Target Customer List for Horizon Bank

Prepared by: North Star Consultants (GROUP 1)

Date: October 20, 2023

Submitted by:

Yian Chen

Sean Adam Kagugube

Emmanuella Acheampong

Susan Arzapalo

Table of Contents:

❖ Introduction

- ➤ About North Star Consultants
 - Mission Statement
 - Vision Statement
 - Our Services
- > Why Invest in us?
- > Customer Background
- ➤ Value Proposition

Phase 2: Target Customer List Development

- ➤ Minimum Deposit Qualification Criteria for New Product
 - Balance Criteria
 - Age Criteria
 - Loan Criteria
- > Justifications for the criteria used
- > Further Considerations/Analysis
- ➤ Our Solution

Reflection

1. Introduction

North Star Consultants

North Star Consultants is a leading consultancy dedicated to designing and implementing cutting-edge Database Management Systems (DBMS) for businesses across various industries. With a team of highly skilled database experts and a passion for data optimization, we specialize in helping financial institutions harness the full potential of their data resources.

Mission Statement:

"Our mission is to offer comprehensive DBMS solutions that enable financial institutions to unlock the true value of their data assets and ensure that our clients thrive in a rapidly evolving financial landscape."

Vision Statement:

"To be the trusted partner of choice for financial institutions seeking excellence in optimizing data storage, retrieval, and analysis, ensuring our clients stay ahead in today's data-driven world".

Our Services:

- **Data Quality Assurance**: We ensure your data is accurate, consistent, and reliable, eliminating errors and inaccuracies.
- **Data Transformation**: We convert raw data into actionable insights, seamlessly integrating data from multiple sources.
- **Insights for Success**: We provide client-centric insights to empower strategic planning and drive success.
- Engaging Reporting: Our visualizations and reports offer clarity and actionable recommendations.

Why Invest in us?

We have a history of over 100+ successful database implementations and satisfied clients in our 2 years of existence. Our solutions are high in demand and our expertise and commitment to excellence ensure investor confidence in our ability to execute and deliver results.

Customer Background

Horizon Bank is a prominent bank with over 50 branches across the United States. The bank wants to launch a new loan product for its loyal customers (more than 2 years with the bank). In addition, the bank is grappling with soaring marketing expenditures and an alarming drain on resources due to inefficient marketing campaigns. Horizon Bank is in dire need of a database with insights for optimizing their marketing strategies targeting the right customers, reducing expenses, and maximizing the effectiveness of their campaigns to maintain a competitive edge in the financial sector.

Value Proposition

Phase_1: To Build a relational database for managing the data set of the bank and leverage advanced data analytics and predictive modeling

Phase_2: To identify customers who qualify for the loan loyalty product for targeted marketing

Phase_3: To provide customer insights to enable the bank to develop a marketing strategy that minimizes expenses while delivering superior results.

Phase 2: Target Customer List Development

In this exciting second phase of our consultancy engagement with Horizon Bank, we are pleased to announce a pivotal project. Building upon the solid foundation of our newly implemented Database Management System (DBMS) in phase 1, Phase 2 focuses on creating a Target Customer List for a groundbreaking term deposit product set to be unveiled.

Horizon Bank's aspirations for this new product are underpinned by a robust set of requirements listed below.

Minimum Deposit Qualification Criteria for New Product:

- 1. **Balance Criteria:** Customers with a minimum monthly balance of \$5,200 or more qualify.
- 2. **Age Criteria:** Customers aged between 18 and 65 are eligible.
- 3. Loan Criteria: Customers must have existing housing or personal loans to qualify.

Justifications for the criteria used:

The selected criteria for this customer qualification process serve multiple essential purposes. Firstly, the financial stability criterion ensures that customers have the means to commit to the new product, thus minimizing financial risk. The age range criterion offers a broad and diverse customer base, optimizing the product's reach. Identifying customers with existing loans, a significant financial relationship indicator, positions them as suitable candidates for additional products. The bank is specifically interested in customers with existing loans because it already has a record of the financial records of these customers and has established a long-lasting relationship with them.

One may ask, why would the bank want to put criteria in place for customers who want to

deposit their funds with the bank. The answer is, since the new term deposit product offers a high interest rate, the bank has put in place these requirements in order to have loyal customers enjoy this high interest rate offer.

By combining these criteria related to customer loyalty, financial capability, and unique preferences, we can now present a finely curated list of potential customers who not only qualify for the new product but are also highly likely to embrace it. This data-driven approach optimizes resource allocation and enhances the prospects of a successful product launch.

Further Considerations/Analysis

- 1. Priority Status: Customers fall into three priority categories: "High_Priority," "Low_Priority," and "Disqualified." The categorization of customers into high-priority (monthly balance over \$20,000), low-priority (balances between \$5,200 and \$20,000) and disqualified (below \$5,200) segments tailors the product to customers' distinct financial profiles and this will help the bank to know which customers to prioritize during its marketing campaign.
- 2. Loan and Age Data Analysis: The average age of customers with existing loans is considered. Analyzing the average age of customers with existing loans (who are the bank's target) provides insights into their life stage and financial commitments, helping predict their inclination toward longer-term deposit options. This data aids in tailoring the new term deposit product to better meet their needs and encourages longer-term deposits.

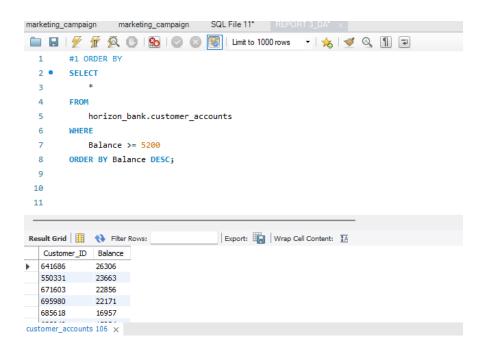
Our Solution

Our approach centers on data-driven precision. To meet these requirements, our team has conducted a comprehensive analysis, employing 28 SQL queries that carefully select individuals who align with the desired criteria.

Our approach centers on data-driven precision. To meet these requirements, our team has conducted a comprehensive analysis, employing 28 SQL queries that carefully select individuals who align with the desired criteria.

Query 1: ORDER BY

Objective: The objective of this query is to assess the balance distribution of the bank's customers with a monthly balance greater than or equal to \$5,200. The aim is to generate a pool of customers who meet the first requirement of having a minimum balance of \$5,200.



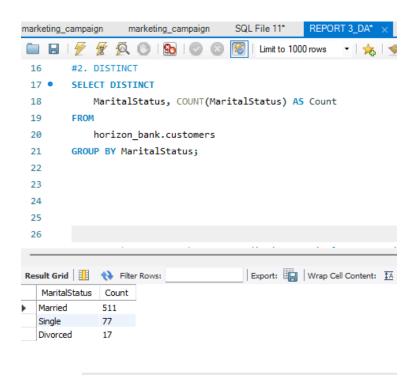
Output Report:

The query filters the dataset to include only those customers with a monthly balance greater than or equal to \$5,200 and then **orders** the balances in descending order. The ordered output reveals a balance distribution ranging between \$5,200 and \$26,306.

From the output, 47 rows were returned out of a total of 605 customers.

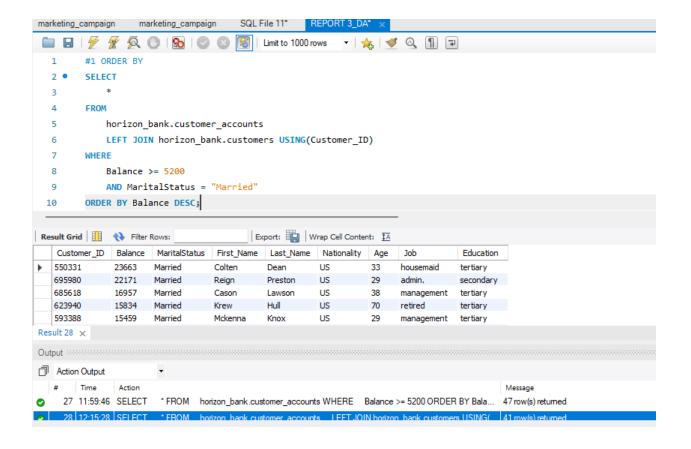
Query 2: DISTINCT

Objective: To analyze the distinct marital statuses of the bank's customers. Chances are that, married couple may want to have long term investments to support the educational needs of their children. It could also be argued that divorced or single customers may have low commitments and would therefore want to invest their surplus funds for a long term.



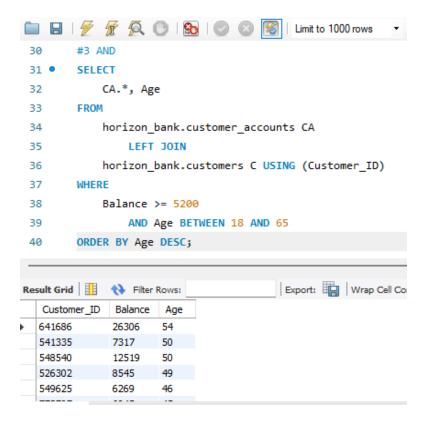
Output Report:

The query identifies the distinct marital segments within the customer dataset and provides a count of customers within each segment. The output revealed that the majority of the bank's customers are married. In light of this, we sought to get the proportion of married customers who met the first requirement (maintained a minimum balance of \$5.200 or more). The findings as revealed in the output below suggested that 41 out of the 47 customers who met the first requirement are married. Please see rows returned (highlighted in blue) from the below screenshot.



Query 3: AND

Objective: Now we moved a step further to identify the customers who meet the second requirement of being 20 years or older in addition to the first minimum balance requirement. The age factor can guide the bank in knowing the likely duration a customer may want to keep funds in term deposit. For example, aged customers may want to liquidate their funds soon to cater for their retirement needs compared to younger customers.

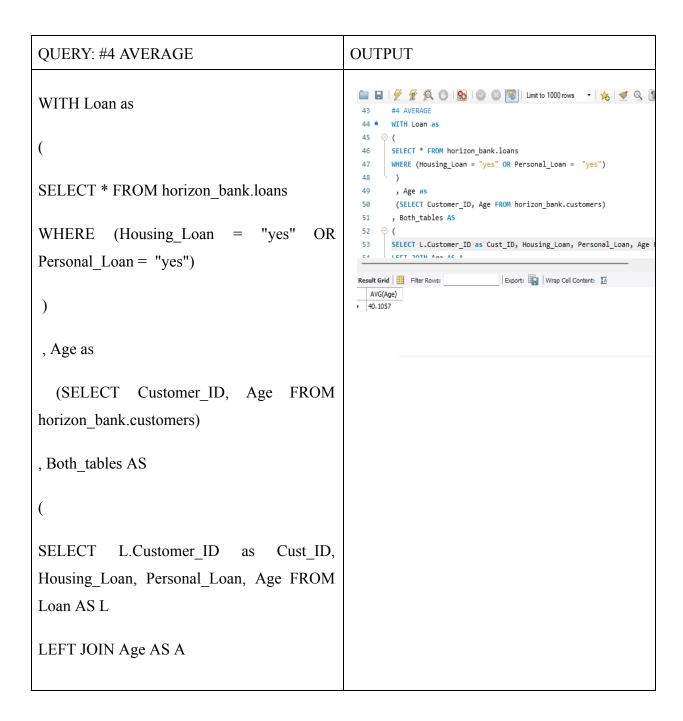


This query selects customers who satisfy two conditions: a monthly balance greater than or equal to \$5,200 and an age between 18 and 65. The results are ordered by balance in descending order.

7 out of the 47 customers were dropped as they were either less than 18 years old or more than 65 years old, leaving just 40 customers to target.

Query 4: AVERAGE

Objective: Since it is also a requirement for a customer to have either a Personal or Housing Loan, the next step seeks to calculate the average age of customers with housing or personal loans.



ON L.Customer_ID = A.Customer_ID
)
SELECT AVG(Age) FROM Both_tables;

Output Report:

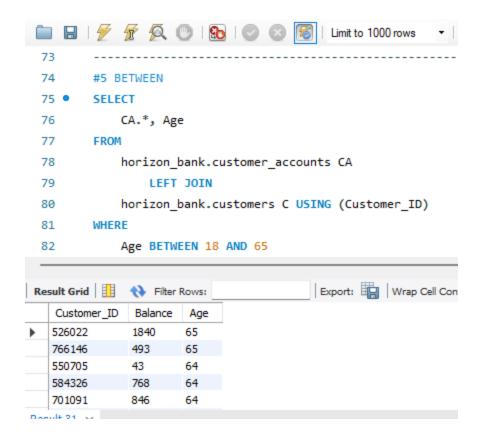
This query calculates the average age of customers who have either a housing loan or a personal loan. The first CTE identifies customers who have either loans from the loan table. The second CTE selects the ages of the customers from the customers table. The third CTE selects the customers who have either loans from the first CTE and joins it to their respective ages from the second CTE. Then finally, the query after the 3 CTEs selects the ages and performs an average function to get the average age.

The result indicates that the average age of the bank's customers who have either loans is 40 years irrespective of whether they have met the first two requirements. That is, it only focuses on finding the average age of customers who have either of the loans.

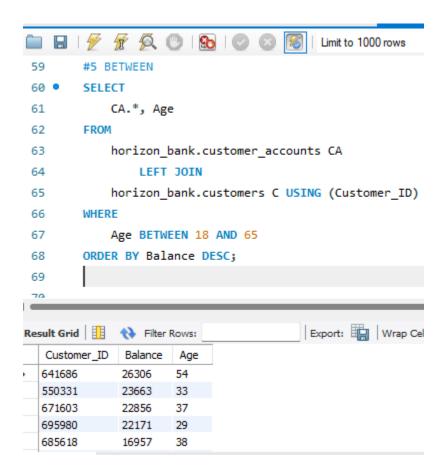
Query: 5 BETWEEN

Objective: The objective is to identify customers aged between 18 and 65 for the second requirement.

This query selects customers whose ages fall within the range of 18 to 65 years. The results are ordered by balance in descending order.



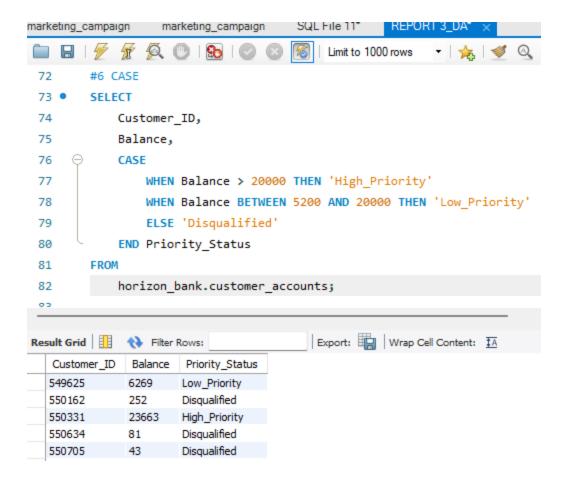
In this result above, we realized that the oldest customer is 65 years old and the youngest customer is 20 years old when ordered by age. So we asked, is age correlated with customers account balance? From the result, age is not correlated with account balance. This is because, we can see a customer who is 64 years but has a balance as low as \$43. Again, when this same output is ordered by account balance instead of age (from the screenshot below), the customer with the highest balance is less than 65 years. Still, this is interesting because one would have inferred that an old customer would be expected to have a lower balance because they may not be engaged in an active income generating job.



Query 6: CASE

Objective: Next, we aimed to categorize customers into different priority groups based on their account balances. Categorizing customers by priority status allows the bank to tailor its product offering and marketing strategies to different customer segments. This approach optimizes resource allocation and enhances the chances of success.

The query below classifies customers into three priority categories: "High_Priority" for balances over \$20,000, "Low_Priority" for balances between \$5,200 and \$20,000, and "Disqualified" for balances below \$5,200.

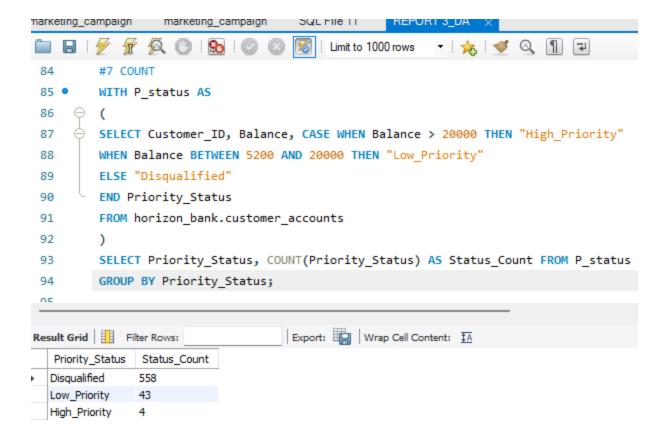


The output below shows the distribution of customers as high priority, low_priority and disqualified customers based on their account balances. Majority of the customers were disqualified as they maintained a low balance.

Query 7: COUNT

Objective: Having classified customers into various priority statuses, we were interested in knowing the number of customers in each priority status

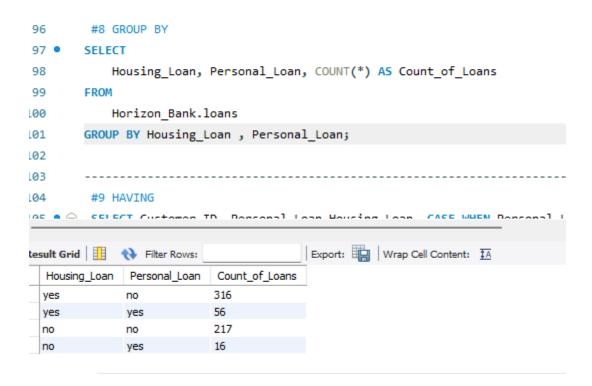
The query below builds on the CASE query above by turning it into a CTE and then writing a select statement from the CTE that counts the number of customers in each priority status category, providing insight into the distribution of customers across these categories.



The result reveals that the majority of the bank's customers fall within the "disqualified" category and only a few fall within the high priority category. However, in total, without factoring age, 47 customers (low and high priority customers) can be targeted by the bank for the new product. This affirms our result from the "order by" query above.

Query 8: GROUP BY

Objective: The objective here is to get an overview of the number of people with or without loans. Grouping and counting loans by type helps the bank assess its loan portfolio and better understand the distribution of different loan products among its customers.



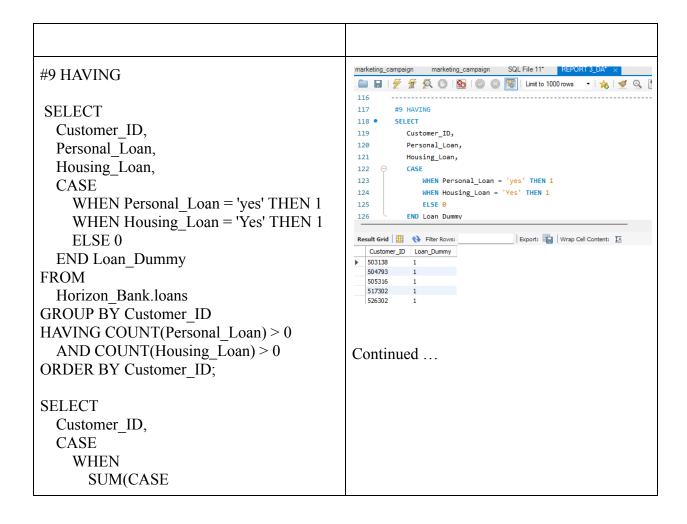
Output Report:

From the above result, we realized that the majority of the bank's customers have housing loans relative to personal loans. Also, the majority of the customers do not have any of the loans. The bank may therefore want to take advantage of this insight to increase its loan base so as to ensure that most of the customers will have loans and subsequently qualify for the new term deposit which comes with a prerequisite of having a loan.

Query 9: HAVING

Objective: Here, we aim to identify customers who have both personal and housing loans. Identifying customers with both types of loans can provide insights into customers' financial commitments. Such customers might be interested in additional financial products or services.

The query below aims to find customers who have both personal and housing loans. Because the responses to whether a customer has a loan or not is categorized as "YES" or "NO", the query first creates a "Loan_Dummy" value for each customer, where 1 indicates the presence of both loan types and 0 indicates otherwise. It then groups these responses by Customer_ID having the sum of loans greater than zero.



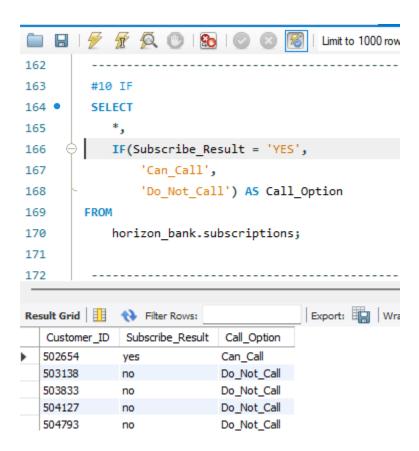
```
WHEN Personal Loan = 'yes'
THEN 1
       ELSE 0
                                         END) > 0
ORDER BY Customer_ID;
     END) > 0
       AND SUM(CASE
       WHEN Housing Loan = 'yes'
THEN 1
       ELSE 0
     END) > 0
   THEN
     1
   ELSE 0
 END AS Loan Dummy
FROM
 Horizon Bank.loans
GROUP BY Customer ID
HAVING SUM(CASE
 WHEN Personal Loan = 'yes' THEN 1
 ELSE 0
END) > 0
 AND SUM(CASE
 WHEN Housing Loan = 'yes' THEN 1
 ELSE 0
END) > 0
ORDER BY Customer_ID;
```

From the output (row returned), we realized that 56 people have both Personal loan and Housing Loan.

Query 10: IF

Objective: This section categorizes customers based on subscription status. This is because the bank may want to reach out to customers who qualify for the new product and from the dataset. Since some customers ticked no when asked to be contacted by the bank through a phone call, we aimed to generate a list of customers, separating them by their call options.

The query below categorizes customers into two groups: "Can_Call" for those with a subscription result of "YES" and "Do Not Call" for those with no.

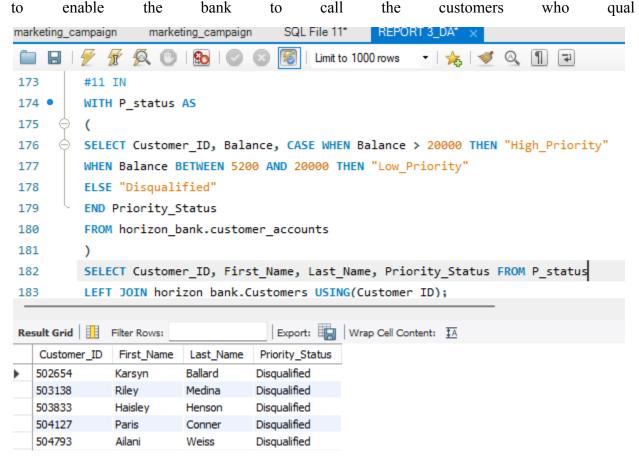


Output Report:

From the result, the majority of the customers requested not to be reached by phone. This indicates that the bank has to find alternative means of reaching out to the customers who qualify for the term deposit product but opted out of call subscription.

Query 11: IN

Objective: This query builds on query 6 and 7 (which categorized customers into various priority statuses) and goes on to identify who the customers in the various categories are. This is



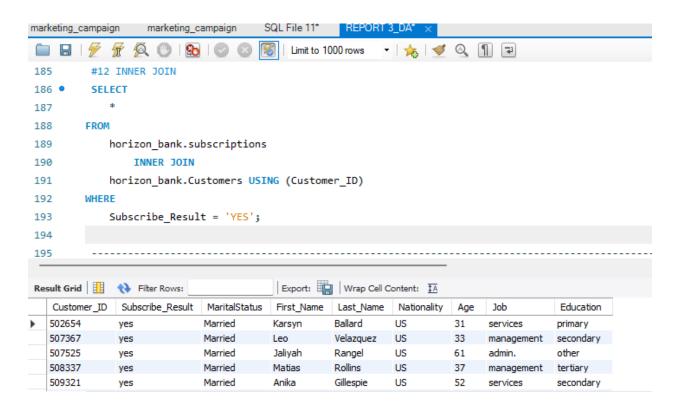
Output Report:

The output from this query is similar to the query on "COUNT" but this query rather than count the number of customers, provides the names of the customers in the various categories by combining the balance and customer tables.

Query 12: INNER JOIN

Objective: This query seeks to combine customer and subscription data to identify only customers who have subscribed "YES" for call options.

The query below performs an inner join between customer and subscription data to identify customers who have subscribed.



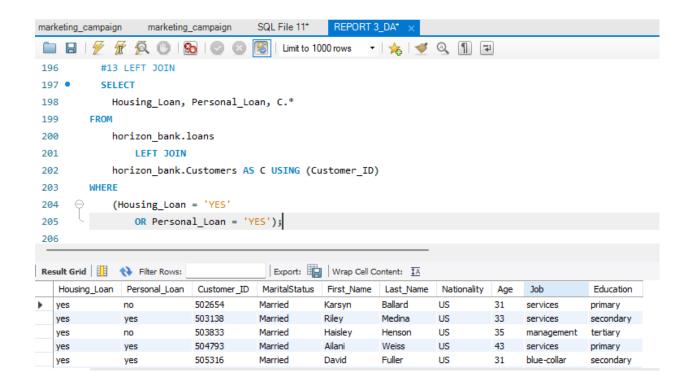
Output Report::

The result gives the bank a fair idea of the number of customers who they can call for marketing campaigns. Only 146 out of the 605 customers opted "YES" to call option.

Query 13: LEFT JOIN

Objective: Having identified the customers with loans from the loans table, this query seeks to combine customer data from the customers table with loan information to give a complete view of who those customers are..

The below query combines customer data with loan information through a left join. It identifies customers who have loans, either housing loans or personal loans.



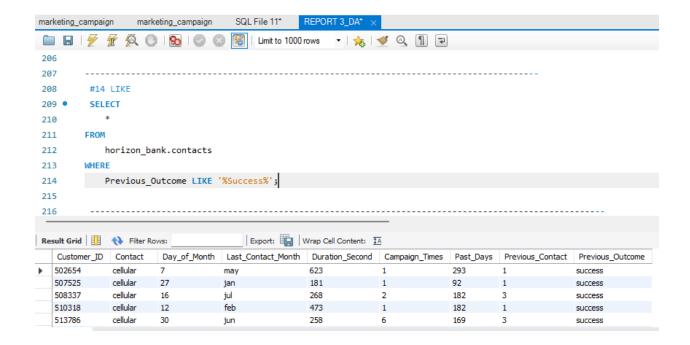
Output Report:

The result is a combination of the loans table and customers table to give details of the customers who have loans with the bank.

Query 14: LIKE

Objective: The aim is to identify customers whom the bank has been successful in contacting for marketing campaigns in the past. Identifying customers with successful contacts can provide insights into effective communication and customer engagement strategies.

The query below selects customers who have had previous contacts with the term "Success" in their outcome from the contacts table.



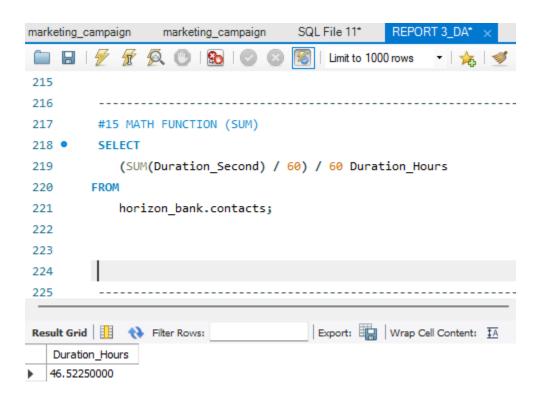
Output Report:

The result shows that only 127 customers out of 605 customers were contacted with success over the last year by the bank. This indicates that communication between the bank and its customers is not really effective.

Query 15: Math Function - SUM

Objective: Having established the number of customers who the bank contacted in the past successfully, we aim to understand the duration of contact time between the bank and its customers. Calculating the total duration of customer contacts provides insights into the volume of interactions and helps in resource allocation for customer support.

The below query calculates the total duration of customer contacts in hours by summing the "Duration_Second" field and dividing by 60 and for duration in minutes and further dividing by 60 for duration in hours.



Output Report:

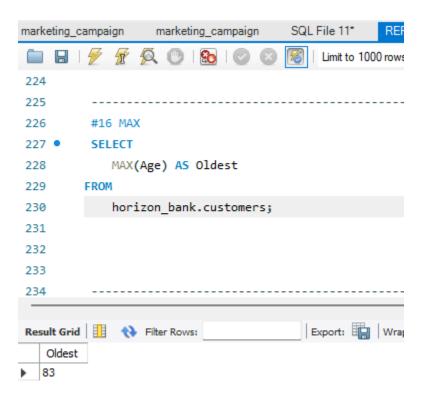
We found that the bank only spent approximately 47 hours in contacting its customers over one year. It could be that there are other means by which the bank communicates with its customers.

However, this requires further investigation.

Query 16: MAX

Objective: The aim here is to find the oldest customer's age. Knowing the age of the oldest customer can help in understanding the longevity of customer relationships and tailoring services for different age groups.

The query below identifies the oldest customer's age within the dataset by using the MAX function.



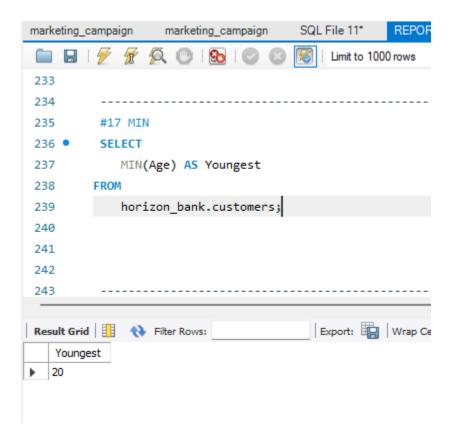
Output Report:

As seen in the results from the query, the oldest customers of the bank are aged 83 years old. However, the dataset doesn't give information about the duration of the relationship with its customers hence we cannot conclude if the customers in this age bracket have the longest relationship with the bank.

Query 17: MIN

Objective: Contrary to the above query which sought to find the age of the oldest customer, the objective here is to find the youngest customer's age. Understanding the age of the youngest customer provides insights into attracting and retaining a younger customer base.

The below query identifies the youngest customer's age within the dataset using the MIN function.



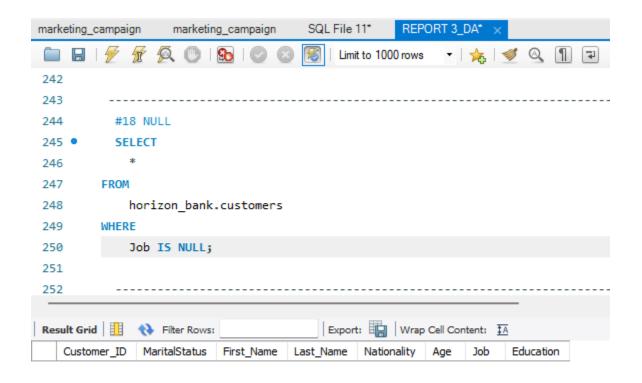
Output Report:

We found out that the youngest customer is 20 years old. This gives a fair change to all young customers to qualify for the new product as the qualifying age starts from 18 years.

Query 18: NULL

Objective: The aim here is to Identify customers with a missing job field. Identifying customers with missing job information is crucial for data completeness and improving customer profiles.

The query below selects customers whose "Job" field is NULL, indicating missing job information.



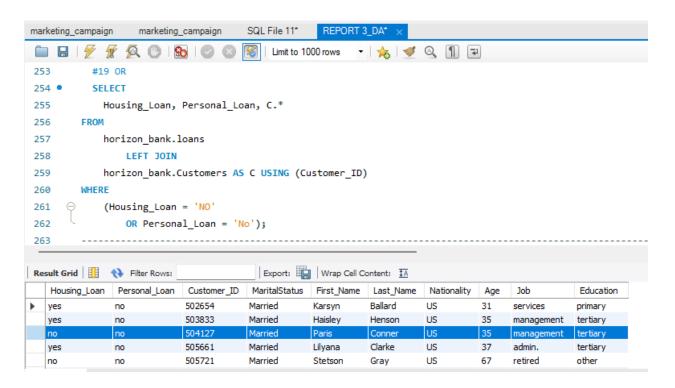
Output Report:

Evidently from the screenshot above with no output, we can conclude that all customers in the bank's database have the job field not being empty. There are no NULLS for all other fields in the database. This suggests complete customer profiles from which good insights can be drawn.

Query 19: OR

Objective: Unlike the query above that identifies customers having only one type of loan, this step identifies customers without loans.

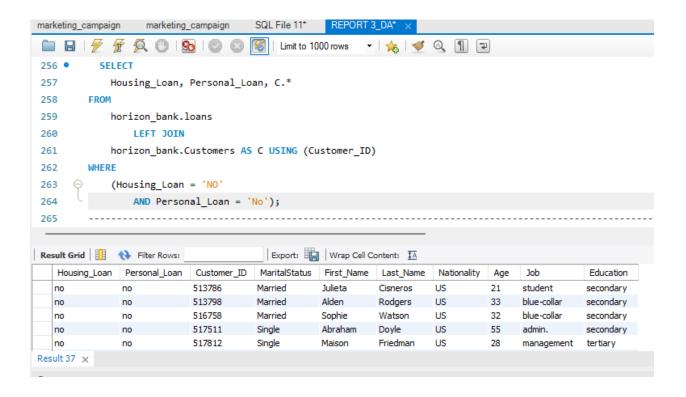
The below query combines the customers table with the loans table and identifies customers without any loans.



Output Report:

The result of this query is quite intriguing. Initially, we anticipated that it would retrieve only customers without any loans. However, we discovered that the query did not only captured customers without any loans, as we expected, but also included customers who had either housing loans or personal loans. This outcome has taught us an important lesson in SQL query logic. When the housing loan status is set to 'no,' the OR function within SQL moves on to the next option, in this case, personal loans to look for customers without personal loans which at the same time result in a 'yes' for housing loans and a 'no' for personal loans in certain cases.

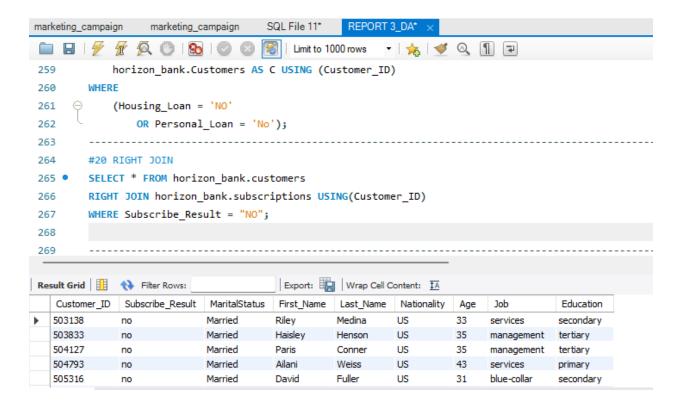
We can conclude that the best syntax to use here is the "AND" syntax which can be stated as WHERE Personal_loan = "NO" and Housing_Loan = "NO" to get customers without loans like the below screenshot shows.



Query 20: RIGHT JOIN

Objective: The objective here is to combine customer and subscription data to identify customers who have not subscribed to a call option. Understanding which customers have not subscribed can assist in targeting these customers with other communication offers and improving conversion rates.

The query below performs a right join between the customers table and subscriptions table, identifying customers who have not subscribed (Subscribe_Result = "NO").

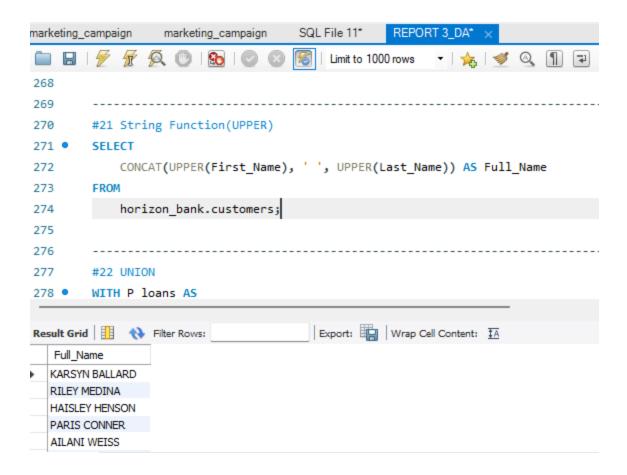


459 out of the 605 customers have "NO" as a subscription result. This is more than 50% of the bank's customers and indicates a low volume of communication between the bank and its customers. This can influence customer churning for some of the customers.

Query 21: String Function - UPPER

Objective: The objective here is to create a list of customer full names in uppercase for data consistency and suitability in various reports and applications.

The query below generates a list of customer full names with uppercase letters using the CONCAT and UPPER syntax.

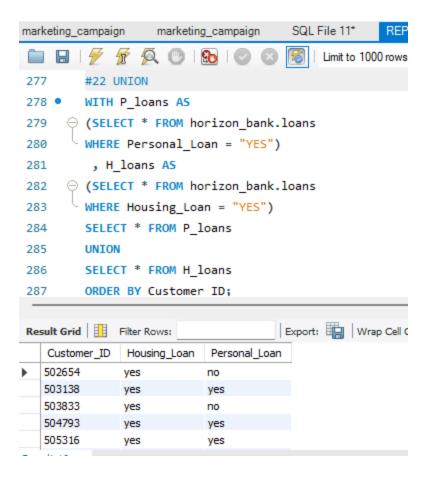


The output is a list of customers' first names combined with their last names in upper case.

Query 22: UNION

Objective: The aim is to combine two subsets of loan data (personal and housing loans) to create a unified list of loans. This is to streamline loan management and analysis processes. Since we didn't have two tables with similar characteristics to perform a union, we resorted to creating CTEs as tables and creating a union out of them.

The query below creates 2 CTEs, one for personal loans and another for housing loans and combines two subsets of loan data through a union. It produces a unified list of loans ordered by customer ID.

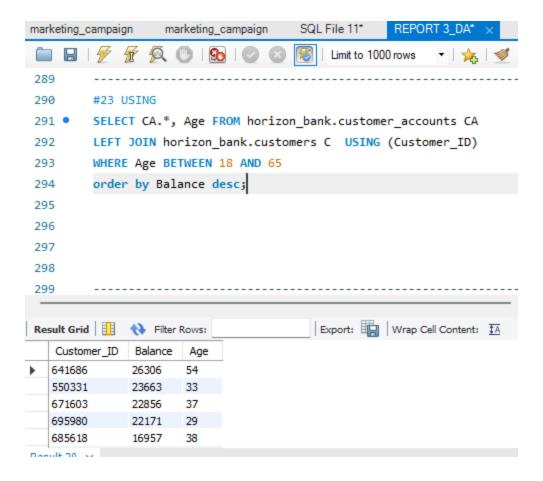


The output is a list of customers with their loans, basically making up the loans table.

Query 23: USING

Objective: The aim is to identify customers aged between 18 and 65.

The query below selects customers aged between 18 and 65, from the "customer_accounts" and joins it to the "customers" tables using Customer ID as a key.

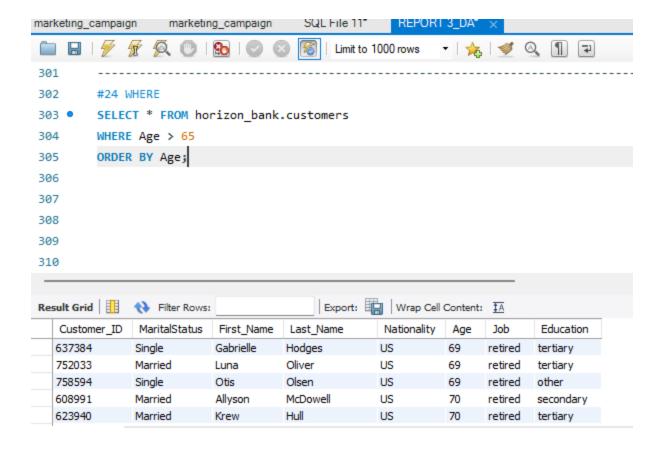


The output is a distribution of customer_IDs and balances together with their ages. Identifying customers with a specific balance threshold is essential for offering tailored financial products and services.

Query 24: WHERE

Objective: The objective is to identify customers aged above 65 years. This is to enable us obtain a view of customers who do not meet the age criteria for qualifying for the new product.

The query below selects customers whose age is greater than 65.



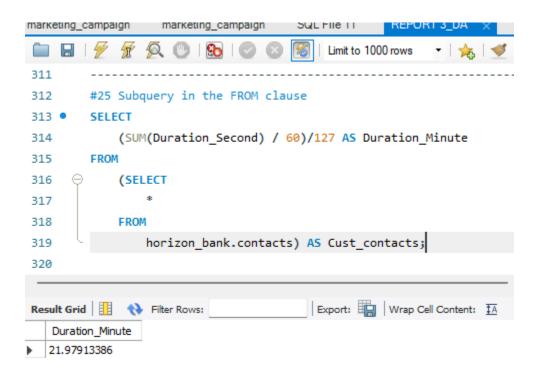
The output revealed 30 customers who are aged above 65 years, most of whom have retired from active business. The bank can provide specialized services, including retirement financial planning, low-risk investments, healthcare financing options and personalized customer service. These services aim to address the unique financial needs and preferences of elderly customers and ensure their financial well-being during retirement.

Query 25: (Subquery in the FROM clause)

Objective: This aim is to calculate the average duration of customer contacts per customer in minutes.

This query calculates the total duration of customer contacts in minutes by summing the

"Duration_Second" field from the subquery and dividing by 60 to convert if from seconds to minutes and then divides the answer by 127 which is the number of people who were contacted.



Output Report:

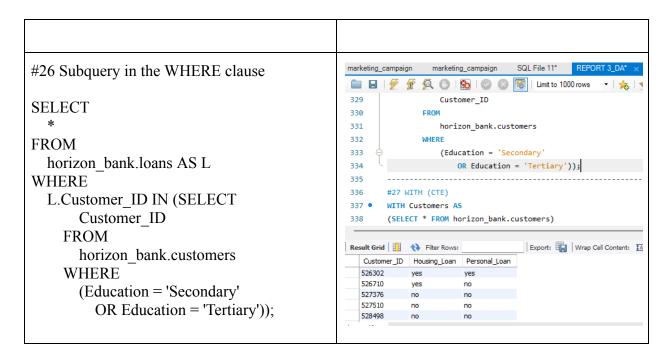
We can infer that, on average, the bank dedicated approximately 21 minutes to engage with each of the 127 customers it contacted. This suggests that the bank values customer interactions and invests time in providing personalized services or addressing their needs. It may also indicate a commitment to quality customer service and a willingness to spend adequate time to address customer inquiries, concerns, or transactions.

Query 26 (Subquery in the WHERE clause)

Objective: The aim here is to identify customers with specific educational backgrounds who have loans. Identifying customers with particular educational backgrounds who have loans is valuable for understanding the loan profiles of different education levels.

The query below relies on a subquery within the WHERE clause to filter the data customers with

specific educational backgrounds who have loans.



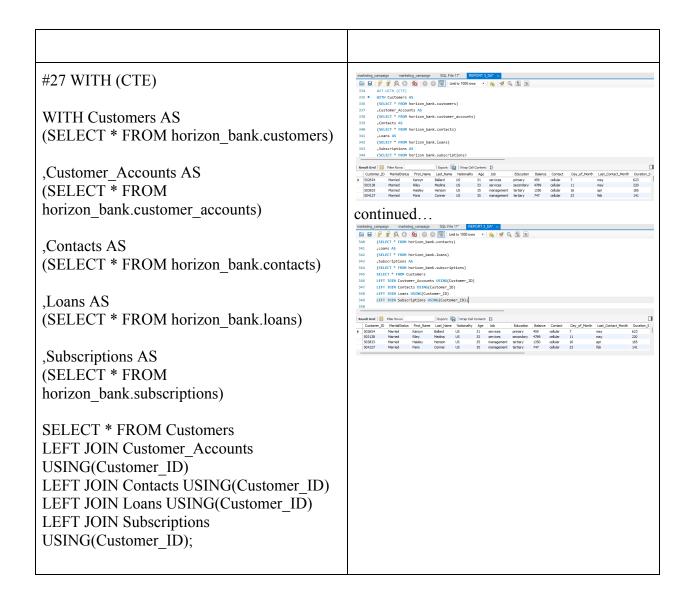
Output Report:

Surprisingly, the majority of customers with educational level as Secondary or Tertiary do not have any of the loans in this category. It could be that they may have alternative funding sources, such as scholarships, grants, or family support, which reduces the need for loans.

Query 27: WITH - CTE

Objective: The aim is to create a comprehensive dataset by joining multiple tables. This simplifies data analysis and provides a holistic view of customer profiles.

The query below creates a comprehensive dataset by combining customers table, customer accounts table, contact table, loan table, and subscription table using Common Table Expressions (CTE).



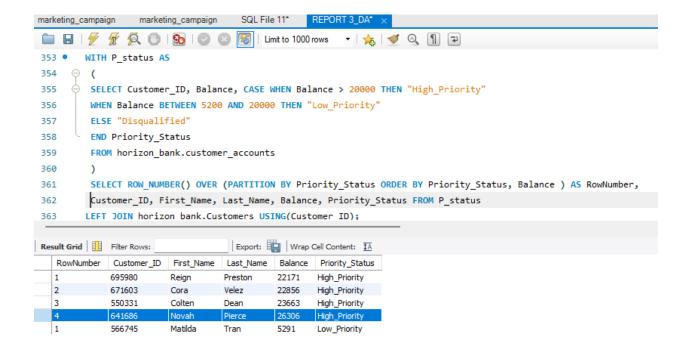
The output provides a simplified table of customers data for easy querying and data analysis.

Query 28: Window Function

Objective: The objective is to assign row numbers within priority status categories to help in

ranking and identifying customers based on their financial status.

The query below assigns row numbers to customers within priority status categories. It is partitioned by priority status and orders by priority status and balance.



Output Report:

The ranking from the table above will help the bank in ranking and identifying customers based on their financial status.

What Next?

Marketing Campaign

Customers who qualify for the product and subscribed to marketing campaigns (and are marked as "Can_Call") should be contacted by phone while those who requested not to be called should be contacted via email for targeted marketing.

Reflection

Reflecting on the bank's current policy that requires a customer to have existing loans to qualify for the new term deposit product seems to exclude a significant portion of potential customers. This approach may be biased against those without loans but with deposit needs. In response, the bank should consider revising this requirement or offering alternatives to cater to these customers' financial needs. For those without loans but with deposit requirements, the bank could provide attractive deposit rates, making it more inclusive.

We also realized that most of the customers of Horizon bank have low balances which currently disqualifies them from accessing the new product. To address this, the bank should engage with these customers to better understand their specific needs. It can create tailored products or establish value chains that redirect customer spending, ensuring more funds remain within the bank's ecosystem. By identifying the spending habits and preferences of its customers, the bank can offer more targeted solutions.

Additionally, offering various subscription options can help expand its contact customer base.

Lastly, given the fact that we had to follow the order of the 28 queries, we were restricted with the output we could generate from a query at different points in time. This had an impact on the consistency of the story we sought to tell with the output of our query given our objective of creating a list of customers who meet the requirements set out by Horizon bank in order to qualify for its new term deposit product.

Nonetheless, below is the query and screenshot of the output of the 23 customers who meet all 3 requirements set by the bank in order to qualify for the new term deposit product ranked in order of priority using their account balances.

```
SELECT
C.Customer_ID,
C.First_Name,
C.Last_Name,
CA.Balance,
```

```
C.Age,

RANK() OVER (ORDER BY CA.Balance DESC) AS BalanceRank

FROM

horizon_bank.customers AS C

INNER JOIN

horizon_bank.customer_accounts AS CA

ON C.Customer_ID = CA.Customer_ID

LEFT JOIN

horizon_bank.loans AS L
```

AND (L.Housing_Loan = 'yes' OR L.Personal_Loan = 'yes');

ON C.Customer_ID = L.Customer_ID

AND C.Age BETWEEN 18 AND 65

WHERE

CA.Balance >= 5200

