Two contrasting patterns of crop seasonal adaptation revealed by a common garden experiment on flood recession sorghum in the Sahel

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Abstract

Studying seasonal adaptation is of great importance for understanding crop diversification. The biological process of seasonal adaptation of sorghum (Sorghum bicolor (L.) Moench.) allowed the development of flood recession agrosystems in the Sahel (West and Central Africa). To analyze the patterns of trait variation linked to this poorly documented process, we evaluated and compared four eco-seasonal groups of sorghum originating from seasonally diversified agrosystems in the same cropping conditions. The muskuwaari and samme groups are composed of landraces adapted to flood recession agrosystems in the Lake Chad Basin (43 accessions) and the Senegal River Valley (14 accessions), respectively. The other two groups are composed of varieties commonly grown in the rainy season (19 accessions from the Lake Chad Basin and 24 accessions from the Senegal River Valley and surrounding regions). Multivariate analyses based on ten agro-morphological traits structured the sample in two main clusters subdivided into four subclusters. This structure was significantly associated with the eco-seasonal groups, highlighting muskuwaari differentiation. Five traits (head appearance, leaf length, leaf number, aboveground biomass, and the root mass to aboveground biomass ratio) supported the differentiation of eco-seasonal groups and backed the hypothesis of seasonal adaptive divergence of sorghum in the Lake Chad Basin. No trait differentiation was detected in the samme group, suggesting that seasonal adaptation in the Senegal River Valley might probably be associated with population plasticity existing in the rainy season groups. These experimental evidences provide new insights for understanding scenarios of seasonal adaptation of sorghum.

Keywords: dry season sorghum; population adaptive divergence; population plasticity seasonal diversification; transplanted sorghum; Lake Chad Basin; Senegal River Valley; muskuwaari, samme.

Introduction

Studying seasonal adaptation is of great importance for understanding crop diversification (Fjellheim et al., 2014). Sorghum (Sorghum bicolor L. Moench) is one of the most drought-tolerant crop species, adapted to semi-arid environments (Borrell et al., 2006). It is an annual grass and an important staple crop species, adapted to semi-arid environments (Borrell et al., 2006). It is the most important food crop in the world, including parts of the USA and Europe. In the Sahel, sorghum is mainly grown in the rainy season (roughly from May to October). The crop displays a wide diversity of landraces belonging to different races and structured in different genetic groups (Deu et al., 2006; Billot et al., 2013). Rainy season cropping systems are dominant worldwide. They are based on clay-loam or sandy clay soils generally located in regions where rainfall is between 400 and 1000 mm (Chantereau et al., 2013). However, some sorghum landraces are grown during the dry season in Sahelian flood recession cropping systems in West and Central Africa. These cultivation areas include the Lake Chad Basin (northern Cameroon, Chad and northeastern Nigeria; Raimond, 1999), the Senegal River Valley (Senegal, Mauritania; Le Roy, 2005) and northern Mali (Harlan and Pasquereau, 1969). Sowing takes place in the middle or at the end of the rainy season, followed by growing and reproductive cycles in the dry season (Bruckmann, 2016). Clayey flooded soils, particularly vertisols, are the main types of soils for these farming practices, which are established in floodplains, wetlands, inland valleys and on lake shores (Sidibe et al., 2016). In India, post-rainy sorghum varieties (rabi) are sown between September and October and grown under receding soil moisture conditions, without irrigation (Patil et al., 2014).
Most dry season sorghum landraces are classified in the durra race and, rarely, in caudatum or kafir (Stemler et al., 1975; Soler et al., 2013; Raimond 1999; Patil et al., 2014; Zach et al., 1996; Naoura et al., 2019). The origin of these landraces is not well known. Studies in the Lake Chad Basin analyzed their genetic relationship with landraces grown in the rainy season (Soler, 2012; Saidou et al., 2014). Dry season sorghum landraces of the Lake Chad Basin have complex local taxonomies (Saidou et al., 2014). The main types are called muskuwaari (northern Cameroon), masakwa (northeastern Nigeria), or berbere (southern Chad) (Raimond, 1999). Seeds are sown in the middle of the rainy season (from mid-August) in small nurseries (sandy soils) close to the village or the field. Seedlings are transplanted to vertisol fields (called karali in northern Cameroon) following flood recession (between October and November). Sowing in the nursery is commonly replicated by farmers two to four times, in order to match a transplanting stage (seedlings of around 40-45 days and 30 to 40 cm tall) to the variable flood recession time. Harvesting is carried out from February to March (Barrault et al., 1972; Caro et al., 1995).

Unlike in the Lake Chad Basin, flood recession sorghum landraces in the Senegal River Valley (samme) are sown directly in the field in flood plains (kolangal) following the recession flow at the end of the rainy season, from October to November (Le Roy, 2005; Bruckmann, 2016). The harvesting period is from March to April (Chantereau, 2002). The adaptation of annual grasses to dry season involves the ability of the plant to respond to a shift in environmental conditions (Fjellheim et al., 2014). For sorghum, this includes adaptation to a water regime based on the water stock remaining in the soil (mainly vertisols), a response to the temperature and photoperiod changes and to specific farmers management practices (particularly in the Lake Chad Basin, where sowing in a nursery and transplanting practices are widespread).

Knowledge about plant adaptation to variations in water regimes and temperatures (including drought) could help to identify some of the adaptive traits linked to seasonal adaptation (Redmann, 1985). Various combinations of morpho-physiological features contribute to the ability of plants to adapt to these variations, including shorter plants, smaller leaf area, increased root biomass and high chlorophyll content; all of which are beneficial under drought stress in order to avoid desiccation (Jaleel et al., 2013). The survival of deep drought (Passioura, 2012). The adaptation of annual grasses to dry season involves the ability of the plant to respond to a shift in environmental conditions (Fjellheim et al., 2014). For sorghum, this includes adaptation to a water regime based on the water stock remaining in the soil (mainly vertisols), a response to the temperature and photoperiod changes and to specific farmers management practices (particularly in the Lake Chad Basin, where sowing in a nursery and transplanting practices are widespread).

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of uncertainty in accession assignation over DAPC using the described ad hoc procedure, we discarded artefact effects of marginal accessions in each cluster and validated these tests (Table S5). Of all the factors, the eco-seasonal group had the highest level of association (P = 1.85 x 10^{-14} and P = 1.16 x 10^{-16} at K=2 and K=4, respectively; Table 4), suggesting a major structuring effect of the eco-seasonal diversification on the variation in agro-physiological traits of sorghum. Muskuwaari accessions were all from the durra race (Table S4; Table S1). The racial differentiation between clusters was supported by the predominance of the durra race in Cluster 1 (Table S4).

**Variation in agro-physiological traits between eco-seasonal groups and between inferred clusters**

The summary statistics of the traits according to the four eco-seasonal groups were computed (Table S2). Two of the traits exhibited a statistically significant distribution shift between eco-seasonal groups for head appearance (P = 3.08 x 10^{-11}) and leaf length (P = 1.51 x 10^{-05}) (Table S6a; Figure 3). Pairwise comparisons suggested that the muskuwaari group was differentiated form the other types (Tukey test, Table S6b). Muskuwaari flowered significantly later (115 days, on average, from sowing to head appearance), compared to the other three groups (average values grouped between 74 and 81 days). This eco-seasonal group also exhibited shorter leaves (LL average of 50.36 cm versus 68.15 for samme, 68.72 cm for rainy season varieties of the Lake Chad Basin, and 72.77 cm for rainy season varieties of the Senegal River Valley).

The evaluated traits statistics were also computed according to the inferred clusters (Table S3). The analysis of variance of the traits according to the inferred clusters (K=2) showed a significant distribution shift between Cluster 1 (mainly composed of muskuwaari) and Cluster 2 for head appearance, leaf length, leaf number, aboveground biomass and the root mass to aboveground biomass ratio (Figure 4; Table S7a).

We also deciphered the differentiations between the four subclusters (Table S7b). The subclusters 1a and 1b containing muskuwaari were differentiated from each other based on leaf width (P = 5.80 x 10^{-05}), root number (P = 1.00 x 10^{-06}), and aboveground biomass (P = 9.04 x 10^{-06}). The Cluster 1a (exclusively composed of muskuwaari) had a larger leaf width (average difference of +1.53 cm), more aboveground biomass (+123.39 g per plant), and a smaller root number (-23.69 roots per plant), compared to the Cluster 1b (Figure S3).

Across the two eco-geographical zones, three traits were significantly shaped by the seasonal type of sorghum, namely head appearance, leaf length and aboveground biomass (Table S6a). Four traits varied between eco-geographical across all accessions (Table S6a).

**Discussion**

**Signal of eco-seasonal structure in sorghum validated with robust statistical procedures**

We used an ad hoc validation procedure to overcome statistical shortcomings in DAPC structure inference, notably the uncertainty in the inference of the number of clusters and in the assignation of accessions to the inferred clusters (Jombart et al., 2010; Soler et al., 2013). Using empirical in silico resampling on our data, we showed that BIC is sensitive to sample size and provides unstable results across runs (Figure 5a). This criterion tended to detect a larger number of clusters (fine level of structure). The DiffGroup criterion (derived from BIC) detected a basic level of structure (small number of clusters), which was consistent across resampled data and less sensitive to sample size (Figure 5a). We therefore opted for the number of clusters suggested by this criterion (K=2). This was also consistent with the PCA projection (Figure 1). All DAPC runs were performed without including a priori eco-seasonal group information. So the assignation of all muskuwaari accessions in the same cluster by DAPC a posteriori indicated an actual phenotypic differentiation (Table 3).

The statistical association of seasonal type, eco-geographical origin, eco-seasonal group, race and country of origin with the phenotypic structure was consistent regardless to the uncertainty in the assignation of marginal accessions in the clusters (Table 4; Table S5).

**Identification of the agro-physiological traits and mechanisms linked to seasonal adaptation**

The reason for using a common garden experiment was to make environmental and cropping conditions similar, in order to enable the analysis of genetic differences between the four studied populations (Vitasse et al., 2009). Each group was represented by an average of 25 ±6.33 accessions (the minimum being n=14; Table 1) representing the diversity of the group.

Over all presented analyses, we identified five traits that differentiated the muskuwaari type (or clusters predominantly composed of muskuwaari) from the other eco-seasonal groups, namely head appearance, leaf length, leaf number, aboveground biomass and the root mass to aboveground biomass ratio. Muskuwaari landraces are very late flowering when sown in the rainy season. During this trial, the muskuwaari landraces sown in July flowered between December and January (around 6 to 7 months’ duration). This behavior has been described in northern Cameroon (Barrault et al., 1972). It was linked to the high photoperiod sensitivity of muskuwaari (Chantereau et al., 2013; Craufurd et al., 1999).

Leaf size reduction is a mechanism for adjusting leaf area to avoid stress (Blum and Sullivan, 1986). Larger values of the root mass to aboveground biomass ratio indicate an increase in root density over leaf area, a phenomenon that offers the advantages of minimizing water loss and maximizing water uptake in a context of water stress (Passioura, 2012). Under the conditions of our trial (no induced water stress), the dry season accessions (particularly muskuwaari from Cluster 1) showed a lower root mass to aboveground biomass ratio, compared to the rainy season accessions. This could indicate more efficient water uptake and use, meaning an ability to produce more aboveground biomass with fewer roots.

Leaf area reduction is one of the mechanisms involved in limiting evapotranspiration (Levitt, 1972). Although this trait did not show a significant difference between the four eco-seasonal groups (Table S6a), it was found that the mean of leaf area of dry season sorghum tended to be lower than that of the rainy season sorghums, particularly for the accessions from the Senegal River Valley (Table S2). Variation in leaf width is associated with drought adaptation in grasses (Redmann 1985).
Table 1. Presentation of the sampled sorghum panel (n=100 accessions)

<table>
<thead>
<tr>
<th>Seasonal Type</th>
<th>Lake Chad Basin (LCB)</th>
<th>Senegal River Valley (SRV)</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Cameroon</td>
<td>Chad</td>
<td>Total</td>
</tr>
<tr>
<td>Rainy season sorghum (RSS)</td>
<td>19</td>
<td>0</td>
<td>19</td>
</tr>
<tr>
<td>Dry season sorghum (DSS)</td>
<td>32</td>
<td>11</td>
<td>43</td>
</tr>
<tr>
<td><strong>Total number of accessions</strong></td>
<td><strong>51</strong></td>
<td><strong>11</strong></td>
<td><strong>62</strong></td>
</tr>
</tbody>
</table>

The number of accessions sampled is presented according to the eco-geographical origin of the accessions (LCB or SRV) and their seasonal type (RSS or DSS). The cross between an eco-geographical origin and a seasonal type defined an eco-seasonal group. In the text, LCB DSS will be referred to as muskuwari and SRV DSS will be referred to as somme.

Fig 1. Projection of the traits and the accessions by principal component analysis (PCA). A. Correlation circle of the 10 considered traits. B. Projection of the 100 accessions, colored according to the eco-seasonal group. Dry season sorghum from Lake Chad basin (left; in red) appeared as the most distinguished group. HA : head appearance, LW : leaf width, LL : leaf length, LN : leaf number, DL : leaf dead percentage, CC : chlorophyll content, LA : leaf area, RN : root number, RM : root dry weight, AB : aboveground biomass.

Table 2. List and definition of the agro-physiological traits.

<table>
<thead>
<tr>
<th>Characters</th>
<th>Code</th>
<th>Category</th>
<th>Measurement method</th>
<th>Unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. Head appearance</td>
<td>HA</td>
<td>Phenology</td>
<td>Number of days from sowing to heading</td>
<td>day</td>
</tr>
<tr>
<td>2. Leaf width</td>
<td>LW</td>
<td>Leaf trait</td>
<td>Width of the third leaf below the last ligulated leaf</td>
<td>cm</td>
</tr>
<tr>
<td>3. Leaf length</td>
<td>LL</td>
<td>Leaf trait</td>
<td>Length of the third leaf below the last ligulated leaf</td>
<td>cm</td>
</tr>
<tr>
<td>4. Leaf number</td>
<td>LN</td>
<td>Leaf trait</td>
<td>Count of the number of harvested leaves (including green, dead and fallen leaves)</td>
<td></td>
</tr>
<tr>
<td>5. Percentage of dead leaves</td>
<td>DL</td>
<td>Leaf trait</td>
<td>Number of dead leaves at harvest divided by total number of observed leaves at harvest</td>
<td>%</td>
</tr>
<tr>
<td>6. Chlorophyll content</td>
<td>CC</td>
<td>Leaf trait</td>
<td>Chlorophyll content in leaf green tissue obtained with SPAD (three measurements per leaf, early in the morning)</td>
<td>-</td>
</tr>
<tr>
<td>7. Leaf area</td>
<td>LA</td>
<td>Leaf trait</td>
<td>Area of all the green leaves of the plant before harvest calculated using the formula of Stickler et al., (1961) (length x width x 0.747)</td>
<td>cm²</td>
</tr>
<tr>
<td>8. Root number</td>
<td>RN</td>
<td>Root trait</td>
<td>Count of the number of nodal roots.</td>
<td>-</td>
</tr>
<tr>
<td>9. Root mass</td>
<td>RM</td>
<td>Root trait</td>
<td>Roots from each individual plant collected and dried separately at 60°C for 72 hours before obtaining dry mass.</td>
<td>g</td>
</tr>
<tr>
<td>10. Aboveground biomass</td>
<td>AB</td>
<td>Biomass</td>
<td>Leaves and stems collected from each individual plant and dried separately at 60°C for 72 hours before obtaining dry mass.</td>
<td>g</td>
</tr>
</tbody>
</table>

Two supplemental traits were computed from the measured traits: the RM/LA ratio (root mass to leaf area) and the RM/AB ratio (root mass to aboveground biomass).
Fig 2. Structure of sorghum panel revealed by discriminant analysis of principal components (DAPC). a) K=2. The two clusters (Cluster 1 and Cluster 2) are represented and distinguished with color codes. b) K=4. Cluster 1 (identified in subplot a) was split into two subclusters: 1a and 1b; Cluster 2 was split into 2a and 2b. Clusters are shown by different colors and inertia ellipses, while dots represent individual accessions. The first two discriminant functions are plotted, explaining a total of 93.63% of the variance.

Table 3. Repartition of the number of accessions between clusters inferred using DAPC and eco-seasonal groups. a. Contingency table between clusters inferred using DAPC and the eco-seasonal groups.

<table>
<thead>
<tr>
<th>Eco-seasonal groups</th>
<th>Clusters (K=2)</th>
<th>Subclusters (K=4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>LCB DSS</td>
<td>1</td>
<td>1a</td>
</tr>
<tr>
<td>LCB RSS</td>
<td>41</td>
<td>1b</td>
</tr>
<tr>
<td>SRV DSS</td>
<td>5</td>
<td>2</td>
</tr>
<tr>
<td>SRV RSS</td>
<td>20</td>
<td>1</td>
</tr>
</tbody>
</table>

| Cluster size        | 50            | 49                |

b. Contingency table between the two main clusters (K=2) and the four subclusters (K=4).

<table>
<thead>
<tr>
<th>Clusters</th>
<th>K = 4</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1a</td>
</tr>
<tr>
<td>K=2</td>
<td>26</td>
</tr>
<tr>
<td></td>
<td>1</td>
</tr>
</tbody>
</table>

a) Repartition of the number of accessions from inferred clusters in the eco-seasonal groups of accession origin. b) Repartition of the number of accessions from the two main inferred clusters (K=2) in the four inferred clusters (K=4). LCB: Lake Chad Basin, SRV: Senegal River Valley, DSS: dry season sorghum, RSS: rainy season sorghum.

Fig 3. Distribution of head appearance and leaf length according to sorghum eco-seasonal group. Muskuwaari group (LCB_DSS) showed a distribution shift, suggesting phenotypic differentiation from rainy season sorghum (later flowering and shorter leaves). Samme group (SRV_DSS) distribution did not differentiate from that of the two rainy season groups.
Table 4. P-values of association between the classification factors and the inferred clusters (Chi-squared test).

<table>
<thead>
<tr>
<th>Inferred clusters (DAPC)</th>
<th>Eco-geographical zone of origin (n=2)</th>
<th>Seasonal type (n=2)</th>
<th>Eco-seasonal group (n=4)</th>
<th>Race (n=8)</th>
<th>Country of origin (n=4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>K=2</td>
<td>1.26 x 10^{-09}</td>
<td>7.25 x 10^{-07}</td>
<td>1.85 x 10^{-14}</td>
<td>1.32 x 10^{-08}</td>
<td>3.27 x 10^{-07}</td>
</tr>
<tr>
<td>K=4</td>
<td>9.34 x 10^{-08}</td>
<td>3.43 x 10^{-07}</td>
<td>1.16 x 10^{-10}</td>
<td>6.88 x 10^{-06}</td>
<td>8.69 x 10^{-07}</td>
</tr>
</tbody>
</table>

These results provided candidate traits and candidate mechanisms for understanding the seasonal adaptation of sorghum. Further physiological studies are needed to validate the mechanisms underlying these features. Genome-wide association studies (GWAS) could assess the link between these traits and the polymorphisms associated with seasonal adaptation (Fjellheim et al., 2014).

Dry season in the Lake Chad Basin and the Senegal River Valley: contrasting patterns of seasonal adaptation?

The differentiations shown in five traits suggested that seasonal adaptation of sorghum in the Lake Chad Basin might be linked to population adaptive divergence of *muskuwari*. This backed the hypothesis of an adaptive evolution that genetically shaped the evolutionary history of *muskuwari* (Saidou et al., 2014).

Dry season sorghums from the Senegal River Valley (*samme*) included diverse races (D, DC, G, GC) and exhibited an agronomic behavior close to that of the rainy season sorghum varieties. This eco-seasonal group did not show any significant trait differences with the rainfed accessions. This suggested that the adaptation of sorghum to dry season cultivation in flood recession systems of the Senegal River Valley might be based on the expression of a population phenotypic plasticity existing in rainy season sorghum.

### Material and methods

### Studied sorghum panel

The studied sorghum panel is composed of four eco-seasonal groups of landraces; *muskuwari* (landraces grown in dry season sorghum in Lake Chad Basin, generally transplanted), *samme* (landraces grown in dry season in Senegal River Valley), rainy season landraces from the Lake Chad Basin and rainy season landraces from the Senegal River Valley and surrounding regions (Table 1). These groups were represented by respective samples of 43, 14, 19, 24 accessions, corresponding to a panel of 100 accessions (see Table S1 for passport data and Figure S1 for the map of locations of origin of the accessions). This panel is part of a larger Sorghum Seasonal Study Panel (S3P; n=153 accessions) evaluated at Bamby (Senegal).

### Field trial

The four eco-seasonal groups were evaluated in a common garden experiment carried out during the rainy season 2016 at the Centre National de Recherche Agronomique of Bamby, ISRA (Senegal). The accessions representing each of the four eco-seasonal groups were randomly assigned to the plots using a design with spatial control of plots heterogeneity. The spatial control was allowed by the mapping of plot position in the field (row and column...
coordinates recorded) to capture the heterogeneity among plots. Seeds were sown on the 28th July 2016 under natural rainfed conditions (5 seeds per hill). A thinning was done 15 days after sowing to keep only two seedlings per hill. All plots received supplemental irrigation at the end of the rainy season to allow for the completion of the cycle of late flowering varieties (last harvest on January the 3rd 2017). Each plot measured 3.2 m x 3.2 m and spaced 1 m from the others. Inside plots, 0.8 m spacing between hills was used in each side. For each accession, 25 hills were sown (one plot per accession). Four plants per accession (from four different hills) were used for the traits measurements and notations.

**Phenotypic traits notation**

Ten traits related to phenology, leaf, root, and plant biomass were measured (Table 2). Two other traits were calculated from the measured traits: the ratio RM/LA (root mass over leaf area) and the ratio RM/AB (root mass over aboveground biomass). Except for the date of heading and leaf chlorophyll content, all traits were measured at harvest period. Leaf chlorophyll content was evaluated before harvest, based on SPAD chlorophyll meter (Süß et al., 2015).

**Phenotypic data modelling and accession BLUP extraction**

The recorded data were analyzed independently for each trait using a linear mixed-effects model accounting for the spatial structure of the design to estimate accession effect while controlling for field heterogeneity. The linear mixed model was set as follows:

\[ Y = \mu + S x + X \beta + e \] (model 1);

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The model was fitted using the package nlme (version 3.1-137) in R (version 3.4.3; R Core Team, 2017). The dry season sorghum group from Senegal River Valley (version 3.1-137) in R (version 3.4.3; R Core Team, 2017).

**Inference of population agro-physiological structure**

Discriminant analysis of principal components (DAPC) (Jombart et al., 2010) was used to infer the structure (adegenet R package) of the sample based on the BLUPs of the 10 measured traits described above (Table 2). Two criteria proposed by Jombart et al. (BIC and DiffNgroup) were used to infer the optimal number of clusters, K (K varied from 1 to 20). We assessed the effect of sample size and the effect of across runs variability on the inference of the optimal number of clusters for each criteria. The procedure was defined as follows: i) randomly sample a subset of accessions of size n with no replacing (n varied from 40 to 99); ii) for the sampled subset, make a run of find.clusters() function and choose the best value of K using BIC and DiffNgroup independently; iii) repeat the analysis 100 times for each value of n; iv) compute the frequency of choice of each possible values of K over the 100 runs for each sample size. Next, DAPC was ran at chosen values of K to make the final clustering and assign accessions into the clusters. One accession over the 100 was discarded from this analysis due to missing data (CM134). The number of principal components retained during the process of DAPC was determined with the cross-validation approach implemented in the function xvalDapc(), with 1000 repetitions. This optimization was tested by varying the number of K only from 2 to 10. The number of principal components was set to eight after exploratory analysis, accounting for about 86% of the total phenotypic variance.

**Test of association between inferred phenotypic structure and sampling factors**

We performed Chi-square tests to assess the association of the inferred phenotypic structure with each of the factors (eco-seasonal groups, race, region of origin and seasonal type). As the assignation of accessions to DAPC clusters could be variable over runs, we validated the robustness of the association using Chi-square test replication based on accessions in silico resampling (Table S5). The resampling was structured with respect to the Euclidian distance of accessions to DAPC clusters’ centre. Our assumption was that subsamples composed of individuals closer to cluster centres are the best-assigned accessions and reflect an unbiased trend of association with the factors.

**Analysis of variance of the traits**

For each trait, an analysis of variance was performed based on two respective models. The first model fitted the effect of eco-seasonal group and its underlying factors (seasonal type and eco-geographical of origin) as follows:

\[ Y_{ijk} = \mu + \alpha_i + \beta_j + \gamma_{ij} + e_{ijk} \] (model 2);

\[ Y_{ijk} = \mu + \alpha_i + \beta_j + \gamma_{ij} + e_{ijk} \] (model 2).

Where \( Y_{ij} \) represents the phenotypic value for the \( i \)th observation in the \( j \)th cluster, \( \mu \) is the intercept, \( \alpha_i \) represents the fixed effect of the cluster \( i \) and \( e_{ij} \) represents the random error.

For the Y vector in these two models, we used the BLUPs previously estimated (model 1a). All the accessions representing an eco-seasonal group were used as random repeats for this group. When the analysis of variance was significant (P<0.05), Tukey’s honestly significant difference (HSD) was used to check pairwise differences.

**Conclusion**

This study showed that the agro-physiological traits diversity of sorghum supports an eco-seasonal structure linked to the diversification of the crop in Lake Chad Basin. Dry season sorghum from the Lake Chad Basin (muskuwaari), and the associated clusters (Cluster 1, Subclusters 1a and 1b), showed a differentiation from rainy season sorghum based on head appearance, leaf length, leaf number, aboveground biomass, and the root mass to aboveground biomass ratio. This pattern of trait shift, observed through a common garden experiment with controlled design, backs a hypothesis of genetic differentiation and adaptive divergence linked to the seasonal diversification of sorghum. The dry season sorghum group from Senegal River Valley (samme), showed no particular trait shift compared to both rainy season groups. This pattern, contrasting with the one
observed in *muskuwaari*, suggested that the seasonal adaptation of *samme* was more likely supported by a population plasticity in the rainy season sorghum. These evidences provided new insights for understanding the seasonal adaptation of sorghum in the flood recession agrosystems and, more widely, the seasonal adaptation of annual crop species.

**AUTHORS’ CONTRIBUTIONS**

AAS, HIJ, OF, AA and DF designed the study. OB, OF, BS and AA conducted the trials. OF and AAS carried out statistical analyses. All the authors took part in discussing the results. AAS, OF, GK and HIJ wrote the paper.

**DECLARATIONS OF INTEREST:** There is no conflict.

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