```
In [9]:
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
In [10]:
data = pd.read csv("housing.csv", delimiter=r"\s+", header=None)
data.head()
Out[10]:
                                                                               13
         0
                   2 3
                                   5
                                                7 8
                                                             10
                                                                     11
                                                                          12
0 0.00632
           18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296.0 15.3 396.90
                                                                        4.98 24.0
1 0.02731
             0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242.0 17.8 396.90 9.14 21.6
2 0.02729
             0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242.0 17.8 392.83 4.03 34.7
3 0.03237
             0.0 \quad 2.18 \quad 0 \quad 0.458 \quad 6.998 \quad 45.8 \quad 6.0622 \quad 3 \quad 222.0 \quad 18.7 \quad 394.63
             0.0 2.18 0 0.458 7.147 54.2 6.0622 3 222.0 18.7 396.90 5.33 36.2
  0.06905
In [13]:
import seaborn as sns
corr_matrix = data.corr()
plt.figure(figsize=(10,5))
sns.heatmap(corr matrix, annot=True)
Out[13]:
<AxesSubplot:>
                                                                               - 1.0
 0 - 1
        -0.2 0.41 -0.056 0.42 -0.22 0.35 -0.38 0.63
                                                0.58 0.29 -0.39 0.46
                                                                   -0.39
             -0.53 <mark>-0.043 -0.52 | 0.31 | -0.57 | 0.66 | -0.31 | -0.31 | -0.39 | 0.18 | -0.41</mark>
                                                                               - 0.8
              1 0.063 0.76 -0.39 0.64 -0.71
                                                0.72 0.38 -0.36 0.6
                                                                              - 0.6
    -0.056-0.043 0.063 1 0.091 0.091 0.087 -0.0990.00740.036 -0.12 0.049 -0.054 0.18
    -0.43
                                                                              - 0.4
             -0.39 0.091 -0.3
                             1
                                 -0.24 0.21 -0.21 -0.29 -0.36 0.13 -0.61
    -0.22 0.31
                                                                    0.7
                                                                               - 0.2
        -0.57 0.64 0.087 0.73 -0.24 1
                                      -0.75
             -0.71 -0.099 -0.77
                                 -0.75
                                           -0.49 -0.53 -0.23
    -0.38
         0.66
                                       1
                                                               -0.5
                                                                              - 0.0
              0.6 -0.0074 0.61 -0.21 0.46
                                                0.91
                                                                    -0.38
 \infty
         -0.31
                                      -0.49
                                            1
                                                          -0.44
         -0.31 0.72 -0.036 0.67 -0.29
                                      -0.53
                                           0.91
                                                 1
                                                          -0.44
                                                                   -0.47
 S)
                                                                               -0.2
              0.38 -0.12 0.19 -0.36 0.26
    0.29 -0.39
                                      -0.23
                                                      1
                                                          -0.18
                                                                   -0.51
 2
                                                                               - -0.4
    1
 П
         -0.41 0.6 -0.054 0.59 -0.61 0.6
                                                         -0.37
                                                                    -0.74
                                                                                -0.6
 12
    -0.39
                                 -0.38 0.25 -0.38 -0.47 -0.51
             -0.48 0.18 -0.43
                             0.7
                                                               -0.74
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                                                                     13
```

i am plotting corelation matrix of data and i found that the 5th column i.e, rm average number of rooms per dwelling has a strong corelation of 0.7 with the 13th column that is medv median value of owner-occupied homes in \$1000s. So i am picking 5th column to predict 13th column

```
In [38]:

X = data.iloc[:,5]
y = data.iloc[:,13]
```

```
In [18]:
```

```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=1)
```

In [21]:

```
reg = LinearRegression()
reg.fit(x_train.values.reshape(-1,1),y_train)
```

Out[21]:

LinearRegression()

In [26]:

```
from sklearn.metrics import mean_squared_error, r2_score

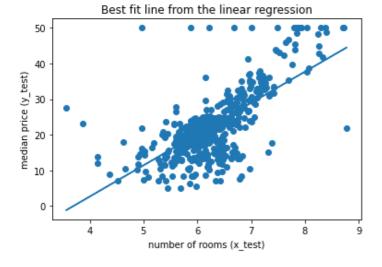
y_pred = reg.predict(x_test.values.reshape(-1,1))
rmse_Score = mean_squared_error(y_test,y_pred, squared=False)
r2_Score = r2_score(y_test,y_pred)

print("rmse_score of the data: ", rmse_Score)
print("r2 score of the data: ", r2_Score)
```

rmse score of the data: 6.383135112649168 r2 score of the data: 0.5877214395051774

In [52]:

```
y_predicted = reg.predict(X.values.reshape(-1,1))
plt.scatter(X,y)
plt.plot(X,y_predicted)
plt.title("Best fit line from the linear regression")
plt.xlabel("number of rooms (x_test)")
plt.ylabel("median price (y_test)")
plt.show()
```



In [122]:

```
from sklearn.preprocessing import PolynomialFeatures

poly = PolynomialFeatures(degree=2)
X_poly = poly.fit_transform(X.values.reshape(-1,1))
```

In [123]:

```
X_poly.shape
```

Out[123]:

(506, 3)

In [124]:

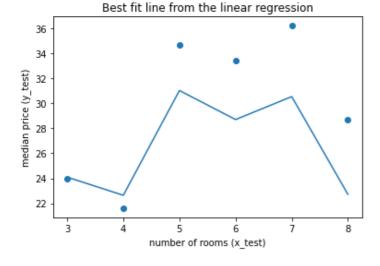
```
x_train, x_test, y_train, y_test = train_test_split(X_poly, y, test_size=0.2, random_sta
te=1)
```

In [125]: reg poly = LinearRegression() reg poly.fit(x train, y train) Out[125]: LinearRegression() In [126]: y_pred = reg_poly.predict(x_test) rmse Score poly = mean squared error(y test, y pred, squared=False) r2_Score_poly = r2_score(y_test,y_pred) print("rmse score of the data: ", rmse Score poly) print("r2 score of the data: ", r2 Score poly) rmse score of the data: 5.799858200692183

r2 score of the data: 0.6596251774934456

In [127]:

```
y predicted = reg poly.predict(X poly)
x \text{ space} = np.arange(3, 9, 1)
plt.scatter(x space, y[0:6])
plt.plot(x_space,y_predicted[0:6])
plt.title("Best fit line from the linear regression")
plt.xlabel("number of rooms (x test)")
plt.ylabel("median price (y_test)")
plt.show()
```



In [128]:

```
poly 20 = PolynomialFeatures(degree=20)
X \text{ poly } 20 = \text{poly.fit transform}(X.\text{values.reshape}(-1,1))
```

In [129]:

```
x train, x test, y train, y test = train test split(X poly 20, y, test size=0.2, random
state=1)
```

In [130]:

```
reg poly 20 = LinearRegression()
reg poly 20.fit(x train, y train)
```

Out[130]:

LinearRegression()

In [131]:

```
y pred = reg poly 20.predict(x test)
```

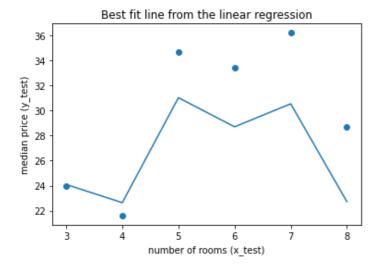
```
rmse_Score_poly_20 = mean_squared_error(y_test,y_pred, squared=False)
r2_Score_poly_20 = r2_score(y_test,y_pred)

print("rmse score of the data: ", rmse_Score_poly_20)
print("r2 score of the data: ", r2_Score_poly_20)
```

rmse score of the data: 5.799858200692183 r2 score of the data: 0.6596251774934456

In [132]:

```
y_predicted = reg_poly_20.predict(X_poly_20)
x_space = np.arange(3, 9, 1)
plt.scatter(x_space,y[0:6])
plt.plot(x_space,y_predicted[0:6])
plt.title("Best fit line from the linear regression")
plt.xlabel("number of rooms (x_test)")
plt.ylabel("median price (y_test)")
plt.show()
```



In [112]:

```
x_multiple = data.iloc[:,[5,11,12]]
y = data.iloc[:,13]
```

In [114]:

```
x_train, x_test, y_train, y_test = train_test_split(x_multiple, y, test_size=0.2, random
_state=1)
reg_multiple = LinearRegression()
reg_multiple.fit(x_train,y_train)
```

Out[114]:

LinearRegression()

In [117]:

```
y_pred = reg_multiple.predict(x_test)
rmse_Score_multiple = mean_squared_error(y_test,y_pred, squared=False)
r2_Score_multiple = r2_score(y_test,y_pred)

print("rmse score of the data: ", rmse_Score_multiple)
print("r2 score of the data: ", r2_Score_multiple)
```

rmse score of the data: 5.474914616536909 r2 score of the data: 0.6966965309874333