# **Advanced Data Analysis**

DATA 71200

Class 1

# **Course Description**

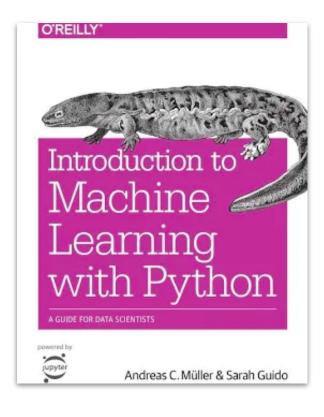
- This course will provide you with skills necessary to apply machine learning techniques to data, and interpret and communicate their results.
- You will also begin to develop intuitions about when machine learning is an appropriate tool versus other statistical methods.
- This course will cover both **supervised methods** (e.g., k-nearest neighbors, naïve Bayes classifiers, decision trees, and support vector machines) and **unsupervised methods** (e.g., principal component analysis and k-means clustering).
  - The supervised methods will focus primarily on "classic" machine learning techniques where features are designed rather than learned, although we will briefly look at recent deep learning models with neural networks.
- This is an applied machine learning class that emphasizes the intuitions and know-how needed to get learning algorithms to work in practice, rather than mathematical derivations.
- The course will be taught in **Python**, primarily using the **scikit-learn** library.

# **Course Objectives**

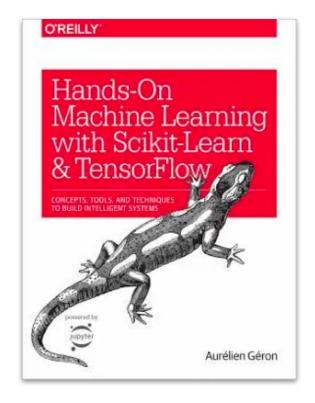
- By the end of the course, you will be able to
  - articulate the main assumptions underlying machine learning approaches
  - demonstrate the basic principles of dataset creation
  - articulate the importance of data representations
  - evaluate machine learning algorithms
  - articulate the difference between supervised and unsupervised learning
  - apply a range of supervised and unsupervised learning techniques

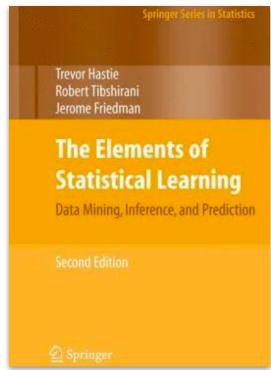
### **Textbooks**

#### Required



#### Recommended





## **Grade Breakdown**

Class Participation	10%
Datacamp Assignments	25%
Project 1: Dataset creation	15%
Project 2: Supervised learning	15%
Project 3: Unsupervised learning	15%
Final Paper	20%

### **Grade Breakdown Details**

#### Class Participation: 10%

The participation grade is a combination of attendance (including arriving on time); attentiveness, engagement, and participation during class; and general preparedness for class discussions.

#### Datacamp Assignments: 25%

 These projects are hands-on activities designed to both provide coding background and reinforce the concepts covered in class.

### **Grade Breakdown Details**

#### Project 1 (Dataset creation): 15%

- Curation and cleaning of a labeled data set that you will use for the supervised and unsupervised learning tasks in project 2 and 3. The dataset can built from existing data and should be stored in your GitHub repostiory.

### Project 2 (Supervised learning): 15%

 Application of two supervised learning techniques on the dataset you created in Project 1. This assignment should be completed as a Jupyter notebook your GitHub repository.

### **Grade Breakdown Details**

### Project 3 (Unsupervised learning): 15%

 Application of two unsupervised learning techniques on the dataset you created in Project 1. This assignment should be completed as a Jupyter notebook your GitHub repository.

#### Final Paper: 20%

- A 5--8 page paper describing the work you did in projects 1--3 (your dataset and your supervised and unsupervised experiments). The paper should describe both what you did technically and what you learned from the relative performance of the machine learning approaches you applied to your dataset. This assignment should be posted as a PDF in your GitHub repository.

### **Course Schedule**

29-Jan Introduction

5-Feb What is Machine Learning?

12-Feb No Class

19-Feb Getting Started with Machine Learning

26-Feb Inspecting Data

4-Mar Representing Data

# **Course Schedule**

11-Mar	Evaluation Methods
18-Mar	Supervised Learning (k-Nearest Neighbors, Linear Models) – <i>Project 1 Due</i>
25-Mar	Supervised Learning (Naive Bayes Classifiers and Decision Trees)
1-Apr	Supervised Learning (Support Vector Machines and Uncertainty estimates from Classifiers)
7-Apr	Unsupervised Learning (Dimensionality Reduction & Feature Extraction, and Manifold Learning) - <i>Project 2 Due</i>

### **Course Schedule**

8-Apr No class

15-Apr No class

22-Apr Unsupervised Learning (Clustering)

29-Apr Deep Learning

6-May Ethics - *Project 3 Due* 

13-May Ethics

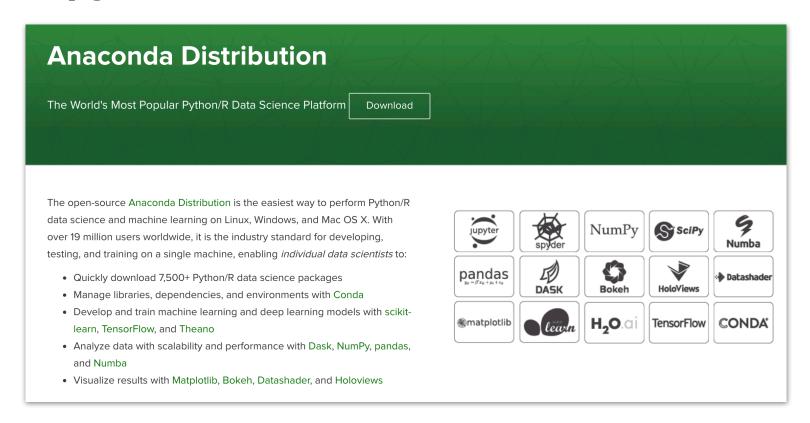
20-May Final Project Due

# **Coding Environment**

#### Python 3

matplotlib, NumPy, Pandas, SciPy scikit learn

#### Jupytr notebooks

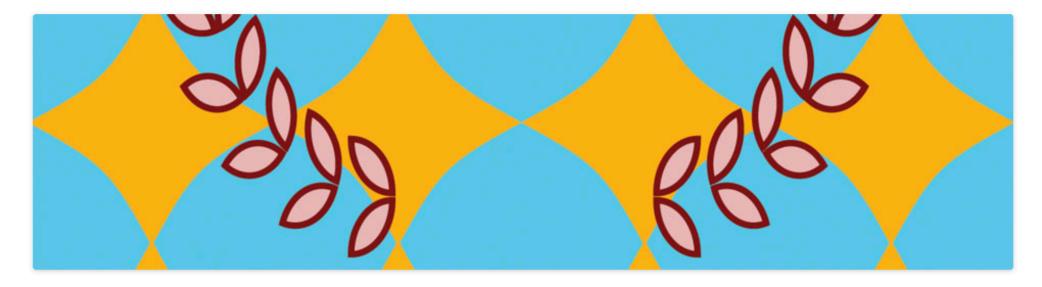


### **Class Website**

#### **DATA 71200 Advanced Data Analysis Methods**

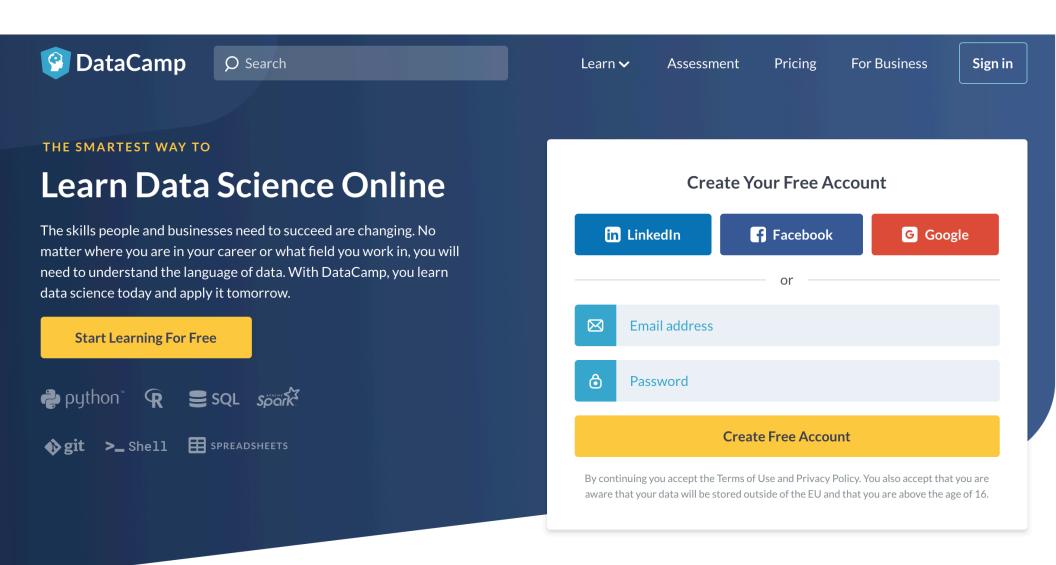
An introduction to supervised and unsupervised machine learning methods

HOME SYLLABUS COURSE SCHEDULE RESOURCES POSTS

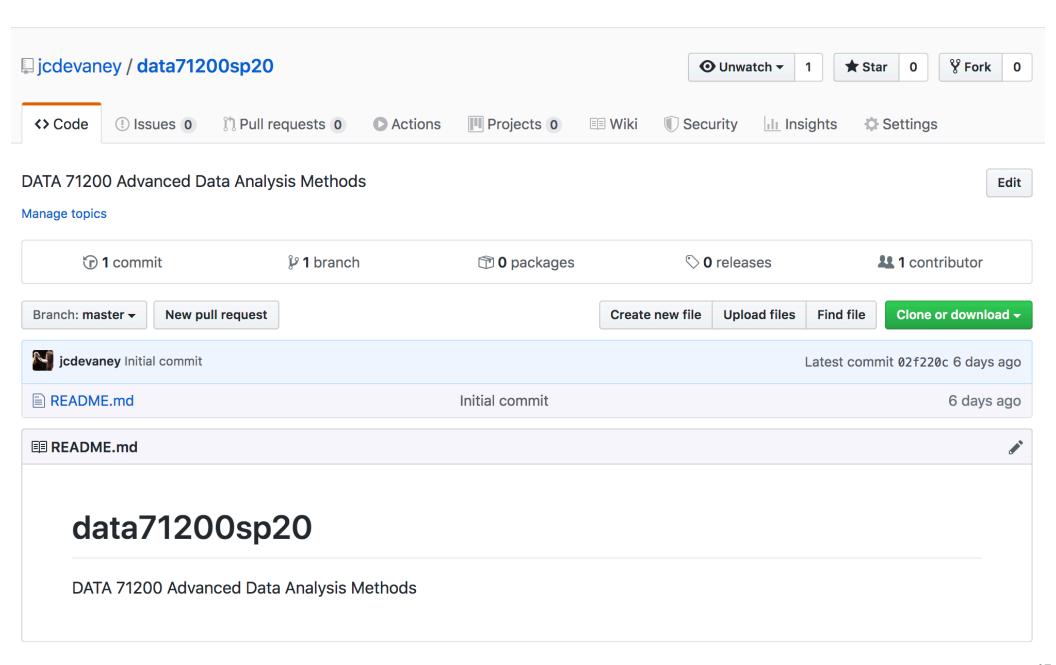


https://data71200sp20.commons.gc.cuny.edu/

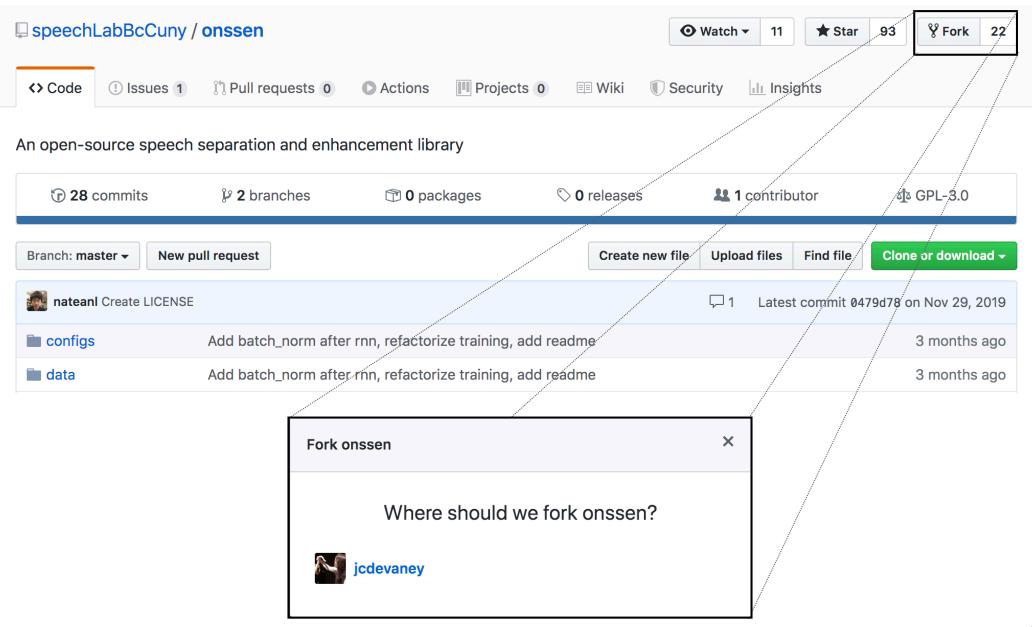
# **Data Camp**



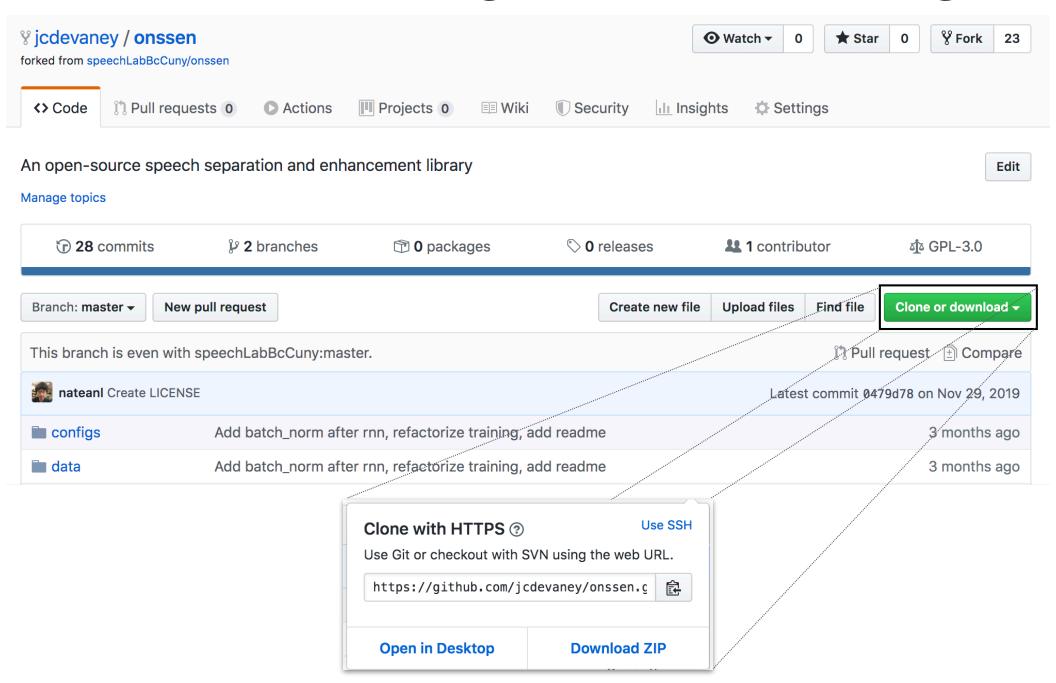
### **GitHub**



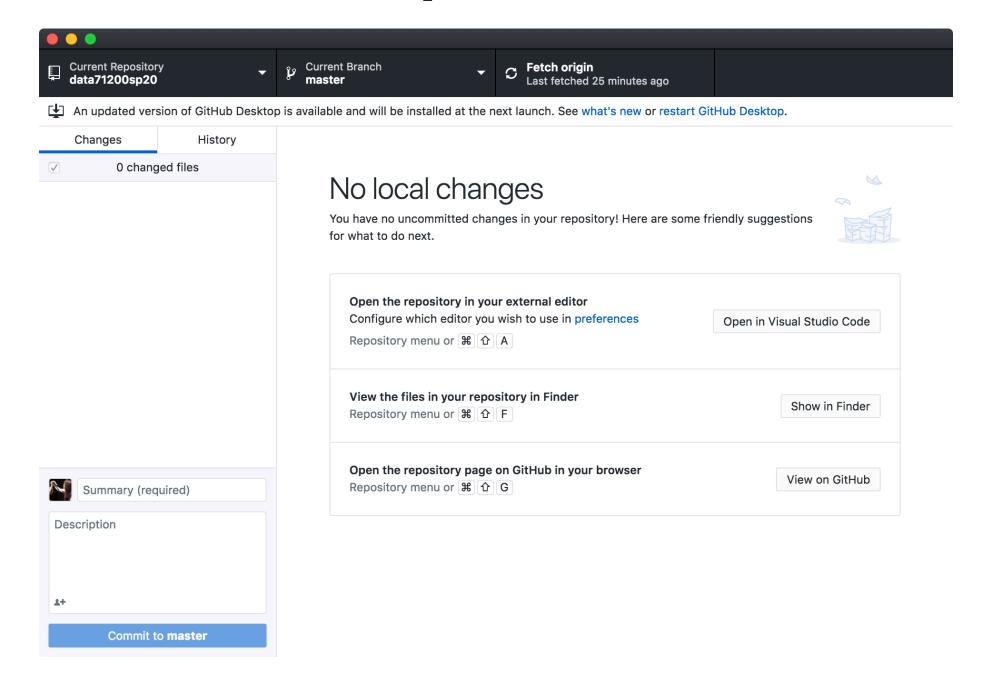
# GitHub - Forking versus Cloning



# GitHub - Forking versus Cloning



# GitHub Desktop



## **Git Command Line**

Git Basics		Rewriting Git History	
git init <directory></directory>	Create empty Git repo in specified directory. Run with no arguments to initialize the current directory as a git repository.	git commitamend	Replace the last commit with the staged changes and last commit combined. Use with nothing staged to edit the last commit's message.
git clone <repo></repo>	Clone repo located at <repo> onto local machine. Original repo can be located on the local filesystem or on a remote machine via HTTP or SSH.</repo>	git rebase <base/>	Rebase the current branch onto <base/> . <base/> can be a commit ID, a branch name, a tag, or a relative reference to HEAD.
git config user.name <name></name>	Define author name to be used for all commits in current repo. Devs commonly useglobal flag to set config options for current user.	git reflog	Show a log of changes to the local repository's HEAD. Addrelative-date flag to show date info orall to show all refs.
git add <directory></directory>	Stage all changes in <directory> for the next commit. Replace <directory> with a <file> to change a specific file.</file></directory></directory>	Git Branches	
git commit -m " <message>"</message>	Commit the staged snapshot, but instead of launching a text editor, use <message> as the commit message.</message>	git branch	List all of the branches in your repo. Add a <branch> argument to create a new branch with the name <branch>.</branch></branch>
git status	List which files are staged, unstaged, and untracked.	git checkout -b    dranch>	Create and check out a new branch named dъгансh>. Drop the -ь flag to checkout an existing branch.
git log	Display the entire commit history using the default format. For customization see additional options.	git merge <branch></branch>	Merge <branch> into the current branch.</branch>
git diff	Show unstaged changes between your index and working directory.	Remote Repositories	
Indoing Changes		Create a new connection to a remote repo. After adding a remote, you can use <name> as a shortcut for <url> in other commands.</url></name>	
git revert <commit></commit>	Create new commit that undoes all of the changes made in <commit>, then apply it to the current branch.</commit>	git fetch <remote> <branch></branch></remote>	Fetches a specific <branch>, from the repo. Leave off <branch> to fetch all remote refs.</branch></branch>
git reset <file></file>	Remove <file> from the staging area, but leave the working directory unchanged. This unstages a file without overwriting any changes.</file>	git pull <remote></remote>	Fetch the specified remote's copy of current branch and immediately merge it into the local copy.
git clean -n	Shows which files would be removed from working directory. Use the -f flag in place of the -n flag to execute the clean.	git push <remote> <branch></branch></remote>	Push the branch to <remote>, along with necessary commits and objects. Creates named branch in the remote repo if it doesn't exist.</remote>

# **Machine Learning**

Human-driven: programmers First widely available Adaptive models: smaller datasets can win and evaluate model output and platforms for creating created new rules to make training data at scale, free(ish) pre-trained models starting with MTurk are state-of-the-art models more accurate 2010-2015 2000-2005 2005-2010 2015-today 1990s

> Rise of Machine Learning: first major training data sets, but they are slow and expensive to create

Human-driven: non-programmers evaluate model output and annotate data, plus first data-hungry neural models