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In [25]: import pandas as pd
                    import numpy as np
                    import matplotlib.pyplot as plt
                    from sklearn.model_selection import train_test_split
                    from sklearn.linear model import LinearRegression
                   from sklearn.preprocessing import PolynomialFeatures
                    from sklearn.tree import DecisionTreeRegressor
                    from sklearn.ensemble import RandomForestRegressor
                   from sklearn.svm import SVR
                    from sklearn.neural_network import MLPRegressor
                    from sklearn.metrics import mean_squared_error, r2_score
                   df = pd.read_csv('scr-dataset.csv')
                   X = df[['x']] values
                   y = df['y'].values
                   X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, rain_test_split(X, y, test_size=0.2, rain_test_size=0.2, rain_test_split(X, y, test_size=0.2, rain_test_split(X, y, test_size=0.2, rain_test_split(X, y, test_size=0.2, rain_test_split(X, y, test_size=0.2, rain_test_split(X, y, y, test_size=0.2
                   models = {
                            'Linear Regression': LinearRegression(),
                            'Polynomial Regression (degree=2)': PolynomialFeatures(degree=2),
                            'Decision Tree': DecisionTreeRegressor(),
                            'Random Forest': RandomForestRegressor(n estimators=100, random state=41
                            'Support Vector Machine': SVR(),
                            'Neural Network': MLPRegressor(max_iter=1000, random_state=42)
                   def fit_and_predict(model, X_train, y_train, X_test, x_value):
                            if isinstance(model, PolynomialFeatures):
                                    X_poly_train = model.fit_transform(X_train)
                                    X_poly_test = model.transform(X_test)
                                    lin_reg = LinearRegression().fit(X_poly_train, y_train)
                                    y_pred = lin_reg.predict(X_poly_test)
                                    y_value_pred = lin_reg.predict(model.transform([[x_value]]))[0]
                            else:
                                    model.fit(X_train, y_train)
                                    y_pred = model.predict(X_test)
                                    y_value_pred = model.predict([[x_value]])[0]
                            return y_pred, y_value_pred
                   x_value = 50
                    predictions = {}
                    for name, model in models.items():
                            y pred, y_value_pred = fit_and_predict(model, X_train, y_train, X_test,
                           mse = mean_squared_error(y_test, y_pred)
                            r2 = r2_score(y_test, y_pred)
                            predictions[name] = (y_value_pred, mse, r2)
                   for name, (y_value_pred, mse, r2) in predictions.items():
                            print(f''\{name\}: \n - Predicted y at x = \{x_value\}: \{y_value_pred\} \n - MSI
                   plt.figure(figsize=(10, 6))
                   plt.scatter(X, y, color='blue', label='Actual Data')
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for name, model in models.items():
    if isinstance(model, PolynomialFeatures):
        X_{poly} = model.transform(X)
        lin_reg = LinearRegression().fit(model.transform(X_train), y_train)
        plt.plot(X, lin_reg.predict(X_poly), label=name)
    else:
        model.fit(X, y)
        plt.plot(X, model.predict(X), label=name)
plt.scatter(x_value, predictions['Linear Regression'][0], color='green', lat
plt.title('y vs x')
plt.xlabel('x')
plt.ylabel('y')
plt.grid(True)
plt.legend()
plt.show()
Linear Regression:
 - Predicted y at x = 50: -0.12139300595210278
 - MSE: 0.816
 - R^2: -0.021
Polynomial Regression (degree=2):
 - Predicted y at x = 50: -0.11411178551411022
 - MSE: 0.815
 - R^2: -0.021
Decision Tree:
 - Predicted y at x = 50: 1.7594025983376151
 - MSE: 0.034
 - R^2: 0.957
Random Forest:
 - Predicted y at x = 50: 1.6997102509493058
 - MSE: 0.017
 - R^2: 0.979
Support Vector Machine:
 - Predicted y at x = 50: 0.012436371614998845
 - MSE: 0.849
 - R^2: -0.063
Neural Network:
 - Predicted y at x = 50: -0.08630193546583581
 - MSE: 0.798
 - R^2: 0.001
```

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