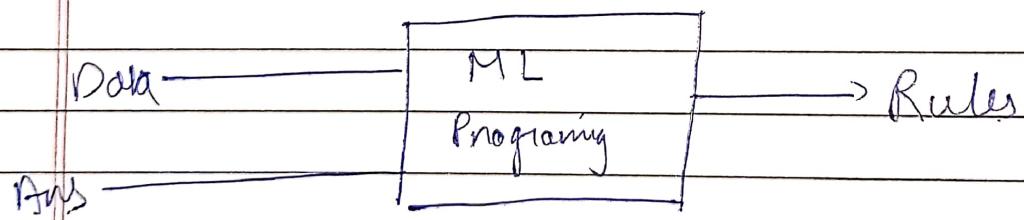
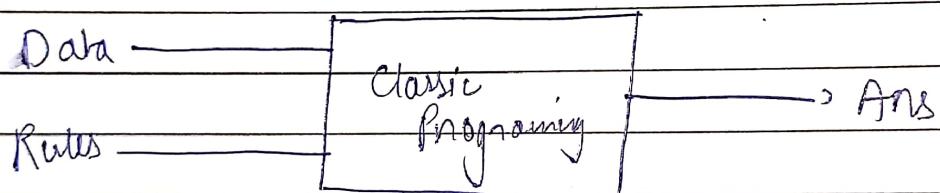
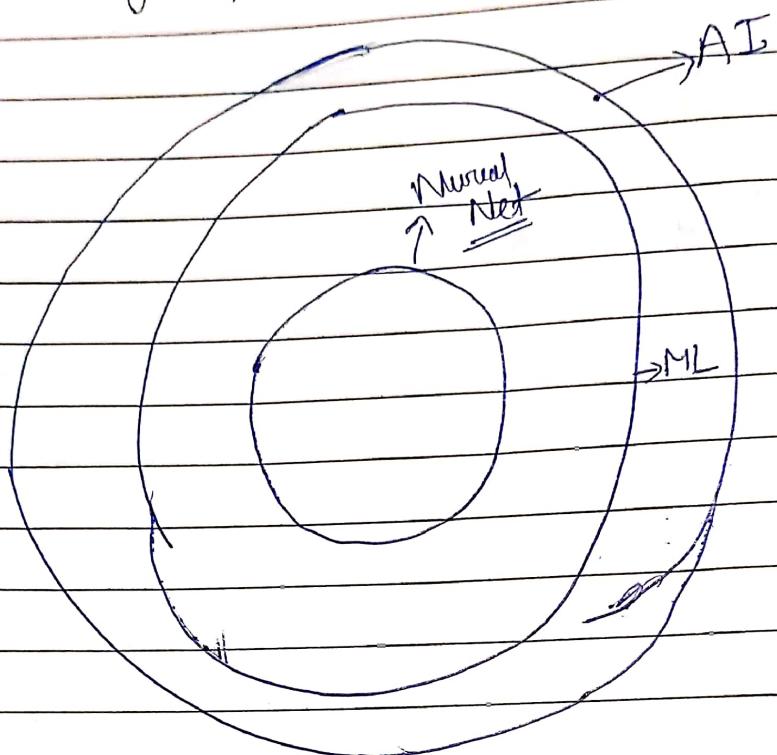


Artificial Intelligence (Personal Study)

Date
Page

Agent: Anything that can be viewed as perceiving its environment through the sensors and acting upon that environment through actuators



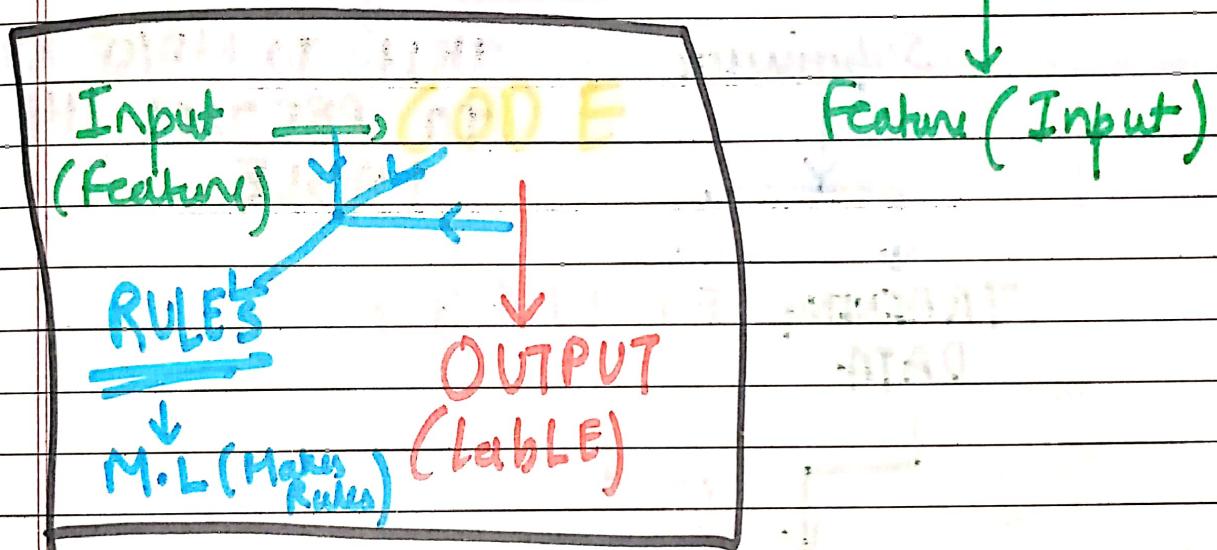
Neural Net → layered

Machine learning → figures out the rules for a given input and output

Data

Ex Table. (for detection of Mid Term grade)

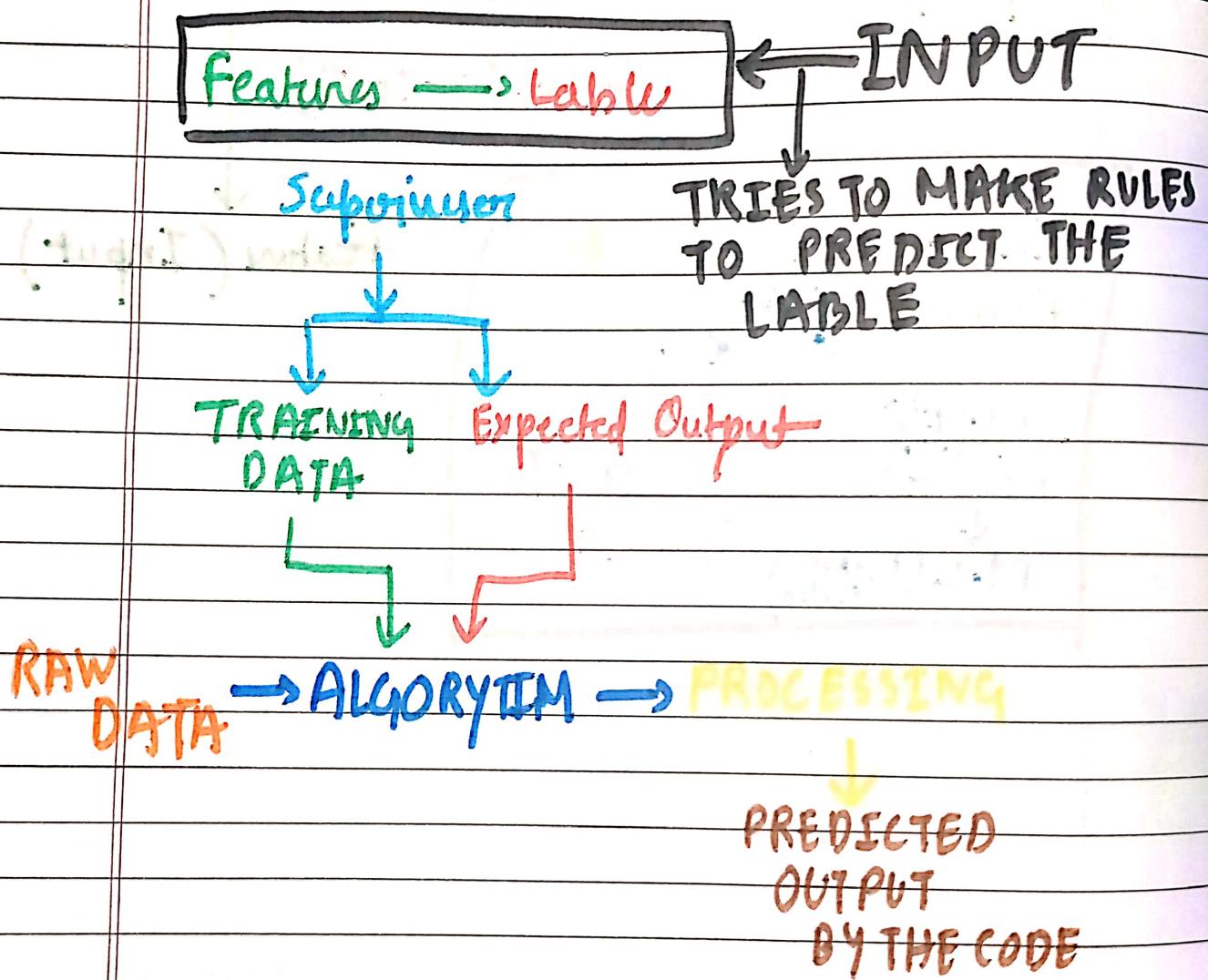
Feature	Mid form 1	Mid form 2	Final	Feature
1.	70	80	77	
2.	60	90	84	
3.	40	50	38	



Different types of M.L.

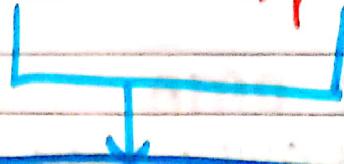
- Unsupervised learning
- Supervised learning
- Reinforcement learning

Supervised LEARNING



IF,

PREDICTED OUTPUT != Expected OUTPUT

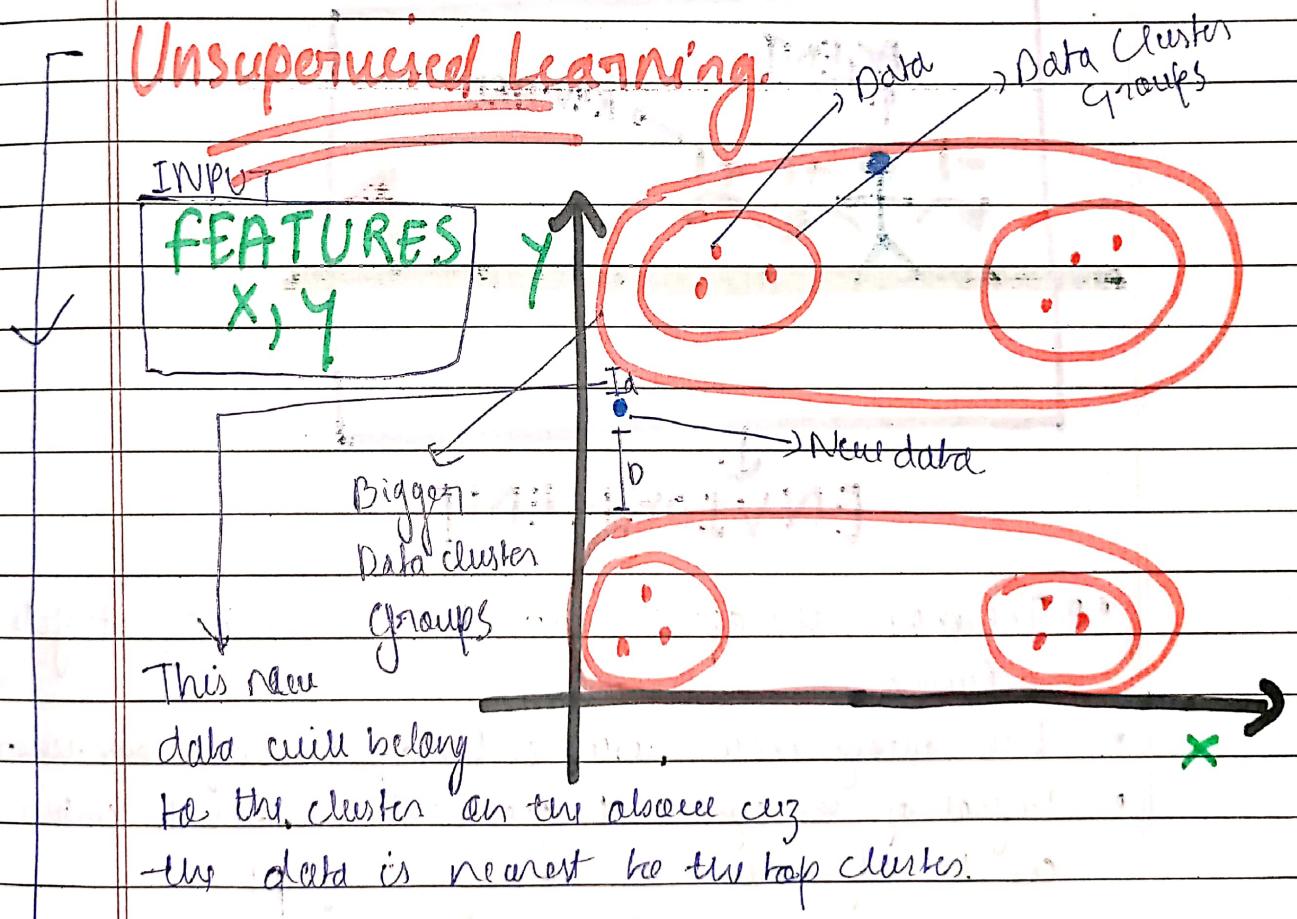


Tweaks Made by the Supervisor to get the correct output

THIS IS REPEATED UNTIL,

PREDICTED OUTPUT == Expected OUTPUT

Unsupervised Learning.

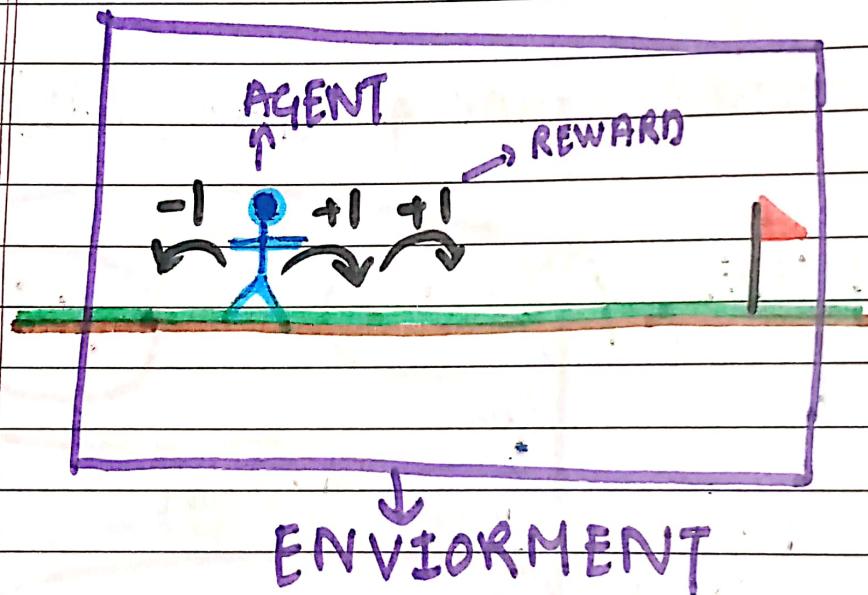


Essentially groups data and data clusters and figures out where new data will go

REINFORCEMENT LEARNING

- NO DATA
- you ONLY have a AGENT
- An Environment

For example (goal is to reach the flag)



- The more the agent moves towards the flag it gets a reward
- Each wrong move results in the agent losing some reward
- The goal of the agent is to get maximum reward which can be only done by reaching the goal.

* GOAL OF THE AGENT IS TO
GET MAXIMUM REWARD

Import Tensorflow as tf

INTRODUCTION TO TENSOR Flow

>>> pip install TENSORFlow

A collection of Data
for our ai
Making Math easy.

Tensor :- Vector generalised in higher dimensions

Data types

Represent dimensions
of the data

float32, int32, string etc.

How to create tensors

Example

String = tf.Variable("this is a string", tf.String)

Number = tf.Variable(1234, tf.int16)

Floating = tf.Variable(3.567, tf.float64)

Rank and Degree of tensor.

RANK / DEGREE

DIMENTION

1D ← rank0-tensor = tf.Variable("str", tf.string)
or scalar

2D ← rank1-tensor = tf.Variable(["Hey", "Hello"], tf.string)

3D ← rank2-tensor = tf.Variable([["test"]], tf.string)

Dimensions can be known by counting the number of nested list.

To find the Rank we can use :-

tf.rank(rank2-tensor)

OUTPUT

<tf.Tensor: Shape=(), dtype=int32, numpy=2>

Rank ←

Example = `tf.Variable([["A", "B"], ["C", "D"]])`

Shape
2D
Dimension

Shape of a tensor

- Simply no. elements in each dimension

To get shape we use:

`Example. Shape`

OUTPUT

Tensor shape `(2, 2)`

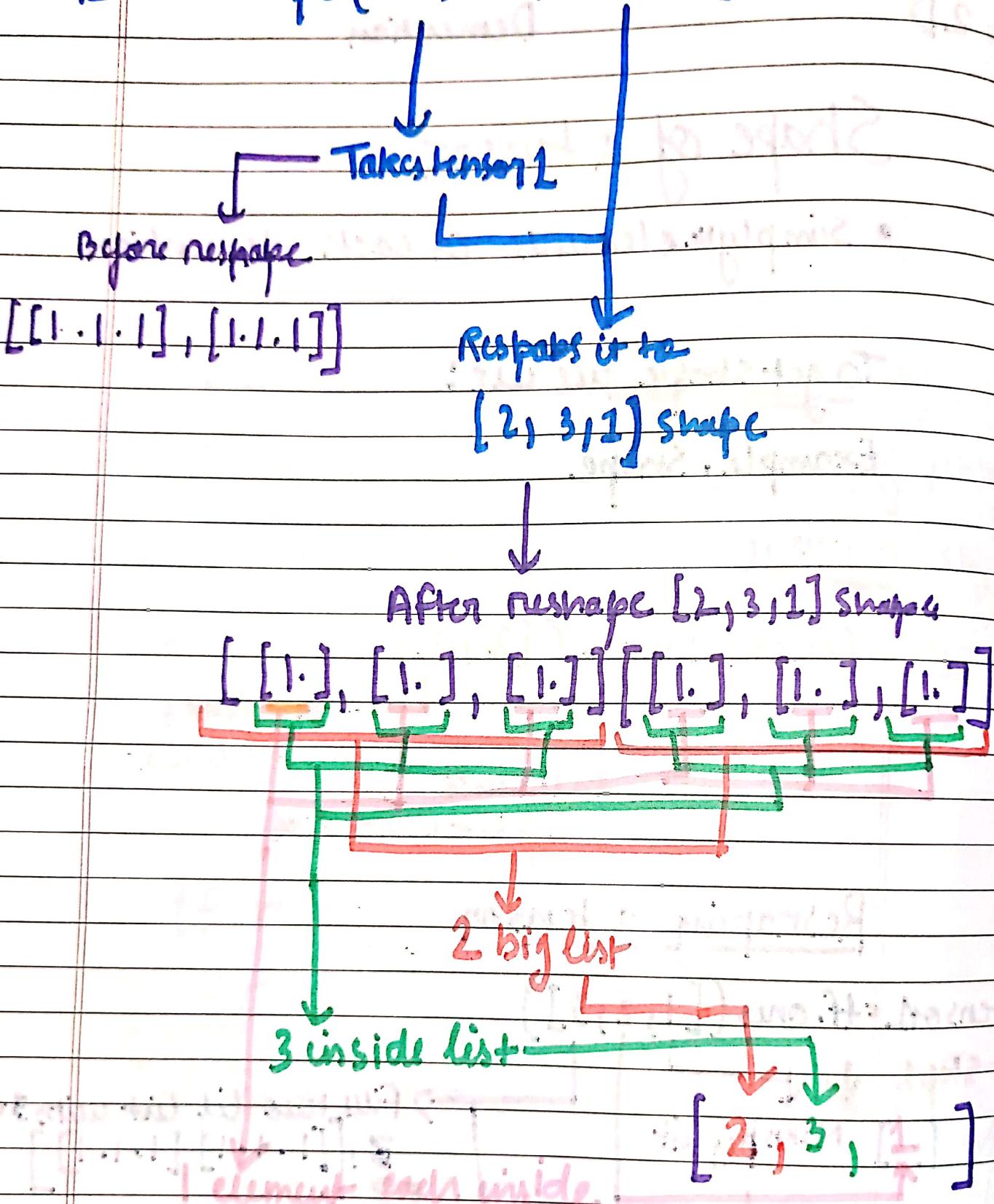
2 elements in first dimension
2 elements in 2nd dimension

Reshaping a tensor

`tensor1 = tf.ones([1, 2, 3])`

Step 1 [] Makes big list
Step 2 Makes two lit list inside big list
Step 3 Fill two lit list with 3 ones
 $3, [[1 \cdot 1 \cdot 1], [1 \cdot 1 \cdot 1]]$

tensor2 = tf.reshape(tensor1, [2, 3, 1])



No first value will be taken as 1

↑

Date _____
Page _____

tensor3 = tf.reshape(tensor2, [3, -1])

↓

- It tells tensor to calculate the data.
Dimension

↓

$$[3, -1] = [3, 3-1] = [3, 2]$$

Tensor2 before modification

$[[[1.], [1.], [1.]] [[1.], [1.], [1.]]]$

New Tensor3

$\begin{bmatrix} [1. 1.] & [1. 1.] & [1. 1.] \end{bmatrix}$

Types of Tensor:-

- Variable } \rightarrow Mutable
 - Constant
 - Placeholder } \rightarrow Immutable
 - SparseTensor

Evaluating a tensor

get value of a tensor, (uses a graph)

use next scalars to eval tensor

With `tf.Session()` as `sess`:

`tensor.eval()`

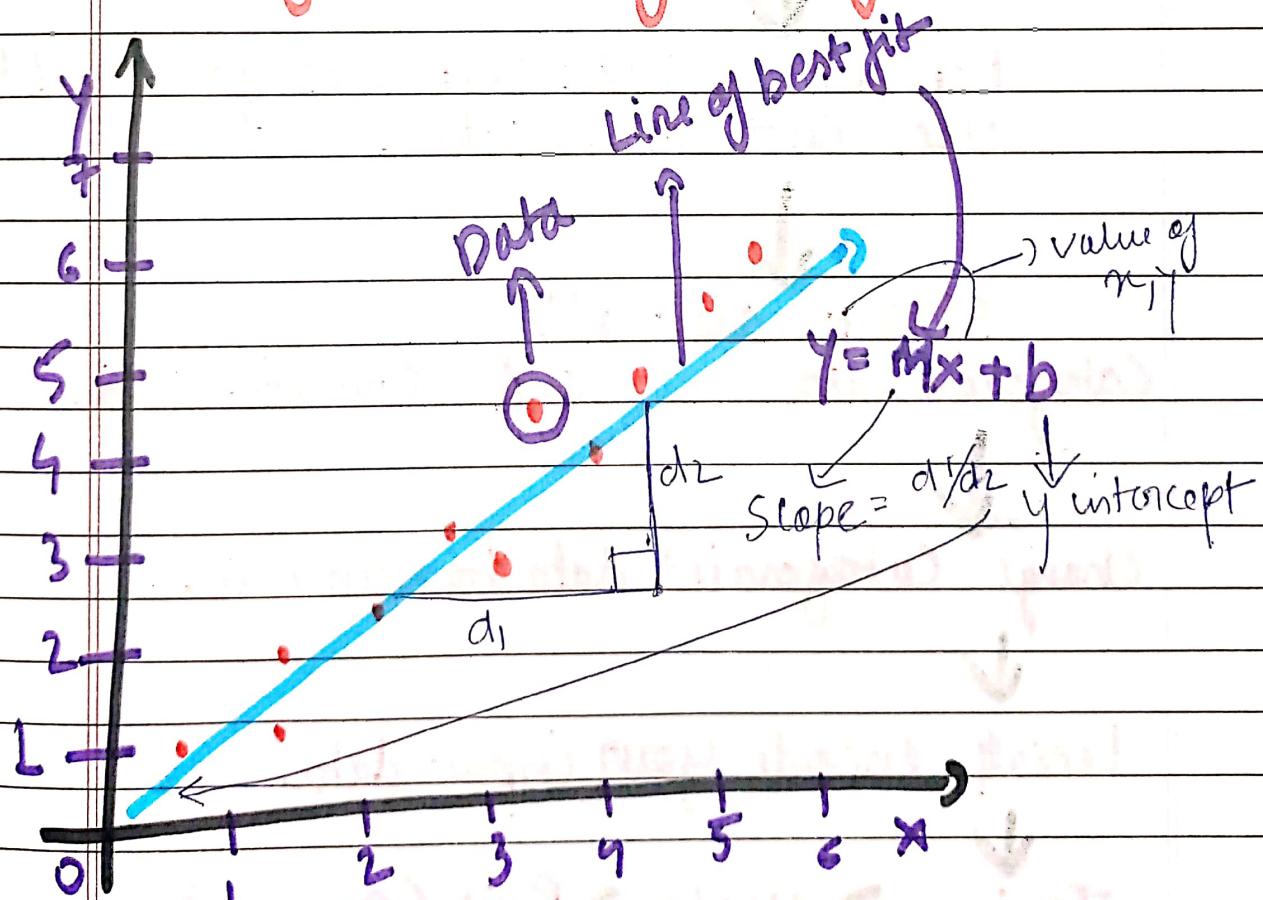
Name of tensor.

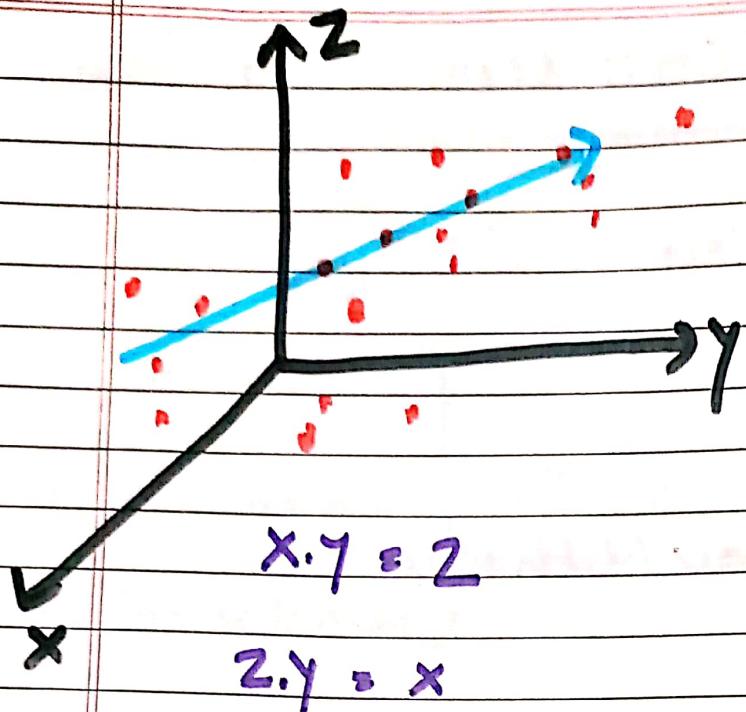
Tensorflow CORE learning algorithms

-) Linear Regression
-) Classification
-) Clustering
-) Hidden Markov Methods

Linear Regression

Takes linearly correlated Data to find the line of best fit and predict using the line of best fit





In CODE

Take a .csv (datasheet)

↓
Pop out all the elements you need to use into a variable

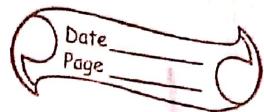
↓
Short out all Categorical Data and Numeric data?

↓
Change Categorical data to numeric.

↓
Lastly, encode your input data.

↓
Train → eval → Print (Accuracy)

CLASSIFICATION



CLASSIFIES DATA INTO DIFFERENT DATA SET (CLASSES)

< TAKE IN CODE >

→ CLASSIFIER TYPE

- DNN CLASSIFIER (DEEP NEURAL NET)
- LINEAR CLASSIFIER

Take the CSV file

↓
specify the class types

↓
pop out the columns you need.

↓
Make an input function

↓
Make feature columns

↓
Make the model

Classifier = tf.estimator.DNNClassifier(
feature_columns=feature_my_feature_columns,
hidden_units=[30, 10],
n_classes=3)

Number of
classes

Number of Nodes in 2nd layer
Number of Nodes in 1st layer
of Neural Net



Train Model

`classifier.train(`

`input_fn = lambda: input_fn(train, train_y),`
`Training = True, steps = 5000)`

Number of times
to be runned

`lambda` defines a single line `map` function
in python



`equal`

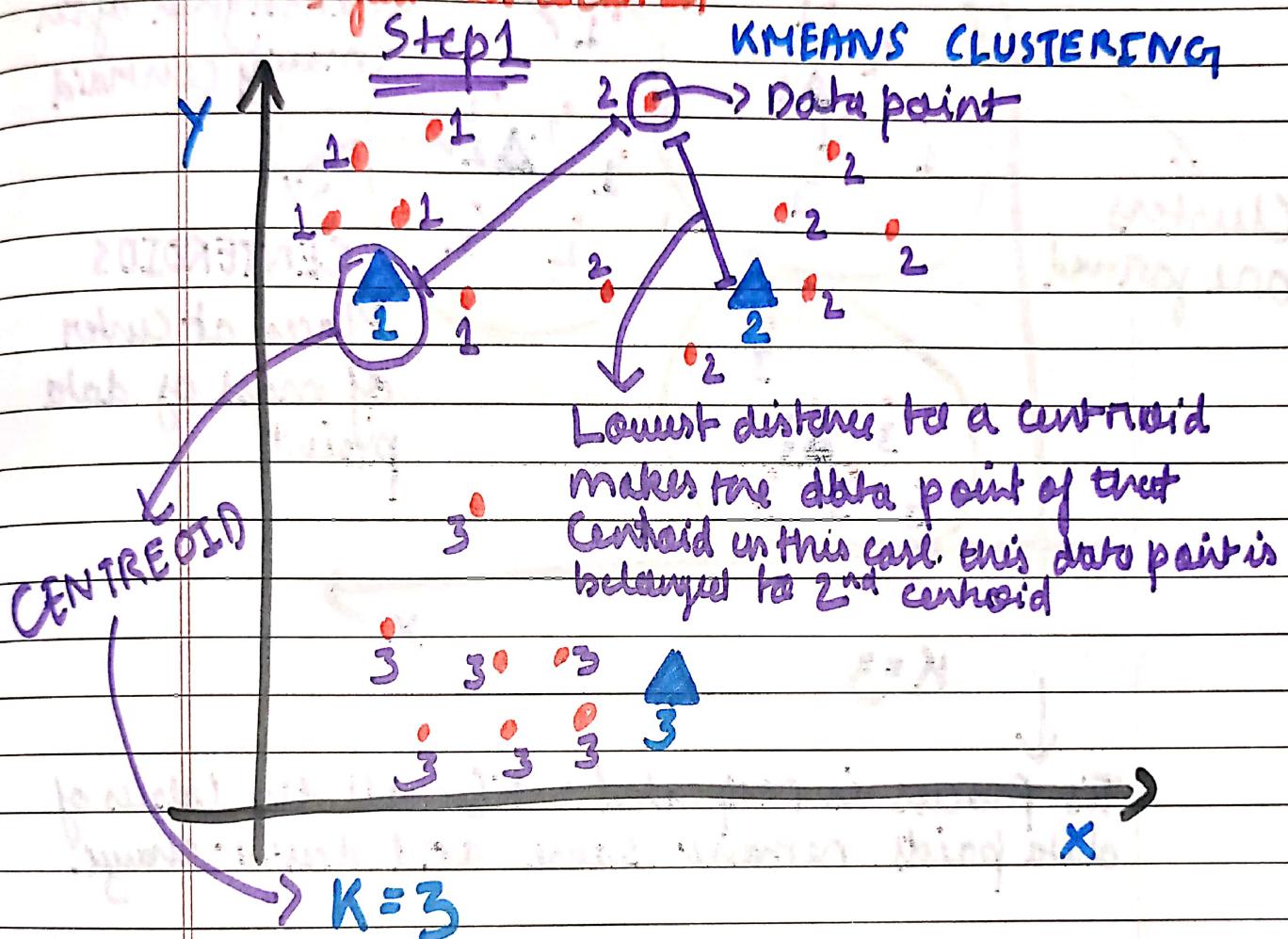


`Predict / Accuracy`

CLUSTERING

Date _____
Page _____

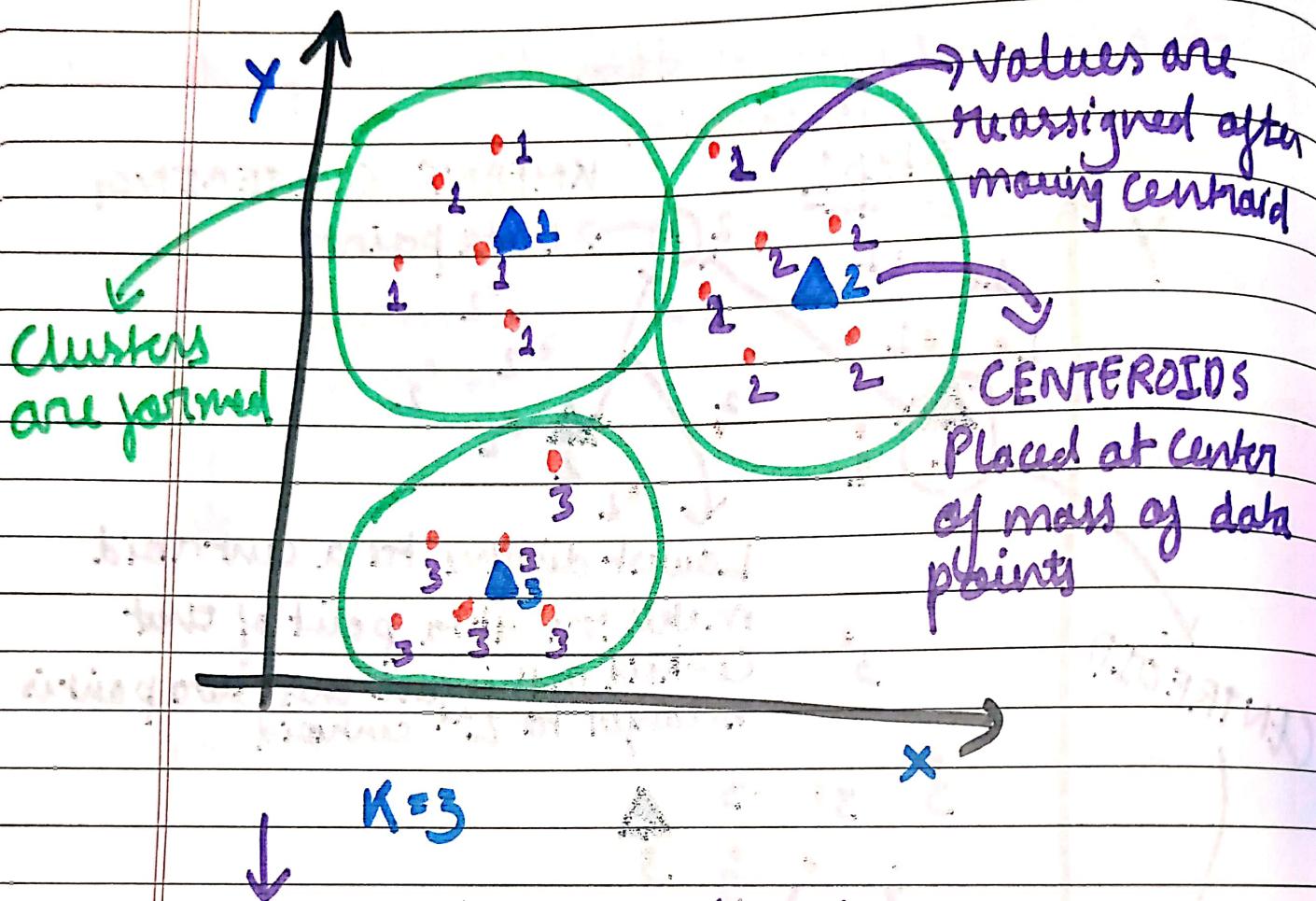
- • Unsupervised learning algorytm.
- • Used when you only have input info and no labels (output) info.
- • Finds cluster of data like data points and tells you its location



K = Number of clusters

Step 2

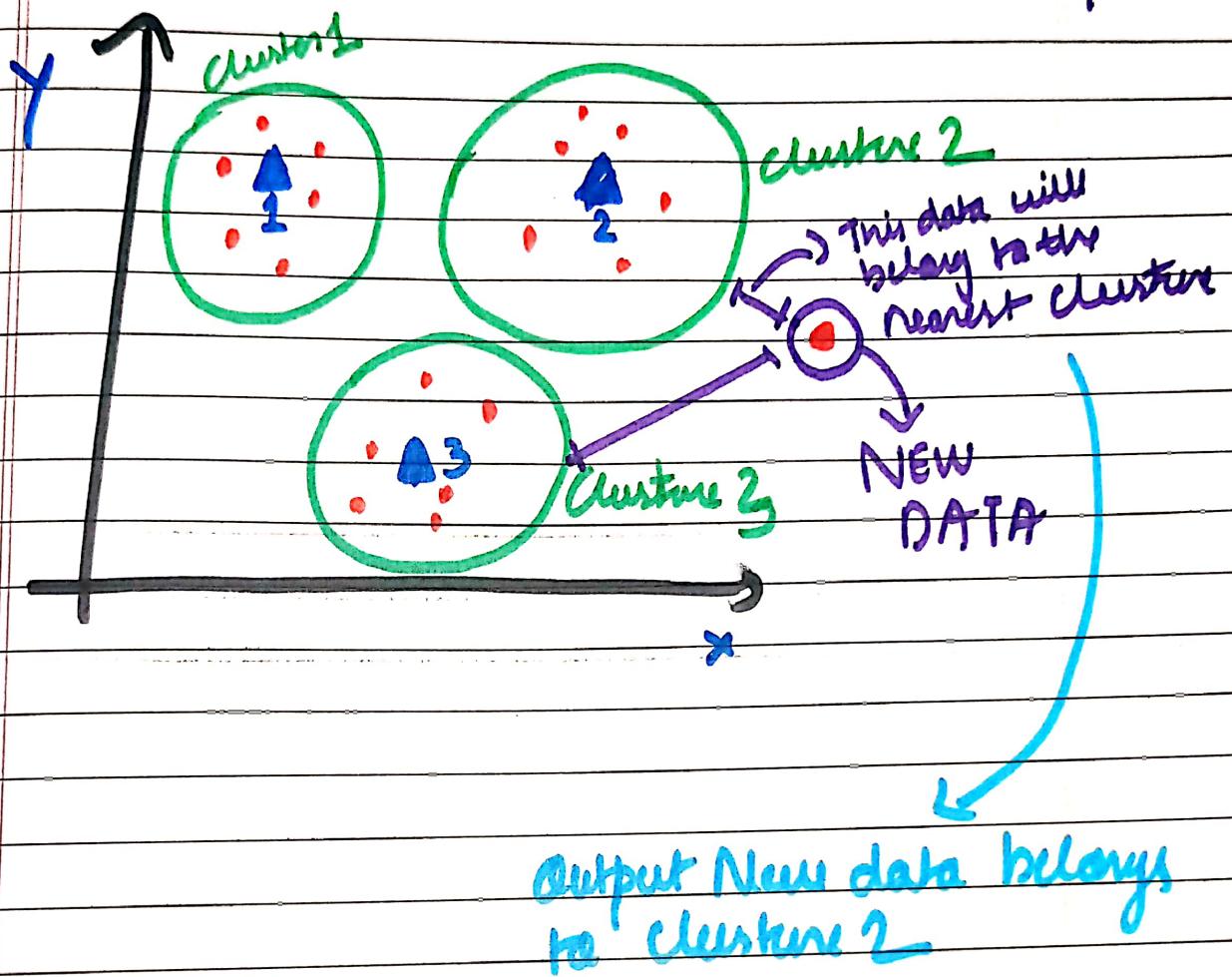
Finds center of mass of Data points
and places centroids in the middle of them



This process is repeated until all the labels of data points remain same and doesn't change

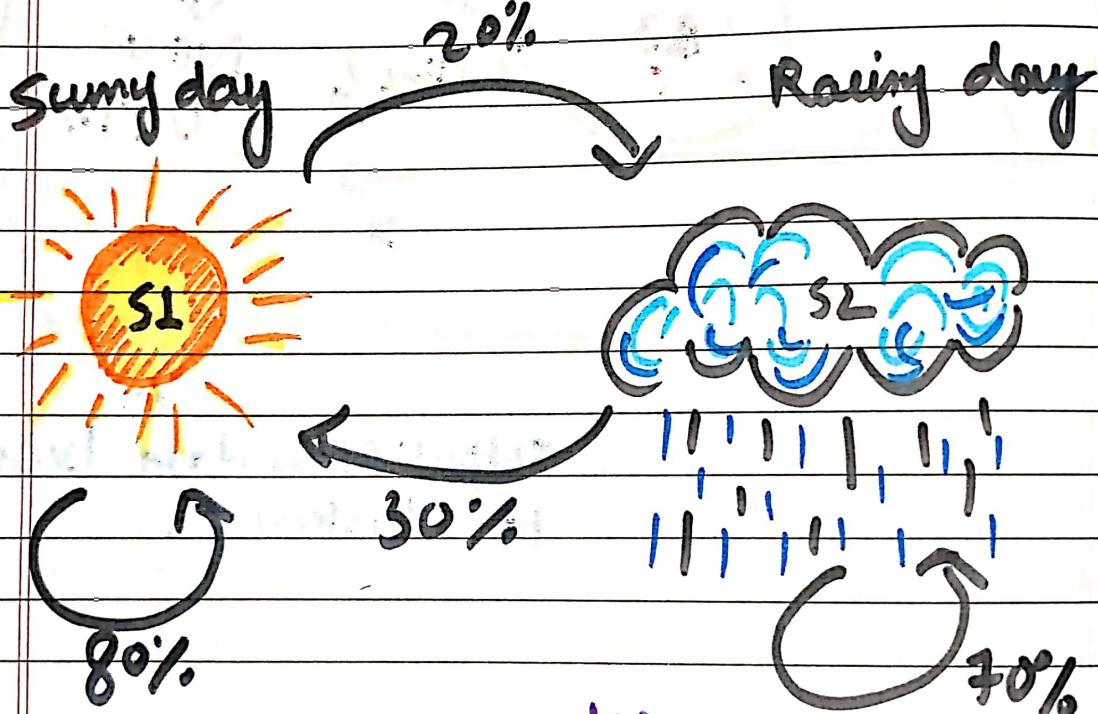
Step 3

Once clusters are formed the new data are assigned to the nearest cluster and labelled the label will be the output



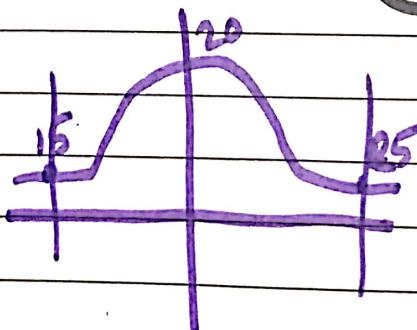
HIDDEN MARKOV MODEL

- ↳ • finds probability of something happening based on observations
- ↓
 - States
 - ↓
 - Observation Distribution
 - ↓
 - Transition Distribution



Observation:

Mean: 20
Min: 15
Max: 25



Observation:
Mean: 5
Min: -5
Max: 15

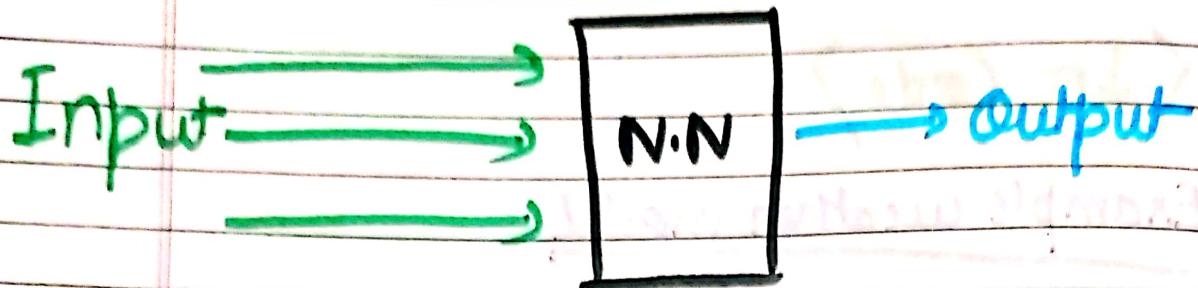
Purpose: Predict future event based on passed events

< In Code >

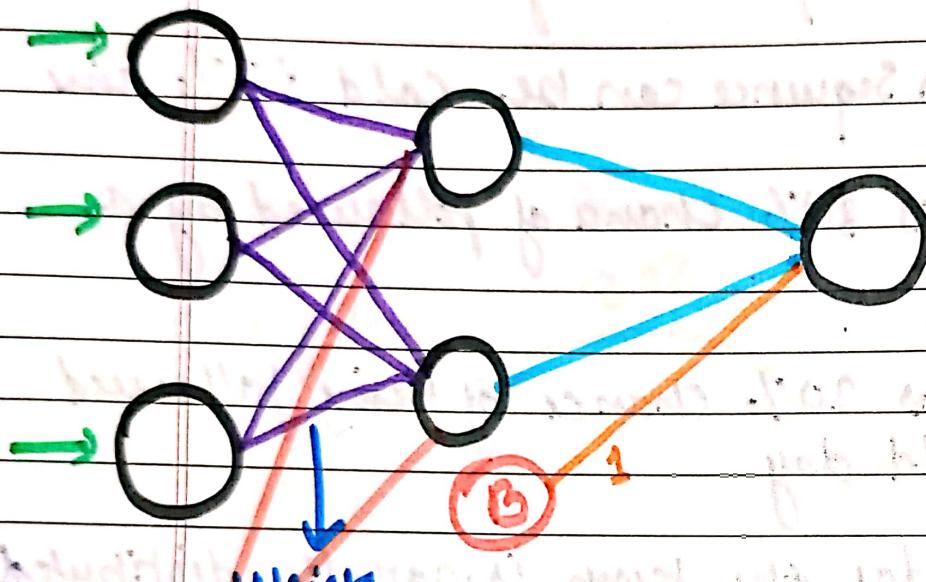
Example weather model.

1. Cold days = 0 / hot days = 1
2. First day in sequence can be cold 80% time
3. Cold day has 30% chance of followed by a hot day
4. Hot day has 20% chance of being followed by a cold day
5. On each day the temp is normally distributed with mean 0 and standard deviation 0 and 5 on a cold day and mean and standard deviation 15 and 10 on a hot day

NEURAL NETWORKS



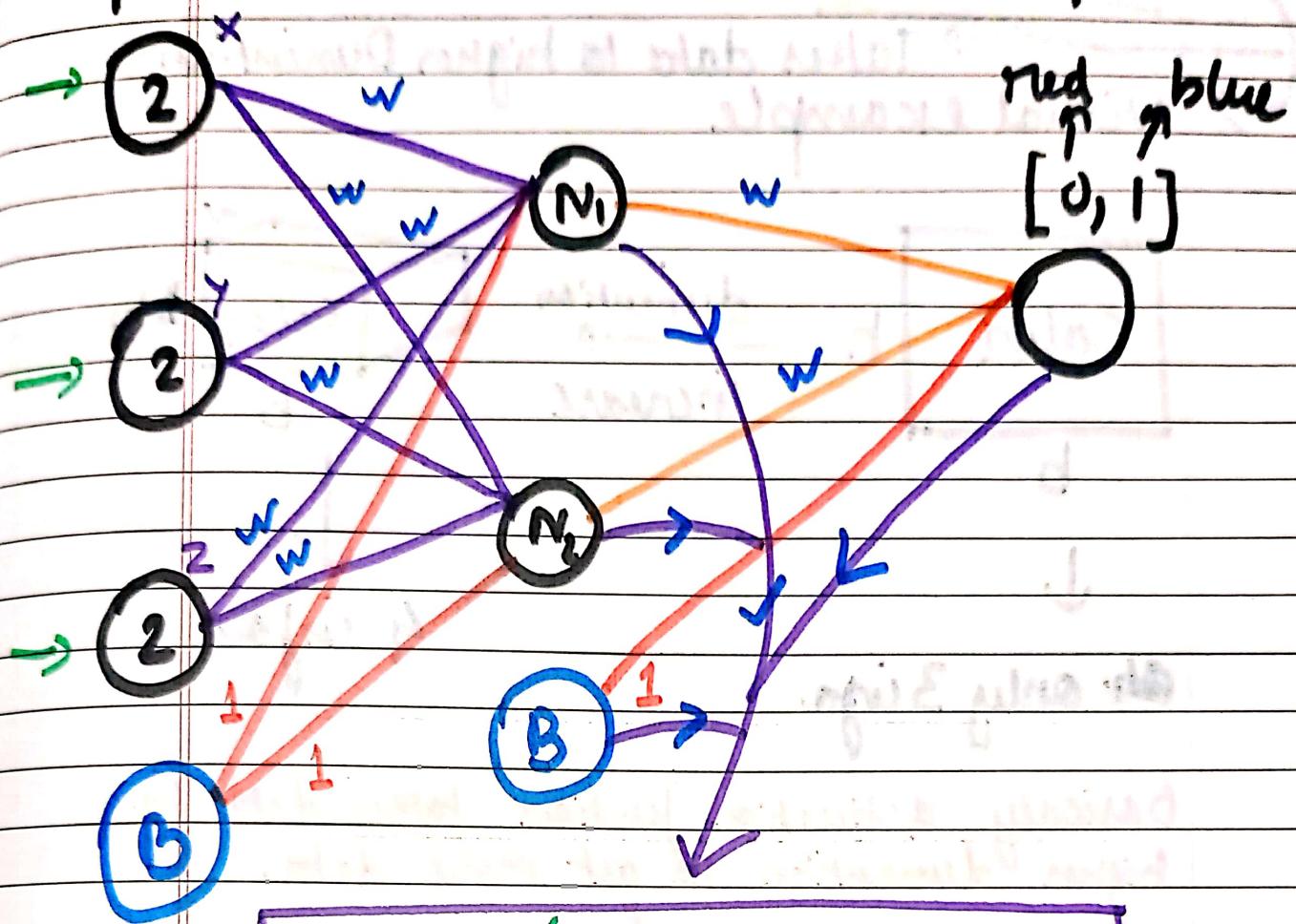
Input Hidden Output



Bias: Constant value

Example: $(x, y, z) \rightarrow [\text{red}, \text{blue}]$

Input Hidden Output



activation function

$$N_x^F = \left(\sum_{i=0}^n w_i x_i + b \right)$$

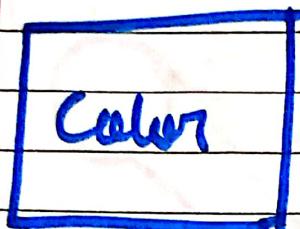
$$= w_x(2) + w_y(2) + w_z(2)$$

= v

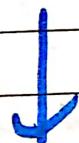
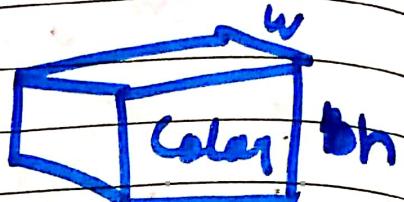
→ Next page

Activation function

→ Takes data to higher Dimension
Visual example



h dimension
increase



4 info

But only 3 info

Basically activation function takes data to higher dimension to get more data

Loss function

Calculates how far output is from expected output and suggest changes

$$(S)_{\text{actual}} + (S)_{\text{expected}} - (S)_{\text{actual}}$$