

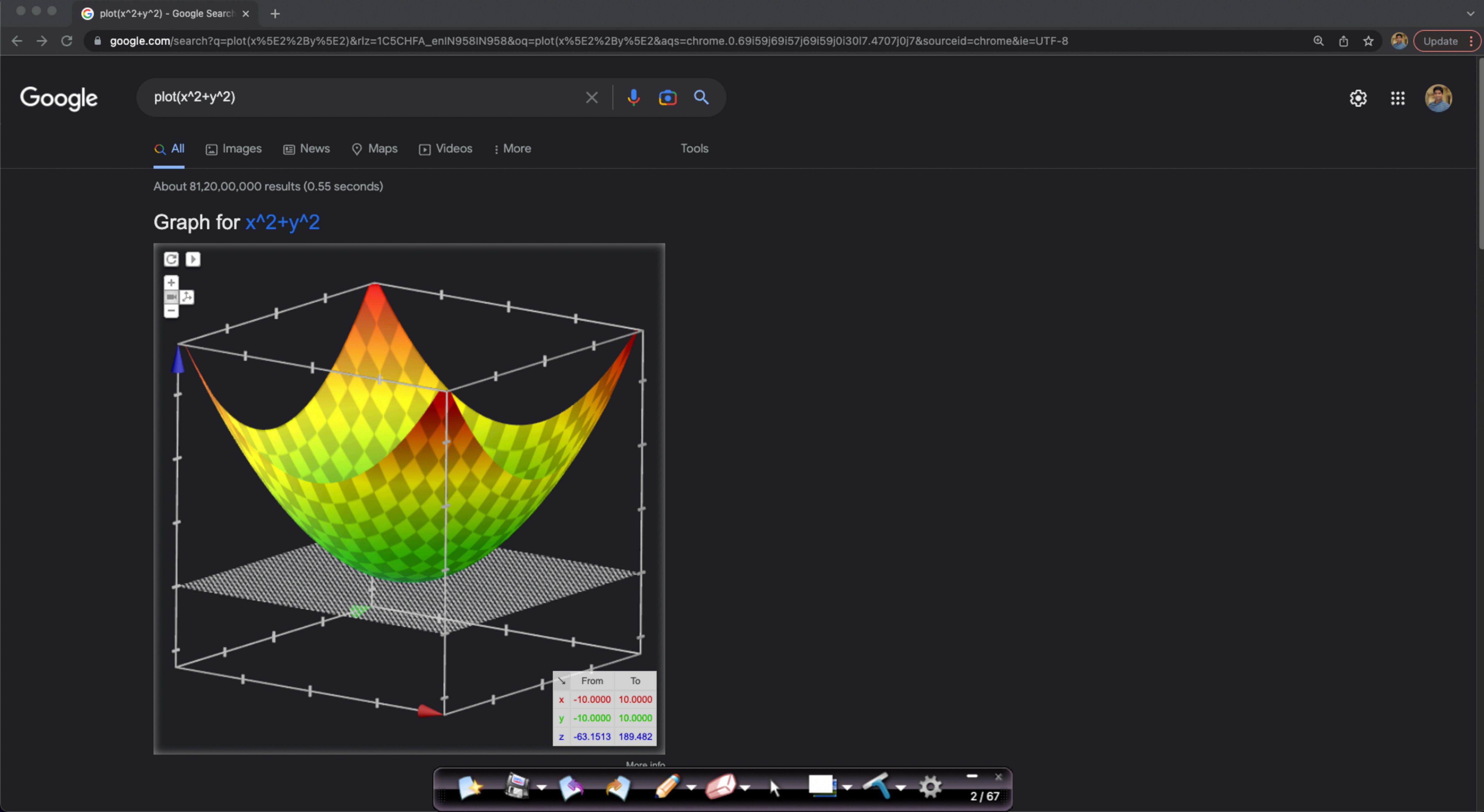
M W F  
MWF

# Constrained Optimization

& SGD

3:15 min

✓ { Request →  $\| \cdot \|_1 : 30\text{PM sharp}$  } ✓  
{ Q&A →  $\| \cdot \|_1 : 30 +$  }



Maxima & Minima

Prev:

-  for  $f(x)$   $\rightarrow$  one-variable

$\eta$ : learning rate

Today {  $\rightarrow$  GD for  $f(\underline{\omega}, \underline{w_0}) \rightarrow$  d+1 variables  $\rightarrow$  Binary classification

Contour  $\nwarrow$  {  $\rightarrow$  Constrained Optimzation  $\rightarrow$  PCA (next class)

$\rightarrow$  SGD & minibatch GD

OPS:  
~~not~~

Topic → Chat (end of topic)

off topic → Ques (end of session)  
Audio

overshooting:

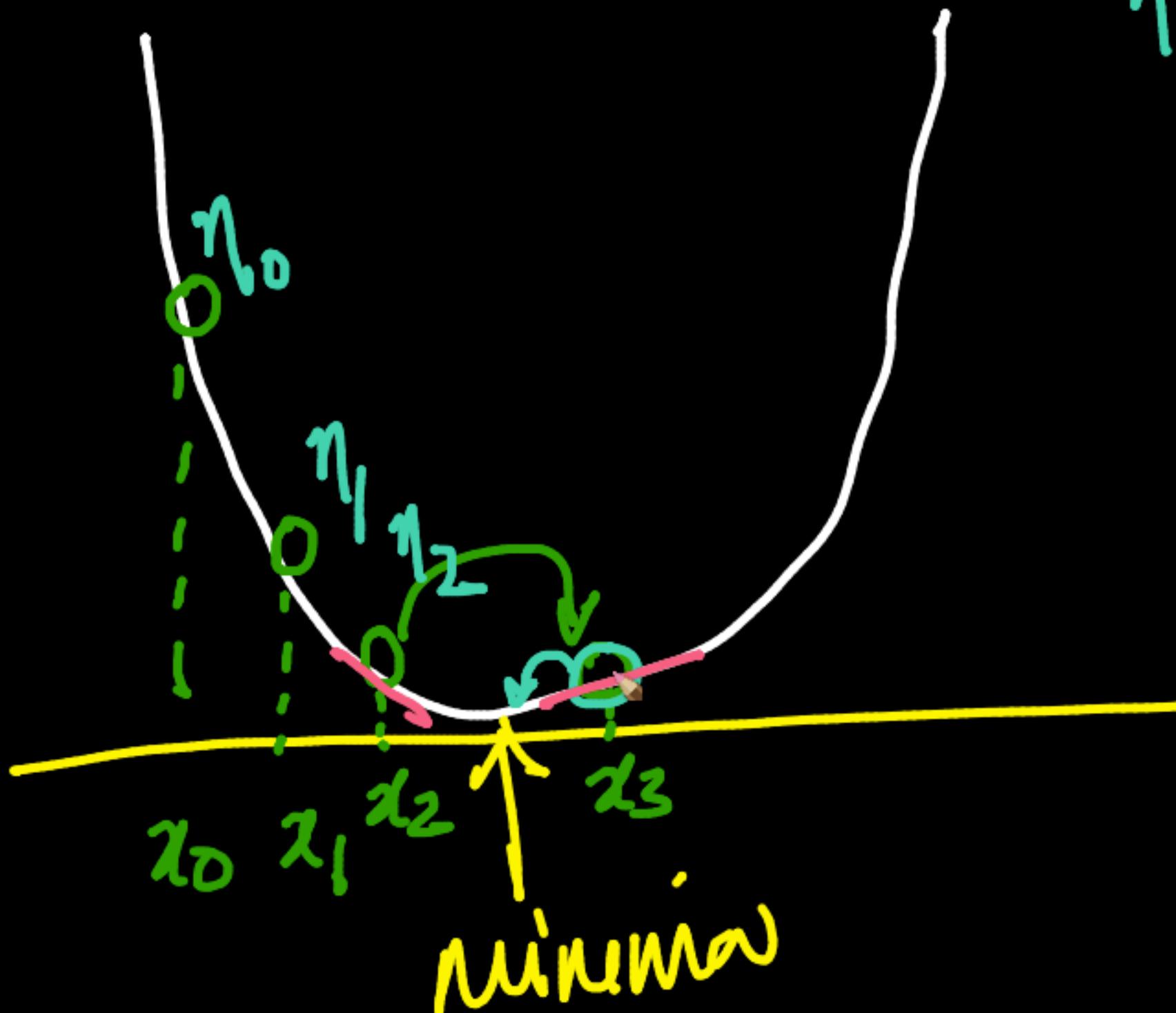
Code Snippet

$$\eta_0 = 0.1 ; \eta_1 = 0.09$$

$\eta_2 = 0.081$   $\rightarrow 0.1$  (const)  
 $\eta$ : learning rate

i: iteration number

$$\left\{ \begin{array}{l} \eta \downarrow \text{as } i \uparrow \\ \eta = 0.1 \\ \eta_{\text{new}} = 0.9 \times \eta_{\text{old}} \end{array} \right.$$



$$x_{i+1} = x_i - \eta \frac{\partial f}{\partial x} \Big|_{x_i}$$

$$\downarrow y = f(i)$$

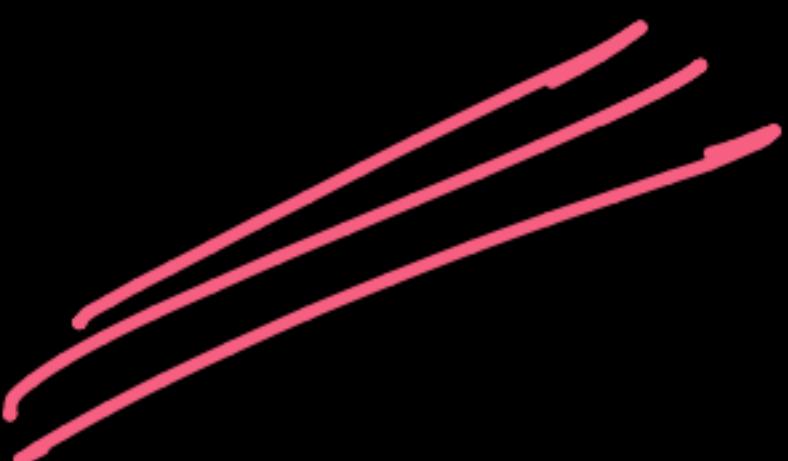
$$\eta = 0.01 \quad ; \quad C = 0.9$$

$\text{while } | df - dx - C | < 0$

$\left\{ \begin{array}{l} \text{update eqn} \\ \eta = \eta \times 0.9 \\ i = i + 1 \end{array} \right.$

$i \uparrow \quad \eta \downarrow$





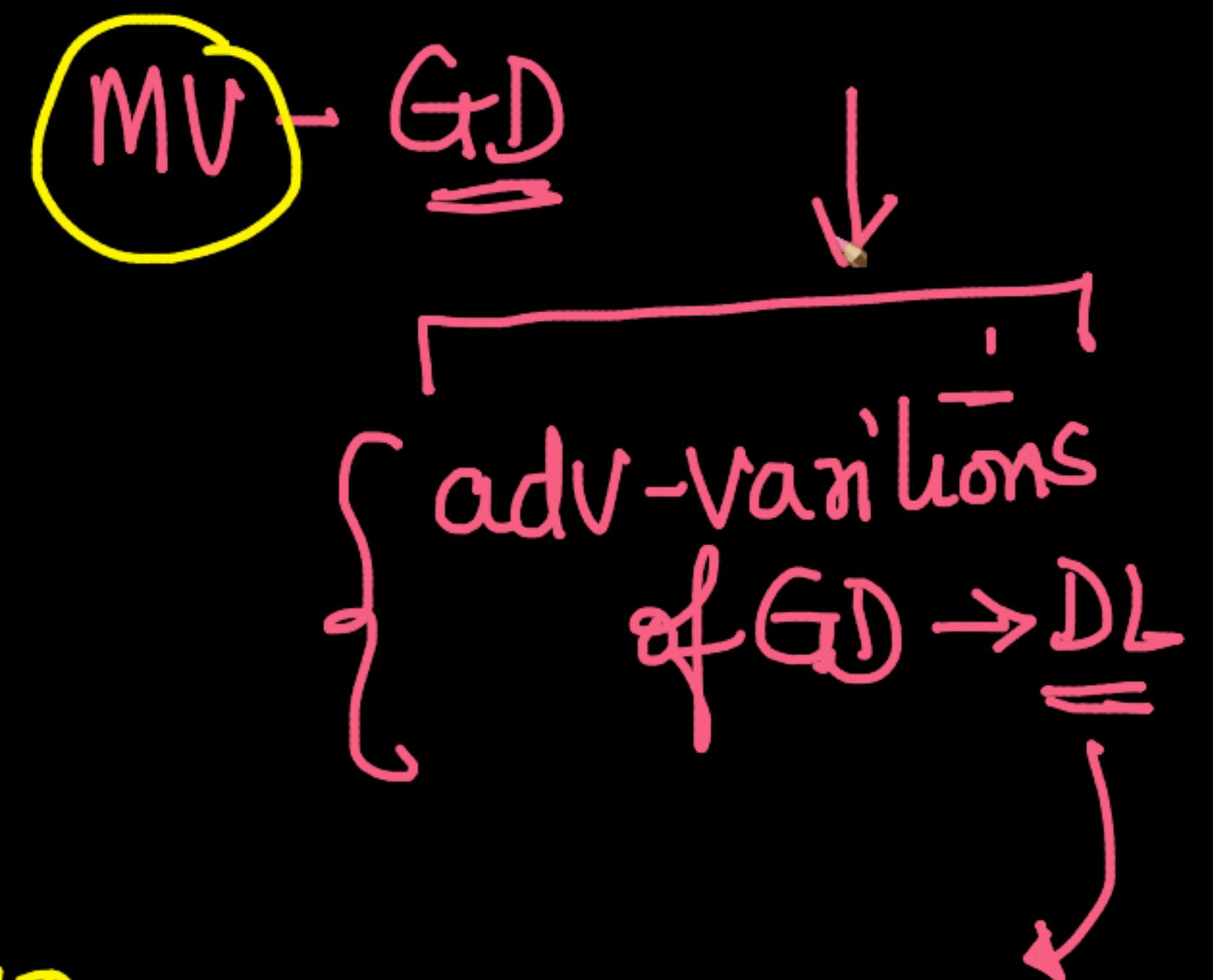
$$z = f(x, y)$$

↓    ↓    2 variables

→ initialize  $x_0$  &  $y_0$  randomly

while  $\frac{\partial f}{\partial x} \approx 0.0$  and  $\frac{\partial f}{\partial y} \approx 0$

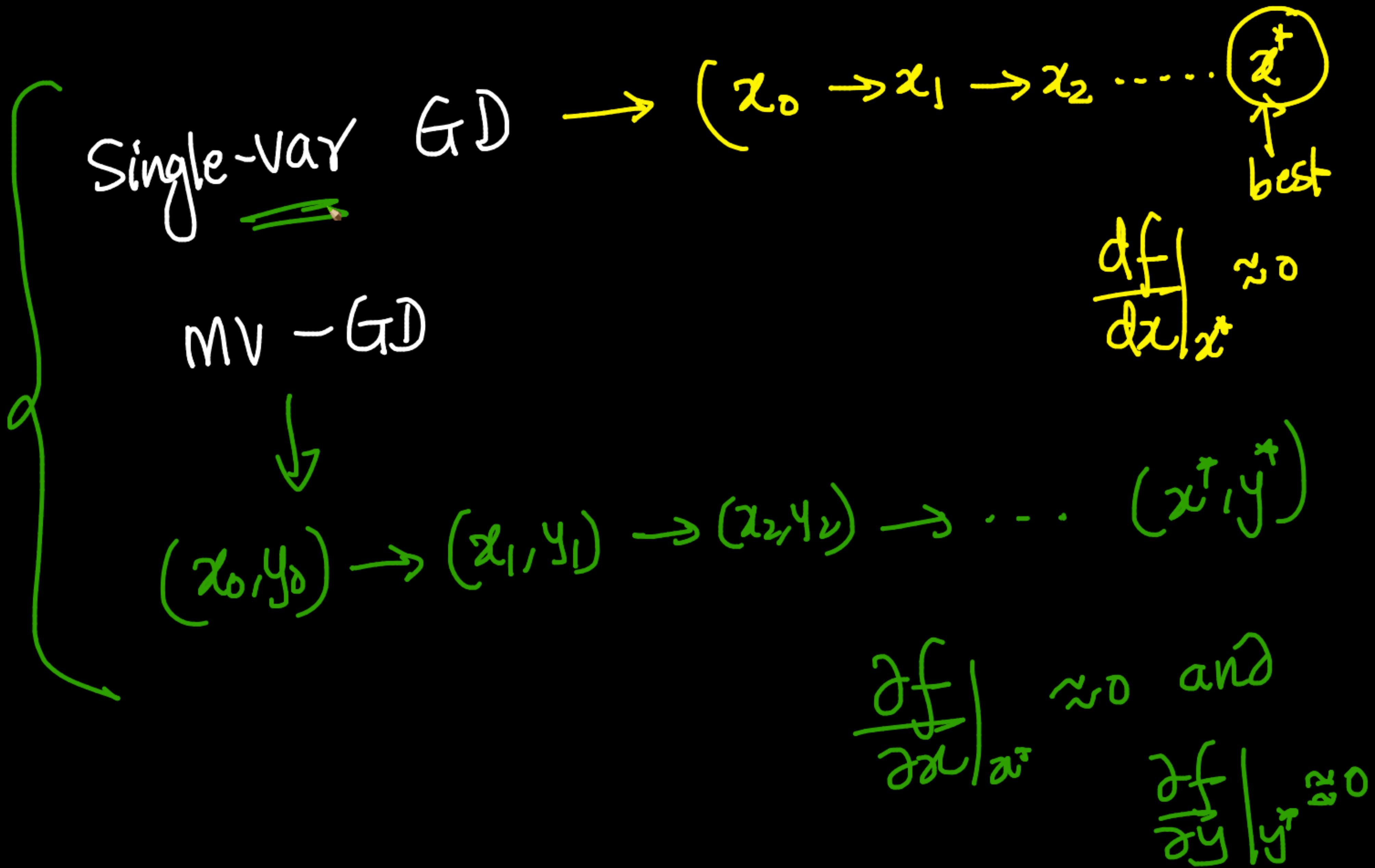
$$\checkmark \quad x_1 = x_0 - \eta \left[ \frac{\partial f}{\partial x} \right]_{x=x_0, y=y_0}$$



$$y_t = y_0 - \eta \left[ \frac{\partial f}{\partial y} \right]_{y=y_0, x=x_0}$$

randomly  
 $(x_0, y_0) \rightarrow (x_1, y_1) \rightarrow (x_2, y_2)$

$$\frac{\partial f}{\partial x} \Big|_{x_0}, \frac{\partial f}{\partial y} \Big|_{y_0}$$



→ **minima** exists (Test)

$$\min f(x, y)$$



MV-GD

$x^*, y^*$

local  
Minima ✓  
Maxima

higher order deriv. less

expensive

$$f(x) \rightarrow$$

$$f(\underline{w}, \underline{w_0}) \rightarrow$$

Comp  
-hull  $\nabla f_{x^*, y^*} = [0]$   
 $f(x^*, y^*) = 10$  (let)

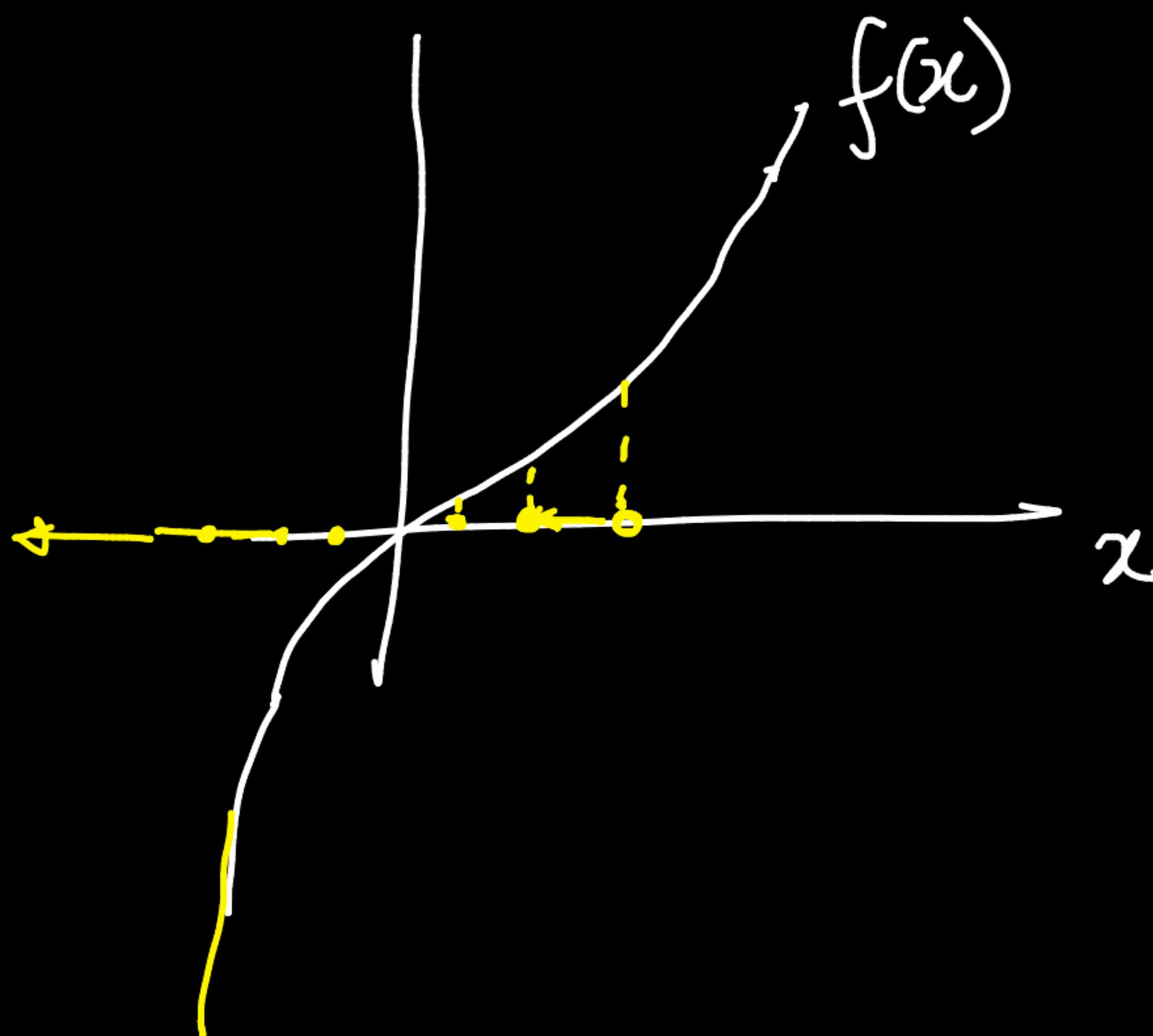
$f(x^* + \delta, y^* + \delta) =$

q  $\Rightarrow$  cannot be minima

l  $\Rightarrow$   $x^*, y^*$  cannot be the  
maximum

GD  $\rightarrow$  ignore saddle points

↓  
minima



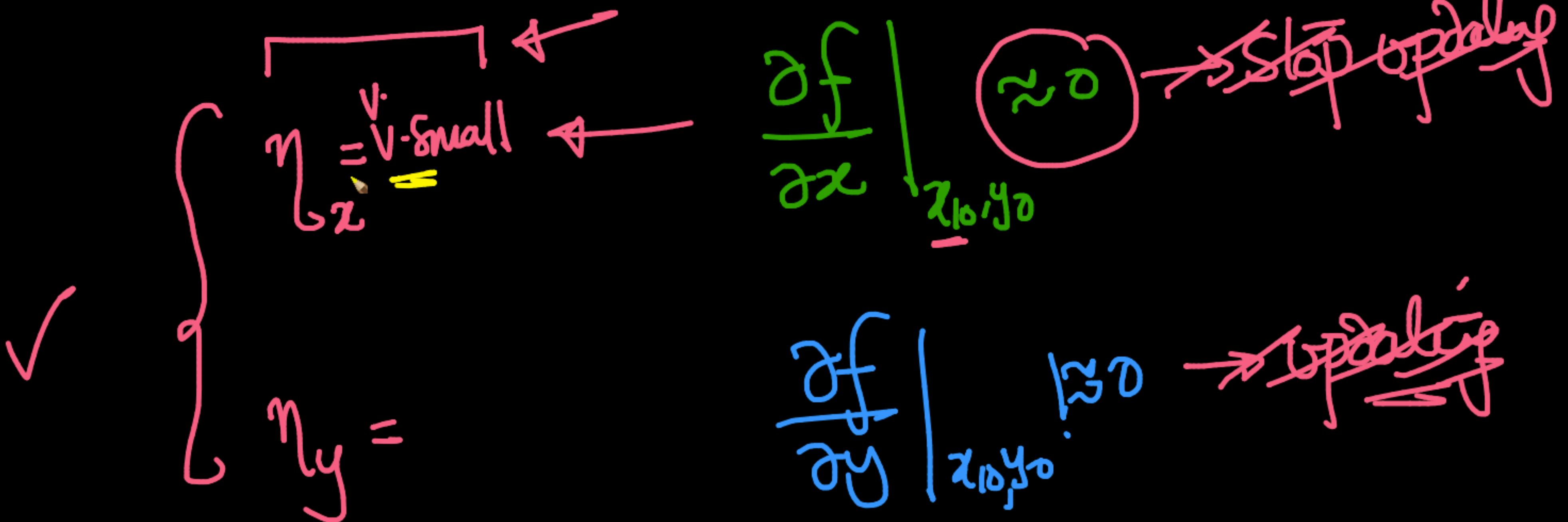
∞ loop

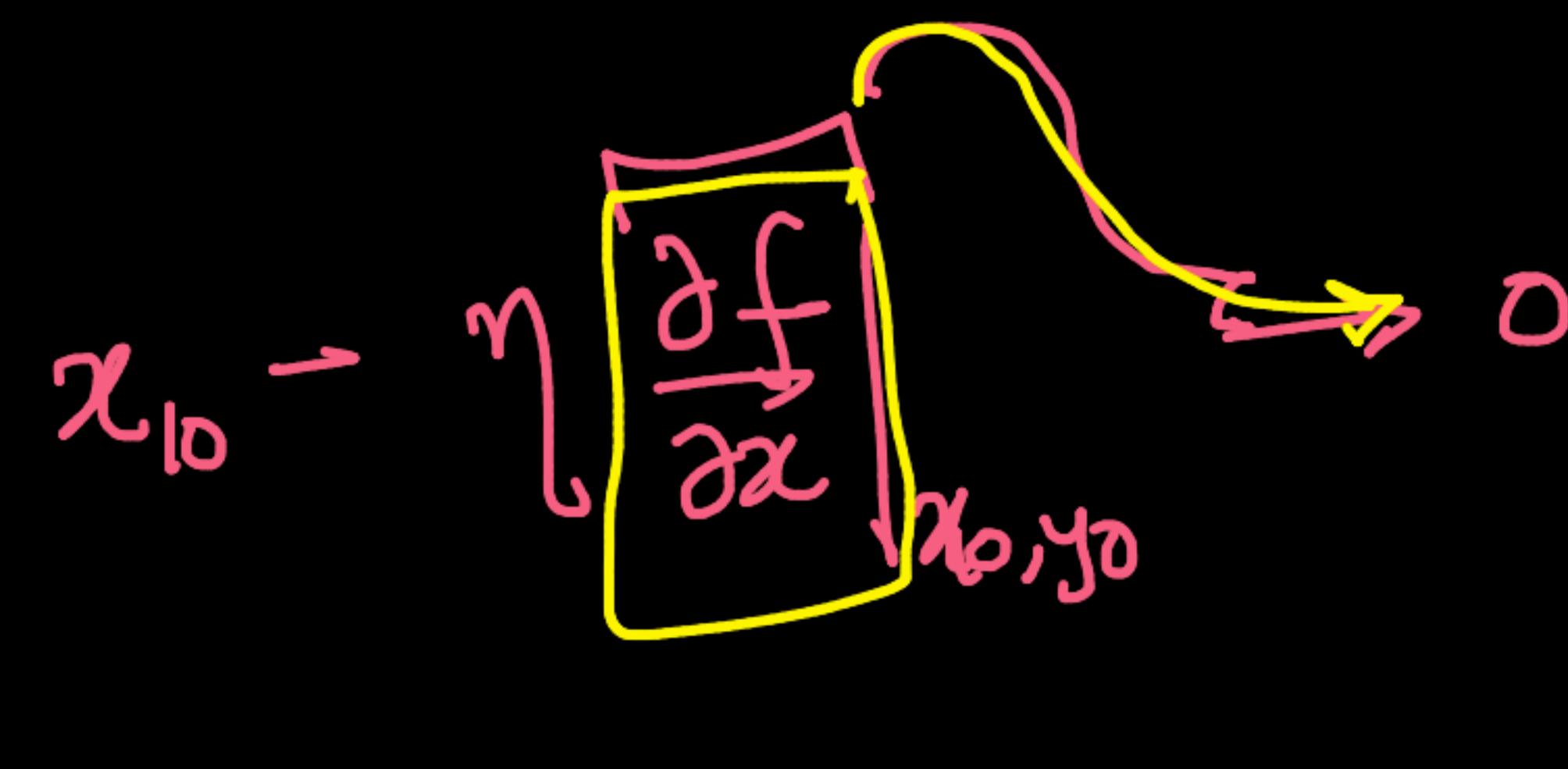
10000-Iterations

$\left. \frac{df}{dx} \right|_{x_{10000}} \approx 0$

$f(x, y)$

$x_0, y_0 \rightarrow x_1, y_1 \rightarrow \dots \rightarrow x_{10}, y_{10}$

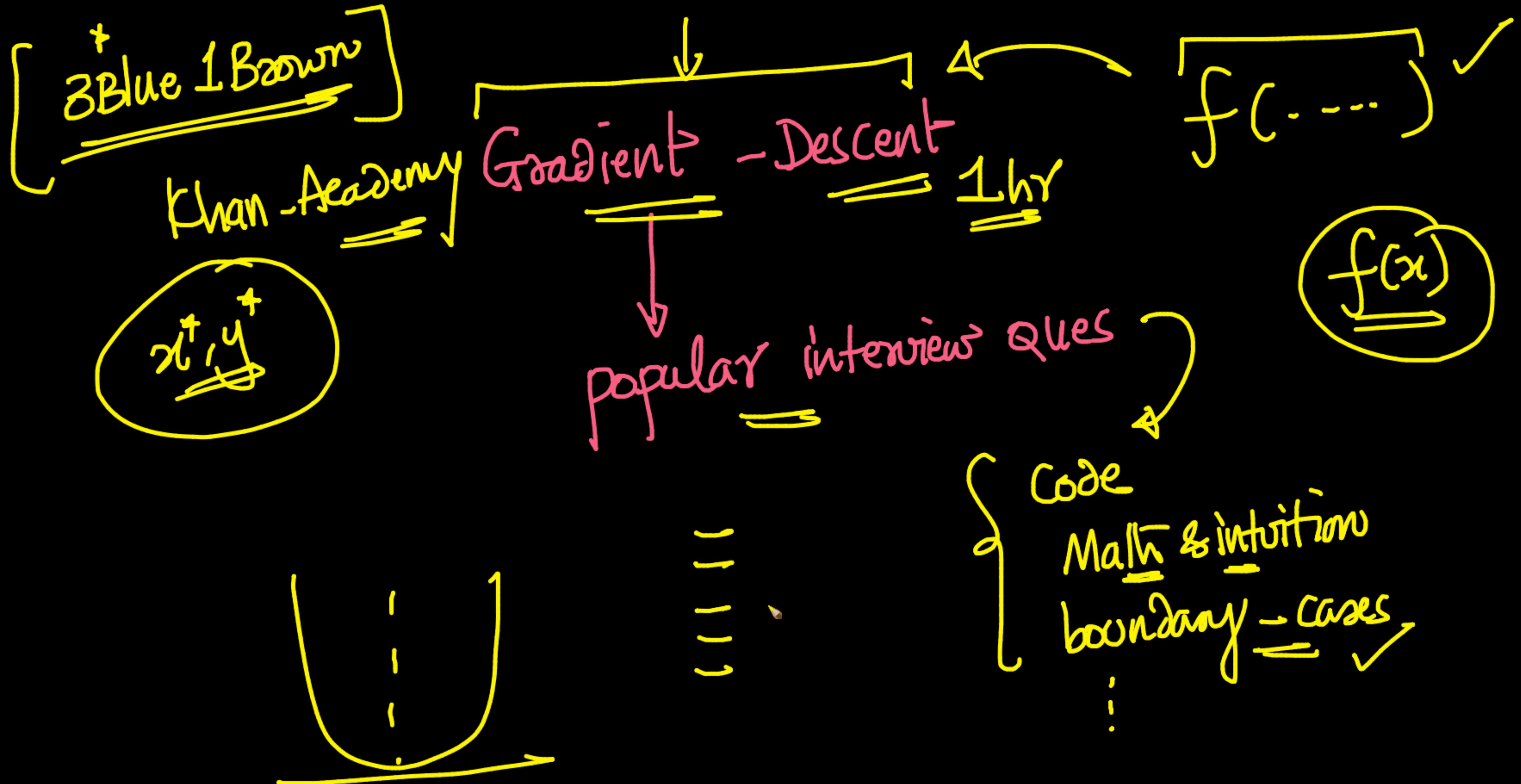


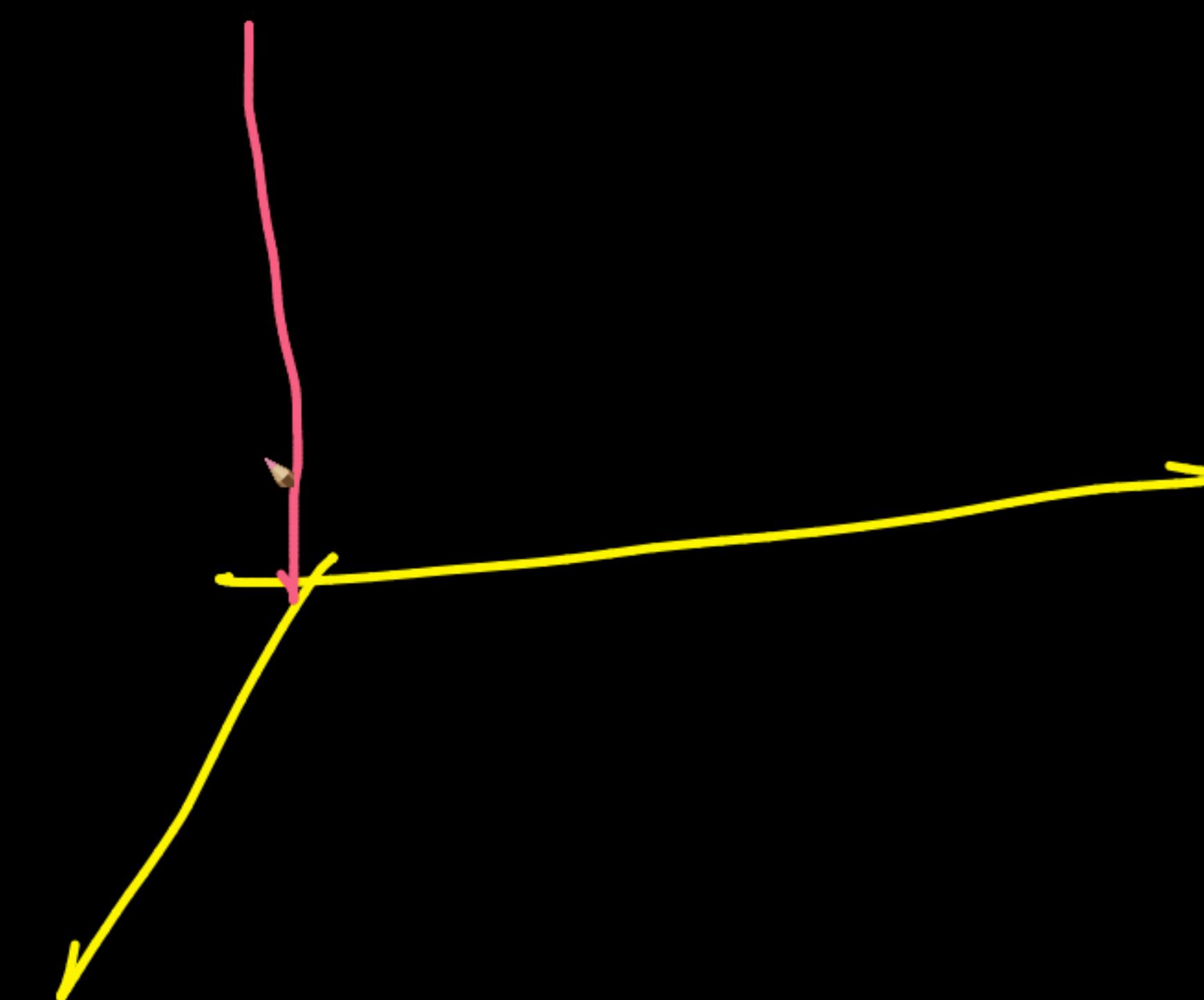


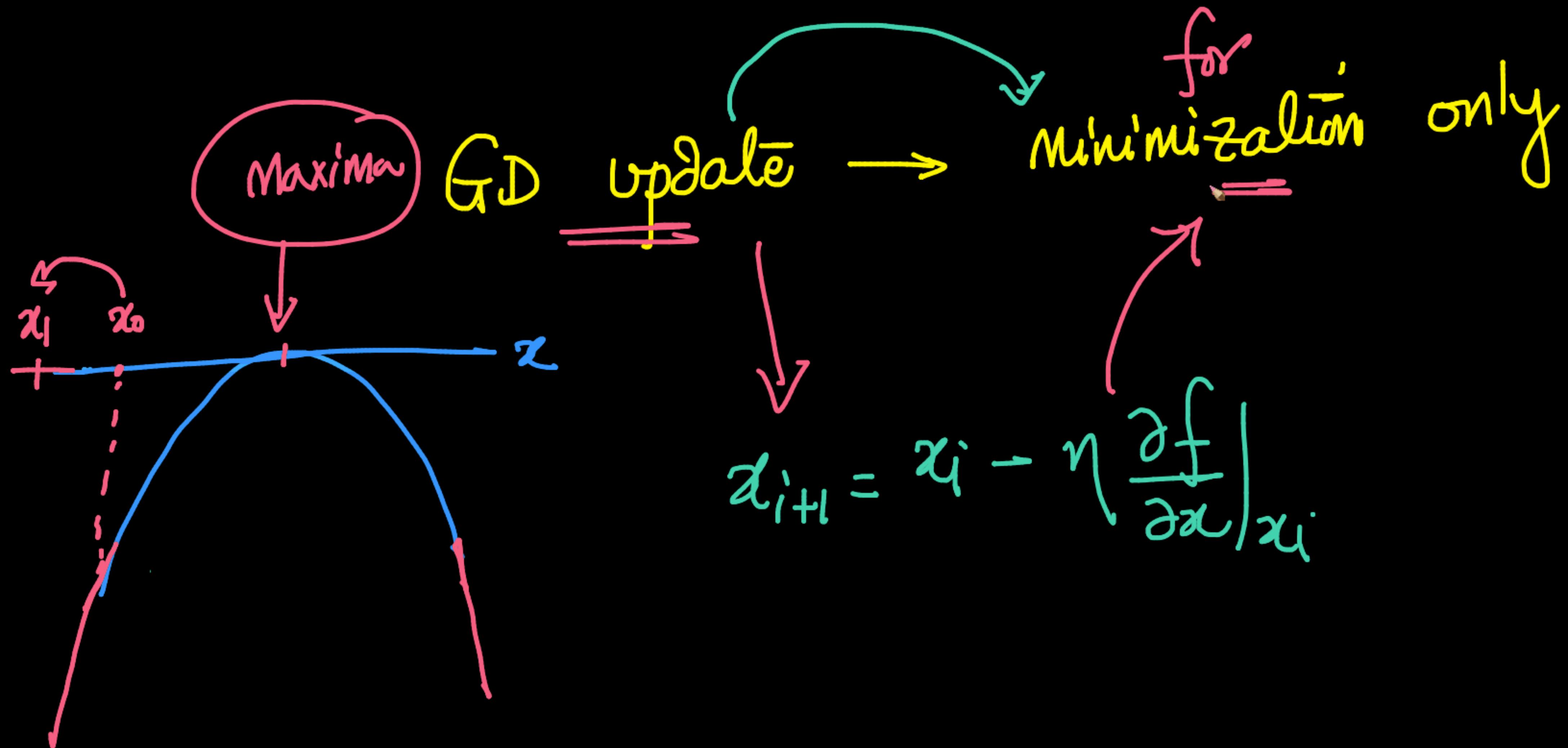
$$\left. \frac{df}{dx} \right|_{x_0, y_0} = 0$$

$$\underbrace{x_{10}, y_0}_{\sim} \rightarrow \underbrace{x_{10}, y_{11}}_{\sim}$$

A diagram showing a function  $f$  with a derivative  $\frac{df}{dx}$  evaluated at  $x_{10}, y_{11}$ . A yellow arrow points from the input  $x_{10}, y_0$  to the output  $x_{10}, y_{11}$ . A yellow circle surrounds the derivative term  $\frac{df}{dx}$ , and a yellow checkmark is placed next to the output  $x_{10}, y_{11}$ .







$$x^* = \max_x f(x) \underset{\approx}{=} x^* = \min_x [-f(x)]$$

ML: min only  
GD-update

Google Search  $x^2+y^2$

plot(x^2+y^2) - Google Search

google.com/search?q=plot(x%5E2%2By%5E2)&rlz=1C5CHFA\_enIN958IN958&oq=plot(x%5E2%2By%5E2&aqs=chrome.0.69i59j69i57j69i59j0i30l7.4707j0j7&sourceid=chrome&ie=UTF-8

Update

All Images News Maps Videos More Tools

About 81,20,00,000 results (0.55 seconds)

Graph for  $x^2+y^2$

$z$ -axis

$x$ -axis

$y$ -axis

$x_0, y_0$

$\downarrow$

$(x_i, y_i)$

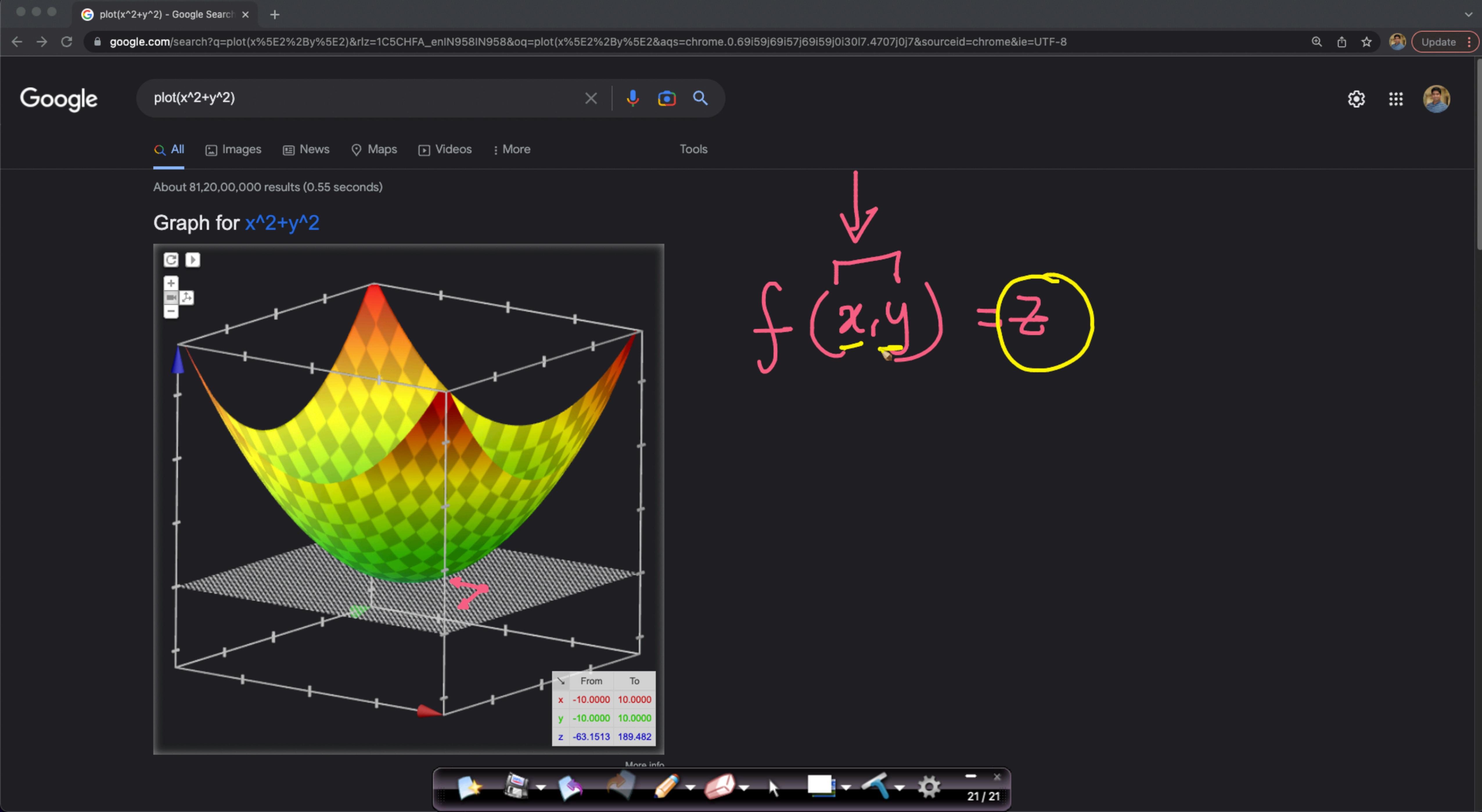
$g_r$

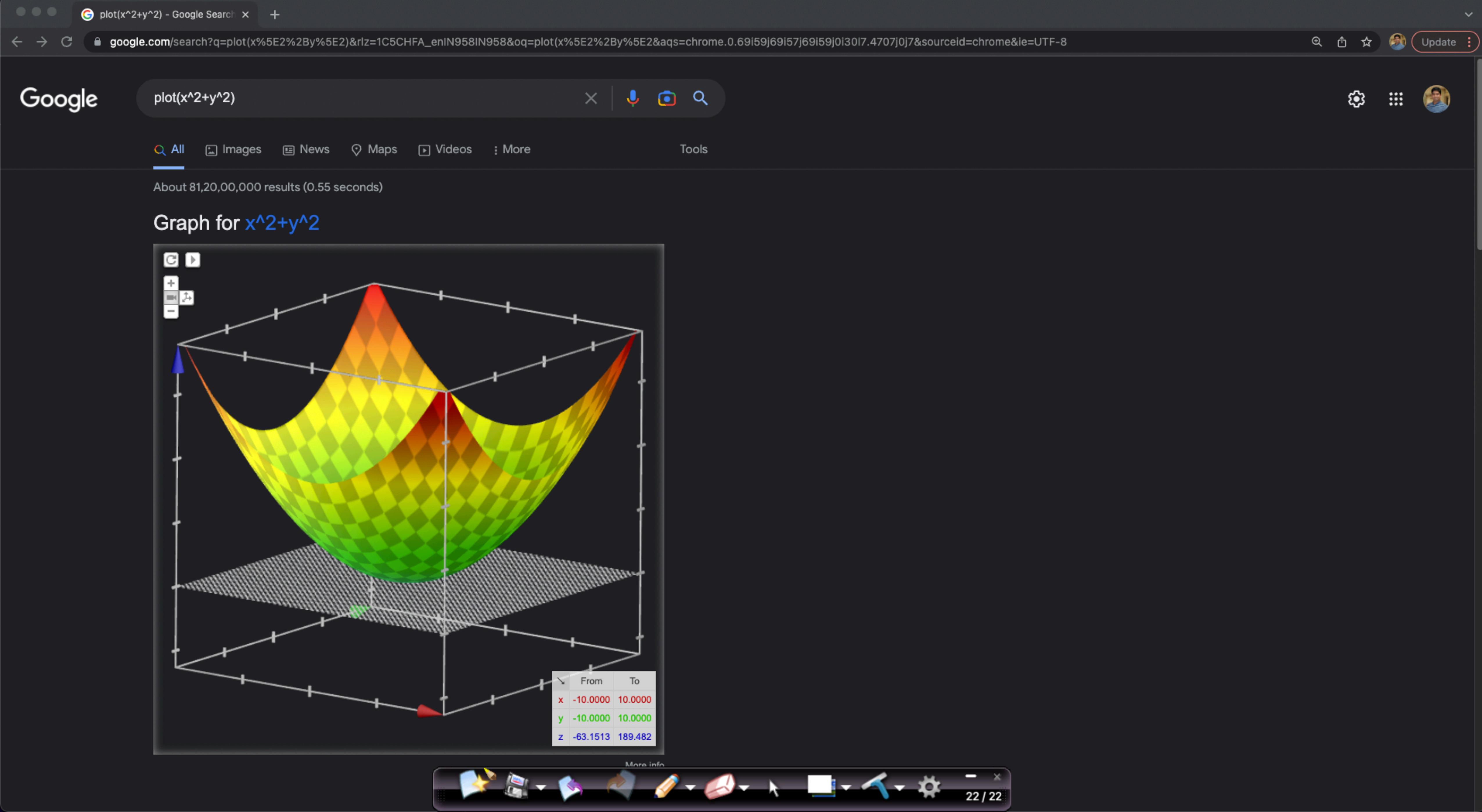
From To

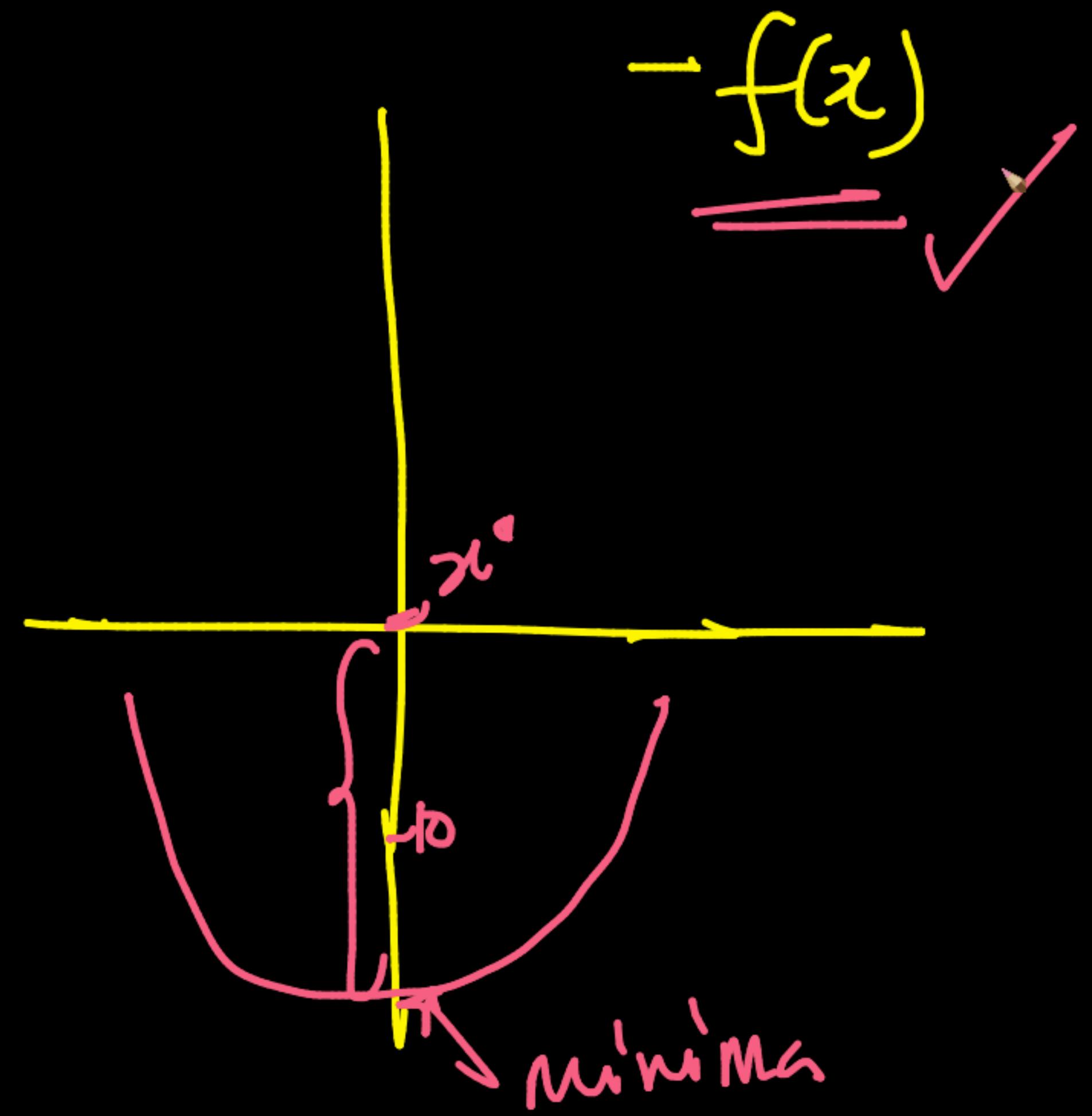
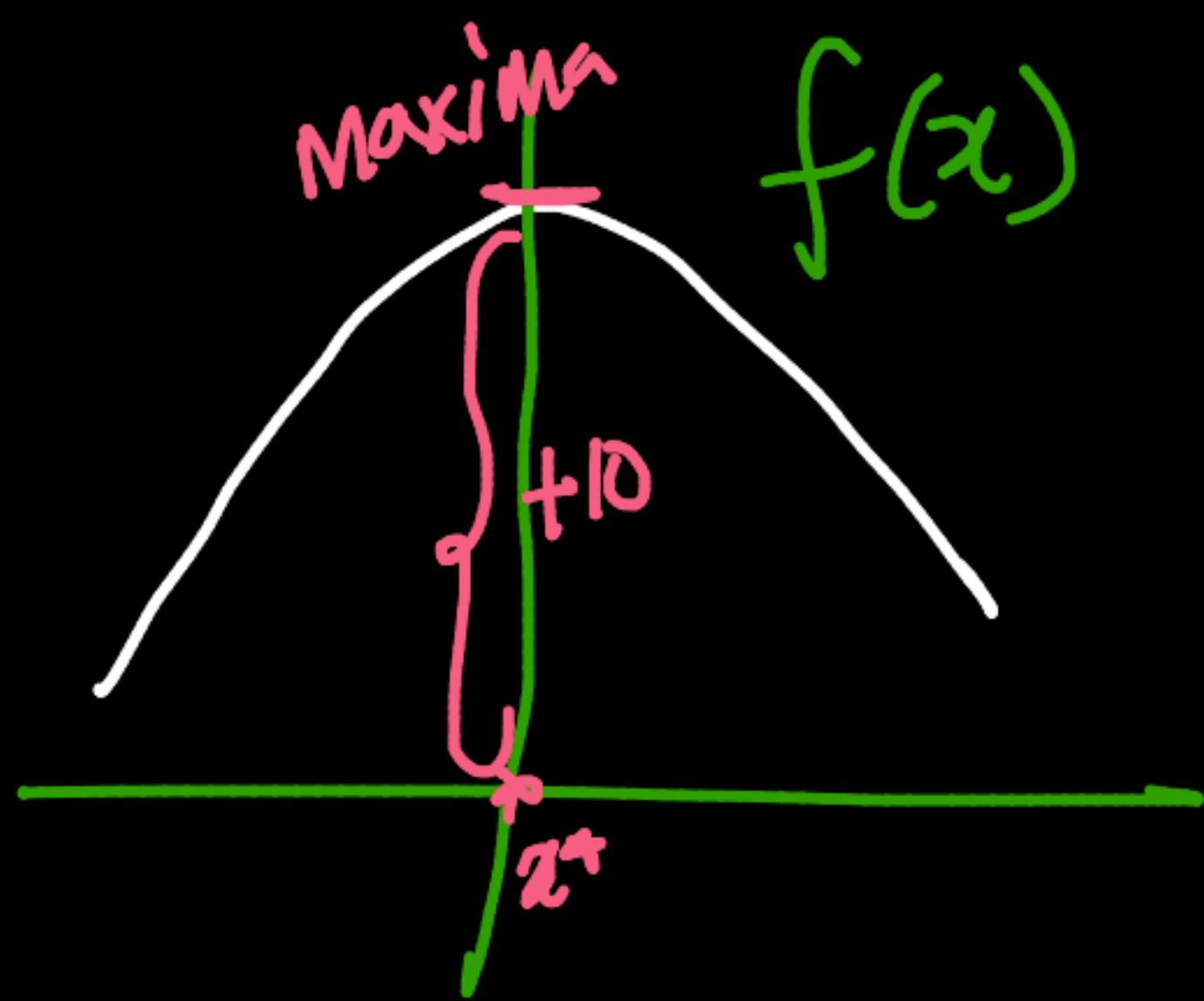
x	-10.0000	10.0000
y	-10.0000	10.0000
z	-63.1513	189.482

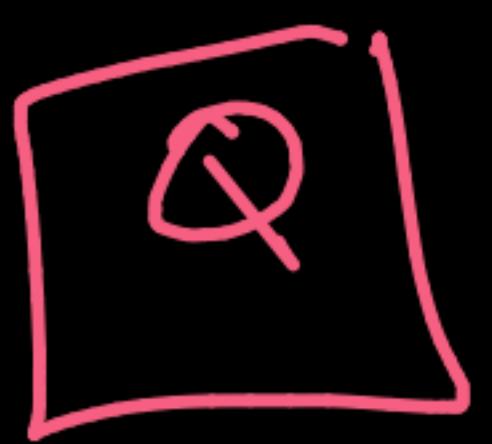
More info

20 / 20









GD-min:

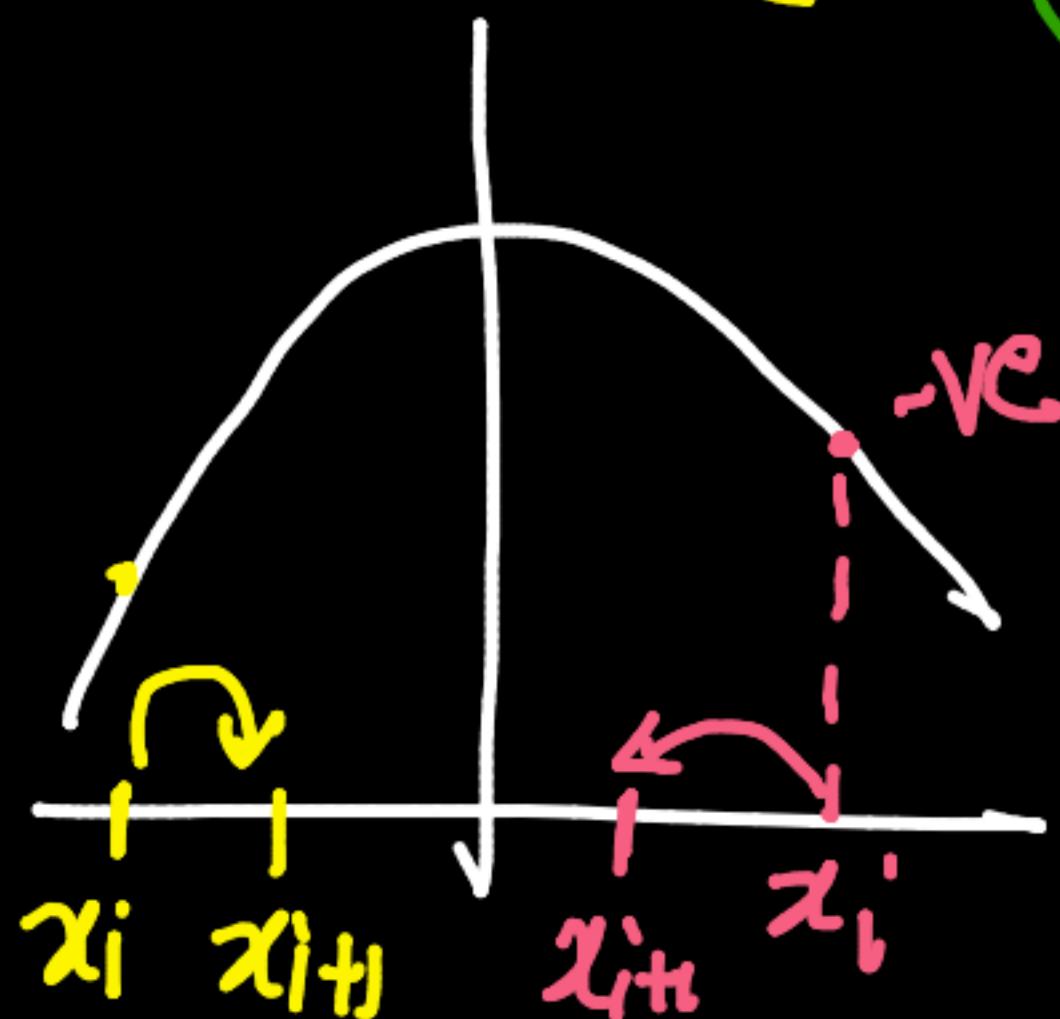
$$x_{i+1} = x_i - \eta \frac{\partial f}{\partial x} \Big|_{x_i}$$

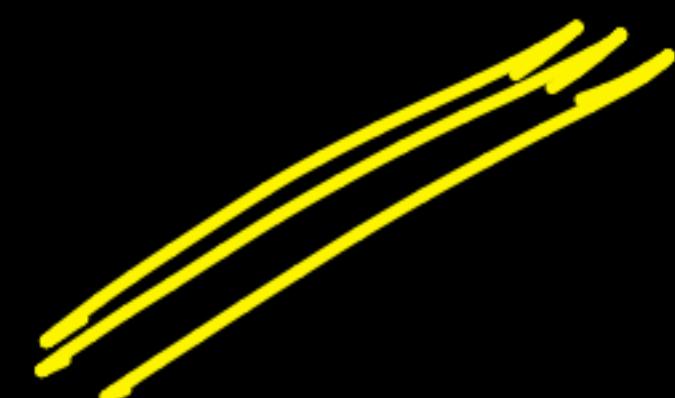
+ve

GD-Max:

$$x_{i+1} = x_i + \eta \frac{\partial f}{\partial x} \Big|_{x_i}$$

-ve





ML:  $\{0,1\}$

$\checkmark \left\{ \begin{array}{l} \text{MIN} \\ \cancel{\text{MAX}} \\ w, w_0 \end{array} \right.$

$$\sum_{i=1}^n -y_i \left( \overbrace{w^T x_i + w_0}^{\text{Ans}} \right) / \|w\|$$

Binary Classfn. Task

$$f(w_0, w_1, \dots, w_d) \xrightarrow{\text{d+1-var}} l_i$$

$$\text{Ans. MIN } w, w_0 \quad \sum_{i=1}^n \left[ -y_i \left( w_0 + w_1 x_{i1} + w_2 x_{i2} + \dots + w_d x_{id} \right) \right]$$

$$\sqrt{w_1^2 + w_2^2 + w_3^2 + \dots + w_d^2}$$

Loss-fn:

$$(x_i, y_i) \xrightarrow{\text{Ans}} \delta$$

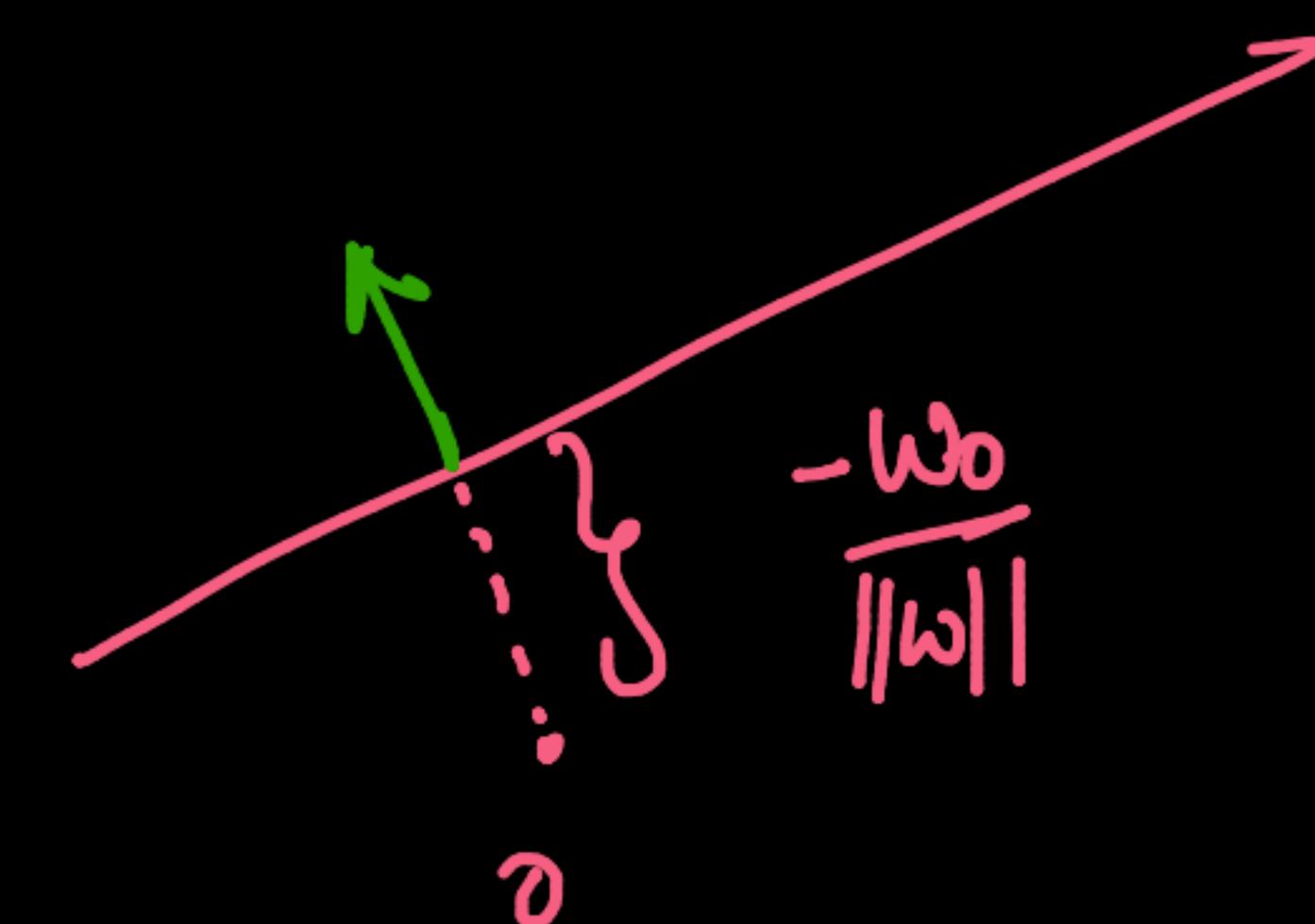
Eqn. of plane:

$$\tilde{w}^T \tilde{x} + \tilde{w}_0$$

$\downarrow$   
 $d$

$$y = Mx + \underline{C}$$

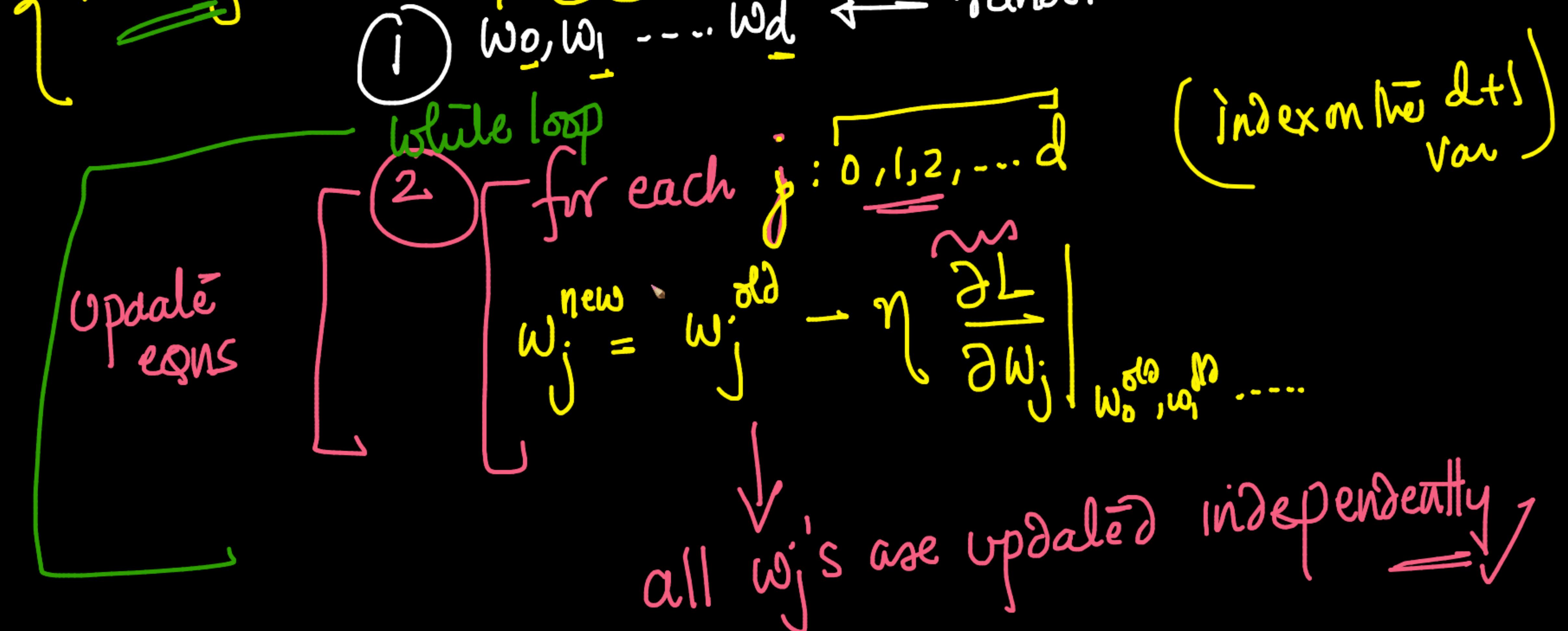
$$\overline{Td}: \underline{w^T x} + \underline{w_0} = 0$$



# Binary-Classif:

i)  $w_0, w_1, \dots, w_d \leftarrow$  random-values

# Binary-Classif:



for all  
↓

$$\forall j=0,1,2\dots d;$$

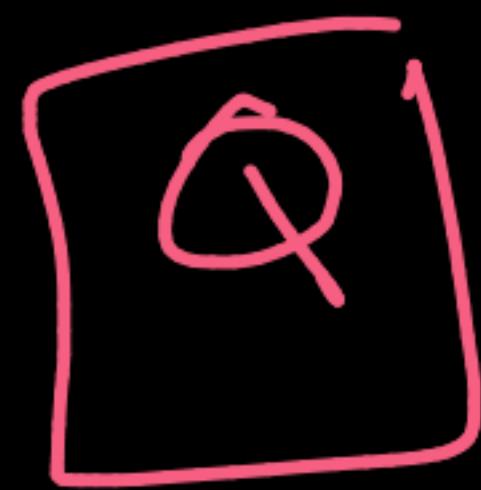
~~$\nabla$~~

$$\frac{\partial L}{\partial w_j}$$

$\approx 0$

easily

{d-variables  
 $d+1 \rightarrow$  variables  
=}



$$\frac{\partial L}{\partial w_1} = ?$$

$$\frac{d}{dx}$$

$$\sum_{i=1}^n$$

{ Quotient - rule }  $\frac{u}{v} = \frac{u'v - uv'}{v^2}$

V-complex

$$L = \sum_{i=1}^n -y_i \left( \underline{w_0 + w_1 x_{i1}} + w_2 x_{i2} + \dots + w_d x_{id} \right)$$

Messy

$$\frac{df+g}{dx}$$

$$\left\{ \frac{w_1^2}{-} + w_2^2 + \dots + w_d^2 \right.$$

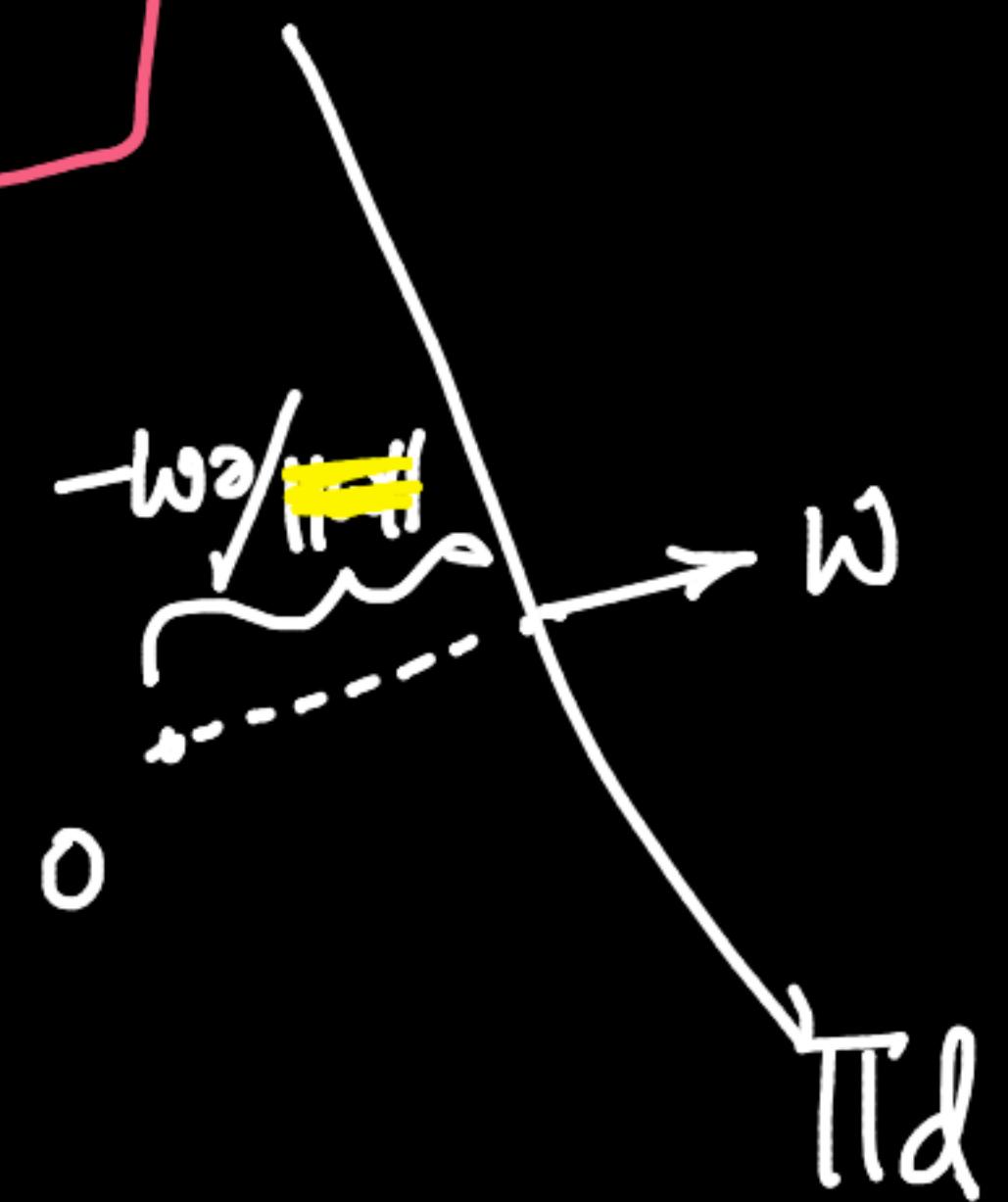
chain rule

$$\frac{f}{g} = \frac{f'}{g} - \frac{fg'}{g^2}$$

Hack of geometry  
& optimization

$$L = - \sum_{i=1}^n y_i \frac{(\omega^T x_i + w_0)}{\|\omega\|}$$

enforce  $\|\omega\| = 1$   
↑  
Unit-vec



Unconstrained

$$\left\{ \begin{array}{l} \min_{w, w_0} L \\ \text{---} \end{array} \right.$$

$$-\sum_{i=1}^n y_i (\underline{w^T x_i + w_0})$$

$\|w\|$   $\rightarrow$  headache

$\pi_d$

Constrained

~~Opt~~  $\left\{ \begin{array}{l} \min_w L \\ \text{---} \end{array} \right.$

$$\approx -\sum_{i=1}^n y_i (\underline{w^T x_i + w_0})$$

such that

$$\|w\| = 1$$

Constraint

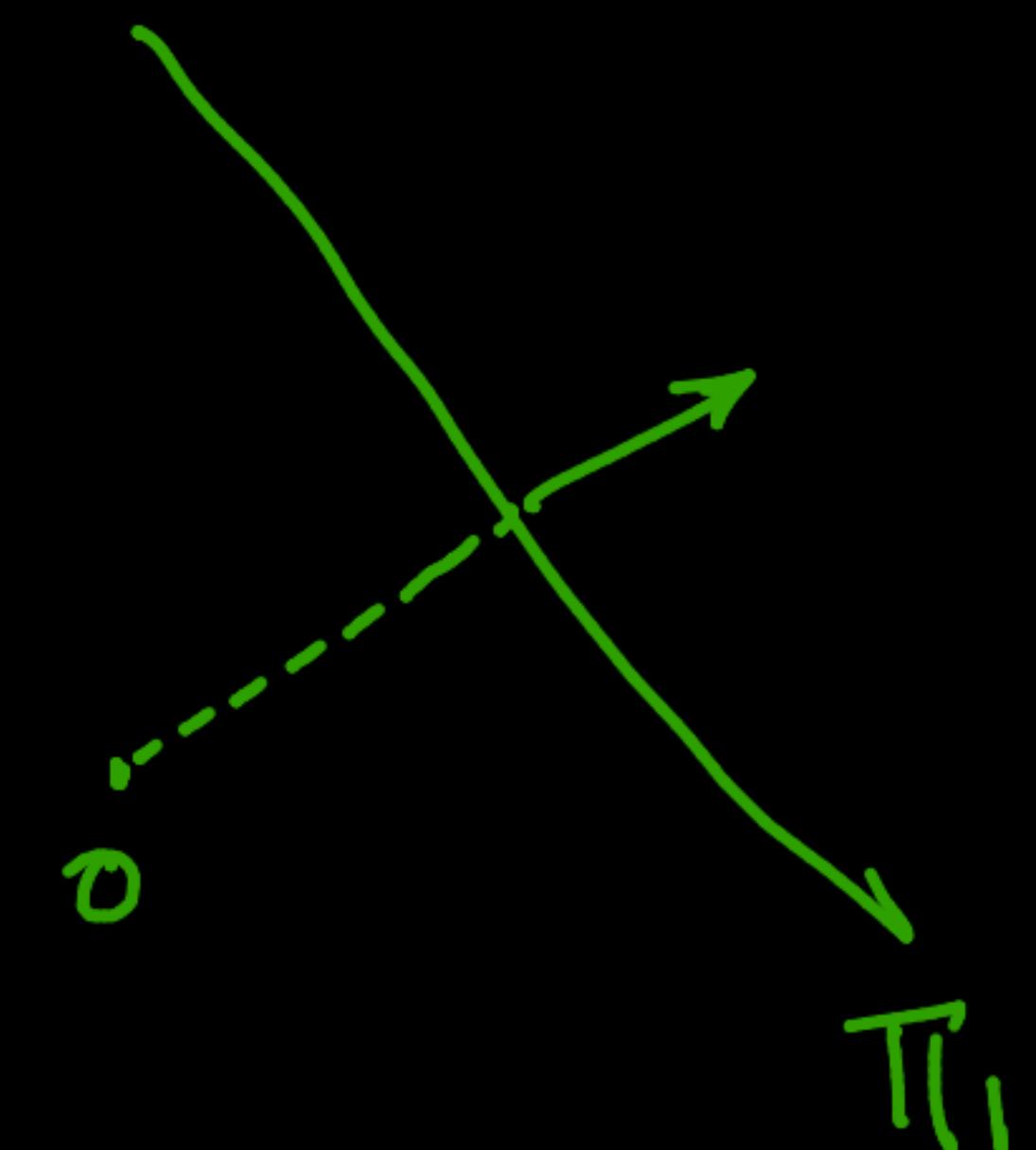
$$\rightarrow \text{dist} : \frac{w_0}{\|w\|}$$

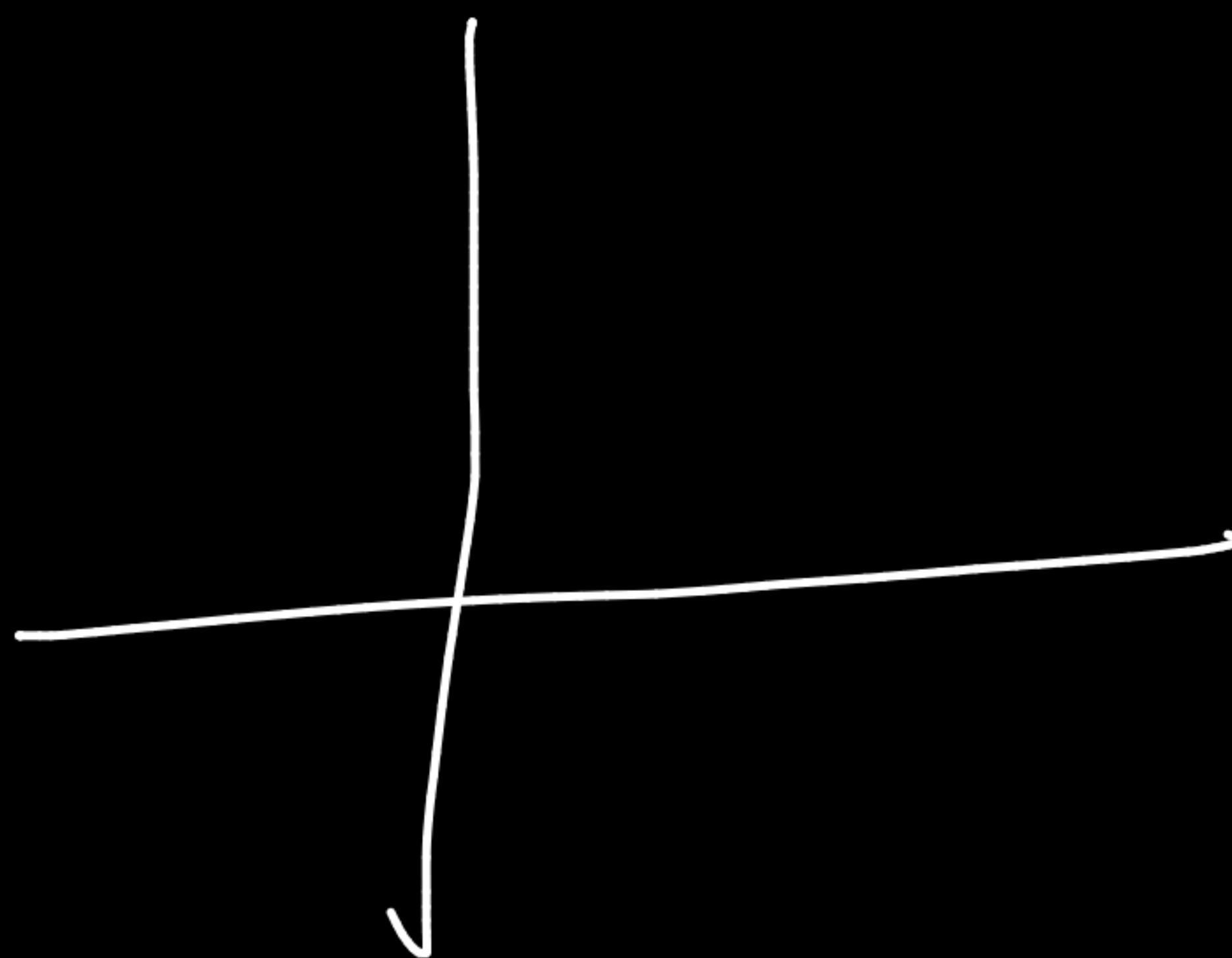

$\pi_1 \rightarrow \left\{ w_0 + w_1 x_1 + w_2 x_2 + \dots + w_d x_d = 0 \right.$

$\pi_2 \rightarrow \left\{ \frac{w'_0}{c} + \frac{w'_1}{c} x_1 + \frac{w'_2}{c} x_2 + \dots + \frac{w'_d}{c} x_d = 0 \right.$

dist:

$$\sqrt{\left(\frac{w_1}{\alpha}\right)^2 + \left(\frac{w_2}{\alpha}\right)^2 + \dots + \left(\frac{w_d}{\alpha}\right)^2}$$





$$\begin{aligned}l_1: \quad & w_0 + \\& w^T x = 0 \\& 2x + 4y + 6 = 0 \\l_2: \quad & x + 2y + 3 = 0\end{aligned}$$





$$x^{\dagger}, y^{\dagger} =$$

$$\min_{x,y} f(x,y)$$

s.t.

$$g(x,y) = c$$
$$g(x,y) \leq c$$

$$(x_0, y_0)$$

$$\begin{array}{l} \xrightarrow{\quad} x^*, y^* \\ \approx \\ \left\{ \begin{array}{l} \min_{x, y} f(x, y) \\ + \lambda (g(x, y) - c) \end{array} \right. \\ \uparrow \\ \lambda > 0 \\ \text{Lagrange Multiplier} \end{array}$$

$$\mathcal{L}(x, y, \lambda) = f(x, y) + \lambda (g(x, y) - c)$$

↑  
Lagrangian fn

eq. constraint  
opt. problem

$$\min_{w, w_0} \sum_{i=1}^n -y_i (w^T x_i + w_0) \rightarrow f(w, \underline{w_0})$$

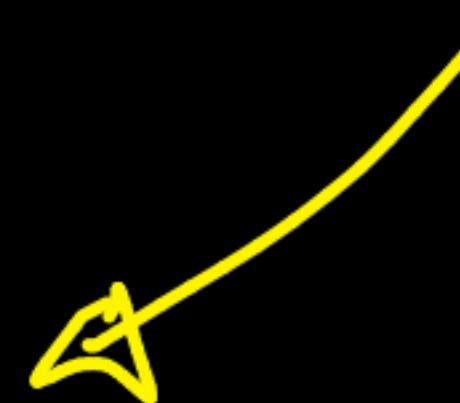
$$\text{s.t. } \|w\|^2 = 1$$

$$\|w\|^2 - 1 = 0$$

$$g(w, w_0)$$

$$0.5(w_1^2 + w_2^2 + \dots + w_d^2) + \dots$$

$w_0, w$



Unconstrained  
opt :

$\tilde{w}, \tilde{w}_0$

$$\min_{w, w_0} \sum_{i=1}^n -y_i (\omega^T x_i + w_0)$$

$$+ \lambda (||\omega||^2 \rightarrow)$$

$$\lambda > 0$$

$x, y^+$

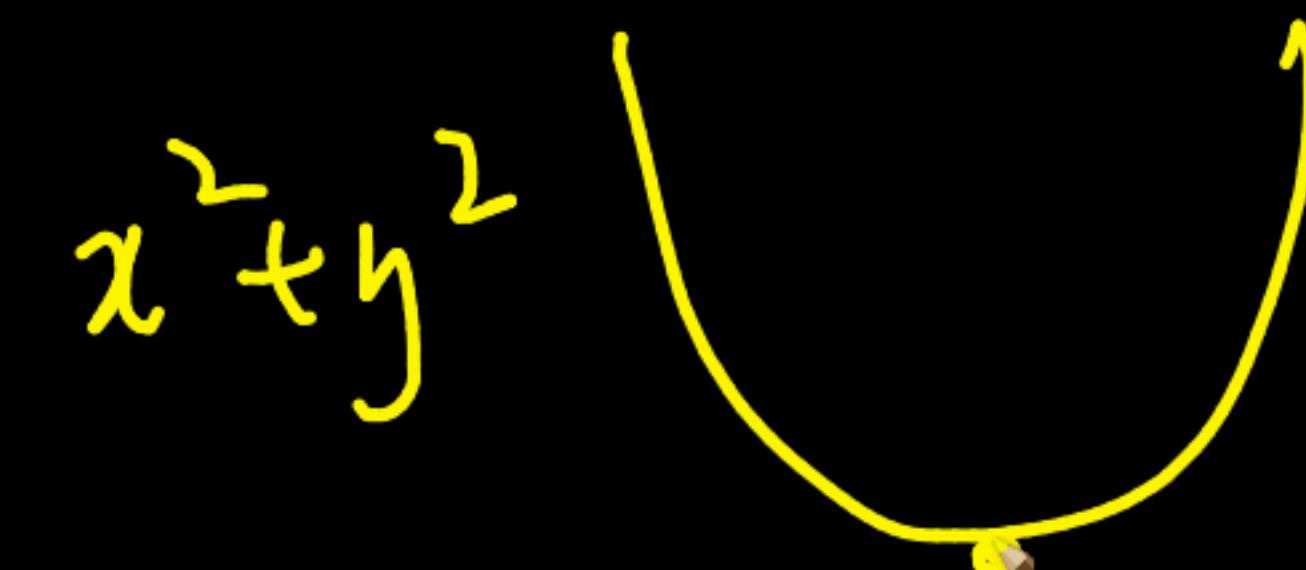
$$\min_{x,y} f(x,y)$$

$$\text{s.t. } g(x,y) = c$$

$$\min_{x,y} f(x,y) + \lambda (g(x,y) - c) = 0$$

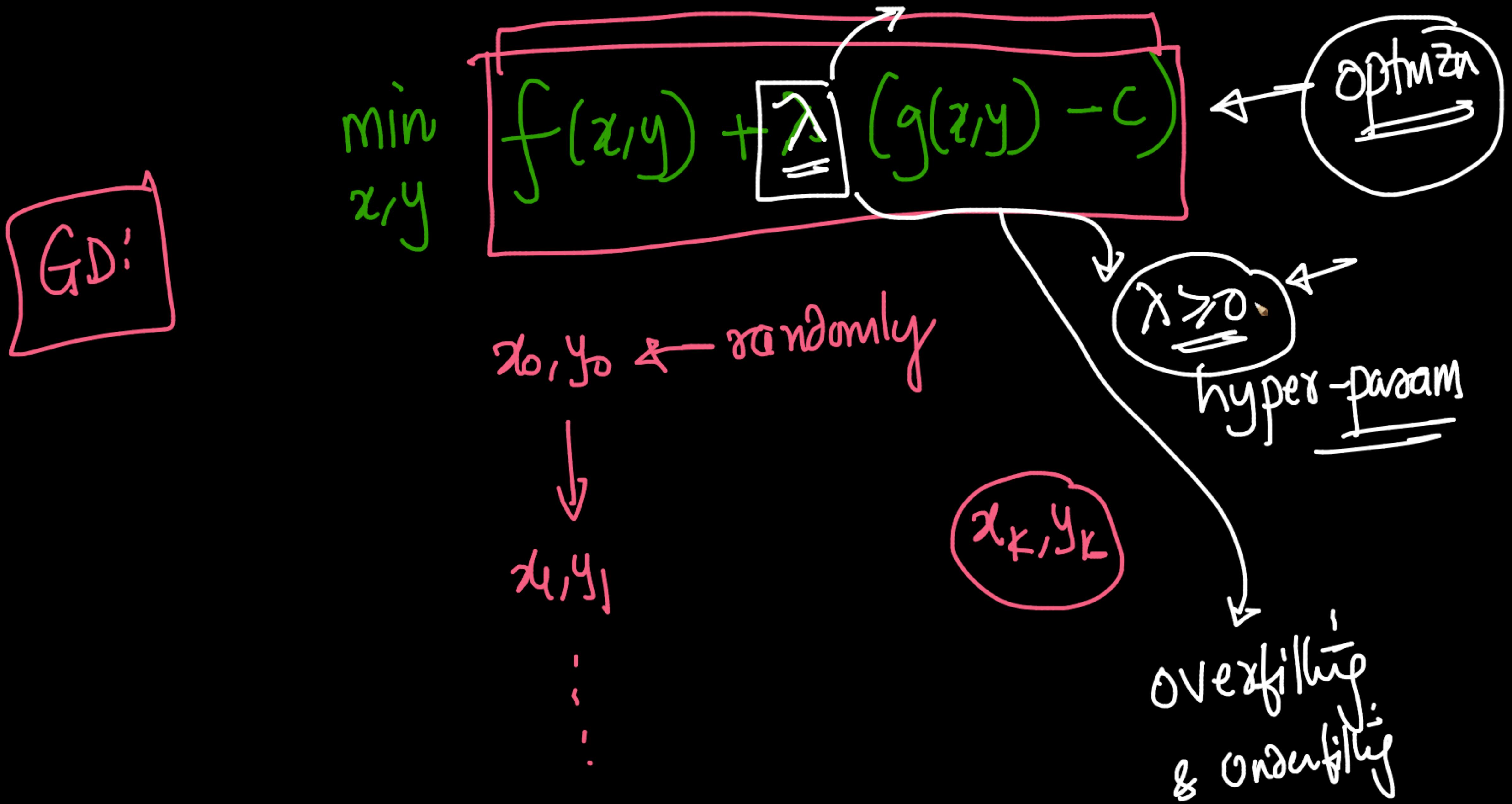
Const. opt

(0,0)



Unconst-opt-pads





ML

$$\left\{ \begin{array}{l} \min_{w, w_0} \sum_{i=1}^n -y_i (\bar{w}^T x_i + w_0) \\ \text{s.t. } \|\bar{w}\|^2 = 1 \end{array} \right.$$

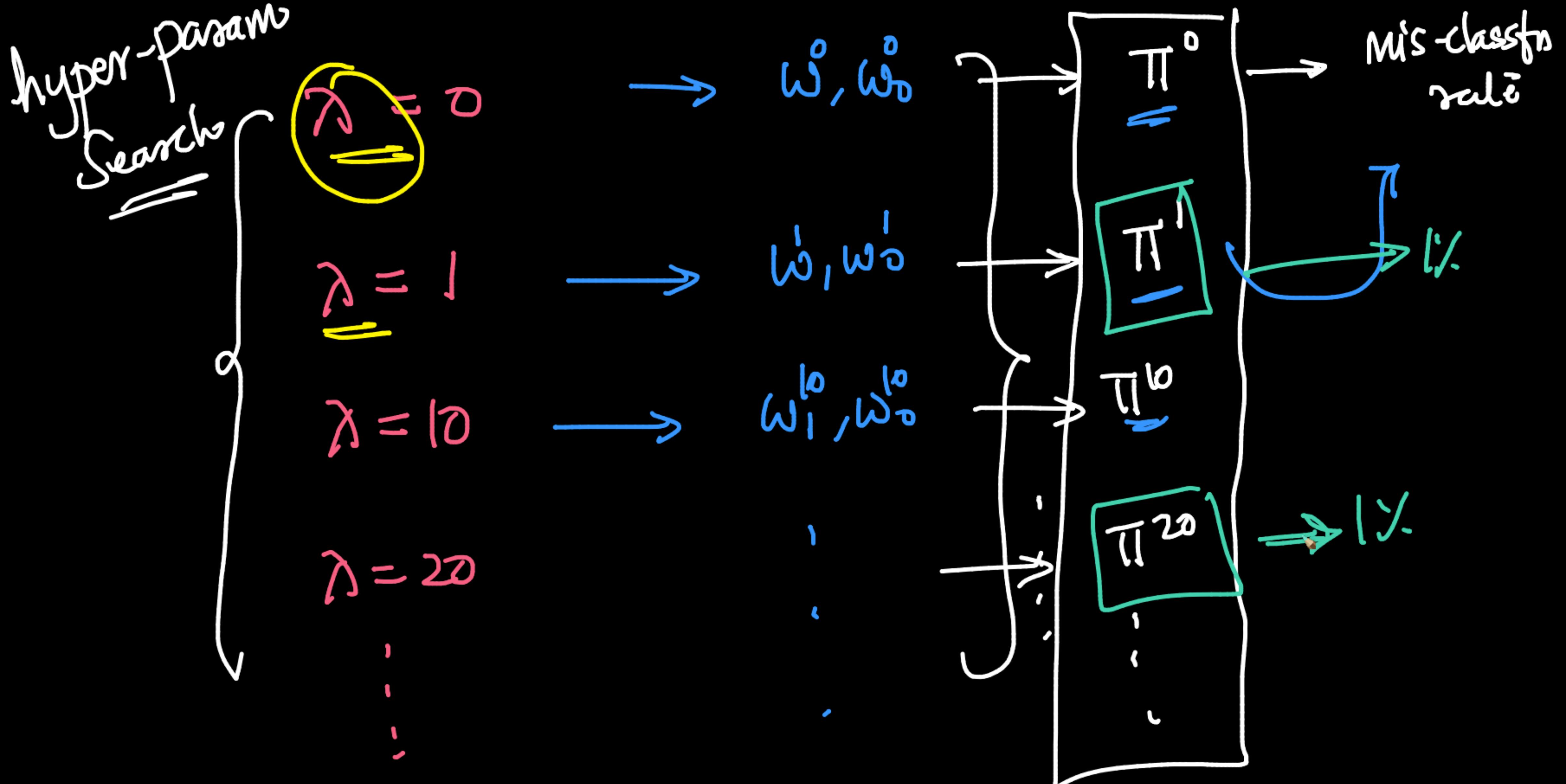
$$\lambda = 10 > 0$$

$$\lambda > 0$$

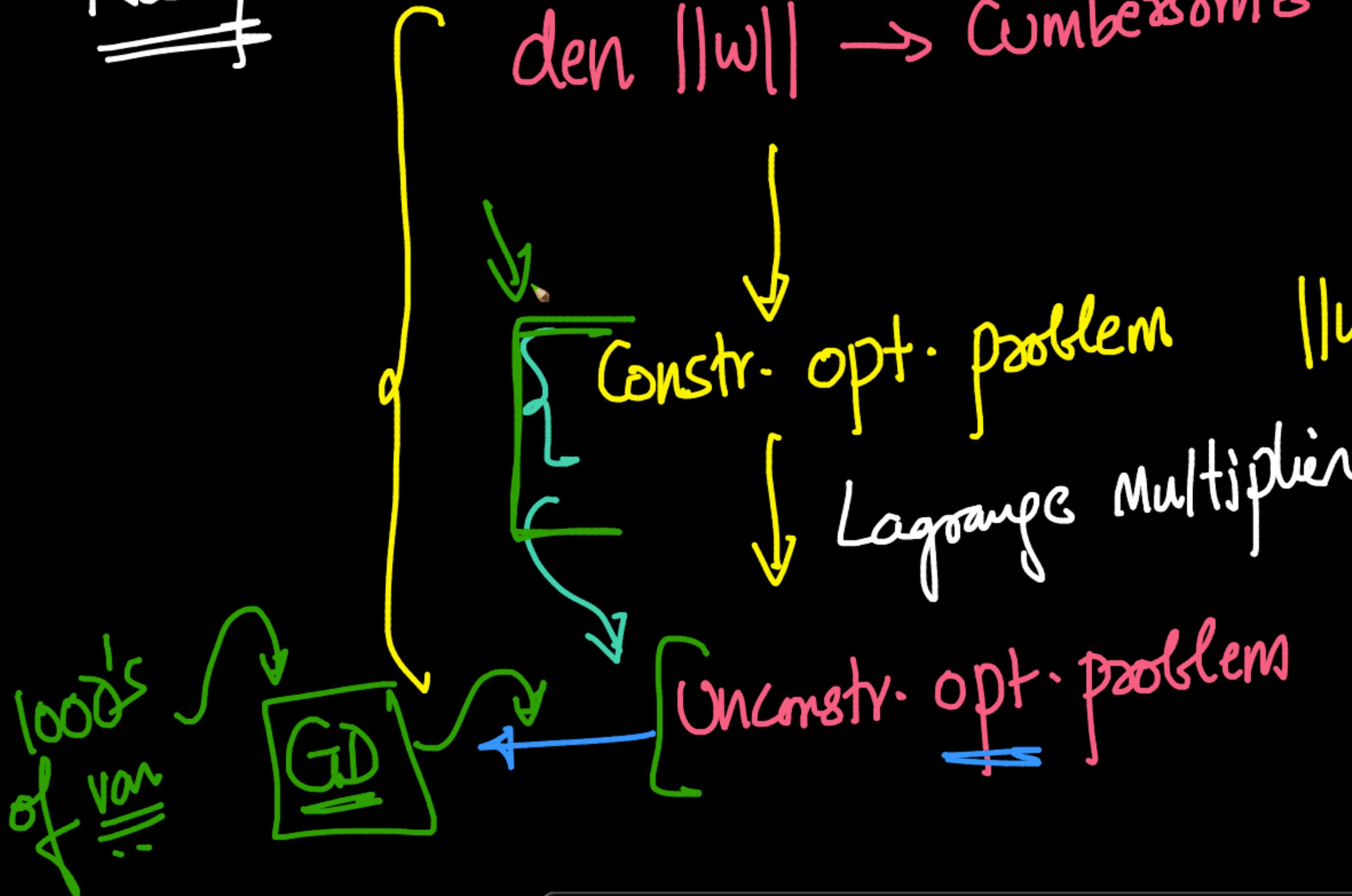
how to  
solve

$$\left\{ \begin{array}{l} \min_{\bar{w}, \bar{w}_0} \sum_{i=1}^n -y_i (\bar{w}^T x_i + \bar{w}_0) + \lambda (\|\bar{w}\|^2 - 1) \end{array} \right.$$

GD  $\rightarrow$  best  $\bar{\bar{w}}, \bar{\bar{w}}_0$

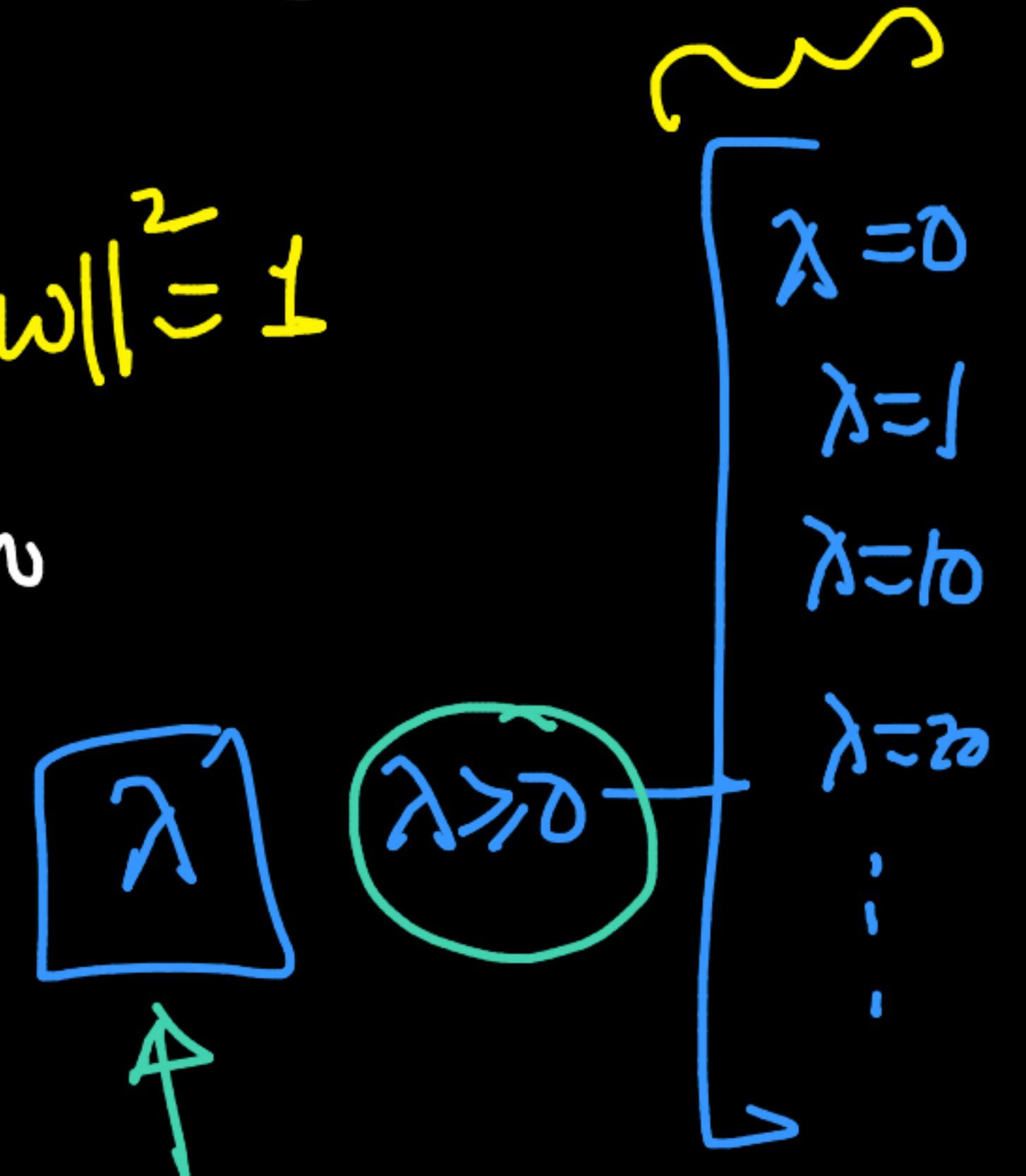


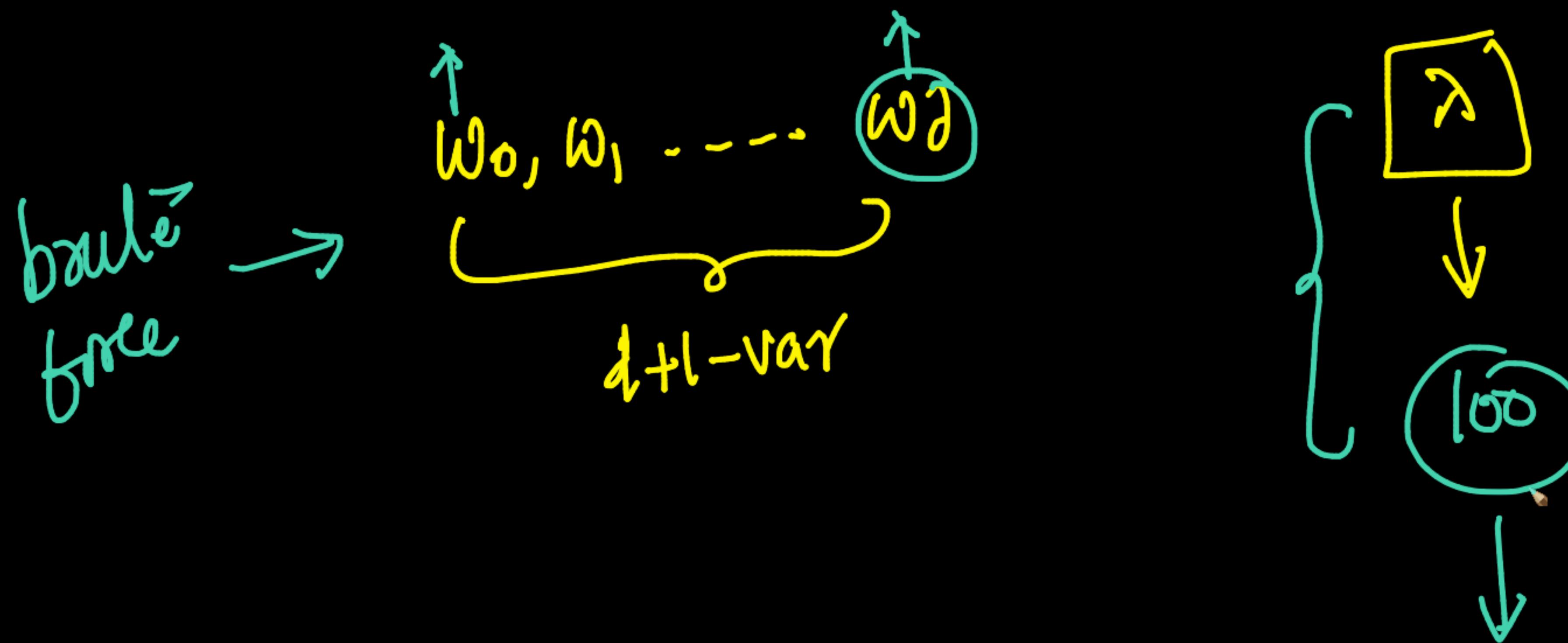
Recap:



$$\frac{\partial L}{\partial w_j}$$

$w, w_0$





random ↗ {  
     $\lambda = 0 \cdot 1$   
     $\lambda = 1$   
     $\lambda = 10$   
     $\lambda = 100$

→ hyper-param tuning

hyper-param Search  
( $\lambda$ )

Consty  
opt

Lagrange fu  $\rightarrow \mathcal{L}(x, y, \lambda)$

$$\min_{x, y} f(x, y) + \lambda (g(x, y) - c)$$

On consty opt

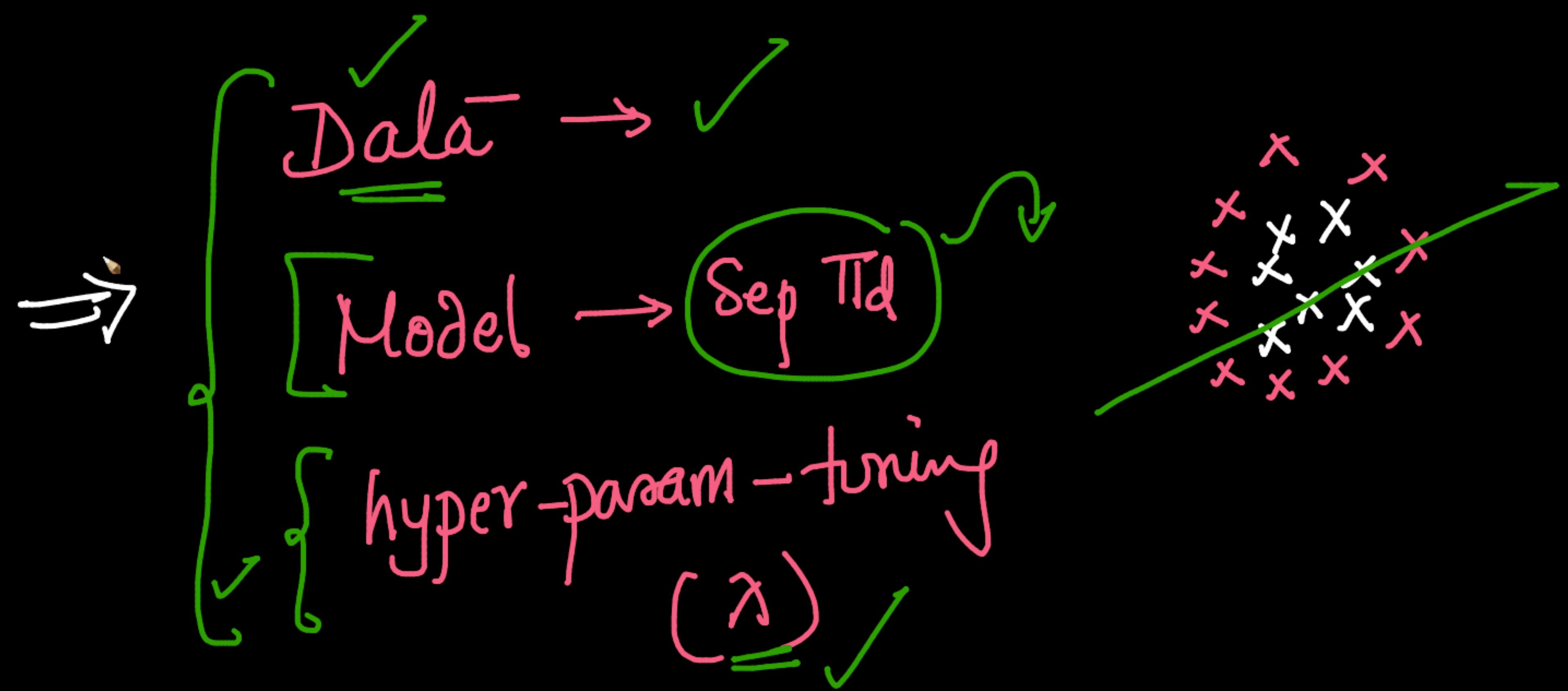
$$\left\{ \begin{array}{l} \frac{\partial \mathcal{L}}{\partial x} = 0 \\ \frac{\partial \mathcal{L}}{\partial y} = 0 \\ \frac{\partial \mathcal{L}}{\partial \lambda} = 0 \end{array} \right.$$

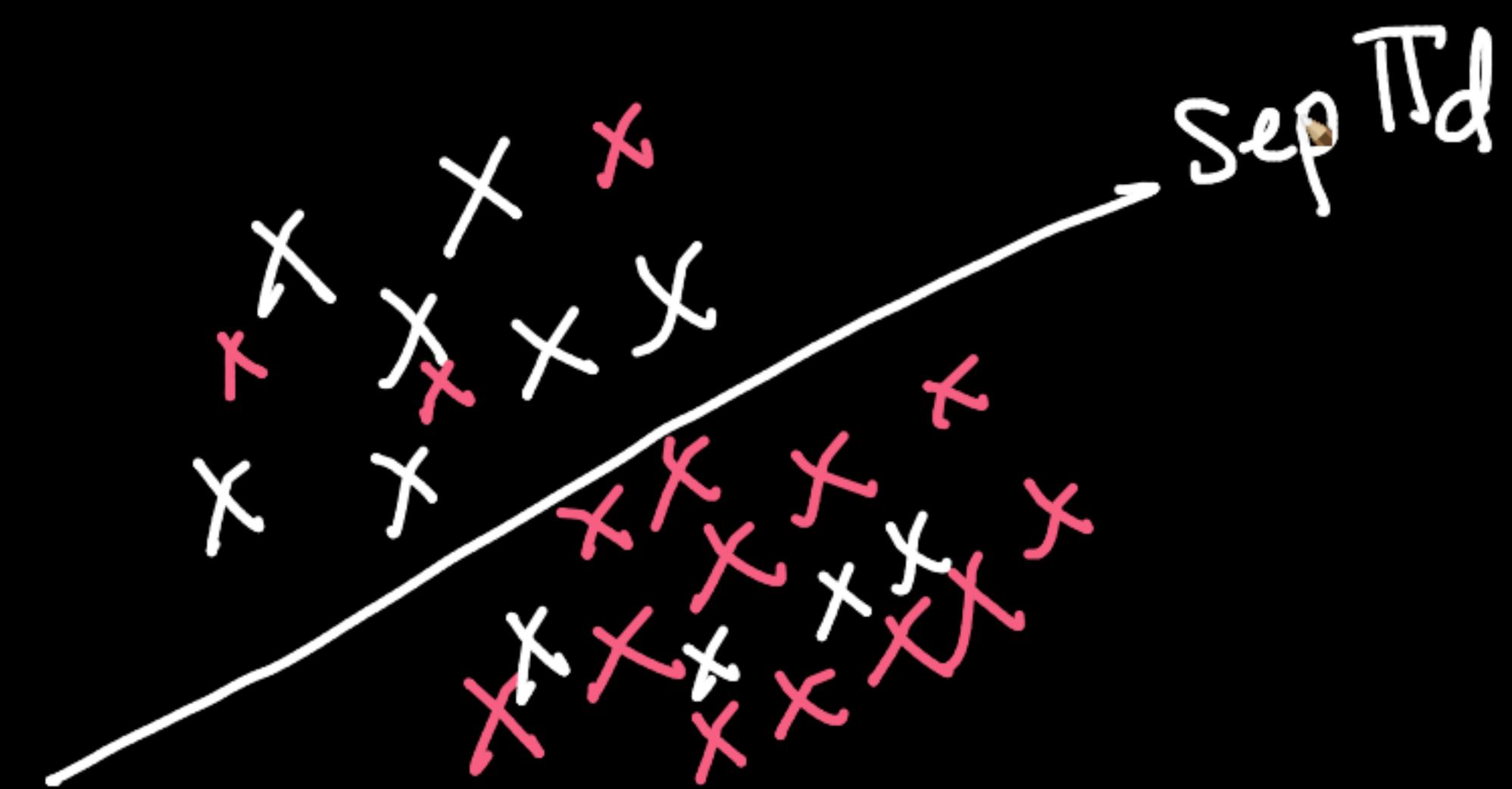
one eqn  $x, y, \lambda$

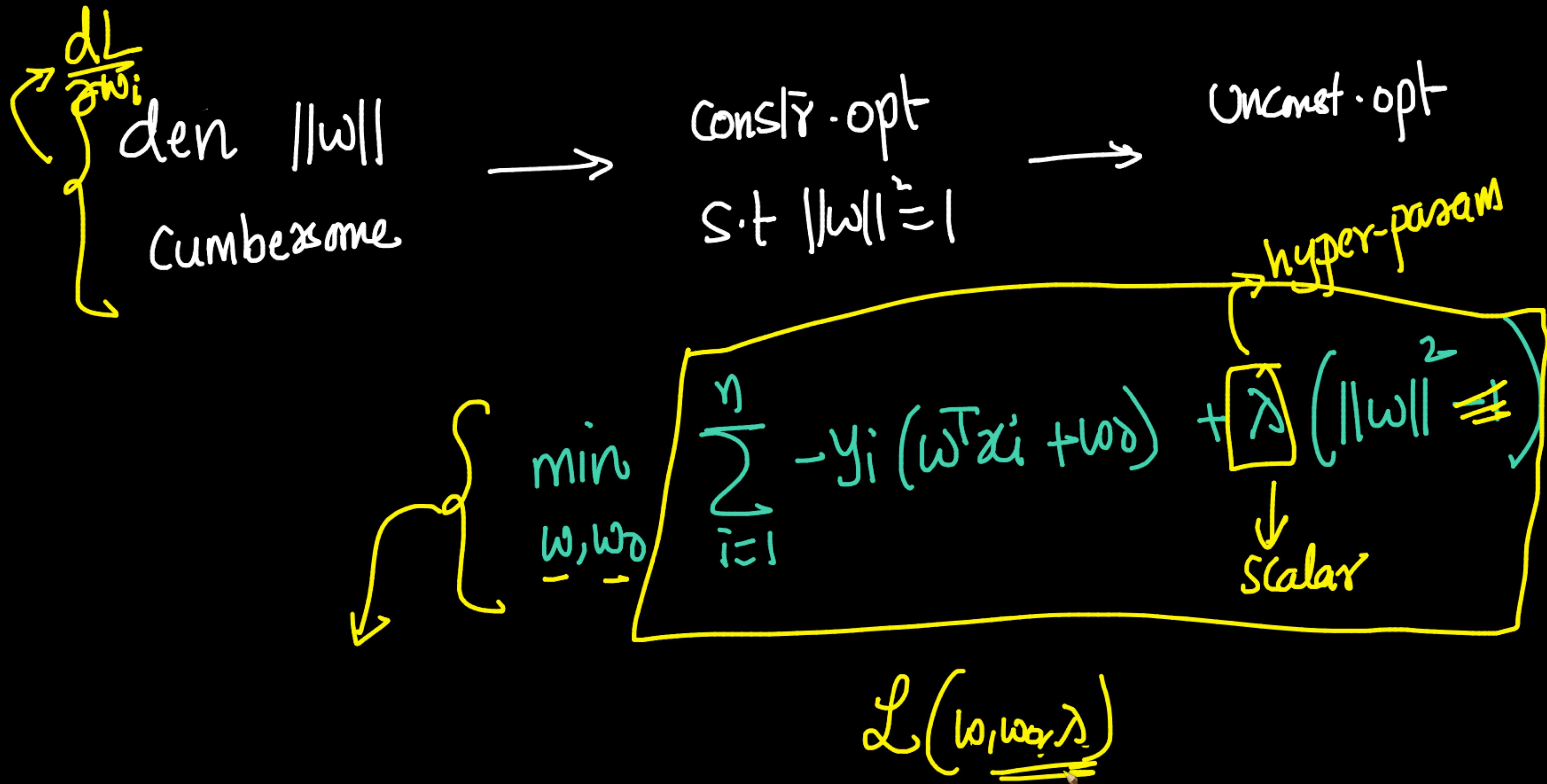
2nd eqn  $x, y, \lambda$

3rd eqn  $x, y, \lambda$

$x, \lambda, y$







$$\begin{array}{l} x=0 \\ y=0 \end{array} \leftarrow \min x^2 + y^2 + c$$

$$\begin{array}{l} x=0 \\ y=0 \end{array} \leftarrow \min x^2 + y^2$$

$$\min_{w, w_0} \sum_{j=1}^n -y_j(w^T x_j + w_0) + \lambda \|w\|$$

$w_0, w_1, \dots, w_d =$

$$\min_{w_0, w_1, \dots, w_d} \sum_{j=1}^n -y_j \left( w_0 + \underbrace{w_1 x_{j1} + w_2 x_{j2} + \dots + w_d x_{jd}}_{-} \right) + \lambda (w_1^2 + w_2^2 + \dots + w_d^2)$$

$\lambda = 10$  (let)  
(const)



$$\frac{\partial \mathcal{L}}{\partial w_0} = \left[ \sum_{i=1}^n -y_i \cdot 1 \right] + \lambda \cdot 0 = \sum_{i=1}^n -y_i$$

$$\sum_{i=1}^n -y_i$$

$\forall j \in 1, 2, \dots, d$

$$\frac{\partial \mathcal{L}}{\partial w_j} = \sum_{i=1}^n (-y_i x_{ij}) + \lambda (2 \cdot w_j)$$

batch GD

$\lambda : 1 \text{ Million}$   
~~1000~~

$d+1$

[ ]

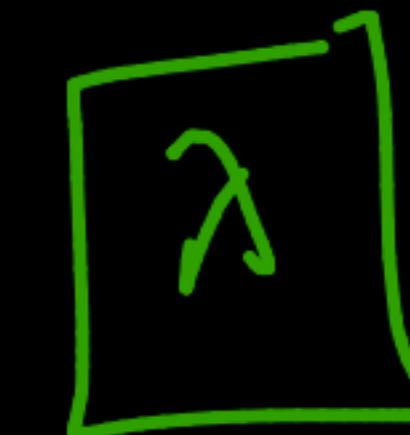
Update eqn

$w_0 \leftarrow$

+  $j = 1, 2, \dots, d$

$w_j \leftarrow$

m-iterations



$\rightarrow w_0, w_1, \dots, w_d$

GD:

$\ln$

each iteration ( $m$ -iter)

( $m$ -iter)

$= 10^5$   
 $= \underline{100,000}$  pts  
,  
n datapoints

$$\frac{\partial L}{\partial w_j}$$

go through

d+1 partial dev  
 $(100) = 10^2$

$10^7$

↓  
10 Million

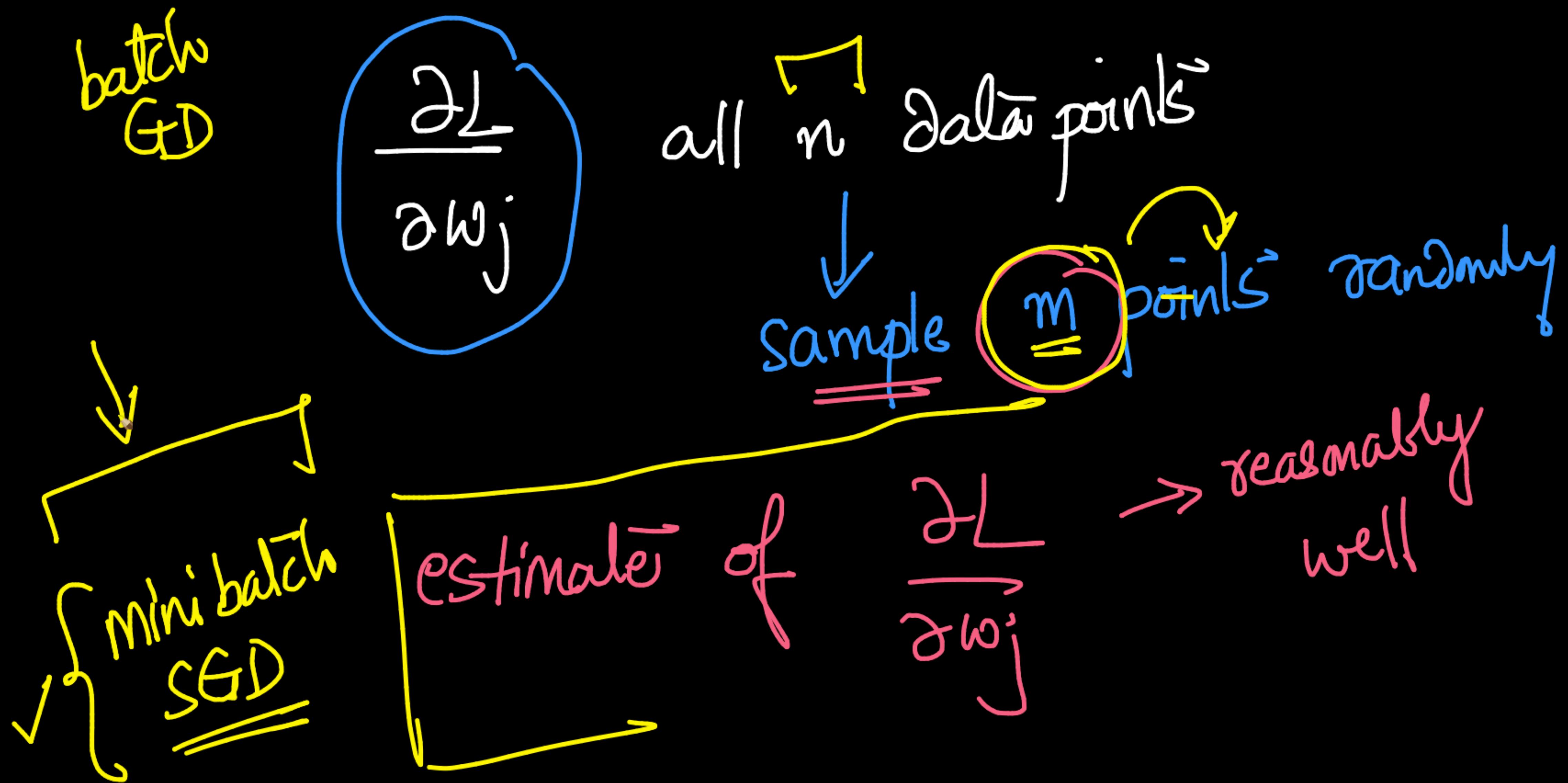
In each iteration

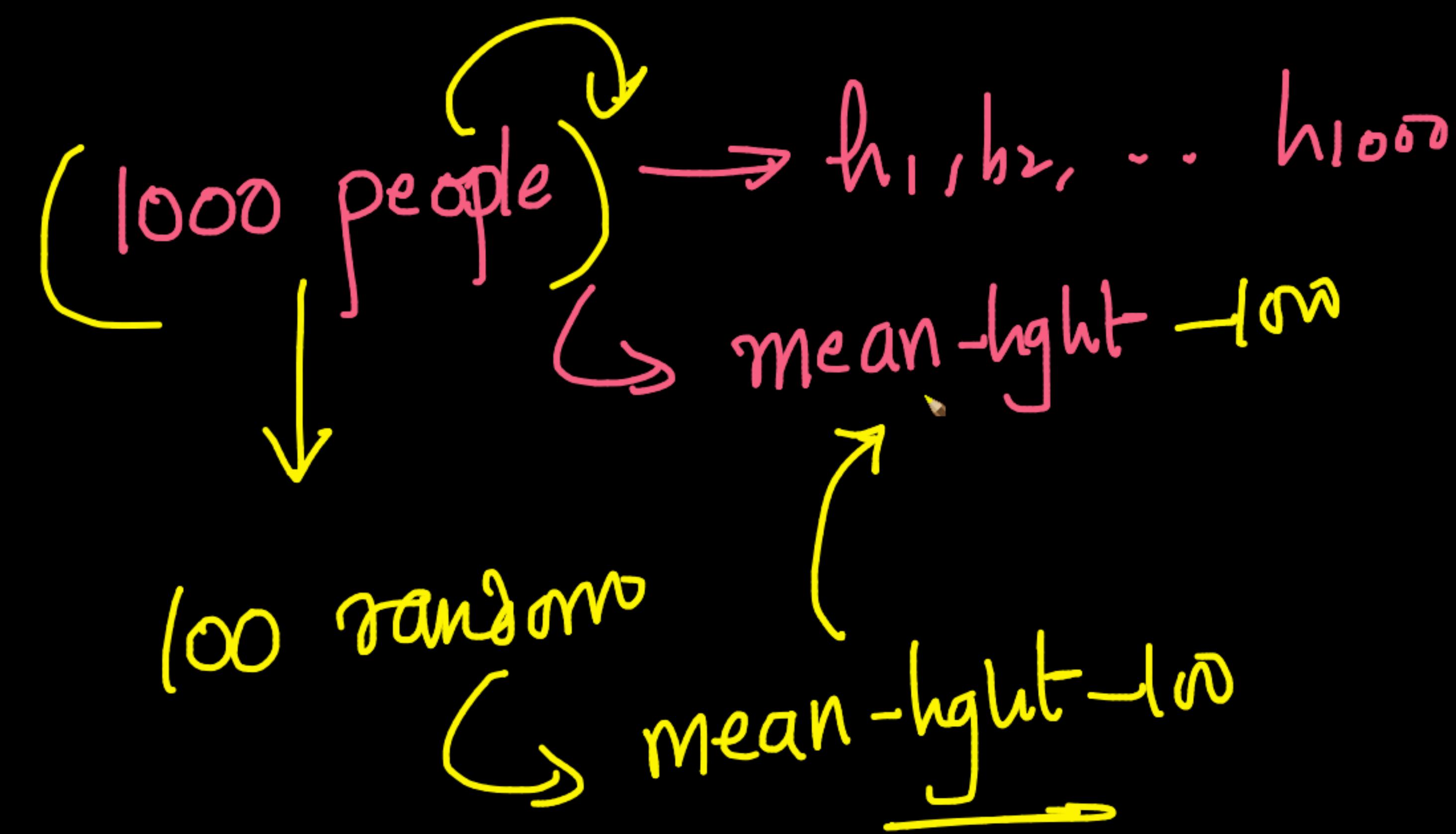
[for each update-equation → dH-variables

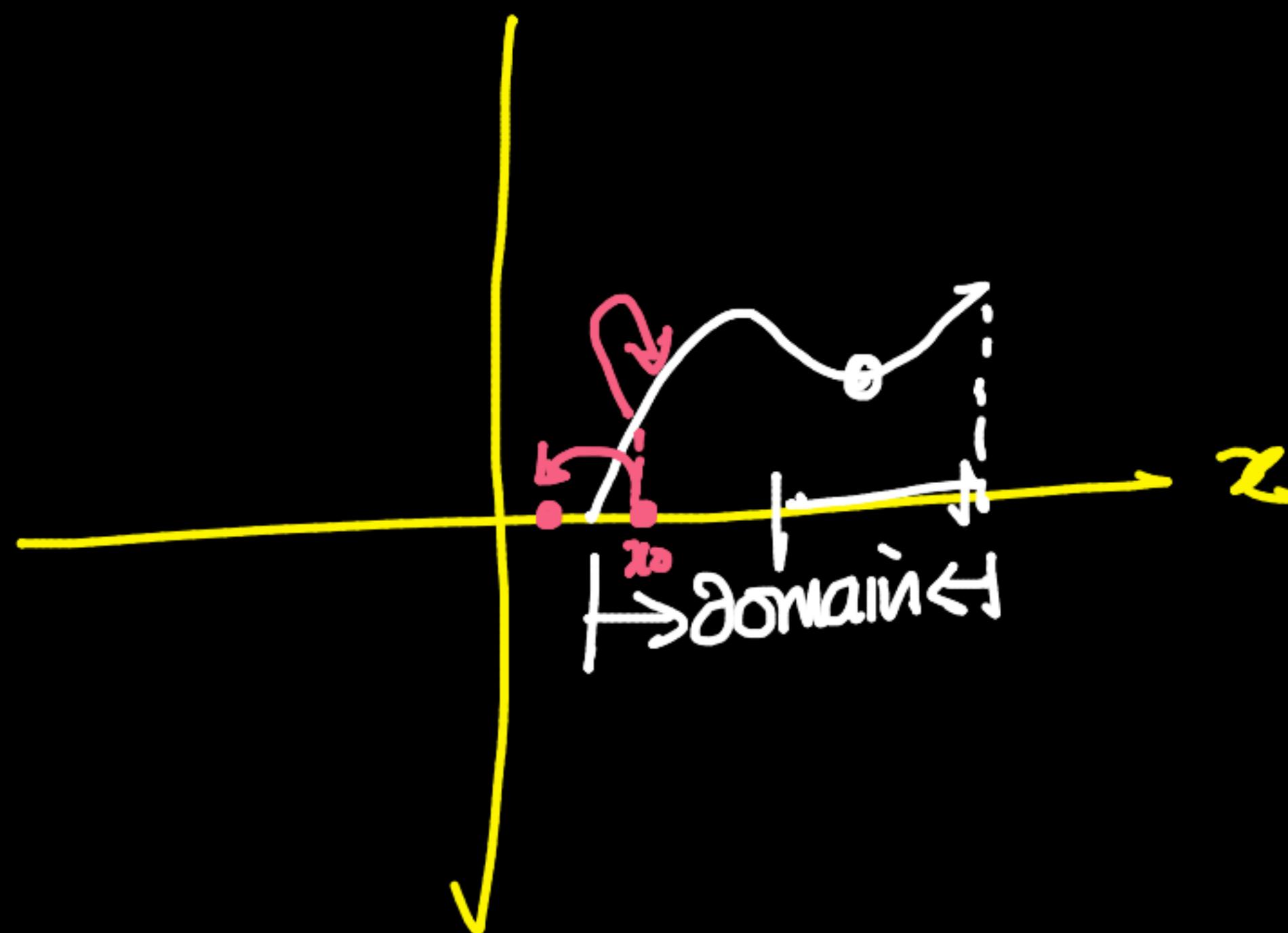
$\frac{\partial L}{\partial w_j}$

all data pls

hack (comp. tech): s





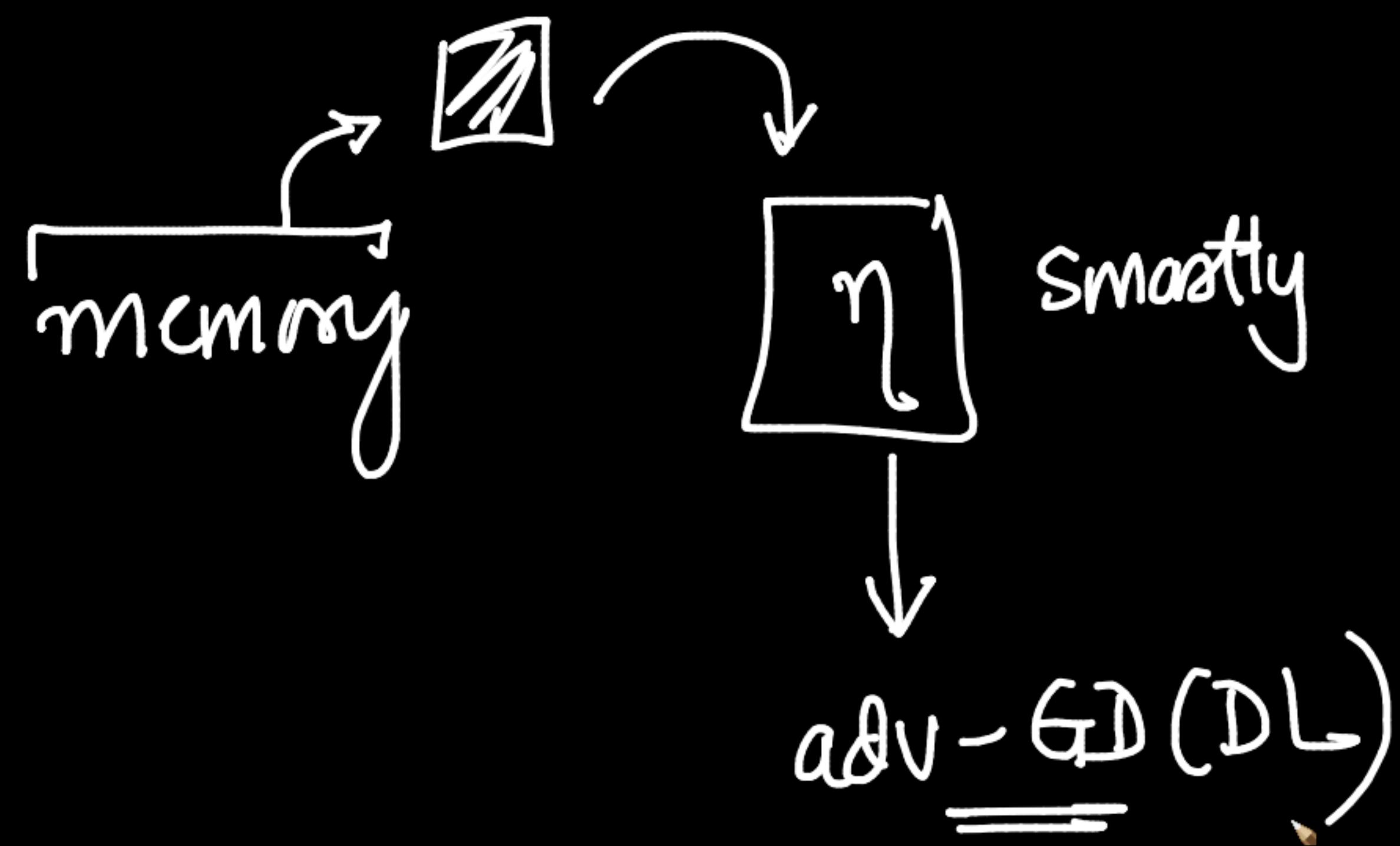


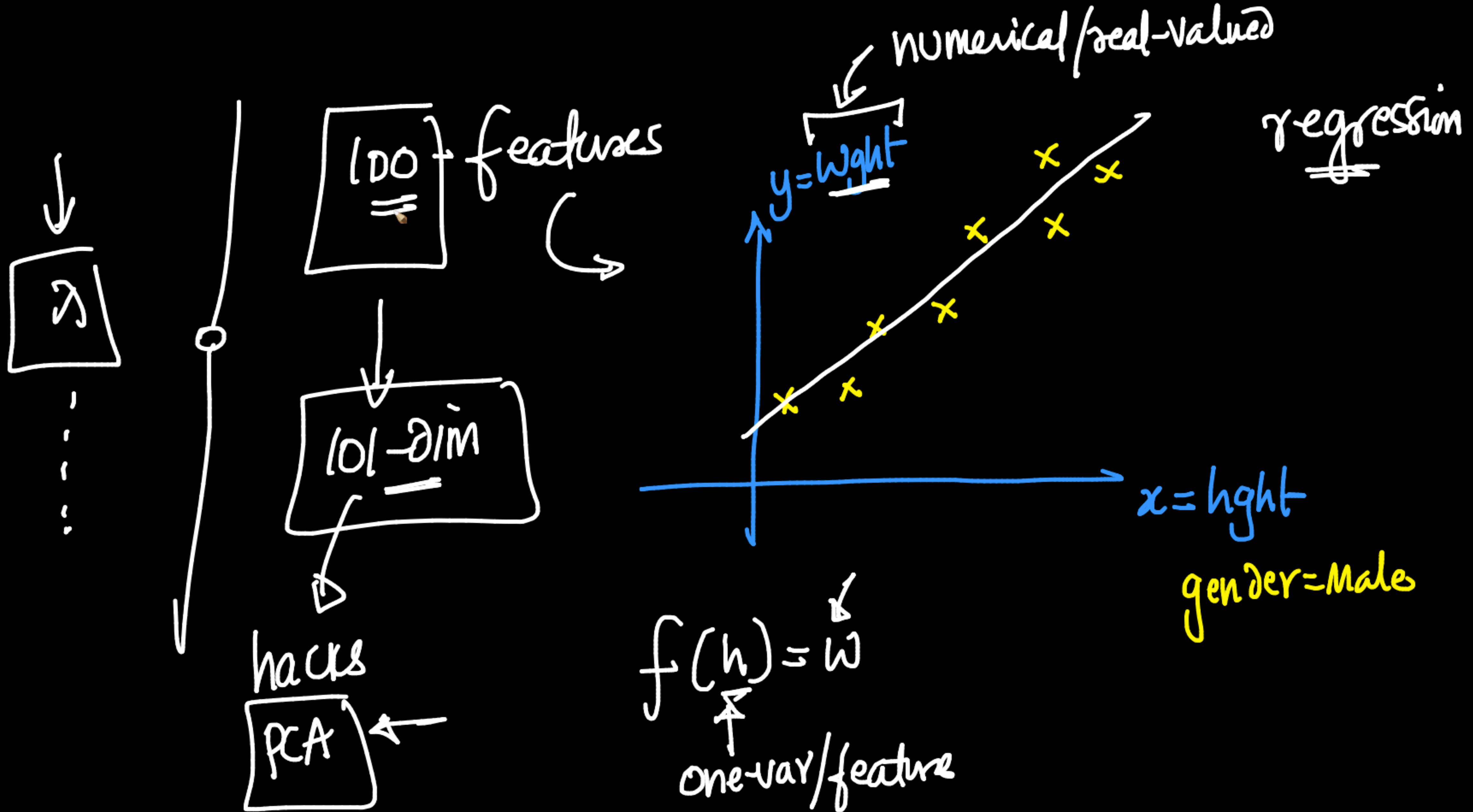
$$\downarrow \\ x_1 = x_0 - \sqrt{\frac{\partial f}{\partial x}} \Big|_{x_0}$$

{ boundary case:

if  $x_1$  outside  
domain

↳  $x_1: \underline{random} \checkmark$





(Q)  $f(x) = \frac{1}{x}$

$$\lim_{x \rightarrow 0} f(x)$$

$$\frac{1}{x}$$

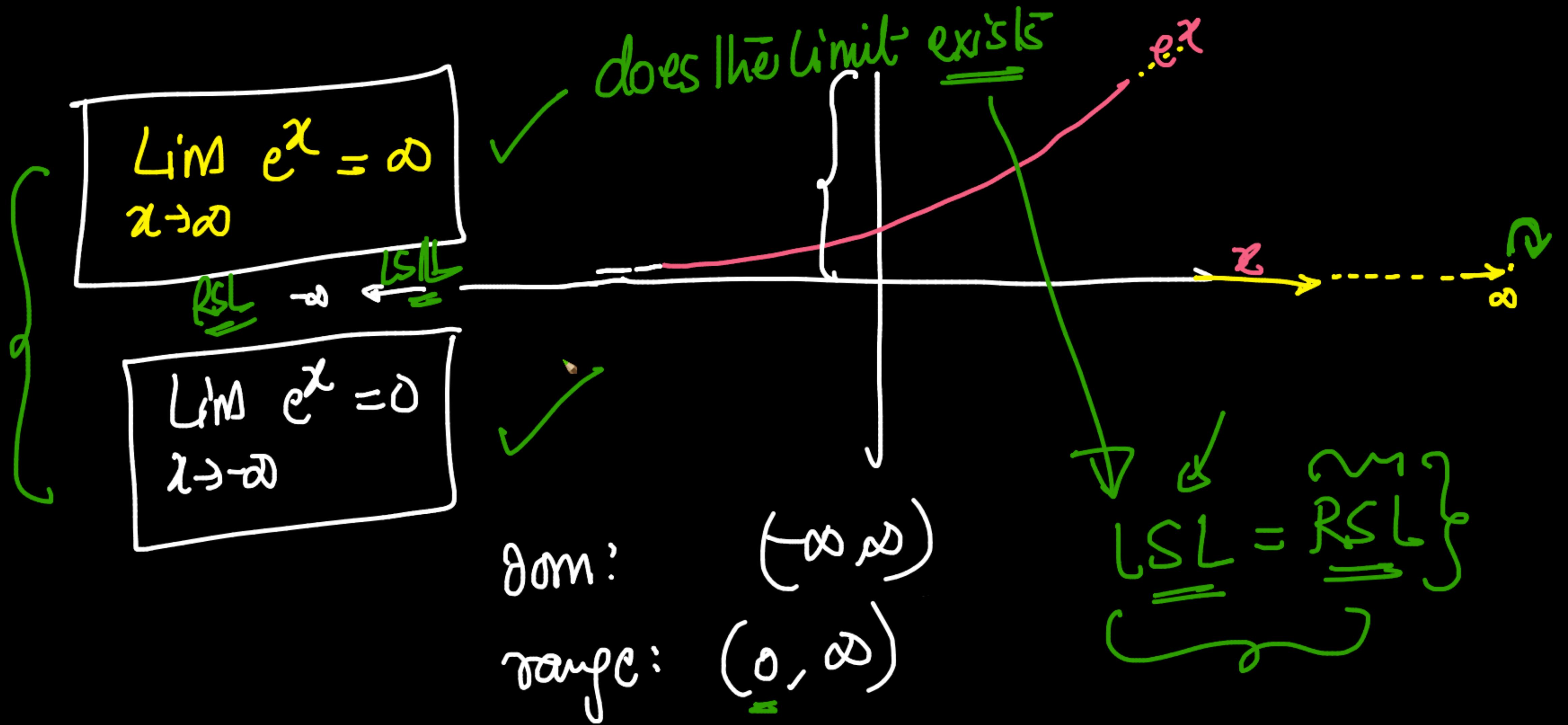
Notation {  $\lim_{x \rightarrow 1} f(x) = 1$

2-sided

$$\lim_{x \rightarrow 1^+} f(x) = 1$$

$$\lim_{x \rightarrow 1^-} f(x)$$





features: state, hght =  $\langle s_1, s_2 \dots s_{30}, \text{hght} \rangle$

output-var: wght



{ numeric  $\rightarrow$  reg.  
Categorical  $\rightarrow$  Classfr

