

DI504 Foundations of Deep Learning

Introduction





This is DI504, Foundations of Deep Learning!

!!! About the Hybrid System !!!



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 - o (preferably took Introduction to Data Science or Data Informatics courses before).



- DI students, before all
 - The aim of this course is to introduce the fundamental concepts of deep learning (including technical or non technical issues) to an audience of interdisciplinary background. It is designed for the DI program students.
 - Other Inf. Inst. program students (such as IS, etc.) are also welcome.
 - Other interdisciplinary program students are also welcome.
 - MMI students have their own course MMI727.
 - EEE students have their own course EE543.
 - CENG students have their own course CEN501.
 - Or if you have taken one of these, I do not recommend you to take this course. METU is full spectacular courses prepared by fantastic lecturers. Why not go take another one?



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 - In other words, basic Programming skills is a must. Python will be our primary choice of environment. MATLAB can be an alternative, if it is where you have some experience.



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 - In other words, basic Programming skills is a must. Python will be our primary choice of environment. MATLAB can be an alternative, if it is where you have some experience.
 - Fundamental concepts of Calculus (matrices, derivation, algebra, etc) will also be very important in understanding the basics.



- This is a both a theoretical and a practical introductory level course for deep learning
 - Courses such as optimization, machine learning can be considered as weak co-requisites.
 - You'd better have either previously passed, or are studying at the same time as another related unit. (at least you will have to)



Outline

- Introduction to deep learning
- (Artificial) Neural Networks
- Learning Deep
- High Level Features
- Performance evaluation
- Introduction to Sequence Models
- Introduction to Deep Natural Language Processing
- Applications for Deep Learning



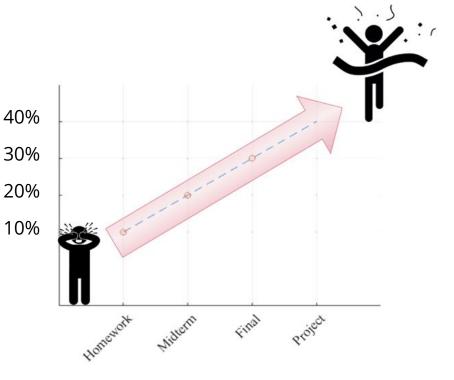
Grading

Project

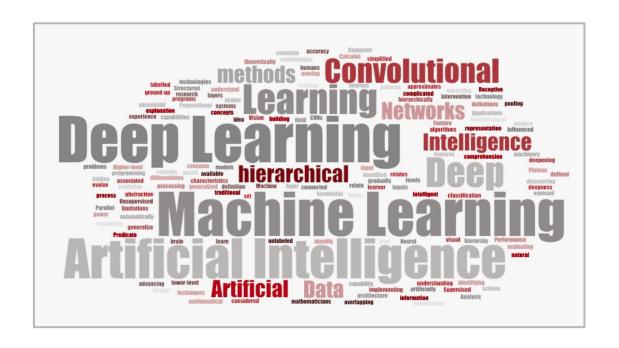
• Final

• Midterm

Homework







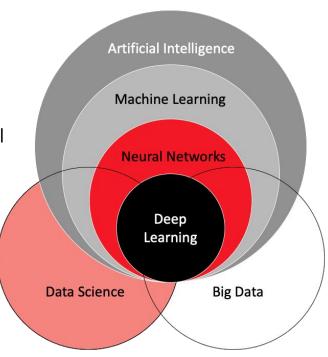


Artificial Intelligence: is human intelligence exhibited by machines.

Machine Learning is an approach to achieve artificial intelligence

Neural Networks (or Nets) is a one of the techniques for implementing machine learning.

Deep Learning is a subfamily of machine learning methods, based on "deep" (many layered) neural networks.

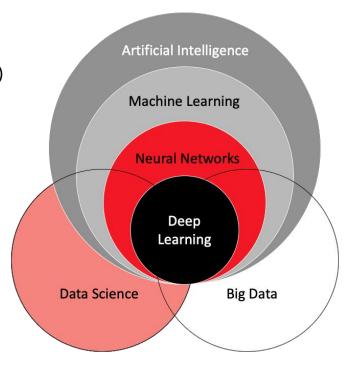




The fields of artificial intelligence (AI), machine learning (ML) and data science have a great deal of overlap, but they are not interchangeable.

There are some nuances between them. Here is a very simplified explanation of how these three areas differ:

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





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• ML algorithms build a model based on sample data, known as "training data", in order to make predictions or decisions.



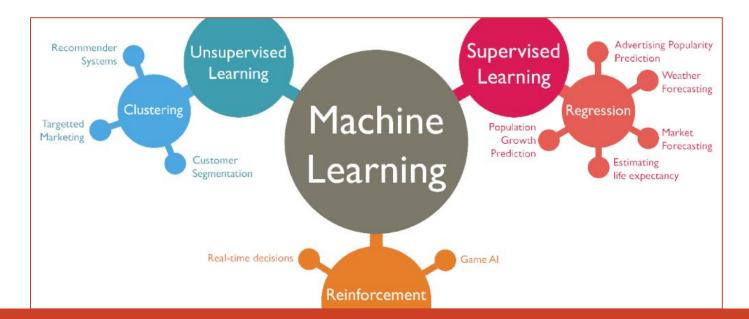
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- ML algorithms build a model based on sample data, known as "training data", in order to make predictions or decisions.
- ML models are not explicitly programmed for tasks, but they are programmed to learn from data.



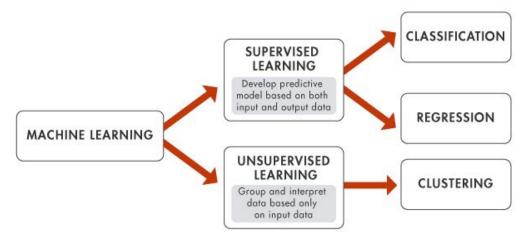
- ML focuses on the development of computer programs that can access data and use it to learn for themselves, using the data.
- ML algorithms are often categorised according to how they are supervised.





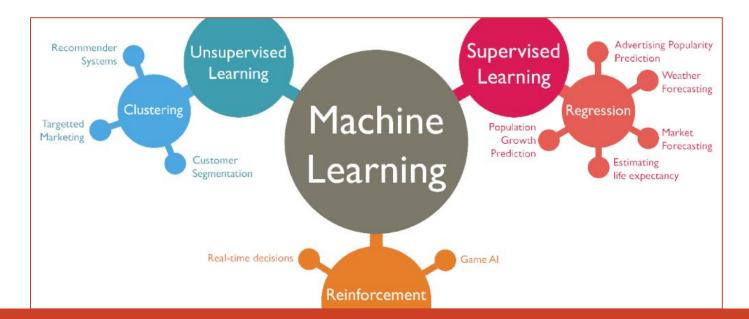
Supervision in ML

- Supervised learning (SL) is the machine learning task of learning a function that maps an input to an output based on example input-output pairs
- Unsupervised learning (UL) is a type of algorithm that learns patterns from untagged data. The
 hope is that, through mimicry, the machine is forced to build a compact internal representation
 of its world and then generate imaginative content.





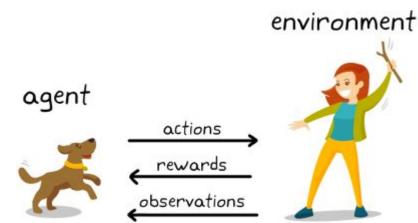
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Reinforcement Learning

- Reinforcement learning (RL) is an area of machine learning concerned with how intelligent
 agents ought to take actions in an environment in order to maximize the notion of cumulative
 reward.
- Reinforcement learning is one of three basic machine learning paradigms, alongside supervised learning and unsupervised learning.

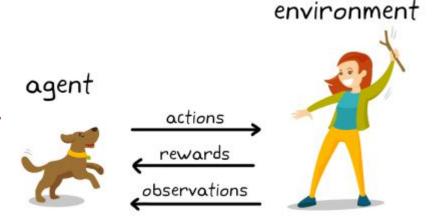




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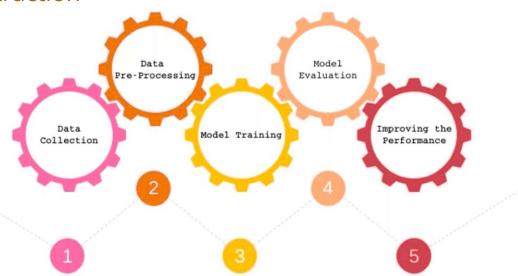
It is outside the scope of DI504.
 Though, MMI706 "REINFORCEMENT LEARNING" is strongly recommended.





ML Workflow

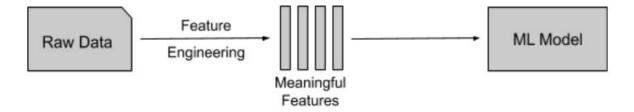
- Data Collection
- 2. Pre-processing / Feature Extraction
- 3. Model Training
- 4. Model Evaluation / Testing
- Fine-Tuning / ParameterOptimization





Features in ML

- In ML a "*feature*" is an individual measurable property or characteristic of a phenomenon, that represents "something" (tangible or not) in the data.
- Features are higher level representations of "raw data".



Before deep learning, in 2000s, the AI society of today focused mainly on "feature engineering",
i.e. finding the right hand-crafted representation for the data that would make our ML model
perform the best.



Features in ML

- Intelligent being understand data in higher representations, sometimes referred to as concepts.
- Features can be of different levels. Edges of an image are low-level features.
- Where as objects in an image are higher level features
- Concepts are the highest level relational features.
- Nobody hears vibrations or frequencies. They hear pitch, timbre, melody, rhythm.
- Nobody sees pixels in an image. They see faces, objects, figures, actions.





Features in ML

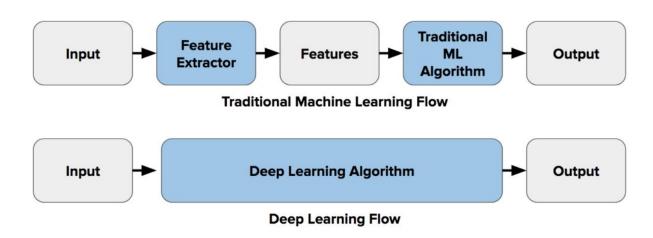


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 - All they see is blonde, brunette, redhead...





- [Wikipedia] Deep learning (DL) is a class of ML algorithms that uses multiple layers to "progressively extract higher-level features from the raw input".
- DL algorithms learn their own features. Hence, they do not require an additional feature extraction step, like traditional ML algorithms do.





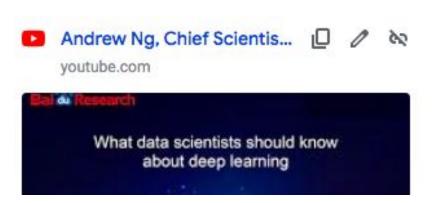
Why DL?



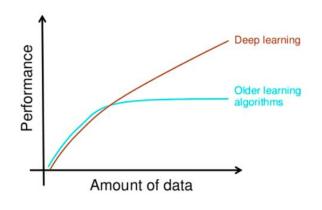
Andrew Ng [@ExtractConf 2015]

"For most flavors of the old generations of learning algorithms performance will plateau. Deep learning is the first class of algorithms that is scalable. Performance just keeps getting better as you feed them more data."

https://www.youtube.com/watch?v=00VN0pGgBZM



Why deep learning

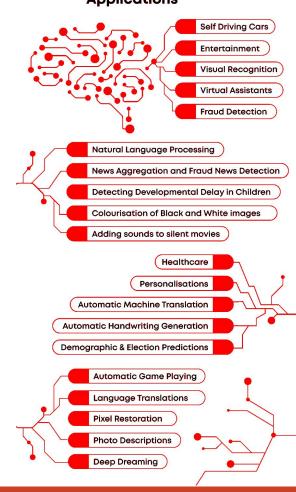




DL Application Domain

 Deep Learning algorithms extract high-level features, complex abstractions as data representations through a hierarchical learning process.

DEEP LEARNINGApplications



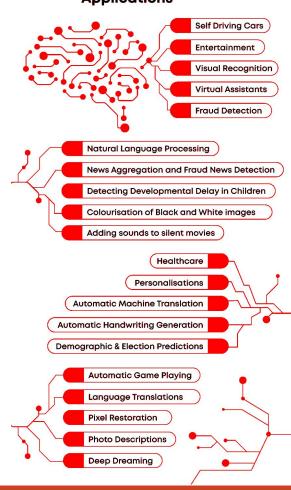


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- Deep Learning algorithms extract high-level features, complex abstractions as data representations through a hierarchical learning process.
- By extracting such features, Deep Learning enables the use of relatively simpler linear models for <u>supervised data analysis</u> tasks, such as classification and prediction, which is important when developing models to deal with the scale of Big Data.

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DEEP LEARNINGApplications

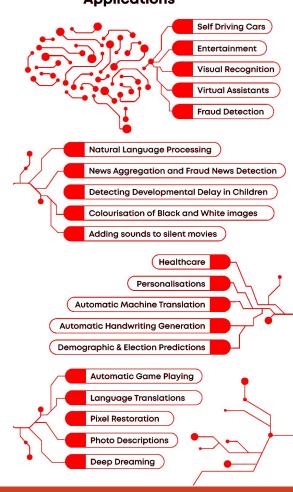




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- A key benefit of Deep Learning is the analysis and learning of massive amounts of unsupervised data, making it a valuable tool for Data Analytics, where raw data is largely unlabeled and un-categorized.

DEEP LEARNINGApplications





DL Practices

When studying deep learning in data science there are three main skills you need to have experience in:

- Data/Feature Analysis/Visualization
- Model Training
- Performance Evaluation



Data/Feature - Analysis/Visualization

If you have taken (I hope you had) *Introduction to Data Science* or *Data Informatics* courses before, you already know that data can be

- Qualitative or Quantitative
- Structured or Unstructured
- Multi-dimensional
- Spatial, Temporal or Spatiotemporal

For each type of data, different DL models are used to extract features. And different analysis/visualization tools are used.



Data/Feature - Analysis/Visualization

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I strongly recommend Python for Data/Feature Analysis/Visualization for DL. MATLAB is a good alternative, with several built-in functions.



Model Training

This is where all the fun takes place.

- Some say it is all local in their PC, all data, all tools etc.
- Some say they have nothing installed in their PC, it is all cloud.
- Which one is better? Where to do it? How to do it?













Model Training Platforms

There are several different alternatives, such as:

- which programming language, Python, MATLAB, C, R?
- which DL library, PyTorch, Keras, Tensorflow, Theano, other?
- which platform, Google colab, Amazon sagemaker, Anaconda Spyder, Pycharm?

All have their pros and cons.

When you start working on one, you will most probably stick to it for a long time. So this decision is important.



Programming Language



I think Python is a winner here, because of many reasons:

- Python offers concise and readable code. While complex algorithms and versatile workflows stand behind DL, Python's simplicity allows developers to build both fast prototypes or reliable systems.
- Python code is understandable by humans, which makes it easier to build models for DL.
- Most state-of-the-art DL is shared in Python. (Maybe this is the most important reason).



DL Library



Python won the battle as the preferred language of DL. The availability of libraries and open source tools make it ideal choice for developing DL models.

But which one?

The battle is mainly between PyTorch vs Tensorflow.

- Both TensorFlow and PyTorch have their advantages as starting platforms to get into neural network programming.
- Traditionally, researchers and Python enthusiasts have preferred PyTorch, while TensorFlow has long been the favored option for building large scale deep learning models for use in production.



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But which one?

- Some libraries have specific advantages in certain applications domains.
- For example "Theano" is preferred by embedded deep learning people.



DL Platform

This one is difficult choice. The first to choose is local vs cloud?

 There are several factors that affects this decision, such as flexibility, security, privacy, cost, ease-of-use...







DL Platform "Cloud"

- For the project using a cloud service is the easiest solution.
- No configurations, installations, etc.
 (Strongly recommended for beginners).
- However, these platforms provide limited GPU-hours for free.





DL Platform "Local"

- If you have a NVIDIA (cuda compatible) GPU, you have everything you need!
 - Check the cuda-compatibility of your GPU, and make sure that it can run the python library (e.g. PyTorch) you will use!
- The downside is you will have to install an integrated development environments (IDE) and learn how to configure this.
- My personal choice: Anaconda-Spyder with Pytorch. Once you set it, it is smooth as MATLAB.



What to do until next week?

- Choose your programming environment and set it up!
- Check the additional reading page for more references and Google around to make a good decision.
- Think of a project idea. Preferably related to your thesis.
 - You don't know how to do it yet, even if it can be done or not. Do some research. Did people do it before? What could they achieve? Maybe you will implement a modification on the problem? Start thinking about it.
 - Never forget what a wise man has once said:

"Don't worry about it if you don't understand."

Andrew Ng



Additional Reading & References

- https://www.youtube.com/watch?v=O0VN0pGgBZM
- https://www.mygreatlearning.com/blog/deep-learning-applications/
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