

# Bonnes pratiques

Chloé-Agathe Azencott

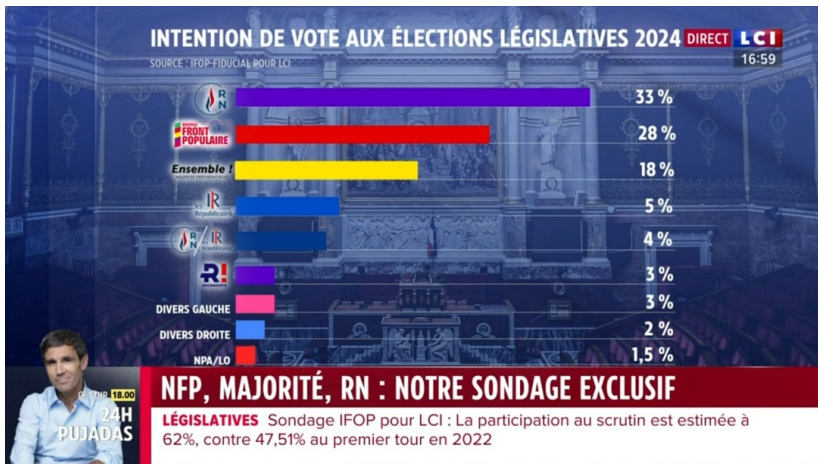
Center for Computational Biology (CBIO)  
Mines Paris PSL – Institut Curie – INSERM U900  
PSL Research University & PR[AI]RIE, Paris, France

Juin 2024

<http://cazencott.info>    [chloe-agathe.azencott@minesparis.psl.eu](mailto:chloe-agathe.azencott@minesparis.psl.eu)    [@cazencott@lipn.info](mailto:@cazencott@lipn.info)

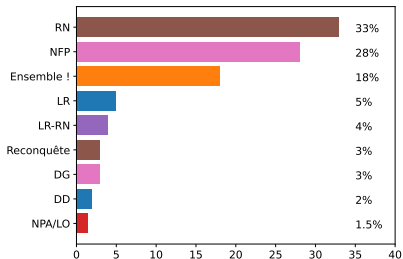
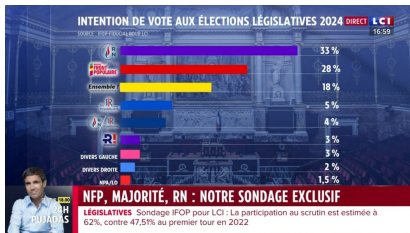
# 1. Visualisation de données

# 1. Exemple 1



TF1, 17 juin 2024, 16h56

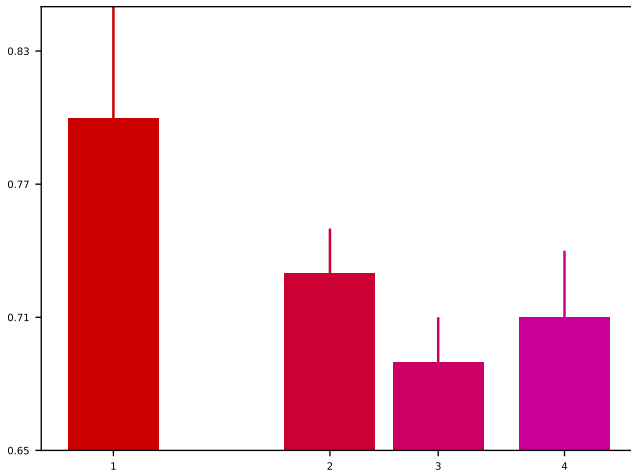
# 1. Exemple 1



## 2. Exemple 2

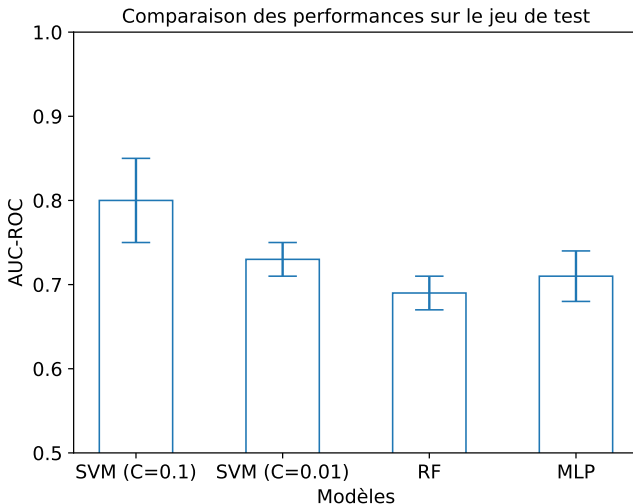
---

Performance de 4 modèles sur un problème d'apprentissage



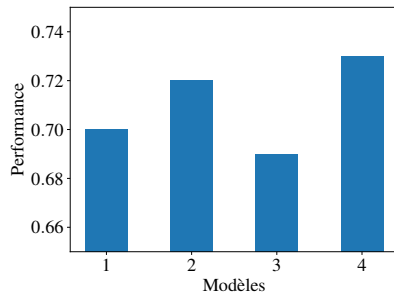
## 2. Exemple 2

Performance de 4 modèles sur un problème d'apprentissage



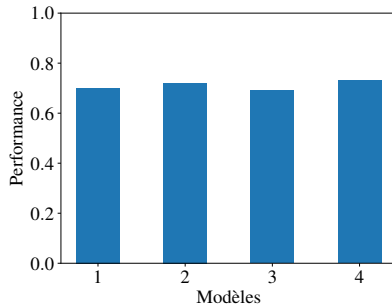
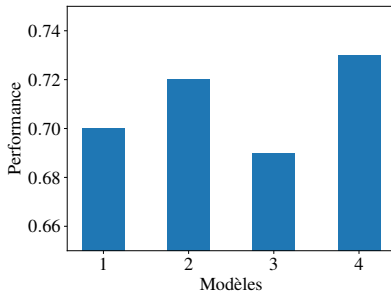
### 3. Choix des axes (1)

---



### 3. Choix des axes (1)

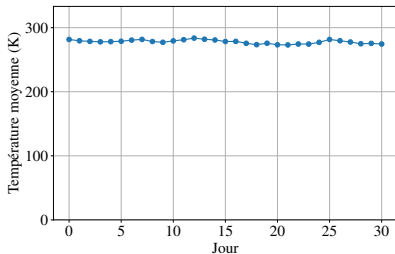
---



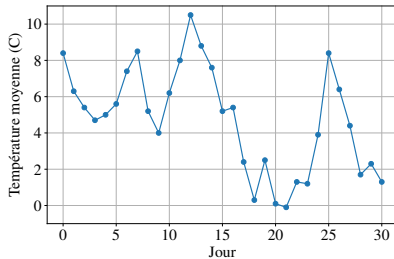
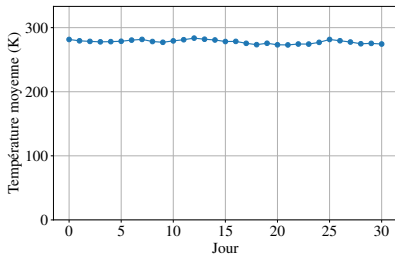


### 3. Choix des axes (2)

---

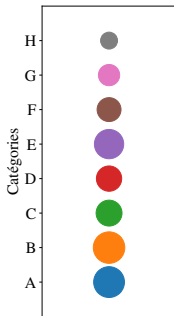


# 3. Choix des axes (2)

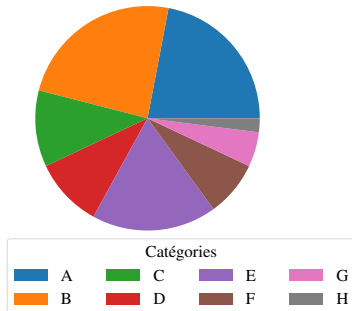
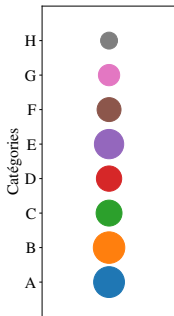


## 4. Proportional ink

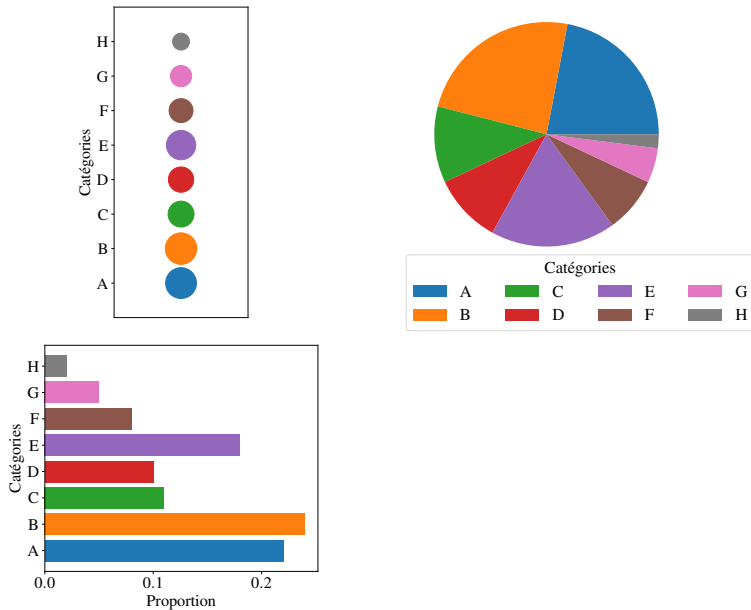
---



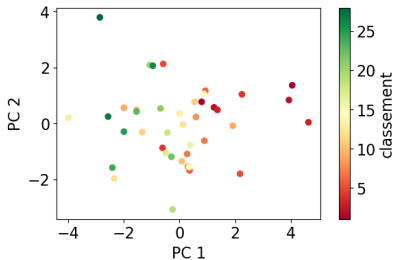
## 4. Proportional ink



# 4. Proportional ink

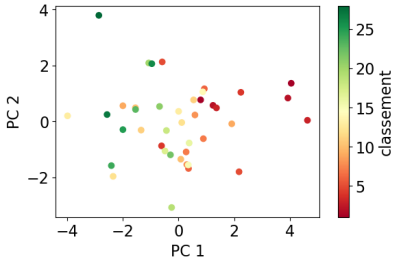


# 5. Dyschromatopie

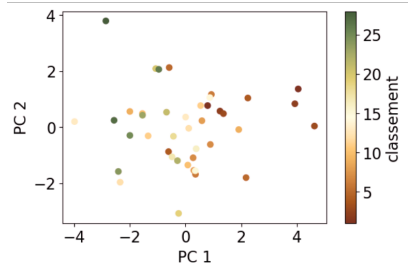


```
plt.scatter(...cmap='RdYlGn')
```

# 5. Dyschromatopie



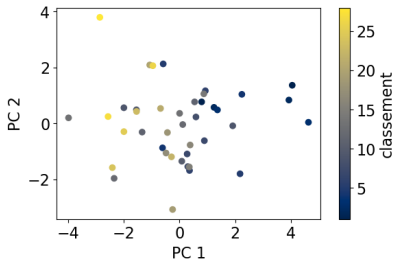
```
plt.scatter(...cmap='RdYlGn')
```



Simulation de deut ranopie par CoBliS

[[lien vers CoBliS](#)]

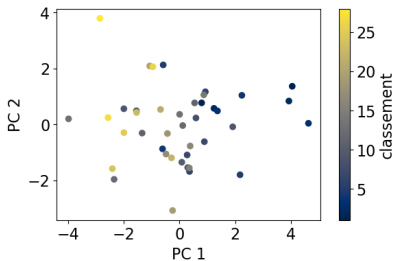
# 5. Dyschromatopie



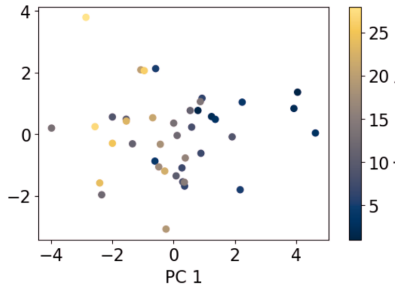
```
plt.scatter(...cmap='cividis')
```



# 5. Dyschromatopie



```
plt.scatter(...cmap='cividis')
```



Simulation de deut ranopie par CoBlis

[lien vers CoBlis]

## 2. Équité des algorithmes

# 1. Recrutement automatisé (Amazon)

---

source: Reuters [\[lien\]](#)

- Système fortement biaisé en faveur des CV déposés par des hommes
- Pourtant le sexe ne faisait pas partie des variables utilisées

## 2. Détection de criminels par leur photo

---

- Article sur arxiv : *Automated Inference on Criminality using Face Images*, Xiaolin Wu & Xi Zhang (2017)
- Motivation : “Unlike a human examiner/judge, a computer vision classifier has absolutely no subjective baggage, having no emotions, no biases whatsoever due to past experience, race, religion, political doctrine, gender, age, etc.”

## 2. Détection de criminels par leur photo

---

- Article sur arxiv : *Automated Inference on Criminality using Face Images*, Xiaolin Wu & Xi Zhang (2017)



(a) Three samples in criminal ID photo set  $S_c$ .

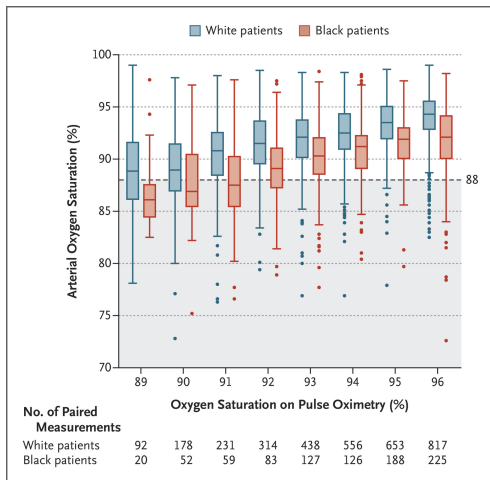


(b) Three samples in non-criminal ID photo set  $S_n$

Figure 1. Sample ID photos in our data set.

### 3. It's not just AI : oxymètres de pouls

- *Racial Bias in Pulse Oximetry Measurement*, Sjoding et al., New England Journal of Medicine, 2020; 383:2477-2478 [lien]



Accuracy of Pulse Oximetry in Measuring Arterial Oxygen Saturation, According to Race.

### 3. Fiabilité

# Attaque (bruit gaussien)



$x$

“panda”

57.7% confidence

+ .007 ×



$\text{sign}(\nabla_x J(\theta, x, y))$

=



$x +$

$\epsilon \text{sign}(\nabla_x J(\theta, x, y))$

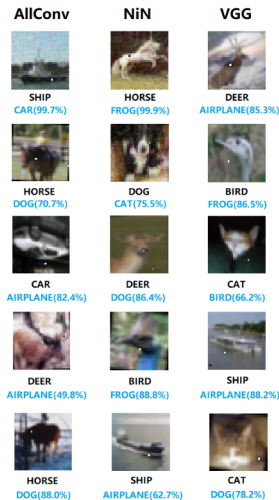
“gibbon”

99.3 % confidence

Goodfellow, Shlens & Szegedy (ICLR 2015)

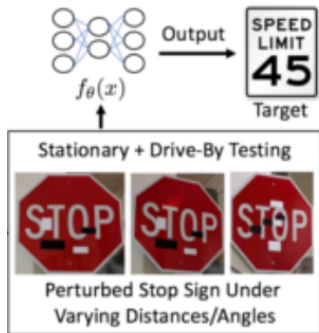


# Attaque (1-pixel)



Su, Vargas & Kouichi (IEEE Transactions on Evolutionary Computation 2019)

# Attaque (monde réel)



Eykholt et al. (CBPR 2018)

## 4. Confidentialité

## 5. Enjeux écologiques