



écoinformatique FAIR\* pour et par les communautés

**Pl@nt**AgroEco

ONE FORE

### **Cooperative learning for biodiversity monitoring:**

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https://www.mambo-project.eu



A citizen science platform that uses machine learning to help people identify plants with their mobile phones





200+ countries - 53K species 7M users ac. - 20 M users / Y. 200k-2M identifications / Day 1.2B identification queries



Nature, walks



**Personal Usage** 



Phytotherapy



#### **Professional Usage**



Agro-ecology



Education, animation



Natural Areas Management



Tourism

Trade

### Key concept of Pl@ntNet: Cooperative Learning Online version



### Key concept of Pl@ntNet: Cooperative Learning Off-line version

Identify plants without connection + resynchronisation



### Contribution

Users can contribute their observations



## Revision

Users can tag, vote, or revise other contributions (image, obs. level) of the members of the PN community







21:29 🔶 🗹 🝭 🔹		😰 🐓 .ii 57% 🛢
← 4 Nom(s) con <sub>Français</sub>	nmun	(s)
<u>Larix decidua Mill.</u>		
Mélèze commun	<b>.</b>	12 Votes
Mélèze d'Europe	<b>*</b> =	8 Votes
Pin de Briançon	<b>≜</b> ≡	6 Votes
Pomme de pins	<b>.</b> ≡	1 Vote
Ajouter un nom	<b>e</b> <sup>-</sup>	1 Vote
Nom commun		

## **Cooperative learning**

The weight of a user in the decision process depends on his estimated expertise



Most probable species  $y = \arg \max_{j} \hat{\eta}_{j}(x)$ 

Validation decision (valid  $\rightarrow$  used by AI)

$$\hat{\eta_y}(x) > \theta$$

Lefort, T., et al., 2024. Cooperative learning of PI@ ntNet's Artificial Intelligence algorithm: how does it work and how can we improve it?. arXiv preprint arXiv:2406.03356. In review in MEE. <u>https://doi.org/10.48550/arXiv.2406.03356</u>

Database



#### 1.2 bil. observations









1 raw observation:

- 1 or more images
- 🛛 organ tags 🔊 🔻 🍎 📕 🌪 😻
- AI-based species prob.
- (geo-localization)

#### 1 valid shared observation

- author name
- aligned species name
- human validation > score
- cc-by-sa license
- High GPS precision
- Region-based filtering matching
- Auto. visual quality inference



	DataStore	Observations -	Queries -	Users -	Projects -	RealTime -	Misc	
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Cakile maritima Scop.

Pierre Bonnet - replication: running prod

0.071830



0.004840

Erodium aethiopicum (Lam.) Brumh. & Thell.

- **Top-4 data provider to GBIF** (world's largest infrastructure for biodiversity data)
- Valid observations + trusted gueries identified by the AI (AI score>0.9)
- Additional quality filters: \* 2 AI classifier (potted & cultivated plants removal),
  - \* Region-based filtering (Kew POWO)

#### **1. Foreground class.**













noplantphoto amphibian noplantphoto bird

noplantphoto fish

noplantphoto mammal noplantphoto man

noplantphoto invertebrate noplantphoto reptile

noplantphoto mushroom



noplantphoto food



noplantphoto\_landscape noplantphoto d rawing bw



noplantphoto microscopy



noplantphoto humanmade noplantphoto\_digital\_docs





plantphoto\_herbarium plantphoto\_root plantphoto seed



noplantphoto

drawing col

plantphoto habit plantphoto bark plantphoto branch

plantphoto bud

plantphoto flower plantphoto fruit plantphoto leaf

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#### 2. Context class.







plantphoto bw



plantphoto montage



plantphoto with human body presence



plantphoto intothewild



plantphoto uniform background macro dissection

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```
],
 "foreground": [
         "name": "leaf",
         "score": 0.6499367952346802
     },
         "name": "man",
         "score": 0.08426956832408905
     },
"context": [
        "name": "human",
        "score": 0.8416258096694946
    },
        "name": "anthropized",
        "score": 0.09990807622671128
    },
```



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https://doi.org/10.15468/mma2ec



ELSEVIER

## Other collaborative tools









# Thank you









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GENCI





