

```
# Handling numerical Data
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler, StandardScaler, Normalizer
```

```
df = pd.read_csv("housing.csv")
df.head(5)
```

| | RM | LSTAT | PTRATIO | MEDV |
|---|-------|-------|---------|----------|
| 0 | 6.575 | 4.98 | 15.3 | 504000.0 |
| 1 | NaN | 9.14 | 17.8 | 453600.0 |
| 2 | 7.185 | 4.00 | 17.8 | 728700.0 |
| 3 | 6.998 | 2.94 | 18.7 | 701400.0 |
| 4 | 7.147 | 5.33 | 18.7 | 760200.0 |

Next steps:

[Generate code with df](#)[New interactive sheet](#)

```
df.describe()
```

| | RM | LSTAT | PTRATIO | MEDV |
|-------|------------|------------|------------|--------------|
| count | 486.000000 | 485.000000 | 486.000000 | 4.880000e+02 |
| mean | 6.220737 | 12.933340 | 18.525720 | 4.543961e+05 |
| std | 0.733110 | 7.094904 | 2.101756 | 1.655058e+05 |
| min | 0.727000 | 1.980000 | 12.600000 | 1.050000e+05 |
| 25% | 5.881000 | 7.370000 | 17.400000 | 3.501750e+05 |
| 50% | 6.185000 | 11.660000 | 19.100000 | 4.389000e+05 |
| 75% | 6.578000 | 17.120000 | 20.200000 | 5.187000e+05 |
| max | 8.398000 | 37.970000 | 22.000000 | 1.024800e+06 |

```
df.duplicated().sum()
```

```
np.int64(0)
```

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

| | 0 |
|---------|---|
| RM | 3 |
| LSTAT | 4 |
| PTRATIO | 3 |
| MEDV | 1 |

```
dtype: int64
```

```
df["RM"] = df["RM"].fillna(df["RM"].mode())
df["LSTAT"] = df["LSTAT"].fillna(df["LSTAT"].mean())
df["PTRATIO"] = df["PTRATIO"].fillna(df["PTRATIO"].median())
```

```
df["MEDV"] = df["MEDV"].fillna(df["MEDV"].median())
```

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

```

      0
RM      0
LSTAT   0
PTRATIO  0
MEDV     0

```

```
dtype: int64
```

```

# min max scaling
scaler = MinMaxScaler()

df_minmax = df.copy()
df_minmax[["RM", "LSTAT", "PTRATIO", "MEDV"]] = scaler.fit_transform(
    df_minmax[["RM", "LSTAT", "PTRATIO", "MEDV"]]
)

print("After MinMax Scaling:")
display(df_minmax.head())

```

After MinMax Scaling:

| | RM | LSTAT | PTRATIO | MEDV |
|---|----------|----------|----------|----------|
| 0 | 0.762352 | 0.083356 | 0.287234 | 0.433790 |
| 1 | 0.716170 | 0.198944 | 0.553191 | 0.378995 |
| 2 | 0.841872 | 0.056127 | 0.553191 | 0.678082 |
| 3 | 0.817494 | 0.026674 | 0.648936 | 0.648402 |
| 4 | 0.836918 | 0.093081 | 0.648936 | 0.712329 |

```
# Standardization or Z score formula.. (x - mean) / std
```

```

std = StandardScaler()

df_std = df.copy()
df_std[["RM", "LSTAT", "PTRATIO", "MEDV"]] = std.fit_transform(
    df_std[["RM", "LSTAT", "PTRATIO", "MEDV"]]
)

print("After Standardization:")
display(df_std.head())

```

After Standardization:

| | RM | LSTAT | PTRATIO | MEDV |
|---|----------|-----------|-----------|-----------|
| 0 | 0.485222 | -1.126769 | -1.542419 | 0.300515 |
| 1 | 0.000000 | -0.537412 | -0.348317 | -0.004628 |
| 2 | 1.320718 | -1.265608 | -0.348317 | 1.660944 |
| 3 | 1.064590 | -1.415780 | 0.081560 | 1.495659 |
| 4 | 1.268670 | -1.077183 | 0.081560 | 1.851659 |

```

# Normalization or L2
norm = Normalizer()

```

```

df_norm = df.copy()
# Impute missing values before normalization
df_norm["RM"] = df_norm["RM"].fillna(df_norm["RM"].mean())
df_norm["LSTAT"] = df_norm["LSTAT"].fillna(df_norm["LSTAT"].mean())
df_norm["PTRATIO"] = df_norm["PTRATIO"].fillna(df_norm["PTRATIO"].mean())
df_norm["MEDV"] = df_norm["MEDV"].fillna(df_norm["MEDV"].mean())

df_norm[["RM", "LSTAT", "PTRATIO", "MEDV"]] = norm.fit_transform(
    df_norm[["RM", "LSTAT", "PTRATIO", "MEDV"]]
)

print("After Normalization:")
display(df_norm.head())

```

After Normalization:

| | RM | LSTAT | PTRATIO | MEDV | |
|---|----------|----------|----------|------|--|
| 0 | 0.000013 | 0.000010 | 0.000030 | 1.0 | |
| 1 | 0.000014 | 0.000020 | 0.000039 | 1.0 | |
| 2 | 0.000010 | 0.000005 | 0.000024 | 1.0 | |
| 3 | 0.000010 | 0.000004 | 0.000027 | 1.0 | |
| 4 | 0.000009 | 0.000007 | 0.000025 | 1.0 | |

df.describe()

| | RM | LSTAT | PTRATIO | MEDV | |
|-------|------------|------------|------------|--------------|--|
| count | 489.000000 | 489.000000 | 489.000000 | 4.890000e+02 | |
| mean | 6.220737 | 12.933340 | 18.529243 | 4.543644e+05 | |
| std | 0.730853 | 7.065766 | 2.095767 | 1.653376e+05 | |
| min | 0.727000 | 1.980000 | 12.600000 | 1.050000e+05 | |
| 25% | 5.884000 | 7.390000 | 17.400000 | 3.507000e+05 | |
| 50% | 6.193000 | 11.720000 | 19.100000 | 4.389000e+05 | |
| 75% | 6.575000 | 17.110000 | 20.200000 | 5.187000e+05 | |
| max | 8.398000 | 37.970000 | 22.000000 | 1.024800e+06 | |