

Introduction to Big Data

- CSCI316 -
**Big Data Mining Techniques and
Implementation**

Statement

I did not invent Big Data nor did anyone else. Big Data just happened. Coined by Roger Mougalias in 2005 and driven by technology and commercial ambitions it has become the new frontier for innovation and competition in a global marketplace.

M. Hagenbuchner

Disclaimer

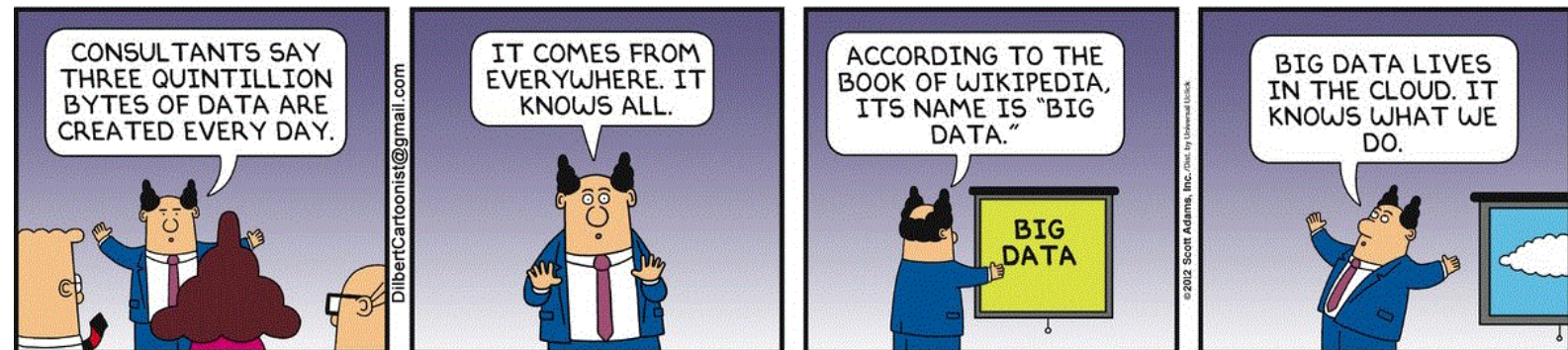
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What is Big Data?

Definition:

Big Data is used in the singular and refers to a collection of data so large and complex, it's impossible to process them with the usual databases and tools. Because of its size, *Big Data* can be hard to capture, store, search, share, analyze and visualize.

Even so the term Big Data is well known its meaning is often not well understood:



What is Big Data?

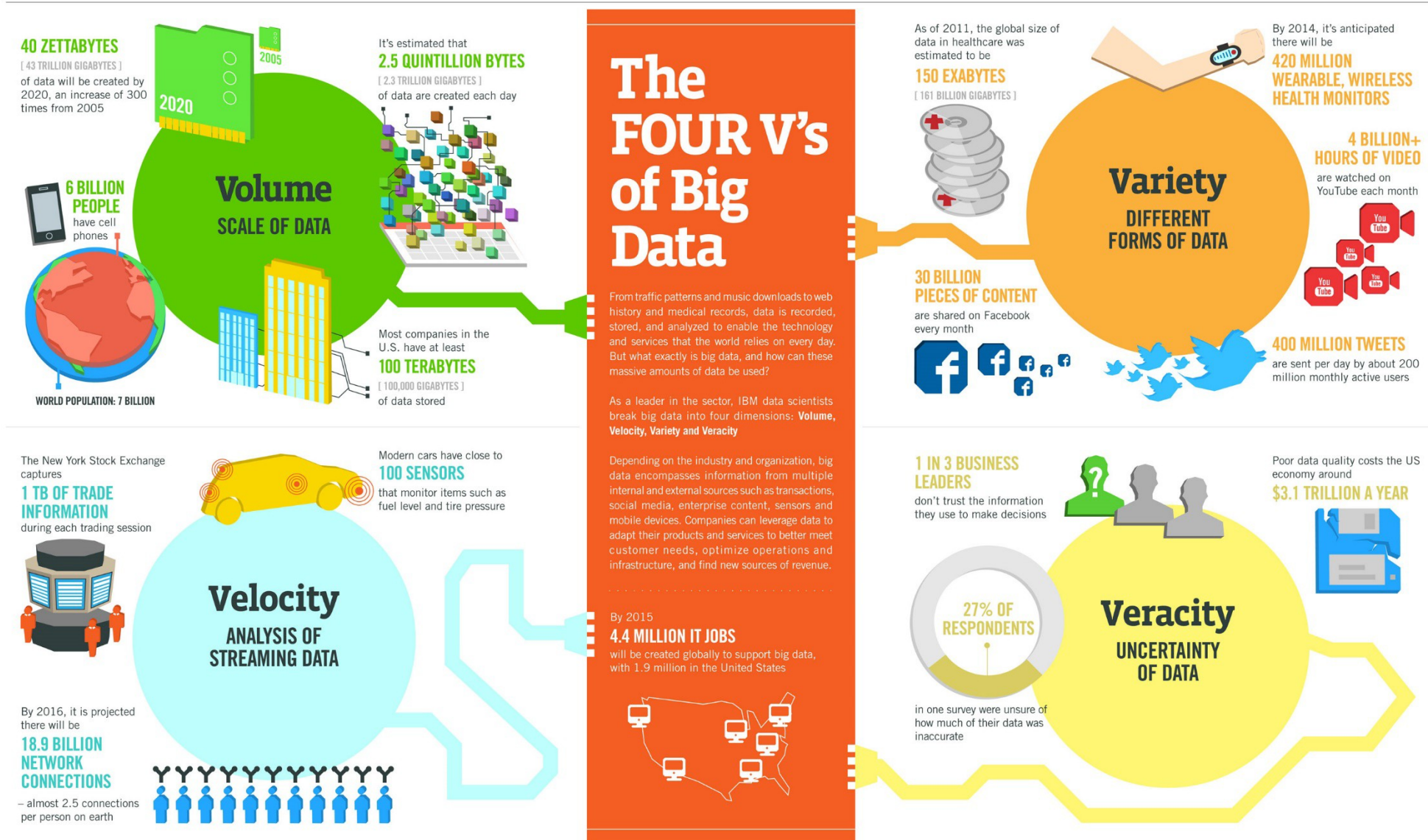
Big Data is also known under alternative terms such as

- Smart Data
- Predictive Analysis
- Data Science
- Massive Data

Big Data has general properties. These are:

- Velocity, Variety, Volume, Veracity
- famously known as the four Vs

What is Big Data? The 4 Vs.



The 4Vs

Volume:

- Concerns the sheer volume of data.
- A typical PC stores TBs of data but data is created at a much higher rate:
 - 3.8 Billion internet users per day (as of 2016)
 - Youtube: 400 hours of videos added every minute (as of 2016)
 - Facebook: 3 million posts per minute (as of 2016)
 - Google: 3,607,080 searches per minute
 - SMS: 15,220,700 texts per minute
 - Instagram: 46,740 pictures per minute.
- In 2016 an estimated 44 Billion GB (Exabyte) of data was created each day, predicted to grow to 463 billion GB by 2025.

The 4Vs

Variety:

- Data comes in many forms and can vary in:
 - Structure: structured (i.e. forms), semi-structured (i.e. newspaper article), unstructured (meta-data).
 - Media: Type of data (i.e. text, multimedia, audio, 3D, geo)
 - Semantic variety: Interpretation of values. (i.e. age=3 vs age=infant, income=55k vs income=above_average)
 - Availability variations: real time (i.e. sensory), intermitted (i.e. satellite data,) or stored (i.e. records).
- Increased data diversity
- Adds complexity

The 4Vs

Velocity:

- Refers to:
 - A. Speed by which data is generated, stored, and analysed.
 - How much data is generated per unit of time?
 - Speed by which results need to become available.
 - May require real time processing (i.e. streaming data).
 - B. Speed by which data changes over time
 - Domain changes, environmental changes, changes in user behaviour, changes in expectations.
 - May require regular update of models

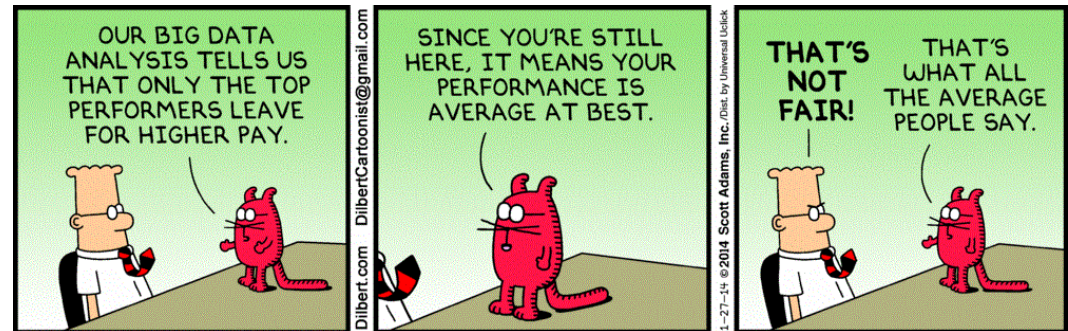
The 4Vs

Veracity:

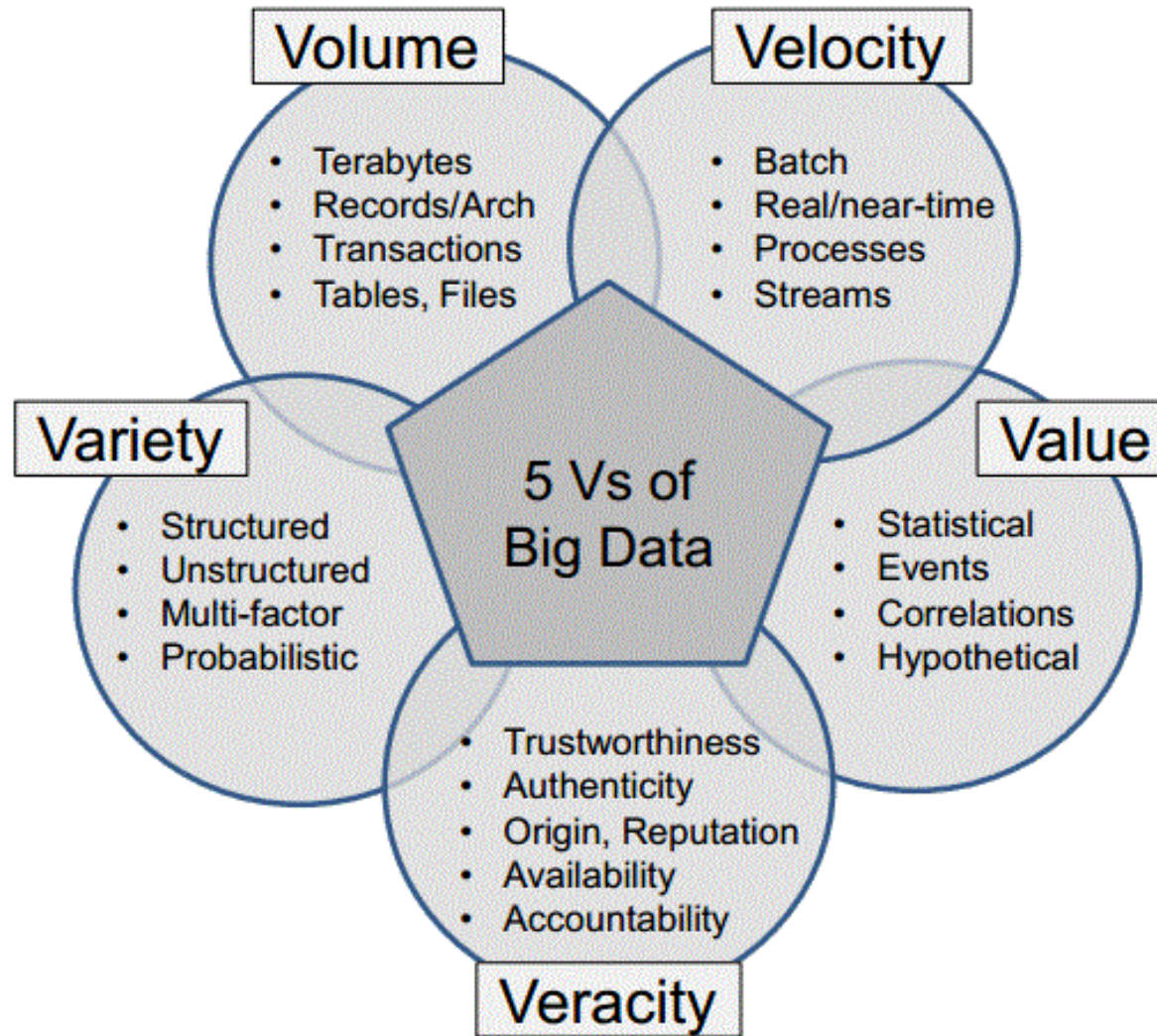
- Refers to data quality, data uncertainty, imprecise data types.
- Data validity:
 - Noise and accuracy of data.
 - Regulated vs unregulated
- Data volatility:
 - Is data collected in the past still valid today?
 - Are results from data collected today valid for future decision making applications?
- Big Data is only as good as the quality of the data (junk in = junk out)

The fifth V?

- The four Vs are said to be fundamental dimensions of Big Data.
- Although **Value** is at the heart of Big Data:
- Refers to the value of Big Data results (the new insights obtained):
 - Academic value: Domain understanding, method development,...
 - Statistical value: To get a better overview
 - Correlations: Discovery of links and relationships.
 - Business value: Buying and selling data, buying and selling results, decision support.
- Note that “value” is in the eye of the beholder:



4 or 5Vs?



From a practical perspective:

- Big data only makes sense when there is value associated with it.
- Volume, Velocity, Variety, and Veracity refer to property of data (the input) whereas Value refers to the envisaged results (the output).

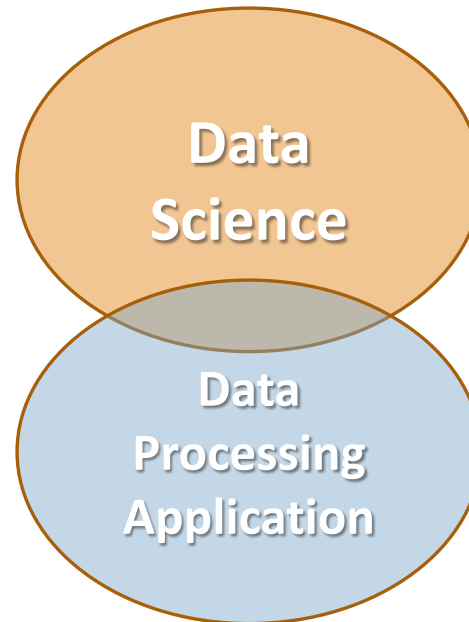
Data Scientist vs. Data Engineer

Data Scientist

- Analyse and model data
- Make prediction based on data
- Build data pipelines to fulfil certain tasks

Data Engineer

- Develop data processing applications
- Deploy the output of data scientists in production



Career Paths and Challenges

- ❖ What Does a Data Scientist Do?
- ❖ Skills Needed to Be a Data Scientist
- ❖ Where Do Data Scientists Work?
- ❖ Related Jobs in Data Science
- ❖ Challenges in Data Science Careers

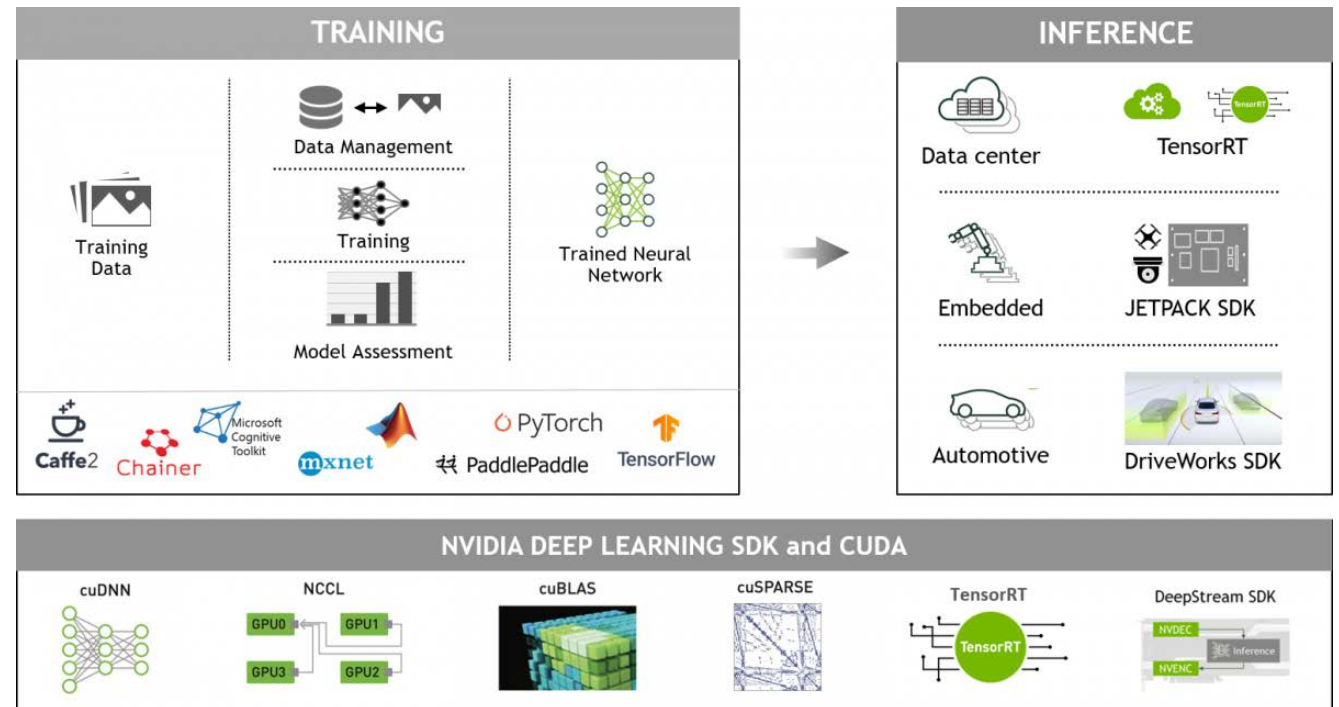
What Does a Data Scientist Do?

Data Collection

Data Preprocessing

Data Visualization

Data Analytics and Application



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Collecting Data from Various Sources

Source: <https://innovativeadagency.com/blog/importance-data-collection/>

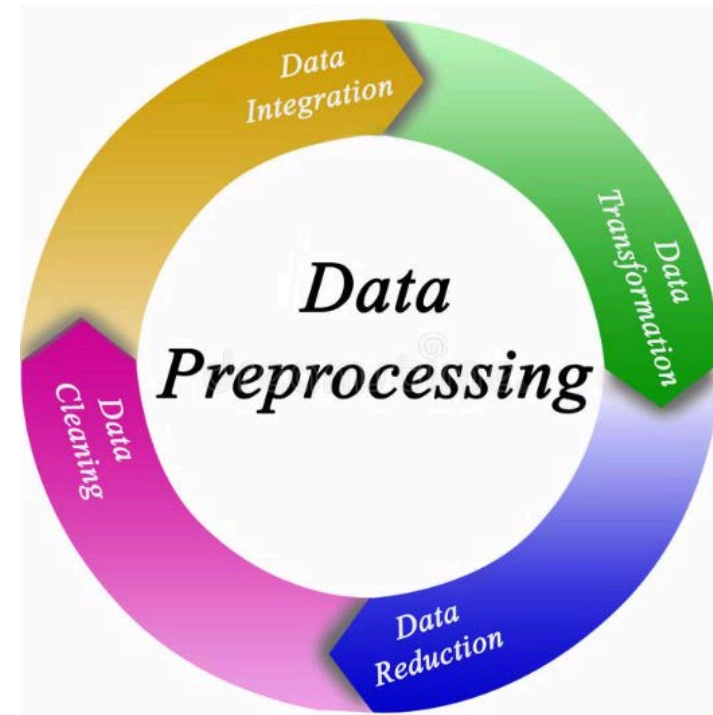
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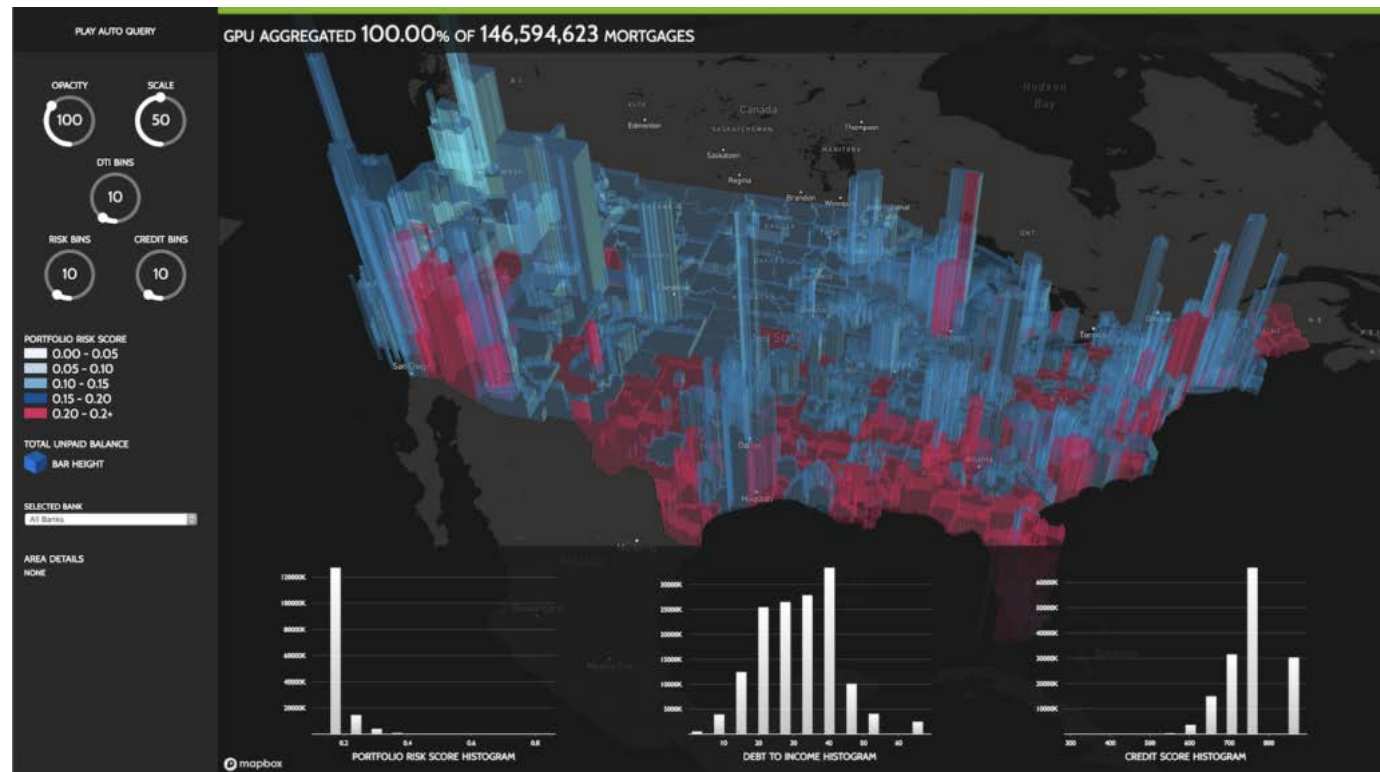
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Source: <https://developer.nvidia.com/blog/gpu-accelerated-analytics-rapids/>

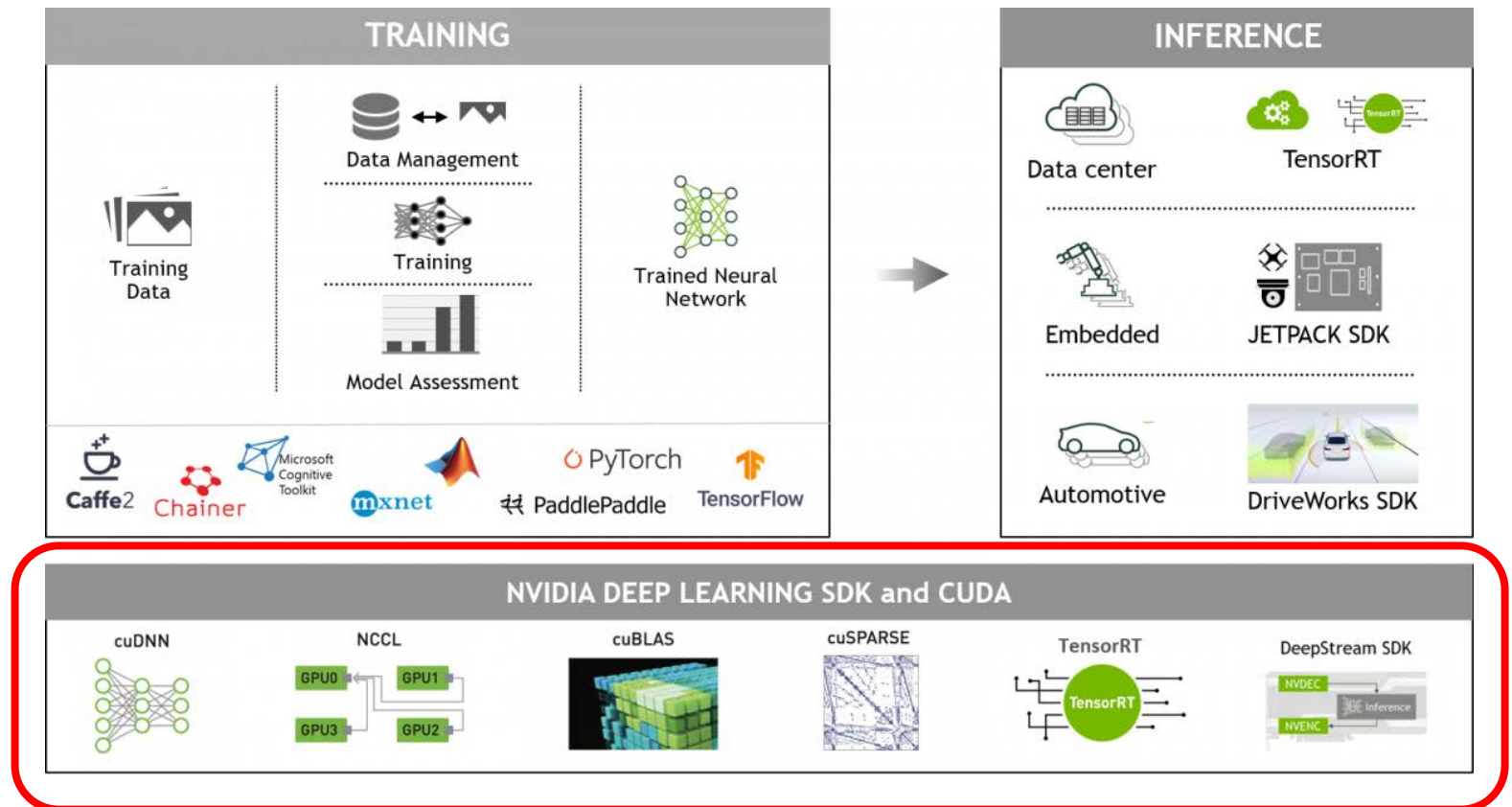
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Skills Needed to Be a Data Scientist

Analytical Skills

Communication Skills

Critical and Logical Thinking Skills

Math Skills

Computer Programming Competency

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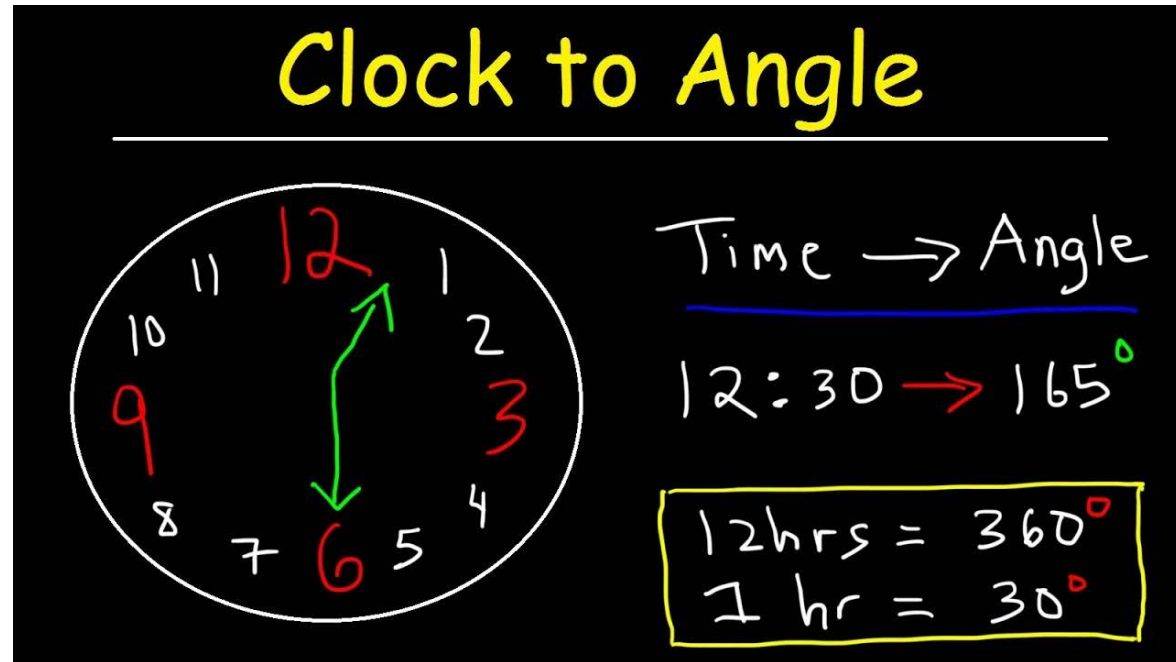
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Source: <https://www.youtube.com/watch?v=LEHYr0XfSyl>

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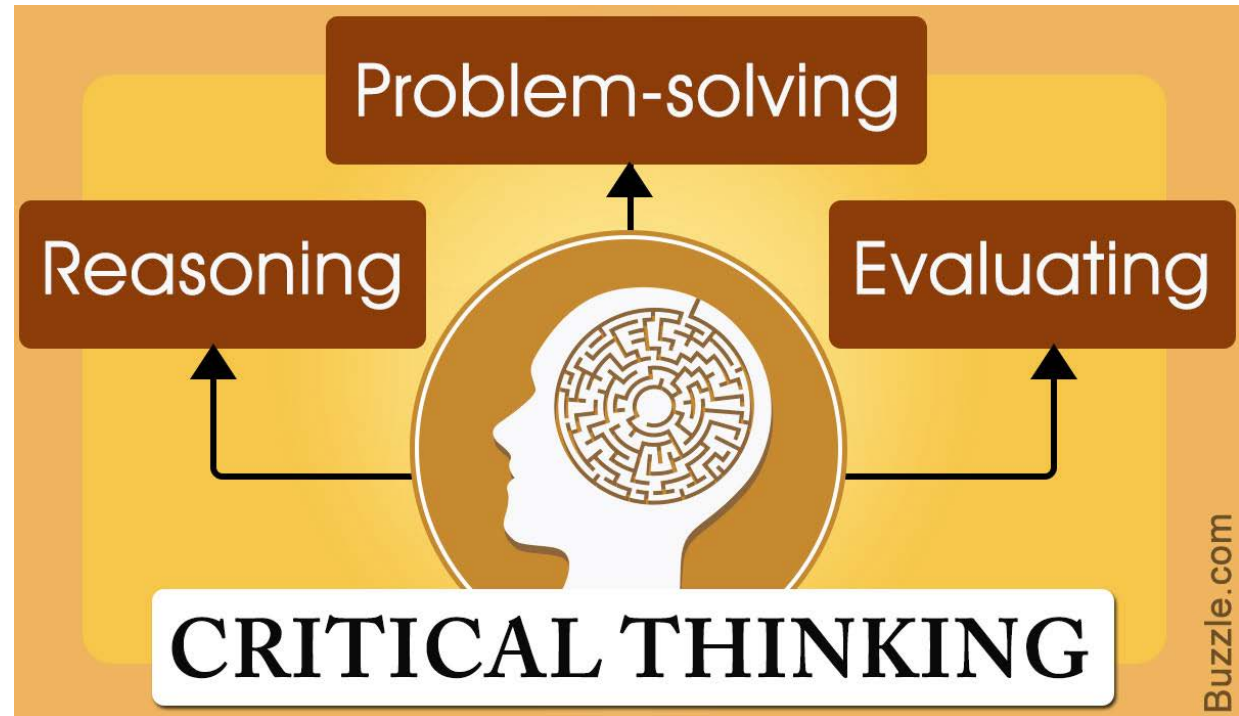
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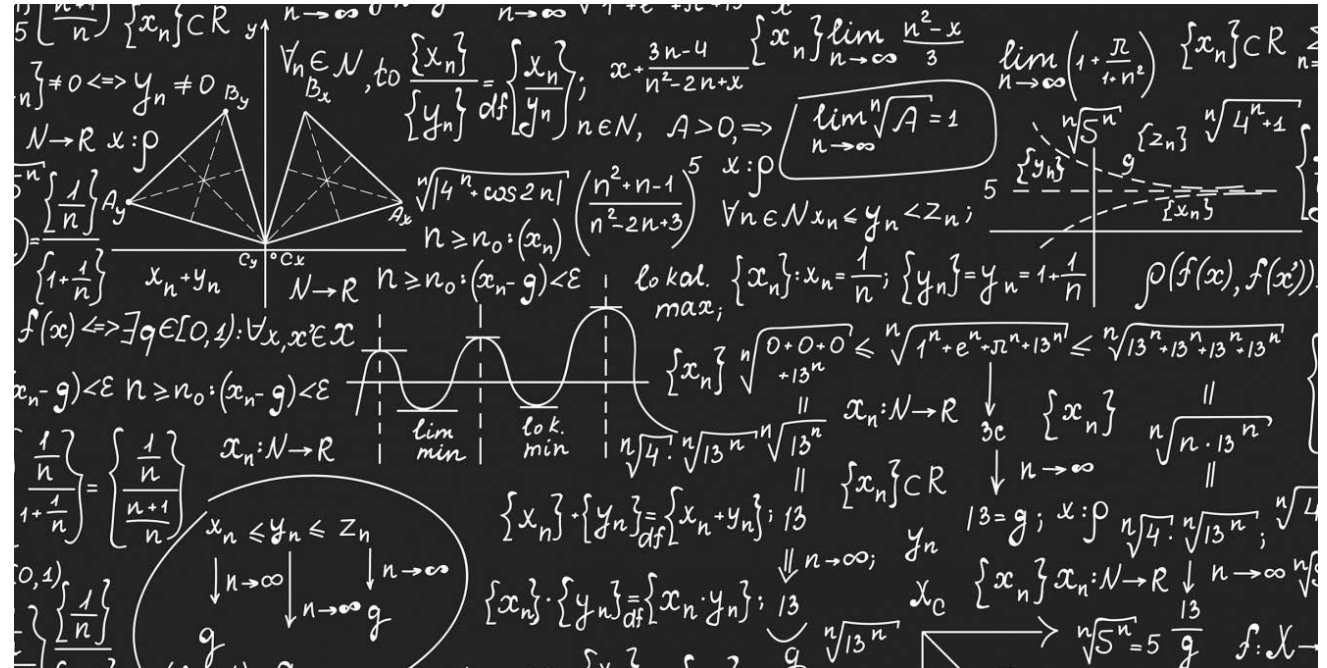
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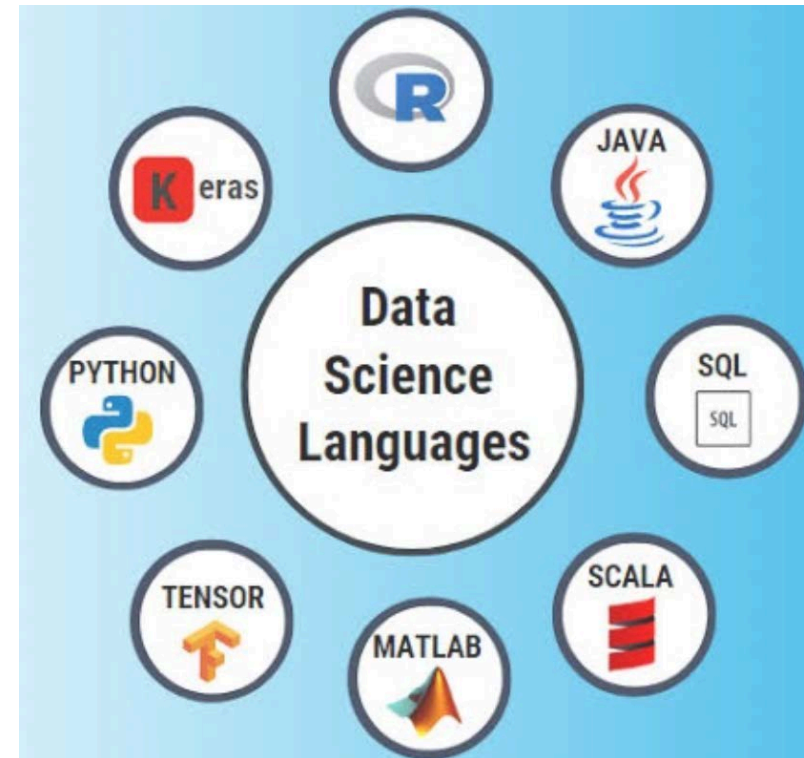
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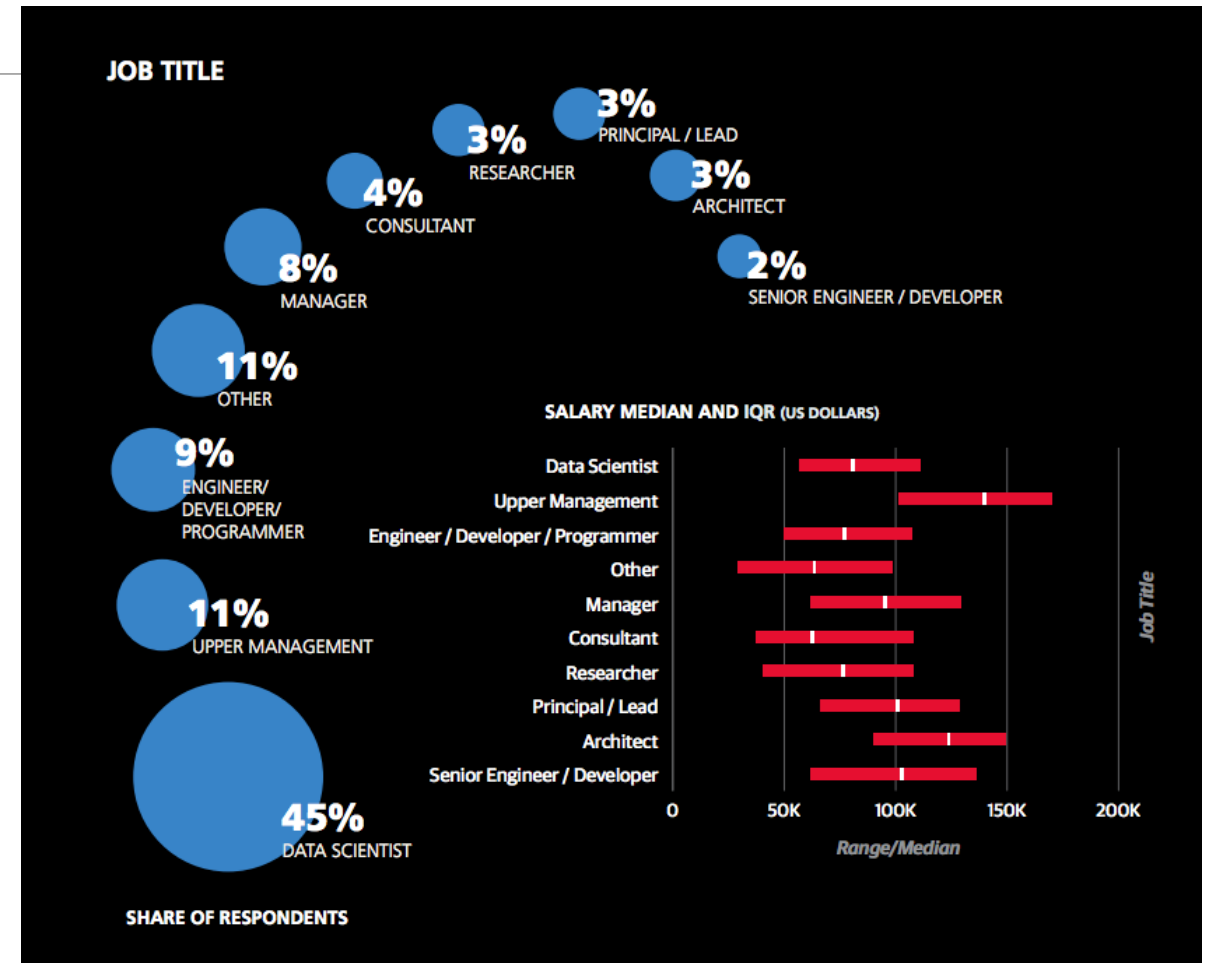


What Salary Can a Data Scientist Earn?

Data Scientist Salary by Job Title

According to an O'Reilly data science salary report, 45 percent of those surveyed said they hold the title of “data scientist.”

In general, the more a data science professional engages in managerial tasks, the higher the salary.



Source: <https://datasciencedegree.wisconsin.edu/data-science/data-scientist-salary/>

Where Do Data Scientists Work?

Academia

- Research and development
- Colleges and universities
- ...

Industry

- Software companies
- Car companies
- Delivery companies
- ...

Related Jobs in Data Science

Data analyst

Research scientist

Machine learning engineer

Big data engineer

...

Data Scientist Job Titles Include:

- Product analyst
- Data analyst
- Research scientist
- Quantitative analyst
- Machine learning engineer
- Data engineer
- Big data engineer
- Back-end engineer
- Natural language processing engineer
- Business analyst
- Statistician
- Economist
- Applied scientist
- Operations research scientist
- Research scientist
- Research engineer
- Machine learning scientist
- Product scientist
- Business intelligence analyst
- Natural Scientist

Challenges in Data Science Careers

Insights not Used in Decision Making

Data Privacy, Veracity, Unavailability

Limitations of tools to scale/deploy

Wrong Questions Asked