

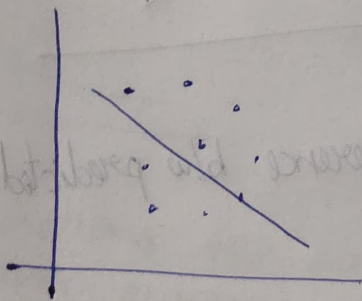
Machine Learning

Machine learning is a program that analyses data and learns to predict the outcome.

Dataset:- In the mind of a computer, a data set is any collection of data. It can be anything from an array to a complete database.

Regression:- It is used when you try to find relationship b/w variables, that relationship is used to predict the outcome of future events.

Linear regression:- It is a statistical technique used to find the relationship between variables to draw a straight line through them.



In algebraic terms, the model would be defined as $y = mx + b$

In ML, we write equation for linear regression model as

$$y' = b + w_1 x_1$$

y' - the output

b - bias (y-intercept) is a parameter of model & is calculated during training.

w_1 - weight (m in algebra)

x_1 - input / feature.

• During training, the model calculates the weight and bias that produce the best model.

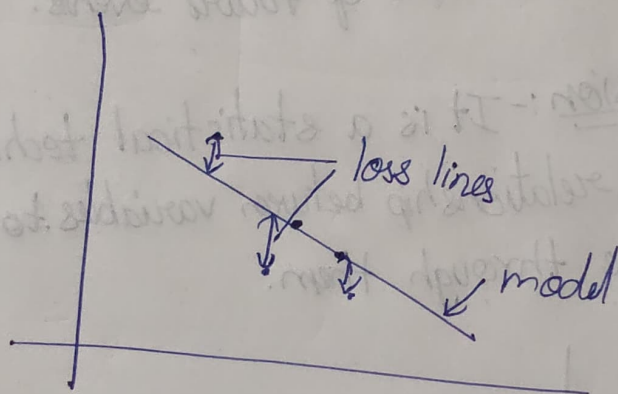
Models with multiple features:-

$$y' = b + w_1 x_1 + w_2 x_2 + w_3 x_3 + \dots + w_n x_n$$

Loss:- Loss is a numerical metric that describes how wrong a model's predictions are.

Loss measures the distance between the model's predictions & actual labels

The goal of training a model is to minimize the loss



Distance of loss is difference b/w predicted & actual value.

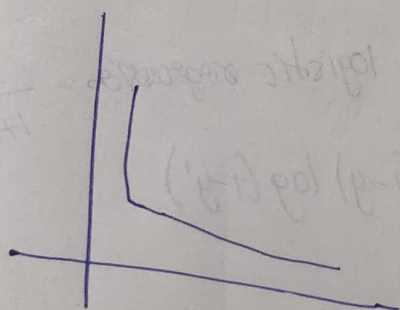
Types of loss:-

- 1) L_1 loss :- $\sum | \text{actual value} - \text{predicted value} |$
- 2) Mean absolute error :- $\frac{1}{N} \sum | \text{actual value} - \text{predicted value} |$
- 3) L_2 loss :- $\sum (\text{actual value} - \text{predicted value})^2$
- 4) Mean squared error :- $\frac{1}{N} \sum (\text{actual value} - \text{predicted value})^2$

Gradient descent is a mathematical technique that iteratively finds the weights and bias that produce the model with the lowest loss.

The model begins training with randomized weights and biases near zero and then ~~for~~ repeats the following steps:

- 1) Calculate the loss with current weight & bias
- 2) Determine the direction to move the weights & bias that reduce loss.
- 3) Move the weight & bias values a small amount in the direction that reduces loss.
- 4) return to step 1 & repeat the process until the model can't reduce the loss any further.



— Typical loss curves

The loss functions for linear models always produce convex surface

Hyperparameters:- These are variables that control different aspects of training. Three common hyperparameters are:

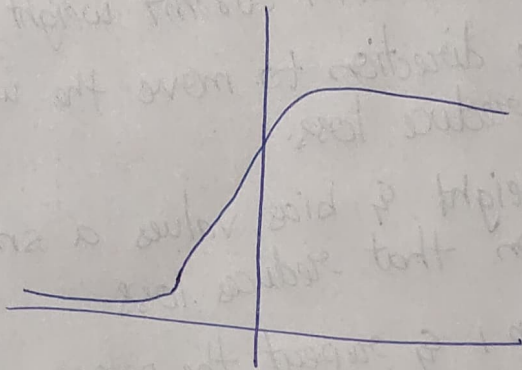
- learning rate
- batch size
- epochs.

Logistic regression:-

Logistic regression that is designed to predict the probability of a given outcome.

Sigmoid function

$$f(x) = \frac{1}{1+e^{-x}}$$



If $z = b + w_1x_1 + \dots + w_nx_n$

log loss = $\sum_{(x,y) \in D} -y \log(y') - (1-y) \log(1-y')$

logistic regression: $\frac{1}{1+e^{-z}}$