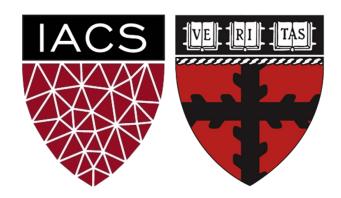
# Lecture 1: Introduction: Virtual Environments, Virtual Machines



Advanced Practical Data Science
Pavlos Protopapas



# Outline

1: Why you should take this class and why not?

2: Who are we?

3: Course structure and activities?

4: Class organization (Workload, Logistics, Grades).

5: Virtual environments.

6: Virtual machines.



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# Why you should take this class

#### Because you want to learn how to:

- Put your model in production
- Integrate and orchestrate applications
- Deploy increasing amount of data
- Take advantage of available models
- Evaluate and debug model using visualization

If you have attended **ComputeFest 2020** and found the topics interesting, this class will also be interesting.

# Why you shouldn't take this class

You are **not** familiar with most of the concepts covered in CS109A/B

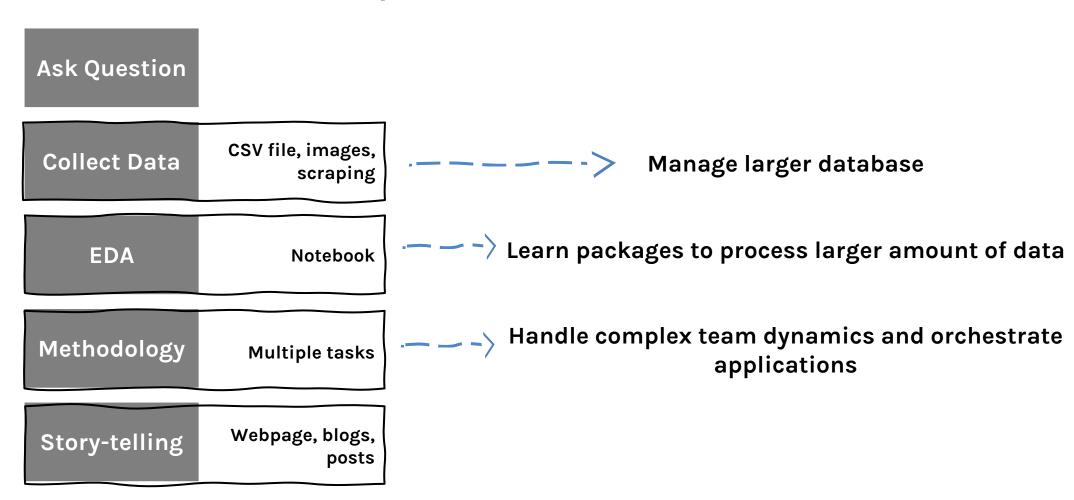
#### For example:

- Basic Machine Learning
- CNNs, RNNs, Autoencoders, GANs, etc
- Basic linux commands

### **Data Science Series to Real World**

#### Data Science Series 109A/B

**Real World** 

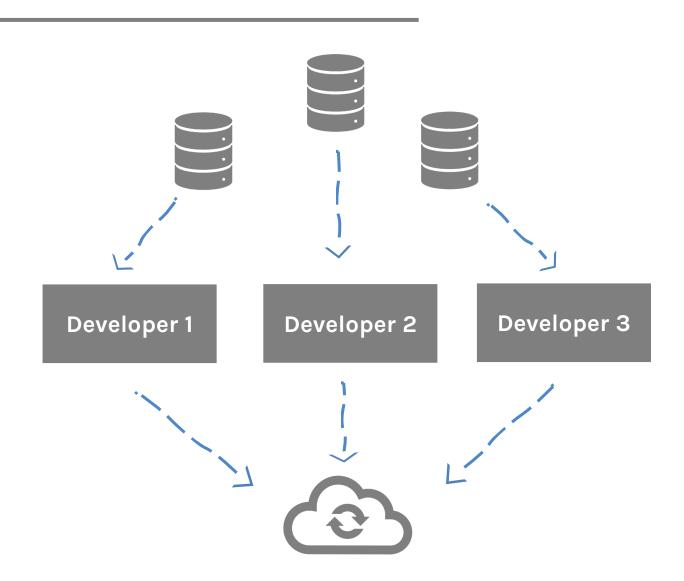


# Data Science Series to Real World (cont)

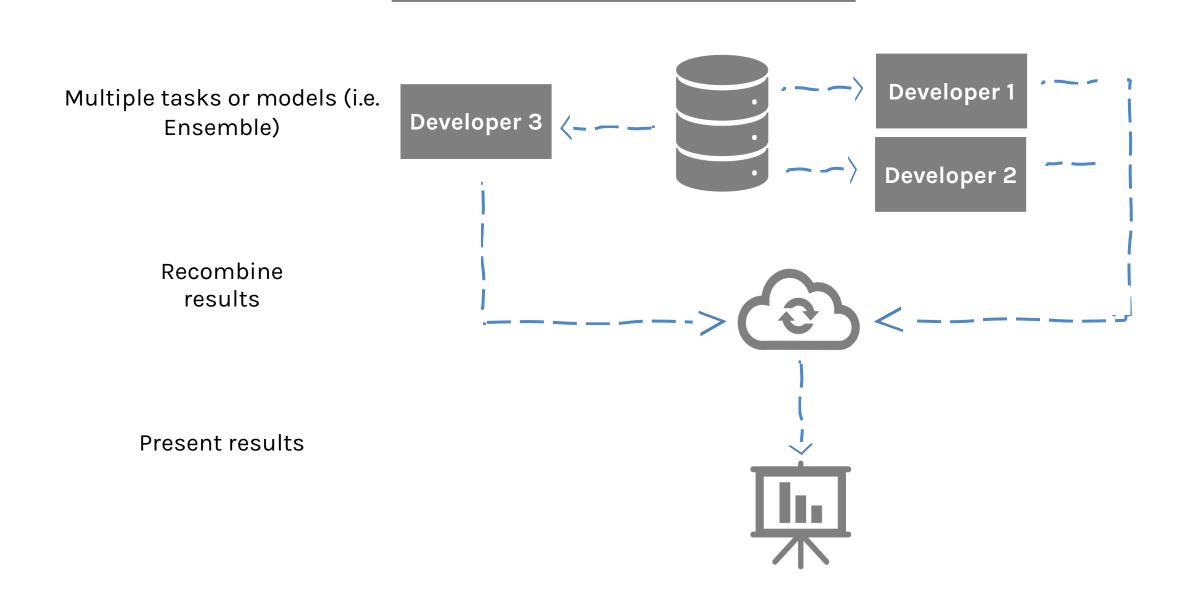
Fragmented database

Multitude requirements and applications

Recombine and deploy



# Data Science Series to Real World (cont)



# Data Science Series to Real World (cont)

Model too expensive to train Or not enough training data

Use pre-trained model

Model (---->

Final Results

Pre Trained Model



Present results

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#### Who?

# **Pavlos Protopapas**

Teaches CS109(a/b), the data science capstone course, and AC295 (advanced practical data science). Research in astrostatistics: machine learning, statistical learning, big data for astronomical problems.

He has picked some new hobbies besides 109s and eating:



Going to BSO (well not anymore), cross country ski (completed Engadin skimarathor cheese making and being a TikToker (check me out @pavlosprotopapas)



### Who? (cont)



Rashmi Banthia
TF for many Data
Science classes here at
Harvard including

CS109A/B.



Yujiao Chen
TFed for CS109A/B.
Currently a Data
Scientist



Hai Bui
Graduate Student
from Bocconi
University in Milan,
currently (not) visiting
MIT.



Javid Lakha
Machine Learning
Engineer at Legatics
(a legal technology
start-up).



# Who? (cont)



Shivas Jayaram
CTO and Co-Founder @
Brain Cradle.



Andrea Porelli
Master's from IACS CSE.



William Palmer
Data Science student at IACS.



**Faras Sadek** 



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# **Course Structure and Activities**

#### Modules:

- Deploy data science (integration + scalability)
- 2. Transfer learning and distillation
- 3. Visualization as investigative tool \* [no presentations or exercises]

#### **Activities:**

lectures, reading and presentations, exercises, forum, practicums, projects

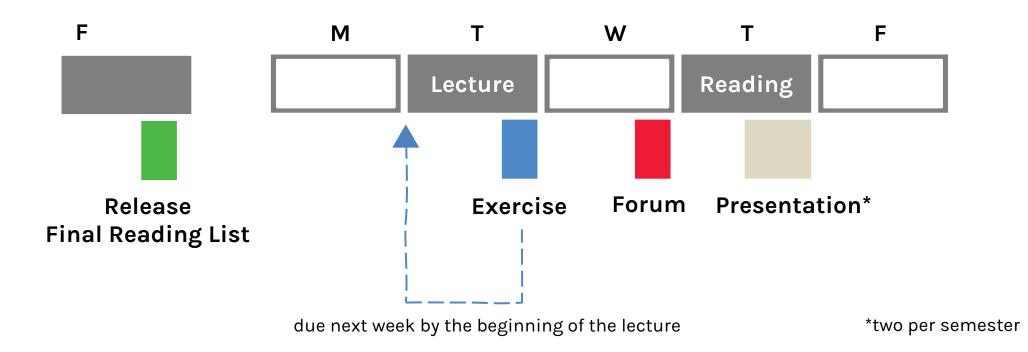
Lectures online: Tuesdays 10:30-11:45 am (repeat 6:00-7:15 pm)

Presentations on Reading and Discussions: Thursdays 10:30-11:45 am (repeat 6:00-7:15 pm)



# **Course Structure and Activities**

# Regular week schedule





# **Topics**

Deploy data science (integration + scalability)

- A. Virtual Environments, Virtual Boxes, and Containers
- B. Kubernetes
- C. Dask



# Topics (cont)

### Transfer learning and distillation

- A. Intro to Transfer Learning: basics and Convolutional Neural Networks review
- B. Transfer Learning across Tasks for images and SOTA Models
- C. Language Models and Transfer Learning with Text Data
- D. Attention and Transformers
- E. Distillation and Compression



# Topics (cont)

### Visualization as investigative tool

- A. Introduction and Overview of Viz for Deep Models: lime and shapley
- B. CNN for Image Data, Activation Maximization and Saliency Maps
- C. Attention for Debugging Language Models



# Calendar

> <u>Link to Calendar <</u>

Week	Date	Lecture #	Topics	Exercise
1	9/3	1	Introduction: Virtual Enviroments and Virtual Boxes	
2	9/8	2	Containers	EX1
	9/10		Use Case: Dockers in a real setting	
3	9/15	3	Kubernetes	EX2
	9/17		Use Case: Kubernetes in a real setting	
4	9/22	4	Dask	EX3
	9/24		Use Case: Dask in a real setting	
5	9/29		Practicum 1: End to end art search engine	Practicum 1
	10/1		Practicum 1	
6	10/6	5	Intro to Transfer Learning: basics and CNNs review	EX4
	10/8		Journal Discussion: Transfer Learning (Statistical approaches to Transfer Learning)	
7	10/13	6	Transfer Learning for Images and SOTA Models	EX5
	10/15		Journal Discussion:	
8	10/20	7	Language Models and Transfer Learning for Text	EX6
	10/22		Journal Discussion	
9	10/27	8	Attention and Transformers	EX7
	10/29		Journal Discussion	
10	11/3	9	Distillation and Compression	EX8
	11/5		Journal Discussion	
11	11/10		Practicum 2	Practicum 2
	11/12		Practicum 2	
12	11/17	10	Introduction and Overview of Viz for Deep Models: lime and shapley	
	11/19	11	CNNs for Image Data, Activation Maximization and Saliency Maps	
13	11/24	12	Attention for Debugging Language Models	
	11/26			
14	12/1		Project	
	12/3		Project	
15	12/8		Project	
	12/11		Final projects presentation	

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### Workload

#### Regular Week

3 hours in class
5 hours reading
5 hours exercise
1 hour forum questions
3 hours presentation\*

~ 16 hours/week

\* 1 presentation per module per group (2 total)

#### Practicum and Project Week

~ 16 hours/week\*\*

\*\* 2 practicums and 1 final project (2 weeks long)

We will be asking for your feedback on the workload



# **Expectations**

# How to read and present class material

- > Link to Reading Guidelines <
- > Link to Presentation Guidelines <



# Logistics

### Fill up forms

Survey

Make group \*

Sign-up presentation\*\*

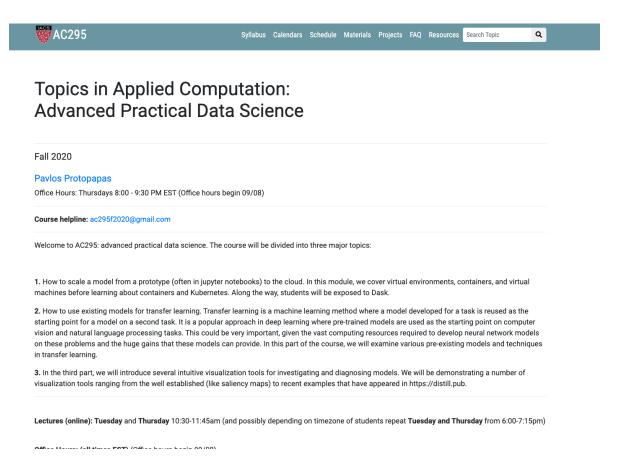


<sup>\*</sup> Fill group components in each row

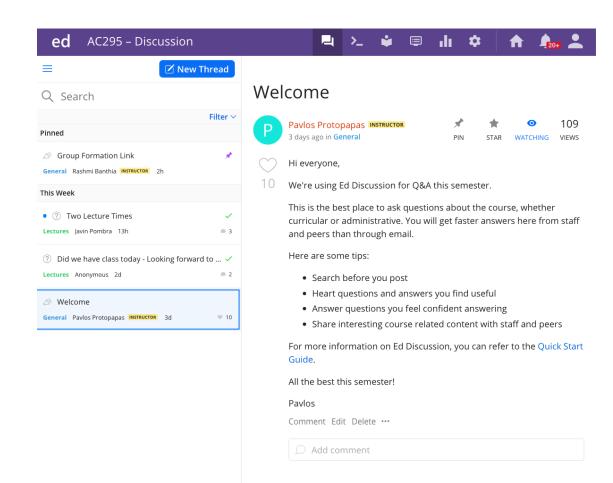
<sup>\*\*</sup> Each group should pick one slot (white background) in each module. We will release presentation slots on Sunday 8PM

# Course Components

Web Page: Syllabus, lecture slides and notebooks



Edstem: Forum and surveys



Github Repo: <a href="https://github.com/Harvard-IACS/2020F-AC295.git">https://github.com/Harvard-IACS/2020F-AC295.git</a>

# Grades

Assignment	Final Grade Weight
Discussion Forum	10%
Exercises	10%
Presentations	15%
Practicums	40%
Final Projects	25%
Total	100%



# **Final Details**

We will be using ED for discussions, announcements and surveys.

Individual,

Submit at Canvas

- Presentations: Group
- Practicums: Group
- Projects: Group

Submissions for presentations, practicums and projects we will be using github (details soon).



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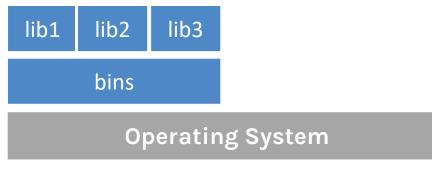
6: Virtual machines.



- Virtual environments help to make development and use of code more streamlined.
- Virtual environments keep dependencies in separate "sandboxes" so you can switch between both applications easily and get them running.
- Given an operating system and hardware, we can get the exact code environment set up using different technologies. This is key to understand the trade off among the different technologies presented in this class.



• Maggie took cs109a, she used to run her Jupyter notebooks from anaconda prompt. Every time she installed a module it was placed in the either of bin, lib, share, include folders and she could import it in and used it without any issue.

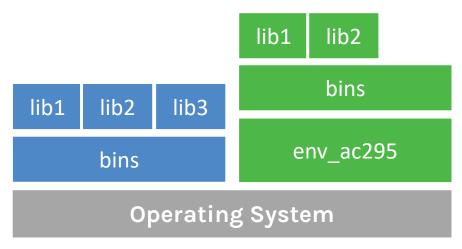


\$ which python
/c/Users/maggie/Anaconda3/python

Maggie



Maggie starts taking ac295, and she thinks that it would be good to isolate the
new environment from the previous environments avoiding any conflict with the
installed packages. She adds a layer of abstraction called virtual environment
that helps her keep the modules organized and avoid misbehaviors while
developing a new project.

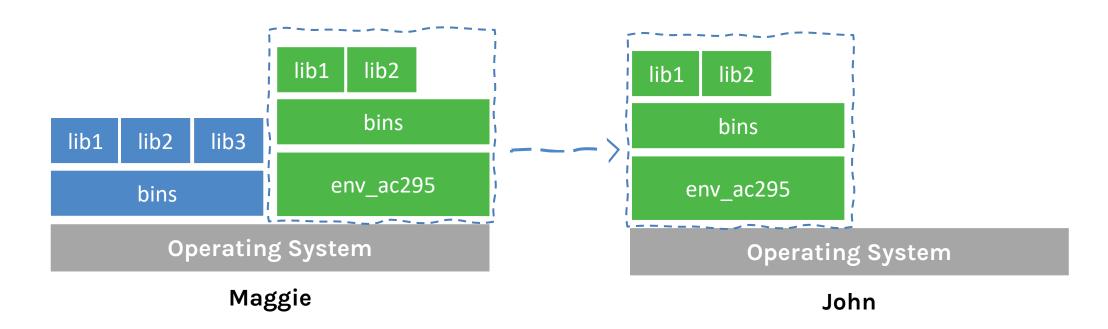


\$ which python
/c/Users/maggie/Anaconda3/envs/env\_ac295/python

Maggie

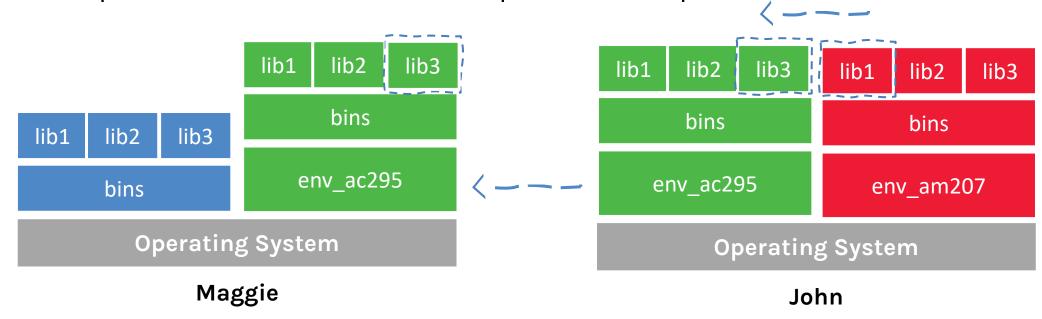


 Maggie collaborates with John for the final project and shares the environment she is working on through .yml file.



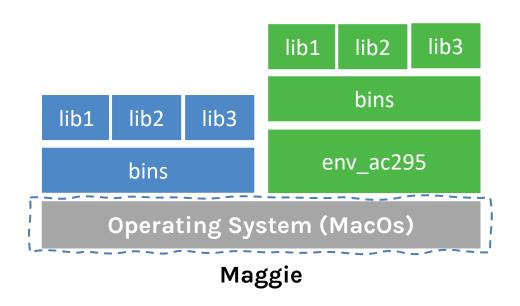


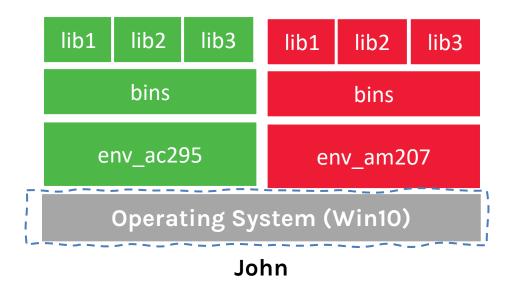
• John experiments a new method he learned in another class and adds a new library to the working environment. After seeing tremendous improvements, he sends Maggie back his code and a new .yml file. She can now update her environment and replicate the experiment.





 What could go wrong? Unfortunately, Maggie and John reproduce different results, and they think the issue relates to their operating systems. Indeed while Maggie has a MacOS, John uses a Win10.







# Virtual environments

#### **Pros**

- Reproducible research
- Explicit dependencies
- Improved engineering collaboration
  - Broader skill set

#### Cons

- Difficulty setting up your environment
  - Not isolation
- Does not work across different OS



# What are virtual environments then?

A virtual environment is a directory with the following components:

- site\_packages/ directory where third-party libraries are installed
- links [really symlinks] to the executables on your system
- some scripts that ensure that the code uses the interpreter and site packages in the virtual environment

> Adapted from CS207 <



### Virtual environments: virtualenv vs conda

#### virtualenv

- virtual environments manager embedded in Python
- incorporated into broader tools such as pipenv
- allow to install modules using pip package manager

#### how to use virtualenv

- create an environment within your project folder virtualenv your\_env\_name
- it will add a folder called environment\_name in your project directory
- activate environment: source env/bin/activate
- install requirements using: pip install package\_name=version
- deactivate environment once done: deactivate



# Virtual environments in practice

#### conda environment

- virtual environments manager embedded in Anaconda
- allow to use both conda and pip to manage and install packages

#### how to use conda

create an environment

```
conda create --name your env name python=3.7
```

it will add a folder located within your anaconda installation

```
/Users/your username /anaconda3/envs/your env name
```

- activate environment conda activate your\_env\_name (should appear in your shell)
- install requirements using conda install package\_name=version
- deactivate environment once done conda deactivate
- duplicate your environment using YAML file conda env export > my environment.yml



# how to use conda

find which environment you are using

conda env list

create an environment

conda create --name your env name python=3.7

it will add a folder located within your anaconda installation

/Users/your username/[opt]/anaconda3/envs/your env name

activate environment

conda activate your env name (should appear in your shell)

install requirements using

conda install package name=version

deactivate environment once done

conda deactivate

- duplicate your environment using YAML file conda env export > my\_environment.yml
- to recreate the environment now use conda env create -f environment.yml



# More on Virtual environments

#### Further readings

- For detailed discussions on similarities and differences among virtualenv and conda <a href="https://jakevdp.github.io/blog/2016/08/25/conda-myths-and-misconceptions/">https://jakevdp.github.io/blog/2016/08/25/conda-myths-and-misconceptions/</a>
- More on venv and conda environments
   https://towardsdatascience.com/virtual-environments-104c62d48c54
   https://towardsdatascience.com/getting-started-with-python-environments-using-conda-32e9f2779307



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# Why should we use virtual machines?

#### Motivation

- We have our isolated systems, and after we set up the environment with our colleagues' machine, we expect to get identical results, right? Unfortunately, it is not always the case. Why? Most likely because we run on a different operating system.
- Even though using virtual environments, we isolate our computations, we might need to use the same operating system that requires running "like if" we are in different machines.
- How can we run the same experiment? Virtual Machines!
- Isolation!



# Why should we use virtual machines? (cont)

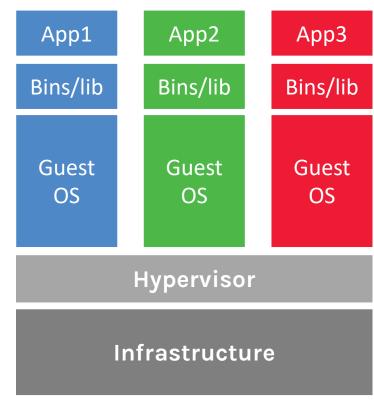
#### **Advantages**

- Full autonomy: it works like a separate computer system; it is like running a computer within a computer.
- Very secure: the software inside the virtual machine cannot affect the actual computer.
- Lower costs: buy one machine and run multiple operating systems.



# What are virtual machines?

- virtual machines have their own virtual hardware: CPUs, memory, hard drives, etc.
- you need a hypervisor that manages different virtual machines on server
- hypervisor can run as many virtual machines as you wish
- operating system is called the "host" while those running in a virtual machine are called "guest"
- You can install a completely different operating system on this virtual machine



**Machine Virtualization** 



# Limitations

- Uses hardware in your local machine
- There is an overhead associated with virtual machines
  - 1. Guest is not as fast as the host system
  - 2. Takes a long time to start up
  - 3. It may not have the same graphics capabilities

This is the second time we are offering the course, so your feedback will improve it for future years.

However, we are making every effort to have a well-organized course and we promise you an exciting semester full of learning!

#### THANK YOU

AC295 Advanced Practical Data Science
Pavlos Protopapas