## Import Dependencies

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.utils import resample
from sklearn.model_selection import train_test_split,GridSearchCV,cross_val_score
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report,confusion_matrix,accuracy_score
import pickle
```

# **Data Collection And Processing**

```
#load data set
data=pd.read_csv('/content/Rainfall.csv')
print(type(data))
```

<class 'pandas.core.frame.DataFrame'>

data.shape

**→** (366, 12)

data.head()

<b>→</b>		dayı	nnoccuno	maytomn	temparature	mintomp	doumaint	humidity	cloud	nainfall
		uay	pressure	шахсешр	temparature	штисешр	иемротис	Humituity	CIOUU	Latiliatt
	0	1	1025.9	19.9	18.3	16.8	13.1	72	49	yes
	1	2	1022.0	21.7	18.9	17.2	15.6	81	83	yes
	2	3	1019.7	20.3	19.3	18.0	18.4	95	91	yes
	3	4	1018.9	22.3	20.6	19.1	18.8	90	88	yes
	4	5	1015.9	21.3	20.7	20.2	19.9	95	81	yes

Next steps:

Generate code with data

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data.tail()

day pressure maxtemp temparature mintemp dewpoint humidity cloud rainfal.

361	27	1022.7	18.8	17.7	16.9	15.0	84	90	ye
362	28	1026.6	18.6	17.3	16.3	12.8	75	85	ye
363	29	1025.9	18.9	17.7	16.4	13.3	75	78	ye
364	30	1025.3	19.2	17.3	15.2	13.3	78	86	ye
365	31	1026.4	20.5	17.8	15.5	13.0	74	66	nc

data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 366 entries, 0 to 365
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	day	366 non-null	int64
1	pressure	366 non-null	float64
2	maxtemp	366 non-null	float64
3	temparature	366 non-null	float64
4	mintemp	366 non-null	float64
5	dewpoint	366 non-null	float64
6	humidity	366 non-null	int64
7	cloud	366 non-null	int64
8	rainfall	366 non-null	object
9	sunshine	366 non-null	float64
10	winddirection	365 non-null	float64
11	windspeed	365 non-null	float64
	67		

dtypes: float64(8), int64(3), object(1)

memory usage: 34.4+ KB

data.columns=data.columns.str.strip()

data.info()

<u>\_</u>

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 366 entries, 0 to 365
Data columns (total 12 columns):

		/ -	
#	Column	Non-Null Count	Dtype
0	day	366 non-null	int64
1	pressure	366 non-null	float64
2	maxtemp	366 non-null	float64
3	temparature	366 non-null	float64
4	mintemp	366 non-null	float64
5	dewpoint	366 non-null	float64
6	humidity	366 non-null	int64
7	cloud	366 non-null	int64
8	rainfall	366 non-null	object
9	sunshine	366 non-null	float64

```
10 winddirection 365 non-null
                                         float64
      11 windspeed
                         365 non-null
                                         float64
     dtypes: float64(8), int64(3), object(1)
    memory usage: 34.4+ KB
data=data.drop(columns=["day"])
data.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 366 entries, 0 to 365
    Data columns (total 11 columns):
          Column
                         Non-Null Count
                                         Dtype
                         -----
          -----
     ---
      0
                         366 non-null
                                         float64
          pressure
                                         float64
      1
         maxtemp
                         366 non-null
      2
         temparature
                         366 non-null
                                         float64
      3
          mintemp
                         366 non-null
                                         float64
      4
         dewpoint
                         366 non-null
                                         float64
      5
          humidity
                         366 non-null
                                         int64
      6
          cloud
                         366 non-null
                                         int64
      7
          rainfall
                                         object
                         366 non-null
      8
          sunshine
                         366 non-null
                                         float64
      9
         winddirection 365 non-null
                                         float64
                         365 non-null
      10 windspeed
                                         float64
     dtypes: float64(8), int64(2), object(1)
     memory usage: 31.6+ KB
```

### data.isnull().sum()

	0
pressure	0
maxtemp	0
temparature	0
mintemp	0
dewpoint	0
humidity	0
cloud	0
rainfall	0
sunshine	0
winddirection	1
windspeed	1

## dtype: int64

	windspeed
0	26.3
1	15.3
2	14.2
3	16.9
4	13.7
361	18.4
362	25.9
363	33.4
364	20.9
365	23.3

366 rows × 1 columns

dtype: float64

data.isna().sum()

	0
pressure	0
maxtemp	0
temparature	0
mintemp	0
dewpoint	0
laa ! al !4	^

```
numicity 0
cloud 0
rainfall 0
sunshine 0
winddirection 1
windspeed 1
```

dtype: int64

data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 366 entries, 0 to 365
Data columns (total 11 columns):
```

#	Column	Non-Null Count	Dtype
0	pressure	366 non-null	float64
1	maxtemp	366 non-null	float64
2	temparature	366 non-null	float64
3	mintemp	366 non-null	float64
4	dewpoint	366 non-null	float64
5	humidity	366 non-null	int64
6	cloud	366 non-null	int64
7	rainfall	366 non-null	object
8	sunshine	366 non-null	float64
9	winddirection	365 non-null	float64
10	windspeed	365 non-null	float64
dtyp	es: float64(8),	int64(2), objec	t(1)
memo	ry usage: 31.6+	KB	

#converting yes & no to 1 and o respectively
data['rainfall']=data['rainfall'].map({'yes':1,'no':0})

data.head()

	pressure	maxtemp	temparature	mintemp	dewpoint	humidity	cloud	rainfall	sunsi
0	1025.9	19.9	18.3	16.8	13.1	72	49	1	
1	1022.0	21.7	18.9	17.2	15.6	81	83	1	
2	1019.7	20.3	19.3	18.0	18.4	95	91	1	
3	101 <u>2</u> 0	<b>၁</b> ၁	20 6	10 1	12 2	۵۸	ጸጸ	1	

J	1010.5	22.0	20.0	13.1	10.0	30	00	1
4	1015.9	21.3	20.7	20.2	19.9	95	81	1

Next steps: Generate code with data

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# Exploraty Data Analysis(EDA)

```
data.shape
```

(366, 11)

#setting plot style for all the plot
sns.set\_style("whitegrid")

data.describe()

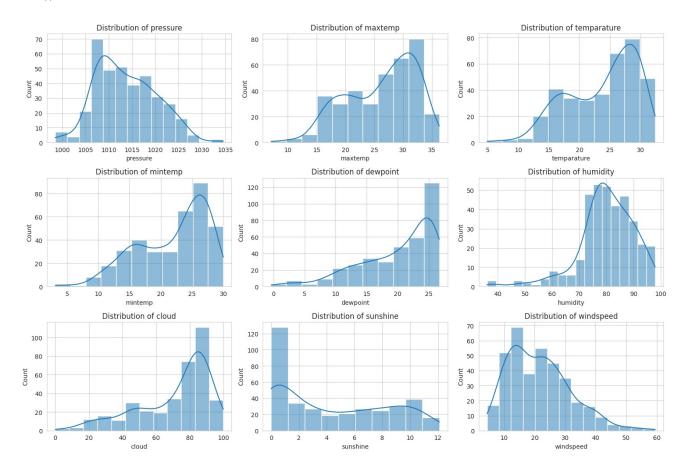
	pressure	maxtemp	temparature	mintemp	dewpoint	humidity	<b>c</b> ]
count	366.000000	366.000000	366.000000	366.000000	366.000000	366.000000	366.000
mean	1013.742623	26.191257	23.747268	21.894536	19.989071	80.177596	71.128
std	6.414776	5.978343	5.632813	5.594153	5.997021	10.062470	21.798
min	998.500000	7.100000	4.900000	3.100000	-0.400000	36.000000	0.000
25%	1008.500000	21.200000	18.825000	17.125000	16.125000	75.000000	58.000
50%	1013.000000	27.750000	25.450000	23.700000	21.950000	80.500000	80.000
75%	1018.100000	31.200000	28.600000	26.575000	25.000000	87.000000	88.000
max	1034.600000	36.300000	32.400000	30.000000	26.700000	98.000000	100.000

data.columns

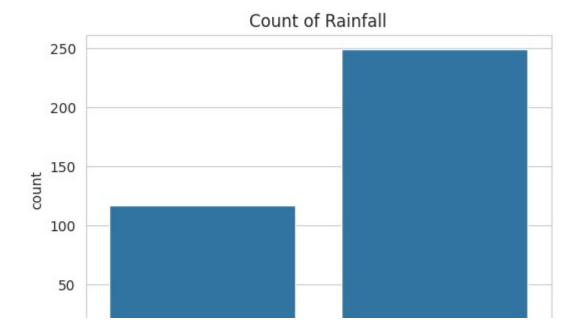
```
plt.figure(figsize=(15,10))
```

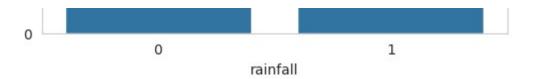
## plt.title(f"Distribution of {column}")

plt.tight\_layout()
plt.show()

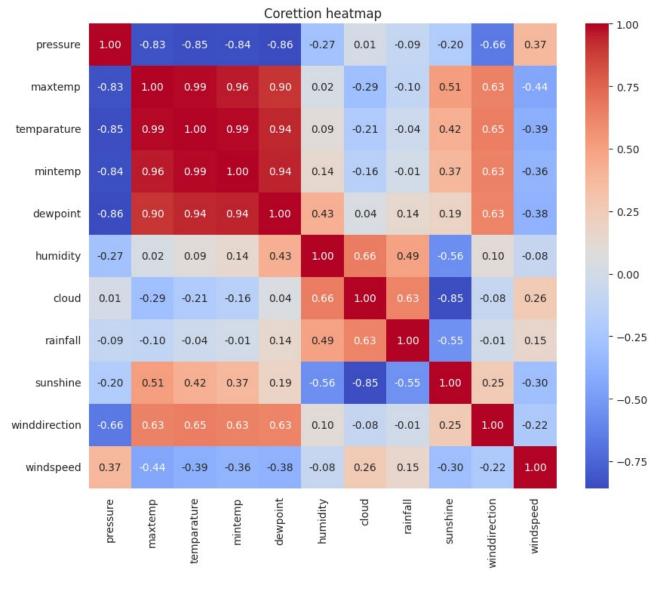


plt.figure(figsize=(6,4))
sns.countplot(x="rainfall",data=data)
plt.title("Count of Rainfall")
plt.show()

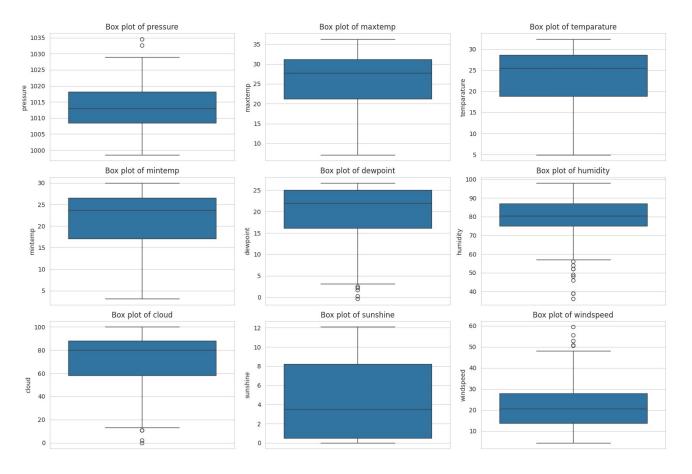




```
#corelation matrix
plt.figure(figsize=(10,8))
sns.heatmap(data.corr(),annot=True,cmap="coolwarm",fmt=".2f")
plt.title("Corettion heatmap")
plt.show()
```



plt.tight\_layout()
plt.show()



#data Preproessing
data=data.drop(columns=['maxtemp','temparature','mintemp'])

data.head()

	pressure	dewpoint	humidity	cloud	rainfall	sunshine	winddirection	windspeed
0	1025.9	13.1	72	49	1	9.3	80.0	26.3
1	1022.0	15.6	81	83	1	0.6	50.0	15.3
2	1019.7	18.4	95	91	1	0.0	40.0	14.2
3	1018.9	18.8	90	88	1	1.0	50.0	16.9
4	1015.9	19.9	95	81	1	0.0	40.0	13.7

Next steps: Generate code with data View recommended plots New interactive sheet

data['rainfall'].value\_counts()

```
rainfall
    1
             249
   0
              117
```

count

dtype: int64

```
#sepaate majority
df_majority=data[data['rainfall']==1]
df_minority=data[data['rainfall']==0]
print(df_majority.shape)
print(df_minority.shape)
     (249, 8)
     (117, 8)
```

df\_majority\_downsample=resample(df\_majority,replace=False,n\_samples=len(df\_minority),rand df\_majority\_downsample.shape

(117, 8)

df\_downsampled=pd.concat([df\_majority\_downsample,df\_minority])

df\_downsampled.shape

(234, 8)

df\_downsampled.head()

	pressure	dewpoint	humidity	cloud	rainfall	sunshine	winddirection	windspee
188	1005.9	25.6	77	53	1	10.5	270.0	11.
9	1017.5	15.5	85	91	1	0.0	70.0	37.
137	1012.3	20.1	80	86	1	0.3	80.0	39.
89	1018.3	16.3	79	89	1	2.4	40.0	14.
157	1008.8	24.7	91	80	1	2.2	20.0	11.:

Next

Generate code with df downsampled ) ( View recommended plots ) ( New interactive sheet

steps:

df\_downsampled =df\_downsampled.sample(frac=1,random\_state=42).reset\_index(drop=True)

df\_downsampled.head()

	pressure	dewpoint	humidity	cloud	rainfall	sunshine	winddirection	windspeed
0	1022.2	14.1	78	90	1	0.0	30.0	28.5
1	1013.4	19.5	69	17	0	10.5	70.0	12.4
2	1006.1	24.4	74	27	0	10.8	220.0	8.7
3	1007.6	24.8	85	84	1	1.8	70.0	34.8
4	1021.2	8.4	66	18	0	10.1	20.0	24.4

Next steps:

Generate code with df\_downsampled

View recommended plots

New interactive sheet

df\_downsampled['rainfall'].value\_counts()

#### count

rainfall					
1	117				
0	117				

dtype: int64

#splitt features and target as x and y
x=df\_downsampled.drop(columns=['rainfall'])
y=df\_downsampled['rainfall']

print(x)

	pressure	dewpoint	humidity	cloud	sunshine	winddirection	windspeed
0	1022.2	14.1	78	90	0.0	30.0	28.5
1	1013.4	19.5	69	17	10.5	70.0	12.4
2	1006.1	24.4	74	27	10.8	220.0	8.7
3	1007.6	24.8	85	84	1.8	70.0	34.8
4	1021.2	8.4	66	18	10.1	20.0	24.4
	• • •					• • •	
229	1008.1	25.4	86	75	5.7	20.0	9.5
230	1010.1	19.9	91	89	0.0	70.0	31.8
231	1020.6	14.7	91	88	0.3	50.0	24.4
222	4000 3	24.4	74	20		10.0	A A

Model Evalution

```
24.1
                                                     5./
                                                                                4.4
     232
            T008.3
                                     /4
                                            29
                                                                   10.0
     233
                        26.1
                                                     2.2
                                                                  160.0
            1005.0
                                    87
                                            82
                                                                               12.6
     [234 rows x 7 columns]
#splitting the data into training data and test data
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
x_test.shape
     (47, 7)
Model Training
rf_model=RandomForestClassifier(random_state=42)
param_grid_rf ={
    'n_estimators':[50,100,200],
    'max_features':['sqrt','log2'],
    'max_depth':[None,10,20,30],
    'min_samples_split':[2,5,10],
    'min_samples_leaf':[1,2,4]
    }
grid_search_rf=GridSearchCV(estimator=rf_model,param_grid=param_grid_rf,cv=5,n_jobs=-1,ve
grid_search_rf.fit(x_train,y_train)
     Fitting 5 folds for each of 216 candidates, totalling 1080 fits
                     GridSearchCV
                                          (i) (?
                   best_estimator_:
               RandomForestClassifier
            RandomForestClassifier ?
best_rf_model=grid_search_rf.best_estimator_
print("best_rf_model:" ,grid_search_rf.best_params_)
     best_rf_model: {'max_depth': 10, 'max_features': 'sqrt', 'min_samples_leaf': 1, 'min_
```

```
cv_scores=cross_val_score(best_rf_model,x_train,y_train,cv=5)
print("CV scores:",cv_scores)
print("Mean CV score:",np.mean(cv_scores))
    CV scores: [0.71052632 0.78947368 0.86486486 0.83783784 0.89189189]
    Mean CV score: 0.818918918919
#test set performance
y_pred=best_rf_model.predict(x_test)
print("Classification Report:\n",classification_report(y_test,y_pred))
print("Confusion Matrix:\n",confusion_matrix(y_test,y_pred))
print("Accuracy Score:",accuracy_score(y_test,y_pred))
    Classification Report:
                   precision recall f1-score support
               0
                       0.77
                                 0.71
                                           0.74
                                                       24
               1
                       0.72
                                 0.78
                                           0.75
                                                       23
                                           0.74
                                                       47
        accuracy
                       0.75
                                 0.75
                                           0.74
                                                       47
       macro avg
                       0.75
    weighted avg
                                 0.74
                                           0.74
                                                       47
    Confusion Matrix:
     [[17 7]
     [ 5 18]]
    Accuracy Score: 0.7446808510638298
Prediction on unknown data
```

#### Load the saves model and file and use it for predition

import pickle

```
#load the trained model and fetures names from the pickel file
with open("model.pkl","rb") as file:
    model_data=pickle.load(file)

model=model_data["model"]
feture_names=model_data['features_names']

input_data=(1015.9,19.9,95,81,0.0,40.0,13.7)
input_df =pd.DataFrame([input_data],columns=feture_names)

prediction=model.predict(input_df)
print("Prediction result:",'Rainfall' if prediction[0]==1 else 'No Rainfall')
    Prediction result: Rainfall
Start coding or generate with AI.
```