

Machine Learning

EE514 – CS535

Overview

Zubair Khalid

School of Science and Engineering Lahore University of Management Sciences

https://www.zubairkhalid.org/ee514 2021.html





Zoom Policies

- Please use Zoom 'raise hand' feature if you have any questions.
- You will be asked to unmute yourself so that you can ask questions or provide your response.
- Please keep your microphone mute at all other times.
- We do not require you to turn-on your video.
- You can also post questions in the Chat. We may respond to the questions offline.

Zoom sessions will be recorded for offline viewing. Please let us know if you have any issues due to the privacy concerns.



About us!



Zubair



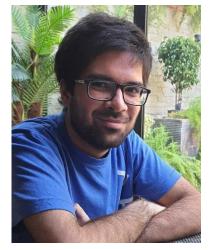




Taimoor







Omer



Mariyam

Alishba



Haseeb

What is this course about?

Introductory course in Machine Learning (ML) – Fundamental topics in

- Supervised learning
- Unsupervised learning

Course Objectives:

- To provide a thorough introduction to ML methods
- To build mathematical foundations of ML and provide an appreciation for its applications
- To provide experience in the implementation and evaluation of ML algorithms
- To develop research interest in the theory and application of ML



Is this course a right choice for you?

Undergraduate students

- Interested in pursuing AI, Deep Learning and/or Machine Learning in their grad school
- Interesting in pursuing a professional career focused on the development of Machine Learning solutions

Graduate students

- Want to do fundamental research in the area of Machine Learning
- Wish to apply Machine Learning in their research work



Course Prerequisites

Undergraduate students

- Linear Algebra (MATH120)
- Probability (MATH230, DISC203, CS501)
- Programming (CS200, EE201)

Graduate students

- Encouraged to revise Linear Algebra and Probability concepts (on-the-fly)

We expect all the students to have good programming skills (in C/Python/MATLAB)

Note on Assignment O!



Learning Interface

Delivery of Content:

<u>Mode:</u> Hybrid (Asynchronous/synchronous) <u>Lectures:</u>

- Pre-recorded (uploaded on YouTube)
- Lectures for the week will be uploaded by Wednesday

Utilization of Lecture-slots:

- Tutorial in the Friday slot (@3:30 PM)
- Quiz (15-20 minutes)



Learning Interface

Communication:

<u>Course Page:</u> https://www.zubairkhalid.org/ee514_2021.html <u>Zoom:</u> Same link for office hours, live sessions, meetings <u>Slack:</u> Course-related questions or discussions. We will try to respond to the queries ASAP. <u>Office Hours:</u> Posted on course page; distributed throughout the week

<u>Email Policy:</u>

Subject:

- 'ML-URGENT-Assignment Clarification'
- 'ML-NOT URGENT-Extend Assignment deadline'

Please **do not** flood my inbox to verify whether I have received your submission in LMS or the submission is late due to last-minute connectivity issues.



Grading Distribution

- Programming Assignments and Homeworks: 30%
 - 5 Programming Assignments
 - 3 Homeworks
- Quizzes: 20% (Almost every week)
- Mid-exam or Mid-Viva: 15%
- Project: 10%
- Final Exam: 25%



Course Polices

- Homework Late Policy
 - 10% per day for 3 days. No submission after 3 days (72 hours)
- Missed Quiz Policy
 - No make-up for quiz
- Plagiarism will be strictly dealt with as per university policies (take it seriously).
- Zero Tolerance for Plagiarism and Cheating
- Re-grading can be requested after grade reporting, within the following time limits:
 - HW and Assignments: 2 days
 - Final Exam: 3 days



Course Polices

Harassment Policy

Harassment of any kind is **unacceptable**, whether it be sexual harassment, online harassment, bullying, coercion, stalking, verbal or physical abuse of any kind. Harassment is a very broad term; it includes both direct and indirect behaviour, it may be physical or psychological in nature, it may be perpetrated online or offline, on campus and off campus. It may be one offense, or it may comprise of several incidents which together amount to sexual harassment. It may include overt requests for sexual favours but can also constitute verbal or written communication of a loaded nature. Further details of what may constitute harassment may be found in the LUMS Sexual Harassment Policy, which is available as part of the university code of conduct.

LUMS has a Sexual Harassment Policy and a Sexual Harassment Inquiry Committee (SHIC). Any member of the LUMS community can file a formal or informal complaint with the SHIC. If you are unsure about the process of filing a complaint, wish to discuss your options or have any questions, concerns, or complaints, please write to the Office of Accessibility and Inclusion (OAI, <u>oai@lums.edu.pk</u>) and SHIC (<u>shic@lums.edu.pk</u>) —both of them exist to help and support you and they will do their best to assist you in whatever way they can.

To file a complaint, please write to <u>harassment@lums.edu.pk</u>.

Course Polices

Help related to equity and Belonging at SSE

SSE's Council on Equity and Belonging is committed to devising ways to provide a safe, inclusive, and respectful learning, living, and working environment for its students, faculty, and staff.

For help related to any such issue, please feel free to write to any member of the school council for help or feedback.

Mental Health Support at LUMS

For matters relating to counselling, kindly email <u>student.counselling@lums.edu.pk</u>, or visit <u>https://osa.lums.edu.pk/content/student-counselling-office</u> for more information.

You are welcome to write to me or speak to me if you find that your mental health is impacting your ability to participate in the course. However, should you choose not to do so, please contact the Counselling Unit and speak to a counsellor or speak to the OSA team and ask them to write to me so that any necessary accommodations can be made.





Course Overview, notation

Supervised Learning Setup

<u>Weeks:</u> 1,2

Components:

• Programming Assignment 1: Intro to Python, Setting up Environment



KNN			
Evaluation N	letrics, Curse of Dimension	ality	

<u>Weeks:</u> 3,4

Components:

- Programming Assignment 2: KNN based (Using Images)
- Homework 1A



Linear Regression
Gradient Descent
Multi-variate Regression
Polynomial Regression
Bias-Variance Trade-off, Regularization

<u>Weeks:</u> 4,5

Components:

- Programming Assignment 3: Regression
- Homework 1B



Logistic Regression

Weeks: 6

Components:

• Programming Assignment 4: Logistic Regression

5 – Bayesian

Framework

Bayes Theorem

Naive Bayes Classification

<u>Weeks:</u> 7,8

Components:

- Programming Assignment 5: Naïve Bayes Classifier (may be merged with Assignment 4)
- Homework 2

6 – Perceptron, SVM and Neural Network

SVM

Neural Networks

<u>Weeks:</u> 9,10,11,12

(Mid-Exam)

Components:

- Programming Assignment 6: Neural Networks
- Homework 3

7 – Clustering

Unsupervised Learning Overview

Clustering (k-means)

<u>Weeks:</u> 13,14

Components:

• Homework 3

Suggested Reference Books



- (CB) Pattern Recognition and Machine Learning, Christopher M. Bishop
- (KM) Machine Learning: a Probabilistic Perspective, Kevin Murphy
- (TM) Machine Learning, Tom Mitchell
- (HTF) The Elements of Statistical Learning: Data mining, Inference, and Prediction, by Hastie, Tibshirani, Friedman
- (DM) Information Theory, Inference, and Learning Algorithms, David Mackay



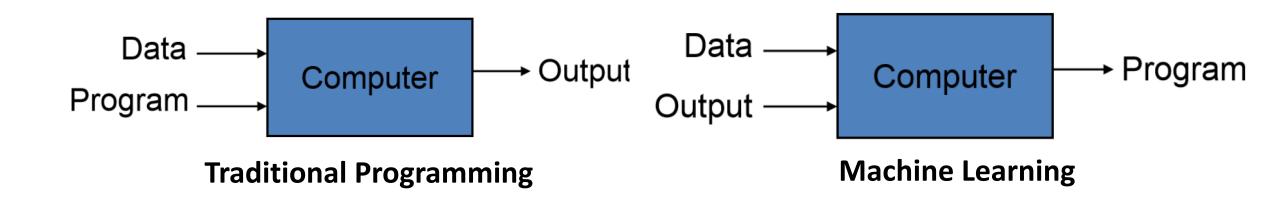
"As to methods, there may be a million and then some, but principles are few. The man who grasps principles can successfully select his own methods."

Ralph Waldo Emerson



What is Machine Learning?

- Automating the process of automation
- Getting computers to program themselves



Given examples (training data), make a machine learn system behavior or discover patterns



Classical Example: Recognize hand-written 2!

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Example Applications

- Medical Diagnosis
- Autonomous Driving
- Information extraction
- Computer/Machine Vision
- Finance
- Web Search
- Robotics
- Social networks
- Production Industry
- Logistics
- Waste Management
- [Your research/favorite area]

Face Recognition: Demo

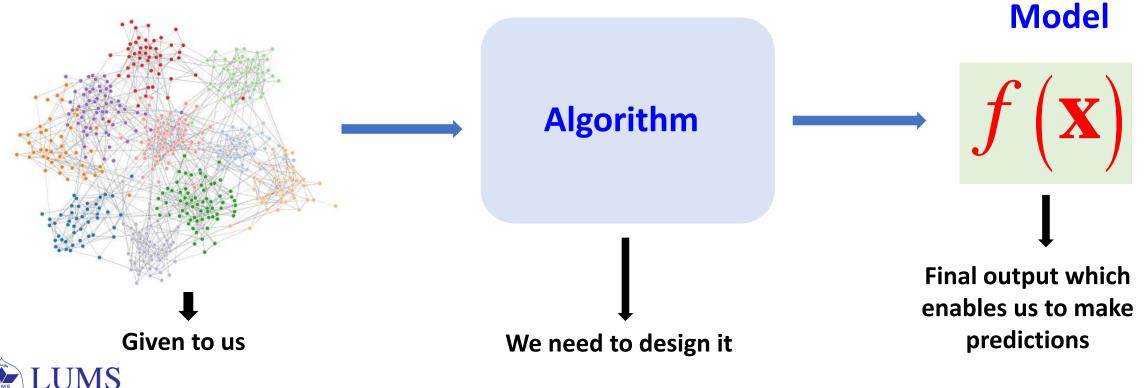


What is Machine Learning?

A Not-for-Profit University

- Study, Design and Analysis of algorithms that improve their performance at some task with experience

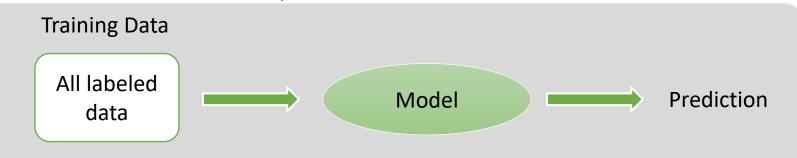
Data



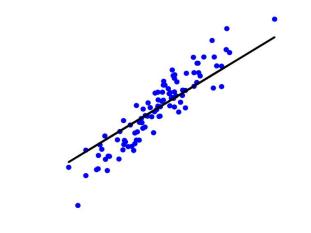
Nature of ML Problems

1. Supervised Learning

The learning algorithm would receive a set of inputs along with the corresponding correct outputs to train a model



<u>Classification: Discrete Prediction</u> Given a data sample, predict its class <u>Regression: Quantitative Prediction on a continuous scale</u> Given a data sample, predict a numerical value

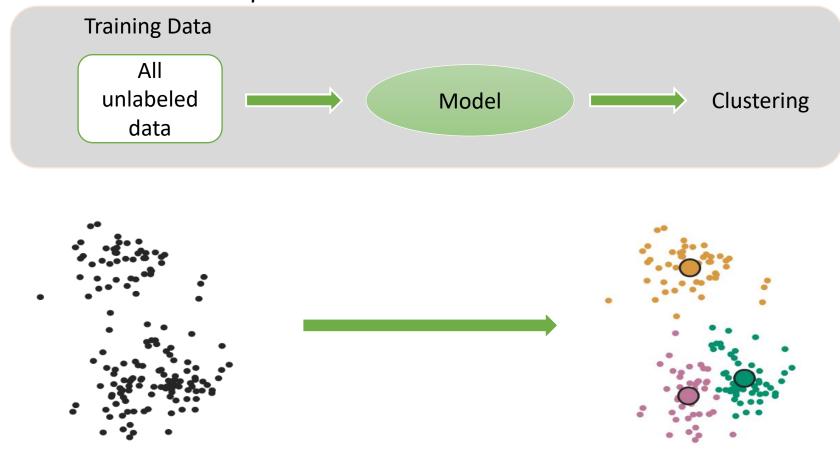




Machine Learning: Overview Nature of ML Problems

2. Unsupervised Learning

The learning algorithm would receive unlabeled raw data to train a model and to find patterns in the data

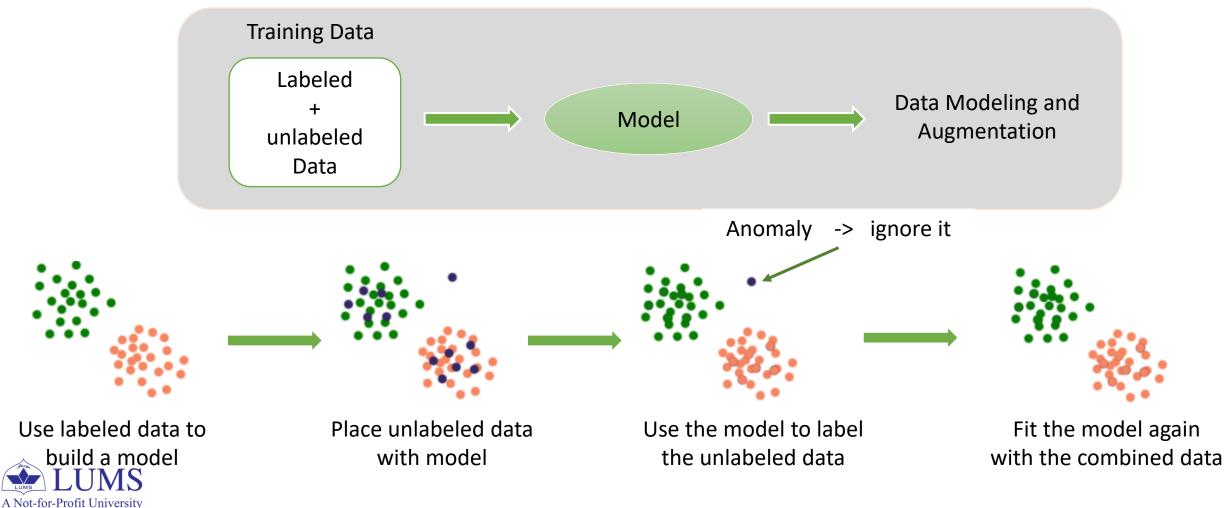




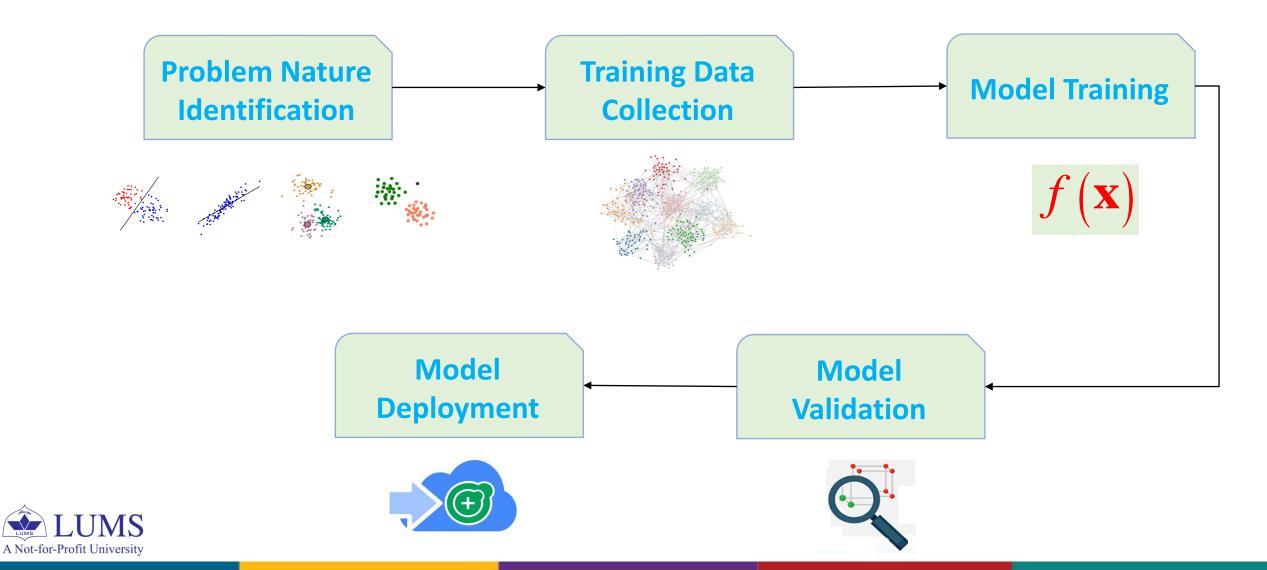
Nature of ML Problems

3. Semi-supervised Learning

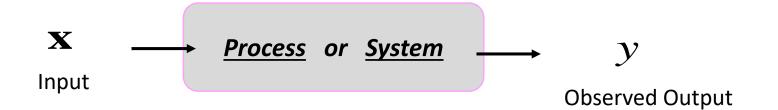
- The learning algorithm receives labeled and unlabeled raw data to train a model
- Main objective is to efficiently accommodate the unlabeled data



Machine Learning: Overview Typical Flow



Training Data Collection

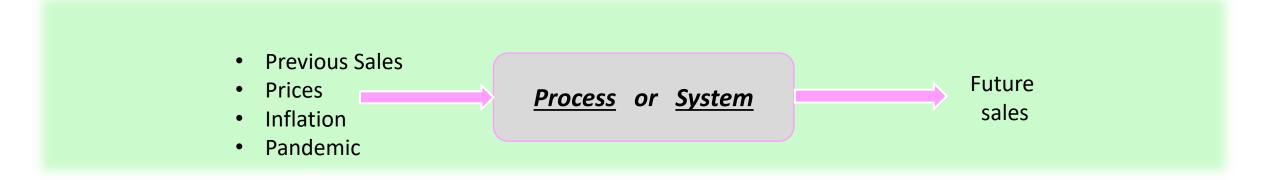


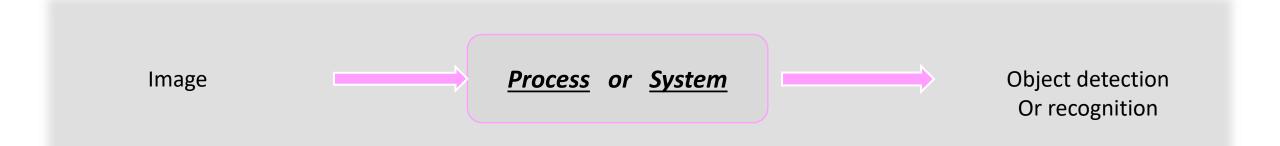
PROCESS or **SYSTEM** : Underlying physical or logical phenomenon which maps our input data to our observed output

Collect the training data by observing our unknown **PROCESS** or **SYSTEM**



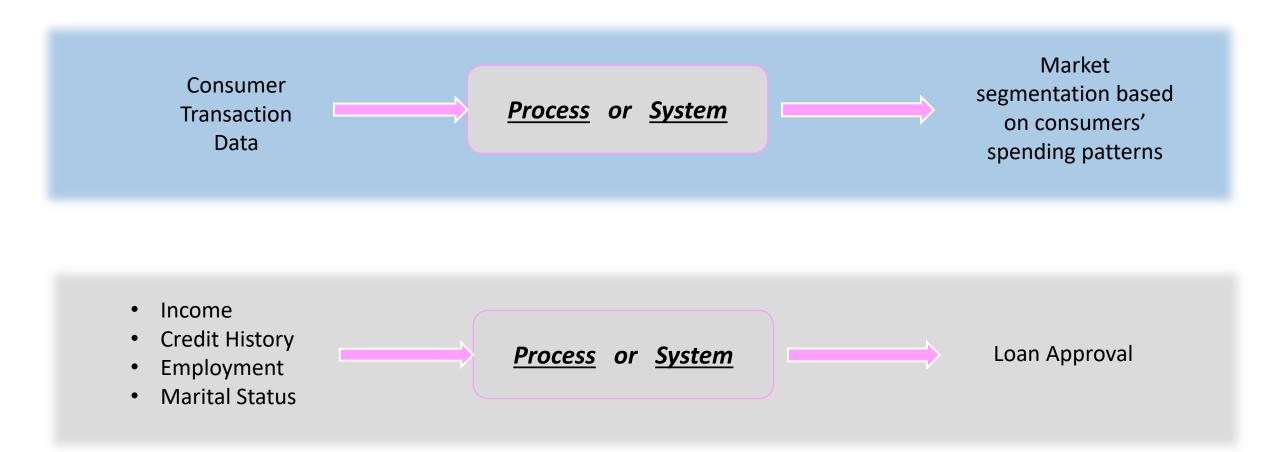
Example Systems







Example Systems





Machine Learning: Overview *Data Types*

Tabulated Data

columns

15

rows

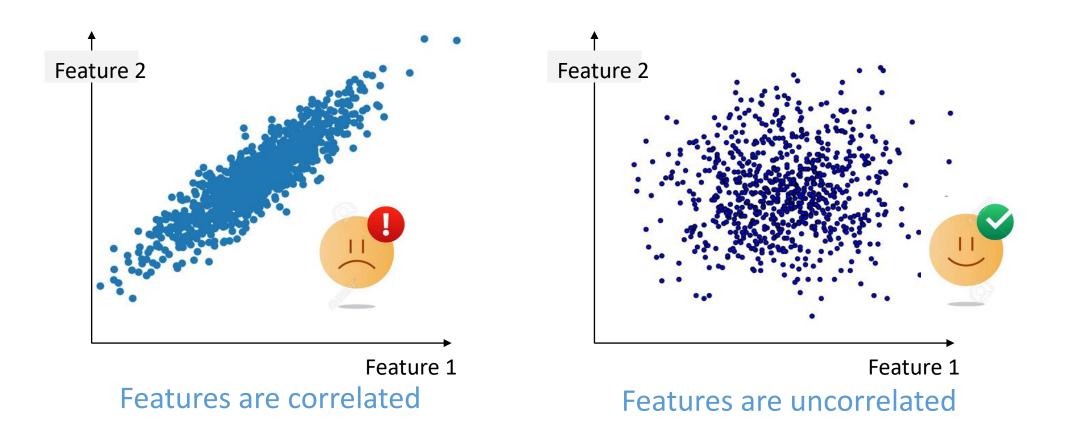
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- Each column is a feature and adds one dimension to the data
- Number of columns define total number of features and hence data dimensionality
- Ideally, these columns should contain different information -> uncorrelated

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	age	workclass	fnlwgt	education	education- num	marital-status	occupation	relationship	race	sex	capital- gain	capital- loss	hours-per- week	native- country	income
view as	II h		l.	I		I	1 111	h.		h	I				Ι.
	39	State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in- family	White	Male	2174	0	40	United- States	<=50K
	50	Self-emp- not-inc	83311	Bachelors	13	Married-civ- spouse	Exec- managerial	Husband	White	Male	0	0	13	United- States	<=50K
	38	Private	215646	HS-grad	9	Divorced	Handlers- cleaners	Not-in- family	White	Male	0	0	40	United- States	<=50K
	53	Private	234721	11th	7	Married-civ- spouse	Handlers- cleaners	Husband	Black	Male	0	0	40	United- States	<=50K
	28	Private	338409	Bachelors	13	Married-civ- spouse	Prof-specialty	Wife	Black	Female	0	0	40	Cuba	<=50K
	37	Private	284582	Masters	14	Married-civ- spouse	Exec- managerial	Wife	White	Female	0	0	40	United- States	<=50K
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Machine Learning: Overview *Data Types*



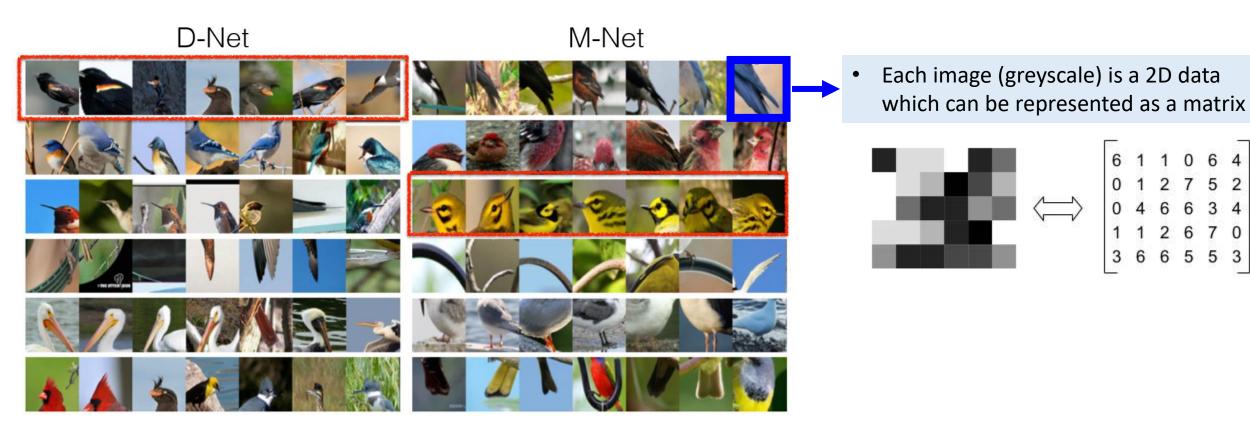
Dimensionality Reduction

Find all those dimensions (features) which carry very less information and discard them



Data Types

Images



4

2

4

0



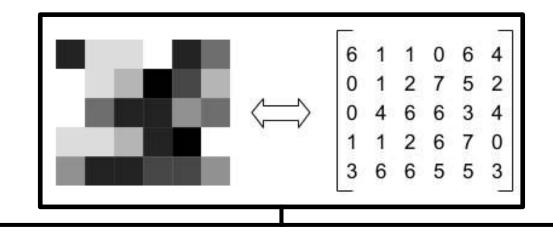
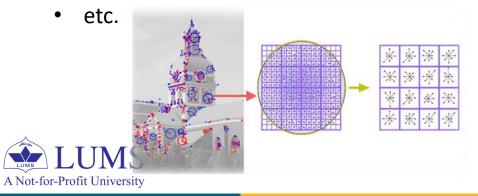


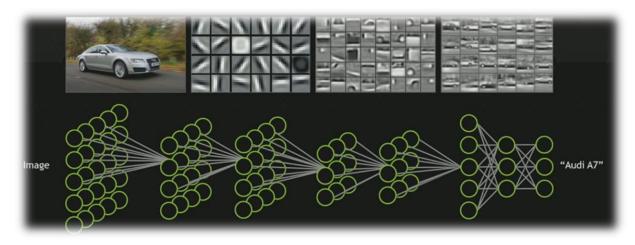
Image Descriptors: Manually extract features

- Speeded-Up Robust Features (SURF)
- Binary Robust Independent Elementary Features (BRIEF)
- Oriented FAST and Rotated BRIEF (ORB)
- Scale-Invariant Feature Transform (SIFT)
- Histogram of Oriented Gradients (HOG)
- HAAR features



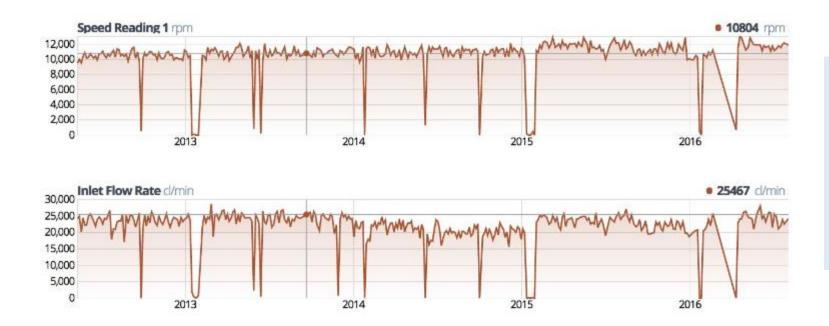
Deep Learning: Let machine learning extract most important features

Neural Networks such as Convolutional Neural Networks (CNN) take raw images as input and first process it inside multiple layers to get best features called intrinsic dimension of the data



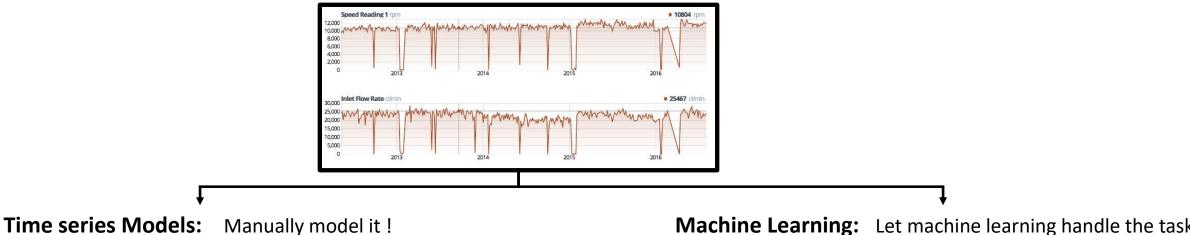
Data Types

Time Series



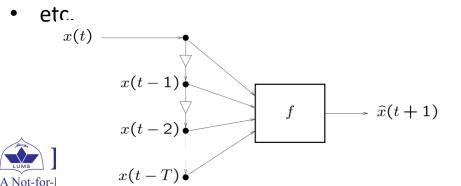
- Much difficult to deal with
 - Enormous Size
 - o Cumbersome Labeling
 - Not so standard feature extraction strategies as in case of images. (Direct modeling of time series is preferred)





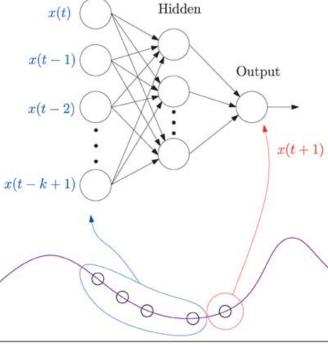
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- Exponential Smoothing (ETS)
- Autoregressive Integrated Moving Average . (ARIMA) models
- Linear regression ٠
- **Generalized Autoregressive Conditional** • Heteroskedasticity (GARCH)
- Bayesian Models •
- Vector auto-regression (VAR) models





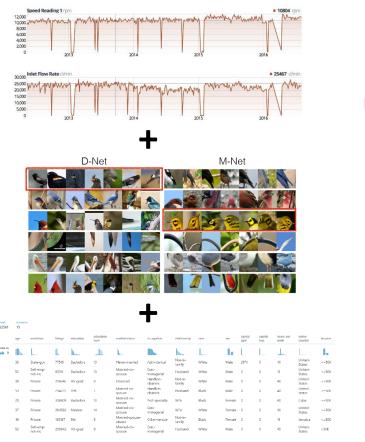
Input



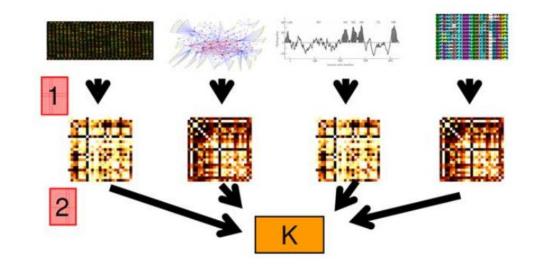
Neural Networks such as **Convolutional Neural** Networks (CNN) take raw waveforms as input and first process it inside multiple layers to get best predictive model for it

Data Types

Heterogeneous data



Multimodal Machine Learning !





About the Instructor

- Assistant Professor, LUMS since 2015
- Post-doctorate 2013–2015, Australian National University (ANU)
- PhD, Australian National University (ANU) 2013

Affiliations:

- Signal, Image and Video Processing Lab, LUMS
- Applied Signal Processing Group, ANU
- Smart Data, Systems and Applications Lab (<u>www.sdsa.lums.edu.pk</u>)

Collaborations: Princeton, UCL, University of Edinburgh, EPFL, ANU

PhD Students: 7 (4 graduated)

Publications: More than 70 (21 Transactions/Journals, 50 Conference proceedings)

Service: Senior Member IEEE and Associate Editor, IEEE Signal Processing Letters



Feedback: Questions or Comments?

Email: <u>zubair.khalid@lums.edu.pk</u>

