

Week #9 Deliverables PDF Document

Team member's details:

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Problem description

- Implementing the solutions in Week 8 deliverables
- Click Here to View Week 8 Deliverables
- <u>Url:</u>

https://docs.google.com/document/d/1JtSPZYQzz qRKpLK50nXwF5J-DQuR4AkAWmd0DUxa_x4/edit ?usp=drivesdc

Data cleansing and transformation done on the data.

- Prior to cleansing the columns had the following skewness:

Column: age Skewness: 0.6993

Interpretation: Moderately Positively Skewed

Column: balance Skewness: 6.5942

Interpretation: Highly Positively Skewed

Column: day

Skewness: 0.0946

Interpretation: Approximately Symmetric (Slightly Positive)

Column: duration Skewness: 2.7715

Interpretation: Highly Positively Skewed

Column: campaign Skewness: 4.7423

Interpretation: Highly Positively Skewed

Column: pdays Skewness: 2.7162

Interpretation: Highly Positively Skewed

Column: previous Skewness: 5.8733

Interpretation: Highly Positively Skewed

['balance', 'duration', 'campaign', 'pdays', 'previous']

- After cleansing the data the columns had the following skewness:

Column: age Skewness: 0.6993

Interpretation: Moderately Positively Skewed

Column: balance Skewness: 0.1212

Interpretation: Approximately Symmetric (Slightly Positive)

Column: day

Skewness: 0.0946

Interpretation: Approximately Symmetric (Slightly Positive)

Column: duration Skewness: -0.4677

Interpretation: Approximately Symmetric (Slightly Negative)

Column: campaign Skewness: 0.7649

Interpretation: Moderately Positively Skewed

Column: pdays Skewness: 1.6881

Interpretation: Highly Positively Skewed

Column: previous Skewness: 2.2485

Interpretation: Highly Positively Skewed

['pdays', 'previous']

- Checking for missing values, with code, we were able to figure out there are no null values:

```
import pandas as pd
df = pd.read_csv('bank.csv', sep=';')
missing_values = df.isnull().any(axis=1).sum() #it will check inside the columns of the dataframes
print (missing_values)
```

- However in the pdays column null values seem to be encoded by the number -1 because you can't contact someone after minus one days.
- This will be handled during construction of the model by creating a binary feature. When we make the model- whenever the model reads a -1 it will be interpreted as a false.

Try at least 2 techniques to clean the data

imputing any techniques.

- We are looking for outliers in the following columns: age, balance, duration, campaign, pdays and previous.
- Outliers techniques solution: for most numerical columns we impute the mean. For pdays and previous we ignore values that encode false.

```
Outliers in column 'age':
                                        Outliers in column 'age':
Number of outliers: 38
                                        Number of outliers: 38
Percentage of outliers: 0.84%
                                        Percentage of outliers: 0.84%
Outliers in column 'balance':
                                        Outliers in column 'balance':
Number of outliers: 506
                                        Number of outliers: 156
Percentage of outliers: 11.19%
                                        Percentage of outliers: 3.45%
Outliers in column 'day':
                                        Outliers in column 'day':
Number of outliers: 0
                                        Number of outliers: 0
Percentage of outliers: 0.00%
                                        Percentage of outliers: 0.00%
Outliers in column 'duration':
                                        Outliers in column 'duration':
Number of outliers: 330
                                        Number of outliers: 34
Percentage of outliers: 7.30%
                                        Percentage of outliers: 0.75%
Outliers in column 'campaign':
                                        Outliers in column 'campaign':
Number of outliers: 318
                                        Number of outliers: 0
Percentage of outliers: 7.03%
                                        Percentage of outliers: 0.00%
Outliers in column 'pdays':
                                        Outliers in column 'pdays':
Number of outliers: 7
                                        Number of outliers: 119
Percentage of outliers: 0.15%
                                        Percentage of outliers: 2.63%
Outliers in column 'previous':
                                        Outliers in column 'previous':
Number of outliers: 34
                                        Number of outliers: 0
Percentage of outliers: 0.75%
                                        Percentage of outliers: 0.00%
                                           - Low percentage of
- High percentage of
                                                outliers after we
   outliers before
```

imputed the mean.

Github Repo link

https://github.com/cralph31/Data-Glacier-Final-Group-Project-Weeks-7-12-Deliver ables