
CDS501: PRINCIPLES & PRACTICES OF DATA SCIENCE & ANALYTICS

Chapter 1: Data Science Process

Outline

- Introduction
- What is Data Science?
- Applications of Data Science
- Stages of Data Science Project

How Much Data Do We Create Every Day?

Introduction


Forbes

May 21, 2018, 12:42am EDT

How Much Data Do We Create Every Day? The Mind-Blowing Stats Everyone Should Read



Bernard Marr Contributor @
Enterprise Tech

 This article is more than 2 years old.

The amount of data we produce every day is truly mind-boggling. There are 2.5 quintillion bytes of data created each day at our current pace, but that pace is only accelerating with the growth of the Internet of Things (IoT). Over the last two years alone 90 percent of the data in the world was generated. This is worth re-reading! While it's almost impossible to wrap your mind around these numbers, I gathered together some of my favorite stats to help illustrate some of the ways we create these colossal amounts of data every single day.



306.4 billion emails are sent everyday.

500 million Tweets are made everyday.

95 million photos and videos are shared every day on Instagram.

1.7MB of data is created every second by every person during 2020.

2.5 quintillion bytes of data are produced by humans every day.

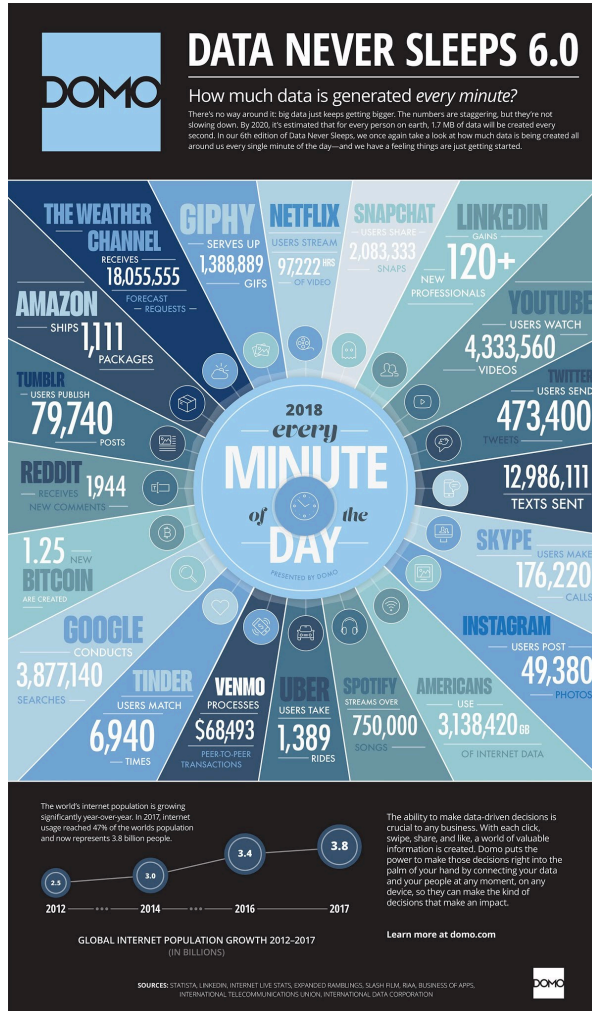
Introduction

1 quintillion = 1 000 000 000 000 000 000 000

A million million million

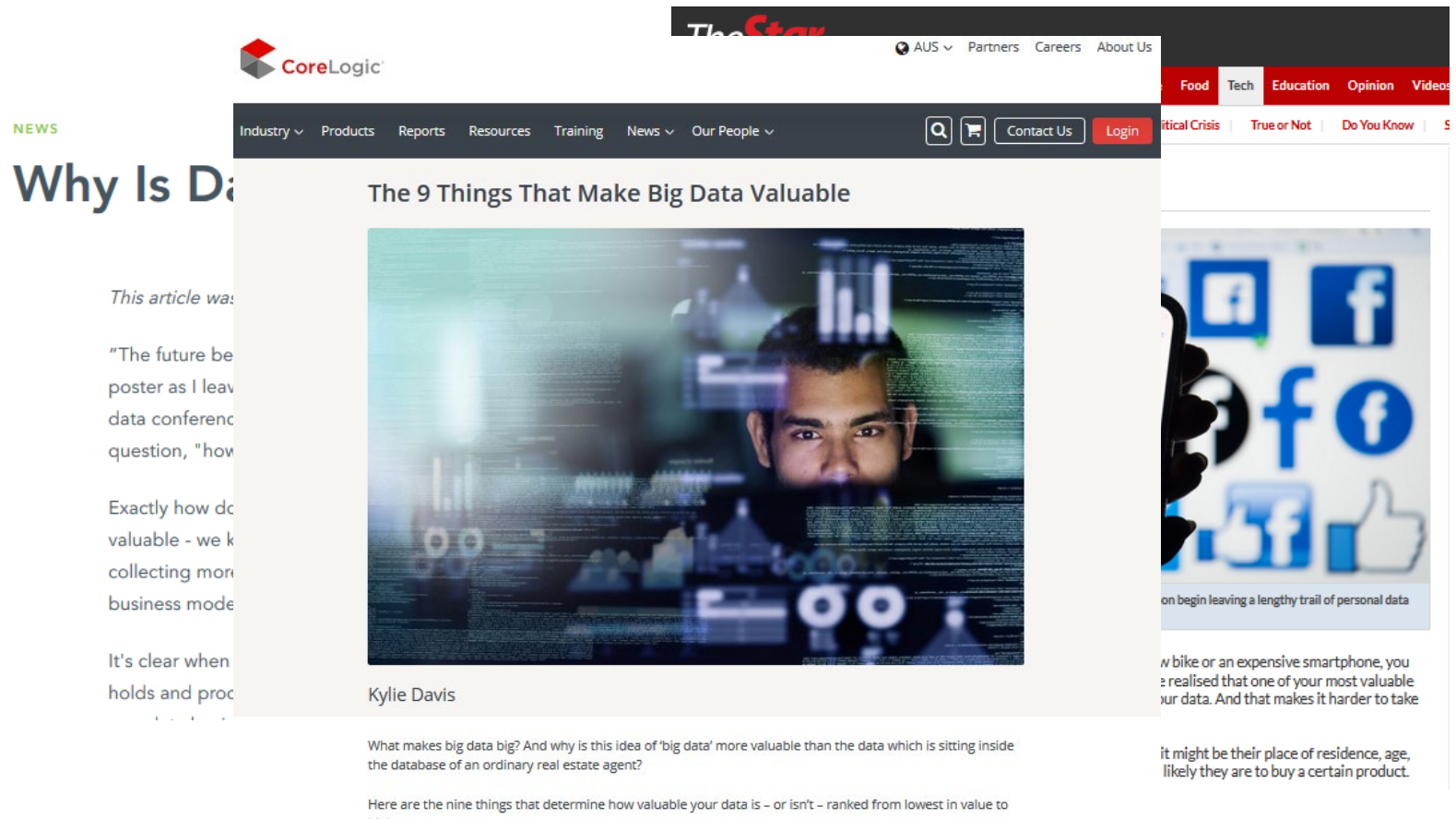
2.5 quintillion = 2 500 000 000 000 000 000 000

Introduction



- Social media
- Online video sharing platform
- Media service provider
- Online retailer
- Mobile payment service

Introduction



CoreLogic

AUS Partners Careers About Us

Food Tech Education Opinion Videos

Critical Crisis True or Not Do You Know

NEWS

Why Is Data Valuable?


This article was...

"The future belongs to those who learn to use data as a poster as I leave the data conference, the question, "how much data is too much?"

Exactly how do we make data valuable - we know we're collecting more data than ever before, but how do we make it valuable in our business model?

It's clear when you look at the data that holds and processes data, it's clear that data is valuable.

The 9 Things That Make Big Data Valuable



Kylie Davis

What makes big data big? And why is this idea of 'big data' more valuable than the data which is sitting inside the database of an ordinary real estate agent?

Here are the nine things that determine how valuable your data is - or isn't - ranked from lowest in value to highest:

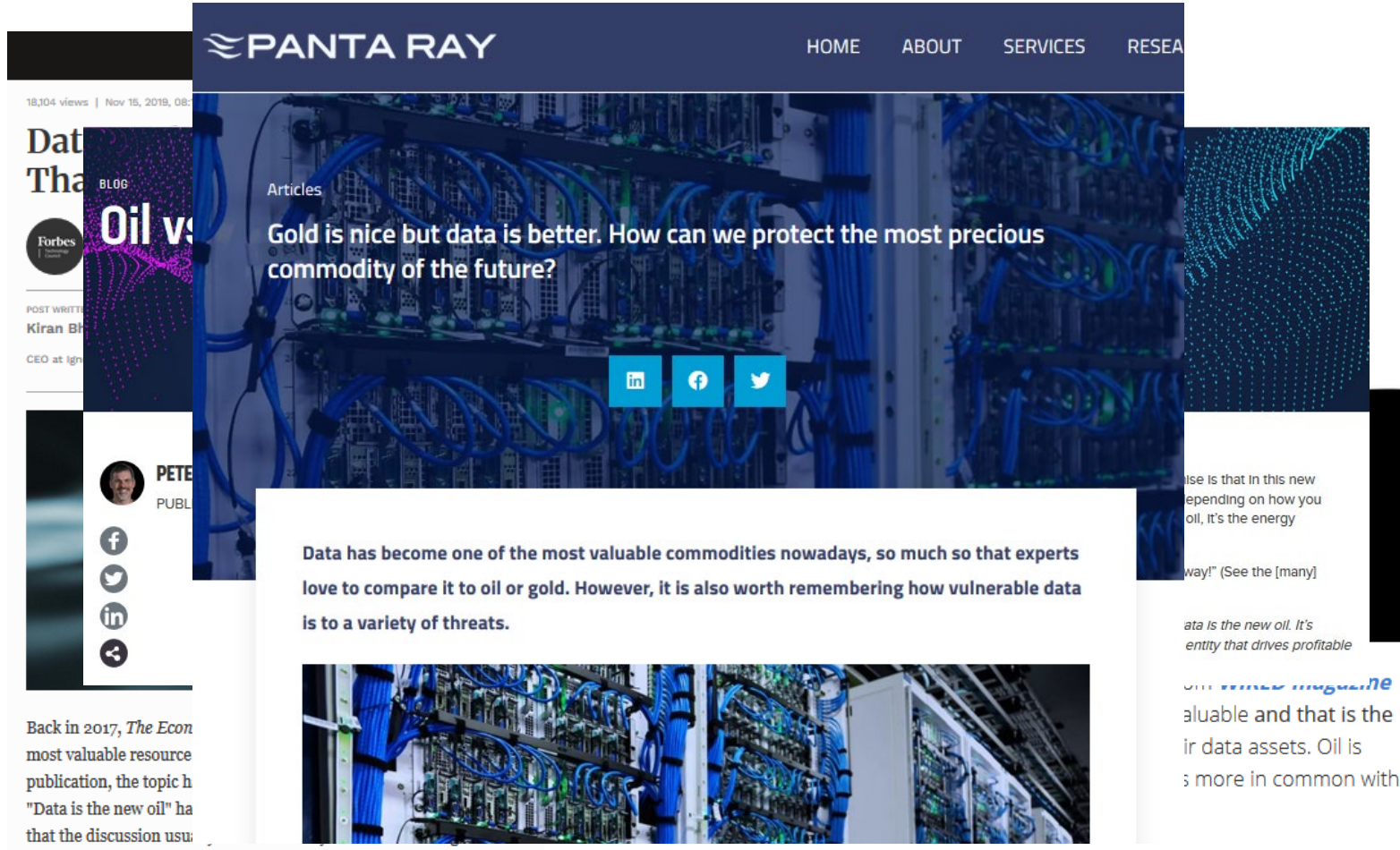
on begin leaving a lengthy trail of personal data

on a bike or an expensive smartphone, you've realised that one of your most valuable assets is your data. And that makes it harder to take

it might be their place of residence, age, gender, likely they are to buy a certain product.

Data is valuable

Introduction



PANTA RAY HOME ABOUT SERVICES RESEARCH

18,304 views | Nov 15, 2019, 08:00

Data That Matters

BLOG

Oil vs Gold

Forbes

POST WRITTEN BY
Kiran Bhargava
CEO at Ignite

PETER
PUBLISHED


f t in

Articles

Gold is nice but data is better. How can we protect the most precious commodity of the future?

LinkedIn Facebook Twitter

Data has become one of the most valuable commodities nowadays, so much so that experts love to compare it to oil or gold. However, it is also worth remembering how vulnerable data is to a variety of threats.



...ise is that in this new
depending on how you
oil, it's the energy

...way!" (See the [many]


...ata is the new oil. It's
entity that drives profitable

...om *TIME* magazine
valuable and that is the
ir data assets. Oil is
s more in common with

Back in 2017, *The Economist* named data the most valuable resource publication, the topic has become "Data is the new oil" has become that the discussion usu


Data is valuable





Introduction



BLOG

Oil vs. Data – Which is more Valuable?

 **PETER SILVA**
PUBLISHED APRIL 09, 2019

It depends who you ask.

In recent years there's been a volley of sorts about data replacing oil as the world's most valuable resource. The basic premise is that in this new digital economy, data and what you extract from that data is similar to oil a century ago. An untapped, massive asset that—depending on how you extract and use it—can have enormous rewards. The raw material's value comes from the refinement into a commodity. For oil, it's the energy extracted; for data, it's in the knowledge extracted.

Economists, professors and even CEOs are touting that data is the new oil in today's economy while others are saying, "no way!" (See the [many] references below for examples.)

The earliest mention of this notion is from 2006. UK Mathematician and architect of Tesco's Clubcard, Clive Humby said, "*Data is the new oil. It's valuable, but if unrefined it cannot really be used. It has to be changed into gas, plastic, chemicals, etc. to create a valuable entity that drives profitable activity; so must data be broken down, analyzed for it to have value.*"

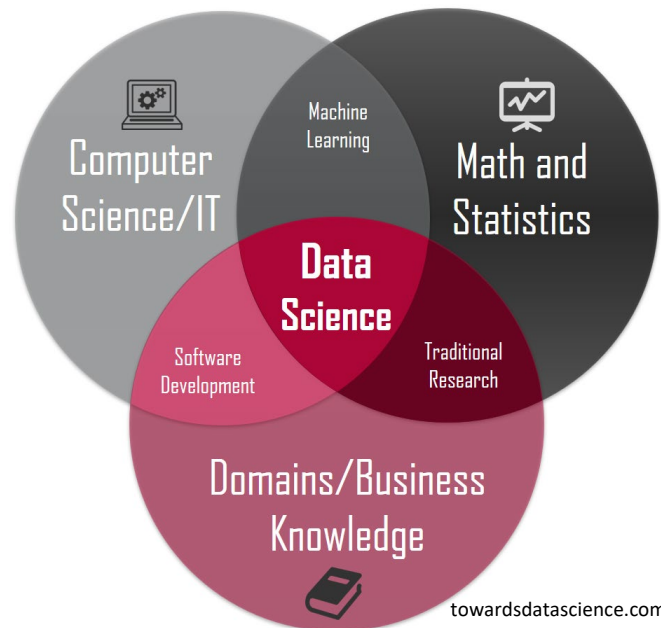
Data is unrefined and cannot really be used

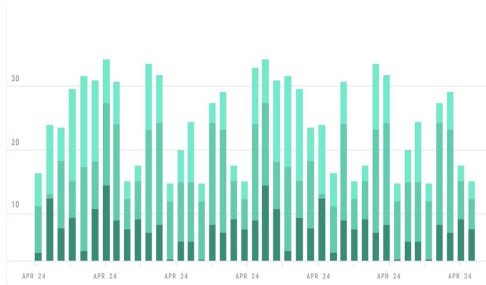
Data needs to be changed into useful form

Manage, Analyze and Model

What is Data Science?

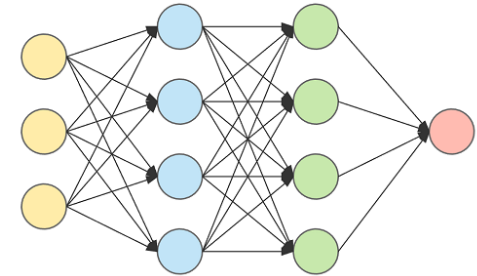
- Data science is a process of extracting knowledge and insight, and transforming hypothesis to actionable predictions from a huge and diverse set of data through **managing**, **analyzing** and **modelling** using various statistical and computing methods





Visualization

Graphical representation of data to understand and communicate information



Machine Learning

Subset of AI to learn and identify patterns in data without human assistance

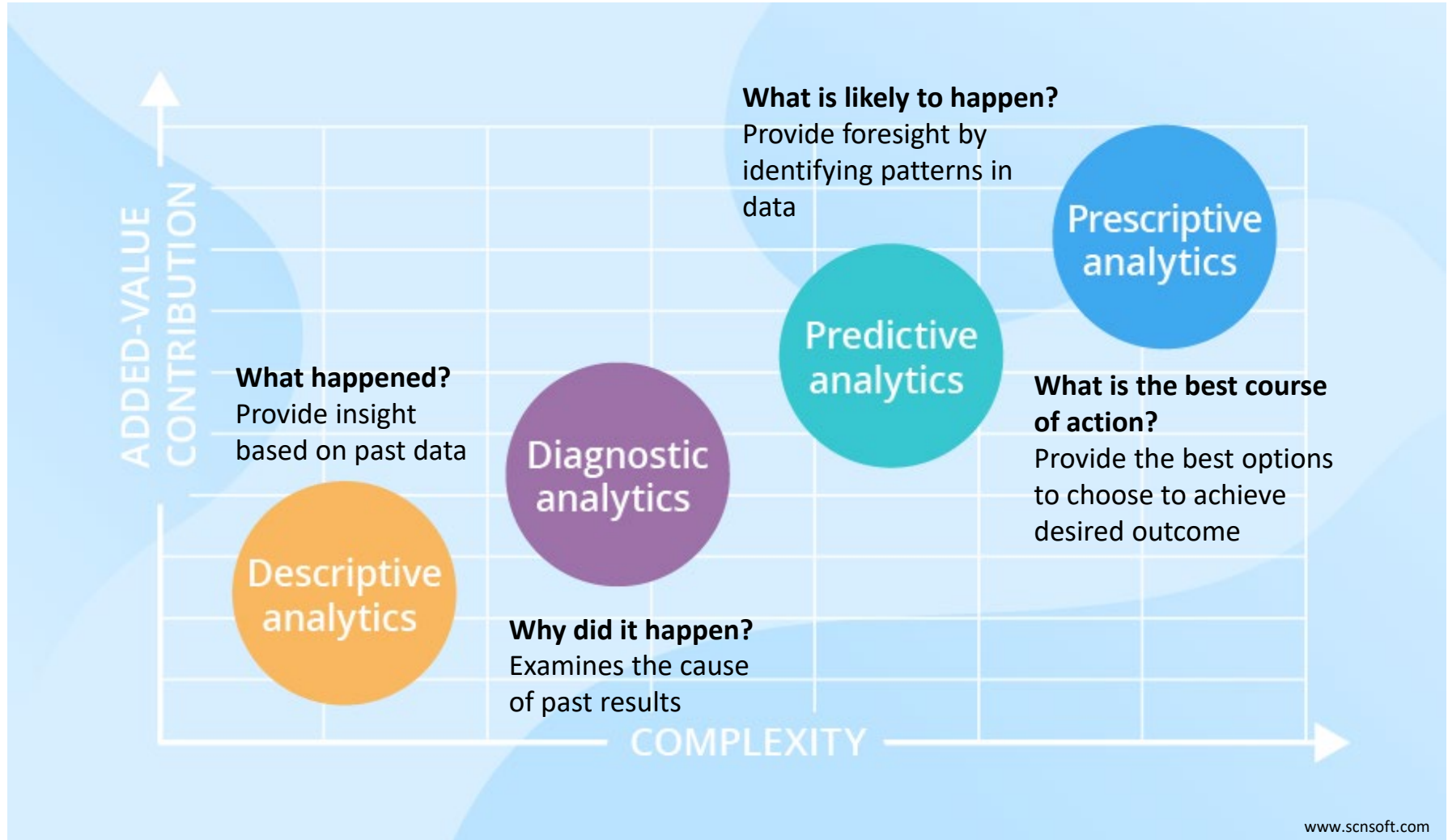
Data Science

Statistics

$$s^2 = \frac{\sum (X - \bar{X})^2}{N - 1}$$

Methods of analyzing and interpreting data

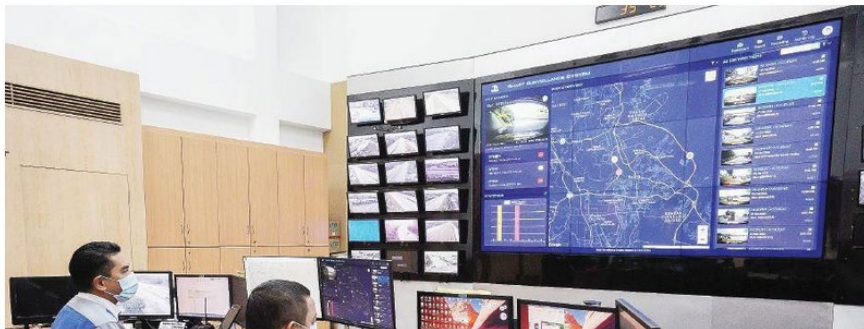
Types of Data Analytic



Applications of Data Science

#TECH: Towards safer highways

By Hanum Afandi - January 18, 2021 @ 3:18pm



Better monitoring

Through the implementation of S3, the level of highway efficiency will be upgraded and the safety of drivers improved. S3 enables the monitoring and detection of accidents, foreign objects, wild animals, potholes, surface cracks and ponding. The system also covers problems such as water spots, guard rail and slope failure, liquid spillage and road signage damage.

It combines technologies like AI and machine learning to provide notification to the relevant parties for further action. Since the launch of S3 on 19 August 2020, 1,303 incidents were detected in the first month alone. So far, the S3 has helped operations in carrying out immediate rectification with the real-time notifications. Fifty per cent of surface damage and highway asset damage were detected by the system and repairs were made immediately.

Improving Safety

To improve security and safety, the company uses the Artificial Intelligence System Analytics (Aisya). By leveraging dashboard cameras and computers installed in every highway patrol car, it is able to obtain images of damage and accidents immediately.

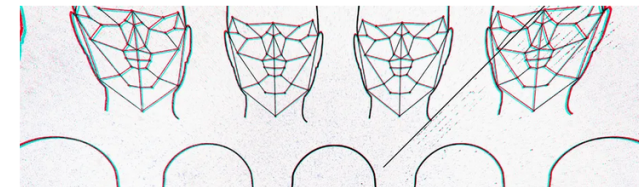
POLICY TECH PRIVACY

Most US government agencies are using facial recognition

A new GAO report finds 19 agencies are using some form of the technology

By Russell Brandom | Aug 25, 2021, 1:23pm EDT

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Applications of Data Science

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The latest from Google Research

Applying Deep Learning to Metastatic Breast Cancer Detection

Friday, October 12, 2018

Posted by Martin Stumpe, Technical Lead and Craig Mermel, Product Manager, Healthcare, Google AI

A pathologist's microscopic examination of a tumor in patients is considered the gold standard for cancer diagnosis, and has a profound impact on prognosis and treatment decisions. One important but laborious aspect of the pathologic review involves detecting cancer that has spread ([metastasized](#)) from the primary site to nearby lymph nodes. Detection of nodal metastasis is relevant for most cancers, and forms one of the foundations of the widely-used [TNM cancer staging](#).

In breast cancer in particular, nodal metastasis influences treatment decisions regarding radiation therapy, chemotherapy, and the potential surgical removal of additional lymph nodes. As such, the accuracy and timeliness of identifying nodal metastases has a significant impact on clinical care. However, studies have shown that [about 1 in 4](#) metastatic lymph node staging classifications would be changed upon second pathologic review, and detection sensitivity of small metastases on individual slides can be [as low as 38%](#) when reviewed under time constraints.

AI breast cancer screening project wins government funding for NHS trial

by [Ryan O'Hare](#)

16 June 2021



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The partnership, which includes Imperial College London, [Google Health](#), Imperial College Healthcare NHS Trust, St George's Hospitals NHS Foundation Trust, and the Royal Surrey NHS Foundation Trust builds on previous work, in which the researchers trained the algorithm on [depersonalised patient data and mammograms from patients](#) in the UK and US.

The findings, [published in Nature](#) in January 2020, showed the AI system was able to correctly identify cancers from the images with a similar degree of accuracy to expert radiologists, and demonstrated potential to assist clinical staff in practice.

LATEST NEWS

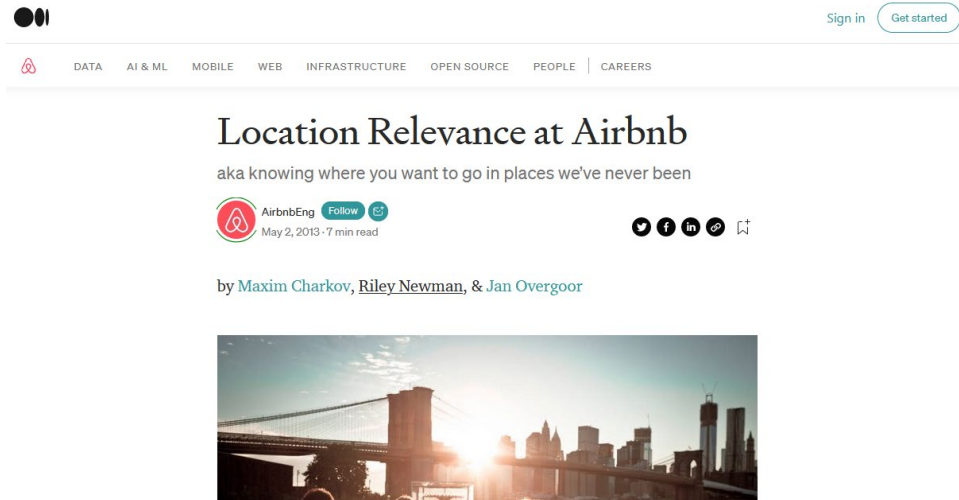


THE KIDS AREN'T ALRIGHT
Children will face huge increases in extreme climate events in their lifetimes

IMPACT GRADUATION
Imperial's flagship BAME talent programme celebrates its seventh cohort

Healthcare Services

Applications of Data Science



Recommendation Systems

Applications of Data Science

- Is there any relationship between weather and sales?
- Is weather is influencing the sales of your shop?



Humidity



Temperature



Sales



Quantity Sold

Applications of Data Science

- Can we detect a person with depression disorder?



Social Media Activity



Medical Data

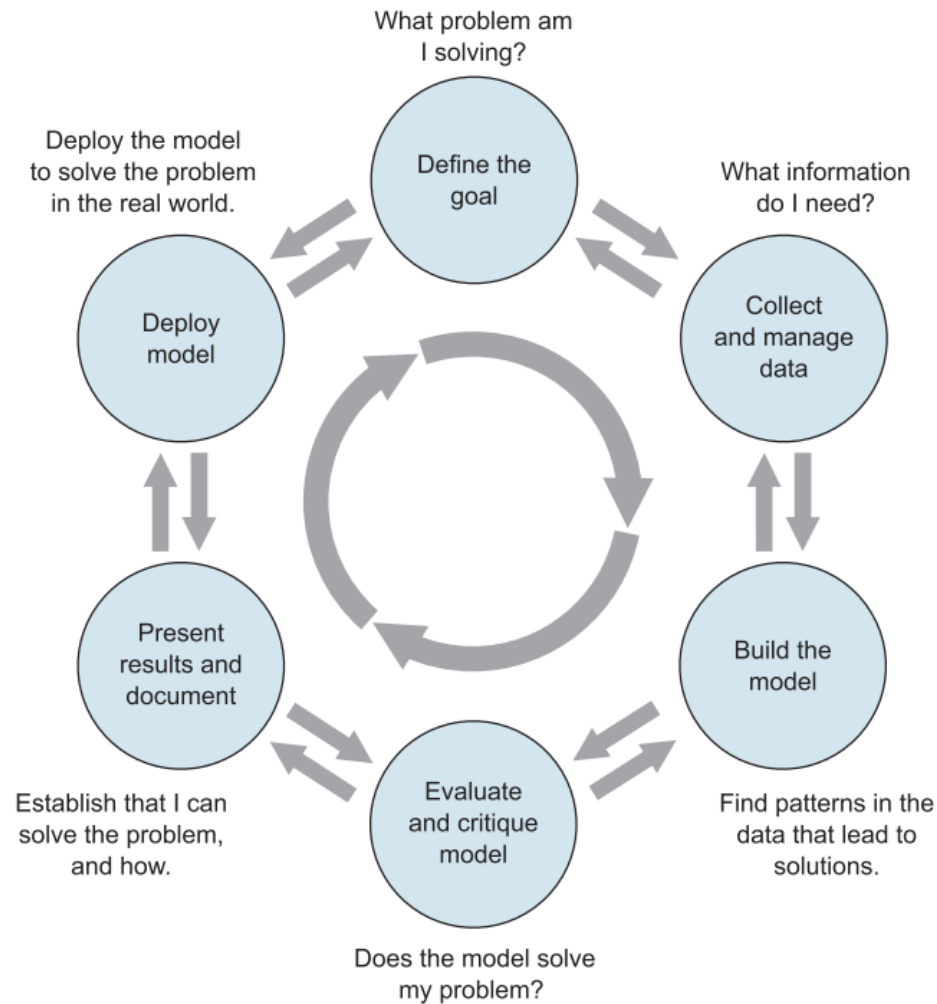


Person with and
without Depression

The Roles in Data Science Projects

Project sponsor	Client	Data scientist	Data architect	Operations
<ul style="list-style-type: none">• Person who wants the result• Decide the project success/failure• Keep them updated	<ul style="list-style-type: none">• The end user• Has interest in your model• Who will be using your model	<ul style="list-style-type: none">• Design the project steps• Apply the process• Pick the statistic and machine learning techniques that will be used	<ul style="list-style-type: none">• Responsible for data and its storage• Manage data warehouse• e.g. database administrator	<ul style="list-style-type: none">• Responsible for acquiring data and deliver the final results• Responsible for deployment

Stages of a Data Science Project



Defining the Goal

- Discuss and work with stakeholders/sponsor to understand and identify business problems
- An online shop is making losses
 - Products – which products are not selling well?
 - Customers – how to identify customers who are more likely to buy?
 - Fraudulent orders – how to identify fraudulent orders?

Defining the Goal

- Define a **specific**, **quantifiable** and **achievable** goal
- "The detection accuracy rate must be at least 85%"

Defining the Goal

- Specific goal allows stopping condition and acceptance criteria to be defined
- Otherwise the project will go unbounded

Data Collection and Management

- The most time-consuming step in the process
- Most important and crucial step

Data Collection and Management

- Identify data that is relevant to the question
- What are the attributes that are related to the target?
- Do you have the attributes that are related to the target?

Data Collection and Management

- Number of products in the order – unusually large orders could be fraudulent
- Type of products in the order – wary of orders that are uncommonly purchased together.
- Billing/Shipping address of the order – not a real location or not a residential location
- Number of receiving orders in a timeframe – placing multiple orders at the same time

Data Collection and Management

- Do you need additional attributes to address problem?
- Do you have sufficient examples or not?

Data Collection and Management

- Remove redundant and irrelevant examples
- The collected data often needs to be cleaned from missing values and outliers,

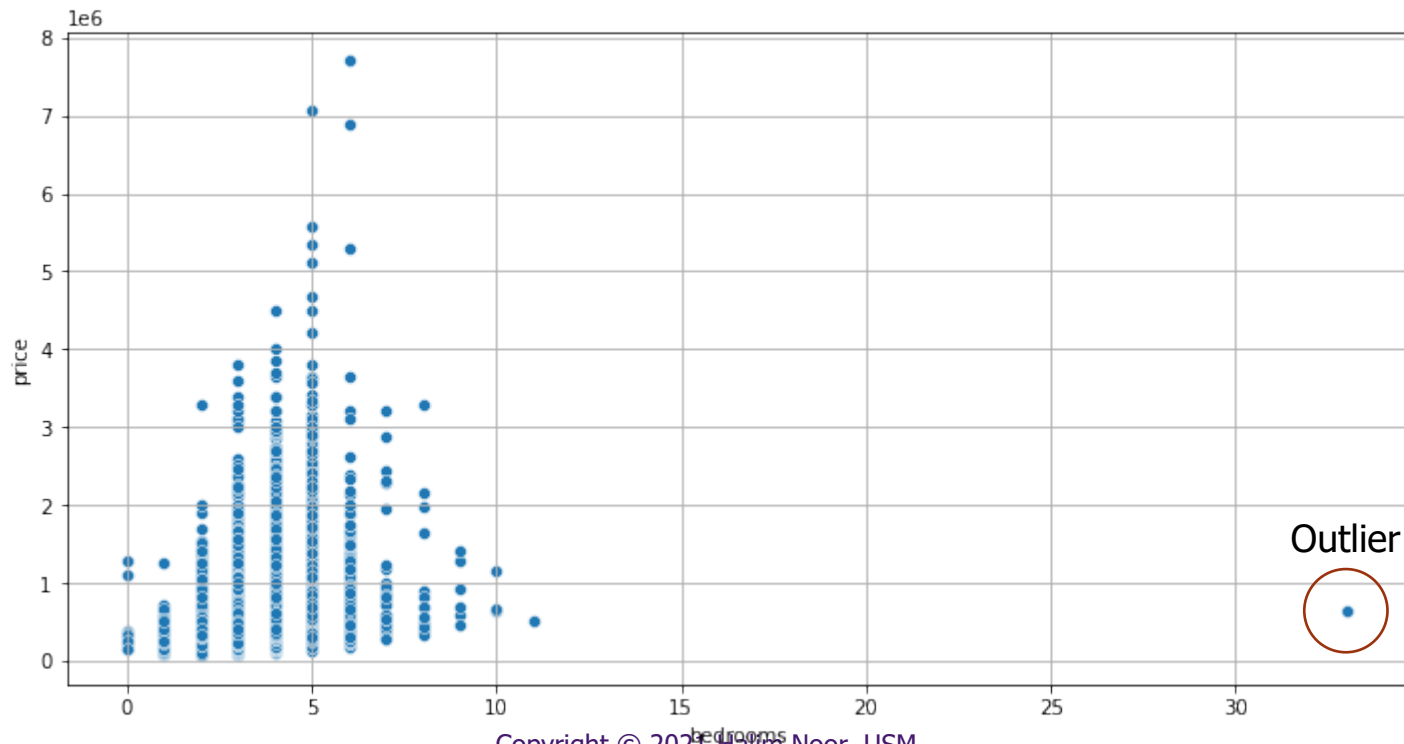
Data Collection and Management

- An attribute of an example (row) has no value
- Are there missing values? How do handle them?

kc_house_data_rev															
	X	id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	grade	sqft_above	sq
1	0	7129300520	20141013T000000	221900	3	1.00	1180	5650	1.0	0	0	3	7	1180	
2	1	6414100192	20141209T000000	538000	3	2.25	2570	7242	2.0	0	0	3	7	2170	
3	2	5631500400	20150225T000000	180000	2	1.00	770	10000	1.0	0	0	3	6	770	
4	3	2487200875	20141209T000000	604000	4	3.00	1960	5000	1.0	0	0	5	7	1050	
5	4	1954400510	20150218T000000	510000	3	2.00	1680	8080	1.0	0	0	3	8	1680	
6	5	7237550310	20140512T000000	1225000	4	4.50	5420	101930	1.0	0	0	3	11	NA	
7	6	1321400060	20140627T000000	257500	3	2.25	1715	6819	2.0	0	0	3	7	1715	
8	7	2008000270	20150115T000000	291850	3	1.50	1060	9711	1.0	0	0	3	7	1060	
9	8	2414600126	20150415T000000	229500	3	1.00	1780	7470	1.0	0	0	3	7	1050	
10	9	3793500160	20150312T000000	323000	3	2.50	1890	NA	2.0	0	0	3	7	1890	
11	10	1736800520	20150403T000000	662500	NA	2.50	3560	9796	1.0	0	0	3	8	1860	
12	11	9212900260	20140527T000000	468000	2	1.00	1160	6000	1.0	0	0	4	7	860	
13	12	114101516	20140528T000000	310000	3	1.00	1430	19901	1.5	0	0	4	7	1430	
14	13	6054650070	20141007T000000	400000	3	1.75	1370	9680	1.0	0	0	4	7	1370	
15	14	1175000570	20150312T000000	530000	5	2.00	1810	4850	1.5	0	0	3	7	1810	
16	15	9297300055	20150124T000000	650000	4	3.00	2950	5000	2.0	0	3	3	9	1980	
17	16	1875500060	20140731T000000	395000	3	2.00	1890	14040	2.0	0	0	3	7	1890	
18	17	6865200140	20140529T000000	485000	4	1.00	1600	4300	1.5	0	0	4	7	1600	
19	18	16000397	20141205T000000	189000	2	1.00	1200	9850	1.0	0	0	4	7	1200	
20	19	7983200060	20150424T000000	230000	3	1.00	1250	9774	1.0	0	0	4	7	1250	

Data Collection and Management

- Values that are significantly differ from others
- Are there outliers? How do we handle them? Should we remove it or keep it?



Data Modeling

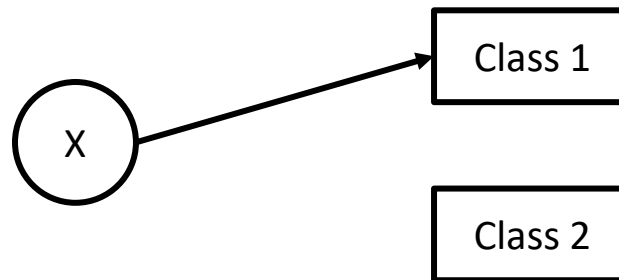
- To confirm our insight and hypothesis about the data
 - Number of rooms and House prices – positive relationship
- Fit the data using linear model
 - If slope is positive = positive relationship
 - If slope is negative = negative relationship

Data Modeling

- A model is an approximation of the data that describes the relationship between the attributes
- A model can be used to make predictions

Data Modeling

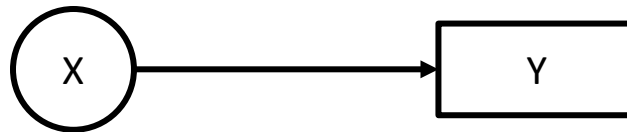
- Classification – deciding if something belongs to one category or another
- Placed order and Fraudulent or Not Fraudulent



X is an order, fraudulent order (Class 1) or valid order (Class 2)
X is social media activity, depression (Class 1) or not (Class 2)

Data Modeling

- Regression (scoring) – estimating a numeric value

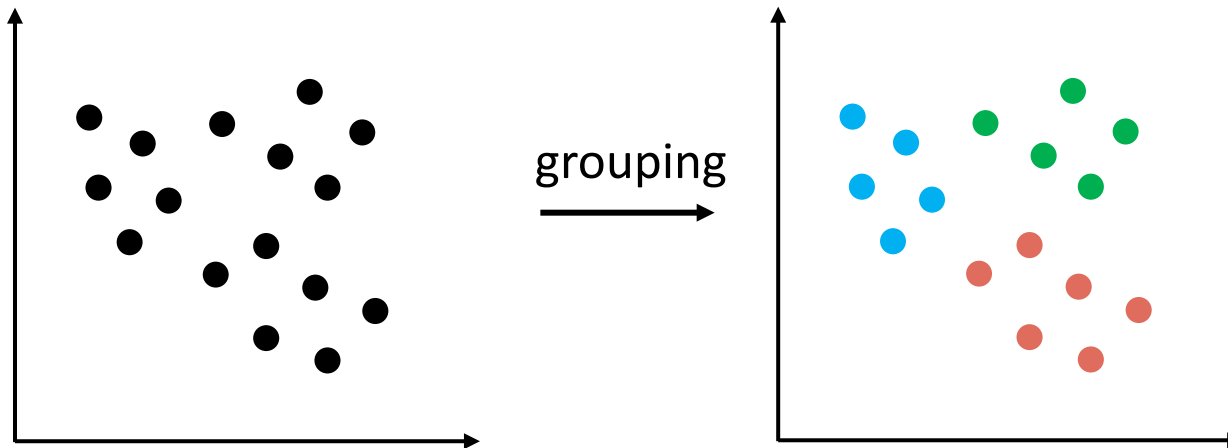


X is house attributes, house price (Y)

X is weather, sales (Y)

Data Modeling

- Clustering – grouping items into most similar groups



Model Evaluation

- Determine if the model meets the goals
- Is it **accurate** enough for your needs? Does it **perform better** than whatever estimate that is currently being used?
- Do the results of the model **make sense** in the context of the problem domain?
- If no, then repeat the Modeling step (or the steps before it)

Presentation and Documentation

- Document the model for those who are responsible for using, running, and maintaining the model once it has been deployed
- Presentation must **not** be technical
- Presentation must **make sense** to the human brain and **easy to understand** – use visualization
- Highlight the most **interesting findings** or recommendation (if any)

Model Deployment

- Ensure the model run smoothly
- Model can be updated when needed
- Monitor the performance of the model
- Why the model's decision is being overridden frequently?
- Is the model incomplete?

Summary

- Data science is a process of extracting knowledge or insight from data
- Data science project involves many roles and skills – back-and-forth between data scientist and project stakeholders
- Project goal must be specific, measurable and quantifiable

End