

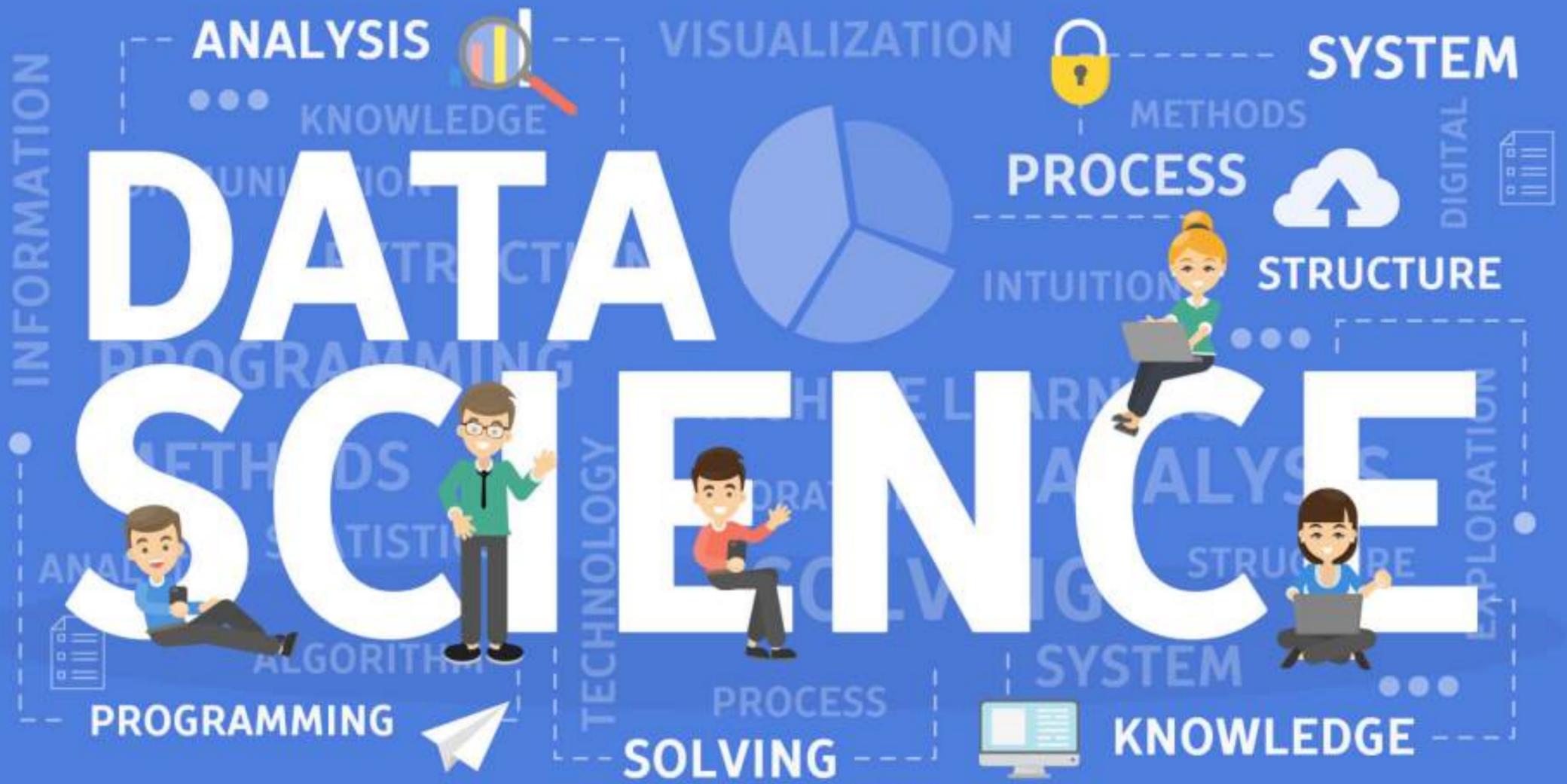
# [IF-5-OT7:TD] Foundation of data engineering

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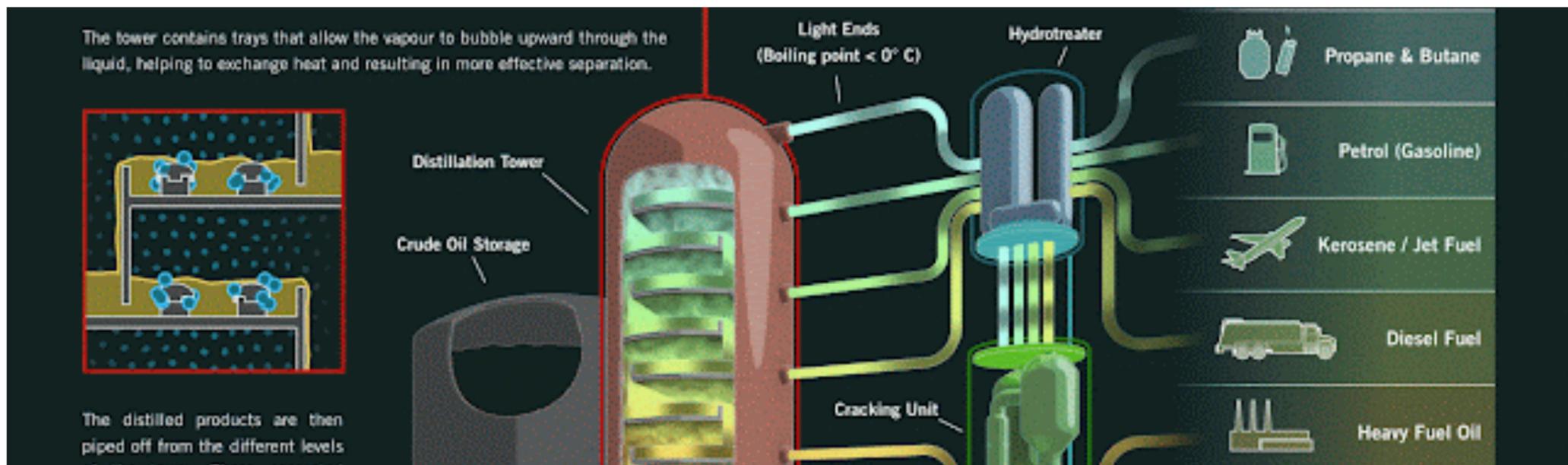


## Quote

“A scientist can discover a new star, but he cannot make one.  
He would have to ask an engineer to do it for him.”

– *Gordon Lindsay Glegg*





...refining crude oil

<sup>01</sup> [Source](#)

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- Discovering what we don't know from data
- Obtaining predictive, actionable insight from data
- Creating Data Products that have business impact now
- Communicating relevant business stories from data
- Building confidence in decisions that drive business value

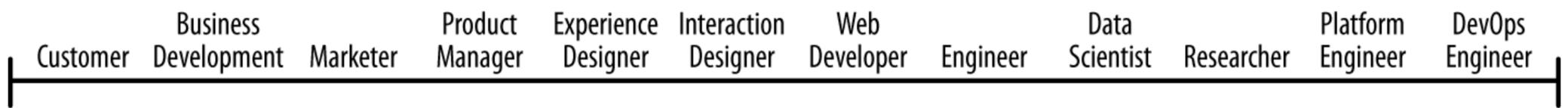
# Data Engineering is...



...build the refinery.

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# Roles in a Data Science Project<sup>02</sup>



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<sup>02</sup> <http://emanueledellavalle.org/slides/dspm/ds4biz.html#25>

# Roles in a Data Science Project<sup>02</sup>



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<sup>02</sup> <http://emanueledellavalle.org/slides/dspm/ds4biz.html#25>

# Data Engineer!

## The Data Engineer

A dedicated specialist that maintain data available and usable by others (Data Scientists).<sup>03</sup>

Data engineers set up and operate the organization's data infrastructure preparing it for further analysis by data analysts and scientists.<sup>03</sup>

Data engineering field could be thought of as a superset of business intelligence and data warehousing that brings more elements from software engineering.<sup>04</sup>

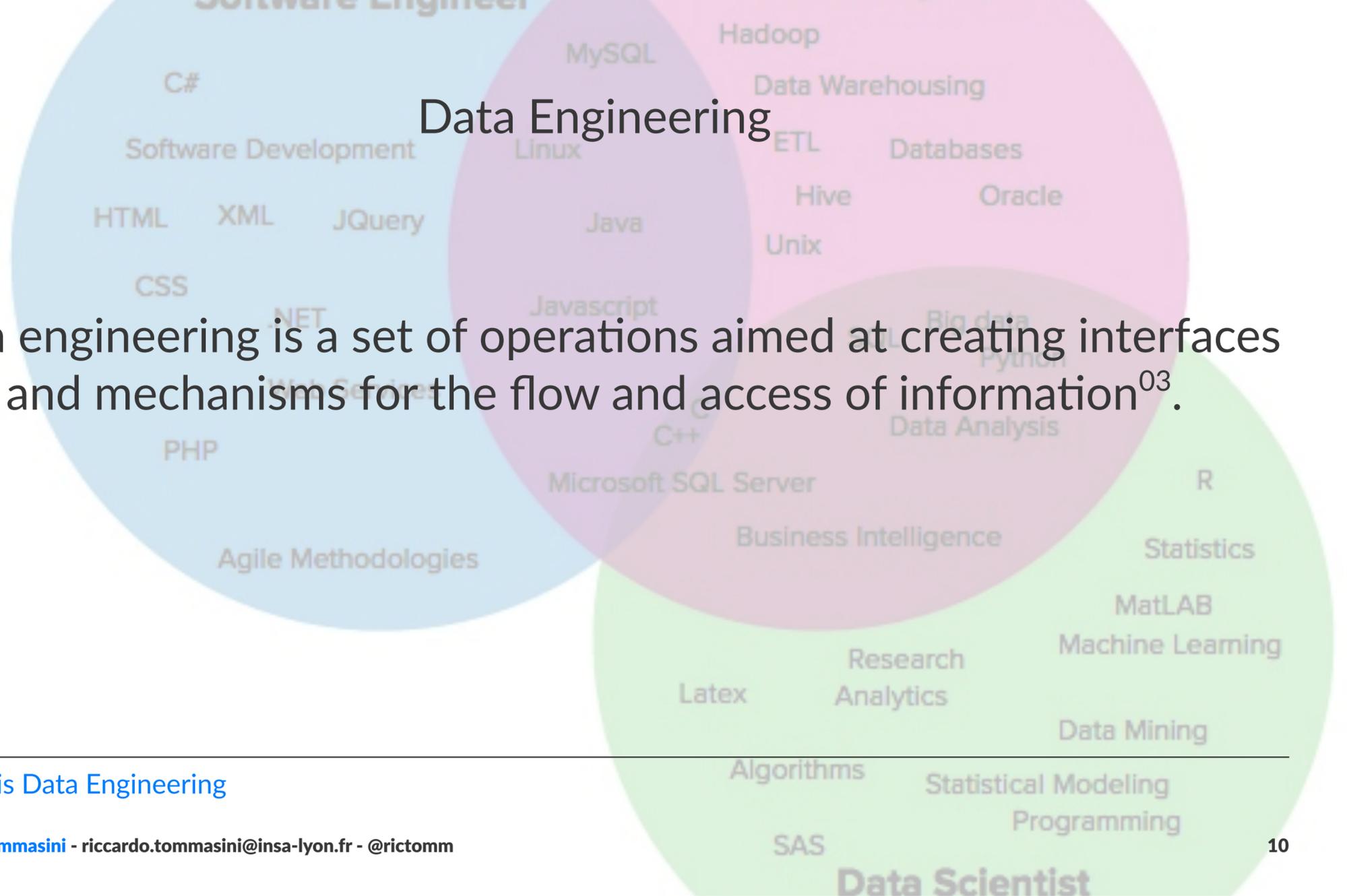
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<sup>03</sup> [What is Data Engineering](#)

<sup>04</sup> [Source: The Rise of Data Engineer](#)

# Data Engineering

Data engineering is a set of operations aimed at creating interfaces and mechanisms for the flow and access of information<sup>03</sup>.



<sup>03</sup> [What is Data Engineering](#)

↻ You Retweeted



**Seth Rosen** @sethrosen · Apr 20

Them: Can you just quickly pull this data for me?

Me: Sure, let me just:

```
SELECT * FROM  
some_ideal_clean_and_pristine.table_that_you_think_exists
```

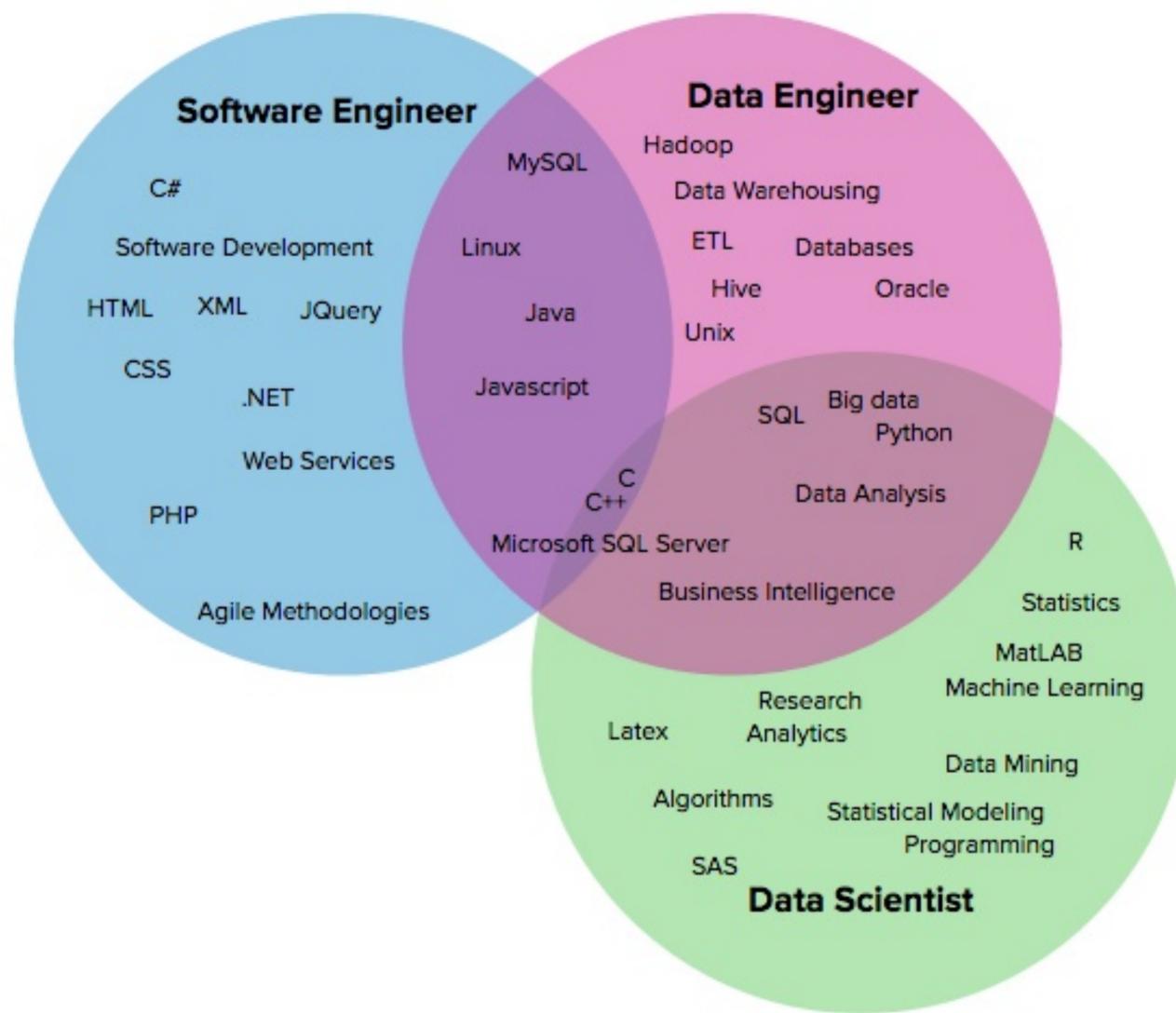
💬 323

↻ 4.4K

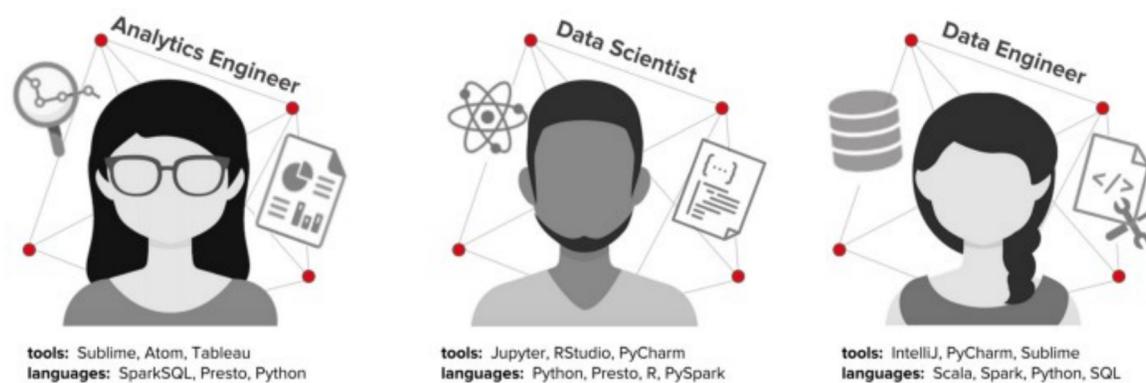
❤️ 28K



[Show this thread](#)



## Netflix's Perspective<sup>05</sup>



- a data engineer might create a new aggregate of a dataset containing trillions of streaming events
- analytics engineer might use that aggregate in a new report on global streaming quality
- a data scientist might build a new streaming compression model reading the report

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<sup>05</sup> [Netflix Innovation](#)

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each of these workflows has multiple overlapping tasks:

# Data Bricks

**The ML Team**

- Subject Matter Experts (SMEs)
  - Deep understanding of business problems
  - Need to be integrated into the ML lifecycle
- Data Scientist
  - Needs modeling skills
  - More importantly need communication skills!
- Data Engineer
  - Manages the efficient flow of data
- ML Engineer (can also be split between ML Architects/ DevOps)
  - Design, manage and scale the ML infrastructure (ML Architect)
  - Deploy and operationalize the models, ensure availability (DevOps)

Diagram labels: Framing the problem, Operationalize the model

databricks

# Google's Two-Cents

## Professional Data Engineer

A Professional Data Engineer enables data-driven decision making by collecting, transforming, and publishing data. A Data Engineer should be able to design, build, operationalize, secure, and monitor data processing systems with a particular emphasis on security and compliance; scalability and efficiency; reliability and fidelity; and flexibility and portability. A Data Engineer should also be able to leverage, deploy, and continuously train pre-existing machine learning models.

The Professional Data Engineer exam assesses your ability to:

- ✓ Design data processing systems
- ✓ Build and operationalize data processing systems
- ✓ Operationalize machine learning models
- ✓ Ensure solution quality

[Register](#)

[FAQs](#)

This exam is available in English and Japanese.

# The Knowledge Scientist<sup>06</sup>



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<sup>06</sup> [The Manifesto](#)

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# Philosophy of (Data) Science<sup>07</sup>



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<sup>07</sup> Data as Fact

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Nowdays we deal with a number of data from different domains.

# What is Data?



# Oxford Dictionary

*Data [**uncountable, plural**] facts or information, especially when examined and used to find out things or to make decisions.*<sup>08</sup>

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<sup>08</sup> Def

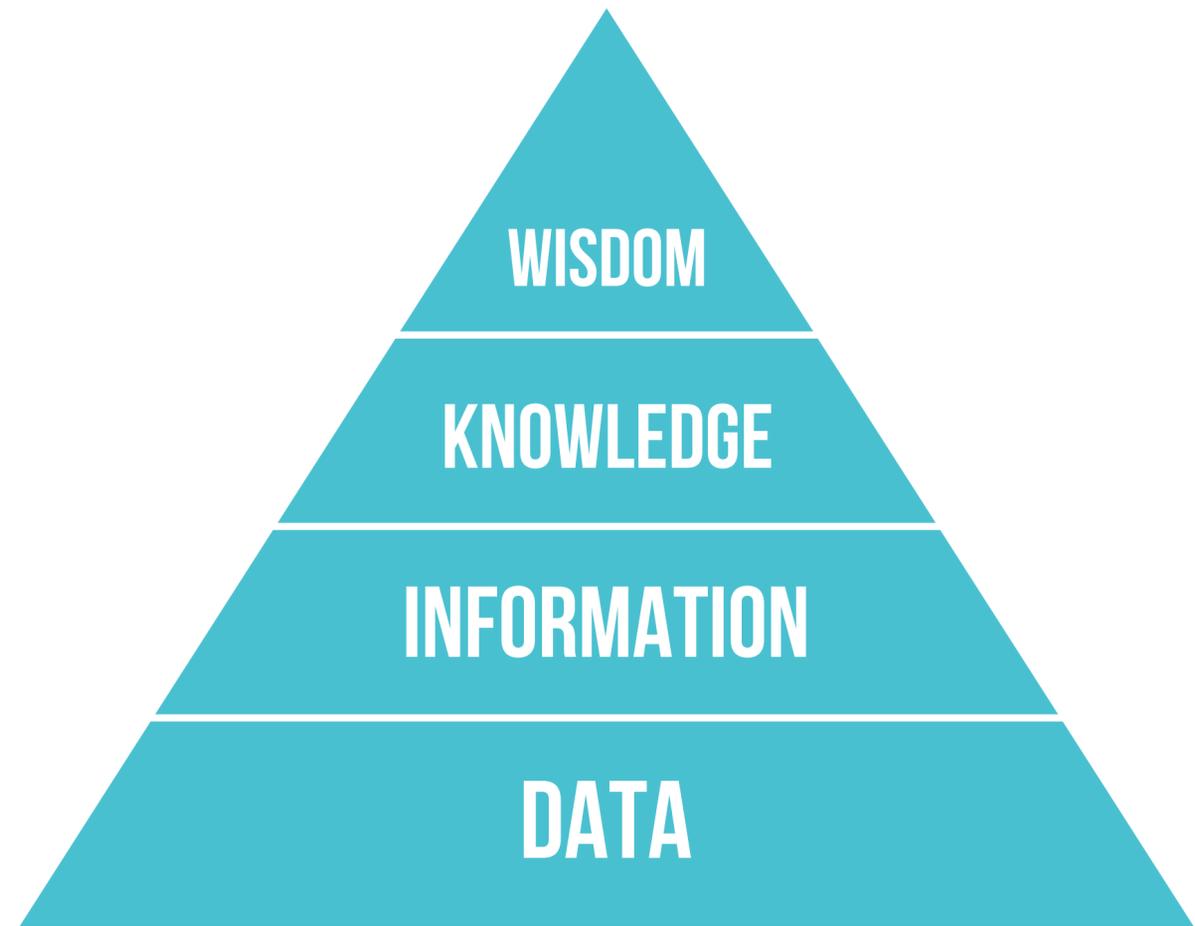
# Wikipedia

Data (treated as singular, plural, or as a mass noun) is any sequence of one or more symbols given meaning by specific act(s) of interpretation <sup>09</sup>

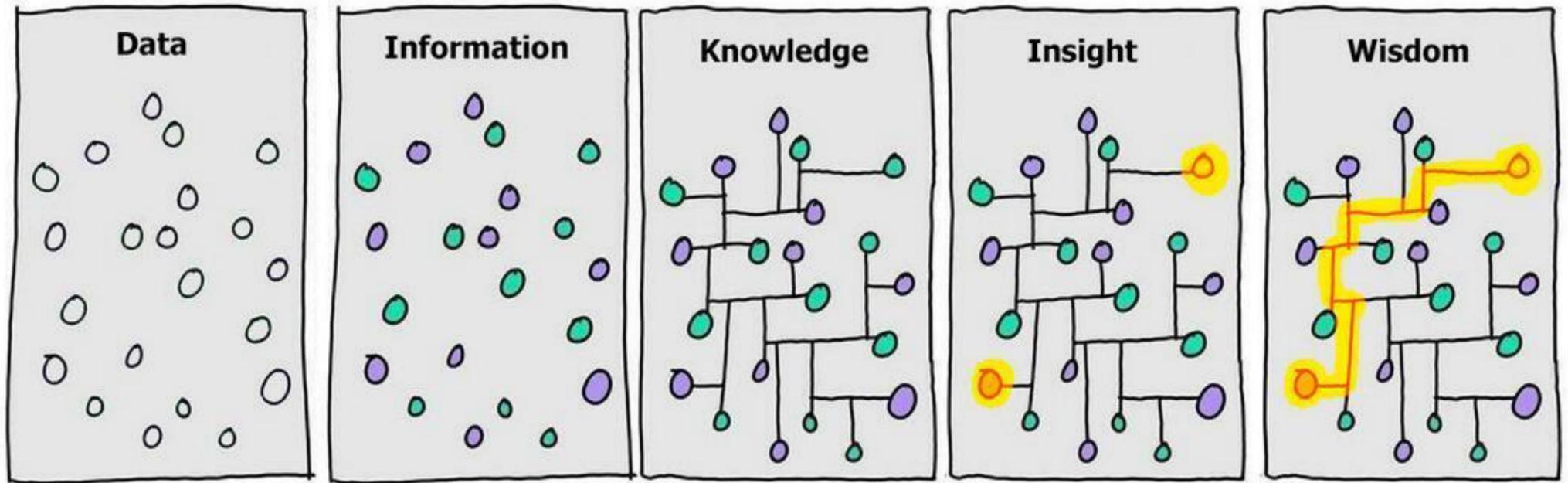
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<sup>09</sup> [Data in Computing](#))

## DIKW Pyramid



# Graph View



## Data about data



# Data Semantics

## semantics

/sɪˈmɑntɪks/ 

*noun*

the branch of linguistics and logic concerned with meaning. The two main areas are *logical semantics*, concerned with matters such as sense and reference and presupposition and implication, and *lexical semantics*, concerned with the analysis of word meanings and relations between them.

- the meaning of a word, phrase, or text.

plural noun: **semantics**

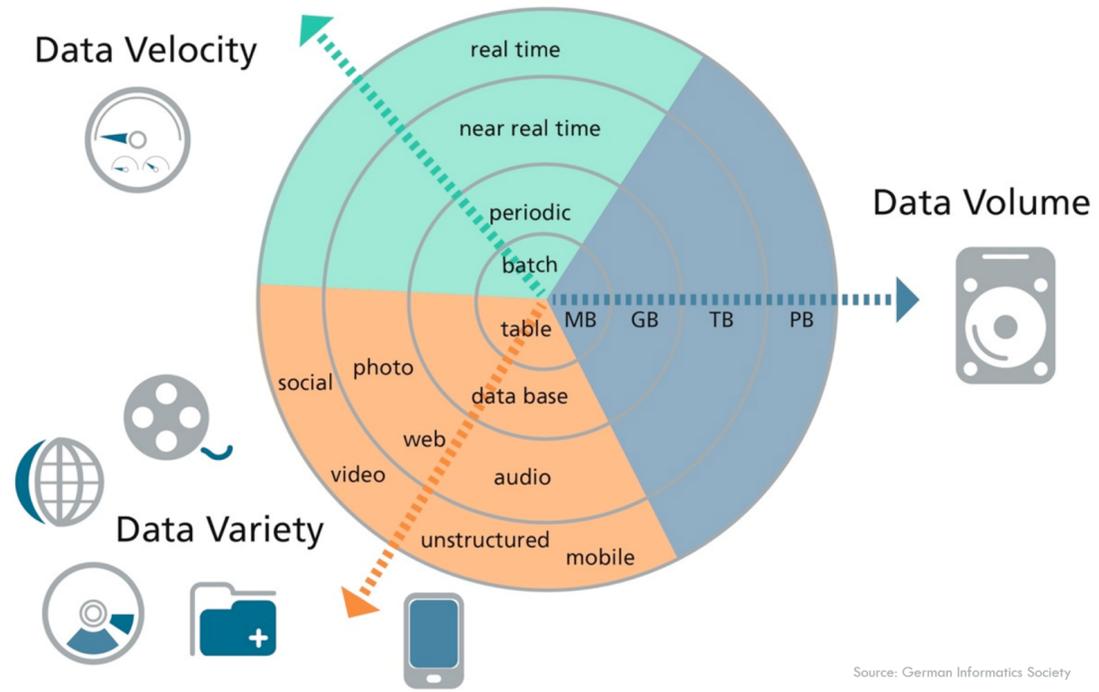
"such quibbling over semantics may seem petty stuff"



Translations, word origin, and more definitions

# Big Data

# Challenges<sup>014</sup>



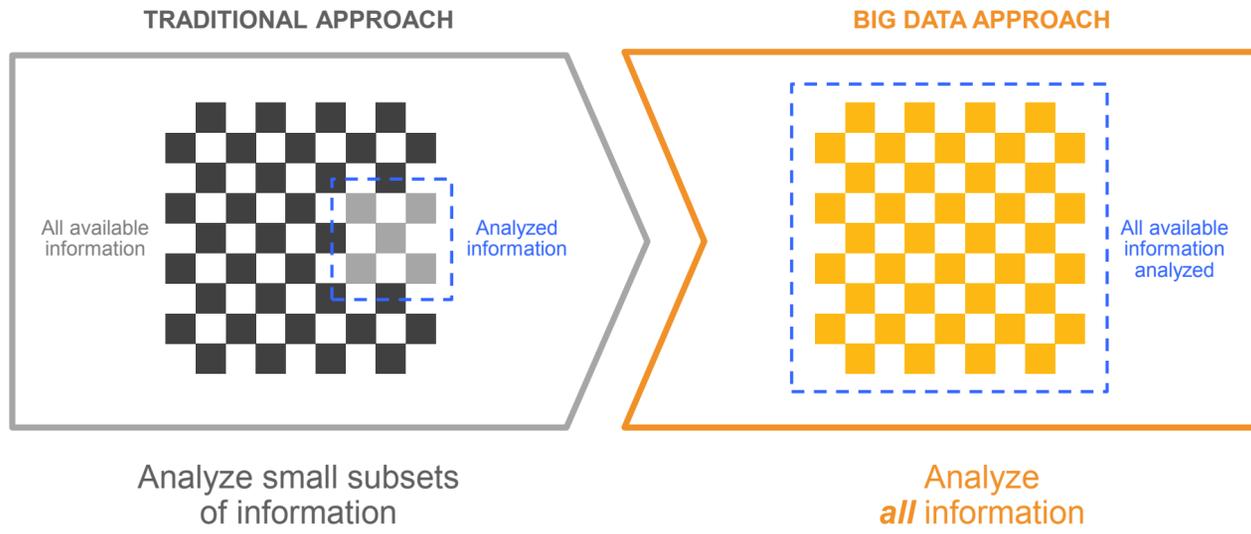
<sup>014</sup> Lanely, 2001

# Paradigm Shift



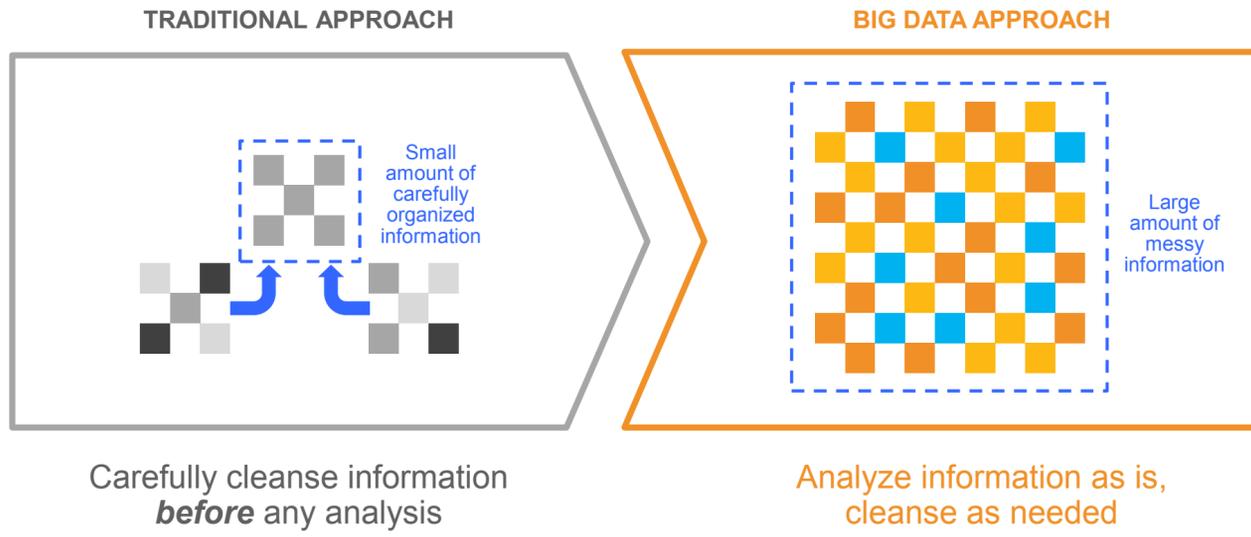
# Paradigm shifts enabled by big data

Leverage more of the data being captured



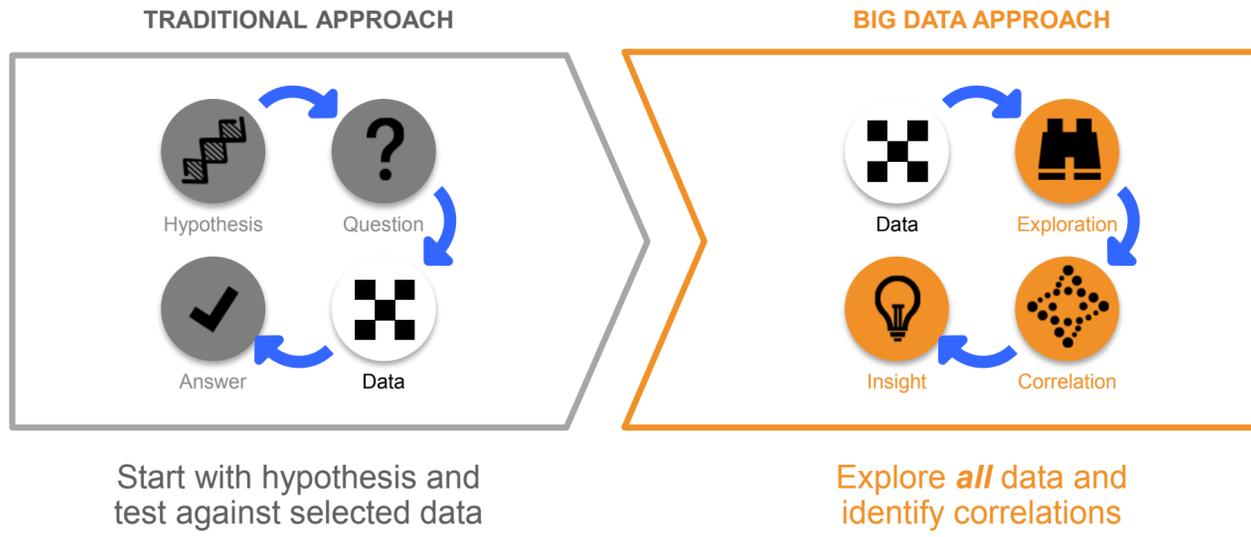
# Paradigm shifts enabled by big data

## Reduce effort required to leverage data



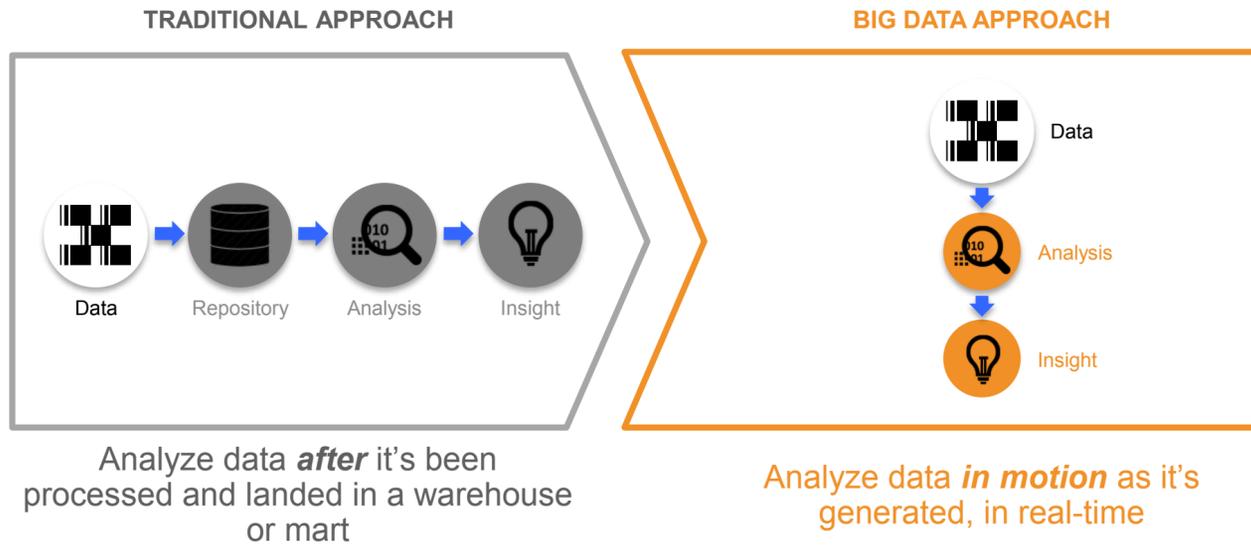
# Paradigm shifts enabled by big data

Data leads the way—and sometimes correlations are good enough



# Paradigm shifts enabled by big data

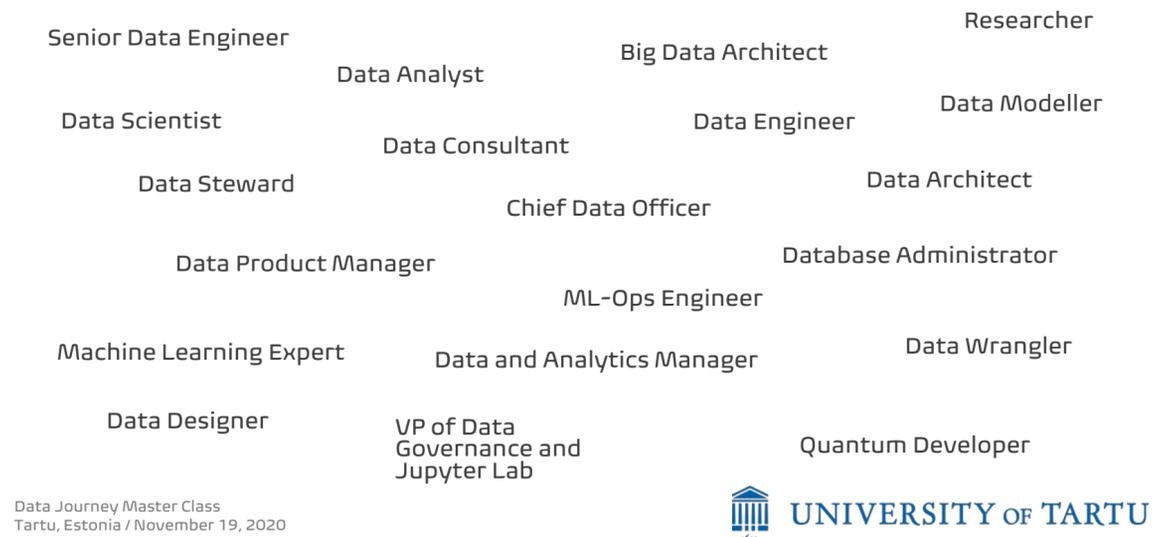
## Leverage data as it is captured



# New Roles

In the context of Big Data, a data engineer must focus on **distributed systems**, and **programming languages** such as Java and Scala.

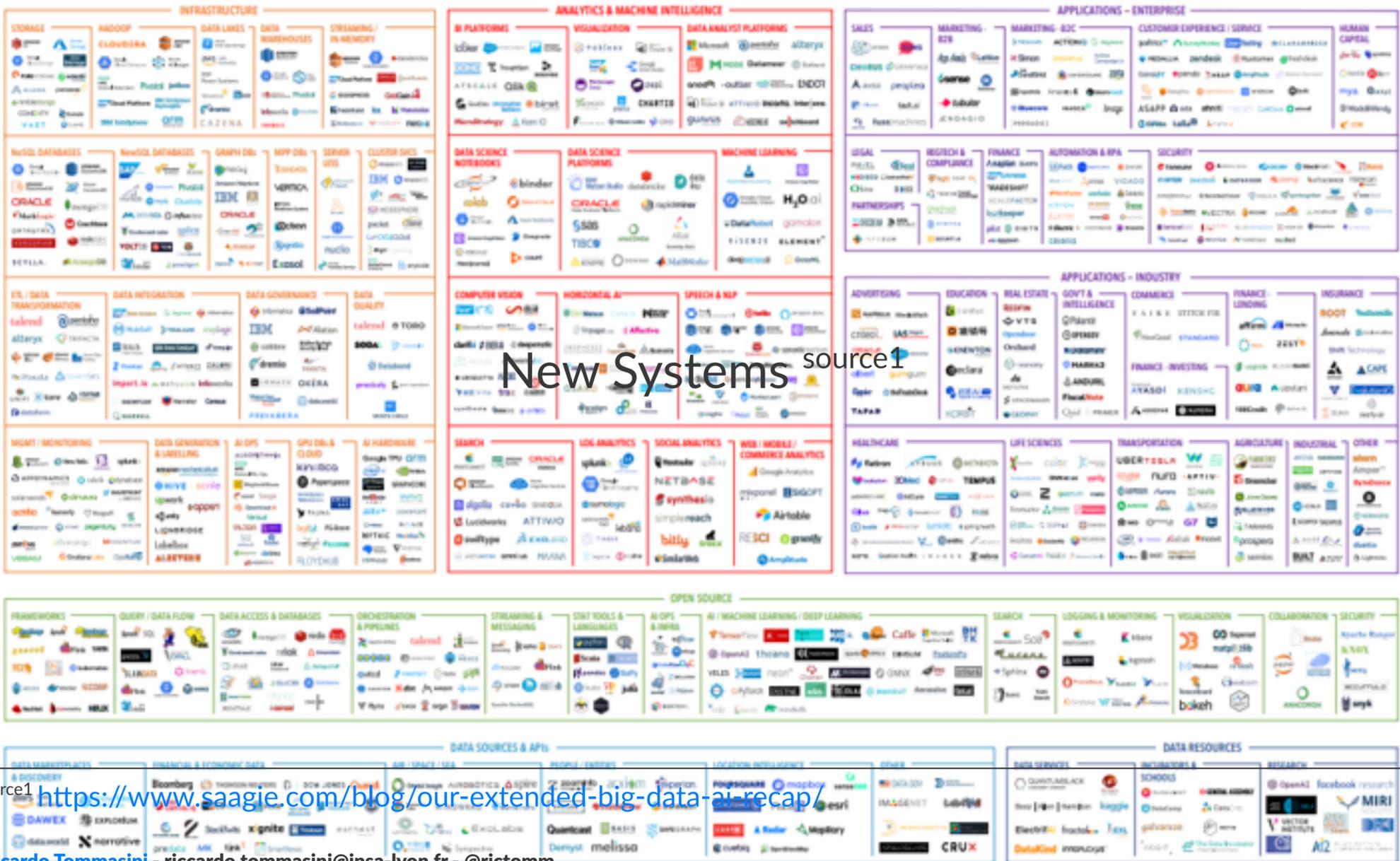
## Profession



## New Tasks

Since data lake are taking data from a wide range of systems, data can be in **structured** or **unstructured** formats, and usually **not clean**, e.g., with missing fields, mismatched data types, and other data-related issues.

Therefore data engineers are challenged with the task of wrangling, cleansing, and integrating data.



New Systems source1

source1 <https://www.saagie.com/blog/our-extended-big-data-ai-recap/>

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# Where is Data?

# Data on the Inside vs Data on the Outside <sup>pt</sup>

	Outside Data	Inside Data
Immutable?	Yes	No
Identity-Based References	Yes	No
Open Schema?	Yes	No
Represent in XML?	Yes	No
Encapsulation Useful?	No	Yes
Long-Lived Evolving Data with Evolving Schema?	No	Yes
Business Intelligence Desirable over Data?	Yes	Yes
Durable Storage in SQL Inside the Service?	Yes: Copy of XML Kept in SQL	Yes 37

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<sup>pt</sup> [Data on the Outside vs Data on the Inside](#) Pat Helland, CIDR 2005

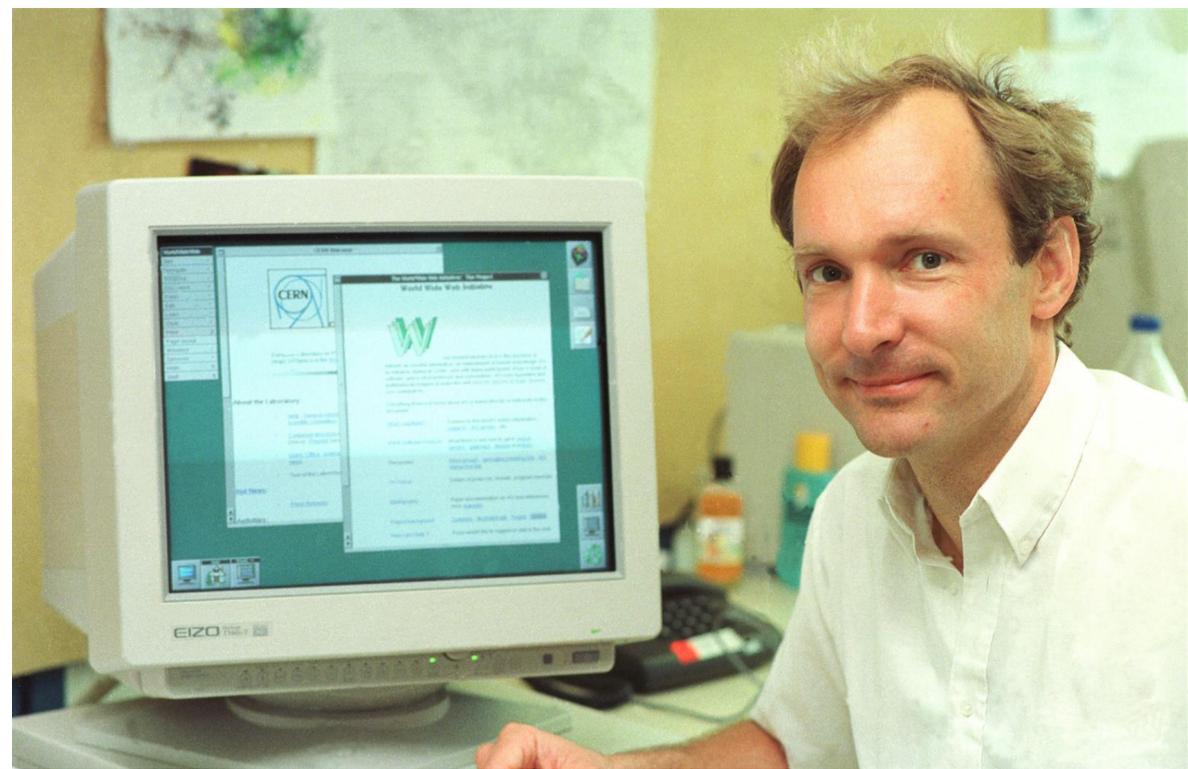
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# The Outside (WEB)

International ecosystem of applications and services that allows us to search, aggregate, combine, transform, replicate, cache, and archive the information that underpins society.

The Web is the result of millions of simple, small-scale interactions between agents and resources that use the founding technologies of HTTP and URI. <sup>121</sup>

The Web is a set of widely commoditised servers, proxies, caches, and content delivery networks [an engineers]



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<sup>121</sup> [Architecture of the World Wide Web, Volume One](#)

# Resources

Resources are the fundamental building blocks

Anything we can expose, i.e., documents, images, videos, audio, devices, people, things...

We can represent them by abstracting the useful information and identifying using a Uniform Resource Identifier (URI)

# URIs

URL format (RFC 2396):

scheme:[//[user:password@]host[:port]][/]path[?query][#fragment]



E.G.:

`git@github.com:nodejs/node.git`

`mongodb://root:pass@localhost:27017/TestDB?options?replicaSet=test`

`http://example.com`

# Representation

Access to a resource is mediated by a representation

This separation is convenient to promote loose-coupling between server (producers) and client (consumers)

Multiple Views and Content Negotiation are the basis for interoperability

# Protocols: HTTP

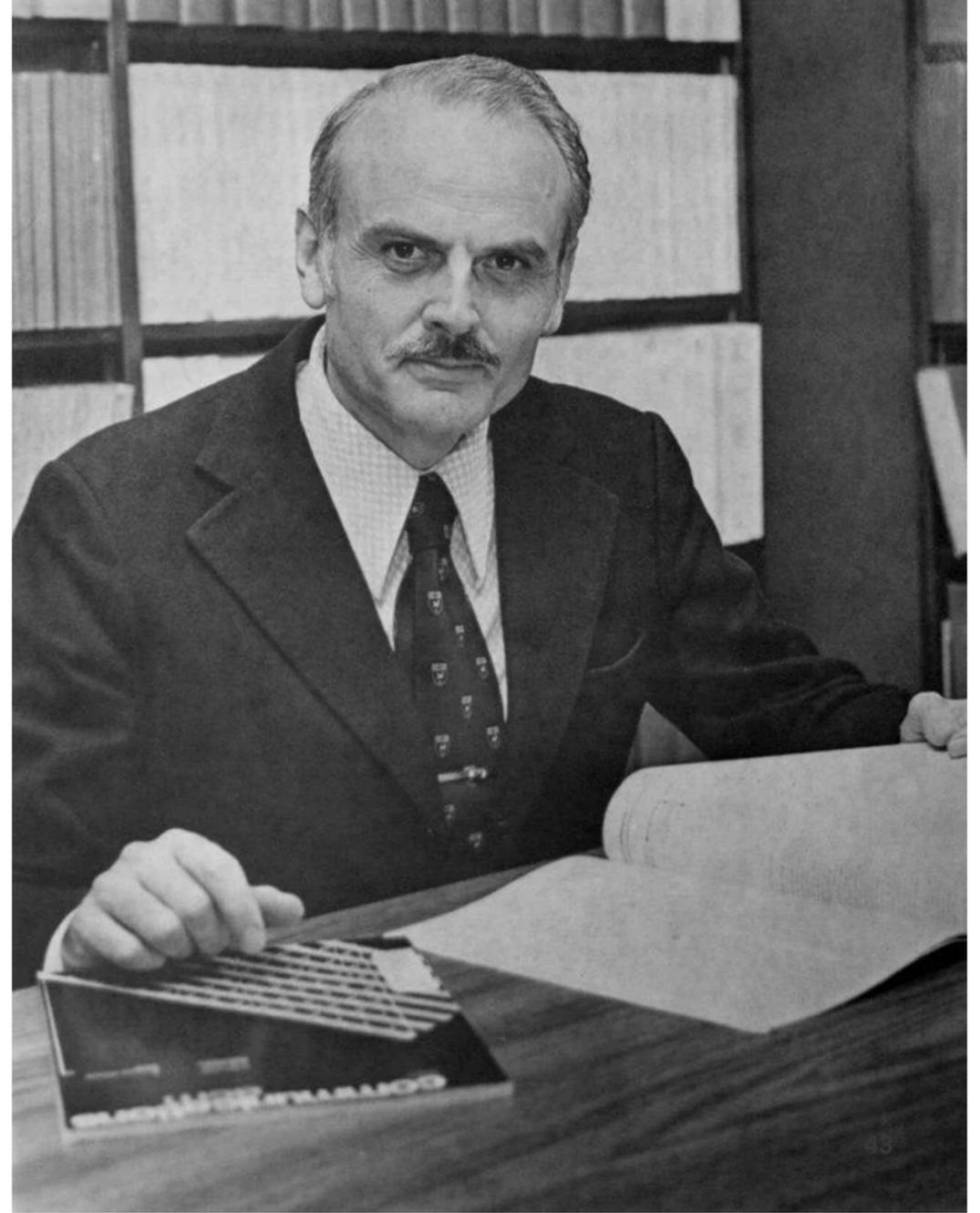
- GET
  - Uniform Interface
  - read-only operation
  - idempotent
- POST
  - like a resource upload
  - idempotent
- DELETE
  - remove resources
  - idempotent
- HEAD
  - HEAD is like GET except it returns only a response code
- PUT
  - the only non-idempotent and unsafe operation is allowed to modify the service in a unique way
- OPTIONS is used to request information about the communication options of the resource

# Databases Management Systems

A database is **an organised collection of structured information, or data**, typically stored electronically in a computer system

Several kind of DBSMs exist. We will survey some of them. It is interesting to know that Edgar F. Codd defined 12+1 rules that make a DBMS relational [link](#)

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## Relational DBMS

- It must be relational as a database and as a management system
- All data should be in table form
- All data should be accessible without ambiguity



# NoSQL Family

Document Database	Graph Databases
  	 
Wide Column Stores	Key-Value Databases
   	    

# Data Warehouse: A Traditional Approach:

A data warehouse is a copy of transaction data specifically structured for query and analysis. — [Ralph Kimball](#)

A data warehouse is a subject-oriented, integrated, time-variant and non-volatile collection of data in support of management's decision making process.-- [Bill Inmon](#)

- A data warehouse is a central repository where raw data is transformed and stored in query-able forms.<sup>03</sup>
- Data Warehouse are still relevant today and their maintenance is part of Data Engineers' responsibilities.
- The warehouse is created with structure and model first before the data is loaded and it is called schema-on-write.

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<sup>03</sup> [What is Data Engineering](#)

# Data Warehouse vs Data Bases

Surprisingly, Data Warehouse isn't a regular database.

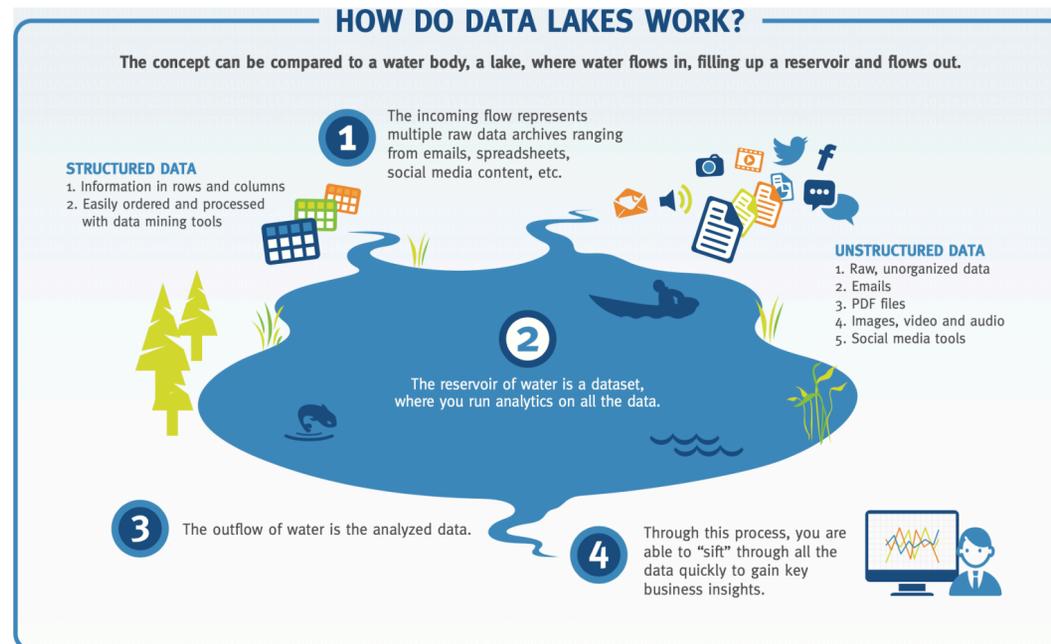
- A database normalizes data separating them into tables and avoiding redundancies
- It supports arbitrary workload and complex queries
- do not store multiple versions of data
- a Data Warehouse uses few tables to improve performance and analytic.
- a Data Warehouse allows simple queries
- supports versioning for complex analysis

# Data Lake

A Data lake is a vast pool of raw data (i.e., data as they are natively, unprocessed). A data lake stands out for its high agility as it isn't limited to a warehouse's fixed configuration<sup>03</sup>.

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<sup>03</sup> [What is Data Engineering](#)



## Full Inforgraphic

- In Data Lake, the raw data is loaded as-is, when the data is used it is given structure, and it is called schema-on-read.
- Data Lake gives engineers the ability to easily change.
- In practice, Data Lake is a commercial term so don't sweat it.

# DATA WAREHOUSE

## Data Lake vs Data Warehouse

# DATA LAKE

- **Structured Data**
- **Schema On Write**
- **Data Pipelines: Extract-Transform-Load**
- **Processing Model: Batch**

VS

- **Unstructured Data**
- **Schema on Read**
- **Data Pipelines: Extract-Load-Transform**
- **Processing Model: Streaming**