

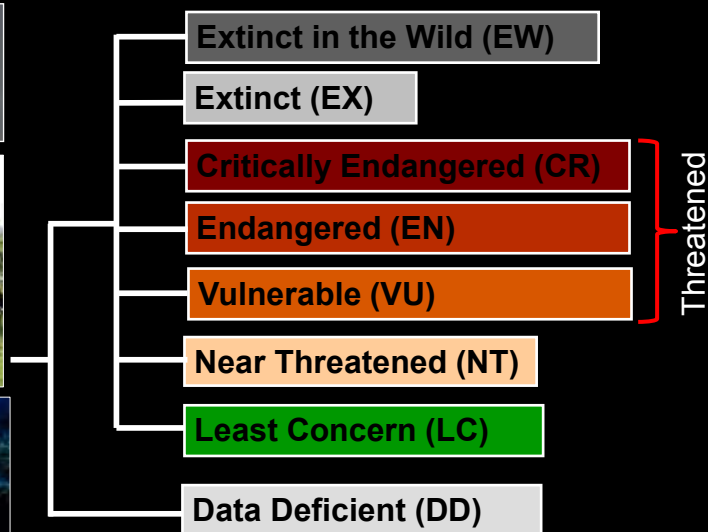
## Predicting the conservation status of IUCN Data Deficient species

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### The IUCN Red List



## Reasons for Data Deficiency

- Type specimens only
- Few records
- Old records (>20 years)
- Unknown provenance
- Taxonomic uncertainty
- Unknown threats
- Unknown population trends



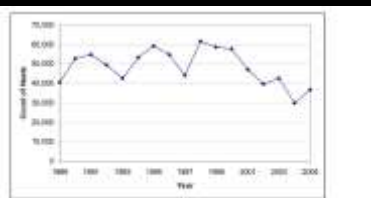

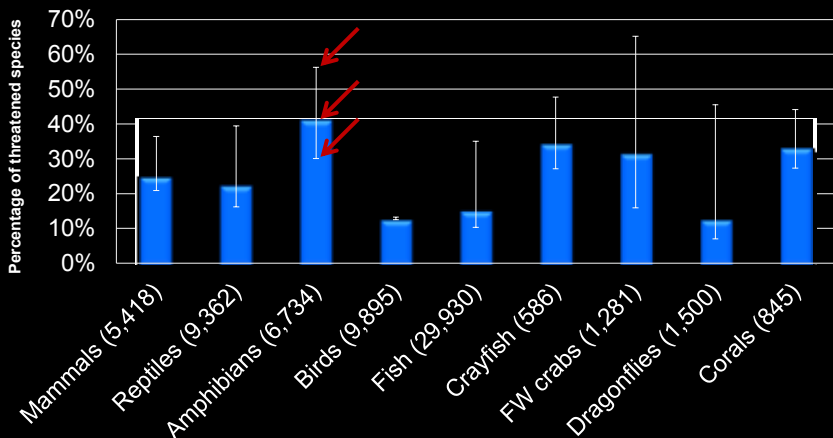




Figure 1. Number of Atlantic Ocean loggerhead sea turtle (*Caretta caretta*) recorded at US beaches (FL, SC, GA) and Panama, Florida (data reaches from 1989 to 2007) (NOAA, in review; PLoS 2007).

## Uncertainty in threat estimates





Taxonomic Group	Total Species	Percentage of Threatened Species (approx.)
Mammals	5,418	25%
Reptiles	9,362	22%
Amphibians	6,734	40%
Birds	9,895	12%
Fish	29,930	15%
Crayfish	586	35%
FW crabs	1,281	30%
Dragonflies	1,500	12%
Corals	845	32%

## Study aims

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- Predict the conservation status of DD species
- Reduce uncertainty in macroecological patterns of threat
- Prioritization for re-assessment/conservation



## Machine Learning Methods

**ZSL**

Classification  
Trees

Random  
Forests

Boosted  
Trees

K-Nearest  
Neighbours

Support  
Vector  
Machines

Neural  
Networks

No Free Lunch Theorem

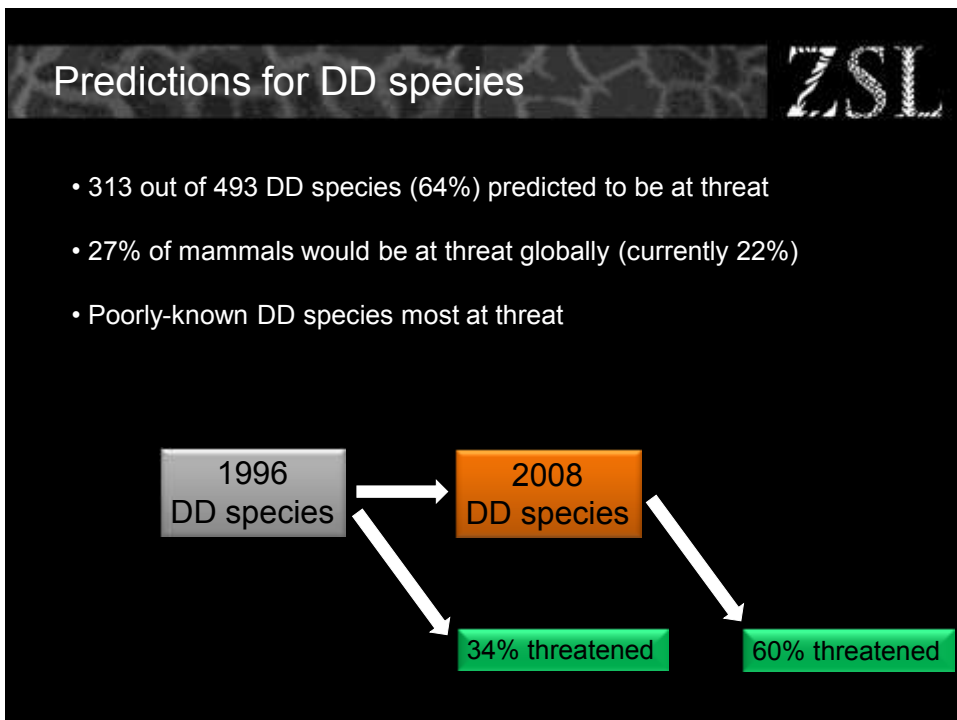
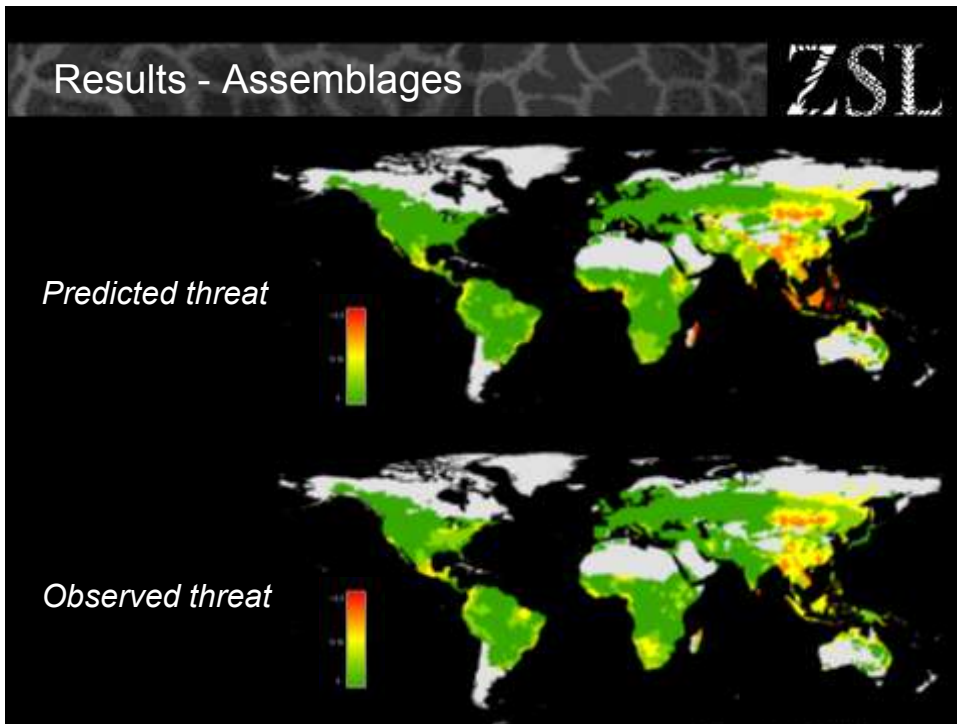
## Materials

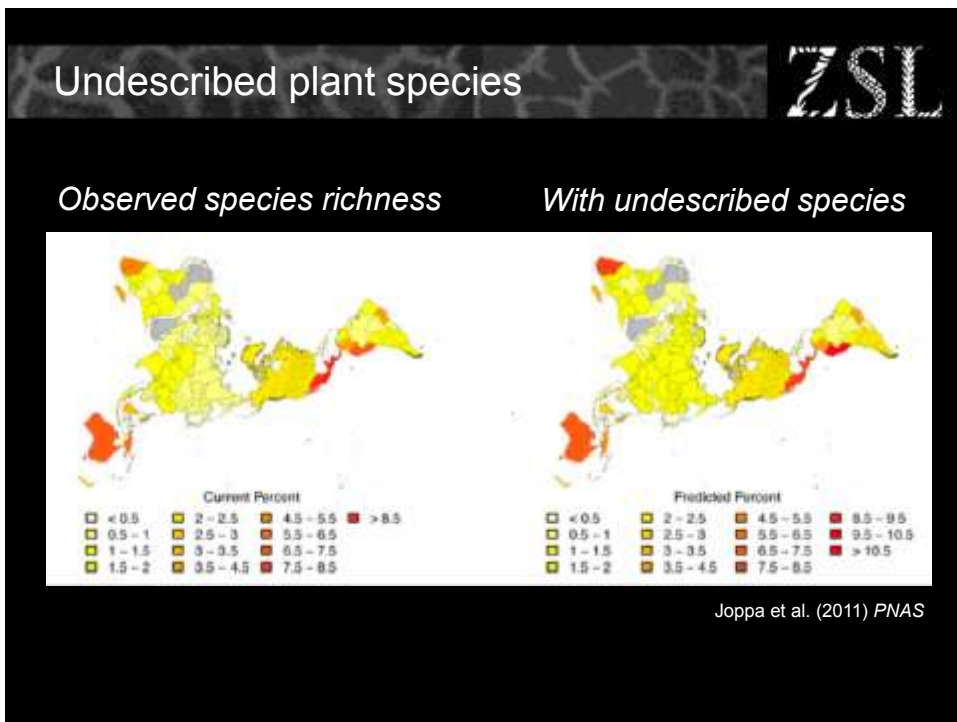
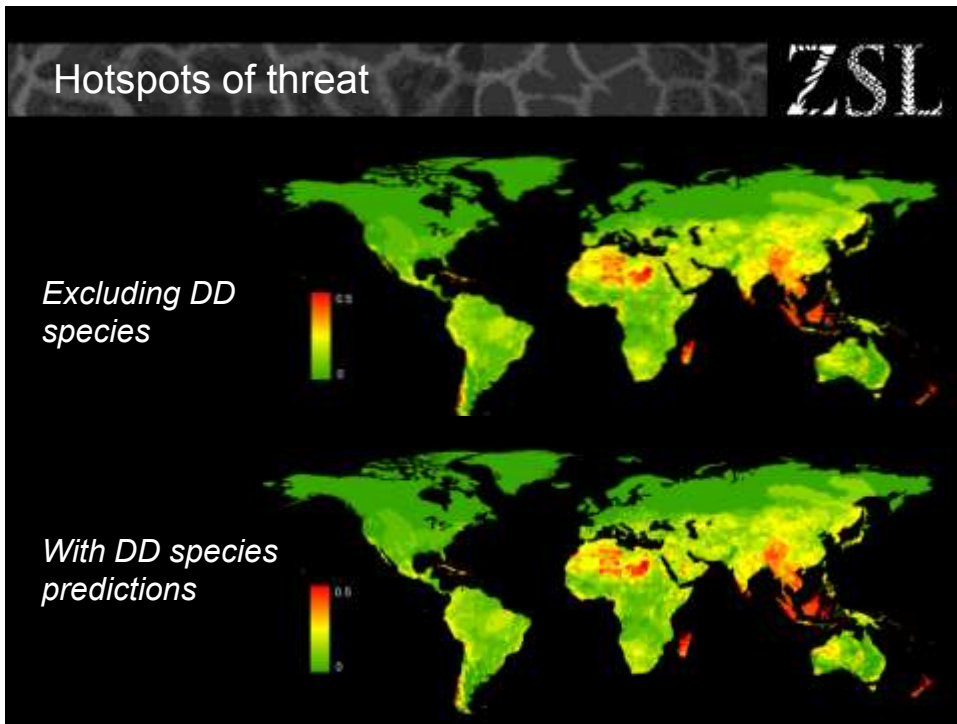
- 4,997 exclusively terrestrial mammal species
- Response variable: IUCN Red List status coded as a binary variable
- Explanatory variables: ~40
  - Life history and ecological (e.g. Body Mass, Litter Size...)
  - Environmental (e.g. PET, Temperature...)
  - Human Impact (e.g. HPD...)
  - Taxonomy
- 5 datasets: Global, Primates, Carnivores, Bats, Rodents

## Results - Species

Dataset	Best Model Type	No. of predictors	No. of non-DD species	Proportion threatened sp.	Accuracy	Specificity	Sensitivity
Global	Random Forest	35	3,967	22.1%	0.898	0.887	0.935
Rodents	Random Forest	29	1,666	17%	0.918	0.933	0.843
Primates	Support Vector Machine	32	304	56.7%	0.803	0.727	0.861
Carnivores	Neural Network	36	188	23.2%	0.913	0.917	0.900
Bats	Boosted Trees	36	828	17%	0.854	0.842	0.914







## Conclusions

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- ML methods show high predictive power for threatened species classification.
- Global conservation priorities unchanged, but species imperilment has been underestimated.
- Poorly-known species are disproportionately at threat.
- Taking into account information on DD species could help conserve the earth's poorly known biodiversity.

## Current work

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- Predicting species extinction risk could offer considerable economic benefits.
- Bayesian Value of Information analysis.





*Thank you for listening!*

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Imperial College  
London

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