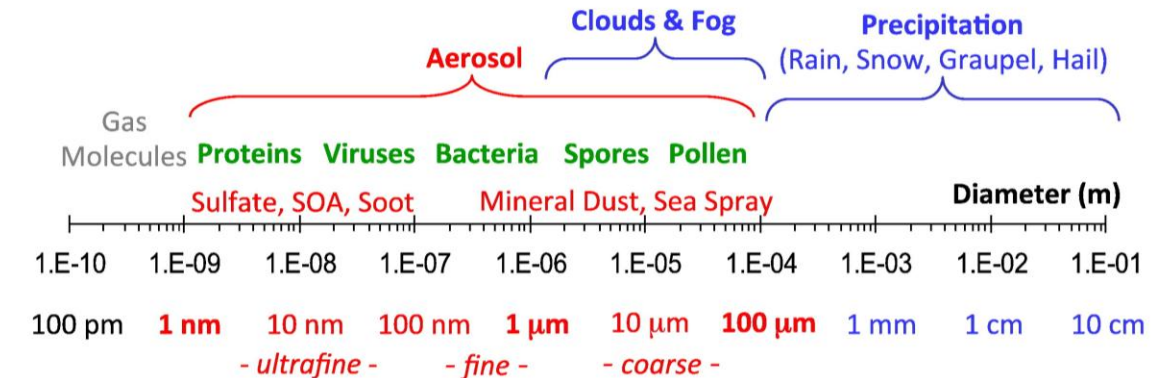
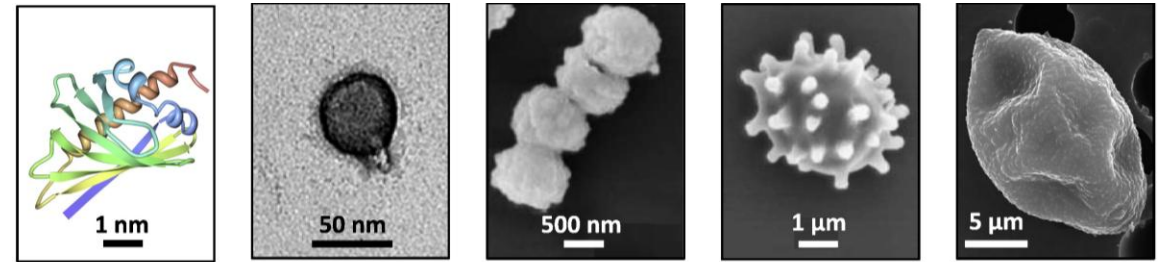
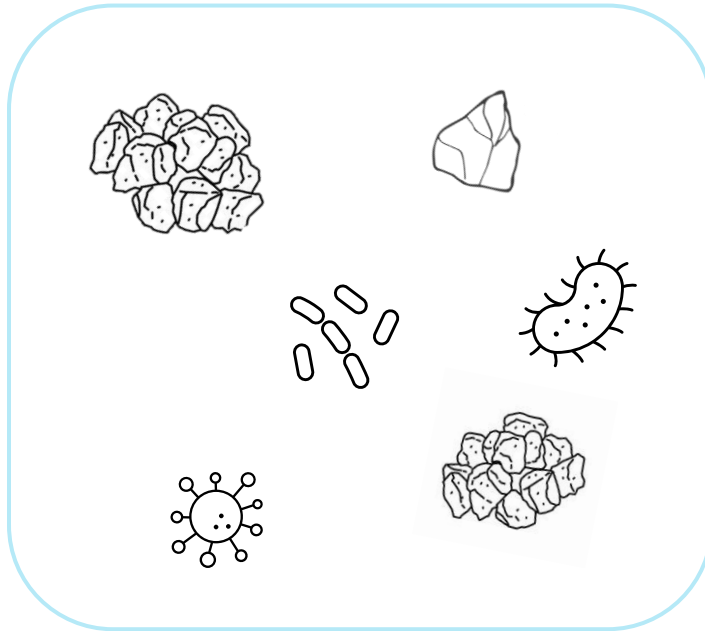
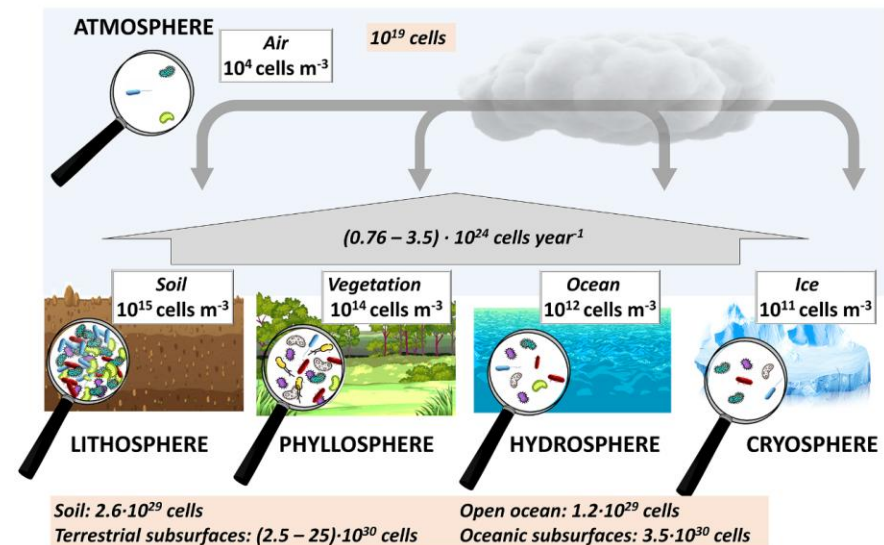


Contexte scientifique

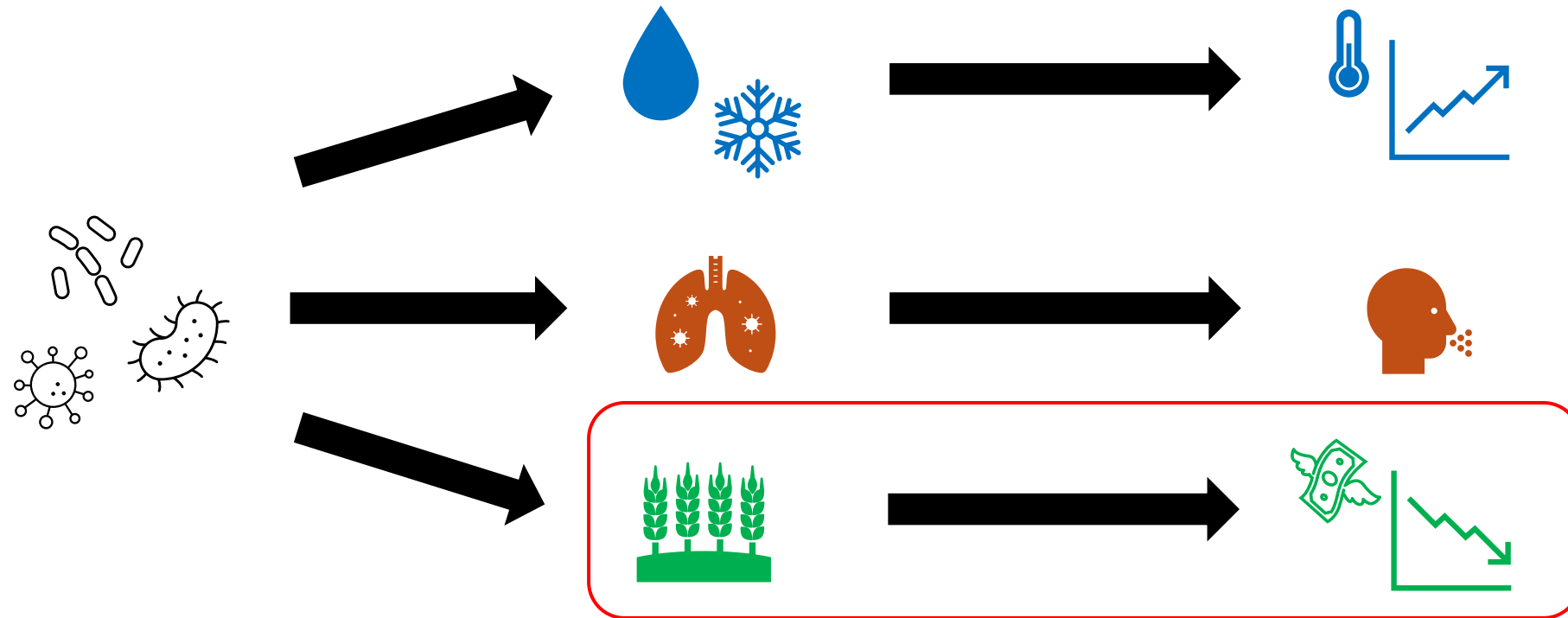


Source: Fröhlich-Nowoisky et al. (2016)



Source: Ervens et al. (2025)

Pourquoi étudie-t-on les bioaérosols ?

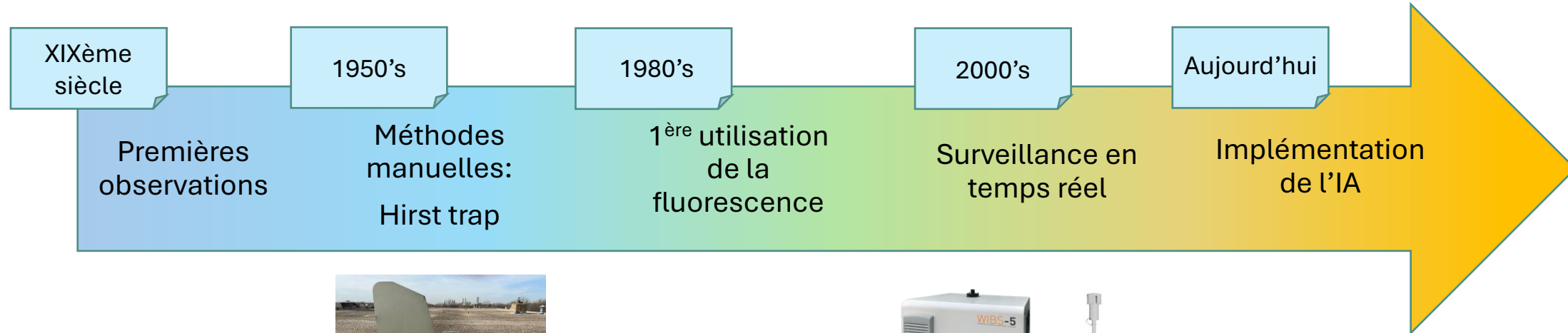


Rouille brune de blé (*Puccinia triticina*)



(haut) *Alternaria linariae*
(bas) Symptômes sur plant de tomate

Mesures In Situ: historique des méthodes d'observation



Carnelley et al., 1887
Pasteur 1862



Hirst, 1952

Pinnick et al., 1995



Erb et al., 2024

WIBS (Healy et al., 2014)
BAA500 (Oteros et al., 2020)
Rapid-E (Šaulienė et al., 2019)
Poleno (Sauvageat et al., 2020)

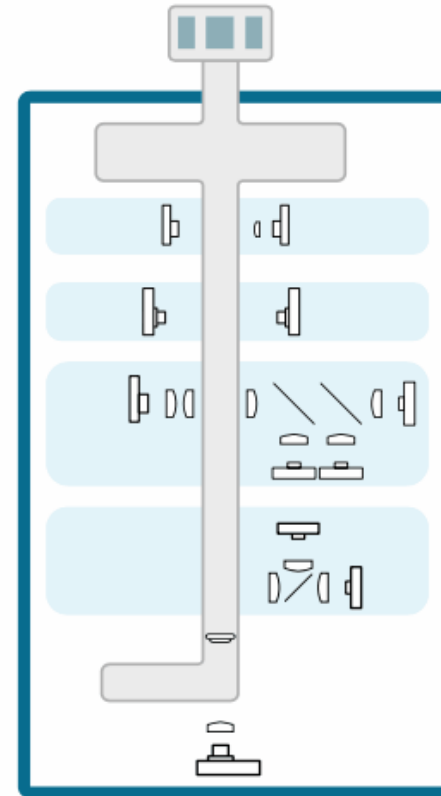
Mesures In Situ : Swisens Poleno Jupiter



Grenoble, 3 février 2025

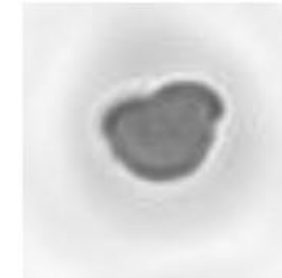
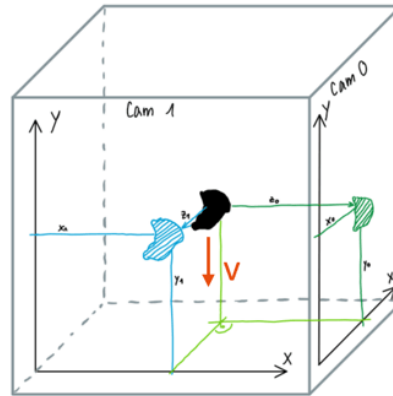


Puy-de-Dôme, 6 février 2025





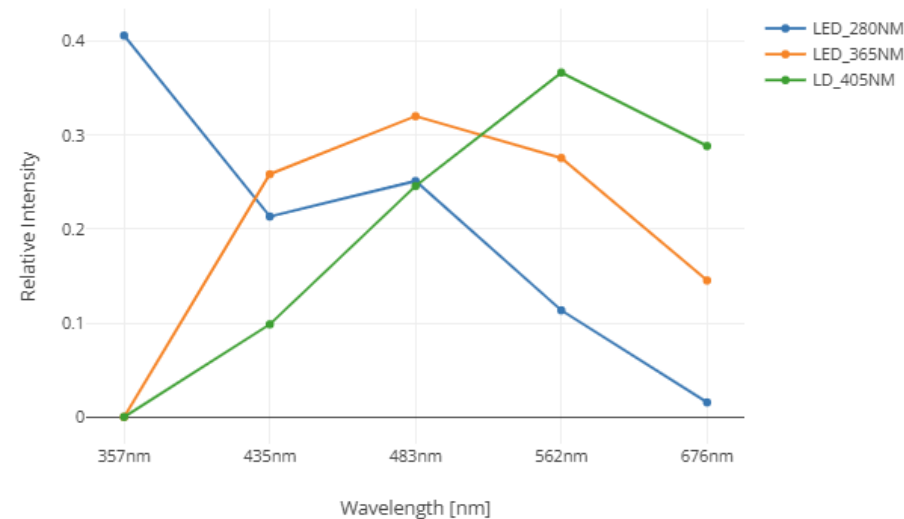
Holographie:



Grand volume de données à traiter

Fluorescence:

Relative Spectra

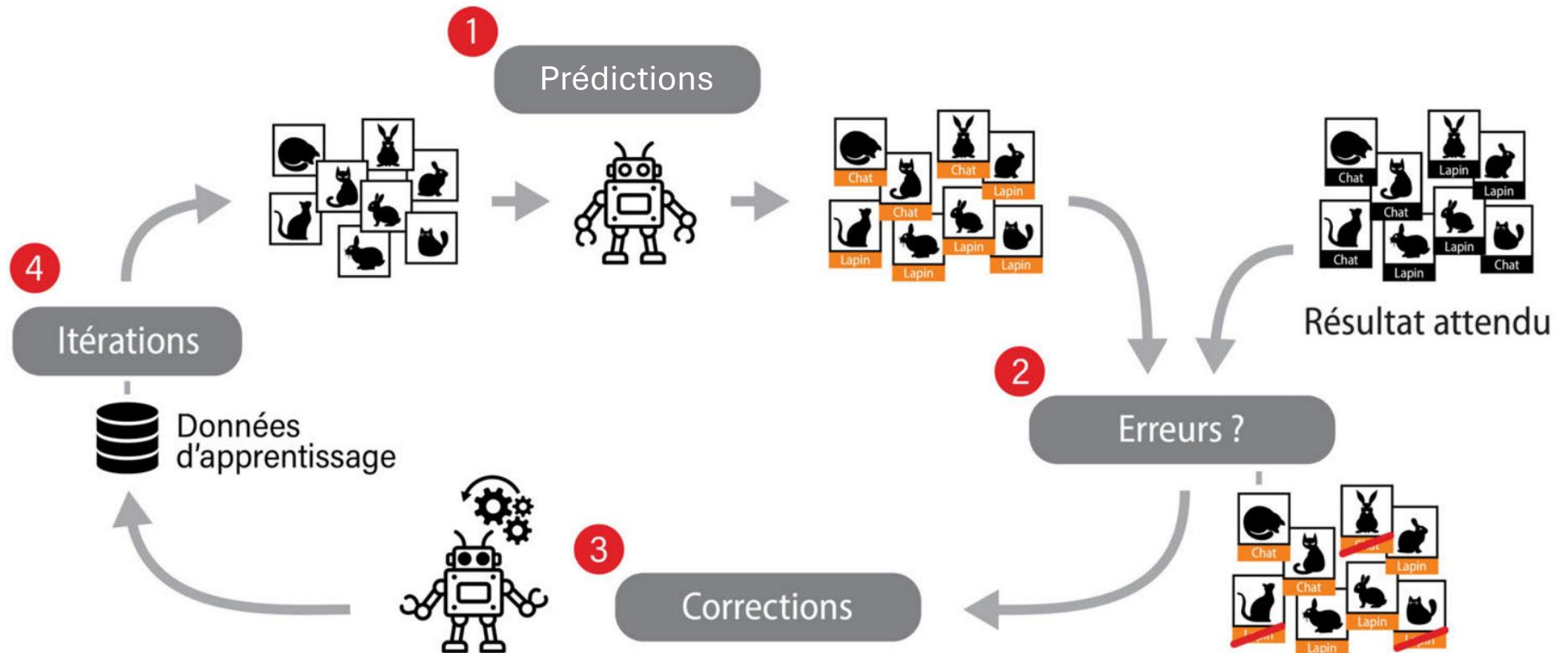


→ Le Machine Learning peut aider à traiter ce problème

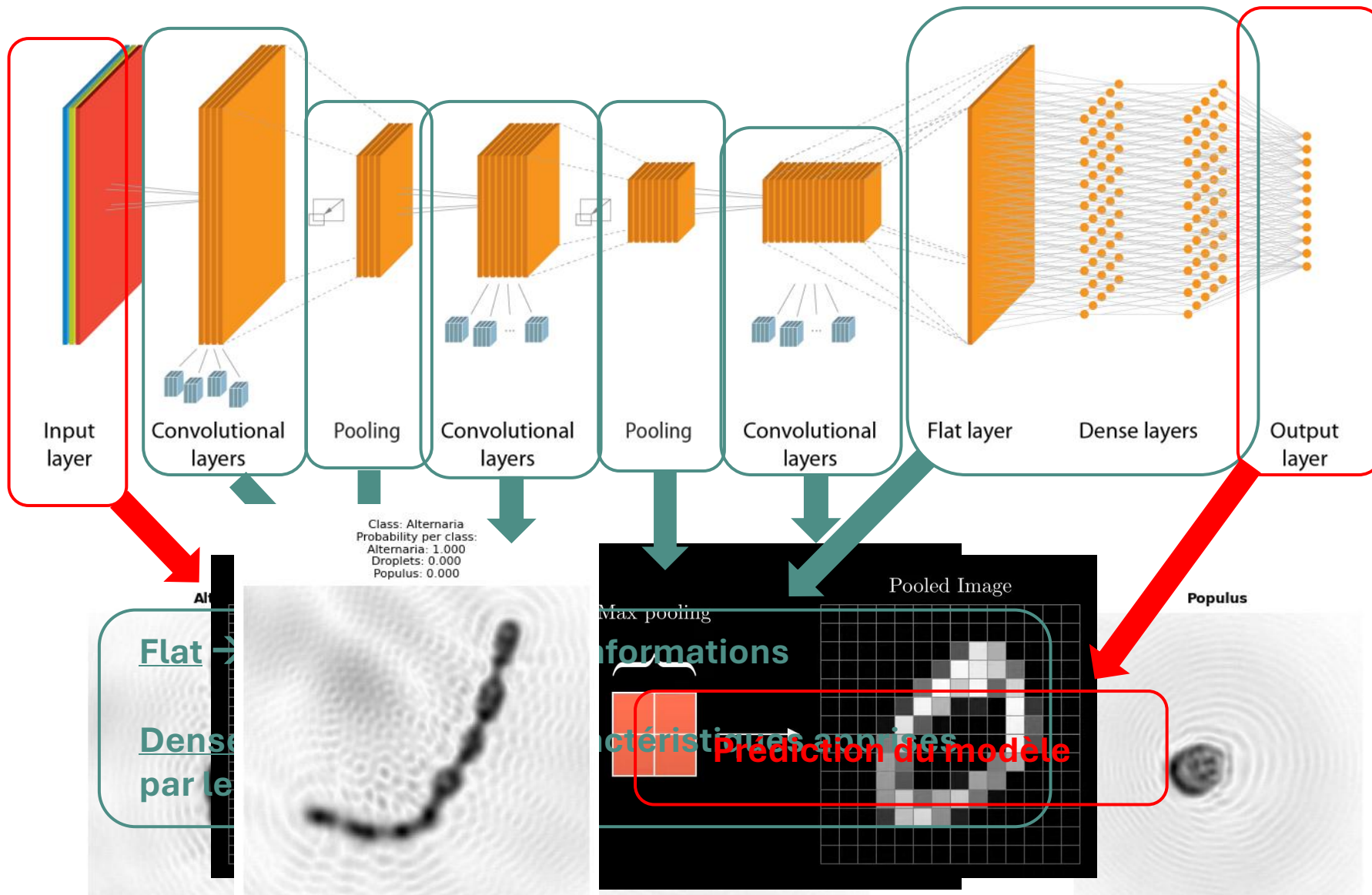
Sources :

- Sauvageat et al. 2020
- Erb et al. 2024
- Swisspoleno Jupiter fonctionnement: <https://kb.swisens.ch/how-does-swisenspoleno-jupiter-work>

Réseau de neurones: époques d'apprentissage



Détail d'une époque

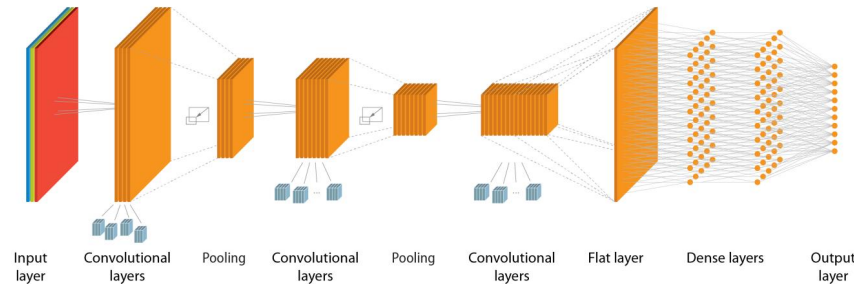


Sources:

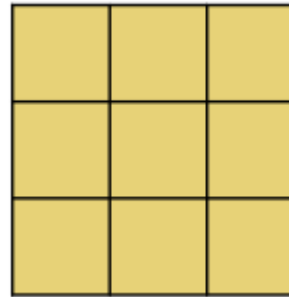
www.fidle.cnrs.fr

https://www.youtube.com/watch?v=ip2HYPC_T9Q

Premier cas d'étude: Apprentissage sur des images simples



Kernel
3×3



Max
Pooling
2×2

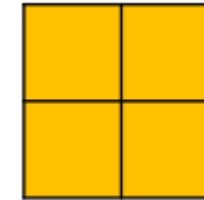
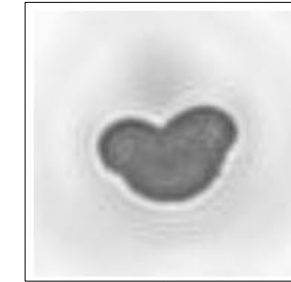


Image en entrée
200×200 pixels



**5 couches
convolutives**

4 MaxPooling

2 Drop Out

```
model = Sequential([
    Conv2D(4, (3,3), activation='relu', input_shape=(200, 200, 1)),
    MaxPooling2D(2,2),

    Conv2D(8, (3,3), activation='relu'),
    MaxPooling2D(2,2),

    Conv2D(16, (3,3), activation='relu'),
    MaxPooling2D(2,2),

    Conv2D(32, (3,3), activation='relu'),
    MaxPooling2D(2,2),
    Dropout(0.3),

    Flatten(),
    Dense(1024, activation='relu'),
    Dropout(0.5),
    Dense(3, activation='softmax')
])
```

Second cas d'étude: Apprentissage sur des paires images

Ajout d'une 4e classe:
Quercus

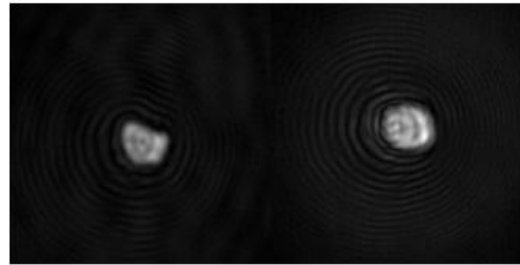


Classes d'entrées:

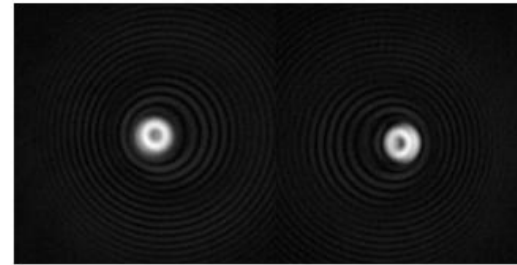
Alternaria (Spore fongique)
Droplet (Hydrométéore)
Populus (Pollen)
+ Quercus (Pollen)

Bonnes prédiction

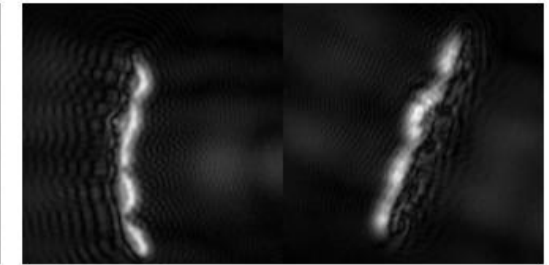
Classe : Populus
Alternaria: 0.000
Droplets: 0.002
Populus: 0.884
Quercus: 0.114



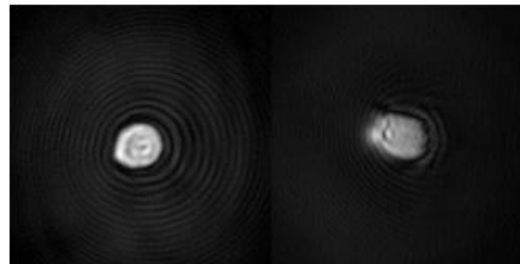
Classe : Droplets
Alternaria: 0.000
Droplets: 0.994
Populus: 0.005
Quercus: 0.002



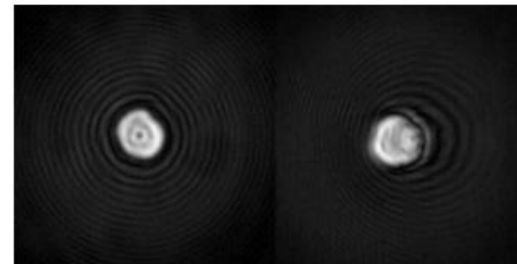
Classe : Alternaria
Alternaria: 1.000
Droplets: 0.000
Populus: 0.000
Quercus: 0.000



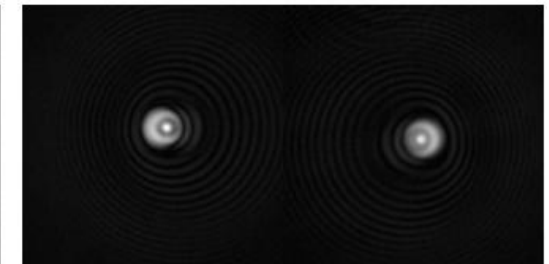
Classe : Quercus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.305
Quercus: 0.695



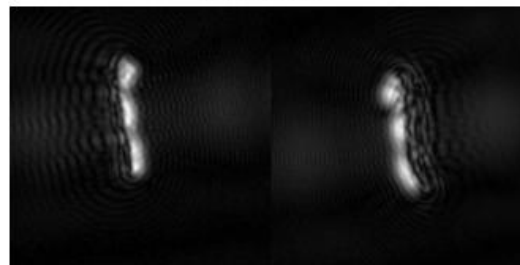
Classe : Quercus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.390
Quercus: 0.610



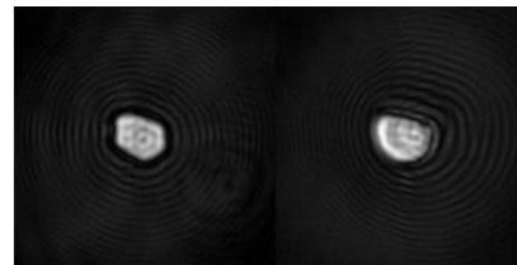
Classe : Droplets
Alternaria: 0.000
Droplets: 1.000
Populus: 0.000
Quercus: 0.000



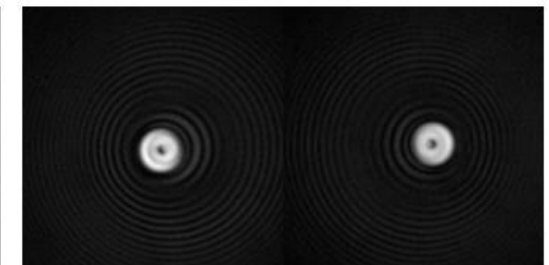
Classe : Alternaria
Alternaria: 1.000
Droplets: 0.000
Populus: 0.000
Quercus: 0.000



Classe : Populus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.810
Quercus: 0.190



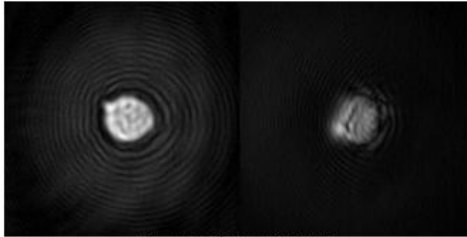
Classe : Droplets
Alternaria: 0.000
Droplets: 1.000
Populus: 0.000
Quercus: 0.000



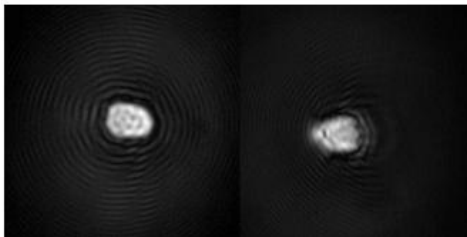
Second cas d'étude: Apprentissage sur des paires images

Erreurs de prédiction

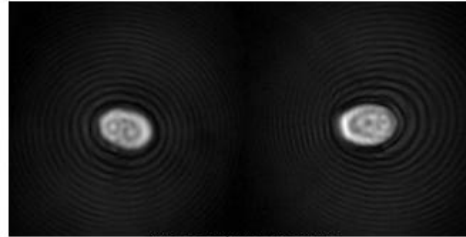
Classe prédite: Quercus
Vraie classe : Populus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.169
Quercus: 0.831



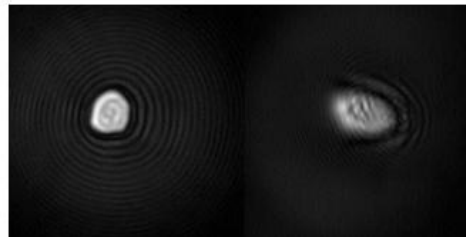
Classe prédite: Quercus
Vraie classe : Populus
Alternaria: 0.001
Droplets: 0.000
Populus: 0.495
Quercus: 0.504



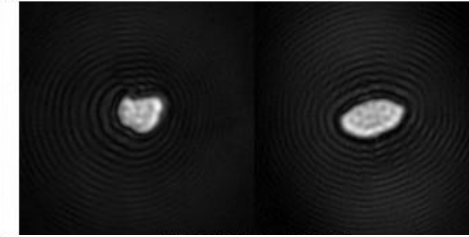
Classe prédite: Populus
Vraie classe : Quercus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.659
Quercus: 0.341



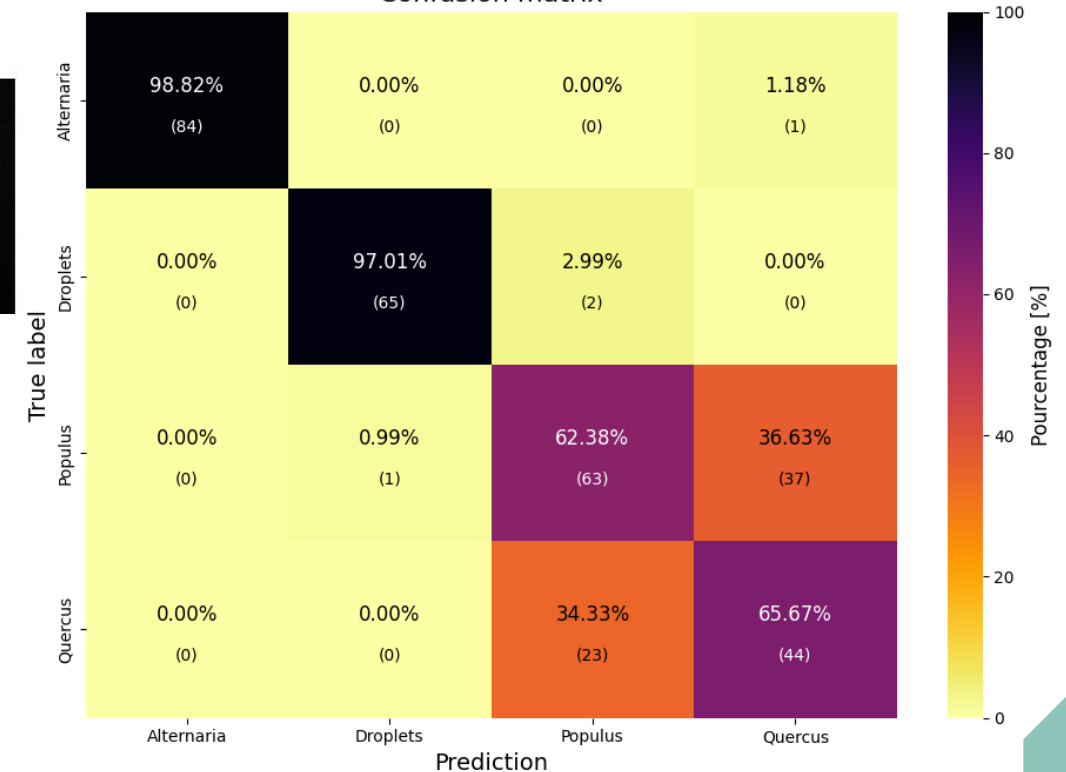
Classe prédite: Populus
Vraie classe : Quercus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.601
Quercus: 0.399



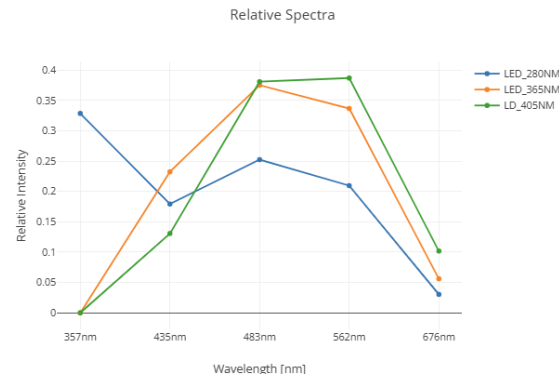
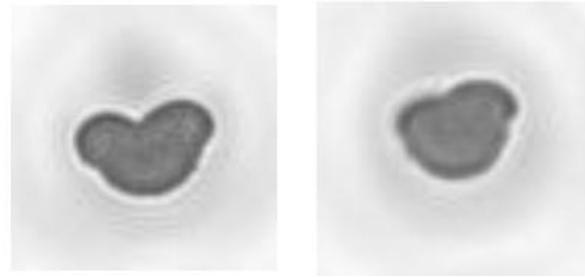
Classe prédite: Populus
Vraie classe : Quercus
Alternaria: 0.000
Droplets: 0.000
Populus: 0.786
Quercus: 0.214



Confusion matrix



Modèle CNN Swisens:



Classification with model pollen_models_first_stage_v3

