Data Science

Introduction

Google Says...

2.5 quintillion bytes of **data** are produced by humans every **day**. If you've wondered how much **data** the average person uses **per** month, you can start by looking at how much **data** is **created** every **day** in **2020** by the average person. This currently stands at 2.5 quintillion bytes **per** person, **per day**. Sep 10, 2020

techjury.net > blog > how-much-data-is-created-every-day

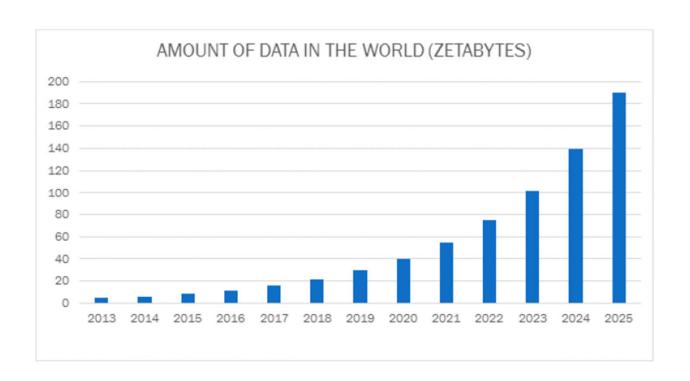
How Much Data Is Created Every Day in 2020? [You'll be ...

What does that mean?

Prefix		Bass 40	Dosimal	English		
Name	Symbol	Base 10	Decimal	Short scale		
yotta	Y	10 ²⁴	1 000 000 000 000 000 000 000 000	septillion		
zetta	Z	10 ²¹	1 000 000 000 000 000 000 000	sextillion		
exa	E	10 ¹⁸	1 000 000 000 000 000 000	quintillion		
peta	Р	10 ¹⁵	1 000 000 000 000 000	quadrillion		
tera	Т	10 ¹²	1 000 000 000 000	trillion		
giga	G	10 ⁹	1 000 000 000	billion		

If data had mass, earth could be a blackhole!

Exponential Growth of Data



What is Data Science

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- Wikipedia
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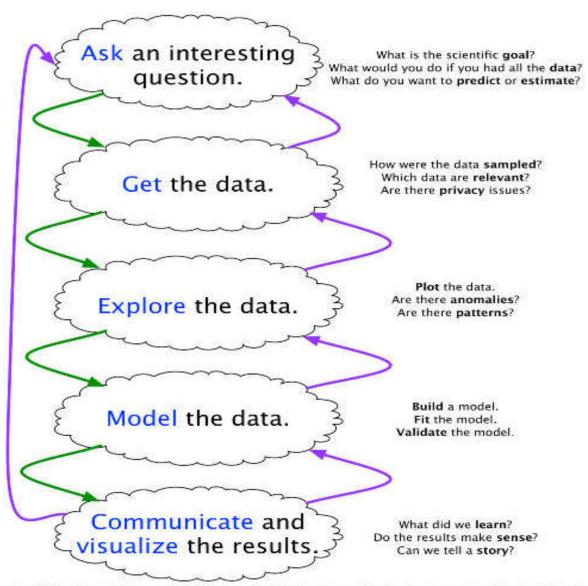
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- Wikipedia
 - "Data Science is the extraction of knowledge from data, which is a continuation of the field of data mining and predictive analytics."
- NIST Big Data Working Group
 - "Data Science is the empirical synthesis of actionable knowledge from raw data through the complete data lifecycle process."

Why Data Science Matters

- It empowers policy makers with quantifiable data driven evidences
- It helps organizations
 - to make better decisions and test those decisions
 - to and make plans for improvement
 - to track trends, analyze user behaviors and define goals
 - to identify and refine target audiences
 - to identify opportunities
 - to design effective business processes and adopt best practices

The Data Science Process



Joe Blitzstein and Hanspeter Pfister, created for the Harvard data science course http://cs109.org/.

The Data Science Process

- The Data Science Process is similar to the scientific process one of observation, model building, analysis and conclusion:
 - Ask questions
 - Data Collection
 - Data Exploration
 - Data Modeling
 - Data Analysis
 - Visualization and Presentation of Results
- Note: This process is by no means linear!

It's all about story! Can we tell a story?

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- In April 2017, Hubway held a Data Visualization Challenge at the Microsoft NERD Center in Cambridge, releasing 5 years of trip data.

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- In April 2017, Hubway held a Data Visualization Challenge at the Microsoft NERD Center in Cambridge, releasing 5 years of trip data.
- The Question
 - What does the data tell us about the ride share program?

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_ 8	seq_id	hubway_id	status	duration	start_date	strt_statn	end_date	end_statn	bike_nr	subsc_type	zip_code	birth_date	gender
0	1	8	Closed	9	7/28/2011 10:12:00	23.0	7/28/2011 10:12:00	23.0	B00468	Registered	'97217	1976.0	Male
1	2	9	Closed	220	7/28/2011 10:21:00	23.0	7/28/2011 10:25:00	23.0	B00554	Registered	'02215	1966.0	Male
2	3	10	Closed	56	7/28/2011 10:33:00	23.0	7/28/2011 10:34:00	23.0	B00456	Registered	'02108	1943.0	Male
3	4	11	Closed	64	7/28/2011 10:35:00	23.0	7/28/2011 10:36:00	23.0	B00554	Registered	'02116	1981.0	Female
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Based on the data, what kind of questions can we ask?

- Who?
 - Who's using the bikes?

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 - Subscribers or one time users?

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- Sometimes the data need a lot of pre-processing.

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 - More during the weekend than on the weekdays?
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- Sometimes the feature we want to explore doesn't exist in the data, and must be engineered!

- Why?
 - For what reasons/activities are people checking out bikes?
- Refine into specific hypotheses
 - More bikes are used for recreation than commute?
 - More bikes are used for touristic purposes?
 - Bikes are use to bypass traffic?

- Why?
 - For what reasons/activities are people checking out bikes?
- Refine into specific hypotheses
 - More bikes are used for recreation than commute?
 - More bikes are used for touristic purposes?
 - Bikes are use to bypass traffic?
- Do we have the data to answer these questions with reasonable certainty?
- What data do we need to collect in order to answer these questions?

- How?
 - Questions that combine variables.
 - How does user demographics impact the duration the bikes are being used?
 Or where they are being checked out?
 - How does weather or traffic conditions impact bike usage?
 - How do the characteristics of the station location affect the number of bikes being checked out?
- How questions are about modeling relationships between different variables.

Communicate the results What is your story? Data Visualization

Contents of the course

- The scope of Data Science
- Descriptive Statistics
- Exploratory Data Analysis, Principles of Visualizing Data
- Data Scraping, Cleaning and Summarization
- Statistical Significance and P-values
- Building Models and Validating Models
- Linear Algebra Review
- Linear Regression and Logistic Regression
- Crowdsourcing and Ensemble Learning
- Large-scale Clustering
- Mining Massive Datasets
- Python for data analysis, data wrangling and modeling.

