

# *E9 205 Machine Learning for Signal Processing*

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## Introduction to Machine Learning of Sensory Signals

05-09-2019

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**Teaching Assistant** - Prachi Singh ([prachisingh@iisc.ac.in](mailto:prachisingh@iisc.ac.in)).

**Class Location** - EE B308

**Web** - [http://leap.ee.iisc.ac.in/sriram/teaching/MLSP\\_19/](http://leap.ee.iisc.ac.in/sriram/teaching/MLSP_19/)

**Timings** - MW 330-500pm. *Fridays (Tentative) 8-9 pm.*





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# Overview

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- ❖ What are the typical real-world signals
- ❖ What is learning
- ❖ Why should we attempt learning of such signals
- ❖ Roadmap of the course



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# Real World Signals

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- ❖ Signal in general is a function  $f : X \rightarrow V$
- ❖ Real World Signals
  - ❖ which we see everyday everywhere
  - ❖ Text, Speech, Image, Videos...
  - ❖ DNA sequence, financial data, weather parameters, neural spike train...
  - ❖ Belonging to / generated by certain category of events.



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# Real World Signals

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- ❖ Types of signals- Continuous and Discrete
- ❖ Observations from real world signals
  - ❖ Information may not be uniform.
  - ❖ Cannot be modeled deterministically.
  - ❖ Affected by noise, sensing equipments.
  - ❖ Missing or hidden variables.



# Real World Signals - Examples

- ❖ Text data
  - ❖ Discrete sequence of items

In the last 29 years, sir has never ever said 'well played' to me because he thought I would get complacent and I would stop working hard.

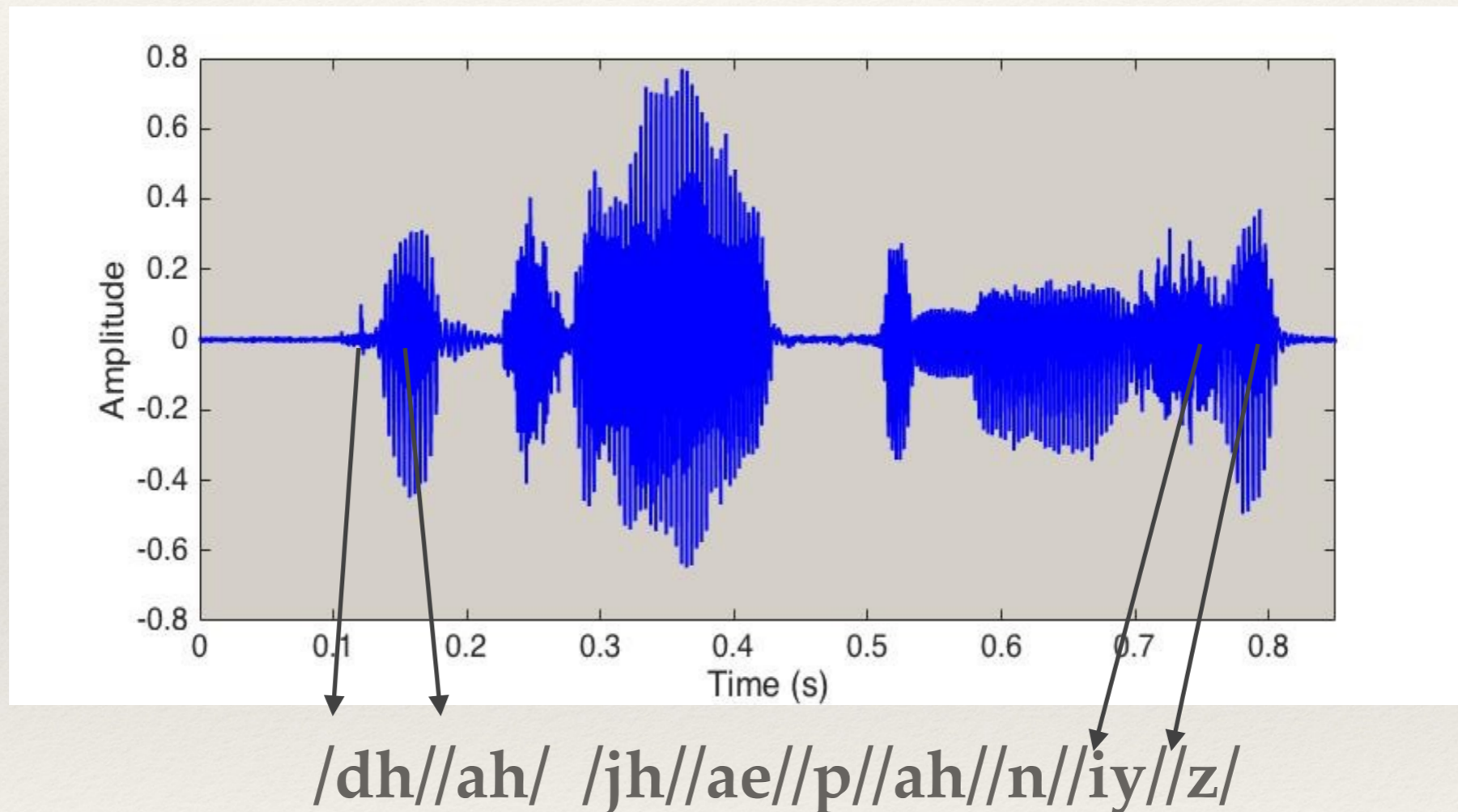
Items - [In] [the] [last] [29] [years] .....

- ❖ Some items carry more **importance** than others.



# Real World Signals - Examples

- ❖ Speech data



- ❖ Phonetic units - underlying hidden variables.



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# Real World Signals - Examples

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- ❖ Images



- ❖ Measurement artifacts - noise.



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# Patterns in Real World Signals

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- ❖ Patterns in real world signals
  - ❖ Caused by various generation processes in the real-world signals.
  - ❖ Hidden from the observation.
  - ❖ Value patterns and geometric patterns.
  - ❖ May be hierarchical in nature.
  - ❖ Manifested as pure patterns or transformed / distorted versions.



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# What is Learning

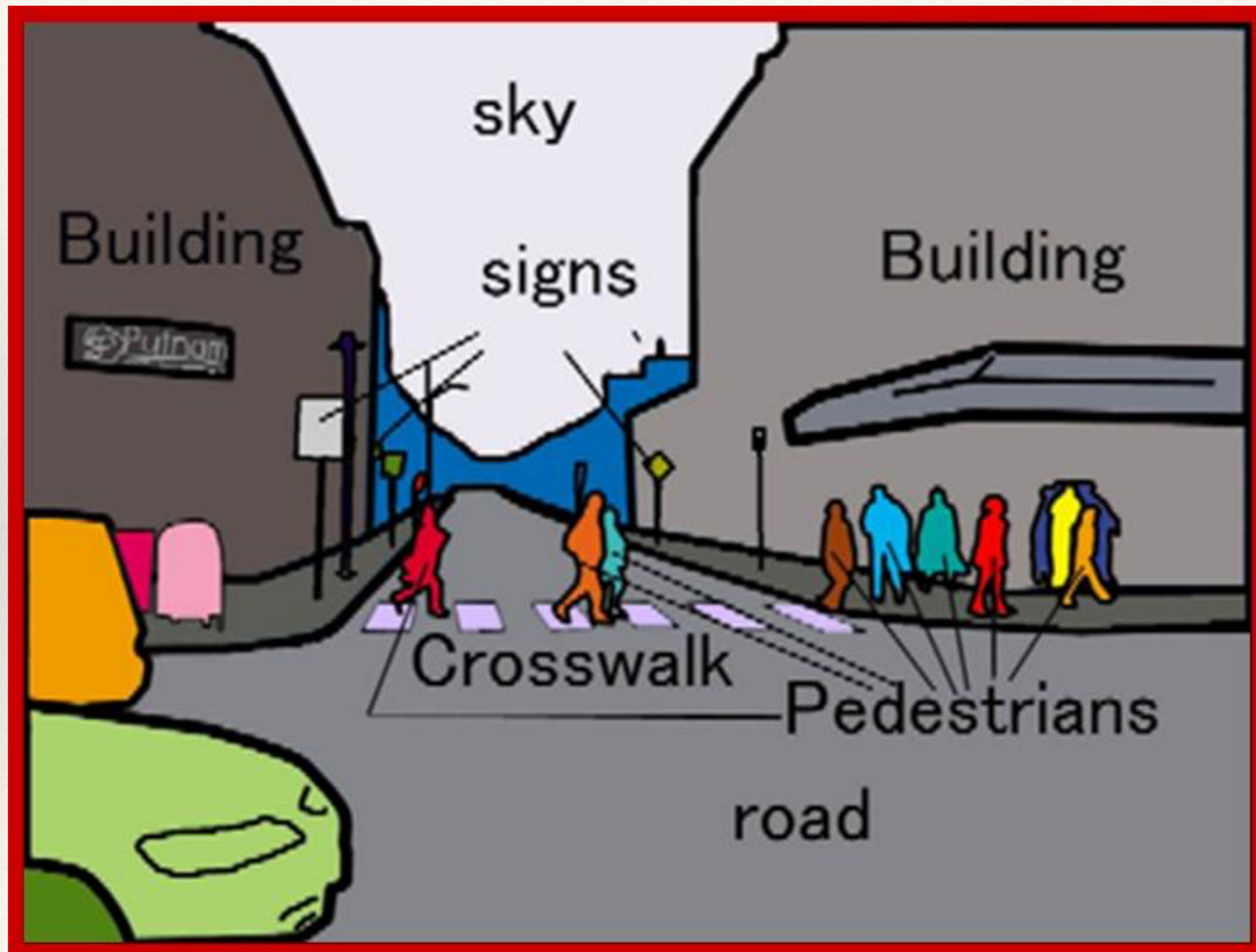
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- ❖ Learning
  - ❖ Process of describing or uncovering the pattern.
  - ❖ Understanding the physical process of generation.
  - ❖ Generalization for prediction, classification, decision making.
  - ❖ Using the data to learn the underlying pattern.
- ❖ Humans are **fundamentally trained** to learn and recognize patterns.



# What is Learning

Object  
Recognition





# What is Learning

Facial Identification



Topic Summarization

The Karnataka government is planning to start an aviation school to help students from lower economic and rural backgrounds become pilots.



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# Machine Learning

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- ❖ Machine Learning
  - ❖ Automatic discovery of patterns.
  - ❖ Motivated by human capabilities to process real world signals.
  - ❖ Mimicking / Extending / Replacing human functions.
  - ❖ Branch of artificial intelligence.
  - ❖ Classification and Regression.



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# Machine Learning - Examples

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## Domain Identification - Blog v/s Chat ?

“I tried these Butterscotch Muffins today and they turned out so good. I had half the pack of butterscotch chips that I bought long back so wanted to use it up.”

"Hey, it's Geoff from yesterday. How's it going?  
Hi there. Don't wanna bother you long, but  
you saw this video?"



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# Machine Learning - Examples

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Did a Human or Machine write this ?

“A shallow magnitude 4.7 earthquake was reported Monday morning five miles from Westwood, California, according to the U.S. Geological Survey. The temblor occurred at 6:25 AM, Pacific time at a depth of 5.0 miles.”

“Kitty couldn’t fall asleep for a long time. Her nerves were strained as two tight strings, and even a glass of hot wine, that Vronsky made her drink, did not help her. Lying in bed she kept going over and over that monstrous scene at the meadow.”

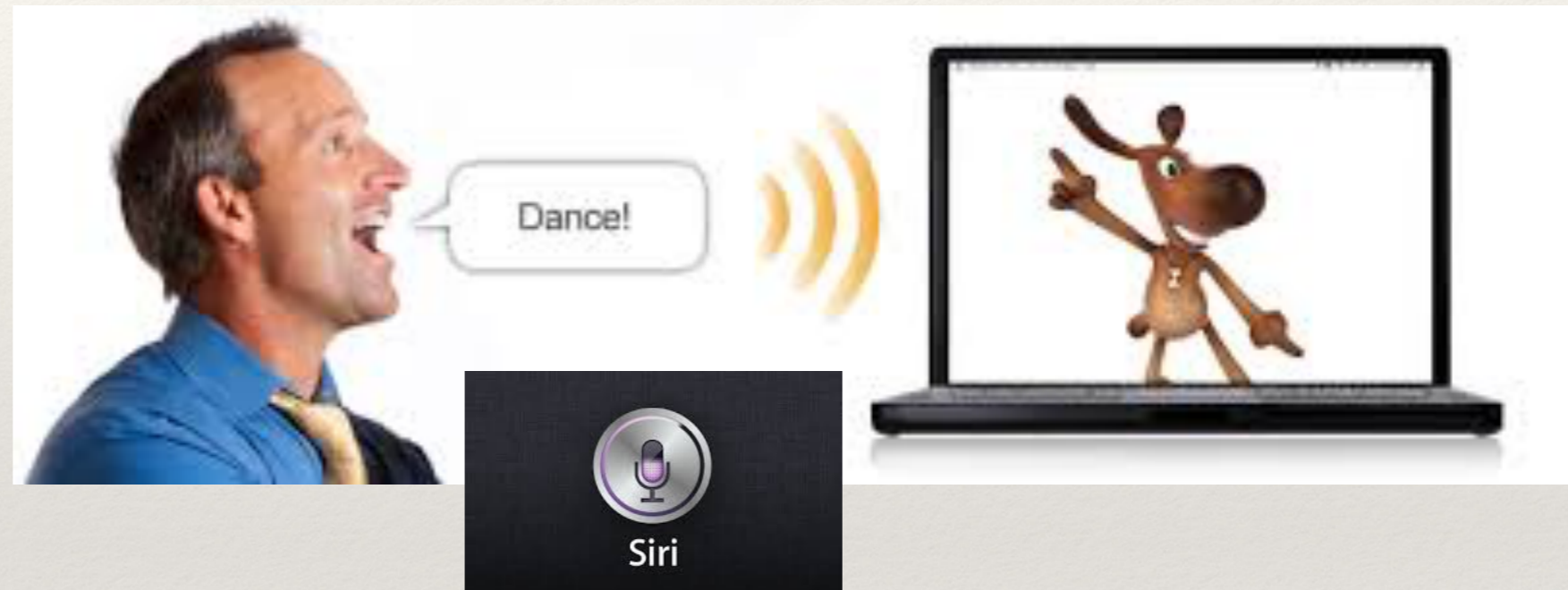
<http://www.nytimes.com/interactive/2015/03/08/opinion/sunday/algorithm-human-quiz.html>





# Machine Learning - Examples

## Speech Recognition



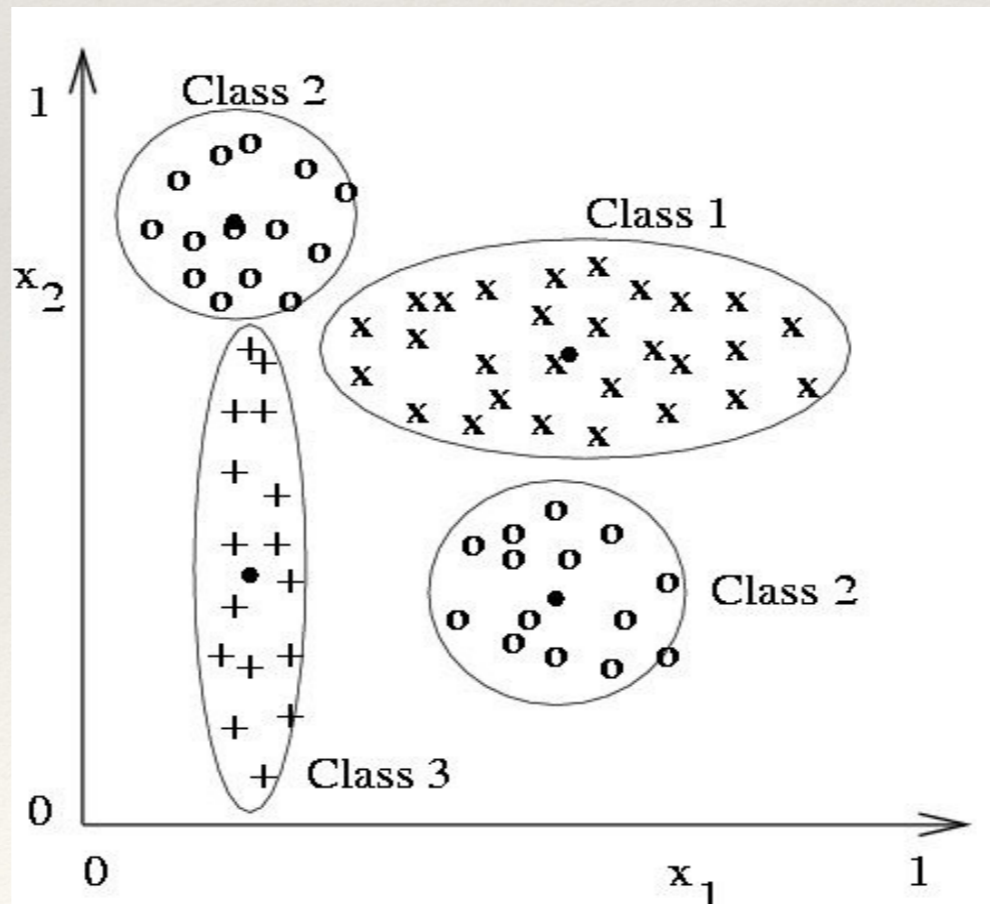
## Sound Synthesis

<http://news.mit.edu/2016/artificial-intelligence-produces-realistic-sounds-0613>



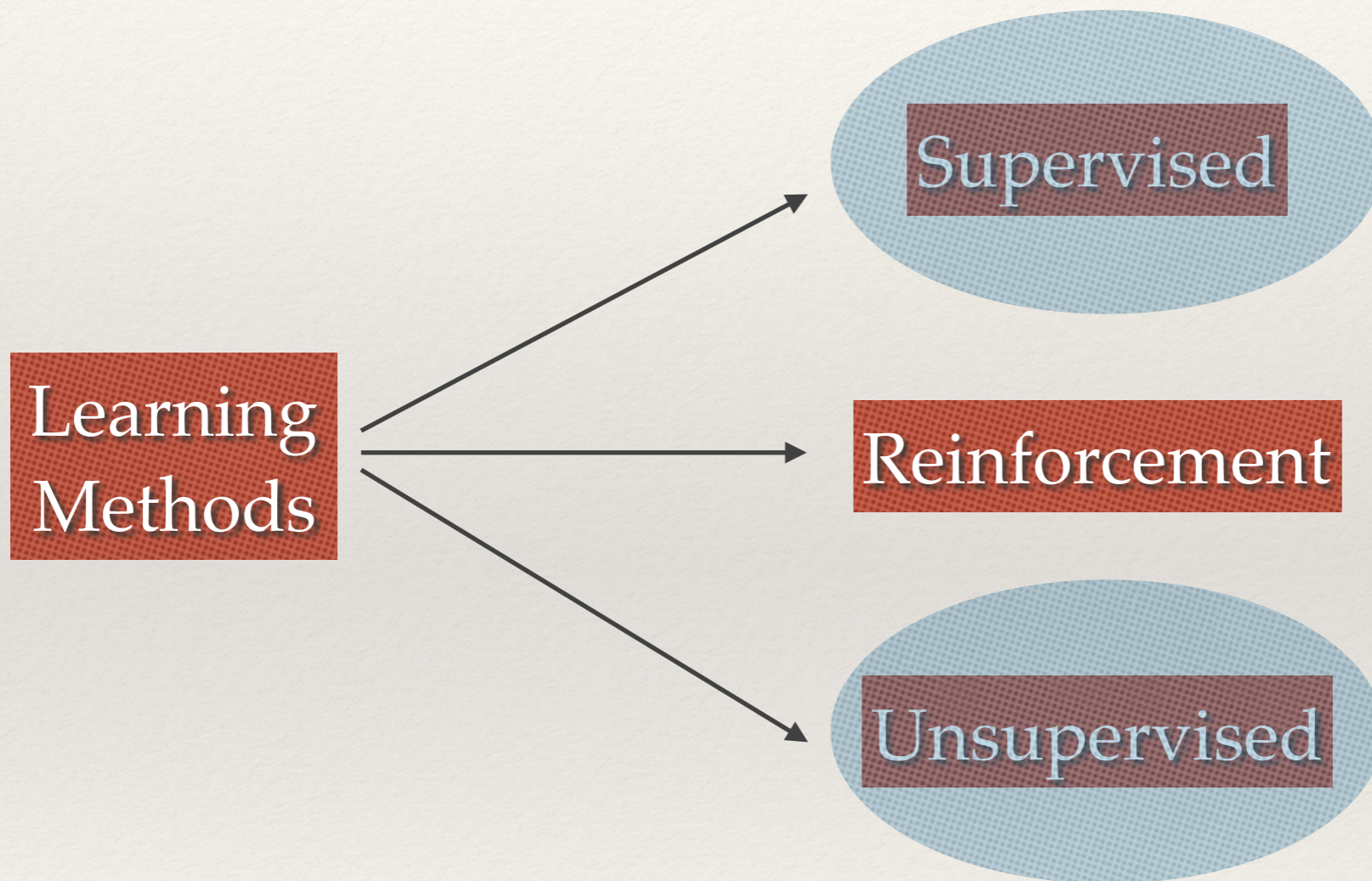
# Machine Learning

- ❖ Traditional approaches to Machine Learning
  - ❖ Rule and heuristic based methodologies
  - ❖ Using small amounts of data.
- ❖ Recently, most problems are addressed as statistical pattern recognition problem with big data.





# Types of Learning





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# Unsupervised Learning

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- ❖ Data is presented without associated output targets
- ❖ Extracting structure from the data.
- ❖ Examples like clustering and segmentation.
- ❖ Concise description of the data - dimensionality reduction methods.



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# Reinforcement Learning

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- ❖ Dynamic environment resulting in triplets - state / action / reward.
- ❖ No optimal action for a given state
- ❖ Algorithm has to learn actions in a way such the expected reward is maximized over time.
- ❖ May also involve minimizing punishment.
- ❖ Reward / punishment could be delayed - learning based on past actions.



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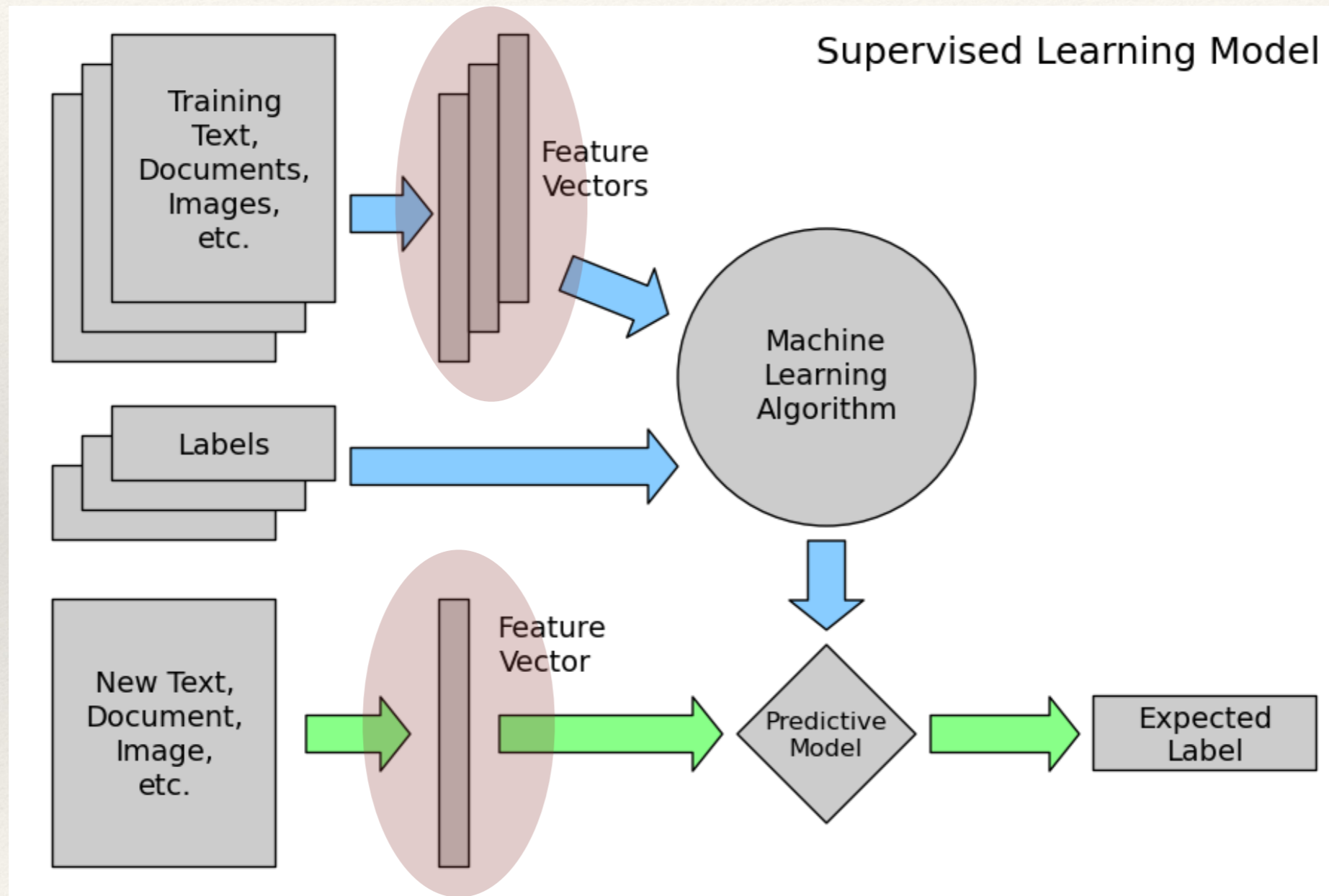
# Supervised Learning

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- ❖ Training data is provided with along with target values (ground truth).
  - ❖ Goal - to learn the mapping function from data to targets.
  - ❖ Use the mapping function to predict unseen / test data samples.
- ❖ Two types based on the structure of the labels.
  - ❖ Classification - discrete number of classes or categories.
  - ❖ Regression - continuous output variables.

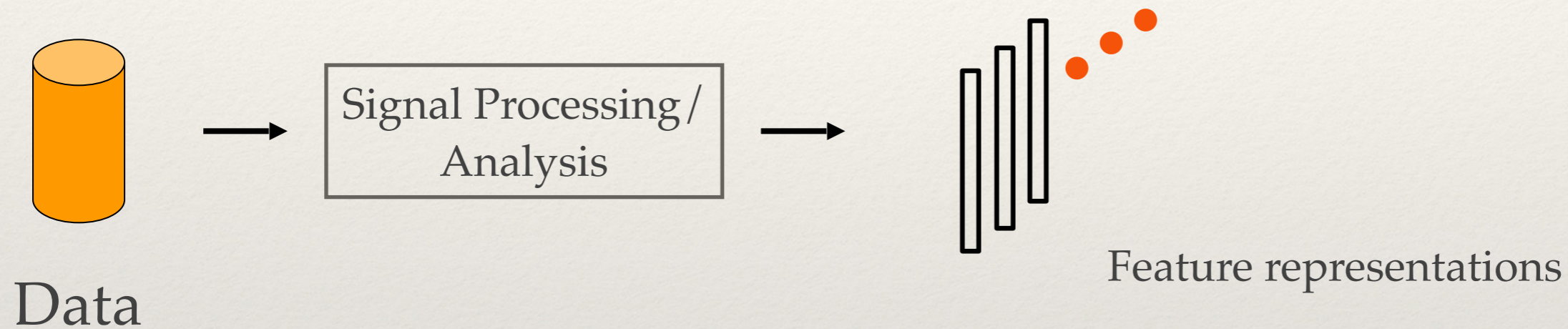


# Supervised Learning





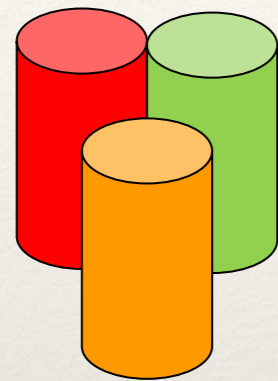
# Course Roadmap



- ❖ Feature Extraction from Text, Speech, Image / Video signals (first 3 lectures).



# Course Roadmap



Data Set

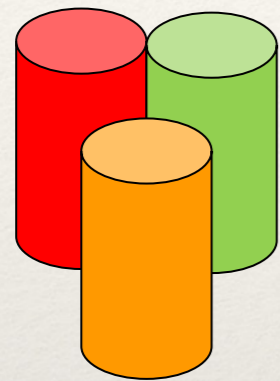
→ Features →

Models for Pattern Recognition

- ❖ Between features and pattern recognition
- ❖ Feature selection, dimensionality reduction.
- ❖ Representation learning.



# Course Roadmap



→ Features →

Models for Pattern  
Recognition

Data Set

- ❖ Modeling the generation of data
  - ❖ Gaussian, Mixture Gaussian, Hidden Markov Models etc.
- ❖ Modeling the separation of data
  - ❖ Support Vector Machines, Deep Neural Networks etc.



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# Course Structure (Rough Schedule)

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- ❖ Signal analysis and processing (1st week)
  - ❖ Text Features, Audio / Speech - spectrograms, Image Features.
- ❖ Basics of Pattern Recognition (2nd week).
  - ❖ Dimensionality reduction, factorization and feature selection.
- ❖ Generative modeling (next 2 weeks)
  - ❖ Gaussian and mixture Gaussian modeling, factor analysis models.
- ❖ Discriminative modeling - Support vector machines (next 2 weeks)
- ❖ Deep Learning (next 6-7 weeks)
- ❖ Unsupervised learning from Deep Models (last 3 weeks)



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# Housekeeping

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- Requisite**
- ❖ Must
    - ❖ Probability / Random process / Stochastic Models
    - ❖ Linear Algebra / Matrix Analysis
  - ❖ Preferred
    - ❖ Intro to Signal Processing
  - ❖ Preferred
    - ❖ Coding in Python
- Grading**
- ❖ Assignments - Theory + Implementation (20%)
  - ❖ Mid-terms (20%)
  - ❖ Project (25%)
  - ❖ Finals (35%)



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# Housekeeping

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## Project and Coding Assignments

- ❖ Coding and submissions
  - ❖ Preferred Language - Python.
- ❖ In class demos and example recipes in python.
- ❖ Final Project - GPU platform can be setup

## Resources

- ❖ Textbooks -
  - ❖ PRML (Bishop), NN (Bishop).
  - ❖ Deep Learning (Goodfellow)
- ❖ Online resources (papers and other textbooks listed in webpage).

Course Webpage

[www.leap.ee.iisc.ac.in/sriram/teaching/MLSP\\_19](http://www.leap.ee.iisc.ac.in/sriram/teaching/MLSP_19)





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# Dates of Various Rituals

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- ❖ 5 Assignments spread over 3 months (roughly one assignment every two weeks).
- ❖ September 1st week - project topic announcements.
- ❖ September 3rd week - 1st Midterm
- ❖ September 4th week - project topic and team finalization and proposal submission. [1 and 2 person teams].
- ❖ October 1st week - Project Proposal
- ❖ October 3rd week - 2nd MidTerm
- ❖ November 1st week - Project MidTerm Presentations.
- ❖ December 1st week - Final Exams
- ❖ December 2nd week - Project Final Presentations.



# Content Delivery

Theory  
and Mathematical  
Foundation

Implementation  
and Understanding

Intuition and  
Analysis

Lecture and  
Beyond

- ❖ Teaching Assistant - Prachi Singh
- ❖ Additional lecture slot on Friday (time ?)
- ❖ Industry research lectures (1-2)



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# Housekeeping

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**No Class on 07-08-2019**

**However, we will meet  
on 12-8-2019 at 330pm.**

**Questions/Comments ?**