

#

Session's Agenda:

- 1 DL Landscape
- 2 ML and DL
- 3 Why DL is Popular
- 4 Application
- 5 DL/ANN history
- 6 Perceptron | MLP
- 7 first use case DL

...LANDSCAPE OF DL...

- # 1 DL is Subfield of AI and ML
- = 2 inspired from human brain \Rightarrow Nervous System \Rightarrow Artificial Neurons \Rightarrow Neural Network

Human Neuron Image= Artificial Neuron Image

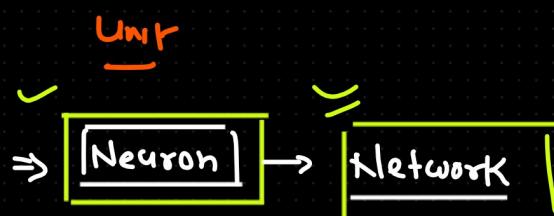
- Deep Learning is the heart of AI

- Deep Learning is subset of Machine Learning which in turn is the subset of Artificial Intelligence. Therefore, we can say that DL is the core of AI.

- Network of Artificial Neurons is called Neural Network

- Neuron is the building block of the neural network

Perceptron is the simplest neural network that is formed with the help of only 1 artificial neuron. Perceptron acts as a building block for the artificial neural network(ANN).



Network formed by the combination of multiple artificial neurons is called Neural Network

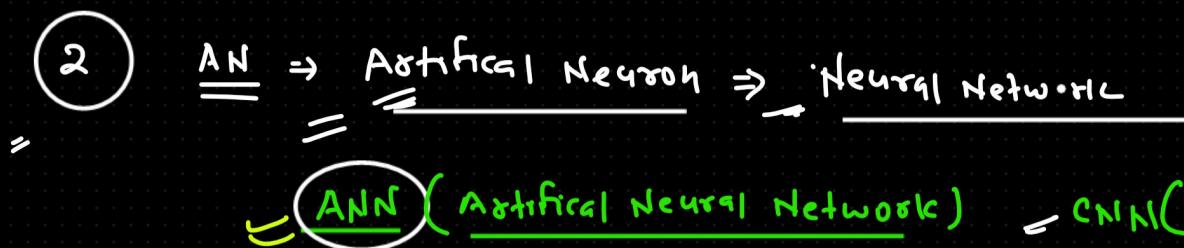
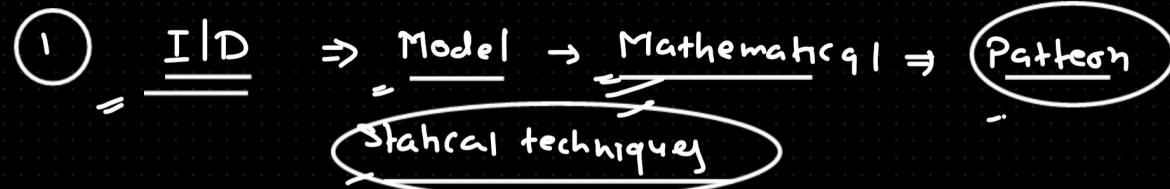


Deep Neural Layer

There are 3 parts/layers of the MLP or ANN: Input layer, Hidden layer(s) and Output layer.

Due to presence of Hidden Layer it acts as an Deep Neural layer

ML vs DL



ANN, CNN, RNN, GAN and RL are the different parts of the Deep Learning where ANN is considered as the basic unit of DL that contributes in all the parts of DL (ANN, CNN, RNN, GAN and RL)

$$\approx \boxed{\text{Convolution}} + \boxed{\text{ANN}}$$



Q-learning and Deep QNN are some famous algorithms used in RL (Reinforcement Learning)

LSTM and GRU are some of the good implementation of RNN

$$\approx \boxed{\text{RNN (Recurrent Neural Network)}}$$

$$\boxed{\text{ANN}} + \text{Recurrence} + \boxed{\text{BPTT}}$$

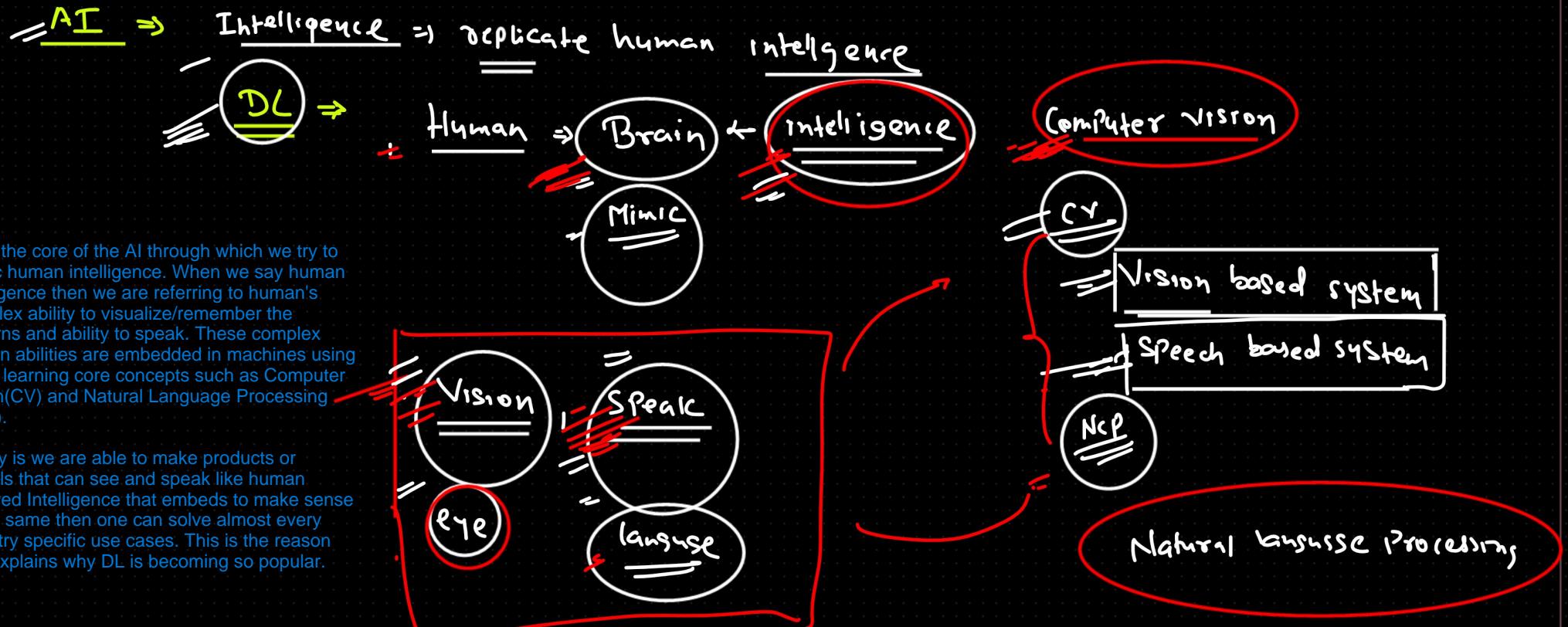
LSTM GRU

BPTT stands for Back Propagation through time

Why this DL is too much famous

Applicable in each and every Domain

Vast amt of Problem, wide Domain of Problem

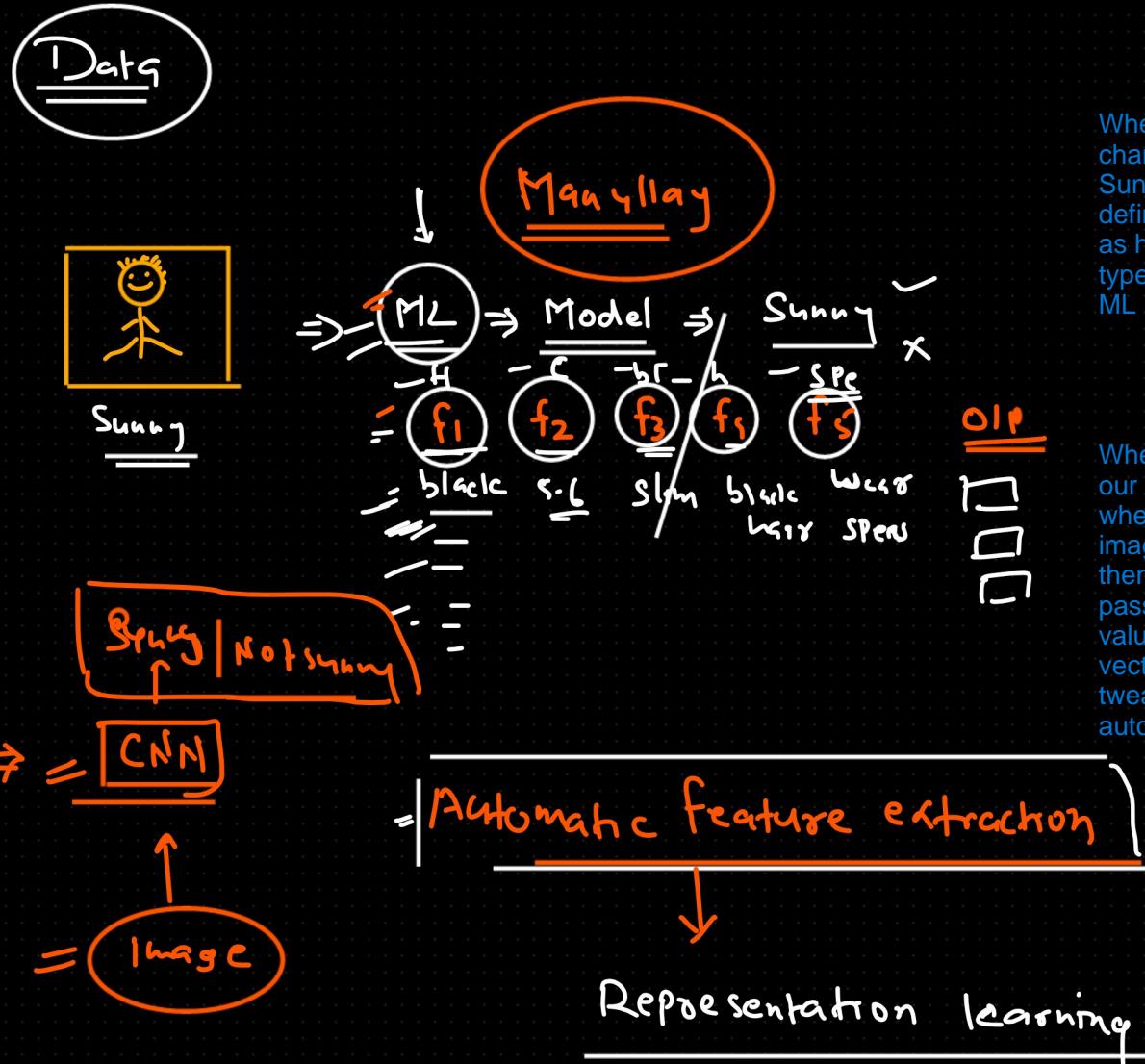


When to use ML or DL :

In ML we need to pass the features manually whereas in DL we don't need to do this. In DL there is automatic feature extraction process through which model itself creates optimal set of features which is also called Representation learning

ML, DL

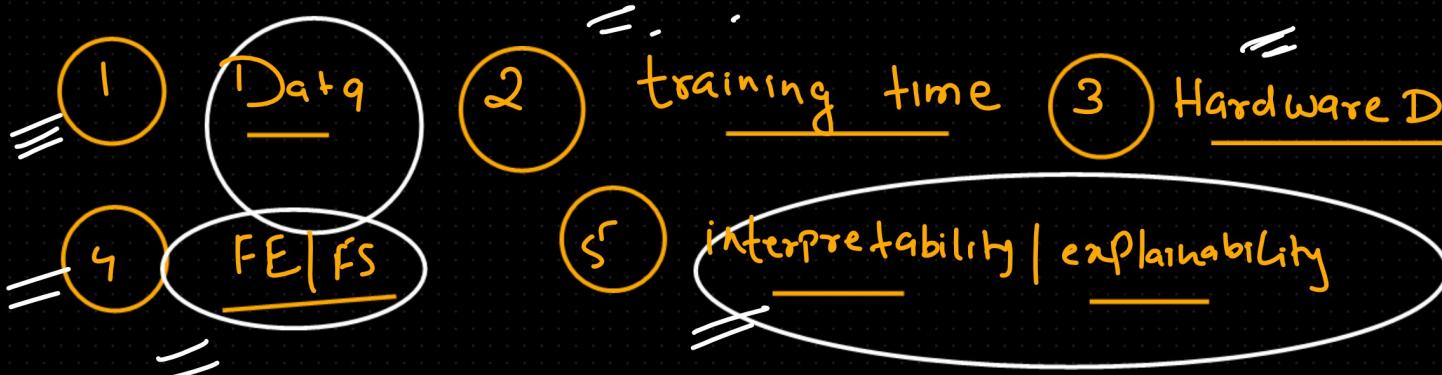
Feature extraction



When we pass the characteristics of Sunny manually by defining feature such as hair, height, body type etc then we use ML

Whereas, if we train our model to identify whether person in image is sunny or not then we are basically passing Pixel RGB values in the form of vector that DL algo is tweaking in an automated fashion

~~X~~ Now we will understand that when to use ML or DL based on the following parameters:

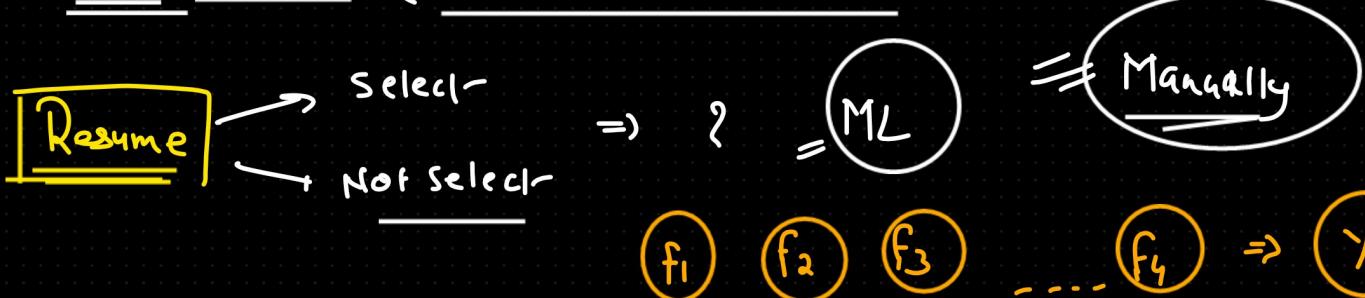


Data:

DL \Rightarrow huge amr of Data

ML \Rightarrow Not required (low amr of Data)

Problem



DL

CNN is the basic unit or building block of CV

\Rightarrow CNN \Rightarrow Image classification
 \Rightarrow Object Detection

Segmentation

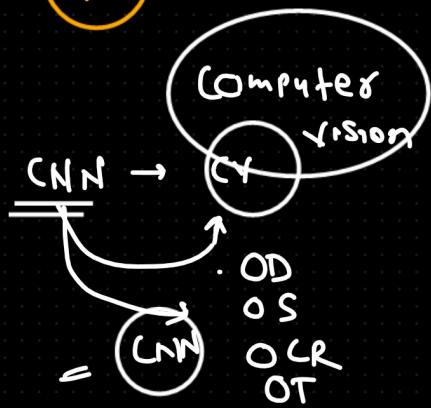
OCR

NLP

RNN

!!!
text-extraction

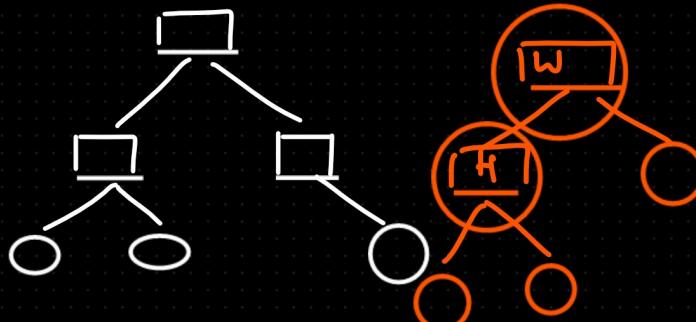
Automate



- 2. Training Time: Since ML models are trained over less data when compared with DL they require then less time. Whereas, DL model can take a lot of time(days, weeks ,months, year). For Eg; ChatGPT is the DL based model that took several months to get trained
- 3. Hardware Dependency: ML models don't have any specific hardware prerequisite and they can simply run over CPU. Whereas. DL models on the other hand have special hardware dependency such as GPU(graphical processing unit), edge devices and TPU.
- 4. Feature Selection or Feature Extraction: This is manual in case of ML whereas in case of DL this is done in an automated fashion that is called Transfer Learning.

ML \Rightarrow Linear reg \Leftrightarrow $y = mx + c$

DT



= weight hypothesis - BMI

$$BMI = m_1 \times W + m_2 \times h + C$$

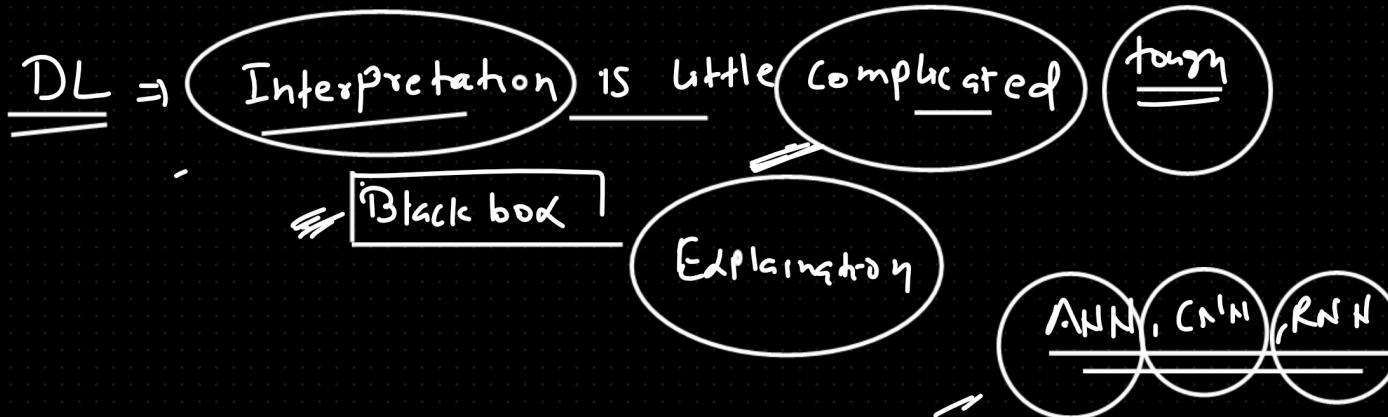
$$0.5 \times W \quad 0.8 \times h$$

In Machine Learning models say for example we are provided with a linear regression based hypothesis where we can easily tell the importance of the features included by comparing the respective weights(Learning Parameter). The feature that is having high importance will be assigned with bigger weight and hence if we talk in terms of Decision tree then that particular feature will be treated as the root node.

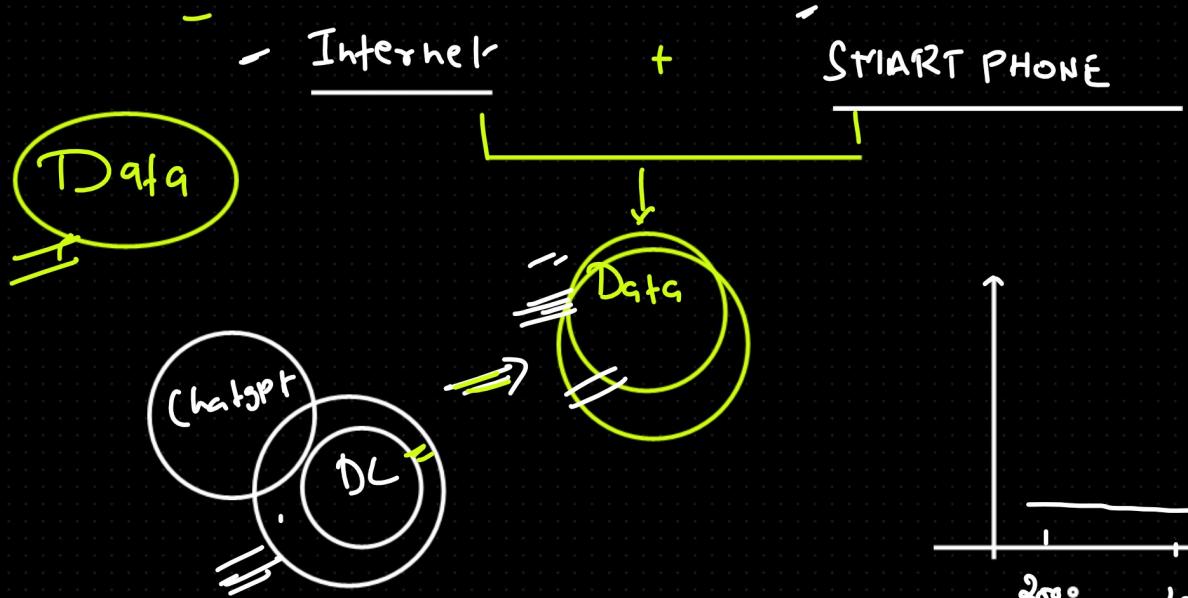
To, conclude we can say that ML models are easy to Interpret and Explain.

Whereas, DL model on the other are little complicated and a bit difficult to Interpret and Explain due to presence of complex network of hidden layers. This is the reason why DL models are also often called as a BLACK BOX.

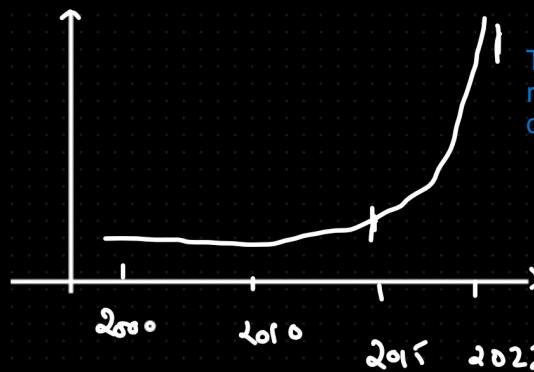
Interpretation | Explain



Understanding the reasons that actually gave the breakthrough in the field of data resulting in the generation of huge amount of data thus expanding the scope of AI, ML and DL:



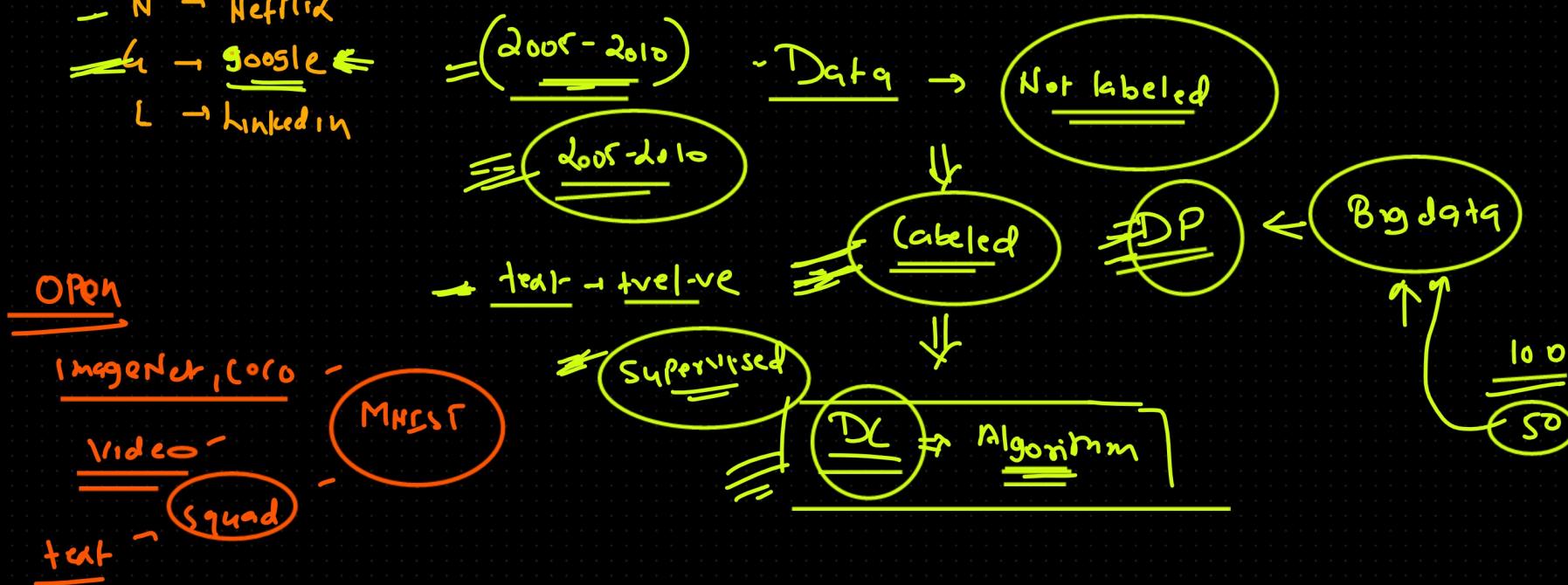
Internet and Smartphones are the ones that has given a breakthrough in the field of data thus generating huge volume of data.

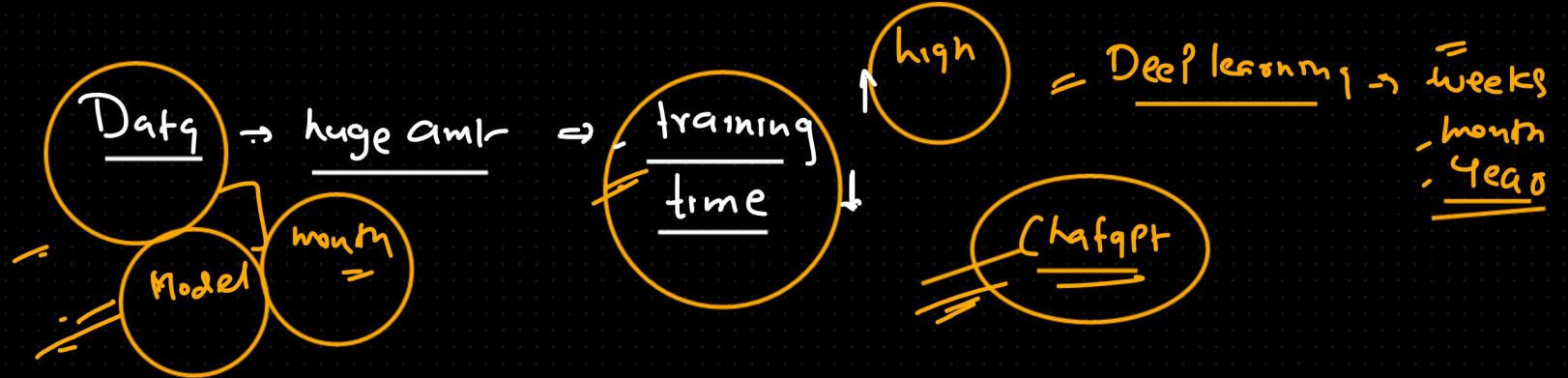


This is the graph that gives the pictorial representation of the increase in the volume of data with the increase in time

- M → Meta, Microsoft
- A → Apple, Amazon
- N → Netflix
- G → Google ← = (2005 - 2010)
- L → LinkedIn

Data coming to tech giants in timeframe 2005-2010 where there was a lot of data which were not labeled. Hence labeling was done and then feeded to the algorithms





Hardware

- ML ⇒ CPU
 - DL ⇒ GPU

CPU

GPU

Edge device

Google

TPU

Parallel

Nvidia

Framework

- ML ⇒ Scikit learn

- DL ⇒ tensorflow
 ↳ google

pytorch
 ↳ fb

In ML we used sklearn or scikit learn pythonic Library extensively for implementing the ML based concepts.

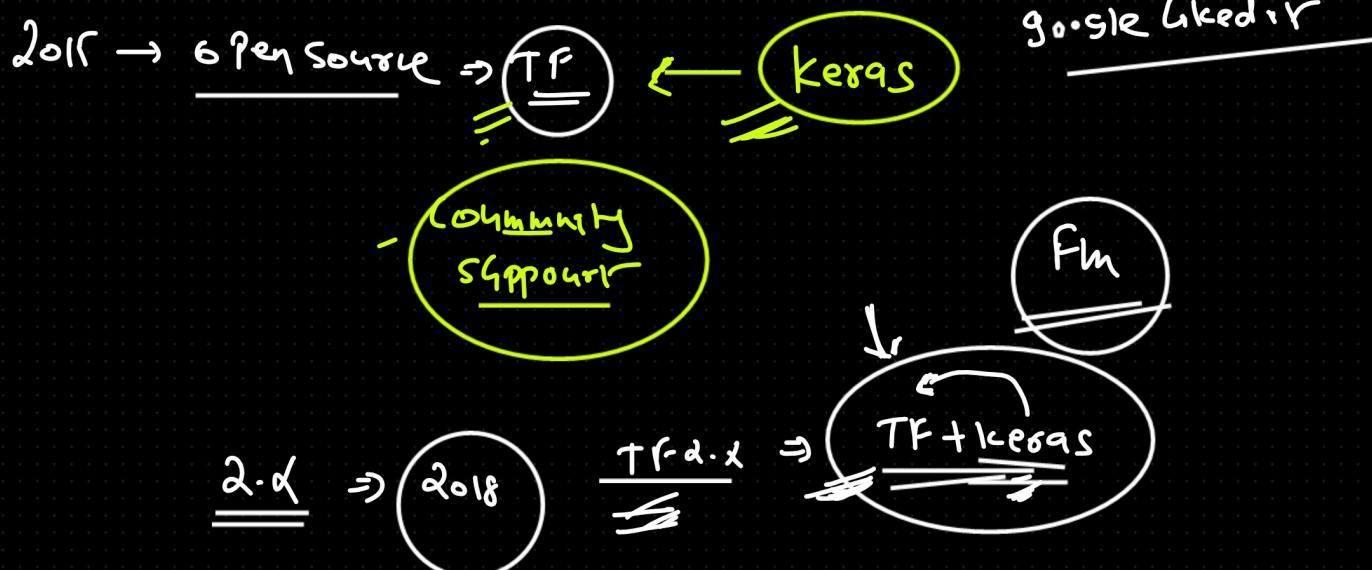
In the same way For implementing DL we are provided with 2 pythonic lib:
 TensorFlow by Google and Pytorch by Facebook



In 2015 tensorflow was made open source with strong community support. But it was difficult to implement for developers. So community made Keras on top of Tensorflow which simplified the process.

Tensorflow can be seen as Backend whereas, Keras as frontend working on top of Tensorflow

Internal Proj. of google



Power \Rightarrow DL

COMMUNITY SUPPORT

TRANSFER LEARNING

Below are some of the examples of DL based models or frameworks which one can download and use as per their need. Also, one can perform fine tuning of these existing models as per the use case

- If someone already trained model on the data you want to use it for your purpose you can do it \rightarrow fine tune Model \rightarrow Download

Image classification \Rightarrow $\begin{cases} \text{ResNet} \\ \text{GoogleNet} \end{cases}$

ResNet or GoogleNet are Image Classification framework/architecture/model that one can download and use for simpler implementation

Object Detection \Rightarrow $\begin{cases} \text{YOLO} \\ \text{SSD} \\ \text{Fast, faster RCNN} \end{cases}$

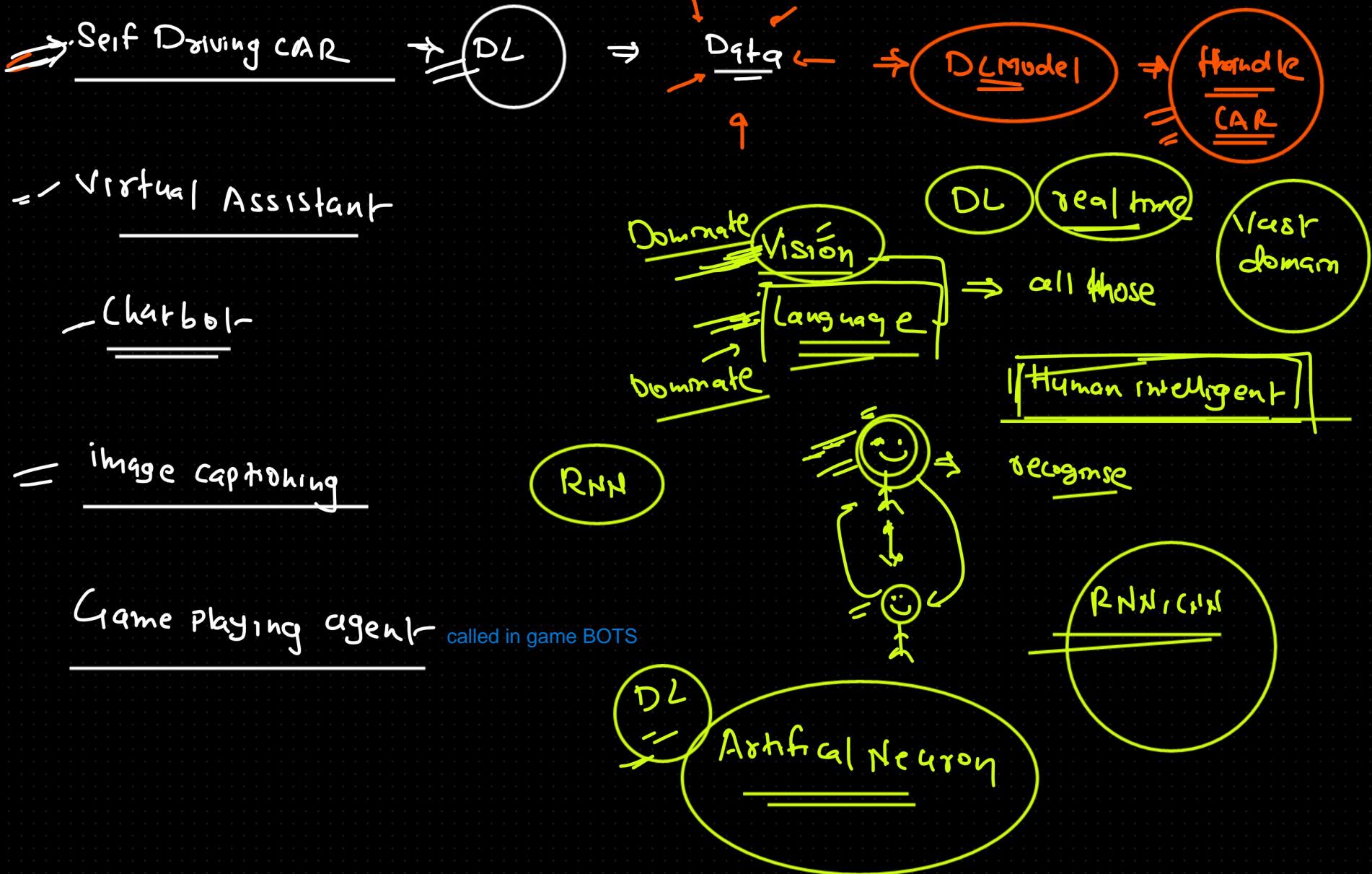
Similarly we have framework for Object Segmentation, Image translation, Speech generation, Text classification etc

Object Segmentation \Rightarrow $\begin{cases} \text{Mask RCNN} \\ \text{U-Net} \\ \text{V-Net} \end{cases}$

Image translation \Rightarrow Pic to Pic

Speech generation \Rightarrow WaveNet

Text classification \Rightarrow BERT



RoadMap \Rightarrow

DL



MULTIlayer Perception

Python \Rightarrow TF, PT

activation

loss

optimizer

Vanishing gradient

Exploding gradient

Regulation, Normalization (L_1, L_2 , BatchN,

Dropouts)

Hyperparameter tuning (kerastuner)

tensorboard

CNN =

- 1 Convolution
- 2 Poolings
- 3 Kernels
- 4 LeNet, Alexnet, Vgg, ResNet, GoogleNet (These are some of the CNN frameworks)
- 5 Practical \Rightarrow CNN App
- 6 Object Detection — SSD
— RCNN
— YOLO
- 7 Obj Seg
- 8 framing, OCR
- 9 GPU, Edge devices
- 10 Projects

OpenCV

RNN

- 1 RNN
- 2 LSTM, GRU
- 3 Encoder, Decoder
- 4 types of RNN
- = (seq to seq)
- 5 attention
- 6 Self attention, transformer
- 7 BERT (LM) =
- 8 GPT
- 9 OpenAI (ChatGPT)
- 11 Project

Wordembedding

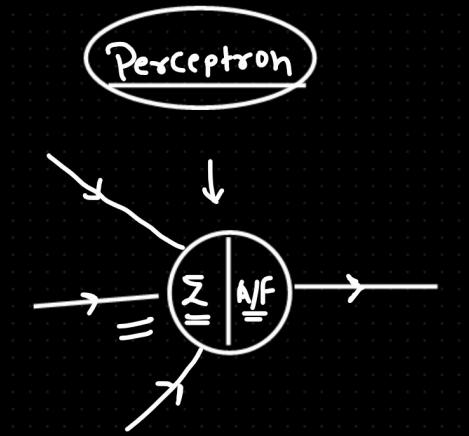
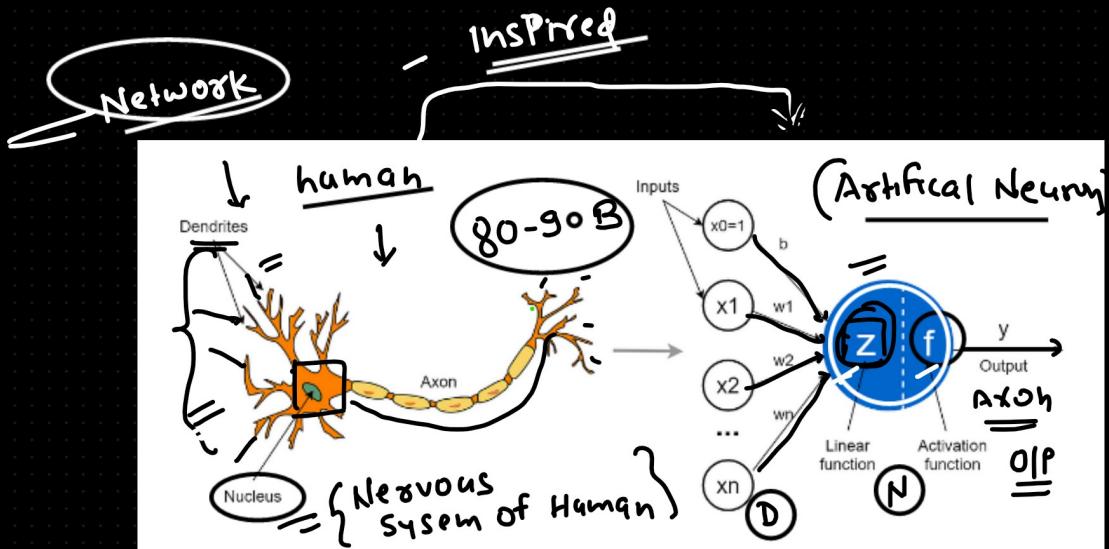
- Word2Vec
- Elmo

= Hugging face

GAN

RL

Array and tensor



Summation + Activation

Height | weight | O/P

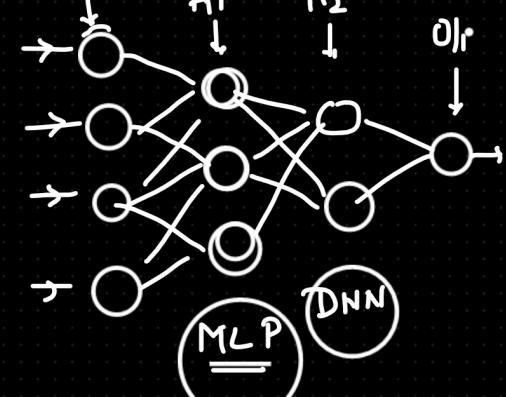
Training \Leftarrow DL
Prediction \Leftarrow DC

{ Feed forward NN }
BP

This is the structure of a Neuron. A neuron is the simplest neural network which is called PERCEPTRON. It acts as an building block for ANN. Go through lecture 2 notes for understanding more

The diagram shows a detailed view of a perceptron. Inputs H and W (represented by arrows) connect to a summation node (Σ). The summation node also receives a bias b. The output of the summation node passes through an activation function (A/F) to produce the final output O/P. The equation for the output is given as:

$$\text{O/P} = \text{Act} (Hw_1 + w_2 + b)$$





Linear reg \Rightarrow $Ax + By + C = 0$ \Rightarrow equation of line

$$y = mx + c$$

Step F/n

$x \geq 0$	<u>Obese</u>
$x < 0$	<u>Not obese</u>

Sigmoid, tanh, Softmax, Step, Linear

loss

BP \Rightarrow GD, SGD, Minibatch, Adagrad, RMS
SGD with Momentum PROP

Adadelta, ADAM

Pytorch, TF