

5rfeuhliq

February 5, 2025

1 Machine Learning Part 2

```
[2]: import pandas as pd
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import FunctionTransformer
from mlxtend.feature_selection import SequentialFeatureSelector
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression

data = pd.read_csv("/Users/ratnadeepgurav/Desktop/AIDS/Study for carrer_
↳material/WSCUBE_Data_Analyst/ML/loan.csv")
data
```

```
[2]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	\
0	LP001002	Male	No	0	Graduate	No	
1	LP001003	Male	Yes	1	Graduate	No	
2	LP001005	Male	Yes	0	Graduate	Yes	
3	LP001006	Male	Yes	0	Not Graduate	No	
4	LP001008	Male	No	0	Graduate	No	
..	
609	LP002978	Female	No	0	Graduate	No	
610	LP002979	Male	Yes	3+	Graduate	No	
611	LP002983	Male	Yes	1	Graduate	No	
612	LP002984	Male	Yes	2	Graduate	No	
613	LP002990	Female	No	0	Graduate	Yes	

	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	\
0	5849	0.0	NaN	360.0	
1	4583	1508.0	128.0	360.0	
2	3000	0.0	66.0	360.0	
3	2583	2358.0	120.0	360.0	
4	6000	0.0	141.0	360.0	

```

..          ...          ...          ...          ...
609          2900          0.0          71.0          360.0
610          4106          0.0          40.0          180.0
611          8072          240.0          253.0          360.0
612          7583          0.0          187.0          360.0
613          4583          0.0          133.0          360.0

```

```

      Credit_History Property_Area Loan_Status
0              1.0      Urban      Y
1              1.0      Rural      N
2              1.0      Urban      Y
3              1.0      Urban      Y
4              1.0      Urban      Y
..          ...          ...          ...
609          1.0      Rural      Y
610          1.0      Rural      Y
611          1.0      Urban      Y
612          1.0      Urban      Y
613          0.0  Semiurban      N

```

[614 rows x 13 columns]

2 Feature Scaling Normalization

```

[75]: ms = MinMaxScaler()

ms.fit(data[["CoapplicantIncome"]])

data['CoapplicantIncome_MinMaxScaling'] = ms.
    ↪transform(data[['CoapplicantIncome']])

data.head(3)

```

```

[75]:   Loan_ID Gender Married Dependents Education Self_Employed \
0  LP001002  Male    No           0 Graduate           No
1  LP001003  Male   Yes           1 Graduate           No
2  LP001005  Male   Yes           0 Graduate           Yes

      ApplicantIncome  CoapplicantIncome  LoanAmount  Loan_Amount_Term \
0              5849              0.0           NaN           360.0
1              4583            1508.0          128.0           360.0
2              3000              0.0           66.0           360.0

      Credit_History Property_Area Loan_Status  CoapplicantIncome_MinMaxScaling
0              1.0      Urban      Y              0.000000
1              1.0      Rural      N              0.036192

```

2 1.0 Urban Y 0.000000

```
[76]: plt.figure(figsize=(12, 5))

# Before
plt.subplot(1, 2, 1)
plt.title("Before")
sns.distplot(data["CoapplicantIncome"], kde=True)

# After
plt.subplot(1, 2, 2)
plt.title("After")
sns.distplot(data["CoapplicantIncome_MinMaxScaling"], kde=True)

plt.show()
```

/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/2589777655.py:6:
UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

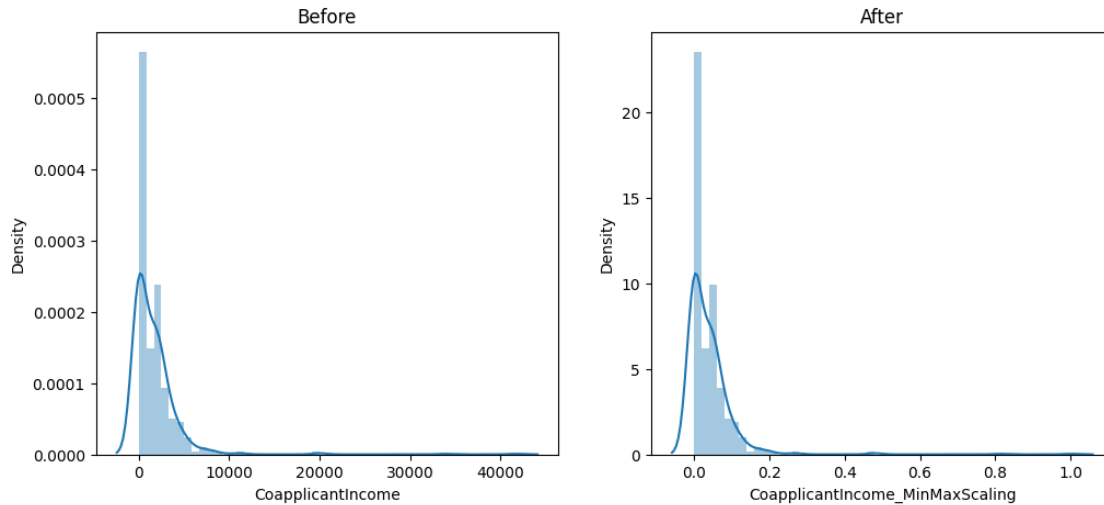
```
sns.distplot(data["CoapplicantIncome"], kde=True)
/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/2589777655.py:11
: UserWarning:
```

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data["CoapplicantIncome_MinMaxScaling"], kde=True)
```



3 Handle Duplicate Data

```
[77]: data.shape
```

```
[77]: (614, 14)
```

```
[27]: data.drop_duplicates(inplace = True)
```

```
[30]: data.shape
```

```
[30]: (614, 14)
```

4 Replace Data Types

```
[34]: data.isnull().sum()
```

```
[34]: Loan_ID          0
      Gender         13
      Married         3
      Dependents     15
      Education        0
      Self_Employed  32
      ApplicantIncome  0
      CoapplicantIncome 0
      LoanAmount      22
      Loan_Amount_Term 14
      Credit_History   50
      Property_Area    0
```

```
Loan_Status                                0
CoapplicantIncome_MinMaxScaling            0
dtype: int64
```

```
[40]: data['Dependents'].value_counts()
```

```
[40]: Dependents
0      360
1      102
2      101
3+       51
Name: count, dtype: int64
```

```
[57]: data["Dependents"].fillna(data["Dependents"].mode()[0],inplace = True)

data["Dependents"].replace("3+", "3",inplace = True)

print("\n\n Values in data :- \n\n",data["Dependents"].value_counts(),"\n\n")

data["Dependents"] = data["Dependents"].astype("int64")

data.info()
```

Values in data :-

```
Dependents
0      360
1      102
2      101
3       51
Name: count, dtype: int64
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 614 entries, 0 to 613
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Loan_ID               614 non-null   object
1   Gender                601 non-null   object
2   Married               611 non-null   object
3   Dependents            614 non-null   int64
4   Education             614 non-null   object
5   Self_Employed         582 non-null   object
6   ApplicantIncome       614 non-null   int64
```

```

7   CoapplicantIncome          614 non-null    float64
8   LoanAmount                 592 non-null    float64
9   Loan_Amount_Term           600 non-null    float64
10  Credit_History             564 non-null    float64
11  Property_Area              614 non-null    object
12  Loan_Status                614 non-null    object
13  CoapplicantIncome_MinMaxScaling 614 non-null    float64
dtypes: float64(5), int64(2), object(7)
memory usage: 67.3+ KB

```

```

/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/3797479550.py:1:
FutureWarning: A value is trying to be set on a copy of a DataFrame or Series
through chained assignment using an inplace method.
The behavior will change in pandas 3.0. This inplace method will never work
because the intermediate object on which we are setting values always behaves as
a copy.

```

For example, when doing `'df[col].method(value, inplace=True)'`, try using `'df.method({col: value}, inplace=True)'` or `df[col] = df[col].method(value)` instead, to perform the operation inplace on the original object.

```

data["Dependents"].fillna(data["Dependents"].mode()[0],inplace = True)
/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/3797479550.py:3:
FutureWarning: A value is trying to be set on a copy of a DataFrame or Series
through chained assignment using an inplace method.
The behavior will change in pandas 3.0. This inplace method will never work
because the intermediate object on which we are setting values always behaves as
a copy.

```

For example, when doing `'df[col].method(value, inplace=True)'`, try using `'df.method({col: value}, inplace=True)'` or `df[col] = df[col].method(value)` instead, to perform the operation inplace on the original object.

```
data["Dependents"].replace("3+", "3", inplace = True)
```

5 Function Transformer (Convert Non-distributable to distributable)

```
[91]: sns.distplot(data['CoapplicantIncome'])
plt.show()
```

```

/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/1526187818.py:1:
UserWarning:

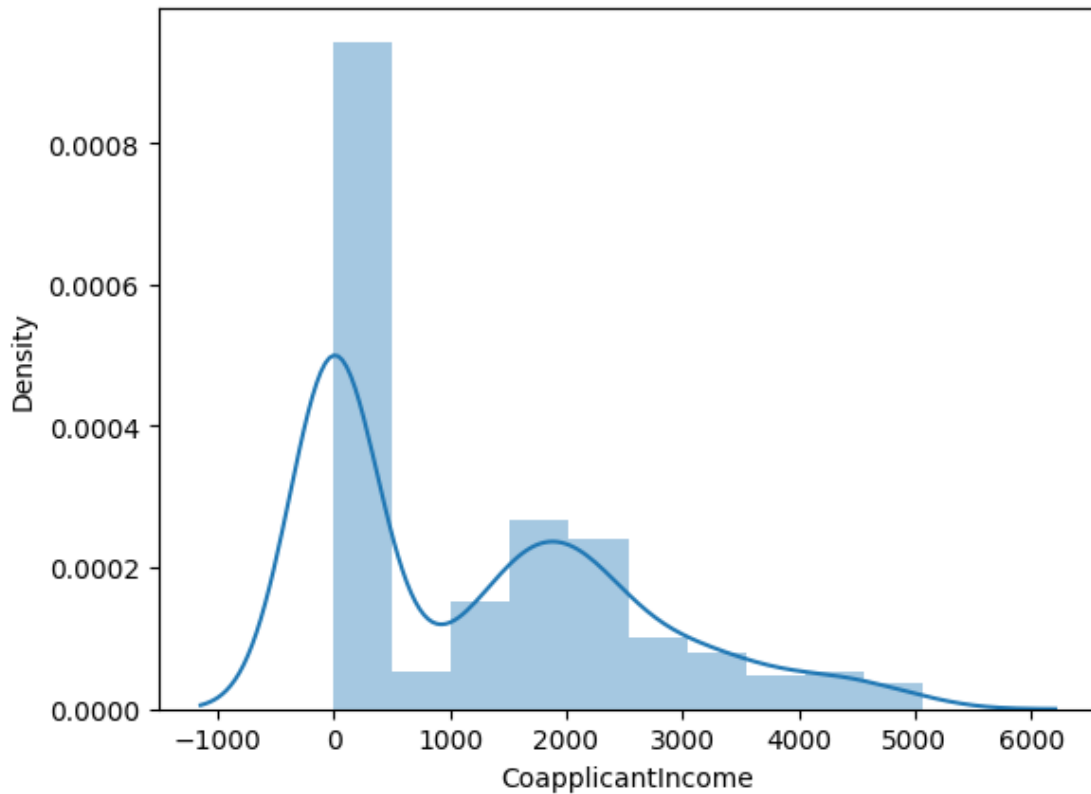
```

`'distplot'` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data['CoapplicantIncome'])
```



```
[90]: q1 = data['CoapplicantIncome'].quantile(0.25)
      q3 = data['CoapplicantIncome'].quantile(0.75)

      IQR = q3 - q1

      min_range = q1 - (1.25 * IQR)
      max_range = q3 + (1.25 * IQR)

      min_range, max_range

      data = data[data["CoapplicantIncome"] < max_range]
```

6 Use Function transformer

```
[103]: ft = FunctionTransformer(func = np.log1p)

ft.fit(data[["CoapplicantIncome"]])
data["Income Function_Transformer"] = ft.transform(data[["CoapplicantIncome"]])

plt.figure(figsize = (13,5))

plt.subplot(1,2,1)
sns.distplot(data["CoapplicantIncome"])
plt.title("Before")

plt.subplot(1,2,2)
sns.distplot(data["Income Function_Transformer"])
plt.title("After")

plt.show()
```

/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/3596817702.py:9:
UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

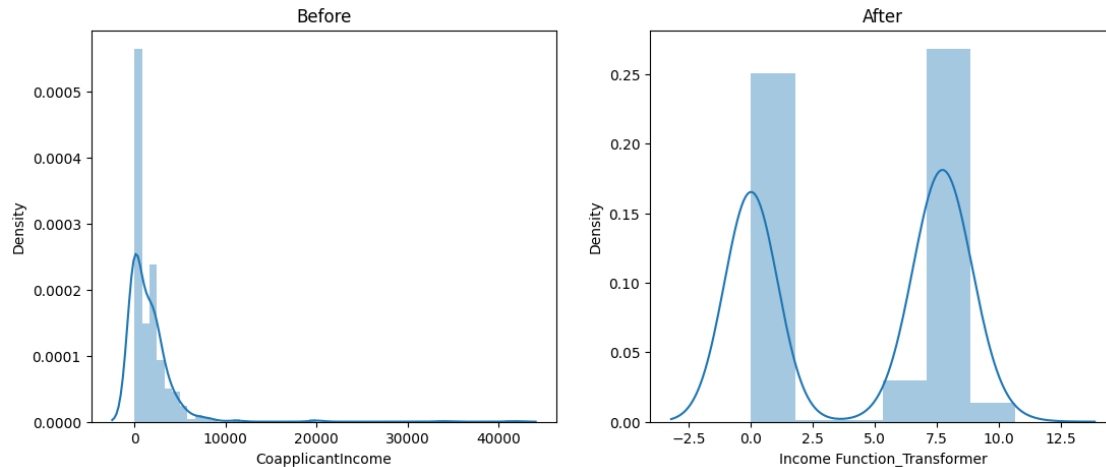
```
sns.distplot(data["CoapplicantIncome"])
/var/folders/c_/rbrshmgx64b9ch2skklfhfbw0000gn/T/ipykernel_1549/3596817702.py:14
: UserWarning:
```

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data["Income Function_Transformer"])
```

7 Feature Selection Technique

8 Forward Elimination

```
[15]: data1 = pd.read_csv("/Users/ratnadeepgurav/Desktop/AIDS/Study for career_
    ↳material/WSCUBE_Data_Analyst/ML/diabetes.csv")

x = data1.iloc[:, :-1]
y = data1["Outcome"]

print(x.shape)

lr = LogisticRegression()

fs = SequentialFeatureSelector(lr, k_features = 5 , forward = True)
fs.fit(x, y)

print("\n\n Features :- ", fs.feature_names, "\n\n")
print("\n\n K - Features :-", fs.k_feature_names_, "\n\n")
print("\n\n K - Score :- ", fs.k_score_, "\n\n")
```

(768, 8)

Features :- ['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness',
'Insulin', 'BMI', 'DiabetesPedigreeFunction', 'Age']

```
K - Features :- ('Pregnancies', 'Glucose', 'Insulin', 'BMI', 'Age')
```

```
K - Score :- 0.7708768355827178
```

9 Regression Analysis (Supervised Learning)

```
[6]: data_clg = pd.read_csv("/Users/ratnadeepgurav/Desktop/AIDS/Study for carrer_
    ↳material/WSCUBE_Data_Analyst/ML/placement.csv")

data_clg.drop(columns = ["iq"],inplace = True)

data_clg.head(3)
```

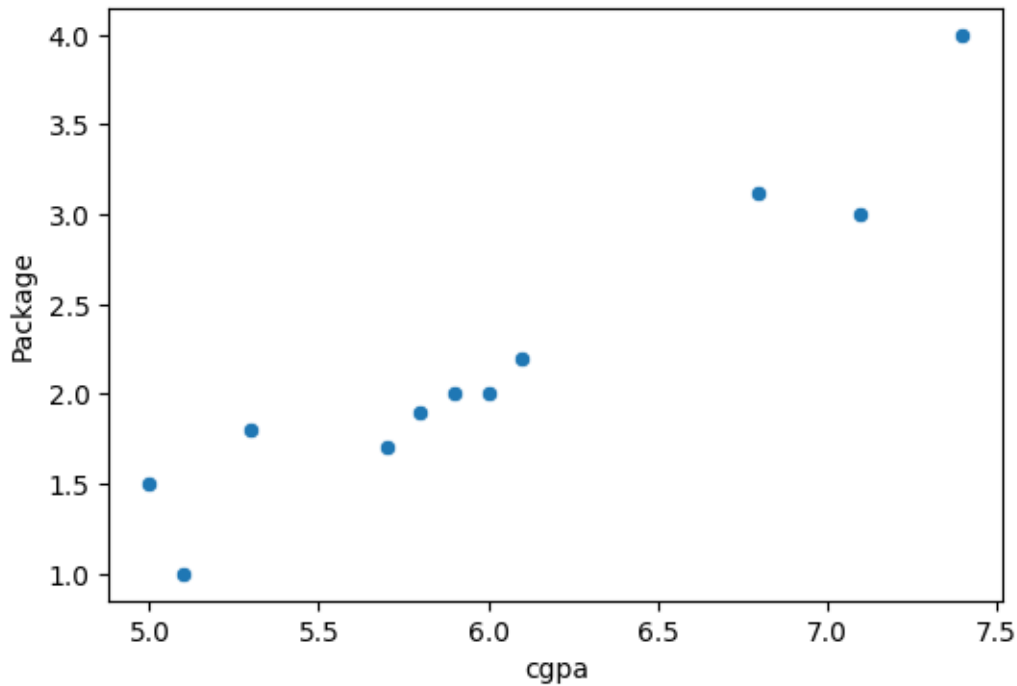
```
[6]:
```

	Unnamed: 0	cgpa	Package
0	0	6.8	3.12
1	1	5.9	2.00
2	2	5.3	1.80

```
[8]: data_clg.isnull().sum()
```

```
[8]: Unnamed: 0    0
cgpa           0
Package        0
dtype: int64
```

```
[13]: plt.figure(figsize = (6,4))
sns.scatterplot(x = "cgpa", y = "Package",data = data_clg)
plt.show()
```



```
[63]: x1 = data_clg[["cgpa"]]
      y1 = data_clg["Package"]

      x_train,x_test,y_train,y_test = train_test_split(x1,y1,test_size=0.
      ↪2,random_state = 42)

      lr = LinearRegression()

      lr.fit(x_train,y_train)

      lr.predict([[5.3]])

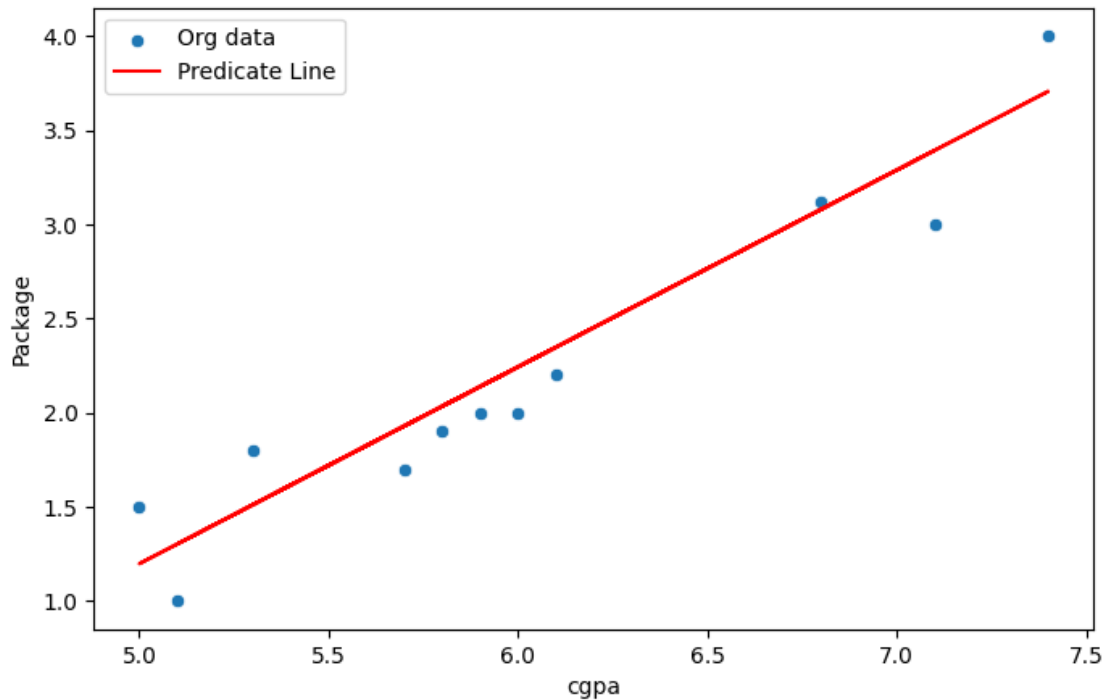
      lr.score(x_test,y_test)*100
```

```
/opt/homebrew/Cellar/jupyterlab/4.2.3/libexec/lib/python3.12/site-
packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid
feature names, but LinearRegression was fitted with feature names
  warnings.warn(
```

```
[63]: 91.32265824841723
```

```
[37]: y_pred = lr.predict(x1)
```

```
[44]: plt.figure(figsize = (8,5))
sns.scatterplot(x = "cgpa", y = "Package",data = data_clg)
plt.plot(data_clg["cgpa"],y_pred,c = "red")
plt.legend(["Org data","Predicate Line"])
plt.show()
```

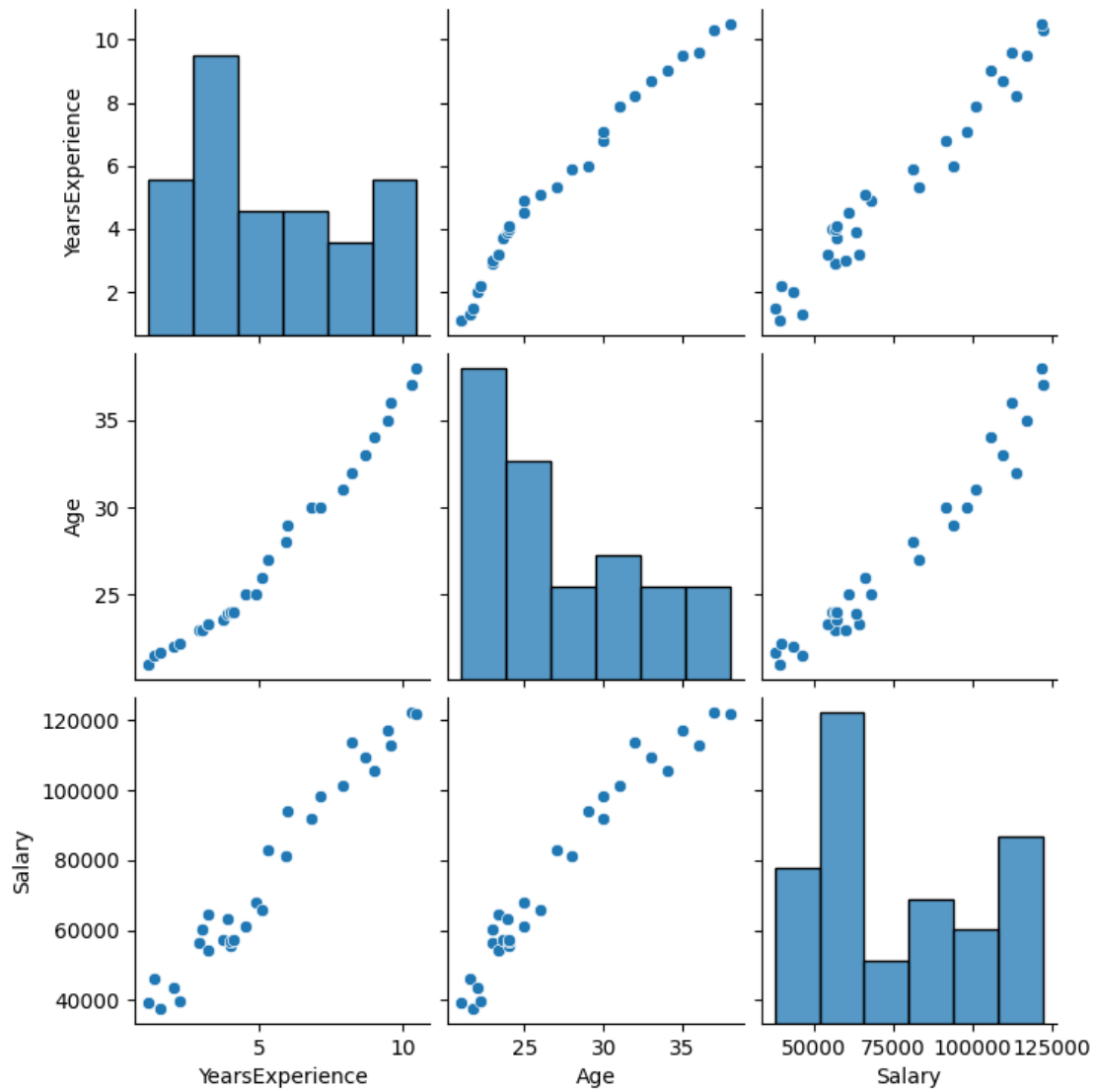


10 Multiple Linear Regression

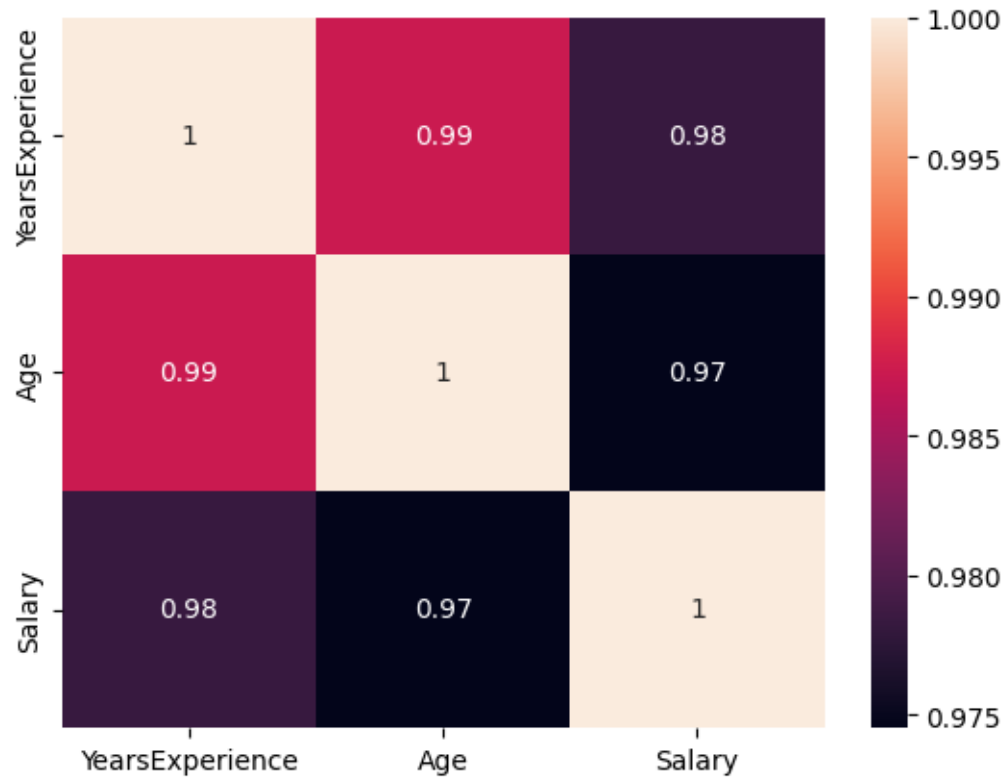
```
[3]: data3 = pd.read_csv("/Users/ratnadeepgurav/Desktop/AIDS/Study for carrer_
↳material/WSCUBE_Data_Analyst/ML/Salary_Data.csv")

data3.isnull().sum()

sns.pairplot(data = data3)
plt.show()
```



```
[4]: sns.heatmap(data=data3.corr(),annot = True)
plt.show()
```



```
[5]: x3 = data3.iloc[:, :-1]
      y3 = data3["Salary"]

      x_train, x_test, y_train, y_test = train_test_split(x3, y3, test_size=0.
      ↪ 2, random_state = 42)

      lr = LinearRegression()

      lr.fit(x_train, y_train)

      lr.score(x_test, y_test)*100

      lr.predict(x3)
```

```
[5]: array([ 38675.56314937,  40935.75217728,  42425.68560928,  45637.01545868,
            47126.94889069,  52598.467768  ,  53086.6826187  ,  54833.367916  ,
            54833.367916  ,  58044.6977654  ,  59791.38306271,  60536.34977871,
            60536.34977871,  61024.56462941,  65544.94268522,  67497.80208802,
            71041.75044243,  74585.69879684,  80082.50655405,  83138.24005776,
            89611.47751637,  91076.12206847,  97549.35952708, 101581.52273219,
            106590.1156387  , 110622.27884381, 115630.87175032, 118686.60525402,
            124671.62786194, 128215.57621634])
```

[]: