

final-we4tech-project

August 21, 2023

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

```
[ ]: # Load the dataset
data = pd.read_csv('/content/B07_General_Election_2019.csv')
```

```
[ ]: # Display basic information about the dataset
print(data.info())
```

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

RangeIndex: 2263 entries, 0 to 2262

Data columns (total 19 columns):

| # | Column | Non-Null Count | Dtype |
|----|---------------------|----------------|---------|
| 0 | STATE | 2263 non-null | object |
| 1 | CONSTITUENCY | 2263 non-null | object |
| 2 | NAME | 2263 non-null | object |
| 3 | WINNER | 2263 non-null | int64 |
| 4 | PARTY | 2263 non-null | object |
| 5 | SYMBOL | 2018 non-null | object |
| 6 | GENDER | 2018 non-null | object |
| 7 | CRIMINAL | 2018 non-null | object |
| 8 | AGE | 2018 non-null | float64 |
| 9 | CATEGORY | 2018 non-null | object |
| 10 | EDUCATION | 2018 non-null | object |
| 11 | ASSETS | 2018 non-null | object |
| 12 | LIABILITIES | 2018 non-null | object |
| 13 | GENERAL | 2018 non-null | object |
| 14 | POSTAL | 2263 non-null | int64 |
| 15 | TOTAL | 2263 non-null | int64 |
| 16 | OVER TOTAL ELECTORS | 2263 non-null | int64 |
| 17 | IN CONSTITUENCY | 2263 non-null | float64 |

```

17 OVER TOTAL VOTES POLLED
IN CONSTITUENCY 2263 non-null float64
18 TOTAL ELECTORS 2263 non-null int64
dtypes: float64(3), int64(5), object(11)
memory usage: 336.0+ KB
None

```

```

[ ]: # Display the first few rows of the dataset
print(data.head())

```

| | STATE | CONSTITUENCY | NAME | WINNER | PARTY | SYMBOL | \ |
|---|---------------|--------------|-----------------------|--------|-------|--------|---|
| 0 | Telangana | ADILABAD | SOYAM BAPU RAO | 1 | BJP | Lotus | |
| 1 | Telangana | ADILABAD | Godam Nagesh | 0 | TRS | Car | |
| 2 | Telangana | ADILABAD | RATHOD RAMESH | 0 | INC | Hand | |
| 3 | Telangana | ADILABAD | NOTA | 0 | NOTA | NaN | |
| 4 | Uttar Pradesh | AGRA | Satyapal Singh Baghel | 1 | BJP | Lotus | |

| | GENDER | CRIMINAL\ncases | AGE | CATEGORY | EDUCATION | \ |
|---|--------|-----------------|------|----------|---------------|---|
| 0 | MALE | 52 | 52.0 | ST | 12th Pass | |
| 1 | MALE | 0 | 54.0 | ST | Post Graduate | |
| 2 | MALE | 3 | 52.0 | ST | 12th Pass | |
| 3 | NaN | NaN | NaN | NaN | NaN | |
| 4 | MALE | 5 | 58.0 | SC | Doctorate | |

| | ASSETS | LIABILITIES | GENERAL\ncases | \ |
|---|-----------------------------|-----------------------------|----------------|---|
| 0 | Rs 30,99,414\n ~ 30 Lacs+ | Rs 2,31,450\n ~ 2 Lacs+ | 376892 | |
| 1 | Rs 1,84,77,888\n ~ 1 Crore+ | Rs 8,47,000\n ~ 8 Lacs+ | 318665 | |
| 2 | Rs 3,64,91,000\n ~ 3 Crore+ | Rs 1,53,00,000\n ~ 1 Crore+ | 314057 | |
| 3 | NaN | NaN | 13030 | |
| 4 | Rs 7,42,74,036\n ~ 7 Crore+ | Rs 86,06,522\n ~ 86 Lacs+ | 644459 | |

| | POSTAL\ncases | TOTAL\ncases | OVER TOTAL ELECTORS \ncases | IN CONSTITUENCY | \ |
|---|---------------|--------------|-----------------------------|-----------------|---|
| 0 | 482 | 377374 | 25.330684 | | |
| 1 | 149 | 318814 | 21.399929 | | |
| 2 | 181 | 314238 | 21.092771 | | |
| 3 | 6 | 13036 | 0.875023 | | |
| 4 | 2416 | 646875 | 33.383823 | | |

| | OVER TOTAL VOTES POLLED \ncases | IN CONSTITUENCY | TOTAL ELECTORS |
|---|---------------------------------|-----------------|----------------|
| 0 | 35.468248 | 1489790 | |
| 1 | 29.964370 | 1489790 | |
| 2 | 29.534285 | 1489790 | |
| 3 | 1.225214 | 1489790 | |
| 4 | 56.464615 | 1937690 | |

```

[ ]: # Summary statistics of numerical columns
print(data.describe())

```

| | WINNER | AGE | GENERAL\nVOTES | POSTAL\nVOTES | TOTAL\nVOTES \ |
|-------|-------------|-------------|----------------|---------------|----------------|
| count | 2263.000000 | 2018.000000 | 2.263000e+03 | 2263.000000 | 2.263000e+03 |
| mean | 0.238179 | 52.273538 | 2.615991e+05 | 990.710561 | 2.625898e+05 |
| std | 0.426064 | 11.869373 | 2.549906e+05 | 1602.839174 | 2.559822e+05 |
| min | 0.000000 | 25.000000 | 1.339000e+03 | 0.000000 | 1.342000e+03 |
| 25% | 0.000000 | 43.250000 | 2.103450e+04 | 57.000000 | 2.116250e+04 |
| 50% | 0.000000 | 52.000000 | 1.539340e+05 | 316.000000 | 1.544890e+05 |
| 75% | 0.000000 | 61.000000 | 4.858040e+05 | 1385.000000 | 4.872315e+05 |
| max | 1.000000 | 86.000000 | 1.066824e+06 | 19367.000000 | 1.068569e+06 |

| | OVER TOTAL ELECTORS \nIN CONSTITUENCY \ |
|-------|---|
| count | 2263.000000 |
| mean | 15.811412 |
| std | 14.962861 |
| min | 0.097941 |
| 25% | 1.296518 |
| 50% | 10.510553 |
| 75% | 29.468185 |
| max | 51.951012 |

| | OVER TOTAL VOTES POLLED \nIN CONSTITUENCY | TOTAL ELECTORS |
|-------|---|----------------|
| count | 2263.000000 | 2.263000e+03 |
| mean | 23.190525 | 1.658016e+06 |
| std | 21.564758 | 3.145187e+05 |
| min | 1.000039 | 5.518900e+04 |
| 25% | 1.899502 | 1.530014e+06 |
| 50% | 16.221721 | 1.679030e+06 |
| 75% | 42.590233 | 1.816857e+06 |
| max | 74.411856 | 3.150313e+06 |

```
[ ]: # Check for missing values
print(data.isnull().sum())
```

| | |
|-----------------|-----|
| STATE | 0 |
| CONSTITUENCY | 0 |
| NAME | 0 |
| WINNER | 0 |
| PARTY | 0 |
| SYMBOL | 245 |
| GENDER | 245 |
| CRIMINAL\nCASES | 245 |
| AGE | 245 |
| CATEGORY | 245 |
| EDUCATION | 245 |
| ASSETS | 245 |
| LIABILITIES | 245 |
| GENERAL\nVOTES | 0 |
| POSTAL\nVOTES | 0 |

```

TOTAL\nVOTES                                0
OVER TOTAL ELECTORS \nIN CONSTITUENCY        0
OVER TOTAL VOTES POLLED \nIN CONSTITUENCY    0
TOTAL ELECTORS                              0
dtype: int64

```

```

[ ]: # Handling missing values
data.dropna(inplace=True)

```

```

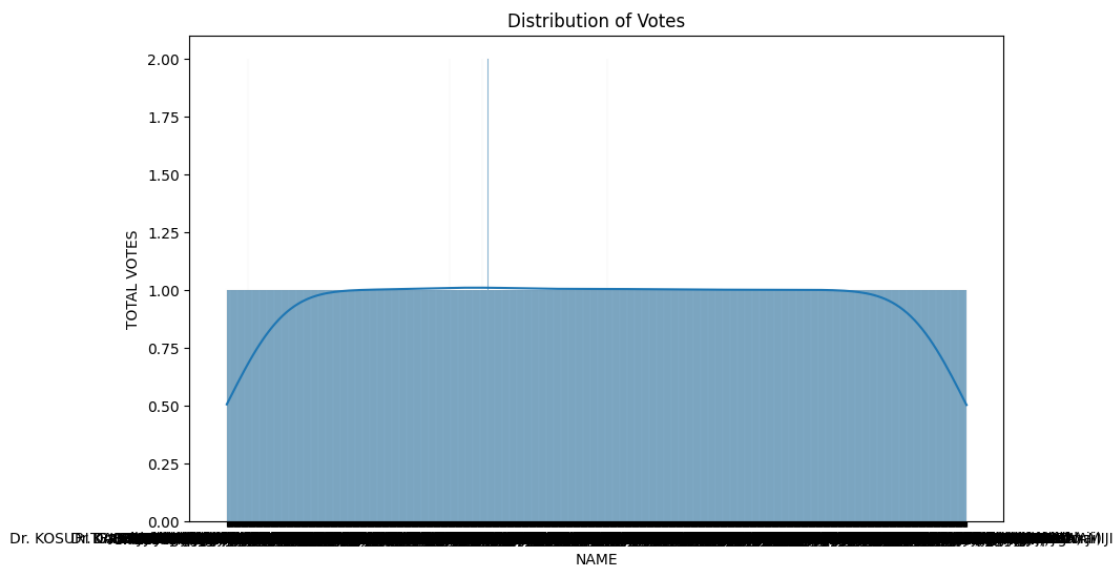
[ ]: # Handling duplicates
data.drop_duplicates(inplace=True)

```

```

[ ]: # Histogram of a numeric variable
plt.figure(figsize=(10, 6))
sns.histplot(data['NAME'], bins=20, kde=True)
plt.title('Distribution of Votes')
plt.xlabel('NAME')
plt.ylabel('TOTAL VOTES')
plt.show()

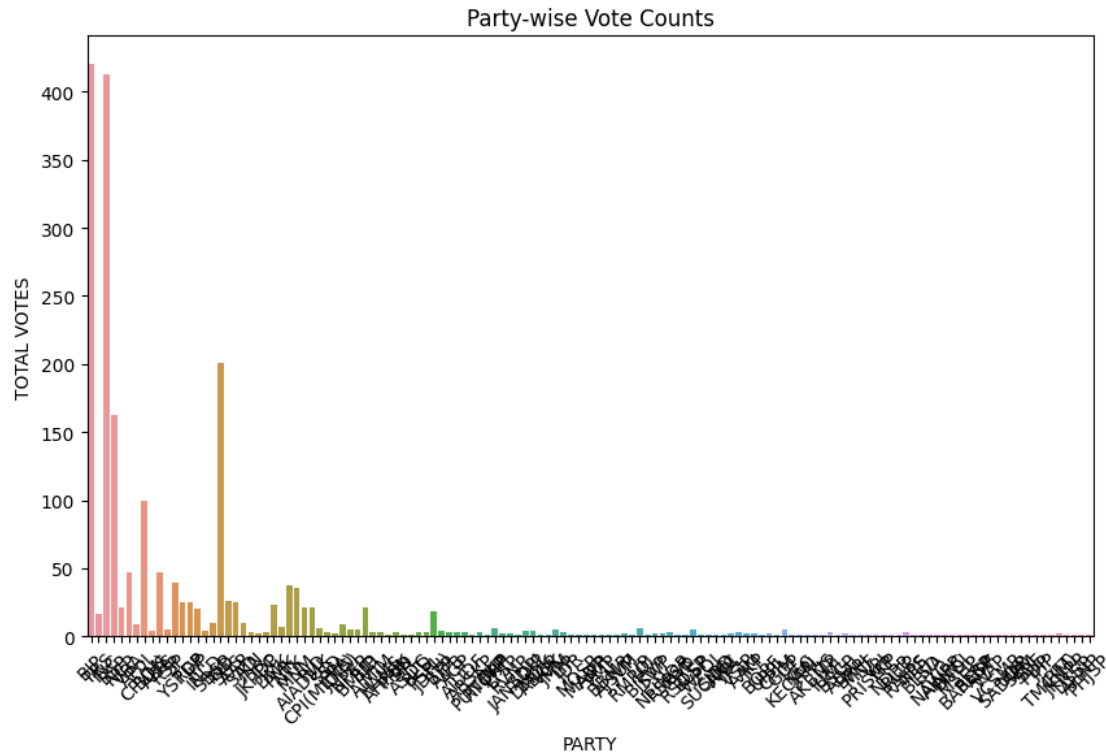
```



```

[ ]: # Count plot of a categorical variable
plt.figure(figsize=(10, 6))
sns.countplot(x='PARTY', data=data)
plt.title('Party-wise Vote Counts')
plt.xticks(rotation=45)
plt.xlabel('PARTY')
plt.ylabel('TOTAL VOTES')
plt.show()

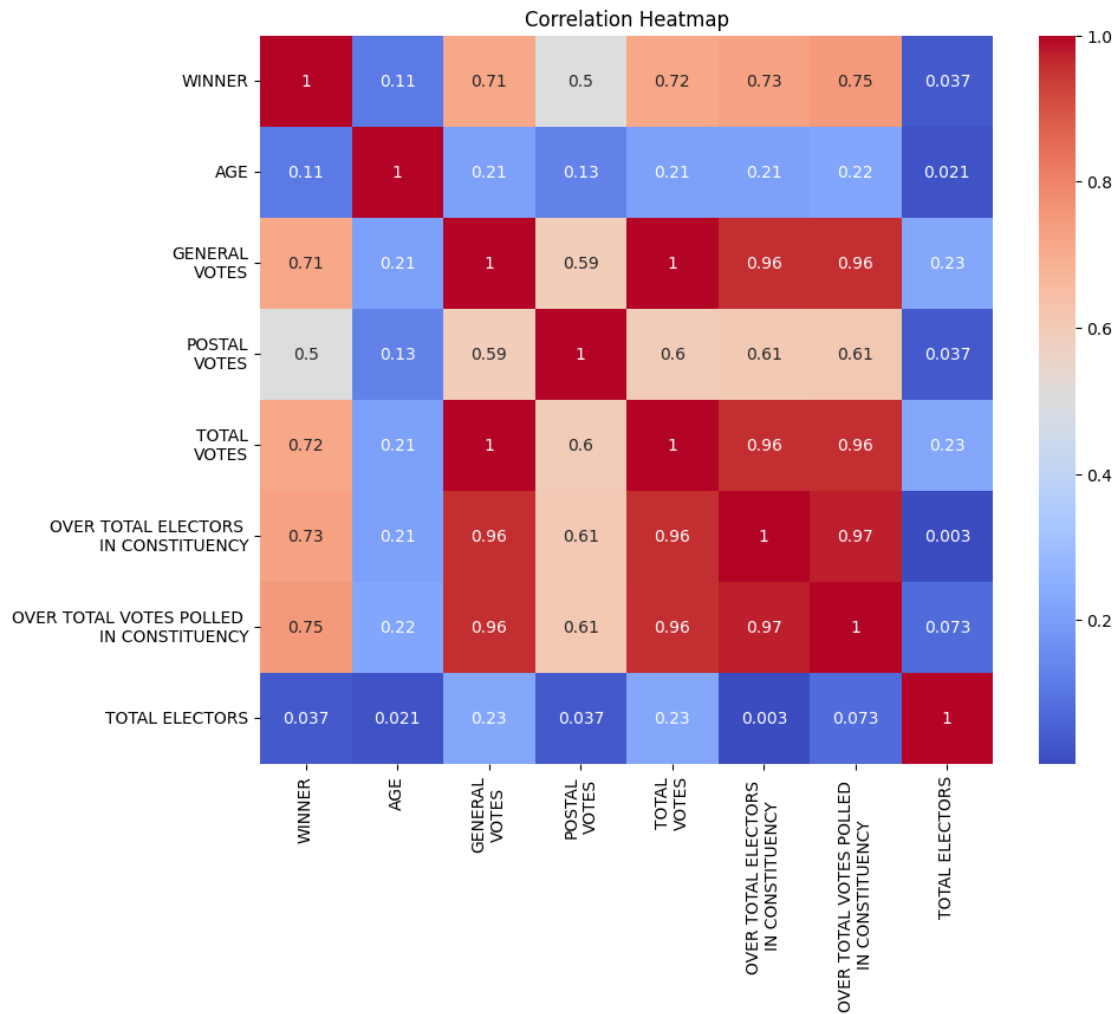
```



```
[ ]: # Correlation heatmap for numeric variables
plt.figure(figsize=(10, 8))
sns.heatmap(data.corr(), annot=True, cmap='coolwarm')
plt.title('Correlation Heatmap')
plt.show()
```

<ipython-input-15-cc94773b3ac5>:3: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.

```
sns.heatmap(data.corr(), annot=True, cmap='coolwarm')
```



```
[ ]: # Countplot of votes by state
plt.figure(figsize=(12, 6))
sns.countplot(data=data, x='STATE', hue='PARTY')
plt.title('Votes by State and Party')
plt.xlabel('State')
plt.ylabel('Number of Votes')
plt.xticks(rotation=45)
plt.legend(title='Party')
plt.show()
```



```
[ ]: c1=0
      c2=0
      c3=0
      for i in data['AGE']:
          if i>=60 :
              c1=c1 + 1
          elif i>30:
              c2=c2+1
          elif i>0:
              c3=c3+1
      print("Age greater than 60:", c1)
      print("Age between 30 and 60:", c2)
      print("Age less than 30:", c3)
```

Age greater than 60: 606
 Age between 30 and 60: 1344
 Age less than 30: 68

```
[ ]: df=data.groupby(['PARTY'])['WINNER'].sum()
      df1=df.sort_values( ascending=False).head(10)
      print(df1)
```

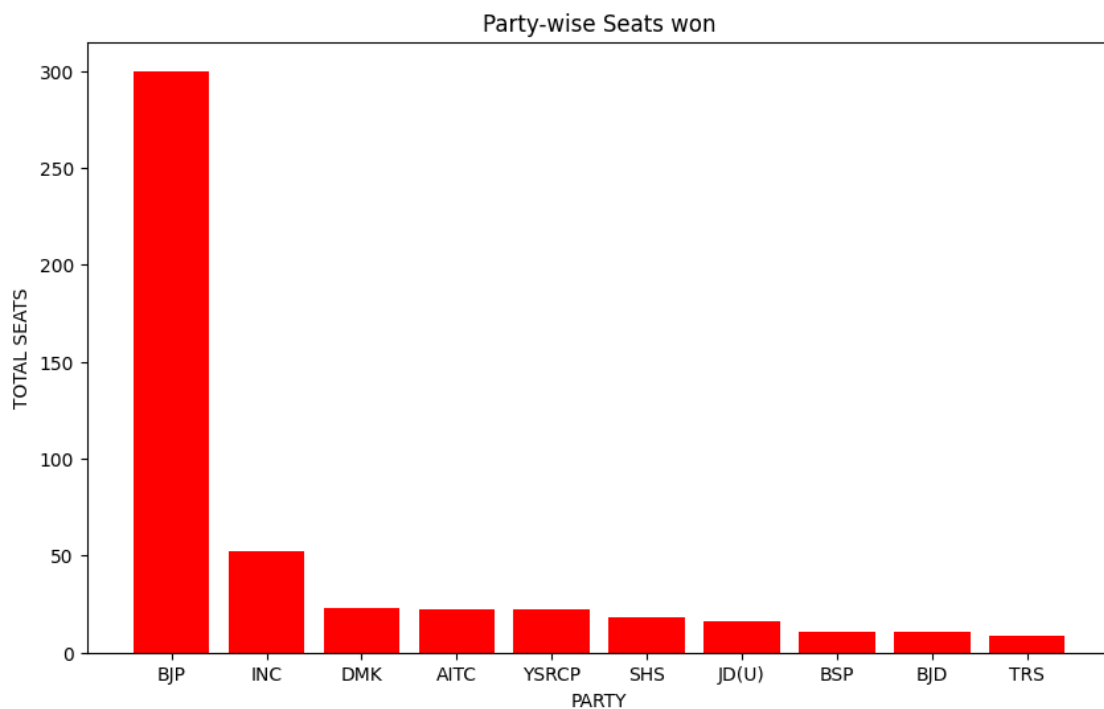
```
PARTY
BJP      300
INC       52
DMK       23
AITC      22
YSRCP     22
SHS       18
JD(U)     16
BSP       11
BJD       11
TRS        9
Name: WINNER, dtype: int64
```

```
[ ]: df2=pd.DataFrame({'PARTY':df1.index, 'SEATS':df1.values})
      print(df2)
```

```
   PARTY  SEATS
0    BJP    300
1    INC     52
2    DMK     23
3    AITC     22
4  YSRCP     22
5    SHS     18
```


| | | |
|---|-------|----|
| 6 | JD(U) | 16 |
| 7 | BSP | 11 |
| 8 | BJD | 11 |
| 9 | TRS | 9 |

```
[ ]: # Count plot of a categorical variable
plt.figure(figsize=(10, 6))
#sns.countplot(x='PARTY', data=df2)
plt.bar(df2['PARTY'], df2['SEATS'], color='r')
plt.title('Party-wise Seats won')
plt.xlabel('PARTY')
plt.ylabel('TOTAL SEATS')
plt.show()
```



```
[ ]: edu=data.groupby(['PARTY', 'EDUCATION'])['NAME'].count()
#edu1=edu.sort_values(ascending=False)
print(edu)
```

| PARTY | EDUCATION | |
|-------|-----------|---|
| AAM | 8th Pass | 1 |
| | Literate | 1 |
| AAP | 10th Pass | 1 |
| | 12th Pass | 5 |
| | 8th Pass | 1 |
| | .. | |

```
YSRCP  Doctorate      1
      Graduate       5
      Graduate Professional  6
      Post Graduate   10
ravp   Doctorate      1
Name: NAME, Length: 372, dtype: int64
```

```
[ ]: cat=data.groupby(['PARTY','CATEGORY'])['NAME'].count()
      #cat1=cat.sort_values(ascending=False)
      print(cat)
```

```
PARTY  CATEGORY
AAM     GENERAL      1
      SC             1
AAP     GENERAL     20
      SC             5
ABGP    ST           1
      ..
YKP     GENERAL      1
YSRCP   GENERAL     20
      SC             4
      ST             1
ravp    GENERAL      1
Name: NAME, Length: 198, dtype: int64
```

```
[ ]: gen=data.groupby(['PARTY','GENDER'])['NAME'].count()
      #gen1=gen.sort_values(ascending=False)
      print(gen)
```

```
PARTY  GENDER
AAM     FEMALE      1
      MALE          1
AAP     FEMALE      3
      MALE         22
ABGP    MALE          1
      ..
WPOI    MALE          1
YKP     MALE          1
YSRCP   FEMALE      4
      MALE         21
ravp    MALE          1
Name: NAME, Length: 171, dtype: int64
```