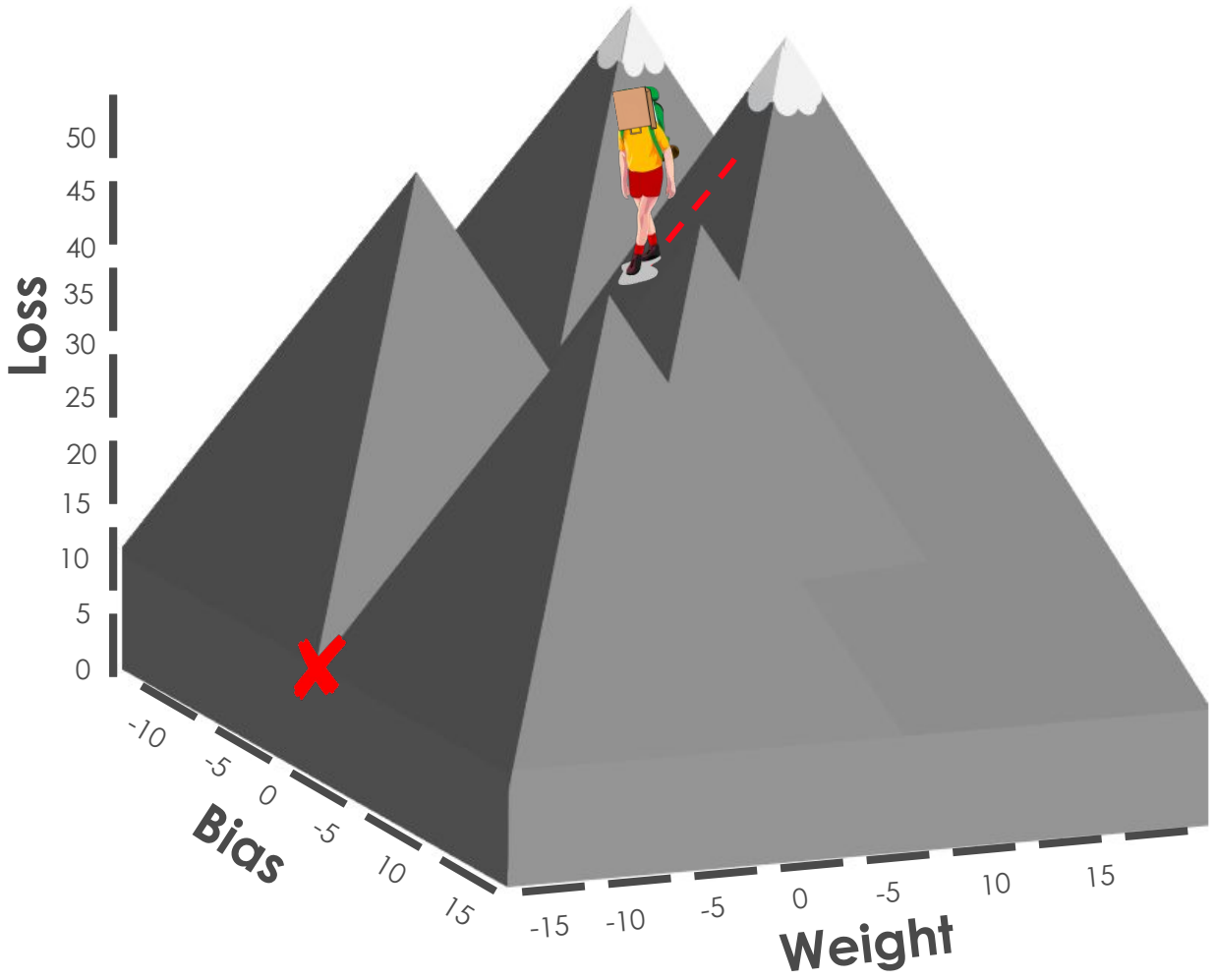
Calculating the Gradient

Deep Learning Pre-Work

Gradient Descent Steps



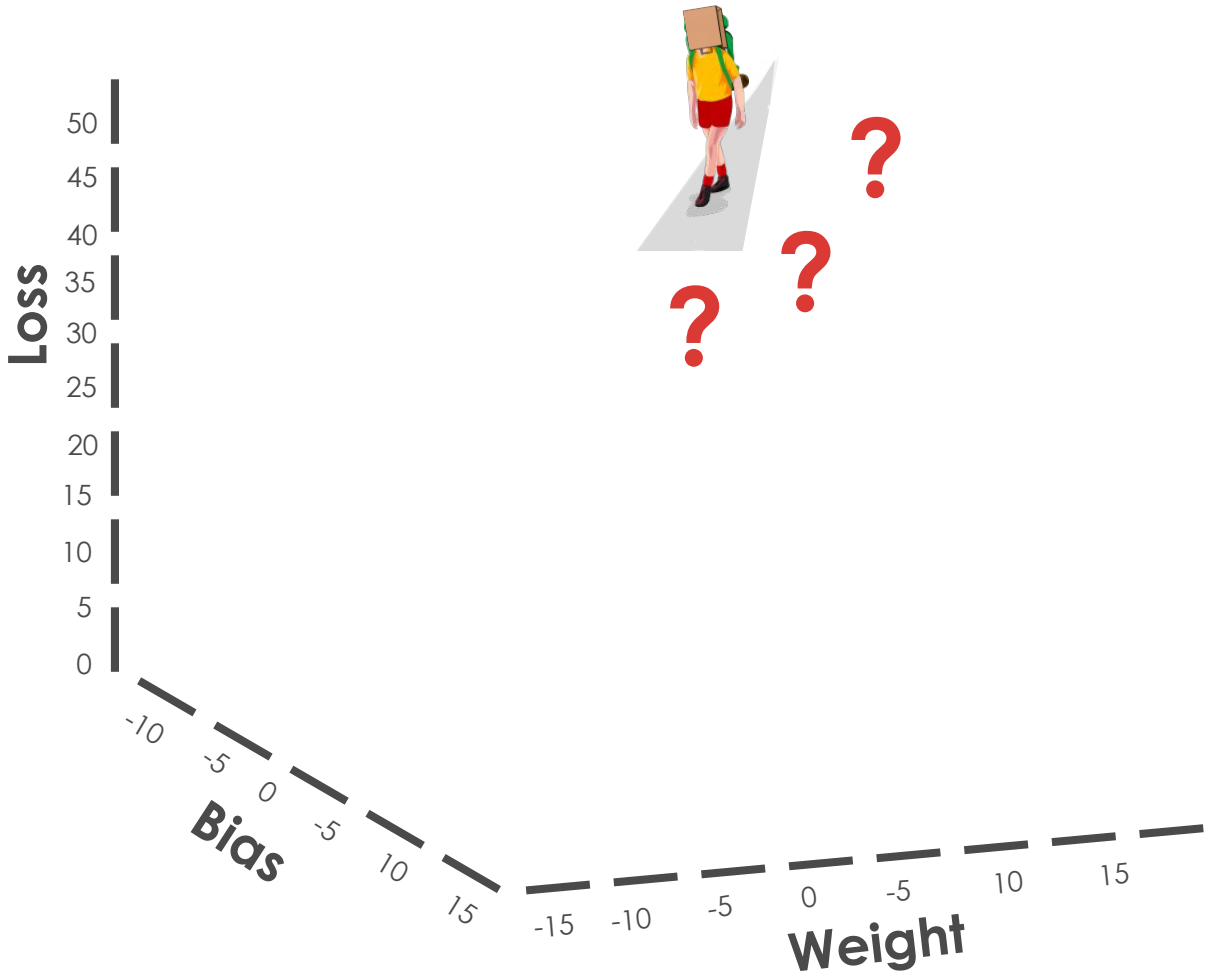
Step 1 Start at a random bias and weight and calculate the loss

Step 2 Take a step in the direction with the steepest gradient

Step 3 Calculate the new loss

Step 4 Repeat steps 2 and 3

Gradient Descent Steps



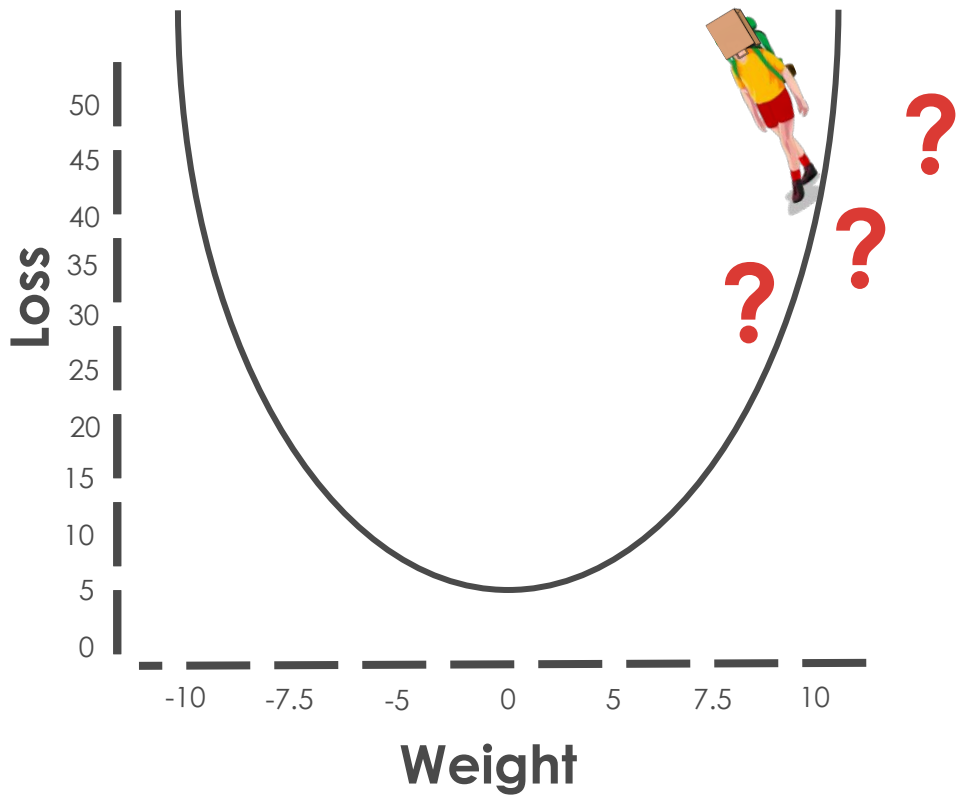
Step 1 **Start at a random bias and weight and calculate the loss**

Step 2 Take a step in the direction with the steepest gradient

Step 3 Calculate the new loss

Step 4 Repeat steps 2 and 3

Gradient Descent Steps



Step 1 Start at a random bias and weight and calculate the loss

Step 2 Take a step in the direction with the steepest gradient

Step 3 Calculate the new loss

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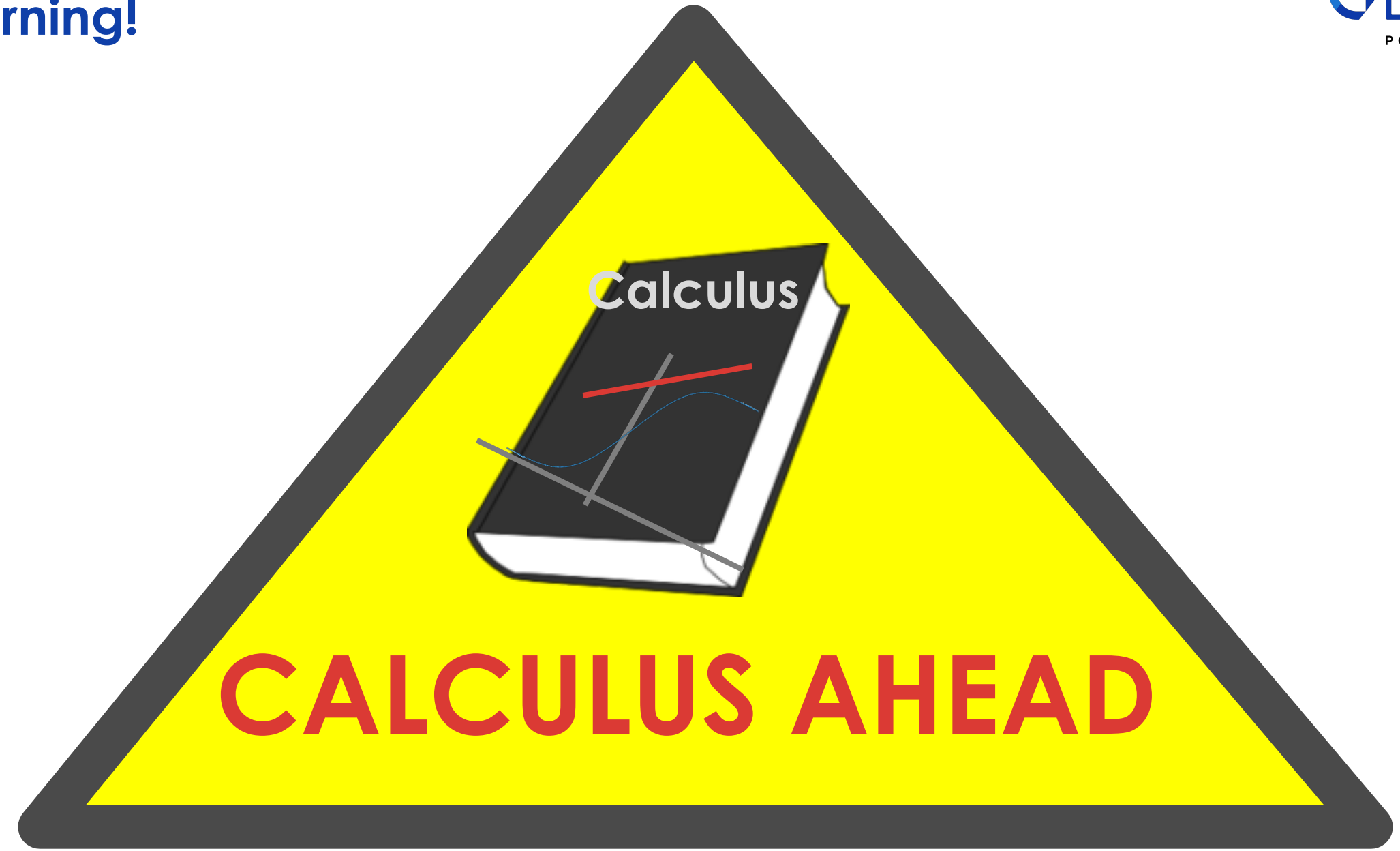
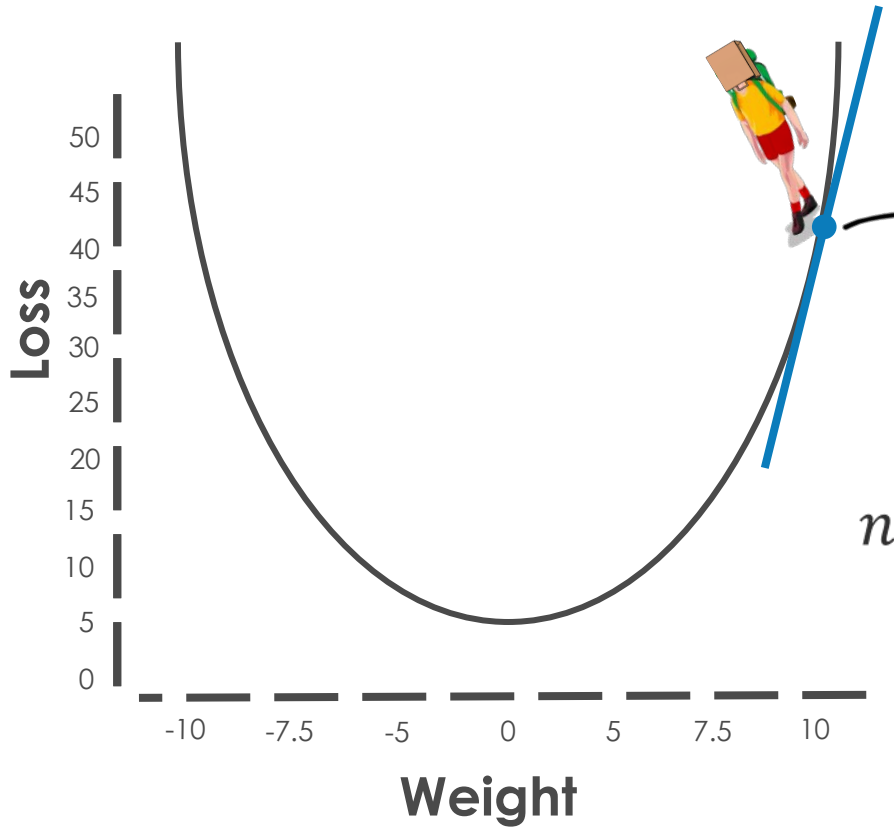
Warning!



Derivatives

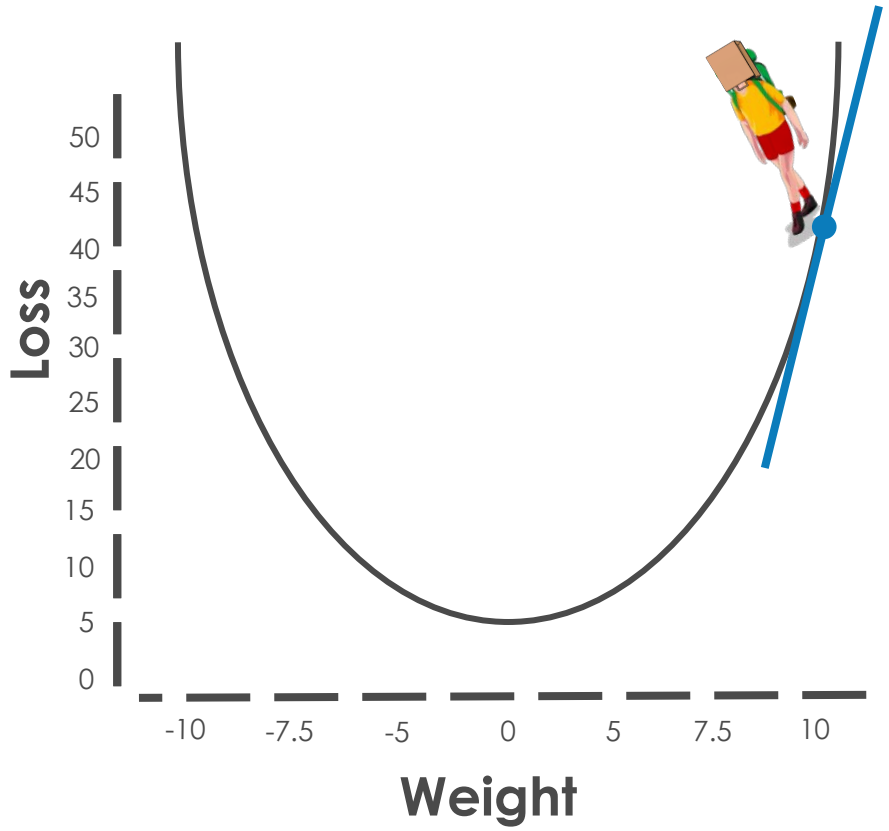
The derivative

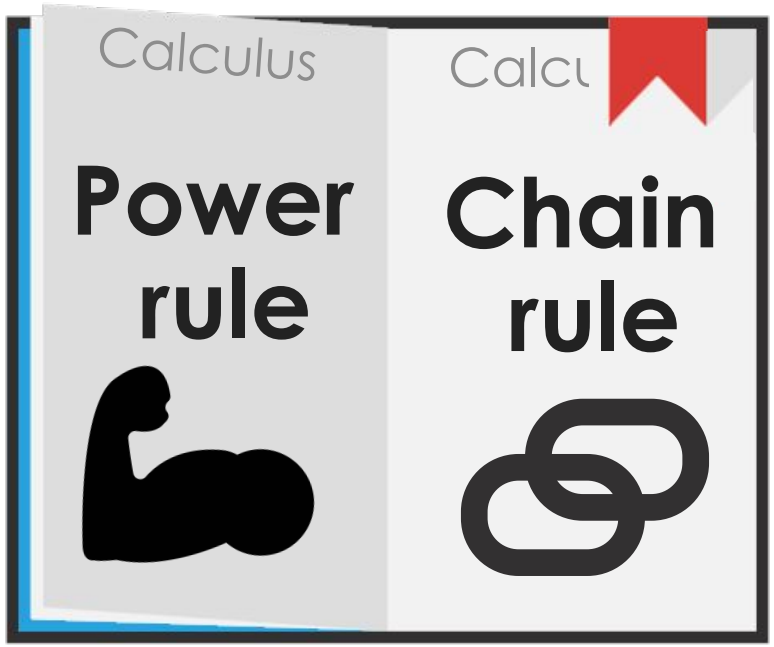


$$\frac{d(loss)}{d(weight)}$$

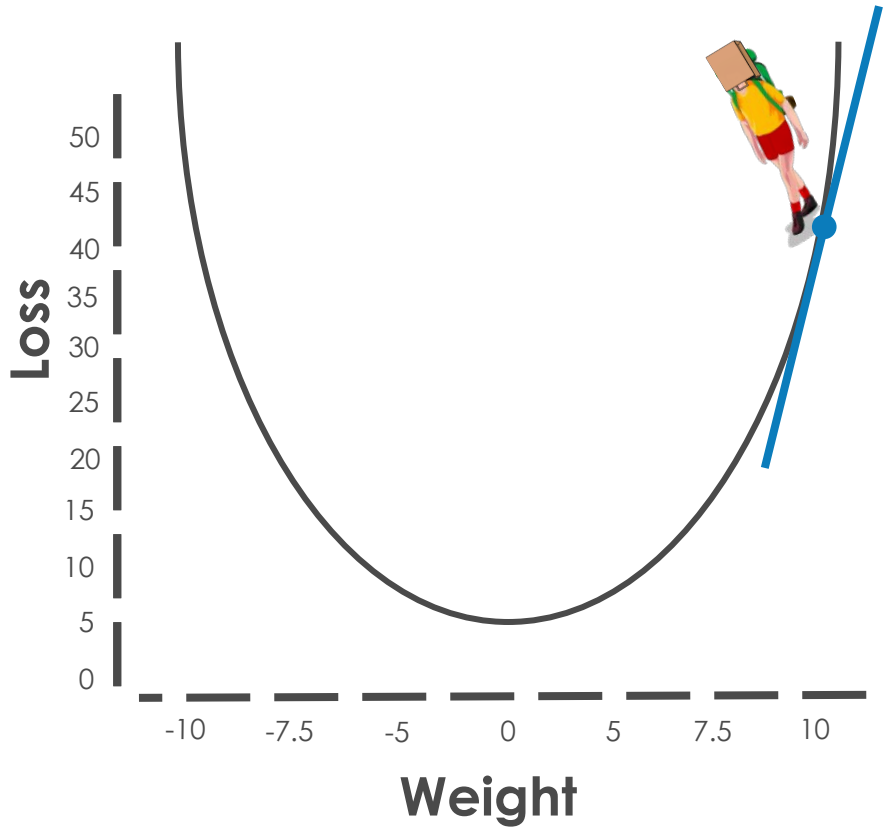
$$new\ weight = weight - \frac{d(loss)}{d(weight)} * learning\ rate$$

Calculating the derivative





Power Rule



$$\frac{d}{d(x)} x^n = nx^{n-1}$$

Power rule

mean ← $\frac{1}{n}\sum$ ← error → squared

$$loss = \frac{1}{n}\sum (\hat{y} - y)^2$$

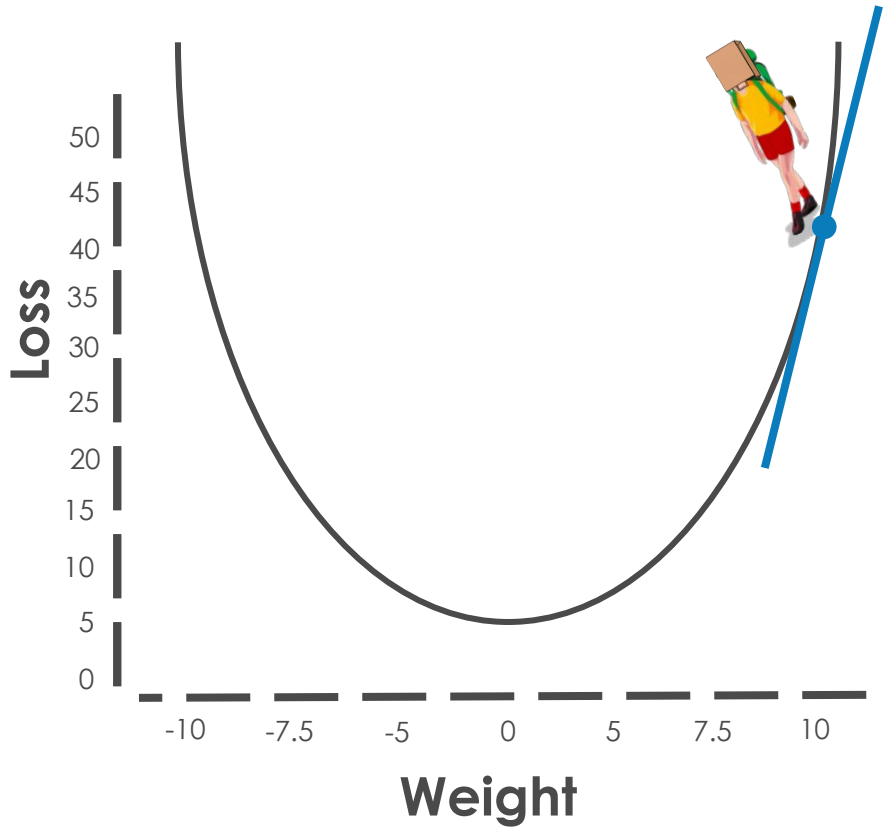
$$loss = error^2$$

$$\frac{d(loss)}{d(error)} = 2erro$$

$$\frac{d(loss)}{d(error)} = 2error$$

Power rule

Chain Rule



$$y = x$$
$$x = z$$

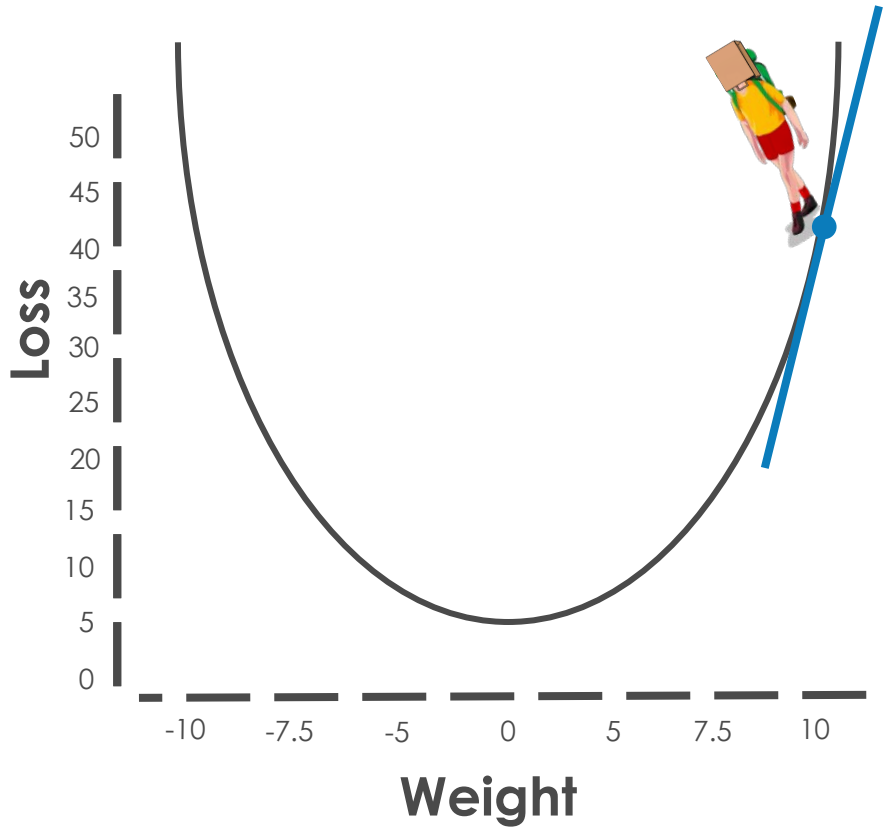
$$\frac{d(y)}{d(z)} = \frac{d(y)}{d(x)} * \frac{d(x)}{d(z)}$$

Chain rule

$$loss = error$$
$$error = weight$$

$$\frac{d(loss)}{d(weight)} = \frac{d(loss)}{d(error)} * \frac{d(error)}{d(weight)}$$

Calculating the gradient



$$\frac{d(loss)}{d(weight)} = \frac{d(loss)}{d(error)} * \frac{d(error)}{d(weight)}$$

$2error$

$$error = x * weight - y$$

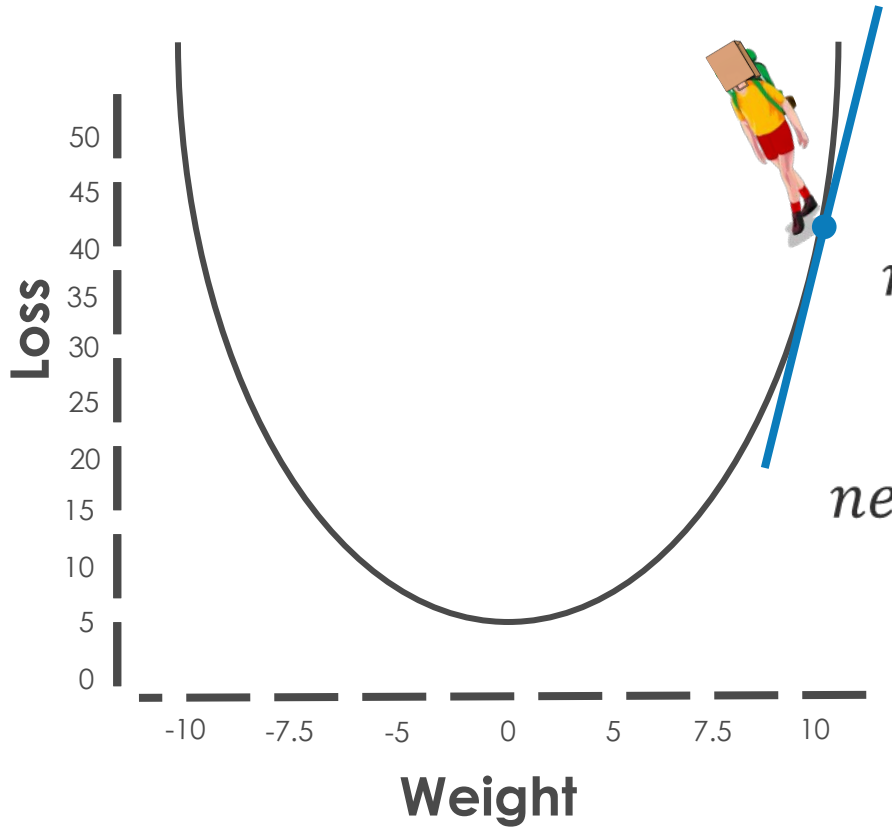
$$x * weight^1$$

$$1 * x * weight^0$$

$$x$$

$$\frac{d(loss)}{d(weight)} = 2error * x$$

Updating the weight



$$\frac{d(loss)}{d(weight)} = 2error * x$$

$$new\ weight = weight - \frac{d(loss)}{d(weight)} * learning\ rate$$

$$new\ weight = weight - 2error * x * learning\ rate$$