# students\_mark\_predictor

#### November 18, 2022

```
[29]: #Import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

### 1 Load Dataset

```
[3]: df.head()
```

```
[3]:
        study_hours
                      student_marks
     0
                6.83
                               78.50
     1
                6.56
                               76.74
                               78.68
                 NaN
     3
                5.67
                               71.82
     4
                8.67
                               84.19
```

```
[4]: df.tail()
```

```
student_marks
[4]:
          study_hours
     195
                  7.53
                                 81.67
     196
                  8.56
                                 84.68
     197
                  8.94
                                 86.75
     198
                  6.60
                                 78.05
     199
                  8.35
                                 83.50
```

```
[5]: df.shape
```

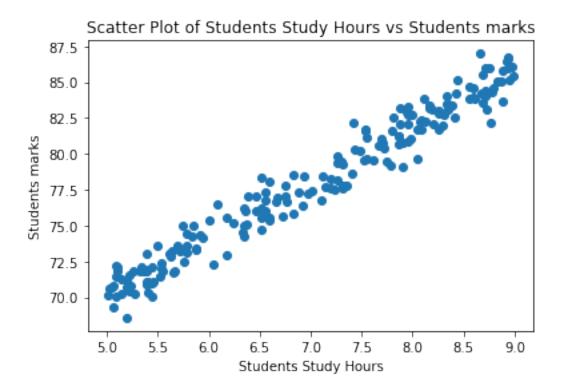
[5]: (200, 2)

#### 1.1 Discover and visualize the data to gain insights

```
[6]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200 entries, 0 to 199
    Data columns (total 2 columns):
         Column
                        Non-Null Count Dtype
         _____
                        -----
     0
         study_hours
                        195 non-null
                                        float64
                                        float64
         student_marks 200 non-null
    dtypes: float64(2)
    memory usage: 3.2 KB
[7]: df.describe()
[7]:
            study_hours
                         student_marks
            195.000000
                             200.00000
     count
               6.995949
                              77.93375
    mean
     std
               1.253060
                               4.92570
    min
               5.010000
                              68.57000
    25%
               5.775000
                              73.38500
    50%
               7.120000
                              77.71000
    75%
               8.085000
                              82.32000
    max
               8.990000
                              86.99000
[8]: plt.scatter(x =df.study_hours, y = df.student_marks)
     plt.xlabel("Students Study Hours")
     plt.ylabel("Students marks")
     plt.title("Scatter Plot of Students Study Hours vs Students marks")
```

plt.show()



## 1.2 Prepare the data for Machine Learning algorithms

```
[9]: # Data Cleaning
     df.isnull().sum()
[10]:
[10]: study_hours
                       5
      student_marks
                       0
      dtype: int64
[11]: df.mean()
[11]: study_hours
                        6.995949
      student_marks
                       77.933750
      dtype: float64
[12]: df2 = df.fillna(df.mean())
[13]:
     df2.isnull().sum()
[13]: study_hours
                       0
      student_marks
                       0
      dtype: int64
```

```
[14]: df2.head()
「14]:
         study_hours student_marks
            6.830000
                              78.50
      1
            6.560000
                              76.74
      2
            6.995949
                              78.68
                              71.82
      3
            5.670000
      4
            8.670000
                              84.19
[15]: # split dataset
[16]: X = df2.drop("student_marks", axis = "columns")
      y = df2.drop("study hours", axis = "columns")
      print("shape of X = ", X.shape)
      print("shape of y = ", y.shape)
     shape of X = (200, 1)
     shape of y = (200, 1)
[17]: from sklearn.model_selection import train_test_split
      X_train, X_test,y_train,y_test = train_test_split(X,y, test_size = 0.2,__
      →random_state=51)
      print("shape of X_train = ", X_train.shape)
      print("shape of y train = ", y train.shape)
      print("shape of X_test = ", X_test.shape)
      print("shape of y_test = ", y_test.shape)
     shape of X_{train} = (160, 1)
     shape of y_{train} = (160, 1)
     shape of X_{test} = (40, 1)
     shape of y_{test} = (40, 1)
     2 Select a model and train it
[18]:  # y = m * x + c 
       from sklearn.linear_model import LinearRegression
       lr = LinearRegression()
[19]: lr.fit(X_train,y_train)
[19]: LinearRegression()
[20]: lr.coef_
[20]: array([[3.93571802]])
[21]: lr.intercept_
```

```
[21]: array([50.44735504])
[22]: m = 3.93
      c = 50.44
      y = m * 4 + c
      У
[22]: 66.16
[23]: lr.predict([[4]])[0][0].round(2)
     C:\ProgramData\Anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X
     does not have valid feature names, but LinearRegression was fitted with feature
     names
       warnings.warn(
[23]: 66.19
[24]: | y_pred = lr.predict(X_test)
      y_pred
[24]: array([[83.11381458],
             [78.9025963],
             [84.57003024],
             [85.82946001],
             [84.72745896],
             [80.75238377],
             [72.84159055],
             [71.66087515],
             [73.23516235],
             [71.66087515],
             [73.47130543],
             [76.38373677],
             [73.23516235],
             [73.58937697],
             [82.95638585],
             [70.40144538],
             [73.23516235],
             [78.74516758],
             [75.55723598],
             [82.68088559],
             [76.65923703],
             [70.48015974],
             [74.77009238],
             [77.98143645],
             [85.59331693],
             [82.56281405],
             [76.42309395],
```

```
[78.39095296],
              [81.38209865],
              [81.73631327],
              [83.15317176],
              [82.20859943],
              [81.10659839],
              [73.58937697],
              [71.1492318],
              [71.89701823],
              [81.53952737],
              [72.60544747],
              [71.93637541]])
[25]: pd.DataFrame(np.c_[X_test, y_test, y_pred], columns = ["study_hours",__

¬"student_marks_original", "student_marks_predicted"])

[25]:
                                                  student_marks_predicted
          study_hours
                        student_marks_original
      0
             8.300000
                                           82.02
                                                                 83.113815
      1
             7.230000
                                           77.55
                                                                 78.902596
      2
             8.670000
                                           84.19
                                                                 84.570030
      3
             8.990000
                                           85.46
                                                                 85.829460
      4
                                           84.03
             8.710000
                                                                 84.727459
      5
             7.700000
                                           80.81
                                                                 80.752384
      6
             5.690000
                                           73.61
                                                                 72.841591
      7
             5.390000
                                           70.90
                                                                 71.660875
      8
             5.790000
                                           73.14
                                                                 73.235162
      9
             5.390000
                                           73.02
                                                                 71.660875
      10
             5.850000
                                           75.02
                                                                 73.471305
      11
             6.590000
                                           75.37
                                                                 76.383737
      12
             5.790000
                                           74.44
                                                                 73.235162
      13
             5.880000
                                           73.40
                                                                 73.589377
                                           81.70
      14
             8.260000
                                                                 82.956386
      15
             5.070000
                                           69.27
                                                                 70.401445
      16
             5.790000
                                           73.64
                                                                 73.235162
      17
             7.190000
                                           77.63
                                                                 78.745168
      18
             6.380000
                                           77.01
                                                                 75.557236
      19
             8.190000
                                           83.08
                                                                 82.680886
      20
              6.660000
                                           76.63
                                                                 76.659237
      21
             5.090000
                                           72.22
                                                                 70.480160
      22
             6.180000
                                           72.96
                                                                 74.770092
      23
             6.995949
                                           76.14
                                                                 77.981436
      24
             8.930000
                                           85.96
                                                                 85.593317
      25
             8.160000
                                           83.36
                                                                 82.562814
      26
             6.600000
                                           78.05
                                                                 76.423094
      27
                                           84.60
             8.790000
                                                                 85.042316
             7.100000
                                           76.76
      28
                                                                 78.390953
```

[85.0423164],

7.860000	81.24	81.382099
7.950000	80.86	81.736313
8.310000	82.69	83.153172
8.070000	82.30	82.208599
7.790000	79.17	81.106598
5.880000	73.34	73.589377
5.260000	71.86	71.149232
5.450000	70.06	71.897018
7.900000	80.76	81.539527
5.630000	72.87	72.605447
5.460000	71.10	71.936375
	7.950000 8.310000 8.070000 7.790000 5.880000 5.260000 5.450000 7.900000 5.630000	7.950000       80.86         8.310000       82.69         8.070000       82.30         7.790000       79.17         5.880000       73.34         5.260000       71.86         5.450000       70.06         7.900000       80.76         5.630000       72.87

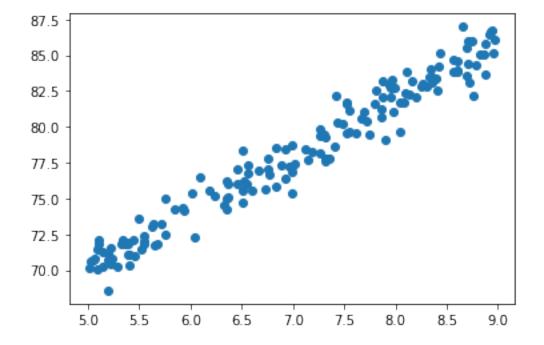
### 2.1 Fine-tune the model

[26]: lr.score(X\_test,y\_test)

[26]: 0.9514124242154464

[27]: plt.scatter(X\_train,y\_train)

[27]: <matplotlib.collections.PathCollection at 0x2c4a2c3d1c0>

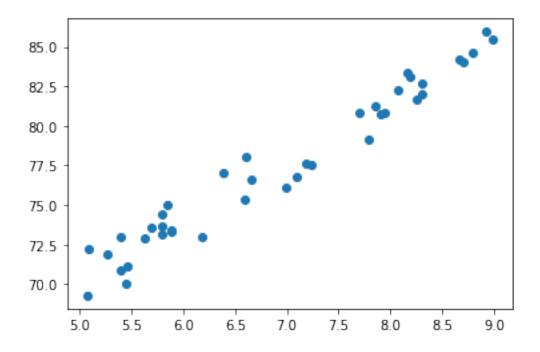


```
[28]: plt.scatter(X_test, y_test)
plt.plot(X_train, lr.predict(X_train), color = "r")
```

```
TypeError
                                          Traceback (most recent call last)
File C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\base.py:
 ⇒3621, in Index.get_loc(self, key, method, tolerance)
   3620 try:
-> 3621
           return self. engine.get loc(casted key)
   3622 except KeyError as err:
File C:\ProgramData\Anaconda3\lib\site-packages\pandas\_libs\index.pyx:136, in_
 →pandas. libs.index.IndexEngine.get loc()
File C:\ProgramData\Anaconda3\lib\site-packages\pandas\_libs\index.pyx:142, in_
 →pandas._libs.index.IndexEngine.get_loc()
TypeError: '(slice(None, None, None), None)' is an invalid key
During handling of the above exception, another exception occurred:
InvalidIndexError
                                          Traceback (most recent call last)
Input In [28], in <cell line: 2>()
      1 plt.scatter(X test, y test)
----> 2 plt.plot(X_train, lr.predict(X_train), color = "r")
File C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\pyplot.py:2757, in_
 oplot(scalex, scaley, data, *args, **kwargs)
   2755 @_copy_docstring_and_deprecators(Axes.plot)
   2756 def plot(*args, scalex=True, scaley=True, data=None, **kwargs):
-> 2757
           return gca().plot(
   2758
                *args, scalex=scalex, scaley=scaley,
                **({"data": data} if data is not None else {}), **kwargs)
   2759
File C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_axes.py:1632,__
 →in Axes.plot(self, scalex, scaley, data, *args, **kwargs)
   1390 """
   1391 Plot y versus x as lines and/or markers.
   1392
   (...)
   1629 (``'green'``) or hex strings (``'#008000'``).
   1630 """
   1631 kwargs = cbook.normalize_kwargs(kwargs, mlines.Line2D)
-> 1632 lines = [*self._get_lines(*args, data=data, **kwargs)]
   1633 for line in lines:
   1634
            self.add_line(line)
File C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py:312, i:
 -_process_plot_var_args.__call__(self, data, *args, **kwargs)
    310
           this += args[0],
    311
            args = args[1:]
--> 312 yield from self._plot_args(this, kwargs)
```

```
File C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py:487, i:
 oprocess plot var args. plot args(self, tup, kwargs, return_kwargs)
                kw[prop name] = val
    486 if len(xy) == 2:
            x = _{check_1d(xy[0])}
--> 487
            y = check 1d(xy[1])
    489 else:
File C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\__init__.py:
 \hookrightarrow1327, in _check_1d(x)
   1321 with warnings.catch_warnings(record=True) as w:
            warnings.filterwarnings(
   1322
                "always",
   1323
   1324
                category=Warning,
   1325
                message='Support for multi-dimensional indexing')
            ndim = x[:, None].ndim
-> 1327
            # we have definitely hit a pandas index or series object
   1328
   1329
            # cast to a numpy array.
   1330
            if len(w) > 0:
File C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\frame.py:3505, in_
 →DataFrame. getitem (self, key)
   3503 if self.columns.nlevels > 1:
            return self._getitem_multilevel(key)
   3504
-> 3505 indexer = self.columns.get_loc(key)
   3506 if is_integer(indexer):
            indexer = [indexer]
   3507
File C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\base.py:
 →3628, in Index.get_loc(self, key, method, tolerance)
   3623
                raise KeyError(key) from err
   3624
            except TypeError:
   3625
                # If we have a listlike key, _check_indexing_error will raise
                # InvalidIndexError. Otherwise we fall through and re-raise
   3626
   3627
                # the TypeError.
                self. check indexing error(key)
-> 3628
   3629
   3631 # GH#42269
File C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\base.py:
 ⇔5637, in Index._check_indexing_error(self, key)
   5633 def _check_indexing_error(self, key):
            if not is_scalar(key):
   5634
                # if key is not a scalar, directly raise an error (the code bel w
   5635
   5636
                # would convert to numpy arrays and raise later any way) -
 →GH29926
-> 5637
              raise InvalidIndexError(key)
```

InvalidIndexError: (slice(None, None, None), None)



#### 2.2 Save Ml Model

```
[]: import joblib
    joblib.dump(lr, "student_mark_predictor.pkl")

[]: model = joblib.load("student_mark_predictor.pkl")

[]: model.predict([[5]])[0][0]
```