

RESEARCH ARTICLE

Nitrogen use efficiency in bread wheat: Genetic variation and prospects for improvement

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Abstract

Nitrogen (N) is one of the primary macronutrients required for crop growth and yield. This nutrient is especially limiting wheat yields in the dry and low fertile agro-ecologies having low N in the root zone soil strata. Moreover, majority of farmers in India and South Asia are small to marginal with meagre capacity to invest in costly nitrogen fertilizers. Therefore, there is an immense need to identify lines that use nitrogen efficiently. A set of 50 diverse wheat genotypes consisting of indigenous germplasm lines (05), cultivars released for commercial cultivation (23) and selected elite lines from CIMMYT nurseries (22) were evaluated in an alpha-lattice design with two replications, a six-rowed plot of 2.5m length for 24 agro morphological, physiological and NUE related traits during two consecutive crop seasons in an N-depleted precision field under two different N levels of 50%-N50 (T1) and 100%-N100 (T2) of recommended N, i.e., 100 kg/ha. Analysis of variance revealed significant genetic variation among genotypes for all the traits studied. About 11.36% yield reduction was observed at reduced N levels. Significant correlations among NUE traits and yield component traits were observed which indicated pivotal role of N remobilization to the grain in enhancing yield levels. Among N-insensitive genotypes identified based on their yielding ability at low N levels, UASBW13356, UASBW13358, UASBW13354, UASBW13357 and KRL1-4 showed their inherent genotypic plasticity toward N application. The genotypes with more yield and high to moderate NUtE can be used as parents for the breeding of N efficient genotypes for marginal agro-ecologies. Low N tolerant genotypes identified from the current investigation may be further utilized in the identification of genomic regions responsible for NUE and its deployment in wheat breeding programs. The comprehensive data of 24 traits under different nitrogen levels for diverse genotypes from India and global sources (mainly CIMMYT) should be useful for supporting breeding for NUE and thus will be of great help for small and marginal farmers in India and South Asia.

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1. Introduction

Nitrogen (N) is an essential nutrient for optimizing plant growth and reproduction and therefore, applying N fertilizer is an essential practice to secure productivity in diverse cropping systems. Plants take N from both atmospheric air and soil minerals, but crop plants are inefficient in the acquisition and utilization of applied nitrogen. The efficient use of N by plants is determined by their ability to absorb naturally available N or applied N fertilizers. The current global nitrogen (N) demand stands at about 110.19 million metric tons with a projected annual increase of ~1.3% in the future [1] which equates to global N fertilizer consumption of more than 110 Mt per annum of which half of the total is being used for the production of major cereal crops i.e., maize, rice, and wheat [2].

Indian agriculture consumes over 17 million tons of N fertilizer per year. Studies have shown that more than 50% of applied N fertilizers are unused by crops [3, 4] and this unabsorbed N fertilizer in the soil is lost to the environment by leaching, denitrification, runoff, and atmospheric release through volatilization and becomes a major environmental concern [5–9]. Besides, an excess usage of N fertilizer not only decreases the efficiency of nutrient use but also affects the rate of economic returns per unit of chemical fertilizer applied. The effect of negative environmental and economic impacts could be reduced through better agronomic practices and by utilizing N efficient lines with improved nitrogen use efficiency (NUE) [10]. Hence, the development of more N-use efficient cultivars is an important research goal for enhancing future food production capabilities for greater agricultural sustainability [11].

Wheat (*Triticum aestivum* L.) is one of the staple food crops that ensures food and nutritional security for the world's poorest people living in different agro-ecological zones. In the Indian agriculture scenario, wheat is grown under diverse production conditions including low fertile and rainfall dependent central and peninsular areas often facing drought stress during different stages of crop growth. In such areas, farmers apply a high dose of N fertilizer for maximizing grain yield (GY), however excessive usage of N fertilizer leads to a low NUE of the crop along with various environmental hazards. Further, many farmers lack knowledge about the ideal dose of fertilizer to harvest the real yield potential. The majority (90%) of such farmers are small and marginal and do not have capacity to take risk of applying high inputs. Hence, yield improvement under low nitrogen input would be extremely beneficial for economic and environmentally sustainable cultivation of wheat crop.

Similar to grain yield, nitrogen use efficiency (NUE) is also a complex trait, which is associated with various morphological, physiological, molecular, and biochemical changes in plants throughout the life cycle. For a clear understanding of this complex nature, studies of various physiological traits and their close correlation with one or more economically important traits like GY are foremost critical. Nevertheless, it will help in selecting low N tolerant/high yielding lines at different N conditions [12]. In general, plant function is always associated with chlorophyll content, which directly indicates the N status of the leaf [13]. Leaf chlorophyll content and photosynthetic capacity are appropriate benchmarks for identifying high NUE (HNUE) genotypes under low N conditions during field trials [14, 15]. Leaf N status was usually measured by using a hand-held optical chlorophyll meter to monitor the leaf nitrogen status and chlorophyll content.

NUE has several definitions based on agronomic, genetic, or physiological studies [16, 17]. NUE is the efficiency of nitrogen recovery from applied fertilizer, or the N available to the crop, giving rise to 33% efficiency of the crop recovery [3, 17, 18] (Alternatively, it is often considered a productivity index and defined as the yield produced per unit of available N to the crop expressed in terms of kg yield per kg of available N [19, 20]. NUE included two components namely, N uptake efficiency (NUpE) and N utilization efficiency (NUtE) and is

mathematically the product of $NU_{pE} \times NU_{tE}$ [21–23]. NU_{pE} is the ability of the plant to take up N from the soil. It is estimated as the ratio of N taken up by the plant per unit of N available from the soil and applied fertilizer and is expressed as kg N (in the crop) per kg N (available). NU_{tE} is the ability to use N to produce grain yield which is estimated as the grain yield produced per unit of N taken up by the plant and is also expressed as kg (grain) per kg N [11, 24].

Studies on genotypic variations under low and recommended N for NUE traits in various crops spurred breeding activities to develop N use-efficient genotypes possessing high yield sustainability under low N conditions [11, 14, 25–32]. There have been continuous improvements in N use efficiency (NUE) of crops over the years along with increase in crop yield [33–36] and extensive utilization of the genetic pool through introgression of landraces and ancestral germplasm is expected to support breeding programs in driving future NUE-based outcomes [36, 37]. The molecular studies identified 333 genomic regions associated to 28 NUE related traits in winter wheat and NUE and its related traits were found to be polygenic in nature and highly influenced by the genetic backgrounds [38].

The high NUE/ N-insensitive genotypes (NIS-top grain yielders) give more or equal GY with minimal application of N fertilizer compared to recommended N fertilizer conditions [39]. The understanding of GY and its associated NUE traits is still lagging in wheat but few studies have explored genotypic variations for NUE at different N levels [40, 41]. Several studies in cereals have suggested an enhanced NUE through improved NU_{tE} under low N conditions [19, 21]. The nitrogen harvest index (NHI) is another useful measure of the efficiency of N use, which is the fraction of total N taken up by the crop that is partitioned to the grain. NHI is independent of yield and uptake efficiency, and a low yielding crop may have a high NHI [42].

The green revolution period in India had focus primarily on the responsiveness of wheat genotypes to high-input supply. In the present scenario of climate change, realizing consistent output (yield and quality) of the genotype under varying levels of nitrogen input is a major challenge to wheat improvement programme. Identifying promising genotypes for enhanced NUE in wheat and estimating its genetic variability can provide useful guidance to breeders [41]. Keeping the high yield when N supply is reduced and/or increasing yield when N supply is continuous may be two strategies for NUE enhancement. Nitrogen production costs, environmental pollution owing to nitrate leaching [43, 44], and greenhouse gas volatilization necessitate improvement of wheat nitrogen use efficiency at a reduced N supply. Simultaneously, GY is a complex trait controlled by a network of multiple traits and their associations. Hence, uncovering the genetic basis of GY, and other related NUE traits under low and high N conditions is a prerequisite to understanding the mechanism and for ideal selection of NUE associated traits. Nitrogen use efficient crop varieties can be a great choice for ensuring sustainability in farming systems and meeting future customer requirements, particularly in the face of changing environment caused by climate change. This study provides useful information for uncovering the physiological and genetic basis of GY and its related traits under low N which further facilitates the development of low N stress resilient wheat cultivars.

2. Material and methods

2.1 Selection of genotypes

The genotypes included in the study consisted of fifty bread wheat genotypes representing high, medium, and low NUE based on multi-location and multi-year evaluation. These genotypes represent indigenous elite lines (DTW2011-56, GW2013-540, MP1293, RAJ4248, UAS323), cultivars released for commercial cultivation (DBW14, DL153-2, GW322, HD2189, HD2733, HD2967, HI1500, HI1530, HPW251, K9107, Kalyansona, KRL1-4, KRL237, MP4010, NP846, PBW175, PBW343, RAJ1972, UAS304, WH147, WH542, WH1021,

WH1022) and selected elite lines from CIMMYT trials/nurseries (UASBW10227, UASBW10453, UASBW11948, UASBW12380, UASBW12758, UASBW12876, UASBW12877, UASBW12878, UASBW13354, UASBW13355, UASBW13356, UASBW13357, UASBW13358, UASBW13359, UASBW13360, UASBW13361, UASBW13362, UASBW13363, UASBW13364, UASBW13365, UASBW13366, UASBW13367). The investigation was carried out for two years during *rabi* 2016–17 and 2017–18 at All India Coordinated Wheat Improvement Project, Main Agricultural Research Station, University of Agricultural Sciences, Dharwad located at 15.48°N latitude and 74.98°E longitude and falls in agro-climatic Zone 8 of Karnataka state in Peninsular India.

2.2 Experimental design, treatment, and crop management

The experiment was laid out in an Alpha-lattice design at two treatment levels of nitrogen (T₁ and T₂) with two replications. The T₁ consisted of soil N + 50% of the recommended nitrogen dose (RND) of 100 kg N ha⁻¹ which is equivalent to 50kg/ha⁻¹ whereas T₂ consisted of soil N +RND of 100 kg N ha⁻¹. The genotypes were randomly allotted to plots in each treatment in each replication and grown in 6 row plots of 2.5 meter length with 20 cm row spacing. The nitrogen fertilizer dose was calculated as per the treatments and was applied to each plot. The soil samples from the experimental sites were collected from 0 to 30 cm soil depth before the start of the field experiment and after harvest of the crop and N was analyzed using standard procedure. All other recommended agronomic package of practices was adopted to raise a good crop.

2.3 Observations

Data on 24 agro-morphological, physiological, and NUE related traits were recorded from phase 41 to phase 92 of the Zadok's scale [45] for wheat plants in the four central rows of each plot. The agro-morphological traits included, days to heading, days to flowering, plant height (cm), number of productive tiller per meter row length, spike length (cm), awn length (cm), spikelet number per spike, grains number per spike, 1000-grains weight (g), grain yield (q ha⁻¹) biomass yield (q ha⁻¹) and harvest index (%); Physiological traits namely, chlorophyll content (SPAD) and normalized difference vegetation index (NDVI) were recorded at booting (SPAD 1, NDVI-1), anthesis (SPAD-2, NDVI-2) and grain filling (SPAD-3, NDVI-3) stages. The chlorophyll content was measured by a Chlorophyll meter or Spad-meter and NDVI was measured by the Trimble GreenSeeker handheld crop sensor using standard procedures. Nitrogen use efficiency (NUE) and its related parameters like nitrogen uptake efficiency (NUpE), nitrogen utilization efficiency (NUtE), nitrogen harvest index (NHI), and total nitrogen uptake (TNUp) were worked out according to [19, 46]. NHI was calculated as [Grain N / (Grain N + straw N)] *100 and expressed in percentage. The grain N (%) and straw N (%) contents were analyzed by the Micro-Kjeldahl digestion method as per the FAO guidelines for plant nutrient analysis [47]. TNUp (kg N ha⁻¹) was determined as the sum of nitrogen in straw and grain at harvest. NUpE was calculated as total N uptake/crop N supply where crop N supply includes fertilizer N + soil mineral N at planting. NUtE was calculated as grain yield/ total N uptake. NUE was measured as Grain yield/ total nitrogen supply from the soil and applied fertilizer. The values for NUE, NUtE and NUpE were expressed in kg kg⁻¹.

2.4 Statistical analysis

Before pooling the data across environments, the ANOVA assumption was tested for its homogeneity using the Bartlett test [48]. The ratio of the highest error means square and the smallest error mean square value was compared with the F table and it showed non-significant

results for all the traits hence, two years of data across environments were pooled. From pooled data, the analysis of variance (ANOVA) was done [49]. Critical difference was estimated to see significant differences among the genotypes for various studied traits. The genotypic and phenotypic coefficients of variation [50] and heritability [51, 52] were worked out and the traits were categorized as low, medium and high based on genetic parameters [53, 54]. The correlation among all the traits studied under both the nitrogen levels was worked out using Karl Pearson's simple correlation coefficient method.

For understanding the effect of nitrogen doses on different genotypes concerning yield as well as NUE, various indexes namely, stress susceptibility index [55], percentage reduction [56] and yield stability index [57] were also estimated to identify promising genotypes. As grain yield and NUE measurement are reflective of each other in totality, grain yield was used to estimate these indices.

3. Results

3.1 Trait specific variability among genotypes

The computed mean sum of squares for various traits namely, chlorophyll content at booting stage (CC-1), anthesis stage (CC-2), grain filling stage (CC-3), NDVI at booting stage (NDVI-1), anthesis stage (NDVI-2), grain filling stage (NDVI-3), days to heading (DH), days to maturity (DM), plant height (PH), number of productive tiller per meter row length (TPM), spike length (SL), awn length (AL), number of spikelet's per spike (SPS), number of grains per spike (GPS), thousand grains weight (TGW), biomass yield (BMY), harvest index (HI), grain protein content (GPC), nitrogen harvest index (NHI), total nitrogen uptake (TNU_p), nitrogen uptake efficiency (NU_pE), nitrogen utilization efficiency (NU_tE) and nitrogen use efficiency (NUE) showed that genotypes differed significantly for all the traits which indicated the existence of variation among genotypes for various traits studied. A wide range was observed for all the traits evaluated under T1 (soil N + 50 kg N ha⁻¹) and T2 (Soil N + 100 kg N ha⁻¹) as shown in Table 1. In general, the mean values for all the traits except HI, NHI, NU_tE and NUE were higher at full/recommended fertilizer dose as compared to reduced nitrogen dose. A similar trend was also observed for trait specific maximum values where all the traits except NDVI-1, SL, HI and NUE showed higher values under T2 as compared to T1 (S1 Table). DMRT encompasses the calculation of numerical thresholds, facilitating the categorization of the disparity between any two treatment means as either statistically significant or not. Treatment means sharing a common letter are deemed non-significantly different at the 5% level of significance. Among the traits examined, NDVI1, AL, HI, NHI, NU_pE, and NUE, did not exhibit significant differences at the 5% level of significance (Table 1).

3.1.1. Yield and agro-morphological traits. The present investigation included 12 agro morphological traits including grain yield. A wide range was observed for these yield and agro-morphological traits under both the nitrogen levels. In T1 (N50), the average GY of the tested genotypes was 28.65 q/ha and ranged from 16.87 to 38.91q/ha whereas in T2 (N100), GY ranged from 20.40 q/ha to 40.36q/ha with a mean of 32.32 q/ha. Nitrogen in limited condition (N50) resulted in 11.36% reduction in GY compared to N100. A higher mean BMY of 105.9 q/ha with a range of 79.1 to 132.4q/ha was recorded in T2 as compared to 90.8 q/ha mean BMY with range of 73.0 to 108.7q/ha under the T1 condition that showed 14.25% reduction. Interestingly, 1.3% lower HI was observed in T2 (31.5%) as compared to T1 (31.9%) condition. TPM, GPS and TGW are major yield component traits that significantly contribute to higher yield realization. The results showed a higher mean of 97.3 TPM, 50.1 GPS and 37.0g TGW in the T2 condition as compared to 86.0 TPM, 47.0 GPS and 34.1g TGW in the T1

Table 1. Pooled ANOVA and genetic parameters for morpho-physiological, yield, yield attributes, NUE and NUE related traits in 50 genotypes of bread wheat under T1 (soil N + 50 kg Nha-1) and T2 (Soil N + 100 kg Nha-1).

Traits	F statistics (Genotypes at 49 df)		Mean		Range		PCV		GCV		h ² bs (%)		Difference (% of HN)	CD @ 5% T1	CD @ 5% T2
	T1	T2	T1	T2	T1	T2	T1	T2	T1	T2	T1	T2			
CC 1	2.54**	5.22**	47.70b	51.63a	42.95–55.93	45.70–59.65	6.7	6.45	4.42	5.32	43.49	67.82	7.61***	4.89	3.84
CC 2	3.70**	6.82**	49.24b	57.68a	42.00–56.45	51.70–67.15	5.6	6.25	4.25	5.39	57.47	74.43	14.63***	3.92	3.66
CC 3	5.46**	5.00**	47.87b	50.96a	40.50–56.00	45.70–57.83	7.03	6.12	5.84	5.00	69.05	66.71	6.06***	3.82	3.62
NDVI 1	4.14**	3.41**	0.54a	0.53a	0.47–0.61	0.48–0.60	5.1	5.64	3.99	4.17	61.06	54.69	-1.89 ^{ns}	0.03	0.04
NDVI 2	3.48**	4.67**	0.57b	0.60a	0.50–0.66	0.53–0.68	6.33	5.70	4.71	4.58	55.37	64.73	5.00***	0.05	0.04
NDVI 3	3.83**	2.86**	0.55b	0.56a	0.48–0.62	0.51–0.65	5.43	5.49	4.16	3.81	58.61	48.15	1.79*	0.04	0.04
DH	14.20**	22.97**	59.42b	62.25a	50.75–65.25	55.25–68.00	5.52	4.66	5.14	4.46	86.84	91.66	4.55***	2.39	1.68
DM	13.93**	31.48**	94.73b	99.99a	81.75–104.50	91.75–110.50	5.25	3.84	4.88	3.72	86.6	93.84	5.26***	3.76	1.92
PH	2.21**	2.28**	74.41b	80.02a	61.48–90.70	63.98–93.98	9.14	9.88	5.62	6.17	37.77	39.01	7.01***	10.89	12.51
TPM	4.50**	2.76**	85.95b	97.32a	60.00–111.25	79.50–121.50	15.68	11.53	12.51	7.89	63.63	46.76	11.68***	16.39	16.46
SL	2.03**	3.03**	7.95b	8.30a	6.84–10.07	6.92–9.96	10.56	9.30	6.16	6.61	34.05	50.4	4.22*	1.44	1.10
AL	2.28**	9.41**	6.27a	6.72a	5.43–8.29	4.72–8.50	12.48	14.15	7.8	12.72	39.02	80.78	6.70 ^{ns}	1.23	0.84
SPS	4.82**	4.37**	16.44b	17.8a	13.20–19.80	15.25–20.30	9.41	5.91	7.63	4.68	65.65	62.77	7.64***	1.83	1.29
GPS	3.68**	4.56**	47.00b	50.1a	37.60–57.50	42.75–57.75	9.34	6.09	7.07	4.87	57.3	64.02	6.19***	5.98	3.67
TGW	3.71**	11.53**	34.10b	36.99a	29.13–43.05	31.58–46.40	10.18	8.23	7.72	7.55	57.53	84.03	7.81***	4.59	2.46
GY	5.62**	7.09**	28.65b	32.32a	16.87–38.91	20.40–40.36	16.22	11.4	13.55	9.89	69.79	75.27	11.36***	5.18	3.94
BMY	4.29**	4.02**	90.81b	105.90a	72.95–108.71	79.13–132.38	11.22	15.6	8.85	12.10	62.22	60.13	14.25***	12.59	21.51
HI	11.14**	8.08**	31.89a	31.49a	20.77–52.12	16.88–50.01	20.47	21.61	18.71	19.08	83.53	77.98	-1.27 ^{ns}	5.38	6.70
GPC	2.66**	2.74**	12.06b	12.56a	10.56–13.29	10.85–13.70	7.49	6.81	5.04	4.65	45.35	46.5	3.98**	1.34	1.26
NHI	6.84**	2.22**	75.83a	73.79a	70.03–78.60	66.09–78.64	2.97	4.67	2.57	2.87	74.52	37.81	-2.76***	2.29	5.80
TNU _p	6.02**	2.57**	94.59b	116.24a	71.86–113.11	88.97–134.78	10.81	9.10	9.15	6.04	71.54	43.97	18.63***	10.96	18.50
NU _p E	6.23**	2.57**	0.52a	0.54a	0.40–0.63	0.42–0.63	14.54	9.67	12.36	6.42	72.34	43.98	3.70*	0.06	0.09
NU _t E	4.20**	5.35**	30.16a	28.08b	22.11–35.15	17.95–35.17	10.63	12.24	8.34	10.13	61.54	68.5	-7.41**	4.03	4.40
NUE	5.83**	8.97**	15.85a	15.16a	9.36–21.54	9.57–18.95	16.54	12.02	13.91	10.75	70.73	79.95	-4.55 ^{ns}	2.85	1.70

PCV, GCV- Phenotypic and genotypic coefficients of variation, h²bs- heritability broad sense, CC1, CC2, CC3 & NDVI1, NDVI2, NDVI3- Chlorophyll contents and NDVI at booting, anthesis and grain filling stages, respectively, DH- Days to heading, DM- Days to maturity, PH- Plant height, TPM- Number of productive tiller per meter, SL- Spike length, AL-Awn length, SPS-Number of spikelet's per spike, GPS-Number of grains per spike, TGW-Thousand grain weight, GY: Grain yield, BMY- Biomass yield, HI- Harvest index, GPC-Grain protein content, NHI- Nitrogen harvest index, TNU_p-Total nitrogen uptake, NU_pE-Nitrogen uptake efficiency, NU_tE-Nitrogen utilization efficiency, NUE-Nitrogen use efficiency

Note: ns: not significant.

*, **, and ***: significant at the level of probability p < 0.05, p < 0.01, and p < 0.001, respectively

Note: Means with the same letter are not significantly different at 5% level

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condition indicating 11.7, 6.19, and 7.81 percent reductions for TPM, GPS and TGW respectively under T1 condition.

The lower mean values for other morphological traits DH, DM, PH, SL, AL and SPS were observed in T1 with 4.6, 5.3, 7.0, 4.2, 6.7 and 7.6 percent reduction, respectively as compared to mean values for these traits in the T2 condition.

3.1.2 Physiological parameters. Chlorophyll content and NDVI values are two physiological parameters that directly indicate the phenotypic nitrogen levels in crop plants based on the greenness of leaves. Chlorophyll content (CC-1, CC-2, CC-3) and NDVI values (NDVI-1, NDVI-2, NDVI-3) were measured at booting, anthesis, and grain filling stages, respectively which also showed a wide range in both conditions. Higher mean values of 51.6, 57.7 and 51.0

chlorophyll content at all three stages were recorded in T2 as compared to 47.7, 49.2 and 47.9 in the T1 condition with a reduction of 7.6%, 14.6% and 6.1%, respectively. Higher chlorophyll content was observed at the anthesis stage in both T1 and T2 conditions. A similar trend was observed for NDVI at anthesis and grain filling stages with 5.0% and 1.8% reduction in the T1 but NDVI at booting stage was a little higher in T1 as compared to T2.

3.1.3 NUE associated traits. Nitrogen based six traits were estimated in the present investigation (Table 1). The NUE in T1 ranged from 9.4 to 21.5 kg grain/kg N (Avg: 15.9 kg grain/kg N) whereas it was with a range of 9.6 to 19.0 kg grain/kg N in T2 (Avg: 15.2 kg grain/kg N) which indicated 4.6% reduction as compared to T1. NUE is mainly dependent on TNU_p, NupE and NutE. The results indicated an increase of TNU_p with increased application of N fertilizer. It ranged from 71.9 to 113.1 kg N/ha (Avg: 94.6 kg grain/kg N) in T1 and decreased by 18.6% as compared with T2 (N100) where it ranged from 89.0 to 134.8 kg N/ha (Avg: 116.2 kg N/ha). Similarly, NupE was with a range of 0.40 to 0.63 (0.52) and 0.42 to 0.63 (0.54) in T1 and T2, respectively indicating 3.7% reduction in T1. A large variation was observed for NUtE of genotypes which ranged from 22.1 to 35.2 kg grain/kg N (Avg: 30.2 kg grain/kg N) under T1 and 18.0 to 35.2 (Avg: 28.1) in T2. The overall mean showed 7.4% reduction in NUtE in T2 due to poor performance of some of the genotypes which was reflected in lower values of range in T2. The NHI varied from 70.0 to 78.6% (Avg: 75.8%) in the T1 condition, whereas it ranged from 66.1 to 78.6% (Avg: 73.8%) in T2 indicating 2.76% reduction as compared to T1. Grain protein content (GPC) is an important trait depending on N uptake and its utilization and in this study, it was approx. 4.0% higher in T2 as compared to the mean of T1. The GPC ranged from 10.9–13.7% (Avg: 12.6%) in T2 as compared to T1 with a range of 10.6 to 13.3% (Avg: 12.1%).

3.2 Genetic parameters

Genetic parameters namely phenotypic (PCV) and genotypic (GCV) coefficients of variation, heritability in the broad sense (h^2_{bs}) were worked out as shown in Table 1.

3.2.1 Coefficients of variation. In this study, phenotypic (PCV) and genotypic (GCV) coefficients of variation were estimated in T1 and T2 conditions. In general, higher PCV values were observed as compared to GCV values at both the nitrogen levels. The PCV values ranged from 2.97 (NHI) to 20.47 (HI) under T1 whereas it ranged from 3.84 (DM) to 21.61 (HI) under T2 condition. Harvest index showed a high PCV value, both for T1 (20.47) and T2 (21.61), respectively. The traits TPM (15.68, 11.53), AL (12.48, 14.15), GY (16.22, 11.4), BMY (11.22, 15.6), NUtE (10.63, 12.24) and NUE (16.54, 12.02) have moderate PCV values (10–20) whereas other traits showed low PCV values (<10) under T1 and T2, respectively. Similarly, GCV ranged from 2.57 and 2.81 for NHI to 18.71 and 21.08 for HI under T1 and T2, respectively. Moderate GCV values for HI (18.71, 19.08) and NUE (13.91, 10.75) were observed under both T1 and T2. Moderate GCV values were also observed for TPM (12.51), GY (13.55), and NU_pE (12.36) under T1 and for AL (12.72), BMY (12.1), and NUtE (10.13) under T2. Other traits showed low GCV values in T1 and T2.

3.2.2 Heritability. The heritability values in broad sense ranged from 34.05% for SL to 86.84% for DH in T1 and 37.81% for NHI to 93.84% for DM in T2 (Table 1). At reduced nitrogen level (T1), the highest heritability was observed for DH (86.84) followed by DM (86.60) and HI (83.53) whereas it was highest for DM (93.84) followed by DH (91.66), TGW (84.03), and AL (80.78) at full nitrogen dose (T2). Additionally, a high heritability (>60%) was observed for CC-3 (69.1, 66.7), SPS (65.7, 62.8), GY (69.8, 75.3), BMY (62.2, 60.1), NUE (70.7, 80.0) and NUtE (61.5, 68.5) at both T1 and T2 conditions, respectively. Traits NHI (74.5), TNU_p (71.5), NU_pE (72.3), TPM (63.6) and NDVI-1 (61.1) showed a high heritability under

T1 only whereas for T2, CC-2 (74.4), CC-1 (67.8), NDVI-2 (64.7) and GPS (64.0) showed a high heritability. The plant height showed a low heritability under both T1 (37.8), and T2 (39.0) conditions.

3.3 Phenotypic correlation

Character association was studied among genotypes using Karl Pearson's simple correlation coefficient method to identify the interrelation among traits for both the T1 and T2. Nitrogen use efficiency was measured as grain yield divided by available N soil (soil N + fertilizer N) due to which the correlation of nitrogen use efficiency and grain yield at both levels of nitrogen treatments showed similar results. The association between grain yield and NUE under the two nitrogen levels is visually depicted in [S1 Fig](#). Therefore, the correlation values with NUE were estimated as shown in [Table 2](#). The results indicated a positive and highly significant correlated response of nitrogen use efficiency with DM, HI, TGW, GPS, SPS, CC-1, CC-2, CC-3, NDVI-1, NDVI-2, NDVI-3 TNU_p, NU_pE and NU_tE under both T1 and T2 conditions. Similar correlations of NUE with TPM, SL, DH, GPC, PH, TPM, SL and BMY were also observed either in T1 or T2 condition. All the six physiological traits namely CC-1, CC-2, CC-3, NDVI-1, NDVI-2 and NDVI-3 showed a significantly positive correlation with each other. Among NUE related traits, significant and positive correlation of GPC with NHI, TNU_p, and NU_pE and of TNU_p with NU_pE and NU_tE was observed under both T1 and T2 conditions. For agro-morphological traits, NUE/grain yield showed highly positive correlations with DM, SPS, GPS, TGW and HI under both nitrogen conditions. Traits like SPS, GPS and TGW have a highly significant and positive correlation with each other under both conditions. DM and DH have highly significant and positive associations in T1 as well as T2 conditions. Besides these, highly significant and positive associations of SL with DM, TPM, SPS, GPS, TGW, HI; TPM with SPS, GPS, HI; HI with SPS, GPS, TGW and DH with PH were observed in T1 condition whereas the similar correlation of DH and DM with SPS and HI were observed in the T2 condition. A highly significant but negative association of HI with BMY was also observed under both nitrogen levels.

The NUE related traits showed a highly significant and positive correlations with yield and its component traits among which GPC with TGW & HI; BMY with TNU_p & NU_pE and NU_tE with SPS, GPS, TGW & HI are prominent ones. All the physiological traits namely chlorophyll content and NDVI at booting, anthesis and grain filling stages showed a highly significant and positive correlations with NUE related traits GPC, TNU_p, and NU_pE under both conditions. NU_tE has a similar association with chlorophyll content at all three stages under the T1 condition and with chlorophyll content and NDVI at the anthesis stage in T2 condition. Considering yield and physiological traits, SPS, GPS, TGW, and HI showed a highly significant and positive associations with all six physiological parameters under both T1 and T2 conditions. Under both conditions, BMY showed significant negative associations with chlorophyll content at all three stages. Significantly positive associations of TPM with all the physiological parameters and SL with chlorophyll content (CC-1, CC-2, CC-3) were observed in T1 only. In T2, significant and positive associations of DH and DM with NDVI and AL with chlorophyll content at different stages were observed. BLUP analysis revealed a positive and weak correlation of GY with GPC ([S2A Fig](#)) and NU_pE ([S2B Fig](#)) under both N levels ($R^2 = 0.09$ at T1 and 0.24 at T2 for GPC and $R^2 = 0.66$ in T1 and T2 ($R^2 = 0.13$ in T2 for NU_pE). Furthermore, a moderately positive correlation was observed between GY and NU_tE at T1 ($R^2 = 0.64$), while a weaker but positive relationship was observed at T2 conditions ($R^2 = 0.57$) ([S2C Fig](#)). Remarkably, the NUE component trait exhibited a perfect positive relationship with grain yield at both T1 and T2 ($R^2 = 1.00$) ([S2D Fig](#)).

Table 2. Phenotypic correlations among different morpho-physiological, yield, yield attributes, NUE and NUE related traits in 50 genotypes of bread wheat under T1 (soil N + 50 kg Nha⁻¹) and T2 (Soil N + 100 kg Nha⁻¹) conditions.

	T1	T2	CC1	CC2	CC3	NDVI1	NDVI2	NDVI3	DH	DM	PH	TPM	SL	AL	SPS	GPS	TGW	BY	HI	GPC	NHI	TNUp	NUPE	NUE	
CC1			0.97**		0.98**	0.33**	0.43**	0.39**	0.14	0.07	-0.02	0.19	0.16	0.22*	0.21*	0.23*	0.54**	-0.23*	0.44**	0.38**	-0.06	0.25*	0.24*	0.21*	0.40**
CC2	0.90**			0.97**	0.97**	0.36**	0.46**	0.42**	0.16	0.10	0.02	0.15	0.17	0.21*	0.25*	0.27**	0.57**	-0.24*	0.47**	0.42**	-0.06	0.25*	0.24*	0.26**	0.45**
CC3	0.82**	0.87**			0.45**	0.35**	0.40**	0.15	0.07	0.02	0.02	0.21*	0.15	0.24*	0.22*	0.24*	0.52**	-0.24*	0.44**	0.43*	-0.03	0.25*	0.24*	0.21*	0.41**
NDVI1	0.38**	0.38**	0.45**			0.83**	0.88**	0.88**	0.35**	0.34**	-0.08	-0.01	0.08	0.19	0.31**	0.31**	0.36**	0.08	0.24*	0.36**	-0.01	0.45**	0.44**	0.11	0.43**
NDVI2	0.40**	0.39**	0.48**	0.56**			0.91**	0.37**	0.36**	0.36**	-0.01	0.05	0.16	0.14	0.45**	0.44**	0.48**	-0.07	0.42**	0.34**	0.06	0.32**	0.31**	0.27**	0.51**
NDVI3	0.43**	0.42**	0.52**	0.73**	0.64**			0.34**	0.34**	0.34**	-0.06	0.06	0.15	0.20*	0.45**	0.43**	0.46**	-0.06	0.39**	0.47**	0.09	0.37**	0.37**	0.25*	0.51**
DH	0.12	0.11	0.06	-0.06	-0.07	-0.04		0.79**	0.16	0.79**	0.16	-0.04	0.21*	-0.05	0.38**	0.38**	0.08	-0.27**	0.40**	0.13	0.07	-0.08	-0.08	0.42**	0.36**
DM	0.06	0.08	0.03	-0.07	-0.11	0.04	0.80**		0.80**	0.16	-0.02	-0.12	0.18	0.00	0.26**	0.24*	0.05	-0.09	0.29**	0.14	0.28**	0.01	0.01	0.41**	0.40**
PH	0.11	0.06	0.04	0.05	0.01	0.00	0.27**	0.20*		0.02	-0.08	-0.08	0.04	0.09	-0.01	-0.01	-0.21*	-0.04	-0.08	-0.08	-0.15	-0.10	-0.09	-0.09	-0.17
TPM	0.28**	0.27**	0.30**	0.31**	0.23*	0.40**			0.02	0.09	0.09		0.01	0.02	0.18	0.17	0.13	-0.15	0.24*	0.19	0.03	0.09	0.07	0.14	0.23*
SL	0.46**	0.48**	0.39**	0.15	0.15	0.11	0.21*	0.29**	0.18	0.34**			0.19	0.17	0.19	0.19	0.06	0.06	0.07	-0.05	0.00	0.09	0.10	0.13	0.20*
AL	0.06	0.11	0.11	0.07	0.16	0.16	0.11	-0.18	-0.05	-0.13	0.08	0.09		-0.03	-0.01	-0.01	0.06	0.23*	-0.10	0.14	-0.04	0.34**	0.34	-0.18	0.07
SPS	0.40**	0.41**	0.40**	0.32**	0.32**	0.31**	0.18	0.19	0.13	0.38**	0.55**	0.08	0.08	0.99**	0.99**	0.33**	-0.26**	0.26**	0.47**	0.14	0.01	0.01	0.04	0.43**	0.47**
GPS	0.42**	0.42**	0.39**	0.28**	0.37**	0.28**	0.18	0.18	0.15	0.40**	0.55**	0.13	0.96**	0.13	0.96**	0.34**	-0.26**	0.26**	0.47**	0.15	0.01	-0.00	0.00	0.44**	0.47**
TGW	0.52**	0.50**	0.60**	0.39**	0.34**	0.38**	-0.05	-0.03	0.12	0.17	0.12	0.17	0.38**	0.08	0.38**	0.35**		-0.33**	0.59**	0.42**	-0.03	0.21*	0.22*	0.36**	0.54**
BY	-0.26**	-0.28**	-0.23*	-0.15	-0.11	-0.07	-0.16	0.08	0.08	0.06	-0.01	0.06	-0.09	0.06	-0.06	-0.06	-0.32**		-0.80**	-0.17	0.10	0.59**	0.58**	-0.62**	-0.24*
HI	0.59**	0.62**	0.58**	0.39**	0.35**	0.39**	0.24*	0.18	0.19	0.43**	0.49**	0.43**	0.49**	-0.04	0.50**	0.48**	0.55**	-0.48**	0.32**	0.36**	0.02	-0.18	-0.17	0.83**	0.74**
GPC	0.44**	0.44**	0.47**	0.31**	0.20*	0.33**	0.11	0.06	-0.02	0.15	-0.02	0.15	0.21	0.14	0.16	0.15	0.36**	-0.36**	-0.23*		0.38**	0.23*	0.24*	0.23*	0.40**
NHI	-0.04	-0.09	0.01	-0.03	0.00	0.04	0.12	0.11	-0.03	0.06	-0.04	0.06	-0.04	0.13	0.04	0.03	-0.16	0.28**	-0.23*	0.29**		-0.17	-0.17	0.32*	0.19
TNUp	0.42**	0.46**	0.45**	0.32**	0.26**	0.39**	0.01	0.18	0.14	0.43**	0.38**	0.43**	0.38**	0.09	0.33**	0.32**	0.36**	0.33**	0.53**	0.32**	-0.14		0.99**	0.38**	0.32**
NUPE	0.43**	0.47**	0.46**	0.33**	0.24*	0.38**	0.03	0.18	0.14	0.43**	0.39**	0.43**	0.39**	0.07	0.32**	0.30**	0.36**	0.30**	0.53**	0.33**	-0.13	0.99**		-0.37**	0.32**
NUtE	0.31**	0.31**	0.30**	0.19	0.22*	0.24*	0.23*	0.22*	0.17	0.41**	0.33**	0.41**	0.33**	-0.15	0.44**	0.42**	0.24**	-0.12	0.75**	-0.15	0.05	0.28**	0.29**		0.72**
NUE	0.48**	0.50**	0.49**	0.33**	0.30**	0.38**	0.18	0.26**	0.20*	0.52*	0.20*	0.52*	0.47**	-0.05	0.49**	0.47**	0.38**	0.10	0.82**	0.12	-0.06	0.79**	0.80**		0.80**

Correlation: T1- below diagonal, T2- Above diagonal

* and **: significant at the level of probability p < 0.05 and p < 0.01, respectively

CC1, CC2, CC3 & NDVI1, NDVI2, NDVI3- Chlorophyll contents and NDVI at booting, anthesis and grain filling stages, respectively, DH- Days to heading, DM- Days to maturity, PH- Plant height, TPM- Number of productive tiller per meter, SL- Spike length, AL- Awn length, SPS-Number of spikelet's per spike, GPS-Number of grains per spike, TGW- Thousand grain weight, BY- Biomass yield, HI- Harvest index, GPC-Grain protein content, NHI- Nitrogen harvest index, TNUp- Total nitrogen uptake, NUPE- Nitrogen uptake efficiency, NUtE- Nitrogen utilization efficiency, NUE- Nitrogen use efficiency

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3.4 Genetic correlation estimates (r_g) and predicted response from direct selection under stress condition (CR_{T1}/R_{T1}) for agro-morphological, physiological and NUE related traits

In the context of this study, the genetic correlations between T2 and T1 were approximately 1.0 for CC1, CC2, TSW, GY, HI, BMY and NUE. The evaluation of indirect selection efficiency under T2 concerning performance at T1 relative to the predicted response from direct selection under T1 (CR_{T1}/R_{T1}) showed values close to 1.0 for CC1, TSW, BMY, GPC, and NUE (Table 3). This implies that utilizing indirect selection under optimal fertilization conditions for these traits would be as effective as direct selection under reduced fertilization. However, for remaining traits, the efficiency of indirect selection would be comparatively lower than direct selection. Regarding the trait CC1, the prospects for indirect selection appeared even more favourable than direct selection ($CR_{T1}/R_{T1} = 1.18$), mainly due to the substantially lower heritability under T1 in contrast to T2.

3.5 Performance of genotypes for yield and NUE related traits

Based on GY, top ten genotypes for high grain yield (S3A Fig) and poor grain yield (S3B Fig) at T1(N50) and T2 (N100) nitrogen levels were identified, respectively. The top 10 highest yielding genotypes and the least 10 poor grain yielding genotypes were identified under N50 conditions which were identified as N-insensitive (NIS) and N-sensitive (NS) genotypes,

Table 3. Genetic correlations (r_g) estimates of analysed traits at two treatments and across the years and 50 wheat genotypes and efficiency of indirect selection under T2 for performance under T1 relative to the predicted response to direct selection under T1 (CR_{T1}/R_{T1}) for 24 traits.

Traits	$r_g \pm S.E.^a$	CR_{T1}/R_{T1}
CC1	1.05 \pm 0.00172	1.18
CC2	0.94 \pm 0.27646	0.89
CC3	0.84 \pm 0.50086	0.82
NDVI1	0.30 \pm 0.06782	0.32
NDVI2	0.44 \pm 0.05840	0.47
NDVI3	0.63 \pm 0.01239	0.66
DH	0.77 \pm 0.97472	0.65
DM	0.72 \pm 0.00886	0.73
PH	0.08 \pm 0.12698	0.09
TPM	0.15 \pm 0.01737	0.15
SL	0.75 \pm 0.00762	0.83
AL	0.48 \pm 0.02052	0.58
SPS	0.74 \pm 0.00245	0.63
GPS	0.75 \pm 0.00369	0.65
TSW	1.00 \pm 0.00021	1.12
GY	0.93 \pm 0.00089	0.82
BMY	1.02 \pm 0.00057	1.07
HI	0.94 \pm 0.01070	0.81
GPC	0.83 \pm 0.00352	1.04
NHI	0.23 \pm 0.02736	0.24
TNUP	0.70 \pm 0.65115	0.59
NU _p E	0.71 \pm 0.00002	0.67
NU _t E	0.85 \pm 0.00221	0.92
NUE	0.93 \pm 0.00001	0.97

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respectively [58]. Out of 10 NIS genotypes, four genotypes (UASBW13356, UASBW13358, UASBW13354 and UASBW13357) were common in the top 10 high yielding genotypes at the N100 level showing genotypic plasticity towards N50 and N100 conditions. Similarly, out of 10 NS genotypes, 7 lines (KRL237, DTW2011-56, HI8730, DL153-2, UASBW13362, WH1021 and MP1293) were also included in the 10 least yielding genotypes in T2. NUE and its associated traits also showed a wide range among genotypes under both T1 and T2 conditions (Table 4). The top 10 promising genotypes for each trait were identified under both conditions. The common genotypes in both T1 and T2 conditions for NUE (UASBW13356, UASBW13358, UASBW13354, and UASBW13357), TNU_p & NU_pE (UASBW13356, PBW175, Raj1972, and WH1022), NHI (UASBW13362), NU_tE (UASBW13356, UASBW13358, UASBW13354, and UASBW13364) and GPC (UASBW13356, UASBW13358, PBW175, K9107, HD2967, and GW322) were identified as potential donors. Based on comparative performance of the genotypes for their NutE under T1 and T2 conditions, the genotypes were classified into four groups—viz., N efficient responsive (NER), N efficient non-responsive (NENR), N inefficient responsive (NIR) and N inefficient non-responsive (NINR) [59, 60]. Subsequently, an evaluation of their genotypic performance for NU_tE was conducted under low (T1- N50) and recommended (T2-N100) N conditions (S4 Fig). The average NU_tE of all 50 genotypes under T1 (30.16) and T2 condition (28.08) were considered as a cut-off for the identification of genotypic efficiency and responsiveness for N use.

Above-average genotypes under T1 were considered efficient (E) whereas below-average genotypes were considered inefficient (I). Similarly, in T2, above-average genotypes were considered responders (R), and below-average genotypes are considered non-responders (NR). In this classification, N was prefixed to indicate nitrogen. Considering together the category in T1 and T2, it was observed that efficient genotypes are higher in the utilization of absorbed N over inefficient genotypes. In this way, 20 genotypes viz., GW322, HD2967, HI1500, Kalyansona, KRL1-4, PBW343, UAS304, UAS323, UASBW13354, UASBW13355, UASBW13356, UASBW13357, UASBW13358, UASBW13359, UASBW13363, UASBW13364, UASBW13367, UASBW10453, UASBW12876, WH147 were categorised as efficient and responsive whereas 18 genotypes namely, HD2189, DBW14, DL153-2, DTW2011-56, HI8730, KRL237, MP1293, MP4010, NP846, PBW175, RAJ1972, RAJ4248, UASBW13360, UASBW13362, UASBW13365, UASBW10227, WH1021, and WH1022 were found as inefficient and non-responsive. Likewise, genotypes GW2013-540, HD2733, UASBW13361, UASBW12758, UASBW11948, UASBW12380, and UASBW12878, WH542, UASBW13366 were classified as efficient but non-responsive and HPW251, K9107, and UASBW12877 were categorised as in-efficient but responsive genotypes to N fertilisers.

3.6 Stress tolerance indices for NUE

The pooled data of grain yield in T1 and T2 for a total 50 genotypes were used to estimate different stress related indices for the identification of promising genotypes that have the least effects of stress conditions, i.e., reduced nitrogen levels (Table 4). Three different indices namely yield stability index (YSI), stress susceptibility Index (SSI), and percent reduction under stress (PR) were estimated among which higher values of YSI and lower values of SSI and PR are desirable for the identification of promising genotypes under stress conditions. Genotypes showed a mean YSI of 0.89 with a range of 0.66 to 1.05 whereas the mean SSI was 0.99 with a range of -0.44 to 2.97. PR among genotypes was 11.33 ranging from -4.97 to 33.86 among all the genotypes, exotic lines showed better performance for YSI (0.90), SSI (0.86), and PR (9.84) in comparison to indigenous Indian germplasm lines which showed mean YSI, SSI, and PR of 0.88, 1.10 and 12.5, respectively. The results indicated that two indigenous genotypes

Table 4. Mean performance for NUE traits under T1 (soil N + 50 kg Nha⁻¹) and T2 (Soil N + 100 kg Nha⁻¹) conditions and stress indices in bread wheat genotypes.

S No.	Genotypes	Gr Yield (q/ha)		NUE related traits								Stress indices		
				NUE		NUpE		NUtE		GPC		YSI	SSI	PR (%)
		T1	T2	T1	T2	T1	T2	T1	T2	T1	T2			
1	DBW 14	23.32	30.28	12.92	14.23	0.50	0.57	25.98	25.26	11.33	13.17	0.77	2.02	22.99
2	DL 153-2	20.60	28.64	11.39	13.45	0.42	0.48	27.12	27.99	11.58	11.61	0.72	2.46	28.08
3	DTW 2011-56	19.16	20.40	10.59	9.57	0.44	0.53	24.01	17.95	12.19	10.85	0.94	0.53	6.07
4	GW 2013-540	28.32	30.33	15.69	14.24	0.49	0.54	32.44	26.91	11.41	13.06	0.93	0.58	6.63
5	GW 322	28.78	31.98	15.96	15.02	0.50	0.48	31.84	31.19	13.04	13.26	0.90	0.88	10.00
6	HD 2189	24.60	29.03	13.63	13.61	0.47	0.49	29.39	27.92	10.56	11.56	0.85	1.34	15.24
7	HD 2733	29.77	35.54	16.44	16.66	0.54	0.63	30.84	26.60	12.84	12.78	0.84	1.42	16.24
8	HD 2967	31.25	37.90	17.29	17.78	0.54	0.59	32.59	30.64	13.07	13.45	0.82	1.54	17.54
9	HI 1500	28.94	30.23	16.01	14.17	0.53	0.48	30.73	32.22	12.32	12.56	0.96	0.37	4.26
10	HI8730	19.27	23.34	10.68	10.97	0.49	0.52	22.11	21.24	11.48	11.28	0.83	1.53	17.42
11	HPW251	26.88	30.61	14.88	14.37	0.51	0.48	29.11	30.28	12.74	13.33	0.88	1.07	12.17
12	K 9107	31.42	37.29	17.38	17.49	0.58	0.57	30.13	31.11	13.10	13.63	0.84	1.38	15.74
13	Kalyansona	30.83	33.26	17.05	15.62	0.57	0.55	30.21	28.66	12.98	12.97	0.93	0.64	7.31
14	KRL 237	16.87	25.51	9.36	11.96	0.40	0.51	23.41	23.32	12.16	12.25	0.66	2.97	33.86
15	KRL1-4	32.15	34.28	17.81	16.08	0.58	0.54	31.06	29.54	11.85	12.22	0.94	0.55	6.21
16	MP 1293	25.24	27.97	13.99	13.12	0.50	0.56	28.29	23.49	11.21	12.02	0.90	0.86	9.75
17	MP 4010	26.42	26.72	14.60	12.54	0.53	0.58	27.61	21.78	12.42	12.17	0.99	0.10	1.12
18	NP846	27.90	27.80	15.46	13.07	0.59	0.56	26.44	23.32	12.72	13.22	1.00	-0.03	-0.37
19	PBW 175	32.36	30.83	17.89	14.49	0.62	0.62	28.89	23.31	12.92	13.63	1.05	-0.44	-4.97
20	PBW 343	28.80	38.45	15.93	18.04	0.51	0.52	31.77	34.92	13.06	13.20	0.75	2.20	25.08
21	RAJ 1972	29.96	32.89	16.57	15.41	0.63	0.62	26.55	24.96	12.55	13.70	0.91	0.78	8.90
22	RAJ 4248	27.93	32.93	15.48	15.42	0.56	0.59	28.01	26.21	12.32	12.76	0.85	1.33	15.16
23	UAS 304	33.87	34.43	18.73	16.15	0.57	0.53	32.98	30.64	11.62	12.20	0.98	0.14	1.63
24	UAS 323	29.25	35.55	16.22	16.68	0.53	0.58	30.46	29.04	10.99	12.68	0.82	1.55	17.71
25	WH 147	25.29	30.17	14.02	14.14	0.45	0.42	31.08	34.01	11.95	11.76	0.84	1.42	16.16
26	WH 542	29.58	32.32	16.37	15.17	0.47	0.55	34.99	27.84	10.70	11.67	0.92	0.75	8.49
27	WH 1021	22.83	26.87	12.61	12.58	0.47	0.45	26.99	27.93	12.21	12.87	0.85	1.32	15.04
28	WH 1022	30.47	36.47	16.85	17.10	0.58	0.62	29.42	27.84	12.94	13.17	0.84	1.44	16.46
29	UASBW13354	34.20	38.76	18.90	18.19	0.57	0.53	33.10	34.61	12.52	12.95	0.88	1.04	11.75
30	UASBW13355	29.10	39.06	16.09	18.32	0.51	0.56	31.47	32.90	12.70	13.60	0.75	2.24	25.50
31	UASBW13356	38.91	40.36	21.54	18.95	0.62	0.59	35.15	32.27	13.29	13.69	0.96	0.32	3.60
32	UASBW13357	33.34	38.08	18.47	17.86	0.61	0.58	30.61	30.86	12.75	12.59	0.88	1.09	12.44
33	UASBW13358	38.12	40.12	21.11	18.83	0.62	0.54	34.02	35.17	13.04	13.34	0.95	0.44	4.98
34	UASBW13359	33.60	35.43	18.59	16.63	0.58	0.55	32.15	33.40	12.99	12.71	0.95	0.46	5.19
35	UASBW13360	29.39	31.48	16.25	14.78	0.56	0.54	29.29	27.84	12.73	13.08	0.93	0.58	6.65
36	UASBW13361	29.09	33.50	16.11	15.71	0.51	0.59	31.83	26.67	10.67	12.86	0.87	1.16	13.16
37	UASBW13362	22.19	26.08	12.28	12.22	0.46	0.54	26.98	22.67	12.17	12.80	0.85	1.31	14.90
38	UASBW13363	27.70	28.98	15.32	13.61	0.49	0.47	31.68	28.76	11.85	12.25	0.96	0.39	4.42
39	UASBW13364	32.02	34.53	17.71	16.17	0.52	0.52	34.38	31.28	12.04	10.90	0.93	0.64	7.26
40	UASBW13365	25.81	28.97	14.31	13.58	0.50	0.53	29.06	25.66	11.80	12.54	0.89	0.96	10.90
41	UASBW13366	29.44	33.22	16.33	15.61	0.47	0.58	33.40	27.22	11.14	12.41	0.89	0.99	11.38
42	UASBW13367	31.05	32.40	17.20	15.23	0.53	0.55	32.68	28.21	11.18	12.97	0.96	0.36	4.14
43	UASBW12758	26.42	31.17	14.64	14.67	0.49	0.59	30.22	24.77	12.15	12.22	0.85	1.34	15.25
44	UASBW10227	28.57	29.13	15.80	13.65	0.54	0.54	29.32	25.18	12.26	11.15	0.98	0.17	1.91

(Continued)

Table 4. (Continued)

S No.	Genotypes	Gr Yield (q/ha)		NUE related traits								Stress indices		
				NUE		NUpE		NUE		GPC		YSI	SSI	PR (%)
		T1	T2	T1	T2	T1	T2	T1	T2	T1	T2			
45	UASBW10453	30.44	35.44	16.85	16.64	0.54	0.58	31.57	29.35	11.04	12.04	0.86	1.24	14.11
46	UASBW11948	29.73	34.42	16.42	16.13	0.51	0.63	31.94	25.67	11.27	11.77	0.86	1.19	13.61
47	UASBW12380	31.00	34.25	17.13	16.06	0.56	0.60	30.89	27.04	12.03	12.13	0.91	0.83	9.49
48	UASBW12876	29.80	33.01	16.48	15.47	0.52	0.48	31.72	32.54	11.14	12.42	0.90	0.85	9.72
49	UASBW12877	28.58	32.57	15.80	15.28	0.55	0.54	29.03	28.75	11.36	12.58	0.88	1.08	12.26
50	UASBW12878	31.84	33.16	17.62	15.54	0.53	0.58	33.36	27.13	10.95	12.13	0.96	0.35	3.96
	Mean	28.65	32.32	15.85	15.16	0.52	0.54	30.16	28.08	12.06	12.56	0.89	0.99	11.33
	CD @ 5%	5.18	3.94	2.85	1.70	0.06	0.09	4.03	4.40	1.34	1.26			

NUE-Nitrogen use efficiency, NUpE-Nitrogen uptake efficiency, NUE-Nitrogen utilization efficiency, GPC-Grain protein content, YSI- Yield stability index, SSI, Stress susceptibility index, PR- Percent reduction under stress

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namely PBW175 and NP846 showed higher yield at reduced nitrogen levels which is also supported by negative SSI and PR values in PBW175 (-0.44 & -4.97) and NP846 (-0.03 & -0.37), respectively. The perusal of results of YSI, SSI, and PR indicated that PBW175, NP846, MP4010, UAS304, and UASBW10227 were promising genotypes that were least affected for grain yield under reduced nitrogen doses. In addition, the indigenous germplasm lines, lines, HI1500, DTW2011-56, KRL1-4, GW2013-540, and exotic germplasm lines UASBW13356, UASBW12878, UASBW13367, UASBW13363, UASBW13358, and UASBW13359 showed promising performance for these three stress indices.

4. Discussion

Nitrogen (N) is often a limiting factor for crop yield and is, therefore, considered as an essential macronutrient required for the plant growth. Globally, the adoption of N-responsive cultivars and extensive application of N fertilizer has augmented food production in the past 50 years of agriculture. This excessive application of N fertilizer is increasing the cost of cultivation and thereby reducing the net profitability at the farm level in addition to negative impacts on the environment [61–65]. The present investigation was driven by a worldwide willingness, especially from a huge number of small and marginal farmers in developing countries, to understand genotype behaviour for the identification and development of genotypes (without compromising GY) with high NUE under low N conditions. As demand for wheat grain is still rising, there is a need to boost productivity and production for the amount of N applied and therefore, identification of genotypes with better NUE is crucial. The immediate aim should, therefore, be to exploit the available variability for NUE in wheat cultivars through an appropriate breeding procedure.

The present study aimed to understand the variability and correlations for NUE traits with various yield and physiological parameters to identify the promising genotypes based on various parameters of nitrogen sensitivity, response to N fertilizers, and susceptibility indices at low N availability. Precised phenotyping under low N input is challenging and influenced by the genotype (G), environment (E), and G × E interactions [66, 67] which makes it difficult to identify nitrogen insensitive genotypes under low N conditions at field level. Presently, very few wheat breeding programs are targeting the development of low N tolerance traits which are a must for sustainable agriculture with minimal negative impacts on the environment.

Analysis of variance indicated a highly significant variation among genotypes at both treatments for all the traits studied and this existence of significant differences for different

attributes is very helpful to improve the trait of interest. Diverse responses have been observed among genotypes for all traits across N levels, despite similar growth conditions and an equal amount of N fertilizer in a given N level. The genotypic variations observed purely reveal trait specific genotype plasticity. These results concur with other reported field experiments on wheat [68, 69].

Wheat is more responsive to a higher dose of nitrogen whereas, sensitive to low nitrogen availability like most cereal crops. Less soil N availability leads to a reduced plant vegetative growth especially tillering which results in reduced grain yield [70]. In this study, reduction in tillering and ultimately yield reduction to the tune of approx. 11.3% was observed at reduced N levels as compared to the recommended dose. Similarly, the low N availability is expressed phenotypically in the form of low chlorophyll content as indicated by lower SPAD and NDVI values which might be due to less N availability in the leaves at lower N levels. Similar results of increased chlorophyll content by increasing N application were reported by [71–73]. Almost all the traits studied except NDVI-1, HI, NUE, NUtE and NHI, showed reduced performance at lower N doses which reflects the genotypic sensitivity to low N availability [74]. When N supply was low, the remobilization of N from the tillers and leaves was efficient and converted into grain N indicating a better NUtE. This suggested that few genotypes have the potential to use nitrogen more efficiently and exhibiting high NUE. The NHI indicates the level of efficiency of plants to use acquired total N for grain formation for which significant genotypic variation was observed at both the N levels [75]. Varied response of genotypes to NUPE, NUtE, and NHI was observed by [28, 33, 76–79] which are in tune with the performance of NUE related traits at varying N levels in the present study. From per cent reduction calculations we came to know that a significant reductions were observed in almost traits at T1 in comparison to T2 except, NHI (2.8%), NUtE (7.4%), and NUE (4.6%) traits which exhibited notably higher mean values at T1 when contrasted with T2. A study was also conducted on similar investigation, revealing a 10% reduction in yield and a 14% decrease in protein content at LN as opposed to HN conditions [80]. Additionally, grain N yield and aboveground N per unit area were both reduced by approximately 20% under LN conditions.

To assess the nature and magnitude of diversity among the genotypes, the phenotypic and genotypic coefficients of variation were estimated. In general, higher PCV values were observed as compared to GCV values at both the nitrogen levels. High PCV and GCV were observed for harvest index whereas other major yield traits and NUE related traits showed moderate values. Similarly, previous studies have established significant genetic variation for NUE-related traits in wheat [81, 82].

In general, moderate to high heritability was recorded under both nitrogen levels. The heritability values in a broad sense ranged from 34.05% to 86.84% in T1 and 37.81% to 93.84% in T2. High heritability (>60%) were observed for CC-3, DH, DM, SPS, GY, BMY, HI, NUE, and NUtE under both conditions. A similar trend was also observed for yield and NUE related traits by [83, 84]. High heritability is useful in laying importance in choice for such traits during selection.

In the present study, nitrogen use efficiency was measured as grain yield divided by available N (soil N + fertilizer N) due to which the correlation of NUE and grain yield at both levels of nitrogen treatments showed similar results. The positive and highly significant correlated response of NUE with DM, HI, TGW, GPS, SPS, CC-1, CC-2, CC-3, NDVI-1, NDVI-2, NDVI-3 TNU_p, NUPE, and NUtE under both the condition suggested improvement of these traits for better NUE in wheat genotypes [85, 86]. Interestingly, highly significant and positive associations between physiological traits and NUE related traits were also observed. The results of BLUP analysis also indicated reliability of NUE related traits with positively moderate values for improvement for grain yield. Similar and contrast regression relationships from BLUP

analysis were also graphically depicted involving grain yield vs GPC, grain yield vs NUpE, and grain yield vs NUtE by [80]. Similar associations of NUE related traits by [46, 70, 79, 87, 88] indicated that these characters should be considered as NUE components in crop breeding that can be harnessed for higher grain yields under both low and high N fertility conditions. The current study highlighted that genetic correlations between T2 and T1 were notably high for CC1, CC2, TSW, GY, HI, BMV, and NUE, indicating the absence of significant genotype by nitrogen (G x N) interactions for these traits. Heritability estimates for the assessed traits either showed similarity across both N treatment levels or were higher at T2 compared to T1. Reflecting this, the indirect selection efficiency (ISE) at T2 for performance at T1, relative to direct selection at T1, was equal to or greater than 1 for CC1, TSW, BMV, GPC, and NUE. This aligns with previous findings from studies by [80, 89], underscoring its significant implications for the selection of wheat cultivars with enhanced nitrogen efficiency.

Across the year on average, higher GY was recorded at higher N doze and large genetic variability among genotypes for GY was observed. In general, increased GY was correlated with the enhanced N application rate, which might be due to sufficient nitrogen availability. Reduced GY by 10% under low N conditions compared to normal conditions was also reported by [65, 89] in wheat. Remarkably, top 10 highest yielding genotypes were identified under N50 conditions as N-insensitive (NIS) among which four genotypes (UASBW13356, UASBW13358, UASBW13354, and UASBW13357) were common in the top 10 high yielding genotypes at the N100 level showing their genotypic plasticity towards varying N levels. These genotypes may be potential genetic resources to breed for tolerance to low N conditions. On the other hand, the top 10 high yielding genotypes in N100 can have the best acceptances for cultivation where soils are fertile and following the ideal N levels for cultivation and could be further used to improve GY, along with higher NUE. Several studies have suggested the utilization of N efficient lines with enhanced GY in the farmer fields which may help to reduce fertilizer input as well as increase the profitability of farm operations [14, 75, 90]. Similarly, selected NIS lines with high NUE will certainly play a role in reducing environmental pollution and could increase economic profit for farmers.

The present study indicated that NUE and its associated traits also showed a wide range among genotypes for all the six NUE related traits and their diagrammatic representation of frequency distributions under T1 and T2 conditions were depicted in S5 Fig. The higher frequency of genotypes in medium to high class based on NUE parameters suggests the potentiality of these tested genotypes. The best performing 10 entries were identified for each trait under T1 and T2 conditions among which common genotypes under both the conditions were identified as promising trait specific genetic resources. Further, UASBW13356, UASBW13358, UASBW13354, PBW 175, Raj 1972, and WH 1022 were found promising for multiple NUE traits and may be utilized extensively for NUE improvement programme in wheat. Determination of the genetic variations for NUE related traits is essential for the selection of efficient genotypes that can be used further in breeding programs to develop low N tolerant material. The concept of genotype grouping is used widely in nutrient use efficiency [91]. Among NUE related traits, nitrogen utilization efficiency has the most significance which is defined as the genotype ability to assimilate and remobilize N ultimately to produce the GY [16, 59]. Large genotypic variations in NUtE have been reported under field/pot screening in various wheat genotypes and other crops [92–94]. Based on NUtE efficiency data in T1 (N50) versus T2 (N100), above and below average genotypes under T1 were considered efficient and inefficient, respectively whereas, above and below average genotypes in T2 were considered as responders, and non-responders, respectively.

Based on NUtE in T1(N50), 29 efficient (E) and 21 inefficient (I) genotypes were identified whereas 23 responsive (R) and 27 non-responsive (NR) genotypes were identified based on

NUtE in T2 (N100), Considering together, all the 50 genotypes were classified into four groups viz., NER (20), NENR (9), NIR (03) and NINR (18) as per [59, 60]. The NENR genotypes showed a progressive performance under low N which may enable breeders to develop efficient genotypes under low input environments. The remaining 20 NER genotypes were most desirable that exhibited a progressive response to increased N availability. These NER genotypes could be the prospective targets for selection for the genetic improvement of wheat for better N utilization. Interestingly, among the 10 NIS genotypes, 8 genotypes namely, UASBW13356, UASBW13358, UASBW13354, UAS304, UASBW13359, UASBW13357, KRL1-4, and UASBW13364 were N efficient and responsive which showed progressive performance in terms of efficient and responsive use of nitrogen. High GY per unit of N consumption could be obtained from the genotypes having more NUtE [19], and thus, breeding for efficient genotypes under low N could be progressed with high NUtE [75, 95, 96]. Similarly, inefficient but responsive genotypes (HPW251, K9107, and UASBW12877) can be used in breeding programs for further improvement. The rest of the inefficient, nonresponsive genotypes are less desirable from the NUE point of view. Overall, it was found that the efficient genotypes are higher in the utilization of absorbed nutrients than inefficient genotypes.

The differential behaviour of genotypes under stress was studied by estimation of three different stress related indices for the identification of promising genotypes that have the least effects of stress conditions, *i.e.*, reduced nitrogen levels. Among these indices, higher values of YSI and lower values of SSI and PR were desirable for the identification of promising genotypes under stress conditions. A wide range was observed among genotypes for these indices and exotic genotypes showed more desirable indices. Based on these indices, the indigenous germplasm lines, HI1500, DTW2011-56, KRL1-4, GW2013-540, and exotic germplasm lines UASBW13356, UASBW12878, UASBW13367, UASBW13363, UASBW13358, and UASBW13359 were found promising. These indices also indicated PBW175, NP846, MP4010, UAS304, and UASBW10227 as the least affected genotypes for grain yield under reduced nitrogen doses.

5. Conclusions

Improving NUE is pivotal for a sustainable crop growth and yield, especially under low nitrogen soils on which depend majority of poor farmers in most developing countries, especially in South Asia. Likewise, improving crop productivity using N fertilization is important for achieving climate resilience. Nevertheless, the genetic improvement of NUE depends on the nature and extent of variation among the germplasm. This study aimed to derive morphologic and agronomical traits associated with NUE in a set of 50 diverse lines under different N levels. Although, the nitrogen limitation has resulted in the reduction of GY, extensive genetic variations for grain yield, NUE, and their associated traits were noted among genotypes at both levels of nitrogen. The association between yield and NUE traits indicated avenues for wheat improvement concerning enhanced NUE and thereby, yield. Different selection parameters for the selection of nitrogen-insensitive genotypes, nitrogen efficient and responsive genotypes, and stress tolerance due to low N availability have identified UASBW13356, UASBW13358, UASBW13354, UASBW13357, and KRL1-4 as promising genotypes that showed their inherent genotypic plasticity toward N application. These findings suggest that genotypes with more yield and high to moderate NUtE can be used as parents for the breeding of N efficient genotypes for marginal agro-ecologies. Furthermore, these nitrogen-use efficient genotypes can be further tested on large scale to know their stability for their release as a variety so that farmers can get the advantage of low input cost. These genotypes have the potential to decrease environmental pollution and high economic costs associated with excess N fertilizer. Additionally, these are the best suggested genotypes for organic farming due to their

inherent potentiality of low input response which may also help the vast number of small and marginal farmers for optimizing the use of fertilizer inputs for economic and environmentally sustainable food production.

Supporting information

S1 Fig. The relationship between grain yield and NUE under two nitrogen levels. The grain yield and NUE under the two nitrogen levels N50 and N100 is showing similar trend as NUE is estimated by dividing grain yield available N soil.

(TIF)

S2 Fig. BLUP analysis for correlation of grain yield with NUE traits. GY showed (D) perfectly positive correlation with NUE, positive but weak correlation with (A) GPC and (B) NUpE under both N levels. GY showed (C) moderately positive correlation under T1 and weaker but positive correlation under T2 with NUtE.

(TIF)

S3 Fig. Top high yielding and poor yielding genotypes under T1 and T2 conditions. Top ten genotypes for high grain yield (A) and poor grain yield (B) at T1(N50) and T2 (N100) nitrogen levels were identified.

(TIF)

S4 Fig. Relationship between genotypic performance for NUtE under low (T1-N50) and recommended (T2-N100) nitrogen conditions. The genotypes were classified into four groups—viz., N efficient responsive (NER), N efficient non-responsive (NENR), N inefficient responsive (NIR) and N inefficient non-responsive (NINR) based on comparative performance of the genotypes for their NUtE under T1 and T2 conditions.

(TIF)

S5 Fig. Frequency distribution for various NUE related traits in wheat under T1 and T2 conditions. Frequency distribution pattern for all the six NUE related traits namely TNUp, NUtE, NUpE, NUE, NHI and GPC was studied under T1 and T2 conditions.

(TIF)

S1 Table. Mean performance of bread wheat genotypes for agro-morphological, physiological and NUE traits under T₁ (soil N + 50 kg Nha⁻¹) and T₂ (Soil N + 100 kg Nha⁻¹) conditions. DH: Days to Heading, DM: Days to maturity, PH: Plant height (cm), NPT: Number of productive tiller per meter, SL: Spike length (cm), AL: Awn length (cm), SPS: spikelet's per spike, GPS: grains per spike, TSW: Thousand seed weight (g), GY: Grain yield (q ha⁻¹), BMY: Biomass yield (q ha⁻¹), HI: Harvest index (%), GPC: Grain protein content (%), CC-1: Chlorophyll content at booting stage, NDVI -1: NDVI at booting stage, CC-2: Chlorophyll content at anthesis stage, NDVI - 2: NDVI at anthesis stage, CC-3: Chlorophyll content at grain filling stage, NDVI -3: NDVI at grain filling stage, NHI: Nitrogen harvest index, TNUp: Total nitrogen uptake (kg N ha⁻¹), NUpE: Nitrogen uptake efficiency (kg N kg⁻¹N), NUtE: Nitrogen utilization efficiency (kg grain kg⁻¹N) and NUE: Nitrogen use efficiency (kg grain kg⁻¹N).

(DOCX)

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References

1. FAO. 2019. World fertilizer trends and outlook to 2022. Pp. 40.
2. Ladha J.K., Tirol-Padre A., Reddy C.K., Cassman K.G., Verma S., Powlson D.S., et al. 2016. Global nitrogen budgets in cereals: A 50-year assessment for maize, rice, and wheat production systems. *Sci. Rep.* 6, 19355. <https://doi.org/10.1038/srep19355> PMID: 26778035
3. Raun W.R., Johnson G.V., 1999. Improving nitrogen use efficiency for cereal production. *Agron. J.* 91, 357–363.
4. Lassaletta L., Billen G., Grizzetti B., Anglade J., Garnier J., 2014. 50 year trends in nitrogen use efficiency of world cropping systems: the relationship between yield and nitrogen input to cropland. *Environ. Res. Lett.* 9, 105011. <https://doi.org/10.1088/1748-9326/9/10/105011>
5. Vitousek P.M., Mooney H.A., Lubchenco J., Melillo J.M., 1997. Human domination of Earth's ecosystems. *Sci.* 277, 494–499. <https://doi.org/10.1126/science.277.5325.494>
6. Zafar J., Muhammad F.C., 2007. Effects of soil and foliar application of different concentrations of NPK and foliar application of $(\text{NH}_4)_2\text{SO}_4$ on growth and yield attributes in wheat (*Triticum aestivum* L). *Pak. J. Agric. Sci.* 13(2), 119–128.
7. Hickman J.E., Palm C.A., Mutuo P., Melillo J.M., Tang J., 2014. Nitrous oxide (N_2O) emissions in response to increasing fertilizer addition in maize (*Zea mays* L.) agriculture in western Kenya. *Nutr. Cycling Agroecosyst.* 100, 177–187. <https://doi.org/10.1007/s10705-014-9636-7>
8. Russo T.A., Tully K., Palm C., Neill C., 2017. Leaching losses from Kenyan maize cropland receiving different rates of nitrogen fertilizer. *Nutr. Cycling Agroecosyst.* 108, 195–209. <https://doi.org/10.1007/s10705-017-9852-z> PMID: 33488271
9. Chattha M.S., Ali Q., Haroon M., Afzal M.J., Javed T., Hussain S., et al. 2022. Enhancement of nitrogen use efficiency through agronomic and molecular based approaches in cotton. *Front. Plant Sci.* 13, 1–24. <https://doi.org/10.3389/fpls.2022.994306> PMID: 36237509
10. Raun W.R., Johnson G.V., Westerman R.L., 1999. Fertilizer nitrogen recovery in long-term continuous winter wheat. *Soil Sci. Soc. Am. J.*, 63, 645–650. <https://doi.org/10.2136/sssaj1999.03615995006300030030x>
11. Hirel B., Le Gouis J., Ney B., Gallais A., 2007. The challenge of improving nitrogen use efficiency in crop plants: towards a more central role for genetic variability and quantitative genetics within integrated approaches. *J. Exp. Bot.* 58, 2369–2387. <https://doi.org/10.1093/jxb/erm097> PMID: 17556767
12. Monostori I., Árendás T., Hoffman B., Galiba G., Gierczik K., Szira F., et al. 2016. Relationship between SPAD value and grain yield can be affected by cultivar, environment and soil nitrogen content in wheat. *Euphytica* 211, 103–112. <https://doi.org/10.1007/s10681-016-1741-z>
13. Yang H., Yang J., Lu Y., He J., 2014. SPAD values and nitrogen nutrition index for the evaluation of rice nitrogen status. *Plant Pro. Sci.* 17, 81–92. <https://doi.org/10.1626/pp.s.17.81>
14. Vijayalakshmi P., Kiran T.V., Kumari B.R., Srikanth B., Rao I.S., Swamy K.N., et al. 2015. Biochemical and physiological characterization for nitrogen use efficiency in aromatic rice genotypes. *Field Crops Res.* 179, 132–143. <https://doi.org/10.1016/j.fcr.2015.04.012>

15. Kiran T.V., Vijayalakshmi P., Rao Y.V., Swamy K.N., Kondamudi R., Srikanth B., et al. 2016. Effects of nitrogen limitation on antioxidant enzymes, chlorophyll content and grain yield of rice genotypes. *Asian Res. J. Agri.* 1, 1–10. <https://doi.org/10.9734/arja/2016/28503>
16. Fageria N.K., Baligar V.C., Li Y.C., 2008. The role of nutrient efficient plants in improving crop yields in the twenty first century. *J. Plant Nutr.* 31, 1121–1157. <https://doi.org/10.1080/01904160802116068>
17. Zhao Y., Islam S., Alhabbar Z., Zhang J., O'Hara G., Anwar M. and Ma W., 2023. Current progress and future prospect of wheat genetics research towards an enhanced nitrogen use efficiency. *Plants*, 12(9), 2–22. <https://doi.org/10.3390/plants12091753> PMID: 37176811
18. Zhang X., Davidson E.A., Mauzerall D.L., Searchinger T.D., Dumas P., Shen Y., 2015. Managing nitrogen for sustainable development. *Nature* 528, 51–59. <https://doi.org/10.1038/nature15743> PMID: 26595273
19. Moll R.H., Kamprath E.J., and Jackson W.A., 1982. Analysis and interpretation of factors which contribute to efficiency of nitrogen utilization. *Agron J.* 74, 562–564. <https://doi.org/10.2134/agronj1982.00021962007400030037x>
20. Barraclough P.B., Howarth J.R., Jones J., Lopez-Bellido R., Parmar S., Shepherd C.E., et al. 2010. Nitrogen efficiency of wheat: Genotypic and environmental variation and prospects for improvement. *Eur. J. Agron.* 33, 1–11. <https://doi.org/10.1016/j.eja.2010.01.005>
21. Good A.G., Shrawat A.K., Muench D.G., 2004. Can less yield more? Is reducing nutrient input into the environment compatible with maintaining crop production? *Trends Plant Sci.* 9, 597–605. <https://doi.org/10.1016/j.tplants.2004.10.008> PMID: 15564127
22. Hakeem K.R., Chandna R., Ahmad A., Qureshi M.I., Iqbal M., 2012. Proteomic analysis for low and high nitrogen-responsive proteins in the leaves of rice genotypes grown at three nitrogen levels. *Appl. Biochem.* 168, 834–850. <https://doi.org/10.1007/s12010-012-9823-4> PMID: 22903322
23. Vijayalakshmi P., Kiran T.V., Rao Y.V., Srikanth B., Rao I.S., Sailaja B., et al. 2013. Physiological approaches for increasing nitrogen use efficiency in rice. *Indian J. Plant Physiol.* 18, 208–222. <https://doi.org/10.1007/s40502-013-0042-y>
24. Ladha J.K., Kirk G.J.D., Bennett J., Peng S., Reddy C.K., Reddy P.M., et al. 1998. Opportunities for increased nitrogen-use efficiency from improved lowland rice germplasm. *Field Crops Res.* 56, 41–71. [https://doi.org/10.1016/S0378-4290\(97\)00123-8](https://doi.org/10.1016/S0378-4290(97)00123-8)
25. Borrell A., Garside A., Fukai S., Reid D., 1998. Season, nitrogen rate, and plant type affect nitrogen uptake and nitrogen use efficiency in rice. *Aust. J. Agric. Res.* 49, 829–843. <https://doi.org/10.1071/A97057>
26. Muchow R.C., 1998. Nitrogen utilization efficiency in maize and grain sorghum. *Field Crops Res.* 56, 209–216. [https://doi.org/10.1016/S0378-4290\(97\)00132-9](https://doi.org/10.1016/S0378-4290(97)00132-9)
27. Inthapanya P., Sihavong P., Sihathep V., Chanphengsay M., Fukai S., Basnayake J., 2000. Genotype differences in nutrient uptake and utilisation for grain yield production of rainfed lowland rice under fertilized and non-fertilized conditions. *Field Crops Res.* 65, 57–68. [https://doi.org/10.1016/S0378-4290\(99\)00070-2](https://doi.org/10.1016/S0378-4290(99)00070-2)
28. Le Gouis J., Beghim D., Heumez E. and Pluchard P., 2000. Genetic differences for nitrogen up take and nitrogen utilization efficiencies in winter wheat. *Eur. J. Agron.* 12, 163–173. [https://doi.org/10.1016/S1161-0301\(00\)00045-9](https://doi.org/10.1016/S1161-0301(00)00045-9)
29. Presterl T., Seitz G., Landbeck M., Thiemt E.M., Schmidt W., Geiger H.H., 2003. Improving nitrogen-use efficiency in European maize. *Crop Sci.* 43, 1259–1265. <https://doi.org/10.2135/cropsci2003.1259>
30. Anbessa Y., Juskiw P., Good A., Nyachiro J., and Helm J., 2009. Genetic variability in nitrogen use efficiency of spring barley. *Crop Sci.* 49, 1259–1269. <https://doi.org/10.2135/cropsci2008.09.0566>
31. Namai S., Toriyama K., Fukuta Y., 2009. Genetic variations in dry matter production and physiological nitrogen use efficiency in rice (*Oryza sativa* L.) varieties. *Breed. Sci.* 59, 269–276. <https://doi.org/10.1270/jsbbs.59.269>
32. Le Gouis J., Gaju O., Hubbart S., Allard V., Orford S., Heumez E., et al. 2010. Genetic improvement for an increased nitrogen use efficiency in wheat. *Asp. Appl. Biol.* 105, 151–158.
33. Ortiz-Monasterio J.I., Sayre K.D., Rajaram S., McMahon M., 1997. Genetic progress in wheat yield and nitrogen use efficiency under four nitrogen rates. *Crop Sci.* 37(3), 898–904. <https://doi.org/10.2135/cropsci1997.0011183X003700030033x>
34. Ciampitti I.A., Vyn T.J., 2012. Physiological perspectives of changes over time in maize yield dependency on nitrogen uptake and associated nitrogen efficiencies: A review. *Field Crops Res.* 133, 48–67. <https://doi.org/10.1016/j.fcr.2012.03.008>
35. Sadras V.O., Lawson C., 2013. Nitrogen and water-use efficiency of Australian wheat varieties released between 1958 and 2007. *Eur. J. Agron.* 46, 34–41. <https://doi.org/10.1016/j.eja.2012.11.008>

36. Cormier F., Foulkes J., Hirel B., Gouache D., <Locco Y. M. and Le Gouis J., 2016, Breeding for increased nitrogen-use efficiency: a review for wheat (*T. aestivum* L.). *Plant Breed.* 135, 255–278. <https://doi.org/10.1111/pbr.12371>
37. Hawkesford M.J., 2014. Reducing the reliance on nitrogen fertilizer for wheat production. *J. Cereal Sci.* 59, 276–283. <https://doi.org/10.1016/j.jcs.2013.12.001> PMID: 24882935
38. Cormier F., Le Gouis J., Dubreuil P., Lafarge S. and Praud S., 2014, A genome-wide identification of chromosomal regions determining nitrogen use efficiency components in wheat (*Triticum aestivum* L.). *Theor. Appl. Genet.* 127(12), 2679–2693. <https://doi.org/10.1007/s00122-014-2407-7> PMID: 25326179
39. Hawkesford M.J., 2017. Genetic variation in traits for nitrogen use efficiency in wheat. *J. Exp. Bot.* 68, 2627–2632. <https://doi.org/10.1093/jxb/erx079> PMID: 28338945
40. Maman N., Mason S.C., Lyon D.J., 2006. Nitrogen rate influence on pearl millet yield, nitrogen uptake, and nitrogen use efficiency in Nebraska. *Commun. Soil Sci. Plant Anal.* 37, 127–141. <https://doi.org/10.1080/00103620500406112>
41. Gaju O., Allard V., Martre P., Snape J.W., Heumez E., Le Gouis J., et al. 2011. Identification of traits to improve the nitrogen-use efficiency of wheat genotypes. *Field Crops Res.* 123, 139–152. <https://doi.org/10.1016/j.fcr.2011.05.010>
42. Hawkesford M.J., Riche A.B., 2020. Impacts of GxExM on Nitrogen Use Efficiency in Wheat and Future Prospects. *Front. Plant Sci.*, 11, 1157. <https://doi.org/10.3389/fpls.2020.01157> PMID: 32903740
43. Pathak R.R., Ahmad A., Lochab S., Raghuram N., 2008. Molecular physiology of plant nitrogen use efficiency and biotechnological options for its enhancement. *Curr. Sci.* 94, 1394–1403.
44. Abdo A.I., Sun D., Li Y., Yang J., Metwally M.S., Abdel-Hamed E.M., et al. 2022. Coupling the environmental impacts of reactive nitrogen losses and yield responses of staple crops in China. *Front. Plant Sci.*, 13, 1–13. <https://doi.org/10.3389/fpls.2022.927935> PMID: 36092406
45. Zadoks J.C., Chang T.T., Konzak C.F., 1974. Decimal code for growth stages of cereals. *Weed Res.* 14, 415–421.
46. Ayadi S., Karmous C., Hammami Z., Tamani N., Yoissef T., Esposito S., et al. 2012. Genetic variability of nitrogen use efficiency components in Tunisian improved genotypes and landraces of durum wheat. *Agril. Sci. Res. J.* 2(11), 591–601.
47. FAO. 2008. FAO fertilizer and plant nutrition bulletin: Guide to laboratory establishment for plant nutrient analysis. FAO, Rome, Italy, p. 203.
48. Snedecor G.W., Cochran W.G., 1989. *Statistical Methods*, Eighth Edition, Iowa State University Press.
49. Panse V.G., Sukhatme P.V., 1967. *Statistical methods for agricultural workers*, ICAR Publication, New Delhi.
50. Burton G.W., 1952. Quantitative inheritance in grasses. *Proc. Int. Grassland Congr.* 1, 277–283.
51. Lush J.L., 1940. Intra-class correlations or regression of offspring on dam as a method of estimating heritability of characteristics. *Am. Soc. Anim. Prod.* 33, 293–301
52. Hanson C.H., Robinson H.F., Comstock R.E., 1956. Biometrical studies in yield of segregating population of Korean lespedeza. *Agronomy J.* 48, 214–318.
53. Sivasubramanian S., Menon M., 1973. Heterosis and inbreeding depression in rice. *Madras Agric. J.* 60, 1139.
54. Robinson H.F., Comstock R.E., Harvey P.H., 1949. Estimates of heritability and degree of dominance in corn. *Agron. J.* 41, 253–259.
55. Fischer R.A., Maurer R., 1978. Drought Resistance in Spring Wheat Cultivars. I Grain Yield Responses. *Aust. J. Agric. Res.* 29, 897–912
56. Choukan R., Taherkhani T., Ghanadha M.R., Khodarahmi M., 2006. Evaluation of drought tolerance in grain maize inbred lines using drought tolerance indices. *Iranian Journal of Crop Sci.* 8(1), 79–89.
57. Bouslama M., Schapaugh W.T., 1984. Stress Tolerance in Soybean. Part 1: Evaluation of Three Screening Techniques for Heat and Drought Tolerance. *Crop Sci.* 24, 933–937. <https://doi.org/10.2135/cropsci1984.0011183X002400050026x>
58. Pujarula V., Pusuluri M., Bollam S., Das R.R., Ratnala R., Adapala G., et al. 2021. Genetic Variation for Nitrogen Use Efficiency Traits in Global Diversity Panel and Parents of Mapping Populations in Pearl Millet. *Front. Plant Sci.* 12, 625915. <https://doi.org/10.3389/fpls.2021.625915> PMID: 33613608
59. Rengel Z., Graham R.D., 1995. Wheat genotypes differ in Zn efficiency when grown in chelate-buffered nutrient solution. *Plant Soil* 176, 307–316. <https://doi.org/10.1007/BF00011795>

60. Worku M., Bänziger M., Erley G.S.A.M., Friesen D., Diallo A.O., Horst W. J., 2007. Nitrogen uptake and utilization in contrasting nitrogen efficient tropical maize hybrids. *Crop Sci.* 47, 519–528. <https://doi.org/10.2135/cropsci2005.05.0070>
61. Raun W.R., Solie J.B., Johnson G.V., Stone M.L., Mullen R.W., Freeman K.W., et al. 2002. Improving nitrogen use efficiency in cereal grain production with optical sensing and variable rate application. *J. Agron.* 94, 815–820. <https://doi.org/10.2134/agronj2002.8150>
62. Rothstein S.J., 2007. Returning to Our Roots: Making Plant Biology Research Relevant to Future Challenges in Agriculture. *Plant cell* 19(9), 2695–2699. <https://doi.org/10.1105/tpc.107.053074> PMID: 17873097
63. Pathak R.R., Lochab S., Raghuram N., 2011. Improving plant nitrogen-use efficiency. *Comprehensive Biotechnology* 4, 209–218. <https://doi.org/10.1016/B978-0-08-088504-9.00472-4>
64. Ali J., Jewel Z.A., Mahender A., Anandan A., Hernandez J., Li Z., 2018. Molecular genetics and breeding for nutrient use efficiency in rice. *Int. J. Mol. Sci.* 19, 1–27. <https://doi.org/10.3390/ijms19061762> PMID: 29899204
65. Hawkesford M.J., Griffiths S., 2019. Exploiting genetic variation in nitrogen use efficiency for cereal crop improvement. *Curr. Opin.* 49, 35–42. <https://doi.org/10.1016/j.pbi.2019.05.003> PMID: 31176099
66. Chen B., Xu K., Li J., Li F., Qiao J., Li H., et al. 2014. Evaluation of yield and agronomic traits and their genetic variation in 488 global collections of *Brassica napus* L. *Genet. Resour. Crop Evol.* 61, 979–999. <https://doi.org/10.1007/s10722-014-0091-8>
67. Rao I.S., Neeraja C.N., Srikanth B., Subrahmanyam D., Swamy K.N., Rajesh K., et al. 2018. Identification of rice landraces with promising yield and the associated genomic regions under low nitrogen. *Sci. Rep.* 8, 9200. <https://doi.org/10.1038/s41598-018-27484-0> PMID: 29907833
68. Sial M.A., Arain M.A., Khanzada S., Naqvi M.H., Dahot M.U., Nizamani N.A., 2005. Yield and quality parameters of wheat genotypes as affected by sowing dates and high temperature stress. *Pak. J. Bot.* 37, 575–584.
69. Belete F., Dechassa N., Molla A., and Tana T., 2018. Effect of nitrogen fertilizer rates on grain yield and nitrogen uptake and use efficiency of bread wheat (*Triticum aestivum* L.) varieties on the Vertisols of central highlands of Ethiopia. *Agric. Food Secur.*, 7, 1–12. <https://doi.org/10.1186/s40066-018-0231-z>
70. Kaur G., Asthir B., Bains N.S., 2015. Nitrogen levels effect on wheat nitrogen use efficiency and yield under field condition. *Afr. J. Agric. Res.* 10(23), 2372–2377.
71. Kitajima K., Hogan K.P., 2003. Increases of chlorophyll a/b ratios during acclimation of tropical woody seedlings to nitrogen limitation and high light. *Plant Cell Environ.* 26, 857–865. <https://doi.org/10.1046/j.1365-3040.2003.01017.x> PMID: 12803613
72. Li Y.Y., Ming B., Fan P.P., Liu Y., Wang K.R., Hou P., et al. 2022. Effects of nitrogen application rates on the spatio-temporal variation of leaf SPAD readings on the maize canopy. *The J. Agric. Sci.*, 160(1–2): 32–44. <https://doi.org/10.1017/S0021859621001052>
73. Pramanik K., Bera A.K., 2013. Effect of seedling age and nitrogen fertilizer on growth, chlorophyll content, yield and economics of hybrid rice (*Oryza sativa* L.). *Int. J. Agron. Plant Prod.* 4, 3489–3499.
74. Paponov I.A., Engels C., 2005. Effect of nitrogen supply on carbon and nitrogen partitioning after flowering in maize. *J. Plant. Nutr. Soil Sci.* 168(4), 447–453.
75. He H., Yang R., Li Y., Ma A., Cao L., Wu X., et al. 2017. Genotypic variation in nitrogen utilization efficiency of oilseed rape (*Brassica napus*) under contrasting N supply in pot and field experiments. *Front. Plant Sci.* 8, 1825. <https://doi.org/10.3389/fpls.2017.01825>
76. Brancourt-Hulmel M., Goussinault D., Lecomte C., Berard P., LeBuanec B., Trottet M., 2003. Genetic improvement of agronomic traits of winter wheat cultivars release in France from 1946 to 1992. *Crop Sci.* 43, 37–45.
77. Muurinen S., Kleemola J., Peltonen-Sainio P., 2007. Accumulation and translocation of nitrogen in spring cereal cultivars differing in nitrogen use efficiency. *Agron. J.* 99, 441–449.
78. Chen C., Han G., He H., Westcott M., 2011. Yield, protein, and remobilization of water soluble carbohydrate and nitrogen of three spring wheat cultivars as influenced by nitrogen input. *Agron. J.* 103, 786–795.
79. Monostori I., Szira F., Tondelli A., Arendas T., Gierczik K., Cattivelli L., et al. 2017. Genome-wide association study and genetic diversity analysis on nitrogen use efficiency in a Central European winter wheat (*Triticum aestivum* L.) collection. *PLoS One* 12, e0189265. <https://doi.org/10.1371/journal.pone.0189265> PMID: 29283996
80. Ivić M., Grljušić S., Plavšić I., Dvojković K., Lovrić A., Rajković B., et al. 2021. Variation for nitrogen use efficiency traits in wheat under contrasting nitrogen treatments in south-eastern Europe. *Frontiers in plant science*, 12, p.682333 <https://doi.org/10.3389/fpls.2021.682333> PMID: 34868096

81. Yousaf A., Atta B. M., Akhter J., Monneveux P., Lateef Z., 2008. Genetic variability, association and diversity studies in wheat (*T. aestivum*) germplasm. Pak. J. Bot. 40, 2087–2097.
82. Kalimullah, Khan S.J., Irfaq M, Rahman H.U., 2012. Genetic variability, correlation and diversity studies in bread wheat (*Triticum aestivum* L.) germplasm. J. Anim. Plant Sci. 22(2), 330–333.
83. Majumder B., Mandal B., Bandyopadhyay P.K., Gangopadhyay A., Mani P.K., Kundu A.L., et al. 2008. Organic amendments influence soil organic carbon pools and rice–wheat productivity. Soil Sci. Soc. Am. J. 72(3), 775–785. <https://doi.org/10.2136/sssaj2006.0378>
84. Nikolic Ana, Andjelkovic V., Dodig D., Mladenović-Drinić S., Kravić N., Ignjatovic-Micic D., 2013. Identification of QTL-s for drought tolerance in maize, II: Yield and yield components. Genetika 45, 141–150. <https://doi.org/10.2298/GENSR1302341N>
85. Rangare N.R., Krupakar A., Kumar A., Singh S.P., 2010. Character association and component analysis in wheat (*Triticum aestivum* L.). Electron. J. Plant Breed. 1(3), 231–238.
86. Baranwal D.K., Mishra V.K., Vishwakarma M.K., Yadav, Punam S., Arun B., 2012. Studies on genetic variability, correlation and path analysis for yield and yield contributing traits in wheat (*T. aestivum* L. em Thell.). Plant Archives 12 (1):99–104.
87. Gorjanovic B.M., Brdar J., Balalic M.K., 2011. Phenotypic variability of bread wheat genotypes for nitrogen harvest index. Genetika 43(2), 419–426.
88. Beche E., Benin G., Lemes C., Munaro L., Marchese J., 2014. Genetic gain in yield and changes associated with physiological traits in Brazilian wheat during the 20th century. Eur. J. Agron. 61, 49–59. <https://doi.org/10.1016/j.eja.2014.08.005>
89. Sarcevic H., Jukic K., Ilic I., Lovric A., 2014. Estimation of quantitative genetic parameters for grain yield and quality in winter wheat under high and low nitrogen fertilization. Euphytica 199, 57–67. <https://doi.org/10.1007/s10681-014-1154>
90. Würschum T., 2012. Mapping QTL for agronomic traits in breeding populations. Theor. Appl. Genet. 125, 201–210. <https://doi.org/10.1007/s00122-012-1887-6> PMID: 22614179
91. Fageria N.K., Baligar V.C., 2003. Methodology for evaluation of lowland rice genotypes for nitrogen use efficiency. J. Plant Nutr. 26, 1315–1333. <https://doi.org/10.1081/PLN-120020373>
92. Bouchet A.S., Nesi N., Bissuel C., Bregeon M., Laripe A., Navier H., et al. 2014. Genetic control of yield and yield components in winter oilseed rape (*Brassica napus* L.) grown under nitrogen limitation. Euphytica 199, 183–205. <https://doi.org/10.1007/s10681-014-1130-4>
93. Ma B.L., Biswas D.K., Herath A.W., Whalen J.K., Ruan S.Q., Caldwell C., et al. 2015. Growth, yield, and yield components of canola as affected by nitrogen, sulfur, and boron application. J. Plant Nutr. Soil Sci. 178, 658–670. <https://doi.org/10.1002/jpln.201400280>
94. Mălinaș A., Vidican R., Rotar I., Mălinaș C., Moldovan C.M. and Proorocu M., 2022. Current status and future prospective for nitrogen use efficiency in wheat (*Triticum aestivum* L.). Plants, 11(2), 2–19. <https://doi.org/10.3390/plants11020217>
95. Fageria N.K., 2014. Nitrogen harvest index and its association with crop yields. J. Plant Nutr., 37, 795–810. <https://doi.org/10.1080/01904167.2014.881855>
96. Nehe A., King J., King I.P., Murchie E.H. and Foulkes M.J., 2022. Identifying variation for N-use efficiency and associated traits in amphidiploids derived from hybrids of bread wheat and the genera *Aegilops*, *Secale*, *Thinopyrum* and *Triticum*. Plos one, 17(4), 1–19. <https://doi.org/10.1371/journal.pone.0266924>