

**PRESENTATION**

# **Data Analysis & Visualization**

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# 1.HR DATA





## 1. Distribution of Job Titles:

### What the visualization shows:

The frequency of different job titles in the dataset.

### Why I used this visualization:

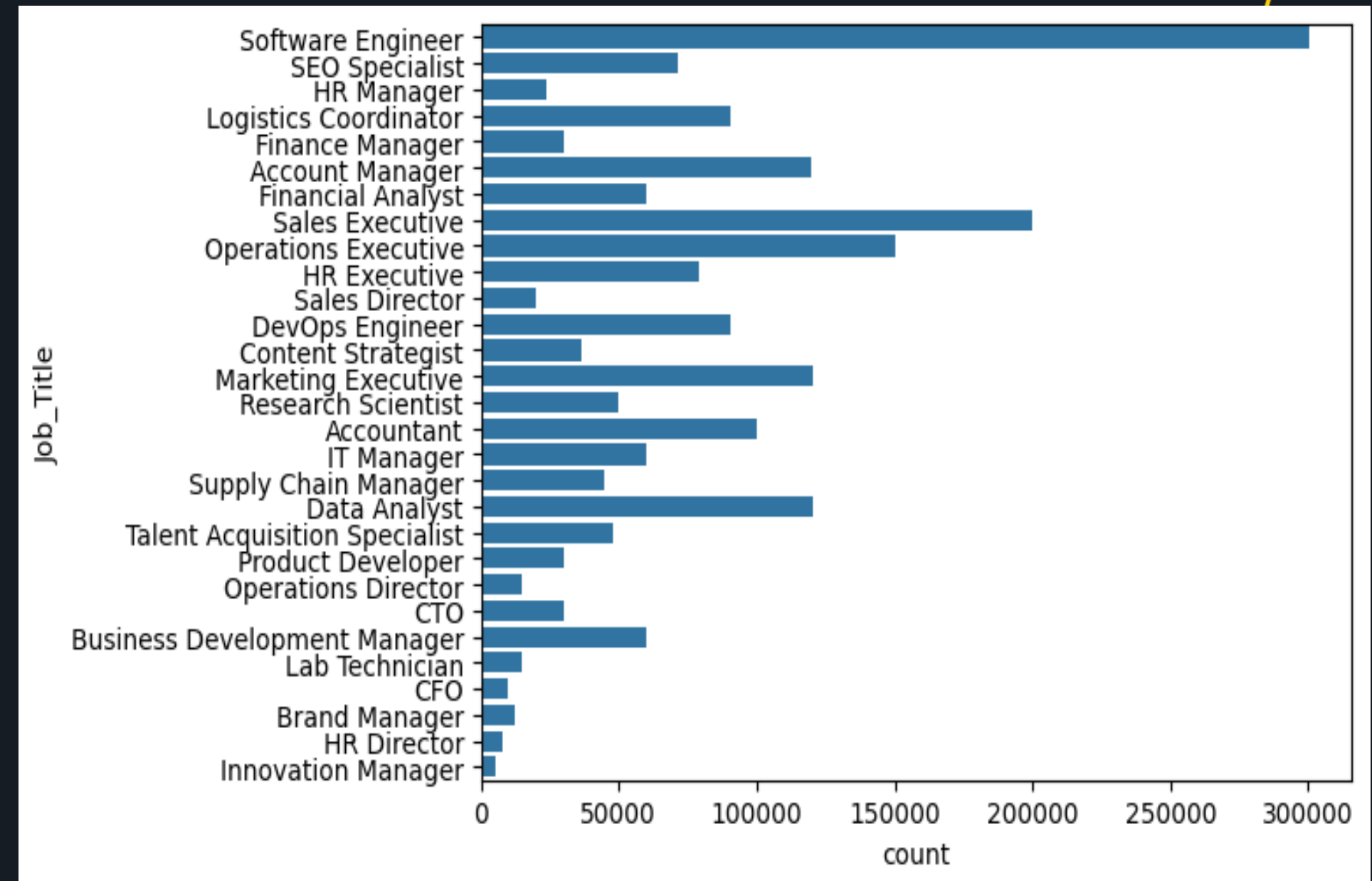
A horizontal bar chart makes it easy to compare the number of employees in each role.

It highlights which job positions are the most common.

### Key takeaway:

*Software Engineer* is by far the most frequent role.

Other roles with high counts include *Sales Executive* and *Operations Executive*.





## 2. Distribution of Employee Status

### What the visualization shows:

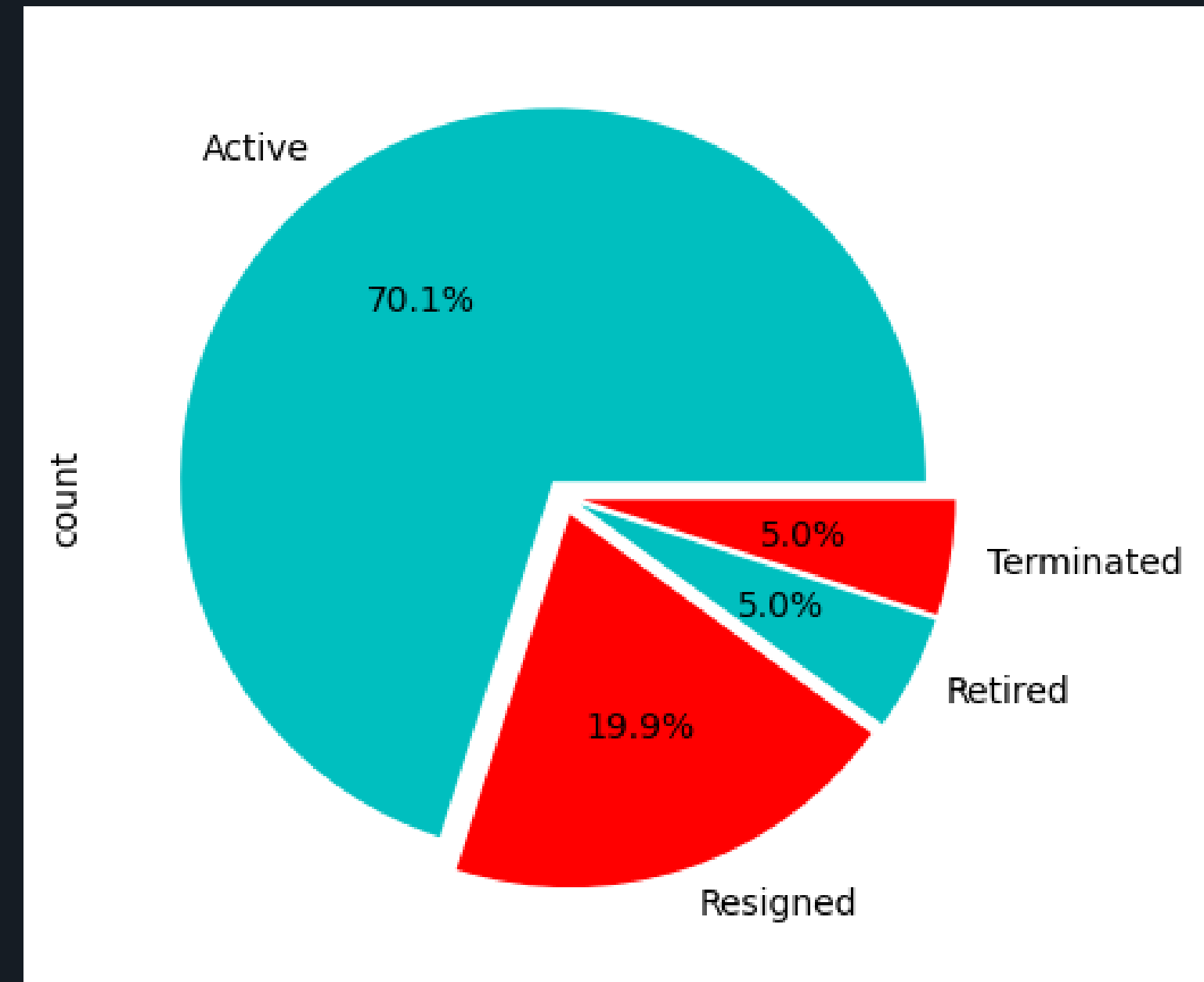
The proportion of employees by status (e.g., Active, Resigned, Retired).

### Why I used this visualization:

A pie chart clearly shows the share of each category and allows quick comparison of employment status across the workforce.

### Key takeaway:

The majority of employees are currently active, while only a small fraction are resigned or retired.





### 3. Employees by Department

#### What the visualization shows:

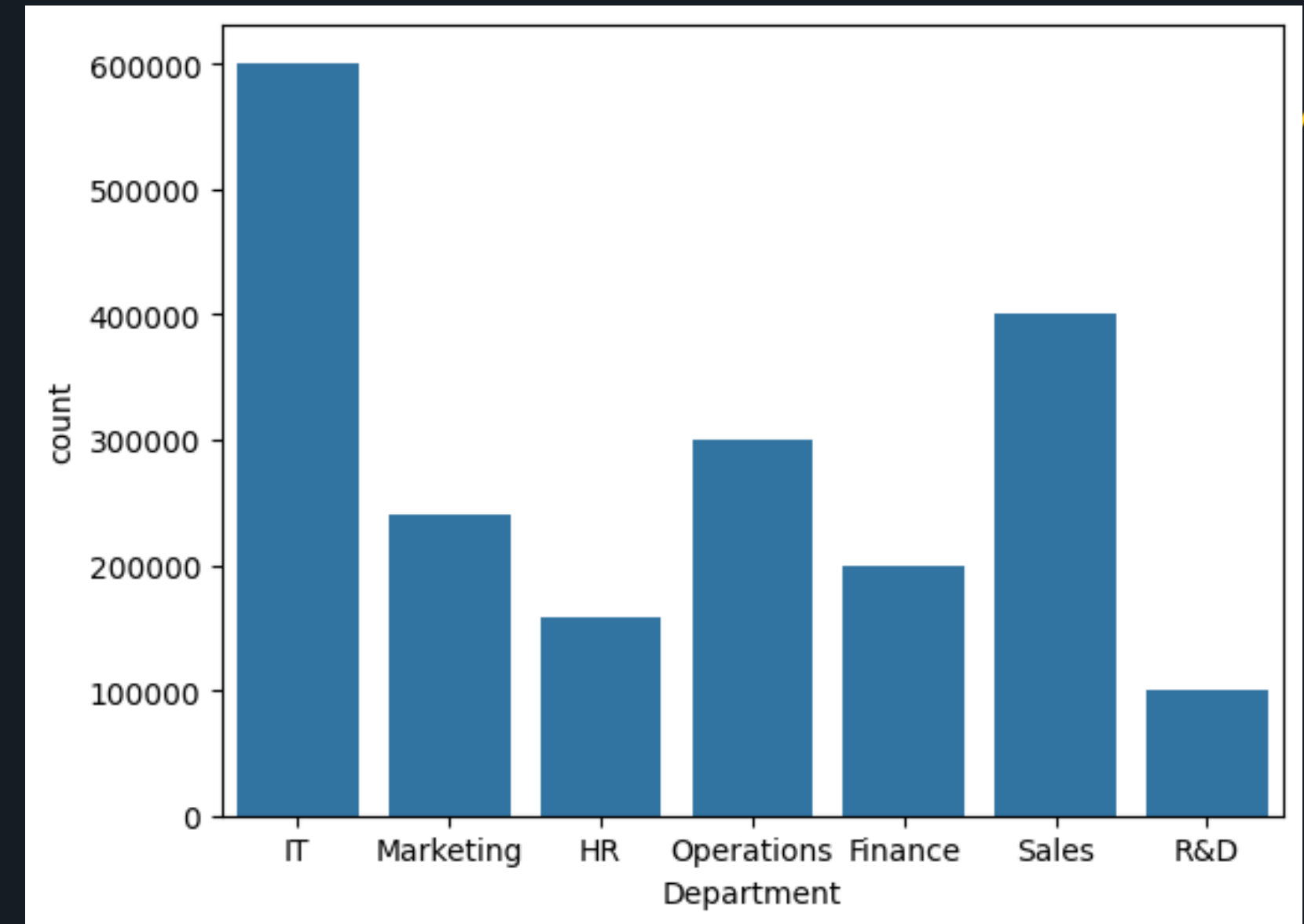
The number of employees in each department.

#### Why I used this visualization:

A bar chart allows us to compare department sizes and identify where the largest groups of employee's work.

#### Key takeaway:

IT and Sales departments have the highest employee counts, while smaller departments like HR or R&D employ fewer people.





## 4. Correlation Heatmap

### What the visualization shows:

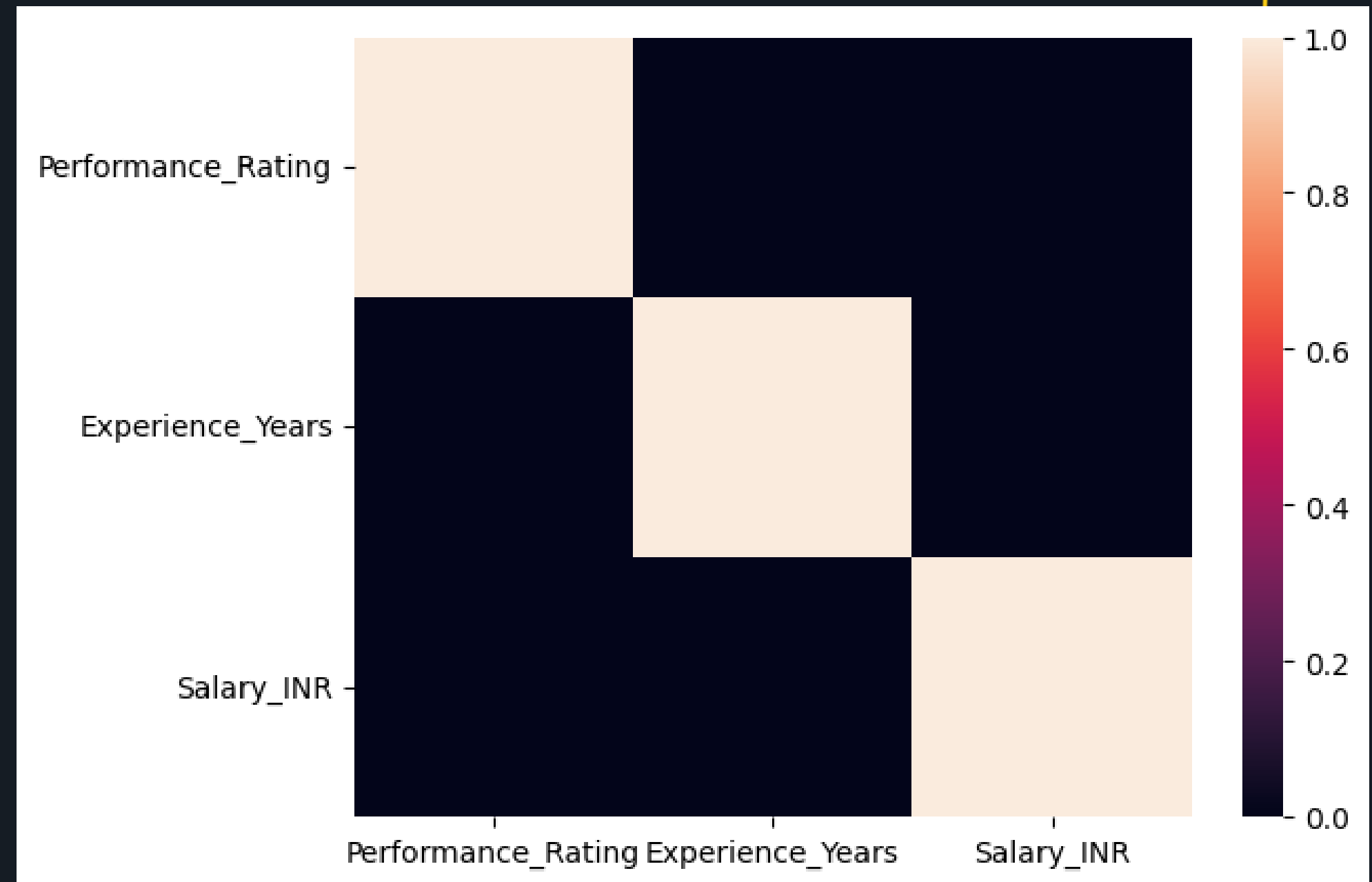
The relationships between numerical variables such as salary, age, performance rating, and years of experience.

### Why I used this visualization:

A heatmap helps identify strong positive or negative correlations, which can reveal important patterns such as whether higher experience leads to higher salary.

### Key takeaway:

Salary is positively correlated with experience and performance rating, suggesting that both factors influence pay.





## 5. Distribution of Hire Years

### What the visualization shows:

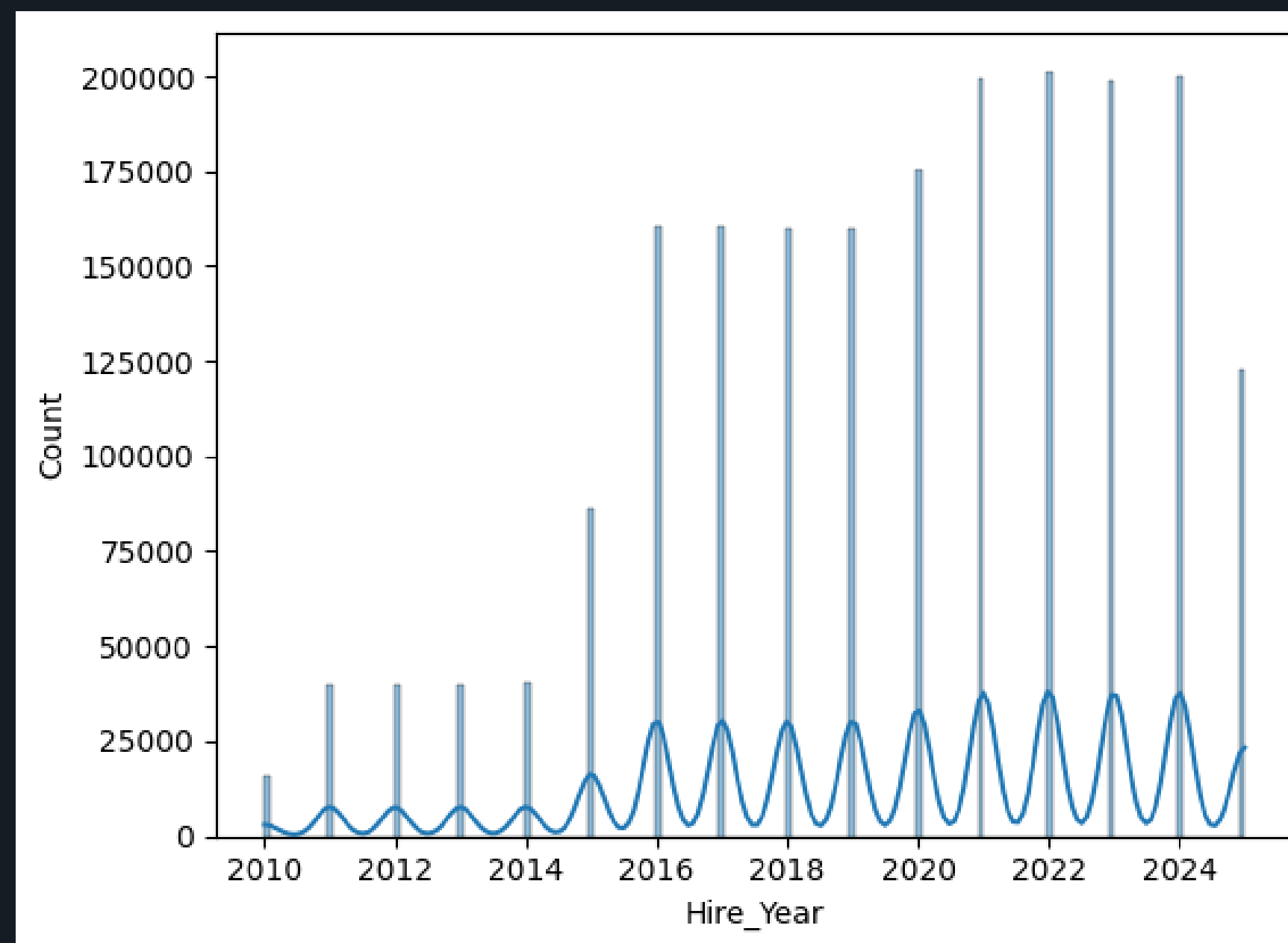
The frequency of employees hired in each year.

### Why I used this visualization:

A histogram highlights hiring trends over time, showing whether recruitment increased or decreased in certain years.

### Key takeaway:

Hiring peaked in specific years, indicating periods of major recruitment activity.





## 6. Average Salary by Department

### What the visualization shows:

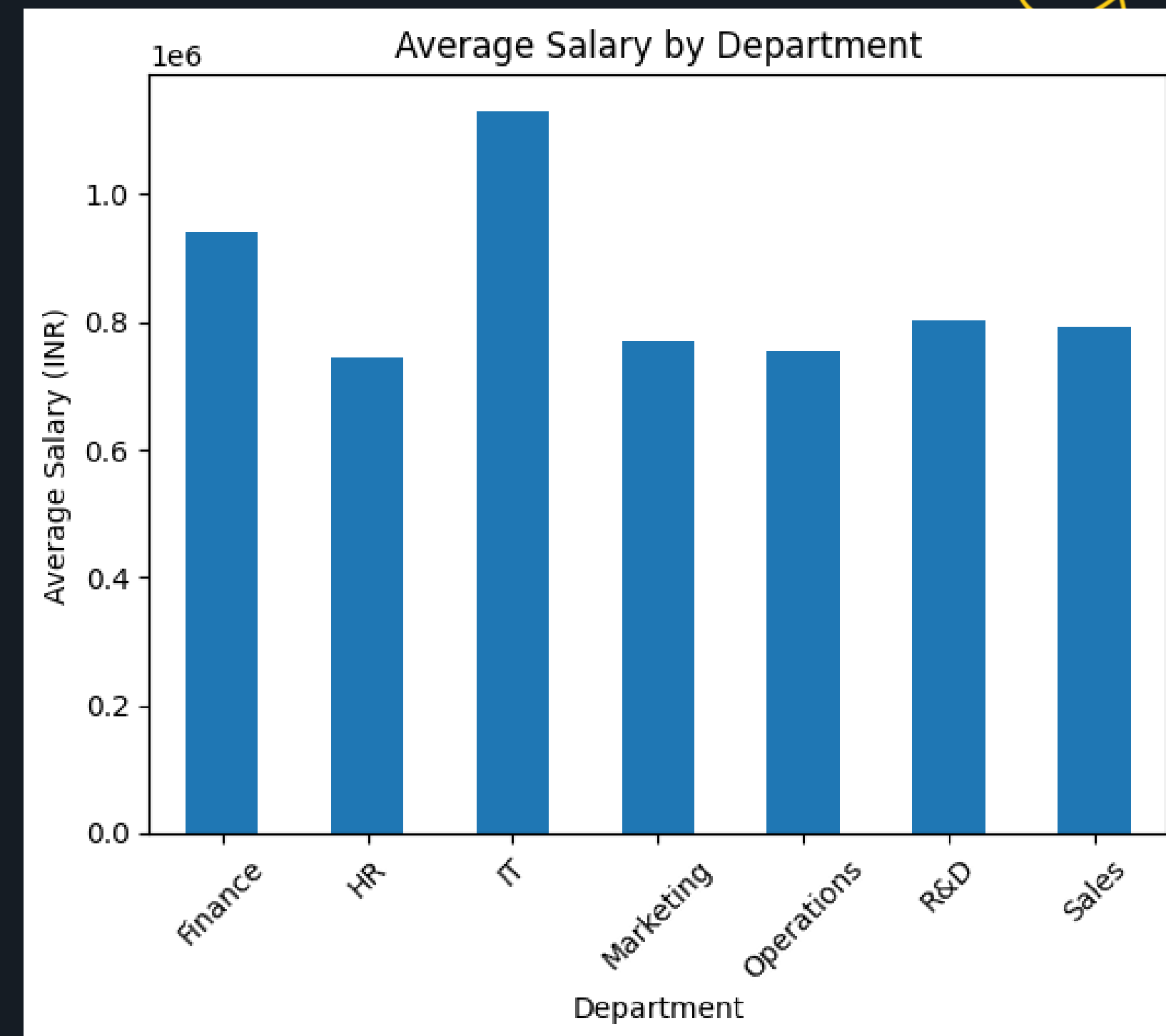
The mean salary for employees grouped by department.

### Why I used this visualization:

A bar chart makes it easy to compare average salaries across departments and see which fields are more highly paid.

### Key takeaway:

Departments such as Finance and R&D have higher average salaries, while departments like Operations and HR have comparatively lower salaries.







## 7. Salary Distribution

### What the visualization shows:

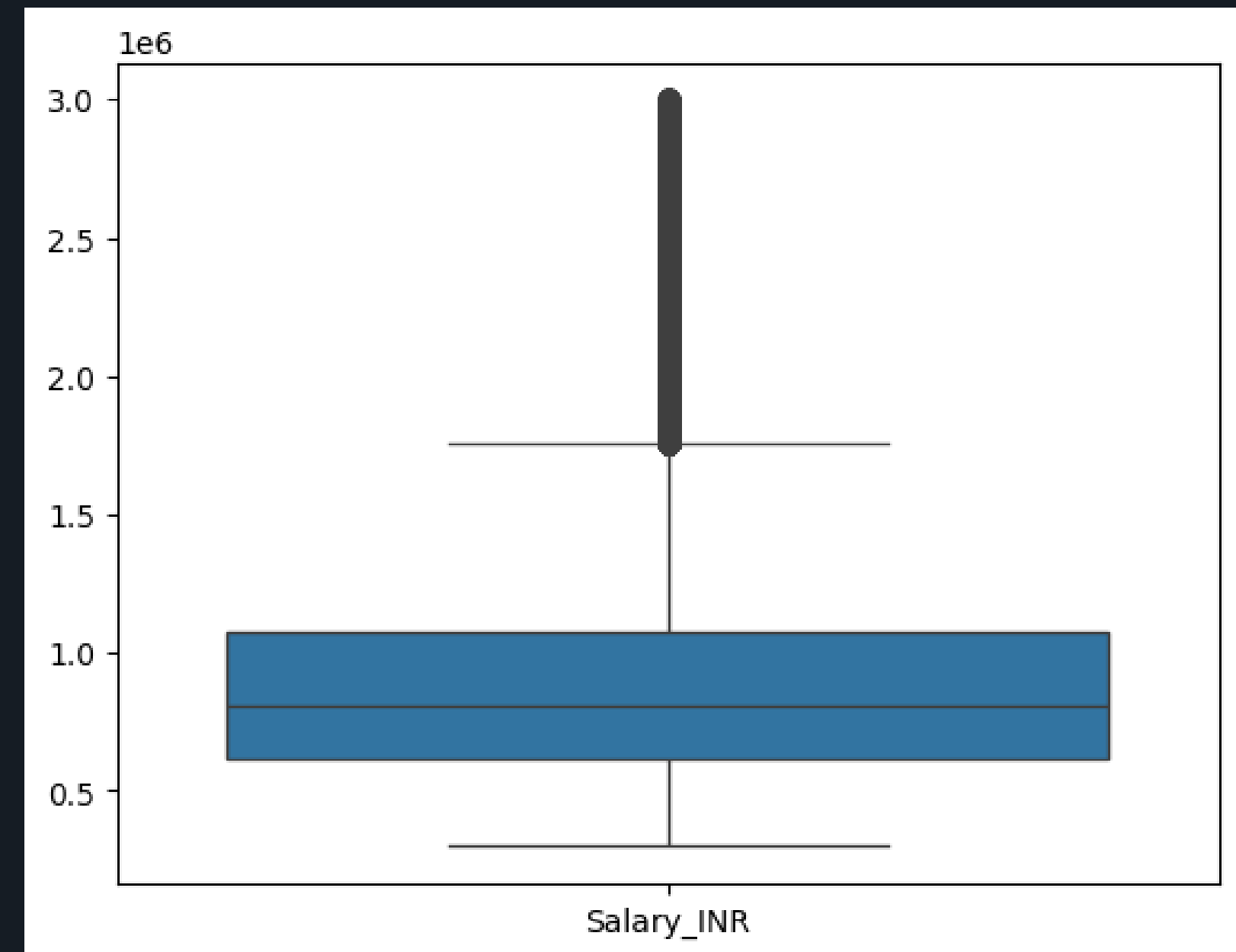
The range of employee salaries in the dataset.

### Why I used this visualization:

A box plot shows the salary spread, highlights outliers, and provides insights into pay inequality.

### Key takeaway:

There is a wide variation in salaries, with most employees earning in a middle range but a few individuals receiving very high salaries.





## 8. Salary vs. Experience

### What the visualization shows:

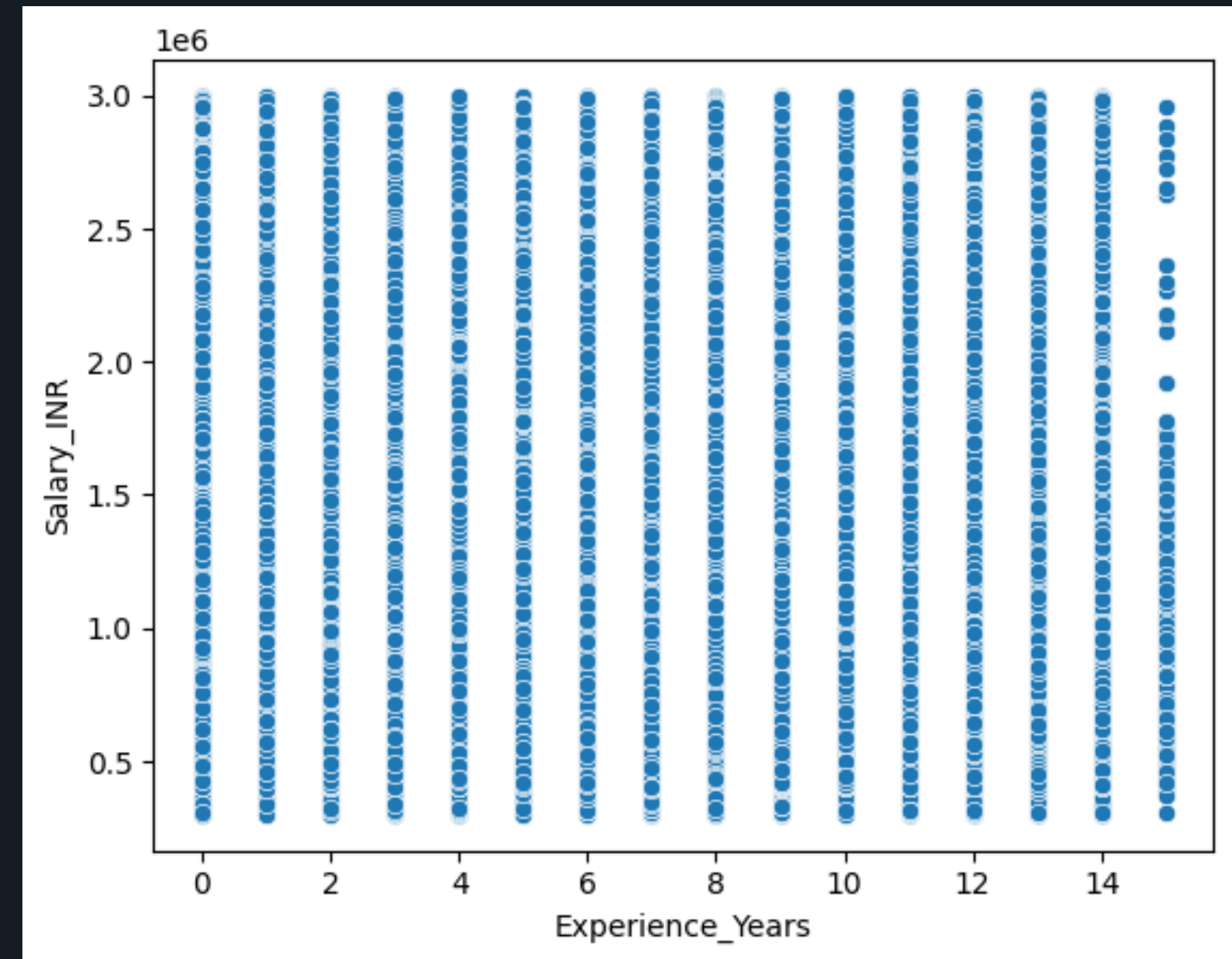
The relationship between years of experience and salary, with a regression line.

### Why I used this visualization:

A scatter plot with regression helps demonstrate trends and shows if experience has an impact on salary growth.

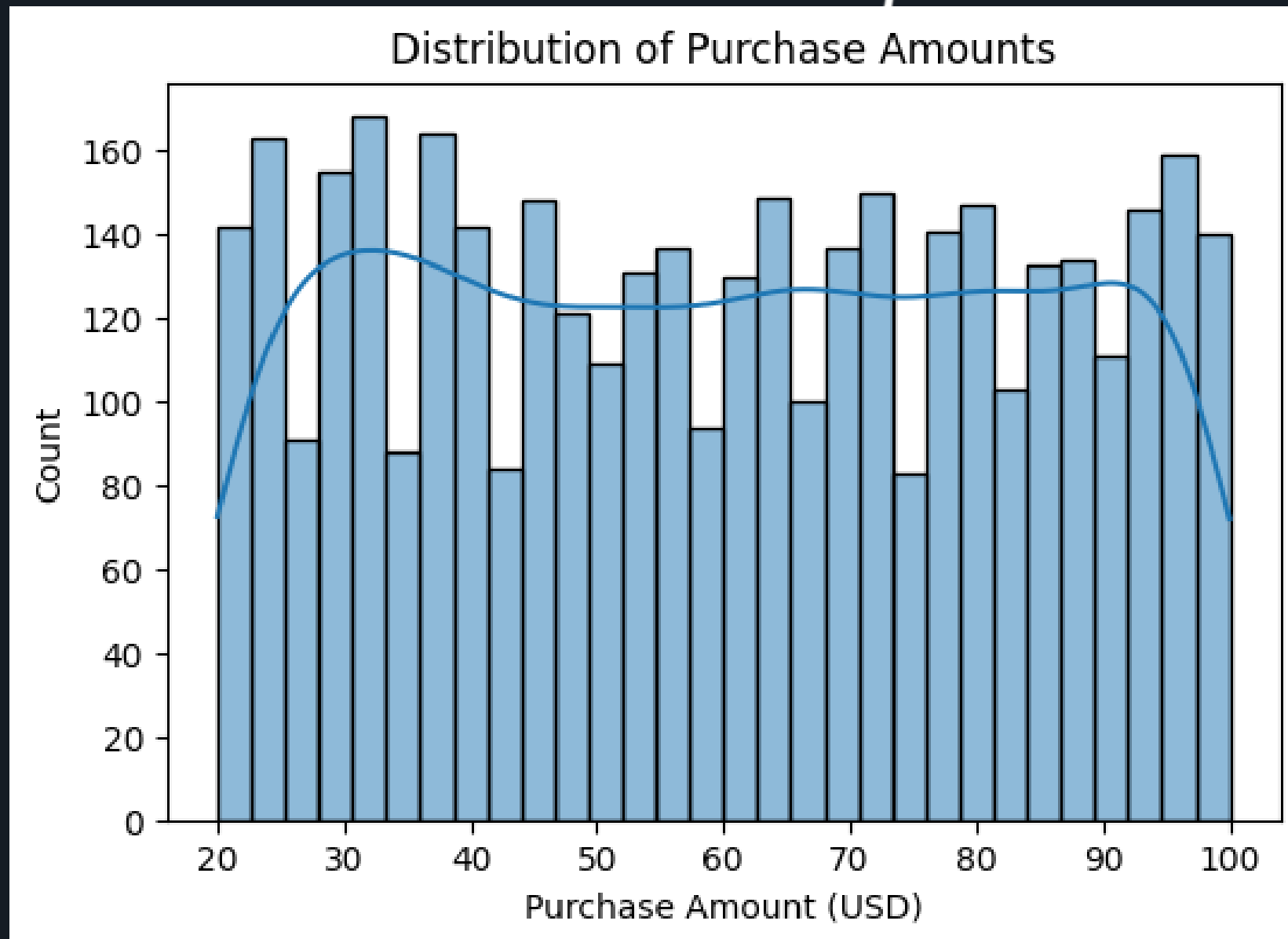
### Key takeaway:

There is a clear positive relationship: employees with more years of experience generally earn higher salaries.



# 2.Shopping Trends





## 1. Distribution of Purchase Amounts

### What the visualization shows:

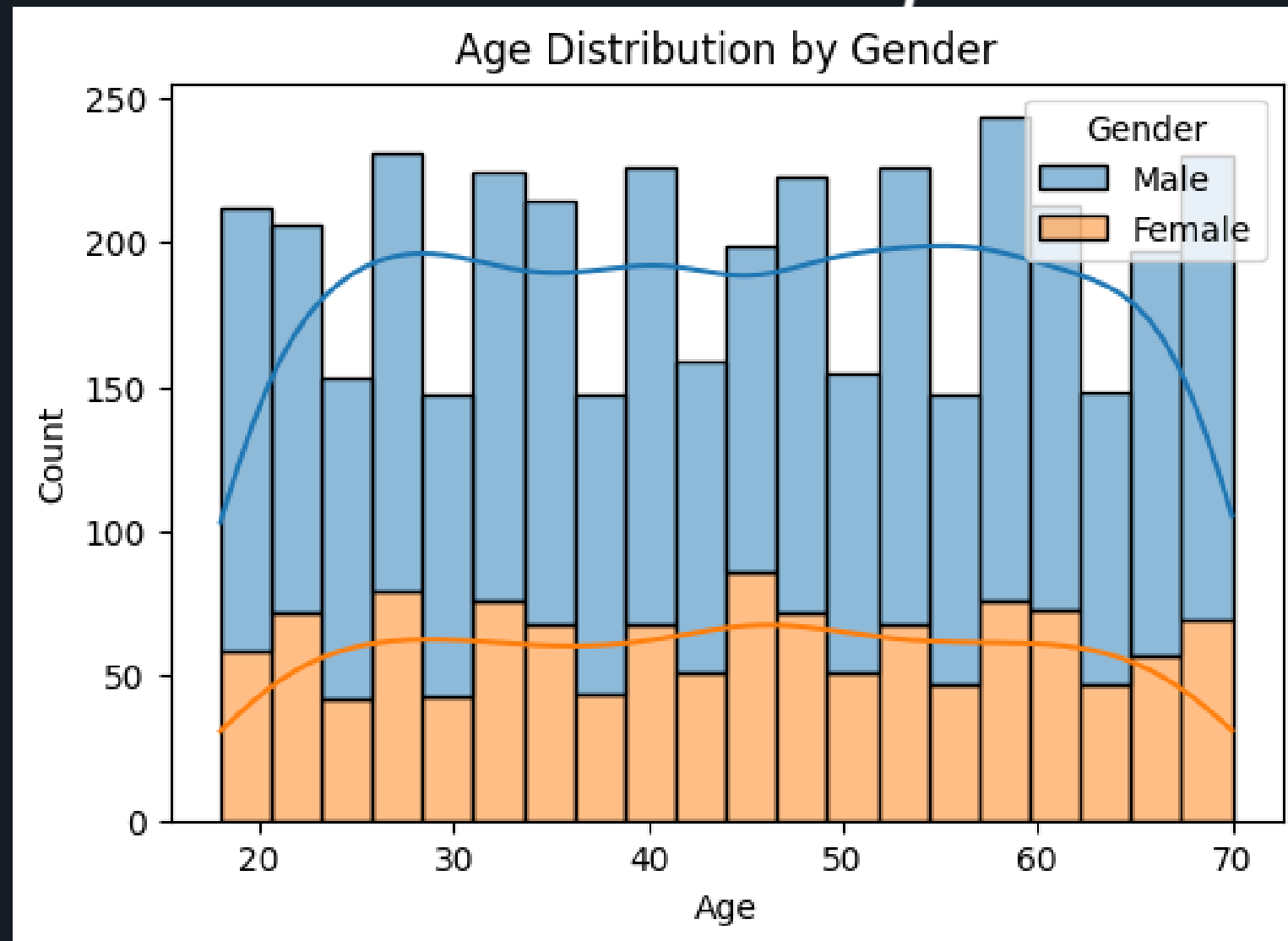
The spread of customer purchase amounts in USD.

### Why I used this visualization:

A histogram helps visualize the overall spending behavior and detect if most customers buy in similar ranges.

### Key takeaway:

Most purchases fall within a mid-range amount, while a few customers spend significantly higher, creating outliers.



## 2. Age Distribution by Gender

### What the visualization shows:

The distribution of customer ages, separated by gender.

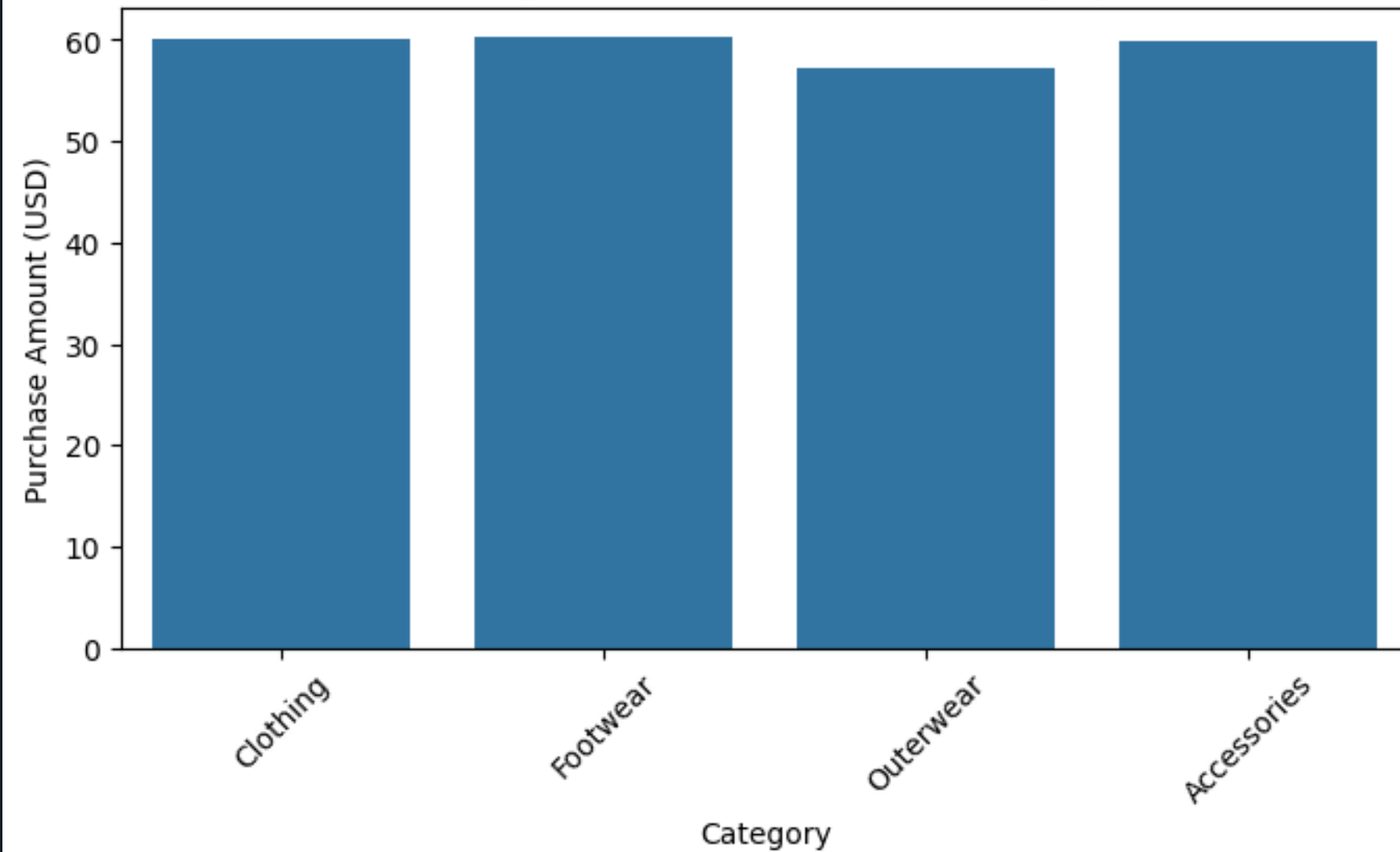
### Why I used this visualization:

Stacked histograms allow comparison of male vs. female customers across different age groups.

### Key takeaway:

Both genders are represented across all ages, with peaks in younger age groups, showing a strong youth customer base.

Average Purchase Amount per Category



### 3. Average Purchase Amount per Category

#### What the visualization shows:

The mean purchase value for each product category.

#### Why I used this visualization:

A bar chart is effective to compare spending across categories.

#### Key takeaway:

Some categories (e.g., Electronics, Fashion) have higher purchase amounts, while others remain lower.

#### 4. Purchase Amount by Payment Method

##### What the visualization shows:

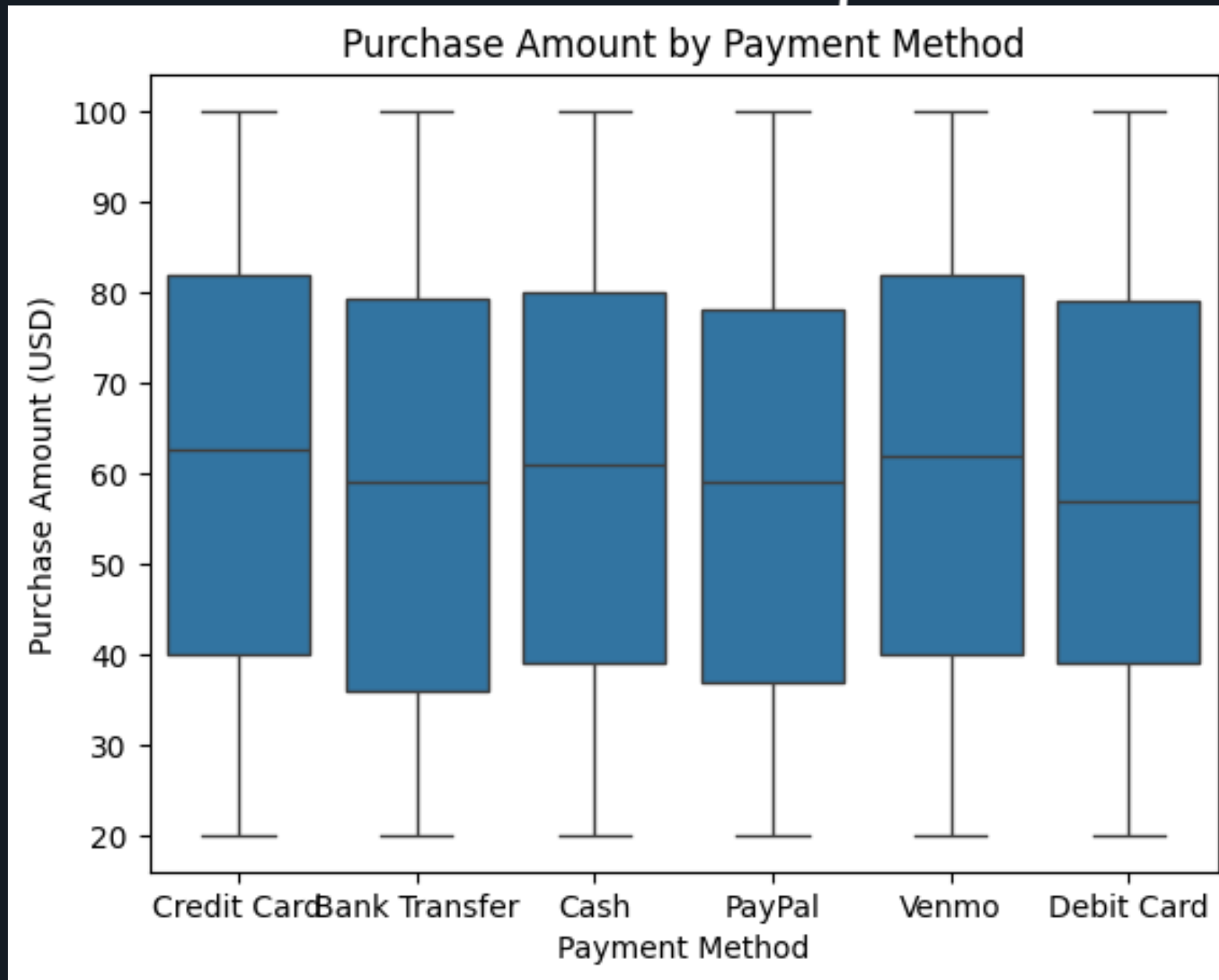
Variation in purchase values depending on payment method.

##### Why I used this visualization:

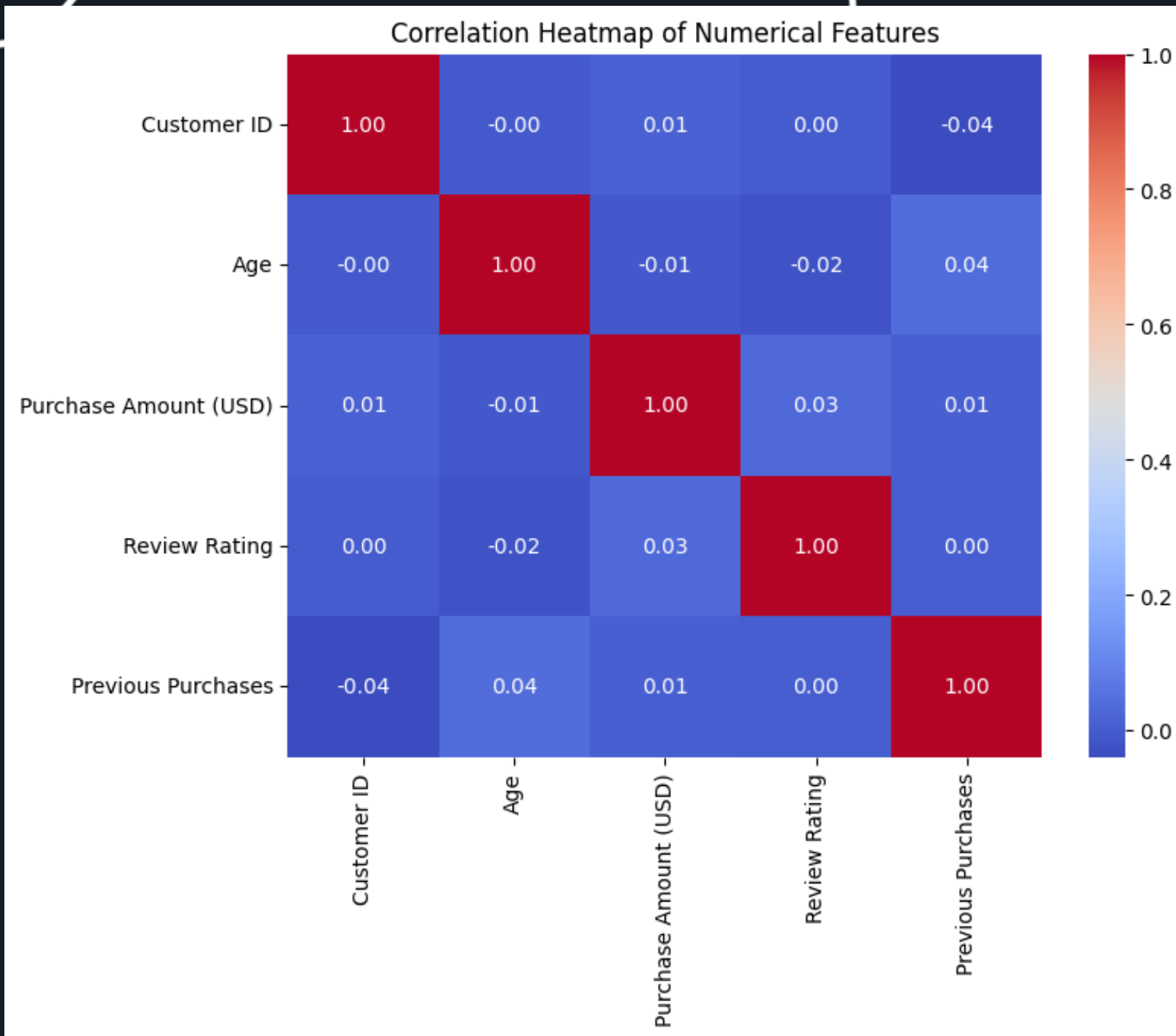
A box plot highlights how different payment types affect spending and reveals outliers.

##### Key takeaway:

Certain payment methods are linked with higher spending, while others are used mainly for smaller transactions.







## 5. Correlation Heatmap of Numerical Features

### What the visualization shows:

Relationships between numerical fields such as Age, Purchase Amount, Review Rating, etc.

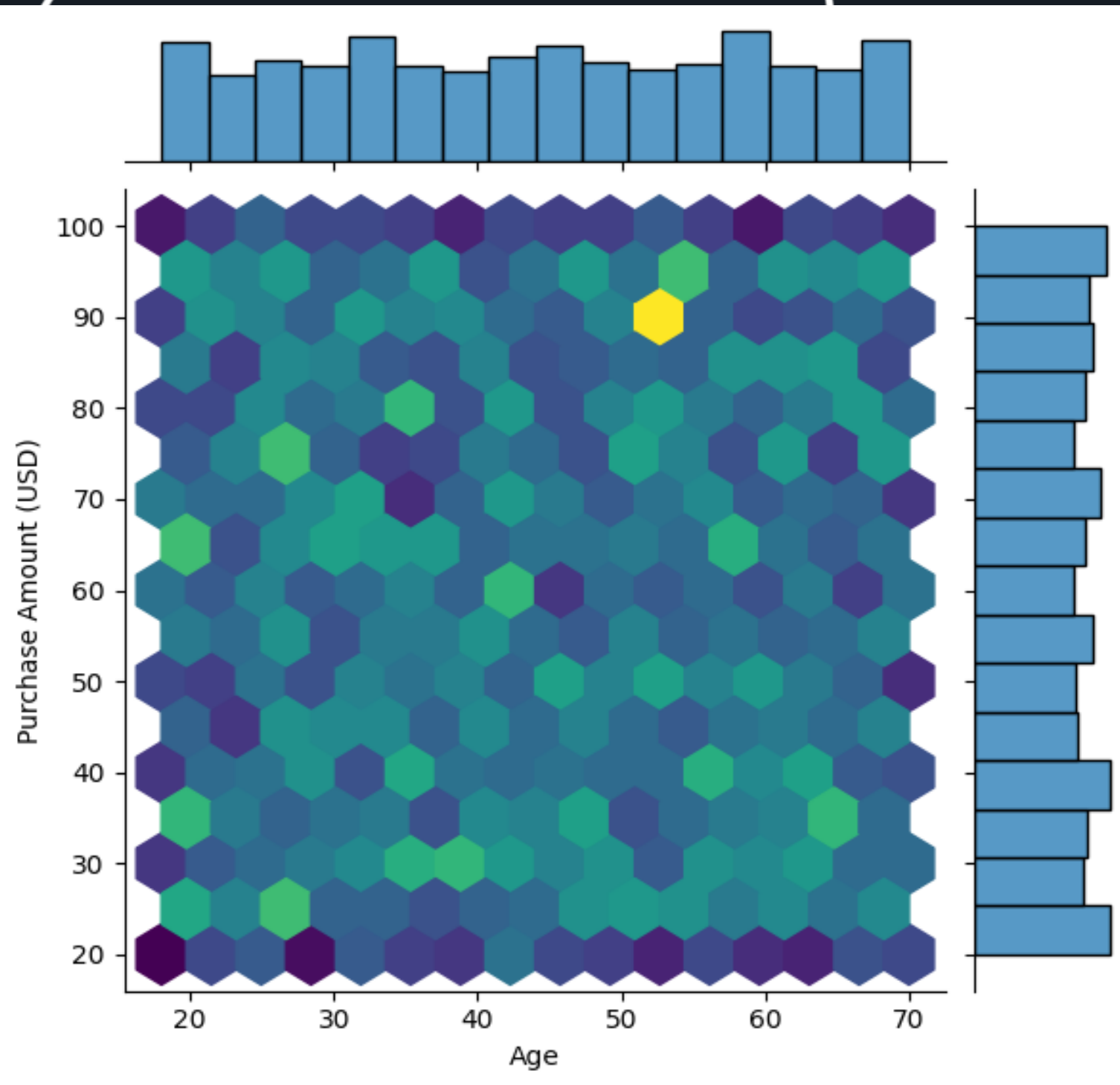
### Why I used this visualization:

A heatmap makes it easy to see which variables are related.

### Key takeaway:

Purchase amount has weak-to-moderate relationships with other features, while some variables (like Age and Rating) show minimal correlation.





## 6. Age vs Purchase Amount

### What the visualization shows:

A scatter plot with side histograms that shows how customer age relates to purchase amount.

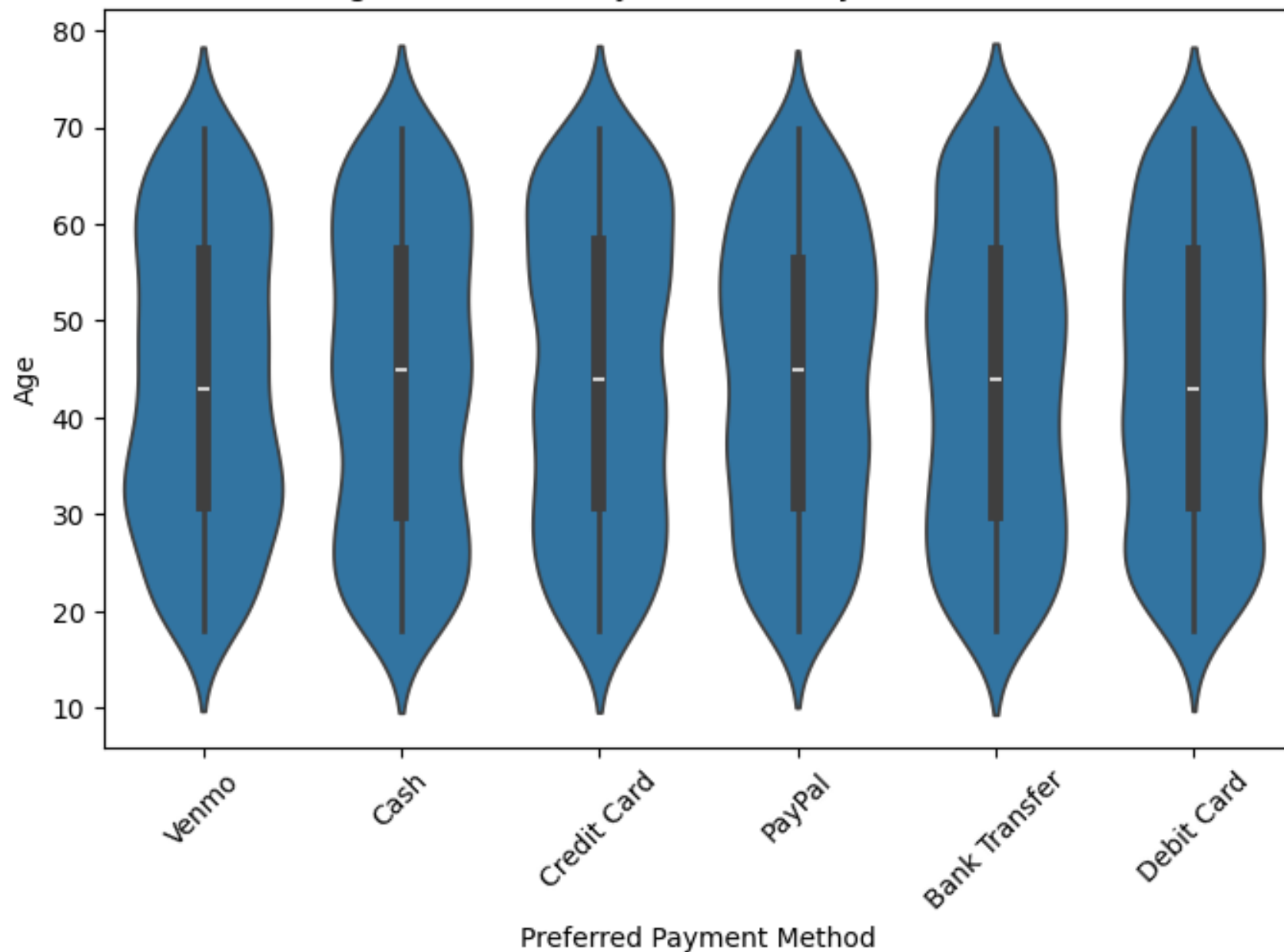
### Why I used this visualization:

Combines two views, overall distributions and the relationship between them.

### Key takeaway:

Middle-aged customers tend to spend more consistently, while younger groups show more variation.

Age Distribution by Preferred Payment Method



## 7. Age by Purchase Method

### What the visualization shows:

Each violin shows the distribution of customer ages for a specific purchase method (e.g., Credit Card, Cash, Online).

### Why I used this visualization:

Highlights which age groups prefer which payment types, and how spread out or concentrated they are.

### Key takeaway:

Younger customers may favor modern payment methods (like online or card), while older customers cluster around traditional methods.

# 3.Supermarket





## 1. Sales Distribution

### What the visualization shows:

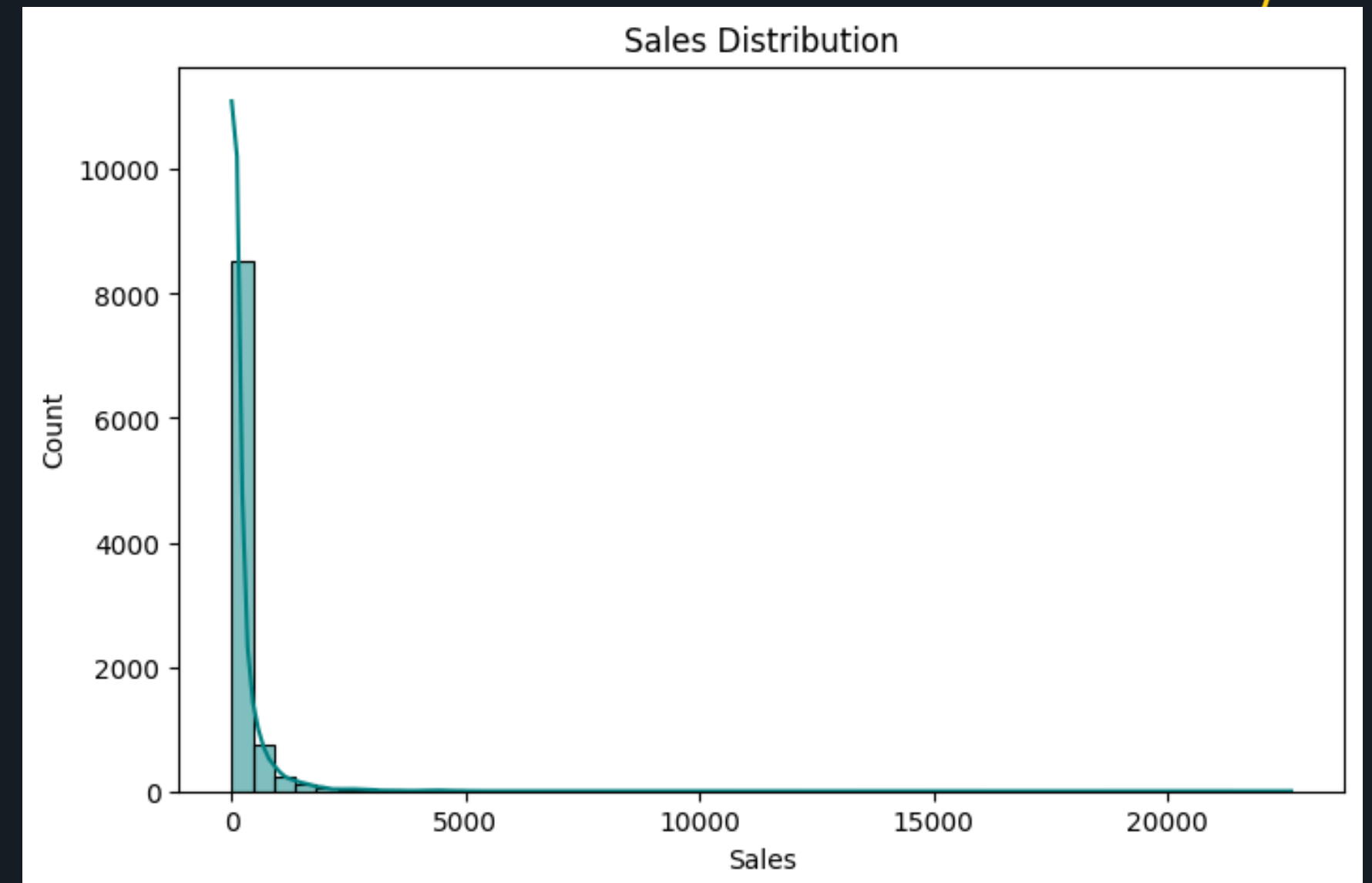
The spread of sales amounts across all transactions.

### Why I used this visualization:

A histogram is useful for spotting common sales values and detecting skewed distributions.

### Key takeaway:

Most sales transactions cluster in a middle range, with fewer very high or very low sales.





## 2. Orders Count by Region

### What the visualization shows:

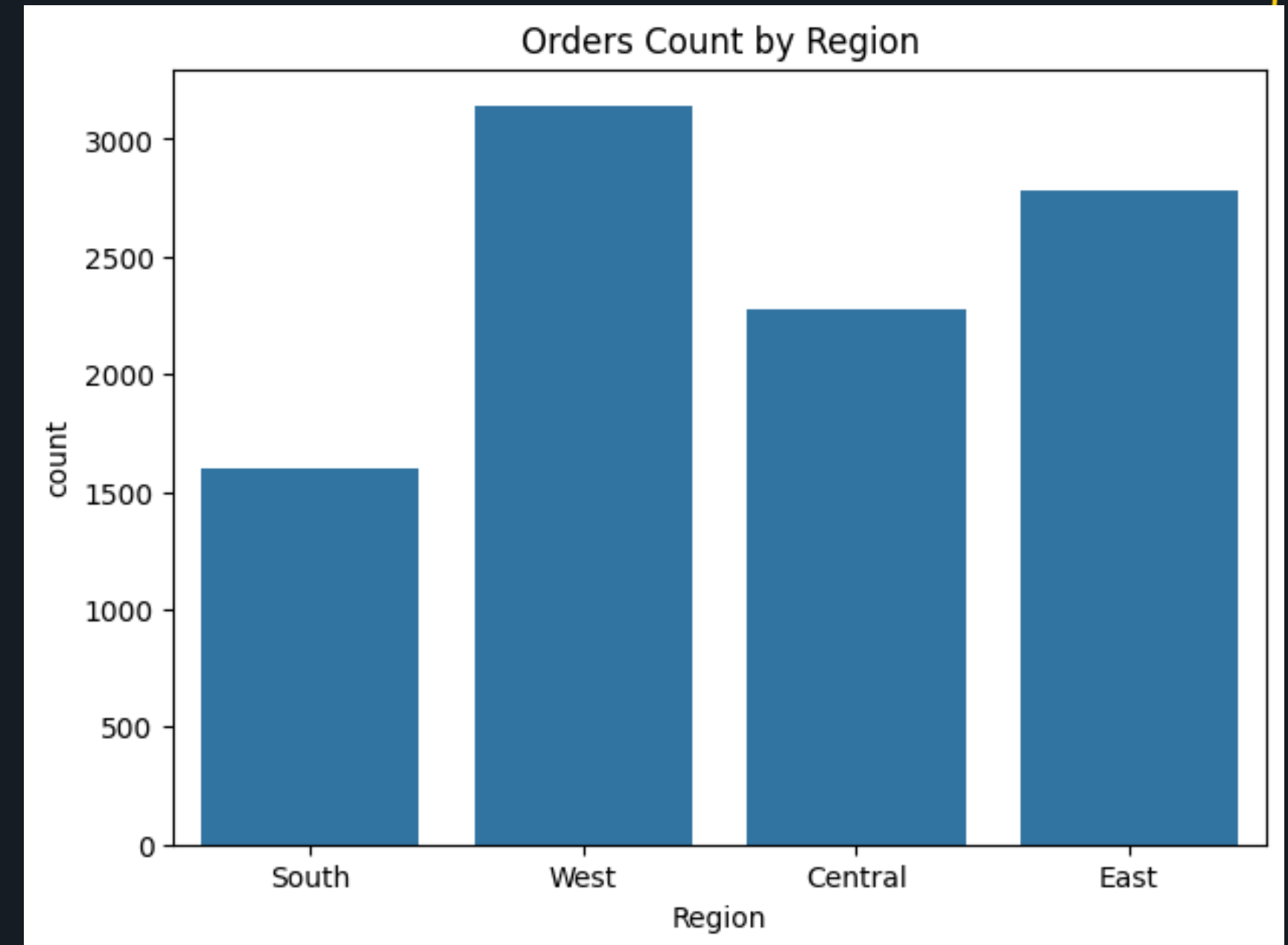
The number of orders placed in each region.

### Why I used this visualization:

A count plot helps identify the busiest regions in terms of customer orders.

### Key takeaway:

Certain regions dominate order counts, highlighting strong customer demand there.





### 3. Total Sales by Region

**What the visualization shows:**

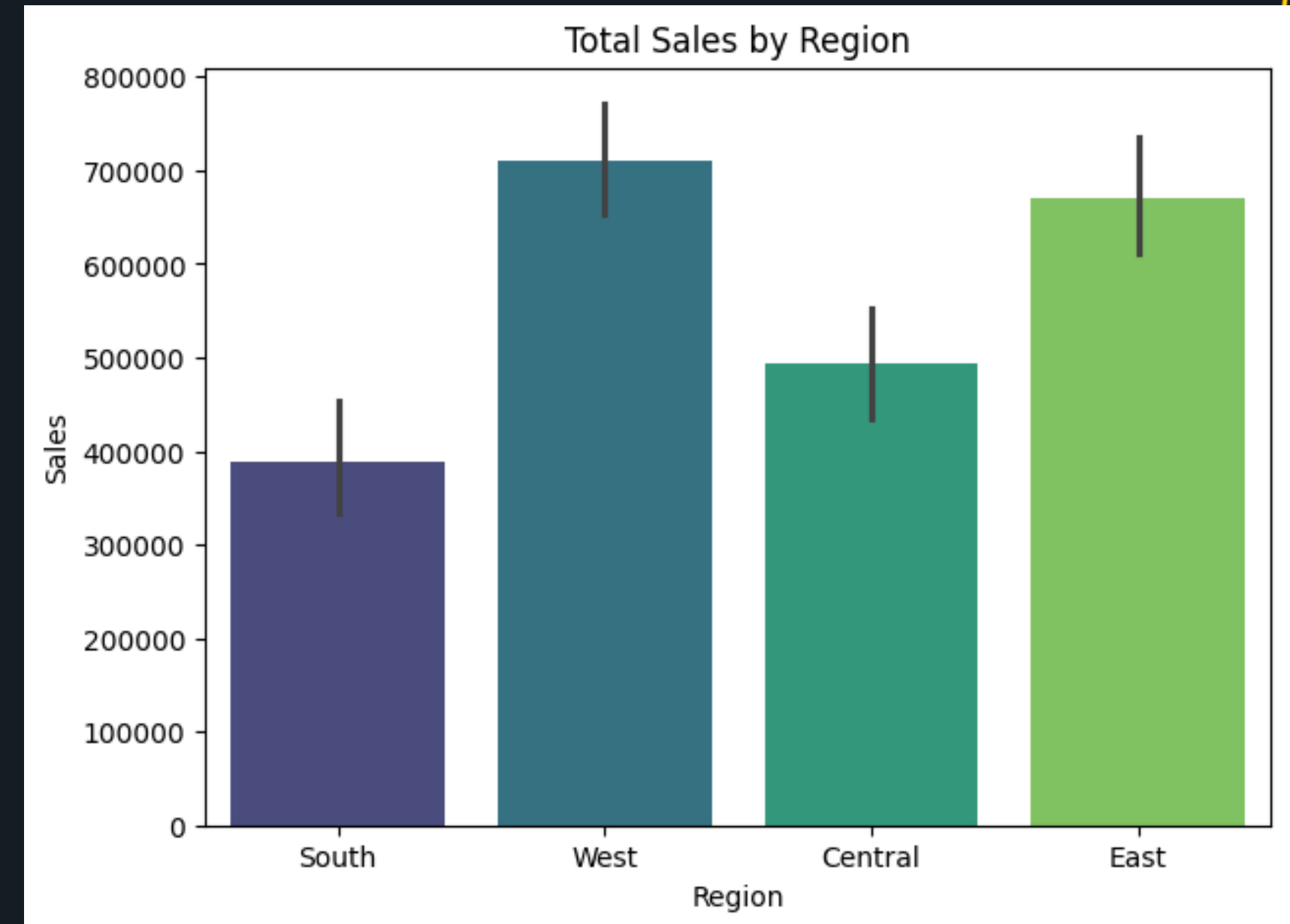
Total revenue generated from each region.

**Why I used this visualization:**

A bar chart allows comparison of how much revenue different regions contribute.

**Key takeaway:**

Some regions not only have more orders but also generate higher total sales, making them top-performing markets.





## 4. Total Sales by Category

### What the visualization shows:

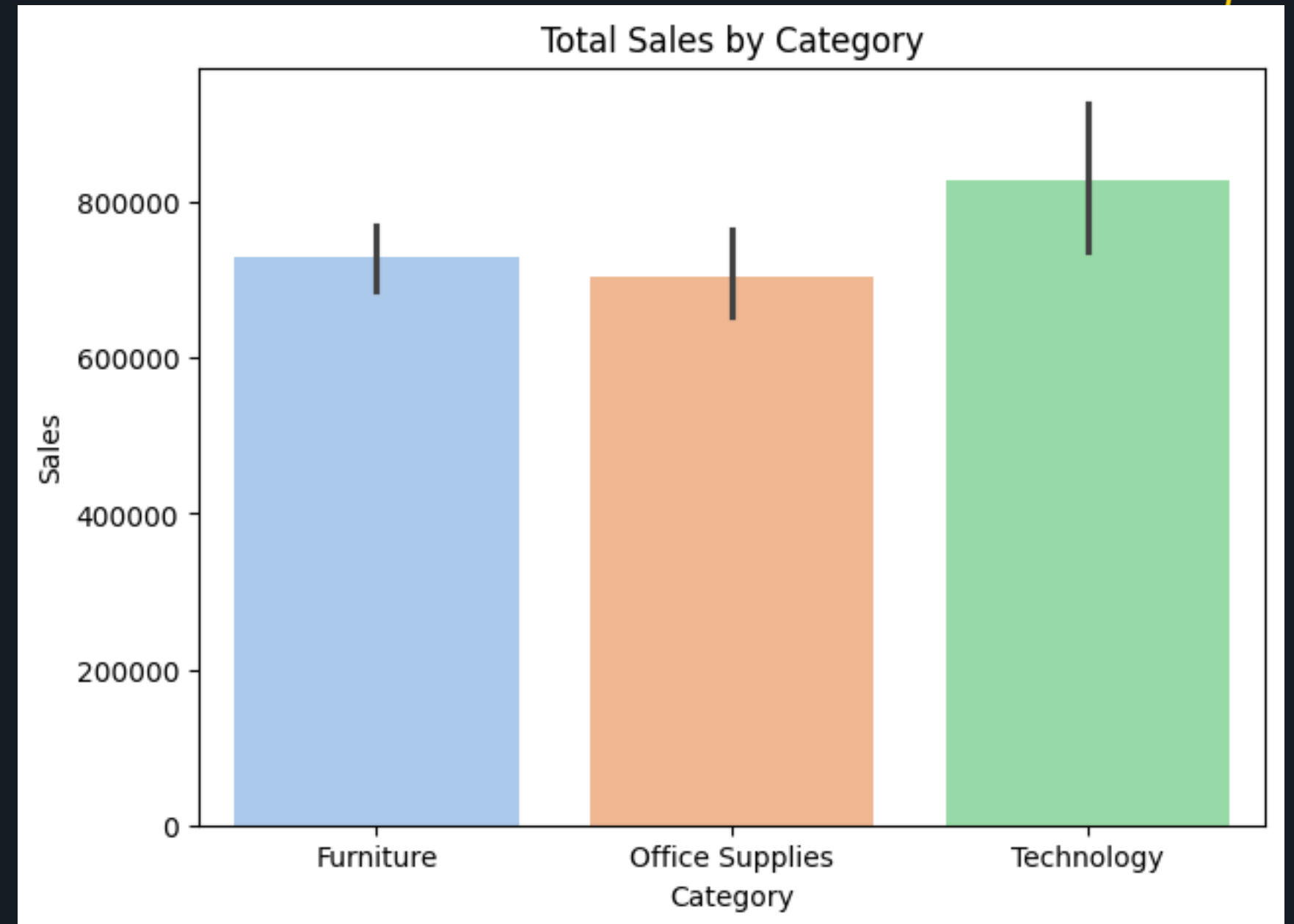
The overall sales split across product categories.

### Why I used this visualization:

Category analysis highlights which products bring the most revenue.

### Key takeaway:

A few categories dominate total sales, suggesting where the business should focus.





## 5. Sales Trend Over Time

### What the visualization shows:

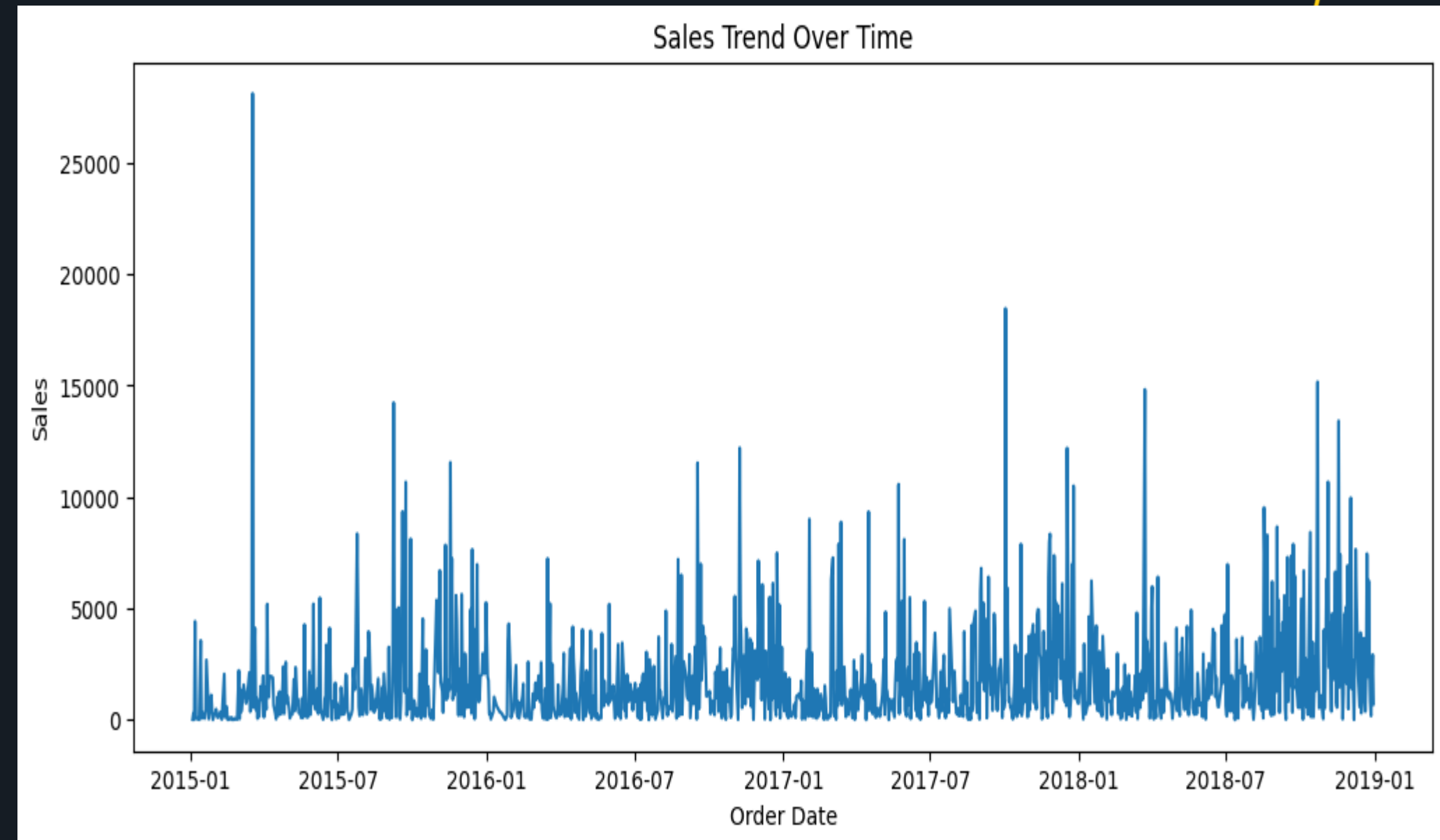
The movement of total sales over time.

### Why I used this visualization:

A line plot shows seasonal peaks, growth patterns, or drops in sales.

### Key takeaway:

Sales increase steadily with clear peaks in certain months, suggesting strong seasonal demand.







## 6. Sales vs Order Size

### What the visualization shows:

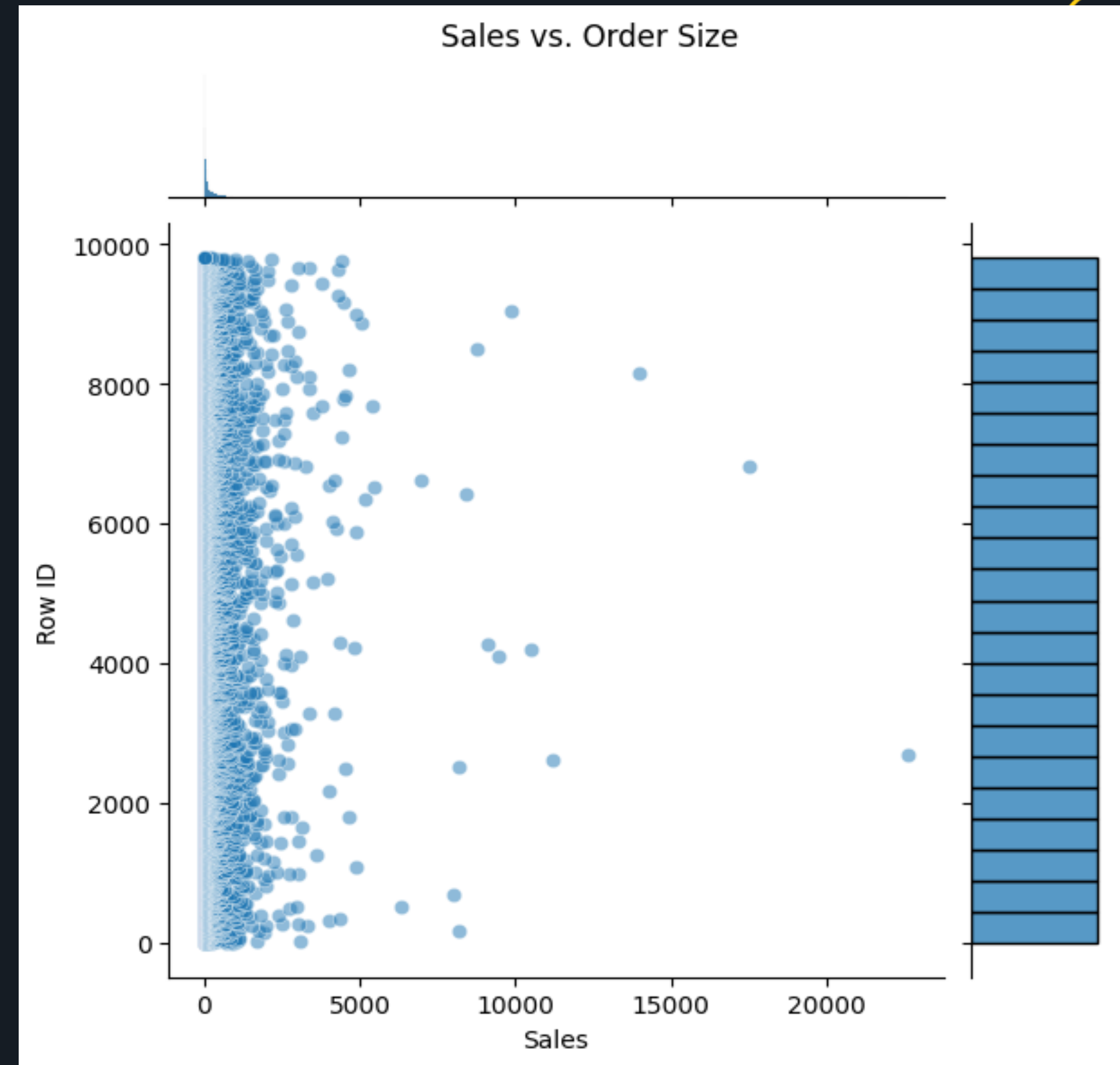
The relationship between the size of an order (number of items) and the total sales value.

### Why I used this visualization:

To check if larger orders always lead to higher revenue, or if small orders can also contribute significantly.

### Key takeaway:

Larger order sizes generally bring more sales, but there are some small orders with high sales due to expensive items.





## 7. Top 10 Sub-Categories by Sales

### What the visualization shows:

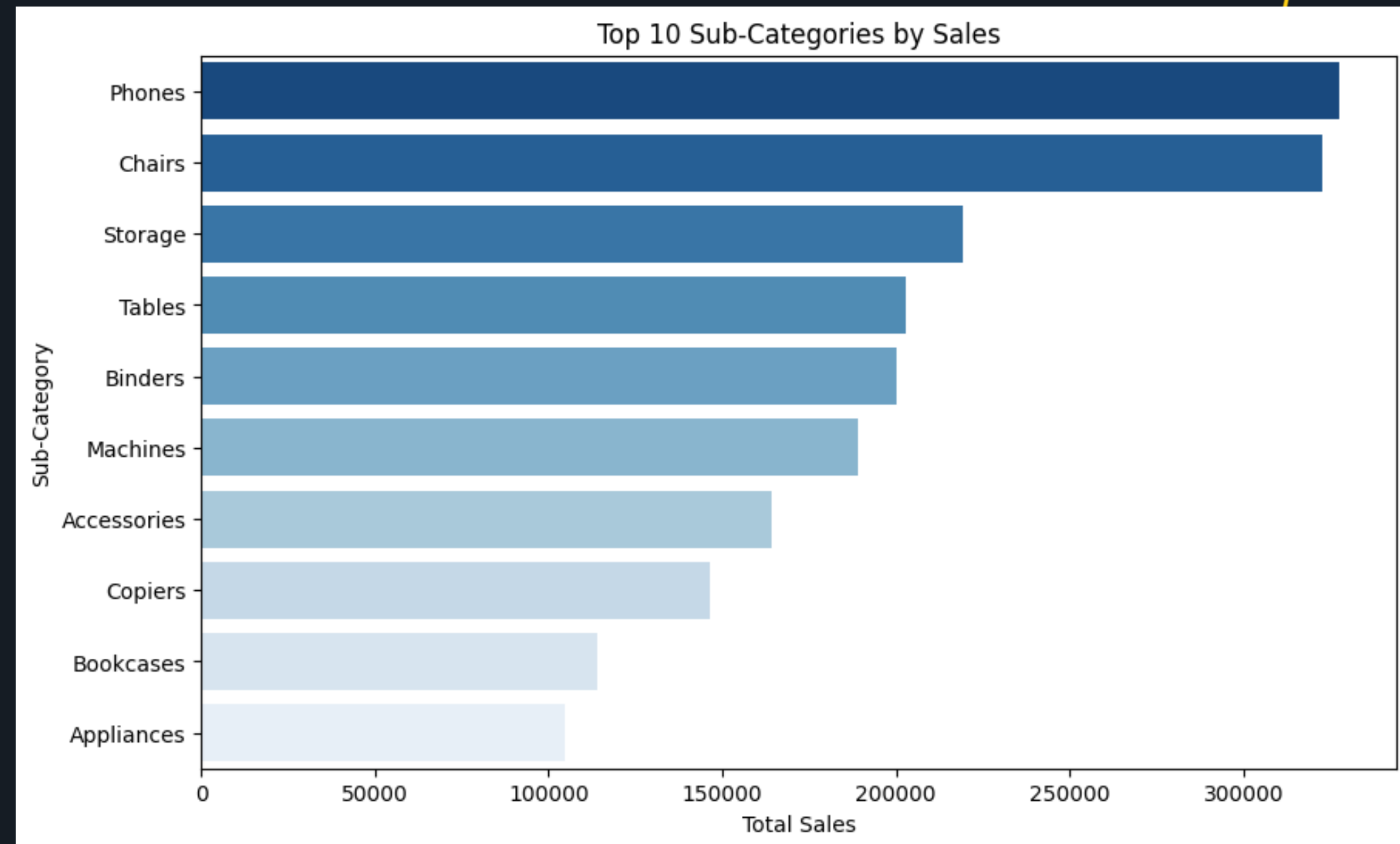
Total or average sales broken down by grouped categories, such as **Customer Segment** or **Product Category**.

### Why I used this visualization:

Grouped bar charts make it easier to compare how different segments contribute to overall sales.

### Key takeaway:

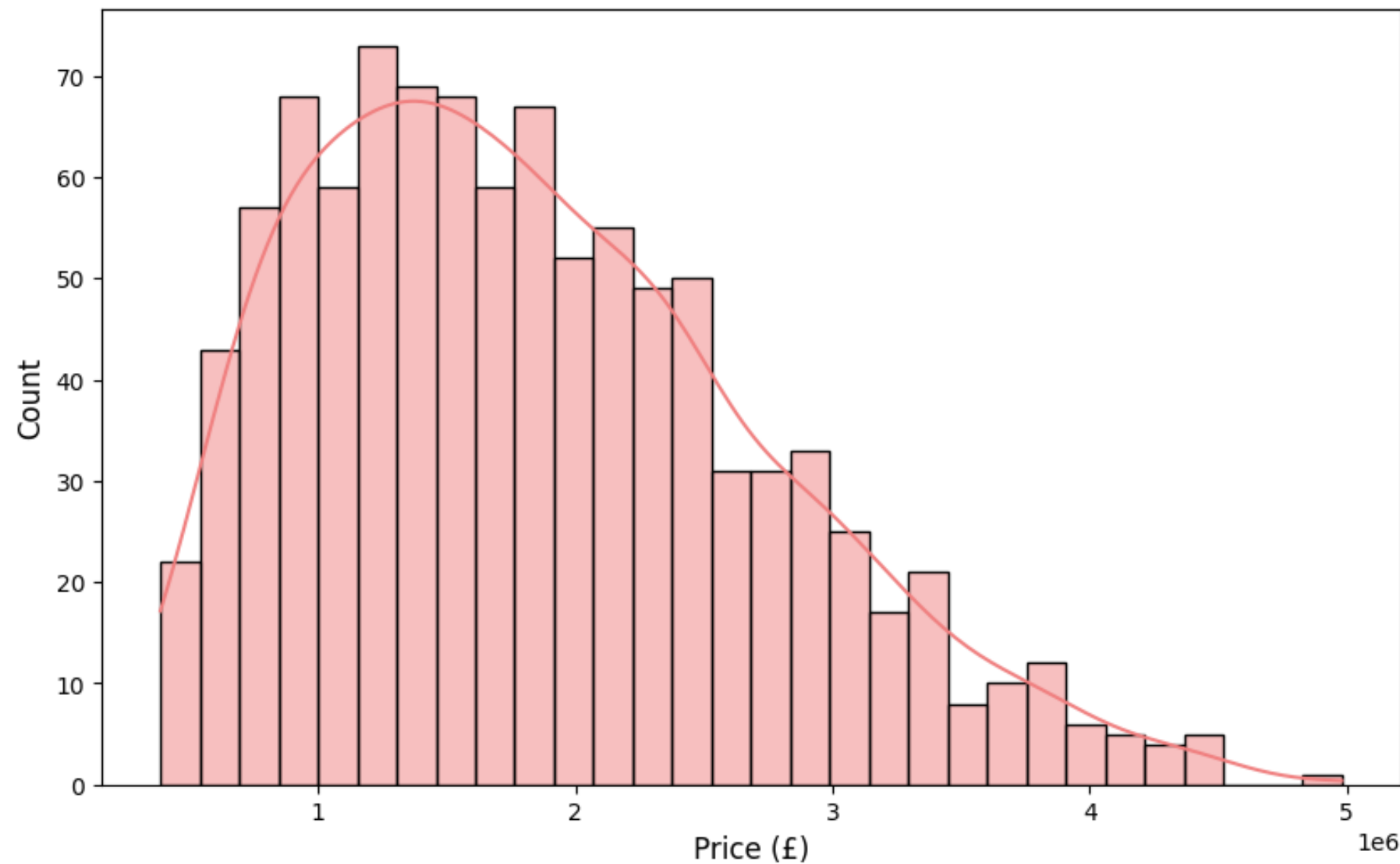
Certain groups (e.g., Corporate customers or Technology products) stand out as stronger revenue drivers compared to others.



# 4.London Houses



Distribution of House Prices



## 1. Distribution of House Prices

### What the visualization shows:

How house prices are distributed across different ranges.

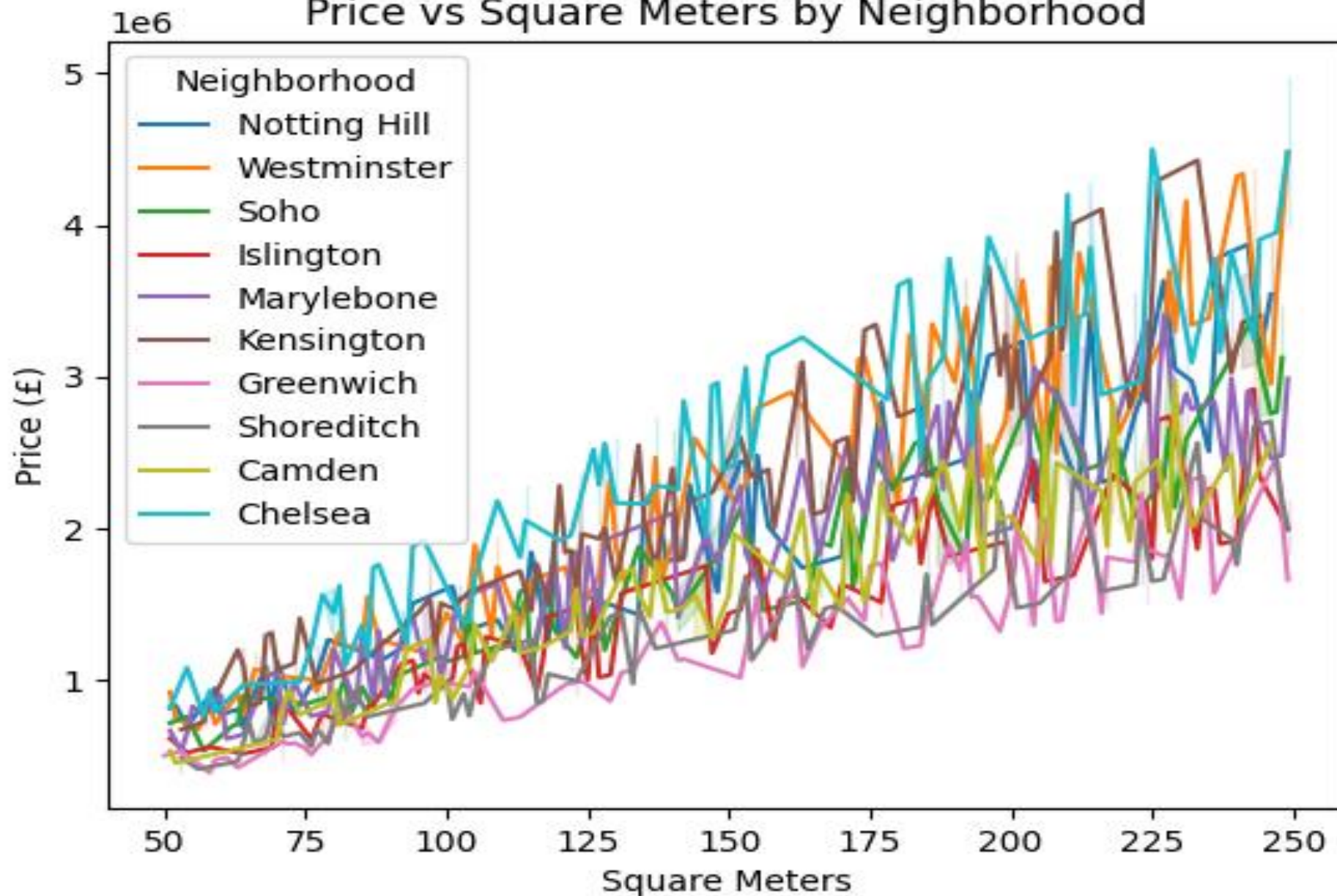
### Why I used this visualization:

A histogram with KDE reveals whether prices are concentrated in certain ranges or widely spread.

### Key takeaway:

Most houses fall into middle price ranges, with fewer very high-priced properties.

Price vs Square Meters by Neighborhood



## 2. Price vs Square Meters by Neighborhood

### What the visualization shows:

The relationship between property size and price, separated by neighborhood.

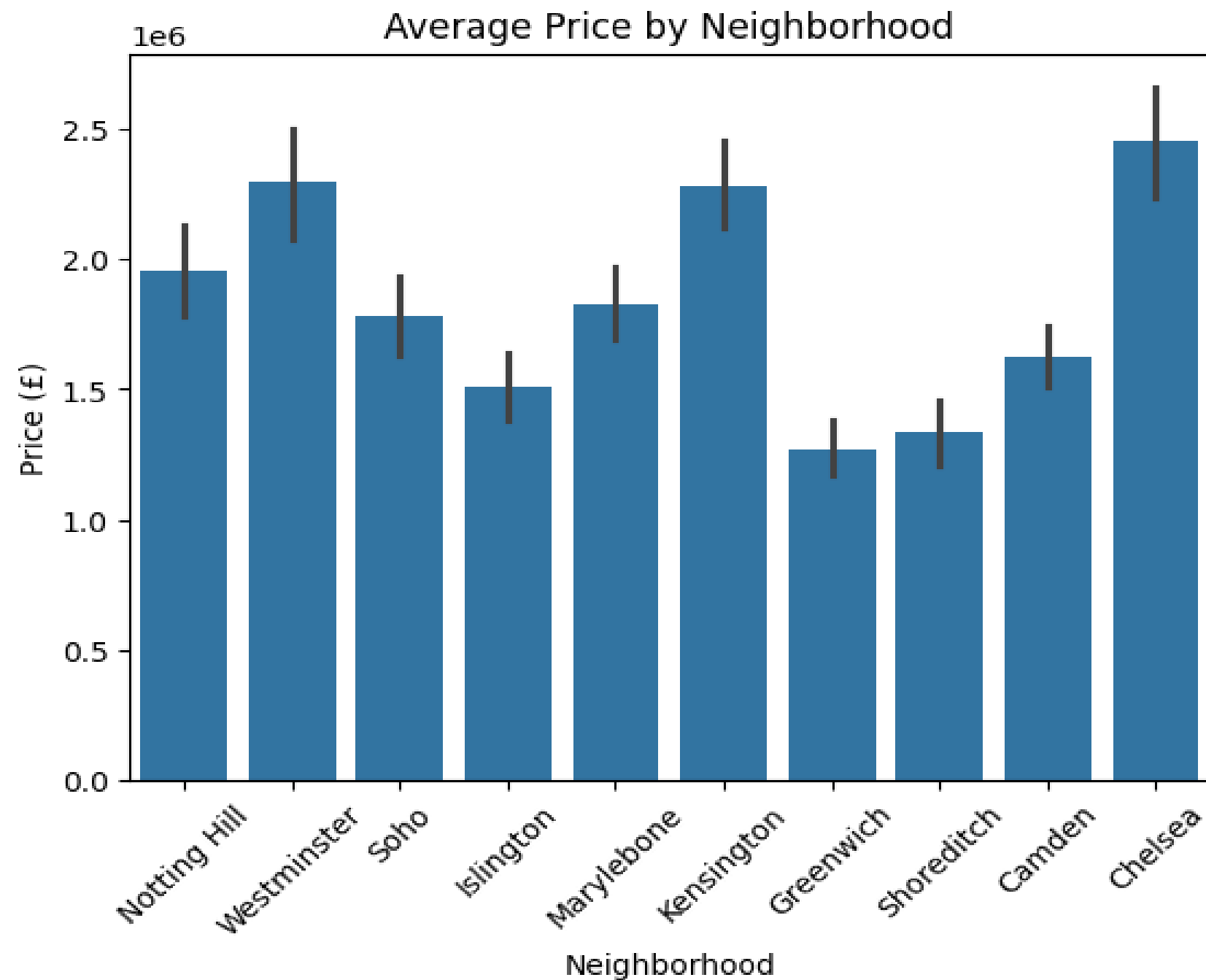
### Why I used this visualization:

To see how price grows with size and whether the trend differs across neighborhoods.

### Key takeaway:

Larger properties generally cost more, but some neighborhoods are more expensive regardless of size.





### 3. Average Price by Neighborhood

#### What the visualization shows:

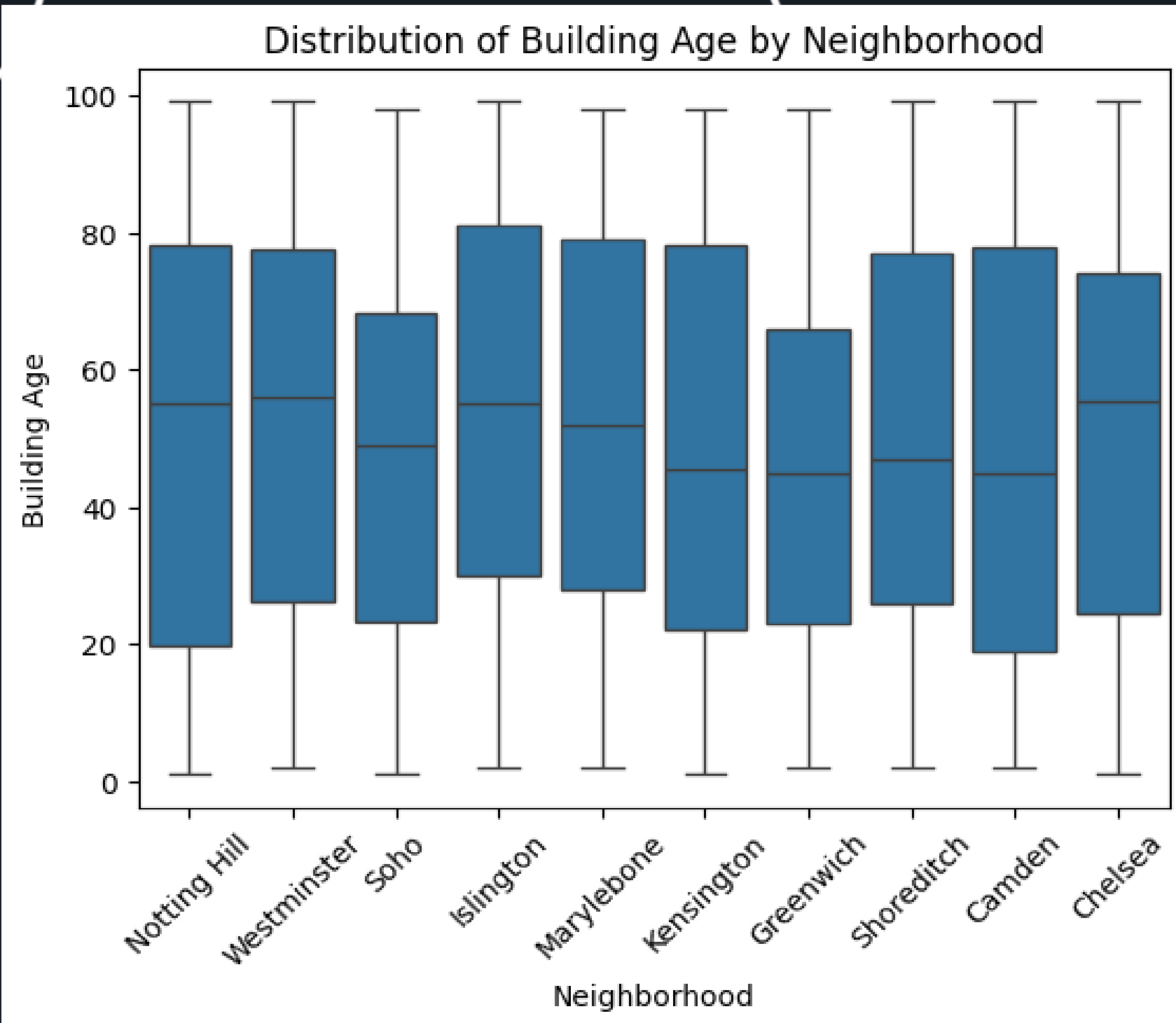
The mean house price in each neighborhood.

#### Why I used this visualization:

Averages give a quick comparison of neighborhood affordability.

#### Key takeaway:

Some neighborhoods consistently have higher prices, standing out as more premium locations.



#### 4. Distribution of Building Age by Neighborhood

##### What the visualization shows:

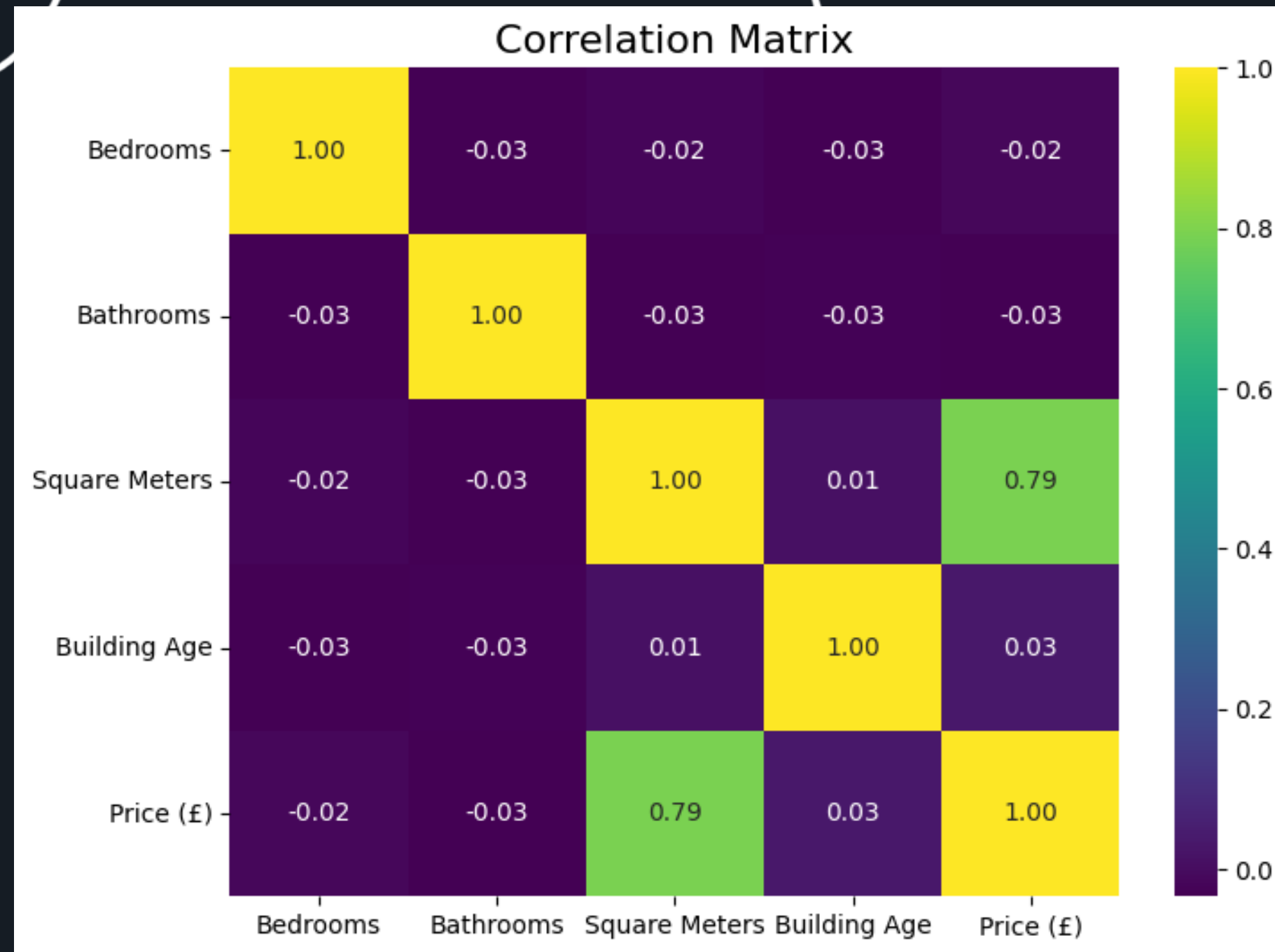
How old the buildings are in different neighborhoods.

##### Why I used this visualization:

A box plot shows medians, ranges, and outliers, highlighting age variation.

##### Key takeaway:

Certain neighborhoods have much older housing stock, while others feature newer developments.



## 5. Correlation Matrix

### What the visualization shows:

The strength of relationships between numeric variables like bedrooms, size, age, and price.

### Why I used this visualization:

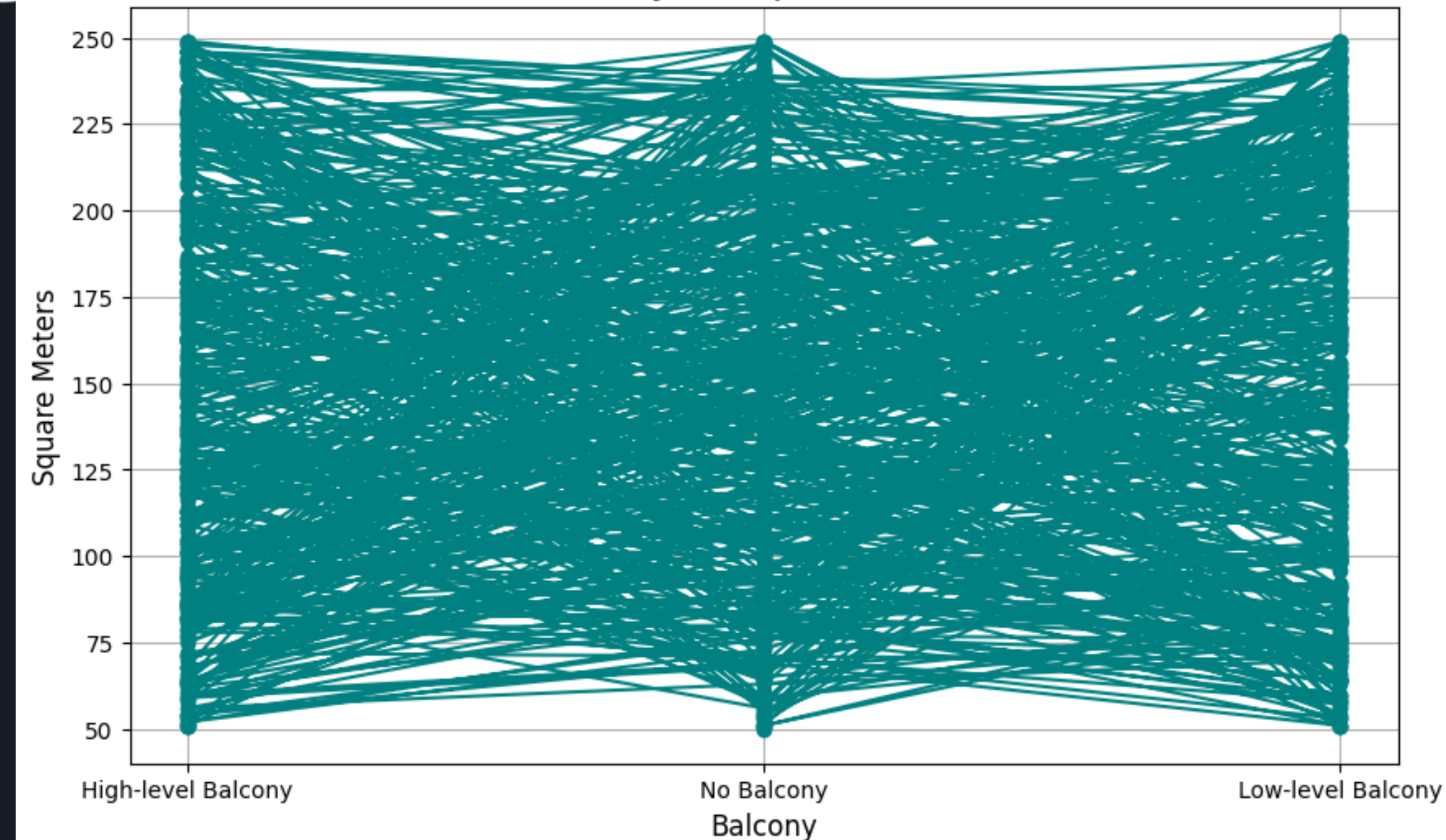
Helps identify which features strongly influence price.

### Key takeaway:

Square meters and number of bedrooms are highly correlated with price, while building age has a weaker effect.



Balcony vs Square Meters



## 6. Balcony vs Square Meters

### What the visualization shows:

The relationship between balcony size/availability and property size.

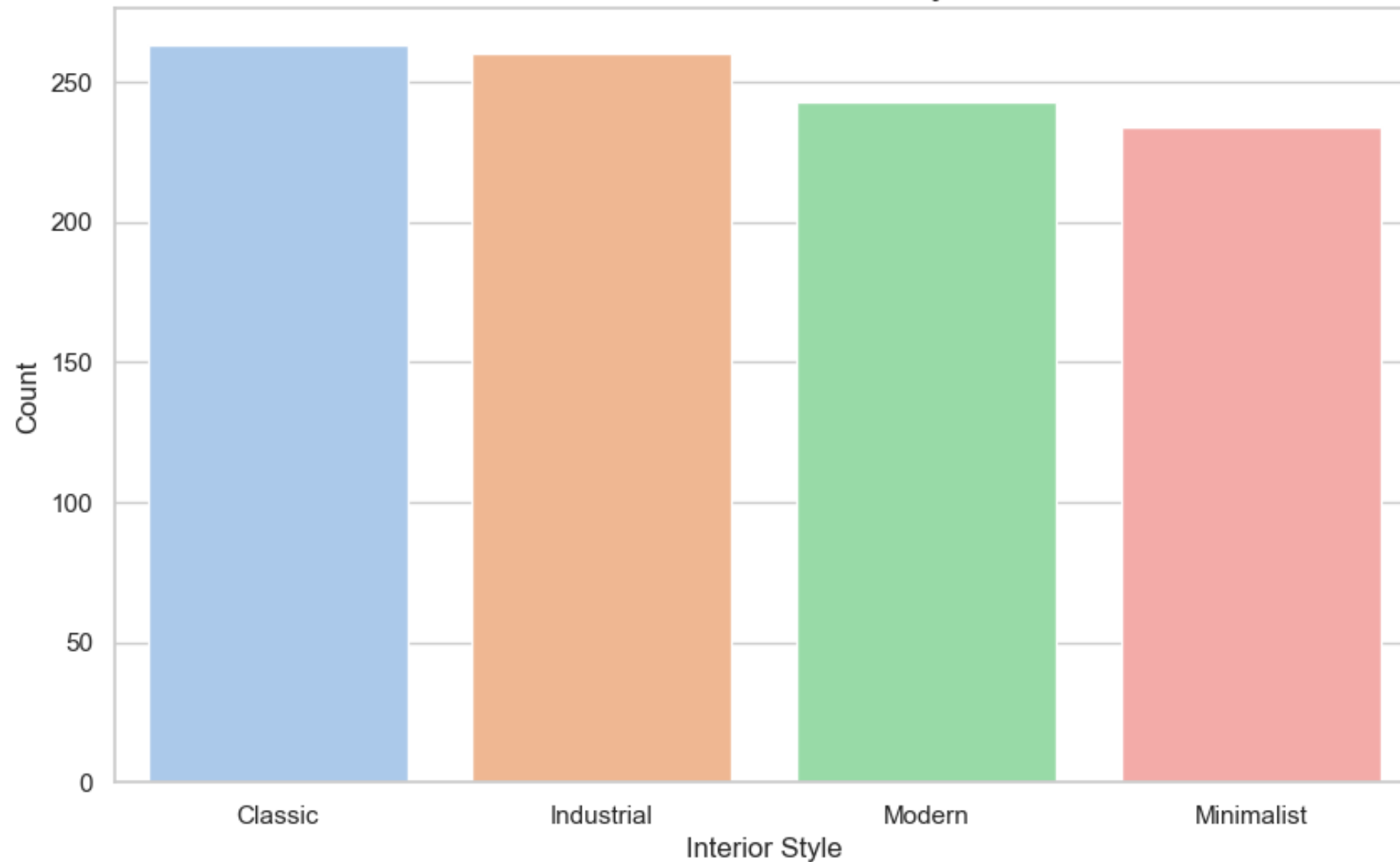
### Why I used this visualization:

To see whether bigger houses are more likely to include balconies.

### Key takeaway:

Larger properties usually include balconies, but smaller ones rarely do.

Distribution of Interior Styles



## 7. Distribution of Interior Styles

### What the visualization shows:

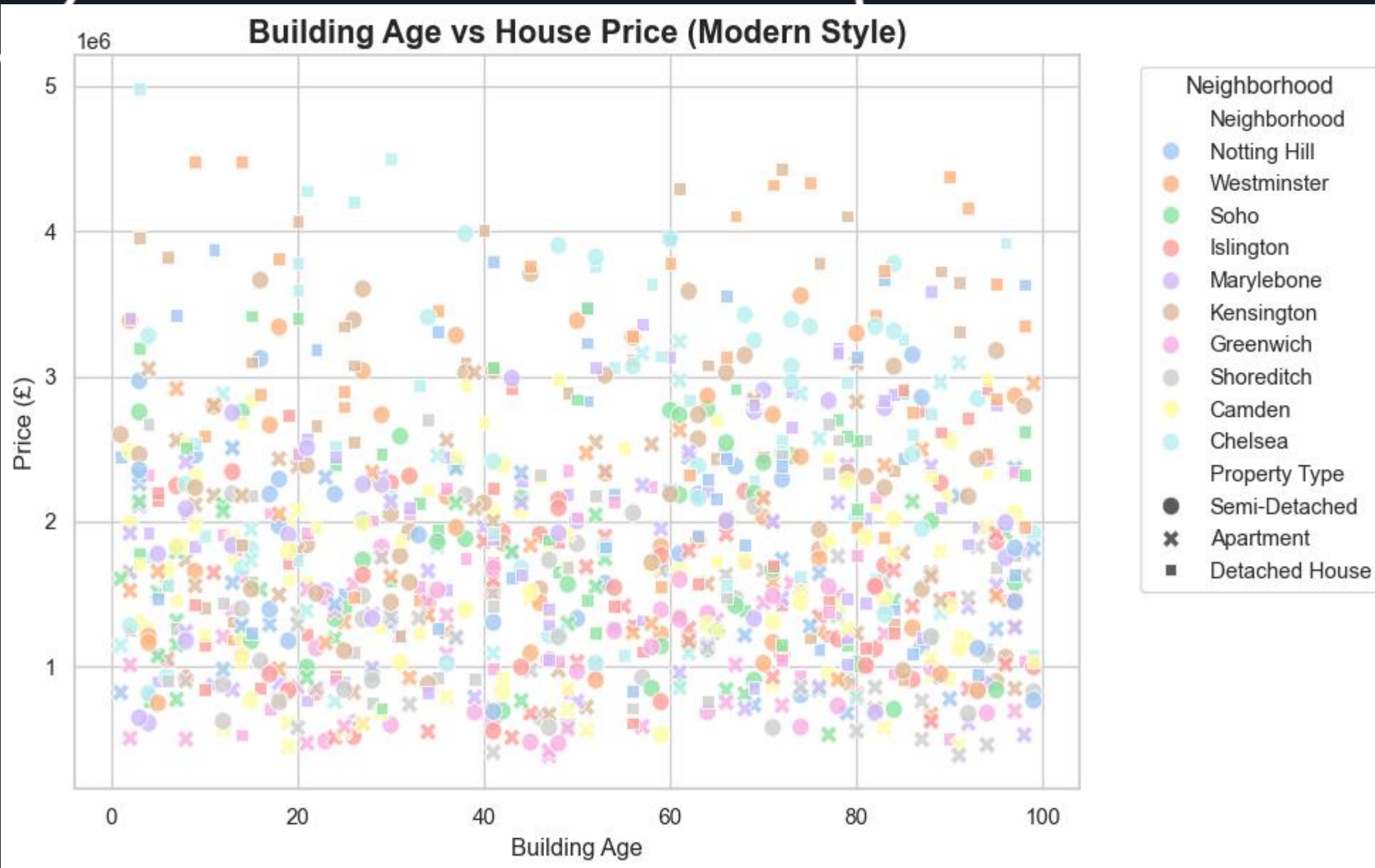
How many properties fall under each interior style (e.g., Modern, Classic).

### Why I used this visualization:

Count plots are great for comparing categorical popularity.

### Key takeaway:

Modern styles dominate the dataset, while other styles are less common.



## 8. Building Age vs House Price

### What the visualization shows:

The relationship between building age and price, with neighborhoods and property types highlighted.

### Why I used this visualization:

Scatter plots reveal patterns and clustering that indicate how age impacts price.

### Key takeaway:

Newer buildings often sell at higher prices, but location (neighborhood) also strongly influences value.

# 5.Used Cars





## 1. Distribution of Purchased Car Prices

### What the visualization shows:

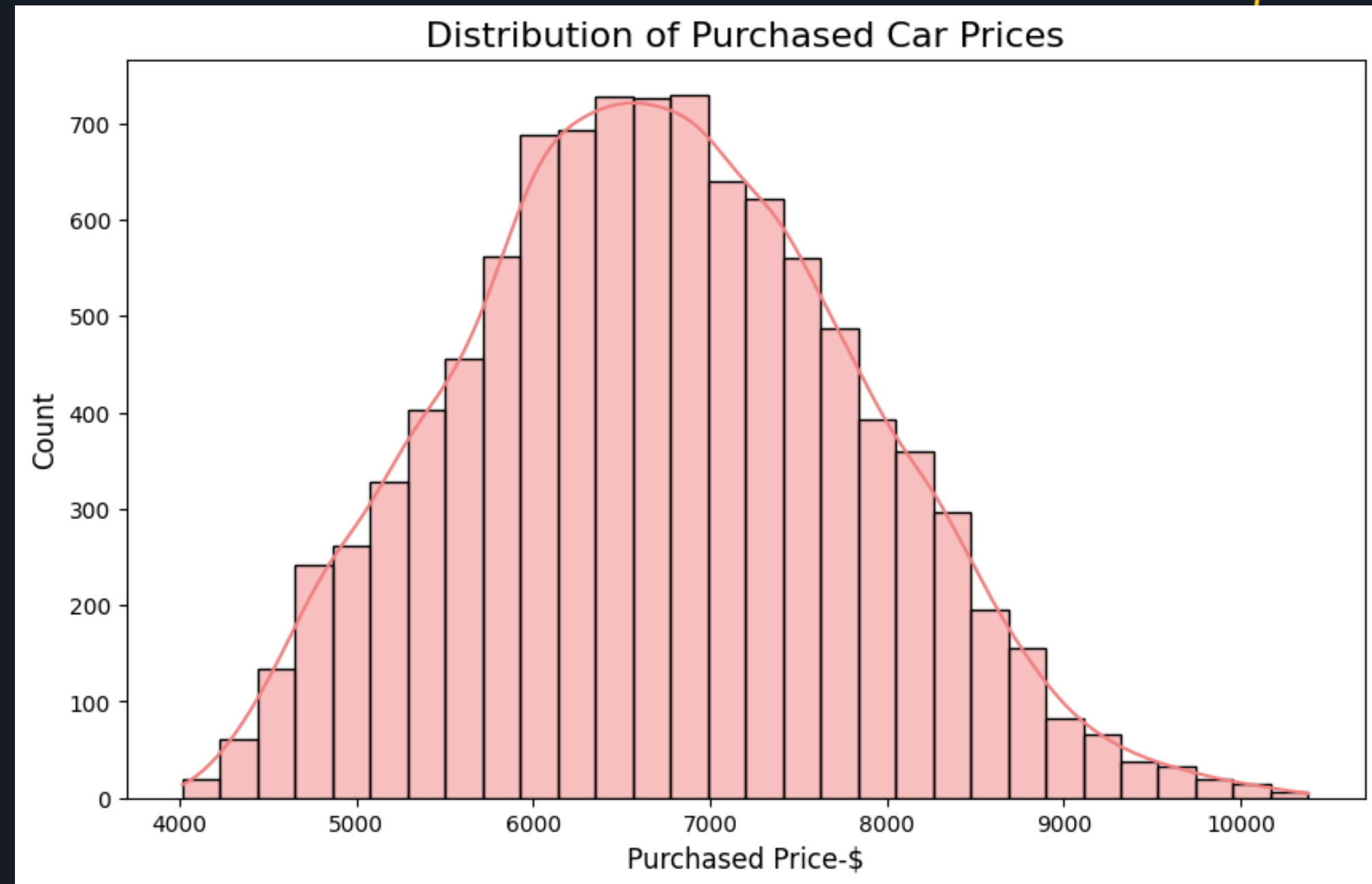
The spread of car purchase prices, with frequency counts and an overlaid smooth curve (KDE).

### Why I used this visualization:

A histogram with KDE highlights the overall distribution, central tendency, and whether prices are skewed.

### Key takeaway:

Most cars fall within a certain purchase price range, with fewer very cheap or very expensive cars.







## 2. Correlation Matrix

### What the visualization shows:

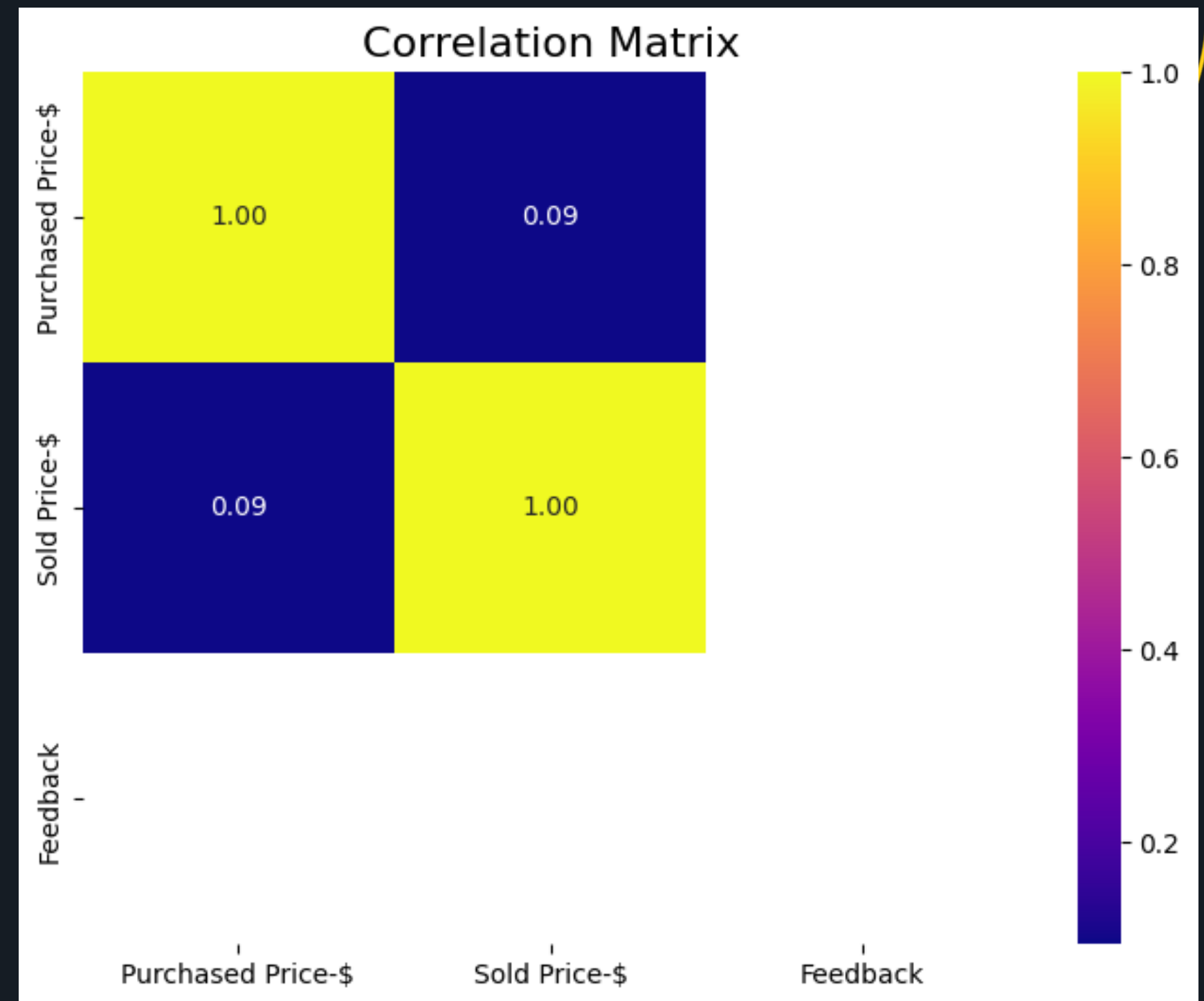
Correlations between numerical features like purchased price, sold price, and feedback.

### Why I used this visualization:

A heatmap makes it easy to see strong positive or negative relationships at a glance.

### Key takeaway:

Purchased price and sold price are strongly correlated, while feedback may have weaker relationships.





### 3. Car Sale Status Distribution

#### What the visualization shows:

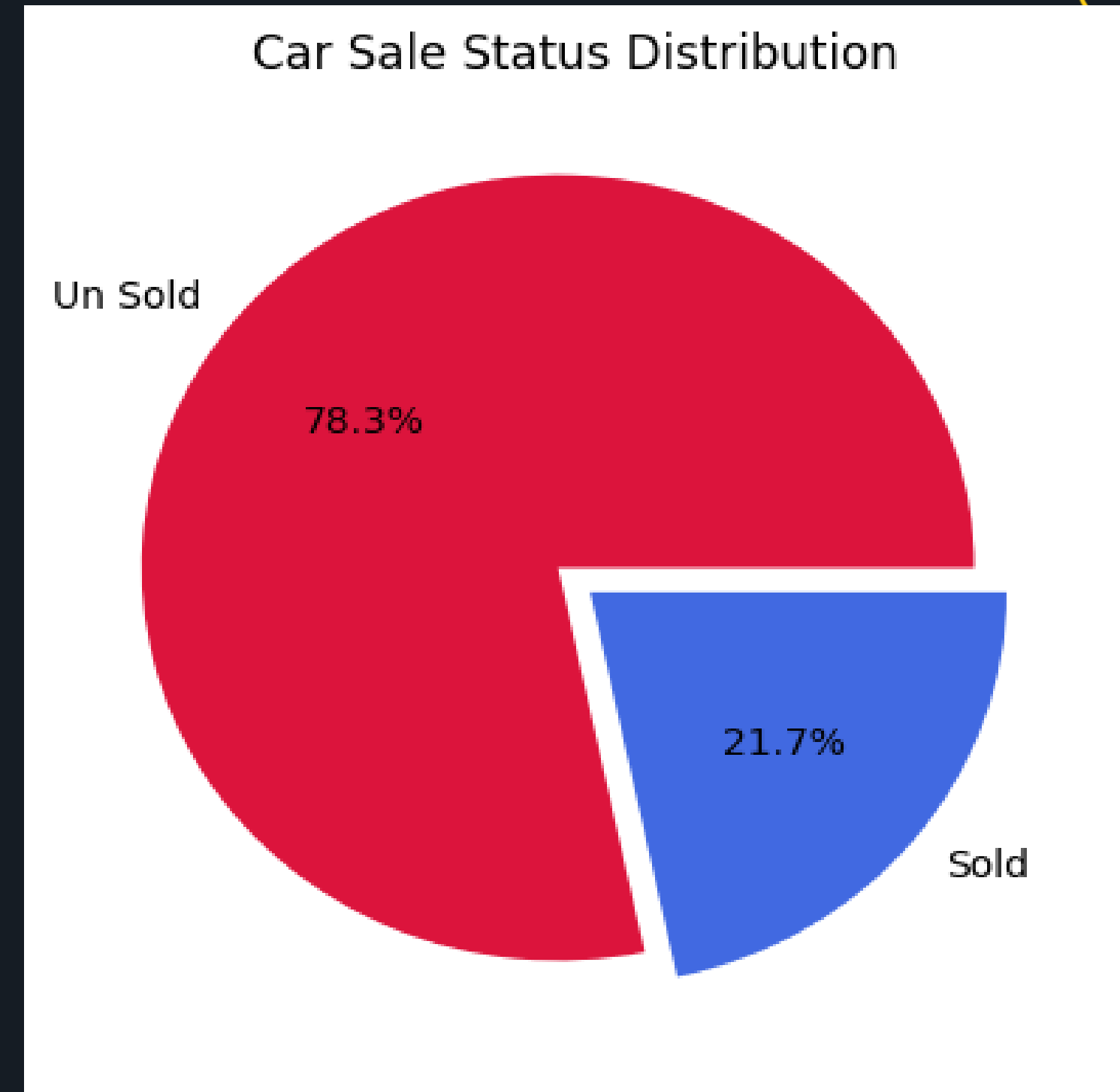
The proportion of cars in different sale statuses (e.g., sold, pending, etc.).

#### Why I used this visualization:

A pie chart is straightforward for showing relative proportions across categories.

#### Key takeaway:

Certain statuses dominate (e.g., most cars are sold), while others are less frequent.









## 5. Distribution of Car Names

### What the visualization shows:

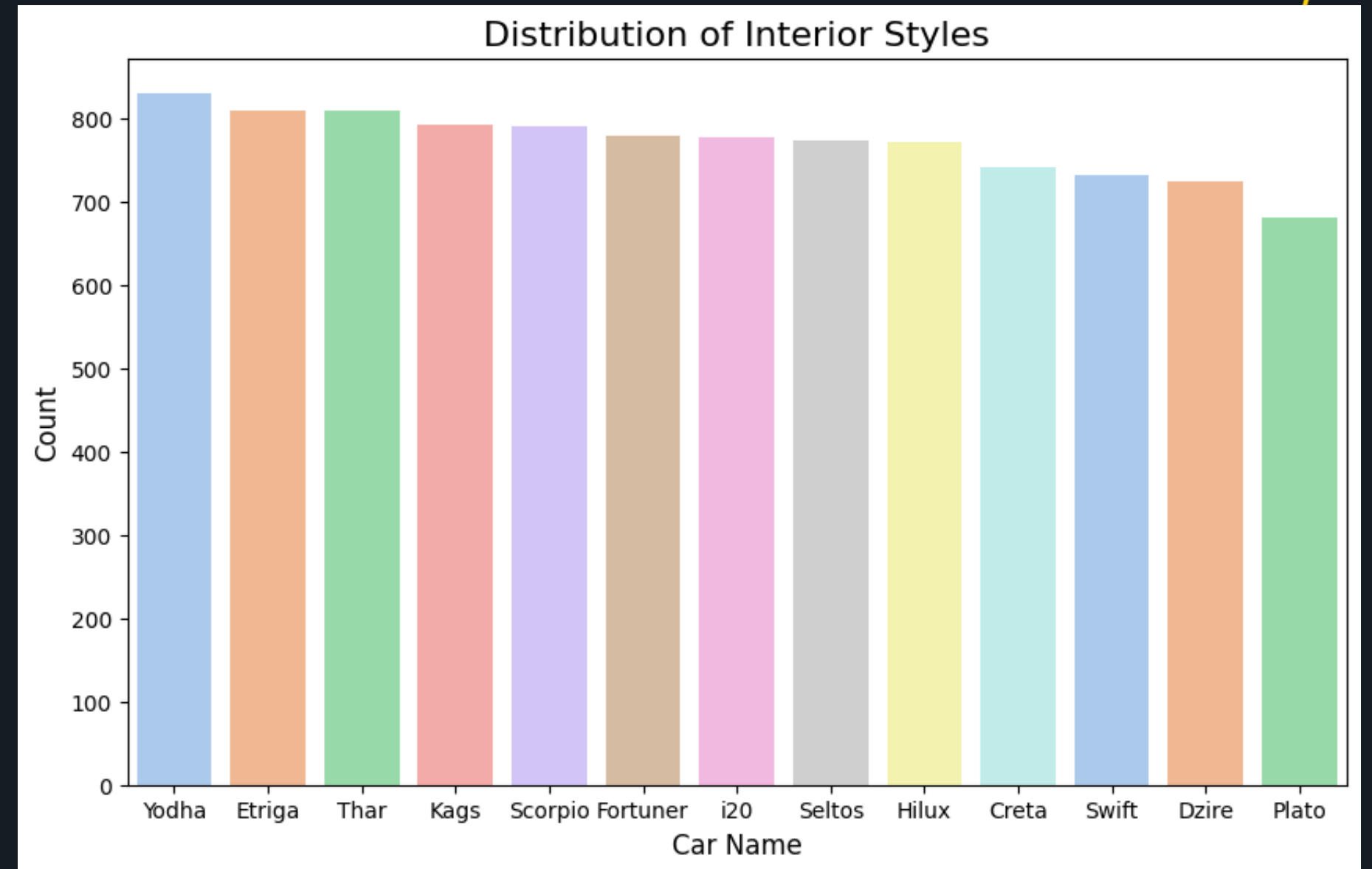
How many times each car model/name appears in the dataset.

### Why I used this visualization:

A count plot clearly shows which car names dominate the dataset.

### Key takeaway:

A few car names are very common, while many appear only rarely.





## 6. Distribution of Number of Seats

### What the visualization shows:

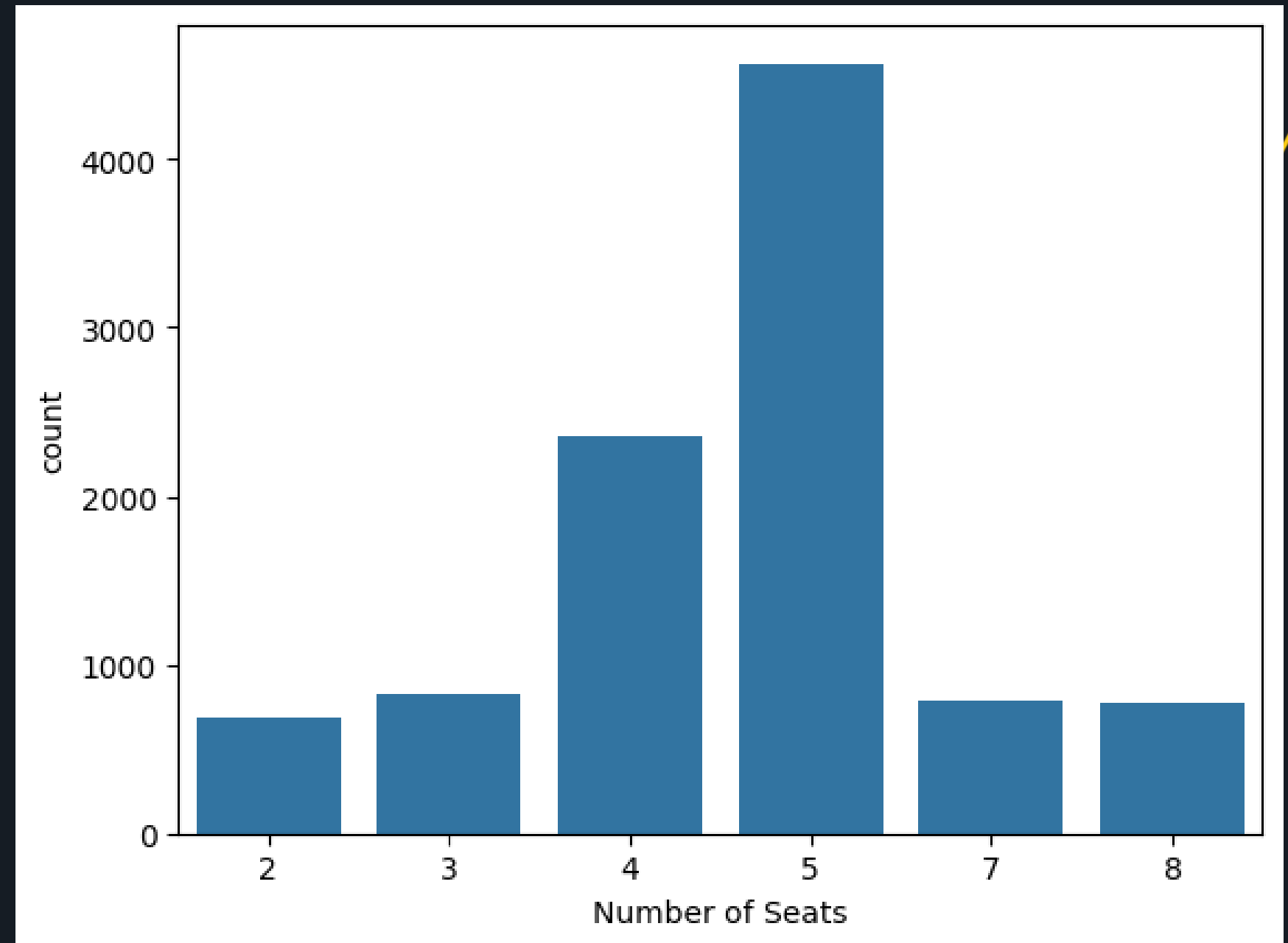
The frequency of cars by their seating capacity.

### Why I used this visualization:

Count plots are effective for categorical or discrete numerical data.

### Key takeaway:

Most cars have a standard seating capacity (like 4–5), with fewer cars outside this range.





## 7. Purchased Price vs. Sold Price

### What the visualization shows:

The relationship between car purchase price and its sold price.

### Why I used this visualization:

A line plot with markers highlights individual transactions and trends.

### Key takeaway:

Sold prices generally follow purchased prices, but some cars resell at higher or lower values than expected.





## 8. Distribution of Manufactured Year

### What the visualization shows:

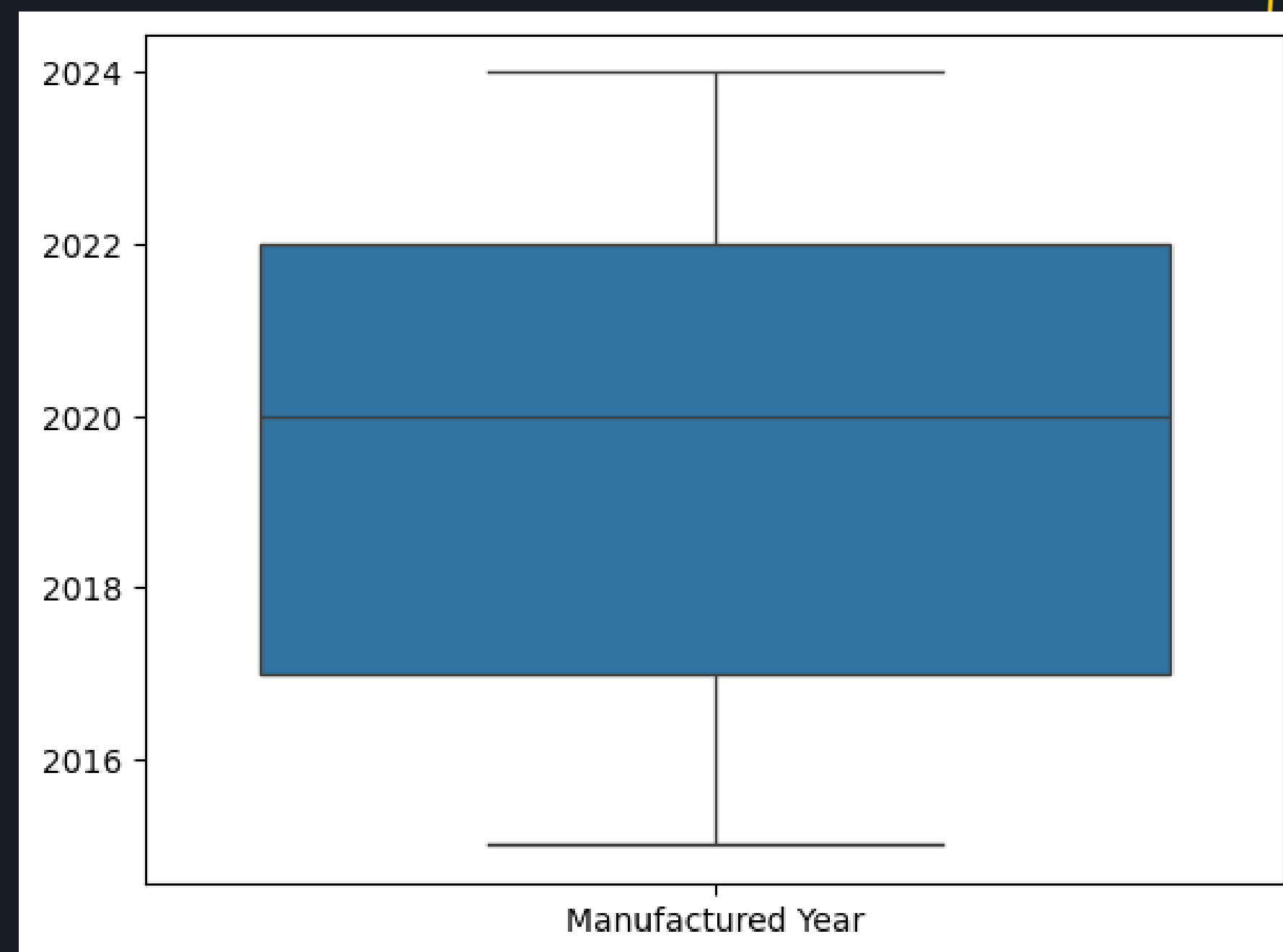
The spread of car manufacturing years, including the median, quartiles, and outliers.

### Why I used this visualization:

A box plot is ideal for detecting outliers and seeing how recent or old the cars in the dataset are.

### Key takeaway:

Most cars were manufactured within a certain modern time range, but there are some outliers representing much older vehicles.



# Thank You

