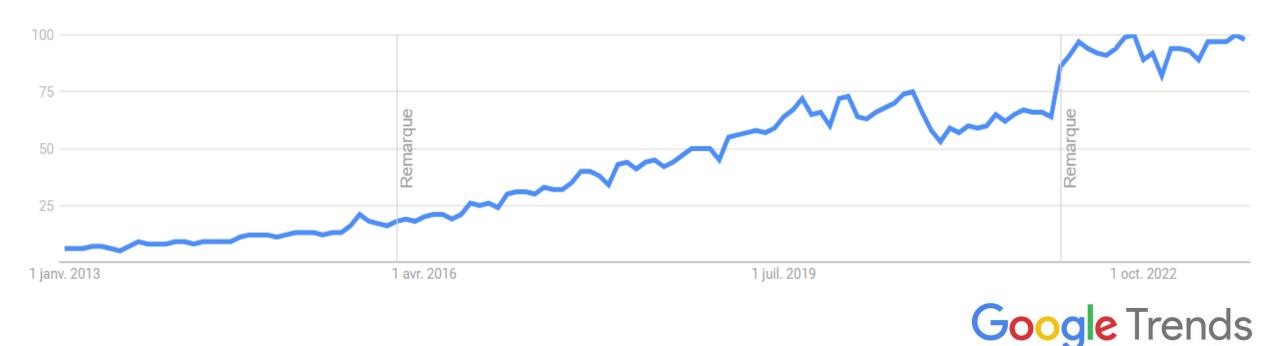
# Introduction to Data Science

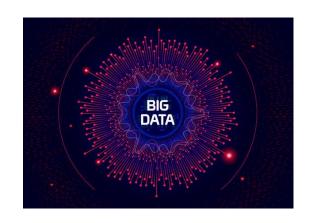


# Data science: Lots of hype recently!

Interest over time. Web Search for « Data science » in the past 10 years.

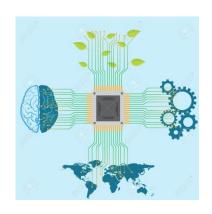


### Why the hype arround Data Science?





- Unprecedented Data
  Generation Rates
- Need for Advanced Analytical Tools



**Technological Advancements** 

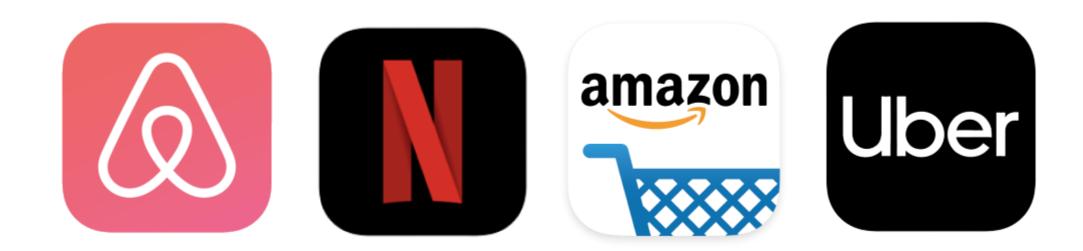
- High-Performance Computing
- Advanced Machine Learning Algorithms



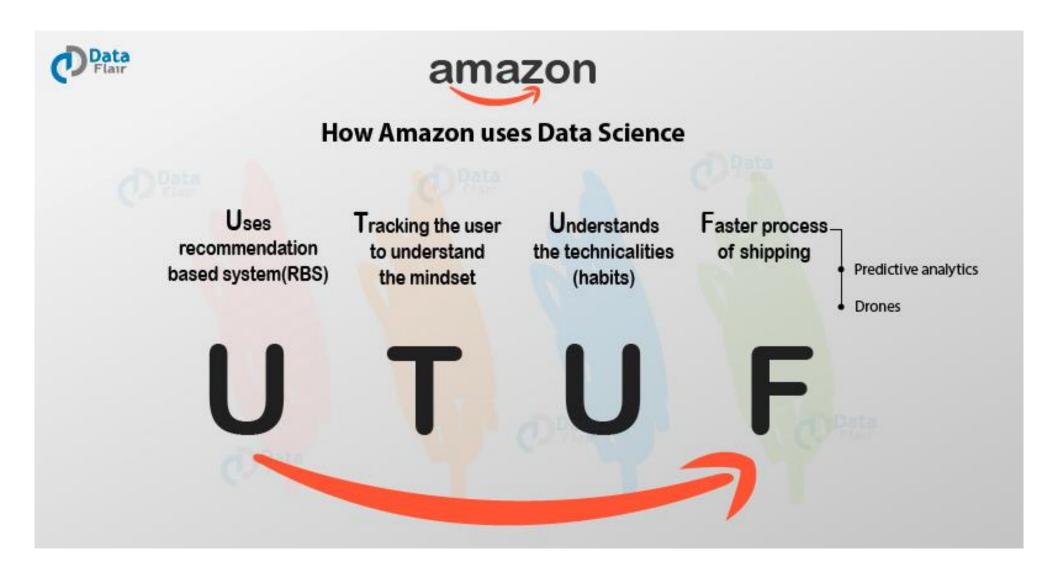
**Real-World Impact** 

- Tangible Benefits in Various Industries
- From Healthcare to Finance to E-Commerce

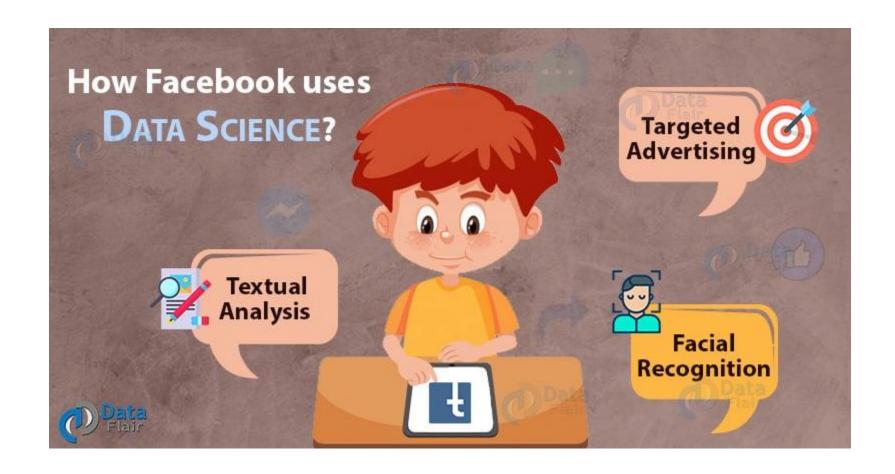
### Some success stories...



# Amazon – Transforming E-commerce with Data Science



# Facebook – Using Data to Revolutionize Social Networking & Advertising



### Uber – Using Data to Make Rides Better

### **Data Science in Uber**



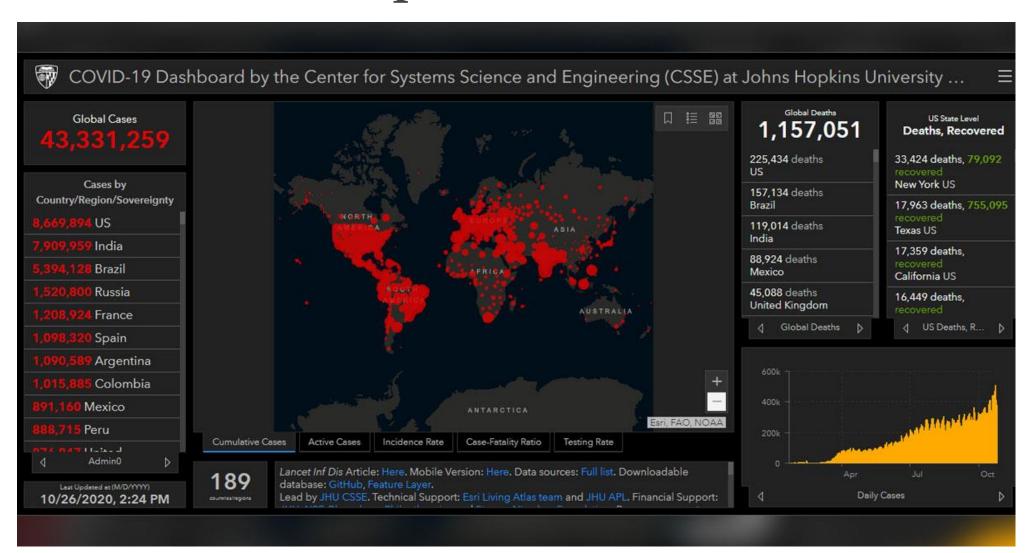
# Spotify – Revolutionizing Music Streaming

#### **Features of Spotify**





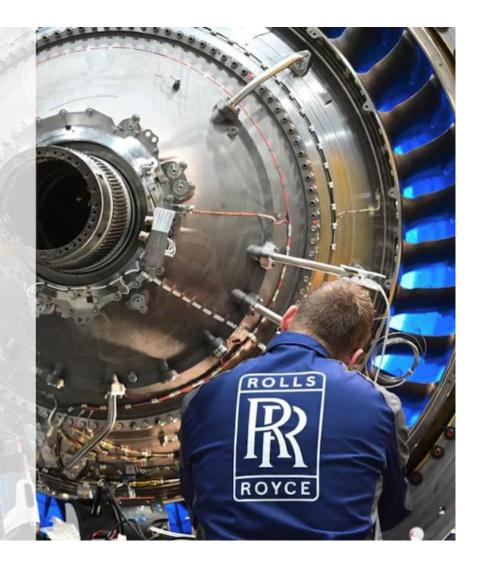
# Covid-19 – Using Data to Track the Pandemic and Forecast Hotspots



# Rolls-Royce – Data science in manufacturing

### How Rolls-Royce 100x'ed the speed of their engineering design processes

By leveling up their data skills, Rolls-Royce could identify and automate manual processes and save its engineers time to work on more valuable initiatives. By increasing the skills of both technical and non-technical roles, Rolls-Royce was able to help everyone better improve their Python, Power Bl, and general data literacy.



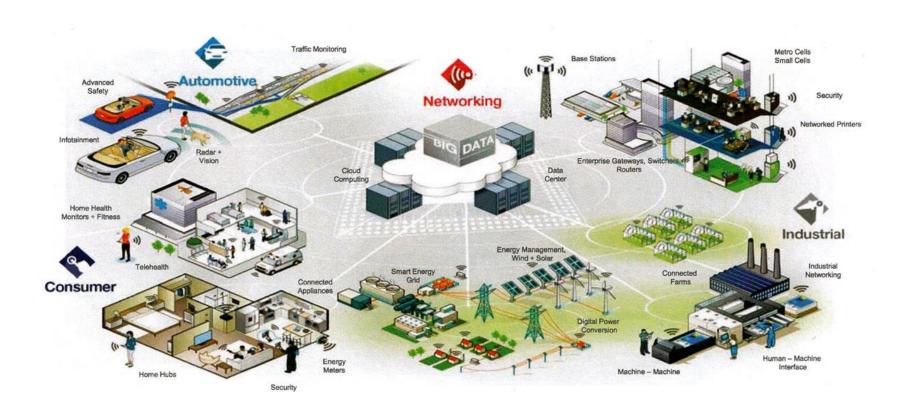
# The power of data



- "In God we trust. All others must bring data."
- "Without data you're just another person with an opinion."

Dr. W. Edwards Deming

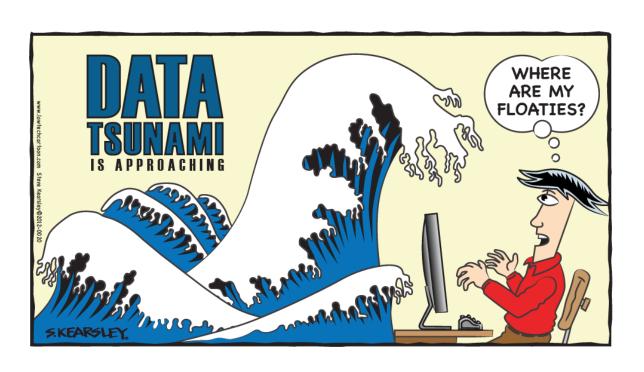
### Where does data come from?

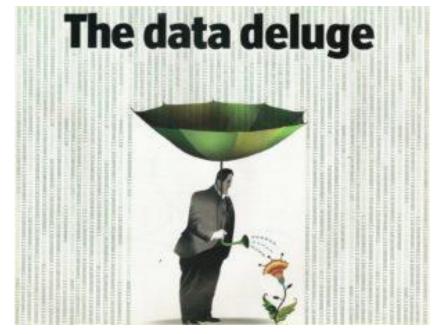




The world is digital: Every interaction, transaction, and communication now leaves a digital footprint.

# The tsunami of data keeps growing!





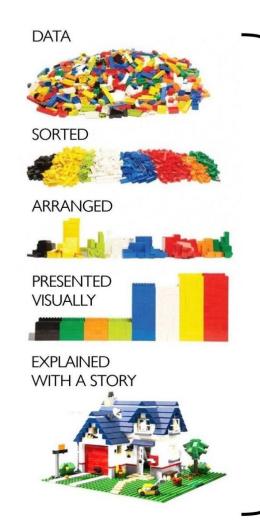
According to the latest estimate, approximately 328.77 million terabytes of data are created each day in 2023; 181 zettabytes of data will be generated in 2025

# Drowning in data but starving for insights

"Data are just summaries of thousands of stories—tell a few of those stories to help make the data meaningful."

~ Dan Heath,

 Data itself is useless. It's only when you analyze it, and make it actionable that it becomes valuable.

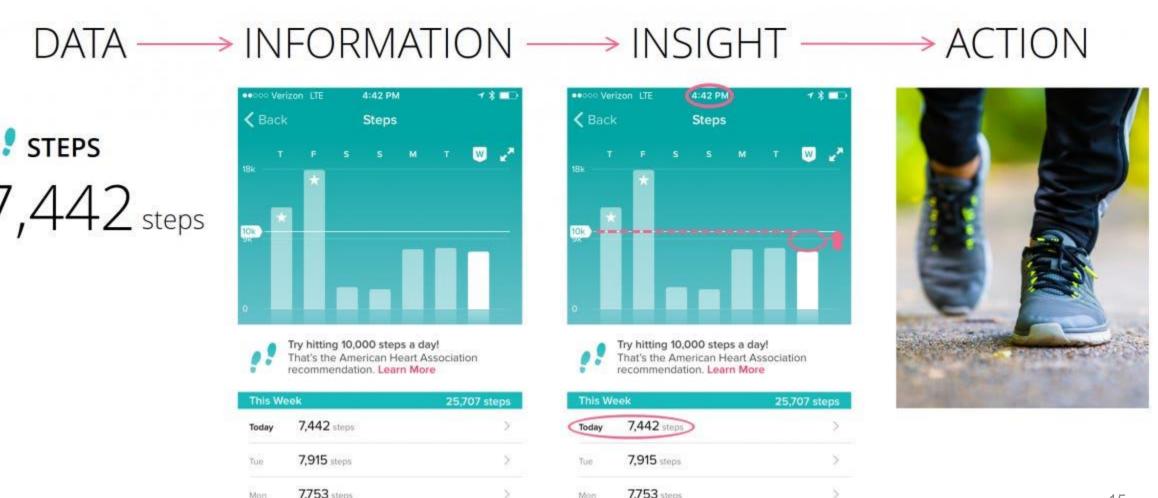


The real value add is in the storytelling to make the data comprehensible and actionable by humans.

Terence Kawaja

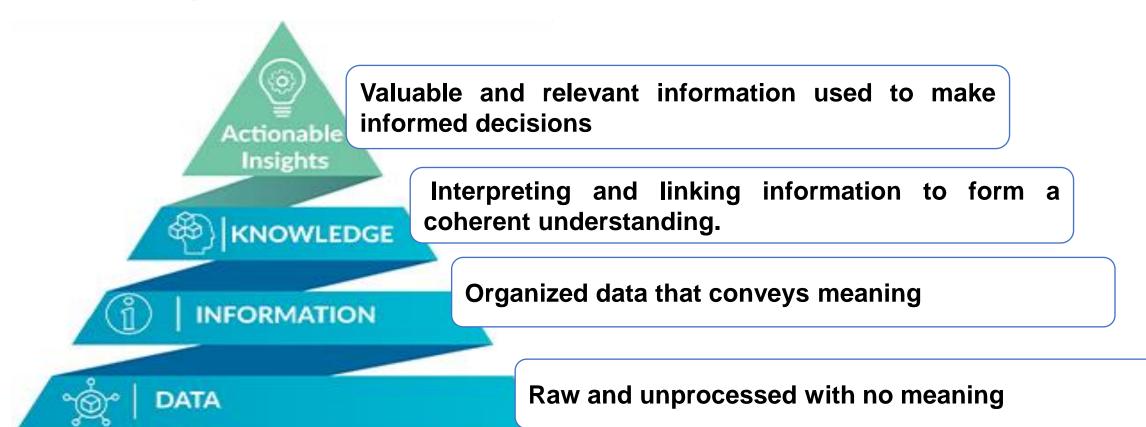


# Drowning in data but starving for insights

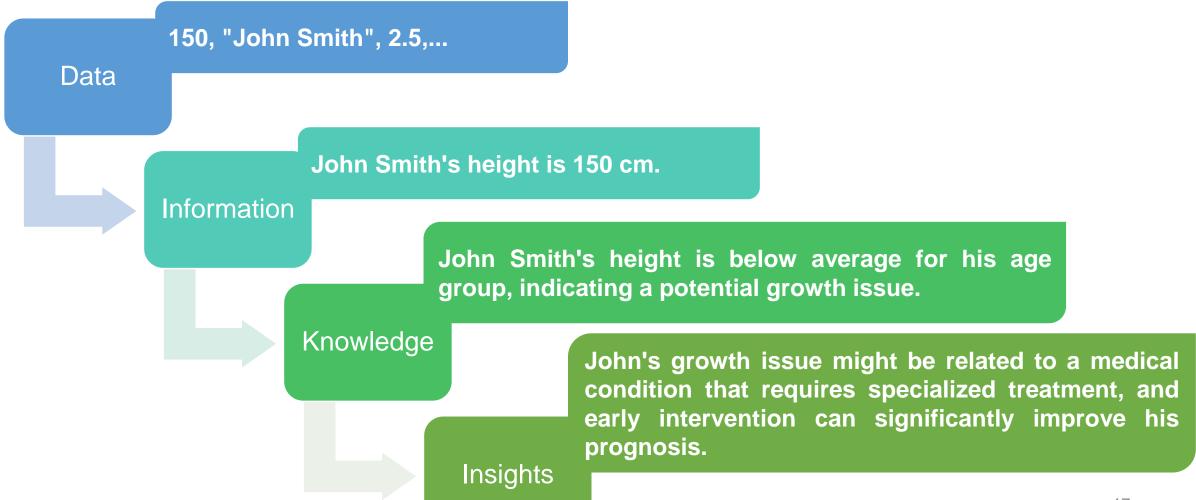


# Towards actionable insights

### The insight pyramid



# Data, Information, Knowledge and Insights



# The quest for Meaning: from data to insights

• Data science is the key to making sense of this digital deluge.

• It is the art of transforming data into actionable insights and valuable knowledge.



### Data science is multidisciplinary

#### Math & Statistics

The mathematical theories and tools are the foundation of data science.

#### **Programming Skills**

Software tools and libraries are necessary to have something done.

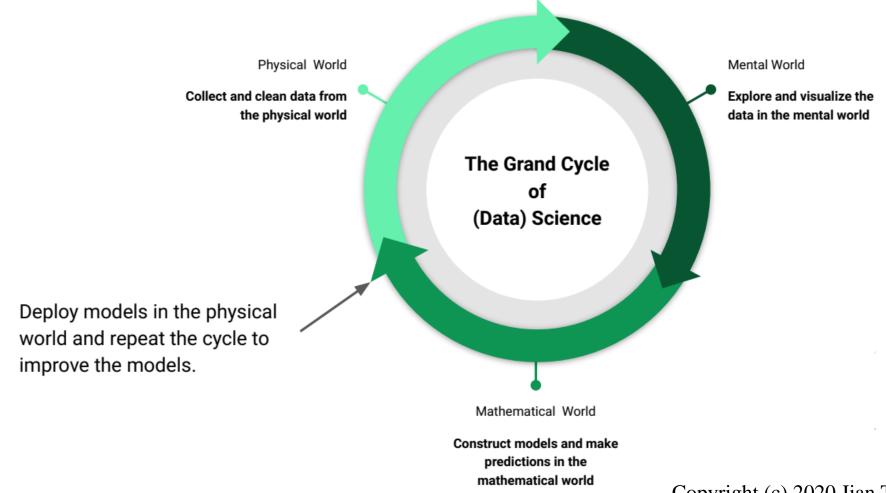
#### **Domain Knowledge**

Domain knowledge is essential to identify and explore questions.

#### **Data Science**

Data science is an interdisciplinary field about processes and systems to extract knowledge or insights from data in various forms, either structured or unstructured.

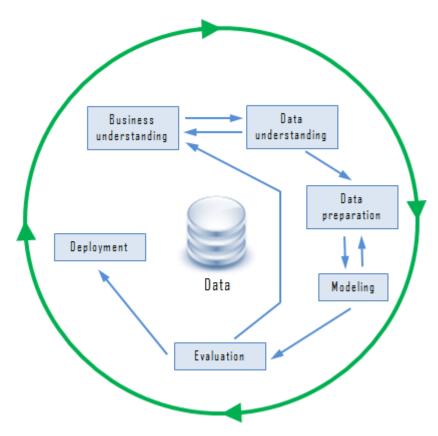
# The data science cycle



### Data science is a process

#### **CRISP-DM**

- Cross-industry standard process for data mining
- An open standard process model that describes common steps followed for data mining and analytics projects
- It is the most widely-used analytics model

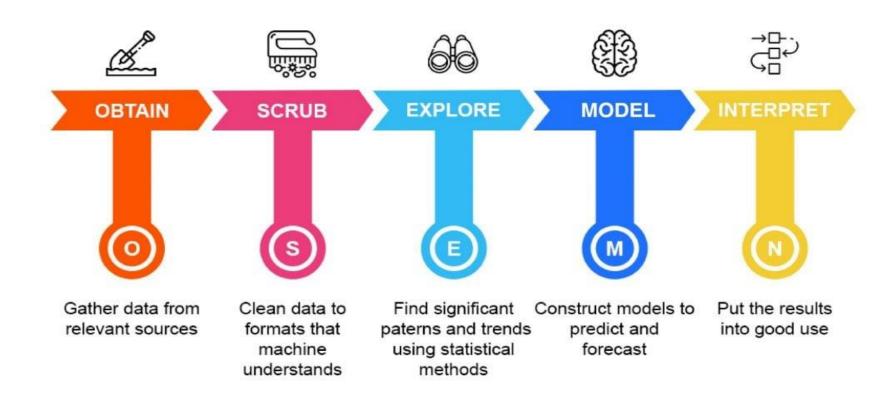


Cross-industry Standard Process for Data Mining (CRISP-DM)

### The OSMEN ("Awesome") DS process!



#### DATA SCIENCE PROCESS



# Prerequisite step: Business understanding

"Far better an approximate answer to the right question, which is often vague, than the exact answer to the wrong question, which can always be made precise."

— John Tukey

What is the Business Question we are trying to answer?

Asking the right question sets up the rest of the path.

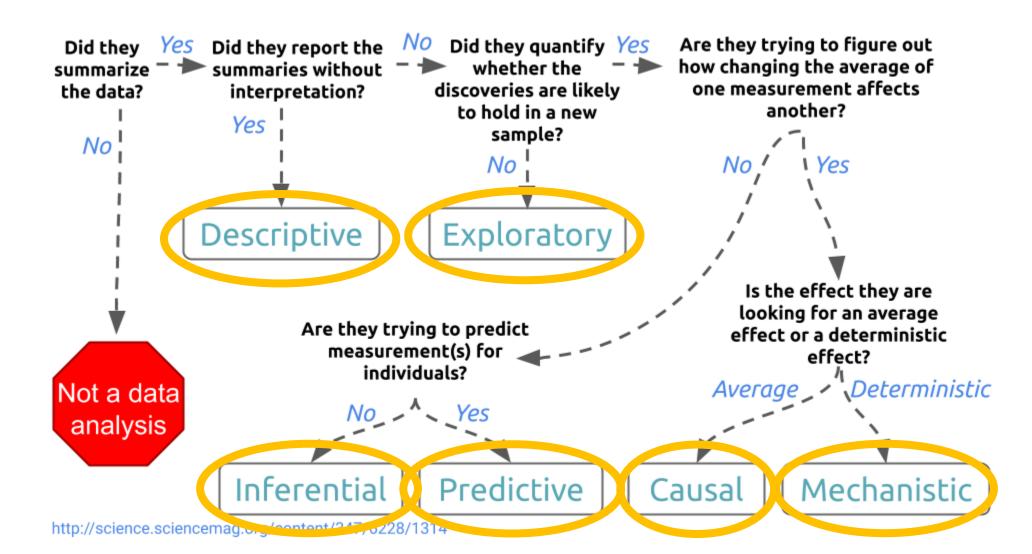
# Questions are important!

"Good data science is more about the questions you pose of the data rather than data munging and analysis." — Riley Newman



- First, formulate the questions that you will use the data to solve.
- The more questions you ask of the data, the more insight you will get.

# Types of Data Science questions



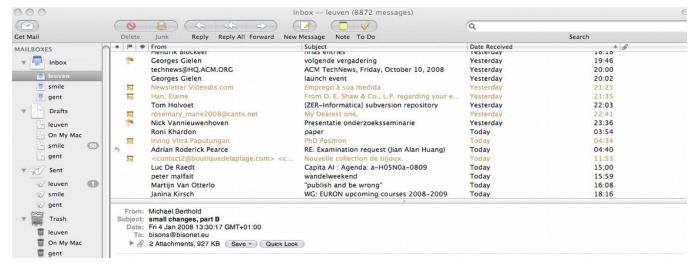
# Types of Data Science questions

- Descriptive: Describe and summarize data (e.g. Covid-19 statistics)
- Exploratory: Search for unknown relationships, new discoveries, without identifying the cause
- Inferential: What data say about larger population..
- Predictive: use Historical data to make predictions for the future
- Causal: Explore causation; what happens to variable X when variable Y change.
- Mechanistic: Determine governing principles; what exact changes in variables lead to changes in other variables.

### Example of Data science workflow

Business question





SPAM email reduces productivity, automatically remove it

How can we effectively distinguish between legitimate and spam emails?

### Obtaining data

- Collect messages, in general and from the user, that are spam (negative) and legitimate (positive): acquisition, annotation, ...
- Given a text message, predict whether it is spam or not
- → text categorization, useful in general
- $\rightarrow$  we want a <u>function</u> from message to  $\{0,1\}$
- → is called binary classification problem

### Scrubing data

Given a raw text, convert string data into numerical data one

- Bag of words, TFIDF, Word2Vec

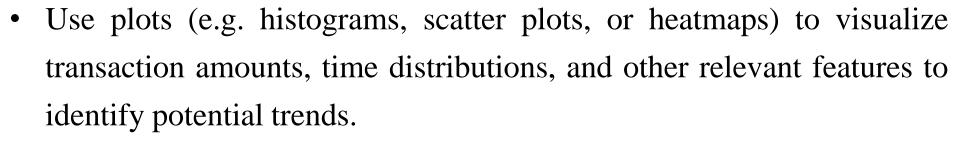
#### **Text Preprocessing**

- 1. Remove Noisy Data: header, footer, HTML, XML, markup data
- 2. Tokenization: word, character, and subword (n-gram characters)
- 3. Normalization: converting all words to lowercases



### • Exploring the cleaned data (EDA)

- Explore to understand
  - the distribution of features,
  - correlations, and patterns related to transactions.



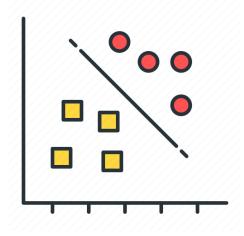


### Modeling

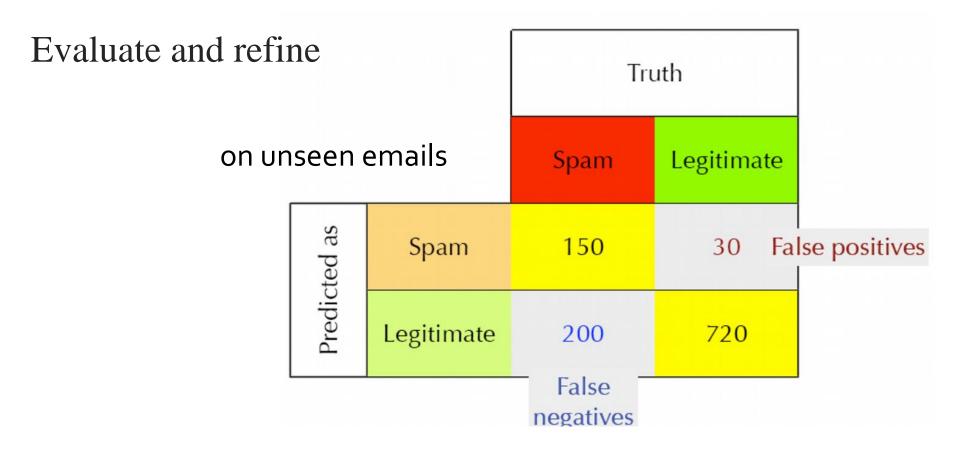
• We could write a rule-based system, such as

if Title.contains("YOU HAVE WON!!!") then return Spam

- train a classifier (e.g. naïve bayes)
- Does it work well?  $\rightarrow$  evaluate



### Modeling



### • Interpreting (Data Storytelling)

Data-driven battle against spam

- Data, the weapon that safeguard inboxes

#### Display:

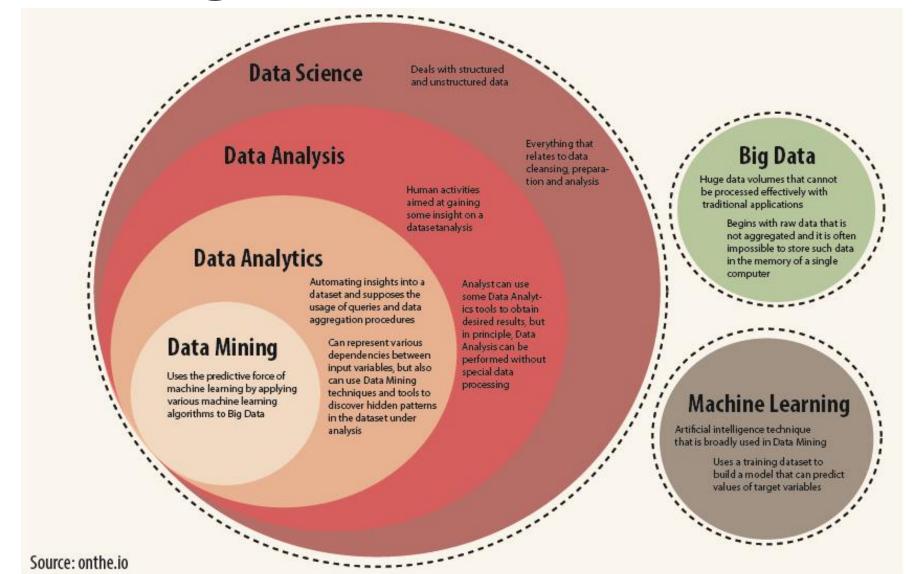
- a word cloud showcasing prevalent words in spam emails.
- bar charts or histograms depicting the frequency of specific terms or patterns in spam versus non-spam emails.
- Confusion matrix or ROC curve illustrating the model's performance in detecting spams.



### DS in summary,

- Form your **Business Problem**
- Obtain your data
- Clean your data,
- Explore your data with visualizations,
- Model your data using statistical or machine learning models,
- iNterpret your analysis results,

# Data Mining & Data Science



# Data Mining & Data science

	Data Mining	Data Science
01	More involved with its processes	Broadly focuses on the science of data
02	Primarily used for business purposes	It is essentially implemented for scientific purposes
03	Data mining is a technique that is a part of the KDD process	Data science is a field of study
04	Primarily deals with structured data	It deals with all types of data - structured, unstructured, or semi-structured
05	It is about extracting valuable information from data	It is about collecting, & processing, analyzing & utilizing data in various operations
06	It is a subset of data science as mining activities are in the pipeline of data science	Involves data scraping, cleaning, visualization, stats, etc. Therefore, it is a superset of data mining
07	Its objective is to realize the value of data & make it usable by extracting important Info.	Objective is to build data- dominant products for a venture

### Course Content

#### • Data Wrangling and Cleaning

- Data collection and preprocessing
- Handling missing data and outliers
- Data transformation and feature engineering

#### • Exploratory Data Analysis (EDA)

- Descriptive statistics and summary metrics
- Correlation analysis and dimensionality reduction

#### Classification and prediction

- Supervised learning : SVM...
- Ensemble learning
- Evaluation Metrics for Classification Models
- Model Overfitting and regularization