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*CORRESPONDENCE Gang Huang HuangGang@shzu.edu.cn Hong-bin Li Nib@shzu.edu.cn

[†]These authors have contributed equally to this work and share first authorship

RECEIVED 30 November 2023 ACCEPTED 18 March 2024 PUBLISHED 08 April 2024

CITATION

Du Z-z, Xu W-b, Wang Y-x, Yan P, Ma Z-c, Huang G and Li H-b (2024) Prediction of the potential distribution area of *Glycyrrhiza inflata* in China using a MaxEnt model. *Front. Ecol. Evol.* 12:1347066. doi: 10.3389/fevo.2024.1347066

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Prediction of the potential distribution area of *Glycyrrhiza inflata* in China using a MaxEnt model

Zhen-zhu Du^{1,2†}, Wen-bin Xu^{1†}, Yu-xia Wang¹, Ping Yan¹, Zhan-cang Ma¹, Gang Huang^{1*} and Hong-bin Li^{1*}

¹College of Life Sciences, Shihezi University, Shihezi, Xinjiang, China, ²Laboratory Equipment Department, Shihezi University, Shihezi, Xinjiang, China

Glycyrrhiza inflata Batalin is an important medical plant of the genus Glycyrrhiza. It is one of the key protected plants in China, distributed in the desert areas of southern Xinjiang and Dunhuang of Gansu Province. It has a strong resistance to drought, heat, and salt stresses, and plays a pivotal role in sand fixtion in desert areas. In this study, based on 157 valid distribution records and eight environmental factors including climate factors and altitude, the potential distribution area of G. inflata in the last glacial maximum, middle Holocen, modern, and future (2050) times in China were predicted, using the optimized MaxEnt model and ArcGis 10.2 software. The results showed that the predicted distribution area was highly consistent with the current distribution range, and the area under the receiver operating characteristic (AUC) curve was 0.986, indicating that the prediction performance was excellent. The key climatic factors affecting the distribution were precipitation in December and the average annual precipitation. Meanwhile, the suitable area of G. inflata in modern times was 1,831,026 km², mainly distributed in Turpan-Hami Basin, Tarim Basin, and Dunhuang of Gansu Province, with Lop Nur Town of Xinjiang as the distribution center. In 2050, the potential suitable area for G. inflata in China will be 1,808,090 km², 250,970 km² of which will be highly suitable, which is $150,600 \text{ km}^2$ smaller than that in modern times, with a reduction rate of 60.0%. Therefore, there is a trend of great reduction in the suitable area of G. inflata. From the last glaciation maximum to the middle Holocene, the geographical distribution center shifted to the southwest margin of the Kumtag Desert, Xinjiang, then later continued to shift to the southwest. This study will provide a basis for understanding the origin and evolution of G. inflata, developing conservation strategies to minimize the impacts of environment change, and utilizing plant resource.

KEYWORDS

Glycyrrhiza inflata, MaxEnt model, potential distribution, geographical distribution center, climate factors

1 Introduction

Environmental factors greatly affect the distribution of species (Zhang et al., 2019). Among them, climate has an immediate and profound impact on geographical distribution and population density (Bellard et al., 2012; Bystriakova et al., 2014). The influence of different climates on species distribution has been used to clarify the relationship between the potential habitat of species and climatic conditions and the origin and evolution of species, which is of great significance for the utilization of species and the prediction of potential geographical distribution areas (Hou et al., 2023). Niche model, a common approach to explore the plantenvironment and plant-climate interactions, is suitable for analyzing the distribution range and spread tendency of plants (Siller-Clavel et al., 2022). Based on the known distribution data of species and related environmental variables, a model can be constructed by a specific algorithm to evaluate the ecological needs of species, and the calculation results can be projected to a specific time and space to predict the actual and potential distribution areas (Beckerman et al., 2002; Phillips et al., 2006; Phillips and Dudík, 2008; Kong et al., 2019; Li et al., 2019; Liao et al., 2020; Qin et al., 2022; Li et al., 2024). This method has been widely used in the prediction of potential distribution area of species, biodiversity conservation, and prediction of invasive areas of alien species under climate change (Wang et al., 2008; Liu et al., 2016; Wang and Yaermaimaiti, 2022). Nowadays, MaxEnt, CLIMEX, DOMAIN, and BIOCLIM models have been widely used, and the MaxEnt model is the most representative and widely used ecological niche model (Zhu et al., 2013; Zhang et al., 2023). MaxEnt models determine the stable relationship between species and environment by calculating the state parameters with maximum entropy in the species-environment interaction system and deduces the suitable distribution of species (Niu et al., 2017; Fu et al., 2023). It can effectively deal with the complex interaction relationship between variables, and its prediction performance is excellent (Sun et al., 2018; Liu et al., 2020; He et al., 2021).

As one of the three medicinal Glycyrrhiza species recorded in the Chinese Pharmacopoeia (Editorial Committee of Flora of China and Chinese Academy of Sciences, 1998), G. inflata is mainly distributed in the Tarim Basin in southern Xinjiang and Turpan-Hami Basin in eastern Xinjiang, and sporadically distributed in Dunhuang, Gansu province (Li et al., 2015). Apart from the strong cold- and drought-resistance, this species also has a strong salinealkali resistance. Therefore, it is an ideal plant to be used to improve the saline-alkali soil (Wang et al., 2022). In addition, G. inflata can be used to expel phlegm, relieve cough, detoxify, and treat parasitic malaria and leishmaniasis in traditional Chinese medical science (Lu, 2015). Glycyrrhiza plants are widely distributed in the temperate desert and the major river basins in Xinjiang, and its habitats are also the first choice for agricultural production. Therefore, from the 1850s to the 1990s, with the large-scale land reclamation in Xinjiang and the extensive collection of wild Glycyrrhiza plants, the distribution area of wild Glycyrrhiza plants reduced by 50%, and the reserves reduced by 75%. Subsequently, since the 1990s, Xinjiang's cultivated land area was expanded again, reaching 54.29% of the total area of Xinjiang (He et al., 2018), resulting in a rapid decrease in the wild *Glycyrrhiza* population. *G. inflata* was listed as a key protected plant in China in 2022 (Wang et al., 2022). Habitat protection and artificial planting are important ways to protect *Glycyrrhiza* plants and meet the market needs, respectively (Jia et al., 2019).

At present, many studies focus mainly on G. inflata's pharmacology and chemical composition (Ablat et al., 2016; Guo et al., 2022; Zeng et al., 2023; Zhurinov et al., 2023). There has been a growing interest in the DIVA-Gis and MaxEnt models for predicting the potential distribution areas of Glvcvrrhiza populations. Some scholars reported that climate change might lead to an increase in suitable habitats for some Glycyrrhiza species in the future, including G. inflata (Huang et al., 2023). In addition, Li et al. (2022) reported that the potential highly suitable habitat for Glycyrrhiza plants would decrease, and suitable area fragmentation would occur, with a steadily increasing trend. Currently, it is important to achieve a win-win relationship between Glycyrrhiza plant resource protection and utilization. However, the influencing factors affecting the spatial distribution of G. inflata are not clear. Therefore, to clarify the distribution of G. inflata and its response to future climate change, based on the existing field specimen collection points of G. inflata, this study used the MaxEnt model to predict the potential distribution areas in the past (last glacial maximum, the middle Holocene), contemporary, and future (RCP2.6-2050), and the ArcGIS software to determine the suitable area and distribution center (Booth et al., 2014). The Jackknife test, response curve and environmental limiting factor research were conducted to determine the climate limiting factors and suitable climatic conditions for G. inflata. The objectives of this study were: (i) To predict the potentially suitable area of *G. inflata* in China under current climate conditions; (ii) to analyze the main environmental factors which influence the geographical distribution of G. inflata; and (iii) to predict and compare the potentially suitable area and the change trends of G. inflata under different climate conditions in the past and future. This study will provide a scientific basis for the development and utilization of G. inflata resources.

2 Data and methods

2.1 Data sources

Moat of the data on *G. inflata* distribution were collected by field surveys. and the latitude and longitude were determined by a GPS locator and recorded. A small amount of distribution data (10.83%) were downloaded from Global Biodiversity Information Network (http://www.gbif.org), Kunming Plant Research Institute Specification Database (http://kun.kingdonia.org), and Chinese Plants (http://frps.iplant.cn). After deleting redundant data, a total of 130 distribution points (Figure 1) were used for later analyses.

The data for a total of 67 climate variables (Table 1) were downloaded from the WorldClim database (http:// www.worldclimorg). The soil data were downloaded from Harmonized World Soil Database v 1.2 (Wieder et al., 2014), Food



and Agriculture Organization of the United Nations (http:// www.fao.org/home/en/). The elevation data were obtained by downloading a 30 m-resolution DEM from the geospatial data cloud platform of Computer Network Information Center, Chinese Academy of Sciences (http://www.gscloud.cn). The GISMap (map based on a geographic coordinate system that is downloaded in real time from special online map services, such as OpenStreetMap) was used to cut climatic layers and convert them into ASC format. The current climatic variables used in the study were downloaded from the World Climatic Database and the spatial resolution of layers was 5 arc (about 9.2 km), climatic data for the Last Glacial Maximum of 22 ka, the Mid Holocene (about 6ka) and the future climate (2050) were from the WorldClim database. Besides, the CCSM4 model and rcp2.6 were selected to predict the future climate (2050) scenario. To maintain the comparability of the models in the spatial and temporal dimensions, the elevation and soil factors were unchanged in the prediction of future geographic distribution. The GCS_WGS_1984 projection coordinate system was adopted in this study.

2.2 Data analyses

2.2.1 Filtering of major environmental factors

In total, 80 environmental variables, including 67 climate variables, 12 soil variables, and altitude, were selected. Since collinearity between variables would lead to over-fitting of the distribution prediction model, principal component analysis and Spearman correlation analysis (Figure 2) were used to select variables with a correlation coefficient less than 0.8, and the variable with the largest contribution to the model was retained among the two variables with a correlation greater than 0.8 (Sun et al., 2023). After filtering, eight environmental factors were selected for model prediction including four climate factors (bio11, bio12, bio19, and Prec12) three soil factors (S-Esp, S-OC, and S-SAND), and one topographic factor (Alt).

2.2.2 Modeling and optimization

The MaxEnt default parameter was used for modeling. The regularization multiplier (RM) and feature combination (FC) parameter (Cobos et al., 2019) of the MaxEnt model were adjusted by using the ENMTools package. The complexity of the model under various parameter conditions was analyzed, and the model parameters with the lowest complexity were selected. Then, the response curve and prediction results were comprehensively analyzed. Finally, the influence of the MaxEnt model complexity closely related to the RM and FC parameters on the prediction results was analyzed. There were five features in MaxEnt, namely linear (linear-L), quadratic (quadratic-Q), fragmentation (hinge-H), product (product-P), and threshold (threshold-T). The ENMTools package can evaluate the complexity of the model by testing the modified Akaike information criterion correction (AICc) value of the MaxEnt model under different parameter conditions. The AIC is a standard to measure the fitting ability of the model, which can balance the complexity and the fitting ability of the model. AIC information quantity criterion gives priority to the model with the minimum AIC value. The AIC value can be calculated by the lambdas file generated after running the MaxEnt model. The RM was set at 0.1-4, increasing by 0.1 each time. A total of 31 feature combinations (FC) were used to evaluate the fitting effect of MaxEnt model on the distribution data of G. inflata. Three-fourths of all distribution points were randomly selected for model construction and optimization, and the remaining points were used for the model test. MaxEnt models were compared and evaluated mainly by response curve and omission rate curve (Wang et al., 2023), and delta AIC was used to test model complexity and degree of fit. Then, the optimal combination for MaxEnt model operation was determined.

Based on AIC information criterion, when the RM of MaxEnt was 2 and the operating features were L and P, there was a minimum AIC value (Figure 3). So, the parameters were selected as the optimal parameter combination to construct MaxEnt model for predicting *G. inflata* distribution.

TABLE 1 Climate variables used for modeling

Variable	Description	Variable	Description
Bio 1	Average annual temperature/°C	Bio 19	Average precipitation in the driest season/mm
Bio 2	Mean monthly temperature difference between day and night/°C	Alt	Altitude/m
Bio 3	Diurnal temperature variation/Annual temperature difference	Tmin1-12	Minimum temperature from January to December/°C
Bio 4	Temperature variation variance/°C	Tmax 1-12	Maximum temperature from January to December/°C
Bio 5	Highest temperature in the hottest month/°C	Tavg 1-12	Average temperature from January to December/°C
Bio 6	Lowest temperature in the coldest month/°C	Prec 1-12	Precipitation from January to December/mm
Bio 7	Annual temperature range/°C	T-SAND	Topsoil sand fraction/% wt.
Bio 8	Average temperature in the wettest season/°C	T-CLAY	Topsoil clay fraction /% wt.
Bio 9	Average temperature in the driest season/°C	T-ECE	Topsoil salinity/ dS/m
Bio 10	Average temperature in the warmest season/°C	T-Esp	Topsoil sodicity/%
Bio 11	Average temperature in the coldest season/°C	T-OC	Topsoil organic carbon/%wt.
Bio 12	Mean annual precipitation/mm	Т-РН-Н2О	Topsoil pH (H ₂ O)/-log(H+)
Bio 13	Precipitation in the wettest month/mm	S-SAND	Subsoil sand fraction% wt.
Bio 14	Precipitation in the driest month/mm	S-CLAY	Subsoil clay fraction% wt.
Bio 15	Precipitation variation variance/mm	S-ECE	Subsoil salinity/ dS/m
Bio 16	Precipitation in the wettest season/mm	S-Esp	Subsoil sodicity/%
Bio 17	Precipitation in the driest season/mm	S-OC	Subsoil organic carbon/%wt.
Bio 18	Average precipitation in the warmest season/mm	S-PH-H2O	Topsoil pH (H ₂ O)/-log(H+)

2.2.3 Prediction of suitable areas and central distribution points of *G. inflata*

The distribution point data of *G. inflata* were input, and the cropped environmental factors were layered into MaxEnt software. Three-fourths of the total points were randomly selected for modeling and 25% of the total points were used to verify the



model. The maximum number of iterations was 1000, and the Bootstrap was 10. Response curves were plotted, and the jackknife method was used to analyze the relationship between environmental factors and *G. inflata* distribution. According to the suitability value, ArcGIS software was used to divide the distribution areas predicted by the constructed model into the following four grades: non-suitable areas (fitness value: 0%-20%), general suitable area (fitness value: 20%-50%), moderately suitable areas (fitness value: 50%-70%), and highly suitable areas (fitness value: 70%-100%). The receiver operating curve value was applied to evaluate the accuracy of the model (Zhang et al., 2020). The ArcGIS (Mondal et al., 2022) was used to narrow the distribution range of suitable areas in different periods to a single central point, and then the changes of distribution centers in different periods were calculated to analyze the evolution trend of *G. inflata* distribution.

3 Results

3.1 Model accuracy detection

The ROC curve was obtained by simulating the training set through the built-in function of MaxEnt software, and the area under the curve was the value of AUC. The AUC value was 0.986, and the standard deviation was ± 0.002 (Figure 4). This indicates that the models had a good stability. According to the evaluation criteria (Liu et al., 2016), the overall prediction accuracy of the model reached an excellent level. This indicates that the model has a high accuracy in predicting the potential suitable areas for *G. inflata*.

3.2 Analysis of major environmental factors

The analysis results of climate variables of *G. inflata* (Figure 5) showed that there were eight major environmental factors: annual



average precipitation(bio12), subsoil sodicity (S-Esp), subsoil sand fraction (S-SAND), precipitation in December (Prec12), average precipitation in the coldest season (bio19), average temperature in the coldest season (bio11), subsoil organic carbon (S-OC), and altitude(Alt). Among them, bio12, Prec12, and bio19 were precipitation factors; S-Esp, S-SAND, and S-OC were soil factors; bio11 was a temperature factor; and Alt was a topographic factor. If a single variable was used for model prediction (Cai et al., 2022), the normalized gain value, test gain value, and AUC value of average annual precipitation was greater than 2.1, the normalized gain value was greater than 1.9, and the area under ROC curve was greater



than 0.95. Therefore, the average annual precipitation is the main climatic limiting factor affecting the distribution of *G. inflata*.

3.3 Changes in potential suitable areas of *G. inflata*

The area of suitable habitats (Figure 6C) for G. inflata in modern times decreased by 401,126.1 km² compared with that in the last glacial maximum (Figure 6A). The range of suitable habitats in modern times shrank in eastern Xinjiang, and the suitable habitat center moved to the vicinity of Lop Nur Town, Xinjiang. However, the area of highly suitable habitats in modern times decreased by 125,174.1 km² compared with that in the last glacial maximum (31.2% of the current), and the highly suitable areas, such as Turpan-Hami Basin in Xinjiang, had little shrinkage, but the area around Tarim Basin in southern Xinjiang had obvious shrinkage. There was also a small area of highly suitable habitats in Alxa League of Inner Mongolia. This indicates the eastward spread of the G. inflata population. The moderately suitable area for G. inflata in modern times greatly reduced by 377,922.7 km² (-125.8% of the current) compared with that in the last glacial maximum. The moderately suitable area in the Tarim Basin in Xinjiang and northern Inner Mongolia in modern times showed a reduction, while the moderately suitable area in Haixi of Qinghai Province was almost unchanged, compared with those in the last glacial maximum. The suitable area for modern G. inflata in modern times increased by 101,970.7 km² (9.0% of the current) compared with that in the last glacial maximum, distributing in the Tarim Basin of Xinjiang. Most of the moderately suitable areas in northern Inner Mongolia during the last glacial maximum changed into general suitable areas in modern times.

The range of suitable habitats in modern times (Figure 6C) increased by 53,252.3 km² compared with that in the middle Holocene (Figure 6B). The distribution of suitable habitats shifted slightly to the southwest, and the center shifted slightly to the southwest in Lop Nur Town, Xinjiang. However, the highly suitable area in modern times decreased by 31,115.7 km² (7.8% of the current) compared with that in the middle Holocene, which mainly appeared in the southern Tarim Basin of Xinjiang, with sporadic reduction in Hotan and Kashgar. Compared with the middle Holocene, the moderately suitable area of G. inflata in modern times increased by 4,089.5 km²(1.4% of the current), mainly in Turpan and Xinjiang, compared with that in the middle Holocene, but reduced in the central Tarim Basin. The suitable area for G. inflata in modern times increased by 80,278.6 km² (7.1% of the current) compared with that in the middle Holocene. The increase was mainly detected in the middle of the Tarim Basin, where the moderate suitable area shifted into the general suitable area.

The range of suitable habitats for *G. inflata* (Figure 6D) in 2050 reduced by 22,936.7 km² (1.3% of the current) compared with that in modern times (Figure 6C), and the range of suitable habitats in 2050 shifted to the southwest. The highly suitable area in 2050 decreased by 150,600.1 km² (mainly in the Turpan and Hami of



FIGURE 5

Jackknife test of the importance of environment variables in MaxEnt model. (S-SAND, subsoil sand fraction; S-OC, subsoil organic carbon; S-Esp, subsoil sodicity; Prec12, precipitation in December; bio19, average precipitation in the coldest season; bio12, average annual precipitation; bio11, average precipitation in the coldest season; alt, altitude).

Xinjiang, 60.0% of the current) compared with that in modern times. The moderately suitable area for *G. inflata* in 2050 increased by 35,916.4 km² (10.7% of the current) compared with that in the modern times due to the highly suitable areas in Turpan and Hami of Xinjiang transforming to moderately suitable areas. However, the moderately suitable area in the central Tarim Basin mainly transformed to a general suitable area. The general suitable area in 2050 increased by 91,746.9 km² (7.5% of the current) compared

with that in modern times, due to the moderately suitable areas transforming to general suitable areas.

The predicted results of suitable areas from the last glacial maximum to 2050 showed that the total suitable area for *G. inflata* decreased significantly until the middle Holocene and then changed little (Figure 7). The suitable areas in modern times reached 1,831,026 km². The general suitable areas increased gradually. This is due to the conversion of moderately suitable and highly suitable



FIGURE 6

Spatial distribution of Glycyrrhiza inflata under climate change scenarios. (A) Potential distribution areas in the last glacial maximum. (B) Potential distribution area in the middle Holocene. (C) Potential distribution area in modern times. (D) Potential distribution area in RCP2.6-2050.



areas to general suitable areas. In the middle Holocene, the moderately suitable areas decreased greatly and then increased slightly, but the highly suitable areas gradually decreased, especially by 2050. In 2050, the potential suitable area for *G. inflata* in China will be 1,808,090 km², of which the highly suitable areas will be 250,970 km², which is 150,600 km² smaller than that in modern times, with a reduction rate of 60.0%. Therefore, there is a trend of great reduction in the suitable areas for *G. inflata*.

3.4 Changes in the distribution center of *Glycyrrhiza inflata*

In the last glacial maximum, the distribution center of *G. inflata* was in the western margin of the Kumtag Desert, Xinjiang, and moved southwest along the Aqik Valley (40.2037° N, 91.9738° E) (Figure 8). By the middle Holocene, the distribution center was in the southeast of Lop Nur Town of Xinjiang (39.9072° N, 90.1943° E), and moved further southeast. Under modern climate conditions, the distribution

center continued to move southwest along the paleochannel for more than 10 kilometers (39.833° N, 90.1201° E). In the future climate scenario (RCP2.6-2050), the distribution center would move to the southwest of Ruoqiang County (39.7589° N, 89.9719° E). In general, the distribution center of *G. inflata* continued to move to the southern margins of the Kumtag Desert in Xinjiang.

The changes in the distribution center of *G. inflata* showed that the distribution center of *G. inflata* moved to the southwest significantly from the last glacial maximum to the middle Holocene. After the middle Holocene, the temperature increased, the glaciers in the middle and high latitude mountains melted, and the climate became warm and humid. However, the distribution center of *G. inflata* was stable in the southeast of Lop Nur Town, Xinjiang, due to its limited influence on the habitats of *G. inflata* (Lu et al., 2013; Niu et al., 2017). *G. inflata* has strong resistance to light, drought, salt, and alkali stresses; therefore, the climate change had little impact on the habitats of *G. inflata* in the northern margin of the Kumtagh Desert and the margin of the Taklimakan Desert. It was also found that the distribution center of *G. inflata* showed a continuous movement to the



southwest as the global temperature continued to rise. In the future, as the global temperature continues to rise, the distribution center of *G. inflata* will stably move to the southwest.

4 Discussion

4.1 Main environmental factors affecting the distribution of *Glycyrrhiza inflata*

At present, the models for the evaluation of suitable plant and animal habitats mainly include niche model (Xu et al., 2015; Grev et al., 2019; Wu et al., 2022; Zhang et al., 2022), mechanism model, and regression model. Among them, the niche model pays more attention to species occurrence sites and environmental variables. To some extent, the MaxEnt model is a representative of the niche model (Wang et al., 2008; Lin et al., 2017). According to the normalized gain value, contribution rate, and single factor response curve in the results obtained by the optimized MaxEnt model, the main environmental factors affecting the distribution of G. inflata are average annual precipitation (bio12) and precipitation in December (Prec12). Precipitation especially is the main environmental factor affecting the distribution of G. inflata. G. inflata is only distributed in arid regions of Central Asia such as Uzbekistan, Turkmenistan, Kyrgyzstan, Tajikistan, and the arid areas of northwest China (West of Hexi Corridor in Gansu Province, Turpan-Hami Basin in eastern Xinjiang, Tarim Basin in southern Xinjiang). It can be seen that G. inflata can adapt to the arid environment. Huang et al. found that the precipitation in the driest month (bio14) had a significant effect on the distribution of *Glycyrrhiza* in China (Huang et al., 2023). This is similar to the results obtained in this study, indicating that precipitation is an important factor affecting the distribution of Glycyrrhiza under global climate change. In this study, the main precipitation variables that affected the distribution of G. inflata were the precipitation in December (32.9%) and the average annual precipitation (29.6%), and the main temperature variable was the average temperature in the coldest season (2.1%). This may be due to the fact that seeds of G. inflata are small, hard, and resistant to water stress (Zhou et al., 2017). After ripening in October, stratification is needed for winter, and soil moisture and temperature are very important for the germination of G. inflata seeds during the snowmelt season in the following year (Wang and He, 2004; Yang et al., 2016). G. inflata is a perennial herb with strong underground rhizomes, through which G. inflata propagated asexually (Wang and He, 2004). The annual precipitation in the Taklamakan Desert is only 20-70 mm, and plants can only grow if they can obtain sufficient water from groundwater, lakes, or rivers (Bruelheide et al., 2010). The lack of rainfall and the low groundwater table make it difficult for G. inflata survive, while too much water causes root rot and even death. Therefore, the clonal reproduction and growth of G. inflata are affected by the average annual precipitation (Zhang and Wang, 2005). From the distribution altitude of G. inflata, it can be seen that there is almost no G. inflata at high altitudes (more than 2000 meters) where there is a lack of heat.

The credibility of the species distribution model to simulate the potential distribution area of species is closely related to the number and distribution range of species samples (Araujo et al., 2005). In this study, the data on valid geographical distribution points obtained from field investigations were used to predict the past, present, and future distribution of G. inflata. In modern times, the suitable areas for G. Inflata were mainly distributed around the Tarim Basin, Tarim River valley, and Konqi River valley in Xinjiang, and Guazhou, Jinta, and Dunhuang in Gansu Province. Suitable areas for G. Inflata were also scattered in the Ejin River basin of Inner Mongolia (Wang et al., 2021). From the perspective of the climate zone, the suitable areas for G. Inflata were concentrated in the middle-temperate arid zone. This is consistent with the actual distribution. In Qaidam Basin of Qinghai Province, there were middle-height suitable areas but no distribution records. This may be related to the high altitude and low temperatures, which are not suitable for the heliophilic properties of G. inflata. The prediction results showed that, in 2050, the total suitable area for G. inflata would decrease and the habitat fragmentation would be serious, especially in Turpan-Hami Basin. This may be related to the exacerbating water shortage in this region in the future (Zhang and Zhang, 2013). In addition, it was also found that the suitable areas were degraded, that is, the highly suitable areas were transferred to moderately and generally suitable areas in a large area, and most suitable areas were degraded to low and non-suitable areas. This is similar to the study results of Ephedra equisetina (Rong et al., 2023), Sabina centrasiatica (Wang et al., 2022), Populus xjrtyschensis (Wang et al., 2022), Kalidium (Liu et al., 2022), and Allium mongolicum (Lang et al., 2023). This may be due to that the increase of CO₂ emissions and human interference such as excessive wild plant collection and land reclamation, which aggravates the habitat fragmentation of G. inflata and leads to the reduction of suitable areas. Yang et al. also verified this point by studying the genetic structure of the G. inflata population (Yang et al., 2016). This study found that the total habitat area of G. inflata showed a gradually decreasing trend. This may be due to the fact that, under the future climate scenario, the habitats of some species will continue to decrease (Wang et al., 2008; Wang and Yaermaimaiti, 2022), and these species will migrate to high latitude and altitude, along with the fragmentation of habitats (Li et al., 2022). However, high altitude and low temperature are limiting factors for G. inflata growth. Therefore, an overall decreasing trend was predicted in the future in this study. This result is contrary to Huang's (Huang et al., 2023). This may be due to a slight increase in precipitation in arid areas due to global warming and glacier melting. However, in arid areas, temperature and evaporation increases, so a slight increase in precipitation can hardly offset the negative effect of temperature rise. This will eventually lead to the degradation of natural vegetation in deserts. Therefore, the prediction of suitable areas for plants in arid areas considering only precipitation difficult.

The climatic change in the Quaternary Ice Age has a great influence on the distribution of species. According to the results of model prediction, the distribution center of *G. inflata* moved to the

10.3389/fevo.2024.1347066

southwest significantly from the last glacial maximum to the middle Holocene. due to the extensive existence of glaciers in the Tibetan Plateau, Kunlun Mountains, Altun Mountain, and Qilian Mountains in the last glacial maximum (Zhu and Qiao, 2016). The cooling of the climate led to the migration of G. inflata from north to south (northern margin of Kumtag Desert). From the mid-Holocene to the future, the change trend of distribution center tended to be stable, but the distribution center showed a trend of continuous southwest movement. This may be related to the continuous southwest shrinkage of the suitable areas for G. inflata, especially the severe shrinkage in the Turpan-Hami Basin. There is no river in the Turpan-Hami Basin. The water storage is small, the precipitation is low, and the evaporation is large, coupled with the excessive use of groundwater by human beings (Fang et al., 2022); therefore, water is the main limiting factor that makes this area gradually unsuitable for the growth of G. inflata. These results are consistent with those of Zhang et al. (2019).

4.3 Suggestions on the *Glycyrrhiza inflata* natural population protection

G. Inflata is widely distributed in arid areas. However, according to this study results, its suitable area will shrink in the future, and the natural population degradation is serious, so it is particularly important to protect the germplasm resources of G. inflata. Considering the distribution characteristics, the change in highly suitable areas in the future, and the stable southeast movement of the distribution center (in the Lop Nur Town in Xinjiang), it is suggested to set up habitat protection zones in the Tarim Basin (Shaya County in north Aksu, Qira county in southern Hotan, east of Korla City, and south of Lop Nur Town), Turpan-Hami Basin (Daheyan in Turpan, Daquanwan in Hami), and Dunhuang in Gansu Province. This is consistent with the conservation recommendations based on the population genetic structure of Yang et al. (2016). According to the field survey results, it was found that there was still a certain G. Inflata distribution in the above areas, and the areas with less human activity can be prioritized as protected areas to maintain its natural habitats. In Daheyan in Turpan, Daquanwan in Hami, Mogao Town in Dunhuang, and Yangtak in Qiemo, G. inflata grows in deserts or salinized deserts. The G. Inflata population is gradually reducing. Therefore, in these places, in situ protection and scientific management is the best choice to preserve the species.

The hinterlands of rivers and tributaries in the arid zone, the main habitats for *G. inflata*, have been reclaimed as farmlands. *G. inflata* is considered a weed in canal banks or roadsides; habitat fragmentation and hybridization are serious. If a large area of farmland is restored to grassland, the cost will be high. At present, various Chinese herbal germplasm gardens and botanical gardens have been established successively. *G. inflata* has been planted in the Turpan Desert Botanical Garden, but the population size is very small. On the one hand, the existing facilities can be selected to increase the number of breeding bases to protect *G. inflata* germplasm resources and genetic homozygosity.

5 Conclusion

G. inflata is the most salt- and drought-tolerant medicinal species of the genus Glycyrrhiza; therefore, G. inflata should be prioritized when considering the maintenance of the ecology of harsh desert habitats. In this study, the MaxEnt model was used to simulate past, present, and future (2050) suitable areas for G. inflata. The findings were as follows: (1) The primary climatic factor influencing the geographic distribution of G. inflata was the average annual precipitation; (2) The Turpan-Hami Basin, Tarim Basin, and Dunhuang of Gansu Province were the main suitable areas for G. inflata in modern times, with Lop Nur Town in Xinjiang as the distribution center; (3) The potential highly suitable areas for G. inflata showed a continuous large-scale reduction trend in the past, present, and future; (4) The geographical distribution center of G. inflata shifted to the southwestern margins of the Kumtag Desert in Lop Nur Town, Xinjiang, and would continuously move southwest. The study results will provide a basis for future research for the conservation, breeding, and development of G. inflata for medical purposes. Long-term research is still necessary to support the scientific protection of G. inflata populations.

Data availability statement

Publicly available datasets were analyzed in this study. Climate variables: WorldClim database, http://www.worldclim.com/. Soil factor data: Harmonized World Soil Database, https://www.fao. org/soils-portal/soil-survey/soil-maps-and-databases/harmonized-world-soil-database-v12/en. Elevation variables: Geospatial Data Cloud Platform of Computer Network Information Center, https://www.gscloud.cn/.

Author contributions

GH: Writing – review & editing. Z-ZD: Conceptualization, Funding acquisition, Methodology, Software, Writing – original draft. W-BX: Conceptualization, Methodology, Writing – original draft. Y-XW: Supervision, Writing – review & editing. PY: Methodology, Writing – review & editing. Z-CM: Methodology, Writing – review & editing. H-BL: Writing – review & editing.

Funding

The author(s) declare financial support was received for the research, authorship, and/or publication of this article. This study was supported by the National Natural Science Foundation of China (Grant number: 31470191), the New Variety Cultivation Project of Shihezi University (Grant number: YZZX202106; ZZZC2022122; XPRU202104), the Science and Technology

Project of Xinjiang Production and Construction Corps (Grant number: 2023AB052), and the Science and Technology Project of Huyanghe City (Grant number: QS2023008).

Acknowledgments

We would like to thank Dr. Shamshidin Abduriyim for his comments on an earlier version and the English of our manuscript.

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fevo.2024.1347066/ full#supplementary-material

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