

14 Terrestrial biodiversity

The following protocol describes the contribution of global terrestrial biodiversity models to ISIMIP2b. Biodiversity is influenced by both climate and land-use change, as well as the biome changes resulting from these drivers. All of these drivers will be considered in terrestrial biodiversity simulations.

Different model types may be used to simulate terrestrial biodiversity, such as correlative species distribution models (SDMs), macroecological species-richness models (MEMs), process-based biodiversity models, and others. There are no restrictions regarding the model type, as long as the methodology has been documented in previous peer-reviewed publications.

In its initial stage, this protocol focuses on correlative SDMs and MEMs; it will be amended with the needs and requirements of other model types as required.

Species distribution models are used to identify the potential climatic niche of a species and so allow to predict a species' probability of occurrence under present and future climatic conditions. Running these models for multiple species, one can aggregate the individual occurrence probabilities to a summed probability of occurrence (a proxy of species richness).

Species distribution data, in combination with the observed climate dataset "EWEMBI" provided by ISIMIP, are used for the initial model construction (i.e., model calibration). Biodiversity projections are then calculated using the ISIMIP2b bias-corrected GCM data.

The effects of biome and land-use changes on biodiversity are currently not considered. In the future, biome and land-use changes may be directly used as predictor variables during model construction.

14.1 Scenarios

Climate scenarios	
picontrol	Pre-industrial climate (year specific for the entire period 1661-2299).
historical	Historical climate.
rcp26	Future climate from RCP2.6.
rcp60	Future climate from RCP6.0.
rcp85	Future climate from RCP8.5.
Human influences scenarios	
nosoc	No human influences considered.

Table 50 ISIMIP2b scenarios for global (and potentially regional) terrestrial biodiversity simulations.

Experiment		Input	Pre-industrial 1660-1860	Historical 1861-2005 ¹	Future 2006-2099 ²	Extended future 2101-2299 ³
I	pre-industrial climate	Climate	picontrol	picontrol	picontrol	picontrol
	no other human influences	Human & LU	nosoc	nosoc	nosoc	nosoc
II	RCP2.6 climate	Climate	Experiment I	historical	rcp26	rcp26
	no other human influences	Human & LU		nosoc	nosoc	nosoc
III	RCP6.0 climate	Climate	Experiment I	Experiment II	rcp60	not simulated
	no other human influences	Human & LU			nosoc	
IV-VII	not simulated					
VIII	RCP8.5 climate	Climate	Experiment I	Experiment II	rcp85	not simulated
	no other human influences	Human & LU			nosoc	

* For now, only correlative species distribution models are considered. Additional scenario combinations will be contributed from other model types in due time.

¹ For the Terrestrial biodiversity sector, “historical” refers to a 30-year period of current conditions (i.e., 1976-2005).

² Within these long-term time periods, biodiversity models will be run for average conditions of selected 30-year periods (2006-2035, 2036-2065, 2066-2095) and the 30-year periods centered around the 1.5°C GCM-specific Global Mean Temperature (GMT) thresholds (1996-2025, 2012-2041, 2018-2047, 2034-2063, 2038-2067, 2042-2071) provided by ISIMIP (<https://www.isimip.org/protocol/temperature-thresholds-and-time-slices/>) are considered.

³ Within this extended-future time period, biodiversity models will be run for average conditions of selected 30-year periods (2086-2115, 2136-2165, 2186-2215, 2236-2265).

14.2 Output data

Table 51 Output variables to be reported by terrestrial biodiversity sector models.

Variable (long name)	Variable name	Unit (NetCDF format)	Resolution	Comments
Amphibian species probability of occurrence	amphibianprob	Probability of occurrence per cell ²	30-year averages of selected time periods ¹ (0.5°x0.5°)	Results from individual SDMs assuming full dispersal ³
Terrestrial bird species probability of occurrence	birdprob			
Terrestrial mammal species probability of occurrence	mammalprob			
Amphibian summed probability of occurrence	amphibiansumprob	Summed probability of occurrence per cell ²	30-year averages of selected time periods ¹ (0.5°x0.5°)	Aggregated results from individual SDMs with different dispersal scenarios ⁴
Terrestrial bird summed probability of occurrence	birdsumprob			
Terrestrial mammal summed probability of occurrence	mammalsumprob			
Summed probability of endemic amphibian species ⁵	endamphibiansumprob			
Summed probability of endemic terrestrial bird species ⁵	endbirdsumprob			
Summed probability of endemic terrestrial mammal species ⁵	endmammalsumprob			
Summed probability of threatened amphibian species ⁶	thramphibiansumprob			
Summed probability of threatened terrestrial bird species ⁶	thrbirdsumprob			
Summed probability of threatened terrestrial mammal species ⁶	thrmammalsumprob			
Amphibian species richness	amphibiansr			
Terrestrial bird species richness	birdsr			

Terrestrial mammal species richness	mammalsr			
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¹ Currently the following 30-year periods (2006-2035, 2036-2065, 2066-2095, 2086-2115, 2136-2165) and the 30-year periods centered around the 1.5°C GCM-specific Global Mean Temperature (GMT) thresholds (1996-2025, 2012-2041, 2018-2047, 2034-2063, 2038-2067, 2042-2071) provided by ISIMIP (<https://www.isimip.org/protocol/temperature-thresholds-and-time-slices/>) are considered.

² For the Maximum Entropy (MaxEnt) model algorithm the output is not probability, but habitat suitability/relative occurrence probability. Values also range between 0 and 1.

³ Probability of occurrence is projected to the currently present and all neighbouring realms of a species and so sort of represents the unlimited dispersal of a species into the future.

⁴ Summed probability of occurrence is calculated for different dispersal scenarios (no dispersal, 0.5*d, 1*d, 2*d, full dispersal). Full dispersal represents the sum of the probability of occurrence output files. No dispersal assumes that species can only be present where they are actually present according to the IUCN and BirdLife range maps. The other three dispersal scenarios consider species-specific dispersal buffers added to the present range, where **d** is the largest diameter of the original range of the species.

⁵ Endemic (range-restricted) species are the smallest ranging 15% of all species.

⁶ Threatened species are all species that are (i) either critically endangered, (ii) endangered or (iii) vulnerable according to their IUCN red list status.

15 References

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