



# Master Informatique EID2

# Deep Learning TP 2 - Perceptron Multi-Couches

```
1 #@title Fichier import
2 import numpy as np
3 from sklearn.datasets import make_blobs
4 import matplotlib.pyplot as plt
5 import pandas as pd
6 from sklearn.neural_network import MLPRegressor
7 from sklearn import datasets
8 from sklearn.neural_network import MLPClassifier
9 from sklearn import decomposition
10 from sklearn import preprocessing
```

### A) Régression avec un Perceptron Multi-Couches

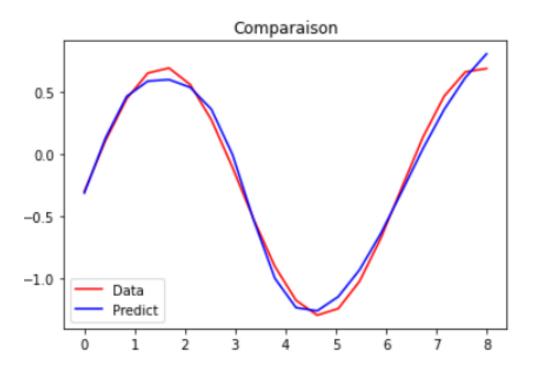
```
1 #@title Création des données
2 x = np.linspace(0,8,20).reshape(-1,1)
3 y = np.sin(x) + np.random.randn(20)*0.2
Création des données
```

### Apprentissage MLPRegressor

```
1 #@title Apprentissage MLPRegressor
2 MLP = MLPRegressor(hidden_layer_sizes=(3,), activation='logistic', learning_rate_init=0.1, max_iter=200)
3 a = MLP.fit(x,y)
4 z = a.predict(x)
```

```
1 #@title Visualisation des données et du predict
2 plt.plot(x[:,0],y[:,0],label='Data',c='r')
3 plt.plot(x[:,0],z[:,0],label='Predict',c='b')
4 plt.title('Comparaison')
5 plt.legend()
6 plt.show()
```

Visualisation des données et du predict



### B) Classification avec un Perceptron Multi-Couches

```
1 #@title Chargement des données Irisdata
                                                      Chargement des données Irisdata
2 iris = datasets.load iris()
3 idx = iris.target != 2
4 irisdata = iris.data[idx].astype(np.float32)
5 iristarget = iris.target[idx].astype(np.float32)
```

### MLPClassifier logistic

```
1 #@title MLPClassifier logistic
2 MLP = MLPClassifier(hidden_layer_sizes=(2,), activation='logistic', learning_rate_init=0.2, max_iter=100)
3 a = MLP.fit(irisdata,iristarget)
4 z = a.predict(irisdata)
5 print('Prediction\n',z)
6 print('Pourcentage d\'erreurs :',float(sum(z != iristarget)/len(z)))
Prediction
 1. 1. 1. 1.]
Pourcentage d'erreurs : 0.0
1 #@title MLPClassifier solver='lbfgs'
2 clf = MLPClassifier(hidden_layer_sizes=(2,), activation='logistic', solver='lbfgs',learning_rate_init=0.1, max_iter=100)
3 a = clf.fit(irisdata,iristarget)
4 z = a.predict(irisdata)
5 print('Prediction\n',z)
```

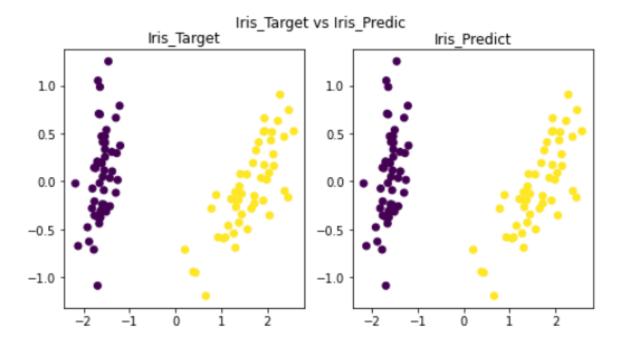
```
6 print('Pourcentage d\'erreurs :',float(sum(z != iristarget)/len(z)))
```

## MLPClassifier solver='lbfgs'

```
Prediction
1. 1. 1. 1. 1
Pourcentage d'erreurs : 0.0
```

```
1 #@title Iris_Target vs Iris_Predic
 2 pca = decomposition.PCA(n_components=2)
 3 irisdata_PCA = pca.fit(irisdata).transform(irisdata)
 4 fig = plt.figure(figsize=(8,4))
 5 fig.suptitle('Iris_Target vs Iris_Predic')
 6 axs = fig.add_subplot(1,2,1)
 7 axs.scatter(irisdata_PCA[:,0],irisdata_PCA[:,1],c=iristarget)
 8 plt.title("Iris_Target")
9 axs = fig.add_subplot(1,2,2)
10 axs.scatter(irisdata_PCA[:,0],irisdata_PCA[:,1],c=z)
11 plt.title("Iris_Predict")
12 plt.show()
```

Iris\_Target vs Iris\_Predic



### Normalisation Iris

# C) Réduction de dimensions avec un Perceptron Multi-Couches

```
1 fig = plt.figure(figsize=(8,4))
 2 fig.suptitle('Reduction de dimension')
 4 MLP_reg = MLPRegressor(hidden_layer_sizes=(2,), activation='identity', learning_rate_init =0.2)
 5 a = MLP_reg.fit(irisdata, irisdata)
 6 X,Y = a.coefs_
 7 Iris_X = np.dot(irisdata,X)
 8 ax1 = fig.add_subplot(1,2,1)
 9 ax1.scatter(Iris_X[:,0],Iris_X[:,1], c=iristarget)
10 plt.title('Linear')
11
12 MLP_reg = MLPRegressor(hidden_layer_sizes=(2,),activation='logistic', learning_rate_init =0.2)
13 a = MLP_reg.fit(irisdata, irisdata)
14 X, Y = a.coefs_
15 Iris_X = np.dot(irisdata,X)
16 ax1 = fig.add_subplot(1,2,2)
17 ax1.scatter(Iris_X[:,0],Iris_X[:,1], c=iristarget)
18 plt.title('Logistic')
```

