# Solutions for Assignment 1

## October 22, 2003

## 1. Answers

### a) False

We will show by contradiction that  $f \notin O(g)$ . Suppose  $f \in O(g)$ , then there exist constants c > 0 and  $n_0 > 0$  such that  $\frac{f(n)}{g(n)} \le c$  for all  $n \ge n_0$ . But

$$\lim_{n \to \infty} \frac{f(n)}{g(n)} = \lim_{n \to \infty} n = \infty$$

which contradicts the previous statement.

#### **b)** False

Take f(n) = 1 and g(n) = n. Then  $f_{min}(n) = 1$ , and

$$(f+g)(n) = f(n) + g(n) = n+1$$

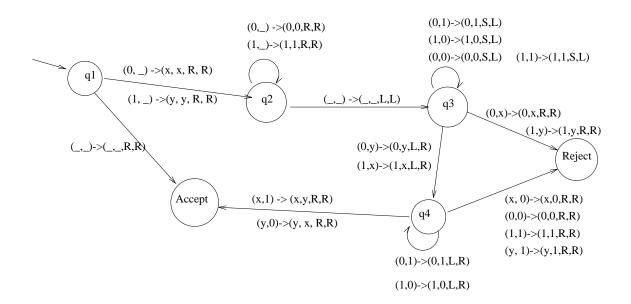
 $f+g \notin \Theta(f_{min})$  since  $f+g \notin O(f_{min})$  (because  $\lim_{n\to\infty} \frac{f(n)+g(n)}{f_{min}(n)} = \lim_{n\to\infty} \frac{n+1}{1} = \infty$ ).

### c) True

Suppose  $f \in O(g)$ . Then we know there exist postive constants c and  $n_0$  such that  $f(n) \leq cg(n)$  for all  $n \geq n_0$ . Thus  $\frac{1}{c}f(n) \leq g(n)$  for all  $n \geq n_0$ ; so there exists a constant  $d = \frac{1}{c} > 0$  such that  $df(n) \leq g(n)$  for all  $n \geq n_0$ , which implies  $g \in \Omega(f)$ .

#### d) False

Counterexample: f(n) = 1, g(n) = n. In this case  $f \in O(g)$  since for c = 1 and  $n_0 = 1$ ,  $f(n) = 1 \le n = 1 \cdot g(n)$  for all  $n \ge 1$ . However  $g \notin O(f)$  since  $\lim_{n\to\infty} \frac{g(n)}{f(n)} = \lim_{n\to\infty} n = \infty$ , which is not below any constant.

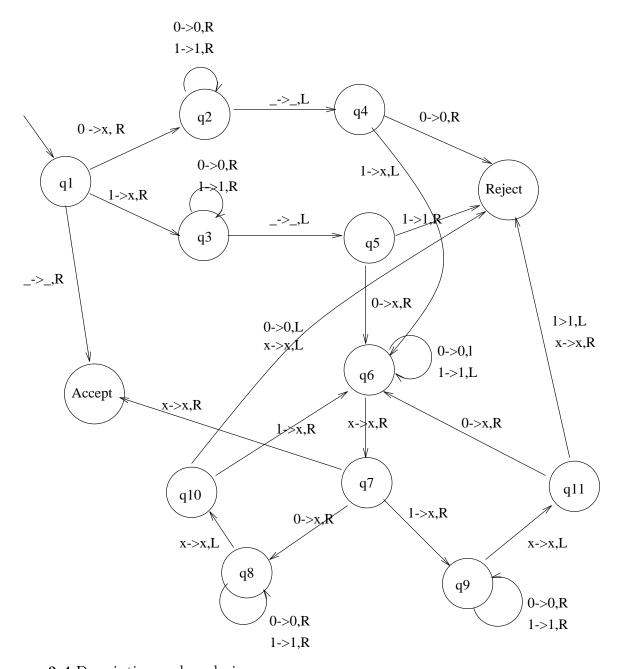


#### 2.2

- Step1 (states q1 and q2): Go over tape 1 writing its conents to tape 2.
- Step2 (state q3): Rewind tape 2 while keeping tape 1 in its last input symbol.
- Step 3 (state q4): Go over both tapes, tape 1 from right to left and tape 2 from left to right, comparing symbols. If symbols are equal at any point, rejest. Stop when reaching leftmost point of tape 1.

Let T(n) be the worst case running time of  $M_2$ . Steps 1, 2, and 3 each consists of a single scanning of the tape so  $T(n) \in \Theta(n)$ 

2.3



## 2.4. Description and analysis

• Step 1: Scan the tape crossing the first and last symbols. If symbols were equal then reject.

- Step 2: Scan the tape again crossing the first symbol that is not an X and the last symbol that is not an X. If original symbols were equal then reject. If no symbol other than X has been found, accept.
- Step 3: GO to step 2.

The first step can be done with a single scan of the tape, so it takes O(n) steps. Similarly step 2 takes O(n) steps. Step 2 is repeated at most  $\frac{n}{2}$  times, since at each step two symbols are transformed into X. Therefore the total running time is in  $O(n^2)$ 

```
2.5 RAM Program
We will use registers:
       r_0 = \text{accumulator}
       r_1 = \text{store index of leftmost symbol to be examined}
       r_2 = store index of rightmost symbol to be examined
1 \text{ LOAD} = 1
2 STORE 1
3 \text{ READ} \uparrow 1
4 STORE 2
5 LOAD 1
6 \text{ ADD} = 2
7 STORE 1
8 LOAD 2
9 STORE ↑ 1
10 ADD 1
11 JZERO 16
12 LOAD 1
13 SUB 1
14 STORE 1
15 JUMP 3
                    // loop 3-15 reads i_1, \ldots, i_n into r_3, \ldots, r_{n+2}
16 LOAD 1
17 \text{ SUB} = 1
                     // r_2 \leftarrow n + 2
18 STORE 2
19 \text{ LOAD} = 3
                     // r_1 \leftarrow 3
20 STORE 1
21 LOAD 2
22 SUB 1
```

```
// if r_2 < r_1 then go to accept
23 JNEG 34
24 LOAD ↑ 1
25 \text{ SUB} \uparrow 2
26 JZR 36
                    // if leftmost symbol = rightmost symbol then reject
27 LOAD 1
28 \text{ ADD} = 1
                      // r_1 \leftarrow r_1 + 1
29 STORE 1
30 LOAD 2
31 \text{ SUB} = 1
                      // r_2 \leftarrow r_2 - 1
32 STORE 2
33 JMP 21
                       // accept
34 \text{ LOAD} = 1
35 HALT
                       // reject
36 \text{ LOAD} = 0
37 HALT
```

#### 2.6 Running Time

The first loop (line 3 to line 15) simply reads the input which takes time O(n). The second loop (line 21 to line 33) runs for  $\left\lceil \frac{n}{2} \right\rceil$  iterations, so it takes time in O(n). The whole program takes time in O(n).

**3.** Let  $A \in P$ ,  $A \neq \Sigma^*$  and  $A \neq \phi$ . Since  $A \in P$ , we know  $A \in NP$ . It remains to show A is NP-hard.

We need to show that  $L \leq_p A$  for all  $L \in NP$ .

Let  $L \in NP$  be an arbitrary language in NP. Since P = NP, we conclude  $L \in P$  and so there exists a polynominal time algorithm D that decides L. We will build a reduction algorithm F (reduction from L to A) in the following way:

```
Algorithm F(x) { Let a be a string in A, Let b be a string in \Sigma^* \setminus A if D(x) = 1 then return a else return b } if x \in L then D(x) = 1 and F(x) = a \in A. if x \notin L, then D(x) = 0 and F(x) = b \notin A.
```

Moreover, since D runs in polynominal time and F simply calls D (polynomial time) and does a constant number of steps, F runs in polynominal

time.

**4.** Since HAMPATH  $\in P$ , there exists a polynomial time algorithm A that decides HAMPATH.

Below we describe the polynomial time algorithm B that finds a hamiltonian path from u to v, if one exists.

```
Algorithm B (G = (V, E), u, v)
1
         if A(G, u, v) = 0 then
2
                output "no hamiltonian path from u to v exists"
3
         else
         \{ n = |V|
4
5
                w = u
6
                for i = 1 to n do
7
                \{ PATH[i] = w;
8
                       G' \leftarrow G \setminus \{w\};
9
                       for each edge (w, t) in G coming out of w do
                        \{ \text{if } (A(G', t, v) = 1) \text{ then } 
10
11
                               \{ w = t;
12
                                  break this loop;
13
14
                       \overset{"}{G} \leftarrow G';
15
16
17
                  return (PATH);
18
19 }
```

The correctness of the algorithm comes from the idea that if there exists a hamiltonian path from u to v, having second vertex t, when we remove u from G, there must be hamiltonian path from t to v in the reduced graph. Iterating this principle, we can determine the sequence of the vertices in the hamiltonian path.

The algorithm runs in polynomial time, for the following reasons:

Let T(n, m) be the worst case running time of algorithm A for a graph with n vertices and m edges.

```
step1 takes time T(n, m)
step 2-5 takes constant time
```

```
loop in line 6 runs n times loop in line 9 runs at most m times therefore line 10 over all iteration run in n \times m \times T(n, m) line 11- 12 run in time n \times m line 8 and 15 are repeated n times and each time it may take O(n \times m), so totally, it is O(n^2 \times m)
```

Thus the running time for B is in  $O(n^2 \times m + n \times m \times T(n, m))$ , since T(n, m) is a polynomial, B runs in polynomial time.

## **5.**Proof:

Step 1 DoubleSAT  $\in NP$ Certificate: Two truth assignments Verification Algorithm:  $A(<\phi>, y_1, y_2)$ 

- Evaluate formula  $\phi$  using truth assignment given in  $y_1$ . If  $(y_1$  is not a satisfying assignment) then return 0;
- Evaluate formula  $\phi$  using truth assignment given in  $y_2$ . If  $(y_2$  is not a satisfying assignment) then return 0;
- If $(y_1 \neq y_2)$  then return 1; else return 0;

 $\phi$  is satisfiable  $\Leftrightarrow$  there exist two distinct truth assignments that satisfy  $\phi \Leftrightarrow$  there exists inputs  $y_1, y_2$  for A that causes A to return 1.

Algorithm A runs in polynomial time since the formula evaluation can be done in linear time with the size of  $\phi$ . Also  $y_1$  and  $y_2$  have size n, the number of variables in  $\phi$ , so these comparison also take linear time.

```
Step 2 We will prove SAT \leq_p DoubleSAT

Step 3 We will describe the reduction algorithm F.

Algorithm F(<\phi>)

{

Let x_1, x_2, ..., x_n be the variable in \phi

Build another formula \phi_2 equivalent to \phi but substituting variables x_1, x_2, ..., x_n by y_1, y_2, ..., y_n respectively.

Create a formula \phi' as the disjunction of \phi and \phi_2: \phi' = \phi \vee \phi_2; return \phi'
```

}

#### Step 4

- If  $\phi \in SAT$ , then there exists a satisfying assignment  $\sigma = (\sigma_1, \sigma_2, ..., \sigma_n)$ , where  $\sigma_i = 0$  or 1 for  $\phi$ . Thus,  $\sigma$  satisfies  $\phi$  and  $\phi_2$ . Let  $\tau = (\tau_1, \tau_2, ..., \tau_n)$ ,  $\tau \neq \sigma$ . Therefore  $\sigma^1 = (\sigma_1, \sigma_2, ..., \sigma_n, \tau_1, \tau_2, ..., \tau_n)$  and  $\sigma^2 = (\tau_1, \tau_2, ..., \tau_n, \sigma_1, \sigma_2, ..., \sigma_n)$  satisfy  $\phi' = \phi \vee \phi_2$ . So  $\phi' \in DoubleSAT$
- If  $\phi \notin SAT$ , then for all truth assignments  $\sigma = (\sigma_1, \sigma_2, ..., \sigma_n)$ ,  $\phi$  evaluates to 0. Therefore, for any truth assignment  $\tau$ ,  $\phi_2$  evaluates to 0. So  $\phi' = \phi \lor \phi_2$  evaluates to 0 for any truth assignment  $(\sigma_1, \sigma_2, ..., \sigma_n, \tau_1, \tau_2, ..., \tau_n)$ . So  $\phi' \notin DoubleSAT$

#### Step 5

 $\phi$  can be copied to  $\phi_2$  in linear time and both can be combined with an or operation in constant time. So F runs in linear time on the size of  $\phi$