

Advanced Algorithms

Lecture #1 1/12/2021

- self-evaluation test.

Diff from A to A' \hookrightarrow

1) Randomized: \Pr [algo correct] $\geq 90\%$.

2) Approx: outputs a'

$$a \leq a' \leq c \cdot a$$

\uparrow approx. \uparrow correct ans.

1 \neq 2: \Pr [algo outputs a' $\left. \begin{array}{l} a \leq a' \leq c \cdot a \end{array} \right\} \geq 90\%$

$c =$ approx. ratio 1% $\rightarrow c = \underline{1.01}$

$c = 2$
 $c = f(\text{input size})$

1) Design algo

2) Analysis: a) correctness
 b) performance. → time.
 → space.
 ↘ communication network us

1) Hashing: $h: U \rightarrow [n] = \{1, 2, \dots, n\}$.

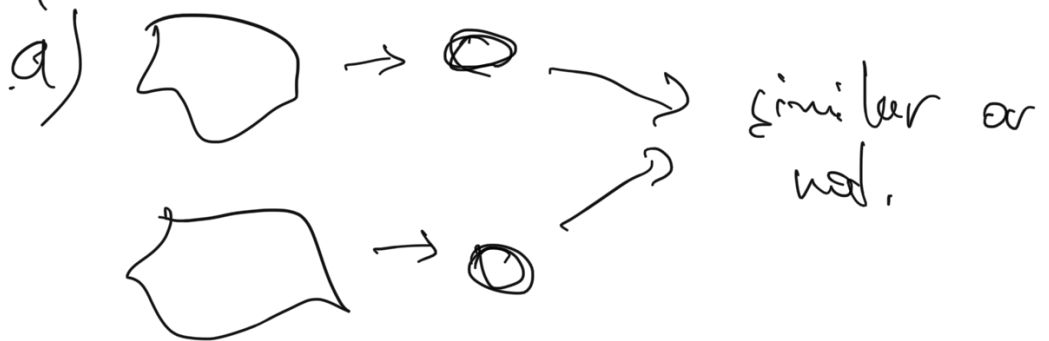
solves Dictionary problem

$O(1)$ expected query time

perfect hashing: $O(1)$ w.c. q. t.

2) Sketching / streaming algos.

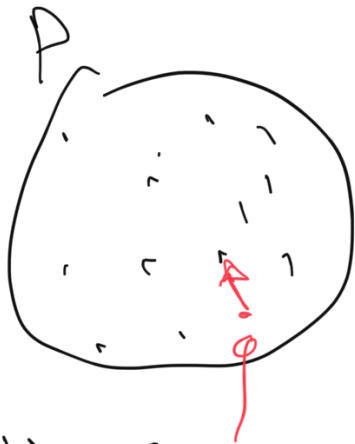
↓
 functional compression



ex: estimate # distinct IPs
seen.

related to high dim. geom.
 \mathbb{R}^d d "large".

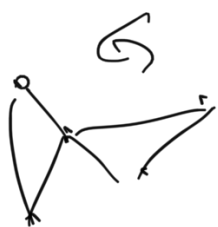
3) Nearest Neighbor Search (NNS)
in high d. space



4) Graphs → advanced algo/method
for graphs.

Max-Flow: - poly-time algo.
- scaling algo's.

5) Spectral graph theory.



spectral clustering algo.

6) Optimization.

$$\min_{x \in \mathbb{R}^d} f(x)$$

$$x \in C$$

Lin. Programming: $f, C \rightarrow$ linear

Inferior Point Method

↳ gradient descent, Newton's method,

Multiplicative weights Update.
Learning from experts.

7) Large-scale models.

- parallel / cluster computing,
for systems like Map Reduce

- I/O external model

→ feeds into cache.

8) ff.

Course Expectations/Del.:

Grading:

- 1) writing 10%
- 2) 5 hw's 55%
- 5 "free" lab days.
- 10% off / day. \rightarrow 5 days la
- 3) project. 35% Team
 - report 1 } logit.
 - & - 2 }
 - final. 25%

Types:

- 1) ready
- 2) implementation. \leftarrow
- 3) research-level.

Prereq:

- strong math backg.
- basic prob., linear algebra
- CS backg.
 - $O(-)$, $\Delta(-)$.

- sorting, bin. search, com. comp.

Problems counting up to n .

space: how many bits will be nec. to count to n .

$\rightarrow \lceil \log_2 n \rceil$ bits to repr. n .

Goal: to do better.

Thm: ~~can't~~ can't do better if algo det. or is exact.

Morris's Algo: get $O(\lg \lg n)$ bits

to count up to n , up to constant-factor approx.

$\rightarrow c = O(1)$

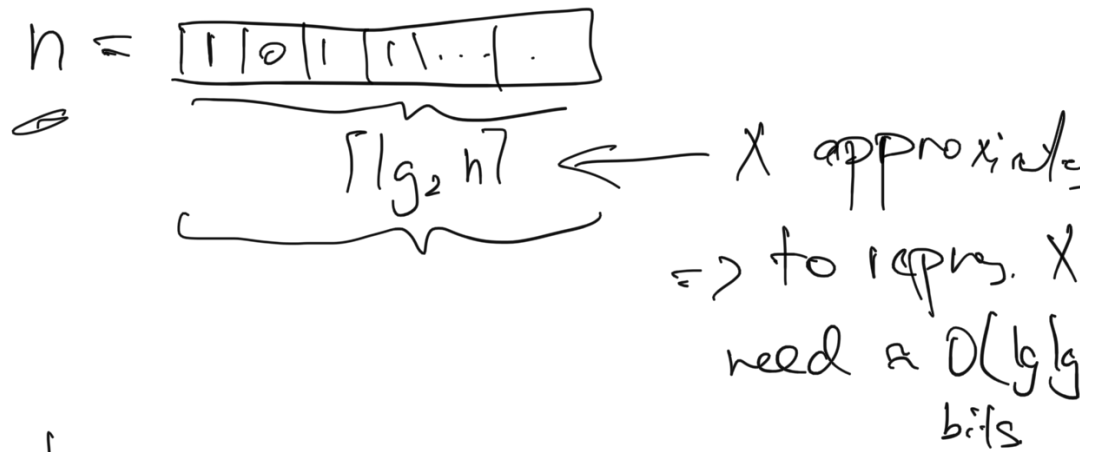
M. 9.1: \rightarrow focus for $X = 0$.

- @ button: $X := \begin{cases} X+1, & \text{with prob } 2^{-X} \\ X, & \text{otherwise} \end{cases}$

- @ end, to output approx to # presses of button:

$$\hat{n} = n^{X-1}$$

" " < " "
↑ the estimate.



Analysis: a) Correctness:

$$\Pr[\hat{n} \approx n] \geq 90\%$$

Probability:

$X \rightarrow$ random variable.

Def: $E[X] = \sum_a a \cdot \Pr[X=a]$ ←

Def: $X = X_1 + X_2$.

$$E[X] = E[X_1] + E[X_2].$$

Concentration bounds:

Lemma: Chernoff's bound

Lemma $\lfloor \text{Markov's inequality} \rfloor$

$X \geq 0$: $\forall \lambda > 0$:

$$\Pr[X > \lambda] \leq \frac{E[X]}{\lambda} \leftarrow$$

$$\Pr[X > c \cdot E[X]] \leq 1/c.$$

Def: Variance: $\text{Var}[X] = E[(X - E[X])^2]$

Lemma [Chebyshev's bound]: $\forall \lambda > 0$.

$$\Pr[|X - E[X]| > \lambda] \leq \frac{\text{Var}[X]}{\lambda^2}$$

\hookrightarrow close or = to actual
val.