

## ASSIGNMENT NO: 6 (Group Activity)

By

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### Problem Statement

Using visualization(s) to tell a story with data

### Theory: Here is one way to start.

Step 1. Pick a domain and data set that you are interested in. <https://www.kaggle.com/> or any other repository Choose the one of greatest interest to you. If you would like to explore a different data set, you are free to do so but be aware of how much work might be needed to clean the data and get it into a usable format.

Step 2. Explore the data. Find a story. Ask questions. Start by asking questions. For example: Is there a relationship between melting point and atomic number? Are the brightness and color of stars correlated? Are there different patterns of nucleotides in different regions in human DNA?

Step 3. Assess the fitness of the data for answering your question. Inspect the data -- it is invariably helpful to first look at the raw values. Does the data seem appropriate for answering your question? If not, you may need to start the process over. If so, does the data need to be reformatted or cleaned prior to analysis? Perform any steps necessary to get the data into shape prior to visual analysis.

Step 4. Create the visualization(s) that tell a story about the data. You will likely need to create several and see what works best. Keep a record of things you tried and discarded.

**Input:** Link of dataset which you have used

<https://www.kaggle.com/hendraherviawan/customer-purchasing-patterns/data?select=data.csv>

Create a Doc/ppt that contains the following information:

### 1. Describe the data set you chose and why?

E-Commerce Data

Actual transactions from UK retailer

Now a days there are many purchase are happening in retail sector so we thought why not to understand customers behaviour towards buying products and their spending pattern.

Analyses for this dataset could include time series, clustering, classification and more.

### 2. What were the question(s) that you set out to answer?

How many orders made by the customers?

Check TOP 5 most number of orders

How much money spent by the customers?

Check TOP 5 highest money spent

How many orders (per month)?

How many orders (per day)?

How many orders (per hour)?

Discover patterns for Unit Price

Discover patterns for each Country

How many orders for each country?

How much money spent by each country?

### **3. What visualization tool(s)/programming did you use?**

We have used python programming language

### **4. Describe any transformations of the dataset that you needed to perform to get the data into the format needed by the visualization tool(s).**

Data Cleaning

Check missing values for each column

Remove rows with missing values

Remove Quantity with negative values

Add the column - amount\_spent

Add the columns - Month, Day and Hour for the invoice

### **5. Write a couple of paragraphs telling the story, describing the visualization, and saying how it answers the questions you posed.**

Added in code and in Result section.

### **6. Include screenshots and brief explanations of some of your draft work to show your process and how you got to your final visualization.**

**Output:** Link of your a story telling based on data.

## Business Intelligence Lab 6 ( Group Activity )

By

- Prasad Khandake PA-12
- Nikhil Munday PA-23
- Masoom Raza PA-17

## Customer Purchasing Patterns

### Context of Data

Company - UK-based and registered non-store online retail

Products for selling - Mainly all-occasion gifts

Customers - Most are wholesalers (local or international)

Transactions Period - **1st Dec 2010 - 9th Dec 2011** (One year)

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

import warnings
# current version of seaborn generates a bunch of warnings that we'll ignore
warnings.filterwarnings('ignore')
sns.set_style('whitegrid')

import missingno as msno # missing data visualization module for Python
import pandas_profiling

import gc
import datetime

%matplotlib inline
color = sns.color_palette()
```

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```
In [2]: pd.set_option('display.max_rows', 10000)
pd.set_option('display.max_columns', 100)
```

```
In [3]: # specify encoding to deal with different formats
df = pd.read_csv('C:/Users/prasa/customer_purchasing_patterns.csv', encoding = 'ISO-8859-1')
```

```
In [4]: df.head()
```

Out[4]:










	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country
0	536365	85123A	WHITE HANGING HEART T-LIGHT HOLDER	6	12/1/2010 8:26	2.55	17850.0	United Kingdom
1	536365	71053	WHITE METAL LANTERN	6	12/1/2010 8:26	3.39	17850.0	United Kingdom
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	12/1/2010 8:26	2.75	17850.0	United Kingdom
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	12/1/2010 8:26	3.39	17850.0	United Kingdom
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	12/1/2010 8:26	3.39	17850.0	United Kingdom

```
In [5]: # change the column names
df.rename(index=str, columns={'InvoiceNo': 'invoice_num',
                              'StockCode': 'stock_code',
                              'Description': 'description',
                              'Quantity': 'quantity',
                              'InvoiceDate': 'invoice_date',
                              'UnitPrice': 'unit_price',
                              'CustomerID': 'cust_id',
                              'Country': 'country'}, inplace=True)
```

```
In [6]: df.head()
```

Out[6]:

	invoice_num	stock_code	description	quantity	invoice_date	unit_price	cust_id	country
0	536365	85123A	WHITE HANGING HEART T-LIGHT HOLDER	6	12/1/2010 8:26	2.55	17850.0	United Kingdom
1	536365	71053	WHITE METAL LANTERN	6	12/1/2010 8:26	3.39	17850.0	United Kingdom
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	12/1/2010 8:26	2.75	17850.0	United Kingdom
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	12/1/2010 8:26	3.39	17850.0	United Kingdom
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	12/1/2010 8:26	3.39	17850.0	United Kingdom

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```
In [7]: df.info()

<class 'pandas.core.frame.DataFrame'>
Index: 541909 entries, 0 to 541908
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype  
---  --
0   invoice_num      541909 non-null  object 
1   stock_code       541909 non-null  object 
2   description       540455 non-null  object 
3   quantity         541909 non-null  int64  
4   invoice_date     541909 non-null  object 
5   unit_price       541909 non-null  float64 
6   cust_id          406829 non-null  float64 
7   country          541909 non-null  object 
dtypes: float64(2), int64(1), object(5)
memory usage: 37.2+ MB
```

## Check missing values for each column






```
In [8]: # check missing values for each column
df.isnull().sum().sort_values(ascending=False)
```

```
Out[8]: cust_id      135080
description  1454
invoice_num    0
stock_code     0
quantity       0
invoice_date   0
unit_price     0
country        0
dtype: int64
```

```
In [9]: # check out the rows with missing values
df[df.isnull().any(axis=1)].head()
```

```
Out[9]:
```

	invoice_num	stock_code	description	quantity	invoice_date	unit_price	cust_id	country
622	536414	22139		56	12/1/2010 11:52	0.00	NaN	United Kingdom
1443	536544	21773	DECORATIVE ROSE BATHROOM BOTTLE	1	12/1/2010 14:32	2.51	NaN	United Kingdom
1444	536544	21774	DECORATIVE CATS BATHROOM BOTTLE	2	12/1/2010 14:32	2.51	NaN	United Kingdom
1445	536544	21786	POLKADOT RAIN HAT	4	12/1/2010 14:32	0.85	NaN	United Kingdom
1446	536544	21787	RAIN PONCHO RETROSPOT	2	12/1/2010 14:32	1.66	NaN	United Kingdom

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```
1446    536544    21787    RAIN PONCHO RETROSPOT    2    12/1/2010 14:32    1.66    NaN    United Kingdom
```

```
In [10]: # change the invoice_date format - String to Timestamp format
df['invoice_date'] = pd.to_datetime(df.invoice_date, format='%m/%d/%Y %H:%M')
```

```
In [11]: # change description - UPPER case to LOWER case
df['description'] = df.description.str.lower()
```

```
In [12]: df.head()
```

```
Out[12]:
```

	invoice_num	stock_code	description	quantity	invoice_date	unit_price	cust_id	country
0	536365	85123A	white hanging heart t-light holder	6	2010-12-01 08:26:00	2.55	17850.0	United Kingdom
1	536365	71053	white metal lantern	6	2010-12-01 08:26:00	3.39	17850.0	United Kingdom
2	536365	84406B	cream cupid hearts coat hanger	8	2010-12-01 08:26:00	2.75	17850.0	United Kingdom
3	536365	84029G	knitted union flag hot water bottle	6	2010-12-01 08:26:00	3.39	17850.0	United Kingdom
4	536365	84029E	red woolly hottie white heart.	6	2010-12-01 08:26:00	3.39	17850.0	United Kingdom

## Remove rows with missing values

```
In [13]: # df_new without missing values
df_new = df.dropna()
```

```
In [14]: # check missing values for each column
df_new.isnull().sum().sort_values(ascending=False)
```

```
Out[14]: invoice_num    0
stock_code    0
description    0
quantity      0
invoice_date   0
unit_price     0
cust_id        0
country        0
dtype: int64
```

In [15]: `df_new.info()`

```
<class 'pandas.core.frame.DataFrame'>
Index: 406829 entries, 0 to 541908
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  -
0   invoice_num      406829 non-null  object
1   stock_code       406829 non-null  object
2   description       406829 non-null  object
3   quantity         406829 non-null  int64
4   invoice_date     406829 non-null  datetime64[ns]
5   unit_price       406829 non-null  float64
6   cust_id          406829 non-null  float64
7   country          406829 non-null  object
dtypes: datetime64[ns](1), float64(2), int64(1), object(4)
memory usage: 27.9+ MB
```

In [16]: `# change columns tyoe - String to Int type`  
`df_new['cust_id'] = df_new['cust_id'].astype('int64')`

In [17]: `df_new.head()`

Out[17]:

	invoice_num	stock_code	description	quantity	invoice_date	unit_price	cust_id	country
0	536365	85123A	white hanging heart t-light holder	6	2010-12-01 08:26:00	2.55	17850	United Kingdom
1	536365	71053	white metal lantern	6	2010-12-01 08:26:00	3.39	17850	United Kingdom
2	536365	84406B	cream cupid hearts coat hanger	8	2010-12-01 08:26:00	2.75	17850	United Kingdom
3	536365	84029G	knitted union flag hot water bottle	6	2010-12-01 08:26:00	3.39	17850	United Kingdom
4	536365	84029E	red woolly hottie white heart.	6	2010-12-01 08:26:00	3.39	17850	United Kingdom

In [18]: `df_new.info()`

```
<class 'pandas.core.frame.DataFrame'>
Index: 406829 entries, 0 to 541908
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  -
0   invoice_num      406829 non-null  object
1   stock_code       406829 non-null  object
2   description       406829 non-null  object
3   quantity         406829 non-null  int64
4   invoice_date     406829 non-null  datetime64[ns]
5   unit_price       406829 non-null  float64
6   cust_id          406829 non-null  int64
7   country          406829 non-null  object
dtypes: datetime64[ns](1), float64(2), int64(2), object(3)
memory usage: 27.0+ MB
```

In [19]: `df_new.describe().round(2)`

Out[19]:

	quantity	unit_price	cust_id
count	406829.00	406829.00	406829.00
mean	12.06	3.46	15287.69
std	248.69	69.32	1713.60
min	-80995.00	0.00	12346.00
25%	2.00	1.25	13953.00
50%	5.00	1.95	15152.00
75%	12.00	3.75	16791.00
max	80995.00	38970.00	18287.00

## Remove Quantity with negative values

In [20]: `df_new = df_new[df_new.quantity > 0]`

In [21]: `df_new.describe().round(2)`

Out[21]:

	quantity	unit_price	cust_id
count	397924.00	397924.00	397924.00
mean	13.02	3.12	15294.32
std	180.42	22.10	1713.17
min	1.00	0.00	12346.00
25%	2.00	1.25	13969.00
50%	6.00	1.95	15159.00
75%	12.00	3.75	16795.00
max	80995.00	8142.75	18287.00

## Add the column - amount\_spent

```
In [22]: df_new['amount_spent'] = df_new['quantity'] * df_new['unit_price']
```

```
In [23]: # rearrange all the columns for easy reference
df_new = df_new[['invoice_num', 'invoice_date', 'stock_code', 'description', 'quantity', 'unit_price', 'amount_spent', 'cust_id', 'country']]
```

## Add the columns - Month, Day and Hour for the invoice

```
In [24]: df_new.insert(loc=2, column='year_month', value=df_new['invoice_date'].map(lambda x: 100*x.year + x.month))
df_new.insert(loc=3, column='month', value=df_new.invoice_date.dt.month)
# +1 to make Monday=1....until Sunday=7
df_new.insert(loc=4, column='day', value=(df_new.invoice_date.dt.dayofweek)+1)
df_new.insert(loc=5, column='hour', value=df_new.invoice_date.dt.hour)
```

```
In [25]: df_new.head()
```

```
Out[25]:
```

	invoice_num	invoice_date	year_month	month	day	hour	stock_code	description	quantity	unit_price	amount_spent	cust_id	country
0	536365	2010-12-01 08:26:00	201012	12	3	8	85123A	white hanging heart t-light holder	6	2.55	15.30	17850	United Kingdom
1	536365	2010-12-01 08:26:00	201012	12	3	8	71053	white metal lantern	6	3.39	20.34	17850	United Kingdom
2	536365	2010-12-01 08:26:00	201012	12	3	8	84406B	cream cupid hearts coat hanger	8	2.75	22.00	17850	United Kingdom
3	536365	2010-12-01 08:26:00	201012	12	3	8	84029G	knitted union flag hot water bottle	6	3.39	20.34	17850	United Kingdom
4	536365	2010-12-01 08:26:00	201012	12	3	8	84029E	red woolly hottie white heart.	6	3.39	20.34	17850	United Kingdom

## Exploratory Data Analysis (EDA)

### How many orders made by the customers?

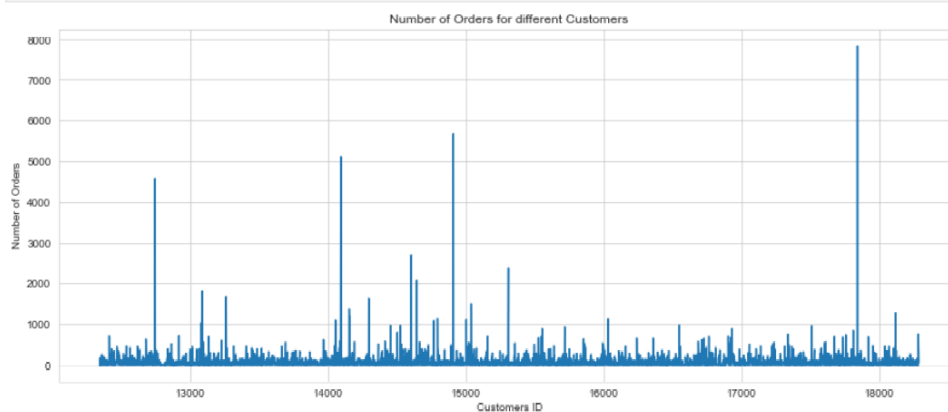
```
In [26]: df_new.groupby(by=['cust_id', 'country'], as_index=False)['invoice_num'].count().head()
```

```
Out[26]:
```

	cust_id	country	invoice_num
0	12346	United Kingdom	1
1	12347	Iceland	182
2	12348	Finland	31
3	12349	Italy	73
4	12350	Norway	17

```
In [27]: orders = df_new.groupby(by=['cust_id', 'country'], as_index=False)['invoice_num'].count()
```

```
plt.subplots(figsize=(15,6))
plt.plot(orders.cust_id, orders.invoice_num)
plt.xlabel('Customers ID')
plt.ylabel('Number of Orders')
plt.title('Number of Orders for different Customers')
plt.show()
```



## Check TOP 5 most number of orders

```
In [28]: print('The TOP 5 customers with most number of orders...')
orders.sort_values(by='invoice_num', ascending=False).head()
```

The TOP 5 customers with most number of orders...

```
Out[28]:
```

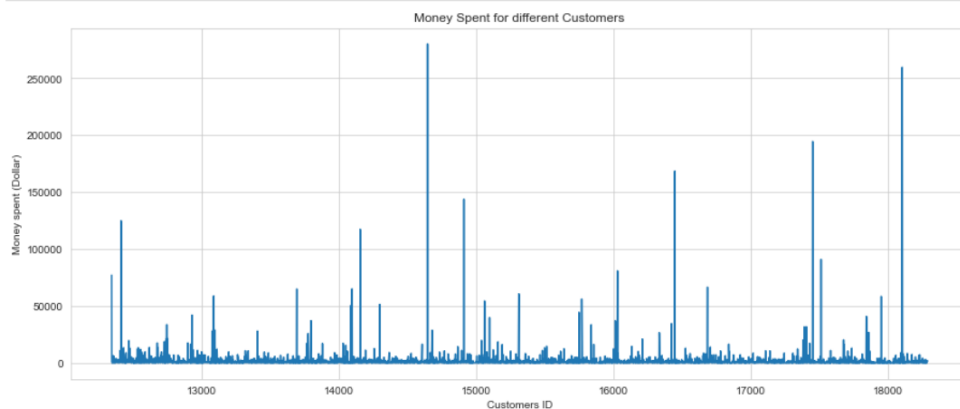
	cust_id	country	invoice_num
4019	17841	United Kingdom	7847
1888	14911	EIRE	5677
1298	14096	United Kingdom	5111
334	12748	United Kingdom	4596
1670	14606	United Kingdom	2700

## How much money spent by the customers?

```
In [29]: money_spent = df_new.groupby(by=['cust_id', 'country'], as_index=False)['amount_spent'].sum()

plt.subplots(figsize=(15,6))
plt.plot(money_spent.cust_id, money_spent.amount_spent)
plt.xlabel('Customers ID')
plt.ylabel('Money spent (Dollar)')
plt.title('Money Spent for different Customers')
plt.show()
```

```
plt.ylabel('Money spent (Dollar)')
plt.title('Money Spent for different Customers')
plt.show()
```



## Check TOP 5 highest money spent

```
In [30]: print('The TOP 5 customers with highest money spent...')
money_spent.sort_values(by='amount_spent', ascending=False).head()
```

The TOP 5 customers with highest money spent...

```
Out[30]:
```


	cust_id	country	amount_spent
1698	14646	Netherlands	280206.02
4210	18102	United Kingdom	259657.30
3737	17450	United Kingdom	194550.79
3017	16446	United Kingdom	168472.50
1888	14911	EIRE	143825.06

## Discover patterns for Number of Orders

• <https://www.kaggle.com/hendraheriawan/customer-purchasing-patterns>

### How many orders (per month)?

```
In [31]: # color available
sns.palplot(color)
```

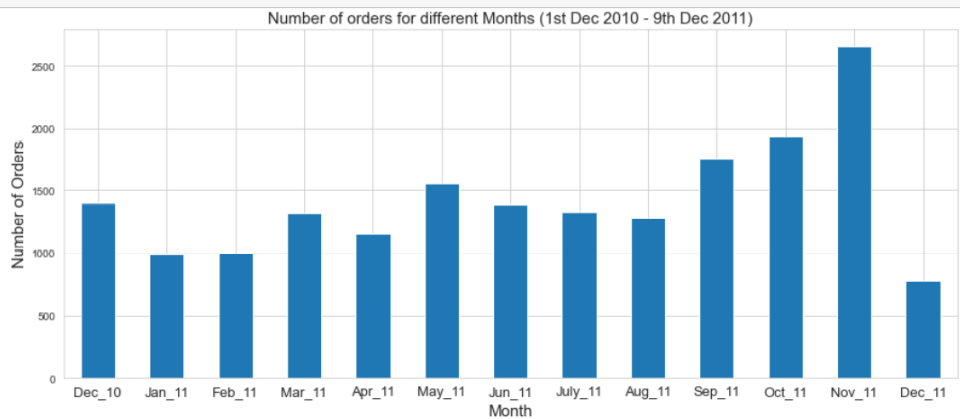


```
In [32]: df_new.head()
```

```
Out[32]:
```

	invoice_num	invoice_date	year_month	month	day	hour	stock_code	description	quantity	unit_price	amount_spent	cust_id	country
0	536365	2010-12-01 08:26:00	201012	12	3	8	85123A	white hanging heart t-light holder	6	2.55	15.30	17850	United Kingdom
1	536365	2010-12-01 08:26:00	201012	12	3	8	71053	white metal lantern	6	3.39	20.34	17850	United Kingdom
2	536365	2010-12-01 08:26:00	201012	12	3	8	84406B	cream cupid hearts coat hanger	8	2.75	22.00	17850	United Kingdom
3	536365	2010-12-01 08:26:00	201012	12	3	8	84029G	knitted union flag hot water bottle	6	3.39	20.34	17850	United Kingdom
4	536365	2010-12-01 08:26:00	201012	12	3	8	84029E	red woolly hottie white heart	6	3.39	20.34	17850	United Kingdom

```
In [33]: ax = df_new.groupby('invoice_num')['year_month'].unique().value_counts().sort_index().plot(kind='bar',color=color[0],figsize=(15,6))
ax.set_xlabel('Month',fontsize=15)
ax.set_ylabel('Number of Orders',fontsize=15)
ax.set_title('Number of orders for different Months (1st Dec 2010 - 9th Dec 2011)',fontsize=15)
ax.set_xticklabels(['Dec_10','Jan_11','Feb_11','Mar_11','Apr_11','May_11','Jun_11','July_11','Aug_11','Sep_11','Oct_11','Nov_11','Dec_11'], rotation='horizontal', fontsize=13)
plt.show()
```



### How many orders (per day)?

```
In [52]: df_new.groupby('invoice_num')['day'].unique().value_counts().sort_index()
```

```
Out[52]:
```

[1]	2863
[2]	3185
[3]	3455
[4]	4033
[5]	2831
[7]	2169

Name: day, dtype: int64

```
-----
TypeError                                Traceback (most recent call last)
pandas\libs\hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.map_locations()

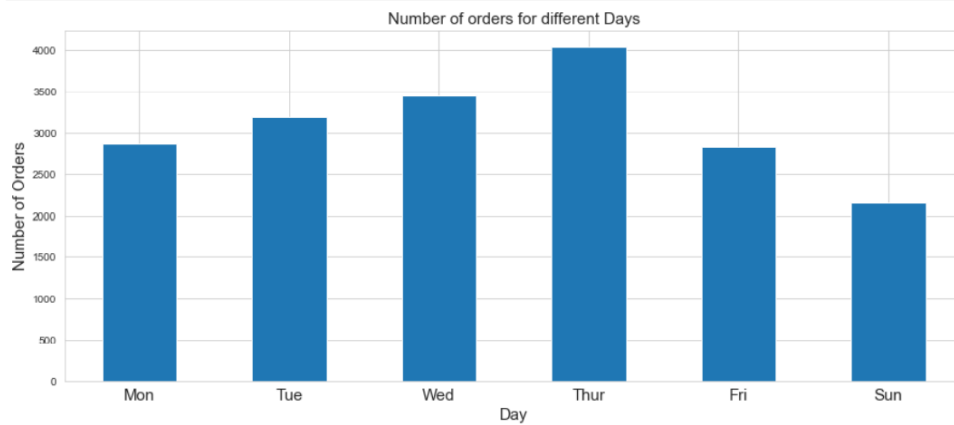
TypeError: unhashable type: 'numpy.ndarray'
Exception ignored in: 'pandas._libs.index.IndexEngine._call_map_locations'
Traceback (most recent call last):
  File "pandas\libs\hashtable_class_helper.pxi", line 4588, in pandas._libs.hashtable.PyObjectHashTable.map_locations
  TypeError: unhashable type: 'numpy.ndarray'
```



```

In [35]: ax = df_new.groupby('invoice_num')['day'].unique().value_counts().sort_index().plot(kind='bar',color=color[0],figsize=(15,6))
ax.set_xlabel('Day',fontsize=15)
ax.set_ylabel('Number of Orders',fontsize=15)
ax.set_title('Number of orders for different Days',fontsize=15)
ax.set_xticklabels(('Mon','Tue','Wed','Thur','Fri','Sun'), rotation='horizontal', fontsize=15)
plt.show()

```



## How many orders (per hour)?

```

In [36]: #df_new.groupby('invoice_num')['hour'].value_counts().iloc[: -1].sort_index()

In [37]: df_new.groupby('invoice_num')['hour'].agg('count').value_counts().iloc[: -1].sort_index()
# .unique()

```

```

Out[37]:
1      1392
2       735
3       638
4       626
5       659
6       595
7       581
8       588
9       589
10      525
11      540
12      480
13      500
14      503
15      531
16      555
17      444
18      423
19      472
20      428
21      393
22      332
23      348
24      312
25      240
26      236
27      227
28      219
29      278
30      209
31      192
32      173
33      160
34      157
35      135
36      116
37      128
38      114

```

```

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In [39]: ax = df_new.groupby('invoice_num')['hour'].agg('count').value_counts().iloc[:1].sort_index().head(20).plot(kind='bar', color='color[0]', figsize=(15,6))
ax.set_xlabel('Hour',fontsize=15)
ax.set_ylabel('Number of Orders',fontsize=15)
ax.set_title('Number of orders for different Hours',fontsize=15)
#ax.set_xticklabels(range(6,21), rotation='horizontal', Fontsize=15)
plt.show()
#

```



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```

## Discover patterns for Unit Price

```
In [40]: df_new.unit_price.describe()
```

```

Out[40]: count    397924.000000
mean         3.116174
std          22.096788
min           0.000000
25%          1.250000
50%          1.950000
75%          3.750000
max         8142.750000
Name: unit_price, dtype: float64

```

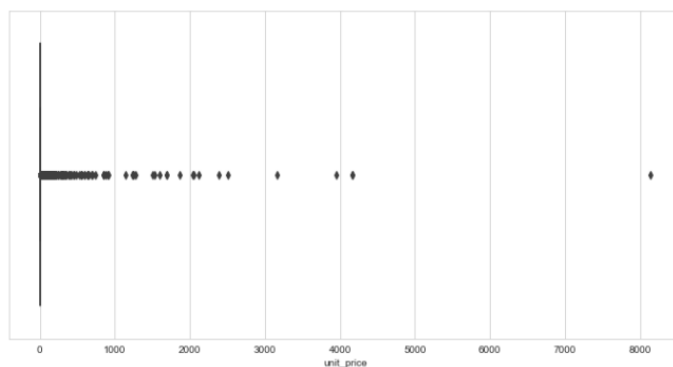
**We see that there are unit price = 0 (FREE items)**

There are some free items given to customers from time to time.

```

In [41]: # check the distribution of unit price
plt.subplots(figsize=(12,6))
sns.boxplot(df_new.unit_price)
plt.show()

```



In [42]:

In [43]:

Out[43]:

In [44]:

Out[44]:

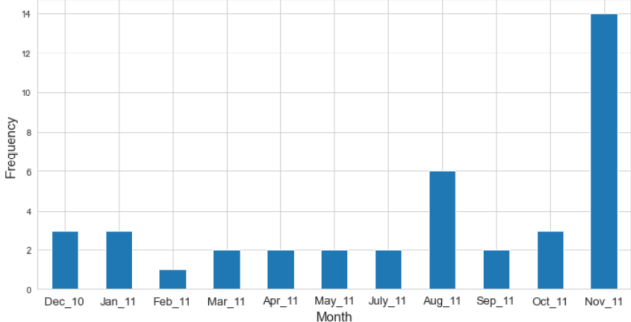
```

201012      3
201101      3
201102      1
201103      2
201104      2
201105      2
201107      2
201108      6
201109      2
201110      3
201111     14
Name: year_month, dtype: int64

```

```
name: year_month, dtype: int64
```

In [45]



Not clear why there are FREE items given to certain customers

- On average, the company gave out 2-4 times FREE items to customers each month (Except in June 2011)

## Discover patterns for each Country

In [46]

Out[46]

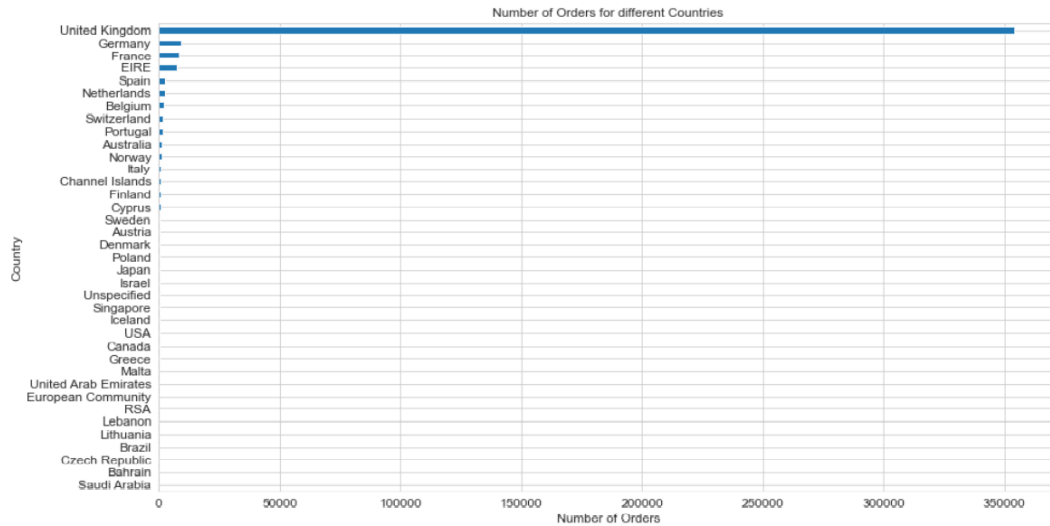
	invoice_num	invoice_date	year_month	month	day	hour	stock_code	description	quantity	unit_price	amount_spent	cust_id	country
0	536365	2010-12-01 08:26:00	201012	12	3	8	85123A	white hanging heart t-light holder	6	2.55	15.30	17850	United Kingdom
1	536365	2010-12-01 08:26:00	201012	12	3	8	71053	white metal lantern	6	3.39	20.34	17850	United Kingdom
2	536365	2010-12-01 08:26:00	201012	12	3	8	84406B	cream cupid hearts coat hanger	6	2.75	22.00	17850	United Kingdom
3	536365	2010-12-01 08:26:00	201012	12	3	8	84029G	knitted union flag hot water bottle	6	3.39	20.34	17850	United Kingdom
4	536365	2010-12-01 08:26:00	201012	12	3	8	84029E	red woolly tooth white heart.	6	3.39	20.34	17850	United Kingdom

## How many orders for each country?

In [47]:

```
group_country_orders = df_new.groupby('country')['invoice_num'].count().sort_values()
# del group_country_orders['United Kingdom']

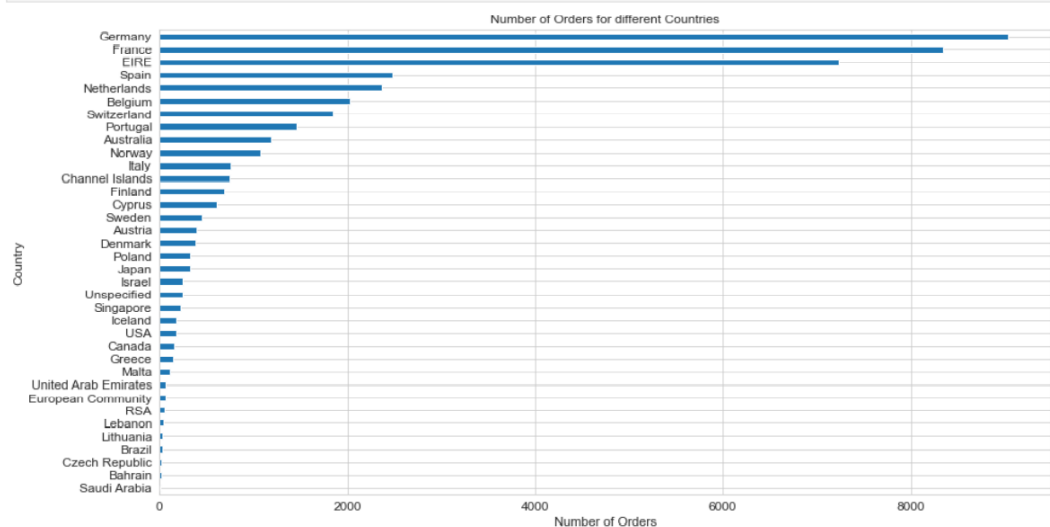
# plot number of unique customers in each country (with UK)
plt.subplots(figsize=(15,8))
group_country_orders.plot(kind='barh', fontsize=12, color=color[0])
plt.xlabel('Number of Orders', fontsize=12)
plt.ylabel('Country', fontsize=12)
plt.title('Number of Orders for different Countries', fontsize=12)
plt.show()
```



In [48]:

```
group_country_orders = df_new.groupby('country')['invoice_num'].count().sort_values()
del group_country_orders['United Kingdom']

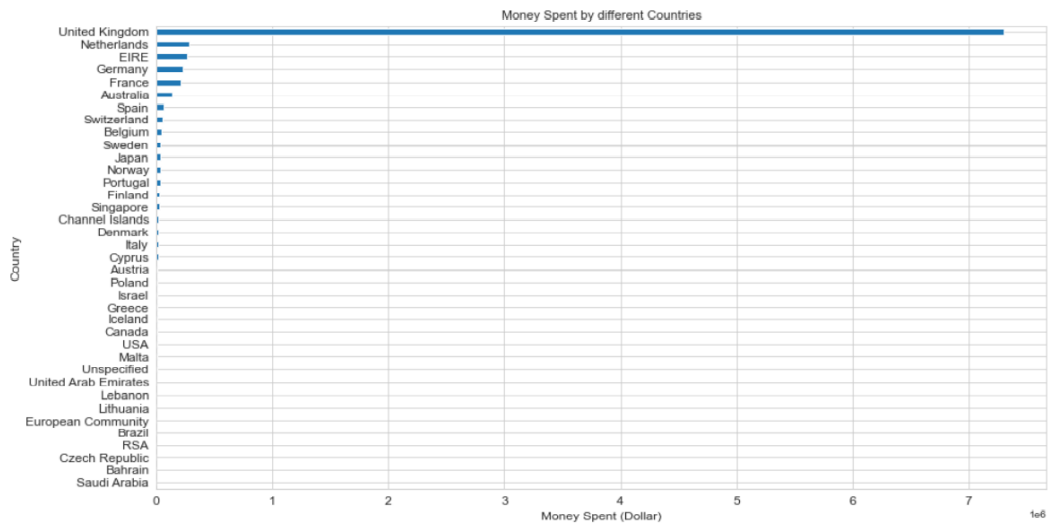
# plot number of unique customers in each country (without UK)
plt.subplots(figsize=(15,8))
group_country_orders.plot(kind='barh', fontsize=12, color=color[0])
plt.xlabel('Number of Orders', fontsize=12)
plt.ylabel('Country', fontsize=12)
plt.title('Number of Orders for different Countries', fontsize=12)
plt.show()
```



## How much money spent by each country?

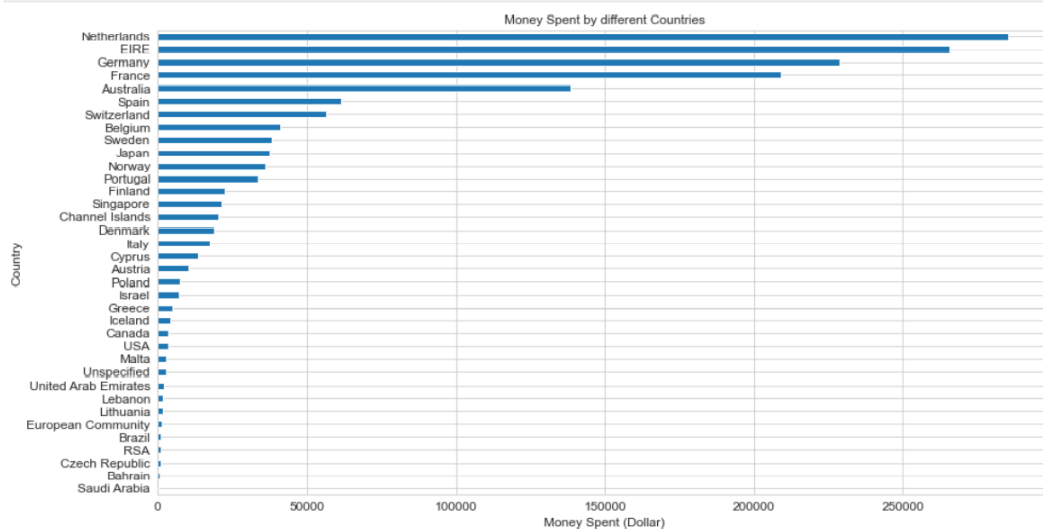
```
In [49]: group_country_amount_spent = df_new.groupby('country')['amount_spent'].sum().sort_values()
# del group_country_orders['United Kingdom']

# plot number of unique customers in each country (with UK)
plt.subplots(figsize=(15,8))
group_country_amount_spent.plot(kind='barh', fontsize=12, color=color[0])
plt.xlabel('Money Spent (Dollar)', fontsize=12)
plt.ylabel('Country', fontsize=12)
plt.title('Money Spent by different Countries', fontsize=12)
plt.show()
```



```
In [50]: group_country_amount_spent = df_new.groupby('country')['amount_spent'].sum().sort_values()
del group_country_amount_spent['United Kingdom']

# plot number of unique customers in each country (without UK)
plt.subplots(figsize=(15,8))
group_country_amount_spent.plot(kind='barh', fontsize=12, color=color[0])
plt.xlabel('Money Spent (Dollar)', fontsize=12)
plt.ylabel('Country', fontsize=12)
plt.title('Money Spent by different Countries', fontsize=12)
plt.show()
```



### Results obtained from Exploratory Data Analysis (EDA)

1. The **customer with the highest number of orders comes from the United Kingdom (UK)**
2. The **customer with the highest money spent on purchases comes from Netherlands**
3. The company receives the highest number of orders from customers in the UK (since it is a UK-based company). Therefore, the TOP 5 countries (including UK) that place the highest number of orders are as below:
  - United Kingdom
  - Germany
  - France
  - Ireland (EIRE)
  - Spain
4. As the company receives the highest number of orders from customers in the UK (since it is a UK-based company), customers in the UK spend the most on their purchases. Therefore, the TOP 5 countries (including UK) that spend the most money on purchases are as below:
  - United Kingdom
  - Netherlands
  - Ireland (EIRE)
  - Germany
  - France
5. **November 2011 has the highest sales**
  - The month with the lowest sales is undetermined as the dataset consists of transactions until 9th December 2011 in December
6. There are **no transactions on Saturday** between 1st Dec 2010 - 0th Dec 2011
7. The number of orders received by the company tends to increase from Monday to Thursday and decrease afterward
8. The company receives the **highest number of orders at 12:00pm**
  - Possibly most customers made purchases during **lunch hour between 12:00pm - 2:00pm**
9. The company tends to **give out FREE items for purchases occasionally each month**
  - However, it is not clear what factors contribute to giving out the FREE items to the particular customers

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9. The company tends to **\*\*give out FREE items for purchases occasionally each month\*\***

- However, it is not clear what factors contribute to giving out the FREE items to the particular customers.

### **Conclusion:**

Hence, understood better the process of using visualizations to tell a story about a data set.