$\left\{ \operatorname{Csc} 80030 \mid \operatorname{Lecture} 6 \right\}$

PROBABILISTIC ANALYSIS & RANDOMIZED ALGORITHMS

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part 0

- ► HOMEWORK: HW3 due next week
- ▶ PROGRAMMING ASSIGNMENT: is out, due Mar 25
- ► TODAY: probabilistic method & derandomization

PART 1 derandomization

Derandomization

- ► NAMELY, remove the randomness from randomized algorithms.
 - ▶ Eliminate errors, from *w.h.p.* to *always*
 - Theoretical interest: does randomness *really* help in the design of algorithms?
- ► TWO TECHNIQUES. method of conditional expectations & pairwise independence.

Finding a Large Cut

▶ THEOREM: Any graph G with m edges has a cut of size $\geq \frac{m}{2}$.

LARGECUT Algorithm

Input: a graph G = ([n], E)

- 1. Flip *n* coins r_1, r_2, \ldots, r_n , put vertex *i* in *S* if $r_i = 0$ and in *T* otherwise.
- **2**. Output (S, T).
- ▶ by averaging argument, there exists a choice of $r_1, r_2, ..., r_n$ that leads to a cut of size at least $\frac{m}{2}$.
- find a good sequence of coin tosses "bit by bit".

LARGECUT tree



QUESTION. How to compute labels of the "internal" nodes?

Deterministic LARGECUT

► DEFINITION. Define the conditional expectation

$$e(r_1, r_2, \dots, r_i) = \mathbb{E}_{R_1, \dots, R_n} \Big[|\operatorname{cut}(S, T)| \mid R_1 = r_1, R_2 = r_2, \dots, R_i = r_i \Big]$$

• BASE CASE.
$$e(\lambda) = |E|/2$$
.

► INDUCTIVE CASE.
$$e(\lambda) = E_{R_1}[e(R_1)].$$

More generally, $e(r_1, \dots, r_i) = E_{R_{i+1}}[e(r_1, \dots, r_i, R_{i+1})]$

Deterministic LARGECUT

Deterministic LARGECUT Algorithm I

1. Set $S = \emptyset, T = \emptyset$

2. For
$$i = 0, \ldots, n - 1$$
:

2.1 If $|\operatorname{cut}(\{i+1\}, S)| > |\operatorname{cut}(\{i+1\}, \overline{S})|$, set $T \leftarrow T \cup \{i+1\}$,

2.2 else set
$$S \leftarrow S \cup \{i+1\}$$
.

REMARK. This is the "natural" greedy algorithm. Method of conditional expectations tells us which objective function to optimize locally. ► ANALYSIS, REVISITED.

$$\mathbb{E}[|\operatorname{cut}(S,T)|] = \sum_{(i,j)\in E} \Pr[R_i \neq R_j] = |E|/2$$

- ► OBSERVATION. Applies for any distribution on r.v.'s (R₁,..., R_n) satisfying Pr[R_i ≠ R_j] = 1/2 for each i ≠ j; that is, *pairwise independent*
 - each R_i is an unbiased random bit;
 - for each $i \neq j$, R_i is independent from R_j
- QUESTION. Can we generate N pairwise independent bits using less than N truly random bits?

Pairwise independent bits

- ► CONSTRUCTION. Let $B_1, ..., B_k$ be *k* independent unbiased random bits. For each nonempty set $A \subseteq [k]$, let R_A be the r.v. $\bigoplus_{i \in A} B_i$.
- ► CLAIM. The $2^k 1$ random variables R_A are pairwise independent unbiased random bits.
 - Clear that each R_A is unbiased.
 - ▶ For pairwise independence, consider any $A \neq A' \subseteq [k]$. A'hen,

$$R_A = R_{A \cap A'} \oplus R_{A \setminus A'}; \quad R_{A'} = R_{A \cap A'} \oplus R_{A' \setminus A}$$

- ► $R_{A \cap A'}, R_{A \setminus A'}, R_{A' \setminus A}$ are independent and at least two are non-empty.
- Hence, R_A , $R_{A'}$ takes each value in $\{0, 1\}^2$ with prob. 1/4.
- ► Can generate N pairwise independent bits from [log(N + 1)] independent random bits.

Deterministic LARGECUT Algorithm II

 For all sequences of bits b₁, b₂,..., b_{⌈log(n+1)⌉}, run the randomized LARGECUT algorithm using coin tosses (r_A = ⊕_{i∈A}b_i)_{A≠∅} and choose the largest cut thus obtained.

PART 2 sample and modify

- STAGE ONE. Construct a random structure that does not have the required properties.
- ► STAGE TWO. Modify the structure to have the required property.
- APPLICATION. Obtain bounds on the size of the largest independent set in a graph (set of vertices with no edges between them).

Existence of large independent sets

- ► THEOREM: Any graph G with n vertices and m edges has an independent set with at least n²/4m vertices.
- ► CONSTRUCTION.
 - 1. Delete each vertex of G (and its incident edges) with prob 1 p.
 - 2. For each remaining edge, remove it and one of its adjacent vertices.
- ► CLAIM. Let X and Y be resp. the # of vertices and edges that survive step 1. Then, construction outputs an I.S. of size at least X Y.
- ► ANALYSIS. E[X] = np and $E[Y] = mp^2$, so $E[X Y] = np mp^2$. maximized at n - 2mp = 0, i.e. p = 2m/n and $E[X - Y] = n^2/4m$.

THE END | next, power of two choices