# A METHOD TO COMPUTE LOWER BOUNDS ON CIRCUIT-SIZE COMPLEXITY

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## INTRODUCTION

In Boolean complexity theory, circuit-size complexity is one of the main targets of research ([1], [3]). Circuit-size complexity (which is also called combinational complexity or network complexity) of a Boolean function f over a base set B is defined to be the least number of gates contained in a Boolean circuit which is composed of gates in B and computes f. For most of the functions it is extremely difficult to get good estimate, not to mention the exact value, on the circuit-size complexity. Especially, good lower bounds are hard to obtain. It is, therefore,

quite welcome to develop any new methods to derive good bounds on circuit-size complexity.

The purpose of this note is to exhibit a technique to derive a lower bound on the circuit-size complexity of a function f by counting, individually, for **each** gate in B the number of occurrences only of that gate appearing in any circuit computing f.

Note that this method is not new: It already appeared in Tiekenheinrich [2]. However, it is rarely used in the literature and so is worth mentioning here.

#### PRINCIPLE

Let  $G=\{g_1,\ g_2,\ \dots,\ g_r\}$  be a base set, i.e., a set consisting of gates (Boolean functions) which can be used in constructing a circuit. Let f be a Boolean function which is realizable by a circuit over G. Denote by  $C_G(f)$  the circuit-size complexity of f over G, and by  $C_{G,g_1}(f)$ ,  $1\leq i\leq r$ , the least number of occurrances of the gate  $g_i$  contained in any circuit computing f over G.

It is clear that

$$c_{G}(f) \ge \sum_{i=1}^{r} c_{G,gi}(f).$$

Consequently, if a lower bound  $m_i$  on  $C_{G,g\,i}(f)$  can be derived for each  $1\leq i\leq r$ , we have a lower bound on  $C_G(f)$  as well. That

is, if

$$C_{G,gi}(f) \ge m_i, \quad 1 \le i \le r,$$

then

$$C_{G}(f) \geq \sum_{i=1}^{r} m_{i}$$
.

#### RESULT

The above principle can be applied to the following function f to get (8/3)n lower bound on the circuit-size complexity over the monotone base {AND, OR}.

Let  $G = \{AND(= \Lambda), OR(= V)\}$  and  $f^{(n)}$  be an n-variable function, n>0, defined as follows.

$$f_{1}^{(n)}(x_{1}, \dots, x_{n}) = \begin{cases} 1 & x_{1} + \dots + x_{n} \ge (1/3)n, \\ 0 & x_{1} + \dots + x_{n} < (1/3)n, \end{cases}$$

$$f_{2}^{(n)}(x_{1}, \dots, x_{n}) = \begin{cases} 1 & x_{1} + \dots + x_{n} > (2/3)n, \\ 0 & x_{1} + \dots + x_{n} \le (2/3)n, \end{cases}$$

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For the functions  $f_1^{(n)}$  and  $f_2^{(n)}$ , we can prove the following.

**Lemma 1**  $C_{G,\Omega R}(f_1^{(n)}) \ge (4/3)n - 2.$ 

Lemma 2  $C_{G,AND}(f_2^{(n)}) \ge (4/3)n - 2.$ 

As

$$f^{(n)}(x_1, \dots, x_n, 1) = f_1^{(n)}(x_1, \dots, x_n)$$

and

$$f^{(n)}(x_1, \dots, x_n, 0) = f_2^{(n)}(x_1, \dots, x_n),$$

we have

$$C_{G,OR}(f^{(n)}) \quad (\ge C_{G,OR}(f_1^{(n)})) \ge (4/3)n - 2$$

and

$$C_{G,AND}(f^{(n)}) ( \ge C_{G,AND}(f_2^{(n)}) ) \ge (4/3)n - 2.$$

The above principle can now be applied to get

**Proposition**  $C_G(f^{(n)}) \ge (8/3)n - 4.$ 

#### PROOF OF LEMMAS

**Proof of Lemma1:** Let  $\gamma$  be an optimal circuit computing  $f_1^{(n)}$  over the base set G, where  $n \ge 2$ .

Fact 1. For every input node  $x_1$  and every path p from  $x_1$  to the output node in  $\gamma$  , there exists at least one OR-gate in p.

**Proof.** Suppose, on the contrary, that there is a path from  $x_i$  to the output node which contains only AND-gates. Then by setting  $x_i = 0$  the output always takes the value 0, contradicting the definition of  $f_1^{(n)}$ .

Fact 2. For some input node  $x_i$  in  $\gamma$ , there exist at least two OR-gates  $g_1$  and  $g_2$  as well as two paths  $p_1$  and  $p_2$  each connecting  $x_i$  and the output node such that

- 1)  $g_i$  lies on  $p_i$  (j = 1, 2) and
- 2) there is no other OR-gate on  $p_i$  between  $x_i$  and  $g_i$  (i = 1, 2).

**Proof.** Assume that the statement is false, and consider, for each i, all the paths from the input node  $x_i$  to the output node. Then all of them must contain the same OR-gate, say  $\hat{g}_i$ , as the closest OR-gate to the input node  $x_i$ . Moreover, for some  $i_0$  the

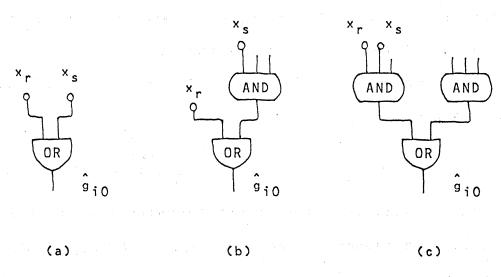


Fig. 1

gate  $\hat{g}_{i0}$  has no OR-gate as its ancestors, that is, there is no OR-gate between  $\hat{g}_{i0}$  and input nodes. The situation is shown in Fig. 1. In cases (a) and (b) the circuit is made not to depend on  $x_s$  by assigning  $x_r = 1$ , and in case (c) it is made not to depend on  $x_s$  by assigning  $x_r = 0$ . In either case this contradicts the definition of the function.

Now, Lemma 1 can be proved by mathematical induction. As the basis, note that

$$C_{G,OR}(f_1^{(1)}) = 0,$$

$$C_{G,OR}(f_1^{(2)}) = 1,$$

$$C_{G,OR}(f_1^{(3)}) = 2.$$

Suppose  $C_{G,QR}(f_1^{(n)}) \ge (4/3)n - 2$  holds for  $n \ge 3$  and consider a circuit  $\gamma^{(n+3)}$  computing  $f_1^{(n+3)}$ . Choose an input node  $x_i$  in  $\gamma^{(n+3)}$  as in Fact 2, and substitute the value 0 to  $x_i$ . By this procedure at least two OR-gates can be eliminated from  $\gamma^{(n+3)}$  without affecting the result of the circuit. Repeat the same procedure to the resulting circuit, eliminating at least two OR-gates again. Finally, substitute the value 1 to any one of the rermaining input nodes. Then what we have is a circuit computing  $f_1^{(n)}$ , because

$$f_1^{(n+3)}(x_1, \dots, x_n, 0, 0, 1) = f_1^{(n)}(x_1, \dots, x_n)$$

holds. (Here, w.l.o.g., the inputs to which the value 0 or 1 is substituted are assumed to be  $x_{n+1}$ ,  $x_{n+2}$  and  $x_{n+3}$ .) By hypothesis, we have

$$C_{G,OR}(f_1^{(n+3)}) \ge ((4/3)n - 2) + 2 + 2$$
  
=  $(4/3)(n+3) - 2$ .

This completes the proof of Lemma 1.

QED

**Proof of Lemma 2:** In a circuit  $\gamma$  over the base  $G=\{AND, OR\}$  computing  $f_2^{(n)}(x_1, \ldots, x_n)$ , replace each AND-gate by an OR-gate and each OR-gate by an AND-gate. Then, by de Morgan's law, the resulting circuit  $\gamma'$  computes  $\overline{f_2^{(n)}(\overline{x}_1, \ldots, \overline{x}_n)}$ . Now it is easy to see that

$$f_2^{(n)}(\bar{x}_1, \dots, \bar{x}_n) = f_1^{(n)}(x_1, \dots, x_n),$$

and Lemma 2 follows from Lemma 1.

QED

#### REFERENCES

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- [2] Tiekenheinrich, J., A 4n-lower bound on the monotone network complexity of a one-output Boolean function, Information Processing Letters 18 (1984) 201-202.
- [3] Wegener, I., The Complexity of Boolean Functions, Wiley-Teubner, 1987.