regression-using-gredient-descent

November 10, 2024

```
import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
    df = pd.read_csv('diabetes.csv')
    x = df['Glucose'].values
[6]: x
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154, 128, 137, 123, 106, 190, 88, 170, 89, 101, 122, 121, 126,
93], dtype=int64)
```

```
[7]: y = df['Outcome']

[8]: y = np.array(y)
y

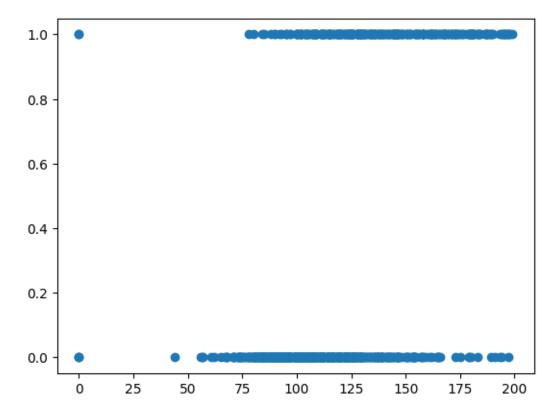
[8]: array([1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0,
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[9]: print(y)

 $0\; 0\; 0\; 1\; 1\; 0\; 0\; 0\; 0\; 0\; 0\; 1\; 0\; 0\; 0\; 1\; 0\; 0\; 0\; 1\; 0\; 0\; 0\; 1\; 0\; 0\; 0\; 1\; 0\; 0\; 0\; 1\; 1\; 0$

```
[10]: plt.scatter(x,y)
plt.show()
```



```
c += y[i]*np.log(p(x[i],b0,b1) + 0.00001) + (1-y[i]) * np.log(1 -u

p(x[i],b0,b1)+0.0001)

return -1*(c/n)
```

```
[14]: def GD(x,y,epoch):
          lemda = 0.001
          b0 = 3.58
          b1 = 2.72
          counter = 1
          c = cost(y,x,b0,b1)
        \# c = cost(y,x,b0,b1)
          while(True):
              c = cost(y,x,b0,b1)
              b0_cost = 0
              b1 cost = 0
              for i in range(len(x)):
                  b1_cost = b1_cost + (-y[i] + p(x[i],b0,b1)) * x[i]
                  b0_cost = b0_cost + (-y[i] + p(x[i],b0,b1))
              b1 = b1 - (1/len(x)) * lemda * b1_cost
              b0 = b0 - lemda * (1/len(x)) *b0_cost
              c1 = cost(y,x,b0,b1)
              print(cost(y,x,b0,b1))
              if(abs(c1-c) < 0.0000001):
                  return (b0,b1)
```

```
[16]: b0,b1 = GD(x,y,60)
```

- 5.974480542310062
- 5.9744781239728555
- 5.9744757057030355
- 5.9744732875006505
- 5.974470869365749
- 5.9744684512983826
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- 5.974456361976256
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- 5.974451526721854
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```
15 b1 = b1 - (1/len(x)) * lemda * b1_cost
         16 b0 = b0 - lemda * (1/len(x)) *b0_cost
     ---> 17 c1 = cost(y,x,b0,b1)
         19 print(cost(y,x,b0,b1))
         21 if(abs(c1-c) < 0.0000001):
     Input In [13], in cost(y, x, b0, b1)
          3 c = 0
          4 for i in range(n):
     \Rightarrow p(x[i],b0,b1)+0.0001)
          6 return -1*(c/n)
     KeyboardInterrupt:
[]: b0,b1 = GD(x,y,1)
[]: def y_pred(b0,b1,x):
        1 = []
        for i in range(len(x)):
           if(p(x[i],b0,b1) >= 0.5):
               1.append(1)
           else:
               1.append(0)
        return (np.array(1))
[ ]: yPred = y_pred(b0,b1,x)
[]: yPred
```