

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

df = pd.read_csv("/content/sustainable_development_report_2023.csv")
print(df.head())
```

```
country_code country region overall_score goal_1_score goal_2_score \
0 FIN Finland OECD 86.760595 99.5750 60.886750
1 SWE Sweden OECD 85.981397 98.8885 63.074125
2 DNK Denmark OECD 85.683637 99.2155 71.025250
3 DEU Germany OECD 83.358447 99.5105 72.366000
4 AUT Austria OECD 82.280189 99.4510 73.067500

goal_3_score goal_4_score goal_5_score goal_6_score ... goal_8_score \
0 95.386385 97.169333 92.11125 94.3276 ... 86.789000
1 96.904000 99.761667 91.44025 95.0576 ... 84.966429
2 95.398500 99.339667 86.99800 90.7316 ... 87.562429
3 93.039357 97.162667 81.92025 88.4434 ... 86.967286
4 92.468000 97.914333 84.57925 92.1636 ... 83.274143

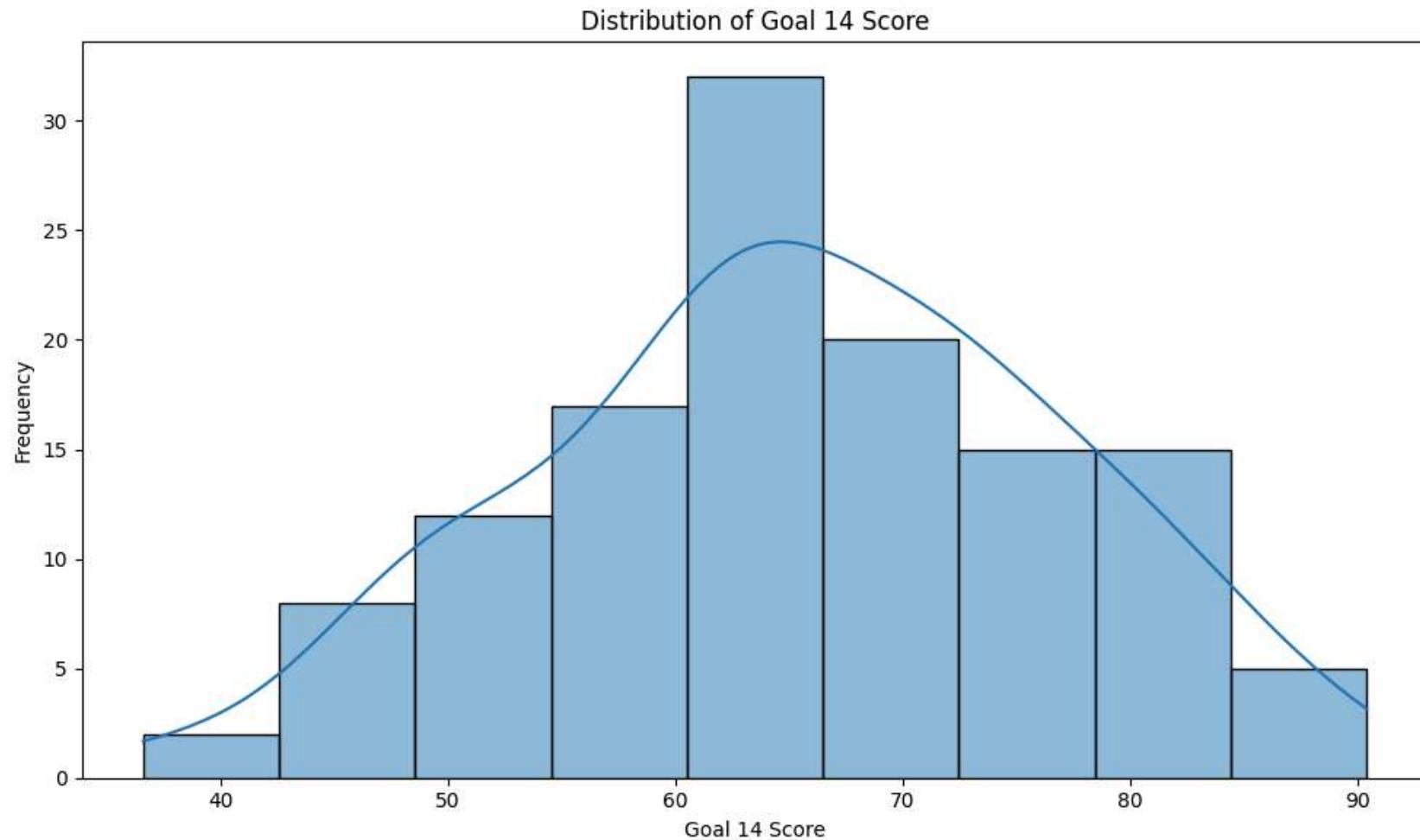
goal_9_score goal_10_score goal_11_score goal_12_score goal_13_score \
0 95.995714 98.4685 91.233750 60.059571 68.793667
1 97.586286 94.9650 90.389250 56.830571 70.031000
2 96.984857 98.1560 93.038500 44.571714 60.780667
3 95.788429 88.1470 90.096500 55.412857 64.002000
4 96.982143 94.6345 92.473667 49.623286 57.332000

goal_14_score goal_15_score goal_16_score goal_17_score
0 87.928000 85.0700 92.521091 75.60100
1 69.348667 80.1882 88.508455 85.77025
2 76.303333 92.7924 93.844909 82.14800
3 73.996000 79.2318 89.457545 84.39025
4 NaN 73.5836 87.911455 71.13025
```

[5 rows x 21 columns]

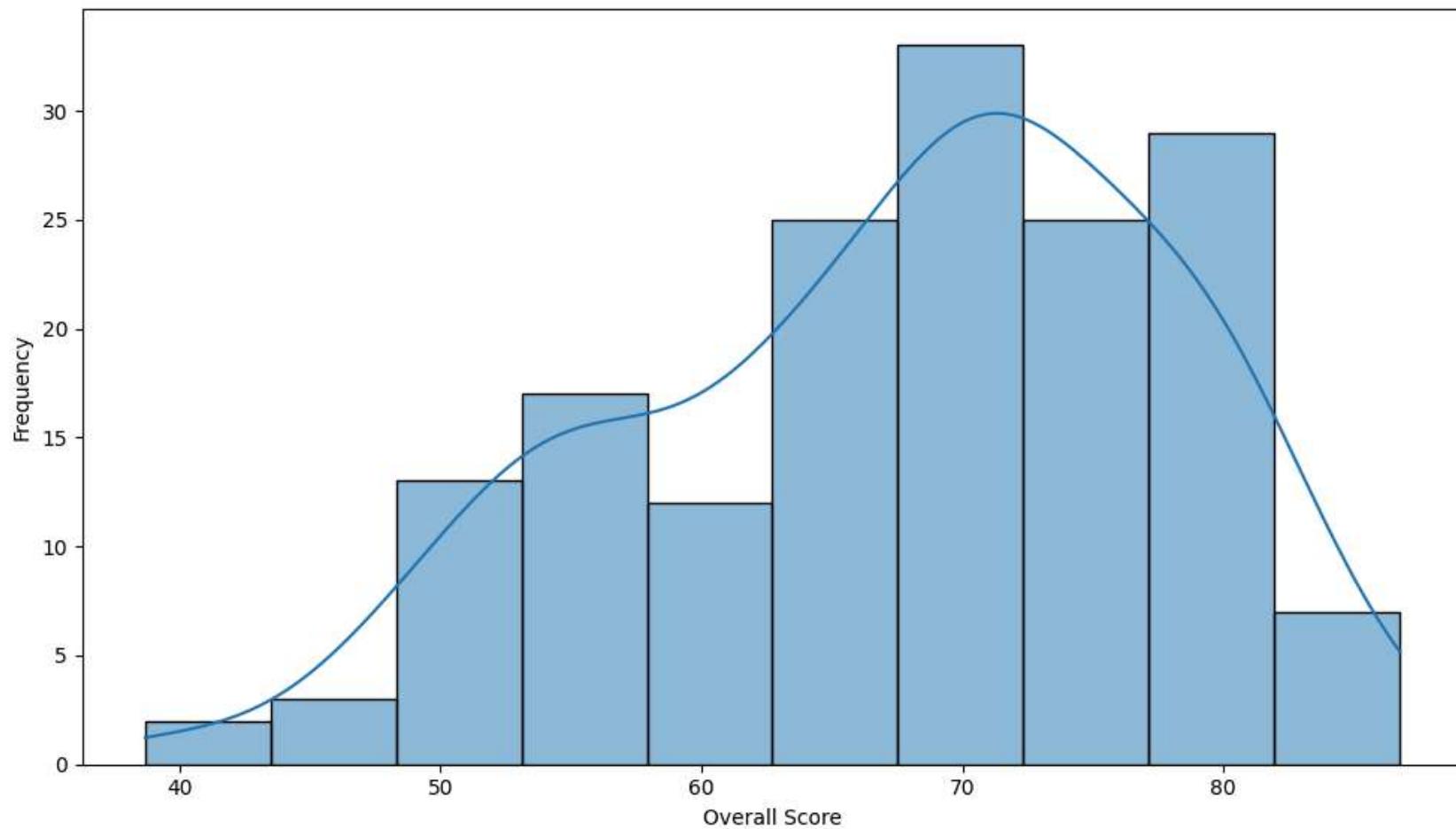
```
# Visualize the distribution of goal_14_score using Seaborn
plt.figure(figsize=(10, 6))
sns.histplot(data=df, x="goal_14_score", kde=True)
plt.title("Distribution of Goal 14 Score")
plt.xlabel("Goal 14 Score")
plt.ylabel("Frequency")
```

```
plt.tight_layout()  
plt.show()
```



```
# Visualize the distribution of overall_score using Seaborn  
plt.figure(figsize=(10, 6))  
sns.histplot(data=df, x="overall_score", kde=True)  
plt.title("Distribution of Overall Score")  
plt.xlabel("Overall Score")  
plt.ylabel("Frequency")  
plt.tight_layout()  
plt.show()
```

Distribution of Overall Score



```
# Check for missing values in the dataset
print("\nMissing values per column:")
print(df.isnull().sum())
```

```
Missing values per column:
country_code      0
country          0
region           0
overall_score    0
goal_1_score     15
goal_2_score     0
goal_3_score     0
goal_4_score     0
```

```
goal_5_score      0  
goal_6_score      0  
goal_7_score      0  
goal_8_score      0  
goal_9_score      0  
goal_10_score     11  
goal_11_score     0  
goal_12_score     0  
goal_13_score     0  
goal_14_score     40  
goal_15_score     0  
goal_16_score     0  
goal_17_score     0  
dtype: int64
```

```
# Show summary statistics for df
print("\nSummary Statistics for the DataFrame:")
display(df.describe())
```

Summary Statistics for the DataFrame:

	overall_score	goal_1_score	goal_2_score	goal_3_score	goal_4_score	goal_5_score	goal_6_score	goal_7_score	goal_8_score	gc
count	166.000000	151.000000	166.000000	166.000000	166.000000	166.000000	166.000000	166.000000	166.000000	166.000000
mean	67.549197	75.234401	59.799100	69.694078	76.512968	63.285420	66.710744	61.413598	71.952935	
std	10.295499	31.169948	10.620853	20.354575	23.181919	16.399691	14.091641	20.364351	10.592308	
min	38.676086	0.000000	19.805800	12.952714	1.232250	13.054750	32.600000	8.697000	39.535000	
25%	60.547488	55.779250	54.007188	51.860089	61.417938	51.046250	55.237250	47.521312	66.426857	
50%	69.376528	93.300500	61.027500	75.437629	84.772875	65.869875	67.878000	68.612750	73.157643	
75%	74.947511	98.950750	67.264335	85.524428	95.644063	76.137000	76.044200	74.364000	79.626036	
max	86.760595	100.000000	83.401125	97.115143	99.761667	94.021667	95.057600	99.550750	93.382750	

```
# Get the OECD average scores
oecd_average_scores = average_score_by_region[average_score_by_region["region"] == "OECD"]

# Transpose the Sub-Saharan Africa average scores for easier comparison
sub_saharan_africa_avg_transposed = average_scores_sub_saharan_africa.to_frame(name="Sub-Saharan Africa")

# Select only the goal columns from the transposed Sub-Saharan Africa data
sub_saharan_africa_avg_goals = sub_saharan_africa_avg_transposed[sub_saharan_africa_avg_transposed.index.str.startswith("goal ")]
```

```
# Transpose the OECD average scores for easier comparison
oecd_average_scores_transposed = oecd_average_scores.set_index("region").T

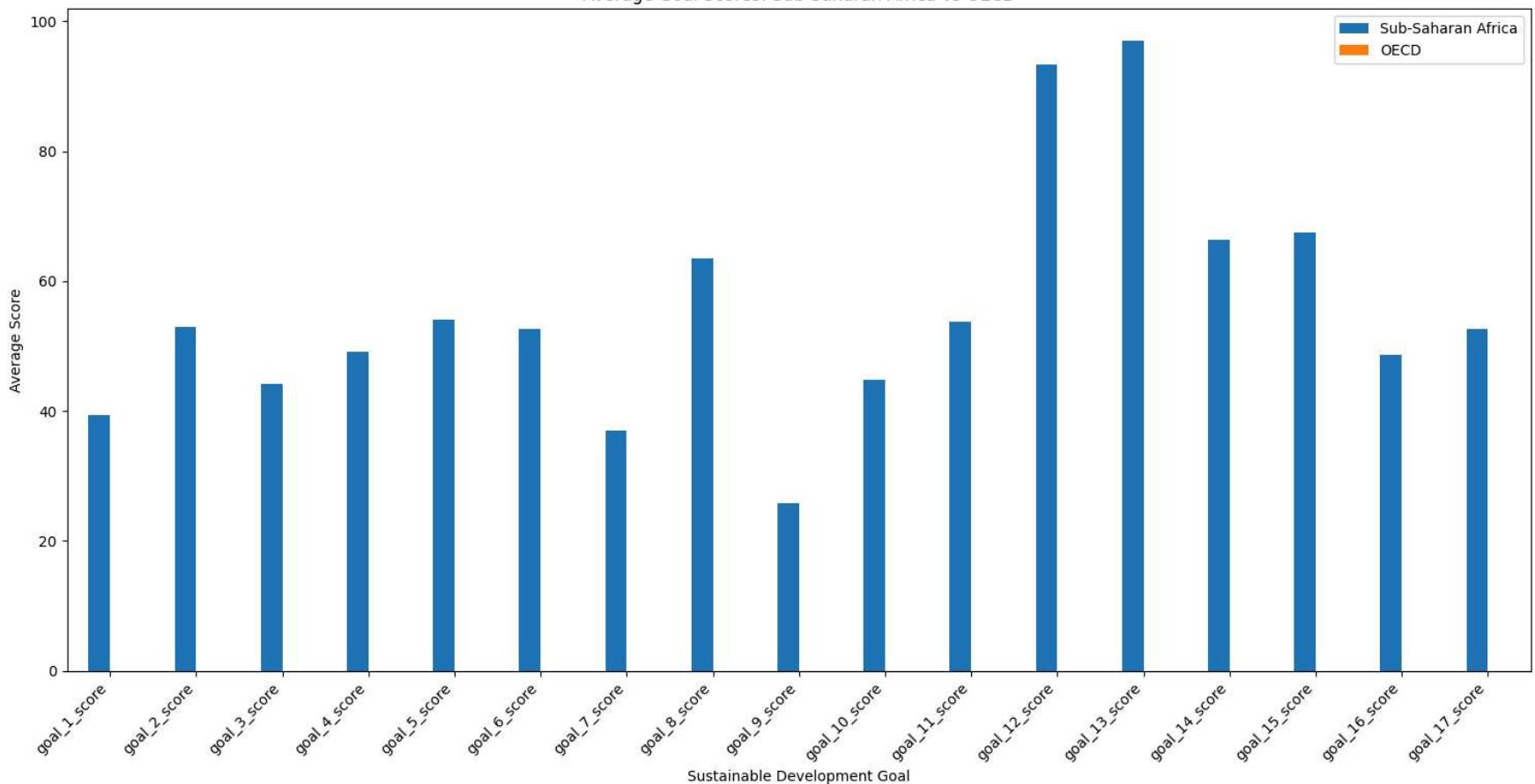
# Select only the goal columns from the transposed OECD data
oecd_average_goals = oecd_average_scores_transposed[oecd_average_scores_transposed.index.str.startswith("goal_")]

# Combine the two transposed DataFrames
comparison_df = pd.concat([sub_saharan_africa_avg_goals, oecd_average_goals], axis=1)

# Plotting the comparison
plt.figure(figsize=(15, 8))
comparison_df.plot(kind='bar', figsize=(15, 8))
plt.title('Average Goal Scores: Sub-Saharan Africa vs OECD')
plt.xlabel('Sustainable Development Goal')
plt.ylabel('Average Score')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```

<Figure size 1500x800 with 0 Axes>

Average Goal Scores: Sub-Saharan Africa vs OECD



```
# Filter data for the Sub-Saharan Africa region
sub_saharan_africa_data = df[df["region"] == "Sub-Saharan Africa"]

# Calculate the average scores for each goal in the Sub-Saharan Africa region
average_scores_sub_saharan_africa = sub_saharan_africa_data.mean(numeric_only=True)

print("\nAverage Scores for Sub-Saharan Africa Region Across All Goals:")
print(average_scores_sub_saharan_africa)
```

Average Scores for Sub-Saharan Africa Region Across All Goals:

overall_score	55.461529
goal_1_score	39.445256

```
goal_2_score      52.988240
goal_3_score      44.102043
goal_4_score      49.195380
goal_5_score      54.055420
goal_6_score      52.640587
goal_7_score      36.977539
goal_8_score      63.414371
goal_9_score      25.760039
goal_10_score     44.866956
goal_11_score     53.809474
goal_12_score     93.407992
goal_13_score     97.047867
goal_14_score     66.334599
goal_15_score     67.490306
goal_16_score     48.723877
goal_17_score     52.586052
dtype: float64
```

```
# Select relevant columns for the top 5 countries (excluding country code, country, and region)
top_5_countries_scores = sorted_data.head(5).drop(columns=['country_code', 'country', 'region'])

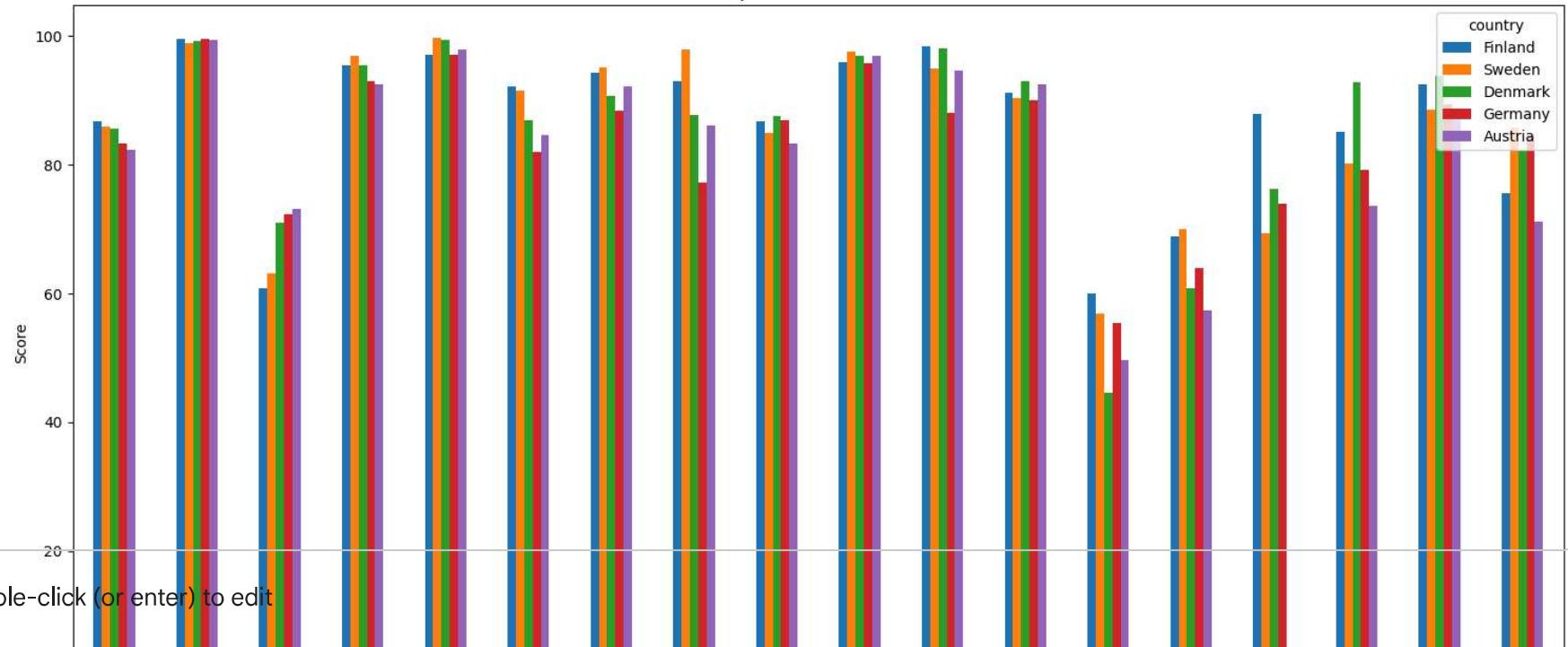
# Transpose the DataFrame for easier plotting
top_5_countries_scores_transposed = top_5_countries_scores.T

# Rename the columns with country names
top_5_countries_scores_transposed.columns = sorted_data.head(5)['country']

# Plotting the scores
plt.figure(figsize=(15, 8))
top_5_countries_scores_transposed.plot(kind='bar', figsize=(15, 8))
plt.title('Scores of Top 5 Countries Across All Goals')
plt.xlabel('Sustainable Development Goal')
plt.ylabel('Score')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```

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Scores of Top 5 Countries Across All Goals



```
# Display the top 5 countries by overall score
print("\nTop 5 Countries by Overall Score:")
print(sorted_data.head())
```

Top 5 Countries by Overall Score:						
	country_code	country	region	overall_score	goal_1_score	goal_2_score
0	FIN	Finland	OECD	86.760595	99.5750	60.886750
1	SWE	Sweden	OECD	85.981397	98.8885	63.074125
2	DNK	Denmark	OECD	85.683637	99.2155	71.025250
3	DEU	Germany	OECD	83.358447	99.5105	72.366000
4	AUT	Austria	OECD	82.280189	99.4510	73.067500

	goal_3_score	goal_4_score	goal_5_score	goal_6_score	...	goal_8_score
0	95.386385	97.169333	92.11125	94.3276	...	86.789000
1	96.904000	99.761667	91.44025	95.0576	...	84.966429
2	95.398500	99.339667	86.99800	90.7316	...	87.562429
3	93.039357	97.162667	81.92025	88.4434	...	86.967286
4	92.468000	97.914333	84.57925	92.1636	...	83.274143

	goal_9_score	goal_10_score	goal_11_score	goal_12_score	goal_13_score
0	95.995714	98.4685	91.233750	60.059571	68.793667

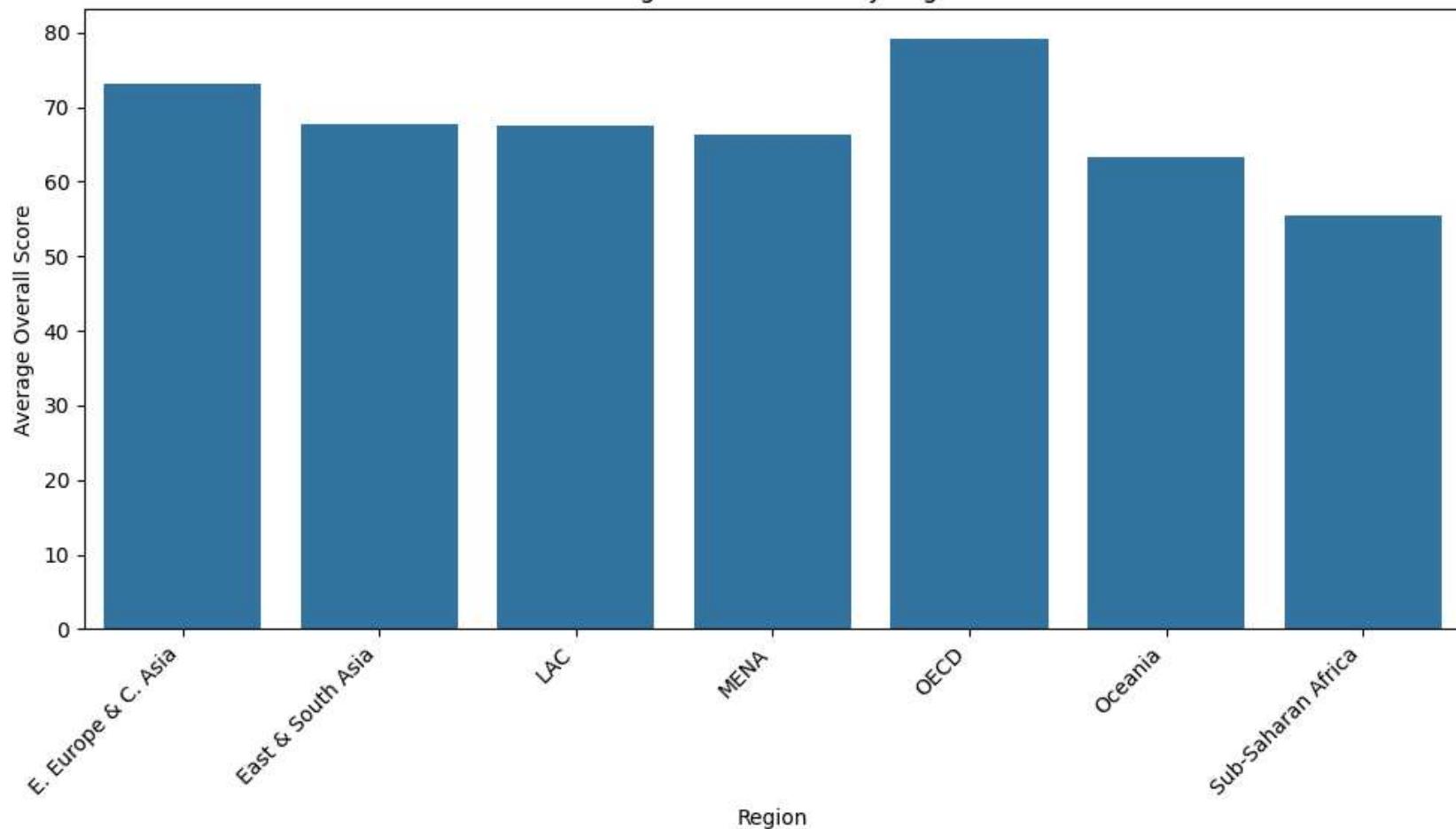
```
1 97.586286    94.9650    90.389250    56.830571    70.031000
2 96.984857    98.1560    93.038500    44.571714    60.780667
3 95.788429    88.1470    90.096500    55.412857    64.002000
4 96.982143    94.6345    92.473667    49.623286    57.332000
```

```
goal_14_score  goal_15_score  goal_16_score  goal_17_score
0      87.928000      85.0700      92.521091      75.60100
1      69.348667      80.1882      88.508455      85.77025
2      76.303333      92.7924      93.844909      82.14800
3      73.996000      79.2318      89.457545      84.39025
4           NaN      73.5836      87.911455      71.13025
```

[5 rows x 21 columns]

```
# Plot average overall score by region
plt.figure(figsize=(10, 6))
sns.barplot(x="region", y="overall_score", data=average_score_by_region)
plt.title("Average Overall Score by Region")
plt.xlabel("Region")
plt.ylabel("Average Overall Score")
plt.xticks(rotation=45, ha="right")
plt.tight_layout()
plt.show()
```

Average Overall Score by Region



```
# Analyze average 'overall_score' by region
average_score_by_region = df.groupby("region")["overall_score"].mean().reset_index()
print("\nAverage Overall Score by Region:")
print(average_score_by_region)
```

Average Overall Score by Region:

region	overall_score
0 E. Europe & C. Asia	73.149119
1 East & South Asia	67.655676
2 LAC	67.598713
3 MENA	66.283838
4 OECD	79.150447