

# Machine Learning Introduction

# Definition of Learning systems

- Tom Mitchell: A Computer learns a Task T from experience E, if its performance P improves with E.

E: Experience of past election results

T: Predicting outcome of next elections for given Party Win, Loss - Classification

P: True +ve : No. of correctly predicted Wins

True -ve: No. of correctly predicted Losses

False +ve: No. of incorrectly predicted Wins

False -ve: No. of incorrectly predicted Losses

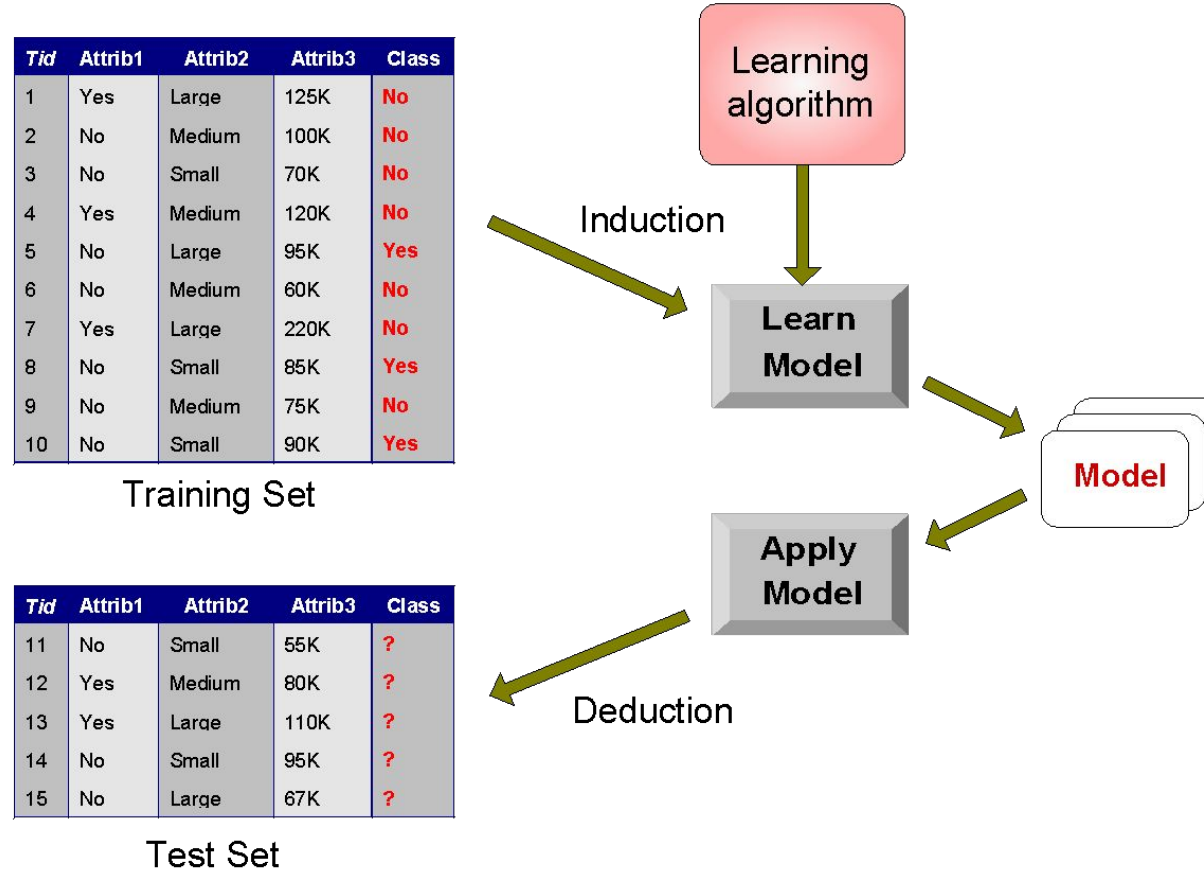
OR

P: Probability of win/loss - Regression

# Purview

- ML – **Data does all the work** to solve problems
- As a programmer, you have to **develop an ML model – find best values of its parameters**
- **Great Applications:** Image recognition, sales prediction, medical diagnosis, Recommender systems
- **And more power when combined with AI:** Robotics, space exploration, game playing, humanoid voice recognition and synthesis, machine translation

# Interaction of Data and Learning Algorithm



# Features

- Regression – Volume of sales. Features??

*Price, quality, time of sale, model of sale ....*

- Can ML deal with infinite features?

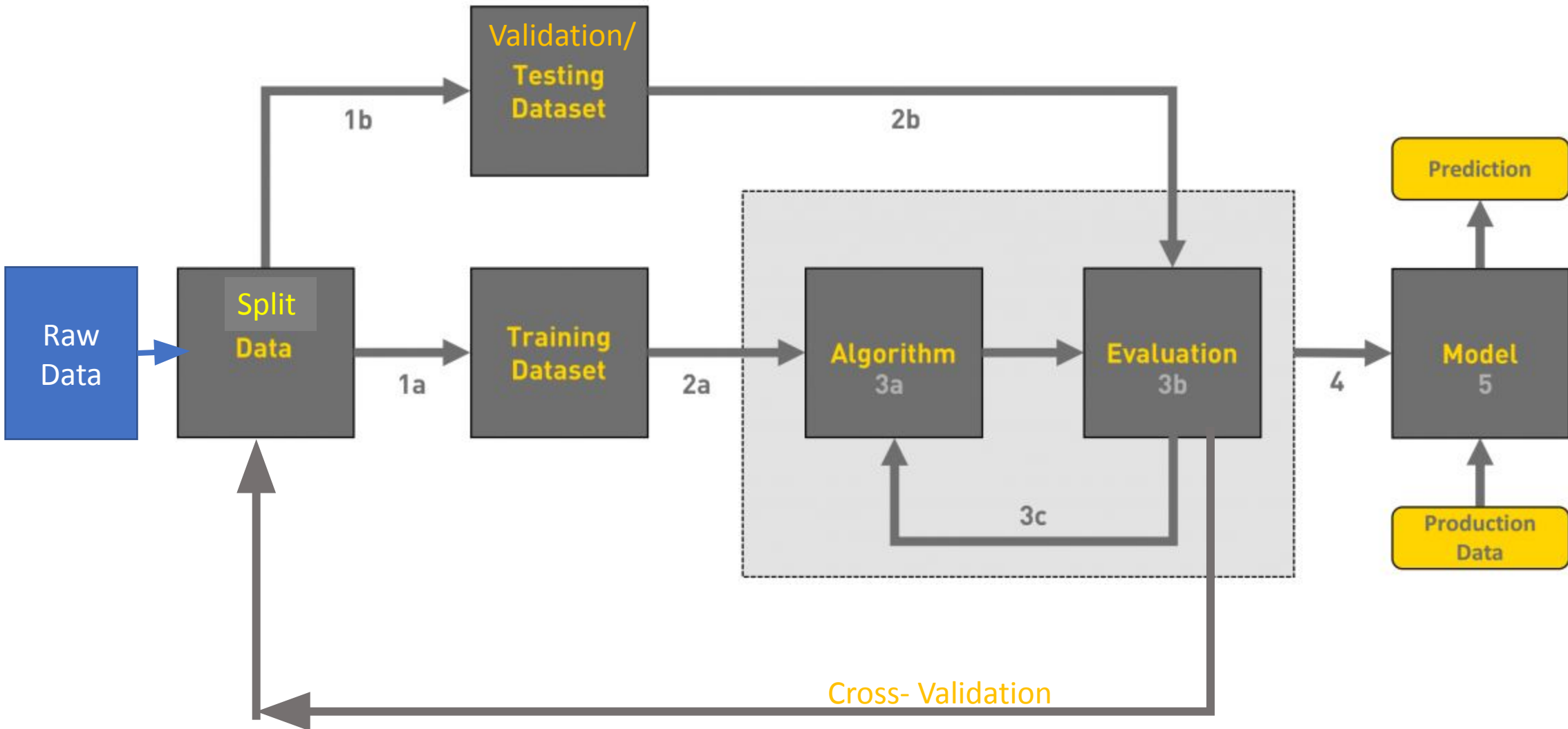
- Classification –Detection of breast cancer?

*Size of tumour, degree of malignancy, medical history, age, thickness, type of tumour, clumpiness*

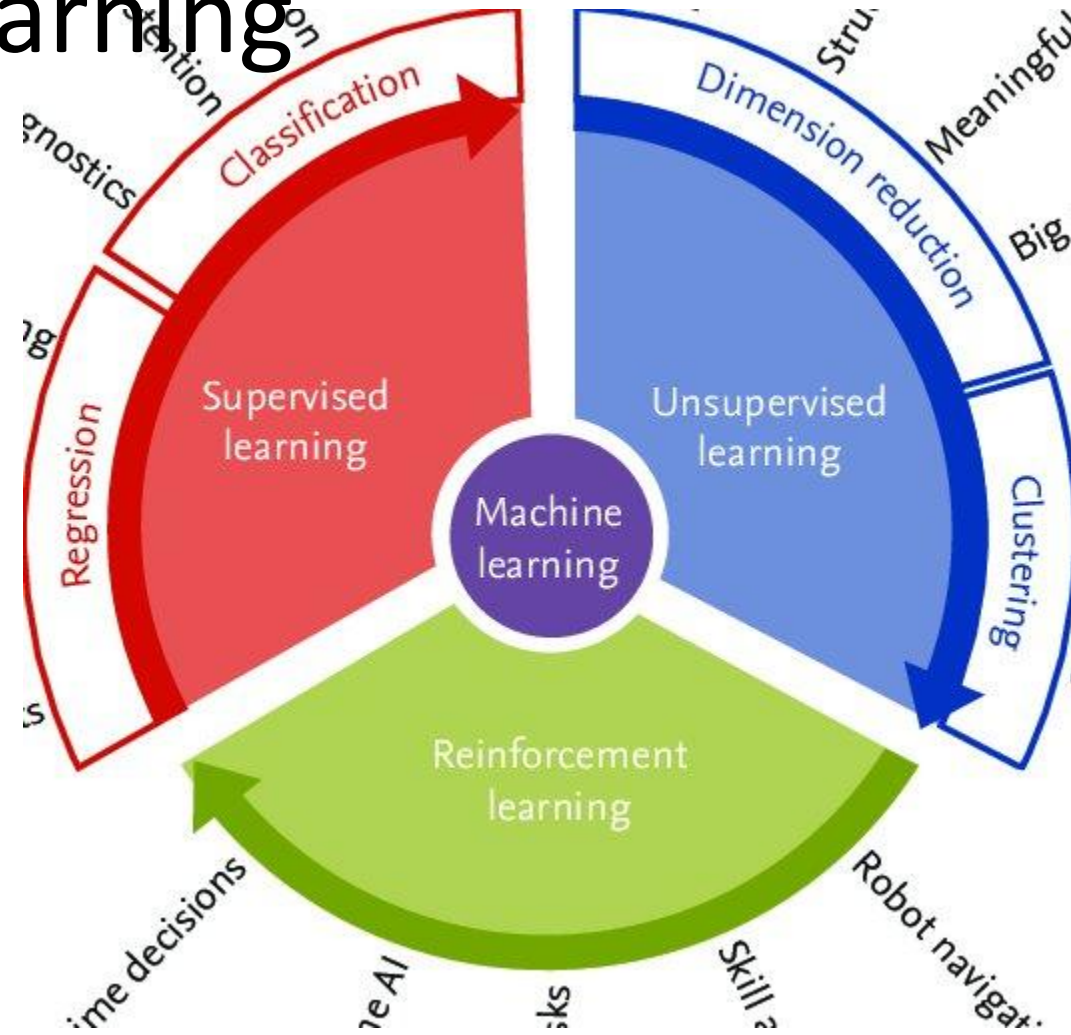
Detection of Depression?

*Age, history, suicidal thoughts, motor skills, speech, interest in activities, anxiety level, psychological parameters*

# Work-Flow of ML



# Types of Learning



# Basic types of learning

- **Supervised:** Given training dataset {features, response}, predict response for new features

Bayesian, Decision Trees, Linear and Logistic Regression, SVM, KNN, ANN, LDA, HMM, CRF

Price of mobile	Make of Mobile	Number of cameras	.....	Volume of sales
9K	Vivo	2		10K
12K	Redme	3		9K
16K	Oppo	4		12K
....	...	...		

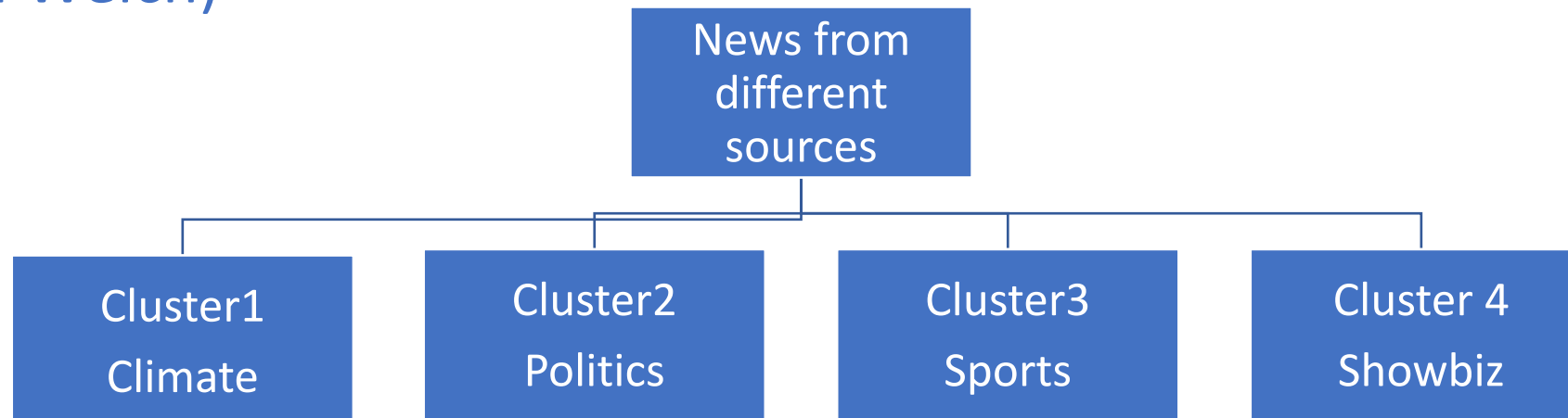
Error reduction based learning



# Unsupervised:

- Given dataset {features}, find the patterns to cluster the data

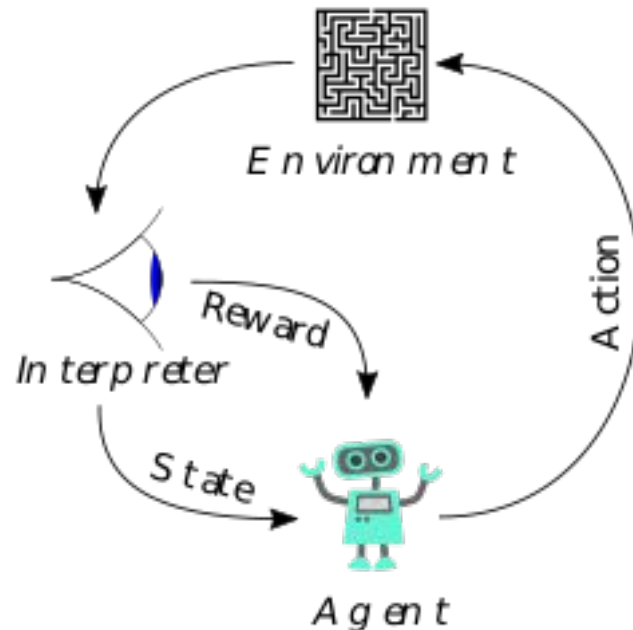
Clustering, Principal Component Analysis, Latent Semantic Analysis (LSA), Association Rules, Self-Organizing Maps (SOM), HMM Training (Baum Welch)



# Reinforcement:

- **Semi-supervised:** Starting from few examples, progressively learn
- Learn from rewards and penalties from a series of interactions with the environment

Q-Learning, Monte Carlo Methods, Markov Decision Processes



# Reinforcement Learning

- Short term v/s long term goals
- Games, Robotics
- Given Action A States S

$$P_a(S, S') = \Pr(S_{t+1} = S' | S_t = S, A_t = a)$$

$$R_a(S, S') = \text{Immediate reward after transition}$$

- (i) Model of environment is there, but analytical solution not known
- (ii) Simulation based optimization
- (iii) To collect info, agent must interact with environment
- (iv) Distributed Applications: a) Ad-hoc networks – Distributed communication b) Multi-robot action – Cooperative tasks with leader election

# The process of undertaking ML based solutions

- Understand a domain
- Incorporate prior knowledge
- Know the goals
- **Data collection, integration, cleaning, completion, pre-processing**
- **Feature Selection**
- Choose and fine tune appropriate learning model
- Interpret results □ measure final performance
- Deploy knowledge gained
- Loop