



NOTE: Implementing these Algorithms is dead easy but understanding them is a bit challenging

## Regression Problems - Linear Regression

A regression problem is when you have one or multiple independent variables and you try to predict one or more dependent variable. The dependent value is a continuous or real value.

Examples of Regression problems :

- 1) Predicting a person's age
- 2) Predicting stock prices
- 3) Predicting price of something

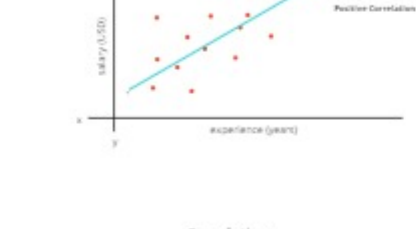
Target Variable or Dependent Variable

Independent Variables

	Age	Experience(yr)	Job	Salary
1)	24	3.5	Software engineer	\$1500
2)	28	5.6	Data scientist	\$6700
3)	35	10.1	ML engineer	125600

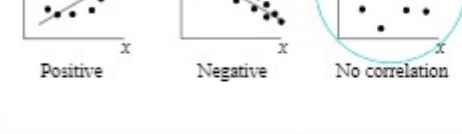
NOTE: in regression, Target Variable is a Continuous value

## Scatterplot in 2D



## Correlation

We plot scatterplots to find the correlation between variables. Correlation is the measure of how the values of variables change in relation to each other



NOTE: For regression problems, it is good if the Independent & Dependent Variable is "correlated".

NOTE: It is compulsory that the Independent and Dependent Variables are Correlated when using Linear Regression.

## Linear Regression

It is the simplest Machine Learning Algorithms for solving Regression problems

Simple Linear Regression (1 Independent & 1 Dependent Variable)

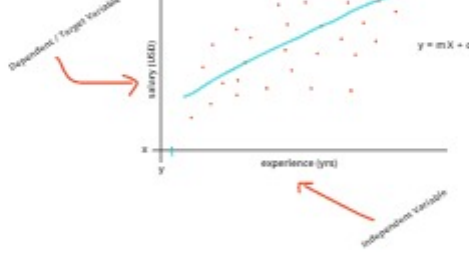
Multiple Linear Regression (Multiple Independent & Dependent Variables)

## Simple Linear Regression

(Data of some employees)

	Age	Experience(yr)	Job	Salary
1)	24	3.5	Software engineer	\$1500
2)	28	5.6	Data scientist	\$6700
3)	35	10.1	ML engineer	125600

predicted value 41500



## Equation of "Line of Best Fit"

Dependent Variable

$$y = mx + c$$

Y = M (x) + C

Y = (0.4) (5) + 10

Slope of line

Independent Variable

Intercept or Bias

NOTE: Our learning parameters are "M" and "C"

## COST FUNCTION or LOSS FUNCTION

It's a method of evaluating how well your algorithm models your dataset. If your model is predicting all values correctly then the Loss will be a lower number, whereas if your model is predicting values incorrectly, the Loss will be higher. As you change pieces of your algorithm to try and improve your model, your loss function will tell you if you're getting anywhere.

NOTE: We have different Cost/Loss functions depending on your problem.

- Binary Cross-Entropy Loss
- Hinge Loss
- Mean Square Error
- Huber Loss

NOTE: Cost functions output a single number representing how well our model is predicting values. Lower cost function is always preferable.

THE GOAL OF TRAINING ANY MODEL IS TO LOWER THE LOSS FUNCTION VALUE.

For Regression Problems the most common LOSS function is "MSE"

## Mean Squared Error

$$mse = \frac{1}{n} \sum_{i=1}^n (y_i - y_{predicted})^2$$

## EXAMPLE of MSE

	Y1	Y
	True Values	Predicted Values
1)	3.5	1.5
2)	4.2	3.5
3)	1.8	2.5
4)	0.5	0.9

$$MSE \text{ LOSS} = \frac{(3.5 - 1.5)^2 + (4.2 - 3.5)^2 + (1.8 - 2.5)^2 + (0.5 - 0.9)^2}{4}$$

$$MSE \text{ LOSS} = \frac{4 + 0.49 + 0.49 + 0.16}{4}$$

$$MSE \text{ LOSS} = 1.285$$

NOTE: We Optimize our Machine Learning Algorithm to Lower the Cost

Now we know what loss function is, i.e we know which "Line of best fit" or which set of parameter values for (m & b) would be good, but how do we choose the most optimal parameter values?

For example, among a set of values for M & B, we can use the LOSS function to find which parameters are the best for our model. But how do we find these "set of values" for our parameters?

This is where Gradient Descent Comes in...

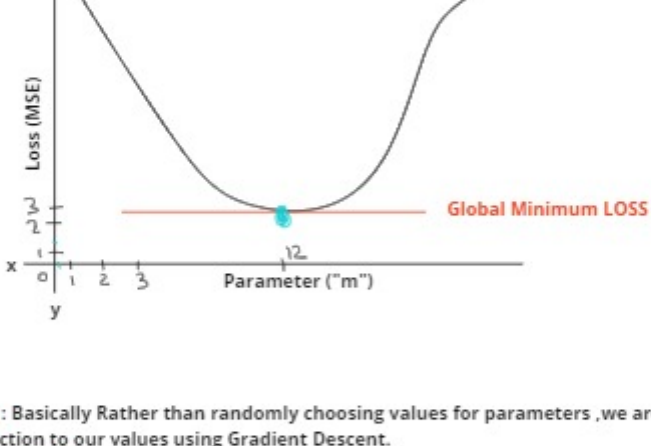
## GRADIENT DESCENT

It is an optimization algorithm which is commonly-used to train machine learning models and neural networks. Training data helps these models learn over time, and the cost function within gradient descent specifically acts as a barometer, gauging its accuracy with each iteration of parameter updates.

In Simple words, "Gradient Descent" helps us find the most optimal parameter values which give the lowest LOSS function value.

## How Gradient Descent Works?

- Step 1] Set all the parameters to 0 at start
- Step 2] Calculate the LOSS
- Step 3] Update the parameter using "UPDATE FORMULA"
- Step 4] Repeat Step 2 and 3 until you reach a Global Minimum Loss or Until your Loss is decreasing.



NOTE: Basically Rather than randomly choosing values for parameters, we are giving a direction to our values using Gradient Descent.

## PARAMETER UPDATE FORMULA

$$w_{t+1} = w_t - \alpha \frac{\partial L}{\partial w_t}$$

New Value for Parameter

Old Value of Parameter

Learning Rate

Derivative of LOSS function w.r.t Parameter

Learning rate is a constant which decides how fast we want the algorithm to learn. A common value for learning rate is 0.01

Example, Lets say the current value of "m" is 0 and LR is 0.01, and also the derivative of "m" is -24 then using the UPDATE FORMULA its value will be:

$$m = m - (0.01) * (-24)$$

$$m = 0 + 0.24$$

$$m = 0.24$$

Basically, Its very simple, we have a LOSS FUNCTION and we try to reduce it by trying a large number of different parameter values.

## Implementing Linear Regression in Python

NOTE: Dont try to understand or implement these ML Algorithms from scratch, Not unless you really need to.