

## Solution Set 7

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Chris Umans

Obviously, if you have not yet turned in Problem Set 7, you shouldn't consult these solutions.

1. (a) Let  $p_i = \Pr_y[f(x+y) - f(y) = i]$ . The probability two random voters disagree is  $2p_0p_1$ . If  $p_0 \geq 1/2$  (so the majority is 0), then the probability a random voter disagrees with the majority is  $p_1 \leq 2p_0p_1$ ; similarly if  $p_1 \geq 1/2$  (so the majority is 1), then the probability a random voter disagrees with the majority is  $p_0 \leq 2p_0p_1$ . So we have

$$\Pr_y[f(x+y) - f(y) \neq \tilde{f}(x)] \leq 2 \Pr_{y,z}[f(x+y) - f(y) \neq f(x+z) - f(z)].$$

Now, if both (1)  $f(x+y) + f(z) = f(x+y+z)$  and (2)  $f(x+z) + f(y) = f(x+y+z)$  hold, then  $f(x+y) - f(y) = f(x+z) - f(z)$ . So at least one of (1) and (2) must fail to hold for the event on the right-hand-side to hold. Thus by a union bound

$$\begin{aligned} \Pr_{y,z}[f(x+y) - f(y) \neq f(x+z) - f(z)] &\leq \Pr_{y,z}[f(x+y) + f(z) = f(x+y+z)] \\ &\quad + \Pr_{y,z}[f(x+z) + f(y) = f(x+y+z)] \end{aligned}$$

and then by Eq. (7.1), the right-hand-side is bounded by  $2\delta$ . It follows that

$$\Pr_y[f(x+y) - f(y) = \tilde{f}(x)] \geq 1 - 4\delta.$$

- (b) We know that  $\Pr_{x,y}[f(x+y) - f(y) = f(x)] \geq 1 - \delta$  by Eq. (7.1). Now for those  $x$  such that  $f(x) \neq \tilde{f}(x)$ , we have

$$\Pr_y[f(x+y) - f(y) = f(x)] = \Pr_y[f(x+y) - f(y) \neq \tilde{f}(x)] \leq 1/2,$$

by the definition of  $\tilde{f}$ . Thus if  $p = \Pr[f(x) \neq \tilde{f}(x)]$

$$\Pr_{x,y}[f(x+y) - f(y) = f(x)] \leq p/2 + (1-p) = 1 - p/2,$$

from which we conclude  $1 - \delta \leq 1 - p/2$ , and thus  $p \leq 2\delta$ .

- (c) Fix  $x, y$ . Following the hint, we have

$$\begin{aligned} \Pr_{w,z}[f(w) + f(z) = f(w+z)] &\geq 1 - \delta \\ \Pr_{w,z}[f(x+w) + f(y+z) = f(x+y+w+z)] &\geq 1 - \delta \\ \Pr_{w,z}[f(x+w) - f(w) = \tilde{f}(x)] &\geq 1 - 4\delta \\ \Pr_{w,z}[f(y+z) - f(z) = \tilde{f}(y)] &\geq 1 - 4\delta \\ \Pr_{w,z}[f(x+y+w+z) - f(w+z) = \tilde{f}(x+y)] &\geq 1 - 4\delta. \end{aligned}$$

So with all but  $14\delta$  probability, *all* of the equations in the above probabilities hold, in which case

$$\tilde{f}(x+y) = f(x+y+w+z) - f(w+z) = f(x+w) + f(y+z) - f(w) - f(z) = \tilde{f}(x) + \tilde{f}(y).$$

Thus

$$\Pr_{w,z}[\tilde{f}(x) + \tilde{f}(y) = \tilde{f}(x+y)] \geq 1 - 14\delta > 0$$

(using the assumption that  $\delta > 1/14$ ). But the event in the probability does not depend on  $w, z$  so it must hold (and the probability must be 1). This is true for all  $x, y$ .

- (d) Completeness is obvious. If  $f$  passes the test with probability  $1 - \delta$ , then by definition

$$\Pr_{x,y}[f(x) + f(y) = f(x+y)] \geq 1 - \delta.$$

We can then say that there exists a *linear* function  $\tilde{f}$  satisfying  $\Pr_x[f(x) = \tilde{f}(x)] \geq 1 - 14\delta$ , because if  $\delta \geq 1/14$ , this is trivially true, and otherwise by part (b) we get that the function  $\tilde{f}$  defined using the majority function agrees with  $f$  on all but a  $2\delta < 14\delta$  fraction of the  $x$ , and by part (c) we get that  $\tilde{f}$  is linear.

2. (a) The probability that  $A$  satisfies a given  $\phi_i$  is at most

$$(1 - \epsilon)^{\log_2 n} \leq e^{-\epsilon \log_2 n} = n^{-\epsilon \log_2 n / \ln n} = n^{-\epsilon \log_2 e} \leq n^{-\epsilon/2}.$$

Define the indicator random variable  $X_i$  to be 1 if  $A$  satisfies  $\phi_i$  and zero otherwise. Notice that  $E[X_i] \leq n^{-\epsilon/2}$ . Define  $X = \sum_i X_i$ , and notice that  $E[X] \leq n^{3-\epsilon/2}$  by linearity of expectations. Applying the Chernoff bound, we find that

$$\Pr[X > n^{3-\epsilon}] < e^{-n^{3-\epsilon}/6} \leq e^{-n^2}$$

as desired.

- (b) It is clear that if  $\phi$  is a YES instance, then every one of the  $\phi_i$  is simultaneously satisfied by some assignment – namely, the one that satisfies all of the clauses of  $\phi$ .

If  $\phi$  is a NO instance, then taking the union bound over all  $2^n$  possible assignments  $A$ , we find that

$$\Pr[\exists A \text{ that satisfies more than } n^{3-\epsilon} \text{ of the } \phi_i] \leq 2^n e^{-n^2} < 1/2,$$

as desired.

- (c) We produce a graph with  $n^3$  sets of nodes. Each node in set  $i$  corresponds to one of the possible satisfying assignments to  $\phi_i$ . Since  $\phi_i$  consists of  $\log_2 n$  clauses with at most 3 variables each, there are at most  $n^3$  nodes in each set, for a total of  $n^6$  nodes in the graph. Now, we connect a node in set  $i$  to a node in set  $j$  (for  $i \neq j$ ) iff the assignments they represent are consistent.

Now, in the positive case, it is clear that  $G$  has a clique of size  $n^3$ , consisting of the nodes representing assignment  $A$  to each of the  $\phi_i$ .

In the negative case, we observe that a clique in  $G$  can have at most 1 node from each set (since there are no edges within the sets), so a clique of size greater than  $n^{3-\epsilon}$  must imply an assignment that is simultaneously consistent with more than  $n^{3-\epsilon}$  of the  $\phi_i$ , a contradiction.

- (d) Note that  $N = n^c$  for some constant  $c$ . Set  $\delta = \epsilon/(c + 1)$ . Given any language  $L \in \mathbf{NP}$  and an input  $x$  we obtain  $\phi$  using the PCP theorem, and then use randomness to construct  $G$  from  $\phi$  as described above. We then run the  $N^\delta$ -approximation algorithm on the instance  $(G, k = n^3)$ . If it returns a clique of size at least  $k/N^\delta > n^{3-\epsilon}$ , then we accept; otherwise we reject.

Now, if  $x \in L$ , then our construction will always produce a graph  $G$  with a clique of size  $n^3$ , and our approximation algorithm is guaranteed to return a clique of size at least  $k/N^\delta$ , and we will accept.

If  $x \notin L$ , then our construction will produce a graph with no clique larger than  $n^{3-\epsilon}$  with probability 1/2 and in this case we will reject (because no clique returned by the approximation algorithm will be large enough).

Thus we have a randomized algorithm that always accepts if  $x \in L$ , and rejects with probability at least 1/2 if  $x \notin L$ . We conclude that  $L \in \mathbf{coRP}$ , and therefore  $\mathbf{NP} \subseteq \mathbf{coRP}$ . Now, we know from the midterm that  $\mathbf{NP} \subseteq \mathbf{coRP} \subseteq \mathbf{BPP}$  implies that  $\mathbf{NP} \subseteq \mathbf{RP}$ . We conclude that  $\mathbf{NP} \subseteq (\mathbf{RP} \cap \mathbf{coRP}) = \mathbf{ZPP}$  as required.

Alternatively, we could argue directly that  $\mathbf{NP} \subseteq \mathbf{coRP}$  implies  $\mathbf{coNP} \subseteq \mathbf{RP}$ , and therefore

$$\mathbf{NP} \subseteq \mathbf{coRP} \subseteq \mathbf{coNP} \subseteq \mathbf{RP} \subseteq \mathbf{NP}$$

and so  $\mathbf{NP} = \mathbf{coRP} = \mathbf{RP} = \mathbf{ZPP}$ .

3. (a) We describe a recursive divide and conquer algorithm. As the base case, if  $n = 1$  then it is easy to evaluate  $f(0)$  and  $f(1)$  with  $O(1)$  operations. If  $n > 1$ , then write  $f(x_1, \dots, x_n) = g(x_2, \dots, x_n) + x_1 h(x_2, \dots, x_n)$ , and recursively compute  $g$  and  $h$  at all of  $\{0, 1\}^{n-1}$ . Note that  $f(0, x_2, \dots, x_n) = g(x_2, \dots, x_n)$  while  $f(1, x_2, \dots, x_n) = g(x_2, \dots, x_n) + h(x_2, \dots, x_n)$ . So we can obtain all of the required evaluations from the values returned by the recursive calls.

Preparing  $g$  and  $h$  for the recursive calls requires  $O(2^n)$  operations (since we just need to go through the coefficients one by one), and computing the evaluations of  $f$  from the returned lists takes  $O(2^n)$  operations (we need to copy one list of size  $2^{n-1}$  and then output the element-wise sum of two lists of size  $2^{n-1}$ ).

Let  $T(n)$  denote the number of operations when there are  $n$  variables. Then we have

$$T(n) \leq 2T(n-1) + O(2^n),$$

from which we conclude  $T(n) = O(n2^n)$ . We know that  $f(x) \leq \text{quasipoly}(|C|)$ , and we are always summing positive numbers, so the maximum magnitude of any integer in these operations is  $\text{quasipoly}(|C|)$ , and arithmetic operations on such integers take time  $O(\text{poly}(\log |C|))$ . The overall running time is  $O(\text{quasipoly}(|C|) + 2^n \cdot \text{poly}(n))$  to obtain the representation in the theorem, plus  $O(2^n \text{poly}(n, \log |C|))$  for evaluating  $f(x)$  at all of  $\{0, 1\}^n$  plus the time to perform  $2^n$  evaluations of  $T$ , each of which takes time  $\text{poly}(\log \ell) = \text{poly}(\log |C|)$ .

- (b) Plug in each of the  $2^{n'}$  possible values, resulting in a new circuit, and let  $C'$  be the OR of these  $2^{n'}$  circuits, which remains an **ACC**-type circuit, of size  $\text{poly}(n) \cdot 2^{n'}$ . Clearly  $C'$  is satisfiable iff  $C$  is. Applying the procedure in the previous part to  $C'$  takes time

$O(2^{n-n'} \text{poly}(n) + \text{quasipoly}(\text{poly}(n) \cdot 2^{n'}))$ . By choosing  $n' = n^\epsilon$  for  $\epsilon$  a sufficiently small constant we can make  $\text{quasipoly}(\text{poly}(n)2^{n'}) < O(2^{\sqrt{n}})$  (say), and the overall running time is thus  $O(2^{n-n^\delta})$  for a constant  $\delta < \epsilon$ .

- (c) Consider the language consisting of pairs  $(C, i)$  where  $C$  is a **SUCCINCT 3-SAT** instances, and the  $i$ -th bit of the lexicographically first satisfying assignment to the 3-SAT formula encoded by  $C$  is one. We claim this language is in  $\mathbf{E}^{\mathbf{NP}}$ . Indeed, in time at most  $2^{|C|} \text{poly}(|C|)$ , we can extract the 3-SAT formula encoded by  $C$ . Then using the *NP* oracle, we can perform a binary search to find the lexicographically first satisfying assignment, if there is one. Then it is easy to accept or reject based on the  $i$ -th bit of this assignment. Since we are assuming  $\mathbf{E}^{\mathbf{NP}} \subseteq \mathbf{ACC}$ , there exists an **ACC** circuit  $W_x$  as described in the problem by hardwiring  $C_x$  as part of the input to the **ACC** circuit decided this language (for inputs of the appropriate length).
- (d) To begin, we perform the **SUCCINCT 3-SAT** reduction from language  $L$ , with input  $x$ , to obtain  $C_x$ . Set  $n = |x|$ . So far this takes polynomial time.

Now, we need to argue that  $D, G, V$  exist. For this we note that the following are functions in  $\mathbf{P}$ :

- given a circuit  $C$  and an input  $x$ , output  $C(x)$
- given a circuit  $C$  and an input  $i$ , output the gate information for gate  $i$  of circuit  $C$
- given a circuit  $C$ , an input  $i$ , and an input  $x$ , output the value of gate  $i$  when evaluating  $C$  on input  $x$ .

Since we are assuming  $\mathbf{E}^{\mathbf{NP}} \subseteq \mathbf{ACC}$ , we certainly have  $P \subseteq \mathbf{ACC}$ . Thus there are polynomial-size families of **ACC** circuits computing each of these functions. Hardwiring  $C_x$  as the circuit  $C$  in the **ACC** circuit of the appropriate size yield the **ACC** circuits  $D$ ,  $G$ , and  $V$ , respectively. As usual, it is more challenging to actually get our hands on these circuits, and for this we use the ability to guess and verify as suggested in the hint.

We now nondeterministically guess  $D, G, V, W_x$ . Given guessed **ACC** circuits  $D, G, V$ , we note the following:

- in polynomial time, we can verify that  $G$  is correct, by running through all of its inputs (there are at most polynomially many) and consulting  $C_x$ ,
- $V$  is correct iff the following holds for all  $x$  and all  $i$ : evaluate  $G(i)$  to determine the gate type of gate  $i$ , and its at most 2 input gates  $j, k$ ; check that  $V(x, i)$ ,  $V(x, j)$  and  $V(x, k)$  are consistent (e.g., if the gate type of gate  $i$  is OR, and the two input gates are gate  $j$  and gate  $k$ , then we check that  $V(x, i) = V(x, j) \vee V(x, k)$ ), and
- $D$  is correct iff for all  $x$ :  $D(x) = V(x, i^*)$  where  $i^*$  is the index of the output gate (we can standardize our gate numbering so this is always gate 0, for example).

Observe that after the universal quantification of  $x, i$ , the checks in the last two bullets can be expressed as an **ACC** circuit with  $\ell = |(x, i)| = n + O(\log n)$  inputs, because in both cases we are performing a constant number of evaluations of **ACC** circuits and using those values on a computation involving at most  $O(\log n)$  bits (which we could even afford to write out as a CNF). Therefore, we can use part (b) to perform these checks in  $O(2^{\ell - \ell^\delta})$  time.

If  $D, G, V$  pass these checks, then we are left checking whether  $W_x$  encodes a satisfying assignment to  $\phi_x$  (the 3-SAT instance succinctly encoded by  $C_x$  – and now  $D$  as well). Recall that  $C_x$  (and  $D$ ) have at most  $m = n + 5 \log n$  inputs. Thus there are at most  $2^m$  clauses in  $\phi_x$ , and  $\phi_x$  involves at most  $2^m$  variables. Thus, given a clause number  $i$ , it takes  $\text{poly}(m) = \text{poly}(n)$  many evaluations of  $D$  to extract a description of clause  $i$ . This consists of the names of the three variables  $j_1, j_2, j_3$  appearing in the clause, and whether or not they are negated. We can then check whether  $W_x(j_1), W_x(j_2), W_x(j_3)$  satisfy the clause. Again, after the universal quantification of the clause number  $i$ , this check can be expressed as an **ACC** circuit with  $m$  inputs, as we are just plugging a sequence of evaluations of  $D$  into  $W_x$ , three times, and possibly negating the results before taking their OR. Therefore, we can use part (b) to perform these checks in  $O(2^{m-m^\delta})$  time.

Altogether, on input  $x$  (an instance of  $L$ , an arbitrary language in  $\mathbf{NTIME}(2^n)$ ), we guess  $\text{poly}(n)$  bits (to describe  $D, G, V, W_x$ ), and perform  $\text{poly}(n)$  deterministic computation (to produce  $C_x$ , to check the correctness of  $G$ , and to set up the **ACC** circuits to be used in the two invocations of part (b)), followed by  $O(2^{\ell-\ell^\delta}) + O(2^{m-m^\delta})$  steps to invoke part (b) twice. Since  $\ell, m \leq n + O(\log n)$ , this last quantity plus the various  $\text{poly}(n)$  quantities is at most  $O(2^{n-n^{\delta'}})$  for some constant  $\delta' > 0$ . We accept iff  $D, G, V$  pass their checks, and  $W_x$  indeed encodes a satisfying for  $\phi_x$ , which happens iff  $x \in L$ .