



ANALYSIS ARTICLE

Bibliographic analysis of modelling weed distribution and invasion with global perspective

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ABSTRACT

Invasive alien weeds are of great concern because of their capability of spreading fast, their high competitiveness and ability to settle in new areas within short period of time. Thus, they are the second biggest threat to biodiversity after habitat destruction. It is therefore necessary to prevent the introduction, establishment, and spread of these invasive alien weeds (IAWs) into newer areas. Ecological niche modelling (ENMs) and species distribution modelling (SDMs) are two commonly used approaches in theoretical and applied studies in ecology to study the species behavior in future climatic conditions. In this study, we undertook a bibliographic analysis of scholarly articles on the modelling studies on species invasion under current and future climatic scenarios. In addition, results of different studies on modelling and prediction on distribution of IAWs on global as well as India level were also discussed. Study revealed that researchers started getting interest and published more work in the subject between 2015 and 2020. The greater number of related articles were published in the subjects such as ecology, biology, habitat and climate change and published mostly by Wiley, Elsevier and Springer publishers. Further, the shortcomings of species distribution modelling and future prospects were also discussed.

Keywords: Bibliographic analysis, Invasive alien weeds, Modelling, Species distribution modelling

INTRODUCTION

Invasive organism is defined as a non-native organism whose introduction causes, or is likely to cause, economic or environmental harm, or harm to human, animal, or plant health (Reaser *et al.* 2020). The invasive alien species are those that are introduced into places outside their natural range, adversely impacting native biodiversity, ecosystem or human well-being. According to Convention on Biological Diversity (2005), invasive alien species are introduced purposefully or accidentally outside their natural habitat, where they exhibit the ability to establish themselves, invade, out-compete native weeds and take over the new environment within short period of time. Thus, they have the potential to harm the biodiversity, ecosystem and human well-being (Ansong and Pickering 2015; Beaumont *et al.* 2014; Kleunen *et al.* 2015). They put significant social, ecological and economic impacts on the invaded environment (Gharde *et al.* 2018). The nature and severity of the impacts of these weeds on society, environment, health and national heritage are of great concern (McNeely *et al.* 2001). They are also highly tolerant to climatic and edaphic changes

and have ability to compete and drive off other species from their habitat. Thus, they are the second biggest threat to biodiversity after habitat destruction. They reduce agricultural yields, and interfere with crop lands, grazing areas, water availability, and contribute to spread of many diseases (Essa *et al.* 2006). Further, their uncontrolled expansion in agriculture ecosystem may cause huge crop yield losses (Chauhan *et al.* 2011; Fahad *et al.* 2015; Parker 2012).

In the era of globalization, it is necessary to prevent the introduction, establishment, and spread of these invasive alien weeds (IAWs) into newer areas (Rao *et al.* 2017). It is usually accepted that prevention before the establishment of the invasive weeds is a much better economic strategy than control or eradication (Seebens *et al.* 2017) after the establishment (Jarnevich *et al.* 2010; Braun *et al.* 2016). Moreover, management of invasive species relies on information about their expected distributional potential and relative abundance under current and future climate scenarios. Therefore, it is important to know the areas which are favorable for occurrence of these species so that planning can be done for appropriate long-term management strategies for the control of these species before its invasion in the new areas. Further, how species will respond to projected future climate change is of fundamental importance for effective management

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and conservation of biodiversity (Hannah *et al.* 2002; Hijmans and Graham 2006). It is reported that some extreme weather events such as droughts and floods may increase due to climate change and can cause huge impacts on the global ecosystem, including rise in sea levels (Lee 2010), change in areas of crop production and spread of species (Kwak *et al.* 2008; Pearson and Dawson 2003). As per estimate reported by Intergovernmental Panel on Climate Change (IPCC), the earth temperature is estimated to increase by about 1.4–5.8°C from 1990 to 2100, whereas precipitation is estimated to increase by up to 1.0% for the mid- and high-latitude regions and 0.3% for the tropical zones (IPCC 2014). It was also confirmed that South Asia will experience a substantial change in its climate during the 21st century. It is established that climate change has already changed many species' behaviors, biodiversity, their distribution and habitat substantially. As the climate is known to be the most significant factor affecting the growth and development (Rosenzweig *et al.* 2001), invasive weeds are heavily influenced by climate change and can extend their range, thereby causing increased damage to ecosystem and agricultural production.

However, the relationship between IAW and climate change is complex (Hellmann *et al.* 2008). Climatic factors are considered as one of the main factors determining the overall distribution of invasive species due to their synergistic effects (Guisan and Thuiller 2005; Bai *et al.* 2013; Gharde *et al.* 2019). However, Sathischandra *et al.* (2014) reported the absence of a linear correlation between the occurrence of weeds and insect pests with climate variables. Hence, with such complexities, there is need for precise prediction on dynamics of IAW under future climate change scenarios in order to manage such weeds (Kariyawasam *et al.* 2019).

To address these questions, we undertook a bibliographic analysis of scholarly articles on the modelling studies on species invasion under current and future climatic scenarios. Specifically, two approaches used for modelling, *viz.* ecological niche modelling and species distribution modelling along with different commonly used algorithms were discussed. Results of different studies on modelling and prediction on distribution of IAWs on global as well as India level were also discussed. Further, this article summarizes shortcomings of species distribution modelling (SDM) and future prospects.

Ecological niche modelling and species distribution modelling

Ecological niche modelling (ENMs) and SDM are two commonly used approaches in theoretical and

applied studies in ecology (Peterson *et al.* 2015). Most common applications of these models are finding suitable sites for species (Guisan and Zimmermann 2000), predicting the impacts of future climate change on species' distributions (Pearson and Dawson 2003), assessing the invasive potential of alien species (Jiménez-Valverde *et al.* 2011), and subsequently the conservation planning (Guisan *et al.* 2013).

However, there is considerable difference between these two approaches. SDM refers to the approach for modelling the objects in G-space (the geographical space occupied by the species), on the other hand, ENM refers to approach for modelling the objects in E-space (all the environmental combinations available in the study region) (Soberón *et al.* 2017). ENM requires an overt estimation of the fundamental niche of the species, and are envisioned to model the processes that defined the area of distribution of the species (Peterson and Soberón 2012). Usually SDM can only target the species' distribution, and preferably must restrict model calibration to accessible areas of the study region, account for true absences and integrate dispersal and colonization abilities (Peterson and Soberón 2012). Three main classes of models are recognized in this field: correlative models, the most commonly used models found in the literature, which estimate the ecological requirements of species by relating their known spatial distributions to a set of environmental/climatic variables (Araújo and Guisan 2006; Franklin 2010); mechanistic models which use exhaustive physiological information and first principles of biophysics (Kearney and Porter 2009); and process-oriented models, which estimate species' distributions in terms of processes, including dispersal capability and biotic interactions (Peterson *et al.* 2015).

Use of ecological niche models (ENMs) plays important role in early detection of IAWs and to identify the ecologically sensitive areas for further monitoring and making necessary control measures (Srivastava *et al.* 2019; Yan *et al.* 2020; Marambe and Wijesundara, 2021). Species distribution modelling makes use of point-occurrence data and raster data layers summarizing environmental information (**Figure 1**). These species distribution models thus infer species' environmental requirements, and have been used to anticipate the geographic potential of species (Wisz *et al.* 2008). These models have become the extensively useful tool to determine the relationships between species and their environments and are used to predict extreme impacts of climate change, biogeographic studies, improve species management and answer conservation biology questions.

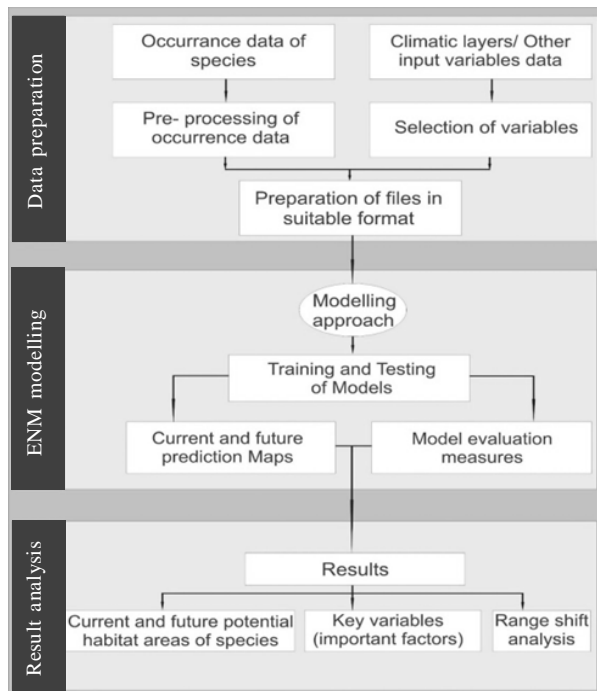


Figure 1. Steps used in species distribution modelling

BIOCLIM, DOMAIN, and Maxent are the frequently used ENMs known for their simplicity and the data accessibility (Elith *et al.* 2011; Katz and Zellmer 2018; Srivastava *et al.* 2019). However, compared to BIOCLIM and DOMAIN, Maxent exhibited much higher predictive performance (Phillips *et al.* 2006; Peterson and Anamza 2015), and it can generate much more robust results especially when applied to small sample sizes (Phillips *et al.* 2006; Elith and Leathwick 2009) when most of the technique fail to produce the adequate results. Thus, Maxent has been successfully applied to model the distributions of invasive species (Srivastava *et al.* 2019).

Additionally, several researchers have used remote sensing data to map the distribution of IAWs using phenology-based approaches (Ishii and Washitani 2013; Bradley *et al.* 2018; Huang and Geiger 2008). Remote detection of IAWs based on their distinct biochemical, physiological and structural traits are important in cases where IAWs and native species have similar phenology (Glenn *et al.* 2005; Mitchell and Glenn 2009; Yang and Everitt 2010). However, success of these approaches depends on the availability of hyperspectral data (Gholizadeh *et al.* 2022).

Species invasion under climate change scenarios

Several studies have been conducted to investigate the potential impact of global climate change on the geographic distribution of IAWs, but the results are somewhat different in each case (Buckley and Csergo 2017; Merow *et al.* 2017).

Some studies reported that climate change may favour the expansion of geographic distribution of these species (Priyanka and Joshi 2013; Banerjee *et al.* 2017; Wei *et al.* 2017; Shrestha *et al.* 2018; Thapa *et al.* 2018); whereas, others have reported that climate change may constrain the geographic distribution of some IAW species (Table 1 and 2) (Bradley 2009; Taylor and Kumar 2013; Roger *et al.* 2015; Allen and Bradley 2016; Manzoor *et al.* 2018). Hence, prediction on areas of high/low invasion risk or varied invasion impact for the future is full of uncertainties and mainly depends upon the invasive species or biome type (Buckley and Csergo 2017).

Bibliographic analysis of articles on weeds distribution modelling

Bibliographic analysis is defined as the evaluation of published scientific literature including articles, books, book chapter and provide a way to measure the impact of publication among the scientific community. Research in SDMs in predicting the future distribution of IAW has expanded significantly over the past few decades with evolution of different algorithms. Hence, the present bibliographic analysis, study was done using the literatures available in the area of modelling of IAW distribution. For this purpose, published research articles were accessed and analysis was done with the help of LENS.ORG free access database. It is meant for search, analyze and manage patent as well as Scholarly data. Here, scholarly data was used for the purpose, and keywords, *viz.* invasive alien plant species climate change future distribution and modelling were used for selection of the articles out of total 252,116,476 scholarly literature available in the LENS.ORG. Using these keywords, 3244 articles have been filtered out. Results were also shown as the impact of the scholarly articles present in the area searched. These includes active author information; citation of the scholarly works; classification of articles based on their research area as well as institution where the work was conducted and publication trend of the documents over the years. These results are presented through Figure 2 to Figure 6.

Figure 2 revealed that researchers started getting interest and published more work between 2015 and 2020. Many documents were published during these years with maximum observed during 2017 with more than 270 articles. These articles were more related to ecology (1360), biology (1218), geography (813), biodiversity (513), introduced species (406), invasive species (398), habitat (343), Climate change (325), ecosystem (302) and rest with other areas of interest (Figure 3). Most of the research were published by Spanish National

Table 1. The global prediction of species invasion under future climate scenario using distribution modelling

Weeds	Region	Contraction/expansion in areas	Reference	Weed type
<i>Ageratina adenophora</i> (Spreng.) R. M. King et H. Rob., <i>Alternanthera philoxeroides</i> (Mart.) Griseb., <i>Ambrosia artemisiifolia</i> L. <i>Mikania micrantha</i> Kunth <i>Parthenium hysterophorus</i> L.	China World and Oman	Expansion Contraction in areas in 2081–2100 at global level. Expansion during 2021–40 and a decrease during 2081–2100	Tu <i>et al.</i> 2021 Amna <i>et al.</i> 2022	Terrestrial Terrestrial
<i>Cynodon dactylon</i> , <i>Cyperus rotundus</i> , <i>Echinochloa colona</i> , <i>Echinochloa crus-galli</i> , <i>Eichhornia crassipes</i> , <i>Eleusine indica</i> , <i>Imperata cylindrica</i> , <i>Lantana camara</i> , <i>Panicum maximum</i> , and <i>Sorghum halepense</i> <i>Parthenium hysterophorus</i> L.	World Bangladesh	Expansion Expansion	Wan and Wang 2019 Masum <i>et al.</i> 2022	Terrestrial and agro-ecosystem Terrestrial
<i>Ageratina adenophora</i>	China	Expansion of the dispersal zone towards the northeast and coastal areas, and a slight contraction in the Yunnan–Guizhou plateau	Zhang <i>et al.</i> 2022	Terrestrial
<i>Ageratina adenophora</i>	Global	Contraction in potential suitable area globally and range expansion in six biodiversity hotspot regions	Changjun <i>et al.</i> 2021	Terrestrial
<i>Ambrosia artemisiifolia</i> , <i>Ambrosia trifida</i> , <i>Symphytotrichum pilosum</i> , <i>Ageratina altissima</i> , <i>Hypochaeris radicata</i> , <i>Lactuca serriola</i> , <i>Paspalum dilatatum</i> , <i>Paspalum distichum</i> , <i>Rumex acetosella</i> , <i>Sicyos angulatus</i> , <i>Solanum carolinense</i> , <i>Solidago altissima</i> <i>Spartina alterniflora</i> Loisel	South Korea China	Expansion Expansion	Adhikari <i>et al.</i> 2022 Yuan <i>et al.</i> 2021	Terrestrial Terrestrial
<i>Verbesina encelioides</i> (Cav.) Benth. & Hook. Fil ex Gray	South Africa	Expansion	Moshobane <i>et al.</i> 2022	Terrestrial
<i>Ageratina adenophora</i> (Sprengel) R. King and H. Robinson	Chitwan–Annapurna Landscape (CHAL) of Nepal	Expansion	Poudel <i>et al.</i> 2020	Terrestrial
<i>Parthenium Hysterophorus</i> L. <i>Apium leptophyllum</i> , <i>Astragalus sinicus</i> , <i>Bromus unioloides</i> , <i>Chenopodium ambrosioides</i> , <i>Coronopus didymus</i> , <i>Gnaphalium calviceps</i> , <i>Lolium multiflorum</i> , <i>Modiola caroliniana</i> , <i>Oenothera laciniata</i> , <i>Paspalum dilatatum</i> , <i>Sida rhombifolia</i> , <i>Silene gallica</i> , <i>Sisymbrium officinale</i> , <i>Sisyrinchium angustifolium</i> , <i>Spergularia rubra</i> , <i>Malva parviflora</i>	Bhutan South Korea	Expansion Expansion	Dorji <i>et al.</i> 2022 Hong <i>et al.</i> 2021	Terrestrial Terrestrial
<i>Urochloa panicoides</i> P. Beauv.	World	Reductions in climate suitability in Brazil, Australia, India, and Africa, and an increase in suitability in Mexico, the United States, European countries, and China	Duque <i>et al.</i> 2022	Terrestrial
<i>Amaranthus palmeri</i>	USA	Northward range expansion and significantly increased suitability across large portions of the U.S. Overall	Runquist <i>et al.</i> 2019	Terrestrial
<i>Parthenium hysterophorus</i>	Chitwan Annapurna Landscape, Nepal	Expansion in the suitable habitat under RCP 4.5 scenario in 2050 and 2070, however decrease in suitable areas under RCP 8.5 scenario in 2050 and 2070	Maharjan <i>et al.</i> 2019	Terrestrial
<i>Alstonia macrophylla</i> Wall., <i>Annona glabra</i> L., <i>Austroeuatorium inulifolium</i> (H.B.K.) R. M. King & H. Rob. <i>Clidemia hirta</i> (L.) D. Don, <i>Dillenia suffruticosa</i> (Griff ex Hook.f. & Thomson) Martelli, <i>Lantana camara</i> L., <i>Leucaena leucocephala</i> (Lam.) de Wit, <i>Mimosa pigra</i> L., <i>Opuntia dillenii</i> (Ker-Gawl.) Haw, <i>Panicum maximum</i> Jacq., <i>Parthenium hysterophorus</i> L., <i>Prosopis juliflora</i> (Sw.) DC., <i>Sphagneticola trilobata</i> (L.) Pruski, <i>Ulex europaeus</i> L.	Sri Lanka	Contraction of the very low class and expansion of the moderate class of suitability.	Kariyawasam <i>et al.</i> 2019	Terrestrial
<i>Myriophyllum aquaticum</i> , <i>Pistia stratiotes</i> , <i>Azolla filiculoides</i> , <i>Eichhornia crassipes</i> , <i>Salvinia molesta</i>	South Africa	Contraction in <i>Myriophyllum aquaticum</i> and <i>Pistia stratiotes</i> suitable areas and expansion in rest three areas	Hoveka <i>et al.</i> 2016	Freshwater weeds
<i>Nitellopsis obtusa</i>	United States	Decrease of the species' suitable range	Romero-Alvarez <i>et al.</i> 2017	Aquatic weed

Weeds	Region	Contraction/expansion in areas	Reference	Weed type
<i>Alternanthera philoxeroides</i> , <i>Ceratophyllum demersum</i> , <i>Crassula helmsii</i> , <i>Elodea canadensis</i> , <i>Hydrilla verticillata</i> , <i>Ludwigia peruviana</i> , <i>Najas minor</i> , <i>Pistia stratiotes</i> , <i>Potamogeton crispus</i> , <i>Sagittaria platyphylla</i> <i>Lantana camara</i> L.	World	Significantly higher climatic suitability for temperate coastal rivers and temperate floodplain rivers	Wang <i>et al.</i> 2017	Freshwater weeds
	World	Climatically suitable areas globally will contract. However, some areas in North Africa, Europe and Australia may become climatically suitable. In South Africa and China, its potential distribution could expand further inland.	Taylor <i>et al.</i> 2012	Terrestrial
<i>Butomus umbellatus</i>	North America	Decrease of suitable areas, though two of three global circulation models predict range expansion across gas emission scenarios	Banerjee <i>et al.</i> 2020	Terrestrial
<i>Ageratum conyzoides</i> , <i>Praxelis clematidea</i> , <i>Solidago canadensis</i> , <i>Anredera cordifolia</i> , <i>Lantana camara</i> , <i>Conyza sumatrensis</i> , <i>Chenopodium ambrosioides</i> , <i>Parthenium hysterophorus</i> , <i>Avena fatua</i> , <i>Pharbitis purpurea</i> , <i>Aster subulatus</i> <i>Mikania micrantha</i>	China	Species will expand northward	Guan <i>et al.</i> 2020	Terrestrial
	South and Southeast Asia, Australia, Oceania and parts of the USA	Predicted to expand toward cold and dry areas of the invasive range	Banerjee <i>et al.</i> 2019	Terrestrial
<i>Lonicera japonica</i>	Forests of the Cumberland Plateau and Mountain Region in the southeast of USA	Expansion	Lemke <i>et al.</i> 2011	Forest land
<i>Chromolaena odorata</i>	World	Expansion	Kriticos <i>et al.</i> 2004	Terrestrial
<i>Amaranthus retroflexus</i> , <i>Amaranthus spinosus</i> , <i>Amaranthus viridis</i> , <i>Bidens pilosa</i> , <i>Conyza bonariensis</i> , <i>Conyza canadensis</i> , <i>Galinsoga parviflora</i> , <i>Physalis angulata</i> <i>Lantana camara</i> L.	China	Expansion	Wan <i>et al.</i> 2017	Terrestrial
	Queensland, Australia	Reduction in climatic suitability	Taylor and Kumar 2013	Terrestrial

Table 2. The prediction of a few weed species invasion in India under future climate scenarios

Weeds	Region	Contraction/expansion in areas	Reference	Weed type
<i>Lantana camara</i>	Jharkhand, eastern India	Expansion up to 20–26% by 2050	Tiwari <i>et al.</i> 2022	Terrestrial
<i>Parthenium hysterophorus</i>	India	Overall decrease in habitat suitability with some highly vulnerable (Western Himalaya) region to its invasion under future climate	Ahmad <i>et al.</i> 2019	Terrestrial
<i>Chromolaena odorata</i> L. (King) & H.E. Robins	India	Higher suitability for species in northeastern states, the central Himalayan provinces and the Western Ghats and Eastern Ghats	Barik and Adhikari 2012	Terrestrial
<i>Ageratina adenophora</i> L., <i>Ageratum conyzoides</i> L., <i>Ageratum houstonianum</i> Mill., <i>Amaranthus spinosus</i> L., <i>Bidens pilosa</i> L., <i>Erigeron karvinskianus</i> DC., <i>Lantana camara</i> L., <i>Parthenium hysterophorus</i> L., <i>Senna occidentalis</i> (L.) Link., <i>Senna tora</i> (L.) Roxb., <i>Xanthium strumarium</i> L. <i>Cassia tora</i> and <i>Lantana camara</i>	Western Himalaya, India	Most of these invasive plants are expected to expand under future climatic scenarios	Thapa <i>et al.</i> 2018	Terrestrial
<i>Chromolaena odorata</i> and <i>Tridax procumbens</i>	India	Distribution ranges of both species could shift in the northern and north-eastern directions in India Both are likely to reduce their potential distribution areas in the future climate	Panda <i>et al.</i> 2018 Panda and Behera 2019	Terrestrial

Research Council in all major areas and it was followed by Stellenbosch University in case of Ecology, Biology and Introduced Species. PloS one, PloS biology, PloS neglected tropical diseases are some of the leading free access journals who

published the work related to modelling IAW invasion. The Wiley, Elsevier, Springer and Wiley-Blackwell are the leading publishers of research results on these areas (Figure 5 and 6).

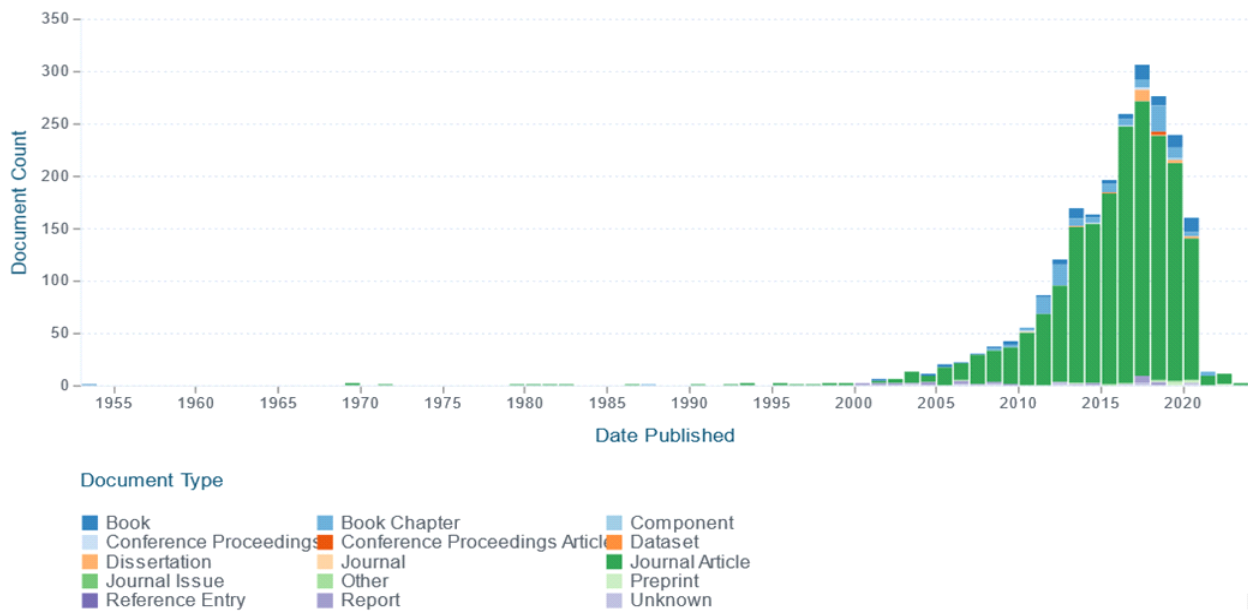


Figure 2. Published literature on modelling weed invasion over the years (from 1955-2022)

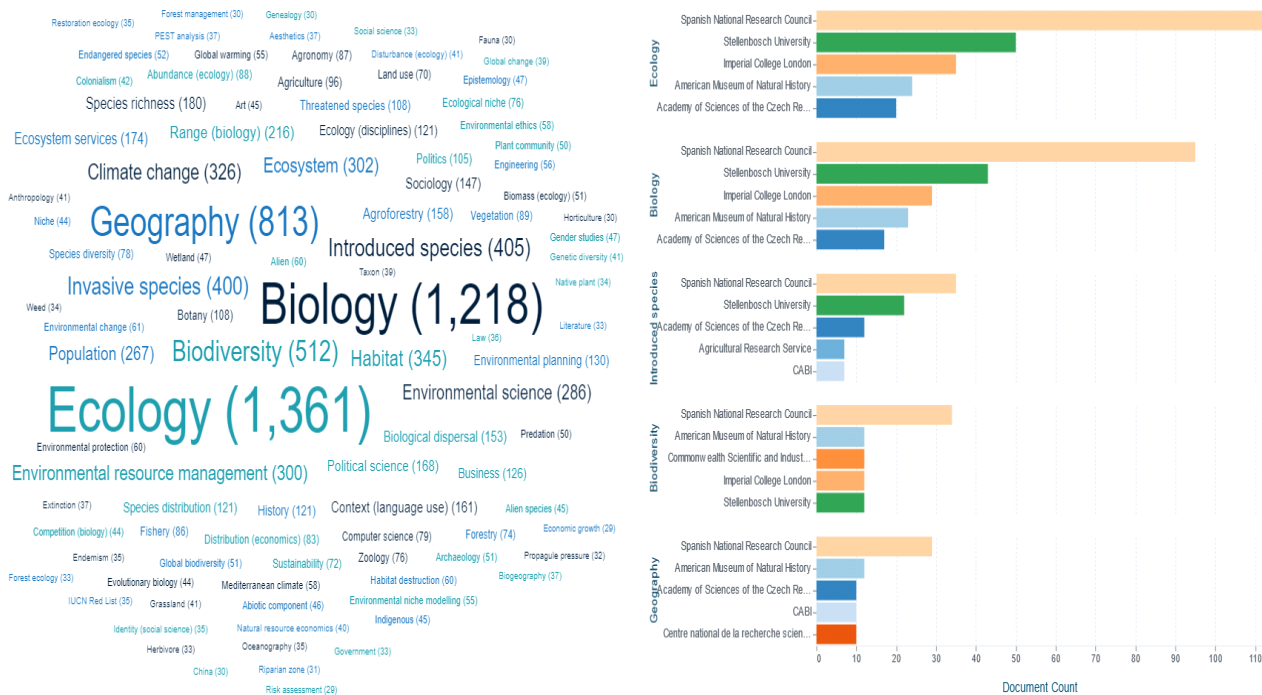


Figure 3. Classification of articles according to the research areas and research institutions

Implications of future climate on invasive alien weeds

The predicted distributional maps, based on the results of many studies, showed that climate change would significantly affect the global distribution of IAWs (Ahmad *et al.* 2019; Guan *et al.* 2020). Many studies have reported a range expansion for invasive species under climate change (Taylor *et al.* 2012, Cunze *et al.* 2013, Buczkowski and Bertelsmeier 2017, Kadioglu and Farooq 2017, Wei *et al.* 2017, Trethowan *et al.* 2011, Bradley BA *et al.*

2012, Bellard 2013, Priyanka and Joshi 2013). The possible reason may be that these species incline to expand their ranges with increasing temperature under climate change scenarios (Ju *et al.* 2015). In particular, the areas adjacent to the current distribution range of the species fall under high risk of invasion (Ahmad *et al.* 2019). However, many researchers reported that climate change may constrain the geographic range of some IAW species (Bradley 2009, Taylor and Kumar 2013, Roger *et al.* 2015, Allen and Bradley 2016, Manzoor *et al.* 2018).

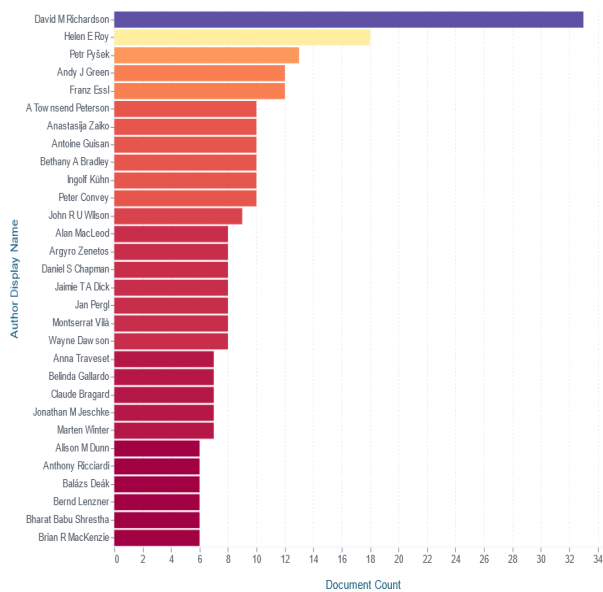


Figure 4. Leading authors who have contributed more to studies on the weed invasion in future climatic conditions

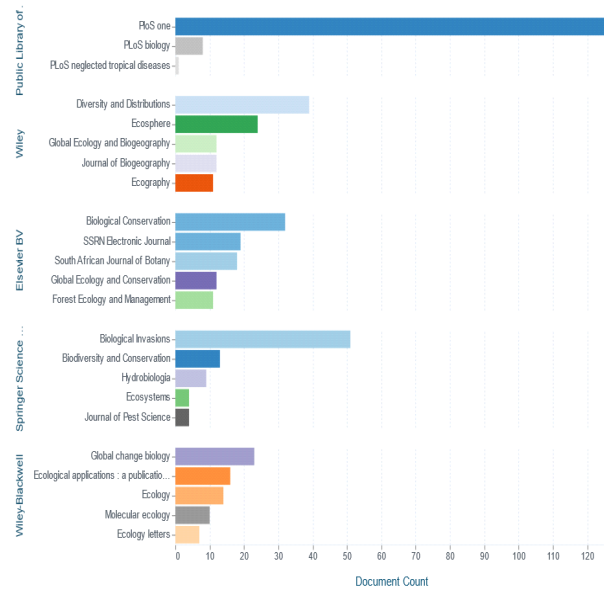


Figure 5. The leading journals which published work on weed invasion under future climatic scenarios

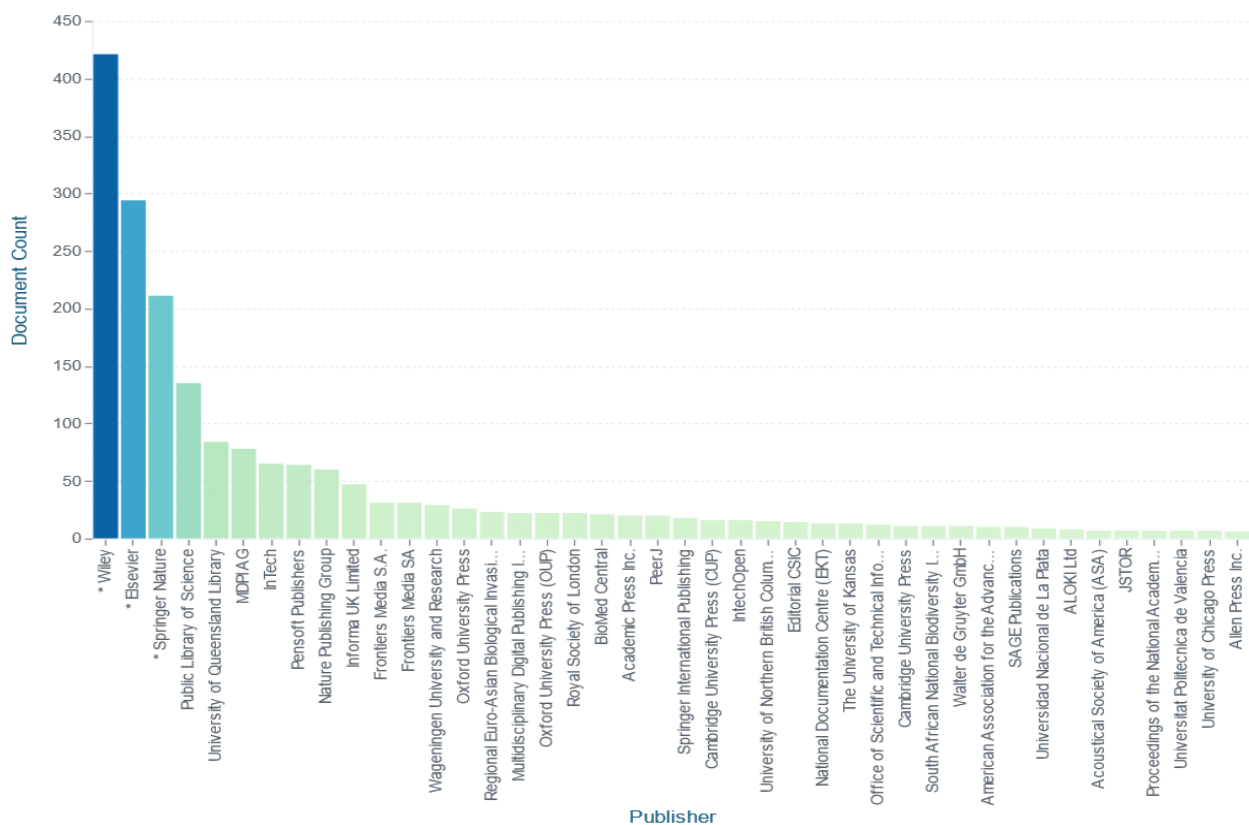


Figure 6. Leading publishers of the work on future distribution of invasive alien weeds

As far as high-risk regions are concerned, alarming situation will put forth the challenges for the policy makers, land resource managers and for other stakeholders to develop effective management plans in order to prevent the introduction, and further if it fails, then to control the further spread of this invasive species in High-Risk zones.

Limitations of SDM and future prospects

Many studies suggest that invasive alien species conquers climatic niches similar to those of its native places in some regions and such a resemblance in the climatic space between the native regions and invaded areas is considered to be critical factor for successful invasion in the non-native places (Becerra Lopez *et al.*

2017; Ficitola *et al.* 2007; Ahmad *et al.* 2019; Banerjee *et al.* 2019). Whereas, in some of the studies it is evident that the species when introduced into a new area, can simultaneously occupy climatic niches different from its native range, thereby it is necessary to reaffirm the fact that same species can exhibit variable invasion niche dynamics in an invaded regions (Wei *et al.* 2017; Becerra Lopez *et al.* 2017; Goncalves *et al.* 2014).

Many studies have successfully identified the invaded/hotspot areas and also determined the effect factors for the spreading of the invasive plants using climate factors (Welk *et al.* 2002; Kriticos *et al.* 2003; Tu *et al.* 2021), while others projected distribution scenarios and invasion trends of the plants in future (Van Wilgen *et al.* 2016; Tu *et al.* 2021). Although species distribution modelling plays a vital role in predicting the future potential distribution range of species under different climatic scenarios, such models are still relied on mainly abiotic factors. But, there are many other essential factors that affect the introduction and invasion process of invasive alien plant species and need to be considered in prediction (Coulin *et al.* 2019), for example, soil type, land use and biotic pressure, and especially anthropogenic factor (Zhu *et al.* 2017). Some biotic factors, such as competition, dispersal ability along with abiotic factors, also affect the potential distribution of species in future (Wisiz *et al.* 2013). Therefore, the future research challenge is to incorporate the biotic factors along with other factors such as land use and land cover changes in the SDMs to have a more sophisticated representation of the species distributions under changing climate (Bellard *et al.* 2016).

Currently, species distribution models (SDMs) are being widely used to predict distribution of IAWs at global (Guisan & Zimmermann 2000; Guisan and Thuiller 2005) as well as regional level. Though, one main challenge with the use of these SDMs is the selection of the most appropriate algorithm and suitable methodology among all available large number of modelling algorithms which are increasing at a rapid pace too (Elith *et al.* 2010). Recent studies revealed the difficulties in making the choice of the appropriate modelling algorithm due to varied performance of different algorithm. To avoid such situation, there is an emerging scientific trend to use several algorithms concurrently [*e.g.* ensemble modelling (Araujo and New 2007; Thuiller *et al.* 2009)] within a consensus modelling framework (Thuiller 2004, Marmion *et al.* 2009). By combining different algorithms for predictions, these ensemble modelling approaches accounts for uncertainties of using single algorithm (Buisson *et al.* 2010,

Grenouillet *et al.* 2011) and hence increasing the predictive power of distribution modelling and projection (Marmion *et al.* 2009).

CONCLUSION

Climate change can cause huge impacts on the global ecosystem, change in areas of crop production and spread of weed species. An understanding the impact of climate change on weed species' future invasion is important for sustainable biodiversity conservation. This study summarized the important issues related to modelling weed invasion in future along with bibliographic analysis of the literatures related to weed invasion in future climate scenarios. The positive and negative economic and ecological consequences of species invasion everywhere are important concerns to all stakeholders of the society. The identification of areas where policies could benefit from synergies between climate, land use change and invasive species management is of prime relevance.

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