The background features a dark gray base with several wavy, organic shapes in shades of brown and gold. A prominent grid pattern is visible in the upper left corner, transitioning into the wavy lines. The overall aesthetic is modern and technical.

# Machine Learning Specialization

# The Cost Function + Gradient Descent

$f_w(x)$  = function of  $x$ , fixed  $w$

$J(w)$  = function of  $w$

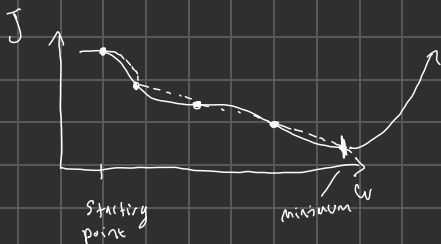
$$J(w) = \frac{1}{2M} \sum_{i=1}^M (f_w(x^{(i)}) - y^{(i)})^2$$

$$= \frac{1}{2M} \sum_{i=1}^M (w x^{(i)} - y^{(i)})^2 = \frac{1}{2M} (0^2 + 0^2 + 0^2)$$

min  $J(w, b)$  Minimize the cost fn

min  $J(w_1, w_2, \dots, w_n, b)$   
 $w_1, w_2, \dots, w_n, b$

- Start with some  $w, b$ , then iteratively change values to minimize the cost fn



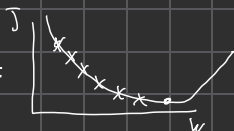
Main update formula:  $w = w - \alpha \frac{d}{dw} J(w, b)$

$x, y \in \mathbb{R}$

learning rate  $\alpha$

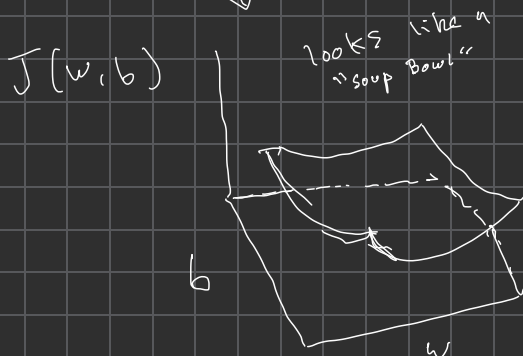
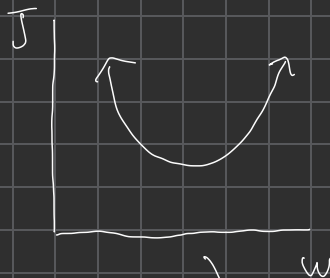
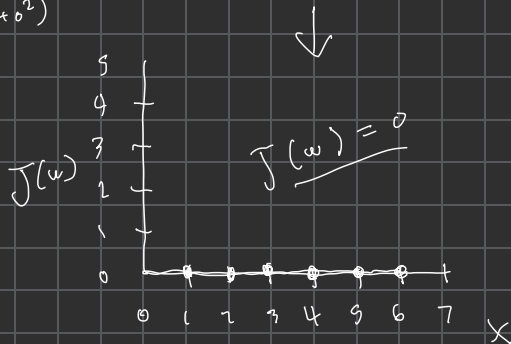
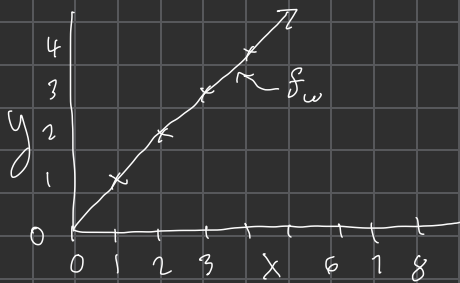
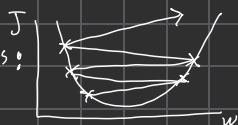
$$b = b - \alpha \frac{\partial}{\partial b} J(w, b)$$

$\alpha$  is small: then lots of small updates



fail to converge and even diverge

$\alpha$  is big, lots of big updates



# Multiple Lin Regression

## House Prediction

- size (sq ft)  $x_j = j^{\text{th}}$  feature
- bedrooms  $n = \text{num features}$
- floors
- Age (years)
- price (\$)  $\vec{x}^{(i)} = i^{\text{th}}$  training example
- $x_j^{(i)} = j^{\text{th}}$  feature of  $i^{\text{th}}$  sample

$$f_{\vec{w}, b}(x) = w_1 x_1 + w_2 x_2 + \dots + w_n x_n + b$$

$$f_{\vec{w}, b}(\vec{x}) = \vec{w} \cdot \vec{x} + b = b + \sum_{j=0}^n w_j x_j$$

## Previous Notation

params:  $w_1, \dots, w_n$

Model:  $f_{\vec{w}, b}(\vec{x}) = w_1 x_1 + \dots + w_n x_n + b$

cost fn:  $J(w_1, w_2, \dots, w_n)$

## Vector Notation

$$\vec{w} = [w_1, w_2, \dots, w_n], b$$

$$f_{\vec{w}, b}(\vec{x}) = \vec{w} \cdot \vec{x} + b$$

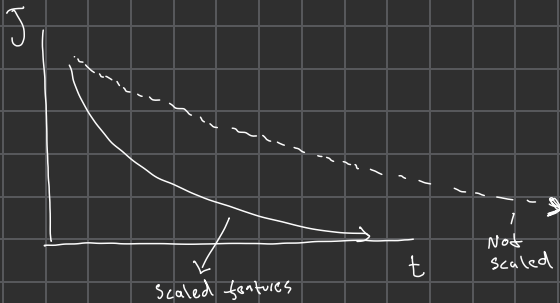
$$J(\vec{w}, b)$$

## Gradient Descent

$$w_j = w_j - d \frac{\partial}{\partial w_j} (J(\vec{w}, b))$$

$$b = b - d \frac{\partial}{\partial b} J(\vec{w}, b)$$

$$w_n = w_n - d \frac{1}{n} \sum_{i=1}^n (f_{\vec{w}, b}(\vec{x}^{(i)}) - y^{(i)}) x_n^{(i)}$$



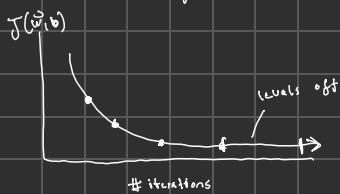
Mean Normalization:

$$x_1 = \frac{x_1 - \mu_1}{\text{max} - \text{min}}$$

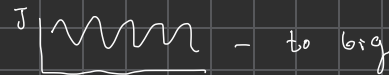
Z-score Normalization:

$$x_1 = \frac{x_1 - \mu_1}{\sigma_1}$$

• make sure gradient descent is working

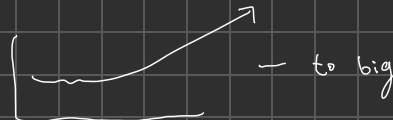


• How to choose a good learning rate?



$$w_1 = w_1 + \alpha d_1$$

$$w_1 = w_1 - \alpha d_1$$



• try a range of values and see what produces a good learning graph

# Polynomial Regression

