

```
In [124... # importing Libraries
import matplotlib.pyplot as plt
import plotly.express as px
import seaborn as sns
import pandas as pd
import numpy as np
import sqlite3
```

```
In [126... # Define the file path for the excel
file_path = r"D:\toki\testings\python\new\fashion(p).xlsx"
df = pd.read_excel(file_path)
```

```
In [127... # Checking if the data frame works
print(df.head())
print(df.info())
```

| | Brand_ID | Brand_Name | Country | Year | Sustainability_Rating | \ |
|---|------------|------------|-----------|------|-----------------------|---|
| 0 | BRAND-0001 | Brand_1 | Australia | 2018 | | D |
| 1 | BRAND-0001 | Brand_1 | Australia | 2018 | | D |
| 2 | BRAND-0002 | Brand_2 | Japan | 2015 | | D |
| 3 | BRAND-0003 | Brand_3 | USA | 2024 | | A |
| 4 | BRAND-0004 | Brand_4 | Italy | 2023 | | D |

| | Material_Type | Eco_Friendly_Manufacturing | Carbon_Footprint_MT | \ |
|---|---------------|----------------------------|---------------------|---|
| 0 | Tencel | No | 1.75 | |
| 1 | Tencel | No | 1.75 | |
| 2 | Vegan Leather | Yes | 124.39 | |
| 3 | Vegan Leather | No | 336.66 | |
| 4 | Bamboo Fabric | No | 152.04 | |

| | Water_Usage_Liters | Waste_Production_KG | Recycling_Programs | Product_Lines | \ |
|---|--------------------|---------------------|--------------------|---------------|---|
| 0 | 4511152.79 | 97844.11 | No | 2 | |
| 1 | 4511152.79 | 97844.11 | No | 2 | |
| 2 | 1951566.31 | 37267.75 | No | 15 | |
| 3 | 467454.52 | NaN | No | 2 | |
| 4 | 899576.90 | 32665.45 | No | 13 | |

| | Average_Price_USD | Market_Trend | Certifications |
|---|-------------------|--------------|----------------|
| 0 | 38.33 | Growing | GOTS |
| 1 | 38.33 | Growing | GOTS |
| 2 | 250.07 | Growing | GOTS |
| 3 | 146.16 | Growing | B Corp |
| 4 | 165.52 | Stable | OEKO-TEX |

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5001 entries, 0 to 5000
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Brand_ID                             5000 non-null   object
1   Brand_Name                           5000 non-null   object
2   Country                              5001 non-null   object
3   Year                                 5001 non-null   int64
4   Sustainability_Rating                 5001 non-null   object
5   Material_Type                         5001 non-null   object
6   Eco_Friendly_Manufacturing            5001 non-null   object
7   Carbon_Footprint_MT                  4998 non-null   float64
8   Water_Usage_Liters                   4999 non-null   float64
9   Waste_Production_KG                  5000 non-null   float64
10  Recycling_Programs                    5001 non-null   object
11  Product_Lines                         5001 non-null   int64
12  Average_Price_USD                     4998 non-null   float64
13  Market_Trend                         5001 non-null   object
14  Certifications                        3997 non-null   object
dtypes: float64(4), int64(2), object(9)
memory usage: 586.2+ KB
None

```

In [128...

```

# Checking for missing values
print("Missing values in each column:\n", df.isnull().sum())

```

Missing values in each column:

| | |
|----------------------------|------|
| Brand_ID | 1 |
| Brand_Name | 1 |
| Country | 0 |
| Year | 0 |
| Sustainability_Rating | 0 |
| Material_Type | 0 |
| Eco_Friendly_Manufacturing | 0 |
| Carbon_Footprint_MT | 3 |
| Water_Usage_Liters | 2 |
| Waste_Production_KG | 1 |
| Recycling_Programs | 0 |
| Product_Lines | 0 |
| Average_Price_USD | 3 |
| Market_Trend | 0 |
| Certifications | 1004 |

dtype: int64

```
In [129... # Removing duplicates
df.drop_duplicates(inplace=True)
```

```
In [130... # Removing extra column
df.drop(columns=['Brand_Name'], inplace=True)
```

```
In [131... # Filling in missing values with error handling
columns_to_fill = {
    'Carbon_Footprint_MT': df['Carbon_Footprint_MT'].mean(),
    'Waste_Production_KG': df['Waste_Production_KG'].median(),
    'Water_Usage_Liters': df['Water_Usage_Liters'].median(),
    'Average_Price_USD': df['Average_Price_USD'].mean(),
    'Certifications': "Not-Certified"
}

for column, value in columns_to_fill.items():
    if column in df.columns:
        df[column].fillna(value, inplace=True)
```

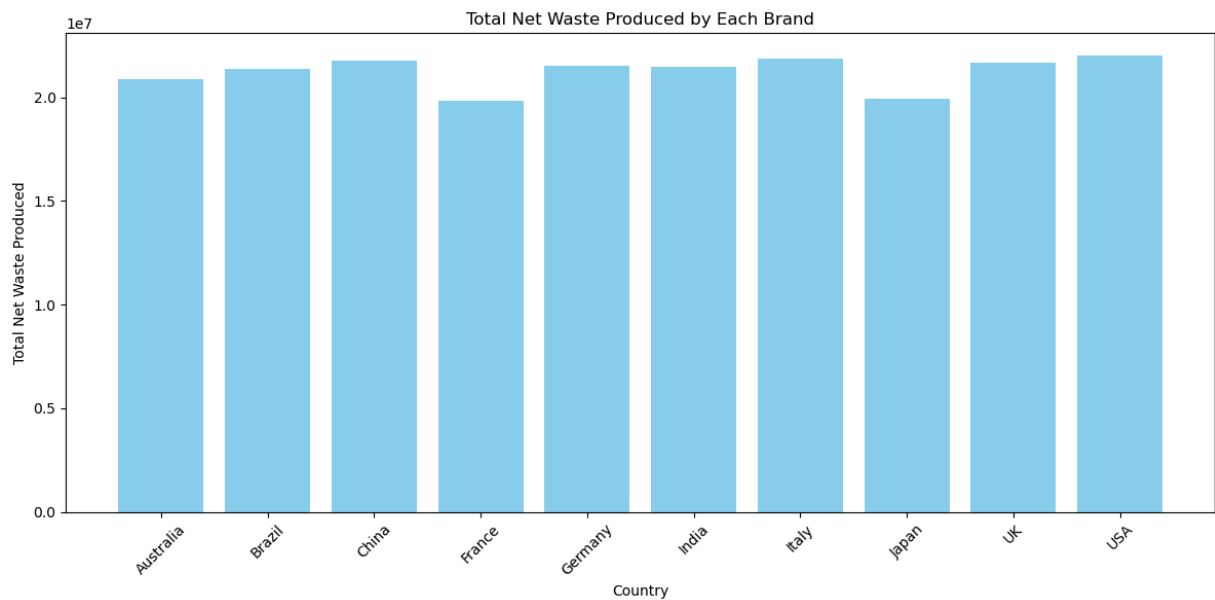
```
In [132... # Generating missing Brand_ID's
df['Brand_ID'] = df['Brand_ID'].fillna(df.apply(lambda row: f"BRAND-{str(row.name).",
```

```
In [133... # Feature engineering #1: Net waste produced
recycle_est = 0.3
df['RecPrgBIN'] = df['Recycling_Programs'].map({'Yes': 1, 'No': 0})
df['Net_WastePD'] = df['Waste_Production_KG'] * (1 - df['RecPrgBIN'] * recycle_est)
```

```
In [134... # Grouping Country with the Net Waste in a new variable
waste_summary = df.groupby('Country')['Net_WastePD'].sum().reset_index()
```

```
In [135... # Bar Chart
plt.figure(figsize=(12, 6))
plt.bar(waste_summary['Country'], waste_summary['Net_WastePD'], color='skyblue')
plt.xlabel('Country')
plt.ylabel('Total Net Waste Produced')
plt.title('Total Net Waste Produced by Each Brand')
```

```
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



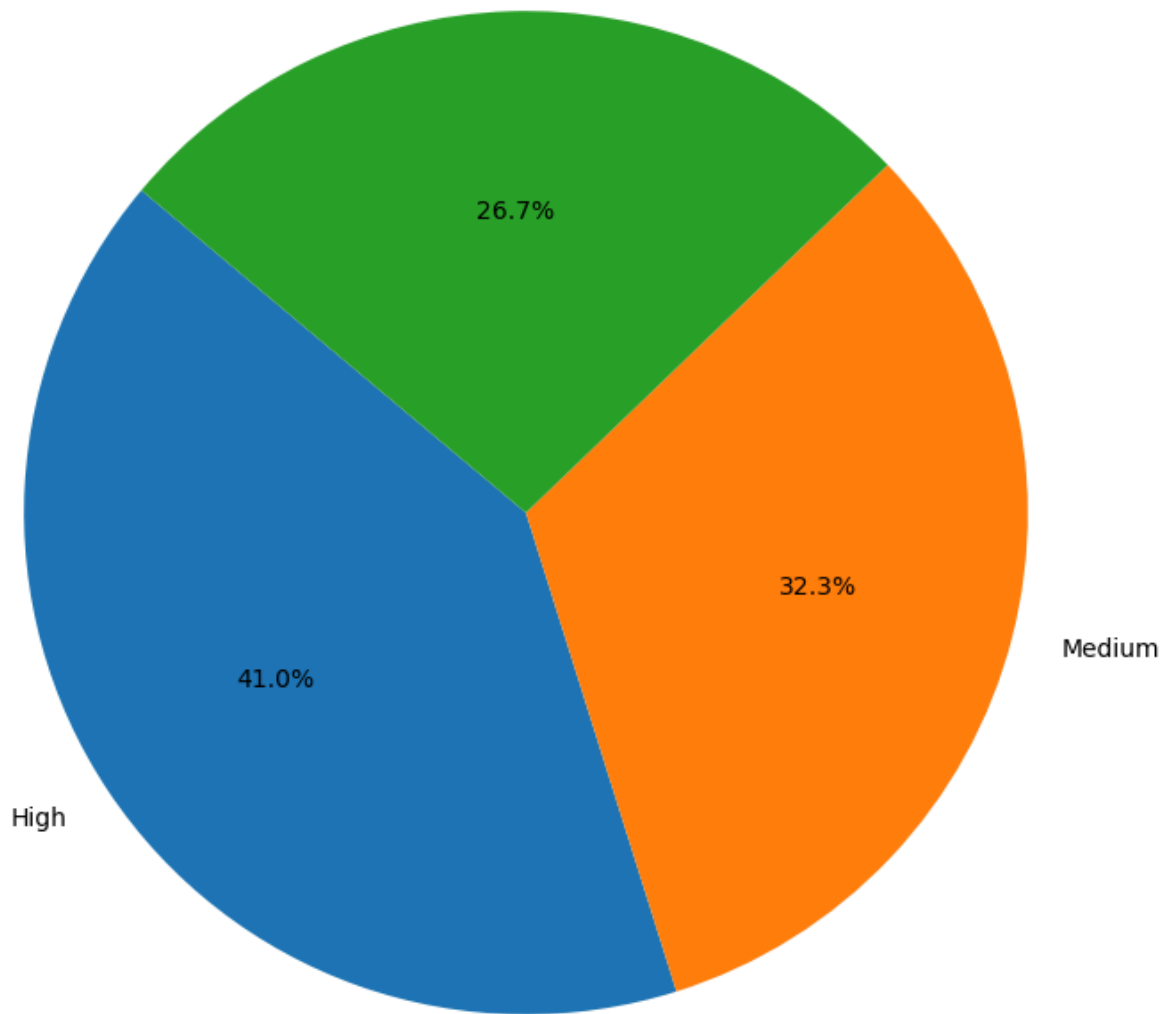
```
In [137... # Feature engineering #2: Price ranges
price_bins = [20, 150, 300, 500]
price_labels = ['Low', 'Medium', 'High']
df['Price_Range'] = pd.cut(df['Average_Price_USD'], bins=price_bins, labels=price_l
```

```
In [145... # Counting NO. of Brands in each Price Range
price_range_counts = df['Price_Range'].value_counts()
```

```
In [147... # Pie Chart

plt.figure(figsize=(8, 8))
plt.pie(price_range_counts, labels=price_range_counts.index, autopct='%1.1f%%', sta
plt.title('Distribution of Brands Across Price Ranges')
plt.axis('equal')
plt.show()
```

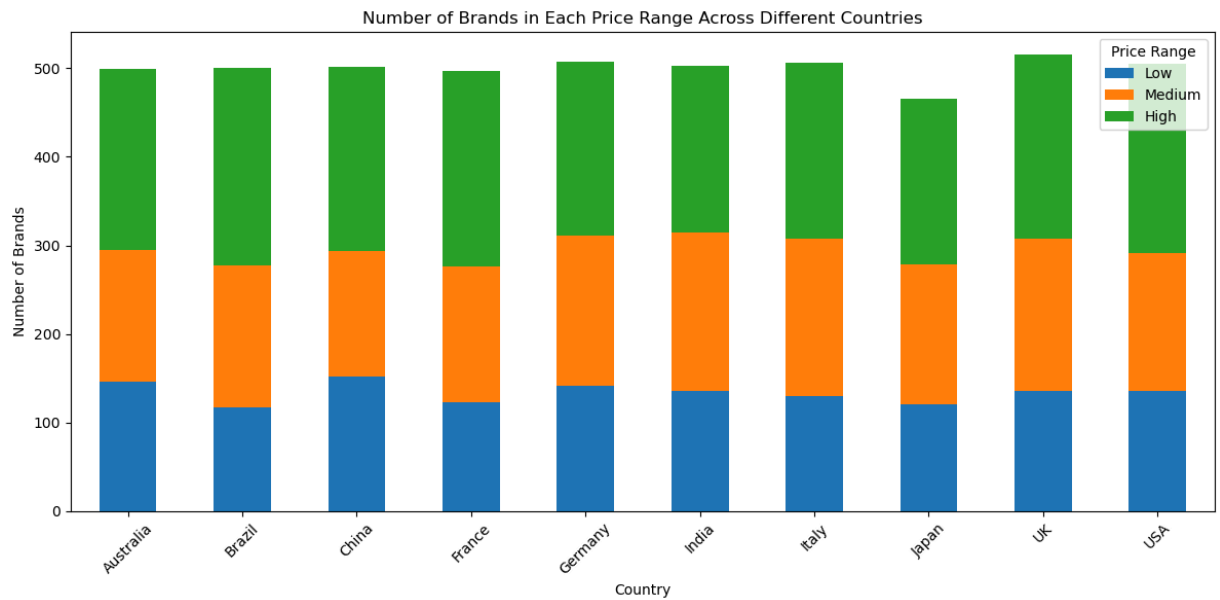
Distribution of Brands Across Price Ranges
Low



```
In [148... # Grouping by Country and Price Range
country_price_range_counts = df.groupby(['Country', 'Price_Range'], observed=False)
```

```
In [151... # Stacked Bar Chart

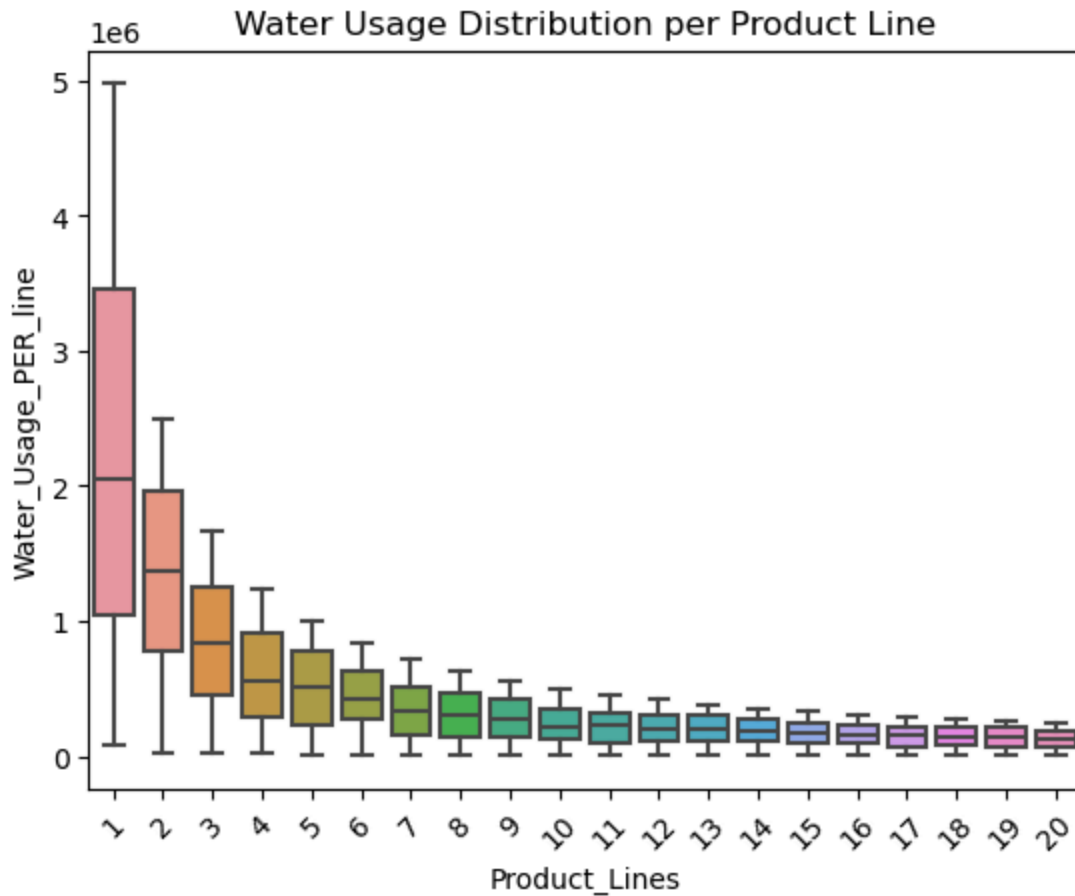
country_price_range_counts.plot(kind='bar', stacked=True, figsize=(12, 6))
plt.title('Number of Brands in Each Price Range Across Different Countries')
plt.xlabel('Country')
plt.ylabel('Number of Brands')
plt.xticks(rotation=45)
plt.legend(title='Price Range')
plt.tight_layout()
plt.show()
```



```
In [152... # Feature engineering #3: Water usage per product line & water usage per dollar
df['Water_Usage_PER_line'] = df['Water_Usage_Liters'] / df['Product_Lines']
df['Water_Usage_PER_dollar'] = df['Water_Usage_Liters'] / df['Average_Price_USD']
```

```
In [154... # Box Plot

sns.boxplot(x='Product_Lines', y='Water_Usage_PER_line', data=df)
plt.title('Water Usage Distribution per Product Line')
plt.xticks(rotation=45)
plt.show()
```



```
In [178... # Round all relevant numeric columns to 2 decimal places
numeric_cols = ['Carbon_Footprint_MT', 'Waste_Production_KG', 'Water_Usage_Liters',
                 'Average_Price_USD', 'Net_WastePD', 'Water_Usage_PER_line', 'Water_
df[numeric_cols] = df[numeric_cols].round(2)
```

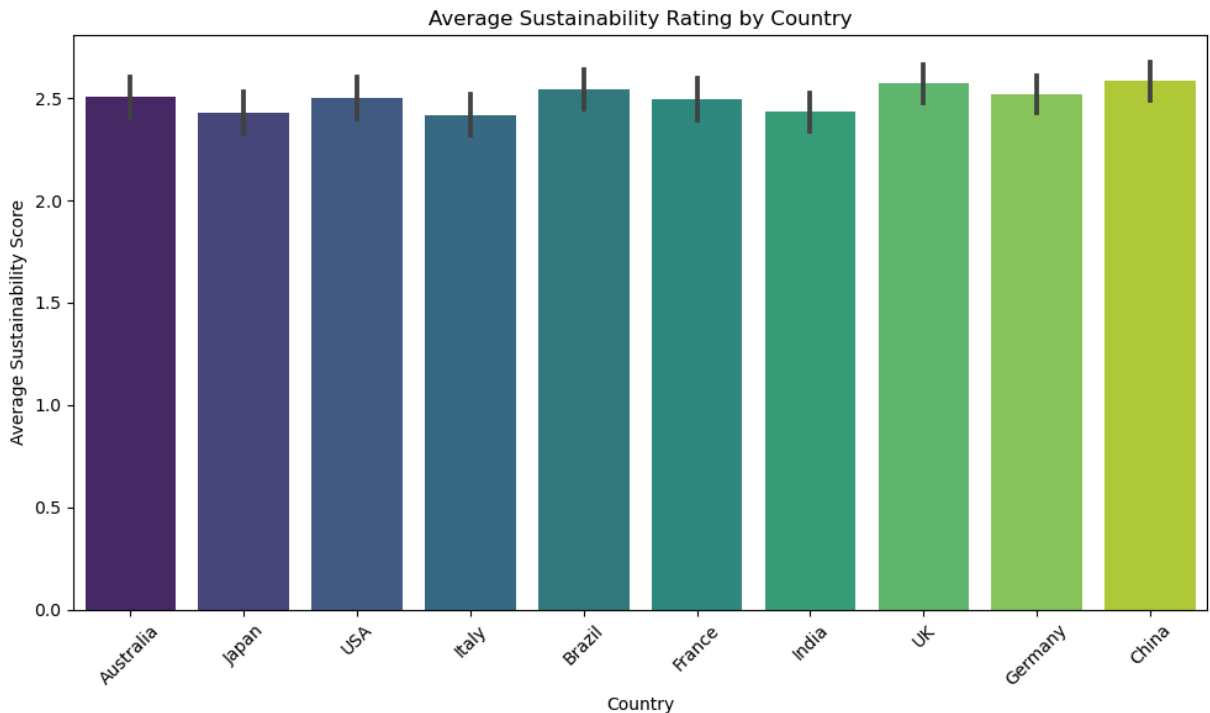
```
In [160... # Exploratory analysis #1: Average sustainability ratings by country
sus_map = {'A': 1, 'B': 2, 'C': 3, 'D': 4}
df['Sustain_Score'] = df['Sustainability_Rating'].map(sus_map)
average_ratings = df.groupby('Country')['Sustain_Score'].mean().reset_index()
sorted_ratings = average_ratings.sort_values(by='Sustain_Score', ascending=False)
print("Average Sustainability Ratings by Country:\n", sorted_ratings)
```

Average Sustainability Ratings by Country:

| | Country | Sustain_Score |
|---|-----------|---------------|
| 2 | China | 2.583665 |
| 8 | UK | 2.574757 |
| 1 | Brazil | 2.544000 |
| 4 | Germany | 2.520710 |
| 0 | Australia | 2.507014 |
| 9 | USA | 2.502970 |
| 3 | France | 2.492958 |
| 5 | India | 2.435388 |
| 7 | Japan | 2.431330 |
| 6 | Italy | 2.418972 |

```
In [161... # Bar Chart
plt.figure(figsize=(10, 6))
sns.barplot(x='Country', y='Sustain_Score', data=df, palette='viridis')
```

```
plt.title('Average Sustainability Rating by Country')
plt.ylabel('Average Sustainability Score')
plt.xlabel('Country')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
In [163... # Exploratory analysis #2: Common materials used by brands with high sustainability
df_pro = df[df['Sustain_Score'].isin([1, 2])]
material_groups = df_pro.groupby('Material_Type')['Brand_ID'].count().reset_index(na
sorted_materials = material_groups.sort_values(by='Brand Count', ascending=False)
print("Common Materials Used by High Sustainability Brands:\n", sorted_materials)
```

Common Materials Used by High Sustainability Brands:

| | Material_Type | Brand Count |
|---|--------------------|-------------|
| 3 | Recycled Polyester | 438 |
| 1 | Hemp | 431 |
| 0 | Bamboo Fabric | 419 |
| 2 | Organic Cotton | 417 |
| 5 | Vegan Leather | 410 |
| 4 | Tencel | 394 |

```
In [166... # Grouping high-end Brands by Material Type to count there use
material_count = df_pro.groupby('Material_Type')['Brand_ID'].count().reset_index(na
```

```
In [167... # Tree Map
fig = px.treemap(material_count,
                  path=['Material_Type'],
                  values='Brand Count',
                  title="Common Materials Used by High-Rated Brands",
                  color='Brand Count',
                  color_continuous_scale='Blues')

fig.show()
```



```
In [168... # Exploratory analysis #3: Which certifications are most common among brands with g
trend_map = {'Growing': 1.5, 'Stable': 1, 'Declining': -0.5}
df['Trend_Map'] = df['Market_Trend'].map(trend_map)
df_grow = df[df['Trend_Map'] == 1.5]
df_stable = df[df['Trend_Map'] == 1]
df_decline = df[df['Trend_Map'] == -0.5]
certified_stableBrands = df_stable.groupby('Certifications')['Brand_ID'].count()
certified_growingBrands = df_grow.groupby('Certifications')['Brand_ID'].count()
certified_declineBrands = df_decline.groupby('Certifications')['Brand_ID'].count()
print('Growing // Certificate Count')
print(certified_growingBrands)
print('Stable // Certificate Count')
print(certified_stableBrands)
print('Decline // Certificate Count')
print(certified_declineBrands)
```

```

Growing // Certificate Count
Certifications
B Corp          312
Fair Trade      334
GOTS            312
Not-Certified   313
OEKO-TEX        341
Name: Brand_ID, dtype: int64
Stable // Certificate Count
Certifications
B Corp          317
Fair Trade      300
GOTS            381
Not-Certified   335
OEKO-TEX        376
Name: Brand_ID, dtype: int64
Decline // Certificate Count
Certifications
B Corp          358
Fair Trade      297
GOTS            357
Not-Certified   356
OEKO-TEX        311
Name: Brand_ID, dtype: int64

```

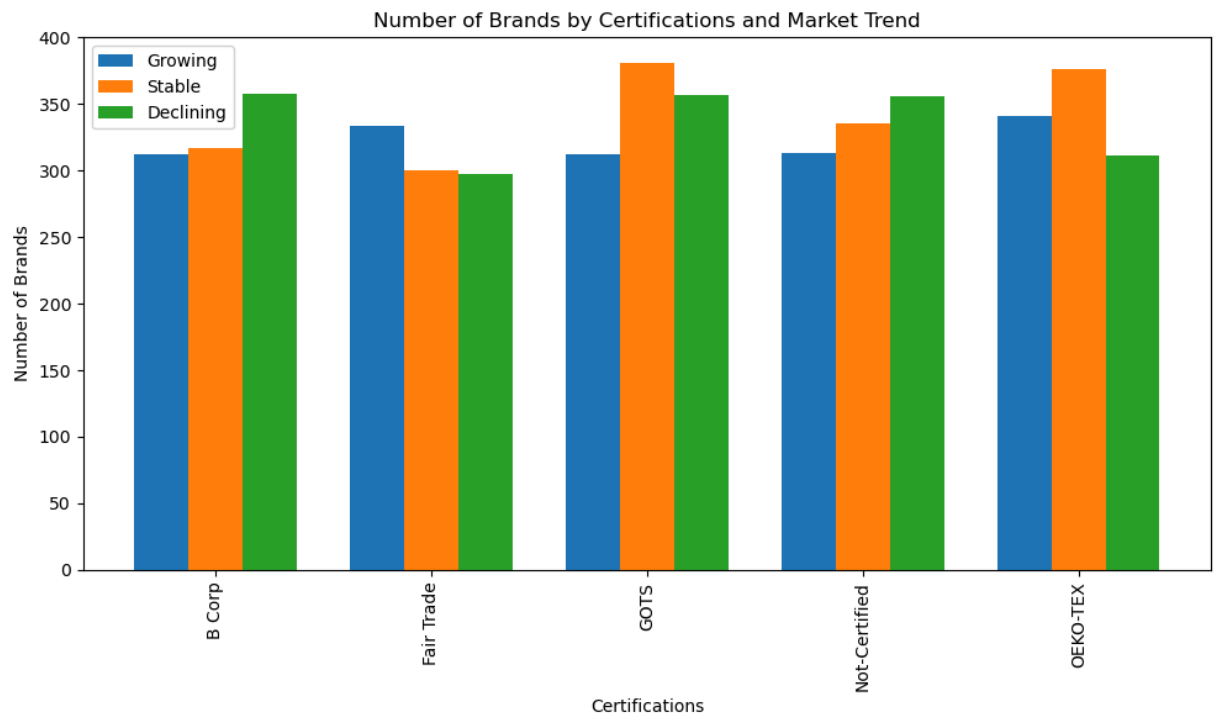
In [169...

```

# Merging the data into a single DataFrame for easy plotting
certified_df = pd.DataFrame({
    'Growing': certified_growingBrands,
    'Stable': certified_stableBrands,
    'Declining': certified_declineBrands
}).fillna(0)

# Grouped Bar Chart
certifications = certified_df.index
x = np.arange(len(certifications)) # Label Locations
width = 0.25 # width of the bars
fig, ax = plt.subplots(figsize=(10, 6))
rects1 = ax.bar(x - width, certified_df['Growing'], width, label='Growing')
rects2 = ax.bar(x, certified_df['Stable'], width, label='Stable')
rects3 = ax.bar(x + width, certified_df['Declining'], width, label='Declining')
ax.set_xlabel('Certifications')
ax.set_ylabel('Number of Brands')
ax.set_title('Number of Brands by Certifications and Market Trend')
ax.set_xticks(x)
ax.set_xticklabels(certifications, rotation=90)
ax.legend()
plt.tight_layout()
plt.show()

```



In [170...

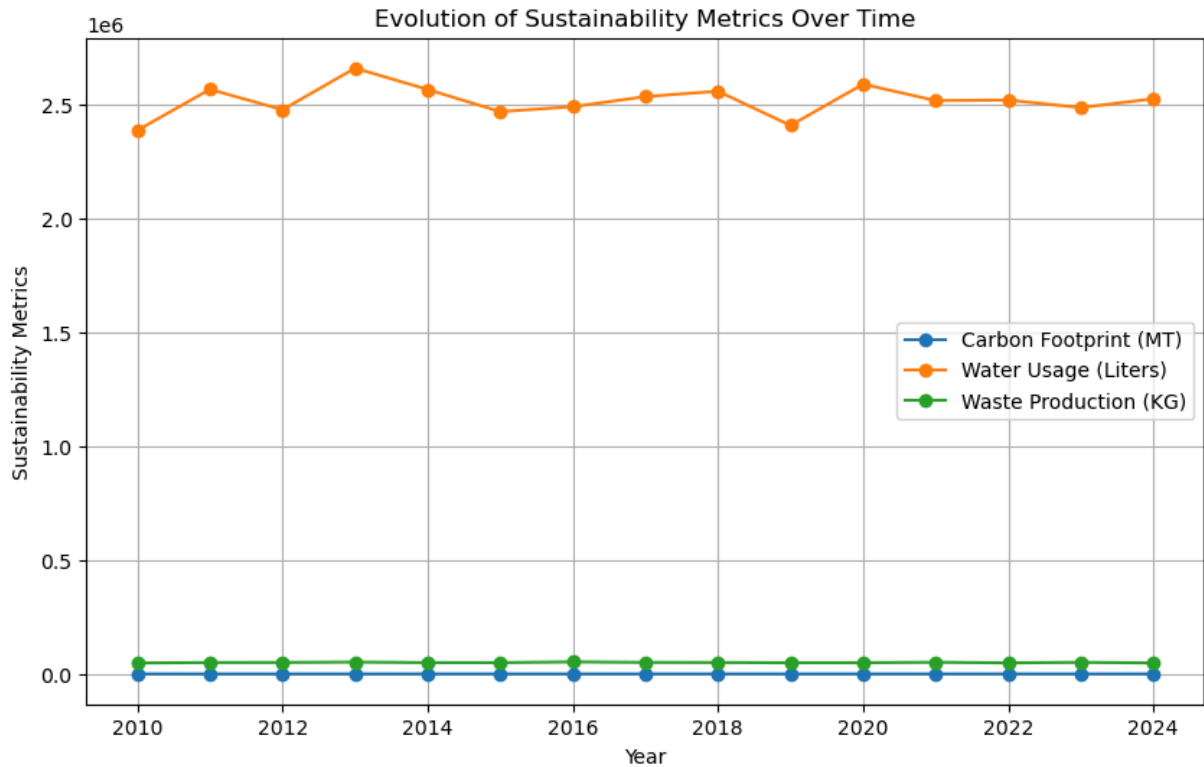
```
# Exploratory analysis #4: How have sustainability metrics evolved over time across
sustainability_metrics = ['Year', 'Carbon_Footprint_MT', 'Water_Usage_Liters', 'Was
sustainability_trends = df[sustainability_metrics].groupby('Year').mean().reset_ind
print(sustainability_trends)
```

| | Year | Carbon_Footprint_MT | Water_Usage_Liters | Waste_Production_KG |
|----|------|---------------------|--------------------|---------------------|
| 0 | 2010 | 252.946400 | 2.387696e+06 | 48101.137657 |
| 1 | 2011 | 239.340173 | 2.567852e+06 | 50077.442968 |
| 2 | 2012 | 258.879286 | 2.478142e+06 | 50388.991964 |
| 3 | 2013 | 242.231550 | 2.660909e+06 | 52072.664195 |
| 4 | 2014 | 255.412387 | 2.567160e+06 | 49759.162810 |
| 5 | 2015 | 247.967710 | 2.469890e+06 | 49564.657839 |
| 6 | 2016 | 263.101724 | 2.491450e+06 | 52905.869885 |
| 7 | 2017 | 246.065702 | 2.535976e+06 | 50709.001287 |
| 8 | 2018 | 253.031183 | 2.559679e+06 | 50267.270000 |
| 9 | 2019 | 245.528079 | 2.409346e+06 | 49037.260199 |
| 10 | 2020 | 251.134524 | 2.590538e+06 | 49314.014082 |
| 11 | 2021 | 250.436534 | 2.518427e+06 | 51247.054688 |
| 12 | 2022 | 246.274955 | 2.520546e+06 | 48641.170593 |
| 13 | 2023 | 244.950781 | 2.487783e+06 | 50955.585435 |
| 14 | 2024 | 255.842066 | 2.526200e+06 | 48302.327305 |

In [184...

```
# Line Plot
plt.figure(figsize=(10, 6))
plt.plot(sustainability_trends['Year'], sustainability_trends['Carbon_Footprint_MT'])
plt.plot(sustainability_trends['Year'], sustainability_trends['Water_Usage_Liters'])
plt.plot(sustainability_trends['Year'], sustainability_trends['Waste_Production_KG'])

plt.xlabel('Year')
plt.ylabel('Sustainability Metrics')
plt.title('Evolution of Sustainability Metrics Over Time')
plt.legend()
plt.grid(True)
plt.show()
```



In [186...

```
# Establishing connection with sqlite3
conn = sqlite3.connect(r"D:\toki\testings\python\new\fashion(p).db")
```

In [188...

```
# Move pandas DataFrame into SQLite
df.to_sql('brands', conn, if_exists='replace', index=False)
```

Out[188...

5000

In [190...

```
# Query the database to double check
cursor = conn.cursor()
cursor.execute("SELECT * FROM brands LIMIT 5;")
rows = cursor.fetchall()
for row in rows:
    print(row)
```

```
('BRAND-0001', 'Australia', 'D', 'Tencel', 'No', 1.75, 4511152.79, 97844.11, 'No',  
2, 38.33, 'Growing', 'GOTS', 0, 97844.11, 'Low', 2255576.4, 117692.48, 4, 1.5)  
( 'BRAND-0002', 'Japan', 'D', 'Vegan Leather', 'Yes', 124.39, 1951566.31, 37267.75,  
'No', 15, 250.07, 'Growing', 'GOTS', 0, 37267.75, 'Medium', 130104.42, 7804.08, 4,  
1.5)  
( 'BRAND-0003', 'USA', 'A', 'Vegan Leather', 'No', 336.66, 467454.52, 50470.95, 'No',  
2, 146.16, 'Growing', 'B Corp', 0, 50470.95, 'Low', 233727.26, 3198.24, 1, 1.5)  
( 'BRAND-0004', 'Italy', 'D', 'Bamboo Fabric', 'No', 152.04, 899576.9, 32665.45, 'N  
o', 13, 165.52, 'Stable', 'OEKO-TEX', 0, 32665.45, 'Medium', 69198.22, 5434.85, 4,  
1.0)  
( 'BRAND-0005', 'USA', 'D', 'Bamboo Fabric', 'Yes', 415.63, 1809219.9, 37295.47, 'Ye  
s', 19, 259.35, 'Stable', 'Fair Trade', 1, 26106.83, 'Medium', 95222.1, 6975.87, 4,  
1.0)
```

```
In [192... # Save cleaned DataFrame to Excel  
df.to_excel(r"D:\toki\testings\python\new\fashion(q).xlsx", index=False)
```

```
In [198... # Close the connection  
conn.close()
```

```
In [ ]: # the end :3
```