

Boolean Circuit Depth (II)

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A Quick Recap

Definition

The **depth** $d(C)$ of a circuit C is the length of the longest path from the output node to an input node. The **size** $L(F)$ of a formula F is the number of its input nodes.

For a function f , the **depth complexity** $d(f)$ is the minimum depth of a circuit computing f and the **size complexity** $L(f)$ is the minimum size of a formula computing f .

The measure $d_m(C)$, $L_m(F)$, $d_m(f)$, and $L_m(f)$ are defined similarly for monotone circuits, formulas, and functions respectively.

Definition

For a Boolean function $f : \{0, 1\}^n \rightarrow \{0, 1\}$ let

$$X = f^{-1}(1) \quad \text{and} \quad Y = f^{-1}(0).$$

We define

$$R_f = \{(x, y, i) \mid x \in X, y \in Y, \text{ and } i \in \{1, \dots, n\} \text{ with } x_i \neq y_i\}.$$

For monotone f we also define

$$M_f = \{(x, y, i) \mid x \in X, y \in Y, \text{ and } i \in \{1, \dots, n\} \text{ with } x_i = 1 \text{ and } y_i = 0\}.$$

Theorem

$$d(f) = D(R_f) \quad \text{and} \quad L(f) = C^P(R_f).$$

Theorem

$$d_m(f) = D(CM_f) \quad \text{and} \quad L_m(f) = C^P(M_f).$$

We prove circuit lower bounds by reductions to lower bounds for communication complexity.

Matching

Given a graph G on n vertices,

$$\text{MATCH}(G) = \begin{cases} 1, & \text{if there is a matching of size } \geq n/3 \text{ in } G, \\ 0, & \text{otherwise.} \end{cases}$$

Theorem

$$d_m(\text{MATCH}) = \Omega(n).$$

Given a directed graph G on n nodes,

$$\text{STCON}(G) = \begin{cases} 1, & \text{if there is a path in } G \text{ from vertex 1 to vertex } n \\ 0, & \text{otherwise.} \end{cases}$$

Theorem

$$d_m(\text{STCON}) = \Omega(\log^2 n).$$

Set Cover

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$$f(z_1, \dots, z_t) = \begin{cases} 1, & \text{if there exists a row } x \text{ of } M_g \text{ such that} \\ & \text{for all } i \text{ we have } (x \in R_i \implies z_i = 1) \\ 0, & \text{otherwise.} \end{cases}$$

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Assume $D(g) = N^2(g)$, then the function f has $t = 2^{N(g)}$ variables and

$$d_m(f) = D(M_f) \geq D(M) \geq D(g) = \log^2 t.$$

Similarly $L(f) = \Omega(t^{\log t})$.

We can write

$$f(z_1, \dots, z_t) \equiv \exists x \in \{0, 1\}^n : \\ [(x \in R_1) \implies (z_1 = 1)] \wedge \dots \wedge [(x \in R_t) \implies (z_t = 1)].$$

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3. and both p and s are polynomially bounded in t .

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SET-COVER

Input: A collection of m sets over a universe of ℓ elements and a number d .

Problem: Is there a subcollection of d sets that covers the whole universe?

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2. For every x_j there are two sets $A_{x_j=1}$ and $A_{x_j=0}$. $A_{x_j=1}$ contains all terms in which x_j appears, and $A_{x_j=0}$ contains all terms in which \bar{x}_j appears.

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3. Finally, set $d = p$.

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Then the cover induces a satisfying assignment, since the universe contains all the terms.

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$$d_m(\text{SET-COVER}) \geq d(f) - O(\log t) = \Omega(\log^2 t).$$

Monotone Constant-Depth Circuits

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However, the depth $d(f)$ is equal to the **round complexity** of the protocol, the number of alternations between the communication from Alice to Bob and the communication from Bob to Alice.

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We show that any depth $k - 1$ formula computing f has size exponential in m .

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It is known that the $(k-1)$ -round communication complexity $D^{k-1}(T_k)$ of T_k is

$$D^{k-1}(T_k) = \Omega(m/\text{polylog}(m)).$$

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Reduction from T_k to M_f (2)

3. Alice computes a string x of length n by putting 1 in all coordinates j for $j \in S_k$ and 0 elsewhere.
4. Bob computes a string y of length n by putting 0 in all coordinates j for $j \in Q_k$ and 1 elsewhere.
5. Finally, Alice and Bob use the protocol for M_f on (x, y) and output the result.

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$f(y) = 0$ By induction on i from $k - 1$ to 1 if each node in Q_{i+1} computes the value 0, then so do all the nodes in Q_i .

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Finally, we prove that there is exactly one j with $x_j = 1$ and $y_j = 0$ by showing that for every $i \in \{1, \dots, k\}$ the set $S_i \cap Q_i$ includes a single node v_i , which is the node in level i that the path from the root reaches.

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- ▶ The case for even i is symmetric.

The lower bound

We conclude for any constant k , the size of any depth $k - 1$ formula for f is

$$C^{P,k-1}(M_f) = \Omega\left(2^{D^{k-1}(M_f)/(k-1)}\right) = \Omega\left(2^{D^{k-1}(T_f)/(k-1)}\right) = \Omega\left(2^{m/\text{polylog}(m)}\right).$$

Small Circuits

Q-Circuits

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The **Q-circuits complexity** of a function f , denoted by $S_Q(f)$, is the minimum cost of a Q-circuit computing f .

Worst-case partition

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Definition

Let $f : \{0, 1\}^m \rightarrow \{0, 1\}$ be a function. Let S and T be a partition of the variables x_1, \dots, x_m into two disjoint sets. The **(deterministic) communication complexity of f between S and T** , denoted $D^{S:T}(f)$, is the complexity of computing f where Alice sees all bits in S , and Bob sees all bits in T .

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The **worst-case communication complexity of f** , denoted by $D^{\text{worst}}(f)$, is the maximum of $D^{S:T}(f)$ over all such partitions.

Lemma

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$$S_Q(f) \geq \frac{D^{\text{worst}}(f)}{c_Q}.$$

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Thus, to simulate each gate, c_Q bits of communication are sufficient.

Because the circuit is of size $S_Q(f)$, the whole simulation uses at most

$$c_Q \cdot S_Q(f)$$

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Then the gate computes whether

$$\sum_{i=1}^t w_i \cdot z_i > \theta.$$

GT by threshold gates

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The “greater than” function

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Assume $x = x_n \cdots x_1$ and $y = y_n \cdots y_1$. Then

$$x > y \iff \sum_{i=1}^n 2^{i-1} x_i + \sum_{i=1}^n -2^{i-1} y_i = x - y > \theta = 0.$$

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We will show that an exponential weight, $W \geq 2^n$ is necessary for computing GT with **a single gate**.

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1. Alice computes $\sum_{z_i \in S} w_{z_i} z_i$ and send the result to Bob.
2. Bob computes $\sum_{z_i \in T} w_{z_i} z_i$, adds the result to the number received from Alice, and compares the sum with θ .

Now we apply

$$S_Q(\text{GT}) \geq D^{\text{worst}}(\text{GT})/c_Q$$

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and $D^{\text{worst}}(\text{GT}) = D(\text{GT}) = n + 1$. Thus the size of any Q -circuit computing GT is at least

$$\frac{n + 1}{\log W + 1}.$$

Depth 2 Threshold Circuits

Lemma

Assume that a function $f : \{0, 1\}^m \rightarrow \{0, 1\}$ can be computed by a depth 2 threshold circuit, whether the total weight of each gate is bounded by W . Then

$$R_{1/2+1/(4W)}^{\text{pub, worst}}(f) \leq \log W + 1.$$

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The function does not change, and the new total weight $W' \leq 4W$.

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Let f_1, \dots, f_t be the functions that are the inputs to the top gate and w_1, \dots, w_t . These functions are either constants or input variables or threshold gates, all satisfy

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 - 3.1 If $b = 0$, then the output is chosen uniformly at random from 0 and 1.
 - 3.2 If $b = 1$, then the output is 1 if $w_i > 0$ and 0 if $w_i < 0$.

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Therefore the total success probability is at least

$$\frac{\alpha}{2} + \frac{1 - \alpha}{2} + \frac{1}{W'} = \frac{1}{2} + \frac{1}{W'}.$$

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That is, provided the weights are small, the size of the circuit has to be exponential.

Thank You!