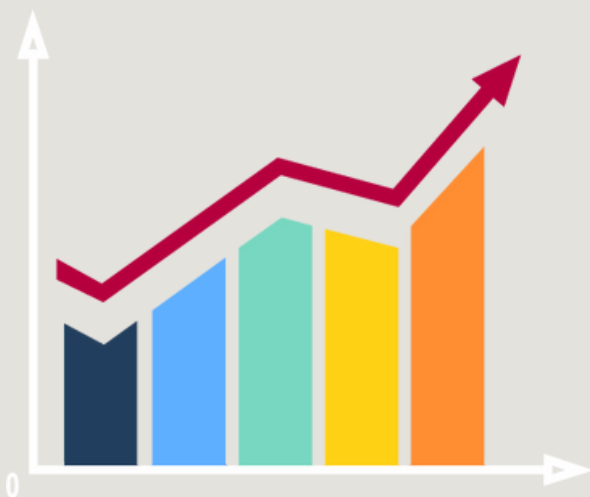


Analyzing Banking Trends: Customer Transactions and Regional Impact

C H E L S E A
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Analyzing Banking Trends: Customer Transactions and Regional Impact

PROJECT DESCRIPTION

In the ever-evolving world of banking and finance, understanding customer behavior and the regional impact of transactions plays a crucial role in decision-making and strategic planning.

This project aims to explore and analyze the vast troves of transaction data to gain valuable insights into customer behavior patterns and their implications on different world regions.

OBJECTIVE

The objective of this project is to explore customer transaction data and identify trends that may impact regional economies and financial systems.



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TECHNICAL TOOLS

- Python
- Microsoft SQL Server

DATA SOURCES

The project leverages three key tables that provide valuable information for analysis:

1. **world_regions**: This table contains data on various world regions and their corresponding codes and names. It serves as a reference to categorize customers based on their regional affiliation.
2. **user_nodes**: This table holds details about consumers' banking nodes, including their unique consumer IDs, associated region IDs, node IDs, start dates, and end dates. This data enables us to identify the specific banking nodes to which customers are connected and their duration of association.
3. **user_transaction**: This table is a repository of customer transactions, containing data such as consumer IDs, transaction dates, types of transactions, and transaction amounts. Analyzing this data allows us to uncover patterns in customer spending and financial behaviors.

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```
In [26]: import pandas as pd
import warnings
warnings.filterwarnings("ignore")

In [27]: # Function to read the CSV file into a DataFrame
def read_csv():
    # read the user_nodes.csv file using pandas library and return it
    df = pd.read_csv('user_nodes.csv')
    return df

In [28]: # Function to check for null (missing) values in the DataFrame
def check_null_values():
    # do not edit the predefined function name
    df = read_csv()
    # Check for null values using the isnull() method and sum them for each column
    null_values = df.isnull().sum()
    return null_values

In [29]: # check for null values
check_null_values()

Out[29]: id_          0
area_id_         0
node_id_         0
act_date         0
deact_date        0
has_loan         0
is_act           0
dtype: int64

In [30]: # Function to check for duplicate rows in the DataFrame
def check_duplicates():
    # do not edit the predefined function name
    df = read_csv()
    # Calculate the number of duplicate rows using the duplicated() method and sum them
    duplicates = df.duplicated().sum()
    return duplicates

In [31]: # check for duplicates
check_duplicates()

Out[31]: 143
```

Each data source was cleaned, checked for null and duplicate values.

Duplicates and null values were then dropped and files were exported to a csv.

```
In [32]: # Function to drop duplicate rows from the DataFrame
def drop_duplicates():
    # do not edit the predefined function name
    df = read_csv()
    # Drop duplicate rows using the drop_duplicates() method with inplace=True
    df.drop_duplicates(inplace=True)
    return df

In [33]: # drop duplicates
drop_duplicates().head()

Out[33]:   id  area_id  node_id  act_date  deact_date  has_loan  is_act
0    1         3         4  02-01-2020  03-01-2020         1         0
1    2         3         5  03-01-2020  17-01-2020         0         1
2    3         5         4  27-01-2020  18-02-2020         0         0
3    4         5         4  07-01-2020  19-01-2020         1         1
4    5         3         3  15-01-2020  23-01-2020         0         1

In [34]: def data_cleaning():
df = drop_duplicates()

# Step 3: Drop specified columns from the DataFrame("has_loan", "is_act")
df.drop(columns=["has_loan", "is_act"], inplace=True)
#Rename columns names id_,area_id_,node_id_,act_date,deact_date to consumer_id,region_id,node_id,start_date,end_date
df.rename(columns={"id_": "consumer_id", "area_id_": "region_id", "node_id_": "node_id",
                  "act_date": "start_date", "deact_date": "end_date"}, inplace=True)
df.to_csv('user_nodes_cleaned.csv', index=False)
return df

In [37]: data_cleaning().info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 3500 entries, 0 to 3499
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype
---  -
0   consumer_id  3500 non-null   int64
1   region_id   3500 non-null   int64
2   node_id     3500 non-null   int64
3   start_date  3500 non-null   object
4   end_date    3500 non-null   object
dtypes: int64(3), object(2)
memory usage: 164.1+ KB
```

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List all regions along with the number of users assigned to each region.

```
SELECT wr.region_name, COUNT(DISTINCT(un.consumer_id)) AS num_user
FROM bank_trends.dbo.world_regions AS wr
LEFT JOIN bank_trends.dbo.user_nodes AS un
ON wr.region_id = un.region_id
GROUP BY wr.region_name
ORDER BY num_user DESC;
```

region_name	num_user
United States	110
Europe	105
Australia	102
Asia	95
Africa	88
China	0
Russia	0

Retrieve the total number of transactions for each region.

```
SELECT wr.region_name, COUNT(ut.consumer_id) AS total_transactions
FROM bank_trends.dbo.user_transaction AS ut
INNER JOIN bank_trends.dbo.user_nodes AS un ON ut.consumer_id = un.consumer_id
INNER JOIN bank_trends.dbo.world_regions AS wr ON un.region_id = wr.region_id
GROUP BY wr.region_name
ORDER BY total_transactions DESC
```

This code shows the total number of transactions for each region. United States has the most transactions and Africa with the least amount of transactions.

	region_name	total_transactions
1	United States	9107
2	Europe	8806
3	Australia	8414
4	Asia	7952
5	Africa	6797

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What is the unique count and total amount for each transaction type?

```
SELECT transaction_type, COUNT(DISTINCT(transaction_type)) AS unique_count, SUM(transaction_amount) AS total_amount
FROM bank_trends.dbo.user_transaction
GROUP BY transaction_type
```

	transaction_type	unique_count	total_amount
1	deposit	1	1359168
2	purchase	1	806537
3	withdrawal	1	793003

This query shows the total amount of deposits, purchases, and withdrawals made.

Write a query to find the total deposit amount for each region in the user_transaction table.

```
SELECT wr.region_name, SUM(ut.transaction_amount) AS total_deposit_amount
FROM bank_trends.dbo.user_transaction AS ut
INNER JOIN bank_trends.dbo.user_nodes AS un ON ut.consumer_id = un.consumer_id
INNER JOIN bank_trends.dbo.world_regions AS wr ON un.region_id = wr.region_id
WHERE ut.transaction_type = 'deposit'
GROUP BY wr.region_name
ORDER BY total_deposit_amount DESC
```

This query shows customers in Europe have deposited the most amount of funds.

region_name	total_deposit_amount
Europe	2120503
United States	2078069
Asia	1853110
Australia	1851703
Africa	1610791

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Calculate the total amount deposited for the top 10 users in the "Europe" region.

```
SELECT TOP 10 un.consumer_id, SUM(ut.transaction_amount) AS total_deposits
FROM bank_trends.dbo.user_transaction AS ut
INNER JOIN bank_trends.dbo.user_nodes AS un ON ut.consumer_id = un.consumer_id
INNER JOIN bank_trends.dbo.world_regions AS wr ON un.region_id = wr.region_id
WHERE wr.region_name = 'Europe' AND ut.transaction_type = 'deposit'
GROUP BY un.consumer_id
ORDER BY total_deposits DESC;
```

Top 10 consumers and the total amount deposited

consumer_id	total_deposits
197	56679
281	53606
110	46039
500	42329
240	40999
76	40467
186	37660
212	37079
456	36911
205	35819

Find the regions with the highest number of nodes assigned to them.

```
SELECT wr.region_name, COUNT(un.node_id) AS num_nodes
FROM bank_trends.dbo.user_nodes AS un
INNER JOIN bank_trends.dbo.world_regions AS wr ON un.region_id = wr.region_id
GROUP BY wr.region_name
HAVING COUNT(un.node_id) = (SELECT TOP 1 COUNT(un2.node_id)
FROM bank_trends.dbo.user_nodes AS un2
GROUP BY un2.region_id
ORDER BY COUNT(un2.node_id) DESC
);
```

region_name	num_nodes
United States	770

The United States has the highest number of nodes assigned.

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Which month had the highest transaction amount for deposits?

```
SELECT DATEPART(MONTH, ut.transaction_date) AS transaction_month,  
       SUM(ut.transaction_amount) AS total_deposit_amount  
FROM bank_trends.dbo.user_transaction AS ut  
WHERE ut.transaction_type = 'deposit'  
GROUP BY DATEPART(MONTH, ut.transaction_date)  
ORDER BY total_deposit_amount DESC;
```

We saw the most deposits in January

transaction_month	total_deposit_amount
1	437894
3	390103
2	357040
4	174131

KEY TAKEAWAYS

- Most prominent transaction type is deposit across all regions.
- United States and Europe stand out with the most substantial deposit volume.
- January is the most popular month for deposits, possibly due to the holiday season.
- There are currently no users assigned in China and Russia, potential areas for growth.