

# Exploring the potential of neural networks for Species Distribution Modeling

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## Introduction to Species Distribution Models (SDMs)

- Relate species occurrence data with environmental variables
- Used to understand and predict the geographic distribution of a species
- Used to support decision-making for conservation and restoration

### Species occurrence data

- 📍 📍 Presence-absence data: systematic field surveys
- 📍 📍 Presence-only data: incidental observations, combined with generated negative samples

### Environmental variables

- ☁️ 🏠 Climate, soil and topography
- 🌿 📱 Vegetation indices, land use, land cover
- 📡 📡 Remote sensing imagery

### Models

- 📊 Traditional statistical methods
- 💻 ML methods
- 🧠 Neural networks, deep learning

## Our work

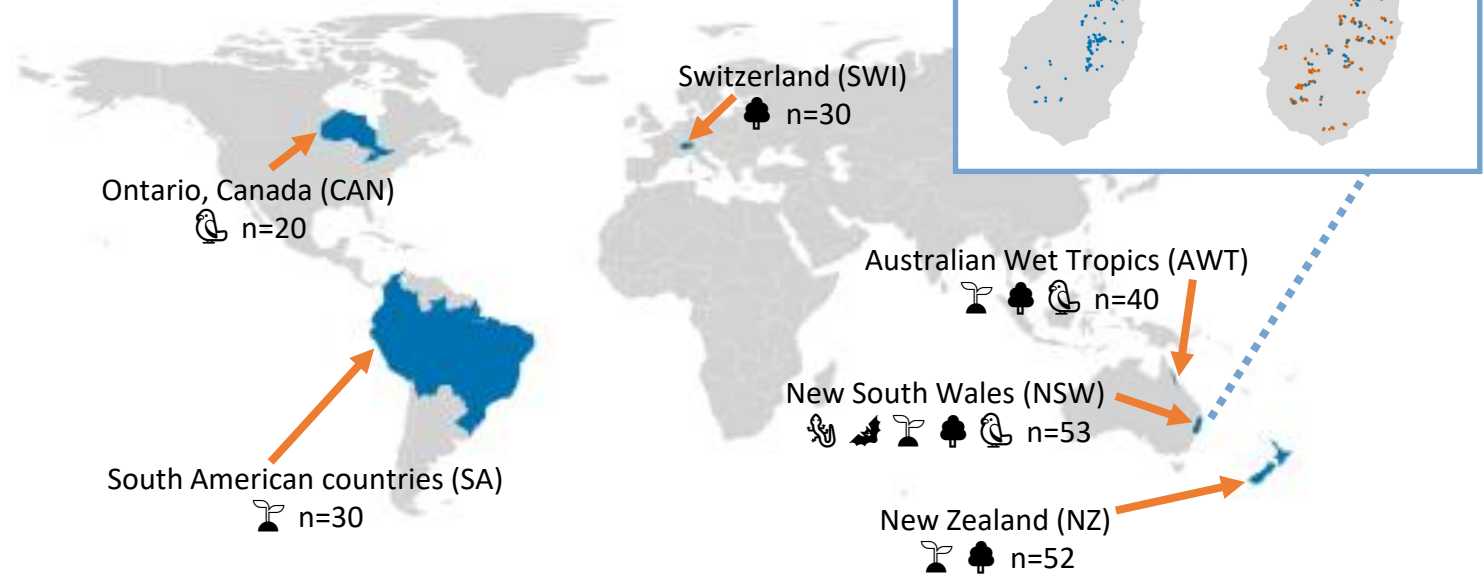
How do neural networks perform compared to well-established methods for SDMs?

**Dataset:** benchmark dataset [1]: tabular data for 225 species from 6 regions. Train on presence-only, test on presence-absence.

**Model:** multi-layer perceptrons (MLPs) with:

- binary classification for single-species models
- multi-label classification for multi-species models

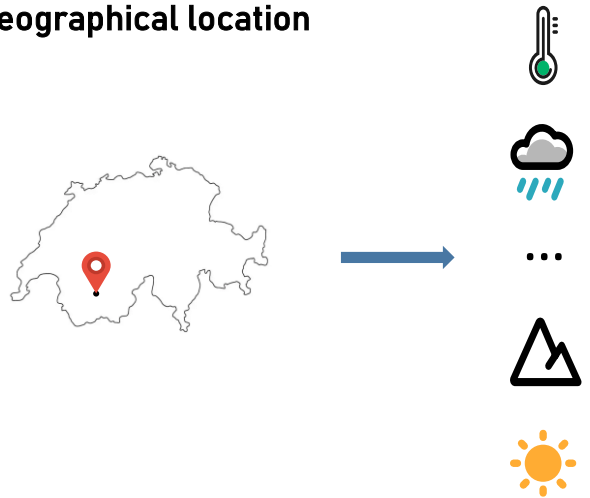
**Loss function:** weighted binary cross-entropy



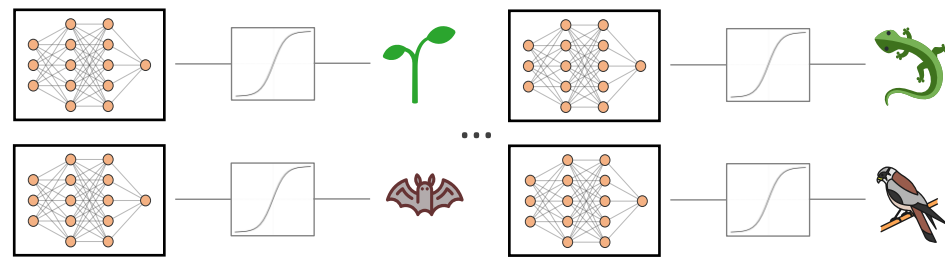
## References

- [1] Elith, Jane, et al. "Presence-only and presence-absence data for comparing species distribution modeling methods." *Biodiversity informatics* 15.2 (2020): 69-80.
- [2] Valavi, Roozbeh, et al. "Predictive performance of presence-only species distribution models: a benchmark study with reproducible code." *Ecological Monographs* 92.1 (2022): e01486.

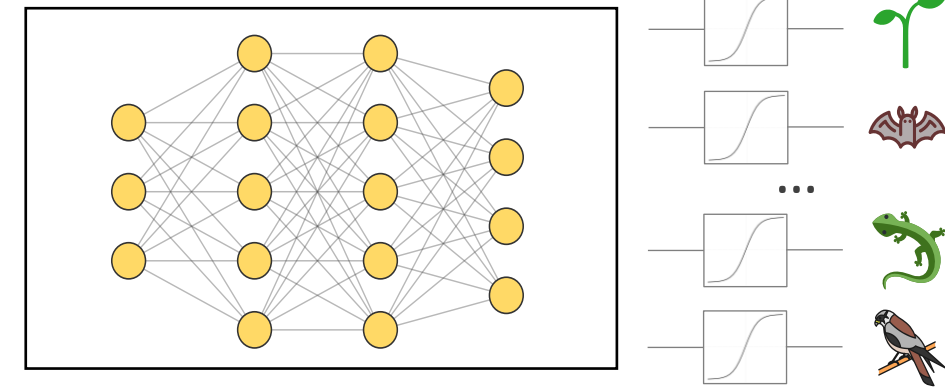
## Environmental conditions in a geographical location



## Single-species model: binary classification



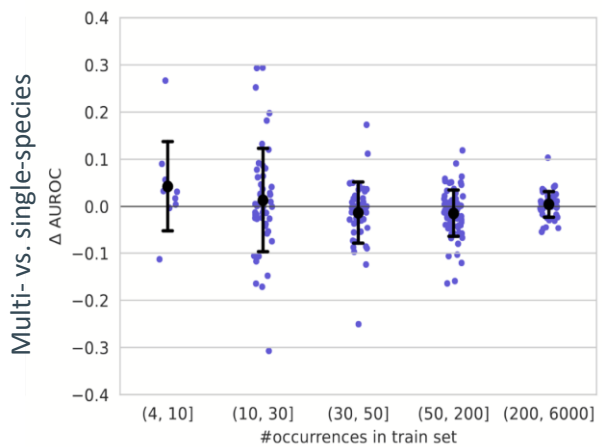
## Multi-species model: multi-label classification



## Results

Mean **AUROC** across species for each region

		AWT	CAN	NSW	NZ	SA	SWI
[2]	MaxEnt	<b>0.686</b>	0.584	0.713	0.738	0.804	0.809
	XGBoost	0.653	0.568	0.706	0.720	0.788	0.815
	Random Forest	0.675	0.572	0.718	0.746	<b>0.813</b>	<b>0.818</b>
	Ensemble	0.683	0.580	<b>0.723</b>	<b>0.749</b>	0.806	0.812
Ours	Single-species MLP	0.666	0.589	0.688	0.715	0.799	0.808
	Multi-species MLP	0.617	<b>0.605</b>	0.708	0.714	0.803	0.815



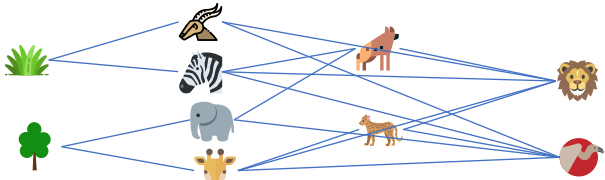
## Future directions

More complex deep learning based SDMs involving and combining:

- **Multi-modal data** providing, for example, geospatial and temporal context



- Biological and ecological information through **knowledge-guided** machine learning



- Transfer learning, model **pre-training**

