$Gradient\ Descent\ algorithm$

$Gradients:\ difference$

$x_1\ and\ \ x_2\ = x_2 - x_1 =\ \Delta x\ =\ the\ change\ in\ x$

$y_1\ and\ \ y_2\ = y_2 - y_1 =\ \Delta y\ =\ the\ change\ in\ y$

$(x_1, x_2)\ and\ \left(y_1, y_2\right)\ \ the\ slope:$

$slope\ = \frac{y_2 - y_1}{x_2 - x_1} = \frac{\Delta y}{\Delta x} = \frac{the\ rate\ of\ change\ in\ y}{the\ rate\ of\ change\ in\ x} = \frac{dy}{dx} = m\ =\ tan(\theta)$

$tan(0) = 0\ \ No\ slope$

$tan(45) = 1\ \ Littile\ slope$

$tan(90) = inf\ \ \ max\ slope\ ,\ imagine\ you\ have\ a\ bike\ \ you\ cant\ drive\ vertical$

$Imagine\ you\ are\ trying\ to\ find\ the\ minimum\ point\ of\ \ y\ = x^2$

$y\ = x^2\ \ \ graph$

$I\ assume\ that\ \ at\ \ x\ =\ 4\ ,\ y\ = x^2\ \ might\ have\ minimum\ point$

$No\ it\ is\ wrong$

$why\ ?:\ \ \ My\ omkar\ sir\ told\ if\ \ any\ equation\ has\ minmum\ point\ ,$

$\quad\ the\ slop\ of\ the\ equation\ at\ particular\ point\ =\ 0$

$y\ = x^2\ \ slope$

$\frac{dy}{dx} = y^{'}\ = 2x\ \ = 2\ *\ 4\ =\ 8$

$jr\ bikas\ understood\ \ x = 4\ \ is\ not\ a\ minimum\ point$

$x = 4\ +\ \ or\ \ 4\ -\ \ \ ====\ \ > uncle\ zubair$

$For\ the\ first\ assumption:\ \ y = x^2\ \ slope$

$\frac{dy}{dx} = y^{'} = 2x\ = 2 * 4 = 8\ \ \ (+\ ve)$

$x_{new} = x_{old}\ -$

$Goal:\ \ Need\ to\ find\ minmum\ point$

$Rule:\ \ at\ minmum\ point\ slope\ of\ the\ equation\ = 0$

$Case\ -\ 1:\ \ \ assume\ \ x = 4\ \ is\ Minimu\ point$

$It\ is\ wrong:\ \ \ y^{'} = 2x = 2 * 4 = 8\ !\ = 0$

$Case\ -\ 2:\ \ x_{new} = x_{old}\ +\ \ or\ \ -$

$In\ case\ -\ 1\ \ slope\ value\ \ is\ (+\ ve)\ \ decrease\ the\ values$

$slope\ value\ \ is\ (-\ ve)\ \ increase\ the\ value$

$Case\ -\ 3:$

$x_{new} = 4\ -\ \ slope\ value$

$x_{new} = 4\ -\ 0.2\ *\ 8 = 4\ -\ 1.6 = 2.4$

$slope = y^{'} = 2 * x = 2 * 2.4 = 4.8$

$x_{new} = x_{old}\ -\ \ 0.2\ *\ slope$

$x_{new} = 2.4\ -\ 0.2\ *\ 4.8 = 2.4\ -\ 0.96 = 1.44$

$slope = y^{'} = 2 * x = 2 * 1.44 = 2.88$

$x_{new} = 1.44 - 0.2 * 2.88 = 0.864$

$4, \; ==== \; > 2.4 === \; > 1.44 === \; > 0.8 ===== \; > 0$

$x_{new} = x_{old}$

$x_{new} = x_{old} - 0.2 * slope$

$Iteration - 1: \quad x = 4$

$$calculate\ slope\ at\ this\ point = y' = 2 * x = 2 * 4 = 8$$

$x_{new} = 4 - 0.2 * 8 = 4 - 1.6 = 2.4$

$Iteration - 2: \quad x = 2.4$

$$calculate\ slope\ at\ this\ point = y' = 2 * 2.4 = 4.8$$

$x_{new} = 2.4 - 0.2 * 4.8 = 2.4 - 0.96 = 1.44$

$Iteration - 2: \quad x = 1.44$

$$calculate\ slope\ at\ this\ point = y' = 2 * 1.44 = 2.88$$

$x_{new} = 1.44 - 0.2 * 2.88 = 1.44 - 0.576 = 0.86$

$x_{new} = x_{old} - learning\ rate * slope$

$\alpha = learning\ rate = (0, 1)$

$Neural\ network: \; Will\ find\ the\ error$

$Goal: \; Need\ to\ Minimise\ the\ cost\ function(J)\ by\ providing\ the\ suitable\ weights$

$$Minimise\ the\ cost\ function = slope\ of\ the\ cost\ function$$

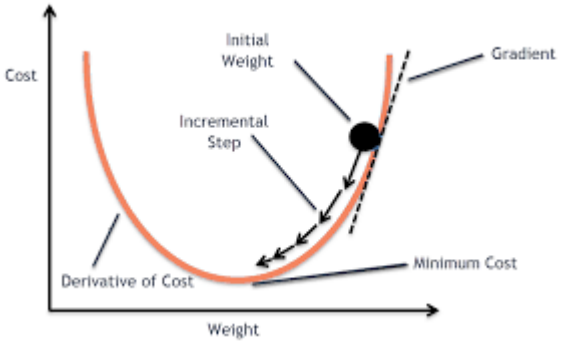
$$\frac{dJ}{dw} = 0$$

$$w_{new} = w_{old} - \ learning\ rate * \frac{dJ}{dw_{at=w_{old}}}$$

$$b_{new} = b_{old} - \ learning\ rate * \frac{dJ}{db_{at=b_{old}}}$$

$$w_{new} = w_{old} - \ \alpha * \frac{dJ}{dw_{at=w_{old}}}$$

$$\frac{dJ}{dw_{at=w_{old}}} = 0 \ then \ w_{new} = w_{old}$$



Sir to what we call learning rate i didn't get network issue plz sir just explain again

Learning rate :

$$w_{new} = w_{old} - \ 0.001 * \frac{dJ}{dw_{at=w_{old}}}$$

Gradient descent algorithm

repeat until convergence {

$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$$

(for $j = 1$ and $j = 0$)

}

Convergence: *reach the minmum point*

1) *Batch Gradient descent*

Imagine you have an image size is $28x28$

$H1$: $128N$

$H2$: $128N$

$H3$: $128N$

O: $10N$

$I\ (784) - H1(128) - H2(128) - H3(128) - O(10N)$: *Output*

Calculate the error

Go back by update the weights

Forward Prpogation + *Backward prpogation* : $1Min$

100 *Images are there* ===== *at a time*

Im asking to 100 *students, go down and come up* :

1 *training iteration is completed* = *epoch*

100 *members go down and come up* : 1 *epoch* : *accuracy mean square error*

I will get error afer 100 *mins*

I will get a chance to do next iteration after 100 *mins*

I will get a chance to update my weights after 100*mins*

Analogy: *pass* X_{test} *to* $DT.predict(X_{test})$

one epoch will not give desired answer:

starting we intilaise weights randomly

at least 10 *epochs we required*

For the entire process: $100 * 10 = 1000min$

2) Mini batch Grdaient descent

100 *members divided into* 10 *batches*

Mini batch Gradient descent

first 10 *mebers will go* : $F + B$

will calculate 10 *members avg error, we try to update the weights*

after 10 *mins only, we started to update the weights*

Next bactch 10 *members use the new updated weights*

will calculate the error of 10 *m, update the weights*

after 10 *batch Forward* + *Back* = 1 *epoch*

3159/3159
1/3159
2/3159
3159/3159

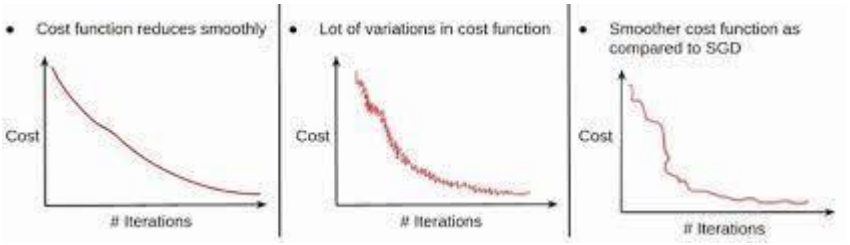
3*rd batch pepole*

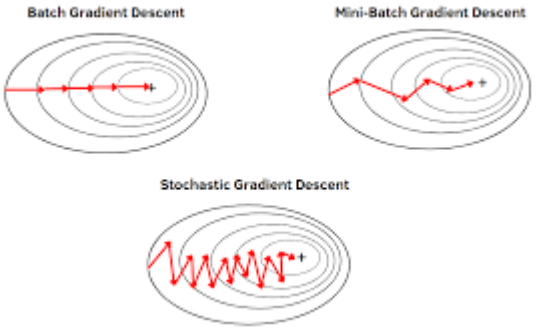
3) Stochastic gradient descent:

every time one input will go and will calculate the error and it come back by update the weights

Error Graphs

For every epoch error will be decrease





Saturday 9 to 10 exam : Statistics

10 to 1 : NLP class

10 to 1 : NLP class

Exponential weighted averages