

# A Smart Solution for Tracking, Monitoring, and Visualizing Bar Lines

DS 5110: Introduction to Data Management and Processing

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# Introduction

## Objectives and Goals

- Simulate data and implement an intuitive database.
- Create a robust and stable back-end for a potential mobile app.
- Mathematically aggregate data on determined intervals.
- Design a mock-up of useful user-interfaces with visualizations.

## Project Scope

- Provide users with real-time line estimates.
- Design a friendly interface.
- Provide bar and restaurant owners with analytics to forecast attendance and sales.
- Ensure the solution is scalable to expand to various types of establishments.

# Literature Review



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- Aggregates data from users who share their phone location.
- Requires multiple searches to find wait times for various bars.

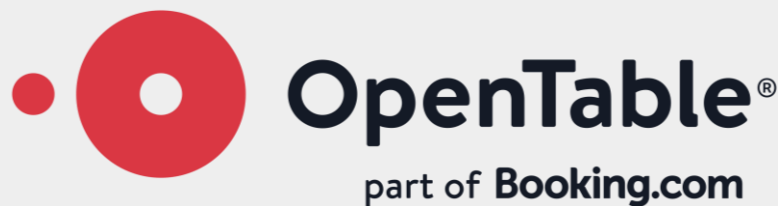
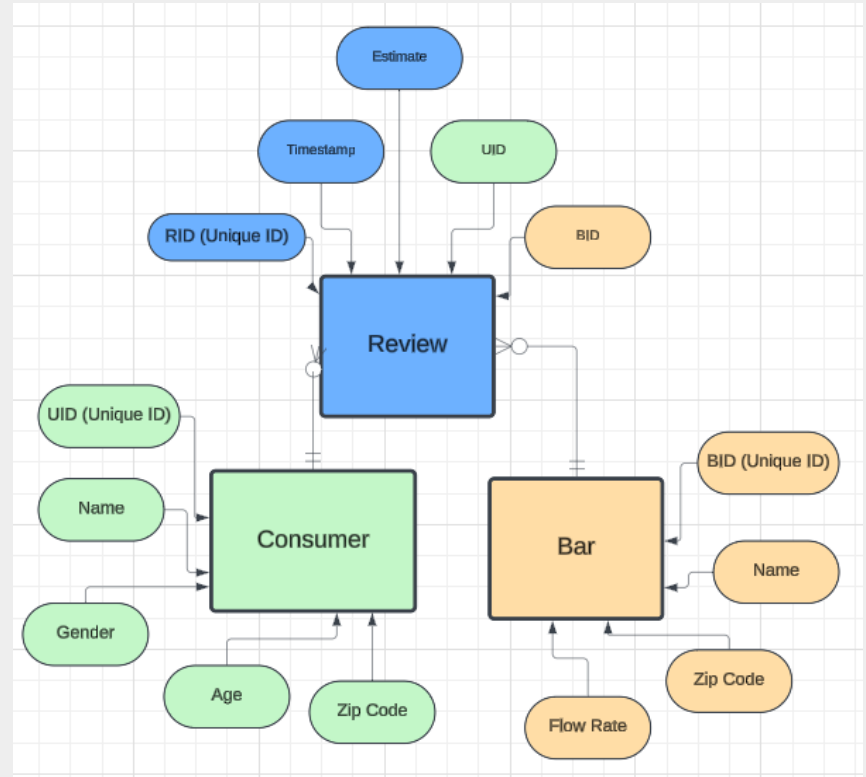


Image courtesy of OpenTable.com

- Primarily designed for restaurant owners to manage queues.
- Limited insights for patrons.

# Methodology

- Design a normalized relational database (3NF).
- Simulate data using Python for our 3 entities: Users, Bars, and 'Reviews'.
- Create a MySQL server to store data.
- Build pipelines for stable data flow.
- Aggregate and visualize data to create reports.



# Methodology contd.

- **Language/packages:**

Python, Pandas, MySQL, sqlalchemy, matplotlib, plotly, ipywidgets

- **Platforms:** JupyterLab and Google Colab (IDE), MySQL Server (storage), phpMyAdmin (DB interface)

```
# Create tables per our ERD
try:
    connection = mysql.connector.connect(
        host='sql5.freemysqlhosting.net',
        user='sql5743967',
        password='FsUwn8fAFT',
        database='sql5743967'
    )

    if connection.is_connected():
        cursor = connection.cursor()

        cursor.execute(create_table_query)
        print("Table created successfully")

except Error as e:
    print("Error:", e)

Table created successfully
```

```
create_table_query = """
CREATE TABLE Bars (
    BID INT PRIMARY KEY,
    Name TEXT NOT NULL,
    Zip_code INT,
    Flow_rate INT
);

CREATE TABLE Users (
    UID INT PRIMARY KEY,
    Name TEXT NOT NULL,
    Gender INT CHECK (Role in (0, 1, 2)),
    Age INT,
    Zip_code INT
);

CREATE TABLE Reviews (
    RID INT PRIMARY KEY,
    Timestamp DATETIME NOT NULL,
    UID INT,
    BID INT,
    Estimate INT,
    FOREIGN KEY (UID) REFERENCES Users(UID),
    FOREIGN KEY (BID) REFERENCES Bars(BID)
);
"""
```

				BID	Name	Zip_code	Flow_rate
<input type="checkbox"/>				563	Harbor & Hops	2127	29
<input type="checkbox"/>				564	Beacon Brews	2127	7
<input type="checkbox"/>				565	Fenway Tavern	2127	13
<input type="checkbox"/>				566	The Green Line Pub	2127	8
<input type="checkbox"/>				567	Back Bay Barrel	2127	40
<input type="checkbox"/>				568	Southie Spirits	2127	34
<input type="checkbox"/>				569	The Freedom Taproom	2127	36
<input type="checkbox"/>				570	Charles River Brew House	2127	40
<input type="checkbox"/>				571	Cobblestone & Co.	2127	22
<input type="checkbox"/>				572	The Bostonian Barrel	2127	31

# Time Complexity Analysis

- Data Loading into and Data Querying from MySQL – full tables
  - $O(n*m)$
  - ( $n$  = rows,  $m$  = columns) Virtually linear time, as we are appending into an existing database and there is only an insertion operation into a tabular format.
- Subquerying Data from MySQL – querying based on criteria
  - $O(\log(n) + r*m)$
  - ( $n$  = rows,  $r$  = matching rows,  $m$  = columns) Log OR linear time, depending on the number of matching rows. MySQL indexes data using binary trees and therefore reduces the search to  $\log(n)$  compared to a non-indexed database that would take  $O(n)$  to search. It then costs  $r*m$  to retrieve the data, where  $r$  is the matching rows and  $m$  is the columns.
- Aggregations – analysis of data
  - $O(\log(n) + r)$
  - ( $n$  = rows,  $r$  = matching rows) This is greater than log time, but less than linear time. Time complexity approaches linear as matching rows increases.  $\log(n)$  to search for matching values, then 1 operation per matching row, or  $r$ .

# Analysis and Results

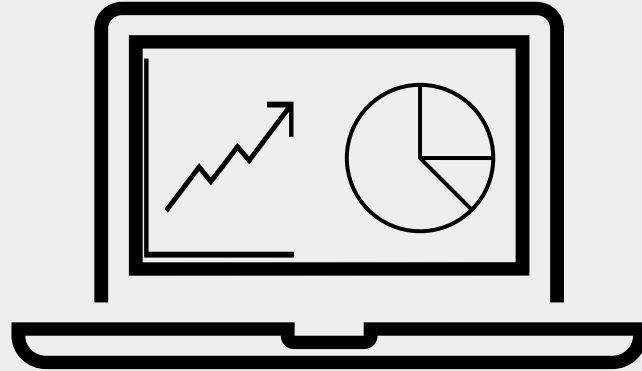
Our Stakeholders Include:

## Patrons



Patrons can access a mobile application to report and view lines

## Bar Owners

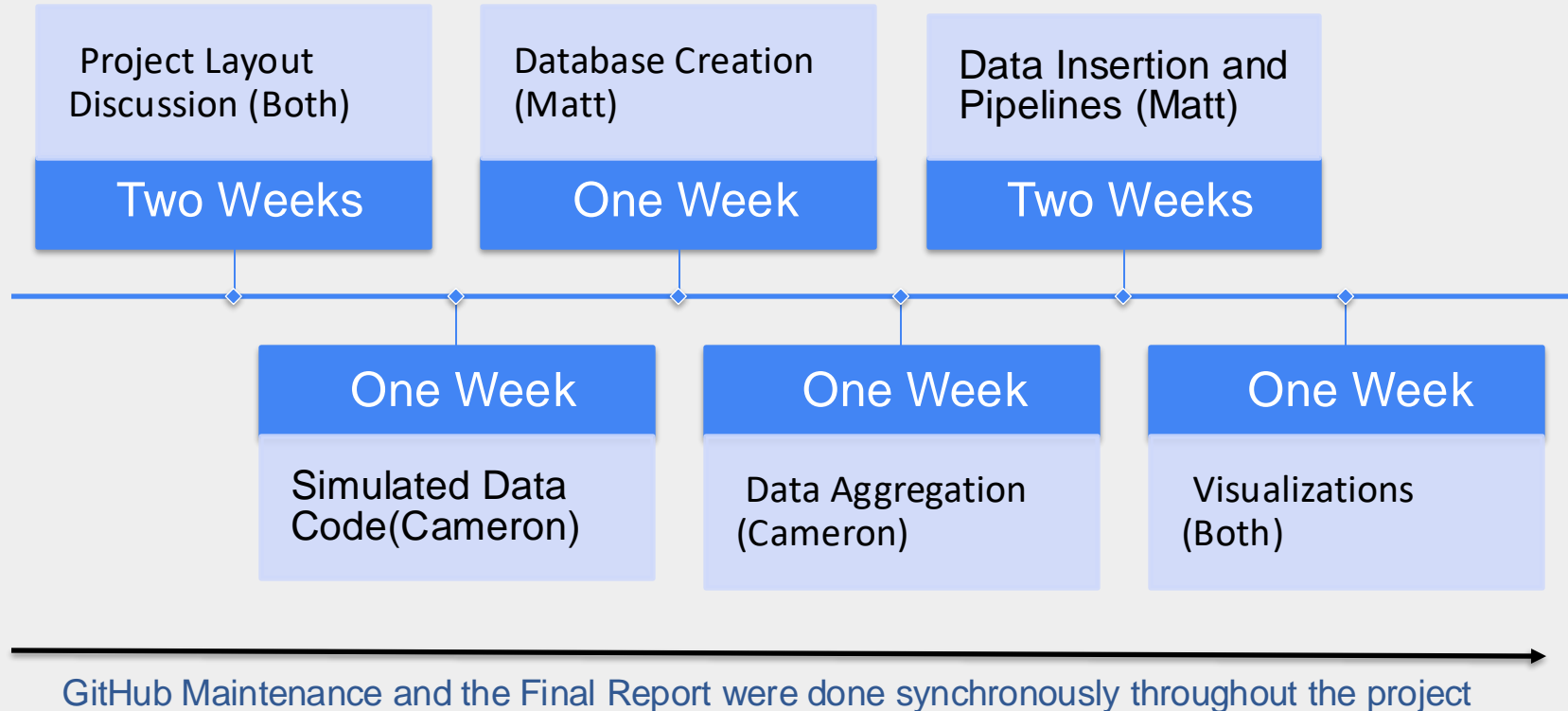


Owners can access trends to enhance their business



# **Code Walkthrough and Demo**

# Time and Responsibilities Breakdown



# Discussion

## Implications of Findings

- Even seemingly simple mobile apps require complex back-end development.
- More "Reviews" per time period smooths averages and reduces dramatic outliers.
- Increasing data size slows down aggregations and calculations.
- Several features could be added to enhance the main project goals.

## Future Work

- Develop an actual mobile app front-end for ease of access and usability.
- Include an estimate from the Bar to augment the accuracy of user-reported line data.
- Modify our aggregation to weight older estimates lower than more recent ones.

# Conclusion

## Streamlined Processes

- A real-time queuing app can improve the bar selection process.

## Sales Growth

- Bar owners benefit from forecasting customer demographics and trends.

## Broad Applications

- The app has potential applications beyond the beverage industry.

# References

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<https://www.geeksforgeeks.org/python-dictionary/>

Google. (n.d.). *About popular times, Wait Times & Visit Duration Data*. Google Business Profile Help.

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*Restaurants and restaurant bookings*. OpenTable. (n.d.). <https://www.opentable.com/>

# Question and Answer

