MARKET ANALYSIS IN BANKING DOMAIN

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Setup

```
bankRDD: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[2] at map at <console>:24
scala> bankRDD.coalesce(1).saveAsTextFile("./Data/BankClean")
 scala> val bankDF = spark.read.
| option("header",true).
| option("inferSchema",true).
| csv("./dat/BankClean/part-00000")
| bankDF: org.apache.spark.sql.DataFrame = [age: int, job: string ... 15 more fields]
scala> bankDF.show(5)
 |age|
                   job|marital|education|default|balance|housing|loan|contact|day|month|duration|campaign|pdays|previous|poutcome| y|
 | 58| management|married| tertiary|
| 44| technician| single|secondary|
| 33|entrepreneur|married|secondary|
| 47| blue-collar|married| unknown|
| 33| unknown| single| unknown|
                                                                                                                      261
151
76
                                                                  29
2
                                                                            yes| no|unknown|
yes| yes|unknown|
                                                                                                    5| may|
5| may|
5| may|
5| may|
                                                                                                                                                          0| unknown| no|
0| unknown| no|
                                                               1506
                                                                             yes| no|unknown|
no| no|unknown|
                                                                                                                      92 |
198 |
                                                                                                                                                          0| unknown| no|
0| unknown| no|
                                                       no i
only showing top 5 rows
scala> bankDF.count
res2: Long = 45211
scala> bankDF.createOrReplaceTempView("bank")
```

Campaign Performance

- Success Rate 11.7%
- Failure Rate 88.3%

Customers Age Statistics

- Minimum 18
- Maximum 95
- Average 40.94

```
scala> spark.sql("""
| SELECT
| MIN(age) A5 min_age,
| MAX(age) A5 max_age,
| ROUND(AVG(age), 1) A5 avg_age
| FROM
| bank
| ""').show
| min_age|max_age|avg_age|
| 18| 95| 40.9|
```

Customer Quality Check (Balance)

 Due to skewness, we can conclude that most of the customers belong to lower than average income strata.

Does Age matter?

• Due to significantly low difference between average age of subscribers vs non-subscribers, we can conclude that age is **not** significant factor for campaign success.

Does Marital status matter?

Single customers
 have higher average
 chance to respond
 positively,
 concluding that
 marital status does
 matter.

```
| Select | Spark.sql(""" | Select | ROUND((allsubscribers.count/all.count)*190.2) AS overall_success_rate, | ROUND((allsubscribers.count/alldivorced.count)*190.2) AS divorced_success_rate, | ROUND((marriedsubscribers.count/allmarried.count)*100.2) AS married_success_rate, | ROUND((marriedsubscribers.count/allsingle.count)*100.2) AS single_success_rate, | ROUND((marriedsubscribers.count/allsingle.count)*100.2) AS single_success_rate, | ROUND((marriedsubscribers.count/allsingle.count)*100.2) AS single_success_rate, | ROUND((married)*100.2) AS single_success_rate, | ROUND((married)*100.2) AS single_success_rate, | ROUND((married)*100.2) AS single_success_rate, | ROUND((married)*100.2) AS count FROM bank, | ROUND((married)*100.2) AS count FROM bank, | ROUND((married)*100.2) Alldivorced, | Married, | ROUND((married)*100.2) AS count FROM bank, | ROUND((married)*100.2) A
```

Does Age + Marital status matter?

 (Old + Married or Divorced)
 customers give
 significantly higher
 positive reponse to
 subscription deposit
 scheme.

```
(SELECT COUNT(*) AS count FROM bankAgeGroupMarital WHERE age_group="middle_age') allmarried, 
(SELECT COUNT(*) AS count FROM bankAgeGroupMarital wHERE ym'yes' Allo age_group='old') oldsubscribers, 
(SELECT COUNT(*) AS count FROM bankAgeGroupMarital wHERE age_group='old') allold="").show
    |youngadult|middleage| old|
                            12.16| 9.21|33.63|
(SELECT COUNT(*) AS count FROM bankAgeGroupMarital WHERE marital='married' AND age_group='old') alloldmarried""").show
    Ivoungadult married|middleage married|old married
                                                                     9.21
                                                                                                                                                         8.591
 scala> spark.sql("""SELECT
                                 spark.sql(***SELECT GUNIT(*) AS count FROM bankAgeGroupMartial WHERE martial='divorced' AND age_group='old' AND age_group='old' AND age_group-'old' AND age_group-'old
   |youngadult_divorced|middleage_divorced|oldage_divorced|
                                                                       10.56
                                                                                                                                                             10.14
                                                                                                                                                                                                                                         38.94
scala> spark.sql("""SELECT
                                  spark.sql(***SELECT COUNT(*) AS count FROM bankageforeupMarttal WHERE marttal='single' AND age_group='old') altoldsagesingle.gep' (SELECT COUNT(*) AS count FROM bankageforeupMarttal WHERE marttal='single' AND age_group-'old') altoldsagesingle.gep' (SELECT COUNT(*) AS count FROM bankageforeupMarttal WHERE marttal='single' AND age_group-'old') altoldsagesingle.gep' And age_group-'old') altoldsagesingle.gep' And age_group-'old' And age_group-'ol
```

Age for target marketing

Old age customers
 i.e 60+ years

END