



EARLY MORNING DELIVERY SUBSCRIPTION

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DMA Group 5



IE6700 Data Management for Analytics Final Project Report

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Submission Date: 02/18/2024

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1 Executive Summary

DawnDash, a leading early morning delivery service, recognized the need for efficient data management to maintain operational excellence and customer satisfaction. This project developed a comprehensive data management solution leveraging relational (MySQL) and NoSQL (MongoDB) databases.

The relational database schema stores data related to users, vehicles, drivers, routes, deliveries, warehouses, products, inventory, suppliers, orders, promotions, subscriptions, additional items, feedback, and payments. Appropriate constraints ensure data integrity and efficient querying.

Data was randomly generated through Python to facilitate future testing and ensure consistency across tables using queries. The data was then implemented in MySQL using DDL statements.

SQL queries were developed for various operations, including simple queries, aggregate functions, joins, nested queries, correlated queries, and the use of EXISTS and ANY. These queries enable effective data retrieval, calculations, and information combination from multiple tables.

A NoSQL solution was implemented in MongoDB for large-scale data processing, real-time analytics, and performance monitoring.

Finally, an application demo was created using Python, establishing a connection between the databases and generating charts (histogram, pie chart, scatter plot, and box plot) for data visualization and analysis.

The combined relational models and databases, along with the Python application, provide DawnDash with a robust, scalable data management solution, ensuring efficient operations, accurate data processing, and valuable insights for continuous improvement in the early morning delivery market.

2 Introduction

In the fast-paced world of early morning transportation, businesses and individuals alike require a reliable and efficient delivery service to meet their urgent needs before daybreak. DawnDash steps in to fill this crucial gap, specializing in swift and dependable deliveries during the early hours. However, with the increasing demand for such services comes the challenge of managing vast amounts of data efficiently.

Theory for Early Morning Delivery:

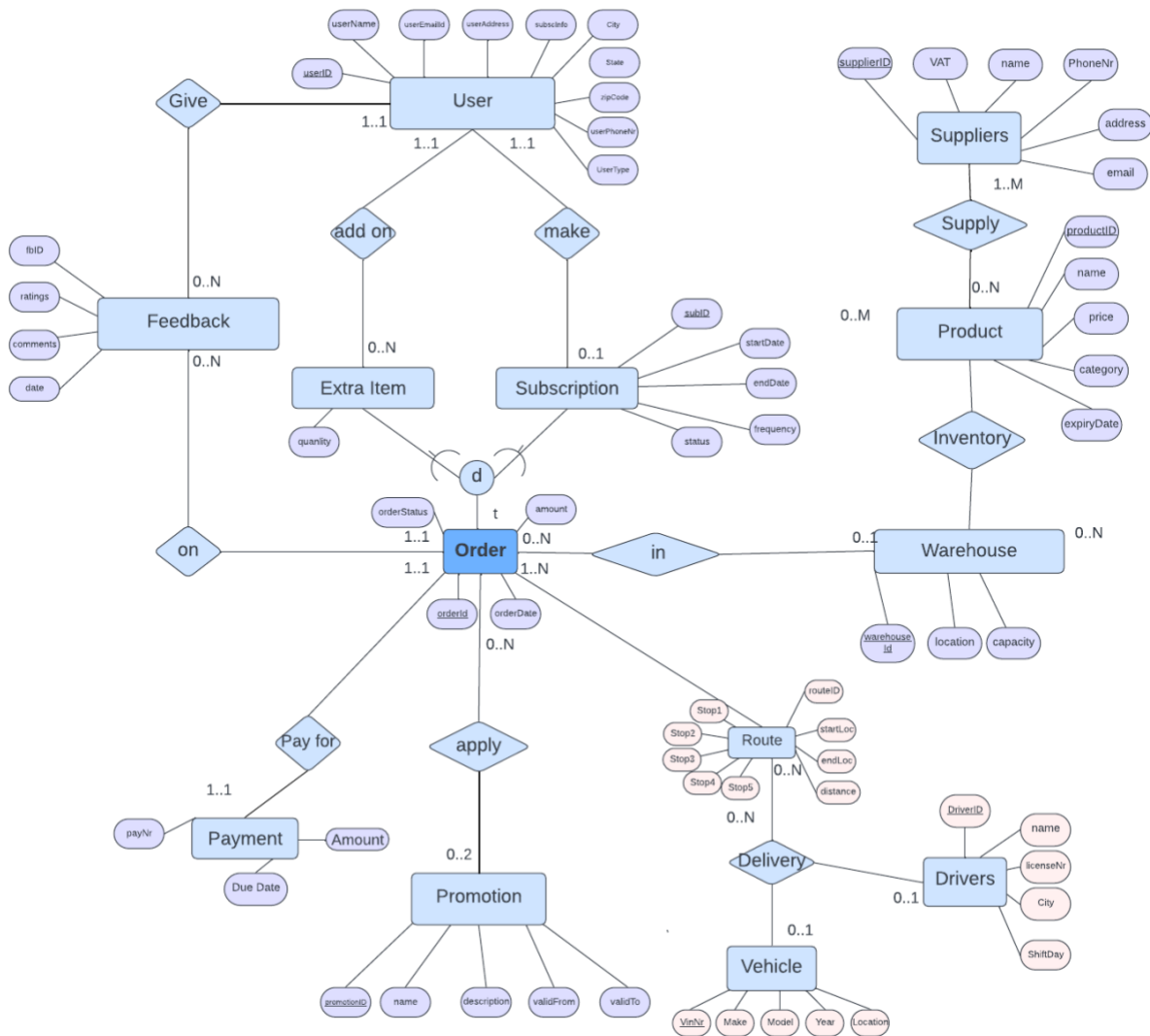
- **Enhanced Customer Experience:** Create a user-friendly platform where customers can subscribe to regular deliveries while having the flexibility to add extra items as per their needs.
- **Efficient Inventory Management:** Implement a robust data management system to track real-time inventory, ensuring accurate availability information and preventing order discrepancies.
- **Optimized Order Processing:** Develop an efficient order processing system that incorporates user preferences, processes orders in a timely manner, and schedules deliveries for the next morning.
- **Effective Communication:** Implement a notification system to keep users informed about order confirmations, delivery status, and reminders to add extra items before the cutoff time.

The business problems we are going to solve by build the database management system:

1. **Real-Time Order Processing:** Implement a system capable of processing delivery orders in real-time, ensuring prompt handling of customer requests as they come in.
2. **Optimized Route Planning:** Develop algorithms to optimize delivery routes based on factors such as distance, traffic conditions, and delivery deadlines, enabling efficient utilization of resources and timely deliveries.
3. **Customer Information Management:** Establish a secure database to store and manage customer information, including delivery preferences, contact details, and payment preferences, to streamline the ordering process and personalize the customer experience.
4. **Inventory Management:** Implement a system to track inventory levels of goods, food items, and groceries, ensuring adequate stock availability and preventing stockouts.
5. **Performance Monitoring:** Develop tools to monitor key performance metrics such as delivery times, customer's feedback/ratings, and driver efficiency, enabling continuous improvement and operational excellence.

3 Conceptual Data Modeling

3.1 EER Diagram



3.2 UML Diagram



4 Mapping Conceptual Model to Relational Model

Note: Primary Keys are underlined, and Foreign Keys are italic.

USER_DATA (UserID, userName, userEmailID, userAddress, City, State, zipCode, PhoneNr, UserType)

VEHICLE (VinNr, Make, Model, Year, Location)

DRIVERS (DriverID, name, LicenseNr, City, ShiftDay)

ROUTE (RouteID, StartLoc, EndLoc, Stop1, Stop2, Stop3, Stop4, Stop5, Distance)

DELIVERY (*RouteID*, *VinNr*, *DriverID*)

- *RouteID* foreign key refers to RouteID in ROUTE, NULL NOT ALLOWED
- *VinNr* foreign key refers to VinNr in VEHICLE, NULL NOT ALLOWED
- *DriverID* foreign key refers to DriverID in DRIVER, NULL NOT ALLOWED

WAREHOUSE (WarehosueID, location, capacity)

PRODUCT (ProductID, name, price, category, expireDate)

INVENTORY (*WarehosueID*, *ProductID*)

- *WarehosueID* foreign key refers to WarehosueID in WAREHOUSE, NULL NOT ALLOWED
- *ProductID* foreign key refers to ProductID in PRODUCT, NULL NOT ALLOWED

SUPPLIERS (SupplierID, VAT, name, PhoneNr, Address, Email)

SUPPLY (*SupplierID*, *ProductID*)

- *SupplierID* foreign key refers to SupplierID in SUPPLIER, NULL ALLOWED
- *ProductID* foreign key refers to ProductID in PRODUCT, NULL ALLOWED

ORDERS (OrderID, *ProductID*, Amount, OrderStatus, OrderDate, *RouteID*, *WarehosueID*, *PromotionID*, *UserID*)

- *RouteID* foreign key refers to RouteID in ROUTE, NULL NOT ALLOWED
- *WarehosueID* foreign key refers to WarehosueID in WAREHOUSE, NULL NOT ALLOWED
- *ProductID* foreign key refers to ProductID in PRODUCT
- *PromotionID* foreign key refers to PromotionID in ALL_PROMOTIONS

ALL_PROMOTION (PromotionID, name, description, VaildFrom, VaildTo)

SUBSCRIPTION (SubID, startDate, endDate, frequency, status, *UserID* , *OrderID*)

- *UserID* foreign key refers to UserID in USER, NULL NOT ALLOWED
- *OrderID* foreign key refers to OrderID in ORDERS, NULL NOT ALLOWED

EXTRAITEM (*UserID*, *OrderID*, *productID*, quantity)

- *UserID* foreign key refers to UserID in USER_DATA, NULL NOT ALLOWED
- *OrderID* foreign key refers to OrderID in ORDERS, NULL NOT ALLOWED
- *ProductID* foreign key refers to ProductID in PRODUCT, NULL NOT ALLOWED

FEEDBACK (fbID, rating, comments, date, *UserID*, *OrderID*)

- *UserID* foreign key refers to UserID in USER_DATA, NULL NOT ALLOWED
- *OrderID* foreign key refers to OrderID in ORDERS, NULL NOT ALLOWED

PAYMENT (PayNr, DueDate, amount, *OrderID*)

- *OrderID* foreign key refers to OrderID in ORDERS, NULL NOT ALLOWED

5 Implementation of Relation Model via MySQL & NoSQL

5.1 MySQL Implementation

5.1.1 Simple Query

1. Find the name, ID, city and state of user from USER_DATA.

```
select userName, userID, City, State  
from user_data;
```

| userName | userID | City | State |
|-----------------|--------|-------------|-------|
| Kelli Bowman | F60638 | New York | NY |
| Julie Bell | B79933 | Los Angeles | CA |
| April Baker | B19439 | Boston | MA |
| Kevin Kramer | F44564 | Boston | MA |
| Angelica Rowe | F48701 | Los Angeles | CA |
| Jacob Pineda | B45710 | Boston | MA |
| Shannon Combs | B15153 | Boston | MA |
| Kenneth Randall | B94401 | New York | NY |
| Dennis Frye | B84602 | New York | NY |
| Matthew Perez | F52104 | Boston | MA |
| Wyatt Bullock | F48704 | New York | NY |
| Brandon Solis | F62140 | Los Angeles | CA |
| John Hendricks | F23522 | New York | NY |

5.1.2 Aggregate Functions

2. Find the total amount of all orders.

```
select sum(Amount) as TotalAmount  
from Orders;
```

| TotalAmount |
|-------------|
| 5596 |

3. Find the average of rating score as AverageRating from feedback

```
select avg(rating) as AverageRating from feedback
```

| AverageRating |
|---------------|
| 2.9950 |

4. Find the Max and Min Price of Products.

select max(price) as MaxPrice,

min(price) as Minprice

from products;

| MaxPrice | Minprice |
|----------|----------|
| 19.94 | 0.52 |

5.1.3 Join Tabela

5. Retrieve a list of all deliveries, including information about the drivers assigned to them

select d.routeId, d.vinnr, d.driverid,

dr.name as DriverName,

dr.licensenr,

dr.city as DriverCity, dr.shiftDay

from delivery d

right join drivers dr

on d.driverid = dr.driverid;

| routeId | vinnr | driverid | DriverName | licensenr | DriverCity | shiftDay |
|---------|------------|----------|-----------------|-----------|-----------------|----------|
| R867 | AQU43894T2 | D9232 | Jessica Burton | 462124470 | New York, NY | Sat |
| R997 | ADQ75725V0 | D9059 | Chase Bishop | 890753334 | New York, NY | Tue |
| R610 | RVU11580A7 | D5971 | Wesley Griffin | 671284320 | Boston, MA | Wed |
| R147 | ESN29654U1 | D6187 | Barry Aguilar | 430222484 | New York, NY | Thu |
| R927 | DKN58297J9 | D4485 | Thomas Hudson | 667723152 | New York, NY | Thu |
| R802 | TEY06042D6 | D2058 | David Waters | 366916335 | Boston, MA | Fri |
| R115 | ZAS38974T3 | D7360 | Richard Barnett | 297163084 | Boston, MA | Thu |
| R649 | RSY37523W4 | D4161 | Donald Sullivan | 621246346 | Los Angeles, CA | Sun |
| R316 | CLI94671A8 | D5796 | Julie Gibson | 219219019 | Boston, MA | Fri |
| R234 | WMX15599Y1 | D9905 | Deborah Oliver | 142918708 | New York, NY | Mon |
| R747 | WMX15599Y1 | D1426 | Jeffery Green | 116578485 | Boston, MA | Wed |
| R810 | WMX15599Y1 | D9366 | Lee Martin | 480599822 | New York, NY | Fri |
| R823 | WMX15599Y1 | D8378 | Logan Jacobs | 631781312 | Los Angeles, CA | Mon |
| R320 | WMX15599Y1 | D8212 | Shawn Frank | 789183483 | Boston, MA | Fri |

6. Use self join to find pairs of routes with similar distances.

```
select r1.routeid as Route1ID,
       r1.distance as Distance1,
       r2.routeid as route2ID,
       r2.distance as distance2
from route r1
join route r2
on r1.routeid < r2.routeid
where abs(r1.distance - r2.distance) < 5;
```

| Route 1ID | Distance1 | route2ID | distance2 |
|-----------|-----------|----------|-----------|
| R342 | 40 | R867 | 37 |
| R811 | 35 | R867 | 37 |
| R250 | 41 | R867 | 37 |
| R772 | 33 | R867 | 37 |
| R119 | 41 | R867 | 37 |
| R287 | 39 | R867 | 37 |
| R212 | 38 | R867 | 37 |
| R409 | 34 | R867 | 37 |
| R605 | 35 | R867 | 37 |
| R687 | 35 | R867 | 37 |
| R215 | 39 | R867 | 37 |
| R709 | 40 | R867 | 37 |
| R601 | 37 | R867 | 37 |
| R376 | 36 | R867 | 37 |

5.1.4 Nested Query

7. Finding routes with distance greater than the average distance.

```
select routeid, distance
from route
where distance > (select avg(distance) from route);
```

| routeid | distance |
|---------|----------|
| R997 | 100 |
| R147 | 95 |
| R927 | 73 |
| R115 | 69 |
| R316 | 77 |
| R823 | 93 |
| R320 | 66 |
| R910 | 90 |
| R944 | 82 |
| R788 | 68 |

8. Retrieve the names of drivers along with the total number of deliveries made by each driver who has received positive feedback.

```
select d.name,
       ( select count(*) from delivery
         where delivery.driverid = d.driverid
       ) as total_deliveries
from drivers d
where d.driverid in
      ( select driverid from feedback
        where rating > 3);
```

| name | total_deliveries |
|-----------------|------------------|
| Jessica Burton | 1 |
| Chase Bishop | 1 |
| Wesley Griffin | 1 |
| Barry Aguilar | 1 |
| Thomas Hudson | 1 |
| David Waters | 1 |
| Richard Barnett | 1 |
| Donald Sullivan | 1 |
| Julie Gibson | 1 |
| Deborah Oliver | 1 |
| Jeffery Green | 1 |
| Lee Martin | 1 |
| Logan Jacobs | 1 |
| Shawn Frank | 1 |

5.1.5 Correlated Query

9. Find the products supplied by each supplier.

```
select s.supplierid, s.name as suppliername,  
       (select group_concat(p.name separator ', ')  
        from products p  
        where exists  
          (select *  
           from supply  
           where supplierid = s.supplierid and productid = p.productid)  
        ) as suppliedproducts  
from suppliers s;
```

| supplierid | suppliername | suppliedproducts |
|------------|-------------------------------|---|
| S33628 | Porter, Smith and White | Cheese, Turkey, Cream, Grape, Apple, Waterm... |
| S31782 | Gonzales, Johnson and Mueller | Pasta, Pancake, Watermelon, Quinoa, Grape, B... |
| S11069 | Durham, Bush and Eaton | Milk, Rice, Buckwheat, Cereal |
| S74683 | Patel LLC | Cucumber, Toast, Ice Cream, Pineapple, Spinac... |
| S15429 | Rodriguez-Mejia | Brownie, Ice Cream, Bread, Quinoa, Rice, Yogu... |
| S70046 | Willis LLC | Cheese, Candy, Sausage, Sausage, Bagel, Bac... |
| S13641 | Farrell-Woods | Toast, Cake, Barley, Tomato, Cheese, Banana,... |
| S57630 | Stanton PLC | Butter, Brownie, Ice Cream |
| S97976 | Peterson-Smith | Banana, Couscous, Tomato, Cucumber, Potato,... |
| S73950 | Carlson PLC | Buckwheat, Quinoa, Cream, Tomato, Egg, Ice ... |
| S32269 | Schmidt-Harmon | Cream, Potato, Bread, Pasta, Butter, Bell Pepp... |
| S56070 | Thomas, Valencia and Bell | Lamb, Cheese, Chicken, Bread, Butter, Yogurt, ... |
| S20148 | Harris, Downs and Cobb | Oatmeal, Lamb, Barley, Cake, Bagel, Buckwhea... |
| S11807 | Torres LLC | Couscous, Tomato, Cucumber, Potato, Cousco... |

5.1.6 Using ALL/ANY/Exists/Not Exists

10. Retrieve Products where the product is not available in inventory.

-- No result because all the products available in inventory


```

SELECT *
FROM Orders o
WHERE NOT EXISTS (
    SELECT *
    FROM Inventory i
    WHERE i.ProductID = o.ProductID
);

```

| OrderID | ProductID | Amount | OrderStatus | OrderDate | RouteID | WarehouseID | PromotionID |
|-----------------|-----------|--------|-------------|-----------|---------|-------------|-------------|
| Toggle wrapping | | | | | | | |

11. Retrive results where a feedback is associated with a given orderID.

```

SELECT * FROM Orders
WHERE EXISTS (
    SELECT *
    FROM feedback
    WHERE feedback.OrderID = Orders.OrderID);

```

| OrderID | ProductID | Amount | OrderStatus | OrderDate | RouteID | WarehouseID | PromotionID |
|---------|-----------|--------|-------------|------------|---------|-------------|-------------|
| O00003 | P75624 | 6 | shipped | 2023-08-03 | R0405 | W120 | PRM5362 |
| O00005 | P29002 | 5 | processing | 2023-02-16 | R0995 | W910 | PRM7272 |
| O00006 | P62249 | 8 | pending | 2023-09-14 | R0099 | W344 | PRM3551 |
| O00007 | P19398 | 6 | processing | 2023-01-23 | R0023 | W393 | PRM4037 |
| O00008 | P42406 | 6 | shipped | 2023-09-03 | R0714 | W414 | PRM8921 |
| O00010 | P25614 | 7 | pending | 2023-01-17 | R0969 | W671 | PRM4826 |
| O00012 | P77631 | 3 | pending | 2023-06-28 | R0363 | W421 | PRM1861 |
| O00013 | P38180 | 6 | shipped | 2023-06-21 | R0316 | W421 | PRM1861 |
| O00014 | P20964 | 1 | pending | 2023-04-17 | R0018 | W120 | PRM1728 |
| O00015 | P60070 | 10 | shipped | 2023-05-15 | R0120 | W351 | PRM2068 |
| O00016 | P30538 | 10 | processing | 2023-03-31 | R0759 | W414 | PRM8106 |
| O00018 | P51407 | 6 | shipped | 2023-10-31 | R0053 | W393 | PRM8206 |
| O00020 | P72064 | 1 | processing | 2023-01-04 | R0035 | W604 | PRM5147 |
| O00021 | P32016 | 9 | pending | 2023-02-24 | R0277 | W181 | PRM6013 |

12. Retrieve amount is greater than any amount for the same product in the orders table.

```
SELECT *
FROM Orders o1
WHERE Amount > ANY (
    SELECT Amount
    FROM Orders o2
    WHERE o1.ProductID = o2.ProductID
);
```

| OrderID | ProductID | Amount | OrderStatus | OrderDate | RouteID | WarehouseID | PromotionID |
|---------|-----------|--------|-------------|------------|---------|-------------|-------------|
| O00001 | P60906 | 5 | shipped | 2023-10-19 | R0156 | W344 | PRM3053 |
| O00005 | P29002 | 5 | processing | 2023-02-16 | R0995 | W910 | PRM7272 |
| O00007 | P19398 | 6 | processing | 2023-01-23 | R0023 | W393 | PRM4037 |
| O00008 | P42406 | 6 | shipped | 2023-09-03 | R0714 | W414 | PRM8921 |
| O00010 | P25614 | 7 | pending | 2023-01-17 | R0969 | W671 | PRM4826 |
| O00012 | P77631 | 3 | pending | 2023-06-28 | R0363 | W421 | PRM1861 |
| O00013 | P38180 | 6 | shipped | 2023-06-21 | R0316 | W421 | PRM1861 |
| O00015 | P60070 | 10 | shipped | 2023-05-15 | R0120 | W351 | PRM2068 |
| O00016 | P30538 | 10 | processing | 2023-03-31 | R0759 | W414 | PRM8106 |
| O00018 | P51407 | 6 | shipped | 2023-10-31 | R0053 | W393 | PRM8206 |
| O00019 | P12112 | 2 | processing | 2023-10-23 | R0155 | W421 | PRM1453 |
| O00021 | P32016 | 9 | pending | 2023-02-24 | R0277 | W181 | PRM6013 |
| O00024 | P88681 | 3 | processing | 2023-12-07 | R0662 | W791 | PRM6878 |
| O00025 | P37225 | 8 | pending | 2023-12-31 | R0038 | W393 | PRM6013 |

5.1.7 Subquery in Select and From

14. Calculate the percentage of total orders that each user has placed.

-- it will be 100 since each customer has only 1 order in our data

```
SELECT UserID,
    UserName,
    (SELECT COUNT(*)
    FROM Orders
```

```

WHERE UserID = u.UserID)

/ (SELECT COUNT(*)

FROM Orders) * 100 AS PercentageOfOrders

FROM user_data u;

```

| UserID | UserName | PercentageOfOrders |
|--------|-----------------|--------------------|
| F60638 | Kelli Bowman | 100.0000 |
| B79933 | Julie Bell | 100.0000 |
| B19439 | April Baker | 100.0000 |
| F44564 | Kevin Kramer | 100.0000 |
| F48701 | Angelica Rowe | 100.0000 |
| B45710 | Jacob Pineda | 100.0000 |
| B15153 | Shannon Combs | 100.0000 |
| B94401 | Kenneth Randall | 100.0000 |
| B84602 | Dennis Frye | 100.0000 |
| F52104 | Matthew Perez | 100.0000 |
| F48704 | Wyatt Bullock | 100.0000 |
| F62140 | Brandon Solis | 100.0000 |
| F23522 | John Hendricks | 100.0000 |
| B62602 | Tiffany Ramos | 100.0000 |

5.2 NoSQL Implementation

5.2.1 Simple Query

1. Finding Drivers from Boston, MA and New York, NY

```
{"$or": [{"City": "Boston, MA"}, {"City": "New York, NY"}]}
```

```
{"DriverID": 1, "ShiftDay": 1, "LicenseNr": 1, "City": 1, "Name": 1}
```

{"\$or": [{"City": "Boston, MA"}, {"City": "New York, NY"}]}
 Explain Reset Find Options

Project {"DriverID": 1, "ShiftDay": 1, "LicenseNr": 1, "City": 1, "Name": 1}

Sort { field: -1 } or [['field', -1]]
 MaxTimeMS 60000

Collation { locale: 'simple' }
 Skip 0 Limit 0

Tell Compass what documents to find (e.g. which movies were released in 2000)
 Generate

ADD DATA EXPORT DATA UPDATE DELETE
 1 - 20 of 30

```

_id: ObjectId('661c46746d0d33371fc6cee5')
DriverID: "D9232"
Name: "Jessica Burton"
LicenseNr: 462124470
City: "New York, NY"
ShiftDay: "Sat"
  
```

```

_id: ObjectId('661c46746d0d33371fc6cee6')
DriverID: "D9059"
Name: "Chase Bishop"
LicenseNr: 890753334
City: "New York, NY"
ShiftDay: "Tue"
  
```

5.2.2 Complex Query

2. Finding the Total Amount of all Orders

```

{
  _id: "$ProductID",
  totalAmount: {
    $sum: "$Price",
  },
}
  
```

Stage 1 \$group
 Output after \$group stage (Sample of 10 documents)

```

1 {
2   _id: "$ProductID",
3   totalAmount: {
4     $sum: "$Price",
5   },
6 }
  
```

```

_id: "P33101"
totalAmount: 2.5
  
```

```

_id: "P87417"
totalAmount: 1.78
  
```

5.2.3 Aggregate

3. Finding how many users are in which city

```

{
  _id: "$City",
  totalUsers: {
    $sum: 1,
  },
}
  
```



```
}
```

The screenshot shows a MongoDB aggregation pipeline editor. On the left, a JSON document is displayed with line numbers 1 through 6. The document is:

```
1 {
2   _id: "$City",
3   totalUsers: {
4     $sum: 1,
5   },
6 }
```

 Above the document, a dropdown menu shows "Stage1" and a toggle switch is turned on. On the right, the output of the "\$group" stage is shown for a sample of 3 documents. It consists of two boxes: the first with

```
_id: "New York"
totalUsers: 61
```

 and the second with

```
_id: "Los Angeles"
totalUsers: 58
```

4. Finding the Split of Business and Family USERS

```
{
```

```
  _id: "$UserType",
```

```
  Count: {
```

```
    $sum: 1,
```

```
  },
```

```
}
```

The screenshot shows a MongoDB aggregation pipeline editor. On the left, a JSON document is displayed with line numbers 1 through 6. The document is:

```
1 {
2   _id: "$UserType",
3   count: {
4     $sum: 1,
5   },
6 }
```

 Above the document, a dropdown menu shows "Stage1" and a toggle switch is turned on. On the right, the output of the "\$group" stage is shown for a sample of 2 documents. It consists of two boxes: the first with

```
_id: "Family User"
count: 100
```

 and the second with

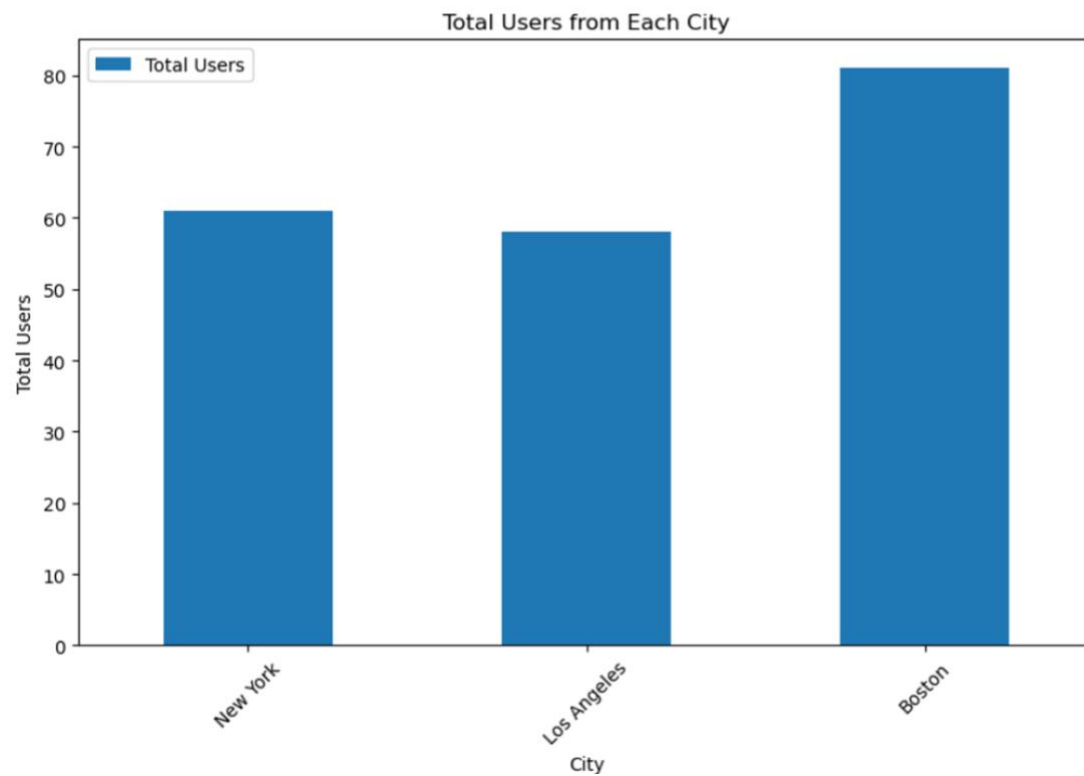
```
_id: "Business User"
count: 100
```

6 Database Access via Python

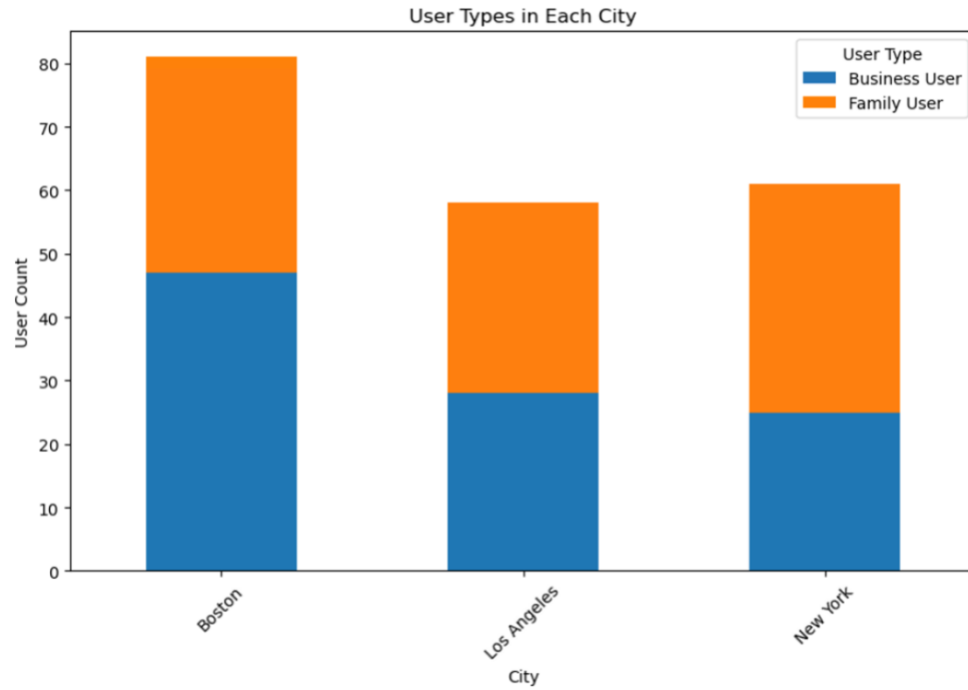
6.1 Histogram

1. Total Users from Each City

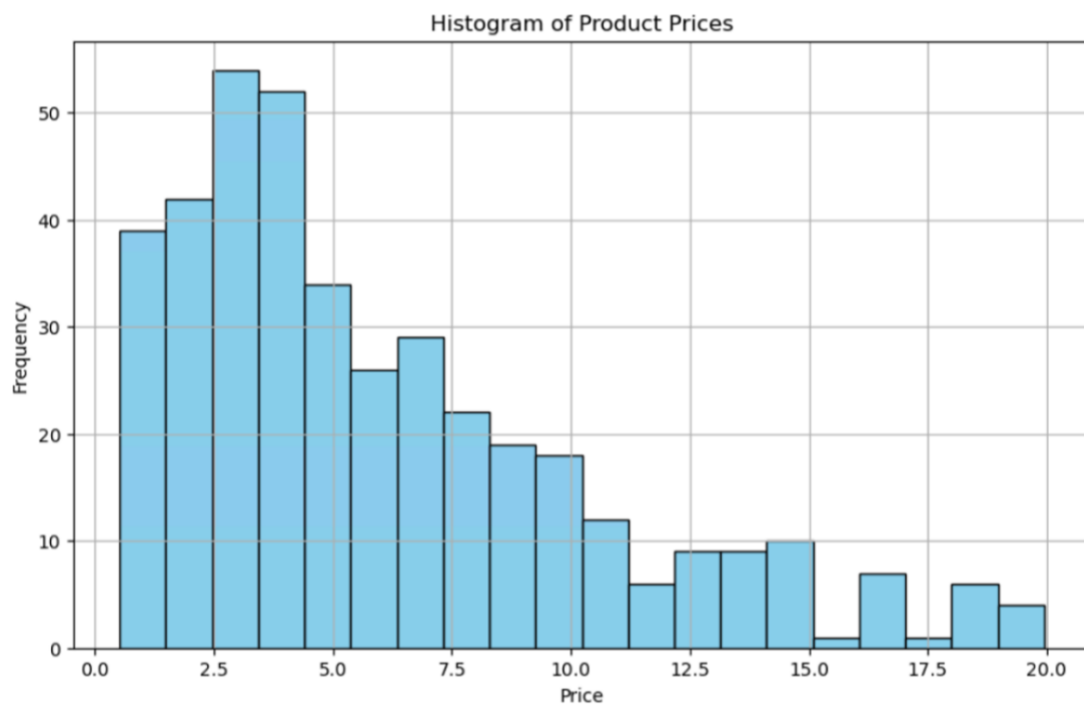
```
Connected to MySQL server
Total users from each city:
  City  Total Users
0  New York         61
1  Los Angeles      58
2    Boston         81
```



2. User Type in each city

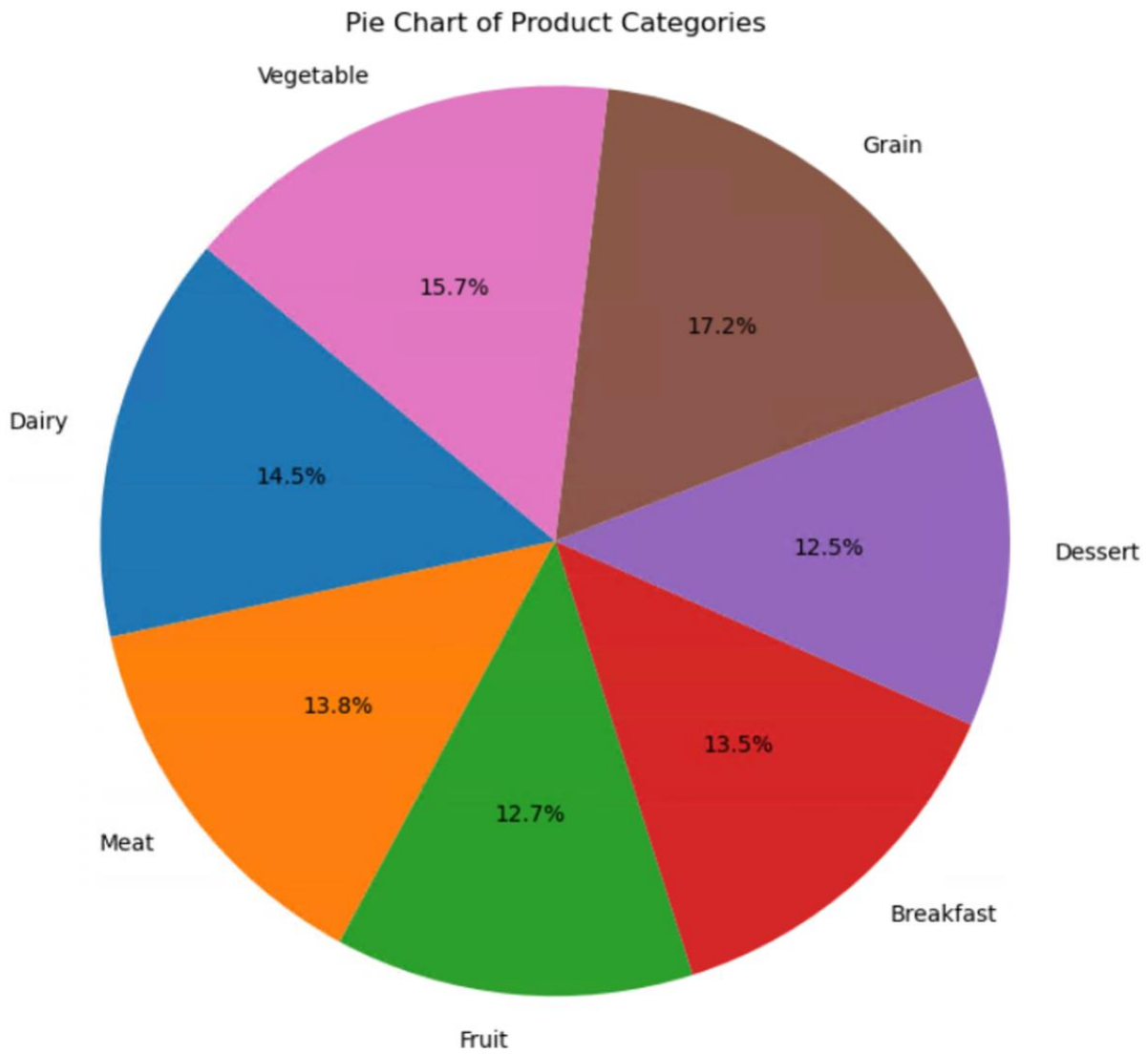


3. Histogram of Product prices



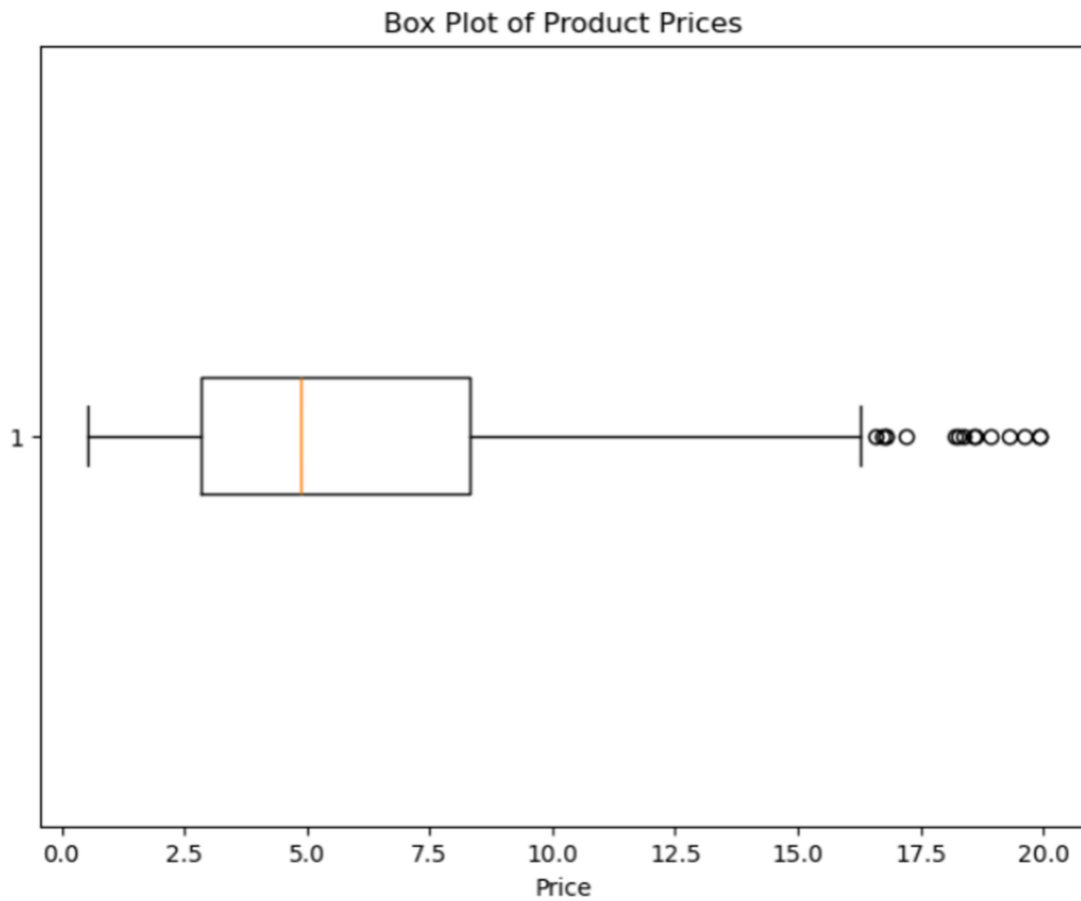
6.2 Pie Chart

4. Product Categories



6.3 Box Plot

5. Product Prices



6.4 Scatter Plot

6.Product Prices across various categories



7 Summary and Recommendation

The DawnDash project has successfully delivered a comprehensive data management system that addresses the company's diverse data storage and processing requirements. The relational database model, implemented in MySQL, provides a robust foundation for storing and managing structured data, while the NoSQL solution, implemented in MongoDB, offers scalability and flexibility for handling large-scale data processing and real-time analytics.

The combination of these technologies, along with the Python application, equips DawnDash with a powerful toolset for efficient data management, accurate data processing, and valuable insights for continuous improvement in the early morning delivery market.

Some improvement can be made by Establishing a data governance framework to ensure data quality, consistency, and accountability across the organization. This framework should encompass data policies, standards, processes, and roles for managing data assets throughout their lifecycle. Additionally, explore the integration of Neo4j, a graph database, and its query language, Cypher, to handle complex relationships and interconnected data. This can be particularly useful for route optimization, network analysis, and recommendation systems, leveraging the strengths of graph databases in representing and querying highly connected data.

Consider leveraging big data technologies like Apache Spark and Hadoop to process and analyze large volumes of data generated by DawnDash's operations. These technologies can provide scalable and distributed computing capabilities, enabling efficient batch processing, real-time stream processing, and advanced analytics on massive datasets.

By implementing these recommendations, DawnDash can further enhance the capabilities of its data management solution, unlocking new opportunities for operational efficiency, data-driven decision-making, and delivering exceptional customer experiences.