

Introduction to Machine Learning

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What is Machine Learning?



- Getting a computer to do something, without explaining to the computer exactly how to do it
 - “Learn from these examples” – **Supervised Learning**
 - e.g., given many examples of pictures of cats and dogs with the label “cat” or “dog”, learn how to identify whether a new, unlabeled image is a cat or a dog
 - “Find patterns in this data” – **Unsupervised Learning**
 - e.g., given information about all your customers, can you group your customers into sets that are similar in some respect, perhaps so you could treat them as a group in a marketing campaign?
 - “Try to achieve this goal by learning from your experience” – **Reinforcement Learning**
 - e.g. get a robot to learn how to climb over a box by trying different movements of its legs to see which ones are most effective

Why is it so powerful?



- You can use (basically) the same algorithms for tasks in multiple domains, e.g.,
 - identifying genes that are similar to each other and identifying customers that are similar to each other
 - predicting which of your customers are likely to unsubscribe from your service and predicting the upcoming weather from current rain radar images
 - identifying tumours in medical images and identifying words spoken to a voice assistant
 - Classifying images of cats and dogs, and classifying email as spam or not spam
 - Teaching a robot to walk, and predicting the next word you will type in a text message
 - Responding to a prompt in ChatGPT and creating an image from a text prompt

Some other definitions of machine learning | epcc |

- “the field of study that gives computers the ability to learn without explicitly being programmed.” – Arthur Samuel, IBM (~1959)
 - This is possibly a paraphrase, but it is widely accepted that the term was coined by Arthur Samuel
- “Machine learning (ML) is a field of study in artificial intelligence concerned with the development and study of statistical algorithms that can effectively generalize and thus perform tasks without explicit instructions” – Wikipedia (accessed November, 2023)
- “Machine learning is a branch of artificial intelligence (AI) and computer science which focuses on the use of data and algorithms to imitate the way that humans learn, gradually improving its accuracy.” – IBM
- “it's the science of getting computers to learn without being explicitly programmed” – Andrew Ng, Stanford/Coursera/...

What do the machines learn?



- Central to Machine Learning is the idea of a **model**
- A model is learned from data using a **machine learning algorithm** and the model can then be used to make predictions/classifications from new, previously unseen data (in the case of supervised learning) or to cluster/group the data (in the case of unsupervised learning) or to decide on a course of action in an observed environment (reinforcement learning)
- Generally speaking, with ML, machines do *not* learn facts and they do *not* learn to reason, they learn models/patterns
 - No reason why this isn't possible in future, but that's not how ML currently works, even in ChatGPT!

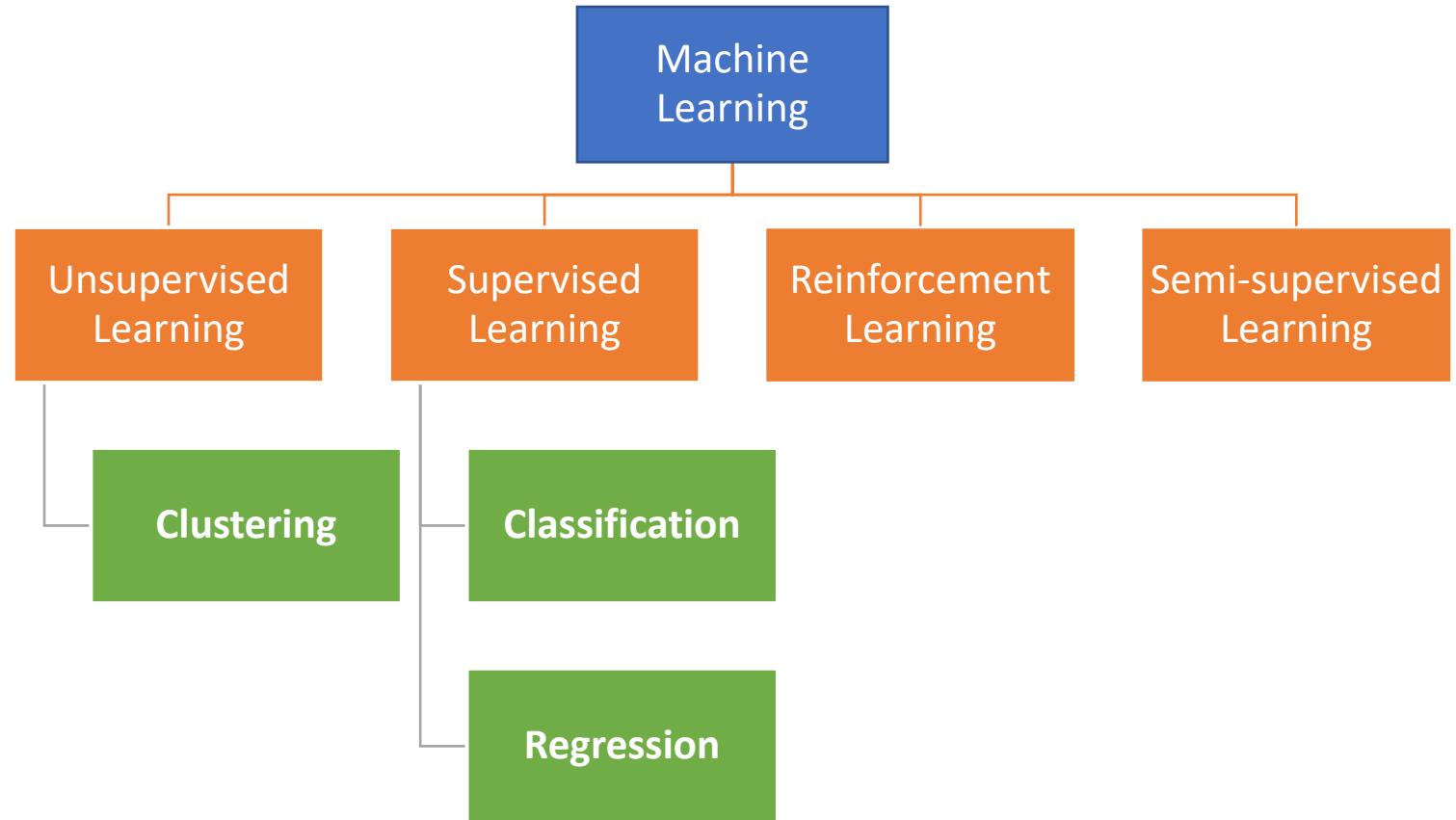
Types of Machine Learning

- Unsupervised Learning

- Clustering
 - kMeans
 - DBSCAN
 - Hierarchical

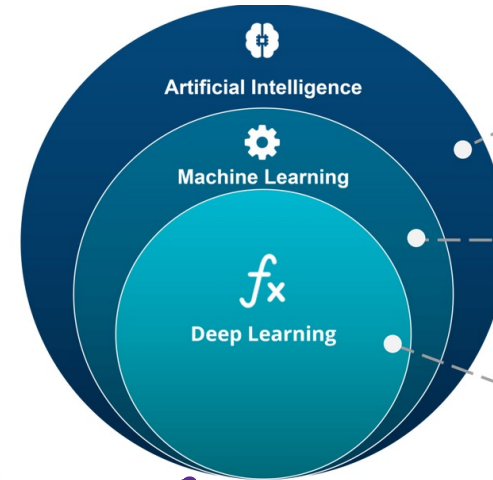
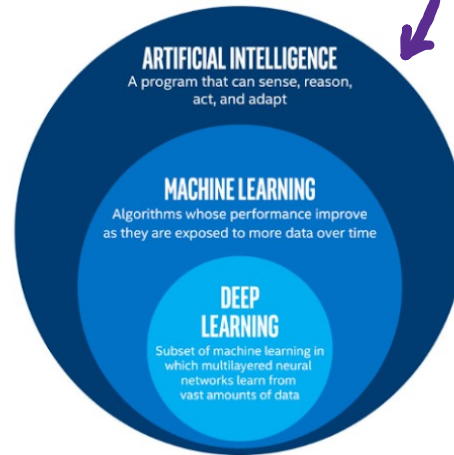
- Supervised Learning

- Regression
 - Linear Regression
- Classification
 - kNN
 - Naïve Bayes
 - Logistic Regression
 - Neural Networks



ML, Deep Learning & AI

from <https://builtin.com/machine-learning/what-is-deep-learning>

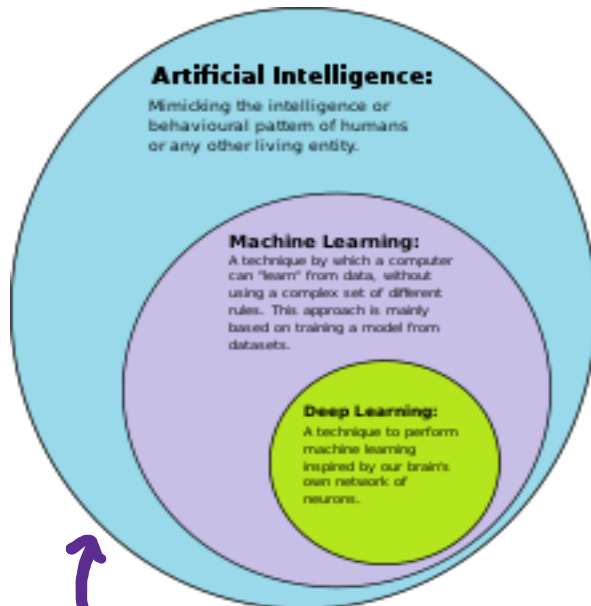


ARTIFICIAL INTELLIGENCE
A technique which enables machines to mimic human behaviour

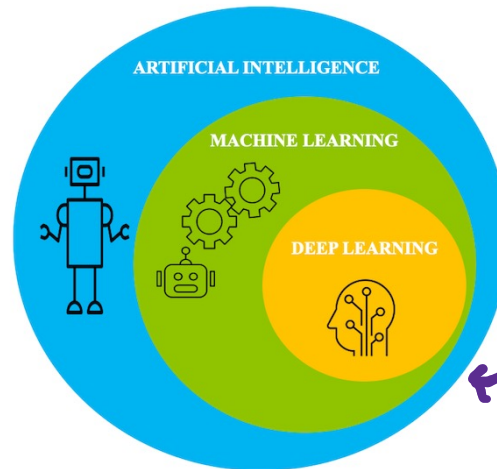
MACHINE LEARNING
Subset of AI technique which use statistical methods to enable machines to improve with experience

DEEP LEARNING
Subset of ML which make the computation of multi-layer neural network feasible

from <https://www.edureka.co/blog/ai-vs-machine-learning-vs-deep-learning/>

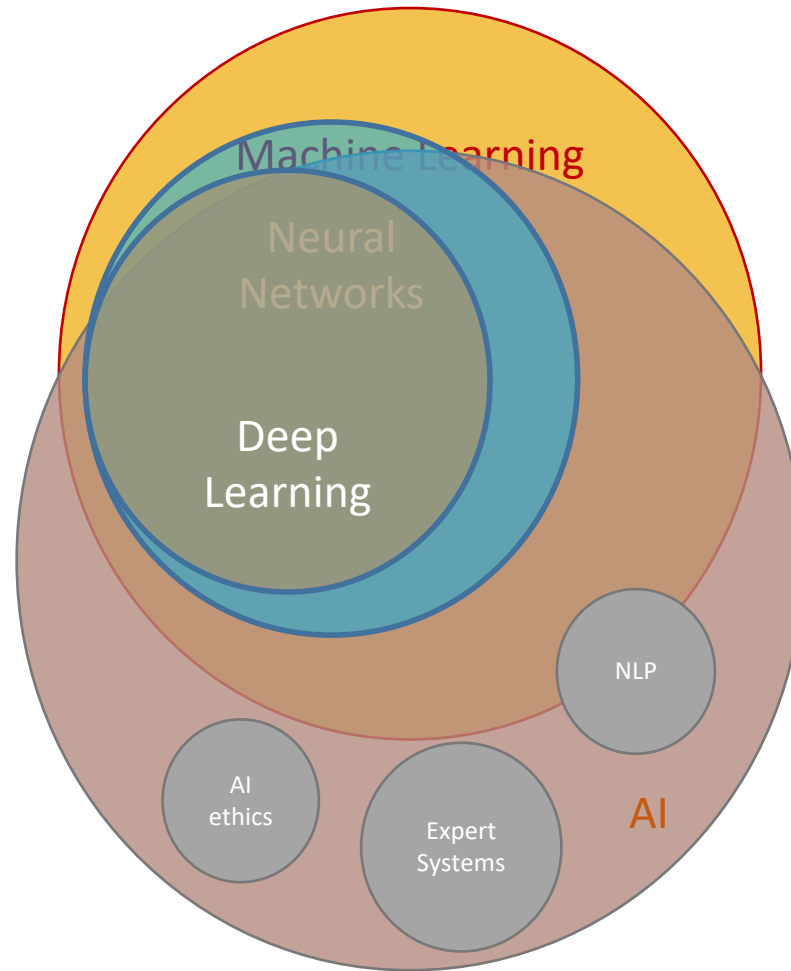


from <https://commons.wikimedia.org/wiki/File:AI-ML-DL.svg>



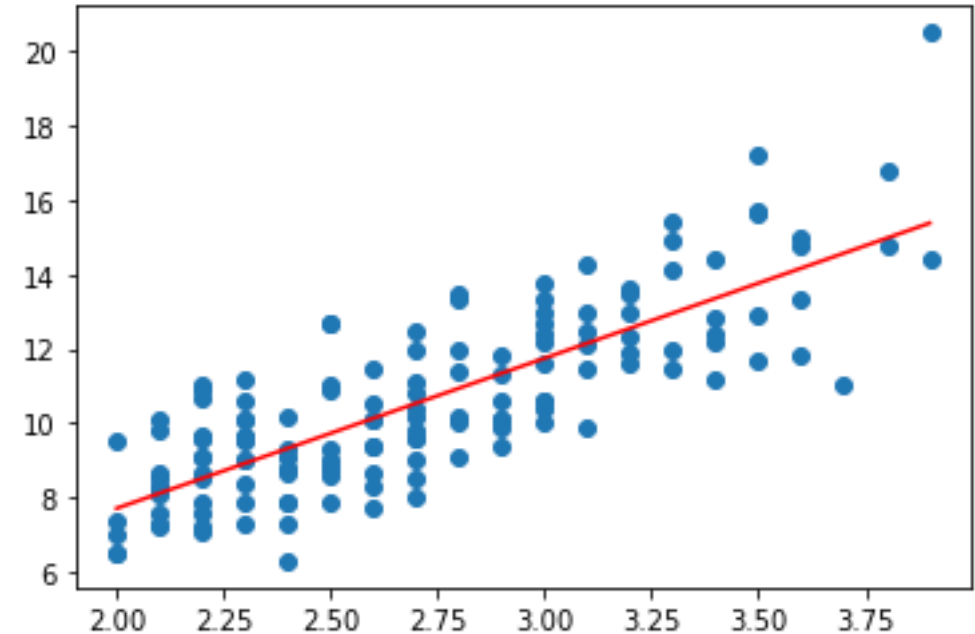
from <https://developer.ibm.com/articles/an-introduction-to-deep-learning/>

...or maybe...



Linear Regression as an example of ML

- Given a new value of data point with known x but unknown y can you predict y ?
- The **model** here is the red line
- The model is **learned** using a process which finds the best fit for the existing datapoints: the “**training data**”
- We sometimes say we **fit** the model to the training data, or that we **train** the model using the training data
- Key point: In ML, like here, we often decide on the shape of the model (here a straight line) but we let the machine learning algorithm find what the *best* straight line is...



Some important ideas



- Models are only as good as the data they're trained on
 - Possibility for bias, changing underlying conditions
 - Rare events can be harder to model (less training data)
- Machine learning models are often opaque:
 - It can be difficult to “understand” why they make a certain prediction
- There are different ways of deciding if a model is “good”
 - Would you prefer your AI poisonous mushroom predictor to be 98% accurate, or would you prefer that it was only accurate 95% of the time but erred on the “safe” side when unsure?
- Training models is often computationally expensive
 - Although modern computing resources make this more accessible than ever
 - Training is usually much more resource intensive than prediction/inference

- The process of training models is often iterative
 - That is, machine learning algorithms often repeatedly use the training data to get better and better models (for some definition of better)
 - Increased training time can lead to better models
- Sometimes in the process of training models, you can build a model which suffers from **overfitting**
 - The model represents the training data very well but does not generalise well to new, unseen data