ITC 6107: Data Storage and Retrieval Techniques Project Winter Term 2020

Compare the performance of a standard RDBMS and SPARK-SQL on medium-size data

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Abstract

The main goal of this project is to compare the performance of a standard RDBMS and SPARK-SQL Framework on medium-size data. This could be achieved by storing and retrieving data from our RDBMS, as well as from our local disk using Spark SQL. For this demonstration we are going to run complex queries both on the RDBMS of our choice and on our local disk using Spark Framework.. Especially for our Spark Framework processing we are going to use the PySpark library with purpose to query our dataset using only one core in the beginning and at a later stage utilizing all the available cores. For our RDBMS processing we will use Microsoft's MSSQL Server 2012, Oracle 11 G R2, MySql 8 and Postgre-sql 10.12 and we will query our data using the classical PL/T-SQL using the appropriate Python Connector for each database. Calculating the execution time for those two cases will give us a crystal clear picture of the performance and speed for our two scenarios (RDBMS / SPARK-SQL) with final goal to run a statistical analysis of the extracted results.

The dataset

The dataset that we are going to use for this project is called Churn Modelling and it is about Banking Data from an Individual Bank. It is freely available on Kaggle (https://www.kaggle.com/kmalit/bank-customer-churn-prediction/data) and it contains about 10,000 rows and 14 feature variables. Each row corresponds to a customer infomation, and includes the following variables:

- 1. RowNumber: simply the number of the dataset's row
- 2. CustomerId: The unique Id for each specific customer
- 3. Surname: The surname of the customer
- 4. CreditScore: The calculated Credit Score of the customer
- 5. Geography: The country of the customer
- 6. Gender: The gender of the customer
- 7. Age: The age of the customer
- 8. Tenure: Number of Tenure payment plan
- 9. Balance: The current balance of customers' deposit
- 10. NumOfProducts: Number of Bank product services
- 11. HasCrCard: The possession of Credit Card
- 12. IsActiveMember: The activity status of the customer
- 13. EstimatedSalary: The estimated salary of the customer
- 14. Exited: The current churn status of the customer

We can very easily notice that this dataset full fills our requirements

- >10000 rows
- >10 attributes
- 3 columns with less than 10 distinct values (Exited, Gender, Geography for instance)

First things first let's see our dataset's description:

```
There are 10000 rows and 14 columns
There are no missing values
                  3 ...
                          9998 9999 100001
     1
           2
[15634602 15647311 15619304 ... 15584532 15682355 15628319]
                           ... 'Kashiwagi' 'Aldridge'
['Hargrave' 'Hill'
                    'Onio'
                                                         'Burbidge'l
[619 608 502 699 850 645 822 376 501 684 528 497 476 549 635 616 653
 726 732 636 510 669 846 577
                               756 571 574 411 591 533 553 520
                                                                 722 475
                                                                          490
 804 582 472 465
                  556 834 660
                               776 829 637
                                           550
                                                698
                                                    585
                                                         788
                                                             655
                                                                 601
                                                                     656
 511 614 742
             687
                  555
                      603
                          751
                               581
                                   735
                                       661 675
                                                738
                                                    813
                                                        657
                                                             604
                                                                 519
                                                                     664
                                                                          678
     416 665
              777
                  543
                      506
                          493
                               652
                                   750
                                       729
                                           646
                                                647
                                                    808
                                                         524
                                                             769
                                                                 730
                                                                     515
                                   479
                                           538
     710
         413
             623
                  670
                      622
                          785
                               605
                                       685
                                                562
                                                    721
                                                         628
                                                             668
                                                                 828
                                                                      674
                                                                          625
                      789
                          589
                               461
                                   584
                                       579
                                           663
                                                682
     770
         758
              795
                  686
                                                    793
                                                         691
                                                             485
                                                                 650
                                                                      754
                                                                          535
                          800
     539
                               683
                                   704
                                       615
                                           667
                                                484
                                                    480
 716
         706
              586
                  631
                      717
                                                         578
                                                             512
                                                                 606
                                                                      597
                                                         620
 514 525
         715
              580
                      521
                          759
                               516
                                   711
                                           643
                                                671
                                                    689
                                                                 572
                  807
                                       618
                                                             676
     694
         547
              594
                  673
                      610
                          767
                               763
                                   712
                                       703
                                           662
                                                659
                                                    523
                                                             545
                                                                 634
                                                         772
 681
     544
         696
              766
                  727
                      693
                          557
                               531
                                   498
                                       651
                                            791
                                                733
                                                    811
                                                         707
                                                             714
                                                                 782
                                                                          799
 602
     744
         588
              747
                  583
                      627
                          731
                               629
                                   438
                                       642
                                           806
                                                474
                                                    559
                                                        429
                                                             680
                                                                 749
                                                                          644
 626 649
         805
              718
                  840
                      630
                          654
                               762
                                   568
                                       613
                                            522
                                                737
                                                    648
                                                        443
                                                             640
                                                                 540
                                                                     460
                                                                          593
 801 611
         802
              745
                 483
                      690
                          492
                               709
                                   705
                                       560
                                            752
                                                701
                                                    537
                                                        487
                                                             596
                                                                 702
                                                                     486
                                                                          724
 548 464
         790
              534
                  748
                      494
                          590
                               468
                                   509
                                       818
                                           816
                                                536
                                                    753
                                                         774
                                                             621
                                                                 569
                                                                     658
                                                                          798
 641 542
         692
             639
                  765
                      570
                          638
                               599
                                   632
                                       779
                                            527
                                                564 833
                                                         504
                                                             842
                                                                 508
                                                                     417
                                                                          598
 741
     607
         761
              848
                  546
                      439
                          755
                               760
                                   526
                                       713
                                            700
                                                666
                                                    566
                                                         495
                                                             688
                                                                 612
                                                                     477
                                                                          427
                                                    786
 839 819
         720
             459
                  503
                      624
                          529
                               563 482
                                       796 445
                                                746
                                                         554
                                                             672
                                                                 787
                                                                     499
                                                                          844
 450 815
         838
             803
                  736
                      633
                          600
                               679
                                   517
                                       792
                                            743
                                                488 421
                                                         841
                                                             708
                                                                 507
                                                                      505
                                                                          456
 435 561
                                                         719
         518
             565
                  728
                      784
                          552 609
                                   764 697
                                            723
                                                551 444
                                                             496
                                                                 541
                                                                     830
                                                                          812
                          826 434 513 478
 677 420 595
             617
                  809
                      500
                                            797
                                                363
                                                    399
                                                        463
                                                             780
                                                                 452
                                                                     575
                                                                          837
                                       768 831 359 820
 794 824 428 823
                  781
                      849 489 431 457
                                                        573
                                                            576
                                                                 558
                                                                     817
                                                                          449
 440 415 821 530
                  350
                      446 425 740 481 783 358 845 451 458 469 423 404
 473 835 466 491
                  351 827 843 365 532 414 453 471 401 810 832 470 447 422
 825 430 436 426 408 847 418 437 410 454 407 455 462 386 405 383 395 467
 433 442 424 448 441 367 412 382 373 419]
```

Image 1: Dataset Description

```
'France'
          'Spain'
                  'Germany']
 'Female'
          'Male'
[42 41 39 43 44 50 29 27 31 24 34 25 35 45 58 32 38 46 36 33 40 51 61 49
 37 19 66 56 26 21 55 75 22 30 28 65 48 52 57 73 47 54 72 20 67 79 62 53
 80 59 68 23 60 70 63 64 18 82 69 74 71 76 77 88 85 84 78 81 92 83]
    1 8
              4 6 3 10
                         5 9
                                0]
      0.
            83807.86 159660.8
                                    57369.61 75075.31 130142.79]
[1 3 2 4]
[1 0]
[1 0]
[101348.88 112542.58 113931.57 ...
                                    42085.58
                                              92888.52
                                                         38190.781
[1 0]
```

Image 2: Dataset Description-2

We can confirm that our dataset consists of 10000 rows and 14 columns and we can also see all the distinct values for each variable contained in our dataframe. As we can see we have a mix of numerical and categorical and values; and we can notice that for some columns like Gender (Male/Female), Geography(France/Spain/Germany) and Exited which is a binary attribute (0/1) we do not have more than 3 different distinct values. On the other hand

some other columns like Estimated Salary and Surname contain multiple values (more than eight).

Before we proceed to our Data Storage and Retrieve step, let's visualize our dataset with purpose to get more familiar with it. This will help us to get a better understanding of our result when we will write our queries.

Proportion of customer churned and retained Exited 20.4% 79.6% Retained

Image 3: Churned Customer Piechart

In the above piechart we can see that the majority of the customers in our dataset are not churners. So what we are going to use for our queries are attributes which are responsible for the percentages we see on our pie.

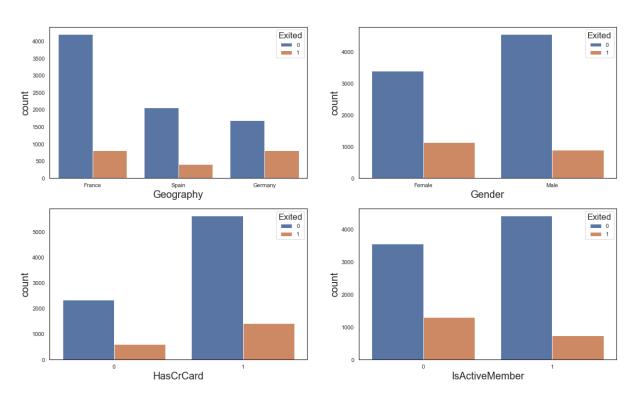


Image 4: Categorical Data Barplots

The above barplots can help us to understand what are the main characteristics of our churners. The majority of our customers and inevitably our churners are located in France and are Males. It is also interesting to notice that the majority of churners have a credit card. Unsurprisingly the inactive members column is higher than the active members, so the less time a customer spends with the bank the more likely he is to become a churner.

For the numerical attributes we created the following box-plots.

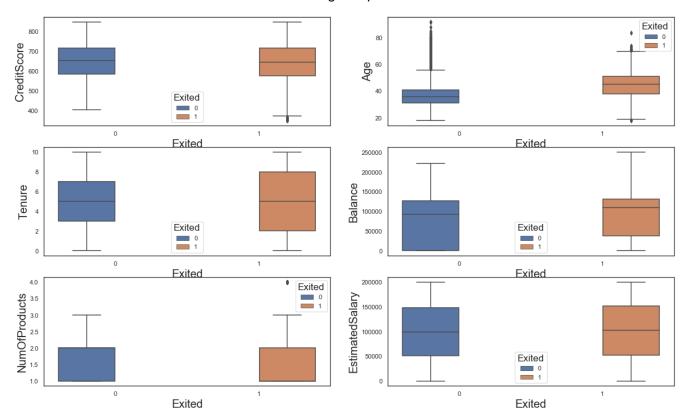


Image 5: Box plots for Numerical Data

If we observe each box-plot carefully we can see that there is no significant difference between the y axis values and that all the boxes (churner and not churner) are pretty much at the same height. The only two variables that are slight different is the Age variable and the Estimated Salary variable, which simply means that older Customers and people with high salary are more likely to leave the bank.

Importing and Storing Data

In this part we are going to see what technological stack did we use to store our data and what are the main differences between each database. Ideally we want our data to be stored exactly as appeared in our Excel File:

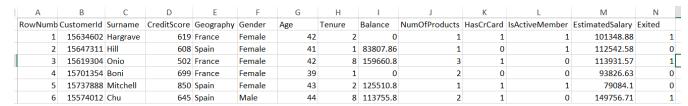


Image 6: Dataset in Excel form

The first database we used for our storage is an SQL Server 2012 database called "ITC6107A1" and it contains one table with the same name as our Excel file "Churn_Modelling" as the following image depicts:

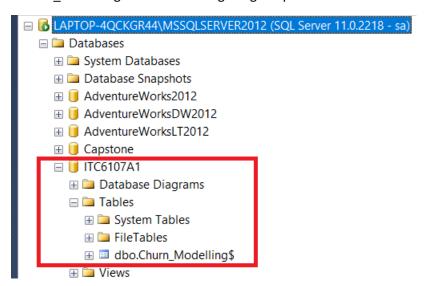


Image 7: SQL Server Menu

Using the Microsoft's Importing Excel Wizard was very easy and it resulted in storing the data exactly as we wanted them to be:

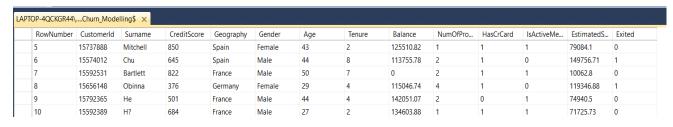


Image 8: SQL Server Churn_Modelling Table

We can query our data stored in our SQL Server database through Python just by installing the "pyodbc" library. Then all we have to do is to create the correct connection string and start querying our data. An Indicative Example is the following:

Image 9: SQL Server connector code

We can see that for the SQL Server queries we have put the correct database name and we have entered also the correct server name as we showed earlier. Then we used the datime library to count the execution and plotting time of each specific query. As we can see the result is the data and on the bottom we can see the execution time.

```
In [123]: runfile('C:/Users/30694/SQLServer.py', wdir='C:/
Users/30694')
(2037, )
-----
The time for Query No#1 is 0:00:00.002138
```

Image 10: SQL Server code results

The very first query we used is "SELECT COUNT(CustomerID) AS Number_of_Exited_Customers FROM churn_modelling WHERE Exited=1" we are going to explain all our queries at a later stage but this simply aggregate query with only one aggregate function will be our basic example for the rest databases in order to have a common line and be sure that the same query produces the exact same results in all our databases.

Similarly we did exactly the same steps for our MySql 8.0.19 Database in the Workbench studio, we used the same database and table names as we used in SQL Server and again we can see that the table's schema is the one we are looking for:

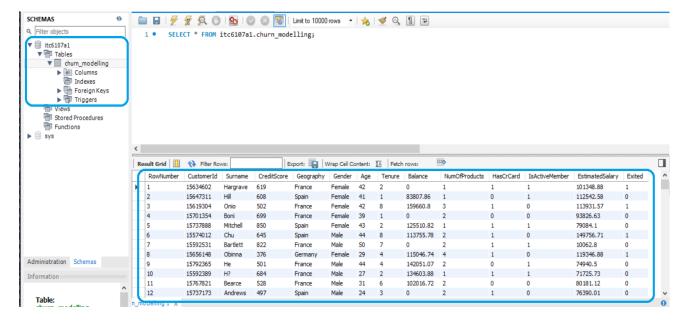


Image 11: MySql Menu

The querying concept via Python is exactly the same with the SQL Server but this time we are going to use the "mysql-connector" library in order to be able to communicate with the MySql Server:

Image 12: MySql Connector code

We can see that we are following the same "clean code" manner in our Python script. This helps us to spot very easily the differences between the each's databases' Python script. The datetime library and the Printing format is exactly the same, the only difference is the definition of the connection string which impacts the declaration of the database's cursor variable. Again we can see that the result is exactly the same '2037' for our counter which means that both databases are at the same state.

The third database that we chose to use for this demonstration is an Oracle 11 G R2 database. Due to the fact that setting up and a defining a new Oracle table is quite complex we are going to use the default "HR" database that comes with the Oracle's Installation but the table's name is going to be exactly the same churn_modelling:

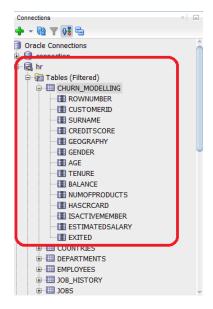


Image 13: Oracle Churn_Modelling Table Schema

The data stored in this table follow exactly the table structure and the order we wanted them to be:

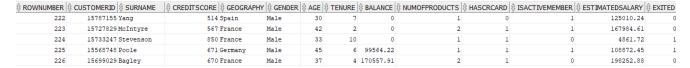


Image 14: Oracle Churn Modelling Table Data

The library that we installed and used this time is the cx_Oracle library. Notice that this time we did not define any new database but we used the "HR" database name because the Oracle's files and databases live on the generic 'Orcl' node database with the credentials we are using. Although Oracle uses PL-SQL instead of T-SQL for our queries we are not going to see many differences.

```
# importing module
import cx_Oracle
from datetime import datetime

con = cx_Oracle.connect('hr/hr@localhost/orcl')

#Query No#1
start=datetime.now()

c = con.cursor()
c.execute('SELECT COUNT(CustomerID) AS Number_of_Exited_Customers FROM churn_modelling WHERE Exited=1')

for result in c:
    print (result) # this only shows the first two columns. To add an additional column you'll need to add

print("------")
print ('The time for Query No#1 is %s' % (datetime.now()-start))
print("-----")
c.close()
```

Image 15: Oracle connector code

If we observe the Python code for the cx_Oracle library we can see that we are opening and closing each connection before and after each query execution respectively and this is because the Oracle consumes a lot of memory not only for just storing the data and living as a service but also during execution time. For the other databases we did not chose to do that because the impact on the final time is minor.

```
In [126]: runfile('C:/Users/30694/Oracle_sources/30694')
(2037,)
The time for Query No#1 is 0:00:00.003989
```

Image 16: Oracle code results

The last database that we are going to use is the PostgreSQL database 10.12 and similarly to what we did for our first two databases we name the database as "ITC6107A1" and the table "Churn_Modeling". The database's management program is the pgAdmin 4 which stands for Postgre-Sql Admin and it is a web based dashboard for creating and managing the database. Unfortunately pgAdmin does not provide any automatic Excel File Import wizard so we had to manually created the table's schema and import the data using a bulk-insert wizard.

The result of the insert operation is the following:

rownumber	i.	customerid	surname	creditscore	geography	gender	A C	age	<i>6</i> 1	tenure	•	balance	numofproducts	hascrcard	9	isactivem
bigint		[PK] bigint	text	bigint	text	text	9	bigint		bigint		double precision	bigint	bigint		bigint
	1	15634602	Hargrave	619	France	Female		4	2		2	0	1		1	^
	2	15647311	Hill	608	Spain	Female		4	1		1	83807.86	1		0	
	3	15619304	Onio	502	France	Female		4	2		8	159660.8	3		1	
	4	15701354	Boni	699	France	Female		3	39		1	0	2		0	
	5	15737888	Mitchell	850	Spain	Female		4	13		2	125510.82	1		1	
	6	15574012	Chu	645	Spain	Male		4	14		8	113755.78	2		1	
	7	15592531	Bartlett	822	France	Male		5	50		7	0	2		1	
	8	15656148	Obinna	376	Germany	Female		2	29		4	115046.74	4		1	
	9	15792365	Не	501	France	Male		4	4		4	142051.07	2		0	
	10	15592389	H?	684	France	Male		2	27		2	134603.88	1		1	
	11	15767821	Bearce	528	France	Male		3	31		6	102016.72	2		0	~

Image 17: PostgreSQL Churn_Modeling table

All the column data types had to be chosen by the database administrator so the bigint and text types were chose very carefully. Of course we can confirm with the following image that we chose the same names for convenience as we did before.



Image 18: PostgreSQL 10 Menu

The library that we used to communicate via Python was psycopg2 and the code format mirrors the previous script we saw. This time are targeting the database "ITC6107A1" and we are defining the username, the password, the post and the port that the connector we need to use in order to exchange information with the database.

*Image 19: Postgre SQL Connector Code*The query results are are always the same.

```
In [129]: runfile('C:/Users/30694/Postgre_sql.py',
(2037,)
------
The time for Query No#1 is 0:00:00.014179
```

Image 20: PostgreSQL Connector Code results

Last but not least we are going to see the setup and the configuration of the Spark Framework. Since our machine consists of a Windows 10 machine with an i7 core 10th generation we need to simply download Spark from the appropriate link (https://spark.apache.org/downloads.html) and extract the contents of the installation folder:

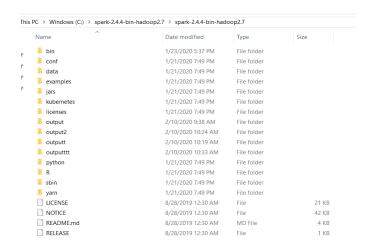


Image 21: Apache Spark Installation

Then we simply need to Edit our Environment variables so that Spark Framework can be ready for usage. First we make sure that the "SPARK_HOME" installation is located in our system variables along with the appropriate JDK Java installation.

And then we define as

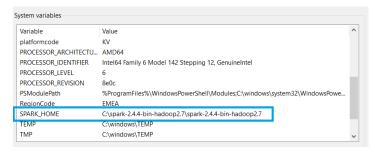


Image 22: SPARK_HOME System Variable

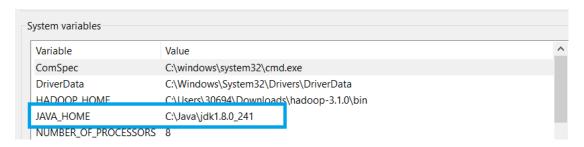


Image 23: JAVA_HOME System Variable

And then we add only the bin folder of our %SPARK_HOME% variable to our system path:

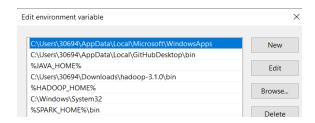


Image 24: System Path Variables

Then for our Query Execution we will install and use the Pyspark Library so that we can write Spark-Sql from our Python Intepreter. There are many possible variations but for this project we are going to use only two. In the first variation we are going to utilize one core and in the second variation we will utilize all the available cores which are eight (8).

The code that we will use to run SQL queries through Spark framework is the following:

```
import svs
from pyspark import SparkContext,SparkConf
import matplotlib.pyplot as plt # plotting
import numpy as np # linear algebra
import pandas as pd
df = pd.read_csv("C:\\Users\\30694\\Downloads\\Churn_Modelling.csv")
def kill_current_spark_context():
   SparkContext.getOrCreate().stop()
kill_current_spark_context()
sc = SparkContext('local[*]', 'FinalProject')
spark=SparkSession.builder.appName("spark sql test2").getOrCreate()
spark_df = spark.createDataFrame(df)
spark_df.registerTempTable("Table")
# Query No# 1
start=datetime.now()
query_result=spark.sql("SELECT COUNT(CustomerID) AS ExitedCustomers FROM Table WHERE Exited=1 ").show()
print ('The time for Query No#1 is %s' % (datetime.now()-start))
```

Image 25: Apache Spark Code

The main idea is the same with the previous scripts when we were using databases but in this scenario we simply define a Spark Context instead of a connection string so that we can start and have a reference variable for the Spark Framework. Once we have done that we define function which simply stops all the previously opened Spark Session, we run it and we create a new session. Then we create a Spark Dataframe from our Python Dataframe and we load that data in an temporary table which lives on our machine's memory. That temporary table will store temporary the data which will be queried at a later stage.

Image 26: Apache Spark Code Results

Queries

Since we have already discussed about the Spark Framework and the different databases that we will query, let's see the SQL Code of those 20 queries. We need to use complex group-by queries along with aggregate functions. Some of the queries use only one aggregate functions and some other use some more complex code like multiple aggregate functions with multiple sql conditions.

As we will see below not all the queries are complex in the same way nor they a contain the group by statement or the same name of aggregate functions. The variety of using queries of medium complexity in the beginning and then increasing the difficulty, will give us a better understanding of how each database and framework responds to each query(easy,medium,hard).

-- Query 1 count how many customers left from the bank

SELECT COUNT(CustomerID) AS Number_of_Exited_Customers FROM churn_modelling WHERE Exited=1

In the first query we are using simply the Count aggregate function, of course we could have added a Group By function but this would have resulted in a very big list of ashes and this is not something aesthetic for the first query which was used as a reference example in the previous chapter.

-- Query 2 Get Minimum, Maximum Average Customer Salary

SELECT AVG(EstimatedSalary) As
Average_Estimated_Customer_Salary,MIN(EstimatedSalary) As
Minimum_Estimated_Customer_Salary ,MAX(EstimatedSalary) As
Maximum_Estimated_Customer_Salary FROM churn_modelling

For the Second Query we want to find some basic statistics for the Estimated Salary because all this aggregate functions together will "slow" the databases' engine response.

-- Query 3 Count Active Members who exited

SELECT COUNT(CustomerID) FROM churn_modelling WHERE Exited=1 AND IsActiveMember=1

In the third query we use the count aggregate function along with two conditions with purpose to make the query a little bit more complex than the first one.

-- Query 4 Male/Female Percentage

SELECT (COUNT(CustomerID)*100/(SELECT COUNT(*) FROM churn_modelling)) AS Total_Male_Percentage FROM churn_modelling WHERE Gender='Male'

For the fourth query we are using a combination of a division of an aggregate function with a select statement. The result will produce a 'heavier' query not only because of the complex calculations but also because we added a where condition

--Query 5 Count people between 20 and 40

SELECT COUNT (CustomerId) AS COUNTERSUM FROM churn_modelling GROUP BY Age,CustomerId HAVING Age BETWEEN 20 AND 40;

For the fifth query we count the number of people between a specific age. We combine a group by query with a Having statement with purpose to filter the age range we are looking for.

-- Query 6 Count Customers from Spain and France

SELECT COUNT(CustomerId), Geography FROM churn_modelling WHERE Geography='Spain' or Geography='France' GROUP BY Geography

For the sixth query we use another group by query but in this case we have a double where query because we are using the word 'or', note that we still use the count aggregate function like the previous query.

-- Query 7 Select Last NAmes of Customers who are young, active member and also have credit card

SELECT CustomerId, Surname from churn_modelling WHERE NOT (HasCrCard=0 AND IsActiveMember=0) AND (Age<40) GROUP BY Surname, CustomerId ORDER BY Surname DESC

For the query number 7 we are using a combination of a Group-Order by statement, generally the Order By increases the response time of the query.

-- Query 8 Get All Customers whos name begins with A and are located in France/Germany

SELECT CustomerId,Surname, Geography FROM churn_modelling WHERE Surname LIKE 'A%' AND (Geography IN('Germany','France')) GROUP BY CustomerId,Surname,Geography ORDER BY CustomerId, Surname DESC,Geography DESC

For the query number 8 we added more columns on the Order By statement so that will increase even more the response time. Not that used also for the first time the LIKE SQL command which filters the data which are followed after the IN statement.

-- Query 9 Get All Customer with Tenure either 2,4,6,8 located in Spain/Germany and there age is only 20 or 30 or 40

SELECT CustomerId,Surname FROM churn_modelling WHERE Tenure IN ('2','4','6','8') AND Geography IN('Spain','Germany') AND AGE IN ('20', '30','40') GROUP BY CustomerId,Surname,Tenure,Geography,Age ORDER BY CustomerId,Surname,Tenure,Geography,Age

The query number 9 is one of the slow ones because as we can see it contains many values for the Where clause and also for the Group-Order By statements.

-- Query 10 Count Spanish Male with Credit Score over 500

SELECT COUNT(CustomerId) FROM churn_modelling WHERE (Gender='Male' AND Geography='Spain') OR(CreditScore>500)

For the Query number 10 we can see again the Count aggregate function but this time with a more complicated where clause. To be more specific we use 2 logical operator the 'AND' the 'OR' word so that the query can search even more within the data.

-- Query 11 Get Customers Over 30 with High Balance and Numb of Products 3 or 4

SELECT CustomerId,NumOfProducts FROM churn_modelling WHERE (Balance>150000 AND Age>30) OR (NumOfProducts IN('3' ,'4')) GROUP BY CustomerId,NumOfProducts

The eleventh query contains a three conditional where clause along with two columns in the group by sections, this also creates a mixed query that will take also a lot of time to be executed because we have the 'OR' logical operator which will fetch even more data.

-- Query 12 Count Customers with Credit Score over 500

SELECT Age, CreditScore FROM churn_modelling WHERE CreditScore>500 GROUP BY Age, CreditScore HAVING Age>(Select AVG(Age) FROM churn_modelling)

For the query number 12 we have added a having statement that compares the age attribute with a whole select statement.

-- Query 13 Min, Max, Avg for grouped Balance, Age, CreditScore, Tenure

SELECT MIN (Balance) AS Minimum_Balance, MAX (Balance) AS Maximum_Balance, AVG (Balance) AS Average_Balance, MIN (Age) AS Minimum_Age, MAX (Age) AS Maximum_Age, AVG (Age) AS Average_Age, MIN (CreditScore) AS Minimum_CreditScore, MAX (CreditScore) AS Maximum_CreditScore, AVG (CreditScore) AS Average_CreditScore, MIN (Tenure) AS Minimum_Tenure, MAX (Tenure) AS Maximum_Tenure, AVG (Tenure) AS Average_Tenure FROM churn_modelling GROUP BY Balance, Age, CreditScore, Tenure

The query number 13 is the longest one because we use many aggregate functions for many columns in our table, so it is the query that takes the longest time to be executed and it contains the most words.

-- Query 14 Get Geography Balance Per Customer

SELECT Geography, Balance, sum(Balance) OVER(PARTITION BY Geography) AS TotalBalPerRegion FROM churn_modelling GROUP BY Geography, Balance

The query number 14 contains a group by clause with the Over Partition By sql command and this help us to create a better grouping over a particular set of rows. The results will be to split the table into geographical regions and then fetch that data to our group by clause.

-- Query 15 Get Customers whose Salary is Greater than Average

SELECT CustomerId,EstimatedSalary FROM churn_modelling GROUP BY CustomerId,EstimatedSalary HAVING EstimatedSalary >(SELECT AVG(EstimatedSalary) FROM churn_modelling)

Similarly to query 12 we have the Having statement in our query but this time we compare our attribute with a whole sql query that contains also an aggregate function.

-- Query 16 Get Customer whose Balance is Greater than Average

SELECT CustomerId, Surname FROM churn_modelling GROUP BY
CustomerId,Surname,Balance,CreditScore HAVING Balance >(SELECT AVG(Balance)FROM churn_modelling) AND CreditScore >(SELECT AVG(CreditScore)FROM churn_modelling)

For the query number 16 again we use many columns in our group by clause and we filter them by using a having command along with a comparison of a single column with another sql query

-- Query 17 Count Customer per Tenure group

SELECT CustomerId, Tenure, count(*) FROM churn_modelling GROUP BY GROUPING SETS ((CustomerId), (Tenure))

For the query number 17 we are replacing our classical group by sql command with the command group by grouping sets, which results into returning to grouped lists for the columns that are contained in the parentheses.

-- Query 18 Get Customers whose total Profit is greater than a Million

SELECT CustomerId, SUM(Balance+EstimatedSalary) as [TotalMoney] FROM churn_modelling GROUP BY CustomerId HAVING SUM(Balance+EstimatedSalary)> 100000 ORDER BY TotalMoney DESC;

The query number 18 contains a SUM function which adds the values of two other columns, this combined with the order by sql command will result also in a complex query.

-- Query 19 Tenure Percentage

select Tenure, count(*) as total_tenure, count(*) * 1.0 / sum(count(*)) over () as ratio from churn_modelling group by Tenure;

For the last before the end query we can see a group by statement which contains a count aggregate functions along with a statistical division of two other functions; the count and the sum function. Notice also the word over which filters the data over a set of rows and does a more elastic search otherwise the percentage would not have been calculated.

-- Query 20 Get Customers with Credit Score greater than average

SELECT CustomerId FROM churn_modelling WHERE CreditScore> (SELECT AVG(CreditScore) FROM churn_modelling)

The final query contains a where clause followed by a whole select statement, that slowers the execution time because the engine will execute first a second select query within the main one.

Statistical Analysis

In this part we are going to see the final results of all those queries that we mentioned before. The analytical SQL code for each database can be found in the Appendix of this assignment. In the below image we can see the database comparison in seconds for each specific query. As we can see the Postgre-SQL database is the slowest one where as the SQL Sever is the fastest.

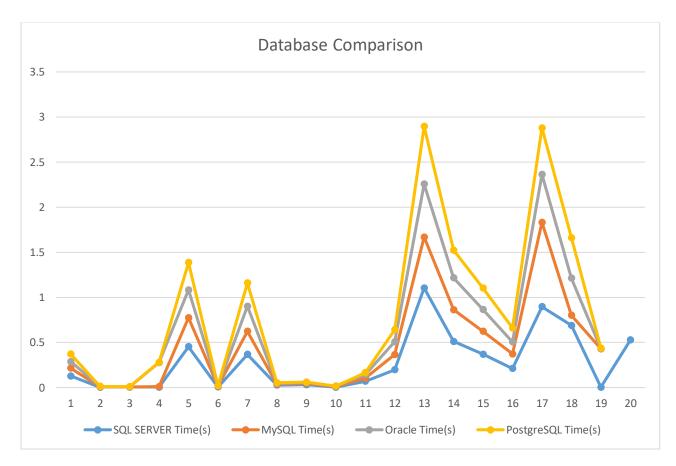


Image 27: Database Comparison Graph

Now let's put the Spark Framework in the previous graph and see how well it did:

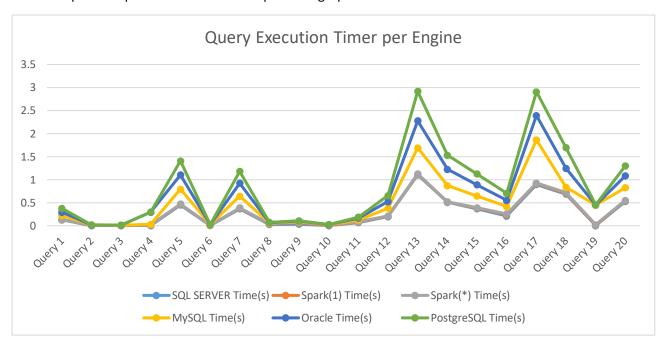


Image 28: Query Execution Time per Engine Graph

We can see very easily that in all the queries the Apache Spark Framework was the fastest (red and grey line) where as the other data storage systems did not perform so well. Of course there are some queries that were not extremely complex so the lines might overlap, so this is also another proof that for the very complex queries the time difference is significant for instance Queries 4-7 and 12-18. For more details for each specific query that overlap for example Queries 2 and 3 we have noted the time for all the databases and the Spark Framework for each specific query:

Queries	SQL SERVER Time(s)	Spark(1) Time(s)	Spark(*) Time(s)	MySQL Time(s)	Oracle Time(s)	PostgreSQL Time(s)
Query 1	0.127657	0.006479	0.001996	0.087764	0.071808	0.083787
Query 2	0.001995	0.004404	0.005163	0.004988	0.001993	0.002981
Query 3	0.001995	0.002017	0.003992	0.003988	0.001996	0.001962
Query 4	0.002991	0.012604	0.006065	0.007979	0.26686	0.001994
Query 5	0.452605	0.012021	0.008639	0.321646	0.30868	0.303162
Query 6	0.004988	0.003641	0.005195	0.008967	0.002958	0.002992
Query 7	0.36709	0.014535	0.00606	0.256336	0.279057	0.256818
Query 8	0.025904	0.014598	0.008779	0.009969	0.008976	0.007978
Query 9	0.032912	0.031357	0.014396	0.012964	0.01097	0.002992
Query 10	0.003989	0.002027	0.009546	0.00698	0.002992	0.001959
Query 11	0.068816	0.013921	0.007816	0.034905	0.039892	0.021941
Query 12	0.196292	0.010943	0.004437	0.169546	0.142619	0.131648
Query 13	1.104802	0.013112	0.009166	0.564542	0.585813	0.640084
Query 14	0.511632	0.005591	0.004598	0.350639	0.353562	0.306751
Query 15	0.369012	0.013273	0.009892	0.253833	0.241354	0.23787
Query 16	0.211435	0.02822	0.018125	0.162051	0.131647	0.154586
Query 17	0.895978	0.021214	0.008369	0.936071	0.531306	0.514897
Query 18	0.689125	0.018209	0.016401	0.111689	0.4109	0.449451
Query 19	0.003989	0.009717	0.009625	0.425121	0.003024	0.003988
Query 20	0.526575	0.018967	0.008166	0.27327	0.254867	0.217923

We have already proved that the Apache Spark Framework is way faster than the rest databases when it comes to query processing. What is left to compare now is to see if the number of cores in our machine play any significant role in the execution time.

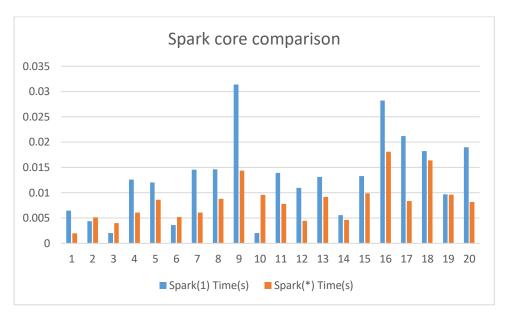


Image 29: Spark Core Comparison

As we can see, utilizing all the available cores in our machine and making them available to the spark frameworks makes it very fast. The orange columns which stand for the multi-core utilization are almost half the size of the blue one which stand for the single-core utilization, so we can conclude that indeed the Spark Framework is very fast compared to other data storage systems but the Spark Framework with multi-core utilization is fast in the scale of microsecond!

Future Work

All in all this assignment simply compared how the different databases and the Spark Framework behave on complex queries. This assignment could be easily extended by examining even more complex SQL structures like stored procedures or functions. It would be also interesting to see how a powerful business server would have behaved with these queries or even a very strong GPU instead of simply CPU.

References

- L. Perkins (2018) Seven Databases in Seven Weeks, 2nd edition, The Pragmatic Bookshelf
- L. Weise (2015), Advanced Data Management, de Gruyter
- J. Lescovec, A. Rajaraman, J.D. Ullman (2014), Mining of Massive Datasets, Cambridge University Press

Apendix SQL SERVER Queries

```
from datetime import datetime
import pyodbc
conn = pyodbc.connect('Driver={SQL Server};'
          'Server=LAPTOP-4QCKGR44\MSSQLSERVER2012;'
          'Database=ITC6107a1;'
          'Trusted_Connection=yes;')
cursor = conn.cursor()
#Query No#1
start=datetime.now()
cursor.execute('SELECT COUNT(CustomerId) AS [Number of Exited Customers] FROM
ITC6107A1.dbo.Churn_Modelling$ WHERE Exited=1 ')
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#1 is %s' % (datetime.now()-start))
print("-----")
#Query No#2
start=datetime.now()
cursor.execute('SELECT AVG(EstimatedSalary) As [Average Estimated Customer
Salary], MIN(EstimatedSalary) As [Minimum Estimated Customer Salary]
,MAX(EstimatedSalary) As [Maximum Estimated Customer Salary] FROM
ITC6107A1.dbo.Churn_Modelling$')
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#2 is %s' % (datetime.now()-start))
print("-----")
```

```
#Query No#3
start=datetime.now()
cursor.execute('SELECT COUNT(CustomerID) FROM ITC6107A1.dbo.Churn_Modelling$
WHERE Exited=1 AND IsActiveMember=1 ')
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#3 is %s' % (datetime.now()-start))
print("-----")
#Query No#4
start=datetime.now()
cursor.execute(""" SELECT (COUNT(CustomerID)*100/(SELECT COUNT(*) FROM
ITC6107A1.dbo.Churn_Modelling$)) AS [Total Male Percentage] FROM
ITC6107A1.dbo.Churn Modelling$ WHERE Gender='Male'""")
for row in cursor:
 print(row)
 print("-----")
print ('The time for Query No#4 is %s' % (datetime.now()-start))
print("-----")
#Query No#5
start=datetime.now()
cursor.execute('SELECT COUNT (CustomerId) AS [COUNTERSUM] FROM
ITC6107A1.dbo.Churn Modelling$ GROUP BY Age, CustomerId HAVING Age BETWEEN 20
AND 40')
for row in cursor:
 print(row)
print("-----")
print ('The time for Query No#5 is %s' % (datetime.now()-start))
print("----")
```

```
#Query No#6
start=datetime.now()
cursor.execute("""SELECT COUNT(CustomerId), Geography FROM
ITC6107A1.dbo.Churn Modelling$ WHERE Geography='Spain' or Geography='France'
GROUP BY Geography""")
for row in cursor:
 print(row)
print("-----")
print ('The time for Query No#6 is %s' % (datetime.now()-start))
print("-----")
#Query No#7
start=datetime.now()
cursor.execute('SELECT CustomerId, Surname from ITC6107A1.dbo.Churn Modelling$
WHERE NOT (HasCrCard=0 AND IsActiveMember=0) AND (Age<40) GROUP BY
Surname, CustomerId ORDER BY Surname DESC')
for row in cursor:
 print(row)
print("-----")
print ('The time for Query No#7 is %s' % (datetime.now()-start))
print("----")
#Query No#8
start=datetime.now()
cursor.execute("""SELECT CustomerId,Surname, Geography FROM
ITC6107A1.dbo.Churn Modelling$ WHERE Surname LIKE 'A%' AND (Geography
IN('Germany', 'France')) GROUP BY CustomerId, Surname, Geography ORDER BY CustomerId,
Surname DESC, Geography DESC""")
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#8 is %s' % (datetime.now()-start))
print("-----")
```

```
#Query No#9
start=datetime.now()
cursor.execute(""" SELECT CustomerId, Surname FROM ITC6107A1.dbo.Churn_Modelling$
WHERE Tenure IN ('2', '4', '6', '8') AND Geography IN('Spain', 'Germany') AND AGE IN ('20',
'30','40') GROUP BY CustomerId, Surname, Tenure, Geography, Age ORDER BY
CustomerId, Surname, Tenure, Geography, Age""")
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#9 is %s' % (datetime.now()-start))
print("----")
#Query No#10
start=datetime.now()
cursor.execute("""SELECT COUNT(CustomerId) FROM ITC6107A1.dbo.Churn Modelling$
WHERE (Gender='Male' AND Geography='Spain') OR(CreditScore>500)""")
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#10 is %s' % (datetime.now()-start))
print("-----")
#Query No#11
start=datetime.now()
cursor.execute(""" SELECT CustomerId, NumOfProducts FROM
ITC6107A1.dbo.Churn_Modelling$ WHERE (Balance>150000 AND Age>30) OR (
NumOfProducts IN('3','4')) GROUP BY CustomerId, NumOfProducts """)
for row in cursor:
 print(row)
```

print("")
print ('The time for Query No#11 is %s' % (datetime.now()-start))
print("")
#Query No#12
start=datetime.now()
cursor.execute('SELECT Age,CreditScore FROM ITC6107A1.dbo.Churn_Modelling\$ WHERE CreditScore>500 GROUP BY Age, CreditScore HAVING Age>(Select AVG(Age) FROM ITC6107A1.dbo.Churn_Modelling\$)')
for row in cursor:
print(row)
print("")
print ('The time for Query No#12 is %s' % (datetime.now()-start))
print("")
#Query No#13
start=datetime.now()
cursor.execute('SELECT MIN (Balance) AS [Minimum Balance], MAX (Balance) AS [Maximum Balance], AVG (Balance) AS [Average Balance], MIN (Age) AS [Minimum Age], MAX (Age) AS [Maximum Age], AVG (Age) AS [Average Age], MIN (CreditScore) AS [Minimum CreditScore], MAX (CreditScore) AS [Maximum CreditScore], AVG (CreditScore) AS [Average CreditScore], MIN (Tenure) AS [Minimum Tenure], MAX (Tenure) AS [Maximum Tenure], AVG (Tenure) AS [Average Tenure] FROM ITC6107A1.dbo.Churn_Modelling\$ GROUP BY Balance,Age,CreditScore,Tenure')
for row in cursor:
print(row)

```
print("----")
print ('The time for Query No#13 is %s' % (datetime.now()-start))
print("----")
#Query No#14
start=datetime.now()
cursor.execute('SELECT Geography, Balance, sum(Balance ) OVER(PARTITION BY Geography )
AS TotalBalPerRegion FROM ITC6107A1.dbo.Churn_Modelling$ GROUP BY Geography,
Balance')
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#14 is %s' % (datetime.now()-start))
print("----")
Query No#15
start=datetime.now()
cursor.execute('SELECT CustomerId, Estimated Salary FROM
ITC6107A1.dbo.Churn_Modelling$ GROUP BY CustomerId,EstimatedSalary HAVING
EstimatedSalary > (SELECT_AVG(EstimatedSalary ) FROM ITC6107A1.dbo.Churn_Modelling$
)')
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#15 is %s' % (datetime.now()-start))
print("----")
```

```
#Query No#16
start=datetime.now()
cursor.execute('SELECT CustomerId, Surname FROM ITC6107A1.dbo.Churn_Modelling$
GROUP BY CustomerId, Surname, Balance, CreditScore HAVING Balance > (SELECT
AVG(Balance)FROM ITC6107A1.dbo.Churn_Modelling$) AND CreditScore >(SELECT
AVG(CreditScore )FROM ITC6107A1.dbo.Churn_Modelling$ )')
for row in cursor:
 print(row)
print ('The time for Query No#16 is %s' % (datetime.now()-start))
print("----")
#Query No#17
start=datetime.now()
cursor.execute('SELECT CustomerId, Tenure, count(*) FROM
ITC6107A1.dbo.Churn Modelling$ GROUP BY GROUPING SETS ( (CustomerId), (Tenure))')
for row in cursor:
 print(row)
print("----")
print ('The time for Query No#17 is %s' % (datetime.now()-start))
print("----")
```

```
#Query No#18
start=datetime.now()
cursor.execute('SELECT CustomerId, SUM(Balance+EstimatedSalary) as [TotalMoney] FROM
ITC6107A1.dbo.Churn Modelling$ GROUP BY CustomerId HAVING
SUM(Balance+EstimatedSalary)> 100000 ORDER BY TotalMoney DESC;')
for row in cursor:
 print(row)
print("-----")
print ('The time for Query No#18 is %s' % (datetime.now()-start))
print("----")
#Query No#19
start=datetime.now()
cursor.execute('select [Tenure], count(*) as total tenure, count(*) * 1.0 / sum(count(*))
over () as ratio from ITC6107A1.dbo.Churn Modelling$ group by [Tenure];')
for row in cursor:
 print(row)
print ('The time for Query No#19 is %s' % (datetime.now()-start))
print("----")
#Query No#20
start=datetime.now()
cursor.execute('SELECT CustomerId FROM ITC6107A1.dbo.Churn_Modelling$ WHERE
CreditScore> (SELECT AVG(CreditScore) FROM ITC6107A1.dbo.Churn_Modelling$ )')
# for row in cursor:
  print(row)
print("----")
print ('The time for Query No#20 is %s' % (datetime.now()-start))
print("-----")
```

Oracle Queries

```
# importing module
import cx_Oracle
from datetime import datetime
con = cx_Oracle.connect('hr/hr@localhost/orcl')
#Query No#1
start=datetime.now()
c = con.cursor()
c.execute('SELECT COUNT(CustomerID) AS Number_of_Exited_Customers FROM
churn modelling WHERE Exited=1') # use triple quotes if you want to spread your query
across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("----")
print ('The time for Query No#1 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#2
start=datetime.now()
c = con.cursor()
c.execute("SELECT AVG(EstimatedSalary) As
Average_Est_Customer_Salary,MIN(EstimatedSalary) As Minimum_Est_Customer_Salary
,MAX(EstimatedSalary) As Maximum_Est_Customer_Salary FROM churn_modelling") # use
triple quotes if you want to spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("----")
print ('The time for Query No#2 is %s' % (datetime.now()-start))
print("----")
```

```
#Query No#3
start=datetime.now()
c = con.cursor()
c.execute('SELECT COUNT(CustomerID) FROM churn_modelling WHERE Exited=1 AND
IsActiveMember=1') # use triple quotes if you want to spread your query across multiple
lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#3 is %s' % (datetime.now()-start))
print("----")
#Query No#4
start=datetime.now()
c = con.cursor()
c.execute("SELECT (COUNT(CustomerID)*100/(SELECT COUNT(*) FROM churn modelling))
AS Total_Male_Percentage FROM churn_modelling WHERE Gender='Male' GROUP BY
CUSTOMERID") # use triple quotes if you want to spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#4 is %s' % (datetime.now()-start))
print("-----")
```

```
#Query No#5
start=datetime.now()
c = con.cursor()
c.execute('SELECT COUNT (CustomerId) AS COUNTERSUM FROM churn modelling GROUP
BY Age, CustomerId HAVING Age BETWEEN 20 AND 40') # use triple quotes if you want to
spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#5 is %s' % (datetime.now()-start))
print("----")
#Query No#6
start=datetime.now()
c = con.cursor()
c.execute("SELECT COUNT(CustomerId), Geography FROM churn_modelling WHERE
Geography='Spain' or Geography='France' GROUP BY Geography") # use triple quotes if you
want to spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("----")
print ('The time for Query No#6 is %s' % (datetime.now()-start))
print("----")
#Query No#7
start=datetime.now()
c = con.cursor()
c.execute("SELECT CustomerId, Surname from churn modelling WHERE NOT ( HasCrCard=0
AND IsActiveMember=0) AND (Age<40) GROUP BY Surname, CustomerId ORDER BY Surname
DESC") # use triple quotes if you want to spread your query across multiple lines
```

for result in c:

print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc. print("-----") print ('The time for Query No#7 is %s' % (datetime.now()-start)) print("-----") #Query No#8 start=datetime.now() c = con.cursor() c.execute("SELECT CustomerId, Surname, Geography FROM churn modelling WHERE Surname LIKE 'A%' AND (Geography IN('Germany', 'France')) GROUP BY CustomerId, Surname, Geography ORDER BY CustomerId, Surname DESC, Geography DESC") # use triple quotes if you want to spread your query across multiple lines for result in c: print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc. print("-----") print ('The time for Query No#8 is %s' % (datetime.now()-start)) print("----") #Query No#9 start=datetime.now() c = con.cursor() c.execute("SELECT CustomerId, Surname FROM churn_modelling WHERE Tenure IN ('2', '4' ,'6' ,'8') AND Geography IN('Spain' ,'Germany') AND AGE IN ('20', '30','40') GROUP BY CustomerId, Surname, Tenure, Geography, Age ORDER BY CustomerId, Surname, Tenure, Geography, Age") # use triple quotes if you want to spread your query across multiple lines for result in c: print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc. print("-----")

print ('The time for Query No#9 is %s' % (datetime.now()-start))

print("")
Query No#10
start=datetime.now()
c = con.cursor()
c.execute("SELECT COUNT(CustomerId) FROM churn_modelling WHERE (Gender='Male' AND Geography='Spain') OR(CreditScore>500)") # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.
print("")
print ('The time for Query No#10 is %s' % (datetime.now()-start))
print("")
#Query No#11
start=datetime.now()
c = con.cursor()
c.execute("SELECT CustomerId,NumOfProducts FROM churn_modelling WHERE (Balance>150000 AND Age>30) OR (NumOfProducts IN('3' ,'4')) GROUP BY
CustomerId,NumOfProducts ") # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', $row[2]$, etc.
print/"
print("")
print ('The time for Query No#11 is %s' % (datetime.now()-start))
print("")

```
#Query No#12
start=datetime.now()
c = con.cursor()
c.execute('SELECT Age,CreditScore FROM churn_modelling WHERE CreditScore>500 GROUP
BY Age, CreditScore HAVING Age>(Select AVG(Age) FROM churn modelling )') # use triple
quotes if you want to spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#12 is %s' % (datetime.now()-start))
print("----")
#Query No#13
start=datetime.now()
c = con.cursor()
c.execute('SELECT MIN (Balance) AS Minimum Balance, MAX (Balance) AS
Maximum_Balance, AVG (Balance) AS Average_Balance, MIN (Age) AS Minimum_Age, MAX
(Age) AS Maximum_Age, AVG (Age) AS Average_Age, MIN (CreditScore) AS
Minimum CreditScore, MAX (CreditScore) AS Maximum CreditScore, AVG (CreditScore) AS
Average_CreditScore , MIN (Tenure) AS Minimum_Tenure, MAX (Tenure) AS
Maximum_Tenure, AVG (Tenure) AS Average_Tenure FROM churn_modelling GROUP BY
Balance, Age, CreditScore, Tenure') # use triple quotes if you want to spread your query across
multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("----")
print ('The time for Query No#13 is %s' % (datetime.now()-start))
print("----")
#Query No#14
start=datetime.now()
c = con.cursor()
c.execute('SELECT Geography, Balance, sum(Balance) OVER(PARTITION BY Geography) AS
TotalBalPerRegion FROM churn_modelling GROUP BY Geography, Balance') # use triple
quotes if you want to spread your query across multiple lines
```

for result in c:

need to add , '-', row[2], etc. print("----") print ('The time for Query No#14 is %s' % (datetime.now()-start)) print("-----") Query No#15 start=datetime.now() c = con.cursor() c.execute('SELECT CustomerId, Estimated Salary FROM churn_modelling GROUP BY CustomerId, EstimatedSalary HAVING EstimatedSalary > (SELECT AVG(EstimatedSalary)) FROM churn modelling)') # use triple quotes if you want to spread your query across multiple lines for result in c: print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc. print("----") print ('The time for Query No#15 is %s' % (datetime.now()-start)) print("----") #Query No#16 start=datetime.now() c = con.cursor() c.execute('SELECT CustomerId, Surname FROM churn_modelling GROUP BY CustomerId, Surname, Balance, CreditScore HAVING Balance > (SELECT AVG(Balance) FROM churn modelling) AND CreditScore >(SELECT AVG(CreditScore)FROM churn modelling)') # use triple quotes if you want to spread your query across multiple lines for result in c: print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc. print("-----") print ('The time for Query No#16 is %s' % (datetime.now()-start))

print (result) # this only shows the first two columns. To add an additional column you'll

print("")
#Query No#17
start=datetime.now()
c = con.cursor()
c.execute('SELECT CustomerId, Tenure, count(*) FROM churn_modelling GROUP BY GROUPING SETS ((CustomerId), (Tenure))') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', $row[2]$, etc.
print("")
print ('The time for Query No#17 is %s' % (datetime.now()-start))
print("")
Query No#18
start=datetime.now()
c = con.cursor()
c.execute('SELECT CustomerId, SUM(Balance+EstimatedSalary) as TotalMoney FROM churn_modelling GROUP BY CustomerId HAVING SUM(Balance+EstimatedSalary)> 100000 ORDER BY TotalMoney DESC') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', $row[2]$, etc.
print("")
print ('The time for Query No#18 is %s' % (datetime.now()-start))

print("")
#Query No#19
start=datetime.now()
c = con.cursor()
c.execute('select Tenure, count(*) as total_tenure, count(*) * 1.0 / sum(count(*)) over () as ratio from churn_modelling group by Tenure ') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', $row[2]$, etc.
print("")
print ('The time for Query No#19 is %s' % (datetime.now()-start))
print("")
#Query No#20
start=datetime.now()
c = con.cursor()
c.execute('SELECT CustomerId FROM churn_modelling WHERE CreditScore> (SELECT AVG(CreditScore) FROM churn_modelling)') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.
print("")
print ('The time for Query No#20 is %s' % (datetime.now()-start))
print("")

MySQL Queries

```
from datetime import datetime
import mysql.connector
mydb = mysql.connector.connect(user='root', password='Bnft!123',
              host='127.0.0.1', database='itc6107a1',
              auth_plugin='mysql_native_password')
mycursor = mydb.cursor()
# Query No# 1
start=datetime.now()
mycursor.execute("SELECT COUNT(CustomerID) AS Number_of_Exited_Customers FROM
churn modelling WHERE Exited=1")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#1 is %s' % (datetime.now()-start))
print("-----")
# Query No# 2
start=datetime.now()
mycursor.execute("SELECT AVG(EstimatedSalary) As
Average_Estimated_Customer_Salary,MIN(EstimatedSalary) As
Minimum_Estimated_Customer_Salary ,MAX(EstimatedSalary) As
Maximum_Estimated_Customer_Salary FROM churn_modelling")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#2 is %s' % (datetime.now()-start))
print("-----")
```

```
# Query No# 3
start=datetime.now()
mycursor.execute("SELECT COUNT(CustomerID) FROM churn_modelling WHERE Exited=1
AND IsActiveMember=1")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#3 is %s' % (datetime.now()-start))
print("-----")
# Query No# 4
start=datetime.now()
mycursor.execute("SELECT (COUNT(CustomerID)*100/(SELECT COUNT(*) FROM
churn modelling )) AS Total Male Percentage FROM churn modelling WHERE
Gender='Male'")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#4 is %s' % (datetime.now()-start))
print("-----")
# Query No# 5
start=datetime.now()
mycursor.execute("SELECT COUNT(CustomerId) AS COUNTERSUM FROM churn_modelling
GROUP BY Age, CustomerId HAVING Age BETWEEN 20 AND 40 ")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#5 is %s' % (datetime.now()-start))
print("-----")
```

```
# Query No# 6
start=datetime.now()
mycursor.execute("SELECT COUNT(CustomerId), Geography FROM churn_modelling
WHERE Geography='Spain' or Geography='France' GROUP BY Geography")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#6 is %s' % (datetime.now()-start))
print("-----")
# Query No# 7
start=datetime.now()
mycursor.execute("SELECT CustomerId, Surname from churn modelling WHERE NOT (
HasCrCard=0 AND IsActiveMember=0) AND (Age<40) GROUP BY Surname, CustomerId
ORDER BY Surname DESC")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#7 is %s' % (datetime.now()-start))
print("-----")
# Query No# 8
start=datetime.now()
mycursor.execute("SELECT CustomerId,Surname, Geography FROM churn_modelling WHERE
Surname LIKE 'A%' AND (Geography IN('Germany', 'France')) GROUP BY
CustomerId, Surname, Geography ORDER BY CustomerId, Surname DESC, Geography DESC")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#8 is %s' % (datetime.now()-start))
print("----")
```

```
# Query No# 9
start=datetime.now()
mycursor.execute("SELECT CustomerId, Surname FROM churn_modelling WHERE Tenure IN
('2','4','6','8') AND Geography IN('Spain','Germany') AND AGE IN ('20', '30','40') GROUP BY
CustomerId, Surname, Tenure, Geography, Age ORDER BY
CustomerId, Surname, Tenure, Geography, Age")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#9 is %s' % (datetime.now()-start))
print("----")
# Query No# 10
start=datetime.now()
mycursor.execute("SELECT COUNT(CustomerId) FROM churn modelling WHERE
(Gender='Male' AND Geography='Spain') OR(CreditScore>500)")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#10 is %s' % (datetime.now()-start))
print("-----")
# Query No# 11
start=datetime.now()
mycursor.execute("SELECT CustomerId, NumOfProducts FROM churn modelling WHERE
(Balance>150000 AND Age>30) OR ( NumOfProducts IN( '3', '4')) GROUP BY
CustomerId, NumOfProducts ")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#11 is %s' % (datetime.now()-start))
print("----")
```

```
# Query No# 12
start=datetime.now()
mycursor.execute("SELECT Age, CreditScore FROM churn_modelling WHERE
CreditScore>500 GROUP BY Age, CreditScore HAVING Age>(Select AVG(Age) FROM
churn_modelling )")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#12 is %s' % (datetime.now()-start))
print("----")
# Query No# 13
start=datetime.now()
mycursor.execute(" SELECT MIN(Balance) AS Minimum_Balance, MAX(Balance) AS
Maximum Balance, AVG(Balance) AS Average Balance, MIN(Age) AS
Minimum Age, MAX(Age) AS Maximum Age, AVG(Age) AS Average Age, MIN(CreditScore)
AS Minimum_CreditScore, MAX(CreditScore) AS Maximum_CreditScore, AVG(CreditScore)
AS Average_CreditScore , MIN(Tenure) AS Minimum_Tenure, MAX(Tenure) AS
Maximum_Tenure, AVG(Tenure) AS Average_Tenure FROM churn_modelling GROUP BY
Balance, Age, Credit Score, Tenure")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#13 is %s' % (datetime.now()-start))
print("----")
# Query No# 14
start=datetime.now()
mycursor.execute("SELECT Geography, Balance, sum(Balance) OVER(PARTITION BY
Geography ) AS TotalBalPerRegion FROM churn_modelling GROUP BY Geography, Balance")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
```

```
print ('The time for Query No#14 is %s' % (datetime.now()-start))
print("-----")
# Query No# 15
start=datetime.now()
mycursor.execute("SELECT CustomerId, Estimated Salary FROM churn_modelling GROUP BY
CustomerId, EstimatedSalary HAVING EstimatedSalary > (SELECT AVG(EstimatedSalary))
FROM churn_modelling )")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#15 is %s' % (datetime.now()-start))
print("-----")
# Query No# 16
start=datetime.now()
mycursor.execute("SELECT CustomerId, Surname FROM churn modelling GROUP BY
CustomerId, Surname, Balance, CreditScore HAVING Balance > (SELECT AVG(Balance ) FROM
churn_modelling ) AND CreditScore >(SELECT_AVG(CreditScore )FROM churn_modelling )")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#16 is %s' % (datetime.now()-start))
print("----")
# Query No# 17
start=datetime.now()
mycursor.execute("SELECT CustomerId, Tenure, count(*) FROM churn modelling GROUP BY
CustomerId, Tenure WITH ROLLUP")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
```

```
print ('The time for Query No#17 is %s' % (datetime.now()-start))
print("-----")
# Query No# 18
start=datetime.now()
mycursor.execute("SELECT COUNT(CustomerID) AS Number of Exited Customers FROM
churn modelling WHERE Exited=1 GROUP BY CustomerId")
myreslt = mycursor.fetchall()
for x in myresult:
print(x)
print("----")
print ('The time for Query No#18 is %s' % (datetime.now()-start))
print("-----")
# Query No# 19
start=datetime.now()
mycursor.execute("SELECT CustomerId, SUM(Balance+EstimatedSalary) as TotalMoney
FROM churn modelling GROUP BY CustomerId HAVING
SUM(Balance+EstimatedSalary)>100000 ORDER BY TotalMoney DESC ")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----")
print ('The time for Query No#19 is %s' % (datetime.now()-start))
print("----")
# Query No# 20
start=datetime.now()
mycursor.execute("SELECT CustomerId FROM churn_modelling WHERE CreditScore>
(SELECT AVG(CreditScore) FROM churn modelling )")
myresult = mycursor.fetchall()
for x in myresult:
print(x)
print("-----") print ('The time for Query No#20 is %s' %
(datetime.now()-start))print("-----")
```

Postgre SQL Queries

```
import psycopg2
from datetime import datetime
connection = psycopg2.connect(database="ITC6107A1", user="sysadmin",
password="Bnft!123", host="127.0.0.1", port=5432)
#Query No#1
start=datetime.now()
c = connection.cursor()
c.execute('SELECT COUNT(customerid) AS Number of exited Customers FROM
churn modeling WHERE exited=1') # use triple quotes if you want to spread your query
across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#1 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#2
start=datetime.now()
c = connection.cursor()
c.execute('SELECT AVG(estimatedsalary) As
Average_Estimated_Customer_Salary,MIN(estimatedsalary) As
Minimum_Estimated_Customer_Salary, MAX(estimatedsalary) As
Maximum Estimated Customer Salary FROM churn modeling') # use triple quotes if you
want to spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#2 is %s' % (datetime.now()-start))
print("-----")
c.close()
```

```
#Query No#3
start=datetime.now()
c = connection.cursor()
c.execute('SELECT COUNT(customerid) FROM churn_modeling WHERE exited=1 AND
isactivemember=1') # use triple quotes if you want to spread your query across multiple
lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#3 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#4
start=datetime.now()
c = connection.cursor()
c.execute("SELECT (COUNT(customerid)*100/(SELECT COUNT(*) FROM churn_modeling ))
AS Total_Male_Percentage FROM churn_modeling WHERE gender='Male' ") # use triple
quotes if you want to spread your query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#4 is %s' % (datetime.now()-start))
print("----")
c.close()
#Query No#5
start=datetime.now()
c = connection.cursor()
c.execute('SELECT COUNT (customerid) AS COUNTERSUM FROM churn modeling GROUP BY
age, customerid HAVING age BETWEEN 20 AND 40 ') # use triple quotes if you want to spread
your query across multiple lines
for result in c:
```

print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.

print("")
print ('The time for Query No#5 is %s' % (datetime.now()-start))
print("")
c.close()
#Query No#6
start=datetime.now()
c = connection.cursor()
c.execute("SELECT COUNT(customerid), geography FROM churn_modeling WHERE geography='Spain' or geography='France' GROUP BY geography ") # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.
print("")
print ('The time for Query No#6 is %s' % (datetime.now()-start))
print("")
c.close()
#Query No#7
start=datetime.now()
c = connection.cursor()
c.execute('SELECT customerid, surname from churn_modeling WHERE NOT (hascrcard=0 AND isactivemember=0) AND (age<40) GROUP BY surname,customerid ORDER BY surname DESC') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.
print("")
print ('The time for Query No#7 is %s' % (datetime.now()-start))
print("")
c.close()

```
Query No#8
start=datetime.now()
c = connection.cursor()
c.execute("SELECT customerid, surname, geography FROM churn_modeling WHERE surname
LIKE 'A%' AND (geography IN('Germany', 'France')) GROUP BY
customerid, surname, geography ORDER BY customerid, surname DESC, geography DESC") #
use triple quotes if you want to spread your query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add, '-', row[2], etc.
print("-----")
print ('The time for Query No#8 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#9
start=datetime.now()
c = connection.cursor()
c.execute("SELECT customerid, surname FROM churn_modeling WHERE tenure IN ('2', '4', '6'
,'8') AND geography IN('Spain','Germany') AND age IN ('20', '30','40') GROUP BY
customerid, surname, tenure, geography, age ORDER BY
customerid, surname, tenure, geography, age") # use triple quotes if you want to spread your
query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#9 is %s' % (datetime.now()-start))
print("-----")
c.close()
```

```
#Query No#10
start=datetime.now()
c = connection.cursor()
c.execute("SELECT COUNT(customerid) FROM churn_modeling WHERE (gender='Male' AND
geography='Spain') OR(creditscore>500)") # use triple quotes if you want to spread your
query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#10 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#11
start=datetime.now()
c = connection.cursor()
c.execute("SELECT customerid,numofproducts FROM churn_modeling WHERE
(Balance>150000 AND age>30) OR (numofproducts IN('3','4')) GROUP BY
customerid, numofproducts ") # use triple quotes if you want to spread your query across
multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#11 is %s' % (datetime.now()-start))
print("----")
c.close()
#Query No#12
start=datetime.now()
c = connection.cursor()
c.execute('SELECT age,creditscore FROM churn modeling WHERE creditscore>500 GROUP
BY age, creditscore HAVING age>(Select AVG(age) FROM churn_modeling)') # use triple
quotes if you want to spread your query across multiple lines
for result in c:
```

print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.

print("")
print ('The time for Query No#12 is %s' % (datetime.now()-start))
print("")
c.close()
#Query No#13
start=datetime.now()
c = connection.cursor()
c.execute('SELECT MIN (Balance) AS Minimum_Balance,MAX (Balance) AS Maximum_Balance, AVG (Balance) AS Average_Balance, MIN (age) AS Minimum_age,MAX (age) AS Maximum_age, AVG (age) AS Average_age, MIN (creditscore) AS Minimum_creditscore,MAX (creditscore) AS Maximum_creditscore, AVG (creditscore) AS Average_creditscore, MIN (tenure) AS Minimum_tenure,MAX (tenure) AS Maximum_tenure, AVG (tenure) AS Average_tenure FROM churn_modeling GROUP BY Balance,age,creditscore,tenure') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.
print("")
print ('The time for Query No#13 is %s' % (datetime.now()-start))
print("")
c.close()
#Query No#14
start=datetime.now()
c = connection.cursor()
c.execute('SELECT geography,Balance,sum(Balance) OVER(PARTITION BY geography) AS TotalBalPerRegion FROM churn_modeling GROUP BY geography, Balance') # use triple quotes if you want to spread your query across multiple lines
for result in c:
print (result) # this only shows the first two columns. To add an additional column you'll need to add , '-', row[2], etc.

```
print("-----")
print ('The time for Query No#14 is %s' % (datetime.now()-start))
print("----")
c.close()
#Query No#15
start=datetime.now()
c = connection.cursor()
c.execute('SELECT customerid, estimated salary FROM churn modeling GROUP BY
customerid, estimatedsalary HAVING estimatedsalary > (SELECT AVG(estimatedsalary)
FROM churn_modeling )') # use triple quotes if you want to spread your query across
multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add, '-', row[2], etc.
print("----")
print ('The time for Query No#15 is %s' % (datetime.now()-start))
print("----")
c.close()
#Query No#16
start=datetime.now()
c = connection.cursor()
c.execute('SELECT customerid, surname FROM churn modeling GROUP BY
customerid, surname, Balance, creditscore HAVING Balance > (SELECT AVG(Balance ) FROM
churn_modeling ) AND creditscore >(SELECT_AVG(creditscore )FROM churn_modeling )') #
use triple quotes if you want to spread your query across multiple lines
for result in c:
 print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("----")
print ('The time for Query No#16 is %s' % (datetime.now()-start))
```

```
c.close()
#Query No#17
start=datetime.now()
c = connection.cursor()
c.execute('SELECT customerid, tenure, count(*) FROM churn_modeling GROUP BY
GROUPING SETS ( ( customerid), (tenure))') # use triple quotes if you want to spread your
query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print ('The time for Query No#17 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#18
start=datetime.now()
c = connection.cursor()
c.execute('SELECT customerid, SUM(Balance+estimatedsalary) as TotalMoney FROM
churn modeling GROUP BY customerid HAVING SUM(Balance+estimatedsalary)> 100000
ORDER BY TotalMoney DESC;') # use triple quotes if you want to spread your query across
multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#18 is %s' % (datetime.now()-start))
print("----")
c.close()
```

```
#Query No#19
start=datetime.now()
c = connection.cursor()
c.execute('select tenure, count(*) as total_tenure, count(*) * 1.0 / sum(count(*)) over () as
ratio from churn_modeling group by tenure') # use triple quotes if you want to spread your
query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("----")
print ('The time for Query No#19 is %s' % (datetime.now()-start))
print("-----")
c.close()
#Query No#20
start=datetime.now()
c = connection.cursor()
c.execute('SELECT customerid FROM churn_modeling WHERE creditscore> (SELECT
AVG(creditscore) FROM churn_modeling )') # use triple quotes if you want to spread your
query across multiple lines
for result in c:
  print (result) # this only shows the first two columns. To add an additional column you'll
need to add , '-', row[2], etc.
print("-----")
print ('The time for Query No#20 is %s' % (datetime.now()-start))
print("----")
```

c.close()

```
Apache Spark Queries – same code for 1 core or multi core only the line :
sc = SparkContext('local[*]', 'FinalProject')
changes from local[*] to local[1] for one core utilization
# ==== Apache Spark Queries ===
from datetime import datetime
from pyspark.sql import SparkSession
import sys
from pyspark import SparkContext,SparkConf
import matplotlib.pyplot as plt # plotting
import numpy as np # linear algebra
import pandas as pd
df = pd.read_csv("C:\\Users\\30694\\Downloads\\Churn_Modelling.csv")
def kill_current_spark_context():
  SparkContext.getOrCreate().stop()
kill_current_spark_context()
sc = SparkContext('local[*]', 'FinalProject')
spark=SparkSession.builder.appName("spark sql test2").getOrCreate()
spark_df = spark.createDataFrame(df)
 spark_df.registerTempTable("Table")
# Query No# 1
start=datetime.now()
query result=spark.sql("SELECT COUNT(CustomerID) AS ExitedCustomers FROM Table
WHERE Exited=1").show()
# query_result.show()
```

print ('The time for Query No#1 is %s' % (datetime.now()-start))

```
# Query No# 2
start=datetime.now()
query_result=spark.sql("SELECT AVG(EstimatedSalary) As
Average Estimated Customer Salary, MIN(EstimatedSalary) As
Minimum_Estimated_Customer_Salary, MAX(EstimatedSalary) As
Maximum_Estimated_Customer_Salary FROM Table")
print ('The time for Query No#2 is %s' % (datetime.now()-start))
# Query No# 3
start=datetime.now()
query result=spark.sql("SELECT COUNT(CustomerID) FROM Table WHERE Exited=1 AND
IsActiveMember=1")
print ('The time for Query No#3 is %s' % (datetime.now()-start))
# Query No# 4
start=datetime.now()
query result=spark.sql("SELECT (COUNT(CustomerID)*100/(SELECT COUNT(*) FROM Table
)) AS Total_Male_Percentage FROM Table WHERE Gender='Male'")
print ('The time for Query No#4 is %s' % (datetime.now()-start))
# Query No# 5
start=datetime.now()
query_result=spark.sql("SELECT COUNT (CustomerId) AS COUNTERSUM FROM Table GROUP
BY Age, CustomerId HAVING Age BETWEEN 20 AND 40 ")
print ('The time for Query No#5 is %s' % (datetime.now()-start))
# Query No# 6
start=datetime.now()
query_result=spark.sql("SELECT COUNT(CustomerId), Geography FROM Table WHERE
Geography='Spain' or Geography='France' GROUP BY Geography ")
print ('The time for Query No#6 is %s' % (datetime.now()-start))
# Query No# 7
start=datetime.now()
query result=spark.sql("SELECT CustomerId, Surname from Table WHERE NOT (
HasCrCard=0 AND IsActiveMember=0) AND (Age<40)GROUP BY Surname, CustomerId ORDER
BY Surname DESC ")
print ('The time for Query No#7 is %s' % (datetime.now()-start))
```

```
# Query No# 8
start=datetime.now()
query_result=spark.sql("SELECT CustomerId,Surname, Geography FROM Table WHERE
Surname LIKE 'A%' AND (Geography IN('Germany', 'France')) GROUP BY
CustomerId, Surname, Geography ORDER BY CustomerId, Surname DESC, Geography DESC")
print ('The time for Query No#8 is %s' % (datetime.now()-start))
# Query No# 9
start=datetime.now()
query_result=spark.sql("SELECT CustomerId,Surname FROM Table WHERE Tenure IN ('2', '4'
,'6' ,'8') AND Geography IN('Spain' ,'Germany') AND AGE IN ( '20', '30','40') GROUP BY
CustomerId, Surname, Tenure, Geography, Age ORDER BY
CustomerId, Surname, Tenure, Geography, Age ")
print ('The time for Query No#9 is %s' % (datetime.now()-start))
# Query No# 10
start=datetime.now()
query_result=spark.sql("SELECT COUNT(CustomerId) FROM Table WHERE (Gender='Male'
AND Geography='Spain') OR(CreditScore>500)")
print ('The time for Query No#10 is %s' % (datetime.now()-start))
# Query No# 11
start=datetime.now()
query_result=spark.sql("SELECT CustomerId,NumOfProducts FROM Table WHERE
(Balance>150000 AND Age>30) OR ( NumOfProducts IN( '3', '4')) GROUP BY
CustomerId, NumOfProducts ")
print ('The time for Query No#11 is %s' % (datetime.now()-start))
# Query No# 12
start=datetime.now()
query result=spark.sql("SELECT Age, CreditScore FROM Table WHERE CreditScore>500
GROUP BY Age, CreditScore HAVING Age>(Select AVG(Age) FROM Table)")
print ('The time for Query No#12 is %s' % (datetime.now()-start))
# Query No# 13
start=datetime.now()
query_result=spark.sql(""" SELECT MIN (Balance) AS Minimum Balance, MAX (Balance) AS
Maximum Balance, AVG (Balance) AS Average Balance,
```

```
MIN (Age) AS Minimum_Age, MAX (Age) AS Maximum_Age, AVG (Age) AS Average_Age,
MIN (CreditScore) AS Minimum_CreditScore, MAX (CreditScore) AS Maximum_CreditScore,
AVG (CreditScore) AS Average CreditScore,
 MIN (Tenure) AS Minimum Tenure, MAX (Tenure) AS Maximum Tenure, AVG (Tenure) AS
Average_Tenure
  FROM Table GROUP BY Balance, Age, CreditScore, Tenure""")
print ('The time for Query No#13 is %s' % (datetime.now()-start))
# Query No# 14
start=datetime.now()
query_result=spark.sql("SELECT Geography,Balance,sum(Balance) OVER(PARTITION BY
Geography ) AS TotalBalPerRegion FROM Table GROUP BY Geography, Balance")
print ('The time for Query No#14 is %s' % (datetime.now()-start))
# Query No# 15
start=datetime.now()
query_result=spark.sql("SELECT CustomerId,EstimatedSalary FROM Table GROUP BY
CustomerId, EstimatedSalary HAVING EstimatedSalary > (SELECT AVG(EstimatedSalary)
FROM Table )")
print ('The time for Query No#15 is %s' % (datetime.now()-start))
# Query No# 16
start=datetime.now()
query_result=spark.sql("SELECT CustomerId, Surname FROM Table GROUP BY
CustomerId, Surname, Balance, CreditScore HAVING Balance > (SELECT AVG(Balance) FROM
Table ) AND CreditScore >(SELECT_AVG(CreditScore )FROM Table )")
print ('The time for Query No#16 is %s' % (datetime.now()-start))
```

```
# Query No# 17
start=datetime.now()
query_result=spark.sql("SELECT CustomerId, Tenure, count(*) FROM Table GROUP BY
GROUPING SETS ( ( CustomerId), (Tenure))")
print ('The time for Query No#17 is %s' % (datetime.now()-start))
# Query No# 18
start=datetime.now()
query_result=spark.sql("""SELECT CustomerId,
    SUM(Balance+EstimatedSalary) as TotalMoney FROM Table
  GROUP BY CustomerId
HAVING SUM(Balance+EstimatedSalary) > 100000
ORDER BY TotalMoney DESC""")
print ('The time for Query No#18 is %s' % (datetime.now()-start))
# Query No# 19
start=datetime.now()
query_result=spark.sql("""select Tenure, count(*) as total_tenure,
    count(*) * 1.0 / sum(count(*)) over () as ratio
from Table
group by Tenure """)
print ('The time for Query No#19 is %s' % (datetime.now()-start))
Query No# 20
start=datetime.now()
query result=spark.sql("SELECT CustomerId FROM Table WHERE CreditScore> (SELECT
AVG(CreditScore) FROM Table )")
# print(query_result)
# for row in query_result:
# print(row)
print ('The time for Query No#20 is %s' % (datetime.now()-start))
```

Code for Data Visualization

```
import matplotlib.pyplot as plt # plotting
import numpy as np # linear algebra
import pandas as pd
import seaborn as sns
sns.set_context("paper", rc={"font.size":18,"axes.titlesize":18,"axes.labelsize":25})
df = pd.read_csv("C:\\Users\\30694\\Downloads\\Churn_Modelling.csv")
sns.set(rc={'figure.figsize':(11.7,8.27),"font.size":20,"axes.titlesize":20,"axes.labelsize":20},st
yle="white")
nRow, nCol = df.shape
print(f'There are {nRow} rows and {nCol} columns')
if pd.isnull(df).sum().all()==0:
  print ('There are no missing values')
else:
  print ('There are missing values')
for col in df:
  print(df[col].unique())
print(df['Exited'].value_counts())
labels = 'Exited', 'Retained'
sizes = [df.Exited[df['Exited']==1].count(), df.Exited[df['Exited']==0].count()]
```

```
explode = (0, 0.1)
fig1, ax1 = plt.subplots(figsize=(10, 8))
ax1.pie(sizes, explode=explode, labels=labels, autopct='%1.1f%%',
    shadow=True, startangle=90)
ax1.axis('equal')
plt.title("Proportion of customer churned and retained", size = 20)
plt.show()
# Get unique count for each variable
df.nunique()
fig, axarr = plt.subplots(2, 2, figsize=(20, 12))
sns.countplot(x='Geography', hue = 'Exited',data = df, ax=axarr[0][0])
sns.countplot(x='Gender', hue = 'Exited',data = df, ax=axarr[0][1])
sns.countplot(x='HasCrCard', hue = 'Exited',data = df, ax=axarr[1][0])
sns.countplot(x='IsActiveMember', hue = 'Exited',data = df, ax=axarr[1][1])
# Relations based on the continuous data attributes
fig, axarr = plt.subplots(3, 2, figsize=(20, 12))
sns.boxplot(y='CreditScore',x = 'Exited', hue = 'Exited',data = df, ax=axarr[0][0])
sns.boxplot(y='Age',x = 'Exited', hue = 'Exited',data = df , ax=axarr[0][1])
sns.boxplot(y='Tenure',x = 'Exited', hue = 'Exited',data = df, ax=axarr[1][0])
sns.boxplot(y='Balance',x = 'Exited', hue = 'Exited',data = df, ax=axarr[1][1])
sns.boxplot(y='NumOfProducts',x = 'Exited', hue = 'Exited',data = df, ax=axarr[2][0])
sns.boxplot(y='EstimatedSalary',x = 'Exited', hue = 'Exited',data = df, ax=axarr[2][1])
```