

✓ Boston Housing Dataset Cleaning Assignment

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```
1 from google.colab import files
2 uploaded = files.upload()
3 import pandas as pd
4 import numpy as np
5 #load Data set
6 df = pd.read_csv(list(uploaded.keys())[0])
7 # Show first rows
8 df.head() # showing first five rows known as headers.
9
```

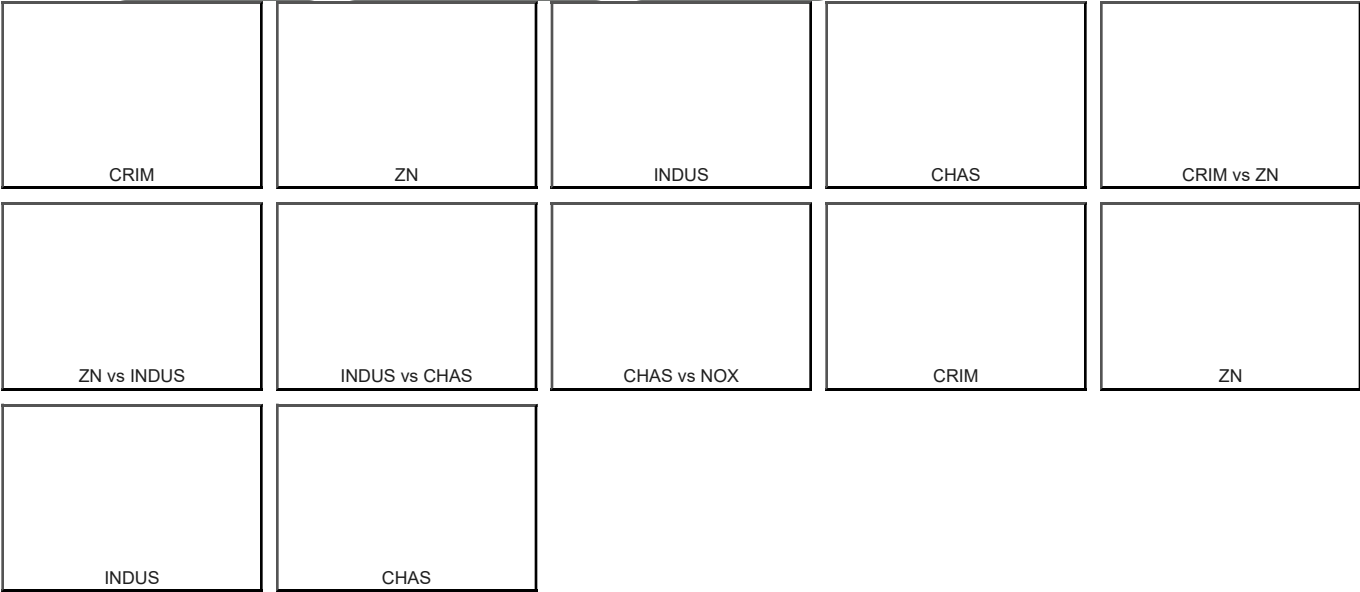
Choose files boston_housing.csv

- **boston_housing.csv**(text/csv) - 35200 bytes, last modified: 30/08/2025 - 100% done

Saving boston_housing.csv to boston_housing.csv

| | CRIM | ZN | INDUS | CHAS | NOX | RM | AGE | DIS | RAD | TAX | PTRATIO | B | LSTAT | MEDV |
|---|---------|------|-------|------|-------|-------|------|--------|-----|-----|---------|--------|-------|------|
| 0 | 0.00632 | 18.0 | 2.31 | 0 | 0.538 | 6.575 | 65.2 | 4.0900 | 1 | 296 | 15.3 | 396.90 | 4.98 | 24.0 |
| 1 | 0.02731 | 0.0 | 7.07 | 0 | 0.469 | 6.421 | 78.9 | 4.9671 | 2 | 242 | 17.8 | 396.90 | 9.14 | 21.6 |
| 2 | 0.02729 | 0.0 | 7.07 | 0 | 0.469 | 7.185 | 61.1 | 4.9671 | 2 | 242 | 17.8 | 392.83 | 4.03 | 34.7 |
| 3 | 0.03237 | 0.0 | 2.18 | 0 | 0.458 | 6.998 | 45.8 | 6.0622 | 3 | 222 | 18.7 | 394.63 | 2.94 | 33.4 |
| 4 | 0.06905 | 0.0 | 2.18 | 0 | 0.458 | 7.147 | 54.2 | 6.0622 | 3 | 222 | 18.7 | 396.90 | 5.33 | 36.2 |

Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)



```
1 # Step 2: Inspect data
2 print("Data Types:\n", df.dtypes)#Analysing DataTypes
3 print("\nMissing Values:\n", df.isnull().sum())#Checking for Missing Values
4 print("\nSummary Statistics:\n", df.describe())#Printing Statics of the Data set.
5
```

NOX float64
RM float64
AGE float64
DIS float64
RAD int64
TAX int64
PTRATIO float64
B float64

```

ZN      0
INDUS   0
CHAS    0
NOX     0
RM      0
AGE     0
DIS     0
RAD     0
TAX     0
PTRATIO 0
B       0
LSTAT   0
MEDV    0
dtype: int64

```

Summary Statistics:

| | CRIM | ZN | INDUS | CHAS | NOX | RM |
|-------|------------|------------|------------|------------|------------|------------|
| count | 506.000000 | 506.000000 | 506.000000 | 506.000000 | 506.000000 | 506.000000 |
| mean | 3.613524 | 11.363636 | 11.136779 | 0.069170 | 0.554695 | 6.284634 |
| std | 8.601545 | 23.322453 | 6.860353 | 0.253994 | 0.115878 | 0.702617 |
| min | 0.006320 | 0.000000 | 0.460000 | 0.000000 | 0.385000 | 3.561000 |
| 25% | 0.082045 | 0.000000 | 5.190000 | 0.000000 | 0.449000 | 5.885500 |
| 50% | 0.256510 | 0.000000 | 9.690000 | 0.000000 | 0.538000 | 6.208500 |
| 75% | 3.677083 | 12.500000 | 18.100000 | 0.000000 | 0.624000 | 6.623500 |
| max | 88.976200 | 100.000000 | 27.740000 | 1.000000 | 0.871000 | 8.780000 |

| | AGE | DIS | RAD | TAX | PTRATIO | B |
|-------|------------|------------|------------|------------|------------|------------|
| count | 506.000000 | 506.000000 | 506.000000 | 506.000000 | 506.000000 | 506.000000 |
| mean | 68.574901 | 3.795043 | 9.549407 | 408.237154 | 18.455534 | 356.674032 |
| std | 28.148861 | 2.105710 | 8.707259 | 168.537116 | 2.164946 | 91.294864 |
| min | 2.900000 | 1.129600 | 1.000000 | 187.000000 | 12.600000 | 0.320000 |
| 25% | 45.025000 | 2.100175 | 4.000000 | 279.000000 | 17.400000 | 375.377500 |
| 50% | 77.500000 | 3.207450 | 5.000000 | 330.000000 | 19.050000 | 391.440000 |
| 75% | 94.075000 | 5.188425 | 24.000000 | 666.000000 | 20.200000 | 396.225000 |
| max | 100.000000 | 12.126500 | 24.000000 | 711.000000 | 22.000000 | 396.900000 |

| | LSTAT | MEDV |
|-------|------------|------------|
| count | 506.000000 | 506.000000 |
| mean | 12.653063 | 22.532806 |
| std | 7.141062 | 9.197104 |
| min | 1.730000 | 5.000000 |
| 25% | 6.950000 | 17.025000 |
| 50% | 11.360000 | 21.200000 |
| 75% | 16.955000 | 25.000000 |
| max | 37.970000 | 50.000000 |

```

1 # Step 3: Handle missing values
2 df = df.fillna(df.mean(numeric_only=True)) # numeric → mean
3 for col in df.select_dtypes(include='object').columns: # categorical → mode
4     df[col] = df[col].fillna(df[col].mode()[0])
5

```

```

1 # Step 4: Detect outliers before handling
2 for col in df.select_dtypes(include=np.number).columns:
3     Q1 = df[col].quantile(0.25)
4     Q3 = df[col].quantile(0.75)
5     IQR = Q3 - Q1
6     lower = Q1 - 1.5 * IQR
7     upper = Q3 + 1.5 * IQR
8     outliers = df[(df[col] < lower) | (df[col] > upper)]
9     print(f"{col}: {len(outliers)} outliers detected")
10

```

```

11 # Step 4b: Handle outliers (capping)
12 for col in df.select_dtypes(include=np.number).columns:
13     Q1 = df[col].quantile(0.25)
14     Q3 = df[col].quantile(0.75)
15     IQR = Q3 - Q1
16     lower = Q1 - 1.5 * IQR
17     upper = Q3 + 1.5 * IQR
18     df[col] = np.where(df[col] < lower, lower,
19                        np.where(df[col] > upper, upper, df[col]))
20

```

```

🔍 CRIM: 0 outliers detected
   ZN: 0 outliers detected
   INDUS: 0 outliers detected
   CHAS: 0 outliers detected
   NOX: 0 outliers detected
   RM: 0 outliers detected
   AGE: 0 outliers detected
   DIS: 0 outliers detected

```

```
RAD: 0 outliers detected  
TAX: 0 outliers detected  
PTRATIO: 0 outliers detected  
B: 0 outliers detected  
LSTAT: 0 outliers detected  
MEDV: 0 outliers detected
```

```
1 # Step 5: Save cleaned dataset  
2 df.to_csv("boston_cleaned.csv", index=False)  
3  
4 files.download("boston_cleaned.csv")  
5
```

