

## Formula lower bounds (lecture notes)

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### 1 The formula balancing lemma

**Definition 1** (Formulas). A *formula* is a circuit  $C: \{0, 1\}^n \rightarrow \{0, 1\}$  in which each gate has fan-out (out-degree) at most 1. In other words, the underlying graph structure is a tree. A *De Morgan formula* is a formula in which the gates are AND gates and OR gates with fan-in two, with literals and constants at the leaves. The *leafsize* of a formula is the number of leaves in the underlying tree, excluding constants.

**Lemma 1** (Formula balancing lemma). *Let  $f: \{0, 1\}^* \rightarrow \{0, 1\}$ . The following are equivalent.*

1.  $f \in \text{NC}^1$ , i.e.,  $f$  can be computed by circuits of depth  $O(\log n)$  and size  $\text{poly}(n)$  over the full binary basis.
2. For every  $n \in \mathbb{N}$ , there is a De Morgan formula  $C_n$  of leafsize  $\text{poly}(n)$  that computes  $f$  restricted to inputs of length  $n$ .

*Proof.* (1  $\implies$  2) If  $f \in \text{NC}^1$ , then  $f$  can be computed by a “De Morgan circuit” (AND/OR gates of fan-in two, with literals and constants at the bottom) of depth  $d = O(\log n)$ . It is straightforward to show by induction on  $d$  that such a circuit can be simulated by a de Morgan formula of leafsize  $2^d$ .

(2  $\implies$  1) Let  $C$  be a De Morgan formula of leafsize  $S = \text{poly}(n)$ . We will show by induction on  $S$  that  $C$  can be computed by a De Morgan formula of depth  $3 \log S$ . By starting at the root and always choosing the child with more leaf descendants, we can find a gate  $u$  in  $C$  with children  $u_L, u_R$  such that  $u$  has at least  $S/2$  leaf descendants, whereas  $u_L$  and  $u_R$  have fewer than  $S/2$  leaf descendants each. Identify  $u$  with the function  $u(x)$  giving the output value at that gate. Let  $C_0$  and  $C_1$  be the formulas obtained from  $C$  by replacing  $u$  and all of its descendants with a 0 and a 1 respectively. Then

$$C(x) = (C_1(x) \wedge u(x)) \vee (C_0(x) \wedge \neg u(x)). \quad (1)$$

By induction, the output values of  $u_L$  and  $u_R$  can be computed by De Morgan formulas of depth at most  $3 \log(S/2)$ , hence  $u(x)$  and  $\neg u(x)$  can be computed by de Morgan formulas of depth  $1 + 3 \log(S/2)$ . Furthermore,  $C_0$  and  $C_1$  have leafsize at most  $S/2$ , so by induction,  $C_0(x)$  and  $C_1(x)$  can be computed by De Morgan formulas of depth  $3 \log(S/2)$ . Therefore,  $C$  can be computed by a De Morgan formula of depth  $3 + 3 \log(S/2) = 3 \log S$ . Note that such a formula necessarily has at most  $O(S^3) = \text{poly}(n)$  gates, hence it shows  $f \in \text{NC}^1$ .  $\square$

Thus, the question of whether  $\text{NC}^1 = \text{P/poly}$  is the question of whether circuits can be converted into formulas with polynomial overhead. The standard conjecture is “no.”

### 2 Subbotovskaya’s lower bound

For a function  $f: \{0, 1\}^n \rightarrow \{0, 1\}$ , let  $L(f)$  denote the minimum leafsize of any De Morgan formula computing  $f$ . It turns out that  $L(\text{PARITY}_n) = \Theta(n^2)$ . In this section, we will prove the weaker bound  $L(\text{PARITY}_n) \geq n^{1.5}$  via a beautiful and powerful technique called *random restrictions*.

**Definition 2** (Restrictions). A *restriction* is a string  $\rho \in \{0, 1, \star\}^n$ . If  $f$  is a function on  $\{0, 1\}^n$ , then  $f|_\rho$  is another function on  $\{0, 1\}^n$ , defined by the rule  $f|_\rho(x) = f(y)$ , where

$$y_i = \begin{cases} \rho_i & \text{if } \rho_i \in \{0, 1\} \\ x_i & \text{if } \rho_i = \star. \end{cases}$$

**Lemma 2** (Assigning a value to a single variable). *Let  $C: \{0, 1\}^n \rightarrow \{0, 1\}$  be a De Morgan formula of size  $S$ , where  $n \geq 2$ . There exists a restriction  $\rho \in \{0, 1, \star\}^n$  such that  $|\rho^{-1}(\{0, 1\})| = 1$  and  $L(C|_\rho) \leq (1 - \frac{1.5}{n}) \cdot S$ .*

*Proof.* If  $C$  is equivalent to a constant or a literal, then the lemma is trivial, so assume otherwise. The first step is to perform some simplifications to  $C$  before applying any restriction. For each subformula of the form  $x_i \wedge g$ , we can replace each occurrence of  $x_i$  in  $g$  with the constant 1, because if  $x_i = 0$ , then the subformula will evaluate to false regardless of what  $g$  does. Similarly, in a subformula of the form  $\neg x_i \wedge g$ ,  $x_i \vee g$ , or  $\neg x_i \vee g$ , we can replace each occurrence of  $x_i$  in  $g$  with an appropriate constant. Then, afterward, we can remove all constants from the formula, because  $0 \wedge g \equiv 0$ ,  $1 \wedge g \equiv g$ ,  $0 \vee g \equiv g$ , and  $1 \vee g \equiv 1$ . After making these simplifications, the new formula  $C'$  still has size at most  $S$ , and now it has the following property: For each vertex  $u$ , if  $\ell$  is a leaf that is a child of  $u$  and  $\ell'$  is a distinct leaf that is a descendant of  $u$ , then  $\ell$  and  $\ell'$  read distinct variables.

Now we are ready to perform the restriction. Pick  $\rho$  uniformly at random among all restrictions such that  $|\rho^{-1}(\{0, 1\})| = 1$ . For each leaf  $\ell$ , we divide into three cases.

- Perhaps  $\rho$  does not assign a value to the variable that  $\ell$  reads. In this case, we define  $K_\ell = \emptyset$ .
- Perhaps  $\rho$  assigns a value to the variable that  $\ell$  reads, making  $\ell$  a constant, but the parent  $u$  of  $\ell$  remains nonconstant. In this case, we define  $K_\ell = \{\ell\}$ .
- Perhaps  $\rho$  assigns a value to the variable  $\ell$  reads that makes both  $\ell$  and its parent  $u$  constant. (Note that  $0 \wedge g \equiv 0$  and  $1 \vee g \equiv 1$  for any  $g$ .) In this case, we define  $K_\ell = \{\ell, \ell'\}$ , where  $\ell'$  is any other leaf that is a descendant of  $u$ .

By construction, the function  $C|_\rho$  can be computed by a De Morgan formula constructed from  $C'$  by replacing some nodes with constants, thereby eliminating all the leaves in  $\bigcup_\ell K_\ell$ . Furthermore, because of the way we constructed  $C'$ , we have  $K_\ell \cap K_{\ell'} = \emptyset$  whenever  $\ell \neq \ell'$ . Therefore,

$$\mathbb{E}[L(C|_\rho)] \leq \mathbb{E}\left[S - \sum_\ell |K_\ell|\right] = S - \sum_\ell \left(1 \cdot \frac{0.5}{n} + 2 \cdot \frac{0.5}{n}\right) = S \cdot \left(1 - \frac{1.5}{n}\right).$$

The best case is at least as good as the average case. □

**Lemma 3** (Non-optimal shrinkage of De Morgan formulas). *Let  $f: \{0, 1\}^n \rightarrow \{0, 1\}$ , let  $p \in [0, 1]$ , and assume that  $pn$  is an integer. There exists a restriction  $\rho \in \{0, 1, \star\}^n$  such that  $|\rho^{-1}(\star)| = pn$  and  $L(f|_\rho) \leq p^{1.5} \cdot L(f)$ .*

*Proof.* Let  $k = pn$ . If  $k = 0$ , the lemma is trivial, so assume  $k \geq 1$ . By applying [Lemma 2](#)  $n - k$  times, we construct a restriction  $\rho$  such that  $|\rho^{-1}(\star)| = pn$  and

$$\begin{aligned} L(f|_\rho) &\leq L(f) \cdot \prod_{i=k+1}^n \left(1 - \frac{1.5}{i}\right) \\ &\leq L(f) \cdot \prod_{i=k+1}^n \left(1 - \frac{1}{i}\right)^{1.5} && \text{(Bernoulli's inequality)} \\ &= L(f) \cdot \left(\prod_{i=k+1}^n \frac{i-1}{i}\right)^{1.5} \\ &= L(f) \cdot (k/n)^{1.5}. \end{aligned} \quad \square$$

**Theorem 1** (Non-optimal formula lower bound for parity).  $L(\text{PARITY}_n) \geq n^{1.5}$ .

*Proof.* By [Lemma 3](#), there exists a restriction  $\rho \in \{0, 1, \star\}^n$  such that  $|\rho^{-1}(\star)| = 1$  and

$$L(\text{PARITY}_n|_\rho) \leq (1/n)^{1.5} \cdot L(\text{PARITY}_n).$$

On the other hand,  $\text{PARITY}_n|_\rho$  is non-constant, so  $L(\text{PARITY}_n|_\rho) \geq 1$ . □

### 3 Near-cubic formula lower bounds

In the previous section, we used two steps to show that there exists an explicit function  $f$  (namely, the parity function) such that  $L(f) \geq n^{1.5}$ :

1. We showed that small De Morgan formulas simplify under random restrictions.
2. We constructed  $f$  such that  $f$  does not simplify under random restrictions.

It turns out that both of the steps above can be improved, as we now discuss.

#### 3.1 Optimal shrinkage of De Morgan formulas

**Definition 3** (Random restrictions). Let  $n \in \mathbb{N}$  and  $p \in [0, 1]$ . We define  $R_p$  to be the distribution over  $\{0, 1, \star\}^n$  defined as follows. To sample  $\rho \sim R_p$ , for each coordinate  $i \in [n]$  independently, set

$$\rho_i = \begin{cases} \star & \text{with probability } p \\ 0 & \text{with probability } (1-p)/2 \\ 1 & \text{with probability } (1-p)/2. \end{cases}$$

**Theorem 2** (Optimal shrinkage of De Morgan formulas [Tal14]). *For every function  $f: \{0, 1\}^n \rightarrow \{0, 1\}$  and every  $p \in [0, 1]$ , we have*

$$\mathbb{E}_{\rho \sim R_p} [L(f|\rho)] \leq O\left(p^2 \cdot L(f) + p \cdot \sqrt{L(f)}\right) \leq O(p^2 \cdot L(f) + 1).$$

The proof of [Theorem 2](#) is omitted.

#### 3.2 Andreev's function

**Theorem 3** (Near-cubic formula lower bound). *For every  $n \in \mathbb{N}$ , there exists a function  $A: \{0, 1\}^{2n} \rightarrow \{0, 1\}$  (“Andreev’s function”) such that  $A \in \mathcal{P}$  and  $L(A) \geq \tilde{\Omega}(n^3)$ .*

*Proof.* Given  $f \in \{0, 1\}^n$  and  $x^{(1)}, \dots, x^{(\log n)} \in \{0, 1\}^{n/\log n}$ , we interpret  $f$  as the truth table of a function  $f: \{0, 1\}^{\log n} \rightarrow \{0, 1\}$ , and we define

$$A(f, x^{(1)}, \dots, x^{(\log n)}) = f(\text{PARITY}_{n/\log n}(x^{(1)}), \dots, \text{PARITY}_{n/\log n}(x^{(\log n)})).$$

Clearly,  $A \in \mathcal{P}$ . To prove the formula lower bound, sample a restriction  $\rho \sim R_p$ , where  $p = \Theta((\log^2 n)/n)$ . On the one hand, by [Theorem 2](#), we have

$$\mathbb{E}[L(A|\rho)] \leq O(1 + p^2 \cdot L(A)) = O\left(1 + \frac{L(A) \cdot (\log n)^4}{n^2}\right).$$

On the other hand, let us show that  $\mathbb{E}[L(A|\rho)] \geq \tilde{\Omega}(n)$ .

After applying  $\rho$ , let us randomly assign values to the remaining variables in the “ $f$ ” portion of the input of  $A$ . This can only make the formula size smaller. By Shannon’s counting argument, with probability at least 0.9, the function  $f$  has circuit complexity  $\Omega(n/\log n)$ , hence it also satisfies  $L(f) \geq \Omega(n/\log n)$ .<sup>1</sup> Meanwhile, the probability that  $\rho$  assigns values to all  $n/\log n$  of the variables in some block  $x^{(i)}$  is at most  $\log n \cdot (1-p)^{n/\log n} \leq \log n \cdot \exp(-pn/\log n) \ll 0.1$ . Assuming this does not occur, it is possible to deterministically assign values to all but one variable in each block  $x^{(i)}$  such that  $\text{PARITY}_n(x^{(i)})$  is simply

<sup>1</sup>In fact, Shannon’s counting argument can be improved for the special case of De Morgan formulas, but let’s just use the bound that we already proved.

a single variable. Consequently, under the resulting restriction  $\rho'$ , the restricted function  $A|_{\rho'}$  is simply  $f$ , applied to a subset of the variables. Thus, we have shown that

$$\Pr[L(A|_{\rho}) \geq \Omega(n/\log n)] \geq 0.8,$$

and hence  $\mathbb{E}[L(A|_{\rho})] \geq \Omega(n/\log n)$  by Markov's inequality. Combined with the upper bound on  $\mathbb{E}[L(A|_{\rho})]$ , this implies  $L(A) \geq \tilde{\Omega}(n^3)$ .  $\square$

It is an open problem to show that some  $h \in \text{NP}$  satisfies  $L(h) \geq n^{3+\Omega(1)}$ .

## References

- [Tal14] Avishay Tal. “Shrinkage of De Morgan Formulae by Spectral Techniques”. In: *Proceedings of the 55th Annual Symposium on Foundations of Computer Science (FOCS)*. 2014, pp. 551–560. DOI: [10.1109/FOCS.2014.65](https://doi.org/10.1109/FOCS.2014.65).