



Master Informatique EID2

Traitement numérique des données

TP5 Descente de gradient

```

1 import argparse
2 import numpy as np
3 import pandas as pd
4 import seaborn as sns
5 import matplotlib.pyplot as plt
6 import sklearn.datasets as datasets
7 import sys
8 import scipy

```

```

1 #####
2 # Descente de gradient #
3 #####
4
5 #-----#
6 ## 1. Calculez l'expression analytique de la fonction  $E(x) = (x - 1)(x - 2)(x - 3)(x - 5)$  ##
7 #-----#
8
9 def E(x):
10     return (x-1)*(x-2)*(x-3)*(x-5)

```

```

1 #-----#
2 ## Dérivée ##
3 #-----#
4
5 def dE(x):
6     return 4*np.power(x,3) - 33*np.power(x,2) + 82*x - 61
7

```

```

1 #-----#
2 ## 2. Implémentez l'algorithme DG sous Python pour la fonction E(x) ##
3 #-----#
4
5 def DG(x0,n,e,t):
6     x =[x0]
7     res= np.array([0,E(x[0])])
8     exmin = E(x[0])
9     i = 1
10    x.append(x[0] - n*dE(x[0]))
11    while i < t and abs(x[i] - x[i-1]) > e :
12        x.append(abs(x[i] - n*dE(x[i])))
13        if exmin > E(x[i]):
14            exmin = E(x[i])
15            pred = [i,E(x[i])]
16            res = np.vstack((res,pred))
17            i = i + 1
18    print ('Max_iteration =',i)
19    print ('Min = ',exmin)
20    return res

```

```

1 #-----#
2 ## 3. testez l'algorithme implémenté en utilisant des exemples d'exécution ##
3 ## 4. Affichez le minimum trouvé, ainsi que E(xmin) et le nombre d'itérations ##
4 #-----#
5
6 print('(a) x0 = 5 et  $\eta = 0.001$ ')
7 a = DG(5,0.001,0.001,1000)
8
9 print('\n(b) x0 = 5 et  $\eta = 0.01$ ')
10 b = DG(5,0.01,0.001,1000)
11
12 print('\n(c) x0 = 5 et  $\eta = 0.1$ ')
13 c = DG(5,0.1,0.001,1000)
14
15 print('\n(d) x0 = 5 et  $\eta = 0.17$ ')
16 d = DG(5,0.17,0.001,1000)

```

```

17
18 #print('\n(e) x0 = 5 et  $\eta = 1$ ')
19 #e = DG(5,1.0,0.001,1000)
20
21 print('\n(f) x0 = 0 et  $\eta = 0.001$ ')
22 f = DG(0,0.001,0.001,1000)
23

```

(a) $x_0 = 5$ et $\eta = 0.001$
Max_iteration = 106
Min = -6.89196485736797

(b) $x_0 = 5$ et $\eta = 0.01$
Max_iteration = 20
Min = -6.913929319624811

(c) $x_0 = 5$ et $\eta = 0.1$
Max_iteration = 1000
Min = -6.630728959124845

(d) $x_0 = 5$ et $\eta = 0.17$
Max_iteration = 1000
Min = -6.538588969216836

(f) $x_0 = 0$ et $\eta = 0.001$
Max_iteration = 151
Min = -1.349463100975843

```

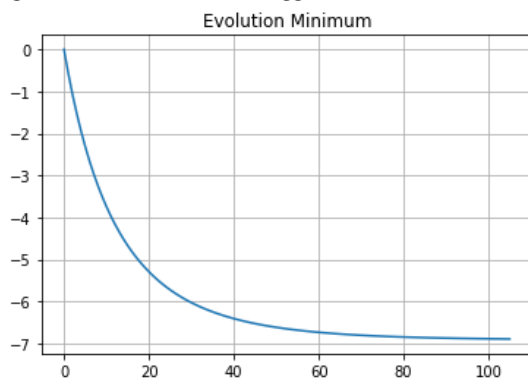
1 #-----#
2 ## 5. Visualisez l'évolution des minimums de la fonction E(x) trouvés au cours des itérations ##
3 #-----#
4
5 def afficherDG(res):
6     plt.plot(res[:,0],res[:,1])
7     plt.title("Evolution Minimum")
8     plt.grid()
9     plt.show()
10
11 print(a[:10])
12 afficherDG(a)
13
14 print(b[:10])
15 afficherDG(b)

```

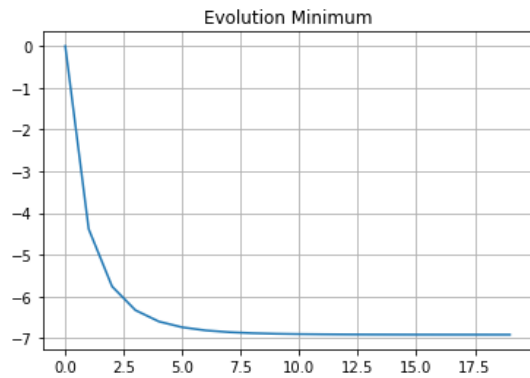
```

[[ 0.      0.      ]
 [ 1.     -0.56114808]
 [ 2.     -1.06646858]
 [ 3.     -1.52269704]
 [ 4.     -1.93560652]
 [ 5.     -2.31016646]
 [ 6.     -2.65067188]
 [ 7.     -2.96084912]
 [ 8.     -3.24394284]
 [ 9.     -3.5027879 ]]

```



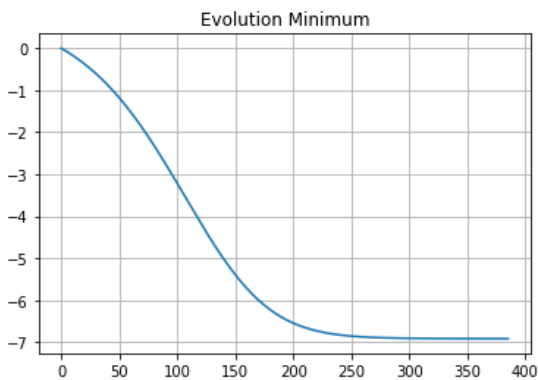
```
[[ 0.      0.      ]
 [ 1.     -4.38349824]
 [ 2.     -5.75937754]
 [ 3.     -6.32923246]
 [ 4.     -6.59920625]
 [ 5.     -6.73772514]
 [ 6.     -6.81259422]
 [ 7.     -6.85454135]
 [ 8.     -6.87865559]
 [ 9.     -6.89278239]]
```



```
1 #-----#
2 ## 6. Testez votre algorithme avec d'autres valeurs de e et nombremax ##
3 #-----#
4 g = DG(3,0.001,0.0001,500)
5 afficherDG(g)
```

Max_iteration = 386

Min = -6.913865530471172



```
1 #-----#
2 ## 1. Calculez les dérivées partielles de la fonction E(a, b) selon a et b ##
3 #-----#
4
5 def modele(X, 0):
6     return X.dot(0)
7
8 def cout(X, y, 0):
9     m = len(y)
10    return 1/(2*m) * np.sum((modele(X, 0) - y)**2)
11
12 def grad(X, y, 0):
13     m = len(y)
14     return 1/m * X.T.dot(modele(X, 0) - y)
15
```

```

16 #-----#
17 ## 2. Implémentez l'algorithme DG ##
18 #-----#
19
20 def DG_R(X, y, n, it):
21     c = np.zeros(it)
22     np.random.seed(0)
23     O = np.random.randn(2,1)
24     for i in range(0, it):
25         O = O - n * grad(X, y, O)
26         c[i] = cout(X, y, O)
27         p = modele(X, O)
28         print("\tDescente pour n =",n,"et nombreMax=",it)
29         plt.figure()
30         plt.scatter(x, y)
31         plt.plot(x, p, c='r')
32     return O, c

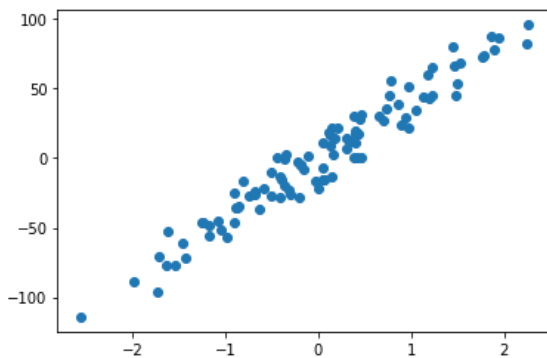
```

```

1 #-----#
2 ## 3. Importez la fonction datasets.make_regression ##
3 #-----#
4 np.random.seed(0)
5 x, y = datasets.make_regression(n_samples=100, n_features=1, noise=10)
6 y = y.reshape(y.shape[0], 1)
7 X = np.hstack((x, np.ones(x.shape)))
8 plt.scatter(x,y)

```

<matplotlib.collections.PathCollection at 0x7f546ab6f898>

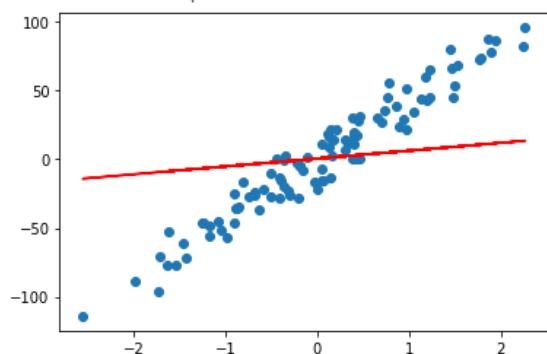


```

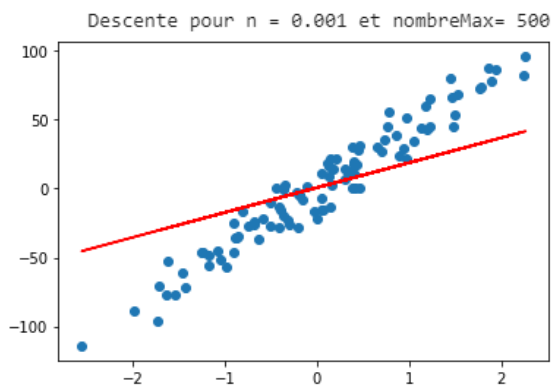
1 #-----#
2 ## 4. Affichez les coefficients trouvés ##
3 #-----#
4 a,a1 = DG_R(X,y,0.001,100)

```

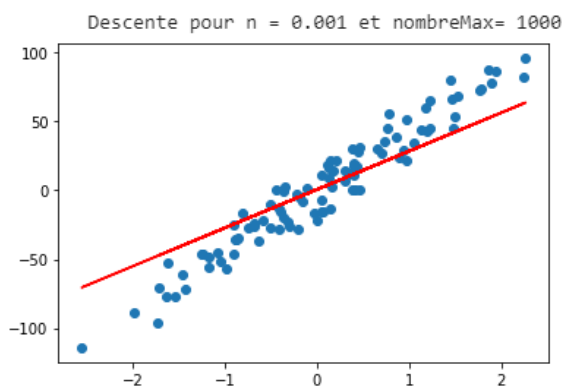
Descente pour n = 0.001 et nombreMax= 100



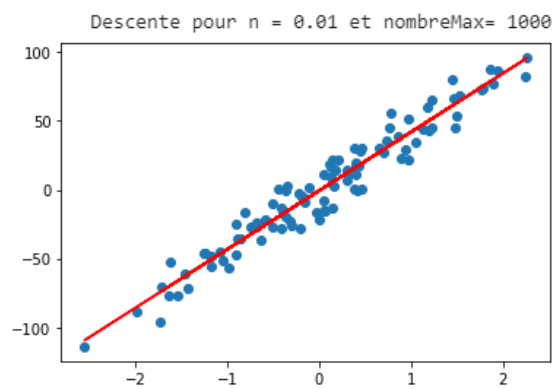
```
1 b = DG_R(X,y,0.001,500)
```



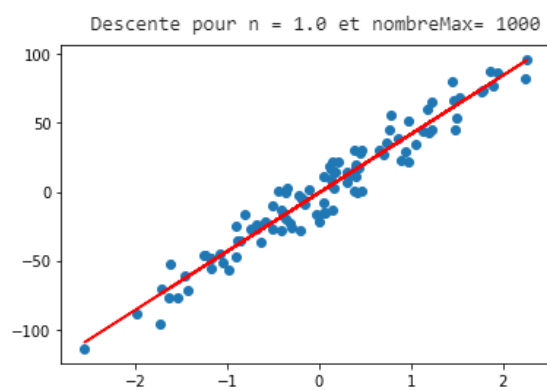
```
1 c = DG_R(X,y,0.001,1000)
```



```
1 d = DG_R(X,y,0.01,1000)
```



```
1 e,e1 =DG_R(X,y,1.0,1000)
```



```
1 #-----#
2 ## 5. Importez la fonction stats.linregress de scipy ##
3 #-----#
4
5 #slope, intercept, r_value, p_value, std_err = scipy.stats.linregress(e, e1)
6
```