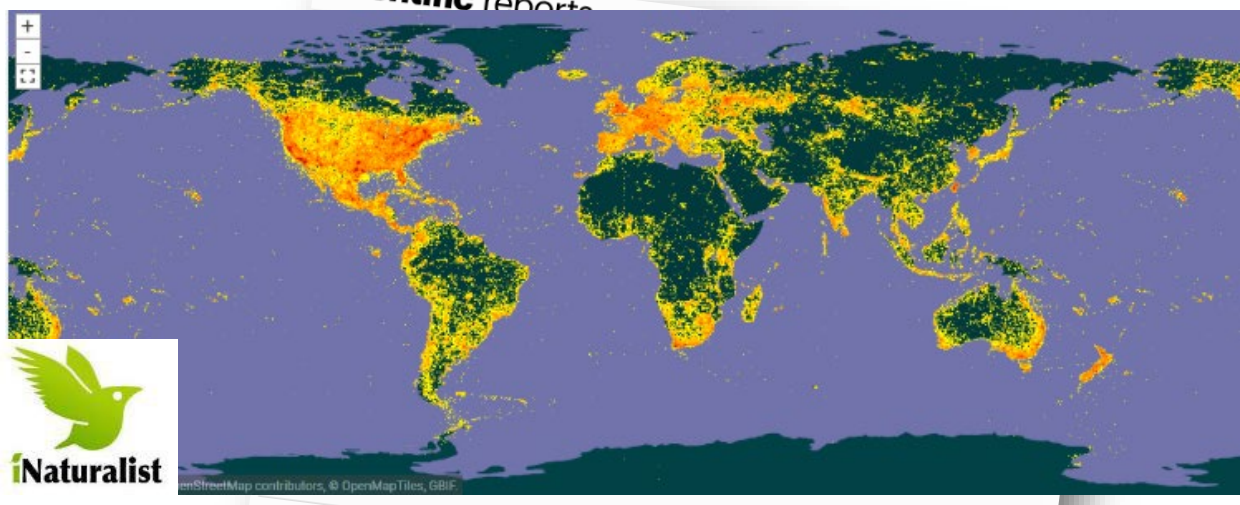
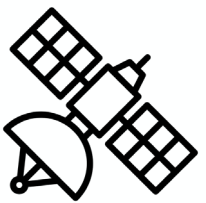


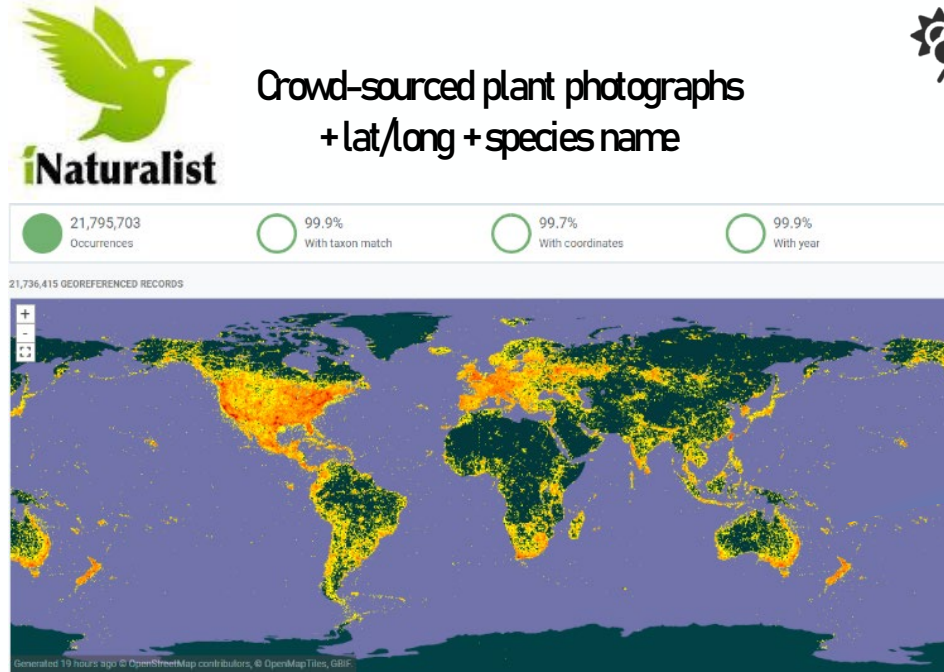
Mapping global distributions of plant functional traits from crowd-sourced plant photographs

Height?
Nitrogen?
Dry matter content?
...

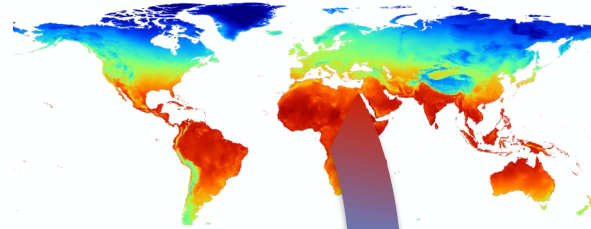


Model approach

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WorldClim Climate data + lat/long



TRY Functional traits + species names

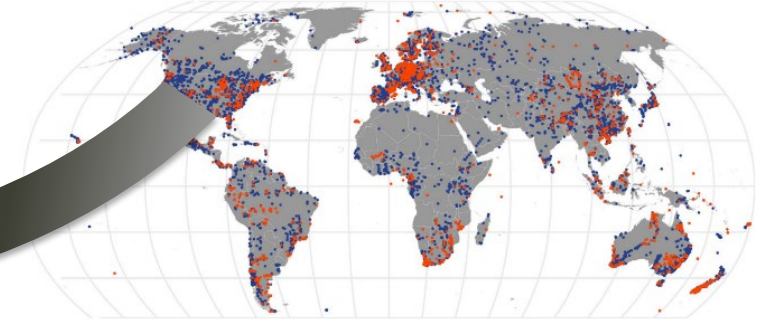


FIGURE 6 Geographic coverage of measurement sites in TRY version 1 (red) and additional measurement sites in TRY version 5 (blue)

TABLE 4 Most often requested traits. The 30 traits with the highest number of requests (status 1 October 2019). Number of requests and in parentheses the percentage relative to all 7,330 requests

Trait	Data type	Number of requests
-------	-----------	--------------------

- Specific Leaf Area
- Leaf area
- Growth height
- Stem density
- Seed mass
- Nitrogen concentration

TABLE 5 Species with best trait coverage. The 30 species with the highest number of traits in the TRY database version 5 and number of traits represented for these species in TRY version 1 and 5. Sorted by the number of traits in TRY 5

Species	Plant growth form	Number of traits	
		TRY 1	TRY 5
<i>Pinus sylvestris</i>	Tree	264	569
<i>Fagus sylvatica</i>	Tree	237	517
<i>Picea abies</i>	Tree	252	475
<i>Quercus robur</i>	Tree	194	435
<i>Acer saccharum</i>	Tree	139	430
<i>Betula pendula</i>	Tree	265	429
<i>Achillea millefolium</i>	Herb	209	403
<i>Acer pseudoplatanus</i>	Tree	186	397
<i>Trifolium pratense</i>	Herb	181	395
<i>Quercus rubra</i>	Tree	190	388
<i>Dactylis glomerata</i>	Herb	193	387
<i>Plantago lanceolata</i>	Herb	156	386
<i>Vaccinium vitis-idaea</i>	Shrub	189	382
<i>Trifolium repens</i>	Herb	173	380
<i>Fraxinus excelsior</i>	Tree	196	378
<i>Acer platanoides</i>	Tree	186	378
<i>Quercus petraea</i>	Tree	194	368

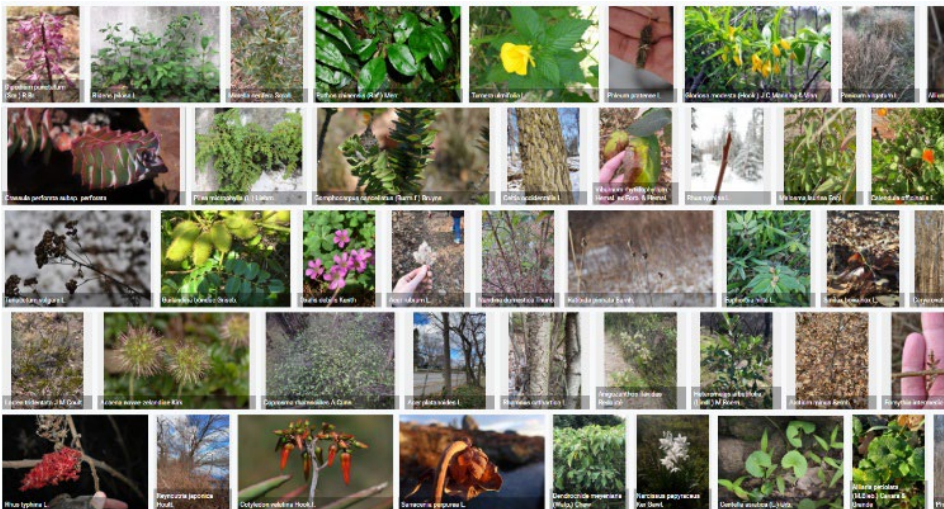
Convolutional Neural Networks

(weakly supervised training)



...predict traits from photos

- CNNEnsembles
- Trait plasticity
- With/without climate



Model Evaluation



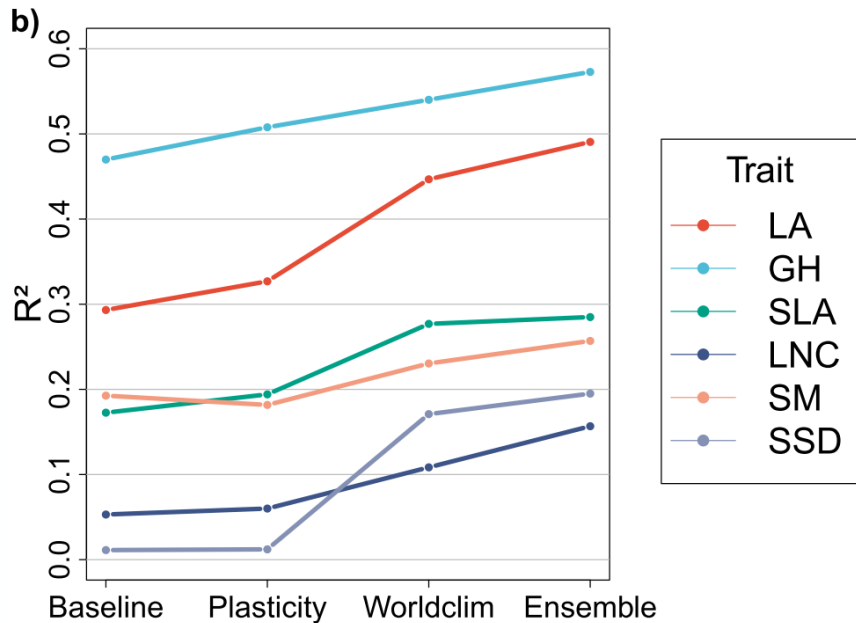
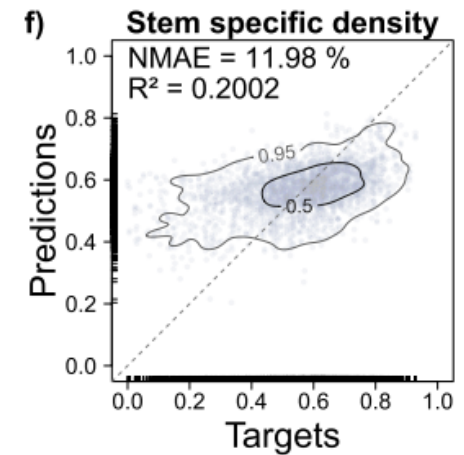
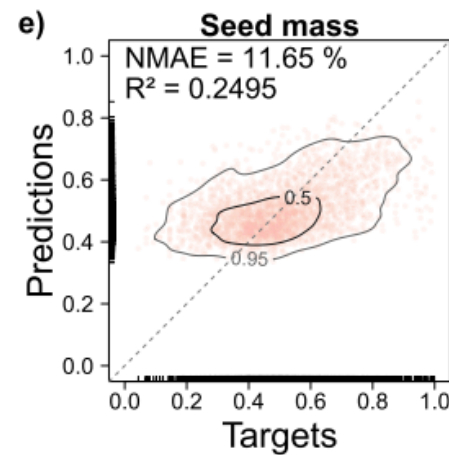
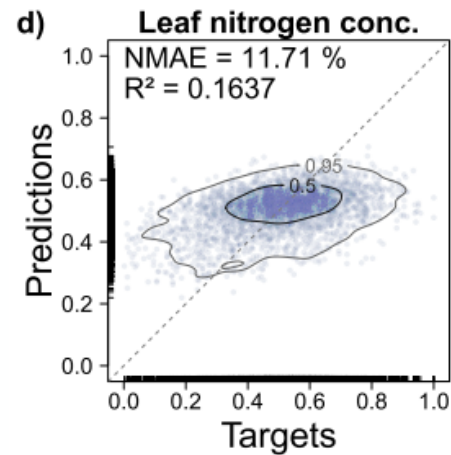
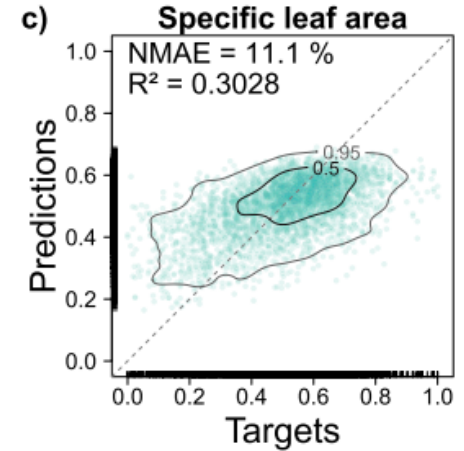
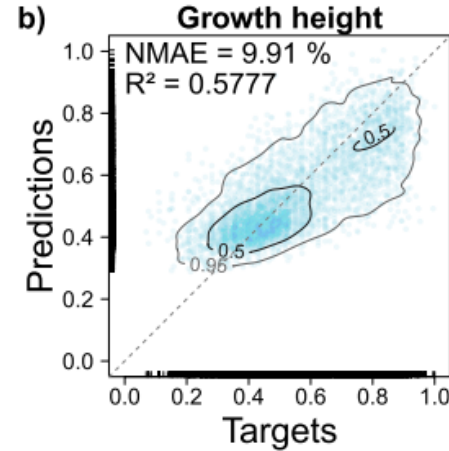
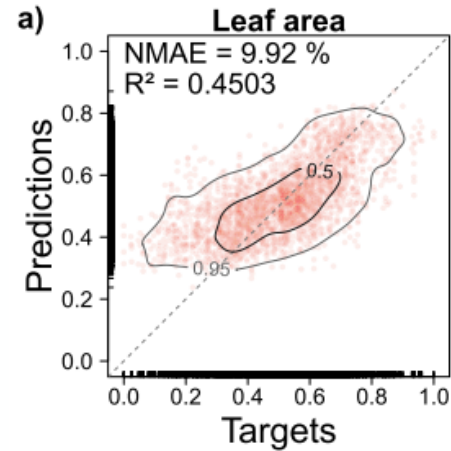
Leaf area = 23.2 cm²
Growth height = 23.7 cm
Specific Leaf Area = 70.2 cm²/mg
...



Leaf area = 12.6 cm²
Growth height = 31.1 cm
Specific Leaf Area = 87.8 cm²/mg
...



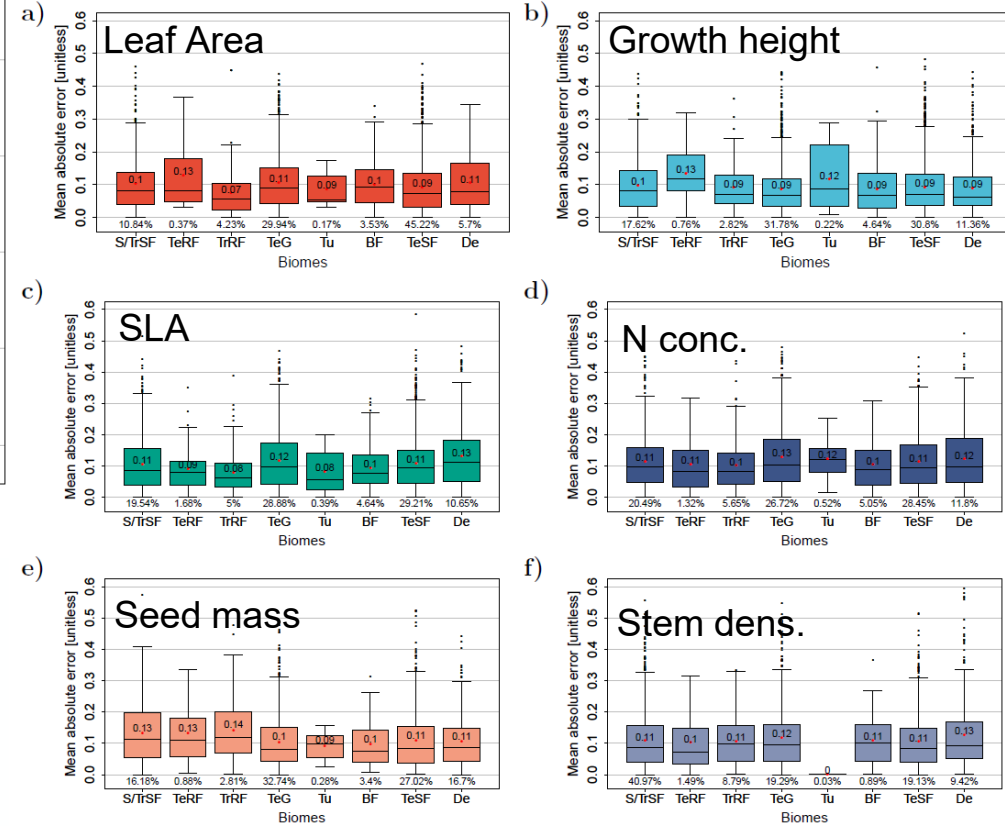
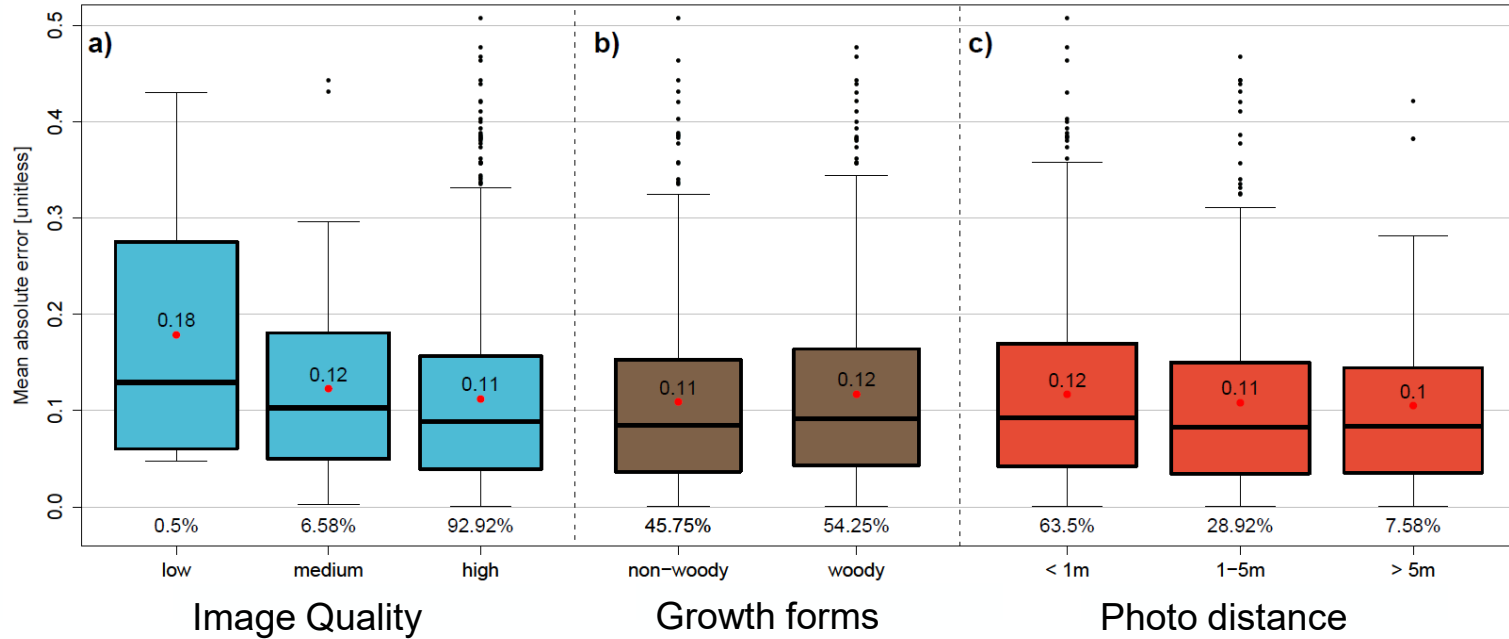
Leaf area = 3.9 cm²
Growth height = 6.3 cm
Specific Leaf Area = 45.1 cm²/mg
...



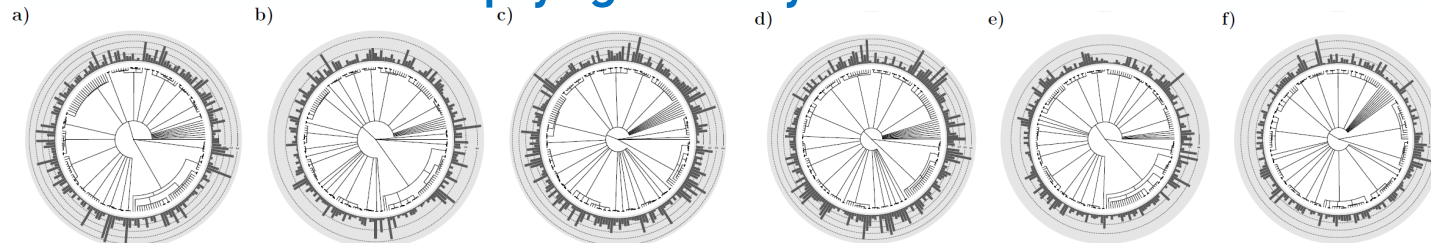
...but its just a **weak validation**.

Model Evaluation

Generalization across **image quality**, **growth form** or **image target distance**...



Prediction errors are **not phylogenetically autocorrelated**.



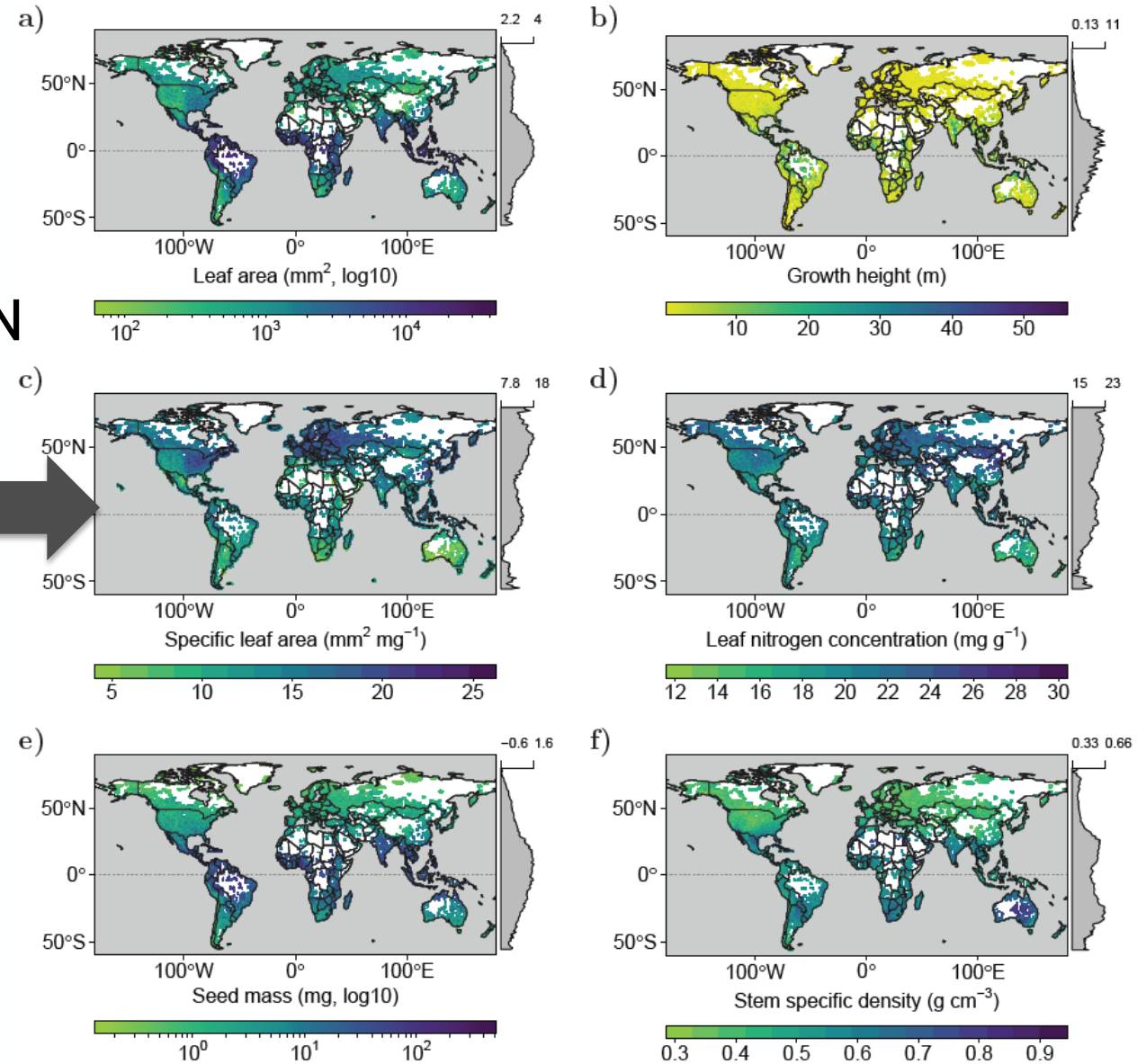
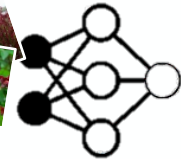
...**low bias across biomes**
indicates **global generalization**.

...from photos to global trait maps

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Trained CNN Ensemble

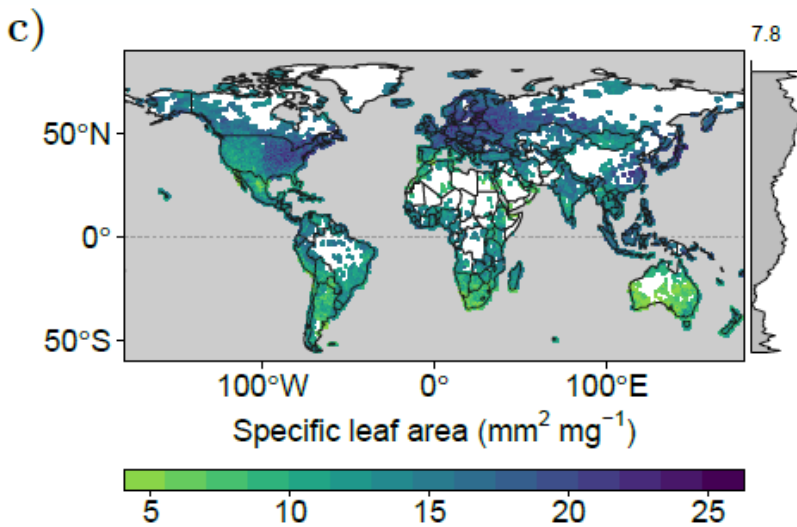
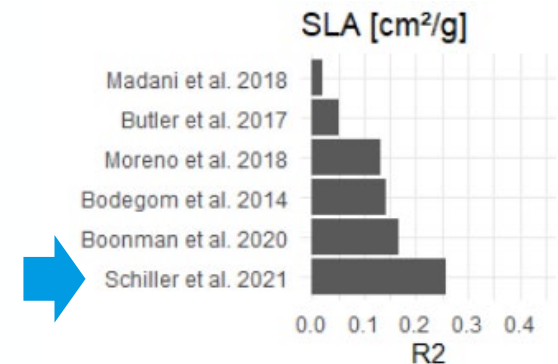
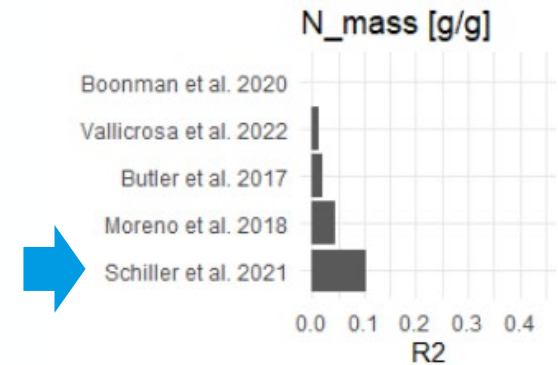
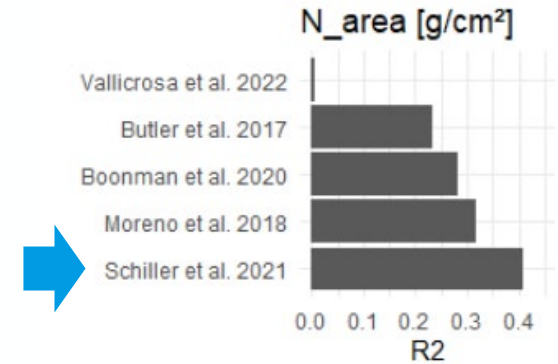


Evaluation of global trait maps

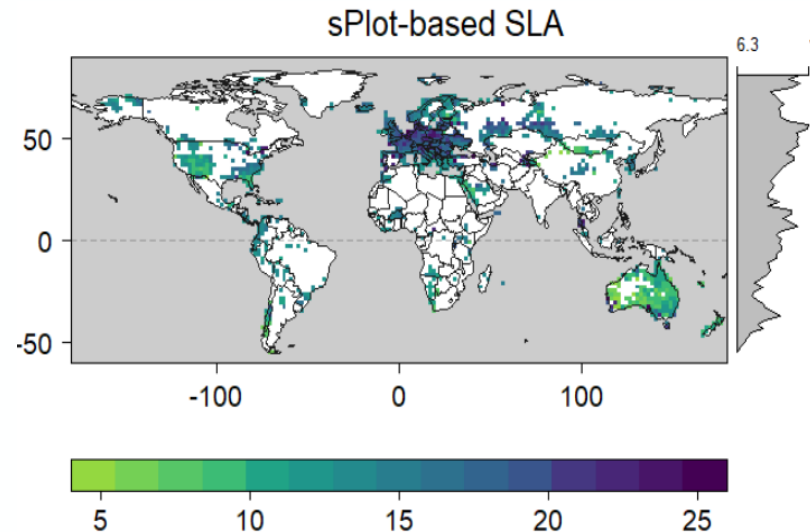
... can we really trust
our trait products?
(Schiller et al. 2021)



... community data of 95104 vegetation plots
... linked with trait data (TRY data)
→ Gridded trait community weighted mean



VS



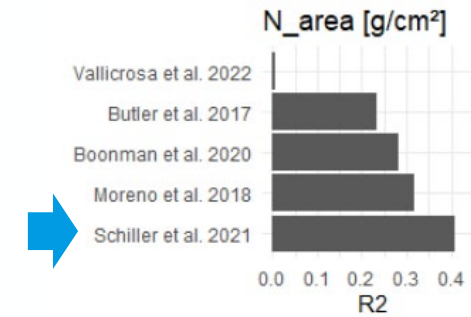
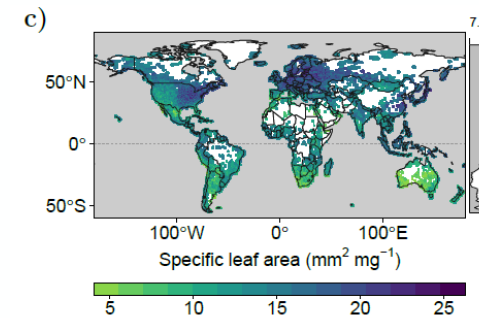
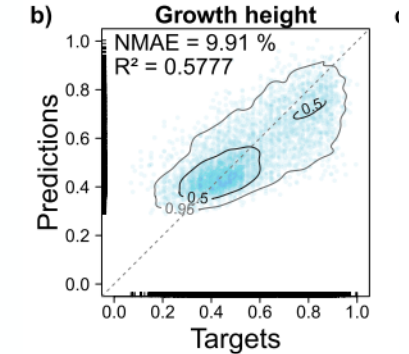
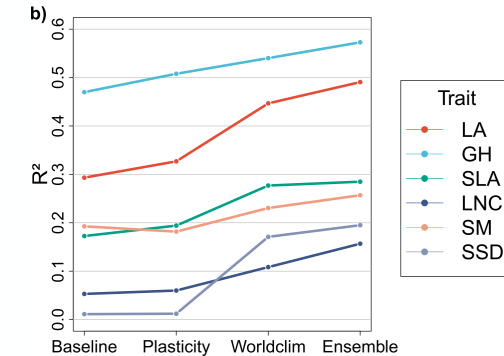
Sabatin et al. 2021, *Biogeography*
Bruehlheide et al. 2019, *JVS*.

...many thanks to Ben Dechant and the sTrait (iDiv) consortium!

Summary

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- **We can ‘see’ traits in (crowd-sourced) plant photographs**
 - ... Model performance increased with integrating trait variation, climate data, and model ensembles
- **Geocoordinates and excessive coverage (constantly increasing) enable global mapping**
 - ... appears to be more accurate than previous maps based on extrapolation methods (according to sPlot data)
- **All resources are openly available:**
 - ... Code: github.com/ChrSchiller/cnn_traits
 - ... Trait maps: tinyurl.com/schiller-trait-maps
 - ... Manuscript: www.nature.com/articles/s41598-021-95616-0



Thank you!

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More interest in global trait maps?

Poster Sophie Wolf (A2.01 Biodiversity, today!)

...from photographs to global maps

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