

weather-predicting-system

July 26, 2023

1 WEATHER PREDICTING SYSTEM

2 IMPORTING DATASET

```
[146]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import preprocessing
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
```

```
[147]: dataset = pd.read_csv("seattle-weather.csv")
dataset.head(10)
```

```
[147]:
```

	date	precipitation	temp_max	temp_min	wind	weather
0	2012-01-01	0.0	12.8	5.0	4.7	drizzle
1	2012-01-02	10.9	10.6	2.8	4.5	rain
2	2012-01-03	0.8	11.7	7.2	2.3	rain
3	2012-01-04	20.3	12.2	5.6	4.7	rain
4	2012-01-05	1.3	8.9	2.8	6.1	rain
5	2012-01-06	2.5	4.4	2.2	2.2	rain
6	2012-01-07	0.0	7.2	2.8	2.3	rain
7	2012-01-08	0.0	10.0	2.8	2.0	sun
8	2012-01-09	4.3	9.4	5.0	3.4	rain
9	2012-01-10	1.0	6.1	0.6	3.4	rain

3 DATA PREPROCESSING

Dimensional check

```
[148]: dataset.shape
```

```
[148]: (1461, 6)
```

```
[149]: dataset.columns
```

```
[149]: Index(['date', 'precipitation', 'temp_max', 'temp_min', 'wind', 'weather'],
dtype='object')
```

```
[150]: dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1461 entries, 0 to 1460
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   date            1461 non-null   object
1   precipitation    1461 non-null   float64
2   temp_max        1461 non-null   float64
3   temp_min        1461 non-null   float64
4   wind            1461 non-null   float64
5   weather         1461 non-null   object
dtypes: float64(4), object(2)
memory usage: 68.6+ KB
```

4 Take care of missing data.

```
[151]: dataset = dataset.dropna()
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1461 entries, 0 to 1460
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   date            1461 non-null   object
1   precipitation    1461 non-null   float64
2   temp_max        1461 non-null   float64
3   temp_min        1461 non-null   float64
4   wind            1461 non-null   float64
5   weather         1461 non-null   object
dtypes: float64(4), object(2)
memory usage: 79.9+ KB
```

Here we have two temperature columns for the prediction we had took avg of the both columns.

```
[152]: avg_temp=dataset[["temp_max","temp_min"]].mean(axis=1)
dataset['avg_temp']=avg_temp
dataset.head()
```

```
[152]:
```

	date	precipitation	temp_max	temp_min	wind	weather	avg_temp
0	2012-01-01	0.0	12.8	5.0	4.7	drizzle	8.90
1	2012-01-02	10.9	10.6	2.8	4.5	rain	6.70

2	2012-01-03	0.8	11.7	7.2	2.3	rain	9.45
3	2012-01-04	20.3	12.2	5.6	4.7	rain	8.90
4	2012-01-05	1.3	8.9	2.8	6.1	rain	5.85

```
[153]: cols = ['temp_max', 'temp_min']
dataset = dataset.drop(cols, axis=1)
dataset.head()
```

```
[153]:
```

	date	precipitation	wind	weather	avg_temp
0	2012-01-01	0.0	4.7	drizzle	8.90
1	2012-01-02	10.9	4.5	rain	6.70
2	2012-01-03	0.8	2.3	rain	9.45
3	2012-01-04	20.3	4.7	rain	8.90
4	2012-01-05	1.3	6.1	rain	5.85

```
[154]: dataset.columns
```

```
[154]: Index(['date', 'precipitation', 'wind', 'weather', 'avg_temp'], dtype='object')
```

```
[155]: dataset=dataset[["date" ,"precipitation", "wind","avg_temp","weather"]]
```

```
[156]: dataset.head()
```

```
[156]:
```

	date	precipitation	wind	avg_temp	weather
0	2012-01-01	0.0	4.7	8.90	drizzle
1	2012-01-02	10.9	4.5	6.70	rain
2	2012-01-03	0.8	2.3	9.45	rain
3	2012-01-04	20.3	4.7	8.90	rain
4	2012-01-05	1.3	6.1	5.85	rain

5 IDENTIFYING THE CORRELATION USING HEAT MAP

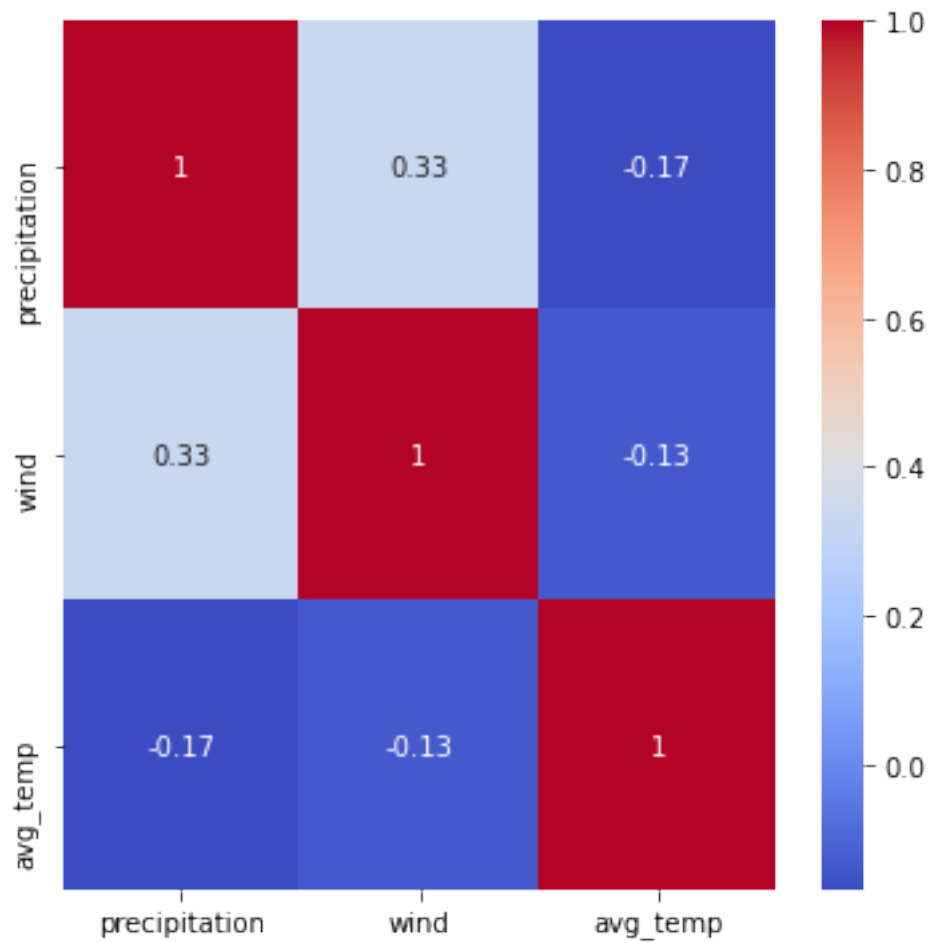
```
[157]: dataset.corr()
```

```
[157]:
```

	precipitation	wind	avg_temp
precipitation	1.000000	0.328045	-0.170465
wind	0.328045	1.000000	-0.132067
avg_temp	-0.170465	-0.132067	1.000000

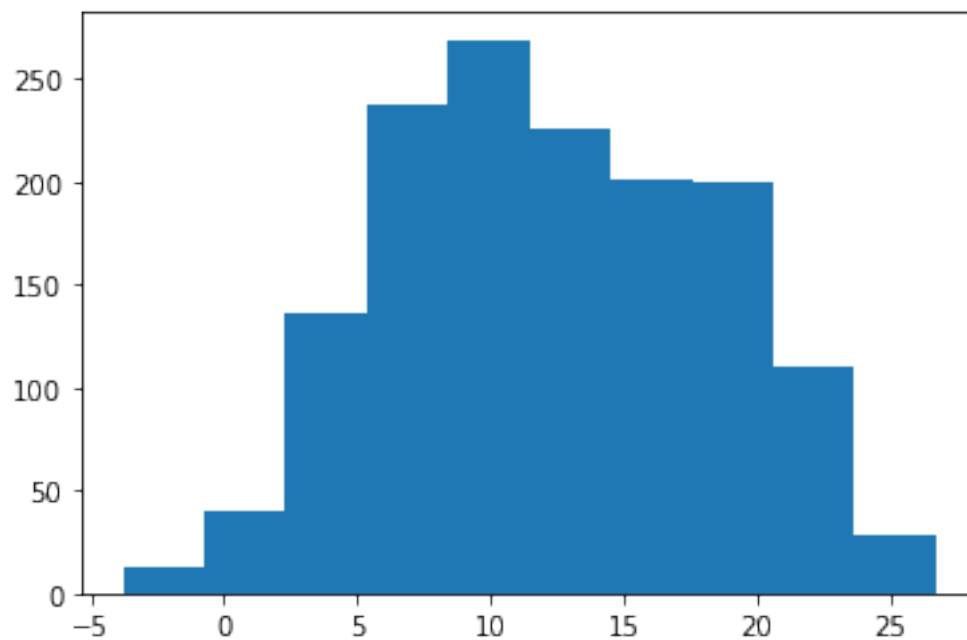
```
[158]: plt.figure(figsize=(6,6))
sns.heatmap(dataset.corr(),annot=True,cmap="coolwarm")
```

```
[158]: <matplotlib.axes._subplots.AxesSubplot at 0x1bf4bca55e0>
```



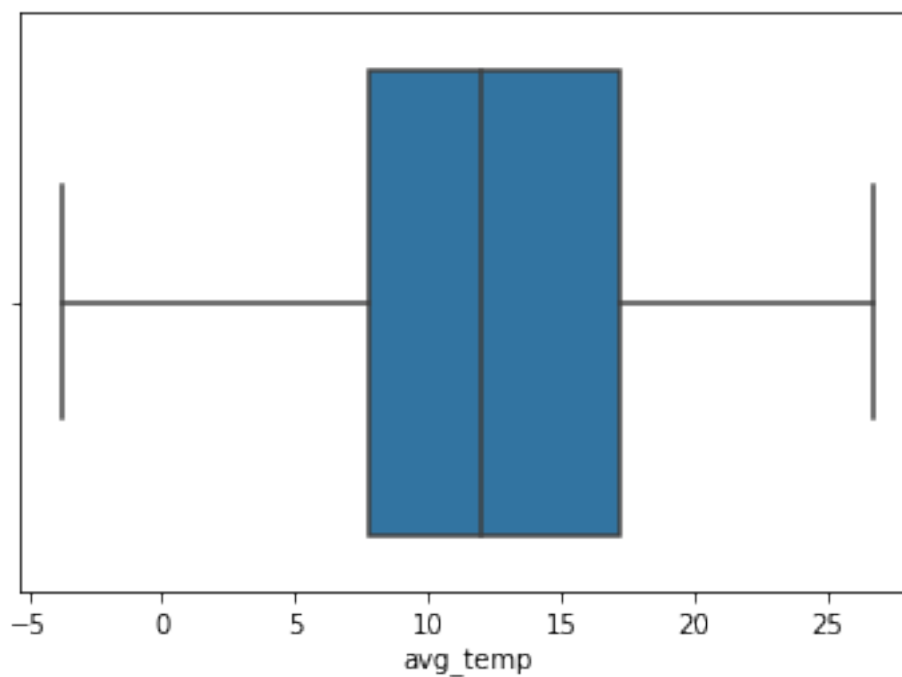
```
[159]: plt.hist(dataset["avg_temp"]) #outliers
```

```
[159]: (array([ 13.,  40., 136., 238., 269., 226., 201., 200., 110.,  28.]),
       array([-3.8 , -0.75,  2.3 ,  5.35,  8.4 , 11.45, 14.5 , 17.55, 20.6 ,
              23.65, 26.7 ]),
       <a list of 10 Patch objects>)
```



```
[160]: sns.boxplot(x=dataset["avg_temp"])
```

```
[160]: <matplotlib.axes._subplots.AxesSubplot at 0x1bf4ba16a90>
```



6 SPLITTING THE DATA

```
[161]: X=dataset.iloc[:,1:4]
X.head()
```

```
[161]:    precipitation  wind  avg_temp
0           0.0    4.7     8.90
1          10.9    4.5     6.70
2           0.8    2.3     9.45
3          20.3    4.7     8.90
4           1.3    6.1     5.85
```

```
[162]: Y=dataset.iloc[:,-1:]
Y.head()
```

```
[162]:    weather
0  drizzle
1    rain
2    rain
3    rain
4    rain
```

7 KNN ALGORITHM

```
[163]: X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size = 0.20,
↳random_state = 0)
```

```
[164]: #we are doing feature scaling to the training and test set of independent
↳variables for reducing the size to smaller values
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

```
[165]: from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors = 5, metric = 'minkowski', p = 2)
↳# n_neighbors is setting as 5, which means 5 neighborhood points are
↳required for classifying a given point.
classifier.fit(X_train, y_train)
#The Minkowski distance or Minkowski metric is a metric in a normed vector
↳space which can be considered as a generalization of both the Euclidean
↳distance and the Manhattan distance.
```

<ipython-input-165-abdf7f822748>:3: 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().

```
classifier.fit(X_train, y_train)
```

```
[165]: KNeighborsClassifier()
```

```
[166]: y_pred = classifier.predict(X_test)
       y_pred
```

```
[166]: array(['sun', 'fog', 'rain', 'sun', 'drizzle', 'rain', 'rain', 'rain',
            'sun', 'sun', 'sun', 'rain', 'sun', 'sun', 'sun', 'sun', 'sun',
            'rain', 'rain', 'sun', 'rain', 'rain', 'rain', 'sun', 'sun', 'sun',
            'sun', 'rain', 'sun', 'sun', 'rain', 'rain', 'rain', 'rain', 'sun',
            'rain', 'sun', 'rain', 'sun', 'rain', 'sun', 'rain', 'rain',
            'rain', 'sun', 'sun', 'drizzle', 'sun', 'sun', 'sun', 'sun', 'sun',
            'rain', 'rain', 'rain', 'sun', 'sun', 'rain', 'sun', 'rain', 'sun',
            'sun', 'sun', 'sun', 'sun', 'rain', 'sun', 'sun', 'sun', 'rain',
            'rain', 'sun', 'rain', 'fog', 'fog', 'sun', 'sun', 'sun', 'rain',
            'rain', 'rain', 'sun', 'rain', 'rain', 'rain', 'sun', 'fog',
            'rain', 'sun', 'rain', 'sun', 'sun', 'rain', 'sun', 'rain', 'rain',
            'sun', 'sun', 'sun', 'rain', 'rain', 'sun', 'sun', 'rain', 'rain',
            'sun', 'rain', 'rain', 'fog', 'sun', 'rain', 'sun', 'rain', 'rain',
            'sun', 'rain', 'sun', 'sun', 'sun', 'rain', 'rain', 'rain', 'rain',
            'rain', 'rain', 'rain', 'sun', 'drizzle', 'sun', 'sun', 'rain',
            'rain', 'sun', 'sun', 'rain', 'fog', 'sun', 'sun', 'sun', 'rain',
            'rain', 'rain', 'sun', 'sun', 'sun', 'sun', 'rain', 'sun', 'rain',
            'sun', 'rain', 'sun', 'sun', 'fog', 'fog', 'sun', 'sun', 'sun',
            'sun', 'sun', 'sun', 'fog', 'sun', 'sun', 'rain', 'rain', 'rain',
            'rain', 'rain', 'rain', 'sun', 'rain', 'sun', 'sun', 'rain',
            'rain', 'sun', 'fog', 'sun', 'sun', 'sun', 'sun', 'sun', 'sun',
            'rain', 'sun', 'sun', 'rain', 'fog', 'sun', 'sun', 'sun', 'sun',
            'drizzle', 'rain', 'rain', 'sun', 'sun', 'sun', 'rain', 'rain',
            'sun', 'sun', 'sun', 'rain', 'sun', 'rain', 'sun', 'sun', 'sun',
            'rain', 'rain', 'rain', 'rain', 'sun', 'sun', 'sun', 'rain',
            'rain', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun', 'sun', 'sun',
            'rain', 'rain', 'sun', 'drizzle', 'rain', 'rain', 'rain', 'rain',
            'fog', 'sun', 'sun', 'rain', 'rain', 'sun', 'sun', 'sun', 'sun',
            'rain', 'sun', 'sun', 'sun', 'sun', 'sun', 'sun', 'sun', 'rain',
            'rain', 'sun', 'fog', 'sun', 'sun', 'sun', 'rain', 'sun', 'fog',
            'fog', 'sun', 'sun', 'rain', 'sun', 'rain', 'sun', 'sun', 'rain',
            'sun', 'rain', 'rain', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun',
            'rain', 'rain', 'rain', 'sun', 'sun', 'sun', 'sun', 'sun', 'rain',
            'sun', 'rain', 'rain', 'sun'], dtype=object)
```

```
[167]: y_test
```

```
[167]:      weather
530      sun
657      sun
459     rain
279      sun
```

```

656      sun
..      ...
440     rain
634      sun
494      sun
61      rain
517      sun

```

```
[293 rows x 1 columns]
```

```
[168]: from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
ac1 = accuracy_score(y_test, y_pred)
```

```
[169]: ac1
```

```
[169]: 0.689419795221843
```

```
[170]: cm
```

```
[170]: array([[ 2,  0,  2,  0,  6],
             [ 1,  3,  3,  0, 22],
             [ 0,  3, 95,  0, 25],
             [ 0,  0,  5,  0,  1],
             [ 2,  9, 12,  0, 102]], dtype=int64)
```

8 NAVIE BAYES

```
[171]: from sklearn.naive_bayes import GaussianNB
classifier = GaussianNB()
classifier.fit(X_train, y_train)
```

```

C:\Users\Dell\anaconda3\lib\site-packages\sklearn\utils\validation.py:73:
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().
    return f(**kwargs)

```

```
[171]: GaussianNB()
```

```
[172]: y_pred = classifier.predict(X_test)
y_pred
```

```
[172]: array(['sun', 'sun', 'rain', 'sun', 'sun', 'rain', 'rain', 'rain', 'sun',
            'rain', 'sun', 'rain', 'sun', 'sun', 'sun', 'sun', 'sun', 'rain',
            'rain', 'rain', 'rain', 'rain', 'rain', 'sun', 'sun', 'sun', 'sun',
```



```

'rain', 'sun', 'sun', 'rain', 'sun', 'rain', 'rain', 'sun', 'rain',
'sun', 'rain', 'sun', 'rain', 'sun', 'rain', 'rain', 'rain', 'sun',
'sun', 'sun', 'sun', 'sun', 'sun', 'sun', 'sun', 'sun', 'rain',
'rain', 'sun', 'sun', 'rain', 'sun', 'rain', 'sun', 'sun', 'sun',
'sun', 'sun', 'rain', 'sun', 'sun', 'sun', 'rain', 'rain', 'rain',
'rain', 'sun', 'sun', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun',
'sun', 'rain', 'rain', 'rain', 'sun', 'sun', 'rain', 'sun', 'rain',
'sun', 'sun', 'rain', 'sun', 'rain', 'rain', 'sun', 'sun', 'sun',
'sun', 'rain', 'sun', 'sun', 'rain', 'rain', 'sun', 'rain', 'rain',
'sun', 'sun', 'rain', 'sun', 'rain', 'sun', 'sun', 'sun', 'sun',
'sun', 'rain', 'rain', 'sun', 'rain', 'rain', 'sun', 'rain', 'sun',
'sun', 'sun', 'sun', 'rain', 'sun', 'rain', 'sun', 'rain', 'rain',
'rain', 'sun', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun', 'sun',
'sun', 'rain', 'rain', 'sun', 'sun', 'sun', 'sun', 'sun', 'sun',
'rain', 'sun', 'rain', 'sun', 'sun', 'rain', 'sun', 'sun', 'sun',
'rain', 'sun', 'rain', 'rain', 'sun', 'rain', 'rain', 'rain',
'sun', 'sun', 'sun', 'rain', 'sun', 'sun', 'sun', 'rain', 'rain',
'sun', 'sun', 'sun', 'sun', 'rain', 'sun', 'sun', 'rain', 'rain',
'rain', 'rain', 'sun', 'rain', 'sun', 'rain', 'rain', 'sun',
'rain', 'sun', 'rain', 'rain', 'sun', 'rain', 'sun', 'sun', 'sun',
'sun', 'sun', 'rain', 'rain', 'sun', 'sun', 'sun', 'sun', 'sun',
'sun', 'rain', 'sun', 'sun', 'sun', 'sun', 'rain', 'sun', 'rain',
'sun', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun', 'sun', 'rain',
'sun', 'rain', 'sun', 'rain', 'rain', 'rain', 'sun', 'sun', 'sun',
'rain', 'sun', 'rain', 'sun', 'sun', 'rain', 'sun'], dtype='<U7')

```

```
[173]: y_test
```

```

[173]:      weather
530      sun
657      sun
459     rain
279      sun
656      sun
..      ...
440     rain
634      sun
494      sun
61      rain
517      sun

```

```
[293 rows x 1 columns]
```

```
[174]: from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
ac2 = accuracy_score(y_test, y_pred)
```

```
[175]: ac2
```

```
[175]: 0.8054607508532423
```

```
[176]: cm
```

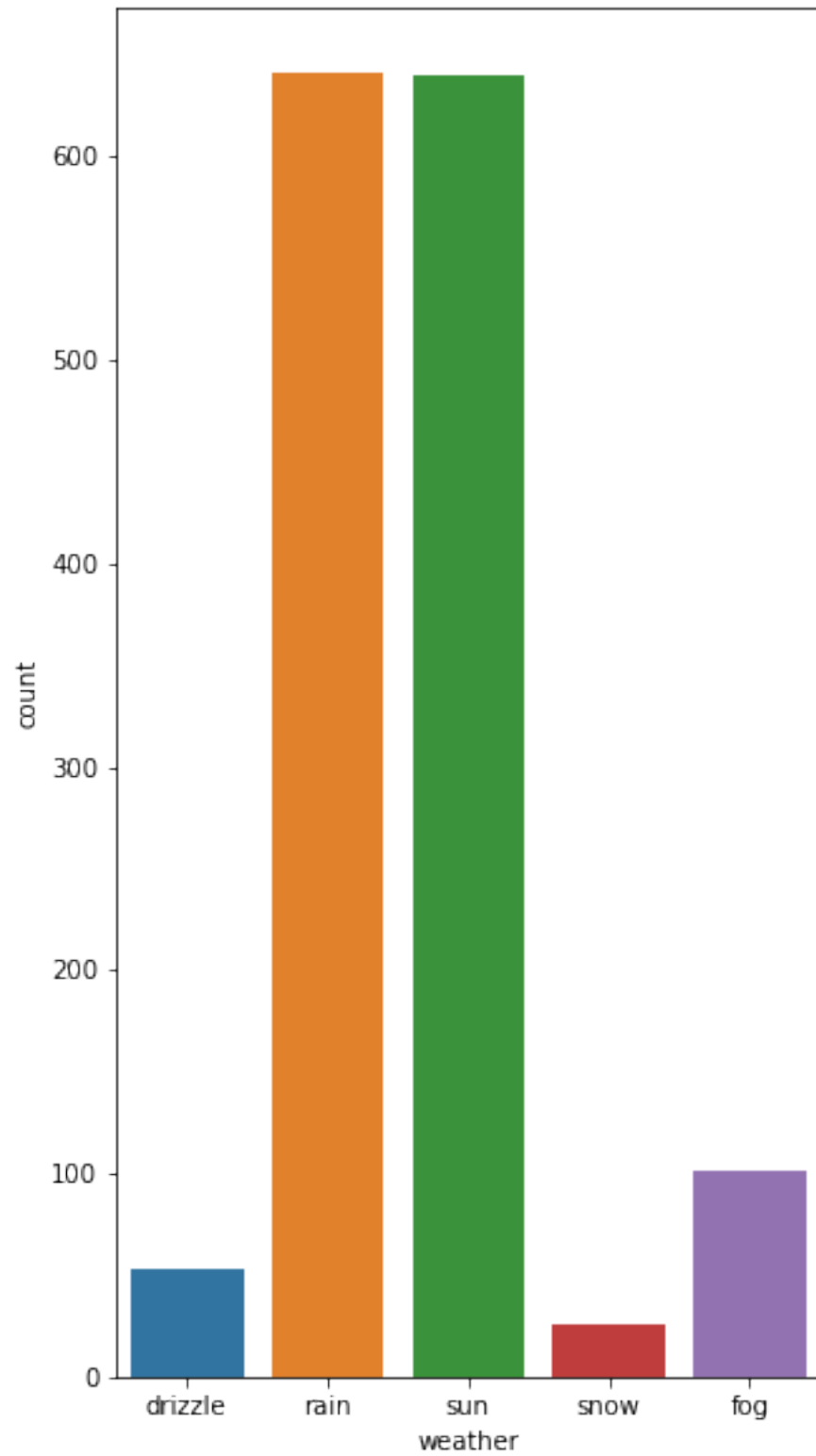
```
[176]: array([[ 0,  0,  0,  0, 10],
        [ 0,  0,  0,  0, 29],
        [ 0,  0, 111,  0, 12],
        [ 0,  0,  6,  0,  0],
        [ 0,  0,  0,  0, 125]], dtype=int64)
```

9 DECISION TREE

```
[177]: dataset['weather'].value_counts()
```

```
[177]: rain      641
sun       640
fog       101
drizzle    53
snow       26
Name: weather, dtype: int64
```

```
[178]: plt.figure(figsize=(5,10));
sns.countplot(dataset['weather']);
```



```
[179]: from sklearn.tree import DecisionTreeClassifier
```

```
[180]: # Create Decision Tree classifier object  
clf = DecisionTreeClassifier()
```

```
[181]: # Train Decision Tree Classifier  
clf = clf.fit(X_train,y_train)
```

```
[182]: #Predict the response for test dataset  
y_pred = clf.predict(X_test)  
y_pred
```

```
[182]: array(['sun', 'fog', 'rain', 'sun', 'sun', 'rain', 'rain', 'rain',  
        'drizzle', 'rain', 'sun', 'rain', 'sun', 'sun', 'sun', 'sun',  
        'sun', 'rain', 'rain', 'rain', 'rain', 'rain', 'rain', 'sun',  
        'sun', 'sun', 'sun', 'rain', 'sun', 'sun', 'rain', 'sun', 'rain',  
        'rain', 'drizzle', 'rain', 'sun', 'rain', 'sun', 'rain', 'sun',  
        'rain', 'rain', 'rain', 'drizzle', 'fog', 'drizzle', 'sun', 'sun',  
        'sun', 'sun', 'sun', 'rain', 'rain', 'rain', 'sun', 'sun', 'rain',  
        'sun', 'rain', 'sun', 'sun', 'sun', 'sun', 'fog', 'rain', 'sun',  
        'sun', 'sun', 'rain', 'snow', 'snow', 'rain', 'sun', 'fog', 'sun',  
        'sun', 'drizzle', 'rain', 'rain', 'sun', 'sun', 'rain', 'rain',  
        'rain', 'sun', 'fog', 'rain', 'sun', 'rain', 'sun', 'sun', 'rain',  
        'sun', 'rain', 'rain', 'sun', 'drizzle', 'sun', 'sun', 'rain',  
        'sun', 'sun', 'rain', 'rain', 'drizzle', 'rain', 'rain', 'fog',  
        'fog', 'rain', 'sun', 'rain', 'fog', 'sun', 'sun', 'sun', 'sun',  
        'rain', 'rain', 'rain', 'rain', 'rain', 'sun', 'rain', 'rain',  
        'sun', 'sun', 'sun', 'rain', 'sun', 'rain', 'sun', 'rain', 'rain',  
        'rain', 'drizzle', 'fog', 'sun', 'fog', 'rain', 'rain', 'sun',  
        'sun', 'sun', 'rain', 'rain', 'drizzle', 'fog', 'sun', 'sun',  
        'sun', 'sun', 'rain', 'drizzle', 'rain', 'sun', 'sun', 'rain',  
        'sun', 'sun', 'sun', 'rain', 'sun', 'rain', 'rain', 'fog', 'rain',  
        'rain', 'rain', 'sun', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun',  
        'fog', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun', 'rain', 'sun',  
        'sun', 'sun', 'rain', 'sun', 'sun', 'sun', 'sun', 'drizzle',  
        'rain', 'rain', 'sun', 'drizzle', 'rain', 'rain', 'rain', 'sun',  
        'sun', 'drizzle', 'rain', 'rain', 'rain', 'sun', 'sun', 'sun',  
        'rain', 'rain', 'rain', 'rain', 'fog', 'sun', 'sun', 'rain',  
        'rain', 'rain', 'sun', 'sun', 'sun', 'rain', 'drizzle', 'sun',  
        'sun', 'sun', 'rain', 'drizzle', 'rain', 'rain', 'rain', 'rain',  
        'rain', 'drizzle', 'rain', 'sun', 'rain', 'rain', 'rain', 'rain',  
        'sun', 'rain', 'rain', 'sun', 'rain', 'sun', 'drizzle', 'sun',  
        'sun', 'sun', 'rain', 'snow', 'drizzle', 'fog', 'sun', 'sun',  
        'sun', 'rain', 'rain', 'fog', 'sun', 'sun', 'sun', 'rain', 'fog',  
        'rain', 'drizzle', 'sun', 'sun', 'sun', 'rain', 'rain', 'sun',  
        'sun', 'rain', 'drizzle', 'rain', 'sun', 'rain', 'rain', 'rain',  
        'sun', 'fog', 'drizzle', 'rain', 'sun', 'rain', 'sun', 'fog',
```

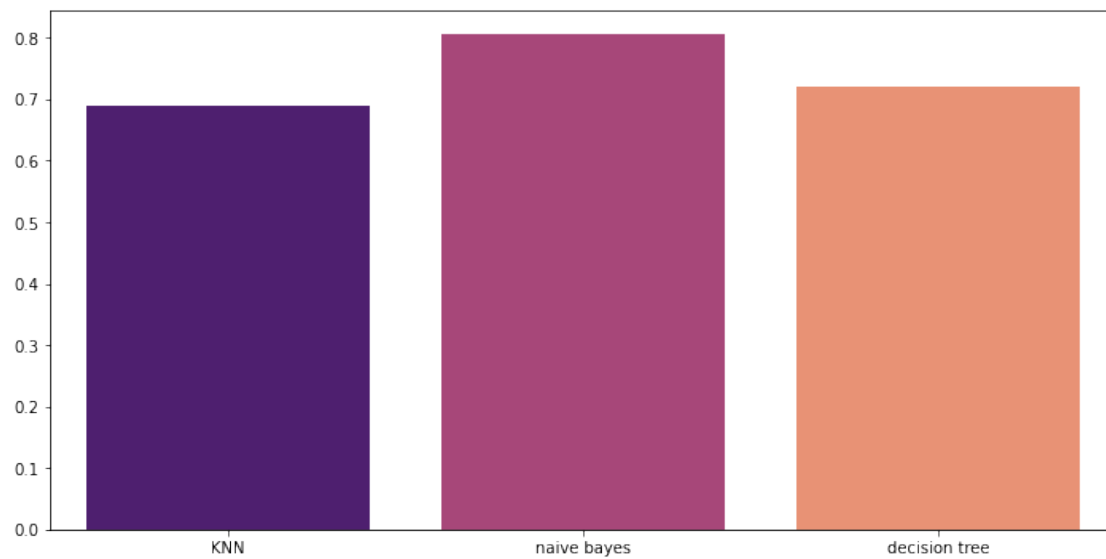
```
'rain', 'sun'], dtype=object)
```

```
[183]: from sklearn import metrics  
ac3= metrics.accuracy_score(y_test, y_pred)  
ac3
```

```
[183]: 0.7201365187713311
```

```
[184]: plt.figure(figsize=(12,6))  
model_acc = [ac1, ac2, ac3]  
model_name = [ 'KNN', 'naive bayes', 'decision tree']  
sns.barplot(x= model_name, y=model_acc, palette='magma')
```

```
[184]: <matplotlib.axes._subplots.AxesSubplot at 0x1bf4bd16340>
```



```
[ ]:
```