

# Weekly Report

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# This Week

- Adjust weights by using stochastic gradient descent.

# Stochastic Gradient Descent

- Let  $z=(x,y)$  be the data input/output pair
- Let  $\text{loss}(\hat{y},y)$  be the loss function
  - In our model, we use least squares function
- Let  $f_w(x)$  be the function parameterized by a weight vector  $w$
- We seek the function  $f_w(x)$  that minimizes  $\text{loss}(f_w(x),y)$  on the sample

# Stochastic Gradient Descent

- At first, let  $w_0 = [1/n, 1/n, \dots]$  where  $n$  is the number of trees
- For each sample  $z_t = (x_t, y_t)$  in the new coming data, update weights vector:
  - $w_{t+1} = w_t - \alpha \nabla_w \text{loss}(f_w(x_t), y_t)$
  - $w_{t+1} = w_t + 2\alpha (y_t - f_w(x_t)) [g(x_t) \cdot w_t]$ 
    - $g(x_t)$  is the vector consisting of outputs from each tree

# Functions

- `update(X, y, alpha)`
  - `X` is the data input
  - `y` is the data output
  - `alpha` is the constant learning rate
  - this function iterates each data point and adjusts the weight vector by stochastic gradient descent method.

# Next Step

- Test the synthetical data