

```
In [ ]: import numpy as np
import matplotlib.pyplot as plt
from numpy.linalg import solve
```

Question 4. Computation problem

• Given function

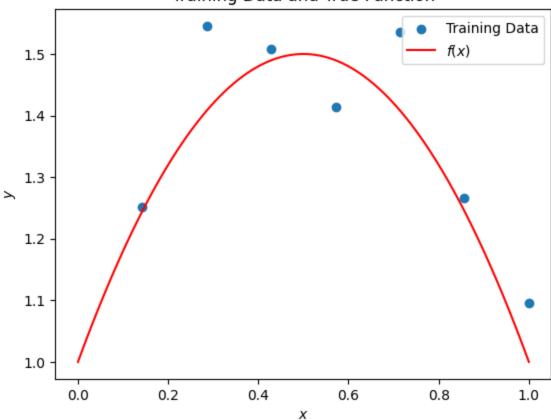
```
In [ ]: def f(x):
    return 1 + 2 * x - 2 * x**2
```

1. Generate training data and plot

```
In []: x_train = np.array([i/7 for i in range(1, 8)])
y_train = f(x_train) + 0.1 * np.random.normal(0, 1, size=len(x_train))

plt.scatter(x_train, y_train, label='Training Data')
x_vals = np.linspace(0, 1, 100)
plt.plot(x_vals, f(x_vals), label='$f(x)$', color='red')
plt.xlabel('$x$')
plt.ylabel('$x$')
plt.legend()
plt.title('Training Data and True Function')
plt.show()
```

Training Data and True Function



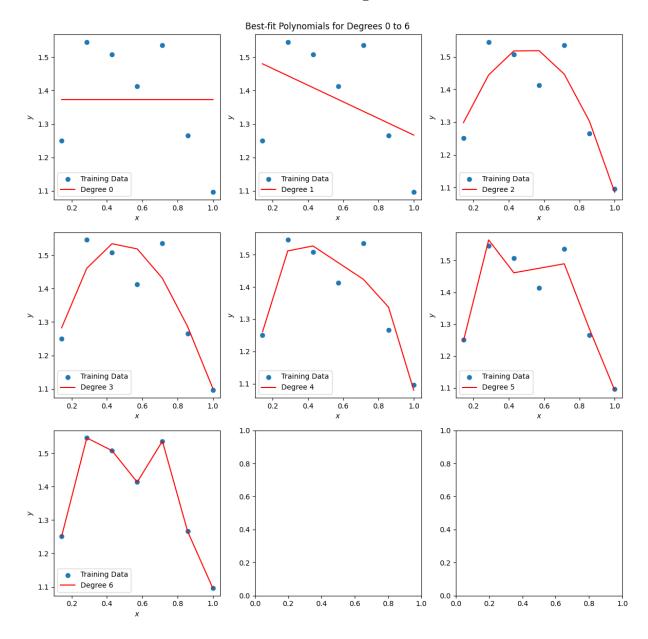
2. Find best-fit polynomials for degrees 0 to 6 - Subplots

```
In [ ]: fig, axs = plt.subplots(3, 3, figsize=(12, 12))
fig.suptitle('Best-fit Polynomials for Degrees 0 to 6')

for d, ax in zip(range(7), axs.flatten()):
    X = np.vander(x_train, d+1, increasing=True)
    theta = solve(X.T @ X, X.T @ y_train)
    y_pred = X @ theta
    ax.scatter(x_train, y_train, label='Training Data')
    ax.plot(x_train, y_pred, label=f'Degree {d}', color='red')
    ax.set_xlabel('$x$')
    ax.set_ylabel('$y$')
    ax.legend()

plt.tight_layout()

plt.show()
```



3.Plot training MSE vs degree

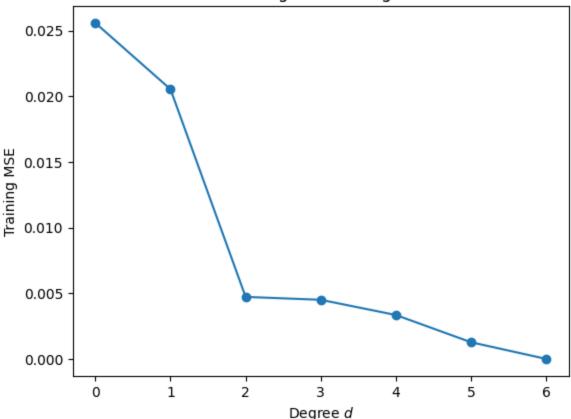
```
In [ ]: degrees = range(7)
    mse_train = []

for d in degrees:
        X_train = np.vander(x_train, d+1, increasing=True)
        theta = solve(X_train.T @ X_train, X_train.T @ y_train)
        y_pred = X_train @ theta
        mse_train.append(np.mean((y_train - y_pred)**2))

plt.plot(degrees, mse_train, marker='o')
    plt.xlabel('Degree $d$')
    plt.ylabel('Training MSE')
    plt.title('Training MSE')
    plt.show()

print(f'The degree {mse_train.index(min(mse_train))} polynomial gives the left
```

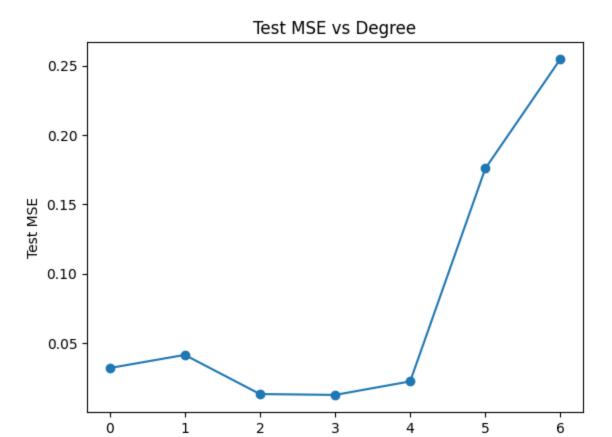




The degree 6 polynomial gives the least MSE: 2.639421255187896e-17

4. Generate test data and plot test MSE vs degree

```
In []: x test = np.random.rand(1000)
        y test = f(x \text{ test}) + 0.1 * np.random.normal(0, 1, size=len(x test))
         degrees = range(7)
        mse_test = []
         for d in degrees:
             X_train = np.vander(x_train, d+1, increasing=True)
            X \text{ test} = \text{np.vander}(x \text{ test, d+1, increasing=True})
             theta = solve(X train.T @ X train, X train.T @ y train)
             y_pred = X_test @ theta
             mse test.append(np.mean((y test - y pred)**2))
         plt.plot(degrees, mse test, marker='o')
         plt.xlabel('Degree $d$')
         plt.ylabel('Test MSE')
         plt.title('Test MSE vs Degree')
         plt.show()
         print(f'The degree {mse test.index(min(mse test))} polynomial gives the leas
```



The degree 3 polynomial gives the least MSE : 0.012373715423148707

Degree d

In []: