CS6713: Scalable Algorithms for Data Analysis

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20-Aug-2022

The course

- Contents:
 - Streaming algorithms (Frequency moments)
 - Sketching
 - Dimension reduction
 - Graph streaming/sketching
- Prerequisite: Undergraduate algorithms, Basics of probability and linear algebra

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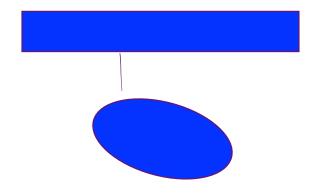
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 - Streaming algorithms (Frequency moments)
 - Sketching
 - Dimension reduction
 - Graph streaming/sketching
- Prerequisite: Undergraduate algorithms, Basics of probability and linear algebra
- Evaluation
 - 35% : Theory assignments
 - 30% : Coding assignments
 - 35% : Mini project

Reference

- Data Stream algorithms, Lecture Notes, Amit Chakrabarti, 2020
- For basics of probability and randomized algorithms:
 Probability and Computing (2nd Edition) by Mitzenmacher and Upfal

Streaming Algorithms

Classical Algorithms: Random Access Model (RAM)



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Classical Algorithms: Random Access Model (RAM)

Streaming Model

- The input consists of m objects/items/tokens e_1, e_2, \ldots, e_m that are seen one by one by the algorithm.
- The algorithm has "limited" memory say for B tokens where B < m (often B << m) and hence cannot store all the input
- Want to compute interesting functions over input

Classical Algorithms: Random Access Model (RAM)

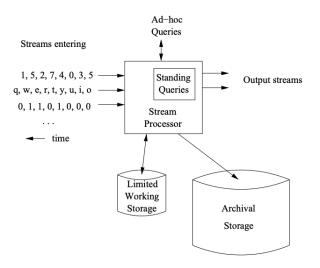
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Some examples:

- Each token is a number from [n]
- High-speed network switch: tokens are packets with source, destination IP addresses and message contents.
- Each token is an edge in graph (graph streams)
- Each token is a point in some feature space
- Each token is a row/column of a matrix

A data stream management system



["Mining of Massive Data Sets" by Leskovec, Rajaraman, Ullman]

Streaming model: motivation/connections

- Very large but slow storage (tape, slow disk) that is suited for sequential access and fast main memory. Read data in one (or more) passes from slow medium.
- Scenarios such as network switches, sensors etc where huge amount of data is flying by and cannot be stored (due to cost or privacy/legal reasons) but one wants only high-level statistics.
- Distributed computing. Data stored in multiple machines.
 Cannot send all data to central location. Streaming algorithms can simulate a class of algorithms that exchange small amount of data.

Finding Majority Element

Given an array A of m integers, output an element that occurs more than m/2 times in A?

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Finding Majority Element

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Algorithm:

• Initialize $c \leftarrow 0$ and s = Null

• For
$$i = 1$$
 to m

• If
$$A[i] = s$$
, then $c \leftarrow c + 1$.

• If
$$A[i] \neq s$$
 and $c > 0$, then $c \leftarrow c - 1$.

• If
$$A[i] \neq s$$
 and $c = 0$, then $c \leftarrow 1$ and $s \leftarrow A[i]$.

• Check whether s is indeed the majority element and output accordingly. 2. 3.3.1.2.1.1.1.5.6 i=11,s=1,c=1

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Definition: For each element q in the array A, let f_q be the frequency of q.

Claim: If there is a majority element q, then algorithm outputs s=q and $c\geq f_q-m/2$.

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Proof:

- I_q : No. of times we increment c when we see q.
- D_q : No. of times we decrement c when we see q.
- I_0 : No. of times we increment c when we see an element $\neq q$
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$$\geq I_{q} + I_{0} - \frac{m}{2} =$$

$$\geq I_{q} + D_{q} - \frac{m}{2} \geq f_{q} - \frac{m}{2}$$

Heavy Hitters Problem and Misra-Gries Algorithm

Heavy Hitters Problem: Find all elements i such that $f_i > m/k$.

Thank You.