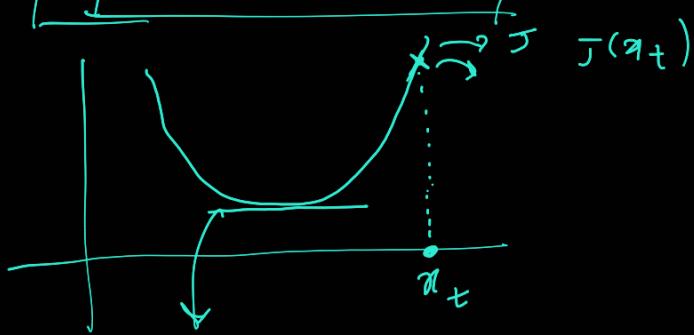
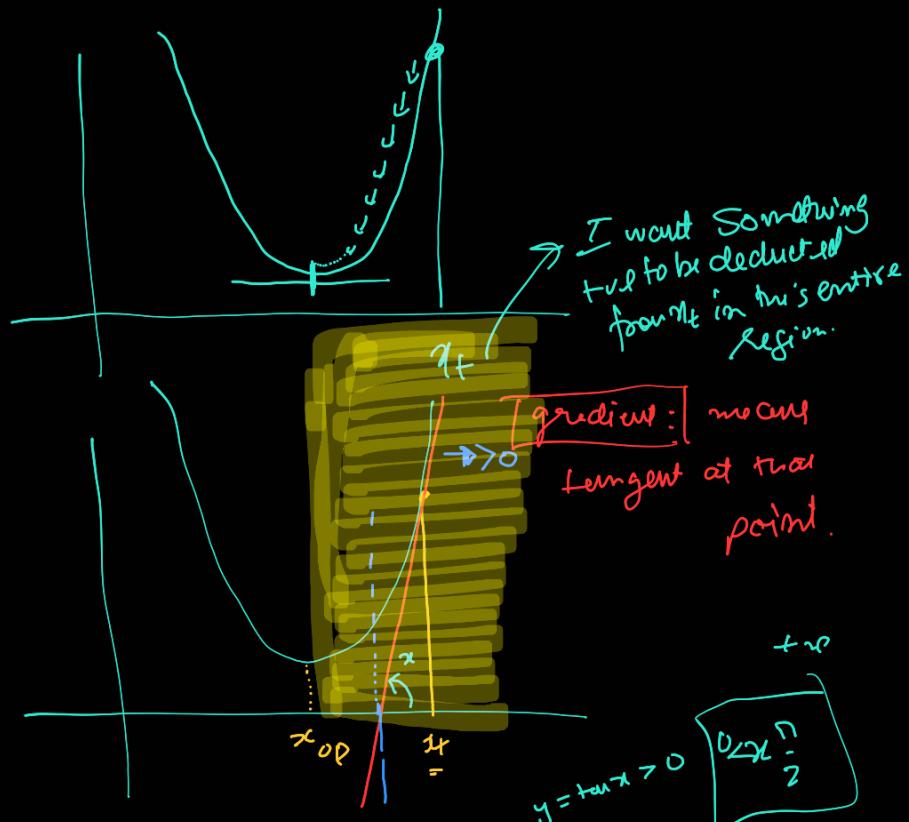


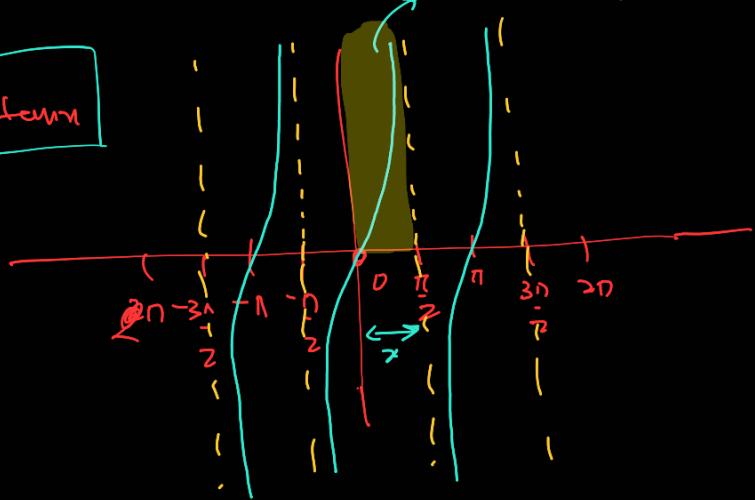
Gradient Descent



My goal is to reach optima.



$y = \text{tanh}$



$$x_{0p} < x_t$$

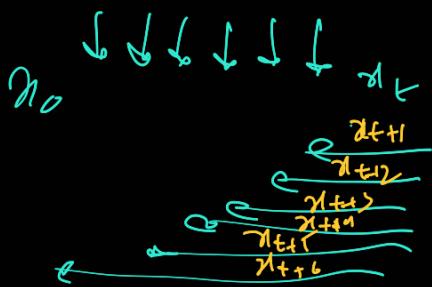
$$\left[x_t - \left(\boxed{\quad} \right) = x_0 \right]$$

+ve
or
-ve

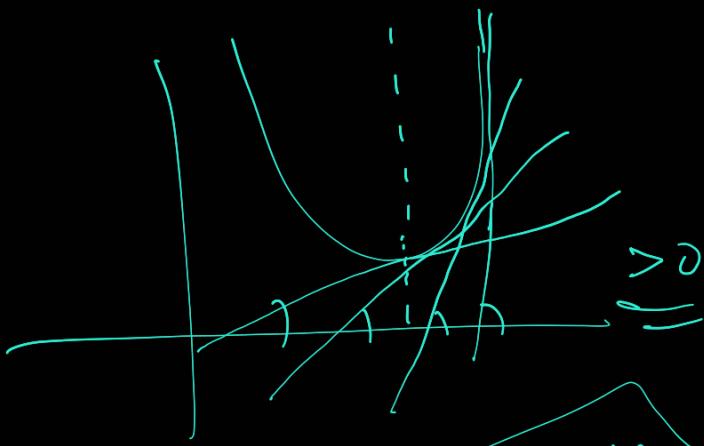
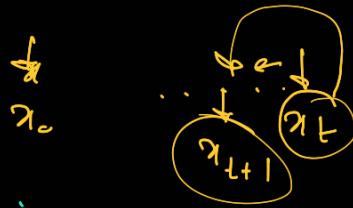
\downarrow
 +
 f_{true}
 x_t

I need to subtract some stuff from x_t so
that I reach x_{optimal} .

The catch is this, that, I want to
reach optimal in a series of steps.
(iteratively)



$\rightarrow x_t - (f_{\text{true}}) = x_{t+1}$



$$x_{t+1} = x_t - (f_{\text{true}})$$

step size
learning rate

$$x_{t+1} = x_t - \eta_t \cdot \nabla f(x_t) \quad (< 1)$$



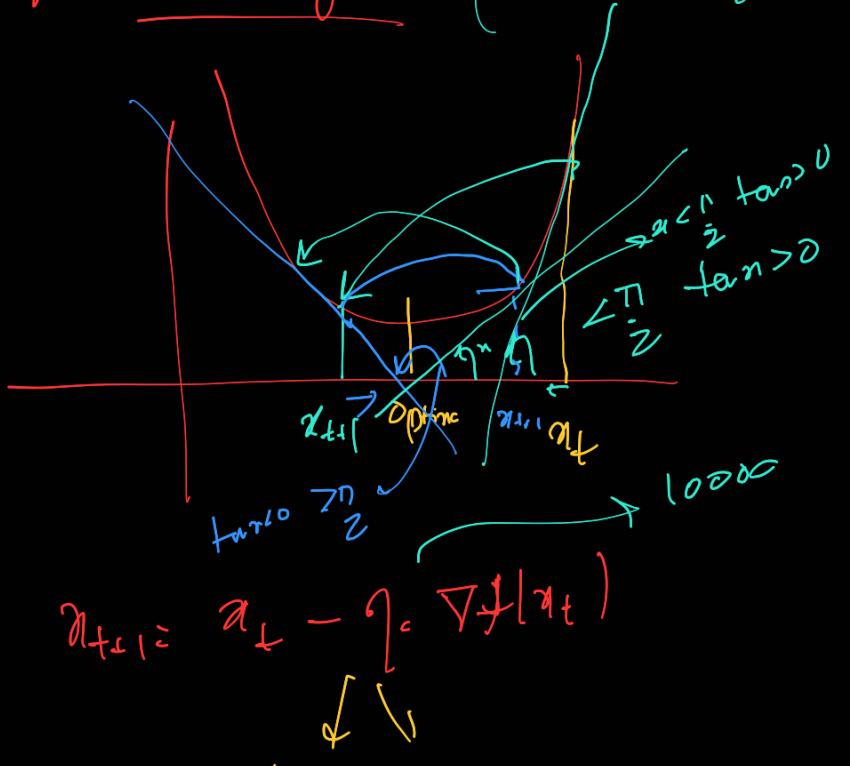


$$\begin{aligned} x_{t+1} &= x_t - \eta \cdot \nabla f(x_t) \\ \theta_j^{(t+1)} &\leftarrow \theta_j^{(t)} - \eta \cdot \nabla_j T(\theta) \Big|_{\theta_t} \end{aligned}$$

In this case
 $\eta_b > \eta_1$

$$\begin{aligned} x_{t+1} &= x_t - (-ve) \\ \theta_j^{(t+1)} &\leftarrow \theta_j^{(t)} - \eta \cdot \nabla_j T(\theta) \Big|_{\theta^{(t)}} \end{aligned}$$

A note on learning rate. (Overshooting minⁿ)



$$x_{t+1} = x_t - \eta \cdot \nabla f(x_t)$$

↓ ↓
 low high

$$x_{t+1} = x_t - []$$

We tend to keep our learning rate low,
so that we do not overshoot "the min"

When to use NEQns > When to use
GD vs Newton's Method.

$$\text{① } \theta = (X^T X)^{-1} X^T Y \quad | \text{ NE}$$

$$\text{② } \theta_j^{(t+1)} \leftarrow \theta_j^{(t)} - \left. \nabla J(\theta) \right|_{\theta^{(t)}} \quad | \text{ GD}$$

$$\text{③ } \theta_j^{(t+1)} \leftarrow \theta_j^{(t)} - \left. (H)^{-1} \nabla J(\theta) \right|_{\theta^{(t)}} \quad | \text{ NM}$$

In NE -> we have inverse.

"Computationally expensive" \Rightarrow $\nabla J = mn$
"Garner" is also difficult.
 \downarrow
Pseudo-inv

$$O(n^3)$$

$$\approx \sqrt{mn}$$

$$m = 1000$$

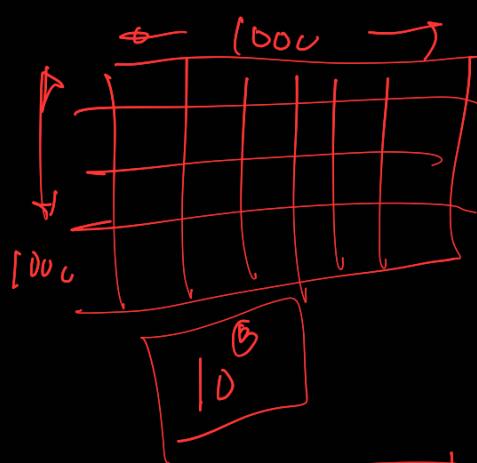
$$O(N^3)$$

$$n = 1000$$

$$\Rightarrow mn = 1000 \times 1000 = 10^6$$

$$O((1^b)^3) = O(1^b)$$

entries in
the matrix



GD

SGD

> Batch

$$Q_j^{(t+1)} \leftarrow Q_j - \eta \nabla J(\phi)$$

~~It is very fast~~

~~But involves second order term~~

~~FP~~

~~GRAD~~
 $\nabla J(\phi)$

\rightarrow Newton

Feynman's Learning Method

Learn in 3 stages (passes)

1. Pass 1 → Just go through the material w/o thinking about it, that what it is trying to say. Some things will go inside head. (20.1.)

2. Pass 2 : Now draw the connections

2. Pass 2: Now pay attention, be more
carefully this time. Try to fill in gaps
from prev pass. Now you will have
proper context, and make some improvements
on context from Pass 1.

3. Pass 3: Now pay most attention,
and try to see holistic picture, relating
everything.

→ DRAFT (Version 4)
→ Well focused

Explain to some "novice"