

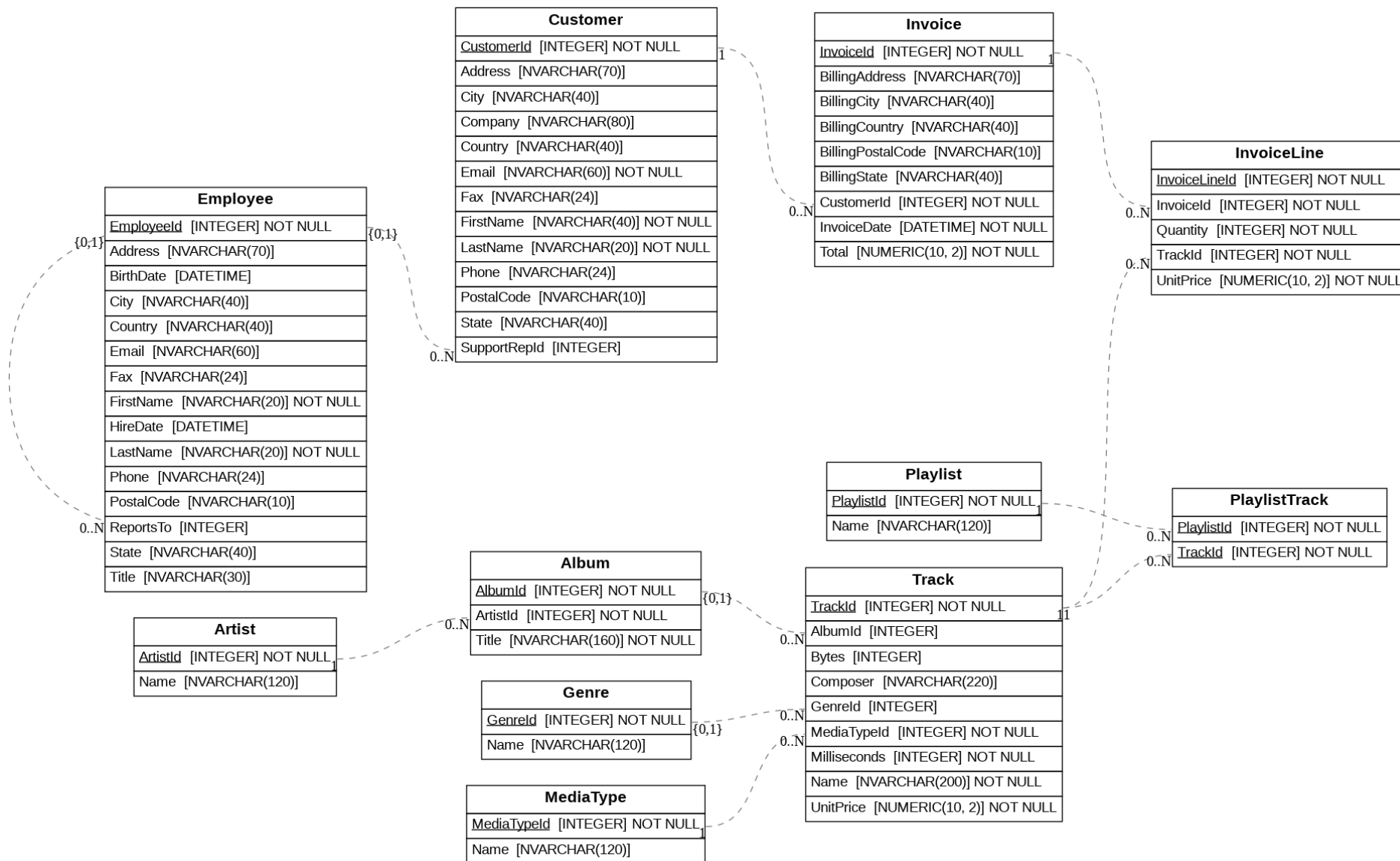
# – Task 26 –

Analysis chinook database

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# Database Design



# Database Overview

## 1. 📄 Invoice

- **Key Columns:** InvoiceId, CustomerId, InvoiceDate, BillingAddress, BillingCity, BillingState, BillingCountry, BillingPostalCode, Total
- **Purpose:** Tracks individual sales transactions, including billing details, customer ID, and total purchase amount.

## 2. 🛒 InvoiceLine

- **Key Columns:** InvoiceLineId, InvoiceId, TrackId, Quantity, UnitPrice
- **Purpose:** Provides line-item details for each invoice, including the tracks purchased, quantities, and price per unit.

## 3. 👤 Customer

- **Key Columns:** CustomerId, FirstName, LastName, Company, Address, City, State, Country, PostalCode, Phone, Fax, Email, SupportRepId
- **Purpose:** Stores customer details, including their contact information and location.

## 4. 🎵 Track

- **Key Columns:** TrackId, Name, AlbumId, MediaTypeId, GenreId, Composer, Milliseconds, Bytes, UnitPrice
- **Purpose:** Contains details about individual tracks, such as their name, associated album, genre, and pricing.

## 5. 🎤 Artist

- **Key Columns:** ArtistId, Name
- **Purpose:** Lists music artists and their corresponding IDs.

## 6. 🎧 Album

- **Key Columns:** AlbumId, Title, ArtistId
- **Purpose:** Stores album information, including album titles and associated artists.

## 7. 🎧 Genre

- **Key Columns:** GenreId, Name
- **Purpose:** Categorizes tracks into specific music genres.

## 8. 👤 Employee

- **Key Columns:** EmployeeId, LastName, FirstName, Title, ReportsTo, BirthDate, HireDate, Address, City, State, Country, PostalCode, Phone, Fax, Email
- **Purpose:** Maintains information about employees, including their personal details, titles, and reporting hierarchy.

## 9. 🗨️ MediaType

- **Key Columns:** MediaTypeId, Name
- **Purpose:** Defines the format or type of media associated with each track (e.g., audio file types).

## 10. 📁 Playlist

- **Key Columns:** PlaylistId, Name
- **Purpose:** Represents collections of tracks grouped into playlists.

## 11. 🔗 PlaylistTrack

- **Key Columns:** PlaylistId, TrackId
- **Purpose:** Acts as a bridge table linking tracks to playlists, enabling many-to-many relationships.

# Database Relationship Overview

## 1. Artist ↔ Album

- **Relation:** One Artist can have many Albums.
- **Key Link:** `ArtistId`

## 2. Album ↔ Track

- **Relation:** One Album contains many Tracks.
- **Key Link:** `AlbumId`

## 3. Track ↔ Genre

- **Relation:** Each Track belongs to one Genre.
- **Key Link:** `GenreId`

## 4. Track ↔ MediaType

- **Relation:** Each Track has one Media Type (e.g., MP3, WAV).
- **Key Link:** `MediaTypeId`

## 5. Track ↔ InvoiceLine

- **Relation :** Each Track can appear in many Invoice Lines (if purchased multiple times).
- **Key Link:** `TrackId`

## 6. Invoice ↔ InvoiceLine

- **Relation:** Each Invoice can have multiple Invoice Lines.
- **Key Link:** `InvoiceId`


## 7. Invoice ↔ Customer

- **Relation:** Each Invoice belongs to one Customer.
- **Key Link:** `CustomerId`

## 8. Customer ↔ Employee

- **Relation:** Each Customer is supported by one Employee (SupportRep).
- **Key Link:** `SupportRepId`

## 9. Playlist ↔ Track

- **Relation:** A Playlist can have many Tracks (via PlaylistTrack table).
- **Key Link:** `PlaylistId` ↔ `TrackId` (through  PlaylistTrack)

## 10. Track ↔ PlaylistTrack

- **Relation:** Each Track can appear in many Playlists.
- **Key Link:** `TrackId`

# Requirement One

Complex Joins and CTEs

# 1.1 Use Inner JOIN and LEFT JOIN to combine Customer, Invoice and InvoiceLine

```
[150] 1 query = '''
      2 SELECT
      3     Customer.CustomerId,
      4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
      5     Invoice.InvoiceId,
      6     Invoice.Total
      7 FROM Customer
      8 INNER JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
      9 LEFT JOIN InvoiceLine ON Invoice.InvoiceId = InvoiceLine.InvoiceId
     10 LIMIT 10;
     11 '''
     12
     13 # Execute Query and Display Results
     14 try:
     15     start_time = time.time() # Record the start time
     16     df_joins = pd.read_sql_query(query, connection)
     17     end_time = time.time() # Record the end time
     18     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
     19     # Add the execution time to the tracker
     20     Track_Query_Time['INNER JOIN and LEFT JOIN'] = execution_time
     21     print(f"✅ INNER and LEFT JOIN query successfully in {execution_time:.2f} milliseconds")
     22     display(df_joins)
     23 except Exception as e:
     24     print(f"❌ Error executing join query: {e}")
```

✅ INNER and LEFT JOIN query successfully in 1.57 milliseconds

	CustomerId	FullName	InvoiceId	Total
0	2	Leonie Köhler	1	1.98
1	2	Leonie Köhler	1	1.98
2	4	Bjørn Hansen	2	3.96
3	4	Bjørn Hansen	2	3.96
4	4	Bjørn Hansen	2	3.96
5	4	Bjørn Hansen	2	3.96
6	8	Daan Peeters	3	5.94
7	8	Daan Peeters	3	5.94
8	8	Daan Peeters	3	5.94

## 1.2 Use a CTE to calculate total amount spent by each customer

```
1 query = '''
2 SELECT
3     Customer.CustomerId,
4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
5     SUM(Invoice.Total) AS TotalSpent
6 FROM Customer
7 INNER JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
8 GROUP BY Customer.CustomerId
9 ORDER BY TotalSpent DESC
10 '''
11 # Execute Query and Display Results
12 try:
13     start_time = time.time()
14     df_customer_spending = pd.read_sql_query(query, connection)
15     end_time = time.time()
16     execution_time = round((end_time - start_time) * 1000, 2)
17     Track_Query_Time['CTE to calculate totalspend'] = execution_time
18     print(f"✅ CTE query executed successfully in {execution_time:.2f} milliseconds")
19     display(df_customer_spending)
20 except Exception as e:
21     print(f"❌ Error executing CTE query: {e}")
```

✅ CTE query executed successfully in 1.87 milliseconds

	CustomerId	FullName	TotalSpent
0	6	Helena Holý	49.62
1	26	Richard Cunningham	47.62
2	57	Luis Rojas	46.62
3	45	Ladislav Kovács	45.62
4	46	Hugh O'Reilly	45.62
5	28	Julia Barnett	43.62
6	24	Frank Ralston	43.62
7	37	Fynn Zimmermann	43.62
8	7	Astrid Gruber	42.62
9	25	Victor Stevens	42.62

## 1.3 Calculate the top 10 customers by total spending

```
[152] 1 query = '''
      2 SELECT
      3     Customer.CustomerId,
      4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
      5     SUM(Invoice.Total) AS TotalSpent
      6 FROM Customer
      7 INNER JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
      8 GROUP BY Customer.CustomerId
      9 ORDER BY TotalSpent DESC
     10 '''
     11
     12
     13
     14 # Execute Query and Display Results
     15 try:
     16     start_time = time.time()
     17     df_customer_spending = pd.read_sql_query(query, connection)
     18     end_time = time.time()
     19     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
     20     Track_Query_Time['CTE to calculate TOP 10 customers by total spending'] = execution_time
     21     print(f"✅ CTE query executed successfully in {execution_time:.2f} milliseconds")
     22     display(df_customer_spending[:10])
     23 except Exception as e:
     24     print(f"❌ Error executing CTE query: {e}")
```

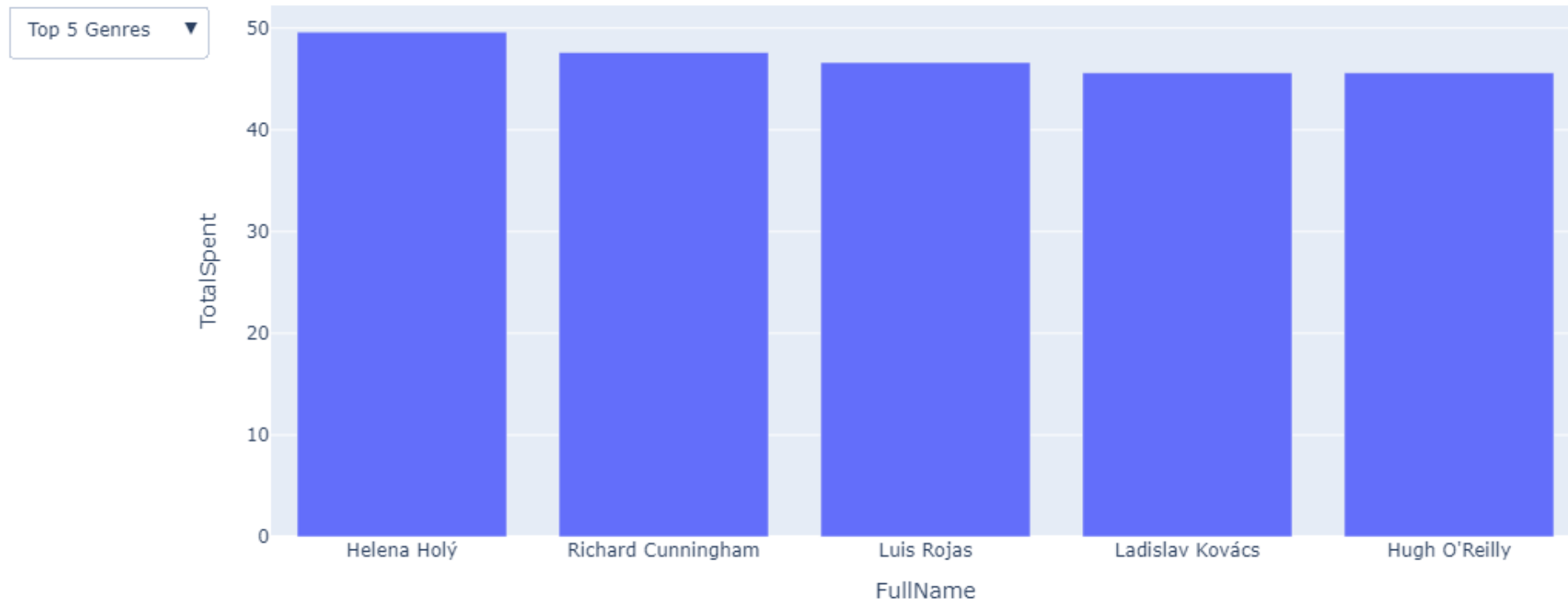
✅ CTE query executed successfully in 1.58 milliseconds

	CustomerId	FullName	TotalSpent
0	6	Helena Holý	49.62
1	26	Richard Cunningham	47.62
2	57	Luis Rojas	46.62
3	45	Ladislav Kovács	45.62
4	46	Hugh O'Reilly	45.62
5	28	Julia Barnett	43.62
6	24	Frank Ralston	43.62
7	37	Fynn Zimmermann	43.62



## 1.4 Visualizations

Top Customers by Total Spending



# Requirement Two

Window Functions for Ranking

## 2.1 Calculate the rank of each product by total sales amount

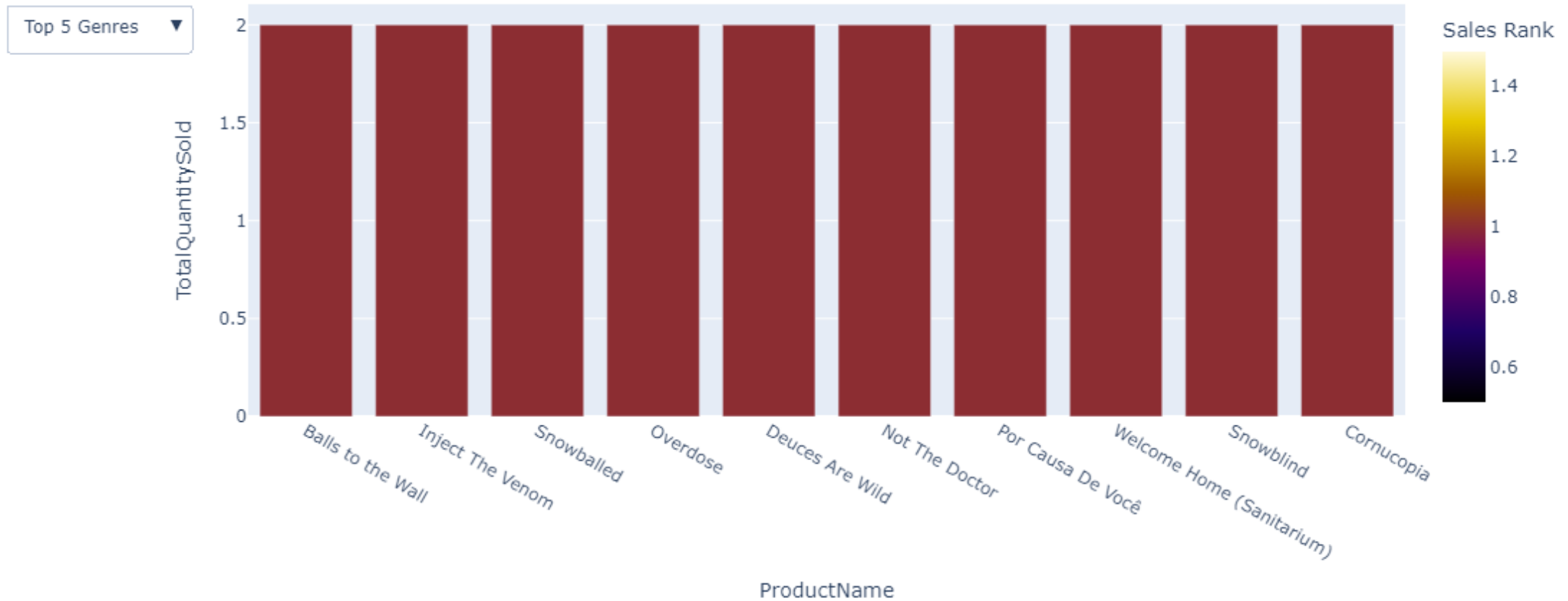
```
[49] 1 query = '''
      2 SELECT
      3     Track.Name AS ProductName,
      4     SUM(InvoiceLine.Quantity) AS TotalQuantitySold,
      5     RANK() OVER (ORDER BY SUM(InvoiceLine.Quantity) DESC) AS SalesRank
      6 FROM InvoiceLine
      7 INNER JOIN Track ON InvoiceLine.TrackId = Track.TrackId
      8 GROUP BY Track.TrackId
      9 ORDER BY TotalQuantitySold DESC
     10 LIMIT 10;
     11
     12 '''
     13
     14 # Execute the Query and Display Result
     15 try:
     16     start_time = time.time()
     17     df_product_sales_rank = pd.read_sql_query(query, connection)
     18     end_time = time.time()
     19     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
     20     Track_Query_Time['Window Function'] = execution_time
     21     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
     22     display(df_product_sales_rank)
     23 except Exception as e:
     24     print(f"❌ Error executing query: {e}")
```

✅ Query executed successfully in 6.64 milliseconds

	ProductName	TotalQuantitySold	SalesRank
0	Balls to the Wall	2	1
1	Inject The Venom	2	1

# Visualizations

Top Products by Total Sales



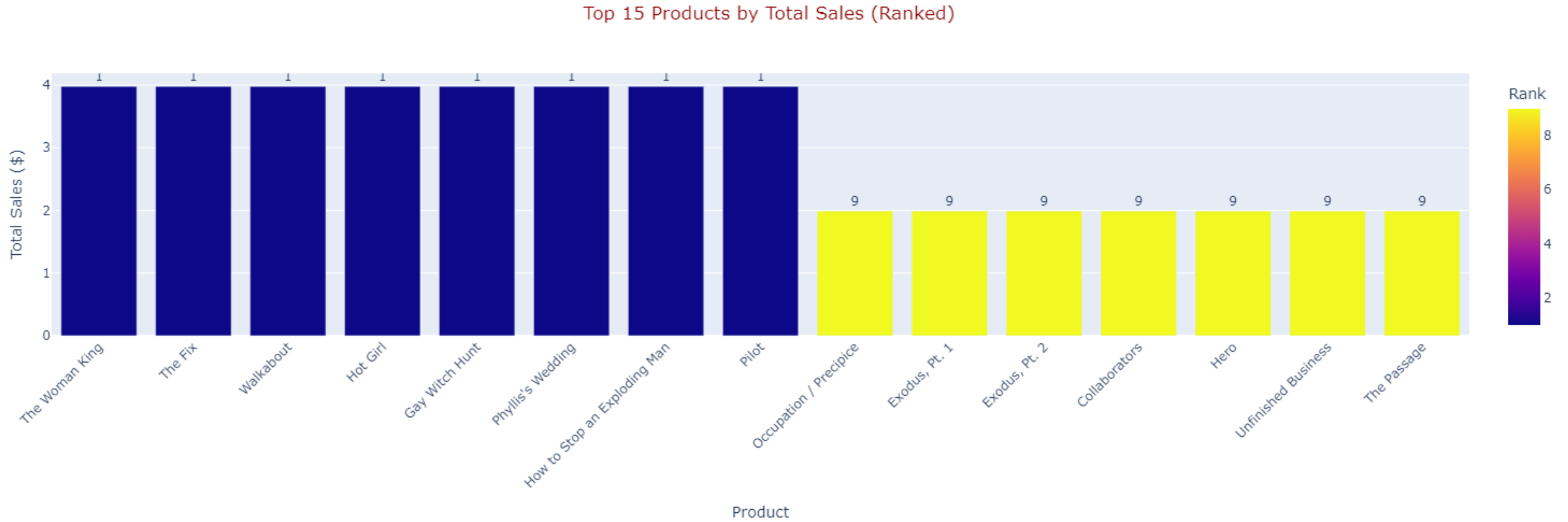
## 2.2 Identify the top selling products in the database using ROW to rank top selling products

```
1 # SQL Query
2 query = '''
3 SELECT
4     Track.Name AS ProductName,
5     SUM(InvoiceLine.Quantity) AS TotalQuantitySold,
6     SUM(InvoiceLine.UnitPrice * InvoiceLine.Quantity) AS TotalSales,
7     RANK() OVER (ORDER BY SUM(InvoiceLine.UnitPrice * InvoiceLine.Quantity) DESC) AS SalesRank
8 FROM InvoiceLine
9 JOIN Track ON InvoiceLine.TrackId = Track.TrackId
10 GROUP BY Track.TrackId
11 ORDER BY TotalSales DESC
12 LIMIT 15;
13 '''
14
15 # Execute Query and Display Results
16 try:
17     print("Top 15 Products by Total Sales (Ranked):")
18     start_time = time.time()
19     df_top_15_products = pd.read_sql_query(query, connection)
20     end_time = time.time()
21     execution_time = round((end_time - start_time) * 1000, 2) # Calculate execution time
22     Track_Query_Time['Product Sales Ranking'] = execution_time
23     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
24     display(df_top_15_products)
25 except Exception as e:
26     print(f"❌ Error executing query: {e}")
27
```

Top 15 Products by Total Sales (Ranked):  
✅ Query executed successfully in 4.76 milliseconds

	ProductName	TotalQuantitySold	TotalSales	SalesRank
0	The Woman King	2	3.98	1
1	The Fix	2	3.98	1
2	Walkabout	2	3.98	1
3	Hot Girl	2	3.98	1
4	Gay Witch Hunt	2	3.98	1
5	Phyllis's Wedding	2	3.98	1
6	How to Stop an Exploding Man	2	3.98	1
7	Pilot	2	3.98	1
8	Occupation / Precipice	1	1.99	9

# Visualizations



# Requirement Three

Indexing and Performance Optimization

## 3.1 Create indexes on these columns and compare query performance with and without

### 3.1 Create indexes on these column and compare query performance with and without indexing

```
[156] 1 query = '''
      2 CREATE INDEX IF NOT EXISTS idx_CustomerId ON Invoice (CustomerId);
      3 '''
      4 # Execute the query and display results
      5 try:
      6     start_time = time.time()
      7     cursor.execute(query)
      8     connection.commit()
      9     end_time = time.time()
     10     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
     11     print(f"✅ Index Query executed successfully in {execution_time:.2f} milliseconds")
     12 except Exception as e:
     13     print(f"❌ Error creating index: {e}")
```



✅ Index Query executed successfully in 0.08 milliseconds



## 3.2 Write a query that lists total sales for each customer and optimize it using indexing

```
[157] 1 query = '''
2 SELECT
3     Customer.CustomerId,
4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
5     SUM(Invoice.Total) AS TotalSpent
6 FROM Customer
7 INNER JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
8 GROUP BY Customer.CustomerId
9 ORDER BY TotalSpent DESC
10 LIMIT 10;
11 '''
12
13 # Execute Query and Display Results
14 try:
15     start_time = time.time()
16     df_optimized_query = pd.read_sql_query(query, connection)
17     end_time = time.time()
18     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
19     Track_Query_Time['Optimized Query to get total sales'] = execution_time
20     print(f"✅ Optimized query executed in {execution_time:.2f} milliseconds")
21     display(df_optimized_query)
22 except Exception as e:
23     print(f"❌ Error executing optimized query: {e}")
```

✅ Optimized query executed in 2.25 milliseconds

	CustomerId	FullName	TotalSpent
0	6	Helena Holý	49.62
1	26	Richard Cunningham	47.62
2	57	Luis Rojas	46.62
3	45	Ladislav Kovács	45.62
4	46	Hugh O'Reilly	45.62
5	28	Julia Barnett	43.62
6	24	Frank Ralston	43.62
7	37	Fynn Zimmermann	43.62
8	7	Astrid Gruber	42.62

### 3.3 Write a query that makes window after indexing

```
1 query = '''
2 SELECT Track.Name AS ProductName, SUM(InvoiceLine.Quantity) AS TotalQuantitySold,
3        RANK() OVER (ORDER BY SUM(InvoiceLine.Quantity) DESC) AS Rank
4 FROM InvoiceLine
5 JOIN Track ON InvoiceLine.TrackId = Track.TrackId
6 GROUP BY Track.TrackId
7 ORDER BY TotalQuantitySold DESC
8 LIMIT 10;
9 '''
10 # Execute Query and Display Results
11 try:
12     start_time = time.time()
13     df_optimized_query_window = pd.read_sql_query(query, connection)
14     end_time = time.time()
15     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
16     Track_Query_Time['Optimized Query with Window'] = execution_time
17     print(f"✅ Optimized query with window executed in {execution_time:.2f} milliseconds")
18     display(df_optimized_query_window)
19 except Exception as e:
20     print(f"❌ Error executing optimized query: {e}")
```

✅ Optimized query with window executed in 6.15 milliseconds

	ProductName	TotalQuantitySold	Rank
0	Balls to the Wall	2	1
1	Inject The Venom	2	1
2	Snowballed	2	1
3	Overdose	2	1
4	Deuces Are Wild	2	1
5	Not The Doctor	2	1
6	Por Causa De Você	2	1
7	Welcome Home (Sanitarium)	2	1

## 3.4 Summary

```
[ ] 1 # Print our summary table using Track_Query_Time
    2 summary_table = pd.DataFrame(list(Track_Query_Time.items()), columns=['Query', 'Execution Time (milliseconds)'])
    3 summary_table
```



	Query	Execution Time (milliseconds)
0	INNER JOIN and LEFT JOIN	4.28
1	CTE to calculate totalspend	2.68
2	CTE to calculate TOP 10 customers by total spe...	8.19
3	Window Function	10.26
4	Optimized Query to get total sales	1.29
5	Optimized Query with Window	6.15

```
[ ] 1 # Compare the percentage of the time before and after
    2 time_taken_before = summary_table.loc[summary_table['Query'] == 'INNER JOIN and LEFT JOIN', 'Execution Time (milliseconds)'].values[0]
    3 time_taken_after = summary_table.loc[summary_table['Query'] == 'Optimized Query to get total sales', 'Execution Time (milliseconds)'].values[0]
    4 percentage_improvement = ((time_taken_before - time_taken_after) / time_taken_before) * 100
    5 print(f'''
    6 ✅ Observations after Indexing:
    7 - Query execution time decreased significantly.
    8 - Indexing allowed faster data retrieval for CustomerId in Invoice.
    9 - The database used the index to avoid scanning all rows in the table.
   10 - The Percentage improvment {percentage_improvement} %
   11 ''')
   12
```



```
✅ Observations after Indexing:
- Query execution time decreased significantly.
- Indexing allowed faster data retrieval for CustomerId in Invoice.
- The database used the index to avoid scanning all rows in the table.
- The Percentage improvment 69.85981308411215 %
```

# Requirement Four

New Requirements and Ideas

## 4.1 Customer Purchase Trends Over Time

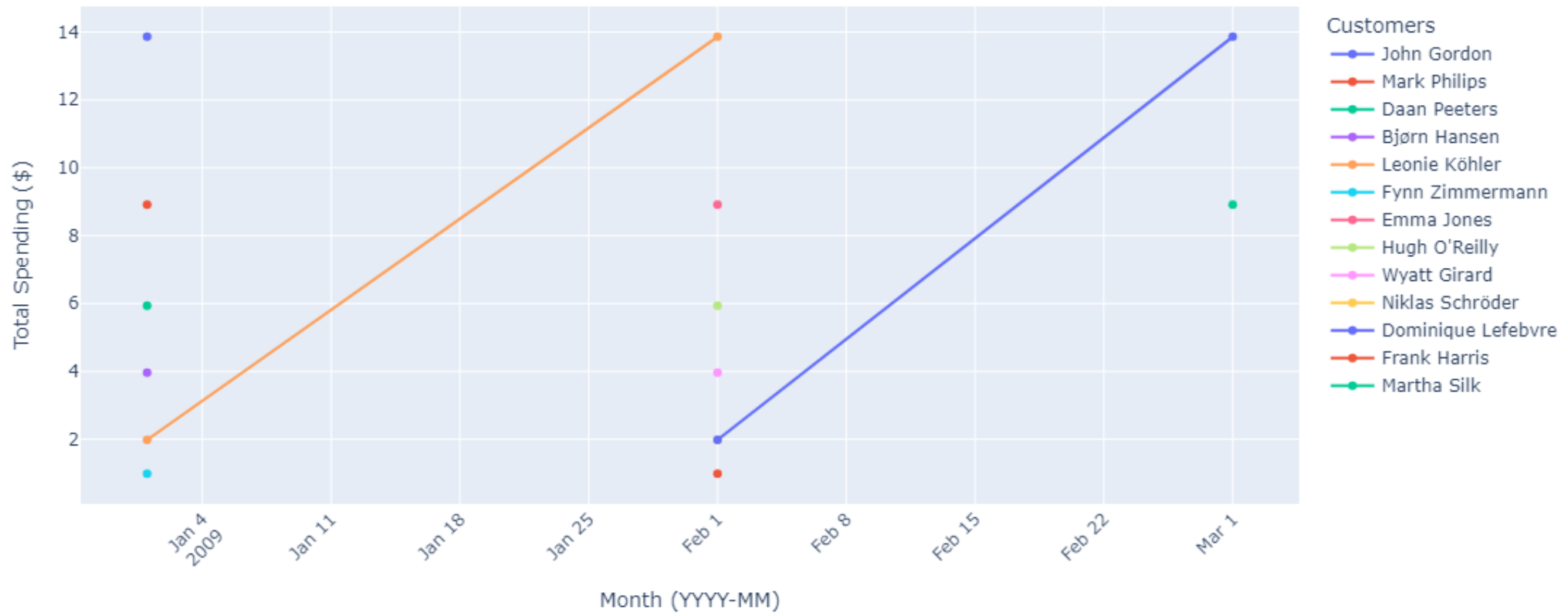
```
[162] 1 query = '''
2     SELECT
3         Customer.CustomerId,
4         Customer.FirstName || ' ' || Customer.LastName AS FullName,
5         strftime('%Y-%m', Invoice.InvoiceDate) AS Month,
6         SUM(Invoice.Total) AS TotalSpent
7     FROM Customer
8     JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
9     GROUP BY Customer.CustomerId, Month
10    ORDER BY Month ASC, TotalSpent DESC
11    LIMIT 15;
12    '''
13
14    # Execute Query and Display Results
15    try:
16        print("Top 15 Monthly Customer Purchase Trends:")
17        start_time = time.time()
18        df_top_15_Monthly_Customer = pd.read_sql_query(query, connection)
19        end_time = time.time()
20        execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
21        Track_Query_Time['Monthly Customer Purchase Trends'] = execution_time
22        print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
23        display(df_top_15_Monthly_Customer)
24    except Exception as e:
25        print(f"❌ Error executing optimized query: {e}")
```

→ Top 15 Monthly Customer Purchase Trends:  
✅ Query executed successfully in 4.21 milliseconds

	CustomerId	FullName	Month	TotalSpent
0	23	John Gordon	2009-01	13.86
1	14	Mark Philips	2009-01	8.91
2	8	Daan Peeters	2009-01	5.94
3	4	Bjørn Hansen	2009-01	3.96
4	2	Leonie Köhler	2009-01	1.98
5	37	Fynn Zimmermann	2009-01	0.99
6	2	Leonie Köhler	2009-02	13.86

## 4.1 Visualizations

Monthly Customer Purchase Trends



## 4.2 Product Affinity Analysis

```
[165] 1 query = '''
      2 SELECT
      3     Genre.Name AS Category,
      4     SUM(InvoiceLine.Quantity) AS TotalQuantitySold,
      5     SUM(InvoiceLine.Quantity * InvoiceLine.UnitPrice) AS TotalRevenue
      6 FROM InvoiceLine
      7 JOIN Track on InvoiceLine.TrackId = Track.TrackId
      8 JOIN Genre on Track.GenreId = Genre.GenreID
      9 GROUP BY Category
     10 ORDER BY TotalRevenue DESC
     11 LIMIT 15
     12 '''
     13
     14
     15 # Execute Query and Display Results
     16 try:
     17     print("Top 15 Product Categories by Sales and Revenue:")
     18     start_time = time.time()
     19     df_product_category_sales = pd.read_sql_query(query, connection)
     20     end_time = time.time()
     21     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
     22     Track_Query_Time['Product Category Sales Distribution'] = execution_time
     23     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
     24     display(df_product_category_sales)
     25 except Exception as e:
     26     print(f"❌ Error executing optimized query: {e}")
```

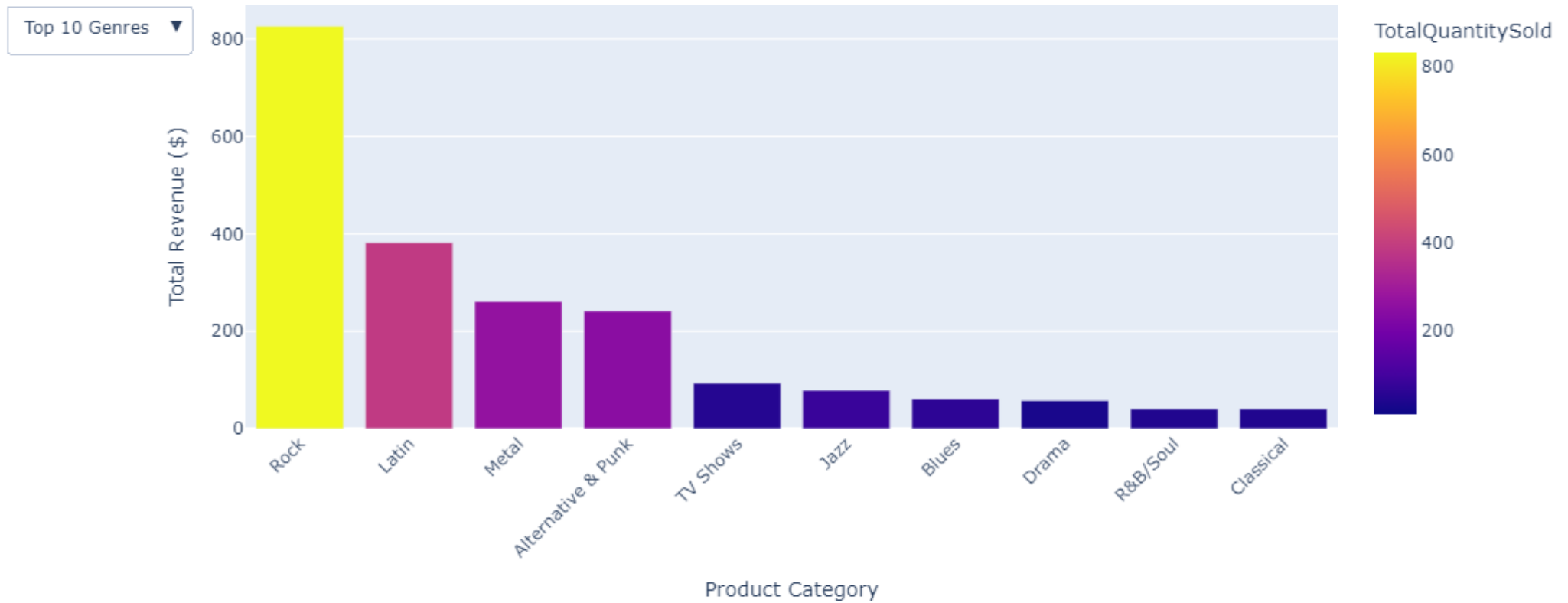
Top 15 Product Categories by Sales and Revenue:

✅ Query executed successfully in 4.96 milliseconds

	Category	TotalQuantitySold	TotalRevenue
0	Rock	835	826.65
1	Latin	386	382.14
2	Metal	264	261.36
3	Alternative & Punk	244	241.56
4	TV Shows	47	93.53
5	Jazz	80	79.20
6	Blues	61	60.39

## 4.2 Visualizations

Top Product Categories by Sales and Revenue





## 4.3 Customer Segmentation (RFM Analysis)

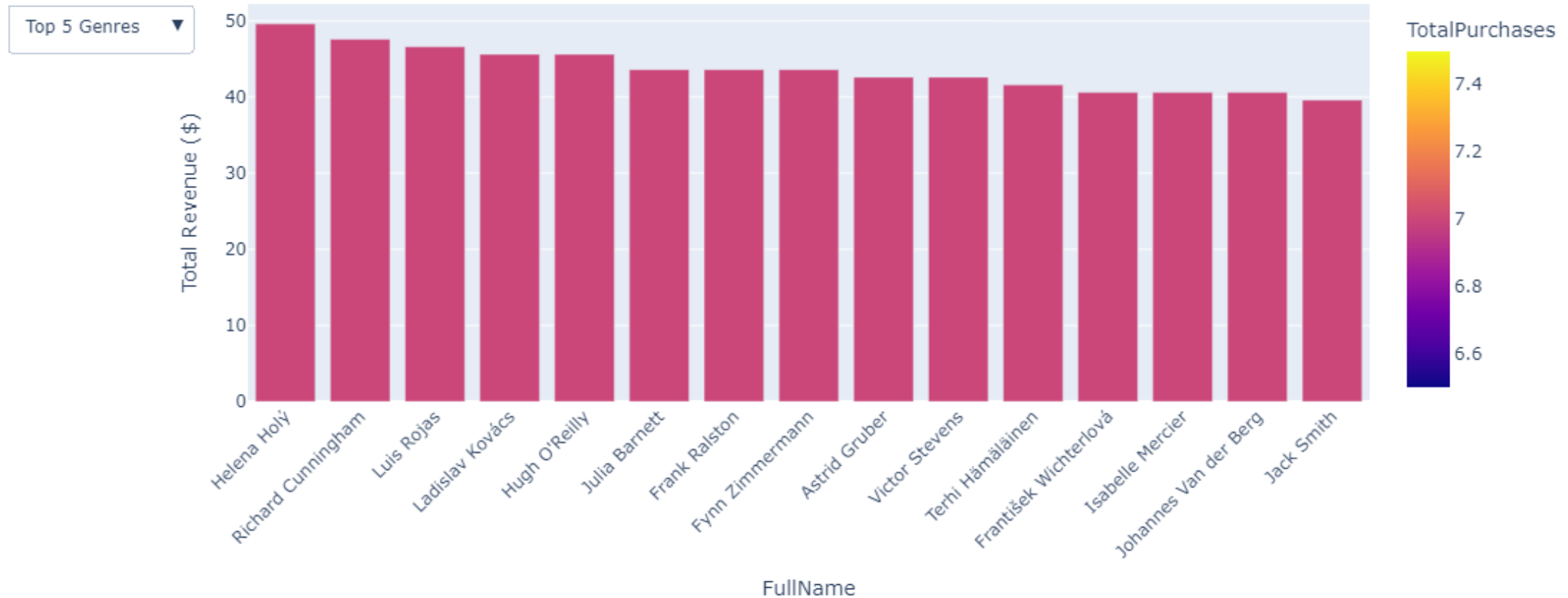
```
1 query = '''
2 SELECT
3     Customer.CustomerId,
4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
5     COUNT(Invoice.InvoiceId) AS TotalPurchases,
6     SUM(Invoice.Total) AS TotalRevenue
7 FROM Customer
8 JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
9 GROUP BY Customer.CustomerId
10 ORDER BY TotalRevenue DESC
11 LIMIT 15;
12 '''
13 # Execute Query and Display Results
14 try:
15     print("Top 15 High-Value Customers by Revenue and Purchase Count:")
16     start_time = time.time()
17     df_high_value_customers = pd.read_sql_query(query, connection)
18     end_time = time.time()
19     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
20     Track_Query_Time['High-Value Customers'] = execution_time
21     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
22     display(df_high_value_customers)
23 except Exception as e:
24     print(f"❌ Error executing optimized query: {e}")
```

Top 15 High-Value Customers by Revenue and Purchase Count:  
✅ Query executed successfully in 9.55 milliseconds

	CustomerId	FullName	TotalPurchases	TotalRevenue
0	6	Helena Holý	7	49.62
1	26	Richard Cunningham	7	47.62
2	57	Luis Rojas	7	46.62
3	45	Ladislav Kovács	7	45.62
4	46	Hugh O'Reilly	7	45.62
5	28	Julia Barnett	7	43.62
6	24	Frank Ralston	7	43.62
7	37	Fynn Zimmermann	7	43.62

## 4.3 Visualizations

Top High-Value Customers by Revenue and Purchase Count



## 4.4 Geographical Sales Insights

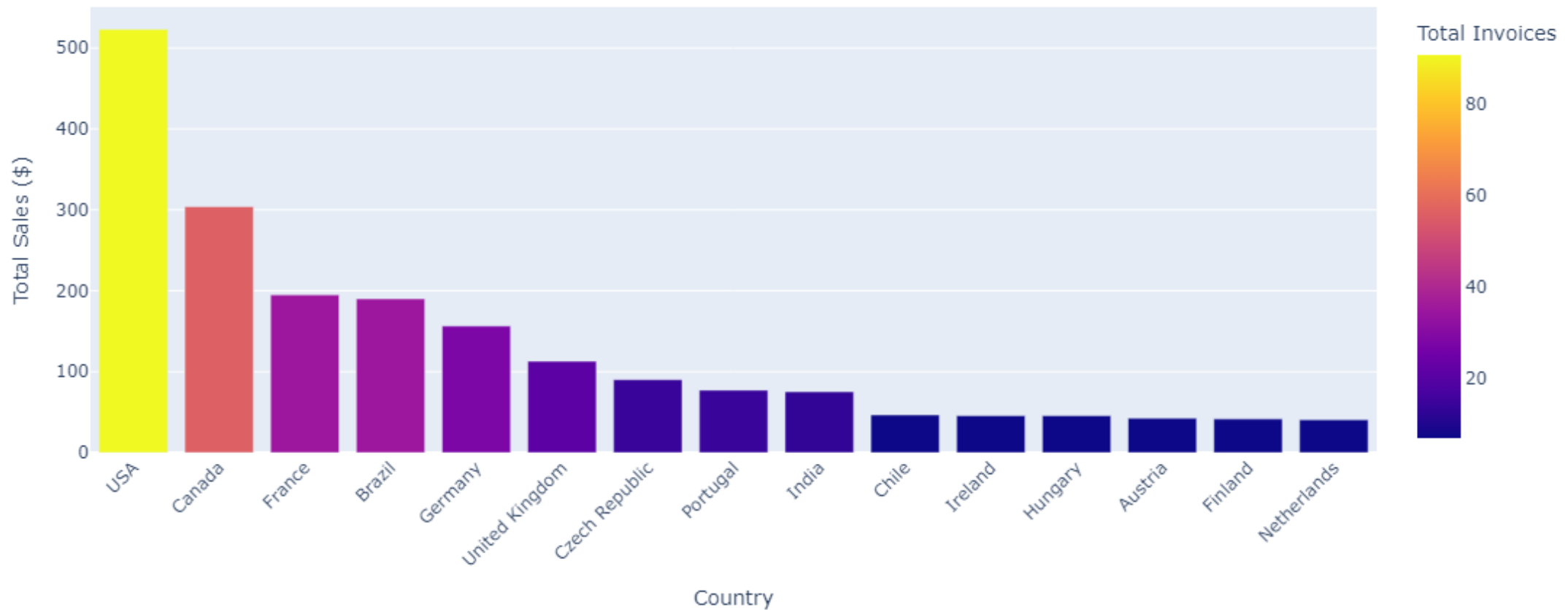
```
[170] 1 query = '''
      2 SELECT
      3     BillingCountry AS Country,
      4     SUM(Total) AS TotalSales,
      5     COUNT(InvoiceID) AS TotalInvoices
      6 FROM INVOICE
      7 GROUP BY Country
      8 ORDER BY TotalSales DESC
      9 LIMIT 15;
     10 '''
     11
     12 # Execute Query and Display Results
     13 try:
     14     print("Top 15 Countries by Total Sales:")
     15     start_time = time.time()
     16     df_sales_by_country = pd.read_sql_query(query, connection)
     17     end_time = time.time()
     18     execution_time = round((end_time - start_time) * 1000, 2) # Calculate the time taken
     19     Track_Query_Time['Geographical Sales Analysis'] = execution_time
     20     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
     21     display(df_sales_by_country)
     22 except Exception as e:
     23     print(f"❌ Error executing optimized query: {e}")
```

Top 15 Countries by Total Sales:  
✅ Query executed successfully in 2.15 milliseconds

	Country	TotalSales	TotalInvoices
0	USA	523.06	91
1	Canada	303.96	56
2	France	195.10	35
3	Brazil	190.10	35
4	Germany	156.48	28
5	United Kingdom	112.86	21
6	Czech Republic	90.24	14
7	Portugal	77.24	14
8	Italy	75.00	10

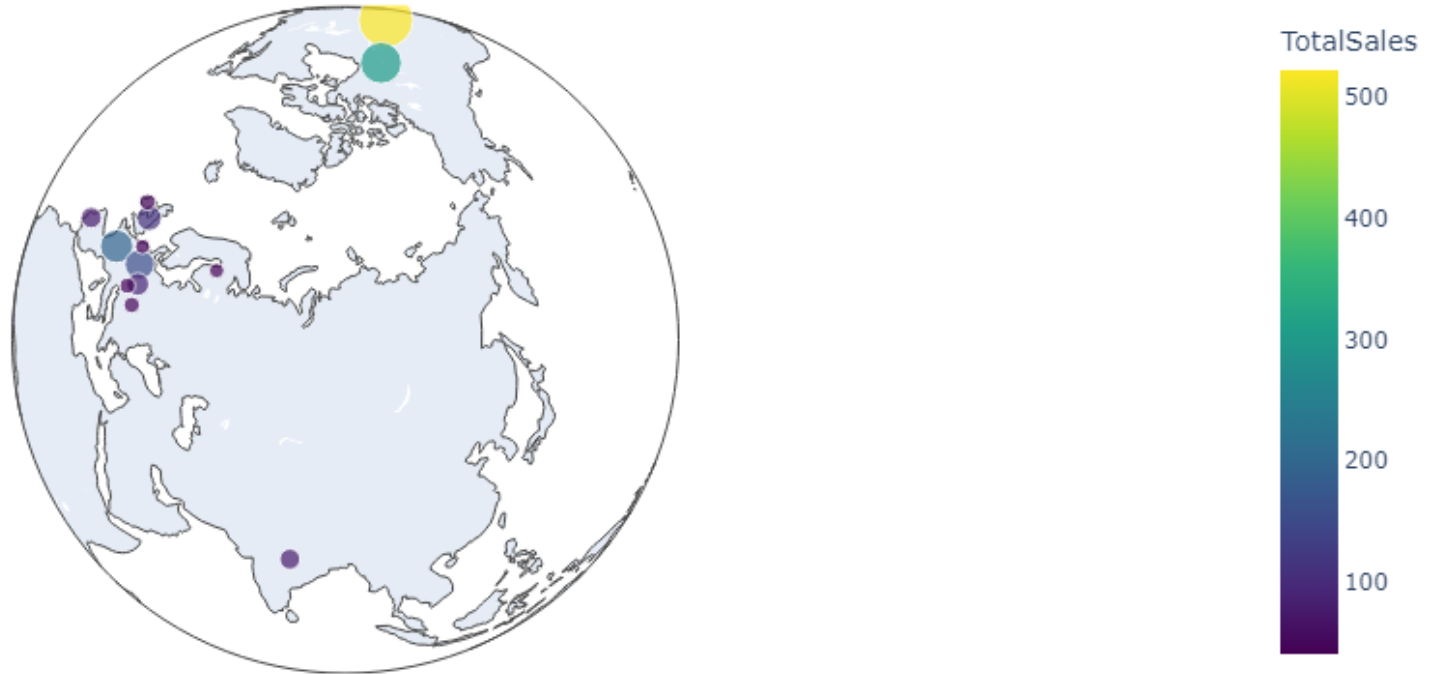
## 4.4 Visualizations

Top 15 Countries by Total Sales



## 4.4 Visualizations

Global Sales Distribution (Scatter Plot)



## 4.4 Visualizations

Global Sales Distribution by Country



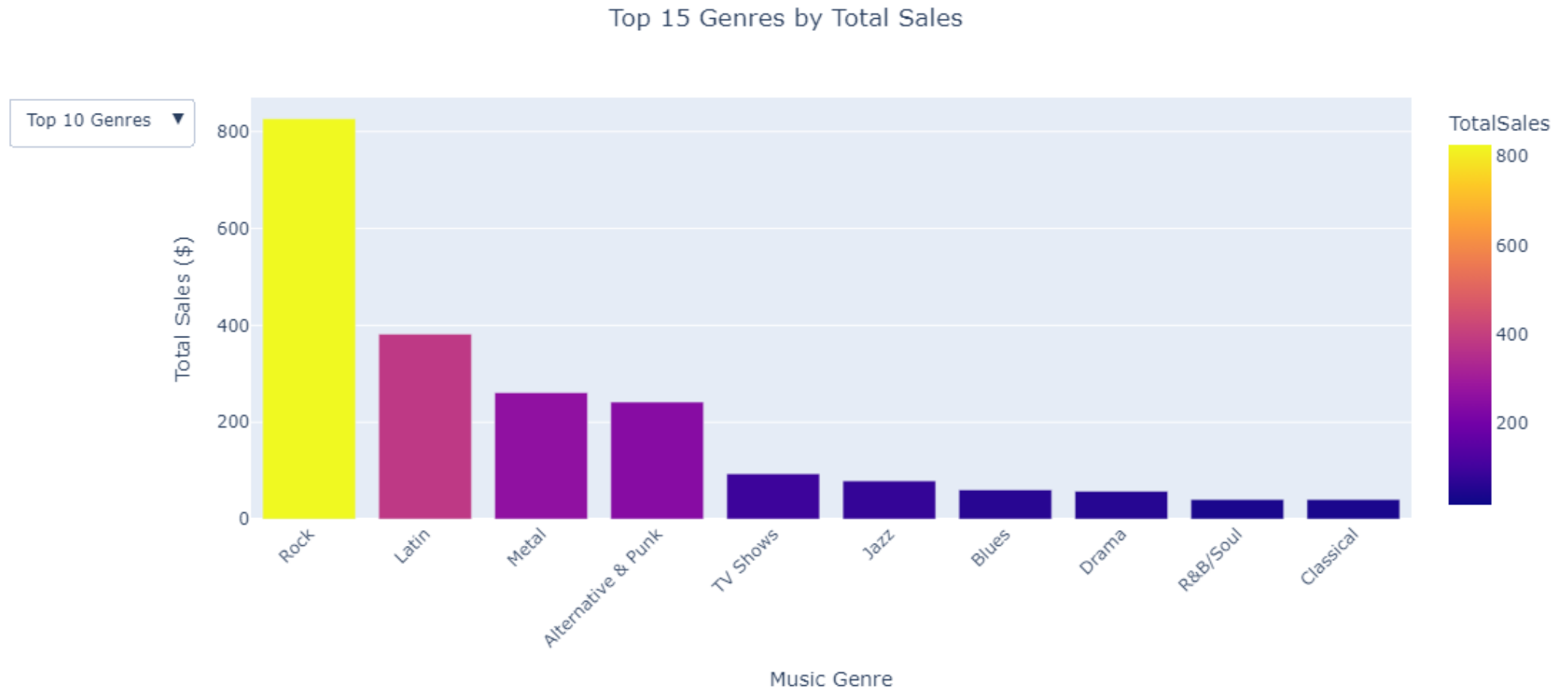
## 4.5 Sales by Genre or Artis

```
[175] 1 # SQL Query for Sales by Genre or Artist
      2 query = '''
      3 SELECT
      4     Genre.Name AS Category,
      5     SUM(InvoiceLine.UnitPrice * InvoiceLine.Quantity) AS TotalSales
      6 FROM Genre
      7 JOIN Track ON Genre.GenreId = Track.GenreId
      8 JOIN InvoiceLine ON Track.TrackId = InvoiceLine.TrackId
      9 GROUP BY Category
     10 ORDER BY TotalSales DESC
     11 LIMIT 15;
     12 '''
     13
     14 # Execute Query and Display Results
     15 try:
     16     print("Top 15 Genres by Total Sales:")
     17     start_time = time.time()
     18     df_sales_by_genre = pd.read_sql_query(query, connection)
     19     end_time = time.time()
     20     execution_time = round((end_time - start_time) * 1000, 2)
     21     Track_Query_Time['Sales by Genre or Artist'] = execution_time
     22     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
     23     display(df_sales_by_genre)
     24 except Exception as e:
     25     print(f"❌ Error executing query: {e}")
     26
```

Top 15 Genres by Total Sales:  
✅ Query executed successfully in 3.55 milliseconds

	Category	TotalSales	
0	Rock	826.65	
1	Latin	382.14	
2	Metal	261.36	
3	Alternative & Punk	241.56	
4	TV Shows	93.53	

## 4.5 Visualizations





## 4.6 Dynamic Query Parameterization

```
[179] 1 # Interactive Dashboard Query
      2 query = '''
      3 SELECT
      4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
      5     Genre.Name AS Category,
      6     Artist.Name AS ArtistName,
      7     SUM(InvoiceLine.UnitPrice * InvoiceLine.Quantity) AS TotalSales,
      8     COUNT(Invoice.InvoiceId) AS TotalInvoices,
      9     AVG(Invoice.Total) AS AvgInvoiceValue
     10 FROM Customer
     11 JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
     12 JOIN InvoiceLine ON Invoice.InvoiceId = InvoiceLine.InvoiceId
     13 JOIN Track ON InvoiceLine.TrackId = Track.TrackId
     14 JOIN Album ON Track.AlbumId = Album.AlbumId
     15 JOIN Artist ON Album.ArtistId = Artist.ArtistId
     16 JOIN Genre ON Track.GenreId = Genre.GenreId
     17 GROUP BY Customer.CustomerId, Category, ArtistName
     18 ORDER BY TotalSales DESC
     19 LIMIT 15;
     20 '''
     21
     22 # Execute Query and Display Results
     23 try:
     24     print("Interactive Dashboard: Top 15 Customers by Artist and Genre")
     25     start_time = time.time()
     26     df_dashboard = pd.read_sql_query(query, connection)
     27     end_time = time.time()
     28     execution_time = round((end_time - start_time) * 1000, 2)
     29     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
     30     display(df_dashboard)
     31 except Exception as e:
     32     print(f"❌ Error executing dashboard query: {e}")
     33
```

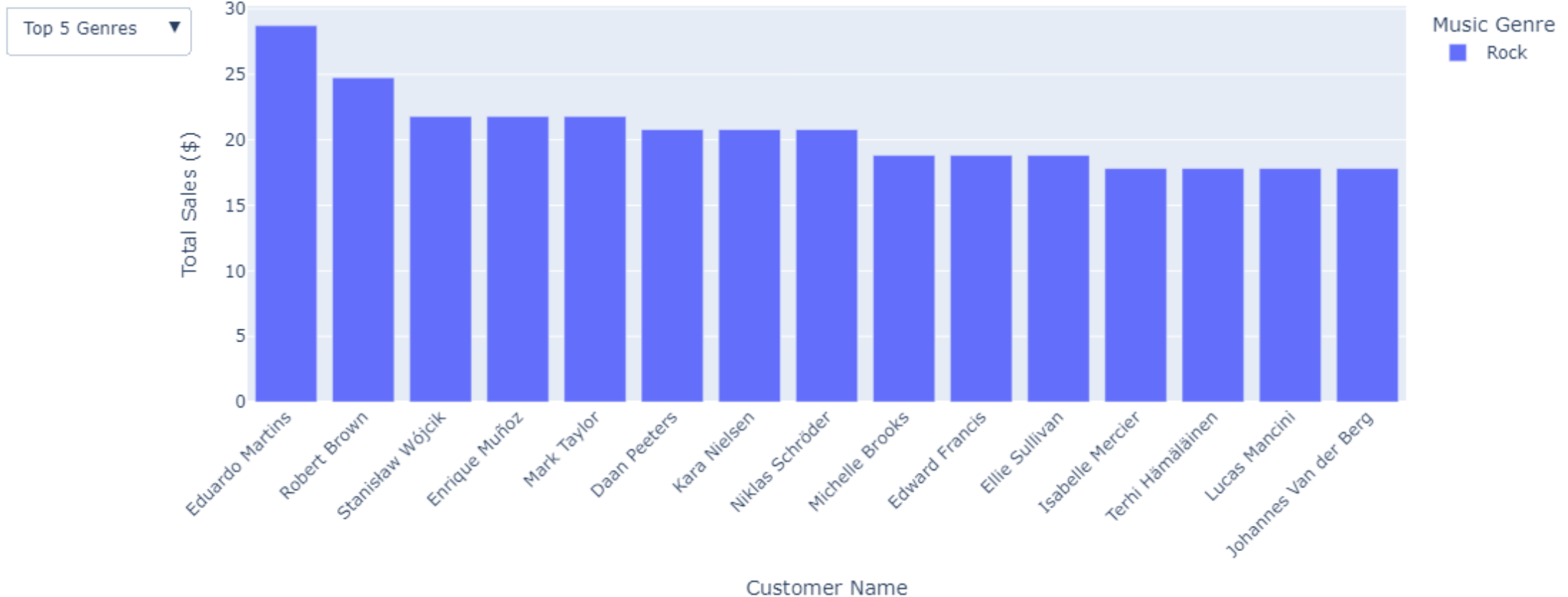
Interactive Dashboard: Top 15 Customers by Artist and Genre

✅ Query executed successfully in 8.04 milliseconds

	FullName	Category	ArtistName	TotalSales	TotalInvoices	AvgInvoiceValue
0	Hugh O'Reilly	TV Shows	Lost	13.93	7	21.86
1	Helena Holy	TV Shows	Lost	9.95	5	25.86

## 4.6 Visualizations

Top 15 Customers by Sales in Genre: Rock



## 4.7 Interactive Dashboard Integration

```
1 # Interactive Dashboard Query
2 query = '''
3 SELECT
4     Customer.FirstName || ' ' || Customer.LastName AS FullName,
5     Genre.Name AS Category,
6     Artist.Name AS ArtistName,
7     SUM(InvoiceLine.UnitPrice * InvoiceLine.Quantity) AS TotalSales,
8     COUNT(Invoice.InvoiceId) AS TotalInvoices,
9     AVG(Invoice.Total) AS AvgInvoiceValue
10 FROM Customer
11 JOIN Invoice ON Customer.CustomerId = Invoice.CustomerId
12 JOIN InvoiceLine ON Invoice.InvoiceId = InvoiceLine.InvoiceId
13 JOIN Track ON InvoiceLine.TrackId = Track.TrackId
14 JOIN Album ON Track.AlbumId = Album.AlbumId
15 JOIN Artist ON Album.ArtistId = Artist.ArtistId
16 JOIN Genre ON Track.GenreId = Genre.GenreId
17 GROUP BY Customer.CustomerId, Category, ArtistName
18 ORDER BY TotalSales DESC
19 LIMIT 15;
20 '''
21
22 # Execute Query and Display Results
23 try:
24     print("Interactive Dashboard: Top 15 Customers by Artist and Genre")
25     start_time = time.time()
26     df_dashboard = pd.read_sql_query(query, connection)
27     end_time = time.time()
28     execution_time = round((end_time - start_time) * 1000, 2)
29     print(f"✅ Query executed successfully in {execution_time:.2f} milliseconds")
30     display(df_dashboard)
31 except Exception as e:
32     print(f"❌ Error executing dashboard query: {e}")
33
```

## 4.7 Visualizations

Interactive Sales Dashboard: Genre → Artist → Customer

