

TP 4 Notebook

December 30, 2019



Département Informatique

MASTER EN SCIENCES ET TECHNIQUES



SYSTÈMES D'INFORMATION DÉCISIONNELS ET IMAGERIE
FACULTÉ DES SCIENCES ET TECHNIQUES ERRACHIDIA



TP4 Maching learning Extraction des caractéristiques & SVM

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Vendredi 27 Decembre 2019

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image

Vous allez trouver la suite ici: [Notebook Online](#)

1 Partie 1 : Extraction de caractéristiques simples

1.0.1 Question 0

```
[27]: from imageio import imread
import matplotlib.pyplot as plt
import numpy as np
#ouverture dune image couleur
im = imread('images/chat1.jpg')
"""Affchage en utilisant une figure de matplotlib"""
plt.imshow(im)
#Suppression des axes
plt.axis('off')
#affchage
plt.show ()
```



1.0.2 Question 1

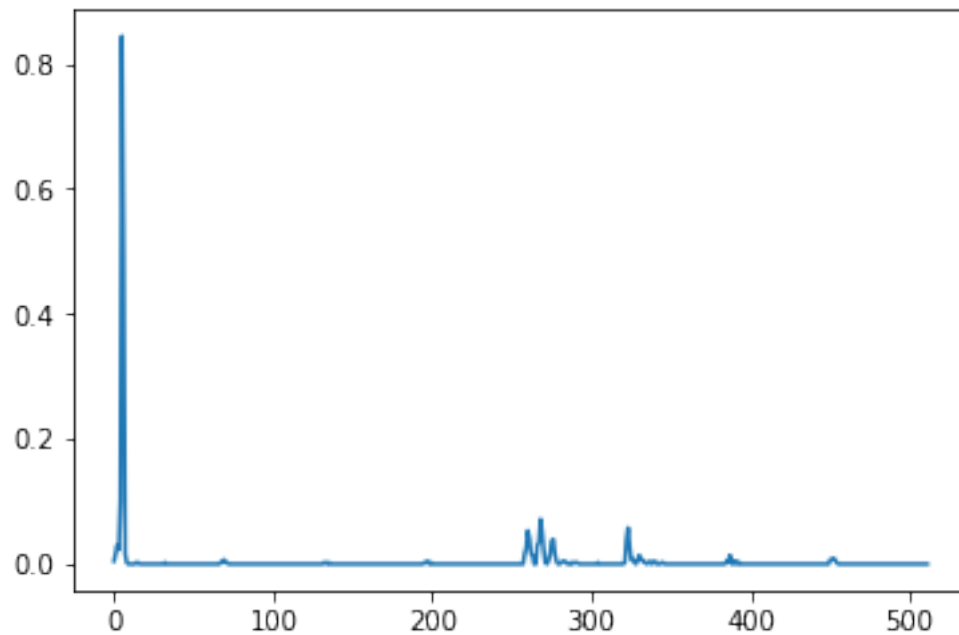
```
[2]: from numpy import resize, shape
def applatir(image):
    image = resize(image, (32, 32))
    return image.flatten()
```

1.0.3 Question 2

```
[3]: import cv2
import os

[4]: def histogramme(image):
    resized = cv2.resize(image, (32, 32))
    bins=(8, 8, 8)
    hsv = cv2.cvtColor(resized, cv2.COLOR_BGR2HSV)
    hist = cv2.calcHist([hsv], [0,1,2], None, bins, [0, 180, 0, 256, 0, 265])
    cv2.normalize(hist, hist)
    return hist.flatten()

[5]: HSVhist = histogramme(im)
plt.plot(HSVhist)
plt.show()
```



1.0.4 Question 3

```
[6]: import os
path = 'images/'
images = ['images/{}'.format(i) for i in os.listdir(path)]
images

[6]: ['images/chien1.jpg',
      'images/chien.jpg',
      'images/chat2.jpg',
```

```
'images/chat1.jpg',  
'images/chien2.jpg',  
'images/chat.jpg',  
'images/chat3.jpg']
```

```
[7]: def images_en_vecteur_naif(images):  
    x = []  
    y = []  
    for image in images:  
        x.append(aplatir(cv2.imread(image)))  
        if 'chat' in image:  
            y.append(1)  
        elif 'chien' in image:  
            y.append(0)  
    return x, y
```

```
[8]: # test de la fonction  
X_naif, y_naif = images_en_vecteur_naif(images)  
X_naif = np.array(X_naif)  
y_naif = np.array(y_naif)  
#y_naif = np.array(y_naif)  
#y_naif = y_naif.reshape(y_naif.size, 1)  
#transformation de la liste obtenu en matrice
```

1.0.5 Question 4

```
[9]: def images_en_vecteur(images):  
    x = []  
    y = []  
    for image in images:  
        x.append(histogramme(cv2.imread(image)))  
        if 'chat' in image:  
            y.append(1)  
        elif 'chien' in image:  
            y.append(0)  
    return x, y
```

```
[10]: #test de la fonction  
X_hist, y_hist = images_en_vecteur(images)  
X_hist = np.array(X_hist)  
y_hist = np.array(y_hist)
```

1.0.6 Question 5

```
[11]: def images_en_vecteur_complet(images):  
    x = []  
    y = []
```

```

for image in images:
    vect1 = applatir(cv2.imread(image))
    vect2 = histogramme(cv2.imread(image))
    x.append(np.concatenate((vect1, vect2), axis=0))
    if 'chat' in image:
        y.append(1)
    elif 'chien' in image:
        y.append(0)
return x, y

```

```

[12]: #test de la fonction
X_complet, y_complet = images_en_vecteur_complet(images)
X_complet = np.array(X_complet)
y_complet = np.array(y_complet)

```

2 Partie 2 : Sélection d'attributs / Réduction de dimension

2.0.1 Question 0

Déjà sur les parties de teste

```

[13]: print(X_naif)
print("Its shape is :", np.shape(X_naif))

```

```

[[140 149 162 ... 236 250 226]
 [ 51 113  89 ... 105  79  50]
 [127 184 229 ... 170 227  83]
 ...
 [206 207 235 ... 237 251 228]
 [246 248 248 ... 236 232 237]
 [ 41  39  39 ...  53  57  54]]
Its shape is : (7, 1024)

```

```

[14]: X_naif[0]

```

```

[14]: array([140, 149, 162, ..., 236, 250, 226], dtype=uint8)

```

```

[15]: print(y_naif)
print("Its shape is :", np.shape(y_naif))

```

```

[0 0 1 1 0 1 1]
Its shape is : (7,)

```

```

[16]: print(X_hist)
print("Its shape is :", np.shape(X_hist))

```

```

[[0.      0.      0.      ... 0.      0.      0.      ]
 [0.      0.      0.      ... 0.      0.      0.      ]

```

```

[0.00544638 0.          0.0490174 ... 0.          0.          0.          ]
...
[0.08579136 0.03431654 0.01143885 ... 0.          0.          0.          ]
[0.          0.          0.          ... 0.          0.          0.          ]
[0.01274118 0.08918826 0.12104121 ... 0.          0.          0.          ]]
Its shape is : (7, 512)

```

```

[17]: print(y_hist)
      print("Its shape is :",np.shape(y_hist))

```

```

[0 0 1 1 0 1 1]
Its shape is : (7,)

```

```

[18]: print(X_complet)
      print("Its shape is :",np.shape(X_complet))

```

```

[[140. 149. 162. ... 0.  0.  0.]
 [ 51. 113.  89. ... 0.  0.  0.]
 [127. 184. 229. ... 0.  0.  0.]
 ...
 [206. 207. 235. ... 0.  0.  0.]
 [246. 248. 248. ... 0.  0.  0.]
 [ 41.  39.  39. ... 0.  0.  0.]]
Its shape is : (7, 1536)

```

```

[19]: print(y_complet)
      print("Its shape is :",np.shape(y_complet))

```

```

[0 0 1 1 0 1 1]
Its shape is : (7,)

```

2.0.2 Question 1

```

[:]:

```

```

[20]: from sklearn.decomposition import PCA

```

```

[21]: # X_naif
      # PCA à 20%
      pca20 = PCA(n_components=int(X_naif.shape[0]*0.8))
      X_naif_ACP20 = pca20.fit_transform(X_naif)
      # PCA à 40%
      pca40 = PCA(n_components=int(X_naif.shape[0]*0.6))
      X_naif_ACP40 = pca40.fit_transform(X_naif)
      # PCA à 60%
      pca60 = PCA(n_components=int(X_naif.shape[0]*0.4))
      X_naif_ACP60 = pca60.fit_transform(X_naif)

```

```
[22]: # X_hist
# PCA à 20%
pca20_hist = PCA(n_components=int(X_hist.shape[0]*0.8))
X_hist_ACP20 = pca20_hist.fit_transform(X_hist)
# PCA à 40%
pca40_hist = PCA(n_components=int(X_hist.shape[0]*0.6))
X_hist_ACP40 = pca40_hist.fit_transform(X_hist)
# PCA à 60%
pca60_hist = PCA(n_components=int(X_hist.shape[0]*0.4))
X_hist_ACP60 = pca60_hist.fit_transform(X_hist)

[23]: # X_complet
# PCA à 20%
pca20_hist = PCA(n_components=int(X_complet.shape[0]*0.8))
X_complet_ACP20 = pca20_hist.fit_transform(X_complet)
# PCA à 40%
pca40_hist = PCA(n_components=int(X_complet.shape[0]*0.6))
X_complet_ACP40 = pca40_hist.fit_transform(X_complet)
# PCA à 60%
pca60_hist = PCA(n_components=int(X_complet.shape[0]*0.4))
X_complet_ACP60 = pca60_hist.fit_transform(X_complet)
```

3 Partie 3 : Classification avec des machines à vecteurs de support

```
[24]: from sklearn import svm
      from sklearn.svm import SVC
      from sklearn.model_selection import cross_validate
      from sklearn import linear_model

[25]: lasso = linear_model.Lasso()
      cv_results = cross_validate(lasso, X_complet, y_complet, cv=3)
      cv_results
```

```
/home/nbuser/anaconda3_420/lib/python3.5/site-
packages/sklearn/linear_model/coordinate_descent.py:491: ConvergenceWarning:
Objective did not converge. You might want to increase the number of iterations.
Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
/home/nbuser/anaconda3_420/lib/python3.5/site-
packages/sklearn/linear_model/coordinate_descent.py:491: ConvergenceWarning:
Objective did not converge. You might want to increase the number of iterations.
Fitting data with very small alpha may cause precision problems.
  ConvergenceWarning)
/home/nbuser/anaconda3_420/lib/python3.5/site-
packages/sklearn/utils/deprecation.py:122: FutureWarning: You are accessing a
training score ('train_score'), which will not be available by default any more
```

```
in 0.21. If you need training scores, please set return_train_score=True
warnings.warn(*warn_args, **warn_kwargs)
```

```
[25]: {'fit_time': array([0.32001901, 0.04164314, 0.06860423]),
      'score_time': array([0.00971746, 0.00038648, 0.0006032 ]),
      'test_score': array([-2.33516706, -1.39038317,  0.          ]),
      'train_score': array([0.98763648, 0.9946662 , 0.98252724])}
```

```
[26]: linear_svc = svm.SVC(kernel='linear')
      print(linear_svc.fit(X_complet, y_complet))
```

```
SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='auto', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```