

# Lecture #1: Introduction to Data Science

Asso. Prof. Huynh Trung Hieu

Sources: Introduction to Data Science

By Pavlos Protopapas, Kevin Rader and Chris Tanner

# Lecture Outline

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- Why data science? Why taking this course?
- What is data science?
- What is this class and what it is not?
- The data science process
- Example

# Why?

## Jobs!

### 50 Best Jobs in America





This report ranks jobs according to each job's Glassdoor Job Score, determined by combining three factors: number of job openings, salary, and overall job satisfaction rating.

Employers: Want to recruit better in 2017? [Find out how.](#)


United States

2017

12k Shares



#### 1 Data Scientist



4.8 / 5  
Job Score


4.4 / 5  
Job Satisfaction


\$110,000  
Median Base Salary

4,184  
Job Openings

View Jobs

#### 2 DevOps Engineer




 Awards

Best Places to Work

Highest Rated CEOs

Best Places to Interview


 Lists

Best Jobs

Best Cities for Jobs

Highest Paying Jobs

Oddball Interview Questions

 Trends

Overview

# Why?

# Jobs!

glassdoor

Jobs

Company Reviews

Salaries

Interviews

Salary Calculator

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Write Review

For Employers

Post Jobs Fr

Q Job Title, Keywords, or Company

Jobs

Location

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## 50 Best Jobs in America for 2019

Best Jobs

2019

United States

Share



	Job Title	Median Base Salary	Job Satisfaction	Job Openings	
#1	Data Scientist	\$108,000	4.3/5	6,510	<a href="#">View Jobs</a>
#2	Nursing Manager	\$83,000	4/5	13,931	<a href="#">View Jobs</a>
#3	Marketing Manager	\$82,000	4.2/5	7,395	<a href="#">View Jobs</a>
#4	Occupational Therapist	\$74,000	4/5	17,701	<a href="#">View Jobs</a>
#5	Product Manager	\$115,000	3.8/5	11,884	<a href="#">View Jobs</a>

# Why?

## Jobs!

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**Best Jobs**

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Overview

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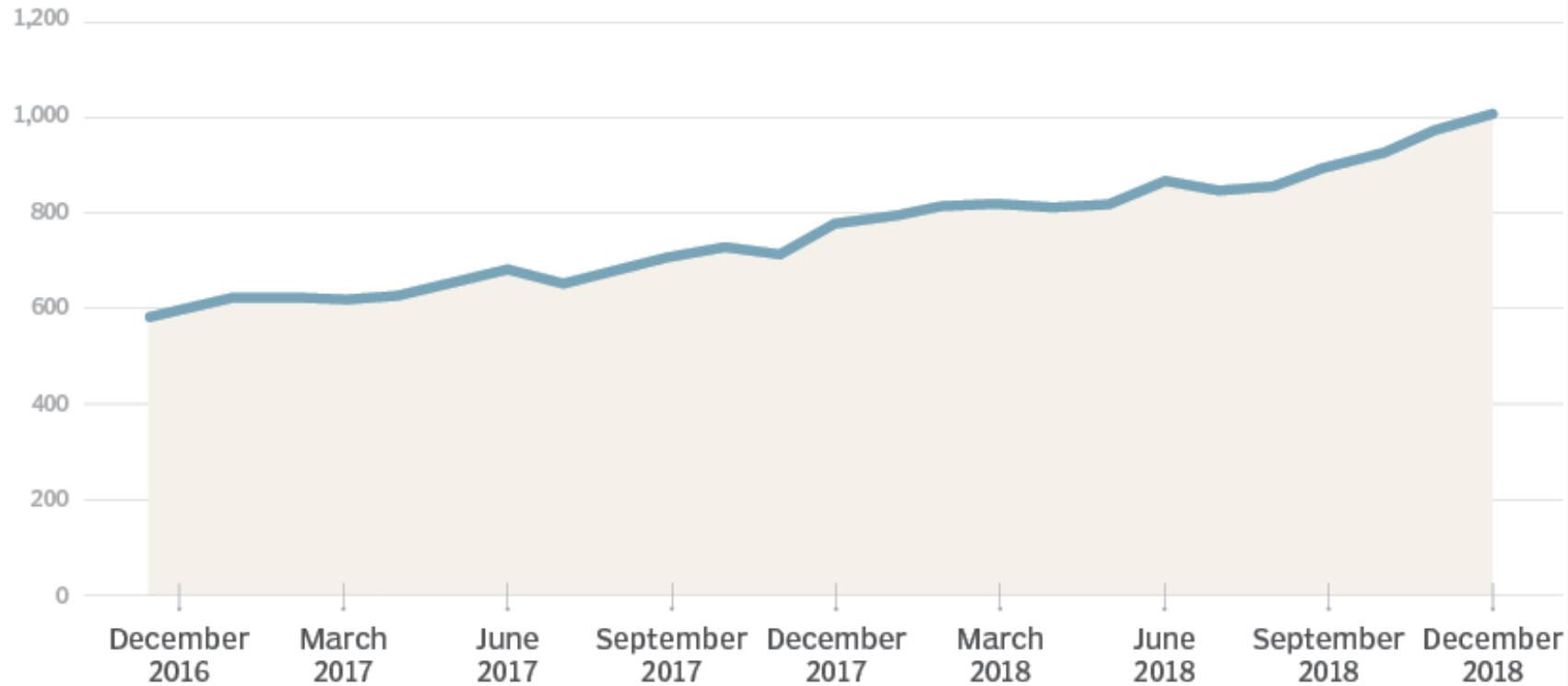
#### 2 DevOps Engineer



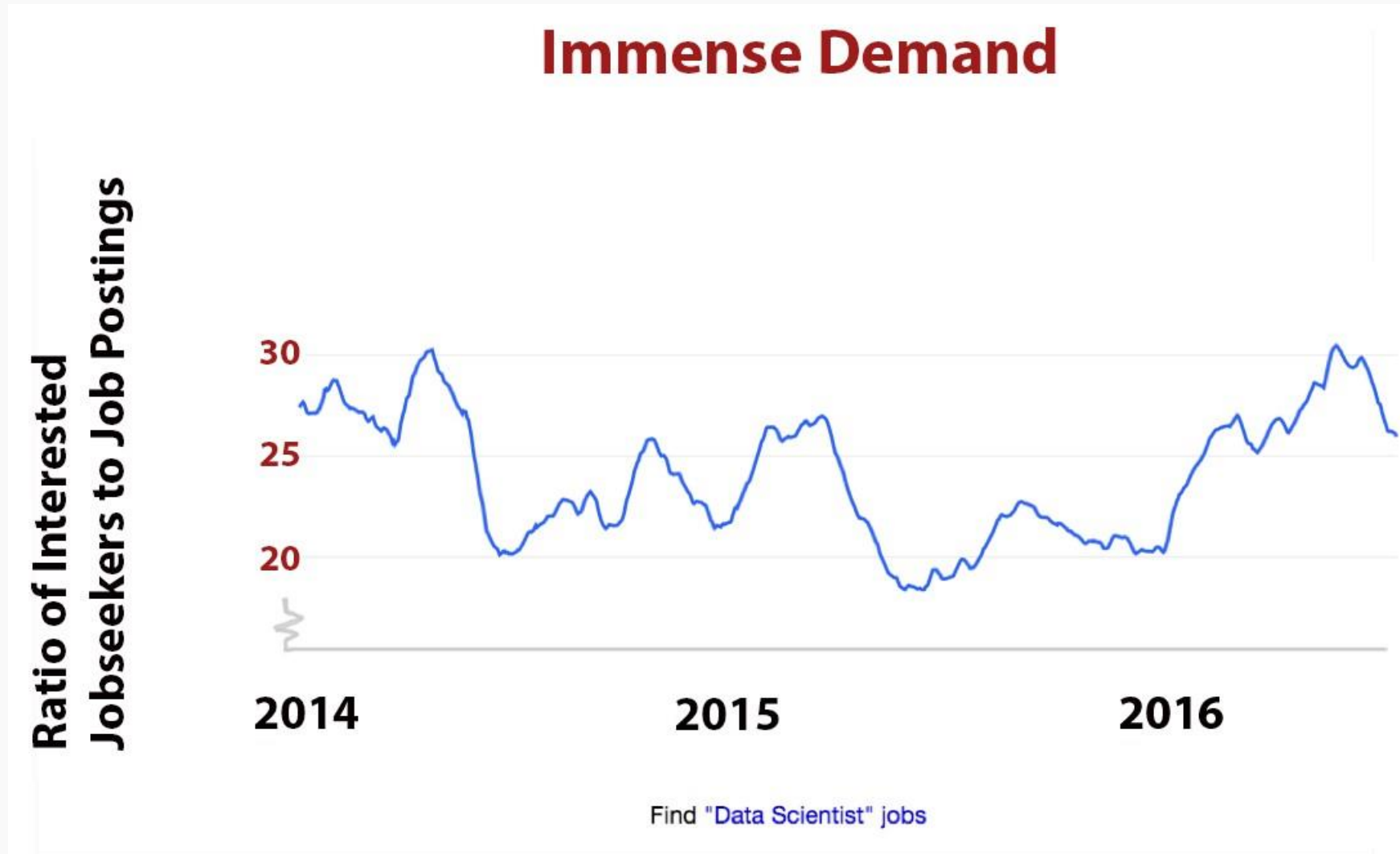
# Why?

## Data scientists are in high demand

Data scientist job postings, per 1 million postings on Indeed



# Why?




# Why?

**itviec** All Jobs IT Companies Blog IT Story data scientist All Cities For Employers Sign In EN

Job Level Salary Range Company Type Clear all filters


**7 data scientist jobs in Vietnam**

**Data Scientist (Python, Database) Up to \$2500**Hot

Sign in to view

- Competitive Salary Up to \$2500
- 13th-15th month salary + bonus, ESOP, Remote
- Leading experts in IT, Finance and Blockchain


Python Data Analyst Database 3d

**Data Scientist**Hot

Sign in to view


Database Python Data Analyst

Fresher Accepted 3d

**[Remote] Senior Python Data Scientist**

Sign in to view

Python Data Analyst Database 23d

**(Mid/Senior/Lead) Data Engineer**Hot

Sign in to view

### Data Scientist (Python, Database) Up to \$2500

DAT Technologies

[Apply Now](#)

Python Database Data Analyst

Sign in to view

Tầng 7, 117 Hoàng Cầu, Dong Da, Ha Noi [See map](#)

At office

3 days ago

#### Top 3 Reasons To Join Us

- Competitive Salary Up to \$2500
- 13th-15th month salary + bonus, ESOP, Remote
- Leading experts in IT, Finance and Blockchain

#### Job Description

DATX Vietnam (DAT Technologies) is leading in utilizing ML and AI in investment not just in Vietnam, but also in other countries such as Australia, and we are moving fast into other big markets like US and EU. We apply ML and AI to help our customers not just saving their time for investing in share market, but also warrant a stable income from their investment.

At DATX Vietnam, our primary goal is to bring share investment to everyone, so that as a community, we all share the wealth and growth of our fast-growing market in Vietnam.

[https://itviec.com/it-jobs/data-scientist?job\\_selected=data-scientist-python-database-up-to-2500-dat-technologies-0241](https://itviec.com/it-jobs/data-scientist?job_selected=data-scientist-python-database-up-to-2500-dat-technologies-0241)



# Why?

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Why do I love data science?

Why are you here?

what my friends think I do



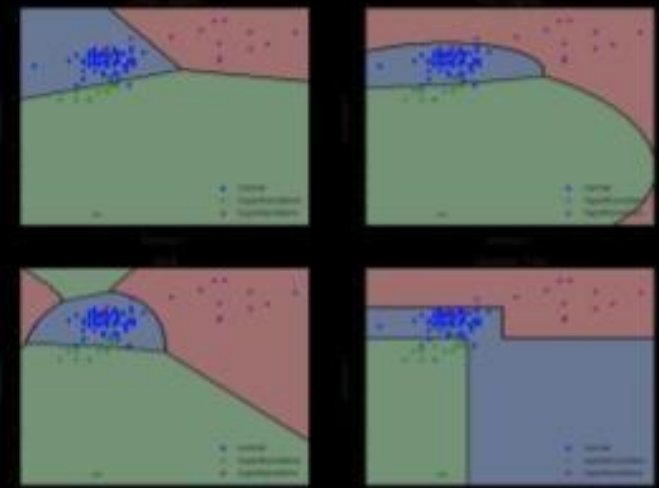
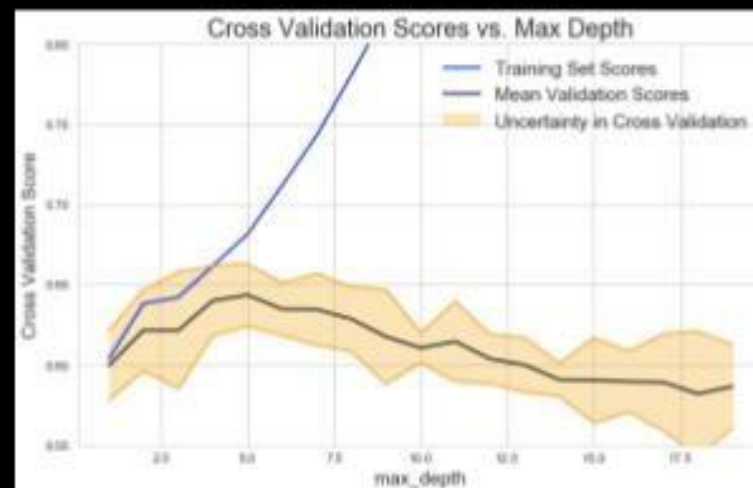
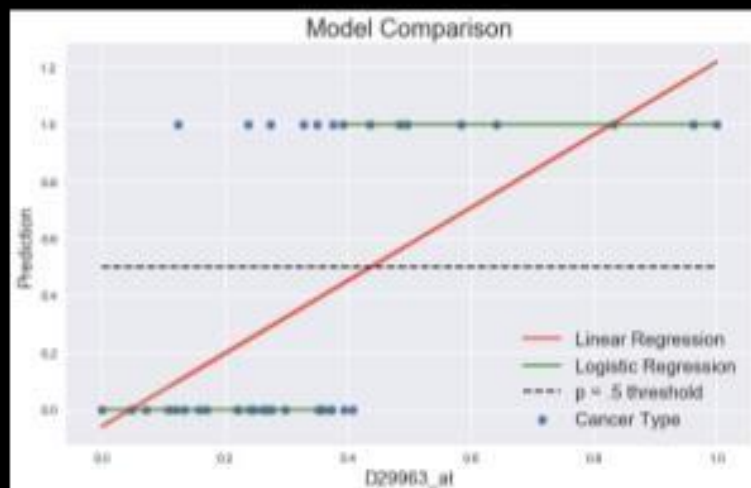
what my family thinks I do



what society thinks I do



what I actually (will) do in Data Science 1



# Why?

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Why are you here?

# A little bit of history

# History

Long time ago (thousands of years) science was only empirical and people counted stars





# History (cont)

Long time ago (thousands of years) science was only empirical and people counted stars or crops



# History (cont)

Long time ago (thousands of years) science was only empirical and people counted stars or crops and used the data to create machines to describe the phenomena



# History (cont)

Few hundred years: theoretical approaches, try to derive equations to describe general phenomena.

1.  $\nabla \cdot \mathbf{D} = \rho_V$
2.  $\nabla \cdot \mathbf{B} = 0$
3.  $\nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t}$
4.  $\nabla \times \mathbf{H} = \frac{\partial \mathbf{D}}{\partial t} + \mathbf{J}$

$$T^2 = \frac{4\pi^2}{GM} a^3$$

can be expressed  
as simply

$$T^2 = a^3$$

If expressed in the following units:

$T$  Earth years

$a$  Astronomical units AU  
( $a = 1$  AU for Earth)

$M$  Solar masses  $M_\odot$

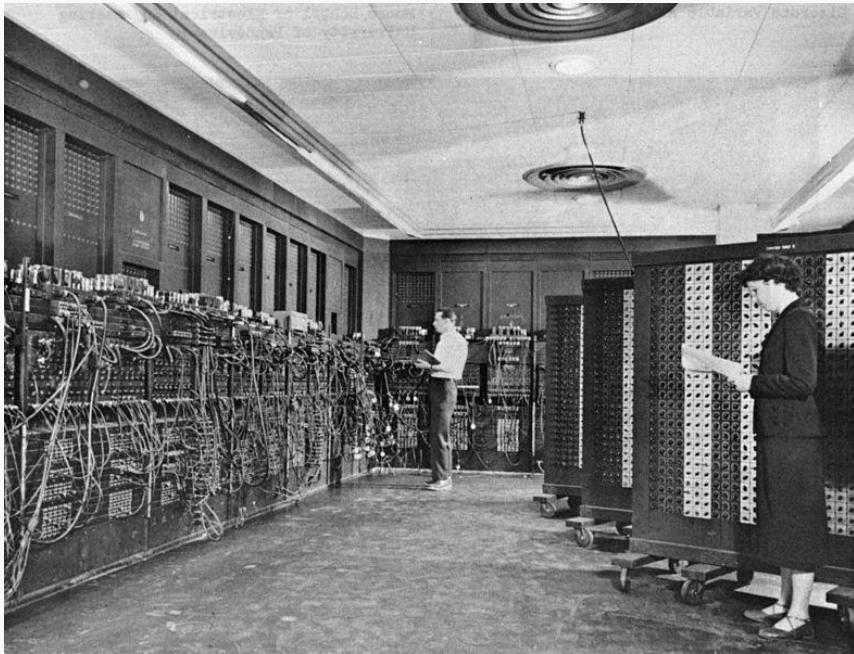
Then  $\frac{4\pi^2}{G} = 1$

$$H(t)|\psi(t)\rangle = i\hbar \frac{\partial}{\partial t} |\psi(t)\rangle$$



# History (cont)

About a hundred years ago: computational approaches



# History (cont)

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And then ... data science

# What is data science?

# What?

## The Data Science Process

Ask an interesting question

Get the Data

Explore the Data

Model the Data

Communicate/Visualize the Results

# What?

## The Data Science Process

Ask an interesting question

Get the Data

Explore the Data

Model the Data

Communicate/Visualize the Results

What is the scientific goal?

What would you do if you had **all** of the data?

What do you want to predict or estimate?

# What?

## The Data Science Process

Ask an interesting question

Get the Data

Explore the Data

Model the Data

Communicate/Visualize the Results

How were the data sampled?

Which data are relevant?

Are there privacy issues?

# What?

## The Data Science Process

Ask an interesting question

Get the Data

Explore the Data

Model the Data

Communicate/Visualize the Results

Plot the data.

Are there anomalies or egregious issues?

Are there patterns?

# What?

## The Data Science Process

Ask an interesting question

Get the Data

Explore the Data

Model the Data

Communicate/Visualize the Results

Build a model.

Fit the model.

Validate the model.



# What?

## The Data Science Process

Ask an interesting question

Get the Data

Explore the Data

Model the Data

Communicate/Visualize the Results

What did we learn?

Do the results make sense?

Can we effectively tell a story?

# What?

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The material of the course will integrate the five key facets of an investigation using data:

1. data collection; data wrangling, cleaning, and sampling to get a suitable data set
2. data management; accessing data quickly and reliably
3. exploratory data analysis; generating hypotheses and building intuition
4. prediction or statistical learning
5. communication; summarizing results through visualization, stories, and interpretable summaries.

# What?

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## Week 1:

Getting ready with python, jupyter notebooks, environments and numpy.

# What?

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## Week 2:

Basic statistics, visualization, pandas and data scraping

# What?

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## Week 3 and 4:

Regression, and sklearn using transportation data:

- knn regression
- Linear and Polynomial Regression
- Multiple Regression
- Model Selection
- Regularization

# What?

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## Week 5:

Exploratory Data Analysis, matplotlib and seaborn:

- Basic concepts of EDA
- Basic concepts of Visualization and Communications

# What?

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## Week 6-7:

Classification, data imputations on Health Data:

- Logistic Regression (linear and polynomial)
- Multiple Logistic Regression
- Missing data and knn classification

# What?

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**Week 8-10:**

EthiCS

PCA and high dimensionality

Decisions trees and ensemble methods :

- Simple Decision Trees for classification and Regression
- Bagging
- Random Forest
- Boosting
- Stacking



# Other topics relating to this course

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## A. Neural Networks:

- CNNs
- RNNs
- Generative models

## B. Unsupervised Clustering

## C. Piecewise Linear Regression

## D. Bayesian Modeling

# Advanced Practical Data Science

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- A. Productions Data Science, from notebooks to the cloud
- B. Big models, transfer learning and architecture learning
- C. Visualization tools for interpreting models
- D. Sequential data, seq2seq with attention, transformers, NLP and time series modeling

# Who?

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## Lecturers:

- Asso. Prof. Huynh Trung Hieu
- Dr. Nguyen Chi Kien

# Related Sections

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## Topics

1. Linear Algebra and Hypothesis Testing: The Short Versions
2. Methods of regularization and their justifications
3. Generalized Linear Models
4. Mathematics of PCA
5. Decision trees and Ensemble method;
6. Stochastic Gradient Descent

# Homework(s)

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There will be 8 homework (not including Homework 0):

- Homework 0
- Homework 1: Web scraping, Beautiful Soup
- Homework 2: Regression kNN and LinReg
- Homework 3: Multi-regression, polynomial reg and model selection
- **Homework 4\*: Log Reg and more**
- Homework 5: PCA and ethics
- Homework 6: Random Forest, Boosting and Neural Networks
- **Homework 7\*: Neural Networks**
- Homework 8: Experimental Design

# Final Project

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There will be a final group project (2-4 students) due during exams period.

- We will provide some pre-defined projects which you could use for your final project.
- In some very special cases you can use your own (public) data set and your own project definition (to be approved by the instructors)

# CS109a: Introduction to Data Science

Fall 2019

[Pavlos Protopapas](#), [Kevin A. Rader](#), and [Chris Tanner](#)

**Lab Leaders:** Chris Tanner and Eleni Kaxiras

Welcome to CS109a/STAT121a/AC209a, also offered by the DCE as CSCI E-109a, Introduction to Data Science. This course is the first half of a one-year course to data science. We will focus on the analysis of data to perform predictions using statistical and machine learning methods. Topics include data scraping, data management, data visualization, regression and classification methods, and deep neural networks (a detailed schedule will be made available soon). You will get ample practice through weekly homework assignments. The class material integrates the five key facets of an investigation using data:

1. data collection - data wrangling, cleaning, and sampling to get a suitable data set
2. data management - accessing data quickly and reliably
3. exploratory data analysis – generating hypotheses and building intuition
4. prediction or statistical learning
5. communication – summarizing results through visualization, stories, and interpretable summaries

Only one of CS 109a, AC 209a, or Stat 121a can be taken for credit. Students who have previously taken CS 109, AC 209, or Stat 121 cannot take CS 109a, AC 209a, or Stat 121a for credit.

**Announcement:** [HW0](#) is now available.

**Lectures:** Mon and Wed 1:30-2:45 pm in Harvard Northwest Building, NW B-103

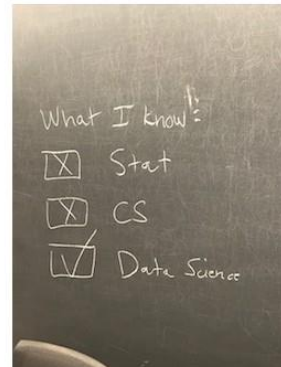
**Labs:** Thur 4:30-5:45 pm in Pierce 301

**Head TFs:** Chris Gumb - **DCE Head TF:** Sol Girouard

**Office Hours:** IACS student lobby in Maxwell-Dworkin's ground. Just follow the signs.

**Online Office Hours zoom link:** <https://harvard-dce.zoom.us/j/7607382317>

Course material can be viewed in the public [GitHub repository](#).



## STANDARD SECTIONS

**Friday 9/13** 10:30–11:45 am 1 Story St. Room 306

**Monday 9/16** 4:30–5:45 pm Science Center 110

Cover the material presented in class. Both standard sections are identical.

## ADVANCED SECTIONS

**Wednesday 9/18** 4:30–5:45 pm 1 Story St. Room 306

Cover a different topic each week and are required for 209a students.

## Instructor Office Hours

**Pavlos & Kevin:** Monday 3–5 pm, IACS Lobby

**Chris:** Wednesday 3–4 pm, Maxwell-Dworkin B125

CS109A, PROTOPAPAS, RADER, TANNER

# The Data Science Process



# The Data Science Process

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The Data Science Process is similar to the scientific process - one of observation, model building, analysis and conclusion:

- Ask questions
- Data Collection
- Data Exploration
- Data Modeling
- Data Analysis
- Visualization and Presentation of Results

**Note:** This process is by no means linear!

# Analyzing Hubway Data

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**Introduction:** Hubway is metro-Boston's public bike share program, with more than 1600 bikes at 160+ stations across the Greater Boston area. Hubway is owned by four municipalities in the area.

By 2016, Hubway operated 185 stations and 1750 bicycles, with 5 million ride since launching in 2011.

**The Data:** In April 2017, Hubway held a Data Visualization Challenge at the Microsoft NERD Center in Cambridge, releasing 5 years of trip data.

**The Question:** What does the data tell us about the ride share program?

# The Data Exploration/Question Refinement Cycle

Our original question: ‘What does the data tell us about the ride share program?’ is a reasonable slogan to promote a hackathon. It is not good for guiding scientific investigation.

Before we can refine the question, we have to look at the data!

	seq_id	hubway_id	status	duration	start_date	strt_statn	end_date	end_statn	bike_nr	subsc_type	zip_code	birth_date	gender
0	1	8	Closed	9	7/28/2011 10:12:00	23.0	7/28/2011 10:12:00	23.0	B00468	Registered	'97217	1976.0	Male
1	2	9	Closed	220	7/28/2011 10:21:00	23.0	7/28/2011 10:25:00	23.0	B00554	Registered	'02215	1966.0	Male
2	3	10	Closed	56	7/28/2011 10:33:00	23.0	7/28/2011 10:34:00	23.0	B00456	Registered	'02108	1943.0	Male
3	4	11	Closed	64	7/28/2011 10:35:00	23.0	7/28/2011 10:36:00	23.0	B00554	Registered	'02116	1981.0	Female
4	5	12	Closed	12	7/28/2011 10:37:00	23.0	7/28/2011 10:37:00	23.0	B00554	Registered	'97214	1983.0	Female

Based on the data, what kind of questions can we ask?

# The Data Exploration/Question Refinement Cycle

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Who? Who's using the bikes?

Refine into specific hypotheses:

- More men or more women?
- Older or younger people?
- Subscribers or one time users?

# The Data Exploration/Question Refinement Cycle

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**Where?** Where are bikes being checked out?

Refine into specific hypotheses:

- More in Boston than Cambridge?
- More in commercial or residential?
- More around tourist attractions?

**Sometimes the data is given to you in pieces and must be merged!**

# The Data Exploration/Question Refinement Cycle

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**When?** When are the bikes being checked out?

Refine into specific hypotheses:

- More during the weekend than on the weekdays?
- More during rush hour?
- More during the summer than the fall?

**Sometimes the feature you want to explore doesn't exist in the data, and must be engineered!**

# The Data Exploration/Question Refinement Cycle

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**Why?** For what reasons/activities are people checking out bikes?

Refine into specific hypotheses:

- More bikes are used for recreation than commute?
- More bikes are used for touristic purposes?
- Bikes are use to bypass traffic?

**Do we have the data to answer these questions with reasonable certainty?**

**What data do we need to collect in order to answer these questions?**

# The Data Exploration/Question Refinement Cycle

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**How?** Questions that combine variables.

- How does user demographics impact the duration the bikes are being used? Or where they are being checked out?
- How does weather or traffic conditions impact bike usage?
- How do the characteristics of the station location affect the number of bikes being checked out?

**How** questions are about modeling relationships between different variables.



# Inspirations for Data Viz/Exploration

So how well did we do in formulating creative hypotheses and manipulating the data for answers?

Check out the winners of the Hubway Challenge:

<http://hubwaydatachallenge.org>

