

sion-tree-algorithm-using-drug200

March 25, 2025

```
[8]: import pandas as pd
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, mean_squared_error
df=pd.read_csv("/content/drug200.csv")
df
```

```
[8]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
0	23	F	HIGH	HIGH	25.355	drugY
1	47	M	LOW	HIGH	13.093	drugC
2	47	M	LOW	HIGH	10.114	drugC
3	28	F	NORMAL	HIGH	7.798	drugX
4	61	F	LOW	HIGH	18.043	drugY
..
195	56	F	LOW	HIGH	11.567	drugC
196	16	M	LOW	HIGH	12.006	drugC
197	52	M	NORMAL	HIGH	9.894	drugX
198	23	M	NORMAL	NORMAL	14.020	drugX
199	40	F	LOW	NORMAL	11.349	drugX

[200 rows x 6 columns]

```
[21]: df.head()
```

```
[21]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
0	23	F	HIGH	HIGH	25.355	drugY
1	47	M	LOW	HIGH	13.093	drugC
2	47	M	LOW	HIGH	10.114	drugC
3	28	F	NORMAL	HIGH	7.798	drugX
4	61	F	LOW	HIGH	18.043	drugY

```
[18]: df.tail(10)
```

```
[18]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
190	58	M	HIGH	HIGH	18.991	drugY
191	23	M	HIGH	HIGH	8.011	drugA
192	72	M	LOW	HIGH	16.310	drugY

193	72	M	LOW	HIGH	6.769	drugC
194	46	F	HIGH	HIGH	34.686	drugY
195	56	F	LOW	HIGH	11.567	drugC
196	16	M	LOW	HIGH	12.006	drugC
197	52	M	NORMAL	HIGH	9.894	drugX
198	23	M	NORMAL	NORMAL	14.020	drugX
199	40	F	LOW	NORMAL	11.349	drugX

```
[20]: df.sample(10)
```

```
[20]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
142	60	M	HIGH	NORMAL	8.621	drugB
58	60	M	NORMAL	NORMAL	10.091	drugX
18	23	M	LOW	HIGH	7.298	drugC
129	32	F	NORMAL	HIGH	7.477	drugX
167	57	F	NORMAL	HIGH	14.216	drugX
124	53	F	HIGH	NORMAL	12.495	drugB
101	45	F	HIGH	HIGH	12.854	drugA
185	57	F	NORMAL	NORMAL	25.893	drugY
128	47	M	LOW	NORMAL	33.542	drugY
3	28	F	NORMAL	HIGH	7.798	drugX

```
[22]: df.isnull()
```

```
[22]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
0	False	False	False	False	False	False
1	False	False	False	False	False	False
2	False	False	False	False	False	False
3	False	False	False	False	False	False
4	False	False	False	False	False	False
..
195	False	False	False	False	False	False
196	False	False	False	False	False	False
197	False	False	False	False	False	False
198	False	False	False	False	False	False
199	False	False	False	False	False	False

[200 rows x 6 columns]

```
[23]: df.duplicated()
```

```
[23]:
```

0	False
1	False
2	False
3	False
4	False
...	

```

195    False
196    False
197    False
198    False
199    False
Length: 200, dtype: bool

```

```
[24]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 6 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   Age             200 non-null   int64
 1   Sex             200 non-null   object
 2   BP              200 non-null   object
 3   Cholesterol     200 non-null   object
 4   Na_to_K         200 non-null   float64
 5   Drug            200 non-null   object
dtypes: float64(1), int64(1), object(4)
memory usage: 9.5+ KB

```

```
[25]: df.describe()
```

```

[25]:
      count    Age      Na_to_K
count  200.000000  200.000000
mean    44.315000   16.084485
std     16.544315    7.223956
min     15.000000    6.269000
25%     31.000000   10.445500
50%     45.000000   13.936500
75%     58.000000   19.380000
max     74.000000   38.247000

```

```
[29]: df.columns
```

```
[29]: Index(['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K', 'Drug'], dtype='object')
```

```
[30]: df.dtypes
```

```

[30]: Age             int64
      Sex             object
      BP              object
      Cholesterol     object
      Na_to_K         float64
      Drug            object

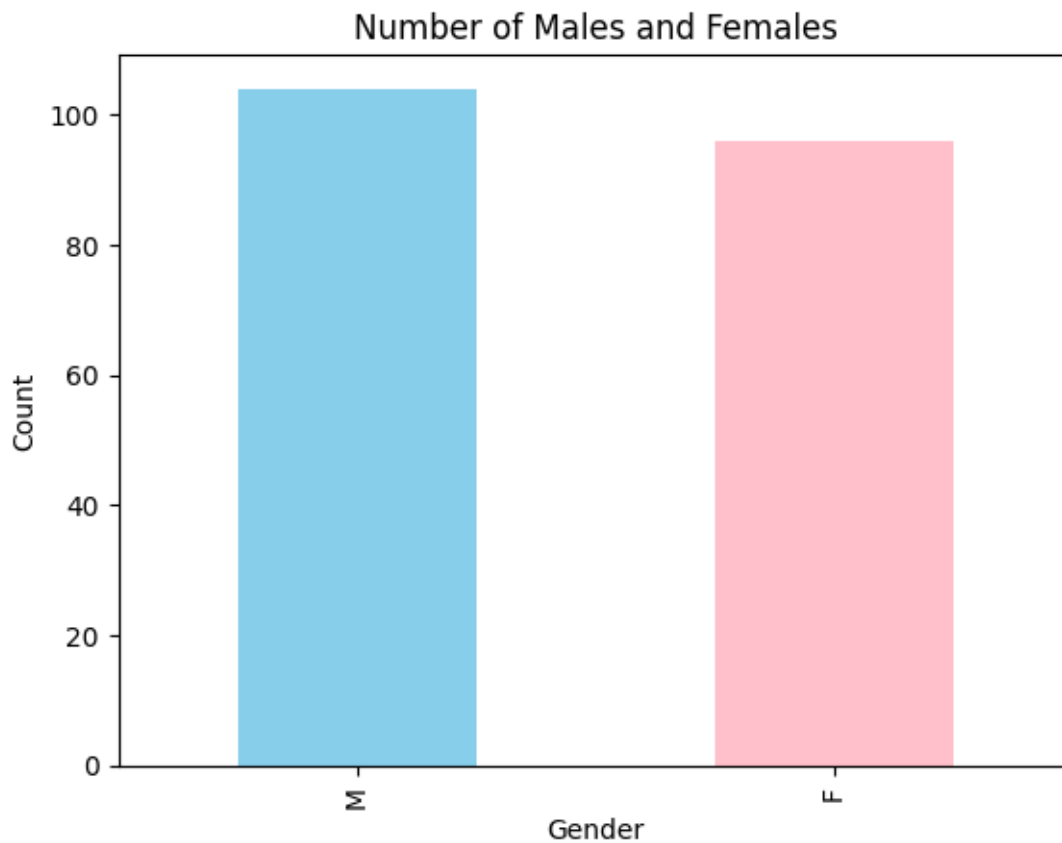
```

dtype: object

```
[31]: df.shape #number rows and columns
```

```
[31]: (200, 6)
```

```
[37]: #find the number of male and female
import matplotlib.pyplot as plt
gender_counts = df['Sex'].value_counts()
gender_counts.plot(kind='bar', color=['skyblue', 'pink'])
plt.title('Number of Males and Females')
plt.xlabel('Gender')
plt.ylabel('Count')
plt.show()
```

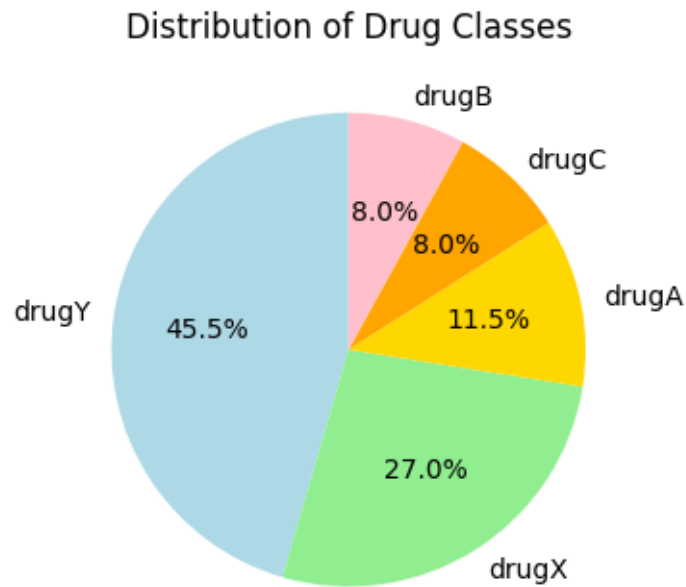


```
[44]: #finding the number of class present in drug column
drug_counts = df['Drug'].value_counts()
plt.figure(figsize=(6, 4))
plt.pie(drug_counts.values, labels=drug_counts.index, autopct='%1.1f%%',
        ↪startangle=90,
```

```

        colors=['lightblue', 'lightgreen', 'gold', 'orange', 'pink'])
plt.title('Distribution of Drug Classes')
plt.show()

```



```

[53]: from sklearn.preprocessing import LabelEncoder
      # Create a LabelEncoder
      le = LabelEncoder()
      df['Sex'] = le.fit_transform(df['Sex'])
      df['BP'] = le.fit_transform(df['BP'])
      df['Cholesterol'] = le.fit_transform(df['Cholesterol'])
      df['Drug'] = le.fit_transform(df['Drug'])
      df

```

```

[53]:
   Age  Sex  BP  Cholesterol  Na_to_K  Drug
0    23    0    0           0    25.355     4
1    47    1    1           0    13.093     2
2    47    1    1           0    10.114     2
3    28    0    2           0     7.798     3
4    61    0    1           0    18.043     4
..  ...  ...  ..          ...      ...
195  56    0    1           0    11.567     2
196  16    1    1           0    12.006     2
197  52    1    2           0     9.894     3
198  23    1    2           1    14.020     3
199  40    0    1           1    11.349     3

```

[200 rows x 6 columns]

```
[47]: X = df[['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K']]
      Y = df['Drug']
```

```
[48]: from sklearn.model_selection import train_test_split
      X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2,
      ↪random_state=42)
```

```
[49]: from sklearn.tree import DecisionTreeClassifier, export_text
      dtree=DecisionTreeClassifier(criterion='gini',max_depth=3, random_state=42)
      dtree.fit(X_train,Y_train)
```

```
[49]: DecisionTreeClassifier(max_depth=3, random_state=42)
```

```
[50]: Y_pred = dtree.predict(X_test)
```

```
[51]: from sklearn.metrics import confusion_matrix
      print("Accuracy:",accuracy_score(Y_test, Y_pred))
      print("\nConfusion Matrix:\n",confusion_matrix(Y_test,Y_pred))
```

Accuracy: 0.875

Confusion Matrix:

```
[[ 6  0  0  0  0]
 [ 0  3  0  0  0]
 [ 0  0  0  5  0]
 [ 0  0  0 11  0]
 [ 0  0  0  0 15]]
```