Introduction to Big Data





Not-So-Traditionally?

Digital Transactions: Online purchases, website visits, social media interactions, financial transactions.

Sensor Data: Internet of Things (IoT) devices, GPS tracking, wearable technology.

Scientific Monitoring: Satellite imagery, climate data.

Text and Media: Online news articles, blog posts, social media streams, videos, images.



Traditionally?

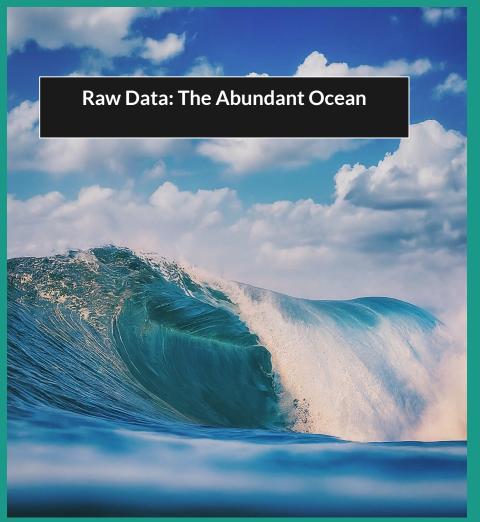
Government Records: Census data, tax records, vital records (births, deaths, marriages), trade statistics, legal documents.

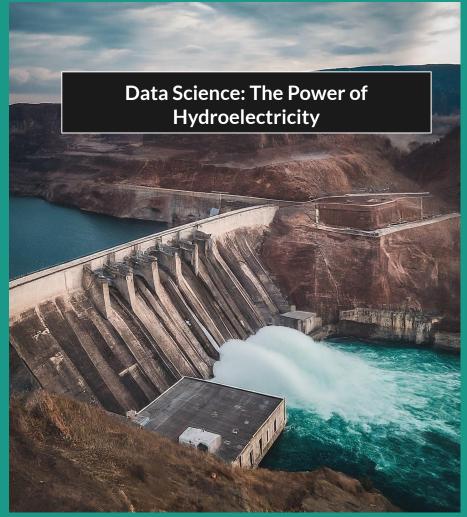
Historical Documents: Newspapers, personal diaries, business records, scientific reports.

Cultural Artifacts: Literature, artwork, archaeological findings.









Introduction to Data Science



Data science is like the science and engineering behind hydroelectricity.

Capturing and Channeling: Just like dams and reservoirs manage water flow.

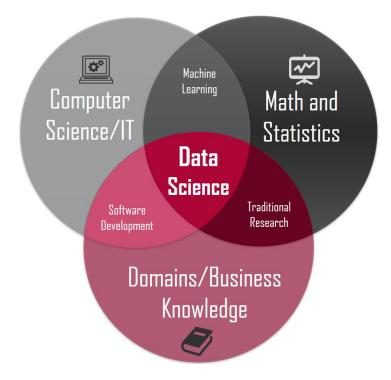
Filtering and Purification: Much like water treatment plants ensure water quality.

Generating Insights (Energy): Reveal hidden patterns – equivalent of turbines converting water flow into usable energy.

Problem-Solving Applications: Inform decision-making, just like hydroelectricity provides light, powers industries.

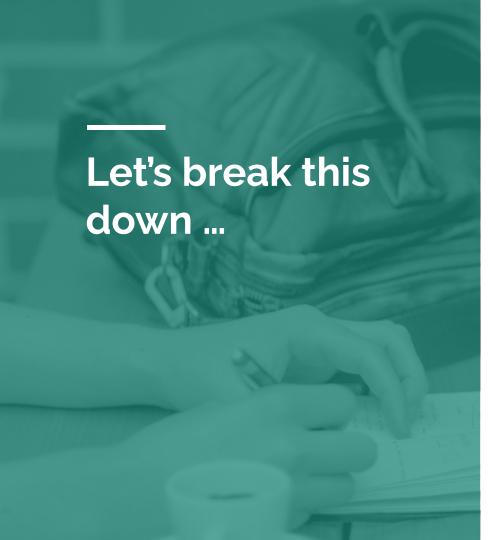


Formally -



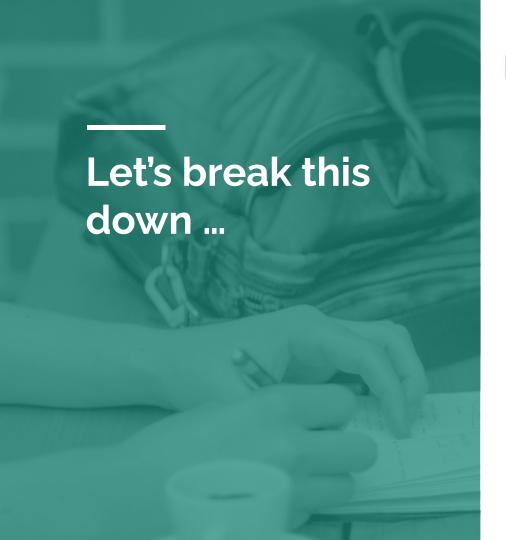
Pic Credits:

https://medium.com/@anuraggandhi29/what-is-datascience-6ac639f830c2



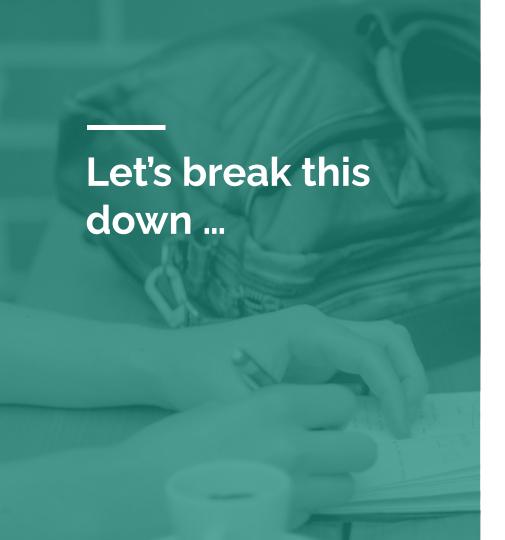
Computer Science

- Algorithms and Data Structures
- Programming
- Databases
- Machine Learning



Mathematics & Statistics

- Statistics
- Probability
- Linear Algebra
- Calculus

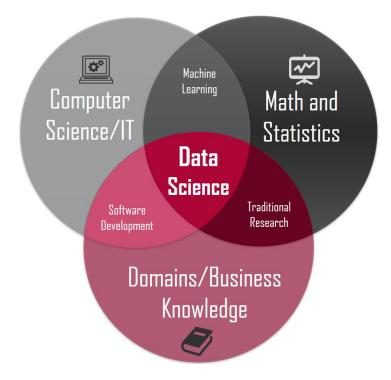


Domain Knowledge

- Understanding the Problem
- Feature Engineering
- Data Storytelling



Formally -



Pic Credits:

https://medium.com/@anuraggandhi29/what-is-datascience-6ac639f830c2

The 5 P's of Data Science

What does it take to go from raw data to useful products?



What does it take to go from raw data to useful products?

What does it take for a chef to bake a cake?

Ingredients!

What does it take to go from raw data to useful products?

What does it take for a chef to bake a cake?

The 5 P's



Purpose - Aims to bake a specific cake

People - The chef himself!

Process - A recipe.

Platforms - Kitchenware, tools (oven, mixer, whisk) etc.

Programmability - Skill - chef's mastery of techniques.

The 5 P's



Purpose - A clearly defined problem.

People - A qualified team.

Process - A well-defined workflow.

Platforms - The right platforms & tools to transform data.

Programmability - The ability to code & automate.

The 5 P's

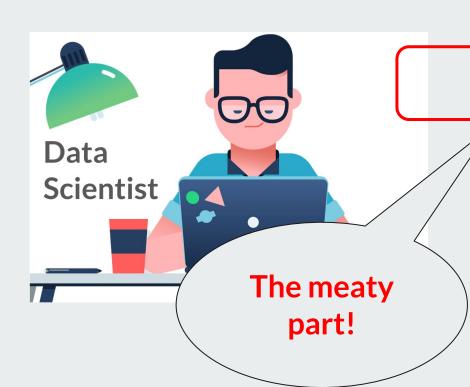
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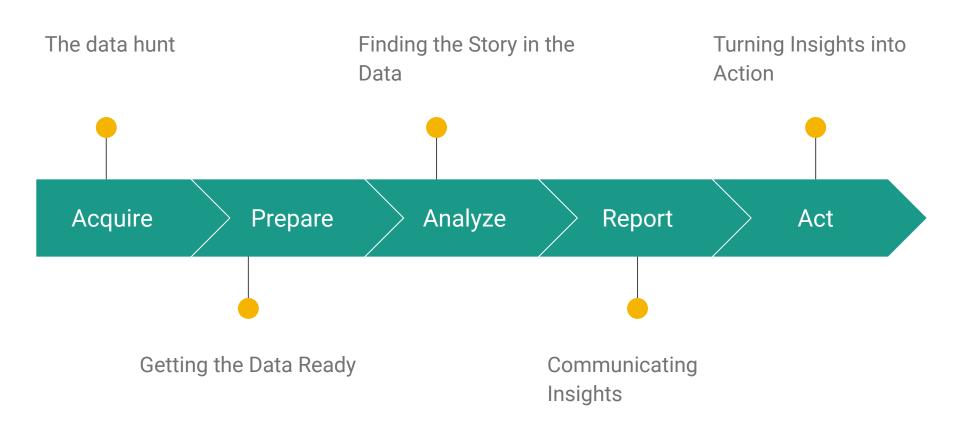
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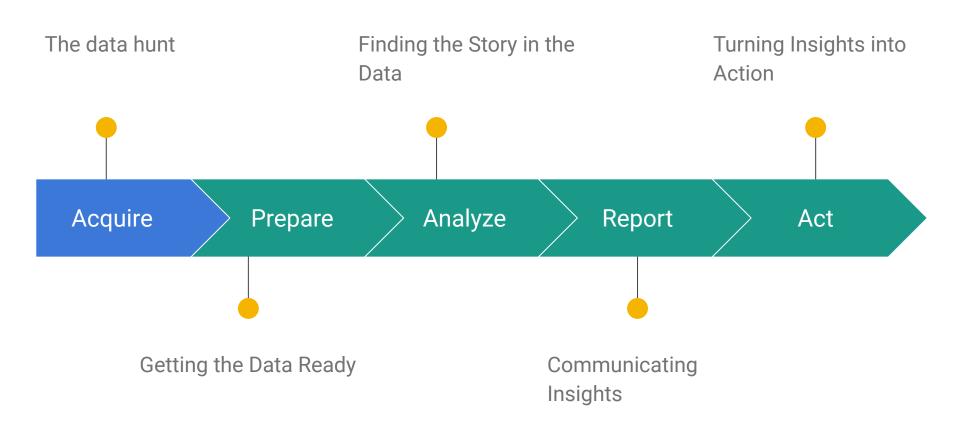
Platforms - The right platforms & tools to transform data.

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Steps in the Data Science Process







- Process of obtaining the data needed
- Variety of sources:
 - Relational Databases
 - NoSql databases
 - Text files
 - Websites



- Process of obtaining the data needed
- Variety of sources:
 - Relational Databases Use SQL
 - NoSql databases Use API & Web Services
 - Text files Scripting languages (js, python, perl, php)
 - Websites Web Services for remote data

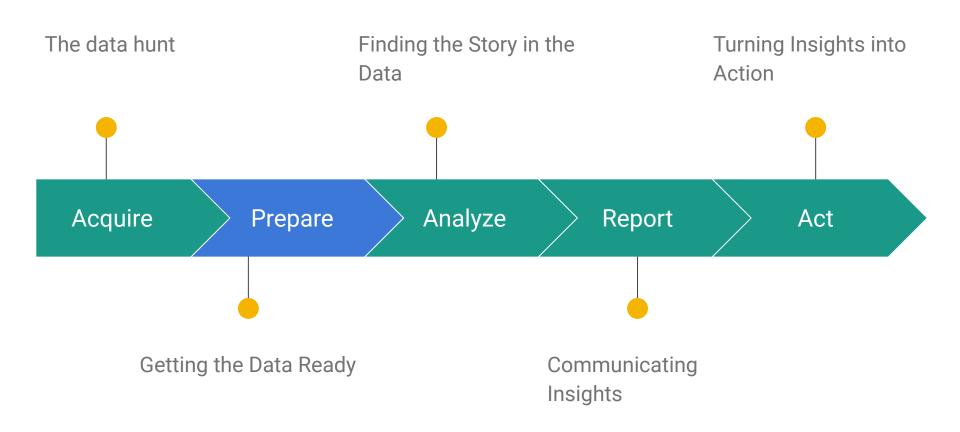
Movies

movieID	movieFullName	movieYear	movieRating	movieGenre
1	8 Mile	2002	7.2	
2	X2	2003		Action
3	Insidious	2010	6.8	Horror
4		1971	5.5	Family
5	Jumper	2008		Action
6	Shining	1980	8.4	
7		2011	7.4	Romance
8	Deadpool	2016	8.1	Action
9	Parasite	2019	8.6	
10	God Father	1972		Crime
11	Titanic	1997	7.8	Romance
12		1994	9.3	Drama

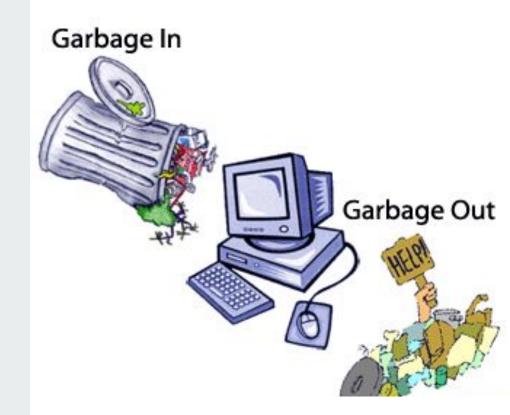
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6	Jumper Shining	G	ive me all mo	vies with ra	ting	> 8
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6		2016	ive me all mo	vies with rat	ting	> 8
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6 7 8	Shining Deadpool	2016	8.1		ting	> 8
6 7 8 9	Shining Deadpool Parasite	2016	8.1	Action	ting	> 8

	Movies			G	ive me all m	novies	with rating > 8	
	movieID	movieFullName	movie	Year	movieRa	ting	movieGenre	
	1	8 Mile	200	92	7.2			
	2	X2	200	93			Action	
	2	Theidious	201	1.0	6 9		Harrar	
SEL	SELECT movieFullName, movieYear				Che	ck <u>vis</u>	ualisation for the	query!
	FROM Movies				Check out https://animatesql.com/			
WHE	WHERE movieFullName IS NOT NULL AND				8.4			
mov	ieRating >	8.0;		20.	7.4		Romance	
	8	Deadpool	201	16	8.1		Action	
	9	Parasite	201	19	8.6			
	10	God Father	197	72			Crime	
	11	Titanic	199	97	7.8		Romance	
	12		199	94	9.3		Drama	



PrepareGetting the Data
Ready



Issues in raw data

Order ID	Customer Name	Order Date	Price (\$)	Country
12345	John Smith	2023-12-15	55.99	USA
98765	jane doe	15/12/2023	12.5	UK
12345	J. Smith	12/15/2023	55.99	US
45678	Sarah Johnson	2023-13-05	-20	Canada
33322	William Lee	null	89.99	Australia

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PrepareGetting the Data Ready

Types of Issues:

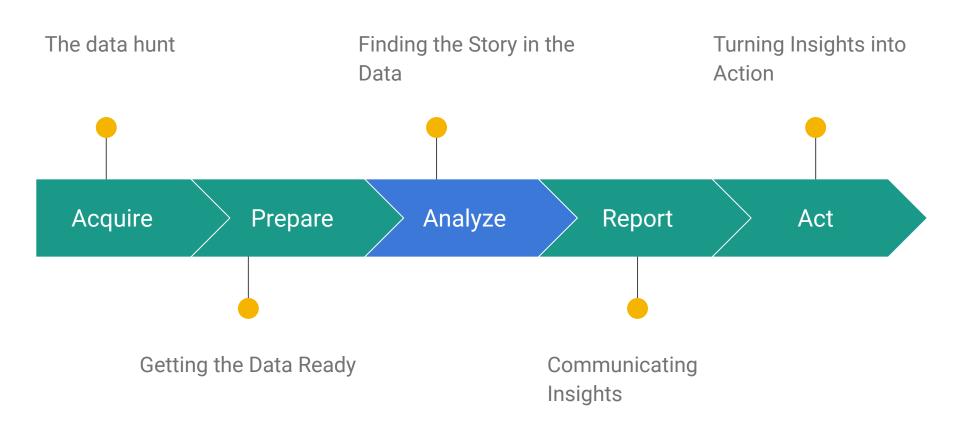
Inconsistent values: Different spellings, date formats.

Duplicate records: Identify and handle them.

Missing values: Deletion, Imputation.

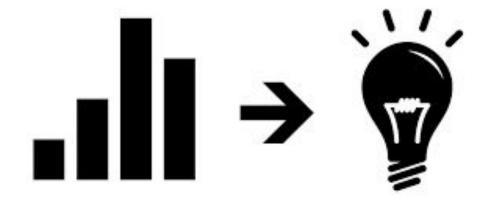
Invalid data: Out-of-range values, errors.

Outliers: Investigate if they're true errors or meaningful extremes.



Analyze

Finding the Story in the Data



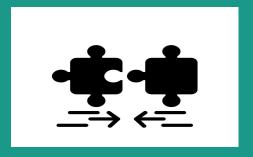
Why do you want to analyse?



What is likely to happen in the future?



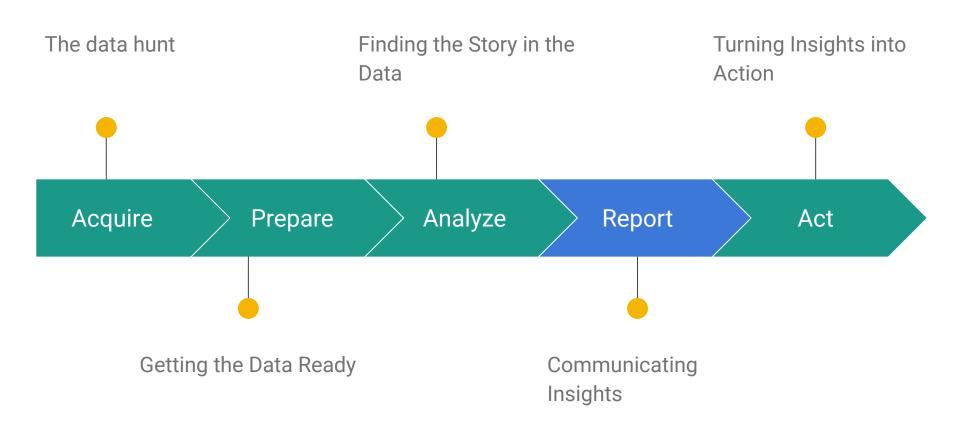
What are the natural divisions within my data?



What items or events tend to occur together?

Common Analysis Types

- Regression: Predicting future values (e.g., Sales forecasting, stock price prediction).
- Clustering: Grouping similar data points together (e.g., Customer segmentation)
- **Association Rule Mining**: Finding patterns of co-occurrence (e.g., "Customers who bought X also bought Y").
- Classification: Predicting categories (Email: spam/not spam).

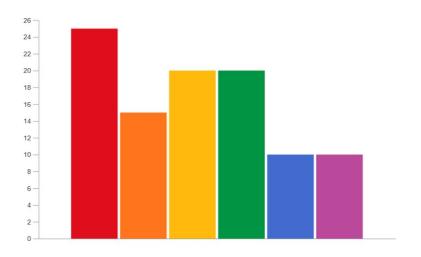


Report Communicating Insights



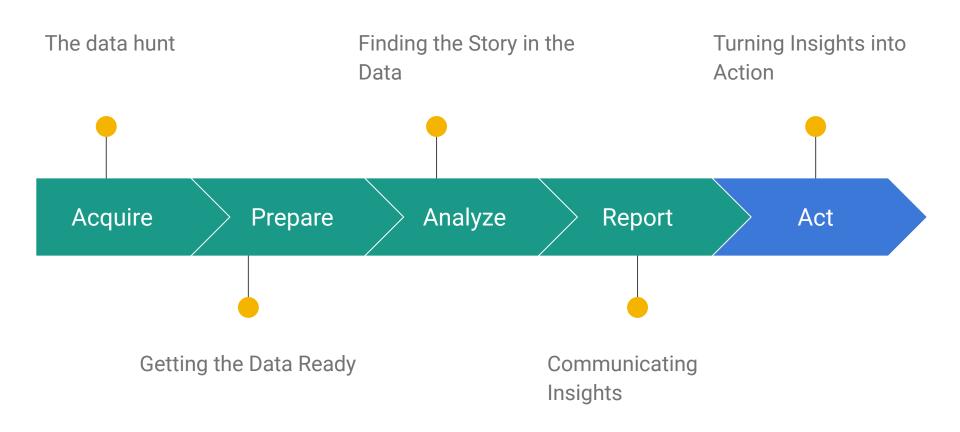
Report - Communicating Insights

- Data alone doesn't create change.
- Presentation is important!









Act

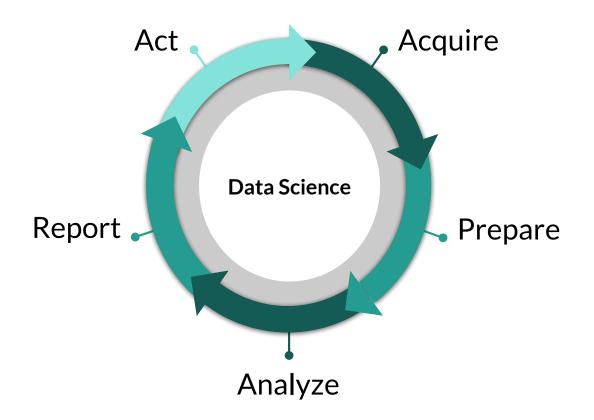
Turning Insights into Action



Act - Turning Insights into Action

- Goal: Use data-driven insights to inform decisions that improve our business, research, or outcomes.
- Decision-making isn't the end Feed new data back into the process.

Takeaway - Data science process is a continuous cycle!



The data science techniques we've discussed work well with datasets that fit on a single machine.

But what happens when the data explodes?

The data scient discussed work on a single material

But what happoexplodes?

Volume

Variety

Speed of information

Big Data!