CS 180 INTRODUCTION TO DATA SCIENCE

INTRODUCTION, COURSE OVERVIEW, OBJECTIVES, HOW TO SUCCEED



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

SOME ANNOUNCEMENTS



OPEN HOUSE for 5x AI & DATA SCIENCE MAJORS!

Pop in to grab food, ask questions, meet the new Data Science Administrator, and learn about new majors, clubs, and events!

Sept 3rd-5th, 8th-10th 8am-5pm West View Building (WVB) #1162

STEM fair



Thurs., September 11 12:00 PM – 1:00 PM EB 204



Thurs., September 18 9:00 AM – 3:00 PM WSC Ballroom & Garden Court

Connect with top employers and secure your next job or internship!

BYU Career Services



BYU Statistics **BYU** Career Services

Data Science and Actuarial Career Fair

Wed., October 1

9:00 AM - 12:00 PM

Lunch*: 12:00 PM - 1:30 PM

Hinckley Center Assembly Hall



Register here!



*RSVP required for lunch.



ANNOUNCEMENTS

Dr. Jake Rhodes

rhodes@stat.byu.edu

WVB 2177

Office Hours: TBD

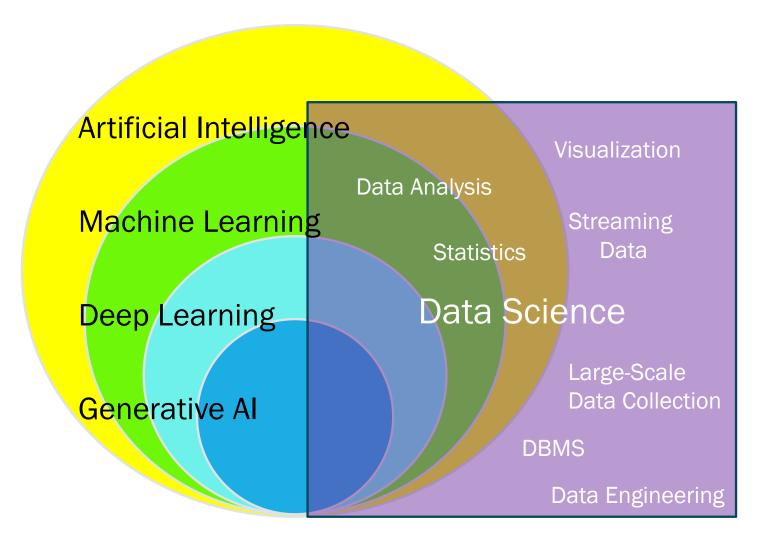
Teaching Assistants (WVB 1151):

- 1. Toby Allen
- 2. Corbin Christiansen
- 3. Michael Jensen
- 4. Spencer Marshall
- 5. Ian Villanueva

Office Hours: Posted on Syllabus



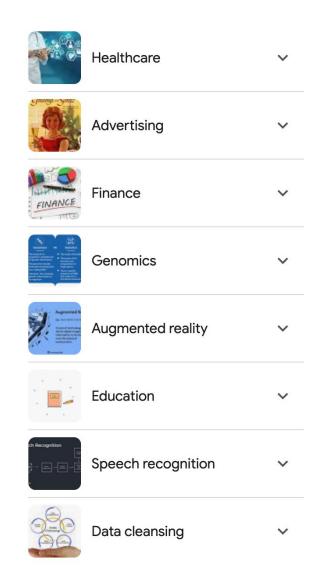
WHAT IS DATA SCIENCE?

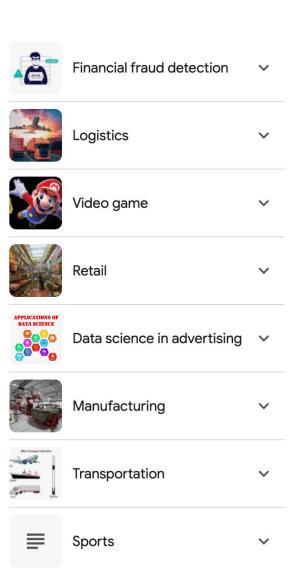


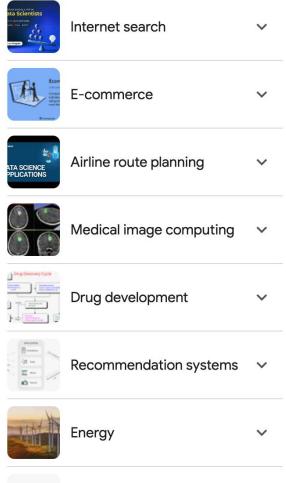
Data Science:

- 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.

WHY DATA SCIENCE?





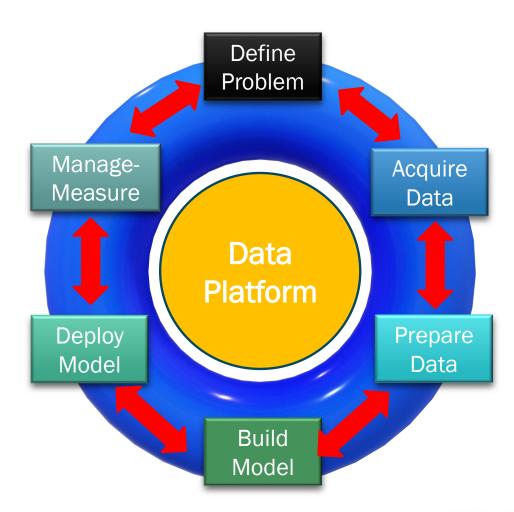


Virtual assistants



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)
- Data Management (SQL)
- Machine Learning basic algorithms
- Use of GenAl tools for language-based problems





GEN AI APPLICATIONS



Code















suno.com

WE WILL COVER A LOT THIS SEMESTER!

Class # \Xi	Week# 束	Month =	Date =	Topic	Reading =	Labs	Due date
1	1	Sep	4	Welcome, Introduction, Course Objectives, DS Lifecycle	Chapter 1 Intro DS	Lab 1: IDE Set Up, GitHub	Sept 6, 2025
2	2	Sep	9	Python setup, Google colab, Github	Chapter 2.1-2.5 Python		
3	2	Sep	11	Numpy, vectorization, reading/writing		Lab 2: Vectorization	Sept 13, 2025
4	3	Sep	16	Pandas, Matplotlib, Seaborn	Chapter 2.6-2.8 Python		
5	3	Sep	18	Data Cleaning and Preparation		Lab 3: Pandas, Numpy	Sept 20, 2025
6	4	Sep	23	Data Acquisition, ETL, Populations, Sampling	Chapter 3 Data Prep		
7	4	Sep	25	Descriptive Statistics	Chapter 4 Prob & Stat	Lab 4: Data Preparation	Sept 27, 2025
8	5	Sep	30	Exploratory Data Analysis (EDA)	Chapter 5 EDA		
9	5	Oct	2	Principles of Data Visualization		Lab 5: Data Visualization	Oct 04, 2025
10	6	Oct	7	Business Intelligence with Tableau	Chapter 6 BI & Tableau		
11	6	Oct	9	Tableau Dashboards, Stories		Lab 6: Data Story Telling with Tableau	Oct 11, 2025
12	7	Oct	14	Data management - databases, SQL queries			
13	7	Oct	16	More SQL Features, Joins	Chapter 7 DB and SQL		
14	8	Oct	21	SQL continued, NoSQL, Connect to Tableau			
15	8	Oct	23	MIDTERM REVIEW		Midterm	Oct 25, 2025
16	9	Oct	28	Overview of AI, ML, DL, GenAI Topics for remainder of semester	Chapter 8 Unsupervised Learning		
		Oct	30	Unsupervised Learning- Kmeans		Lab 7: Data Engineering with SQL	Nov 1, 2025
17	9	OCC	30				
17 18	9	Nov	4	Unsupervised Learning- Hierarchical, DBSCAN	Chapter 9 Supervised Learn		
				Unsupervised Learning- Hierarchical, DBSCAN Supervised Learning: Part 1	Chapter 9 Supervised Learn Chapter 10 Decision Trees	Lab 8: Cluster Analysis	Nov 08, 2025
18	10	Nov	4			Lab 8: Cluster Analysis	Nov 08, 2025
18 19	10 10	Nov Nov	4	Supervised Learning: Part 1	Chapter 10 Decision Trees Chapter 11 Regression (optional)	Lab 8: Cluster Analysis Lab 9: ML Classification/Regression	Nov 08, 2025 Nov 15, 2025
18 19 20 21 22	10 10 11	Nov Nov	4 6 11	Supervised Learning: Part 1 Supervised Learning: Part 2	Chapter 10 Decision Trees Chapter 11 Regression (optional)		
18 19 20 21	10 10 11 11	Nov Nov Nov	4 6 11 13	Supervised Learning: Part 1 Supervised Learning: Part 2 Evaluation of models, comparing performance	Chapter 10 Decision Trees Chapter 11 Regression (optional) Chapter 12 Eval		
18 19 20 21 22	10 10 11 11 11	Nov Nov Nov Nov	4 6 11 13 18	Supervised Learning: Part 1 Supervised Learning: Part 2 Evaluation of models, comparing performance ANN, Multi-Layer Perceptron, Backpropagation Deep Learning GenAl - Introduction	Chapter 10 Decision Trees Chapter 11 Regression (optional) Chapter 12 Eval	Lab 9: ML Classification/Regression	Nov 15, 2025
18 19 20 21 22 23	10 10 11 11 12 12	Nov Nov Nov Nov Nov	4 6 11 13 18 20	Supervised Learning: Part 1 Supervised Learning: Part 2 Evaluation of models, comparing performance ANN, Multi-Layer Perceptron, Backpropagation Deep Learning	Chapter 10 Decision Trees Chapter 11 Regression (optional) Chapter 12 Eval Chapter 13 ANN	Lab 9: ML Classification/Regression	Nov 15, 2025
18 19 20 21 22 23	10 10 11 11 12 12 13	Nov Nov Nov Nov Nov	4 6 11 13 18 20 25	Supervised Learning: Part 1 Supervised Learning: Part 2 Evaluation of models, comparing performance ANN, Multi-Layer Perceptron, Backpropagation Deep Learning GenAl - Introduction	Chapter 10 Decision Trees Chapter 11 Regression (optional) Chapter 12 Eval Chapter 13 ANN	Lab 9: ML Classification/Regression	Nov 15, 2025
18 19 20 21 22 23 24 - 25 26	10 10 11 11 12 12 13 13 14	Nov Nov Nov Nov Nov Nov Dec Dec	4 6 11 13 18 20 25 27	Supervised Learning: Part 1 Supervised Learning: Part 2 Evaluation of models, comparing performance ANN, Multi-Layer Perceptron, Backpropagation Deep Learning GenAl - Introduction No Class. Thanksgiving Holiday.	Chapter 10 Decision Trees Chapter 11 Regression (optional) Chapter 12 Eval Chapter 13 ANN Chapter 14 GenAl	Lab 9: ML Classification/Regression	Nov 15, 2025
18 19 20 21 22 23 24 -	10 10 11 11 12 12 13 13	Nov Nov Nov Nov Nov Nov Nov Dec	4 6 11 13 18 20 25 27 2	Supervised Learning: Part 1 Supervised Learning: Part 2 Evaluation of models, comparing performance ANN, Multi-Layer Perceptron, Backpropagation Deep Learning GenAl - Introduction No Class. Thanksgiving Holiday. GenAl - Applications	Chapter 10 Decision Trees Chapter 11 Regression (optional) Chapter 12 Eval Chapter 13 ANN Chapter 14 GenAl	Lab 9: ML Classification/Regression Lab 10: MLP and Backpropagation	Nov 15, 2025 Nov 22, 2025



COURSE GOALS

- Enhance your ability to quickly learn and master new concepts
- Utilize cutting-edge data science languages and tools to extract insights from data
- Effectively communicate insights through data visualization, dashboards, data storytelling
- Think critically about conclusions drawn from data analysis
- Recognize the potential to do good by applying Data Science to address critical global problems



WE ARE A LEARNING COMMUNITY

- Definition: A group of people who come together with the shared goal of learning, growing, and improving—often through collaboration, discussion, and mutual support. [ChatGPT]
- Get to know each other. How?
- At the beginning of every class, introduce yourself to those around you.
- Benefits: More effective learning, better performance, build a social network, better sense of belonging, and it's more fun!



AI "POLICY"

- Generative AI tools like ChatGPT, Gemini, Claude, Grok, etc., can be helpful to assistants in the learning process.
- If misused, GenAI can become a crutch, lead to sloppy thinking, and even a substitute for engagement with real people.
- We want you to have the best learning experience, resulting in foundational understanding and deep skill development.
- We recommend limiting GenAl use to answering questions about topics, looking up syntax for code, and understanding code examples, but not using GenAl to answer coding questions end-to-end.
- We trust you will follow this recommendation (aka policy). We will not police you, but if it is obvious, we will bring this to your attention.
- To help you better learn, we will have portions of the mid-term and final that you can't do with AI. As a result, it will benefit you to know the concepts to the point where you do not need AI.

UPCOMING ASSIGNMENTS

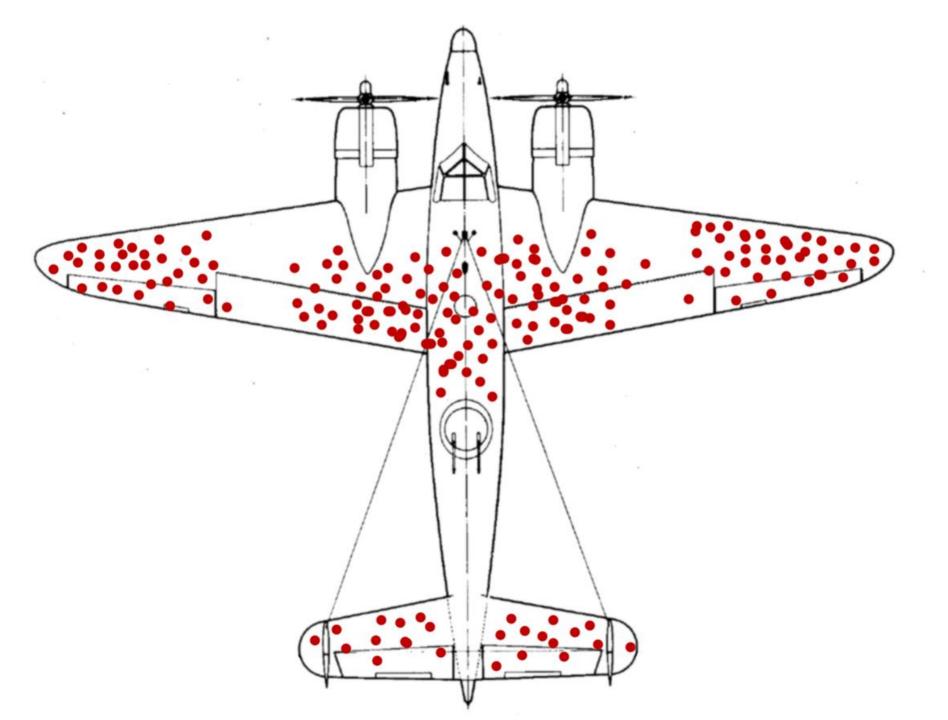
- 1. Sign up for zyBook. How? Go to the first reading assignment in Canvas and click "Load Chapter 1 Reading..." This will initiate the process of buying the zyBook. You should not have to do this again.
- 2. Reading Assignment: Chapter 1. Points automatically accrue as you do the activities in the book.
- 3. Data Science Lab 1: Intro to Colab. Set up a Python Development Environment. Programming assignments will be turned in via Google Colab notebooks. However, if you are planning a career requiring programming, I recommend using a professional IDE, such as VS Code. If needed, we'll go through the setup quickly next week.
- 4. Getting Set Up "Assignment": This will help you (and me) hit the ground running!





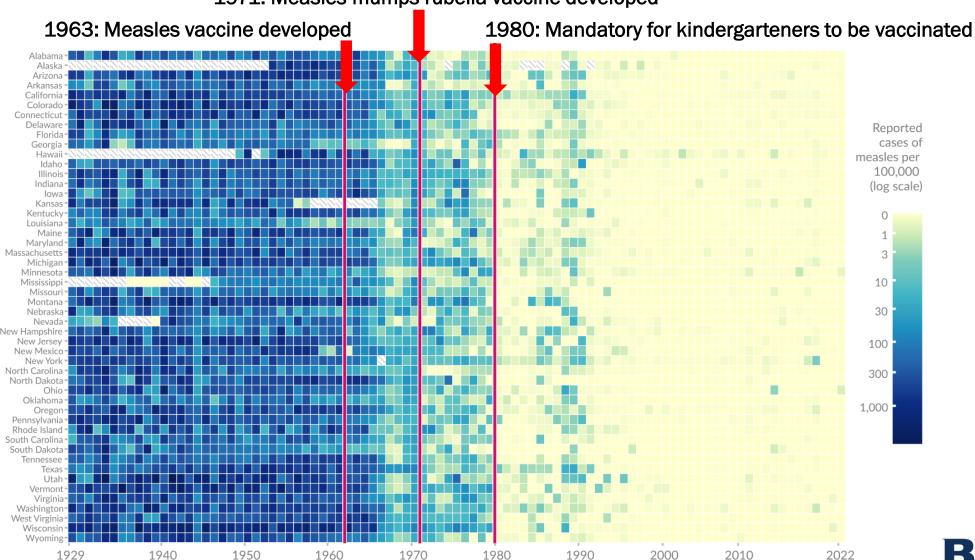
- What do you see?
- What should you do?
- Analyze or act?





BYU

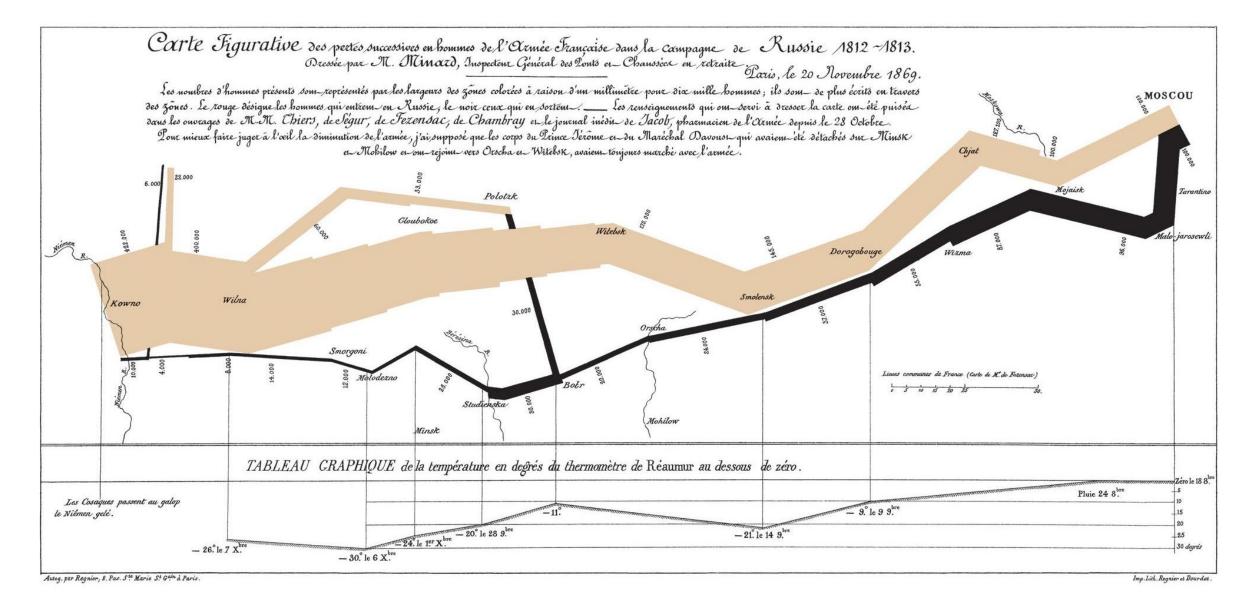




Data source: Project Tycho (2018); Centers for Disease Control and Prevention (1959–2022)

US States

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?

	Ме	n	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?

	Ме	n	Women	
	Applicants	Admitted	Applicants	Admitted
Total	8442	44%	4321	25%

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

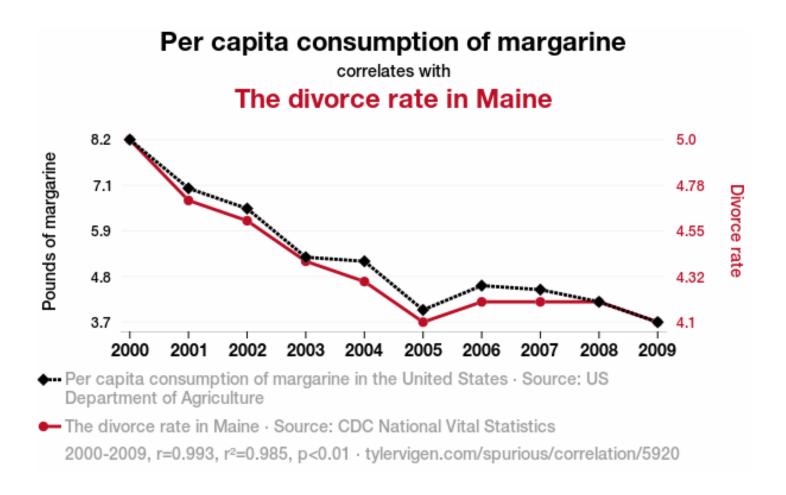
Donartment	Men		Women	
Department	Applicants	Admitted	Applicants	Admitted
A	825	62%	108	82%
В	560	63%	25	68%
С	325	37%	593	34%
D	417	33%	375	35%
E	191	28%	393	24%
F	373	6%	341	7%

SPURIOUS CORRELATIONS



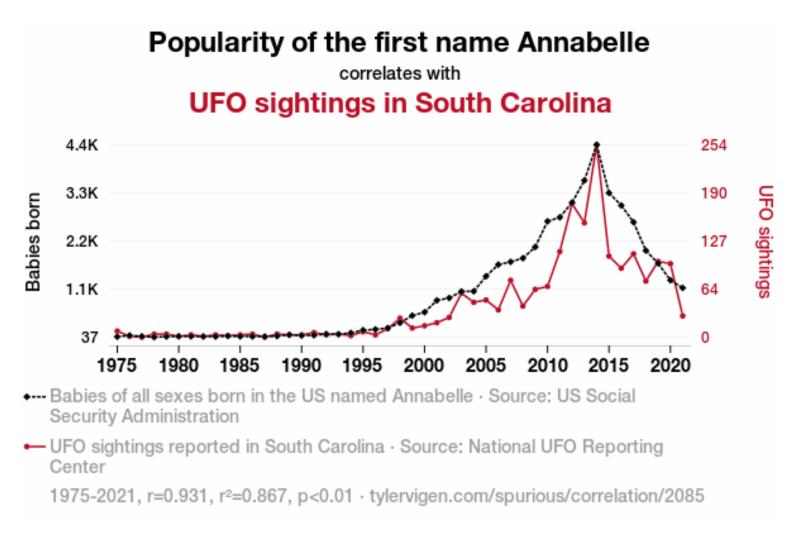


SPURIOUS CORRELATIONS



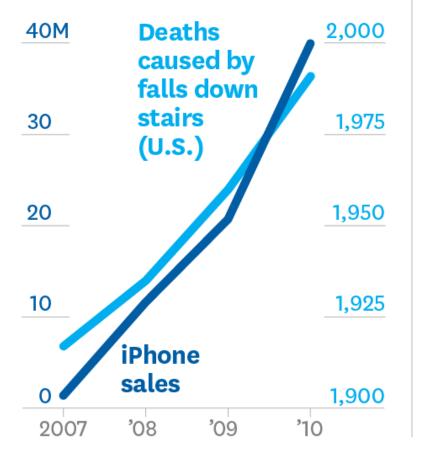


SPURIOUS CORRELATIONS

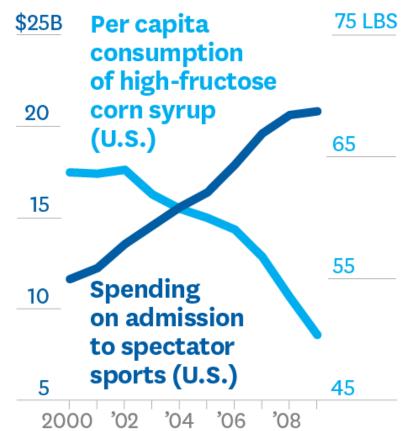




MORE IPHONES MEANS MORE PEOPLE DIE FROM FALLING DOWN STAIRS



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

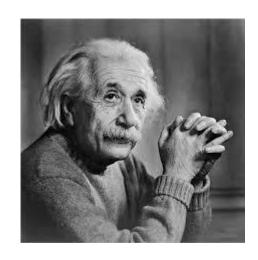
- What is the core problem?
- What processes, systems, orgs are affected?
- If solved, what is business value?
- How can problem be scoped?
- How is value measured?
- Characterize problem domain
- Is this a datadriven problem?
- What data is needed? (prelim)



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

THINKING ABOUT THE PROBLEM AND 5 MINUTES THINKING ABOUT SOLUTIONS.

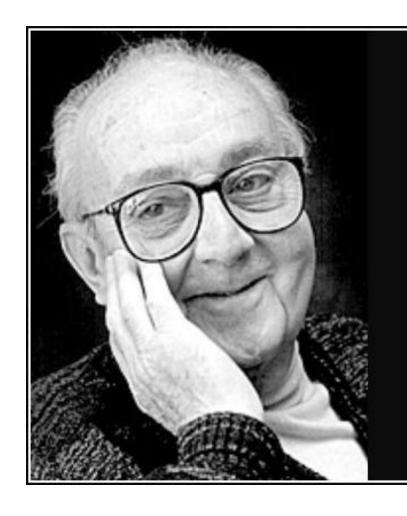




Albert Einstein



4. BUILD THE MODEL



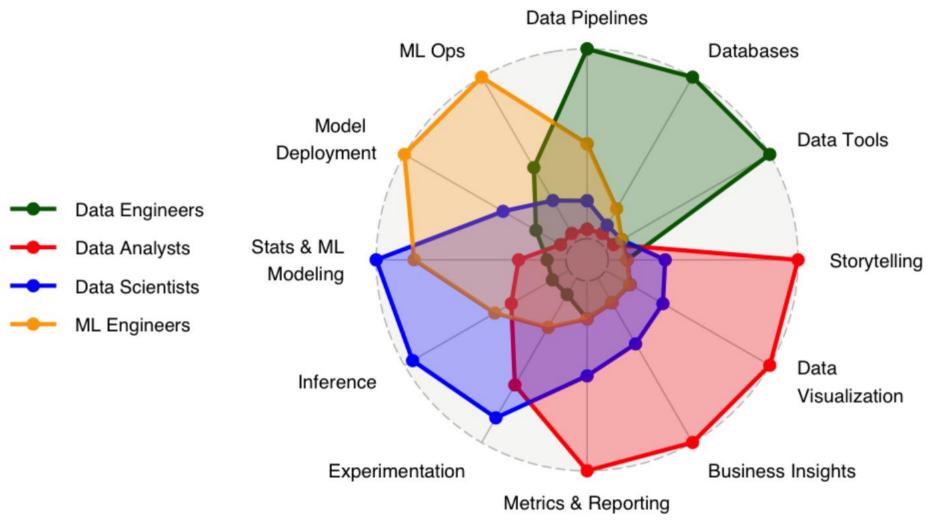
All models are approximations.
Essentially, all models are wrong, but some are useful. However, the approximate nature of the model must always be borne in mind.

— George E. P. Box —

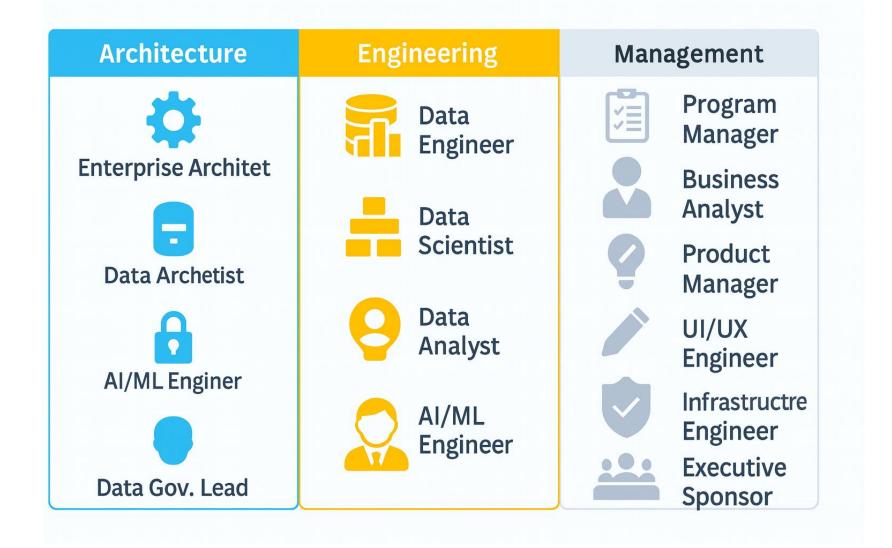
AZ QUOTES



SPIDER CHART OF RELATIVE SKILLS FOR KEY DATA ROLES



ROLES ON A DATA SCIENCE APPLICATION PROJECT

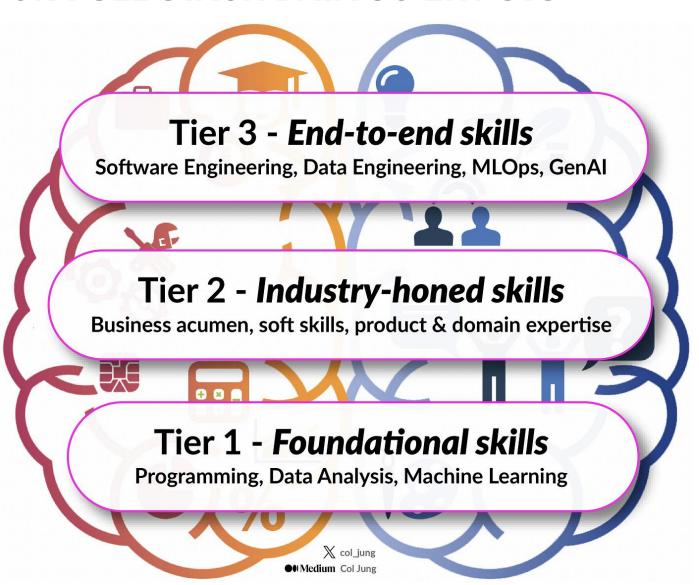


EXTENDED ROLES ON A DATA SCIENCE APPLICATION PROJECT

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.

SKILL SETS FOR FULL STACK DATA SCIENTISTS

"Evolution of Data
Science: New Age Skills for the Modern End-to-End Data
Scientist," by Col Jung, Medium, July 23, 2024.





THE GOAL: FULL-STACK DATA SCIENTIST AND BEYOND

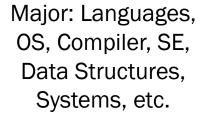
Data Engineer: ETL Pipelines, SQL, Data Prep

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

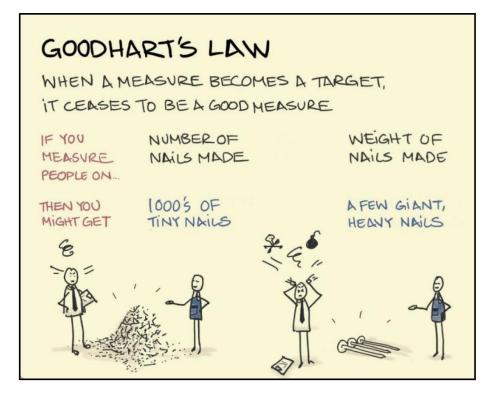
Social

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 ©

