



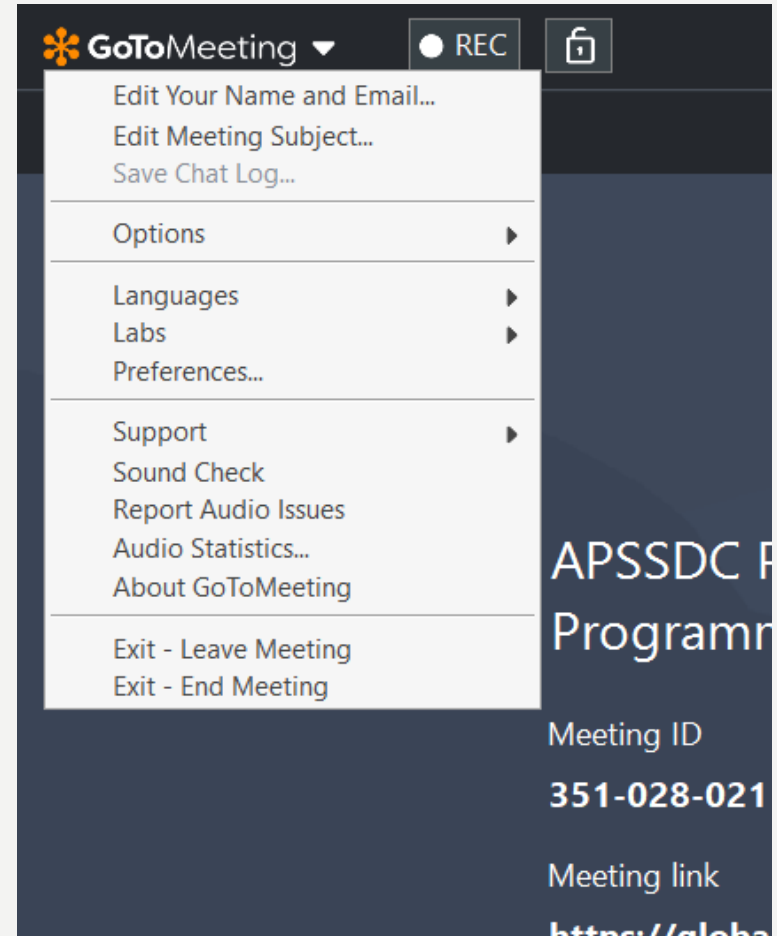
**Skill AP**  
**APSSDC**

# AN INTRODUCTION TO MACHINE LEARNING



# For Attendance and Verification Purpose

**RollNo-Name-  
CollegeCode/  
CollegeName  
And RegisteredEmail  
ID**



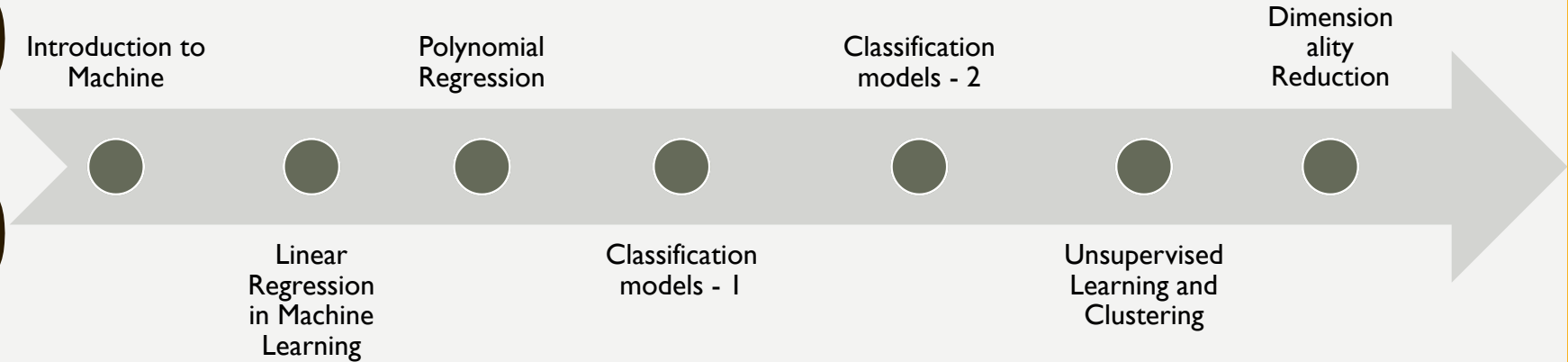
By Anil Kumar APSSDC

# SESSION RESOURCES

<https://bit.ly/apssdc-ml-ab2>

# MACHINE LEARNING USING PYTHON

## AGENDA



# DAY1 AGENDA

What is  
Machine  
Learning

Machine  
Learning  
Classification

Types of  
Algorithms

Data  
Importing  
and  
manipulating

# WHAT IS THIS FRUIT?



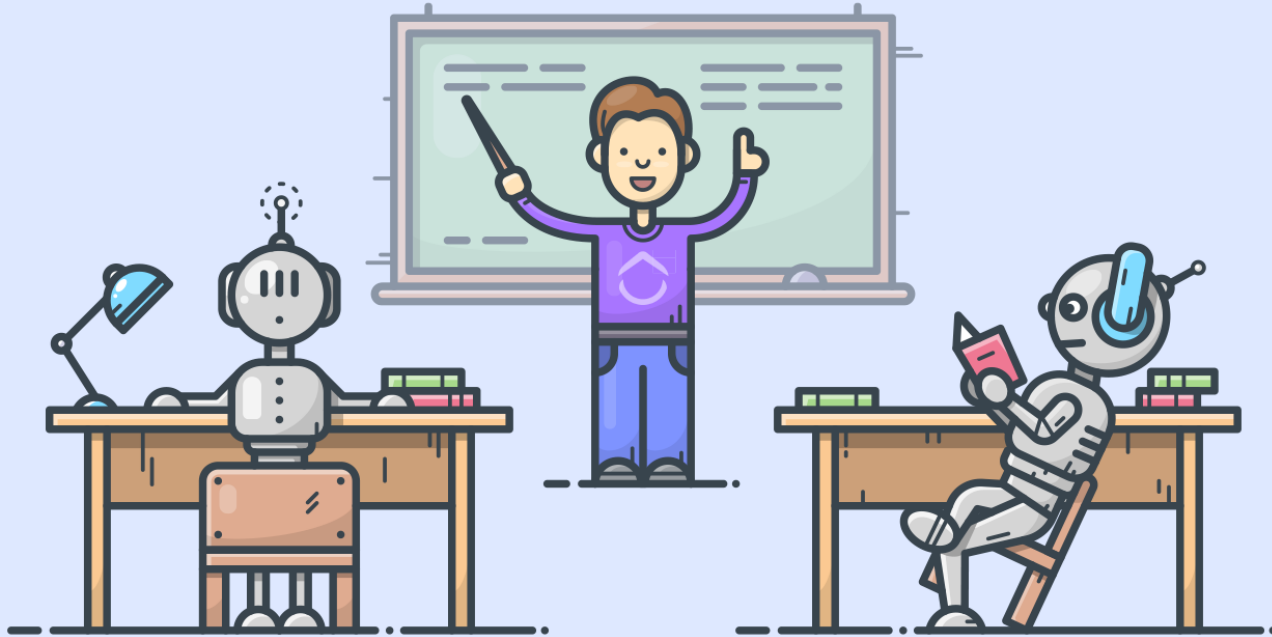
- Shape - Love symbol
- Color - Red
- Weight – >500grms

# WHAT IS THIS FRUIT?

- PAPPLE -> pears + Apple
- White color dots



# WHAT MACHINE LEARNING ?





“A computer program is said to learn from experience **E** with respect to some class of tasks **T** and performance measure **P**, if its performance at tasks in **T**, as measured by **P**, improves with experience **E**.”

— Tom Mitchell, Professor at Carnegie Mellon University

# WHY YOU ARE THE THIS THIS SESSION

- For learning this machine learning program sir
- To gain knowledge
- to learn ml programing using python programing
- **To know about how the algorithms are used for making the machine learn**
- **i want to read the IMAGE and to make analysis**
- **do project on ML**
- **To develop skills of using recent machine learning software for solving practical problems.**

# PRESENT TRENDING TECHNOLOGIES

## Python

- **AI & ML**
- **Big Data → Pyspack**
- **IoT**
- **Block Chain**
- **Deep Learning**
- **Quantum Computing, QML, QCML → Qiskit**
- **NLP, CV**
- **Web Development**
- **Data Science, Data Analyst, Data Engineer**
- **Cyber Security**
- **Devops Engineer**
- **Cloud Computing**

# WHAT IS **ARTIFICIAL INTELLIGENCE, MACHINE LEARNING AND DEEP LEARNING**



# ARTIFICIAL INTELLIGENCE

Artificial Intelligence (**AI**) is the science of making things smart. Can be defined as:

**“Human intelligence exhibited by machines”**

A broad term for getting computers to perform human tasks. The scope of AI is disputed and constantly changing over time.

# AI: COMMON USE CASES

- Object recognition
- Speech recognition / Sound detection
- Natural Language Processing / Sentiment analysis
- Creative (e.g. Style Transfer – Learning to draw an image in the style of an artist)
- Prediction – given some inputs, what is the expected output for unseen examples
- Translation between languages
- Restoration / Transformation – e.g. taking an image and using ML to figure out what should be there, or generating faces based on what it knows face to be.
- Some [AI Examples](#)

# MACHINE LEARNING

- Machine Learning (**ML**) can be defined generally as:

**“An approach to achieve AI through systems that can learn from experience to find patterns in a set of data”**

ML involves **teaching a computer to recognize patterns by example, rather than programming it with specific rules.** These patterns can be found within data. In other words, ML is about creating algorithms (or a set of rules) that learn complex functions (or patterns) from data and make predictions on it –a form of “narrow AI”

# A DIFFERENT WAYS OF DOING THINGS

Write a computer program  
with **explicit rules** to follow

```
if email contains V!agrå  
    then mark is-spam;  
if email contains ...  
if email contains ...
```

**Traditional Programming**

Write a computer program  
to **learn from examples**

```
try to classify some emails;  
change self to reduce errors;  
repeat;
```

**Machine Learning Programs**



# DEEP LEARNING

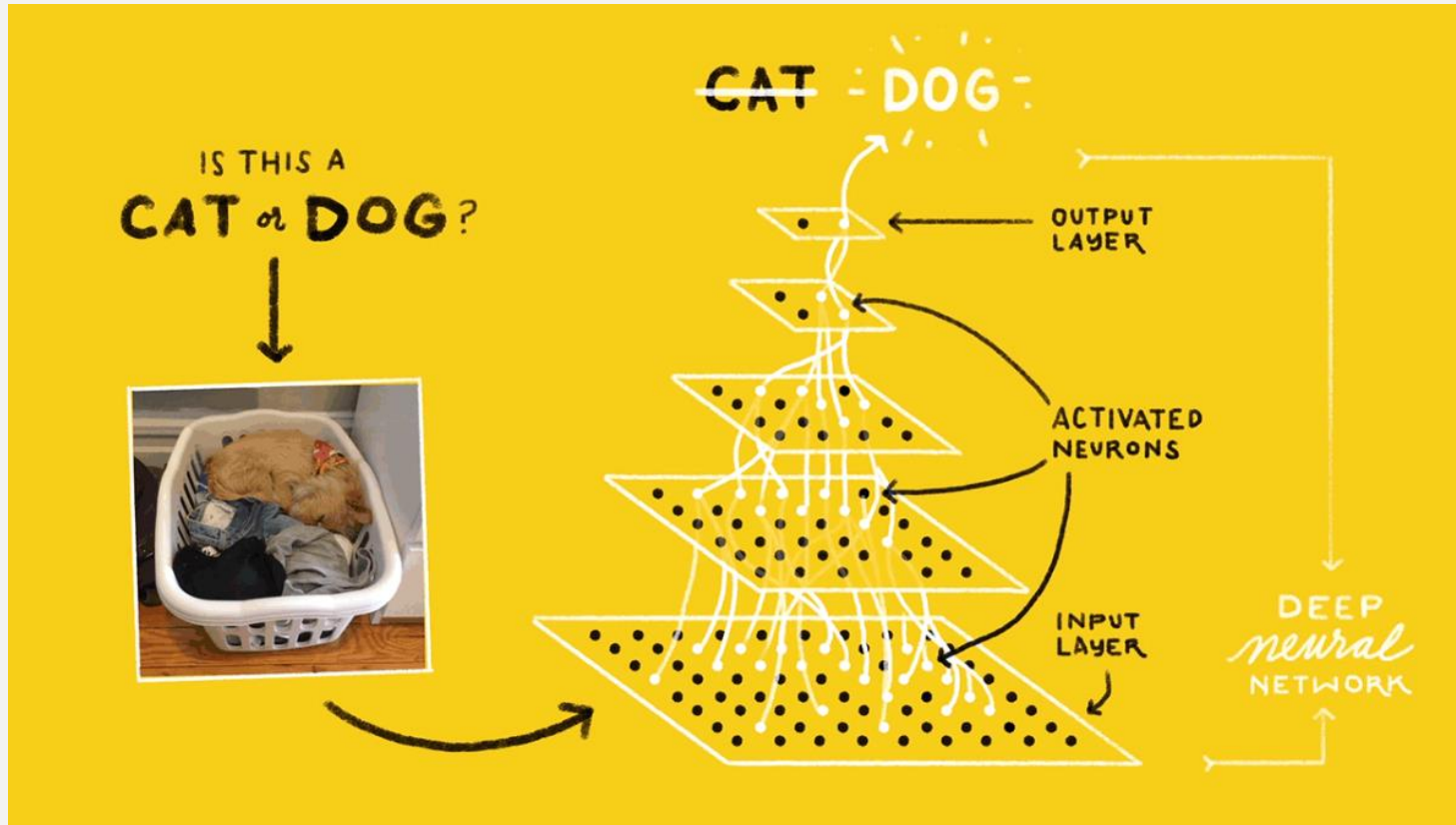
- Deep Learning (**DL** from here on) can be defined generally as:

**“A technique for implementing Machine Learning”**

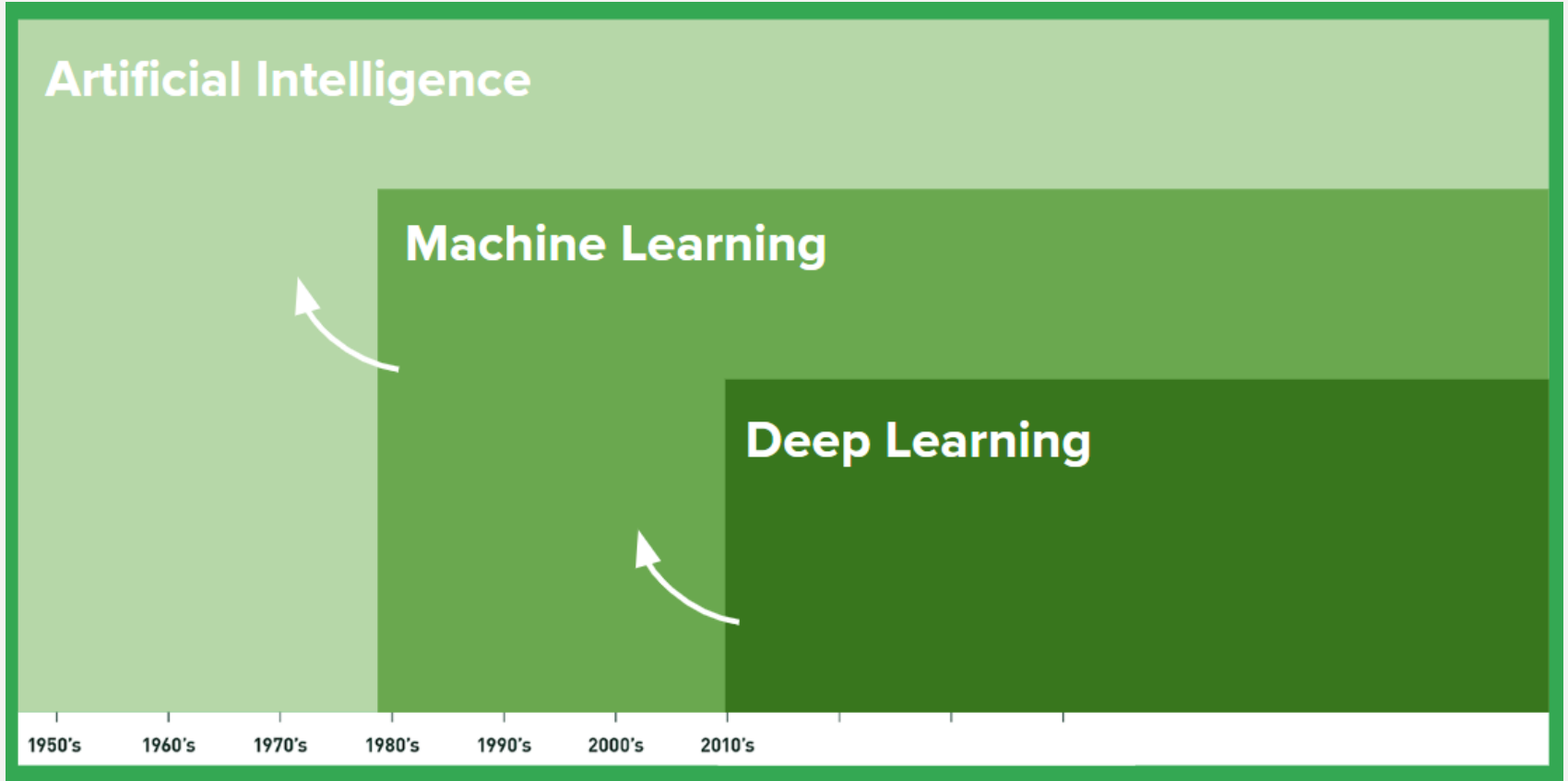
One such DL technique is a concept known as **deep learning Neural networks (DNNs)** which you may have heard of.

Essentially DL in the context of DNNs is where the code structures you write **are arranged in the layers that loosely mimic the human brain, learning patterns of patterns.**

# SIMPLE NEURAL NETWORK

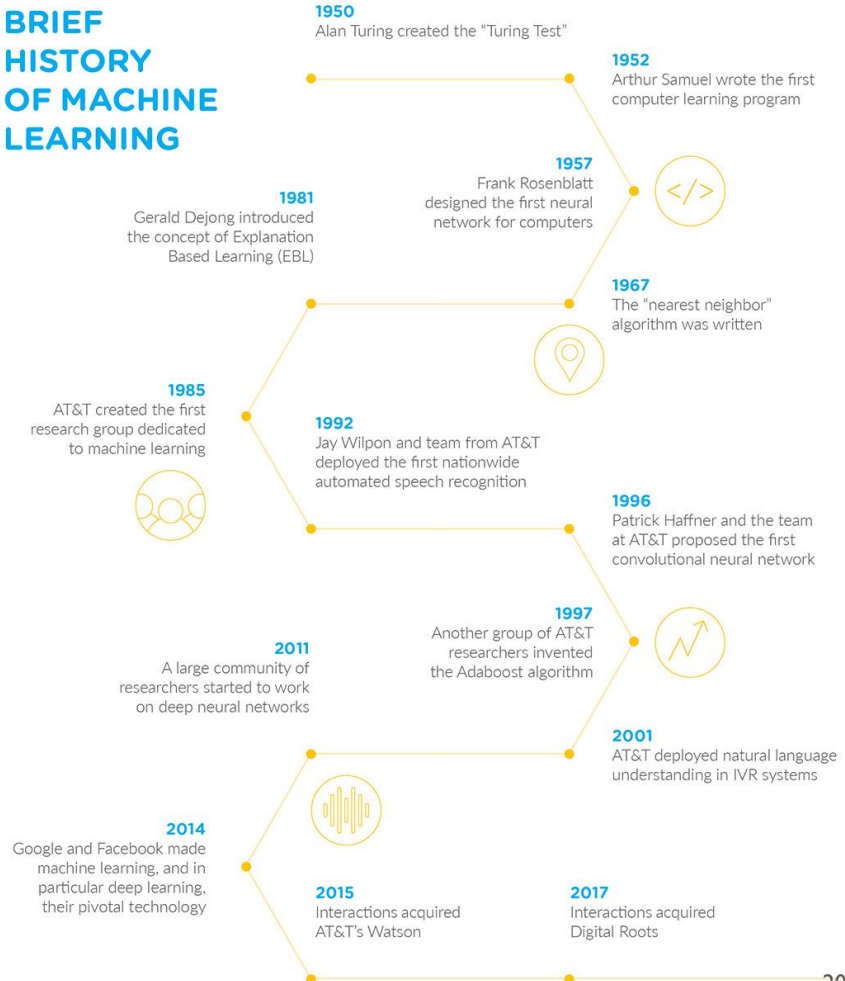


# SUMMARY



# HISTORY

## BRIEF HISTORY OF MACHINE LEARNING

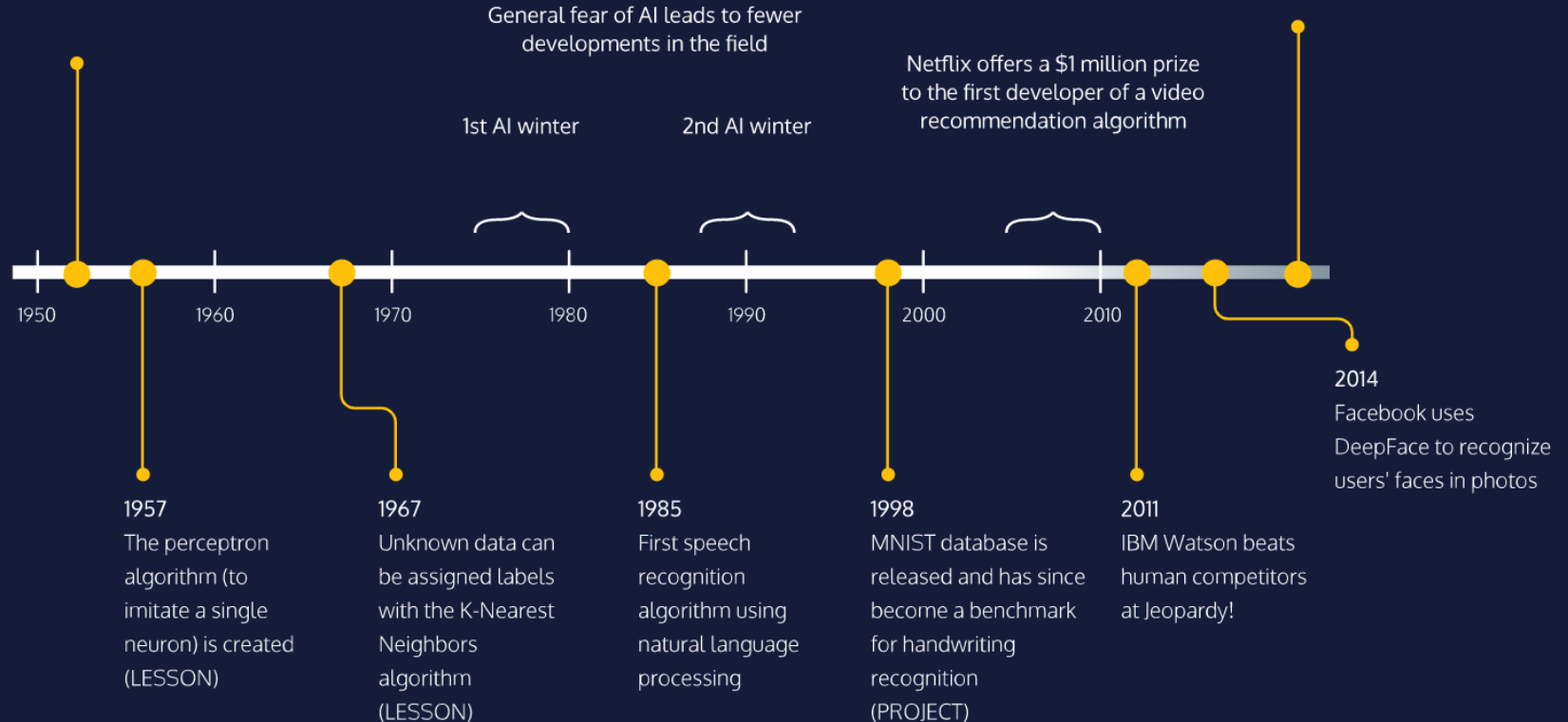


1950

Alan Turing publishes  
Computing Machinery  
and Intelligence  
"Can machines think?"

2016

Google's AlphaGo  
beats professional  
players at Go  
A notoriously difficult  
board game studied  
for centuries



# FEW OTHER DEFINITIONS

“Machine learning is the hot new thing”

— John L. Hennessy, President of Stanford (2000–2016)

“A breakthrough in machine learning would be worth ten Microsoft”

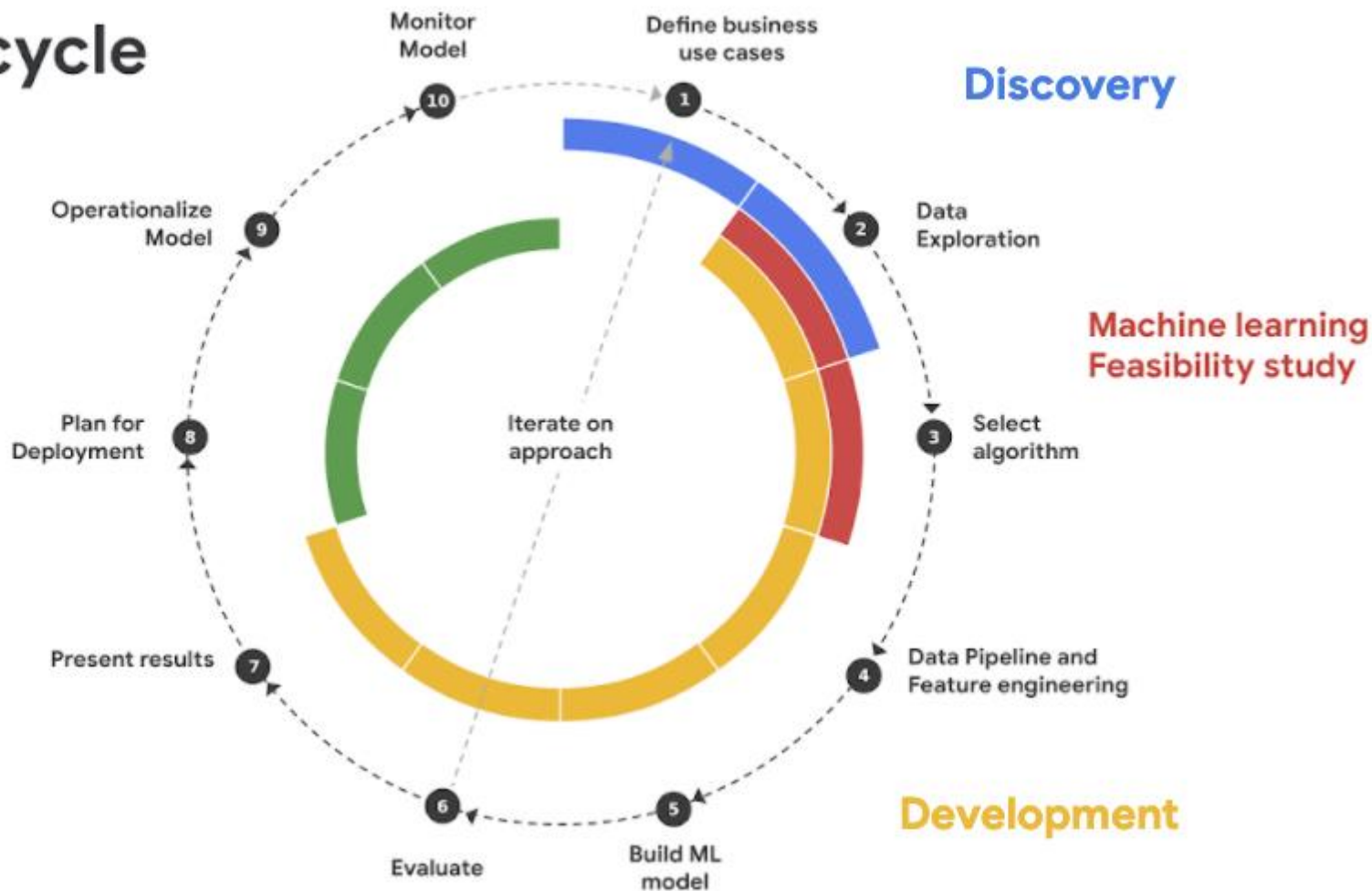
— Bill Gates, Microsoft Co-Founder

“Machine learning is the field of study that gives computers the ability to learn without being explicitly programmed”

— Arthur Samuel's

# ML Lifecycle

Deployment

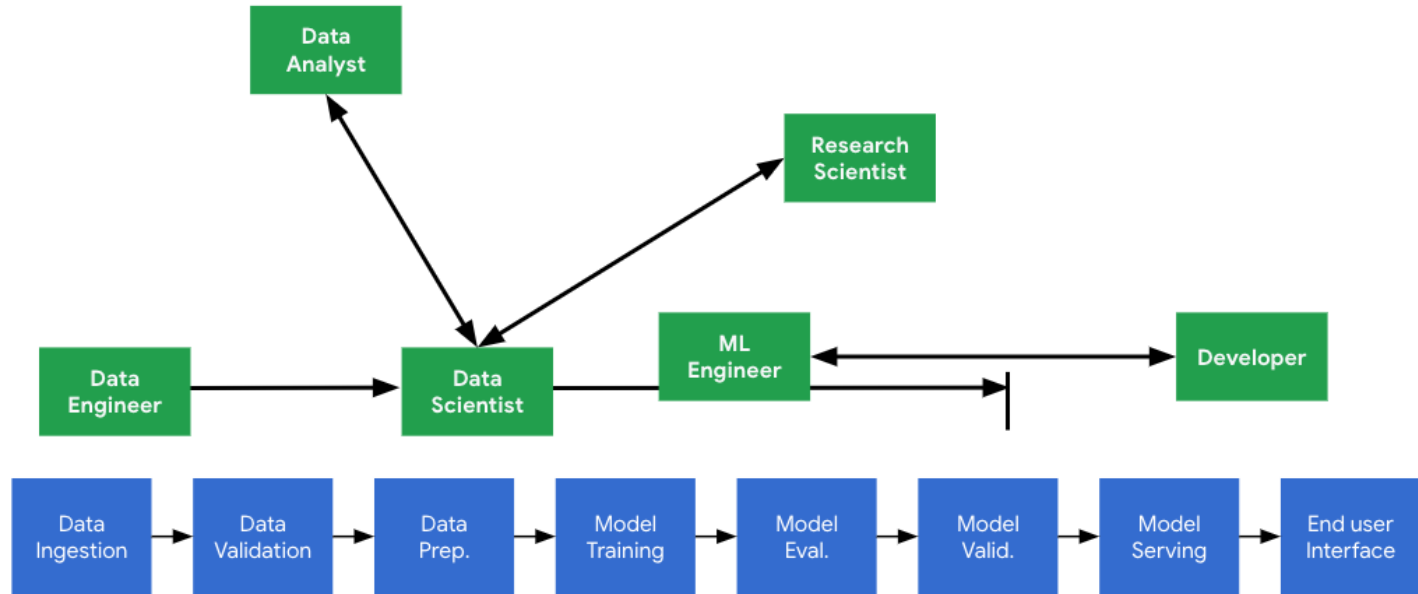


# THE NEED FOR MACHINE LEARNING DESIGN PATTERNS

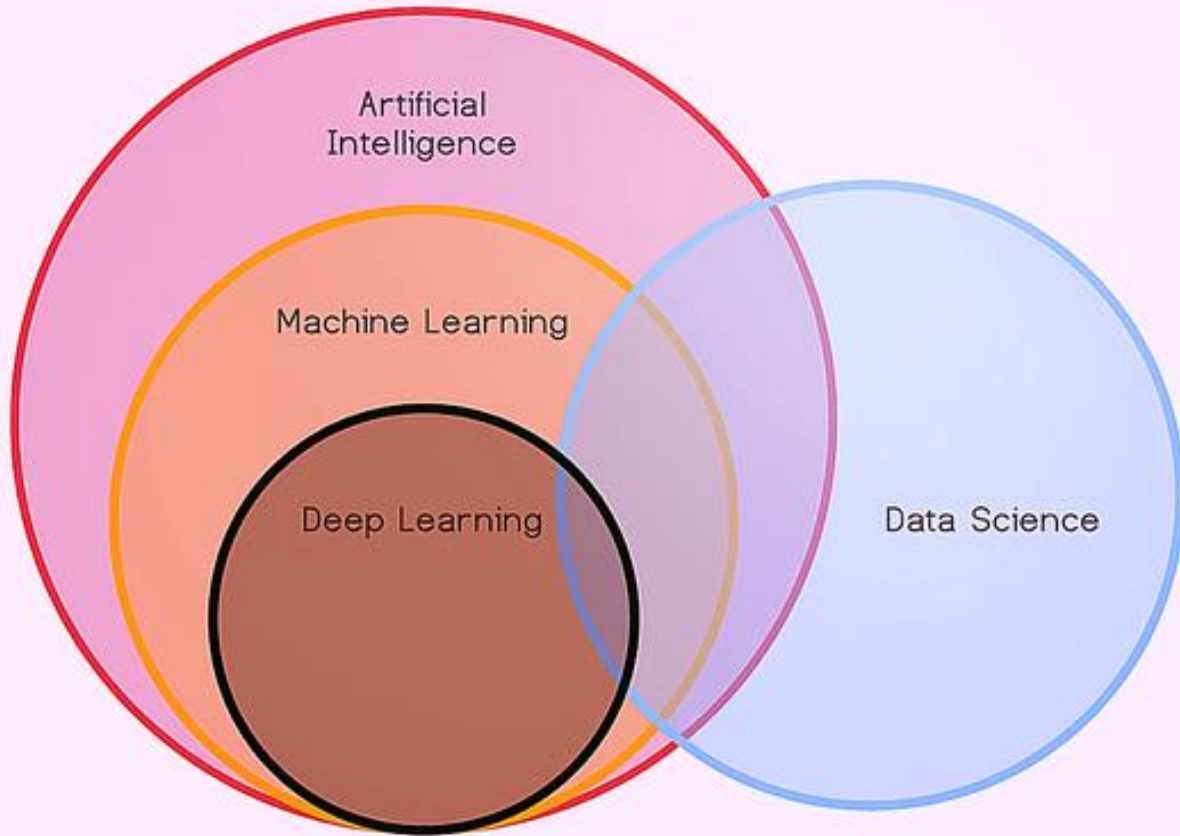
Business

R & D

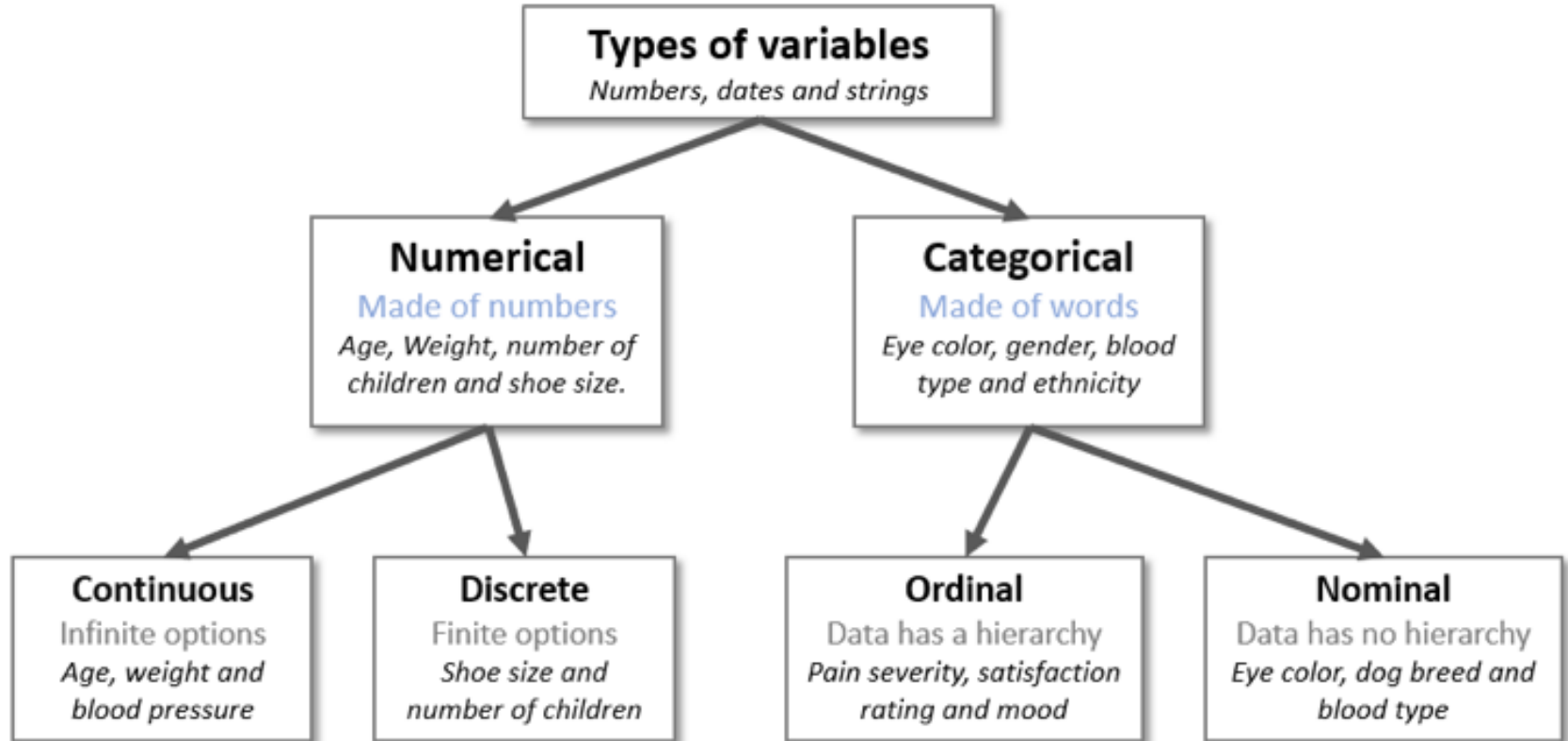
Engineering





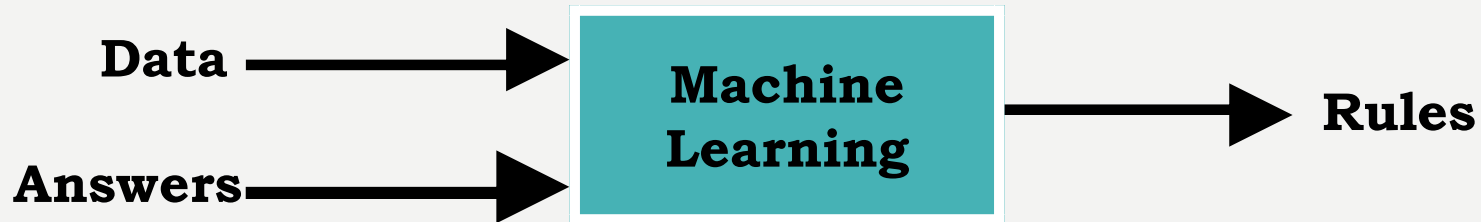


# TYPES OF VARIABLES



[https://en.wikipedia.org/wiki/Statistical\\_data\\_type](https://en.wikipedia.org/wiki/Statistical_data_type)

# CLASSICAL PROGRAMMING VS MACHINE LEARNING

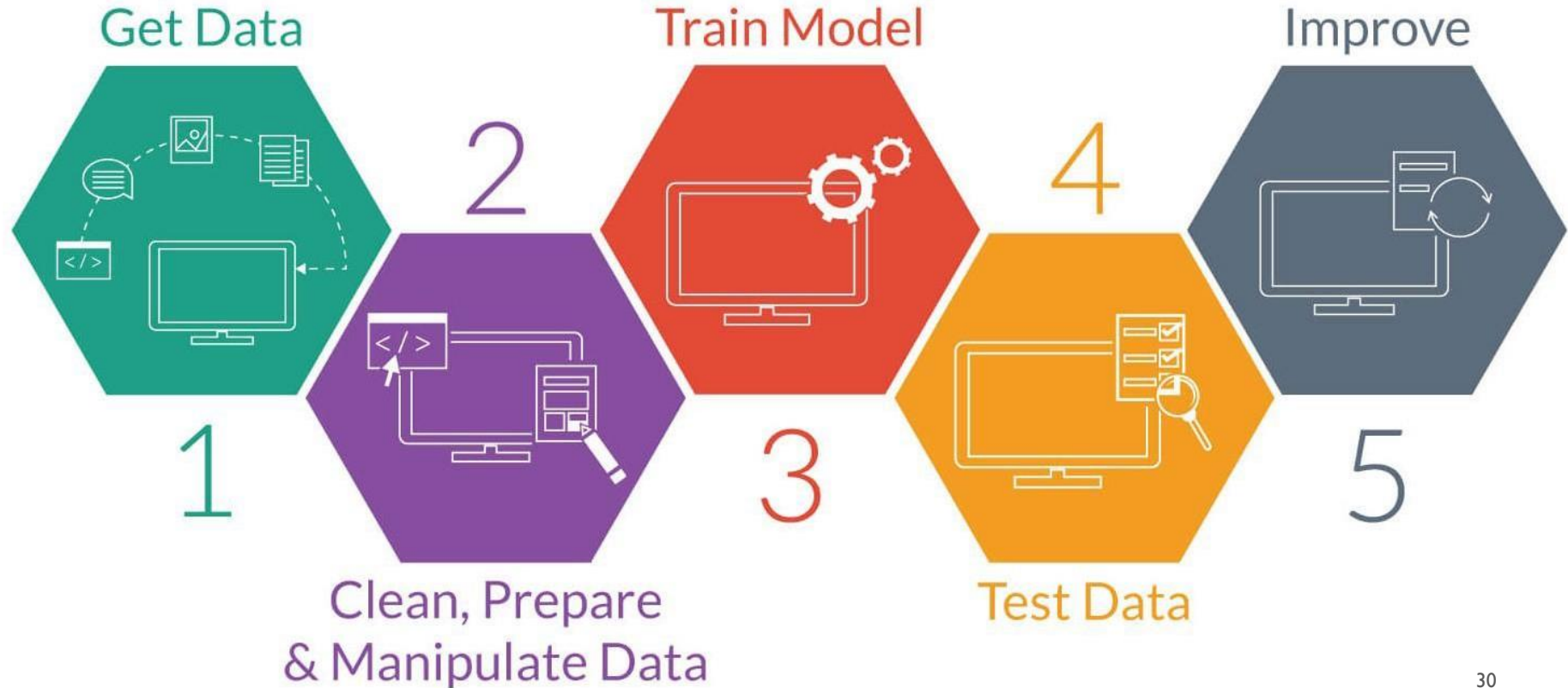


# PACKAGES FOR ML IN PYTHON



# HOW TO CHOOSE DATA TO TRAIN THE MODEL

# MACHINE LEARNING PROCESS



# FEATURES / ATTRIBUTES

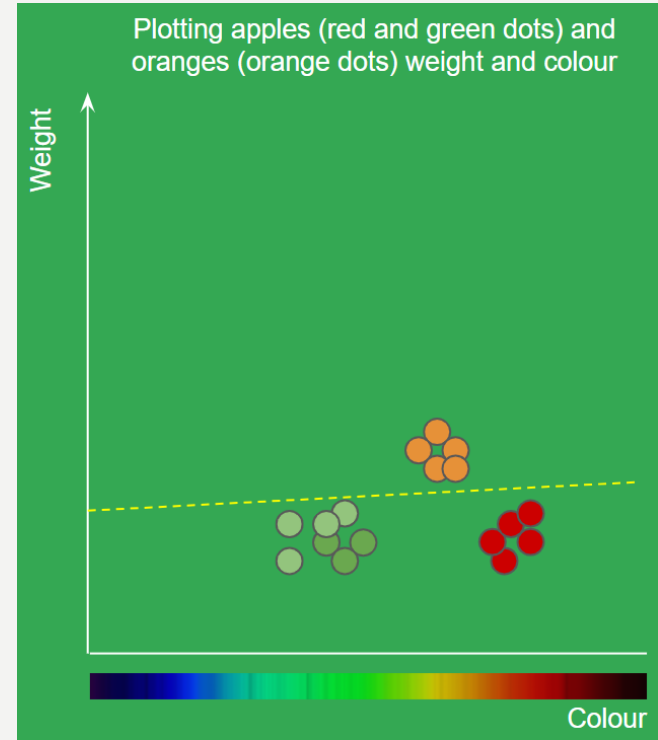
- **Features (aka attributes)** are used to train an **ML system**. They are the properties of the things you are trying to learn about.



# FEATURES / ATTRIBUTES

Taking fruit as an example. Features of a fruit might be weight and color. 2 features, would mean there are 2 dimensions. A 2D system may be plotted on a graph if features are represented in a numerical way.

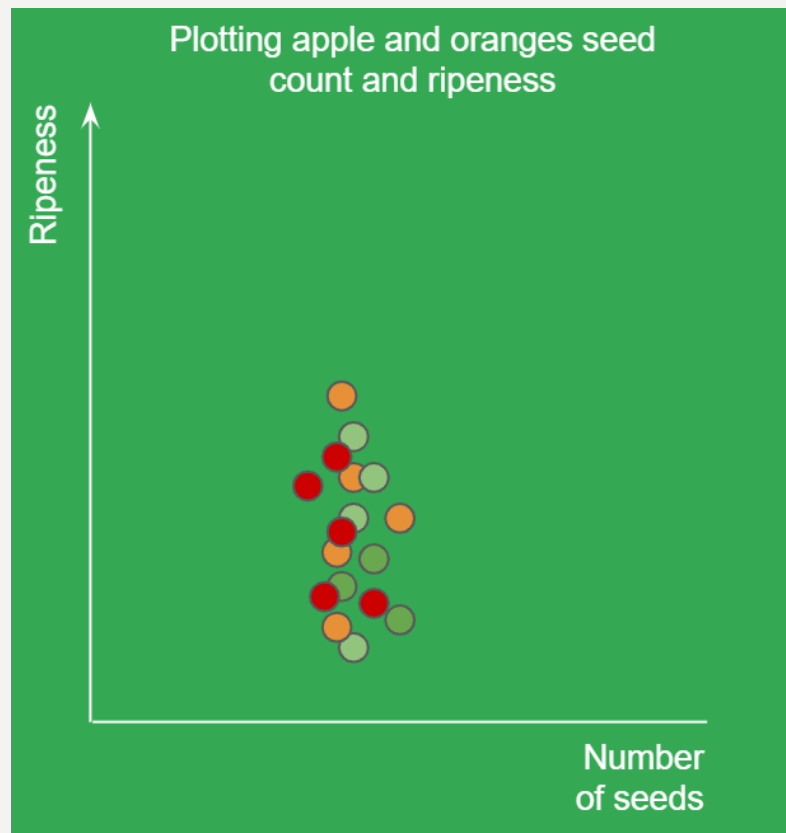
In the plot on the right, the ML system can learn to split the data up with a line to separate apples from oranges. This **can now be used to make future classifications** when we plot new points the system has not seen (anything above is orange, below is apple)





# FEATURES / ATTRIBUTES

- **Choosing useful features can have a big impact on the quality of the ML system.** Some features may not be useful enough to separate the data points.
- In this example we take bad features of fruits(ripeness and seed count) that do not allow us to learn any distinguishing factors for the fruit.



# WHAT ML CANNOT PREDICT STUFF IT DOESN'T KNOW ABOUT

Lets say you teach an ML system about animals like this:

**Number of Legs, Color, Weight, Animal:**

- 4, Black, 10KG, Dog
- 2, Orange, 5KG, Chicken

If you now present it with a Cow: 4 legs, black, 200KG it would predict “Dog”. This is because it only knows about dogs and chickens and this was the closest match.

# HOW ML SYSTEMS ARE TRAINED (LEARNING STYLE)

# MACHINE LEARNING TYPES

## Supervised Learning

- Makes machine Learn explicitly
- Data with clearly defined output is given
- Direct feedback is given
- Predicts outcome/future
- Resolves classification and regression problems



## Unsupervised Learning

- Machine understands the data (Identifies patterns/structures)
- Evaluation is qualitative or indirect
- Does not predict/find anything specific



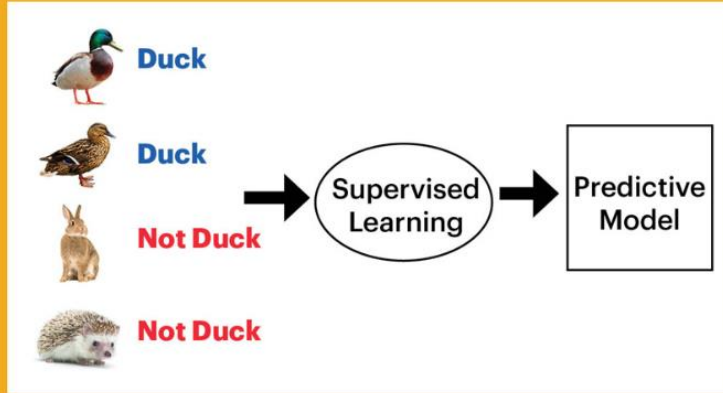
## Reinforcement Learning

- An approach to AI
- Reward based learning
- Learning form +ve & +ve reinforcement
- Machine Learns how to act in a certain environment
- To maximize rewards

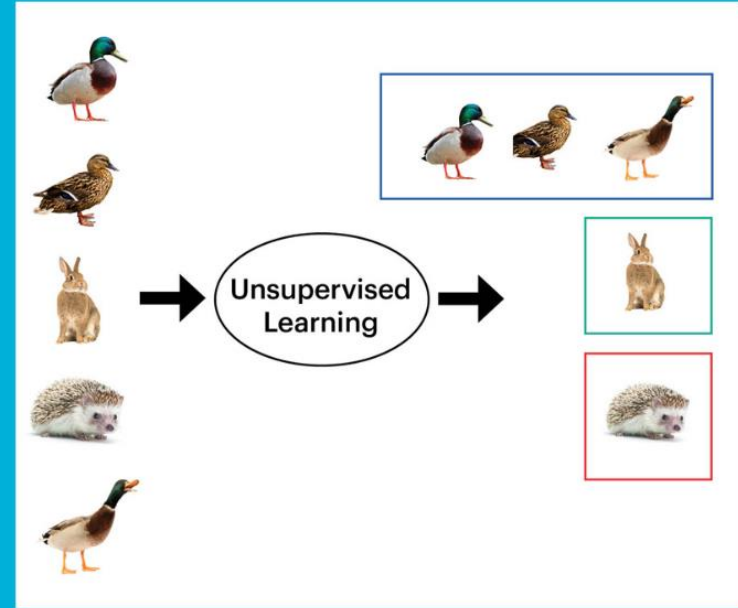


# SUPERVISED VS UNSUPERVISED

## Supervised Learning (Classification Algorithm)

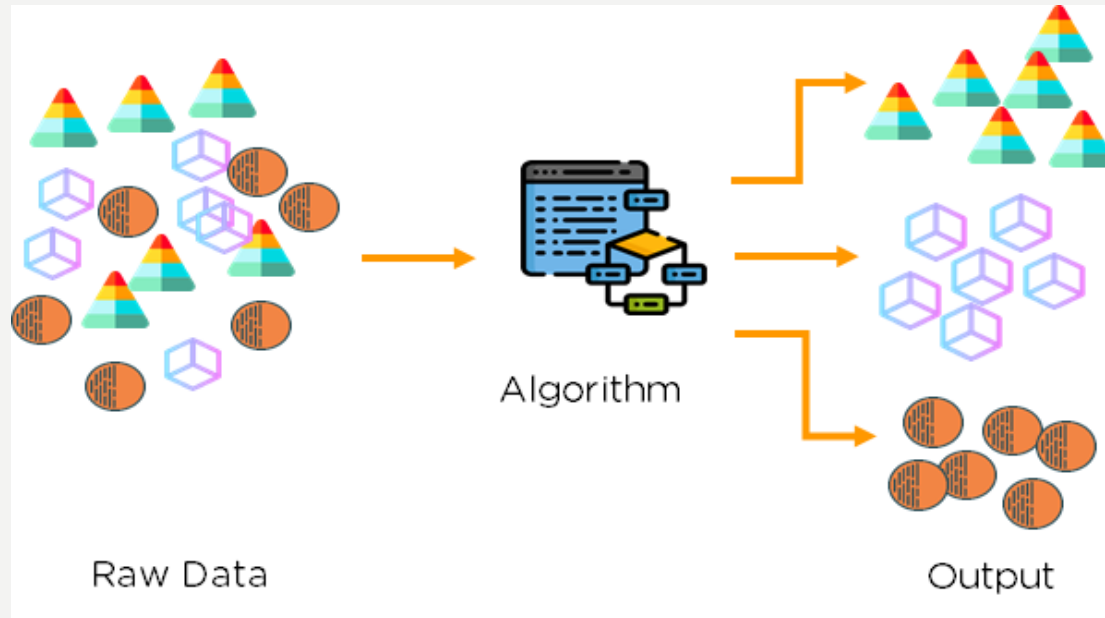


## Unsupervised Learning (Clustering Algorithm)



# UNSUPERVISED LEARNING

Unsupervised learning model learns through observation and finds structures in the data. When the model is feed data, it automatically finds patterns and relationships in the data by creating clusters in it. What it cannot do is adding labels to the cluster. Like the picture shown below.



By Anil Kumar APSSDC

# MACHINE LEARNING ALGORITHMS

## SUPERVISED

### Regression

- Linear Regression
- Polynomial Regression

### Classification

- Linear Classifiers
  - Logistic Regression
- K - Nearest Neighbors
- Decision Trees
- Random Forest
- Support Vector Machines

# CLASSIFICATION VS REGRESSION



Student Profile



*Predicting Student*  
**Pass Or Fail**



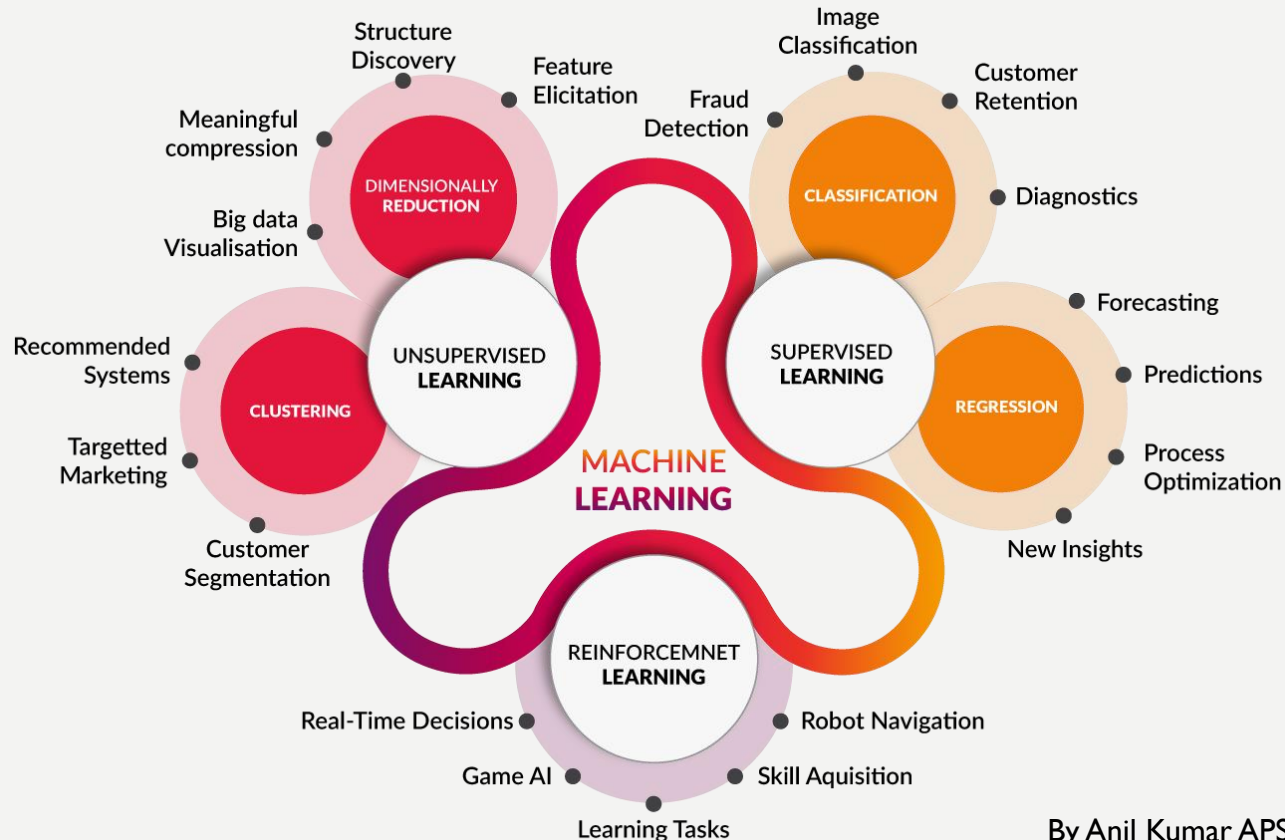
Student Profile



*Predicting Student Marks*  
**Percentage**



# MACHINE LEARNING CATEGORIES



# UNSUPERVISED

## Clustering Types

- Hierarchical clustering
- K-means clustering
- DBSCAN
- Spectral clustering

## Dimensionality Reduction

- Principal Component Analysis
- Independent Component Analysis
- randomized SVD

# CLASSIFICATION



0 0 0 0 0 0 0 0 0 0 0 0 0 0 0  
1 1 1 1 1 1 1 1 1 1 1 1 1 1 1  
2 2 2 2 2 2 2 2 2 2 2 2 2 2 2  
3 3 3 3 3 3 3 3 3 3 3 3 3 3 3  
4 4 4 4 4 4 4 4 4 4 4 4 4 4 4  
5 5 5 5 5 5 5 5 5 5 5 5 5 5 5  
6 6 6 6 6 6 6 6 6 6 6 6 6 6 6  
7 7 7 7 7 7 7 7 7 7 7 7 7 7 7  
8 8 8 8 8 8 8 8 8 8 8 8 8 8 8  
9 9 9 9 9 9 9 9 9 9 9 9 9 9 9

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# FRAUD DETECTION



# HOUSE PRICE PREDICTION





# STOCK PREDICTION





# CUSTOMER PREDICTION



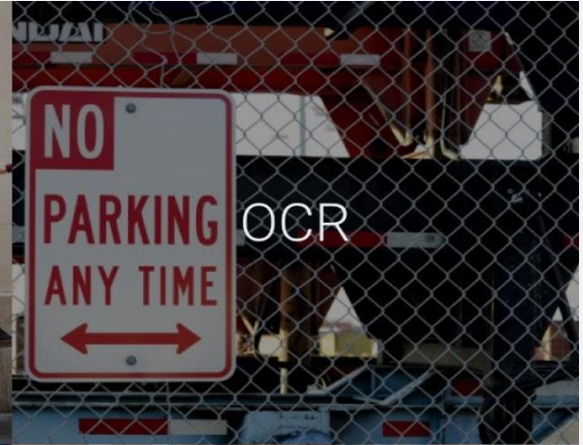




Label  
Detection



Face  
Detection



OCR



Explicit Content  
Detection



Landmark  
Detection

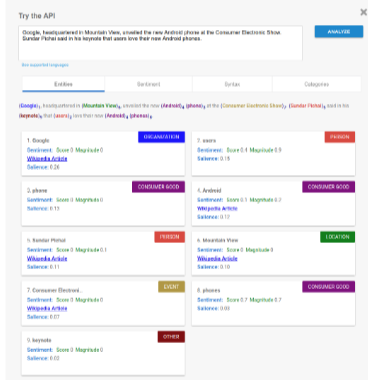


Logo Detection

Google

# Natural Language API

Derive insights from unstructured text using Google machine learning.



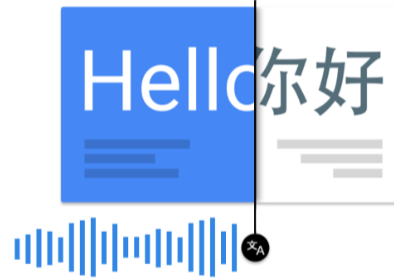
## Entity Recognition

Identify entities and label by types such as person, organization, location, events, products and media.



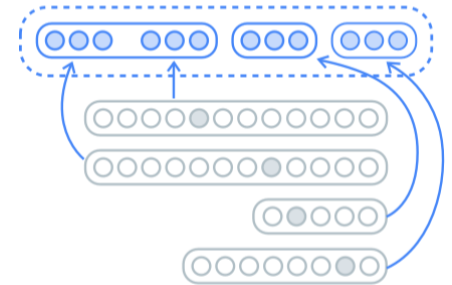
## Sentiment Analysis

Understand the overall sentiment expressed in a block of text.



## Multi-Language Support

Enables you to easily analyze text in multiple languages including English, Spanish and Japanese.



## Syntax analysis

Extract tokens and sentences, identify parts of speech (PoS) and create dependency parse trees for each sentence.

# REFERENCES

- Machine Learning in 45 minutes by Jason Mayes, Senior Creative Engineer at Google
  - **Video:** <https://www.youtube.com/watch?v=X4I9QmcSEYo>
  - **Slides:** <https://goo.gl/fGJ8HJ>