

# DATA SCIENCE

Brian Chung

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## WHO ARE WE?

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### **BRIAN CHUNG, INSTRUCTOR**

Brian is a researcher in the field of quantitative finance. He has worked at Citadel, LLC researching trading signals and building prediction models.

He graduated with a BS in Electrical Engineering from University of Illinois-Urbana Champaign and an MS from Stanford University. When not in front of a computer, he enjoys motorcycling, CrossFit, and cooking with various gadgets.

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## WHO ARE WE?

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### **SCOTT LITTLE, EXPERT IN RESIDENCE**

Scott Little is a data scientist who likes working with physical sensor data. Recently, he completed a project that predicts solar power from satellite imagery and ground photometer sensors. He has a PhD in Physics from the University of Toledo, where he specialized in thin-film photovoltaic solar cells. For fun he enjoys cycling, dreaming, electronics, quadcopters, neurohacking and making things at Pumping Station: One, the local hackerspace.

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# WHO ARE YOU?

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3 minutes:

- ▶ Turn to a person next to you and share your answers
- ▶ You will introduce them to the class 😊

Questions:

- ▶ What is your name?
- ▶ What industry do you work in or what field do you study?
- ▶ What are you most excited to learn in this class?
- ▶ What is a hobby or interest of yours?

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# AGENDA

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- Logistics
- Course Philosophy
- What is Data Science?
- Machine Learning taxonomy
- Project Discussion

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# LOGISTICS

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## EXERCISE #1: BOOKMARK THIS PAGE

**[HTTPS://GITHUB.COM/BRIANCHANDBOUND/GA-DS](https://github.com/Brianchandbound/GA-DS)**

The course website has all the information regarding logistics. If you have a course question not answered, please email [gadschicago@gmail.com](mailto:gadschicago@gmail.com)

### Website Topics:

Course logistics

Schedule

Project

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## **ADDITIONAL COURSE EXPECTATIONS**

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**Attendance / late policy**

**Computer / Phone Use**

**Participation before, during, and outside of class**

**Requesting help & Getting feedback**

**Treating other students with respect and helping others**

# COURSE PHILOSOPHY

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THIS IS NOT THE END

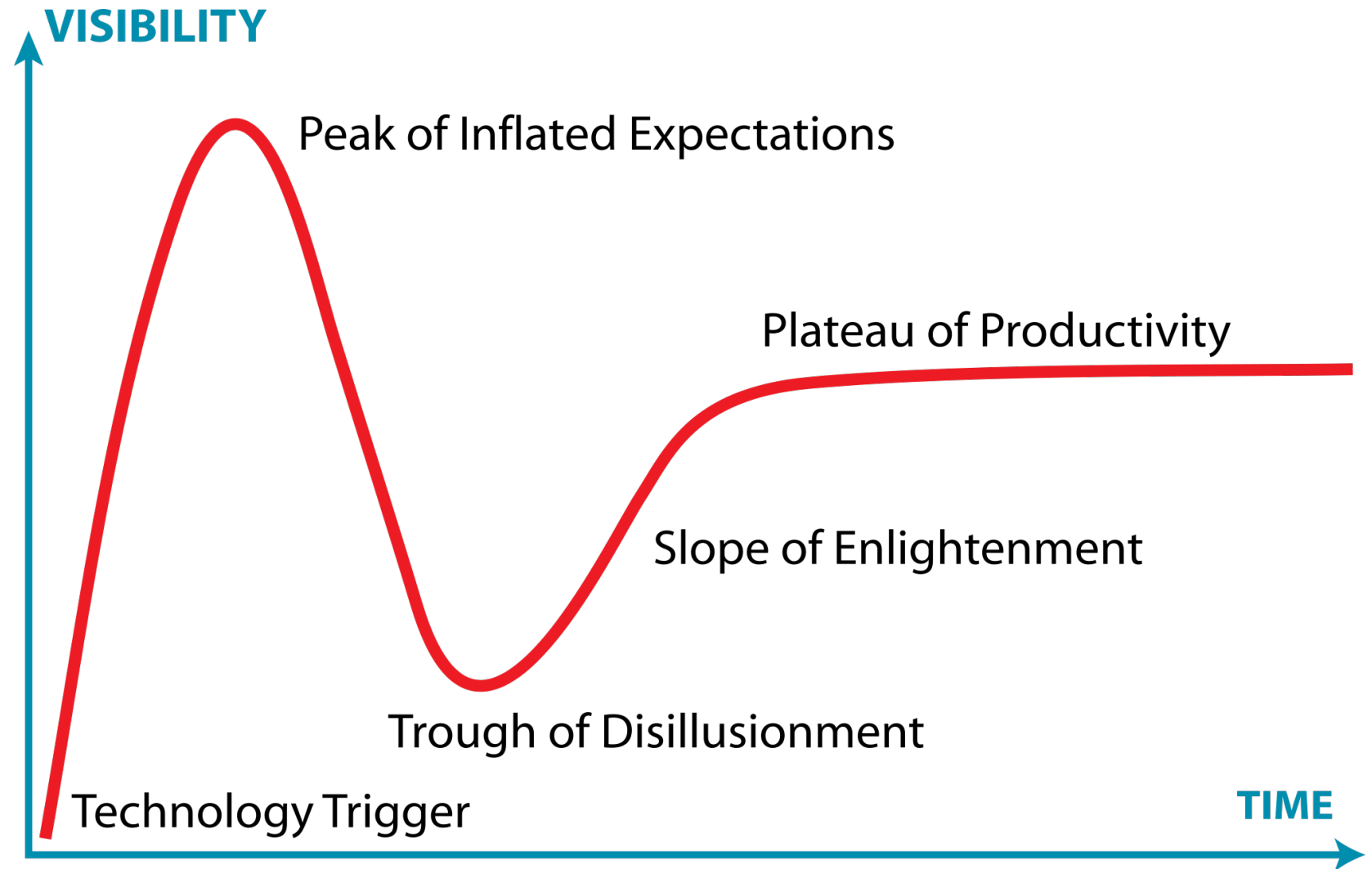




# COURSE PHILOSOPHY

**THIS IS NOT THE END**

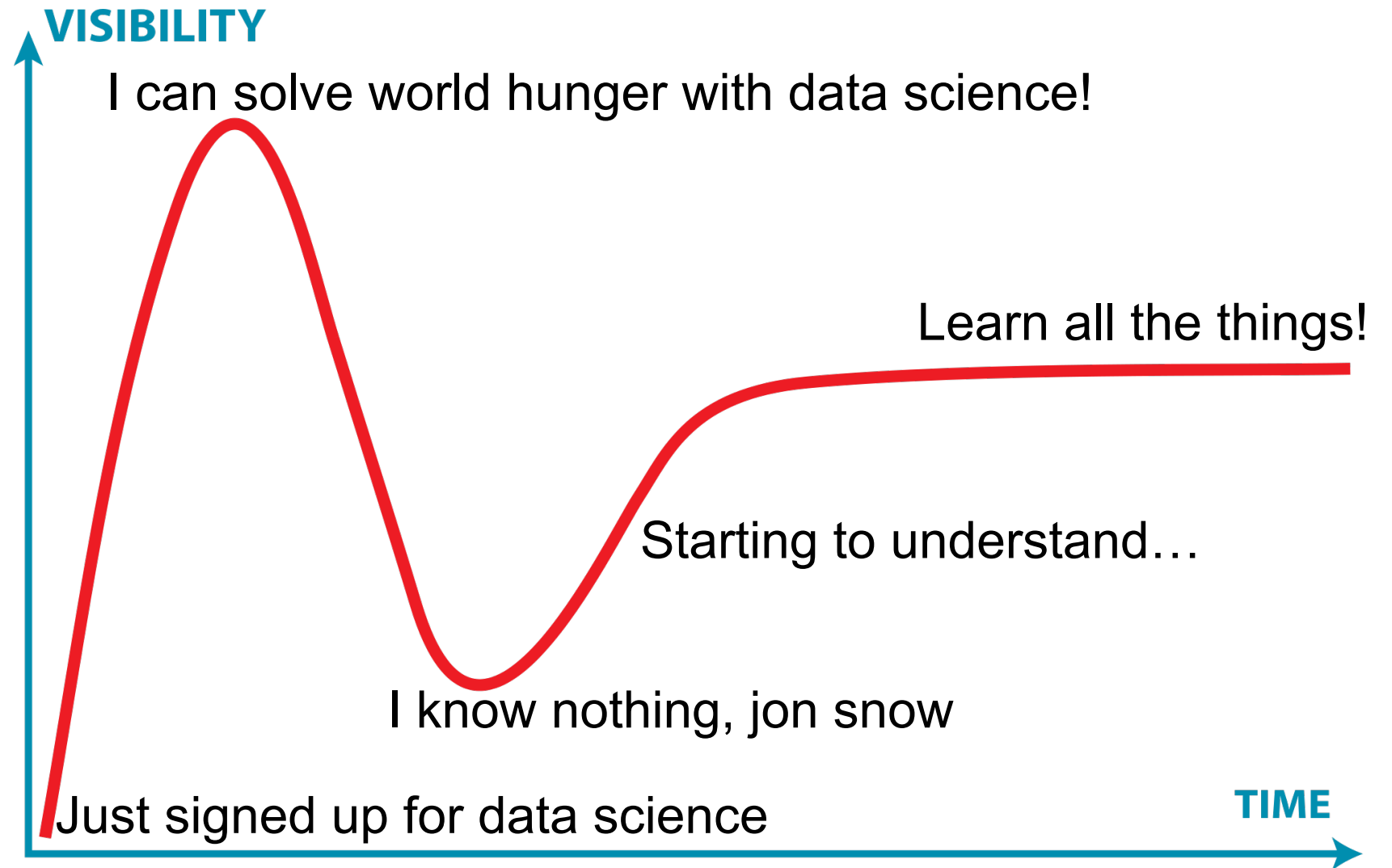
**DATA SCIENCE IS HARD**



# COURSE PHILOSOPHY

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DATA SCIENCE IS HARD



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# COURSE PHILOSOPHY

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**THIS IS NOT THE END**

**DATA SCIENCE IS HARD**

**SEEK AND YE SHALL FIND  
(HELP)**



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# COURSE PHILOSOPHY

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**THIS IS NOT THE END**

**DATA SCIENCE IS HARD**

**SEEK AND YE SHALL FIND  
(HELP)**

**LEARN BY DOING**



# WHAT IS DATA SCIENCE?

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A set of tools and techniques used to extract useful information from data

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An interdisciplinary, problem-solving oriented subject

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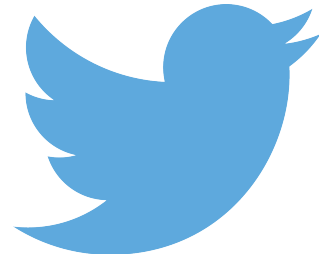
The application of statistical techniques to model practical problems



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## WHO USES DATA SCIENCE? TL;DR EVERYONE

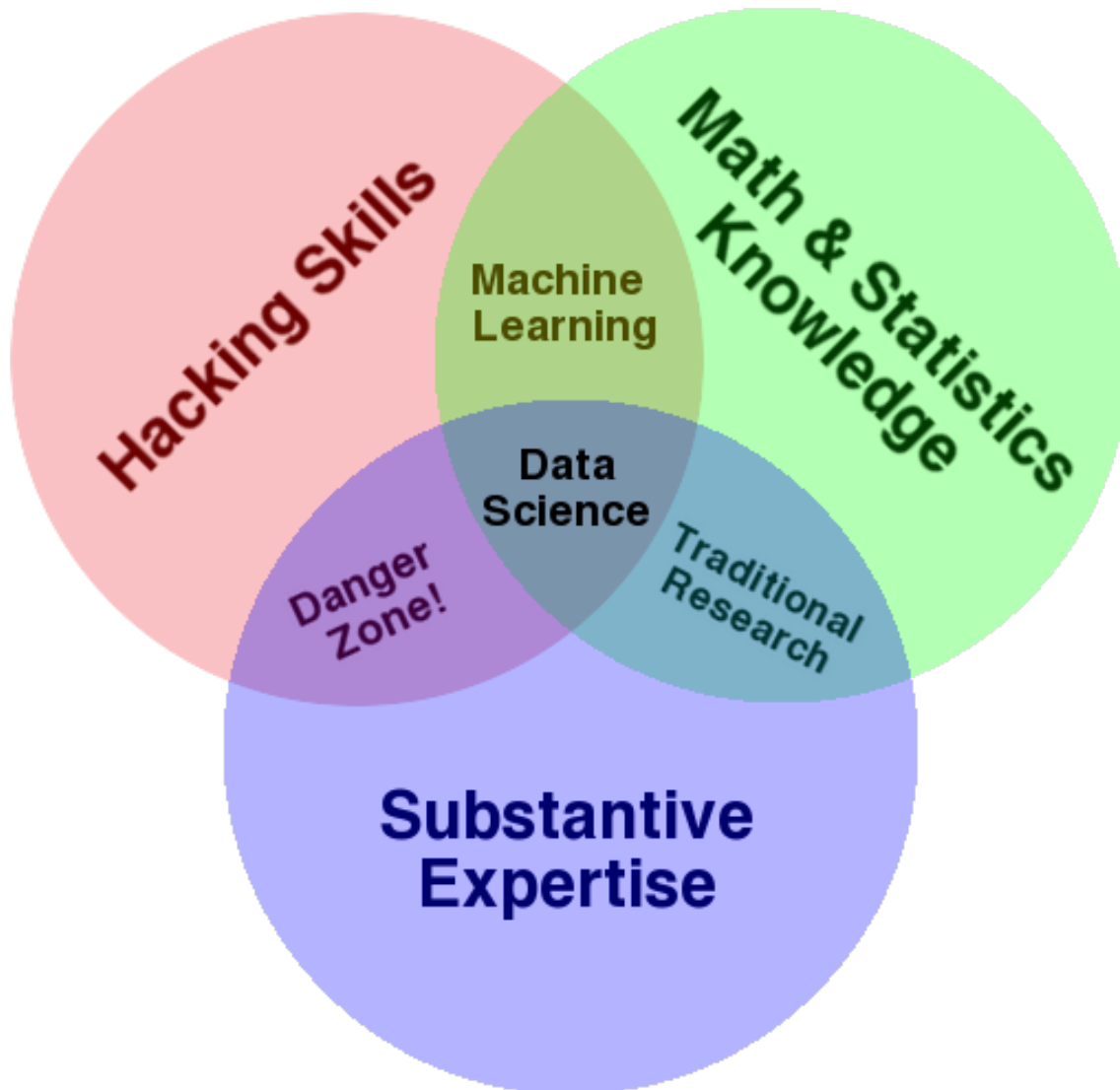
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The Netflix logo, featuring the word "NETFLIX" in white, bold, sans-serif capital letters on a red rectangular background.The Google logo, consisting of the word "Google" in its multi-colored, rounded sans-serif font.The LinkedIn logo, featuring the word "Linked" in black and "in" in white inside a blue square, followed by a registered trademark symbol.The Facebook logo, featuring the word "facebook" in white, lowercase, sans-serif font on a blue rectangular background.The Microsoft logo, consisting of four colored squares (red, green, blue, yellow) arranged in a 2x2 grid, followed by the word "Microsoft" in a grey, sans-serif font.The CIVIS ANALYTICS logo, featuring the word "CIVIS" in large, orange, sans-serif capital letters, with the word "ANALYTICS" in smaller, grey, sans-serif capital letters below it.The Nest logo, featuring the word "nest" in a blue, lowercase, rounded sans-serif font, with a small trademark symbol (TM) to the right.

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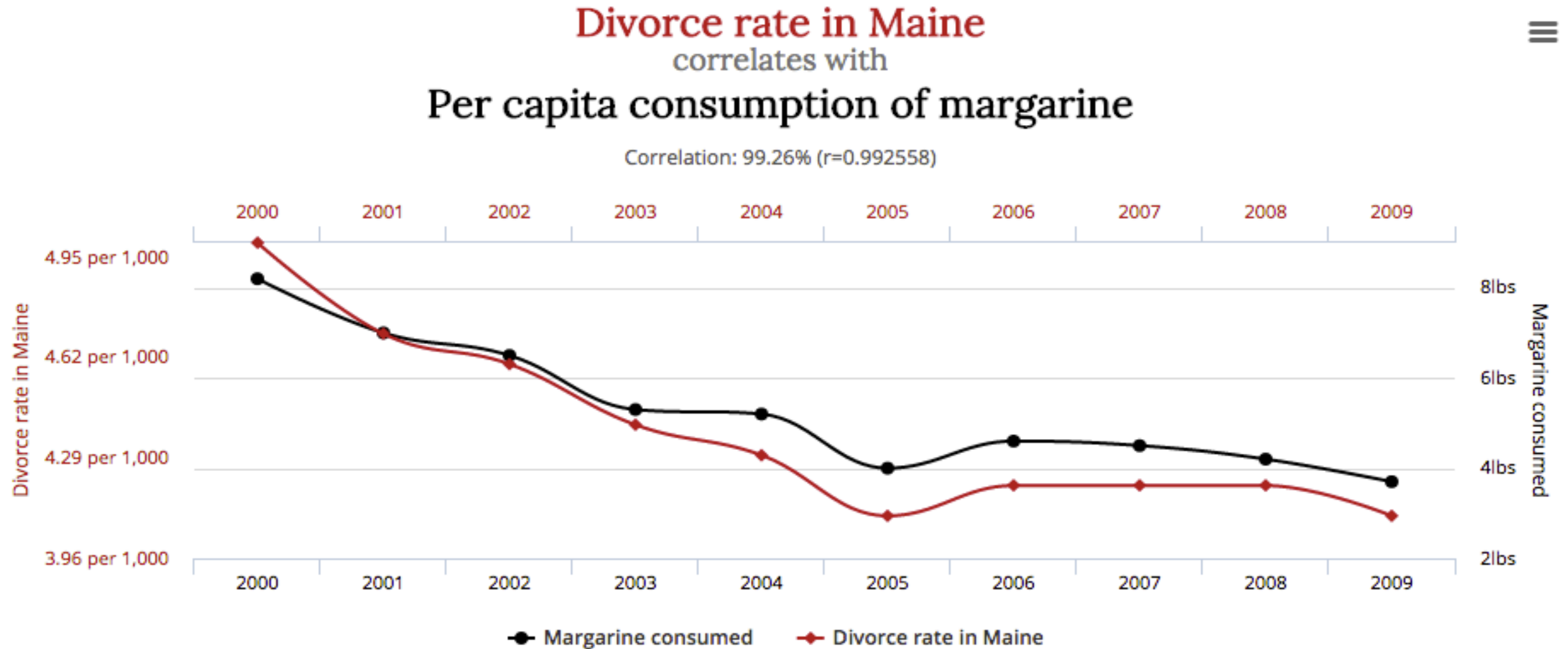
# WHAT QUALITIES MAKE UP A DATA SCIENTIST?

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- Hacking skills
- Math and Stats knowledge
- Substantive expertise

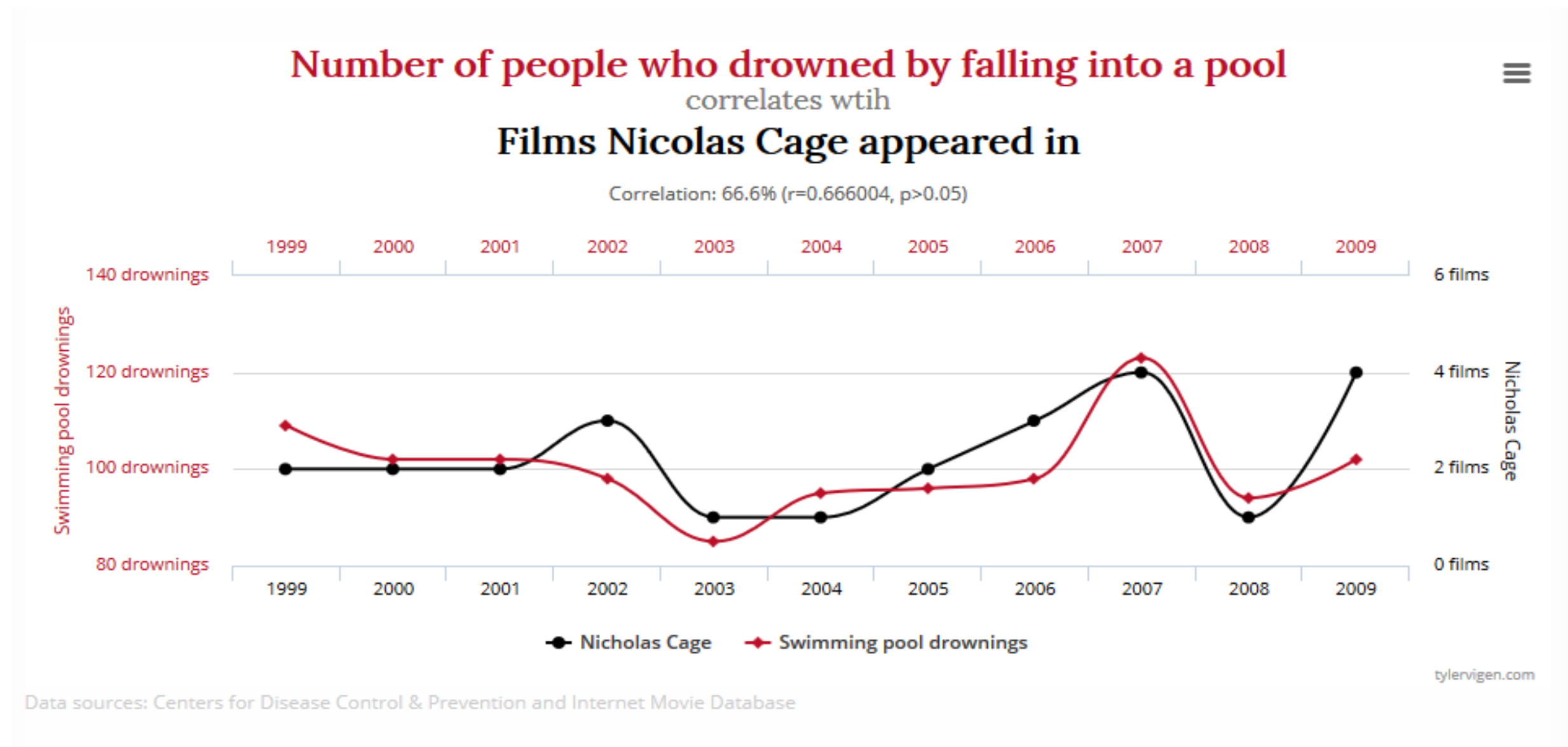
# WHAT QUALITIES MAKE UP A DATA SCIENTIST?



tylervigen.com

Data sources: National Vital Statistics Reports and U.S. Department of Agriculture

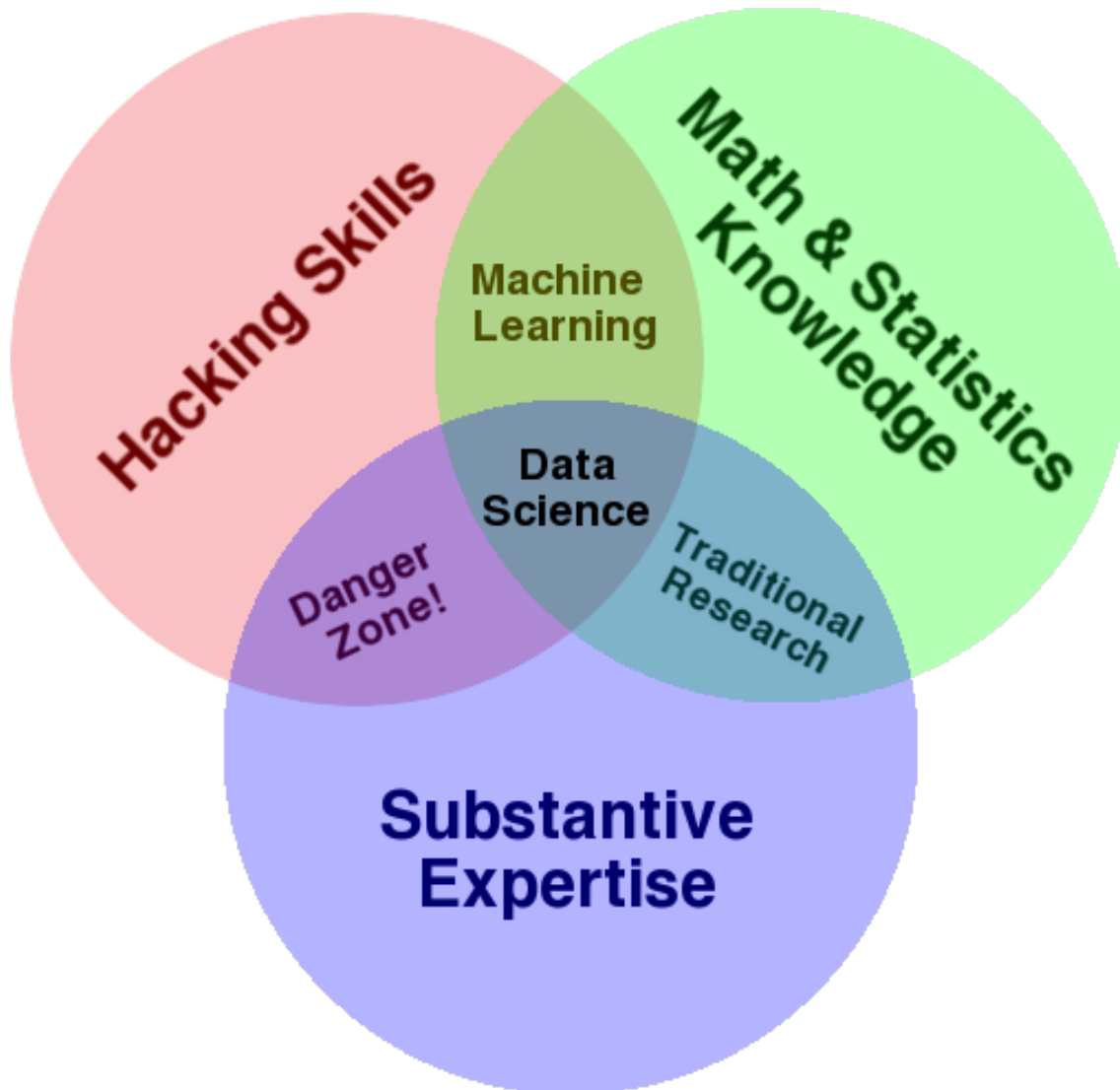
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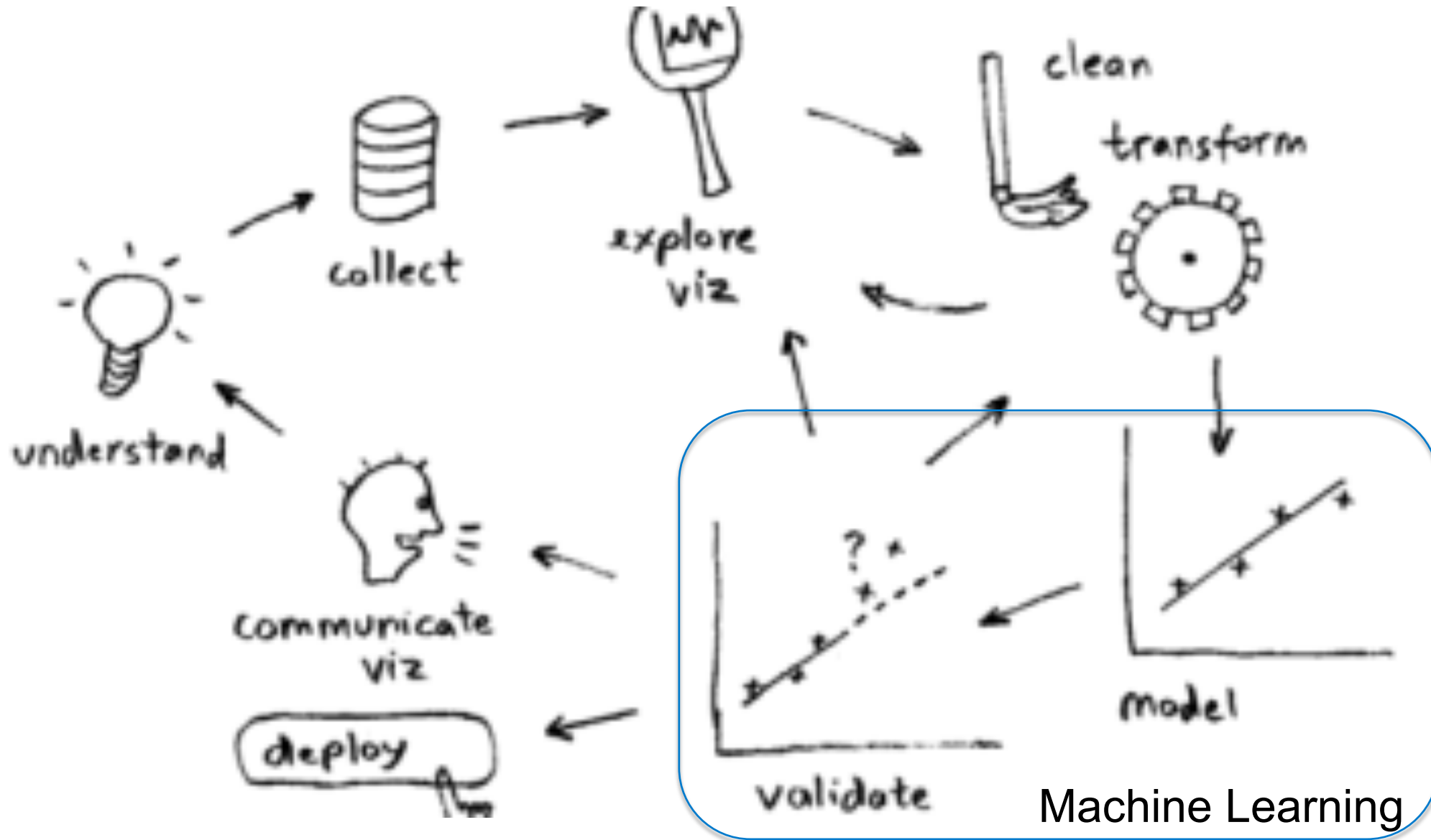
# WHAT QUALITIES MAKE UP A DATA SCIENTIST?

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- Hacking skills
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- Lastly.....Communication skills!

# DATA SCIENCE WORKFLOW



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# DATA SCIENCE WORKFLOW EXAMPLE

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## PROBLEM: HOW MUCH SHOULD I CHARGE FOR A NEW CPU?

**Understand:** Can my previous CPU sales help predict future \$ sales? I would like to predict \$ Sales based on known quantities

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# DATA SCIENCE WORKFLOW EXAMPLE

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i.e. number of cores, clock speed, L1 and L2 cache sizes, number of competing chips, \$ sales



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**Model:** You'll learn how to do this 😊

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**Model:** You'll learn how to do this 😊

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**Communicate:** Great! So the \$Sales of a new CPU can be predicted based on a mixture of Gaussian variables based on logarithmic cpu clock speed,  $10.45 * \# \text{ of cores}$ ,  $(\# \text{Cores})^2$ , and  $\exp(\# \text{ of competing chips})$ .

Now, how do you communicate this to a non-technical audience?

# PROBLEM: HOW WOULD YOU IMPLEMENT “MORE ITEMS TO CONSIDER” ON AMAZON.COM?

### Frequently Bought Together

Total price: **\$154.18**

[Add all three to Cart](#)

[Add all three to List](#)

These items are shipped from and sold by different sellers. [Show details](#)

- ☒ **This Item:** Beats Solo2 Wired On-Ear Headphones - Black **\$139.10**
- ☒ Matte Zipper Earphones Carrying Case for Beats Monster by Dr.Dre Studio, Solo Wireless, Solo, Solo... **\$7.99**
- ☒ Original Replacement Cable/Wire For Beats By Dre Headphones Solo/Studio/Pro/Detox/Wireless... **\$7.09**

V-MODA Crossfade M-100 3D Custom Headphones  
★★★★★ (1161)  
\$310.00 **\$199.99**

[Add feedback](#)

### Customers Who Bought This Item Also Bought

Matte Zipper Earphones Carrying Case for Beats Monster by Dr.Dre Studio, Solo Wireless, Solo, Solo...  
★★★★☆ 95  
**#1 Best Seller** in Headphone Cases  
**\$7.99**

Original Replacement Cable/Wire For Beats By Dre Headphones Solo/Studio/Pro/Detox...  
★★★★☆ 653  
**\$7.09** *Prime*

Beats Solo HD On-Ear Headphone (Discontinued by Manufacturer - Black) wired  
★★★★☆ 1,064  
**\$127.99** *Prime*

Black Menba Matte Zipper Earphones Carrying Case/ Pouch/ Box for Beats & Monster  
★★★★★ 2

Beats Solo HD On-Ear Headphone - Light Blue (Certified Refurbished)  
★★★★☆ 76  
**\$139.99** *Prime*

Bluecell Protection Carrying Hard Case/Bag for Monster Dr Dre Beats Solo/Studio Headphone  
★★★★☆ 419  
**\$4.35**

Official Monster Beats By Dr. Dre 3.5mm In ear/earbuds Stereo Headset for HTC Red...  
★★★★☆ 334  
**\$54.90**



# MACHINE LEARNING

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# WHAT IS MACHINE LEARNING?

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from Wikipedia:

Machine learning explores the study and construction of algorithms that can *learn from* and make predictions on data.



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# WHAT IS MACHINE LEARNING?

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“A computer program is said to learn from experience **E** with respect to some set of tasks **T** and performance measure **P**, if its performance at tasks **T**, as measured by **P**, improves with experience **E**.”



Tom Mitchell,  
Professor CMU

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## WHAT IS MACHINE LEARNING?

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“A computer program is said to learn from experience **E** with respect to some set of tasks **T** and performance measure **P**, if its performance at tasks **T**, as measured by **P**, improves with experience **E**.”

“A student is said to learn from the General Assembly **Data Science Course** with respect to some set of **homeworks** and measured by **grades**, if its performance at **homeworks** as measured by **grades**, improves throughout the **course**”

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# WHAT IS MACHINE LEARNING?

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from Wikipedia:

Machine learning explores the study and construction of algorithms that can *learn from* and make predictions on data.

“The core of machine learning deals with **representation** and **generalization**...”

**Representation** – extracting a mathematical structure from data

**Generalization** – making predictions from data

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# TAXONOMY OF MACHINE LEARNING PROBLEMS

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***Supervised***

Labeled examples - Making Predictions (**generalization**)

***Unsupervised***

No labeled examples - Discovering patterns (**representation**)

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# TAXONOMY OF MACHINE LEARNING PROBLEMS

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## ***Supervised Example***

Jim is 30 years old and can eat 4 donuts. Sally can eat 2 donuts and is 60 years old. Bobby is 15 years old. How many donuts can he probably eat?

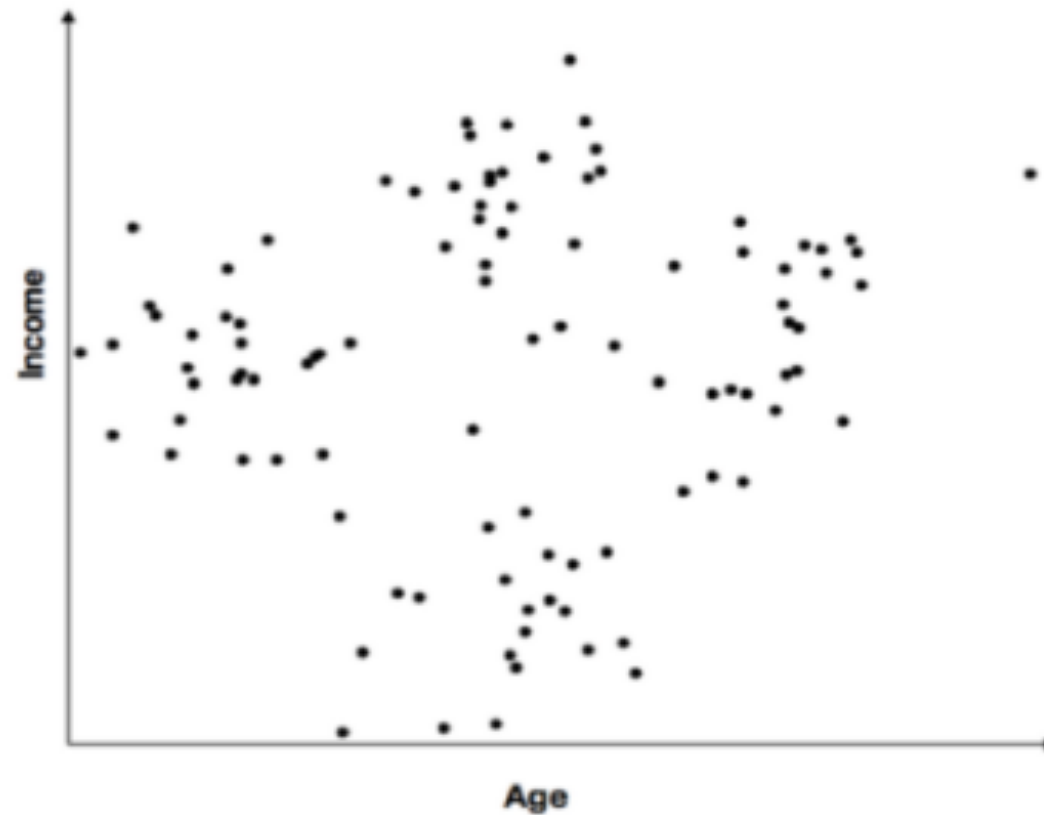
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# TAXONOMY OF MACHINE LEARNING PROBLEMS

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## *Unsupervised Example*

Can we find structure to unlabeled data?

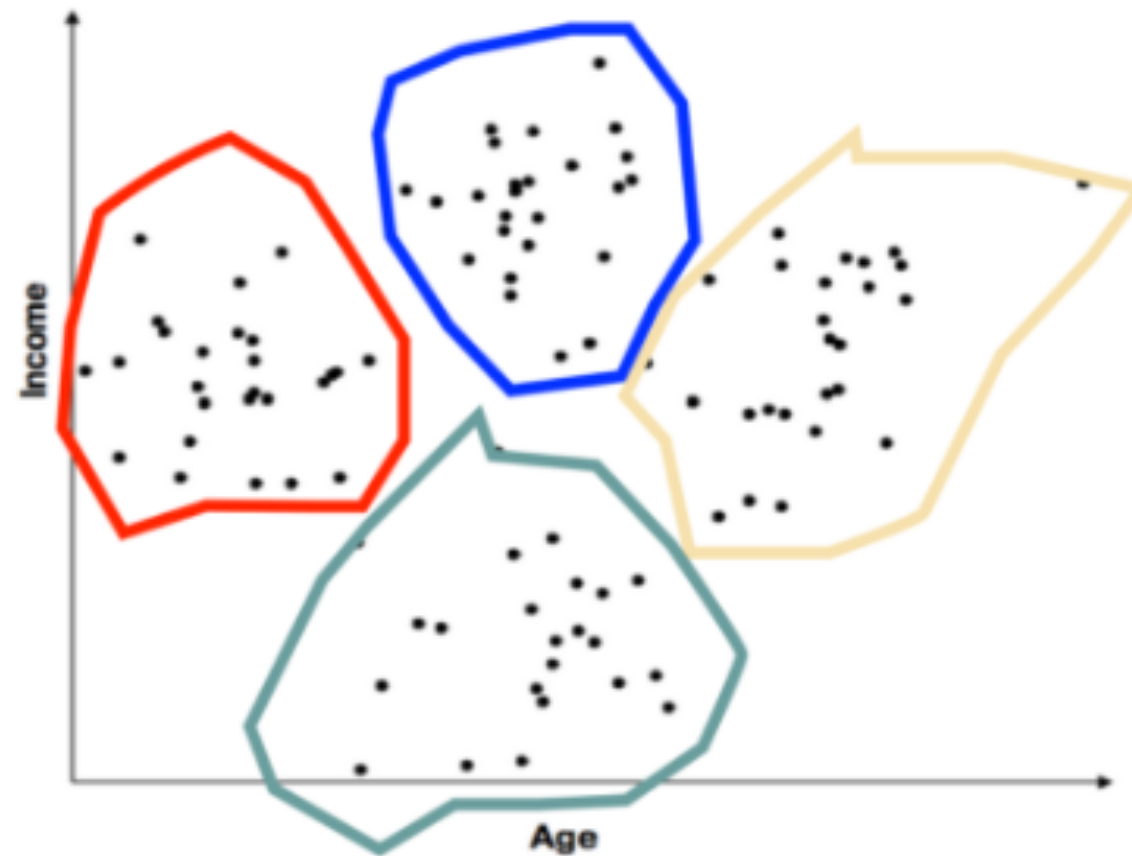


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# TAXONOMY OF MACHINE LEARNING PROBLEMS

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## *Unsupervised Example*



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# TAXONOMY OF MACHINE LEARNING PROBLEMS

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***Continuous***

***Categorical***

**Quantitative**  
(ordered data, age,  
Height, salary, etc.)

**Qualitative**  
(sets, yes/no, vote, etc.)



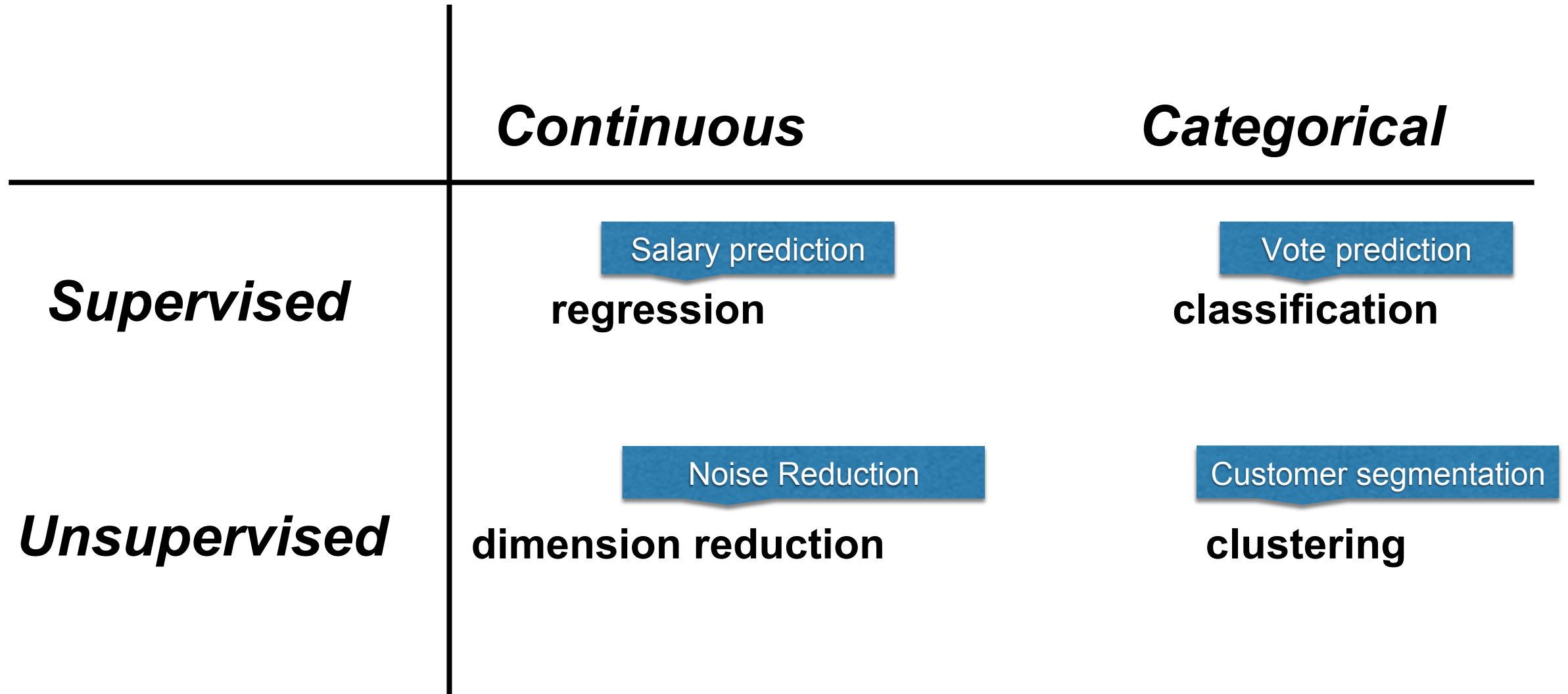
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# TAXONOMY OF MACHINE LEARNING PROBLEMS

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	<i>Continuous</i>	<i>Categorical</i>
<i>Supervised</i>	regression	classification
<i>Unsupervised</i>	dimension reduction	clustering

# TAXONOMY OF MACHINE LEARNING PROBLEMS



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## **SUPERVISED OR UNSUPERVISED?**

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**You want to determine whether an email is spam or not**

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## **SUPERVISED OR UNSUPERVISED?**

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**You want to group Amazon customers together based on their previous purchases, location, and number of visits to the website so you can advertise to them specifically**

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## **SUPERVISED OR UNSUPERVISED?**

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**You want to predict the rating of a Netflix movie**

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## **SUPERVISED OR UNSUPERVISED EXERCISE**

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**In a group, answer what kind of ML problems these can be classified as:**

- Pandora Music Recommendation (i.e. What songs would you like)
- Digit recognition (i.e. post office performs digit recognition on mail)
- Predicting likelihood (i.e. probability) of a student passing high school
- You want to automatically reduce noise in your dataset
- You want to predict whether someone prefers Chevy or Ford based on their level of Car knowledge (1-10), age, and whether they like LS engines or Coyote engines

**Homework 1 on Github – Due Dec 9 before class!**

**Exit tickets (This is Lesson 1)**