

PYSPARK CASE STUDY

Title: Online Banking Analysis

Data and Files Collected:

loan.csv

credit card.csv

txn.csv

In loandata.csv file

1.number of loans in each category

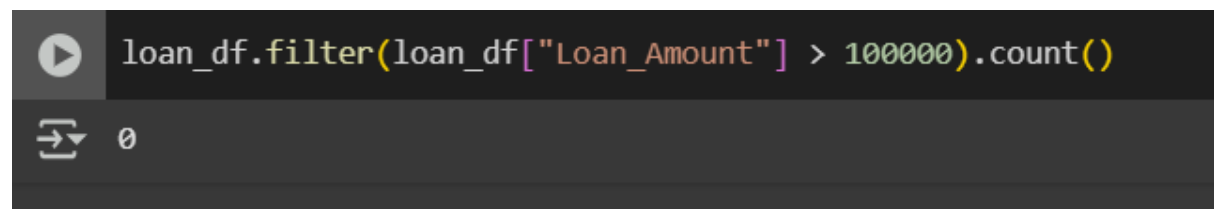
```
loan_df.groupBy("Loan_Category").count().show()
```

```
# 1. Number of loans in each category
loan_df.groupBy("Loan_Category").count().show()
```

Loan_Category	count
HOUSING	67
TRAVELLING	53
BOOK STORES	7
AGRICULTURE	12
GOLD LOAN	77
EDUCATIONAL LOAN	20
AUTOMOBILE	60
BUSINESS	24
COMPUTER SOFTWARES	35
DINNING	14
SHOPPING	35
RESTAURANTS	41
ELECTRONICS	14
BUILDING	7
RESTAURANT	20
HOME APPLIANCES	14

2.number of people who have taken more than 1 lack loan

```
loan_df.filter(loan_df["Loan_Amount"] > 100000).count()
```

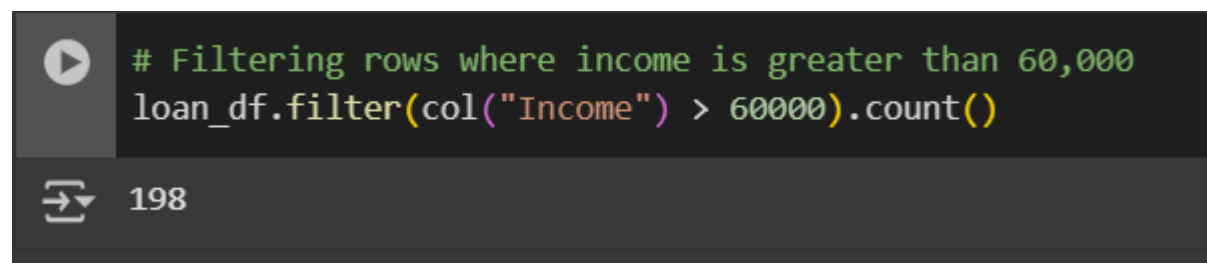


```
loan_df.filter(loan_df["Loan_Amount"] > 100000).count()
```

0

3.number of people with income greater than 60000 rupees

```
loan_df.filter(col("Income") > 60000).count()
```

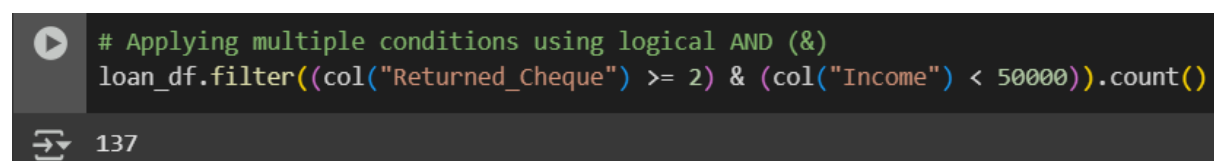


```
# Filtering rows where income is greater than 60,000  
loan_df.filter(col("Income") > 60000).count()
```

198

4.number of people with 2 or more returned cheques and income less than 50000

```
loan_df.filter((col("Returned_Cheque") >= 2) &  
(col("Income") < 50000)).count()
```



```
# Applying multiple conditions using logical AND (&  
loan_df.filter((col("Returned_Cheque") >= 2) & (col("Income") < 50000)).count()
```

137

5.number of people with 2 or more returned cheques and are single

```
loan_df.filter((col("Returned_Cheque") >= 2) &  
(col("Marital_Status") == "Single")).count()
```

```
# Filtering by returned cheques and marital status
loan_df.filter((col("Returned_Cheque") >= 2) & (col("Marital_Status") == "Single")).count()

0
```

6.number of people with expenditure over 50000 a month

```
loan_df.filter(col("Expenditure") > 50000).count()
```

```
# Filtering by returned cheques and marital status
loan_df.filter((col("Returned_Cheque") >= 2) & (col("Marital_Status") == "Single")).count()

0
```

7.number of members who are eligible for credit card

```
loan_df.filter((col("Use_Frequency") > 10) &
(col("Debt_Record") == "Good")).count()
```

```
# Example Filter: Eligible if Debt_Record is good and they use banking services frequently
loan_df.filter((col("Use_Frequency") > 10) & (col("Debt_Record") == "Good")).count()

0
```

In credit.csv file

1.credit card users in Spain

```
credit_df.filter(col("Geography") == "Spain").count()
```

```
# 1. Show number of credit card users located in Spain

credit_df.filter(col("Geography") == "Spain").count()

2477
```

2.number of members who are eligible and active in the bank

```
credit_df.filter((col("CreditScore") >= 650) &
(col("IsActiveMember") == 1)).count()
```

```
credit_df.filter((col("CreditScore") >= 650) & (col("IsActiveMember") == 1)).count()
```

```
2672
```

In Transactions file

1.Maximum withdrawal amount in transactions

```
txn_df.select(max("WITHDRAWAL_AMT").alias("Max_Wit
hdrawal_Amount")).show()
```

```
#1. Maximum Withdrawal Amount
txn_df.select(max("WITHDRAWAL_AMT").alias("Max_withdrawal_Amount")).show()
```

```
+-----+
|Max_Withdrawal_Amount|
+-----+
|          4.594475464E8|
+-----+
```

2.minimum withdrawal amount of an account in txn.csv

```
txn_df.select(min("WITHDRAWAL_AMT").alias("Min_With
drawal_Amount")).show()
```

```
#Minimum Withdrawal Amount
txn_df.select(min("WITHDRAWAL_AMT").alias("Min-Withdrawal_Amount")).show()
```

Min-Withdrawal_Amount
0.01

3.maximum deposit amount of an account

```
txn_df.select(max("DEPOSIT_AMT").alias("Max_Deposit_Amount")).show()
```

```
#Maximum Deposit Amount
txn_df.select(max("DEPOSIT_AMT").alias("Max_Deposit_Amount")).show()
```

Max_Deposit_Amount
5.448E8

4.minimum deposit amount of an account

```
txn_df.select(min("DEPOSIT_AMT").alias("Min_Deposit_Amount")).show()
```

```
# Minimum Deposit Amount
txn_df.select(min("DEPOSIT_AMT").alias("Min_Deposit_Amount")).show()
```

Min_Deposit_Amount
0.01

5.sum of balance in every bank account

```
txn_df.groupBy("Account_No") \
    .sum("BALANCE_AMT") \
    .withColumnRenamed("sum(BALANCE_AMT)",
        "Total_Balance") \.show()
```

#Sum of Balance Amount in Each Bank Account

```
txn_df.groupBy("Account_No") \
    .sum("BALANCE_AMT") \
    .withColumnRenamed("sum(BALANCE_AMT)", "Total_Balance") \
    .show()
```

Account_No	Total_Balance
409000438611	-2.49486577068339...
1196711	-1.60476498101275E13
1196428	-8.1418498130721E13
409000493210	-3.27584952132095...
409000611074	1.615533622E9
409000425051	-3.77211841164998...
409000405747	-2.43108047067000...
409000362497	-5.2860004792808E13
409000493201	1.0420831829499985E9
409000438620	-7.12291867951358...

6.Number of transaction on each date

```
txn_df.groupBy("VALUE_DATE") \
    .agg(count("*").alias("Transaction_Count")) \
    .orderBy("VALUE_DATE") \
    .show()
```

Number of Transactions Happening on Each Date

```
txn_df.groupBy("VALUE_DATE") \
    .agg(count("*").alias("Transaction_Count")) \
    .orderBy("VALUE_DATE") \
    .show()
```

VALUE_DATE	Transaction_Count
1-Apr-17	1
1-Aug-15	75
1-Aug-16	85
1-Aug-17	65
1-Aug-18	144
1-Dec-15	96
1-Dec-16	106
1-Dec-17	45
1-Dec-18	97
1-Feb-16	97
1-Feb-17	81
1-Feb-18	87
1-Feb-19	79
1-Jan-15	3
1-Jan-16	59
1-Jan-18	53
1-Jan-19	57
1-Jul-15	25
1-Jul-16	111
1-Jul-17	243

only showing top 20 rows

7. List of customers with withdrawal amount more than 1 lakh

```
txn_df.filter(col("WITHDRAWAL_AMT") > 100000) \
    .select("Account_No", "WITHDRAWAL_AMT",
"VALUE_DATE") \
    .orderBy(col("WITHDRAWAL_AMT").desc()) \
    .show()
```

```
# Customers With Withdrawal Amount Greater Than ₹1,00,000

txn_df.filter(col("WITHDRAWAL_AMT") > 100000) \
    .select("Account_No", "WITHDRAWAL_AMT", "VALUE_DATE") \
    .orderBy(col("WITHDRAWAL_AMT").desc()) \
    .show()
```

Account_No	WITHDRAWAL_AMT	VALUE_DATE
1196711	4.594475464E8	26-Jun-18
1196711	4.482072231E8	26-May-17
409000438620	4.0E8	8-Mar-16
409000425051	3.54E8	31-Oct-18
1196711	2.671403184E8	20-Jun-15
409000438611	2.4E8	31-Mar-16
1196711	2.021E8	3-May-16
409000438620	2.0E8	11-Mar-16
409000438620	2.0E8	18-Mar-16
1196711	2.0E8	21-Oct-15
1196711	2.0E8	3-Oct-15
409000405747	1.7E8	30-Jan-16
1196711	1.54E8	24-Sep-15
1196711	1.5E8	22-Aug-15
1196711	1.5E8	17-Oct-15
1196711	1.5E8	7-Apr-16
1196428	1.5E8	13-Apr-16
409000362497	1.413662392E8	16-Aug-16
409000362497	1.317762365E8	14-Sep-16
409000362497	1.316962119E8	10-Oct-16

only showing top 20 rows