

# Market Analysis in Banking Domain

## (Source Code)

```
// hdfs dfs -copyToLocal /user/gkesapragadateksystems/Project/dataset_bank-  
full.csv ./
```

```
// spark-shell
```

```
// 1. Load data and create Spark data frame
```

```
val lines = sc.textFile("/user/gkesapragadateksystems/Project/dataset_bank-  
full.csv")
```

```
val bank = lines.map(x => x.split(";"))
```

```
val bfields = bank.mapPartitionsWithIndex { (idx, iter) => if (idx == 0)  
iter.drop(1) else iter }
```

```
//Define Class for the schema
```

```
case class Bank(age:Int, job:String, marital:String, education:String, dft:String,  
balance:Int,
```

```
housing:String, loan:String, contact:String, day:Int, month: String, duration:Int,  
campaign:Int, pdays:Int,
```

```
previous:Int, poutcome:String, y:String)
```

```
val bankrdd = bfields.map(  
x => Bank(x(0).replaceAll("\\\"", "").toInt, x(1).replaceAll("\\\"", ""),  
x(2).replaceAll("\\\"", ""), x(3).replaceAll("\\\"", ""), x(4).replaceAll("\\\"", ""),  
x(5).toInt, x(6).replaceAll("\\\"", ""),  
x(7).replaceAll("\\\"", ""), x(8).replaceAll("\\\"", ""), x(9).toInt,  
x(10).replaceAll("\\\"", ""), x(11).toInt, x(12).toInt, x(13).toInt, x(14).toInt,  
x(15).replaceAll("\\\"", ""),
```

```
x(16).replaceAll("\\", ""))
```

```
val df = bankrdd.toDF()
```

```
df.registerTempTable("bank")
```

```
//spark.conf.set("spark.sql.crossJoin.enabled", "true")
```

**// 2. Give marketing success rate. (No. of people subscribed / total no. of entries)**

```
val sqlContext = new org.apache.spark.sql.SQLContext(sc)
```

```
val success = sqlContext.sql("select (a.subscribed/b.total)*100 as success_percent  
from (select count(*) as subscribed from bank where y='yes') a,(select count(*) as  
total from bank) b").show()
```

**// 2a Give marketing failure rate**

```
val failure = sqlContext.sql("select (a.not_subscribed/b.total)*100 as  
failure_percent from (select count(*) as not_subscribed from bank where y='no')  
a,(select count(*) as total from bank) b").show()
```

**// 3. Maximum, Mean, and Minimum age of average targeted customer**

```
df.select(max($"age")).show()
```

```
df.select(min($"age")).show()
```

```
df.select(avg($"age")).show()
```

```
// df.select("age").summary().show()
```

#### **// 4. Check quality of customers by checking average balance, median balance of customers**

```
df.select(avg($"balance")).show()
```

```
val median = sqlContext.sql("SELECT percentile_approx(balance, 0.5) FROM bank").show()
```

#### **// 5. Check if age matters in marketing subscription for deposit**

```
val age = sqlContext.sql("select age, count(*) as number from bank where y='yes' group by age order by number desc").show()
```

#### **// 6. Check if marital status mattered for subscription to deposit.**

```
val marital = sqlContext.sql("select marital, count(*) as number from bank where y='yes' group by marital order by number desc").show()
```

```
//df.groupBy($"y".alias("Did the customer subscribed")).agg(count($"marital").alias("marital count")).show()
```

#### **// 7. Check if age and marital status together mattered for subscription to deposit scheme**

```
val age_marital = sqlContext.sql("select age, marital, count(*) as number from bank where y='yes' group by age,marital order by number desc").show()
```

**// 8. Do feature engineering for column—age and find right age effect on campaign**

```
import scala.reflect.runtime.universe
import org.apache.spark.SparkConf
import org.apache.spark.SparkContext
import org.apache.spark.sql.DataFrame
import org.apache.spark.sql.SQLContext
import org.apache.spark.sql.functions.mean

val ageRDD = sqlContext.udf.register("ageRDD",(age:Int) => {
  if (age < 20)
    "Teen"
  else if (age > 20 && age <= 32)
    "Young"
  else if (age > 33 && age <= 55)
    "Middle Aged"
  else
    "Old"
})

//Replacing old “age” column with new “age” column
val banknewDF = df.withColumn("age",ageRDD(df("age")))
banknewDF.registerTempTable("bank_new")
```

**//Running a query to see the age group which subscribed the most. We see it's 'Middle-Aged'**

```
val age_target = sqlContext.sql("select age, count(*) as number from bank_new  
where y='yes' group by age order by number desc ").show()
```

**//Pipeline**

```
import org.apache.spark.ml.feature.StringIndexer
```

```
val ageInd = new StringIndexer().setInputCol("age").setOutputCol("ageIndex")
```

```
//Fitting the model
```

```
var strIndModel = ageInd.fit(banknewDF)
```

```
strIndModel.transform(banknewDF).select("age", "ageIndex").show(5)
```

**So we can conclude from the Feature Engineering that It is the 'Middle Aged' people between age 33 and 55 who should be the targeted customers as they subscribe the most**