

# Data Cleaning & Integrity Procedures

In [3]: *# Import necessary libraries*

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

In [4]: *# Read the file*

```
df = pd.read_csv("train.csv", delimiter=";")
df.head()
```

Out[4]:

	age	job	marital	education	default	balance	housing	loan	cont
--	-----	-----	---------	-----------	---------	---------	---------	------	------

0	58	management	married	tertiary	no	2143	yes	no	unkn
1	44	technician	single	secondary	no	29	yes	no	unkn
2	33	entrepreneur	married	secondary	no	2	yes	yes	unkn
3	47	blue-collar	married	unknown	no	1506	yes	no	unkn
4	33	unknown	single	unknown	no	1	no	no	unkn

In [5]: *# Rows & Columns of data*

```
df.shape
```

Out[5]: (45211, 17)

In [6]: *# Distribution of Values & Data types*

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 45211 entries, 0 to 45210
Data columns (total 17 columns):
#   Column      Non-Null Count  Dtype
---  -
0   age         45211 non-null  int64
1   job         45211 non-null  object
2   marital     45211 non-null  object
3   education   45211 non-null  object
4   default     45211 non-null  object
5   balance     45211 non-null  int64
6   housing     45211 non-null  object
7   loan        45211 non-null  object
8   contact     45211 non-null  object
9   day         45211 non-null  int64
10  month       45211 non-null  object
11  duration    45211 non-null  int64
12  campaign    45211 non-null  int64
13  pdays      45211 non-null  int64
14  previous    45211 non-null  int64
15  poutcome    45211 non-null  object
16  y           45211 non-null  object
dtypes: int64(7), object(10)
memory usage: 5.9+ MB
```

In [7]: *# Checking for Missing(null) values*

```
df.isnull().sum()
```

```
Out[7]: age         0
        job         0
        marital     0
        education   0
        default     0
        balance     0
        housing     0
        loan        0
        contact     0
        day         0
        month       0
        duration    0
        campaign    0
        pdays      0
        previous    0
        poutcome    0
        y           0
        dtype: int64
```

In [8]: *# Descriptive statistics for numerical columns (detects outliers or un*

```
df.describe()
```

Out [8]:

	age	balance	day	duration	campaign
<b>count</b>	45211.000000	45211.000000	45211.000000	45211.000000	45211.000000
<b>mean</b>	40.936210	1362.272058	15.806419	258.163080	2.763841
<b>std</b>	10.618762	3044.765829	8.322476	257.527812	3.098021
<b>min</b>	18.000000	-8019.000000	1.000000	0.000000	1.000000
<b>25%</b>	33.000000	72.000000	8.000000	103.000000	1.000000
<b>50%</b>	39.000000	448.000000	16.000000	180.000000	2.000000
<b>75%</b>	48.000000	1428.000000	21.000000	319.000000	3.000000
<b>max</b>	95.000000	102127.000000	31.000000	4918.000000	63.000000

In [19]: *# Check for duplicate rows*

```
duplicate_rows = df.duplicated()
print("Total duplicate rows:", duplicate_rows.sum())
```

Total duplicate rows: 0

## Data Validity Check

In [36]: 

```
unknown_contact = df[df['contact'] == "unknown"]
unknown_contact.head()
```

Out [36]:

	age	job	marital	education	default	balance	housing	loan	contact
<b>0</b>	58	management	married	tertiary	no	2143	yes	no	unknown
<b>1</b>	44	technician	single	secondary	no	29	yes	no	unknown
<b>2</b>	33	entrepreneur	married	secondary	no	2	yes	yes	unknown
<b>3</b>	47	blue-collar	married	unknown	no	1506	yes	no	unknown
<b>4</b>	33	unknown	single	unknown	no	1	no	no	unknown

In [38]: 

```
unknown_new = unknown_contact[unknown_contact['campaign'] == 32]
unknown_new.head()
```

Out [38]:

	age	job	marital	education	default	balance	housing	loan	
<b>3331</b>	50	entrepreneur	married	primary	no	461	yes	no	u
<b>3483</b>	59	management	married	tertiary	no	2319	yes	no	u
<b>3529</b>	53	blue-collar	married	secondary	no	1140	yes	no	u
<b>4020</b>	42	self-employed	married	tertiary	no	1932	yes	no	u
<b>8238</b>	38	admin.	married	primary	no	0	yes	no	u