

Class 10: Agenda and Questions

1 Announcements

- HW4 due tomorrow!
- HW5 out now!

2 Warm-Up**Group Work**

1. Show that, in any undirected, unweighted graph $G = (V, E)$ with no self-loops, there is a cut with at least $|E|/2$ edges that cross it. (Recall that a *cut* in G is just a partition of the vertices $V = S \cup \bar{S}$, and that an edge $\{u, v\}$ crosses the cut if $u \in S$ and $v \in \bar{S}$ or the other way around).
2. Let φ be a 3-CNF formula. That is, φ is the AND of a bunch of clauses like $(x \vee \bar{y} \vee \bar{z})$ where \bar{x} means “not x ”, \vee means “OR” and \wedge means “AND”. For example:

$$\varphi = (x_1 \vee \bar{x}_2 \vee x_3) \wedge (x_2 \vee \bar{x}_4 \vee x_5) \wedge \cdots \wedge (x_{23} \vee \bar{x}_1 \vee \bar{x}_5).$$

Suppose that each clause has three distinct literals that appear in it. (e.g., $(x_1 \vee x_1 \vee x_1)$ is not allowed).

Given an *assignment* σ to the variables x_1, x_2, \dots (eg, $x_1 = TRUE, x_2 = FALSE$, etc), we say that a clause of φ is *satisfied* by σ if that clause evaluates to TRUE.

Show that any 3-CNF formula φ has an assignment σ so that at least 7/8 of the clauses are satisfied.

3 Recap/Questions

Any questions from the minilectures and/or the quiz? (The probabilistic method; Ramsey numbers; Independent sets)

4 Derandomization via conditional expectation

In class today, we’ll explore a general way to turn an existence proof—like the ones from your warm-up exercise—into an algorithm. This is called “Derandomization via conditional expectation.”

Group Work

Our goal in this group work is to find an efficient, deterministic algorithm to find a cut (S, \bar{S}) so that the number of edges crossing the cut is at least $|E|/2$. In general, finding a cut with the *maximum* number of edges crossing it is NP-hard; but this will at least find a large-ish cut.

Note: There is a straightforward deterministic greedy algorithm to do this. Here, we'll see a way to derive a deterministic algorithm using conditional expectations.

1. Let $G = (V, E)$ be as in warm-up question 1. Suppose the vertices are ordered $V = \{v_1, v_2, \dots, v_n\}$.

Suppose that $S \subseteq V$ is chosen uniformly at random (that is, each v_i is included in S independently with probability $1/2$). Let X be the number of edges crossing the cut (S, \bar{S}) .

Convince yourself that $\mathbb{E}[X | v_1 \in S] = |E|/2$.

2. Suppose that you have made some choices for v_1, v_2, \dots, v_{t-1} (eg, $v_1 \in S, v_2 \notin S, v_3 \in S, \dots, v_{t-1} \in S$), so that

$$\mathbb{E}[X | \text{choices for } v_1, \dots, v_{t-1}] \geq \frac{|E|}{2}.$$

Show that **either**

$$\mathbb{E}[X | \text{choices for } v_1, \dots, v_{t-1}; \text{ and } v_t \in S] \geq \frac{|E|}{2}$$

or

$$\mathbb{E}[X | \text{choices for } v_1, \dots, v_{t-1}; \text{ and } v_t \notin S] \geq \frac{|E|}{2}$$

3. Again, suppose that you have made choices for v_1, \dots, v_{t-1} so that

$$\mathbb{E}[X | \text{choices for } v_1, \dots, v_{t-1}] \geq \frac{|E|}{2}.$$

Show how to *deterministically, efficiently* make a choice for v_t so that

$$\mathbb{E}[X | \text{choices for } v_1, \dots, v_{t-1}; \text{ and } v_t] \geq \frac{|E|}{2}.$$

4. Building on your method above, design an algorithm to make a choice for v_1 , and then v_2 , and then v_3 , and so on, so that eventually you have (efficiently, deterministically) found a set S so that at least $|E|/2$ edges cross the cut (S, \bar{S}) .

[Solutions and discussion of the general paradigm of derandomization via conditional expectation; see lecture notes]

Group Work

1. Let φ be a 3-CNF formula with n variables and m clauses, and 3 distinct variables in each clause. Use the method of derandomization via conditional expectation to give an efficient (polynomial in n, m) deterministic algorithm to find an assignment to φ so that at least a $7/8$ -fraction of the clauses are satisfied.
2. (If time) There is also a natural greedy algorithm for this problem:
 - For $i = 1, 2, \dots, n$:
 - Assign x_i to be whichever value makes the most currently unsatisfied clauses true (breaking ties arbitrarily).

In the previous example (maximizing the size of a cut), the algorithm we came up with was secretly the natural greedy algorithm. Is your algorithm from the previous part the same as this natural greedy algorithm? Is it better or worse?