

# A Switching Lemma for DNFs over $\mathbb{F}_p$ : the Canonical Decision Tree Approach

## Abstract

We prove a switching lemma for DNFs over  $\mathbb{F}_p$  via the canonical decision tree (CDT) method of Razborov, obtaining the  $M$ -independent bound

$$\Pr[\text{DT}(f|_\rho) \geq s] \leq \left( \frac{2pwq}{1-q} \right)^s$$

for any width- $w$  DNF  $f$  with *any number of terms*  $M$ , where  $\rho$  is a random restriction keeping each variable alive with probability  $q$ . The key observation is that the  $p$ -ary branching of the CDT over  $\mathbb{F}_p$  is effectively binary: all  $p-1$  nonzero branches lead to the same continuation, so the encoding requires only  $2w$  (not  $pw$ ) bits per step. For  $p=2$  this recovers the classical bound  $(4wq/(1-q))^s$ . As a corollary, depth- $d$  circuits over  $\mathbb{F}_p$  computing generalized parity require size  $\exp(\Omega_p(n^{1/(d-1)}))$ .

## 1 Introduction

Håstad's switching lemma [1] is the cornerstone of  $\text{AC}^0$  lower bounds. For a width- $w$  CNF or DNF  $f$  on  $\{0,1\}^n$  and a random restriction  $\rho_q$  keeping each variable alive with probability  $q$ :

$$\Pr[\text{DT}(f|_\rho) \geq s] \leq (Cwq)^s,$$

where  $C$  is an absolute constant and DT denotes decision tree depth. Two features are essential: the bound controls DT depth (not merely certificate complexity), and it is *independent* of the number of clauses  $M$ .

Extending this to  $\mathbb{F}_p$  is nontrivial. Over  $\mathbb{Z}_p^n$  (for a prime  $p$ ), a literal is  $\ell_i(x) = \mathbf{1}[x_i \neq 0]$ , and random restrictions assign each variable to: alive with probability  $q$ , or dead with a uniform value in  $\{0, \dots, p-1\}$  with probability  $(1-q)/p$  each. The recent work [6] proved the single-gate switching lemma  $(C_p q K/s)^s$  for fan-in  $K$  gates, which already yields tight  $\text{AC}^0$  lower bounds.

In this note we prove the full  $M$ -independent DNF/CNF switching lemma over  $\mathbb{F}_p$ , adapting Razborov's canonical decision tree proof [2, 3].

**Theorem 1** (Main). *Let  $f = T_1 \vee \dots \vee T_M$  be a width- $w$  DNF over  $\mathbb{F}_p^n$  (each term  $T_j = \bigwedge_{i \in S_j} \ell_i$  with  $|S_j| \leq w$ ). For  $\rho \sim \rho_q$ :*

$$\Pr[\text{DT}(f|_\rho) \geq s] \leq \left( \frac{2pwq}{1-q} \right)^s.$$

*An identical bound holds for width- $w$  CNFs by duality.*

For  $p=2$  and small  $q$ , this gives  $(4wq)^s$ , recovering the classical bound.

**Remark 2** (Comparison with Boolean). In the Boolean case ( $p=2$ ), the constant in the switching lemma is  $C \approx 4-7$  depending on the proof method. Our proof gives  $C_p = 2p$ ; for  $p=2$  this is  $C_2 = 4$ , matching Razborov's original argument. Whether  $C_p$  can be improved to  $O(1)$  independent of  $p$  is an interesting open question.

## 2 Setup

### 2.1 Literals and formulas over $\mathbb{F}_p$

We work over  $\mathbb{Z}_p^n$  for a prime  $p$ . A *literal* is  $\ell_i(x) = \mathbf{1}[x_i \neq 0]$ . A *width- $w$  DNF* is  $f = T_1 \vee \dots \vee T_M$  where each *term* is  $T_j = \bigwedge_{i \in S_j} \ell_i$  with  $|S_j| \leq w$ . Term  $T_j$  evaluates to 1 iff all variables in  $S_j$  are nonzero.

### 2.2 Random restrictions

A *random restriction*  $\rho \sim \rho_q$  independently sets each variable  $x_i$  to:

- alive (denoted  $*$ ) with probability  $q$ ;
- dead with value  $v \in \{0, \dots, p-1\}$  with probability  $(1-q)/p$  each.

We write  $A(\rho)$  for the set of alive variables and  $\sigma(\rho)$  for the dead assignment.

### 2.3 Effect on terms

Under restriction  $\rho$ , a term  $T_j$  is:

- *falsified*: some variable in  $S_j$  is dead with value 0;
- *fully satisfied*: every variable in  $S_j$  is dead with a nonzero value;
- *active*: no variable in  $S_j$  is dead-0, and at least one variable is alive.

The restricted function  $f|_\rho$  depends only on the alive variables.

## 3 The canonical decision tree over $\mathbb{F}_p$

**Definition 3** (CDT for DNFs). *Given a width- $w$  DNF  $f = T_1 \vee \dots \vee T_M$  and a restriction  $\rho$ , the canonical decision tree  $\mathcal{T}(f|_\rho)$  is defined recursively:*

1. *If some term  $T_j$  is fully satisfied under  $\rho$ : output 1 (leaf, depth 0).*
2. *Otherwise, let  $T_j$  be the first active term (smallest  $j$ ). Query the smallest-index alive variable  $x_i \in S_j$ .*
  - **Branch**  $x_i = 0$ : term  $T_j$  is falsified (as are all other terms containing  $x_i$ ). Recurse on  $f|_{\rho'}$  where  $\rho'$  extends  $\rho$  by setting  $x_i = 0$ .
  - **Branch**  $x_i \neq 0$ : literal  $\ell_i$  is satisfied. Recurse on  $f|_{\rho'}$  where  $\rho'$  extends  $\rho$  by setting  $x_i = 1$  (any nonzero value).
3. *If no active term exists (all terms falsified): output 0 (leaf, depth 0).*

The depth  $\text{DT}(f|_\rho)$  is the depth of  $\mathcal{T}(f|_\rho)$ .

**Proposition 4** (Binary effective branching). *In the CDT, the subtrees rooted at branches  $x_i = v$  for  $v = 1, 2, \dots, p-1$  are all identical. Hence the CDT is effectively a binary tree.*

*Proof.* The CDT state depends only on which terms are active and which variables are alive. For a variable  $x_i$  in active term  $T_j$ :

- Value 0:  $T_j$  becomes falsified.
- Any value  $v \neq 0$ : literal  $\ell_i = \mathbf{1}[x_i \neq 0]$  becomes 1. Term  $T_j$  either becomes satisfied (if  $x_i$  was the last alive variable) or continues with one fewer alive variable. The effect is the same for all  $v \neq 0$ .  $\square$

## 4 Proof of Theorem 1

We adapt Razborov's encoding argument [2, 3].

**Bad restrictions.** Let  $\text{BAD}_s = \{\rho : \text{DT}(f|_\rho) \geq s\}$ . For  $\rho \in \text{BAD}_s$ , there exists a root-to-node path in  $\mathcal{T}(f|_\rho)$  of depth  $s$ .

**Extracting a path.** Fix  $\rho \in \text{BAD}_s$ . By Proposition 4, the CDT is an effectively binary tree. Choose the *lexicographically first* (left-first) path of depth  $s$  in the CDT. This path queries variables  $x_{v_1}, \dots, x_{v_s}$  in terms  $T_{\alpha_1}, \dots, T_{\alpha_s}$ , with branch indicators  $b_1, \dots, b_s \in \{0, 1\}$  (where 0 = “value is 0” and 1 = “value is nonzero”).

**Injection.** Define  $\Phi(\rho) = (\tilde{\rho}, \tau)$  where:

- $\tilde{\rho}$  agrees with  $\rho$  except each  $x_{v_i}$  is set to *dead with value 1* (instead of alive).
- The *encoding* is  $\tau = (k_1, b_1, \dots, k_s, b_s)$ , where  $k_i = \text{pos}(x_{v_i}, S_{\alpha_i}) \in \{0, \dots, w-1\}$  is the position of  $x_{v_i}$  within term  $T_{\alpha_i}$  (in sorted order).

**Preservation of falsified terms.** We kill to value 1 (nonzero). This does not falsify any term, since falsification requires a dead variable with value 0. Formally: for any term  $T_j$ , if  $T_j$  is falsified under  $\rho$  (has some dead variable with value 0), then  $T_j$  is still falsified under  $\tilde{\rho}$  (the same dead-0 variable persists). Conversely, if  $T_j$  is not falsified under  $\rho$ , the only change is that some alive variables become dead-1, which does not create any dead-0 variable.

Hence: *the set of falsified terms is identical under  $\rho$  and  $\tilde{\rho}$* .

**Injectivity.** Given  $(\tilde{\rho}, \tau)$ , we reconstruct  $\rho$  iteratively:

1. Initialize  $\rho := \tilde{\rho}$ . Let  $\mathcal{F}_0$  be the set of *base-falsified* terms (those with a dead-0 variable under  $\tilde{\rho}$ ). Initialize the *CDT-falsified* set  $\mathcal{F} := \mathcal{F}_0$  and a pointer to the first non-falsified term.
2. For  $i = 1, \dots, s$ :
  - (a) *Determine the current term.* Let  $T_{\alpha_i}$  be the first term (smallest index) not in  $\mathcal{F}$ .
  - (b) *Identify the variable.* The  $k_i$ -th variable (sorted order) in  $T_{\alpha_i}$  is  $x_{v_i}$ . This variable is currently dead with value 1 in  $\rho$ .
  - (c) *Revive.* Set  $\rho(x_{v_i}) := *$  (alive).
  - (d) *Update CDT-falsified set.* If  $b_i = 0$  (zero branch), add to  $\mathcal{F}$  *every* term  $T_j$  containing  $x_{v_i}$ . If  $b_i = 1$  (nonzero branch),  $\mathcal{F}$  is unchanged and we remain at term  $T_{\alpha_i}$  (the next step will continue querying this term unless it becomes fully satisfied).
3. Output  $\rho$ .

This procedure is well-defined because:

- *Step 2(a):* the base-falsified set  $\mathcal{F}_0$  is identical under  $\tilde{\rho}$  and  $\rho$ , since we only change dead-1 to alive (never creating or removing dead-0 variables). The CDT-falsified terms added in step 2(d) exactly mirror the CDT’s own state: on the original  $\rho$ , querying  $x_{v_i}$  and branching  $x_{v_i} = 0$  falsifies every term containing  $x_{v_i}$ , not just the current one.
- *Step 2(b):* the variable at position  $k_i$  in the current term is dead-1 (it was set to dead-1 by the forward map and has not been revived in a previous step, since the CDT queries each variable at most once).

Since the procedure uniquely determines  $\rho$  from  $(\tilde{\rho}, \tau)$ , the map  $\Phi$  is injective on each fiber  $\text{BAD}_\tau := \{\rho \in \text{BAD}_s : \Phi(\rho) \text{ has encoding } \tau\}$ .

**Encoding count.**

$$|\{\tau\}| \leq (2w)^s,$$

since each step contributes  $w$  choices for the position  $k_i$  and 2 choices for the branch  $b_i$ . Crucially,  $b_i$  records only “zero vs. nonzero” (not the specific nonzero value), which suffices by Proposition 4.

**Probability ratio.** Each  $x_{v_i}$  changes from alive (probability  $q$ ) to dead with value 1 (probability  $(1 - q)/p$ ):

$$\frac{\Pr[\rho]}{\Pr[\tilde{\rho}]} = \prod_{i=1}^s \frac{q}{(1 - q)/p} = \left(\frac{pq}{1 - q}\right)^s.$$

**Combining.**

$$\begin{aligned} \Pr[\text{BAD}_s] &= \sum_{\tau} \Pr[\text{BAD}_\tau] = \sum_{\tau} \sum_{\rho \in \text{BAD}_\tau} \Pr[\rho] \\ &= \sum_{\tau} \sum_{\rho \in \text{BAD}_\tau} \Pr[\tilde{\rho}] \cdot \left(\frac{pq}{1 - q}\right)^s \\ &\leq \sum_{\tau} 1 \cdot \left(\frac{pq}{1 - q}\right)^s = (2w)^s \cdot \left(\frac{pq}{1 - q}\right)^s = \left(\frac{2pwq}{1 - q}\right)^s. \end{aligned}$$

## 5 AC<sup>0</sup> lower bound

**Corollary 5.** *For any prime  $p$  and constant depth  $d \geq 2$ , any depth- $d$  circuit over  $\mathbb{F}_p$  computing  $\text{PAR}_p(x) = \mathbf{1}[\sum_i x_i \not\equiv 0 \pmod{p}]$  has size  $\exp(\Omega_p(n^{1/(d-1)}))$ .*

*Proof.* The argument is standard. Let  $C$  be a depth- $d$  circuit of size  $S$  with bottom fan-in  $w$ . Apply  $d - 1$  rounds of random restrictions with  $q = c/(pw)$  for a small constant  $c$ .

At each round, every bottom gate (a width- $w$  DNF or CNF) satisfies

$$\Pr[\text{DT} \geq s] \leq (2pwq/(1 - q))^s = (2c/(1 - q))^s \leq (3c)^s$$

for small  $q$ . Setting  $s = \lceil C' \ln S \rceil$  for large  $C'$ , this is  $< 1/S^2$ . By a union bound over  $S$  gates, all gates simplify to DT depth  $\leq s$  with positive probability.

Each gate is replaced by its depth- $s$  decision tree expansion, reducing the circuit depth by 1 and setting the new bottom fan-in to  $s = O(\ln S)$ . After  $d - 1$  rounds:

$$n' = n \cdot \prod_{i=1}^{d-1} q_i \geq \frac{n}{(Cp)^{d-1} \cdot w \cdot (\ln S)^{d-2}}$$

alive variables survive. For  $\text{PAR}_p$  to remain non-constant, we need  $n' \geq 1$ , giving  $w \cdot (\ln S)^{d-2} \leq n/(Cp)^{d-1}$ . Since  $w \leq S$ , this forces  $S \geq \exp(\Omega_p(n^{1/(d-1)}))$ .  $\square$

## 6 Discussion

### 6.1 The key observation: binary effective branching

The CDT for a DNF over  $\mathbb{F}_p$  branches  $p$  ways at each query, but  $p-1$  of these branches are identical (all nonzero values have the same effect on clause satisfaction). This is the crucial observation: the encoding needs only 1 bit per step to specify the branch (zero vs. nonzero), rather than  $\log_2 p$  bits.

In contrast, for more general “ $\mathbb{F}_p$ -valued” gate types (e.g., functions that distinguish between different nonzero values), the  $p$ -ary branching would be genuinely  $p$ -ary, and the encoding would require  $\log_2 p$  bits per step. This would give a bound of  $(p^2 wq/(1-q))^s$ .

### 6.2 Killing to value 1

The choice to kill staircase variables to dead-1 (any fixed nonzero value) is essential. Killing to value 0 (as in the injection of [7]) preserves the set of *surviving* clauses, but the CDT structure changes because the active terms differ. Killing to a nonzero value preserves the set of *falsified* terms (since falsification requires a dead-0 variable), which is what the CDT-based inverse needs.

### 6.3 Open problems

1. **Optimal constant.** Our bound gives  $C_p = 2p$ . For  $p = 2$ , this is  $C_2 = 4$ , matching Razborov’s proof. Can  $C_p$  be made  $O(1)$ ? In the Boolean case, Razborov’s bound of 4 is not tight; improved bounds give  $C_2 \leq 7/2$  [4]. The *correct* constant over  $\mathbb{F}_p$  is likely  $O(p)$  since the probability ratio  $pq/(1-q)$  inherently involves  $p$ .
2. **Projections and depth hierarchy.** RST [5] extended the switching lemma to random *projections* to prove average-case depth hierarchy for the Boolean PH relative to a random oracle. Extending this to  $\mathbb{F}_p$  would give analogous results for  $\mathbb{F}_p$ -valued computation models.
3. **ACC<sup>0</sup>.** Our switching lemma handles AND/OR gates over  $\mathbb{F}_p$ . The frontier of ACC<sup>0</sup>-style lower bounds is ACC<sup>0</sup> (circuits with MOD <sub>$m$</sub>  gates). Connecting the  $\mathbb{F}_p$  framework to ACC<sup>0</sup> lower bounds remains a major open problem.

## References

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