

Fashion Collections using R

Created by-VRINDA GOEL





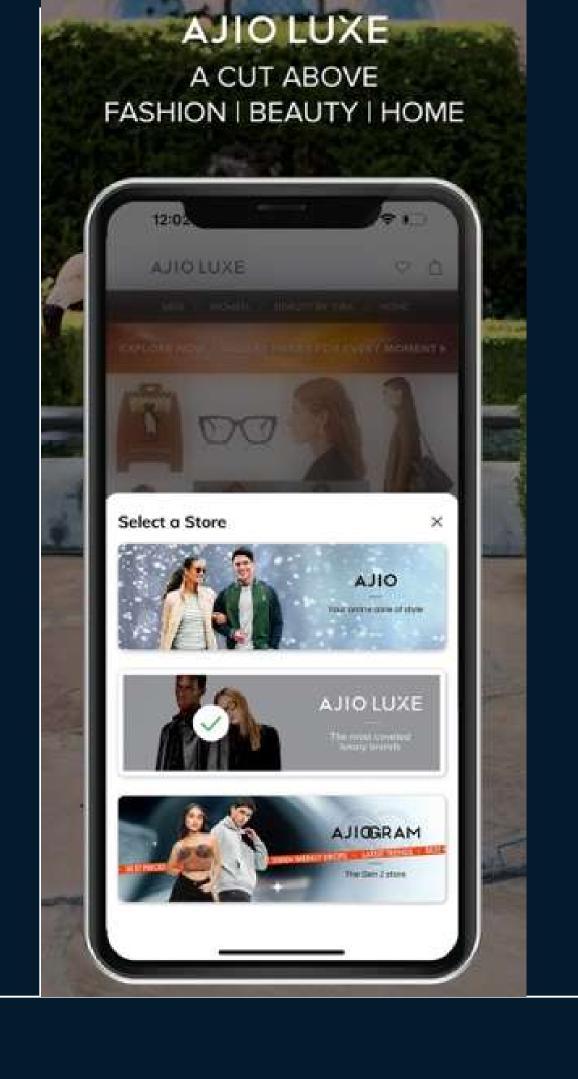
INTRODUCTION

This project aims to analyze AJIO's dataset to identify key trends in pricing, discounts, product categories, and regional offerings. By using descriptive statistics, visualizations, and statistical methods like correlation, the study explores relationships among variables and provides / actionable insights to enhance business strategies and customer engagement.

OBJECTIVE

The primary objective of this project is to analyze AJIO's product data to understand various aspects such as pricing, discounts, product characteristics, and seller attributes. Specific goals include:

- Exploring the distribution of prices and discounts.
- Identifying trends in product attributes, such as material, category, and shipping time.
- Analyzing relationships between numerical variables like price and discount.
- Evaluating brand popularity and country-level insights.



RESEARCH METHODOGLY

- Data Collection: Data was obtained from AJIO, focusing on product listings with attributes like price, discount, category, and brand.
- Data Cleaning:Missing values were handled using na.omit().Duplicates were removed using the distinct() function.Column names were standardized for consistency.
- Exploratory Data Analysis (EDA):Univariate, bivariate, and multivariate analyses were conducted using descriptive statistics and visualizations. Variables were analyzed for their relationships and distributions.
- Statistical Analysis: Techniques like correlation and regression analysis were applied to uncover deeper insights.

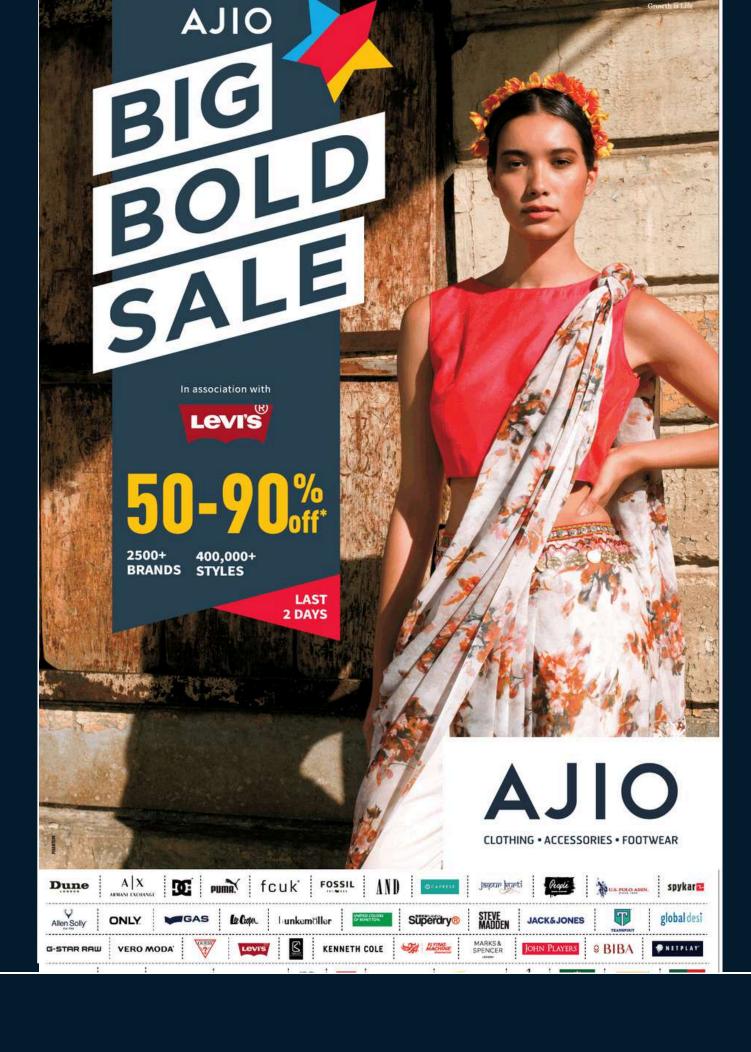


Data and Variables

Data Description

The dataset contains product-related information with the following key variables:

- price_usd: Price of the product.
- discount_usd: Discount offered on the product.
- product_type: Category or type of the product.
- product_material: Material of the product.
- brand_name: Brand offering the product.
- shipping_time: Average time for delivery.
- country: Origin of the seller.



Variables for Analysis

- Dependent Variable: price_usd.
- Independent Variables:
 discount_usd,
 product_material,
 product_category,
 shipping_time.





STATISTICS TOOLS AND TECHNIQUES FOR DATA ANALYSIS

Descriptive Analysis

- 1. Summarized data using functions like summary() for an overview of the dataset.
- 2. Visualizations (e.g., histograms and bar plots) to understand distributions.

 Correlation Analysis
- Pearson correlation coefficient was calculated to measure the strength of the linear relationship between price and discount (cor(price_usd, discount_usd)).



EXPLORATORY DATA ANALYSIS

- 1. Univariate Analysis: Histograms and bar charts revealed trends in discounts, product types, and top-selling countries.
- 2. Bivariate Analysis: Scatter plots and correlation analysis highlighted the relationship between price and discount.
- 3. Multivariate Analysis: Faceted plots and grouped summaries showed how product attributes like material, shipping time, and condition impacted pricing.



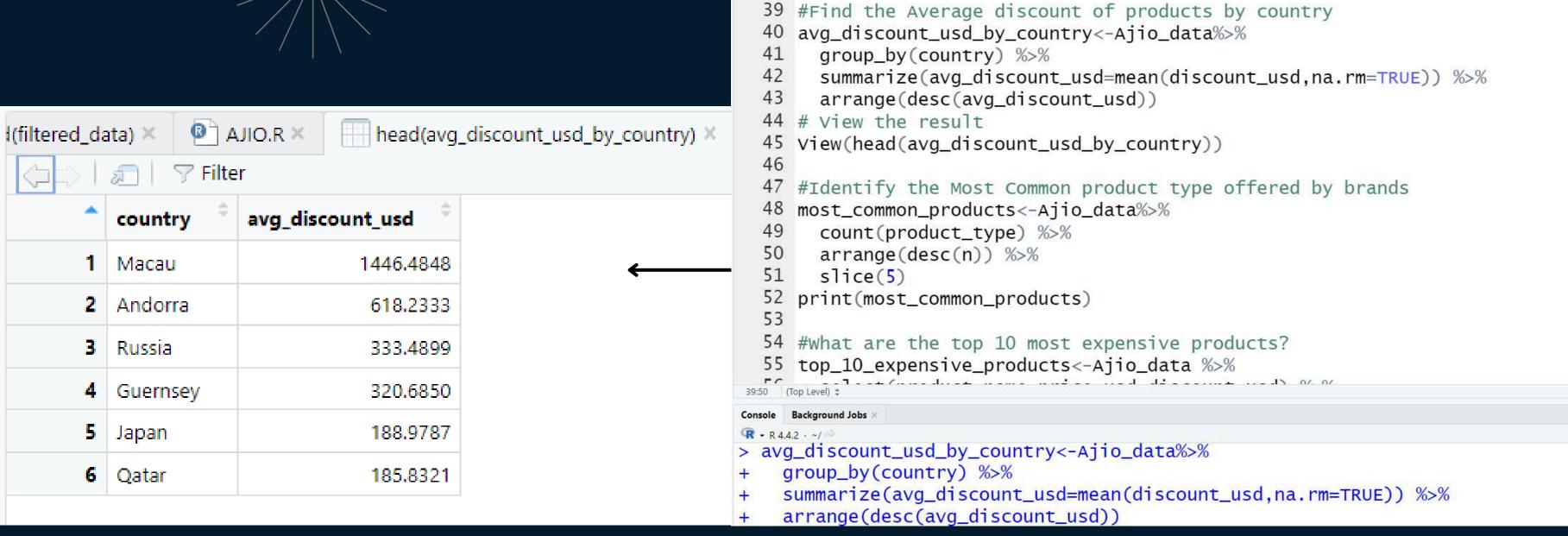
Find the Average discount of products by country?

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head(avg_discount_usd_by_country) ×





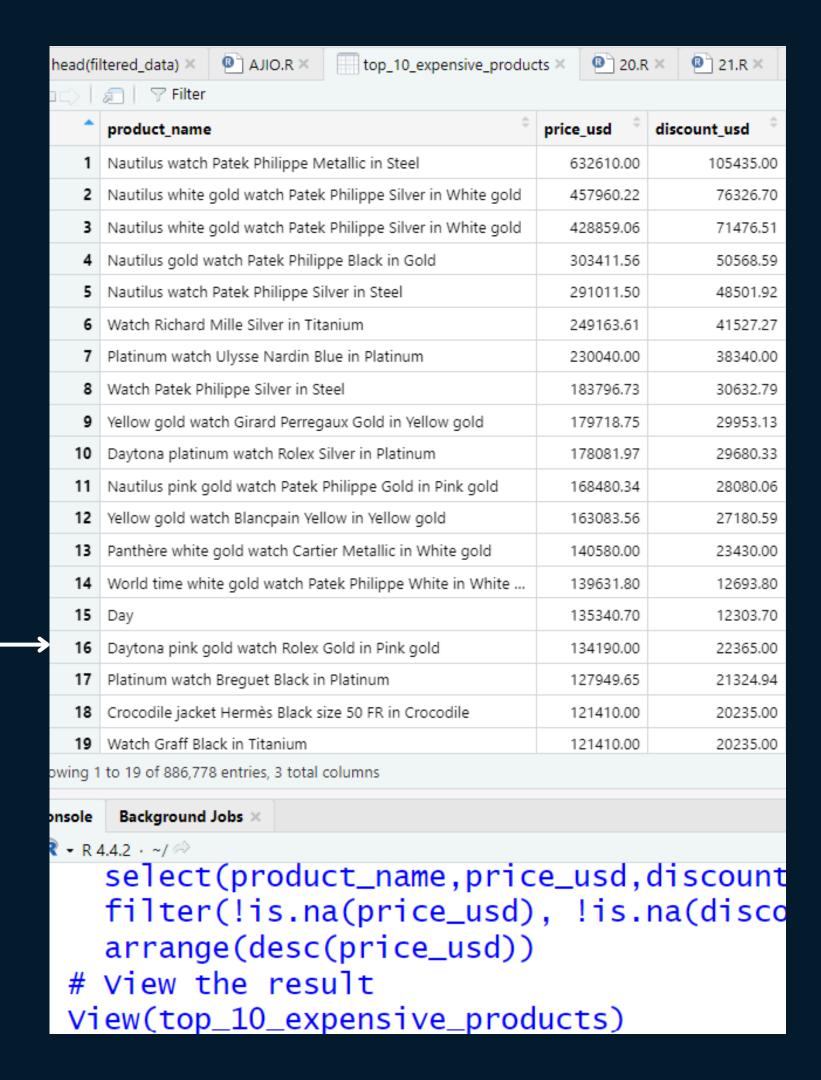
Identify the most common product type offered by brands?



```
l(filtered_data) × 🚇 AJIO.R × head(avg_discount_usd_by_country) × 🚇 20.R × 🚇 21.R × 🚇 21.1.R × 🚇 20.1.R ×
#Identify the Most Common product type offered by brands
     most_common_products<-Ajio_data%>%
        count(product_type) %>%
        arrange(desc(n)) %>%
        slice(5)
     print(most_common_products)
  54 #what are the top 10 most expensive products?
  55 top_10_expensive_products<-Ajio_data %>%
        select(product_name,price_usd,discount_usd) %>%
       filter(!is.na(pride_usd), !is.na(discount_usd))%>%
        arrange(desc(price_usd))
  59 # View the result
  60 View(top_10_expensive_products)
  62 #what are the top 10 most cheapest products?
     top_10_cheapest_products<-Ajio_data %>%
52:28 (Top Level) $
Console Background Jobs
R → R 4.4.2 · ~/ △
    arrange(desc(n)) %>%
    slice(5)
  print(most_common_products)
  product_type
         Shirt 21587
```

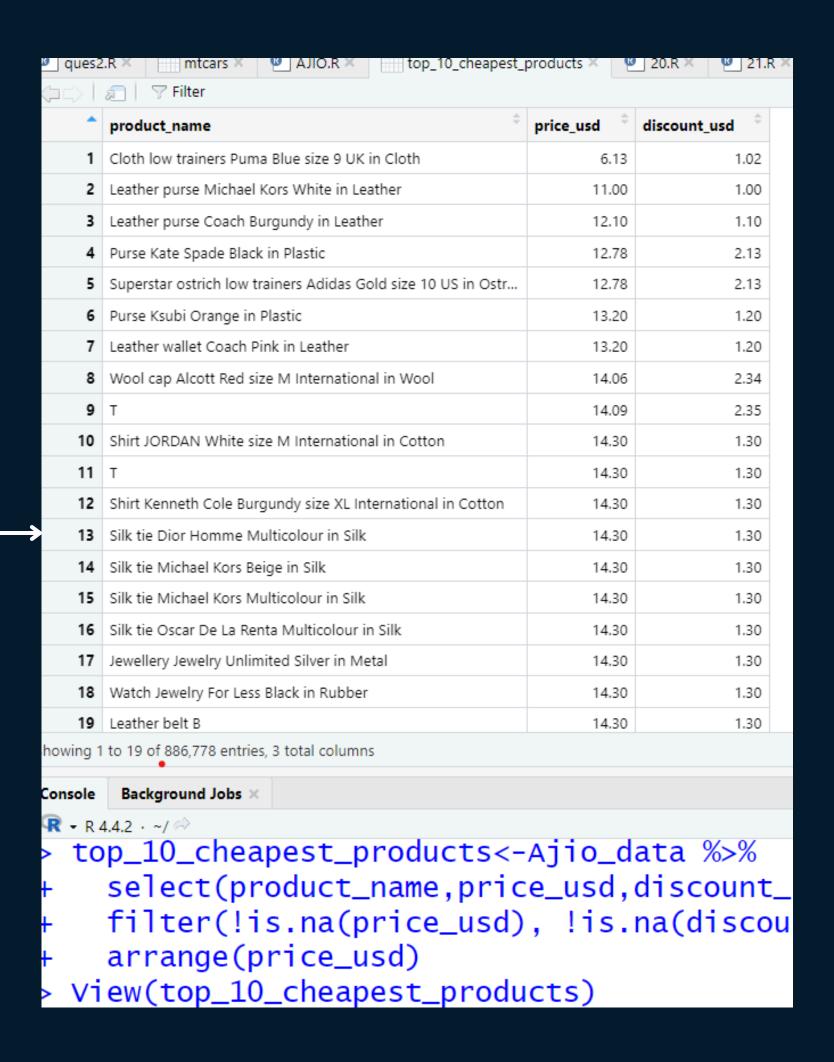
What are the top 10 most expensive products?

```
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54 #what are the top 10 most expensive products?
   top_10_expensive_products<-Ajio_data %>%
     select(product_name,price_usd,discount_usd) %>%
     filter(!is.na(price_usd), !is.na(discount_usd))%>%
     arrange(desc(price_usd))
59 # View the result
   View(top_10_expensive_products)
   #what are the top 10 most cheapest products?
   top_10_cheapest_products<-Ajio_data %>%
     select(product_name,price_usd,discount_usd) %>%
     filter(!is.na(price_usd), !is.na(discount_usd))%>%
     arrange(price_usd)
   # View the result
   View(top_10_cheapest_products)
   #Find the top 10 Highest-Brand Name by product type
top_10_expensive_products<-Ajio_data %>%
  select(product_name,price_usd,discount_usd) %>%
  filter(!is.na(price_usd), !is.na(discount_usd))%>%
  arrange(desc(price_usd))
```



Ques:what are top most cheapest products?

```
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  62 #what are the top 10 most cheapest products?
    top_10_cheapest_products<-Ajio_data %>%
       select(product_name,price_usd,discount_usd) %>%
       filter(!is.na(price_usd), !is.na(discount_usd))%>%
       arrange(price_usd)
    # View the result
    View(top_10_cheapest_products)
  70 #Find the top 10 Highest-Brand Name by product type
  71 top_10_highest_brand_name<-Ajio_data%>%
       select(product_type,brand_name,price_usd)%>%
       filter(!is.na(brand_name), !is.na(price_usd))%>%
       arrange(desc(brand_name))%>%
       slice(1:10)
    #view the result
     View(top_10_highest_brand_name)
> View(top_10_expensive_products)
> top_10_cheapest_products<-Ajio_data %>%
   select(product_name, price_usd, discount_usd) %>%
   filter(!is.na(price_usd), !is.na(discount_usd))%>%
   arrange(price_usd)
```



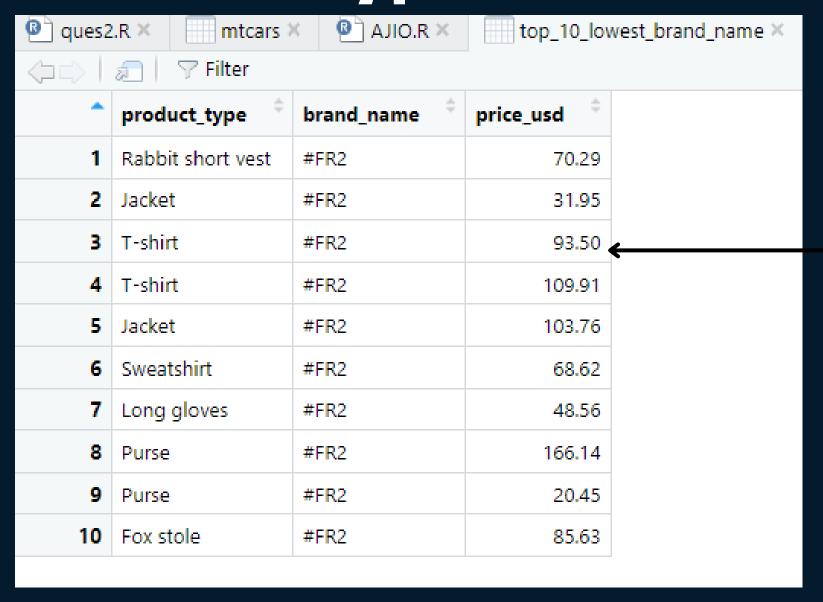
Find the top 10 Highest Brand Name by Product Type?

| a ques2 | .R × mtcars) | ⓐ AJIO.R × | top_10_hig | ghest_brand_name × | |
|---------|----------------|-------------------------|------------------------|--------------------|--|
| | | | | | |
| ^ | product_type | brand_name [‡] | price_usd [‡] | | |
| 1 | Sweatshirt | Ølaf | 77.56 | | |
| 2 | Straight pants | Études Studio | 98.41 | | |
| 3 | Trousers | Études Studio | 89.46 | | |
| 4 | Linen blouse | Études Studio | 52.50 | ← | |
| 5 | Sweatshirt | Études Studio | 83.07 | | |
| 6 | T-shirt | Études Studio | 76.68 | | |
| 7 | T-shirt | Études Studio | 49.84 | | |
| 8 | Sweatshirt | Études Studio | 83.07 | | |
| 9 | Pull | Études Studio | 76.68 | | |
| 10 | Wool trousers | Études Studio | 115.02 | | |
| | | | | | |



```
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62 #what are the top 10 most cheapest products?
   top_10_cheapest_products<-Ajio_data %>%
     select(product_name,price_usd,discount_usd) %>%
     filter(!is.na(price_usd), !is.na(discount_usd))%
     arrange(price_usd)
  # View the result
68 View(top_10_cheapest_products)
69
70 #Find the top 10 Highest-Brand Name by product typ
  top_10_highest_brand_name<-Ajio_data%>%
     select(product_type,brand_name,price_usd)%>%
     filter(!is.na(brand_name), !is.na(price_usd))%>%
     arrange(desc(brand_name))%>%
     slice(1:10)
76 #view the result
77 View(top_10_highest_brand_name)
sole Background Jobs
▼ R 4.4.2 · ~/ 👄
top_10_highest_brand_name<-Ajio_data%>%
  select(product_type,brand_name,price_usd)%>%
  filter(!is.na(brand_name), !is.na(price_usd))%>%
  arrange(desc(brand_name))%>%
  slice(1:10)
```

Find the Top 10 Lowest Brand Name by Product Type?





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mtcars × @ AJIO.R × @ 20.R ×
                               @ 21.1.R × @ 20.1.R ×
                          @ 21.R ×
18
  79 #Find the Top 10 lowest-Brand Name by product type
     top_10_lowest_brand_name<-Ajio_data%>%
       select(product_type,brand_name,price_usd)%>%
       filter(!is.na(brand_name), !is.na(price_usd))%>%
       arrange(brand_name)%>%
       slice(1:10)
     #view the result
     View(top_10_lowest_brand_name)
  87
     #what is the correlation between price and discount
     correlation <- cor(Ajio_data$price_usd, Ajio_data$c
  90
                         = "pearson")
     print(correlation)
  92
    #what is the range of price and discount
     range_price_usd <- range(Ajio_data$price_usd)</pre>
     print(paste("Range of Price:", diff(range_price_use
     Background Jobs
R - R 4.4.2 · ~/ @
> top_10_lowest_brand_name<-Ajio_data%>%
    select(product_type,brand_name,price_usd)%>%
    filter(!is.na(brand_name), !is.na(price_usd))%>%
    arrange(brand_name)%>%
    slice(1:10)
```

What is the correlation between price and discount?



```
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  87
     #what is the correlation between price and discount
     correlation <- cor(Ajio_data$price_usd, Ajio_data$discount_usd, method
                        = "pearson")
  91 print(correlation)
  93 #what is the range of price and discount
     range_price_usd <- range(Ajio_data$price_usd)</pre>
     print(paste("Range of Price:", diff(range_price_usd)))
     range_discount_usd <- range(Ajio_data$discount_usd)</pre>
     print(paste("Range of Discount:", diff(range_discount_usd)))
  99 #Analyze the impact of product material on Price of product
 100 product_material_impact<-Ajio_data %>%
       group_by(product_material)%>%
 101
       summarize(avg_price_usd=mean(pride_usd,na.rm = TRUE))
 102
 103 # View the result
 101 Viou(product material impact)
Console Background Jobs
R - R 4.4.2 · ~/ @
correlation"
> correlation <- cor(Ajio_data$price_usd, Ajio_data$discount_usd, method
                    = "pearson")
 print(correlation)
   0.9408314
```

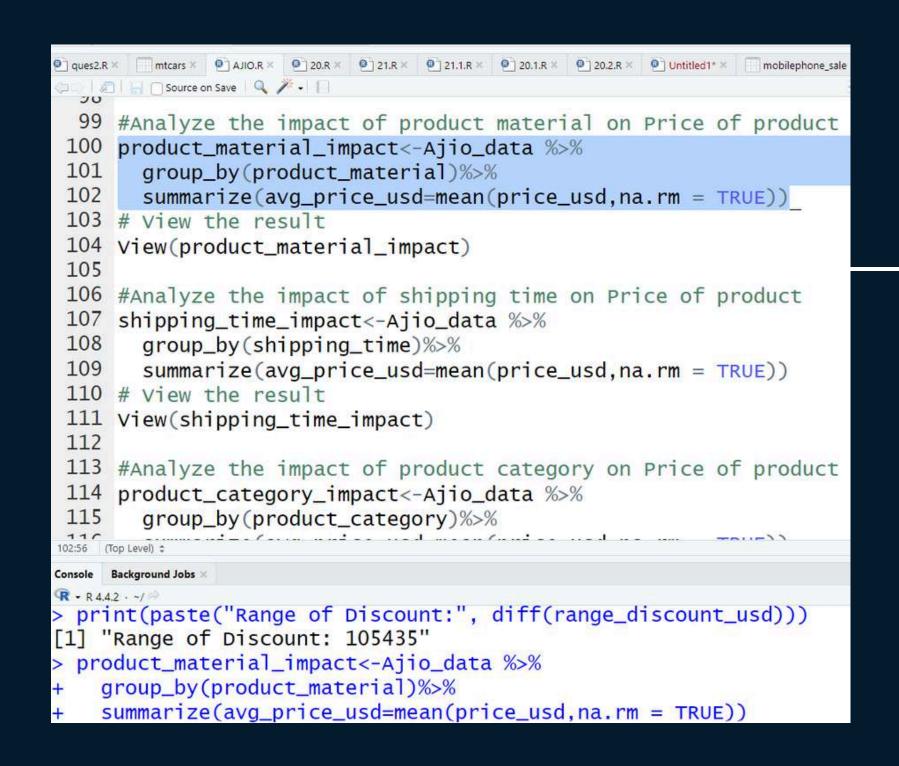
What is the range of price and product?



```
mtcars × B AJIO.R × B 20.R ×

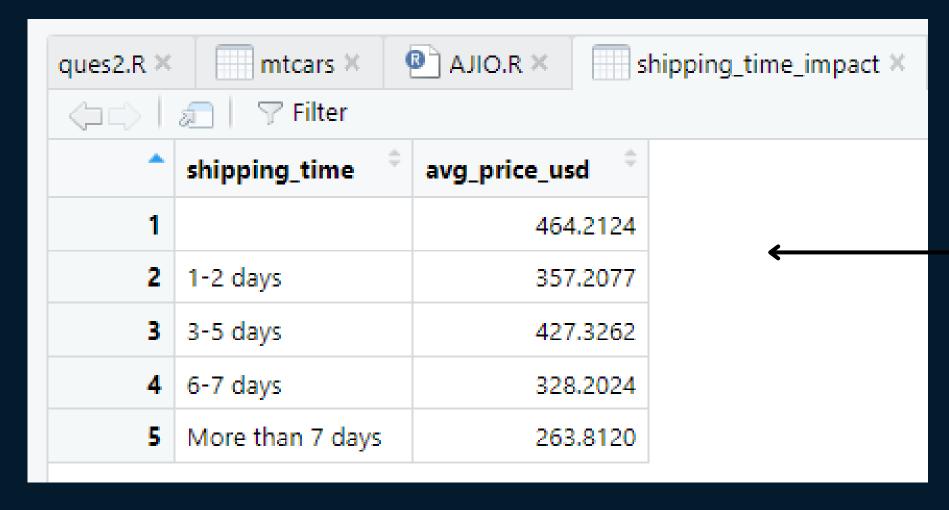
    □ 21.R × □ 21.1.R × □ 20.1.R × □ 20.2.R × □ 20
 92
      93 #what is the range of price and discount
                range_price_usd <- range(Ajio_data$price_usd)</pre>
                print(paste("Range of Price:", diff(range_price_usd)))
               range_discount_usd <- range(Ajio_data$discount_usd)</pre>
                print(paste("Range of Discount:", diff(range_discount_usd)))
      98
      99 #Analyze the impact of product material on Price of product
               product_material_impact<-Ajio_data %>%
                      group_by(product_material)%>%
   101
                      summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
   102
   103 # View the result
              View(product_material_impact)
   105
   106 #Analyze the impact of shipping time on Price of product
                shipping_time_impact<-Ajio_data %>%
   108
                      group_by(shipping_time)%>%
                      summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
             (Top Level) $
              Background Jobs
 R - R 4.4.2 · ~/ →
> print(paste("Range of Price:", diff(range_price_usd)))
[1] "Range of Price: 632603.87"
> range_discount_usd <- range(Ajio_data$discount_usd)</pre>
> print(paste("Range of Discount:", diff(range_discount_usd)))
[1] "Range of Discount: 105435"
```

Ques:- Analyze the impact of product material on price of product



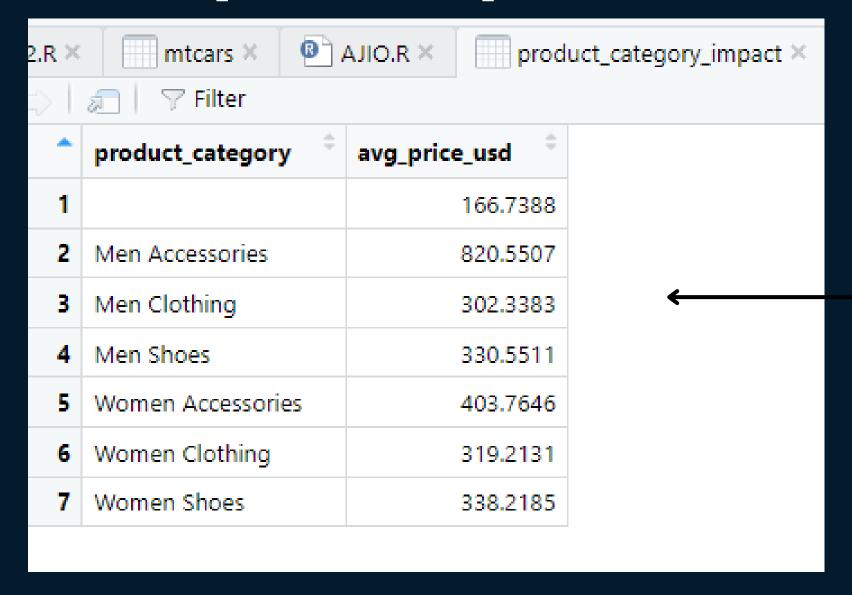
| ques2.R × mtcars × | | | product_material_impact \times | | |
|----------------------|--------------------|--------------------------|----------------------------------|--|--|
| (□□) @□ ♥ Filter | | | | | |
| ^ | product_material | avg_price_usd $^{\circ}$ | | | |
| 1 | | 109.2700 | | | |
| 2 | Alligator | 1837.2400 | | | |
| 3 | Astrakhan | 645.3091 | | | |
| 4 | Beaver | 761.7240 | | | |
| 5 | Cashmere | 608.1975 | | | |
| 6 | Ceramic | 2615.6752 | | | |
| 7 | Chain | 1053.1495 | | | |
| 8 | Chinchilla | 1153.2249 | | | |
| 9 | Cloth | 319.5333 | | | |
| 10 | Cotton | 226.6185 | | | |
| 11 | Cotton - elasthane | 160.6911 | | | |
| 12 | Crocodile | 1464.3303 | | | |
| 13 | Crystal | 488.5524 | | | |
| 14 | Denim - Jeans | 258.1757 | | | |
| 15 | Eel | 324.1735 | | | |
| 16 | Exotic leathers | 461.5733 | | | |
| 17 | Faux fur | 338.9097 | | | |
| 18 | Fox | 533.0806 | | | |
| 19 | Fur | 494.8707 | | | |

Analyze the impact of Shipping Time on price of product



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105
106 #Analyze the impact of shipping time on Price of product
     shipping_time_impact<-Ajio_data %>%
108
       group_by(shipping_time)%>%
       summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
109
110 # View the result
111 View(shipping_time_impact)
112
113 #Analyze the impact of product category on Price of product
    product_category_impact<-Ajio_data %>%
      group_by(product_category)%>%
115
       summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
116
117 # View the result
118 View(product_category_impact)
119
120 #univariate Analysis
121 # Distribution of a numerical variable(eg-discount)
122 ggplot(Ajio_data, aes(x=discount_usd)) +
    Background Jobs
> # View the result
> View(Ajio_data)
> shipping_time_impact<-Ajio_data %>%
   group_by(shipping_time)%>%
   summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
```

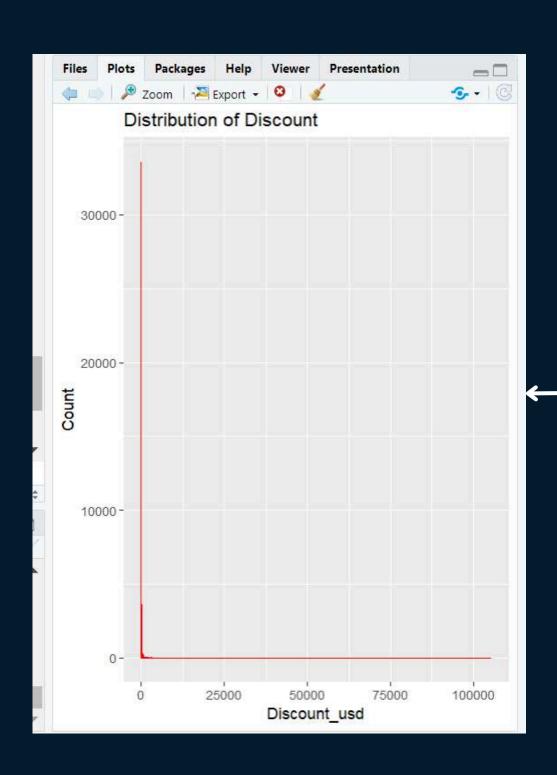
Analyze the impact of Product Category on price of product



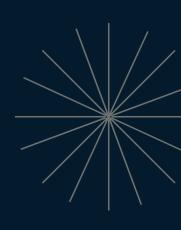


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 105
 106 #Analyze the impact of shipping time on Price of product
 107 shipping_time_impact<-Ajio_data %>%
 108
       group_by(shipping_time)%>%
       summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
 109
 110 # View the result
 111 View(shipping_time_impact)
 112
 113 #Analyze the impact of product category on Price of product
 114 product_category_impact<-Ajio_data %>%
       group_by(product_category)%>%
 115
 116
       summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
 117 # View the result
 118 View(product_category_impact)
 119
 120 #univariate Analysis
 121 # Distribution of a numerical variable(eg-discount)
     ggplot(Ajio_data, aes(x=discount_usd)) +
    summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
> View(shipping_time_impact)
> product_category_impact<-Ajio_data %>%
    group_by(product_category)%>%
    summarize(avg_price_usd=mean(price_usd,na.rm = TRUE))
```

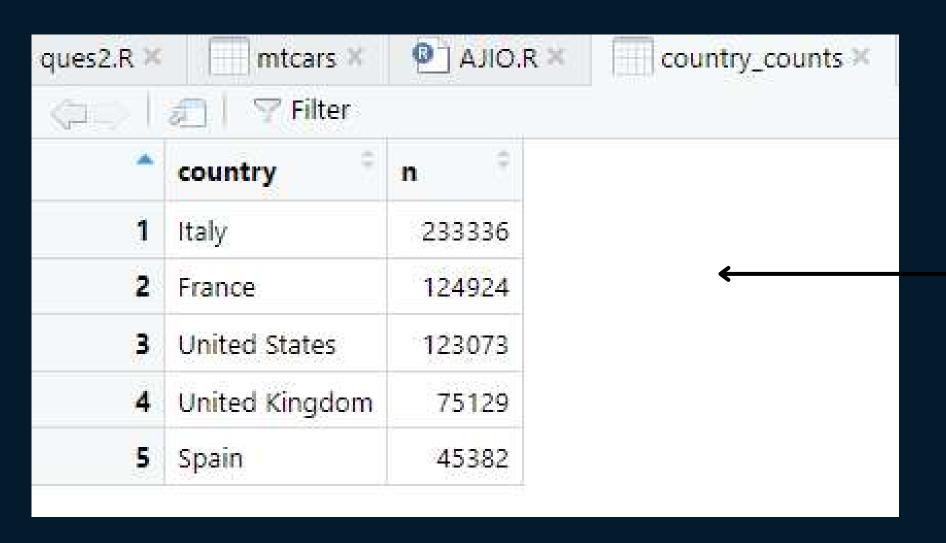
Ques:- Histogram of single variable discount



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120 #univariate Analysis
121 # Distribution of a numerical variable(eg-discount)
122 ggplot(Ajio_data, aes(x=discount_usd)) +
       geom_histogram(binwidth=0.5, fill="blue",
124
                       color="red") +
       labs(title="Distribution of Discount",
125
126
            x="Discount_usd", y="Count")
127
128 # Calculate the count of each product material
129 country_counts <- Ajio_data %>%
130
       count(country) %>%
       arrange(desc(n)) %>%
131
       slice(1:5) # Keep only the top 5 product material
133 # View the result
134 View(country_counts)
135
136 # Create the bar plot for the top 5 country
126:36 (Top Level) $
Console Background Jobs
> ggplot(Ajio_data, aes(x=discount_usd)) +
   geom_histogram(binwidth=0.5, fill="blue",
                    color="red") +
   labs(title="Distribution of Discount",
        x="Discount_usd", y="Count")
```



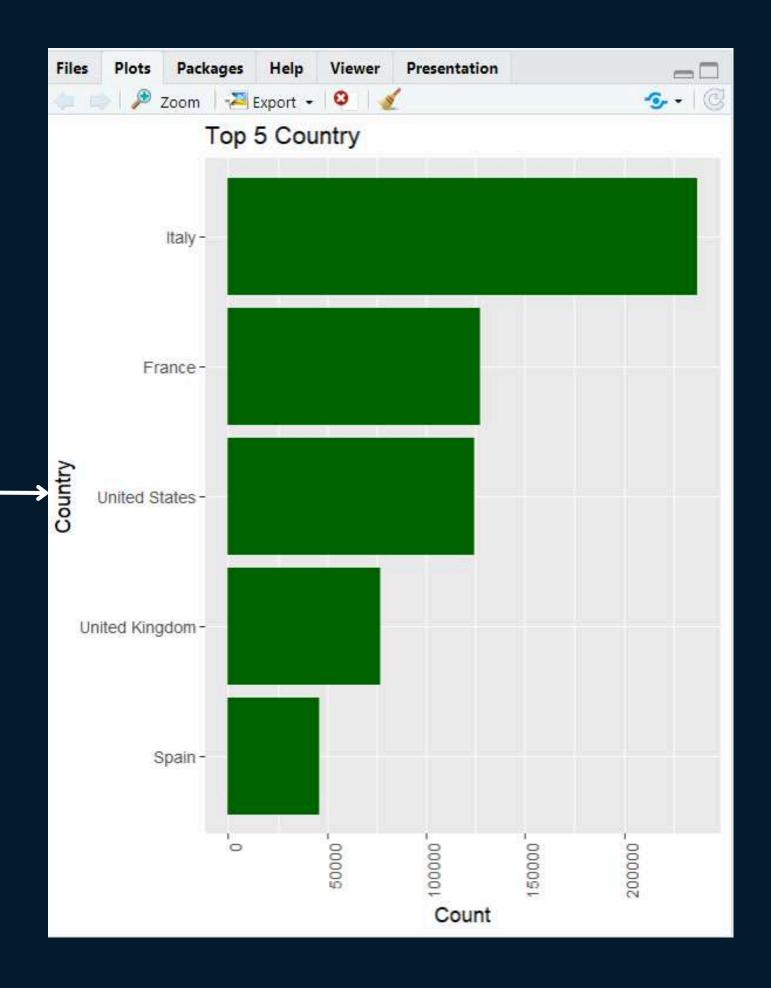
Calculate the count of each product material?



```
Ajio_data
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127
    # Calculate the count of each product material
    country_counts <- Ajio_data %>%
130
       count(country) %>%
131
      arrange(desc(n)) %>%
132
      slice(1:5) # Keep only the top 5 product materia
133 # View the result
134 View(country_counts)
135
    # Create the bar plot for the top 5 country
    ggplot(country\_counts, aes(x = reorder(country, n))
138
      geom_bar(stat = "identity", fill = "pink") +
139
      labs(title = "Top 5 Country", x = "Country", y =
140
      theme(axis.text.x = element_text(angle = 90, hjus
141
      coord_flip()
142
143 #Bivariate Analysis
    # Scatter plot between two numerical variables (pr
onsole Background Jobs
R - R 4.4.2 · ~/ 🖘
        x="Discount_usd", y="Count")
 country_counts <- Ajio_data %>%
   count(country) %>%
   arrange(desc(n)) %>%
   slice(1:5)
```

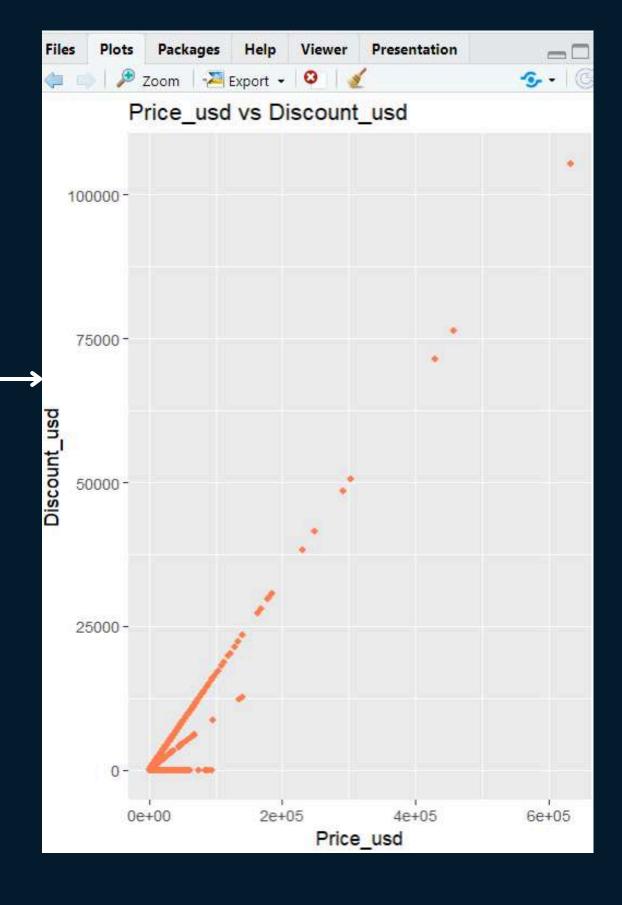
Bar Plot for Top 5 Country

```
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    Source  
27
28 # Calculate the count of each product material
29 country_counts <- Ajio_data %>%
                count(country) %>%
131
                arrange(desc(n)) %>%
                 slice(1:5) # Keep only the top 5 product material
.33 # View the result
.34 View(country_counts)
135
.36 # Create the bar plot for the top 5 country
          ggplot(country\_counts, aes(x = reorder(country, n), y = n)) +
                 geom_bar(stat = "identity", fill = "darkgreen") +
                 labs(title = "Top 5 Country", x = "Country", y = "Count") +
                 theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
41
                 coord_flip()
42
43 #Bivariate Analysis
44 # Scatter plot between two numerical variables (price and disc
 ggplot(country\_counts, aes(x = reorder(country, n), y = n)) +
      geom_bar(stat = "identity", fill = "darkgreen") +
      labs(title = "Top 5 Country", x = "Country", y = "Count") +
       theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
       coord_flip()
```



Ques:- between price and discount

```
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     COULUTITE (
L42
L43 #Bivariate Analysis
L44 # Scatter plot between two numerical variables (price and discount)
L45 ggplot(Ajio_data, aes(x=price_usd, y=discount_usd)) +
     geom_point(color="coral") +
L47
     labs(title="Price_usd vs Discount_usd", x="Price_usd", y="Discount_usd")
L48
L49 #Calculate the count and percentage of each product category
   product_category_counts <- Ajio_data %>%
L51
     count(product_category) %>%
     mutate(percentage = n / sum(n) * 100)
L53
L54 # Create a pie chart with percentages
   ggplot(product\_category\_counts, aes(x = "", y = percentage,
L56
                                        fill = product_category)) +
L57
     geom_bar(width = 1, stat = "identity") +km
     coord_polar(theta = "y")
sole Background Jobs
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  coord_flip()
ggplot(Ajio_data, aes(x=price_usd, y=discount_usd)) +
  geom_point(color="coral") +
  labs(title="Price_usd vs Discount_usd", x="Price_usd", v="Discount_usd")
```



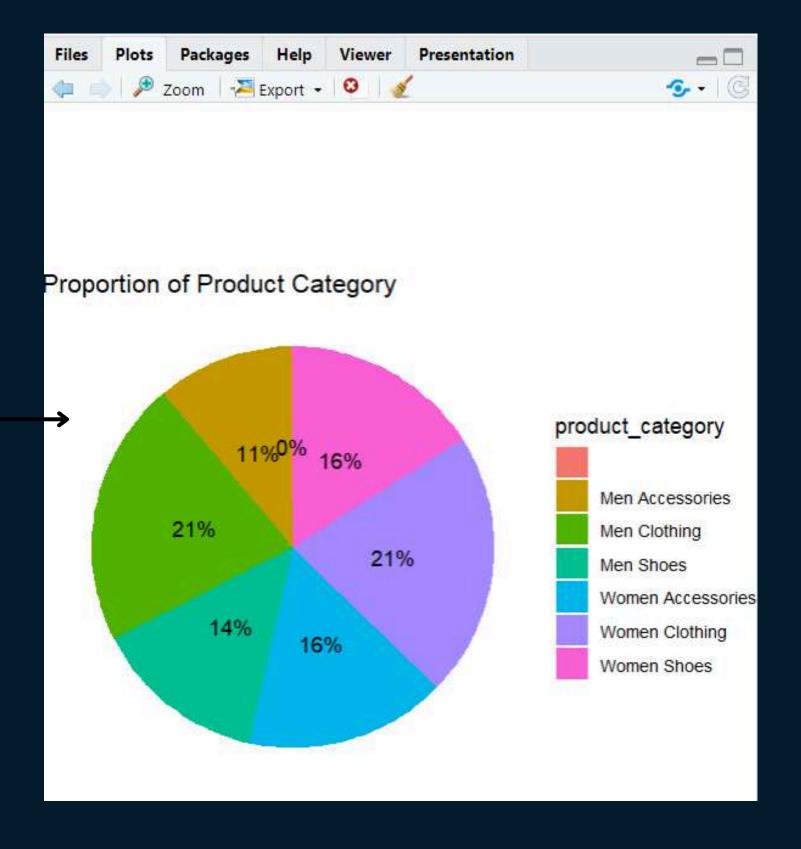
Calculate the count and percentage of each product category?



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142
 143 #Bivariate Analysis
    # Scatter plot between two numerical variables (price and dis
     ggplot(Ajio_data, aes(x=price_usd, y=discount_usd)) +
       geom_point(color="coral") +
 146
       labs(title="Price_usd vs Discount_usd", x="Price_usd", y="D
 147
 148
 149 #calculate the count and percentage of each product category
     product_category_counts <- Ajio_data %>%
 151
       count(product_category) %>%
 152
       mutate(percentage = n / sum(n) * 100)
     print(product_category_counts)
154
     # Create a pie chart with percentages
    Background Jobs
R • R 4.4.2 · ~/ ∅
> print(product_category_counts)
   product_category
                          n percentage
                      1183 0.1313694
2
   Men Accessories 98776 10.9688467
       Men clothing 193082 21.4413102
          Men Shoes 125425 13.9281566
 Women Accessories 146018 16.2149617
     Women clothing 192197 21.3430330
        Women Shoes 143833 15.9723225
```

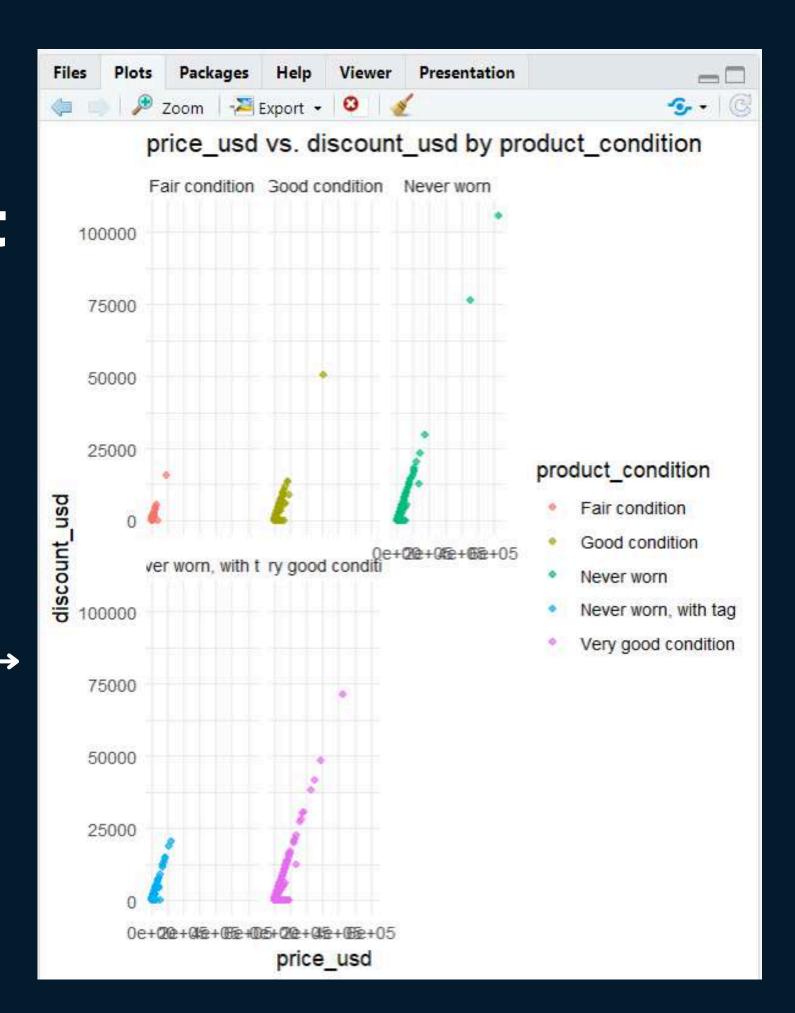
Ques:- Pie Chart of Product Category

```
mutate(percentage = n / sum(n) * 100)
153 print(product_category_counts)
154
155 # Create a pie chart with percentages
     ggplot(product\_category\_counts, aes(x = "", y = percentage,
                                      fill = product_category)) +
157
158
      geom_bar(width = 1, stat = "identity") +
      coord_polar(theta = "y") +
 159
160
      geom_text(aes(label = paste0(round(percentage,0), "%")),
                position = position_stack(viust = 0.5)) +
 161
      labs(title = "Proportion of Product Category") +
 162
      theme_void()
 163
 164
165 #Faceted plot to analyze multiple subsets
    Background Jobs
R + R 4.4.2 · ~/ ∅
Error: object 'km' not found
 ggplot(product\_category\_counts, aes(x = "", y = percentage,
                                   fill = product_category)) +
   geom_bar(width = 1, stat = "identity") +
   coord_polar(theta = "y") +
   geom_text(aes(label = paste0(round(percentage,0), "%")),
             position = position_stack(vjust = 0.5)) +
   labs(title = "Proportion of Product Category") +
   theme_void()
```



Faceted plot to analyze price vs discount across different product condition

```
@ ques2.R × mtcars × @ AJIO.R* × @ 20.R × @ 21.R × @ 21.1.R × @ 20.1.R × Ajio_data × @ 20.2.R × @ mobilephone.R
   🗐 📗 🗌 Source on Save
       tneme_vola()
 164
     #Faceted plot to analyze multiple subsets
     #(e.g., price vs discount across different product condition)
     ggplot(Ajio_data, aes(x = price_usd, y = discount_usd)) +
 168
       geom_point(aes(color = product_condition), alpha = 0.7) +
       facet_wrap(~ product_condition) +
 169
       labs(title = "price_usd vs. discount_usd by product_condition
 170
       x = "price_usd", y = "discount_usd") +
171
 172
       theme_minimal()
 173
 174
 175
 176
               position = position_stack(vjust = 0.5)) +
    labs(title = "Proportion of Product Category") +
    theme_void()
 ggplot(Ajio_data, aes(x = price_usd, y = discount_usd)) +
    geom_point(aes(color = product_condition), alpha =0.7) +
    facet_wrap(~ product_condition) +
    labs(title = "price_usd vs. discount_usd by product_condition"
    x = "price_usd", y = "discount_usd") +
    theme_minimal()
```



Findings

- Pricing and Discounts:
- -The range of prices and discounts was significant, highlighting product diversity.
- -A moderate positive correlation was found between price and discount.
 - Product Trends:
- -Certain materials and categories were associated with higher average prices.
- -Shipping time influenced product pricing, with faster shipping times correlating with higher prices.







- -The most common product types and their associated brands were identified.
- -Countries with the highest average discounts were determined, indicating regional pricing strategies.

Visual Insights:

- -Scatter plots and faceted plots revealed trends in price vs. discount for different product conditions.
- -Pie charts illustrated the distribution of product categories

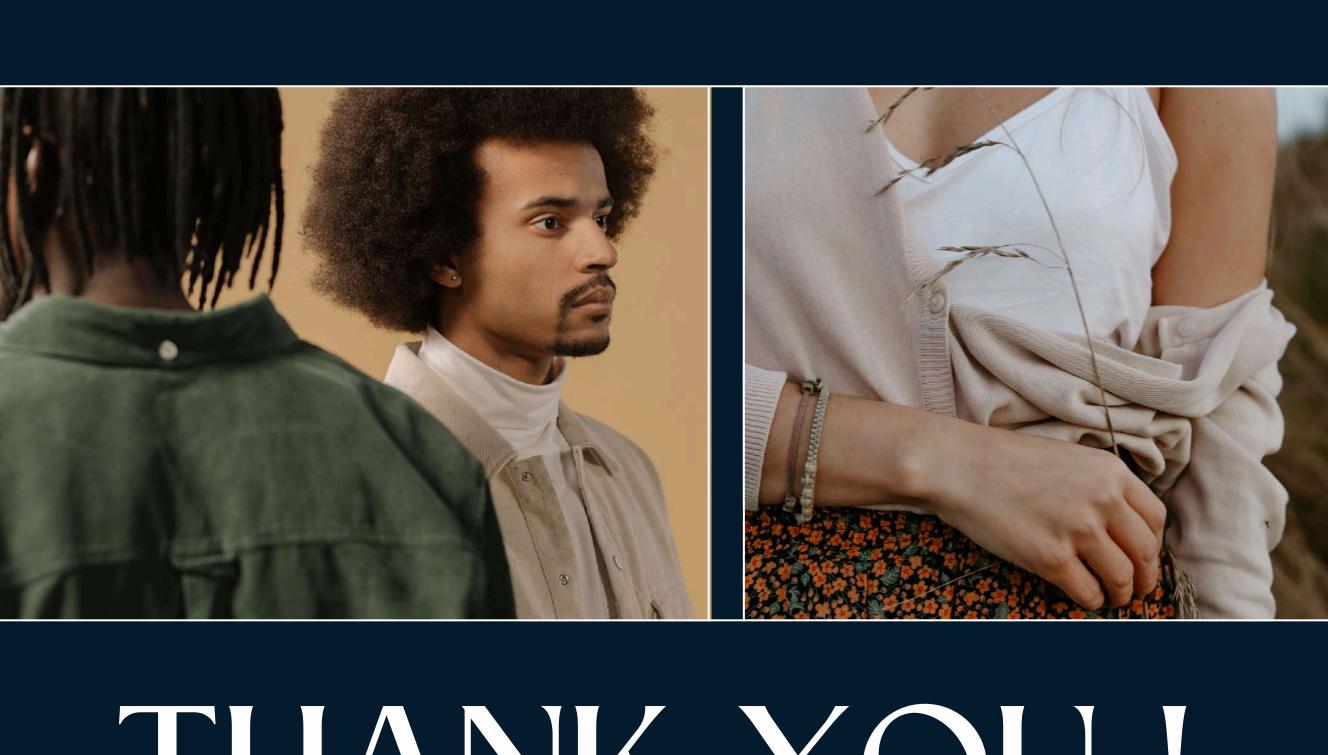




Conclusion



This project provided valuable insights into AJIO's product data, aiding in better understanding market trends and customer preferences. The analysis highlights opportunities for optimizing pricing strategies and product offerings.



THANK YOU!

