend using the adjusted mean
28 value of 0.862 g/m²—based on the carbon content for West African grasses—for calculating emissions for West
29 African savannas

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31 Key Words: Savanna fires, methane, emission factors, combustion efficiency, Africa

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35 1 Introduction

36

37 The African savannas are the Earth's most extensively and frequently burned regions (Giglio et al., 2010) and
38 account for some 64% of the global extent of area burnt annually (Grégoire et al., 2013). Indeed, African savanna
39 fires regularly burn such large areas that they are visible from space, so much so that NASA scientists refer to Africa
40 as the “burn center of the planet” (National Aeronautics and Space Administration, 2005). Savanna fires are a major
41 source of greenhouse gases (GHGs) including carbon dioxide, carbon monoxide, methane, and nitrous oxide
42 (Koppmann et al., 2005).

43 Methane, a critical GHG, is responsible for about 20% of the warming induced by long-lived gases.
44 Although most sources and sinks of methane are known, their relative contributions to atmospheric methane levels
45 remain highly uncertain (Kirschke et al., 2013, Saunio et al., 2016, 2020). Our lack of understanding of the global
46 methane cycle contributed to the recent “methane enigma” a dramatic observed decline in the rate of increase in
47 atmospheric methane, which triggered a search for “missing methane” (Heimann, 2011). Although the decrease was
48 originally and mistakenly attributed to a decline in fossil fuel burning and a shift in farming practices (Kirschke et
49 al., 2013), it was eventually determined that the missing methane was due to a decline in area burned by savanna
50 fires. As NASA researchers determined, the missing methane from a drop in savanna burned area caused a decrease
51 of 3.7 Tg CH₄ per year—a value nearly twice the decrease expected (Worden et al., 2018).

52 The “missing methane” event demonstrates two important aspects of emissions from savanna fires. First,
53 these emissions are significant, so much so that they can offset increases from the key sources (fossil fuels and
54 agriculture). Second, our knowledge of the processes and factors that regulate the amount of methane emissions
55 from savannas is limited to the point that a large decrease went virtually undetected. Although eventually scientists
56 discovered the source, there remains high level of uncertainty for many key variables that determine the amounts of
57 methane emitted from savanna fires (Worden et al., 2018). In addition, there remains concern about the gap between
58 “top-down” (atmospheric measurements) and “bottom-up” (land-based models) estimates of global methane
59 emissions, which differ by thirty percent; Saunio et al (2020) suggest the reason is an overestimation of emissions
60 from bottom-up models. There is thus a need to improve land-based estimates of emissions from savanna fires.

61 The precise emissions from savanna fires depend on a variety of factors including those associated with
62 fuel, specifically vegetation type (the mix of grassy, leafy and woody fuels) and fuel moisture (a function of climate
63 and soil and fire season) as well as factors directly related to a fire's properties. In general, the crucial parameters for
64 determining GHG emissions from fires include burned area (BA), fuel consumption (FC), and the species specific
65 emission factor (EF), usually defined as the amount of gas or particle mass emitted per kg of dry fuel burned,
66 expressed in units of g/kg dry matter (van Leeuwen and van der Werf, 2011).

67 By one estimate savanna fires contribute 62% (4.92 PgCO₂ per year) of gross global mean fire emissions
68 (Lipsett-Moore et al., 2018). Due to their high rates of burning and vast extents, savannas are thought to hold
69 potential as major carbon sinks, if the fire regime could be modified to reduce emissions. The most commonly
70 proposed change in the regime to reduce the impacts of fires is to shift burning to an earlier period in the dry season
71 because early fires generally burn less completely and more patchily. Indeed, Lipsett-Moore (et al., 2018) recently



72 argued that there are “global opportunities for significant emissions reductions by simply shifting the fire period in
73 African savannas to early dry season” (1).

74 Yet, although scientists and policy makers increasingly recognize the important role these fires play in the
75 global carbon cycle, there are few accurate estimates of their emissions especially in terms of the key factors that
76 determine the type and quantity of GHG emissions. Critically, most studies of emission are global scale and use
77 average biome level EFs. EFs show large variability, however, between and within biomes due to differences in fuel
78 type and composition, burning conditions, and tree density (Andreae and Merlet, 2001; Korontzi, 2005; van
79 Leeuwen and van der Werf, 2011). There are few regionally specific emissions estimates because accurate
80 quantification of such emissions is difficult, being dependent upon reliable estimation of the various parameters,
81 many of which require intense fieldwork (Russell-Smith et al., 2009).

82 Nowhere is this truer than for West Africa, the continent’s most active fire region. To date, measurements
83 of emissions from African savannas are limited to a few broad-scale studies, largely based in the continent’s
84 southeast that rarely adequately account for changes in fuel classes, seasonality, or a host of other key factors
85 including fire type and intensity (Bonsang et al., 1995; Lacaux et al., 1995; Hoffa et al., 1999; Korontzi 2005).
86 Indeed, the most recent catalogs of EFs and fuel consumption (FC) for savannas includes a *single* data point from
87 West Africa (van Leeuwen et al., 2014; Andreae 2019). Studies from other regions find there is great variation in
88 study results (Russel-Smith 2009; van Leeuwen and van der Werf 2011); and, as Murphy et al., (2012) note, the
89 variability *between* samples collected *within* fires was greater than the differences between fires of different season.
90 These authors were unable to draw general conclusions about seasonal variation in methane emissions and EFs.
91 Among the key issues cited were the variations in the fraction of tree-leaf litter in the fuels of different savanna
92 environments.

93 In fact, there is very little data in the literature on fine fuel mixtures (the primary fuel for savanna fires)
94 used to estimate EFs in Africa, although the amount of woody vegetation clearly affects the emissions from
95 savannas (Korontzi, 2005; van Leeuwen and van der Werf, 2011). In the Brazilian cerrado, for example, Vernooij et
96 al. (2020) found that the seasonal effect on methane EF was stronger in more woody savanna vegetation with LDS
97 fires having 20% lower EF than EDS ones in shrub dominated areas.

98 Fuel moisture is also an issue; Russel-Smith et al. (2009) note there are currently no comprehensive
99 measurements of the seasonality of emissions gas composition, yet fuel moisture is a key determinant. This is a
100 critical problem because although evidence suggests that early dry season (EDS) fires consume less biomass and
101 burn more patchily; they also tend to have a lower combustion efficiency than later fires due to their higher fuel
102 moisture levels. A lower combustion efficiency theoretically causes a higher emission factor for CH₄. Indeed, one
103 study in Africa finds that the bulk of CH₄ emissions come from EDS fires (Hoffa et al., 1999) because the decrease
104 in area burned is more than offset by the increase in the CH₄EF. As such, whether a shift to an earlier fire regime
105 will result in a decrease in methane emissions for a given savanna must be determined empirically and proposed
106 policies to apply generalized findings from one continent to another may not achieve desired emissions reductions.

107 In sum, while savannas undoubtedly harbor great theoretical potential to sequester more carbon, and admit
108 less through a change in fire regime, there exists a great deal of uncertainty as to what the *actual* carbon shifts might



109 be, should fire regimes be altered. Elsewhere we identify five key sources of uncertainty largely arising from the
110 complexity of the patterns of savanna vegetation and fire regimes as well as biases associated with a lack of
111 consideration of the actual human fire setting and land management practices in these complex landscapes (Laris et
112 al., 2020, Laris 2020). It is clear that any effort to predict future changes in emissions or to implement policy to
113 reduce emissions requires more detailed information on how emissions vary according to the key factors noted
114 above, many of which are a function of human land management practices.

115 This study aims to fill a knowledge gap by incorporating data on a variety of human burning practices, the
116 characteristics of the fire regimes they produce, the vegetation conditions on the landscapes burned and the resulting
117 emissions of key GHG gases. Through a novel geographic approach, we designed our experiments to gather data in
118 ways that reflect actual *on-the-ground* burning practices of people living in working landscapes at two mesic
119 (precipitation > 750 mm) savanna sites in Mali, West Africa. By “working landscapes,” we mean savanna lands that
120 are occupied and worked by people as opposed to areas managed as reserves (e.g., Charnley et al., 2014); the latter
121 are most often used in fire research. The biomass (fuels) in working landscapes are a function of land use practices
122 including rotational agriculture, annual burning, and animal grazing and can differ significantly from those found on
123 non-working lands, and these can affect fire intensity, combustion completeness and combustion efficiency with
124 implications for gas emissions. The burning regimes studied, which are determined by such factors as seasonality,
125 time of day (ambient weather), fire type (with or counter to the wind), grass type and woody vegetation cover, were
126 selected to reflect *local practices* and based on over a decade of field and remotely sensing research.

127 To determine the factors that influence fire emissions of methane gas we conducted 97 experimental burns
128 using a field-based method to measure key factors. Vegetation plots, fire timing, and season (early, middle or late)
129 were selected based on local burning practices. We collected canister samples of smoke emissions for 36 of these
130 fires during the early and middle seasons, which we report on here. We also collected data for savanna type, grass
131 type, biomass composition and amount consumed; scorch height, speed of fire front, fire type and ambient air
132 conditions for two mesic savanna sites in Mali.

133

134 **2 Study Area and Methods**

135 We based our research in two working landscapes located in the Sudanian savanna of southern Mali (Fig.
136 1). We chose areas with annual precipitation over 900 mm because they burn frequently. The climate is divided into
137 two seasons: a wet period from approximately June through October and a dry season from November through May.
138 We also recognize cool dry period from approximately November through February and a hot dry period from
139 March through May. This distinction is important because the Harmattan wind, which is dry and desiccates
140 vegetation while creating unique fire weather, dominates in the cool season. The mean annual rainfall is 991 mm for
141 Tabou and 1,177 mm for Faradiélé (based on data from the nearby urban centers of Bamako (latitude: 12.64°,
142 longitude: -8.00°) and Bougouni (latitude: 11.42°, longitude: -7.47°) for each study area respectively) (Henry,
143 2011). The fire season follows the rains and typically runs from November through April, with the majority of fires
144 occurring in late December and early January.

145



146

147 Fig 1. Study areas in southern Mali

148

149 The vegetation is southern Sudanian savanna and is predominantly composed of a mixture of grasses, trees,
150 and shrubs in a complex mosaic. The landscape heterogeneity is a function of topography, underlying soil and
151 hydrology, as well agricultural uses, the combinations of which produces unique patterns of land cover (Duvall,
152 2011; Laris, 2011).

153

154 **2.1 Data Collection**

155 We studied 97 experimental fires. Data on the following variables were collected in the field for each fire:
156 average plot biomass, grass percentage of biomass, grass species, biomass consumed, fuel moisture, wind speed,
157 scorch height, ambient humidity, temperature, fire type, time of day, fire duration, burn efficiency or patchiness and
158 fire season. Vegetation characteristics including grass type (annual or perennial), grass species, and leaf height were
159 also recorded for each site. Fuel load (plot biomass) was measured in each of the experimental plots by delineating
160 three representative pre-fire quadrats of 1 x 1 m. Grasses were cut at the base using a scythe and weighed with an
161 electronic balance to determine the average. When present, we weighed leaf litter separately. Most grasses burned
162 were fully cured; however, for those that were not, we cut a sample and weighed wet, then dried and reweighed to
163 determine the cure rate, which was taken as the average for the plot. Fuel moisture content was calculated for each
164 plot using the method developed by Viney (1991).

165 We used a Kestrel 5500 Weather Meter station (KestrelMeters.com, Boothwyn, Pennsylvania, USA) to
166 collect wind speed, ambient humidity, and temperature during the burning of each plot. We recorded values every
167 five seconds and averaged them for each burn. The weather station was placed up wind and near each
168 experimental plot 2 m off the ground in an open area. Wind direction relative to the direction of each fire was
169 recorded.

170 We noted ignition time and each fire was timed until the flaming front reached the end of the 10 meter
171 plot. We set the majority of fires in late afternoon, which is in accordance with local practice, although we set
172 some fires earlier for comparative purposes. Post-fire ash and any unburned material were weighed for areas of
173 similar composition to the 1 m x 1 m pre-fire quadrats to determine the amount of biomass consumed. Scorch
174 height was averaged for each plot by measuring the height of scorch marks on several small trees. Burn
175 efficiency—a measure of the patchiness of the burn—was estimated by two observers.

176

177 **2.2 Plot design**

178 We selected plots to represent an array of savanna vegetation types dominated by different amounts of
179 woody cover and grass species. To aid in the selection of the burn plots, we used a long-term fire database to select
180 sites with known fire seasonality—fires known to burn during the early, mid, or late fire season on an annual basis
181 (Laris, 2011). We divided the sites into plots of 10 x 10 meters and applied fire treatments of head and back burns.
182 Fire timing was set according to the historical pattern of burning with early fires set in November through



183 December, middle fires in January, and late fires in late-February and March (Laris et al., 2016). We conducted
184 multiple burns per site to account for plot level heterogeneity. Plots at each site were located in close proximity with
185 attention paid to maintaining consistency in grass type and woody cover. Head and backfire plots were located
186 directly adjacent.

187

188 **2.3 Field data analysis**

189 To quantify intensity we used Byram's (1959) fire-line intensity, which is defined:

190

$$191 \quad (1) \quad I = Hwr$$

192

193 where I is Byram's fireline intensity (kW/m), H is the net low heat of combustion (kJ/kg), w is the
194 fuel consumed in the active flaming front (kg/m²), and r is the linear rate of fire spread (m/sec¹). The net
195 low heat of combustion (H) was selected following Williams et al. (1998) with 20,000 kJ/kg as an
196 appropriate value for savanna fires.¹ The amount of fuel consumed was calculated by subtracting the
197 average ash and unburned material remaining in three quadrats per plot from the pre-fire measurement of
198 dry biomass. Variable r was derived from the time it took for the base of the first flaming front to reach the
199 end of the 10 m plot. We calculated fire-line intensity for all samples possessing all the variables for
200 analysis. Finally, combustion completeness was calculated by dividing the biomass consumed by the pre-
201 fire biomass.

202 Fuel moisture content for the cured fuels was calculated using the method developed by Viney
203 (1991) based on McArthur (1967) for savanna fuels:

204

$$205 \quad (2) \quad m = 5.658 + (0.04651H) + \left[\frac{(0.0003151H^3)}{T} \right] - (0.1854T^{0.77}),$$

206

207 where H is relative humidity and T is ambient temperature at the time of the burn. We calculated dry biomass
208 weight by subtracting the fuel moisture content from the wet biomass weight and the amount of fuel consumed
209 was by subtracting the average ash and unburned fuels remaining in three quadrats per plot from the pre-fire dry
210 biomass weight.²

211

212 **2.4 Gas emissions sampling and analysis**

¹ We used the value of 20,000 kJ/kg following Williams et al. (1998) (230) who note: "Given the range and lack of consistency between studies in the value of H , and, in the view of the authors, the misleading precision implied by values rounded to the nearest 100 kJ/kg, 20,000 kJ/kg is within the range of reported values, and is easy and convenient to apply."

² We note that while a portion of the unburned material was likely mineral as opposed to carbon based ash, the majority of the unburned fuels were in the form of grass and leaf remains.



213 We collected samples of smoke from a sample of 36 of our experimental fires for early and mid-season
214 along with background air samples for each different site prior to burning. Samples were collected in stainless steel
215 vacuum canisters by mounting the canister on a pole and holding the canister with open flow-restricting valve about
216 40 centimeters above the flame. Once filled, the canisters were shipped directly back to California for analysis at the
217 laboratory of the Department of Chemistry at the University of California at Irvine. Mass Spectrometer (MS),
218 thermal conductivity and flame ionization after separation by gas chromatography were used to detect, inventory
219 and measure the quantities of the different species of gas contained in the samples (Katzenstein et al., 2003; Kone et
220 al 2020).

221

222 We calculated EF as:

223

224 (3)
$$EF_x = F_c 1000 \frac{MM_x}{MM_{\text{carbon}}} \frac{C_x}{C_T}$$

225

226 EF_x is the emissions factor for species x (g/kg). F_c is the mass fraction of carb in in the fuel for which we use the
227 value of 0.5 (the majority of studies find the carbon fraction to vary between 0.425 and 0.50; the latter is used most
228 often for purposes of comparison (Ward et al., 1996) although Lacaux et al., (1995) found a value of 0.425 for West
229 Africa). MM is the molecular mass of species x (g), and 1000 g/kg is a conversion factor. MM_{carbon} is the molecular
230 mass of carbon (12 g), and C_x/C_T is the ratio of the number of moles of species x in the emissions sample divided by
231 the total number of moles of carbon, calculated as:

232

233

234 (4)
$$\frac{C_x}{C_T} = \frac{ER_{x/CO_2}}{\sum_{j=1}^n (NC_j ER_{j/CO_2})}$$

235

236 Where ER_{x/CO_2} is the emissions ratio of species x to CO_2 , NC_j is the number of carbon atoms in compound j and the
237 sum is over all carbonaceous species (approximated as CO_2 , CO and CH_4 for this study).

238

239 The general equation used to quantify the gas species emitted from vegetation fires is the basic biomass
240 burning emissions model of the Intergovernmental Panel on Climate Change (IPCC) (IPCC 2003: 49; IPCC 2006:
241 A2.13):

242

243 (5)

$$\text{Emission (tons)} = \text{Area (ha)} * \text{Fuel (tons/ha)} * \text{Completeness (\%)} * \text{Emission Factor (g.kg}^{-1}\text{)} * 10^{-3}$$

244

245 Here Emission is the gas or aerosol flux in tons (t); Area is the total area burnt in hectares (ha); Fuel load is the amount
246 of burnt biomass in tons per hectares (tons/ha); Combustion completeness is the fraction of fuel burnt in percent (%);
247 Emission factor of a gas is the amount of this gas generated when one kilogram of fuel is burnt. We have revised the
248 formula to include seasonally specific values for area, fuel load, combustion completeness and emission factor and



249 add the variable burn efficiency (BE) which is the fraction of the surface area burned by the fire. We suggest the
250 following revision for determining emissions *by fire season* in savannas (E_s):

251
252 (6) $Emission_s \text{ (tons)} = BA_s(\text{ha}) * FL_s(\text{tons/ha}) * CC_s(\%) * EF_{x_s} * BE_s(\%)$
253

254 Here, BA is burned area percent FL is fuel load, CC is combustion completeness, EF_x is emission factor of species x
255 and BE is burn efficiency. We propose using *seasonal* values for these key factors because these variables vary
256 significantly by season as a function of ambient weather as well as fuel moisture and fuel type (fraction of leaf litter
257 or shrubs) and fuel conditions. We have added BE_s because most estimates of burned area are based on satellite
258 image analysis, which is too coarse to determine the actual surface area burned accurately (Russell-Smith et al
259 2009). Note that even medium resolution burned area estimates using Landsat contains errors in the percent area
260 burned due to the fragmented patterns fires create across a heterogeneous savanna landscape. We did not determine
261 total BA for this work, but have done so previously using Landsat data covering the study area (Laris 2011). As
262 such, we present our results in terms of emissions per meter-squared according to season of the burn.

263

264 **3 Results**

265 **3.1 Plot Characteristics**

266 The mean plot characteristics for biomass and weather conditions demonstrate the importance of the inclusion
267 of the mid-season in this study (Table 1). Average temperature generally increases over the course of the dry season,
268 but dips by 0.4°C in mid-season, which is an established phenomenon in West African climates. Average humidity
269 decreases as the dry season progresses, but the mid-season humidity is considerably more variable than in the early
270 and especially the late seasons. Calculated fuel moisture content declines over the course of the dry season from a
271 mean of 4.62% in the EDS, to 4.09% in the MDS and 3.65% LDS. When combined with the measured mean cure
272 rates for wet grasses; however, the total fuel moisture was 8.41% in the EDS, 12.04% and 3.65% in the LDS. Mean
273 wind speed peaks mid-season during the Harmattan, although the wind speeds are relatively low—classified as a light
274 breeze on the Beaufort scale. The percent grass of the total plot biomass is greatest in the early season, while the total
275 biomass (total fine fuels—grasses and leaves) is higher later in the dry season, reflecting an increase in leaf litter as
276 the dry season progresses. The increase in dry biomass also reflects the changes in species types burned—the taller
277 perennials often burn later in the dry season. Some perennials are too moist to burn during the early months of the dry
278 season and burn less completely in the mid-season due to higher moisture content.

279

280 Table 1. Mean Plot Characteristics by Study Period (standard deviations in parentheses) (2014-2016, Mali)

281

282



283 3.2 Fire Characteristics

284 The characteristics of the fires also vary by season (Laris et al., 2020) (Table 2). The mean BE increased as
285 the dry season progressed to a near complete burn by the late season (85.3% to 92.3 to 99.2) as expected due to the
286 gradual desiccation of the biomass. CC increased slightly from early to mid before increasing substantially in late
287 season (85.1% to 86.4 to 92.8). Both variables showed greater variability in the early and mid-seasons and much less
288 variability in the late season. Burn time and scorch height (both closely related to intensity) show a different pattern.
289 The middle season has the lowest mean values for scorch height and highest variability, while burns also take
290 substantially longer and are more variable in the MDS. This is in spite of higher wind speeds, perhaps reflecting the
291 slight drop and greater variability in temperatures as well as the higher moisture content of perennial grasses burned
292 in the MDS. Fire intensity was highest in late season (288 kW/m), as expected but surprisingly slightly lower in the
293 mid-season compared to the early season (242 Kw/m to 221). While the minimum intensity increased non-
294 monotonically over the fire season, dipping in mid-season, the maximum intensity decreased monotonically. The
295 standard deviation of the seasonal intensity values indicates high variability, especially for EDS. The high variation
296 in intensity values with respect to the mean reflects the wide variety of fuel, weather and fire conditions. Calculated
297 intensity values ranged from 46.6 to 829.6 kWm⁻¹ for head fires and 10.8 to 460.3 kWm⁻¹ for back-fires.

298

299 Table 2. Mean Fire Characteristics by Study Period (2014-2016, Mali)

300

301

302 3.3 Methane Emissions and MCE by Fire Type and Season

303

304

305 The findings from the analysis of the 36 fires for which canisters of smoke were collected fires are
306 shown in Table 3 along with results for fire type and season for the larger data set (n=97). The mean EF for methane
307 was 3.47 g/kg and the mean MCE was 0.90, which is considered on the cusp of flaming and smoldering. The results
308 indicate that fire type has a larger impact on methane EF than fire season. Head fires had much higher methane EF
309 values than backfires (4.9 to 2.5 g/kg) and this held regardless of fire season. MCE was also lower for head- than for
310 back-fires. The mean EF for early season was 3.71 g/kg about 30% higher than the mid-season value of 2.86. MCE
311 was also lower for the EDS as expected. The data also show that fire type has a large influence on fire intensity as
312 expected; head fire mean intensity was much greater than that for back-fires (313.4 kW/m to 109.0). In addition, we
313 found a large variation in the fire-line intensity especially for head fires.

314 We next used the mean values for methane EFs and the FL, BE and CC data for the complete data set to
315 determine methane emissions per meter-squared (Table 3). For the purposes of comparison, we used the mid-season
316 values for methane EF for the late-season emissions calculations.³ We found methane emission densities of 0.981
317 g/m² in the EDS, 1.030 g/m² in the MDS and 1.102 g/m² in the LDS for an average of 1.022 g/m². Based on our past
318 research, about 49.5% of the total area burned is EDS 39.5% is MDS and 11% LDS (Laris 2011). If we then use the
lower value of carbon content from Lacaux et al 1995 (West Africa), we estimate the average methane emissions per

³ We used the methane EF value from mid-season for the late season emissions calculation (which potentially overestimates LDS emissions because fuel moisture drops further in the late season likely affecting MCE and methane EF). All other values used in the calculation were from the late season burn experiments.



319 fire in southern Mali to be 0.862 g/m^2 . Finally, we found emissions from head fires were nearly double those from
320 backfires based on an average fuel load (1.635 g/m^2 to 0.948 g/m^2).

321

322 Table 3. Methane Emission Factor (EF), Combustion Completeness (MCE) and Key Fire Characteristics Emissions
323 by Fire Type by Study Period (standard deviations in parentheses) (2014-2016, Mali)

324

325

326 **4 Discussion**

327 The findings from our study indicate that fires set in the mid-dry season (the peak moment in local burning)
328 differ from those set in the early or late dry season (the seasons used in most fire studies) in key ways. MDS fires
329 have lower intensity than both EDS and LDS fires while CC increased slightly from early to mid-season with a
330 larger jump in late season. Mean values for BE gradually increase from early to middle to late season as expected
331 due to the gradual drying of the biomass. Fuel loads increased in the MDS as did the percentage of leaf litter in the
332 total biomass. These factors had differing effects on seasonal methane emissions.

333 Our study finds that methane EF means were highest for EDS (as expected) and drop by about 23% by the
334 MDS. We also found, however, that fire type had a greater impact on methane EF than season; head fire methane
335 EFs were nearly double those for back-fires. In general, methane EFs increased as fire intensity increased and head
336 fires, which have higher fire intensity had higher methane EF regardless of season.

337 Our results compare favorably with the biome averages from Andreae (2019). Andreae's savanna biome
338 mean MCE was $0.94 (\pm 0.02)$ and mean EF CH_4 was $2.7 (\pm 2.2) \text{ g/kg}$, which compare with our values for our MCE
339 (f) of 0.90 and methane EF (f) of 3.47 (Table 4). If we use the lower percent carbon value for West African savannas
340 (42.5% based on Lacaux et al 1995 as opposed to 50% used for comparative purposes), then our Methane EF values
341 are quite close to the biome means at 3.15 g/kg . It is not surprising that our values had a lower MCE and higher EF
342 CH_4 than the biome means because our values were based on emissions from "wooded mesic savannas" as opposed
343 to the "grassland" values used in most savanna biome estimates. Wooded savannas contain small trees, shrubs and
344 leaf litter, which tend to reduce MCE and increase methane EF values (e.g., Vernooij et al., 2020) due to the fraction
345 of ligneous fuel in the mix (22% on average in our case). Indeed, we suspect the reason head-fires have significantly
346 higher methane EFs than back-fires is that head fires have a higher scorch height and burn a greater amount of fresh
347 green leaves in the juvenile tree canopy. In addition, we note Wooster (2011) found that ground based
348 measurements were generally higher for carbon-based EFs than airborne ones used in most previous studies.

349 Our results compare more favorably for seasonal changes in methane EFs found by Korontzi's for East
350 African woodland-savanna. Importantly, our data is more in line with Kornontzi's values for "woodlands"
351 (Miombo) than for "grassland" (Damba) savannas (Table 4). We attribute this to the fact that both areas have mesic
352 rainfall regimes and high tree and shrub cover. As noted, we found the percent grass of the total plot biomass is
353 greatest in the EDS, while the total biomass is higher in the MDS reflecting a tripling of leaf litter biomass.

354

355 Table 4. A comparison of Methane EF values for seasonal savanna and woodland fire studies

356



357 Kornontzi (2005) argued that because EDS fires have a much lower combustion completeness values—near
358 zero in the early EDS to 80% by LDS—that total emissions from EDS fires would be less than those for LDS ones.
359 This in spite the fact that they found EDS fires have higher methane EFs by a factor of 1.47 for woodlands and 3.93
360 for grasslands (compared with our ratio of 1.30). They also found that fuel load increased by about twenty percent
361 from EDS to LDS with a big increase in MDS (we found a smaller increase). Finally, they found that methane EFs
362 were at their peak in MDS as opposed to LDS. As a result, Korontzi concludes that for SE Africa, early fires
363 produce lower methane emissions than either mid or late season fires in stark contrast to our results. Korontzi also
364 found the regional average CH₄ emission densities more than doubled from 0.24 g/m² in the EDS to 0.55 g/m² in the
365 LDS. By comparison, we found smaller increases by season—about 20% for the entire fire season (EDS-LDS). The
366 larger range in emission density values estimated by Korontzi derive largely from the higher range of CC used in
367 their model. Korontzi (2005) also notes that the lower emission densities for CH₄ in the EDS were mainly a result of
368 the larger effect of the increased fuel moisture content on lowering the combustion completeness compared with its
369 effect on MCE. They note this was opposite of what they found for their grassland fires indicating that the amount of
370 woody vegetation and leaf litter are critical determinants of EFs.

371 It is important to note; however, that Korontzi's values for early season CC were derived using a model and
372 based on fuel moisture levels alone (as compared to our study based on empirical results). We argue that, practically
373 speaking, people do not set fires when grasses are too moist to carry fire in West Africa. As such, we argue that
374 using CC values less than 50%, while theoretically useful, are not at all practical because people are unlikely to set
375 such fires as they would not achieve the desired goals of burning (Laris 2005, 2011). Indeed, it is probable that a
376 fire, which burns less than 50% of biomass, *can burn a second time* late in the season because a fire consuming such
377 a small fraction will not break fuel connectivity. It is also important to note that breaking fuel connectivity is a key
378 reason for setting early fires and a critical reason that a seasonal-mosaic fire regime burns less total area (Laris et al.,
379 2018). As such, although theoretically possible, we do not agree that using such low CC values is reasonable for
380 determining emissions from fires in working African landscapes. Indeed, we have rarely seen burned landscapes,
381 even EDS ones, with more than 50% of the biomass standing post fire.

382 In terms of BE or “burn patchiness”, our values are similar to Yates et al (2015) who found patchiness was
383 79% in EDS and 97% in LDS for a large study in Australian savannas (our values were 85% to 92% to 99%
384 respectively for EDS to MDS to LDS). We argue that our values combining burn efficiency and combustion
385 completeness give reasonable estimates of biomass consumed for the *actual* burning conditions in West African
386 savannas.

387 Korontzi (2005) also notes that woodland fires dominate burning in SE Africa and thus they use the
388 woodland as supposed to grassland values. Based on our fieldwork, however, grassland fires make up the bulk of
389 *early* season burning because woodland and wooded-savanna areas have higher fuel moisture content making them
390 difficult to burn in the EDS. Moreover, while Korontzi found fuel load increased by about 20% from EDS to LDS
391 we did not; fuel load values changed very little because biomass is being consumed by animals on working lands
392 throughout the dry season, thus although leaf litter increased, total biomass did not continue to increase after the
393 mid-season.



394 There are but a few studies of African savanna emission factors for which head and backfire data are
395 available. Wooster et al., (2011) found mean EF methane of 2.5 ± 0.9 g/kg for their ground-based study in SE Africa.
396 They too found mean EF methane for head fires to be higher than the backfire mean, 3.35 g/kg to 1.88 g/kg. Several
397 laboratory results also support our finding that MCE and EF methane are function of fire type. Keene et al. (2006)
398 used laboratory fires of fuels from SE Africa and found that the type of fire (backing, heading, or mixed) as well as
399 fuel moisture (in part a function of season) influenced MCE. The lowest MCE values they recorded were all for
400 heading fires with relatively low moisture content while MCE fell and EF methane rose as fuel moisture increased.
401 Similarly, Surawski et al., (2015) found that heading fires exhibited the lowest MCE and higher methane EF.

402 As noted above, we find the middle season to have unique fire, fuel and ambient weather characteristics.
403 These coupled with the practices of local inhabitants to set fires progressively to grasses as they gradually become
404 just dry enough to burn, creates the conditions under which emissions should be determined for West African
405 landscapes. EDS fires primarily burn shorter (1-1.5 m tall) annual grasses, which dry earlier. Early fires are limited
406 to these dry annual grasses because at this time of year the taller perennials (up to 3 m) are often too moist to carry a
407 fire (Laris et al 2020). As the dry season progresses past mid-December (for our region), the number of fires and
408 area burned increases rapidly with a peak in late December.

409 When comparing our results for EDS and MDS fires, we find strong evidence for an emissions trade-off.
410 That is, while EDS fires have a much lower BE than MDS fires (85% to 92%) they have only slightly lower CC
411 (85% to 86%), as such, the lower amount of biomass burned by EDS fires is offset by the 30% higher methane EF
412 resulting in the higher methane emission density values for the EDS. Thus, we disagree with the policy suggestion
413 put forth by Lipsett-Moore (2018) who promote increased early burning in African savannas to reduce methane
414 emissions. While it is theoretically possible that very early fires would burn a lower fraction of the landscape than
415 we have observed, we argue that such a policy is just as likely to cause an *increase* in methane emissions due to
416 higher methane EF of very early burning (see Korontzi 2005).

417 Finally, it is important to note that higher intensity head fires would be required to increase the burned area
418 of moist perennial grasses in the EDS⁴ and because head-fires have a methane EF nearly double that of backfires,
419 burning with head fires would likely counter any advantage of burning early to reduce emissions. In addition, local
420 inhabitants would be very reluctant to set such fires due to the increased risk that setting head fires could damage
421 field crops, which remain unharvested in the EDS.

422

423 **5 Conclusions**

424 This study finds that when fires are set in working landscapes in accordance with well-documented burning
425 practices of West African people, methane EFs ranged from 3.71 g/kg in the early dry season to 2.86 g/kg in the
426 mid-dry season. We also found that fire type has a greater effect on methane emissions than fire season with head-
427 fires having significantly methane EF compared to back-fires (4.9 to 2.9 g/kg). We note that we are unaware of any
428 estimates for area burned according to fire type for any of the world's savannas. We thus recommend using the

⁴ We made several attempts to burn perennial grasses in December and could not get them to ignite. Only under windy, head fire conditions will perennial grasses burn in the EDS.



429 adjusted mean value of 0.862 g/m^2 —based on the carbon content for West African grasses—for calculating
430 emissions for West African savannas. We suspect based on interview results and field observations in West Africa
431 (Laris 2002, 2011) that early- and mid-season fires tend to be planned burns and are thus likely set as backfires
432 (although fires can shift direction and winds can change) while late fires are more accidental and potentially burn as
433 head fires. As such, the actual mean might be lower given that EDS and MDS fires comprise the bulk of the fires in
434 the region. In terms of methane emission densities by season, we find no significant difference between the values
435 for early and mid-season fires with a ten percent increase occurring in the late season due primarily to the higher
436 CC and BE for this season. It is possible; however, that the higher LDS emissions might simply be an outcome of
437 using the methane EF from the mid-season and that the actual late-season methane EF would be lower offsetting the
438 increase in fuel consumed. Indeed we conclude that due to the *emissions trade-off*—methane EF declines as the dry
439 season progresses while fuel consumed ($\text{BE} \cdot \text{CC}$) increases—there is only a small difference in emissions density
440 values by season ($< 10\%$) for the region.

441 It is important to reiterate that several key findings of this study arise from documented burning practices of
442 people living in working landscapes. People set fires in West Africa later in the day resulting in fires with lower
443 intensity due to lower wind and air temperature, and higher humidity; and people set predominantly backfires all of
444 which contribute to lower intensity burning. In addition, we note that the fuel loads we recorded are nearly 50%
445 lower on working savanna lands compared to reserve lands used in some other studies (Laris et al 2020). Lower fuel
446 loads also produce lower intensity fires, which produce lower methane emissions. Finally, the number of fires peaks
447 in the West African region in the MDS (Laris et al., 2016), which, we find, have a lower methane EF and emissions
448 density than EDS fires.

449 In conclusion, our study finds that several factors influence the emissions from savanna fires including the
450 fire season, fuel load and type, and, most importantly, fire type. Each of these factors are a function of human land
451 and fire management practices. We also conclude there is an emissions trade-off in setting fires earlier and, as such,
452 a policy to increase the amount of early burning in West African would be very difficult to implement because much
453 burning is already “early” and because earlier burning of uncured grasses would likely result in higher methane EFs.
454 Moreover, any policy aimed at increasing the amount of early burning would likely require setting head fires, which
455 would increase burn efficiency and combustion completeness further negating the effects of any reduction in burned
456 area while also causing an undesired increase in uncontrolled fires.

457



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618 **Author Contributions**

619 PL was principal investigator on the project, supervised all aspects of the research and wrote the manuscript. MK
620 was involved in the fieldwork and the gathering cleaning and organizing of all data as well as commenting on
621 manuscript. FD was head of the field research team and advisor on the field. RJ and LY were involved in data
622 organization and analysis as well as discussing and commenting on manuscript.

623

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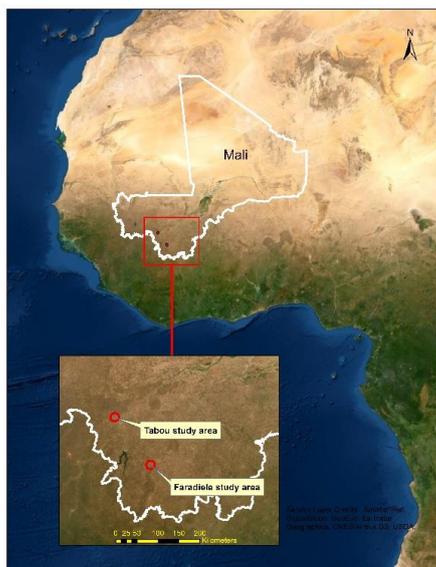
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630 **Figures and Tables**

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632

633 Fig 1. Study areas in southern Mali

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635 Table 1. Mean Plot Characteristics by Study Period (standard deviations in parentheses) (2014-2016, Mali)

Mean plot characteristics	Early	Middle	Late
Dry biomass (g/m ²)	376 (125)	445 (201)	423 (203)
Grass biomass (percent)	90.39 (18.68)	75.25 (23.74)	78.30 (22.98)
Total Fuel moisture (percent)	8.41 (0.57)	12.04 (0.73)	3.65 (0.26)
Temperature (° Celsius)	32.45 (3.37)	32.03 (3.32)	36.80 (3.28)
Relative humidity (percent)	29.62 (5.12)	20.64 (7.92)	19.34 (1.48)
Wind speed (meters/second)	1.16 (0.55)	1.45 (0.60)	0.85 (0.53)

636

637 Table 2. Mean Fire Characteristics by Study Period (2014-2016, Mali)

Mean fire characteristics	Early	Middle	Late
Spread rate (meters/second)	0.031 (0.02)	0.028 (0.03)	0.034 (0.01)
Burn time (seconds)	599.7 (488)	736.6 (591)	357.6 (172)
Scorch Height (meters)	1.39 (0.52)	1.31 (0.68)	1.70 (0.63)
Burn efficiency (%)	85.3 (12.44)	92.3 (11.42)	99.2 (1.06)
Combustion completeness (%)	85.1 (12.50)	86.4 (11.68)	92.8 (3.49)
Byrams Fire Intensity (kWm ⁻¹)	242.1 (259.58)	221.2 (226.91)	288.1 (213.87)

638



639 Table 3. Methane Emission Factor (EF), Combustion Completeness (MCE) and Key Fire Characteristics Emissions
 640 by Fire Type by Study Period (standard deviations in parentheses) (2014-2016, Mali)
 641

Fire and Emissions (n)	Head (47)	Back (50)	Early (34)	Middle (45)	Late (18)	Mean (97)
Modified Combustion Efficiency (MCE) n=36	0.88 (0.19)	0.92 (0.20)	0.88 (0.18)	0.92 (0.19)	NA	0.90 (0.18)
Methane Emissions Factor (EF CH ₄ g/kg) n=36	4.89 (1.64)	2.92 (2.59)	3.71 (2.81)	2.86 (1.93)	NA	3.30 (2.36)
Byrams Fire Intensity (Kw/m)	366.9 (287.4)	124.8 (81.9)	242.1 (259.6)	221.2 (226.9)	288.1 (213.87)	242.1 (235.0)
Burn efficiency (%)	93.6 (8.2)	91.0 (14.0)	85.3 (12.44)	92.3 (11.42)	99.2 (1.06)	91.4 (11.7)
Combustion Completeness (%)	86.7 (10.6)	87.9 (12.1)	85.1 (12.50)	86.4 (11.68)	92.8 (3.49)	87.3 (11.0)
Methane emissions (g/m ²)	1.635 (0.797)	0.948 (0.388)	0.981 (0.488)	1.030 (0.386)	1.102 (0.506)	1.022 (0.446)

642

643 Table 4. A comparison of Methane EF values for seasonal savanna and woodland fire studies

Methane EF	Early	Mid/Late	Ratio	Mean
This study Wooded savanna	3.71	2.86	1.30	3.30 (± 1.9)
Korontzi (2005) Woodlands	3.82	2.61	1.47	3.22
Korontzi (2005) Grasslands	3.14	0.80	3.93	1.97
Andrea (2019) Savanna biome	NA	NA	NA	2.71 (± 2.2)

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