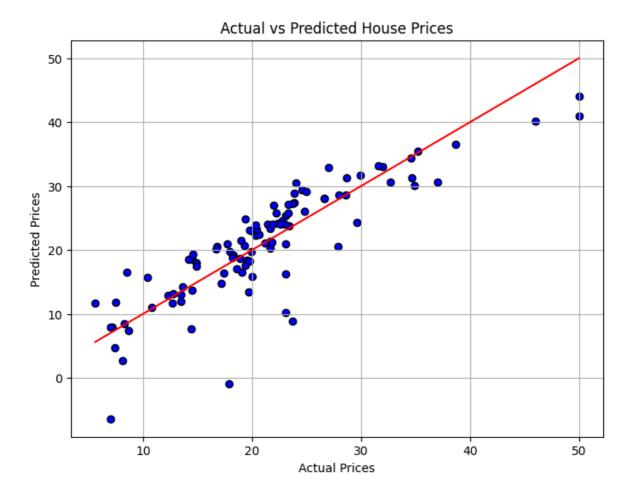
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```
In [5]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression
         from sklearn import metrics
         df=pd.read_csv('BostonHousing.csv')
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         print("\nMissing values:")
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         df.dropna(inplace=True)
In [ ]:
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```
In [ ]: X = df.drop('medv', axis=1)
        y = df['medv']
In [ ]: X_train, X_test, y_train, y_test = train_test_split(
            X, y, test_size=0.2, random_state=42)
In [ ]: model = LinearRegression()
        model.fit(X_train, y_train)
Out[]: ▼ LinearRegression
        LinearRegression()
In [ ]: y_pred = model.predict(X_test)
In [ ]: mae = metrics.mean_absolute_error(y_test, y_pred)
        mse = metrics.mean_squared_error(y_test, y_pred)
        rmse = np.sqrt(mse)
        r2 = metrics.r2_score(y_test, y_pred)
In [ ]: print(f"\nModel Performance:")
        print(f"Mean Absolute Error (MAE): {mae:.2f}")
        print(f"Mean Squared Error (MSE): {mse:.2f}")
        print(f"Root Mean Squared Error (RMSE): {rmse:.2f}")
        print(f"R2 Score: {r2:.2f}")
       Model Performance:
       Mean Absolute Error (MAE): 3.24
       Mean Squared Error (MSE): 20.69
       Root Mean Squared Error (RMSE): 4.55
       R<sup>2</sup> Score: 0.72
In [ ]: plt.figure(figsize=(8,6))
        plt.scatter(y_test, y_pred, color='blue', edgecolors='black')
        plt.xlabel("Actual Prices")
        plt.ylabel("Predicted Prices")
        plt.title("Actual vs Predicted House Prices")
        plt.plot([min(y_test), max(y_test)], [min(y_test), max(y_test)], color='red') #
        plt.grid(True)
        plt.show()
```

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In []: