ORIGINAL RESEARCH

OpenAccess WILEY

Evaluation of resistance and determination of stability of different sugar beet (*Beta vulgaris* **L.) genotypes in rhizomania-infected conditions**

Abazar Rajab[i1](#page-0-0) | **Masoud Ahmad[i2](#page-0-1)** | **Mohsen Bazrafsha[n3](#page-0-2)** | **Mehdi Hassani[4](#page-0-3)** | **Ali Saremirad[1](#page-0-0)**

1 Sugar Beet Seed Institute (SBSI), Agricultural Research, Education and Extension Organization (AREEO), Karaj, Iran

2 Khorasan Razvi Agricultural and Natural Resources Research and Education Center, Agricultural Research, Education and Extension Organization (AREEO), Mashhad, Iran

3 Fars Agricultural and Natural Resources Research and Education Center, Agricultural Research, Education and Extension Organization (AREEO), Shiraz, Iran

4 Hamedan Agricultural and Natural Resources Research and Education Center, Agricultural Research, Education and Extension Organization (AREEO), Hamedan, Iran

Correspondence

Abazar Rajabi, Sugar Beet Seed Institute (SBSI), Agricultural Research, Education and Extension Organization (AREEO), Karaj, Iran. Email: a.rajabi@areeo.ac.ir

Abstract

Plant diseases are considered one of the main factors reducing yield and quality of crops, which are constantly developing and creating more virulent races and cause the resistance of more genes to break. Identifying resistance sources and including them in breeding programs will improve resistant genotypes. Rhizomania is the most common, widespread, and devastating disease of sugar beet in Iran and worldwide. Breeding genotypes with disease resistance genes is one of the most important ways to deal with this destructive disease. Twenty sugar beet genotypes along with five controls were evaluated in a randomized complete block design with four replications in rhizomania-infected conditions in four regions of Mashhad, Shiraz, Miandoab, and Hamedan for 2 years. The results of genotypic reaction to rhizomania showed that the genotypes with resistance reaction were much more frequent than those with susceptibility reaction. The analysis of multiplicative effects of the AMMI model showed that the first six components were significant and explained 98.80% of the interaction variations. The biplot obtained from the mean white sugar yield and the first interaction principal component confirmed the superiority of the RM5 genotype due to its high white sugar yield and stability in infected conditions. The results obtained from the first three principal components biplot showed that the RM9 genotype with a mean white sugar yield of 11.91 t. ha⁻¹ was a genotype with vast general stability in all disease-infected environments. Based on the results of the MTSI index, RM3, RM17, RM9, RM13, and RM15 are introduced as stable genotypes under rhizomaniainfected conditions. In conclusion, it seems that the studied genotypes have valuable and useful genes inherited from their parents to deal with rhizomania disease. Applying these genotypes in sugar beet breeding programs can effectively prevent the threat of rhizomania.

KEYWORDS

adaptation, gene, infection, race, susceptibility

This is an open access article under the terms of the [Creative Commons Attribution](http://creativecommons.org/licenses/by/4.0/) License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited.

© 2022 The Authors. *Food Science & Nutrition* published by Wiley Periodicals LLC.

1 | **INTRODUCTION**

According to United Nations [\(2019](#page-11-0)) investigations, the world population is expected to increase from 7.7 billion people in 2019 to 8.5 in 2030, 9.7 in 2050, and 10.9 in 2100. Naturally, with the growth of population, the demand for food will increase significantly. Sugar is a common molecule that has been of special nutritional importance in all eras (Eggleston, [2019](#page-10-0)). This molecule acts as a tonic matter and provides a large part of the energy in the human diet. Sugar beet is an important agricultural product which is exclusively used in the sugar industry (Akyüz & Ersus, [2021](#page-10-1)) and is considered one of the most important sources of sugar production after sugar cane (Monteiro et al., [2018](#page-10-2); Ribeiro et al., [2016](#page-11-1)) so that currently, it accounts for 20%–30% of the world sugar production (Iqbal & Saleem, [2015](#page-10-3); Monteiro et al., [2018](#page-10-2); Ribeiro et al., [2016](#page-11-1)). Sugar is the main product of sugar beet, but, in addition, many other by-products such as molasses, marc, and ethyl alcohol are also extracted from this plant during the sugar production process (Tomaszewska et al., [2018](#page-11-2)). In addition, the leaf of this plant has protein compounds (Lammens et al., [2012](#page-10-4); Tenorio et al., [2017](#page-11-3)) and balanced amino acids (Akyüz & Ersus, [2021](#page-10-1); Kiskini et al., [2016](#page-10-5)), which reveals the nutritional quality of sugar beet leaves. Therefore, considering the importance of sugar beet in the nutrition of human societies, it is necessary to pay attention to its yield and quality.

Agricultural production is influenced by biotic and abiotic stresses. These factors challenge the plant's quantitative and qualitative production and destroy a considerable part of agricultural products annually. Diseases and pests are among the most important biotic stresses that affect both quantitative and qualitative aspects of plant products (Abdallah et al., [2003](#page-10-6); Oerke, [2006](#page-10-7)). Sugar beet (*Beta vulgaris* L.) is no exception to this rule and is influenced by pathogens. Rhizomania is one of the most common diseases of sugar beet in the world (Galein et al., [2018](#page-10-8)), which can cause a sharp decrease in yield and product quality (Rezaei, [2007](#page-11-4)). The disease is caused by *Beet Necrotic Yellowing Vein Virus* (BNYVV) and its vector is *Polymyxa betae* Keskin (Tamada, [1975](#page-11-5); Tamada & Baba, [1973](#page-11-6)), which has potentially been a destructive disease of sugar beet on a global scale (McGrann et al., [2009](#page-10-9)) in such a way that the amount of damage caused by it may lead to the destruction of sugar beet fields on a large scale. The first report of the disease published in the world was related to northern Italy in 1952 (Pavli et al., [2011](#page-10-10)). This disease in Iran was reported for the first time in Fars province by Izadpanah et al. [\(1996](#page-10-11)); subsequently, its existence was proved in sugar beet fields of most parts of the country (Arjmand & Ahun Manesh, [1996](#page-10-12)). Of course, the spread of rhizomania is not unique to Iran, and currently, this disease is known as the most important cause of damage in sugar beet fields in most countries of the world (Galein et al., [2018](#page-10-8); Lennefors, [2006](#page-10-13); Tamada, [1999\)](#page-11-7). Therefore, if effective methods are not used to control it, it will significantly reduce crop yield and quality (Rezaei, [2007](#page-11-4)).

Genetic, chemical, agronomical, and biological control methods have been presented against rhizomania disease. Among the mentioned methods, agronomical and biological methods are not very important due to their low ability to control the disease and compared to the genetic and chemical control methods, they are less used. Control of the disease by chemical compounds is not always practical due to the high cost, the risk of the developing of pathogen's resistance to it, and the possibility of environmental pollution; On the other hand, due to the soil-borne nature of rhizomania disease and the ineffectiveness of conventional methods of combating soilborne diseases, the pioneers of breeding disease-resistant varieties consider genetic resistance as the most effective way to reduce the damage caused by such diseases. In this regard, it is most important to investigate the amount of genetic diversity in cultivars and lines and carry out genetic studies to identify and select disease-resistant genotypes; Therefore, the present study was conducted to evaluate sugar beet genotypes in terms of resistance to rhizomania disease and yield stability to use resistant genotypes in breeding programs and also to introduce them for cultivation in infected environments.

2 | **MATERIALS AND METHODS**

2.1 | **Plant materials and details of experiments**

A set of 20 genotypes along with domestic resistant control of SBSI038 and four foreign-resistant controls of Lexia, Baloo, Poseidon, and BTS335 formed the present study's genetic material (Table [1](#page-2-0)). The plant materials were cultivated in a randomized complete block design with four replications in each of the four agricultural research stations of Mashhad, Shiraz, Miandoab, and Hamedan, Iran. These research stations are naturally infected with rhizomania disease. To ensure the infection of the experiments with rhizomania, the Sharif susceptible variety was cultivated around the experiments. The specifications of the research stations are presented in Table [2.](#page-2-1)

2.2 | **Disease assessment and measurement of quantitative and qualitative root traits**

The disease severity was recorded at harvest stage according to the Luterbacher et al. ([2005](#page-10-14)) method on a scale of 1–9 at two agricultural research stations in Mashhad and Shiraz. The scores 1 and 9 represent the lowest and highest disease infection, respectively. Although the experiments were also conducted at Miandoab and Hamadan agricultural research stations under disease-infected conditions, the data on the infection severity were not recorded in these two stations.

After harvesting and recording of disease severity and root yield, the roots were washed and a pulp sample was randomly prepared from the roots of each plot; the pulp samples were then examined at the quality control laboratory for quality characteristics including sugar content, alpha-amino N, sodium (Na⁺), and potassium (K⁺) elements (Kunz et al., [2002](#page-10-15)). Finally, the obtained values were used to estimate other characteristics such as sugar yield, molasses sugar, **TABLE 1** Specifications of the studied sugar beet genotypes

Genotype code	Origin	Genotype code	Origin	Genotype code	Origin
RM ₁	F-21076	RM10	F-21085	RM19	F-21095
RM ₂	F-21077	RM11	F-21086	RM20	F-21096
RM ₃	F-21078	RM12	F-21087	SBSI038	Domestic check
RM4	F-21079	RM13	F-21088	Lexia	Foreign check
RM ₅	F-21080	RM14	F-21089	Baloo	Foreign check
RM ₆	F-21081	RM15	$F - 21090$	Poseidon	Foreign check
RM7	F-21082	RM16	F-21091	BTS335	Foreign check
RM ₈	F-21083	RM17	F-21092	$\overline{}$	$\overline{}$
RM ₉	F-21084	RM18	F-21093	٠	۰

TABLE 2 Geographical characteristics of the experimental research stations

white sugar content, white sugar yield, and extraction coefficient of sugar, based on Equations [1 to 5](#page-2-2), respectively (Cook & Scott, [1993](#page-10-16); Reinfeld et al., [1974](#page-11-8)).

$$
SY = RY \times SC \tag{1}
$$

$$
MS = 0.0343(K^{+} + Na^{+}) + 0.094 \text{(alpha amino N)} - 0.31 \qquad (2)
$$

$$
WSC = SC - (MS + 0.6)
$$
 (3)

$$
WSY = WSC \times RY \tag{4}
$$

$$
ECS = \left(\frac{\text{WSC}}{\text{SC}}\right) \times 100\tag{5}
$$

where SY is sugar yield (t. ha⁻¹), RY is root yield (t. ha⁻¹), SC is sugar content (%), MS is molasses sugar (%), K⁺ is potassium (meq.100 g^{-1}), Na⁺ is sodium (meg.100 g^{-1}), alpha-amino-N is nitrogen (meg.100 g^{-1}), WSC is white sugar content (%), WSY is white sugar yield (t. ha⁻¹), and ESC is extraction coefficient of sugar (%).

2.3 | **Statistical analysis**

Before any analysis, the homogeneity of the variances of experimental errors was checked with Bartlett's test (Bartlett, [1937\)](#page-10-17). After the homogeneity of error variances was confirmed, a combined variance analysis was performed on each trait's data. Since the white sugar yield includes the values of other studied traits, it is considered a significant and final trait; Therefore, Additive Main Effect and Multiplicative Interactions (AMMI) stability analysis was performed

in terms of this trait. Equation [6](#page-2-3) was used to perform stability analysis by the AMMI method (Gauch, [1992\)](#page-10-18):

$$
Y_{ge} = \mu + \alpha_g + \beta_e + \sum_{n} \lambda_n \gamma_{gn} \delta_{en} + \rho_{ge}
$$
 (6)

Where Y_{ge} is the yield of genotype *g* in environment *e*; μ is the grand mean; α_g is the genotype deviation from the grand mean; β_e is the environment deviation; λ_n is the singular value for IPC_n and correspondingly λ_n^2 is its eigenvalue; $\gamma_{\rm gn}$ is the eigenvector value for genotype g and component n ; δ_{en} is the eigenvector value for environment *e* and component *n*, with both eigenvectors scaled as unit vectors; and ρ_{ge} is the residual. By performing AMMI analysis of variance using R software, the eigenvalues were obtained for each genotype and environment, and by drawing their biplots, the general and specific adaptability of the genotypes was determined. During this study, 13 statistics obtained from the AMMI model were used to identify the stable genotype in disease-infected conditions through Equations 7 to 18:

$$
AMGE = \sum_{n=1}^{N} \lambda_n \gamma_{in} \delta_{jn}
$$
 (7)

$$
ASI = \sqrt{\left[PC_1^2 \times \theta_1^2\right] + \left[PC_2^2 \times \theta_2^2\right]}
$$
 (8)

$$
ASV = \sqrt{\left[\frac{SSIPCA1}{SSIPCA2} (IPCA_{1score)}\right]2 + (IPCA2score)2}
$$
 (9)

$$
\text{STAB} = \sum_{n=1}^{N'} \lambda_n \gamma_{in}^2 \tag{10}
$$

$$
AV_{AMGE} = \sum_{j=1}^{E} \sum_{n=1}^{N'} \left| \lambda_n \gamma_{in} \delta_{jn} \right| \tag{11}
$$

$$
Da = \sqrt{\sum_{n=1}^{N} (\lambda_n \gamma_{in})^2}
$$
 (12)

$$
D_Z = \sum_{n=1}^{N'} \gamma_{in}^2
$$
 (13)

$$
EV = \sum_{n=1}^{N} \frac{\gamma_{in}^2}{n}
$$
 (14)

$$
FA = \sum_{n=1}^{N'} \lambda_n^2 \gamma_{in}^2
$$
 (15)

$$
MASI = \sqrt{\sum_{n=1}^{N'} PC_n^2 \times \theta_1^2}
$$
 (16)

$$
MASV = \sqrt{\sum_{n=1}^{N'} \left(\frac{SSIPC_n}{SSIPC_{n+1}}\right) \times (PC_n)^2 + (PC_{N'})^2}
$$
(17)

$$
SIPC_i = \sum_{n=1}^{N} | \lambda_n^{0.5} \gamma |
$$
 (18)

$$
Z_{\rm a} = \sum_{\rm n=1}^{\rm N'} |\theta_{\rm n} \gamma_{\rm in}| \tag{19}
$$

where AMGE is the sum across environments of GEI modeled by AMMI (Sneller et al., [1997](#page-11-9)), ASI is the AMMI stability index (Jambhulkar et al., [2014](#page-10-19)), ASV is the AMMI stability value (Purchase et al., [2000](#page-11-10)), ASTAB is AMMI-based stability parameter (Rao & Prabhakaran, [2005](#page-11-11)), AV_{AMGE} is the sum across environments of the absolute value of GEI modeled by AMMI (Zali et al., [2012](#page-11-12)), Da is Annicchiarico's D parameter (Annicchiarico, [2002](#page-10-20)), Dz is Zhang's D parameter (Zhang et al., [1998\)](#page-11-13), EV is the average of the squared eigenvector values (Zobel, [1994\)](#page-11-14), FA is the stability measure based on fitted AMMI model (Raju, [2002](#page-11-15)), MASI is the modified AMMI stability index (Ajay et al., [2018](#page-10-21)), MASV is the modified AMMI stability value (Zali et al., [2012](#page-11-12)), SIPC is the sum of the absolute values of the IPC scores (Sneller et al., [1997](#page-11-9)), and Za is the absolute value of the relative contribution of IPCAs to the interaction (Zali et al., [2012](#page-11-12)).

MTSI index was computed to calculate the mean yield and simultaneous stability of root yield, sugar yield, sugar content, white sugar yield, white sugar content, sodium, potassium, alpha-amino nitrogen, extraction coefficient of sugar, and molasses sugar based on Equation 20 (Olivoto et al., [2019](#page-10-22)):

MTSI_i =
$$
\left[\sum_{j=1}^{f} (y_{ij} - y_j)^2\right]^{0.5}
$$
 (20)

where MTSI*ⁱ* is the multi-trait stability index of the genotype *i*, *𝛾ij* is the score of the genotype i in the factor j , and γ_j is the score of the ideal

genotype in the factor *j*. Scores were calculated based on factor analysis for genotypes and traits.

3 | **RESULTS AND DISCUSSION**

3.1 | **Assessment of genotypic response to disease**

The results of examining the reaction of genotypes against rhizomania disease based on the Luterbacher et al. ([2005](#page-10-14)) method are given in Table [3](#page-4-0). Examining the frequency of sensitivity/resistance of genotypes against rhizomania in each of the research stations for 2 years showed that the genotypes are different in terms of reaction to this disease; the majority of them had a response in the semi-resistant to semi-sensitive range. According to the evaluation of genotypes against the disease in Mashhad in 2018, five of the genotypes under study, including RM1, RM7, RM10, RM14, and RM19, along with two controls, Baloo and BTS335, had healthy roots and no hairy root or color change; in other words, they had a complete resistance reaction. Therefore, these genotypes can carry genes that are involved in the observed resistance. Regarding the reaction of genotypes to rhizomania in Mashhad in 2019, only one genotype (RM17) showed a complete resistance reaction, and the other genotypes along with the controls showed a range of semi-resistant to semi-sensitive reactions. As is evident in the results, the genotypes that were completely resistant in 2018 showed different degrees of semi-resistant to semi-sensitive reaction in 2019. The reason can be attributed to the environmental conditions and the development of new isolates of the disease, which caused the absence of complete resistance of genotypes. Based on the results obtained from the infection severity of the genotypes to the disease in Shiraz, none of the studied genotypes had a complete resistance reaction to the disease during the 2 years of the experiment, and they had a reaction in the semi-resistant to semi-sensitive range. What is certain is that during both years, the infection severity of genotypes in Shiraz was higher than that in Mashhad. Therefore, it can be acknowledged that the environmental conditions in Shiraz for the development and establishment of more virulent isolates of rhizomania were more favorable than the environmental conditions in Mashhad; Therefore, genotypes should be recommended for cultivation in Shiraz that have more effective resistance genes against the disease to prevent its development and causing heavy damages due to the reduction of sugar content and root yield.

3.2 | **Combined analysis of variance**

Bartlett's test (Bartlett, [1937\)](#page-10-17) confirmed the uniformity of error variances in different trials. Therefore, to determine the genotypexenvironment interaction on the data obtained from the root yield, sugar yield, sugar content, white sugar content, white sugar yield, $Na^+, K^+,$ alpha-amino N, molasses sugar, and extraction coefficient of sugar, a combined analysis of variance was performed (Table [4](#page-4-1)). The main effects of year, location, and genotype for all the mentioned traits **TABLE 3** Reaction of sugar beet genotypes against rhizomania in each of the agricultural research stations

TABLE 4 Combined analysis of variance for the studied traits of sugar beet genotypes

Note: *, **Significant at 5% and 1% probability levels, respectively.

Abbreviation: ns, nonsignificant.

were significant at 1% probability level. Two-way interactions of year × location (except for K+), year × genotype, and location × genotype were significant for all studied traits at 1% probability level. The three-way interaction of genotype × year × location was significant at 1% and 5% probability levels for all studied traits. The significance of the interactions is due to the large variations in genotypes across the years and locations under investigation, as well as variation in the relative rank of genotypes.

3.3 | **AMMI analysis**

According to the AMMI model results (Table [5](#page-5-0)), the genotype × year × location interaction for white sugar yield was significant at 1% probability level. The percentage of variance explained by genotype × year × location interaction was equal to 10.65%. In a study conducted on maize (*Zea mays* L.) using the AMMI method, the variance explained by the genotypexenvironment interaction was estimated as 7.84% (Basafa & Taherian, [2016](#page-10-23)). The multiplicative effect of AMMI model was decomposed into interaction principal components. Based on these results (Table [5](#page-5-0)), the first five components had a significant interaction at 1% probability level, and the sixth component had a significant interaction at 5% probability level. The first component explained 45.50% of the interaction variations. The second to sixth principal components were able to explain 19.20%, 14.60%, 9.30%, 5.40%, and 4.80% of the variations related to the genotypexyearxlocation interaction, respectively. These components together with the first component explained 98.80% of the total variations of genotypexyearxlocation interaction. The residual sum of squares (Noise) from the AMMI model with the lowest mean of square was found to be nonsignificant, which indicates the considerable accuracy of this model (Anandan & Eswaran, [2009\)](#page-10-24). In a study conducted using the AMMI model, Mostafavi and Saremirad ([2021](#page-10-25)) stated that the first principal component of the interaction was significant and explained about 63% of the data variation. Karimizadeh et al. ([2008](#page-10-26)) showed that the five main components of the interaction explained 90.30% of the variations of genotype \times environment interaction. Omrani et al. ([2019](#page-10-27)) showed that the first four components together explained 83% of the genotype × environment interaction variations.

To consider yield stability and specific adaptation of genotypes to the studied areas, the biplot of white sugar yield with the first principal component (Figure [1a](#page-6-0)) and a biplot of the two first principal components (Figure [1b](#page-6-0)) was used. Also, for maximum certainty, the biplot of the first three main components explaining about 80% of the variations was used. According to the biplot of average white

TABLE 5 Analysis of variance of genotype × environment interaction for white sugar yield of sugar beet genotypes based on AMMI model

Note: ns, *, **: nonsignificant and significant at 5% and 1% probability levels, respectively.

sugar yield versus the first principal component of the interaction, the genotype that has a higher amount in terms of white sugar yield (horizontal axis) and a lower value in terms of the first component of the genotype \times environment interaction (vertical axis) will be more favorable. Based on this, among the genotypes, RM5, and among the environments, Mashhad in 2019 were recognized as the most stable genotype and environment, respectively, due to its white sugar yield being higher than the total average and the low value of the first interaction component. If the genotype and environment have the same sign in terms of the first component of the interaction, they will have a positive interaction, and if they do not have the same sign in terms of the mentioned component, they will have a negative interaction. Mashhad and Miandoab environments with white sugar yields above the total average had positive interaction with RM13, RM2, RM12, RM11, RM19, Lexia, RM17, RM16, RM14, RM20, RM9, Baloo, RM10, and RM3 and negative interaction with RM4, RM6, RM7, SBSI038, RM18, RM8, Poseidon, BTS335, RM1, RM5, and RM15. The environments of Shiraz and Hamadan had a situation opposite to that of Mashhad and Miandoab so that they showed a similar interaction. These environments interacted positively with RM4, RM6, RM7, SBSI038, RM18, RM8, Poseidon, BTS335, RM1, RM5, and RM15 and negatively with RM13, RM2, RM12, RM11, RM19, Lexia, RM17, RM16, RM14, RM20, RM9, Baloo, RM10, and RM3.

Figure [1b](#page-6-0) shows the values of the first and second principal components of the genotype \times environment interaction for genotypes and environments. A total of 64.70% of the variations related to the multiplicative effect was explained by this biplot. According to this biplot, the genotypes that are located in the vicinity of a place have specific adaptation to that environment, whereas the genotypes that are close to the origin of the coordinates have general adaptation. Accordingly, there is considerable specific adaptation between RM13 with Mashhad (2018) and Miandoab (2018), RM5 with Mashhad (2019), RM11 and RM10 with Miandoab (2019), RM15 with Hamedan (2019), and SBSI038 with Hamedan (2018), Shiraz (2018) and Shiraz (2019). RM3 followed by RM15, BTS335, and RM9 had general adaptation due to being close to the origin of coordinates. To ensure the maximum reliability of the results, the biplot of the first three principal components (Figure [1c](#page-6-0)) was also used to identify genotypes with high general adaptability, justifying about 80% of the variations. Based on the results of this biplot, RM9 with an average white sugar yield of 11.91 t . ha⁻¹ was recognized as a genotype with wide general adaptability because it had values of three components close to zero. None of the environments under study had values of the first, second, and third components of the genotype × environment interaction close to zero (origin of coordinates), which indicates that the environments have the potential to create interaction.

The results of the average white sugar yield of genotypes and different stability statistics of AMMI analysis can be seen in Table [6.](#page-7-0) The average white sugar yield of genotypes in all environments was estimated at 12.24 t. ha−1. RM13 and RM5 had the highest white sugar yields of 14.44 and 14.31 t . ha⁻¹, respectively. The lowest white sugar yield belonged to SBSI038 and Baloo with

FIGURE 1 (a) Scatter plot for genotypes and environments derived from yield means and a first principal component; (b) Scatter plot for genotypes and environments derived from the first two interaction principal components; (c) Scatter plot for genotypes and environments derived from the first three interaction principal components.

[|] 1409

TABLE 6 White sugar yield mean and different AMMI stability parameters in the studied sugar beet genotypes **TABLE 6** White sugar yield mean and different AMMI stability parameters in the studied sugar beet genotypes

Fitted AMMI Model; MASI, Modified AMMI Stability Index; MASV, Modified AMMI Stability Value; SIPC, Sums of the Absolute Value of the IPC Scores; Za, Absolute Value of the Relative Contribution of Fitted AMMI Model; MASI, Modified AMMI Stability Index; MASV, Modified AMMI Stability Value; SIPC, Sums of the Absolute Value of the IPC Scores; Za, Absolute Value of the Relative Contribution of Environments of Absolute Value of GEI Modeled by AMMI; DA, Annicchiarico's D Parameter; DZ, Zhang's D Parameter; EV, Averages of the Squared Eigenvector Values; FA, Stability Measure Based on Environments of Absolute Value of GEI Modeled by AMMI; DA, Annicchiarico's D Parameter; DZ, Zhang's D Parameter; EV, Averages of the Squared Eigenvector Values; FA, Stability Measure Based on IPCAs to the Interaction. IPCAs to the Interaction.

values of 8.78 and 9.93 $t.$ ha⁻¹, respectively. The white sugar yield of RM20, RM11, and RM3 was within the average white sugar yield of all genotypes in all environments with values of 12.16, 12.14, and 12.43t. ha⁻¹, respectively. According to the AMGE stability index, the highest stability with a lower amount of this statistic was observed in RM18. Based on ASI and MASI statistics, RM3 was the most stable with the lowest values for these two statistics. Meanwhile, RM4 was known as the most unstable genotype. The results obtained using ASTAB, AVAMGE, DA, DZ, EV, FA MASV, SIPC, and Za statistics indicated that RM9 and BTS335 with the lowest values for these statistics were the most stable genotypes. Based on the mentioned statistics, RM6, RM4, Baloo, and RM20 with the highest values were identified as the most unstable genotypes. The results of the present study were somewhat similar to the findings of Cheloei et al. ([2020](#page-10-28)); Karimizadeh et al. ([2016](#page-10-29)); and Sharifi et al. ([2017](#page-11-16)). They acknowledged that the most accurate model in the AMMI decomposition could be predicted using the first two principal components. Despite the different stability analysis methods, the AMMI model provides useful information to achieve accurate results (Sharifi et al., [2017](#page-11-16)). Based on the results of the present study, the majority of stable genotypes based on different AMMI stability statistics had an average white sugar yield around the total average. Meanwhile, Ajay et al. ([2020](#page-10-30)) reported that according to these 12 AMMI stability statistics (especially SIPC, MASI, and MASV statistics), high-yielding genotypes can be identified.

3.4 | **MTSI analysis**

Factor analysis was done based on principal component analysis, and interpretation of results was performed after Varimax rotation. The results of factor analysis are presented in Table [7](#page-8-0). Factors with eigenvalues greater than 1 were selected, and the variance of each factor was expressed as a percentage, which indicates its importance in interpreting the overall variations in the data. In this analysis, three independent factors explained 90.83% of data variation. The first factor explained 42.79% of data variance and had an eigenvalue of 3.85. This factor had high and negative coefficients for white sugar content, Na^+ , K^+ , sugar content, extraction coefficient of sugar, and molasses sugar. The second factor with an eigenvalue of 2.56 and justification of 28.53% of total variance, included high and positive factor coefficients for root yield, white sugar yield, and sugar yield. The third factor explained 12.57% of data variation, and with an eigenvalue of 1.13 displayed a high and negative factor coefficient for alpha-amino N. The MTSI stability index of the studied genotypes was calculated based on factor scores of the three mentioned factors.

According to the MTSI index, if the genotype value is less than this index, it is less distant from the ideal genotype. On the other hand, if the MTSI value is higher for the genotype, it means that it is more distant from the ideal genotype and should not be selected (Olivoto et al., [2019\)](#page-10-22). In Figure [2a](#page-9-0), the experimental genotypes are

TABLE 7 Eigenvalues, relative and cumulative variance as well as factor coefficients after varimax rotation in factor analysis based on principal component analysis

ranked from the highest to the lowest value of the MTSI index so that the genotype with the highest value of MTSI is in the center and the genotype with the lowest value of MTSI is located in the outermost circuit. Based on this, by applying a selection pressure of 20%, RM3 ranked first and RM17, RM9, RM13, and RM15 ranked next as the most ideal stable genotypes in terms of all traits. Comparison of the value of traits in the selected genotypes based on MTSI with other genotypes showed that the mean value of root yield, sugar yield, white sugar yield, white sugar content, and extraction efficiency of sugar has been increased in selected genotypes. This increase in the value of traits was aimed at the intended goals. The goal that is followed in Na⁺, K^+ , alpha-amino N, and molasses sugar is to reduce their value, and the selected genotypes showed have a lower value in terms of these traits. In general, the selected genotypes caused a favorable selection differential in all traits except sugar content and white sugar content (Table [8\)](#page-9-1). In the selected genotypes, all traits had high heritability. SBSI038 genotype had the highest value of the MTSI stability index.

Figure [2b](#page-9-0) presents the strengths and weaknesses of the selected genotypes based on the contribution of each factor in the MTSI index. The smallest the proportion explained by a factor (closer to the external edge), the closer the traits within that factor are to the ideotype (Olivoto et al., [2021](#page-10-31)). The dashed line shows the theoretical value if all the factors had contributed equally (Olivoto et al., [2021](#page-10-31)). RM3 and RM15, which had the lowest value in the first factor for sugar content, Na^+ , K^+ , white sugar content, extraction coefficient of sugar, and molasses sugar, and the highest factor coefficients in this factor, are close to the ideal genotype. The ideal genotype is defined according to the traits contained in each factor and the goals that are intended to improve those traits. RM15, RM17, RM3, and

FIGURE 2 (a) Ranking of genotypes in ascending order based on MTSI index and (b) Strengths and weaknesses of selected genotypes as the ratio of each factor in the calculated MTSI index.

Abbreviations: SD, selection differential; SD perc, selection differential in percentage; SG, selection gain; SG perc, selection gain in percentage; Xo, original value; Xs, selected value.

RM9 had the lowest contribution to the second factor. As a result, these genotypes were very close to the ideal genotype in terms of root yield, white sugar yield, and sugar yield. In other words, these genotypes had high values of root yield, sugar yield, and white sugar yield. RM13, RM17, and RM9 had high strength to the third factor. In other words, they had low alpha-amino N values and stability. These results were consistent with the findings of Sharifi et al. ([2021](#page-11-17)) who used the MTSI stability index for the evaluation of yield and other agronomic traits in a set of rice genotypes and showed that this index is well able to identify superior genotypes in terms of yield stability and other agronomic traits.

4 | **CONCLUSION**

Rhizomania disease is one of the main factors in reducing quantitative and qualitative yield of sugar beet. Breeding genotypes with disease resistance genes is one of the most important solutions to deal with this destructive disease. Identifying sources of resistance and including them in breeding programs will improve resistant genotypes. For this purpose, in the present study, the resistance

level of sugar beet genotypes against rhizomania disease was investigated in terms of the infection severity and quantitative and qualitative traits. The analysis of variance for the studied traits in different areas infected with the disease showed significant genetic diversity among the genotypes. The results of the genotypes evaluation in terms of reaction to rhizomania disease showed that the number of genotypes with resistance reaction is much more than the number of genotypes with sensitivity reaction. Examining the genotypes yield stability in disease-infected environments by different statistics introduced somewhat different genotypes as stable ones in terms of white sugar yield, but overall, the majority of stability analysis methods agreed on the stability of RM18, RM3, RM9, and BTS335; However, the MTSI stability statistic considering all the traits presented somewhat different results and introduced RM3, RM17, RM9, RM13, and RM15 as stable genotypes in terms of all the investigated traits under disease-infected conditions.

ACKNOWLEDGMENTS

The authors are thankful to technicians at Sugar Beet Seed Institute (SBSI), Karaj, Iran for their assistance in conducting the experiments.

FUNDING INFORMATION

This work was supported by the Sugar Beet Seed Institute (SBSI), Karaj, Alborz, Iran. The funding body was involved in the material creation.

CONFLICT OF INTEREST

The authors report no conflicts of interest in this work. The authors alone are responsible for the content and writing of this article.

DATA AVAILABILITY STATEMENT

The data that support the findings of this study are available from the corresponding author upon reasonable request.

ORCID

Abazar Rajabi <https://orcid.org/0000-0001-6492-3734> *Ali Saremirad* <https://orcid.org/0000-0002-2775-0288>

REFERENCES

- Abdallah, J. M., Goffinet, B., Cierco-Ayrolles, C., & Pérez-Enciso, M. (2003). Linkage disequilibrium fine mapping of quantitative trait loci: A simulation study. *Genetics Selection Evolution*, *35*(6), 513–532.
- Ajay, B., Aravind, J., & Abdul Fiyaz, R. (2018). *Ammistability: Additive main effects and multiplicative interaction model stability parameters*. R Package Version, 11.
- Ajay, B., Bera, S., Singh, A., Kumar, N., Gangadhar, K., & Kona, P. (2020). Evaluation of genotype*×* environment interaction and yield stability analysis in peanut under phosphorus stress condition using stability parameters of AMMI model. *Agricultural Research*, *9*(4), 477–486.
- Akyüz, A., & Ersus, S. (2021). Optimization of enzyme assisted extraction of protein from the sugar beet (*Beta vulgaris* L.) leaves for alternative plant protein concentrate production. *Food Chemistry*, *335*, 127673.
- Anandan, A., & Eswaran, R. (2009). *Genotype by environment interactions of rice (Oryza sativa L.) hybrids in the east coast saline region of Tamil Nadu*. Paper presented at the proceeding of 2nd interaction Rice congress.
- Annicchiarico, P. (2002). *Genotype x environment interactions: Challenges and opportunities for plant breeding and cultivar recommendations*. Food & Agriculture org.
- Arjmand, M. N., & Ahun Manesh, A. (1996). Rhizomania, a new disease of sugar beet. *Sugar Beet*, *12*(1), 62–71.
- Bartlett, M. S. (1937). *Properties of sufficiency and statistical tests*. Proceedings of the Royal Society of London. Series A-Mathematical and Physical Sciences, 160(901), 268–282.
- Basafa, M., & Taherian, M. (2016). Analysis of stability and adaptability of forage yield among silage corn hybrids. *Journal of Crop Breeding*, *8*(19), 185–191.
- Cheloei, G., Ranjbar, G. A., Babaeian Jelodar, N., Bagheri, N., & Noori, M. Z. (2020). Using AMMI model and its parameters for yield stability analysis of rice (*Oryza sativa* L.) advanced mutant genotypes of Tarrom-Mahalli. *Iranian Journal of Genetics and Plant Breeding*, *9*(1), 70–83.
- Cook, D., & Scott, R. (1993). *The sugar beet crop: Science into practice*. Champan and Hall Press.
- Eggleston, G. (2019). History of sugar and sweeteners. In *Chemistry's role in food production and sustainability: Past and present*. *1314*, pp. 63– 74. American Chemical Society.
- Galein, Y., Legrève, A., & Bragard, C. (2018). Long term management of rhizomania disease—Insight into the changes of the beet necrotic yellow vein virus RNA-3 observed under resistant and

non-resistant sugar beet fields. *Frontiers in Plant Science*, *9*, 795. <https://doi.org/10.3389/fpls.2018.00795>

- Gauch, H. (1992). *Statistical analysis of regional yield trials: AMMI analysis of factorial designs*. Elsevier Science Publishers.
- Iqbal, M. A., & Saleem, A. M. (2015). Sugar beet potential to beat sugarcane as a sugar crop in Pakistan. *American-Eurasian Journal of Agricultural and Environmental Sciences*, *15*(1), 36–44.
- Izadpanah, K., Hashemi, P., Kamran, R., Pakniat, M., Sahandpour, A., & Masoumi, M. (1996). Extensive existence of root canal disease in Fars. *Journal of Plant Diseases*, *32*, 200–206.
- Jambhulkar, N., Bose, L., & Singh, O. (2014). *AMMI stability index for stability analysis*. Central Rice Research Institute, Cuttack, Orissa, 35, 15–15.
- Karimizadeh, R., Asghari, A., Chinipardaz, R., Sofalian, O., & Ghaffari, A. (2016). Determining yield stability and model selection by AMMI method in rain-fed durum wheat genotypes. *Turkish Journal of Field Crops*, *21*(2), 174–183.
- Karimizadeh, R., Dehghani, H., & Dehghanpour, Z. (2008). Use of AMMI method for estimating genotype-environment interaction in early maturing corn hybrids. *Seed and Plant Journal*, *23*(4), 531–546.
- Kiskini, A., Vissers, A., Vincken, J.-P., Gruppen, H., & Wierenga, P. A. (2016). Effect of plant age on the quantity and quality of proteins extracted from sugar beet (*Beta vulgaris* L.) leaves. *Journal of Agricultural and Food Chemistry*, *64*(44), 8305–8314.
- Kunz, M., Martin, D., & Puke, H. (2002). Precision of beet analyses in Germany explained for polarization. *Zuckerindustrie*, *127*(1), 13–21.
- Lammens, T., Franssen, M., Scott, E., & Sanders, J. (2012). Availability of protein-derived amino acids as feedstock for the production of biobased chemicals. *Biomass and Bioenergy*, *44*, 168–181.
- Lennefors, B.-L. (2006). *Molecular breeding for resistance to rhizomania in sugar beets*. Vol. 2006.
- Luterbacher, M., Asher, M., Beyer, W., Mandolino, G., Scholten, O., Frese, L., & Slyvchenko, O. (2005). Sources of resistance to diseases of sugar beet in related Beta germplasm: II. Soil-borne diseases. *Euphytica*, *141*(1), 49–63.
- McGrann, G. R., Grimmer, M. K., Mutasa-Gottgens, E. S., & Stevens, M. (2009). Progress towards the understanding and control of sugar beet rhizomania disease. *Molecular Plant Pathology*, *10*(1), 129–141.
- Monteiro, F., Frese, L., Castro, S., Duarte, M. C., Paulo, O. S., Loureiro, J., & Romeiras, M. M. (2018). Genetic and genomic tools to asssist sugar beet improvement: The value of the crop wild relatives. *Frontiers in Plant Science*, *9*, 74–85.
- Mostafavi, K., & Saremirad, A. (2021). Genotype - environment interaction study in corn genotypes using additive main effects and multiplicative interaction method and GGE- biplot method. *Journal of Crop Production*, *14*(3), 1–12. [https://doi.org/10.22069/](https://doi.org/10.22069/ejcp.2022.17527.2293) [ejcp.2022.17527.2293](https://doi.org/10.22069/ejcp.2022.17527.2293)
- Oerke, E.-C. (2006). Crop losses to pests. *The Journal of Agricultural Science*, *144*(1), 31–43.
- Olivoto, T., Lúcio, A. D. C., da Silva, J. A. G., Sari, B. G., & Diel, M. I. (2019). Mean performance and stability in multi-environment trials II: Selection based on multiple traits. *Agronomy Journal*, *111*(6), 2961– 2969.<https://doi.org/10.2134/agronj2019.03.0221>
- Olivoto, T., Nardino, M., Meira, D., Meier, C., Follmann, D. N., de Souza, V. Q., … Baretta, D. (2021). Multi-trait selection for mean performance and stability in maize. *Agronomy Journal*, *113*(5), 3968–3974.
- Omrani, S., Omrani, A., Afshari, M., Saremirad, A., Bardehji, S., & Foroozesh, P. (2019). Application of additive Main effects and multiplicative interaction and Biplot graphical analysis multivariate methods to study of genotype-environment interaction on safflower genotypes grain yield. *Journal of crop Breeding*, *11*(31), 153–163.
- Pavli, O. I., Stevanato, P., Biancardi, E., & Skaracis, G. N. (2011). Achievements and prospects in breeding for rhizomania resistance in sugar beet. *Field Crops Research*, *122*(3), 165–172.
- Purchase, J. L., Hatting, H., & van Deventer, C. S. (2000). Genotype *×* environment interaction of winter wheat (*Triticum aestivum* L.) in South Africa: II. Stability analysis of yield performance. *South African Journal of Plant and Soil*, *17*(3), 101–107. [https://doi.](https://doi.org/10.1080/02571862.2000.10634878) [org/10.1080/02571862.2000.10634878](https://doi.org/10.1080/02571862.2000.10634878)
- Raju, B. (2002). A study on AMMI model and its biplots. *Journal of the Indian Society of Agricultural Statistics*, *55*, 297–322.
- Rao, A., & Prabhakaran, V. (2005). Use of AMMI in simultaneous selection of genotypes for yield and stability. *Journal of the Indian Society of Agricultural Statistics*, *59*, 76–82.
- Reinfeld, E., Emmerich, A., Baumgarten, G., Winner, C., & Beiss, U. (1974). *Zur Voraussage des Melassez zuckersaus Ruben analysen Zucker*. Chapman & Hall, World Crop Series.
- Rezaei, J. (2007). *Evaluation of resistance of sugar beet commercial cultivars to rhizomania in field conditions*. Retrieved from Khorasan Razavi, [Iran](http://iran)
- Ribeiro, I. C., Pinheiro, C., Ribeiro, C. M., Veloso, M. M., Simoes-Costa, M. C., Evaristo, I., & Ricardo, C. P. (2016). Genetic diversity and physiological performance of Portuguese wild beet (*Beta vulgaris* spp. *maritima*) from three contrasting habitats. *Frontiers in Plant Science*, *7*(1), 1293.
- Sharifi, P., Abbasian, A., & Mohaddesi, A. (2021). Evaluation the mean performance and stability of Rice genotypes by combining features of AMMI and BLUP techniques and selection based on multiple traits. *Plant Genetic Researches*, *7*(2), 163–180.
- Sharifi, P., Aminpanah, H., Erfani, R., Mohaddesi, A., & Abbasian, A. (2017). Evaluation of genotype*×* environment interaction in rice based on AMMI model in Iran. *Rice Science*, *24*(3), 173–180.
- Sneller, C., Kilgore-Norquest, L., & Dombek, D. (1997). Repeatability of yield stability statistics in soybean. *Crop Science*, *37*(2), 383–390.
- Tamada, T. (1975). *Beet necrotic yellow vein virus*. CMI/AAB description of plant viruses, 144, 1–4.
- Tamada, T. (1999). *Benny viruses. Encyclopedia of virology* (Vol. *2*). Academic Press.
- Tamada, T., & Baba, T. (1973). Beet necrotic yellow vein virus from rizomania-affected sugar beet in Japan. *Japanese Journal of Phytopathology*, *39*(4), 325–332 _321.
- Tenorio, A. T., Schreuders, F., Zisopoulos, F., Boom, R., & Van der Goot, A. (2017). Processing concepts for the use of green leaves as raw materials for the food industry. *Journal of Cleaner Production*, *164*, 736–748.
- Tomaszewska, J., Bieliński, D., Binczarski, M., Berlowska, J., Dziugan, P., Piotrowski, J., Stanishevsky, A., & Witońska, I. (2018). Products of sugar beet processing as raw materials for chemicals and biodegradable polymers. *RSC Advances*, *8*(6), 3161–3177.
- United Nations. (2019). *World population prospects 2019: Highlights*. Department of Economic and Social Affairs, Population Division [https://www.un.org/development/desa/publications/world-popul](https://www.un.org/development/desa/publications/world-population-prospects-2019-highlights.html) [ation-prospects-2019-highlights.html](https://www.un.org/development/desa/publications/world-population-prospects-2019-highlights.html)
- Zali, H., Farshadfar, E., Sabaghpour, S. H., & Karimizadeh, R. (2012). Evaluation of genotype*×* environment interaction in chickpea using measures of stability from AMMI model. *Annals of Biological Research*, *3*(7), 3126–3136.
- Zhang, Z., Cheng, L., & Zhonghuai, X. (1998). Analysis of variety stability based on AMMI model. *Zuo wu xue bao*, *24*(3), 304–309.
- Zobel, R. (1994). *Stress resistance and root systems*. Paper presented at the proceedings of the workshop on adaptation of plants to soil stress. Institute of Agriculture and Natural Resources, University of Nebraska-Lincoln.

How to cite this article: Rajabi, A., Ahmadi, M., Bazrafshan, M., Hassani, M., & Saremirad, A. (2023). Evaluation of resistance and determination of stability of different sugar beet (*Beta vulgaris* L.) genotypes in rhizomania-infected conditions. *Food Science & Nutrition*, *11*, 1403–1414. [https://doi.org/10.1002/](https://doi.org/10.1002/fsn3.3180) [fsn3.3180](https://doi.org/10.1002/fsn3.3180)