week3 code

August 7, 2024

```
[1]: import pandas as pd
     import numpy as np
[2]: df = pd.read_csv('diabetes_csv.csv')
[3]: df.head
[3]: <bound method NDFrame.head of
                                           Pregnancies
                                                         Glucose
                                                                   BloodPressure
     SkinThickness
                     Insulin
                                BMI
                             148
                                              72
                                                               35
                                                                          0
                                                                             33.6
     1
                     1
                              85
                                              66
                                                               29
                                                                          0
                                                                             26.6
     2
                     8
                             183
                                              64
                                                                0
                                                                          0
                                                                            23.3
     3
                     1
                              89
                                              66
                                                               23
                                                                         94 28.1
     4
                     0
                             137
                                              40
                                                               35
                                                                        168 43.1
     . .
                                                               •••
     763
                    10
                             101
                                              76
                                                                        180 32.9
                                                               48
     764
                     2
                             122
                                              70
                                                               27
                                                                          0 36.8
     765
                     5
                             121
                                              72
                                                               23
                                                                        112 26.2
     766
                     1
                             126
                                              60
                                                                0
                                                                          0
                                                                             30.1
     767
                                              70
                                                                            30.4
                     1
                              93
                                                               31
                                                                          0
          DiabetesPedigreeFunction
                                            Outcome
                                       Age
                               0.627
     0
                                        50
                                                   1
                               0.351
                                                   0
     1
                                        31
     2
                               0.672
                                        32
                                                   1
     3
                               0.167
                                        21
                                                   0
     4
                               2.288
                                        33
                                                   1
     763
                               0.171
                                        63
                                                   0
     764
                               0.340
                                        27
                                                   0
     765
                                                   0
                               0.245
                                        30
     766
                               0.349
                                        47
                                                   1
     767
                               0.315
                                        23
                                                   0
     [768 rows x 9 columns]>
```

1. 1. Load data in Pandas.

[4]: df.head

[4]:	<pre><bound method="" ndframe.head="" of<="" pre=""></bound></pre>		Pregnancies	Glucose	BloodPressure			
	SkinThickness	Insuli	in BMI	\				
	0	6	148		72	35	0	33.6
	1	1	85		66	29	0	26.6
	2	8	183		64	0	0	23.3
	3	1	89		66	23	94	28.1
	4	0	137		40	35	168	43.1
	• •		••	•••	•••	•••	•••	
	763	10	101		76	48	180	32.9
	764	2	122		70	27	0	36.8
	765	5	121		72	23	112	26.2
	766	1	126		60	0	0	30.1
	767	1	93		70	31	0	30.4
	Diabetes	Pedigree	Function	Age	Outcome			
	0		0.627	50	1			
	1		0.351	31	0			
	2		0.672	32	1			
	3		0.167	21	0			
	4		2.288	33	1			
			•••		•••			

0.171

0.340

0.245

0.349

0.315

[768 rows x 9 columns]>

1. 2. Drop columns that aren't useful.

[5]: df.corr()

		·					
[5]:		Pregnancie	s Glucose	e BloodPressure	SkinThickness	\	
	Pregnancies	1.00000	0 0.129459	0.141282	-0.081672		
	Glucose	0.12945	9 1.000000	0.152590	0.057328		
	BloodPressure	0.14128	2 0.152590	1.000000	0.207371		
	SkinThickness	-0.08167	2 0.057328	0.207371	1.000000		
	Insulin	-0.07353	5 0.331357	7 0.088933	0.436783		
	BMI	0.01768	3 0.221071	0.281805	0.392573		
	${\tt DiabetesPedigreeFunction}$	-0.03352	3 0.137337	7 0.041265	0.183928		
	Age	0.54434	1 0.263514	1 0.239528	-0.113970		
	Outcome	0.22189	8 0.466581	0.065068	0.074752		
		Insulin	BMI I	DiabetesPedigreeF	unction \		
	Pregnancies	-0.073535	0.017683	-0	-0.033523		
	Glucose	0.331357	0.221071	0	0.137337		
	BloodPressure	0.088933	0.281805	0	.041265		

```
SkinThickness
                                0.436783 0.392573
                                                                    0.183928
      Insulin
                                1.000000 0.197859
                                                                    0.185071
      BMI
                                0.197859 1.000000
                                                                    0.140647
      DiabetesPedigreeFunction 0.185071 0.140647
                                                                    1.000000
                               -0.042163 0.036242
                                                                    0.033561
      Outcome
                                0.130548 0.292695
                                                                    0.173844
                                           Outcome
                                     Age
                                0.544341 0.221898
     Pregnancies
      Glucose
                                0.263514 0.466581
     BloodPressure
                                0.239528 0.065068
      SkinThickness
                               -0.113970 0.074752
      Insulin
                               -0.042163 0.130548
     BMI
                                0.036242 0.292695
     DiabetesPedigreeFunction 0.033561 0.173844
                                1.000000 0.238356
      Outcome
                                0.238356 1.000000
 [6]: df = df.drop(df['Glucose'])
 [7]: df.columns
 [7]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
             'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
            dtype='object')
     1. 3. Drop rows with missing values. df.dropna(axis=0)
     1. 4. Create dummy variables.
 [8]: features =
       →['Pregnancies','Glucose','BloodPressure','SkinThickness','Insulin','BMI','DiabetesPedigreeF
 [9]: outcomes = ['Outcome']
[10]: X = df[features]
      y = df[outcomes]
[11]: X.columns
[11]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
             'BMI', 'DiabetesPedigreeFunction', 'Age'],
            dtype='object')
[12]: y.columns
[12]: Index(['Outcome'], dtype='object')
```

```
1. 5. Take care of missing data.
[13]: df.isnull().sum()
[13]: Pregnancies
                                   0
      Glucose
                                   0
      BloodPressure
                                   0
      SkinThickness
                                   0
      Insulin
                                   0
      BMI
                                   0
      DiabetesPedigreeFunction
                                   0
      Age
                                   0
      Outcome
      dtype: int64
[14]: X=X.fillna(0)
     1.6 Convert the data frame to NumPy.
[15]: ser = np.array(df, dtype='int32')
[16]: ser
[16]: array([[ 1, 85, 66, ...,
                                   0,
                                       31,
                                             0],
                                   Ο,
                                       32,
             [ 8, 183, 64, ...,
                                             1],
             [ 1, 89, 66, ...,
                                   Ο,
                                       21,
                                             0],
             [ 5, 121, 72, ...,
                                  Ο,
                                       30,
                                             0],
             [ 1, 126, 60, ...,
                                   Ο,
                                       47,
                                             1],
             [ 1, 93, 70, ...,
                                       23,
                                             0]], dtype=int32)
                                   Ο,
[17]: ser.shape
[17]: (632, 9)
[18]: print(ser.dtype)
      print(type(ser))
     int32
     <class 'numpy.ndarray'>
     1. 7. Divide the data set into training data and test data.
[19]: from sklearn.model_selection import train_test_split
[20]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.05,_
       →random_state=0)
[21]: X_train.shape
[21]: (600, 8)
```

```
[22]: y_train.shape
```

[22]: (600, 1)

0.0.1 2. a. Construct a CSV file with the following attributes:

Study time in hours of ML lab course (x) Score out of 10 (y) The dataset should contain 10 rows.

	Hours	Score
0	2	1
1	4	2
2	6	3
3	8	4
4	10	5
5	12	6
6	14	7
7	16	8
8	18	9
9	20	10

b. Create a regression model and display the following: Coefficients: B0 (intercept) and B1 (slope)

RMSE (Root Mean Square Error)

Predicted responses

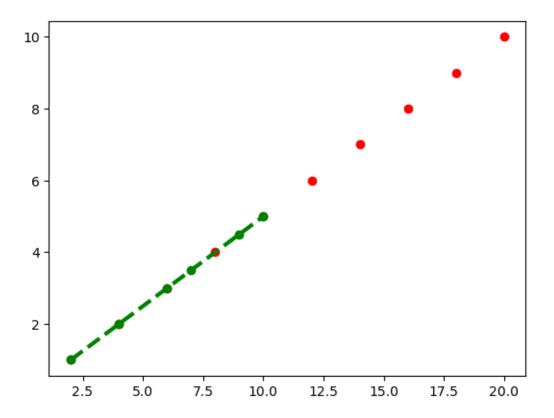
```
[24]: def linear_regression(X, Y):
    n = len(X)

# Calculate means
mean_X = sum(X) / n
mean_Y = sum(Y) / n

# Calculate the slope (b) and intercept (a)
```

```
numerator = sum((X[i] - mean_X) * (Y[i] - mean_Y) for i in range(n))
          denominator = sum((X[i] - mean_X) ** 2 for i in range(n))
          num=0
          den = 0
          for i in range(n):
              num+=(X[i] - mean_X)*(Y[i] - mean_Y)
              den+=(X[i]-mean_X)**2
           b = numerator / denominator
            a = mean Y - b * mean X
          \#eqn : y = a0 + a1.x
          a1 = num/den
          a0 = mean_Y - (a1*mean_X)
          return a0, a1
      a, b = linear_regression(X, y)
      print("Intercept (a):", a)
      print("Slope (b):", b)
     Intercept (a): 0.0
     Slope (b): 0.5
[25]: \#linear\ model\ eqn:\ y = a0 + a1*x \Rightarrow y = 0 + 0.5*x
 []:
[26]: def getPredictions(a, b, X_vals, y_vals):
          for x in X_vals:
              y = a + (b*x)
              y_vals.append(y)
[27]: x_{test} = X
      y_test = []
      getPredictions(a,b,x_test, y_test)
      print("predict y: ")
      print(y_test)
      print("actual y: ")
      print(df['Score'])
      def getError(y_test, y_train):
              err = 0
```

```
for i in range(len(y_test)):
                  err = err + (y_train[i] - y_test[i])**2
              err = err**0.5
              print ("error: ", err)
      getError(y_test, df['Score'] )
     predict y:
     [1.0, 2.0, 3.0, 4.0, 5.0, 6.0, 7.0, 8.0, 9.0, 10.0]
     actual y:
     0
           1
     1
           2
     2
           3
     3
           4
     4
           5
     5
           6
     6
           7
     7
           8
     8
           9
          10
     Name: Score, dtype: int64
     error: 0.0
[28]: x_{test} = [2,4,6,7,9,10]
      y_{test} = []
      for x in x_test:
          y = a + (b*x)
          y_test.append(y)
      print("predicted: ")
      print((x_test))
      print((y_test))
     predicted:
     [2, 4, 6, 7, 9, 10]
     [1.0, 2.0, 3.0, 3.5, 4.5, 5.0]
[29]: #### plotting input data on scatter plot
[30]: import matplotlib.pyplot as plt
      plt.plot(df['Hours'], df['Score'], 'ro')
      plt.plot(x_test, y_test, 'g', linestyle='dashed', marker='o', linewidth=3)
      plt.show()
```

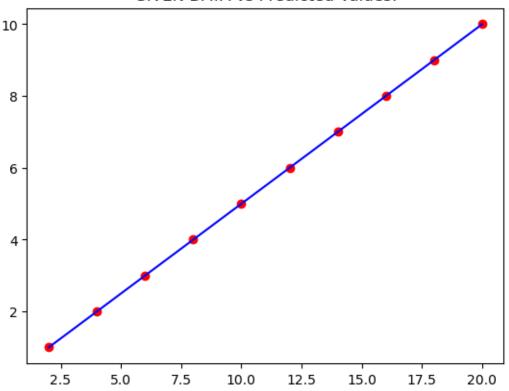


\mathbf{rmse}

Create a scatter plot of the data points in red color and plot the graph of x vs. predicted y in blue color.

```
[32]: plt.plot(df['Hours'],df['Score'], 'ro')
   plt.title("GIVEN DATA vs Predicted Values:")
   plt.plot(df['Hours'], df['Score'], 'b')
   plt.show()
```

GIVEN DATA vs Predicted Values:



Calculus Method

```
import numpy as np

np.random.seed(0) # For reproducibility
X = 2 * np.random.rand(100, 1)
y = 4 + 3 * X + np.random.randn(100, 1)

X_b = np.c_[np.ones((100, 1)), X]

theta = np.random.randn(2, 1) # Random initialization

learning_rate = 0.01
n_iterations = 1000
m = len(X_b)

for iteration in range(n_iterations):
    gradients = 2/m * X_b.T.dot(X_b.dot(theta) - y)
    theta = theta - learning_rate * gradients

print("Estimated coefficients:")
print(f"Intercept (beta_0): {theta[0][0]}")
```

```
print(f"Slope (beta_1): {theta[1][0]}")
      X_{\text{new}} = \text{np.array}([[0], [2]])
      X_{new_b} = np.c_{np.ones((2, 1)), X_{new}}
      y_predict = X_new_b.dot(theta)
      print("Predictions:")
      print(y_predict)
     Estimated coefficients:
     Intercept (beta_0): 4.20607718142562
     Slope (beta_1): 2.9827303563323175
     Predictions:
     [[ 4.20607718]
      [10.17153789]]
     2. e. Compare the coefficients obtained using both methods and compare them with
     the analytical solution.
[34]: print('difference bw intercepts: ')
      print(theta[0][0] - a)
     difference bw intercepts:
     4.20607718142562
[35]: print('diff bw slopes: ')
      print(theta[1][0] - b)
     diff bw slopes:
     2.4827303563323175
     2. f. Test your model to predict the score obtained when the study time of a student
     is 10 hours.
 []: test_x = 10
      test_y = a + (b*test_x)
      print(f'predicted value for x = {test_x} is',test_y)
```

[]: