fdi-province

August 24, 2024

```
[174]: %%HTML
       <script src="require.js"></script>
      <IPython.core.display.HTML object>
[175]: # Library cell
       import pandas as pd
       import geopandas as gpd
       import geoplot
       import geoplot.crs as gcrs
       import matplotlib.pyplot as plt
       import plotly.express as px
       # ignore warnings
       import warnings
       warnings.filterwarnings('ignore')
       import plotly.io as pio
       pio.renderers.default='notebook'
[176]: # Function cell
       # Function cell
       ## Find non-numeric values
       def find_non_numeric_values(df):
           non_numeric_columns = df.select_dtypes(include=['object']).columns
           non_numeric_values = {}
           for col in non_numeric_columns:
               # Change the column to numeric type, if it isn't numeric, it will be
        \hookrightarrow converted to NaN
               temp_col = pd.to_numeric(df[col], errors='coerce')
               # Fill the NaN values with the original values
               non_numeric_data = df[temp_col.isna() & df[col].notna()]
               if not non_numeric_data.empty:
                   non_numeric_values[col] = non_numeric_data[col].tolist()
           return non_numeric_values
       ## Remove non-numeric values
       def remove commas and convert(df):
           non_numeric_columns = df.select_dtypes(include=['object']).columns
```

```
for col in non_numeric_columns:

# Check if the column contains any non-numeric values

try:

# Remove commas from the column

temp_col = df[col] = df[col].str.replace(',', '')

temp_col_numeric = pd.to_numeric(temp_col, errors='raise')

# If the column can be converted to numeric, replace the original_

column with the new column

df[col] = temp_col_numeric

except ValueError:

# If the column contains non-numeric values, keep it

continue

return df
```

0.1 LOAD DATA

<class 'geopandas.geodataframe.GeoDataFrame'>

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

```
[178]:
          Order
                        Provinces Number of new projects \
                 TP. Ho Chi Minh
       0
              1
                                                       836
       1
              2
                        Hai Phong
                                                        52
       2
              3
                           Ha Noi
                                                       453
       3
              4
                       Binh Duong
                                                       256
              5
                         Dong Nai
                                                        91
```

```
Adjusted capital (million USD) \
0 619.07
1 429.24
2 504.47
3 641.97
4 921.05
```

```
Number of times of capital contribution to buy shares \
       0
                                                          1935
                                                             27
       1
       2
                                                            228
       3
                                                             25
                                                             55
         Value of capital contribution, share purchase\n(million USD)
                                                                           Year
       0
                                                       1802.56
                                                                            2016
       1
                                                         96.34
                                                                            2016
       2
                                                        367.21
                                                                            2016
       3
                                                         94.72
                                                                            2016
                                                        273.44
                                                                            2016
[179]: df.tail()
[179]:
                     Provinces Number of new projects
            Order
       436
               437
                      Ha Giang
                                                    NaN
       437
               438
                      Lai Chau
                                                    NaN
       438
               439
                       Lao Cai
                                                    NaN
       439
               440
                    Quang Binh
                                                    NaN
       440
               441
                        Son La
                                                    NaN
           Newly registered capital (million USD) Adjusted project number \
       436
                                                 NaN
                                                                           NaN
       437
                                                                          NaN
                                                 NaN
       438
                                                 NaN
                                                                           NaN
       439
                                                 NaN
                                                                           NaN
       440
                                                 NaN
                                                                          NaN
           Adjusted capital (million USD)
       436
                                        NaN
       437
                                        NaN
       438
                                        NaN
       439
                                        NaN
       440
                                        NaN
           Number of times of capital contribution to buy shares
       436
                                                              NaN
       437
                                                              NaN
       438
                                                              NaN
       439
                                                              NaN
       440
                                                              NaN
           Value of capital contribution, share purchase\n(million USD)
                                                                             Year
       436
                                                              NaN
                                                                              2022
       437
                                                                              2022
                                                              NaN
```

438	NaN	2022
439	NaN	2022
440	NaN	2022

0.2 Analyze the data

```
[180]: # Drop column Order
      n_df = df.drop(columns=['Order'])
      # Show shape data
      print(n_df.shape, end='\n ----- \n')
      # Show info data
      print(n_df.info(), end='\n -----\n')
      # Check for Duplicate
      print(n_df.nunique(), end='\n ----- \n')
      # Check data exist nan or not (bool)
      print(n_df.isnull().any(), end='\n ----- \n')
      # Check for missing value
      print(n_df.isna().sum(), end='\n -----\n')
      (441, 8)
      <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 441 entries, 0 to 440
     Data columns (total 8 columns):
      # Column
                                                                    Non-Null Count
     Dtype
                                                                    _____
         ----
        Provinces
                                                                    441 non-null
     object
          Number of new projects
                                                                    386 non-null
      1
     object
          Newly registered capital (million USD)
                                                                    387 non-null
     object
          Adjusted project number
                                                                    344 non-null
     object
          Adjusted capital (million USD)
                                                                    344 non-null
     object
          Number of times of capital contribution to buy shares
      5
                                                                    378 non-null
     object
          Value of capital contribution, share purchase
      (million USD) 377 non-null
                                   object
      7
          Year
                                                                    441 non-null
      int64
     dtypes: int64(1), object(7)
     memory usage: 27.7+ KB
     None
```

Provinces	63
Number of new projects	100
Newly registered capital (million USD)	351
Adjusted project number	76
Adjusted capital (million USD)	282
Number of times of capital contribution to buy shares	99
Value of capital contribution, share purchase\n(million USD)	330
Year	7
dtype: int64	
Provinces	False
Number of new projects	True
Newly registered capital (million USD)	True
Adjusted project number	True
Adjusted capital (million USD)	True
Number of times of capital contribution to buy shares	True
Value of capital contribution, share purchase\n(million USD) Year	True False
	raise
dtype: bool	
Provinces	0
Number of new projects	55
Newly registered capital (million USD)	54
Adjusted project number	97
Adjusted capital (million USD)	97
Number of times of capital contribution to buy shares	63
Value of capital contribution, share purchase\n(million USD)	64
Year	0
dtype: int64	

Observations

- The shape of dataset fdi_provinces_en.csv is 441 rows and 8 columns
- Only Year column dftype int, so we will convert some columns to numeric for consistency to calculate and explore the data.
- Check all columns to get boolean values indicating if missing values exist and determine which columns have missing values
- Only Year column not exist missing value.

0.3 Data Cleaning

0.3.1 Step-by-step

- 1. Get all "not numeric" from all columns with func find_non_numeric_values()
- 2. Format numeric with func remove_commas_and_convert()
- 3. Remove special character
- 4. Fill all NaN to 0

- 5. Drop Province and Year column for consistency data to numeric
- 6. Re-execute find_non_numeric_values() to check result
- 7. Random select rows to print for review

```
[181]: ## Check for not numeric value
non_numeric_dict = find_non_numeric_values(n_df)
if non_numeric_dict:
    for col, values in non_numeric_dict.items():
        print(f"Column '{col}' have values not numeric:")
        print(values)
else:
    print("No non-numeric values found.")
```

Column 'Provinces' have values not numeric: ['TP. Ho Chi Minh', 'Hai Phong', 'Ha Noi', 'Binh Duong', 'Dong Nai', 'Bac Giang', 'Bac Ninh', 'Long An', 'Ha Nam', 'Tay Ninh', 'Phu Yen', 'Quang Ninh', 'Ba Ria - Vung Tau', 'Hai Duong', 'Tien Giang', 'Hung Yen', 'Ha Tinh', 'Vinh Phuc', 'Nam Dinh', 'Tra Vinh', 'Can Tho', 'Thanh Hoa', 'Phu Tho', 'Thai Nguyen', 'Vinh Long', 'Quang Nam', 'Binh Phuoc', 'Da Nang', 'Ninh Binh', 'Ninh Thuan', 'Binh Dinh', 'Nghe An', 'Hau Giang', 'Khanh Hoa', 'Thai Binh', 'Tuyen Quang', 'Quang Binh', 'Lam Dong', 'Ben Tre', 'Ca Mau', 'Thua Thien Hue', 'Lao Cai', 'Quang Ngai', 'Dong Thap', 'Ha Giang', 'An Giang', 'Hoa Binh', 'Lang Son', 'Binh Thuan', 'Kon Tum', 'Soc Trang', 'Kien Giang', 'Quang Tri', 'Yen Bai', 'Dak Lak', 'Dak Nong', 'Gia Lai', 'Bac Kan', 'Bac Lieu', 'Dien Bien', 'Cao Bang', 'Lai Chau', 'Son La', 'TP. Ho Chi Minh', 'Bac Ninh', 'Thanh Hoa', 'Binh Duong', 'Khanh Hoa', 'Ha Noi', 'Nam Dinh', 'Dong Nai', 'Kien Giang', 'Tay Ninh', 'Hai Phong', 'Bac Giang', 'Ba Ria - Vung Tau', 'Hung Yen', 'Binh Phuoc', 'Long An', 'Quang Ngai', 'Hai Duong', 'Ninh Thuan', 'Ha Nam', 'Yen Bai', 'Ben Tre', 'Ninh Binh', 'Phu Tho', 'Quang Binh', 'Vinh Phuc', 'Binh Dinh', 'Tien Giang', 'Tra Vinh', 'Da Nang', 'Quang Nam', 'Vinh Long', 'Nghe An', 'Thai Nguyen', 'Thai Binh', 'Ha Tinh', 'Dong Thap', 'Lam Dong', 'Dak Lak', 'Quang Ninh', 'Hoa Binh', 'Binh Thuan', 'Can Tho', 'Dak Nong', 'Ca Mau', 'Soc Trang', 'Lao Cai', 'Son La', 'Cao Bang', 'An Giang', 'Thua Thien Hue', 'Dien Bien', 'Ha Giang', 'Quang Tri', 'Lang Son', 'Tuyen Quang', 'Phu Yen', 'Kon Tum', 'Hau Giang', 'Bac Lieu', 'Bac Kan', 'Gia Lai', 'Lai Chau', 'Ha Noi', 'TP. Ho Chi Minh', 'Hai Phong', 'Binh Duong', 'Ba Ria - Vung Tau', 'Dong Nai', 'Thua Thien Hue', 'Bac Ninh', 'Tay Ninh', 'Long An', 'Hai Duong', 'Bac Giang', 'Binh Phuoc', 'Hung Yen', 'Quang Nam', 'Thai Nguyen', 'Ha Nam', 'Ninh Thuan', 'Quang Ninh', 'Ben Tre', 'Vinh Phuc', 'Bac Lieu', 'Quang Ngai', 'Thanh Hoa', 'Kien Giang', 'Da Nang', 'Nam Dinh', 'Tien Giang', 'Hoa Binh', 'Ninh Binh', 'Vinh Long', 'Phu Tho', 'Binh Dinh', 'Tra Vinh', 'Ha Tinh', 'Khanh Hoa', 'Soc Trang', 'Thai Binh', 'Dak Nong', 'Ca Mau', 'Can Tho', 'Quang Binh', 'Dak Lak', 'Tuyen Quang', 'Nghe An', 'Binh Thuan', 'Phu Yen', 'Lang Son', 'Kon Tum', 'Lam Dong', 'Yen Bai', 'Dong Thap', 'Hau Giang', 'An Giang', 'Son La', 'Lao Cai', 'Quang Tri', 'Ha Giang', 'Bac Kan', 'Cao Bang', 'Dien Bien', 'Lai Chau', 'Gia Lai', 'Ha Noi', 'TP. Ho Chi Minh', 'Binh Duong', 'Dong Nai', 'Bac Ninh', 'Hai Phong', 'Tay Ninh', 'Bac Giang', 'Ba Ria - Vung Tau', 'Ha Nam', 'Long An', 'Hai Duong', 'Thai Nguyen', 'Vinh Phuc', 'Da Nang', 'Hung Yen', 'Binh Phuoc', 'Tien Giang', 'Thanh Hoa',

```
'Phu Tho', 'Thua Thien Hue', 'Nghe An', 'Quang Ninh', 'Phu Yen', 'Khanh Hoa',
'Quang Nam', 'Binh Thuan', 'Vinh Long', 'Ninh Binh', 'Quang Ngai', 'Ninh Thuan',
'Bac Lieu', 'Soc Trang', 'Tra Vinh', 'Binh Dinh', 'Ca Mau', 'Hau Giang', 'Can
Tho', 'Thai Binh', 'An Giang', 'Nam Dinh', 'Ben Tre', 'Ha Tinh', 'Lam Dong',
'Kien Giang', 'Quang Tri', 'Tuyen Quang', 'Dong Thap', 'Yen Bai', 'Kon Tum',
'Bac Kan', 'Dak Lak', 'Lang Son', 'Dien Bien', 'Lao Cai', 'Quang Binh', 'Ha
Giang', 'Son La', 'Dak Nong', 'Cao Bang', 'Lai Chau', 'Hoa Binh', 'Gia Lai',
'TP. Ho Chi Minh', 'Bac Lieu', 'Ha Noi', 'Ba Ria - Vung Tau', 'Binh Duong', 'Hai
Phong', 'Dong Nai', 'Bac Ninh', 'Bac Giang', 'Long An', 'Ha Nam', 'Ben Tre',
'Tay Ninh', 'Vinh Phuc', 'Hai Duong', 'Quang Ninh', 'Hung Yen', 'Thai Nguyen',
'Binh Phuoc', 'Thanh Hoa', 'Phu Tho', 'Quang Binh', 'Vinh Long', 'Da Nang',
'Nghe An', 'Tien Giang', 'Dak Nong', 'Quang Ngai', 'Tra Vinh', 'Ninh Binh',
'Binh Thuan', 'Nam Dinh', 'Thai Binh', 'Khanh Hoa', 'Soc Trang', 'Binh Dinh',
'Can Tho', 'Thua Thien Hue', 'Ca Mau', 'Hoa Binh', 'Quang Tri', 'Lam Dong',
'Dong Thap', 'Ha Tinh', 'Kien Giang', 'Tuyen Quang', 'Gia Lai', 'Yen Bai<sup>'</sup>, 'Lao
Cai', 'An Giang', 'Hau Giang', 'Dak Lak', 'Son La', 'Phu Yen', 'Kon Tum', 'Dien
Bien', 'Cao Bang', 'Quang Nam', 'Ninh Thuan', 'Bac Kan', 'Ha Giang', 'Lai Chau',
'Lao Cai', 'Hai Phong', 'Long An', 'TP. Ho Chi Minh', 'Binh Duong', 'Bac Ninh',
'Ha Noi', 'Dong Nai', 'Can Tho', 'Bac Giang', 'Quang Ninh', 'Tay Ninh', 'Vinh
Phuc', 'Hung Yen', 'Phu Tho', 'Thai Binh', 'Dak Lak', 'Binh Phuoc', 'Ba Ria -
Vung Tau', 'Hai Duong', 'Nghe An', 'Ha Nam', 'Thai Nguyen', 'Thanh Hoa', 'Thua
Thien Hue', 'Da Nang', 'Ninh Binh', 'Hau Giang', 'Kon Tum', 'Nam Dinh', 'Tien
Giang', 'Binh Dinh', 'Quang Tri', 'Ninh Thuan', 'Quang Binh', 'Yen Bai', 'Vinh
Long', 'Dong Thap', 'Quang Ngai', 'Quang Nam', 'Khanh Hoa', 'Ca Mau', 'Binh
Thuan', 'Lam Dong', 'Tra Vinh', 'An Giang', 'Ha Tinh', 'Lao Cai', 'Dak Nong',
'Kien Giang', 'Lang Son', 'Gia Lai', 'Phu Yen', 'Cao Bang', 'Lai Chau', 'Soc
Trang', 'Bac Lieu', 'Hoa Binh', 'Tuyen Quang', 'Ben Tre', 'Bac Kan', 'Dien
Bien', 'Ha Giang', 'Son La', 'TP. Ho Chi Minh', 'Binh Duong', 'Quang Ninh', 'Bac
Ninh', 'Hai Phong', 'Ha Noi', 'Thai Nguyen', 'Dong Nai', 'Bac Giang', 'Ba Ria -
Vung Tau', 'Nghe An', 'Long An', 'Hung Yen', 'Phu Tho', 'Tay Ninh', 'Ha Nam',
'Hai Duong', 'Thai Binh', 'Ha Tinh', 'Vinh Phuc', 'Binh Phuoc', 'Tien Giang',
'Thua Thien Hue', 'Can Tho', 'Vinh Long', 'Da Nang', 'Soc Trang', 'Thanh Hoa',
'Ninh Thuan', 'Quang Ngai', 'Quang Nam', 'Ninh Binh', 'Nam Dinh', 'Binh Dinh',
'Binh Thuan', 'An Giang', 'Dak Lak', 'Yen Bai', 'Khanh Hoa', 'Kien Giang', 'Tra
Vinh', 'Ben Tre', 'Phu Yen', 'Lang Son', 'Hoa Binh', 'Tuyen Quang', 'Dak Nong',
'Kon Tum', 'Quang Tri', 'Hau Giang', 'Ca Mau', 'Gia Lai', 'Lam Dong', 'Bac
Lieu', 'Bac Kan', 'Cao Bang', 'Dien Bien', 'Dong Thap', 'Ha Giang', 'Lai Chau',
'Lao Cai', 'Quang Binh', 'Son La']
Column 'Number of new projects' have values not numeric:
['-','-','-','-','-','-','-','-
', ' = ', ' = ', ' = ', ' = ']
Column 'Newly registered capital (million USD)' have values not numeric:
['2,313.95', '3,159.40', '1,356.46', '2,584.86', '1,111.25', '2,134.30',
'1,002.38', '1,342.30', ' - ', '5,041.05', '1,216.58', '1,803.51', '1,382.06',
'1,841.35', '1,546.59', '1,296.70', '4,000.00', '1,064.13', ' - ', ' - ', '
- ', ' - ', ' - ', ' - ', ' - ', '3,518.84', '1,170.51', '1,316.82',
'1,011.55', '- ', '- ', '- ', '- ', '- ', '- ',
```

```
' - ', ' - ', ' - ']
         Column 'Adjusted project number' have values not numeric:
         ['-','-','-','-','-','-','-','-','-
          |, | _ |, | _ |, | _ |, | _ |, | _ |, | _ |, | _ |, | _ |
          ', ' = ', ' = ', ' = ', ' = ']
         Column 'Adjusted capital (million USD)' have values not numeric:
         ['1,000.11', '2,888.34', '1,117.00', '1,829.64', '1,140.00', ' - ',
         '1,261.91', '1,489.66', ' - ', ' - ', ' - ', ' - ', ' - ', ' - ', '
         _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` |, | _ ` 
         - ', '2,727.59', '1,124.28', ' - ', ' - ', ' - ', ' - ', ' - ', ' -
           , ' - ', ' - ', ' - ', ' - ', ' - ', ' - ', ' - ', ' - ', ' - ', ' -
          ', '- ', '- ', '- ', '1,600.72', '- ', '1,685.63', '1,212.16', '-
          ', ' - ', ' - ', ' - ', ' - ']
         Column 'Number of times of capital contribution to buy shares' have values not
         numeric:
         ['2,788', '3,710', '1,351', '5,720', '3,640', '---', '---', '2,289', '--
         ', '- ', '- ', '2,411', '- ', '- ']
         Column 'Value of capital contribution, share purchase
         (million USD)' have values not numeric:
         ['3,191.90', '1,703.13', '4,993.11', '6,472.60', '5,595.33', '1,009.55',
          '3,177.38', ' - ', '1,611.82', ' - ', '2,217.58', '1,927.21', ' - ', ' -
          ', '- ', '- ', '1,738.61', '1,196.48', '- ', '- ']
[182]:  ## Drop comma value
          n_df = remove_commas_and_convert(n_df)
[183]: # Drop ' - ' value
          ### Drop ' - ' value column 'Number of new projects'
          n_df['Number of new projects'] = n_df['Number of new projects'].
           replace(to_replace=r'[^0-9.]', value=0, regex=True)
          ### Drop ' - ' value column 'Newly registered capital (million USD)'
          n_df['Newly registered capital (million USD)'] = n_df['Newly registered capital ∪
           →(million USD)'].replace(to_replace=r'[^0-9.]', value=0, regex=True)
          ### Drop ' - ' value column 'Adjusted project number'
          n df['Adjusted project number'] = n df['Adjusted project number'].
           →replace(to_replace=r'[^0-9.]', value=0, regex=True)
          ### Drop ' - ' value column 'Adjusted capital (million USD)'
          n_df['Adjusted capital (million USD)'] = n_df['Adjusted capital (million USD)'].
           ### Drop ' - ' value column 'Number of times of capital contribution to buy
            ⇔shares'
```

'1,909.08', '2,181.17', '1,139.00', ' - ', ' - ', ' - ', ' - ', ' - ',

```
[184]: ## Check for not numeric value
non_numeric_dict = find_non_numeric_values(n_df)
if non_numeric_dict:
    for col, values in non_numeric_dict.items():
        print(f"Column '{col}' have values not numeric:")
        print(values)
else:
    print("No non-numeric values found.")
```

Column 'Provinces' have values not numeric: ['TP. Ho Chi Minh', 'Hai Phong', 'Ha Noi', 'Binh Duong', 'Dong Nai', 'Bac Giang', 'Bac Ninh', 'Long An', 'Ha Nam', 'Tay Ninh', 'Phu Yen', 'Quang Ninh', 'Ba Ria - Vung Tau', 'Hai Duong', 'Tien Giang', 'Hung Yen', 'Ha Tinh', 'Vinh Phuc', 'Nam Dinh', 'Tra Vinh', 'Can Tho', 'Thanh Hoa', 'Phu Tho', 'Thai Nguyen', 'Vinh Long', 'Quang Nam', 'Binh Phuoc', 'Da Nang', 'Ninh Binh', 'Ninh Thuan', 'Binh Dinh', 'Nghe An', 'Hau Giang', 'Khanh Hoa', 'Thai Binh', 'Tuyen Quang', 'Quang Binh', 'Lam Dong', 'Ben Tre', 'Ca Mau', 'Thua Thien Hue', 'Lao Cai', 'Quang Ngai', 'Dong Thap', 'Ha Giang', 'An Giang', 'Hoa Binh', 'Lang Son', 'Binh Thuan', 'Kon Tum', 'Soc Trang', 'Kien Giang', 'Quang Tri', 'Yen Bai', 'Dak Lak', 'Dak Nong', 'Gia Lai', 'Bac Kan', 'Bac Lieu', 'Dien Bien', 'Cao Bang', 'Lai Chau', 'Son La', 'TP. Ho Chi Minh', 'Bac Ninh', 'Thanh Hoa', 'Binh Duong', 'Khanh Hoa', 'Ha Noi', 'Nam Dinh', 'Dong Nai', 'Kien Giang', 'Tay Ninh', 'Hai Phong', 'Bac Giang', 'Ba Ria - Vung Tau', 'Hung Yen', 'Binh Phuoc', 'Long An', 'Quang Ngai', 'Hai Duong', 'Ninh Thuan', 'Ha Nam', 'Yen Bai', 'Ben Tre', 'Ninh Binh', 'Phu Tho', 'Quang Binh', 'Vinh Phuc', 'Binh Dinh', 'Tien Giang', 'Tra Vinh', 'Da Nang', 'Quang Nam', 'Vinh Long', 'Nghe An', 'Thai Nguyen', 'Thai Binh', 'Ha Tinh', 'Dong Thap', 'Lam Dong', 'Dak Lak', 'Quang Ninh', 'Hoa Binh', 'Binh Thuan', 'Can Tho', 'Dak Nong', 'Ca Mau', 'Soc Trang', 'Lao Cai', 'Son La', 'Cao Bang', 'An Giang', 'Thua Thien Hue', 'Dien Bien', 'Ha Giang', 'Quang Tri', 'Lang Son', 'Tuyen Quang', 'Phu Yen', 'Kon Tum', 'Hau Giang', 'Bac Lieu', 'Bac Kan', 'Gia Lai', 'Lai Chau', 'Ha Noi', 'TP. Ho Chi Minh', 'Hai Phong', 'Binh Duong', 'Ba Ria - Vung Tau', 'Dong Nai', 'Thua Thien Hue', 'Bac Ninh', 'Tay Ninh', 'Long An', 'Hai Duong', 'Bac Giang', 'Binh Phuoc', 'Hung Yen', 'Quang Nam', 'Thai Nguyen', 'Ha Nam', 'Ninh Thuan', 'Quang Ninh', 'Ben Tre', 'Vinh Phuc', 'Bac Lieu', 'Quang Ngai', 'Thanh Hoa', 'Kien Giang', 'Da Nang', 'Nam Dinh', 'Tien Giang', 'Hoa Binh', 'Ninh Binh', 'Vinh Long', 'Phu Tho', 'Binh Dinh', 'Tra Vinh', 'Ha Tinh', 'Khanh Hoa', 'Soc Trang', 'Thai Binh', 'Dak Nong', 'Ca Mau', 'Can Tho', 'Quang Binh', 'Dak Lak', 'Tuyen Quang', 'Nghe An', 'Binh

Thuan', 'Phu Yen', 'Lang Son', 'Kon Tum', 'Lam Dong', 'Yen Bai', 'Dong Thap', 'Hau Giang', 'An Giang', 'Son La', 'Lao Cai', 'Quang Tri', 'Ha Giang', 'Bac Kan', 'Cao Bang', 'Dien Bien', 'Lai Chau', 'Gia Lai', 'Ha Noi', 'TP. Ho Chi Minh', 'Binh Duong', 'Dong Nai', 'Bac Ninh', 'Hai Phong', 'Tay Ninh', 'Bac Giang', 'Ba Ria - Vung Tau', 'Ha Nam', 'Long An', 'Hai Duong', 'Thai Nguyen', 'Vinh Phuc', 'Da Nang', 'Hung Yen', 'Binh Phuoc', 'Tien Giang', 'Thanh Hoa', 'Phu Tho', 'Thua Thien Hue', 'Nghe An', 'Quang Ninh', 'Phu Yen', 'Khanh Hoa', 'Quang Nam', 'Binh Thuan', 'Vinh Long', 'Ninh Binh', 'Quang Ngai', 'Ninh Thuan', 'Bac Lieu', 'Soc Trang', 'Tra Vinh', 'Binh Dinh', 'Ca Mau', 'Hau Giang', 'Can Tho', 'Thai Binh', 'An Giang', 'Nam Dinh', 'Ben Tre', 'Ha Tinh', 'Lam Dong', 'Kien Giang', 'Quang Tri', 'Tuyen Quang', 'Dong Thap', 'Yen Bai', 'Kon Tum', 'Bac Kan', 'Dak Lak', 'Lang Son', 'Dien Bien', 'Lao Cai', 'Quang Binh', 'Ha Giang', 'Son La', 'Dak Nong', 'Cao Bang', 'Lai Chau', 'Hoa Binh', 'Gia Lai', 'TP. Ho Chi Minh', 'Bac Lieu', 'Ha Noi', 'Ba Ria - Vung Tau', 'Binh Duong', 'Hai Phong', 'Dong Nai', 'Bac Ninh', 'Bac Giang', 'Long An', 'Ha Nam', 'Ben Tre', 'Tay Ninh', 'Vinh Phuc', 'Hai Duong', 'Quang Ninh', 'Hung Yen', 'Thai Nguyen', 'Binh Phuoc', 'Thanh Hoa', 'Phu Tho', 'Quang Binh', 'Vinh Long', 'Da Nang', 'Nghe An', 'Tien Giang', 'Dak Nong', 'Quang Ngai', 'Tra Vinh', 'Ninh Binh', 'Binh Thuan', 'Nam Dinh', 'Thai Binh', 'Khanh Hoa', 'Soc Trang', 'Binh Dinh', 'Can Tho', 'Thua Thien Hue', 'Ca Mau', 'Hoa Binh', 'Quang Tri', 'Lam Dong', 'Dong Thap', 'Ha Tinh', 'Kien Giang', 'Tuyen Quang', 'Gia Lai', 'Yen Bai', 'Lao Cai', 'An Giang', 'Hau Giang', 'Dak Lak', 'Son La', 'Phu Yen', 'Kon Tum', 'Dien Bien', 'Cao Bang', 'Quang Nam', 'Ninh Thuan', 'Bac Kan', 'Ha Giang', 'Lai Chau', 'Lao Cai', 'Hai Phong', 'Long An', 'TP. Ho Chi Minh', 'Binh Duong', 'Bac Ninh', 'Ha Noi', 'Dong Nai', 'Can Tho', 'Bac Giang', 'Quang Ninh', 'Tay Ninh', 'Vinh Phuc', 'Hung Yen', 'Phu Tho', 'Thai Binh', 'Dak Lak', 'Binh Phuoc', 'Ba Ria -Vung Tau', 'Hai Duong', 'Nghe An', 'Ha Nam', 'Thai Nguyen', 'Thanh Hoa', 'Thua Thien Hue', 'Da Nang', 'Ninh Binh', 'Hau Giang', 'Kon Tum', 'Nam Dinh', 'Tien Giang', 'Binh Dinh', 'Quang Tri', 'Ninh Thuan', 'Quang Binh', 'Yen Bai', 'Vinh Long', 'Dong Thap', 'Quang Ngai', 'Quang Nam', 'Khanh Hoa', 'Ca Mau', 'Binh Thuan', 'Lam Dong', 'Tra Vinh', 'An Giang', 'Ha Tinh', 'Lao Cai', 'Dak Nong', 'Kien Giang', 'Lang Son', 'Gia Lai', 'Phu Yen', 'Cao Bang', 'Lai Chau', 'Soc Trang', 'Bac Lieu', 'Hoa Binh', 'Tuyen Quang', 'Ben Tre', 'Bac Kan', 'Dien Bien', 'Ha Giang', 'Son La', 'TP. Ho Chi Minh', 'Binh Duong', 'Quang Ninh', 'Bac Ninh', 'Hai Phong', 'Ha Noi', 'Thai Nguyen', 'Dong Nai', 'Bac Giang', 'Ba Ria -Vung Tau', 'Nghe An', 'Long An', 'Hung Yen', 'Phu Tho', 'Tay Ninh', 'Ha Nam', 'Hai Duong', 'Thai Binh', 'Ha Tinh', 'Vinh Phuc', 'Binh Phuoc', 'Tien Giang', 'Thua Thien Hue', 'Can Tho', 'Vinh Long', 'Da Nang', 'Soc Trang', 'Thanh Hoa', 'Ninh Thuan', 'Quang Ngai', 'Quang Nam', 'Ninh Binh', 'Nam Dinh', 'Binh Dinh', 'Binh Thuan', 'An Giang', 'Dak Lak', 'Yen Bai', 'Khanh Hoa', 'Kien Giang', 'Tra Vinh', 'Ben Tre', 'Phu Yen', 'Lang Son', 'Hoa Binh', 'Tuyen Quang', 'Dak Nong', 'Kon Tum', 'Quang Tri', 'Hau Giang', 'Ca Mau', 'Gia Lai', 'Lam Dong', 'Bac Lieu', 'Bac Kan', 'Cao Bang', 'Dien Bien', 'Dong Thap', 'Ha Giang', 'Lai Chau', 'Lao Cai', 'Quang Binh', 'Son La']

[185]: ## Drop missing value fill with 0 ### Number of new projects

```
n df['Number of new projects'] = n df['Number of new projects'].fillna(0)
       ### Newly registered capital (million USD)
       n_df['Newly registered capital (million USD)'] = n_df['Newly registered capital_
       ⇔(million USD)'].fillna(0)
       ### Adjusted project number
       n df['Adjusted project number'] = n df['Adjusted project number'].fillna(0)
       ### Adjusted capital (million USD)
       n df['Adjusted capital (million USD)'] = n df['Adjusted capital (million USD)'].
        ofillna(0)
       ### Number of times of capital contribution to buy shares
       n df['Number of times of capital contribution to buy shares'] = n df['Number of ⊔
        otimes of capital contribution to buy shares'].fillna(0)
       ### Value of capital contribution, share purchase\n(million USD)
       n df['Value of capital contribution, share purchase\n(million USD)'] = ∪
        \neg n_{df}['Value of capital contribution, share purchase \n(million USD)'].
        →fillna(0)
[186]: n df.sample(n=10)
[186]:
             Provinces Number of new projects Newly registered capital (million USD) \
             Ninh Binh
                                                                                120.01
       217
                                             7
       171 Binh Thuan
                                             0
                                                                                     0
       5
             Bac Giang
                                            53
                                                                                937.51
       112
              An Giang
                                                                                   0.2
                                             1
       432
               Bac Kan
                                             0
                                                                                      0
       254
                Ha Noi
                                           496
                                                                                711.81
       277
           Tien Giang
                                             9
                                                                                108.19
       181
               Lao Cai
                                             0
                                                                                      0
       396
               Ha Tinh
                                                                                   275
                                             1
       101
               Dak Lak
                                             2
                                                                                  49.5
           Adjusted project number Adjusted capital (million USD) \
       217
                                                              23.04
       171
                                  1
                                                               0.41
       5
                                 20
                                                              71.06
       112
                                  0
                                                                  0
       432
                                  0
                                                                  0
       254
                                                            1261.91
                                158
       277
                                 11
                                                              55.28
       181
                                  0
                                                                  0
       396
                                  0
                                                                  0
       101
                                  1
                                                                  9
           Number of times of capital contribution to buy shares \
       217
                                                             10
       171
                                                             7
       5
                                                             10
```

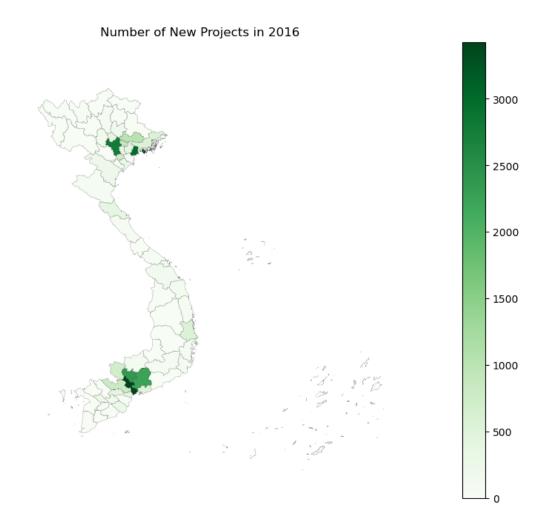
```
112
                                                              4
       432
                                                              0
       254
                                                            751
       277
                                                              6
       181
                                                              2
       396
                                                              2
       101
                                                              2
           Value of capital contribution, share purchase\n(million USD) Year
       217
                                                           4.92
                                                                            2019
       171
                                                          18.71
                                                                            2018
       5
                                                           1.46
                                                                            2016
       112
                                                           7.73
                                                                            2017
       432
                                                              0
                                                                            2022
       254
                                                        1611.82
                                                                            2020
       277
                                                           3.69
                                                                            2020
       181
                                                           0.84
                                                                            2018
       396
                                                           1.13
                                                                            2022
       101
                                                           1.75
                                                                            2017
[187]: ## Data consistency
       cols_to_convert = n_df.columns.drop(['Provinces', 'Year'])
       n_df[cols_to_convert] = n_df[cols_to_convert].apply(pd.to_numeric,__
        →errors='coerce')
       ## Check for missing value
       print(n_df.isnull().values.any())
       print(n_df.isna().sum())
      False
      Provinces
                                                                         0
      Number of new projects
                                                                         0
      Newly registered capital (million USD)
                                                                         0
      Adjusted project number
                                                                         0
      Adjusted capital (million USD)
                                                                         0
      Number of times of capital contribution to buy shares
                                                                         0
      Value of capital contribution, share purchase\n(million USD)
                                                                         0
      Year
                                                                         0
      dtype: int64
      0.4 Visualization
[188]: # Create a new column for the total FDI
       n_df['Total FDI'] = n_df['Newly registered capital (million USD)'] + 
        \rightarrow n_df['Adjusted capital (million USD)'] + n_df['Value of capital_
        ⇔contribution, share purchase\n(million USD)']
       # Merge data with geodata
       fullData = data.merge(
```

```
n_df,
left_on=['Name'], # identifier from geodataframe
right_on=['Provinces'] # identifier from dataframe
)
```

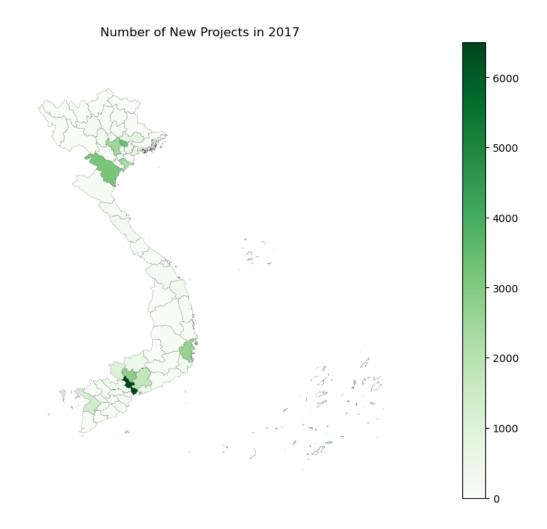
0.4.1 Plot map chart of dataset

```
[189]: years = fullData['Year'].unique()
       fullData['Total FDI'] = pd.to_numeric(fullData['Total FDI'], errors='coerce')
       for year in years:
           # Filter data for each year
           data_year = fullData[fullData['Year'] == year]
           # Plot with geoplot for each year
           plt.figure(figsize=(12, 8))
           geoplot.choropleth(
               data_year,
               projection=gcrs.AlbersEqualArea(),
               hue="Total FDI",
               cmap='Greens',
               linewidth=0.1,
               edgecolor='black',
               legend=True,
               figsize=(12, 8)
           plt.title(f"Number of New Projects in {year}")
           plt.show()
```

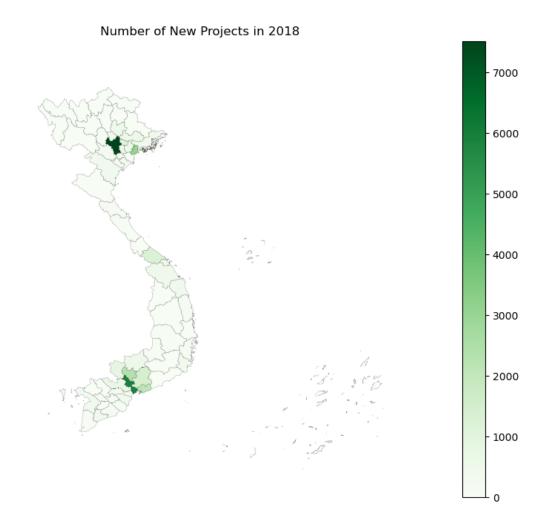
<Figure size 1200x800 with 0 Axes>



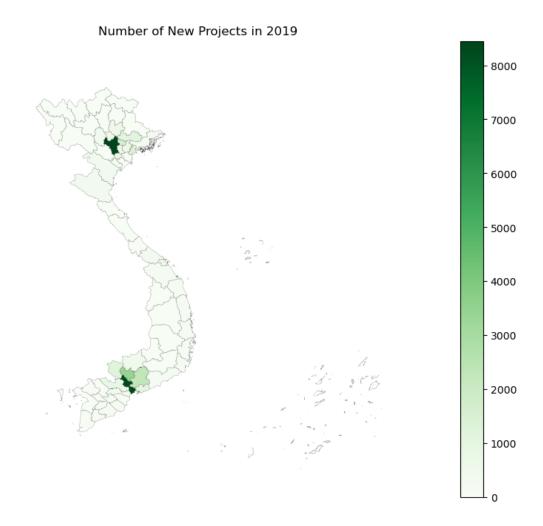
<Figure size 1200x800 with 0 Axes>



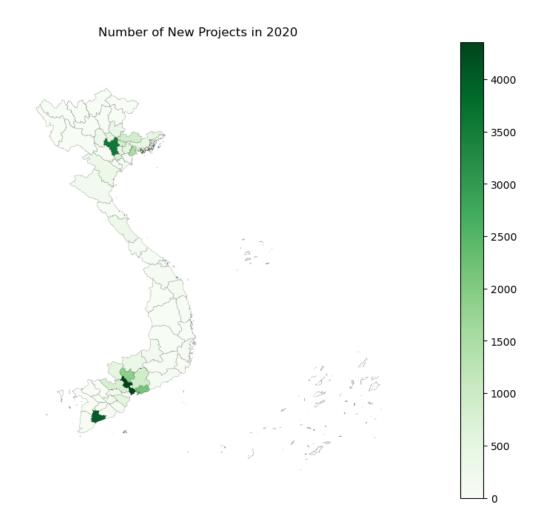
<Figure size 1200x800 with 0 Axes>



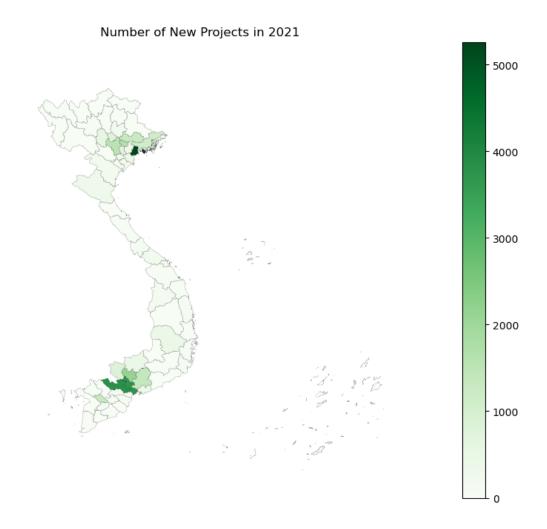
<Figure size 1200x800 with 0 Axes>



<Figure size 1200x800 with 0 Axes>

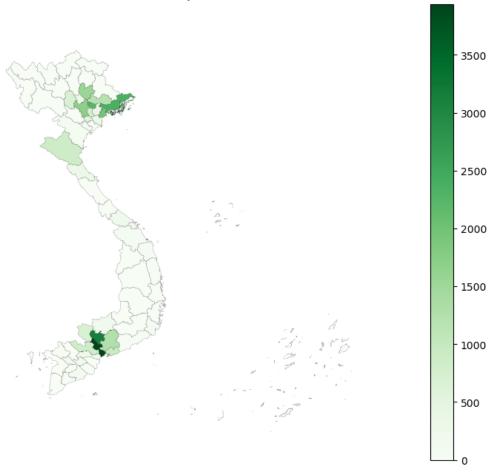


<Figure size 1200x800 with 0 Axes>



<Figure size 1200x800 with 0 Axes>

Number of New Projects in 2022

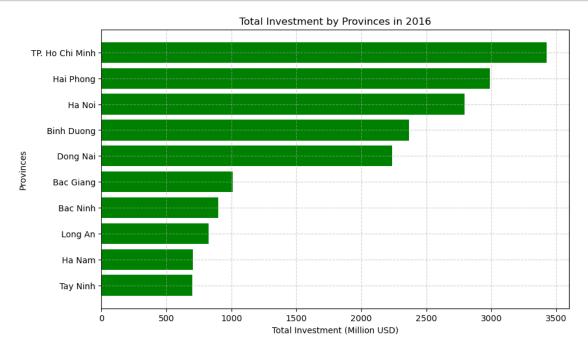


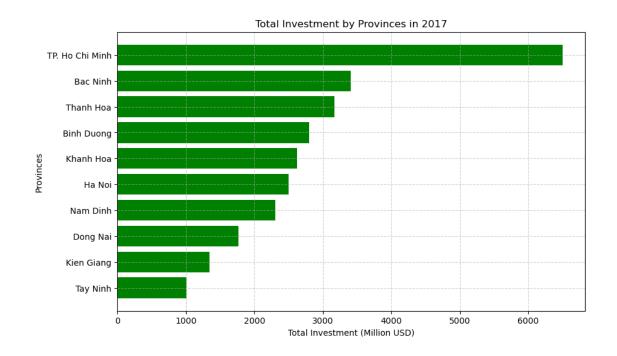
Note: From the chart above, we can see that foreign investment in Vietnam is concentrated in major cities such as Ho Chi Minh City, Hanoi, Hai Phong, etc. However, a positive sign is that there is also investment spread across various provinces.

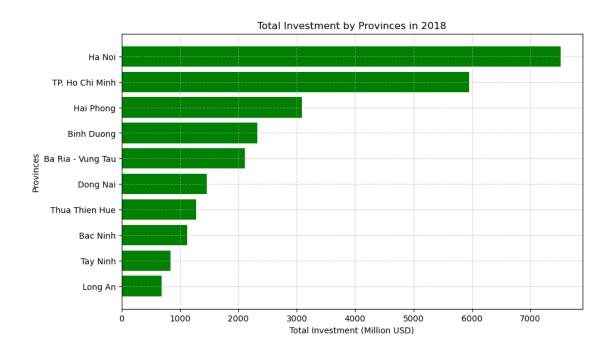
0.4.2 Detail with buble chart

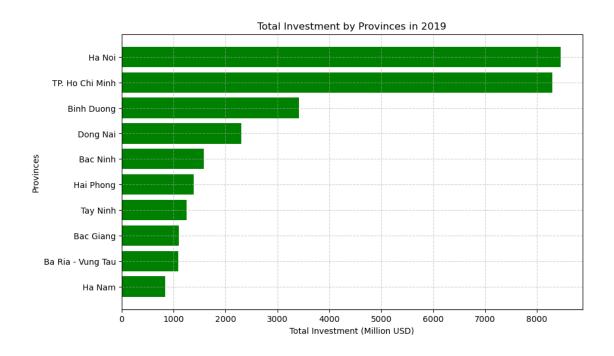
0.4.3 Top 10 provinces with the largest total FDI

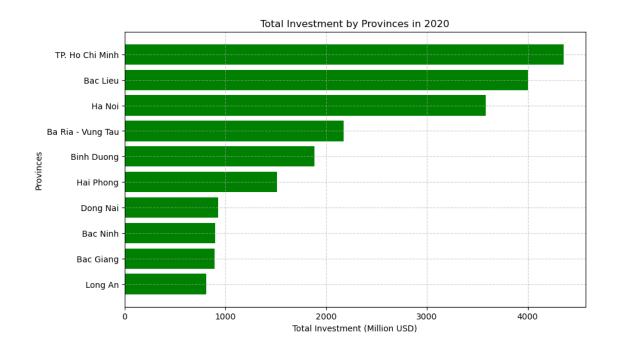
```
[191]: years = n_df['Year'].unique()
       # Loop through each year and plot the total investment by industry
       for year in years:
           # Filter data by year
           df_year = n_df[n_df['Year'] == year]
           # Sort the data by Total FDI
           df_year.sort_values('Total FDI', ascending=False, inplace=True)
           # Select the top 10 provinces with the highest Total FDI
           df_year_top10 = df_year.head(10)
           # Set axis values
           x = df_year_top10['Provinces'].values
           y = df_year_top10['Total FDI'].values
           # Plot
           plt.figure(figsize=(10, 6))
           plt.barh(x, y, color='Green')
           plt.xlabel('Total Investment (Million USD)')
           plt.ylabel('Provinces')
           plt.title(f'Total Investment by Provinces in {year}')
           plt.gca().invert_yaxis()
           plt.grid(True, linestyle='--', alpha=0.6)
           plt.show()
```

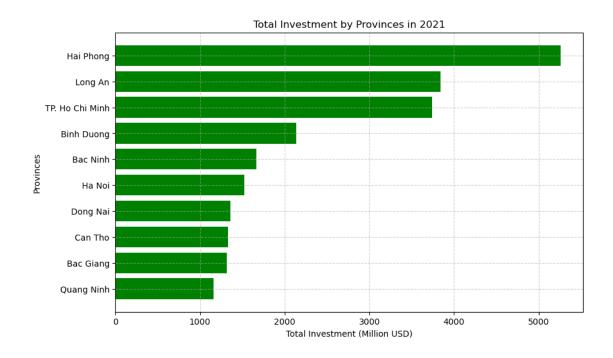


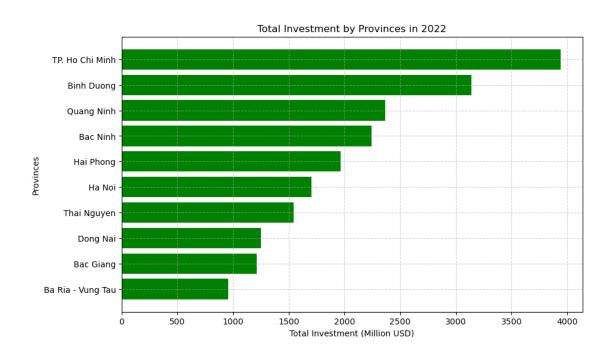












From the ranking plot and multiple barh , we can have below observations: - There is still volatility and changes in the rankings: this indicates the diversity of investment sectors, opportunities, and potential existing across different provinces in Vietnam. - There are new names appearing in some years: this suggests that the development is on a significant upward trend and continues to attract investors.