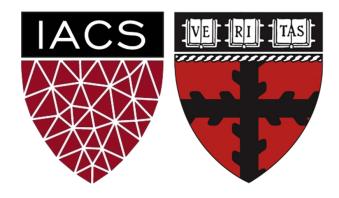
# Lecture 1: Introduction

AC295

# 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: Expectations

5: Workload

6: Logistics

7: Grades



# 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

# Why you shouldn't take this class

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

#### For example:

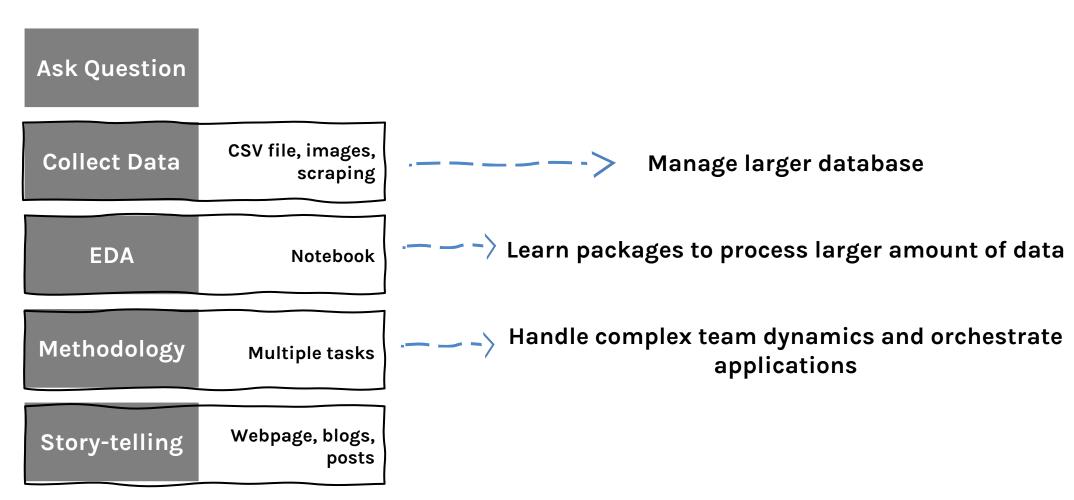
- Basic Machine Learning
- CNNs, RNNs, Autoencoders, GANs, etc
- Basic linux commands

Remember, this course will be offered again in the fall!

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

#### Data Science Series 109A/B

Real World

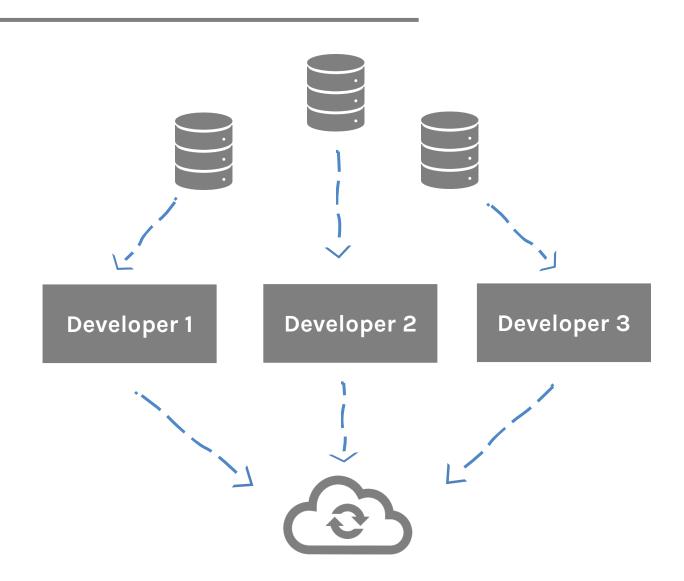


# Data Science Series to Real World (cont)

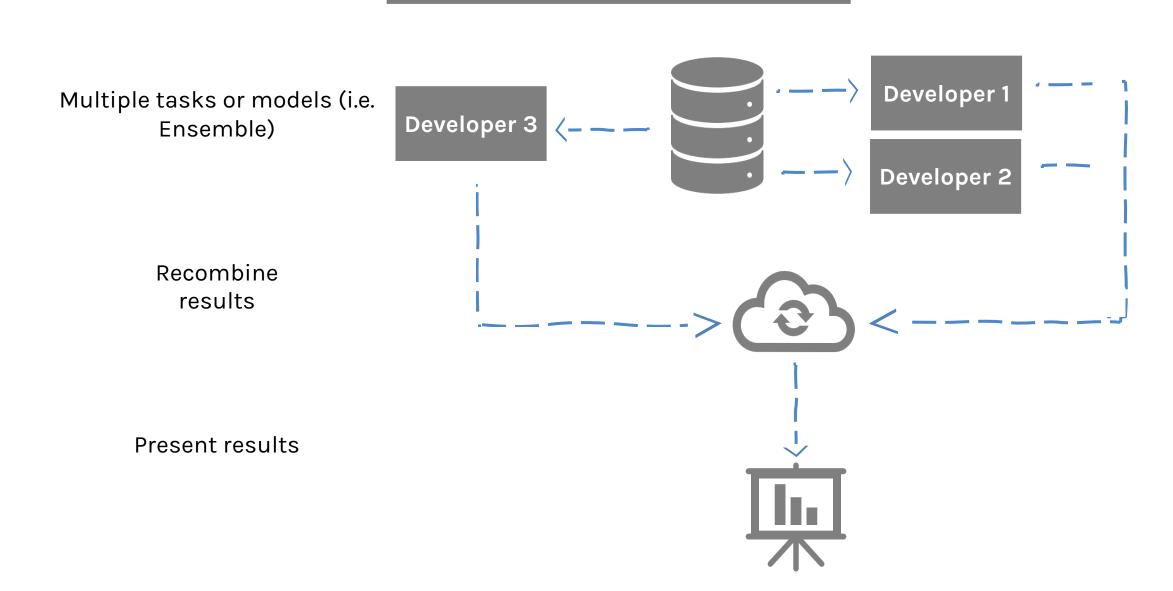
Fragmented database

Multitude requirements and applications

Recombine and deploy



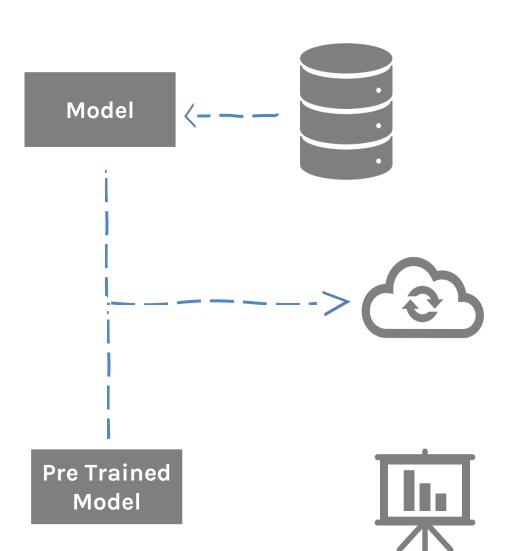
# Data Science Series to Real World (cont)



### Data Science Series to Real World (cont)

Model too expensive to train Or not enough training data

Use pre-trained model



Final Results

Present results

### 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 (see you there), cross country ski (completed Engadin skimarathon), cheese making and being a TikToker (check me out @pavlosprotopapas)



### Who? (cont)

#### Michael S. Emanuel

After 17 years in finance, mainly fixed income portfolio management, Michael started a second career and is completing the Masters of Data Science program at Harvard. He is a father of two small children who occasionally crash IACS events and enjoys distance running and classical music.





### Who? (cont)

### **Andrea Porelli**

Urban planner turned into data hacker. He likes to break things just for the sake of putting them back together (most of the time). Committed to apply Data Science to change something. So far, he managed to change himself the most –thanks IACS– and look forward to pass it over.





### Who? (cont)

### Giulia Zerbini

Agency in Milan, MA Graduate at Politecnico di Milano. Designing and developing visualizations and interfaces based on data. Passionate about using visualizations for discovering patterns in data and communicating information in intuitive terms to a broad audience.





### **Course Structure and Activities**

#### Modules:

- Deploy data science (integration + scalability)
- 2. Transfer learning and distillation
- 3. Visualization as investigative tool

#### **Activities:**

lectures, reading discussions, exercises, quizzes, practicums, projects

Lectures: Tuesday and Thursday 4:30-5:45 pm in Cruft 309

Office Hours: TBD



# **Topics**

Deploy data science (integration + scalability)

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



# Topics (cont)

### Transfer learning and distillation

- A. Basic Transfer Learning and SOTA Models
- B. Transfer Learning across Tasks
- C. Distillation and Compression



# Topics (cont)

### Visualization as investigative tool

- A. Introduction and Overview of Viz for Deep Models
- B. Convolutional Neural Networks for Image Data
- C. Recurrent Neural Networks for Text Data



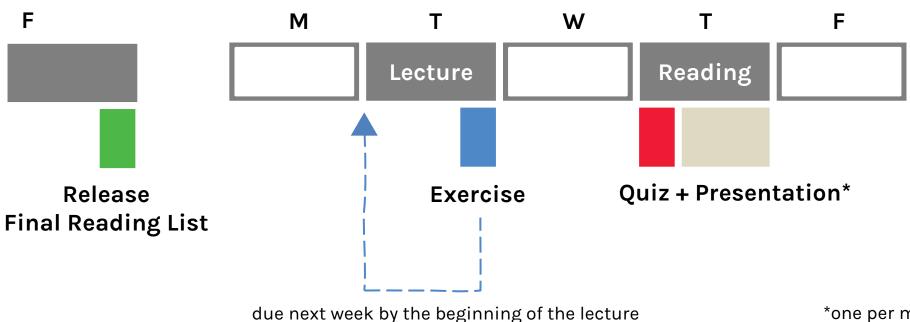
# Calendar

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

Week	Date	Lecture #	Topics
1	1/28	1	Introduction
	1/30	2	Set up
2	2/4	3	Virtual Environments, Virtual Boxes, and Containers
	2/6	4	Journal Discussion
3	2/11	5	Kubernetes
	2/13	6	Journal Discussion
4	2/18	7	Dask
	2/20	8	Journal Discussion
5	2/25	9	Practicum
	2/27	10	Practicum
6	3/3	11	Intro to Transfer Learning: Basic Transfer Learning and SOTA Models
	3/5	12	Journal Discussion
7	3/10	13	Transfer Learning across Tasks
	3/12	14	Journal Discussion
	3/17		Spring Recess
	3/19		
8	3/24	15	Distillation and Compression
	3/26	16	Journal Discussion
9	3/31	17	Practicum
9	4/2	18	Practicum
10	4/7	19	Introduction and Overview of Viz for Deep Models
10	4/9	20	Journal Discussion
11	4/14	21	Convolutional Neural Networks for Image Data
	4/16	22	Journal Discussion
12	4/21	23	Recurrent Neural Networks for Text Data
	4/23	24	Journal Discussion
13	4/28	25	Practicum
	4/30	26	Practicum
14	5/5	27	Project
14	5/7	28	Project
15	5/12	29	Project
10	5/14	30	Project

# **Course Structure and Activities**

### Regular week schedule



\*one per module per group



### Workload

### Regular Week

3 hours in class

3 hours reading

2 hours exercise

4 hours presentation\*

~ 12 hours/week

\* 1 presentation per module per group (3 total)

### Practicum and Project Week

~ 15 hours/week\*\*

\*\* 3 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

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

# Grades

Assignment	Final Grade Weight
Quizzes	9%
Exercises	9%
Presentations	15%
Practicums	45%
Projects	20%
Participation	2%
Total	100%



### Final Details

- We will be using ED for discussions, announcements and quizzes.
- Submissions for exercises, reports, presentations etc we will be using github (details soon).



This is the first time we are offering the course, so your feedback will be vital in tuning it this year and improving 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