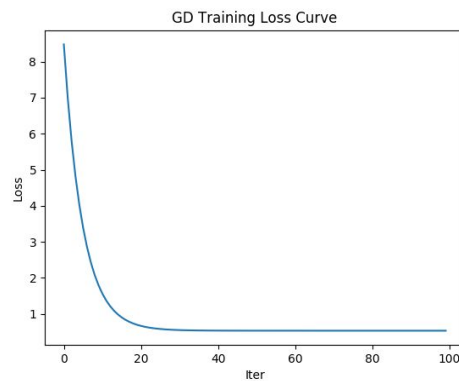


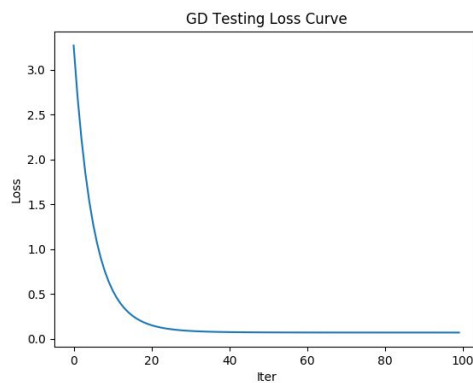
## Part. 1, Coding:

### 1. Learning Curve

- Training loss curve



- Testing loss curve



### 2. Mean Square Error

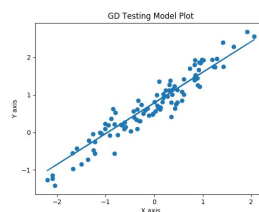
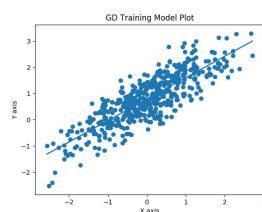
- Training MSE: 0.5347
- Testing MSE: 0.0687

```
Training loss 0.5347280122225704, Testing loss 0.06869637127070814
```

### 3. Weights and Intercepts

- Weights: 0.8180
- Intercepts: 0.7845

```
Model Weight and Intercepts: [[0.8179535 0.7845594]]
```



### 4. Difference between GD, Mini-Batch GD, SGD

- Gradient Descent

- feed all the training data into model at one time, and compute the gradient, then doing backpropagation.
- Mini-Batch Gradient Descent
  - divide the training data into some batches, and feed one batch at one time, then compute gradient and doing backpropagation.
- Stochastic Gradient Descent
  - feed one data into the model at one time, just like mini-batch gradient descent with batch size 1.

## Part. 2, Questions:

$$\begin{aligned}
 p(\text{guava}) &= p(R)p(\text{guava}|R) + p(G)p(\text{guava}|G) \\
 &\quad + p(B)p(\text{guava}|B) \\
 &= 0.2 \cdot \frac{3}{10} + 0.4 \cdot \frac{4}{20} + 0.4 \cdot \frac{2}{4} \\
 &= 0.06 + 0.08 + 0.2 \\
 &= \underline{\underline{0.34}} \#
 \end{aligned}$$

$$\begin{aligned}
 p(B | \text{apple}) &= \frac{p(\text{blue, apple})}{p(\text{apple})} = \frac{p(\text{apple}|B)p(B)}{p(\text{apple})} \\
 &= \frac{0.4 \cdot \frac{2}{4}}{p(R)p(\text{apple}|R) + p(G)p(\text{apple}|G) + p(B)p(\text{apple}|B)} \\
 &= \frac{0.2}{0.2 \cdot \frac{3}{10} + 0.4 \cdot \frac{12}{20} + 0.4 \cdot \frac{2}{4}} = \frac{0.2}{0.06 + 0.24 + 0.2} \\
 &= \frac{0.2}{0.5} = \underline{\underline{0.4}} \#
 \end{aligned}$$

1.

$$\begin{aligned}
 \text{var}[f] &= E[(f(x) - E[f(x)])^2] \\
 &= E[f(x)^2 - 2f(x)E[f(x)] + E[f(x)]^2] \\
 &= E[f(x)^2] - 2E[f(x)E[f(x)]] + E[f(x)]^2 \\
 &= E[f(x)^2] - 2E[f(x)]^2 + E[f(x)]^2 \\
 &= E[f(x)^2] - E[f(x)]^2 \#
 \end{aligned}$$

2.