# Machine Learning (CCS4340)

# Lab 03: Mean or Median Imputation

Mean / median imputation consists in replacing missing values by the variable's mean or median.

In this Lab, we will perform mean and median imputation utilizing pandas, Scikit-learn and Feature-engine.

```
import pandas as pd

# to split the data sets:
from sklearn.model_selection import train_test_split

# to impute missing data with sklearn:
from sklearn.impute import SimpleImputer
from sklearn.compose import ColumnTransformer

# to impute missing data with Feature-engine:
from feature_engine.imputation import MeanMedianImputer
```

#### **Load Data**

```
data = pd.read_csv("credit_approval_uci.csv")
In [2]:
         data.head()
Out[2]:
           A1
                 A2
                       A3 A4 A5 A6 A7
                                            A8
                                                  A9 A10 A11 A12 A13
                                                                           A14 A15 target
            b 30.83 0.000
                                                                         202.0
                                                                                  0
         0
                                           1.25
                            u
                                    W
            a 58.67 4.460
                                        h 3.04
                                                                           43.0
                                                                                560
                            u
                                g
                                    q
         2
                                        h NaN NaN NaN
                                                                       g 280.0
                                                                                824
            a 24.50
                     NaN
                                                             0
                                                                                         1
                            u
                                    q
            b 27.83 1.540
                                           3.75
                                                                       q 100.0
                                                                                  3
           b 20.17 5.625
                                        v 1.71
                                                        f
                                                             0
                                                                       s 120.0
                                                                                  0
                                                                                         1
```

#### Split data in train and test sets

```
0.008282
        Α1
Out[4]:
               0.022774
        Α3
               0.140787
        Α4
               0.008282
        Α5
               0.008282
        Α6
               0.008282
        Α7
               0.008282
               0.140787
        Α9
               0.140787
               0.140787
        A10
        A11
               0.000000
        A12
               0.000000
        A13
               0.000000
        A14
               0.014493
        A15
               0.000000
        dtype: float64
```

### Select the variables to impute

```
In [5]: numeric_vars = X_train.select_dtypes(exclude="0").columns.to_list()
numeric_vars

Out[5]: ['A2', 'A3', 'A8', 'A11', 'A14', 'A15']
```

# Mean / median imputation with pandas

pd.fillna https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.fillna.html

```
In [6]: # Learn the variables median values:
         median_values = X_train[numeric_vars].median().to_dict()
         median_values
Out[6]: {'A2': 28.835, 'A3': 2.75, 'A8': 1.0, 'A11': 0.0, 'A14': 160.0, 'A15': 6.0}
In [7]: # Replace missing data by the median:
         X_train = X_train.fillna(value = median_values) # Fill NA/NaN values using the spe
         X test = X test.fillna(value = median values)
In [8]: # Corroborate absense of missing values:
         X_train[numeric_vars].isnull().sum()
         #X_test[numeric_vars].isnull().sum()
Out[8]:
        Α3
               0
        Α8
               0
        A11
               0
        A14
               0
        A15
        dtype: int64
```

#### Mean / median imputation with Scikitlearn

SimpleImputer https://scikit-

learn.org/stable/modules/generated/sklearn.impute.SimpleImputer.html

```
In [16]: # Split data into train and test set:
          X_train, X_test, y_train, y_test = train_test_split(data.drop("target", axis=1), delicated.
          numeric_vars
Out[16]: ['A2', 'A3', 'A8', 'A11', 'A14', 'A15']
In [17]: # Make a list with the non-numerical variables:
          remaining_vars = [var for var in X_train.columns if var not in numeric_vars]
          remaining vars
Out[17]: ['A1', 'A4', 'A5', 'A6', 'A7', 'A9', 'A10', 'A12', 'A13']
In [18]: # Set up the imputer to replace missing data with the median:
          imputer = SimpleImputer(strategy="median")
          # Indicate which variables to impute:
          ct = ColumnTransformer([("imputer", imputer, numeric_vars)], remainder="passthroug|
          # Find the median value per variable:
          ct.fit(X_train)
                  ColumnTransformer
Out[18]:
                imputer
                             remainder
           ▶ SimpleImputer
                             ▶ passthrough
In [19]:
          # Check the median that will be used in the imputation:
          ct.named_transformers_.imputer.statistics_
         array([ 28.835, 2.75 , 1. , 0. , 160. ,
                                                                      1)
Out[19]:
In [20]: # Replace missing data:
          X train = ct.transform(X train)
          X test = ct.transform(X test)
          X_train
Out[20]: array([[46.08, 3.0, 2.375, ..., 't', 't', 'g'],
                 [15.92, 2.875, 0.085, ..., 'f', 'f', 'g'], [36.33, 2.125, 0.085, ..., 't', 'f', 'g'],
                 [19.58, 0.665, 1.665, ..., 'f', 'f', 'g'],
                 [22.83, 2.29, 2.29, ..., 't', 't', 'g'],
                 [40.58, 3.29, 3.5, ..., 'f', 't', 's']], dtype=object)
```

```
# Convert returned array to a pandas dataframe:
In [21]:
          X_train = pd.DataFrame(X_train, columns=numeric_vars + remaining_vars,)
          X_train.head()
               A2
                                            A15 A1 A4 A5 A6 A7 A9 A10 A12 A13
Out[21]:
                     A3
                           A8 A11
                                     A14
          0 46.08
                     3.0 2.375
                                8.0
                                    396.0 4159.0
                                                       u
                                                           g
                                                                C
                                                                                       g
          1 15.92 2.875 0.085
                                0.0
                                   120.0
                                              0.0
                                                           g
                                                               q
                                                                                       g
          2 36.33 2.125 0.085
                                1.0
                                     50.0 1187.0
                                                                                  f
                                                   b
                                                                        t
                                                               W
                                                                    ٧
                                                       У
                                                           р
                                                                                       g
          3 22.17 0.585
                           0.0
                                0.0
                                    100.0
                                              0.0
                                                           р
                                                               ff
                                                                                       g
          4 57.83
                    7.04
                          14.0
                                6.0 360.0 1332.0
                                                   b
                                                       u
                                                               m
                                                                                       g
                                                           g
In [27]: # Corroboratw absence of missing values:
          X_train[numeric_vars].isnull().sum()
                 0
Out[27]:
          Α3
                 0
          Α8
                 0
          A11
                 0
          A14
                 0
          A15
                 0
          dtype: int64
In [30]: # Convert returned array to a pandas dataframe:
          X_test = pd.DataFrame(X_test, columns = numeric_vars +remaining_vars)
          X_test.head()
Out[30]:
               A2
                    A3
                          A8 A11
                                    A14
                                            A15 A1
                                                            A6 A7
                                                                     A9 A10 A12 A13
                                                     Α4
                                                         A5
          0 45.83
                  10.5
                          5.0
                               7.0
                                     0.0
                                            0.0
                                                      u
                                                              q
                                                                                 t
                                                                                      g
                                                          g
          1 64.08 20.0
                         17.5
                               9.0
                                     0.0 1000.0
                                                  b
                                                                   h
                                                              Χ
                                                                                 t
                                                                                      g
          2 31.25 3.75 0.625
                               9.0 181.0
                                            0.0
                                                  а
                                                          g
                                                              CC
                                                                  h
                                                                                 t
                                                                                      g
          3 39.25
                    9.5
                          6.5
                              14.0 240.0 4607.0
                                                                                 f
                                                  b
                                                                                      g
                                                              m
                    2.0
          4 26.17
                          0.0
                               0.0 276.0
                                             1.0
                                                               j
                                                                   j
                                                                                 t
                                                          g
                                                                                      g
In [31]: # Corroborate absence of missing values:
          X test[numeric vars].isnull().sum()
          Α2
                 0
Out[31]:
          Α3
                 0
          8A
                 0
          A11
          A14
                 0
          A15
          dtype: int64
```

# Mean / Median imputation with Featureengine

MeanMedianImputer https://feature-engine.readthedocs.io/en/latest/api\_doc/imputation/MeanMedianImputer.html

```
In [32]: # Split data into train and test set:
                             X_train, X_test, y_train, y_test = train_test_split(data.drop("target", axis=1), delicated train_test_split(data.drop("target"), delicated train_test_split(data.drop("ta
In [33]: # Set up the imputer to replace missing data with the median:
                             imputer = MeanMedianImputer(imputation_method="median", variables=numeric_vars)
                             # Find the median values:
                             imputer.fit(X_train)
Out[33]:
                                                                                                                 MeanMedianImputer
                            MeanMedianImputer(variables=['A2', 'A3', 'A8', 'A11', 'A14', 'A15'])
In [34]: # The median values per variable:
                             imputer.imputer_dict_
Out[34]: {'A2': 28.835, 'A3': 2.75, 'A8': 1.0, 'A11': 0.0, 'A14': 160.0, 'A15': 6.0}
In [36]: # Replace missing data with the median:
                             X train = imputer.transform(X train)
                             X_test = imputer.transform(X_test)
In [37]: # Corroborate absence of missing values:
                             X train[numeric vars].isnull().sum()
Out[37]: A2
                                                 0
                            Α3
                                                 0
                            Α8
                                                 0
                            A11
                                                 0
                            A14
                                                 0
                            A15
                             dtype: int64
In [38]: # Corroborate absence of missing values:
                             X test[numeric vars].isnull().sum()
                            Α2
                                                 0
Out[38]:
                             Α3
                                                 0
                            Α8
                                                 0
                            A11
                                                 0
                             A14
                                                  0
                             A15
                             dtype: int64
```