

# Homework 1

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Github: [https://github.com/jaskinkabir/Intro\\_ML/tree/main/HM1](https://github.com/jaskinkabir/Intro_ML/tree/main/HM1)

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In [ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

filePath="D3.csv"

df = pd.read_csv(filePath, index_col=False)

X_df = df.iloc[:, :-1]

data = df.to_numpy()

m = data.shape[0]
```

```
In [ ]: def gen_data(df: pd.DataFrame):
    X_df = df.iloc[:, :-1]
    data = X_df.to_numpy()

    Y = df.iloc[:, -1].to_numpy().reshape(-1,1)

    X0 = np.ones((data.shape[0], 1))
    X = np.hstack((X0, data))
    return X, Y
```

```
In [ ]: def compute_cost(X, Y, theta):
    predictions = X.dot(theta)
    errors = np.subtract(predictions, Y)
    sqrErrors = np.square(errors)
    J = (1/ (2*m) ) * np.sum(sqrErrors)

    return J, predictions
```

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In [ ]: def grad_desc(X: np.ndarray, Y: np.ndarray, theta: np.ndarray, alpha: float, iterations):
    m = len(Y)

    cost_history = np.zeros(iterations)

    for i in range(iterations):
        predictions = X.dot(theta)
        errors = np.subtract(predictions, Y)
```

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sum_deltas = (alpha/m) * X.T.dot(errors)

theta -= sum_deltas

sqrErrors = np.square(errors)
cost_history[i]= (1/ (2*m) ) * np.sum(sqrErrors)

return theta,cost_history

```

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In [ ]: def print_model(X, Y, theta, name='Y'):

    cost, pred = compute_cost(X,Y,theta)

    model_str = ""

    for i in range(theta.shape[0]-1, -1, -1):
        if i == 0:
            model_str += f"{round(theta[i,0], 3)}"
            break
        model_str += f"{round(theta[i,0], 3)}*X{i} + "

    print(f"Model:\n    {name} = {model_str}")
    print(f"    cost = {round(cost, 3)}")
    return pred

```

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In [ ]: # Sequential Training
X,Y = gen_data(df)

iterations = 150
alpha = 0.1

fig, ax = plt.subplots(2,2, squeeze=True)

axes: np.ndarray[plt.Axes] = [ax[0,0], ax[0,1], ax[1,0], ax[1,1]]
colors = ['r', 'g', 'b']

iters = range(iterations)
for i in range(1,4):

    X_i = np.hstack(( X[:,0].reshape(-1,1), X[:,i].reshape(-1,1) ))

    theta_col = np.zeros((2,1))
    theta_col, cost_history = grad_desc(X_i, Y, theta_col, alpha, iterations)

    pred = print_model(X_i, Y, theta_col, f"y{i}")
    pred = X_i.dot(theta_col)

    axes[i-1].scatter(X[:,i], Y, label=f"Ground Truth", color='m', marker='+')
    axes[i-1].scatter(X[:,i], pred, label=f"y{i}", color=colors[i-1], marker='+')
    axes[i-1].legend()
    axes[i-1].set_xlabel(f"X{i}")
    axes[i-1].set_title(f"y{i} vs. X{i}")

    axes[3].plot(iters, cost_history, label=f"X{i}", color=colors[i-1])

```

```
plt.title("Isolated Training Costs")
plt.xlabel("Iteration")
plt.ylabel("Cost")
plt.legend()

print('\n')
# print_model(X, Y, theta)
plt.tight_layout()

plt.show()
```

Model:

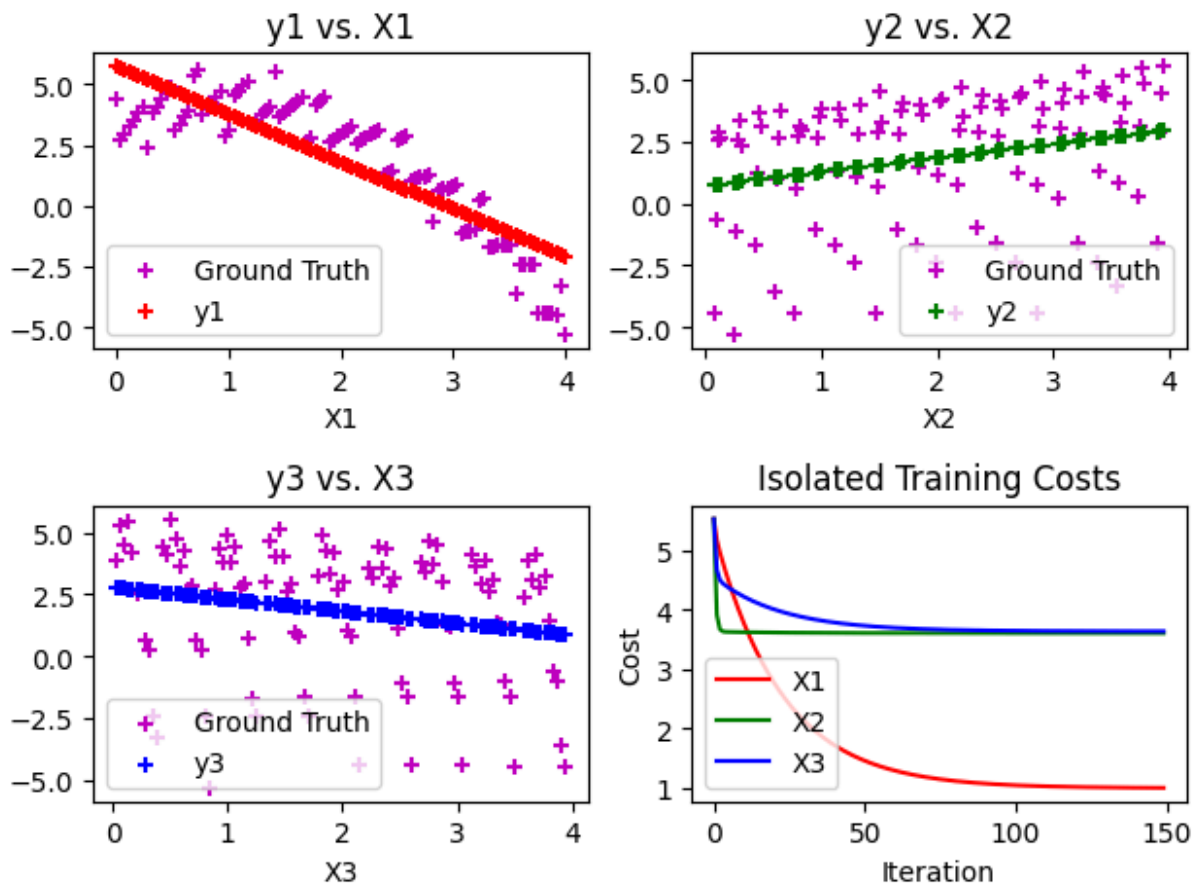
$y_1 = -1.96 \cdot X_1 + 5.725$   
cost = 0.99

Model:

$y_2 = 0.564 \cdot X_1 + 0.72$   
cost = 3.599

Model:

$y_3 = -0.486 \cdot X_1 + 2.784$   
cost = 3.63



```
In [ ]: #Simultaneous Descent
theta = np.zeros((X.shape[1], 1))

# 300 iters vs. 1000 iters only has a difference of .004 in cosst
iterations = 250
# Model still converges at max alpha
```

```

alpha = 0.1

theta, cost_history = grad_desc(X, Y, theta, alpha, iterations)

plt.plot(range(iterations), cost_history)
plt.title("Three-Dimensional Training Cost")
plt.xlabel("Iteration")
plt.ylabel("Cost")

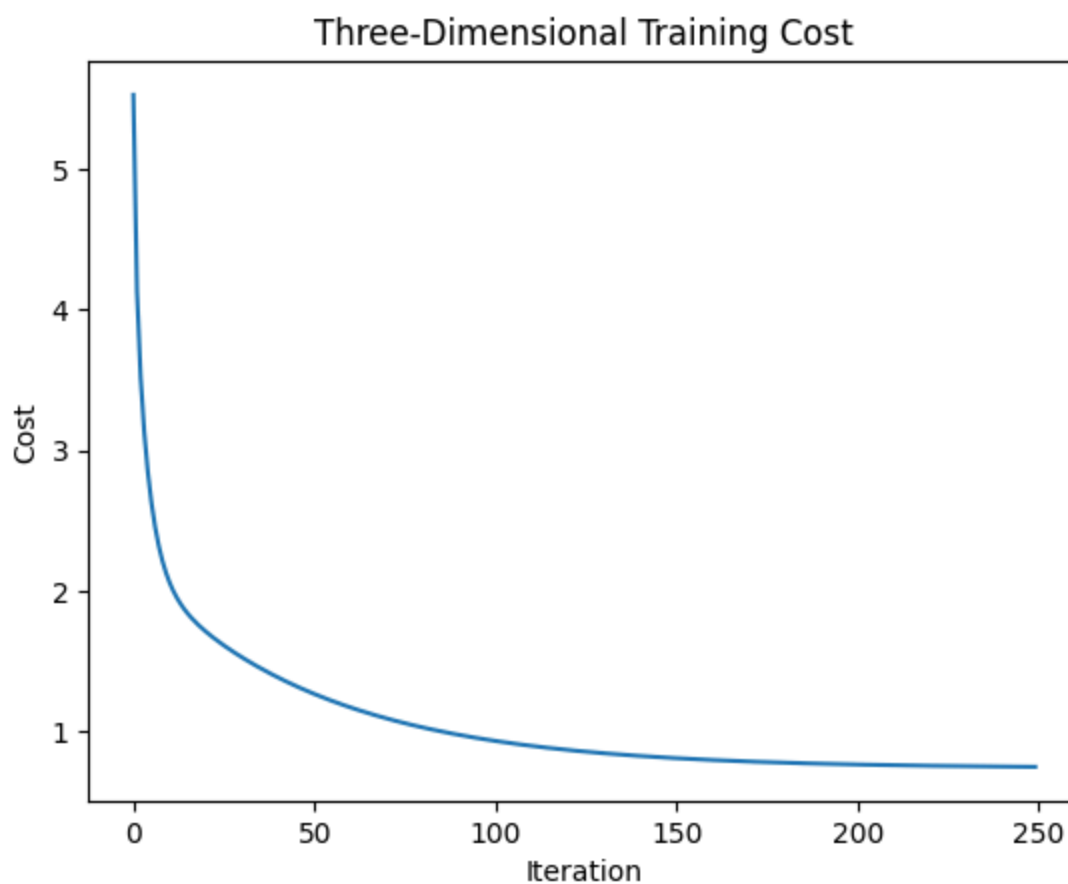
print_model(X, Y, theta, 'y')
plt.show()

```

Model:

$$y = -0.203 \cdot X_3 + 0.603 \cdot X_2 + -1.944 \cdot X_1 + 4.89$$

cost = 0.748



```

In [ ]: X_pred = np.array([
#   X0,1,2,3
[1,1,1,1],
[1,2,0,4],
[1,3,2,1],
])
X_pred.dot(theta)

```

```

Out[ ]: array([[3.34542262],
[0.18869109],
[0.06048419]])

```