#### **Subject: Data Mining and Business Intelligence (2170715)**

**Open Ended Problem** 

Title: Churn Modeling Using Random Forest (classification problem)

**Group No: 1** 

#### **Enrollment No. Name**

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### Importing the libraries

```
In [13]:
```

```
import numpy as np
import pandas as pd
```

### **Data Preprocessing**

#### Importing the dataset

```
In [4]:
```

```
dataset = pd.read_csv('Churn_Modelling.csv')
X = dataset.iloc[:, 3:-1].values
y = dataset.iloc[:, -1].values
print(X)
print(y)

[[619 'France' 'Female' ... 1 1 101348.88]
[608 'Spain' 'Female' ... 0 1 112542.58]
[502 'France' 'Female' ... 1 0 113931.57]
...
[709 'France' 'Female' ... 0 1 42085.58]
[772 'Germany' 'Male' ... 1 0 92888.52]
[792 'France' 'Female' ... 1 0 38190.78]]
[1 0 1 ... 1 1 0]
```

#### **Encoding categorical data**

#### 1. Label Encoding the "Gender" column

```
In [5]:
```

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
X[:, 2] = le.fit_transform(X[:, 2])
print(X)

[[619 'France' 0 ... 1 1 101348.88]
  [608 'Spain' 0 ... 0 1 112542.58]
  [502 'France' 0 ... 1 0 113931.57]
...
  [709 'France' 0 ... 0 1 42085.58]
  [772 'Germany' 1 ... 1 0 92888.52]
  [792 'France' 0 ... 1 0 38190.78]]
```

#### 2. One Hot Encoding the "Geography" column

```
In [6]:
```

```
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [1])], remaind
er='passthrough')
X = np.array(ct.fit_transform(X))
print(X)

[[1.0 0.0 0.0 ... 1 1 101348.88]
[0.0 0.0 1.0 ... 0 1 112542.58]
[1.0 0.0 0.0 ... 1 0 113931.57]
...
[1.0 0.0 0.0 ... 0 1 42085.58]
[0.0 1.0 0.0 ... 1 0 92888.52]
[1.0 0.0 0.0 ... 1 0 38190.78]]
```

#### **Feature Scaling**

```
In [7]:
```

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X = sc.fit transform(X)
print(X)
[ 0.99720391 -0.57873591 -0.57380915 ... 0.64609167 0.97024255
   0.021886491
 [-1.00280393 -0.57873591 \ 1.74273971 \ \dots \ -1.54776799 \ 0.97024255
   0.21653375]
 [0.99720391 - 0.57873591 - 0.57380915 \dots 0.64609167 - 1.03067011
   0.2406869 1
 [0.99720391 - 0.57873591 - 0.57380915 ... -1.54776799 0.97024255
 -1.008643081
 [-1.00280393 1.72790383 -0.57380915 ... 0.64609167 -1.03067011
 -0.125230711
 [0.99720391 - 0.57873591 - 0.57380915 ... 0.64609167 - 1.03067011
  -1.0763697611
```

## Splitting the dataset into the Training set and Test set

```
In [8]:
```

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
m_state = 0)
```

# Training the Random Forest Classification model on the Training set

```
In [9]:
```

```
from sklearn.ensemble import RandomForestClassifier
classifier = RandomForestClassifier(n_estimators = 1000, criterion = 'entropy',
random_state = 0)
classifier.fit(X_train, y_train)
```

#### Out[9]:

RandomForestClassifier(criterion='entropy', n\_estimators=1000, rando
m state=0)

#### **Prediction**

```
In [10]:
```

```
y_pred = classifier.predict(X_test)
```

### **Making the Confusion Matrix**

```
In [11]:
```

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
print(cm)
```

```
[[1525 70]
[ 199 206]]
```