



Machine Learning AIML CZG565 M1: Introduction

Dr. Pankaj Agarwal, Professor & Guest faculty



Profile: Dr. Pankaj Agarwal

Total Experience: More than 23 years of Experience in Academics & Industry.

Current Status: Guest Faculty at BITS Pilani and Professor & Dean (Engineering)

Areas of interest & research: Data Science & Machine Learning algorithms, Algorithms Design & Analysis, Data Analytics, Soft computing Techniques, NLP, Computer Vision & Deep Learning with Python Programming, Tableau, KNIME etc

Research Publications: more than 60 research publications in international journals/conferences.

Patents: 10 research patents including Australian, German & Indian patents.

Certifications: IBM certified data Science Professional & Microsoft certified Programmer.

Consultant & Trainer: BITS Pilani, Ureka International, Princeton Hive, Great Learning.

Books: Design & Analysis of Algorithms, .NET,SPM, RDBMS, C programming, Computer

Organization, Cyber Security

Research Guidance: more than 100 M.Tech Students (BITS Pilani), 5 Ph.D students completed



Disclaimer and Acknowledgement



- The content for these slides has been obtained from books and various other source on the Internet
- I hereby acknowledge all the contributors for their material and inputs.
- I have provided source information wherever necessary
- I have added and modified the content flow to suit the requirements of the course and for ease of class presentation
- Students are requested to refer to the textbook and detailed content of this presentation deck over canvas

Course Introduction



Objective of course

- Introduction to the basic concepts and techniques of Machine Learning
- Gain experience in basics of doing independent study and research in the field of Machine Learning
- Develop skills of using recent machine learning software tools to evaluate learning algorithms and model selection for solving practical problems

Focus of this course

- Strong Mathematical Foundations of ML algorithms
- Structured Data Analytics
- IDD (Independent & Identically Distributed Data)

Topics not expected of this course

- Unstructured Data Analytics
- Time Series/Sequence Data Analytics
- Deep Learning

Course Plan



M1	Introduction
M2	Machine learning Workflow
M3	Linear Models for Regression
M4	Linear Models for Classification
M5	Decision Tree
M6	Instance Based Learning
M7	Support Vector Machine
M8	Bayesian Learning
M9	Ensemble Learning
M10	Unsupervised Learning
M11	Machine Learning Model Evaluation/Comparison

Text books and Reference book(s)



- T1 Tom M. Mitchell: Machine Learning, The McGraw-Hill Companies
- R1 Christopher M. Bishop: Pattern Recognition & Machine Learning, Springer P. Tan, et al. Introduction to Data Mining, Pearson
- R2 C.J.C. BURGES: A Tutorial on Support Vector Machines for Pattern Recognition,
- R3 Kluwer Academic Publishers, Boston.

Evaluation scheme

- Quiz (10% Best 2 of 3 quizzes)
- Assignment (20% 1 Progressive Group Assignment)
- Mid-semester exam (30%)
- Comprehensive exam (40%)

Pre-requisites

- Linear algebra: vector/matrix manipulations, properties
- Calculus: partial derivatives
- Probability: common distributions; Bayes Rule
- Statistics: mean/median/mode; maximum likelihood



Lab Plan

Lab No.	Lab Objective	
1	End to End Machine Learning	
2	Linear Regression and Gradient Descent Algorithm	
3	Logistic Regression Classifier	
4	Decision Tree	
5	Naïve Bayes Classifier	
6	Random Forest	

- Labs not graded
- Most of the Lab recordings available at CSIS virtual labs
- Webinars will be conducted for lab sessions
- Labs will be conducted in Python

Agenda



- What is Machine Learning?
- Why Machine Learning is important?
- Types of Machine Learning
- Application Areas
- Issues in Machine Learning
- Demo Case study

- The term Machine Learning was first coined by **Arthur Samuel** in the year 1959, an IBM employee and pioneer in the field of computer gaming and artificial intelligence
- Humans can learn from their experiences with their learning capability,
- But can a machine also learn from experiences or past data like a human does?
- It is the branch of Al mainly concerned with the development of algorithms which allow a computer to learn from the **data** and **past** experiences on their own.
- Machine Learning is a system that can learn from example through self-improvement and without being explicitly coded by programmer.
- Machine learning uses various algorithms for building mathematical models and making predictions using historical data or information

ML – What, When, Where?



Definition by Tom Mitchell (1998):

"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."

Example: playing checkers.

E = the experience of playing many games of checkers

T = the task of playing checkers.

P = the probability that the program will win the next game.



- To have a learning problem, we must identify
 - The class of tasks
 - The measure of performance to be improved
 - Source of experience

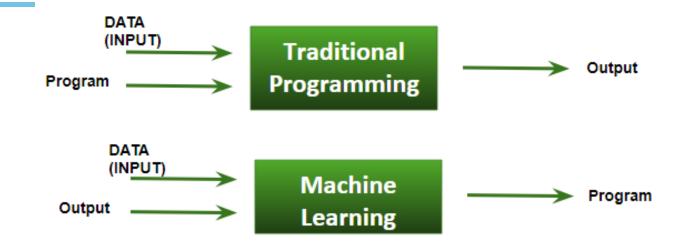
Modern-day machine learning has two objectives, one is to **classify data** based on models which have been developed, the other purpose is to **make predictions** for future outcomes based on these models.



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Traditional Programming VS Machine Learning?



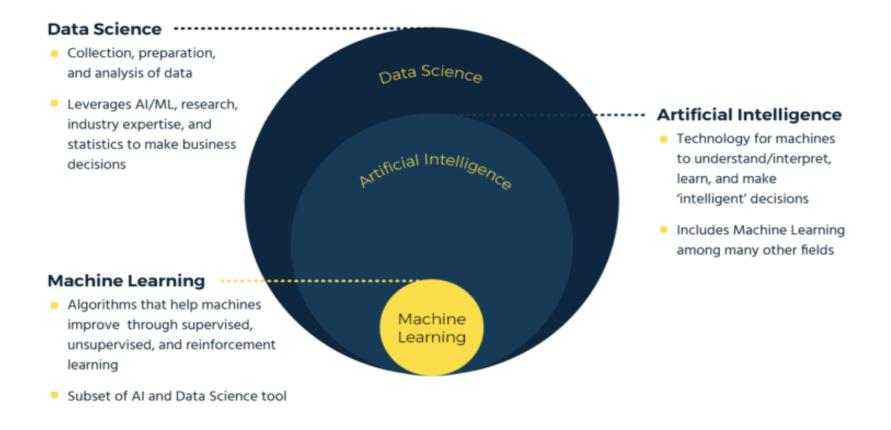


Basic Difference in ML and Traditional Programming?

Traditional Programming: We feed in DATA (Input) + PROGRAM (logic), run it on machine and get output.

Machine Learning: We feed in DATA(Input) + Output, run it on machine during training and the machine creates its own program(logic), which can be evaluated while testing.

Al vs. Data Science vs. Machine Learning



Defining the Learning Tasks

Improve on task T, with respect to performance metric P, based on experience E

Example 1

T: Recognizing hand-written words

P: Percentage of words correctly classified

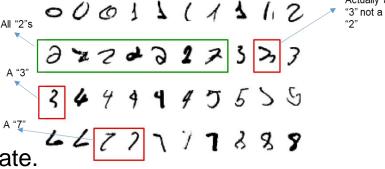
E: Database of human labelled images of handwritten words

Example 2

T: Categorize email messages as spam or legitimate.

P: Percentage of email messages correctly classified.

E: Database of emails, some with human-given labels



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Defining the Learning Tasks

Improve on task T, with respect to performance metric P, based on experience E

Example 3

T: Playing Checkers

P: Percent of games won against opponents

E: Games Played against itself





Example 4

T: Drive on public four-lane highways using vision sensors.

P: Average distance travelled before an error (as judged by human).

E: A sequence of images and steering commands recorded while observing a human driver



Applications - Perspectives

Object Categorization Prediction



Medical Diagnosis

Transaction Analysis

Recommendation System

Speech – Text Processing

Sequence

Forecasting
Medical Research
Recommendation System
Content Management





Applications - Perspectives

Planning

Navigation

Path Finding

Gaming

Controlling

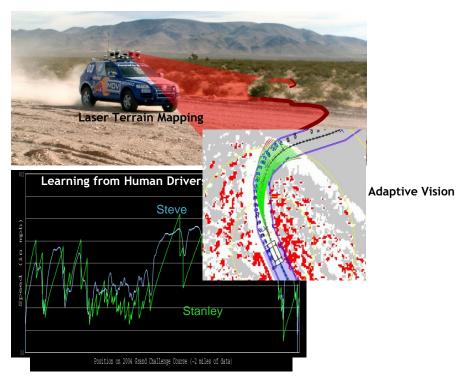


Problem Solving

Pattern Recognition

Optimization

Decision System



Why ML



When Do We Use Machine Learning?

ML is used when:

Human expertise does not exist (navigating on Mars)

Humans can't explain their expertise (Biometrics)

- Models must be customized (personalized medicine)
- Learning isn't always useful:
 - There is no need to "learn" to calculate payroll









Need For Machine Learning?



- Some tasks cannot be defined well, except by examples.
- Relationships and correlations can be hidden within large amounts of data.
- Machine Learning may be able to find these relationships.
- The amount of knowledge available about certain tasks might be too large for explicit encoding by humans (e.g., medical diagnostic).
- We generate around 2.5 quintillion bytes (1 Quintillion is 10^18) of data every single day!
- Increase in Data Generation: Due to excessive production of data, we need a method that can be used to structure, analyze and draw useful insights from data. ML uses data to solve problems and find solutions to the most complex tasks faced by organizations.
- Improve Decision Making: By making use of various algorithms, Machine Learning can be used to make better business decisions. For example, Machine Learning is used to forecast sales, predict downfalls in the stock market, identify risks and anomalies, etc.

Types of ML



Types: Inputs: Based on level of supervision

Feedback	No Feedback	Delayed Feedback (rewards/penalty)
Supervised	Unsupervised	Reinforcement

Machine Learning - Examples

Objective: Employability Prediction

Features / Attributes / Predictors

- ✓ CGPA
- ✓ Communication Skills
- ✓ Aptitude
- ✓ Programming Skills

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S.No.	CGPA	Communication Skills	Aptitude	Programming Skills	Job Offered?
1 —	9.1	Average	Good	Excellent	Yes J
2	8.4	Good	Good	Good	Yes
3	8.3	Poor	Average	Average	No
4 _	7.1	Average	Good	Average	No
5 -	8.2	Good	Excellent	Excellent	No

3

4

Machine Learning - Examples



Objective: Market Segmentation Study

Features / Attributes / Predictors

- √ Family income
- ✓ # of visits in a month
- ✓ Average money spent in a month
- ✓ Zip code

Customers for a retailer may fall into

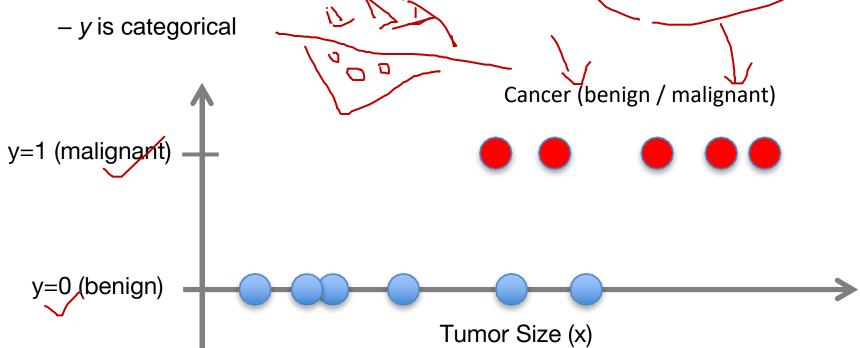
- ✓ two groups say big spenders and low spenders
- ✓ three groups say big spenders, medium spenders and low spenders
- ✓ Four groups,

S.No	Zip Code	Family Income	# of visits in a month	Average Money Spent in a month
1	500078	11,50,000	4	8,000

Supervised Learning: Classification

GOAL: Previously unseen records should be assigned a class as accurately as possible.

- Given $(x_1, y_1), (x_2, y_2), ..., (x_n, y_n)$
- Learn a function f (x) to predict y given x

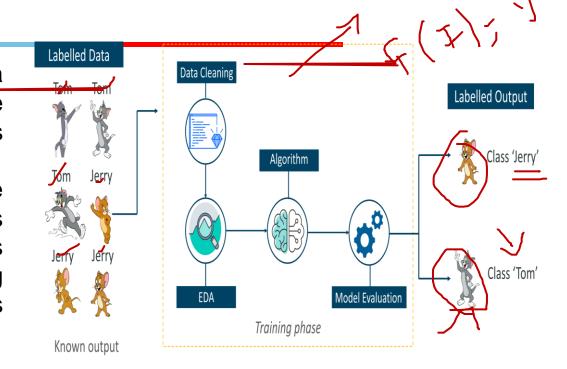


Slide Credit: Eric Eaton



Supervised Learning

- Supervised learning is a technique in which we train the machine using data which is well labeled.
- Applications in which the training data comprises examples of the input vectors along with their corresponding target vectors are known as supervised learning problems.



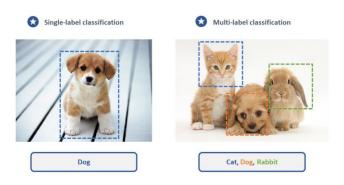
- The training data provided to the machines work as the supervisor that teaches the machines to predict the output correctly.
- It applies the same concept as a student learns in the supervision of the teacher.
- As input data is fed into the model, it adjusts its weights until the model has been fitted appropriately.
- Supervised machine learning algorithms define models that capture relationships among data.



Classification Applications

- Google Image Classification
- Face recognition system
- Spam filters Specific Controls
- Document tagging
- Fraud detection





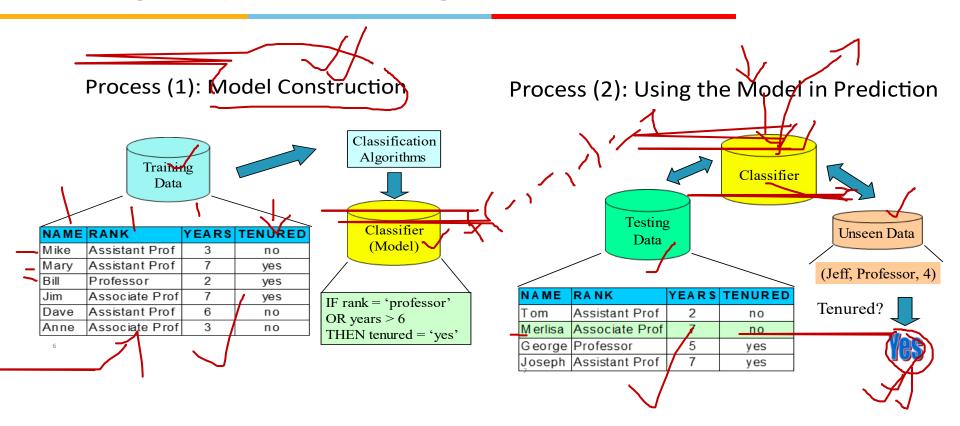


Supervised Learning Techniques / Algorithms

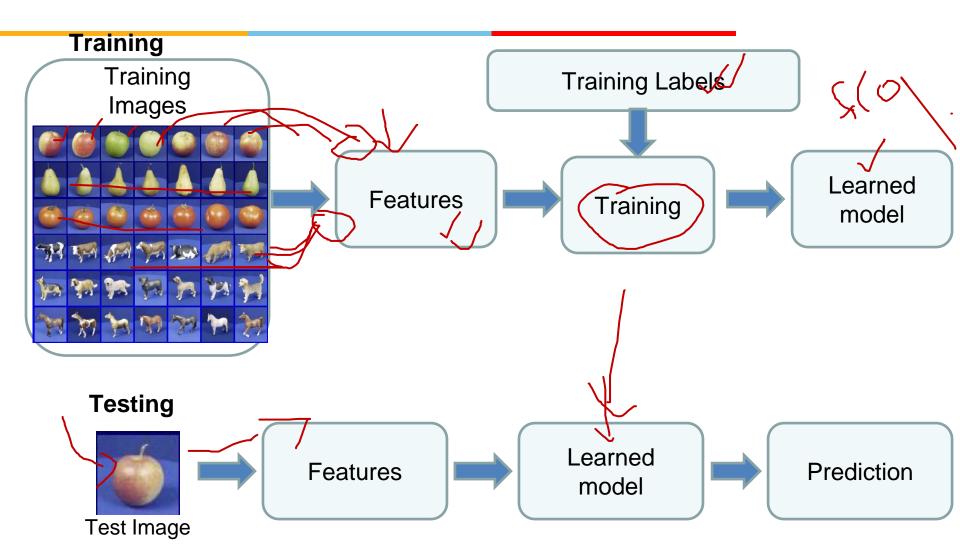
- Linear Regression
- Logistic Regression
- Naïve Bayes Classifiers
- Support Vector Machines (SVMs)
- Decision Trees and Random Forests
- Neural networks



Working of Supervised Learning Method



A Typical Supervised Learning Workflow (for Classification)

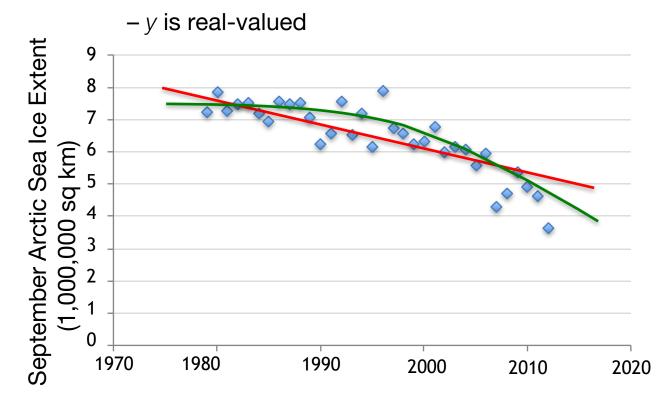


Slide credit: D. Hoiem and L. Lazebnik

Supervised Learning: Regression

GOAL: Previously unseen records should be assigned a value as accurately as possible.

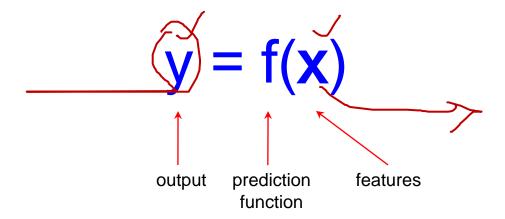
- Given $(x_1, y_1), (x_2, y_2), ..., (x_n, y_n)$
- Learn a function f (x) to predict y given x



Year Number

Data from G. Witt. Journal of Statistics Education, Volume 21, Number 1 (2013) Slide Credit: Eric Eaton

Regression

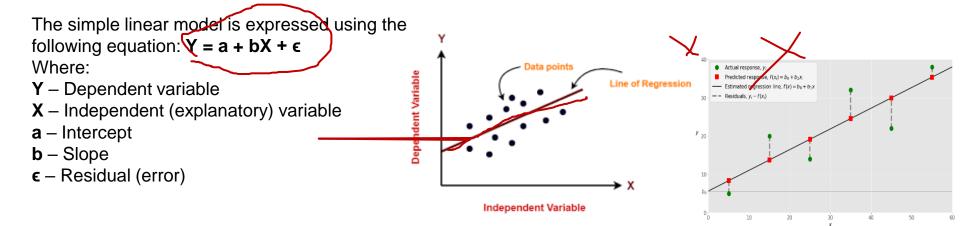


- Training: given a training set of labeled examples {(x₁,y₁), ..., (x_N,y_N)},
 estimate the prediction function f by minimizing the prediction error on the
 training set
- Testing: apply f to a never before seen test example x and output the predicted value y = f(x)

Regression



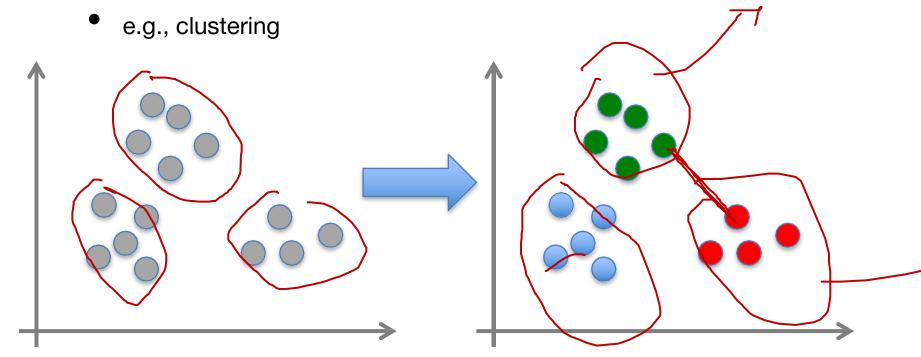
- Regression analysis is used to find trends in data.
- Typically, you need regression to answer whether and how some phenomenon influences the other or how several variables are related.
- It is used for the prediction of continuous variables, such as Weather forecasting, Market Trends, etc.
- For example, you can use it to determine *if* and *to what extent* the experience or gender impact salaries.
- For example, if you've been putting on weight over the last few years, it can predict how much you'll
 weigh in ten years time if you continue to put on weight at the same rate
- Essentially, regression is the "best guess" at using a set of data to make some kind of prediction.
- It's fitting a set of points to a graph.



Unsupervised Learning

GOAL: Intra cluster distances are minimized and inter cluster distances are maximized

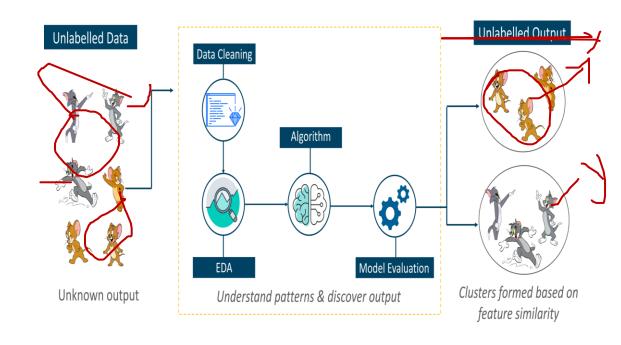
- Given $x_1, x_2, ..., x_n$ (without labels)
- Output hidden structure behind the x's



Slide Credit: Eric Eaton



Unsupervised Learning



For example, it identifies prominent features of Tom such as pointy ears, bigger size, etc, to understand that this image is of type 1. Similarly, it finds such features in Jerry and knows that this image is of type 2. Therefore, it classifies the images into two different classes without knowing who Tom is or Jerry is.

Unsupervised Learning-Example

Let's, take the case of a baby and her family dog.



She knows and identifies this dog. Few weeks later a family friend brings along a dog and tries to play with the baby.



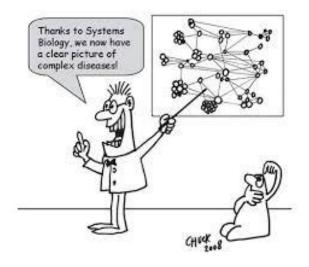
Baby has not seen this dog earlier. But it recognizes many features (2 ears, eyes, walking on 4 legs) are like her pet dog. She identifies the new animal as a dog. This is unsupervised learning, where you are not taught but you learn from the data (in this case data about a dog.)

Had this been supervised learning, the family friend would have told the baby that it's a dog.



Unsupervised Learning Applications

- Personalized recommendation system
- Targeted marketing
- Spam Filters
- Content Management News hosted in Web
- Campaigning



Unsupervised Learning Techniques

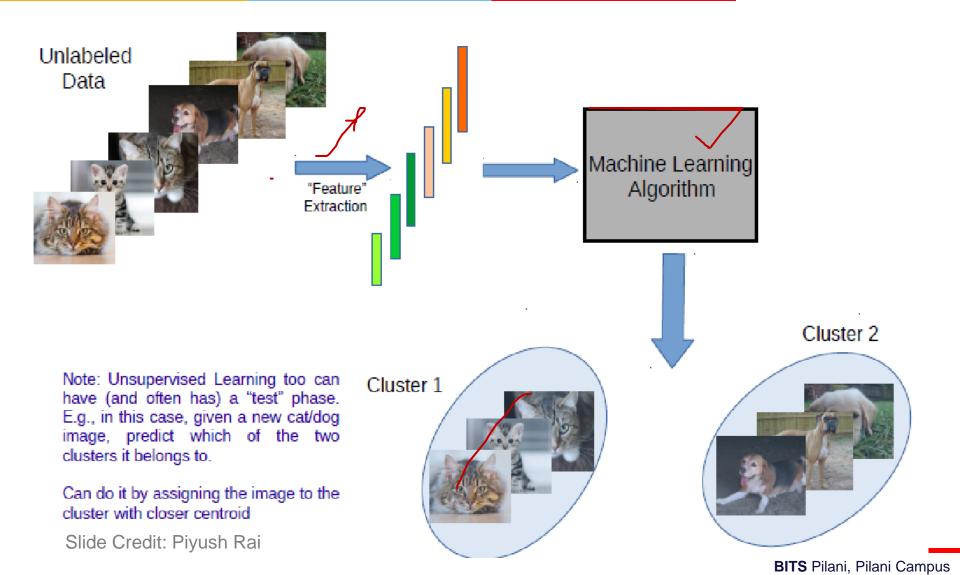
Clustering

- k-Means
- Hierarchical Cluster Analysis
- Expectation Maximization

Visualization and dimensionality reduction

- Principal Component Analysis (PCA)
- Kernel PCA
- Locally-Linear Embedding (LLE)
- t-distributed Stochastic Neighbor Embedding (t-SNE)

A Typical Unsupervised Learning Workflow (for Clustering)



Reinforcement Learning



- Reinforcement Learning is a part of Machine learning where an agent is put in an environment and he learns to behave in this environment by performing certain actions and observing the rewards which it gets from those actions.
- Imagine that you were dropped off at an isolated island! What would you do?
- As time passes by, you will learn how to live on the island. You will explore the environment, understand the climate condition, the type of food that grows there, the dangers of the island, etc. This is exactly how Reinforcement Learning works,
- it involves an **Agent** (you, stuck on the island) that is put in an unknown environment (island), where he must learn by observing and performing actions that result in rewards.
- Reinforcement Learning is mainly used in advanced Machine Learning areas such as self-driving cars, AplhaGo, etc.

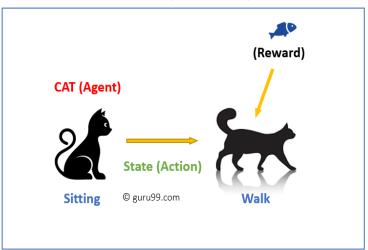


How Reinforcement Learning works?



- Consider the scenario of teaching new tricks to your cat
- As cat doesn't understand English or any other human language, we can't tell her directly what to do. Instead, we follow a different strategy.
- We emulate a situation, and the cat tries to respond in many different ways. If the cat's response is the desired way, we will give her fish.
- Now whenever the cat is exposed to the same situation, the cat executes a similar action with even more enthusiastically in expectation of getting more reward(food).
- That's like learning that cat gets from "what to do" from positive experiences.
- At the same time, the cat also learns what not do when faced with negative experiences.
- There is no supervisor, only a real number or reward signal
- Sequential decision making
- •Time plays a crucial role in Reinforcement problems
- •Feedback is always delayed, not instantaneous
- Agent's actions determine the subsequent data it receives

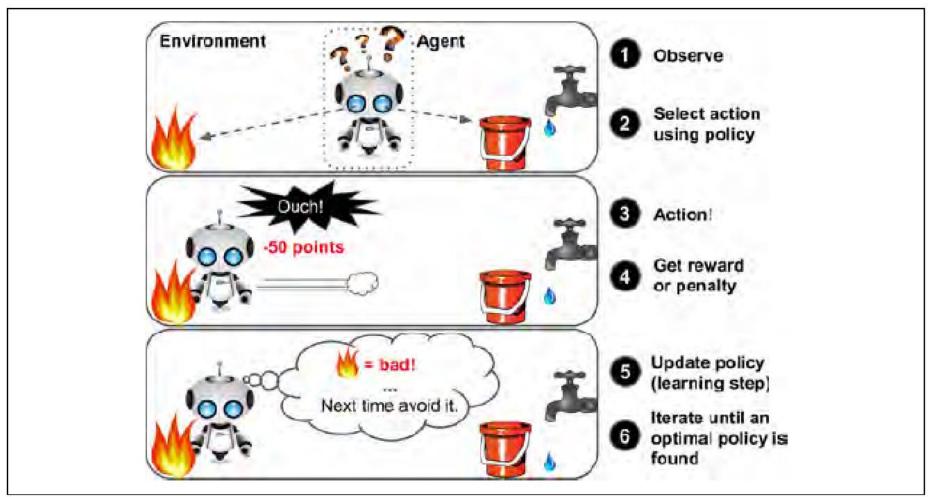
House (environment)







Reinforcement Learning



Some Examples of ML Applications



Traffic Alerts (Maps): Google Maps is probably the app we use whenever we go out and require assistance in directions and traffic. "Despite the Heavy Traffic, you are on the fastest route". But, How does it know that?

•Everyone using maps is providing their location, average speed, the route in which they are traveling which in turn helps Google collect massive Data about the traffic, which makes them predict the upcoming traffic and adjust your route according to it. Social Media (Facebook): One of the most common applications of Machine Learning is Automatic Friend Tagging Suggestions in Facebook or any other social media platform. Facebook uses face detection and Image recognition to automatically find the face of the person which matches it's Database and hence suggests us to tag that person based on DeepFace.

Transportation and Commuting (Uber): If you have used an app to book a cab, you are already using Machine Learning to an extent. It provides a personalized application which is unique to you. Automatically detects your **location** and provides options to either go home or office or any other frequent place based on your **History and Patterns.**

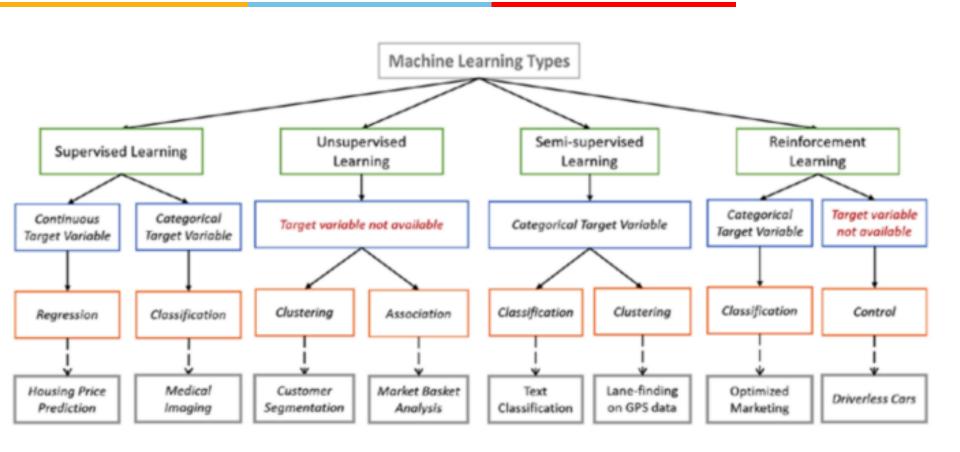
Some Examples of ML Applications



- Amazon's Alexa: The infamous Alexa, which is based on Natural Language Processing and Machine Learning is an advanced level Virtual Assistant that does more than just play songs on your playlist. It can book you an Uber, connect with the other IoT devices at home, track your health, etc.
- Google's Spam Filter: Gmail makes use of Machine Learning to filter out spam messages. It uses Machine Learning algorithms and Natural Language Processing to analyze emails in real-time and classify them as either spam or non-spam.
- Products Recommendations: Suppose you check an item on Amazon, but you do not buy it then and there. But the next day, you're watching videos on YouTube and suddenly you see an ad for the same item. You switch to Facebook, there also you see the same ad. So how does this happen?
- Self Driving Cars: Machine Learning plays a very important role in Self Driving Cars and I'm sure you guys might have heard about Tesla. The leader in this business and their current Artificial Intelligence is driven by hardware manufacturer NVIDIA, which is based on Unsupervised Learning Algorithm. they didn't train their model to detect people or any object as such. The model works on Deep Learning and it crowdsources data from all of its vehicles and its drivers.



Summary: Types of Learning







Open source ML programming tools

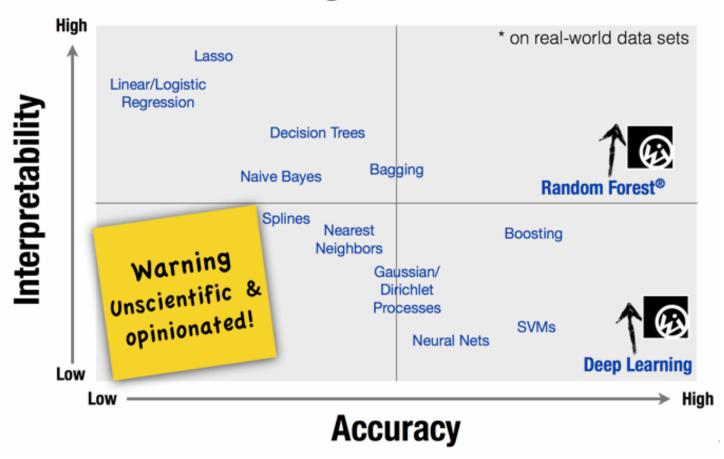
		Platform		Algorithms or Features
S	Scikit Learn	Linux, Mac OS, Windows	Python, C, C++	Classification, Regression, Clustering Preprocessing, Model Selection Dimensionality reduction.
F	PyTorch	Linux, Mac OS, Windows	Python, C++	Autograd Module, Optimization Module NN Module
T	ensorFlow	Linux, Mac OS, Windows	Python, C++	Provides a library for dataflow programming.
V	Veka	Linux, Mac OS, Windows	Java	Data preparation, Classification Regression, Clustering, Visualization Association rules mining



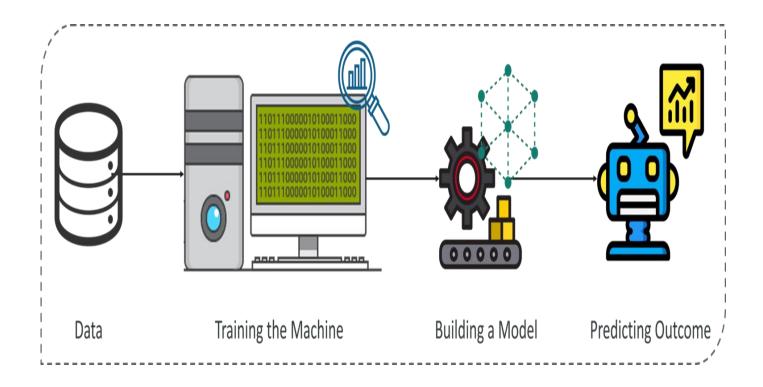
Open source ML programming tools

Colab	Cloud Service	-	Supports libraries of PyTorch, Keras, TensorFlow, and OpenCV
Apache Mahout	Cross-platform	Java Scala	Preprocessors, Regression Clustering, Recommenders Distributed Linear Algebra.
Accors.Net	Cross-platform	C#	Classification, Regression, Distribution Clustering, Hypothesis Tests & Kernel Methods, Image, Audio & Signal & Vision
Shogun	Windows Linux, UNIX Mac OS	C++	Regression, Classification, Clustering Support vector machines. Dimensionality reduction, Online learning etc.
Keras.io	Cross-platform	Python	API for neural networks

ML Algorithmic Trade-Off



ML workflow



ML workflow

- 1. Should I use ML on this problem?
 - Is there a pattern to detect?
 - Can I solve it analytically?
 - Do I have data?
- 2. Gather and organize data.
- 3. Preprocessing, cleaning, visualizing.
- 4. Choosing a model, loss, regularization, ...
- 5. Optimization
- 6. Hyper parameter search.
- 7. Analyze performance and mistakes, and iterate back to step 5 (or 3)

Example: Car Price prediction based on Mileage

- Define the Objective
- Data Gathering: survey, Past Purchase data
- Data Preprocessing
 - training set; test set
 - representation of input features; output
- Exploratory Data Analysis
- Choose form of model: linear regression
- System's performance evaluation: objective function
- Optimize performance by setting appropriate parameters: Optimization
- Evaluate on test set: generalization

X Mileage	Y Car Price	H0	Error L1
9.8	10.48	9	2
9.12	1.75	8	7
9.5	6.95	9	3
10	2.51	9	7

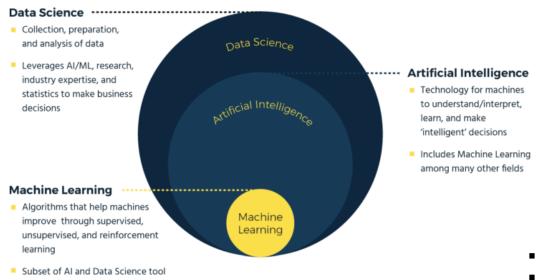
Related Fields

Data Mining Vs Machine Learning



- Data mining is designed to extract the rules from large quantities of data,
- while machine learning teaches a computer how to learn and comprehend the given parameters.
- Or to put it another way, data mining is simply a method of researching to determine a particular outcome based on the total of the gathered data.
- On the other side of the coin, we have machine learning, which trains a system to perform complex tasks and uses harvested data and experience to become smarter.
- Data mining relies on vast stores of data (e.g., <u>Big Data</u>), which then, in turn, is used to make forecasts for businesses and other organizations.
- Machine learning, on the other hand, works with algorithms, not raw data.
- Data mining relies on human intervention and is ultimately created for use by people.
- Whereas machine learning's whole reason for existing is that it can teach itself and not depend on human influence or actions.
- Also, data mining is a process that incorporates two elements: the database and machine learning.
- The former provides data management techniques, while the latter supplies data analysis techniques.
- data mining can't learn or adapt, whereas that's the whole point with machine learning. Data mining follows pre-set rules and is static, while machine learning adjusts the algorithms as the right circumstances manifest themselves

Al vs. Data Science vs. Machine Learning



Artificial Intelligence uses logic and decision trees.

Machine Learning uses statistical models.

Data Science deals with structured and unstructured data.

- Data science produces insights
- Machine learning produces predictions
- Artificial intelligence produces actions

- Data science is a field that studies data and how to extract meaning from it,
- whereas machine learning is a field devoted to understanding and building methods that utilize data to improve performance or inform predictions.
- Machine learning is a branch of artificial intelligence.

Machine learning and artificial intelligence (AI) have dominated parts of data science, playing a critical role in data analytics and business intelligence. Machine learning automates the process of data analysis and goes further to make predictions based on collecting and analyzing large amounts of data on certain populations. Models and algorithms are built to make this happen.

Data Science

Field that determines the processes, systems, and tools needed to transform data into insights to be applied to various industries.

Skills needed:

- Statis
- Data visualization
- Coding skills (Python/F
- County skills (Fytholly R
- SOL/NoSOL
- · Data wrangling

Machine learning is part of data science. Its algorithms train on data delivered by data science

to "learn." Skills needed:

- Math, statistics, and probability
- Comfortable working with data
- Programming skills

Machine Learning

Field of artificial intelligence (AI) that gives machines the human-like capability to learn and adapt through statistical models and algorithms.

Skills needed:

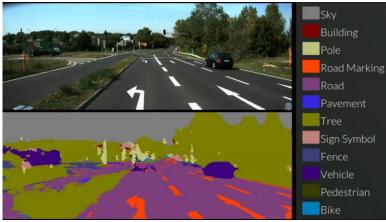
- Programming skills (Python, SQL, Java)
- Statistics and probability
- Prototyping
- Data modeling



Common Use cases - Security & Transaction Domain



Sentiment analysis on Product review of Mobile phone



- Self Driving Cars
- Fraud Detection in Banking
- Email Filtering
- Dynamic Pricing in Travel

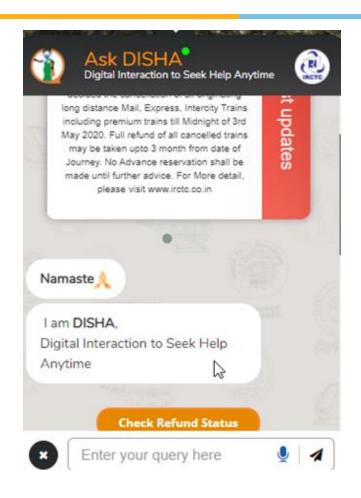
Derived Applications:

- > Cyber Security
- > Video Surveillance
- > Object Detection





Common Use cases - Customer Support Systems





- Apple's Siri
- Google Assistant
- Amazon's Alexa
- Google Duplex
- Microsoft's Cortana
- Samsung's Bixby

Derived Applications:

- > Customer Support Query (Voice vs Text)
- > Chatbots



Common Use cases - Recommendation Engines



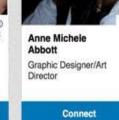
- E-commerce sites like Amazon and Flipkart
- Book sites like Goodreads
- Movie services like IMDb and Netflix
- Hospitality sites like MakeMyTrip, Booking.com, etc.
- Retail services like StitchFix
- Food aggregators like Zomato and Uber Eats

People You May Know



Maris Cohen
Communications Planner
at Carat USA

Connect





Mercedes Jester Customer Service Team Lead at John Wiley and Sons

Connect



Susan Lynch
Assistant Property
Manager at The Bozzuto
Group

Connect

Derived Applications:

- > Personalized Marketing
- > Personalized Banking

MCQ



What is the primary goal of machine learning?

- a) Minimize training time
- b) Maximize model complexity
- c) Minimize prediction accuracy
- d) Maximize generalization to new data

Answer: d) Maximize generalization to new data

Which of the following is a type of unsupervised learning?

- a) Regression
- b) Classification
- c) Clustering
- d) Reinforcement learning

Answer: c) Clustering

What does the bias-variance tradeoff refer to in machine learning?

- a) Balancing model complexity and interpretability
- b) Balancing training time and prediction time
- c) Balancing the model's ability to fit training data and generalize to new data
- d) Balancing the tradeoff between precision and recall

Answer: c) Balancing the model's ability to fit training data and generalize to new data

MCQ



What is the main difference between supervised and unsupervised learning?

- a) Supervised learning requires labeled data, while unsupervised learning does not.
- b) Unsupervised learning requires labeled data, while supervised learning does not.
- c) Supervised learning is used for classification, while unsupervised learning is used for regression.
- d) Unsupervised learning is used for classification, while supervised learning is used for clustering.

Answer: a) Supervised learning requires labeled data, while unsupervised learning does not.

What is the purpose of a validation set in machine learning?

- a) To train the model
- b) To test the model
- c) To fine-tune hyperparameters and avoid overfitting
- d) To evaluate the model's performance on unseen data

Answer: c) To fine-tune hyperparameters and avoid overfitting

Which of the following is an example of a classification problem?

- a) Predicting house prices
- b) Identifying spam emails
- c) Estimating stock prices
- d) Forecasting temperature

Answer: b) Identifying spam emails

References

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Thank you!