Data Science Modelling, concepts, techniques

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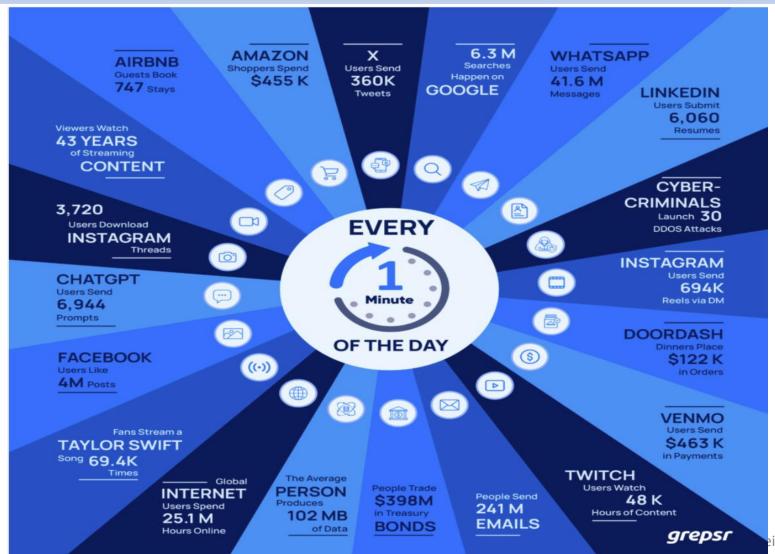
Goals

- ✓ Know what Data Science is and learn the basic algorithms
- ✓ Perform Data Science techniques
- ✓ Know how to apply algorithms to real-world applications

Chapter 1. Introduction

- ➤ Why Data Science ?
- ➤ What is Data Science ? How big is Data Science
- ➤ Characteristics of Data Science
- ➤ Top most popular Data Science algorithms
- ➤ Major issues in Data Science

Why Data Science



ikhi, Sep-2024, SBUK

Why Data Science

Air Pollution Forest Fire Detection Monitoring of combustion gases and preemptive fire conditions to define alert zones. Wine Quality Enhancing grapes and grapevine health. Offspring Care Control of growing conditions of the offspring in

Control of CO, emissions of factories, pollution emitted by cars and toxic gases generated in

Monitoring soil moisture and trunk diameter in vineyards to control the amount of sugar in

animal farms to ensure its survival and health.

Sportsmen Care

Vital signs monitoring in high performance centers and fields.

Structural Health

Monitoring of vibrations and material conditions in buildings, bridges and historical monuments.

Smartphones Detection

Detect iPhone and Android devices and in general any device which works with Wifi or Bluetooth interfaces.

Perimeter Access Control

Access control to restricted areas and detection of people in non-authorized areas.

Radiation Levels

Distributed measurement of radiation levels in nuclear power stations surroundings to generate leakage alerts.

Electromagnetic Levels

Measurement of the energy radiated by cell stations and WiFi routers.

Traffic Congestion

Monitoring of vehicles and pedestrian affluence to optimize driving and walking

Smart Roads

Warning messages and diversions according to climate conditions and unexpected events like accidents or traffic jams.

Smart Lighting

Intelligent and weather adaptive lighting in street lights.

Intelligent Shopping

Getting advices in the point of sale according to customer habits, preferences, presence of allergic components for them or expiring dates.

Noise Urban Maps

Sound monitoring in bar areas and centric zones in real time.

Water Leakages

Detection of liquid presence outside tanks and pressure variations along pipes.

Vehicle Auto-diagnosis

Information collection from CanBus to send real time alarms to emergencies or provide advice to drivers.

Item Location

Search of individual items in big surfaces like warehouses or harbours.

Golf Courses hi, Sep-20

in the city.

Selective irrigation in dry zones to reduce the water resources required in the green.

Waste Management

to optimize the trash collection routes.

Smart Parking

Detection of rubbish levels in containers

Monitoring of parking spaces availability

Water Quality

Study of water suitability in rivers and the sea for fauna and eligibility for drinkable

4. SBUK

Quality of Shipment Conditions Monitoring of vibrations, strokes, container openings or cold chain maintenance for insurance purposes.

Some links of Data Science

https://everysecond.io

https://www.worldometers.info/coronavirus

http://irsc.ut.ac.ir

https://finance.yahoo.com/quote/BTC-USD/history

Why Data Science?

- The Explosive Growth of Data: from terabytes to petabytes
 - Business: Web, e-commerce, transactions, stocks, ...
 - Science: Remote sensing, bioinformatics, scientific simulation, ...
 - Society and everyone: news, digital cameras, YouTube
 - Healthcares, recording patient symptom using online monitoring

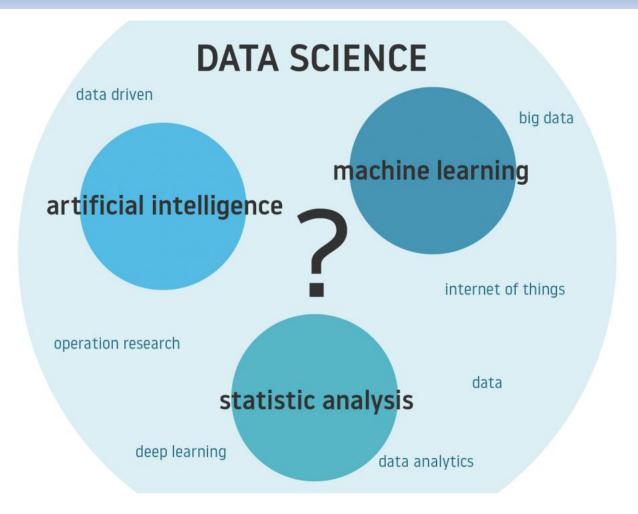
Evolution of Sciences (from empirical sciences to data science)

- Before 1600, **empirical science**
- 1600-1950s, theoretical science
 - Each discipline has grown a *theoretical* component. Theoretical models often motivate experiments and generalize our understanding.
- 1950s-1990s, computational science
 - Over the last 50 years, most disciplines have grown a third, computational branch (e.g. empirical, theoretical, and computational ecology, or physics, or linguistics.)
 - Computational Science traditionally meant simulation. It grew out of our inability to find closed-form solutions for complex mathematical models.
- 1990-now, data science/ Big Science
 - The flood of data from new scientific instruments and simulations
 - The ability to economically store and manage petabytes of data online
 - The Internet and computing Grid that makes all these archives universally accessible
 - Scientific info. management, acquisition, organization, query, and visualization tasks scale almost linearly with data volumes. Data Science is a major new challenge!

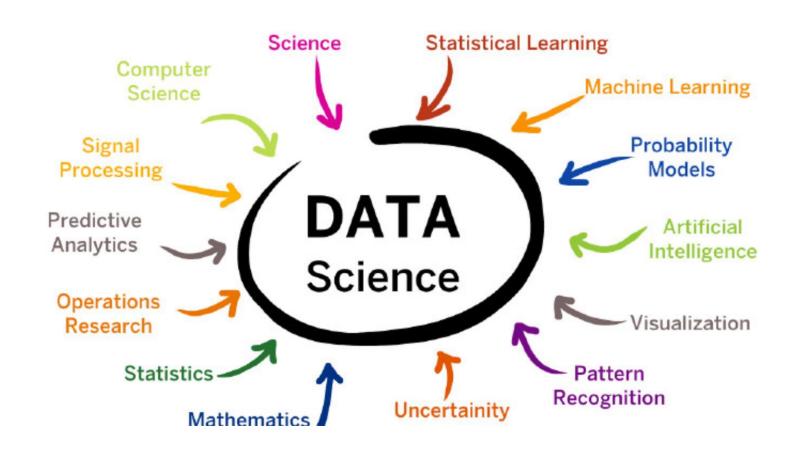
Characteristics of Data Science in big data?



Data Science: Confluence of Multiple Disciplines



Data Science: Confluence of Multiple Disciplines



Types of Variables

- *Nominal*: Name only--Gender, hair color, ethnicity
- *Ordinal*: Nominal categories with an implied order--Low, medium, high.
- *Discrete*: Reflects a number obtained by counting—no decimal.
- *Continuous*: Reflects a measurement; the number of decimal places depends on the precision of the measuring device.

Types of Variables

There are two basic types of variables:

categorical and numerical.

Categorical Variables: variables defined by the classes or categories into which an individual member falls.

Numerical Variables: variables to which a number is assigned as a quantitative value.

Data and Variables

Data are often discussed in terms of variables, where a variable is:

Any characteristic that *varies* from one member of a population to another.

A simple example is height in centimeters, which varies from person to person.

Definition of Variables in a data Matrix

- AGE: Age in years
- BMI: Body mass index, weight/height² in kg/m²
- FFNUM: The average number of times eating "fast food" in a week
- TEMP: High temperature for the day
- GENDER: 1- Female 0- Male
- EXERCISE LEVEL: 1- Low 2- Medium 3- High
- QUESTION: Compared to others, what is your satisfaction rating of the National Practitioner Data Bank?
 - 1- Very Satisfied 2- Somewhat Satisfied 3- Neutral
 - 4- Somewhat dissatisfied 5- Dissatisfied

Data table/Data matrix

OBS	AGE	ВМІ	FFNUM	TEMP(°F)	GENDER	EXERCISE LEVEL	QUESTION
1	26	23.2	0	61.0	0	1	1
2	30	30.2	9	65.5	1	3	2
3	32	28.9	17	59.6	1	3	4
4	37	22.4	1	68.4	1	2	3
5	33	25.5	7	64.5	0	3	5
6	29	22.3	1	70.2	0	2	2
7	32	23.0	0	67.3	0	1	1
8	33	26.3	1	72.8	0	3	1
9	32	22.2	3	71.5	0	1	4
10	33	29.1	5	63.2	1	1	4
11	26	20.8	2	69.1	0	1	3
12	34	20.9	4	73.6	0	2	3
13	31	36.3	1	66.3	0	2	5
14	31	36.4	0	66.9	1	1	5
15	27	28.6	2	70.2	1	2	2
16	36	27.5	2	68.5	1	3	3
17	35	25.6	143	67.8	1	3	4

Reading Web data

Co to the following address

https://raw.githubusercontent.com/amrrs/sample revenue dashboard shiny/master/recommendation.csv

Sheikhi, Nov.-2024, SBUK

Account, Product, Region, Revenue-----→name of attributes Axis Bank, FBB, North, 2000 HSBC,FBB,South,30000 SBI,FBB,East,1000 ICICI,FBB,West,1000 Bandhan Bank, FBB, West, 200 Axis Bank, SIMO, North, 200 HSBC,SIMO,South,300 SBI,SIMO,East,100 ICICI,SIMO,West,100

Thanks for your attention