1.Import necessary library

import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

%matplotlib inline

import seaborn as sns

from sklearn import preprocessing

from sklearn import model_selection

from sklearn import metrics

from sklearn import linear_model

from sklearn import ensemble

from sklearn import tree

from sklearn import svm

import xgboost

from google.colab import drive

drive.mount('/content/drive')

Mounted at /content/drive

df=pd.read_csv("/content/drive/MyDrive/Colab Notebooks/Rainfall weather.csv")

2.analyse the data

df.head()

Date	Location	MinTemp	MaxTemp		Rainfall Evaporation			Sunshine		
	Wind Gust Dir	WindGustSpeed WindDi		ir9am		Humidity9am		Humidity3pm		
	Pressure9am	Pressure3pm	Cloud9a	am	Cloud3	pm	Temp9	am	Temp3	pm
	RainToday	RainTomorrow								
0	2008-12-01	Albury 13.4	22.9	0.6	NaN	NaN	W	44.0	W	
	71.0 22.0	1007.7 1007.1	8.0	NaN	16.9	21.8	No	No		

1	2008-1	2-02	Albury	7.4	25.1	0.0	NaN	NaN	WNW	44.0	NNW	
	44.0	25.0	1010.6	1007.8	NaN	NaN	17.2	24.3	No	No		
2	2008-1	2-03	Albury	12.9	25.7	0.0	NaN	NaN	WSW	46.0	W	
	38.0	30.0	1007.6	1008.7	NaN	2.0	21.0	23.2	No	No		
3	2008-1	2-04	Albury	9.2	28.0	0.0	NaN	NaN	NE	24.0	SE	
	45.0	16.0	1017.6	1012.8	NaN	NaN	18.1	26.5	No	No		
4	2008-1	2-05	Albury	17.5	32.3	1.0	NaN	NaN	W	41.0	ENE	
	82.0	33.0	1010.8	1006.0	7.0	8.0	17.8	29.7	No	No		

5 rows × 23 columns

3. Handling missing values

df.isnull().sum()*100/len(df)

Date 0.000000

Location 0.000000

MinTemp 1.020899

MaxTemp 0.866905

Rainfall 2.241853

Evaporation 43.166506

Sunshine 48.009762

WindGustDir 7.098859

WindGustSpeed 7.055548

WindDir9am 7.263853

WindDir3pm 2.906641

WindSpeed9am 1.214767

WindSpeed3pm 2.105046

Humidity9am 1.824557

Humidity3pm 3.098446

Pressure9am 10.356799

Pressure3pm 10.331363

Cloud9am 38.421559

Cloud3pm 40.807095

Temp9am 1.214767

Temp3pm 2.481094

RainToday 2.241853

RainTomorrow 2.245978

dtype: float64

df.isnull().sum()

Date 0

Location 0

MinTemp 1485

MaxTemp 1261

Rainfall 3261

Evaporation 62790

Sunshine 69835

WindGustDir 10326

WindGustSpeed 10263

WindDir9am 10566

WindDir3pm 4228

WindSpeed9am 1767

WindSpeed3pm 3062

Humidity9am 2654

Humidity3pm 4507

Pressure9am 15065

Pressure3pm 15028

Cloud9am 55888

Cloud3pm 59358

Temp9am 1767

Temp3pm 3609

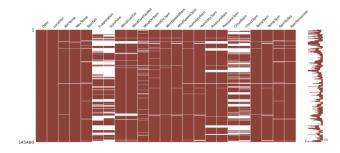
RainToday 3261

RainTomorrow 3267

dtype: int64

import missingno as msno

msno.matrix(df, color= (0.55, 0.255, 0.225), fontsize=16)



df cat = df[['RainToday', 'WindGustDir', 'WindDir9am', 'WindDir3pm']]

df.drop(columns=['Evaporation', 'Sunshine', 'Cloud9am', 'Cloud3pm'], axis=1, inplace=True)

df.drop(columns=['RainToday', 'WindGustDir', 'WindDir9am', 'WindDir3pm'], axis=1, inplace=True)

df['MinTemp'].fillna (df['MinTemp'].mean(), inplace=True)

df['MaxTemp'].fillna(df['MaxTemp'].mean(), inplace=True)

df['Rainfall'].fillna (df['Rainfall'].mean(), inplace=True)

df['WindGustSpeed'].fillna (df['WindGustSpeed'].mean(), inplace=True)

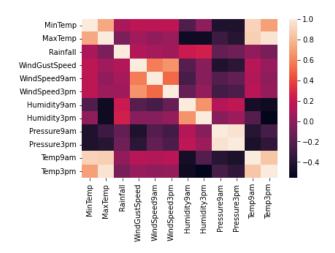
df['WindSpeed9am'].fillna (df['WindSpeed9am'].mean(), inplace=True)

df['WindSpeed3pm'].fillna (df['WindSpeed3pm'].mean(), inplace=True)

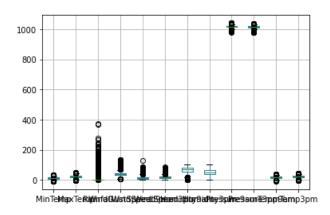
```
df['Humidity9am'].fillna (df['Humidity9am'].mean(), inplace=True)
df['Humidity3pm'].fillna (df['Humidity3pm'].mean(), inplace=True)
df['Pressure9am'].fillna (df['Pressure9am'].mean(), inplace=True)
df['Pressure3pm'].fillna (df[ 'Pressure3pm'].mean(), inplace=True)
df['Temp9am'].fillna (df['Temp9am']. mean(), inplace=True)
df['Temp3pm'].fillna(df['Temp3pm'].mean(), inplace=True)
#Loading the names of categorical columns
cat_names=df_cat.columns
# intializing the simple imputer for missing categorical values
import numpy as np
from sklearn.impute import SimpleImputer
imp_mode= SimpleImputer(missing_values=np.nan, strategy='most_frequent')
# fitting and transforming the missing data
df_cat=imp_mode.fit_transform(df_cat)
# converting array to dataframe
df_cat= pd.DataFrame(df_cat, columns=cat_names)
# concatinating the categorical and numeric
df = pd.concat([df, df_cat], axis=1)
4. Data visulization
df.corr()
                              Rainfall WindGustSpeed
                                                            WindSpeed9am WindSpeed3pm
MinTemp
               MaxTemp
       Humidity9am Humidity3pm Pressure9am Pressure3pm Temp9am
                                                                                   Temp3pm
MinTemp
               1.000000
                              0.733400
                                             0.102706
                                                            0.172553
                                                                           0.173404
                      -0.230970
                                     0.005995
                                                    -0.423584
       0.173058
                                                                    -0.433147
                                                                                   0.897692
       0.699211
```

MaxTemp	0.733400	1.000000	-0.074040	0.065895	0.014	294
0.0497 0.9687		927 -0.498	3760 -0.30	8309 -0	.396622	0.879170
Rainfall 0.1027	706 -0.074	040 1.000	000 0.126	6446 0.	085925	0.056527
0.2213	380 0.2489	905 -0.159	9055 -0.11	9541 0.	011069	-0.077684
WindGustSpee					000000	0.577319
0.6572 0.0318		964 -0.02!	5355 -0.42	5760 -0	.383938	0.145904
WindSpeed9ar	m0.173404	0.014294	0.085925	0.577319	1.000	000
0.5124 0.0044		271 -0.030	0887 -0.21	5339 -0	.165388	0.127592
WindSpeed3pr	m0.173058	0.049717	0.056527	0.657243	0.512	427
1.0000 0.0275		458 0.016	275 -0.27	7604 -0	.239659	0.161060
Humidity9am	-0.230970	-0.497927	0.221380	-0.207964	-0.268	3271 -
0.143458 0.490709	1.000000	0.659072	0.131503	0.176009	-0.469	9641 -
Humidity3pm	0.005995	-0.498760	0.248905	-0.025355	-0.030	887
0.0162 -0.555		1.000	000 -0.02	5848 0.	048695	-0.216964
Pressure9am	-0.423584	-0.308309	-0.159055	-0.425760	-0.215	- 339
0.277604 0.265532	0.131503	-0.025848	1.000000	0.959662	-0.397	'131 -
Pressure3pm	-0.433147	-0.396622	-0.119541	-0.383938	-0.165	388 -
0.239659 0.360707	0.176009	0.048695	0.959662	1.000000	-0.441	.459 -
Temp9am	0.897692	0.879170	0.011069	0.145904	0.127	592
0.1610 0.8461	060 -0.469 141	641 -0.216	5964 -0.39	7131 -0	.441459	1.000000
Temp3pm	0.699211	0.968713	-0.077684	0.031884	0.004	476
	-0.490					

cor = df.corr()sns.heatmap(data=cor,xtickla



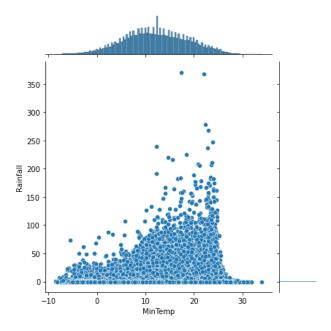
df.boxplot()



sns.jointplot(df["MinTemp"],df['Rainfall'])

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

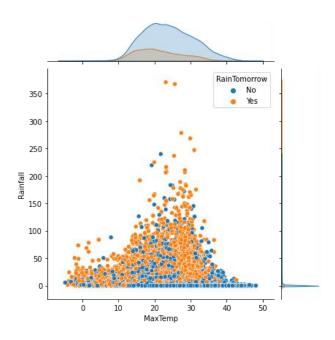
FutureWarning

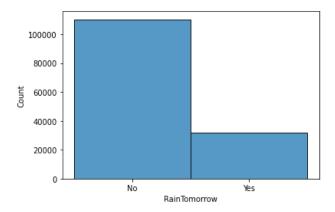


sns.jointplot(df["MaxTemp"],df['Rainfall'],hue=df['RainTomorrow'])

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

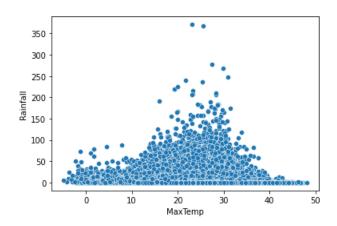




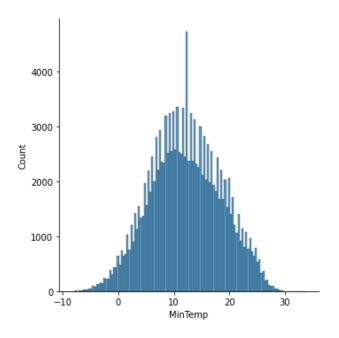
sns.scatterplot(df['MaxTemp'],df['Rainfall'])

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning



sns.displot(df['MinTemp'])



5. Spliting the dataset into Dependent and independent variable

```
y=df['RainTomorrow']
x=df.drop('RainTomorrow',axis=1)
6.Feature Scaling
```

from sklearn.preprocessing import StandardScaler
y=df['RainTomorrow']
x=df.drop('RainTomorrow',axis=1)
names=x.columns
names
Index(['Date', 'Location', 'MinTemp', 'MaxTemp', 'Rainfall', 'WindGustSpeed',
 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm',
 'Pressure9am', 'Pressure3pm', 'Temp9am', 'Temp3pm', 'RainToday',
 'WindGustDir', 'WindDir9am', 'WindDir3pm'],
 dtype='object')

sc=StandardScaler()

7.splitting the data into train and test

sklearn import model_selection

x_train,x_test,y_train,y_test