

E-commerce Dataset Exploration

Using R & Tableau

Purpose of exploration: To extract any actionable and relevant data from data sample and provide findings and recommendations.

About the data: This dataset appears to be from an Orders database table. It contained 541909 records for 8 fields -

- InvoiceNo: *Order number*
- StockCode: *Item stock code*
- Description: *Item description*
- Quantity: *Quantity of item*
- InvoiceDate: *Date of order - with Day/Month written in USA format*
- UnitPrice: *Price of item*
- CustomerID: *Customer Identifier*
- Country: *Order country*

Added:

- Total Amount: *Quantity of item * Unit Price*
- Month: *Month of order*
- Day of the Week: *Day of order*
- Year: *Year of order*
- Time of the Day: *Hour of day of order*

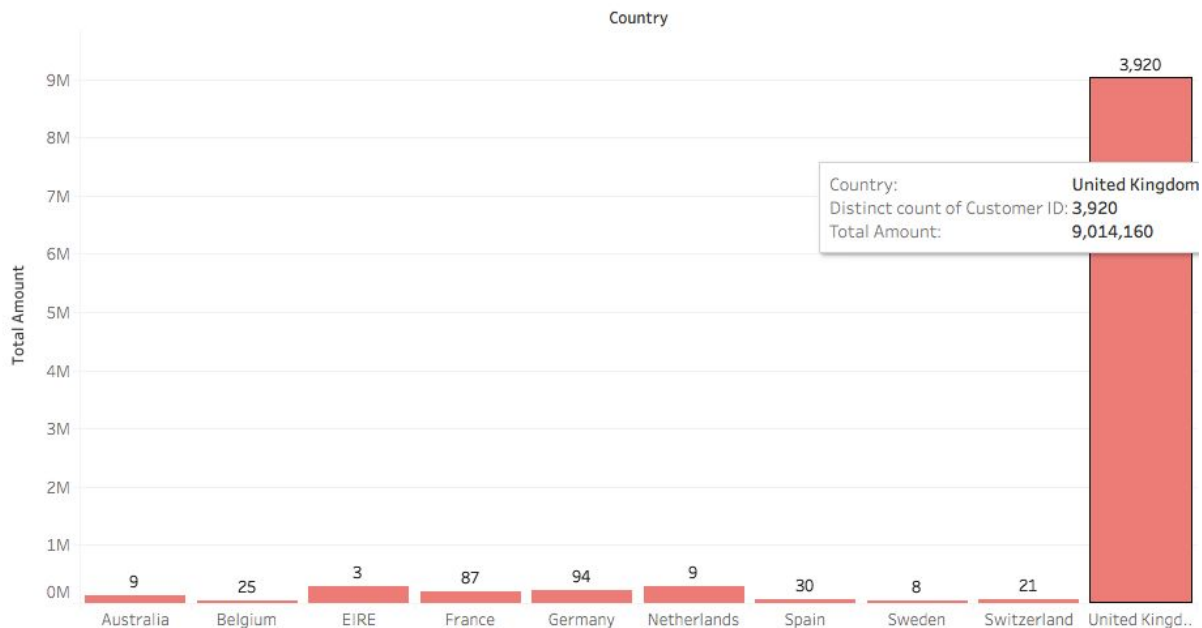
This document covers some of the Findings (1A-F), a Summary (2) and Data Pre-processing steps (3).

1. Findings

A. Most Revenue Generating Countries - Top 10

From the 34 unique Country values in the dataset - this diagram displays the Top 10 per revenue. The United Kingdom is by far the most revenue generating country.

Top 10 Countries Per Revenue Generated -with unique Customer Count



NOTE: The number on top of each block represents the number of unique customer in which that revenue was generated.

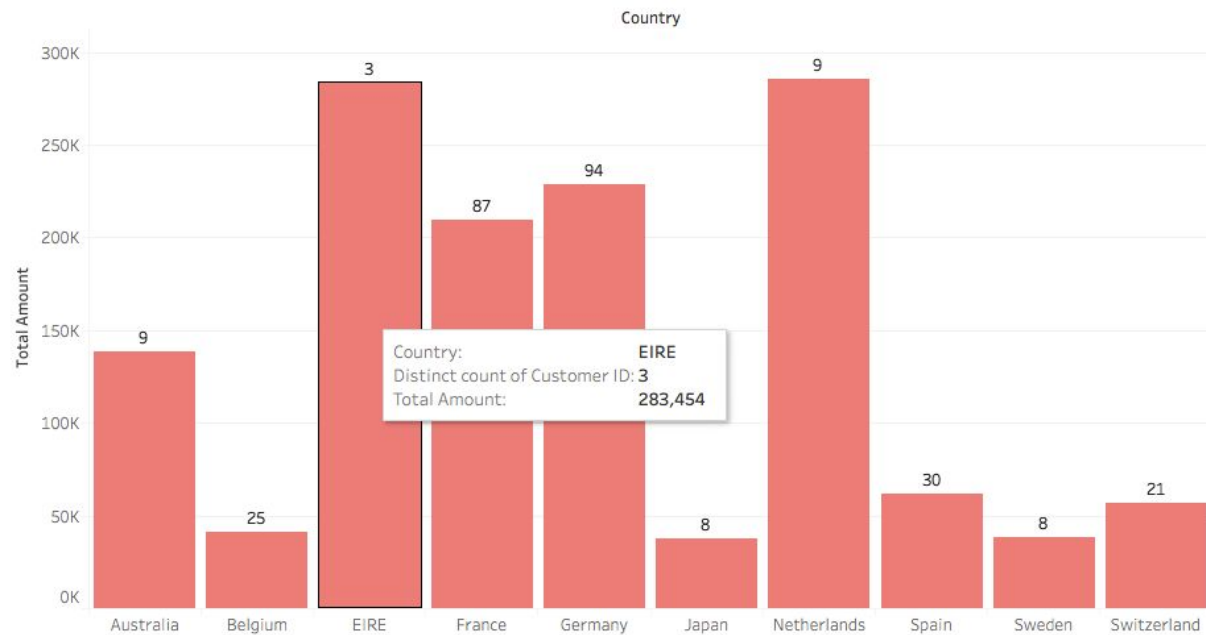
For the UK - this means an average of approx \$2,299 per customer.

Now excluding the UK - taking a closer look at the Top 10, in the diagram below.

We can see that Eire(Ireland) and the Netherlands are the next two most revenue generating - this is especially interesting as they have a very low amount of customers for this revenue. Ireland has just 3 and the Netherlands has 9.

I noted that Eire has approx. 100 CustomerIDs missing.

Top 10 Countries Per Revenue Generated -with unique Customer Count



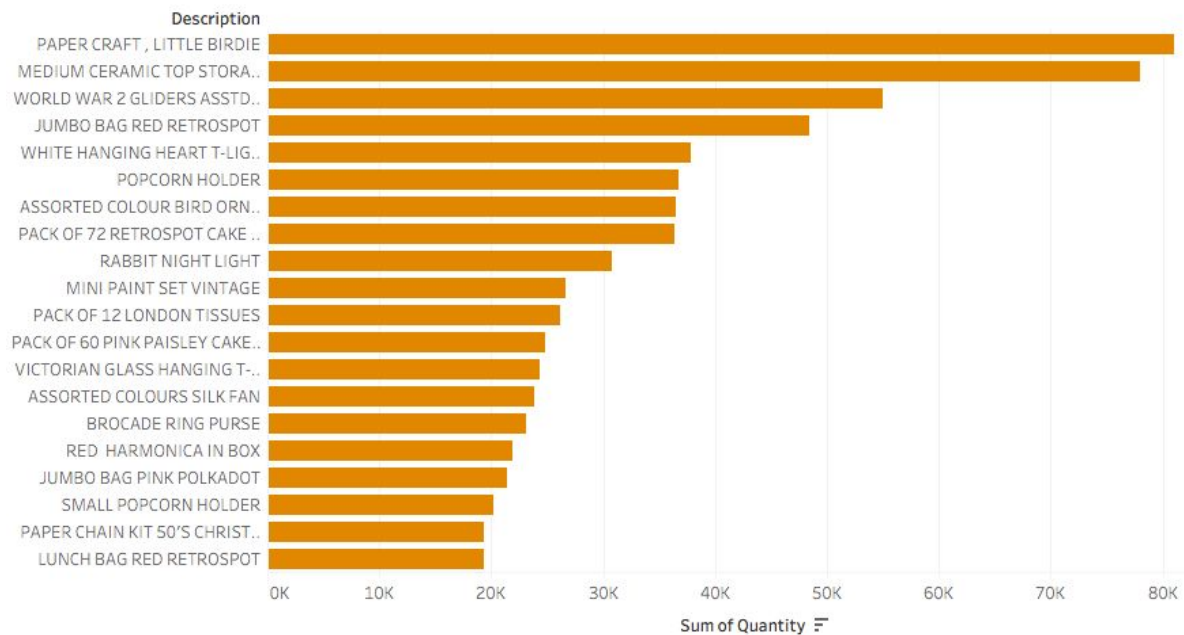
Based on these results - a small number of customers can make up a large amount of Revenue. Customer Retention for these countries could be a focus - and also looking for new customers who plan large orders.

B. Most Popular Products - Top 20

Purely looking at which Products have been sold the most by Quantity - the diagram below shows:

'PAPER CRAFT, LITTLE BIRDIE' and 'MEDIUM CERAMIC TOP STORAGE JAR' are the most Popular Products - way ahead of the other Top 18.

Top 20 Most Popular Products - by Quantity

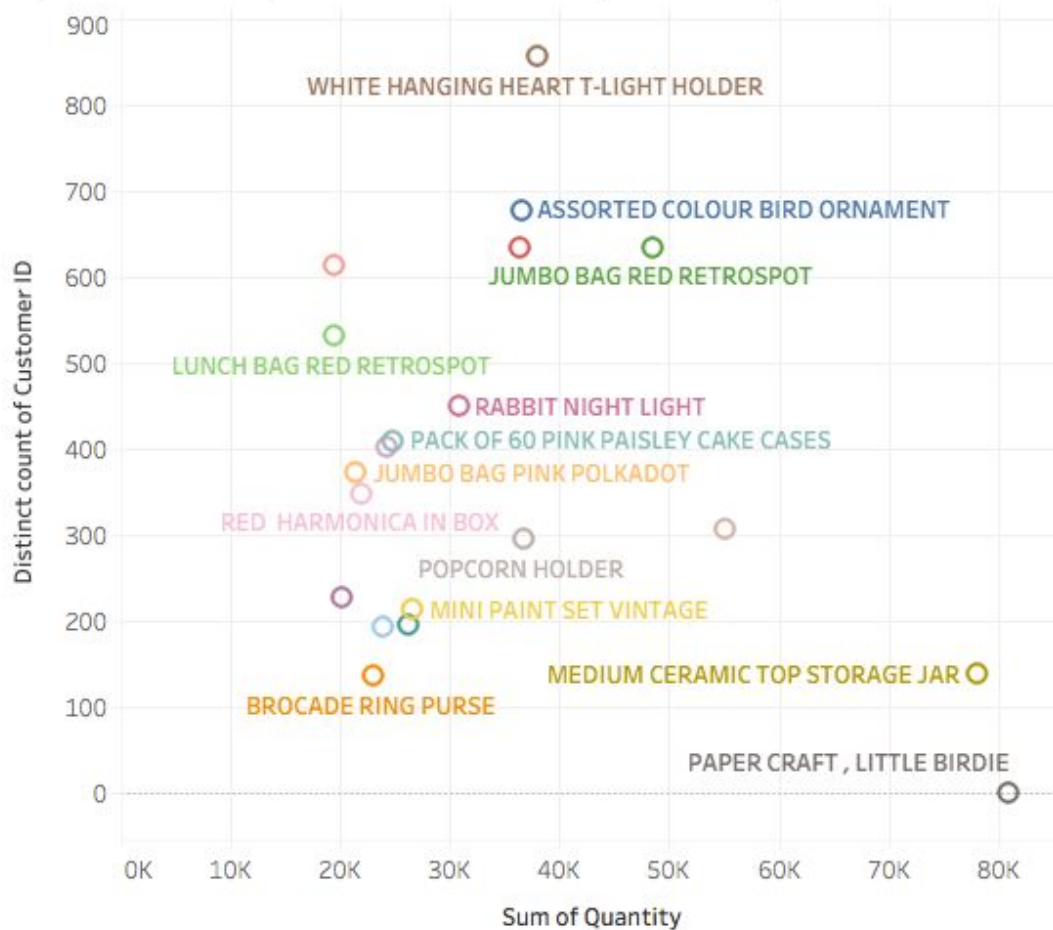


When another dimension is added, unique CustomerID, the data is represented quite differently. It appears that 'PAPER CRAFT, LITTLE BIRDIE' is a complete outlier - as there is only 1 unique CustomerID. 'MEDIUM CERAMIC TOP STORAGE JAR' also has a low customer base.

The most popular items per unique customers orders are:

- WHITE HANGING HEART T-LIGHT HOLDER
- ASSORTED COLOUR BIRD ORNAMENT
- JUMBO BAG RED RETROSPOT

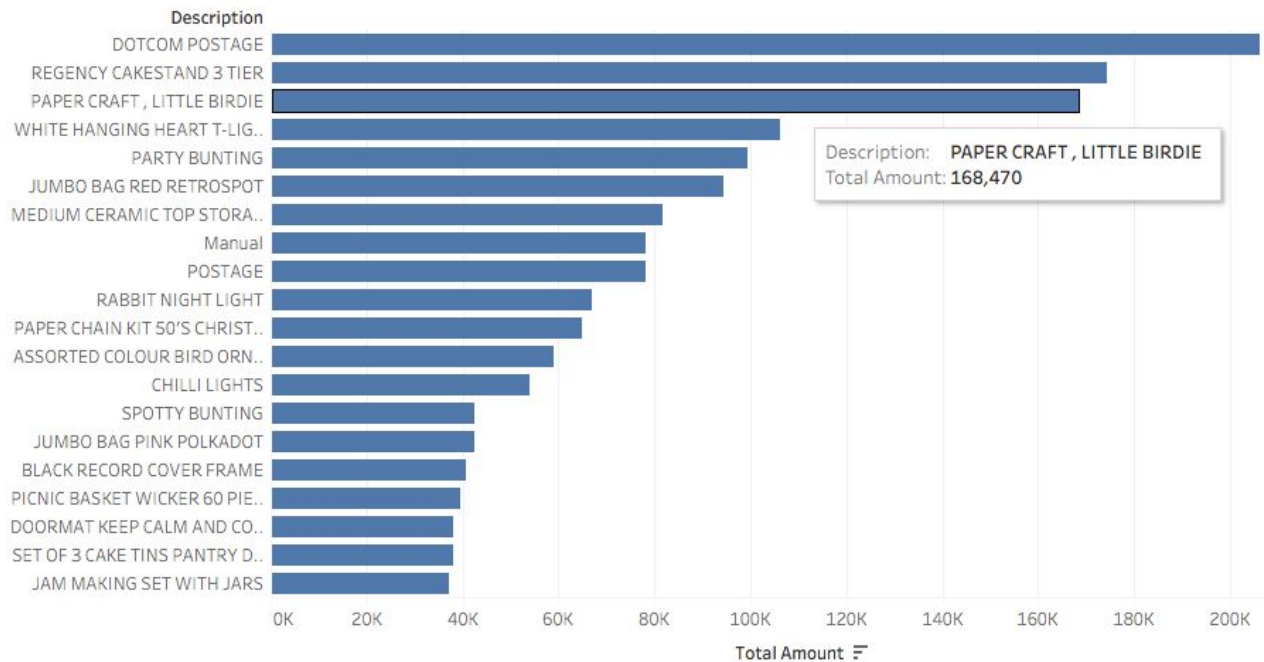
Top 20 Most Popular Products - by Quantity and CustomerID



Knowing your most popular products and the average quantity of sales per week/month can allow you to predict necessary stock levels.

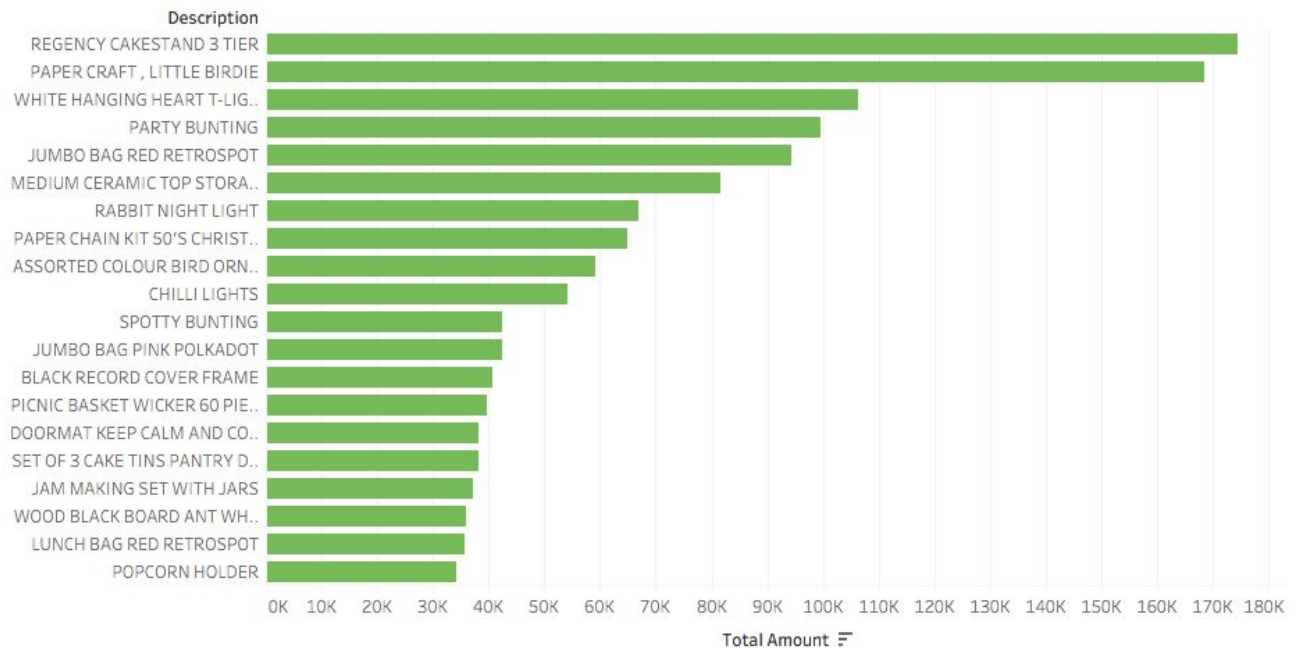
C. Most Profitable Products - Top 20

Top 20 Most Profitable Products



In the diagram above - 'DOTCOM POSTAGE', 'Manual' and 'POSTAGE' do not sound like actual products that the business sells. Therefore I excluded them for the diagram below. This can be easily adjusted once these products are confirmed as genuine.

Top 20 Most Profitable Products

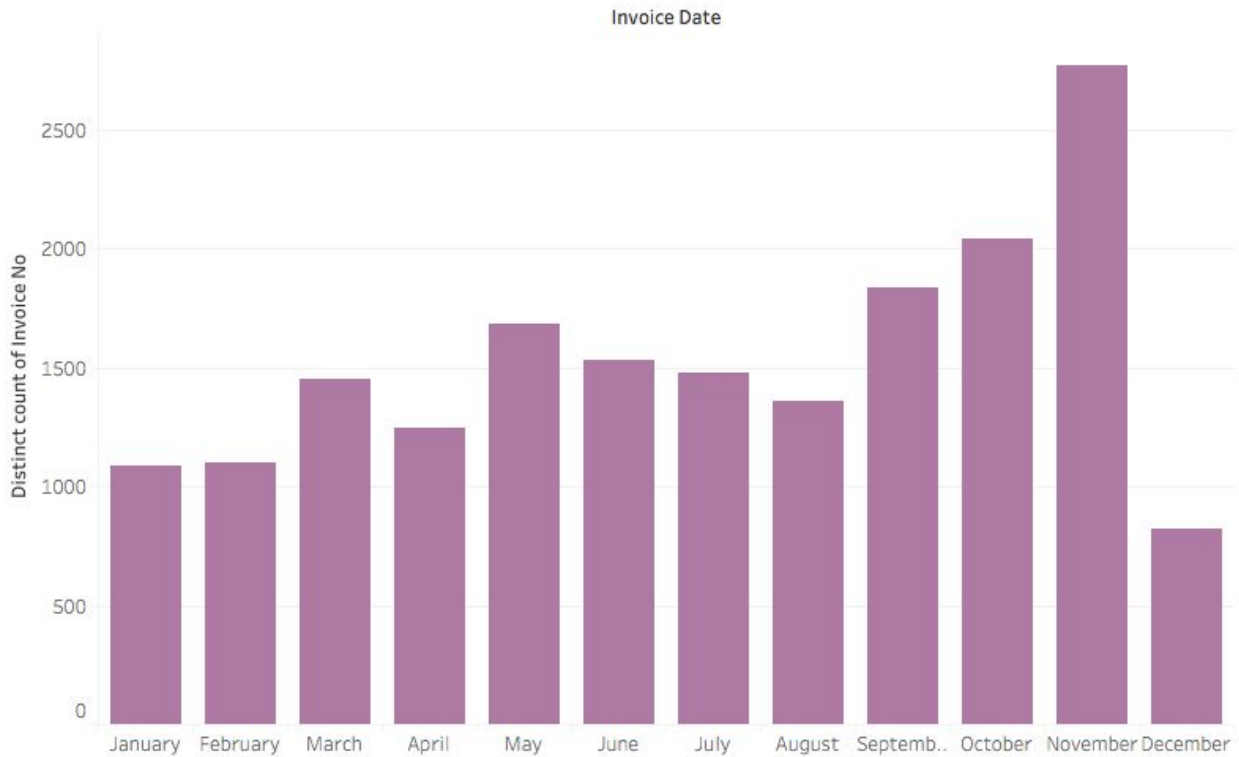


'REGENCY CAKESTAND 3 TIER' is the most profitable Product.

D. Busiest Time of the Year - Seasonality check

September, October and November show an upward trend in quantity of Orders. As these Products may be gifts - this makes sense with gift giving season being in December.

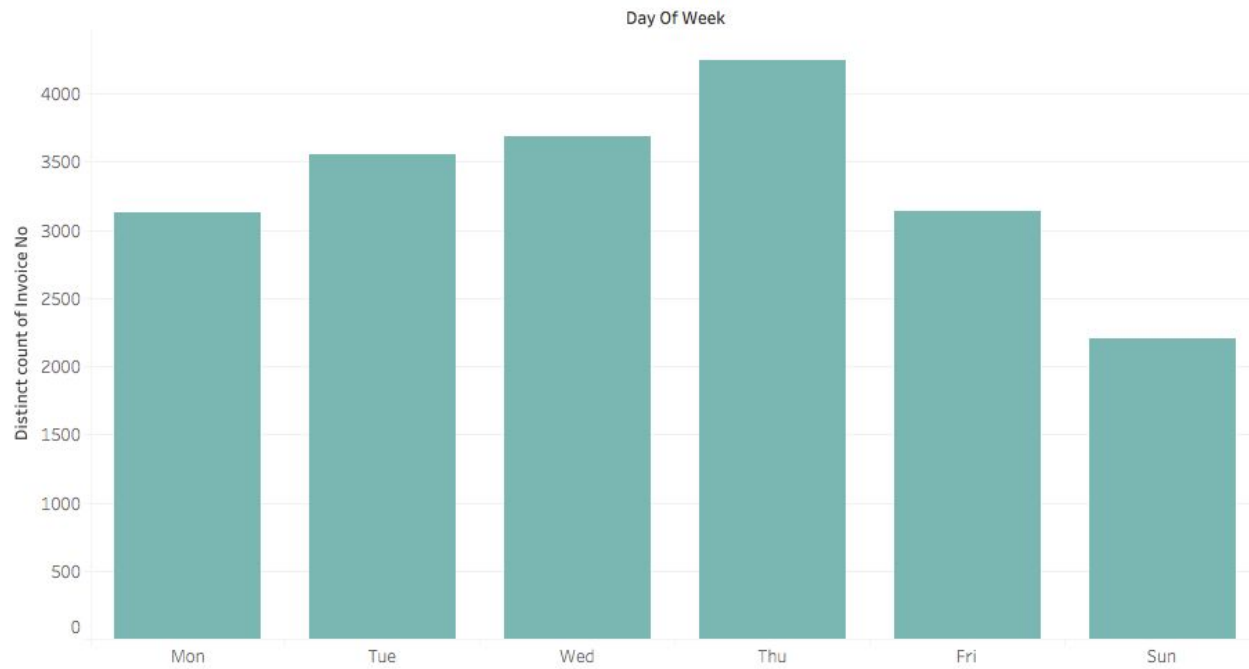
Busiest Month of the Year - 2011



E. Busiest Days of the Week

There are no records for orders on a Saturday. Wednesday and Thursdays are the most popular days for orders.

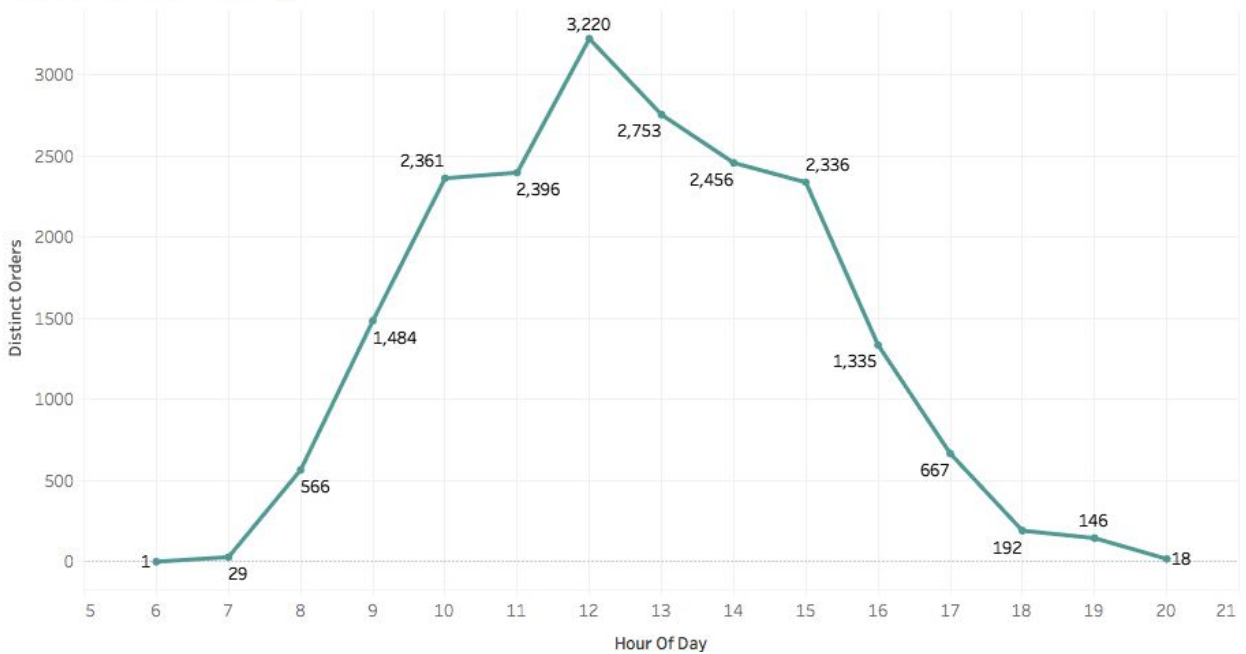
Busiest Day of the Week



F. Busiest Hours of the Day

12 noon is when there is a serious spike in Orders. From 10am to 3pm are the busiest hours of the day.

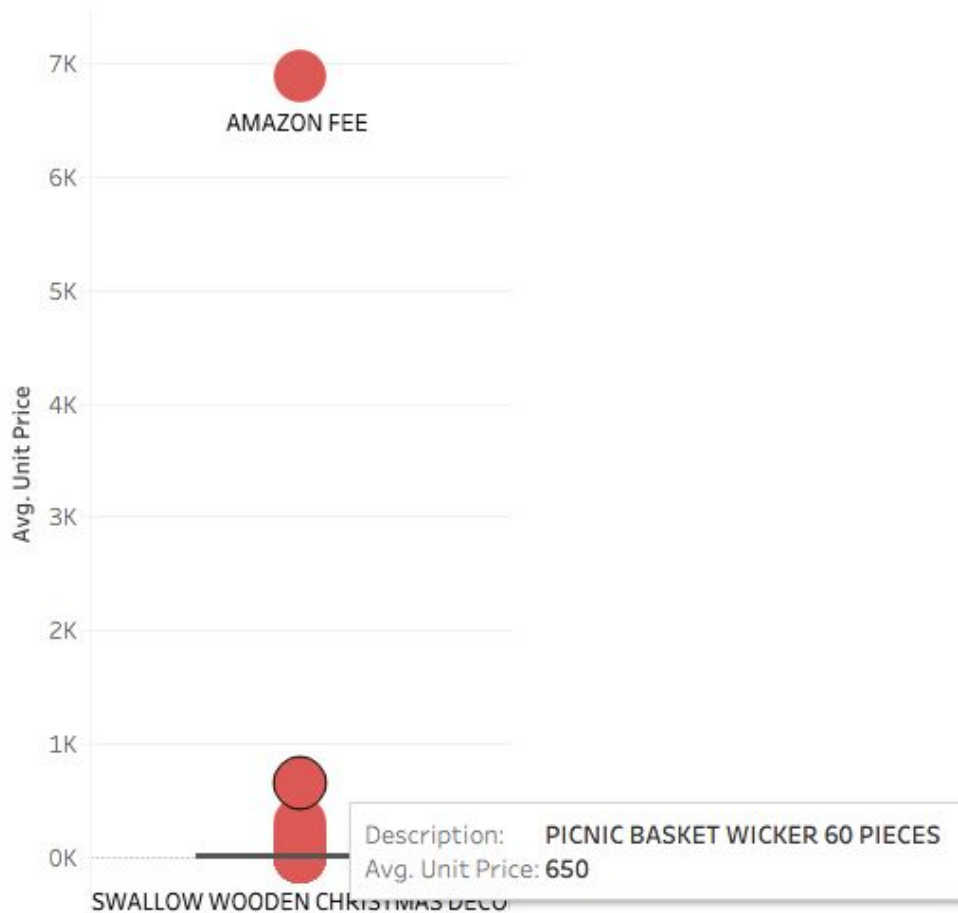
Busiest Time of the Day



For an online business it may be important to ensure that the website is able to handle the number of customers are these peak hours of the day.

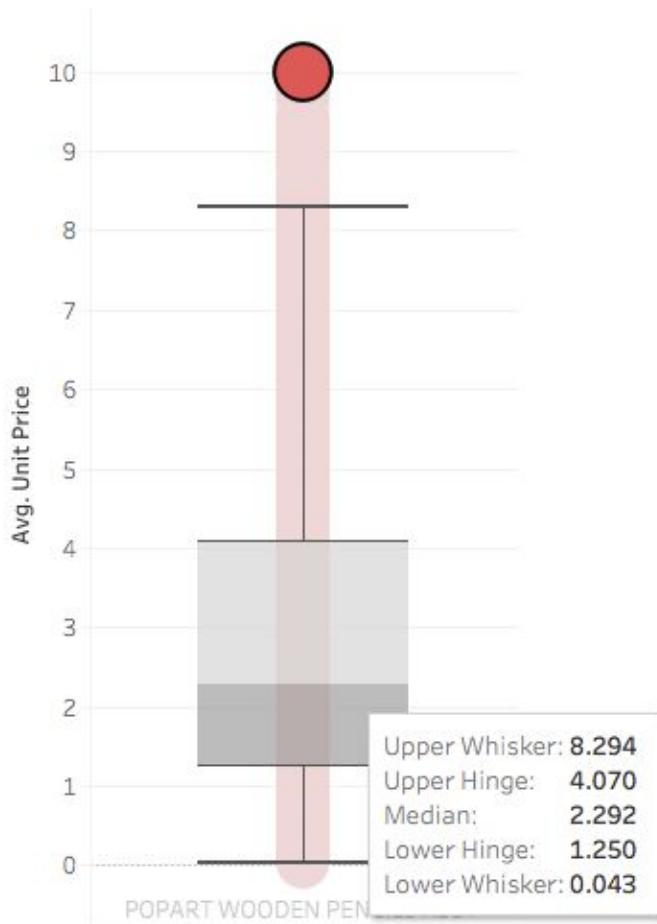
G. Some Average Price per Product Outliers

Average Price per Product



Filtering this diagram for Average Unit price < \$10 provides a better view:

Average Price per Product



This shows that the majority of Products that the business sells are of low unit price, but there are a number of outliers that skew the data.

2. Summary/Notable Findings

- There are no recorded sales on any Saturdays in entire dataset - Is there a reason for this?
- There is a clear spike in Orders at 12pm - are stocks levels high at this time? Can our server handle a large number of visitors?
- November 2011 had the most Orders
- The most ordered Product has only 1 CustomerID
- ¼ of CustomerIDs are missing from the dataset - Can customers be segmented by this fact? Are these one-time customers?
- 2515 records had a UnitPrice = 0. Is there a business reason for this?
- The average price per Product is less than \$3. There are some huge outliers with prices >\$1000. These products could be further analysed.

3. Data Pre-Processing

Firstly, opening the file in Excel allowed to see how it was laid out and gain a greater understanding.

It looked like an **Orders** database table.

There were **541909 records** for **8 fields**:

- InvoiceNo: *Order number*
- StockCode: *Item stock code*
- Description: *Item description*
- Quantity: *Quantity of item*
- InvoiceDate: *Date of order - with Day/Month written in USA format*
- UnitPrice: *Price of item*
- CustomerID: *Customer Identifier*
- Country: *Order country*

Then, loading the .csv file to R Studio.

```
> glimpse(ecom_data)
Observations: 541,909
Variables: 8
$ InvoiceNo   <chr> "536365", "536365", "536365", "536365", "536365", "536365", "53636...
$ StockCode  <chr> "85123A", "71053", "84406B", "84029G", "84029E", "22752", "21730",...
$ Description <chr> "WHITE HANGING HEART T-LIGHT HOLDER", "WHITE METAL LANTERN", "CREA...
$ Quantity   <dbl> 6, 6, 8, 6, 6, 2, 6, 6, 6, 32, 6, 6, 8, 6, 6, 3, 2, 3, 3, 4, 4, 6,...
$ InvoiceDate <dtm> 2010-12-01 08:26:00, 2010-12-01 08:26:00, 2010-12-01 08:26:00, 20...
$ UnitPrice  <dbl> 2.55, 3.39, 2.75, 3.39, 3.39, 7.65, 4.25, 1.85, 1.85, 1.69, 2.10, ...
$ CustomerID <dbl> 17850, 17850, 17850, 17850, 17850, 17850, 17850, 17850, 17850, 130...
$ Country    <chr> "United Kingdom", "United Kingdom", "United Kingdom", "United King...
```

In interpreting the data presented - some business logic validation is important. E.g. Order Quantity and UnitPrice should be positive values, as well as checking for missing data.

Validation Checks

Group 1: Missing data/NA fields

Checking for missing data per column:

```
> sapply(ecom_data, function(x) sum(is.na(x)))
InvoiceNo StockCode Description Quantity InvoiceDate UnitPrice CustomerID Country
0          0          1454          0          0          0          135080      0
```

There were 1454 records missing Description and 135,080 records missing CustomerID - removing all is ~25% of the dataset, which seemed like a lot so I decided to leave them for the moment and do some further analysis.

Group 2: Quantity less than or equal to 0

I omitted any records in this group, as it does not make much business sense for this analysis.

```
> sum(ecom_data$Quantity<=0)
[1] 10624
> ecom_data_clean1 = subset(ecom_data, ecom_data$Quantity > 0)
> sum(ecom_data_clean1$Quantity<=0)
[1] 0
```

Group 3: UnitPrice less than or equal to 0

```
> sum(ecom_data_clean1$UnitPrice<=0.001)
[1] 1185
> ecom_data_clean2 = subset(ecom_data_clean1, ecom_data_clean1$UnitPrice > 0.001)
> sum(ecom_data_clean2$UnitPrice<=0.001)
[1] 0
```

It is interesting that there are 2515 records where the unit price is equal to 0. I omitted them for this analysis but perhaps they could be used for a future analysis of how customers react to receiving 'free' items, if that is what they are.

Note - there was a crossover between records in Group 2 and 3.

The new dataset then contained 530,100 records - and no longer had any NA values for Description.

```
> glimpse(ecom_data_clean2)
Observations: 530,100
Variables: 8
$ InvoiceNo   <chr> "536365", "536365", "536365", "536365", "536365", "536365", "536365", "536366", "536366", "536367", "53636...
$ StockCode  <chr> "85123A", "71053", "84406B", "84029G", "84029E", "22752", "21730", "22633", "22632", "84879", "22745", "22...
$ Description <chr> "WHITE HANGING HEART T-LIGHT HOLDER", "WHITE METAL LANTERN", "CREAM CUPID HEARTS COAT HANGER", "KNITTED UN...
$ Quantity   <dbl> 6, 6, 8, 6, 6, 2, 6, 6, 6, 6, 32, 6, 6, 8, 6, 6, 3, 2, 3, 3, 4, 4, 6, 3, 3, 3, 3, 24, 24, 12, 12, 24, 48, 24,...
$ InvoiceDate <dtm> 2010-12-01 08:26:00, 2010-12-01 08:26:00, 2010-12-01 08:26:00, 2010-12-01 08:26:00, 2010-12-01 08:26:00, ...
$ UnitPrice  <dbl> 2.55, 3.39, 2.75, 3.39, 3.39, 7.65, 4.25, 1.85, 1.85, 1.69, 2.10, 2.10, 3.75, 1.65, 4.25, 4.95, 9.95, 5.95...
$ CustomerID <dbl> 17850, 17850, 17850, 17850, 17850, 17850, 17850, 17850, 17850, 13047, 13047, 13047, 13047, 13047, 13047, 1...
$ Country    <chr> "United Kingdom", "United Kingdom", "United Kingdom", "United Kingdom", "United Kingdom", "United Kingdom"...
```

```
> supply(ecom_data_clean2, function(x) sum(is.na(x)))
InvoiceNo  StockCode Description  Quantity InvoiceDate  UnitPrice  CustomerID  Country
0          0          0          0          0          0          132220      0
```

Create New Fields

To make analysis easier I created 5 calculated fields.

TotalAmount: $Quantity * UnitPrice$


```
> ecom_data_clean2$TotalAmount <- ecom_data_clean2$Quantity * ecom_data_clean2$UnitPrice
> head(ecom_data_clean2,10)
# A tibble: 10 x 9
  InvoiceNo StockCode Description      Quantity InvoiceDate      UnitPrice CustomerID Country TotalAmount
  <chr>      <chr>      <chr>          <dbl> <dtm>          <dbl>      <dbl> <chr>      <dbl>
1 536365    85123A    WHITE HANGING HEART...      6 2010-12-01 08:26:00      2.55      17850 United K...    15.3
2 536365    71053    WHITE METAL LANTERN      6 2010-12-01 08:26:00      3.39      17850 United K...    20.3
3 536365    84406B    CREAM CUPID HEARTS ...      8 2010-12-01 08:26:00      2.75      17850 United K...     22
4 536365    84029G    KNITTED UNION FLAG ...      6 2010-12-01 08:26:00      3.39      17850 United K...    20.3
5 536365    84029E    RED WOOLLY HOTTIE W...      6 2010-12-01 08:26:00      3.39      17850 United K...    20.3
6 536365    22752    SET 7 BABUSHKA NEST...      2 2010-12-01 08:26:00      7.65      17850 United K...    15.3
7 536365    21730    GLASS STAR FROSTED ...      6 2010-12-01 08:26:00      4.25      17850 United K...    25.5
8 536366    22633    HAND WARMER UNION J...      6 2010-12-01 08:28:00      1.85      17850 United K...    11.1
9 536366    22632    HAND WARMER RED POL...      6 2010-12-01 08:28:00      1.85      17850 United K...    11.1
10 536367    84879    ASSORTED COLOUR BIR...     32 2010-12-01 08:34:00      1.69      13047 United K...    54.1
```

Month: Parsed from InvoiceDate

```
> ecom_data_clean2$Mnth <- month(ecom_data_clean2$InvoiceDate)
> head(ecom_data_clean2,10)
# A tibble: 10 x 10
  InvoiceNo StockCode Description      Quantity InvoiceDate      UnitPrice CustomerID Country TotalAmount Mnth
  <chr>      <chr>      <chr>          <dbl> <dtm>          <dbl>      <dbl> <chr>      <dbl> <chr>
1 536365    85123A    WHITE HANGING ...      6 2010-12-01 08:26:00      2.55      17850 United ...    15.3    12
2 536365    71053    WHITE METAL LA...      6 2010-12-01 08:26:00      3.39      17850 United ...    20.3    12
3 536365    84406B    CREAM CUPID HE...      8 2010-12-01 08:26:00      2.75      17850 United ...     22     12
4 536365    84029G    KNITTED UNION ...      6 2010-12-01 08:26:00      3.39      17850 United ...    20.3    12
5 536365    84029E    RED WOOLLY HOT...      6 2010-12-01 08:26:00      3.39      17850 United ...    20.3    12
6 536365    22752    SET 7 BABUSHKA...      2 2010-12-01 08:26:00      7.65      17850 United ...    15.3    12
7 536365    21730    GLASS STAR FRO...      6 2010-12-01 08:26:00      4.25      17850 United ...    25.5    12
8 536366    22633    HAND WARMER UN...      6 2010-12-01 08:28:00      1.85      17850 United ...    11.1    12
9 536366    22632    HAND WARMER RE...      6 2010-12-01 08:28:00      1.85      17850 United ...    11.1    12
10 536367    84879    ASSORTED COLOU...     32 2010-12-01 08:34:00      1.69      13047 United ...    54.1    12
```

```
> table(ecom_data_clean2$Mnth)

 1      2      3      4      5      6      7      8      9     10     11     12
34306 27105 35803 29095 36164 35977 38644 34483 49259 59304 83369 66591
```

Day of the Week: Parsed from InvoiceDate

```
> ecom_data_clean2$DayOfWeek <- wday(ecom_data_clean2$InvoiceDate, week_start = getOption("lubridate.
week.start", 7), label = TRUE, abbr = TRUE)
```

```
> head(ecom_data_clean2,10)
# A tibble: 10 x 11
  InvoiceNo StockCode Description      Quantity InvoiceDate      UnitPrice CustomerID Country TotalAmount Mnth DayOfWeek
  <chr>      <chr>      <chr>          <dbl> <dtm>          <dbl>      <dbl> <chr>      <dbl> <dbl> <ord>
1 536365    85123A    WHITE HANGIN...      6 2010-12-01 08:26:00      2.55      17850 United...    15.3    12 Wed
2 536365    71053    WHITE METAL ...      6 2010-12-01 08:26:00      3.39      17850 United...    20.3    12 Wed
3 536365    84406B    CREAM CUPID ...      8 2010-12-01 08:26:00      2.75      17850 United...     22     12 Wed
4 536365    84029G    KNITTED UNIO...      6 2010-12-01 08:26:00      3.39      17850 United...    20.3    12 Wed
5 536365    84029E    RED WOOLLY H...      6 2010-12-01 08:26:00      3.39      17850 United...    20.3    12 Wed
6 536365    22752    SET 7 BABUSH...      2 2010-12-01 08:26:00      7.65      17850 United...    15.3    12 Wed
7 536365    21730    GLASS STAR F...      6 2010-12-01 08:26:00      4.25      17850 United...    25.5    12 Wed
8 536366    22633    HAND WARMER ...      6 2010-12-01 08:28:00      1.85      17850 United...    11.1    12 Wed
9 536366    22632    HAND WARMER ...      6 2010-12-01 08:28:00      1.85      17850 United...    11.1    12 Wed
10 536367    84879    ASSORTED COL...     32 2010-12-01 08:34:00      1.69      13047 United...    54.1    12 Wed
```

```
> table(ecom_data_clean2$DayOfWeek)
```

Sun	Mon	Tue	Wed	Thu	Fri	Sat
63904	93135	99459	92315	101007	80280	0

There was no orders on Saturdays which seems unusual.

It is assumed that the ordering system does not allow users to make orders on Saturday or that there is some missing data from this dataset.

Year: Parsed from InvoiceDate

```
> ecom_data_clean2$Yr <- year(ecom_data_clean2$InvoiceDate)
```

```
> head(ecom_data_clean2,10)
```

A tibble: 10 x 12

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	TotalAmount	Mnth	DayOfWeek	Yr
	<chr>	<chr>	<chr>	<dbl>	<dtm>	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<ord>	<dbl>
1	536365	85123A	WHITE HANGING H...	6	2010-12-01 08:26:00	2.55	17850	United ...	15.3	12	Wed	2010
2	536365	71053	WHITE METAL LAN...	6	2010-12-01 08:26:00	3.39	17850	United ...	20.3	12	Wed	2010
3	536365	84406B	CREAM CUPID HEA...	8	2010-12-01 08:26:00	2.75	17850	United ...	22	12	Wed	2010
4	536365	84029G	KNITTED UNION F...	6	2010-12-01 08:26:00	3.39	17850	United ...	20.3	12	Wed	2010
5	536365	84029E	RED WOOLLY HOTT...	6	2010-12-01 08:26:00	3.39	17850	United ...	20.3	12	Wed	2010
6	536365	22752	SET 7 BABUSHKA ...	2	2010-12-01 08:26:00	7.65	17850	United ...	15.3	12	Wed	2010
7	536365	21730	GLASS STAR FROS...	6	2010-12-01 08:26:00	4.25	17850	United ...	25.5	12	Wed	2010
8	536366	22633	HAND WARMER UNI...	6	2010-12-01 08:28:00	1.85	17850	United ...	11.1	12	Wed	2010
9	536366	22632	HAND WARMER RED...	6	2010-12-01 08:28:00	1.85	17850	United ...	11.1	12	Wed	2010
10	536367	84879	ASSORTED COLOUR...	32	2010-12-01 08:34:00	1.69	13047	United ...	54.1	12	Wed	2010

```
> unique(ecom_data_clean2$Yr)
```

[1] 2010 2011

Time of the Day: Parsed from InvoiceDate

```
> ecom_data_clean2$HourOfDay <- hour(ecom_data_clean2$InvoiceDate)
```

```
> head(ecom_data_clean2,10)
```

A tibble: 10 x 13

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	TotalAmount	Mnth	DayOfWeek	Yr	HourOfDay
	<chr>	<chr>	<chr>	<dbl>	<dtm>	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<ord>	<dbl>	<int>
1	536365	85123A	WHITE HANGING HEART T-LI...	6	2010-12-01 08:26:00	2.55	17850	United Ki...	15.3	12	Wed	2010	8
2	536365	71053	WHITE METAL LANTERN	6	2010-12-01 08:26:00	3.39	17850	United Ki...	20.3	12	Wed	2010	8
3	536365	84406B	CREAM CUPID HEARTS COAT ...	8	2010-12-01 08:26:00	2.75	17850	United Ki...	22	12	Wed	2010	8
4	536365	84029G	KNITTED UNION FLAG HOT W...	6	2010-12-01 08:26:00	3.39	17850	United Ki...	20.3	12	Wed	2010	8
5	536365	84029E	RED WOOLLY HOTTIE WHITE ...	6	2010-12-01 08:26:00	3.39	17850	United Ki...	20.3	12	Wed	2010	8
6	536365	22752	SET 7 BABUSHKA NESTING B...	2	2010-12-01 08:26:00	7.65	17850	United Ki...	15.3	12	Wed	2010	8
7	536365	21730	GLASS STAR FROSTED T-LIG...	6	2010-12-01 08:26:00	4.25	17850	United Ki...	25.5	12	Wed	2010	8
8	536366	22633	HAND WARMER UNION JACK	6	2010-12-01 08:28:00	1.85	17850	United Ki...	11.1	12	Wed	2010	8
9	536366	22632	HAND WARMER RED POLKA DOT	6	2010-12-01 08:28:00	1.85	17850	United Ki...	11.1	12	Wed	2010	8
10	536367	84879	ASSORTED COLOUR BIRD ORN...	32	2010-12-01 08:34:00	1.69	13047	United Ki...	54.1	12	Wed	2010	8

I checked that the HourOfDay field made sense by viewing all of the unique values. There was only 1 entry for 6AM - I filtered this record out to ensure the parsing had worked as intended. It looked good.

```
> table(ecom_data_clean2$HourOfDay)
```

6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
1	379	8800	33700	47821	56139	77120	71001	65936	76246	53369	27562	7709	3515	802

```
> filter(ecom_data_clean2, ecom_data_clean2$HourOfDay == 6)
```

A tibble: 1 x 13







	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	TotalAmount	Mnth	DayOfWeek	Yr	HourOfDay
	<chr>	<chr>	<chr>	<dbl>	<dtm>	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<ord>	<dbl>	<int>
1	563597	22852	DOG BOWL VINTAGE CREAM	1	2011-08-18 06:20:00	4.25	14305	United Kingdom	4.25	8	Thu	2011	6

The dataset then had 530,100 records and 13 variables. I rechecked any missing values. CustomerID remains the only field with missing values. This looked good.


```
> sapply(ecom_data_clean2, function(x) sum(is.na(x)))
```

InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID
0	0	0	0	0	0	132220
Country	TotalAmount	Mnth	DayOfWeek	Yr	HourOfDay	
0	0	0	0	0	0	

I tested loading this file set to Tableau. Tableau took 'InvoiceNo' as an Integer value. When I explored this further I realised that there was 1 record that contained a letter in the 'InvoiceNo'

		Sort fields	Data source order		<input type="checkbox"/> Show aliases	<input type="checkbox"/> Show hidden fields	1		rows
Abc Ecom data cleaned.csv	Abc Ecom data cleaned.csv	Abc Ecom data cleaned.csv	# Ecom data cleaned.csv	 Ecom data cleaned.csv	# Ecom data cleaned.csv	# Ecom data cleaned.csv	 Ecom data cleaned.csv	# Ecom data cleaned.csv	
Invoice No	Stock Code	Description	Quantity	Invoice Date	Unit Price	Customer ID	Country	Total Amount	
A563185	B	Adjust bad debt	1	12/08/2011 14:50:00	11,062.06	null	United Kingdom	11,	

I decided to remove this record - based on the 'Adjusted bad debt' Description field, it did not seem to fit in with the other records. I removed this in R Studio as seen below:

```
> which(ecom_data_cleaned$InvoiceNo=='A563185')
[1] 292814
> ecom_data_cleaned[292814,]
      InvoiceNo StockCode Description Quantity InvoiceDate UnitPrice CustomerID Country
292814  A563185      B Adjust bad debt      1 2011-08-12 14:50:00 11062.06      NA United Kingdom
      TotalAmount Mnth DayOfWeek Yr HourOfDay
292814  11062.06    8      Fri 2011      14
> ecom_data_cleaned %>% slice(292813:292815)
      InvoiceNo StockCode Description Quantity InvoiceDate UnitPrice CustomerID
1    563184    82482 WOODEN PICTURE FRAME WHITE FINISH      4 2011-08-12 14:50:00      2.55    17516
2    A563185      B Adjust bad debt      1 2011-08-12 14:50:00 11062.06      NA
3    563188    79160 HEART SHAPE WIRELESS DOORBELL      48 2011-08-12 15:00:00      1.69    15606
      Country TotalAmount Mnth DayOfWeek Yr HourOfDay
1 United Kingdom      10.20    8      Fri 2011      14
2 United Kingdom 11062.06    8      Fri 2011      14
3 United Kingdom      81.12    8      Fri 2011      15
> ecom_data_cleaned_2 = subset(ecom_data_cleaned[-c(292814),])
> which(ecom_data_cleaned_2$InvoiceNo=='A563185')
integer(0)
```

Finally, I loaded the clean dataset in Tableau - it contained 530099 records.