

Analysis of sales of an international company

Target: understand which regions, products, and channels generate the most profit and provide business recommendations.

```
In [16]: #We import the libraries necessary for work
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

#Setting up graph display
plt.style.use('ggplot')
plt.rcParams['figure.figsize'] = (10,5)

#Connecting the dataset
df = pd.read_csv('sales.csv')
#Checking if the dataset has Loaded
df.head()
```

	region	country	item_type	sales_channel	order_priority	order_date	order_id
0	Australia and Oceania	Tuvalu	Baby Food	Offline	H	2010-05-28	669165933
1	Central America and the Caribbean	Grenada	Cereal	Online	C	2012-08-22	963881480
2	Europe	Russia	Office Supplies	Offline	L	2014-05-02	341417157
3	Sub-Saharan Africa	Sao Tome and Principe	Fruits	Online	C	2014-06-20	514321792
4	Sub-Saharan Africa	Rwanda	Office Supplies	Offline	L	2013-02-01	115456712

```
In [17]: #Initial data review
print(df.shape)
print(df.columns)
df.info()
print(df.describe())
```

```
(100, 17)
Index(['region', 'country', 'item_type', 'sales_channel', 'order_priority',
       'order_date', 'order_id', 'ship_date', 'units_sold', 'unit_price',
       'unit_cost', 'total_revenue', 'total_cost', 'total_profit', 'month',
       'year', 'shipping_days'],
      dtype='object')
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 17 columns):
 #   Column           Non-Null Count  Dtype  
 ---  --  
 0   region          100 non-null    object  
 1   country         100 non-null    object  
 2   item_type        100 non-null    object  
 3   sales_channel   100 non-null    object  
 4   order_priority  100 non-null    object  
 5   order_date       100 non-null    object  
 6   order_id         100 non-null    int64  
 7   ship_date        100 non-null    object  
 8   units_sold      100 non-null    int64  
 9   unit_price       100 non-null    float64 
 10  unit_cost        100 non-null    float64 
 11  total_revenue   100 non-null    float64 
 12  total_cost       100 non-null    float64 
 13  total_profit    100 non-null    float64 
 14  month            100 non-null    object  
 15  year             100 non-null    int64  
 16  shipping_days   100 non-null    int64  
dtypes: float64(5), int64(4), object(8)
memory usage: 13.4+ KB
      order_id  units_sold  unit_price  unit_cost  total_revenue \
count  1.000000e+02  100.000000  100.000000  100.000000  1.000000e+02
mean   5.550204e+08  5128.710000  276.761300  191.048000  1.373488e+06
std    2.606153e+08  2794.484562  235.592241  188.208181  1.460029e+06
min    1.146066e+08  124.000000   9.330000   6.920000   4.870260e+03
25%   3.389225e+08  2836.250000  81.730000  35.840000  2.687212e+05
50%   5.577086e+08  5382.500000  179.880000  107.275000  7.523144e+05
75%   7.907551e+08  7369.000000  437.200000  263.330000  2.212045e+06
max   9.940222e+08  9925.000000  668.270000  524.960000  5.997055e+06

      total_cost  total_profit  year  shipping_days
count  1.000000e+02  1.000000e+02  100.000000  100.000000
mean   9.318057e+05  4.416820e+05  2013.230000  23.360000
std    1.083938e+06  4.385379e+05  2.088231   14.742586
min    3.612240e+03  1.258020e+03  2010.000000  0.000000
25%   1.688680e+05  1.214436e+05  2012.000000  9.750000
50%   3.635664e+05  2.907680e+05  2013.000000  23.500000
75%   1.613870e+06  6.358288e+05  2015.000000  36.250000
max   4.509794e+06  1.719922e+06  2017.000000  50.000000
```

In [18]: `#We remove spaces and convert everything to lowercase for easier work.`
`df.columns = df.columns.str.strip().str.replace(' ', '_').str.lower()`

`#Removing complete duplicates`
`df = df.drop_duplicates()`

In [19]: `#Converting date column types to datetime, and incorrect dates in NaT`
`df['order_date'] = pd.to_datetime(df['order_date'], errors='coerce', dayfirst=True)`
`df['ship_date'] = pd.to_datetime(df['ship_date'], errors='coerce', dayfirst=False)`

```
#Checking the number of empty dates
df['order_date'].isnull().sum(), df['ship_date'].isnull().sum()
```

Out[19]: (np.int64(0), np.int64(0))

```
In [30]: #Adding additional columns in the form of month and year
df['month'] = df['order_date'].dt.to_period('M')
df['year'] = df['order_date'].dt.year

#We calculate delivery time in days
df['shipping_days'] = (df['ship_date'] - df['order_date']).dt.days

#We calculate how many days it takes for delivery on average
average_shipping_days = round(df['shipping_days'].mean())
print(f"Average delivery time: {average_shipping_days} days")
```

Average delivery time: 23 days

```
In [21]: #Basic business metrics
total_revenue = df['total_revenue'].sum()
total_profit = df['total_profit'].sum()
unique_orders = df['order_id'].nunique()

#Calculate the average bill
order_revenue = df.groupby('order_id')['total_revenue'].sum()
average_order_value = order_revenue.mean()

#Information output
print(f"Total revenue: {total_revenue:.2f}")
print(f"Total profit: {total_profit:.2f}")
print(f"Number of unique orders: {unique_orders:,}")
print(f"Average order value: {average_order_value:.2f}")
```

Total revenue: 137,348,768.31

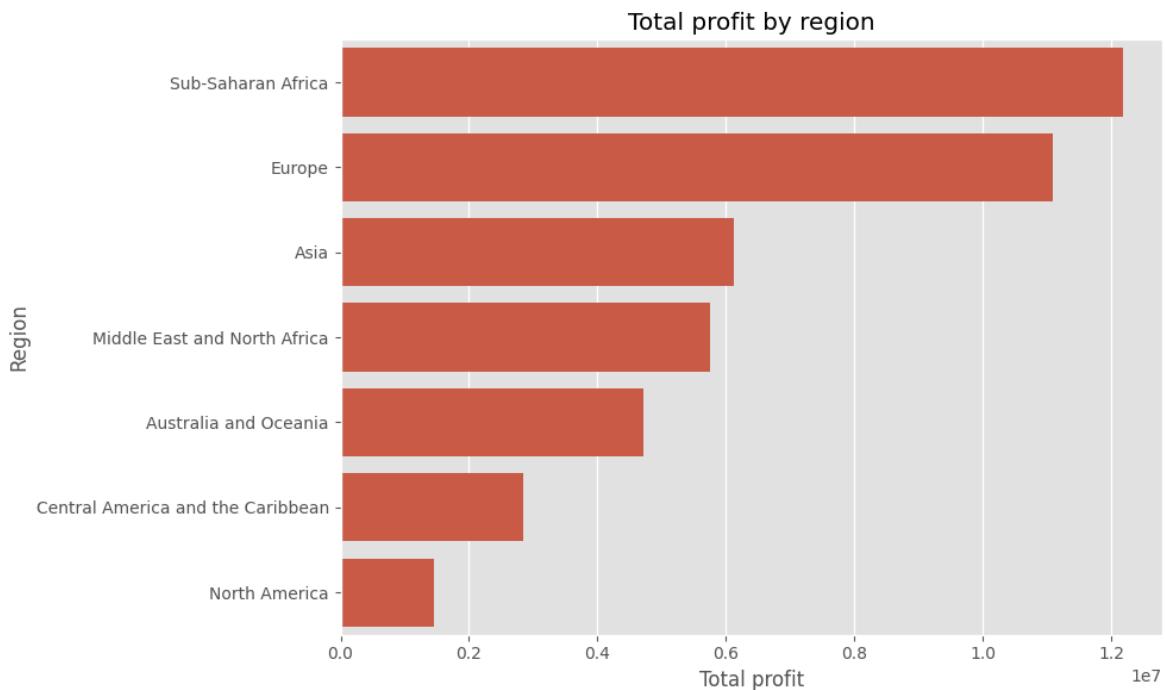
Total profit: 44,168,198.40

Number of unique orders: 100

Average order value: 1,373,487.68

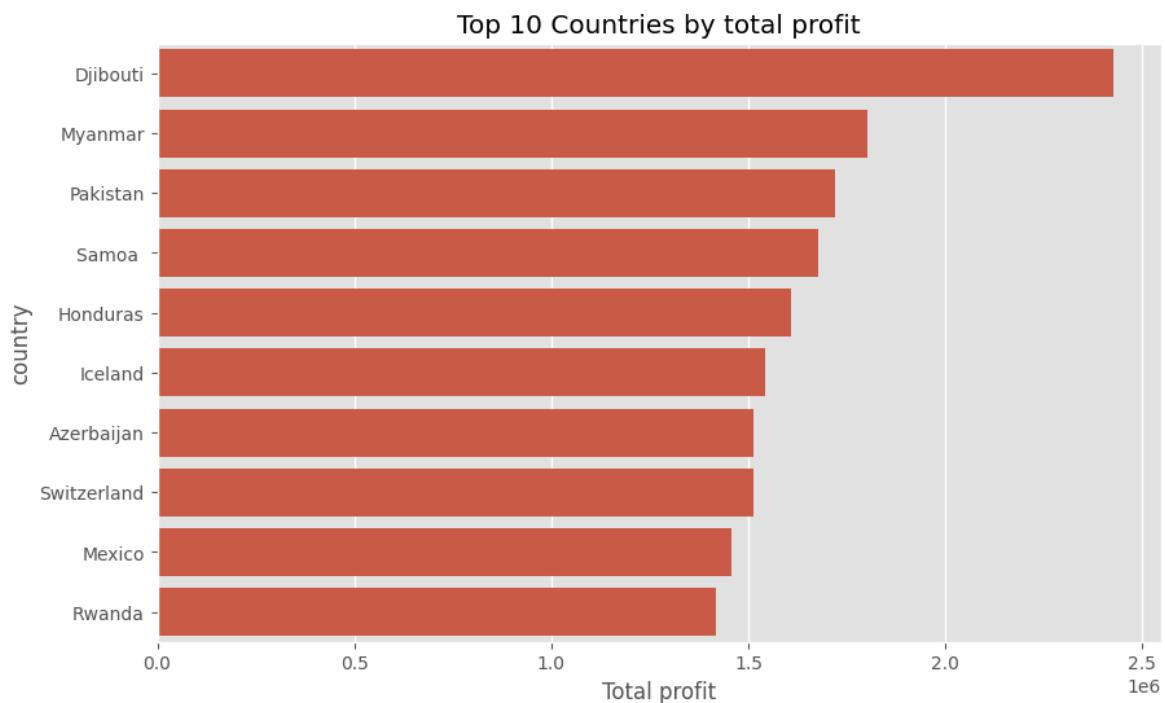
```
In [22]: #We calculate profit by region
region_profit = df.groupby('region')['total_profit'].sum().sort_values(ascending=True)
region_profit

#Creating a graph
plt.figure(figsize=(10, 6))
sns.barplot(x=region_profit.values, y=region_profit.index)
plt.title('Total profit by region') # исправлено: title, не titel
plt.xlabel('Total profit')
plt.ylabel('Region')
plt.tight_layout()
plt.show()
```



```
In [23]: #We count the top 10 countries by income
top_countries = df.groupby('country')['total_profit'].sum().nlargest(10)
top_countries

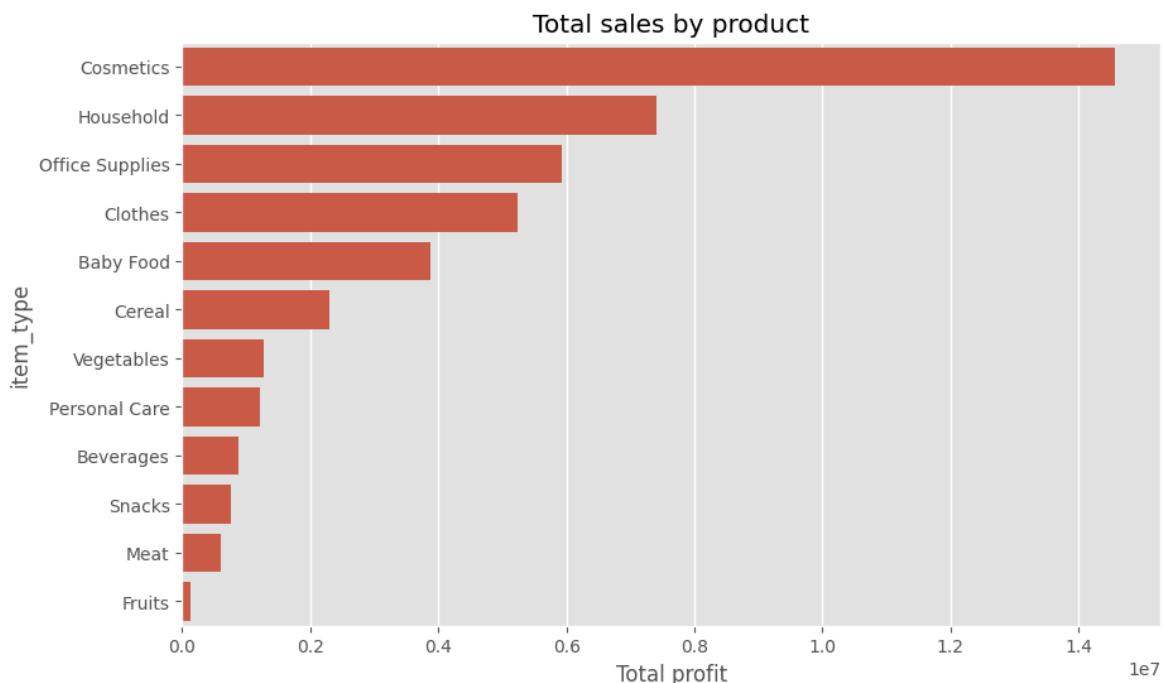
#Creating a graph
plt.figure(figsize=(10,6))
sns.barplot(x=top_countries.values, y=top_countries.index)
plt.title('Top 10 Countries by total profit')
plt.xlabel('Total profit')
plt.show()
```



```
In [24]: #Analysis of product categories by profit
item_profit = df.groupby('item_type')['total_profit'].sum().sort_values(ascending=True)
item_profit

#Creating a graph
```

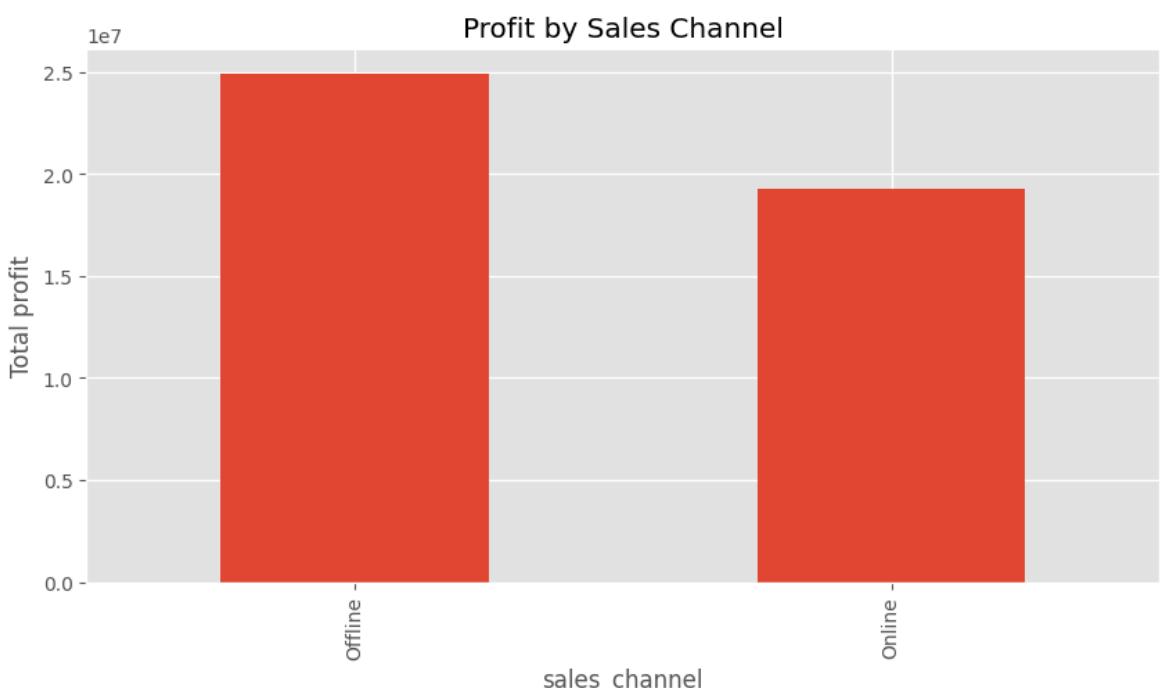
```
plt.figure(figsize=(10,6))
sns.barplot(x=item_profit.values, y=item_profit.index)
plt.title('Total sales by product')
plt.xlabel('Total profit')
plt.show()
```



In [25]:

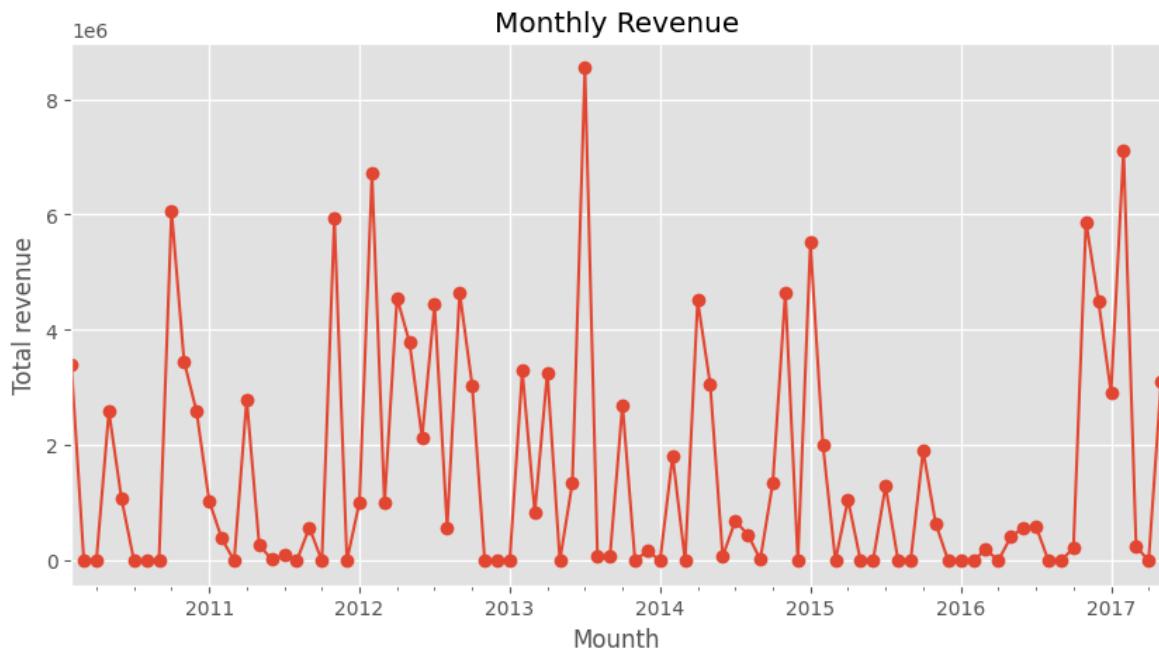
```
#Comparison of offline and online sales
channel_profit = df.groupby('sales_channel')['total_profit'].sum()
channel_profit

#Visualization
channel_profit.plot(kind='bar')
plt.title('Profit by Sales Channel')
plt.ylabel('Total profit')
plt.show()
```



```
In [26]: #Comparison of profits by month
monthly_revenue = df.set_index('order_date').resample('ME')[['total_revenue']].sum()
monthly_revenue.plot(marker='o')

#Visualization
plt.title('Monthly Revenue')
plt.xlabel('Mounth')
plt.ylabel('Total revenue')
plt.show()
```



```
In [27]: #Impact of order category on profit
priority_profit = df.groupby('order_priority')['total_profit'].mean().sort_values()
priority_profit

#Visualization
sns.barplot(x=priority_profit.index, y=priority_profit.values)
plt.title('Average Profit by Order Priority')
plt.ylabel('Average Profit')
plt.show()
```



```
In [28]: #Saving the modified dataset  
df.to_csv('sales.csv', index=False)
```

Sales Analysis Summary

Based on the performed data analysis, the following key insights were identified:

1. Regional Performance

- The most profitable region is **Sub-Saharan Africa**, followed by **Europe**, which generates almost twice as much profit as third-place Asia.
- **North America** is the least profitable region, showing significantly weaker results compared to others.

2. Country-Level Insights

- **Djibouti** stands out as the top-performing country with a noticeable lead in total profit.
- Other countries show more balanced and stable performance, without large deviations.

3. Product Categories

- **Cosmetics** is the best-performing category, generating **over 14 million in profit**, more than double the next category - Household products.
- **Household products** take second place , while **Meat** and **Fruits** show very low profitability and could be reconsidered in the product lineup.

4. Sales Channels

- **Offline sales outperform online**, but online channels still contribute solid revenue and growth potential.

5. Overall Financials

- **Total Revenue:** 137,348,768.31
- **Total Profit:** 44,168,198.40
- **Average Order Value:** 1,373,487.68

Business Recommendations

1. **Expand operations in Sub-Saharan Africa and Europe**, as they deliver the highest profit margins.
Consider investing more in logistics and advertising in these regions to strengthen market dominance.
2. **Reassess North American strategy**, as this region underperforms significantly.
Investigate pricing, competition, or supply chain inefficiencies.
3. **Increase focus on high-profit product categories**, especially **Cosmetics** and **Household goods**.

Expand inventory and marketing campaigns in these segments.

4. **Review low-performing categories** such as **Meat** and **Fruits** — either improve their margins (e.g., through cost optimization) or reduce their presence in the catalog.
5. **Support both sales channels**, but prioritize **offline**, which currently delivers higher profit.
At the same time, develop **online sales** as a long-term growth area.
6. **Evaluate and optimize delivery performance.**
The current **average shipping time is 23 days**, which is relatively long for most business contexts.
It is recommended to **analyze the supply chain** and identify opportunities to reduce shipping delays,
especially if faster delivery could lead to higher customer satisfaction and repeat purchases.

Summary

The company demonstrates strong sales and profit performance in key regions and categories.

Focusing on the most profitable markets and optimizing low-performing areas could significantly increase overall profitability.