

1 **Title: Effect of crop management and climatic factors on weed control in sugarcane**
2 **intercropping systems.**

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17

18 **Abstract**

19 **Context**

20 Cover crops in intercropping systems have a high potential to manage weeds in tropical
21 agroecosystems as an herbicide alternative. However, it still needs to be better investigated in
22 sugarcane cropping systems.

23 **Objective**

24 This study assesses the weed control ability of cover crops in sugarcane intercropping systems
25 depending on crop management and climate in tropical conditions.

26 **Methods**

27 This study was based on a set of eight experiments, including ten cover crop species,
28 performed from 2012 to 2021 in three areas in Reunion Island (France) and comparing three
29 inter-row management treatments: i) chemical weed control, ii) cover crops sown in the inter-
30 row, and iii) low weed control. Mean weed and cover crop ground cover were measured
31 during the entire crop cycle, and the complementary weed management practices (herbicide,
32 manual weeding) were also recorded. Sugarcane yield and crop management were evaluated
33 to assess the economic impact of a change in practice.

34 **Results**

35 Mean weed ground cover was reduced by 51 % in the cover crop and chemical treatments
36 compared to low control. Additionally, the cover crop treatment had an herbicide treatment
37 frequency index lower by 37% compared to chemical treatment but a higher number of
38 manual weedings. In all treatments, weed cover increased with rainfall and temperature. In the
39 cover crop treatment, a structural equation modeling approach allowed us to highlight the
40 influence of cover crop (sowing date and species) and sugarcane management (crop cycle -
41 plant or ratoon crop) on weed cover. Weed cover decreased with an increase in cover crop
42 ground cover, which was higher for the early sowing date. Additionally, weed cover was

43 lower in ratoon crops than plant crops but increased over time since intercropping was
44 implemented. Finally, sugarcane yield was lower by 6% in cover crop treatment compared to
45 chemical treatment. The change in operating costs and revenues between these crop
46 management induced a decrease in profits in intercropping compared to chemical cropping
47 systems, variable according to the degree of mechanization.

48 **Conclusions**

49 This study highlights the predominance of management factors, compared with climate, to
50 explain the effectiveness of weed control in sugarcane intercropping systems.

51 **Implications**

52 Due to the change in weed cover over time, there is a need to assess the long-term
53 sustainability of this practice and the potential change in weed communities.

54

55 **Keywords:** *Saccharum officinarum*; companion plants; interplanting; living mulch; weed
56 management

57

58 **1. Introduction**

59 Over the past few decades, the use of pesticides in agriculture has surged worldwide, mainly
60 due to the expansion of agricultural land and increased application rates (Brühl and Zaller,
61 2022). However, concerns about pesticides' environmental and health impacts have spurred
62 efforts to explore alternative weed management strategies (FAO, 2019). The emergence of
63 herbicide-resistant weeds is particularly concerning, which poses a significant challenge to
64 weed control (Heap, 2014). In this context, research and development efforts have focused on
65 evaluating non-chemical approaches to weed management (Jacquet et al., 2022).

66 In tropical and temperate climates, weeds are a significant biotic constraint on agricultural
67 productivity, reducing crop yield and increasing production costs (FAO, 2017; Oerke, 2006).
68 In recent years, the use of cover crops in intercropping systems gained attraction as a
69 promising alternative to herbicides for weed control while also delivering other benefits, such
70 as improved soil fertility and erosion control (Garcia et al., 2018; Gu et al., 2021; Kocira et
71 al., 2020; Scavo et al., 2022; Verret et al., 2017). There is broad agreement that a vigorous
72 cover crop will suppress weeds and reduce herbicide use (Bhaskar et al., 2018; Christina et
73 al., 2021b; Mennan et al., 2020; Ranaivoson et al., 2019).

74 Sugarcane is a globally significant crop, serving as the primary source of sugar, ethanol, and
75 electric energy in tropical and sub-tropical regions (de Miranda and Fonseca, 2020). Given the
76 increasing demand for ethanol and electricity as potential substitutes for fossil fuels to reduce
77 greenhouse gas emissions (Goldemberg et al., 2014), the sustainability of sugarcane
78 production came under scrutiny (Bordonal et al., 2018), particularly concerning herbicide use.
79 While several studies have highlighted the potential benefits of intercropping sugarcane with
80 cover crops to diversify production (Geetha et al., 2019), increase economic returns per unit
81 of land (Rehman et al., 2014), enhance soil fertility (Solanki et al., 2017), or reduce fertilizer
82 inputs (Prellwitz and Coelho, 2011; Tian et al., 2020), only a limited number have examined

83 the effectiveness of cover crops for inter-row weed management (e.g., Geetha et al., 2019).
84 Although many sugarcane cultivars are now resistant to pests and pathogens (Sundar et al.,
85 2015), weed management remains a major production constraint (Suganthi et al., 2019).
86 As shown in temperate cropping systems, various climatic and management constraints
87 influence intercropping performance for weed control (Gardarin et al., 2022). In the case of
88 tropical sugarcane cropping systems, climatic factors such as temperature and precipitation
89 could affect the growth and development of both the sugarcane, the cover crop, and the
90 weeds, potentially altering the timing and intensity of weed competition. Also, management
91 factors such as mulching (Carvalho et al., 2017) and cover crop management practices could
92 affect weed control in sugarcane intercropping systems. For example, the choice of cover crop
93 species and the timing of its planting can affect its competitiveness with weeds and its ability
94 to suppress weed growth (Dayoub et al., 2022; Liebman and Dyck, 1993). Additionally, weed
95 control in intercropping systems often relies on complementary pre-emergent herbicide
96 application before cover crop sowing and manual weed control (Mansuy et al., 2019), which
97 must be considered to assess the ability of cover crops to manage weeds in the inter-row.
98 Thus, to optimize the performance of sugarcane intercropping systems for weed control, it is
99 essential to consider and address the range of climatic and technical constraints that may
100 influence their effectiveness.

101 Sugarcane is the dominant agricultural industry on Reunion Island, with more than half of
102 agricultural land devoted to it, making it a crucial economic, social, and environmental
103 resource (Leung, 2015). The island's tropical climate is favorable to weed growth, with a daily
104 delay in weed control resulting in yield losses of 0.1-0.2 Mg ha⁻¹ d⁻¹ and an absence of weed
105 control leading to 80% yield loss (Marnotte et al., 2008). As a result, chemical weed control is
106 the primary strategy used to control weeds, making the sugarcane sector the largest consumer
107 of herbicides on the island (Antoir et al., 2016). However, the withdrawal of many chemical

108 herbicides and government incentives to reduce the use of pesticides by 50% by 2025 under
109 the EcoPhyto program have prompted the sector to explore alternative weed control methods
110 (Antoir et al., 2016). While the efficacy of cover crops for weed control in sugarcane inter-
111 rows has been recently tested on Reunion Island (Mansuy et al., 2019), the impact of the
112 island's high spatial climate variability remains poorly understood.

113 A better understanding of the factors influencing weeds x cover crops x sugarcane interactions
114 in intercropping is crucial for developing effective strategies to reduce herbicide use while
115 maintaining yield (Viaud et al., 2023), particularly in tropical islands with diverse
116 environmental conditions. The objectives of this study were to i) assess the weed response to
117 sugarcane intercropping compared to monocropping, ii) assess the changes in herbicide
118 application and manual weeding practices in the intercropping system compared to
119 monocropping, and iii) disentangle the effects of the cover crop, the complementary weed
120 control management, the climate, and the cover crop management on weeds in the sugarcane
121 inter-row. We gathered data from a network of 8 intercrop experiments performed on Reunion
122 Island from 2012 to 2021 to achieve these objectives.

123

124 2. Material & Methods

125

126 2.1. Experimental network

127 Reunion Island is located in the Southwest of the Indian Ocean near Madagascar. Sugarcane
128 agricultural areas are located from sea level to 800 m a.s.l, in dry (West Coast) and high
129 rainfall conditions (East Coast, Christina et al., 2021a). Eight field experiments were
130 conducted over nine years (2012-2021) in three locations on Reunion Island (Table 1). Five
131 experiments were conducted on-station in the North of the Island on a Nitisol (WRB
132 classification). Three experiments were conducted on-farm, one in the East on a Ferralsol and
133 two in the West on an Andic Cambisol. The length of each experiment varied between one to
134 seven years. In this study, a trial defines an experiment performed in a specific year (28
135 trials). Mean annual temperatures (T_{MEAN}) varied from 19.6 to 26.1°C, and annual rainfall
136 varied from 718 to 2611 mm y^{-1} , depending on trials. Soil chemical characteristics were
137 measured at each experiment's beginning and listed in Table A1. The weather variables were
138 obtained from the Meteor software (<https://smartis.re/METEOR>), which interpolates daily
139 weather variables from the Meteo France and CIRAD weather station network.

140

141 2.2. Experimental design, sugarcane management and yield

142 Each experiment consisted of a randomized block sugarcane (*Saccharum officinarum*) to
143 compare three inter-row management treatments: a treatment with chemical weed control in
144 the whole plot (CH), a treatment with a cover crop sown each year in the inter-row (CC), and
145 a weedy treatment with low weed control in the sugarcane inter-row (LC). Depending on the
146 experiments, each treatment was repeated in two to four blocks, with each elementary plot
147 varying from 60 to 780 m² (Table A2). The sugarcane was planted with a 1.5m row distance

148 in all experiments. All experiments were planted with the sugarcane R579 cultivar except for
149 the “P12” experiment using the R587 cultivar (Table A2). All experiments were irrigated
150 except for “Canecoh_OF”, located in the East rainy part of the island. Mineral NPK
151 fertilization was supplied to reach the optimum yield according to recommendations provided
152 by the Serdaf software based on soil analysis (Versini et al., 2018). NPK fertilization varied
153 depending on trials, with an average of 180 kg N ha⁻¹, 100 kg P ha⁻¹, and 250 kg K ha⁻¹. In
154 most trials, fertilization has been split with a first application one month after planting or
155 harvest and a second at 2-3 months (details available in the data paper Ngaba et al., 2023).
156 Sugarcane stalk fresh mass (hereafter sugarcane yield) was measured at harvest by sampling
157 all sugarcane stalks on the two rows in the middle of the plot (Ngaba et al., 2023). Part of the
158 yield data (only legume intercropping treatments) were included in the meta-analysis of Viaud
159 et al. (2023).

160

161

162 **Table 1.** Description of the experimental network used in this study with the experiment
163 location, trial harvest year (including crop cycle), cover crop species sown in the trial and
164 their sowing dates (months after planting or harvest, MAP/H), mean temperature (T_{MEAN} , °C)
165 and rainfall (mm y^{-1}) over the sugarcane cycle, and trial identification (ID). Different sowing
166 dates in the same trial indicated different sowing date treatments. In the cover crops species,
167 “Mixture” indicated a mixture of cover crops (detailed in Table A3).

Location, Experiment (Latitude, Longitude, Altitude)	Harvest year	Cover crop species	Sowing date (MAP/H)	T_{MEAN} (°C)	Rainfall (mm y^{-1})	ID
North “P12” (-20.903, 55.532, 70m a.s.l)	2018 (P)	<i>Vigna unguiculata</i>	2.8	26.0	1421	1
	2019 (R1)	<i>V. unguiculata</i>	1.6	25.0	1260	2
	2020 (R2)	<i>V. unguiculata</i>	1.6	23.9	1284	3
North “P41” (-20.903, 55.528, 68m a.s.l)	2015 (P)	<i>Desmodium intortum</i> , <i>Canavalia ensiformis</i>	1.1/ 2.3/ 3.3	24.0	2008	4
	2016 (R1)	<i>D. intortum</i> , <i>C. ensiformis</i>	2.3	24.2	1117	5
	2017 (R2)	<i>Guizotia abyssinica</i> , <i>C. ensiformis</i> , <i>Avena sativa</i> , <i>V. unguiculata</i> , <i>Crotalaria juncea</i> , <i>Crotalaria spectabilis</i>	2.3	24.2	1350	6
	2018 (R3)	<i>G. abyssinica</i> , <i>C. ensiformis</i> , <i>Brassica carinata</i> , <i>V. unguiculata</i> , <i>C. juncea</i> , <i>C. spectabilis</i>	1.6	25.0	1528	7
	2019 (R4)	<i>G. abyssinica</i> , <i>C. ensiformis</i> , <i>B. carinata</i> , <i>V. unguiculata</i> , <i>C. juncea</i> , <i>C. spectabilis</i>	1.1	25.6	1327	8
	2020 (R5)	<i>G. abyssinica</i> , <i>C. ensiformis</i> , <i>B. carinata</i> , <i>V. unguiculata</i> , <i>Raphanus sativus</i> , Mixture	2.6	25.0	718	9
North “P25_montagne” (-20.902, 55.531, 50m a.s.l)	2018 (P)	<i>C. ensiformis</i> , Mixture	2.7	24.4	1527	11
	2019 (R1)	<i>C. ensiformis</i> , Mixture	1.3	25.6	1310	12
	2020 (R2)	<i>C. ensiformis</i> , Mixture	1.3	25.0	736	13
North “P25_LEG” (-20.902, 55.531, 50m a.s.l)	2012 (P)	<i>V. unguiculata</i> , <i>C. ensiformis</i> , <i>Dolichos lablab</i> , <i>D. intortum</i> , <i>Cajanus cajan</i>	0.4 /1.9	23.9	1569	14
	2013 (R1)	<i>V. unguiculata</i> , <i>C. ensiformis</i> , <i>D. intortum</i>	0.7 /1.8	24.4	1418	15
	2014 (R2)	<i>V. unguiculata</i> , <i>C. ensiformis</i> , <i>D. intortum</i>	0.5 /2	24.4	1023	16
	2015 (R3)	<i>V. unguiculata</i> , <i>C. ensiformis</i> , <i>D. intortum</i>	0.7 /1.6	24.3	1941	17
	2016 (R4)	<i>V. unguiculata</i> , <i>C. ensiformis</i>	1.8	24.1	1063	18
North “P25_ITK” (-20.902, 55.531, 50m a.s.l)	2013 (R1)	<i>V. unguiculata</i>	0.8	24.2	1103	19
	2014 (R2)	<i>V. unguiculata</i> , Mixture	0.6 /2.3	24.5	1033	20
	2015 (R3)	<i>B. juncea</i> , Mixture	0.7 /1.6	24.4	1974	21
East “Canecoh_OF” (-21.047, 55.682, 170m a.s.l)	2016 (R4)	<i>V. unguiculata</i> , <i>C. ensiformis</i>	1.5	22.8	1218	22
	2017 (P)	<i>C. ensiformis</i> , Mixture	3	22.8	1505	23
	2018 (R1)	<i>C. ensiformis</i> , Mixture	1.6	22.6	1289	24
	2019 (R2)	<i>C. ensiformis</i> , Mixture	2.1	22.9	2501	25
West “Ecocanne_AC” (-21.067, 55.280, 545m a.s.l)	2020 (R3)	<i>C. ensiformis</i> , Mixture	1.5	22.4	2611	26
	2015 (R1)	<i>D. intortum</i> , <i>C. ensiformis</i>	0 /1.3	20.3	1241	27

West "Ecocanne_OB" (-21,066, 55,285, 575m a.s.l)	2014 (R2)	<i>D. intortum, C. ensiformis</i>	0.3 /1.4	19.6	860	28
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168 * Crop cycle: P = plant crop, R1 = first ratoon crop, R2 = second ratoon crop...

169

170 2.3. Weed management, HTFI, and manual weeding

171 Weed management of all treatments included: the application of a pre-emergent herbicide
172 after sugarcane planting or harvest in the whole plot, a manual or chemical weeding of the
173 sugarcane row to limit the impact on sugarcane production (throughout the crop cycle), and a
174 specific manual weeding of the weeds considered as the most harmful in the whole plot (in
175 particular, tall grasses: *Panicum maximum* and *Rottboellia cochinchinensis*). In the chemical
176 control treatment, additional post-emergent herbicides were systematically applied to limit
177 ground cover by weeds below the threshold of 30% ground cover (sugarcane sector
178 recommendations). Additional manual weeding were occasionally performed in the cover
179 crop treatment when ground cover by weeds exceeded 30%. An Herbicide Treatment
180 Frequency Index (HTFI) was calculated as the sum of the ratio of the herbicide dose applied
181 divided by the recommended dose across each herbicide application in a specific plot
182 (Halberg, 1999). The HTFI was only available in 81% of trials. Details of the chemical
183 compounds used for herbicides are available in the data paper (Ngaba et al., 2023), and are
184 summarized in Table A6. Information regarding manual weeding included: manual weeding
185 in the row only, manual weeding in the inter-row only, and manual weeding in the whole plot.
186 In the present study, the number of manual weeding in the inter-row (MW_{IR}) was considered
187 as the sum of those in the whole plot and the inter-row, while the number of manual weeding
188 in the row (MW_R) was considered as the sum of those in the whole plot and sugarcane row.
189 The number of manual weeding was only available in 89% of trials. In most cases, when an
190 herbicide was applied, or a manual weeding was performed during the trials, all plots from the

191 same inter-row management treatment were concerned. Consequently, the mean values per
 192 treatment and trial were used in the following analyses.

193

194 **2.4. Cover crop management**

195 The experimental network tested ten cover crop species (Table 2). They included six
 196 Fabaceae, three Brassicaceae, and one Asteraceae plant, as well as mixtures (Table A3).
 197 Cover crops were sown manually or mechanically in two to three lines depending on trials
 198 (Table A3). The sowing date depended on trials (Table 1, Table A3). At the end of their
 199 development, cover crops died gradually with sugarcane closure or were mechanically
 200 destroyed before canopy closure in a few treatments (14% of treatments, Table A3). The
 201 sowing densities were determined based on previous studies and literature (Christina et al.,
 202 2021b; Table A3). In mixtures, the sowing density of each cover crop was its density in pure
 203 crop divided by the number of present species (Négrier et al., 2023).

204

205 **Table 2.** Cover crop species sown in the experimental network: species, common name,
 206 family, area of origin, and the number of trials including these species (Ntrial) is indicated.

Species	Common name	Family	Origin	Ntrial
<i>Brassica carinata</i>	Abyssinian mustard	Brassicaceae	Temperate	4
<i>Brassica juncea</i>	Brown mustard	Brassicaceae	Temperate	1
<i>Canavalia ensiformis</i>	Jack bean	Fabaceae	Tropical	19
<i>Crotalaria juncea</i>	Sunn hemp	Fabaceae	Tropical	2
<i>Crotalaria spectabilis</i>	Showy crotalaria	Fabaceae	Tropical	2
<i>Desmodium intortum</i>	Greenleaf	Fabaceae	Tropical	6
<i>Dolichos lablab</i>	Hyacinth bean	Fabaceae	Tropical	1
<i>Guizotia abyssinica</i>	Niger	Asteraceae	Temperate	4
<i>Raphanus sativus</i>	Radish	Brassicaceae	Temperate	2
<i>Vigna unguiculata</i>	Cowpea	Fabaceae	Tropical	13
Mixture of cover crops	Detailed in Table A3			10

207

208 **2.5. Ground cover by cover crop and weeds**

209 The ground cover by cover crops and weeds was estimated in each elementary plot (excluding
210 borders) using a visual notation method according to a rating scale ranging from 0 to 100%
211 and described in Table A4. This method makes it possible to assess the ground cover by
212 species in multi-species crops and was used in previous studies (Christina et al., 2021b;
213 Mansuy et al., 2019; Marnotte, 1984; Négrier et al., 2023). In the trial plots, ground cover by
214 i) cover crops and ii) weeds was performed every month during the first six months and then
215 every two months up to the sugarcane harvest. Ground cover was then linearly interpolated
216 between two dates of measurement to calculate a mean ground cover by cover crops (COV_{CC})
217 or weeds (COV_{WEED}) between two sugarcane harvest dates. The current study will not present
218 the ground cover results by weed species. Nonetheless, a description of weed species
219 observed in the trials is available in Table A5.

220

221 **2.6. Economic assessment**

222 An economic analysis of the three systems (chemical, cover crop, and low control) was
223 carried out to assess the impact on profits of a change in practices. The economic analysis was
224 carried out using the OTECAS software developed by the eRcane Institute as part of the
225 agricultural innovation and transfer network (2022 version, [https://coatis.rita-
226 dom.fr/wakka.php?wiki=RechercheFacette&q=otecas&facette=](https://coatis.rita-dom.fr/wakka.php?wiki=RechercheFacette&q=otecas&facette=)). This software, developed
227 specifically for the sugarcane industry in La Réunion, calculates revenue and operating costs
228 according to crop management and yields. Three types of farms were considered: i) a farm
229 managed manually (weeding with a backpack sprayer and manual sowing of cover crops), ii)
230 a farm managed with a micro-tractor for weeding and sowing, iii) a farm managed with a
231 tractor for weeding and sowing. These three types of farms currently exist in La Réunion.

232 Only the variable operating costs across the cropping systems were considered in the
233 expenses: the purchase of herbicides (Table A6) and seeds (Table A7), as well as labor and
234 mechanization costs (taking account of equipment depreciation, Table A8). For each plot in
235 the trial network, the costs of herbicide purchase, seed purchase, number of chemical weeding
236 operations, and number of manual (and some mechanical) weeding operations were
237 calculated. Each farm's labor and mechanization costs for herbicide application were
238 calculated differently, considering the number of weed control operations (in the trials) and
239 the type of weed control (backpack sprayer, micro-tractor, or tractor). The costs of sowing
240 cover crops were calculated differently for the three farms, taking into account the type of
241 sowing (manual, micro-tractor, or tractor) as well as the mulch management in the row
242 spacing, which must be removed for manual or micro-tractor sowing. The revenue from
243 sugarcane production in each trial was calculated with OTECAS based on the fresh stalk
244 yield, sugarcane richness, and additional subsidies in La Réunion. For this calculation, we
245 considered the same size of 5 ha for all three types of farms with manual harvest. The profits
246 were calculated as the difference between revenue and operating costs.

247

248 **2.7. Data analysis**

249 Prior to analysis, a few elementary plots were removed due to cover crop failure due to
250 untested factors: accidental herbicide application in the inter-row (14 out of 607 plots), no
251 emergence due to irrigation failure or excessive water stress (60 out of 607 plots), cover crop
252 destruction due to bio-aggressors (6 out of 607 plots). All analyses were performed using R
253 4.1 (R Development Core Team, 2022).

254 For all statistical analyses, the following procedure was made. If necessary, to ensure residue
255 normality, the target variable was transformed using the Box-Cox function (powerTransform

256 function from car R package, Fox et al., 2023) based on a Gaussian law. Then, the
257 identification of the random effects (among trial identification, location, year, and nested
258 effects) was realized by comparing the linear mixed model (lme function, nlme R package,
259 Pinheiro et al., 2022) and the only fixed model (gls function, nlme R package) fitted using the
260 restricted maximum likelihood (REML) estimation and compared using the Bayesian
261 information criterion (BIC). Then, the fixed effect selection was performed using a stepwise
262 downward procedure based on BIC from the model with all fixed effects and interactions (lme
263 function) and fitted using the maximum likelihood (ML) estimation (Zuur et al., 2009).
264 Finally, the model parameters of the resulting model were fitted using REML estimation. The
265 assumptions of homogeneity and normality were checked graphically using histograms of
266 residuals, quantile-quantile plots of Pearson's residuals versus standardized Gaussian sample,
267 and plots of residuals against fitted values. Pairwise comparisons among treatments were
268 performed using the emmeans function with a Tukey p adjustment method and a 0.05
269 probability threshold (emmeans R package, Lenth et al., 2023). The models' mean predicted
270 and confidence interval values depending on factors were calculated using the emmeans
271 function.

272 In the first step, the influence of the inter-row treatment (low control, cover crop, or
273 chemical), the experiment identification, the mean temperature (T_{MEAN}), the cumulated
274 rainfall over the crop cycle, the crop cycle (plant or ratoon crop), the irrigation management,
275 and their interaction on COV_{WEED} (transformation $\lambda=0.39$) were tested as fixed effects using a
276 linear mixed model with the trial identification (ID) as a random effect. The same analysis
277 was performed for HTFI (transformation $\lambda=0.46$) and the number of manual weedings in the
278 inter-row (MW_{IR} , transformation $\lambda=0.46$) and in the row (MW_R , transformation $\lambda=0.50$). Note
279 that analyses on MW_{IR} and MW_R were also realized without transformations with a
280 generalized linear mixed model (glmer function) with a Poisson law which gave the same

281 results. The linear analyses with a Gaussian law and transformation were chosen for these two
 282 variables to be applied in the following structural equation model. Similarly, the influence of
 283 the inter-row treatment on the different operating costs and revenue was assessed with the
 284 same mixed analysis with the trial identification as a random effect. The average predicted
 285 values by the statistical models for costs and revenue depending on the inter-row treatment
 286 were presented.

287 In the cover crop treatment, COV_{WEED} could result from the influence of different weed
 288 control management: COV_{CC} , MW_{IR} , and HTFI, as well as other biotic and abiotic factors.
 289 COV_{CC} , MW_{IR} , and HTFI could themselves be influenced by climate and management
 290 through weed pressure. Consequently, the influence of climate (T_{MEAN} and Rainfall),
 291 sugarcane management (crop cycle, irrigation, number of years since the implementation of
 292 intercropping - Y_{IC}), cover crop (CC) management (CC species - $CC_{species}$ and sowing date –
 293 CC_{Sowing_date}), MW_{IR} , and HTFI on COV_{CC} and COV_{WEED} was tested using a mixed structural
 294 equation model (SEM, psem function from the piecewiseSEM R package, Lefcheck, 2016) to
 295 take into account the indirect effect of factors on COV_{WEED} through COV_{CC} , HTFI, and
 296 MW_{IR} . The initial SEM included four submodels (hypotheses):

$$COV_{WEED} \sim COV_{CC} + MW_{IR} + HTFI + Y_{IC} + T_{MEAN} + Rainfall + CC_{species} + CC_{Sowing_date} \\ + Cycle + Irrigation, \quad random = 1|ID$$

$$COV_{CC} \sim MW_{IR} + Y_{IC} + T_{MEAN} + Rainfall + CC_{species} + CC_{Sowing_date} + Cycle + Irrigation, \\ random = 1|ID$$

$$HTFI \sim Y_{IC} + T_{MEAN} + Rainfall + CC_{species} + CC_{Sowing_date} + Cycle + Irrigation, \\ random = 1|ID$$

$$MW_{IR} \sim Y_{IC} + T_{MEAN} + Rainfall + CC_{species} + CC_{Sowing_date} + Cycle + Irrigation, \\ random = 1|ID$$

297 Following a similar procedure to the linear mixed model (Box-Cox transformation, ML
298 estimation for fixed effect selection, and REML estimation for the final parameter
299 estimations), insignificant effects ($\alpha > 0.05$), as well as unlinked variables, were progressively
300 removed up to reach the global goodness-fit-criteria (Fisher's C test).

301

302 **2.8. Data availability**

303 All the data, including observations and sugarcane management, were described in a
304 companion paper (Ngaba et al., 2023), and are freely available in the CIRAD dataverse
305 “Agroecological Practices to reduce wEED infestAtion in the tropIcS” (APEEDAIS,
306 <https://dataverse.cirad.fr/dataverse/APEEDAIS>). The dataset links separated per trial were
307 listed in supplementary materials.

308

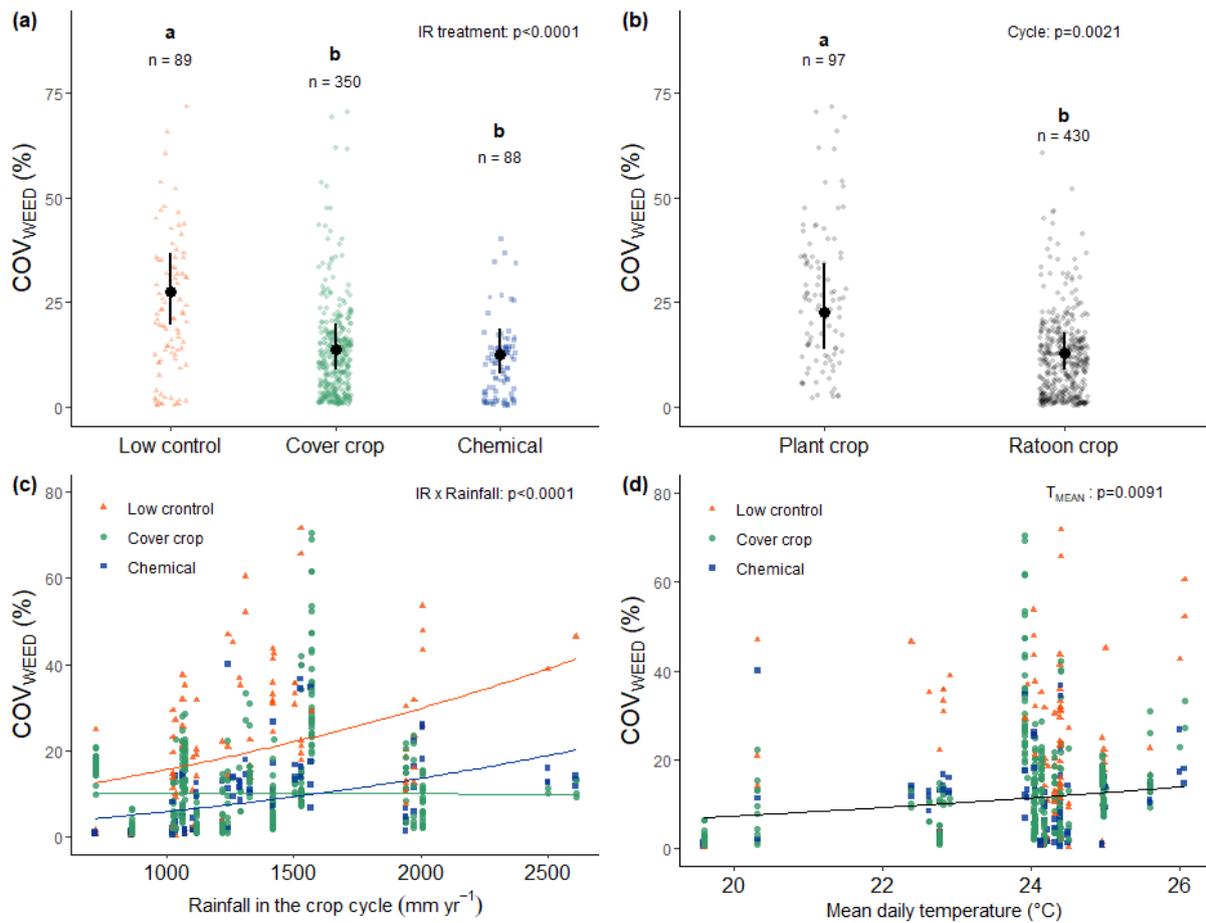
309 **3. Results**

310 **3.1. COV_{WEED} response to inter-row treatment, climate, and sugarcane management**

311 Mean annual ground cover by weeds (COV_{WEED}) varied from 0 to 75% depending on trial and
312 inter-row treatment (Fig. 1). COV_{WEED} was influenced by the inter-row treatment ($p < 0.0001$),
313 T_{MEAN} ($p = 0.0091$), the total rainfall ($p = 0.0127$), the crop cycle ($p = 0.0021$), the interactions
314 between the treatment and the rainfall ($p < 0.0001$), and the interaction between the experiment
315 and the inter-row treatment ($p = 0.0068$, Table A9). On average, COV_{WEED} was reduced by
316 51% in both cover crop and chemical treatments, compared to low control treatment (Table
317 A10, Fig. 1a). Nonetheless, COV_{WEED} was not influenced by the inter-row treatment in two
318 experiments (P25_ITK and Ecocanne_OB) out of eight, and all treatments were similar in
319 both experiments (Table A11). COV_{WEED} was 43% lower in ratoon crops than in plant crops,
320 with mean values of 13% and 23% in ratoon crops (characterized by a sugarcane trash mulch

321 cover) and plant crops (a first-year without mulch cover), respectively (Table A9, Fig 1b).
 322 COV_{WEED} increased with the total rainfall over the crop cycle in both low control and
 323 chemical treatment but not in the cover crop treatment (Table A9, Fig. 1c). COV_{WEED}
 324 increased with T_{MEAN} in all treatments (Table A9, Fig. 1d).

325



326

327 **Fig. 1.** Mean ground cover by weeds (COV_{WEED}) response to inter-row management
 328 treatment (IR, a), sugarcane crop cycle (plant or ratoon crop, b), rainfall (c), and mean daily
 329 temperature (T_{MEAN} , d) during the crop cycle. In (a) and (b), the mean values (black point)
 330 and confidence intervals (black line) predicted by the mixed model with the trial as a random
 331 effect are presented. The number of measurements in each category is indicated by “n”.
 332 Letters indicated significant differences in the pairwise comparisons (Table A10). In (c) and

333 (d), the lines represented the predicted response by the model after back-transformation. P
334 values of the fixed effects in the variance analysis are indicated (Table A9).

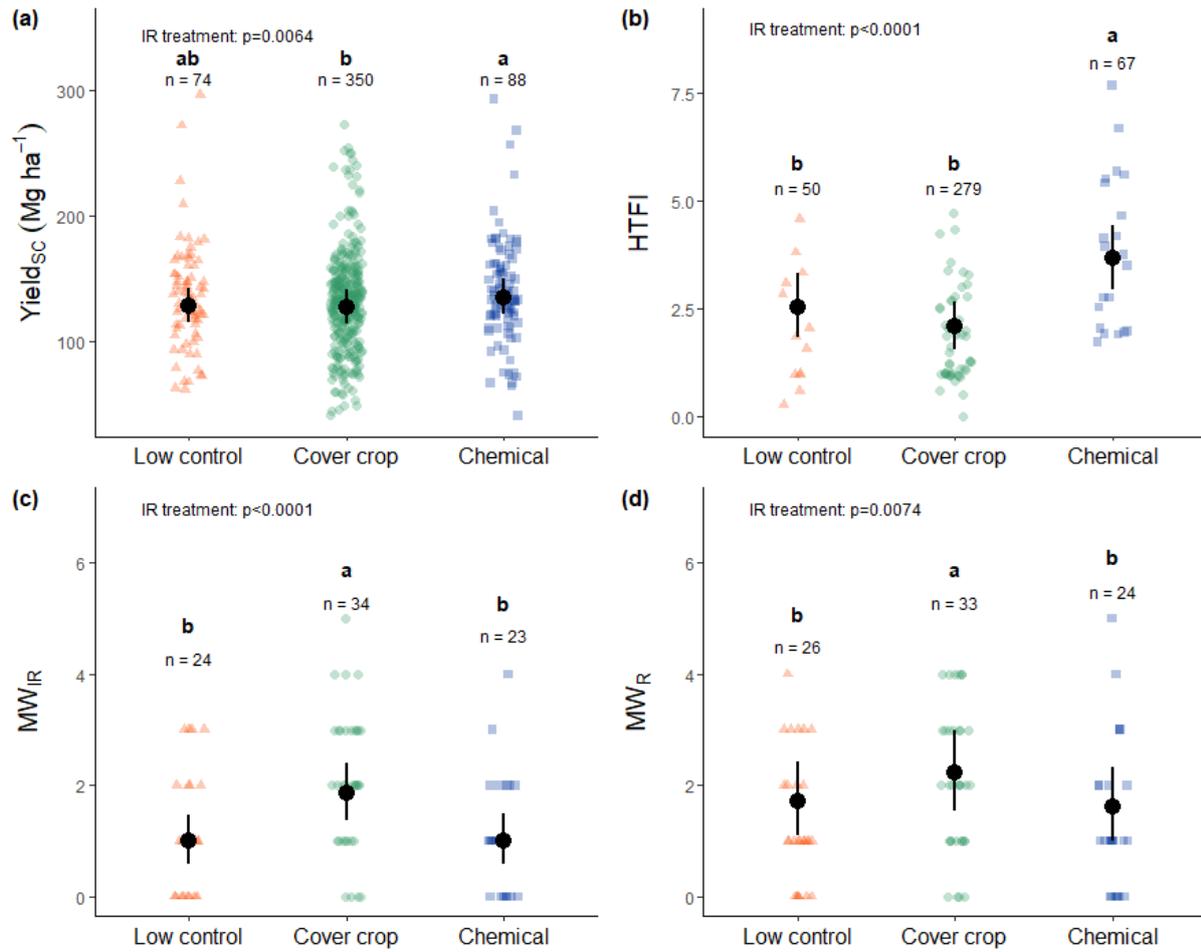
335

336 **3.2. Yield_{SC}, HTFI and manual weeding response to inter-row treatment**

337 The sugarcane yield (Yield_{SC}), the herbicide frequency treatment index (HTFI) and the
338 number of manual weeding in the sugarcane inter-row (MW_{IR}) and in the row (MW_R) were
339 influenced by the inter-row treatment (Fig. 2, Table A9, A10). Average Yield_{SC} across trials
340 were 135, 127, and 128 Mg FM ha⁻¹ in chemical, cover crop, and low control treatments,
341 respectively (Fig. 2a). Yield_{SC} was significantly lower by 6% in the cover crop treatment
342 compared to chemical one. The HTFI was reduced by 37% in the cover crop and low control
343 treatments compared to chemical treatment (Fig. 2b). Nonetheless, the number of manual
344 weeding in the cover crop treatment increased by 86% and 34% in the inter-row and row,
345 respectively (Fig. 2c,d), compared to both low control and chemical treatments. No climate or
346 sugarcane management influences were observed on HTFI, MW_{IR}, and MW_R.

347

348



349

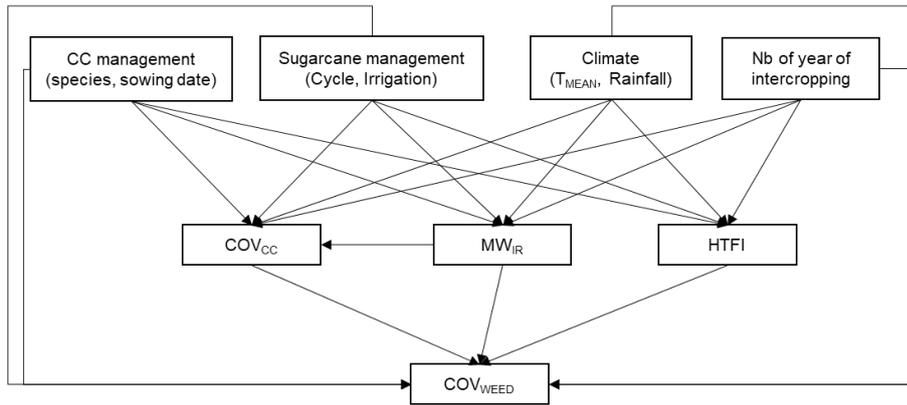
350 **Fig. 2.** Sugarcane yield (Yield_{SC}, a), herbicide treatment frequency index (HTFI, b), and
 351 number of manual weedings in the inter-row (MW_{IR}, b) and row (MW_R, c) depending on the
 352 inter-row management treatment. The mean values (black point) and confidence intervals
 353 (black line) predicted by the linear mixed models with the trial as a random effect are
 354 presented. The number of measurements in each category is indicated by “n”. Letters
 355 indicated significant differences in the pairwise comparisons (Table A10). P values of the
 356 fixed effects in the variance analysis are indicated (Table A9).

357

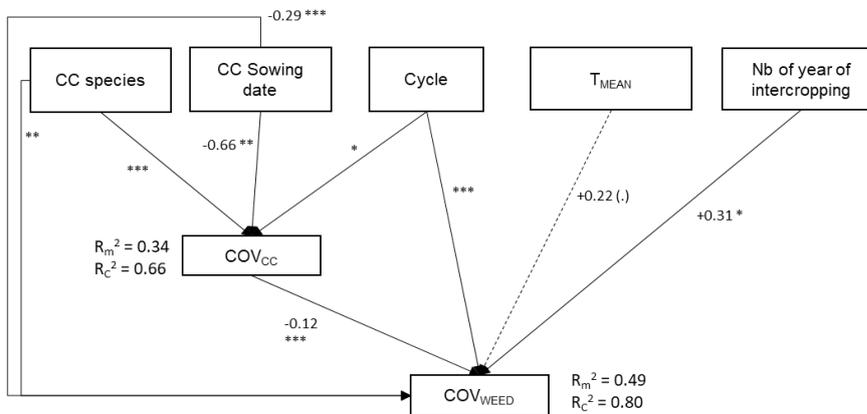
358 **3.3. Structural equation models on COV_{CC} and COV_{WEED}**

359 In the cover crop treatment, COV_{WEED} was directly influenced by COV_{CC} , the crop cycle, the
360 number of years since the implementation of intercropping (Y_{IC}), the cover crop sowing date
361 and species, and COV_{CC} was directly influenced by cover crop sowing date, species, and crop
362 cycle (Fig. 3b, Table A12, A15). COV_{WEED} was also indirectly influenced by cover crop
363 sowing date and species through COV_{CC} . These retained paths respected the Fisher global
364 goodness-of-fit criteria in the SEM with all the data (Fischer C = 0.76, $p=0.94$, Table A12) as
365 well as in SEM using only plant crop data (Fischer C = 2.9, $p=0.23$, Table A13) or ratoon
366 crop data (Fischer C = 3.9 $p=0.42$, Table A14). These fixed factors explained 49% and 34%
367 of COV_{WEED} and COV_{CC} variability, respectively (Fig. 3b). Nonetheless, the respective
368 influence of each factor depended on the crop cycle (Fig. 3c,d). In plant crops, the cover crop
369 sowing date affected COV_{CC} while not in ratoon crops, and this factor had a prominent effect
370 on COV_{WEED} in plant crops compared to ratoon crops (see estimates in Table A13 and A14).
371 The cover crop species directly influenced COV_{WEED} in ratoon crops but not plant crops.

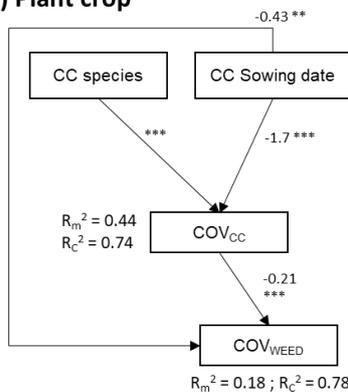
(a) Hypothesis



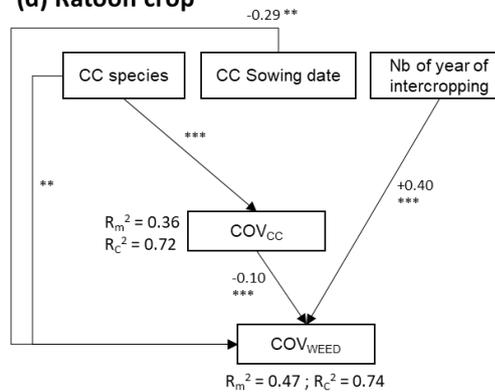
(b) Cover crop treatment



(c) Plant crop



(d) Ratoon crop



372

373 **Fig. 3.** Structural equation model (SEM, a) showing direct and indirect effects of cover crop
 374 management, weed management, and environmental conditions on the ground cover by weeds
 375 (COV_{WEED}) and cover crop (COV_{CC}) in the cover crop treatment considering the hypothesis
 376 (a), final SEM with all data (b), plant crop data (c) or ratoon crop data (d). Arrows represent
 377 the structural equation with variables: mean annual temperature (T_{MEAN}), rainfall, crop cycle,

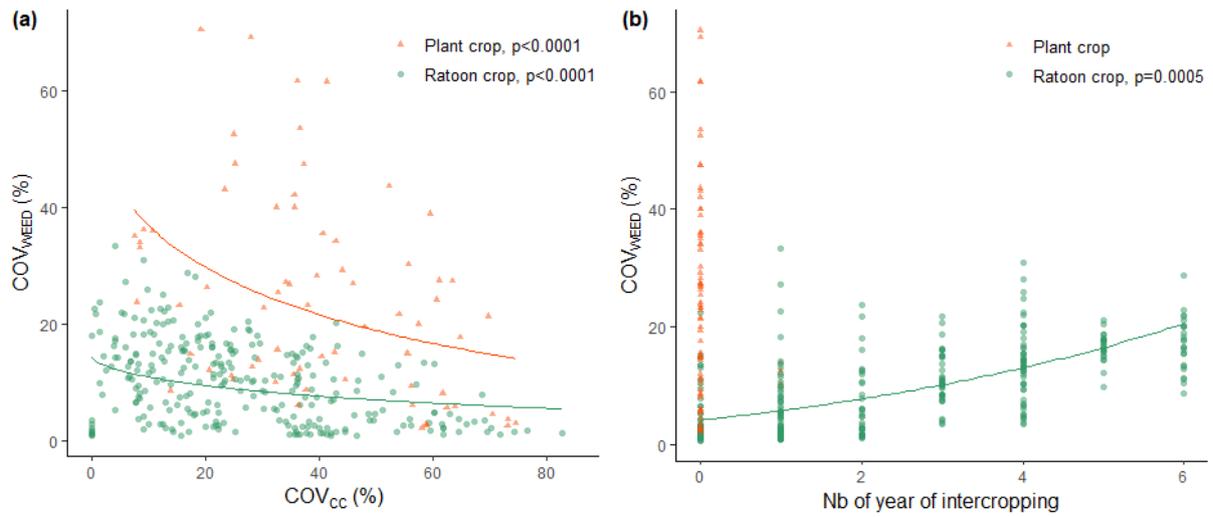
378 irrigation management, cover crop (CC) species, and sowing date, number (nb) of years since
379 the implementation of intercropping, the number of manual weedings in the inter-row
380 (MW_{IR}), and the herbicide treatment frequency index (HTFI). In (b), (c), and (d), the arrows
381 represent the significant result of the analysis. The mode estimates are given for quantitative
382 variables (after Box-Cox transformation), and p values are indicated by ***, **, *, and (.)
383 when lower than 0.001, 0.01, 0.05, and 0.1, respectively. Marginal R² (R_m²) and conditional
384 R² (R_C²) are given. Details on model estimates for qualitative variables and variance analyses
385 are given in Supplementary Material (Table A12 to A14).

386

387 **3.4. Weeds response to cover crops growth and species**

388 A negative impact of COV_{CC} on COV_{WEED} was shown in the structural equation model in
389 both plant and ratoon crops (Fig. 3). The SEM predicted a decrease in COV_{WEED} from 39 to
390 14% in plant crops, and 14 to 5% in ratoon crops, with increasing COV_{CC} (Fig. 4a). COV_{CC}
391 was influenced by the cover crop species in both plant and ratoon crops (Fig. 3, 5a). In
392 contrast, COV_{WEED} was only influenced by the cover crop species in the ratoon crops (Fig. 3,
393 5b). In plant crops, the highest COV_{CC} were found in cover crop mixture (56%), and *V.*
394 *unguiculata* (49%), followed by *C. ensiformis* (42%), and the lowest COV_{CC} was found in *D.*
395 *intortum* (29%, Fig. 5a). In ratoon crops, cover crop species with the highest COV_{CC} was *C.*
396 *ensiformis* (35%) followed by *C. juncea* (33%), *C. spectabilis* (29%) and *R. sativus* (28%).
397 The lowest COV_{CC} were found for *G. abyssinica* (11%), *V. unguiculata* (13%), and *B. juncea*
398 (near 0%). In ratoon crops, COV_{WEED} was similar among all CC species except for *C.*
399 *ensiformis*, which has a lower COV_{WEED} (8%) than *V. unguiculata* (11%, Fig. 5b).

400



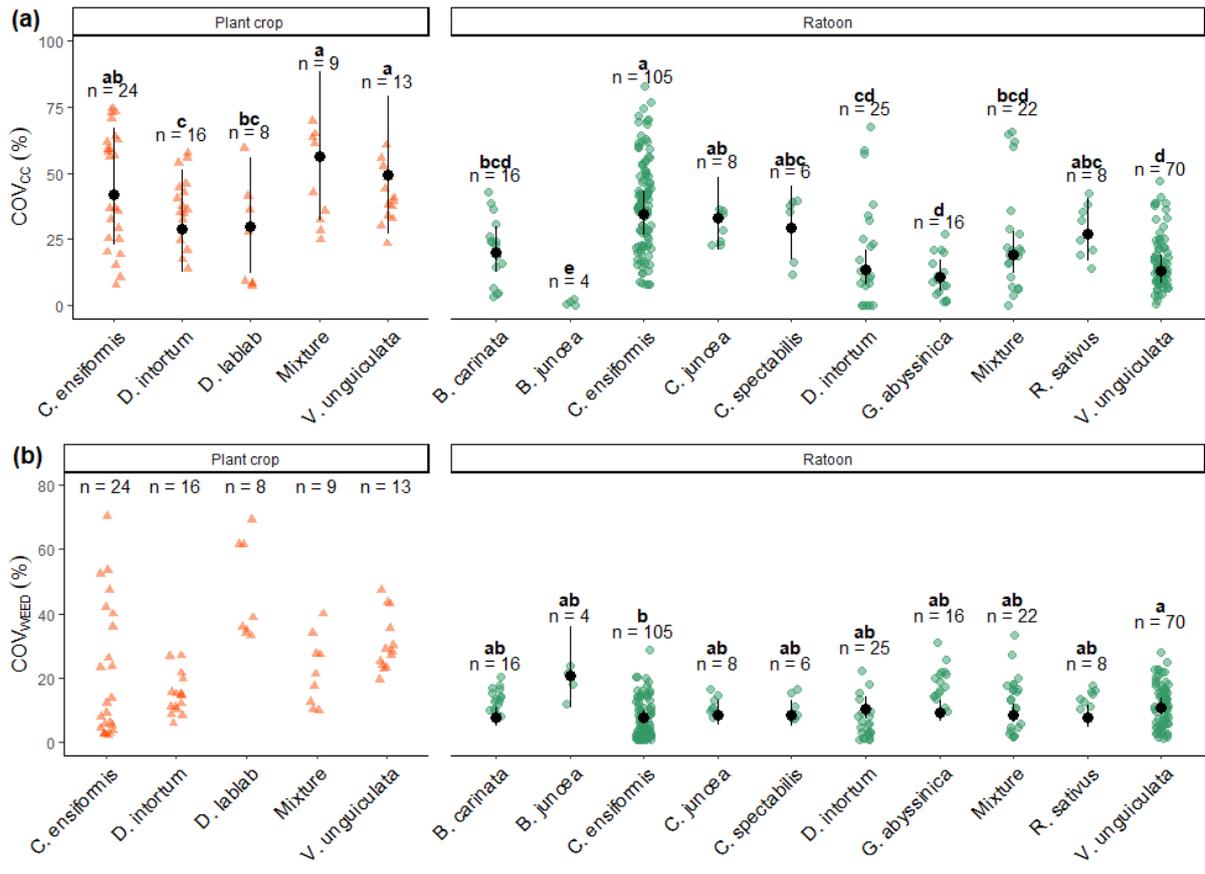
401

402 **Fig. 4.** Mean ground cover by weeds (COV_{WEED}) in the cover crop treatment in the inter-row
 403 in response to the mean ground cover by cover crop (COV_{CC} , a) and the number (Nb) of years
 404 since the implementation of intercropping (b). The lines represented the predicted response by
 405 the SEM models after back-transformation, and the P values are indicated.

406

407

408



409

410 **Fig. 5.** Mean ground cover by cover crop (COV_{CC} , a) and weeds (COV_{WEED} , b) in the cover
 411 crop treatment depending on the cover crop species in plant and ratoon crops. The mean
 412 values (black point) and confidence intervals (black line) predicted by the SEM are presented
 413 depending on cover crop species. The number of measurements in each category is indicated
 414 by “n”. Letters indicated significant differences in the pairwise comparison with a Tukey p-
 415 adjustment method.

416

417

418

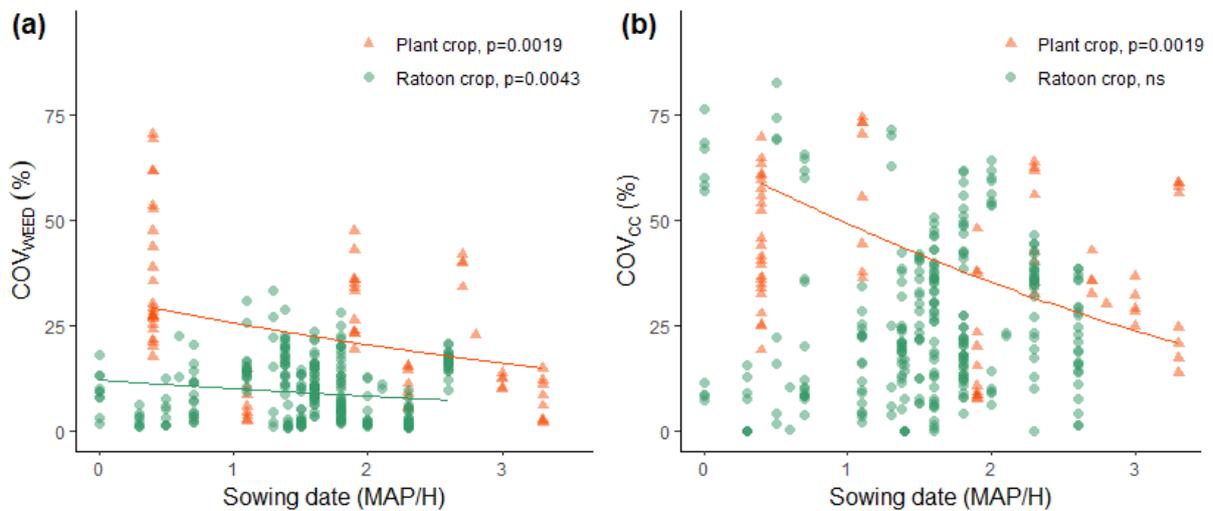
419

420

421 **3.5. Weeds response to cover crop management**

422 The structural equation model showed a direct effect of cover crop sowing date on COV_{WEED}
423 in plant and ratoon crops (Fig. 3, 6a) and a direct effect of the number of years of
424 intercropping on COV_{WEED} (Fig. 3, 4b). COV_{CC} was also influenced by the sowing date but
425 only in plant crops (Fig. 3, 6b). In plant crops, the impact of the sowing date on COV_{WEED}
426 was higher than in ratoon (see coefficient in Fig. 3c,d). COV_{WEED} decreased in plant crops by
427 around 5% per month after sugarcane planting and by around 2% in ratoon (Fig. 5a).
428 Similarly, COV_{CC} decreased by around 13% per sowing month after planting in plant crops
429 (Fig. 5b). COV_{WEED} also increased with the number of years since the implementation of
430 intercropping by around 3% per years (Fig. 4b).

431



432

433 **Fig. 6.** Mean ground cover by weeds (COV_{WEED} , a) and cover crop (COV_{CC} , b) in the cover
434 crop treatment depending on the sowing date of the cover crop (in months after sugarcane
435 planting or harvest, MAP/H). The lines represented the predicted response by the SEM
436 models after back-transformation, and the P values are indicated.

437

438 **3.6. Changes in costs, revenue, and profits**

439 Total operating costs in the cover crop treatment increased by 1762, 1149, and 196 € ha⁻¹
440 compared to chemical treatments in manual farms, farms with micro-tractors, and farms with
441 tractors, respectively (Table 3). This increase was mainly due to the additional cost of sowing
442 and mulch management. The decrease in operating costs for chemical weeding (herbicide
443 purchase and herbicide application) in CC compared to CH was relatively similar to the
444 increase in costs due to seed purchase and additional manual weed weeding. By opposition,
445 the costs in LC were slightly lower than CH due to decreased herbicide use. Nonetheless, in
446 both CC and LC, the revenue was lower by 884 and 609 € ha⁻¹, respectively, compared to CH
447 due to decreased sugarcane yield. Consequently, the profits were lower in CC and LC
448 treatments than in CH. While the type of farm had little impact on costs in the LC treatment,
449 the highest profit reductions were observed in manual farms and farms with micro-tractors for
450 the CC treatment.

451

452

453 **Table 3.** Average operating costs and revenue (€ ha⁻¹) depending on the cropping systems:
 454 chemical weeding (CH), cover crop sown in the inter-row (CC), or low control in the inter-
 455 row (LC), for three types of farms (manual, with micro-tractors or tractors). The change in
 456 costs, revenue, and profits (difference between revenue and costs) between CC or LC
 457 compared to CH are presented.

		Manual farms			Farms with micro-tractors			Farms with tractors		
		CH	CC	LC	CH	CC	LC	CH	CC	LC
Operating costs (€ ha ⁻¹)	Herbicide	160	87	92	160	87	92	160	87	92
	Herbicide application	1056	838	952	569	452	513	246	196	222
	Additional weeding	381	451	352	381	451	352	381	451	352
	Seed purchase	0	172	0	0	172	0	0	172	0
	Sowing	0	1130	0	0	416	0	0	77	0
	Mulch management	0	681	0	0	681	0	0	0	0
	Total costs	1597	3359	1396	1110	2259	957	787	983	666
Revenue (€ ha ⁻¹)	Yield revenue	13862	12978	13253	13862	12978	13253	13862	12978	13253
Change (€ ha ⁻¹)	Change in costs		1762	-201		1149	-153		196	-121
	Change in revenue		-884	-609		-884	-609		-884	-609
	Change in profits		-2646	-408		-2033	-456		-1080	-488

458

459

460

461 **4. Discussion**

462 **4.1. Weed control by cover crops in sugarcane intercropping systems**

463 In our cover crop treatments, weed control was performed by combining preliminary
464 herbicide application (whole plot), cover crop development, and manual weeding
465 (particularly for the tall grasses considered as strongly harmful). Nonetheless, our approach
466 highlights the prominent role of cover crop development in reducing weed infestation in our
467 tropical conditions at the same level as chemical treatments.

468 Despite having a limited number of studies in the sugarcane literature examining the
469 effectiveness of cover crops for inter-row weed management, these findings agree with
470 previous studies showing that cover crops showed weed suppression efficiency higher than
471 30% in India (Geetha et al., 2019), and that sugarcane intercropping allow to limit weed
472 biomass at the same level than chemical treatments (Kaur et al., 2016, 2015; Mahadevaswamy
473 and Martin, 2002). Similar observations were made in other intercropping systems in
474 temperate climatic conditions, with a decrease in weed biomass by 56% compared to non-
475 weeded cropping systems in a meta-analysis on various cash crops (mainly wheat, maize, and
476 barley, Verret et al., 2017).

477 The negative relationship between the weeds and cover crop ground covers observed in our
478 study highlights how the ability of cover crops to limit weed growth is directly linked to its
479 development in the inter-row. These results agree with the broad agreement on cover crop
480 fallow or intercropping that high biomass, cover, or a fast ground cover by cover crops
481 increases weed control (Gardarin et al., 2022; Osipitan et al., 2019). Even if only the mean
482 ground cover by cover crops was tested in this study, the main species used in intercropping
483 are known to have high biomass and rate of increase in ground cover in Reunion Island
484 (Christina et al., 2021b; Négrier et al., 2023).

485 While the choice of cover crop species appeared important regarding cover crop development
486 in our study, the impact on weed growth was relatively constant. Such a statement on weed
487 growth can result from the impact of complementary manual weedings, even if we did not see
488 a link between MW_{IR} and COV_{WEED} in our SEM approach. Both legumes (*C. ensiformis*,
489 *Crotalaria* species) and non-legume species (*R. sativus*) showed high covering ability. In
490 particular, *C. ensiformis* have a high growth rate and biomass production, which can explain
491 its smothering effect (Lawson et al., 2006). Soares et al. (2022) also showed that *C. ensiformis*
492 reduced the diversity of weed species. *C. juncea* has a potential allelopathic effect that can
493 suppress weeds and the main crops (Skinner et al., 2012). The weed-smothering efficiency of
494 sunn hemp (*C. juncea*) and cowpea (*V. unguiculata*) was also observed in previous studies in
495 India (Geetha et al., 2019). *B. carinata* and the mixtures of cover crops were also adapted to
496 limit weed growth. Nonetheless, other species, such as *D. intortum* or *G. abyssinica*, were less
497 adapted to intercropping.

498

499 **4.2. Climate, resource, and weed control in intercropping systems**

500 In our climatic conditions, rainfall and temperature influenced weed development. Such a
501 response was expected in our tropical conditions (Peters et al., 2014). Nonetheless, while
502 COV_{WEED} increased in the low control and chemical treatments with rainfall, it did not change
503 in the cover crop treatments. These results suggest that other factors limit weed development
504 in the cover crop treatments.

505 In fertilized tropical conditions, one hypothesis is that light partitioning is a primary factor
506 determining the competition between weeds and cover crops (Christina et al., 2021b; Tardy et
507 al., 2015). Light competition can explain the differences in weed cover observed in our study
508 in response to cover crop ground cover and between the plant (slow sugarcane canopy

509 closure, around six months after planting) and ratoon crops (fast sugarcane canopy closure,
510 around 3-4 months after harvest), even if a strong effect of mulch should also be expected
511 (Carvalho et al., 2017). The plant or ratoon crop cycle also influenced cover crop growth in
512 our study, illustrating the importance of light competition between sugarcane and cover crops
513 (Viaud et al., 2023). Previous studies have also shown that cover crops limit the emergence of
514 weeds due to a decrease in the quantity and quality of light transmitted through the cover
515 (Benvenuti et al., 2001; Juroszek and Gerhards, 2004). Accordingly, the cover crop sowing
516 date strongly influenced crop development in plant crops in our study compared to ratoon
517 crops, where the sugarcane dominancy limits cover crop growth. Nonetheless, as suggested by
518 our SEM results, the response of weeds to cover crop sowing dates is more complex. In our
519 study, when the cover crop sowing date was delayed, chemical weeding was performed before
520 sowing, thus decreasing the mean weed ground cover over the crop cycle (direct effect). An
521 opposite effect also occurred; mean cover crop cover decreased by delaying the sowing date
522 (and so increased weed ground cover over the crop cycle, indirect effect).

523 Our spatially limited study did not allow us to assess the influence of soil characteristics on
524 the weed-cover crop competition in sugarcane intercropping systems. Nonetheless, Viaud et
525 al. (2023) showed a higher impact of intercropping on sugarcane yield with lower soil carbon
526 content in our tropical conditions, suggesting that soil nutrient competition could occur
527 despite fertilization. The influence of soil resource availability on the weeds x cover crops x
528 main crop competition process has to be further investigated as it is still poorly understood
529 and often limited to nitrogen (Corre-Hellou et al., 2011; Hauggaard-Nielsen et al., 2001).
530 Crop modeling could give helpful insights into understanding how and when competition for
531 resources occurred between plants over the sugarcane growth cycle and also help disentangle
532 the effects of these competitions regarding climate or crop management effects.

533

534 **4.3. Sugarcane intercropping management.**

535 As shown in our study, the combination of cover crops and manual weeding makes it
536 possible to reduce by 37% the frequency of herbicide application in sugarcane cropping
537 systems. However, intercropped species must be carefully selected, and their management
538 adapted because of the risk of excessive inter-specific competition on sugarcane yield
539 depending on soil and climatic conditions (Viaud et al., 2023).

540 Considering cover crop species selection and management, our study suggests that many
541 species would be adapted to sugarcane intercropping regarding weed control. Nonetheless,
542 irrigation failure can affect cover crop emergence and development, highlighting the need to
543 select drought-resistant species such as *C. ensiformis*. Some species, such as *C. juncea*, can
544 also be infested by aphids limiting their development. Cover crop mixtures could help
545 maintain ground cover in case of choosing one sub-optimal species (Négrier et al., 2023) and
546 increase mixture resilience to unexpected events (Elhakeem et al., 2021). Regarding sowing
547 management, our study suggests that early sowing will provide higher cover crop
548 development in plant crops. However, a meta-analysis suggested that early sowing will also
549 induce a higher sugarcane yield reduction in intercropping compared to conventional
550 management (Viaud et al., 2023). Our analysis did not allow us to accurately compare cover
551 crop sowing management, such as the number of sown lines or the sowing method (manual
552 vs. mechanical). Nonetheless, cover crop sowing geometry should be assessed further as
553 previous studies have shown that it can increase cover crop growth with little impact on
554 sugarcane yield (Nadeem et al., 2020; Tian et al., 2020).

555 While using cover crops limits weed growth in the inter-row, weeding sugarcane row remains
556 a major issue. Herbicides can be sprayed in the sugarcane row to control weeds but with a
557 high risk of affecting cover crop development. Consequently, to limit weed growth in the row
558 and very harmful weed species in the inter-row, more manual weeding were performed in the

559 sugarcane row and inter-row compared to chemical control. Such interventions increase weed
560 control costs, but these are, on average, offset by lower herbicide costs. Adding all the costs
561 and changes in revenue linked to yields, profits from intercropping would be lower than for
562 chemical farming in La Réunion. The main reason for this is the cost of sowing and the
563 decrease in yield. Consequently, our results suggest that by adapting the crop management to
564 reduce the impact on yield (Viaud et al., 2023) and favoring sowing directly through the
565 sugarcane mulch (as in farms with tractors), the profits could be less impacted by a change of
566 practice from chemical to intercropping. In other countries, some studies have shown that
567 additional returns from the intercrop could also increase the net profits compared to sugarcane
568 monocropping (Kaur et al., 2015; Nazir et al., 2002; Singh et al., 2008). Therefore, harvesting
569 the cover crop in La Réunion could be evaluated to see the impact on the economic balance
570 (considering additional labor and mechanization costs). Finally, while this study focused on
571 weed control in sugarcane intercropping systems, crop diversification is known to have the
572 potential to provide various ecosystem services (Beillouin et al., 2021; Isbell et al., 2017;
573 Tamburini et al., 2020), some of which have already been studied in sugarcane intercropping
574 systems: associated soil biodiversity (Pang et al., 2022), soil fertility and greenhouse gas
575 emissions (Luo et al., 2016), or soil erosion (Wang et al., 2022). One way of encouraging the
576 adoption of intercropping by improving its profitability would be to work on payments for
577 ecosystem services (Mota et al., 2023; Schomers and Matzdorf, 2013).

578 In opposition to the long-term effect of legume intercropping on soil fertility (Luo et al.,
579 2016), the long-term effect of intercropping on weed pressure has not been investigated in
580 sugarcane intercropping systems. Our results highlighted how the increase in weed cover with
581 years of intercropping questioned the long-term sustainability of this practice. Indeed, they
582 can explain the previous results of increased sugarcane yield loss through years in sugarcane-
583 legume intercropping systems compared to monocropping (Viaud et al., 2023). The authors'

584 hypothesis to explain such an impact was an increasing weed pressure over time. This
585 increase could result from a weed seed bank enrichment and/or the selection of highly
586 competitive weeds, as seen in other temperate cover cropping systems mainly under rotations
587 (Graziani et al., 2012; Mohler et al., 2018). Nonetheless, while some studies attempted to
588 assess the response in weed communities in short-term intercropping (Gomez and Gurevitch,
589 1998; Poggio, 2005; Stefan et al., 2021), the long-term impact of intercropping on change in
590 weed communities still need to be assessed.

591

592 **Conclusions**

593 Sugarcane intercropping in Reunion Island allowed reducing weed cover by 51% compared to
594 low control treatments and reduced herbicide treatment frequency index by 37% compared to
595 chemical treatments, at the cost of an increased number of manual weedings and a 6%
596 decrease in sugarcane yield. In intercropping, crop management significantly influences weed
597 cover more than the climate in our tropical conditions. Changes in operating costs and
598 revenue led to a reduction in profits compared with chemical treatment, which varied
599 according to the type of farm (manual or mechanized). The lack of a relationship between
600 manual weedings and weed cover suggests investigating the effects of further reduction of
601 costly manual weedings in future research. Additionally, the increase in weed cover with
602 years of intercropping highlights the need to assess the long-term impact of intercropping on
603 weed communities to assess this practice's sustainability.

604

605

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616

617 **Author contributions**

618 **Mathilde Soulé:** Investigation, Data Duration, Writing - Original Draft, Vizualisation. **Alizé**
619 **Mansuy:** Conceptualization, Methodology, Investigation, Project administration, Funding
620 acquisition. **Julien Chetty:** Conceptualization, Methodology, Investigation. **Sandrine**
621 **Auzoux:** Data Curation, Visualization. **Pauline Viaud:** Formal analysis, Data Curation,
622 Visualization. **Marion Schwartz & Aude Ripoché:** Resources, Visualization. **Benjamin**
623 **Heuclin:** Formal analysis. **Mathias Christina:** Methodology, Formal analysis, Data Curation,
624 Writing - Original Draft, Visualization, Supervision. **All authors:** Writing - Review &
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