# Course Admin

EL-GY 6143/CS-GY 6923: INTRODUCTION TO MACHINE LEARNING PROF. PEI LIU





# People

- ☐ Professor: Pei Liu, peiliu@nyu.edu
  - $\circ~$  370 Jay Street  $9^{th}$  Floor (Only available on Zoom during Fall 2021)
- ☐ Head TA:
  - Mustafa Ozkoc, mfo254@nyu.edu
  - Host office hours regarding homeworks and labs
- ☐ There will be several other graders as well





# Course Learning Objectives

- ☐ Formulate a task as a machine learning problem
  - Identify learning objectives, source of data, models, ...
- □ Load, pre-process and extract features from common data sources
  - images, text, audio, ...
- ☐ Mathematically describe simple models of the data
- ☐ Fit the models to data and use models for prediction and estimation
  - Use common tools
- Evaluate goodness of fit and refine models
- Evaluate the performance of methods using statistical techniques





# **Grad vs Undergrad**

- □Class is simultaneously offered at the graduate and undergraduate level
- □ Undergrad EE-UY/CSE-UY 4563: Intro to Machine Learning
  - Covers fundamental algorithms and some analysis
  - In depth coverage of software tools including python, Google Cloud, Tensorflow
  - Python-based lab exercises + mandatory project
- ☐ Grad EL-GY 6143: Intro to Machine Learning
  - More algorithms and more mathematical analysis. Faster paced.
  - Software tools must be learned at home. Less coverage in class
  - Python-based lab exercises + optional project
- Lecture notes are mostly common with supplementary material for grad students indicated
- ☐ Many labs are common





#### Texts and Other Resources

- □Undergrad: James, Witten, Hastie and Tibshirani, "An Introduction to Statistical Learning",
  - http://www-bcf.usc.edu/~gareth/ISL/ISLR%20First%20Printing.pdf
  - Very clear explanation of concepts.
  - But examples are in R. And there is no review of probability
- ☐ Grad: Hastie, Tibshirani, Friedman, "Elements of Statistical Learning"
  - https://web.stanford.edu/~hastie/Papers/ESLII.pdf
  - More advanced text with more analysis
- Raschka, "Python Machine Learning", 2015.
  - http://file.allitebooks.com/20151017/Python%20Machine%20Learning.pdf
  - Excellent examples of using Python
- ☐ Bishop, "Pattern Recognition and Machine Learning" (more advanced)
- □ Coursera course: Generally do not cover probability
- ☐ Undergrad probability





#### More Resources

- ☐ Entertaining and very good deep learning lectures by Siraj Raval
  - https://www.youtube.com/channel/UCWN3xxRkmTPmbKwht9FuE5A
- ☐ Universite de Paris labs:
  - https://github.com/m2dsupsdlclass/lectures-labs
  - Focus on deep learning
  - Similar format to this class
- ☐ Andrew Ng's machine learning class:
  - https://www.coursera.org/learn/machine-learning
  - A little less mathematical than this class
- ☐ Many, many others online...





### Pre-Requisites

- ☐ Undergrad probability required for both UG and Grad version:
  - Basics of random variables, densities, Gaussian distributions, correlation, expectation, conditional densities, Bayes' theorem
  - Will provide a short review
  - NYU classes: Data analysis or Intro Probability are sufficient
- Undergraduate calculus and linear algebra
  - Vectors, matrices, partial derivatives, gradients.
  - Again, we will provide a brief review
  - A good review of linear algebra by 3Blue1Brown https://youtube.com/playlist?list=PLZHQObOWTQDPD3MizzM2xVFitgF8hE\_ab
- No machine learning experience is necessary
  - If you have ML experience, do NOT take this class.
  - Take Graduate probability (Fall) then Advanced machine learning (Spring)





### Pre-Requisites Programming

#### Python

- All labs are in python, similar to object-oriented MATLAB, but many more libraries.
- And free!

#### ☐What you need to know

- You do not need to know python before class. But, we will go over it quickly.
- You should have experience in some programming language (eg. MATLAB).
- You should know or being willing to learn object oriented programming

#### Resources:

- Installing python and ipython notebook (make sure you install Version 3.6)
  http://jupyter-notebook-beginner-guide.readthedocs.io/en/latest/index.html
- Python tutorial: <a href="https://docs.python.org/3/tutorial/">https://docs.python.org/3/tutorial/</a>
- Numpy: http://cs231n.github.io/python-numpy-tutorial/





# Grading: Graduate

- ☐ Midterm 30%, Final 30%, Labs, HW 20%, Project: 20%
- ☐ Labs: Simple python exercises
  - Given as jupyter notebook that you complete.
- ■Midterms & final
  - Each over approx. 6-7 weeks
  - Close book and no electronic aids.
  - Follows homework and quiz problems + some very basic python questions
- ☐ Final project:
  - Use machine learning in some interesting way.
  - Must use data and python analysis.
  - Provide final report.





### Machine Learning Project

- ☐ Perform an interesting machine learning task of your choice ☐ Many possible areas:
  - Machine vision, brain-computer interfaces, natural language processing, sentiment analysis, ...
  - Anything that interests you
- ☐ Groups of 2 preferred
  - In NYU Classes, join a group "project1, project2, ..."
  - Submit all material as that group
- ☐ Use real data
  - UCI ML repository
  - Google BigQuery data
  - etc
- ☐ Write code
- ☐ Place all material in a github repo (including documentation) and submit only github repo





# **Project Grading**

#### **□** Formulation

• How well did you formulate the problem? Was it clear? Was that tied to the right objective?

#### Approach

• Does your approach properly solve your problem? Was that made clear?

#### ☐ Evaluation and Interpretation

- Did you comprehensively test the results? How well did you select / create the data?
- Did you test against alternative approaches?

#### Presentation

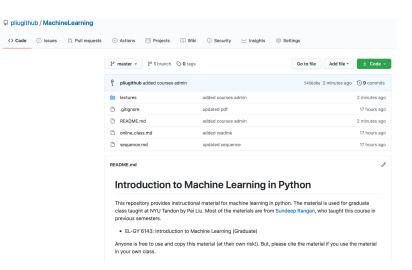
- Were the ideas clear? Were all the details conveyed. Did you highlight the main points?
- You can select a number of formats. Whatever makes sense. A github page

#### **□**Bonus

Given for particularly hard / novel research

### Github

- ☐ Labs and demo posted on github
- https://github.com/pliugithub/MachineLearning
- □ Also includes instruction for installing software
- ■Several tutorials of github on the web.
- ☐ Available on Windows, Mac and Unix.
- ☐But, you can just clone the repo



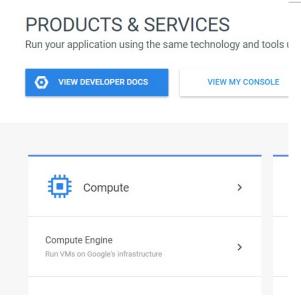


# Google Cloud Platform



- ■All labs in this class can be run on either:
  - Your own computer: Windows, MAC
  - Google Cloud Platform (GCP)
- □GCP pros and cons:
  - Access to powerful machines / large storage for projects.
    Includes GPUs
  - Access to many services such as BigQuery
  - Can scale your computational resources
  - But, somewhat harder to sync editors / debuggers
- ☐Getting started: <a href="https://cloud.google.com/">https://cloud.google.com/</a>
- ☐ Instructions on

https://github.com/pliugithub/MachineLearning/tree/master/GCP



### Other Software

- □On your machine (local or GCP), you will need to install several pieces of software:
- ☐ Python with various packages
  - Make sure you get 3.7
  - Anaconda
  - Jupyter notebook
  - See notes in NYU Classes
- ☐Tensorflow and Keras (needed only later in the class)
- ☐Git hub
  - Guides: https://guides.github.com/
  - Available on Windows, Mac or Linux (including GCP instances)
  - All demos will be available on: <a href="https://github.com/pliugithub/MachineLearning">https://github.com/pliugithub/MachineLearning</a>

# Grading: Undergraduate

- ☐ Midterm 1: 25%, Midterm 2: 25%, Labs, HW: 25%, Final project: 25%
- □ Labs: Simple python exercises
  - Given as jupyter notebook that you complete.
- **■**Midterms
  - Each over approx. 3-4 weeks of material
  - Closed book with cheat sheet.
  - Follows homework and quiz problems + some very basic python questions
- ☐ Final project:
  - Use machine learning in some interesting way.
  - Must use data and python analysis.
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