

CMPS 460 – Spring 2022

MACHINE

LEARNING

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1.b

Admistrivia







Day	Time	Activity	Room
Sun-Tue-Thr	12:00pm-12:50pm	Class	B05-0237
TBD	TBD	Office Hour	My office

- O Other times are available by appointment
- o Best way to contact me is by email



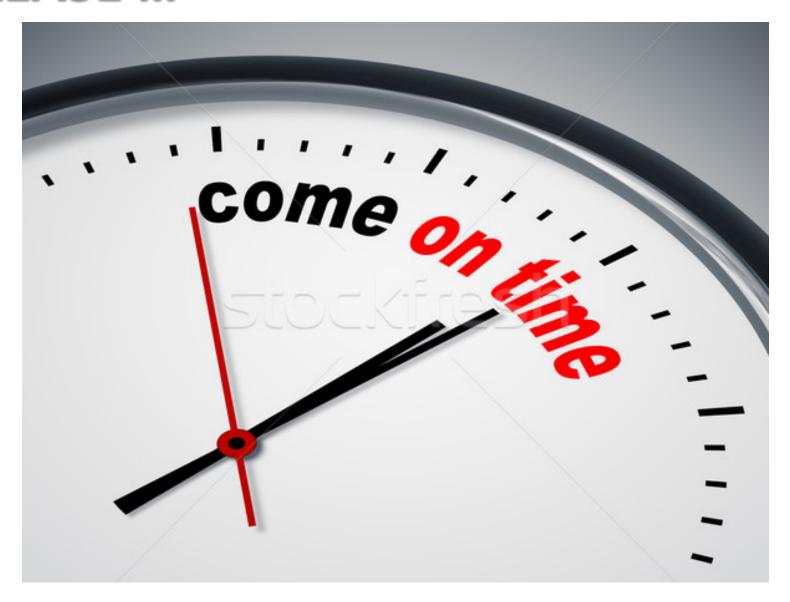


Day	Time	Activity	Room
Sun-Tue-Thr	1:00pm-1:50pm	Class	C11-0112
TBD	TBD	Office Hour	TBD

- O Other times are available by appointment
- o Best way to contact me is by email

PLEASE ...





While in class!



- Phones, etc.
 - Vibrate or off
 - Do not use during class
 - no texting, no tweeting, no updating your status, ...
- Computers
 - Okay to take notes or give presentations
 - Okay to look up stuff related to class when asked to
 - Nothing else
- Other devices
 - Same basic idea

Class page on Blackboard



- Planning to use it extensively!
- Syllabus (tentative)
- Announcements
- Slides
- Resources
- Videos
- Assignments
- Handouts
- Due dates

Prerequisites



No ML/DM/IR/BD background is assumed/required

- Calculus, linear algebra, probability
 - review resources provided

- Python Programming
 - if not, you will pick it up quickly (resources provided)

Course Goals



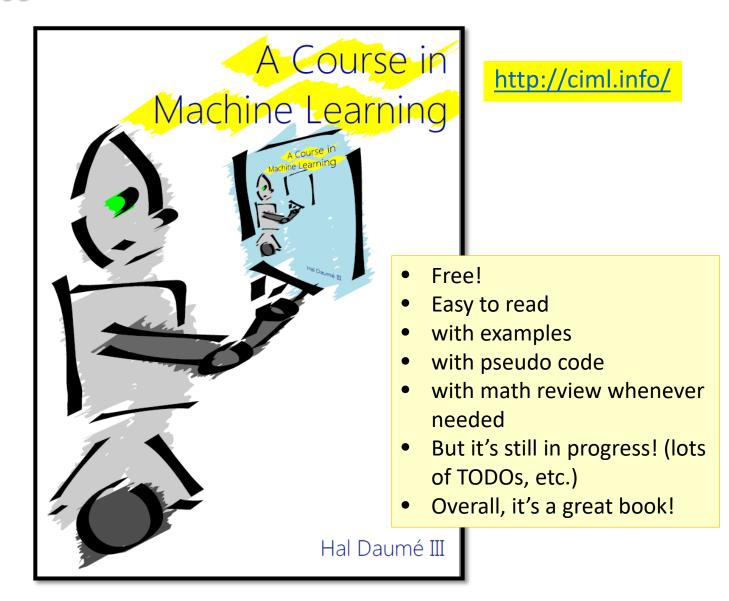
 Broad overview of existing methods for machine learning and an introduction to adaptive systems in general.

- By the end of the semester, you should be able to
 - Look at a problem
 - Identify if ML is an appropriate solution
 - If so, identify what types of algorithms might be applicable
 - Apply those algorithms

- This course is not
 - A survey of ML algorithms
 - A tutorial on ML frameworks such as TensorFlow, PyTorch...

Textbook





Topics



Foundations of Supervised Learning

- Decision trees and inductive bias
- Geometry and nearest neighbors
- Perceptron
- Practical concerns: feature design, evaluation, debugging
- Multiclass classification

Advanced Supervised Learning

- Linear models and gradient descent
- Support Vector Machines & Kernels
- Naive Bayes models and probabilistic modeling
- Neural networks (& Deep Learning)

Topics



Unsupervised learning

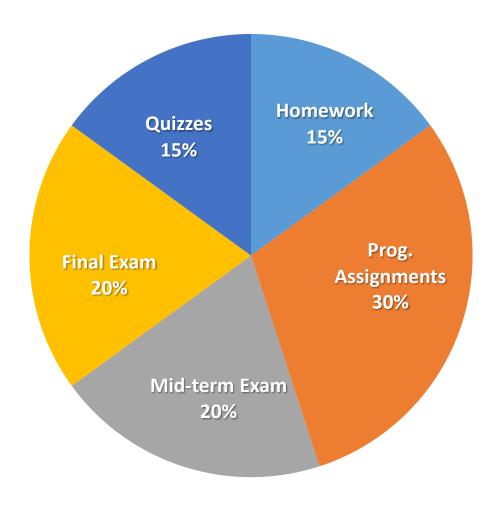
- K-means
- PCA

Selected advanced topics (as time permits)

- Ensemble methods
- Expectation maximization
- Intro to reinforcement learning







Grade Distribution



- Homework: 15%
 - **-** 4-5
 - for individuals
- **Quizzes: 15%**
 - 4, best 3, no makeup
- Programming Assignments: 30%
 - 3 assignments
 - In teams of 2
 - in Python
- Exams: 40%
 - Mid-Term Exam: 20%
 - Final Exam: 20%
 - Not cumulative
 - Both are closed book (with 1 page cheat-sheet)

What I expect from you



- Work hard!
- Review your math (calculus, linear algebra, probability)
- Do a fair amount of programming
- Attend the lectures
 - Interact in class: ask and answer questions, take notes if necessary
 - Come to class prepared
 - Do the required readings **before** class
- Complete HW independently & complete Programming Assignments collaboratively with your mate.
 - Plan your time ahead!
 - Start from day 1
 - Don't wait to the last minute for submissions





is a **VERY SERIOUS** matter!

Honesty



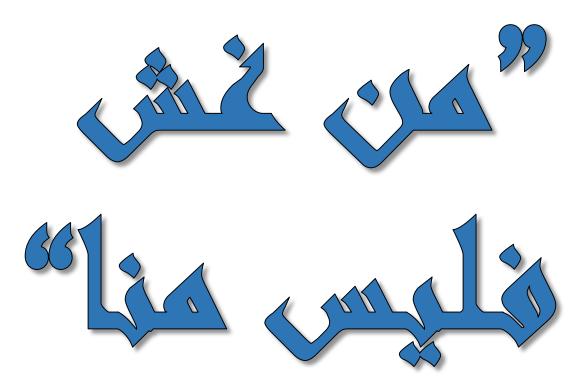


- Copying from each other OR from external sources IS
 CHEATING
- If happened in <u>one</u> problem or part, you will get <u>ZERO</u> in the <u>entire</u> homework or project!
 - Both cheater & cheated-from will get ZERO!
- Assignments for individuals are assignments for individuals, not teams!
- Handed-in work must be your own
 - Do not copy, do not let someone read your work
- Plagiarism and other forms of cheating will result in an F



And more importantly ...

مالم ميلا علا مله عليه وسلم



Whoever deceives, he/she is not one of us (any more)





