

# Financial Data Analysis

Power BI | DAX | Analysis

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# Agenda

Credit card Usage and financial metrics analysis



**We are making use of the Power BI's very rich data analysis feature that is DAX.**



**DATASRT LINK.**

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# Introduction and Objective ::

- In this analysis, we will explore credit card usage patterns and essential financial metrics to gain a comprehensive understanding of customer behavior, credit utilization, and delinquency risk within a banking institution.
- Using Power BI with DAX functions, we will calculate key metrics that provide a dynamic view of customer interactions and financial health. These calculations inform actionable KPIs, which help in assessing customer engagement and identifying high-risk segments.
- Our insights will drive targeted strategies to enhance customer retention and strengthen overall financial performance, positioning the bank for sustainable growth and risk mitigation.

# Data Overview :

This analysis combines Credit Card and Customer datasets to examine clients' credit behavior, demographics, and risk factors.

## 1. Credit Card Dataset:

- **Key Info:** Includes details like Credit\_Limit, Total\_Revolving\_Bal, Total\_Trans\_Amt, and Avg\_Utilization\_Ratio.
- **Focus:** Tracks spending habits, credit utilization, and risk indicators (e.g., Delinquent\_Acc).
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## 2. Customer Dataset:

- **Key Info:** Demographic and financial attributes such as Income, Education\_Level, and Cust\_Satisfaction\_Score.
- **Focus:** Provides insights into customer profiles and satisfaction.

# KEY ANALYSIS CONDUCTED :

- **Financial Metrics:** Running totals, moving averages, and growth rates for transactions.
- **Risk Indicators:** Delinquency rates, credit risk scores, and high-risk client flags.
- **Behavioral Insights:** Retention, churn indicators, and transaction patterns.
- **Correlations:** Examined income vs. credit limit and loan approvals vs. credit limits.

## Running Total of Credit Card Transactions

```
running_total =  
  
CALCULATE(SUM(credit_card[Total_Trans_Amt]), FILTER(all(credit_card),  
credit_card[Week_Start_Date] <= max(credit_card[Week_Start_Date])))
```

**Calculate the 4-week moving average of the creditLimit for each client.**

```
moving_average =  
  
var window_4_weeks = DATESINPERIOD('calendar'[Date],MAX('calendar'[Date]),-28,DAY)  
  
var sales = CALCULATE(SUM(credit_card[Credit_Limit]),window_4_weeks)  
  
var distinct_week = CALCULATE(DISTINCTCOUNT('calendar'[weeknum]),window_4_weeks)  
  
return DIVIDE(sales,distinct_week)
```

## Calculate the mom% growth and wow% growth on transaction amount.

mom%growth =

```
var previous_month = CALCULATE(SUM(credit_card[Total_Trans_Amt]),DATEADD('calendar'[Date],-1,MONTH))
```

```
return DIVIDE(SUM(credit_card[Total_Trans_Amt])-previous_month,previous_month,0)
```

wow%growth =

```
var previous_week = CALCULATE(SUM(credit_card[Total_Trans_Amt]),DATEADD('calendar'[Date],-7,DAY))
```

```
return DIVIDE(SUM(credit_card[Total_Trans_Amt])-previous_week,previous_week,0)
```



## Calculate Customer Acquisition Cost (CAC) as a Ratio of Transaction Amount.

```
ac_trans_amt = DIVIDE(  
    SUM(credit_card[Customer_Acq_Cost]),  
    SUM(credit_card[Total_Trans_Amt]))
```

**Calculate the yearly average of  
avg\_utilization\_ratio for all  
clients.**

```
avg_utilization_ratio =  
| AVERAGE(credit_card[Avg_Utilization_Ratio])/  
| DISTINCTCOUNT(credit_card[current_year])
```

**Calculate the percentage of  
Interest\_Earned compared to  
Total\_Revolving\_Bal for each client.**

```
interest_by_revolving_bal = DIVIDE(  
    SUM(credit_card[Interest_Earned]),  
    SUM(credit_card[Total_Revolving_Bal]),0)
```

## Calculate Top 5 Clients by Total Transaction Amount.

```
top_5_clients_by_transaction_amt =  
  
TOPN(5,  
SUMMARIZE(credit_card,credit_card[Client_Num],"total_amount",SUM(credit_card[Total_Trans_Amt])),  
[total_amount],DESC)
```

**Identify clients whose  
Avg\_Utilization\_Ratio  
exceeds 80%.**

```
avg_uti_exceedes_80% = IF([avg_utilization_ratio] > 0.8 , True, FALSE)
```

**Customer Churn Indicator: Create a KPI that flags clients who have not made any transactions (Total\_Trans\_Amt = 0) in the last 6 months.**

```
no_trans_in_last_6_months =  
  
var months_6 = CALCULATE(SUM(credit_card[Total_Trans_Amt]),  
    | | | DATESINPERIOD('calendar'[Date],MAX('calendar'[Date]),-6,MONTH))  
|  
return IF(ISBLANK(months_6),True,False)
```

**Delinquency Rate: Calculate the percentage of clients with Delinquent\_Acc > 0.**

```
delinquency_rate =  
  
var delinquent_accounts = CALCULATE(COUNTROWS(credit_card), credit_card[Delinquent_Acc] > 0)  
  
var total_accounts = COUNTROWS(credit_card)  
  
return DIVIDE(delinquent_accounts, total_accounts, 0)
```

**Credit Risk Score: Create a score for each client based on their Avg\_Utilization\_Ratio, Delinquent\_Acc, and Total\_Revolving\_Bal.**

Normalising the column  
Total\_revolving balance to --->  
values between 0-1

```
normalize_revolving_blanca =  
  
var min_value = MIN(credit_card[Total_Revolving_Bal])  
  
var max_value = MAX(credit_card[Total_Revolving_Bal])  
  
return DIVIDE(credit_card[Total_Revolving_Bal]-min_value,max_value-min_value,0)
```

Giving weightages to each of  
the columns : --->  
Avg\_Utilization\_ratio = 50%

```
credit_risk_score =  
|  
0.5 * credit_card[Avg_Utilization_Ratio] +  
0.3 * credit_card[Delinquent_Acc] +  
0.2 * credit_card[normalize_revolving_blanca]
```



**Income vs Credit Limit Correlation:**  
**Show the correlation between Income**  
**and Credit\_Limit for all clients.**

**In the Home Tab in Report view :**

**Go to Quick measures -> Under Calculation -> In mathematical operations (Correlation coefficient) ->**

**In category : Client\_Num**

**On X-Axis : Summation of Income**

**On Y axis : Summation of credit\_card**

**0.13**

Income and Credit\_Limit  
correlation for Client\_Num

**Average Customer Satisfaction Score by Credit Card Category: Calculate the average Cust\_Satisfaction\_Score by Card\_Category.**

```
average_score_by_card_category =
```

```
SUMMARIZE(credit_card,credit_card[Card_Category],  
"Avg score",ROUND(AVERAGE('customer data'[Cust_Satisfaction_Score]),2))
```

**Loan Approval vs Credit Limit: Analyze how Credit\_Limit affects Personal\_loan approval by calculating the average credit limit for clients with and without loans.**

```
personal_loan_yes = CALCULATE(AVERAGE(credit_card[Credit_Limit]),'customer data'[Personal_loan] = "yes")
```

```
1 personal_loan_no = CALCULATE(AVERAGE(credit_card[Credit_Limit]),'customer data'[Personal_loan] = "no")
```

**High Risk Clients Flag: Create a flag for clients whose Total\_Revolving\_Bal exceeds 90% of their Credit\_Limit and who have a high Avg\_Utilization\_Ratio.**

```
flag_exceeds_90%_of_credit_limit =  
  
var credit_limit_90 = credit_card[Credit_Limit] * 0.9  
  
return if(credit_card[Total_Revolving_Bal] > credit_limit_90  
&& credit_card[Avg_Utilization_Ratio] > 0.5, True, False)
```

## CONCLUSION :

- The analysis provided deep insights into customer spending, identifying patterns and usage trends that inform better financial management and credit policies.
- Key metrics, calculated with DAX in Power BI, highlighted significant factors such as high-risk segments, credit utilization rates, and delinquency risk.
- Actionable insights emerged to support strategies for reducing credit defaults, optimizing credit offerings, and improving customer retention.
- These findings enable the bank to enhance customer satisfaction and loyalty, while aligning with goals for sustainable growth and risk management.

# Let's Connect !!

**power BI | DAX | Data Analysis**

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