FOUNDATIONS OF MACHINE LEARNING

ANUSH SANKARAN

OVERVIEW OF THE COURSE

Week	Topics
Week 1	Intro to ML Discovering ML Use Cases & ML in Business
Week 2	Python- Hands On Supervised Learning & Regression
Week 3	Neural Network - 1 Neural Network -2 (Bias, Variance) & Hands ON
Week 4	Kernel Learning & SVM Practical Advice for ML projects.
Week 5	Boosting Decision Trees, Random Forest, & xgBoost
Week 6	Unsupervised Learning Clustering & Dimensionality Reduction
Week 7	Time Series Data Analysis Imputation & Prediction Systems
Week 8	ML Use Cases from Products & Research

COURSE OUTCOMES

- Understand the fundamental concepts of different machine learning models
 - Supervised learning
 - Unsupervised learning
- Ability to formulate a business problem as machine learning task. Identify machine learning opportunities in businesses.
- Appreciate the challenging involved in data driven machine learning problems
- Ability to manage the building of tools and products that involves different aspects of machine learning

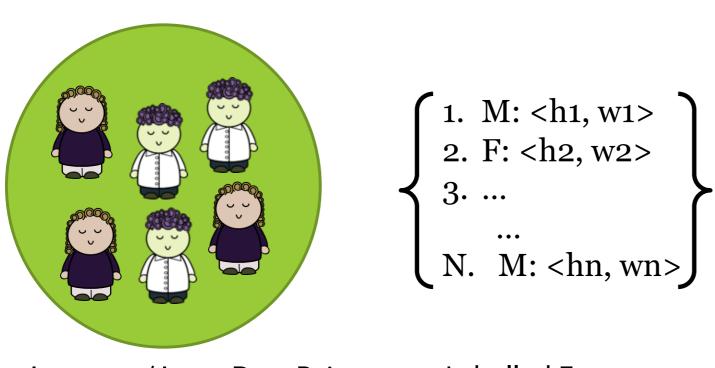
EASY LOGISTICS: GITHUB

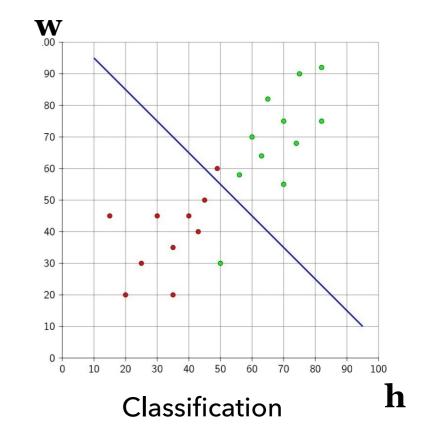
- Github Repo: https://github.com/goodboyanush/isme-bangalore-Oct-Nov-2019
- Lectures slides, Hands-on code, Assignment solutions
- Have any doubt in my lectures or assignments?
 - Go ahead and create an issue in the repo!
 - I will try to answer them asap!
 - Everyone will be benefitted by the questions asked by one

WEEK 1: INTRODUCTION TO MACHINE LEARNING BUILDING THE ML MINDSET IN BUSINESS

WHAT IS MACHINE LEARNING?

• Learn a classifier: learn a mapping function





Instances/Input Data Points

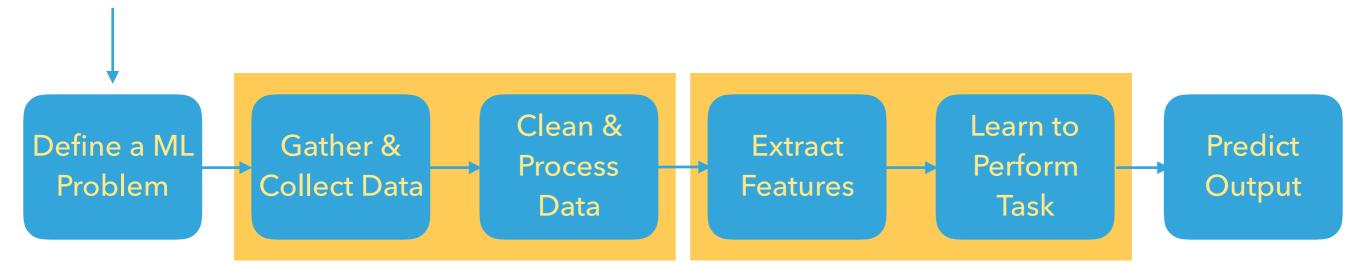
Labelled Features

- 1. **Boundary**: Can be linear or non-linear boundary
- 2. **Method**: Can be a generative classifier or discriminative classifier

Examples: Naïve Bayes, Decision trees, Neural Network, Support Vector Machines etc.

MACHINE LEARNING PIPELINE

What are we focussing on today's lecture?



- 1. Articulate the problem (task)
- 2. Data Drive Strategy: Look for labelled data
- 3. Design your data for the task
- 4. Determine easily obtained inputs
- 5. Determine easily quantifiable outputs

COMMON LINGO

Input data/ Features

Output Task Label

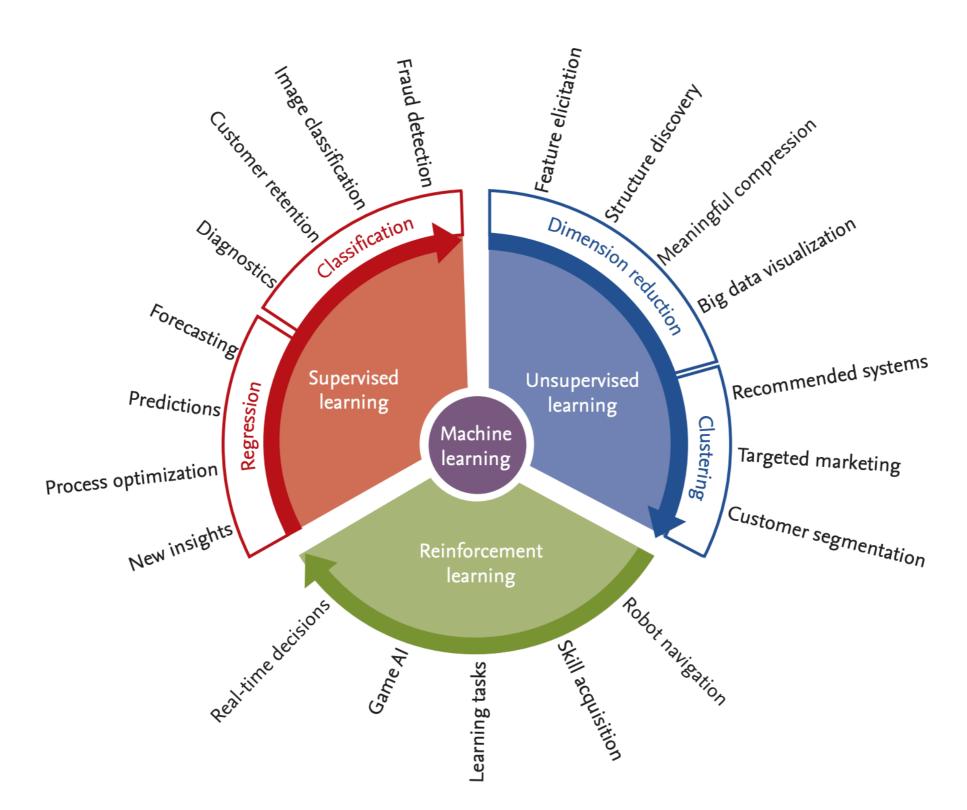
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LP001005	Male	Yes	0	Graduate	Yes	3000	0	66	360	1	Urban	Y
LP001006	Male	Yes	0	Not Graduate	No	2583	2358	120	360	1	Urban	Y
LP001008	Male	No	0	Graduate	No	6000	0	141	360	1	Urban	Y
LP001011	Male	Yes	2	Graduate	Yes	5417	4196	267	360	1	Urban	Y
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LP001014	Male	Yes	3+	Graduate	No	3036	2504	158	360	0	Semiurban	N
LP001018	Male	Yes	2	Graduate	No	4006	1526	168	360	1	Urban	Y
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LP001029	Male	No	0	Graduate	No	1853	2840	114	360	1	Rural	N
LP001030	Male	Yes	2	Graduate	No	1299	1086	17	120	1	Urban	Y
LP001032	Male	No	0	Graduate	No	4950	0	125	360	1	Urban	Υ
LP001036	Female	No	0	Graduate	No	3510	0	76	360	0	Urban	N

Seen data / Training data

Unseen data/Test data

LP001038 Ma	ale	Yes	0 Not Graduate	No	4887	0	133	360	1 Rural
LP001043 Ma	ale	Yes	0 Not Graduate	No	7660	0	104	360	0 Urban
LP001046 Ma	ale	Yes	1 Graduate	No	5955	5625	315	360	1 Urban
LP001047 Ma	ale	Yes	0 Not Graduate	No	2600	1911	116	360	0 Semiurban

DIFFERENT TYPES OF ML ALGORITHMS



THE ML MINDSET

"Machine Learning changes the way you think about a problem.

The focus shifts from a mathematical science to a natural science, running experiments and using statistics, not logic, to analyse its results."

Peter Norvig

THE ML MINDSET

Step	Example
1. Set the research goal.	I want to predict how heavy traffic will be on a given day.
2. Make a hypothesis.	I think the weather forecast is an informative signal.
3. Collect the data.	Collect historical traffic data and weather on each day.
4. Test your hypothesis.	Train a model using this data.
5. Analyze your results.	Is this model better than existing systems?
6. Reach a conclusion.	I should (not) use this model to make predictions, because of X, Y, and Z.
7. Refine hypothesis and repeat.	Time of year could be a helpful signal.

Get Comfortable with Some Uncertainty!

REGRESSION - LINGO

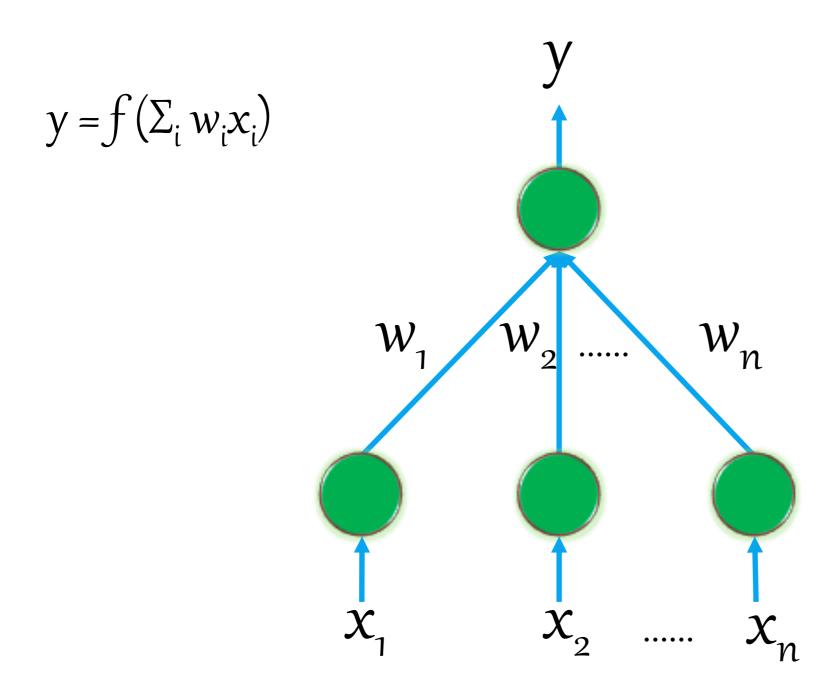
$$Y = X_1 + X_2 + X_3$$

Dependent Variable Independent Variable

Outcome Variable Predictor Variable

Response Variable Explanatory Variable

LINEAR REGRESSION - ACTIVATION FUNCTION

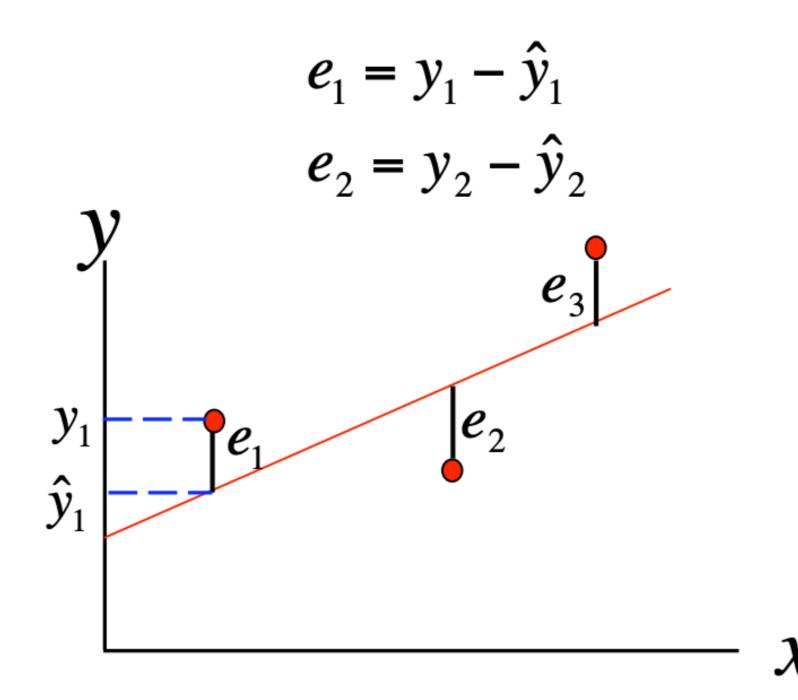


$$Error = (y - y')^2$$

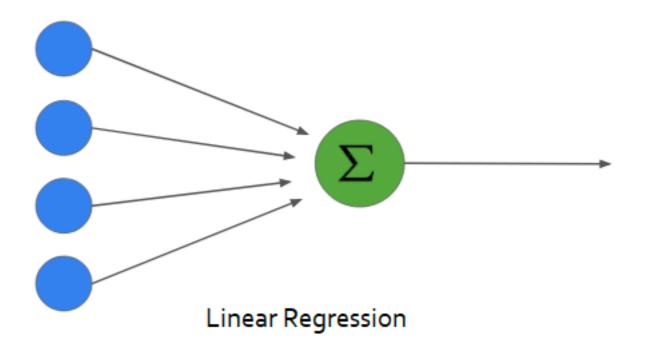
Residual Sum of Squares (RSS)

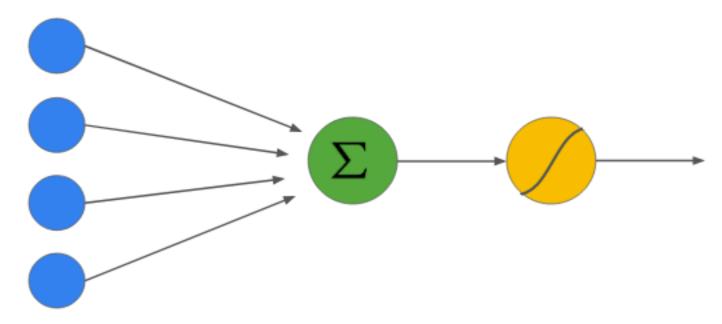
Update w in such a way that the error is minimized!

RESIDUAL SUM OF SQUARES



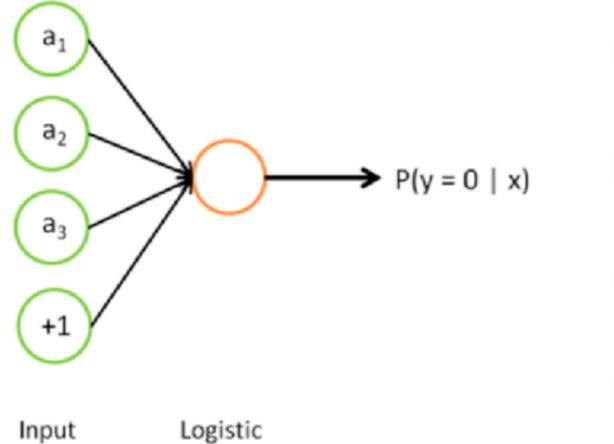
LINEAR VS. LOGISTIC REGRESSION





Logistic Regression

NEURAL NETWORKS

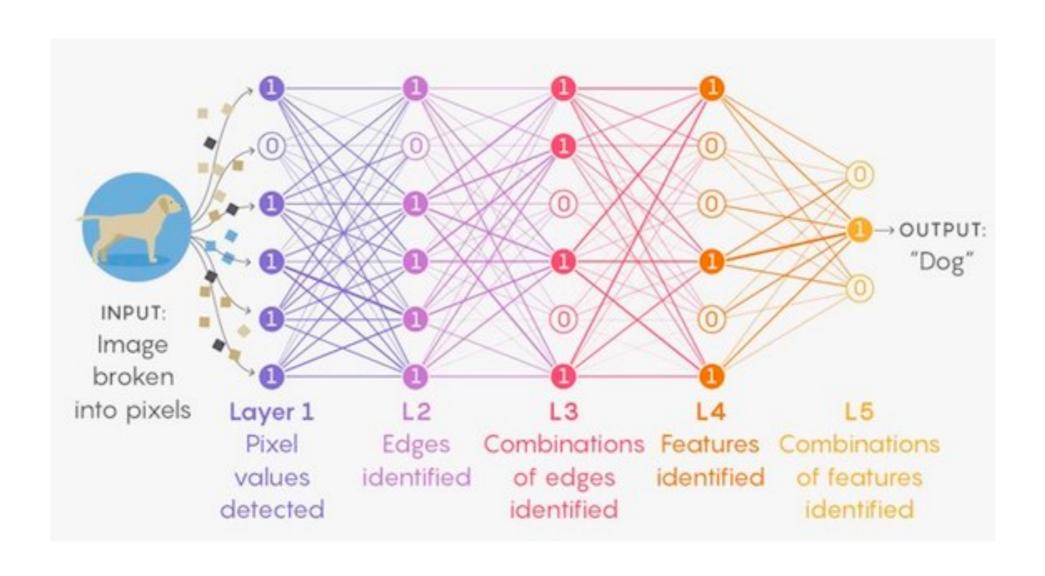


Input Logistic (features) classifier x_1 x_2 $a_2^{(2)}$ $a_3^{(2)}$ $a_3^$

Logistic Regression

Neural Network

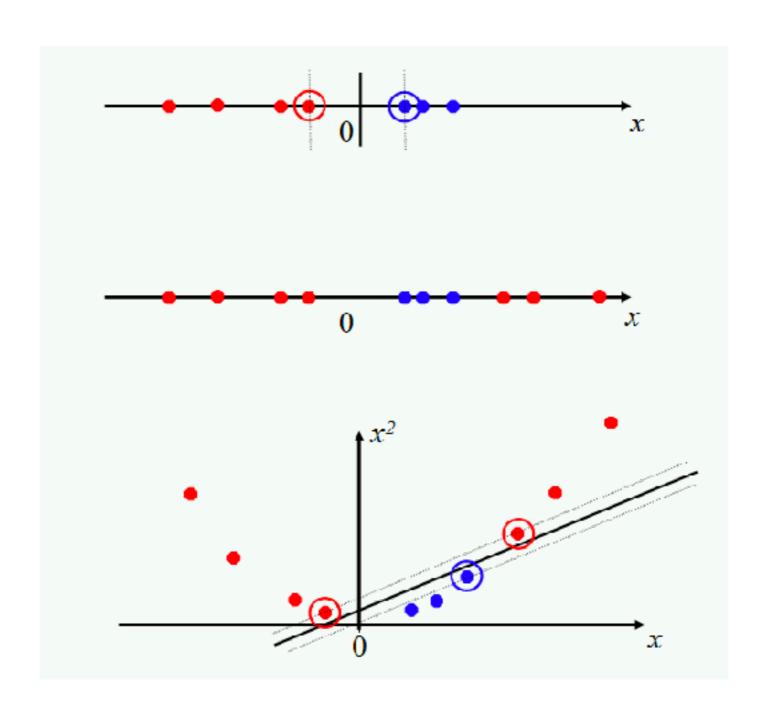
NEURAL NETWORKS



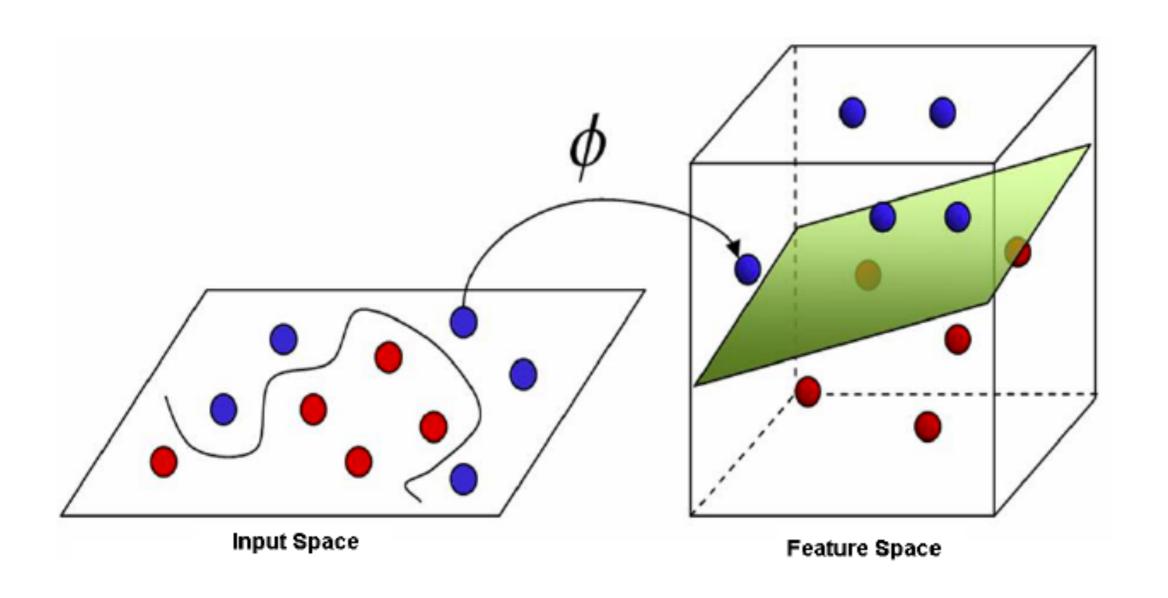
THANK YOU - NEXT WEEK

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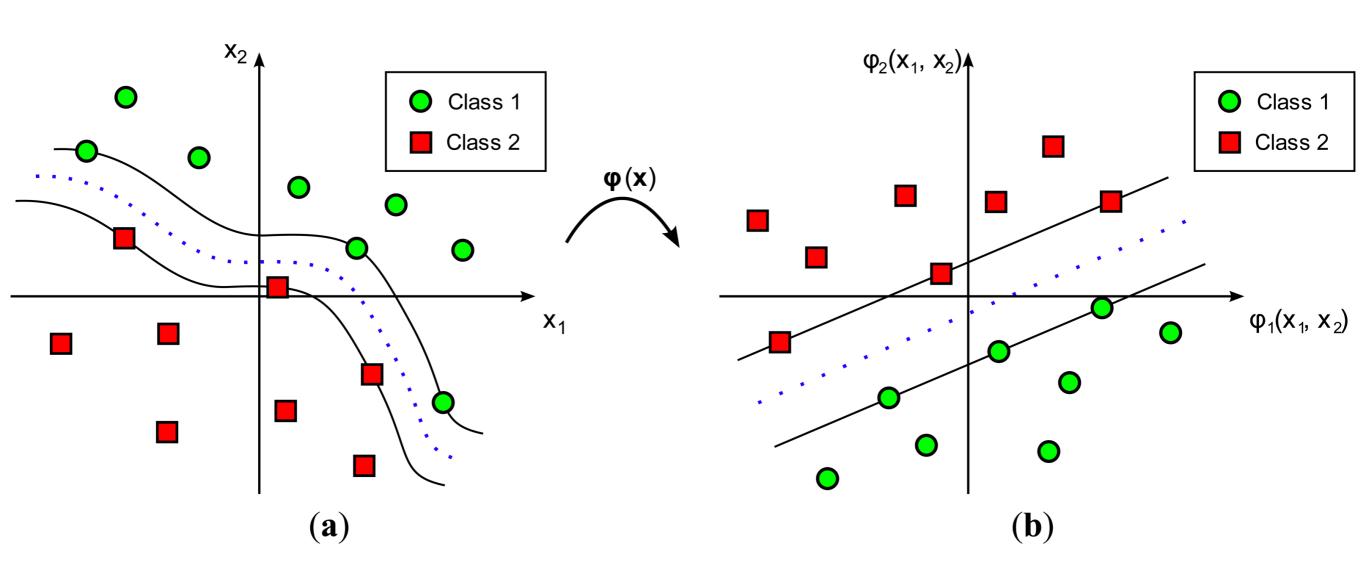
KERNEL TRICK



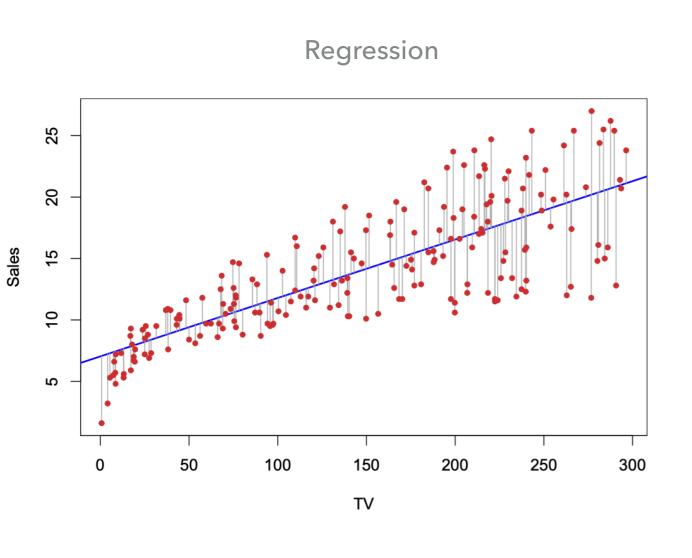
KERNEL LEARNING

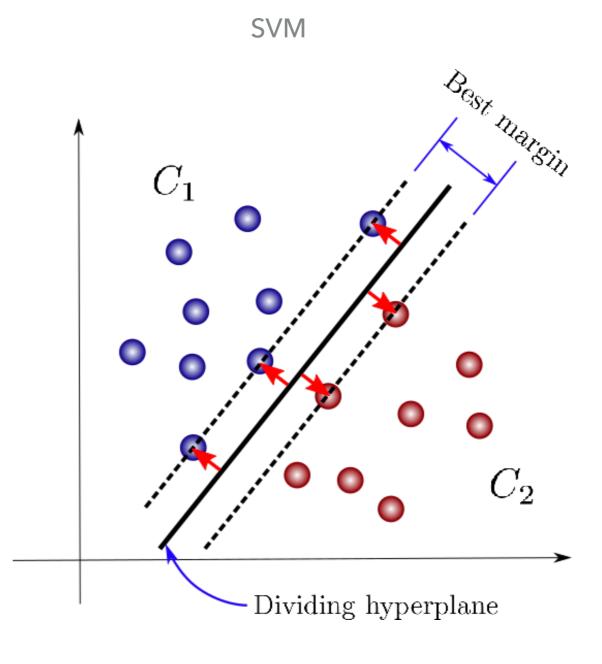


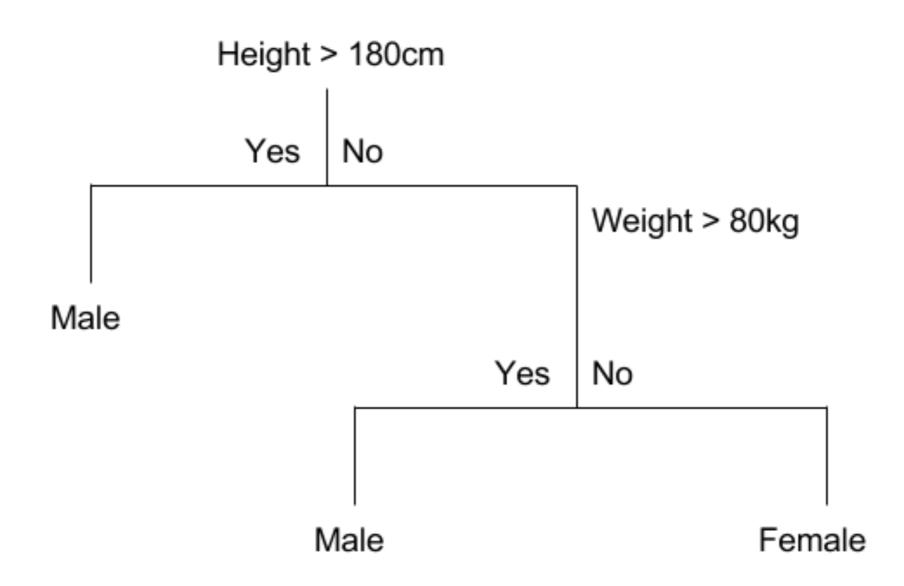
KERNEL LEARNING



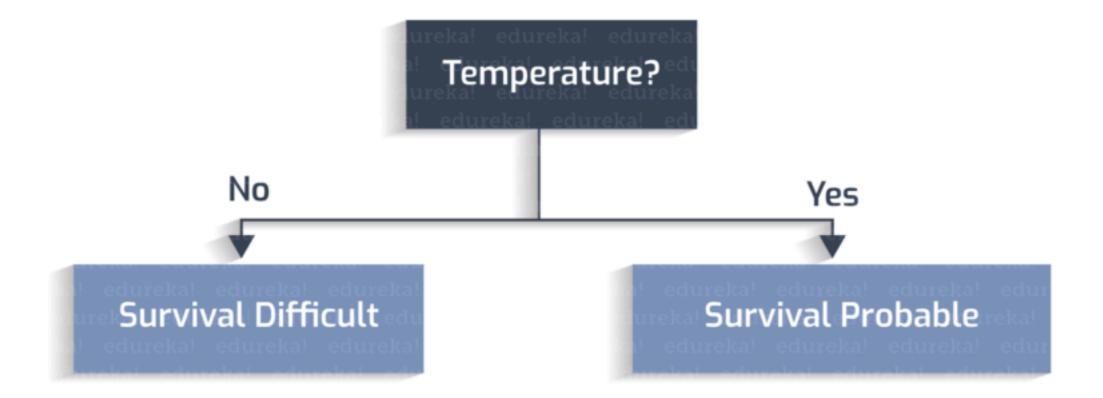
SUPPORT VECTOR MACHINES







edureka!



edureka!

