

# online\_retail

July 16, 2023

## 1 Portfolio Project: Online Retail Exploratory Data Analysis with Python

### 1.1 Overview

In this project, I will step into the shoes of an entry-level data analyst at an online retail company, helping interpret real-world data to help make a key business decision.

### 1.2 Case Study

In this project, i was working with transactional data from an online retail store. The dataset contains information about customer purchases, including product details, quantities, prices, and timestamps. Your task is to explore and analyze this dataset to gain insights into the store's sales trends, customer behavior, and popular products.

By conducting exploratory data analysis, you will identify patterns, outliers, and correlations in the data, allowing you to make data-driven decisions and recommendations to optimize the store's operations and improve customer satisfaction. Through visualizations and statistical analysis, you will uncover key trends, such as the busiest sales months, best-selling products, and the store's most valuable customers. Ultimately, this project aims to provide actionable insights that can drive strategic business decisions and enhance the store's overall performance in the competitive online retail market.

### 1.3 Project Objectives

1. Describe data to answer key questions to uncover insights
2. Gain valuable insights that will help improve online retail performance
3. Provide analytic insights and data-driven recommendations

### 1.4 Dataset

The dataset you will be working with is the “Online Retail” dataset. It contains transactional data of an online retail store from 2010 to 2011. The dataset is available as a .xlsx file named **Online Retail.xlsx**. it can also be downloaded [here](#).

The dataset contains the following columns:

- InvoiceNo: Invoice number of the transaction
- StockCode: Unique code of the product
- Description: Description of the product
- Quantity: Quantity of the product in the transaction
- InvoiceDate: Date and time of the transaction
- UnitPrice: Unit price of the product
- CustomerID: Unique identifier of the customer
- Country: Country where the transaction occurred

## 1.5 Tasks

You may explore this dataset in any way you would like - however if you'd like some help getting started, here are a few ideas:

1. Load the dataset into a Pandas DataFrame and display the first few rows to get an overview of the data.
2. Perform data cleaning by handling missing values, if any, and removing any redundant or unnecessary columns.
3. Explore the basic statistics of the dataset, including measures of central tendency and dispersion.
4. Perform data visualization to gain insights into the dataset. Generate appropriate plots, such as histograms, scatter plots, or bar plots, to visualize different aspects of the data.
5. Analyze the sales trends over time. Identify the busiest months and days of the week in terms of sales.
6. Explore the top-selling products and countries based on the quantity sold.
7. Identify any outliers or anomalies in the dataset and discuss their potential impact on the analysis.
8. Draw conclusions and summarize your findings from the exploratory data analysis.

## 1.6 Task 1: Load the Data

```
[48]: import pandas as pd
import numpy as np
import seaborn as sns                    #visualisation
import matplotlib.pyplot as plt         #visualisation
%matplotlib inline
sns.set(color_codes=True)
```

```
[64]: df = pd.read_excel("Online Retail.xlsx")
# To display the top 5 rows
df.head(5)
```

```
[64]: InvoiceNo StockCode Description Quantity \
0    536365    85123A  WHITE HANGING HEART T-LIGHT HOLDER        6
1    536365    71053           WHITE METAL LANTERN              6
2    536365    84406B  CREAM CUPID HEARTS COAT HANGER           8
```

3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6

	InvoiceDate	UnitPrice	CustomerID	Country
0	2010-12-01 08:26:00	2.55	17850.0	United Kingdom
1	2010-12-01 08:26:00	3.39	17850.0	United Kingdom
2	2010-12-01 08:26:00	2.75	17850.0	United Kingdom
3	2010-12-01 08:26:00	3.39	17850.0	United Kingdom
4	2010-12-01 08:26:00	3.39	17850.0	United Kingdom

```
[62]: df.tail(5)
```

```
[62]:
```

	InvoiceNo	StockCode	Description	Quantity	\
541904	581587	22613	PACK OF 20 SPACEBOY NAPKINS	12	
541905	581587	22899	CHILDREN'S APRON DOLLY GIRL	6	
541906	581587	23254	CHILDRENS CUTLERY DOLLY GIRL	4	
541907	581587	23255	CHILDRENS CUTLERY CIRCUS PARADE	4	
541908	581587	22138	BAKING SET 9 PIECE RETROSPOT	3	

	InvoiceDate	UnitPrice	CustomerID	Country
541904	2011-12-09 12:50:00	0.85	12680.0	France
541905	2011-12-09 12:50:00	2.10	12680.0	France
541906	2011-12-09 12:50:00	4.15	12680.0	France
541907	2011-12-09 12:50:00	4.15	12680.0	France
541908	2011-12-09 12:50:00	4.95	12680.0	France

```
[51]: df.dtypes
```

```
[51]: InvoiceNo          object
      StockCode        object
      Description      object
      Quantity         int64
      InvoiceDate      datetime64[ns]
      UnitPrice        float64
      CustomerID       float64
      Country          object
      dtype: object
```

## 2 Task 2: Clean the Data

```
[52]: df=df[['CustomerID','InvoiceNo','StockCode','Quantity','UnitPrice','Description','Country']]
```

```
[53]: df.head(10)
```

```
[53]:
```

	CustomerID	InvoiceNo	StockCode	Quantity	UnitPrice	\
0	17850.0	536365	85123A	6	2.55	
1	17850.0	536365	71053	6	3.39	
2	17850.0	536365	84406B	8	2.75	
3	17850.0	536365	84029G	6	3.39	
4	17850.0	536365	84029E	6	3.39	
5	17850.0	536365	22752	2	7.65	
6	17850.0	536365	21730	6	4.25	
7	17850.0	536366	22633	6	1.85	
8	17850.0	536366	22632	6	1.85	
9	13047.0	536367	84879	32	1.69	

	Description	Country
0	WHITE HANGING HEART T-LIGHT HOLDER	United Kingdom
1	WHITE METAL LANTERN	United Kingdom
2	CREAM CUPID HEARTS COAT HANGER	United Kingdom
3	KNITTED UNION FLAG HOT WATER BOTTLE	United Kingdom
4	RED WOOLLY HOTTIE WHITE HEART.	United Kingdom
5	SET 7 BABUSHKA NESTING BOXES	United Kingdom
6	GLASS STAR FROSTED T-LIGHT HOLDER	United Kingdom
7	HAND WARMER UNION JACK	United Kingdom
8	HAND WARMER RED POLKA DOT	United Kingdom
9	ASSORTED COLOUR BIRD ORNAMENT	United Kingdom

```
[54]: df.shape
```

```
[54]: (541909, 7)
```

```
[55]: duplicate_rows_df = df[df.duplicated()]
print("number of duplicate rows: ", duplicate_rows_df.shape)
```

```
number of duplicate rows: (5269, 7)
```

```
[56]: df.count()
```

```
[56]: CustomerID      406829
InvoiceNo      541909
StockCode      541909
Quantity      541909
UnitPrice      541909
Description      540455
Country      541909
dtype: int64
```

```
[57]: df = df.drop_duplicates()
df.head(5)
```

```
[57]: CustomerID InvoiceNo StockCode Quantity UnitPrice \
0      17850.0    536365    85123A         6         2.55
1      17850.0    536365     71053         6         3.39
2      17850.0    536365    84406B         8         2.75
3      17850.0    536365    84029G         6         3.39
4      17850.0    536365    84029E         6         3.39
```

```

                                Description          Country
0  WHITE HANGING HEART T-LIGHT HOLDER  United Kingdom
1                WHITE METAL LANTERN  United Kingdom
2    CREAM CUPID HEARTS COAT HANGER  United Kingdom
3  KNITTED UNION FLAG HOT WATER BOTTLE  United Kingdom
4    RED WOOLLY HOTTIE WHITE HEART.  United Kingdom
```

```
[58]: df.count()
```

```
[58]: CustomerID      401603
InvoiceNo      536640
StockCode      536640
Quantity      536640
UnitPrice      536640
Description      535186
Country      536640
dtype: int64
```

```
[59]: print(df.isnull().sum())
```

```
CustomerID      135037
InvoiceNo         0
StockCode         0
Quantity         0
UnitPrice         0
Description      1454
Country          0
dtype: int64
```

```
[60]: df = df.dropna()      # Dropping the missing values.
df.count()
```

```
[60]: CustomerID      401603
InvoiceNo      401603
StockCode      401603
Quantity      401603
UnitPrice      401603
Description      401603
Country      401603
dtype: int64
```

```
[47]: print(df.isnull().sum())
```

```
InvoiceNo      0
StockCode      0
Description    1454
Quantity       0
InvoiceDate    0
UnitPrice      0
CustomerID    135080
Country        0
dtype: int64
```

### 3 Task3: Data visualization & Analysis

```
[65]: TotalAmount = df['Quantity'] * df['UnitPrice']
df.insert(loc=5, column='TotalAmount', value=TotalAmount)
```

```
[66]: new_df = df
      ↪ df[['CustomerID', 'InvoiceNo', 'StockCode', 'Quantity', 'TotalAmount', 'InvoiceDate', 'Country']]

new_df2 = df.copy()
```

```
[67]: new_df.head()
```

```
[67]:
```

	CustomerID	InvoiceNo	StockCode	Quantity	TotalAmount	InvoiceDate	\
0	17850.0	536365	85123A	6	15.30	2010-12-01 08:26:00	
1	17850.0	536365	71053	6	20.34	2010-12-01 08:26:00	
2	17850.0	536365	84406B	8	22.00	2010-12-01 08:26:00	
3	17850.0	536365	84029G	6	20.34	2010-12-01 08:26:00	
4	17850.0	536365	84029E	6	20.34	2010-12-01 08:26:00	

```
Country
0 United Kingdom
1 United Kingdom
2 United Kingdom
3 United Kingdom
4 United Kingdom
```

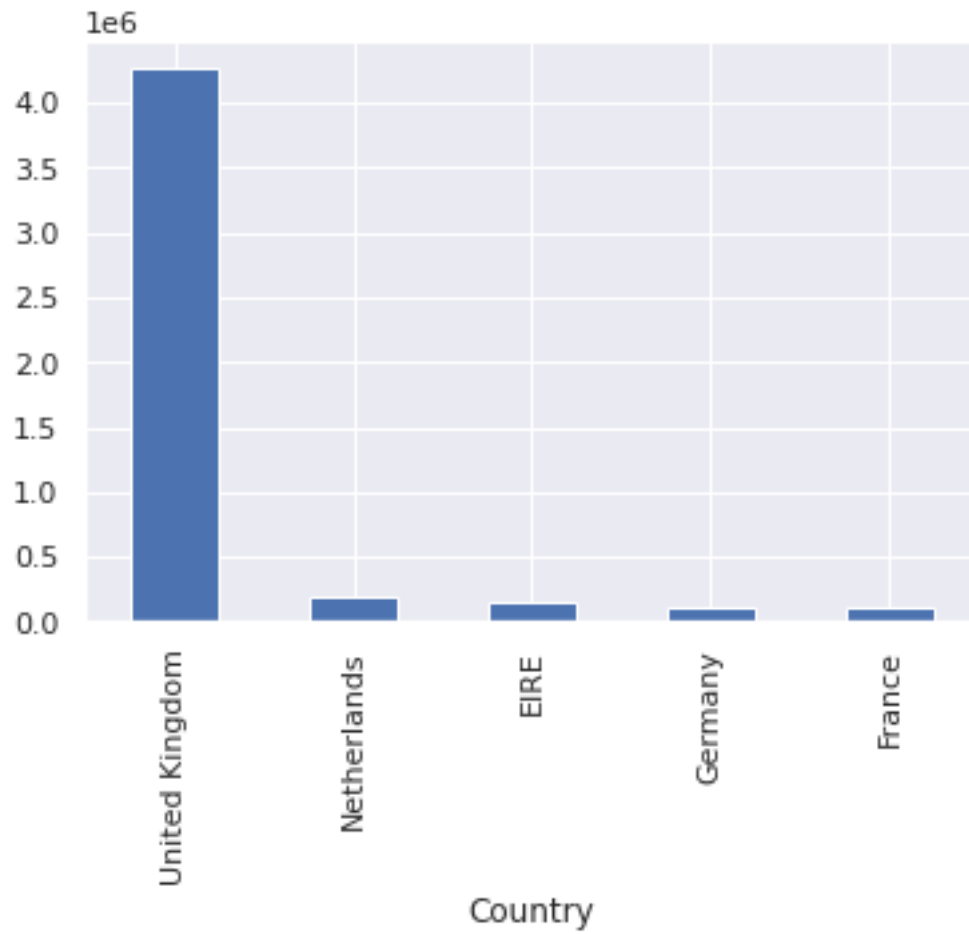
#### 3.0.1 Exploratory Data Analysis(EDA)

```
[69]: country_price = new_df.groupby('Country')['Quantity'].sum().
      ↪ sort_values(ascending = False)
country_price
# Grouping countries by TotalAmount of sales
```

```
[69]: Country
      United Kingdom      4263829
      Netherlands        200128
      EIRE                142637
      Germany            117448
      France             110480
      Australia          83653
      Sweden             35637
      Switzerland        30325
      Spain              26824
      Japan              25218
      Belgium            23152
      Norway             19247
      Portugal           16180
      Finland            10666
      Channel Islands     9479
      Denmark            8188
      Italy              7999
      Cyprus             6317
      Singapore          5234
      Austria            4827
      Hong Kong          4769
      Israel             4353
      Poland             3653
      Unspecified        3300
      Canada             2763
      Iceland            2458
      Greece             1556
      USA                1034
      United Arab Emirates 982
      Malta              944
      Lithuania          652
      Czech Republic     592
      European Community 497
      Lebanon            386
      Brazil             356
      RSA                352
      Bahrain            260
      Saudi Arabia        75
      Name: Quantity, dtype: int64
```

```
[70]: country_price[:5].plot(kind = 'bar')
      # Top 5 Companies with high number of purchase
```

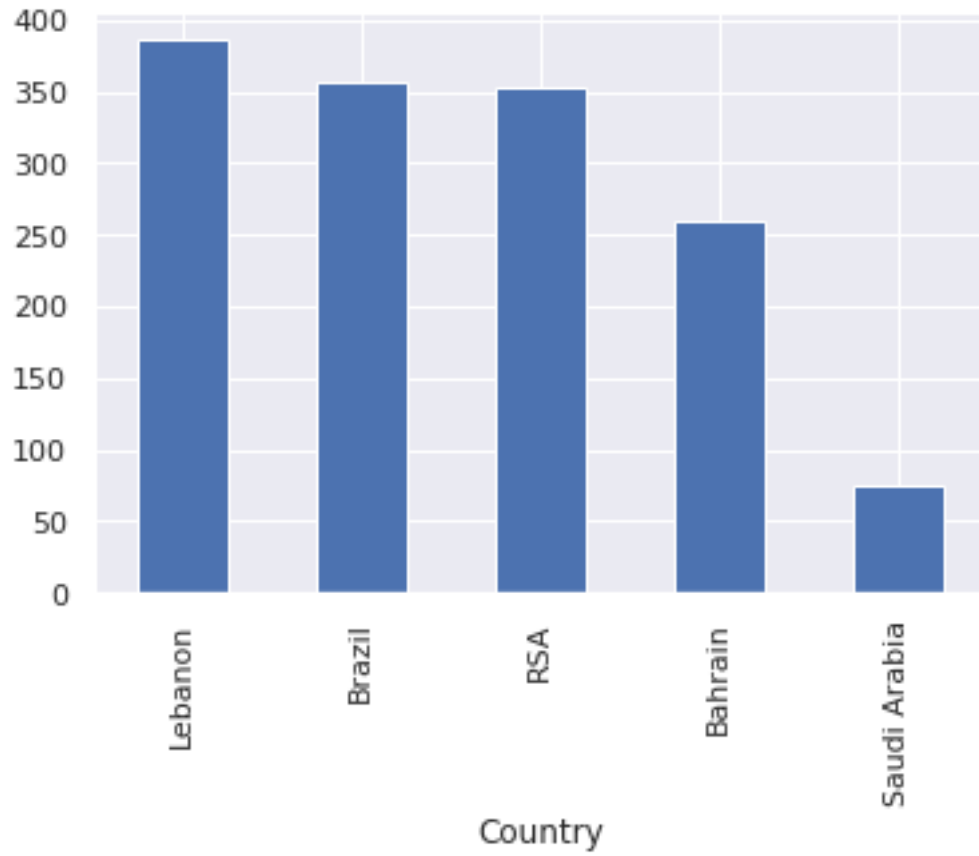
```
[70]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7d4be3e1d0>
```



```
[71]: # 5 Compaines with least number of purchase  
country_price[33:].plot(kind = 'bar')
```

```
[71]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7d543e5bd0>
```





```
[72]: # Adding year feature to the dataset

timest = new_df['InvoiceDate'].dt.year

new_df['Year'] = timest

new_df.head()
```

```
[72]:
```

	CustomerID	InvoiceNo	StockCode	Quantity	TotalAmount	InvoiceDate	\
0	17850.0	536365	85123A	6	15.30	2010-12-01 08:26:00	
1	17850.0	536365	71053	6	20.34	2010-12-01 08:26:00	
2	17850.0	536365	84406B	8	22.00	2010-12-01 08:26:00	
3	17850.0	536365	84029G	6	20.34	2010-12-01 08:26:00	
4	17850.0	536365	84029E	6	20.34	2010-12-01 08:26:00	

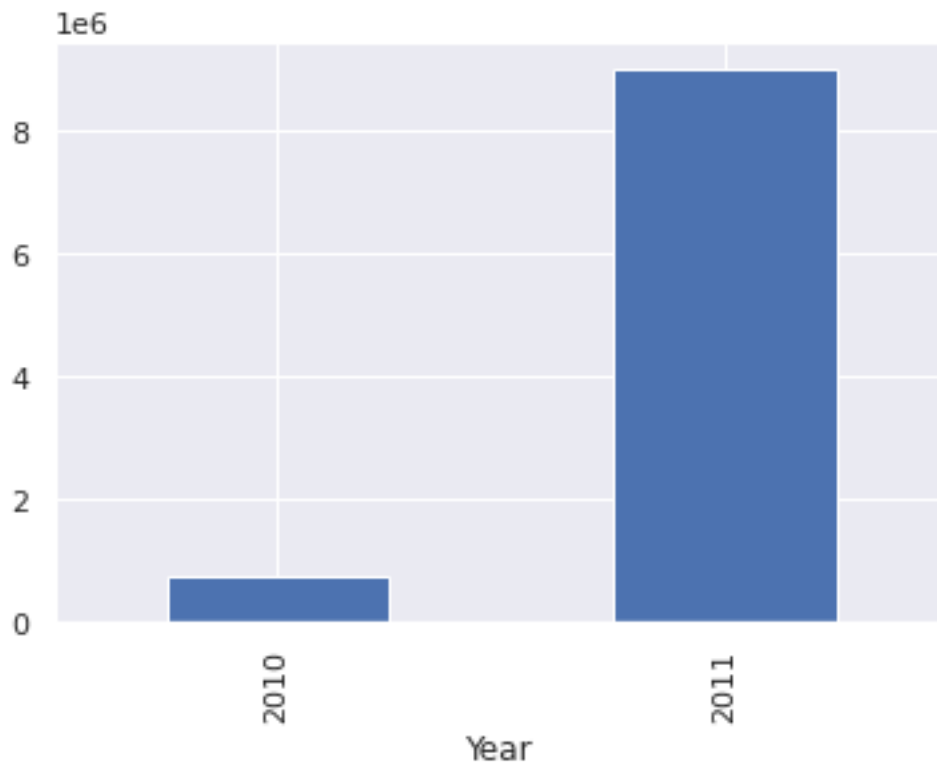
	Country	Year
0	United Kingdom	2010
1	United Kingdom	2010
2	United Kingdom	2010
3	United Kingdom	2010

4 United Kingdom 2010

```
[73]: # Total sales for different years
```

```
new_df.groupby('Year')['TotalAmount'].sum().plot(kind = 'bar')
```

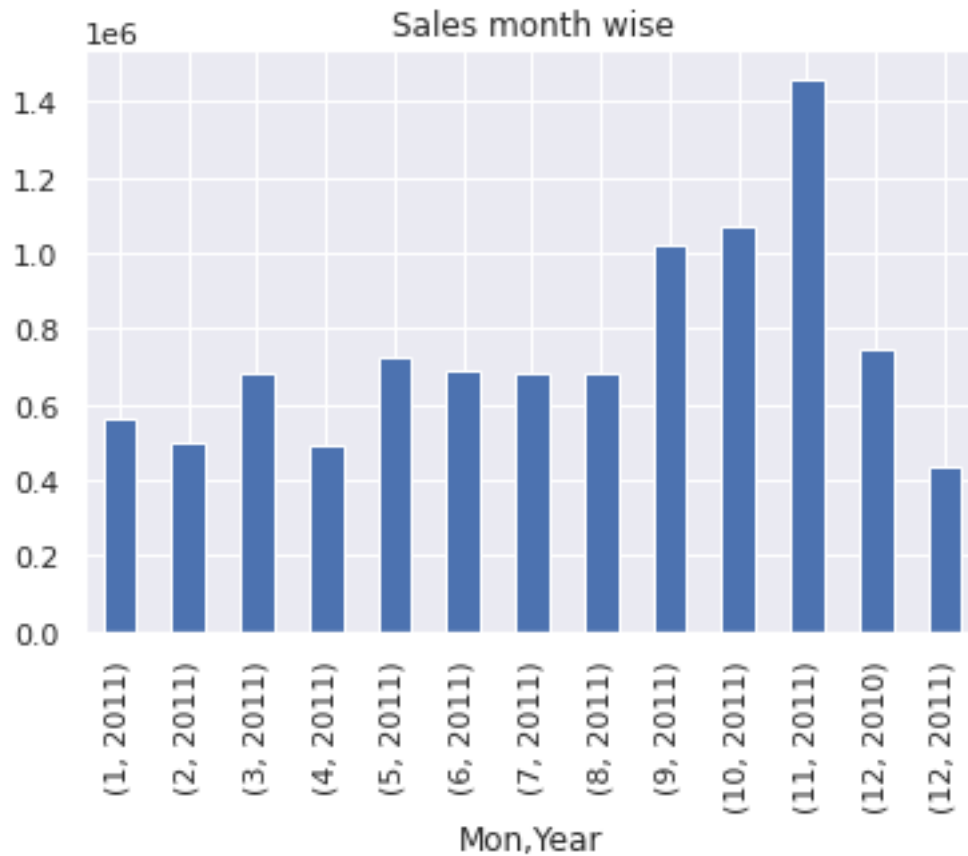
```
[73]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7d5d6ef210>
```



```
[74]: # Sales for different months
```

```
new_df['Mon'] = new_df['InvoiceDate'].dt.month
new_df['month'] = new_df['InvoiceDate'].dt.month_name()
new_df.groupby(['Mon', 'Year'])['TotalAmount'].sum().plot(kind = 'bar', title = 'Sales month wise')
```

```
[74]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7d4a4b1a90>
```



[75]: *# Checking why dec 2011 has a drop comparing to nov 2011*

```
get_2011 = new_df[(new_df['Year'] == 2011)]
get_dec2011 = get_2011[(new_df['month'] == 'December')]
get_dec2011 = get_dec2011['InvoiceDate'].dt.date.unique()
get_dec2011
```

```
[75]: array([datetime.date(2011, 12, 1), datetime.date(2011, 12, 2),
          datetime.date(2011, 12, 4), datetime.date(2011, 12, 5),
          datetime.date(2011, 12, 6), datetime.date(2011, 12, 7),
          datetime.date(2011, 12, 8), datetime.date(2011, 12, 9)],
          dtype=object)
```

### 3.1 Answer :

### 3.2 Performance Analysis

Sales Performance can be seen with

Number of sales every month

Number of sales every year

We see that in 2010 we have sales only for dec and in 2011 we have sales for all months

We can see that from September to November we have very good sales

We could see that DEC 2010 we have sales of 748957 and in DEC 2011 we have sales of 433668 which is a huge drop when analyzed further found out that We have only data upto 9th on dec 2011, so we find a sales drop in the month of dec 2011

## 4 Task 4:Analysis

potential areas of improvement for the business

```
[76]: new_df.head()
```

```
[76]:
```

	CustomerID	InvoiceNo	StockCode	Quantity	TotalAmount	InvoiceDate	\
0	17850.0	536365	85123A	6	15.30	2010-12-01 08:26:00	
1	17850.0	536365	71053	6	20.34	2010-12-01 08:26:00	
2	17850.0	536365	84406B	8	22.00	2010-12-01 08:26:00	
3	17850.0	536365	84029G	6	20.34	2010-12-01 08:26:00	
4	17850.0	536365	84029E	6	20.34	2010-12-01 08:26:00	

	Country	Year	Mon	month
0	United Kingdom	2010	12	December
1	United Kingdom	2010	12	December
2	United Kingdom	2010	12	December
3	United Kingdom	2010	12	December
4	United Kingdom	2010	12	December

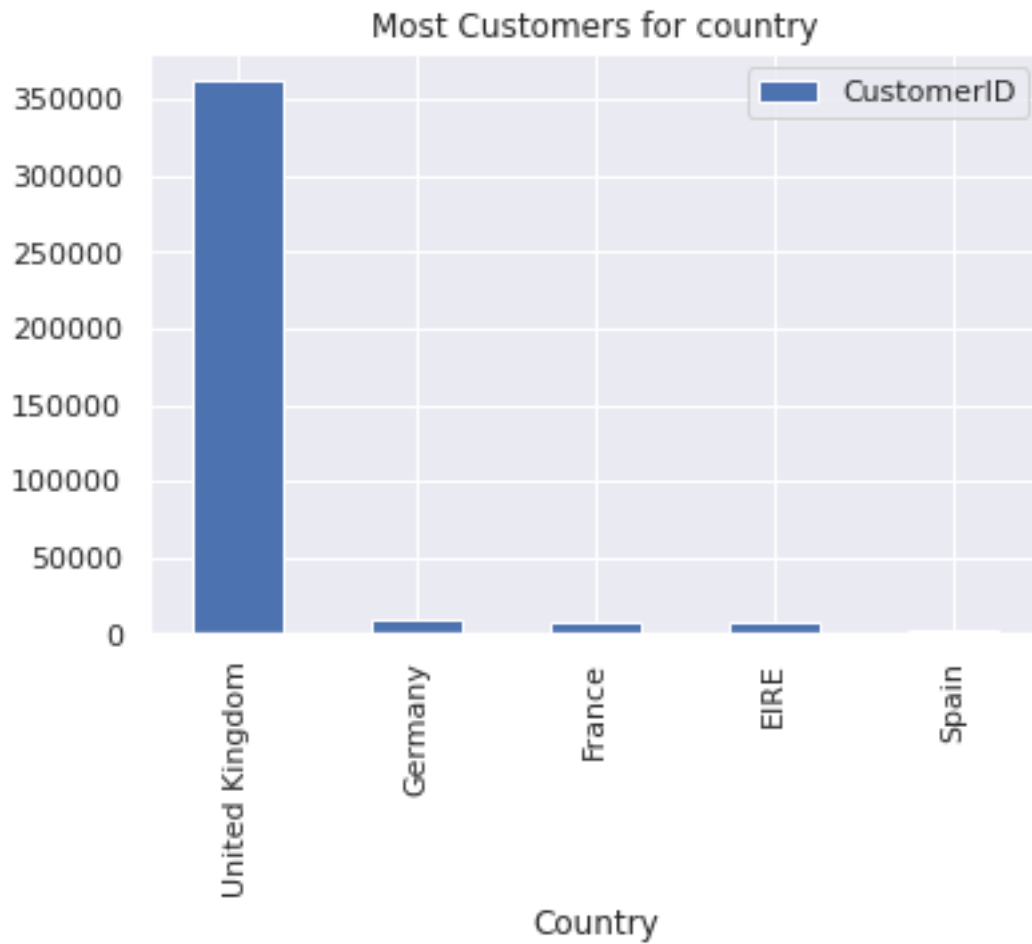
```
[77]: new_df = new_df.dropna()
new_df.isnull().sum()
```

```
[77]: CustomerID      0
InvoiceNo         0
StockCode         0
Quantity          0
TotalAmount       0
InvoiceDate       0
Country           0
Year              0
Mon               0
month             0
dtype: int64
```

```
[78]: #Countries with more number of customers
cus_id = pd.DataFrame(new_df.groupby('Country')['CustomerID'].count().
↪sort_values(ascending = False))
```

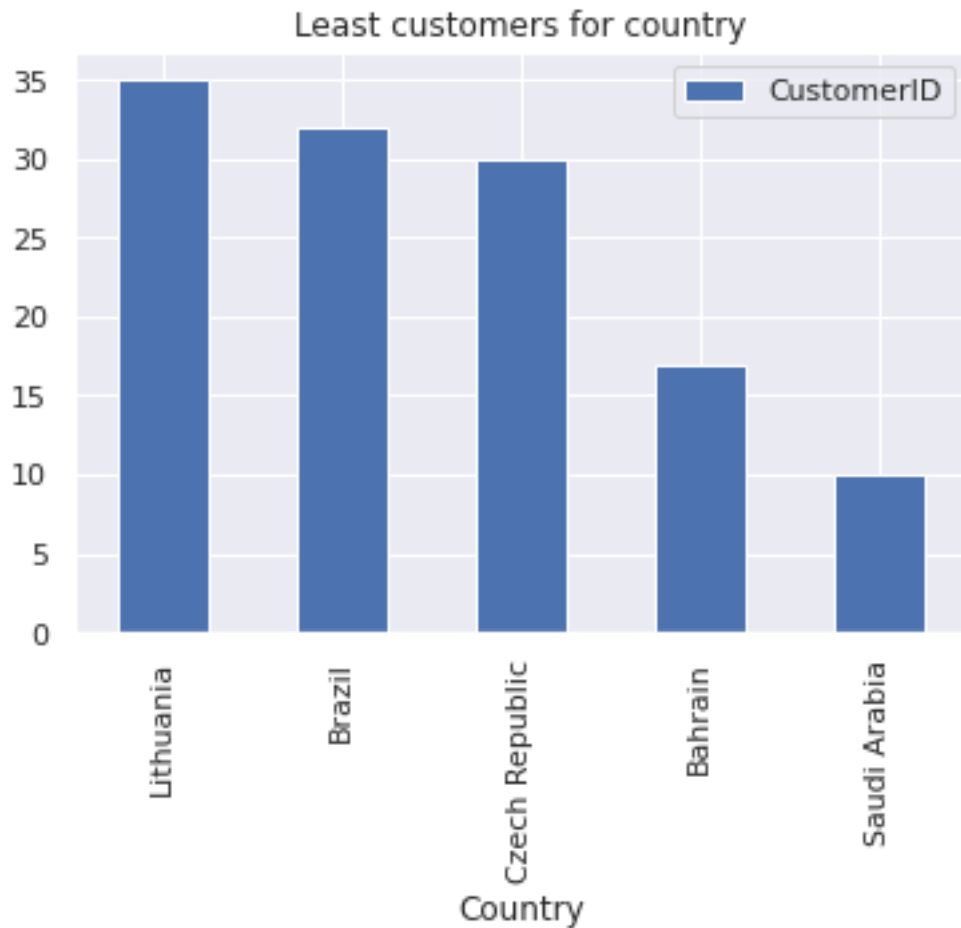
```
cus_id[:5].plot(kind = 'bar', title = 'Most Customers for country')
```

[78]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7f7d4aeb1b50>



```
[79]: # Countries with less number of customers
cus_id[-5:].plot(kind = 'bar', title = 'Least customers for country')
```

[79]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7f7d52018990>



## 5 Answer-

5.0.1 We see that september to december we have very high sales

We can concentrate on improving the sales for the other 8 months

We find very less number of customers in Lithuania, Brazil, Czech Republic, Bahrain, Saudi Arabia

We have very less sales for Lebanon, Brazil, RSA, Bahrain, Saudi Arabia.

We can concentrate on improving their sales

[ ]:

[ ]:

[ ]:

[ ]: