

11– Tree health and mortality mapping

- Relevance of tree mortality mapping
- Towards of global tree mortality mapping
- The platform deadtrees.earth
- Recent findings



Chair of Sensorbased Geoinformatics (geosense)
www.geosense.uni-freiburg.de

Relevance of tree mortality mapping



Relevance of tree mortality mapping



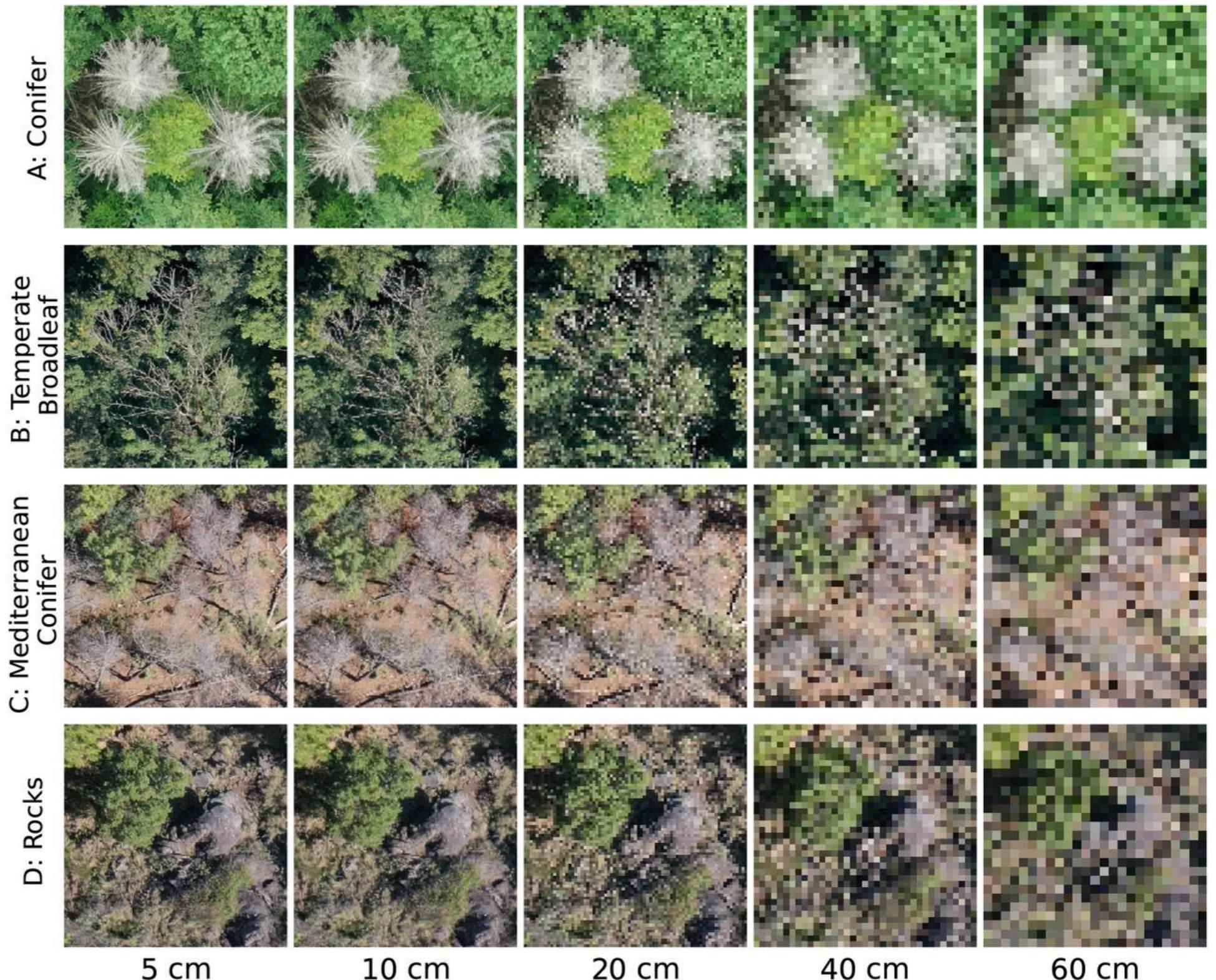
Relevance of tree mortality mapping



- Large-scale assessment of tree mortality only possible with remote sensing (satellites, airplanes,)
- Translating remote sensing signals into tree mortality patterns requires „training data“ (supervised learning)
- Inventory data could be a valuable resource. However, forestry is known to be a conservative/restrictive field, where institutions and companies do not or cannot share inventory data
- What to use instead of inventory data?

Relevance of tree mortality mapping

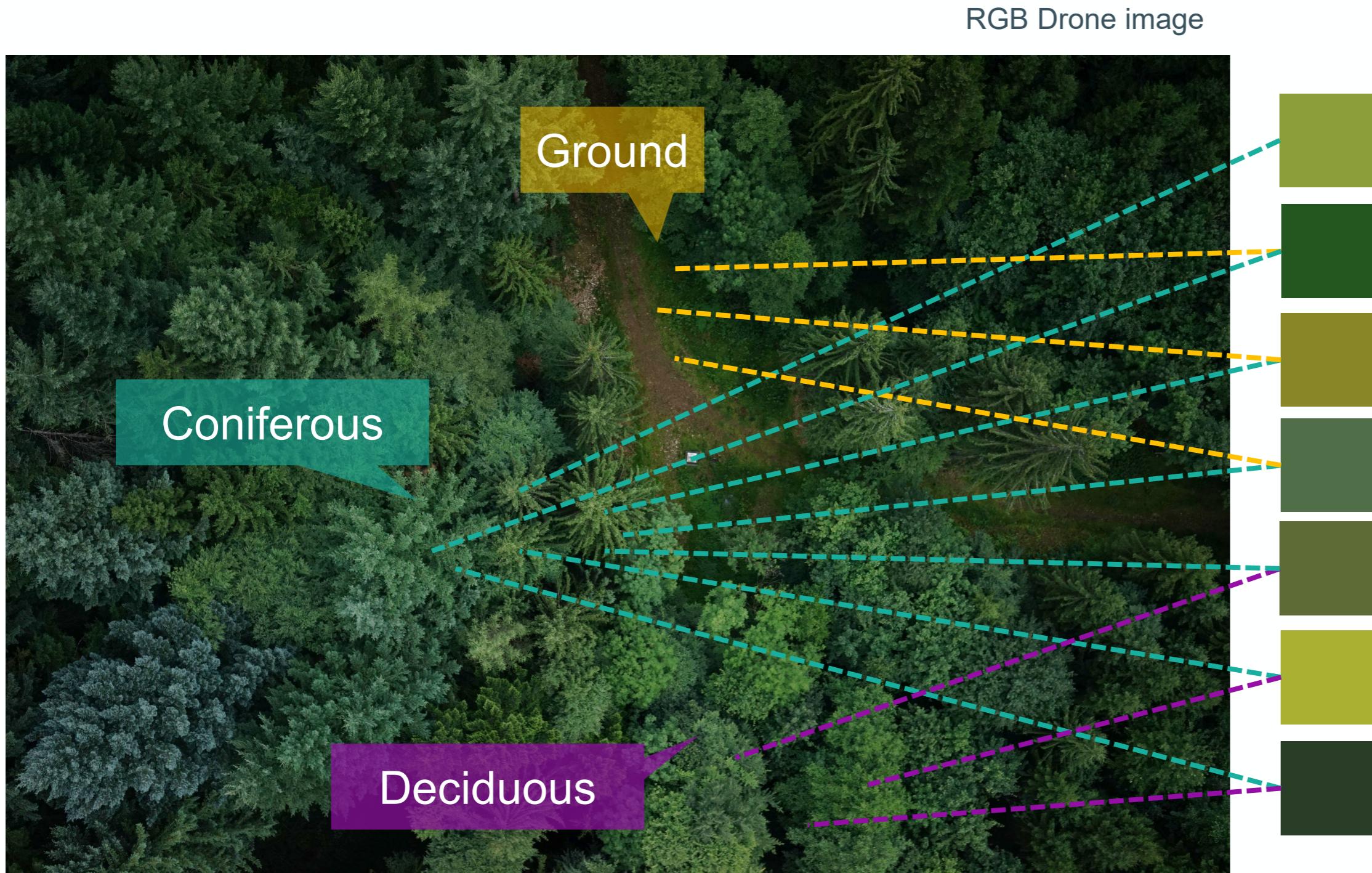
- Aerial imagery is a promising, widely available resource to detect dead trees (at least in the vegetation season).
- The spatial resolution can be of cardinal importance.
- Drones enable to create aerial imagery with unprecedeed resolution.
- Automatization is key.



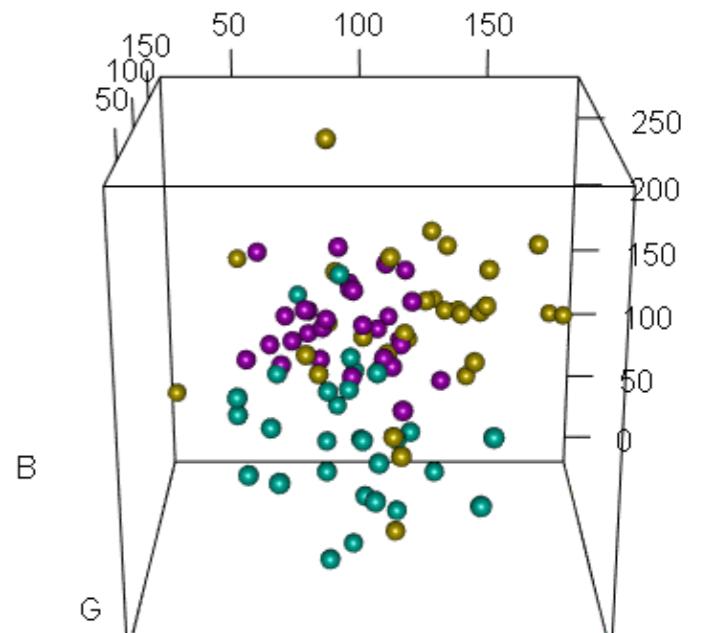
11 – Tree health and mortality mapping

Recap pattern recognition

Recap pattern recognition

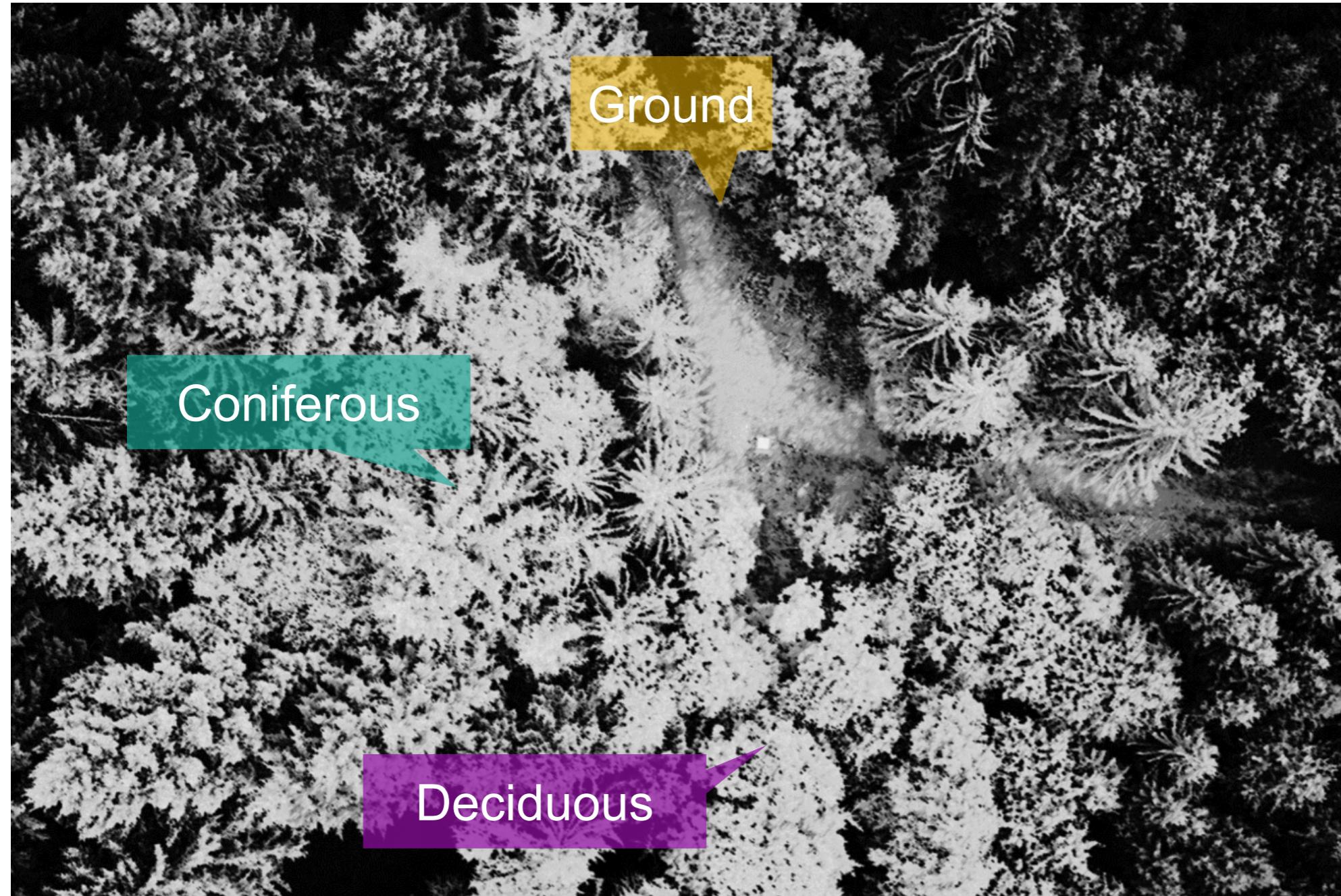


RGB color space:



- Conifer
- Deciduous
- Ground

Recap pattern recognition



Texturmetriken

- Entropy
- Variance
- Mean
- Homogeneity
- etc...

input

3 ₀	3 ₁	2 ₂	1	0
0 ₂	0 ₂	1 ₀	3	1
3 ₀	1 ₁	2 ₂	2	3
2	0	0	2	2
2	0	0	0	1

output

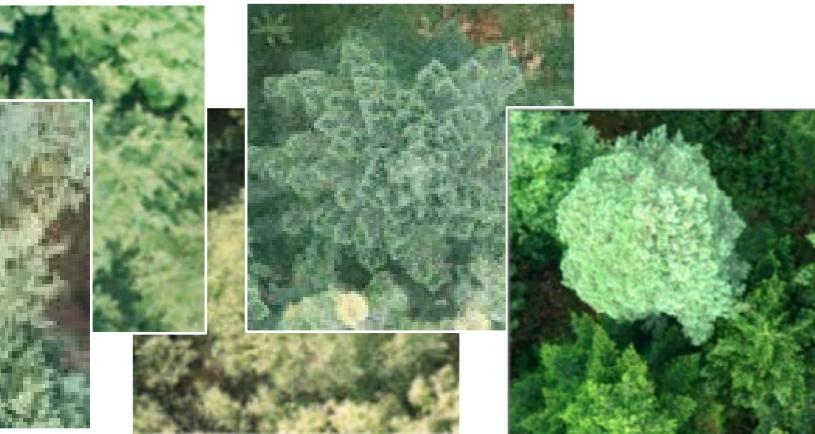
12.0	12.0	17.0
10.0	17.0	19.0
9.0	6.0	14.0

Haralick, R. M., & Shanmugam, K. (1973). Textural features for image classification. *IEEE Transactions on systems, man, and cybernetics*, (6), 610-621.

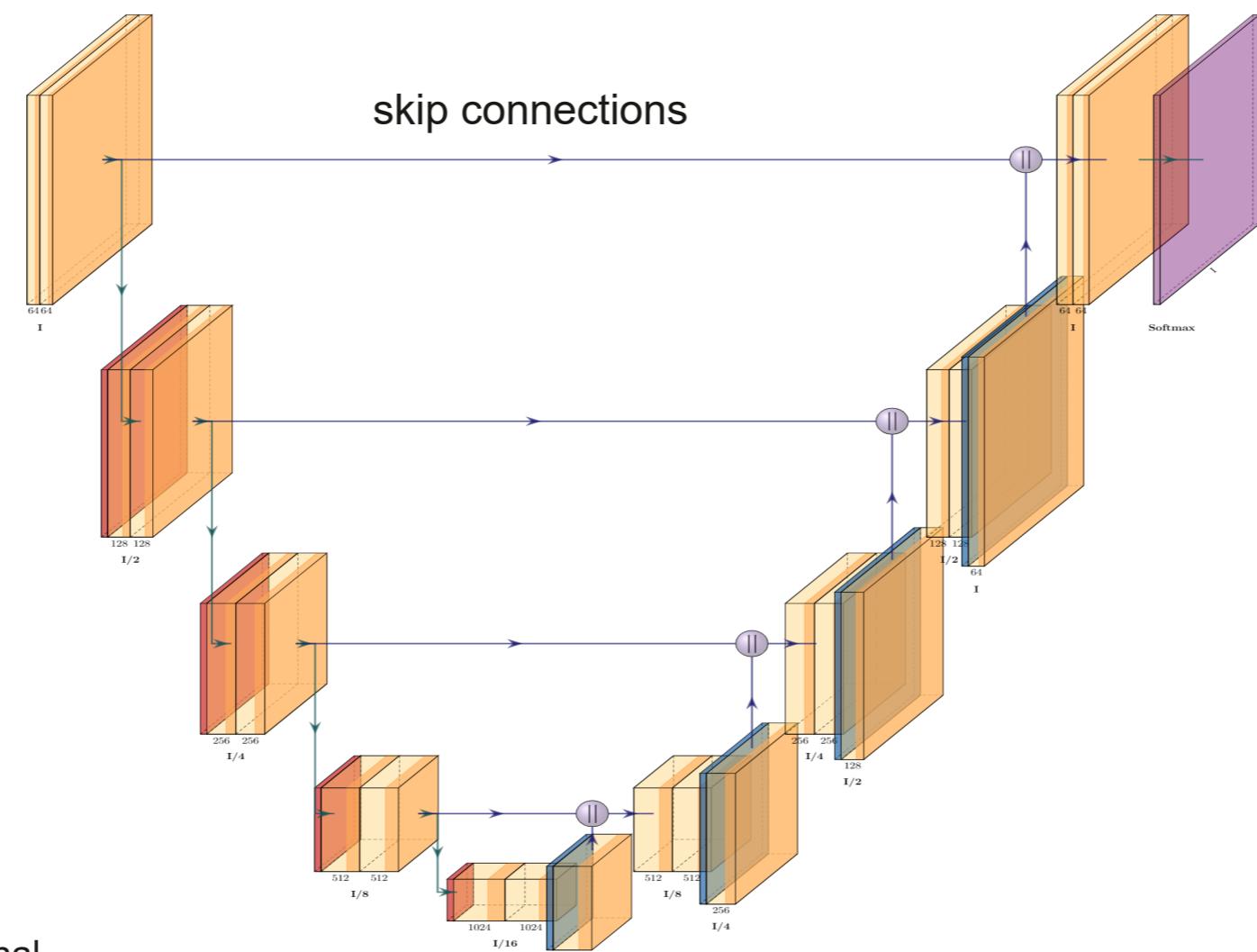
Recap pattern recognition

Image Segmentation with Convolutional Neural Networks (CNN, Unet)

Predictors (x, image data)



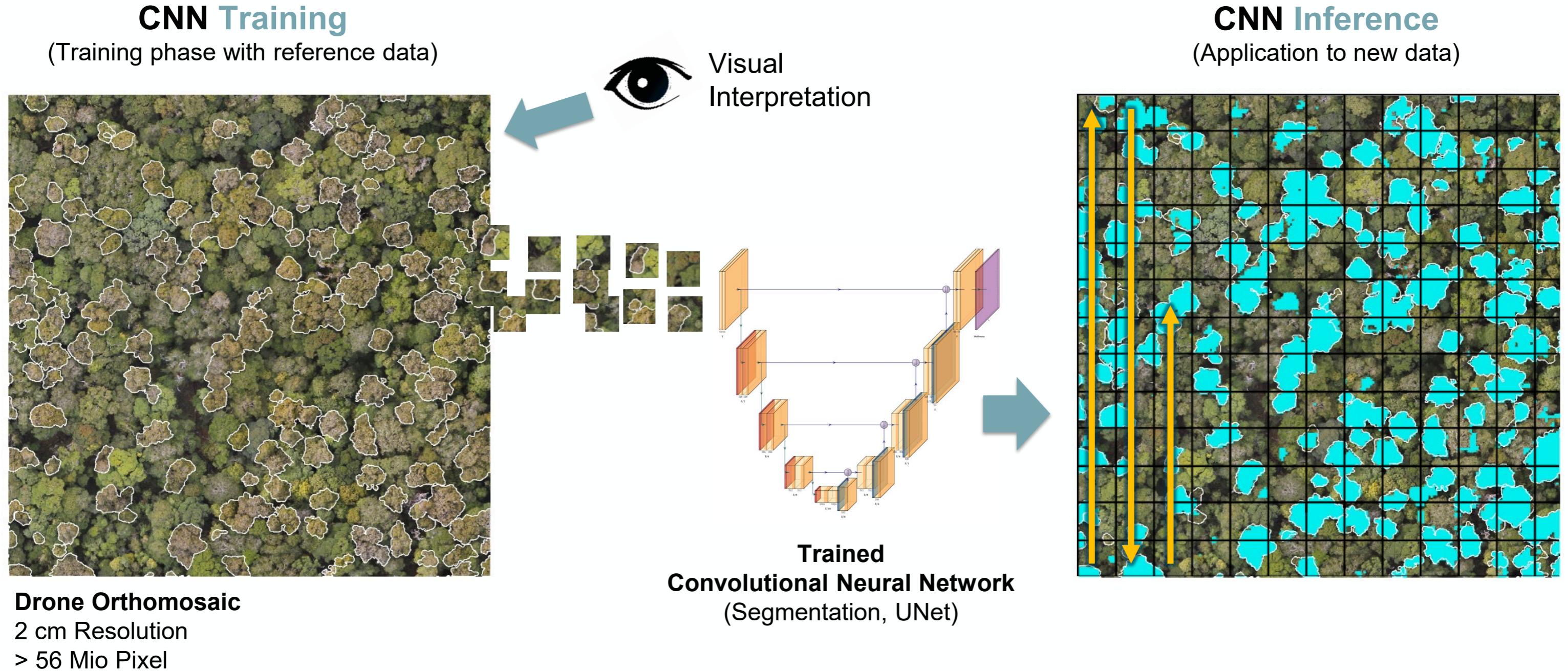
Output: Segmentierung



<https://github.com/Harislqbal88/PlotNeuralNet>

Kattenborn et al. 2022; Review on Convolutional Neural Networks for vegetation remote sensing, ISPRS.

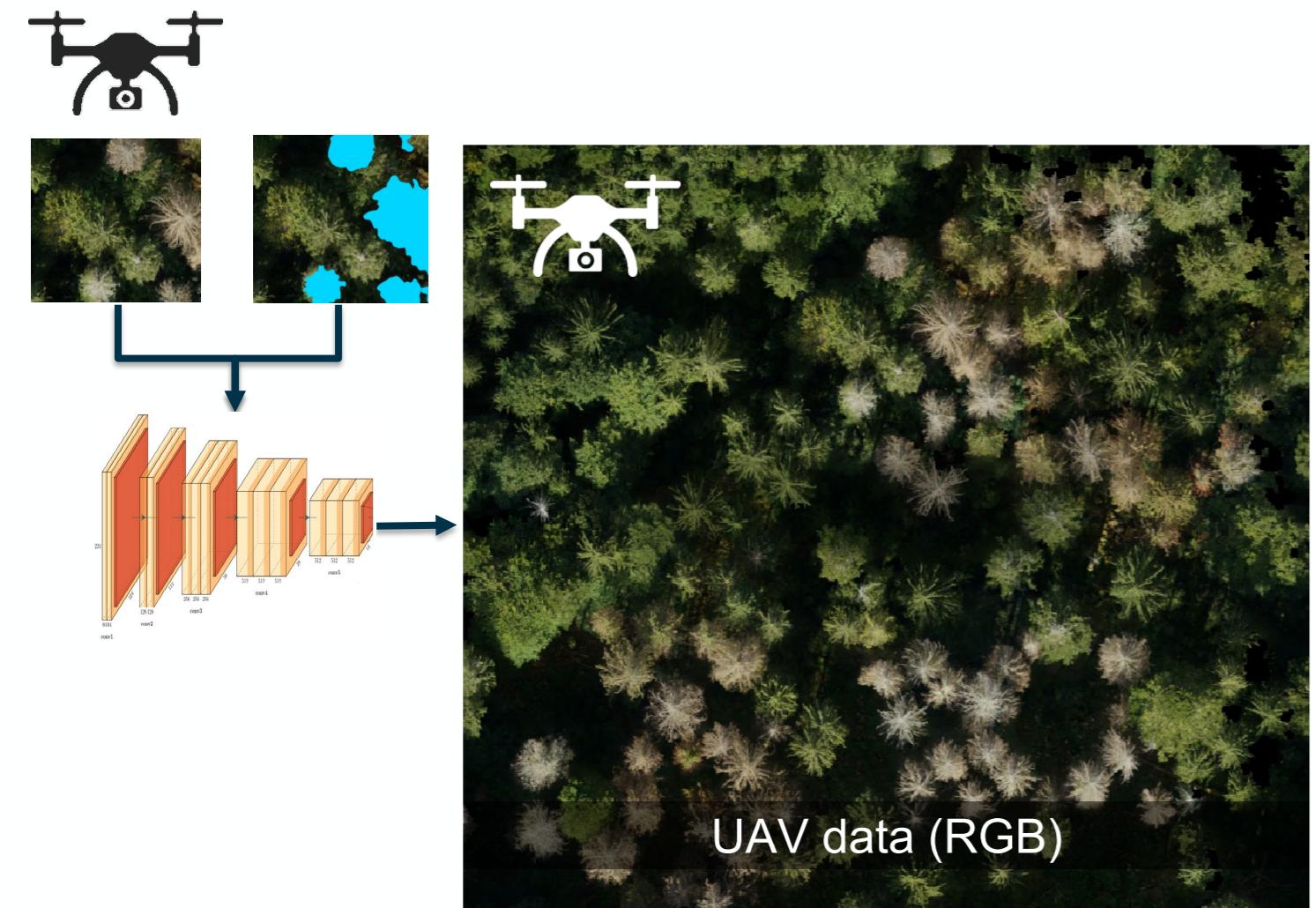
Recap pattern recognition



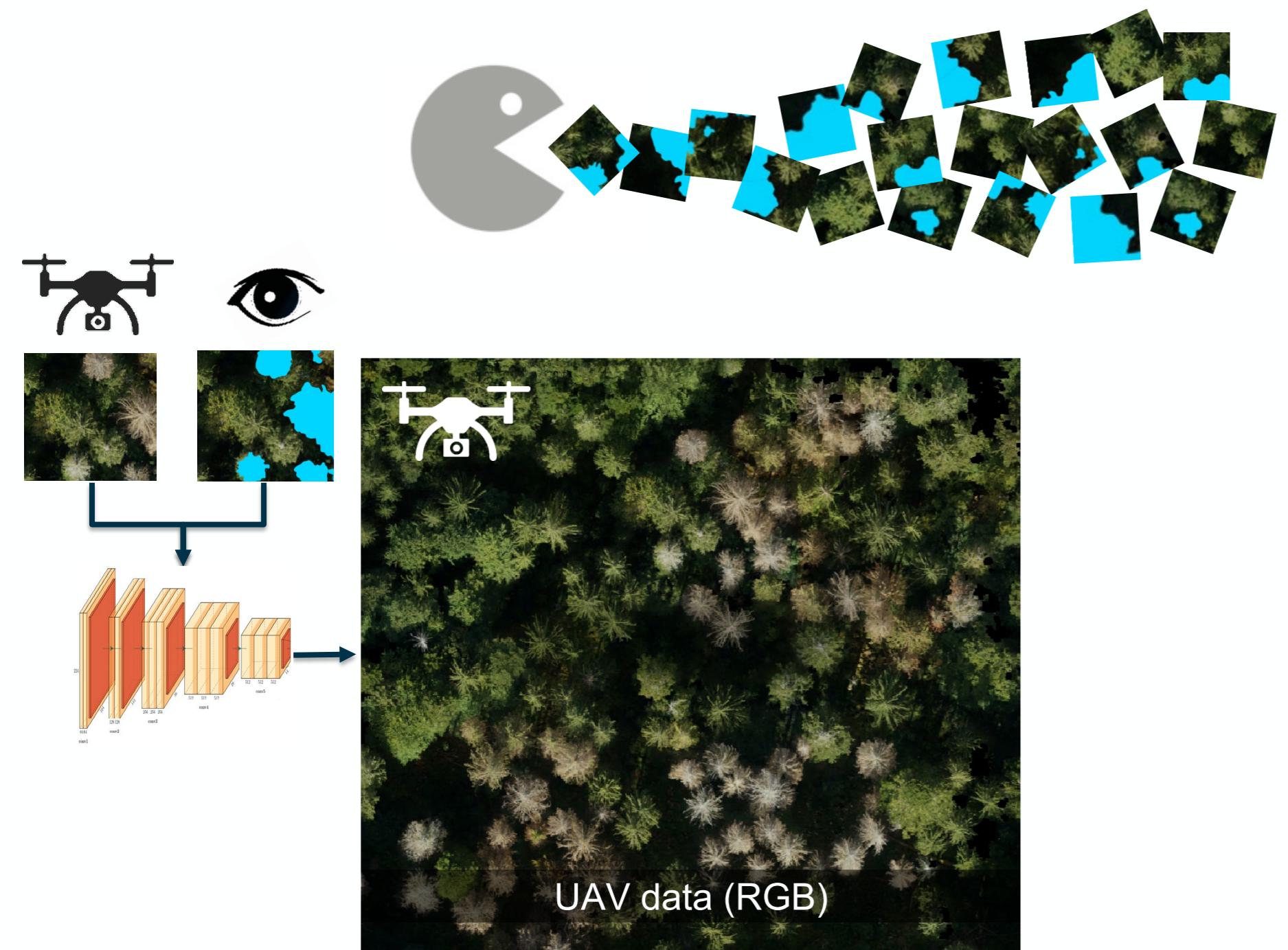
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Towards of global tree mortality mapping

Towards global tree mortality mapping

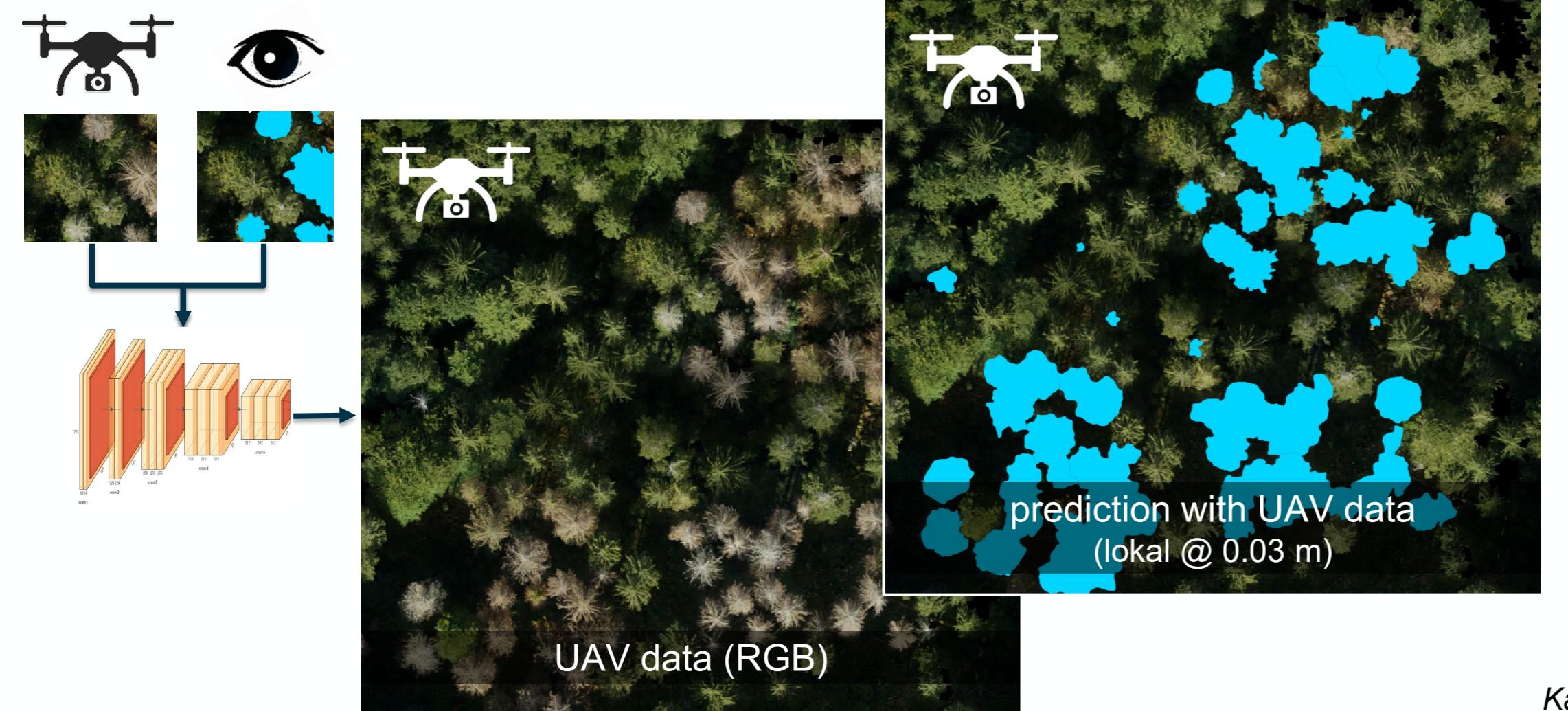


Towards global tree mortality mapping



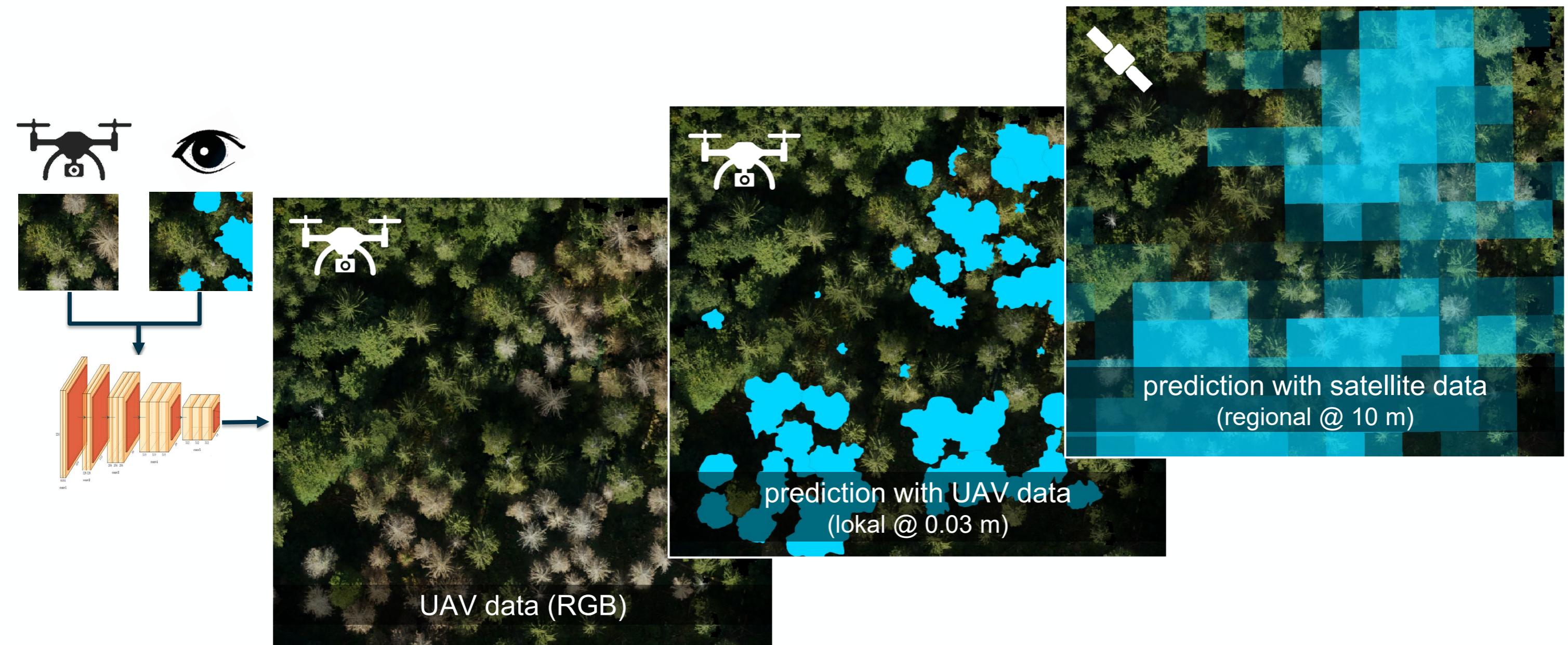
Kattenborn et al. (2019) *Scientific reports*
Schiefer et al. (2023) *ISPRS Open*

Towards global tree mortality mapping

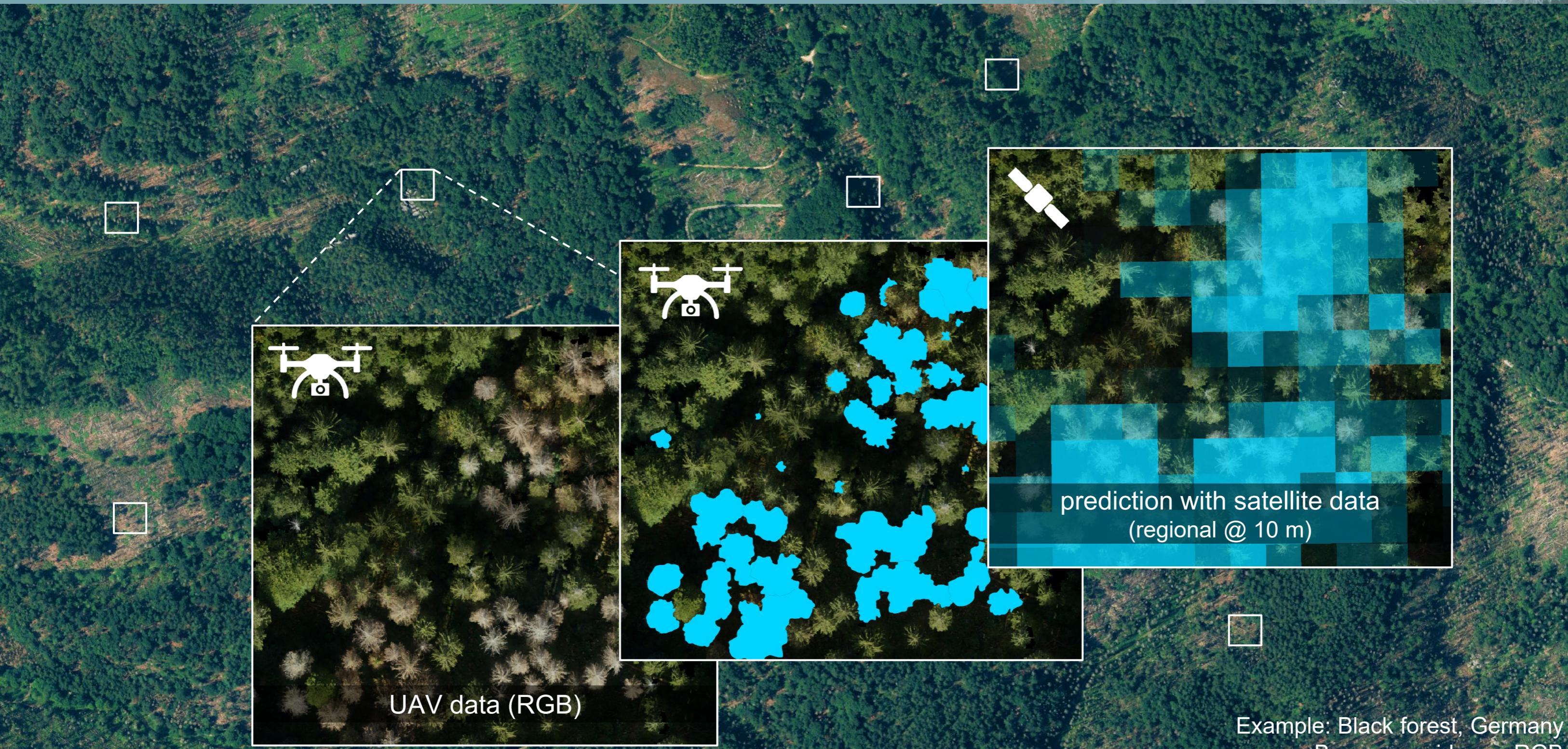


Kattenborn et al. (2019) *Scientific reports*
Schiefer et al. (2023) *ISPRS Open*

Towards global tree mortality mapping

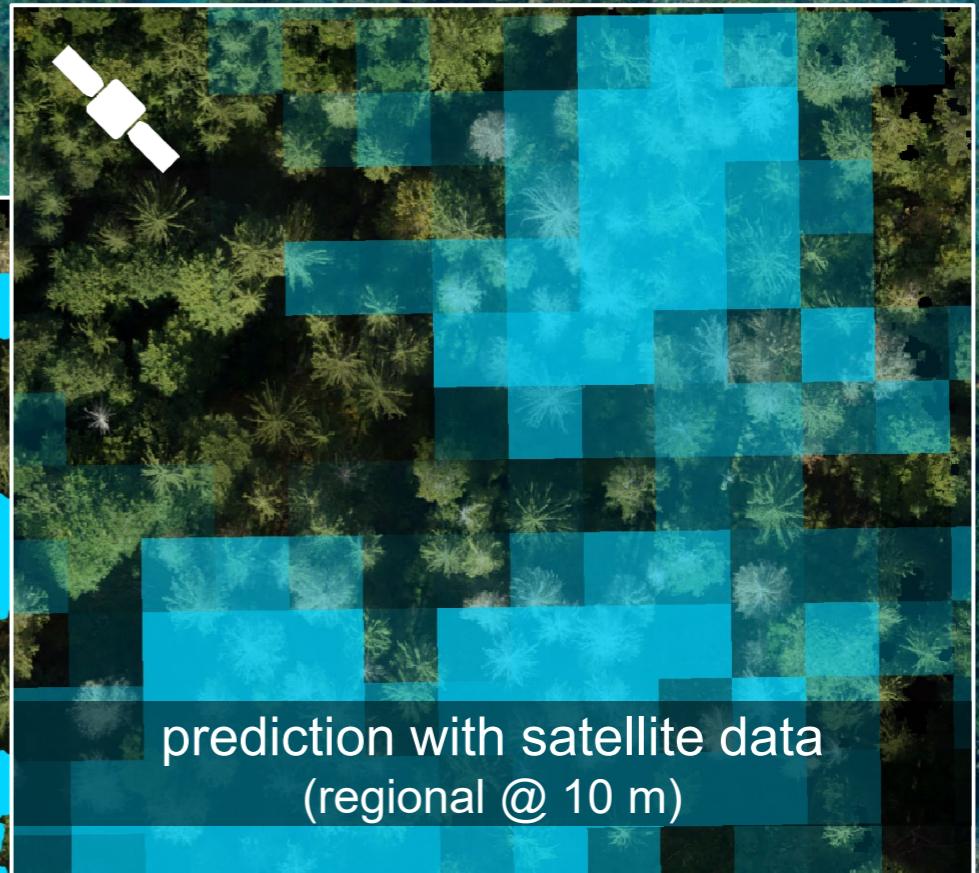
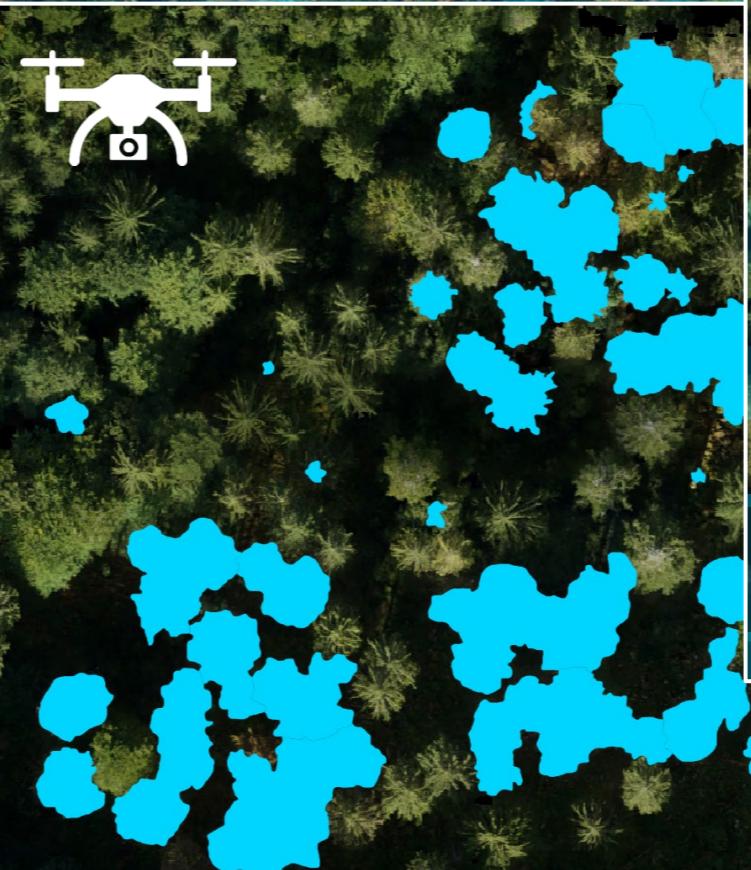
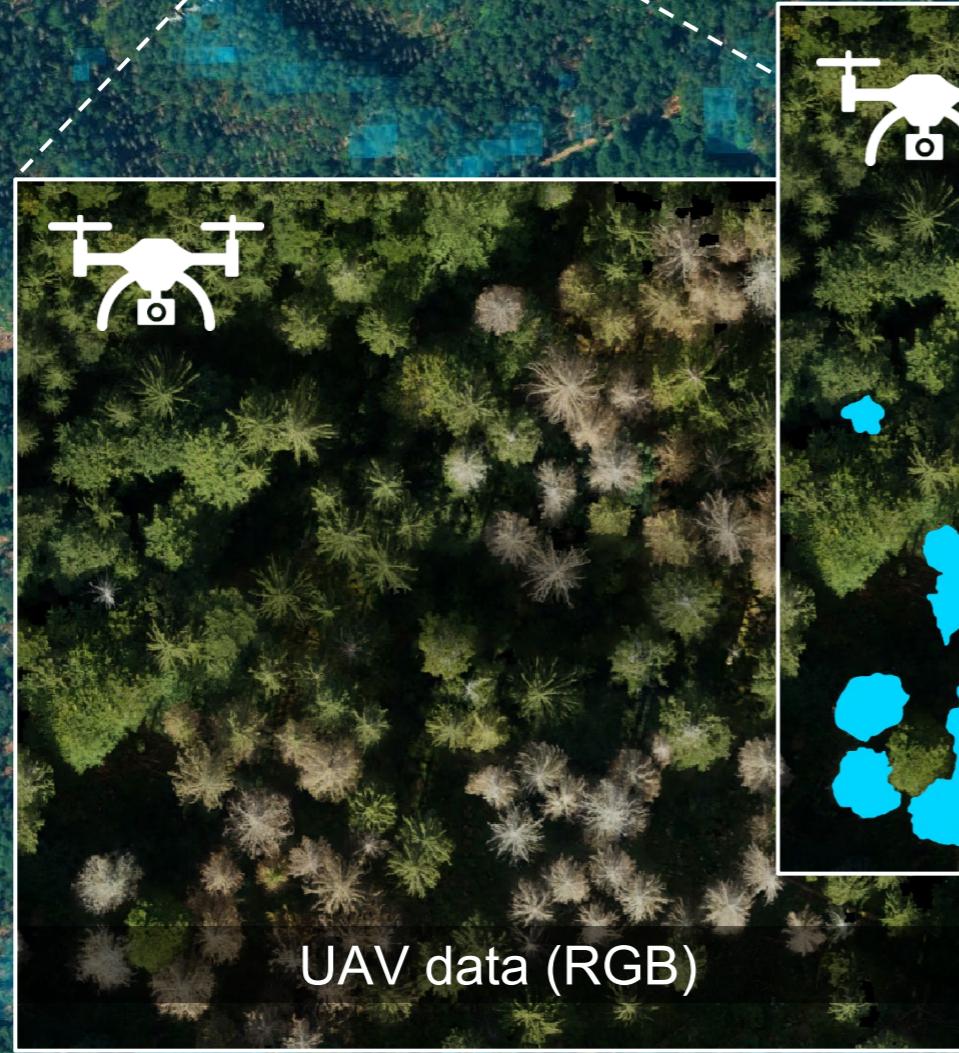


Towards of global tree mortality mapping



Towards of global tree mortality mapping

2018



Towards of global tree mortality mapping

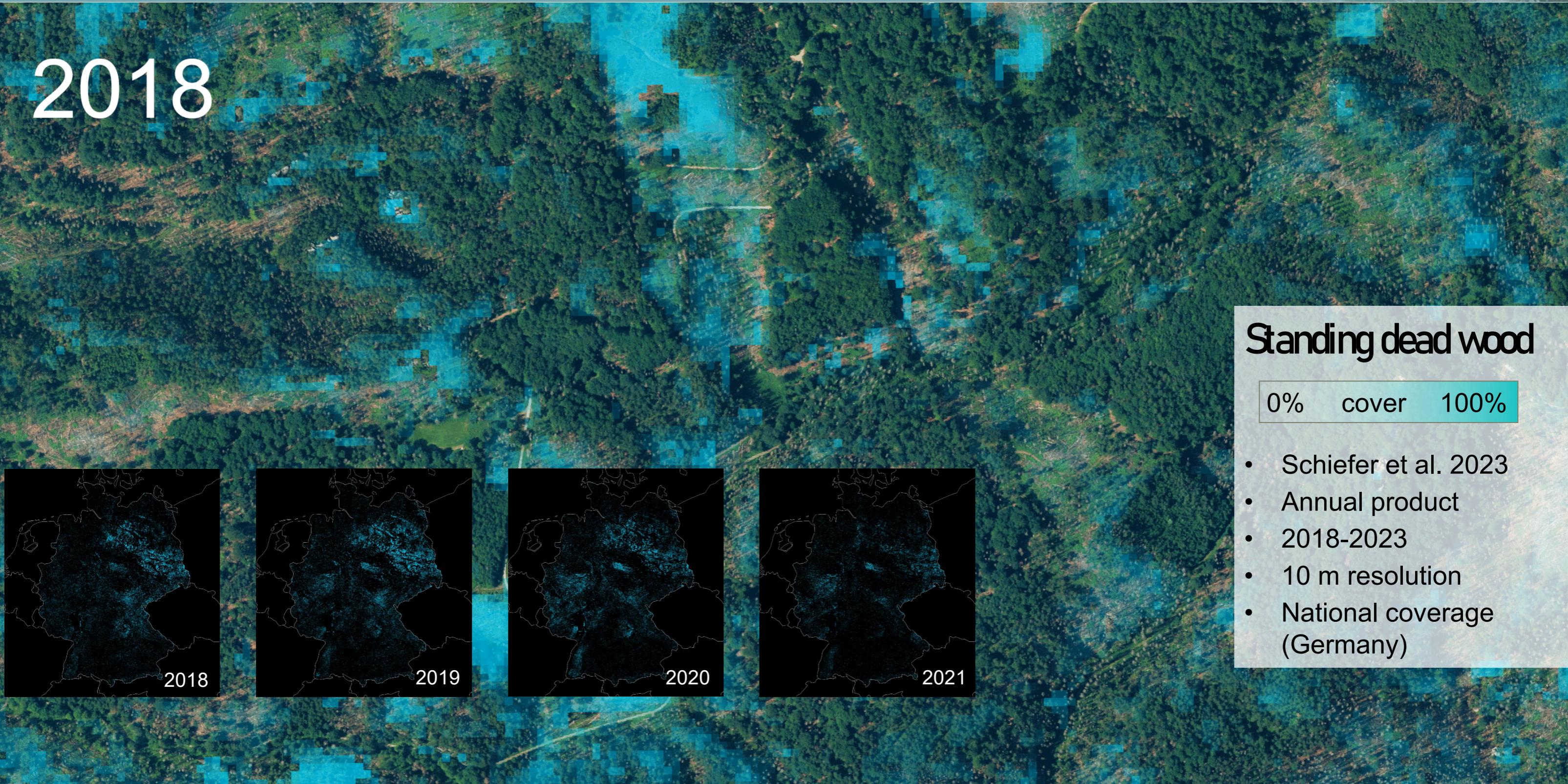
2018

Standing dead wood

0% cover 100%

- Schiefer et al. 2023
- Annual product
- 2018-2023
- 10 m resolution
- National coverage (Germany)

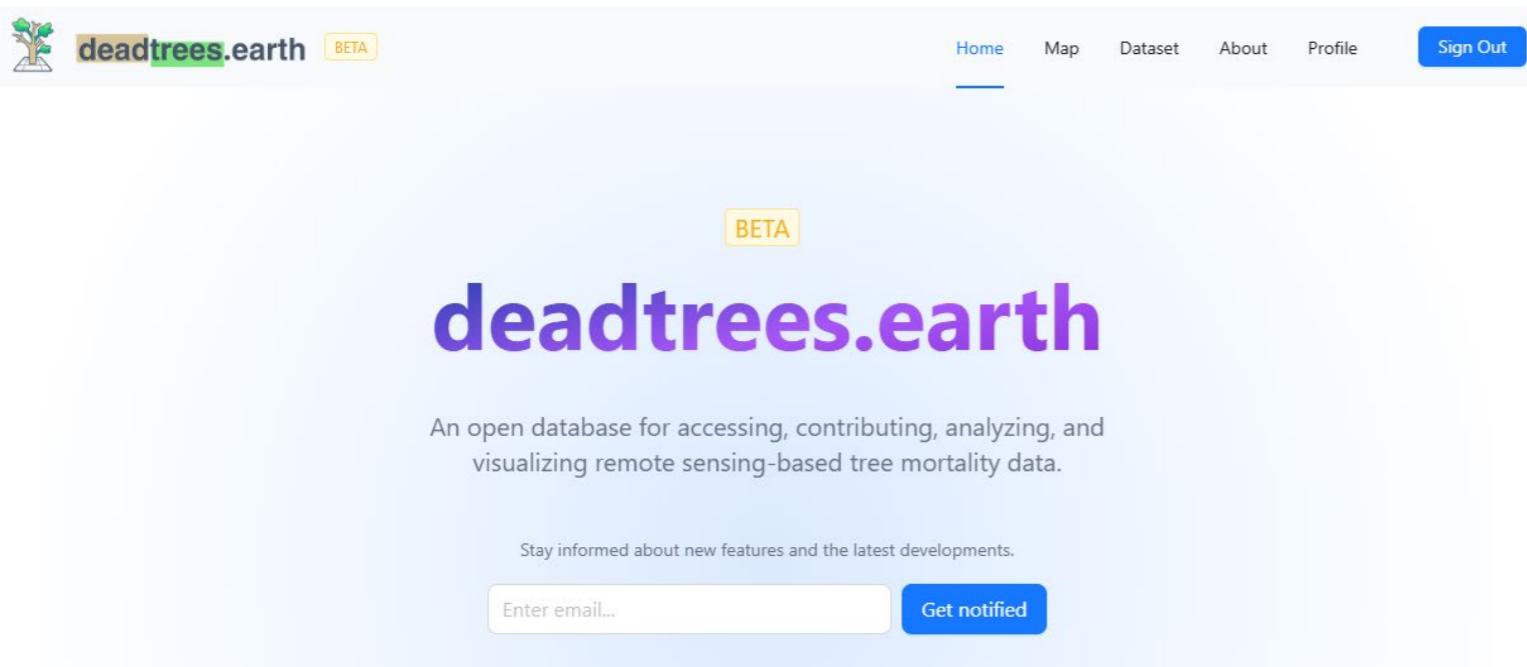
Towards of global tree mortality mapping



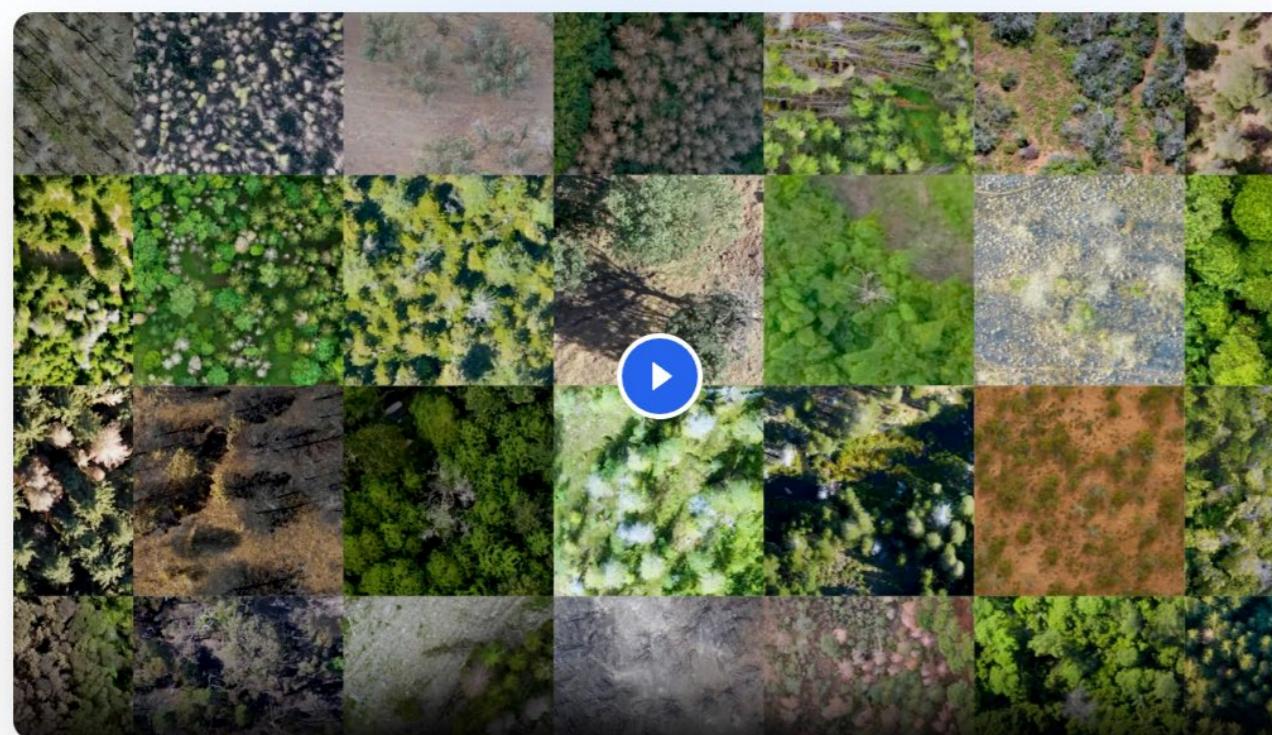
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The platform
deadtrees.earth

The platform deadtrees.earth



The screenshot shows the homepage of the deadtrees.earth website. At the top, there is a navigation bar with links for Home, Map, Dataset, About, Profile, and Sign Out. A "deadtrees.earth" logo with a small tree icon and a "BETA" button is on the left. Below the navigation, the word "deadtrees.earth" is prominently displayed in large purple letters. A sub-headline reads: "An open database for accessing, contributing, analyzing, and visualizing remote sensing-based tree mortality data." Below this, there is a call-to-action section with a text input field labeled "Enter email..." and a blue "Get notified" button. A "BETA" button is also present here.



OUR SERVICES TO THE COMMUNITY

Revealing tree mortality patterns

By integrating Earth observation, machine learning, and ground-based data sources, this initiative aims to bridge the existing gaps in understanding global tree mortality dynamics, fostering a comprehensive and accessible resource for researchers and stakeholders alike.



Open access community effort

Upload and download your aerial imagery with optional delineations of standing deadwood. Every contributor will be credited and invited to collaborate.



Automatic dead tree detection

Automatic detection (semantic segmentation) of dead trees in uploaded aerial imagery through a generic detection computer vision model.



Large-scale tree mortality map

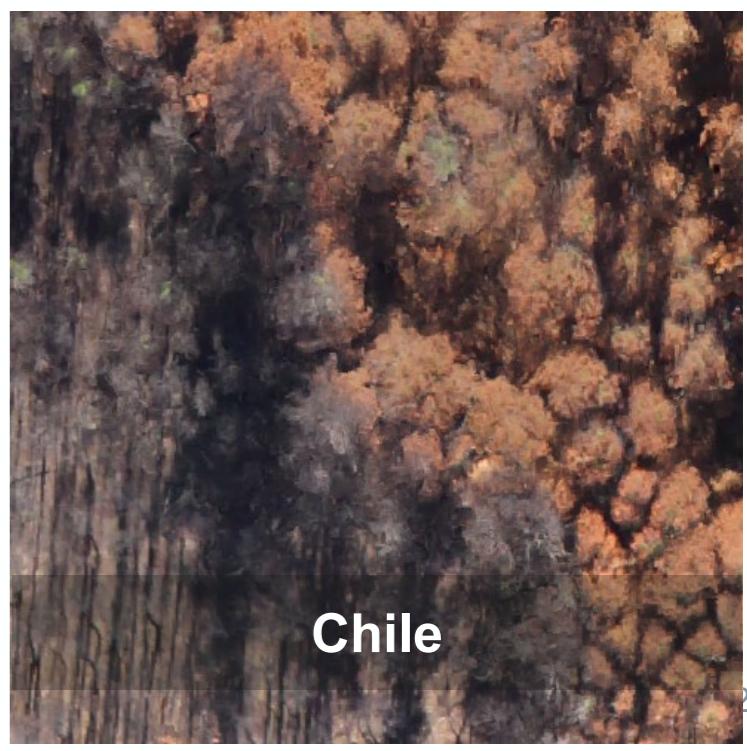
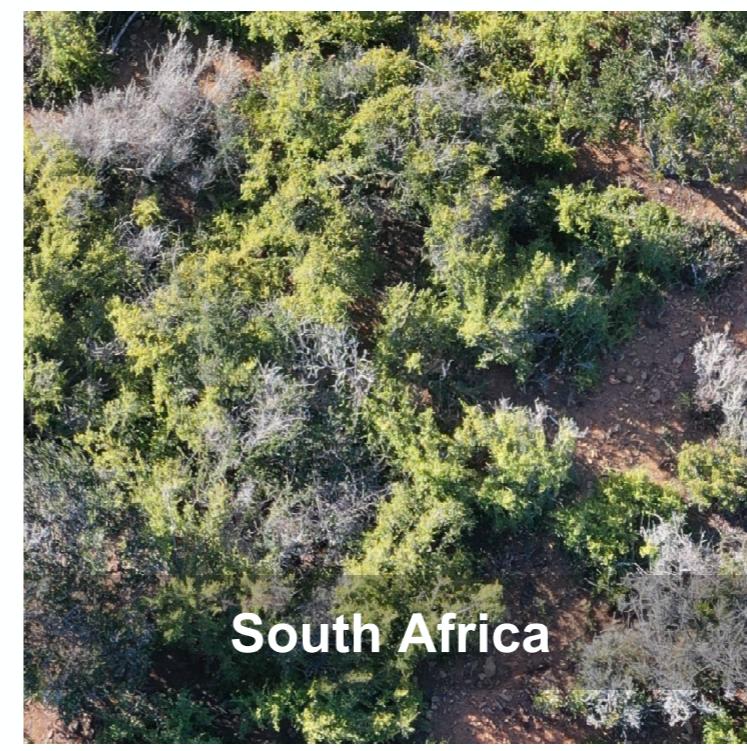
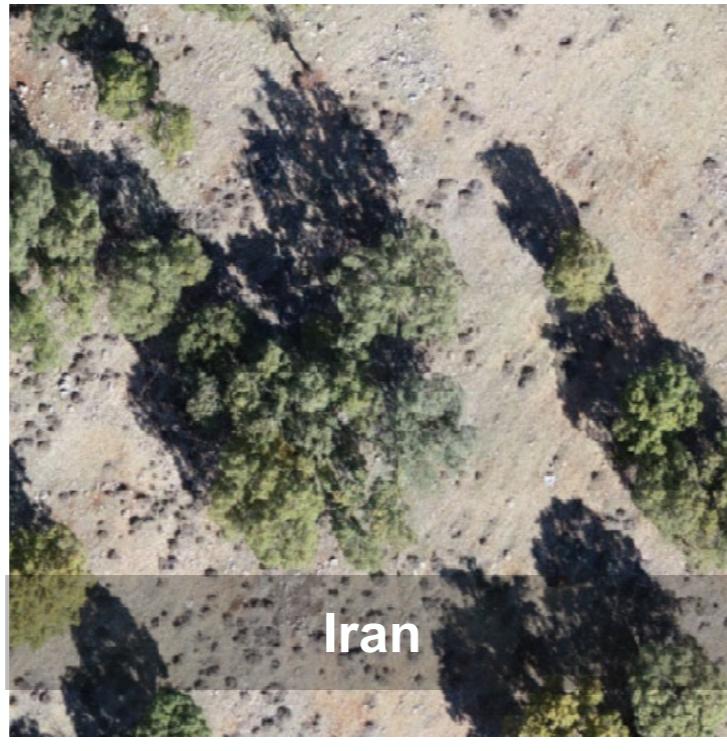
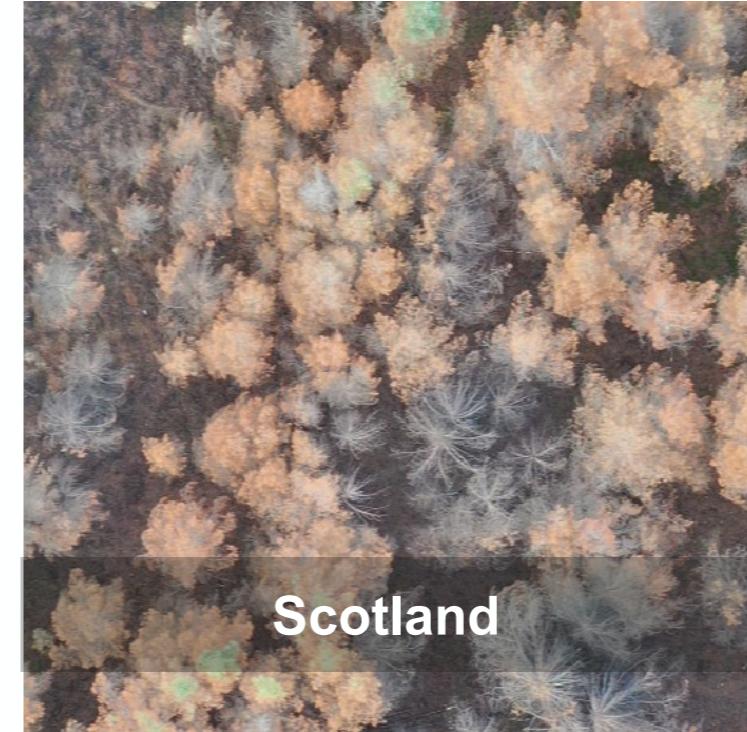
Embedded visualization and download of extensive spatiotemporal tree mortality products derived from extrapolating standing deadwood using Earth observation data.



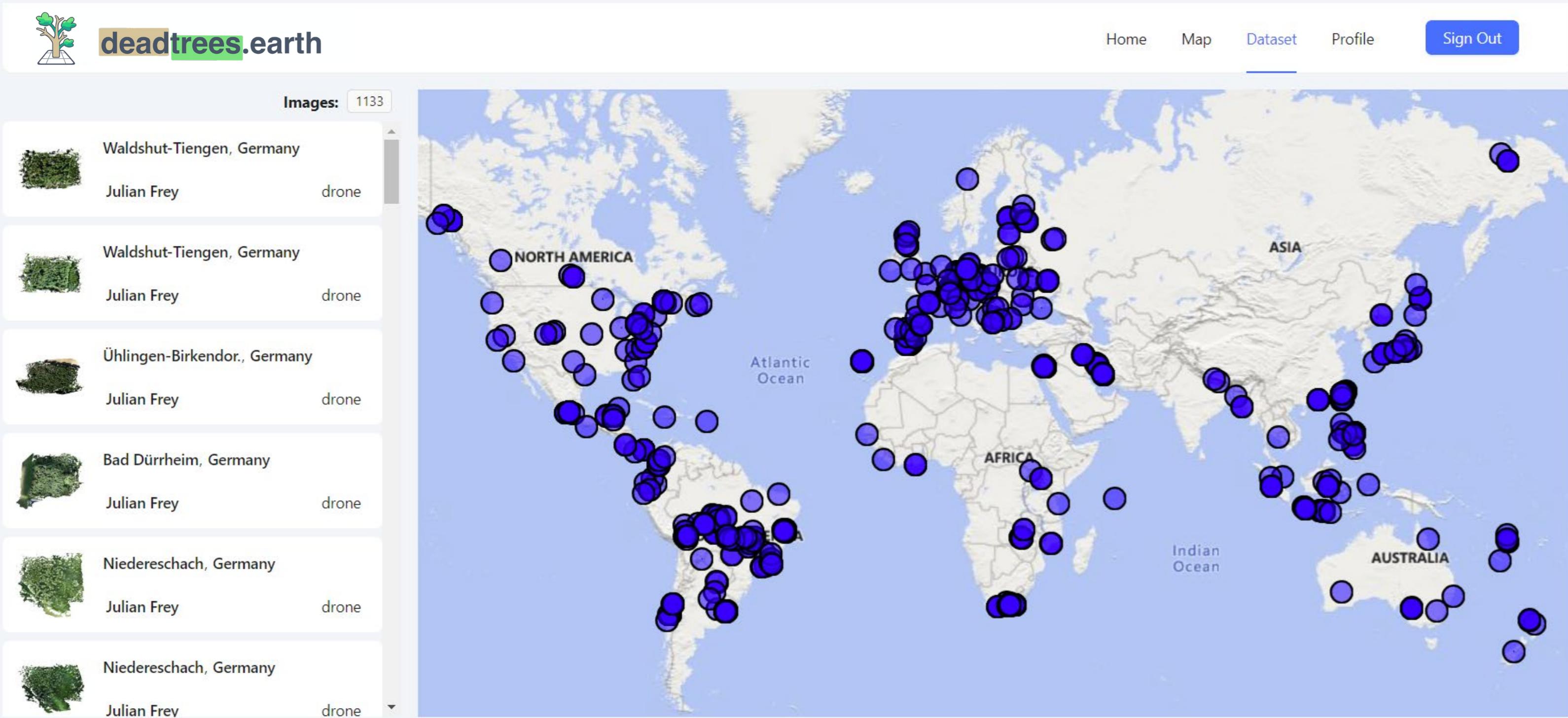
Analysis ready training data

High-resolution aerial imagery of forests worldwide together with delineated standing deadwood which can be used for training your own AI models.

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Acquisition Date: October 4, 2019

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Platform: drone

File Size: 602 MB

Spectral Properties : RGB

Label Source: visual_interpretation

Label Type: semantic_segmentation

Label Quality: 3



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Label Source: visual_interpretation

Label Type: semantic_segmentation

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Satellite Streets



Share of standing deadwood (%)

100% -
50% -
0% -

Deadwood for 2018

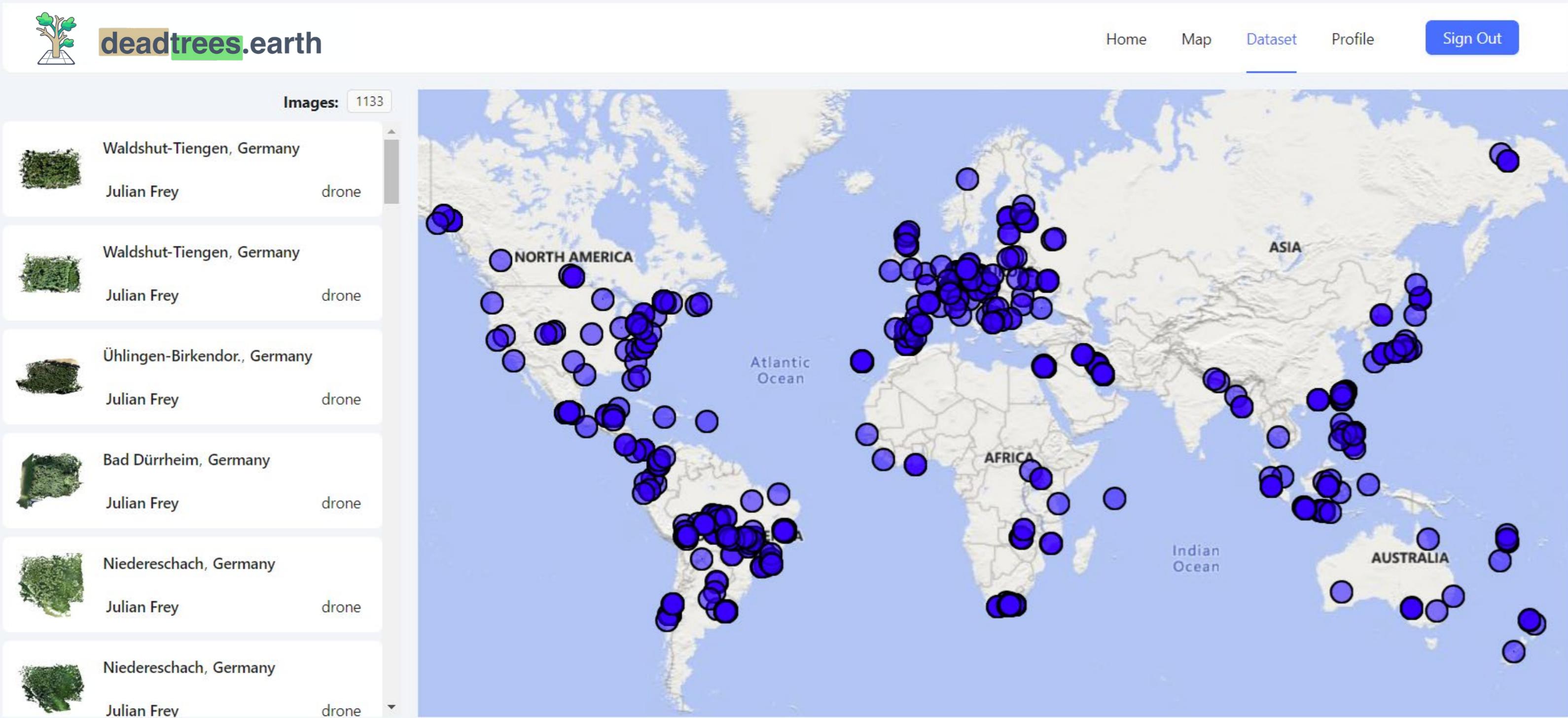
Drone-based label opacity

Satellite-based prediction opacity

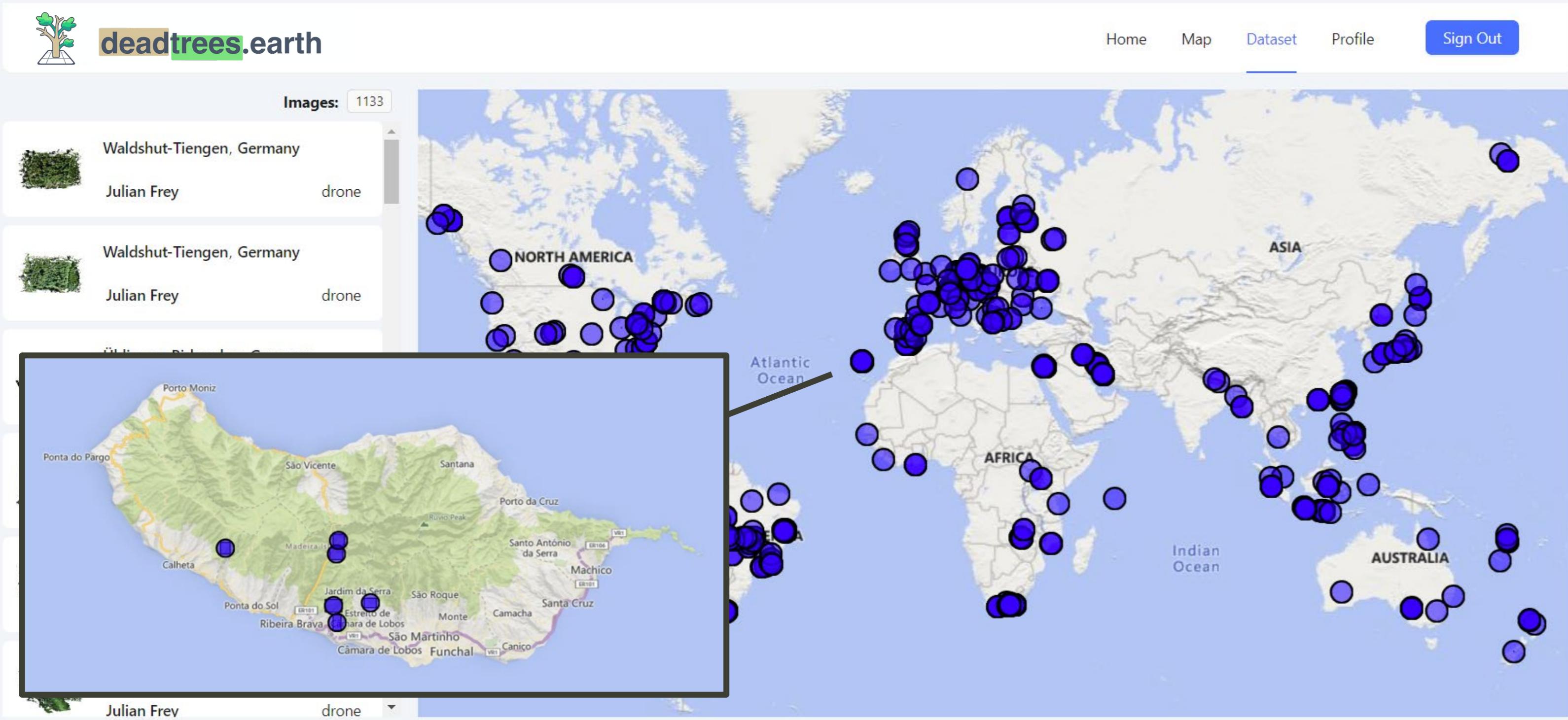
Year 2018 2019 2020 2021

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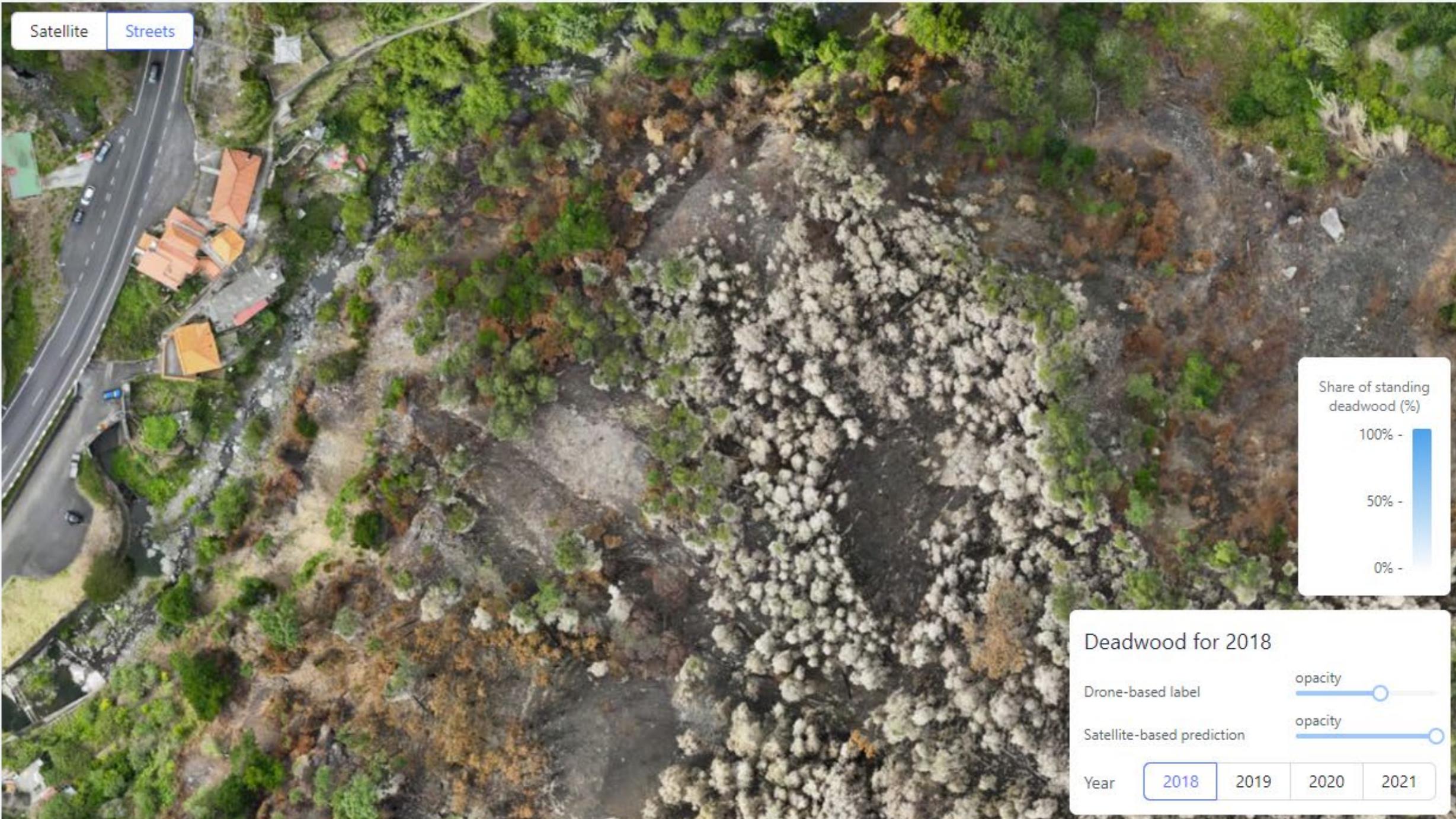


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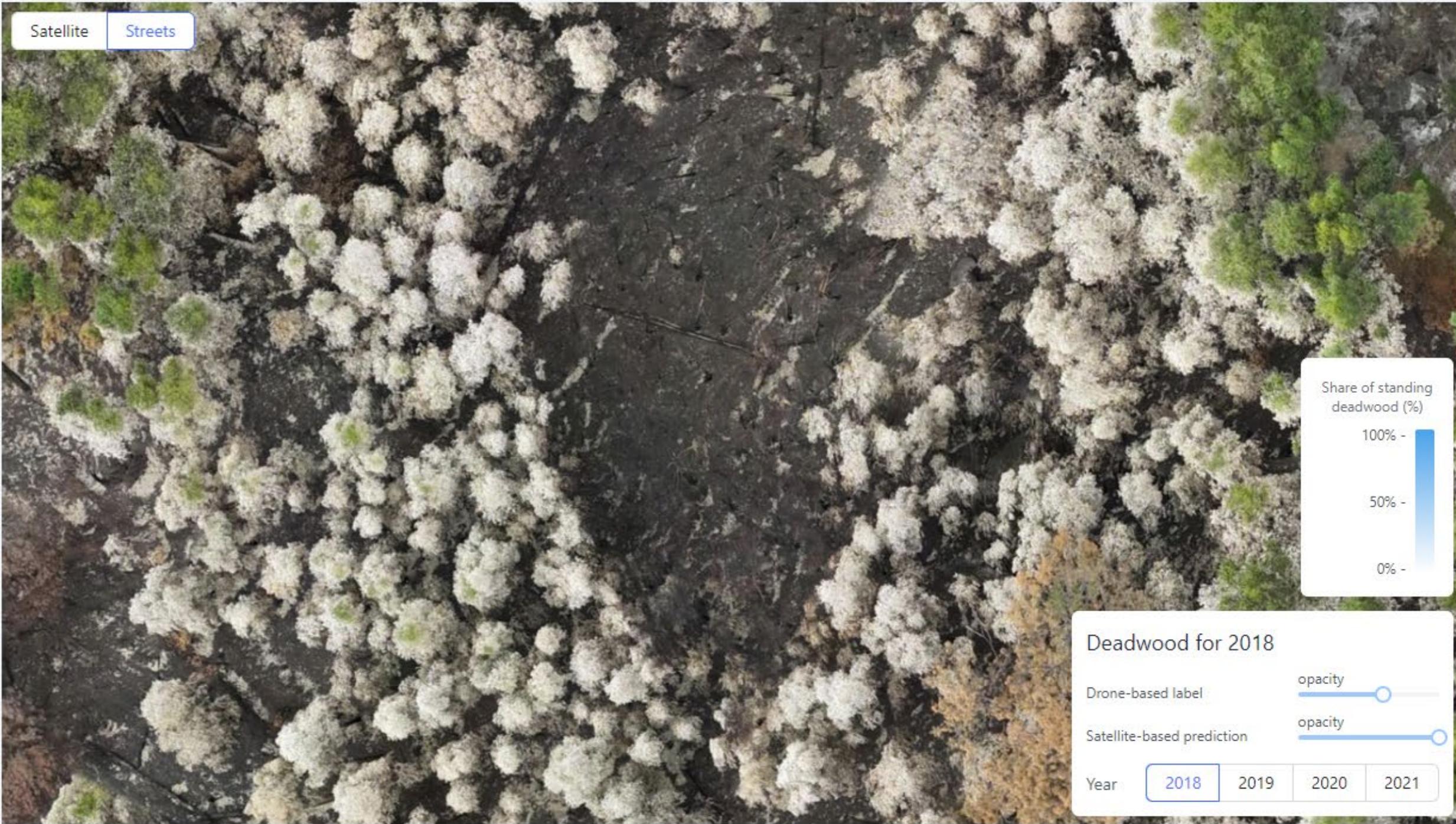
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ID	Date	File	License	Platform	Map	Status
3322	6/9/2024	20240906_mavic3e_madeira_f2_lombo_do_moreilo.tif	CC BY	drone		
3310	11/9/2024	202409111811_009_mavic3e_f6_lombo_do_moreilo.tif	CC BY	drone		
3330	10/9/2024	202409101531_008_madeira_f5_levada_do_norte.tif	CC BY	drone		
2744	6/9/2024	20240906_mavic3e_madeira_f1_clip.tif	CC BY	drone		
3327	6/9/2024	20240906_mavic3e_madeira_f3_jardim_da_serra.tif	CC BY	drone		

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ID	Date	File
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3327	6/9/2024	20240906_mavic3e_madeira_f3_jardim_da_serra.tif

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ID	Date	File
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2744	6/9/2024	20240906_mavic3e_madeira_f1_clip.tif
3327	6/9/2024	20240906_mavic3e_madeira_f3_jardim_da_serra.tif



Upload

You can upload
still in development.
contact us.

Drone or airborne orthomosaic

- Simple RGB imagery, <= 10 cm resolution.
- Any forest or biome type

Optionally: Reference data on dead trees

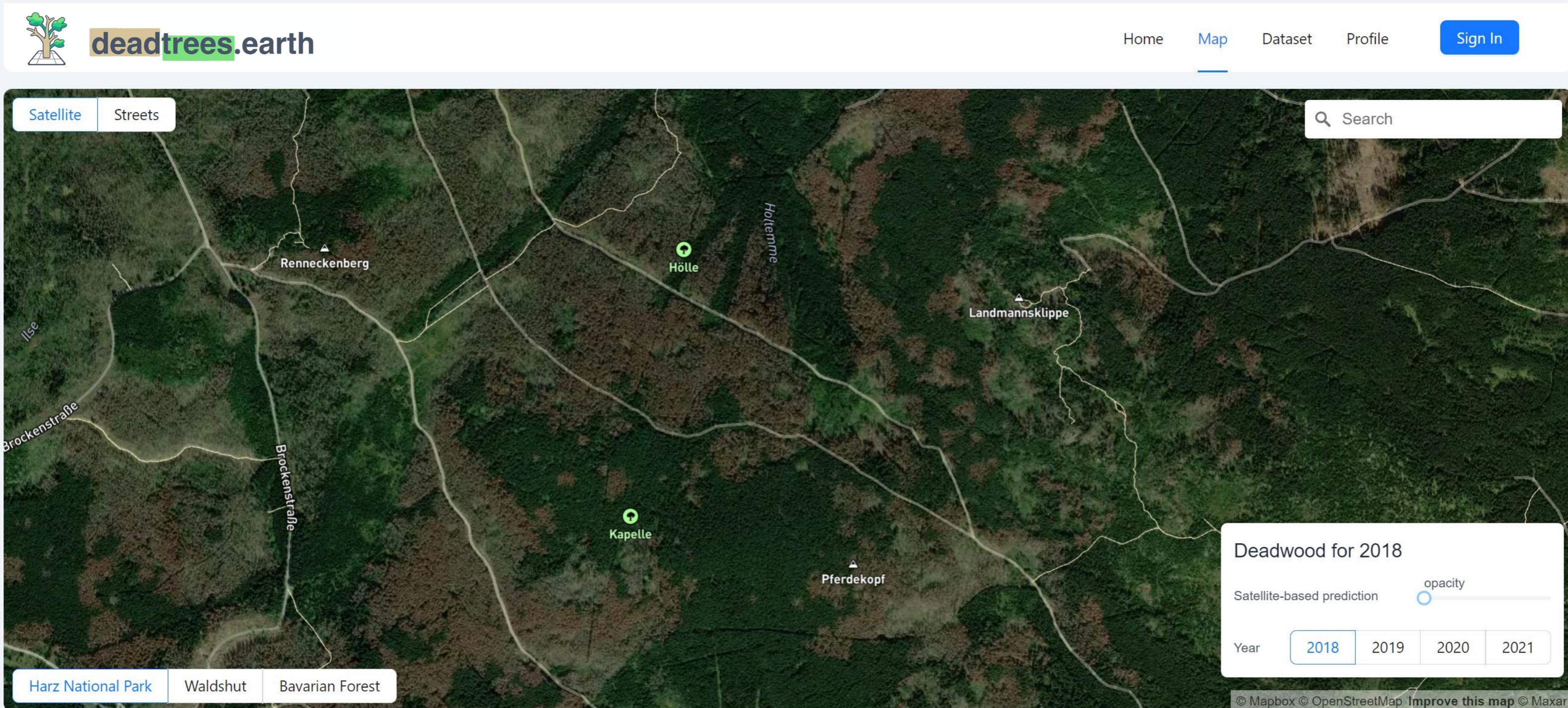
- e.g. polygons or points
- Canopies or individual trees

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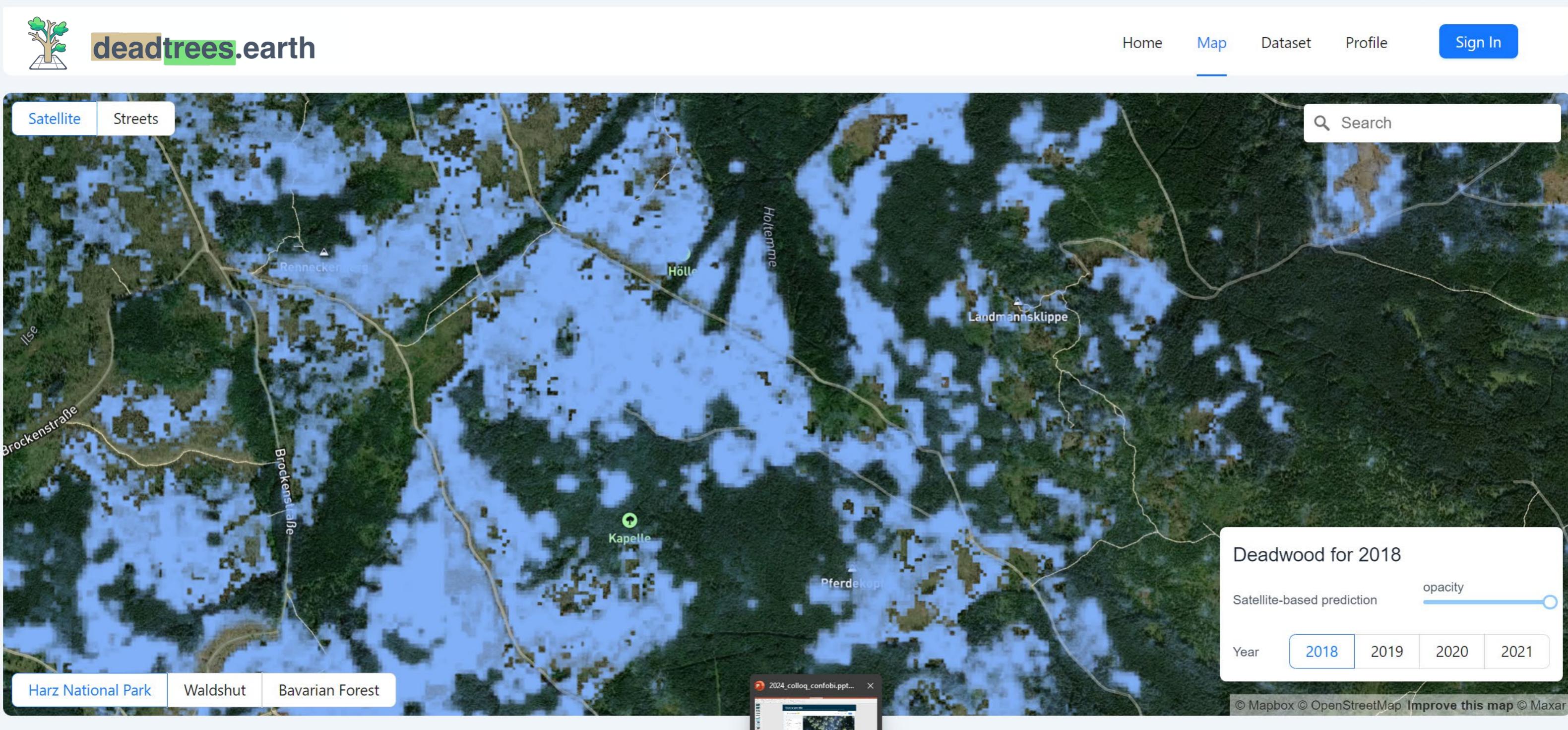
Upload



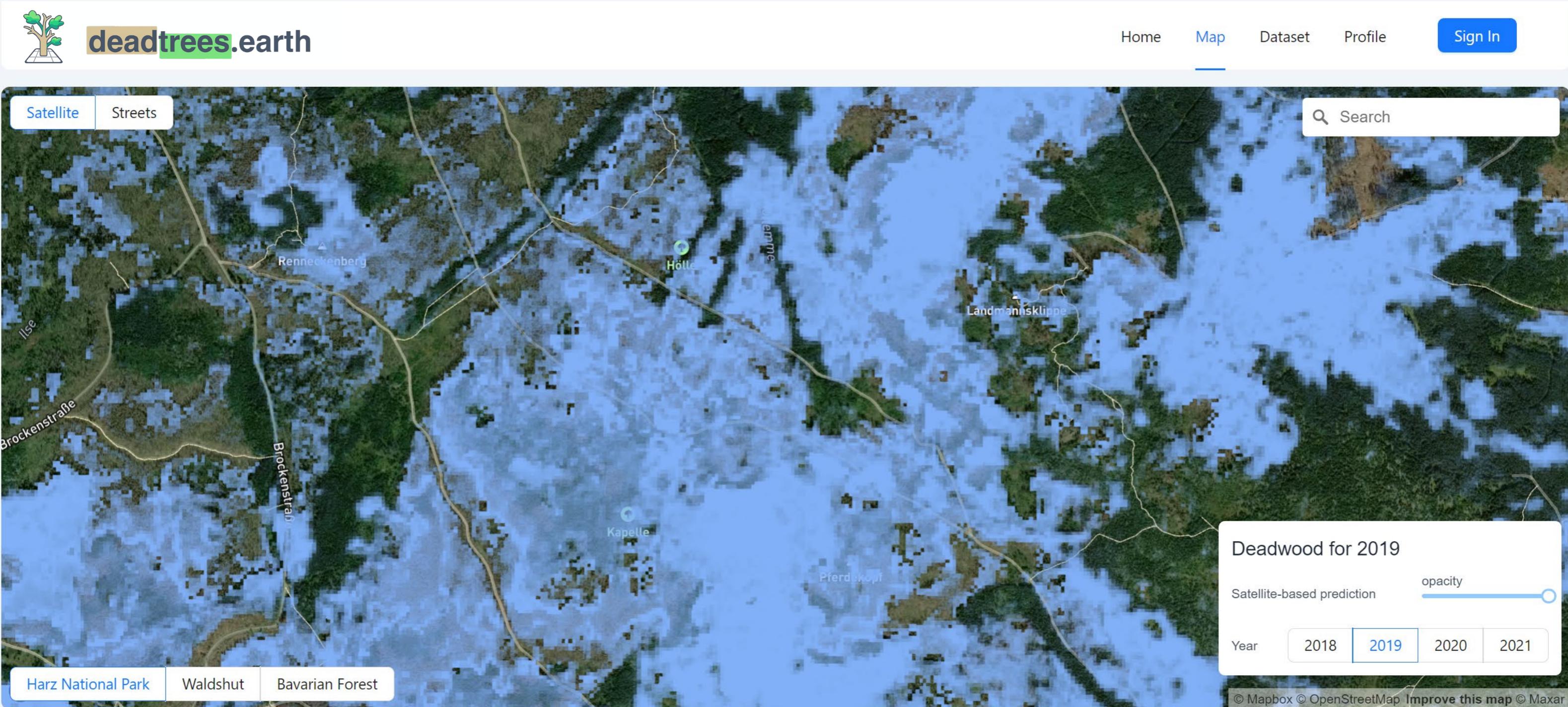
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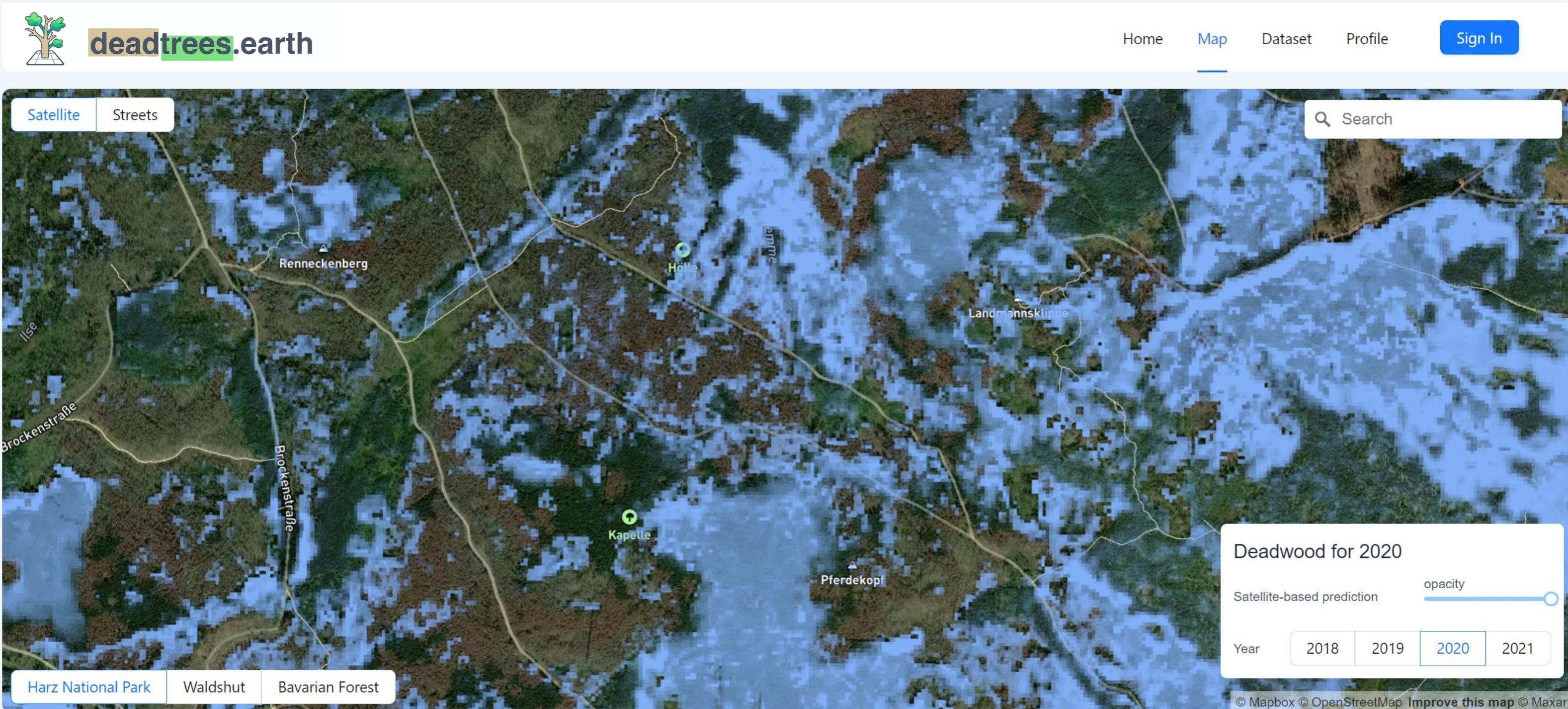
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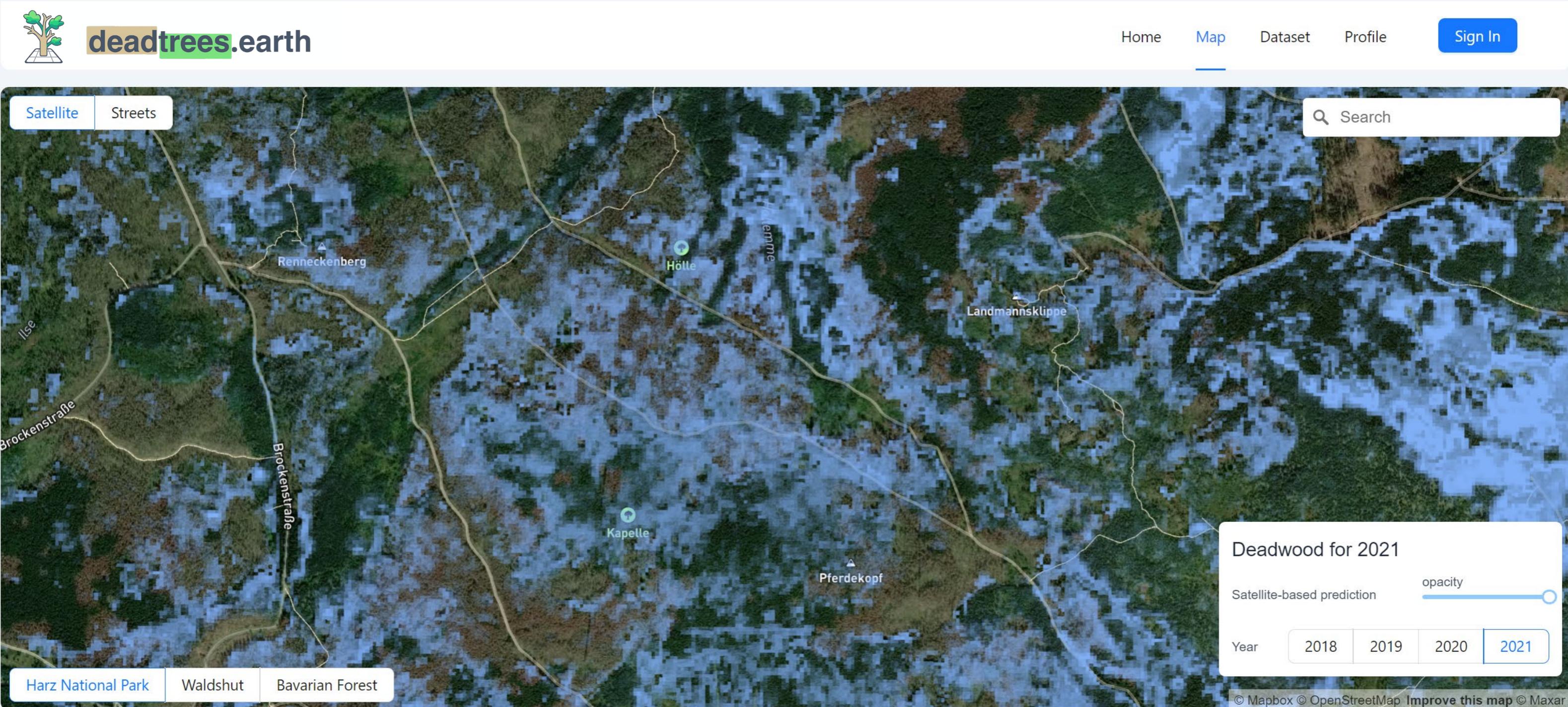
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The platform **deadtrees.earth**

- A **dynamic & open database** for monitoring tree mortality worldwide
- Integrating drone, airborne and satellite data
- Road map:
 - **Q3 2024:** upload & download functionality for aerial imagery and mortality reference data
 - **Q4 2024:** Automated segmentation of dead tree crowns in drone and airborne data
 - **Q1 2025:** Towards **Global satellite-based products** on tree mortality data
 - **Q2 2025:** AI-ready datasets (e.g. for PyTorch)
 - **Q3 2025:** Towards **Attribution** of tree mortality
 - **Q4 2025:** Towards **Forecasting** of tree mortality



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11 – Tree health and mortality mapping

Recent findings

JOURNAL ARTICLE

Large-scale remote sensing reveals that tree mortality in Germany appears to be greater than previously expected

Felix Schiefer , Sebastian Schmidlein, Henrik Hartmann, Florian Schnabel, Teja Kattenborn

Forestry: An International Journal of Forest Research, cpae062,
<https://doi.org/10.1093/forestry/cpae062>

Published: 21 December 2024 Article history ▾

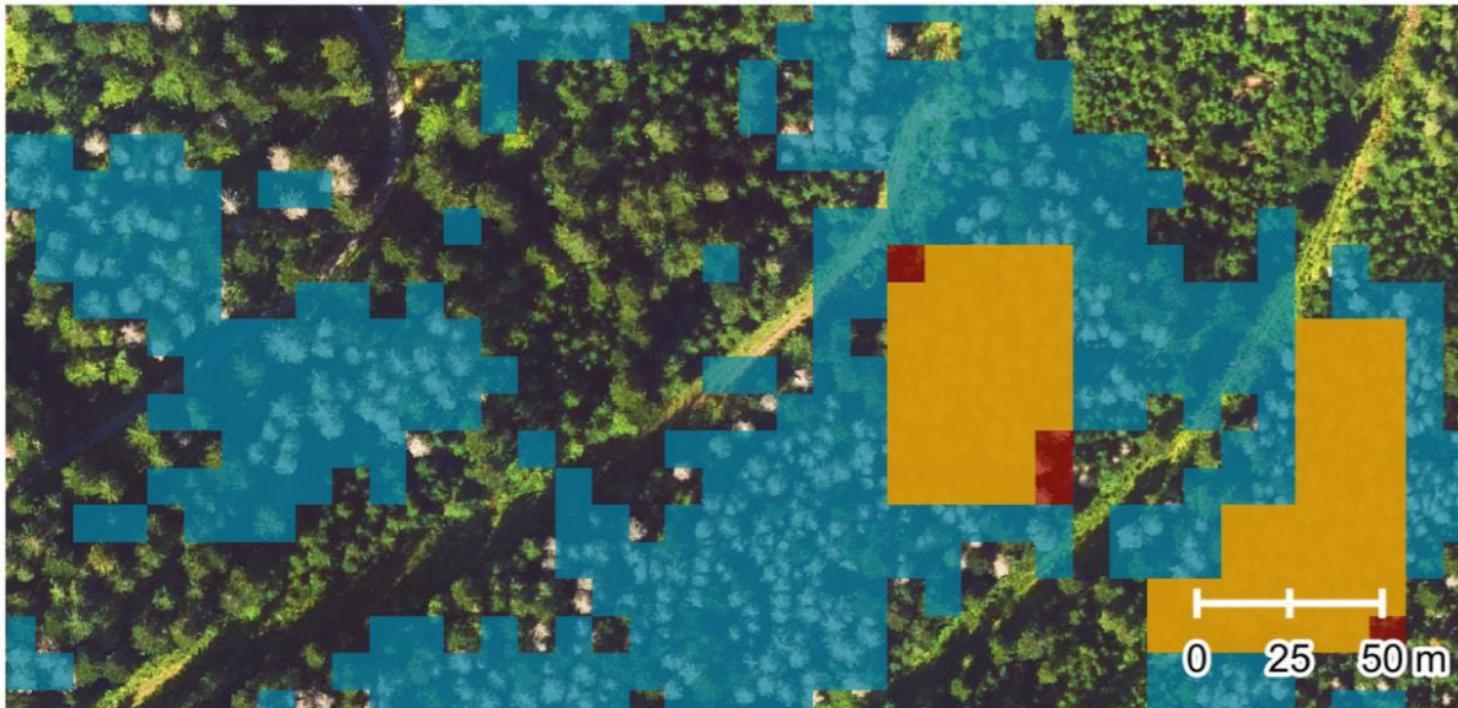
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Abstract

Global warming poses a major threat to forests and events of increased tree mortality are observed globally. Studying tree mortality often relies on local-level observations of dieback while large-scale analyses are lacking. Satellite remote sensing provides the spatial coverage and sufficiently high temporal and spatial resolution needed to investigate tree mortality at landscape-scale. However, adequate reference data for training satellite-based models are scarce. In this study, we employed the first maps of standing deadwood in Germany for the years 2018–2022 with 10 m spatial resolution that were created by using tree mortality observations spotted in hundreds of drone images as the reference. We use these maps to study spatial and temporal patterns of tree mortality in Germany and analyse their biotic and abiotic

- According to satellite-based assessments, tree mortality in Germany underestimated
- Tree mortality assessed across main tree species
- Main drivers revealed
 - Consecutive droughts
 - Early frost

Recent findings

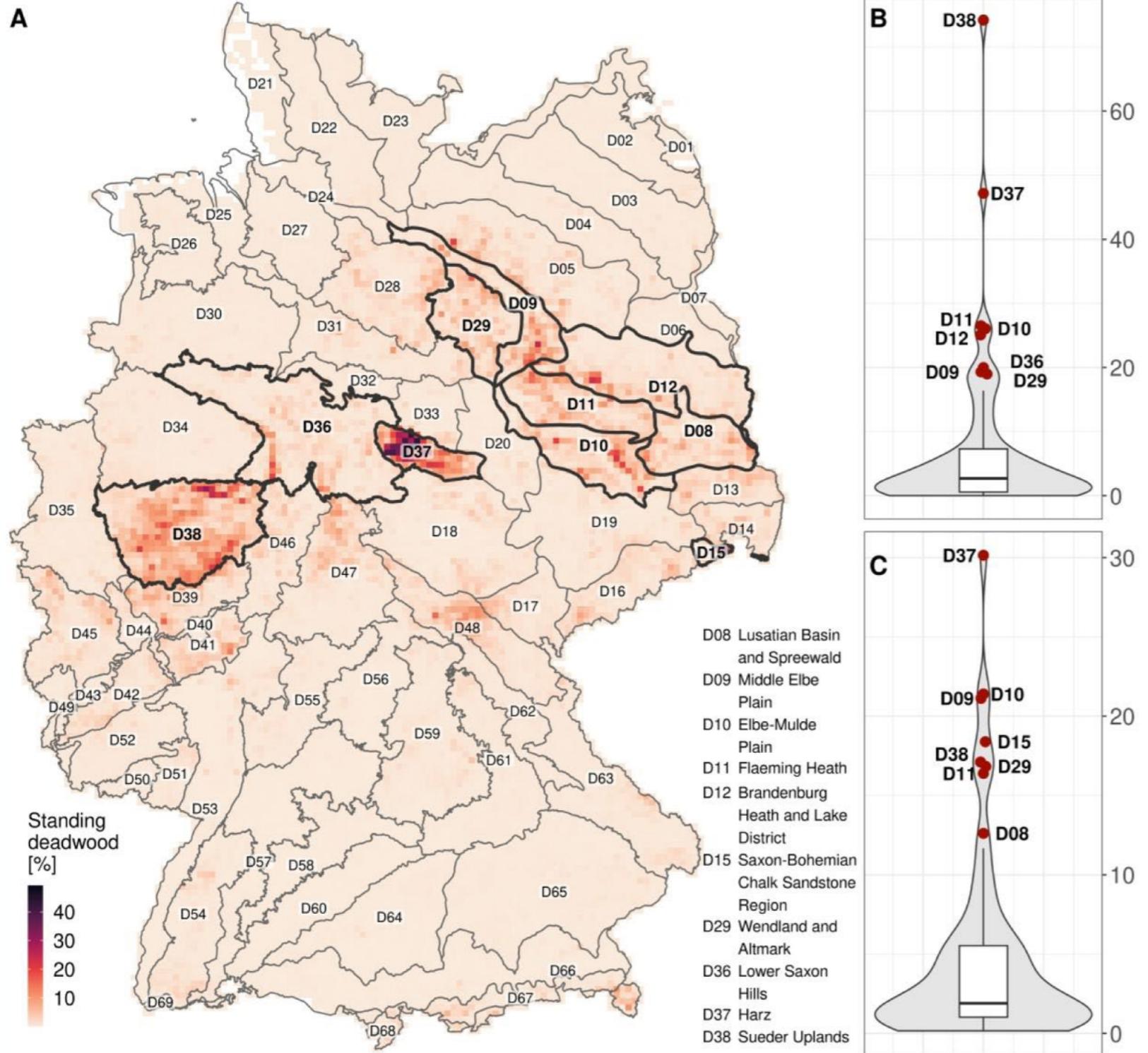


Pixel classified dead in:

Schiefer et al. (2023) Hansen et al. (2013) Both

- Previous remote sensing products (e.g. Global Forest Watch) are not sensitive to tree mortality patterns, as they are
 - Often not stand-replacing
 - Scattered (not always clear patches)

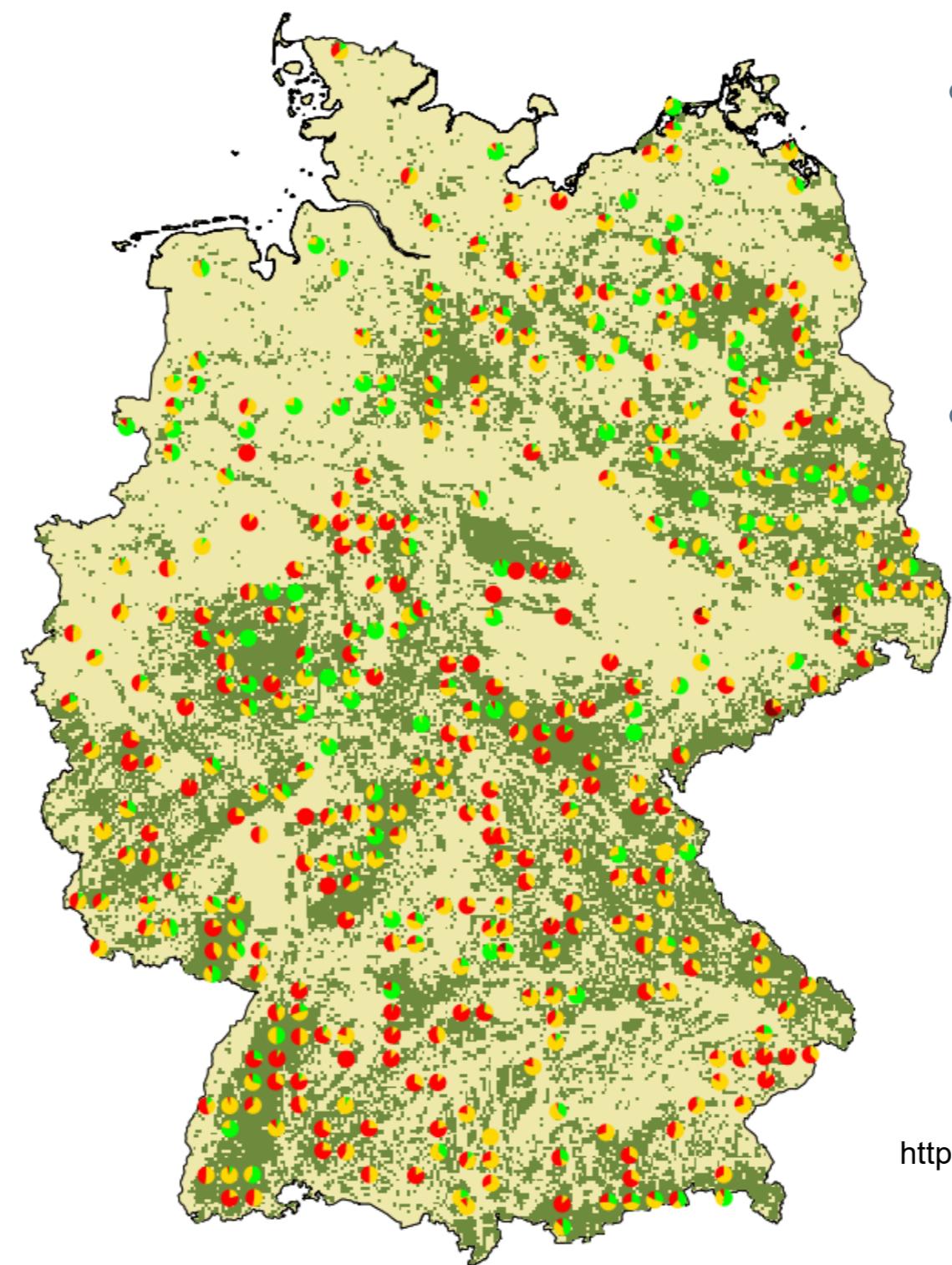
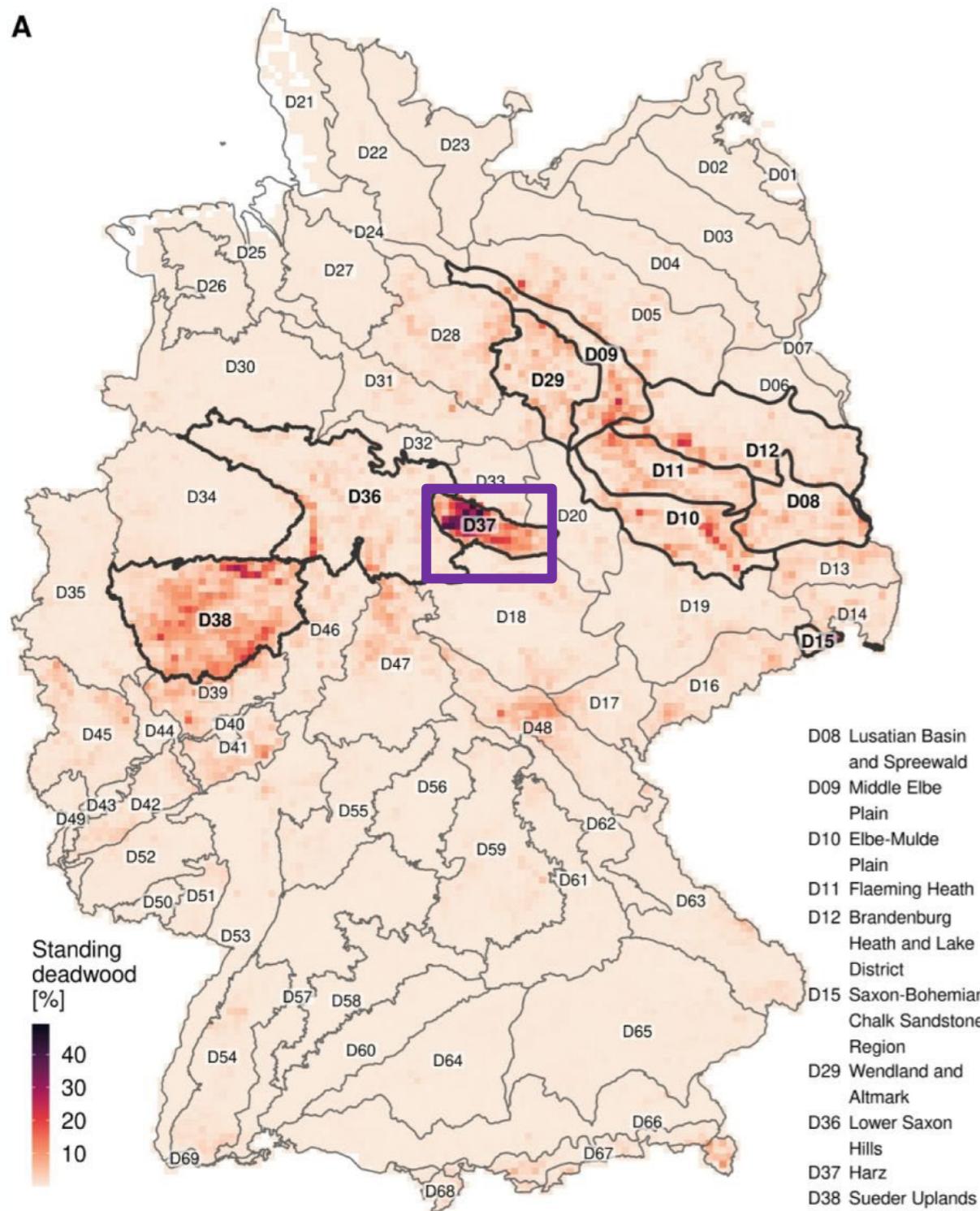
Recent findings



- Tree mortality differs in space
- Hotspots of tree mortality
 - Harz mountains
 - Sueder Upplands
 - Elbe-Mulde Plain

Recent findings

A

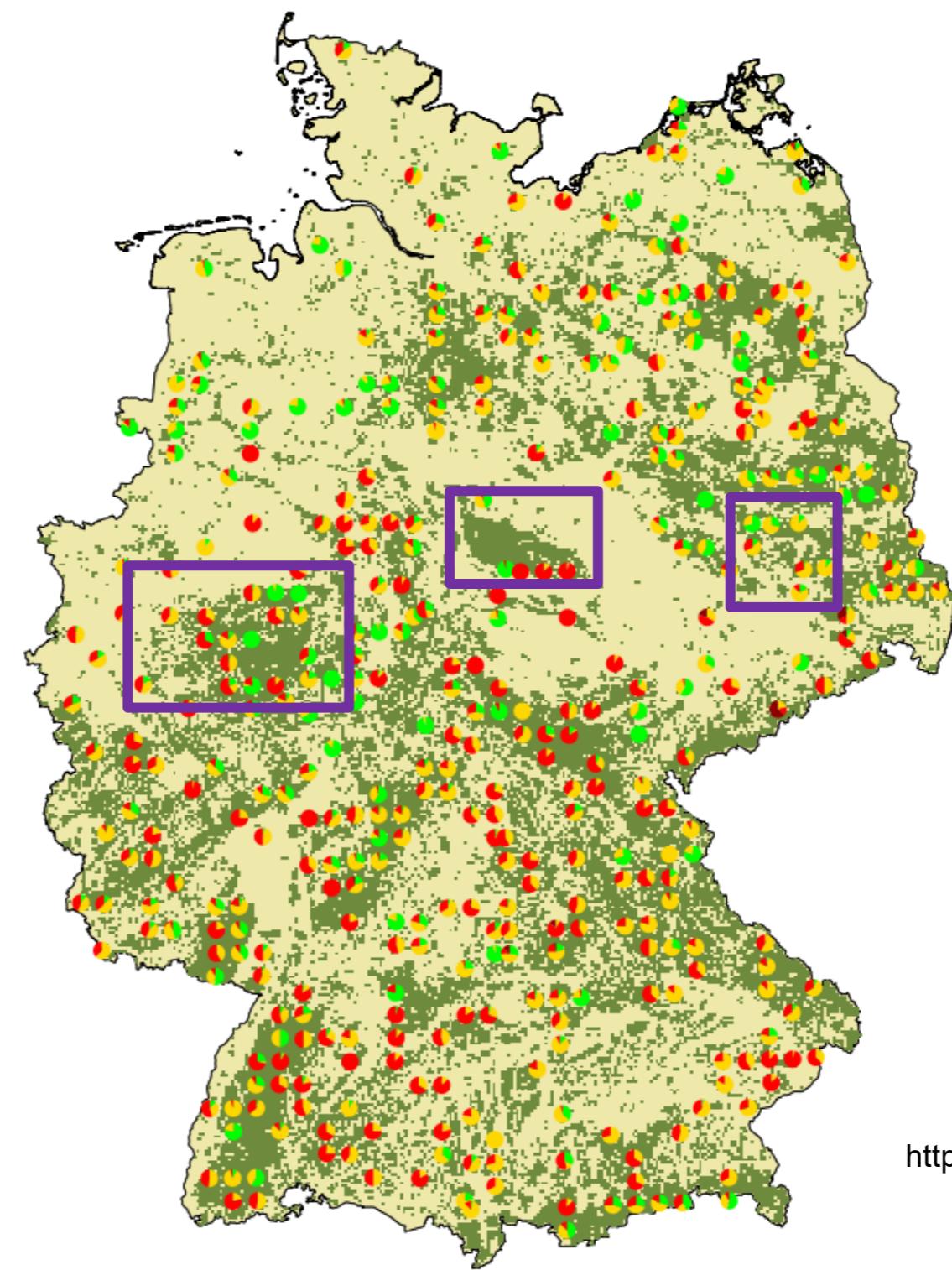
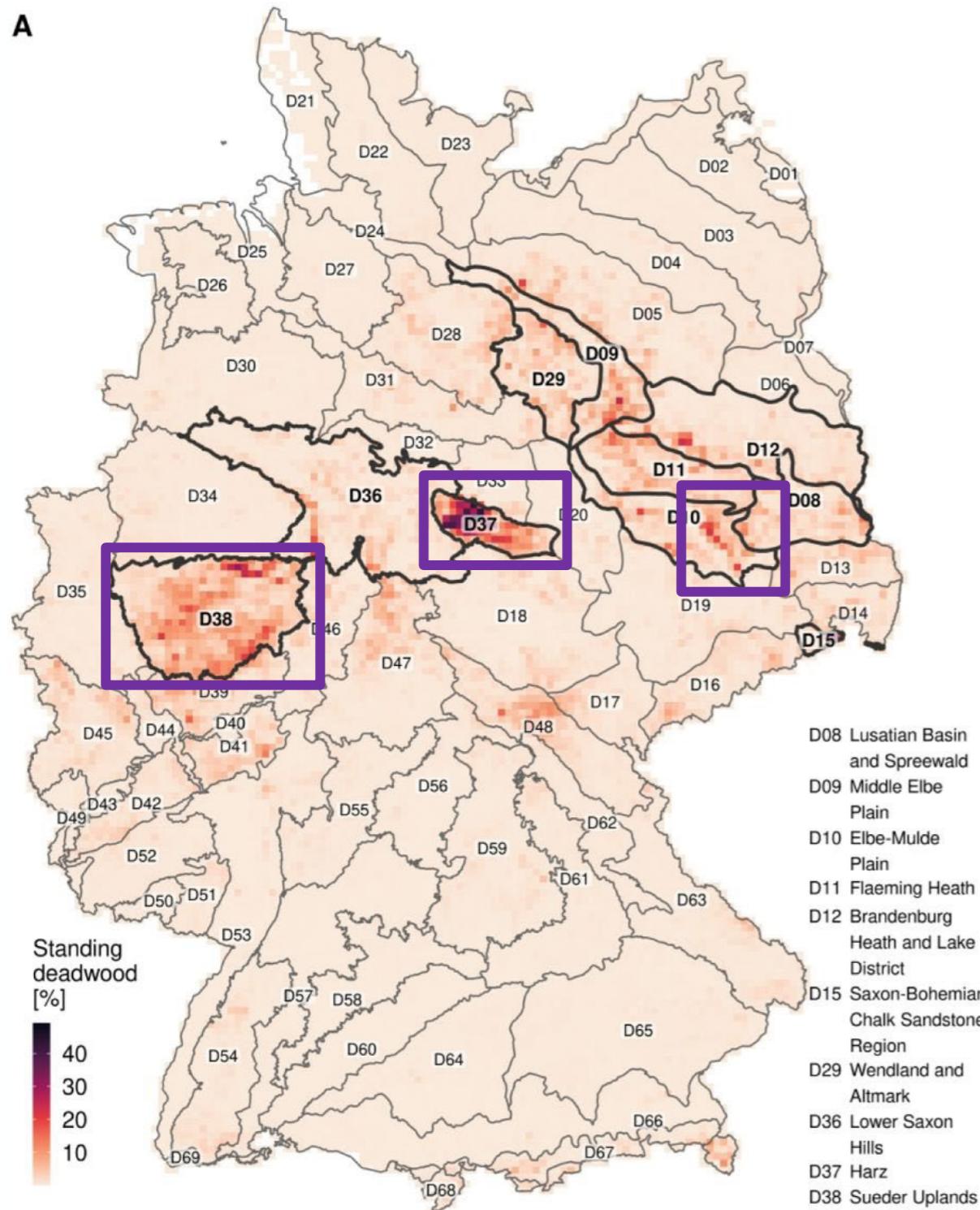


- Forest condition survey does often not cover hot spots
- Therefore, Forest condition survey can largely underestimate forest mortality

https://wo-apps.thuenen.de/apps/wze_maps/

Recent findings

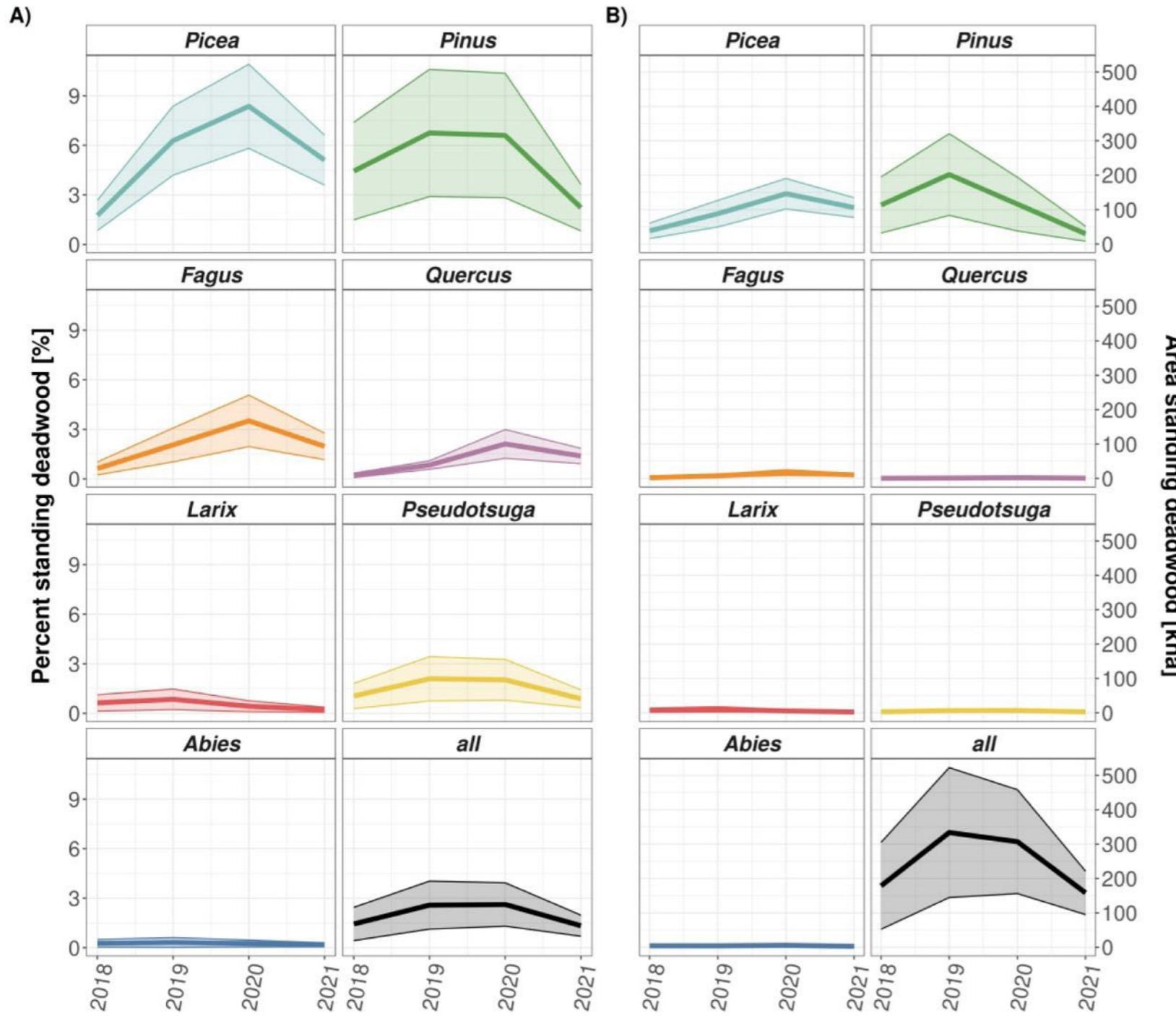
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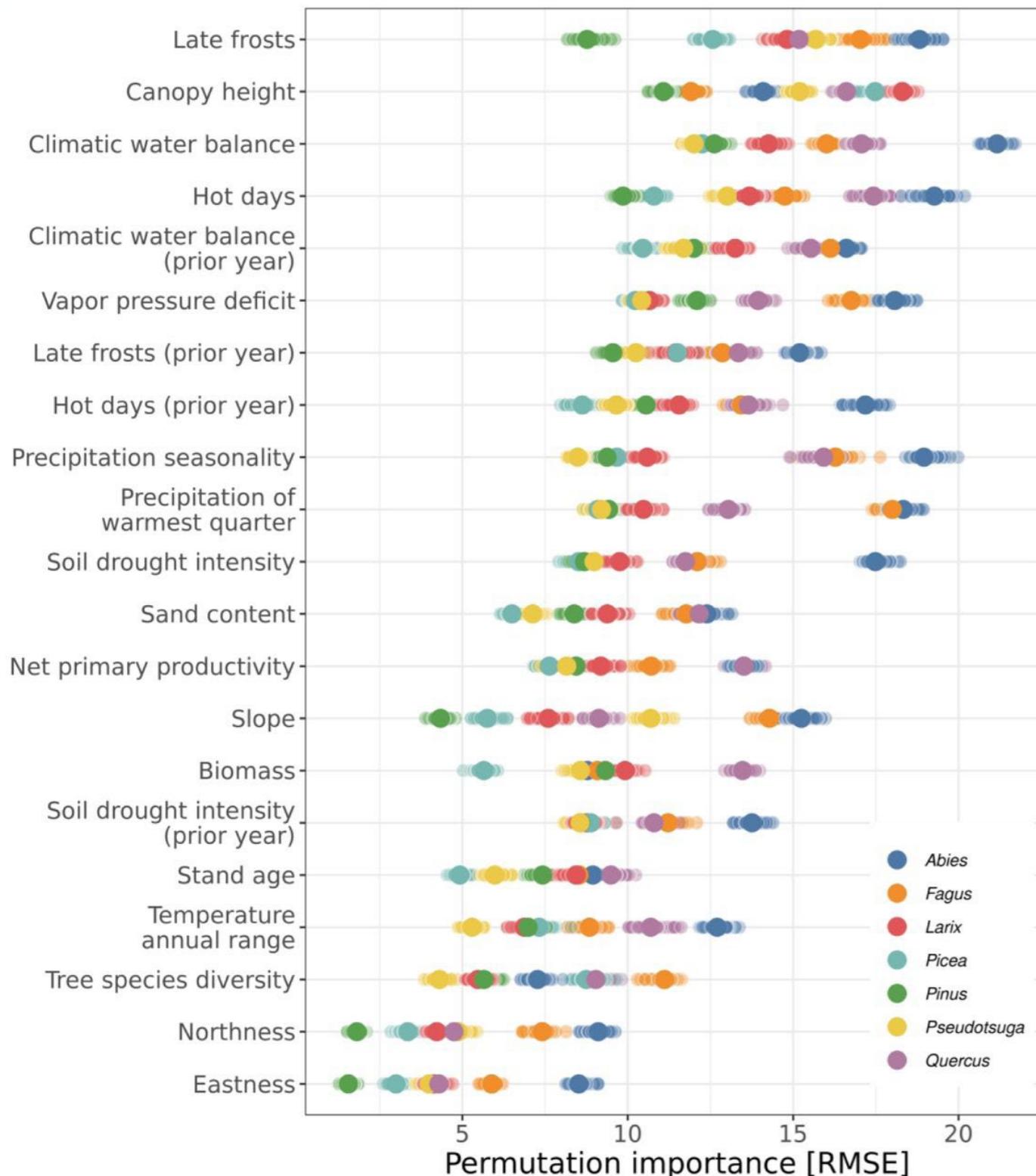
https://wo-apps.thuenen.de/apps/wze_maps/

Recent findings



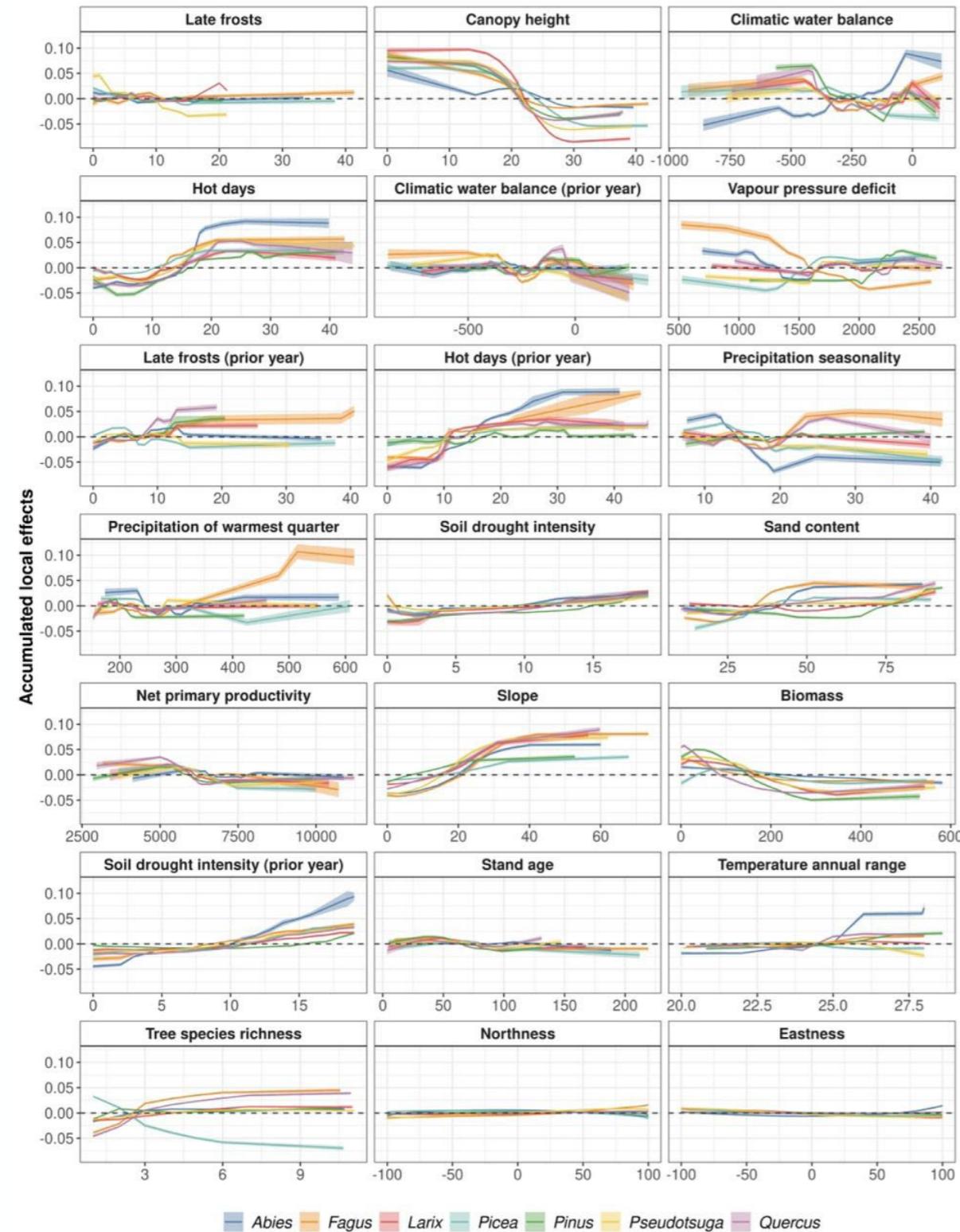
- Tree mortality differs strongly accross species
- Imprint of total share of tree species (tree species selection).

Recent findings



- Drivers per tree species show varying effects
- Some drivers accumulate over multiple years (e.g. reoccurring droughts).
- Tree mortality is an complex interplay of multiple drivers

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Recent findings



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Forest dieback in drinking water protection areas – a hidden threat to water quality

Carolin Winter, Sarina Müller, Teja Kattenborn, Kerstin Stahl, Kathrin Szillat, Markus Weiler, Florian Schnabel

doi: <https://doi.org/10.1101/2024.08.07.606951>

This article is a preprint and has not been certified by peer review [what does this mean?].



Abstract Full Text Info/History Metrics

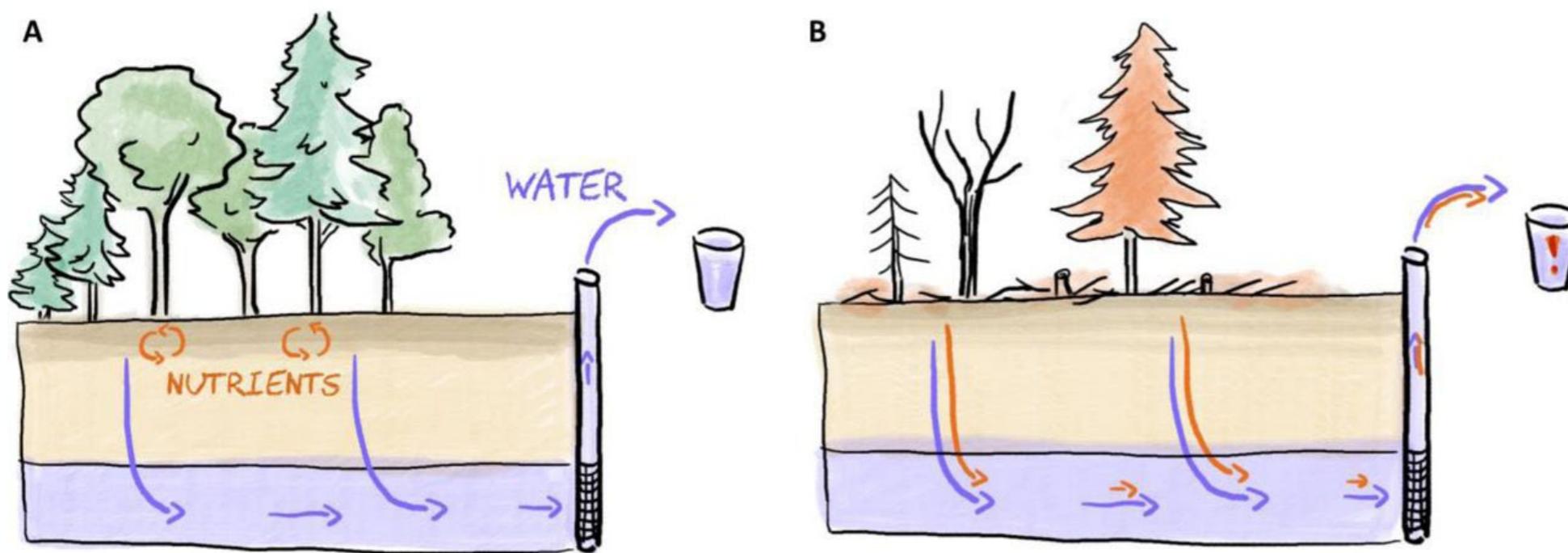
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ABSTRACT

For centuries, forests have been considered a natural safeguard for drinking water quality. We challenge this view in light of the rising frequency of climate extremes, such as droughts coinciding with high temperatures, which have caused unprecedented pulses of forest dieback globally. Drought-induced forest diebacks may jeopardize the crucial role of forests in protecting water quality, potentially even turning forests into sources of contamination. To underscore the critical importance of the topic, here we provide the first comprehensive assessment of forest cover, type, and dieback (assessed as canopy cover loss) across drinking Water Protection Areas (WPAs) in Germany, one of the countries hit most severely by the unprecedented

- Case study on NO₃ concentrations in ground water after forest loss
- Focus on Germany
- Time frame: 2018-today

Recent findings

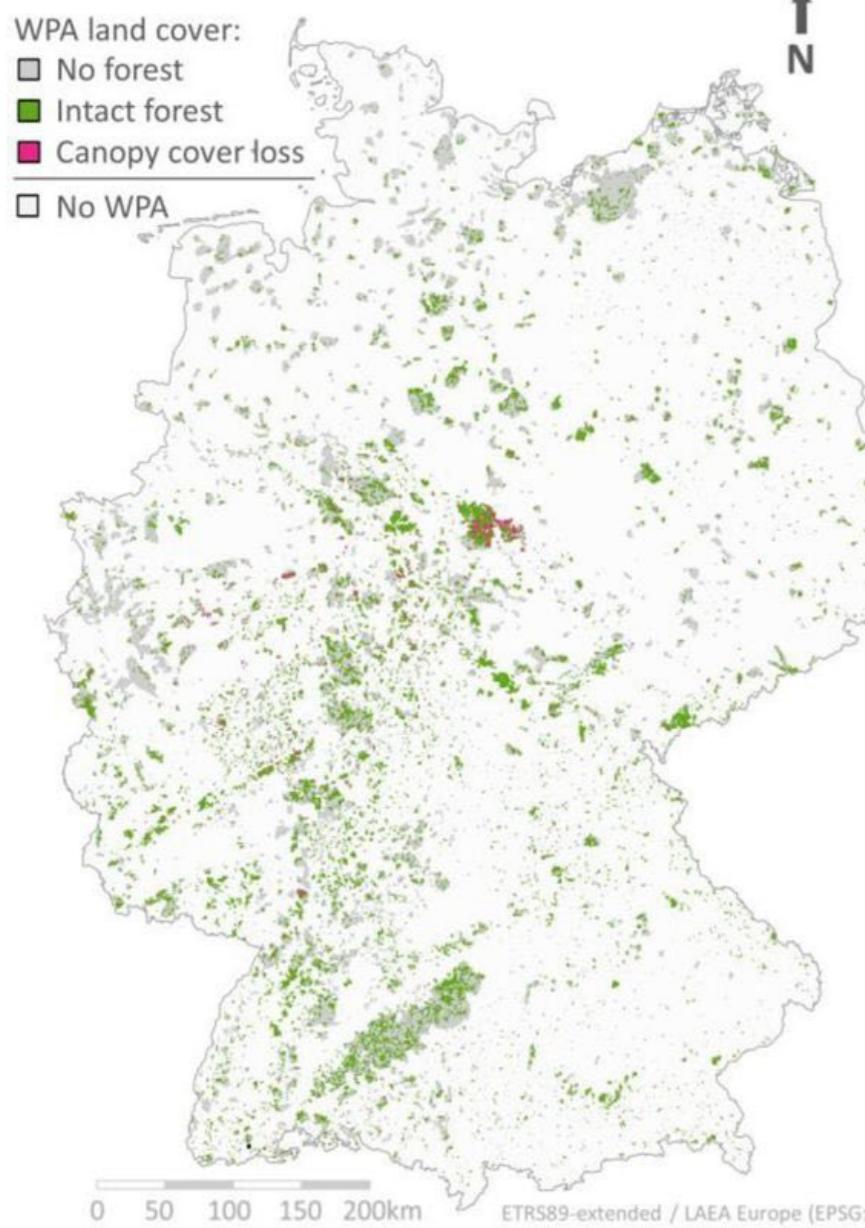


- Intact ecosystems do commonly not leach large amounts of NO₃
- Disturbed ecosystems can leach large quantities of NO₃ into the ground water

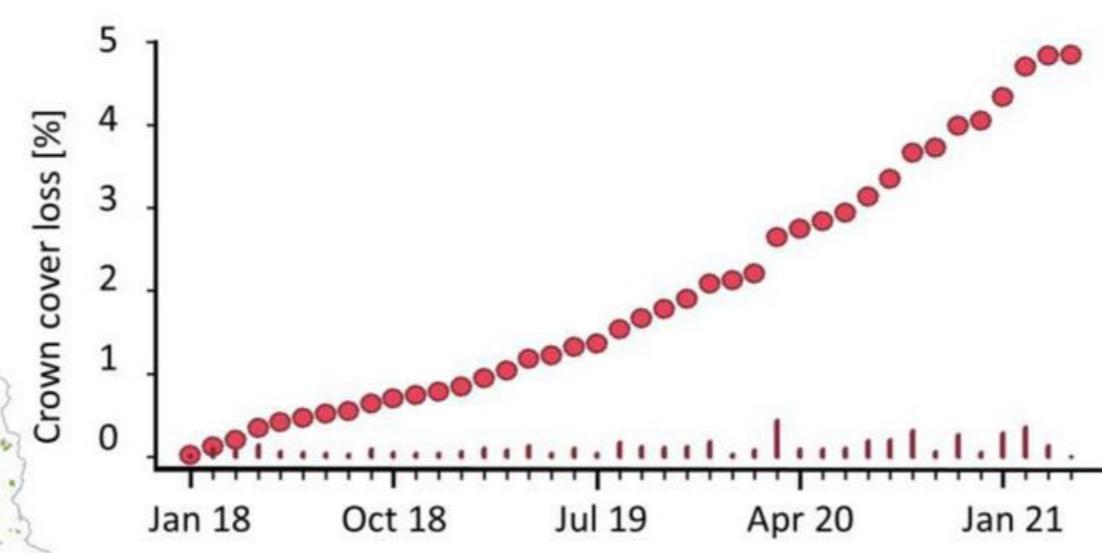
Schematic view on the role of forests as a safeguard of groundwater quality and related drinking water quality.

Recent findings

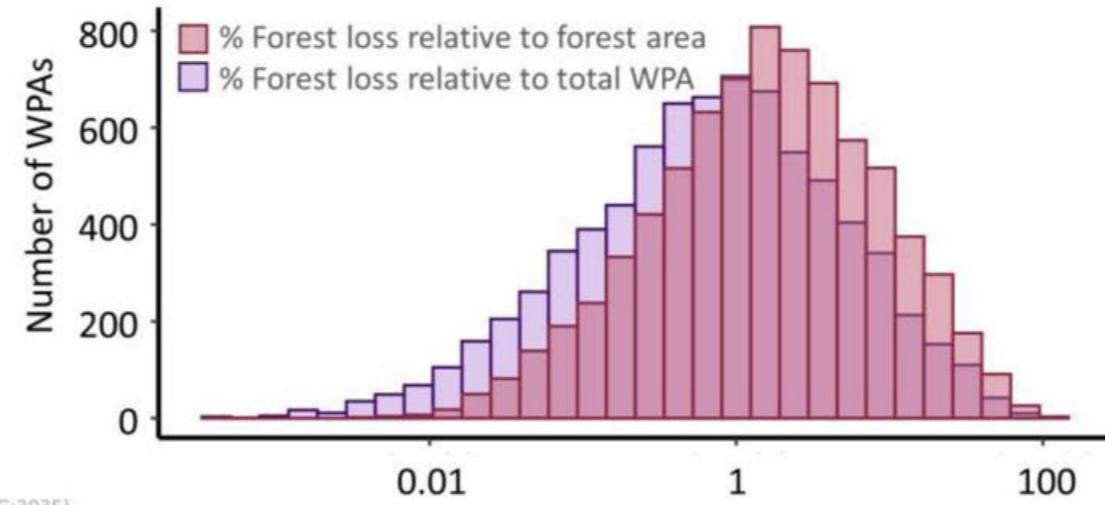
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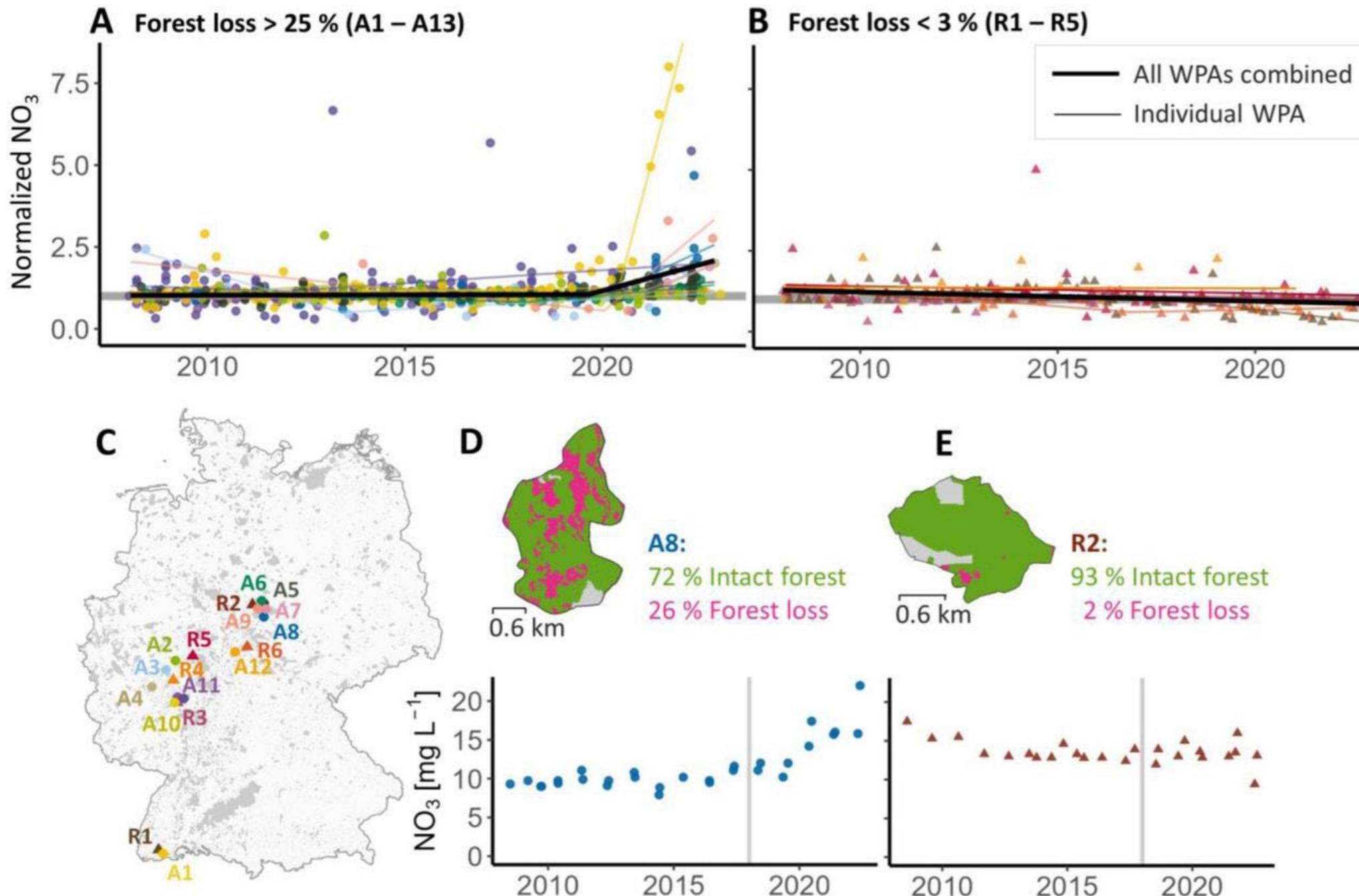
C



- Most water protection areas (WPA) are forested
- Crown cover loss increased in WPA since 2018

Canopy cover loss across German WPAs, with A) a map of intact forest and canopy cover loss between January 2018 and April 2020, B) the temporal resolution of canopy cover loss as absolute (lines) and cumulative (dots) percentage, and C) the percentage of canopy cover loss relative to the forested area of a WPA and relative to the entire area of the WPA.

Recent findings



- Increased NO₃ in areas with forest loss > 25 %
- No increase in areas where forest loss < 3 %

Normalized groundwater nitrate concentrations between 2008 and 2022 in forested sites that were A) affected by <25% canopy cover loss or B) reference sites with <3% canopy cover loss. Thin colored lines depict the area-specific linear models with or without a breakpoint. Thick black lines depict the linear models across all sites with >25% or <3% canopy cover loss. Panel C) depicts a map with the locations of the canopy cover loss and reference sites. Panels D) and E) show two exemplary sites, one with substantial canopy cover loss and a significant increase in nitrate concentration after 2018 (D), and the other a reference site with marginal canopy cover loss and no significant breakpoint after 2018 (E).