

Reading: Text: chapter 1, Paper: *An introduction to randomized algorithms*, Karp 1991.

Logistics

The essentials:

- Website: linked to from my homepage
- Office Hours: by appointment
- Project: due at end of quarter
- Problem Sets: two problem sets, due in class
- Readings: from Randomized Algorithms, current research papers

The prereqs:

- discrete math: graph theory, big-oh notation, basic probability (random variables, expectation, variance, basic bounds)
- algorithms: run-time analysis, dynamic programming, LP-based algorithms, basic NP-hardness

Overview

A *randomized* algorithm takes:

- input
- string of random bits
[[*For now, assume endless supply of truly random bits*]]

Note: same input may produce different outputs.

Advantages:

- reduced execution time/space requirements, simple to analyze/implement
- reduces det alg with bad worst-case behavior to alg that performs well on every input with high probability

Techniques:

- **foiling the adversary:**

Example: Playing the lottery:

- always pick the same number → adversary can guarantee you always lose
- pick a random number → you win sometimes

Idea:

- An algorithm is a zero-sum game between
 - * an adversary providing the inputs and
 - * the algorithm designer providing alg

- * with adversary payoff being running time of alg
 - a randomized alg is a mixed strategy in the game
 - mixed strategies can guarantee higher payoffs.
- **random sampling:**
 - Example:** Sensitivity analysis:
 - given complicated system $f(x_1, \dots, x_n)$ (boolean function), how sensitive system is to failure of i 'th component x_i
 - sample inputs $x \in \{0, 1\}^n$ with $x_i = 0$ and test for what fraction does $f = 0$
 - Idea:** random sample from population is representative of population as a whole
 - **abundance of witnesses:**
 - Example:** Primality testing:
 - a factor p of a number n is a *witness* that n is composite
 - test random numbers to see if they are factors
 - Idea:** Witnesses
 - hard to find deterministically
 - if abundant enough, can sample and get one whp.
 - **fingerprinting and hasing:**
 - Example:** Testing membership:
 - want to maintain set of objects
 - to implement “add”, must test if object is already in set
 - map objects to small set of buckets based on “fingerprints”
- compare new object to those in its bucket
- Idea:**
 - represent a long string by a short fingerprint
 - use fingerprint to reduce search space/input size, or to test equality
- **random re-ordering:**
 - Example:** Binary search trees:
 - arrange input data into binary search tree
 - given order may create very unbalanced tree with naive alg
 - a random re-ordering is likely to give balanced tree with naive alg
 - Idea:** after re-ordering, input is unlikely to be pathological for naive algorithm
 - **load balancing**
 - Example:** Machine scheduling:
 - send n printing jobs to n printers
 - pick arbitrary printer – worst-case load n
 - pick random printer – balls and bins, expected load $\log n$
 - Idea:**
 - randomization spreads load among resources
 - good for distributed environments
 - **markov chains**
 - Example:** PageRank:
 - want to calculate probability random surfer lands at given page

- simulate random walk for “long enough” and count fraction of time spent at given page

Idea:

- many walks are rapidly mixing
- can use walks to efficiently sample from subspace
- useful for counting problems

Application: Sorting

Problem: Given

- a set S of n numbers

Output

- a list of members of S in ascending order

Algorithm:

- find *pivot* element $y \in S$ s.t. half of S is smaller than y
- partition $S \setminus \{y\}$ into S_1 and S_2 s.t.
 - S_1 is elts in S smaller than y
 - S_2 is elts in S larger than y
- recursively sort S_1 and S_2

Analysis: $T(n)$ is running time on input size n

- time to find y : cn for constant c
- time to partition: $(n - 1)$ (compare each elt to y)

so

$$T(n) \leq 2T(n/2) + (c + 1)n$$

which has solution $T(n) = c'n \log n$.

Question: How to find y ?

Idea:

- running time good so long as S_1 and S_2 are *approximately* same size

Example: if aim for partition such that $|S_1| \leq 3n/4$ and $|S_2| \leq 3n/4$, then

- $T(n) \leq 2T(3n/4) + (c + 1)n$ has solution $T(n) = O(n \log n)$

- there are $n/2$ pivots y whose partitions are like this

[To see this, imagine sorted array of elts and observe that middle half have this property.]

- choose a random pivot element y uniformly from S and hope to get lucky often enough

Analysis: (randomized alg):

Question: How many comparisons *in expectation*?

Def: For $1 \leq i \leq n$, let $S_{(i)}$ denote the element of *rank* i (the i 'th smallest element) in S .

- $S_{(1)}$ is smallest elt of S
- $S_{(n)}$ is largest elt of S

Def: Let X_{ij} be indicator random variable that $S_{(i)}$ and $S_{(j)}$ are compared by alg.

- $X_{ij} = 1$ means $S_{(i)}$ and $S_{(j)}$ were compared, sorted directly
- $X_{ij} = 0$ means $S_{(i)}$ and $S_{(j)}$ were not compared, sorted implicitly

Then total number of comparisons is

$$\sum_{i=1}^n \sum_{j>i} X_{ij}$$

and expectation number is

$$E\left[\sum_{i=1}^n \sum_{j>i} X_{ij}\right] = \sum_{i=1}^n \sum_{j>i} E[X_{ij}].$$

Def: Let p_{ij} be probability $S_{(i)}$ and $S_{(j)}$ are compared by alg. Then

$$E[X_{ij}] = 1 \times p_{ij} + 0 \times (1 - p_{ij}) = p_{ij}.$$

Think of steps performed by alg. in a tree of pivots (DRAW TREE) and consider level-order traversal.

- there is a comparison between $S_{(i)}$ and $S_{(j)}$ iff $S_{(i)}$ or $S_{(j)}$ is chosen as pivot *before* any elt of rank between i and j
- consider permutation defined by order in which elts were chosen as pivots and note any elt in $\{S_{(i)}, \dots, S_{(j)}\}$ equally likely to be first one chosen as pivot

Thus,

$$p_{ij} = \frac{2}{j - i + 1}.$$

[Is permutation defined above a uniformly random one? No, e.g., perm $S_{(3)}S_{(1)}S_{(2)}S_{(4)}S_{(5)}$ can't happen. Prob only uniform over 1st elt.]

To conclude,

$$\begin{aligned} E &= \sum_{i=1}^n \sum_{j=i+1}^n p_{ij} \\ &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j - i + 1} \\ &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \\ &\leq 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} \\ &= O(n \log n) \end{aligned}$$

Note:

- expected running time holds for every input *[realization of running time depends only on random choices made by alg, not on input itself]*
- we bounded expected running time, but can make stronger statement that running time is close to expectation *with very high probability* (for every input)
- choosing random pivot required $\log |S|$ random bits; sometimes not so easy
 - picking random real number from $[0, 1]$
 - simulating coin flip with bias $p \neq 2^{-k}$