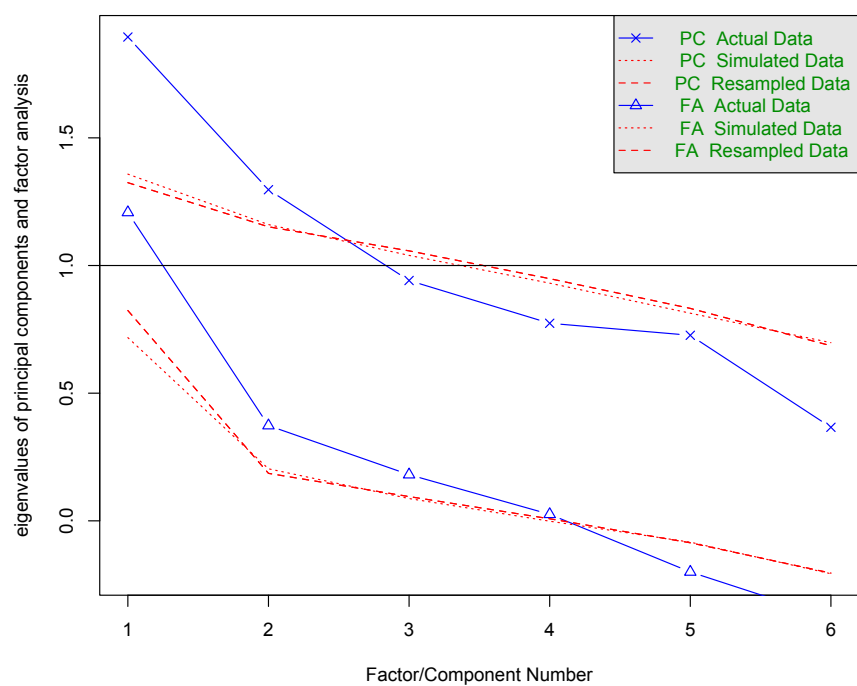


1 **Apêndice 3.3**

2 **Confirmation of Environmental Constructs**

3 We validated the choice of two environmental constructs using the routines for scree plots compared
4 to random parallel matrices (using the function *fa.parallel*), the Very Simple Structure (VSS) and
5 Velicer’s Minimum Average Partial (MAP) criteria using the function *vss*. The parallel analysis
6 suggests that the number of factors is three and the number of components is two (Figure S1).



7
8 Figure S1. Parallel analysis scree plots of environmental constructs from rural interviewees and
9 environmental specialists in the Brazilian Cerrado savannas.

10 A principal component analysis reveals that the first principal component is linked to a
11 sustainability/biodiversity conservation awareness and the second to climate change awareness
12 (Figure S2).

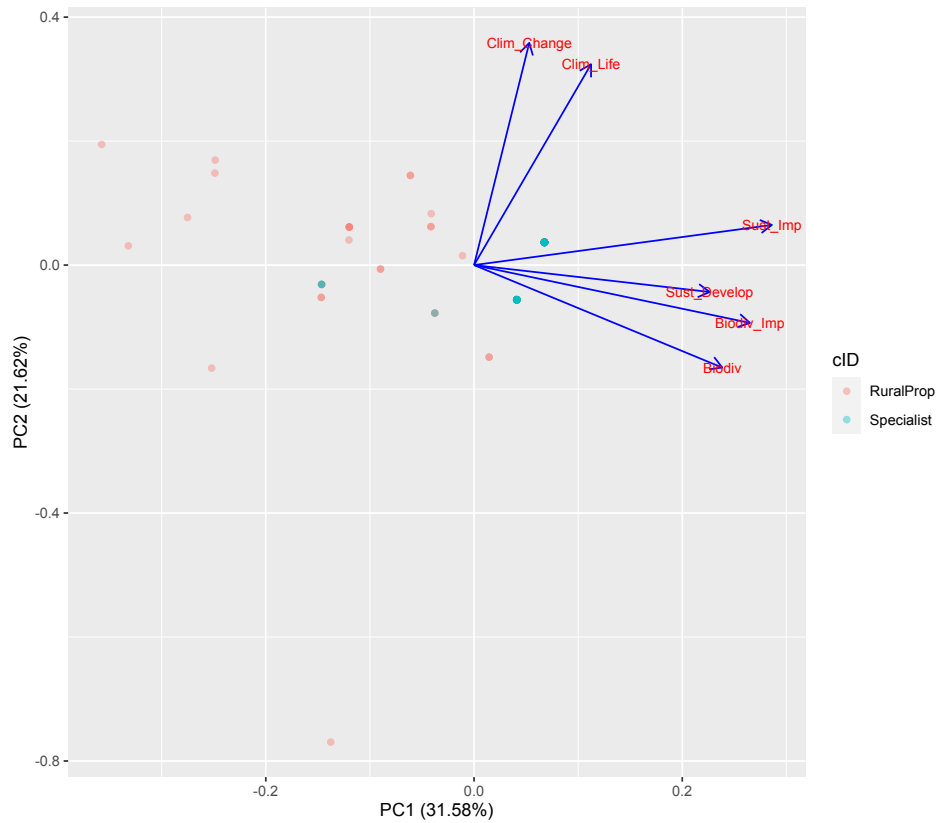
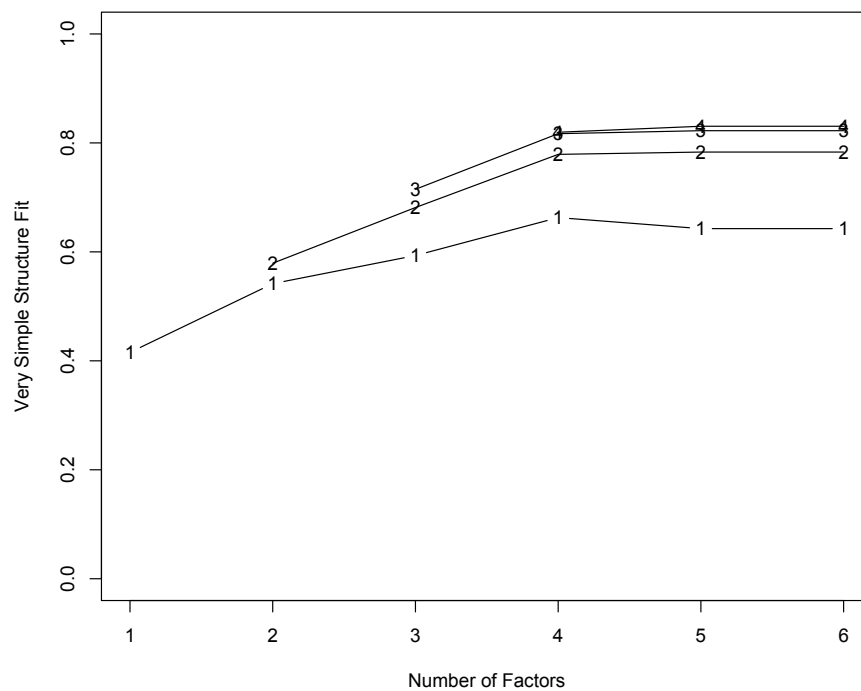


Figure S2. Biplot of principal component analysis (PCA) of environmental constructs from rural interviewees and environmental specialists in the Brazilian Cerrado savannas.

The VSS criterion achieves its maximum with four factors (Figure S3) in every complexity level. The Velicer's MAP achieves its minimum (0.069) with one factor.



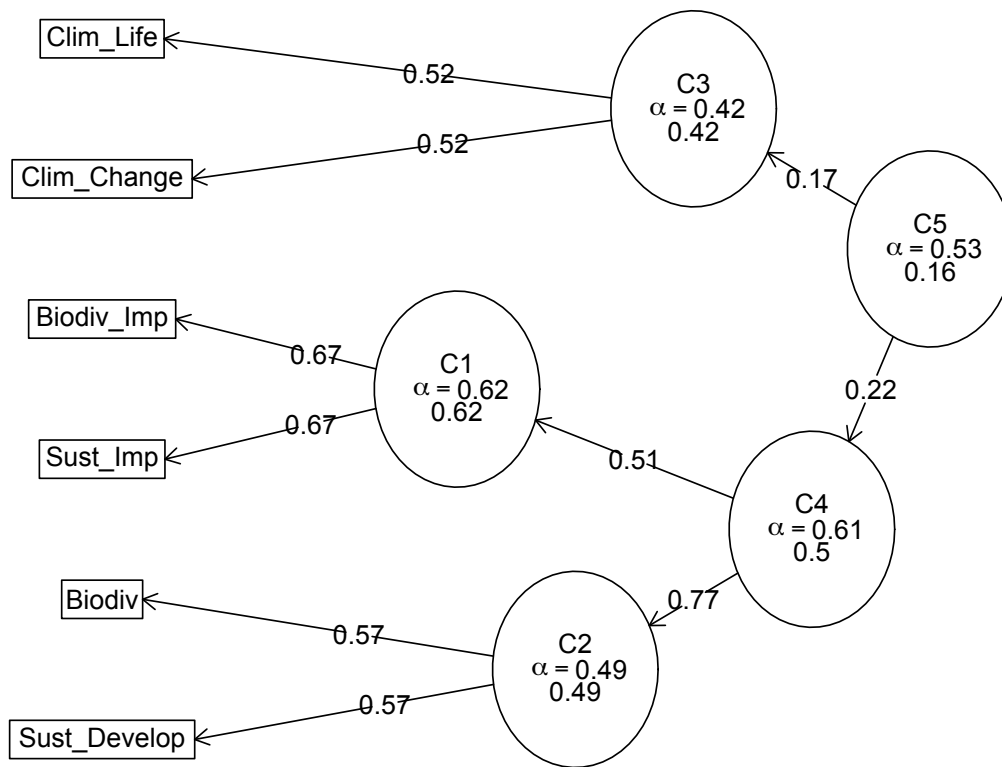
18

19 Figure S3. Very Simple Structure (VSS) fits with different complexities of an exploratory factor
 20 analysis to depict environmental constructs from rural interviewees and environmental specialists in
 21 the Brazilian Cerrado savannas.

22 We also made hierarchical cluster analyses (using the function *iclust*), essentially a VSS factor model
 23 of complexity one. We could recover the climate change and sustainability/conservation awareness
 24 constructs with clusters C3 and C4, respectively (Figure S4). We can further split the C4 cluster in
 25 one cluster of importance value and another of knowledge (C1 and C2, respectively).

26

27



28

29 Figure S4. Hierarchical cluster analysis of the survey items to depict environmental constructs from
 30 rural interviewees and environmental specialists in the Brazilian Cerrado savannas. The values in
 31 the arrows indicate the correlations between the clusters and items. Alpha and Beta values are
 32 estimates of the reliability and general factor saturation of the test.

33

34 Later, we calculated McDonald's omega estimates of general and total factor saturation and
 35 hierarchical factor analyses with two and three factors using the function *omega*. The omega measures
 36 the reliability of the general model considering the hierarchical structure of the inventory (McDonald,
 37 1999). The general model with two specific factors recovers only one general factor (Figure S5) with
 38 omega total = 0.61 and hierarchical = 0.07. However, with three specific factors the omega total
 39 increases to 0.73 and hierarchical to 0.41 with the general factor relating more to the
 40 sustainability/conservation awareness construct (Figure S6). Despite the data suggest more than two
 41 constructs, we preferred to use a simpler model because they are hierarchically related, as the bi-

factor and cluster analyses reveal. Therefore, our exploratory analyses agree with our hypothesis of two environmental constructs, one related to sustainability/conservation awareness and another to climate change awareness.

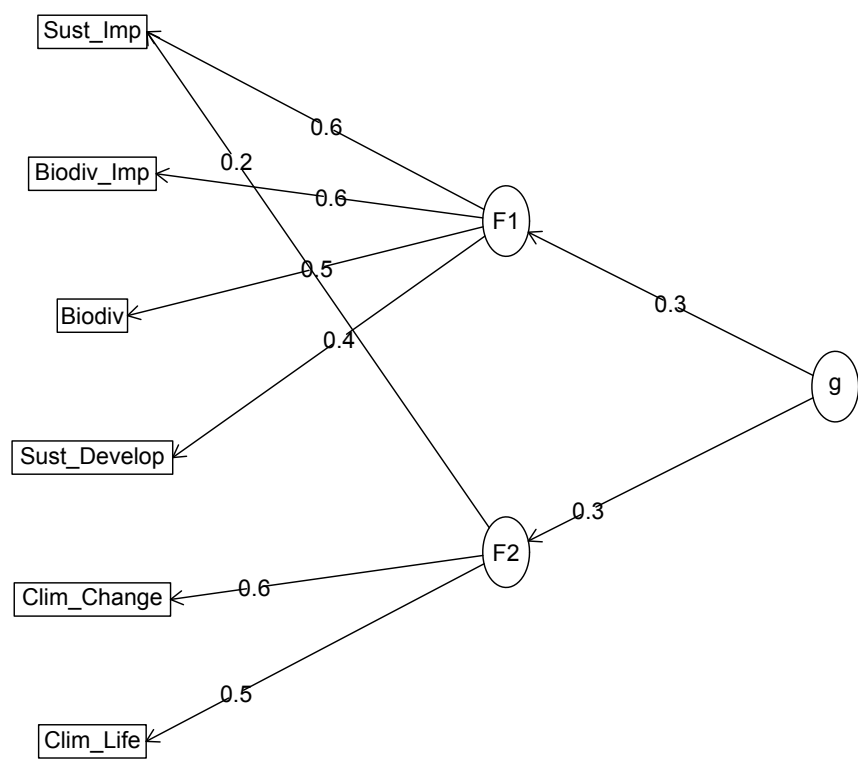


Figure S5. Hierarchical factor analysis with two factors (F1 and F2) and a general factor (g) of the survey items to depict environmental constructs from rural interviewees and environmental specialists in the Brazilian Cerrado savannas.

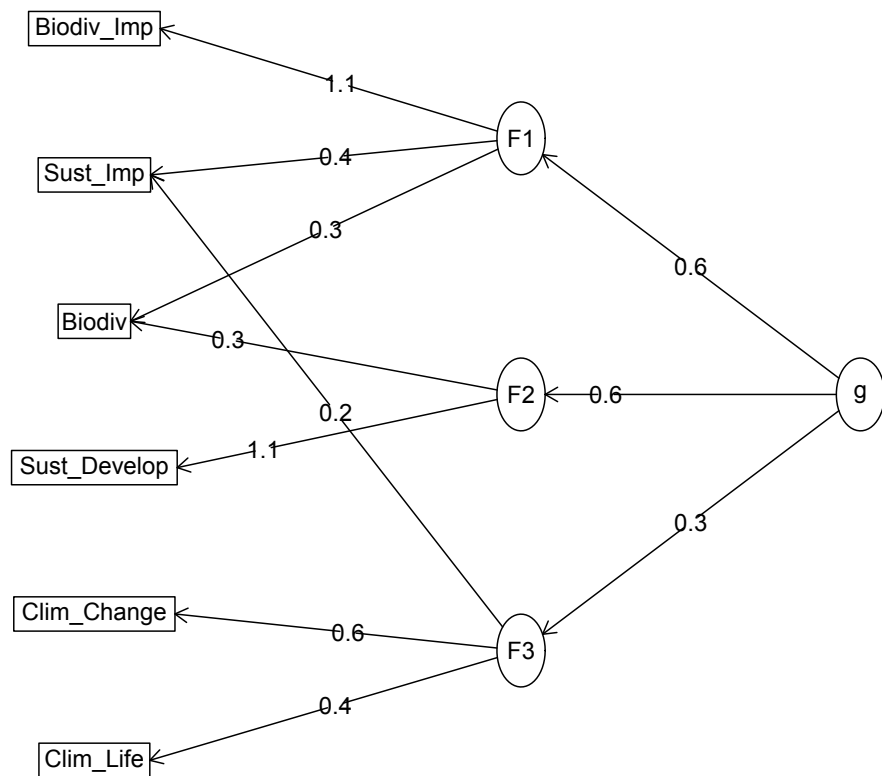


Figure S6. Hierarchical factor analysis with three specific factors (F1, F2, and F3) and a general factor (g) of the survey items to depict environmental constructs from rural interviewees and environmental specialists in the Brazilian Cerrado savannas.

We can observe the factor analysis recovers the same constructs the PCA and cluster analysis revealed (Figure S5). We also present some summary statistics from the factor analysis with two factors we used in the subsequent analyses (Table S1). The two factors had a correlation of 0.17.

Table S1. Summary statistics of the exploratory factor analysis of the survey items to depict environmental constructs from rural interviewees and environmental specialists in the Brazilian Cerrado savannas.

Variable/Statistic	Factor 1	Factor 2	h^2	u^2	Complexity
Sust_Develop	0.4	0.03	0.17	0.83	1
Biodiv	0.53	-0.15	0.28	0.72	1.2
Sust_Imp	0.59	0.26	0.47	0.53	1.4
Biodiv_Imp	0.62	-0.05	0.37	0.63	1
Clim_Change	-0.09	0.56	0.3	0.7	1.1
Clim_Life	0.05	0.46	0.22	0.78	1
SS loadings	1.2	0.62	-	-	-
Proportion of variance	0.2	0.1	-	-	-
Proportion explained	0.66	0.34	-	-	-
Correlation of regression scores with factors	0.81	0.69	-	-	-
Multiple R^2 of scores with factors	0.66	0.48	-	-	-
Minimum correlation of possible factor scores	0.31	-0.04	-	-	-

63

64 **Consistency Ratios of the Analytical Hierarchical Process (AHP)**

65 We present the variation of the priority rankings of the general priorities (Figures S7 and S8) and fire
66 management goals (Figures S9 and S10) with the total sampling and removing individuals with low
67 consistency ratio (truncated). We can observe the main patterns in prioritization do not change
68 between rural residents and environmental specialists. The truncated data resemble the Bayesian
69 multilevel model estimates, indicating the inter-individual variance and inconsistencies were
70 appropriately accounted.

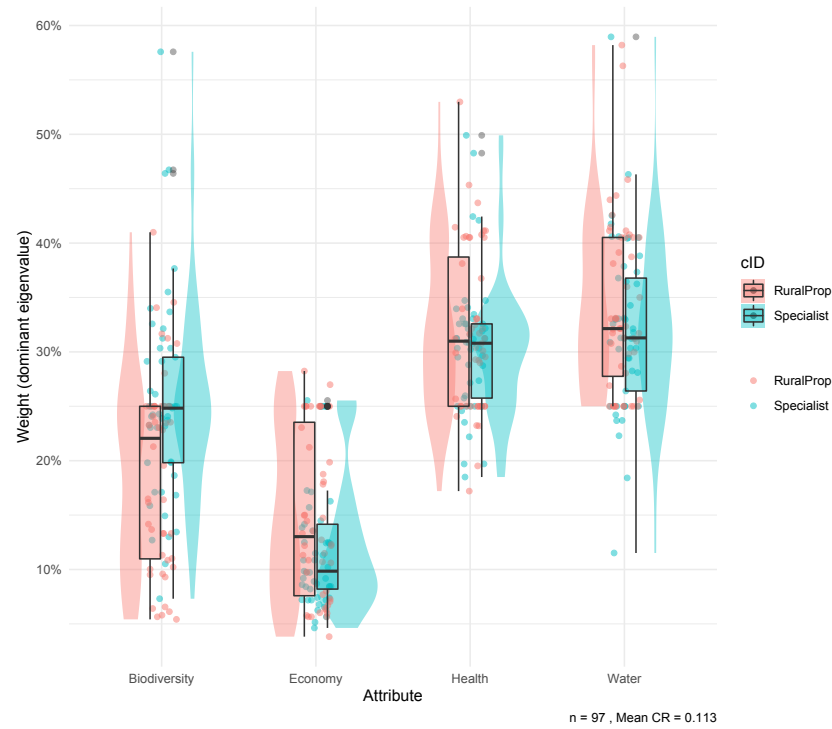


Figure S7. Boxplot and density of the total sampled ranking weights of rural interviewees and environmental specialists to prioritize general priorities in the Brazilian Cerrado savannas.

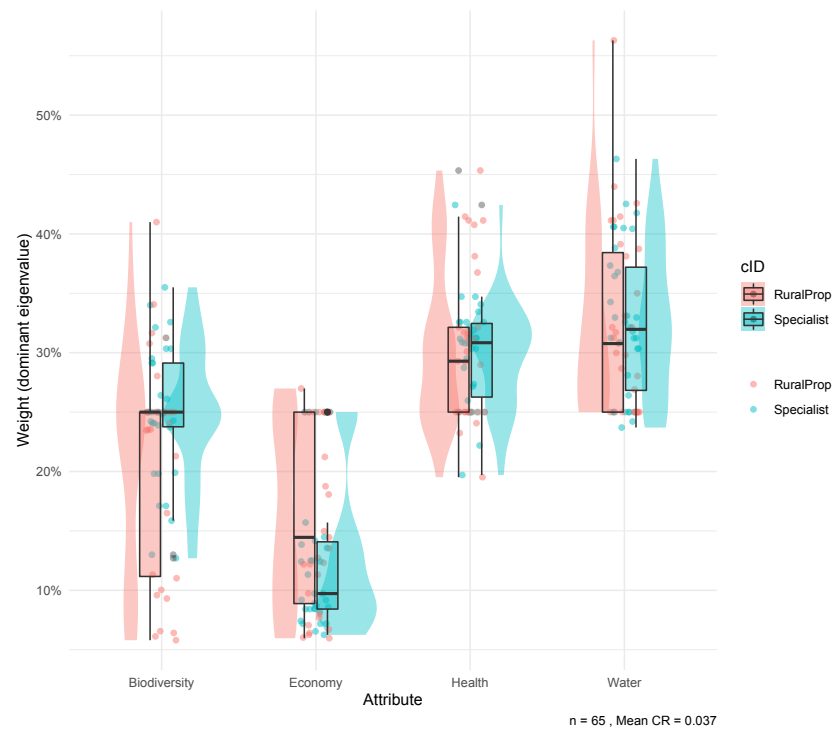
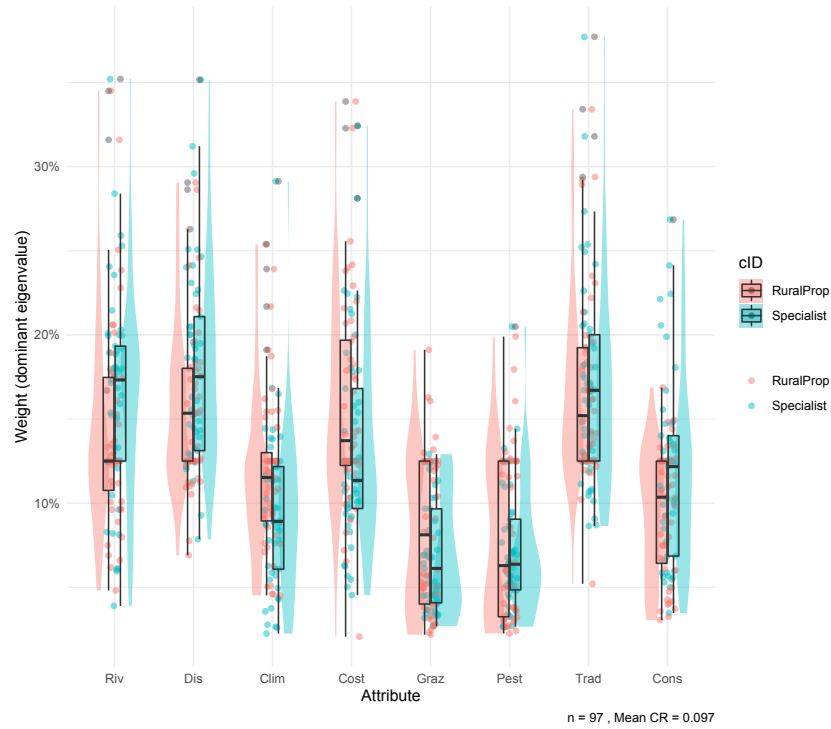
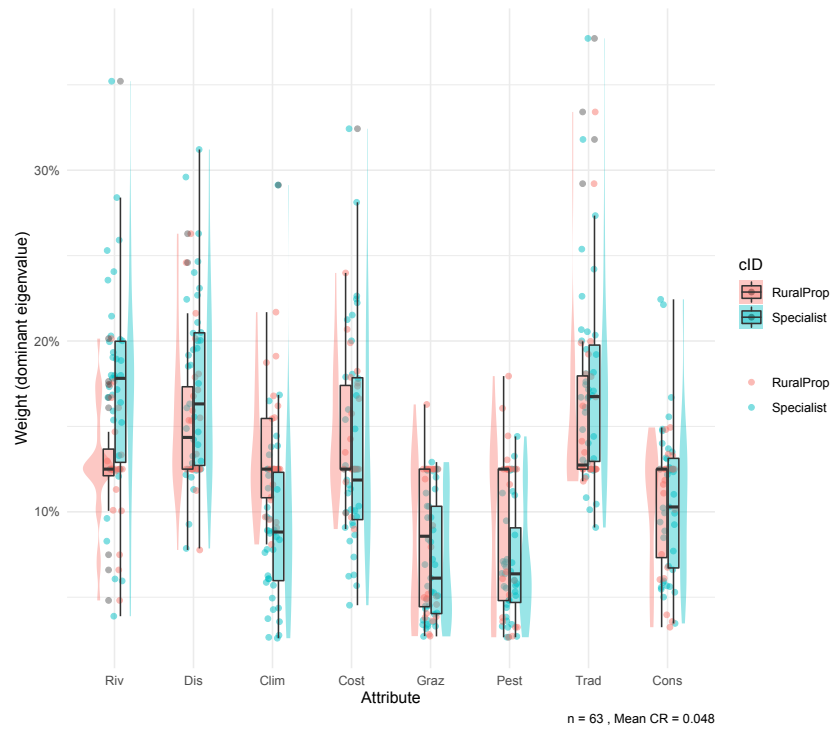


Figure S8. Boxplot and density of the truncated ranking weights of rural interviewees and environmental specialists to prioritize general priorities in the Brazilian Cerrado savannas.



77

78 Figure S9. Boxplot and density of the total sampled ranking weights of rural interviewees and environmental specialists
79 to prioritize fire management goals in the Brazilian Cerrado savannas.



80

81 Figure S10. Boxplot and density of the truncated ranking weights of rural interviewees and environmental specialists to
82 prioritize fire management goals in the Brazilian Cerrado savannas.

83 **Leave-One-Out (LOO) Cross Validations**

84 In our Bayesian multilevel models to assess group (rural interviewees and environmental specialists),
85 socioeconomic (gender, age, income, and educational level), and environmental constructs
86 (sustainability/conservation and climate change awareness) effects on the probability of rural
87 interviewees and environmental specialists to prioritize general priorities (Table S2) and fire
88 management goals (Table S3) and judge fire importance use (Table S4) we compared with Leave-
89 One-Out (LOO) cross validations. We can observe that all the effects we assessed on the general
90 priorities have better explanation than a null model (Table S2). The socioeconomic effects alone are
91 better than a full model with all predictors (Table S2).

92
93 Table S2. Bayesian Leave-One-Out estimates and standard errors of the expected log pointwise predictive density
94 (ELPD) of models assessing *group* (rural interviewees and environmental specialists), socioeconomic (*gender*, *age*,
95 *income*, and *educ*—educational level), and environmental constructs (*sust/cons*—sustainability/conservation and *clim*—
96 climate change awareness) effects on the probability of rural interviewees and environmental specialists to prioritize
97 general priorities ($AHP_{general}$) in the Brazilian Cerrado savannas. Higher ELPD estimates mean higher predictive
98 ability.

Model	ELPD estimate	SE	ELPD diff.	SE diff.
$AHP_{general} \sim gender + age + income + educ$	356.9	14.2	0	0
$AHP_{general} \sim group + sust/cons + clim + gender + age + income + educ$	350.6	15	-6.3	6.1
$AHP_{general} \sim group$	348.7	11.8	-8.2	6.3
$AHP_{general} \sim sust/cons + clim$	344.8	12.8	-12.1	8
$AHP_{general} \sim 1$	343.4	12.5	-13.5	7.6

99

100 Regarding the ranking prioritization of fire management goals, we can observe that the environmental
101 constructs do not affect them significantly (Table S3).

102

103 Table S3. Bayesian Leave-One-Out estimates and standard errors of the expected log pointwise predictive density
 104 (ELPD) of models assessing *group* (rural interviewees and environmental specialists), socioeconomic (*gender*, *age*,
 105 *income*, and *educ*—educational level), and environmental constructs (*sust/cons*—sustainability/conservation and *clim*—
 106 climate change awareness) effects on the probability of rural interviewees and environmental specialists to prioritize
 107 fire management goals (AHP_{fire}) in the Brazilian Cerrado savannas. Higher ELPD estimates mean higher predictive
 108 ability.

Model	ELPD estimate	SE	ELPD diff.	SE diff.
$AHP_{fire} \sim group + sust/cons + clim + gender + age + income + educ$	1113.7	17.2	0	0
$AHP_{fire} \sim gender + age + income + educ$	1112.4	16.1	-1.3	8.3
$AHP_{fire} \sim group$	1104	15.7	-9.7	11.3
$AHP_{fire} \sim I$	1101.7	15.1	-12	11.1
$AHP_{fire} \sim sust/cons + clim$	1097.1	17.1	-16.6	10.4

109

110 Like the main priorities, all the effects we studied have significant effects on the individuals’
 111 judgment of fire importance (Table S4).

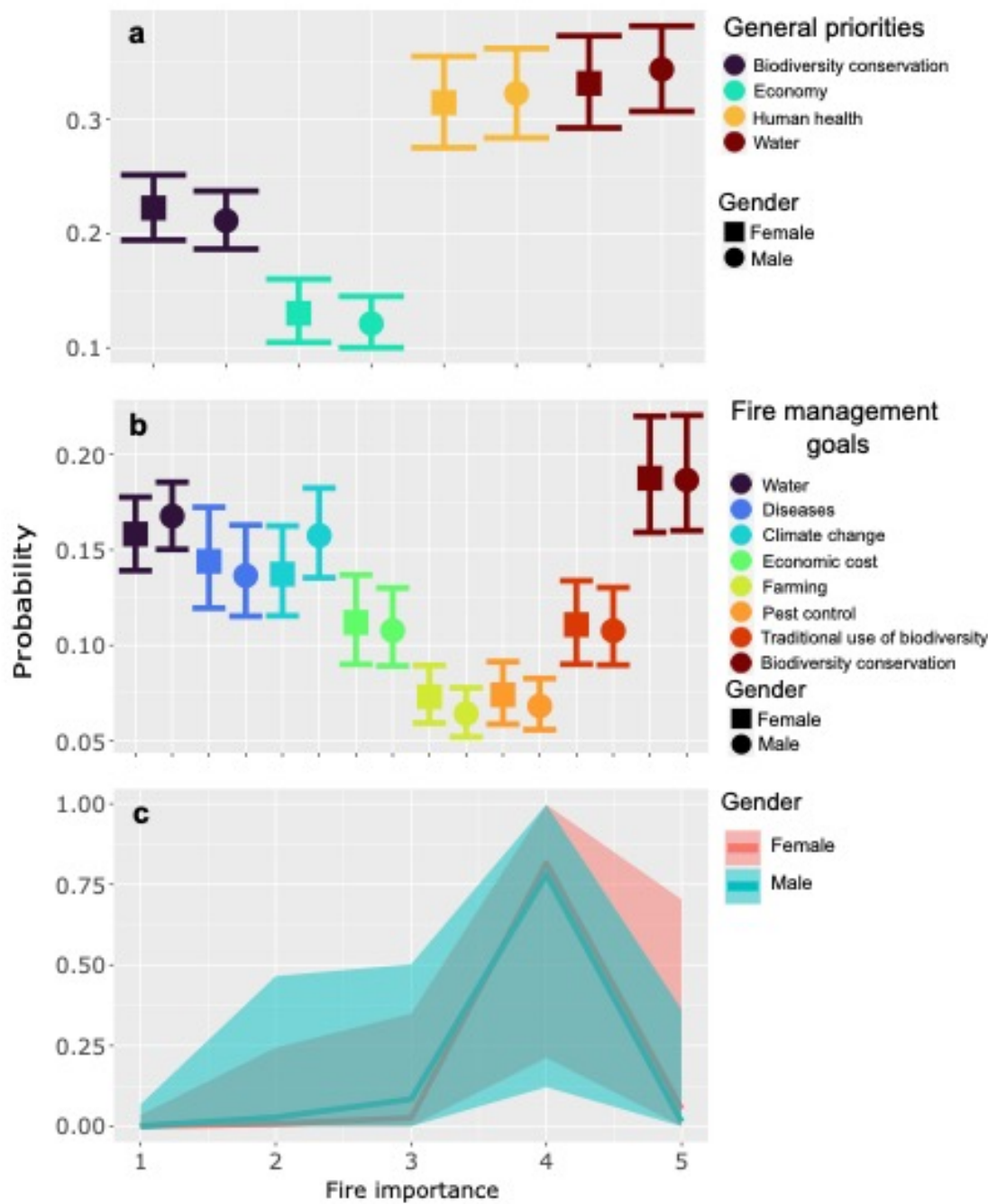
112 Table S4. Bayesian Leave-One-Out estimates and standard errors of the expected log pointwise predictive density
 113 (ELPD) of models assessing *group* (rural interviewees and environmental specialists), socioeconomic (*gender*, *age*,
 114 *income*, and *educ*—educational level), and environmental constructs (*sust/cons*—sustainability/conservation and *clim*—
 115 climate change awareness) effects on the probability of rural interviewees and environmental specialists to judge fire
 116 importance in the Brazilian Cerrado savannas. Higher ELPD estimates mean higher predictive ability.

Model	ELPD estimate	SE	ELPD diff.	SE diff.
$Fire\ importance \sim group + sust/cons + clim + gender + age + income + educ$	-76.9	5.7	0	0
$Fire\ importance \sim gender + age + income + educ$	-104.6	5.7	-27.7	4.2
$Fire\ importance \sim group$	-116.8	7.1	-39.9	2.8
$Fire\ importance \sim sust/cons + clim$	-129.4	5.4	-52.5	4
$Fire\ importance \sim I$	-133.8	4.2	-56.9	4.1

117

118 **Gender effects**

119 We did not find differences between genders on the probability of rural interviewees and
120 environmental specialists to prioritize general priorities (Figure S11a) and fire management
121 (Figure S11b) and judge fire importance use (Figure S11c).



122

123 Figure S11. Differences between genders on the probability to prioritize general priorities (a) and fire management
124 goals (b) and judge fire importance use (c) in the Brazilian Cerrado savannas.

125 **Deepening the prioritization of fire management goals (AHP_{fire})**

126 To deepen our comprehension of the choices of prioritization regarding the fire management goals
127 (AHP_{fire}) from rural interviewees and environmental specialists, we also investigated some factors.
128 For rural interviewees, we tested whether their perceptions about their attitudes towards fire
129 frequency use, applying early fires, and avoiding late fires explained their fire management goals
130 ranking. For environmental specialists, we tested whether their perceptions of fire importance for
131 biodiversity, fire regimes effects (current, early-, and late-dry season fires), and IFM (use in
132 protected and private areas, and its current evaluation) explained their fire management goals
133 ranking.

134 Rural interviewees who used fire more frequently had higher priorities for the fire
135 management goal to decrease human diseases in detriment to biodiversity conservation (Fig. S12a).
136 Rural interviewees who had higher predispositions to apply early dry season fires ranked higher the
137 biodiversity conservation and decrease the occurrence of pulmonary diseases and lower the goals to
138 mitigate climate change and protect the traditional use of biodiversity (Fig. S12b). Rural
139 interviewees who had higher attitudes to avoid late dry season fires prioritized more fire
140 management goals to conserve biodiversity in detriment to control pests (Fig. S12c).

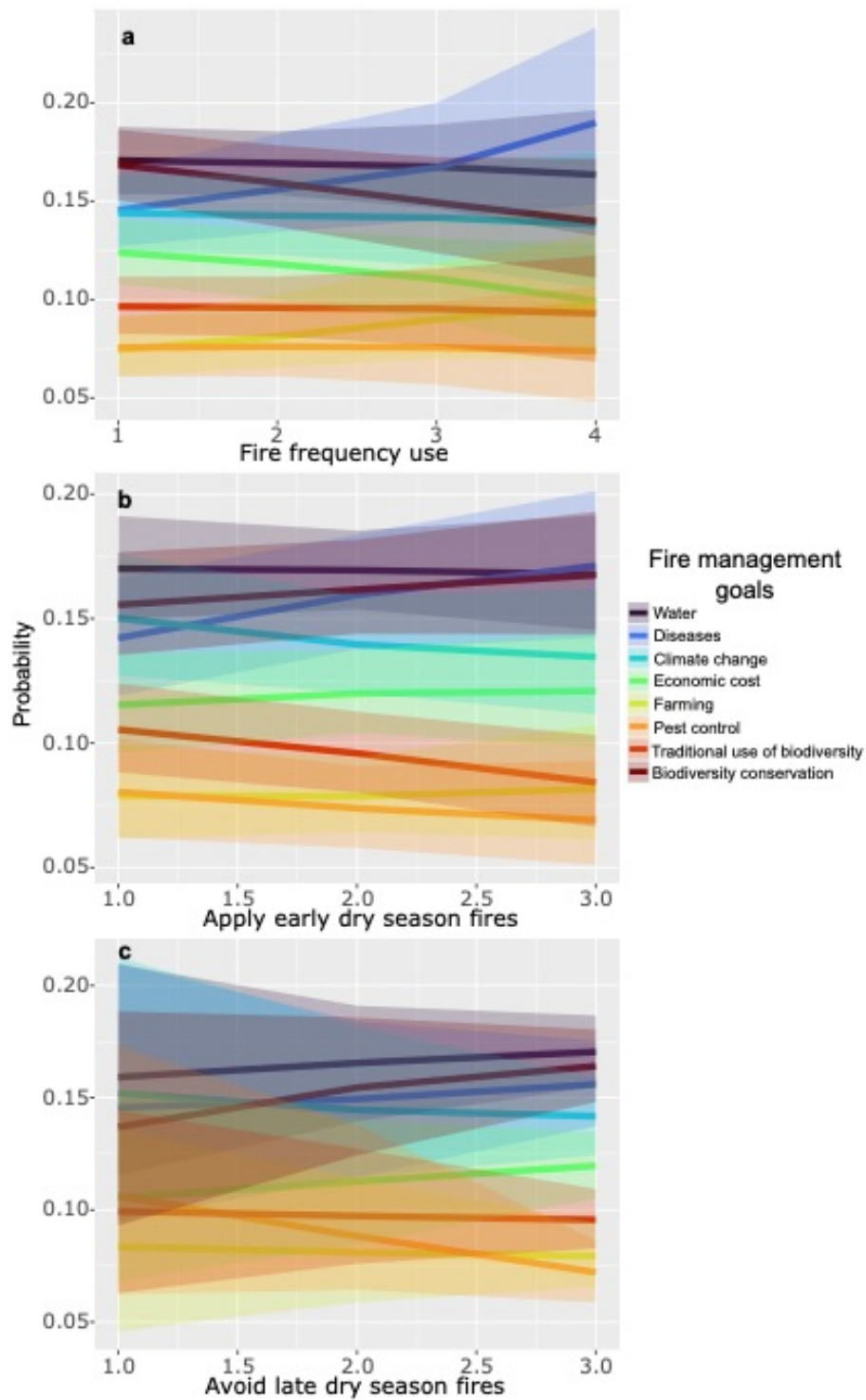


Figure S12. Relationship between rural interviewees' attitudes towards fire frequency use (a), applying early dry season fires (b), and avoiding late dry season fires (c) with the probability to prioritize fire management goals in the Brazilian Cerrado savannas.

Specialists who considered fire highly important to biodiversity ranked higher goals related to the protection of water resources, conservation of biodiversity, and its traditional use (Fig. S13a). Specialists who thought current fire regimes are beneficial to the Cerrado ecosystem ranked lower the fire management goals to protect water courses and higher to protect the traditional use of biodiversity (Fig. S13b). Specialists who thought early dry season fires are beneficial to the Cerrado ecosystem ranked higher the fire management goals related to tackle climate change and protect the traditional use of biodiversity in detriment to biodiversity conservation and protection of watercourses (Fig. S13c). Specialists who stated late dry season fires are beneficial to the ecosystem ranked lower the fire management goals related to tackle climate change and decrease human pulmonary diseases (Fig. S13d).

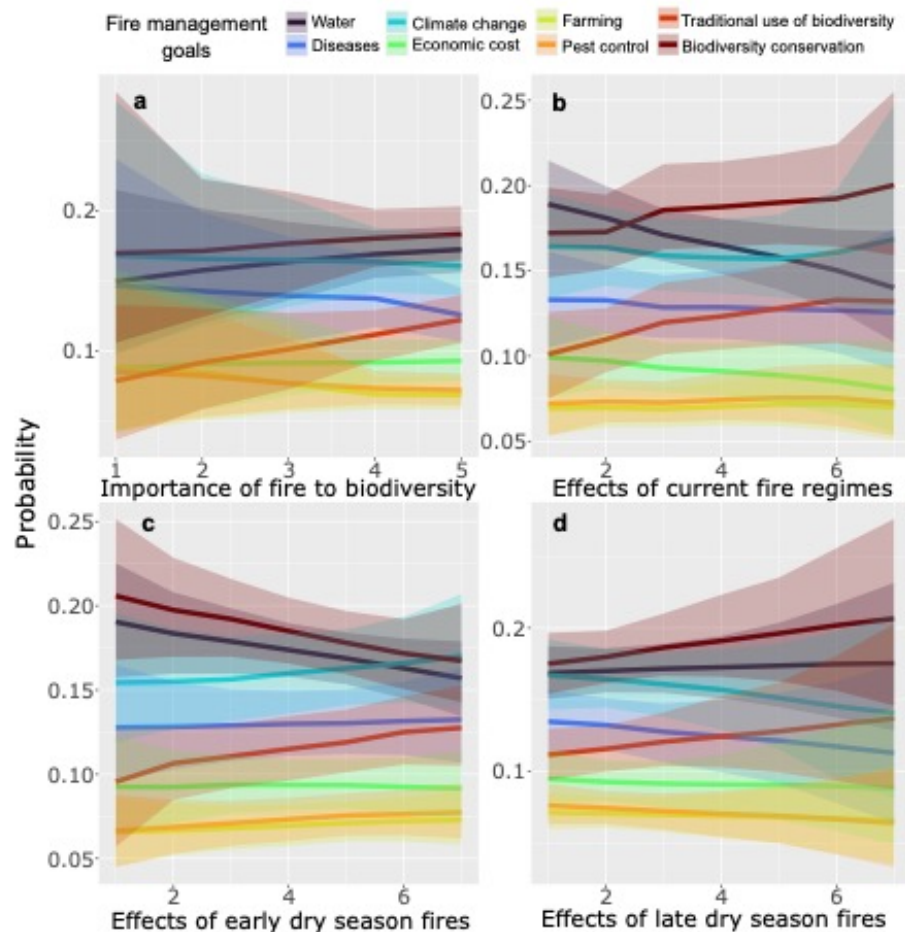
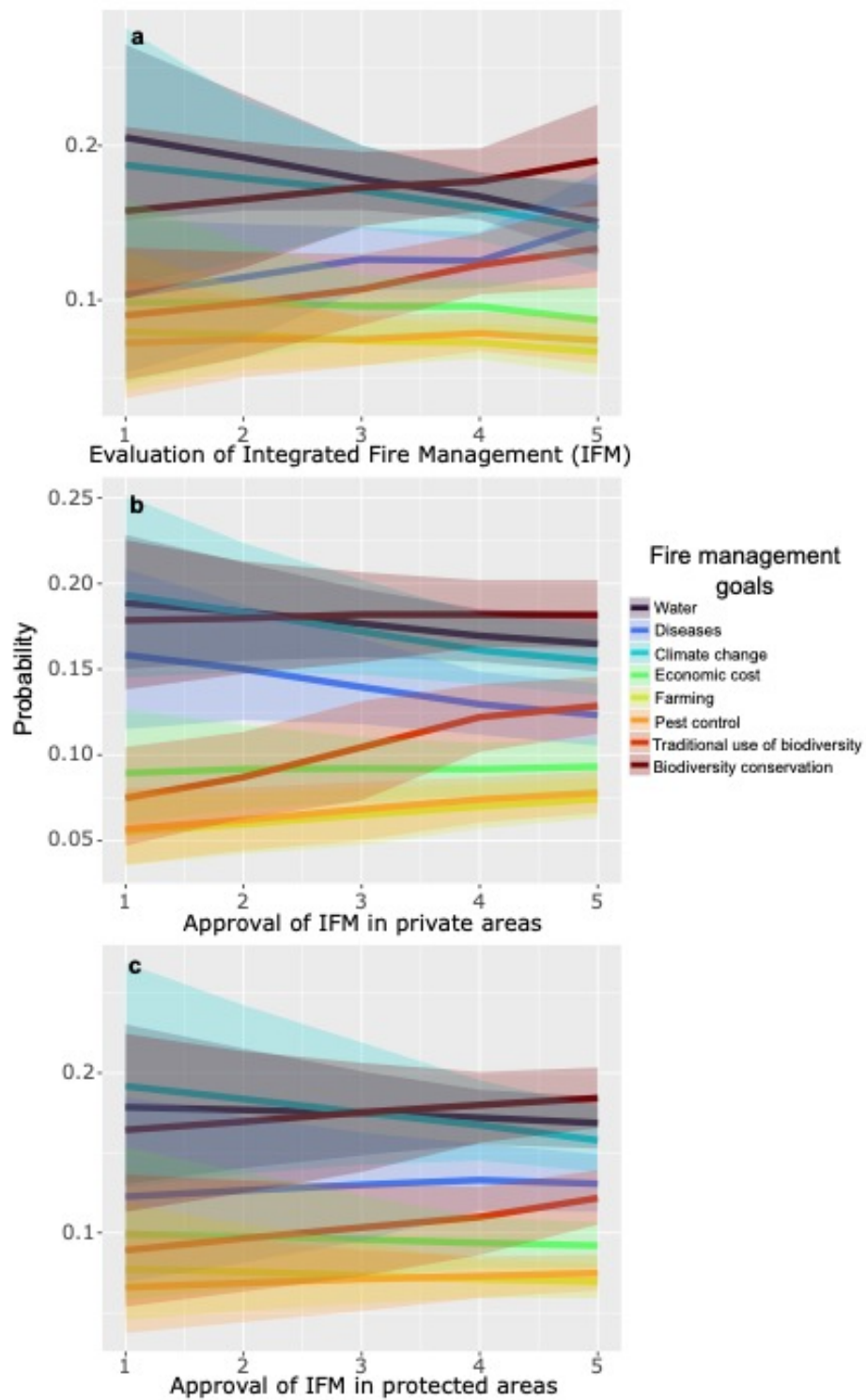


Figure S13. Relationship between environmental specialists' perceptions of the importance of fire to biodiversity (a), effects of current fire regimes (b), early dry season fires (c), and late dry season fires (d) with the probability to prioritize fire management goals in the Brazilian Cerrado savannas.

160 The degree of approval of the current IFM is positively related to prioritizing fire
161 management goals to conserve biodiversity and protect the traditional use of biodiversity in
162 detriment to decrease climate change and protect the water courses (Fig. S14a). People who had
163 higher approval of IFM use in private areas ranked higher the traditional use of biodiversity and
164 lower the ones related to decreasing human pulmonary diseases and climate change (Fig. S14b).
165 Specialists who highly approved the use of IFM in protected areas ranked higher the fire
166 management goals to conserve biodiversity and to protect the traditional use of biodiversity in
167 detriment to climate change (Fig. S14c).



168

169 Figure S14. Relationship between environmental specialists' evaluations of the current Integrated Fire Management (a)
 170 and the approval of its use in private (b) and protected areas (c) in the Brazilian Cerrado savannas with the probability
 171 to prioritize fire management goals in the Brazilian Cerrado savannas.