

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
from sklearn.preprocessing import StandardScaler
```

```
url = "https://raw.githubusercontent.com/anbarasang2024aids-dev/foml/references/main/data.csv"
df = pd.read_csv(url)
print("✅ Dataset Loaded Successfully!")
print("Shape:", df.shape)
print(df.head())
```

✅ Dataset Loaded Successfully!

Shape: (511, 14)

	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

```
df = df.dropna()
print("\nAfter removing missing values:", df.shape)
```

After removing missing values: (506, 14)

```
df.columns = [col.strip().lower() for col in df.columns]
```

```
print("\nData Types:\n", df.dtypes)
```

Data Types:

crim	float64
zn	float64
indus	float64
chas	int64
nox	float64
rm	float64
age	float64
dis	float64
rad	int64
tax	int64
ptratio	float64
b	float64
lstat	float64
medv	float64
dtype:	object

```
if 'location' in df.columns:
    df = pd.get_dummies(df, columns=['location'], drop_first=True)
```

```
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, random_state=42)
```

```
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
```

```
X_test = scaler.transform(X_test)
```

```
model = RandomForestRegressor(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
```

RandomForestRegressor ⓘ ⓘ  
RandomForestRegressor(random\_state=42)

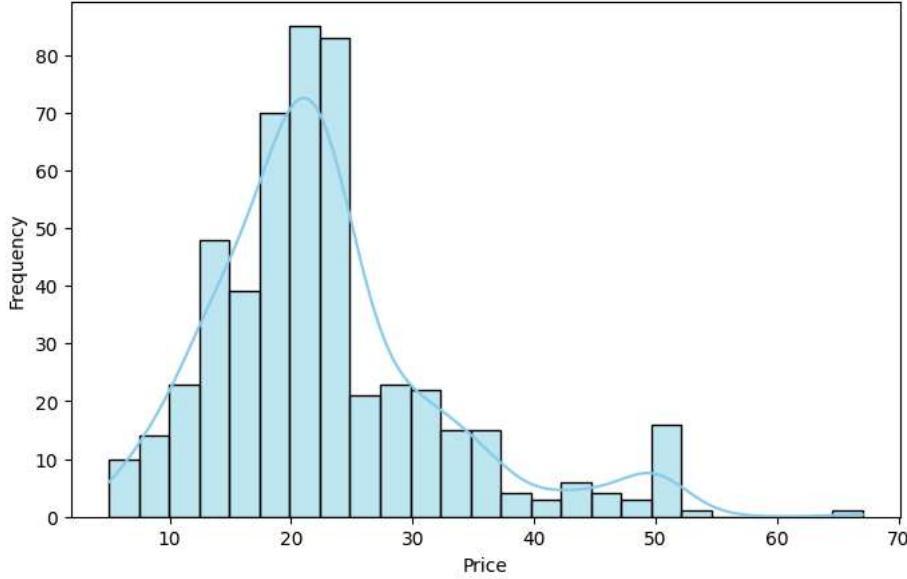
```
y_pred = model.predict(X_test)
```

```
mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
```

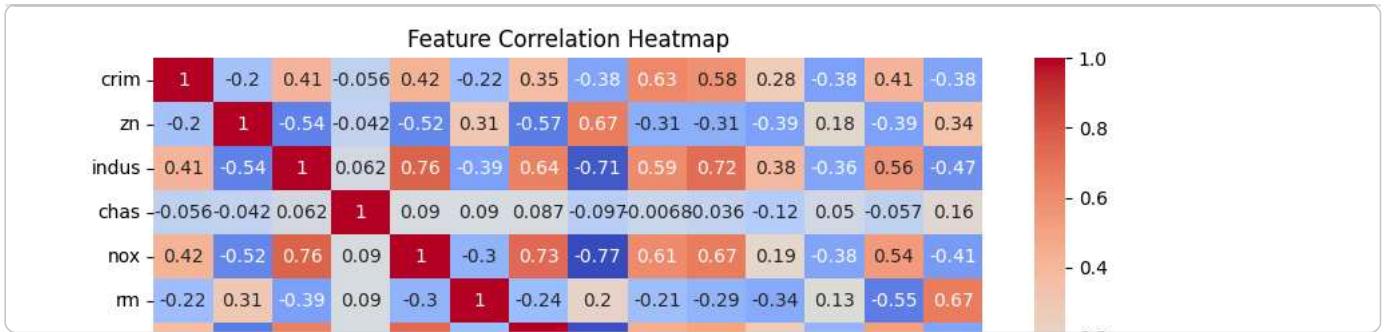
```
mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
```

```
plt.figure(figsize=(8,5))
sns.histplot(df['medv'], kde=True, bins=25, color='skyblue')
plt.title("Distribution of Land Prices")
plt.xlabel("Price")
plt.ylabel("Frequency")
plt.show()
```

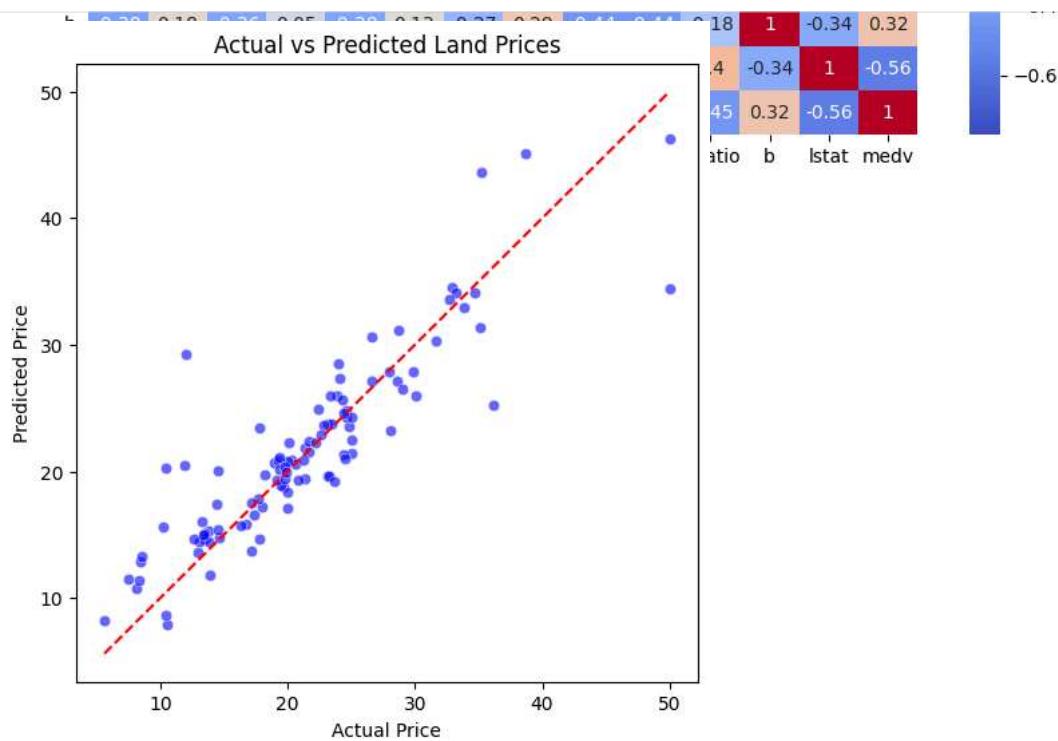
Distribution of Land Prices



```
plt.figure(figsize=(10,6))
sns.heatmap(df.corr(), annot=True, cmap='coolwarm')
plt.title("Feature Correlation Heatmap")
plt.show()
```



```
plt.figure(figsize=(6,6))
sns.scatterplot(x=y_test, y=y_pred, color='blue', alpha=0.6)
plt.plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--') # Diagonal line
plt.xlabel("Actual Price")
plt.ylabel("Predicted Price")
plt.title("Actual vs Predicted Land Prices")
plt.show()
```



```
importances = model.feature_importances_
indices = np.argsort(importances)[::-1]
```

```
plt.figure(figsize=(10,6))
plt.title("Feature Importance in Land Price Prediction")
plt.bar(range(len(indices)), importances[indices], color='teal', align='center')
plt.xticks(range(len(indices)), np.array(X.columns)[indices], rotation=90)
plt.tight_layout()
plt.show()
```

