

In [1]: *###Electric Vehicle Charging Stations in Canada: Data Analysis and Visualization*
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In []: *##Data Source:*
#The dataset used in this analysis was obtained from Natural Resources Canada’s
\$(NRCan) open data portal: Electric Charging and Alternative Fuelling Stations Locator.

In [3]: *## Objective*
#This project analyzes electric vehicle (EV) charging station data in Canada using the dataset provided by Natural Resources Canada (NRCan).

Key Findings:
#- Growth Trends: A significant rise in EV charging stations since 2019, reflecting increased adoption of electric vehicles.
#- Regional Insights: Provinces like Quebec, Ontario, and British Columbia lead in both Level 2 and DC fast charging infrastructure.
#- Connector Types: The J1772 connector dominates the charging network, followed by CHAdeMO and CCS connectors.
#- Top Charging Networks: Flo, Tesla, and Circuit électrique dominate the Canadian EV market.

Impact:
#This analysis highlights Canada’s readiness to support the growing EV market, with strong investments in both standard and fast charging infrastructure.
#It also identifies gaps in charging accessibility in smaller provinces, providing actionable insights for future network expansion.

In [139...
Import essential Libraries
import pandas **as** pd
import matplotlib.pyplot **as** plt
import seaborn **as** sns
import numpy **as** np

Load the NRCan dataset
data_file = **"data.csv"** *# Replace with your CSV file's path*
df = pd.read_csv(**"data.csv"**,low_memory=**False**)

Check the dataset information
print(df.info())

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 31562 entries, 0 to 31561
Data columns (total 83 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   Fuel Type Code                           31562 non-null  object
1   Station Name                             31562 non-null  object
2   Street Address                           31562 non-null  object
3   Intersection Directions                   3415 non-null   object
4   City                                     31562 non-null  object
5   State                                   31562 non-null  object
6   ZIP                                     31562 non-null  object
7   Plus4                                   0 non-null      float64
8   Station Phone                           31185 non-null  object
9   Status Code                             31562 non-null  object
10  Expected Date                           0 non-null      float64
11  Groups With Access Code                  31562 non-null  object
12  Access Days Time                         28355 non-null  object
13  Cards Accepted                           3439 non-null   object
14  BD Blends                               0 non-null      float64
15  NG Fill Type Code                       0 non-null      float64
16  NG PSI                                  0 non-null      float64
17  EV Level1 EVSE Num                      56 non-null     float64
18  EV Level2 EVSE Num                      26060 non-null  float64
19  EV DC Fast Count                        6553 non-null   float64
20  EV Other Info                           0 non-null      float64
21  EV Network                             31562 non-null  object
22  EV Network Web                          29550 non-null  object
23  Geocode Status                          31561 non-null  object
24  Latitude                                31562 non-null  float64
25  Longitude                                31562 non-null  float64
26  Date Last Confirmed                     31444 non-null  object
27  ID                                       31562 non-null  int64
28  Updated At                             31562 non-null  object
29  Owner Type Code                         8709 non-null   object
30  Federal Agency ID                       4 non-null      float64
31  Federal Agency Name                     4 non-null      object
32  Open Date                              31498 non-null  object
33  Hydrogen Status Link                    0 non-null      float64
34  NG Vehicle Class                        0 non-null      float64
35  LPG Primary                             0 non-null      float64
36  E85 Blender Pump                        0 non-null      float64
37  EV Connector Types                      31562 non-null  object
38  Country                                 31562 non-null  object
39  Intersection Directions (French)         263 non-null   object
40  Access Days Time (French)               28027 non-null  object
41  BD Blends (French)                      0 non-null      float64
42  Groups With Access Code (French)        31562 non-null  object
43  Hydrogen Is Retail                      0 non-null      float64
44  Access Code                             31562 non-null  object
45  Access Detail Code                      1165 non-null   object
46  Federal Agency Code                     4 non-null      object
47  Facility Type                           6682 non-null   object
48  CNG Dispenser Num                       0 non-null      float64
49  CNG On-Site Renewable Source             0 non-null      float64
50  CNG Total Compression Capacity           0 non-null      float64
51  CNG Storage Capacity                     0 non-null      float64
52  LNG On-Site Renewable Source             0 non-null      float64
53  E85 Other Ethanol Blends                 0 non-null      float64
54  EV Pricing                              5552 non-null   object
55  EV Pricing (French)                     5493 non-null   object
56  LPG Nozzle Types                        0 non-null      float64
57  Hydrogen Pressures                       0 non-null      float64
58  Hydrogen Standards                       0 non-null      float64
59  CNG Fill Type Code                       0 non-null      float64
60  CNG PSI                                  0 non-null      float64
61  CNG Vehicle Class                       0 non-null      float64
62  LNG Vehicle Class                       0 non-null      float64
63  EV On-Site Renewable Source              58 non-null     object
64  Restricted Access                        5568 non-null   object
65  RD Blends                               0 non-null      float64
66  RD Blends (French)                      0 non-null      float64
67  RD Blended with Biodiesel               0 non-null      float64
68  RD Maximum Biodiesel Level              0 non-null      float64
69  NPS Unit Name                           0 non-null      float64
70  CNG Station Sells Renewable Natural Gas  0 non-null      float64
71  LNG Station Sells Renewable Natural Gas  0 non-null      float64
72  Maximum Vehicle Class                    4070 non-null   object
73  EV Workplace Charging                    31562 non-null  bool
74  Funding Sources                          0 non-null      float64
75  EV J1772 Connector Count                 31562 non-null  int64
76  EV J1772 Power Output (kW)              20986 non-null  float64
77  EV CCS Connector Count                   31562 non-null  int64
78  EV CCS Power Output (kW)                 3543 non-null   float64
79  EV CHAdeMO Connector Count               31562 non-null  int64
80  EV CHAdeMO Power Output (kW)            2635 non-null   float64
81  EV J3400 Connector Count                 31562 non-null  int64
82  EV J3400 Power Output (kW)              363 non-null    float64
dtypes: bool(1), float64(43), int64(5), object(34)
memory usage: 19.8+ MB
None
```

```
In [141... # Step 1: Identifying Null Values in the Dataset
# This step helps us find missing data in each column to decide on further cleaning actions
null_values = df.isnull().sum() # Count of null values in each column
print(null_values)
```

Fuel Type Code	0
Station Name	0
Street Address	0
Intersection Directions	28147
City	0
State	0
ZIP	0
Plus4	31562
Station Phone	377
Status Code	0
Expected Date	31562
Groups With Access Code	0
Access Days Time	3207
Cards Accepted	28123
BD Blends	31562
NG Fill Type Code	31562
NG PSI	31562
EV Level1 EVSE Num	31506
EV Level2 EVSE Num	5502
EV DC Fast Count	25009
EV Other Info	31562
EV Network	0
EV Network Web	2012
Geocode Status	1
Latitude	0
Longitude	0
Date Last Confirmed	118
ID	0
Updated At	0
Owner Type Code	22853
Federal Agency ID	31558
Federal Agency Name	31558
Open Date	64
Hydrogen Status Link	31562
NG Vehicle Class	31562
LPG Primary	31562
E85 Blender Pump	31562
EV Connector Types	0
Country	0
Intersection Directions (French)	31299
Access Days Time (French)	3535
BD Blends (French)	31562
Groups With Access Code (French)	0
Hydrogen Is Retail	31562
Access Code	0
Access Detail Code	30397
Federal Agency Code	31558
Facility Type	24880
CNG Dispenser Num	31562
CNG On-Site Renewable Source	31562
CNG Total Compression Capacity	31562
CNG Storage Capacity	31562
LNG On-Site Renewable Source	31562
E85 Other Ethanol Blends	31562
EV Pricing	26010
EV Pricing (French)	26069
LPG Nozzle Types	31562
Hydrogen Pressures	31562
Hydrogen Standards	31562
CNG Fill Type Code	31562
CNG PSI	31562
CNG Vehicle Class	31562
LNG Vehicle Class	31562
EV On-Site Renewable Source	31504
Restricted Access	25994
RD Blends	31562
RD Blends (French)	31562
RD Blended with Biodiesel	31562
RD Maximum Biodiesel Level	31562
NPS Unit Name	31562
CNG Station Sells Renewable Natural Gas	31562
LNG Station Sells Renewable Natural Gas	31562
Maximum Vehicle Class	27492
EV Workplace Charging	0
Funding Sources	31562
EV J1772 Connector Count	0
EV J1772 Power Output (kW)	10576
EV CCS Connector Count	0
EV CCS Power Output (kW)	28019
EV CHAdeMO Connector Count	0
EV CHAdeMO Power Output (kW)	28927
EV J3400 Connector Count	0
EV J3400 Power Output (kW)	31199

dtype: int64

In [143...
pd.set_option('display.max_rows', None)

In [145...
Calculate the percentage of missing data for better understanding
total_rows = len(df) # Total number of rows in the dataset
null_percentage = (null_values / total_rows) * 100 # Percentage of missing data
print(null_percentage)

Fuel Type Code 0.000000
Station Name 0.000000
Street Address 0.000000
Intersection Directions 89.180027
City 0.000000
State 0.000000
ZIP 0.000000
Plus4 100.000000
Station Phone 1.194474
Status Code 0.000000
Expected Date 100.000000
Groups With Access Code 0.000000
Access Days Time 10.160953
Cards Accepted 89.103986
BD Blends 100.000000
NG Fill Type Code 100.000000
NG PSI 100.000000
EV Level1 EVSE Num 99.822571
EV Level2 EVSE Num 17.432355
EV DC Fast Count 79.237691
EV Other Info 100.000000
EV Network 0.000000
EV Network Web 6.374754
Geocode Status 0.003168
Latitude 0.000000
Longitude 0.000000
Date Last Confirmed 0.373867
ID 0.000000
Updated At 0.000000
Owner Type Code 72.406692
Federal Agency ID 99.987327
Federal Agency Name 99.987327
Open Date 0.202775
Hydrogen Status Link 100.000000
NG Vehicle Class 100.000000
LPG Primary 100.000000
E85 Blender Pump 100.000000
EV Connector Types 0.000000
Country 0.000000
Intersection Directions (French) 99.166719
Access Days Time (French) 11.200177
BD Blends (French) 100.000000
Groups With Access Code (French) 0.000000
Hydrogen Is Retail 100.000000
Access Code 0.000000
Access Detail Code 96.308852
Federal Agency Code 99.987327
Facility Type 78.828972
CNG Dispenser Num 100.000000
CNG On-Site Renewable Source 100.000000
CNG Total Compression Capacity 100.000000
CNG Storage Capacity 100.000000
LNG On-Site Renewable Source 100.000000
E85 Other Ethanol Blends 100.000000
EV Pricing 82.409226
EV Pricing (French) 82.596160
LPG Nozzle Types 100.000000
Hydrogen Pressures 100.000000
Hydrogen Standards 100.000000
CNG Fill Type Code 100.000000
CNG PSI 100.000000
CNG Vehicle Class 100.000000
LNG Vehicle Class 100.000000
EV On-Site Renewable Source 99.816235
Restricted Access 82.358532
RD Blends 100.000000
RD Blends (French) 100.000000
RD Blended with Biodiesel 100.000000
RD Maximum Biodiesel Level 100.000000
NPS Unit Name 100.000000
CNG Station Sells Renewable Natural Gas 100.000000
LNG Station Sells Renewable Natural Gas 100.000000
Maximum Vehicle Class 87.104746
EV Workplace Charging 0.000000
Funding Sources 100.000000
EV J1772 Connector Count 0.000000
EV J1772 Power Output (kW) 33.508650
EV CCS Connector Count 0.000000
EV CCS Power Output (kW) 88.774476
EV CHAdeMO Connector Count 0.000000
EV CHAdeMO Power Output (kW) 91.651353
EV J3400 Connector Count 0.000000
EV J3400 Power Output (kW) 98.849883
dtype: float64

```
In [147... # Remove columns with more than 80% missing data (example threshold)
columns_to_drop = null_percentage[null_percentage > 80].index # Identify columns to drop
df = df.drop(columns=columns_to_drop, axis=1) # Drop the identified columns
print(f"Columns dropped: {list(columns_to_drop)}")
```

Columns dropped: ['Intersection Directions', 'Plus4', 'Expected Date', 'Cards Accepted', 'BD Blends', 'NG Fill Type Code', 'NG PSI', 'EV Level1 EVSE Num', 'E V Other Info', 'Federal Agency ID', 'Federal Agency Name', 'Hydrogen Status Link', 'NG Vehicle Class', 'LPG Primary', 'E85 Blender Pump', 'Intersection Direc tions (French)', 'BD Blends (French)', 'Hydrogen Is Retail', 'Access Detail Code', 'Federal Agency Code', 'CNG Dispenser Num', 'CNG On-Site Renewable Sourc e', 'CNG Total Compression Capacity', 'CNG Storage Capacity', 'LNG On-Site Renewable Source', 'E85 Other Ethanol Blends', 'EV Pricing', 'EV Pricing (Frenc h)', 'LPG Nozzle Types', 'Hydrogen Pressures', 'Hydrogen Standards', 'CNG Fill Type Code', 'CNG PSI', 'CNG Vehicle Class', 'LNG Vehicle Class', 'EV On-Site R enewable Source', 'Restricted Access', 'RD Blends', 'RD Blends (French)', 'RD Blended with Biodiesel', 'RD Maximum Biodiesel Level', 'NPS Unit Name', 'CNG St ation Sells Renewable Natural Gas', 'LNG Station Sells Renewable Natural Gas', 'Maximum Vehicle Class', 'Funding Sources', 'EV CCS Power Output (kW)', 'EV CH AdeMO Power Output (kW)', 'EV J3400 Power Output (kW)']

```
In [149... df.info() #checking the info after dropping key columns containing null values >80%
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 31562 entries, 0 to 31561
Data columns (total 34 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Fuel Type Code                        31562 non-null  object
1   Station Name                          31562 non-null  object
2   Street Address                        31562 non-null  object
3   City                                  31562 non-null  object
4   State                                 31562 non-null  object
5   ZIP                                   31562 non-null  object
6   Station Phone                         31185 non-null  object
7   Status Code                           31562 non-null  object
8   Groups With Access Code               31562 non-null  object
9   Access Days Time                      28355 non-null  object
10  EV Level2 EVSE Num                   26060 non-null  float64
11  EV DC Fast Count                     6553 non-null   float64
12  EV Network                           31562 non-null  object
13  EV Network Web                       29550 non-null  object
14  Geocode Status                       31561 non-null  object
15  Latitude                             31562 non-null  float64
16  Longitude                             31562 non-null  float64
17  Date Last Confirmed                  31444 non-null  object
18  ID                                    31562 non-null  int64
19  Updated At                           31562 non-null  object
20  Owner Type Code                       8709 non-null   object
21  Open Date                            31498 non-null  object
22  EV Connector Types                   31562 non-null  object
23  Country                              31562 non-null  object
24  Access Days Time (French)            28027 non-null  object
25  Groups With Access Code (French)     31562 non-null  object
26  Access Code                           31562 non-null  object
27  Facility Type                         6682 non-null   object
28  EV Workplace Charging                 31562 non-null  bool
29  EV J1772 Connector Count              31562 non-null  int64
30  EV J1772 Power Output (kW)            20986 non-null  float64
31  EV CCS Connector Count                31562 non-null  int64
32  EV CHAdeMO Connector Count            31562 non-null  int64
33  EV J3400 Connector Count              31562 non-null  int64
dtypes: bool(1), float64(5), int64(5), object(23)
memory usage: 8.0+ MB
```

In [151... df.head() #getting top few rows in th dataset

Out[151...

	Fuel Type Code	Station Name	Street Address	City	State	ZIP	Station Phone	Status Code	Groups With Access Code	Access Days Time	...	Access Days Time (French)	Groups With Access Code (French)	Access Code	Facility Type	EV Workplace Charging	EV J1772 Connector Count	EV J1772 Power Output (kW)	Co
0	ELEC	Ramada	1319 2nd St W	Brooks	AB	T1R 1P7	403-362-6440	E	Public - Call ahead	24 hours daily; see front desk for access	...	24 heures par jour; voire le concierge pour accès	Public - Appeler à l'avance	public	HOTEL	False	1	NaN	
1	ELEC	Davis Chevrolet	149 E Lake Crescent NE	Airdrie	AB	T4A 2H9	403-948-6909	E	Public	24 hours daily; for client use only	...	24 heures par jour; à l'usage des clients seul...	Public	public	CAR_DEALER	False	1	NaN	
2	ELEC	Gasonic Instruments	8-823 41st Ave NE	Calgary	AB	T2E 6Y3	403-276-2201	E	Public - Limited hours	8am-5pm M-F; remaining hours and all weekends	...	8:00-17:00 LUN-VEN; disponible au public le re...	Public - Heures limitées	public	OFFICE_BLDG	True	1	NaN	
3	ELEC	International Motor Cars	7220 Railway St SE	Calgary	AB	T2H 3A8	844-288-8918	E	Public - Call ahead	24 hours daily	...	Accessible 24 heures par jour	Public - Appeler à l'avance	public	CAR_DEALER	False	1	NaN	
4	ELEC	Residence Inn	3710 Market St SE	Calgary	AB	T3M 2P2	587-349-8633	E	Public	24 hours daily; for guest use only; parking fe...	...	24 heures par jour; des clients seulement; fra...	Public	public	HOTEL	False	1	NaN	

5 rows × 34 columns

In [153...

```
# Dropping irrelevant columns that are not needed for the analysis
columns_to_remove = [
    'Station Phone',          # Contact details, not relevant for analysis
    'Groups With Access Code', # Access code information, likely not useful
    'EV Network Web',         # Website Links, not needed
    'Geocode Status',         # Geocoding status, unnecessary for the objectives
    'ID',                      # Unique ID, no analytical value
    'Access Days Time (French)', # French duplicate of 'Access Days Time'
    'Groups With Access Code (French)', # French duplicate of access-related column
]

# Drop the columns from the dataset
df = df.drop(columns=columns_to_remove, axis=1)

# Check the updated DataFrame
print(f"Remaining columns: {df.columns}")
```

Remaining columns: Index(['Fuel Type Code', 'Station Name', 'Street Address', 'City', 'State', 'ZIP', 'Status Code', 'Access Days Time', 'EV Level2 EVSE Num', 'EV DC Fast Count', 'EV Network', 'Latitude', 'Longitude', 'Date Last Confirmed', 'Updated At', 'Owner Type Code', 'Open Date', 'EV Connector Types', 'Country', 'Access Code', 'Facility Type', 'EV Workplace Charging', 'EV J1772 Connector Count', 'EV J1772 Power Output (kW)', 'EV CCS Connector Count', 'EV CHAdEMO Connector Count', 'EV J3400 Connector Count'], dtype='object')

```
In [155... # Extracting the year from 'Open Date'
df['Year']=pd.to_datetime(df['Open Date'], errors='coerce').dt.year.astype('Int64')
```

```
In [157... # Sort the DataFrame by the 'Year' column in ascending order
df_sorted = df.sort_values(by='Year', ascending=True)

# Display the first few rows to confirm sorting
df_sorted.head()
```

Out[157...

	Fuel Type Code	Station Name	Street Address	City	State	ZIP	Status Code	Access Days Time	EV Level2 EVSE Num	EV DC Fast Count	...	Country	Access Code	Facility Type	EV Workplace Charging	EV J1772 Connector Count	EV J1772 Power Output (kW)	EV CCS Connector Count	CH Co
190	ELEC	Paul Sadlon Motors Inc	550 Bayfield St	Barrie	ON	L4M 5A2	E	Dealership business hours	3.0	NaN	...	CA	public	CAR DEALER	False	1	NaN	0	
191	ELEC	Paul Sadlon Motors Inc	550 Bayfield St	Barrie	ON	L4M 5A2	E	Dealership business hours	3.0	NaN	...	CA	public	CAR DEALER	False	1	NaN	0	
192	ELEC	Paul Sadlon Motors Inc	550 Bayfield St	Barrie	ON	L4M 5A2	E	Dealership business hours	3.0	NaN	...	CA	public	CAR DEALER	False	1	NaN	0	
649	ELEC	Gabriel Ford Lincoln	7100 rue Saint-Jacques	Montreal	QC	H4B 1V2	E	Dealership business hours; For client use only	1.0	NaN	...	CA	public	CAR DEALER	False	1	NaN	0	
403	ELEC	Chevrolet GMC Roberval	321 boul. Marcotte	Roberval	QC	G8H 1Z4	E	24 hours daily; for client use only	1.0	1.0	...	CA	public	CAR DEALER	False	0	NaN	1	

5 rows × 28 columns

```
In [159... # Additional columns to drop
more_columns_to_remove = [
    'Facility Type',          # May not directly affect trends or visuals
    'EV J1772 Power Output (kW)', # Technical details about power output
    'Updated At',             # Metadata, not relevant for analysis
]

# Drop these columns from the DataFrame
df = df.drop(columns=more_columns_to_remove, axis=1)

# Check the remaining columns to confirm
print(f"Remaining columns: {df.columns}")
```

Remaining columns: Index(['Fuel Type Code', 'Station Name', 'Street Address', 'City', 'State', 'ZIP', 'Status Code', 'Access Days Time', 'EV Level2 EVSE Num', 'EV DC Fast Count', 'EV Network', 'Latitude', 'Longitude', 'Date Last Confirmed', 'Owner Type Code', 'Open Date', 'EV Connector Types', 'Country', 'Access Code', 'EV Workplace Charging', 'EV J1772 Connector Count', 'EV CCS Connector Count', 'EV CHAdEMO Connector Count', 'EV J3400 Connector Count', 'Year'], dtype='object')

```
In [161... # Check the unique values in the Access Code column
print(df['Access Code'].unique())

['public']
```

```
In [163... df = df.drop(columns=['Access Code']) #as it contains only public unique value
```

```
In [165... df.isnull().sum()/len(df)*100 #checking the null percentage again
```

Out[165...
Fuel Type Code 0.000000
Station Name 0.000000
Street Address 0.000000
City 0.000000
State 0.000000
ZIP 0.000000
Status Code 0.000000
Access Days Time 10.160953
EV Level2 EVSE Num 17.432355
EV DC Fast Count 79.237691
EV Network 0.000000
Latitude 0.000000
Longitude 0.000000
Date Last Confirmed 0.373867
Owner Type Code 72.406692
Open Date 0.202775
EV Connector Types 0.000000
Country 0.000000
EV Workplace Charging 0.000000
EV J1772 Connector Count 0.000000
EV CCS Connector Count 0.000000
EV CHAdeMO Connector Count 0.000000
EV J3400 Connector Count 0.000000
Year 0.202775
dtype: float64

In [167...
Fill categorical columns with 'Unknown'
df['Access Days Time'] = df['Access Days Time'].fillna('Unknown')
df['Owner Type Code'] = df['Owner Type Code'].fillna('Unknown')

In [169...
Fill numerical columns with median
- Median is chosen because it is robust to outliers and represents the central tendency
df['EV Level2 EVSE Num'] = df['EV Level2 EVSE Num'].fillna(df['EV Level2 EVSE Num'].median())

Fill missing values in 'EV DC Fast Count' with 0
- 0 is used because missing values likely mean the absence of DC fast chargers
df['EV DC Fast Count'] = df['EV DC Fast Count'].fillna(0)

Fill missing values in 'Open Date' with the placeholder 'Unknown'
- 'Unknown' is used as a string placeholder because:
1. Open Date might not always be available in the dataset.
2. This avoids dropping rows while clearly indicating missing values.
df['Open Date'] = df['Open Date'].fillna('Unknown')

Fill missing values in 'Date Last Confirmed' with the most frequent value (mode)
- Mode is chosen because:
1. It's the most common value and a reasonable assumption for missing data.
2. It prevents introducing misleading or biased information into the dataset.
df['Date Last Confirmed'] = df['Date Last Confirmed'].fillna(df['Date Last Confirmed'].mode()[0])

Check for remaining null values (to verify)
print(df.isnull().sum())

Fuel Type Code 0
Station Name 0
Street Address 0
City 0
State 0
ZIP 0
Status Code 0
Access Days Time 0
EV Level2 EVSE Num 0
EV DC Fast Count 0
EV Network 0
Latitude 0
Longitude 0
Date Last Confirmed 0
Owner Type Code 0
Open Date 0
EV Connector Types 0
Country 0
EV Workplace Charging 0
EV J1772 Connector Count 0
EV CCS Connector Count 0
EV CHAdeMO Connector Count 0
EV J3400 Connector Count 0
Year 64
dtype: int64

In [171...
Drop rows where 'Year' is NaN
- The 'Year' column is critical for time-series analysis.
- Rows with missing 'Year' values (~0.2% of the dataset) were dropped
to ensure the accuracy and clarity of trends over time.
df = df.dropna(subset=['Year'])

print(df.isnull().sum())

Fuel Type Code 0
Station Name 0
Street Address 0
City 0
State 0
ZIP 0
Status Code 0
Access Days Time 0
EV Level2 EVSE Num 0
EV DC Fast Count 0
EV Network 0
Latitude 0
Longitude 0
Date Last Confirmed 0
Owner Type Code 0
Open Date 0
EV Connector Types 0
Country 0
EV Workplace Charging 0
EV J1772 Connector Count 0
EV CCS Connector Count 0
EV CHAdeMO Connector Count 0
EV J3400 Connector Count 0
Year 0
dtype: int64

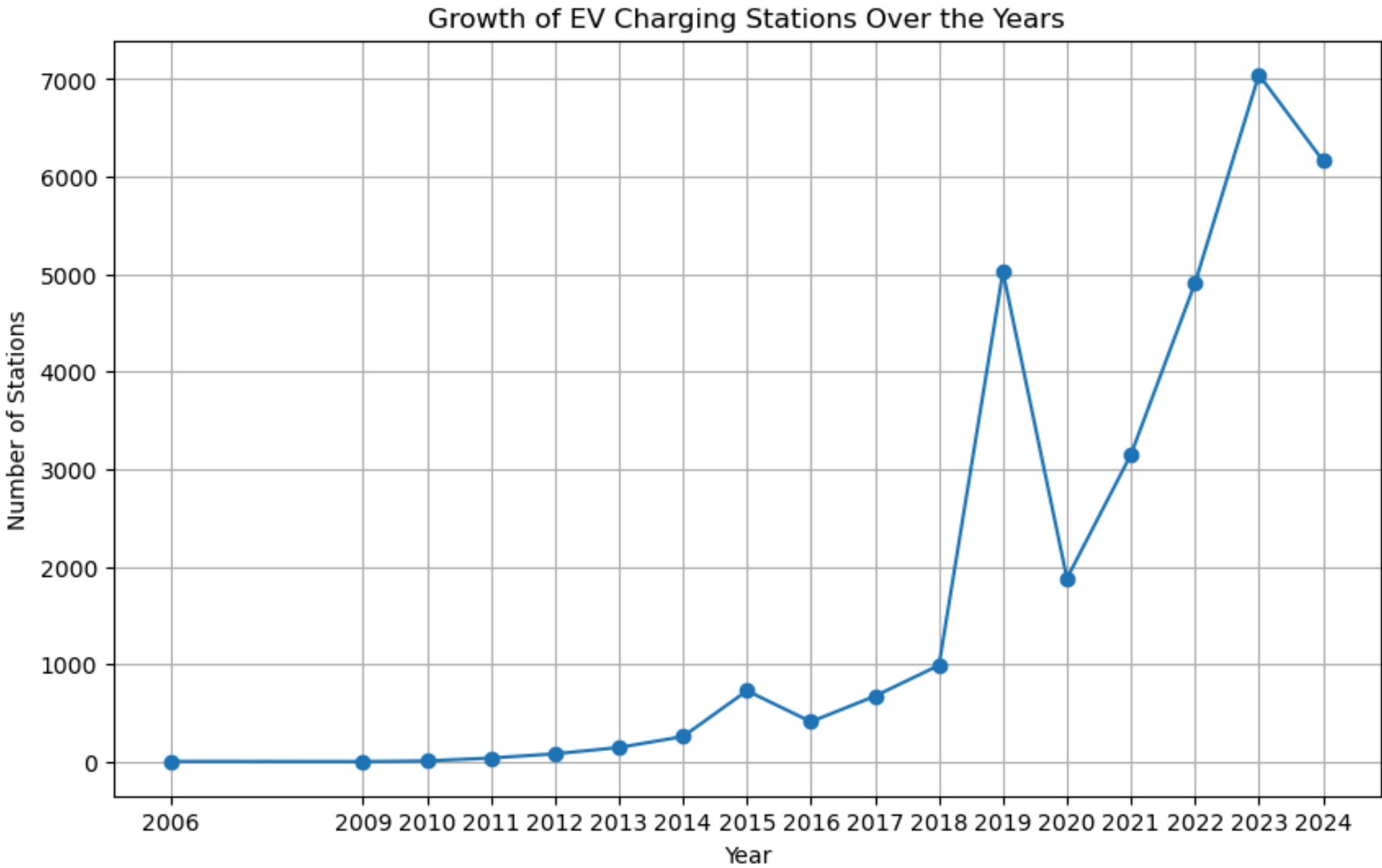
In []:

In []:

In [69]: *#Data Exploration & visulization*

In [173... *# Group data by Year and count the number of stations*
Using .size() here instead of .count() because:
- .size() counts all rows, regardless of column values
- We've already handled missing values by replacing NaN with 'Unknown',
so .size() and .count() would give the same result in this case.
stations_by_year = df.groupby('Year').size()

In [175... *# Plot year-by-year trends*
plt.figure(figsize=(10, 6))
stations_by_year.plot(kind='line', marker='o', title='Growth of EV Charging Stations Over the Years')
plt.xlabel('Year')
plt.ylabel('Number of Stations')
plt.xticks(stations_by_year.index)
plt.grid()
plt.show()

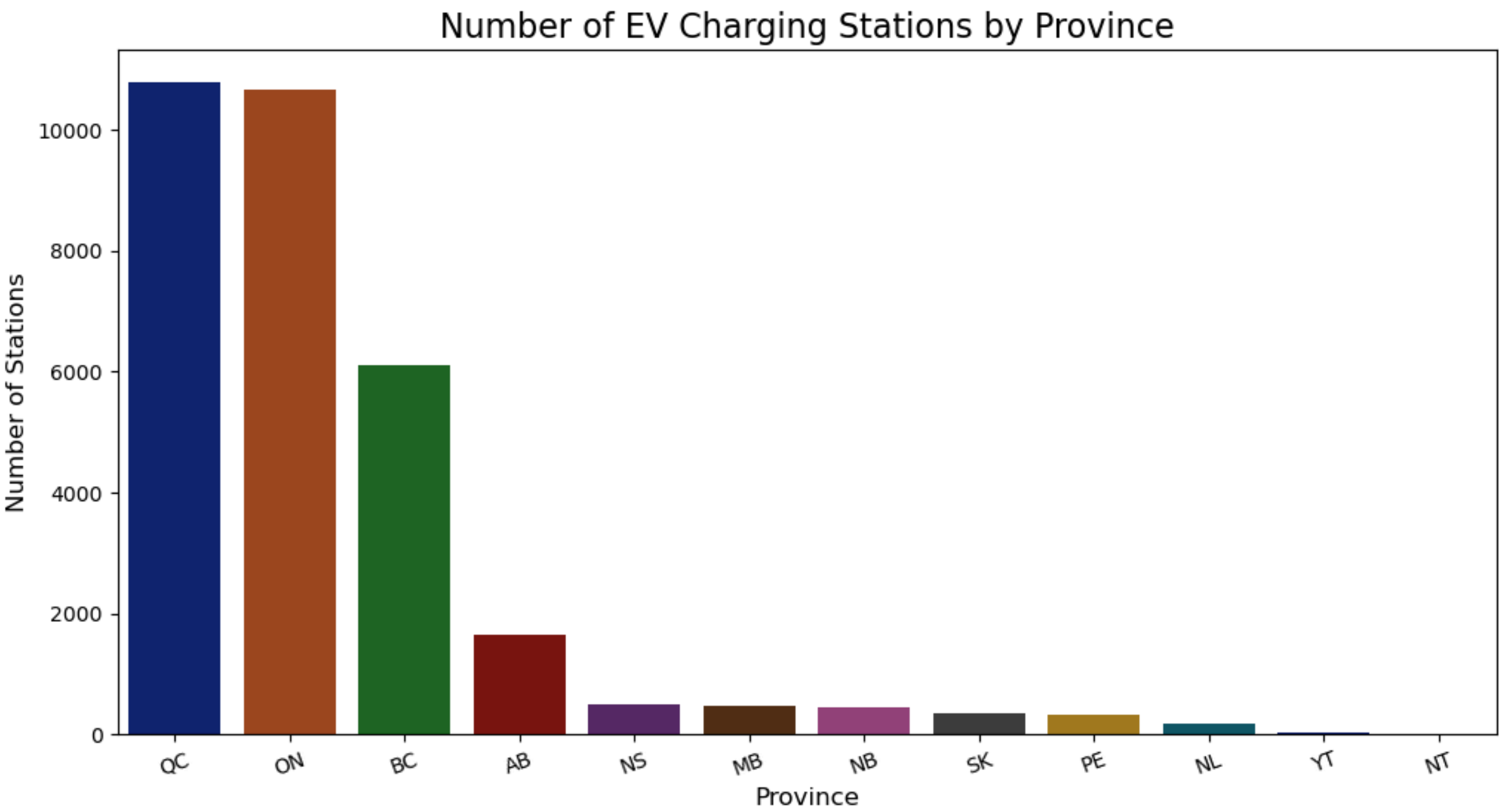


In []: *#Observation:*
#The line chart illustrates the steady growth of EV charging stations in Canada over the years.
#Starting with modest growth until 2018, there is a significant spike in the number of stations from 2019 onward,
#reflecting increased investment in EV infrastructure.
#The dip in 2020 could be attributed to the COVID-19 pandemic, which likely impacted infrastructure projects.
#However, the rapid growth from 2021 onward showcases a strong push toward supporting EV adoption and expanding charging accessibility across the country.

In []:

In [47]: *# Count the number of stations per province*
stations_by_province = df['State'].value_counts()

In [241... *# Bar plot for province-wise distribution*
plt.figure(figsize=(12, 6))
sns.barplot(x=stations_by_province.index, y=stations_by_province.values, hue=stations_by_province.index,palette='dark')
#Assigning a dummy hue to use palette &avoid warning
plt.title('Number of EV Charging Stations by Province', fontsize=16)
plt.xlabel('Province', fontsize=12)
plt.ylabel('Number of Stations', fontsize=12)
plt.xticks(rotation=20)
plt.show()

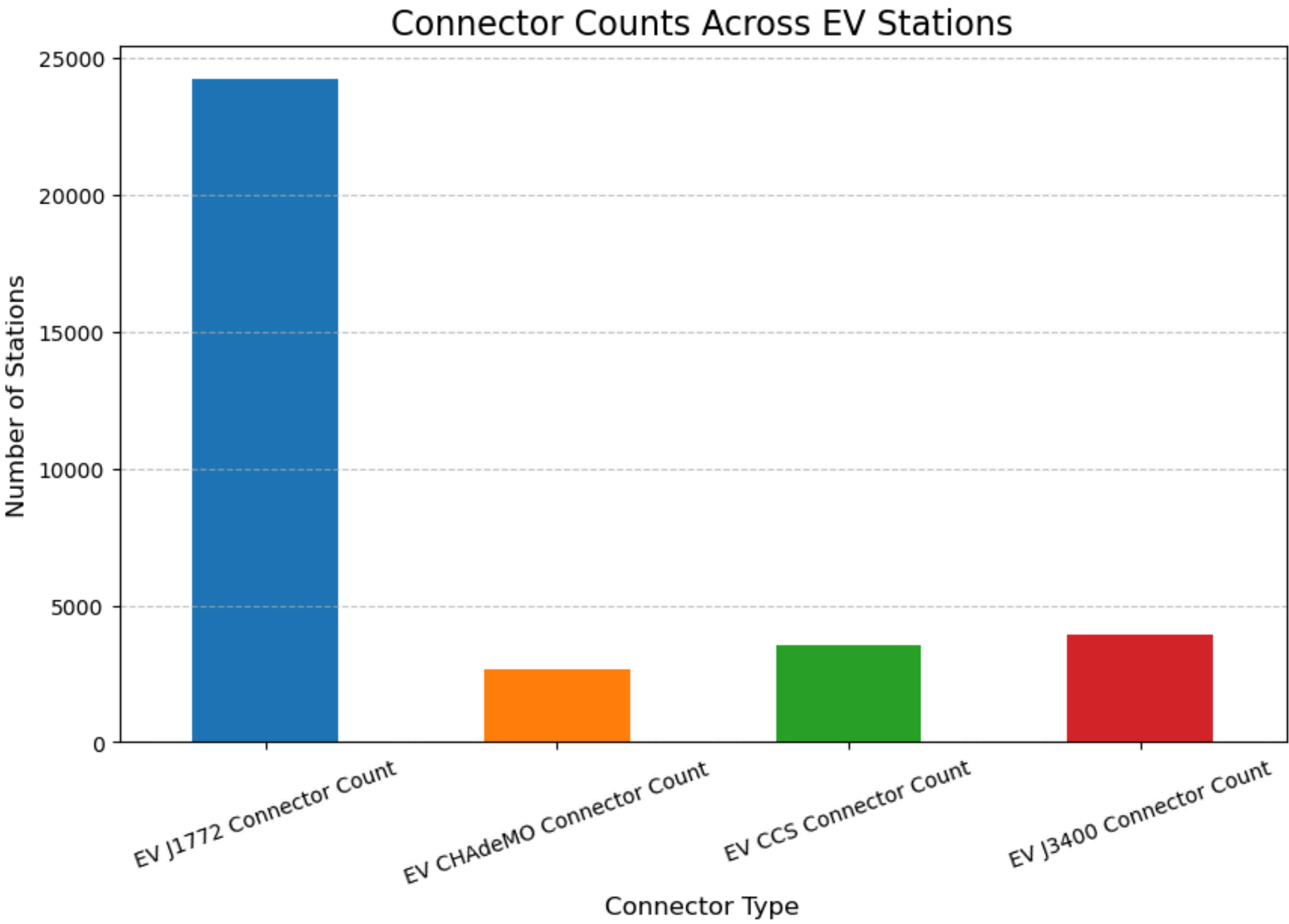


In []: *##Observation = The bar chart highlights the distribution of EV charging stations across provinces. Quebec, Ontario, and British Columbia have the highest number of stations, showing a strong focus on EV infrastructure in these regions.*

In []:

```
In [243... # Sum the counts of each connector type
connector_counts_by_type = df[['EV J1772 Connector Count', 'EV CHAdeMO Connector Count', 'EV CCS Connector Count', 'EV J3400 Connector Count']].sum()

# Plot as a bar chart
plt.figure(figsize=(10, 6))
connector_counts_by_type.plot(kind='bar', color=['#1f77b4', '#ff7f0e', '#2ca02c', '#d62728'])
plt.title('Connector Counts Across EV Stations', fontsize=16)
plt.xlabel('Connector Type', fontsize=12)
plt.ylabel('Number of Stations', fontsize=12)
plt.xticks(rotation=20)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```



In []: *##Observation:*
#The above bar chart displays the distribution of connector counts across EV stations in Canada.
#The J1772 connector dominates, reflecting its widespread use in Level 2 charging infrastructure.
#CHAdeMO, CCS, and J3400 connectors are less common and are critical for fast charging, supporting diverse EV models and use cases.
#This distribution underscores the focus on both standard and fast charging infrastructure.

In []:

```
In [107... print(df["EV Connector Types"].unique()) #different types of EV connectors
```

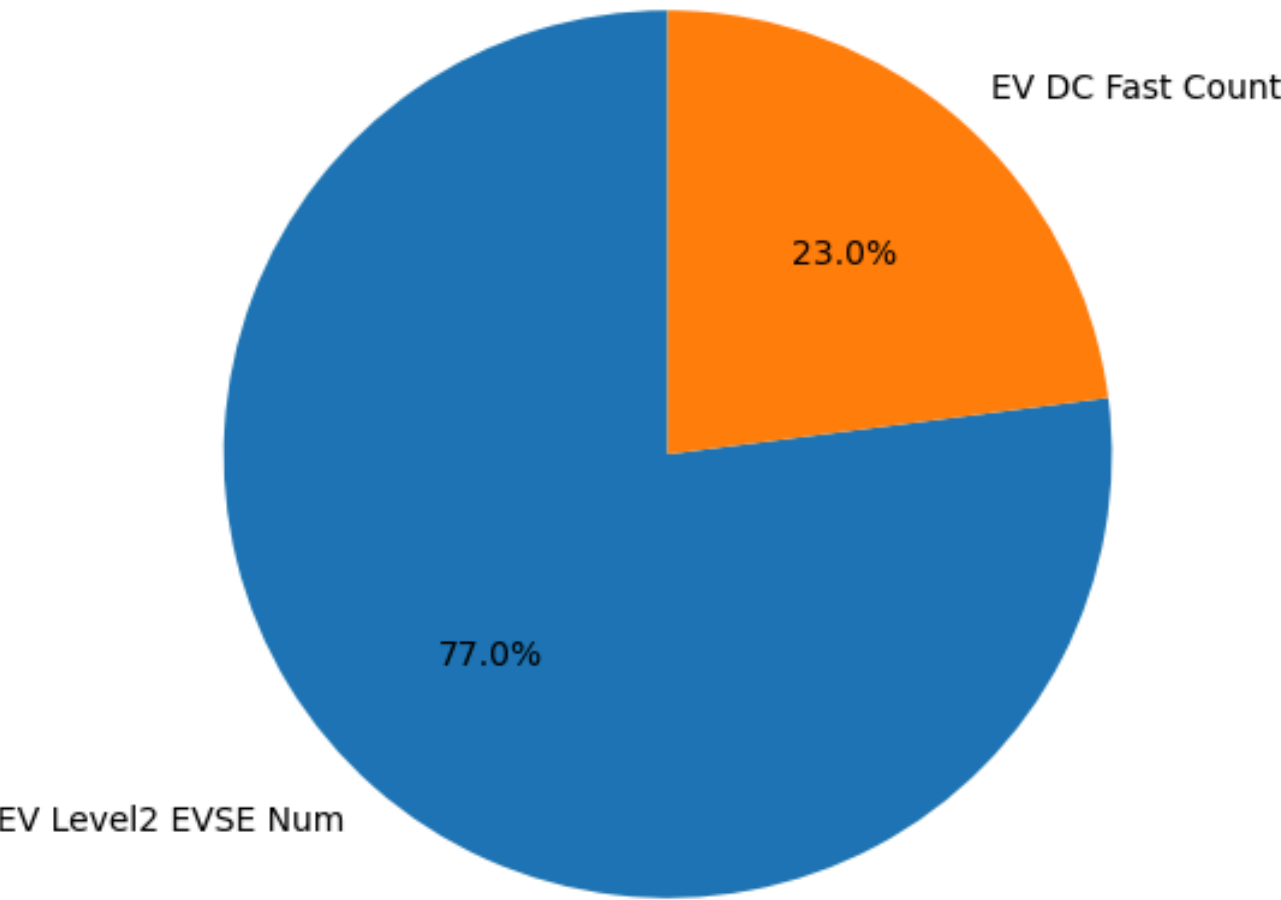
```
['J1772' 'CHADEMO J1772 J1772COMBO NEMA520' 'CHADEMO J1772COMBO'
'J1772COMBO' 'TESLA' 'CHADEMO J1772 J1772COMBO' 'J1772 TESLA'
'J1772 J1772COMBO' 'J1772 NEMA515' 'CHADEMO J1772' 'CHADEMO'
'J1772 NEMA520' 'CHADEMO J1772 NEMA520' 'NEMA520' 'J1772COMBO TESLA'
'CHADEMO J1772COMBO TESLA' 'J1772 NEMA1450']
```

In [193...

```
# Sum Level 2 and DC Fast charging ports
charging_counts = df[['EV Level2 EVSE Num', 'EV DC Fast Count']].sum()

# Plot a pie chart
plt.figure(figsize=(8, 6))
charging_counts.plot(kind='pie', autopct='%1.1f%%', startangle=90, colors=['#1f77b4', '#ff7f0e'])
plt.title('Proportion of Level 2 vs DC Fast Charging Ports', fontsize=16)
plt.ylabel('') # Removing the y-axis label
plt.show()
```

Proportion of Level 2 vs DC Fast Charging Ports



In []:

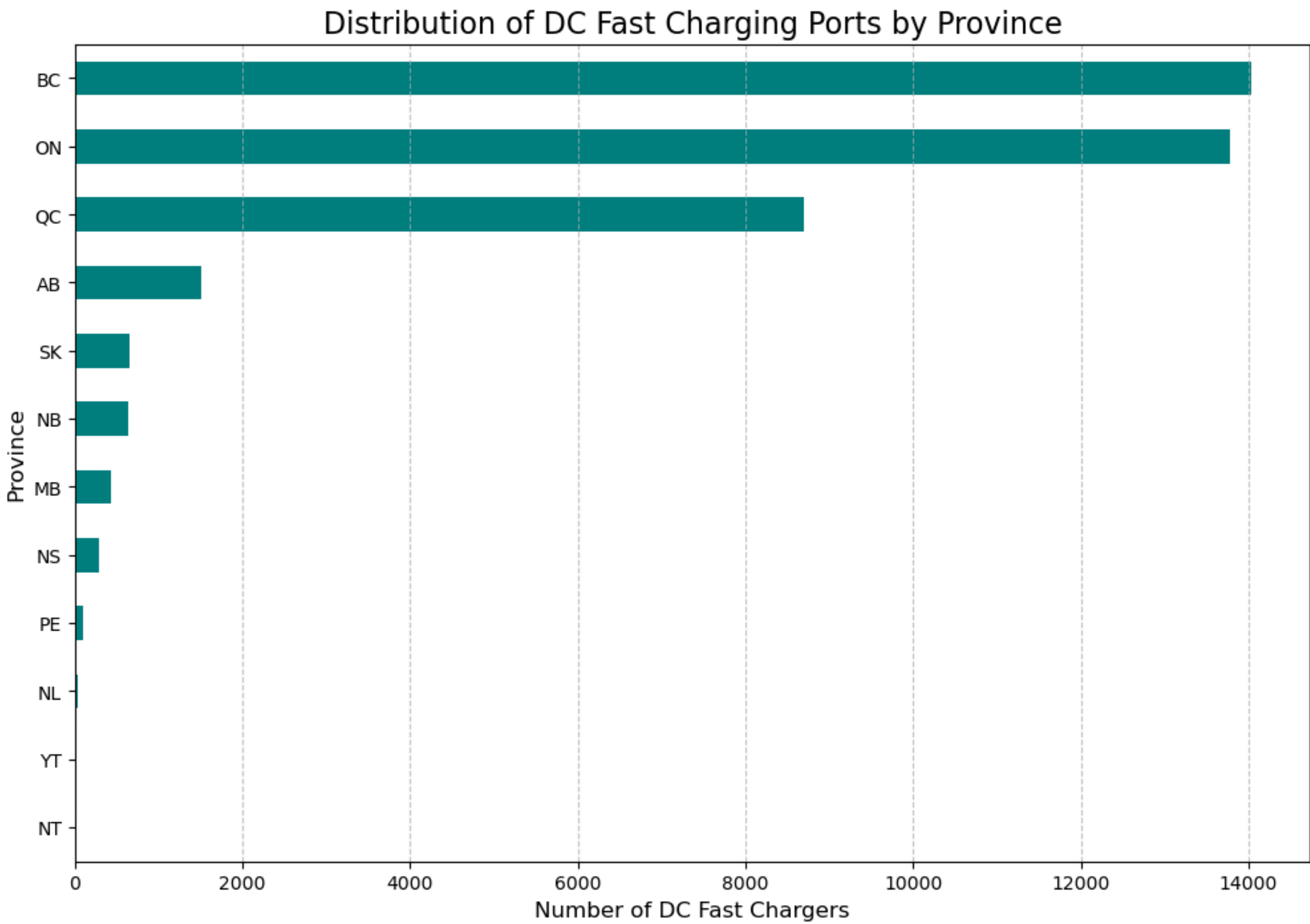
```
#Observation:
#The above pie chart shows the proportion of Level 2 and DC fast chargers in Canada’s EV charging network.
#Level 2 chargers dominate, accounting for the majority of charging infrastructure,
#while DC fast chargers make up a smaller but critical proportion for rapid charging needs.
```

In []:

In [195...

```
# Group by State and sum DC fast chargers
dc_fast_by_state = df.groupby('State')['EV DC Fast Count'].sum().sort_values()

# Plot a horizontal bar chart
plt.figure(figsize=(12, 8))
dc_fast_by_state.plot(kind='barh', color='teal')
plt.title('Distribution of DC Fast Charging Ports by Province', fontsize=16)
plt.xlabel('Number of DC Fast Chargers', fontsize=12)
plt.ylabel('Province', fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()
```

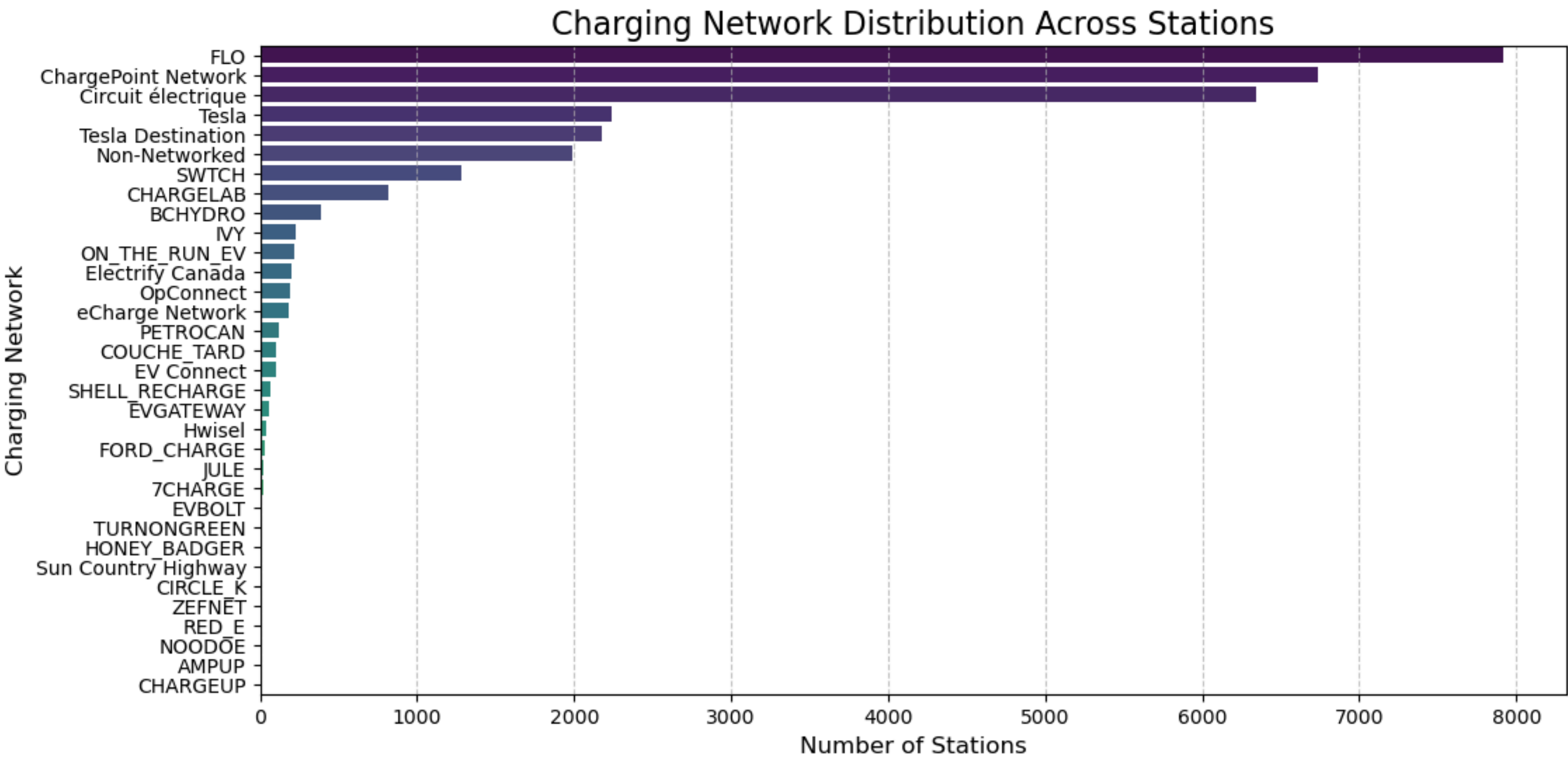


In []: *#Observation:*
#The above horizontal bar chart highlights the distribution of DC fast chargers across provinces.
#Quebec, Ontario, and British Columbia lead in fast charging infrastructure,
#while smaller provinces have limited availability, potentially requiring more investments.

In []:

```
In [213... # Count stations by network
network_counts = df['EV Network'].value_counts()

# Plot a horizontal bar chart for clarity
plt.figure(figsize=(12, 6))
sns.barplot(x=network_counts.values, y=network_counts.index, hue=network_counts.index, palette='viridis')
plt.title('Charging Network Distribution Across Stations', fontsize=16)
plt.xlabel('Number of Stations', fontsize=12)
plt.ylabel('Charging Network', fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()
```



In []: *#Observation:*
#The horizontal bar chart shows the distribution of charging stations across networks.
#Flo and Tesla dominate, showcasing their extensive infrastructure.
#Other networks contribute significantly in specific regions, supporting diverse EV user needs.

In []:

```
In [215... # Calculate the total number of stations per network
overall_top_networks = df['EV Network'].value_counts().head(5).index
```

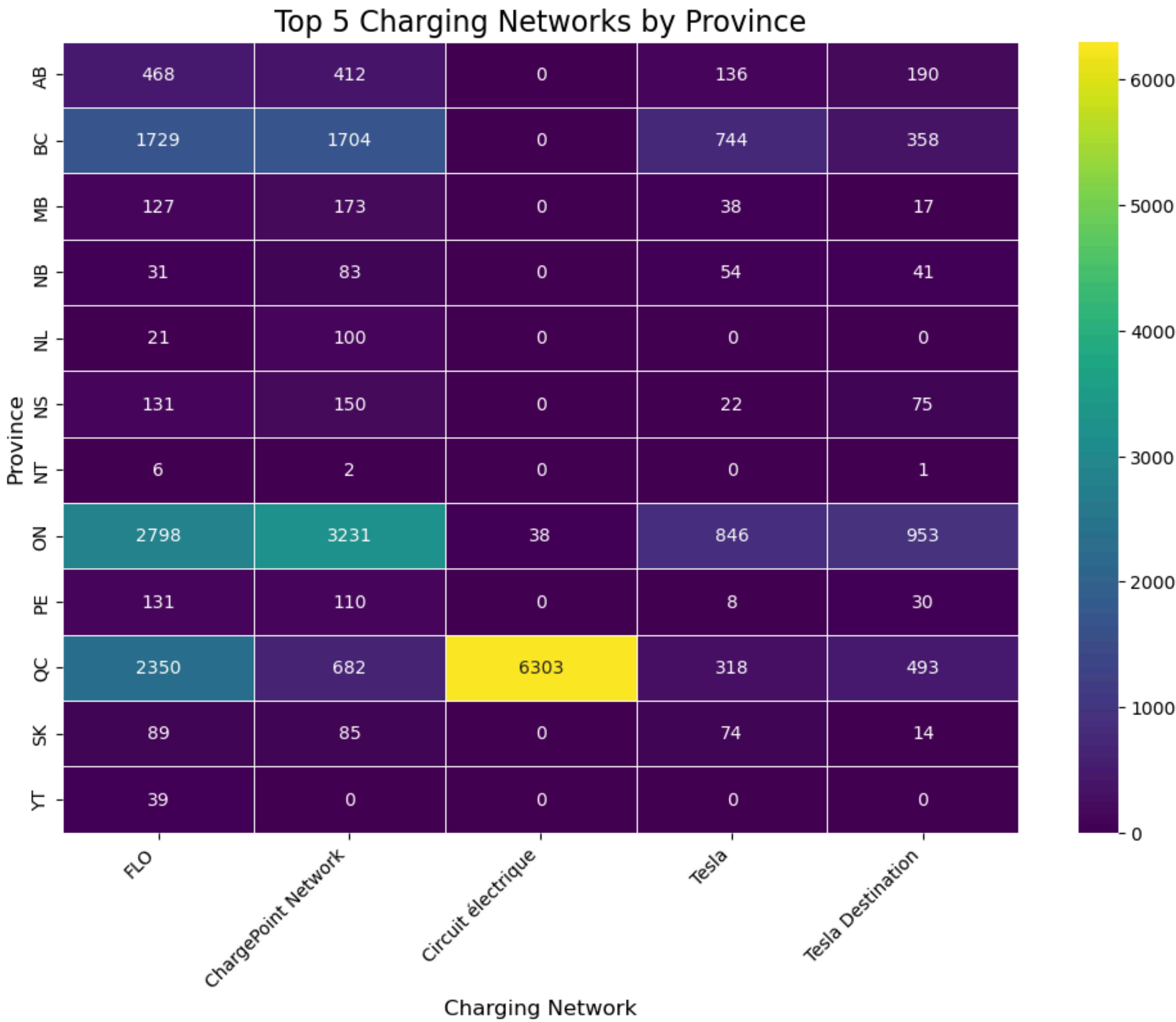
```
print("Top 5 Charging Networks:", list(overall_top_networks))
```

Top 5 Charging Networks: ['FLO', 'ChargePoint Network', 'Circuit électrique', 'Tesla', 'Tesla Destination']

```
In [217... # Group by State and EV Network and count the stations
network_by_state = df.groupby(['State', 'EV Network']).size().unstack(fill_value=0)

# Filter for only the top 5 networks
top_network_by_state = network_by_state[overall_top_networks]
```

```
In [227... # Plot the heatmap
plt.figure(figsize=(12, 8))
sns.heatmap(top_network_by_state, cmap="viridis", annot=True, fmt="d", linewidths=.5, cbar=True)
plt.title('Top 5 Charging Networks by Province', fontsize=16)
plt.xlabel('Charging Network', fontsize=12)
plt.ylabel('Province', fontsize=12)
plt.xticks(rotation=45, ha='right')
plt.show()
```

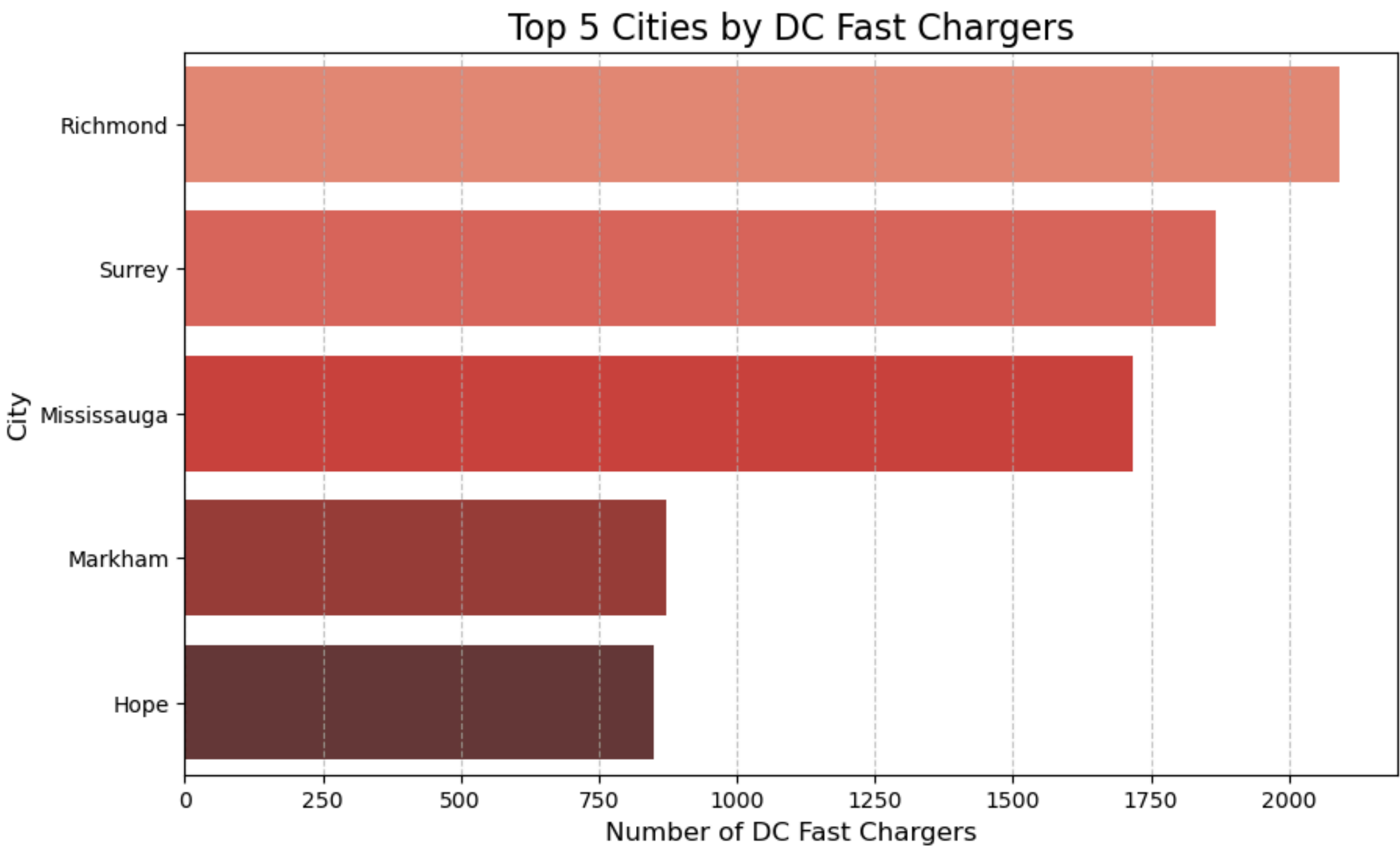


```
In [ ]: #Observation:
#The above heatmap shows the distribution of the top 5 charging networks across provinces.
#FLO is the most dominant network, with significant presence across most provinces.
#Tesla has a strong presence in regions like Ontario and British Columbia.
#Circuit électrique dominating in quebec and ChargePoint Network dominating in Ontario, reflecting their partnerships in key regions.
```

```
In [ ]:
```

```
In [239... # Group by city and sum DC fast chargers
top_cities_dc_fast = df.groupby('City')['EV DC Fast Count'].sum().sort_values(ascending=False).head(5)

# Plot the top 5 cities by DC fast chargers
plt.figure(figsize=(10, 6))
sns.barplot(x=top_cities_dc_fast.values, y=top_cities_dc_fast.index, hue=top_cities_dc_fast.index, palette="Reds_d")
plt.title('Top 5 Cities by DC Fast Chargers', fontsize=16)
plt.xlabel('Number of DC Fast Chargers', fontsize=12)
plt.ylabel('City', fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()
```

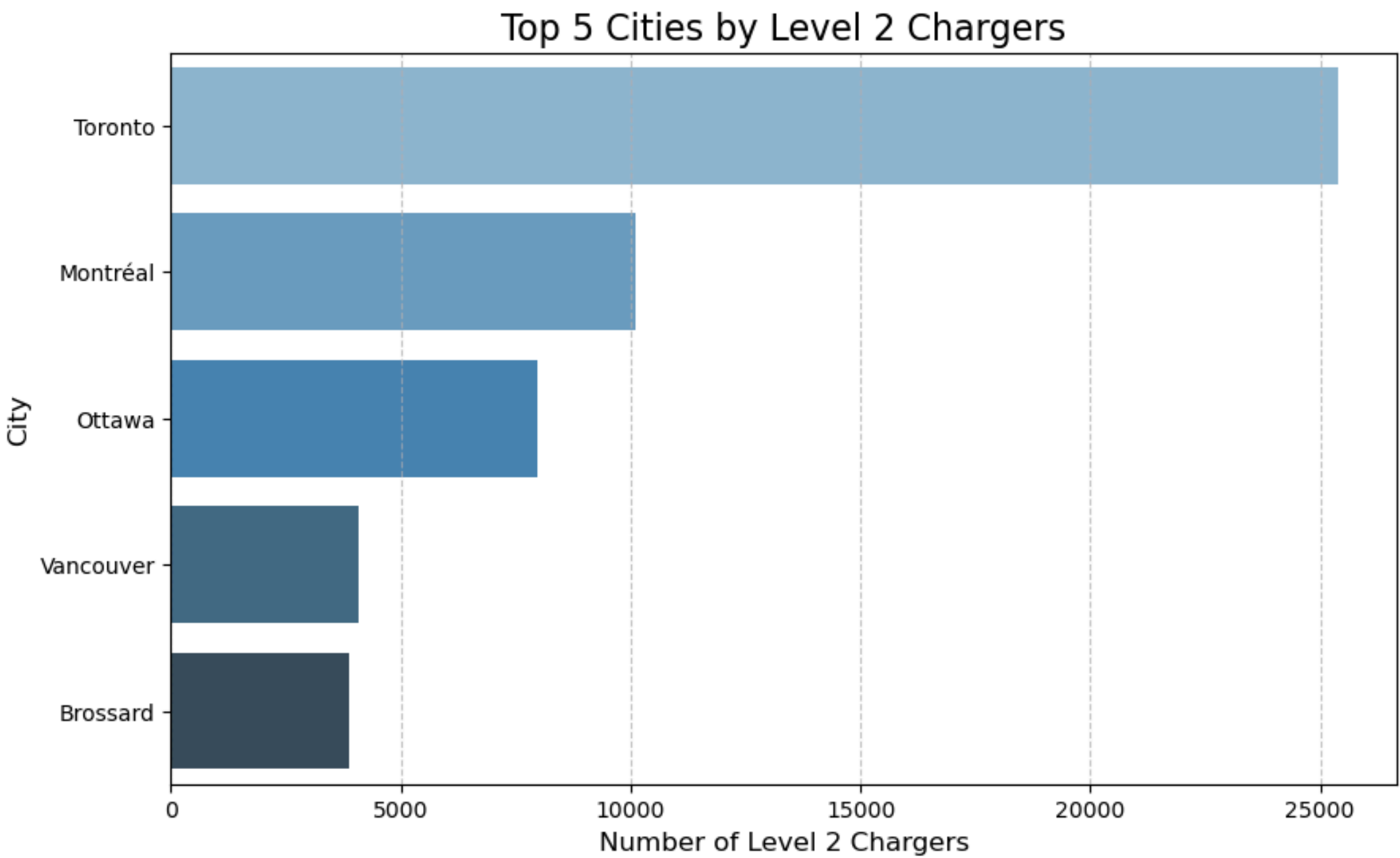



In []: *#Observation:*
#The above chart shows the top 5 cities with the highest number of DC fast chargers.
#These cities are crucial for enabling long-distance EV travel and reducing range anxiety.

In []:

```
In [235... # Group by city and sum Level 2 chargers
top_cities_level2 = df.groupby('City')['EV Level2 EVSE Num'].sum().sort_values(ascending=False).head(5)

# Plot the top 5 cities by Level 2 chargers
plt.figure(figsize=(10, 6))
sns.barplot(x=top_cities_level2.values, y=top_cities_level2.index, hue=top_cities_level2.index,palette="Blues_d")
plt.title('Top 5 Cities by Level 2 Chargers', fontsize=16)
plt.xlabel('Number of Level 2 Chargers', fontsize=12)
plt.ylabel('City', fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()
```



In []: *#Observation:*
#The above chart highlights cities with the highest capacity for Level 2 chargers,
#reflecting their focus on standard EV charging infrastructure.