TDMA 2023 - Traitement des Données Massives et Apprentissage - applications en géophysique, écologie et SHS

L'ÉCOLOGIE PRÉDICTIVE À L'HEURE DES DONNÉES MASSIVES

Wilfried THUILLER & Sara Si-MOUSSI

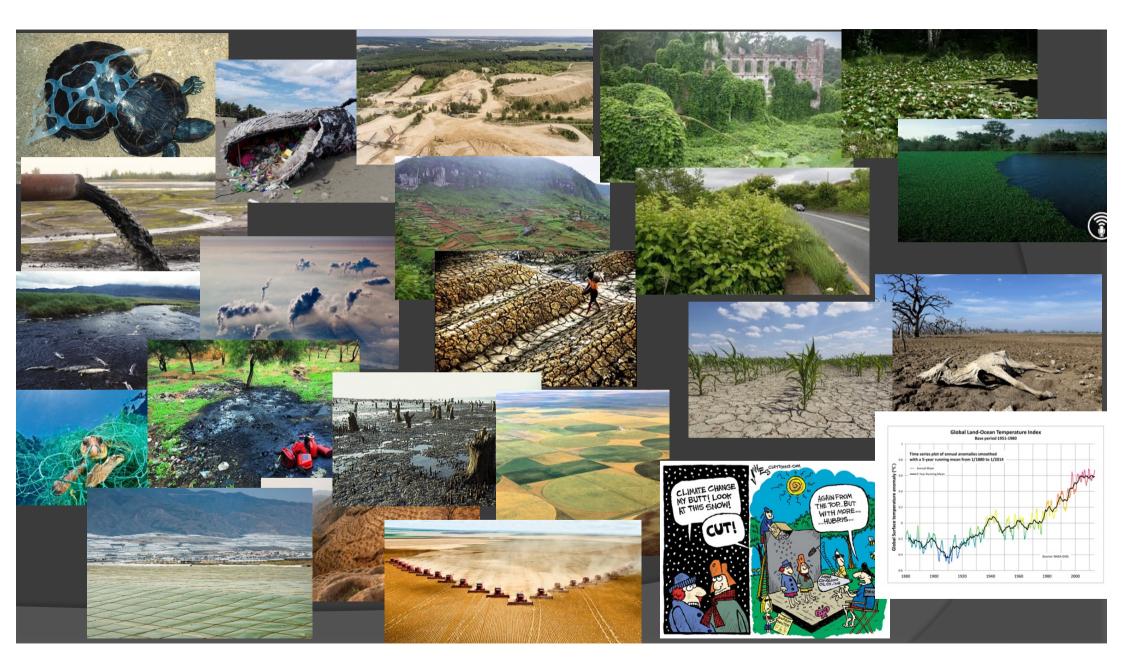
Laboratoire d'Ecologie Alpine Grenoble, France wilfried.thuiller@univ-grenoble-alpes.fr





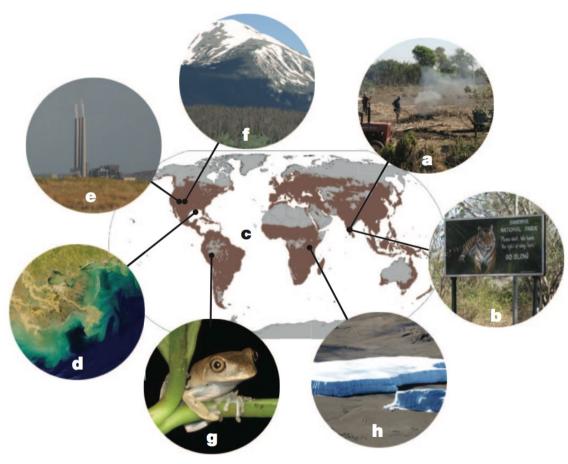
CITES Laboratoire d'Ecologie Alpine

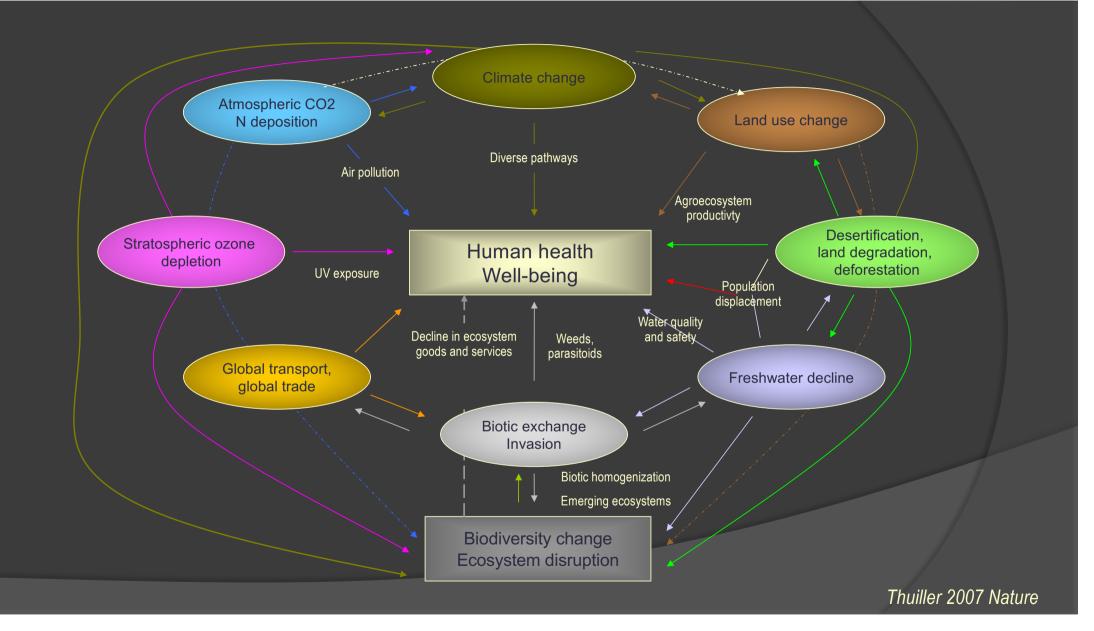
NATURA CONNEC



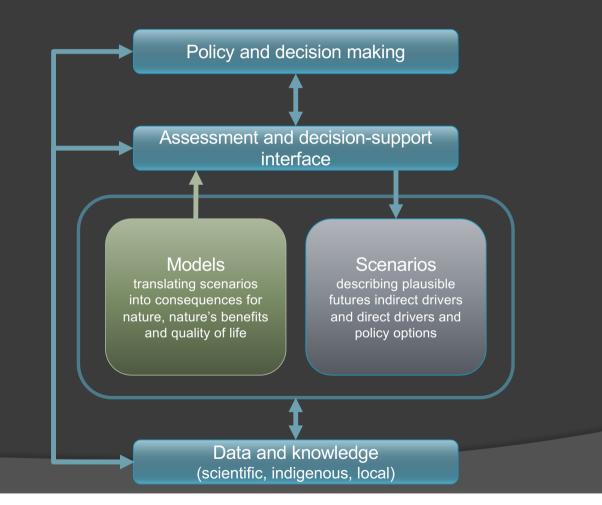
Has the Earth's sixth mass extinction already arrived?

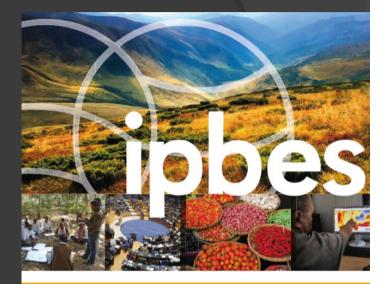
Anthony D. Barnosky^{1,2,3}, Nicholas Matzke¹, Susumu Tomiya^{1,2,3}, Guinevere O. U. Wogan^{1,3}, Brian Swartz^{1,2}, Tiago B. Quental^{1,2}†, Charles Marshall^{1,2}, Jenny L. McGuire^{1,2,3}†, Emily L. Lindsey^{1,2}, Kaitlin C. Maguire^{1,2}, Ben Mersey^{1,4} & Elizabeth A. Ferrer^{1,2}





The need for biodiversity models and scenarios





The methodological assessment report on SCENARIOS AND MODELS OF BIODIVERSITY AND ECOSYSTEM SERVICES



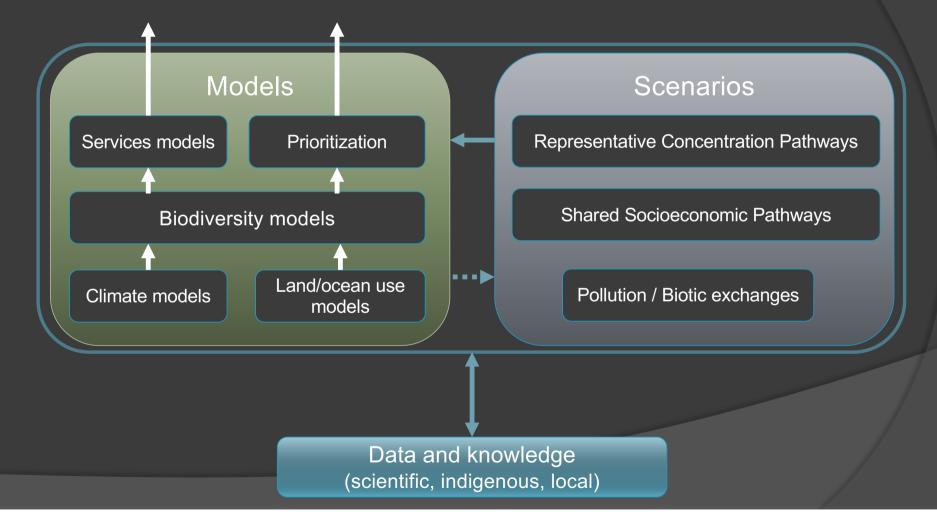
The need for biodiversity models and scenarios

Models translating scenarios into consequences for nature, nature's benefits and quality of life

Scenarios describing plausible futures indirect drivers and direct drivers and policy options

Data and knowledge (scientific, indigenous, local)

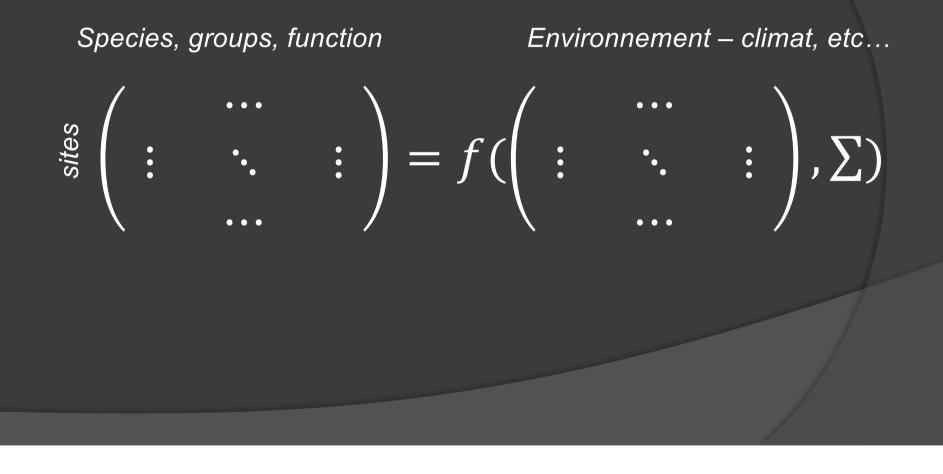
The need for biodiversity models and scenarios



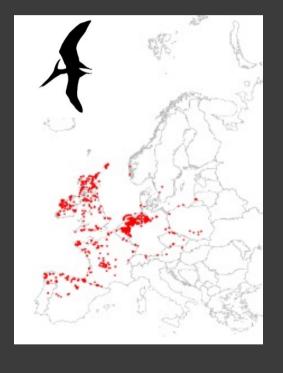
Biodiversity models

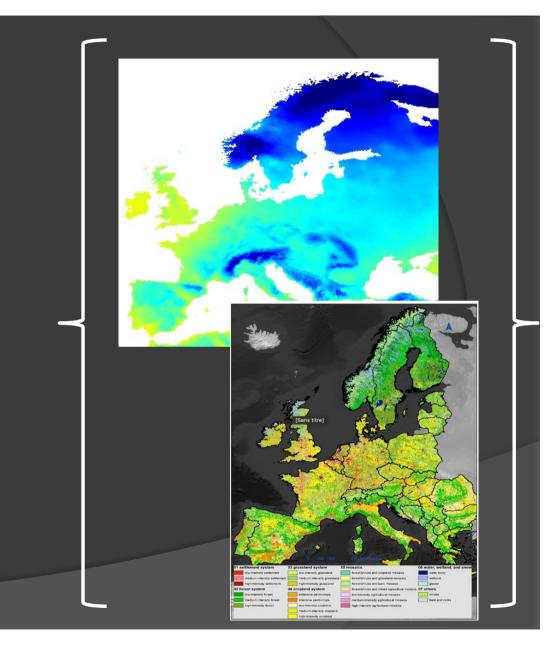
$\overline{Biodiversity} = f(X_i, \Sigma)$

Biodiversity models



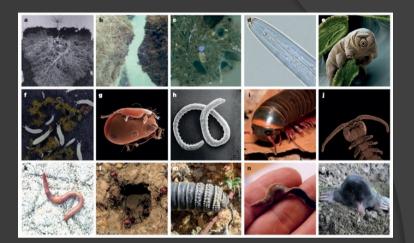
Biodiversity models





Case study 1

Predicting soil diversity in space across the French Alps





Case study 2

Predicting habitat in space across Europe

BIODIVERSITY

Tracking, targeting, and conserving soil biodiversity

A monitoring and indicator system can inform policy

By Carlos A. Guerra, Richard D. Bardgett, Lucrezia Caon, Thomas W. Crowther, Manuel Delgado-Baguerizo, Luca Montanarella, Laetitia M. Navarro, Alberto Orgiazzi, Brajesh K. Singh, Leho Tedersoo, Ronald Vargas-Rojas, Maria J. I. Briones, François Buscot, Erin K. Cameron. Simone Cesarz, Antonis Chatzinotas, Don A. Cowan, Ika Djukic, Johan van den Hoogen, Anika Lehmann, Fernando T. Maestre, César Marín, Thomas Reitz, Matthias C. Rillig, Linnea C. Smith, Franciska T. de Vries, Alexandra Weigelt, Diana H. Wall, Nico Eisenhauer



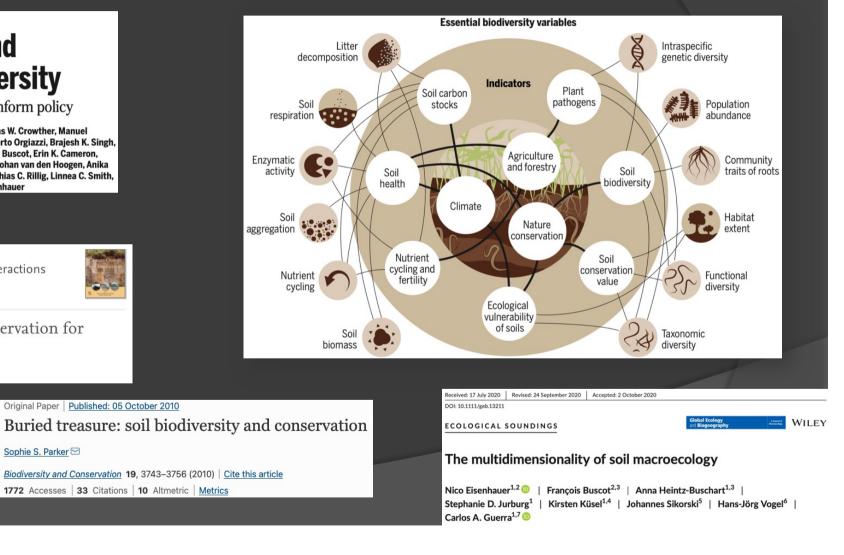
Climate Change and Soil Interactions 2020, Pages 1-19



Sophie S. Parker 🖂

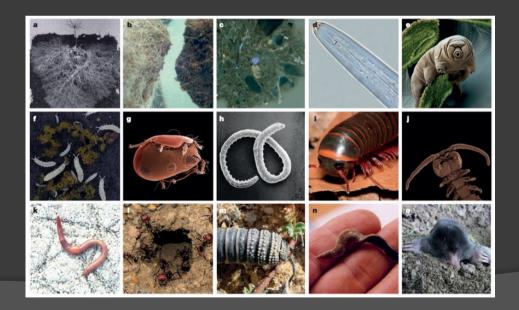
Chapter 1 - Soil biodiversity conservation for mitigating climate change

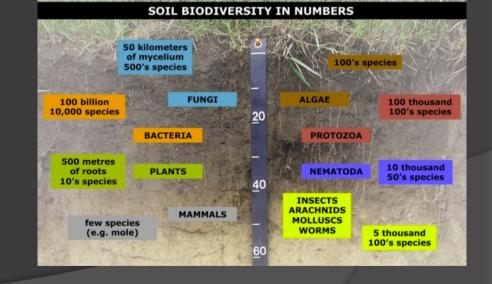
Jan Frouz 1, 2



Major issues in predicting and conserving soil biodiversity:

- Soil biodiversity is a complex system comprising thousand of species in a given sample
- Soil biodiversity is multi-trophic



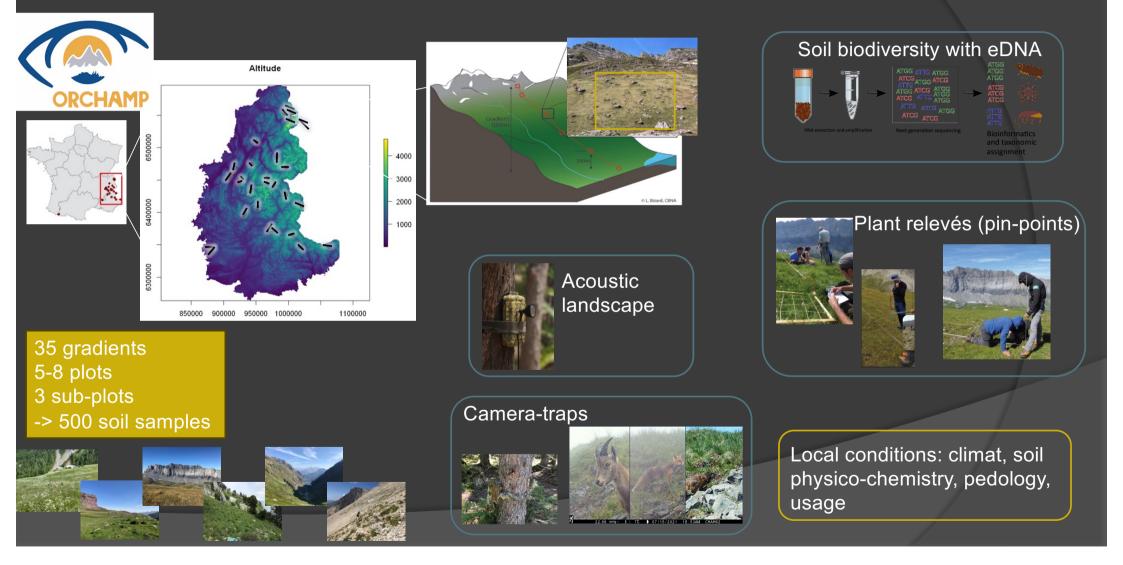


Major issues in predicting and conserving soil biodiversity:

- Soil biodiversity is a complex system comprising thousand of species in a given sample
- Soil biodiversity is multi-trophic
- Strong knowledge gaps:
 - Taxonomy -> Linean shortfall
 - Distribution -> Wallacean shortfall
 - Abundance -> Prestonian shortfall
 - Species traits -> Raunkiæran shortfall
 - Species interactions -> Eltonian shortfall

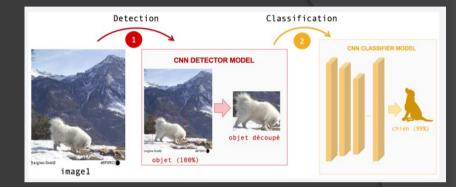
Seven Shortfalls that Beset Large-Scale Knowledge of Biodiversity

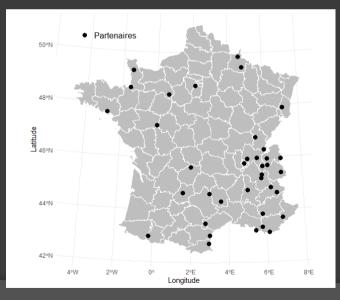
Joaquín Hortal,^{1,2,3,*} Francesco de Bello,^{4,5} José Alexandre F. Diniz-Filho,² Thomas M. Lewinsohn,⁶ Jorge M. Lobo,¹ and Richard J. Ladle^{7,8,*}



Camera-traps



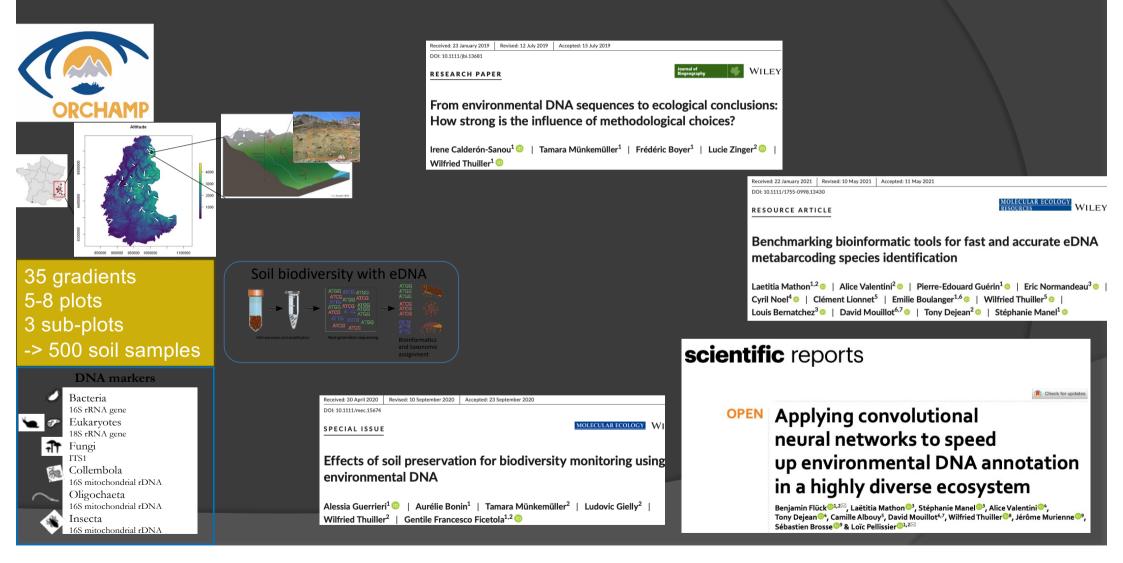


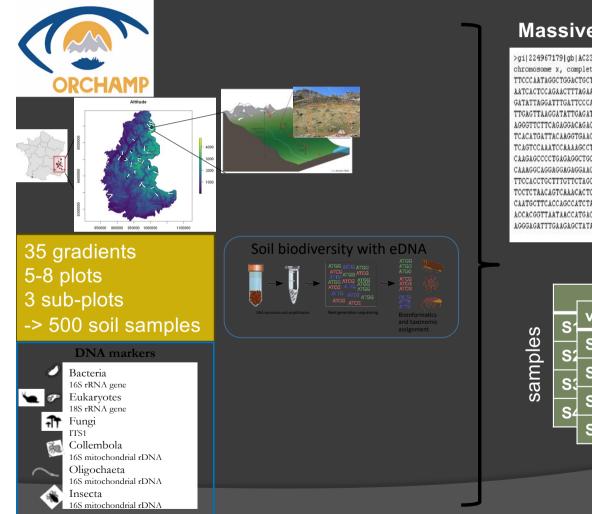


DeepFaune Accueil Données et Modèle Logiciel Partenariat Equipe Contact



https://www.deepfaune.cnrs.fr

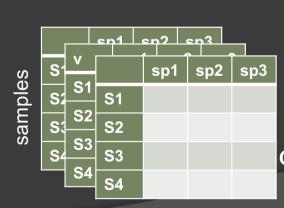




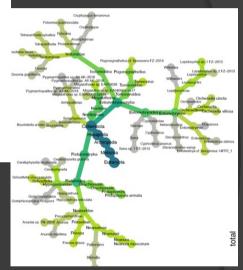
Massive data

>qi|224967179|qb|AC234315.2| Homo sapiens FOSMID clone ABC14-50190700J6 from chromosome x, complete sequence

TTCCCAATAGGCTGGACTGCTTACCACCCCATGTGGCCTCAAAGAGCTCCAGTCACTCCTTTACGAACCC AATCACTCCAGAACTTTAGAACAAAGTTTCTGAGTTACTCCTTGTAATAGGCTAAATAATGGCTCCCAAA GATATTAGGATTTGATTCCCAGAACCTATAAATATTACCTTATTTGGAAAACGGTTCTTAGCAGATGTGA TTGAGTTAAGGATATTGAGATGCAGAGATTATTTTAGATTATCTAGACTATCTGGGTGGATGTATTGGTC AGGGTTCTTCAGAGGACAGAGCCAATAGGATATATGTATATAAAAAAGGGAGTTAATTAGGGAGAATTGGC TCACATGATTACARGGTGAAGTCCCACGATAGGCCGTCTGCAAACTGGGGAGAGAAGCTAGTTGTGTGGG TCAGTCCAAATCCAAAAGCCTCAAAACTGGAGAAGCTGACAGTACAAGCCCTAGTCTGAGGCCAAAGGTC CAAGAGCCCCTGAGAGGCTGCTGCTGCAAGTTCCAGAGTCCAAAGGTTAACAAACCTGAAGTCTGGTGTG CAAAGGCAGGAGGAGGAGGAAGCAGACAGGAAGGAAAGCAAACCAGACTCAGCAAGAAAGCTGCTGTTC TTCCACCTGCTTTGTTCTAGCCACGCTGGCAGTCAATTGCATGGTGCCCATCCACACGAGGGTGGATCT TCCTCTAACAGTCAAACACTGACTCAAATGTCATCTTCTCTGGCAACACCCCTCACAGACACACCCCAGAAA CAATGCTTCACCAGCCATCTATGCAGCCCTCAATCCAGTCAAGGTGACACCTAATGGTTAA



Different taxonomic resolutions



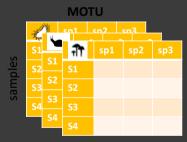
Community matrix per marker

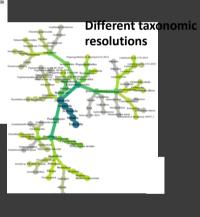
The quest

Massive data

>gi|224967179|gb|AC234315.2| Homo sapiens FOSMID clone ABC14-50190700J6 from chromosome x, complete sequence

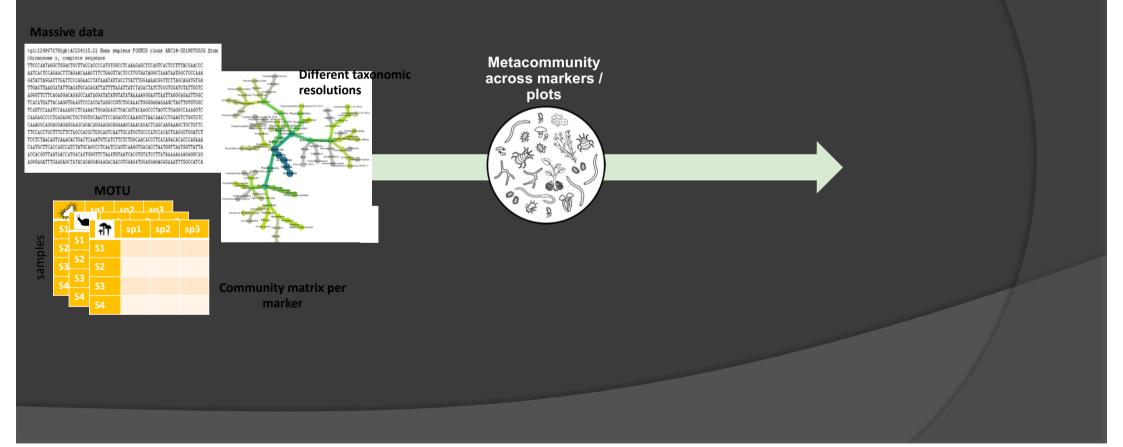
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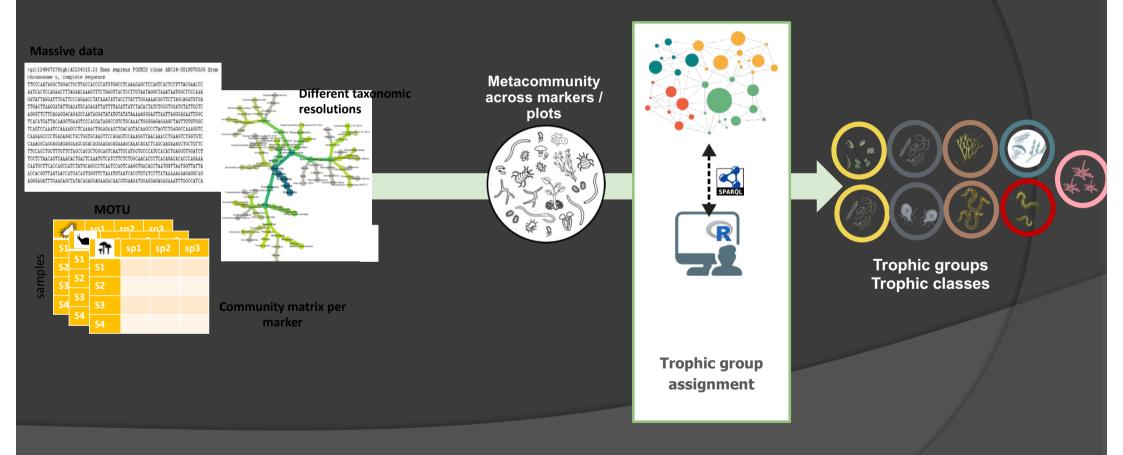


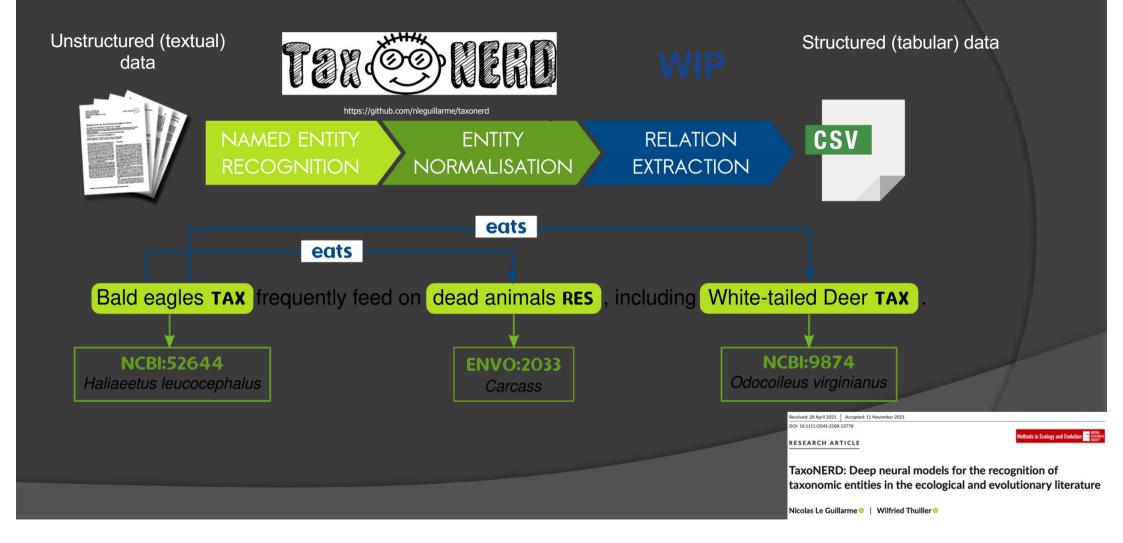
Community matrix per marker

The quest



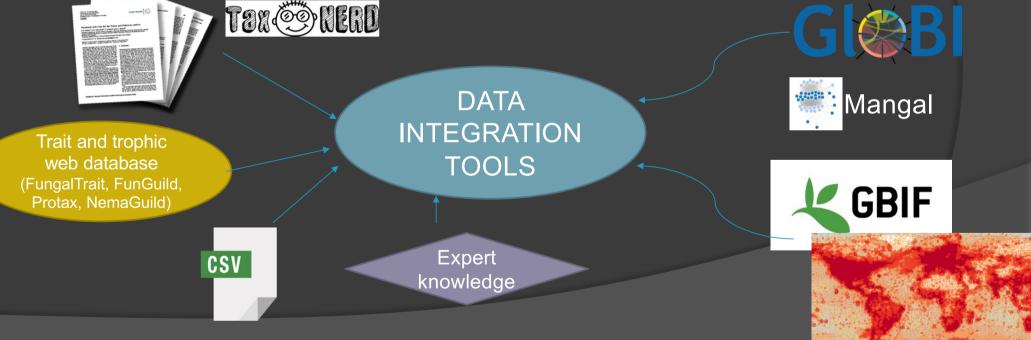
The quest



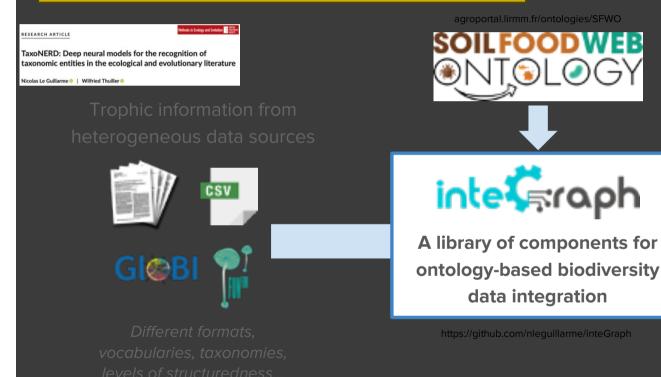


- How to represent and map the knowledge





Trophic knowledge graph creation



GRATIN : a GRAph of Trophic INformation



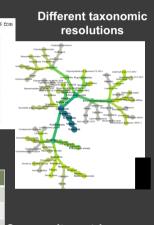


Single format, vocabulary, taxonomy, contains both stated and inferred facts

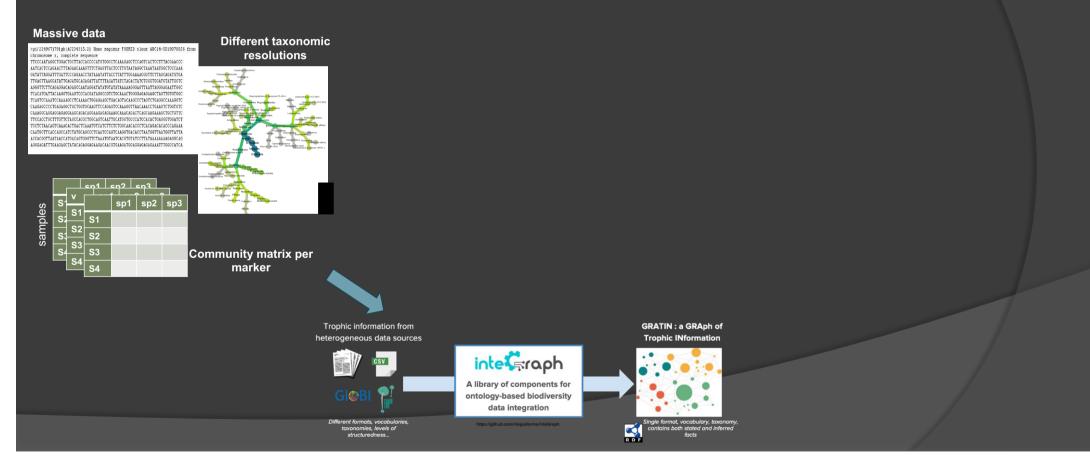
Le Guillarme et al. (2021). Building a Trophic Knowledge Graph to Support Soil Food Web Reconstruction. In S4BioDiv 2021: 3rd International Workshop on Semantics for Biodiversity Le Guillarme et al. (2021) STWO: an Ontology for Soil Food Web Reconstruction. In *S4BioDiv 2021: 3rd International Workshop on Semantics for Biodiversity*

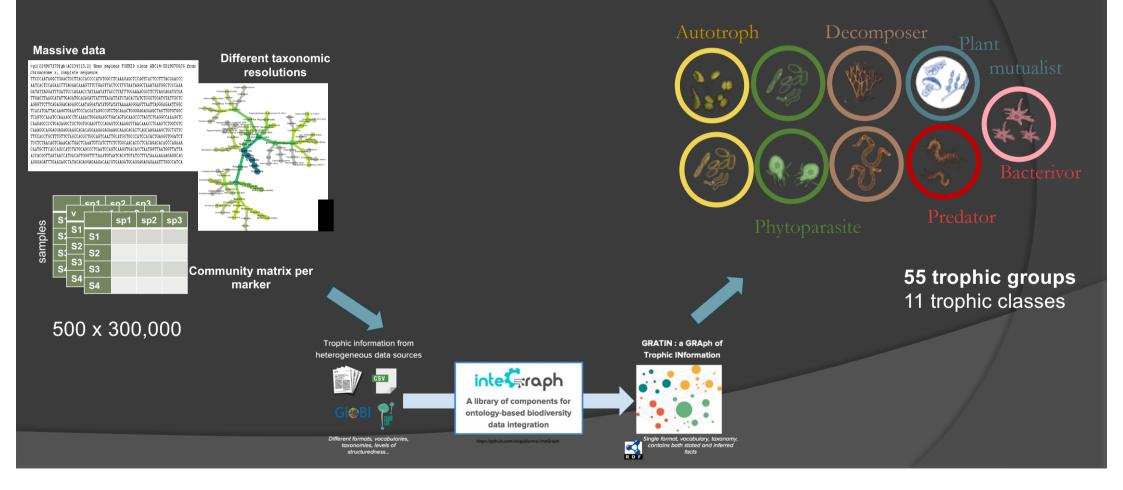
Massive data

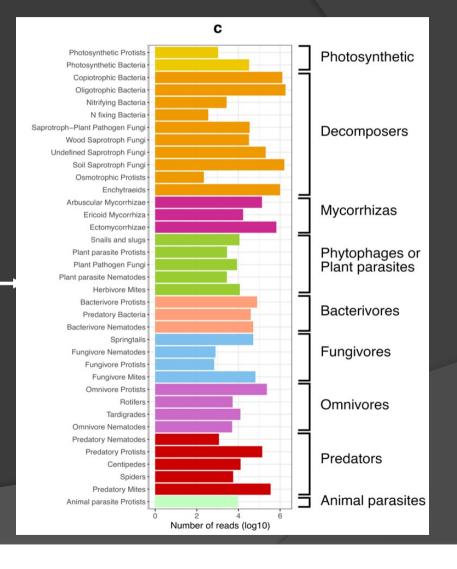
i|224967179|gb|AC234315.2| Homo sapiens FOSMID clone ABC14-50190700J6 fro me x, complete sequence ICACTCCAGARCTITAGARCARRGTITCTGAGTTACTCCTTGTRATAGGCTARATARTGGCTCCCA. TATTAGGATITGATTCCCAGAACCTATAAATATTACCTTATTTGGAAAACGGTTCTTAGCAGATGTGA TTAAGGATATTGAGATGCAGAGATTATTTTAGATTATCTAGACTATCTGGGTGGATGTATTGG ITCTTCAGAGGACAGAGCCARTAGGATATATGTATATAAAAAGGGAGTTAATTAGGGAGAATT CATGATTAC ARGETGA AGTOCCACGAT AGGCCGTCTGCAA ACTGGGGAGAGA AGCTAGTTG CARATCCARRAGCCTCARRACTGGAGAAGCTGRCAGTACAAGCCCTAGTCTGAGGCCARR RECCCCTGRGRGGCTGCTGCTGCRRGTTCCRGRGTCCRRRGGTTRRCARRCCTGRGTCTGGTGT GAGGAGAGGAAGCAGACAGGAAGAGAGAAAGCAAACAGACTCAGCAAGAAAGCTGCTGT CACCTGCTTTGTTCTAGCCACGCTGGCAGTCAATTGCATGGTGCCCATCCACACTGAGGGTGGATCT TOTARCAGTORARCECTGACTCARATOTORTOTOTOTOGOARCACCOCTCACAGACACCCCAGARA TGCTTCACCAGCCATCTATGCAGCCCTCAATCCAGTCAAGGTGACACCCTCAATGGTTAATGGTTATTA cn2 cn2

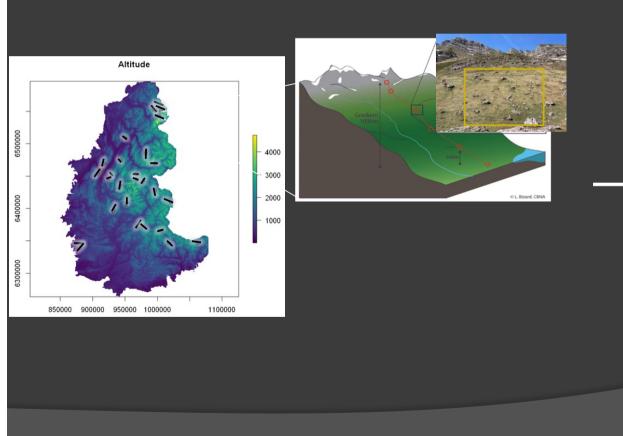


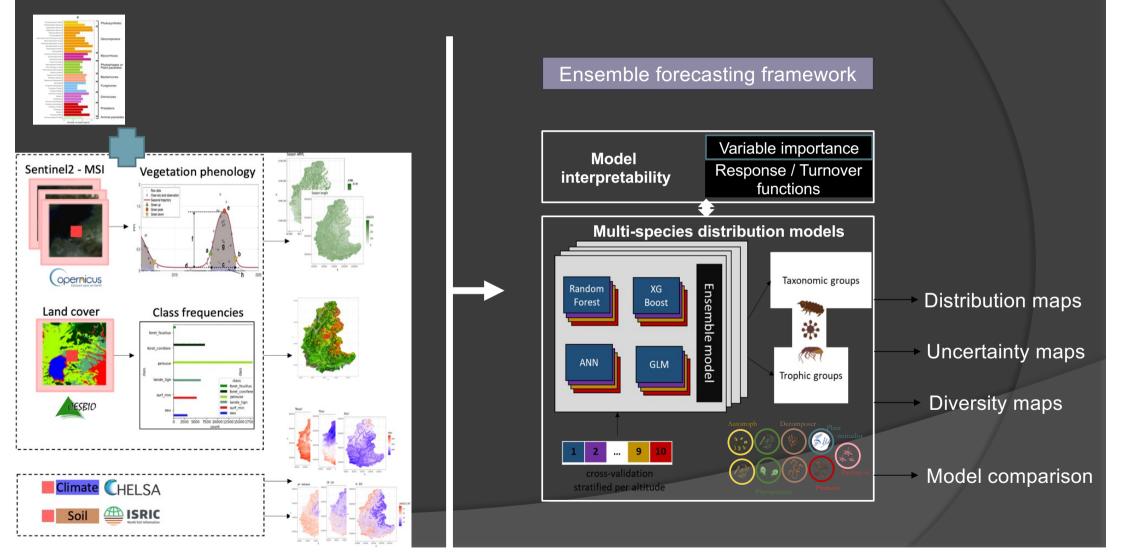
Community matrix per marker

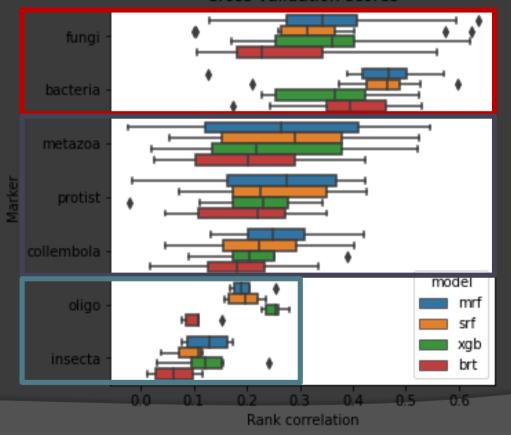










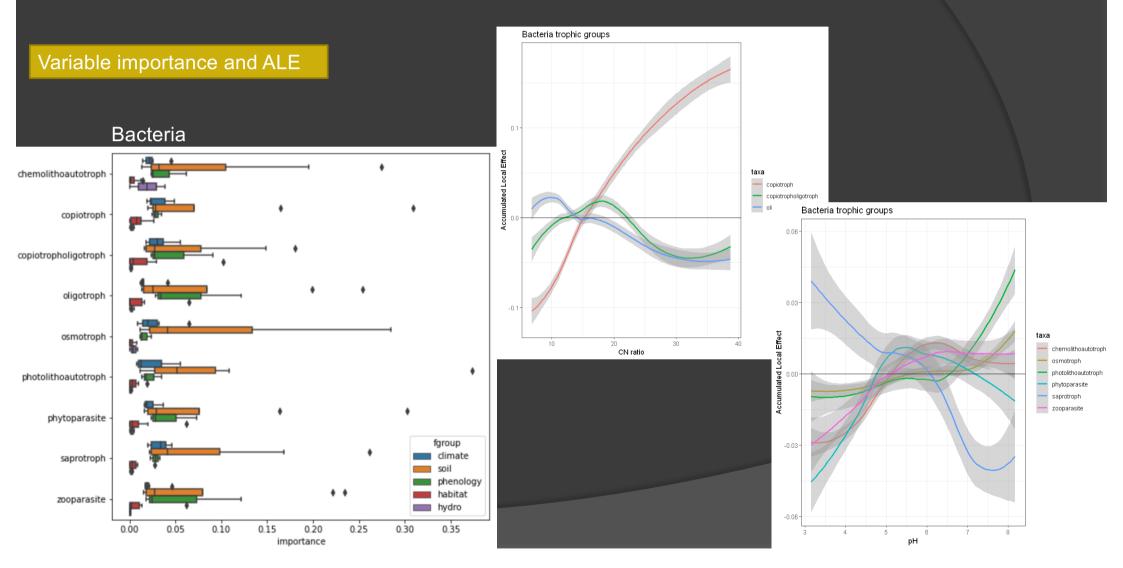


Cross-validation scores

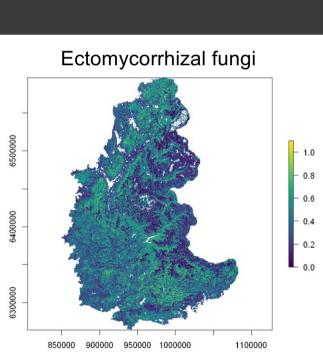
Very good predictions for microbes

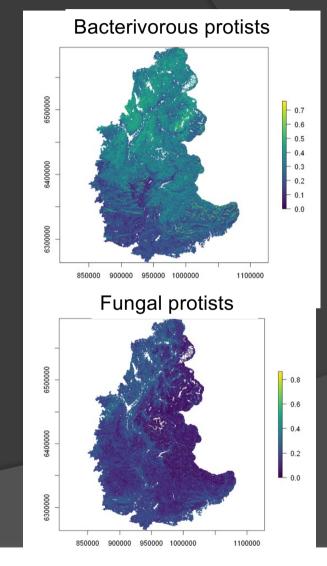
Good predictions for micro-to macro organisms

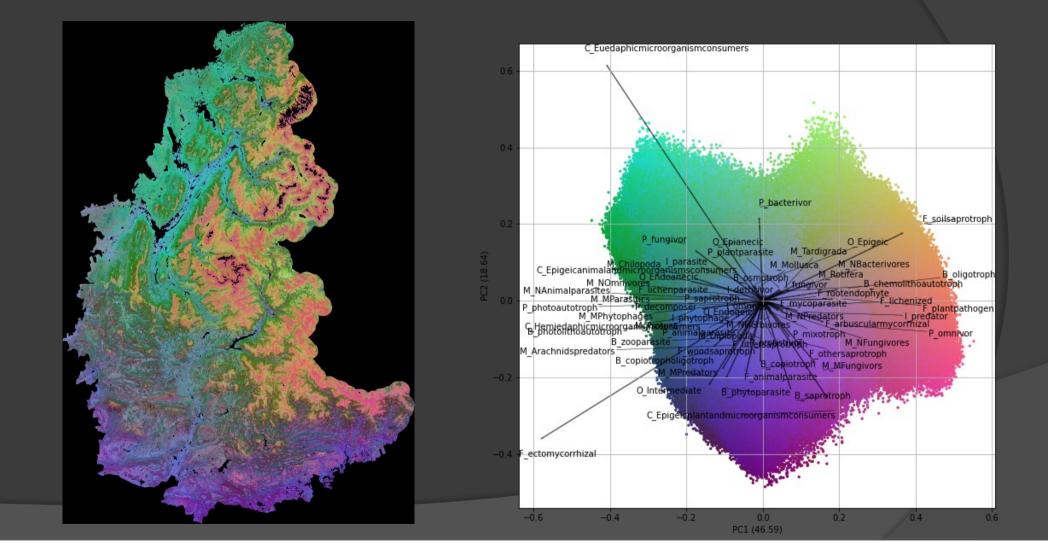
Moderate predictions for macroorganisms



Trophic group distributions Saprotrophic bacteria 6500000 - 0.8 0.6 0.4 6400000 6500000 0.2 0.0 6300000 6400000 850000 900000 950000 1000000 1100000 6300000







Case study 1

Predicting soil diversity in space across the French Alps





Case study 2

Predicting habitat in space across Europe

European & regional habitat modelling



European & regional habitat modelling

Context:

- Strong decline in biodiversity partly due to the degradation of the ecosystems and habitats that support it (27% with unfavorable conservation status).
- EU's biodiversity strategy for 2030 and the new Nature Restoration law aims to protect nature and reverse the degradation of ecosystems (restore > 15%).

-> WANTED! Information about distribution and extent of habitats and ecosystems to inform policy makers.

-> Remote-sensing offers a unique opportunity to monitor the extend and state of habitats.

EUNIS habitat classification system



















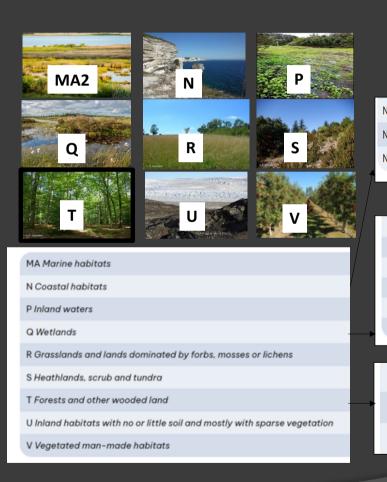
Code 2021	Name 2021
MA	Marine habitats (#11)
N	Coastal habitats (#25)
Р	Freshwater habitats(#10)
Q	Wetlands (#20)
R	Grasslands and lands dominated by forbs, mosses or lichens (#54)
S	Heathland, scrub and tundra (#43)
Т	Forest and other wooded land (#46)
U	Inland habitats with no or little soil and mostly with sparse vegetation(#26)
V	Vegetated man-made habitats (#12)

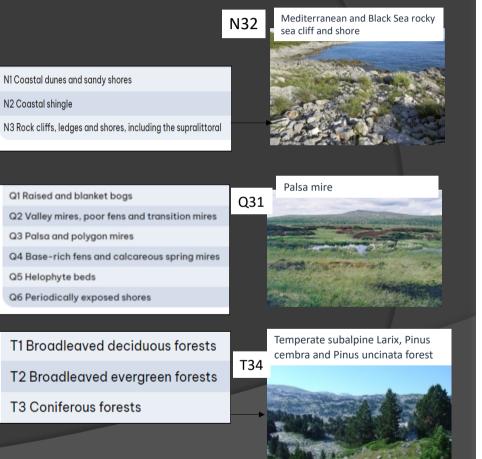
VEGETATION SURVEY 🖻 Open Access 💿 😧

EUNIS Habitat Classification: Expert system, characteristic species combinations and distribution maps of European habitats

Milan Chytrý 🔀 Lubomír Tichý, Stephan M. Hennekens, Ilona Knollová, John A. M. Janssen,







EUNIS level 1

EUNIS level 2

EUNIS level 3

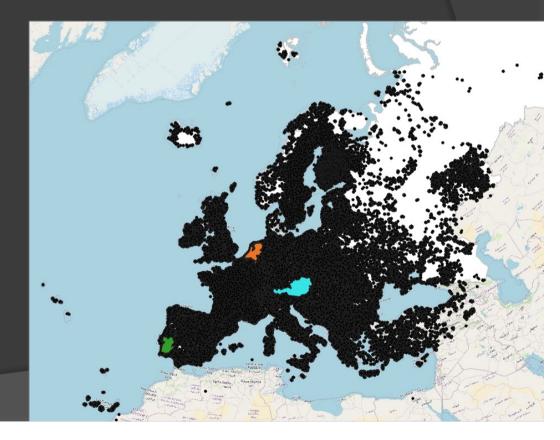
DATA

Training:

Eur. Vegetation Archive (EVA) 2M+ plots, only part used 2000-todate Location uncertainty<100m

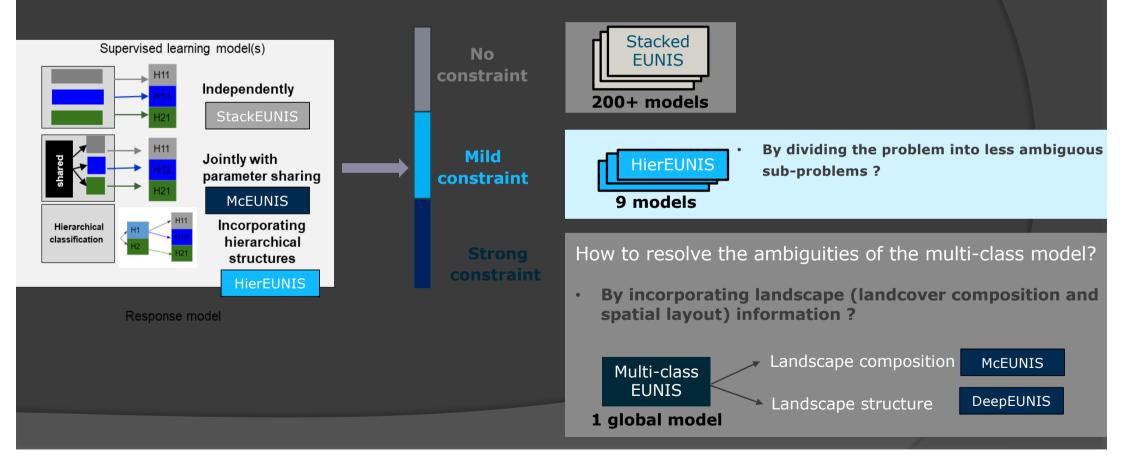
Evaluation with local EUNIS maps:

- Netherlands
- Austria
- South Portugal

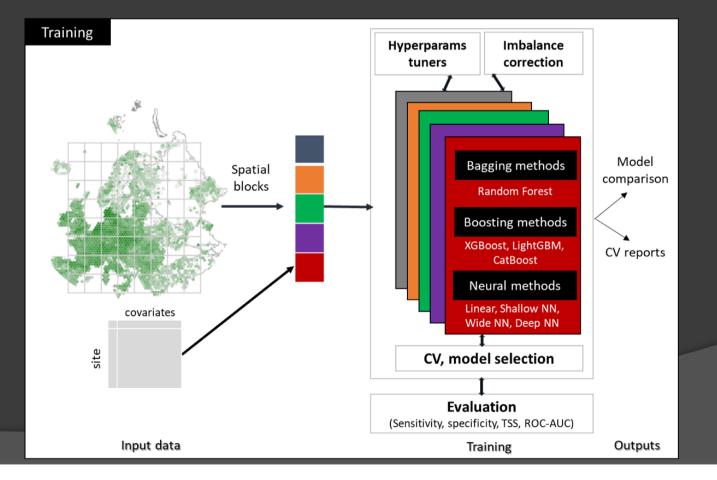


- Large-scale assessment of habitat modelling approaches
- Use of advanced RS products such as HR-VPP
- Evaluation of a broad range of algorithmic approaches and classification strategies
- Production of high resolution maps of EUNIS habitats

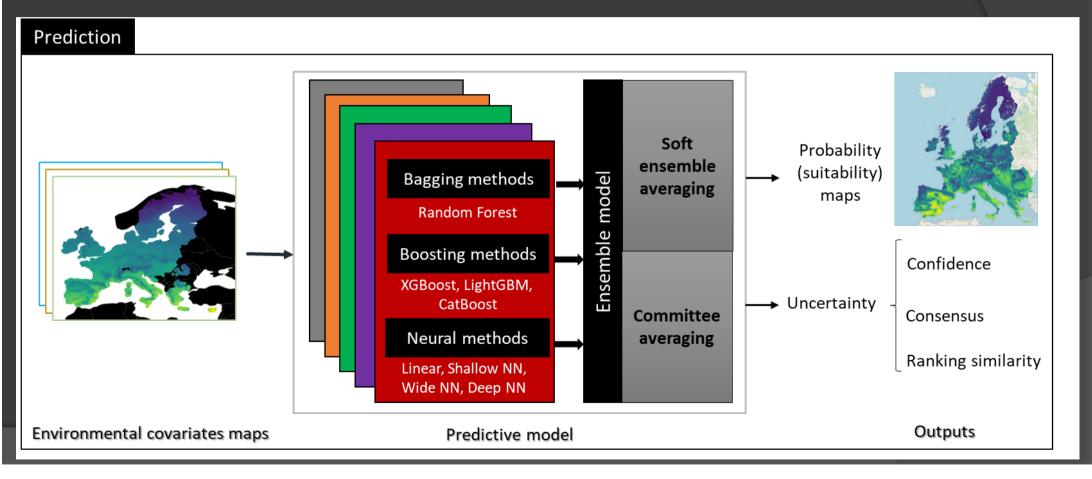
Question: What is the best way to model all EUNIS habitats at level 3 ?



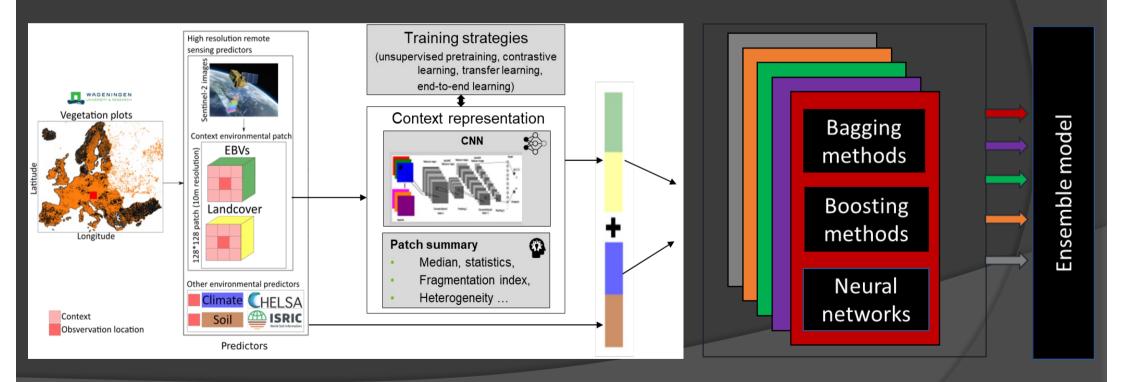
General modelling framework – training phase on point data



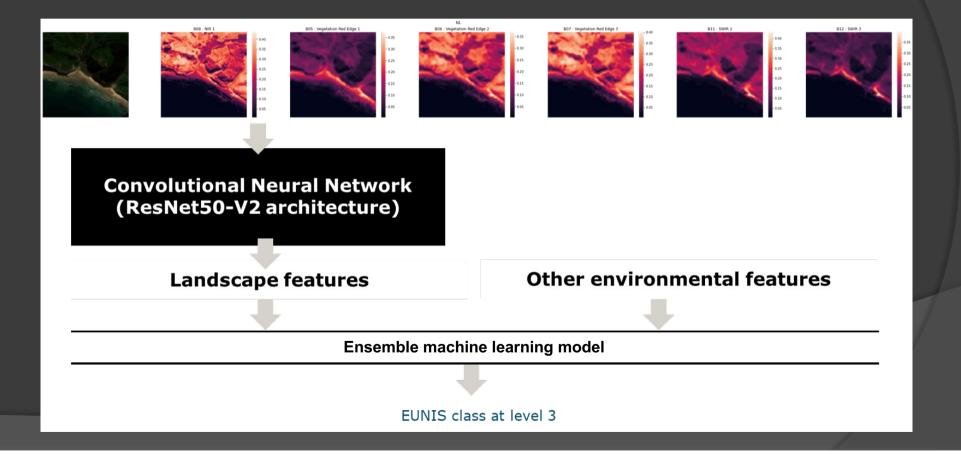
General modelling framework – predicting phase on raster data



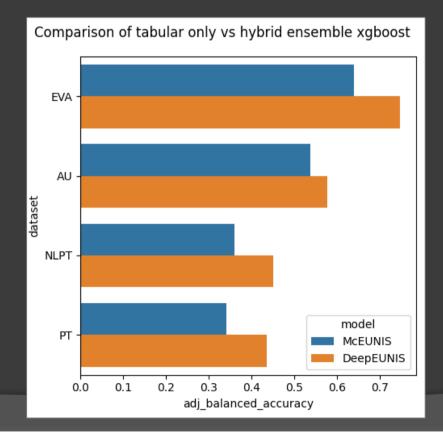
Landscape representation



Learning landscape features using a deep learning model trained on MSI patches



McEUNIS vs DeepEUNIS: does incorporating multi-spectral information and its spatial layout improve habitat discrimination ?



Yes! In general, accounting for habitat structure through MSI information and DL model work better

McEUNIS vs DeepEUNIS: does incorporating multi-spectral information and its spatial layout improve habitat discrimination ?

