

1. Geospatial Analysis: Utilize distribution_centers.csv and users.csv for mapping and analysing the geographic distribution of users and distribution centres.

For first problem firstly I have imported all necessary libraries along with datasets and did some EDA in the data and fetch insights from it following the problem statement

```
In [107]: 1 import numpy as np
          2 import pandas as pd
          3 import seaborn as sns
          4 import matplotlib.pyplot as plt
          5 from mpl_toolkits.basemap import Basemap
          6 import matplotlib.pyplot as plt
          7 import seaborn as sns
          8 sns.set_style("whitegrid")
          9 from mpl_toolkits.basemap import Basemap
         10 from geopy.geocoders import Nominatim
```

Geospatial Analysis:

```
In [3]: 1 # Load data
        2 distribution_centers = pd.read_csv(r"C:\Users\Harshit\Downloads\data\data\distribution_centers.csv")
        3 users = pd.read_csv(r"C:\Users\Harshit\Downloads\data\data\users.csv")
```

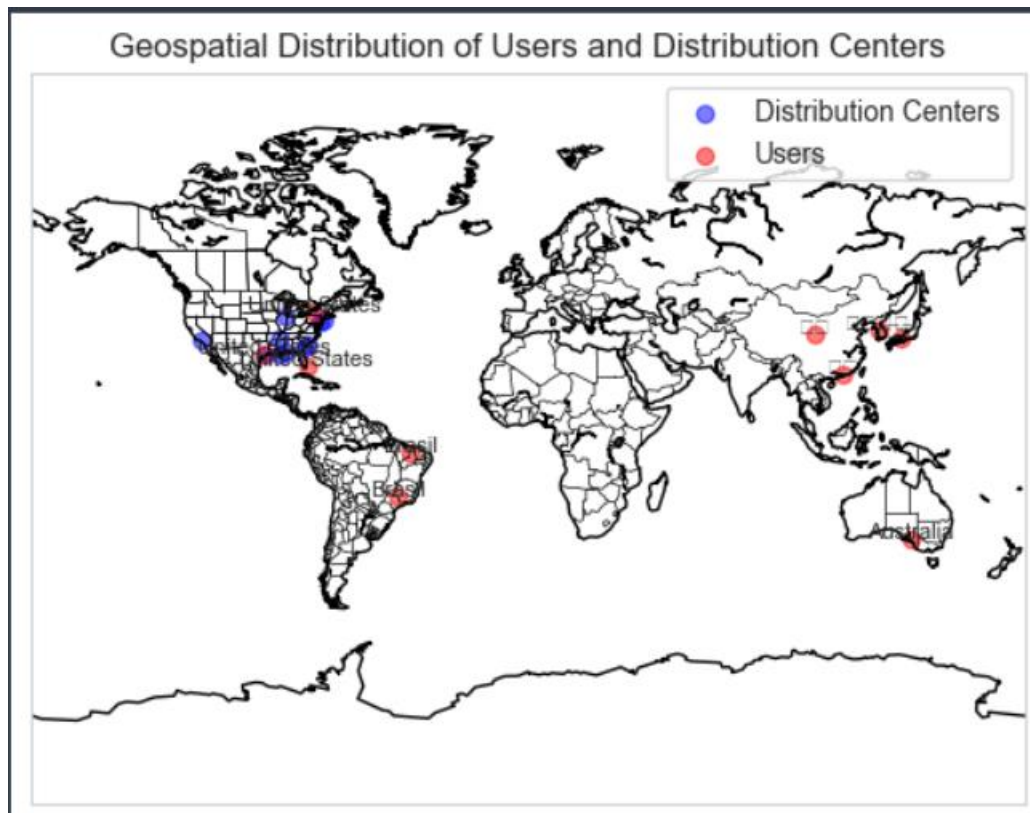
```
In [4]: 1 distribution_centers.head()
```

	id	name	latitude	longitude
0	1	Memphis TN	35.1174	-89.9711
1	2	Chicago IL	41.8369	-87.6847
2	3	Houston TX	29.7604	-95.3698
3	4	Los Angeles CA	34.0500	-118.2500
4	5	New Orleans LA	29.9500	-90.0667

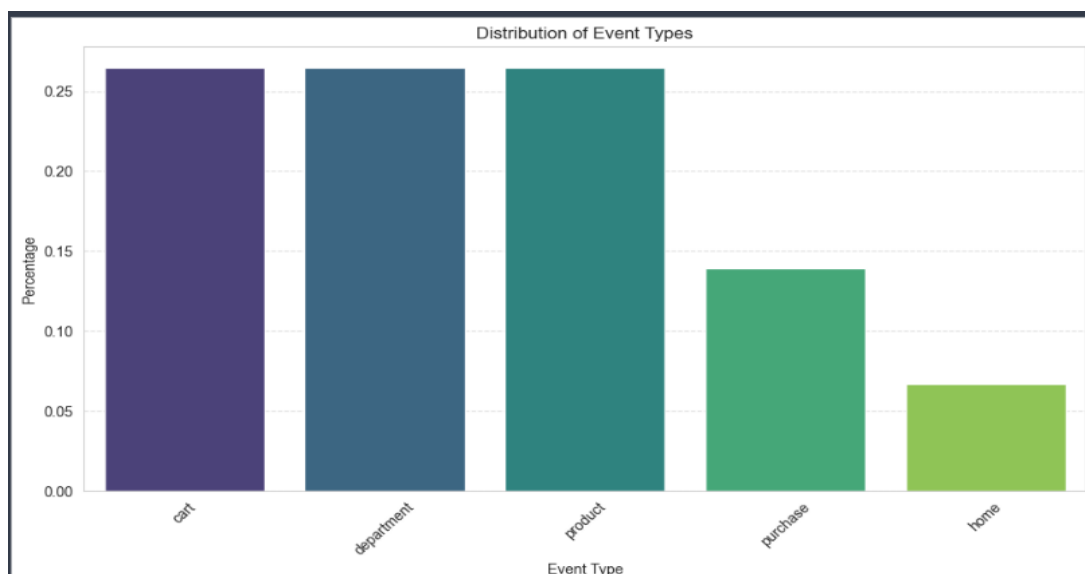
```
In [5]: 1 users.head()
```

	id	first_name	last_name	email	age	gender	state	street_address	postal_code	city	country
0	457	Timothy	Bush	timothybush@example.net	65	M	Acre	87620 Johnson Hills	69917-400	Rio Branco	Brasil
1	6578	Elizabeth	Martinez	elizabethmartinez@example.com	34	F	Acre	1705 Nielsen Land	69917-400	Rio Branco	Brasil
2	36280	Christopher	Mendoza	christophermendoza@example.net	13	M	Acre	125 Turner Isle Apt. 264	69917-400	Rio Branco	Brasil
3	60193	Jimmy	Conner	jimmyconner@example.com	64	M	Acre	0966 Jose Branch Apt. 008	69917-400	Rio Branco	Brasil
4	64231	Natasha	Wilson	natashawilson@example.net	25	F	Acre	20798 Phillip Trail Apt. 392	69917-400	Rio Branco	Brasil

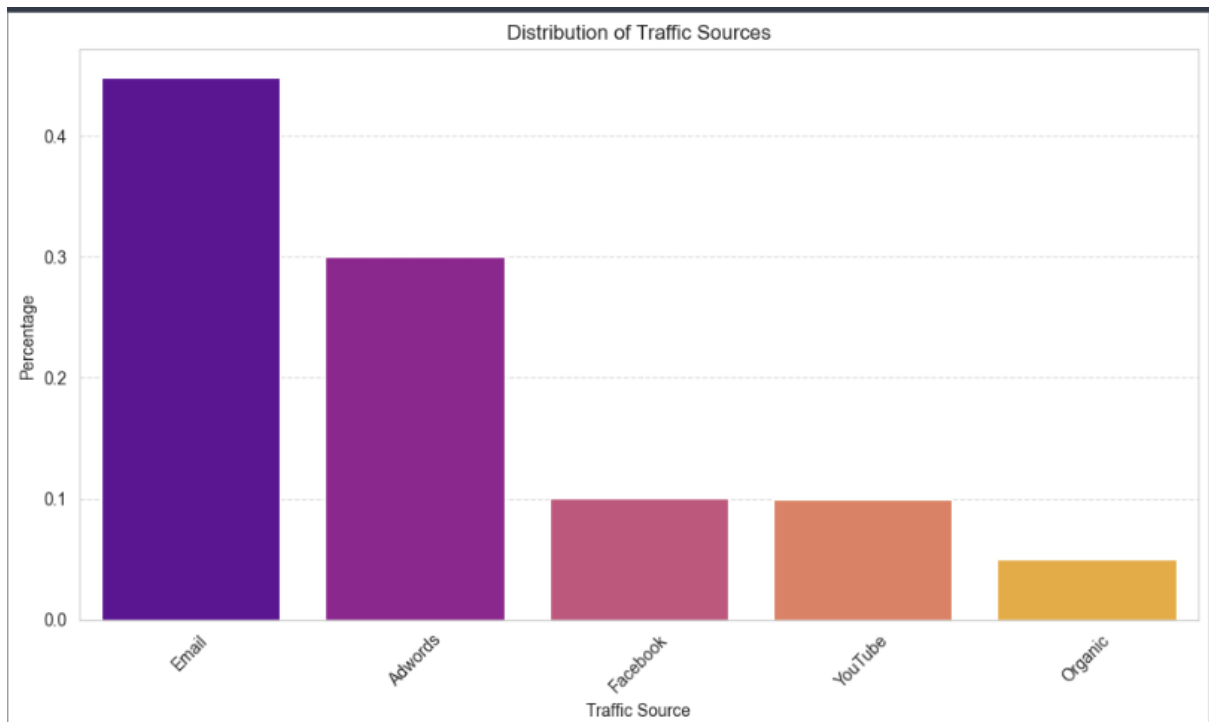
And now as given below visualization we can easily see the geographical distribution and centres.



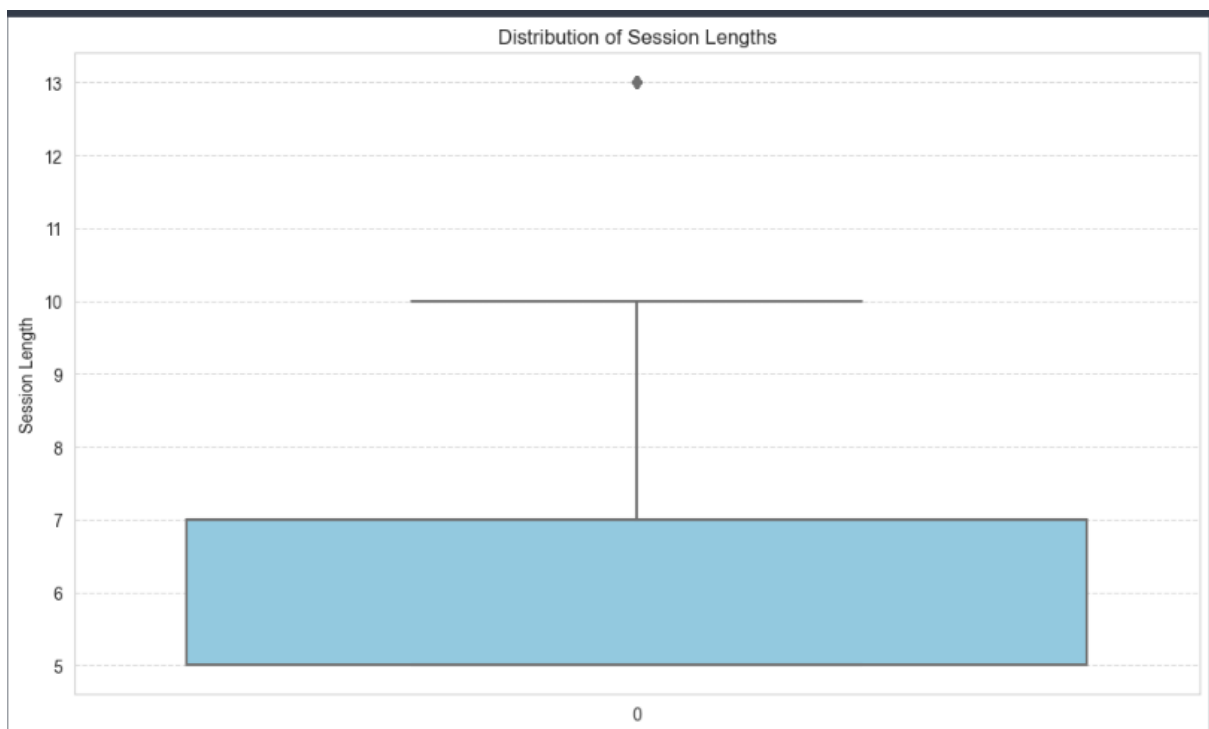
2. User Behaviour Analysis: Use events.csv to analyse user behaviour, including session patterns, traffic sources, and event types.



Event type distribution



Traffic source distribution



Session length distribution using a boxplot

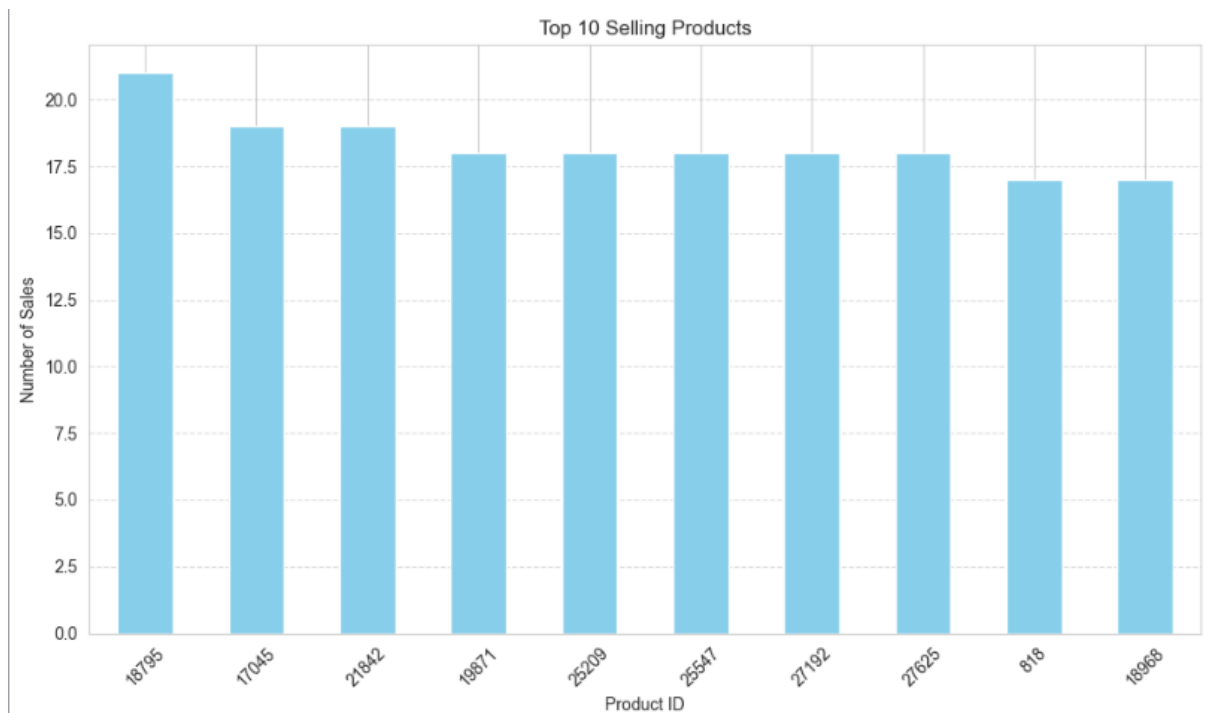
3. Sales and Revenue Analysis: Leverage order_items.csv and inventory_items.csv to analyse product sales, revenue, and profitability.

This is the table of inventory items in which there was missing value at very high percentage in the column name sold_at so I have filled that NaN value with “Not Sold”.

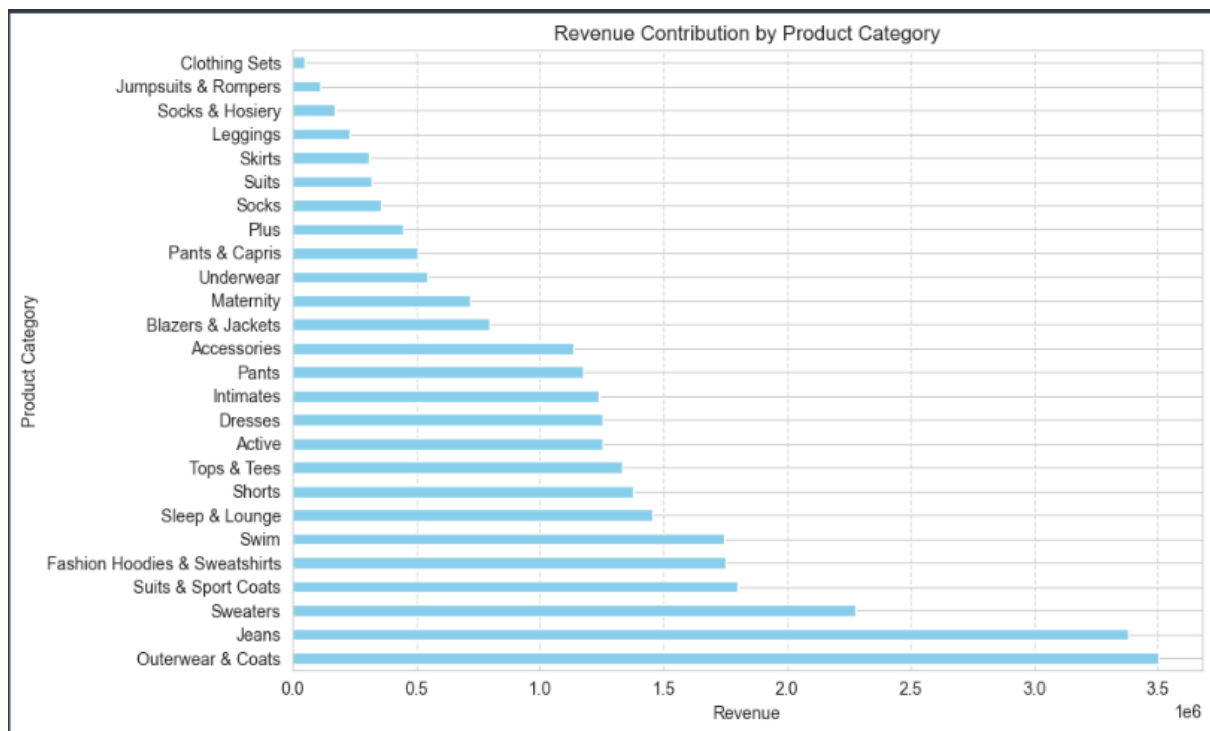
```
In [115]: 1 inventory_items.head()
```

	id	product_id	created_at	sold_at	cost	product_category	product_name	product_brand	product_retail_price
0	67971	13844	2022-07-02 07:09:20+00:00	2022-07-24 06:33:20+00:00	2.76804	Accessories	(ONE) 1 Satin Headband	Funny Girl Designs	6.99
1	67972	13844	2023-12-20 03:28:00+00:00	Not Sold	2.76804	Accessories	(ONE) 1 Satin Headband	Funny Girl Designs	6.99
2	67973	13844	2023-06-04 02:53:00+00:00	Not Sold	2.76804	Accessories	(ONE) 1 Satin Headband	Funny Girl Designs	6.99
3	72863	13844	2021-10-16 22:58:52+00:00	2021-11-22 02:19:52+00:00	2.76804	Accessories	(ONE) 1 Satin Headband	Funny Girl Designs	6.99
4	72864	13844	2021-08-07 16:33:00+00:00	Not Sold	2.76804	Accessories	(ONE) 1 Satin Headband	Funny Girl Designs	6.99

Bar plot for top-selling products



Revenue for each product category



Total Profit for Each Year

```
Total Profit for Each Year:
year
2018    -5.871489e+03
2019    -5.648322e+05
2020    -1.261111e+07
2021    -1.403636e+07
2022    -1.646481e+07
2023    -2.255754e+07
2024    -6.352898e+05
dtype: float64
```

4. Product Performance Analysis: Explore products.csv to analyse product performance, including costs, categories, and popularity.

Here's how data looks like:

	id	cost	category	name	brand	retail_price	department	sku	distribution_center
0	13842	2.51875	Accessories	Low Profile Dyed Cotton Twill Cap - Navy W39S55D	MG	6.25	Women	EBD58B8A3F1D72F4206201DA62FB1204	1
1	13928	2.33835	Accessories	Low Profile Dyed Cotton Twill Cap - Putty W39S55D	MG	5.95	Women	2EAC42424D12436BDD6A5B8A88480CC3	1
2	14115	4.87956	Accessories	Enzyme Regular Solid Army Caps-Black W35S45D	MG	10.99	Women	EE364229B2791D1EF9355708EFF0BA34	1
3	14157	4.64877	Accessories	Enzyme Regular Solid Army Caps-Olive W35S45D (...)	MG	10.99	Women	00BD13095D06C20B11A2993CA419D16B	1
4	14273	6.50793	Accessories	Washed Canvas Ivy Cap - Black W11S64C	MG	15.99	Women	F531DC20FDE20B7ADF3A73F52B71D0AF	1

Analysis based on given problem statement

```

1 # Display results
2 print(f"Total cost of all products: ${total_cost:.2f}")
3 print(f"Average cost per product: ${average_cost:.2f}")
4 print("\nCategory Analysis:")
5 print(category_counts)
6 print(f"\nTotal retail price of all products: ${total_retail_price:.2f}")
7 print(f"Average retail price per product: ${average_retail_price:.2f}")
8 print("\nDepartment Analysis:")
9 print(department_counts)

```

Total cost of all products: \$829389.27
Average cost per product: \$28.48

Category Analysis:

```

category
Intimates      2363
Jeans          1999
Tops & Tees    1868
Fashion Hoodies & Sweatshirts 1866
Swim           1798
Sleep & Lounge 1771
Shorts         1765
Sweaters       1737
Accessories    1559
Active         1432
Outerwear & Coats 1420
Underwear      1088
Pants          1041
Dresses        955
Socks          905
Maternity      898
Plus           758
Suits & Sport Coats 739
Socks & Hosiery 666
Pants & Capris  613
Leggings       564
Blazers & Jackets 561
Skirts         367
Suits          188
Jumpsuits & Rompers 162
Clothing Sets  37
Name: count, dtype: int64

```

Total retail price of all products: \$1724491.17
Average retail price per product: \$59.22

Department Analysis:

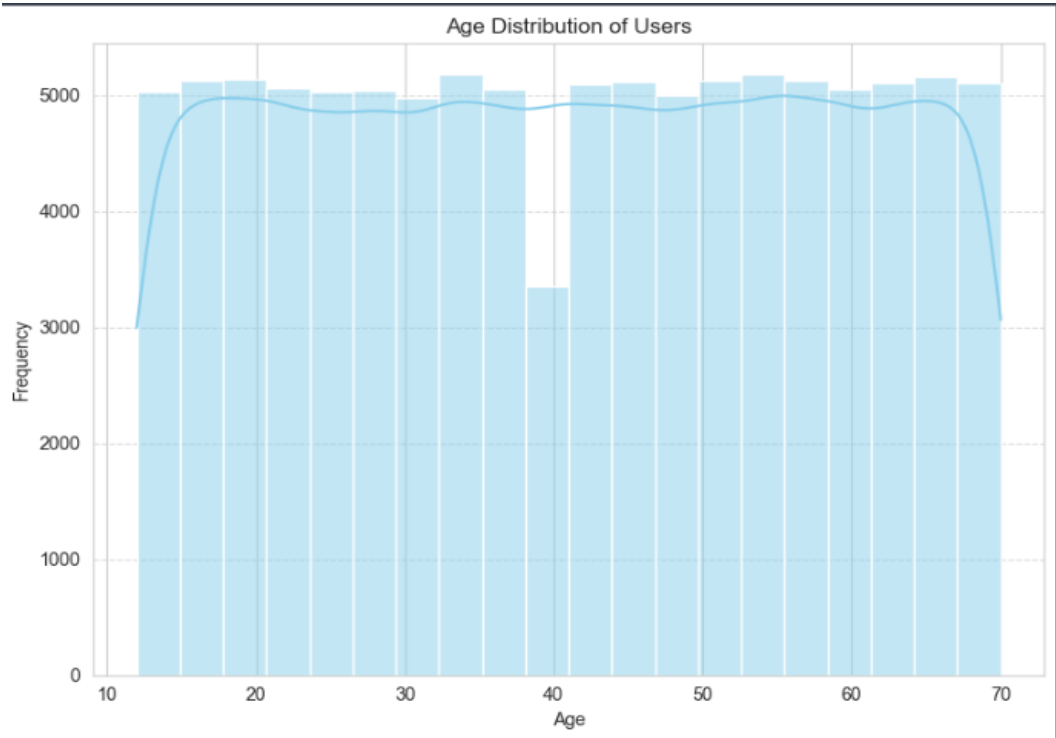
```

department
Women    15989
Men      13131
Name: count, dtype: int64

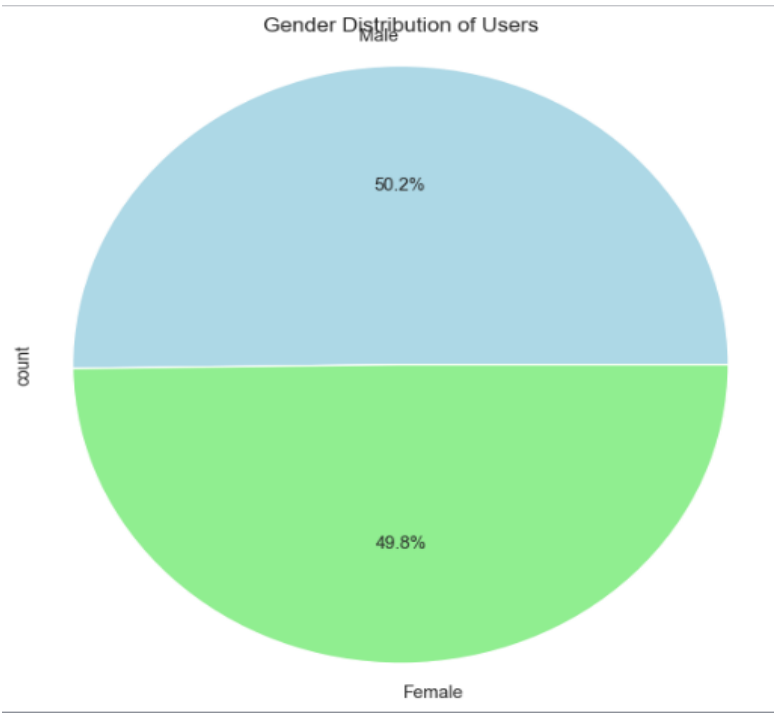
```

5. User Demographics Analysis: Use users.csv to analyse user demographics, such as age, gender, and location.

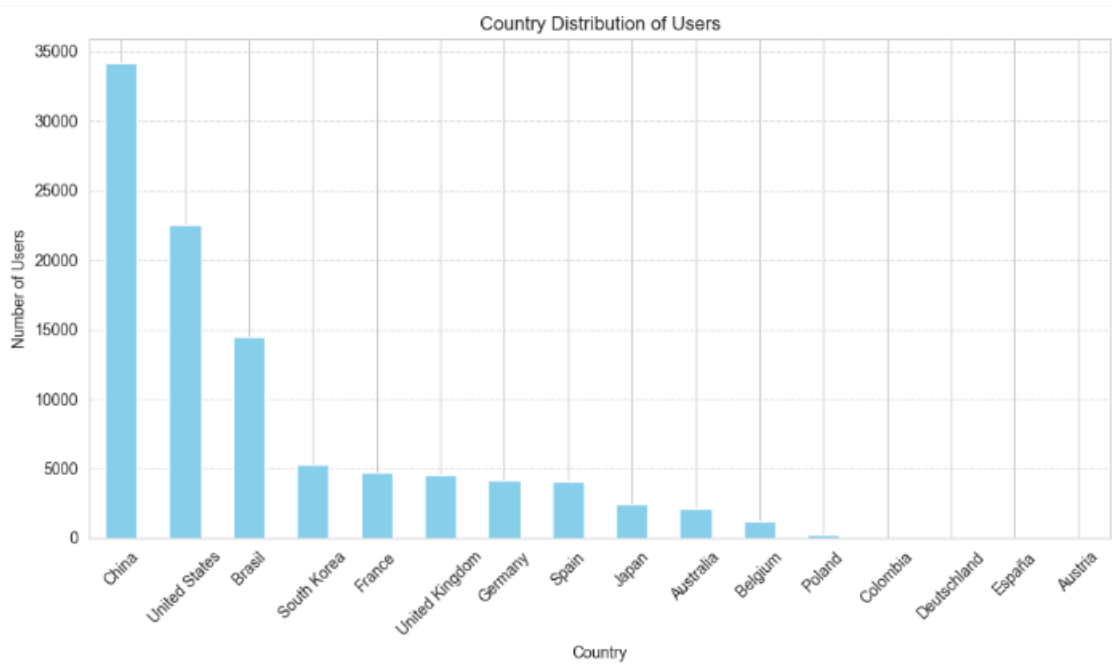
Age Distribution



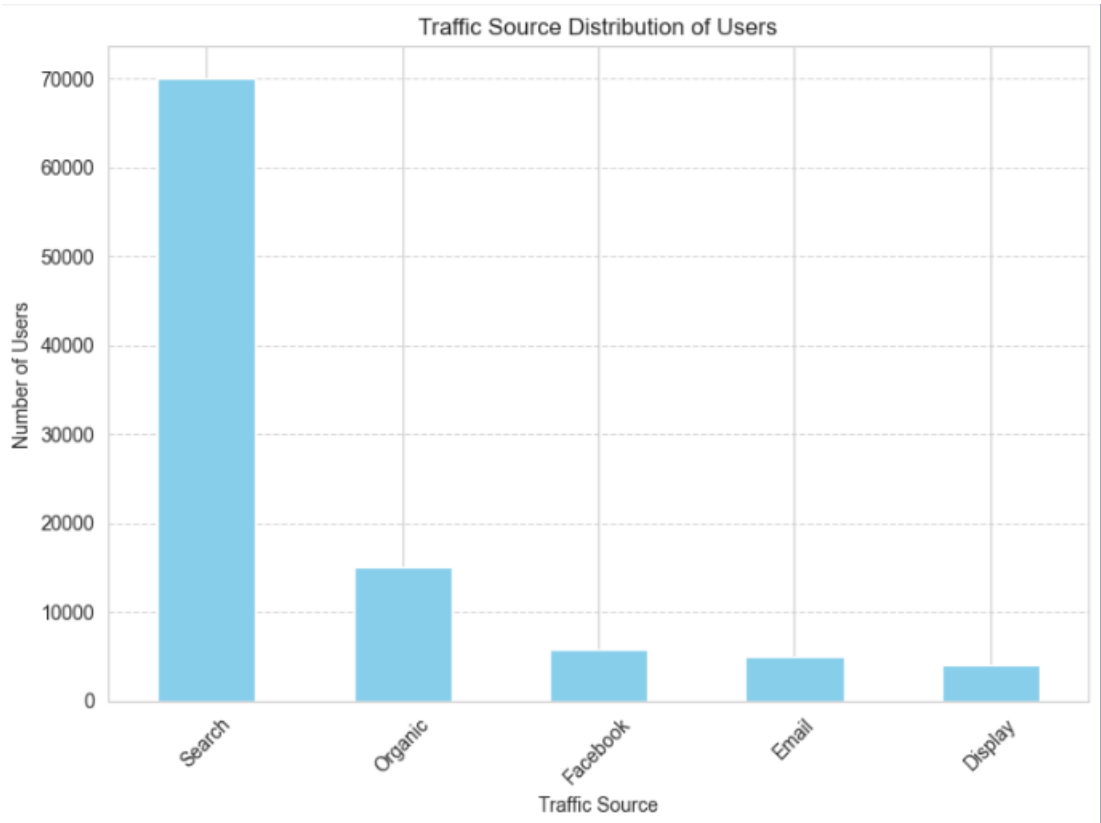
Gender Distribution



Country Distribution



Traffic Source Distribution



Displaying Results

```
Average age of users: 41.05  
Median age of users: 41.0  
Mode age of users: 33
```

Gender Analysis:

```
gender  
F    50208  
M    49792  
Name: count, dtype: int64
```

Country Analysis:

```
country  
China          34150  
United States  22522  
Brasil         14507  
South Korea    5316  
France         4700  
United Kingdom 4561  
Germany        4155  
Spain          4062  
Japan          2438  
Australia      2146  
Belgium        1185  
Poland         235  
Colombia       17  
Deutschland    2  
España         2  
Austria        2  
Name: count, dtype: int64
```

Traffic Source Analysis:

```
traffic_source  
Search    70075  
Organic   15110  
Facebook  5816  
Email     4947  
Display   4052  
Name: count, dtype: int64
```

6. Order Fulfilment Analysis: Analyse order_items.csv and orders.csv to understand order fulfilment timelines and status.

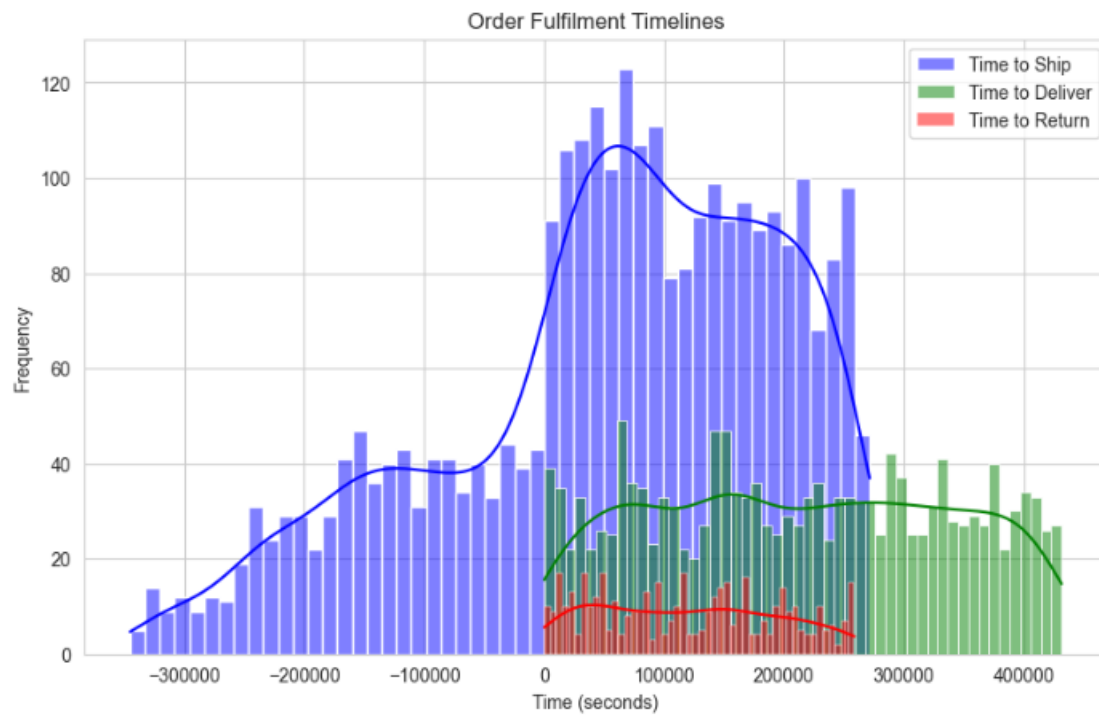
Time to ship

```
Time to Ship:  
count      2871.000000  
mean       53657.491815  
std        142040.469327  
min        -345197.000000  
25%        -26866.000000  
50%         72266.000000  
75%        165445.500000  
max        271546.000000  
Name: time_to_ship_seconds, dtype: float64
```

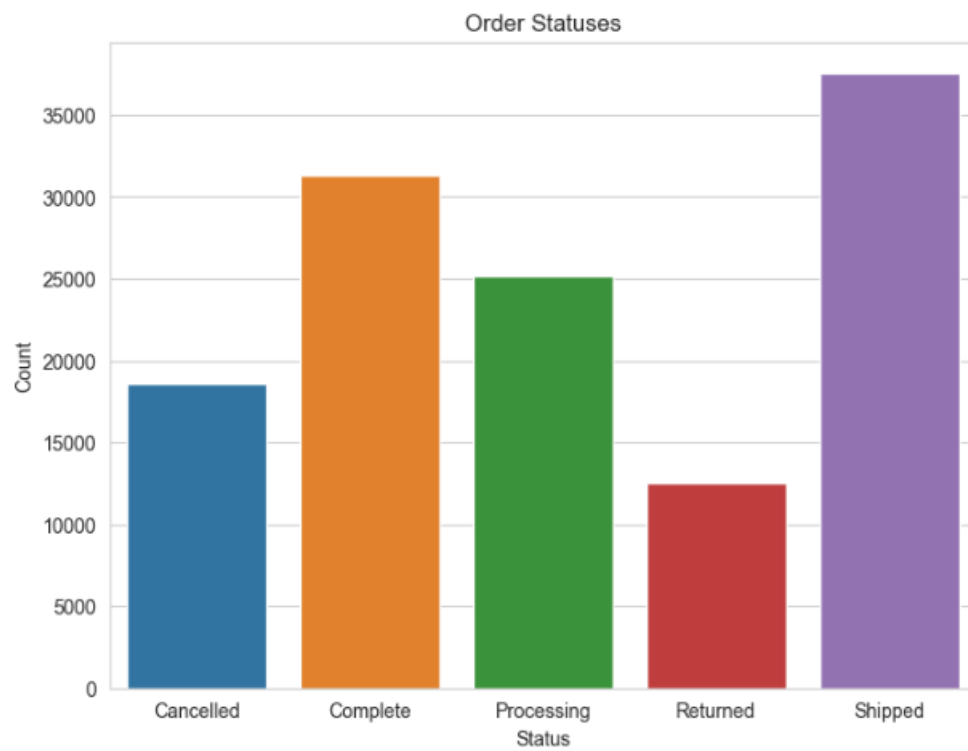
```
Time to Deliver:  
count      1555.000000  
mean       213992.951768  
std        123626.093424  
min          0.000000  
25%        105630.000000  
50%        213000.000000  
75%        320400.000000  
max        431580.000000  
Name: time_to_deliver_seconds, dtype: float64
```

```
Time to Return:  
count       448.000000  
mean       119374.553571  
std        74751.930437  
min          0.000000  
25%        50865.000000  
50%        116670.000000  
75%        181335.000000  
max        258660.000000  
Name: time_to_return_seconds, dtype: float64
```

Order fulfilment timelines



Order status



Return status

