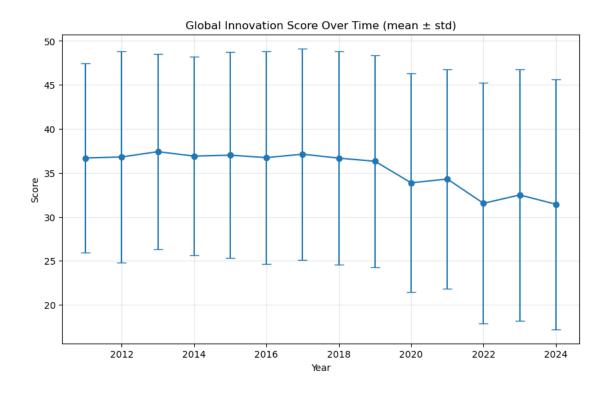
## Global Innovation Index (GII) Day 3 EDA

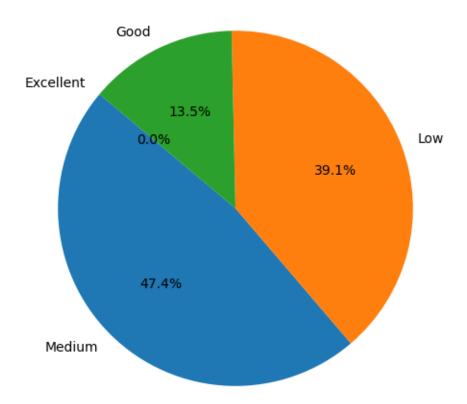
June 21, 2025

```
[1]: import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
[45]: df = pd.read_csv(r'../data/GII_2011_2024_long_format.tsv', sep='\t')
      print(df.head())
      print(df.info())
      df_clean = df[(df['Rank'] != -1) & (df['Score'] != -1)].copy()
      print(f"Number of countries: {df_clean['Country'].nunique()}, number of records_
       →after cleaning: {len(df_clean)}")
       Country
               Year Rank Score
     0
           AGO
                2011
                       -1 -1.00
               2012
                      135 22.20
     1
           AGO
     2
           AGO
                       135 23.46
                2013
     3
           AGO 2014
                       135 23.82
                2015
                       120 26.20
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 2086 entries, 0 to 2085
     Data columns (total 4 columns):
                  Non-Null Count Dtype
          Column
         _____
                   -----
          Country 2086 non-null
                                  object
                   2086 non-null
                                  int64
      1
          Year
      2
          Rank
                   2086 non-null
                                  int64
          Score
                   2086 non-null
                                  float64
     dtypes: float64(1), int64(2), object(1)
     memory usage: 65.3+ KB
     None
     Number of countries: 149, number of records after cleaning: 1862
```

```
[31]: # ... (code to load and clean the data)
      df_cleaned = df.dropna(subset=['Rank', 'Score']).copy()
[47]: | # === Add logical classification (Region, Performance Category) ===
      region_mapping = {
          'IRL': 'Europe', 'GBR': 'Europe', 'DEU': 'Europe', 'FRA': 'Europe',
          'USA': 'North America', 'CAN': 'North America', 'MEX': 'North America',
          'CHN': 'Asia', 'JPN': 'Asia', 'KOR': 'Asia', 'SGP': 'Asia',
          'BRA': 'South America', 'ARG': 'South America', 'CHL': 'South America',
          'EGY': 'Africa', 'ZAF': 'Africa', 'KEN': 'Africa', 'NGA': 'Africa'
      df_clean['Region'] = df_clean['Country'].map(region_mapping).fillna('Other')
      df_clean['Performance_Category'] = pd.cut(
          df_clean['Score'], bins=[0, 25, 50, 75, 100], labels=['Low', 'Medium', __
      )
[49]: # === Global trend analysis ===
      plt.figure(figsize=(10, 6))
      yearly_avg = df_clean.groupby('Year')['Score'].agg(['mean', 'std'])
      plt.errorbar(yearly_avg.index, yearly_avg['mean'], yerr=yearly_avg['std'],_u
      →marker='o', capsize=5)
      plt.title('Global Innovation Score Over Time (mean ± std)')
      plt.xlabel('Year')
      plt.ylabel('Score')
      plt.grid(alpha=0.3)
      plt.show()
```

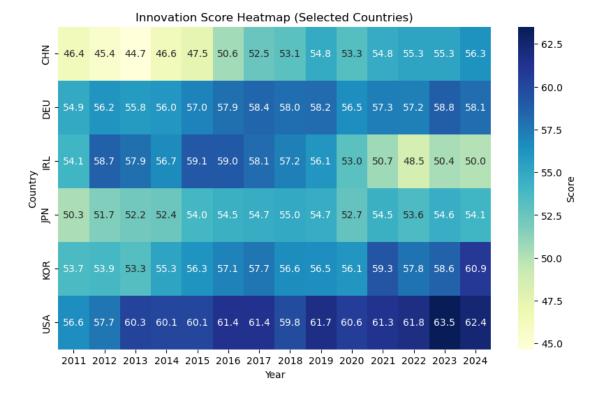


## Performance Categories - 2024

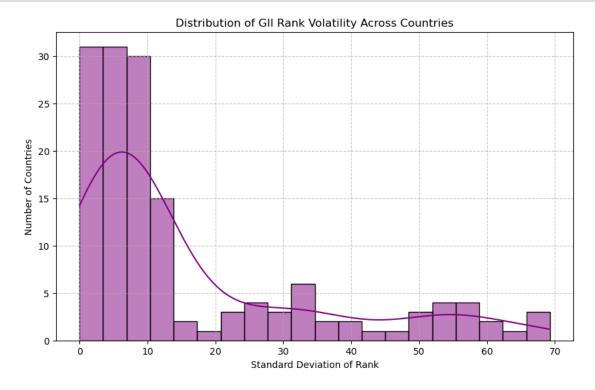


```
[69]: | # === 10 Top and bottom countries 2024 ===
      top_2024 = df_clean[df_clean['Year'] == 2024].nsmallest(10, 'Rank')
      bottom_2024 = df_clean[df_clean['Year'] == 2024].nlargest(10, 'Rank')
      print(" Top 10 Countries 2024:\n", top_2024[['Country', 'Rank', 'Score']])
      print("\n Bottom 10 Countries 2024:\n", bottom_2024[['Country', 'Rank', _
       Top 10 Countries 2024:
           Country Rank Score
                          67.5
     363
              CHE
                      1
     1791
              SWE
                      2
                          64.5
     1987
              USA
                      3
                         62.4
     377
              CHL
                      4
                         32.6
              SGP
                      4
                         61.2
     1721
     685
                         61.0
              GBR
                      5
     1063
              KOR
                      6
                         60.9
     629
                      7
                          59.4
              FIN
     1455
              NLD
                      8
                          58.8
     503
              DEU
                          58.1
      Bottom 10 Countries 2024:
           Country Rank Score
                          10.2
     13
              AGO
                    133
                          11.2
     1413
              NER
                    132
                         11.8
     1259
              MLI
                    131
     615
              ETH
                    130
                         12.3
                         12.8
     167
              BFA
                    129
     1329
              MOZ
                    128
                         13.1
     125
              BDI
                    127
                         13.2
                    126
                         13.2
     1343
              MRT
     1287
              MMR
                    125
                          13.8
                          14.0
     1441
              NIC
                    124
[71]: # === Volatility analysis ===
      volatility = df_clean.groupby('Country').agg({
          'Score': ['std', 'mean'], 'Rank': ['std', 'mean'], 'Year': 'count'})
      volatility.columns = ['Score_Std', 'Score_Mean', 'Rank_Std', 'Rank_Mean', '
      volatility = volatility[volatility['Years_Count'] >= 5]
      volatility = volatility.sort_values('Score_Std', ascending=False)
      print("\n Most volatile countries (Score):\n", volatility.head())
```

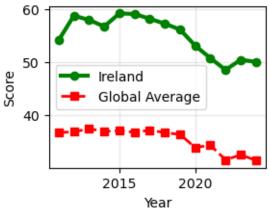
```
Most volatile countries (Score):
               Score_Std Score_Mean
                                                  Rank_Mean Years_Count
                                       Rank_Std
     Country
     AGO
               6.484687
                          18.135000
                                      5.221863 131.125000
                                                                       8
               5.976184
     MOZ
                          22.446923 13.022170
                                                113.923077
                                                                      13
     MLI
               5.623512
                          21.949286
                                      8.562954
                                                117.357143
                                                                      14
     UGA
               5.483335
                          23.867143 10.875924
                                                108.142857
                                                                      14
     GTM
               5.450769
                          24.943571 10.807252
                                                102.214286
                                                                      14
[83]: # === Heatmap of selected countries ===
```



```
[37]: # Visualize distribution of rank volatility
plt.figure(figsize=(10, 6))
sns.histplot(rank_stability, bins=20, kde=True, color='purple')
plt.title('Distribution of GII Rank Volatility Across Countries')
plt.xlabel('Standard Deviation of Rank')
plt.ylabel('Number of Countries')
plt.grid(True, linestyle='--', alpha=0.7)
plt.show()
```



## Ireland Performance vs Global Average



```
[33]: # Distribution of Scores for a recent year (e.g., 2024)
year_to_examine = 2024
df_year = df_cleaned[df_cleaned['Year'] == year_to_examine]['Score'].dropna()

if not df_year.empty:
    plt.figure(figsize=(10, 6))
    sns.histplot(df_year, kde=True, bins=15, color='orange')
    plt.title(f'Distribution of GII Scores in {year_to_examine}')
    plt.xlabel('GII Score')
    plt.ylabel('Number of Countries')
    plt.grid(axis='y', linestyle='--', alpha=0.7)
    plt.show()
else:
    print(f"No valid score data for year {year_to_examine}.")
```

