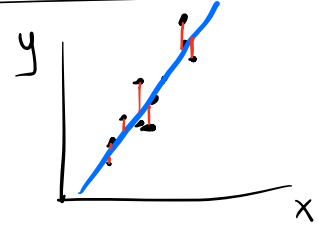


# ML1 - Optimization with Gradient Descent

Linear regression

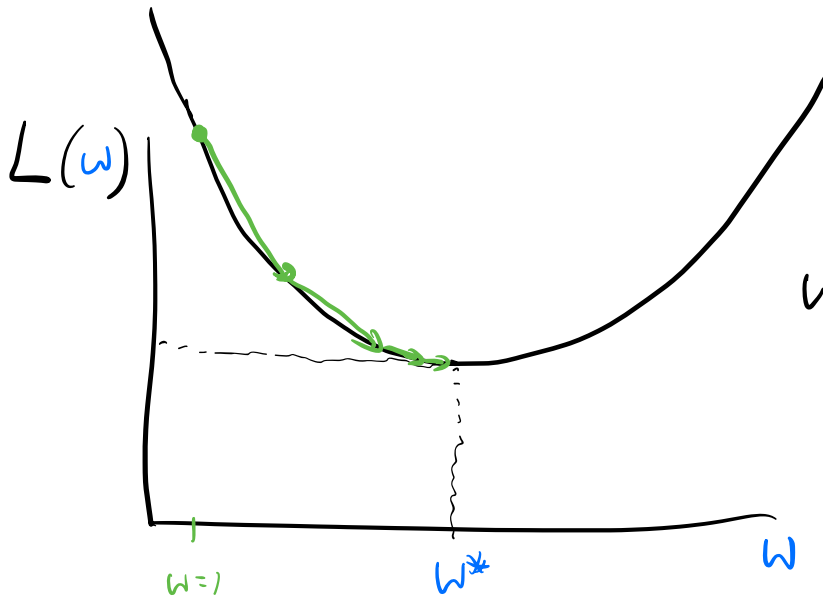
Data:  $\{(x_i, y_i) : 1 \leq i \leq n\}$



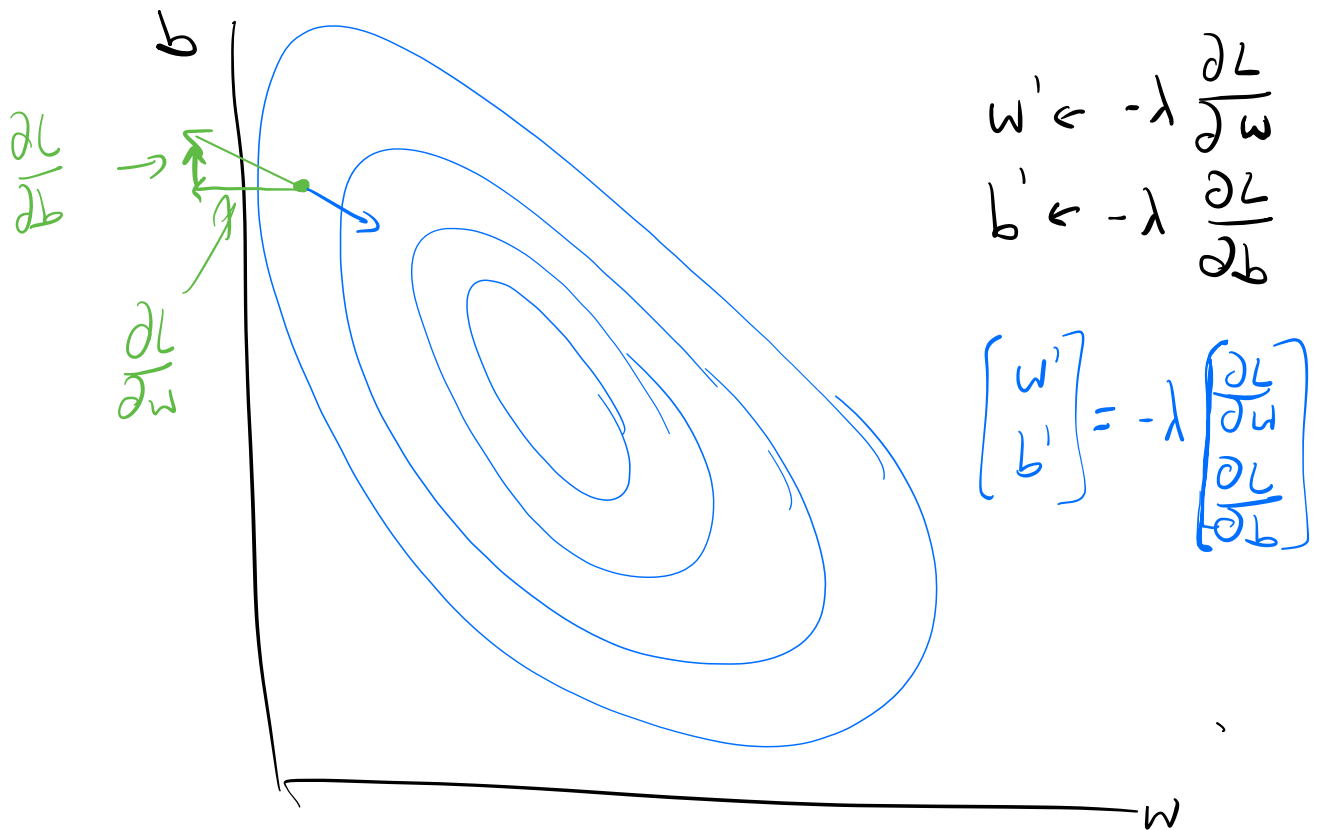
Model:  $\hat{y} = \boxed{w}x + \boxed{b}$  (univariate linear regression)

Loss:  $\frac{1}{2n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$  (mean squared error)

$$L(w, b) = \frac{1}{2n} \sum (y_i - (w \underline{x}_i + b))^2$$



$$w' \leftarrow -\lambda \frac{d}{dw} L(w)$$



$$\frac{\partial}{\partial \hat{y}_i} (y_i - \hat{y}_i)^2 = y_i - \underline{2\hat{y}_i} + \underline{\hat{y}_i^2}$$

$$2\hat{y}_i - 2y_i$$

$$2(\hat{y}_i - y_i)$$

$$\frac{\partial}{\partial w} (wx_i + b) = \boxed{x_i}$$

$$\frac{1}{n} \sum x_i (\hat{y}_i - y_i)$$

# Gradient Descent Algorithm:

$w, b \leftarrow w_{\text{init}}, b_{\text{init}}$

while True:

$\hat{y} \leftarrow \text{model}(x, w, b)$

$\text{loss} \leftarrow \text{loss}(y, \hat{y})$

$dw \leftarrow \text{grad\_wrt\_w}(\hat{y}, x, y)$

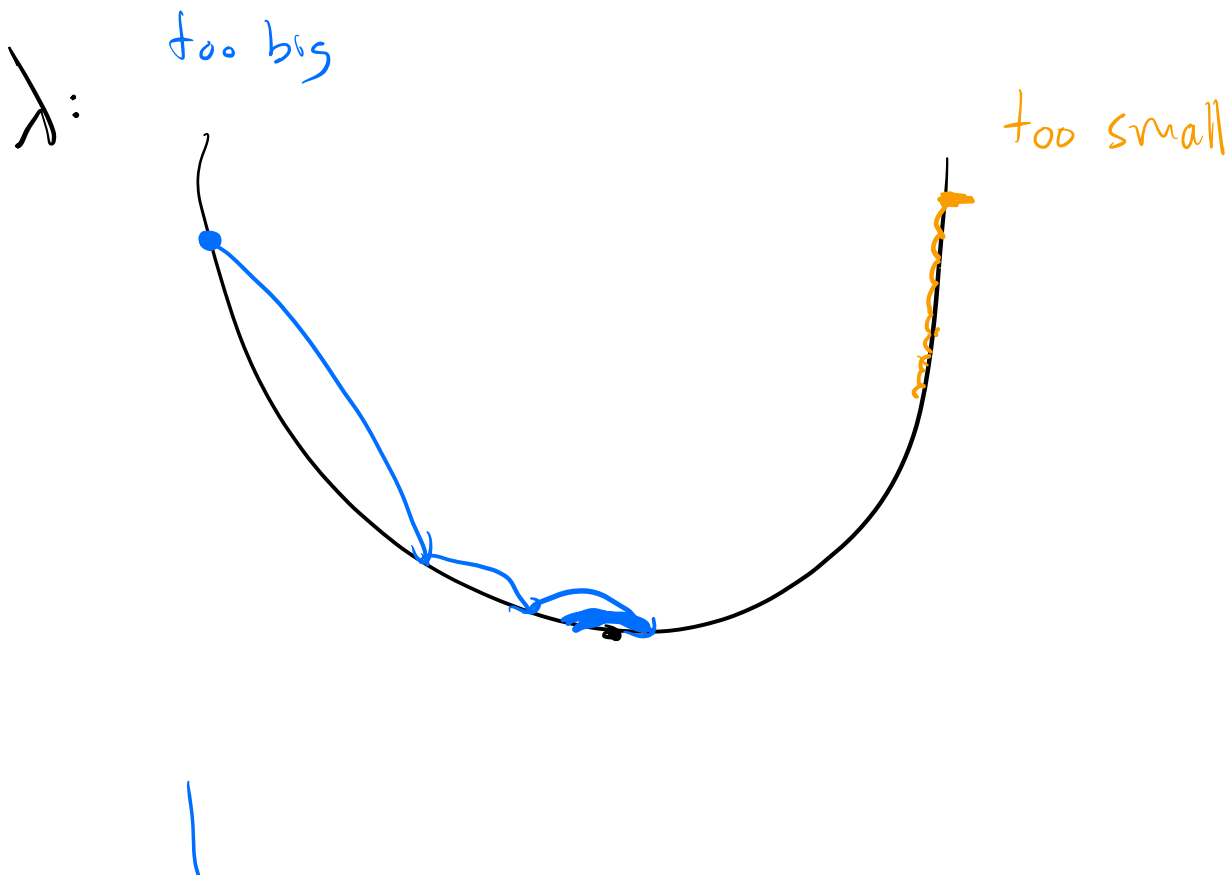
$db \leftarrow \text{grad\_wrt\_b}(\hat{y}, x, y)$

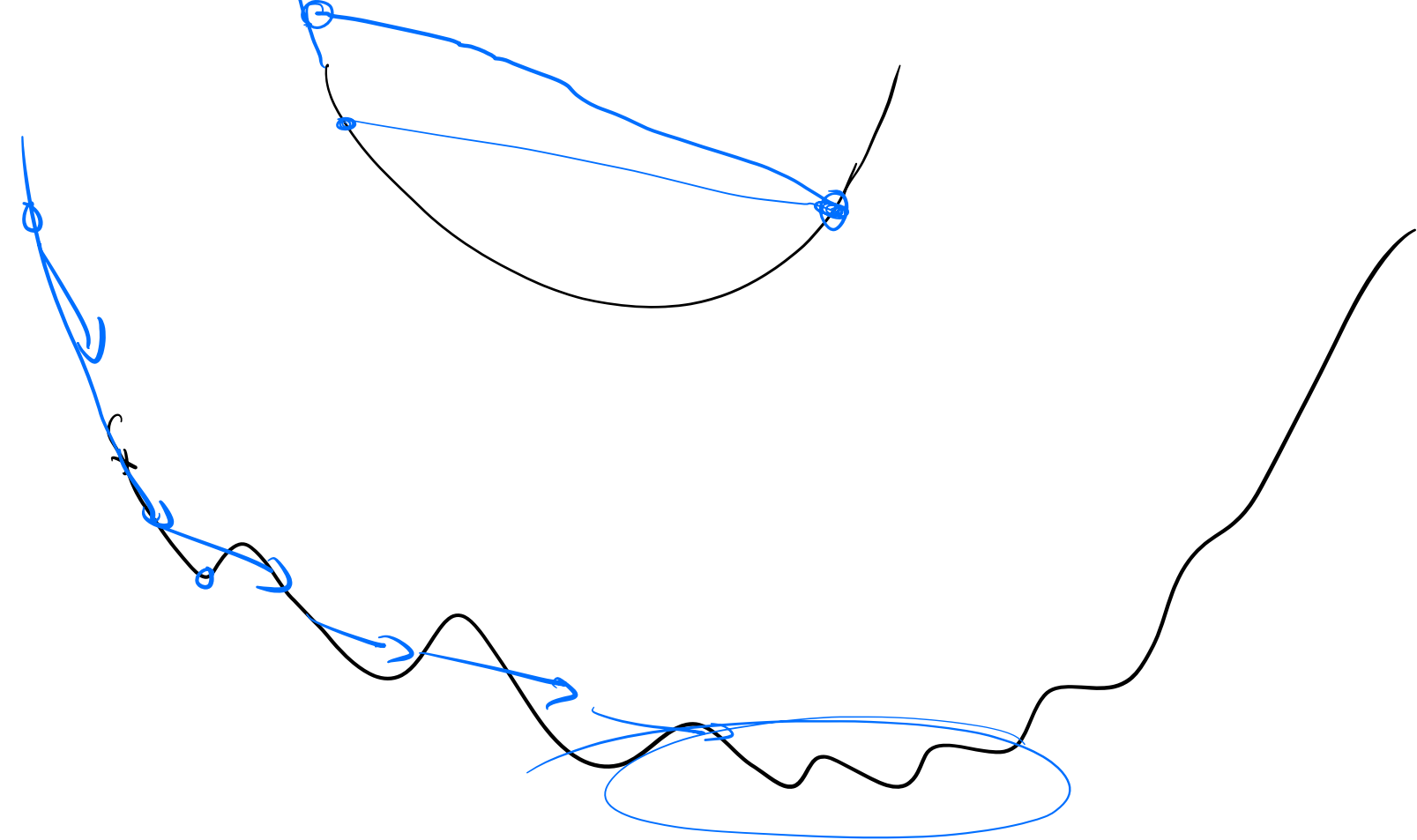
if  $\text{norm}\begin{pmatrix} dw \\ db \end{pmatrix} < \epsilon$ :

return  $w, b$

$w \leftarrow -\lambda \cdot dw$

$b \leftarrow -\lambda \cdot db$

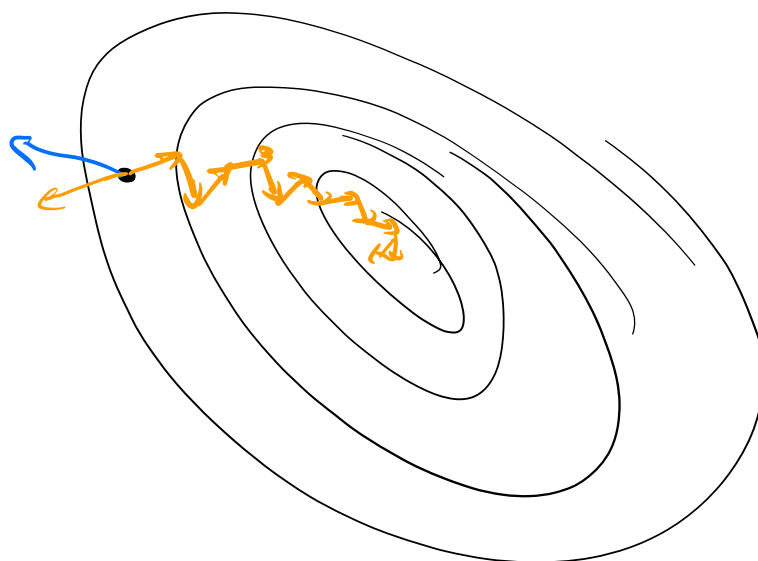




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Stochastic gradient descent

1 random pt per step



Minibatch GD

Choose batch of pts