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2 **So Close Yet So Different: Cultural Differences Among Farmers**
3 **in Central Kenya Affect Their Knowledge of Sorghum (*Sorghum***
4 ***bicolor* [L.] Moench) Landrace Identification**

5 V. LABEYRIE^{*,1,2}, J. I. KAMAU³, C. DUBOIS^{4,5}, X. PERRIER^{4,5}, AND C. LECLERC^{2,3}

6 ¹CIRAD, UPR GREEN, F-34398, Montpellier, France

7 ²GREEN, University of Montpellier, CIRAD, Montpellier, France

8 ³KALRO-Genetic Resources Research Institute, P.O. Box 30148-0100, Nairobi, Kenya

9 ⁴CIRAD, UMR AGAP, F-34398, Montpellier, France

Q2 10 ⁵AGAP, University of Montpellier, CIRAD, INRA, Montpellier SupAgro, Montpellier, France

11 ^{*}Corresponding author; e-mail: vanesse.labeurie@cirad.fr

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13 **So Close Yet So Different: Cultural Differences among Farmers in Central Kenya Affect Their**
14 **Knowledge of Sorghum (*Sorghum bicolor* [L.] Moench) Landrace Identification.** Whether knowledge
15 of landrace identification is shared among farmers in rural societies is a matter of debate in crop diversity
16 research, and the influence of culture on knowledge heterogeneity remains largely misunderstood. This
17 study analyzes the heterogeneity of farmers' knowledge of crop landrace identification, and investigates
18 factors involved in its patterns. It especially explores the effect of cultural differences by comparing how three
19 ethnolinguistic groups identify and name sorghum diversity in the Mount Kenya region. A set of 293
20 panicles representing sorghum diversity in the study area was presented for identification to 96 farmers
21 randomly selected in the three groups. A subset of 287 panicles was scored for morphological characteristics
22 using 16 qualitative descriptors, and neutral genetic diversity of 170 of them was described using 18 SSR
23 genetic markers. Distance-based analyses were applied to analyze knowledge patterns within and between
24 groups and to describe the structure of sorghum morphological and genetic diversity. Results show that the
25 degree of heterogeneity of knowledge among farmers varies strongly according to both their ethnolinguistic
26 membership and panicle characteristics, despite their high geographic proximity. The effect of farmers'
27 experience of landraces and of pathways for social learning on inter-individual variations of knowledge is
28 discussed.

29 **Si proches et pourtant si différents: diversité culturelle et identification du sorgho (*Sorghum bicolor***
30 **[L.] Moench) dans la région du mont Kenya.** Un débat majeur dans le domaine de l'étude de la diversité
31 cultivée est de savoir si l'identification des variétés locales est consensuelle au sein des sociétés rurales.
32 L'influence des facteurs culturels sur l'hétérogénéité des connaissances reste en effet largement méconnue
33 dans ce domaine. Cet article analyse l'hétérogénéité des connaissances des agriculteurs concernant
34 l'identification des variétés de sorgho cultivées dans une localité de la région du Mont Kenya, et examine
35 l'effet des différences culturelles en comparant comment trois groupes ethnolinguistiques identifient les
36 différents morphotypes de sorgho. Un échantillon de 293 panicules représentant la diversité du sorgho dans
37 la zone d'étude a été présenté pour identification à 96 agricultrices sélectionnées au hasard dans les trois
38 groupes. Les caractéristiques morphologiques d'un sous-ensemble de 287 panicules ont été évaluées à l'aide
39 de 16 descripteurs qualitatifs, et la diversité génétique neutre de 170 d'entre elles a été décrite au moyen de

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40 18 marqueurs génétiques SSR. Des analyses basées sur des mesures de distance ont été utilisées pour analyser
41 l'hétérogénéité des connaissances au sein des groupes et entre eux, et pour décrire la structure de la diversité
42 morphologique et génétique du sorgho. Les résultats montrent que le degré d'hétérogénéité des
43 connaissances des agriculteurs varie fortement selon qu'ils appartiennent ou non au même groupe
44 ethnolinguistique et ce malgré leur grande proximité géographique, et qu'il varie aussi en fonction des
45 caractéristiques des panicules. Le rôle de l'expérience des agriculteurs concernant les différentes variétés et de
46 l'apprentissage social sont finalement discutés.

47 **Key Words:** Local ecological knowledge, identification, landrace, sorghum, Africa, ethnobotany.

48

50 Introduction

51 Humans have been in close relation to crops for
52 about 13,000 years, and a large share of the world
53 population continues to entertain this interdependent
54 relation in small-scale farming systems
55 (Diamond 2002; Gepts 2004). In these systems,
56 morphological diversity within crop species is remarkable,
57 as stressed early by Darwin in *Origin of the Species*
58 (1859). This diversity is generated and shaped by various
59 evolutionary mechanisms, which are strongly influenced
60 by human management practices (Hodgkin et al. 2007).
61 These practices especially play on selection (Boster 1985;
62 Louette and Smale 2000), as well as on genetic drift. They
63 also have an impact on both seed- and pollen-mediated
64 gene flows (Barnaud et al. 2007; McGuire 2008;
65 vom Brocke et al. 2003).

66 Landraces selected and identified using vernacular
67 names by farmers are assumed to be distinct management
68 units, organizing the continuum of phenotypic and genetic
69 diversity within crop species (Badstue et al. 2007; Bellon
70 and Brush 1994; Harlan et al. 1976). In small-scale farming
71 systems, knowledge of landrace diversity is of upmost
72 importance for the survival of societies as part of their
73 adaptive strategy. Humans have defined discrete categories
74 into the crop diversity continuum just like they did for the
75 rest of their environment, and have been using them as
76 frames for reasoning and management (Atran and Medin
77 2008). These folk taxonomies involve identification, naming,
78 and classification processes that are interrelated. According
79 to Friedberg (1991), identification is a perceptual process
80 through which farmers assign a plant to a class based on its
81 perceived characteristics. Naming is the process through
82 which these classes are labeled, mainly for communication
83 purposes, which suppose that individuals exchanging
84 information and planting material share a common
85 nomenclature. Last, classification refers to the multi-level
86 organization of classes.

91 A big debate exists concerning whether farmers' knowledge
92 of landrace identification is homogeneous in farming societies,
93 i.e., if farmers agree on the identification, naming, and
94 classification of crop diversity. Indeed, the impressive
95 number of landrace names inventoried on-farm in most
96 studies questioned whether a collective consensus exists
97 among farmers for landrace identification and naming
98 (Jarvis et al. 2008; Sadiki et al. 2007). Furthermore,
99 studies conducted on various species in different countries
100 reported a poor match between farmers' landrace
101 nomenclature and the structure of phenotypic and genetic
102 diversity (e.g., Barnaud et al. 2007; Quiros et al. 1990;
103 Salick et al. 1997). These studies highlighted the lack of
104 scientific knowledge concerning farmers' landrace taxonomy.
105 They raised debate concerning the existence of a common
106 reference frame for landrace identification and naming
107 shared by farmers on the local scale. They also raised
108 questions concerning factors involved in inter-individual
109 variations of taxonomy on the local scale.

110 This issue was first tackled by Boster in his work
111 on cassava landrace identification among Aguaruna
112 farmers (Boster 1986), based on the cultural consensus
113 theory (Romney et al. 1986). Assuming that landraces
114 are categories that constitute "a frame for storing and
115 conveying experience and information" (Bulmer 1974
116 in Friedberg 1991), he hypothesized that farmers should
117 exchange information and share a common experience
118 of the landraces in order to agree on their identification
119 and naming based on their phenotypic characteristics.
120 He showed that differences of knowledge of landrace
121 identification among farmers were associated with cultural
122 differences, knowledge heterogeneity being higher
123 between kinship groups than within. This indicates that
124 the level of knowledge homogeneity among farmers for
125 landraces' identification is proportional to their collective
126 experience of crop characteristics and to the intensity of
127 information exchange among them, and thus to

- 133 the strength of their social ties and cultural
134 proximity.
- 135 Since then, cultural consensus theory has been
136 applied to a variety of local ecological knowledge
137 domains, showing that the ways people think of and
138 classify the natural world differ across cultures.
139 Atran and Medin (2008) notably showed that dif-
140 ferent cultural groups living in the same agro-
141 ecological zones present noteworthy differences of
142 knowledge concerning their environment. These
143 studies revealed that culture influences the way
144 human societies classify the continuum of biological
145 variability surrounding them, but this issue was
146 rarely investigated with crop landraces. Most work
147 on landrace identification by farmers was conducted
148 in the frame of crop diversity studies (Nuijten and
149 Almekinders 2008; Sadiki et al. 2007), and inter-
150 individual variations of knowledge of landrace iden-
151 tification were not documented. Indeed, most stud-
152 ies document the landrace name given by one farm-
153 er to identify a plant (Soler et al. 2013), or rely on a
154 focus group to document what is the consensual
155 name for it within the community (Mucioki et al.
156 2014). Our understanding of the identification,
157 naming, and classification of crop intraspecific di-
158 versity by rural societies hence remains limited de-
159 spite its importance regarding crop genetic resources
160 conservation and property rights issues (Lapeña and
161 Halewood 2016).
- 162 The way farmers identify and classify landraces is
163 expected to influence their management of crop
164 infra-specific diversity, and differences of
165 management should therefore exist between
166 cultural groups. The work of Perales et al. (2005)
167 brought a key contribution to this issue, indicating
168 that two geographically close ethnolinguistic groups
169 in Mexico have divergent seed selection practices as
170 they maintain morphologically distinct maize pop-
171 ulations despite gene flows between them. Their
172 results suggest that differences in knowledge of crop
173 landrace management exist between ethnolinguistic
174 groups despite their geographical proximity. Our
175 study aims at testing this hypothesis by character-
176 izing differences in knowledge concerning sorghum
177 (*Sorghum bicolor* [L.] Moench–Poaceae) landrace
178 identification between three ethnolinguistic groups
179 in a locality of the Mount Kenya region. We tested
180 three main hypotheses following Bosters' work on
181 farmers' knowledge concerning landrace identifica-
182 tion. First, farmers belonging to the same
183 ethnolinguistic group display more similar
184 knowledge than those belonging to different groups
185 because they exchange more information. Second,
186 ethnolinguistic groups display different levels of
187 knowledge heterogeneity for the identification of
188 the different landraces because they differ in their
189 experience of their characteristics. Third,
190 ethnolinguistic groups differ in the landrace names
191 they associate with the different morphotypes be-
192 cause information exchange is limited between
193 them.
- 194 This work builds on a previous study conducted
195 in the same area, in which samples from 14 sor-
196 ghum landraces named by farmers in the three
197 ethnolinguistic groups were collected (Labeyrie
198 et al. 2014). Genetic analysis showed that the plants
199 collected belong to four main genetic clusters cor-
200 responding to differences in phenology, as well as in
201 the origins and history of the landraces. These four
202 clusters respectively correspond to (i) short-cycle
203 local landraces, (ii) long-cycle landraces, (iii) an
204 introduced variety released by research several de-
205 cades ago (*Kaguru*), and (iv) another introduced
206 variety that has just diffused in the area (*Gadam*).
207Q3 The first two clusters included several morphotypes
208 associated with different landrace names, the occur-
209 rence frequency of which differed significantly
210 among ethnolinguistic groups. Based on these re-
211 sults, our article addresses two main questions: do
212 the three ethnolinguistic groups differ in their
213 knowledge concerning sorghum landrace identifica-
214 tion, and does their respective knowledge vary ac-
215 cording to sorghum genetic and morphological
216 characteristics?
- ## 217 Material and Methods
- ### 218 STUDY SITE
- 219 The study site was located in the Eastern Prov-
220 ince of Kenya at the boundary between Tharaka-
221 Nithi and Embu counties (0° 24' S, 37° 46' E). We
222 focused on a contact zone between Chuka, Tharaka,
223 and Mbeere ethnolinguistic groups (Fig. 1). The
224 study site presents uniform agro-ecological condi-
225 tions, at an altitude of about 900 meters (m) above
226 sea level, and with mean monthly temperature rang-
227 ing between 21.7 and 23.9 degrees Celsius (°C)
228 (Jaetzold et al. 2007). The mean rainfall is about
229 700–800 millimeters (mm) per year, distributed
230 across two rainy seasons with the long rains

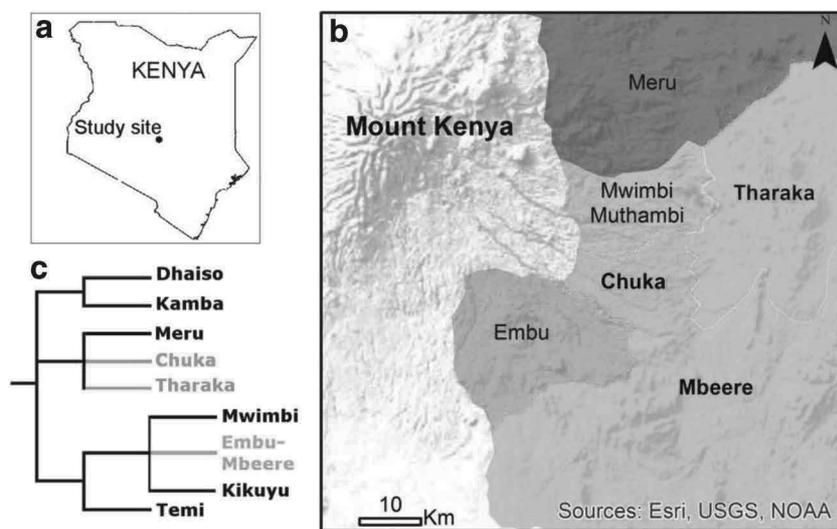


Fig. 1. Study site location. **a** Location of the study site in Kenya. **b** Geographic distribution of ethnolinguistic groups in the Eastern Mount Kenya region. **c** Linguistic classification of Central Kenya languages according to Hammarström et al. (2015).

231 occurring from March to May and the short rains
232 from October to December.

233 According to oral history, people started to mi-
234 grate to the study area by the end of the nineteenth
235 century. The Chuka would have been the first to
236 settle in this area about one century ago, while the
237 Tharaka and the Mbeere probably settled more
238 recently, but information is lacking concerning their
239 history. The Chuka, Tharaka, and Mbeere groups
240 present cultural and linguistic differences
241 (Hammarström et al. 2015). They speak distinct
242 languages of the central Bantu cluster, which are,
243 however, largely inter-intelligible. Members of each
244 group believe in their common ascendance on
245 which is based their distinct ethnolinguistic identity
246 (Heine and Möhlig 1980; Middleton 1953).
247 Tharaka and Chuka were allied in the past and
248 consider themselves to be blood brothers, or
249 *gishiaro*, in Kimeru language (Fadiman 1993), and
250 they have limited relationship with the Mbeere
251 (Glazier 1970; Mwaniki 1973). Intermarriage is
252 frequent between the Tharaka and the Chuka,
253 while it is very uncommon between both groups
254 and the Mbeere (Labeyrie et al. 2016a). This rela-
255 tionship system is reflected by the geographical
256 organization of the three groups, the Tharaka and
257 the Chuka being spatially mixed and settled in the

northern part of the study site, whereas the Mbeere
258 are located separately from the two other groups in
259 the southern part of the area. The maintenance of
260 this geographical partition among ethnolinguistic
261 groups results mainly from the combination of
262 ethnolinguistic endogamy and patrilocal residence,
263 implying that most married men settle near the
264 compound of their father (Middleton 1953, pers.
265 obs.).
266

ETHICS STATEMENT

267
268 This work was conducted in collaboration with
269 the KALRO National Genebank of Kenya that has
270 the national mandate for the collection of plant
271 genetic resources and the documentation of accom-
272 panying information. Institutional and administra-
273 tive procedures were carefully followed prior to
274 undertaking the study, and dedicated committees
275 in KALRO granted approval for our research activ-
276 ities. We followed recommendations of the ISE
277 Code of Ethics, and the involvement of team mem-
278 ber natives from the study region contributed to
279 ensure that local procedures, rules, and customs
280 were respected, and that authorizations were
281 granted from legitimate authorities. First, govern-
282 ment administrative and local community

283 representatives were informed and kept updated of
 284 the activities, and their consent was sought before
 285 conducting the research. Then, the study objective
 286 and the future utilization of data were explained to
 287 farmers and their prior informed consent was ob-
 288 tained verbally before undertaking interviews and
 289 crop collection. Activities were not conducted
 290 where such consent was not granted.

291 DATA COLLECTION

292 Sorghum panicles were sampled on-farm in Jan-
 293 uary and July 2011 in the three ethnolinguistic
 294 groups following the strategy described previously
 295 (Labeyrie et al. 2014), which aimed at representing
 296 the diversity of the sorghum landraces named by
 297 farmers. Seeds from the collected panicles were
 298 sown in October 2011 in an experimental field
 299 under controlled and uniform growing conditions.
 300 A total of 293 descendants were sampled to maxi-
 301 mize the range of morphological variability and
 302 presented to a panel of farmers for identification.
 303 Out of this set, 287 panicles were scored for mor-
 304 phological descriptors as six panicles were too de-
 305 graded at the end of the survey to be scored. One-
 306 hundred seventy plants in this subset were scored
 307 for neutral genetic microsatellite SSR markers, se-
 308 lected to represent the diversity of sorghum land-
 309 races named by farmers in the three ethnolinguistic
 310 groups (Labeyrie et al. 2014).

311 *Landrace Identification Experiment*

312 The set of 293 panicles harvested in the experi-
 313 mental field was presented to a panel of informants
 314 from the three ethnolinguistic groups. Thirty-two
 315 female informants were randomly chosen in each
 316 group, and their ethnolinguistic group was record-
 317 ed. Only women were interviewed because they are
 318 in charge of sorghum seed selection, sowing, har-
 319 vesting, and trading according to the local gendered
 320 division of labor (pers. obs.). Following the proce-
 321 dure used by Boster (1986), each informant was
 322 independently asked to identify each of the 293
 323 panicles that were successively presented to her. A
 324 field assistant recorded the name used by each in-
 325 formant to identify each panicle. Spelling standard-
 326 ization was later done to ensure that differences were
 327 not due to variation in pronunciation among
 328 informants.

Morphological and Genetic Characterization 329

330 Out of the 293 panicles harvested, 287 were
 331 scored for 16 qualitative morphological descrip-
 332 tors at the Kenya Agricultural and Livestock
 333 Research Organization, Genetic Resources Re-
 334 search Institute, Muguga. The study was limit-
 335 ed to the characteristics of the panicles because
 336 the selection of seeds by farmers is done on the
 337 panicle only, at home before threshing, and
 338 thus without considering the characteristics of
 339 the whole plant. Only qualitative descriptors
 340 were scored because they are the main criteria
 341 on which farmers base their perceptual distinc-
 342 tiveness (Gibson 2009). The 16 descriptors were
 343 selected for their polymorphism in the sorghum
 344 population studied and their ease of scoring. They
 345 included the main criteria that farmers reported
 346 using for identifying their sorghum landraces, ac-
 347 cording to information collected during semi-
 348 directive interviews.

349 Traits scored (Electronic Supplementary Ma-
 350 terial [ESM] Appendix 1) concerned the whole
 351 panicle shape, seed characteristics (color, lateral
 352 shape, shattering, endosperm texture, sub-coat pre-
 353 sence, and pericarp thickness) and glume character-
 354 istics (color, opening, adherence, covering, awning,
 355 hairiness, texture, presence of a transversal wrinkle,
 356 and pedicelate spikelet). Parts of these descriptors
 357 were selected among those recommended by the
 358 IPGRI (1993), and more precise descriptors of seed
 359 and spikelets were added among some of those used
 360 by Snowden (1936). Procedures of double charac-
 361 terization of a set of panicles randomly selected
 362 made it possible to ensure the consistency of oper-
 363 ators in scoring morphological traits, and double
 364 data entry was performed to limit typing errors.

365 DNA extraction, amplification, migration, and
 366 alleles' size scoring for 18 microsatellite SSR loci
 367 were done on 170 plants out of the 287 for which
 368 panicles were morphologically scored. The study of
 369 the genetic diversity of the sorghum population,
 370 including these 170 individuals, was conducted in
 371 a previous study, and full methodological details are
 372 provided in Labeyrie et al. (2014).

STATISTICAL ANALYSES 373

374 Statistical analyses were conducted to describe,
 375 on the one hand, the patterns of sorghum genetic

376	and morphological diversity and, on the other hand,	named inconsistently, i.e., no consensus exists	424
377	the patterns of knowledge heterogeneity among	among farmers for its identification.	425
378	informants.		
379	<i>Analysis of Sorghum Genetic and Morphological</i>	<i>Distance-Based Analysis of Knowledge on Panicle</i>	426
380	<i>Diversity</i>	<i>Identification</i>	427
381	First, the structure of sorghum genetic diversity	In this paper, differences of knowledge between	428
382	was described. A discriminant analysis of principal	ethnolinguistic groups were measured by compar-	429
383	components (DAPC, Jombart et al. 2010) was used	ing, on the one hand, their level of knowledge	430
384	to identify and then describe clusters from the ge-	heterogeneity, i.e., inter-individual variations in	431
385	netic diversity of the 170 sorghum panicles collect-	the identification of panicles, and on the other	432
386	ed. The <i>K</i> -means method was performed prior to	hand, their landrace identification, i.e., the name	433
387	running the discriminant analysis using the algo-	they associated to each panicle. First, we analyzed	434
388	rithm included in DAPC function, and the optimal	patterns of informants' knowledge for the whole	435
389	number of clusters to describe the diversity was	panicle set ($n = 287$), and then separately for subsets	436
390	determined based on the Bayesian information cri-	of panicles corresponding to each of the genetic	437
391	terion (BIC) curve. Analyses were performed using	clusters.	438
392	the R package <i>adeigenet</i> , version 2.0.1 (R. Core	The heterogeneity of knowledge within the	439
393	Team 2016; Jombart 2008).	group for the identification of the panicle set was	440
394	Second, the morphological diversity of the 287	measured as the dispersion of informants' answers.	441
395	panicles was described by performing a principal	This heterogeneity degree was compared between	442
396	coordinates analysis (PCoA) on the morphological	the Chuka, Tharaka, and Mbeere groups. For this	443
397	dissimilarity matrix. Dissimilarity between panicle	purpose, a similarity index was computed for each	444
398	pairs was computed with the simple matching in-	pair of informants as the proportion of panicles they	445
399	dex, i.e., dividing the number of traits for which	named identically, and a distance index was then	446
400	both panicles share the same modalities by the total	constructed by subtracting the similarity index to	447
401	number of traits. Correspondence between the	one. Average within-group dispersion, i.e., the av-	448
402	structure of sorghum morphological and genetic	erage distance of individuals to group centroid in	449
403	diversity was assessed by displaying genetic clusters	the space of the simple matching distance index, was	450
404	in colors on the PCoA scatterplot. Analyses were	used as a measure of knowledge heterogeneity with-	451
405	performed using the R package <i>ade4</i> version 1.7-6	in groups. We conducted analyses to assess farmers'	452
406	(Dray and Dufour 2007).	knowledge heterogeneity for the identification of	453
407	<i>Measurement of Informants' Consistency in Naming</i>	the whole panicle set on one hand, and of each set	454
408	<i>Panicles</i>	corresponding to each of the four genetic groups on	455
409	We computed the number of informants in	the other hand. Knowledge heterogeneity degree	456
410	each ethnolinguistic group who cited the same	was compared among ethnolinguistic groups by	457
411	name to identify each sorghum panicle. A	testing if the average within-group dispersion was	458
412	cross-table was built by a group, crossing the	equivalent among groups through running an anal-	459
413	list of the 293 panicles in one way and the list	ysis of multivariate homogeneity of groups' disper-	460
414	of the names given for identification in the	sions (PERMDISP2), which was performed on the	461
415	second way, to calculate the number of infor-	distance matrix (Anderson 2006). Further, pairwise	462
416	nants at each combination of panicle x name.	Tukey's honest significant difference (HSD) tests	463
417	We considered that a panicle was named con-	were performed to test for the significance of	464
418	sistently in a given group, i.e., knowledge het-	pairwise differences in the average dispersion be-	465
419	erogeneity was low within it, when more than	tween ethnolinguistic groups. PERMDISP2 and	466
420	half of informants (60%) used the same name	Tukey's HSD analyses were run globally on the	467
421	to identify this panicle. On the contrary, when	whole panicle set.	468
422	less than 60% of informants used the same	Then, we tested if significant differences in land-	469
423	name, we considered that the panicle was	race identification exist between ethnolinguistic	470
		groups, i.e., if they used different names to identify	471
		panicles in each set. This was done by testing if the	472
		centroids were equivalent for all groups, using a	473

474 non-parametric multivariate analysis of variance
 475 (ADONIS, Anderson 2001). This test is a multi-
 476 variate equivalent of ANOVA, based on the com-
 477 parison of within- and between-group average dis-
 478 persion. ADONIS results can be confidently
 479 interpreted only if within-group dispersion is equiv-
 480 alent among ethnolinguistic groups, which is tested
 481 by PERMDISP2. A principal coordinates analysis
 482 (PCoA) was conducted on the knowledge distance
 483 matrix to visualize knowledge patterns within and
 484 between groups. All distance-based analyses were
 485 performed using the R package *vegan* version 2.4-
 486 0 (Oksanen et al. 2012).

487 Both the analysis of multivariate homogeneity of
 488 groups' dispersions (PERMDISP2) and the non-
 489 parametric multivariate analysis of variance
 490 (ADONIS) were run separately for each genetic
 491 cluster after performing them on the whole panicle
 492 set ($n = 287$). The correspondence between the
 493 structure of sorghum morphological diversity and
 494 knowledge patterns was assessed by displaying the
 495 "consistent" names (i.e., used by more than 60% of
 496 informants) using different colors on the PCoA
 497 scatterplot of morphological distances for each
 498 ethnolinguistic group.

499 Results

500 KNOWLEDGE HETEROGENEITY WITHIN AND 501 BETWEEN ETHNOLINGUISTIC GROUPS

502 A large number of landrace names were cited
 503 during the identification experiment (Chuka, 30
 504 names; Tharaka, 36; Mbeere, 39), but only a few
 505 names were used consistently at least one time, i.e.,
 506 by more than 60% of the informants, to identify the
 507 same panicle (Chuka, 7 names; Tharaka, 6; Mbeere,
 508 5). Most landrace names used consistently were
 509 common to the three ethnolinguistic groups, but
 510 some were peculiar to one or two groups. In most
 511 cases, each consistent name was used to identify
 512 several panicles, but some were used to identify only
 513 one or two panicles.

514 Significant differences of knowledge heterogene-
 515 ity level were observed between ethnolinguistic
 516 groups for the identification of the whole panicle
 517 set ($n = 293$). Indeed, within-group dispersion (dis-
 518 tance to a group's centroid) differed significantly
 519 between groups, as indicated by PERMDISP2 re-
 520 sults ($F = 8.55$, p value < 0.001). A Tukey HSD test
 521 further showed that the mean within-group disper-
 522 sion was significantly lower in the Chuka group

(0.27) than in the Tharaka (0.39) and Mbeere 523
 (0.38) groups, indicating a higher knowledge 524
 homogeneity in the former group. Differences of 525
 knowledge between the Mbeere and the two other 526
 groups are displayed along the second PCoA axis 527
 (Fig. 2), but the significance of these differences 528
 cannot be assessed based on ADONIS results be- 529
 cause of differences in within-group dispersion. 530
 Furthermore, two different knowledge subgroups 531
 are distinguished within the Mbeere group along 532
 the first PCoA axis, and part of the Tharaka appears 533
 close to the Chuka, while the rest presents differ- 534
 ences in knowledge. 535

536 Strong differences in consistency were observed 536
 among panicles. The proportion of informants who 537
 used the same landrace name varied among pani- 538
 cles, ranging from a minimum of 19% in the Chuka 539
 and Mbeere groups and 25% in the Tharaka group, 540
 to a maximum of 100% in the three groups (ESM 541
 Fig. 1). Overall, the proportion of panicles identi- 542
 fied consistently varied strongly among 543
 ethnolinguistic groups, with 80% in the Chuka 544
 group, 40% in Tharaka, and 32% in Mbeere ($n =$ 545
 293 panicles). 546

547 Farmers' knowledge was homogeneous within 547
 and between ethnolinguistic groups for some pani- 548
 cles, which were identified highly consistently in all 549
 groups. This was especially noticed for a set of 550
 panicles named *Kaguru* by a large majority of infor- 551
 mants (Fig. 3). In other cases, the level of knowledge 552
 heterogeneity differed between groups as some pani- 553
 cles were identified highly consistently in one 554
 group and not in the others. This was especially 555
 striking for a set of panicles named *Gadam* by most 556
 Chuka informants, but identified inconsistently in 557
 the two other groups. A similar situation was ob- 558
 served for panicles consistently named *Murugue* and 559
Mugeta by the Chuka and Tharaka, and for panicles 560
 consistently named *Ngirigacha* by the Mbeere, 561
 which were identified inconsistently in the other 562
 groups. Last, a high degree of knowledge heteroge- 563
 neity was observed both within and between 564
 ethnolinguistic groups for some panicles that were 565
 named inconsistently in all groups. 566

567 DIFFERENCES IN KNOWLEDGE ACCORDING TO 568 PANICLES' CHARACTERISTICS

569 An adequate number of classes to describe the 569
 genetic diversity in our dataset was $K = 4$ according 570
 to the BIC criterion in K -means algorithm. Genetic 571
 clusters matched partially with the structure of mor- 572
 phological diversity, some genetic clusters being 573

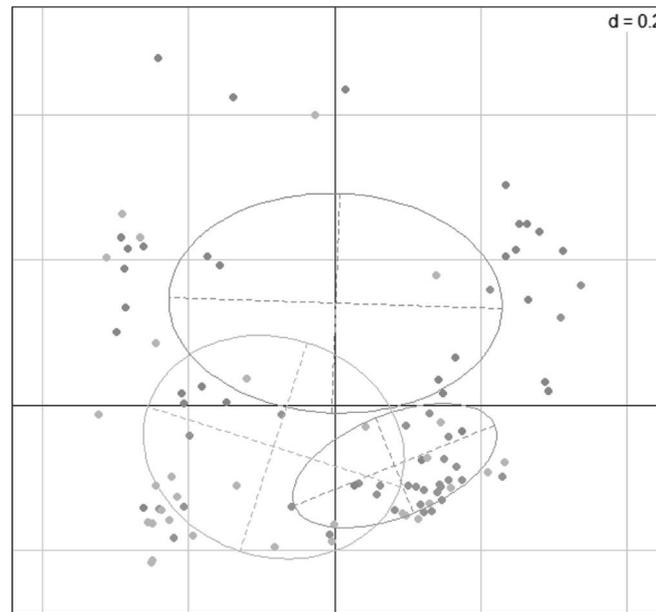


Fig. 2. Knowledge similarity among individuals according to their ethnolinguistic membership. Plot of the two first axis of the PCoA based on knowledge distance matrix between informants ($n = 96$ informants; the first component expresses 23% of the total variation, and the second one expresses 17%). Ethnolinguistic groups are displayed in colors (red, Chuka; green, Tharaka; blue, Mbeere).

574 morphologically distinct while others overlapped
 575 (Fig. 4). G2 was highly homogeneous and distinct
 576 morphologically, while G1, G3, and G4 were hetero-
 577 geneous and overlapped. Overlap was especially
 578 high between G3 and G4.

579 Knowledge heterogeneity was similar among
 580 ethnolinguistic groups for G1, G3, and G4
 581 clusters, as no significant differences of answer
 582 dispersion were detected by PERMDISP2
 583 ($1.41 < F < 1.93$, p value > 0.1). Differences of
 584 knowledge heterogeneity between ethnolinguistic
 585 groups were significant only for the genetic cluster
 586 G2 ($F = 11.04$, p value < 0.001 ; ESM Appendix 2).
 587 Indeed, the large majority of Chuka informants
 588 agree on the identification of panicles assigned to
 589 the G2 genetic group that they named *Gadam*,
 590 while Mbeere and Tharaka informants presented a
 591 high level of knowledge heterogeneity.

592 Differences of knowledge on panicle identifica-
 593 tion were observed between ethnolinguistic
 594 groups for some genetic clusters (G3 and G4)
 595 while not for others (G1). For G1, no signifi-
 596 cant differences of identification were observed
 597 (ADONIS: $F = 1.71$, p value = 0.044; ESM Fig. 2)
 598 as knowledge homogeneity was very high both
 599 within and between groups. Indeed, this cluster

600 was mainly composed of panicles consistently
 601 named *Kaguru* by informants in all groups (Fig. 5).
 602 An ADONIS test on centroid difference between
 603 groups was significant for three clusters (G2 to G4).
 604 Significant differences of groups' centroid for G3
 605 and G4 traduce differences of identification among
 606 groups because within-group dispersion was similar
 607 for these clusters. However, such interpretation
 608 cannot be applied to G2 because significant differ-
 609 ences of within-group dispersion were detected by
 610 PERMDISP2.

611 Ethnolinguistic groups differed significantly in
 612 their identification of panicles in G3 (ADONIS:
 613 $n = 97$, $F = 7.89$, p value < 0.001). This genetic
 614 cluster comprised three morphological sub-clusters
 615 corresponding to different landrace names that were
 616 identified with very different consistency levels by
 617 the different groups (Fig. 6). A first sub-cluster was
 618 composed of a large number of panicles consistently
 619 named *Muruge* by the Chuka, while the Tharaka
 620 and Mbeere identified consistently only a small part
 621 of them. A second sub-cluster included panicles
 622 consistently named *Mugeta* by the Tharaka and to
 623 a lesser extent by the Chuka, while the Mbeere were
 624 inconsistent in their identification. Last, a third sub-
 625 cluster included a few panicles consistently named

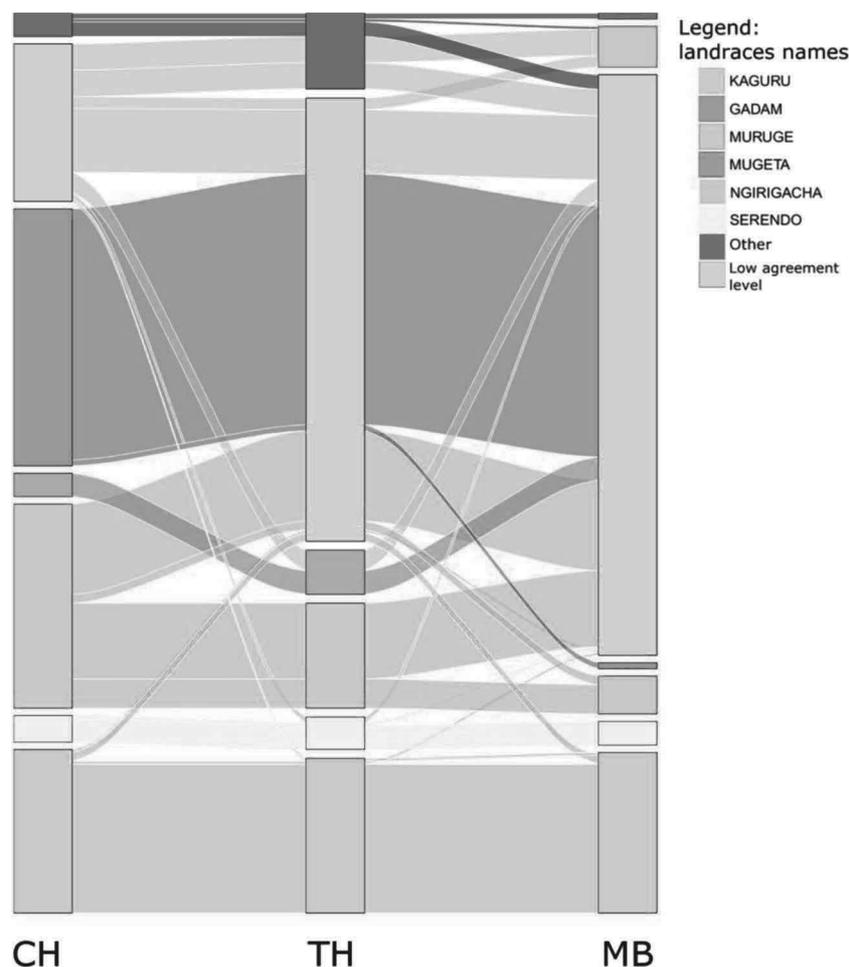


Fig. 3. Comparison of panicle identification between ethnolinguistic groups. Boxes represent the proportion of panicles (y -axis) identified consistently (in colors) or not (in gray) in each ethnolinguistic group (x -axis). Flows among boxes in the different ethnolinguistic groups represent the share of panicles that were identified similarly or differently between groups. For instance, the width of the flow between box A in group X and box B in group Y represents the proportion of panicles identified as landrace A in group X that was identified as landrace B in group Y.

626 *Ngirigacha* by the Mbeere only. Ethnolinguistic
 627 groups also differed significantly in their identifica-
 628 tion of panicles in G4 (ADONIS: $n = 17$, $F = 3.62$,
 629 p value < 0.001). This cluster mainly included pan-
 630 icles identified consistently as *Muruge*, and, in a
 631 lesser extent, *Serendo* and *Kaguru* in proportions
 632 differing among groups. In addition, G3 and G4
 633 also included a large share of panicles identified
 634 inconsistently in proportions that varied strongly
 635 between groups. These panicles were displayed in
 636 between the morphological groups corresponding
 637 to panicles identified consistently, for which

knowledge homogeneity was high, on the PCoA
 of morphological distance.

G3 and G4 genetic clusters displayed a large
 morphological heterogeneity and overlapped. Part
 of the plants in these clusters hence presented mor-
 phological similarities despite their genetic differ-
 ences. Interestingly, some panicles presenting genetic
 differences were named similarly because of their
 morphological similarity. In particular, panicles
 named *Muruge* in all ethnolinguistic groups present-
 ed morphological similarity although they belong to
 both G3 and G4 genetic clusters. Similarly, some

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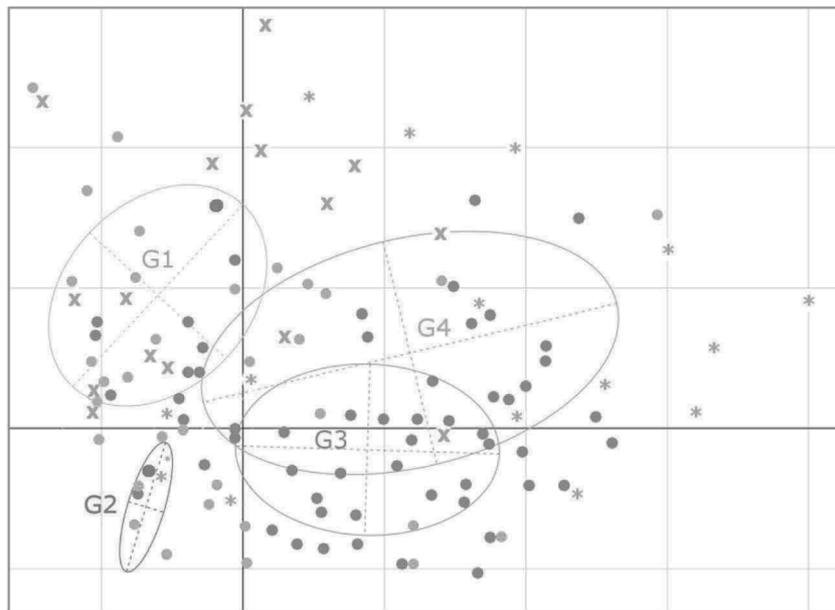


Fig. 4. Match between genetic and morphological diversity. Plot of the first two axes of the PCoA based on panicles' morphological traits ($n = 287$ panicles, 16 traits; the first component expresses 26% of the total variation, and the second one expresses 14%). Genetic clusters are displayed in colors ($n = 170$, panicles morphotyped but not genotyped are displayed in gray).

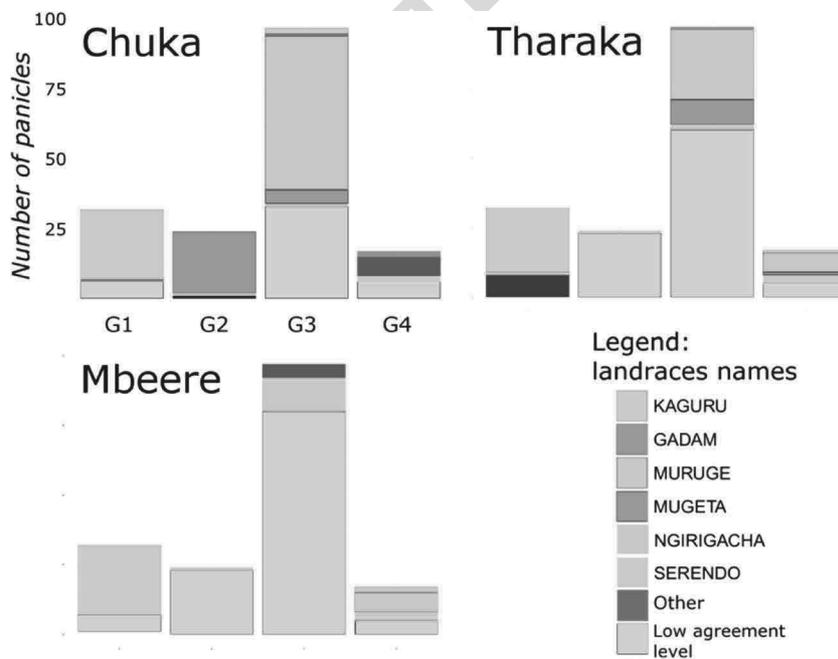


Fig. 5. Match between genetic clusters and landrace identification. Bars represent the number of panicles identified consistently (> 60% of informants, in colors) and inconsistently (< 60%, in black) in each genetic cluster.

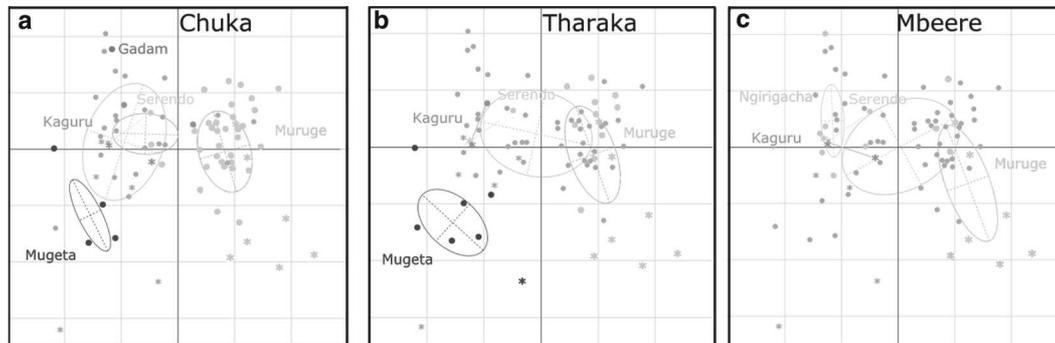


Fig. 6. Correspondence between morphological diversity and landrace identification for panicles assigned to G3 (dots) and G4 (stars) genetic clusters ($n = 114$ panicles, 16 traits). Plot of the two first axes of the PCoA based on panicle morphological traits (variability expressed: 1st Co = 23%, 2nd Co = 15%). Colors correspond to landrace names identified consistently in each ethnolinguistic group, and panicles identified inconsistently are displayed in gray.

650 panicles assigned to G4 but morphologically similar
651 to G1 were named *Kaguru*.

652 **Discussion**

653 In this paper, we analyzed farmers' knowledge of
654 sorghum landrace identification in an
655 ethnolinguistic contact zone. Our results show that
656 knowledge varies according to both individuals'
657 cultural identity and panicle characteristics. We first
658 assessed whether farmers within ethnolinguistic
659 groups shared knowledge concerning landrace identifi-
660 cation by measuring within-group knowledge
661 heterogeneity, and then tested whether the names
662 used to identify panicles differed between groups.
663 Results showed that groups differed significantly in
664 their level of knowledge heterogeneity, the Chuka
665 displaying the highest homogeneity for the identifi-
666 cation of the whole panicle set. Furthermore,
667 within-group knowledge heterogeneity varied
668 strongly among panicles on one hand, and differed
669 between groups for some panicles on the other
670 hand. This indicates that knowledge heterogeneity
671 within and between ethnolinguistic groups is related
672 to panicle characteristics.

673 We further conducted analyses to test for the
674 effect of panicle genetic and morphological charac-
675 teristics on knowledge heterogeneity within and
676 between ethnolinguistic groups. Results first
677 showed that within-group knowledge heterogeneity
678 differed between groups for only one genetic cluster
679 (G2), indicating that Chuka shared common
680 knowledge on the identification of panicles in this
681 cluster, while Tharaka and Mbeere do not. Second-
682 ly, we found that knowledge was highly similar and

homogeneous among ethnolinguistic groups for the
683 identification of one cluster (G1), while significant
684 differences were observed among them for the identifi-
685 cation of two clusters (G3 and G4). Further-
686 more, ethnolinguistic groups differ in their level of
687 knowledge heterogeneity for the identification of
688 the different morphotypes within G3 and G4.
689

The knowledge patterns we described reveal in-
690 sights on farmers' experience of the different land-
691 races, and on the diffusion of knowledge within and
692 between ethnolinguistic groups. Indeed, according
693 to Boster (1986), landrace identification, naming,
694 and classification regarding their morphological
695 characteristics are socially learned and further con-
696 structed by individuals through their direct experi-
697 ence with the plants. He identified three major
698 processes involved in inter-individual differences of
699 knowledge concerning landrace identification: (i)
700 differences among individuals in their learning
701 sources and pathways; (ii) differences in their expe-
702 rience of the landrace and its morphological charac-
703 teristics; and (iii) differences in the time or the
704 willingness individuals have for acquiring experi-
705 ence in this domain. As there is little support for
706 significant differences of inter-individual variations
707 of time and willingness to learn between
708 ethnolinguistic groups, the knowledge patterns we
709 observed reflect, on the one hand, modalities of
710 knowledge transmission within and between groups
711 and, on the other hand, differences between groups
712 in their level of experience concerning landrace
713 identification. This leads us to discuss knowledge
714 patterns regarding the history and characteristics of
715 panicles and knowledge transmission modalities
716 within and between groups (Cavalli-Sforza and
717

718	Feldman 1981; Leclerc and Coppens	Mbeere groups indicates that they are not yet famil-	769
719	d'Eeckenbrugge 2012; Reyes-García et al. 2009).	iar with its characteristics and name.	770
720	PANICLE CHARACTERISTICS AND HISTORY IN	Clusters G3 and G4 included panicles identified	771
721	RELATION TO ETHNOLINGUISTIC GROUPS	consistently with various landrace names as well as	772
722	Panicles presented to informants display different	panicles identified inconsistently, in proportions	773
723	morphological characteristics, and have different	that vary strongly between ethnolinguistic groups.	774
724	histories and origins. These differences help explain	This variety of landrace names was associated with	775
725	the strong variations of within-group knowledge	differences in panicles' morphological characteristics	776
726	heterogeneity between panicles and its variations	within these clusters, the different landrace names	777
727	between ethnolinguistic groups for the same panicle.	corresponding to different morphotypes. Knowl-	778
728	Indeed, farmers share knowledge on panicles	edge heterogeneity was similar among groups for	779
729	presenting morphological characteristics with which	these genetic clusters, but it varied between	780
730	they are collectively familiar. Several characteristics	morphotypes within genetic clusters, especially	781
731	of landraces can contribute to building collective	G3. Ethnic groups appeared to be respectively fam-	782
732	knowledge, such as the time since they are cultivat-	iliar with the identification of different	783
733	ed, their popularity, and whether they can be easily	morphotypes, some being identified highly consis-	784
734	identified and distinguished based on their morpho-	tently in one ethnolinguistic group while not in	785
735	logical characteristics.	others and conversely. The Chuka in particular	786
736	A previous genetic study, which included the 170	consistently identified a set of morphologically sim-	787
737	individuals we analyzed here, showed that sorghum	ilar panicles as <i>Muruge</i> , while knowledge within the	788
738	landraces in our study area present different histo-	two other groups was more heterogeneous. A sim-	789
739	ries and agronomic characteristics (Labeyrie et al.	ilar situation was observed for a set of panicles	790
740	2014). It identified four genetic clusters matching	identified as <i>Mugeta</i> by most Tharaka informants.	791
741	with those we identified, showing that clusters G1	This probably results from differences in experience	792
742	and G2 were introduced varieties released by the	that ethnolinguistic groups developed over time	793
743	formal breeding system, whereas G3 and G4 were	concerning these different landraces. This hypothe-	794
744	local landraces with different agro-morphological	sis is supported by previous results showing that	795
745	characteristics.	<i>Muruge</i> and <i>Mugeta</i> landraces were respectively in-	796
746	Knowledge was highly homogeneous within	troduced by the Chuka and the Tharaka in the	797
747	and between groups for the identification of	study area, which explains their respective higher	798
748	G1 panicles that were named <i>Kaguru</i> by most	knowledge uniformity for the identification of cor-	799
749	informants. By contrast, differences were observed	responding morphotypes (Labeyrie et al. 2014,	800
750	among groups for G2 as knowledge was highly	2016b).	801
751	homogeneous in the Chuka group where it was	Last, our study showed that knowledge hetero-	802
752	named <i>Gadam</i> , while it was heterogeneous in the	geneity was very high in all ethnolinguistic groups	803
753	Mbeere and Tharaka groups. Differences in the	for some panicles. Our hypothesis is that these	804
754	dates of dissemination of these two improved var-	panicles present a combination of traits that do	805
755	ieties likely explain why G1 was identified consis-	not correspond to the traits on which the landrace	806
756	tently in all groups whereas G2 was only in the	classification system used by most informants is	807
757	Chuka group. Indeed, <i>Kaguru</i> was introduced sev-	based. Either these panicles may result from crosses	808
758	eral decades ago and has been widely cultivated and	and combine morphological characteristics of sever-	809
759	sold in markets in the area, while <i>Gadam</i> was re-	al landraces, which is confusing for informants, or	810
760	leased in the area only two years before our study.	they may belong to recently introduced or rare	811
761	Our results further suggest that G2 introduction	landraces, whose characteristics are not yet familiar	812
762	started in the Chuka group, who display a strong	to most informants.	813
763	collective experience of its identification. Such high	The existence of a collective coherence among	814
764	knowledge uniformity may result from massive dis-	farmers for landrace identification has been debated	815
765	semination of G2 under the well-defined " <i>Gadam</i> "	in several studies (Sadiki et al. 2007). Some studies	816
766	name by Kenyan agricultural extension services in	reported high knowledge heterogeneity among	817
767	the Chuka group. By contrast, knowledge hetero-	farmers, such as Salick et al. (1997), on cassava	818
768	geneity for G2 identification in the Tharaka and	nomenclature among Amuesha in the Peruvian	819
		Amazon. Others found higher knowledge homoge-	820
		neity within villages than between villages, and	821

822 postulated that geographic distance that limits seed
823 circulation also limits knowledge transmission
824 (Nuijten and Almekinders 2008). However, knowl-
825 edge of different farmers for the identification of the
826 same plant was not measured since Bosters' work
827 was based on the cultural consensus framework
828 (1986). Our study shows that farmers' consistency
829 in identifying crops varies strongly depending on
830 plants' characteristics and history, but varies also
831 according to the cultural background of farmers.
832 Hence, farmers living in the same location can
833 present strong differences of knowledge because of
834 their cultural differences. Our results are thus in line
835 with those of Perales et al. (2005) on maize in
836 Mexico, and contribute to explaining the differ-
837 ences in selection practices they observed between
838 adjacent ethnolinguistic groups.

839 KNOWLEDGE TRANSMISSION MODALITIES

840 Knowledge heterogeneity within and between
841 ethnolinguistic groups indicates that communica-
842 tion and landrace circulation is limited between
843 them. Indeed, we would expect high knowledge
844 homogeneity both within and between groups if
845 circulation was not limited (Romney et al. 1986).
846 Sharing a common nomenclature depends on the
847 path for learning and knowledge transmission. As
848 landrace names are used for communicating, learn-
849 ing what is the culturally appropriate name to label a
850 given morphotype is essential for farmers. Socially
851 related farmers exchange more information
852 concerning landraces than farmers that are not re-
853 lated, and they are hence more likely to share a
854 common taxonomy (Boster 1986). First, the overall
855 higher knowledge homogeneity in the Chuka group
856 indicates more intense knowledge transmission be-
857 tween individuals within this group than in the two
858 others. Interestingly, this can be put in relation to
859 results from a previous study in the same area
860 (Labeyrie et al. 2016a), which showed that the seed
861 circulation network was more dense and cohesive in
862 the Chuka group than in the two others, possibly
863 because it was the first group to settle in the study
864 area. It is likely that these properties of seed cir-
865 culation networks enhance knowledge homogeniza-
866 tion between individuals concerning landraces.

867 Second, our results indicate that different modal-
868 ities of knowledge diffusion exist for the different
869 genetic clusters and morphotypes. Knowledge for
870 G1 identification was highly homogeneous, indicat-
871 ing intense knowledge diffusion within and be-
872 tween ethnolinguistic groups. These results are in

line with those of previous studies showing that G1 873
corresponds to *Kaguru* improved variety, which is 874
cultivated by most farmers and sold on the market 875
(Labeyrie et al. 2014). Then, major differences of 876
knowledge heterogeneity for G2 between the 877
Chuka and the other groups suggest first that 878
sources of knowledge differed between them, as 879
diffusion of the *Gadam* improved variety appears 880
to have been bound to the former group. Further- 881
more, knowledge diffusion appeared limited be- 882
tween groups for this variety, but this is not surpris- 883
ing as time since its introduction was very short 884
when the study was conducted. Knowledge patterns 885
concerning identification of the different 886
morphotypes within G3 indicate a higher knowl- 887
edge similarity between Chuka and Tharaka than 888
between these groups and the Mbeere. Higher 889
knowledge similarity was especially observed be- 890
tween the two former groups for the identification 891
Muruge and *Mugeya* landraces. This is likely related 892
to the intensity of interpersonal seed exchanges 893
between Chuka and Tharaka, which is the major 894
seed circulation modality for these varieties, while 895
exchanges with the Mbeere were very rare (Labeyrie 896
et al. 2016a). These seed circulation patterns were 897
linked to a strong alliance relationship between 898
Chuka and Tharaka groups. 899

Several studies reported a coincidence between 900
seed circulation networks and knowledge patterns. 901
For instance, cassava circulation was found to be 902
more intense among kin, who also display more 903
similar knowledge on landrace identification among 904
the Aguaruna in the Amazon (Boster 1986), and 905
similar results were observed among members of the 906
same village for rice in Gambia (Nuijten and 907
Almekinders 2008). Other studies showed that peo- 908
ple exchanging more seeds were also the most 909
knowledgeable, for instance among home garden 910
keepers in the Catalan Pyrenees (Calvet-Mir et al. 911Q4
2012), or among *caboclo* cassava farmers in Brazilian 912
Amazonia (Kawa et al. 2013). Our results are partly 913
in line with these studies, as we found a correspon- 914
dence between the structure of seed circulation 915
networks and knowledge patterns for some land- 916
races, but interestingly, this was not the case for 917
others. For instance, knowledge for *Gadam* identi- 918
fication was strongly shared by the Chuka but not 919
with the Tharaka, despite our previous study which 920
had not detected any limitation to seed circulation 921
between both groups. In addition, knowledge for 922
Kaguru identification was shared by the three 923
groups, despite this same study showing that seed 924
circulation was limited between the Mbeere and the 925

926 two other groups. An explanation for such a dis-
 927 crepancy between the structure of seed circulation
 928 networks among farmers and their knowledge pat-
 929 terns could be that they obtain a large share of their
 930 seed on the local market as well as through exten-
 931 sion services and NGOs. That is especially the case
 932 for introduced varieties released by research (unpub-
 933 lished results). Furthermore, these results suggest
 934 that farmers learn about landrace identification
 935 through different pathways depending on its nature,
 936 i.e., local or introduced, and that seed and knowl-
 937 edge of landraces do not necessarily circulate
 938 through the same channels.

939 **Conclusion**

940 This study shows that knowledge concerning
 941 sorghum landrace identification differs between ad-
 942 jacent ethnolinguistic groups and that it varies ac-
 943 cording to landraces' characteristics. First, our re-
 944 sults indicate that farmers in each ethnolinguistic
 945 group present a high level of collective consistency
 946 in naming what they consider their own sorghum
 947 landraces and anciently introduced varieties. This
 948 suggests that the consensus level of landrace identi-
 949 fication reflects how long it has been cultivated by
 950 the human group. Second, our results suggest that
 951 social pathways for learning could play a major role
 952 in shaping knowledge, as geographically close cul-
 953 tural groups present major differences in knowledge
 954 on landraces.

955 These results open perspectives to understand
 956 farmers' seed selection practices, a major driver of
 957 crop evolution and adaptation in situ. Indeed, the
 958 way individuals perceive, represent, and classify
 959 their environment affects their management prac-
 960 tices (Atran and Medin 2008). The effect of inter-
 961 cultural differences was especially observed in maize
 962 seed selection practices in Mexico, where popula-
 963 tions of this crop present divergent morphological
 964 characteristics between adjacent villages (Pressoir
 965 and Berthaud 2004) and ethnolinguistic groups
 966 (Perales et al. 2005). Our study suggests that such
 967 divergence in crop selection practices could result
 968 from differences in the identification and classifica-
 969 tion of landraces by the different human groups. It
 970 advocates for further integration of anthropology in
 971 crop diversity research as crops not only are biolog-
 972 ical objects but also bear the imprint of the societies
 973 in which they are grown, exchanged, and selected
 974 (Harlan 1975).
 975

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