



# DST Banking Case Study

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# Problem Statements

1. The Loans Division of ABC Bank want to know the accounts who are likely to default in repaying the loans when the contract ends.
2. Identify whether any demographic factors is showing strong association with loan defaulters.
3. Identify if there is any relation between the type of credit card issued to a disponentwith his average credit balance maintained in transaction account.
4. Visual dashboard to show -
  - Distribution of client age as of their first transaction date
  - Distribution of client age by gender from their first transaction date
  - Distribution of clients by districts and average salary
  - Distribution of clients by average transaction amount
  - Any other interesting visual representation of the findings

# Technology Used

## Python

Pandas , Seaborn and sklearn

## IDE

Anaconda Distribution , jupyter and spyder

## Database

MS Access : As the data set are in TXT format and Tableau has good integration with Access TXT

Excel : Used it for a simple date formatting calculation on Transaction table

## Dashboard and Visual Report

Tableau Online

Dashboard link : <https://eu-west-1a.online.tableau.com/#/site/sudipbairagi/views/BankCaseStudy/Dashboard1>

# Problem 1 : Predict Loan Defaulter

## Logical Steps and code snippet

- Import all TXT data into MS Access and Table
- Write SQL Query to join Loan and District table and generate out in CSV : LoanToDemographDistinctMap.xls
- Panda Code : BankingDefaulterPrediction.py

### Algorithm used: **K Nearest Neighbour (KNN)**

This algorithm is selected to classify A,B,C and D type Loan Payers with District demographic data.

```
df = pd.read_excel('LoanToDemographDistinctMap.xls')
```

**Data Cleansing** : To replace NaN with 0 and Scale the data using a StandardScaler and remove non-numeric column from DataFrame.

```
df.fillna(value=0, inplace=True)
```

```
scaled_DemoGraphNumeric = scaler.transform(DemoGraphNumeric)
```

```
scaled_DemoGraphNumeric_feature = pd.DataFrame(scaled_DemoGraphNumeric, columns=DemoGraphNumeric.columns)
```

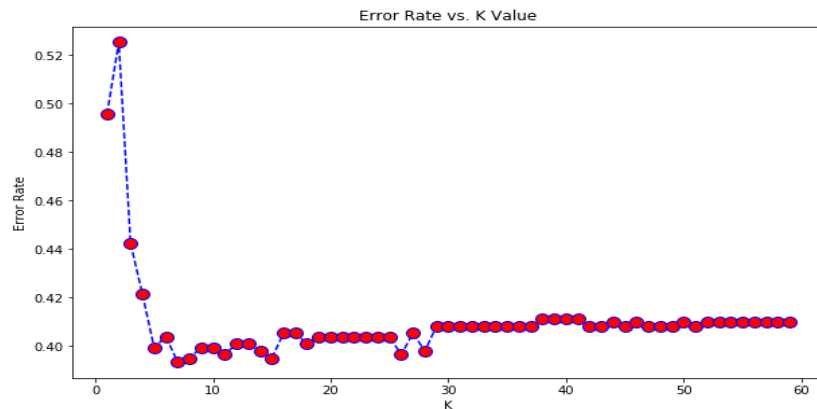
### **Prediction Model : KNN with K=10**

```
knn = KNeighborsClassifier(n_neighbors=10)
```

```
knn.fit(scaled_DemoGraphNumeric_feature, df['status'])
```

### **Find Best Fit K value**

```
error_rate = []  
for i in range(1,60):  
    knn = KNeighborsClassifier(n_neighbors=i)  
    knn.fit(scaled_DemoGraphNumeric_feature, df['status'])  
    pred_i = knn.predict(X_test)  
    error_rate.append(np.mean(pred_i != y_test))  
plt.figure(figsize=(10,6))  
plt.plot(range(1,60), error_rate, color='blue', linestyle='dashed', marker='o',  
         markerfacecolor='red', markersize=10)
```



# Problem 1 : Classification Report and Confusion Matrix

Confusion Matrix :

```
[[ 44  0 158  0]
 [  6  0  24  0]
 [ 37  0 362  0]
 [  5  0  40  0]]
```

Classification Report:

	precision	recall	f1-score	support
A	0.48	0.22	0.30	202
B	0.00	0.00	0.00	30
C	0.62	0.91	0.74	399
D	0.00	0.00	0.00	45
avg / total	0.51	0.60	0.52	676

Result :

- With optimal value  $K=10$  , the Classification report could not determine the bad defaulter Type B .
- The reason may be because of number of records available for Defaulter against good loan payer .

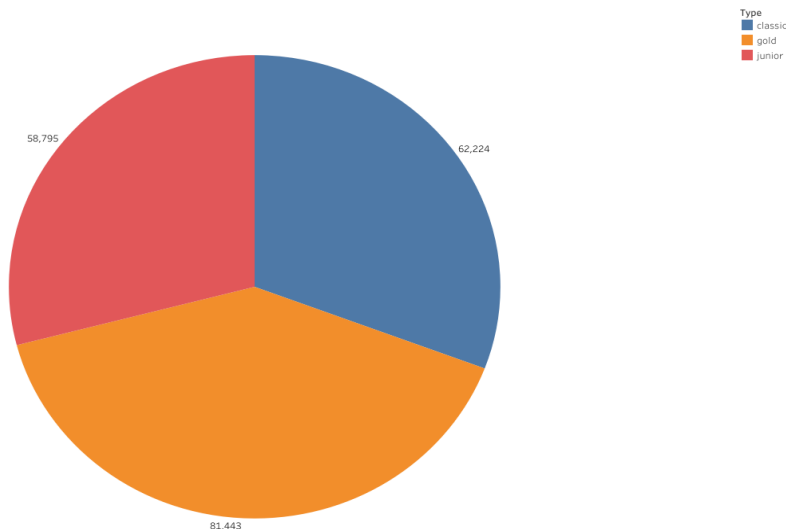
# Problem Card to Avg Loan Balance

## Logical Steps and code snippet

- Import all TXT data into MS Access and Table
- Write SQL Query to find the Last Balance from the Transaction Table : AvgBalancePerAccountQuery

Report **Tableau Online link** : <https://eu-west-1a.online.tableau.com/#/site/sudipbairagi/views/BankCaseStudy/BalanceMaintainedPerCardType?:iid=1>

Avg Balance Maintained Per Card Type



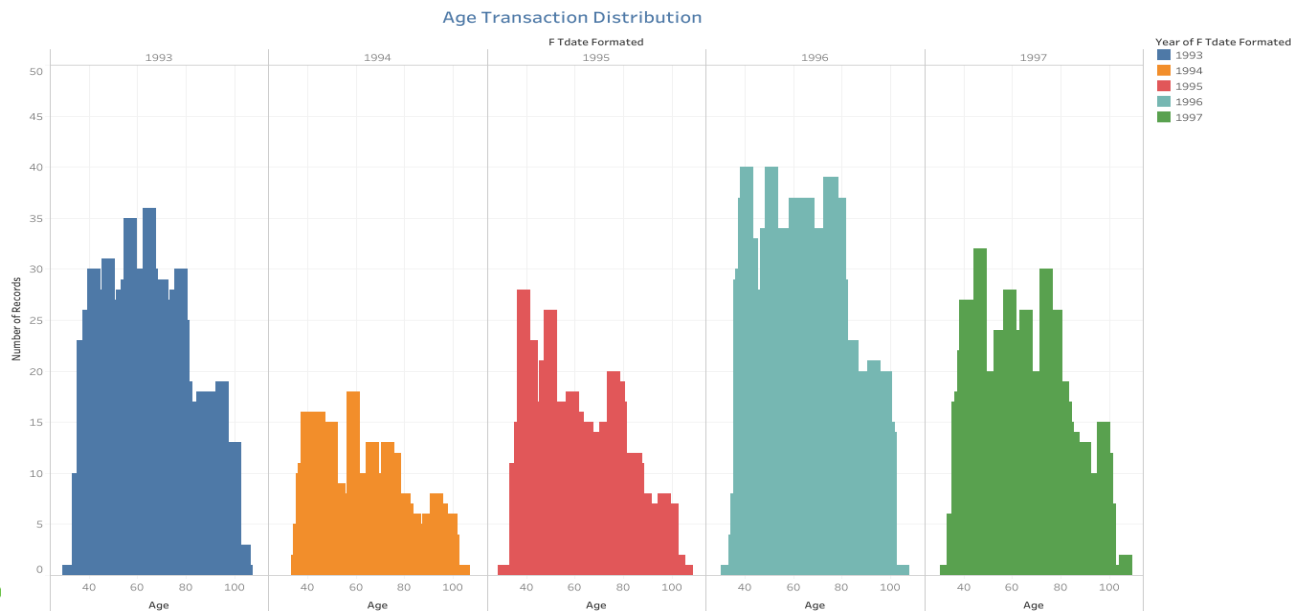
The Report in Tableau is interactive one. Putting cursor on the report shows more data and can be drilled down .

# Client age as distribution of their first transaction date

## Logical Steps and code snippet

- Import all TXT data into MS Access and Table
- Write SQL Query to find the Age of each Client : **ClientBirthDateResult**
- The Date Number fields is also need to decode to find Male and Female (50+M)
- **Tableau Calculated Fields** are created to calculate the Age using Tableau in-build functions .

**Tableau Online link** :<https://eu-west-1a.online.tableau.com/#/site/sudipbairagi/views/BankCaseStudy/AgeTransactionDistribution?:iid=1>



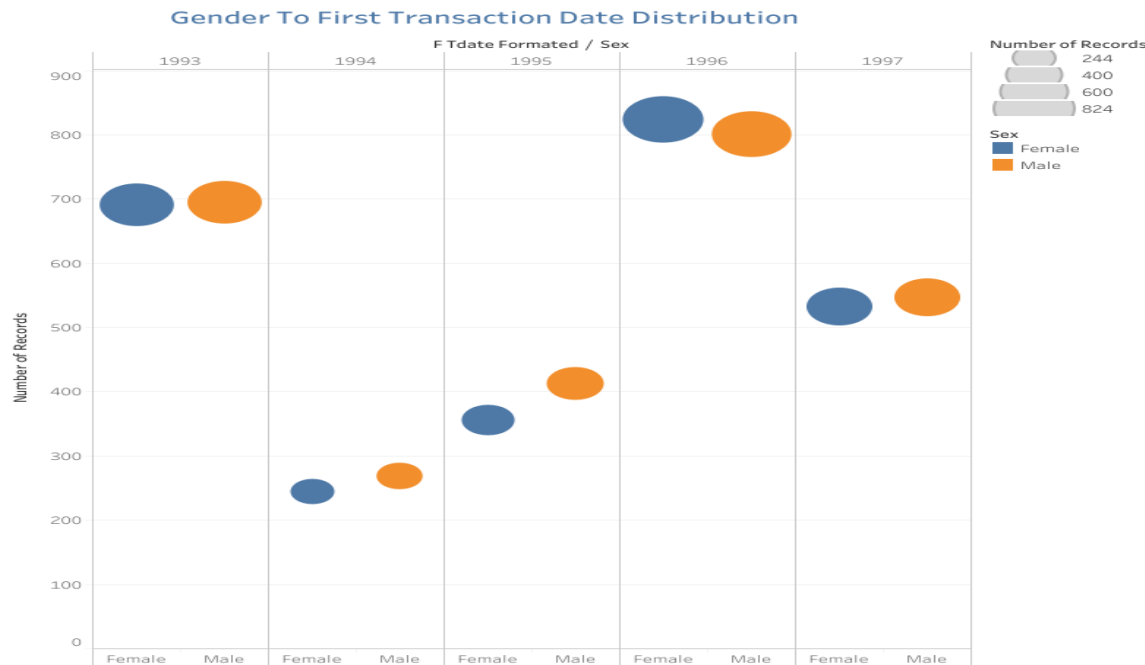
The Report in Tableau is interactive one. Putting cursor on the report shows more data and can be drilled down . In 1994 it is least amount . In 1996 it is the Highest amount . 40-50 Years age is having highest First Transaction Counts.

# Client Gender as distribution of their first transaction date

## Logical Steps and code snippet

- Import all TXT data into MS Access and Table
- Write SQL Query to find the Age of each Client : **ClientBirthDateResult**
- **Write MS Function to calculate Gender from Birth Number** : **ClientBirthDate** Module -> **clientSex**
- The Date Number fields is also need to decode to find Male and Female (50+M)

**Tableau Online link** :<https://eu-west-1a.online.tableau.com/#!/site/sudipbairagi/views/BankCaseStudy/GenderFirstTransactionDistribution?iid=1>



The Report in Tableau is interactive one. Putting cursor on the report shows more data and can be drilled down . In 1996 Female had more first transaction than Male.

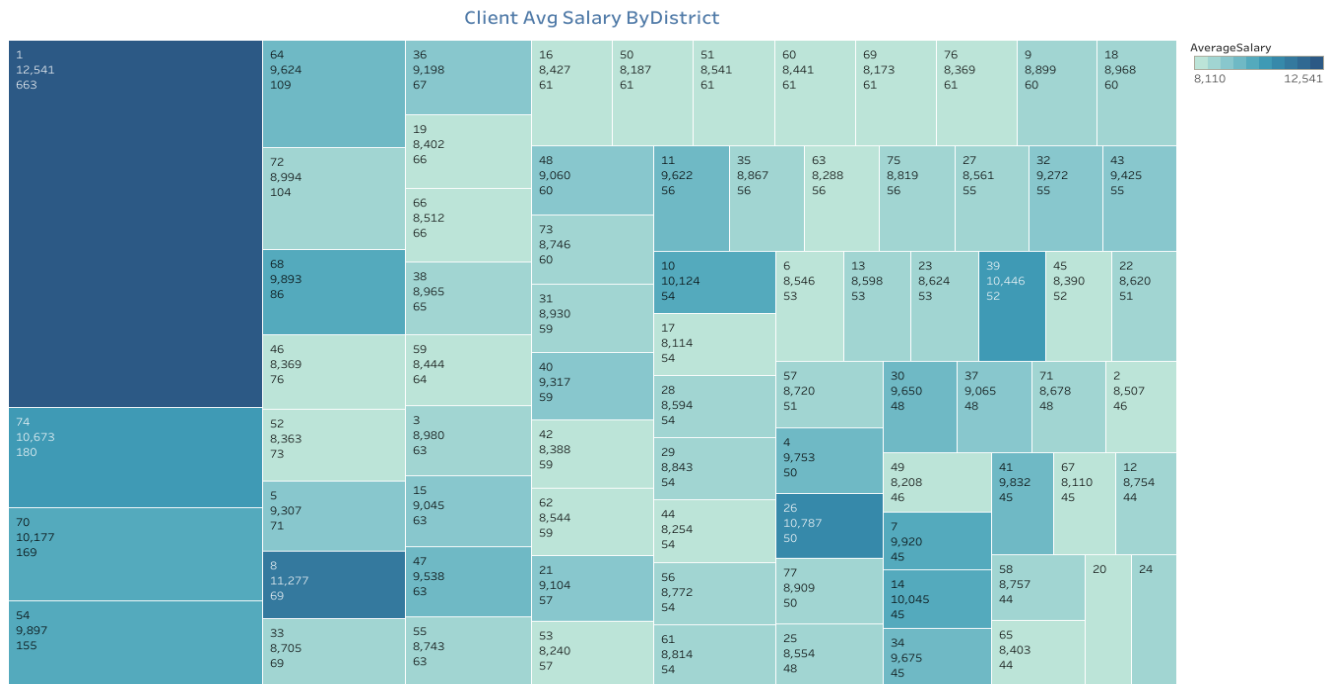


# Distribution of clients by districts and average salary

## Logical Steps and code snippet

- Import all TXT data into MS Access and Table
- Write SQL Query to find the join District

**Tableau Online link :** <https://eu-west-1a.online.tableau.com/#!/site/sudipbairagi/views/BankCaseStudy/ClientAvgSalaryByDistrict?.iid=1>

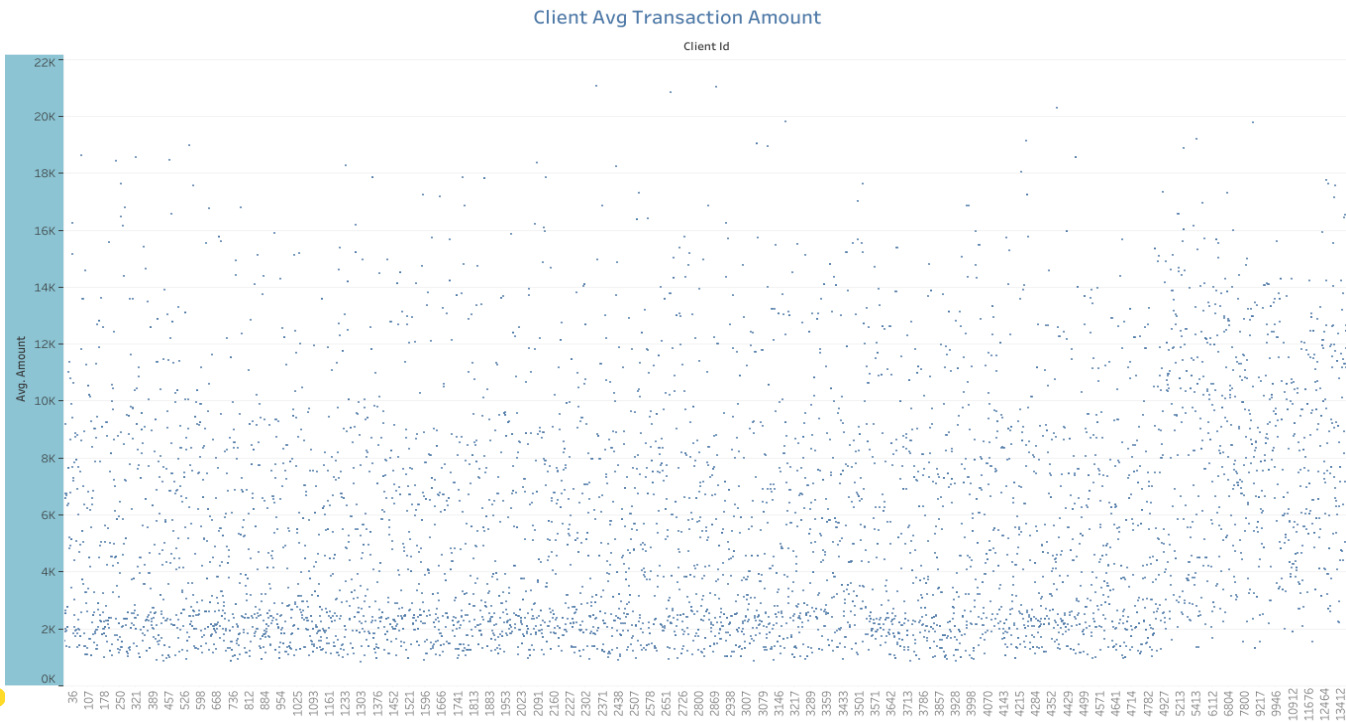


# Distribution of clients by average transaction amount

## Logical Steps and code snippet

- Import all TXT data into MS Access and Table
- Write SQL Query to find the join Transaction with Disp and Client

**Tableau Online link :** <https://eu-west-1a.online.tableau.com/#!/site/sudipbairagi/views/BankCaseStudy/ClientAvgTrnAmount?:iid=1>



The Report in Tableau is interactive one. Putting cursor on the report shows more data and can be drilled down . As there is huge amount of transaction we can see most of the avg amount is around 2K .

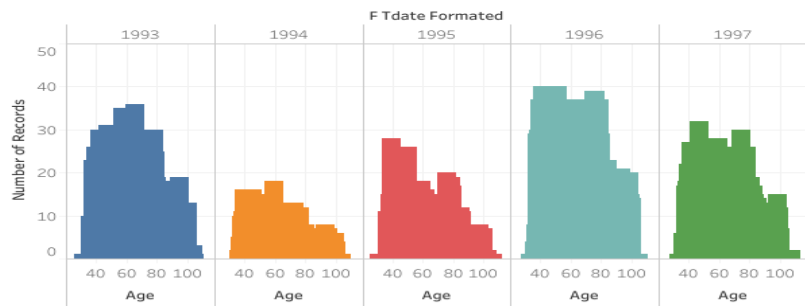
# Interactive Dashboard - Tableau

## Logical Steps and code snippet

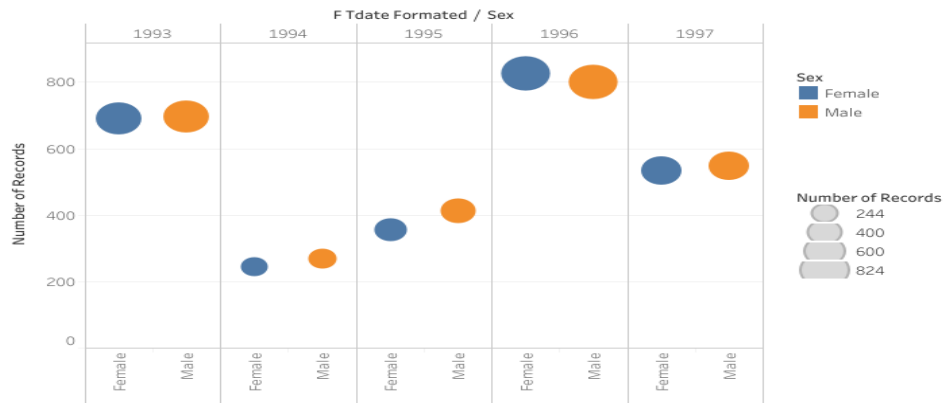
- Create a dashboard including all Worksheet in Tableau

Tableau Online link :<https://eu-west-1a.online.tableau.com/#!/site/sudipbairagi/views/BankCaseStudy/Dashboard1?:iid=1>

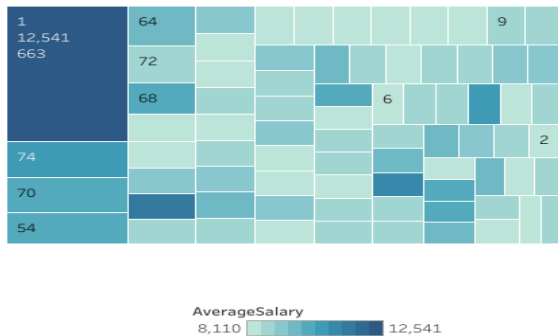
Age Transaction Distribution



Gender To First Transaction Date Distribution



Client Avg Salary ByDistrict



Client Avg Transaction Amount

