



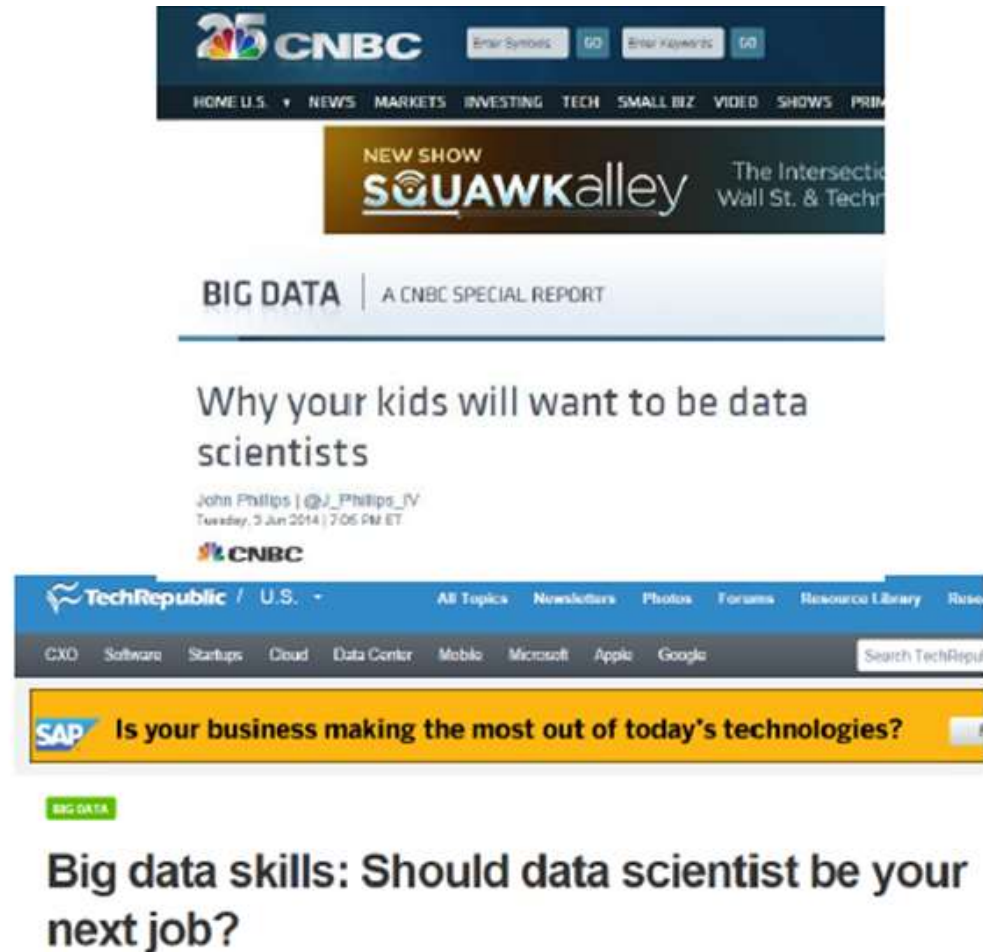
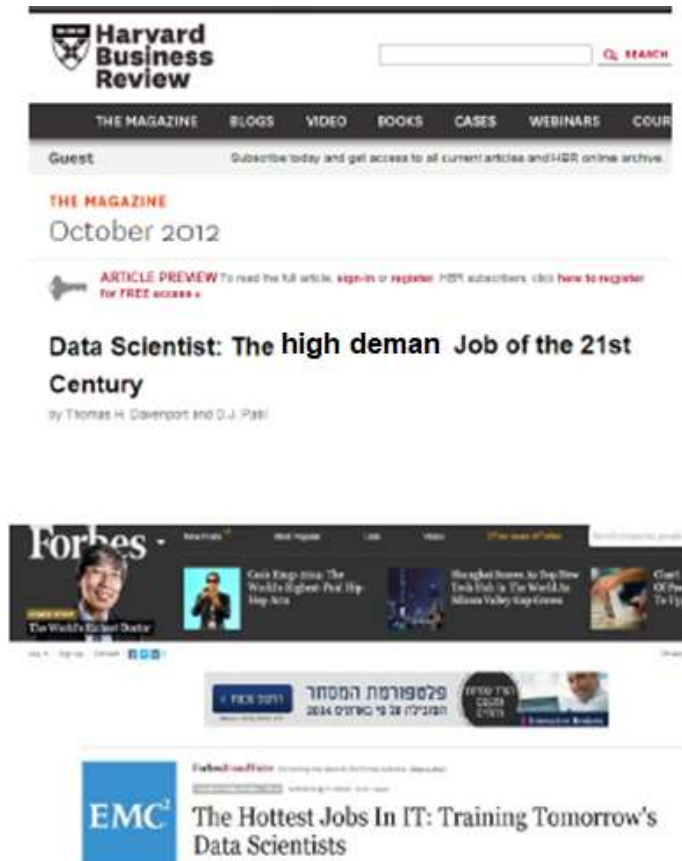
Big Data Analytics

Dr. Muhammad Affan Alim



Data Scientists are in high demand

-



Also in academia

WHITE HOUSE TO UNIVERSITIES: WE NEED MORE DATA SCIENTISTS

NEW YORK UNIVERSITY, UNIVERSITY OF CALIFORNIA-BERKELEY, AND THE UNIVERSITY OF WASHINGTON ARE LAUNCHING A \$37.8 MILLION PROJECT TO BOOST THE NUMBERS OF AMERICAN DATA SCIENTISTS

BY NEAL UNGERLEIDER

It's official: America needs more data scientists. This week, a \$37.8 million project

Berkeley Research

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Must Science & Society Journal Volume 88, Issue 4



How Big Data Can Transform Society for the Better

The digital traces we leave behind each day reveal more about us than we know. This could become a privacy nightmare—or it could be the foundation of a healthier, more prosperous world.

By Neal Ungerleider



RESEARCH CENTERS IN THE FIELD OF DATA SCIENCE

Center for Data Science (CDS)

The NYU Center for Data Science (CDS) is a first-of-its-kind for New York University's university-wide research in data science. It was established to help advance NYU's goal of creating the country's leading data science training and research for IBM, among other firms and professionals with tools to harness the power of big data.

LEARN MORE

Center for the Promotion of Research Involving Innovative Statistical Methodology (CPRIISM)

The Center for the Promotion of Research Involving Innovative Statistical Methodology (CPRIISM) is a new center dedicated to improving the quality of research in quantitative social, behavioral, and health and policy studies.

500k

The world's 500,000+ data science jobs are large enough to fill 10,000 football fields. (Source: Analytics)

75%

75% of digital information is generated by academic, which represents a huge opportunity for 80% of digital data to come from the U.S. (Source: Analytics)

UNIVERSITY of WASHINGTON

eScience Institute
Supporting Data-Driven Discovery in All Fields

WHO WE ARE

New Ph.D. Tracks in "Big Data"

Pays Well

Big Data, Big Paycheck

Median salary for analytics professionals and those specifically within data science, by level of experience.



Note: Data do not include managers Source: Burtch Works

The Wall Street Journal

Data Science: Why all the Excitement?




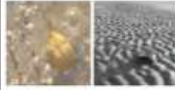































e.g.,
Google Flu Trends:

Detecting outbreaks
two weeks ahead
of CDC data

New models are estimating
which cities are most at risk
for spread of the Ebola virus.

The unreasonable effectiveness of Deep Learning (CNNs)

- 2012 Image-net challenge: Classify 1 million images into 1000 classes.

 Cliff dwelling L2 11.0% - Mah. 99.9%	 horseshoe crab 0.99%	 African elephant 0.99%	 mongoose 0.94%	 Indian elephant 0.88%	 dingo 0.87%	L2
	 cliff 0.07%	 dam 0.00%	 stone wall 0.00%	 brick 0.00%	 castle 0.00%	Mah.
 Gondola L2 4.4% - Mah. 99.7%	 shopping cart 1.07%	 unicycle 0.84%	 covered wagon 0.83%	 garbage truck 0.79%	 forklift 0.78%	L2
	 dock 0.11%	 canoe 0.03%	 fishing rod 0.01%	 bridge 0.01%	 boathouse 0.01%	Mah.
 Palm L2 6.4% - Mah. 98.1%	 crane 0.87%	 stupa 0.83%	 roller coaster 0.79%	 bell cote 0.78%	 flagpole 0.75%	L2
	 cabbage tree 0.81%	 pine 0.30%	 pandanus 0.14%	 iron tree 0.07%	 logwood 0.06%	Mah.

Where does data come from?



“Big Data” Sources

It's All Happening On-line



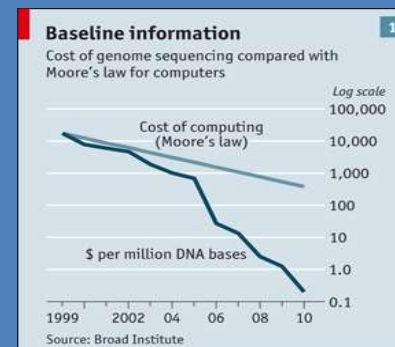
Every:
Click
Ad impression
Billing event
Fast Forward, pause,...
Server request
Transaction
Network message
Fault
...



Internet of Things / M2M



Health/Scientific Computing

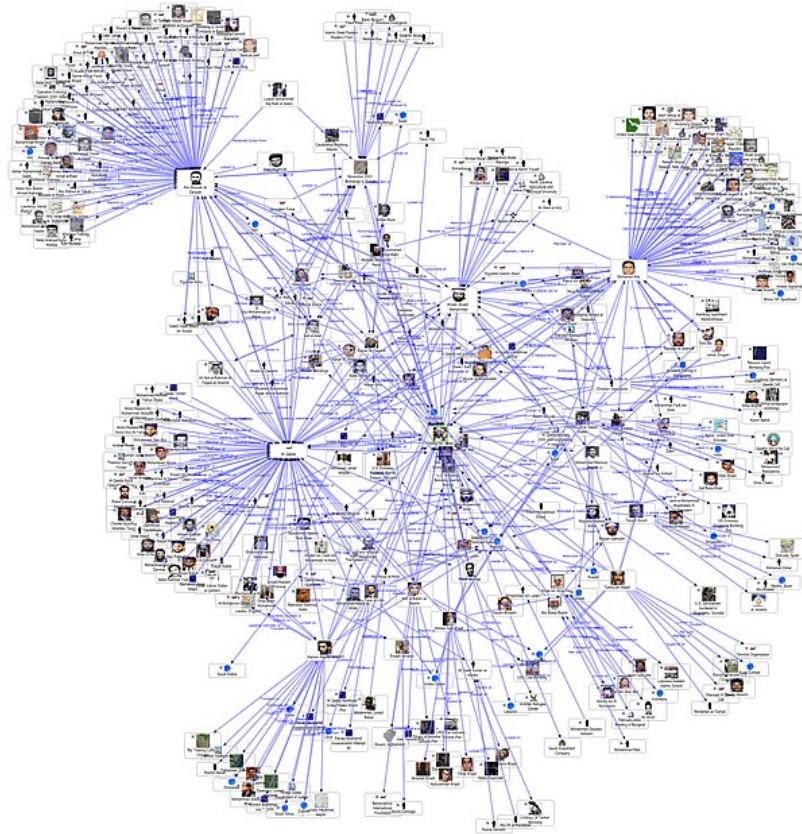


Graph Data

Lots of interesting data has a graph structure:

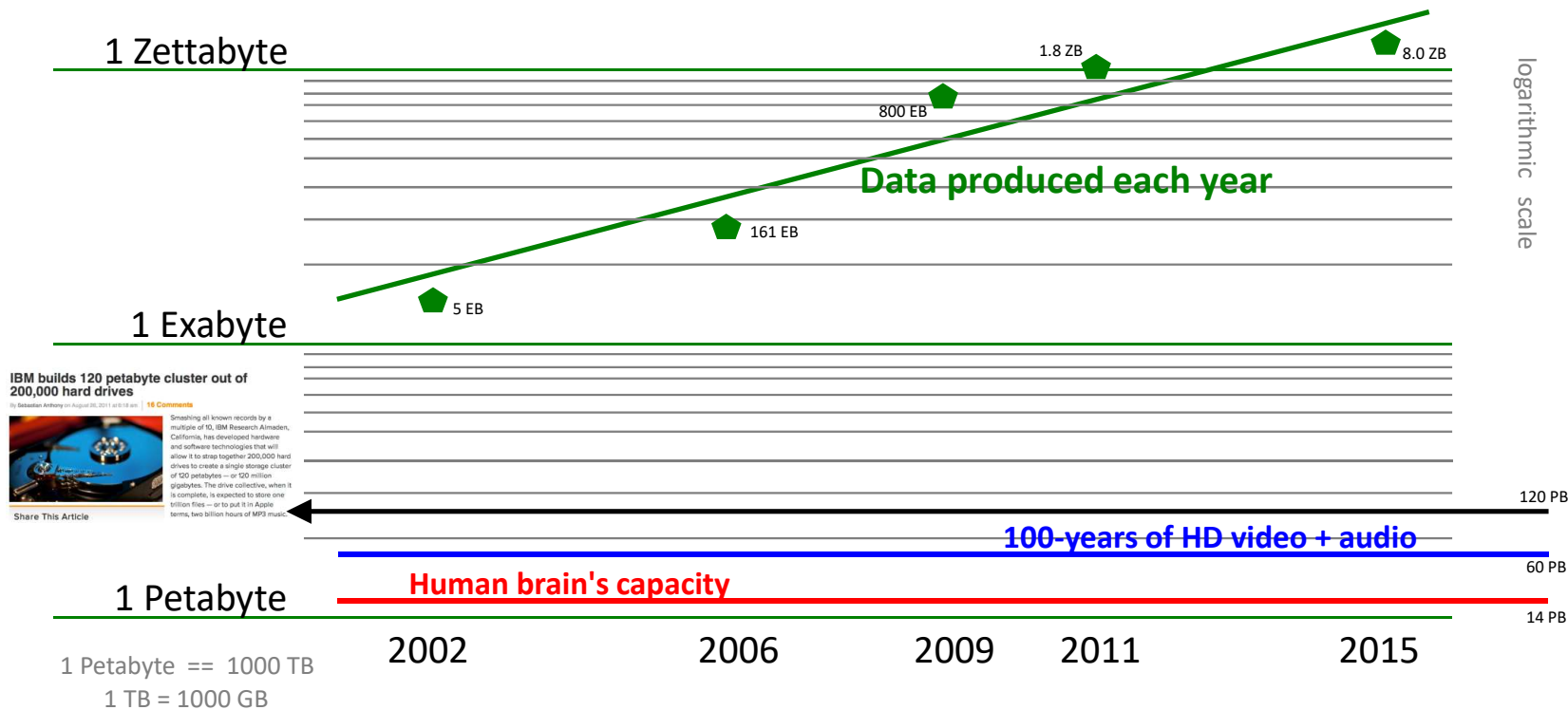
- Social networks
- Communication networks
- Computer Networks
- Road networks
- Citations
- Collaborations/Relationships
- ...

Some of these graphs can get quite large (e.g., Facebook* user graph)



Data, data everywhere...

I here's certainly a lot of it!



References

(2015) 8 ZB: <http://www.emc.com/collateral/analyst-reports/idc-extracting-value-from-chaos-ar.pdf>

(2011) 1.8 ZB: <http://www.emc.com/leadership/programs/digital-universe.htm>

(2009) 800 EB: <http://www.emc.com/collateral/analyst-reports/idc-digital-universe-are-you-ready.pdf>

(2006) 161 EB: <http://www.emc.com/collateral/analyst-reports/expanding-digital-idc-white-paper.pdf>

(2002) 5 EB: <http://www2.sims.berkeley.edu/research/projects/how-much-info-2003/execsum.htm>

(life in video) 60 PB: in 4320p resolution, extrapolated from 16MB for 1:21 of 640x480 video (w/sound) – almost certainly a gross overestimate, as sleep can be compressed significantly!

(brain) 14 PB: <http://www.quora.com/Neuroscience-1/How-much-data-can-the-human-brain-store>

Data Science – A Definition

- **Data Science** is the science which uses computer science, statistics and machine learning, visualization and human-computer interactions to collect, clean, integrate, **analyze, visualize, interact with data to** create data products.

Goal of Data Science

Turn **data** into **data products**

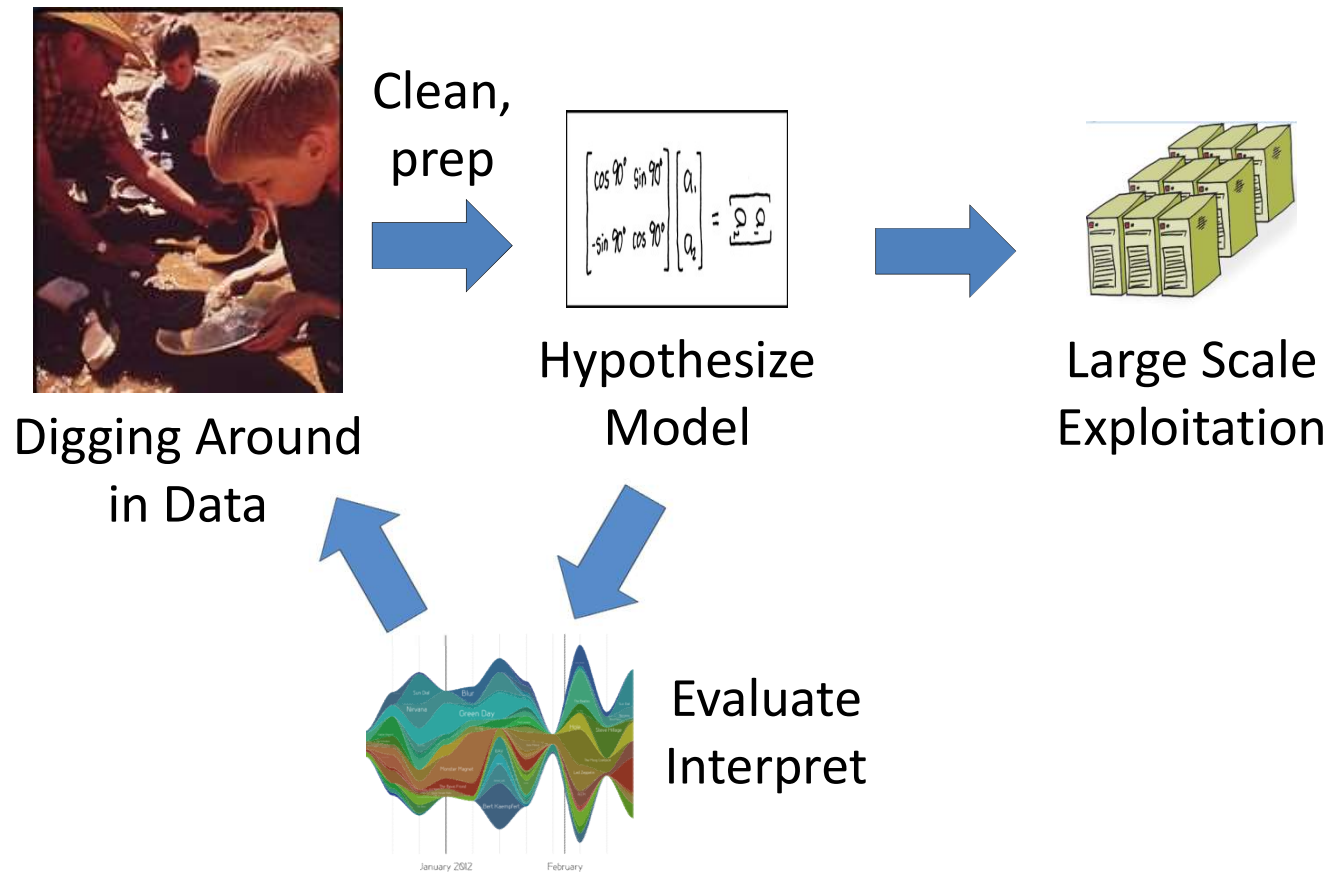
How to use data?

Data => exploratory analysis => knowledge models => product / decision making

Data => predictive models => evaluate / interpret => product / decision making

Data Scientist's Practice

-



Example data science applications

Marketing: predict the characteristics of high life time value (LTV) customers, which can be used to support customer segmentation, identify upsell opportunities, and support other marketing initiatives

Logistics: forecast how many of which things you need and where will we need them, which enables learn inventory and prevents out of stock situations

Healthcare: analyze survival statistics for different patient attributes (age, blood type, gender, etc.) and treatments; predict risk of re-admittance based on patient attributes, medical history, etc.

Example data science applications cont...

Healthcare: analyze survival statistics for different patient attributes (age, blood type, gender, etc.) and treatments; predict risk of re-admittance based on patient attributes, medical history, etc.

Some more examples

Transaction Databases → Recommender systems (NetFlix), Fraud Detection (Security and Privacy)

Wireless Sensor Data → Smart Home, Real-time Monitoring, Internet of Things

Text Data, Social Media Data → Product Review and Consumer Satisfaction (Facebook, Twitter, LinkedIn), E-discovery

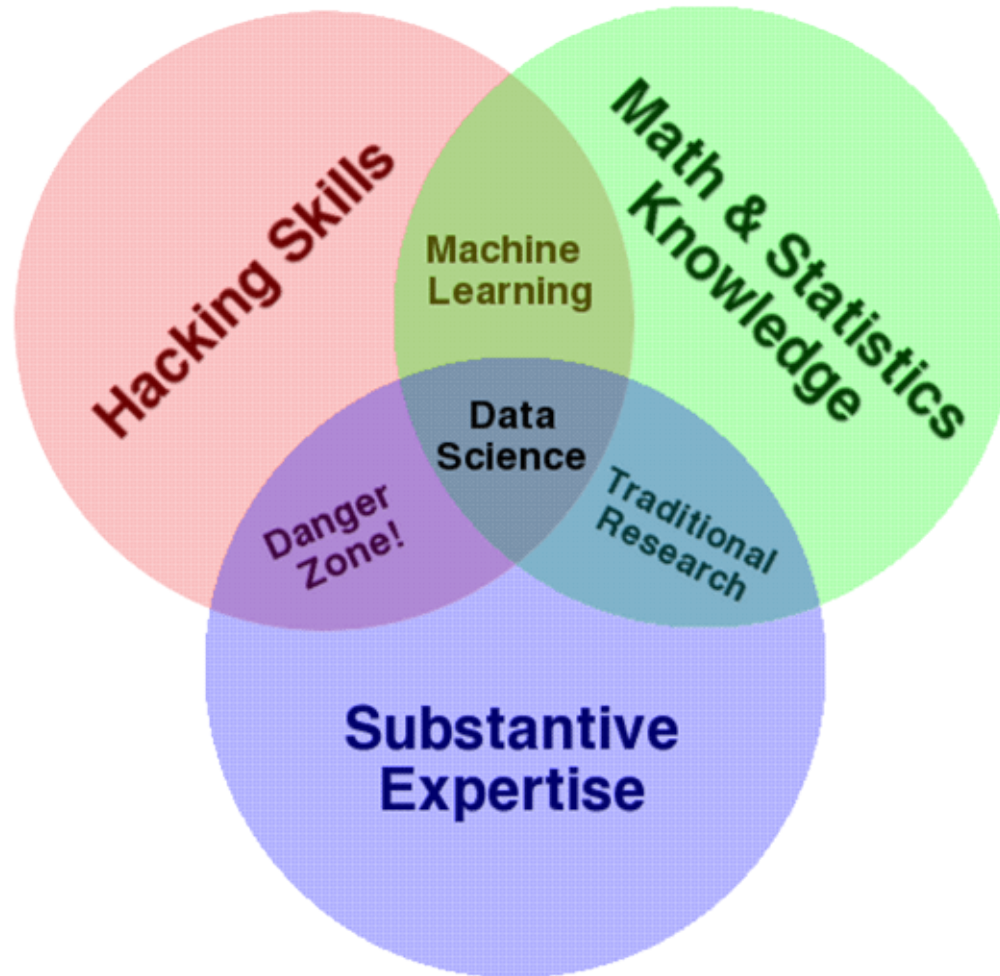
Software Log Data → Automatic Trouble Shooting (Splunk)

Genotype and Phenotype Data → Epic, 23andme, Patient-Centered Care, Personalized Medicine




Data Science – One Definition

-



What's Hard about Data Science

- Overcoming assumptions
 - Making ad-hoc explanations of data patterns
 - Overgeneralizing
 - Communication
 - Not checking enough (validate models, data pipeline integrity, etc.)
 - Using statistical tests correctly
 - Prototype → Production transitions
 - Data pipeline complexity (who do you ask?)
- 

About the course

- A mixture of theory and practice
- Introductory, broad overview of subjects
- Focus on practical aspects, but not on ever-changing technology and tools
- Seminar style - I am here to learn as well as to teach
- Language choice: python

Relatively easy to learn (for computer scientist) compared to R (more popular among statisticians)

Open source means easy access (as opposed to SAS or MATLAB)

[Which one is more frequently used in data science?](#)

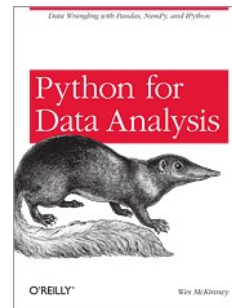
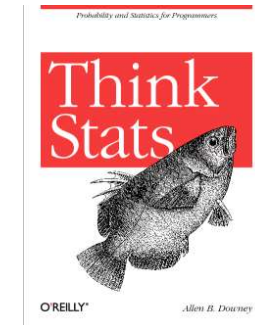
Textbook

- **Required:** Python Data Science Handbook (**PDSH**)
- by Jake VanderPlas
- **Optional:**

Data Science from Scratch (**DSS**) by Joel Grus

Python for Data Analysis (**PDA**) by Wes McKinney

Free e-book: Think Stats (**TS**) by Allen B. Downey.



Brief introduction of Python

- Invented in the Netherlands, early 90s by Guido van Rossum
- Open sourced from the beginning
- Considered a scripting language, but is much more

No compilation needed

Scripts are evaluated by the interpreter, line by line

Functions need to be defined before they are called

Installing the Anaconda

- Watch the video
- <https://www.youtube.com/watch?v=G3Lt1JWBvL8>



Logout

```
In [6]: from platform import python_version  
print(python_version())
```

3.7.3

In []:

Introduction to NumPy

- NumPy (short for Numerical Python) provides an efficient interface to **store and operate** on dense data buffers.
- NumPy arrays from the core of nearly the entire **ecosystem** of data science tools in Python
- If you followed the installation the **Anaconda stack**, you already have NumPy

Jupyter NumPy Practice Part

Last Checkpoint: 41 minutes ago (unsaved changes)



Logout

File Edit View Insert Cell Kernel Widgets Help

Trusted

Python 3



```
In [6]: from platform import python_version
        print(python_version())
```

3.7.3

```
In [15]: import numpy
         print(numpy.version.version)
```

1.16.4

NumPy cont...

- By convention, you'll find that most people in the SciPy/PyData world will import NumPy using np as an alias:
- In[2]: `import numpy as np`

Features

Example Data

Training Examples:

	Action	Author	Thread	Length	Where
e1	skips	known	new	long	Home
e2	reads	unknown	new	short	Work
e3	skips	unknown	old	long	Work
e4	skips	known	old	long	home
e5	reads	known	new	short	home
e6	skips	known	old	long	work

New Examples:

e7	???	known	new	short	work
e8	???	unknown	new	short	work

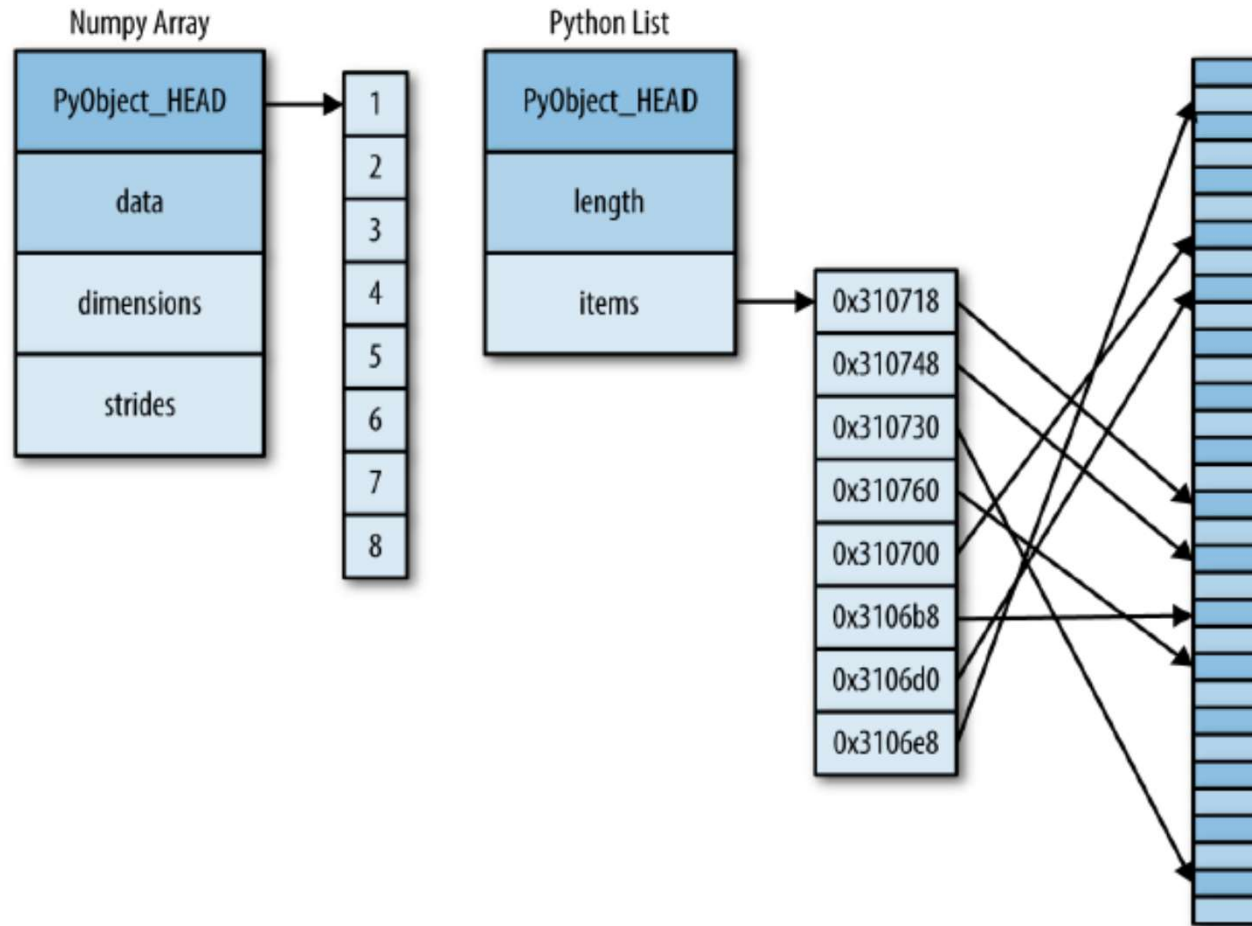
Why is NumPy Faster Than Lists?

- NumPy arrays are stored at one continuous place in memory unlike lists, so processes can access and manipulate them very efficiently.
- This behavior is called **locality of reference** in computer science.
- This is the main reason why NumPy is faster than lists. Also it is optimized to work with latest CPU architectures.

C

- The **Python list**, on the other hand, contains a pointer to a **block of pointers**, each of which in turn points to a full Python object like the Python integer.
- **Fixed-type NumPy-style** arrays lack this flexibility, but are much more efficient for **storing and manipulating** data

Difference between NumPy array and Python List



Creating arrays using NumPy

- First, we can use `np.array` to create arrays from **Python lists**:

```
In[8]: # integer array:  
       np.array([1, 4, 2, 5, 3])
```

```
Out[8]: array([1, 4, 2, 5, 3])
```

- Remember that unlike **Python lists**, **NumPy** is constrained to arrays that all contain the same type.

Creating arrays using NumPy cont...

- If types **do not match**, NumPy will **upcast** if possible (here, integers are upcast to floating point):


```
In[9]: np.array([3.14, 4, 2, 3])
```

```
Out[9]: array([ 3.14,  4.  ,  2.  ,  3.  ])
```

- If we want to explicitly set the data type of the resulting array, we can use the dtype keyword:

```
In[10]: np.array([1, 2, 3, 4], dtype='float32')
```

```
Out[10]: array([ 1.,  2.,  3.,  4.], dtype=float32)
```



Creating arrays using NumPy cont...

- NumPy arrays can explicitly be **multidimensional**; here's one way of initializing a **multidimensional array** using a list of lists:

```
In[11]: # nested lists result in multidimensional arrays  
        np.array([range(i, i + 3) for i in [2, 4, 6]])
```

```
Out[11]: array([[2, 3, 4],  
                [4, 5, 6],  
                [6, 7, 8]])
```

Data Science

Revision

Data Science – A Definition

- **Data Science** is the science which uses computer science, statistics and machine learning, visualization and human-computer interactions to collect, clean, integrate, **analyze, visualize, interact with data to** create data products.

Data Science - Definition

- Data Science is an interdisciplinary field about **processes** to extract the **knowledge or insights** and **predict** from data in various forms, either structured or unstructured



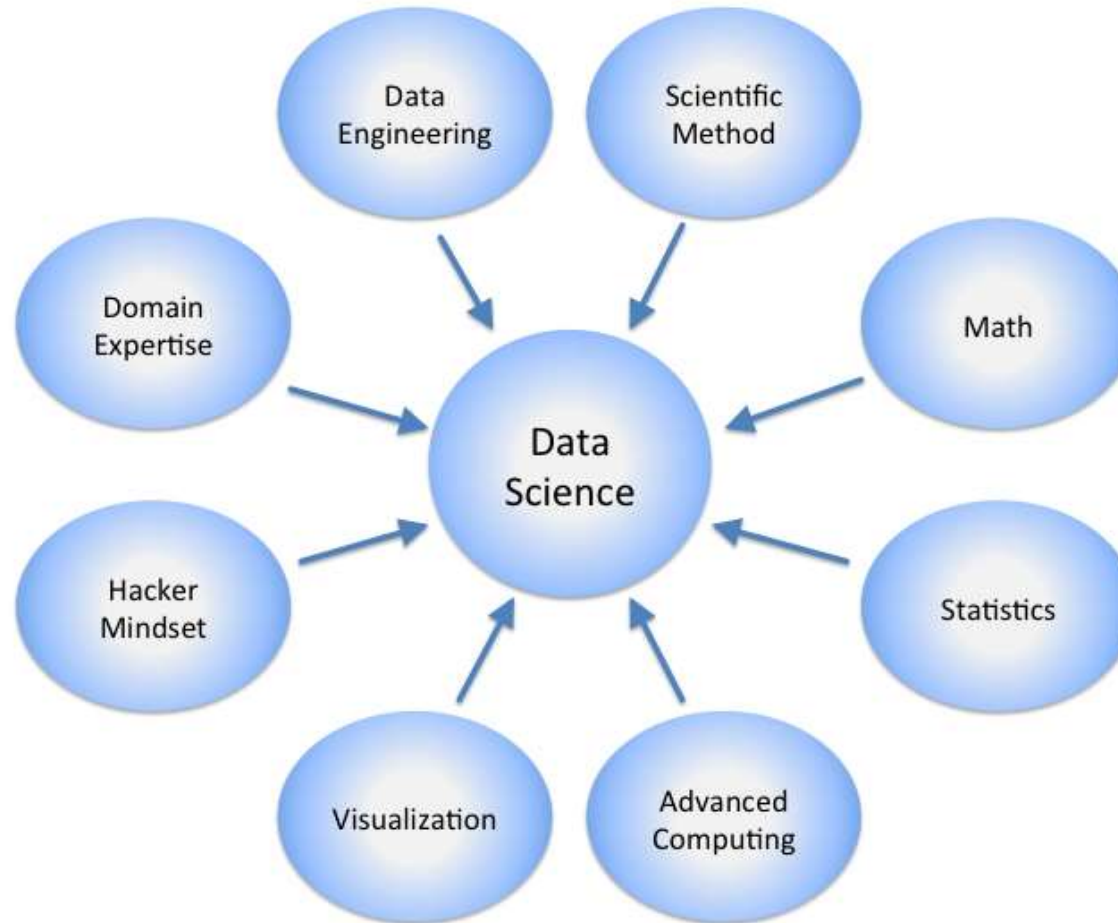
Data Analyst

- **Data analytics** is the science of examining raw data with the purpose of drawing conclusions about that information.
- **Data analytics** is used in many industries to allow companies and organization to make better decisions and in the sciences to verify or disprove existing models or theories

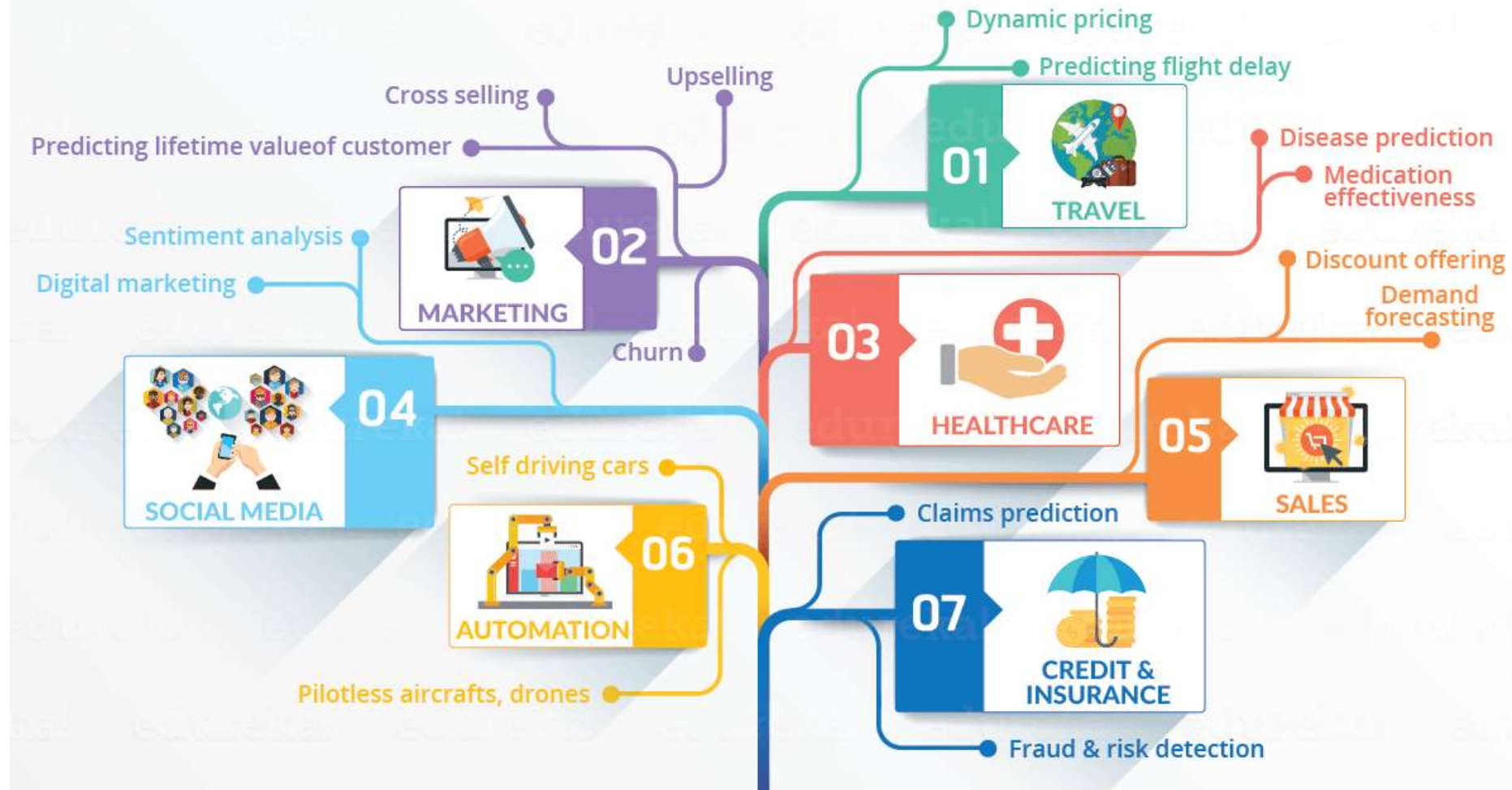


What to know of a data scientist

-



Targeted field of Data Science



Popular tools for Data Science

- **The Ten Most Common
Data Science Skills in Job Postings**

Skill	Percentage of Job Listings
Python	72%
R	64%
SQL	51%
Hadoop	39%
Java	33%
SAS	30%
Spark	27%
Matlab	20%
Hive	17%
Tableau	14%

Source: Glassdoor Economic Research.

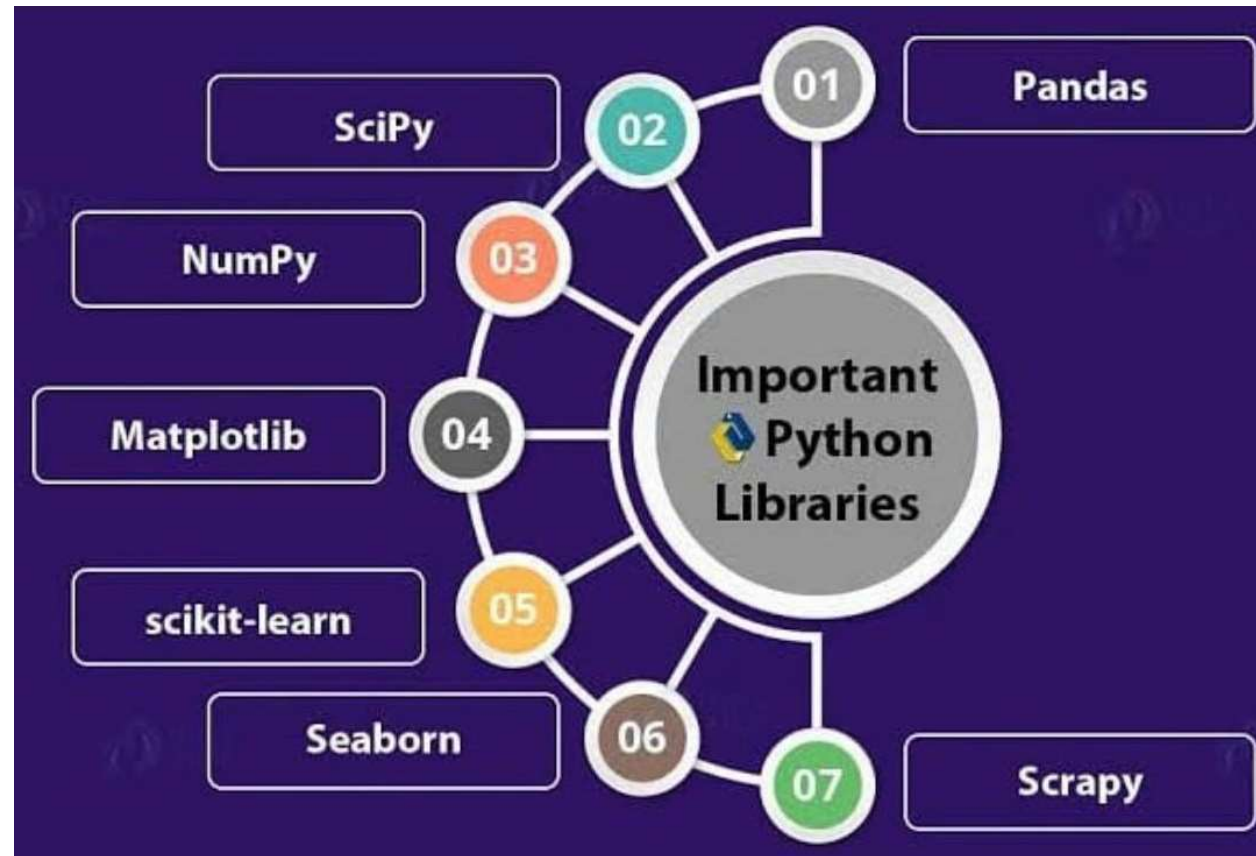
glassdoor®

Discussion of our courses

- Python for Data Science
- Feature Engineering
- Machine Learning towards Data Science
- No SQL for BIG Data
- Real Time Application for Data Science

Python for Data Science

- As requirement of job market, python is the top most popular language
- Due to this reason we have selected python in our course



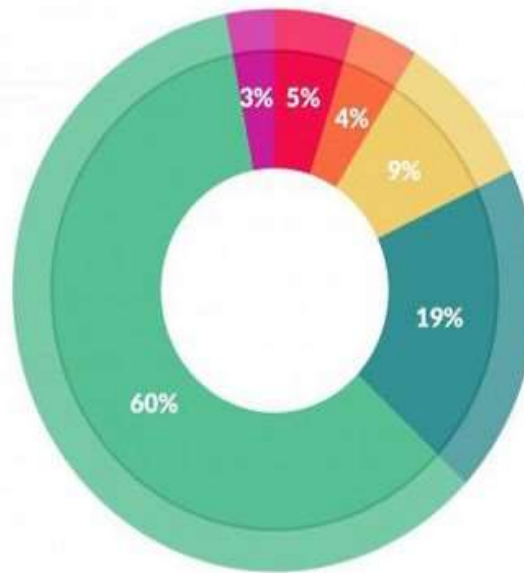
Feature Engineering

- Feature engineering is the process of using **domain knowledge** of the data to create features that make machine learning algorithms work
- Feature Engineering is a process of extracting useful feature from raw data using math, statistics and domain knowledge
-



Feature Engineering cont...

- Feature Engineering takes 80% energies of a modeling.
- Good feature engineering leads to good model insight and prediction



What data scientists spend the most time doing

- Building training sets: 3%
- Cleaning and organizing data: 60%
- Collecting data sets: 19%
- Mining data for patterns: 9%
- Refining algorithms: 4%
- Other: 5%

Feature Engineering cont...

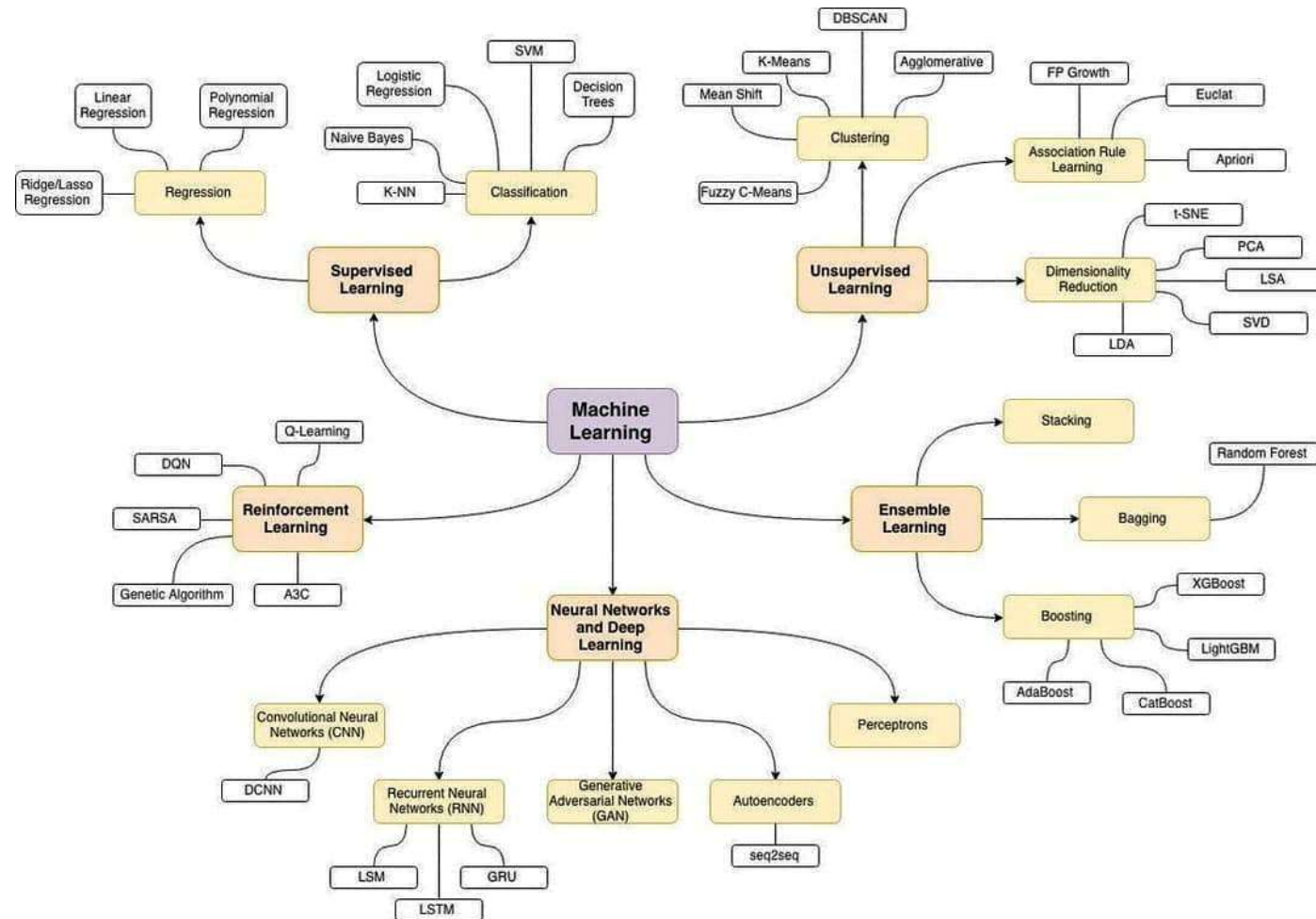
- **List of Techniques**

1. Imputation
2. Handling Outliers
3. Binning
4. Log Transform
5. One-Hot Encoding
6. Grouping Operations
7. Feature Split
8. Scaling
9. Extracting Date

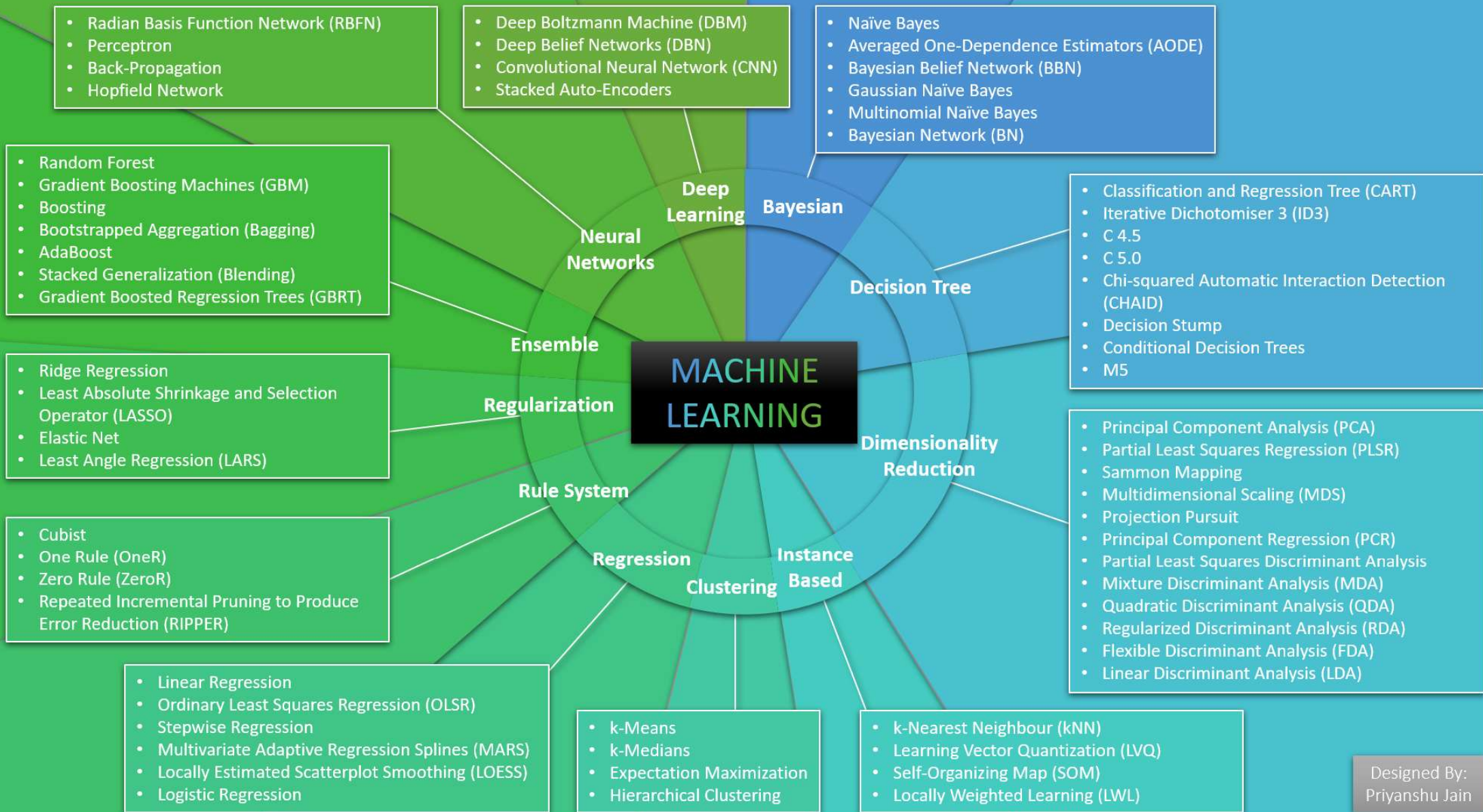
I think the best way to achieve expertise in feature engineering is practicing different techniques on various datasets and observing their effect on model performances

Machine Learning towards Data Science

- Machine learning is a backbone of Data Science.
- Even individual machine learning has good job market
- It helps modeling, feature extraction, feature reduction, etc.

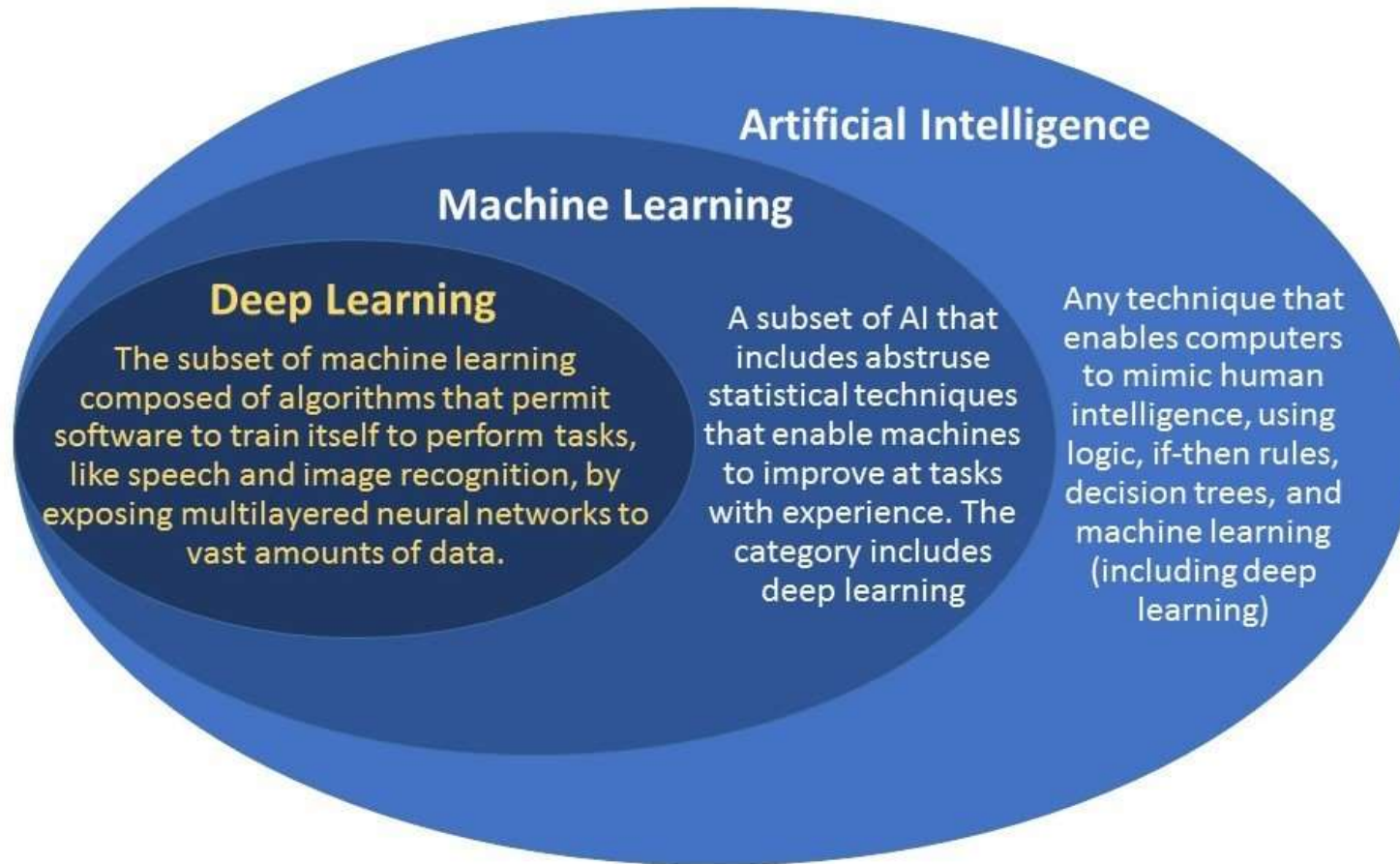


MACHINE LEARNING



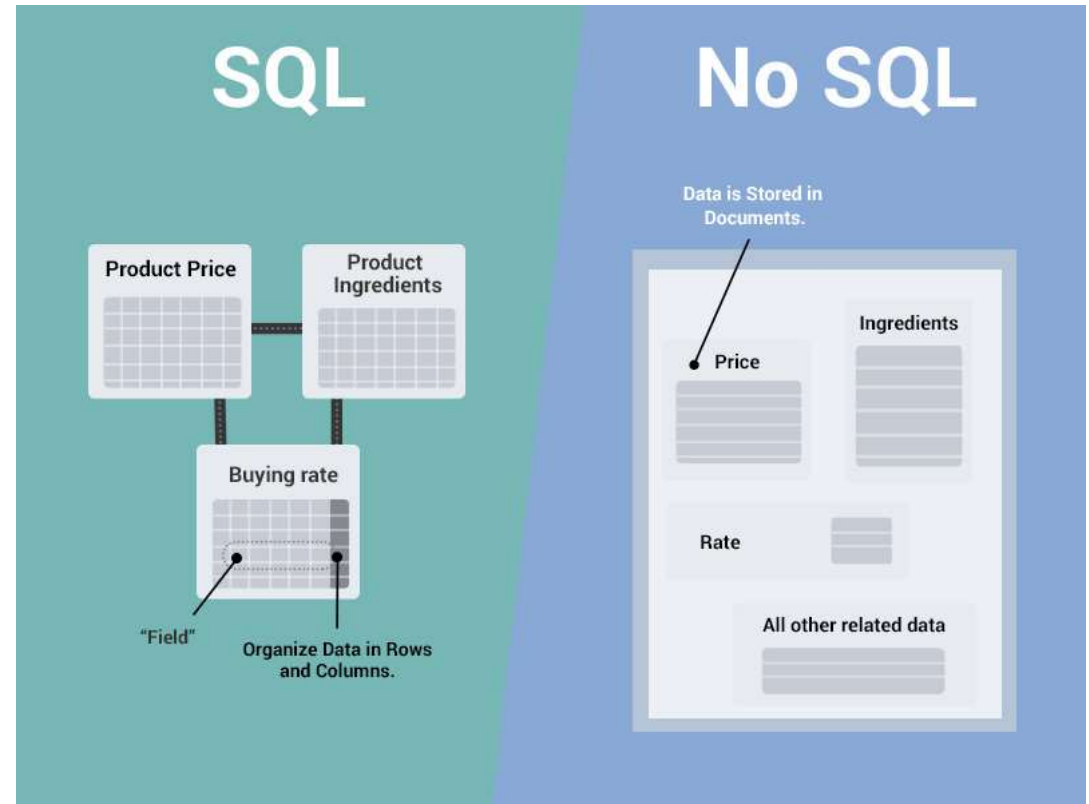
Designed By:
Priyanshu Jain

Differences between AI, ML, and DL,



No SQL for big DATA

- Some time we need to access the partial data
- Whole available data will be stored in no SQL
- Mongo DB is very popular and easy to integrate with python.



Real Time Application for Data Science

Applications of Data Science in Finance

Risk Analytics

Real-Time
Analytics

Consumer
Analytics

Customer Data
Management

Providing
Personalized
Services

Fraud
Detection

Algorithmic
Trading



Data Wrangling and Munging

- Data **wrangling**, sometimes referred to as **data munging**, is the process of **transforming and mapping data** from one "raw" data form into another format with the intent of making. **It more appropriate and valuable for a variety of downstream purposes such as analytics.**
- A **data wrangler** is a person who performs these transformation operations.

Data Wrangling and Munging cont...

- This may include further **munging**, data visualization, data aggregation, training a statistical model, as well as many other potential uses.
- Data munging as a process typically follows a set of general steps which begin with extracting the data in a raw form from the data source, "munging" the raw data using algorithms (e.g. sorting) or parsing the data into predefined data structures, and finally depositing the resulting content into a data sink for storage and future use.