

Session
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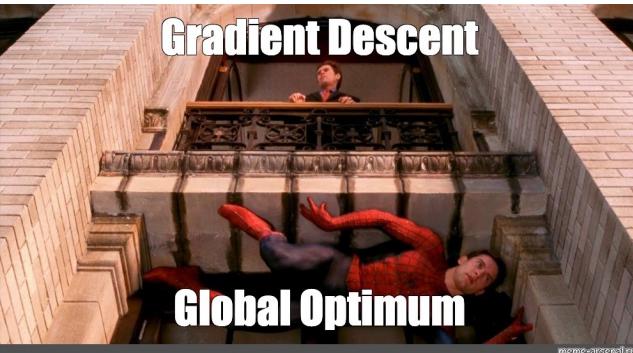
OPTIMIZATION - 4

CONSTRAINED
OPTIMIZATION

Feb 15,
2024

Gradient Descent

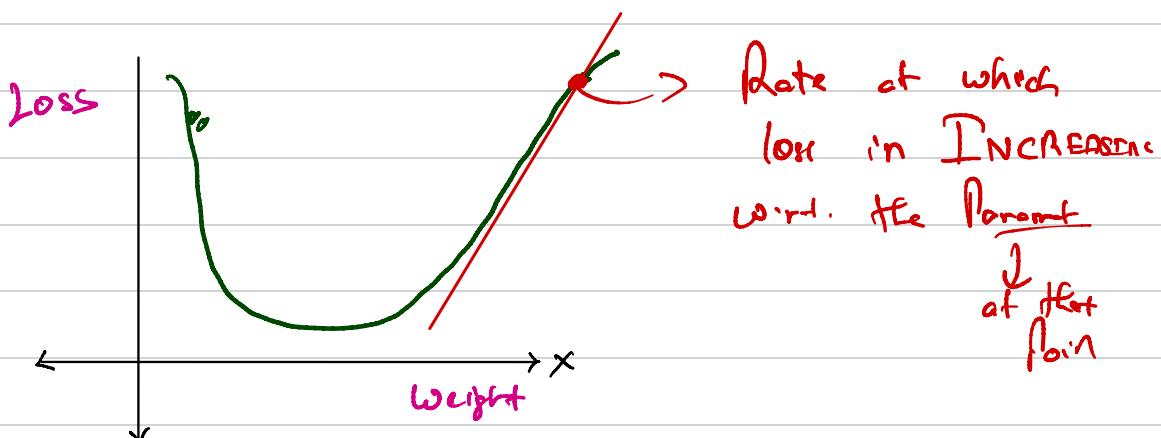
Global Optimum



AGENDA

- ① Optimization Problem
- ② How To Solve CONSTRAINED OPTIMIZATION PROBLEM

GRADIENT DESCENT



$$w_{\text{new}} = w_0 - \eta \frac{\partial L}{\partial w_0}$$

Gradient descent

multiple diff ways of how to calculate gradient

$$\text{Ex. Q. } f(n_1, n_2, n_3, n_4) = a_1 n_1 + a_2 n_2 + a_3 n_3$$

$$\vec{a}^T \cdot \vec{n}$$

$$\frac{\partial f}{\partial n} = \vec{a}^T$$

$$\frac{\partial f}{\partial x_1} = a_1 \quad / \quad \frac{\partial f}{\partial x_2} = a_2 \dots$$

$$\frac{\partial f}{\partial \vec{x}} = \vec{a}$$

$$E_{n-2} \ L \rightarrow f(x_1, x_2 \dots x_n) = \vec{a}^T \cdot \vec{x}$$

$\frac{\partial f}{\partial x} = \vec{a}$

∇f

OPTIMIZATION PROBLEM

$$\min_{\vec{w}, w_0} - \sum y_i \left(\frac{w^T \cdot x_i + w_0}{\|w\|} \right)$$

$$f(\omega_0, \omega_1, \dots, \omega_n)$$

$$= f(\omega_1 + \Delta\omega_1, \omega_2 + \Delta\omega_2, \dots, \omega_n + \Delta\omega_n)$$

Δ

$$= f(\omega_1, \omega_2, \dots)$$

$$\underline{\Delta \approx 0}$$

Derivatives do this

$$\frac{\partial f}{\partial \omega_1} = \frac{- \sum y_i (\omega_1 n_1 + \omega_2 n_2 + \dots + \omega_n)}{\sqrt{\omega_1^2 + \omega_2^2 + \dots + \omega_n^2}}$$

$$\frac{f(\omega_1)}{g(\omega_1)} \Rightarrow \text{Computationally } \textcircled{J. \text{ expensive}}$$

$$\omega_1 x_1 + \omega_2 x_2 - \omega_0 = 0$$

↓ ↓ ↓
 3 4 4

⇒

$$\frac{3x + 4y + 4}{\|\omega\|} = 0$$

$$\omega_1 = 3/5$$

$$\omega_2 = 4/5$$

$$\omega_0 = 1/5$$

$$\|\omega\| = \sqrt{\left(\frac{3}{5}\right)^2 + \left(\frac{4}{5}\right)^2}$$

$$= 1$$

$$\frac{3}{5}x + \frac{4}{5}y + \frac{4}{5} = 0$$

$$\min_{\vec{\omega}, \omega_0} \left\{ -\sum y_i (\omega^T x_i + \omega_0) \right\} \quad \begin{array}{l} f(\vec{\omega}, \omega_0) \\ \text{as } \|\omega\|_1 = 1 \end{array}$$



$$\frac{\partial f}{\partial w}$$

$$\frac{\partial F}{\partial w_j} \Rightarrow - \sum_{i=1}^n y_i (w_1 n_1 + w_2 n_2 - \cancel{w_0 n_0})$$

$$\frac{\partial F}{\partial w_1} \Rightarrow - \sum_{i=1}^n y_i n_1$$

$$w_{j\text{new}} = w_{j\text{old}} - \eta \left(- \sum_{i=1}^n y_i \cdot y_j \right)$$

$\sum_{j=1}^n$

$$w_1 = w_1 - \eta \left(- \sum_{i=1}^n y_i \cdot n_1 \right)$$

Gradient at 1st iteration



my weight change

and norm $\neq 1$

We can't do G.D. (Gradient descent)

Unconstrained loss function is

$$\min_{\vec{w}, w_0} - \sum y_i \left(\frac{\vec{w}^T \cdot \vec{x}_i + w_0}{\|\vec{w}\|} \right)$$

Constrained loss function

$$\min_{\vec{w}, w_0} - \sum y_i \left(\frac{\vec{w}^T \cdot \vec{x}_i + w_0}{\|\vec{w}\|} \right) \quad f(\vec{w})$$

$\underbrace{\text{S.T.}}_{\text{such that}} \quad \underbrace{\|\vec{w}\| = 1}_{\|\vec{w}\| - 1 = 0} \quad g(\vec{w})$

LAGRANGES MULTIPLIERS

→ Converts constrained eqn into unconstrained eqn

$$\rightarrow f(\vec{x}) + \lambda g(\vec{x}) \quad \left. \begin{array}{l} \text{Symbol for} \\ \text{Lagrange's multipl.} \end{array} \right\} \text{Unconstrained eqn}$$

Constrained optimization: $\min_{\theta} f(\theta) ; \text{ s.t. } g_1(\theta) = 0$
 $g_2(\theta) = 0$
 $g_3(\theta) = 0$

L-multiplication

new loss function

$\min_{\theta} f(\theta) + \lambda_1 \times g_1(\theta) + \lambda_2 \times g_2(\theta) + \lambda_3 \times g_3(\theta)$

L-multiplication

Ex.

$\min_{x,y} f(x,y) = x^2 + y^2$

s.t. $x + 2y - 1 = 0$

Perform partial derivatives w.r.t. x, y, λ

new car = $x^2 + y^2 + \lambda (x + 2y - 1)$

$$\frac{\partial f(x, y, z)}{\partial x} = 2x + \lambda = 0$$

$$\Rightarrow x = -\underline{2z}$$

$$\frac{\partial f(x, y, z)}{\partial y} = 2y + 2\lambda = 0$$

$$\Rightarrow y = \underline{-\lambda}$$

$$\frac{\partial f(x, y, z)}{\partial z} \Rightarrow \underline{x + 2y - 1} = 0$$

$$x = \frac{1}{5}, y = \frac{2}{5}, \lambda = \underline{-\frac{2}{5}}$$

$$\begin{array}{l} x = -2z \\ \lambda = -y \\ x + 2y - 1 = 0 \end{array} \quad \left/ \right. \quad \begin{array}{l} \lambda y = f_{xz} \\ y = \underline{\frac{2z}{5}} \end{array}$$

$$x + 2y - 1 = 0$$

$$x + 2(2z) - 1 = 0$$

$$x + 4z - 1 = 0$$

$$x = \underline{\frac{1}{5}}$$

INTUITION

→ In general you get constraint \rightarrow can
→ can't
think
yourself

→ Let's say you're trying \rightarrow lowest point
of hill-slope

Can't explore anywhere

Some constraint (stay on specific path)
Forces you to find lowest pt
meanwhile also staying on path

$$\omega_r = \omega_r - \frac{\partial L}{\partial \omega_i} \times \eta$$

$$\lambda = \lambda - \frac{\partial L}{\partial \lambda} \times \eta$$

My orig can become.

$$L = \min_{\vec{\omega}, \omega_0, \lambda} - \sum_{i=1}^n (\vec{\omega}^T \cdot \mathbf{x}_i + \omega_0) y_i + \lambda (||\omega||_1)$$

Θ . Find minima S.T.

$$f(x) = x^2 - 3x - 3$$

S.T. constraint = $-x^2 + 2x + 3 = 0$

new can

$$h(x) = x^2 - 3x - 2 + \lambda (-x^2 + 2x + 3) = 0$$

$$\frac{\partial h}{\partial x} \Rightarrow 2x - 3 - 2\lambda x + 2\lambda = 0$$

$$\frac{\partial h}{\partial \lambda} \Rightarrow -x^2 + 2x + 3 = 0$$

$$\text{Open } -3 - 2\lambda \geq 0 + 2\lambda$$

$$\begin{aligned}
 & -\lambda^2 + 2\lambda + 3 \\
 \Rightarrow & -\lambda^2 + 3\lambda - \lambda + 3 \\
 = & (\lambda+1)(-\lambda+3) \\
 \Rightarrow & \lambda = \underline{\underline{-1, 3}}
 \end{aligned}$$

for $\lambda = 3$

$$2x3(1-3) + 2\lambda - 3 = 0$$

$$\begin{aligned}
 \lambda = \frac{3}{4} & \quad \left(6 - 6\lambda + 2\lambda - 3 = 0 \right. \\
 & \quad \left. 3 - 4\lambda = 0 \right. \\
 & \quad \underline{\underline{\frac{3}{4} = \lambda}}
 \end{aligned}$$

for $\lambda = -1$

$$\Rightarrow 2\lambda(1-\lambda) + 2\lambda - 3 = 0$$

$$\begin{aligned}
 \lambda = \frac{5}{4} & \quad \left(2(-1)(1-\lambda) + 2\lambda - 3 = 0 \right. \\
 & \quad \left. -2 + 2\lambda + 2\lambda - 7 = 0 \right. \\
 & \quad \underline{\underline{\frac{5}{4} = \lambda}}
 \end{aligned}$$

$$\lambda = 3, \frac{3}{4}$$

$$\lambda = -1, \frac{5}{4}$$

$$x^2 - 3x - 2 + \lambda(-x^2 + 2x + 3) = 0$$

(1) $x = 3, \lambda = \frac{3}{4}; \quad \underline{\underline{-3}}$

(2) $x = -1, \lambda = \frac{5}{4}; \quad (1)$

Can

$$L = \min_{\vec{\omega}, \omega_0, \lambda} - \sum_{i=1}^n (\vec{\omega}^T x_i + \omega_0) y_i + \lambda (||\vec{\omega}|| - 1)$$

$$\frac{\partial L}{\partial \vec{\omega}}$$

$$\frac{\partial L}{\partial \omega_0}$$

$$\frac{\partial L}{\partial \lambda}$$

$$a_n x + b_n y + c = f(x)$$