nhn5dfexg

April 8, 2024

[1]:
```
# Aim: To perform Simple Linear Regression and Find out Coefficient of it.
```

[2]:
```
# Name: Mandar K Satpute
# Sub: Big Data Analytics (ET-Lab 2)
# Section : B
# Roll no: 54
```

[3]:
```
import os
```

[4]:
```
import pandas as pd
```

[5]:
```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import seaborn as sns
import matplotlib.pyplot as plt
```

[6]:
```
from sklearn.linear_model import LogisticRegression  # for Logistic Regression␣
 ↪algorithm
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris
from sklearn.linear_model import LinearRegression
from sklearn.metrics import accuracy_score
from sklearn.metrics import mean_squared_error, r2_score
```

[7]:
```
iris=load_iris()
X = iris.data  # Features
y = iris.target
dir(iris)
```

[7]:
```
['DESCR',
 'data',
 'data_module',
 'feature_names',
 'filename',
 'frame',
 'target',
 'target_names']
```

[8]:
```
os.getcwd()
```

[8]:
```
'C:\\Users\\hp\\Desktop\\BDA practicals(ET-2)'
```

[9]:
```
df=pd.read_csv("C://Users//hp//Desktop//IRIS.csv")
```

[10]:
```
df.head()
```

[10]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

[11]:
```
df.tail()
```

[11]:

	sepal_length	sepal_width	petal_length	petal_width	species
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

[12]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 #   Column        Non-Null Count  Dtype
---  ------        --------------  -----
 0   sepal_length  150 non-null    float64
 1   sepal_width   150 non-null    float64
 2   petal_length  150 non-null    float64
 3   petal_width   150 non-null    float64
 4   species       150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

[13]:
```
df.describe()
```

[13]:

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000

```
50%         5.800000     3.000000      4.350000     1.300000
75%         6.400000     3.300000      5.100000     1.800000
max         7.900000     4.400000      6.900000     2.500000
```

[14]:
```
df.isnull()
```

[14]:
```
     sepal_length  sepal_width  petal_length  petal_width  species
0           False        False         False        False    False
1           False        False         False        False    False
2           False        False         False        False    False
3           False        False         False        False    False
4           False        False         False        False    False
..            …            …             …            …        …
145         False        False         False        False    False
146         False        False         False        False    False
147         False        False         False        False    False
148         False        False         False        False    False
149         False        False         False        False    False

[150 rows x 5 columns]
```

[15]:
```
df.isna().sum()
```

[15]:
```
sepal_length    0
sepal_width     0
petal_length    0
petal_width     0
species         0
dtype: int64
```

[16]:
```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,␣
↪random_state=40)
```

[17]:
```
model = LinearRegression()

# Train the model using the training sets
model.fit(X_train, y_train)

# Make predictions using the testing set
y_pred = model.predict(X_test)

# Coefficients
print('Coefficients:', model.coef_)
```

```
Coefficients: [-0.1502982  -0.04339123  0.25345042  0.58205165]
```

[18]:
```
mse = mean_squared_error(y_test, y_pred)
print('Mean squared error: %.2f' % mse)

# Calculate coefficient of determination (R^2 score)
r2 = r2_score(y_test, y_pred)
print('Coefficient of determination (R^2 score): %.2f' % r2)
```

```
Mean squared error: 0.04
Coefficient of determination (R^2 score): 0.94
```

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