Genome-wide association analysis to delineate high-quality SNPs for seed micronutrient density in chickpea (*Cicer arietinum* L.)

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Chickpea is the most important nutrient-rich grain legume crop in the world. A diverse core set of 147 chickpea genotypes was genotyped with an Axiom®50K CicerSNP array and trait phenotyped in two different environments for four seed micronutrients (Zn, Cu, Fe and Mn). The trait data and high-throughput 50K SNP genotypic data were used for the genome-wide association study (GWAS). The study led to the discovery of genes/QTLs for seed Zn, Cu, Fe and Mn concentrations in chickpea. The analysis of seed micronutrient data revealed significant differences for all four micronutrient concentrations (P ≤ 0.05). The mean concentrations of seed Zn, Cu, Fe and Mn pooled over the 2 years were 65.9 ppm, 63.8 ppm 146.1 ppm, and 27.0 ppm, respectively. The analysis of results led to the identification of 35 SNPs significantly associated with seed Zn, Cu, Fe and Mn concentrations. Among these 35 marker-trait associations (MTAs), 5 were stable (consistently identified in different environments), 6 were major (explaining more than 15% of the phenotypic variation for an individual trait) and 3 were both major and stable MTAs. A set of 6 MTAs, MTAs (3 for Mn, 2 for Fe, and 1 for Cu) reported by us during the present study have been also reported in the same/almost same genomic regions in earlier studies and therefore declared as validated MTAs. The stable, major and validated MTAs identified during the present study will prove useful in future chickpea molecular breeding programs aimed at enhancing the seed nutrient density of chickpea.

Chickpea (*Cicer arietinum* L.; 2n = 2x = 16) is one of the most important grain legume crops in the world and is well known for its health and nutritional benefits. As far as importance is concerned, the crop holds second place with respect to its importance and third place in production among the pulses grown worldwide¹⁰,¹¹. The world population is growing at a very fast rate and is believed to escalate the mark of 9 billion by 2050¹². To meet the dietary demands of this expanding population, it is important to develop nutrient-dense crops for the poorer sections of society. Micronutrient elements such as zinc (Zn), iron (Fe), copper (Cu), magnesium (Mg), phosphorous (P), potassium (K) and manganese (Mn) play vital roles in the metabolism and physiological processes of both plants and humans⁶–⁷. The deficiency of these nutrient elements is caused mainly by non-availability or their low concentrations in the staple diets, resulting in serious health issues for billions of people worldwide⁸. The dietary importance of chickpea is well known across the global human population¹⁰. As far...
as bio-fortification through marker-assisted breeding (MAB) is concerned, the complex genetic inheritance pattern and the gene regulatory functions of micronutrients such as Zn and Fe concentrations are of paramount importance. These micronutrients in chickpea are complex quantitative in nature, showing minor genotypic effects but a higher magnitude of environmental effects\(^1\). The same was also validated in another study where it was demonstrated that genotypic contribution to total variability is limited compared to year and location effects\(^2\). The concentration of these minerals is also influenced by the seed size, seed type (Desi vs Kabuli) and some time by seed coat color\(^3,10\).

Quantitative trait loci (QTLs) for Fe and Zn were reported using a genotyping by sequencing (GBS) approach\(^4\). Biofortification has an outstanding role in increasing the concentration of micronutrients and their bioavailability in crop plants, especially in grain legumes, which have considerable importance in the human diet\(^5\). Regardless of the importance of the process of biofortification, there has been limited research on the development of micronutrient-rich crops over the last half century\(^6\). As far as research on biofortification in legumes is concerned, its focus during the last few decades has been limited to micronutrients such as seed Zn and seed Fe concentrations only and more recently, the focus has been shifted to nutrients such as Cu, P and K\(^5,7\).

Genome-wide association studies (GWASs) are considered one of the paramount and emerging approaches of gene/QTL discovery for the genetic dissection of quantitative traits, including seed micronutrients, in crop plants\(^8,9\). GWAS serves as an alternative method of gene discovery to QTL mapping that involves the use of biparental mapping and eradicates several drawbacks of the latter in the identification of QTLs/gene\(^8\). GWASs have been successfully used in the discovery of genes/QTLs for a variety of traits, including seed micronutrients, in cereals\(^8\) as well as legumes\(^2,5,6,26\). Association-mapping studies have brought insights into how to improve the concentration of mineral nutrients in crops such as rice\(^27\), barley\(^28\), field pea\(^29\), common bean\(^30\) and chickpea\(^31,32\). Recently, in the context of climate change, efforts were also made to identify the effect of drought and heat on the concentration of micronutrients, and MTAs were also reported\(^31\). For GWAS, several genotyping platforms based on SNPs have been developed over the years and have been used successfully for the identification of genes for different traits in crop plants\(^33\). The availability of the draft genome sequence of chickpea\(^33\) and large-scale resequencing efforts\(^34-36\) have led to the discovery of large-scale SNP markers for the development of SNP genotyping platforms such as Axiom®CicerSNP Array\(^37\).

In the present study, we report the marker trait associations (MTAs) for micronutrients using the core set of germplasm and the Axiom®CicerSNP Array. The results of the present study will encourage the effective utilization of chickpea genetic resources, opening possibilities for biofortification programs in chickpea in the near future.

**Results**

**Variability for seed Zn, Cu, Fe and Mn concentrations.** We estimated the concentrations of micronutrients such as Zn, Cu, Fe and Mn in all 147 chickpea seed samples based on 2017–2018 and 2018–2019 as well as pooled samples from both years. We observed a large variation in the concentration of the estimated micronutrients between the years as well as the pooled samples. Based on the samples obtained during 2017–2018, the Zn concentration ranged from 14.56 to 204.39 ppm with a mean of 45.31 ppm, the Cu concentration ranged from 4.23 to 185.47 ppm (mean 60.99 ppm), the Fe concentration ranged from 7.93 to 190.3 ppm (mean 134.24 ppm) and the Mn concentration ranged from 9.07 to 54.29 ppm (mean 25.62 ppm) (Table 1). Similarly, analysis of seed micronutrient concentrations in Year II revealed that the Zn concentration ranged from 9.36 to 205.86 ppm (mean 46.58 ppm), Cu concentration ranged from 3.66 to 190.36 ppm (mean 66.81 ppm), Fe concentration ranged from 100.36 to 127.2 ppm (mean 132.29 ppm) and Mn concentration ranged from 4.86 to 122.86 ppm (mean 28.68 ppm) (Table 1). The mean values of Zn, Cu, Fe and Mn concentrations in Environment III (pooled data) are presented in Fig. 1. Levene’s test conducted for the trait data of two different environments of seed Zn, Cu, Fe and Mn concentrations revealed non-significant differences in variances (P for Zn = 0.48; P for Cu = 0.99; P for Fe = 0.99 and P value for Mn = 0.98) of the trait data of two environments/experiments. Thus, the results indicated that the variances are homogeneous between the two experiments. Combined ANOVA indicated significant interactions (P ≤ 0.5%) between genotype and environment (year) for all four micronutrients (Zn, Cu, Fe and Mn). The mean micronutrient concentrations of the genotypes varied significantly. The maximum coefficient of variation was observed for seed Mn (2.68%), followed by Zn (1.58%), Cu (1.40%) and Fe (0.52%) (Table 2).

**Single nucleotide polymorphism (SNP) distribution on the chickpea genome.** We genotyped the 147 chickpea genotypes with an Axiom® CicerSNP Array comprising 50,590 SNPs, and SNPs were called following the Axiom® best practice workflow\(^29\). We excluded SNPs with ambiguous bases as well as all loci with more than 10% missing data. Furthermore, we also excluded monomorphic SNPs and SNPs with a PIC value < 0.349.

| Value | Zn (ppm) | Cu (ppm) | Fe (ppm) | Mn (ppm) |
|-------|----------|----------|----------|----------|
| Year 2017 | Year 2018 | Pooled data | Year 2017 | Year 2018 | Pooled data | Year 2017 | Year 2018 | Pooled data |
| Min. | 14.5 | 9.3 | 11.9 | 4.2 | 3.6 | 3.9 | 79.3 | 100.3 | 97.1 | 9.0 | 4.8 | 10.1 |
| Max. | 204.3 | 205.8 | 205.1 | 185.4 | 190.3 | 183.9 | 198.2 | 127.2 | 194.5 | 54.2 | 122.8 | 79.0 |
| Avg ± SD | 45.3 ± 26.0 | 46.5 ± 27.4 | 45.9 ± 26.5 | 60.9 ± 40.3 | 66.8 ± 41.8 | 63.8 ± 40.2 | 143.2 ± 29.4 | 132.2 ± 29.0 | 146.1 ± 22.4 | 25.6 ± 8.9 | 28.6 ± 17.7 | 27.0 ± 10.3 |

Table 1. Trait variation in the concentrations of seed Zn, Cu, Fe and Mn concentration in the chickpea core set evaluated during the year 2017–2018, year 2018–2019 and data pooled over environments.
As a result, we obtained a total of 7277 high-quality SNPs that were used for further analysis (Table 3, Fig. 2). On average, we identified 5186 SNPs per pseudomolecule, while the number of SNPs on each pseudomolecule ranged from 119 (Ca8) to 1686 (Ca1) (Table 3). The PIC values of SNPs varied from 0.364 to 0.368. Among the
7277 SNP markers, 920 (12.64%) had a maximum PIC value of 0.375, whereas 200 (2.74%) had a minimum PIC value of 0.350, while the majority of the SNPs 6,157 (84.60%) had a PIC value between 0.350 and 0.375.

**Diversity analysis.** We identified two subpopulations in the chickpea core set based on principal component analysis (PCA) (Fig. 3). Analysis of molecular variance (AMOVA) among the two subpopulations identified based on PCA indicated that the major source of variance was the genotypes, not the populations. The
Significant MTAs for all four micronutrients (P ≥ 10⁻³) on all eight pseudomolecules. The highest number of very good strategy for diet diversification and alleviating micronutrient malnutrition10–13.

One of the major global health issues that need immediate attention is micronutrient malnutrition38. Iron and several abnormalities in important life processes, such as collagen and elastin biosynthesis, iron mobilization, and wound healing40. To overcome or mitigate this problem of micronutrient malnutrition, strategies such as mineral supplementation, dietary diversification and biofortification need to be implemented41. Chickpea is the most important nutrient-rich grain legume crop, and incorporation of this crop in human diets is considered a very good strategy for diet diversification and alleviating micronutrient malnutrition10–13.

Discussion

The MTAs identified in more than one environment were grouped as stable/promising MTAs, while MTAs with greater than 15% phenotypic variation were grouped as major MTAs. Out of the 35 identified associated SNPs, 5 SNPs were stable/promising, 6 MTAs were found to be major MTAs, and 3 MTAs were both stable and major. Among the stable/promising MTAs, 4 MTAs were for Cu (Affx_123261919, Affx_123255840, Affx_123294330 and Affx_123280871), explaining 15.6%, 17.3%, 14.05% and 18.25% of the average phenotypic variation, while no MTAs were found for Fe and Mn. A set of 6 major MTAs (explaining more than 15% of phenotypic variation) were identified, of which 5 MTAs were for Cu (Affx_123259344, Affx_123261919, Affx_123255840, Affx_123280871 and Affx_123280871). The remaining major MTA (Affx_123275255) was identified for Fe, with an average phenotypic variation of 19.9%. Of 5 major MTAs identified for seed Cu concentration, 3 MTAs (Affx_123261919, Affx_123255840 and Affx_123280871) were both stable and major. Affx_123280871 explained the maximum phenotypic variation (18.25%). The list of these markers and their chromosomal location and association with the trait of interest is presented in Table 6.

**Table 4.** Summary of the different genetic diversity parameters used to infer the extent of genetic diversity available in different chickpea sub-populations. N sample size, Na number of different alleles, Ne number of effective alleles, I Shannon’s information index, Ho observed heterozygosity, He expected heterozygosity, %P variation explained, SE standard error.
Recent advances in genomics and next-generation sequencing have enabled the availability of ample genomics resources, making crop genomics resources rich. Conventional QTL mapping approaches to identify genomic regions have been restricted to biparental mapping populations at less resolution. The availability of reliable markers will help in tailoring climate-smart and nutrient-rich varieties that cater to both the demand and health aspects of humans through genomics-assisted breeding programs. The availability of genetic and genomic resources will pave the way for the discovery of useful/stable/major genes/QTLs/markers for seed micronutrients in chickpea through a variety of approaches. A variety of molecular markers have been developed in almost all crop plants and these markers have been used for various purposes including the study of genetic diversity, study of population structure and also in gene discovery programs. However, despite the abundance of different types of markers available, SNPs have witnessed exponential utilization for a variety of diverse applications since their discovery.

Table 5. Details of SNPs showing significant associations with seed Zn, Cu, Fe and Mn concentrations. The table shows the linkage group, position, environment of detection, and phenotypic variation explained (R^2) by the marker-trait associations/SNPs identified during the present study. The table also highlighted the correspondence (validation) of some of the identified SNPs with some earlier identified genes/QTLs/SNPs for seed micronutrients.

| Marker     | Linkage group | Position (current study) | Environment 2017 | Environment 2018 | Pool | P value (range) | R^2 (Range) | Trait | Comparison with earlier studies | Position mapped in earlier studies |
|------------|---------------|--------------------------|------------------|------------------|------|----------------|-------------|-------|-------------------------------|----------------------------------|
| Affx_123247267 | 7             | 29,434,198               | +                | −                | −    | 2.5 × 10^{-4} | 11.6        | Zn    | −                             | −                                 |
| Affx_123295749 | 7             | 40,625,899               | +                | −                | −    | 3.3 × 10^{-4} | 11.3        | Zn    | −                             | −                                 |
| Affx_123241958 | 7             | 42,022,431               | −                | +                | −    | 7.6 × 10^{-4} | 9           | Zn    | −                             | −                                 |
| Affx_123261732 | 4             | 34,010,123               | +                | +                | −    | 4.5 × 10^{-4} | 9.4–10.8    | Zn    | −                             | −                                 |
| Affx_123243695 | 1             | 2,896,714                | −                | +                | −    | 8.3 × 10^{-4} | 9.4         | Zn    | −                             | −                                 |
| Affx_123258734 | 3             | 27,840,598               | +                | −                | −    | 2.4 × 10^{-5} | 14.2        | Fe    | −                             | −                                 |
| Affx_123282040 | 4             | 41,579,601               | −                | +                | −    | 5.8 × 10^{-5} | 12.9        | Fe    | Also reported associated with seed Fe | 41,702,329 and 41,702,981        |
| Affx_123293935 | 6             | 46,127,728               | −                | +                | −    | 5.8 × 10^{-4} | 12.8        | Fe    | −                             | −                                 |
| Affx_123243960 | 6             | 58,400,703               | +                | +                | −    | 3.3 × 10^{-5} | 11.3        | Zn    | −                             | −                                 |
| Affx_123240923 | 6             | 12,583,357               | −                | +                | −    | 3.7 × 10^{-5} | 19.9        | Fe    | −                             | −                                 |
| Affx_123266414 | 7             | 26,226,362               | −                | +                | −    | 1.2 × 10^{-4} | 14.8        | Fe    | −                             | −                                 |
| Affx_123272321 | 1             | 2,899,986                | −                | +                | −    | 2.4 × 10^{-4} | 13.9        | Fe    | Also reported associated with seed Fe | 3,212,701                         |
| Affx_123255008 | 8             | 9,393,627                | +                | −                | −    | 8.9 × 10^{-4} | 12.1        | Fe    | −                             | −                                 |
| Affx_123292401 | 7             | 30,961,497               | −                | +                | −    | 3.8 × 10^{-4} | 4.5–11.3    | Mn    | −                             | −                                 |
| Affx_123261947 | 7             | 3,431,242                | −                | +                | −    | 5.4 × 10^{-6} | 8.1–9.9     | Mn    | −                             | −                                 |
| Affx_123296790 | 3             | 37,944,759               | +                | −                | −    | 4.4 × 10^{-5} | 5.2–11.1    | Mn    | Also reported associated with seed Mn | 37,831,272                        |
| Affx_123293942 | 3             | 16,499,935               | −                | +                | −    | 1.7 × 10^{-5} | 7.9–11.5    | Mn    | −                             | −                                 |
| Affx_123291734 | 4             | 44,921,315               | +                | +                | −    | 4.6 × 10^{-6} | 7.2–11.0    | Mn    | −                             | −                                 |
| Affx_123258734 | 3             | 16,986,881               | −                | +                | −    | 8.0 × 10^{-5} | 10–12.7    | Mn    | −                             | −                                 |
| Affx_123272607 | 7             | 2,896,714                | −                | +                | −    | 2.5 × 10^{-4} | 8.6–11.0    | Mn    | −                             | −                                 |
| Affx_123268837 | 2             | 10,705,659               | −                | +                | −    | 7.7 × 10^{-5} | 5.3–10.3    | Mn    | Also reported associated with seed Mn | 16,018,646                        |
| Affx_123282242 | 1             | 45,882,207               | −                | +                | −    | 3.2 × 10^{-5} | 8.5–10.6    | Mn    | Also reported associated with seed Mn | 43,150,243                        |
| Affx_123280871 | 6             | 58,080,360               | +                | −                | −    | 4.8 × 10^{-5} | 16.7–18     | Cu    | −                             | −                                 |
| Affx_123243430 | 6             | 58,809,225               | +                | +                | −    | 7.1 × 10^{-5} | 12.9–15.2   | Cu    | −                             | −                                 |
| Affx_123280871 | 6             | 58,080,360               | +                | +                | −    | 4.8 × 10^{-5} | 18.1–18.4   | Cu    | −                             | −                                 |
| Affx_123243430 | 6             | 58,809,225               | +                | +                | −    | 7.1 × 10^{-5} | 14.4–15.2   | Cu    | −                             | −                                 |
| Affx_123255840 | 2             | 8,354,799                | +                | +                | −    | 5.8 × 10^{-5} | 15.0–19.6   | Cu    | −                             | −                                 |
| Affx_123299489 | 2             | 31,215,660               | −                | +                | −    | 6.3 × 10^{-5} | 14.5–15.7   | Cu    | −                             | −                                 |
| Affx_123261919 | 1             | 28,692,545               | +                | +                | −    | 2.5 × 10^{-5} | 14.3–16.9   | Cu    | −                             | −                                 |
| Affx_123252076 | 4             | 7,906,273                | +                | −                | −    | 6.4 × 10^{-5} | 13.2–13.3   | Cu    | −                             | −                                 |
| Affx_123268893 | 3             | 15,048,315               | +                | −                | −    | 3.6 × 10^{-5} | 13.0–13.3   | Cu    | −                             | −                                 |
| Affx_123250833 | 5             | 28,265,735               | +                | +                | −    | 9.1 × 10^{-5} | 12.8–13.1   | Cu    | Also reported associated with seed Cu | 28,456,066                        |
| Affx_123251463 | 5             | 1,981,497                | +                | −                | −    | 7.7 × 10^{-5} | 13.9–15.5   | Cu    | −                             | −                                 |
| Affx_123259344 | 8             | 1,362,575                | +                | +                | −    | 5.0 × 10^{-5} | 14.6–16.0   | Cu    | −                             | −                                 |
Figure 4. Circular Manhattan plots of showing marker-trait associations (MTAs) for seed Cu concentrations. The eight different colors represent 8 different chickpea chromosomes. The three concentric circles (each having 8 colors) shows MTAs identified for seed Cu concentrations in two environments and data pooled over environments.

Table 6. Details of the stable, major, and both major & stable marker-trait associations (MTAs) identified during the present study for seed Cu, Zn, and Fe concentration.
Our trait variability results indicated that the genotype × environment interactions were significant, which indicates that the significant proportion of seed micronutrient concentrations in chickpea is dependent on the soil conditions, temperature, precipitation, and various crop management practices and that significant variation also exists between the genotypes. A similar effect of location and year on seed micronutrient concentrations has been reported in rice, field pea, chickpea, common bean and lentil. In the present study, the concentrations of different micronutrients obtained were similar to the results observed in earlier studies. The candidate lines identified for different micronutrients should be used in chickpea breeding programs and would also serve as useful genetic resources for gene discovery and transcriptomic studies. Diversity analysis was conducted to identify molecular genetic variations present in chickpea germplasm, revealing that indigenous chickpea genotypes are more diverse in comparison to exotic ones, which may be due to the genetic bottleneck in exotic lines or a lower number of lines included in the present study. The availability of lines carrying contrasting genetic variations with different genetic parameters will benefit breeders in chickpea breeding programs. The PCA clustered the germplasm set into two groups, indicating the existence of two subpopulations for the set of genotypes used. Similar results have been reported in some other earlier studies in chickpea making use of SSR markers.

We report significant MTAs associated with key micronutrients through SNP data based on Axiom/CicerSNP Array and 2 years of phenotyping of core collections of chickpea. To optimize the results, we utilized 7% high-quality SNPs (PIC > 0.349%) after deep filtration to precisely map and identified 35 highly associated SNPs for four seed micronutrients (9 each with Fe and Mn, 5 for Zn, and 12 for Cu) in chickpea with an average of ~ 9 SNPs/trait. Quantitative trait loci for seed micronutrients have also been identified in different earlier studies in chickpea (2 QTLs for Cu and 24 QTLs for Zn and Fe). In addition, several important genes/QTLs for different micronutrients (Zn, Fe, Cu, Mg, P, K and Mn) have also been reported in some other earlier studies in chickpea. For seed Cu concentration, we identified significant MTAs on LG1, 2, 3, 4, 5, 6 and 8, while as MTAs were identified on LG5 and LG7 only in an earlier study. For seed Zn concentrations, we identified significant MTAs on LG1, 4 and 7, while as in an earlier study, MTAs have been reported on LG4 and LG5. For seed Fe concentration, we identified MTAs on LG1, 3, 4, 5, 6, 7 and 8, while as MTAs have also been reported for seed Fe concentrations on LG1, 4 and LG5 in an earlier study. For Mn concentration, we identified significant MTAs on LG1, 2, 3, 4 and 7, while MTAs have been reported on LG1, 2, 3, 4, 6 and 7 in an earlier study. The comparison of the results revealed that several unique and new MTAs on different chromosomes not reported on these chromosomes in earlier studies have been reported during the present study (Table 5). On the other hand, a set of six MTAs (2 for Fe, 3 for Mn and 1 for Cu) reported by us during the present study have been also reported in the same/almost same genomic regions in earlier studies. Therefore, these MTAs have been declared as validated MTAs. All 35 SNPs identified during the present study were detected on all eight chickpea linkage groups/chromosomes (LG1–LG8). Genomic regions identified in the present study have also been flanked by QTLs responsible for seed micronutrients in chickpea; for example, QTLs for seed Fe were present on Ca1, Ca3, Ca4, Ca5 and Ca7. This could be due to the different genetic backgrounds of the germplasm used in our study. The present study also identified QTLs at different chromosomal positions not reported earlier, suggesting that we have identified novel QTLs linked to seed micronutrients in chickpea. QTLs for all four micronutrients have also been mapped in different crops, including lentil for Zn (on chromosome 1, 2, 3, and 4), and Fe (on chromosome 3, 5 and 6); pea for Zn (on chromosome 2, 3, 5 and 7); and Medicago truncatula for Zn (on chromosomes 1, 2, 3, 4, 5 and 7); Fe (on chromosomes 1, 2, 3, 5 and 7); and Medicago truncatula for Fe (on chromosomes 1, 2, 3, 5, 6 and 7). The markers found to have a significant association with the trait should be used in marker-assisted selection for biofortified chickpea to mitigate the problem of micronutrient malnutrition or hidden hunger.

Conclusions
Our study is the first of its kind that presents the features of a large population of chickpea genotypes in different environments grown in the region of Kashmir, providing an understanding of the influence of environmental and growing conditions on the seed micronutrient concentrations of a crop. This study provides insight into the genetic basis of variability for seed micronutrient concentrations in chickpea and the potential of GWAS in unravelling marker trait associations in economically important crops. Because of their importance, stable MTAs should be recommended for chickpea molecular breeding programs aimed at enhancing the seed micronutrient concentrations of local chickpea landraces preferred by consumers. The stable and major MTAs identified during the present study will prove useful in breeding programs aimed at enhancing seed micronutrient concentrations in chickpea.

Methods
The plant material used, trait data recorded, genotyping done and data analysis conducted in this study has been presented under separate headings below. All the methods in this manuscript were carried out in accordance with relevant guidelines and regulations.

Plant material. We prepared a chickpea core set comprising 147 genotypes (128 Desi, 12 Kabuli and 7 intermediate pea shaped) based on SSR marker data generated on 384 composite sets (Supplementary Table S3). The germplasm lines were obtained from the International Crops Research Institute for Semi-arid Tropics (ICRISAT) Hyderabad, Indian Council of Agricultural Research-Indian Institute of Pulse Research (ICAR-IIPR), Kanpur, and National Bureau of Plant Genetic Resources (NBPGR), New Delhi and Rafi Ahmad Kidwai, RAK College of Agriculture, Rajmata Vijayaraje Scindia Krishi Vishwa Vidyalaya, Sehore, Madhya Pradesh. However, it is...
important to mention that “The material was not collected by us but was procured from the national/international institutions after signing a proper Material Transfer Agreement (MTA) with these institutions”.

**Trait phenotyping.** Trait phenotyping to quantify the seed micronutrient concentration for four micronutrients Zn, Fe, Cu, and Mn was performed using the core set, grown in two replicated field trials arranged in an augmented block design with a planting density of 10 seeds per row per genotype in rows of 4 m long with a row-to-row spacing of 30 cm. The trial was conducted for two consecutive years, 2017 and 2018, in the research fields of the Faculty of Agriculture (FoA), SKUAST-K, Sopore, Kashmir, India. The concentrations of four micronutrients, Zn, Fe, Cu, and Mn were determined following the procedure described by Ref. Data analysis. Analysis of variance. Based on the mixed-effect model for three different replications and 2 years using the statistical functions package of R software according to 5% significance levels, analysis of variance (ANOVA) was conducted to analyze differences in the seed nutrient concentrations. The diverse statistical measures, including minimum, maximum, average value, standard deviation and coefficient of variation (CV), among the diverse accessions were measured. The effect of genotype, environment (year) and their combined effect (G × E) was also analysed. Levene’s test was also conducted to test the homogeneity of variance of genotypes in two different environments/experiments using an established procedure in Microsoft Excel (https://www.youtube.com/watch?v=HJo91wZNY). The test is used to test whether the trait data collected in two different experiments/environments are significantly different. A P value < 0.05 indicates significant differences between the variances (heterogeneous variances), while P > 0.05 indicates nonsignificant differences between the trait values of two experiments (homogeneous variances).

Diversity analysis and population differentiation. Diversity analysis was performed using genotypic data generated by a set of 7277 SNPs on 147 chickpea genotypes. The software program GenAIEx version 6.5 was utilized to statistically analyze the genotypic data, and parameters such as the total number of alleles (Na), the number of effective alleles (Ne) and Shannon’s information index were calculated. To conduct principal coordinate analysis (PCA), the genotypic data (multiple loci and multiple genotypes) were also analyzed by GenAIEx version 6.5 software. For the construction of the PCA plots, a pairwise genetic distance matrix was calculated. The same software program GenAIEx version 6.5 was used to compute the other genetic diversity parameters, such as genetic identity, gene flow (Nm), and pairwise Nei’s unbiased genetic distance. Using the software package Power-Marker version 3.25, we calculated the polymorphic information content (PIC) for each SNP marker locus.

**Genome-wide association analyses.** To identify markers associated with seed Zn, Fe, Cu, and Mn concentrations, genome-wide association study (GWAS) was conducted. GAPIT (The Genome Association and Predic-
tion Integrated Tool) version 3.0, R package (http://zzlab.net/GAPIT) was used for analyzing the trait phenotyping data (concentration of micronutrients analyzed through AAS) and genotyping data (obtained through the use of Axiom Cicer SNP Array). The model used in GWAS was MLM (mixed linear model), as it has proven quite useful for controlling the relatedness and population structure within GWAS. The population structure in MLM is considered a fixed effect, whereas kinship among individuals is incorporated as the variance–covariance structure of the random effect for the individuals64. SNPs showing minor allele frequencies below 0.05 were eliminated from the dataset. In total, 41,489 polymorphic markers were retained for the 147 genotyped accessions. To show the distribution of the P-values for the SNP markers, Manhattan plots were generated (Figs. 4, 5, 6, 7). The significant association between SNP and trait of interest was declared at a threshold value of P (≥ 10^-3).

Figure 5. Circular Manhattan plots of showing marker-trait associations (MTAs) for seed Fe concentrations. The eight different colors represent 8 different chickpea chromosomes. The three concentric circles (each having 8 colors) shows MTAs identified for seed Fe concentrations in two environments and data pooled over environments.
Figure 6. Circular Manhattan plots of showing marker-trait associations (MTAs) for seed Mn concentrations. The eight different colors represent 8 different chickpea chromosomes. The three concentric circles (each having 8 colors) shows MTAs identified for seed Mn concentrations in two environments and data pooled over environments.
Figure 7. Circular Manhattan plots of showing marker-trait associations (MTAs) for seed Zn concentrations. The eight different colors represent 8 different chickpea chromosomes. The three concentric circles (each having 8 colors) shows MTAs identified for seed Zn concentrations in two environments and data pooled over environments.

Declaration. All the methods in this manuscript were carried out in accordance with relevant guidelines and regulations.

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References
1. Kumar, A., Choudhary, A. K., Solanki, R. K. & Pratap, A. Toward marker-assisted selection in pulses: A review. Plant Breed. 130, 297–313 (2011).
2. Varshney, R. K. et al. Achievements and prospects of genomics-assisted breeding in three legume crops of the semiarid tropics. Biotechnol. Adv. 31, 1120–1134 (2013).
3. Godfray, H. C. J. Food security: The challenge of feeding 9 billion people. Science 327, 812–818 (2010).
4. Bohra, A. et al. Genetics-and genomics-based interventions for nutritional enhancement of grain legume crops: Status and outlook. J. Appl. Genet. 56, 151–161 (2015).
5. Esin, O. et al. Association mapping of loci linked to copper, phosphorus, and potassium concentrations in the seeds of C. arietinum and C. reticulatum. Genomics 111, 1873–1881 (2019).
6. Nur, K. et al. Association mapping of magnesium and manganese concentrations in the seeds of C. arietinum and C. reticulatum. Genomics 112(2), 1633–1642 (2020).
7. Nur, K. et al. Identification of SNP markers associated with iron and zinc concentrations in cicer seeds. Curr. Genomics 21(3), 212–223 (2020).
8. Huang, Y. et al. Association mapping of quantitative trait loci for mineral element contents in whole grain rice (Oryza sativa L.). J. Agric. Food Chem. 63, 10885–10892 (2015).
9. Thavarajah, D. & Thavarajah, P. Evaluation of chickpea (Cicer arietinum L.) micronutrient composition: Biofortification opportunities to combat global micronutrient malnutrition. Food Res. Int. 49(1), 99–104 (2012).
10. Bueckert, R. A. et al. Phytic acid and mineral micronutrients in field-grown chickpea (Cicer arietinum L.) cultivars from western Canada. Eur. Food Res. Technol. 233(2), 203–212 (2011).
11. Van den Berg, J. G., Grussak, M. A. & McGee, R. J. Mineral concentrations of chickpea and lentil cultivars and breeding lines grown in the US Pacific Northwest. Crop J. 6(3), 253–262 (2018).
12. Ray, H. et al. Mineral micronutrient content of cultivars of field pea, chickpea, common bean, and lentil grown in Saskatchewan, Canada. Crop Sci. 54(4), 1698–1708 (2014).
13. Fayaz, H. et al. Characterization of chickpea gene pools for nutrient concentrations under agro-climatic conditions of North-Western Himalayas. Plant Genet. Resour. 17(5), 464–467 (2019).
14. Sivasakthi, K. et al. Functional dissection of the chickpea (Cicer arietinum L.) stay-green phenotype associated with molecular variation at an ortholog of Mendel’s gene for cotyledon color: Implications for crop production and carotenoid biofortification. Int. J. Mol. Sci. 20(22), 5562 (2019).
15. Sab, S. Genetic variation and association mapping for 12 agronomic traits in indica rice. BMC Genomics 16(1), 1 (2015).
16. de Valença, A., Bake, A., Brouwer, I. & Giller, K. Agronomic biofortification of crops to fight hidden hunger in sub-Saharan Africa. Glob. Food Sec. 12, 8–14 (2017).
17. Bouis, H. E. & Saltzman, A. Improving nutrition through biofortification: A review of evidence from HarvestPlus, 2003 through 2016. Glob. Food Sec. 12, 49–58 (2017).
18. Mamo, B. E., Barber, B. L. & Steffenson, B. J. Genome-wide association mapping of zinc and iron concentration in barley landraces from Ethiopia and Eritrea. J. Cereal Sci. 60, 497–506 (2014).
19. Lu, Q. et al. Genetic variation and association mapping for 12 agronomic traits in indica rice. BMC Genomics 16(1), 1 (2015).
20. Mir, R. R., Zaman-Allah, M., Sreenivasulu, N., Trethowan, R. & Varshney, R. K. Integrated genomics, physiology and breeding approaches for improving drought tolerance in crops. Theor. Appl. Genet. 125(4), 625–645 (2012).
21. Gupta, P. K., Kulwal, P. L. & Jaialwal, V. Association mapping in crop plants: Opportunities and challenges. Adv. Genet. 85, 109–147 (2014).
22. Leamy, L. J., Zhang, H., Li, C., Chen, C. Y. & Song, B. H. A genome-wide association study of seed composition traits in wild soybean (Glycine soja). BMC Genom. 18(1), 18 (2017).
23. Li, Y. H. et al. Genome-wide association mapping of QTL underlying seed oil and protein contents of a diverse panel of soybean accessions. Plant Sci. 266, 95–101 (2018).
24. García-Oliveira, A. L., Chander, S., Ortiz, R., Menkir, A. & Gedil, M. Genetic basis and breeding perspectives of grain iron and zinc enrichment in cereals. Front. Plant Sci. 9, 937. https://doi.org/10.3389/fpls.2018.00937 (2018).
25. Blair, M. W. et al. QTL for seed iron and zinc concentration and content in a Mesoamerican common bean (Phaseolus vulgaris L.) population. Theor. Appl. Genet. 121, 1059–1070 (2010).
26. Upadhyaya, H. et al. Functional dissection of seed-iron and zinc concentrations in chickpea. Sci. Rep. 6, 24050. https://doi.org/10.1038/srep24050 (2016).
27. Shao, Y. et al. Association mapping of grain color, phenolic content, flavonoid content and antioxidant capacity in dehulled rice. Theor. Appl. Genet. 122, 1005–1016 (2011).
28. Diapari, M., Sinhdu, A., Warkentin, T. D., Bett, K. & Tar'an, B. Population structure and marker-trait association studies of iron, zinc and selenium concentrations in seed of field pea (Pisum sativum L.). Mol. Breed. 35, 1–14 (2015).
29. Blair, M. W., Wu, X., Bhandari, D. & Astudillo, C. Genetic dissection of ICP-detected nutrient accumulation in the whole seed of common bean (Phaseolus vulgaris L.). Front. Plant Sci. https://doi.org/10.3389/fpls.2016.00219 (2016).
30. Diapari, M. et al. Genetic diversity and association mapping of iron and zinc concentrations in chickpea (Cicer arietinum L.). Genome 57(8), 459–468 (2014).
31. Samineni, S. et al. Impact of heat and drought stresses on grain nutrient content in chickpea: Genome-wide marker-trait associations for protein, Fe and Zn. Environ. Exp. Bot. https://doi.org/10.1016/j.envexpbot.2021.104688 (2021).
32. Gupta, P. K., Kulwal, P. L. & Jaiswal, V. Association mapping in plants in the post-GWAS genomics era. Adv. Genet. 104, 75–154 (2019).
33. Varshney, R. K. et al. Draft genome sequence of chickpea (Cicer arietinum) provides a resource for trait improvement. Nat. Biotechnol. 31, 240–246 (2013).
34. Thudi, M. et al. Whole genome resequencing reveals genome-wide variations among parental lines of 16 mapping populations in chickpea (Cicer arietinum L.). BMC Plant Biol. 16, 10 (2016).
35. Thudi, M. et al. Recent breeding programs increased genetic diversity in both desi and kabuli varieties of chickpea (Cicer arietinum L.). Sci. Rep. 6, 38636 (2016).
36. Varshney, R. K. et al. A chickpea genetic variation map based on the sequencing of 3,366 genomes. Nature. https://doi.org/10.1038/s41586-021-04066-1 (2021).
37. Roorikwal, M. et al. Development and evaluation of high density Axiom® Cicer SNP Array for high resolution genetic mapping and applications in chickpea. Plant Biotechnol. J. 16(4), 890–901 (2018).
38. Fletcher, R. J., Bell, I. P. & Lambert, J. P. Public health aspects of food fortification: A question of balance. Proc. Nutr. Soc. 63, 605–614 (2004).
39. Aziz, F. & Chaudhary, K. Life threatening nutritional deficiencies in a dialysis patient. Hemodial. Int. 21, E50–E53 (2017).
40. Fleeer-Greaves, J. H., Sanjeevi, N. & Lee, J. J. Global perspectives on trace element requirements. J. Trace Elem. Med. Biol. 31, 135–141 (2015).
41. White, P. J. & Broadley, M. R. Biofortifying crops with essential mineral elements. Trends Plant Sci. 10, 586–593 (2005).
42. Ibrahim, A. K. et al. Principles and approaches of association mapping in plant breeding. Crop Sci. 53, 212–224 (2020).
43. Mir, R. R., Hiremath, P. J., Riera-Lizarazu, O. & Varshney, R. K. Evolving molecular marker technologies in plants: From RFLPs to GBS. In Diagnostics in Plant Breeding (eds Lubberstedt, T. & Varshney, R. K.) 229–247 (Springer, 2013).
44. Mir, R. R. & Varshney, R. K. Future prospects of molecular markers in plants. In Molecular Markers in Plants (ed. Henry, R. J.) 169–190 (Blackwell Publishing Ltd, 2013).
45. Gupta, P. K., Rustgi, S. & Mir, R. R. Array-based high-throughput dna markers and genotyping platforms for cereal genetics and genomics. In Cereal Genomics II (eds Gupta, P. K. & Varshney, R. K.) (Springer, 2013).
46. Gupta, P. K., Kulwal, P. L. & Mir, R. R. QTL Mapping: Methodology and applications in cereal breeding. In Cereal Genomics II (eds Gupta, P. K. & Varshney, R. K.) (Springer, 2013).
47. Iyagi, S. et al. Plant microRNAs: Biogenesis, gene silencing, web-based analysis tools and their use as molecular markers. J Biotech 35(4), 413 (2019).
48. Iyagi, S. et al. Development and use of miRNA-derived SSR markers for the study of genetic diversity, population structure, and characterization of genotypes for breeding heat tolerant wheat varieties. PLoS ONE https://doi.org/10.1371/journal.pone.023103 (2021).
49. Sihag, P. et al. Discovery of miRNAs and development of heat-responsive miRNA-SSR markers for characterization of wheat germplasm for terminal heat tolerance breeding. Front. Genet. (Plant Genomics). https://doi.org/10.3389/fgene.2021.699420 (2021).
50. Kumar, S., Kumar, M., Mir, R. R., Kumar, R. & Kumar, S. Advances in molecular markers and their use in genetic improvement of wheat. In Physiological, Molecular, and Genetic Perspectives of Wheat Improvement (eds Wani, S. H. et al.) 139–174 (Springer, 2021).
51. Sagane, V. et al. Development and characterization of nitrogen and phosphorus use efficiency responsive genic and miRNA derived SSR markers in wheat. *Heredity*. https://doi.org/10.1038/s41437-022-00506-4 (2022).
52. Garrido-Cardenas, J. A., Mesa-Valle, C. & Manzano-Agugliaro, F. Trends in plant research using molecular markers. *Planta* **247**(3), 543–557 (2018).
53. Norton, G. J. et al. Genome-wide association mapping of grain arsenic, copper, molybdenum and zinc in rice (*Oryza sativa* L.) grown at four international field sites. *PLoS ONE* **9**, e89685 (2014).
54. Upadhyaya, H. D. et al. Genetic dissection of seed-iron and zinc concentrations in chickpea. *Sci. Rep.* **6**, 24050 (2016).
55. Zia-Ul-Haq, M. et al. Nutritional and compositional study of desi chickpea (*Cicer arietinum* L.) cultivars grown in Punjab, Pakistan. *Food Chem.* **105**(4), 1357–1363 (2007).
56. Fayaz, H., et al. Assessment of molecular genetic diversity of 384 chickpea genotypes and development of core set of 192 genotypes for chickpea improvement programs. *Genet. Resour. Crop Evol.* https://doi.org/10.1007/S10722-021-01296-0 (2021).
57. Mir, A. H. et al. SSR markers in revealing extent of genetic diversity and phylogenetic relationships among chickpea core collection accessions for Western Himalayas. *Plant Mol. Biol. Rep.* https://doi.org/10.12103/rs.3.rs-1334460/v1 (2022).
58. Khazaeei, H. et al. Marker–trait association analysis of iron and zinc concentration in lentil (*Lens culinaris* Medik.) seeds. *Plant Genome*. https://doi.org/10.3835/plantgenome2017.02.0007 (2017).
59. Ma, Y. et al. Genome-wide SNP identification, linkage map construction and QTL mapping for seed mineral concentrations and contents in pea (*Pisum sativum* L.). *BMC Plant Biol.* **17**(1), 43 (2017).
60. Sankaran, R. P., Huguet, T. & Grusak, M. A. Identification of QTL affecting seed mineral concentrations and content in the model legume *Medicago truncatula*. *Theor. Appl. Genet.* **119**(2), 241–253 (2009).
61. Singh, D., Chonkar, P. K. & Dwivedi, B. S. *Manual on Soil* (Westville Publishers, 2005).
62. Peakall, R. & Smouse, P. E. *GENALEX 6: Genetic analysis in Excel*. Population genetic software for teaching and research. *Mol. Ecol. Notes* **6**, 288–295 (2006).
63. Liu, K. & Muse, S. V. Powermarker: An integrated analysis environment for genetic marker analysis. *Bioinformatics* **21**, 2128–2129. https://doi.org/10.1093/bioinformatics/bti282 (2005).
64. Zhang, Z. et al. Mixed linear model approach adapted for genome-wide association studies. *Nat. Genet.* **42**, 355–360 (2010).

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Author contributions

R.R.M., H.F. and M.T. conceived the experiments and supervised all work. R.K.V., M.T. and A.C. helped in genotyping. S.T., R.P. helped in micro-nutrient data generation. S.T. and H.F. performed genotypic data analysis. R.R.M., H.F. and M.T. conceived the experiments and supervised all work. R.K.V., M.T. and A.C. helped in genotyping. S.T., R.P. helped in micro-nutrient data generation. S.T. and H.F. performed genotypic data analysis. R.R.M., S.A., A.A.W., M.A.B. and M.T. contributed to data analysis and interpretation of results. M.T., R.P., U.K., A.C. and R.K.V. reviewed and edited the manuscript draft. H.F., S.T., and R.R.M. drafted the manuscript with input from the other co-authors. All authors read and approved the manuscript.

Competing interests

The authors declare no competing interests.

Additional information

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