Paleoecological niche modeling of *Epiophlebia* (Epiophlebioptera: Epiophlebiidae) reveals continuous distribution during the Last Glacial Maximum

Aaron M. Goodman a,b,c,*, Christopher D. Beatty a,d, Sebastian Büsse e, Hidenori Ubukata f, Toshiyuki Miyazaki g, Mary E. Blair h & Jessica L. Ware a,b

a American Museum of Natural History, Division of Invertebrate Zoology, NY, NY, 10024
b Richard Gilder Graduate School, Department of Invertebrate Zoology, American Museum of Natural History, NY, NY, USA, 10024
c Graduate School and University Center, City University of New York, NY, NY, 10016
d Department of Biology, Stanford University, Stanford, CA, USA
e Department of Cytology and Evolutionary Biology, Zoological Institute and Museum, University of Greifswald, Germany
f Professor Emeritus, Hokkaido University of Education, Sapporo, Japan
g Kansai Research Group of Odonatology, Osaka, Japan
h Center for Biodiversity and Conservation, the American Museum of Natural History, New York, NY, USA, 10024
*Corresponding author: Email: agoodman@amnh.org

Abstract. Disjunct biogeographic patterns of similar species remain enigmatic within evolutionary biology. Disparate distributions typically reflect species responses to major historical events including past climate change, tectonics, dispersal, and local extinction. Paleo-ecological niche modeling (PaleoENM) has proven useful in inferring the causes of disjunct distributions within charismatic and well-studied taxa including mammals, plants, and birds, but remains under-explored in insects. The relictual Asian dragonfly genus *Epiophlebia* (Suborder Epiophlebioptera: Epiophlebiidae) allows us a novel opportunity to explore PaleoENM in the context of disjunct distributions due to their endemism to the Japanese islands, Himalayas, China, and North Korea. The aim of this paper is to investigate the potential causes behind the modern distribution of *Epiophlebia* by inferring the historical range of these species within the Last Glacial Maximum (LGM), thereby highlighting the utility of PaleoENM in the context of odonate biogeography. Our results indicate possible past routes of gene flow of *Epiophlebia* during the LGM due to high habitat suitability of the genus stretching from the Himalayas to Japan. Furthermore, our results predict several unsampled areas which have the potential to harbor new populations of the genus.

Key words. Odonata, Anisozygoptera, dragonfly, Epiprocta, *Epiophlebia*, Ecological Niche Modeling, Maxent

Introduction

Dragonflies and damselflies (Order: Odonata Fabricius, 1793) represent an overlooked yet invaluable study system for systematics, ecology, and biogeography (Córdoba-Aguilar et al., 2023). Their evolutionary history stretches back ~ 360 mya, with ~ 6405 extant and fossil species currently described (https://www.odonata-central.org/app/#/wol/, accessed July 25th 2023; Nicholson et al., 2014). Habitats range from single-stream endemics to transoceanic pandemics, and odonate natural history connects both freshwater and terrestrial habitats (Collet et al.,
Paleoecological niche modeling of *Epiophlebia* is the most extensively studied species, with nymphs preferring streams with temperatures as low as 0.6°C in the winter and 10–22°C in the summer, with an extreme record of 23°C, with altitudes ranging from 1–2100 m (Asahina, 1948; Gose, 1953; Ishida et al., 1959; Okazawa, 1974; Tokuyama, 1992; Ubukata & Miyazaki, in press; Yamaura et al., 2009). Both *E. superstitus* and *E. laidlawi* possess the longest development time of any odonate ranging from 5–8 years for both species (Dorji et al., 2020; Ozono et al., 2021).

Studies by Brockhaus & Hartmann (2009), and J. K. Li et al. (2012) suggest that the families Epiophlebiidae and the closely related extinct Stenophlebiidae belonged to an ‘archeo-paleoarctic dragonfly fauna’, formed in the Jurassic at the breakup of Pangaea, with subsequent intermingling of oriental fauna in the Tertiary. However, phylogenetic and morphological analysis by Büsse et al. (2015) and Büsse (2016) suggest that the extant species of *Epiophlebia* constitute minor variants, and indeed genetic variation levels suggest a younger age of extant *Epiophlebia* than the Jurassic. Büsse et al. (2012) hypothesized that *Epiophlebia* might have possessed a continuous distribution within the last glacial maximum (LGM) (21 ka) allowing for a more recent genetic admixture to occur. Widespread cooling and the presence of a land bridge connecting Kyushu to Korea within the LGM might have allowed populations of *Epiophlebia* to be more widespread (Bintanja et al., 2005; Clark et al., 2009; Schmidt & Hertzberg, 2011; Tomita et al., 1975). Subsequent warming and sea level rise after the LGM might have resulted in *Epiophlebia* retreating to cooler and more elevated areas including foothill streams fed by groundwater, resulting in the disjunct populations currently seen. Furthermore, Büsse & Ware (2022) hypothesized other glacial refugia which have the climatic conditions suitable for harboring unsampled populations of current or new species of *Epiophlebia*.

The aim of this paper is to infer the historical range of *Epiophlebia* utilizing ecological niche modeling (ENM, or niche-based species distribution modeling) techniques.
ENM attempts to estimate the abiotic niche of a species by correlating occurrence and environmental data to generate an envelope of habitat suitability (Guisan & Thuiller, 2005; Guisan & Zimmerman, 2000; Zurell et al., 2020). Occurrence data is commonly obtained from natural history collections and citizen-science observations, while environmental data is derived from field stations or remote sensing in the form of climatic variables. Paleo-Ecological Niche Modeling (PaleoENM) provides the opportunity to estimate historical distributional patterns of a species by projecting modern day models onto paleoclimate layers, derived from a range of sources including global circulation models (GCM), ice sheet isotope data, or sedimentological proxies; such paleo-environmental layers can extend back into the Miocene (Brame & Stigall, 2014; Brown et al., 2018; Dudei & Stigall, 2010; Myers et al., 2015; Purcell & Stigall, 2021; Stigall, 2012). We combined verified occurrence records for all *Epiophlebia* species with averaged sets of environmental predictor variables to generate models for both modern-day and potential paleoenvironment distributions of the genus. We used these distribution models to not only infer unsampled areas of potentially suitable habitat for *Epiophlebia* but to determine if a continuous distribution of *Epiophlebia* between the Himalayas, Japan, and northern China existed within the LGM, suggesting a plausible route of gene flow.

**Methods**

**Occurrence records**

We acquired occurrence records of adult *Epiophlebia* species from the Global Biodiversity Information Facility (GBIF: doi:10.15468/dl.fq5r7), as well as various published and unpublished datasets. We selected occurrences from GBIF possessing preserved museum samples and research grade observations, which include verified latitude and longitude coordinates, a photograph of the sighting, observation date, and at least ¾ agreement on species identification by the community. We filtered occurrences by removing sightings from erroneous localities (middle of the ocean, locations of large museums). We acquired additional localities of *E. laidlawi* within India, Nepal, and Bhutan from Brockhaus & Hartmann (2009) and references cited therein, Dorji (2015), Tani & Miyatake (1979), Asahina (1961a, 1961b), and personal data from Büsses et al. (2012), as well as uncatalogued specimens housed within the Zoological Museum, Cambridge, UK, and the National Museum of Natural History, Smithsonian, USA. We acquired occurrence records of nymphs of *E. superstes*, focusing on maximum and minimum altitudes and temperate regions acquired by HU through a questionnaire survey sent out to more than 100 Japanese dragonfly researchers, as well as a substantial literature search. We also analyzed distribution maps of *E. superstes* nymphs and adults updated by TM’s published and unpublished datasets. Details of both data collection types are explained in detail within Ubukata & Miyazaki (in press). Finally, only two localities are known for *E. sinensis* within Eastern China and North Korea, which we acquired from J. K. Li et al. (2012) and Fleck et al. (2013). We opted not to include the locality for the type specimen of *E. diana* due to the misplacement of the specimen (Büsses & Ware, 2022), as well as the vagueness of the locality being collected by “Dr. David C. Graham in the mountainous regions of western Szechuan” (Needham, 1930).

We generated three datasets consisting of occurrences for *E. superstes* (n = 447), *E. laidlawi* (n = 102), and a combined dataset of all three species (n = 551). Although qualitative data on *Epiophlebia* suggests all three species inhabit similar niches (high altitude, cold streams), an assumption of genus-level ENMs is niche similarity among all species, as the same environmental data are inputted into a singular model. As such, we generated species-level ENMs for *E. superstes* and *E. laidlawi* to support our approach and check for similarity in environmental responses compared to our combined model. We were unable to generate models for *E. sinensis* due to a very low sample size (n = 2).

**Environmental data**

We acquired averaged sets of environmental predictor variables for modeling consisting of purely bioclimatic variables. All analyses were conducted using the statistical programming language R v. 4.1.2 (R Core Team, 2021), and all layers are in a latitude/longitude coordinate system using WGS84 datum. We acquired environmental rasters at 2.5 arc-second resolution (~ 5 km at the equator) from the CHELSA climate database v2.1 (Karger et al., 2017, 2023). We downloaded the ‘Anthropocene’ bioclimatic dataset (1979–2013) consisting of 19 bioclimatic variables extrapolated from monthly minimum and maximum temperature, humidity, mean, and coefficient of variation of annual solar radiation. Variables follow Worldclim v2 (Fick & Hijmans, 2017). Although occurrence records for *Epiophlebia* species range from the past 100 years, we considered the broad patterns observed within these long-term datasets as sufficient to accurately represent the environmental conditions for the taxa sampled.

**Species distribution modeling**

Before modeling, we processed our occurrences to account for sampling bias, delineated a study extent to sample background points, and omitted highly correlated environmental variables. We spatially thinned occurrences to the resolution of our environmental variables to reduce the effects of sampling bias and artificial clustering (Veloz, 2009) using the ENMtools package (Warren et al., 2010) (Table 1). Since *Epiophlebia* are highly restrictive in their habitat requirements, we chose a study extent to include potentially unsampled areas between Nepal and Japan yet excluding large
areas outside the species’ dispersal limitations (Peter-son & Soberón, 2012), defined as point buffers around all localities buffered by 5 degrees (approx. 500 km at equator) (Fig. 1). Within this extent, we randomly sampled 50,000 background points for modeling and extracted their environmental values. We used these values to calculate correlations between variables using the ‘vifcor’ and ‘vifstep’ functions in the ‘usdm’ package (Naimi, 2017) and filtered out variables with correlation coefficients higher than 0.7 and a VIF threshold of 10. Finally, we removed bioclimatic variables 8, 9, 18, 19 after visual inspection of interpolation discontinu-ities (Booth, 2022). Although such spatial artefacts were minor within our initial study extent, we observed ma-jor breaks of climate smoothing when we extrapolated our model to a larger spatial extent which included gla-cial refugia hypothesized by Büsse & Ware (2022).

To model the distribution of *Epiophlebia*, we used the presence-background algorithm Maxent v3.4.4 (Phillips et al., 2017), which remains one of the top-performing models for fitting ENMs with background data (Valavi et al., 2021). We used the R Package ‘ENMeval’ 2.0.0 (Kass et al., 2021) for model building, parameterization, evaluation with different complexity settings, and re-porting of results. Since our occurrences varied greatly in space, we partitioned our data using the ‘leave-one-
out’ strategy (referred to as “jackknife” in ‘ENMeval’) to fully encapsulate the climatic ranges of each species, and each occurrence. We clamped our models by omitting ranges of environmental data which fall outside of our training data, to prevent projection of our model to non-analogue climates. All final models were fitted to the full dataset.

We tuned model complexity to find optimal settings (Radosavljevic & Anderson, 2014; Warren & Seifert, 2011). For tuning, in order of increasing complexity, we chose the feature classes linear (L), quadratic (Q), and hinge (H), as well as regularization multipliers one through five (higher numbers penalize complexity more).

We assessed our model using averages of threshold-dependent (omission rate) and threshold-independent (AUC) discrimination performance metrics calculated on withheld validation data, (Warren & Seifert, 2011). We assessed the 10-percentile omission rate, which sets a threshold as the lowest suitability value for occurrences after removing the lowest 10% suitability values (Kass et al., 2021; Radosavljevic & Anderson, 2014). Validation AUC is a measure of discrimination accuracy that can be used to make relative comparisons between ENMs with different settings fit on the same data (Lobo et al., 2008; Radosavljevic & Anderson, 2014). If a series of models possessed near-identical AUC values and 10-percentile omission rates, we chose the simplest model, or the one with the fewest non-zero lambda values (model coefficients).

To investigate model behavior, we examined predictor variable importance values and marginal response to suitability. Permutation importance is calculated by randomly permuting the values of all variables but one, building a new model, then calculating the difference between each model’s training AUC and that of the global model (Phillips, 2021). Marginal response curves show the modeled relationship of each variable individually with the occurrence data when all other variables are held constant and are affected by the complexity of the model settings (Phillips et al., 2017).

We made habitat suitability predictions for Epiophlebia using our ‘Anthropocene’ environmental predictor variables. Predictions were estimated using a new study extent, generated as a bounding box encapsulating all the glacial refugia hypothesized by Büsse & Ware (2022). We chose this new study extent for our predictions to estimate potential routes of dispersal of Epiophlebia during the LGM, as well as to identify potentially new under-sampled refugia hypothesized by Büsse & Ware (2022). Maxent raw predictions were transformed to a scale of 0–1 to approximate probability of occurrence using the ‘cloglog’ transformation (now referred to as our continuous prediction) (Phillips et al., 2017). We also generated a threshold prediction, calculated from the 10-percentile omission rate from our model evaluation. To estimate habitat suitability of Epiophlebia within the geologic past, we generated predictions using environmental predictor variables from the PaleoClim dataset (Brown et al., 2018). We generated predictions within eight distinct time periods: late-Holocene (Megalayan: 0.3–4.2 ka), mid-Holocene (Northgrippian: 4.2–8.326 ka), early Holocene (Greenlandian: 8.326–11.7 ka), Younger Dryas Stadial (11.7–12.9 ka), Bølling-Allerød (12.9–14.7 ka), Heinrich Stadial (14.7–17.0 ka), the Last Glacial Maximum (21 ka), and the Last Interglacial (130 ka), generated from the CHELSA database.

Finally, we calculated multivariate similarity surface (MESS) maps (Elith et al., 2010) to detect areas with novel climatic conditions between our modern-day and paleoclimate variables. MESS calculates environmental similarity by extrapolation of environmental values from our occurrence points to different climate conditions. In the case of our data, we calibrated our occurrence points by extracting environmental values from our ‘Anthropocene’ variables. We then calculated similarity between our occurrences and each of our paleoclimate layers. Negative MESS values indicate geographic regions of our paleoclimate variables falling outside our ‘Anthropocene’ variables (i.e., extrapolation), while positive values indicate regions which fall within our ‘Anthropocene’ variables (interpolation).

Results

Optimal model settings for Epiophlebia exhibited robust results based on a 10-percentile omission rate and mean validation AUC. Based on the results of the collinearity analysis, we used the following six predictor variables to build all models: temperature seasonality (bio04), maximum temperature of the warmest month (bio05), mean diurnal range (bio02), precipitation of the driest month (bio14), precipitation of the wettest month (bio13), and precipitation seasonality (bio15). Our optimal model was complex, with linear, quadratic, and hinge feature classes and a regularization multiplier of 2 (lower complexity penalty). Our mean validation AUC was very high (0.92), and a 10% omission rate that was low (0.12) relative to its expected value of 0.1. Furthermore, the optimal model possessed an AICc of 3628.27, a delta AICc of 0, and 32 non-zero lambda values, indicating a slight overfitting of model prediction. Individual models of E. superstes and E. laidlawi exhibited similar robust results (Table 1).

Temperature seasonality (bio04) possessed the highest variable importance (46%), followed by maximum temperature of the warmest month (bio05) (30%), mean diurnal range (bio02) (14%), precipitation of the driest month (bio14) (6%), while precipitation of the wettest month (bio13), and precipitation seasonality (bio15) possessed < 2% (Supplementary Fig. S1). Temperature seasonality (bio04), mean diurnal range (bio02), maximum temperature of the warmest month (bio05), and precipitation seasonality (bio15) expressed quadratic relationships with suitability. Precipitation of the driest month (bio14) also expressed a quadratic/lin-
ear relationship but with jagged optimal ranges mostly likely due to the disjunct spread of occurrence points. To preserve significant digits, bioclimatic-based temperature variables are multiplied by 10, while temperature seasonality is multiplied by 100 (O’Donnell, 2012). As such, we report the absolute values of temperature-related variables in relation to suitability. Precipitation of the wettest month (bio13) expressed a positive linear relationship with suitability, saturating at 750 mm (Supplementary Fig. S1). Suitability was highest at 50–80°C for temperature seasonality (bio04) (calculated as the standard deviation of average monthly temperature x 100), 7.5°C for mean diurnal range (bio02), 20–30°C for maximum temperature of the warmest month (bio05), 10 mm and 260 mm for precipitation of the driest month (bio14), > 750 mm for precipitation of the wettest month (bio13), and 50 mm for precipitation seasonality (bio15).

Individual models for E. superstes and E. laidlawi exhibited similar variable importance compared to our combined model, but in some cases, response curves of E. superstes and E. laidlawi exhibited opposite response curves (Supplemental Figs S1–3). Within all three models, maximum temperature of the warmest month (bio05) exhibited high variable importance. However, E. laidlawi exhibited a quadratic relationship with suitability, while E. superstes exhibited a negative linear relationship; the combined model also exhibited a quadratic relationship. Precipitation of the driest month (bio14) was also shared among our three models, however E. laidlawi expressed a negative linear relationship, E. superstes exhibited a positive linear relationship, with suitability increasing with lower rainfall, and the combined model exhibited a jagged quadratic relationship. Finally, precipitation seasonality (bio15) expressed a quadratic relationship within E. laidlawi and our combined model, but a negative linear relationship in the case of higher precipitation seasonality within E. superstes. Some variables were shared between individual species and our combined model such as temperature seasonality (bio04) which expressed a negative linear relationship within E. superstes but a quadratic relationship within our combined model. Mean diurnal range (bio02) was shared between E. laidlawi and our combined model but expressed a negative linear and quadratic relationship, respectively.

Predictions over the study extent for our ‘Anthropocene’ environmental variables revealed areas of high suitability across the Himalayas (Nepal, Tibet, Bhutan, northern India, and Myanmar), eastern coastal China (Fujian province), central China (Sichuan province), and all the Japanese Islands—these regions range between 1200–3000 m, except in Japan and North Korea, consistent with previous sightings and the biology of Epiphebia. We also observed sparse areas of high suitability within coastal South Korea and the Kamchatka peninsula of Russia. However, suitability was low within North Korea and northeast China, where sightings of E. sinensis have occurred (Fig. 2A). Areas of high suitability partially overlap with hypothesized refugia of Epiphebia from Büsse & Ware (2022), particularly within the areas of near Taiwan (Sino-Pacific refuge), northern Vietnam, Laos, and Myanmar (Yunnanian refuge), Nepal and Bhutan (Nepalese refuge). Such areas of suitability from our continuous predictions are congruent with threshold maps generated from our 10% omission rates, however coastal South Korea, the Kamchatka peninsula, and southwestern Hokkaido (where E. superstes inhabits) were classified as ‘unsuitable’ (Fig. 2B).

Predictions generated from our paleoclimate variables demonstrated variability in suitability over our time intervals compared to our modern-day ‘Anthropocene’ variables, suggesting that the occupancy of the various glacial refugia differed since the LGM (Fig. 3). Suitability shifted southward within the Holocene (0.3–12.9 ka) into Myanmar, eastern India, and Laos (Yunnanian refuge) (Figs 3B–E). Within the Bølling-Allerød (12.9–14.7 ka), suitability within the Yunnanian refuge began to decrease, but suitability extended eastward off the coast of China due to increased glaciation, recession of the ocean, and subsequent expansion of landmass into the East China Sea (Fig. 3F). Also, a land bridge was observed within Hokkaido, connecting Northern Japan with Russia, suggesting the beginning of the Würm glaciation (Clark et al., 2009); however, suitability was low within this region. Expansion of landmass continued within the Heinrich Stadial (14.7–17.0 ka), creating a land bridge between Coastal China, Taiwan, the Korean Peninsula, and the Yellow Sea with suitability being high within these areas, suggesting an increased importance of the Sino-Pacific refuge and reduced importance of the Sino-Tibetan refuge (Fig. 3G). Within the LGM, a complete land bridge formed, connecting northwestern coastal Japan (Northern Kyushu) to the southernmost point of South Korea with suitability being high within this area (Fig. 3H). However, suitability was significantly lower throughout central China and most of the moun-

Threshold (10-percentile omission rate) predictions revealed slightly different patterns to our continuous predictions, with areas considered suitable being considerably larger across the time periods sampled (Fig. 4). Areas considered suitable remained relatively constant until the Bølling-Allerød (12.9–14.7 ka), overlapping with the threshold prediction of our ‘Anthropocene’ variables; the exception being the Kamchatka refuge becoming more suitable along with eastern India, Myanmar, Laos, northern Vietnam, the land bridge connecting Coastal China, Taiwan, the Korean Peninsula, and the Yellow Sea (Figs 4B–F). Within the Heinrich Stadial (14.7–17.0 ka), areas considered suitable decreased within central China, as well as Kamchatka, suggesting that the Kamchatka refuge possessed high importance within the historic past but not recently (Fig. 4G). Within the LGM, suitable areas encapsulated most of coastal China while Kamchatka was completely unsuitable. Suitability expanded eastward back into
Figure 2. Maxent prediction for *Epiophlebia* projected using our ‘Anthropocene variables’, with hypothesized refugia for *Epiophlebia* adapted from Büsse & Ware (2022). Predictions were derived from the optimal model using the criterion of AUC<sub>test</sub> values being the highest, and 10% omission rate being the lowest. Predictions were transformed using the ‘cloglog’ function on Maxent v3.4.4 (Phillips et al., 2017), in which raw values are converted to a range of 0–1 to approximate a probability of occurrence (A). We also generated a threshold prediction, calculated from the 10-percentile omission rate from our model evaluation (B). Brighter colors (yellow, green, blue) indicate areas of higher suitability (higher probability of occurrence), while darker colors (violet) indicate areas of lower suitability (lower probability of occurrence). For threshold predictions bright colors indicate areas deemed ‘suitable’ while colder colors indicate areas deemed ‘unsuitable’. Orange points are our thinned dataset of *Epiophlebia* occurrences.
Central China (Sino-Tibetan refuge) with pockets of suitable habitat reaching as far west as Pakistan, Afghanistan, Tajikistan, Kyrgyzstan, Kazakhstan, and Uzbekistan (Fig. 4H). Overall, suitability within the LGM reveals a potential continuous route of genetic admixture that is lost within modern-day disjunct populations (Fig. 4H). During the Last Glacial Maximum, and the Last Interglacial (130 ka), suitable habitat for *Epiophlebia* expanded south to Thailand, and southern Vietnam (Fig. 4I).

MESS analyses indicated that most of the study area presented novel paleoclimate conditions throughout our time intervals (Fig. 5). Climatic similarity within our modern-day ‘Anthropocene’ variables were highest within central and eastern coastal China, Taiwan, Korea, Japan, Sakhalin, and the Russian coastal area on the opposite bank, Kamchatka, eastern India, Myanmar, and Laos producing similar results from our suitability predictions (Fig. 5A). However, similar climatic conditions within eastern China (Fujian province) shifted eastward and decreased going backward within our time intervals, resulting in extremely similar climatic conditions within the land bridge connecting southern Japan with coastal Korea, eastern China, and Taiwan during the LGM (Fig. 4H).

**Discussion**

*Environmental preferences of Epiophlebia*

Within our model, it appears that monthly temperature extremes and seasonality are the main variables driving suitability, suggesting that large-scale fluctuations in temperature better explain distributions for *Epiophlebia*. Variables which possessed high importance within our models were recovered with the same importance within ENMs of both Nearctic and Palearctic species (Abbott et al., 2022; Kalkman et al., 2022). The four most important variables exhibited quadratic relationships with suitability, however some response curves were jagged, which can be explained by the disparate localities of *Epiophlebia*, prompting the model to encapsulate a large spectrum of environmental gradients for extrapolation of suitability.

It is interesting that both *E. superstes* and *E. laidlawi*, despite the substantial geographic gap between their ranges, exhibited similar importance in bioclimatic variables. Even more interesting is the contradictory patterns within their responses to such bioclimatic variables. Since genus-level maxent models assume similar niches of all species included, the argument can

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**Figure 3.** Maxent predictions for *Epiophlebia* projected to the modern day using our ‘Anthropocene variables’ (A) and the late-Holocene (Meghalayan: 0.3–4.2 ka) (B), mid-Holocene (Northgrippian: 4.2–8.326 ka) (C), early Holocene (Greenlandian: 8.326–11.7 ka) (D), Younger Dryas Stadial (11.7–12.9 ka) (E), Bolling-Allerød (12.9–14.7 ka) (F), Heinrich Stadial (14.7–17.0 ka) (G) from our PaleoClim variables, and the Last Glacial Maximum (LGM) (21 ka) (H) and Last Interglacial (LIG) (130 ka) (I) generated from the CHELSA database (H) (Brown et al., 2018). Predictions were derived from the optimal model using the criterion of AUC test values being the highest, and 10% omission rate being the lowest. Predictions were transformed using the ‘cloglog’ function in Maxent v3.4.4 (Phillips et al., 2017), in which raw values are converted to a range of 0–1 to approximate a probability of occurrence. Brighter colors (yellow, green, blue) indicate areas of higher suitability (higher probability of occurrence), while darker colors (violet) indicate areas of lower suitability (lower probability of occurrence).
be made that our combined model portrays a biased picture of the distribution of *Epiophlebia* due to differing species-level response curves. However, we justify combining all three species as the importance of climatic variables did not differ significantly among *E. superstes* and *E. laidlawi*, suggesting such variables may possess evolutionary and phylogenetic importance as both species are sister taxa (Büsse & Ware, 2022).

Optimal suitability for temperature seasonality (standard deviation of the mean monthly temperature) was highest at 5°C suggesting that *Epiophlebia* exhibits a narrow threshold to temperature fluctuation. Since both *E. laidlawi* and *E. superstes* prefer to inhabit the headwaters of mountainous streams (Asahina, 1961a, 1961b; Brockhaus & Hartmann, 2009), we hypothesize that *Epiophlebia* are highly susceptible to large scale temperature perturbations due to the species-level adaptation to cool mesic refugia on the tops of the Himalayas and throughout Japan. Although annual average temperature for *E. sinensis* is about 2–3°C, we accredit the discrepancy of temperature differences among our species due to the large-scale extrapolation of our model to encapsulate the environmental preferences of all three species. Optimal suitability for mean temperature of the warmest month was 20–30°C which supports the average temperature preferences for *E. laidlawi* (Brockhaus & Hartmann, 2009).

In the case of *E. superstes*, the optimal suitability for average temperature of the warmest month is 15.8–27.2°C (Ubukata and Miyazaki, in press), slightly lower than *E. laidlawi*. Utilizing regression data derived from altitude and the Japan Meteorological Society’s Automated Meteorological Data Acquisition System (AMeDAS), Ubukata & Miyazaki (in press) explored the relationship between temperature and distribution of *E. superstes*. In terms of larval data, *E. superstes* can live in torrents with average annual temperatures of 2.4–17.2°C, average temperatures of the coldest and warmest months of -9.5–7.5°C and 15.8–27.2°C, respectively, and average daily minimum temperatures of the coldest month from -17.7–3.0°C. In terms of adult data, the minimum temperatures were estimated for a torrent (780 m) in Hokkaido as follows: average annual temperature is 1.8°C, average temperature of the coldest and warmest months is -11.5 and 15.8°C, respectively (coinciding with the Honshu highlands, the highest altitude records of

![Thresholded maxent predictions for *Epiophlebia* projected to the modern day using our ‘Anthropocene variables’ (A), and the late-Holocene (Meghalayan: 0.3–4.2 ka) (B), mid-Holocene (Northgrippian: 4.2–8.326 ka) (C), early Holocene (Greenlandian: 8.326–11.7 ka) (D), Younger Dryas Stadial (11.7–12.9 ka) (E), Bølling-Allerød (12.9–14.7 ka) (F), Heinrich Stadial (14.7–17.0 ka) (G) from our PaleoClim variables, and the Last Glacial Maximum (LGM) (21 ka) (H) and Last Interglacial (LIG) (130 ka) (I) generated from the CHLSA database (H) (Brown et al., 2018). Predictions were derived from the optimal model using the criterion of AUC test values being the highest, and 10% omission rate being the lowest. Threshold predictions were calculated from the 10-percentile omission rate from our model evaluation. Bright colors (yellow) indicate areas deemed ‘suitable’ while colder colors (violet) indicate areas deemed ‘unsuitable’.](image-url)
larvae within Japan), and average daily minimum temperature of the coldest month is -17.1°C. If adult and larval data are combined, the lowest limits of average annual and coldest month temperature ranges are lowered by the collection record in Hokkaido. Overall, it appears that E. superstes can live within warmer climates (especially in the summer) than its continental conspecifics.

Although precipitation seasonality (bio15), and precipitation of the driest and wettest months (bio13 and 14) possessed minimal importance within model building, we still consider such variables vital in determining distribution of the genus due to the disparate elevational and resultant rainfall regimes present within each species’ range. Previous ENM work on E. laidlawi in the context of future climate change demonstrated that annual precipitation possesses high importance within model building (Dorji et al., 2020; Shah et al., 2012). Localities of E. sinensis within the Changbaishan Mountains experience roughly 2000 mm of snow, short hot summers, with mean annual precipitation of 800 mm (J. K. Li et al., 2012). In the case of E. superstes, the geologic features of Japan are particularly interesting, as variability of rainfall on both sides of Japan have allowed the species to reside nearly at sea level.

Comparing the estimated summer water temperatures of rivers of Japan (Amano & Mochizuki, 2011) to known occurrences of E. superstes, we find that records of the species are absent or rare when temperatures exceed 30°C. However, when summer water temperatures are below 25°C, there are records of E. superstes even at low-altitude mountain streams (1–30 m) near the coast (e.g., the Owase region of Mie Prefecture and the northeastern corner of Toyama Prefecture; Ishida et al. [1959], Futahashi et al. [2004]). A common characteristic of these areas is that annual precipitation exceeds 3000 mm (see Japan Meteorological Agency Website: data.jma.go.jp/stats/etrn/view/atlas/docs2020/png_small/prec/precipitation_13.png), most likely due to humid oceanic winds hitting high elevation mountains (> 1000 m) supplying cool spring water throughout the year.

Figure 5. Multivariate environmental similarity surface (MESS) analysis described in Elith et al. (2010) and calculated using the ‘ENMeval’ package in R using our ‘Anthropocene’ variables (A), late-Holocene (Meghalayan: 0.3–4.2 ka) (B), mid-Holocene (Northgrippian: 4.2–8.326 ka) (C), early Holocene (Greenlandian: 8.326–11.7 ka) (D), Younger Dryas Stadial (11.7–12.9 ka) (E), Bølling-Allerød (12.9–14.7 ka) (F), Heinrich Stadial (14.7–17.0 ka) (G) from our PaleoClim variables, and the Last Glacial Maximum (LGM) (21 ka) (H) and Last Interglacial (LIG) (130 ka) (I) generated from the CHELSA database (H) (Brown et al., 2018). Negative scores (shown in orange) indicate novel climate conditions (i.e., bioclimatic values that fall outside the range of the ‘Anthropocene’ variables for each PaleoClim time interval). Positive scores (shown in purple) indicate climate conditions similar to our ‘Anthropocene’ variables.
The Japanese archipelago consists of the long and narrow island of Honshu and three large islands connected to it across straits. Honshu has a maximum altitude of 3776 m (Mt. Fuji), with the other three islands possessing mountains ranging from 1900–2200 m in altitude. Except for some coastal plains, all Japanese mountains are adjacent to the ocean, some of which possess steep slopes covered in temperate forests (subarctic forest in the north). The elevational gradients of such mountains with abundant rainfall have resulted in cool water (< 22°C), allowing larvae of *E. superstes* to inhabit mountain streams even at lower elevation. One such example is Mt. Inaodake, in the southernmost part of Kyushu's Osumi Peninsula, which is 959 m in elevation, and 2–3 km from the coast where imagines *E. superstes* were collected at 427 m on the northern slope (Matsuhira, 1996); the average annual precipitation at the nearest AMeDAS point (Tashiro) is 2880 mm.

Except for Hokkaido, Japan experiences abundant rainfall during the rainy season from June to July. Since Japan is located between the Pacific Ocean to the southeast and the Sea of Japan to the northwest, differences in rainfall patterns occur along both sides of the archipelago. On the Pacific side of southern Honshu, Shikoku, and Kyushu, summer breezes carry moisture resulting in abundant rainfall. During the winter, northwest monsoons bring heavy snowfall to the mountains closer to the Sea of Japan. Such deep snowfall can be seen even in the early spring, where it melts, supplying rivers and tributaries with cold water. Such complex patterns in precipitation allow constant replenishment of groundwater, permitting *E. superstes* to inhabit both mountain and coastal torrents along the Pacific and the Sea of Japan sides of Japan.

Within mainland Japan (Hokkaido, Honshu, Kyushu, and Shikoku), Chiba Prefecture is the only prefecture where *E. superstes* is absent (highest elevation 408 m). Ubukata & Miyazaki (in press) analyzed the factors contributing to the absence as follows. Although climate warming since the last ice age has caused a rise of about 6°C by the 20th century, temperatures in the lowlands of Chiba Prefecture remain within the habitable mean air temperature range for *E. superstes*. Notwithstanding that this region faces the ocean, there are no mountains exceeding 500 m, so there is not so much (less than 2000 mm) annual precipitation, making it difficult to provide a stable supply of cool (22°C or less) running water to mountain streams. This results in the estimated summer water temperature of this area exceeding 30°C.

**Modern day predictions of Epiophlebia**

Our models support the findings of Büsse & Ware (2022), in which we estimate high suitability for *Epiophlebia* within the Japanese, Nepalese, and Yunnanian refugia, overlapping previous records of *E. superstes* and *E. laidlawi* respectively. Furthermore, Büsse & Ware (2022) hypothesized that due to the presence of other glacial refugia throughout Asia (Lattin, 1967), other *Epiophlebia* habitats may exist within the Sino-Pacific refuge, the Sindhisian refuge (northwest China), the Mongolian refuge, as well as the Manchurian and Kamchatka refugia (see Büsse & Ware [2022]). Our results partially support the hypothesis of Büsse & Ware (2022) in which suitability for *Epiophlebia* was high within the Sino-Pacific refuge off the coast of Taiwan and eastern China, and the Sindhisian refuge bordering eastern India, Myanmar, and Vietnam (Fig. 2). Furthermore, we recorded large swaths of high suitability within central China as well as the Sino-Tibetan refuge, elucidating potentially undersampled habitats for *Epiophlebia* (Fig. 2). Interestingly, the type specimen for the recently demoted *E. diana* was collected within the western Sichuan mountains (Carle, 2012; Needham, 1930). Designation of *E. diana* to *E. laidlawi* supports our findings by pushing the range of *Epiophlebia within* central China, increasing the support of suitable habitat eastward. However more occurrence records are required to verify as the type locality for *E. diana* was too vague to incorporate within or analysis.

**Predictions of Epiophlebia** also partially refute the hypotheses of Büsse & Ware (2022), in which suitability was low within the Sindhisian, Mongolian, Manchurian, and Kamchatka refugia (Fig. 2). Although our continuous predictions suggest intermediate suitability within Northern Russia (Kamchatka refuge) and coastal North Korea (Manchurian refuge), our threshold predictions classify these areas as unsuitable, most likely due to the stringency of our 10% threshold. However, the Manchurian and Kamchatka refugia are climatically similar to the Japanese, Nepalese, and Yunnanian refugia (Fig. 5A), and individuals of *E. sinensis* were recorded at 500 m within North Korea (Fleck et al., 2013; J. K. Li et al., 2012). Furthermore, it is unlikely that Kamchatka refugia harbors *Epiophlebia*, as multiple generations of populations would be required to travel north along the coast of Okhostk, passing through the base of the Kamchatka Peninsula even if the distance between the tip of the Kamchatka Peninsula and its base is somewhat shortened due to the sea level fall in the LGM.

We therefore hypothesize that glacial refugia may still harbor *Epiophlebia* but will require sampling to verify. In the case of the Sindhisian and Mongolian refugia we hypothesize that such areas may not possess the environmental requirements of *Epiophlebia* due to their strict winter and summer temperature ranges (Brockhaus & Hartmann, 2009). Our MESS analysis supports this claim, as the Sindhisian and Mongolian refugia possess very dissimilar climates compared to the Japanese, Nepalese and Yunnanian refugia.

**Paleoclimate predictions of Epiophlebia**

Our paleoclimate predictions revealed the potential for a continuous distribution of *Epiophlebia* between the Himalayas, Japan, and northern China within the LGM, suggesting a continuous route of gene flow among the
disparate current populations. The connection between Japan, mainland China, and the Himalayas within the LGM has been well documented, referred to as the Sino-Japanese floristic region (Büsse et al., 2012; Ikeda, 1998). Our models suggest that a continuous route of dispersal was plausible. Although our continuous predictions restrict suitability from Japan to coastal China down to Taiwan, our threshold predictions reveal large swaths of suitability within central China as well as the Himalayas and parts of western India (Figs 3, 4). Differences of both prediction types can be accredited to the geographic separation of all three species included in our model, as made evident by our MESS analysis display large variation in climate since the LGM. Our continuous predictions emphasize the highest suitable areas within our study extent, while our thresholded maps flatten our predictions based on omission, correcting any uncertainty. As such, we discuss two distinct dispersal scenarios of *Epiophlebia* since the LGM, based on both prediction types (Fig. 6).

According to our continuous predictions, the first hypothesis is that *Epiophlebia* possessed a continuous distribution from Japan to coastal China, but climatic conditions allowed for expansion as the Japanese land bridge closed. Starting in the Heinrich Stadial (17.0–14.7 ka), suitability within the Sino-Pacific, Sino-Tibetan, Yunnanian, and Nepalese refugia increased, potentially acting as a stepping stone for dispersal for *Epiophlebia* westward. Concurrently, the coastal regions connecting Japan to Korea and China were closing, isolating populations of *Epiophlebia* still in Japan. Suitability shifted northward within the Younger Dryas Stadial (11.7–12.9 ka), expanding into Central China (Sichuan and Fujian provinces) throughout the rest of the Holocene, resulting in the modern-day distribution (Fig. 6A). According to our threshold predictions, the second hypothesis is that *Epiophlebia* possessed a continuous distribution throughout the Himalayas to Japan but became isolated due to the closing of the land bridge connecting Japan to the rest of Asia. Throughout the rest of the Holocene, suitability decreased resulting in the occupancy of *Epiophlebia* to the modern-day glacial refugia (Fig. 6B). Our MESS analysis seems to support the former of these two hypotheses in which climate similarity is highest along the coasts of Japan, Korea, and China within the LGM, and decreases over time as climate similarity increases within central China until the modern-day. Regardless of these two hypotheses, it seems that the glacial refugia of Asia acted as retreats for *Epiophlebia* during times of warming, resulting in the current disjunct populations currently seen.

An interesting observation within our predictions is the ephemeral use of the Manchurian and Kamchatka glacial refugia since the LGM. According to our threshold predictions, both glacial refugia possessed suitable habitat from 4.2–17.0 ka, suggesting they acted as potential routes of dispersal in the past, but have since become obsolete. However, distances between adjacent islands in the Kuril Islands which connect northern Japan (Hokkaido) to the Kamchatka peninsula are mostly 20–75 km, distances which are probably beyond the dispersal capability of *Epiophlebia*. Sado Island is separated from Honshu (Japan’s main island) by 32 km, with a maximum water depth of 200 m. Although the island possessed suitable habitat for *Epiophlebia* within our threshold predictions, the species is absent from the island. Assuming a sea level drop of 140 m within the LGM, the distance between Sado and Honshu was greater than 10 km (Kawauchi, 2024) providing indirect evidence of the dispersal capability of *Epiophlebia*. Furthermore, the remote island of Oki does possess *Epiophlebia*, and was connected to the mainland within the LGM. Finally, Hokkaido and Honshu were hypothesized to be connected by a land bridge during the LGM (Yashima and Miyachi, 1990). As such, a land bridge within the LGM would still not connect Hokkaido to the Kuril Islands except for Kunashir and Iturup, preventing *Epiophlebia* from dispersing further north.

Increased sampling and population genetics of *Epiophlebia* would illuminate if both refugia were effective habitats throughout the Holocene. Paleo-ecological niche modeling and genetic studies in the context of Asian and Eurasian refugia are rare except in plants and mammals (Anijalg et al., 2018; J. Li et al., 2016; Tang et al., 2018; Wang et al., 2016), the few studies pertaining to insects are in the context of pests (Song et al., 2018).

**Conclusion**

ENMs of odonates are still gaining popularity within the scientific literature, which have stressed the importance of their conservation in the context of fire ecology, climate change, and anthropogenic disruption (Abbott et al., 2022; Bush et al., 2014; Cancellario et al., 2022; Collins, 2014; Collins & McIntyre, 2015; Dorji et al., 2020; Goodman et al., 2022; Kalkman et al., 2022; Nidup et al., 2020). Modeling of future odonate distributions has noted habitat loss or extinction within specialized endemic riverine species, and northward range shifting in high dispersal species (Boys & Adam, 2021; Flenner et al., 2010; Hassall & Thompson, 2008; Jaeschke et al., 2013; Kietzka et al., 2018; Simaika & Samways, 2015). A recent ENM study of *E. laidlawi* by Nidup et al. (2020) noted uphill shifting of the species by 500 m by the year 2080, due to temperature increases at the lower elevational range of the species at 2000 m. However, permanent snow cover within the Himalayas occurs at 4000 m, prohibiting further northern expansion of the species to higher elevations. Furthermore, Nidup et al. (2020) estimate 83–90% habitat loss of *E. laidlawi* by 2080, stressing the importance of conservation for not only the habitats of *E. laidlawi*, but utilizing ENMs to highlight areas of high conservation value within data poor countries (Dorji et al., 2020). *Epiophlebia* is a specialized and endemic group, highly susceptible to environmental perturbation due to their narrow habitat ranges, small population sizes, and partivoltinism. Our predictions here are hypotheses that can be evaluated,
refined, and tested. One cannot know whether a given species has responded to climatic changes in the same fashion across its history, which means that our hypotheses are based on current population data and expanding this dataset may result in modifications to our models. Future studies could utilize our models as a baseline for predictions of future habitat suitability of the genus in the face of habitat destruction.

In our study we present the first present-day distribution model of the relict dragonfly genus *Epiophlebia*.

Figure 6. Hypothesized distributions of *Epiophlebia* based on our continuous (‘cloglog’) (A) and thresholded (10% omission) (B) predictions since the LGM. Colors represent distributional range within the LGM (hatched yellow), Heinrich Stadial—Younger Dryas (17.0–11.7 ka) (hatched blue), and the Younger Dryas to modern day (11.7–0) (hatched violet). Arrows represent dispersal events to each new range within each time period, while rectangles represent vicariance events. Illustration of *Epiophlebia laidlawi* from Gyeltshen et al. (2017).
Our results indicate several unsampled areas (central China and Sino-Tibetan refuge) which have the potential to harbor new populations of the genus. Furthermore, our results support the hypothesis of Büsse et al. (2012) in which a continuous route of gene flow among disparate populations of *Epiophlebia* was possible during the LGM. Previous research postulated that *Epiophlebia* originated in the Jurassic within Pangaea (Brockhaus & Hartmann, 2009). Although this might be true for fossil ancestors of *Epiophlebia*, molecular data from Büsse et al. (2012) and Büss & Ware (2022) refute this claim, providing an alternative that extant species originated much more recently. Our data provide an additional layer of support for Büsse et al. (2012) in which genetic admixture of *Epiophlebia* was possible during the last ice age.

**Competing interests**

The authors declare no competing interests.

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**Author contributions**

Conceptualization: AMG, CDB, SB, HU, JW; Methodology: AMG, MEB; Software: AMG; Validation: AMG, CDB, SB, HU, TM, MEB, JW; Formal analysis: AMG; Investigation: AMG; Resources: JW; Data Curation: AMG; Writing—Original Draft: AMG; Writing—Review and Editing: AMG, CDB, SB, HU, TM, MEB JW; Visualization: AMG; Supervision: CDB, SB, HU, TM, MEB, JW; Project Administration: JW.

**References**


**Supplementary Material**

Supplemental Figure S1. Permutation analysis and response curves of the most important variables from our optimal maxent model of *Epiophlebia*. Maxent calculates permutation importance by changing the values of each environmental variable at random, then calculating the difference using the AUC from the ‘training data’. The values of each environmental variable are randomly permuted within the training presence and background data, the resultant drop in training AUC is calculated, then normalized to percentages (Phillips et al., 2017). Bar graphs represent normalized percentages of variables possessing non-zero lambda values. Response curves show how each environmental variable individually affects the maxent prediction in terms of increasing or decreasing suitability. Behavior of response curve is dictated by model complexity (feature classes and regularization multipliers). Variables follow Worldclim v2 (Fick & Hijmans, 2017). Bioclimatic variable bio04 (Temperature Seasonality) has been multiplied by 100 to keep significant figures (O’Donnell & Ignizio, 2012).

Supplemental Figure S2. Permutation analysis and response curves of the most important variables from our optimal maxent model of *Epiophlebia laidlawi*. Maxent calculates permutation importance by changing the values of each environmental variable at random, then calculating the difference using the AUC from the ‘training data’. The values of each environmental variable are randomly permuted within the training presence and background data, the resultant drop in training AUC is calculated, then normalized to percentages (Phillips et al., 2017). Bar graphs represent normalized percentages of variables possessing non-zero lambda values. Response curves show how each environmental variable individually affects the maxent prediction in terms of increasing or decreasing suitability. Behavior of response curve is dictated by model complexity (feature classes and regularization multipliers). Variables follow Worldclim v2 (Fick & Hijmans, 2017). Bioclimatic variable bio04 (Temperature Seasonality) has been multiplied by 100 to keep significant figures (O’Donnell & Ignizio, 2012).

Supplemental Figure S3. Permutation analysis and response curves of the most important variables from our optimal maxent model of *Epiophlebia superstes*. Maxent calculates permutation importance by changing the values of each environmental variable at random, then calculating the difference using the AUC from the ‘training data’. The values of each environmental variable are randomly permuted within the training presence and background data, the resultant drop in training AUC is calculated, then normalized to percentages (Phillips et al., 2017). Bar graphs represent normalized percentages of variables possessing non-zero lambda values. Response curves show how each environmental variable individually affects the maxent prediction in terms of increasing or decreasing suitability. Behavior of response curve is dictated by model complexity (feature classes and regularization multipliers). Variables follow Worldclim v2 (Fick & Hijmans, 2017). Bioclimatic variable bio04 (Temperature Seasonality) has been multiplied by 100 to keep significant figures (O’Donnell & Ignizio, 2012).