

## Selection Indices Studies in Maize (*Zea mays* L.) under Multiple Abiotic Stress Conditions

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### Abstract

The present investigation was undertaken on maize (*Zea mays* L.) under two abiotic stress conditions viz. low-nitrogen (low soil fertility) and excess soil moisture conditions (waterlogging) during *kharif* 2010. Studies were carried out for developing suitable selection indices in relation to multiple abiotic stress tolerance. The experimental material consisted of twelve lines and four testers and their 48 single crosses planted in Randomized Block Design. Analysis of variance revealed significance for all the traits studied in both the conditions, indicating there by the existence of high genetic variability in the genotypes. In low-N conditions, the best performing line was  $L_5$  followed by and among testers  $T_3$  ranked highest. Among the crosses,  $L_{11} \times T_4$  ranked highest followed by  $L_8 \times T_1$  and  $L_{11} \times T_3$ . In waterlogging conditions,  $L_6$  performed better followed by  $L_5$ . Among the crosses  $L_6 \times T_4$  ranked highest. Economic weights were assigned to various traits under low-nitrogen and excess soil moisture (ESM) conditions. Among the crosses, aggregate score were higher in most of the cases but general trend was that ESM traits had the lower score values. Crosses  $L_{11} \times T_4$ ,  $L_8 \times T_1$ ,  $L_{11} \times T_3$ ,  $L_7 \times T_2$  and  $L_6 \times T_2$  performed well in low-nitrogen conditions for the assigned selection criteria. In waterlogging trials,  $L_6 \times T_4$ ,  $L_7 \times T_2$  and  $L_4 \times T_1$  performed excellent while crosses  $L_5 \times T_1$ ,  $L_{12} \times T_2$ ,  $L_5 \times T_4$ , and  $L_3 \times T_3$  ranked very low in ESM trials. Thus for ESM tolerance, the crosses  $L_6 \times T_4$  and  $L_7 \times T_2$  are the best available crosses according to the given selection criteria and for low-nitrogen tolerance,  $L_{11} \times T_4$  and  $L_8 \times T_1$  were the best crosses.

### 1. Introduction

Crowned with the rhetoric phrase the queen of cereal maize (*Zea mays* L.) is the world's most important cereal and occupies the apex position in terms of production at the world level. It is an economically important cereal crop for human food, animal feed, fiber and a range of industrial uses. Maize crop grown in tropics during summer-rainy season occasionally face extreme climatic conditions and various biotic or abiotic stresses that severely limit crop growth and development and eventually yield potential. Among the abiotic stresses, low nitrogen and Excess Soil Moisture (ESM) are the most important constraints for maize production and productivity in the world and Asian region. Indian soils have been characterized as being low in organic matter and N. Tropical rain fed soils in semi-arid regions are typically low in organic matter, often containing less than 1% organic matter, with a total N generally not exceeding 0.1%. About 77% of these rain fed soils are classified as low to very low in available N (Katyal et al., 1994). Recent statistics on N fertilizer consumption pattern showed that average

application of N in developed countries is 250 kg ha<sup>-1</sup>, while in developing countries it is 82 kg ha<sup>-1</sup>, and in Sub-Saharan African countries it is as low as 5.0 kg ha<sup>-1</sup> (SAA, 2002). The trend of low N application contributes to low maize yield of about 1–2 t ha<sup>-1</sup> in developing countries (CIMMYT, 1994), which is in stark contrasts to yields reported from research stations in the same countries ranging from 4–12 t ha<sup>-1</sup> (CIMMYT, 1995, 1996). Yields in Africa are considerably lower than the world average because the cultivation of maize is often prone to low soil fertility (primarily N but also P and other nutrient deficiencies) in addition to biotic stresses (FAO, 2010). Additional reasons for low-N fertility in the tropics include poor soil types with low N mineralization, high run off or leaching of applied fertilizer with heavy rainfall due to light textured soils, and poor uptake of applied N due to problems of water stress.

In India, excess soil moisture caused by flooding, waterlogging, high water table or heavy soil texture is also one of the most important constraints for crop production where about 8.5 mha of arable land is in the grip of this problem. In



case of June planting it may coincide with flowering which may interfere with the normal pollination behavior and seed setting (Savita et al., 2004). In India out of total 6.6 mha area of maize over 2.5 mha is prone to face excessive soil moisture/ waterlogging conditions, which causes on average 25–30% loss of national maize production almost every year (DMR, 2001). The maize crop suffers badly whenever it encounters temporary excess soil moisture (ESM) conditions during the monsoon season or grown in poorly drained converted paddy fields after a rainy season rice crop, a practice more common in Thailand, Japan and Eastern China (Shimizu, 1992). Maize plants are injured more and greater yield losses occur when flooded at early stages. Therefore, it would be desirable to develop maize cultivars with increased resistance to low N and ESM conditions which is desirable for sustainable production systems with improved yield, without any threat on environment and ecosystems.

Population improvement of a crop is the primary objective of a plant-breeding program. However, the progress in any breeding program depends primarily upon the genetic diversity and the effectiveness of the selection procedure involved. Besides other selection method used for the Population improvement, selection indices are considered as an aid to the breeder for simultaneous selection of multiple traits (Smith, 1936). This tool can help the breeder in spotting the desirable genotype/family of a crop species in a population improvement program. Smith (1936) and Hazel (1943) illustrated the procedure for constructing a selection index that gives maximum advance from selection. Selection indices provide useful information about which traits to be combined (Banziger and Lafitte, 1997). Selection indices have been an effective selection criterion to increase grain yield in corn (Modaressi et al., 2004). Many other Researchers have used selection indices as an effective selection criterion in their breeding programs on different crops (Vikram and Roy, 2003; Xie et al., 1998; Dolan et al., 1996). Keeping this in mind, the proposed study was carried out with the objective to construct selection indices with regard to low N and ESM tolerance.

## 2. Materials and Methods

The present investigation was undertaken with twelve inbred lines and four testers which were initially screened for their performance in low-N and ESM tolerance and normal conditions at the Crop Research Center of Govind Ballabh Pant University of Agriculture and Technology, Pantnagar, Uttarakhand, India. These inbred lines were then crossed in line×tester fashion at Pantnagar to generate  $F_{1s}$ . The experimental material thus consisted of twelve lines, four testers and  $48F_{1s}$ . Field trials were laid down in low-N, excess soil moisture and normal conditions during *khari*f

2010 at Pantnagar. The details of the parental lines have been indicated in the Table 1 (Table 1: Details of the parental lines).

Table 1: Details of the parental lines

Parents	Pedigree	Coded pedigree
1.	POB. 33 C <sub>3</sub> -12-2-1-1-2-2	L <sub>1</sub>
2.	POB. 33 C <sub>3</sub> -12-2-1-2-2-5	L <sub>2</sub>
3.	POB. 33 C <sub>3</sub> -142-1-6-1-1-4	L <sub>3</sub>
4.	POB. 45 C <sub>8</sub> -86-1-3-7-6-4	L <sub>4</sub>
5.	POB. 45 C <sub>8</sub> -45-2-6-1-2-7	L <sub>5</sub>
6.	POB. 45 C <sub>8</sub> -269-2-4-6-3-3	L <sub>6</sub>
7.	POB. 45 C <sub>8</sub> -86-1-1-7-5-1	L <sub>7</sub>
8.	CLG 1708-1-1-9	L <sub>8</sub>
9.	POB. 45 C <sub>8</sub> -45-2-6-1-1-1	L <sub>9</sub>
10.	POB. 45 C <sub>8</sub> -86-1-3-4-5-2	L <sub>10</sub>
11.	POB. 45 C <sub>8</sub> -86-1-3-2-2-5	L <sub>11</sub>
12.	POB. 45 C <sub>8</sub> -269-2-4-6-6-1	L <sub>12</sub>
13.	POB. 445 × 58-6-3-B-B-B	T <sub>1</sub>
14.	POB. 446-74-2-B-B-B	T <sub>2</sub>
15.	CML-421	T <sub>3</sub>
16.	CML-423	T <sub>4</sub>

For the present study, parents along with all the developed generations were grown in Randomized Block Design in one row plot of 5 m length with 3 replications following spacing of 75 cm between rows and 25 cm between plants. The crop management practices were applied in the three environments as per the recommendations. In low N condition, 40 kg N ha<sup>-1</sup> was applied. In ESM trials, waterlogging treatment was given at knee high growth stage for 6 days, by keeping continuous submergence with an average depth of ponding of about 5 cm. After 6 days of ponding, water was drained out of the plots. The data were recorded on days to 50% tasseling, days to 50% silking, anthesis, silking interval, plant height, ear height, cob length, cob diameter, 100-kernel weight, grain yield, nodes bearing adventitious roots, leaf senescence and number of ears per plant. Appropriate statistical and biometrical analysis for selection index (Smith-Hazel) was carried out. A selection index with many traits is likely to have low heritability (Bernardo and Yu, 2007). Genotypic and phenotypic variance-covariance matrices were developed to aid calculation of index coefficients. Estimated indices were calculated by the method described by Smith (1936).

Application of discriminate function as a basis for making selection on several characters simultaneously is aimed at discriminating the desirable genotypes from undesirable genotypes on the basis of their phenotypic performance.



Later on, Hazel (1943) developed a simultaneous selection model following path analysis approach.

Smith (1936) defined the genetic worth of an individual (H) as:

$$H = a_1 G_1 + a_2 G_2 + \dots + a_n G_n = \sum_{i=1}^n a_i G_i$$

Where,  $G_1, G_2, G_n$  are the genotypic (breeding) values of individual characters and  $a_1, a_2, \dots$ , signify their relative economic importance. Another function (I) based on the phenotypic performance of various characters, is defined as:

$$I = b_1 p_1 + b_2 p_2 + \dots + b_n p_n$$

Where,  $b_1, b_2, \dots, b_n$  are to be estimated such that the correlation between H and I, i.e.  $r(H, I)$ , becomes maximum. Once such function is obtained, discrimination of good genotypes from the undesirable ones will be possible on the basis of phenotypic performance, i.e.  $p_1, p_2, \dots, p_n$  directly.

The maximization of  $r(H, I)$  leads to a set of simultaneous equation which upon solving give the desired estimate of  $b_i$  values.

Considering 4 characters, the simultaneous equations look like as follows:

$$b_1 \times 11 + b_2 \times 12 + b_3 \times 13 + b_4 \times 14 = a_1 G_{11} + a_2 G_{12} + a_3 G_{13} + a_4 G_{14}$$

$$b_1 \times 21 + b_2 \times 22 + b_3 \times 23 + b_4 \times 24 = a_1 G_{21} + a_2 G_{22} + a_3 G_{23} + a_4 G_{24}$$

$$b_1 \times 31 + b_2 \times 32 + b_3 \times 33 + b_4 \times 34 = a_1 G_{31} + a_2 G_{32} + a_3 G_{33} + a_4 G_{34}$$

$$b_1 \times 41 + b_2 \times 42 + b_3 \times 43 + b_4 \times 44 = a_1 G_{41} + a_2 G_{42} + a_3 G_{43} + a_4 G_{44}$$

Which in matrix form become :

$$\begin{bmatrix} x_{11} & x_{12} & x_{13} & x_{14} \\ x_{21} & x_{22} & x_{23} & x_{24} \\ x_{31} & x_{32} & x_{33} & x_{34} \\ x_{41} & x_{42} & x_{43} & x_{44} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \begin{bmatrix} G_{11} & G_{12} & G_{13} & G_{14} \\ G_{21} & G_{22} & G_{23} & G_{24} \\ G_{31} & G_{32} & G_{33} & G_{34} \\ G_{41} & G_{42} & G_{43} & G_{44} \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix}$$

The solution of these equations gives the estimates of  $b_i$  values in the following manner :

$$b = X^{-1}Ga$$

Where,  $b$  is the column vector,  $X^{-1}$  is the inverse of phenotypic variance and covariance matrix and  $a$  is the column vector for economic weights.

The various steps involved in the construction of selection indices are described below:

#### (i) Estimation of variance and covariance matrix

Genotypic and phenotypic variances and covariance's were determined using Randomized Complete Block Design with analysis of variance for each trait and covariance between every pairs of traits.

The expected mean sums of squares are as follows:

$$E(MS_g) = \sigma_g^2 + r\sigma_e^2$$

$$E(MS_e) = \sigma_e^2$$

And, therefore,  $\sigma_g^2 = \frac{MS_g - MS_e}{r}$ , where  $r$  is number of replications.

Thus, the estimated genotypic variance is  $\sigma_g^2$  and the environmental variance is  $\sigma_e^2$ . The phenotypic variance,  $\sigma_p^2$ , is calculated as  $\sigma_g^2 + \sigma_e^2$ .

Similarly, expectations of mean sum of products are calculated as follows:

$$E(MSP_g) = \sigma e_1 e_2 + r\sigma g_1 g_2$$

$$E(MSP_e) = \sigma e_1 e_2$$

$$\text{And hence, } \sigma g_1 g_2 = \frac{MSP_g - MSP_e}{r}$$

Thus, the estimated genotypic covariance is  $\sigma g_1 g_2$ , and the environmental covariance is  $\sigma e_1 e_2$ . The phenotypic covariance,  $\sigma p_1 p_2$  is calculated as  $\sigma g_1 g_2 + \sigma e_1 e_2$ . The genotypic variance and covariance matrix is constructed as follows:

$$\begin{bmatrix} \sigma_{g1}^2 & \sigma_{g1g2} & \sigma_{g1g3} & \sigma_{g1g4} \\ \sigma_{g1g2} & \sigma_{g2}^2 & \sigma_{g2g3} & \sigma_{g2g4} \\ \sigma_{g1g3} & \sigma_{g2g3} & \sigma_{g3}^2 & \sigma_{g3g4} \\ \sigma_{g1g4} & \sigma_{g2g4} & \sigma_{g3g4} & \sigma_{g4}^2 \end{bmatrix}$$

And genotypic variance and covariance matrix is constructed as follows:

$$\begin{bmatrix} \sigma_{p1}^2 & \sigma_{p1p2} & \sigma_{p1p3} & \sigma_{p1p4} \\ \sigma_{p1p2} & \sigma_{p2}^2 & \sigma_{p2p3} & \sigma_{p2p4} \\ \sigma_{p1p3} & \sigma_{p2p3} & \sigma_{p3}^2 & \sigma_{p3p4} \\ \sigma_{p1p4} & \sigma_{p2p4} & \sigma_{p3p4} & \sigma_{p4}^2 \end{bmatrix}$$

#### (ii) Setting of simultaneous equations

Assuming that all characters are economically equally important,  $a_1 = a_2 = a_3 = a_4 = 1$ , we can write down the simultaneous equation (in a matrix form) by using the values given in above two matrices.

$$\begin{bmatrix} \sigma_{p1}^2 & \sigma_{p1p2} & \sigma_{p1p3} & \sigma_{p1p4} \\ \sigma_{p1p2} & \sigma_{p2}^2 & \sigma_{p2p3} & \sigma_{p2p4} \\ \sigma_{p1p3} & \sigma_{p2p3} & \sigma_{p3}^2 & \sigma_{p3p4} \\ \sigma_{p1p4} & \sigma_{p2p4} & \sigma_{p3p4} & \sigma_{p4}^2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \begin{bmatrix} \sigma_{g1}^2 & \sigma_{g1g2} & \sigma_{g1g3} & \sigma_{g1g4} \\ \sigma_{g1g2} & \sigma_{g2}^2 & \sigma_{g2g3} & \sigma_{g2g4} \\ \sigma_{g1g3} & \sigma_{g2g3} & \sigma_{g3}^2 & \sigma_{g3g4} \\ \sigma_{g1g4} & \sigma_{g2g4} & \sigma_{g3g4} & \sigma_{g4}^2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

#### (iii) The solution of the equation for $b_i$ values

As we know:  $b = X^{-1}Ga$

We first require to inverse the phenotypic variance-covariance matrix. Now the inverse ( $X^{-1}$ ) is multiplied with the genotypic



variance and covariance matrix, i.e.  $G$ . The resultant matrix i.e.  $X'G$  is multiplied by a vector  $a_i$  to get  $b_i$  values.

(iv) *The selection index and selection criterion*

The mathematical description of the function (I) is known as selection index.

$$I = b_1 p_1 + b_2 p_2 + \dots + b_n p_n$$

Using this function, the selection criterion or index value for each individual is determined as follows:

Genotype	Characters	b	Selection criterion
	$X1 \quad X2 \quad X3 \quad X4$		
$p1$	$X11 \quad X21 \quad X31 \quad X41$	$b_1$	$SC1$
$p2$	$X12 \quad X22 \quad X32 \quad X42$	$b_2$	$SC2$
$p3$	$X13 \quad X23 \quad X33 \quad X43$	$b_3$	$SC3$
$p4$	$X14 \quad X24 \quad X34 \quad X44$	$b_4$	$SC4$
$\dots$	$\dots \quad \dots \quad \dots \quad \dots$	$\dots$	$\dots$
$pn$	$X1n \quad X2n \quad X3n \quad X4n$	$b_n$	$SCn$

### 3. Results and Discussion

Analysis of variance for all the characters were carried out and it was found that mean squares for all the genotypes studied were highly significant under normal, low-N and ESM conditions, clearly indicating the existence of genetic variability in the genotypes. In normal conditions, among parents,  $L_5$  ranked first and  $L_9$  ranked last among lines and in case of testers  $T_1$  ranked first and  $T_4$  ranked last. In low-N conditions,  $L_5$  performed better followed by  $L_2$  and the least rank was shown by  $L_9$  in case of lines and in testers  $T_3$  ranked highest while  $T_4$  ranked last as indicated in Table 2 (Table 2: Selection criterion and ranking of parents and crosses in maize for four traits under normal, high N and low-N conditions). Among the crosses,  $L_{11} \times T_4$  ranked highest followed by  $L_8 \times T_1$  and  $L_{11} \times T_3$  and the least performing cross for this particular selection index was  $L_5 \times T_4$ . In ESM conditions,  $L_6$  performed better followed by  $L_5$  and the least rank was shown by  $L_9$  in case of lines as indicated in Table 3 (Table 3: Selection

Table 2 : Selection criterion and ranking of parents and crosses in maize for four traits under normal (high N) and low-N conditions

Variable / character	$\beta_i$ value				Variable / character	$\beta_i$ value			
	normal		low-N			normal		low-N	
	score	rank	score	rank		score	rank	score	rank
L <sub>1</sub>	8849.18	62	6622.45	61	L <sub>5</sub> T <sub>1</sub>	20494.14	40	20491.50	27
L <sub>2</sub>	11902.97	59	16698.69	47	L <sub>5</sub> T <sub>2</sub>	24227.27	15	24200.58	6
L <sub>3</sub>	7327.56	63	7102.92	60	L <sub>5</sub> T <sub>3</sub>	25050.14	10	23955.92	10
L <sub>4</sub>	13421.64	56	8809.45	57	L <sub>5</sub> T <sub>4</sub>	14619.18	55	10470.03	55
L <sub>5</sub>	20932.03	36	16728.90	46	L <sub>6</sub> T <sub>1</sub>	19114.48	46	16503.29	48
L <sub>6</sub>	16314.75	52	15150.55	51	L <sub>6</sub> T <sub>2</sub>	26356.79	6	24581.19	5
L <sub>7</sub>	12749.13	57	6551.71	62	L <sub>6</sub> T <sub>3</sub>	24695.18	12	24184.04	7
L <sub>8</sub>	9338.65	60	6062.84	63	L <sub>6</sub> T <sub>4</sub>	23211.62	24	22421.66	17
L <sub>9</sub>	5047.73	64	1275.16	64	L <sub>7</sub> T <sub>1</sub>	20532.68	39	13517.20	52
L <sub>10</sub>	15869.99	53	12327.51	53	L <sub>7</sub> T <sub>2</sub>	26785.70	4	26442.40	4
L <sub>11</sub>	16702.96	51	7832.50	59	L <sub>7</sub> T <sub>3</sub>	22729.83	26	18789.70	35
L <sub>12</sub>	12087.18	58	8505.43	58	L <sub>7</sub> T <sub>4</sub>	25187.81	9	21260.09	23
T <sub>1</sub>	22535.39	27	19543.95	32	L <sub>8</sub> T <sub>1</sub>	26564.27	5	26868.71	2
T <sub>2</sub>	18819.55	47	17558.67	42	L <sub>8</sub> T <sub>2</sub>	20884.16	37	17813.30	40
T <sub>3</sub>	21065.57	33	20886.25	26	L <sub>8</sub> T <sub>3</sub>	22244.48	29	19923.14	45
T <sub>4</sub>	9147.12	61	8899.67	56	L <sub>8</sub> T <sub>4</sub>	23030.11	25	22386.47	18
L <sub>1</sub> T <sub>1</sub>	21696.24	30	20309.81	28	L <sub>9</sub> T <sub>1</sub>	24492.03	13	18147.27	38
L <sub>1</sub> T <sub>2</sub>	24417	14	21888.40	20	L <sub>9</sub> T <sub>2</sub>	23692.94	20	23333.35	14
L <sub>1</sub> T <sub>3</sub>	28386.52	3	24063.96	9	L <sub>9</sub> T <sub>3</sub>	20991.55	35	19993.80	30
L <sub>1</sub> T <sub>4</sub>	19472.11	44	19912.76	31	L <sub>9</sub> T <sub>4</sub>	19815.38	43	18561.58	37
L <sub>2</sub> T <sub>1</sub>	18017.78	50	17340.87	43	L <sub>10</sub> T <sub>1</sub>	23503.16	22	23416.53	13
L <sub>2</sub> T <sub>2</sub>	20142.32	42	15982.90	49	L <sub>10</sub> T <sub>2</sub>	21366.22	31	20931.92	25

Continue...



Variable / character	$\beta_i$ value				Variable / character	$\beta_i$ value			
	normal		low-N			normal		low-N	
	score	rank	score	rank		score	rank	score	rank
L <sub>2</sub> T <sub>3</sub>	20353.96	41	19242.79	33	L <sub>10</sub> T <sub>3</sub>	24699.01	11	24130.03	8
L <sub>2</sub> T <sub>4</sub>	24191.44	16	23647.40	12	L <sub>10</sub> T <sub>4</sub>	25357.74	8	21583.88	21
L <sub>3</sub> T <sub>1</sub>	22523.35	28	21503.89	22	L <sub>11</sub> T <sub>1</sub>	23300.98	23	18919.04	34
L <sub>3</sub> T <sub>2</sub>	19214.70	45	18767.49	36	L <sub>11</sub> T <sub>2</sub>	21270.54	32	17976.47	39
L <sub>3</sub> T <sub>3</sub>	18447.71	48	15158.63	50	L <sub>11</sub> T <sub>3</sub>	28683.45	2	26610.25	3
L <sub>3</sub> T <sub>4</sub>	23682.85	21	22918.61	15	L <sub>11</sub> T <sub>4</sub>	32176.19	1	27387.26	1
L <sub>4</sub> T <sub>1</sub>	25470.92	7	22797.26	16	L <sub>12</sub> T <sub>1</sub>	18179.58	49	17696.19	41
L <sub>4</sub> T <sub>2</sub>	20737.84	38	21213.44	24	L <sub>12</sub> T <sub>2</sub>	14797.99	54	11219.38	54
L <sub>4</sub> T <sub>3</sub>	21034.69	34	17170.95	44	L <sub>12</sub> T <sub>3</sub>	24028.84	19	20309.17	29
L <sub>4</sub> T <sub>4</sub>	24088.85	17	23797.00	11	L <sub>12</sub> T <sub>4</sub>	24057.94	18	22213.98	19

Table 3: Selection criterion and ranking of parents and crosses in maize for ten traits under normal and ESM conditions

Variable / character	$\beta_i$ value				Variable / character	$\beta_i$ value			
	normal		low-N*			normal		low-N*	
	score	rank	score	rank		score	rank	score	rank
L <sub>1</sub>	3346.19	62	1367.91	60	L <sub>5</sub> T <sub>1</sub>	7141.33	40	829.601	63
L <sub>2</sub>	3983.94	59	3389.20	52	L <sub>5</sub> T <sub>2</sub>	8583.03	16	5853.49	19
L <sub>3</sub>	2543.73	63	1342.96	61	L <sub>5</sub> T <sub>3</sub>	8698.80	9	6110.88	12
L <sub>4</sub>	4763.65	56	3924.67	42	L <sub>5</sub> T <sub>4</sub>	5255.76	55	1757.92	59
L <sub>5</sub>	7126.36	41	4229.90	40	L <sub>6</sub> T <sub>1</sub>	6754.48	46	4772.63	31
L <sub>6</sub>	5922.43	51	4469.16	35	L <sub>6</sub> T <sub>2</sub>	9009.48	6	6742.36	5
L <sub>7</sub>	4699.29	57	2627.95	55	L <sub>6</sub> T <sub>3</sub>	8291.04	17	6480.97	9
L <sub>8</sub>	3513.99	60	2173.47	56	L <sub>6</sub> T <sub>4</sub>	8157.46	22	7763.04	1
L <sub>9</sub>	2108.32	64	540.42	64	L <sub>7</sub> T <sub>1</sub>	7856.38	29	3614.23	49
L <sub>10</sub>	5713.92	53	3965.73	41	L <sub>7</sub> T <sub>2</sub>	9074.43	4	7248.57	2
L <sub>11</sub>	5835.42	52	2112.85	57	L <sub>7</sub> T <sub>3</sub>	7875.79	28	4370.82	37
L <sub>12</sub>	4282.29	58	1903.35	58	L <sub>7</sub> T <sub>4</sub>	8526.91	10	6060.50	13
T <sub>1</sub>	8086.28	26	6546.67	8	L <sub>8</sub> T <sub>1</sub>	9021.61	5	6026.82	16
T <sub>2</sub>	6403.47	50	5236.52	27	L <sub>8</sub> T <sub>2</sub>	7315.56	38	5799.06	20
T <sub>3</sub>	7321.12	37	6971.58	4	L <sub>8</sub> T <sub>3</sub>	7749.57	30	3731.47	46
T <sub>4</sub>	3489.73	61	6117.76	11	L <sub>8</sub> T <sub>4</sub>	8093.24	24	5241.24	26
L <sub>1</sub> T <sub>1</sub>	7570.80	31	5331.36	25	L <sub>9</sub> T <sub>1</sub>	8417.63	13	6038.15	15
L <sub>1</sub> T <sub>2</sub>	8477.53	12	5402.93	22	L <sub>9</sub> T <sub>2</sub>	8164.11	21	6203.02	10
L <sub>1</sub> T <sub>3</sub>	9710.50	3	5396.58	23	L <sub>9</sub> T <sub>3</sub>	7416.92	35	4270.82	39
L <sub>1</sub> T <sub>4</sub>	6894.50	43	4976.63	29	L <sub>9</sub> T <sub>4</sub>	6876.76	44	4283.38	38
L <sub>2</sub> T <sub>1</sub>	6447.36	49	3472.04	51	L <sub>10</sub> T <sub>1</sub>	8213.88	19	6614.17	7
L <sub>2</sub> T <sub>2</sub>	7462.63	33	3327.55	53	L <sub>10</sub> T <sub>2</sub>	7337.47	36	3702.33	47
L <sub>2</sub> T <sub>3</sub>	7091.92	42	3478.18	50	L <sub>10</sub> T <sub>3</sub>	8512.62	11	5031.58	28
L <sub>2</sub> T <sub>4</sub>	8415.32	14	4475	34	L <sub>10</sub> T <sub>4</sub>	8791.91	7	6005.61	17
L <sub>3</sub> T <sub>1</sub>	7934.82	27	4613.34	32	L <sub>11</sub> T <sub>1</sub>	8176.15	20	6052.27	13
L <sub>3</sub> T <sub>2</sub>	6816.85	45	5357.07	24	L <sub>11</sub> T <sub>2</sub>	7485.62	32	3828.57	44

continue...





Variable / character	$\beta_i$ value				Variable / character	$\beta_i$ value			
	normal		low-N*			normal		low-N*	
	score	rank	score	rank		score	rank	score	rank
$L_3T_3$	6598.16	47	2992.24	54	$L_{11}T_3$	9834.54	2	5720.87	21
$L_3T_4$	8153.21	23	4475.00	33	$L_{11}T_4$	10735.98	1	6005.60	18
$L_4T_1$	8706.99	8	7071.64	3	$L_{12}T_1$	6498.81	48	3911.99	43
$L_4T_2$	7289.06	39	3688.21	48	$L_{12}T_2$	5444.21	54	1042.60	62
$L_4T_3$	7430.32	34	378.79	45	$L_{12}T_3$	8247.38	18	4461.88	36
$L_4T_4$	8389.48	15	4885.98	30	$L_{12}T_4$	8091.42	25	6632.28	6

\* This is low nitrogen application trial. In this trial, no nitrogen was applied to provide the abiotic stress of low soil fertility

criterion and ranking of parents and crosses in maize for ten traits under normal and ESM conditions). Among the crosses  $L_6 \times T_4$  ranked highest and the least performing cross for this particular selection index was  $L_5 \times T_1$ .

Economic weights assigned to various traits under low-N and ESM conditions in the present study have been presented in as indicated in Table 4 (Table 4: Simultaneous selection indices for normal and low-N trials in maize) and Table 5 (Table 5: Simultaneous selection indices for normal and ESM trials in maize), respectively.

Among the parents, none of the parents ranked high among all genotypes according to the assigned selection criteria in normal low-N and ESM conditions. Among the crosses aggregate score were higher in most of the cases but general trend was that ESM traits had the lower score values. Crosses  $L_{11} \times T_4$ ,  $L_8 \times T_1$ ,  $L_{11} \times T_3$ ,  $L_7 \times T_2$ , and  $L_6 \times T_2$  performed well in both normal and low-N conditions for the assigned selection criteria while crosses  $L_5 \times T_4$ , and  $L_{12} \times T_2$  performed poorly in low-N conditions. In ESM trials,  $L_6 \times T_4$ ,  $L_7 \times T_2$ , and  $L_4 \times T_1$  performed excellent while crosses  $L_5 \times T_1$ ,  $L_{12} \times T_2$ ,  $L_5 \times T_4$ , and  $L_3 \times T_3$  ranked very low in ESM trials. Now for ESM tolerance we can deduce from the results that crosses  $L_6 \times T_4$  and  $L_7 \times T_2$  are the best available crosses according to the given selection

Table 4: Simultaneous selection indices for normal and low-N trials in maize

Variable/character	Economic weight	$\beta_i$ value	
		Normal	ESM*
ASI	2	-21.83	34.04
Yield (kg ha <sup>-1</sup> )	5	4.87	4.90
Leaf senescence	1	-56.02	14.47
Number of ears plant <sup>-1</sup>	1	190.13	175.18

\* This is excess soil moisture trials. In this trial, waterlogging treatment was given at knee high growth stage for 6 days by keeping continuous submergence with an average depth of ponding of about 5 cm. After 6 days of ponding, water was drained out of the plots.

Table 5: Simultaneous selection indices for normal and ESM trials in maize

Variable/character	Economic weight	$\beta_i$ value	
		Normal	ESM
Days to 50% tasseling	1	4.331	49.222
Days to 50% silking	1	-13.752	-48.945
ASI	0.7	-23.832	50.842
Plant height	1	2.325	0.897
Ear height	1	4.011	2.782
Cob length	1	29.787	-3.385
Cob diameter	1	-91.623	11.315
100 kernel weight	1	22.626	1.437
Yield (kg ha <sup>-1</sup> )	1.6	1.512	1.556
Nodes bearing adventitious roots	1.3	-36.061	14.192

criteria. For low-N tolerance,  $L_{11} \times T_4$  and  $L_8 \times T_1$  were the best crosses. Similar studies with different objectives were conducted by Brim et al. (1959), Mulamba and Mock (1978), Kauffmann and Dudley (1979), Nawar et al. (1991) Banziger and Lafitte (1997), Edmeades et al. (1999); Modaressi et al. (2004); Kebede (2007); Lone (2006) in different stress conditions.

#### 4. Conclusion

It is desirable to develop maize cultivars with increased resistance to abiotic stress for sustainable production. For this, selection indices are considered as an aid for simultaneous selection of multiple traits which can help the breeder in spotting the desirable genotype/family of a crop species in a population improvement program. The study carried out illustrates how to construct selection indices with regard to low N and ESM tolerance. Thus, it has been recognized that most rapid improvement in the economic value is expected from selection applied simultaneously to all the characters which determine the economic value of a plant, provided



appropriate weights are assigned to each character.

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