Algorithmic Foundations of Data Science

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Room 5.03, Informatics Forum



Drop-in sessions

• We run a weekly drop-in sessions to help you better understand the

teaching materials.

- Time: Thursdays 3pm 4pm
- Place: Absorb@Appleton Tower (the café area outside the Lecture Theature).
- Organiser: Peter Macgregor (<u>P.MacGregor-4@sms.ed.ac.uk</u>)



- Tutorials will take place on Fridays from week 2.
- Tutors: Ruochun Jin, Bogdan Manghiuc, and He Sun.
- Content of the Tutorials: Discuss homework solutions
- All the enrolled students will be allocated into one of the four tutorial groups.
- We'll do the tutor-tutee allocation early next week. Please make sure to register the course by the end of this week.



- Coursework online on **18 October**, and deadline day **8 November**.
- The coursework will consist of two parts:
 - **Part A** is to test your understanding of the materials learned in class.

You're expected to prove/disprove mathematical claims, design

algorithms and analyse their time complexity.

• **Part B** involves programming. Your code will be judged by its

correctness and runtime.



Now let us start.

- Most people outside Edinburgh call *some subject* **computer science**.
- People around the world call *some fancier subject* **data science**.

BUT, if we follow the same rule,

- we should have called astronomy **telescope science**.
- we should have called biology microscope science.

The use of Natural Philosophy in history

From the ancient world to the 19th century, the term "natural philosophy" was the common term used to describe the practice of studying nature. It was in the 19th century that the concept of "science" received its modern shape with new titles emerging such as "biology" and "biologist", "physics" and "physicist" among other technical fields and titles; Issac Newton's book Philosophiae Naturalis Principia Mathematica (1687), whose title translates to "Mathematical Principles of Natural Philosophy", reflects the then-current use of the words "natural philosophy", akin to "systematic study of nature".

My understanding about Data Science

• Data Science is at its rather early stage, the stage in which we couldn't even

find a way to name it without using the word "Science".

• Looking back to history, the early stage of a research field is usually the moment when big ideas and breakthrough have occurred.

• Therefore, you should attend AFDS if you're willing to be a great scientist.

What should you attend AFDS? (cont.)

• Several topics covered in AFDS already have wide practical applications.

• Some more frontier topics covered in AFDS are excellent mathematical training for you, and could have big industrual impacts for you in five to ten years.

• A good university education is not only to help you find a job next year, but also to ensure that you'll be highly-qualified in a decade.

What will be covered in the course? Here are some sample problems.

Problem 1: Streaming Algorithms

Background: big data in the 21st century



Genome sequences for many species are available: each megabytes to gigabytes in size.



There are 5 billion global mobile phone users.



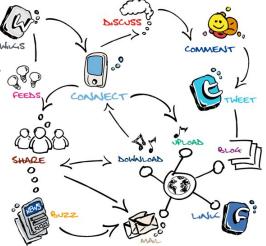
300 hours of videos uploaded per minute



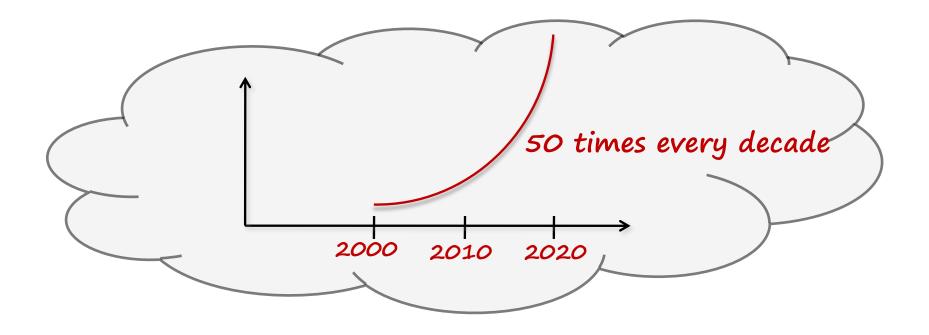
Twitter produces over 500 million tweets per day.



There are about 2.23 billion monthly active users in Facebook.

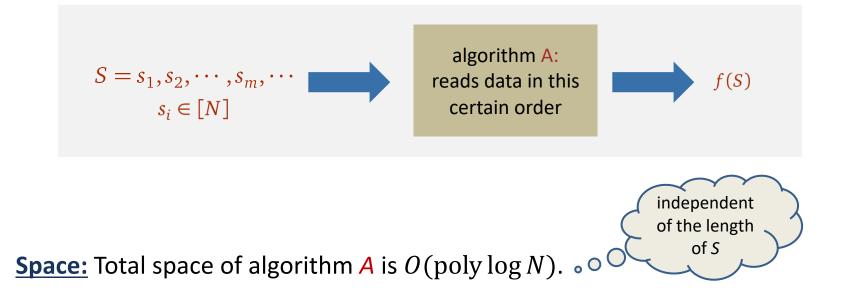


Background: big data in 21st century



- Storing the entire data is usually not possible anymore. Instead, we can only afford to store *partial information* of the input dataset.
- Good approximate solutions are usually sufficient for most practical applications!

Data streaming algorithms: model



Quick update: Processing time of every arriving item should be fast.

Image that the input is the IP addresses visiting UoE website.

Approximate guarantee: With probability 99%, the algorithm's output should be very close to the right answer, e.g., the error is within $\pm 10\%$.

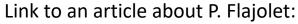
Flajolet's Loglog counting algorithm



distinct words.



FIGURE 1. The LOGLOG Algorithm with m = 256 condenses the whole of Shakespeare's works to a table of 256 "small bytes" of 4 bits each. The estimate of the number of distinct words in this run is $n^{\circ} = 30897$ (the true answer is n = 28239), which represents a relative error of +9.4%.



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https://rjlipton.wordpress.com/2011/03/27/philippe-flajolet-1948-2011/



Philippe Flajolet (1948 -2011)

Topics covered for streaming algorithms

 Two central techniques for designing streaming algorithms: Sampling, and Sketching.





- Streaming algorithms for computing certain statistical information:
 - The number of distinct elements
 - The number of occurrence of frequently occurring items

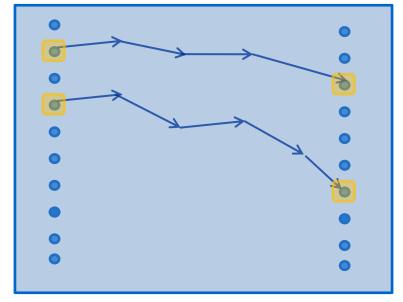
Problem 2: Expander & Super concentrators

Super concentrators

Given n cities in the north and n cities in the south, construct a highway network, such that for any k cities in the north and k cities in the south, there are k vertex disjoint paths.

> "disjoint" => efficiency of transportation, no delay # of edges ⇔ construction cost





Construct a network (directed graph) with *n* input nodes and *n* output nodes, such that for any *K* input nodes, and any *K* output nodes, there are *K* disjoint paths connecting them.

A complete bipartite graph is an example, but too "expensive".

For any n, there is a super concentrator with 28n edges.

Super concentrators (cont.)

"Finally, the <u>super concentrators</u> constructed by Valiant in the context of computational complexity established the fundamental role of <u>expander graphs</u> in computation."



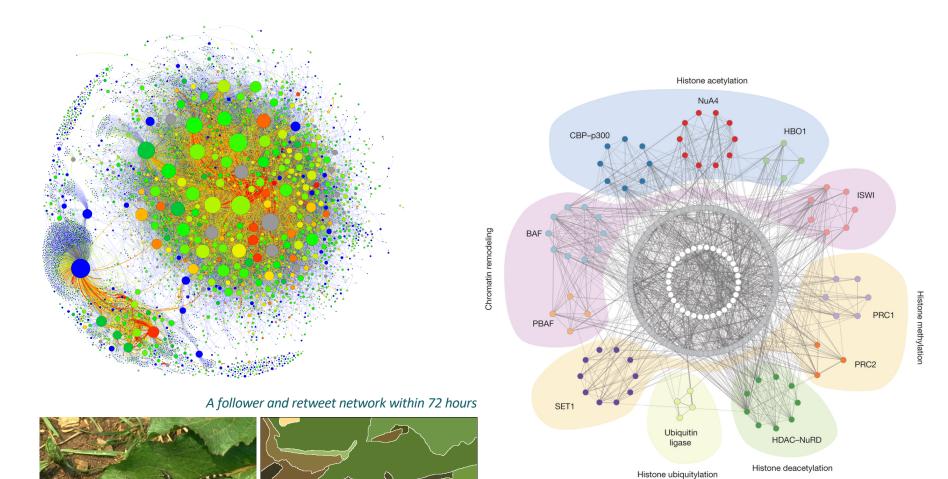
2010 ACM Turing Award Citation

Most of Valiant's work on super concentrators was done when he was at UoE.

In this course we will discuss various aspects of expander graphs, and their applications in Data Science.

Problem 3: Graph Clustering

Applications of clustering

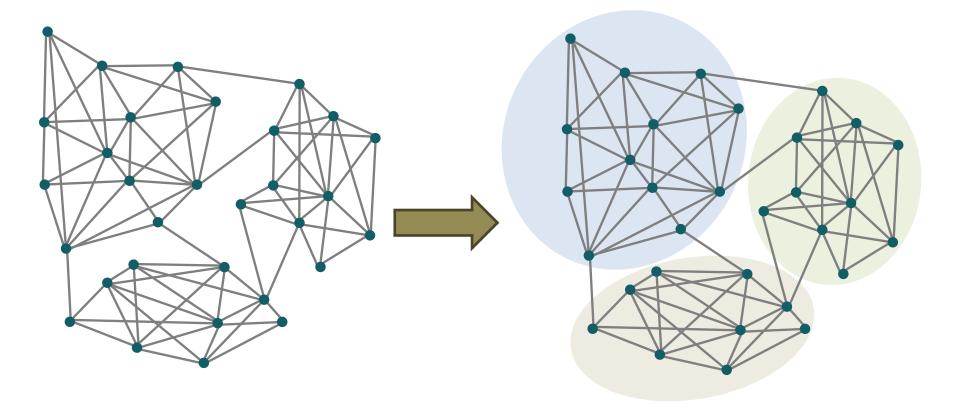


A protein–protein interaction network for the 425 human chromatin factors screened [Huang et al., 2013].

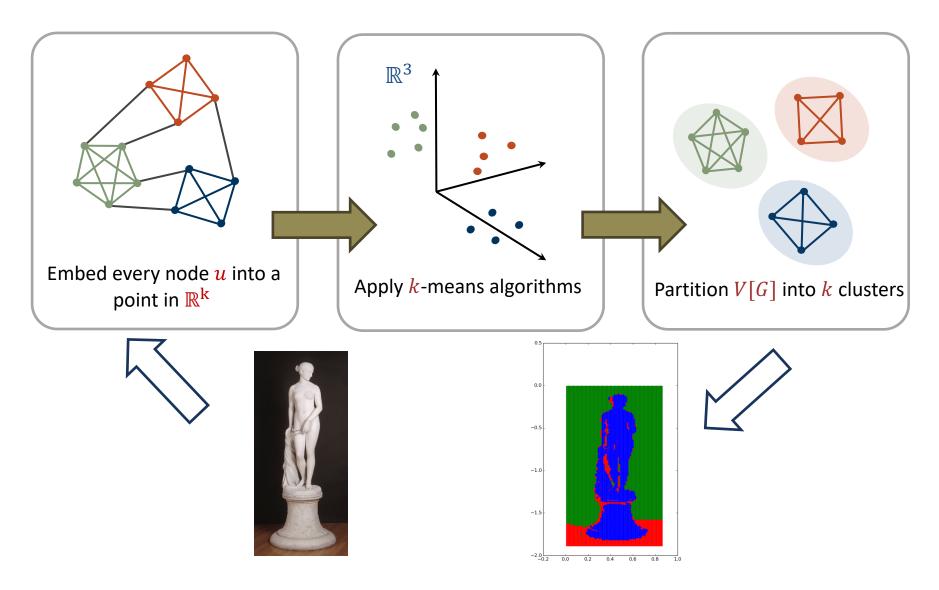
Segmentation in Computer Vision

Graph clustering

Partition a graph into different clusters.



Spectral clustering



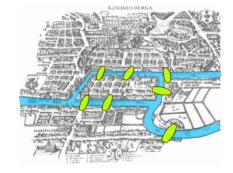
We will discuss how similar algorithms can be analysed.

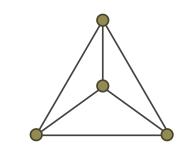
A crucial theme around our discussion:

Relations among graphs, matrices, and geometric objects

Graphs, matrices, and geometric objects







- Coloring
- matching
- Hamiltonian Cycles
- Spanning Trees

Since 1700s

Leonhard Euler (1707-1783)

Seven Bridges of Königsberg 1736

$$L = \begin{pmatrix} 3 & -1 & -1 & -1 \\ -1 & 3 & -1 & -1 \\ -1 & -1 & 3 & -1 \\ -1 & -1 & -1 & 3 \end{pmatrix}$$

- Eigenvalues and Eigenvectors
- Rank, and kernel

- The isoperimetric problem
- Sobolev inequalities
- Heat equations

Construct a graph

- \approx Construct a matrix
- ≈ Construct a geometric object