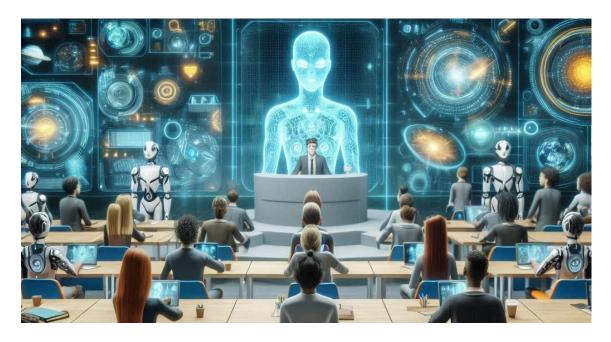
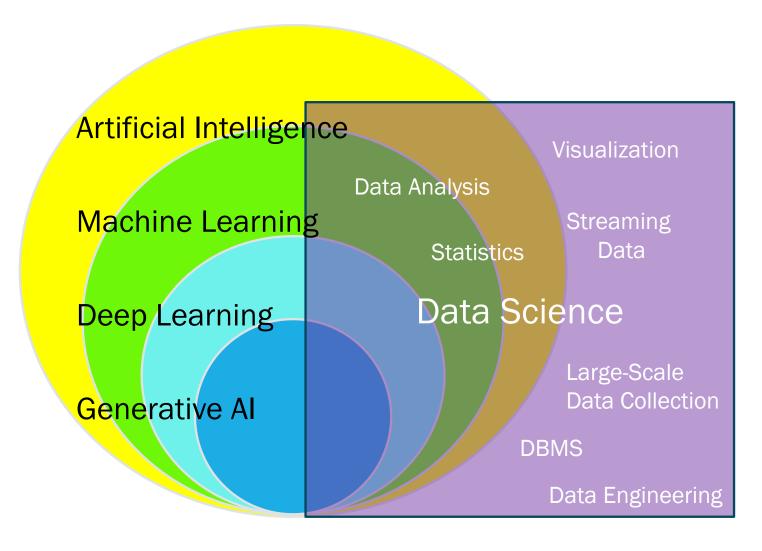
# CS 180 INTRODUCTION TO DATA SCIENCE

INTRODUCTION, COURSE OVERVIEW, OBJECTIVES, HOW TO SUCCEED



Created by DALL-E Prompt: "Artificial Intelligence Classroom"

#### WHAT IS DATA SCIENCE?



#### **Data Science:**

- Scope: focuses on data as a whole, including data collection, processing, analysis, storage, and management.
- concerned with extracting knowledge and actionable insights from data.
- Techniques: Uses data collection, data cleaning, data transformation, statistical analysis, data visualization, data management and data engineering tools.

# THIS IS DATA SCIENCE!

- What do you see?
- What should you do?
- Analyze the data, or act?



# **ANNOUNCEMENTS**

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WVB 2177

Office Hours: TBD

# **Teaching Assistants:**

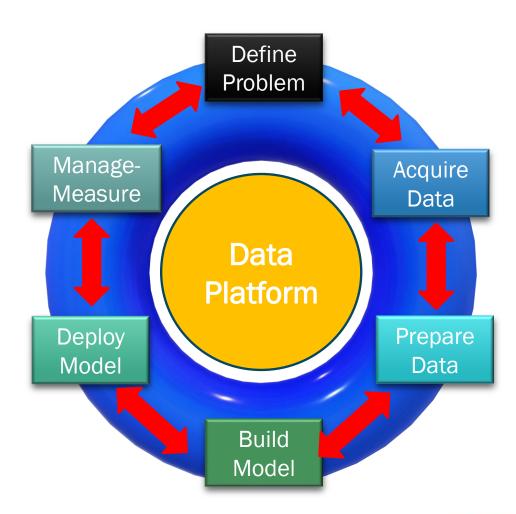
- 1. Bowen Liu
- 2. Patrick Willmott
- 3. Caleb Christensen
- 4. Justin Krogel
- 5. Nathan Roberts
- 6. Eden Evans
- 7. Nathan Roberts
- 8. Cameron Betteridge

Office Hours: (See syllabus)



# **TECH STUFF WE'LL LEARN**

- The Data Science Lifecycle process
- Using Python for data science (Numpy,
   Pandas, Matplotlib, Scikit-Learn, and more)
- How to prepare data for analysis
- How to explore data for insights
- Data Visualization (Python and Tableau)
- Machine Learning basic algorithms
- Use of GenAl tools for language-based problems





#### **WE'LL ALSO TACKLE DATA LITERACY**

- Mental frameworks for decomposing data science problems
- Critical thinking about potential conclusions of an analysis
- Potential pitfalls of overreliance on unreliable data
- When is someone lying to you with statistics?
- => data literacy assignments



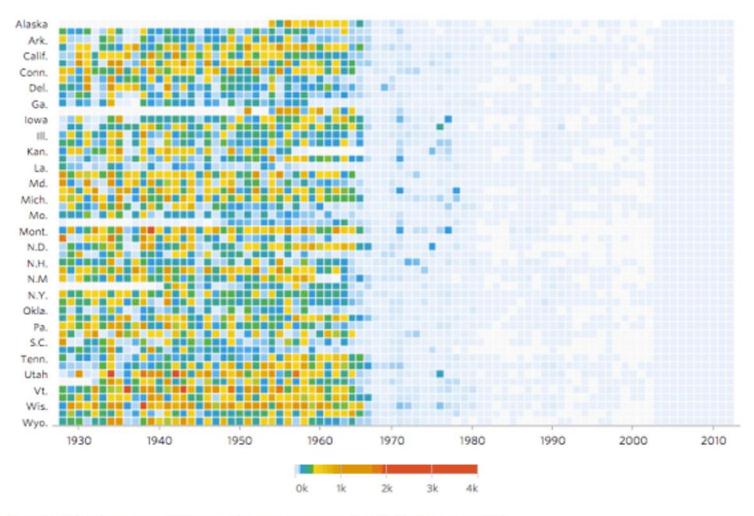
#### **COURSE GOALS**

- Begin the journey to become a "Full-Stack" Data Scientist by:
  - Using advanced tools to extract insights from data
  - Solving novel problems with core data science principles
  - Communicating insights effectively through visualization
  - Critically evaluating data-driven conclusions





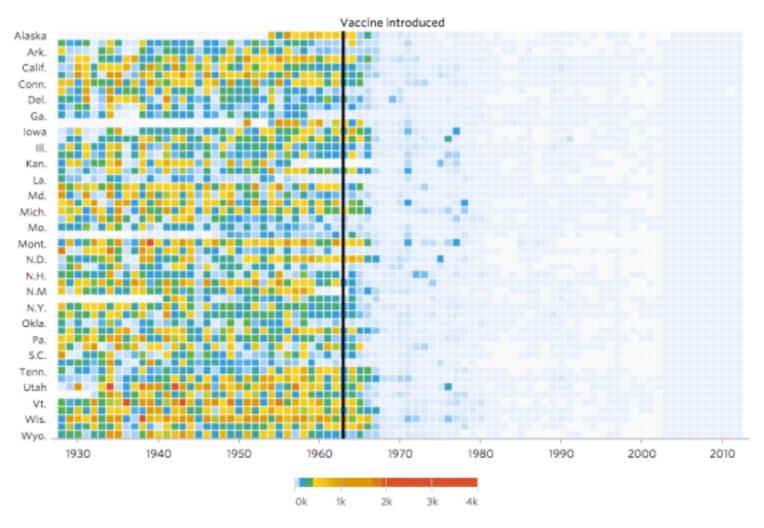
#### Measles



Note: CDC data from 2003-2012 comes from its Summary of Notifiable Diseases, which publishes yearly rather than weekly and counts confirmed cases as opposed to provisional ones.

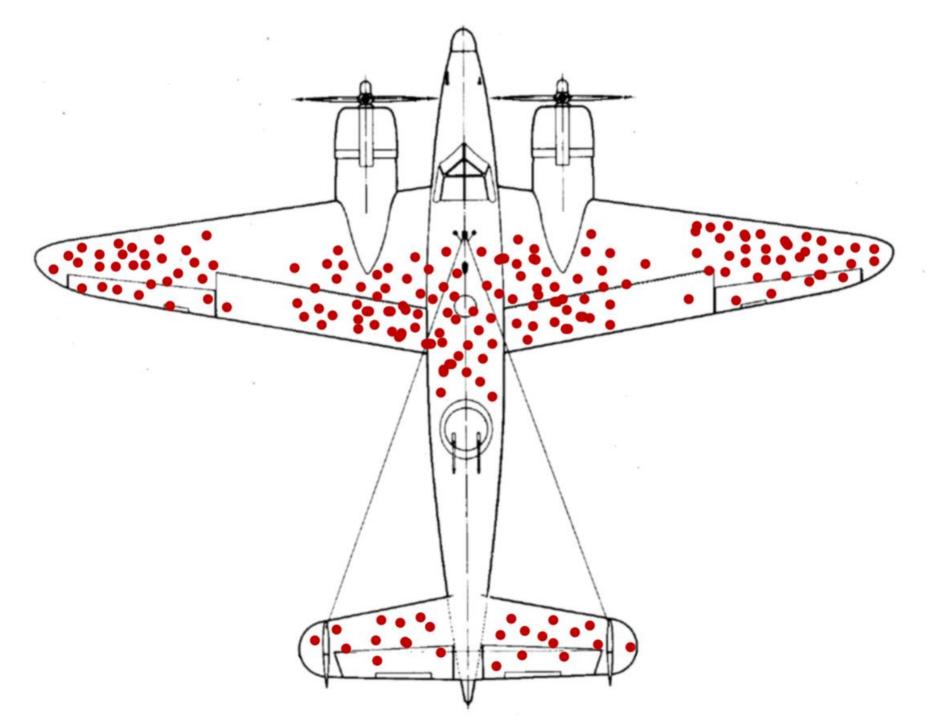


#### Measles



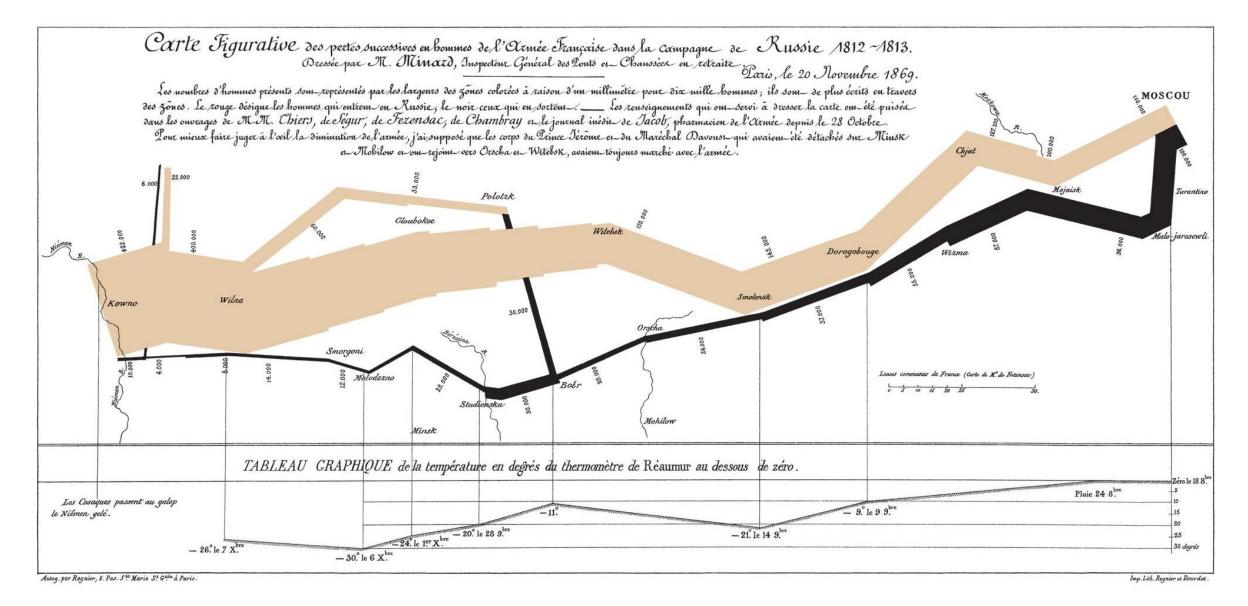
Note: CDC data from 2003-2012 comes from its Summary of Notifiable Diseases, which publishes yearly rather than weekly and counts confirmed cases as opposed to provisional ones.





# **BYU**

#### NAPOLEON'S DISASTROUS INVASION OF RUSSIA IN 1812



# **GENDER BIAS AT BERKLEY (1973)**

Are men applying to Berkeley more likely to get in than women?

	Men		Women	
	Applicants	Admitted	Applicants	Admitted
Total	8442	44%	4321	35%



# **GENDER BIAS AT BERKLEY (1973)**

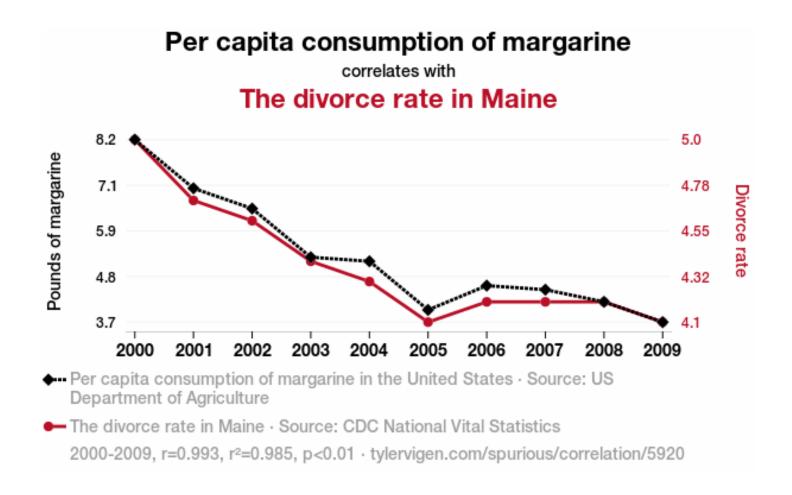
# Are men applying to Berkeley more likely to get in than women?

	Men		Women	
	Applicants	Admitted	Applicants	Admitted
Total	8442	44%	4321	25%

- Departments have different acceptance rates,
- More women applied to departments with lower acceptance rates

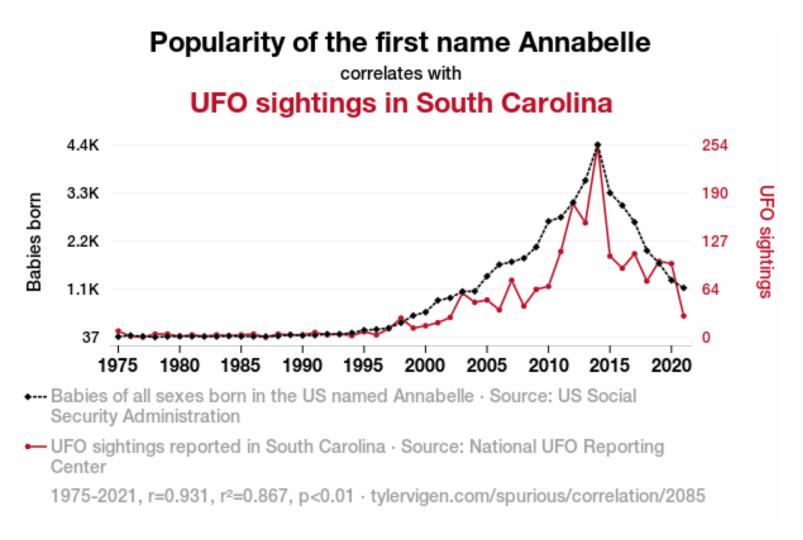
Donartment	Ме	n	Women	
Department	Applicants	Admitted	Applicants	Admitted
A	825	62%	108	82%
В	560	63%	25	68%
С	325	37%	593	34%
D	417	33%	375	35%
E	191	28%	393	24%
F	373	6%	341	7%

# **SPURIOUS CORRELATIONS**



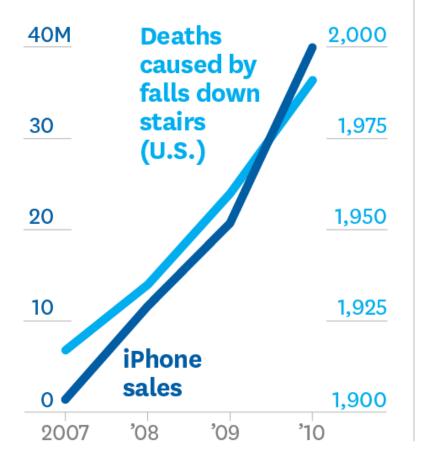


# **SPURIOUS CORRELATIONS**

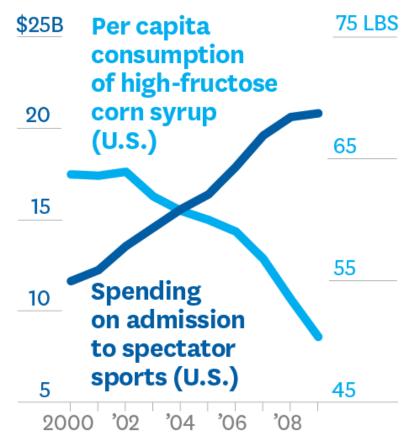




#### MORE IPHONES MEANS MORE PEOPLE DIE FROM FALLING DOWN STAIRS



#### LET'S CHEER ON THE TEAM, AND WE'LL LOSE WEIGHT



# TO INCREASE AUTO SALES, MARKET TRIPS TO UNIVERSAL ORLANDO



# **ZYBOOK DATA SCIENCE LIFECYCLE FOR DATA ANALYSIS**

Table 1.4.1: Data science lifecycle.

Step	Description
Step 1: Gathering data	Identify available and relevant data; gather new data if needed.
Step 2: Cleaning data	Reformat datasets, create new features, and address missing values.
Step 3: Exploring data	Create data visualizations and calculate summary statistics to explore potential relationships in the dataset.
Step 4: Modeling data	Use modeling skills and content knowledge to fit and evaluate models, measure relationships, and make predictions.
Step 5: Interpreting data	Describe and interpret conclusions from data through written reports and presentations.



# **DEFINE THE PROBLEM**

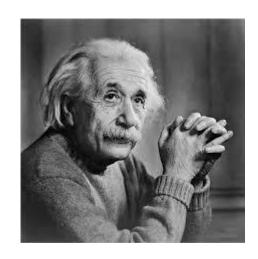
- What is the core problem?
- What processes, systems, orgs are affected?
- If solved, what is business value?
- How can problem be scoped?
- How is value measured?
- Characterize problem domain
- Is this a datadriven problem?
- What data is needed? (prelim)



IF I HAD AN HOUR TO SOLVE A PROBLEM I'D SPEND 55 MINUTES

THINKING ABOUT THE PROBLEM AND 5 MINUTES THINKING ABOUT SOLUTIONS.

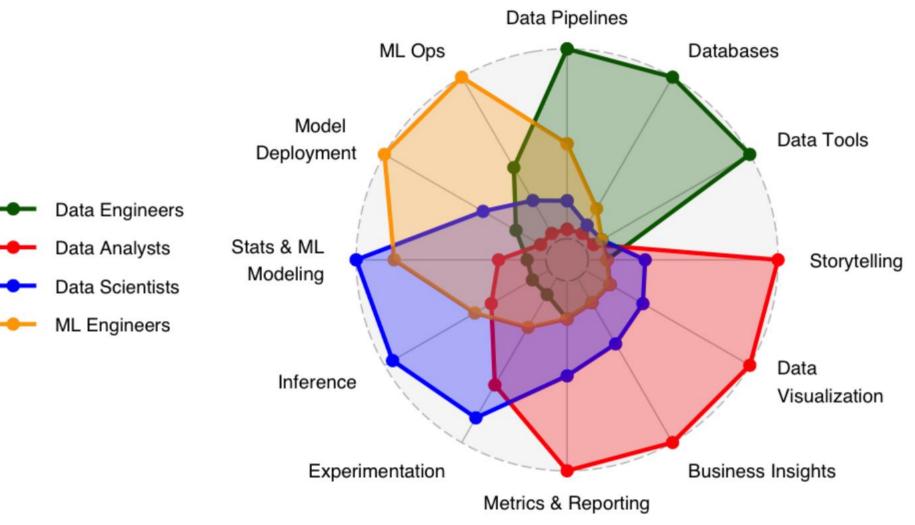




**Albert Einstein** 



# SPIDER CHART OF RELATIVE SKILLS FOR KEY DATA ROLES



# **EXTENDED ROLES ON A DATA SCIENCE APPLICATION PROJECT**

Role	Description
Data Engineer	Builds data pipelines, joins tables, converts data formats, prepares data for use by Data Scientists.
Data Scientist	Prepares data for modeling, extracts features, builds models
Data Analyst	Expert in SQL, BI, Excel, Analyzing data but not necessarily a domain expert (Tableau, PowerBI most popular tools)
AI/ML Engineer	Builds ML pipeline, integrate enterprise systems, monitor & manage models, skilled software engineer & data scientist
Enterprise Architect	Integrates DS applications into enterprise system (e.g., microservices, API gateway, event brokers, etc.)
Data Architect	Defines data management system architecture, data model
Data Governance Lead	Responsible for meta data, data catalog, data access, change management policies
Business Analyst	Expert in a particular domain (e.g., Finance), can use BI tools and Excel if set-up by the Data Analyst.
Program Manager	Keeps track of projects, personnel, budgets, identifies conflicts, dependencies, resource constraints
Application Engineer	Expert in technology for a particular domain or problem (e.g., Finance, Marketing, Sales, Manufacturing, etc.)
UI/UX Engineer	Designs, prototypes, and builds the user interface (mobile and web)
Product Manager	Responsible for overall product design, prioritization, deployment and assuring business value
Infrastructure Engineer	Expert in cloud and data management infrastructure
Security/Privacy Engineer	Assures application architecture is compliant with Enterprise security and privacy standards
Executive Sponsor	Oversees application development. Responsible for resourcing. Communicates with Executive Leadership.

#### THE GOAL: FULL-STACK DATA SCIENTIST AND BEYOND

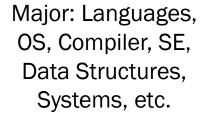
Data Engineer: ETL Pipelines, SQL, Data Prep

Data Analyst: Visualization, EDA, Reports

Data Scientist: Build Predictive, Generative Models

ML Engineer: Deploy, Monitor Models, Apps

**Data Scientist** 



Minor: Math, Lang, Business, Econ, History, etc.

General Ed: History, Languages, Math, Science

Gospel: BofM, New Test, Old Test, D&C

**University Student** 

Mental

**Emotional** 

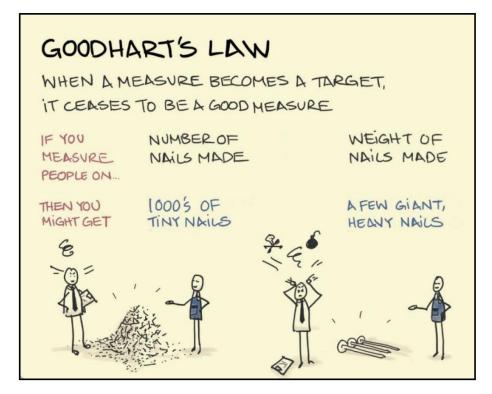
Social

Physical

Spiritual

**Human Being** 

# **GOOD TO KNOW...**



Students understand this law very well. It is easy to get caught up focusing on getting the "A" instead of mastering the knowledge or skill.

# Don't let school... get in the way of your education ©

