

# CS 615 – Deep Learning

Introduction

# Objectives

- ML vs DL
- Course Overview

# AI vs ML vs DL vs DS (oh my!)

- First off, what's the difference between Artificial Intelligence (AI), Machine Learning (ML), Deep Learning (DL), and Data Science (DS)?



# AI vs ML vs DL vs DS (oh my!)

- Artificial Intelligence is a large domain
- It includes a lot of topics, but in a nutshell describes a set of algorithms that attempt to make decisions.

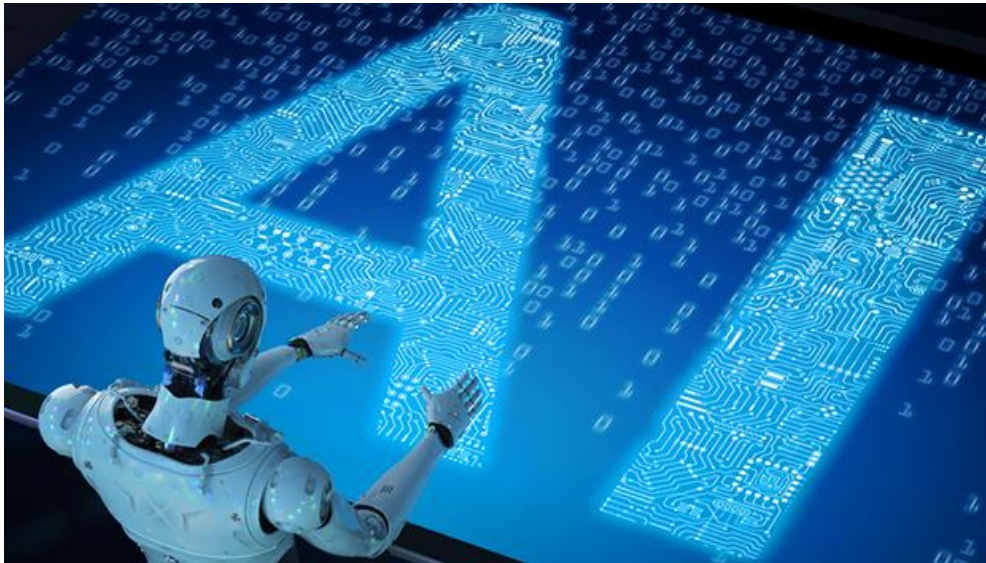
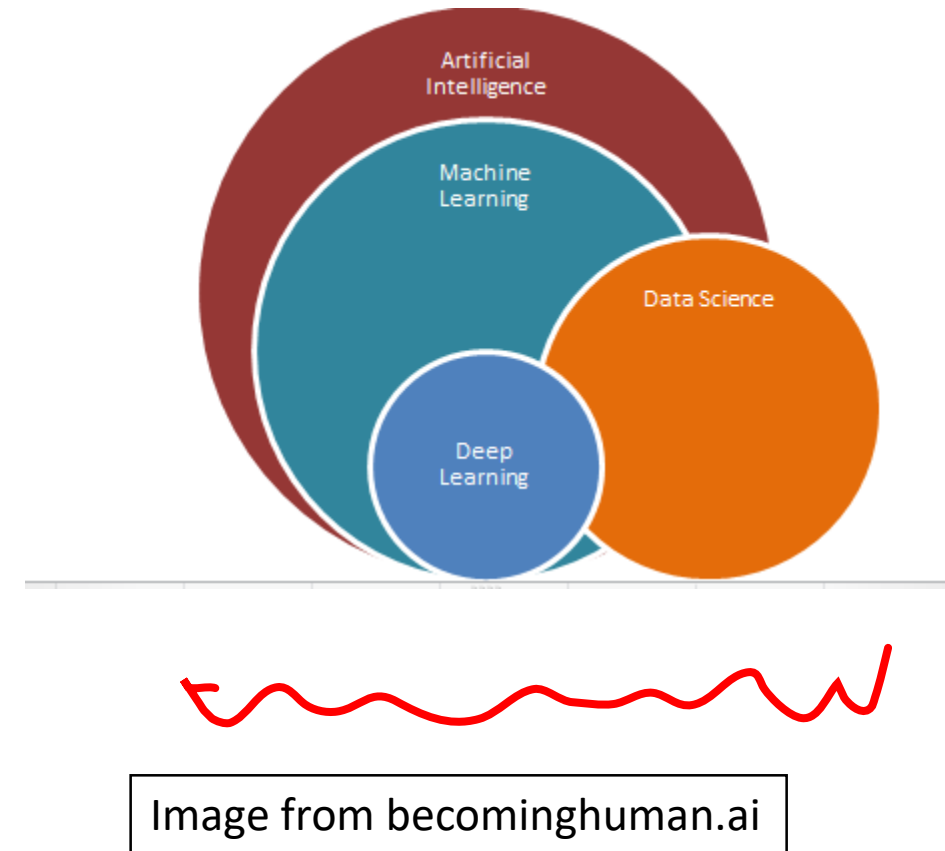


Image from [newyork.cbslocal.com](http://newyork.cbslocal.com)

# AI vs ML vs DL vs DS (oh my!)

- Machine Learning is a *subset* of AI, where algorithms attempt to build (learn) a system using previous data, that can make predictions on new data.
- Deep Learning is a *subset* of Machine Learning
  - A particular set of ML Algorithms
- Data Science is the analysis and use of data
  - This includes stuff from AI/ML/DL but also includes other things.
  - Like simple statistical analysis, database management, etc...



# CS613 vs CS615

- Here at Drexel we opted to make ML and DL separate and independent courses.
- Why?
  - For one thing, 10 weeks isn't enough to cover both
  - On a more logistic note, this way we can alternate when they're offered.
- Therefore there will be some redundancy.
  - Which isn't necessary bad.
  - But sorry for those of you who took CS613 (hope I don't bore you too much)



# Preliminaries

- So we'll start off with some fundamental ML stuff needed for DL
- In particular
  - Basic Terminology
  - Evaluation methods
- But first let's get into some course logistics (yay!)



# Administrative Stuff...



# Contacts

- Instructor:
  - Matt Burlick, [mjburlick@drexel.edu](mailto:mjburlick@drexel.edu), 3675 Market Street, Room 1139
- Teacher Assistant:
  - Maryam Daniali, [md3464@drexel.edu](mailto:md3464@drexel.edu)
- Office Hours will be on Zoom!
  - These will be one-on-one meetings with a waiting room.
  - Links to the rooms are in the Office Hours sections in Bblearn.
  - To the right are the times.
- Lectures:
  - Section 001 Thursdays 6:30pm-9:20pm      Bblearn Collaborate
  - Section 900 Online      N/A

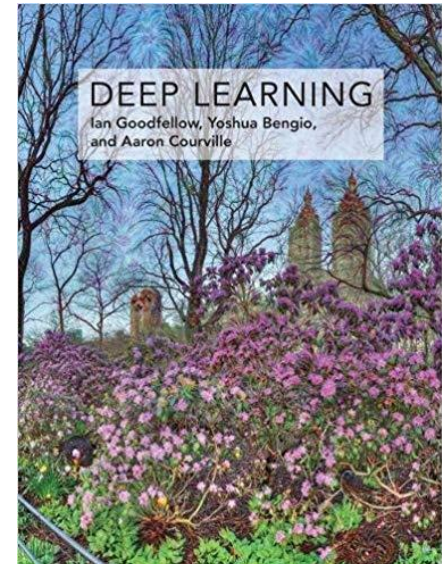
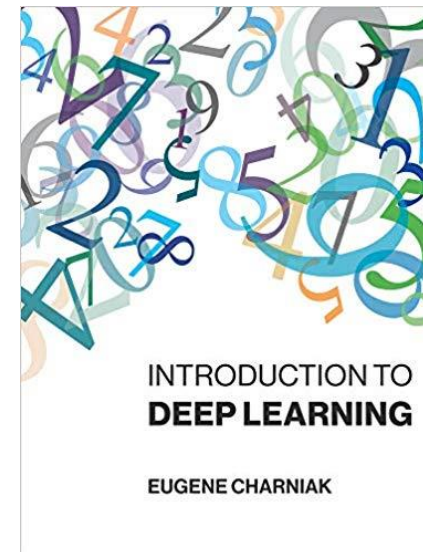
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5pm	Daniali			Burlick	
6pm			Daniali		
7pm			Daniali		
8pm					
9pm				Burlick	
10pm				Burlick	

# Pre-Requisites

- CS 520 (Computer Science Foundations)
- CS 570 (Programming Foundations)
- CS 571 (Advanced Programming Techniques)
- The idea is that you should be a proficient programmer such that you can pick up a new language “on the fly” and use it as a tool.
- You should also be comfortable with linear algebra, probability, statistics, and calculus.
  - There’s an area on Bblearn with “Math Primers”. Review those ASAP

# Course Resources

- Official Textbook:
  - Deep Learning, by Ian Goodfellow, et al, The MIT Press. ISBN: 978-0262035613
- Recommended Textbooks:
  - Introduction to Deep Learning by Eugene Charniak, ISBN: 978-0-262-03951-2
- Blackboard
- Zoom (office hours)
- Slack
  - Use this as your first place to pose questions
    - Hopefully not just I can help
  - But don't post code.
  - [drexelcs615spring2020.slack.com](https://drexelcs615spring2020.slack.com)



# Course Software

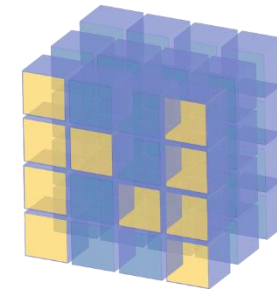
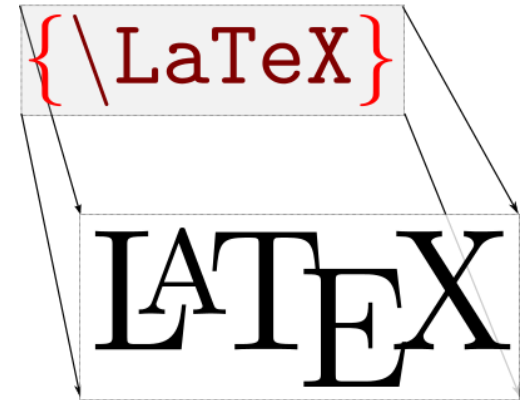


- Programming Environment

- Your choice (most choose Python or Matlab)
- I recommend MatLab (programming environment)
  - Obtain for free from <http://drexel.edu/irt/computers-software/software/>

- Typesetting Environment

- LaTeX
  - First download LatTex itself (warning, it's huge!)
    - <https://latex-project.org/ftp.html>
  - Then (optionally) get a IDE wrapper for it
    - <http://www.xm1math.net/texmaker/download.html#window>
    - <https://github.com/TeXworks/texworks/releases>
  - Or use an online Latex typesetter!
    - [www.overleaf.com](http://www.overleaf.com)



NumPy

# Evaluation

• Homework Sets	30%	$3 \times 10\%$
• Midterm Presentation	10%	
• Exam	30%	
• Final Project	30%	

# Homework Sets

- Include theoretical questions as well as implementation of DL algorithms
- Submission is to be made on Blackboard as a single compressed file consisting of:
  - PDF with your solutions to the theory questions
    - **Must be typeset with LaTeX**
  - README text file on how to run your code
  - Makefile, in necessary
  - Source code
- Note: If you work in any language other than Matlab, make sure that your code works on tux. We can't be expected to have all your dependencies, etc. on our local system.

# Exam

- Based on theory and mathematics.
- Similar to theory questions that are part of assignments.

# Project Proposal

- Present a slidedeck containing the following content:
  - What problem are you tackling?
  - Where are you getting your data from and what does it look like?
  - What resources are you basing your approach off of?
  - What your basic approach?



# Final Project

- There are three components:
  - Presentation
  - Paper
  - Source code
- Presentation
  - Slide deck should cover
    - What problem are you tackling?
    - Where are you getting your data from and what does it look like?
    - What resources are you basing your approach off of?
    - Details of your approach/algorithm.
    - Results and observations
    - Future Extensions

# Final Project

- Paper
  - Should be a Latex typeset “conference style” paper
  - Sections should include:
    - Abstract
    - Background
    - Related work
    - Various sections about your approach, including all applicable mathematics. Imagine that someone wants to be able to recreate your work
    - Evaluation
      - Including info about your dataset
      - Applicable figures
      - Applicable tables/stats
    - Conclusions
    - Future Work
    - Bibliography

# Course Policies

- Assignments are to be done individually unless otherwise noted
- While you are encouraged to use a versioning system like github or bitbucket, please make your work for this course **private**.
- Any dispute about an assignment grade must be formally made (email) and resolved within 5 days of receiving your grade. After this period your grade cannot be adjusted.
- You will lose 1% for every hour late (round up) on an assignment up to 48hrs (after which you will receive a zero)

# Math, Math, Math,....

- This is mostly an applied linear algebra and multi-variate calculus class.
- If you aren't interested (and good at) that stuff, this course may not be for you.



# Notation/Mathematics/Matlab

- I have placed on Blackboard a number of resources to help you review the expected and needed math as well as get you started in Matlab (if you so choose).
- Math Primers
  - Course Notation – There will be a lot of symbols used in this course. This document tries to give you an overview of them.
  - Similarity and Distance Functions – Often we will need to compute the distance and/or similarity between observations. This document includes several commonly used ones.
  - Calculus
  - Linear Algebra
  - Probability and Statistics

# Notation/Mathematics/Matlab

- I have placed on Blackboard a number of resources to help you review the expected and needed math as well as get you started in Matlab (if you so choose)
- Programming References:
  - Matlab Functions – Here's a list of most of the Matlab functions I used in developing this course.
  - Python Functions – Here's a list of most of the Python functions/libraries I used in developing this course.