



# ONLINE SALES DATASET



# INTRODUCTION

## Purpose and Scope

To analyze online sales data for insights into product performance, regional trends, and discounts. Focus on categories, regions, sales, country, pricing, and discounts.

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# LOAD & VIEW DATASET

```
Sales_Data <- read.csv("C:/Users/Vanshika Gupta/Desktop/online  
sale dataset.csv")  
View(Sales_Data)
```

	Sales	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	Discount	PaymentMethod
1	221958	White Mug	38	01-01-2020	1.71	37039	Australia	0.470000	Bank Transfer
2	771155	White Mug	18	01-01-2020	41.25	19144	Spain	0.190000	paypal
3	231932	Headphones	49	01-01-2020	29.11	50472	Germany	0.350000	Bank Transfer
4	465838	Desk Lamp	14	01-01-2020	76.68	96586	Netherlands	0.140000	paypal
5	359178	USB Cable	-30	01-01-2020	-68.11	NA	United Kingdom	1.501433	Bank Transfer
6	744167	Office Chair	47	01-01-2020	70.16	53887	Sweden	0.480000	Credit Card
7	210268	USB Cable	25	01-01-2020	85.74	46567	Belgium	0.150000	Bank Transfer
8	832180	Notebook	8	01-01-2020	95.65	75098	Norway	0.040000	Bank Transfer
9	154886	Wireless Mouse	19	01-01-2020	98.19	87950	Belgium	0.050000	paypal
10	237337	Headphones	40	01-01-2020	98.17	39718	Italy	0.160000	Bank Transfer
11	621430	Notebook	49	01-01-2020	87.56	13030	United Kingdom	0.190000	paypal
12	187498	Office Chair	41	01-01-2020	59.51	32466	Australia	0.390000	Bank Transfer
13	999159	Blue Pen	41	01-01-2020	25.59	89794	Australia	0.010000	Credit Card
...	.....	.....	..	.....	....	.....	...	.....	.....

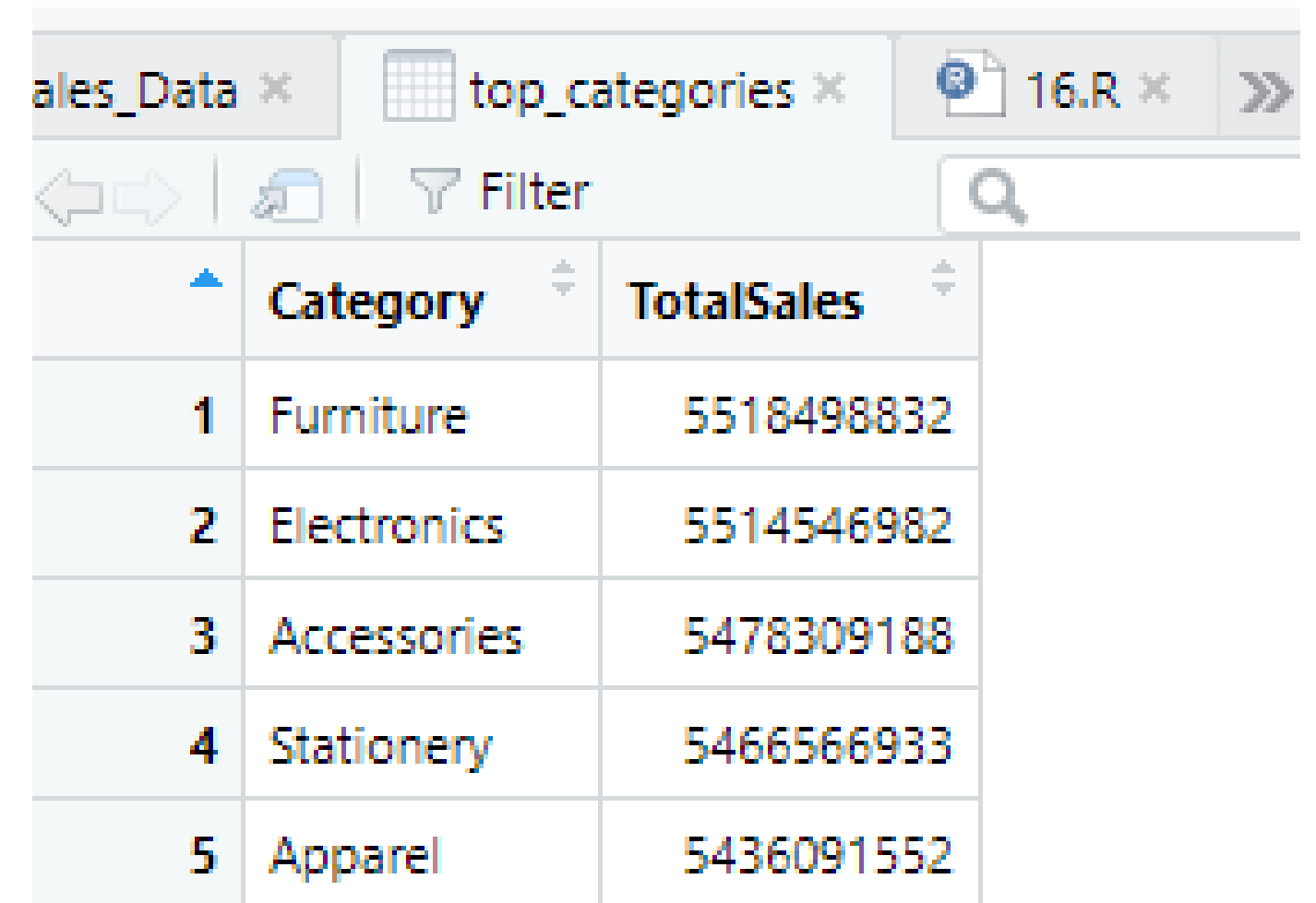


# KEY FINDINGS AND VISUALIZATION



# TOP 5 CATEGORIES BY SALES

```
top_categories <- Sales_Data %>%  
  group_by(Category) %>%  
  summarise(TotalSales = sum(Sales)) %>%  
  arrange(desc(TotalSales)) %>%  
  head(5)
```



The image shows a screenshot of the RStudio interface. The top pane displays the R script used to generate the table. The bottom pane shows the resulting data frame, 'top\_categories', which contains the top 5 categories by total sales. The table has two columns: 'Category' and 'TotalSales'. The rows are ordered from highest to lowest total sales.

	Category	TotalSales
1	Furniture	5518498832
2	Electronics	5514546982
3	Accessories	5478309188
4	Stationery	5466566933
5	Apparel	5436091552



# DISCOUNT IMPACT- SALES VOLUME AND REVENUE WITH DISCOUNTS

```
discount_impact <- Sales_Data %>%  
filter(Discount > 0) %>%  
group_by(Category) %>%  
summarise(TotalDiscountedSales =  
sum(Quantity),  
TotalRevenueWithDiscounts = sum  
((Quantity * UnitPrice) *  
(1 - Discount), na.rm = TRUE))
```

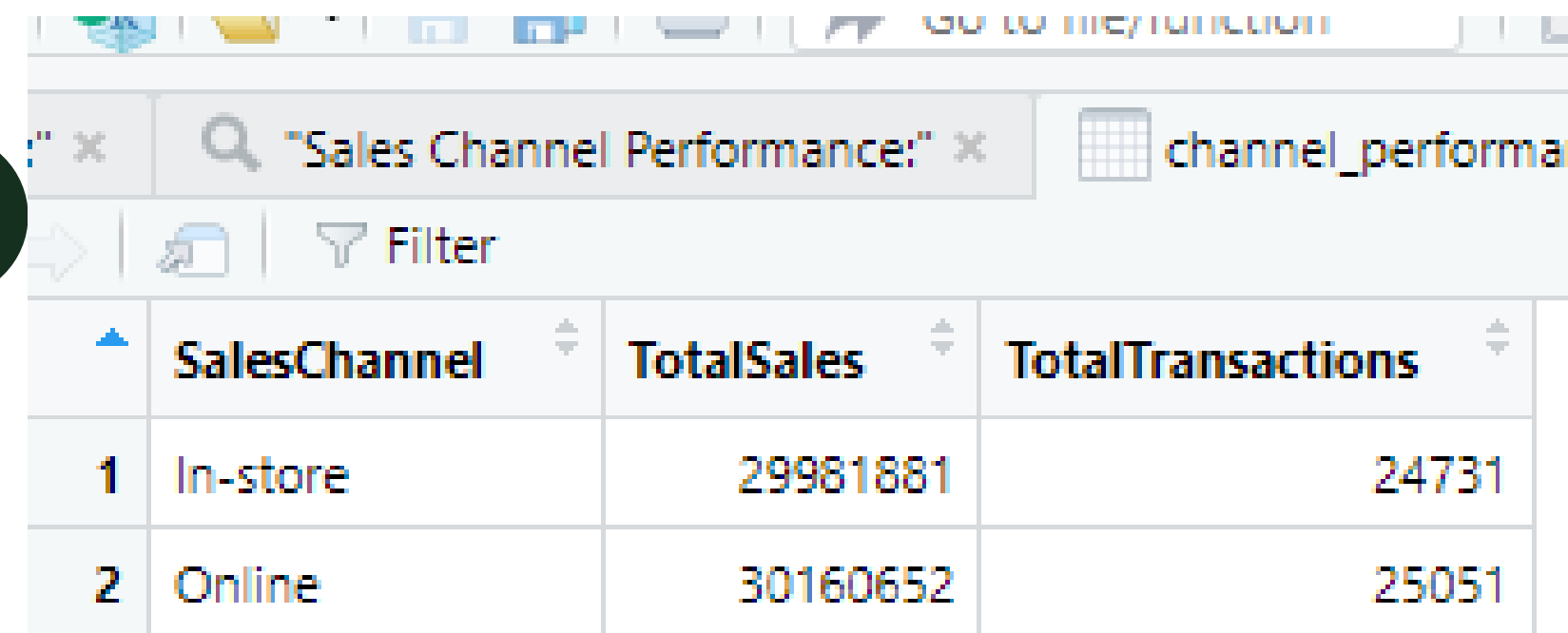
```
View(discount_impact)
```

Reve... x discount_impact x "Shipping Efficiency by Category:" x			
Filter			
	Category	TotalDiscountedSales	TotalRevenueWithDiscounts
1	Accessories	218939	8593149
2	Apparel	219543	8553277
3	Electronics	217191	8503881
4	Furniture	223156	8674973
5	Stationery	222658	8551424

# SALES CHANNEL PERFORMANCE: COMPARE ONLINE VS IN-STORE SALES

```
channel_performance <- Sales_Data %>%  
  group_by(SalesChannel) %>%  
  summarise(TotalSales = sum(Quantity *  
    UnitPrice, na.rm = TRUE),  
    TotalTransactions = n())
```

```
View(channel_performance)
```



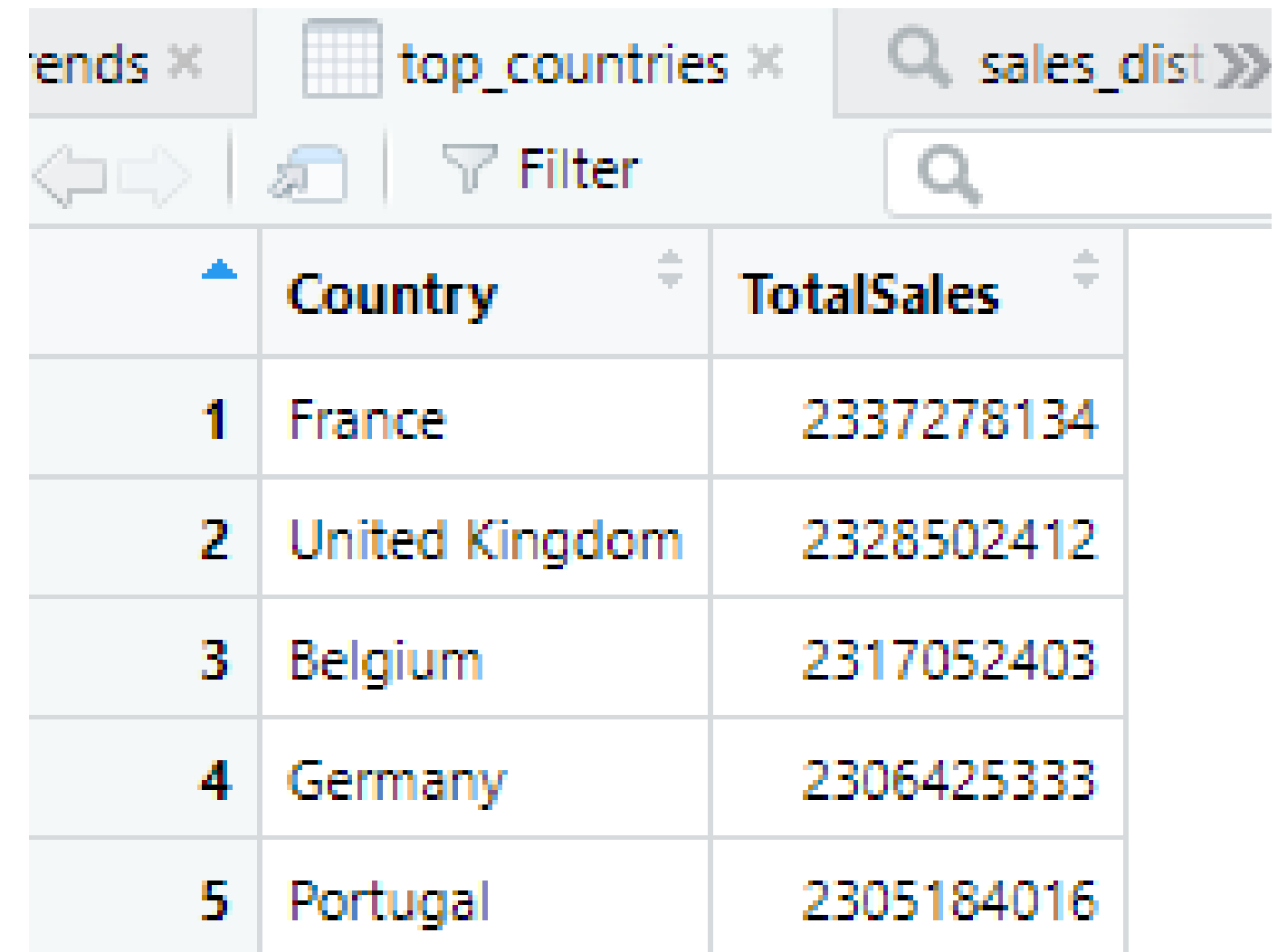
The screenshot shows a RStudio window with a tab titled "Sales Channel Performance:". Below the tab is a toolbar with a "Filter" button. The main area displays a data table with the following content:

	SalesChannel	TotalSales	TotalTransactions
1	In-store	29981881	24731
2	Online	30160652	25051

# TOP 5 COUNTRIES BY TOTAL SALES

```
top_countries <- Sales_Data %>%  
  group_by(Country) %>%  
  summarise(TotalSales = sum(Sales)) %>%  
  arrange(desc(TotalSales)) %>%  
  head(5)
```

```
View(top_countries)
```



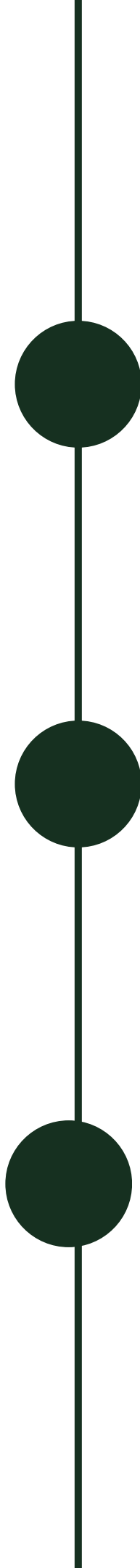
The screenshot shows a data visualization tool interface. At the top, there are tabs for 'ends', 'top\_countries', and 'sales\_dist'. Below the tabs is a toolbar with navigation arrows, a filter icon, and a search bar. The main area displays a table with the following data:

	Country	TotalSales
1	France	2337278134
2	United Kingdom	2328502412
3	Belgium	2317052403
4	Germany	2306425333
5	Portugal	2305184016



# SALES TRENDS OVER TIME

```
sales_trends <- Sales_Data %>%  
group_by(InvoiceDate) %>%  
summarise(DailySales = sum(Sales))  
View(sales_trends)
```

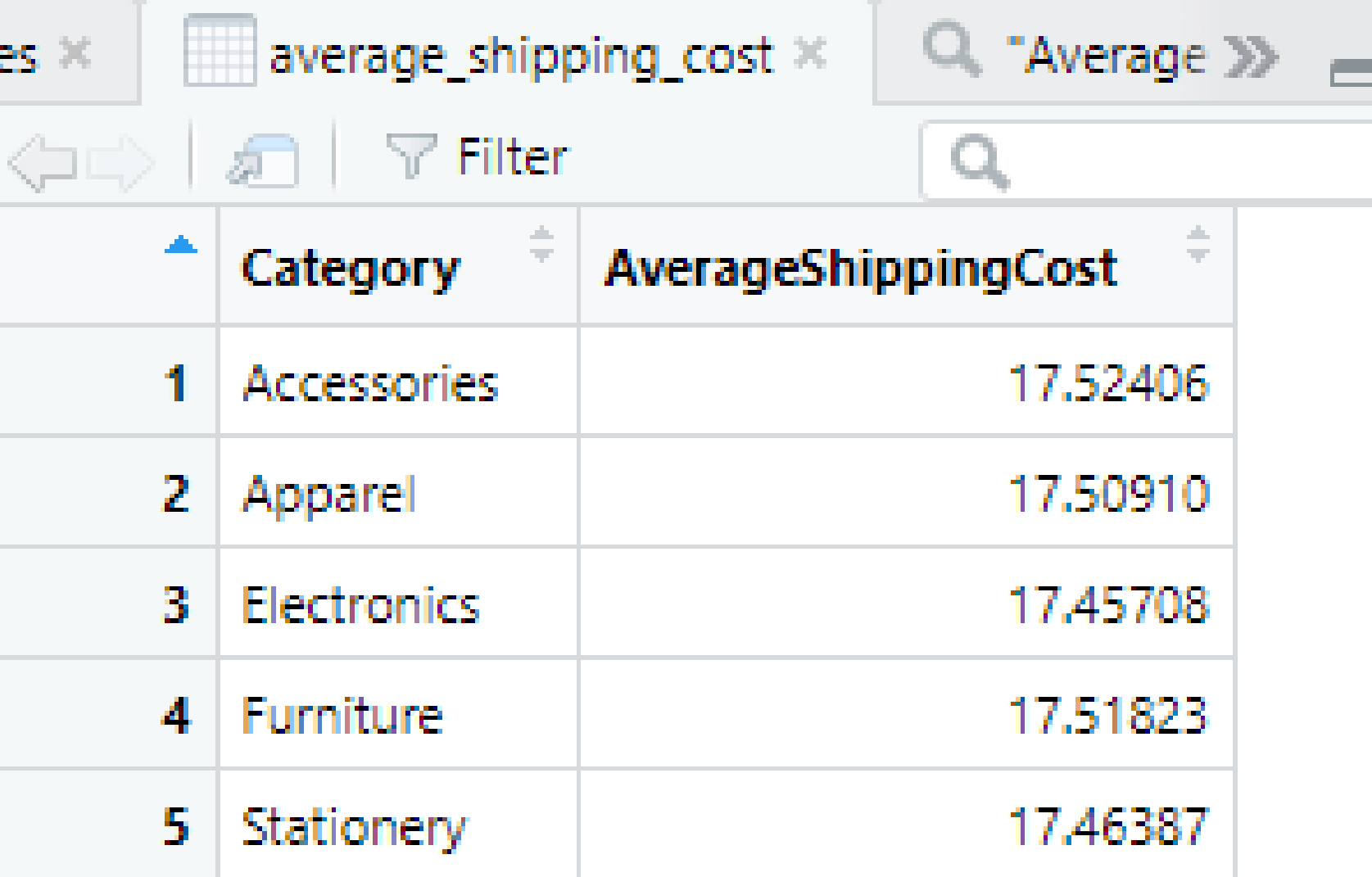


	InvoiceDate	DailySales
1	01-01-2020	11148427
2	01-01-2021	14734509
3	01-01-2022	13100345
4	01-01-2023	11740875
5	01-01-2024	10841146
6	01-01-2025	12933091
7	01-02-2020	11609871
8	01-02-2021	11145803
9	01-02-2022	13094878
10	01-02-2023	12855673
11	01-02-2024	13208560
12	01-02-2025	14002336
13	01-03-2020	13780702
14	01-03-2021	13054213

# CALCULATE AVERAGE SHIPPING COST PER CATEGORY

```
average_shipping_cost <- Sales_Data %>%  
  group_by(Category) %>%  
  summarise(AverageShippingCost =  
    mean(ShippingCost, na.rm = TRUE))
```

```
View(average_shipping_cost)
```



The screenshot shows a data visualization interface with a table titled 'average\_shipping\_cost'. The table has two columns: 'Category' and 'AverageShippingCost'. The data is as follows:

	Category	AverageShippingCost
1	Accessories	17.52406
2	Apparel	17.50910
3	Electronics	17.45708
4	Furniture	17.51823
5	Stationery	17.46387

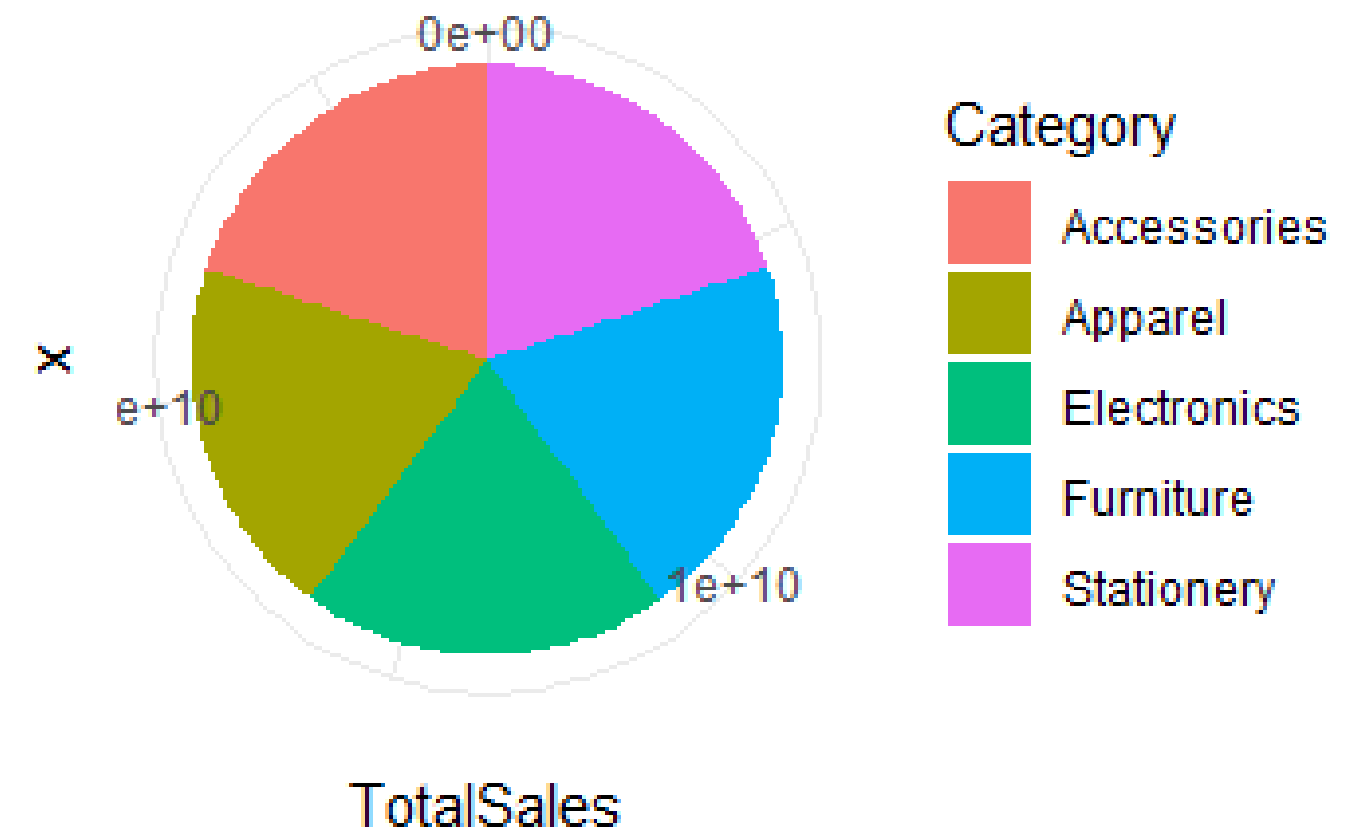
# PIE CHART - CATEGORY SALES CONTRIBUTION

```
category_sales <- Sales_Data %>%  
group_by(Category) %>%  
summarise(TotalSales = sum(Sales))
```

```
ggplot(category_sales, aes(x = "", y = TotalSales,  
fill = Category)) +  
geom_bar(stat = "identity", width = 1) +  
coord_polar("y") +  
labs(title = "Category Sales Distribution") +  
theme_minimal()
```

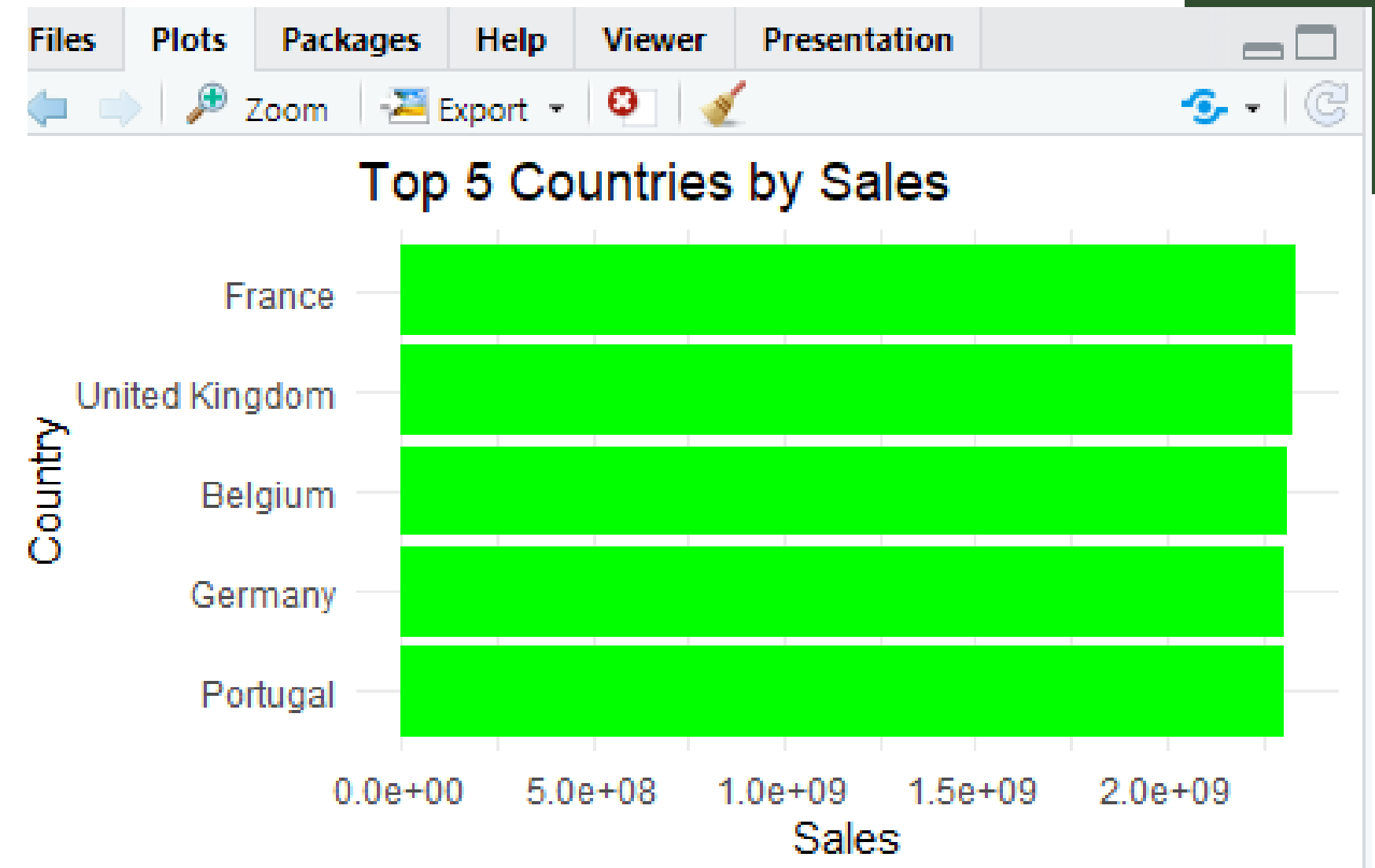


Category Sales Distribution



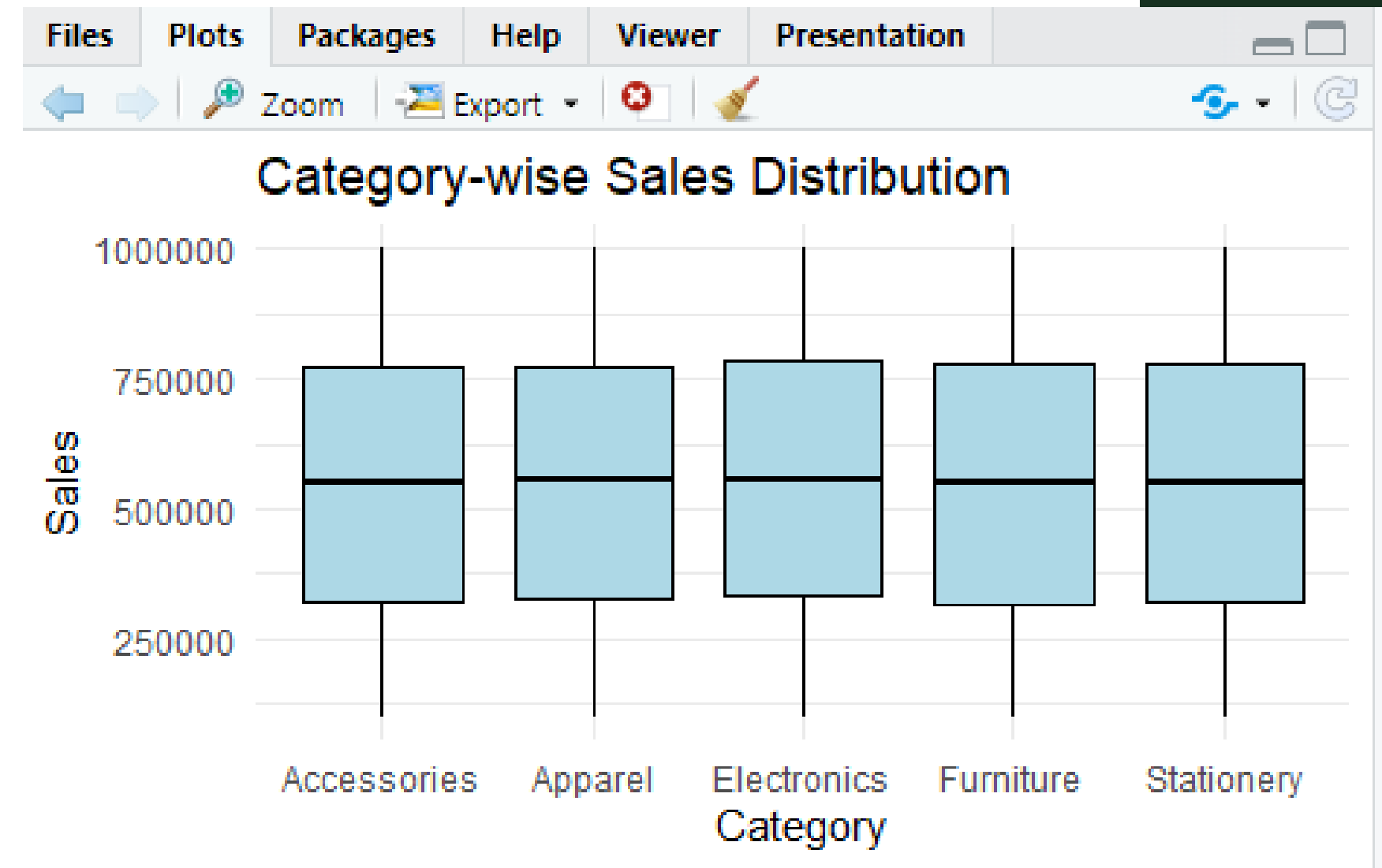
# BAR PLOT FOR TOP COUNTRIES

```
ggplot(top_countries, aes(x = reorder  
(Country, TotalSales), y = TotalSales)) +  
geom_bar(stat = "identity", fill = "green") +  
coord_flip() +  
labs(title = "Top 5 Countries by Sales",  
x = "Country", y = "Sales") +  
theme_minimal()
```



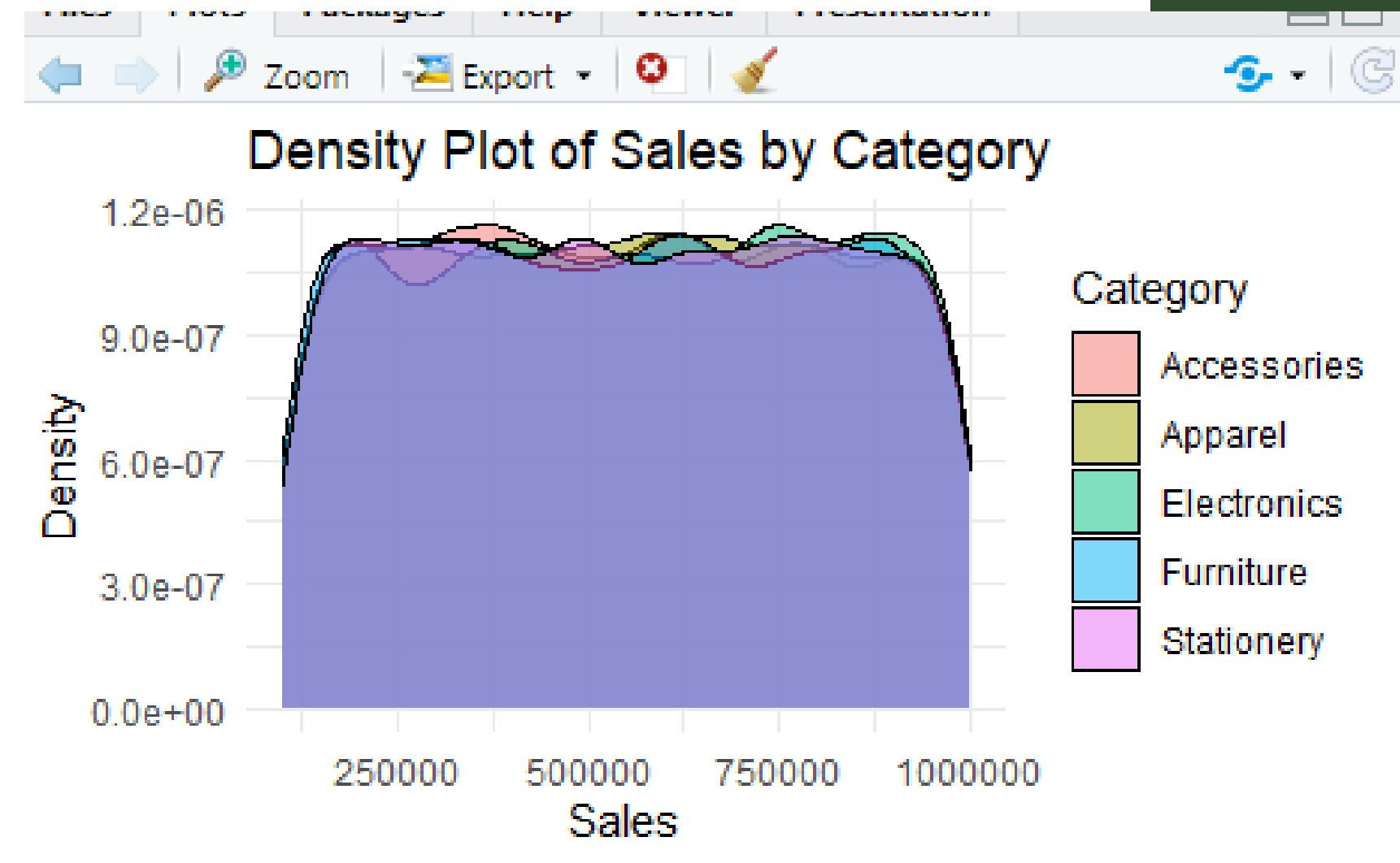
# BOX PLOT CATEGORY-WISE SALES DISTRIBUTION

```
ggplot(Sales_Data, aes(x = Category, y = Sales)) +  
  geom_boxplot(fill = "lightblue", color = "black") +  
  labs(title = "Category-wise Sales Distribution",  
        x = "Category", y = "Sales") +  
  theme_minimal()
```



# DENSITY PLOT: SALES DISTRIBUTION

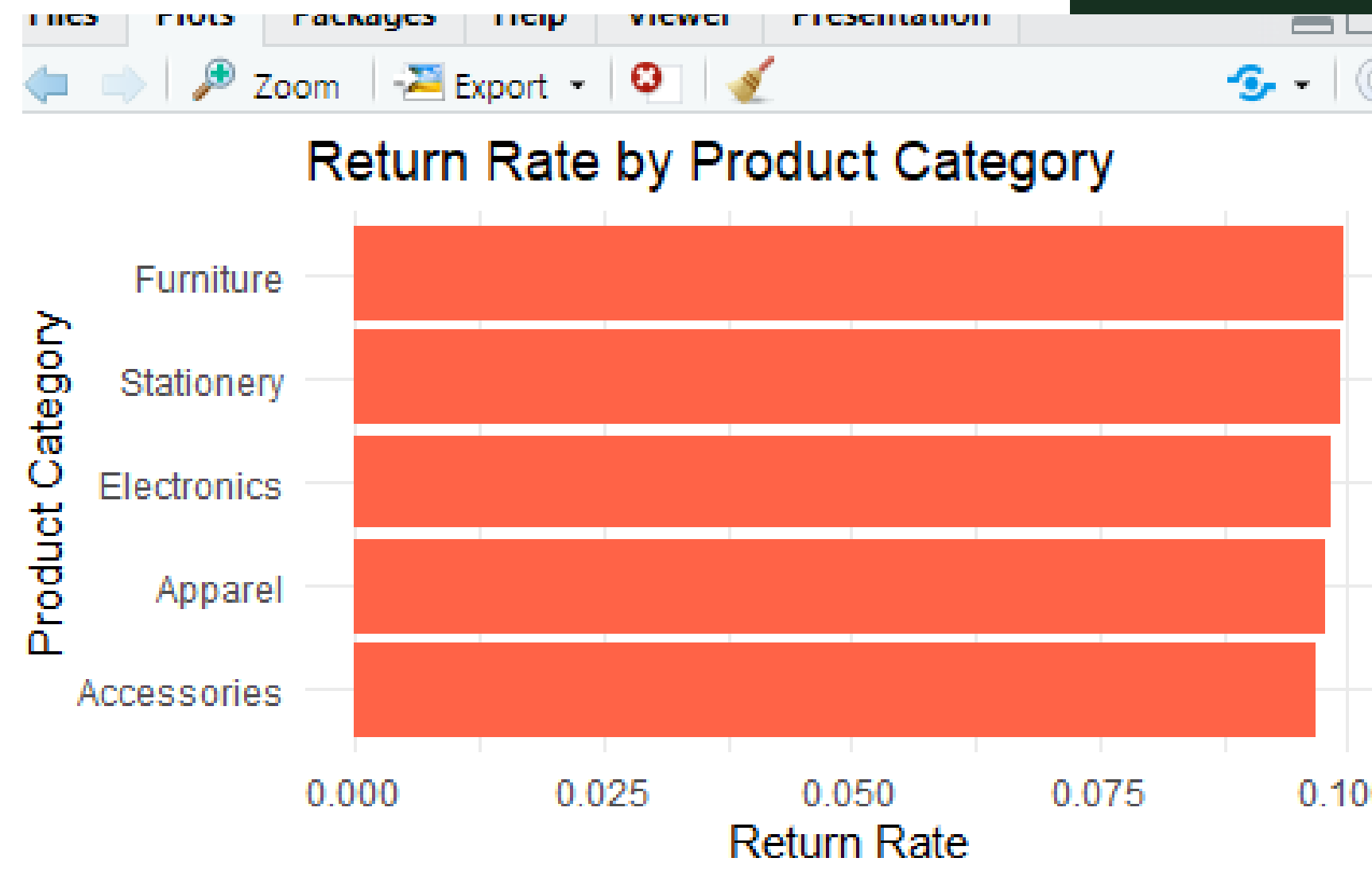
```
ggplot(Sales_Data, aes(x = Sales, fill = Category)) +  
  geom_density(alpha = 0.5) +  
  labs(title = "Density Plot of Sales by Category",  
        x = "Sales", y = "Density") +  
  theme_minimal()
```





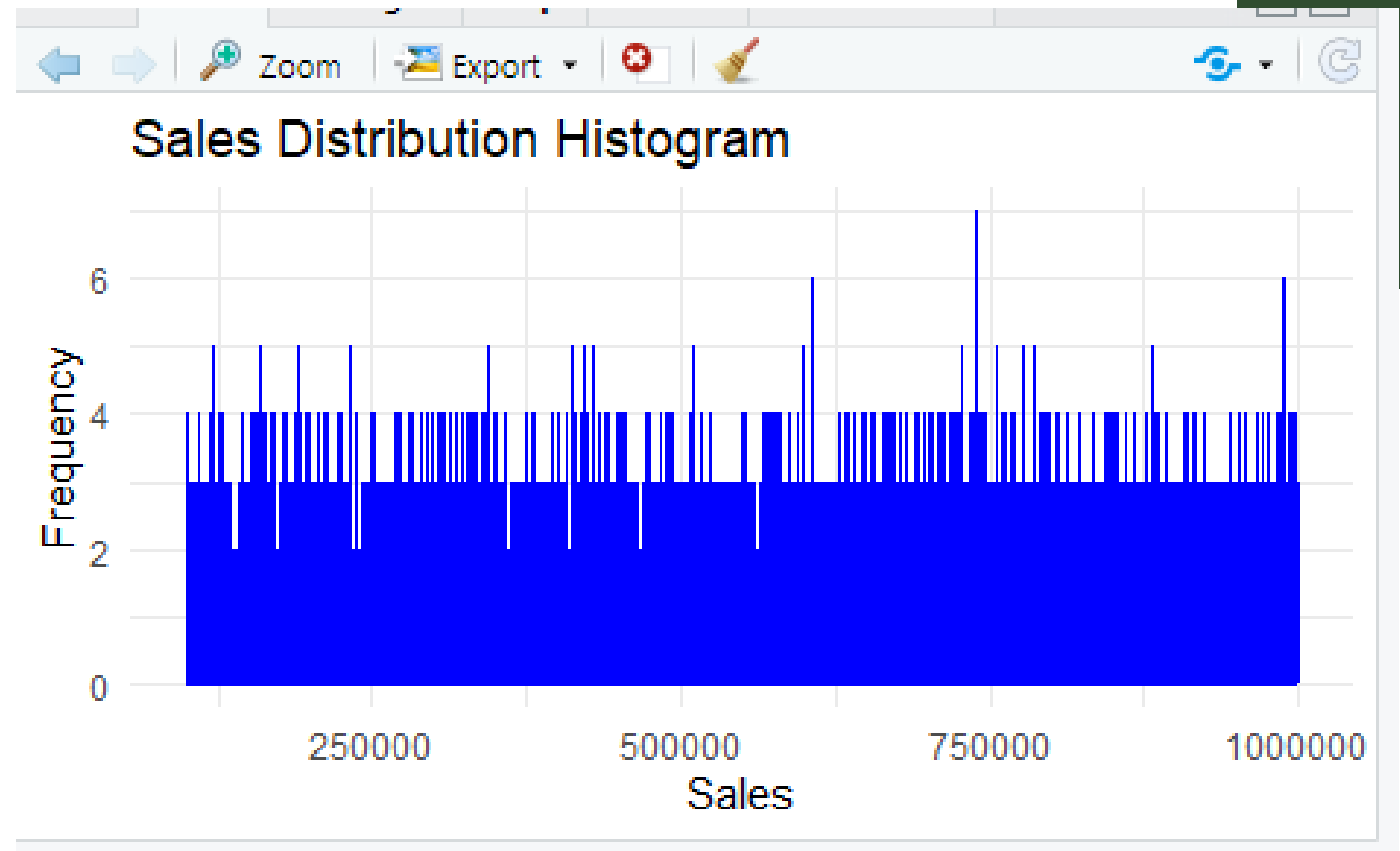
# RETURN RATE BY PRODUCT CATEGORY

```
return_rate_by_category <- Sales_Data %>%  
  group_by(Category) %>%  
  summarise(ReturnCount = sum(ReturnStatus ==  
    "Returned"),  
    TotalCount = n(), ReturnRate = ReturnCount / TotalCount)  
%>% arrange(desc(ReturnRate))  
  
ggplot(return_rate_by_category, aes(x = reorder(Category,  
  ReturnRate), y = ReturnRate)) +  
  geom_bar(stat = "identity", fill = "tomato") +  
  coord_flip() +  
  labs(title = "Return Rate by Product Category", x = "Product  
  Category", y = "Return Rate") +  
  theme_minimal()
```



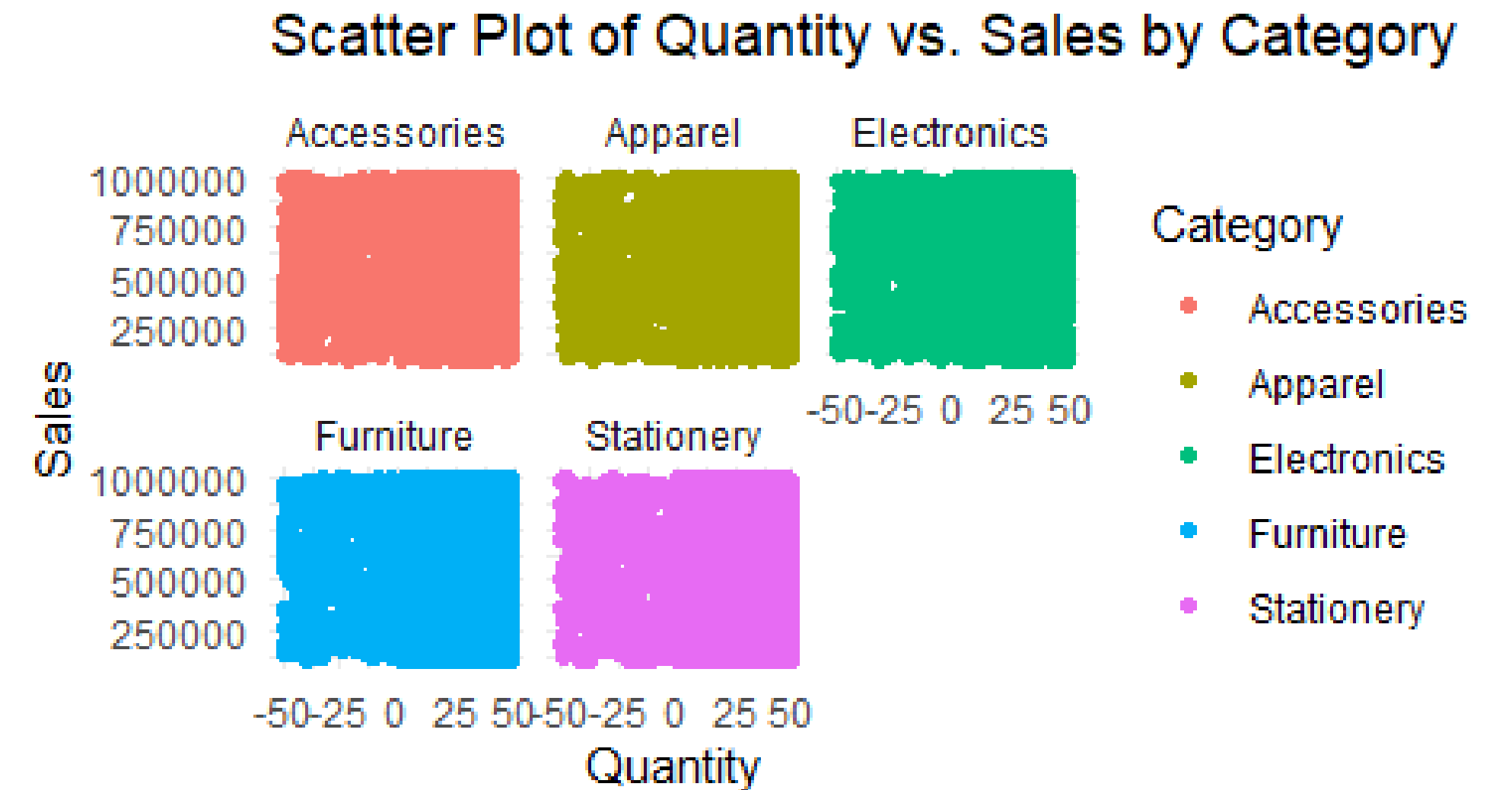
# HISTOGRAM OF SALES DISTRIBUTION

```
ggplot(Sales_Data, aes(x = Sales)) +  
  geom_histogram(binwidth = 10,  
    fill = "orange", color = "blue") +  
  labs(title = "Sales Distribution Histogram",  
    x = "Sales", y = "Frequency") +  
  theme_minimal()
```



# FACETED SCATTER PLOT- QUANTITY VS. SALES BY CATEGORY

```
ggplot(Sales_Data, aes(x = Quantity,  
y = Sales, color = Category)) +  
geom_point() +  
facet_wrap(~ Category) +  
labs(title = "Scatter Plot of Quantity vs.  
Sales by Category",  
x = "Quantity", y = "Sales") +  
theme_minimal()
```



# CONCLUSION

**THE ANALYSIS OF THE ONLINE SALES DATASET HAS PROVIDED VALUABLE INSIGHTS INTO VARIOUS ASPECTS OF SALES PERFORMANCE.**

**THE STUDY UNDERSCORES THE IMPORTANCE OF DATA-DRIVEN STRATEGIES IN OPTIMIZING SALES PERFORMANCE AND ENHANCING CUSTOMER EXPERIENCES. FUTURE RECOMMENDATIONS INCLUDE EXPLORING PREDICTIVE ANALYTICS FOR FORECASTING TRENDS AND REFINING DISCOUNT STRATEGIES FOR MAXIMIZED PROFITABILITY.**

**THANK YOU**