Habitat suitability model of endangered *Latidens salimalii* and the probable consequences of global warming

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Abstract
Salim Ali’s fruit bat, *Latidens salimalii*, is a monotypic endangered fruit bat endemic to Western Ghats (WG) with an ambiguous distribution. The distribution range, habitat suitability, and biology of this species are still uncertain. Endemic species inhabiting the high elevation of WG like *L. salimalii* are threatened due to climatic change and seeks urgent management interventions. Hence, we developed a habitat suitability model for *L. salimalii* using MaxEnt in the current climate condition and projected their distribution for three Representation Concentration Pathway (RCP 4.5, 6.0, and 8.5) climate scenarios of the 2070 time frame. The results show that 9531 km² of habitat in WG is suitable for *L. salimalii* at present, while all the future scenarios estimates propose complete loss of highly suitable habitat. The significant factors influencing the distribution of *L. salimalii* are the precipitation of the driest month, tree density, rain in the coldest quarter, canopy height, and altitude. The study pioneers in predicting the suitable habitat and emphasis the need to develop strategies for the long-term conservation of endangered *L. salimalii* in WG under global warming scenarios.

Keywords Climate change · Endemic species · Habitat loss · Maxent · Western Ghats

Introduction

Bats are a highly threatened group of mammals fronting population drop owing to several reasons. These keystone species play a pivotal role in the modulating ecosystem in which they live. They are responsible for pollination and long-distance dispersal of seeds and fruits, by which it influences the lifecycle of tropical plants (Castaño et al. 2018). Reducing the abundance of such vital species can lead to diminished seed dispersal and pollination, thereby upsetting the complete balance of the ecosystem dynamics (Mokany et al. 2013; Castaño et al. 2018). Bats face massive perils due to emerging diseases, depletion of the habitat, hazardous effects of pesticides, and timber harvesting. The endemic Salim Ali’s fruit bat *Latidens salimalii* Thonglongya (1972), is a critically endangered nocturnal animal distributed above an altitude of 800 MSL (Molur and Vanitharani 2008) in the Western Ghats (WG). Approximately, 1200 individuals of the species are only cited by different authors from various localities in Southern WG (Thonglongya 1972; Bates and Harrison 1997; Ghosh et al. 1999; Vanitharani et al. 2004; Agoramoorthy and Hsu 2005; Wordley et al. 2016; Joy et al. 2019).
**Methodology**

**Study locale**

The WG is an extraordinarily delicate assorted ecosystem accommodating rich endemic and imperilled flora and fauna, making it a hotspot of diversity (Cincotta et al. 2000; Myers et al. 2000). The 1600 km long mountain chain on the west coast of peninsular India is unique owing to diverse topography, varying altitude, diverging climate, and an array of habitats. Humid tropical conditions prevail in the lower elevations, while the temperate climate is predominant in the upper reaches with an annual average temperature of 15 °C. The WG receives both the Northeast and Southwest monsoon and receives the average rainfall of 300-400 cm. In the present scenario, the WG is confronting severe threats from the developmental activities, deforestation, forest encroachment, infra-structure developmental actions, agricultural expansion, hydro-electric projects, mining, timber logging, forest product extraction, and other human-induced stress (Menon and Bawa 1997; Priti et al. 2016; Sen et al. 2016; Sony et al. 2018; Raman et al. 2020). Identical to other tropical montane ecosystems, WG is in a rapid transforming stage, but the exact pace and form are poorly understood. The deviations in the land use and land cover pattern have a strong influence on this fragile ecosystem (Sukumar et al. 1995; Menon and Bawa 1997). The habitat alteration for human needs, unempirical tourism projects, and the spread of exotic or invasive species already drove many species to extinction. The diminution of shola-grasslands and replacement with exotics dominates the landscape modification with erratic backlashes.

**Occurrence records**

We conducted an in-depth survey on the western slopes of the Southern WG to assess the presence of *Latidens salimalii*. The survey sampled 38 locations across the southern slopes to determine the presence of this species during 2017–2018. In 2019, we repeated the sampling of *L. salimalii* along with the Kerala Forest and Wildlife Department in selected protected reserves. A four-shelved mist-net of 38 mm mesh size is placed in the sampling localities to monitor the bats following the protocols of Sikes and...
to 16 from 69 (Fig. 1).

multiple occurrences, and reduced the incidence records spatially auto-correlated incidence points and tumbling the Lammens et al. 2015) in the R platform. We eliminated the L. salimalii autocorrelation between the sightings of was published records (Supplementary table S1). The spatial Earth. We compiled 69 records from our survey and other were also collected and geo-referenced using the Google lan and Marimuthu 2003a, b; Vanitharani et al. 2004; Ago- ramoorthy and Hsu 2005; Molur and Vanitharani 2008; Babu et al. 2013; Wordley et al. 2016; Joy et al. 2019) were also collected and geo-referenced using the Google Earth. We compiled 69 records from our survey and other published records (Supplementary table S1). The spatial autocorrelation between the sightings of L. salimalii was rectified by thinning using the package “spThin” (Aiello-Lammens et al. 2015) in the R platform. We eliminated the spatially auto-correlated incidence points and tumbling the multiple occurrences, and reduced the incidence records to 16 from 69 (Fig. 1).

Environmental variables

We used the tree density, canopy height, elevation, and the nineteen bioclimatic variables as the predictor variables for distribution modeling. We downloaded the elevation layer from the NOAA website (https://www.ngdc.noaa.gov/mgg/topo/globe.html), the tree density from Crowther et al. (2015), global forest canopy heights from Simard et al. (2011), and the 19 bioclimatic predictor variables from the Worldclim database (https://www.worldclim.org) at 1 km scale. We selected only the variables having limited effects on the L. salimalii predictive models to get an optimum predictive model with reduced masking effects. Veloz (2009) reported that even mildly correlated layers could influence the precision of species distribution models. Hence we choose the final set of the environmental variable after accounting for multi-collinearity to generate a model with high accuracy. We finalised seven variables for the Habitat suitability modeling (HSM) by estimating the variance inflation factors (VIFs) in R package “usdm” (Naimi 2013), ignoring the variables scored higher than 5 (VIFs > 5). We used the bio3 (isothermality), bio14 (pre- cipitation of driest month), bio18 (precipitation of warm- est quarter), and bio19 (precipitation of coldest quarter), one topographic layer (altitude), and two vegetation layers (tree density and canopy height) to develop habitat suitability models. bio3 (isothermality) is (bio2/bio7 × 100) in which bio2 = mean diurnal range (mean of monthly (max temp – min temp)), bio7 = temperature annual range.

Modeling and corroboration

MaxEnt software Version 3.3.3k (Phillips et al. 2006) works on the maximum-entropy is used to develop the habitat suit- ability model of L. salimalii. This model also has proven its efficiency in monitoring bats in high elevation ranges (Debata et al. 2019). The software uses environment variables and geo-referenced localities to predict log transformations or raw exponential outputs. The models having maximum entropy is the best model, subjected to certain constraints. MaxEnt is used universally as the preferred modeling choice for the extrapolation of species distribution with high precision (Bosso et al. 2018; Soucy et al. 2018; Zhang et al. 2018). The software initiates running, assuming a uniform distribution, and performs continuous iterations, which increases the suitable location possibility (Merow et al. 2013). We identified all the selected seven variables to run the model as continuous variables. The logistic output was used, which is the estimated probability of the binary argument of a species based on the given environmental variables (Merow et al. 2013). Logistic output also permits us to make a dissimilarity between the appropriateness of different areas.

We adjusted the existing models by shifting the regular- ization multiplier (rm) values. We transformed the model complexity by changing different feature types like linear (L), product (P), quadratic (Q), and hinge (H). Using the “ENMeval” (Muscarella et al. 2014) package in R platform, we developed forty-eight (48) models with varying levels of complexity and different regularization multiplier values. Among the developed models, the best fit model was chosen based on the lowest corrected Akaike Index Criterion (L-Q and rm 0.5) for the current habitat modeling (Supplemen- tary Table 2). We generated a bias grid using the R pack- age “KernSmooth” (Wand, 1994) by calculating the Gaussian kernel density of sampling localities, considering the potential bias in sampling location data. We run the model for L. salimalii with 5000 iterations, and 16 replicates with a subsampling procedure, among which 20% of iterations were for training, and the remaining (80%) was for testing. The significance of the contribution of all the seven envi- ronmental variables was estimated using the Jackknife proto- col. We mapped the Maxent outputs using the geographic information system application QGIS ver 3.14.15. The EOO (Extent of Occurrence) of L. salimalii was determined using the software GeoCAT (https://geocat.kew.org). The change in suitable habitats under future climatic conditions was estimated using DIVA-GIS (https://www.diva-gis.org).

Habitat alteration forecasting in the climate change scenario

The future climatic projections were derived using the bio- climatic variables for three Representative Concentration
Fig. 1 Locations of *Latidens salimalii* in the WG mountain chains in India
Pathways (RCP 4.5, RCP 6.0, and RCP 8.5). We followed the Hadley Global Environment Model 2-Atmosphere Ocean (HADGEM2-AO) of 30 arc-second (~1 km) spatial resolution, as presented in the fifth assessment report (AR5) of the Intergovernmental Panel for Climate Change (IPCC 2014). RCP 4.5 represents an optimistic emission scenario, where emissions will peak around 2040 and then decline. RCP 6.0 is also a stabilizing emission scenario where the emission will get stabilized after the 2060s. RCP 8.5 assumes increased greenhouse gas emissions throughout the twenty-first century (Sharma et al. 2017). Projections are made up to 2070 in the WG region using the future climatic variables downloaded from the Consultative Group on International Agricultural Research’s (CGIAR) Research Program on Climate Change, Agriculture and Food Security (CCAFS) climate data archive (https://www.ccafs-climate.org/data). We treated the topographic features for future projection as static and dynamic non-climatic variables are expected to change in the future (ex. vegetation cover) were omitted from this analysis.

Model appraisal

Two metrics, the area under receiver operating characteristic curve (AUC) and the true skill statistics (TSS), were used for model appraisal. AUC is a threshold-independent metric that measures the model’s ability to distinguish between random and background points to evaluate the model performances. A high AUC score does not always reflect that the models are highly informative (Philips et al. 2006), and the evaluation based on the AUC score alone is not reliable (Austin 2007; Lobo et al. 2008). Hence, we also estimated the TSS scores, which is a threshold dependent measure of accuracy. The formula sensitivity + specificity − 1 is used to calculate TSS, where sensitivity and specificity are measured based on the probability threshold for which their sum is maximized (Allouche et al. 2006).

Results

We recorded *L. salimalii* from the Thondiyar of Periyar Tiger Reserve (PTR) and Silent Valley National Park (SVNP). Our identification is the pioneering one establishing endangered *L. salimalii* from PTR and SVNP (Fig. 2). The observation extends the range of *L. salimalii* towards the north of Palghat gap. The morphometric measurements of the specimen observed from these reserves and the first specimen from the High Wavy Mountains remain almost similar (Table 1).
Model performance

The model performance was evaluated based on the AUC and TSS values. The best performing model had an average AUC of 0.95 and a TSS value of 0.80. The accuracy of the predicted model is 0.80, the sensitivity is 1, and the specificity is 0.80. These values indicate that the model performed well in predicting suitable habitat of *L. salimalii* in the WG region of peninsular India.

Important environmental variables

From the seven environmental variables, the most significant variable identified to influence the spatial distribution of *L. salimalii* is the precipitation of the driest month (58.6%), tree density (34%), elevation (3.4%), and the rainfall of the coldest quarter (3.2%). The cumulative contribution of these variables is 99.2%. These variables form the factors responsible for the highest permutation importance setting their most tremendous significance in the final model (Table 2). The Jackknifing protocol indicates that these variables significantly contribute to the predicted model compared to the other variables (Fig. 3). The logistic output (interpreted as portability of presence) of *L. salimalii* peaked towards the higher value of the variables altitude, bio14, bio18, bio3, and tree density. The other variable, like bio 19 and canopy height, indicated a negative influence of species distribution with increasing values. It clarifies that the preferred vegetation type by *L. salimalii* is short, stunted montane one than the tall trees in the lower tropical climate. Similarly, the local storms in the unexpected seasons other than the traditional monsoon are also not favourable for survival.

Currently suitable habitat of *L. salimalii*

The current model revealed that the regions having appropriate and optimal environmental conditions for the species lie mostly in the southern WG, especially to the south of the Palghat gap (Fig. 4). The Mukurthi National Park, Silent Valley National Park, Indira Gandhi (Anamalai) Tiger Reserve (TR), Eravikulam National Park, Pamadam shola National Park, Anamudishola National Park, Mathikettan-shola National Park, Periyar Tiger Reserve, Shenduruny Wildlife Sanctuary, and Kalakkad Mundanthurai Tiger Reserve are highly suitable habitat for survival and reproduction of *L. salimalii* under the current scenario. Overall potential distribution areas of *L. salimalii* in WG are estimated to be 21,089 km², which are found higher than the EOO (17,183.13 km²). Of this total suitable area, 9531 km² (45.19%) has a very high potential as the critical habitat of *L. salimalii* (Table 3).

| Measurements | Periyar Tiger Reserve | Silent Valley National Park | Bates and Harrison (1997) |
|--------------|-----------------------|----------------------------|---------------------------|
| Forearm length | 69.3 | 70.2 | 67.3 |
| Head-body length | 111.8 | 108.7 | 106.7 |
| Ear | 17.8 | 17.1 | 16.8 |
| Tibia | 33.1 | 32.3 | NA |
| Hind foot | 13.2 | 13.1 | 13.4 |
| Third metacarpal | 48.3 | 47.9 | 47.4 |
| Fourth metacarpal | 44.5 | 43.9 | 45.4 |
| Fifth metacarpal | 46.7 | 44.7 | 45.1 |

Table 2. Contribution percentage and permutation importance of correlated environmental variables used for HSM

| Variable | Percent contribution | Permutation importance |
|----------|----------------------|------------------------|
| bio_14 [precipitation of driest month(mm)] | 58.6 | 37.5 |
| Tree_density | 34 | 29.5 |
| Altitude | 3.4 | 10.3 |
| bio_19 [precipitation of coldest quarter (mm)] | 3.2 | 19.8 |
| Canopy_height | 0.4 | 1.2 |
| bio_3 (isothermality) | 0.3 | 1.5 |
| bio_18 [precipitation of warmest quarter(mm)] | 0.1 | 0.1 |

Changes in habitat suitability and loss in protected areas

We estimated the future changes in potential habitat by taking the difference between the areas under current and future distribution (Table 3). The model predictions under the RCP 4.5, RCP 6.0, and RCP 8.5 scenarios for the 2070 time frame revealed that the areas currently identified as suitable would become increasingly unsuitable for *L. salimalii* by 2070 (Fig. 5). The relatively optimistic climate change scenario (RCP 4.5) predicted a decline of high and very-high potential areas likely by 81.30% and 99.78%, respectively. The extreme climate change scenario RCP 8.5 predicted the maximum range loss of high potential areas (99.57%) and high potential (100%). The Mukurthi National Park, Silent Valley National Park, Anamalai Tiger Reserve (ATR), Eravikulam National Park, Mathikettan-shola National Park, Periyar Tiger Reserve, Shenduruny Wildlife Sanctuary, and Neyyar Wildlife Sanctuary, and Kalakkad Mundanthurai Tiger Reserve are all highly suitable habitat for survival and reproduction of *L. salimalii* (Table 3).
Fig. 3 Response curves showing relationships between the important environmental variables and probability of presence of *Latidens salimalii* and Jackknife test showing the relative importance of different environmental predictors. The values shown are an average of ten replicates of cross-validated Maxent models.
Fig. 4 Habitat suitability for *L. salimalii* in WG under the current climatic conditions, vegetation and topographic feature.
Tiger Reserve assessed as highly vulnerable to extreme climates. We predict the complete absence of suitable and very high suitable habitat for *L. salimalii* in the future scenario under RCP 8.5%. These protected areas may experience drastic habitat loss for *L. salimalii* due to future climate modification.

**Discussion**

*Latidens salimalii* is a non-charismatic species, and the research on its ecology is rare with the actual distributive ranges remain indefinite to date. The study predicts the distribution range of *L. salimalii* on either side of the Palghat gap with more distribution towards the southern side. The distribution towards the north of the Palghat gap was a speculation, and we validated the presence from the Silent Valley National Park. The previous studies believe that the principal distribution range is towards the south of Palghat (Molur et al. 2002; Molur and Vanitharani 2008; Wordley et al. 2016). The Agastialmalai forms the core habitat of *L. salimalii* (Agoramoorthy and Hsu 2005; Molur and Vanitharani 2008), with the range extends to the Valparai plateau and Malayatoor forest division. The predictive map indicates that the suitable predicted habitat range spreads over the Nilgiris up to Karnataka. Roosting sites are primarily noted as caves and abandoned buildings inside the forested areas (Singaravelan and Marimuthu 2003a; Vanitharani et al. 2004; Agoramoorthy and Hsu 2005; Joy et al. 2019) and feed on specific floral species (Vanitharani et al. 2004; Wordley et al. 2016). The model prediction postulates the preference of the bat towards dense wooded high altitude vegetation and the specific monsoon patterns prevailing in the region. The floral structure prevailing in the unique climax community of Western Ghats is so typical with highly endemic fruit yielding evergreen and moist deciduous vegetation. The distribution of *L. salimalii* shows a negative correlation with canopy height, after an optimum range, indicates the preference towards stunted montane foliage. The studies on the structure of high altitude climax vegetation of WG (Jose et al. 1994; Bunyan et al. 2012) clearly showed that the height of the trees rarely exceed 12 m (stunted) and have dense canopy cover.

Debata et al. (2019) observed that the elevation and habitat act as chief determinants of chiropteran distribution, similar to our observations. The high altitudes as generally less disturbed and have rocks, crevices, and caves to provide enough space to roost and rear young ones. The trophic structure gets modified so to enable them to depend more on the native flora. The availability of food determines the biomass of folivorous and frugivorous species in a region (Chapman et al. 2005). The tree diversity in a temperate climate is comparatively fewer than in a tropical environment. The relatively lesser floral diversity creates a strong dependency between the endemic bats and native flora (Castanño et al. 2018). The changing floral pattern, according to altitudinal gradation, restricts the range of habitat specialist bats to a specific trophic structure. Global warming can kindle the range reduction of an endemic habitat specialist (Bellard et al. 2012), as the escalation of temperature, in turn, can restrict the montane climax vegetation to a short-range. The grid-based models predicted by Gopalkrishnan et al. (2011) pointed out that 77% of grids under the forest cover may face threats due the climate change shortly. Mokany et al. (2013) reported that plants with small fruits suffer long-distance seed dispersal under climate change, contributing to extensive compositional alterations in the forest flora. Our results by HSM suggest that there will be no suitable habitat of *L. salimalii* under the RCP 8.5 scenario, which could cause high extinction risk.

RameshKumar et al. (2004) predicted that the warming of the Indian sub-continent would occur shortly with an increase of temperature, both day and night. The increase in the night temperature will be more than the diurnal temperature in the coming years. Climate factors such as temperature and precipitation are strongly correlated to the biogeographical patterns of a species (Sousa-Guedes et al. 2020). The *L. salimalii* is influenced by rainfall and altitude, which also designates that the species is climate-dependent. In terms of precipitation in the future years, there will be an increase of unpredicted intense rainfall in WG and Peninsular India (RameshKumar et al. 2004). The anticipated increment of precipitation in the WG may seriously influence

| Habitat suitability       | Current km² | 2070 RCP4.5 | 2070 RCP6.0 | 2070 RCP8.5 |
|---------------------------|-------------|-------------|-------------|-------------|
| Least potential (0–0.13)  | 112,912     | 119,363 (−5.71%) | 124,822 (10.55%) | 130,338 (15.43%) |
| Moderate potential (0.13–0.25) | 4762     | 8965 (−88.26%) | 7473 (56.93%) | 3207 (−32.65%) |
| Good potential (0.25–0.5)  | 6796        | 5743 (15.49%) | 2364 (−65.21%) | 1255 (−81.53%) |
| High potential (0.5–0.75)   | 5037        | 942 (81.30%)  | 363 (−92.79%) | 223 (−95.57%)  |
| Very-high potential (0.75–1) | 4494      | 10 (99.78%)  | 0 (−100.00%) | 0 (−100.00%)  |
The atmospheric CO₂ concentration, temperature, or precipitation will directly affect metabolic and developmental rates in many animals (Hughes 2000). The reproductive requirements, habitat selection, and foraging patterns of bats may also have unequal effects due to changes in temperature (Sherwin et al. 2012). *Latidens salimalii* is a cave-dwelling species and is very sensitive to disturbances (Vanitharani et al. 2004) can be more influenced by the changing climate. The bats have specialised roosting types, specifically in cave and tree (Sherwin et al. 2012). At different stages of life-history, they may face significant risk when there is an alteration in vegetation and climate. According to the present projections, the twenty-first century will confront a quick climate change that could influence the distribution and diversity of climate specialist animals. The Intergovernmental Panel on Climate Change (IPCC) gauged a severe danger of termination for 20–30% of species with the ascent of 2–3 °C temperature globally (Stocker et al. 2013; Warren et al. 2013). Under the evolving climate, the atmospheric pressure alters (Shukla et al. 2003) over WG, which upsets the too delicate ecological balance accelerating the increased extinction of dependent fauna and change of floral composition.

A mutualistic dependency exists between native flora and the critically dependent frugivorous fauna. This association is crucial, as even a small imbalance can lead to the extermination of both the species (Jordano et al. 2011). Hence, any impact on the endemic or native flora is reflected by the extinction of the habitat specialist, either through climatic variation or other synthetic reasons. The fundamental inter-dependence is vital for numerous ecological processes. The co-evolution, interactions, colonization, and large-scale range dynamics, life-history, and seasonal changes in natural habitats seem related (Jordano et al. 2011). Many species respond to climate change by shifting their distribution pattern to suitable other habitats for existence (Mokany et al. 2013). The alteration in the climatic conditions determines the productivity as well as the presence of native high-altitude flora. Studies by Gopalakrishnan et al. (2011) and Chaturvedi et al. (2011), based on the vulnerability index, estimated that the WG is highly vulnerable to climate changes than the Northeastern forests. The modification of

**Fig. 5** Predicted future distribution model of *L. salimalii* under RCP 4.5 (a), RCP 6.0 (b) and RCP 8.5 (c) emission scenarios. The logistic output raster classified into five potential classes viz., very high potential (>0.75), high potential (0.5–0.75), good potential (0.25–0.5), moderate potential (0.13–0.25) and least potential (<0.13).
climate can adversely affect the regional water resources (Wagner and Weitzman 2015), thereby impacting local diversity unfavorably. Trophic structure alterations are visible from the climatically sensitive areas (Walther et al. 2002). The transformed climatic environments upset and prolong the growing seasons, which give an upper hand for the invasive or non-native species. The tropical montane habitats like WG hosts many threatened taxa with restricted distribution, which can be vulnerable to local extinction due to increased human footprint and spread of invasive species (Arasumani et al. 2019). The pace of spreading invasive species (flora and fauna) can be at a high rate in a changing environment, as they quickly occupy novel niches. The successful invaders are phenologically flexible species that inhabit regions where endemic species thrive (Moran and Alexander 2014). The exotic invader flora has identical characteristics of endemic flora and a synchronously dispersal pattern. The external factors like climatic forcing act as a significant factor in altering the trophic structure by replacement (Walther et al. 2002). Due to complex topography and biogeographic history, fragile montane ecosystems respond quickly to climate change (Sukumar et al. 1995).

The current study is the pioneer one to demonstrate the potential risk and anticipates climatic impact on the viability of L. salimalii. The reported threats to the species are hunting for medicinal value and timber logging (Molur and Vanitharani 2008). In the western slopes of the WG, after the outbreak of the Nippa epidemic and COVID 19 pandemic, bats in general, are facing threats from the local populace. Bats were reported as carriers of viruses for many emerging zoonotic diseases and closely related to the corona virus like SARS (Wang et al. 2006; Lau et al. 2020). Frugivorous bats host the Nippa virus (Yadav et al. 2012), and the virus is spreading through secretions. No zoonotic epidemic is recorded from the roosting regions of L. salimalii, as the human may not prefer to share the native berry fed by L. salimalii. The HSM on Nilgiritragus hylocrius (Sony et al. 2018), Herpestes fuscus fuscus (Raman et al. 2020), and Piper nigrum (Sen et al. 2016) also predict a complete loss of suitable habitats in the southern WG in the future emission scenario underlines the change of floral texture. This alteration on the indigenous floral composition can be through the succession of exotics or resistive species over the native flora. The evolution of a new floral pattern is a threat to the endemic fauna–floral interaction dynamics that can lead to the culmination of reliant fauna. The transformation predicted for the L. salimalii, is a predictive picture for the changing trophic composition, indicating the changing floral design of the WG. Hence, we expect the southern WG is changing in the changing climatic scenario, contributed by a multitude of reasons.

In conclusion, global warming is a threat to climate specialist endemic fauna existing in the high altitude regions. These threats reach its summit with the changing land use and land cover patterns. The protection of L. salimalii in the changing climate scenario is probable only by protecting the native fragile ecosystem of WG anticipating the change in the climate. The study practically implies the conservation of the entire Western Ghats is by protecting its unique landscape, preservation of native flora, check on the exotic invasion and anthropogenic impediments. Intense surveys on the predicted habitat will help to locate more populations and comprehend the ecology of this lesser-known species. Understanding the people’s perception, especially in the pandemic situation, creating awareness and involvement of the local community can engender better results towards conservation. We suggest exhaustive research on the ecology of this species to develop a suitable management plan encompassing the micro and macro variables to enrich habitat quality and quantity.

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