CH: - 5 Regression Analysis

1.Supervised Learning ¶

Regression(integer or float)

- Linear
- 1.)simple
- 2.)multiple
- polynomial
- 1.)simple
- 2.)multiple

Classification(categorical data)

- KNN
- · Decision Tree

equation

- y = mx+b (linear)
- y = a+bx+cx1+dx2+ex3 (linear)
- $y = a+bx+cx^2+dx^3$ (polynomial)
- $y = a+bx1+cx2^2+dx3^3$ (polynomial)

Simple Linear Regression

scikit library

preprocessing

```
In [100]: import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
df = pd.read_csv("Book1.csv")
df.head()
```

Out[100]:

	cgpa	package
0	6.89	3.26
1	5.12	1.98
2	7.82	3.25
3	7.42	3.67
4	6.94	3.57

In [105]: help(train_test_split)

```
Help on function train_test_split in module sklearn.model_selection._split:
train_test_split(*arrays, **options)
    Split arrays or matrices into random train and test subsets
    Quick utility that wraps input validation and
    ``next(ShuffleSplit().split(X, y))`` and application to input data
    into a single call for splitting (and optionally subsampling) data in a
    oneliner.
    Read more in the :ref:`User Guide <cross_validation>`.
    Parameters
    -----
    *arrays : sequence of indexables with same length / shape[0]
        Allowed inputs are lists, numpy arrays, scipy-sparse
        matrices or pandas dataframes.
    test_size : float or int, default=None
        If float, should be between 0.0 and 1.0 and represent the proportion
        of the dataset to include in the test split. If int, represents the
        absolute number of test samples. If None, the value is set to the
        complement of the train size. If ``train_size`` is also None, it wil
1
        be set to 0.25.
    train_size : float or int, default=None
        If float, should be between 0.0 and 1.0 and represent the
        proportion of the dataset to include in the train split. If
        int, represents the absolute number of train samples. If None,
        the value is automatically set to the complement of the test size.
    random_state : int or RandomState instance, default=None
        Controls the shuffling applied to the data before applying the spli
t.
        Pass an int for reproducible output across multiple function calls.
        See :term:`Glossary <random state>`.
    shuffle : bool, default=True
        Whether or not to shuffle the data before splitting. If shuffle=Fals
e
        then stratify must be None.
    stratify : array-like, default=None
        If not None, data is split in a stratified fashion, using this as
        the class labels.
    Returns
    splitting : list, length=2 * len(arrays)
        List containing train-test split of inputs.
        .. versionadded:: 0.16
            If the input is sparse, the output will be a
            ``scipy.sparse.csr_matrix``. Else, output type is the same as th
            input type.
    Examples
```

```
>>> import numpy as np
              >>> from sklearn.model_selection import train_test_split
              >>> X, y = np.arange(10).reshape((5, 2)), range(5)
              >>> X
              array([[0, 1],
                     [2, 3],
                      [4, 5],
                      [6, 7],
                     [8, 9]])
              >>> list(y)
              [0, 1, 2, 3, 4]
              >>> X_train, X_test, y_train, y_test = train_test_split(
                      X, y, test_size=0.33, random_state=42)
               . . .
              >>> X_train
              array([[4, 5],
                      [0, 1],
                      [6, 7]]
              >>> y_train
              [2, 0, 3]
              >>> X test
              array([[2, 3],
                      [8, 9]])
              >>> y_test
              [1, 4]
              >>> train_test_split(y, shuffle=False)
              [[0, 1, 2], [3, 4]]
In [106]: | 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)
          x_test.shape
Out[106]: (40, 1)
In [107]: 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)
          y_train.shape
Out[107]: (160,)
In [108]: | 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)
          y_test.shape
Out[108]: (40,)
In [109]: from sklearn.model selection import train test split
          from sklearn.linear_model import LinearRegression
          x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
          lr=LinearRegression()
          model = lr.fit(x_train,y_train) # creating a model or equation
          lr.coef
Out[109]: array([0.56004401])
```

```
In [110]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    lr=LinearRegression()
    model = lr.fit(x_train,y_train) # creating a model or equation
    lr.intercept_
```

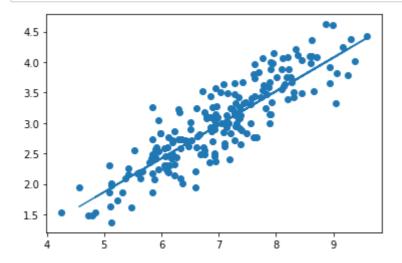
Out[110]: -1.057454060871304

posting

```
In [111]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    lr=LinearRegression()
    model = lr.fit(x_train,y_train) # creating a model or equation
    y_pred = lr.predict(x_test)
    y_pred
```

```
Out[111]: array([2.55731678, 3.16105985, 3.90881503, 2.9671974 , 2.3579154 , 3.6595633 , 2.97273633, 4.28546208, 3.49893441, 3.32168874, 3.21644913, 3.49893441, 3.42692836, 1.95357372, 3.16659878, 2.68471211, 2.92842491, 4.04174928, 3.07243702, 2.9671974 , 1.9701905 , 2.82872422, 3.53216798, 3.73156936, 2.13081939, 2.61270606, 3.37153909, 2.98381418, 2.52962215, 3.43246729, 3.35492231, 3.60971296, 2.65701747, 3.10567058, 2.9671974 , 2.97273633, 2.38561004, 2.96165847, 1.95911264, 3.52662905])
```

```
In [112]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    import matplotlib.pyplot as plt
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    lr=LinearRegression()
    model = lr.fit(x_train,y_train) # creating a model or equation
    y_pred = lr.predict(x_test)
    plt.scatter(df['cgpa'],df['package'])
    plt.plot(x_test,y_pred)
    plt.show()
```



```
from sklearn.model_selection import train_test_split
In [113]:
          from sklearn.linear_model import LinearRegression
          from sklearn import metrics
          import math
          x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
          lr=LinearRegression()
          model = lr.fit(x_train,y_train)
          y_pred = lr.predict(x_test)
          print("MAE",metrics.mean_absolute_error(y_test,y_pred))
          print("MSE", metrics.mean squared error(y test, y pred))
          print("RMSE",math.sqrt(metrics.mean_squared_error(y_test,y_pred)))
          print("R2score", metrics.r2_score(y_test,y_pred)) # x depend on y
          MAE 0.28331977292553595
          MSE 0.13470894990622276
          RMSE 0.3670271787023718
          R2score 0.6419528555237273
In [114]: import numpy as np
          test_data = np.array([9.3,7.8])
          print(test_data)
          print(test_data.shape)
          [9.3 7.8]
          (2,)
In [115]: import numpy as np
          test_data = test_data.reshape(-1,1)
          print(test_data)
          print(test_data.shape)
          [[9.3]
           [7.8]]
          (2, 1)
In [116]: lr.predict(test_data)
Out[116]: array([4.3834217 , 3.48369105])
```

Multiple Linear Regression

```
In [1]:
        # Target is selling_price.
        import pandas as pd
        df = pd.read csv("car data.csv")
        df.isna().sum()
Out[1]: Car_Name
                          0
        Year
                          0
                          0
        Selling Price
        Present_Price
                          0
        Kms Driven
         Fuel_Type
                          0
        Seller_Type
                          0
        Transmission
                          0
        Owner
        dtype: int64
```

```
In [7]: import pandas as pd
    df = pd.read_csv("car data.csv")
    df = df.drop("Car_Name",axis=1)
    df
```

Out[7]:

	Year	Selling_Price	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owı
0	2014	3.35	5.59	27000	Petrol	Dealer	Manual	
1	2013	4.75	9.54	43000	Diesel	Dealer	Manual	
2	2017	7.25	9.85	6900	Petrol	Dealer	Manual	
3	2011	2.85	4.15	5200	Petrol	Dealer	Manual	
4	2014	4.60	6.87	42450	Diesel	Dealer	Manual	
296	2016	9.50	11.60	33988	Diesel	Dealer	Manual	
297	2015	4.00	5.90	60000	Petrol	Dealer	Manual	
298	2009	3.35	11.00	87934	Petrol	Dealer	Manual	
299	2017	11.50	12.50	9000	Diesel	Dealer	Manual	
300	2016	5.30	5.90	5464	Petrol	Dealer	Manual	

301 rows × 8 columns

In [8]: df["Age"] = 2025-df["Year"]
df

Out[8]:

	Year	Selling_Price	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owı
0	2014	3.35	5.59	27000	Petrol	Dealer	Manual	
1	2013	4.75	9.54	43000	Diesel	Dealer	Manual	
2	2017	7.25	9.85	6900	Petrol	Dealer	Manual	
3	2011	2.85	4.15	5200	Petrol	Dealer	Manual	
4	2014	4.60	6.87	42450	Diesel	Dealer	Manual	
296	2016	9.50	11.60	33988	Diesel	Dealer	Manual	
297	2015	4.00	5.90	60000	Petrol	Dealer	Manual	
298	2009	3.35	11.00	87934	Petrol	Dealer	Manual	
299	2017	11.50	12.50	9000	Diesel	Dealer	Manual	
300	2016	5.30	5.90	5464	Petrol	Dealer	Manual	

301 rows × 9 columns

In [9]: df = df.drop("Year",axis=1)
df

Out[9]:

	Selling_Price	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owner	Ą
0	3.35	5.59	27000	Petrol	Dealer	Manual	0	
1	4.75	9.54	43000	Diesel	Dealer	Manual	0	
2	7.25	9.85	6900	Petrol	Dealer	Manual	0	
3	2.85	4.15	5200	Petrol	Dealer	Manual	0	
4	4.60	6.87	42450	Diesel	Dealer	Manual	0	
296	9.50	11.60	33988	Diesel	Dealer	Manual	0	
297	4.00	5.90	60000	Petrol	Dealer	Manual	0	
298	3.35	11.00	87934	Petrol	Dealer	Manual	0	
299	11.50	12.50	9000	Diesel	Dealer	Manual	0	
300	5.30	5.90	5464	Petrol	Dealer	Manual	0	

301 rows × 8 columns

In [10]: pd.get_dummies(df)

Out[10]:

IG	Fuel_Type_Diesel	Fuel_Type_Petrol	Seller_Type_Dealer	Seller_Type_Individual	Transmission_A
0	0	1	1	0	
0	1	0	1	0	
0	0	1	1	0	
0	0	1	1	0	
0	1	0	1	0	
0	1	0	1	0	
0	0	1	1	0	
0	0	1	1	0	
0	1	0	1	0	
0	0	1	1	0	
4					•

```
In [11]: df = pd.get_dummies(df,drop_first=True) # only one column remove from each cat
df
```

Out[11]:

	Selling_Price	Present_Price	Kms_Driven	Owner	Age	Fuel_Type_Diesel	Fuel_Type_Petrol
C	3.35	5.59	27000	0	11	0	1
1	4.75	9.54	43000	0	12	1	0
2	7.25	9.85	6900	0	8	0	1
3	2.85	4.15	5200	0	14	0	1
4	4.60	6.87	42450	0	11	1	0
296	9.50	11.60	33988	0	9	1	0
297	4.00	5.90	60000	0	10	0	1
298	3.35	11.00	87934	0	16	0	1
299	11.50	12.50	9000	0	8	1	0
300	5.30	5.90	5464	0	9	0	1

301 rows × 9 columns

```
In [12]: # target is selling price
y = df['Selling_Price']
x = df.drop('Selling_Price',axis=1)
x.shape
```

Out[12]: (301, 8)

```
In [13]: y = df['Selling_Price']
x = df.drop('Selling_Price',axis=1)
y.shape
```

Out[13]: (301,)

```
In [30]: 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_stat
    x_train.shape
```

Out[30]: (240, 8)

```
In [16]: x_test.shape
```

Out[16]: (61, 8)

```
In [17]: y_train.shape
```

Out[17]: (240,)

```
In [18]: y_test.shape
Out[18]: (61,)
In [31]: from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.coef_
Out[31]: array([ 4.28802981e-01, -5.92316903e-06, -8.49009727e-01, -4.06411225e-01,
                 2.86056285e+00, 7.03103550e-01, -1.07477347e+00, -1.50897975e+00])
In [32]: |print(x.columns)
         Index(['Present_Price', 'Kms_Driven', 'Owner', 'Age', 'Fuel_Type_Diesel',
                'Fuel_Type_Petrol', 'Seller_Type_Individual', 'Transmission_Manual'],
               dtype='object')
In [33]: from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.intercept_
Out[33]: 6.850757321843104
In [52]: from sklearn import metrics
         import math
         y_pred = lr.predict(x_test)
         y_pred
Out[52]: array([10.57889241, 0.71754255, 4.23613904, 5.17855444,
                                                                    9.75533583,
                 4.20615652, 2.6748089, 7.63221623, 0.17032381,
                                                                    5.13283747,
                 6.15727726, 6.44545506, 2.11900126, 7.6644102,
                                                                    1.91710055,
                 1.71680788, 2.02672159, 1.85294787, 9.56717323,
                                                                    4.23076317,
                 1.48724915, 9.36189904, 1.46103587, 9.54572517,
                                                                    0.82954545,
                 8.32266051, 1.53698563, -3.16293717, 4.2127377,
                                                                    2.10385587,
                              3.71284702, 5.54538889, 7.6834828, -1.89757368,
                 3.42279444,
                 7.06591847, 8.46027656, 5.77704136, 6.10531922, 6.52919752,
                16.08564344, 2.07410694, 1.05462465, -0.44710144,
                                                                    7.06940454,
                             0.98609923, 7.08961114, 14.16420523,
                 6.73548589,
                                                                    3.00434598,
                 8.2871215 , -0.87862345, 8.8814273 , 1.15099681,
                                                                    2.11848057,
                -0.82641923, 0.69287067, 9.88583992, -0.45444272, -2.41051663,
                10.28627971])
In [53]:
         from sklearn import metrics
         import math
         y_pred = lr.predict(x_test)
         print("MAE : ",metrics.mean_absolute_error(y_test,y_pred))
         print("MSE : ",metrics.mean_squared_error(y_test,y_pred))
         print("RMSE : ",math.sqrt(metrics.mean_squared_error(y_test,y_pred)))
         print("R2score : ",metrics.r2 score(y test,y pred))
         MAE: 1.2453565634870642
         MSE: 2.750689262139753
         RMSE: 1.658520202511791
```

R2score: 0.8502332355855595

```
In [54]: # when column value is given
    import numpy as np
    test_data = np.array([6.7, 10000,0, 7, 1,0, 0, 1])
    test_data = test_data.reshape(-1,8)
    test_data

Out[54]: array([[6.7e+00, 1.0e+04, 0.0e+00, 7.0e+00, 1.0e+00, 0.0e+00, 0.0e+00]])

In [57]: import numpy as np
    test_data = np.array([6.7, 10000,0, 7, 1,0, 0, 1])
    test_data = test_data.reshape(-1,8)
    lr.predict(test_data)
Out[57]: array([8.17121013])
```

Advertising csv

```
In [58]:
         import pandas as pd
         df = pd.read_csv("advertising.csv")
         df.isna().sum()
Out[58]: TV
                      0
                      0
         Radio
         Newspaper
                      0
         Sales
         dtype: int64
In [66]: y = df["Sales"]
         x = df.drop("Sales",axis=1)
        from sklearn.model_selection import train_test_split
In [67]:
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
         x_train.shape
Out[67]: (160, 3)
In [68]: x test.shape
Out[68]: (40, 3)
In [69]: y_train.shape
Out[69]: (160,)
In [70]: y_test.shape
Out[70]: (40,)
```

```
In [74]:
         from sklearn.linear model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.coef
Out[74]: array([ 0.05413044,  0.1096093 , -0.00071742])
In [75]: from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.intercept_
Out[75]: 4.601987623229277
In [76]: | from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         y_pred = lr.predict(x_test)
         y_pred
Out[76]: array([14.06714329, 17.00172214, 20.74018701, 20.14958878, 21.0236248,
                16.76572535, 15.76327412, 8.03654982, 17.11011956, 21.34362676,
                11.28608236, 11.06766208, 15.39394793, 8.81462303, 24.55268079,
                17.07362544, 24.88331835, 10.42926552, 6.09130108, 24.92061384,
                 9.86831195, 13.35784443, 18.10206165, 10.148075 , 9.51161825,
                10.5969708 , 15.79896773 , 23.52774486 , 20.65344938 , 18.79356837 ,
                19.4127532 , 15.18136138, 19.20252368, 12.03755096, 20.51207209,
                15.53065475, 16.36903336, 24.50947859, 14.4166112, 12.55345811])
In [77]: from sklearn import metrics
         import math
         y_pred = lr.predict(x_test)
         print("MAE : ",metrics.mean_absolute_error(y_test,y_pred))
         print("MSE : ",metrics.mean_squared_error(y_test,y_pred))
         print("RMSE : ",math.sqrt(metrics.mean squared error(y test,y pred)))
         print("R2score : ",metrics.r2 score(y test,y pred))
                1.383308862610568
         MSE: 2.886920336419673
         RMSE: 1.6990939751584293
         R2score: 0.8982801816554089
```

Simple Ploynomial regression

```
In [78]: import pandas as pd
    df = pd.read_csv("temp.csv")
    df
```

Out[78]:

	sno	Temperature	Pressure
0	1	0	0.0002
1	2	20	0.0012
2	3	40	0.0060
3	4	60	0.0300
4	5	80	0.0900
5	6	100	0.2700

```
In [80]: df = df.drop("sno",axis=1)
df
```

Out[80]:

	Temperature	Pressure
0	0	0.0002
1	20	0.0012
2	40	0.0060
3	60	0.0300
4	80	0.0900
5	100	0.2700

```
In [81]: y = df["Pressure"]
x = df.drop("Pressure",axis=1)
```

```
In [83]: from sklearn.preprocessing import PolynomialFeatures
poly = PolynomialFeatures(degree=2)
    x_poly = poly.fit_transform(x)
    x_poly
```

```
In [84]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x_poly,y,train_size=0.2)
```

```
In [85]: from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr.fit(x_train,y_train)
lr.coef_
```

Out[85]: array([0., 0., 0.])

```
In [86]: lr.intercept_
```

Out[86]: 0.0002