

1.Importing Necessary Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import sklearn
import scipy
import seaborn as sns
import missingno as msno
```

2.Importing the Dataset

```
data=pd.read_csv("E:\BM_Project\weatherAUS.csv")
```

3.Analyse the data

```
data.describe()

Out[86]:
```

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustSpeed	WindSpeed9am	WindSpeed3pm	Humidity9am	Humidity3pm	Pressure9am	Pressure3pm
count	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000	142193.000000
mean	12.131807	23.174186	2.326738	3.129340	3.988338	37.377792	13.869248	18.292855	67.984915	50.175964	917.357684	917.357684
std	6.440548	7.194768	8.426426	4.166674	4.686855	16.433198	8.954477	9.075870	20.416089	22.071424	303.403553	303.403553
min	-8.500000	-4.800000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	7.500000	17.900000	0.000000	0.000000	0.000000	30.000000	7.000000	11.000000	56.000000	35.000000	1011.000000	1011.000000
50%	12.000000	22.000000	0.000000	1.600000	0.200000	37.000000	13.000000	17.000000	70.000000	51.000000	1016.700000	1016.700000
75%	16.800000	28.000000	0.600000	5.400000	8.700000	46.000000	19.000000	24.000000	83.000000	65.000000	1021.800000	1021.800000
max	33.900000	48.100000	371.000000	145.000000	14.500000	135.000000	130.000000	87.000000	100.000000	100.000000	1041.000000	1041.000000

```
data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 142193 entries, 0 to 142192
Data columns (total 14 columns):
#   Column              Non-Null Count  Dtype
---  ---
0   Date                 142193 non-null    object
1   Location             142193 non-null    object
2   MinTemp              142193 non-null    float64
3   MaxTemp              142193 non-null    float64
4   Rainfall             142193 non-null    float64
5   Evaporation           142193 non-null    float64
6   Sunshine             142193 non-null    float64
7   WindGustDir          142193 non-null    object
8   WindGustSpeed        142193 non-null    int64
9   WindDir9am           142193 non-null    object
10  WindDir3pm           142193 non-null    object
11  WindSpeed9am         142193 non-null    int64
12  WindSpeed3pm         142193 non-null    int64
13  Humidity9am          142193 non-null    int64
14  Humidity3pm          142193 non-null    int64
15  Pressure9am          142193 non-null    float64
16  Pressure3pm          142193 non-null    float64
17  Cloud9am             142193 non-null    int64
18  Cloud3pm             142193 non-null    int64
19  Temp9am              142193 non-null    float64
20  Temp3pm              142193 non-null    float64
21  RainToday            142193 non-null    object
22  RISK_MM              142193 non-null    float64
23  RainTomorrow         142193 non-null    object
dtypes: float64(10), int64(7), object(7)
memory usage: 26.0+ MB
```

```
data.head()

Out[88]:
```

Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	...	Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm
0 12-2008	Albury	13.4	22.9	0.6	0.0	0.0	W	44	W	...	22	1007.7	1007.1	8	0
1 02-2008	Albury	7.4	25.1	0.0	0.0	0.0	WNW	44	NNW	...	25	1010.6	1007.8	0	0
2 03-2008	Albury	12.9	25.7	0.0	0.0	0.0	WSW	46	W	...	30	1007.6	1008.7	0	2
3 04-2008	Albury	9.2	28.0	0.0	0.0	0.0	NE	24	SE	...	16	1017.6	1012.8	0	0
4 12-2008	Albury	17.5	32.3	1.0	0.0	0.0	W	41	ENE	...	33	1010.8	1006.0	7	8

5 rows × 16 columns

```
data.shape

Out[89]: (142193, 16)
```

4.Handling Missing values

Checking whether Null values exit

```
data.isnull().sum()

Out[90]:
```

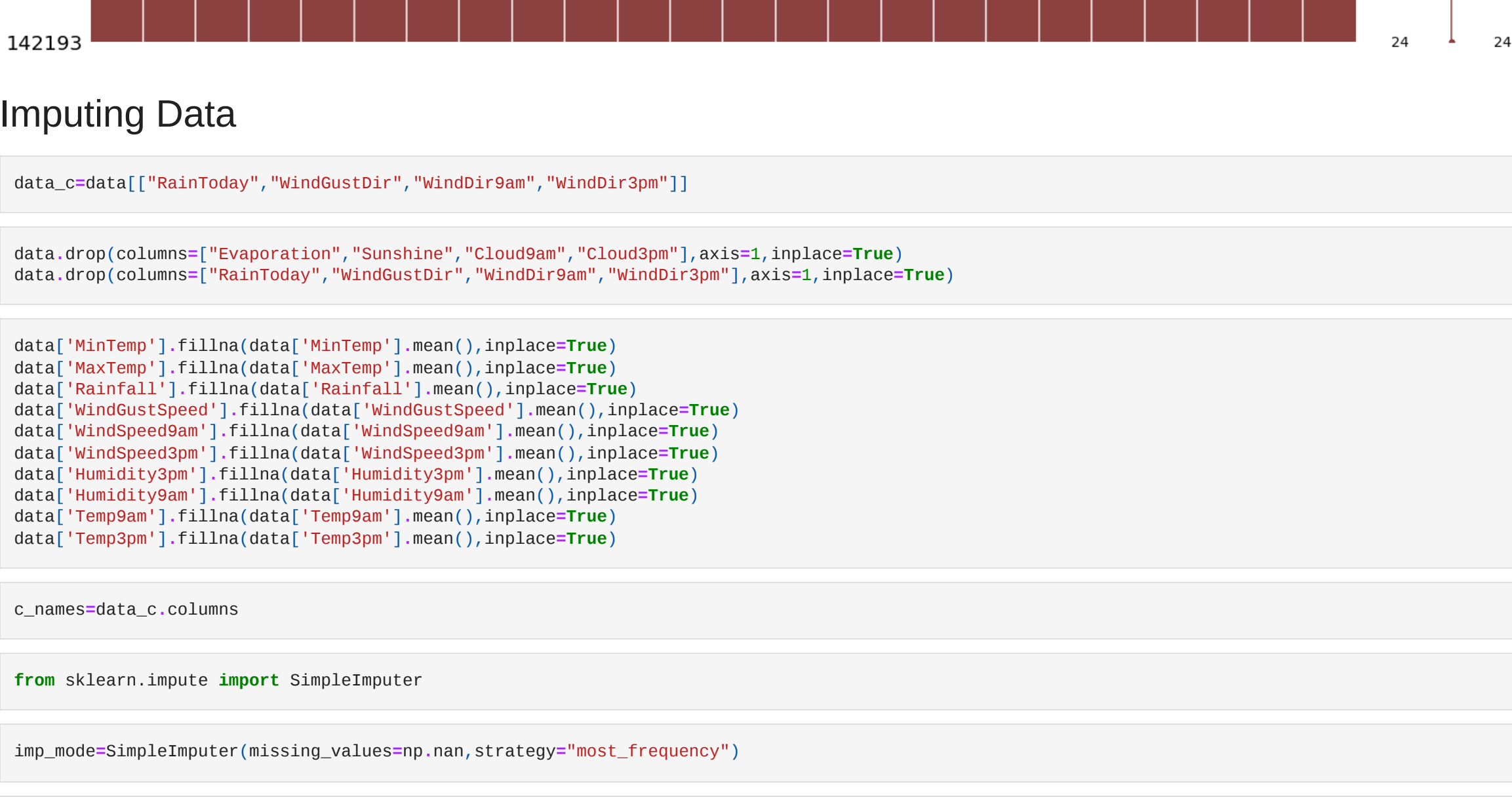
Date	0
Location	0
MinTemp	0
MaxTemp	0
Rainfall	0
Evaporation	0
Sunshine	0
WindGustDir	0
WindGustSpeed	0
WindDir9am	0
WindDir3pm	0
WindSpeed9am	0
WindSpeed3pm	0
Humidity9am	0
Humidity3pm	0
Pressure9am	0
Pressure3pm	0
Cloud9am	0
Cloud3pm	0
Temp9am	0
Temp3pm	0
RainToday	0
RISK_MM	0
RainTomorrow	0
dtype:	int64

Dealing with missing values

```
import missingno as msno

msno.matrix(data,color=(0.55,0.255,0.255),fontsize=16)

<AxesSubplot:~>
```



Imputing Data

```
data_data[data["RainToday"]=="WindGustDir", "WindDir9am", "WindDir3pm"]

data.drop(columns=["Evaporation", "Sunshine", "Cloud9am", "Cloud3pm"], axis=1, inplace=True)
data.drop(columns=["RainToday", "WindGustDir", "WindDir9am", "WindDir3pm"], axis=1, inplace=True)

data["MinTemp"].fillna(data["MinTemp"].mean(), inplace=True)
data["MaxTemp"].fillna(data["MaxTemp"].mean(), inplace=True)
data["Rainfall"].fillna(data["Rainfall"].mean(), inplace=True)
data["WindGustSpeed"].fillna(data["WindGustSpeed"].mean(), inplace=True)
data["WindSpeed9am"].fillna(data["WindSpeed9am"].mean(), inplace=True)
data["WindSpeed3pm"].fillna(data["WindSpeed3pm"].mean(), inplace=True)
data["Humidity9am"].fillna(data["Humidity9am"].mean(), inplace=True)
data["Humidity3pm"].fillna(data["Humidity3pm"].mean(), inplace=True)
data["Temp9am"].fillna(data["Temp9am"].mean(), inplace=True)
data["Temp3pm"].fillna(data["Temp3pm"].mean(), inplace=True)

c_names=data.c.columns

from sklearn.impute import SimpleImputer

imp_model=SimpleImputer(missing_values=np.nan, strategy="most_frequent")

data_c=pd.DataFrame(data.c, columns=c_names)

data=pd.concat([data_data,c], axis=1)
```

5.Data Visualization

Data Correlation

```
data.corr()

Out[101]:
```

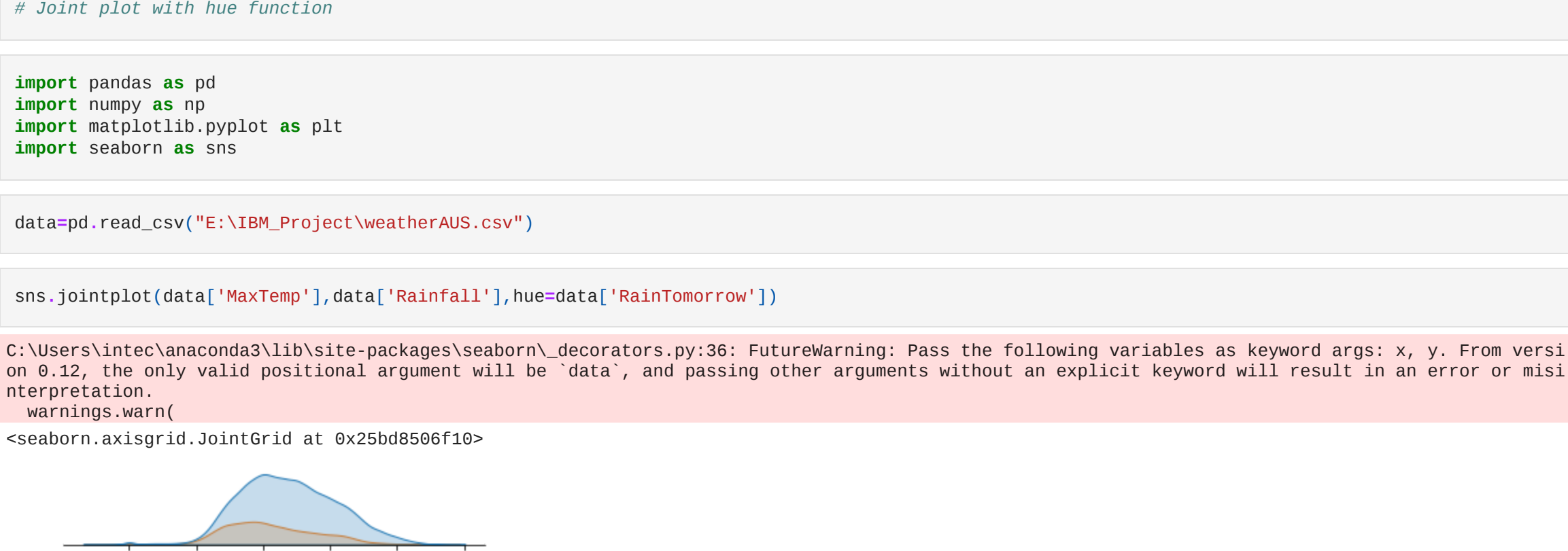
	MinTemp	MaxTemp	Rainfall	WindGustSpeed	WindSpeed9am	WindSpeed3pm	Humidity9am	Humidity3pm	Pressure9am	Pressure3pm	Temp9am	Temp3pm	RISK_MM
MinTemp	1.000000	0.722484	0.022444	0.130381	0.181565	0.173819	-0.180448	-0.000596	0.128941	0.127805	0.691762	0.627278	0.12141
MaxTemp	0.722484	1.000000	-0.072987	0.086229	0.021062	0.061310	-0.431398	-0.459034	0.096821	0.096162	0.666982	0.678366	-0.04374
Rainfall	0.022444	-0.072987	1.000000	0.094167	0.084768	0.052425	0.202095	0.212678	-0.004443	0.009495	-0.077041	0.30011	0.00000
WindGustSpeed	0.130381	0.086229	0.094167	1.000000	0.520705	0.608283	-0.167990	-0.001323	0.175633	0.177270	0.121768	0.113040	0.11701
WindSpeed9am	0.181565	0.021062	0.084768	0.520705	1.000000	0.513031	-0.234094	-0.013411	0.130373	0.130637	0.137968	0.106267	0.06652
WindSpeed3pm	0.173819	0.061310	0.052425	0.608283	0.513031	1.000000	-0.116146	0.081109	0.187028	0.202051	0.165344	0.106350	0.04377
Humidity9am	0.180448	-0.431398	0.202095	-0.167990	-0.234094	-0.116146	1.000000	0.813533	-0.007560	-0.014754	-0.366070	0.389508	0.15252
Humidity3pm	0.000596	-0.459034	0.009495	-0.001323	-0.013411	0.081109	0.613533	1.000000	0.015457	0.029630	-0.056036	0.347502	0.27851
Pressure9am	0.128941	0.096821	-0.004443	0.175633	0.130373	0.187028	-0.007560	0.015457	1.000000	0.983630	0.129621	0.137550	-0.00547
Pressure3pm	0.127805	0.096162	-0.004443	0.177270	0.130537	0.202051	-0.014754	0.029630	0.983630	1.000000	0.121841	0.144317	-0.00562
Temp9am	0.691762	0.666982	0.009495	0.121758	0.137968	0.164344	-0.366070	-0.205338	0.129621	0.121341	1.000000	0.764786	0.04501
Temp3pm	0.627278	0.678366	0.077041	0.113040	0.106267	0.106350	-0.389508	-0.347502	0.137550	0.144317	0.764786	1.000000	-0.00862
RISK_MM	0.121411	-0.043754	0.300183	0.117074	0.066521	0.043727	0.152556	0.278517	-0.005457	-0.005636	0.045691	-0.008622	1.00000

Heat map

```
cor=data.corr()

sns.heatmap(data=cor,xticklabels=cor.columns.values,yticklabels=cor.columns.values)

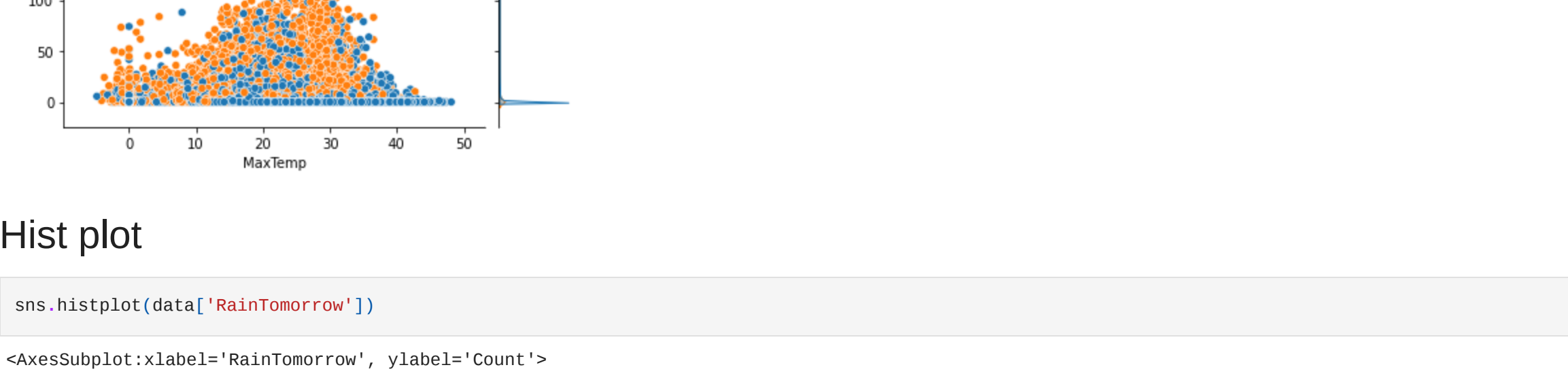
<AxesSubplot:~>
```



Box plot

```
data.boxplot()

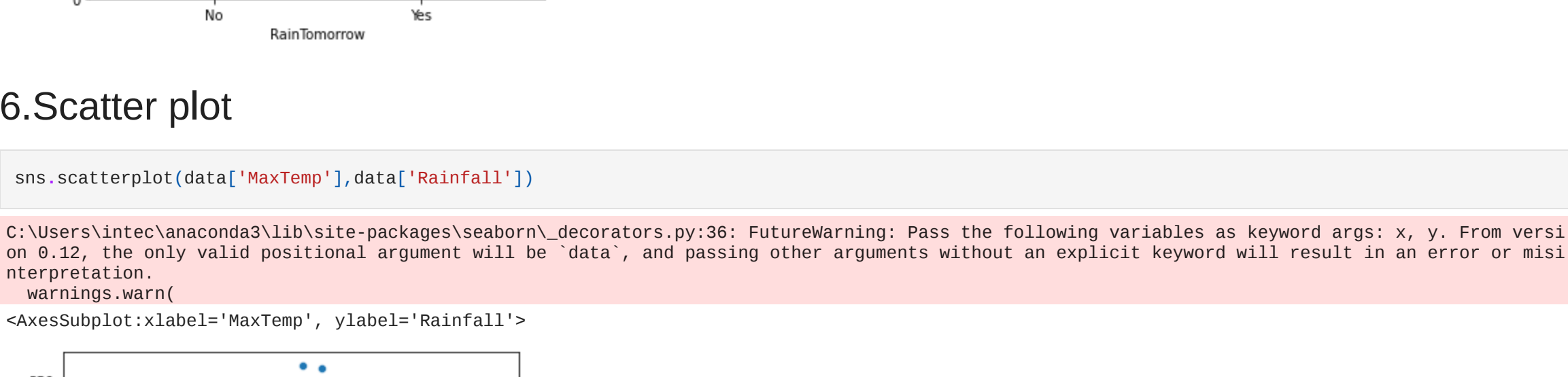
<AxesSubplot:~>
```



Joint plot

```
sns.jointplot(data["MinTemp"],data["Rainfall"])

C:\Users\intecanaconda\lib\site-packages\seaborn\decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From vers
on 0.12.0, the only valid positional argument will be 'data', and passing other arguments without an explicit keyword will result in an error or missi
ng interpretation.
  warnings.warn(
<seaborn.axisgrid.JointGrid at 0x25bbb1b5c08>
```



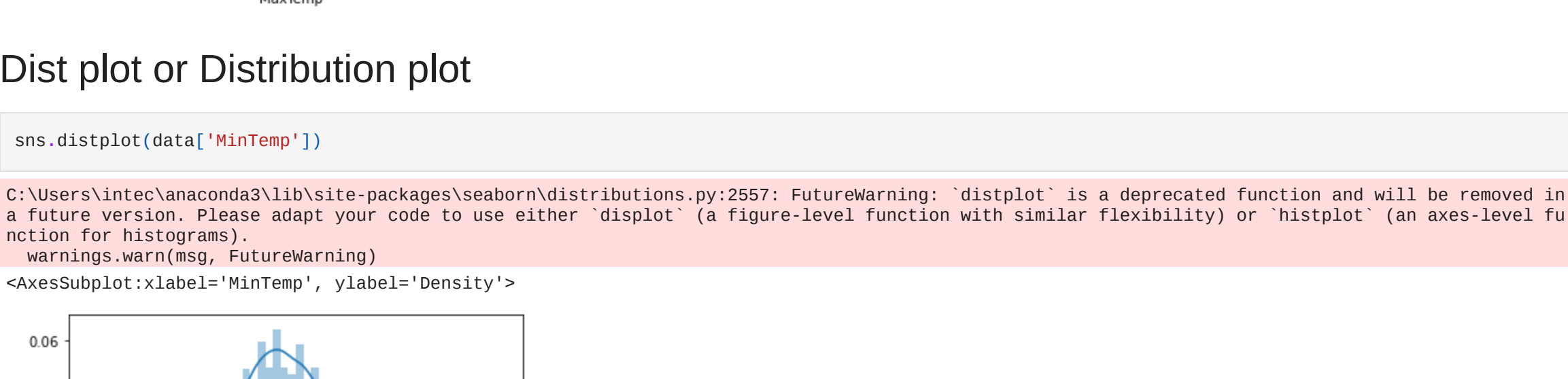
Joint plot with hue function

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

data=pd.read_csv("E:\BM_Project\weatherAUS.csv")

sns.jointplot(data["MaxTemp"],data["Rainfall"],hue=data["RainTomorrow"])

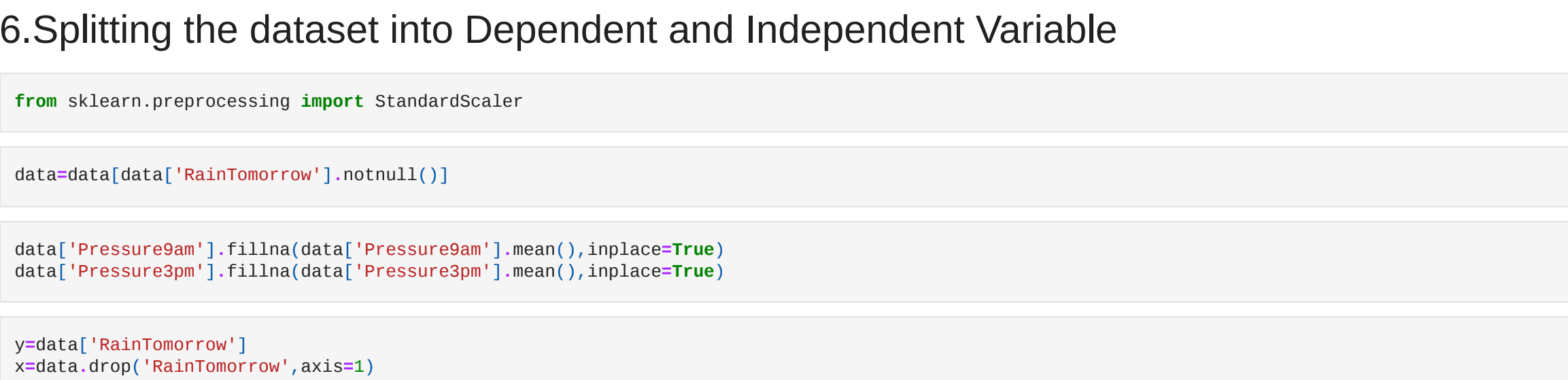
C:\Users\intecanaconda\lib\site-packages\seaborn\decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From vers
on 0.12.0, the only valid positional argument will be 'data', and passing other arguments without an explicit keyword will result in an error or missi
ng interpretation.
  warnings.warn(
<seaborn.axisgrid.JointGrid at 0x25bbb1b5c08>
```



Hist plot

```
sns.histplot(data["RainTomorrow"])

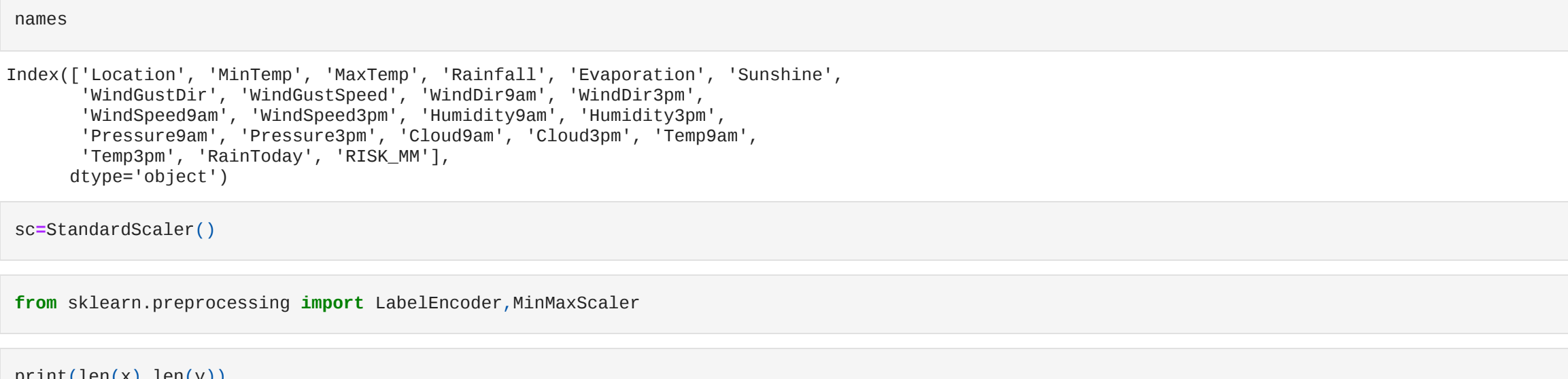
<AxesSubplot:~>
```



6.Scatter plot

```
sns.scatterplot(data["MaxTemp"],data["Rainfall"])

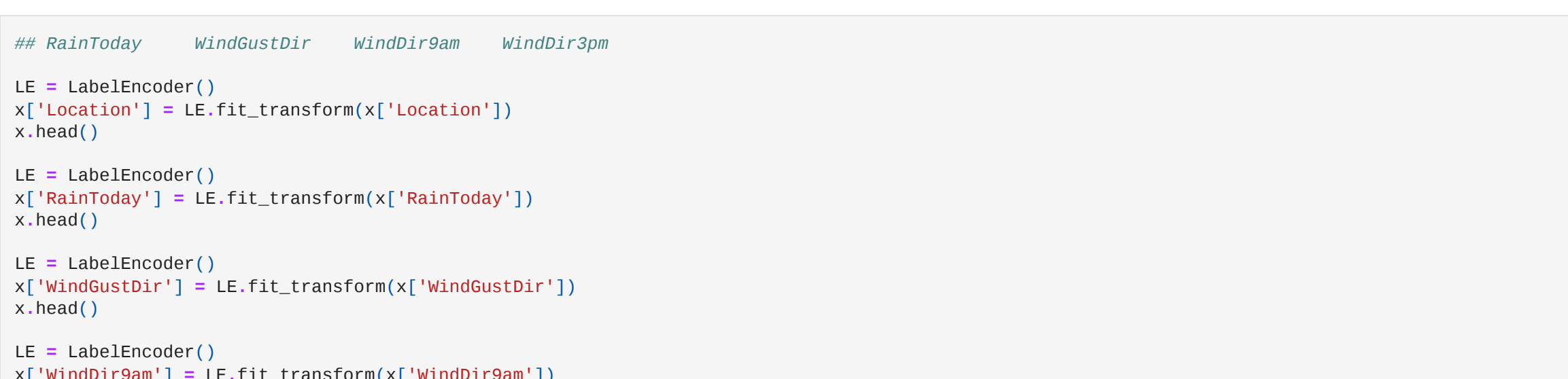
C:\Users\intecanaconda\lib\site-packages\seaborn\decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From vers
on 0.12.0, the only valid positional argument will be 'data', and passing other arguments without an explicit keyword will result in an error or missi
ng interpretation.
  warnings.warn(
<AxesSubplot:~>
```



Dist plot or Distribution plot

```
sns.distplot(data["MinTemp"])

C:\Users\intecanaconda\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: 'distplot' is a deprecated function and will be removed in
a future version. Please adapt your code to use either 'displot' (a figure-level function with similar flexibility) or 'histplot' (an axes-level fu
nction for histograms).
  warnings.warn(FutureWarning)
<AxesSubplot:~>
```



6.Splitting the dataset into Dependent and Independent Variable

```
from sklearn.preprocessing import StandardScaler

data=data[data["RainTomorrow"]>0]

data["Pressure9am"].fillna(data["Pressure9am"].mean(),inplace=True)
data["Pressure3pm"].fillna(data["Pressure3pm"].mean(),inplace=True)

y=data["RainTomorrow"]
x=data.drop(["RainTomorrow",axis=1)

set(y)

{'No', 'Yes'}

x=x.drop(["Date",axis=1)

names=x.columns

names

Location
MinTemp
MaxTemp
Rainfall
Evaporation
Sunshine
WindGustDir
WindGustSpeed
WindDir9am
WindDir3pm
WindSpeed9am
WindSpeed3pm
Humidity9am
Humidity3pm
Pressure9am
Pressure3pm
Cloud9am
Cloud3pm
Temp9am
Temp3pm
RainToday
RISK_MM
dtype: object
```

```
sc=StandardScaler()

from sklearn.preprocessing import LabelEncoder,MinMaxScaler

print(len(x),len(y))

142193 142193
```

```
from sklearn.preprocessing import LabelEncoder, MinMaxScaler

print(len(x),len(y))

142193 142193
```

```
LE = LabelEncoder()
x["Location"] = LE.fit_transform(x["Location"])
x.head()

LE = LabelEncoder()
x["RainToday"] = LE.fit_transform(x["RainToday"])
x.head()

LE = LabelEncoder()
x["WindGustDir"] = LE.fit_transform(x["WindGustDir"])
x.head()

LE = LabelEncoder()
x["WindDir9am"] = LE.fit_transform(x["WindDir9am"])
x.head()

LE = LabelEncoder()
x["WindDir3pm"] = LE.fit_transform(x["WindDir3pm"])
x.head()
```

```
Location MinTemp MaxTemp Rainfall Evaporation Sunshine WindGustDir WindGustSpeed WindDir9am WindDir3pm ... Humidity9am Humidity3pm Pressure9am Pressure3pm
0 2 0.196908 -0.038109 -0.204820 -0.751043 -0.850637 14 0.402991 14 15 ... -0.147682 -1.276568 0.297764 0.303160
1 2 0.734993 0.267670 -0.276125 -0.751043 -0.850637 15 0.402991 7 16 ... -1.174810 -1.240605 0.307322 0.303475
2 2 0.119275 0.531064 -0.276125 -0.751043 -0.850637 16 0.524696 14 16 ... -1.486957 -0.914107 0.297435 0.308451
3 2 0.455212 0.670742 -0.276125 -0.751043 -0.850637 5 -0.814062 10 1 ... -1.125829 -1.548413 0.330394 0.320011
4 2 0.835952 1.268400 -0.157450 -0.751043 -0.850637 14 0.220433 2 8 ... 0.686476 -0.778184 0.307982 0.299522
```

```
LE = LabelEncoder()
y=pd.factorize(y)
y = LE.fit_transform(y)

C:\Users\intecanaconda\lib\site-packages\sklearn\preprocessing\_label.py:115: DataConversionWarning: A column-vector y was passed when a 1d array
was expected. Please change the shape of y to (n_samples, ), for example using.ravel().
  y = column_or_1d(y, warn=True)

print(len(x),len(y))

142193 142193
```

```
sc=sc.fit_transform(x)

x=sc.fit_transform(x)

x[5]
```

```
array([[ 1.5276045, -0.19690834, -0.2048206, -0.751043, -0.2042066, -0.75104289,
-0.85063691, 1.17249935, -0.49209181, 1.29872549, 1.3614584,
0.68466812, 0.62882836, 0.14768248, -1.27656756, 0.29774422,
0.30316051, 1.07626959, -0.90606966, -0.90609923, 0.07696059,
-0.48941675, -0.27844993],
[ 1.5276045, -0.73499294, 0.26766959, -0.27612488, -0.75104289,
-0.85063691, 1.17249923, -0.49209181, -0.34767966, 1.57244417,
-1.10216181, 0.48946384, -1.17480996, -1.34864472, 0.30732249,
0.30547456, -0.88184222, -0.85063696, -0.84847489, -0.49452087,
-0.48941675, -0.27844993],
[ 1.5276045, -0.19690834, 0.35106379, -0.27612488, -0.75104289,
-0.85063691, 1.17249935, 1.22828666, -1.54841324, 0.33039418,
0.32010653, -0.88184222, -0.85063696, -0.84847489, 0.69837321,
-0.48941675, -0.27844993],
[ 1.5276045, -0.73499294, 0.26766959, -0.27612488, -0.75104289,
-0.85063691, 1.17249935, 1.22828666, -1.54841324, 0.33039418,
0.32010653, -0.88184222, -0.85063696, -0.84847489, 0.69837321,
-0.48941675, -0.27844993],
[ 1.5276045, -0.73499294, 0.26766959, -0.27612488, -0.75104289,
-0.85063691, 1.17249935, 1.22828666, -1.54841324, 0.33039418,
0.32010653, -0.88184222, -0.85063696, -0.84847489, 0.69837321,
-0.48941675, -0.27844993]])
```

8.Splitting the data into Train and Test

```
from sklearn import model_selection

x_train,x_test,y_train,y_test=model_selection.train_test_split(x,y,test_size=0.2,random_state=0)
```

9.Training and Testing the model

```
from sklearn.ensemble import RandomForestClassifier

from sklearn.ensemble import GradientBoostingClassifier
```

```
rfc=RandomForestClassifier()

gbc=GradientBoostingClassifier()

np.any(np.isnan(x))

False

gbc.fit(x_train,y_train)

GradientBoostingClassifier()

RFC.fit(x_train,y_train)

RandomForestClassifier()
```

```
x.isnull().any()

False
Location False
MinTemp False
MaxTemp False
Rainfall False
Evaporation False
Sunshine False
WindGustDir False
WindGustSpeed False
WindDir9am False
WindDir3pm False
WindSpeed9am False
WindSpeed3pm False
Humidity9am False
Humidity3pm False
Pressure9am False
Pressure3pm False
Cloud9am False
Cloud3pm False
Temp9am False
Temp3pm False
RainToday False
RainTomorrow False
dtype: bool
```

```
Location Location False
MinTemp MinTemp False
MaxTemp MaxTemp False
Rainfall Rainfall False
Evaporation Evaporation False
Sunshine Sunshine False
WindGustDir WindGustDir False
WindGustSpeed WindGustSpeed False
WindDir9am WindDir9am False
WindDir3pm WindDir3pm False
WindSpeed9am WindSpeed9am False
WindSpeed3pm WindSpeed3pm False
Humidity9am Humidity9am False
Humidity3pm Humidity3pm False
```