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

df = pd.read_csv('boston.csv')
df.head()
```

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296.0	15.3
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242.0	17.8
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242.0	17.8
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222.0	18.7
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222.0	18.7

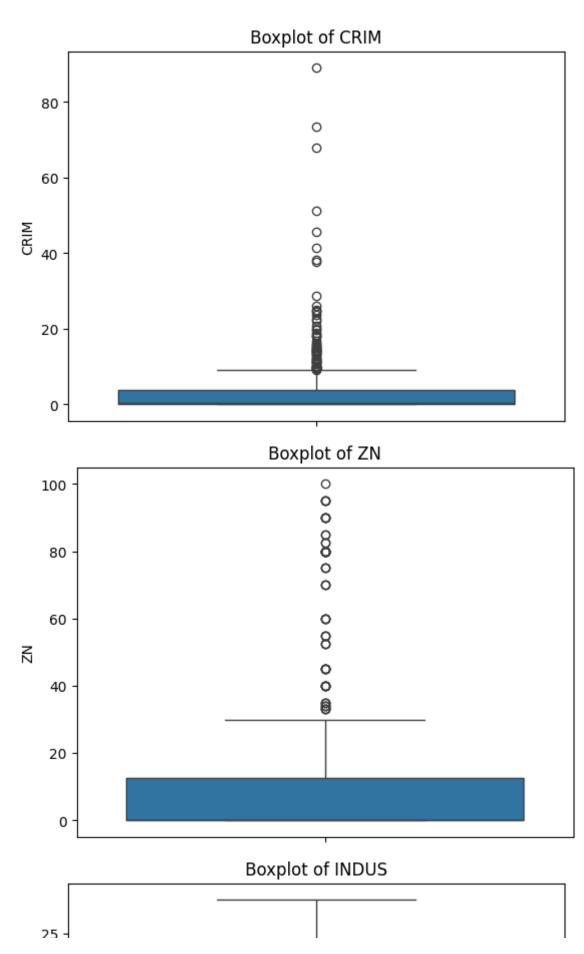
```
df.isnull().sum()
     CRIM
     ZN
                 0
     INDUS
                 0
     CHAS
     NOX
     RM
                 0
     AGE
                 0
     DIS
                 0
     RAD
                 0
     TAX
     PTRATIO
     В
     LSTAT
                 0
    MEDV
     dtype: int64
numerical_features = []
```

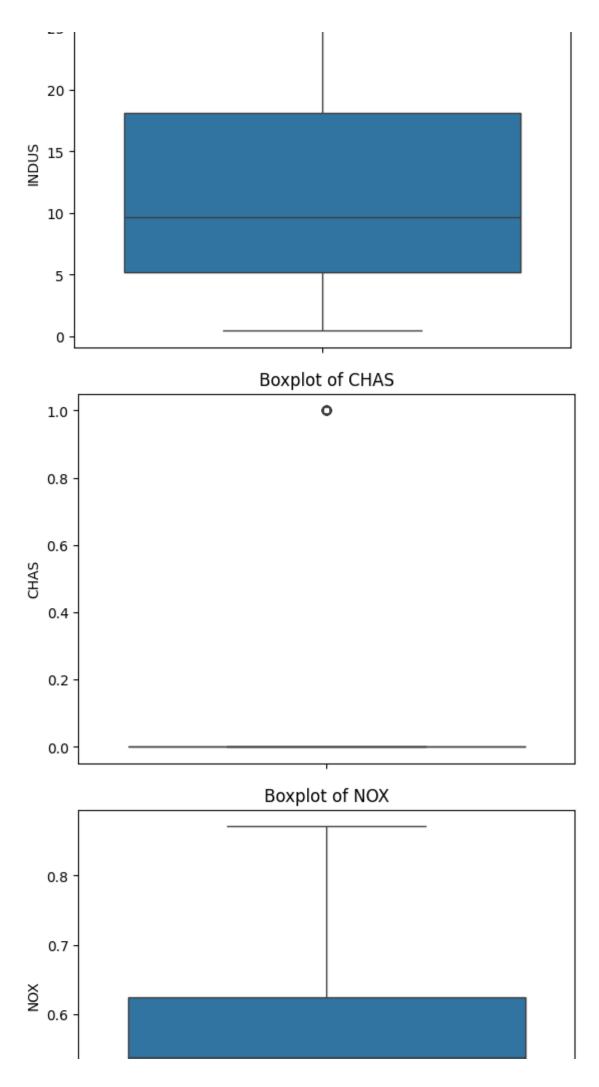
```
categorical_features = []

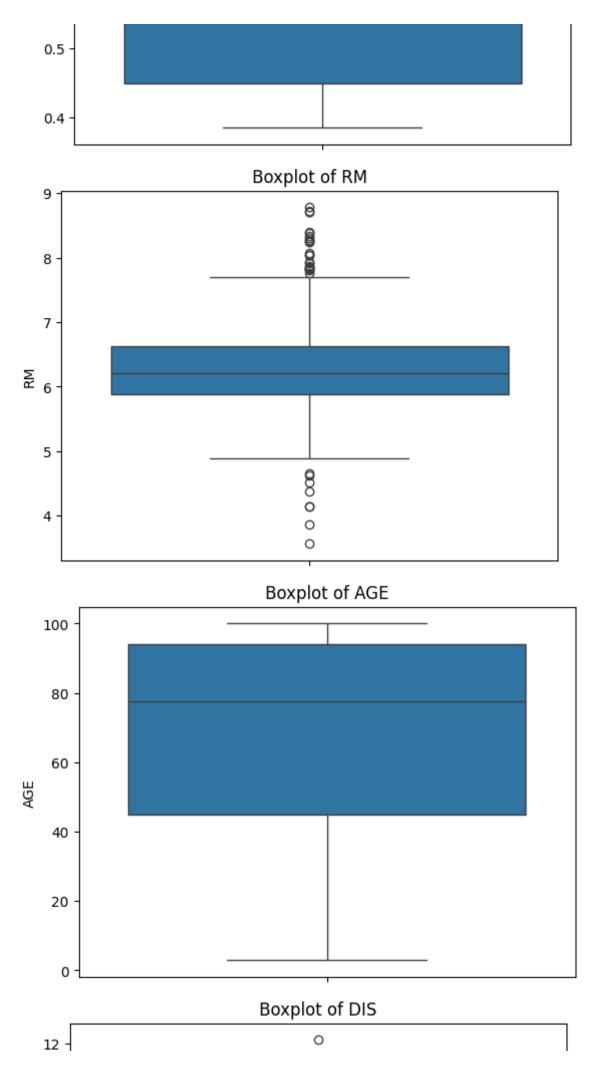
for f in df.columns:
    if df[f].dtype != '0':
        numerical_features.append(f)
    else:
        categorical_features.append(f)

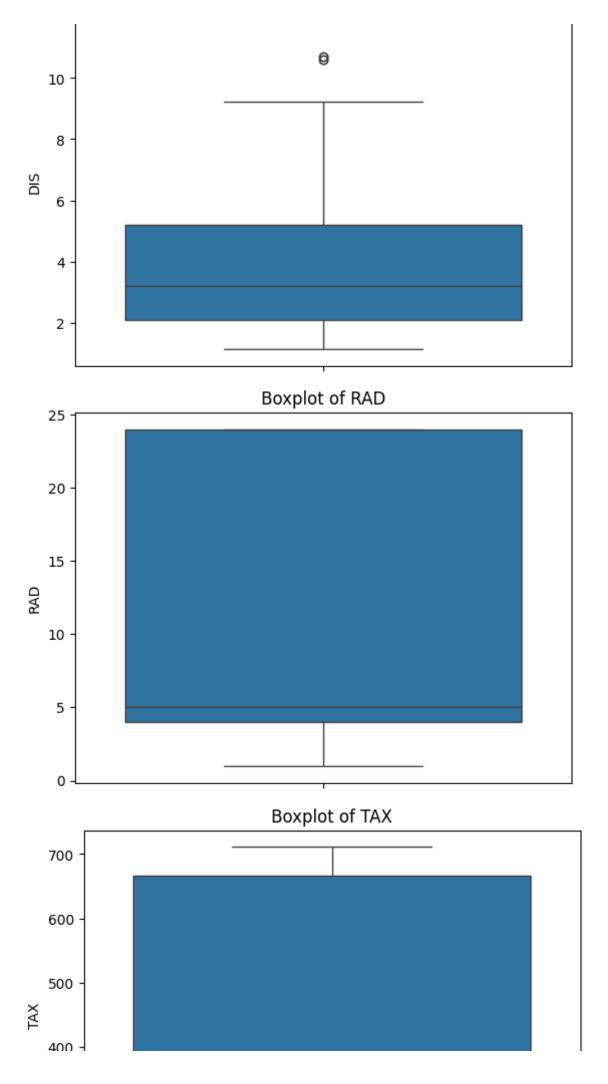
print(numerical_features)
print(categorical_features)
    ['CRIM', 'ZN', 'INDUS', 'CHAS', 'NOX', 'RM', 'AGE', 'DIS', 'RAD', 'TAX', 'I
    []
```

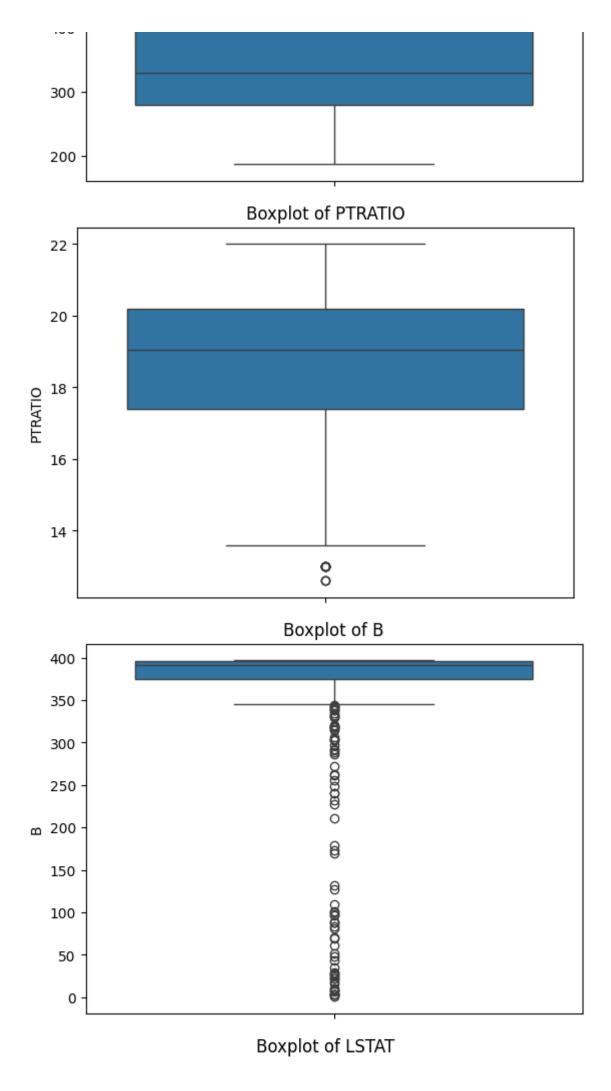
```
for f in df.columns:
    sns.boxplot(df[f])
    plt.title(f'Boxplot of {f}')
    plt.show()
```

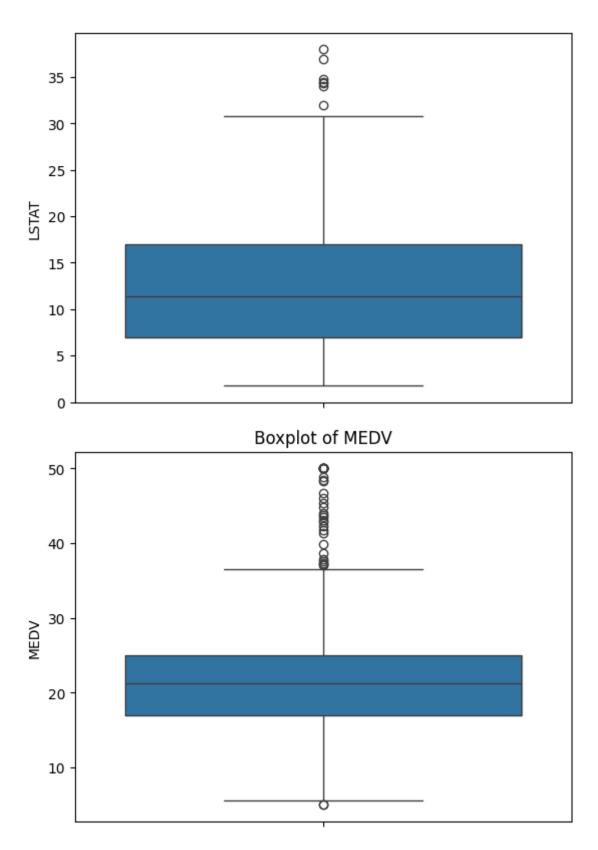






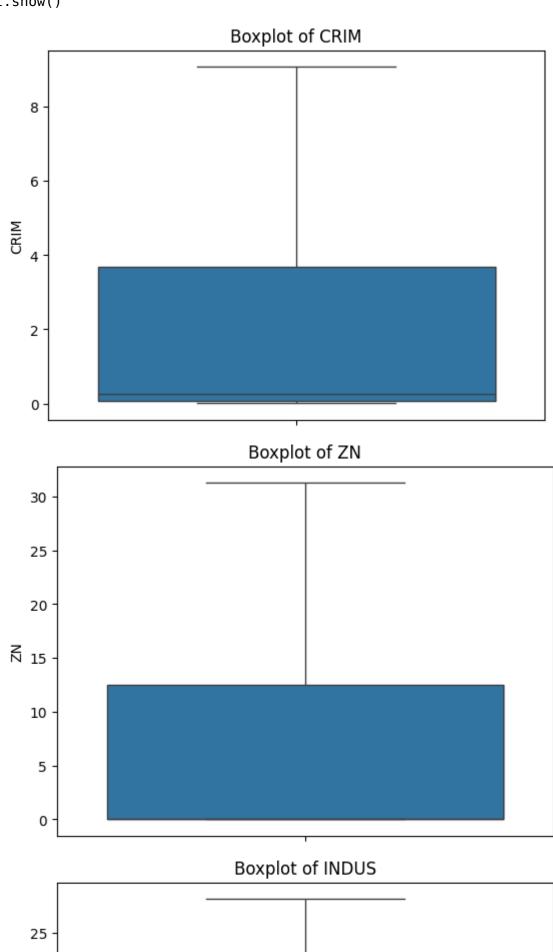


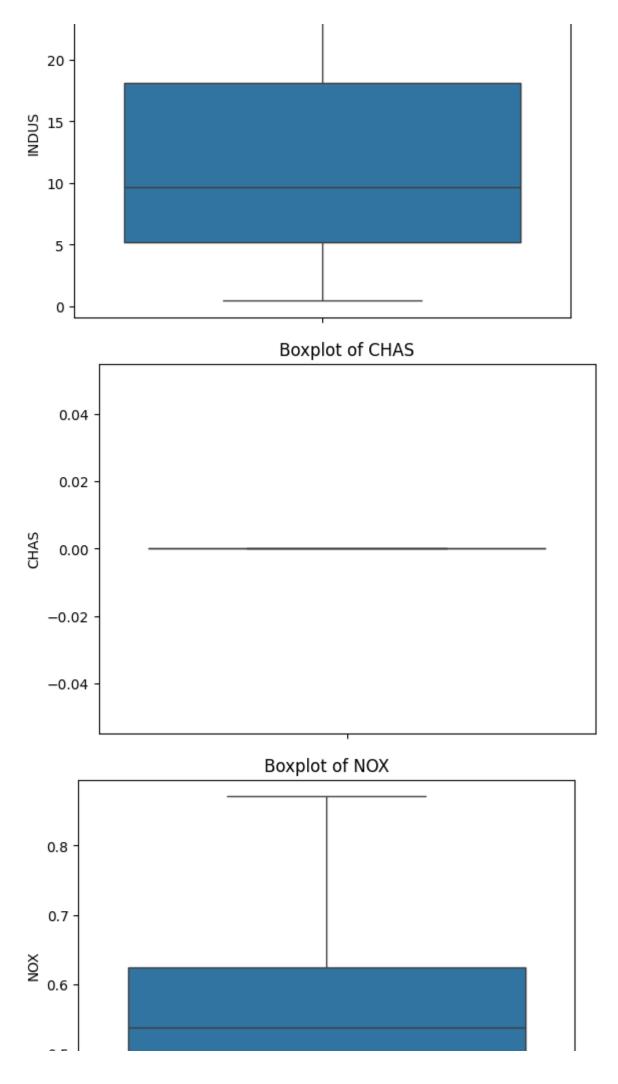


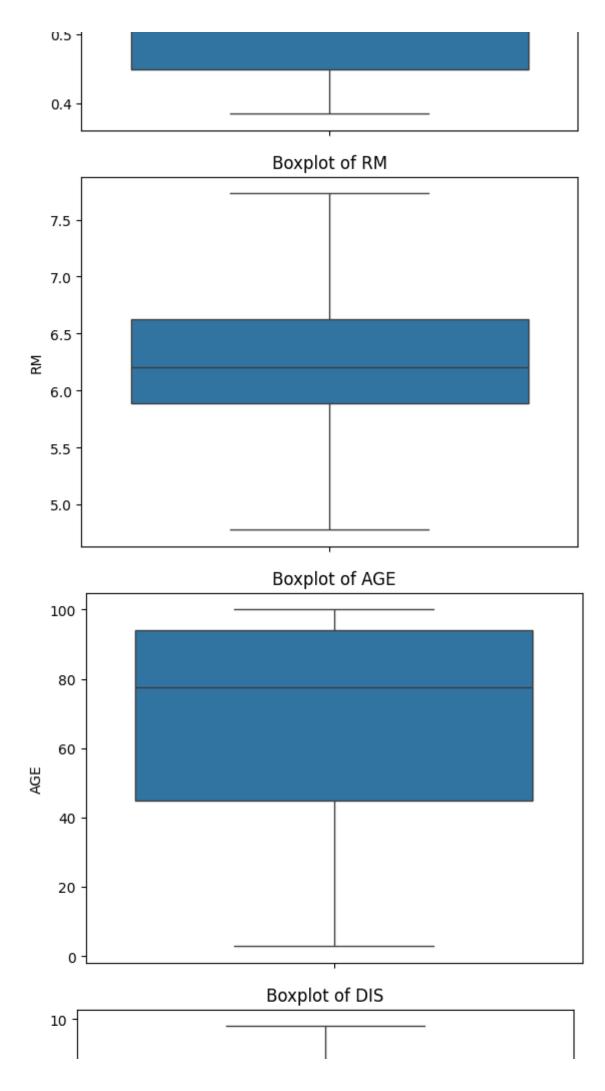


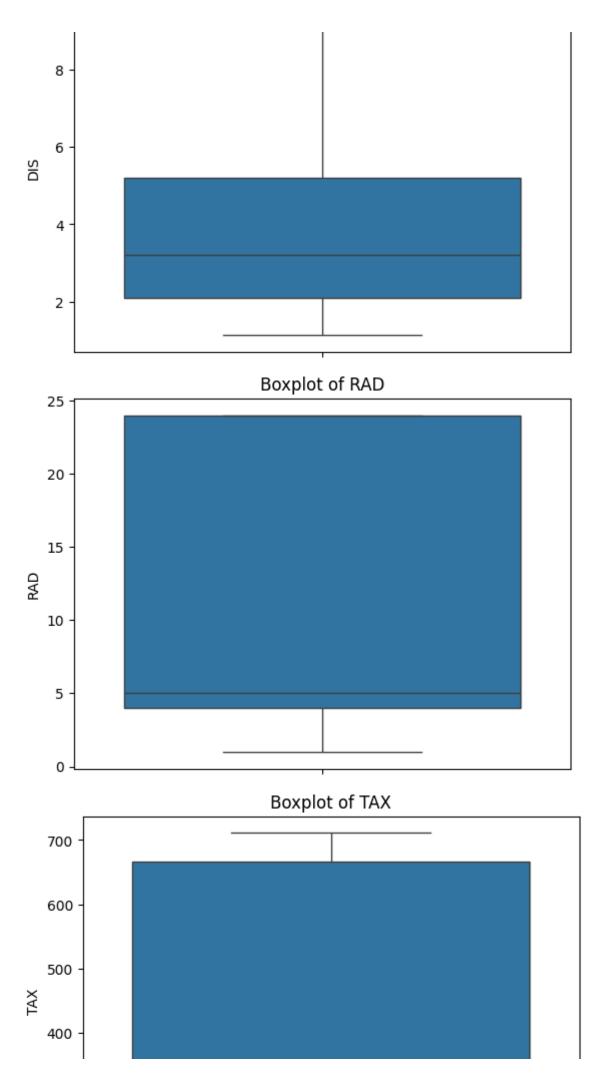
```
numericalCols = df.select_dtypes(include='number').columns
for f in numericalCols:
    q3 = df[f].quantile(0.75)
    q1 = df[f].quantile(0.25)
    iqr = q3 - q1
    upper_bound = q3 + (1.5 * iqr)
    lower_bound = q1 - (1.5 * iqr)
    df.loc[df[f] >= upper_bound, f] = upper_bound
    df.loc[df[f] <= lower_bound, f] = lower_bound</pre>
```

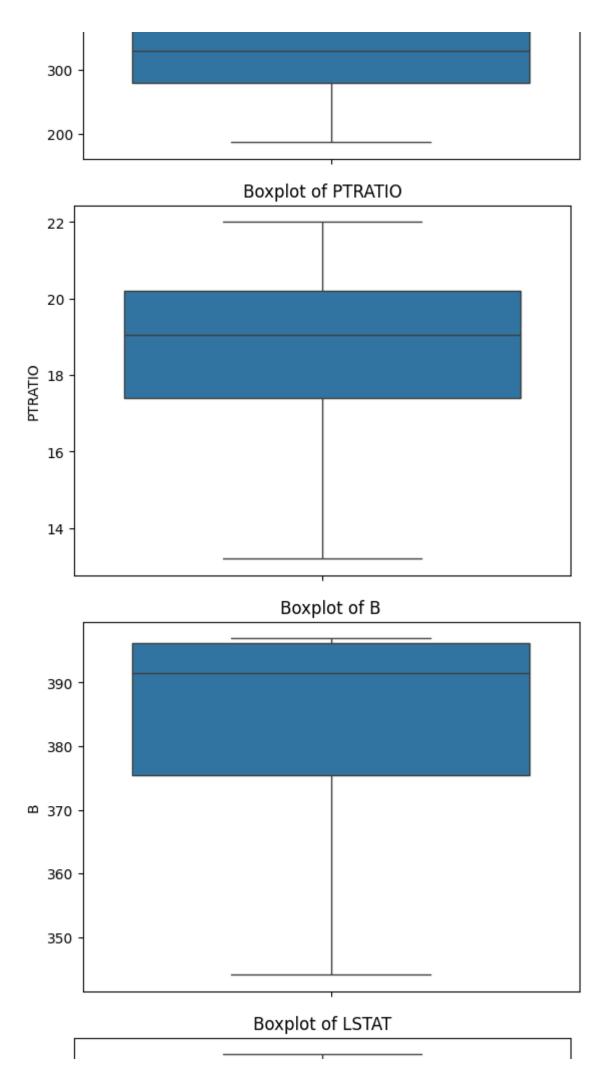
```
for f in df.columns:
   sns.boxplot(df[f])
  plt.title(f'Boxplot of {f}')
  plt.show()
```

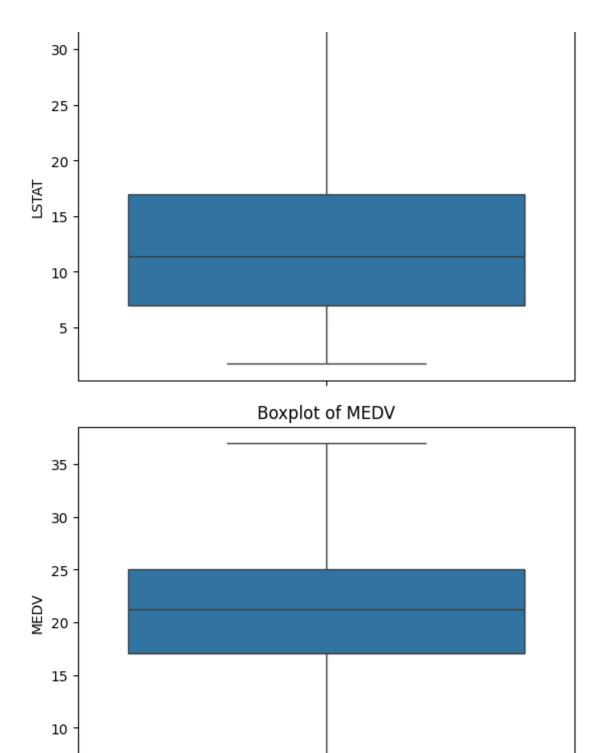






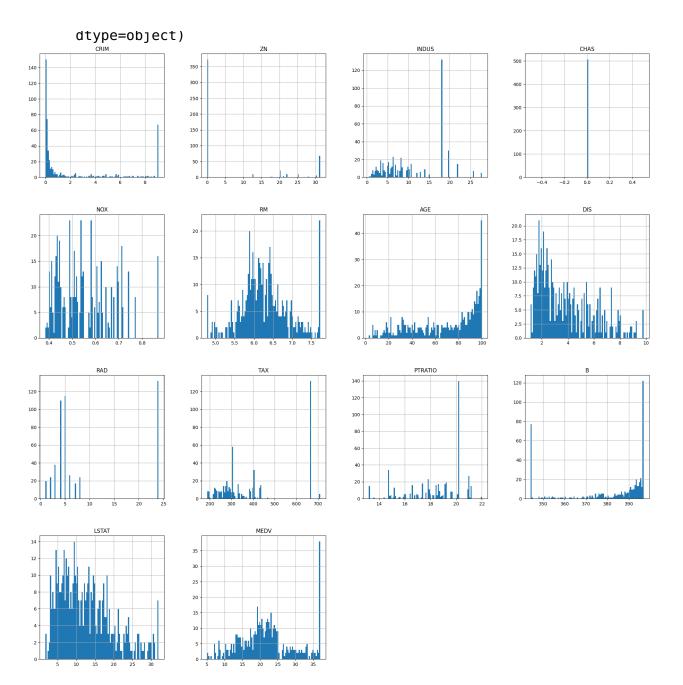






## df.hist(figsize=(24,24), bins=100)

5



```
df.skew()
    CRIM
                1.282313
    ZN
                1.261340
    INDUS
                0.295022
    CHAS
                0.000000
    NOX
                0.729308
    RM
                0.296640
    AGE
               -0.598963
    DIS
               0.908467
    RAD
               1.004815
    TAX
               0.669956
    PTRATIO
               -0.762495
               -1.164208
    LSTAT
                0.808671
    MEDV
                0.353614
    dtype: float64
# Model building
from sklearn.model selection import train test split
from sklearn.linear model import LinearRegression
from sklearn.model selection import GridSearchCV
from sklearn.metrics import mean squared error
X = df.drop('MEDV', axis=1)
y = df['MEDV']
X_train, X_test, y_train, y_test = train test split(X, y, test size=0.2, rando
regr = LinearRegression()
regr.fit(X train, y train)
     ▼ LinearRegression
     LinearRegression()
# predicting the model output
y pred = regr.predict(X test)
mse = mean squared error(y test, y pred)
print(f'Mean Squared Error: {mse}')
    Mean Squared Error: 13.264821903384027
print("Tuned Weights:")
for i in range(len(X.columns)):
    print(f"{X.columns[i]}: {regr.coef [i]}")
print(f"Intercept: {regr.intercept }")
    Tuned Weights:
    CRIM: -0.47702088562571165
    ZN: 0.04146279314123744
    INDUS: -0.021440789997329568
    CHAS: 8.340550472496489e-14
    NOX: -14.041056617657127
    RM: 2.7458448668745974
    ΔGF: - A AA272259A8A48AAA872
```

DIS: -1.0342233144791397 RAD: 0.28211799925832115 TAX: -0.008092199582586938 PTRATIO: -0.8728299586706918 B: 0.006769390573664431

B: 0.006769390573664431 LSTAT: -0.4916902018584183 Intercept: 37.97381451281785