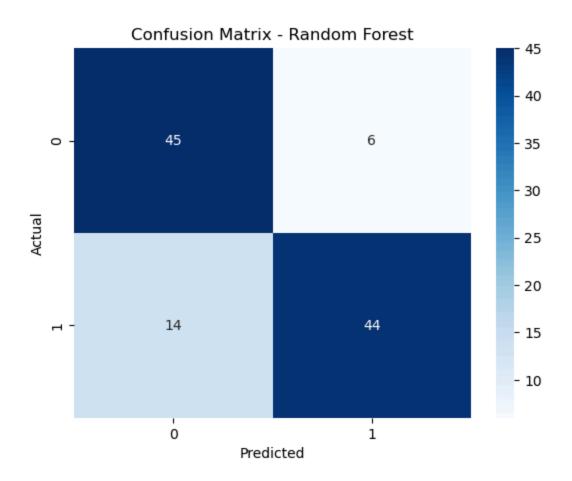


```
In [18]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.model selection import train test split
         from sklearn.preprocessing import LabelEncoder, StandardScaler
         from sklearn.linear model import LogisticRegression
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.metrics import accuracy score, classification report, confusion m
In [22]: #Load and inspect data
         data = pd.read csv("Housing.csv")
         #Clean data
         data = data.dropna()
         data.columns = data.columns.str.strip().str.lower()
         print(data.head())
             price area bedrooms bathrooms stories mainroad questroom basement
       0 13300000 7420
                                4
                                            2
                                                      3
                                                             yes
                                                                        no
                                                                                 no
       1 12250000 8960
                                 4
                                            4
                                                      4
                                                             yes
                                                                        no
                                                                                 no
                                 3
                                            2
       2 12250000 9960
                                                      2
                                                             yes
                                                                        no
                                                                                yes
       3 12215000 7500
                                 4
                                             2
                                                      2
                                                             yes
                                                                        no
                                                                                yes
                                             1
                                                      2
       4 11410000 7420
                                                             yes
                                                                       yes
                                                                                yes
         hotwaterheating airconditioning parking prefarea furnishingstatus
       0
                                     yes
                                                 2
                                                        yes
                                                                   furnished
       1
                                     yes
                                                 3
                                                                   furnished
                      no
                                                         no
       2
                                                 2
                                                        yes
                                                              semi-furnished
                      no
                                      no
       3
                      no
                                      yes
                                                 3
                                                        yes
                                                                   furnished
                                                                   furnished
                      nο
                                     yes
                                                        no
In [26]: #Create target label (binary classification)
         #Above-median price = High (1), else Low (0)
         data['price category'] = np.where(data['price'] > data['price'].median(), 1, 6
         #Encode categorical variables
         le = LabelEncoder()
         for col in data.select dtypes(include=['object']).columns:
             data[col] = le.fit transform(data[col])
         #features and target
         X = data.drop(columns=['price', 'price category'])
         y = data['price_category']
         #Split data
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, rando
         #Scale features
         scaler = StandardScaler()
         X train = scaler.fit transform(X train)
         X test = scaler.transform(X test)
         #Train models
```

```
models = {
             "Logistic Regression": LogisticRegression(),
             "Random Forest": RandomForestClassifier(n estimators=100, random state=42)
         }
         for name, model in models.items():
             model.fit(X train, y train)
             y pred = model.predict(X test)
             acc = accuracy score(y test, y pred)
             print(f"\n{name}")
             print(f"Accuracy: {acc:.2f}")
             print(classification report(y test, y pred))
         #Confusion Matrix for Random Forest
         rf = models["Random Forest"]
         y pred rf = rf.predict(X test)
         cm = confusion_matrix(y_test, y_pred_rf)
       Logistic Regression
       Accuracy: 0.84
                     precision recall f1-score support
                          0.78
                                    0.92
                                              0.85
                                                          51
                  0
                  1
                          0.92
                                    0.78
                                              0.84
                                                          58
                                              0.84
                                                         109
           accuracy
                                    0.85
                                              0.84
                                                         109
                          0.85
          macro avg
                          0.86
                                    0.84
                                              0.84
                                                         109
       weighted avg
       Random Forest
       Accuracy: 0.82
                     precision recall f1-score support
                  0
                          0.76
                                    0.88
                                              0.82
                                                          51
                  1
                          0.88
                                    0.76
                                              0.81
                                                          58
                                              0.82
                                                         109
           accuracy
                          0.82
                                    0.82
                                              0.82
                                                         109
          macro avg
                                    0.82
                                              0.82
       weighted avg
                          0.83
                                                         109
         sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
In [28]:
         plt.title("Confusion Matrix - Random Forest")
         plt.xlabel("Predicted")
         plt.ylabel("Actual")
```

plt.show()



```
importances = pd.Series(rf.feature_importances_, index=X.columns)
importances = importances.sort_values(ascending=False) * 100 # Convert to %
importances.head(10).plot(kind='bar', color='teal')
plt.title("Top 10 Feature Importances (%)")
plt.ylabel("Importance (%)")
plt.xlabel("Features")
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```

