

## IMPORT OF MODULES FOR ANALYSIS

```
In [36]: # Import necessary Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import geopandas as gpd
import plotly.express as px
```

```
In [57]: ts = pd.read_csv("D:/VISUALISATION_WORKSHOP/tselection/DATASET/Telangana_Assembly_election_2023.csv")
df = pd.read_csv("D:/VISUALISATION_WORKSHOP/tselection/DATASET/Telangana_Assembly_election_2023.csv")
#Importing the spatial data
gdf = gpd.read_file("D:/VISUALISATION_WORKSHOP/tselection/DATASET/telangana_ac.geojson.txt")
#reference file
no_gdf = pd.read_csv("D:/VISUALISATION_WORKSHOP/tselection/DATASET/Telangana_Constituency_No.csv")
```

```
In [40]: ts.head()
```

Out[40]:

	Unnamed: 0	S.N.	Candidate	Party	EVM Votes	Postal Votes	Total Votes	% of Votes	Constituency
0	0	1	GUVVALA BALARAJU	Bharat Rashtra Samithi	65661	350	66011	33.74	82 - Achampet
1	1	2	CHIKKUDU VAMSHI KRISHNA	Indian National Congress	113761	1576	115337	58.96	82 - Achampet
2	2	3	DEVANI SATHYANARAYANA ALIAS DEVANI SATISH MADIGA	Bharatiya Janata Party	4201	66	4267	2.18	82 - Achampet
3	3	4	MOTHUKURI NAGARJUN	Bahujan Samaj Party	1174	15	1189	0.61	82 - Achampet
4	4	5	KUNDA MALLIKARJUN	Yuga Thulasi Party	239	0	239	0.12	82 - Achampet

```
In [41]: ts.tail()
```

Out[41]:

	Unnamed: 0	S.N.	Candidate	Party	EVM Votes	Postal Votes	Total Votes	% of Votes	Constituency
2404	2404	19	RAMULU HUGGELLY	Independent	665	3	668	0.32	38 - Zahirabad
2405	2405	20	SHIVASHANKAR	Independent	137	0	137	0.07	38 - Zahirabad
2406	2406	21	SRINIVAS	Independent	85	0	85	0.04	38 - Zahirabad
2407	2407	22	HEMANAND TABLA	Independent	82	0	82	0.04	38 - Zahirabad
2408	2408	23	NOTA None of the Above		611	2	613	0.29	38 - Zahirabad

```
In [58]: #Extracting the number from constituency column
df['Constituency_no'] = df['Constituency'].str.extract(r'(\d+)')

#Extracting the text from constituency column
df['Constituency_Alphabets'] = df['Constituency'].str.extract(r'\d+ - (.+)$')

#dropping columns
df = df.drop(['Unnamed: 0', 'S.N.', 'Constituency'], axis = 1)
#Renaming columns
df.rename(columns={'Constituency_Alphabets': 'Constituency'}, inplace=True)
#head data
print("Sample of data")
df.head(3)
```

Sample of data

Out[58]:

	Candidate	Party	EVM Votes	Postal Votes	Total Votes	% of Votes	Constituency_no	Constituency
0	GUVVALA BALARAJU	Bharat Rashtra Samithi	65661	350	66011	33.74	82	Achampet
1	CHIKKUDU VAMSHI KRISHNA	Indian National Congress	113761	1576	115337	58.96	82	Achampet
2	DEVANI SATHYANARAYANA ALIAS DEVANI SATISH MADIGA	Bharatiya Janata Party	4201	66	4267	2.18	82	Achampet

```
In [59]: gdf_merged = pd.merge(gdf, no_gdf, on = 'id', suffixes = ('', '_no_gdf'))

#Constituency no column as Integer
df['Constituency_no'] = df['Constituency_no'].astype(int)

spatial_df_d = gdf_merged.loc[:,['id', 'assembly', 'constituency_no', 'geometry']]

spatial_df_total = pd.merge(spatial_df_d, df, left_on = 'constituency_no', right_on = 'Constituency_no')

spatial_df = spatial_df_total.loc[:,['id', 'constituency_no', 'Constituency', 'geometry']]

spatial_df = spatial_df.drop_duplicates()
print("Sample of Spatial data")
spatial_df.head()
```

Sample of Spatial data

Out[59]:

	id	constituency_no	Constituency	geometry
0	1	3	Bellampalli	MULTIPOLYGON (((79.48229 19.19617, 79.48282 19...
14	2	8	Boath	MULTIPOLYGON (((78.34869 19.88415, 78.34989 19...
25	3	2	Chennur	MULTIPOLYGON (((79.78546 19.05654, 79.78529 19...
40	4	6	Khanapur	MULTIPOLYGON (((79.01319 19.21401, 79.01499 19...
52	5	4	Mancherla	MULTIPOLYGON (((79.15034 19.09220, 79.15041 19...

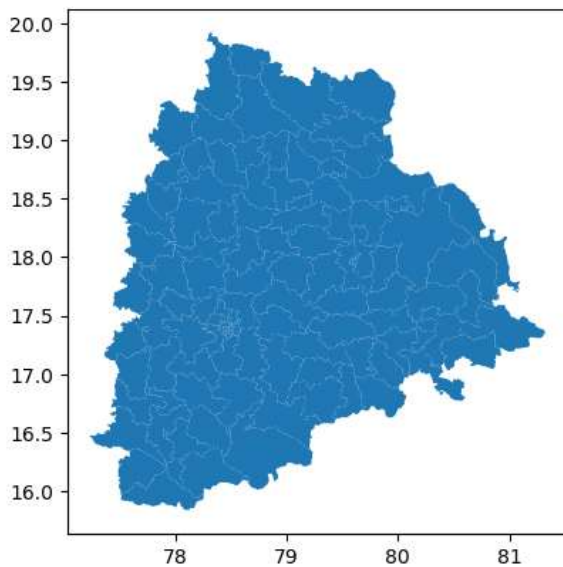
## # TELANGANA MAP WITH GEOPANDAS

```
In [60]: print('Map of Telangana State with Constituency boundaries')

spatial_df.plot()
```

Map of Telangana State with Constituency boundaries

Out[60]: <Axes: >



In [ ]:

```
In [3]: #FINDING THE DESCRIBE
ts.describe()
```

Out[3]:

	Unnamed: 0	S.N.	EVM Votes	Postal Votes	Total Votes	% of Votes
count	2409.000000	2409.000000	2409.000000	2409.000000	2409.000000	2409.000000
mean	1204.000000	12.148194	9646.899543	83.218348	9730.117891	4.939797
std	695.562722	8.479849	25608.097526	282.182086	25833.309113	12.762166
min	0.000000	1.000000	17.000000	0.000000	17.000000	0.010000
25%	602.000000	6.000000	151.000000	0.000000	152.000000	0.080000
50%	1204.000000	11.000000	393.000000	1.000000	394.000000	0.200000
75%	1806.000000	17.000000	1388.000000	7.000000	1394.000000	0.730000
max	2408.000000	49.000000	187327.000000	4501.000000	187999.000000	64.890000

```
In [6]: ts.dtypes
```

Out[6]:

Unnamed: 0	int64
S.N.	int64
Candidate	object
Party	object
EVM Votes	int64
Postal Votes	int64
Total Votes	int64
% of Votes	float64
Constituency	object

dtype: object

```
In [10]: #FINDING THE MEAN
mean_values = ts.mean(numeric_only=True)
print("Mean:")
print(mean_values)

#here showing the mean from evm votes , postal votes , total votes and percentage of votes.
```

Mean:  
Unnamed: 0      1204.000000  
S.N.            12.148194  
EVM Votes      9646.899543  
Postal Votes    83.218348  
Total Votes    9730.117891  
% of Votes      4.939797  
dtype: float64

```
In [13]: #finding the median
median_values = ts.median(numeric_only=True)
print("\nMedian:")
print(median_values)

#here showing the median from evm votes , postal votes , total votes and percentage of votes.
```

Median:  
Unnamed: 0      1204.0  
S.N.            11.0  
EVM Votes      393.0  
Postal Votes    1.0  
Total Votes    394.0  
% of Votes      0.2  
dtype: float64

```
In [15]: #finding the mode
mode_values = ts.mode().iloc[0] # Using iloc[0] to handle multiple modes in a column
print("\nMode:")
print(mode_values)

#here, it is showing the mode (center) value from dataset
```

```
Mode:
Unnamed: 0          0
S.N.                1.0
Candidate           NOTA
Party              Independent
EVM Votes          128.0
Postal Votes        0.0
Total Votes        149.0
% of Votes          0.04
Constituency      49 - Lal Bahadur Nagar
Name: 0, dtype: object
```

```
In [17]: #finding the
# unimodal = means single mode value
# bimodal = two modes from the dataset
# multimodal = two or more modes from dataset

mode_counts = ts.apply(lambda x: x.value_counts().iloc[0])

# Identify if the distribution is unimodal, bimodal, or multimodal
num_modes = mode_counts.value_counts().index.size
if num_modes == 1:
    print("Unimodal distribution")
elif num_modes == 2:
    print("Bimodal distribution")
else:
    print("Multimodal distribution")

# Find the value that repeats the maximum number of times
max_repeats = mode_counts.max()
max_repeats_value = mode_counts[mode_counts == max_repeats].index[0]

print("Value that repeats maximum times:", max_repeats_value)
```

```
Multimodal distribution
Value that repeats maximum times: Party
```

```
In [19]: # Calculate the variance for numerical columns
variance_values = ts.var(numeric_only=True)

print("Variance:")
print(variance_values)
```

```
Variance:
Unnamed: 0    4.838075e+05
S.N.          7.190785e+01
EVM Votes     6.557747e+08
Postal Votes   7.962673e+04
Total Votes    6.673599e+08
% of Votes     1.628729e+02
dtype: float64
```

```
In [20]: # Find the maximum value for numerical columns
max_values = ts.max()

# Find the minimum value for numerical columns
min_values = ts.min()

print("Maximum values:")
print(max_values)
print("\nMinimum values:")
print(min_values)
```

```
Maximum values:
Unnamed: 0          2408
S.N.                49
Candidate          ZEENATH BEGUM
Party              Yuva Taram Party
EVM Votes          187327
Postal Votes        4501
Total Votes        187999
% of Votes          64.89
Constituency       99 - Ghanpur (Station)
dtype: object
```

```
Minimum values:
Unnamed: 0          0
S.N.                1
Candidate          A ANJANEYA CHARY
Party              Aabaad Party
EVM Votes          17
Postal Votes        0
Total Votes         17
% of Votes          0.01
Constituency       1 - Sirpur
dtype: object
```

```
In [24]: # Calculate the skewness for numerical columns
skewness_values = ts.skew(numeric_only=True)

print("Skewness:")
print(skewness_values)
```

```
Skewness:
Unnamed: 0    0.000000
S.N.          1.011864
EVM Votes     3.030310
Postal Votes  5.971769
Total Votes   3.029969
% of Votes    2.773123
dtype: float64
```

```
In [25]: # Classify skewness
negative_skew = skewness_values[skewness_values < 0].index.tolist()
positive_skew = skewness_values[skewness_values > 0].index.tolist()
normal_skew = skewness_values[skewness_values.abs() < 0.5].index.tolist()

print("Negative Skewness:")
print(negative_skew)
print("\nPositive Skewness:")
print(positive_skew)
print("\nApproximately Symmetric (Skewness close to 0):")
print(normal_skew)
```

```
Negative Skewness:
[]
```

```
Positive Skewness:
['S.N.', 'EVM Votes', 'Postal Votes', 'Total Votes', '% of Votes']
```

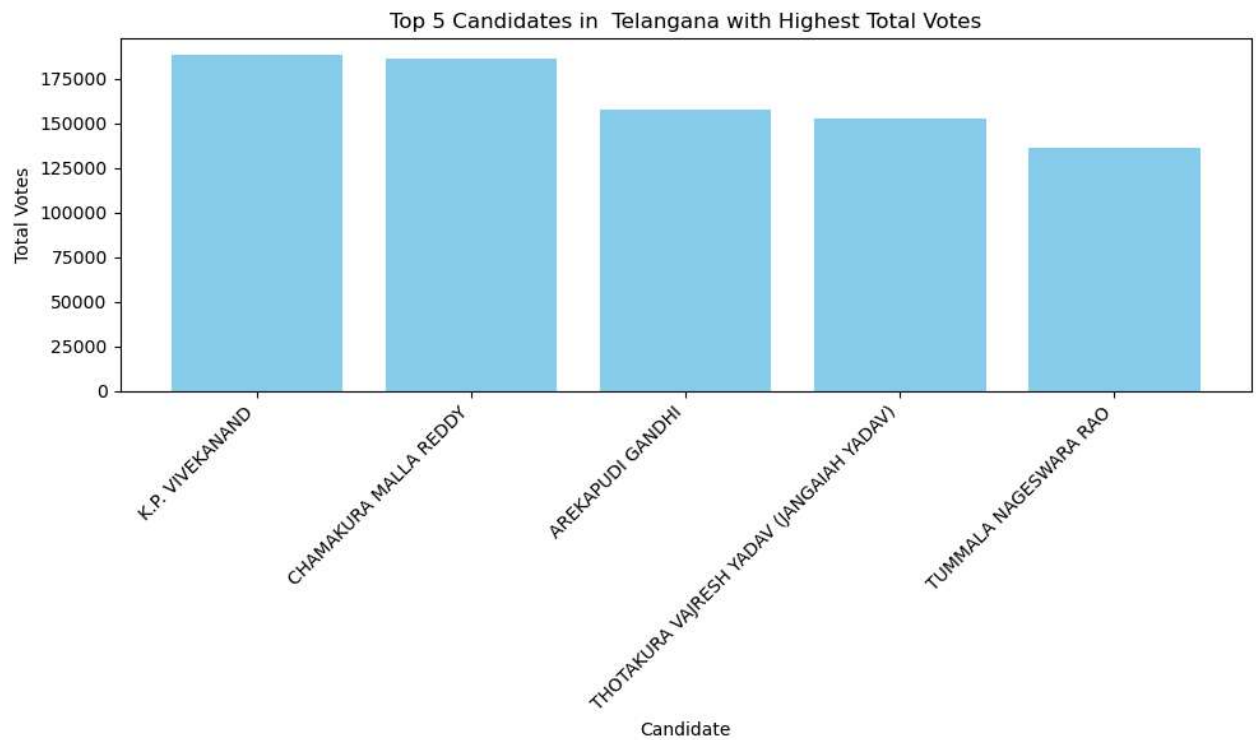
```
Approximately Symmetric (Skewness close to 0):
['Unnamed: 0']
```

```
In [30]: # Sort the DataFrame by total votes in descending order
sorted_df = ts.sort_values(by='Total Votes', ascending=False)

# Select the top 5 candidates
top_5 = sorted_df.head(5)

# Create a bar plot
plt.figure(figsize=(10, 6))
plt.bar(top_5['Candidate'], top_5['Total Votes'], color='skyblue')
plt.xlabel('Candidate')
plt.ylabel('Total Votes')
plt.title('Top 5 Candidates in Telangana with Highest Total Votes')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()

# Display the plot
plt.show()
```



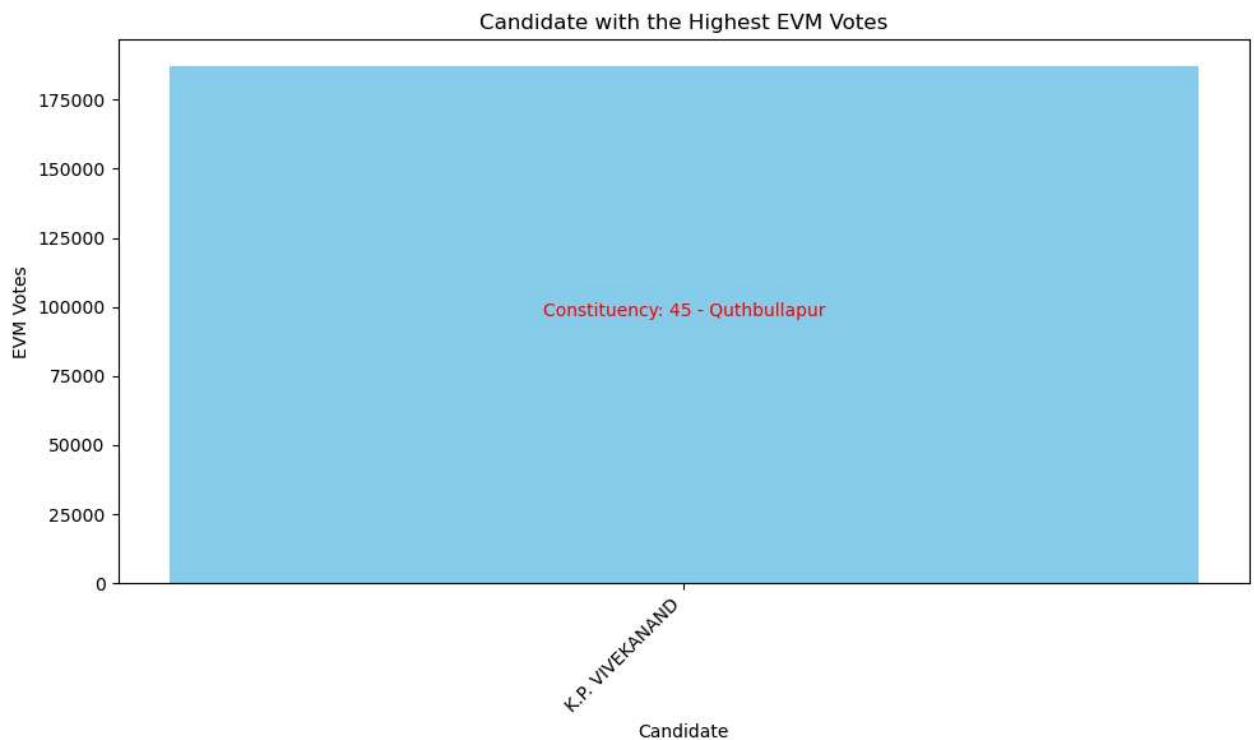
```
In [33]: # Find the candidate with the highest number of EVM votes
candidate_highest_evm_votes = ts.loc[ts['EVM Votes'].idxmax()]

# Get the name and constituency of the candidate
candidate_name = candidate_highest_evm_votes['Candidate']
constituency = candidate_highest_evm_votes['Constituency']
evm_votes = candidate_highest_evm_votes['EVM Votes']

# Create a bar plot
plt.figure(figsize=(10, 6))
plt.bar(candidate_name, evm_votes, color='skyblue')
plt.xlabel('Candidate')
plt.ylabel('EVM Votes')
plt.title('Candidate with the Highest EVM Votes')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()

# Annotate the constituency
plt.annotate(f'Constituency: {constituency}', xy=(0.5, 0.5), xycoords='axes fraction', ha='center', va='center', fontweight='bold', color='red')

# Display the plot
plt.show()
```



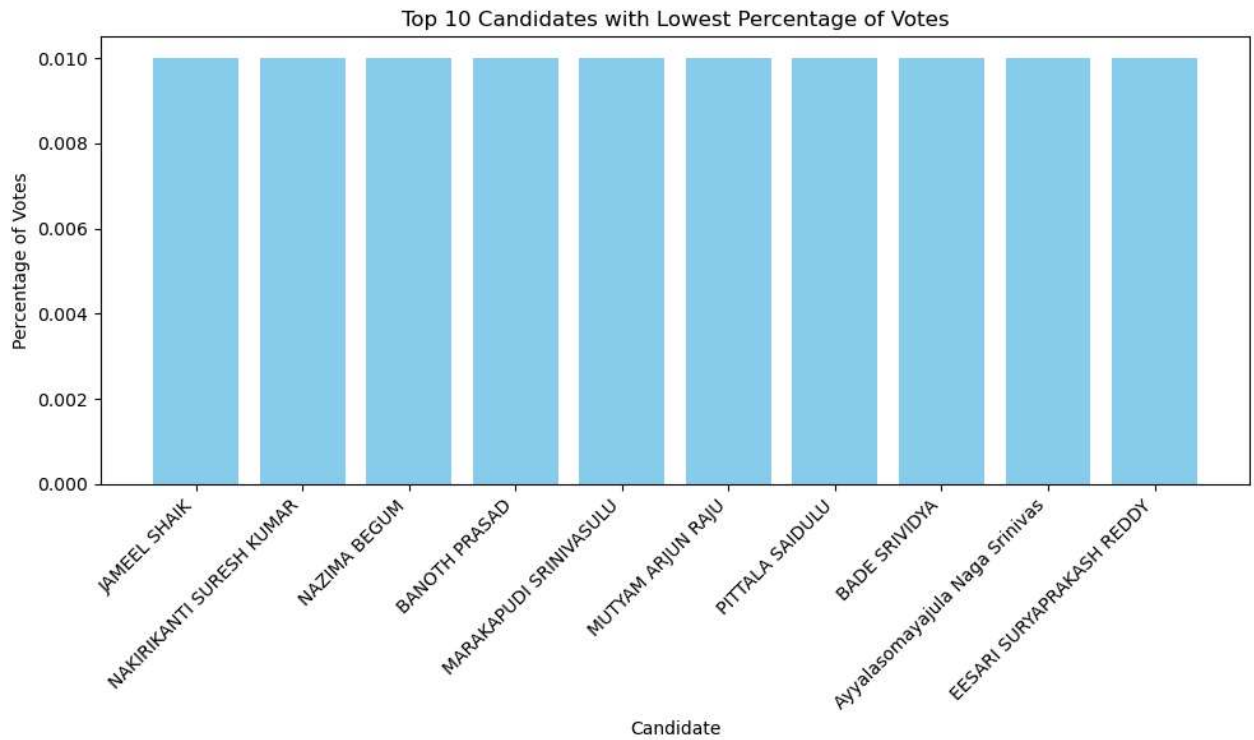
```
In [35]: # Print the column names
print(ts.columns)

Index(['Unnamed: 0', 'S.N.', 'Candidate', 'Party', 'EVM Votes', 'Postal Votes',
      'Total Votes', '% of Votes', 'Constituency'],
      dtype='object')
```

```
In [36]: # Sort the DataFrame by percentage of votes in ascending order and select the top 10
top_10_lowest_percentage = ts.nsmallest(10, '% of Votes')

# Create a bar plot
plt.figure(figsize=(10, 6))
plt.bar(top_10_lowest_percentage['Candidate'], top_10_lowest_percentage['% of Votes'], color='skyblue')
plt.xlabel('Candidate')
plt.ylabel('Percentage of Votes')
plt.title('Top 10 Candidates with Lowest Percentage of Votes')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()

# Display the plot
plt.show()
```



```
In [40]: # Sort the DataFrame by percentage of votes in ascending order and select the top 10
top_10_lowest_percentage = ts.nsmallest(10, '% of Votes')

# Display the candidate names and their corresponding constituency names with the lowest percentage of votes
candidate_and_constituency_names = top_10_lowest_percentage[['Candidate', 'Constituency']]
print("Candidate names and their corresponding constituency names with the lowest percentage of votes:")
print(candidate_and_constituency_names.to_string(index=False))
```

Candidate names and their corresponding constituency names with the lowest percentage of votes:

Candidate	Constituency
JAMEEL SHAIK	112 - Khammam
NAKIRIKANTI SURESH KUMAR	112 - Khammam
NAZIMA BEGUM	112 - Khammam
BANOTH PRASAD	112 - Khammam
MARAKAPUDI SRINIVASULU	112 - Khammam
MUTYAM ARJUN RAJU	112 - Khammam
PITTALA SAIDULU	90 - Kodad
BADE SRIVIDYA	90 - Kodad
Ayyalasomayajula Naga Srinivas	49 - Lal Bahadur Nagar
EESARI SURYAPRAKASH REDDY	49 - Lal Bahadur Nagar



```
In [41]: # Descriptive statistics for integer columns
print("Descriptive statistics for integer columns:")
print(ts[['EVM Votes', 'Postal Votes', 'Total Votes']].describe())

# Descriptive statistics for float columns
print("\nDescriptive statistics for float columns:")
print(ts['% of Votes'].describe())

# Count of unique values for object columns
print("\nCount of unique values for object columns:")
print(ts[['Candidate', 'Party', 'Constituency']].nunique())

# Create histograms for integer columns
plt.figure(figsize=(12, 6))
plt.subplot(1, 3, 1)
sns.histplot(ts['EVM Votes'], kde=True, color='skyblue')
plt.title('EVM Votes Distribution')
plt.subplot(1, 3, 2)
sns.histplot(ts['Postal Votes'], kde=True, color='salmon')
plt.title('Postal Votes Distribution')
plt.subplot(1, 3, 3)
sns.histplot(ts['Total Votes'], kde=True, color='green')
plt.title('Total Votes Distribution')
plt.tight_layout()
plt.show()

# Create histogram for float column
plt.figure(figsize=(8, 6))
sns.histplot(ts['% of Votes'], kde=True, color='purple')
plt.title('Percentage of Votes Distribution')
plt.show()

# Create bar plot for top parties
top_parties = ts['Party'].value_counts().nlargest(10)
plt.figure(figsize=(10, 6))
top_parties.plot(kind='bar', color='orange')
plt.title('Top 10 Parties by Count')
plt.xlabel('Party')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

Descriptive statistics for integer columns:

	EVM Votes	Postal Votes	Total Votes
count	2409.000000	2409.000000	2409.000000
mean	9646.899543	83.218348	9730.117891
std	25608.097526	282.182086	25833.309113
min	17.000000	0.000000	17.000000
25%	151.000000	0.000000	152.000000
50%	393.000000	1.000000	394.000000
75%	1388.000000	7.000000	1394.000000
max	187327.000000	4501.000000	187999.000000

Descriptive statistics for float columns:

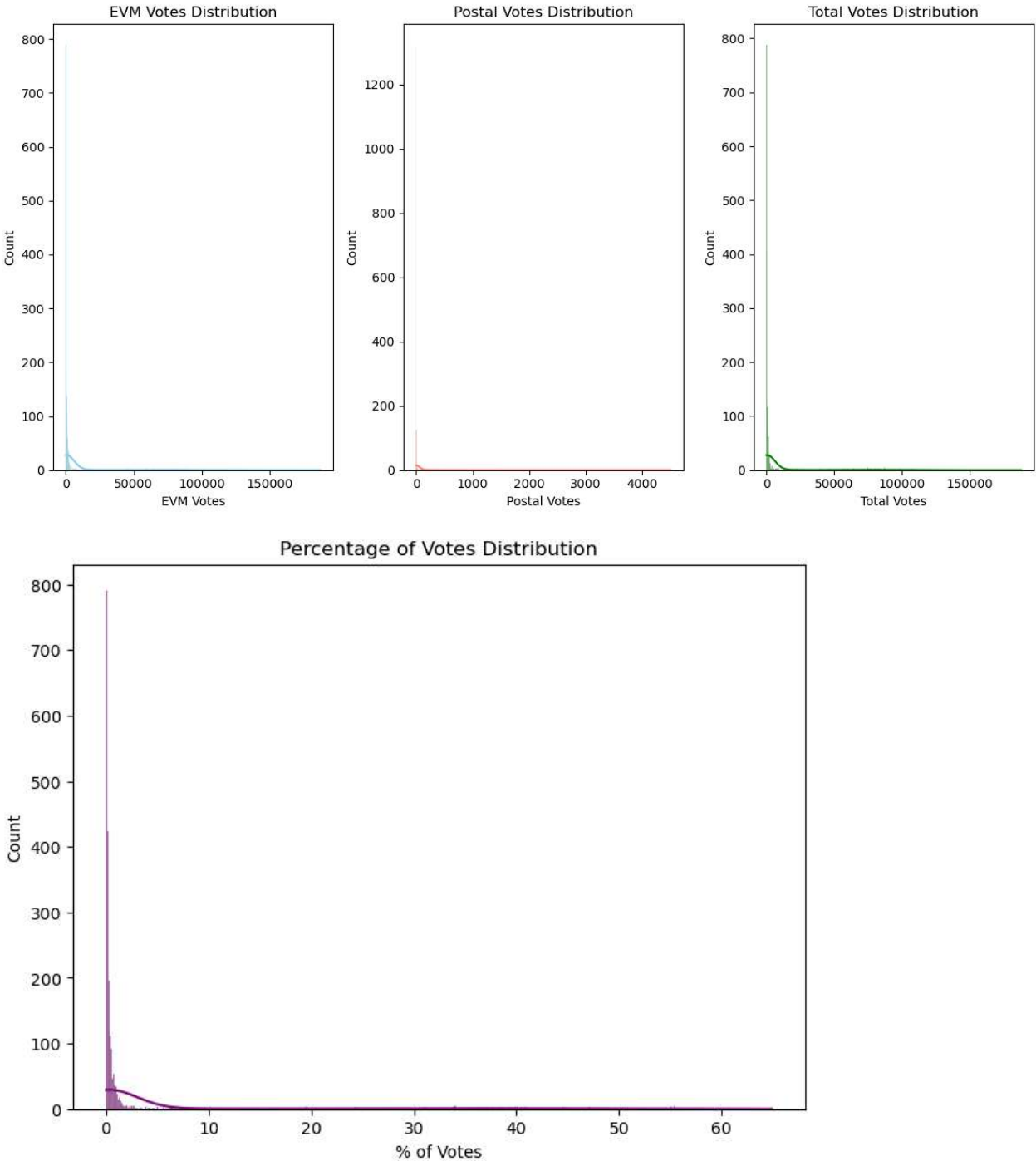
count	2409.000000
mean	4.939797
std	12.762166
min	0.010000
25%	0.080000
50%	0.200000
75%	0.730000
max	64.890000

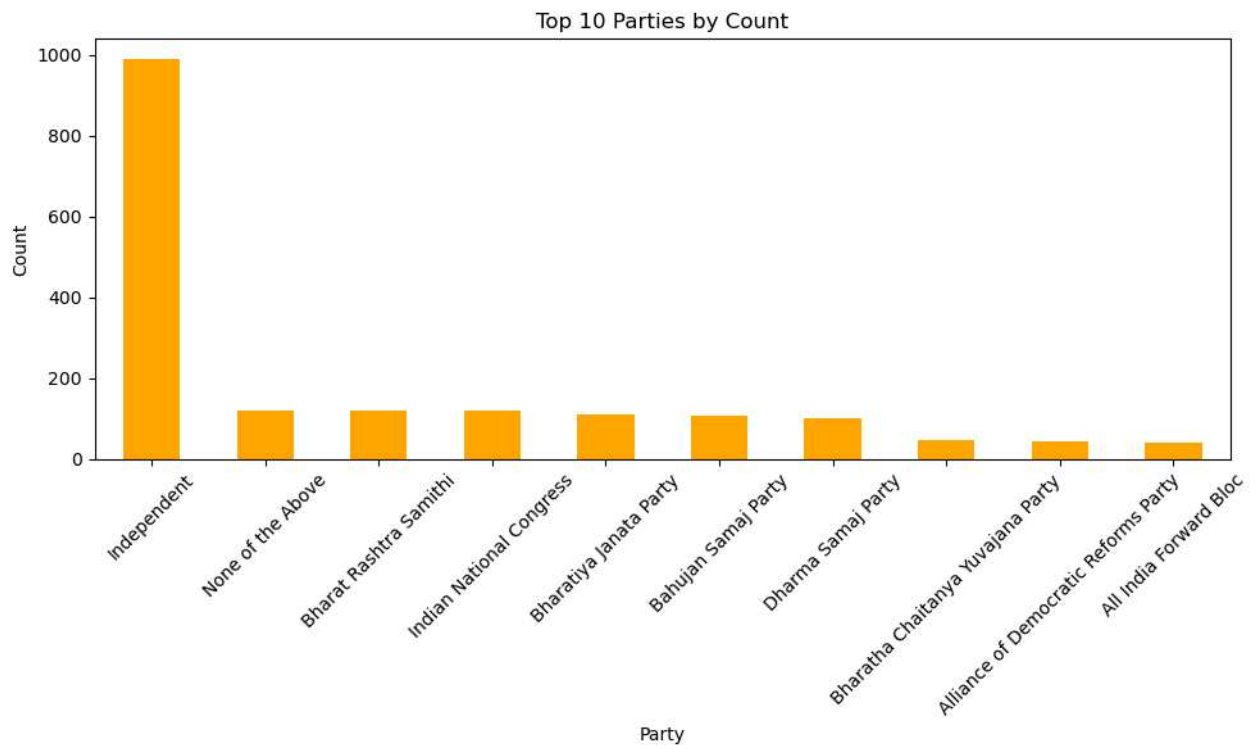
Name: % of Votes, dtype: float64

Count of unique values for object columns:

Candidate	2265
Party	105
Constituency	119

dtype: int64





```
In [42]: # Filter the DataFrame for independent candidates
independent_candidates = ts[ts['Party'] == 'Independent']

# Sort the DataFrame by total votes in descending order and select the top 10 independent candidates
top_10_independent = independent_candidates.nlargest(10, 'Total Votes')

# Display the required information
required_info = top_10_independent[['Candidate', 'Constituency', 'Total Votes', '% of Votes']]
print("Top 10 Independent Candidates with Highest Total Votes:")
print(required_info)
```

Top 10 Independent Candidates with Highest Total Votes:

	Candidate	Constituency	Total Votes	% of Votes
141	KOTNAKA VIJAY KUMAR	5 - Asifabad	16469	8.92
969	Karne Shireesha @Barrelakka	85 - Kollapur	5754	2.99
558	GADDA. SATHISH	32 - Husnabad	5104	2.47
2072	PILLI SAI KUMAR	33 - Siddipet	4970	2.74
1369	MANMOHAN JADHAV	10 - Mudhole	4939	2.44
1738	MADAVAPEDDI VENKAT REDDY	100 - Palakurthi	4146	1.88
1903	SOMARAPU SATYANARAYANA	23 - Ramagundam	4048	2.65
406	Sherla Mahendar	22 - Dharmapuri	3847	2.12
1381	Maddila Venkateshwarlu	109 - Mulug	3709	1.98
191	PUTTA BHASKAR	14 - Banswada	3671	2.29

```
In [45]: # Filter the DataFrame for candidates belonging to Bharat Rashtra Samithi (BRs)
brs_df = ts[ts['Party'] == 'Bharat Rashtra Samithi']

# Calculate the total votes for Bharat Rashtra Samithi (BRs)
total_votes_brs = brs_df['Total Votes'].sum()

# Display the total votes for Bharat Rashtra Samithi (BRs)
print("Total Votes for Bharat Rashtra Samithi (BRs):", total_votes_brs)
```

Total Votes for Bharat Rashtra Samithi (BRs): 8753924

```
In [50]: ts.dtypes
```

```
Out[50]: Unnamed: 0      int64
S.N.          int64
Candidate     object
Party         object
EVM Votes     int64
Postal Votes  int64
Total Votes   int64
% of Votes    float64
Constituency  object
dtype: object
```

```
In [51]: # Calculate the total votes from all parties
total_votes_all_parties = ts['Total Votes'].sum()

# Display the total votes from all parties
print("Total Votes from all parties:", total_votes_all_parties)
```

Total Votes from all parties: 23439854

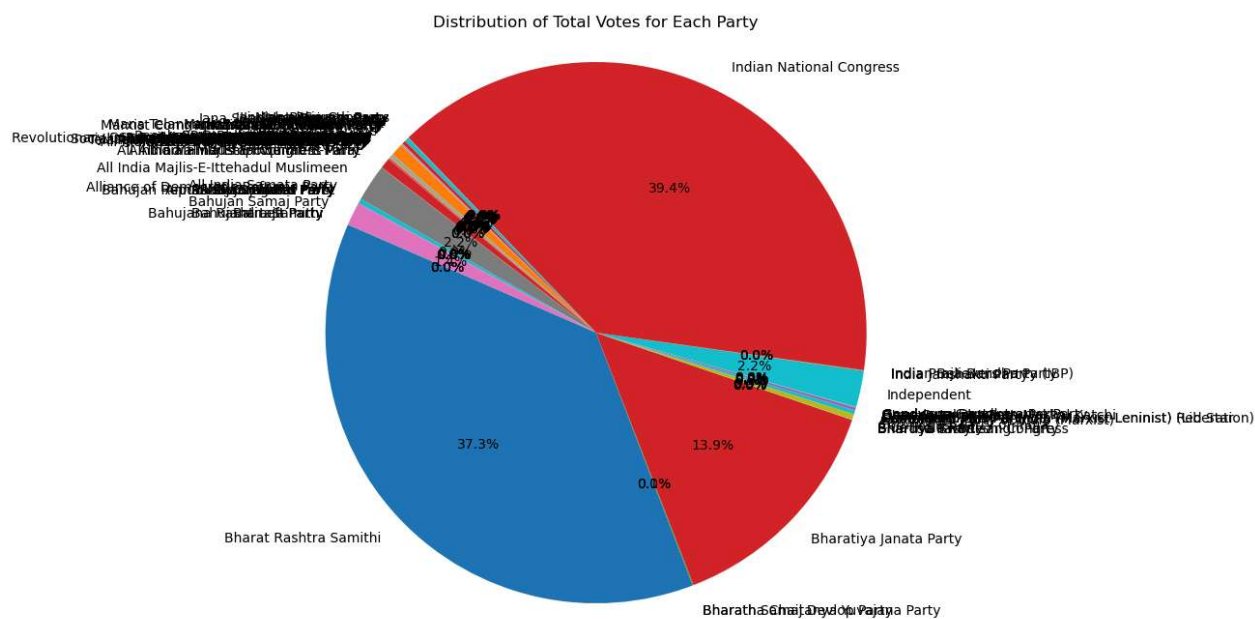
```
In [53]: percentage_votes_brs = (total_votes_brs / total_votes_all_parties) * 100

# Display the percentage
print("BRs vote percentage relative to contested candidates:", percentage_votes_brs)
```

BRs vote percentage relative to contested candidates: 37.34632476806382

```
In [54]: # Group the DataFrame by Party and sum the Total Votes for each party
party_votes = ts.groupby('Party')['Total Votes'].sum()

# Plotting the pie chart
plt.figure(figsize=(10, 8))
plt.pie(party_votes, labels=party_votes.index, autopct='%1.1f%%', startangle=140)
plt.title('Distribution of Total Votes for Each Party')
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.show()
```



```
In [55]: # Filter the DataFrame to include only rows where the candidate is NOTA
nota_df = ts[ts['Candidate'] == 'NOTA']

# Group the filtered DataFrame by constituency and candidate, and sum the votes
nota_votes_by_constituency = nota_df.groupby(['Constituency', 'Candidate'])['Total Votes'].sum()

# Find the constituency where NOTA votes are the highest
constituency_with_highest_nota_votes = nota_votes_by_constituency.idxmax()

# Retrieve the names of candidates and their votes for all candidates in the constituency with highest NOTA votes
candidates_in_constituency = ts[ts['Constituency'] == constituency_with_highest_nota_votes[0]][['Candidate', 'Total '

# Display the results
print("Constituency with the highest NOTA votes:", constituency_with_highest_nota_votes[0])
print("Candidates and their votes in the constituency:")
print(candidates_in_constituency)
```

Constituency with the highest NOTA votes: 45 - Quthbullapur

Candidates and their votes in the constituency:

	Candidate	Total Votes
1839	KUNA SRISAILAM GOUD	102423
1840	MOHAMMED AHMED LAMRA	1759
1841	K.P. VIVEKANAND	187999
1842	KOLAN HANMANTH REDDY	101554
1843	MEKALA KARTHIK YADAV	746
1844	CHOUDHARY GARI SWATIKA	247
1845	THOTA SUVARNA	256
1846	DHONTULA RAMESH MUDIRAJ	137
1847	MOHAMMED WAJID	150
1848	RAVINDAR	199
1849	D. DURGA RAO	299
1850	MUTHYAPAGA SHIVA KUMAR	435
1851	MOHAMMED MAHISAN	279
1852	BAGILI SRINIVAS REDDY	502
1853	SAI KUMAR PANTHULA	603
1854	NOTA	4079

## # PARTY wise contested bar plot

```
In [61]: #Party wise candidates contested
contest = df.groupby('Party')['Candidate'].count().reset_index()
contest = contest.sort_values(by = 'Candidate', ascending = False)

#removing the independent and None of the above

contest = contest.loc[~((contest['Party'] == 'Independent') | (contest['Party'] == 'None of the Above'))]
contest.columns = ['Party', 'Candidates_contested']

top10_contest = contest.head(10)

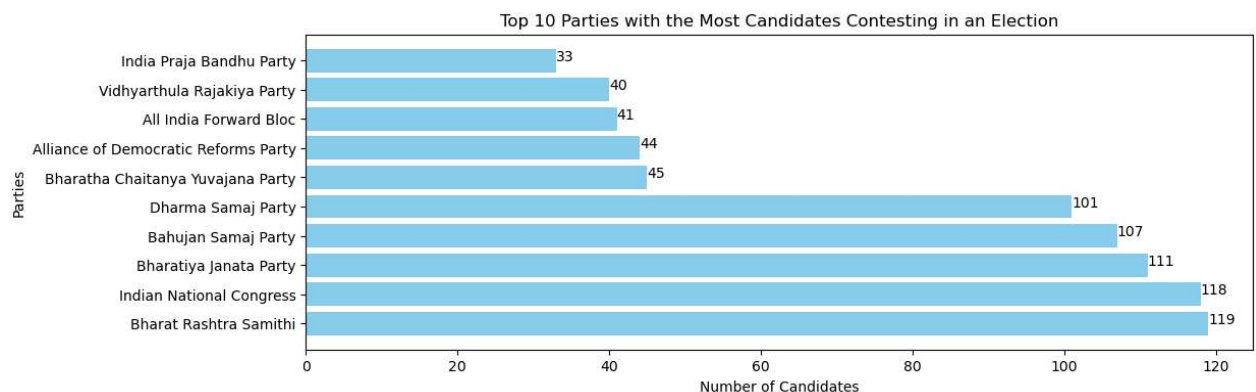
#Plotting the Top10 Parties with most candidates participating in an election

plt.figure(figsize = (12,4))
plt.title("Top 10 Parties with the Most Candidates Contesting in an Election")

# Adding Labels
plt.xlabel('Number of Candidates')
plt.ylabel('Parties')

plt.barh(top10_contest['Party'], top10_contest['Candidates_contested'], color='skyblue')

for index, value in enumerate(top10_contest['Candidates_contested']):
    plt.text(value, index, str(value))
```



## # Cosntituency wise candidates

```
In [62]: #Cosntituency wise candidates

const = df.groupby('Constituency')['Candidate'].count().reset_index()

const = const.sort_values(by = 'Candidate', ascending = False)

const.columns = ['Constituency', 'Candidates contested']

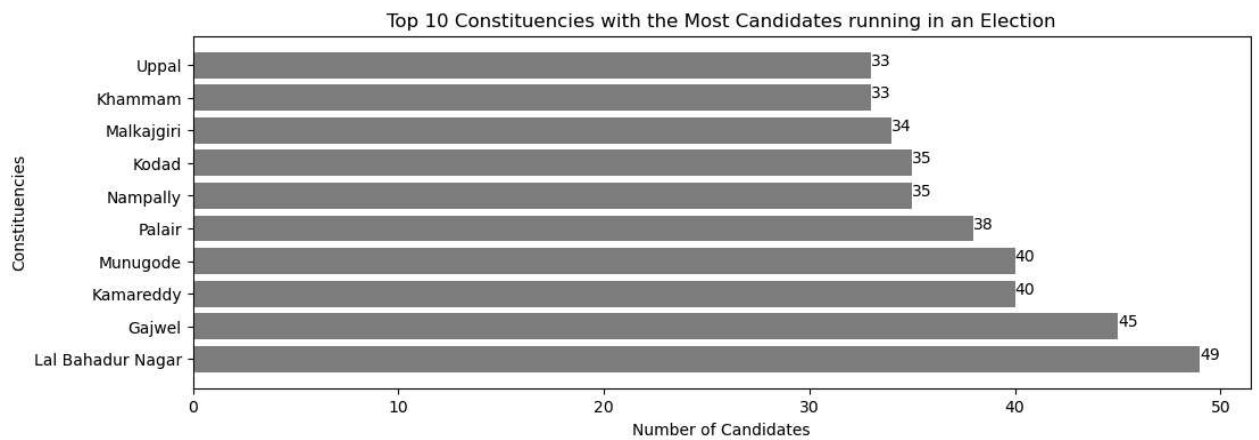
top10_const = const.head(10)

#Plotting the bar for constituency wise candidates
plt.figure(figsize = (12,4))
plt.title("Top 10 Constituencies with the Most Candidates running in an Election")

plt.barh(top10_const['Constituency'], top10_const['Candidates contested'], color='grey')

# Adding Labels
plt.xlabel('Number of Candidates')
plt.ylabel('Constituencies')

for index, value in enumerate(top10_const['Candidates contested']):
    plt.text(value, index, str(value))
```



```
In [65]: #Winning status of each candidate
df['result'] = df.groupby('Constituency')['Total Votes'].transform(max) == df['Total Votes']
df['result'] = df['result'].replace({True: "Win", False: "Loss"})

#Winners
won_candidates = df[df['result'] == 'Win']

#Party wise no of winners
party_won = won_candidates.groupby(['Party']).count().sort_values(by = 'Total Votes', ascending = False).reset_index()

party_won = party_won.loc[:, ['Party', 'Candidate']]

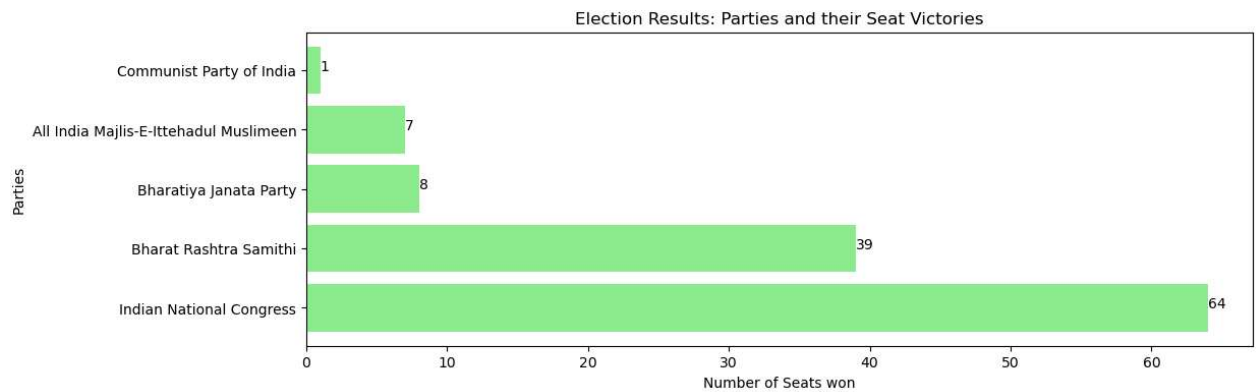
#Renaming the columns
column_mapping2 = {'Party': 'party', 'Candidate': 'candidates_won'}
party_won = party_won.rename(columns = column_mapping2, inplace = False)

#Plotting Parties with highest winning seats
plt.figure(figsize = (12,4))
plt.title("Election Results: Parties and their Seat Victories")

plt.barh(party_won['party'], party_won['candidates_won'], color='lightgreen')

# Adding Labels
plt.xlabel('Number of Seats won')
plt.ylabel('Parties')

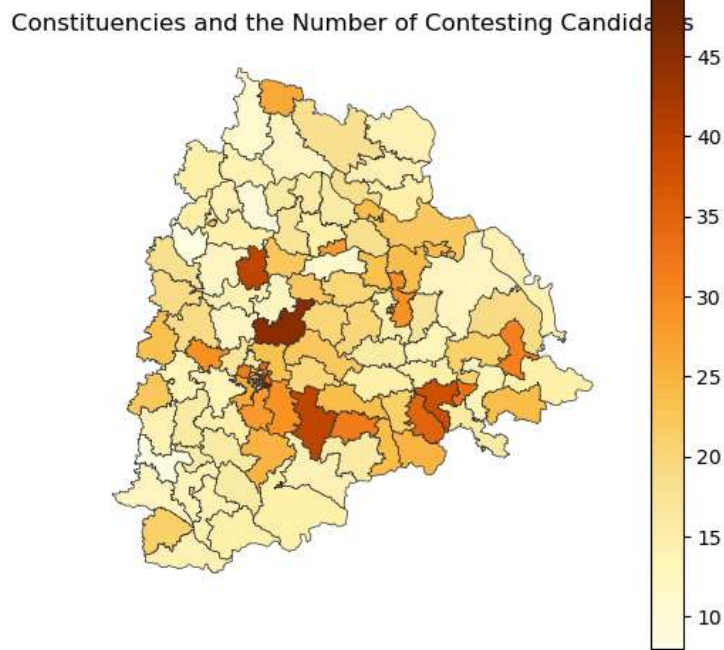
for index, value in enumerate(party_won['candidates_won']):
    plt.text(value, index, str(value))
```



**# ts map analytics using geopandas**

```
In [63]: #Merging the data
map_const = pd.merge(spatial_df,const, on='Constituency')

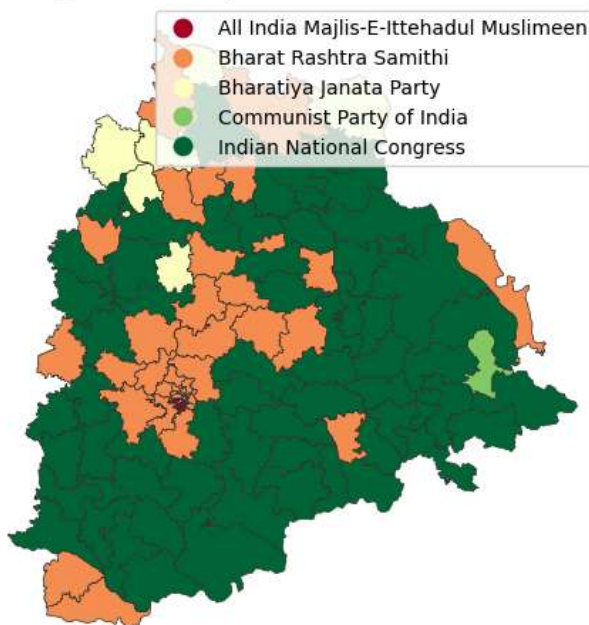
#Map plot
fig, ax = plt.subplots(1, figsize=(6, 6))
plt.title("Constituencies and the Number of Contesting Candidates")
ax.axis('off')
fig = map_const.plot(column='Candidates contested', cmap='YlOrBr', linewidth=0.5, ax=ax, edgecolor='0.2', legend=True)
```



```
In [73]: #Merging the data
map_won = pd.merge(spatial_df,won_candidates, on='Constituency')

#Map plot
fig2, ax = plt.subplots(1, figsize=(6, 6))
plt.title("Winning Political Party in Each Electoral Constituency")
ax.axis('off')
fig2 = map_won.plot(column='Party', cmap='RdYlGn', linewidth=0.5, ax=ax, edgecolor='0.2', legend = True)
```

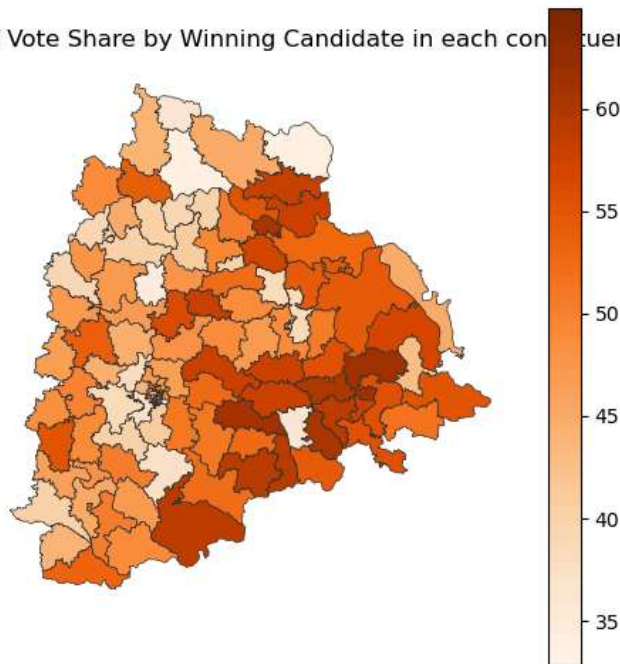
Winning Political Party in Each Electoral Constituency





```
In [74]: #Map plot
fig3, ax = plt.subplots(1, figsize=(6, 6))
plt.title("Percentage of Vote Share by Winning Candidate in each constituency ")
ax.axis('off')
fig3 = map_won.plot(column='% of Votes', cmap='Oranges', linewidth=0.5, ax=ax, edgecolor='0.2', legend = True)
```

Percentage of Vote Share by Winning Candidate in each constituency



## # voing in the capital city

```
In [75]: df_city = won_candidates.copy()
df_city['Constituency_no'] = df_city['Constituency_no'].astype(int)

#Constituency numbers within Hyderabad from 57-71. Lets filter the Hyderabad data
Capital_city = df_city[(df_city['Constituency_no'] >= 57) & (df_city['Constituency_no'] <= 71)]

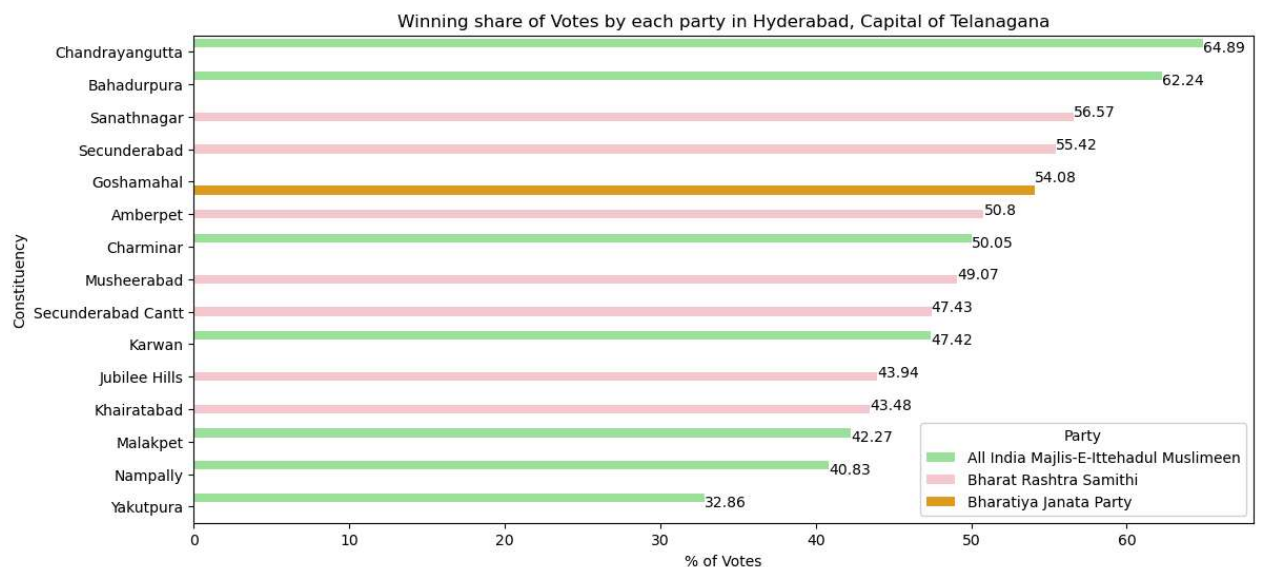
Capital_city = Capital_city.sort_values(by = '% of Votes', ascending = False)

plt.figure(figsize = (13, 6))

# Create a dictionary to map parties to specific colors
party_colors = {'Bharat Rashtra Samithi': 'pink', 'All India Majlis-E-Ittehadul Muslimeen': 'lightgreen', 'Bharatiya

plt.title('Winning share of Votes by each party in Hyderabad, Capital of Telanagana')
sns.barplot(x='% of Votes', y='Constituency', hue='Party', data=Capital_city, dodge=True, palette=party_colors)

for index, value in enumerate(Capital_city['% of Votes']):
    plt.text(value, index, str(value))
```



```
In [76]: #Merging the data
map_capital = pd.merge(spatial_df,Capital_city, on='Constituency')

map_capital['geometry'] = map_capital['geometry'].to_crs(epsg=3395)
#Map plot
fig3, ax = plt.subplots(1, figsize=(6, 6))
plt.title("Election Result in Hyderabad")
ax.axis('off')
fig3 = map_capital.plot(column='Party', cmap='RdYlGn', linewidth=0.5, ax=ax, edgecolor='0.2', legend = True)

for x, y, label in zip(map_capital.geometry.centroid.x, map_capital.geometry.centroid.y, map_capital['Constituency']):
    ax.text(x, y, label, fontsize=8, ha='center', va='center', color='black')
```

