```
In [57]:
           import pandas as pd
           import seaborn as sns
           import numpy as np
           import matplotlib.pyplot as plt
           from sklearn.preprocessing import MinMaxScaler
           from sklearn.model_selection import train_test_split
           from sklearn.metrics import mean_squared_error
           from sklearn.linear_model import LinearRegression
In [58]:
           df = pd.read_csv('../Data/BostonHousing.csv')
Out[58]:
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               0.00632
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                                                               4.9671
                                                                           242
                0.02729
                         0.0
                                7.07
                                           0.469
                                                  7.185
                                                         61.1
                                                               4.9671
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                                           0.573 6.030 80.8 2.5050
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         506 rows × 14 columns
In [59]:
           df.isnull().sum()
          crim
Out[59]:
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          lstat
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          medv
          dtype: int64
In [65]:
           scale = MinMaxScaler()
           Scaled_data = df.iloc[:,:-1] = scale.fit_transform(df.iloc[:,:-1])
           df
```

Out[65]:		crim	zn	indus	chas	nox	rm	age	dis
	0	0.000000	0.18	0.067815	0	0.314815	0.577505	0.641607	0.269203
	1	0.000236	0.00	0.242302	0	0.172840	0.547998	0.782698	0.348962
	2	0.000236	0.00	0.242302	0	0.172840	0.694386	0.599382	0.348962
	3	0.000293	0.00	0.063050	0	0.150206	0.658555	0.441813	0.448545
	4	0.000705	0.00	0.063050	0	0.150206	0.687105	0.528321	0.448545
	•••								
	501	0.000633	0.00	0.420455	0	0.386831	0.580954	0.681771	0.122671
	502	0.000438	0.00	0.420455	0	0.386831	0.490324	0.760041	0.105293
	503	0.000612	0.00	0.420455	0	0.386831	0.654340	0.907312	0.094381
	504	0.001161	0.00	0.420455	0	0.386831	0.619467	0.889804	0.114514
	505	0.000462	0.00	0.420455	0	0.386831	0.473079	0.802266	0.125072

506 rows × 14 columns

```
plt.figure(figsize=(10, 8))
sns.heatmap(df.corr(), annot=True)
plt.show()
```



```
In [62]: x = df.iloc[:,:-1]
y = df[["rm","b","zn"]]
```

```
X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.2
model=LinearRegression()
model.fit(X_train,y_train)
y_test_pred = model.predict(X_test)
y_train_pred = model.predict(X_train)
comparison=pd.DataFrame({"Actual":y_test.values.flatten(),"Predicted":y
comparison
```

```
Out[62]:
                           Predicted
                 Actual
            0 0.547040 5.470397e-01
            1 0.996470 9.964698e-01
            2 0.000000 4.926481e-16
              0.612569 6.125695e-01
             1.000000 1.000000e+00
          301 0.997705 9.977054e-01
         302 0.000000 8.702437e-16
         303 0.522131 5.221307e-01
         304 0.965530 9.655303e-01
         305 0.000000 3.439613e-16
         306 rows × 2 columns
In [63]:
          rmse_test = np.sqrt(mean_squared_error(y_test, y_test_pred))
          print(f"Test RMSE: {rmse test}")
          rmse_train = np.sqrt(mean_squared_error(y_train, y_train_pred))
          print(f"Train RMSE: {rmse train}")
        Test RMSE: 7.71351067881442e-16
        Train RMSE: 7.434339382369642e-16
In [64]:
          comparison = pd.DataFrame({
              "Actual": y_test.values.flatten(),
              "Predicted": y_test_pred.flatten()
          })
          columns = ["rm", "b", "zn"]
          train_colors = ['blue', 'green', 'orange']
          test_colors = ['cyan', 'lightgreen', 'gold']
          fig, axes = plt.subplots(1, 3, figsize=(18, 6))
          for i, col in enumerate(columns):
              axes[i].scatter(y_train[col], y_train_pred[:, i], color=train_color
              axes[i].scatter(y_test[col], y_test_pred[:, i], color=test_colors[:
              min_val = min(y[col].min(), y[col].min())
              \max_{val} = \max(y[col].max(), y[col].max())
              axes[i].plot([min_val, max_val], [min_val, max_val], color='black',
              axes[i].set_xlabel("Actual Values")
              axes[il.set vlabel("Predicted Values")
```

```
axes[i].set_title(f"Actual vs Predicted for {col}")
axes[i].legend()
axes[i].grid(True)

plt.tight_layout()
plt.show()
```

