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[20]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, mean_absolute_error
from sklearn.model_selection import train_test_split

In [2]: dfgpa = pd.read_csv("D:\\\\24 - Machine_Learning\\download files\\LR1.csv")
dfgpa

Out[2]:
   SAT  GPA
0  1714  2.40
1  1664  2.52
2  1760  2.54
3  1685  2.74
4  1693  2.83
...    ...
79 1936  3.71
80 1810  3.71
81 1987  3.73
82 1962  3.76
83 2050  3.81
84 rows x 2 columns

In [4]: dfgpa.head(10)

Out[4]:
   SAT  GPA
0  1714  2.40
1  1664  2.52
2  1760  2.54
3  1685  2.74
4  1693  2.83
5  1670  2.91
6  1764  3.00
7  1764  3.00
8  1792  3.01
9  1850  3.01

In [5]: dfgpa.describe()

Out[5]:
   SAT      GPA
count  84.000000  84.000000
mean   1845.273810  3.330238
std    104.530661  0.271617
min    1634.000000  2.400000
25%    1772.000000  3.190000
50%    1846.000000  3.380000
75%    1934.000000  3.502500
max    2050.000000  3.810000

In [7]: dfgpa.shape

Out[7]:
(84, 2)

In [8]: x = dfgpa['SAT']
y = dfgpa['GPA']

In [9]: plt.scatter(x, y)
plt.xlabel('SAT', fontsize = 13)
plt.ylabel('GPA', fontsize = 13)
plt.show()

In [11]: sns.regplot(x,y,color='purple')
D:\24-Annaconda\lib\site-packages\seaborn\decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be 'data', and passing other arguments without an explicit keyword will result in an error or misinterpretation.
warnings.warn(
<matplotlib.figure.Figure at 0x1714>

In [15]: x.shape

Out[15]:
(84,)

In [12]: x.head()

Out[12]:
0    1714
1    1664
2    1760
3    1685
4    1693
Name: SAT, dtype: int64

In [13]: X=x.values.reshape(-1, 1)
# 1D to 2D to 2D

In [14]: X.shape

Out[14]:
(84, 1)

In [18]: X

Out[18]:
0    1714
1    1664
2    1760
3    1685
4    1693
...
79    1936
80    1810
81    1987
82    1962
83    2050
Name: SAT, length: 84, dtype: int64

In [17]: X

Out[17]:
array([[1714],
       [1664],
       [1760],
       [1685],
       [1693],
       [1670],
       [1764],
       [1764],
       [1792],
       [1850],
       [1730],
       [1775],
       [1735],
       [1732],
       [1773],
       [1872],
       [1755],
       [1674],
       [1842],
       [1786],
       [1761],
       [1722],
       [1663],
       [1697],
       [1874],
       [1826],
       [1787],
       [1821],
       [2020],
       [1794],
       [1769],
       [1934],
       [1750],
       [1855],
       [1800],
       [1849],
       [1808],
       [1954],
       [1777],
       [1831],
       [1865],
       [1859],
       [1966],
       [1702],
       [1900],
       [1925],
       [1824],
       [1956],
       [1897],
       [1979],
       [1802],
       [1855],
       [1907],
       [1624],
       [1879],
       [1887],
       [1730],
       [1953],
       [1781],
       [1891],
       [1964],
       [1880],
       [1893],
       [2841],
       [1893],
       [1832],
       [1890],
       [1834],
       [1861],
       [1931],
       [1933],
       [1778],
       [1975],
       [1994],
       [2021],
       [2015],
       [1997],
       [2020],
       [1843],
       [1936],
       [1810],
       [1987],
       [1962],
       [2060]], dtype=int64)

In [22]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=41)

In [23]: X_train.shape

Out[23]:
(67, 1)

In [24]: X_test.shape

Out[24]:
(17, 1)

In [25]: LR = LinearRegression()
LR.fit(X_train, y_train)
# fit is used for training the data and it is supervised learning

Out[25]:
LinearRegression()

In [ ]: # So here our MODEL is created and ready for testing

In [26]: y_pred = LR.predict(X_test)
# predict function is used for testing

In [27]: y_test

Out[27]:
75    3.62
24    3.24
67    3.54
54    3.44
32    3.29
6    3.00
42    3.38
78    3.71
43    3.39
0    2.40
2    2.54
69    3.58
45    3.40
37    3.34
38    2.37
27    3.28
48    3.41
Name: GPA, dtype: float64

In [28]: y_pred
# used for getting the predicted value

Out[28]:
array([3.61756384, 3.55190026, 3.48783823, 3.39975294, 3.23319167,
       3.21557461, 3.53908785, 3.34209712, 3.11627847, 3.13548708,
       3.20948844, 3.48389258, 3.47342428, 3.51986925, 3.23639477,
       3.366863 , 3.36451981])

In [29]: acc = mean_squared_error(y_test, y_pred)
acc
# here we calculating the error i.e. difference between actual value and predicted value

Out[29]:
0.08496837191499861

In [30]: weights = LR.coef_
intercept = LR.intercept_
print(weights, intercept)
[0.00160155] 0.398439136988436

In [31]: LR.score(X_test, y_test)

Out[31]:
0.2594693604787399

In [32]: plt.scatter(X_test, y_test)
plt.plot(X_test, y_pred, color='green')
plt.show()

In [33]: df = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})
df

Out[33]:
   Actual  Predicted
75    3.62    3.617564
24    3.24    3.551900
67    3.54    3.487838
54    3.44    3.399753
32    3.29    3.233192
6     3.00    3.215575
42    3.38    3.539088
78    3.71    3.342097
43    3.39    3.116278
0     2.40    3.135497
2     2.54    3.209166
69    3.58    3.483034
45    3.40    3.473424
37    3.34    3.519869
38    3.37    3.236395
27    3.28    3.306863
48    3.41    3.364519

In [34]: df2 = df.head(25)
df2.plot(kind='bar', figsize=(10,10))
plt.grid(which='major', linestyle='-', linewidth='0.5', color='green')
plt.grid(which='minor', linestyle=':', linewidth='0.5', color='black')
plt.show()

In [35]: new_data = pd.DataFrame([2115,980])
new_data

Out[35]:
   0
0  2115
1    900

In [36]: LR.predict(new_data)

Out[36]:
array([3.77771893, 3.83183482])

In [37]: LR.predict([[4562]])

Out[37]:
array([7.69671345])

In [38]: print('Mean Absolute Error: ',mean_absolute_error(y_test,y_pred))
print('Mean Squared Error: ',mean_squared_error(y_test,y_pred))
print('Root Mean Squared Error: ',np.sqrt(mean_squared_error(y_test,y_pred)))
Mean Absolute Error: 0.2023950168348122
Mean Squared Error: 0.08496837191499861
Root Mean Squared Error: 0.2913994115814687

In [39]: dfgpa['GPA'].mean()

Out[39]:
3.330238095238094

In [41]: plt.scatter(x,y)
yhat = LR.coef_*x+LR.intercept_
fig = plt.plot(x, yhat, lw=2, c='orange', label='Regression Line')
plt.xlabel('SAT', fontsize=12)
plt.ylabel('GPA', fontsize=12)
plt.show()

In [42]: plt.scatter(x,y)
yhat = LR.coef_*x+LR.intercept_
fig = plt.plot(x, yhat, lw=2, c='orange', label='Regression Line')
plt.xlabel('SAT', fontsize=12)
plt.ylabel('GPA', fontsize=12)
plt.show()
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