



# Relationships

MINERVA  
FRANCO

PROJECT 3  
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Tiny blobs, medium blobs, purge,  
binary identities, fakers, floaters  
and other super unserious SQL  
operators

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# I. Timeline



**DATABASE: CHINOOK**

**11 Tables, 69 Columns**

**8 Employees, 59 Customers**

**Data collected from Jan 2009 thru Dec 2013**

**24 Countries, USA has the most customers (13)**



## II. Complex Joins: Uniting Your Data

Complex joins are essential for combining information from related tables.

CTEs (Common Table Expressions) make complex joins more readable.

INNER JOIN: Returns only matching rows from both tables.

```
WITH EmployeesCTE AS (  
    SELECT  
        EmployeeId, LastName, FirstName  
    FROM employees  
),  
CustomersCTE AS (  
    SELECT  
        CustomerId, FirstName, LastName  
    FROM customers  
)  
SELECT  
    ec.EmployeeId, ec.LastName, ec.FirstName,  
    cc.CustomerId, cc.FirstName, cc.LastName  
FROM CustomersCTE AS cc  
INNER JOIN EmployeesCTE AS ec  
ON cc.SupportRepId = ec.EmployeeId  
;
```

WITH

CTE name

CTE body

CTE usage

Inner Join

# III. Predictive Models: Looking Ahead

```
# # Execute the SQL query and load data into a DataFrame
query = """
SELECT
    t.TrackId,
    t.Name AS TrackName,
    g.Name AS Genre,
    mt.Name AS MediaType,
    SUM(ii.Quantity) AS TotalSales
FROM
    tracks AS t
JOIN
    genres AS g ON t.GenreId = g.GenreId
JOIN
    media_types AS mt ON t.MediaTypeId = mt.MediaTypeId
JOIN
    invoice_items AS ii ON t.TrackId = ii.TrackId
GROUP BY
    t.TrackId, g.Name, mt.Name
ORDER BY
    TotalSales DESC
;
"""
df = pd.read_sql_query(query, conn)
conn.close()
```

**Data Extraction:** Used 3 joins to gather data from 4 tables.

**Model:** Random Forest Regressor does not require one-hot encoding categorical variables (genre, media type)

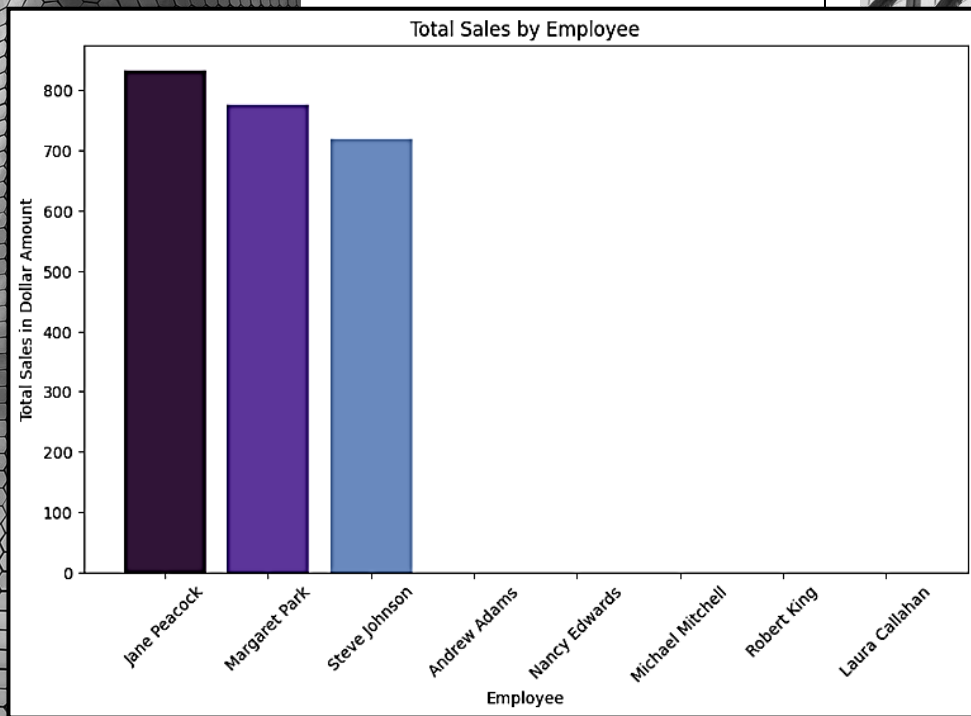
**Business need:** Predict which genres and media types will sell best to plan next album launches.

**Question:** How do I access data on genres and media types & determine if features are predictive?

**Outcome:** The model's performance is evaluated using Mean Squared Error (MSE)=0.14 indicating genre & media type can be good indicators of sales.

## IV. Employee Performance Analysis:

All NPCs , no squad.



**Business Need:** Understanding which employees generate the most sales to recognize top performers and inform training / incentive programs.

**Question:** How do I calculate total sales for employees?

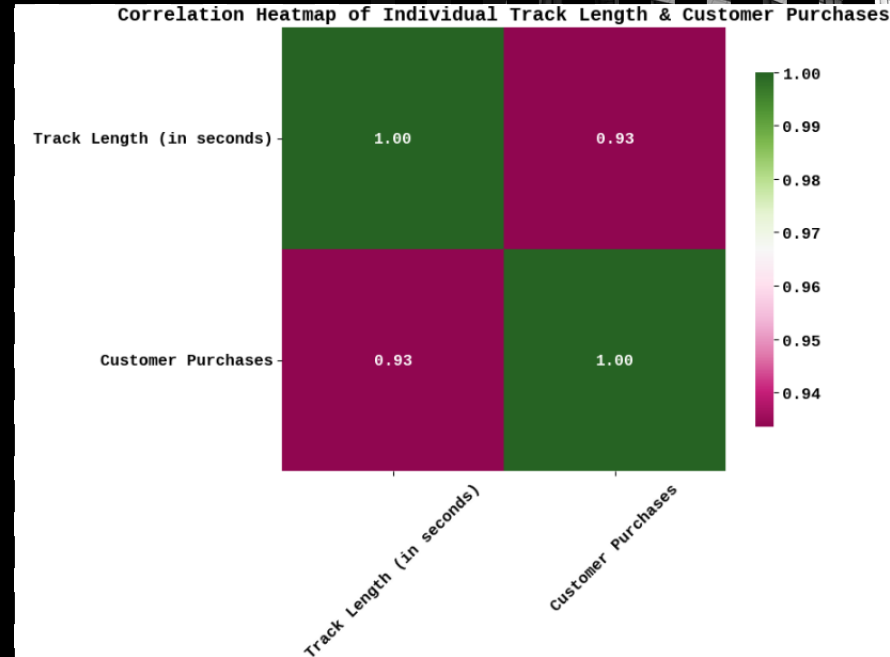
**Method:** Left joined 3 tables to aggregate sales data by employee, showing how much each employee has sold based on the invoices associated with customers they support.



# V: Track Length Analysis: Correlation Station

**Business need:** Analyze track length preferences to understand customer engagement and guide music production strategies

**Question:** How do I retrieve a list of tracks along with the distribution of track length across different music genres?





Thank you for listening

Questions?







# Credits

<https://www.sqlitetutorial.net/wp-content/uploads/2018/03/chinook.zip>

<https://www.collidu.com/presentation-sql-joins>

**Follow me for more dating tips.**

## II. Complex Joins: Uniting Your Data

```
8 %%script sqlite3 --column --header chinook.db
9 WITH EmployeesCTE AS (
10     SELECT
11         EmployeeId, LastName, FirstName, Title, ReportsTo
12     FROM employees
13 ),
14 CustomersCTE AS (
15     SELECT
16         CustomerId, FirstName, LastName, SupportRepId
17     FROM customers
18 )
19 SELECT
20     ec.EmployeeId, ec.LastName, ec.FirstName, ec.Title, ec.ReportsTo,
21     cc.CustomerId, cc.FirstName, cc.LastName
22 FROM CustomersCTE AS cc
23 INNER JOIN EmployeesCTE AS ec
24 ON cc.SupportRepId = ec.EmployeeId
25 ;
```

## III. Predictive Models: Looking Ahead

```
# # Execute the SQL query and load data into a DataFrame
query = """
SELECT
    t.TrackId,
    t.Name AS TrackName,
    g.Name AS Genre,
    mt.Name AS MediaType,
    SUM(ii.Quantity) AS TotalSales
FROM
    tracks AS t
JOIN
    genres AS g ON t.GenreId = g.GenreId
JOIN
    media_types AS mt ON t.MediaTypeId = mt.MediaTypeId
JOIN
    invoice_items AS ii ON t.TrackId = ii.TrackId
GROUP BY
    t.TrackId, g.Name, mt.Name
ORDER BY
    TotalSales DESC
;
"""

df = pd.read_sql_query(query, conn)
conn.close()
```

#### IV. Employee Performance Analysis:

All NPCs , no squad.

```
query = """
SELECT
    Employees.EmployeeId,
    Employees.FirstName,
    Employees.LastName,
    SUM(invoices.Total) AS TotalSales

FROM Employees
LEFT JOIN customers ON EmployeeId = customers.CustomerId
LEFT JOIN invoices ON customers.CustomerId = invoices.CustomerId
GROUP BY Employees.EmployeeId, Employees.FirstName, Employees.LastName
ORDER BY TotalSales DESC;
"""

# Execute query and store results in a DataFrame
df = pd.read_sql_query(query, conn)
```

#### V: Track Length Analysis: Correlation Station

```
# Query to get track lengths and invoice item costs
query = '''
SELECT
    t.TrackId,
    t.Milliseconds / 1000.0 AS TrackLengthSeconds,
    ii.UnitPrice AS InvoiceItemCost
FROM
    tracks AS t
JOIN
    invoice_items AS ii ON t.TrackId = ii.TrackId
JOIN
    invoices AS i ON ii.InvoiceId = i.InvoiceId;
'''

# Load the results into a DataFrame
df_tracks = pd.read_sql_query(query, conn)

# Close the database connection
conn.close()

# Calculate the correlation between Track Length and Invoice Item Cost
correlation = df_tracks['TrackLengthSeconds'].corr(df_tracks['InvoiceItemCost'])
print("Correlation between Track Length (in seconds) and Invoice Item Cost:",
      correlation)
```

## SQL JOINS



SELECT \* FROM  
TableA a **LEFT JOIN**  
TableB b ON a.Key =  
b.Key



SELECT \* FROM  
TableA a **RIGHT JOIN**  
TableB b ON a.Key =  
b.Key



SELECT \* FROM  
TableA a **FULL  
OUTER JOIN** TableB  
b ON a.Key = b.Key



SELECT \* FROM  
TableA a **INNER JOIN**  
TableB b ON a.Key =  
b.Key



SELECT \* FROM  
TableA a **LEFT JOIN**  
TableB b ON a.Key =  
b.Key WHERE b.Key  
IS NULL



SELECT \* FROM  
TableA a **RIGHT JOIN**  
TableB b ON a.Key =  
b.Key WHERE a.Key IS  
NULL



SELECT \* FROM TableA a  
**FULL OUTER JOIN** TableB  
ON a.Key = b.Key WHERE  
a.Key IS NULL OR b.Key IS  
NULL

# Meet Our Team



NAME SURNAME  
Title



NAME SURNAME  
Title



NAME SURNAME  
Title