

Multiparty Karchmer – Wigderson Games and Threshold Circuits

Alexander Kozachinskiy 

Department of Computer Science, University of Warwick, Coventry, UK

Alexander.Kozachinskiy@warwick.ac.uk

Vladimir Podolskii 

Steklov Mathematical Institute, Russian Academy of Sciences, Moscow, Russia

podolskii@mi-ras.ru

Abstract

We suggest a generalization of Karchmer – Wigderson communication games to the multiparty setting. Our generalization turns out to be tightly connected to circuits consisting of threshold gates. This allows us to obtain new explicit constructions of such circuits for several functions. In particular, we provide an explicit (polynomial-time computable) log-depth monotone formula for Majority function, consisting only of 3-bit majority gates and variables. This resolves a conjecture of Cohen et al. (CRYPTO 2013).

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1 Introduction

Karchmer and Wigderson established tight connection between circuit depth and communication complexity [11] (see also [12, Chapter 9]). Namely, they showed that for each Boolean function f one can define a communication game which communication complexity *exactly* equals the depth of f in the standard De Morgan basis. This discovery turned out to be very influential in Complexity Theory. A lot of circuit depth lower bounds as well as formula size lower bounds rely on this discovery [10, 13, 5, 7, 4]. Karchmer – Wigderson games have been used also in adjacent areas like Proof Complexity (see, e.g., [14]).

Karchmer – Wigderson games represent a deep connection of *two-party* communication protocols with De Morgan circuits. Loosely speaking, in this connection one party is responsible for \wedge gates and the other party is responsible for \vee gates. In this paper we address the question of what would be a natural generalization of Karchmer – Wigderson games to the multiparty setting. Is it possible to obtain in this way a connection with other types of circuits?

We answer positively to this question: we suggest such a generalization and show its connection to circuits consisting of *threshold gates*. To motivate our results we first present applications we get from this new connection.



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1.1 Applications to circuits

There are two classical constructions of $O(\log n)$ -depth monotone formulas for the Majority function, MAJ_{2n+1} . The one was given by Valiant [15]. Valiant used probabilistic method which does not give an explicit construction. The other construction is the AKS sorting network [1]. This construction actually gives polynomial-time computable $O(\log n)$ -depth $O(n \log n)$ -size monotone circuit for MAJ_n .

Several authors (see, e.g., [6, 3]) noticed that the Valiant's probabilistic argument actually gives a $O(\log n)$ -depth formula for MAJ_n , consisting only of MAJ_3 gates and variables. Is it possible to construct a $O(\log n)$ -depth circuit for MAJ_{2n+1} , consisting only of MAJ_3 gates and variables, *deterministically in polynomial time*?¹

This question was stated as a conjecture by Cohen et al. in [3]. First, they showed that the answer is positive under some cryptographic assumptions. Secondly, they constructed (unconditionally) a polynomial-time computable $O(\log n)$ -depth circuit, consisting only of MAJ_3 gates and variables, which coincides with MAJ_n for all inputs in which the fraction of ones is bounded away from $1/2$ by $2^{-\Theta(\sqrt{\log n})}$.

We show that the conjecture of Cohen et al. is true (unconditionally).

► **Theorem 1.** *There exists polynomial-time computable $O(\log n)$ -depth formula for MAJ_{2n+1} , consisting only of MAJ_3 gates and variables.*

In the proof we use the AKS sorting network. In fact, one can use any polynomial-time computable construction of $O(\log n)$ -depth monotone circuit for MAJ_{2n+1} . We also obtain the following general result:

► **Theorem 2.** *If there is a monotone formula (i.e., formula, consisting of \wedge, \vee gates and variables) for MAJ_{2n+1} of size s , then there is a formula for MAJ_{2n+1} of size $O(s \cdot n^{\log_2(3)}) = O(s \cdot n^{1.58\dots})$, consisting only of MAJ_3 gates and variables.*

Transformation from the last theorem, however, is not efficient. We can make this transformation polynomial-time computable, provided $\log_2(3)$ is replaced by $1/(1 - \log_3(2)) \approx 2.71$. In turn, we view Theorem 2 as a potential approach to obtain super-quadratic lower bounds on monotone formula size for MAJ_{2n+1} . However, this approach requires better than $n^{2+\log_2(3)}$ lower bound on formula size of MAJ_{2n+1} in the $\{\text{MAJ}_3\}$ basis. Arguably, this basis may be easier to analyze than the standard monotone basis. The best known size upper bounds in the $\{\wedge, \vee\}$ basis and the $\{\text{MAJ}_3\}$ basis are, respectively, $O(n^{5.3})$ and $O(n^{4.29})$ [8]. Both bounds are due to Valiant's method (see [8] also for the limitations of Valiant's method).

We also study a generalization of the conjecture of Cohen et al. to threshold functions. By THR_a^b we denote the following Boolean function:

$$\text{THR}_a^b: \{0, 1\}^b \rightarrow \{0, 1\}, \quad \text{THR}_a^b(x) = \begin{cases} 1 & x \text{ contains at least } a \text{ ones,} \\ 0 & \text{otherwise.} \end{cases}$$

For some reasons (to be discussed below) a natural generalization would be a question of whether THR_{n+1}^{kn+1} can be computed by a $O(\log n)$ -depth circuit, consisting only of THR_2^{k+1} gates and variables (initial conjecture can be obtained by setting $k = 2$). This question was also addressed by Cohen et al. in [3]. First, they observed that there is a construction of depth $O(n)$ (and exponential size). Secondly, they gave an explicit construction of depth $O(\log n)$, which coincides with THR_{n+1}^{kn+1} for all inputs in which the fraction of ones is bounded away from $1/k$ by $\Theta(1/\sqrt{\log n})$.

¹ Note that AKS sorting network does not provide a solution because it consists of \wedge and \vee gates.

However, no exact (even non-explicit) construction with sub-linear depth or sub-exponential size was known. In particular, Valiant's probabilistic construction does not work for $k \geq 3$. Nevertheless, in this paper we improve depth $O(n)$ to $O(\log^2 n)$ and size from $\exp\{O(n)\}$ to $n^{O(1)}$ for this problem:

► **Theorem 3.** *For any constant $k \geq 3$ there exists polynomial-time computable $O(\log^2 n)$ -depth polynomial-size circuit for THR_{n+1}^{kn+1} , consisting only of THR_2^{k+1} gates and variables.*

1.2 Applications to Multiparty Secure Computations

The conjecture stated in [3] was motivated by applications to Secure Multiparty Computations. The paper [3] establishes an approach to construct efficient multiparty protocols based on protocols for a small number of players. More specifically, in their framework one starts with a protocol for a small number of players and a formula F computing a certain boolean function. Then one combines a protocol for a small number of players with itself recursively, where the recursion mimics the formula F .

It is shown in [3] that from our result it follows that for any n there is an explicit polynomial size protocol for n players secure against a passive adversary that controls any $t < \frac{n}{2}$ players. It is also implicit in [3] that from Theorem 3 for $k = 3$ it follows that for any n there is a protocol of size $2^{O(\log^2 n)}$ for n players secure against an active adversary that controls any $t < \frac{n}{3}$ players. An improvement of the depth of the formula in Theorem 3 to $O(\log n)$ would result in a polynomial size protocol. We refer to [3] for more details on the secure multiparty computations.

1.3 Multiparty Karchmer – Wigderson games

We now reveal a bigger picture to which the above results belong to. Namely, they can be put into framework of multiparty Karchmer – Wigderson games.

Before specifying how we define these games let us give an instructive example. Consider ordinary monotone Karchmer – Wigderson game for MAJ_{2n+1} . In this game Alice receives a string $x \in \text{MAJ}_{2n+1}^{-1}(0)$ and Bob receives a string $y \in \text{MAJ}_{2n+1}^{-1}(1)$. In other words, the number of ones in x is at most n and the number of ones in y is at least $n + 1$. The goal of Alice and Bob is to find some coordinate i such that $x_i = 0$ and $y_i = 1$. Next, imagine that Bob flips each of his input bits. After that parties have two vectors in both of which the number of ones is at most n . Now Alice and Bob have to find any coordinate in which both vectors are 0.

In this form this problem can be naturally generalized to the multiparty setting. Namely, assume that there are k parties, and each receives a Boolean vector of length $kn + 1$ with at most n ones. Let the task of parties be to find a coordinate in which all k input vectors are 0. How many bits of communication are needed for that?

For $k = 2$ the answer is $O(\log n)$, because there exists a $O(\log n)$ -depth monotone circuit for MAJ_{2n+1} and hence the monotone Karchmer – Wigderson game for MAJ_{2n+1} can be solved in $O(\log n)$ bits of communication. For $k \geq 3$ we are only aware of a simple $O(\log^2 n)$ -bit solution based on the binary search.

Now, let us look at the case $k \geq 3$ from another perspective and introduce multiparty Karchmer – Wigderson games. Note that each party receives a vector on which THR_{n+1}^{kn+1} equals 0. The goal is to find a common zero. Note that we can consider a similar problem for any function f satisfying so-called Q_k -property: any k vectors from $f^{-1}(0)$ have a common zero. In the next definition we define Q_k -property formally and also introduce related R_k -property.

► **Definition 4.** Let Q_k be the set of all Boolean functions f satisfying the following property: for all $x^1, x^2, \dots, x^k \in f^{-1}(0)$ there is a coordinate i such that $x_i^1 = x_i^2 = \dots = x_i^k = 0$.

Further, let R_k be the set of all Boolean functions f satisfying the following property: for all $x^1, x^2, \dots, x^k \in f^{-1}(0)$ there is a coordinate i such that $x_i^1 = x_i^2 = \dots = x_i^k$.

For $f \in Q_k$ let Q_k -communication game for f be the following communication problem. In this problem there are k parties. The j th party receives a Boolean vector $x^j \in f^{-1}(0)$. The goal of players is to find any coordinate i such that $x_i^1 = x_i^2 = \dots = x_i^k = 0$.

Similarly we can define R_k -communication games for functions from R_k . In the R_k -communication games the objective of parties is slightly different: their goal is to find any coordinate i and a bit b such that $x_i^1 = x_i^2 = \dots = x_i^k = b$.

Self-dual functions belong to R_2 and monotone self-dual functions belong to Q_2 . It is easy to see that R_2 -communication games are equivalent to Karchmer – Wigderson games for self-dual functions (one party should flip all the input bits). Moreover, Q_2 -communication games are equivalent to monotone Karchmer – Wigderson games for monotone self-dual functions.

In this paper we consider R_k -communication games as a multiparty generalization of Karchmer – Wigderson games. In turn, Q_k -communication games are considered as a generalization of *monotone* Karchmer – Wigderson games. To justify this choice one should relate them to some type of circuit complexity.

1.4 Connection to threshold gates and the main result

Every function from Q_k can be *lower bounded* by a circuit, consisting only of THR_2^{k+1} gates and variables. More precisely, let us write $C \leq f$ for a Boolean circuit C and a Boolean function f if for all $x \in f^{-1}(0)$ we have $C(x) = 0$. Then the following proposition holds:

► **Proposition 5 ([3]).** *The set Q_k is equal to the set of all Boolean functions f for which there exists a circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables.*

There is a similar characterization of the set R_k .

► **Proposition 6.** *The set R_k is equal to the set of all Boolean functions f for which there exists a circuit $C \leq f$, consisting only of THR_2^{k+1} gates and literals².*

The proof from [3] of Proposition 5 with obvious modifications also works for Proposition 6.

Given $f \in Q_k$, what is the minimal depth of a circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables? We show that this quantity is equal (up to constant factors) the communication complexity of Q_k -communication game for f .

► **Theorem 7.** *Let $k \geq 2$ be any constant. Then for any $f \in Q_k$ the following two quantities are equal up to constant factors:*

- *the communication complexity of Q_k -communication game for f ;*
- *minimal d for which there exists a d -depth circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables.*

Similar result can be obtained for R_k -communication games.

² We stress that negations can only be applied to variables but not to THR_2^{k+1} gates.

► **Theorem 8.** *Let $k \geq 2$ be any constant. Then for any $f \in R_k$ the following two quantities are equal up to constant factors:*

- *the communication complexity of R_k -communication game for f ;*
- *minimal d for which there exists a d -depth circuit $C \leq f$, consisting only of THR_2^{k+1} gates and literals.*

Proofs of both theorems are divided into two parts:

- (a) transformation of a d -depth circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables (literals), into a $O(d)$ -bit protocol computing $Q_k(R_k)$ -communication game for f ;
- (b) transformation of a d -bit protocol computing $Q_k(R_k)$ -communication game for f into a d -depth circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables (literals).

The first part is simple and the main challenge is the second part. Later in this paper (Section 6) we also formulate refined versions of Theorems 7 and 8. Namely, we refine these theorems in the following two directions. Firstly, we take into account circuit size and for this we consider dag-like communication protocols. Secondly, we show that transformations (a-b) can be done in polynomial time (under some mild assumptions).

We derive our upper bounds on the depth of MAJ_{2n+1} and THR_{n+1}^{kn+1} (Theorems 1 and 3) from Theorem 7. We first solve the corresponding Q_k -communication games with small number of bits of communication. Namely, for the case of MAJ_{2n+1} we use AKS sorting network to solve the corresponding Q_2 -communication game with $O(\log n)$ bits of communication. For the case of THR_{n+1}^{kn+1} with $k \geq 3$ we solve the corresponding Q_k -communication game by a simple binary search protocol with $O(\log^2 n)$ bits of communication. This is where we get depth $O(\log n)$ for Theorem 1 and depth $O(\log^2 n)$ for Theorem 3. Again, some special measures should be taken to make the resulting circuits polynomial-time computable and to control their size³.

1.5 Our techniques: $Q_k(R_k)$ -hypotheses games

As we already mentioned, the hard part of our main result is to transform a protocol into a circuit.

For this we develop a new language to describe circuits, consisting of threshold gates. Namely, for every f in $Q_k(R_k)$ we introduce the corresponding $Q_k(R_k)$ -hypotheses game for f . We show that strategies in these games exactly capture depth and size of circuits, consisting only of THR_2^{k+1} gates and variables (literals). It turns out that strategies are more convenient than circuits to simulate protocols, since they operate in the same top-bottom manner.

Once we establish the equivalence of circuits and hypotheses games, it remains for us to transform a communication protocol into a strategy in a hypotheses games. This is an elaborate construction that is presented in Propositions 20 and 24. Below in this section we introduce hypotheses games and as an illustration sketch the construction of a strategy in a hypothesis game that is used in the proof of Theorem 1.

Here is how we define these games. Fix $f: \{0, 1\}^n \rightarrow \{0, 1\}$. There are two players, Nature and Learner. Before the game starts, Nature privately chooses $z \in f^{-1}(0)$, which then can not be changed. The goal of Learner is to find some $i \in [n]$ such that $z_i = 0$. The game proceeds in rounds. At each round Learner specifies $k + 1$ families $\mathcal{H}_0, \mathcal{H}_1, \dots, \mathcal{H}_k \subset f^{-1}(0)$ to Nature. We understand this as if Learner makes the following $k + 1$ hypotheses about z :

³ We should only care about the size in case of Theorem 3, because depth $O(\log n)$ immediately gives polynomial size.

$“z \in \mathcal{H}_0”$,
 $“z \in \mathcal{H}_1”$,
 \vdots
 $“z \in \mathcal{H}_k”$.

Learner loses immediately if less than k hypotheses are true, i.e., if the number of $j \in \{0, 1, \dots, k\}$ satisfying $z \in \mathcal{H}_j$ is less than k . Otherwise Nature points out to some hypothesis which is true. In other words, Nature specifies to Learner some $j \in \{0, 1, \dots, k\}$ such that $z \in \mathcal{H}_j$. The game then proceeds in the same manner for some finite number of rounds. At the end Learner outputs an integer $i \in [n]$. We say that Learner wins if $z_i = 0$.

It is not hard to show that Learner has a winning strategy in Q_k -hypotheses game for f if and only if $f \in Q_k$. Since we will use similar arguments in the paper, let us go through the “if” part: if $f \in Q_k$, then Learner has a winning strategy. Denote by \mathcal{Z} be the set of all z 's which are compatible with Nature's answers so far. At the beginning $\mathcal{Z} = f^{-1}(0)$. If $|\mathcal{Z}| \geq k + 1$, Learner takes any distinct $z^1, z^2, \dots, z^{k+1} \in \mathcal{Z}$ and makes the following hypotheses:

$“z \neq z^1”$,
 $“z \neq z^2”$,
 \vdots
 $“z \neq z^{k+1}”$.

At least k hypotheses are true, and the Nature's response strictly reduces the size of \mathcal{Z} . When the size of \mathcal{Z} becomes k , Learner is ready to give an answer due to Q_k -property of f .

This strategy requires exponential in n number of rounds. This can be easily improved to $O(n)$ rounds. Indeed, instead of choosing $k + 1$ distinct elements of \mathcal{Z} split \mathcal{Z} into $k + 1$ disjoint almost equal parts. Then let the i th hypotheses be “ z is not in the i th part”. Nature's response to this reduces the size of \mathcal{Z} by a constant factor, until the size of \mathcal{Z} is k .

For $f \in Q_k$ we can now ask what is the minimal number of rounds on in a Learner's winning strategy. The following proposition gives an exact answer:

► **Proposition 9.** *For any $f \in Q_k$ the following holds. Learner has a d -round winning strategy in Q_k -hypotheses game for f if and only if there exists a d -depth circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables.*

Proposition 9 is the core result for our applications. For instance, we prove Theorem 1 by giving an explicit $O(\log n)$ -round winning strategy of Learner in Q_2 -hypotheses game for MAJ_{2n+1} . Let us now sketch our argument (the complete proof can be found in Section 4).

Assume that Nature's input vector is z . We notice that in $O(\log n)$ rounds one can easily find *two* integers $i, j \in [2n + 1]$ such that either $z_i = 0$ or $z_j = 0$. However, we need to know for sure. For that we take any polynomial time computable $O(\log n)$ -depth monotone formula F for MAJ_{2n+1} (for instance one that can be obtained from the AKS sorting network). We start to descend from the output gate of F to one of F 's inputs. Throughout this descending we maintain the following invariant. If g is the current gate, then either $g(z) = 0 \wedge z_i = 0$ or $g(\neg z) = 1 \wedge z_j = 0$ (here \neg denotes bit-wise negation). It can be shown that in one round one can either exclude i or j (which will already give us an answer) or replace g by some gate which is fed to g . If we reach an input to F , we output the index of the corresponding variable.

Similarly one can define R_k -hypotheses game for any $f: \{0, 1\}^n \rightarrow \{0, 1\}$. In R_k -hypotheses game Nature and Learner play in the same way except that now Learner's objective is to find some pair $(i, b) \in [n] \times \{0, 1\}$ such that $z_i = b$. The following analog of Proposition 9 holds:

► **Proposition 10.** *For any $f \in R_k$ the following holds. Learner has a d -round winning strategy in R_k -hypotheses game for f if and only if there exists a d -depth circuit $C \leq f$, consisting only of THR_2^{k+1} gates and literals.*

1.6 Organization of the paper

In Section 2 we give Preliminaries. In Section 3 we define $Q_k(R_k)$ -hypotheses games formally and derive Proposition 9 and 10. In Section 4 we obtain our results for Majority function (Theorems 1 and 2) using simpler arguments than in our general results. Then in Section 5 we prove these general results (Theorems 7 and 8). In Section 6 we refine Theorems 7 and 8 in order to take into account the circuit size and computational aspects (Theorems 27 and 30 below). In Section 7 we derive Theorem 3 and provide another proof for Theorem 1. Finally, in Section 8 we formulate some open problems.

2 Preliminaries

Let $[n]$ denote the set $\{1, 2, \dots, n\}$ for $n \in \mathbb{N}$. For a set W we denote the set of all subsets of W by 2^W . For two sets A and B by A^B we mean the set of all functions of the form $f: B \rightarrow A$.

We usually use subscripts to denote coordinates of vectors. In turn, we usually use superscripts to numerate vectors.

We use standard terminology for Boolean formulas and circuits [9]. We denote the size of a circuit C by $\text{size}(C)$ and the depth by $\text{depth}(C)$. By De Morgan formulas/circuits we mean formulas/circuits consisting of \wedge, \vee gates of fan-in 2 and literals (i.e., we assume that negations are applied only to variables). By monotone formulas/circuits we mean formulas/circuits consisting of \wedge, \vee gates of fan-in 2 and variables. We also consider formulas/circuits consisting only of THR_2^{k+1} gates and variables (literals). We stress that in such circuits we do not use constants. Allowing literals as inputs we allow to apply negations only to variables. We also assume that negations in literals do not contribute to the depth of a circuit.

We use the notion of deterministic communication protocols in the multiparty *number-in-hand* model. However, to capture the circuit size in our results we consider not only standard *tree-like* protocols, but also *dag-like* protocols. This notion was considered by Sokolov in [14]. We use slightly different variant of this notion, arguably more intuitive one. In the next subsection we provide all necessary definitions. To obtain a definition of a standard protocol one should replace dags by binary trees.

2.1 Dags and dag-like communication protocols

We use the following terminology for directed acyclic graphs (dags). Firstly, we allow more than one directed edge from one node to another. A terminal node of a dag G is a node with no out-going edges. Given a dag G , let

- $V(G)$ denote the set of nodes of G ;
- $T(G)$ denote the set of terminal nodes of G .

For $v \in V(G)$ let $\text{Out}_G(v)$ be the set of all edges of G that start at v . A dag G is called t -ary if every non-terminal node v of G we have $|\text{Out}_G(v)| = t$. An ordered t -ary dag is a t -ary dag G equipped with a mapping from the set of edges of G to $\{0, 1, \dots, t-1\}$. This

mapping restricted to $Out_G(v)$ should be injective for every $v \in V(G) \setminus T(G)$. The value of this mapping on an edge e will be called the *label* of e . In terms of labels we require for ordered t -ary dags that any t edges, starting at the same node, have different labels.

By a path in G we mean a sequence of edges $\langle e_1, e_2, \dots, e_m \rangle$ such that for every $j \in [m-1]$ edge e_j ends in the same node in which e_{j+1} starts. Note that there may be two distinct paths visiting same nodes (for instance, there may be two parallel edges from one node to another).

We say that a node w is a descendant of a node v if there is a path from v to w . We call w a successor of v if there is an edge from v to w . A node s is called *starting node* if any other node is a descendant of s . Note that any dag has at most one starting node.

If a dag G has the starting node s , then by depth of $v \in V(G)$ we mean the maximal length of a path from s to v . The depth of G then is the maximal depth of its nodes.

Assume that $\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_k, \mathcal{Y}$ are some finite sets.

- **Definition 11.** A k -party dag-like communication protocol π with inputs from $\mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_k$ and with outputs from \mathcal{Y} is a tuple $\langle G, P_1, P_2, \dots, P_k, \phi_1, \phi_2, \dots, \phi_k, l \rangle$, where
- G is an ordered 2-ary dag with the starting node s ;
 - P_1, P_2, \dots, P_k is a partition of $V(G) \setminus T(G)$ into k disjoint subsets;
 - ϕ_i is a function from $P_i \times \mathcal{X}_i$ to $\{0, 1\}$;
 - l is a function from $T(G)$ to \mathcal{Y} .

The depth of π (denoted by $\text{depth}(\pi)$) is the depth of G . The size of π (denoted by $\text{size}(\pi)$) is $|V(G)|$.

The underlying mechanics of the protocol is as follows. Parties descend from s to one of the terminals of G . If the current node v is not a terminal and $v \in P_i$, then at v the i th party communicates a bit to all the other parties. Namely, the i th party communicates the bit $b = \phi_i(v, x)$, where $x \in \mathcal{X}_i$ is the input of the i th party. Among the two edges, starting at v , parties choose one labeled by b and descend to one of the successors of v along this edge. Finally, when parties reach a terminal t , they output $l(t)$.

We say that $x \in \mathcal{X}_i$ is i -compatible with an edge e from v to w if one of the following two condition holds:

- $v \notin P_i$;
- $v \in P_i$ and e is labeled by $\phi_i(v, x)$.

We say that $x \in \mathcal{X}_i$ is i -compatible with a path $p = (e_1, e_2, \dots, e_m)$ of G if for every $j \in [m]$ it holds that x is i -compatible with e_j . Finally, we say that $x \in \mathcal{X}_i$ is i -compatible with a node $v \in V(G)$ if there is a path p from s to v such that x is i -compatible with v .

We say that an input $(x^1, x^2, \dots, x^k) \in \mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_k$ visits a node $v \in V(G)$ if there is a path p from s to v such that for every $i \in [k]$ it holds that x^i is i -compatible with p . Note that there is unique $t \in T(G)$ such that (x^1, x^2, \dots, x^k) visits t .

To formulate an effective version of Theorems 7 and Theorem 8 we need the following definition.

- **Definition 12.** The light form of a k -party dag-like communication protocol $\pi = \langle G, P_1, P_2, \dots, P_k, \phi_1, \phi_2, \dots, \phi_k, l \rangle$ is a tuple $\langle G, P_1, P_2, \dots, P_k, l \rangle$.

I.e., to obtain the light form of π we just forget about $\phi_1, \phi_2, \dots, \phi_k$. In other words, the light form only contains the underlying graph of π , the partition of non-terminal nodes between parties and the labels of terminals. On the other hand, in the light form there is no information at all how parties communicate at the non-terminal nodes.

Protocol π computes a relation $S \subset \mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_k \times \mathcal{Y}$ if the following holds. For every $(x^1, x^2, \dots, x^k) \in \mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_k$ there exist $y \in \mathcal{Y}$ and $t \in T(G)$ such that (x^1, \dots, x^k) visits t , $l(t) = y$ and $(x^1, x^2, \dots, x^k, y) \in S$.

Using language of relations, we can formally define Q_k - and R_k -communication games. Namely, given $f: \{0, 1\}^n \rightarrow \{0, 1\}$, $f \in Q_k$, we define Q_k -communication game for f as the following relation:

$$S \subset \underbrace{f^{-1}(0) \times \dots \times f^{-1}(0)}_k \times [n],$$

$$S = \{(x^1, \dots, x^k, j) \mid x_j^1 = \dots = x_j^k = 0\}.$$

Similarly, given $f: \{0, 1\}^n \rightarrow \{0, 1\}$, $f \in R_k$, we define R_k -communication game for f as the following relation:

$$S \subset \underbrace{f^{-1}(0) \times \dots \times f^{-1}(0)}_k \times ([n] \times \{0, 1\}),$$

$$S = \{(x^1, \dots, x^k, (j, b)) \mid x_j^1 = \dots = x_j^k = b\}.$$

It is easy to see that a dag-like protocol for S can be transformed into a tree-like protocol of the same depth, but this transformation can drastically increase the size.

3 Formal treatment of $Q_k(R_k)$ -hypotheses games

Fix $f \in Q_k$, $f: \{0, 1\}^n \rightarrow \{0, 1\}$. Here we define Learner's strategies in Q_k -hypotheses game for f formally. We consider not only tree-like strategies but also dag-like. To specify a Learner's strategy S in Q_k -hypotheses game we have to specify:

- An ordered $(k + 1)$ -ary dag G with the starting node s ;
- a subset $\mathcal{H}_j(p)$ for every $j \in \{0, 1, \dots, k\}$ and for every path p in G from s to some node in $V(G) \setminus T(G)$;
- a number $i_t \in [n]$ for every terminal t .

The underlying mechanics of the game is as follows. Let Nature's vector be $z \in f^{-1}(0)$. Learner and Nature descend from s to one of the terminals of G . More precisely, a position in the game is determined by a path p , starting at s . If the endpoint of p is not a terminal, then Learner specifies some sets $\mathcal{H}_0(p), \mathcal{H}_1(p), \dots, \mathcal{H}_k(p)$ as his hypotheses. If less than k of these sets contain z , then Nature wins. Otherwise Nature specifies some $j \in \{0, 1, \dots, k\}$ such that $z \in \mathcal{H}_j(p)$. Among $k + 1$ edges that start at the endpoint of p players choose one which is labeled by j . After that they extend p by this edge. At some point parties reach some terminal t (i.e., the endpoint of p becomes equal t). Then the game ends and Learner output i_t .

We stress that Learner's output depends only on t but not on a path to t (unlike Learner's hypotheses). This property will be crucial in establishing connection of Q_k -hypotheses games to circuits.

We now proceed to a formal definition of what does it mean that S is winning for Learner.

We say that $z \in f^{-1}(0)$ is *compatible* with a path $p = \langle e_1, \dots, e_m \rangle$, starting in s , if the following holds. If p is of length 0, then every $z \in f^{-1}(0)$ is compatible with p . Otherwise for every $i \in \{1, \dots, m\}$ it should hold that $z \in \mathcal{H}_j(\langle e_1, \dots, e_{i-1} \rangle)$, where j is the label of edge e_i . Informally this means that Nature, having z on input, can reach a position in the game which corresponds to a path p .

We say that strategy S is winning for Learner in Q_k -hypotheses game for f if for every path p , starting at s , and for every $z \in f^{-1}(0)$, compatible with p , the following holds:

- if the endpoint of p is not a terminal, then the number of $j \in \{0, 1, \dots, k\}$ such that $z \in \mathcal{H}_j(p)$ is at least k ;
- if the endpoint of p is $t \in T(G)$, then $z_{i_t} = 0$.

We will formulate a stronger version of Proposition 9. For that we need the notion of the *light form* of the strategy S . Namely, the light form of S is its underlying dag G equipped with a mapping which to every $t \in T(G)$ assigns i_t . In other words, the light form contains a “skeleton” of S and Learner’s outputs in terminals (and no information about Learner’s hypotheses).

We can identify the light form of any strategy S with a circuit, consisting only of THR_2^{k+1} gates and variables. Namely, place THR_2^{k+1} gate in every $v \in V(G) \setminus T(G)$ and for every $t \in T(G)$ place a variable x_{i_t} in t . Set s to be the output gate.

► **Proposition 13.** *For all $f \in Q_k, f: \{0, 1\}^n \rightarrow \{0, 1\}$ the following holds:*

- (a) *if S is a Learner’s winning strategy in Q_k -hypotheses game for f , then its light form, considered as a circuit C consisting only of THR_2^{k+1} gates and variables, satisfies $C \leq f$.*
- (b) *Assume that $C \leq f$ is a circuit, consisting only of THR_2^{k+1} gates and variables. Then there exists a Learner’s winning strategy S in Q_k -hypotheses game for f such that the light form of S coincides with C .*

We omit the proof of (b) as in the paper we only use (a).

Proof of (a) of Proposition 13. For a node $v \in V(G)$ let $f_v: \{0, 1\}^n \rightarrow \{0, 1\}$ be the function, computed by the circuit C at the gate, corresponding to v .

We shall prove the following statement. For any path p , starting in s , and for any z which is compatible with p it holds that $f_v(z) = 0$, where v is the endpoint of p . To see why this implies $C \leq f$ take any $z \in f^{-1}(0)$ and note that z is compatible with the path of length 0. The endpoint of such path is s and hence $0 = f_s(z) = C(z)$.

We will prove the above statement by the backward induction on the length of p . The longest path p ends in some $t \in T(G)$. By definition $f_t = x_{i_t}$. On the other hand, since S is winning, $z_{i_t} = 0$ for any z compatible with p . In other words, $f_t(z) = 0$ for any z compatible with p . The base is proved.

Induction step is the same if p ends in some other terminal. Now assume that p ends in $v \in V(G) \setminus T(G)$. Take any $z \in f^{-1}(0)$ compatible with p . Let p_j be the extension of p by the edge which starts at v and is labeled by $j \in \{0, 1, \dots, k\}$. Next, let v_j be the endpoint of p_j (nodes v_0, v_1, \dots, v_k are successors of v). Since S is winning, the number of $j \in \{0, 1, \dots, k\}$ such that $z \in \mathcal{H}_j(p)$ is at least k . Hence by definition the number of $j \in \{0, 1, \dots, k\}$ such that z is compatible with p_j is at least k . Finally, by the induction hypothesis this means that the number of $j \in \{0, 1, \dots, k\}$ such that $f_{v_j}(z) = 0$ is at least k . On the other hand:

$$f_v = \text{THR}_2^{k+1}(f_{v_0}, f_{v_1}, \dots, f_{v_k}).$$

Therefore $f_v(z) = 0$, as required. ◀

One can formally define analogues notions for R_k -hypotheses games. We skip this as modifications are straightforward and only formulate an analog of Proposition 13.

► **Proposition 14.** *For all $f \in R_k, f: \{0, 1\}^n \rightarrow \{0, 1\}$ the following holds:*

- (a) *if S is a Learner’s winning strategy in R_k -hypotheses game for f , then its light form, considered as a circuit C consisting only of THR_2^{k+1} gates and literals, satisfies $C \leq f$.*
- (b) *Assume that $C \leq f$ is a circuit, consisting only of THR_2^{k+1} gates and literals. Then there exists a Learner’s winning strategy S in R_k -hypotheses game for f such that the light form of S coincides with C .*

► **Remark 15.** It might be unclear why we prefer to construct strategies instead of constructing circuits directly, because beside the circuit itself we should also specify Learner’s hypotheses. The reason is that strategies can be seen as *proofs* that the circuit we construct is correct.

4 Results for Majority

Proof of Theorem 1. There exists an algorithm which in $n^{O(1)}$ -time produces a monotone formula F of depth $d = O(\log n)$ computing MAJ_{2n+1} . Below we will define a strategy S_F in the Q_2 -hypotheses game for MAJ_{2n+1} . Strategy S_F will be winning for Learner. Moreover, its depth will be $d + O(\log n)$. In the end of the proof we will refer to Proposition 13 to show that S_F yields a $O(\log n)$ -depth polynomial-time computable formula for MAJ_{2n+1} , consisting only of MAJ_3 gates and variables.

Strategy S_F has two phases. The first phase does not use F at all, only the second phase does. The objective of the first phase is to find some distinct $i, j \in [2n + 1]$ such that either $z_i = 0 \wedge z_j = 1$ or $z_i = 1 \wedge z_j = 0$, where z is the Nature's vector. This can be done as follows.

► **Lemma 16.** *One can compute in polynomial time a 3-ary tree T of depth $O(\log n)$ with the set of nodes $v(T)$ and a mapping $w: v(T) \rightarrow 2^{[2n+1]}$ such that the following holds:*

- if r is the root of T , then $w(r) = [2n + 1]$;
- if v is not a leaf of T and v_1, v_2, v_3 are 3 children of v , then every element of $w(v)$ is covered at least twice by $w(v_1), w(v_2), w(v_3)$;
- if l is a leaf of T , then $w(l)$ is of size 2.

Proof. We start with a trivial tree, consisting only of the root, to which we assign $[2n + 1]$. Then at each iteration we do the following. We have a 3-ary tree in which nodes are assigned to some subsets of $[2n + 1]$. If every leaf is assigned to a set of size 2, we terminate. Otherwise we pick any leaf l of the current tree which is assigned to a subset $A \subset [2n + 1]$ of size at least 3. We split A into 3 disjoint subsets A_1, A_2, A_3 of sizes $\lfloor |A|/3 \rfloor, \lfloor |A|/3 \rfloor$ and $|A| - 2\lfloor |A|/3 \rfloor$. We add 3 children to l (which become new leaves) and assign $A_1 \cup A_2, A_1 \cup A_3, A_2 \cup A_3$ to them.

It is easy to verify that the sizes of $A_1 \cup A_2, A_1 \cup A_3, A_2 \cup A_3$ are at least 2 and at most $\frac{4}{3} \cdot |A|$. Hence the size of the set assigned to a node of depth h is at most $(\frac{4}{3})^h \cdot (2n + 1)$. This means that the depth of the tree is at any moment at most $\log_{5/4}(2n + 1) = O(\log n)$. Therefore we terminate in $3^{O(\log n)} = n^{O(1)}$ iterations, as at each iteration we add 3 new nodes. Each iteration obviously takes polynomial time. ◀

We use T to find two $i, j \in [2n + 1]$ such that either $z_i = 0$ or $z_j = 0$. Namely, we descend from the root of T to one of its leaves. Learner maintains an invariant that the leftmost 0-coordinate of z is in $w(v)$, where v is the current node of T . Let v_1, v_2, v_3 be 3 children of v . Learner for every $i \in [3]$ makes a hypothesis that the leftmost 0-coordinate of z is in $w(v_i)$. Due to the properties of w at least two hypotheses are true. Nature indicates some v_i for which this is true, and Learner descends to v_i . When Learner reaches a leaf, he knows a set of size two containing the leftmost 0-coordinate of z . Let this set be $\{i, j\}$.

We know that either z_i or z_j is 0. Thus $z_i z_j \in \{00, 01, 10\}$. At the cost of one round we can ask Nature to identify an element of $\{00, 01, 10\}$ which differs from $z_i z_j$. If 10 is identified, then $z_i z_j \in \{00, 01\}$, and hence $z_i = 0$, i.e., we can already output i . Similar thing happens when 01 is identified. Finally, if 00 is identified, then the objective of the first phase is fulfilled and we can proceed to the second phase.

The second phase takes at most d rounds. In this phase Learner produces a sequence $g_0, g_1, \dots, g_{d'}$, $d' \leq d$ of gates of F , where the depth of g_i is i , the last gate $g_{d'}$ is an input variable (i.e., a leaf of F) and each $g \in \{g_0, g_1, \dots, g_{d'}\}$ satisfies:

$$(g(z) = 0 \wedge z_i z_j = 01) \vee (g(\neg z) = 1 \wedge z_i z_j = 10). \quad (1)$$

Here $\neg z$ denotes the bit-wise negation of z .

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At the beginning Learner sets $g_0 = g_{\text{out}}$ to be the output gate of F . Let us explain why (1) holds for g_{out} . Nature's vector is an element of $\text{MAJ}_{2n+1}^{-1}(0)$. I.e., the number of ones in z is at most n . In turn, in $\neg z$ there are at least $n + 1$ ones. Since g_{out} computes MAJ_{2n+1} , we have that $g_{\text{out}}(z) = 0$ and $g_{\text{out}}(\neg z) = 1$. In turn, by the first phase it is guaranteed that $z_i z_j = 01 \vee z_i z_j = 10$.

Assume now that the second phase is finished, i.e., Learner has produced some $g_{d'} = x_k$ satisfying (1). Then by (1) either $g_{d'}(z) = z_k = 0$ or $g_{d'}(\neg z) = (\neg z)_k = 1$. In both cases $z_k = 0$, i.e., Learner can output k .

It remains to explain how to fulfill the second phase. It is enough to show the following. Assume that Learner knows a gate g_l of F of depth l satisfying (1) and that g_l is not an input variable. Then in one round he can either find a gate g_{l+1} of depth $l + 1$ satisfying (1) or give a correct answer to the game.

The gate g_{l+1} will be one of the two gates which are fed to g_l . Assume first that g_l is an \wedge -gate and $g_l = u \wedge v$. From (1) we conclude that from the following 3 statements exactly 1 is true for z :

$$u(z) = 0 \text{ and } z_i z_j = 01, \quad (2)$$

$$u(z) = 1, v(z) = 0 \text{ and } z_i z_j = 01, \quad (3)$$

$$u(\neg z) = v(\neg z) = 1 \text{ and } z_i z_j = 10. \quad (4)$$

At the cost of one round Learner can ask Nature to indicate one statement which is false for x . If Nature says that (2) is false for z , then (1) holds for $g_{l+1} = v$. Next, if Nature says that (3) is false for z , then (1) holds for $g_{l+1} = u$. Finally, if Nature says that (4) is false for z , then we know that $z_i z_j = 01$, i.e., Learner can already output i .

In the same way we can deal with the case when g_l is an \vee -gate and $g_l = u \vee v$. By (1) exactly 1 of the following 3 statements is true for z :

$$u(z) = v(z) = 0 \text{ and } z_i z_j = 01, \quad (5)$$

$$u(\neg z) = 1 \text{ and } z_i z_j = 10, \quad (6)$$

$$u(\neg z) = 0, v(\neg z) = 1 \text{ and } z_i z_j = 10. \quad (7)$$

Similarly, Learner asks Nature to indicate one statement which is false for z . If Nature says that (5) is false for z , then $z_i z_j = 10$, i.e., Learner can output j . Next, if Nature says that (6) is false for z , then (1) holds for $g_{l+1} = v$. Finally, if Nature says that (7) is false for z , then (1) holds for $g_{l+1} = u$.

Thus S_F is a $O(\log n)$ -depth winning strategy of Learner. Apply Proposition 13 to S_F . We get a $O(\log n)$ -depth formula $F' \leq \text{MAJ}_{2n+1}$, consisting only of MAJ_3 gates and variables. In fact, F' computes MAJ_{2n+1} . Indeed, $F' \leq \text{MAJ}_{2n+1}$ means that F' outputs 0 on every input with at most n ones. On the other hand, F' consists of MAJ_3 gates and hence F' computes a self-dual function. Therefore F' outputs 1 on every input with at least $n + 1$ ones.

It remains to explain how to compute F' in polynomial time. To do so we have to compute in polynomial time the light form of S_F , i.e., the underlying tree of S_F and the outputs of Learner in the leafs. It is easy to see that one can do this as follows.

First, compute F and compute T from Lemma 16. For each leaf l of T do the following. Let $w(l) = \{i, j\}$. Add 3 children to l . Two of them will be leafs of S_F , in one Learner outputs i and in the other Learner outputs j . Attach a tree of F to the third child. Then add to each non-leaf node of F one more child so that now the tree of F is 3-ary. Each added child is a leaf of S_F . If a child was added to an \wedge -gate, then Learner outputs i in this child.

In turn, if a child was added to an \vee gate, then Learner outputs j in it. Finally, there are leaves that were in F initially, each labeled by some input variable. In these nodes Learner outputs the index of the corresponding input variable. \blacktriangleleft

Proof of Theorem 2. How many rounds takes the first phase of the strategy S_F from the previous proof? Initially the left-most 0-coordinate of z takes $O(n)$ values. At the cost of one round we can shrink the number of possible values almost by a factor of $3/2$. Thus the first phase corresponds to a ternary tree of depth $\log_{3/2}(n) + O(1)$. The size of that tree is hence $3^{\log_{3/2}(n)+O(1)} = O(n^{1/(1-\log_3(2))}) = O(n^{2.70951\dots})$. To some of its leaves we attach a tree of the same size as the initial formula F . As a result we obtain a formula F' of size $O(n^{2.70951\dots} \cdot s)$ for MAJ_{2n+1} , consisting of MAJ_3 gates and variables (here s is the size of the initial formula F).

Let us show that we can perform the first phase in $\log_2(n) + O(1)$ rounds. This will improve the size of the previous construction to $O(3^{\log_2(n)+O(1)} \cdot s) = O(n^{\log_2(3)} \cdot s)$. However, the construction with $\log_2(n) + O(1)$ rounds will not be explicit. We need the following Lemma:

► **Lemma 17.** *There exists a formula D with the following properties:*

- *formula D is a complete ternary tree of depth $\lceil \log_2(n) \rceil + 10$;*
- *every non-leaf node of D contains a MAJ_3 gate and every leaf of D contains a conjunction of 2 variables;*
- *$D(x) = 0$ for every $x \in \{0, 1\}^{2n+1}$ with at most n ones.*

Let us at first explain how to use formula D from Lemma 17 to fulfill the first phase. Recall that our goal is to find two indices $i, j \in [2n + 1]$ such that either $z_i = 0$ or $z_j = 0$. To do so Learner descends from the output gate of D to some of its leaves. He maintains an invariant that for his current gate g of D it holds that $g(z) = 0$. For the output gate the invariant is true because by Lemma 17 D is 0 on all Nature's possible vectors. If we reached a leaf so that g is a conjunction of two variables z_i and z_j , then the first phase is fulfilled (by the invariant $z_i \wedge z_j = 0$). Finally, if g is a non-leaf node of D , i.e., a MAJ_3 gate, then we can descend to one of the children of g at the cost of one round without violating the invariant. Indeed, as $g(z) = 0$, then the same is true for at least 2 children of g . For each child g_i of g Learner makes a hypotheses that $g_i(z) = 0$. Any Nature's response allows us to replace g by some g_i .

Proof of Lemma 17. We will show existence of such D via probabilistic method. Namely, independently for each leaf l of D choose $(i, j) \in [2n + 1]^2$ uniformly at random and put the conjunction $z_i \wedge z_j$ into l . It is enough to demonstrate that for any $x \in \{0, 1\}^{2n+1}$ with at most n ones it holds that $\Pr[D(x) = 1] < 2^{-2n-1}$.

To do so we use the modification of the standard Valiant's argument. For any fixed x let p be the probability that a leaf l of D equals 1 on x . This probability is the same for all the leaves and is at most $1/4$. Now, $\Pr[D(x) = 1]$ can be expressed exactly in terms of p as follows:

$$\Pr[D(x) = 1] = \underbrace{f(f(f(\dots f(p))))}_{\lceil \log_2(n) \rceil + 10},$$

where $f(t) = t^3 + 3t^2(1 - t) = 3t^2 - 2t^3$. Observe that $3f(t) \leq (3t)^2$. Hence

$$3\Pr[D(x) = 1] \leq (3p)^{2^{\lceil \log_2(n) \rceil + 10}} \leq (3/4)^{1000n} < (1/2)^{-2n-1}. \quad \blacktriangleleft$$

5 Proof of the main theorem

Theorem 7 follows from Proposition 18 (Subsection 5.1) and Proposition 20 (Subsection 5.2). In turn, Theorem 8 follows from Proposition 19 (Subsection 5.1) and Proposition 24 (Subsection 5.2).

5.1 From circuits to protocols

► **Proposition 18.** *For any constant $k \geq 2$ the following holds. Assume that $f \in Q_k$ and $C \leq f$ is a circuit, consisting only of THR_2^{k+1} gates and variables. Then there is a protocol π , computing Q_k -communication game for f , such that $\text{depth}(\pi) = O(\text{depth}(C))$.*

Proof. Let the inputs to parties be $z^1, \dots, z^k \in f^{-1}(0)$. Parties descend from the output gate of C to one of the inputs. They maintain the invariant that for the current gate g of C it holds that $g(z^1) = g(z^2) = \dots = g(z^k) = 0$. If g is not yet an input, then g is a THR_2^{k+1} gate and $g = \text{THR}_2^{k+1}(g_1, \dots, g_{k+1})$ for some gates g_1, \dots, g_{k+1} . For each z^i we have $g(z^i) = \text{THR}_2^{k+1}(g_1(z^i), \dots, g_{k+1}(z^i)) = 0$. Hence for each z^i there is at most one gate out of g_1, \dots, g_{k+1} satisfying $g_j(z^i) = 1$. This means that in $O(1)$ bits of communication parties can agree on the index $j \in [k+1]$ satisfying $g_j(z^1) = g_j(z^2) = \dots = g_j(z^k) = 0$.

Thus in $O(\text{depth}(\pi))$ bits of communication they reach some input of C . If this input contains the variable x_l , then by the invariant $z_l^1 = z_l^2 = \dots = z_l^k = 0$, as required. ◀

Exactly the same argument can be applied to the following proposition.

► **Proposition 19.** *For any constant $k \geq 2$ the following holds. Assume that $f \in R_k$ and $C \leq f$ is a circuit, consisting only of THR_2^{k+1} gates and literals. Then there is a protocol π , computing R_k -communication game for f , such that $\text{depth}(\pi) = O(\text{depth}(C))$.*

5.2 From protocols to circuits

► **Proposition 20.** *For every constant $k \geq 2$ the following holds. Let $f \in Q_k$. Assume that π is a communication protocol computing Q_k -communication game for f . Then there is a circuit $C \leq f$, consisting of THR_2^{k+1} gates and variables, such that $\text{depth}(C) = O(\text{depth}(\pi))$.*

Proof. In the proof we will use the following terminology for strategies in Q_k -hypotheses game. Fix some strategy S . A *current play* is a finite sequence $r_1, r_2, r_3, \dots, r_j$ of integers from 0 to k . By r_i we mean Nature's response in the i th round. Given a current play, let $\mathcal{H}_0^i, \dots, \mathcal{H}_k^i \subset f^{-1}(0)$ be $k+1$ hypotheses Learner makes in the i th round according to S if Nature's responses in the first $i-1$ rounds were r_1, \dots, r_{i-1} . If after that Nature's response is r_i , then Nature's input vector z satisfies $z \in H_{r_i}^i$. We say that $z \in f^{-1}(0)$ is *compatible* with the current play r_1, \dots, r_j if $z \in H_{r_1}^1, \dots, z \in H_{r_j}^j$. Informally, this means that Nature, having z on input, can produce responses r_1, \dots, r_j by playing against strategy S .

Set $d = \text{depth}(\pi)$. By Proposition 13 it is enough to give a $O(d)$ -round winning strategy of Learner in the Q_k -hypotheses game for f . Strategy proceeds in d iterations, each iteration takes $O(1)$ rounds.

As the game goes on, a sequence of Nature's responses r_1, r_2, r_3, \dots is produced. Assume that $r_1, \dots, r_{h'}$ are Nature's responses in the first h iteration (here h' is the number of rounds in the first h iterations). Given any r_1, r_2, r_3, \dots , by \mathcal{Z}_h we denote the set of all $z \in f^{-1}(0)$ which are compatible with $r_1, \dots, r_{h'}$. We also say that elements of \mathcal{Z}_h are *compatible with the current play after h iterations*.

Let V be the set of all nodes of the protocol π and let T be the set of all terminals of the protocol π .

Consider a set $\mathcal{Z} \subset f^{-1}(0)$, a set of nodes $U \subset V$ and a function $g: \mathcal{Z} \rightarrow C$, where $|C| = k$. A g -profile of a tuple $(z^1, \dots, z^k) \in \mathcal{Z}$ is a vector $(g(z^1), \dots, g(z^k)) \in C^k$.

We say that $g: \mathcal{Z} \rightarrow C$ is *complete* for \mathcal{Z} with respect to the set of nodes U if the following holds. For every vector $\bar{c} \in C^k$ there exists a node $v \in U$ such that all tuples from \mathcal{Z}^k with g -profile \bar{c} visit v in the protocol π .

We say that a set of nodes $U \subset T$ is complete for \mathcal{Z} if there exists $g: \mathcal{Z} \rightarrow C$, $|C| = k$ which is complete for \mathcal{Z} with respect to U .

Note that we can consider only complete sets of size at most k^k . Formally, if U is complete for \mathcal{Z} , then there is a subset $U' \subset U$ of size at most k^k which is also complete for \mathcal{Z} . Indeed, there are k^k possible g -profiles and for each we need only one node in U .

► **Lemma 21.** *Assume that $U \subset T$ is complete for $\mathcal{Z} \subset f^{-1}(0)$. Then there exists $i \in [n]$ such that $z_i = 0$ for every $z \in \mathcal{Z}$.*

Proof. If \mathcal{Z} is empty, then there is nothing to prove. Otherwise let $g: \mathcal{Z} \rightarrow C$, $|C| = k$ be complete for \mathcal{Z} with respect to U . Take any vector $\bar{c} = (c_1, \dots, c_k) \in C^k$ such that $\{c_i \mid i \in [k]\} = g(\mathcal{Z})$. There exists a node $v \in U$ such that any tuple from \mathcal{Z}^k with g -profile \bar{c} visits v . Note that v is a terminal of π and let i be the output of π in v . Let us show that for any $z \in \mathcal{Z}$ it holds that $z_i = 0$. Indeed, note that there exists a tuple $\bar{z} \in \mathcal{Z}^k$ which includes z and which has g -profile \bar{c} . This tuple visits v . Since π computes Q_k -communication game for f , every element of the tuple \bar{z} should have 0 at the i th coordinate. In particular, this holds for z . ◀

After d iterations Learner should be able to produce an output. For that there should exist $i \in [n]$ such that for any $z \in \mathcal{Z}_d$ it holds that $z_i = 0$. We will use Lemma 21 to ensure that. Namely, we will ensure that there exists $U \subset T$ which is complete for \mathcal{Z}_d . Learner achieves this by maintaining the following invariant.

Let us say that a set of nodes U is h -low if every element of U is either a terminal or a node of depth at least h .

► **Invariant 22.** *There is a h -low set U which is complete for \mathcal{Z}_h .*

This invariant implies that Learner wins in the end, as any d -low set consists only of terminals.

A 0-low set which is complete for $\mathcal{Z}_0 = f^{-1}(0)$ is a set consisting only of the starting node of π .

Assume that Invariant 22 holds after h iterations. Let us show how to perform the next iteration to maintain the invariant. For that we need a notion of *communication profile*.

A communication profile of $z \in f^{-1}(0)$ with respect to a set of nodes $U \subset V$ is a function $p_z: U \rightarrow \{0, 1\}$. For $v \in U$ the value of $p_z(v)$ is defined as follows. If v is a terminal, set $p_z(v) = 0$. Otherwise let $i \in [k]$ be the index of the party communicating at v . Set $p_z(v)$ to be the bit transmitted by the i th party at v on input z . I.e., p_z for every $v \in U$ contains information where the protocol goes from the node v if the party, communicating at v , has z on input.

We also define a communication profile of the tuple $(z^1, \dots, z^k) \in (f^{-1}(0))^k$ as $(p_{z^1}, \dots, p_{z^k})$.

► **Lemma 23.** *Let $(z^1, \dots, z^k), (y^1, \dots, y^k) \in (f^{-1}(0))^k$ be two inputs visiting the same node $v \in V \setminus T$. Assume that their communication profiles with respect to $\{v\}$ coincide. Then these two inputs visit the same successor of v .*

Proof. Let their common communication profile with respect to $\{v\}$ be (p_1, \dots, p_k) . Next, assume that i is the index of the party communicating at v . Then the information where these inputs descend from v is contained in p_i . ◀

Here is what Learner does during the $(h + 1)$ st iteration. He takes any h -low U of size at most k^k which is complete for \mathcal{Z}_h . Then he takes any $g: \mathcal{Z}_h \rightarrow C$, $|C| = k$ which is complete for \mathcal{Z}_h with respect to U . He now devises a new function g' taking elements of the set \mathcal{Z}_h on input. The value of $g'(z)$ is a pair $(p_z, g(z))$, where p_z is a communication profile of z with respect to U . There are at most $2^{|U|} \leq 2^{k^k}$ different communication profiles with respect to U . Hence $g'(z)$ takes at most $2^{k^k} \cdot k = O(1)$ values.

At each round of the $(h + 1)$ st iteration Learner asks Nature to identify some pair (p, c) , where $p: U \rightarrow \{0, 1\}$ and $c \in C$, such that $g'(z) \neq (p, c)$ for the Nature's vector z . Namely, we take any $k + 1$ values of g' which are not yet rejected by Nature and ask Nature to reject one of them. We do so until there are only k possible values $(p_1, c_1), \dots, (p_k, c_k)$ left. This takes $O(1)$ rounds and the $(h + 1)$ st iteration is finished. Any $z \in f^{-1}(0)$ which is compatible with the responses Nature' gave during the $(h + 1)$ st iteration in the current play satisfies $g'(z) \in C' = \{(p_1, c_1), \dots, (p_k, c_k)\}$. In particular, any $z \in \mathcal{Z}_{h+1}$ satisfies $g'(z) \in C'$. I.e., the restriction of g' to \mathcal{Z}_{h+1} is a function of the form $g': \mathcal{Z}_{h+1} \rightarrow C'$. Let us show that $g': \mathcal{Z}_{h+1} \rightarrow C'$ is complete for \mathcal{Z}_{h+1} with respect to some $(h + 1)$ -low set U' . This will ensure that Invariant 22 is maintained after $h + 1$ iterations.

We define U' as follows. Take any vector $\bar{c} \in (C')^k$. It is enough to show that all the inputs from $(\mathcal{Z}_{h+1})^k$ with g' -profile \bar{c} visit the same node v' which is either a terminal or of depth at least $h + 1$. Then we just set U' to be the union of all such v' over all possible g' -profiles.

All the tuples from $(\mathcal{Z}_{h+1})^k$ with the same g' -profile visit the same node $v \in U$. This is because g' -profile of a tuple determines its g -profile (the value of g' determines the value of g), and hence we can use Invariant 22 for \mathcal{Z}_{h-1} here. If v is a terminal, there is nothing left to prove. Otherwise, note that g' -profile of a tuple also determines its communication profile with respect to U and hence with respect to $\{v\} \subset U$. Therefore all the tuples with the same g' -profile by Lemma 23 visit the same successor of v . ◀

With straightforward modifications one can obtain a proof of the following:

► **Proposition 24.** *For every constant $k \geq 2$ the following holds. Let $f \in R_k$. Assume that π is a dag-like protocol computing R_k -communication game for f . Then there is a circuit $C \leq f$, consisting of THR_2^{k+1} gates and literals, satisfying $\text{depth}(C) = O(\text{depth}(\pi))$.*

► **Corollary 25** (Weak version of Theorem 3). *For any constant $k \geq 2$ there exists $O(\log^2 n)$ -depth formula for THR_{n+1}^{kn+1} , consisting only of THR_2^{k+1} gates and variables.*

Proof. We will show that there exists $O(\log^2 n)$ -depth protocol π computing Q_k -communication game for THR_{n+1}^{kn+1} . By Proposition 20 this means that there is a $O(\log^2 n)$ -depth formula $F \leq \text{THR}_{n+1}^{kn+1}$, consisting only of THR_2^{k+1} gates and variables. It is easy to see that F actually coincides with THR_{n+1}^{kn+1} . Indeed, assume that $F(x) = 0$ for some x with at least $n + 1$ ones. Then it is easy to construct x^2, \dots, x^k , each with n ones, such that there is no common 0-coordinate for x, x^2, \dots, x^k . On all of these vectors F takes value 0. However, the function computed by F should belong to Q_k (Proposition 5).

Let π be the following protocol. Assume that the inputs to parties are $x^1, x^2, \dots, x^k \in \{0, 1\}^{kn+1}$, without loss of generality we can assume that in each x^r there are exactly n ones. For $x \in \{0, 1\}^{kn+1}$ define $\text{supp}(x) = \{i \in [kn + 1] \mid x_i = 1\}$. Let T be a binary rooted tree of

depth $d = \log_2(n) + O(1)$ with $kn + 1$ leaves. Identify leaves of T with elements of $[kn + 1]$. For a node v of T let T_v be the set of all leaves of T which are descendants of v . Once again, we view T_v as a subset of $[kn + 1]$.

The protocol proceeds in at most d iterations. After i iterations, $i = 0, 1, 2, \dots, d$, parties agree on a node v of T of depth i , satisfying the following invariant:

$$\sum_{r=1}^k |\text{supp}(x^r) \cap T_v| < |T_v|. \tag{8}$$

At the beginning Invariant (8) holds just because v is the root, $T_v = [kn + 1]$ and each $\text{supp}(x^r)$ is of size n .

After d iterations $v = l$ is a leaf of T . Parties output l . This is correct because by (8) we have $|T_l| = 1 \implies |\text{supp}(x^r) \cap T_l| = 0 \implies x_l = 0$ for every $r \in [k]$.

Let us now explain what parties do at each iteration. If the current v is not a leaf, let v_0, v_1 be two children of v . Each party sends $|\text{supp}(x^r) \cap T_{v_0}|$ and $|\text{supp}(x^r) \cap T_{v_1}|$, using $O(\log n)$ bits. Since T_{v_0} and T_{v_1} is a partition of T_v , we have:

$$\sum_{b=0}^1 \sum_{r=1}^k |\text{supp}(x^r) \cap T_{v_b}| = \sum_{r=1}^k |\text{supp}(x^r) \cap T_v| < |T_v| = \sum_{b=0}^1 |T_{v_b}|.$$

Thus the inequality:

$$\sum_{r=1}^k |\text{supp}(x^r) \cap T_{v_b}| < |T_{v_b}| \tag{9}$$

is true either for $b = 0$ or for $b = 1$. Let b^* be the smallest $b \in \{0, 1\}$ for which (9) is true. Parties proceed to the next iteration with v being replaced by v_{b^*} .

There are $d = O(\log n)$ iterations, at each parties communicate $O(\log n)$ bits. Hence π is $O(\log^2 n)$ -depth, as required. \blacktriangleleft

► **Remark 26.** Strategy from the proof of Proposition 20 is efficient only in terms of the number of rounds. In the next section we give another version of this strategy. This version will ensure that circuits we obtain from protocols for Q_k -communication games are not only low-depth, but also polynomial-size and explicit. For that, however, we require a bit more from the protocol π .

6 Effective version

Fix $f \in Q_k$. We say that a dag-like communication protocol π *strongly* computes Q_k -communication game for f if for every terminal t of π , for every $x \in f^{-1}(0)$ and for every $i \in [k]$ the following holds. If x is i -compatible with t , then $x_j = 0$, where $j = l(t)$ is the label of terminal t in the protocol π .

Similarly, fix $f \in R_k$. We say that a dag-like communication protocol π *strongly* computes R_k -communication game for f if for every terminal t of π , for every $x \in f^{-1}(0)$ and for every $i \in [k]$ the following holds. If x is i -compatible with t , then $x_j = b$, where $(j, b) = l(t)$ is the label of terminal t in the protocol π .

Strong computability is close to the notion of computability that Sokolov gave in [14] for general relations. Strong computability implies more intuitive notion of computability that we gave in the Preliminaries. The opposite direction is false in general.

Next we prove an effective version of Proposition 20.

► **Theorem 27.** *For every constant $k \geq 2$ there exists a polynomial-time algorithm A such that the following holds. Assume that $f \in Q_k$ and π is a dag-like protocol which strongly computes Q_k -communication game for f . Then, given the light form of π , the algorithm A outputs a circuit $C \leq f$, consisting only of THR_2^{k+1} gates and variables, such that $\text{depth}(C) = O(\text{depth}(\pi))$, $\text{size}(C) = O(\text{size}(\pi)^{O(1)})$.*

Proof. We will again give a $O(d)$ -round winning strategy of Learner in the Q_k -hypotheses game for f . Now, however, we should ensure that the light form of our strategy is of size $O(\text{size}(\pi)^{O(1)})$ and can be computed in time $O(\text{size}(\pi)^{O(1)})$ from the light form of π . Instead of specifying the light form of our strategy directly we will use the following trick. Assume that Learner has a *working tape* consisting of $O(\log \text{size}(\pi))$ cells, where each cell can store one bit. Learner memorizes all the Nature's responses so that he knows the current position of the game. But he *does not* store the sequence of Nature's responses on the working tape (there is no space for it). Instead, he first makes his hypotheses which depend on the current position. Then he receives a Nature's response $r \in \{0, 1, \dots, k\}$. And then he *modifies* the working tape, but the result should depend only on the current content of the working tape and on r (and not on the current position in a game). Moreover, we will ensure that modifying the working tape takes $O(\text{size}(\pi)^{O(1)})$ time, given the light form of π .

The main purpose of the working tape manifests itself in the end. Namely, at some point Learner decides to stop making hypotheses. This should be indicated on the working tape. More importantly, Learner's output should depend only on the content of working tape in the end (and not on the whole sequence of Nature's responses). Moreover, this should take $O(\text{size}(\pi)^{O(1)})$ time to compute that output, given the light form of π .

If a strategy satisfies these restrictions, then its light form is computable in $O(\text{size}(\pi)^{O(1)})$ time given the light form of π . Indeed, the underlying dag will consist of all possible configurations of the working tape. There are $O(\text{size}(\pi)^{O(1)})$ of them, as working tape uses $O(\log \text{size}(\pi))$ bits. For all non-terminal configurations c we go through all $r \in \{0, 1, \dots, k\}$. We compute what would be a configuration c_r of the working tape if the current configuration is c and Nature's response is r . After that we connect c to c_0, c_1, \dots, c_k . Finally, in all terminal configurations we compute the outputs of Learner. This gives a light form of our strategy in $O(\text{size}(\pi)^{O(1)})$ time.

Let V be the set of nodes of π and T be the set of terminals of π . Strategy proceeds in d iterations, each taking $O(1)$ rounds. We define sets \mathcal{Z}_h exactly as in the proof of Proposition 20. We also use the same notion of communication profile. However, we define completeness in a different way. First of all, instead of working with sets of nodes with no additional structure we will work with *multidimensional arrays* of nodes. Namely, we will consider k -dimensional arrays in which every dimension is indexed by integers from $[k]$. Formally, such arrays are functions of the form $M: [k]^k \rightarrow V$. We will use notation $M[c_1, \dots, c_k]$ for the value of M on $(c_1, \dots, c_k) \in [k]^k$.

Consider any $\mathcal{Z} \subset f^{-1}(0)$. We say that $g: \mathcal{Z} \rightarrow [k]$ is complete for \mathcal{Z} with respect to a multidimensional array $M: [k]^k \rightarrow V$ if for every $(c_1, \dots, c_k) \in [k]^k$, for every $i \in [k]$ and for every $z \in \mathcal{Z}$ the following holds. If $c_i = g(z)$, then z is i -compatible with $M[c_1, \dots, c_k]$.

We say that a multidimensional array $M: [k]^k \rightarrow V$ is complete for \mathcal{Z} if there exists $g: \mathcal{Z} \rightarrow [k]$ which is complete with respect to M .

To digest the notion of completeness it is instructive to consider the case $k = 2$. In this case M is a 2×2 table containing four nodes of π . The function $g: \mathcal{Z} \rightarrow [2]$ is complete for \mathcal{Z} with respect to M if the following holds. First, for every $z \in \mathcal{Z}$ two nodes in the $g(z)$ th row of M should be 1 -compatible with z . Second, for every $z \in \mathcal{Z}$ two nodes in the $g(z)$ th column of M should be 2 -compatible with z .

Let us now establish an analog of Lemma 21.

► **Lemma 28.** *Assume that $M: [k]^k \rightarrow T$ is complete for $\mathcal{Z} \subset f^{-1}(0)$. Let l be the output of π in the terminal $M[1, 2, \dots, k]$. Then $z_l = 0$ for every \mathcal{Z} .*

Proof. Since π strongly computes Q_k -communication game for f , it is enough to show that every $z \in \mathcal{Z}$ is i -compatible with $M[1, 2, \dots, k]$ for some i . Take $g: \mathcal{Z} \rightarrow [k]$ which is complete for \mathcal{Z} with respect to M . By definition z is $g(z)$ -compatible with $M[1, 2, \dots, k]$. ◀

We now proceed to the description of the Learner's strategy. The working tape of Learner consists of:

- an integer $iter$;
- a multidimensional array $M: [k]^k \rightarrow V$;
- $O(1)$ additional bits of memory.

Integer $iter$ will be at most $d \leq \text{size}(\pi)$ so to store all this information we need $O(\log(\text{size}(\pi)))$ bits, as required. Integer $iter$ always equals the number of iterations performed so far (at the beginning $iter = 0$). The array M changes only at the moments when $iter$ is incremented by 1. So let M_h denote the content of the array M when $iter = h$.

We call an array of nodes h -low if every node in it is either terminal or of depth at least h . Learner maintains the following invariant.

► **Invariant 29.** M_h is h -low and M_h is complete for \mathcal{Z}_h .

At the beginning Learner sets every element of M_0 to be the starting node of π so that Invariant 29 trivially holds.

Note that every node in M_d is a terminal of π . After d iterations Learner outputs the label of terminal $M_d[1, 2, \dots, k]$ in the protocol π . As M_d is complete for \mathcal{Z}_d due to Invariant 29, this by Lemma 28 will be a correct output in the Q_k -hypotheses game for f . Obviously producing the output takes polynomial time given the light form of π and the content of Learner's working tape in the end.

Now we need to perform an iteration. Assume that h iterations passed and Invariant 29 still holds. Let U_h be the set of all nodes appearing in M_h . Take any function $g: \mathcal{Z}_h \rightarrow [k]$ which is complete for \mathcal{Z}_h with respect to M_h .

For any $z \in f^{-1}(0)$ we denote by p_z a communication profile of z with respect to U_h . Recall that p_z is an element of $\{0, 1\}^{U_h}$, i.e., a function from U_h to $\{0, 1\}$. At each round of the $(h + 1)$ st iteration Learner asks Nature to specify some pair $(p, c) \in \{0, 1\}^{U_h} \times [k]$ such that $(p_z, g(z)) \neq (p, c)$, where z is the Nature's vector. Learner stores each (p, c) using his $O(1)$ additional bits on the working tape. Learner can do this until there are only k pairs from $(p_1, c_1), \dots, (p_k, c_k) \in \{0, 1\}^{U_h} \times [k]$ left which are not rejected by Nature. When this moment is reached, the $(h + 1)$ st iteration is finished. The iteration takes $2^{|U_h|} \cdot k - k = O(1)$ rounds, as required. For any z compatible with the current play after $h + 1$ iterations we know that $(p_z, g(z))$ is among $(p_1, c_1), \dots, (p_k, c_k)$, i.e.,

$$(p_z, g(z)) \in \{(p_1, c_1), \dots, (p_k, c_k)\} \text{ for all } z \in \mathcal{Z}_{h+1}. \quad (10)$$

Learner writes $(p_1, c_1), \dots, (p_k, c_k)$ on the working tape (all the pairs that were excluded are on the working tape and hence he can compute the remaining ones). Learner then computes a $(h + 1)$ -low array M_{h+1} which will be complete for \mathcal{Z}_{h+1} . To compute M_{h+1} he will only need to know $M_h, (p_1, c_1), \dots, (p_k, c_k)$ (this information is on the working tape) and the light form of π .

Namely, Learner determines $M_{h+1}[d_1, \dots, d_k]$ for $(d_1, \dots, d_k) \in [k]^k$ as follows. Consider the node $v = M_h[c_{d_1}, \dots, c_{d_k}]$. If v is a terminal, then set $M_{h+1}[d_1, \dots, d_k] = v$. Otherwise let $i \in [k]$ be the index of the party communicating at v . Look at p_{d_i} , which can be considered as a function of the form $p_{d_i}: U_h \rightarrow \{0, 1\}$. Define $r = p_{d_i}(v)$. Among two edges, starting at v , choose one which is labeled by r . Descend along this edge from v and let the resulting successor of v be $M_{h+1}[d_1, \dots, d_k]$.

Obviously, computing M_{h+1} takes $O(\text{size}(\pi)^{O(1)})$. To show that Invariant 29 is maintained we have to show that **(a)** M_{h+1} is $(h+1)$ -low and **(b)** M_{h+1} is complete for \mathcal{Z}_{h+1} .

The first part, **(a)**, holds because each $M_{h+1}[d_1, \dots, d_k]$ is either a terminal or a successor of a node of depth at least h . For **(b)** we define the following function:

$$g': \mathcal{Z}_{h+1} \rightarrow [k], \quad g'(z) = i, \text{ where } i \text{ is such that } (p_z, g(z)) = (p_i, c_i).$$

By (10) this definition is correct. We will show that g' is complete for \mathcal{Z}_{h+1} with respect to M_{h+1} .

For that take any $(d_1, \dots, d_k) \in [k]^k, z \in \mathcal{Z}_{h+1}$ and $i \in [k]$ such that $d_i = g'(z)$. We shall show that z is i -compatible with a node $M_{h+1}[d_1, \dots, d_k]$. By definition of g' we have that $g(z) = c_{d_i}$. As by Invariant 29 function g is complete for \mathcal{Z}_h with respect to M_h , this means that z is i -compatible with $v = M[c_{d_1}, \dots, c_{d_k}]$. If v is a terminal, then $M_{h+1}[d_1, \dots, d_k] = v$ and there is nothing left to prove.

Otherwise $v \in V \setminus T$. Let j be the index of the party communicating at v . By definition $M_{h+1}[d_1, \dots, d_k]$ is a successor of v . If $j \neq i$, i.e., not the i th party communicates at v , then any successor of v is i -compatible with z . Finally, assume that $j = i$. Node $M_{h+1}[d_1, \dots, d_k]$ is obtained from v by descending along the edge which is labeled by $r = p_{d_i}(v)$. Hence to show that z is i -compatible with $M_{h+1}[d_1, \dots, d_k]$ we should verify that at v on input z the i th party transmits the bit r . For that again recall that $g'(z) = d_i$, which means by definition of g' that $p_z = p_{d_i}$. I.e., p_{d_i} is the communication profile of z with respect to U_h . In particular, the value $r = p_{d_i}(v)$ is the bit transmitted by the i th party on input z at v , as required. ◀

In the same way one can obtain an analog of the previous theorem for the R_k -case.

► **Theorem 30.** *For every constant $k \geq 2$ there exists a polynomial-time algorithm A such that the following holds. Assume that $f \in R_k$ and π is a dag-like protocol which strongly computes R_k -communication game for f . Then, given the light form of π , the algorithm A outputs a circuit $C \leq f$, consisting only of THR_2^{k+1} gates and literals, such that $\text{depth}(C) = O(\text{depth}(\pi))$, $\text{size}(C) = O(\text{size}(\pi)^{O(1)})$.*

7 Derivation of Theorems 1 and 3

In this section we obtain Theorems 1 and 3 by devising protocols strongly computing the corresponding Q_k -communication games. Unfortunately, establishing strong computability requires diving into straightforward but tedious technical details, even for simple protocols.

Alternative proof of Theorem 1. We will show that there exists $O(\log n)$ -depth protocol π with polynomial-time computable light form, strongly computing Q_2 -communication game for MAJ_{2n+1} . By Theorem 27 this means that there is a polynomial-time computable $O(\log n)$ -depth formula $F \leq \text{MAJ}_{2n+1}$, consisting only of MAJ_3 gates and variables. From self-duality of MAJ_{2n+1} and MAJ_3 it follows that F computes MAJ_{2n+1} .

Take a polynomial-time computable $O(\log n)$ -depth monotone formula F' for MAJ_{2n+1} . Consider the following communication protocol π . The tree of π coincides with the tree of F' . Inputs to F' will be leafs of π . In a leaf containing input variable x_i the output of the protocol π is i . Remaining nodes of π are \wedge and \vee gates. In the \wedge gates communicates the first party, while in the \vee gates communicates the second party.

Fix an \wedge gate g (which belongs to the first party). Let g_0, g_1 be gates which are fed to g , i.e., $g = g_0 \wedge g_1$. There are two edges, starting at g , one leads to g_0 (and is labeled by 0) and the other leads to g_1 (and is labeled by 1). Take an input $a \in \text{MAJ}_{2n+1}^{-1}(0)$ to the first party. On input a at the gate g the first party transmits the bit $r = \min\{c \in \{0, 1\} \mid g_c(a) = 0\}$. If the minimum is over the empty set, then we set $r = 0$.

Take now an \vee gate h belonging to the second party. Similarly, there are two edges, starting at h , one leads to h_0 (and is labeled by 0) and the other leads to h_1 (and is labeled by 1). Here h_0, h_1 are two gates which are fed to h , i.e., $h = h_0 \vee h_1$. Take an input $b \in \text{MAJ}_{2n+1}^{-1}(0)$ to the second party. On input b at the gate h the second party transmits the bit $r = \min\{c \in \{0, 1\} \mid h_c(-b) = 1\}$. If the minimum is over the empty set, then we set $r = 0$. Here \neg denotes the bit-wise negation. Description of the protocol π is finished.

Clearly, the protocol π is of depth $O(\log n)$ and its light form is polynomial-time computable. It remains to argue that the protocol strongly computes Q_2 -communication game for MAJ_{2n+1} . Nodes of the protocol may be identified with the gates of F' . Consider any path $p = \langle e_1, \dots, e_m \rangle$ in the protocol π . Assume that e_j is an edge from g^{j-1} to g^j and g^0 is the output gate of F' . We shall show that the following: if $a \in \text{MAJ}_{2n+1}^{-1}(0)$ is 1-compatible with p , then $g^0(a) = g^1(a) = \dots = g^m(a) = 0$. Indeed, $g^0(a) = 0$ holds because F' computes MAJ_{2n+1} . Now, assume that $g^j(a) = 0$ is already proved. If g^j is an \vee gate, then $g^{j+1}(a) = 0$ just because g^{j+1} feeds to g^j . Otherwise g^j is an \wedge gate which therefore belongs to the first party. Let $r \in \{0, 1\}$ is the label of the edge e_{j+1} . Note that $g^{j+1} = g_r^j$, where g_0^j, g_1^j are two gates which are fed to g^j . Since a is 1-compatible with p , it holds that r coincides with the bit that the first party transmits at g^j on input a , i.e., with $\min\{c \in \{0, 1\} \mid g_c^j(a) = 0\}$. The set over which the minimum is taken is non-empty because $g^j(a) = 0$. In particular r belongs to this set, which means that $g^{j+1}(a) = g_r^j(a) = 0$, as required.

Similarly one can verify that if $b \in \text{MAJ}_{2n+1}^{-1}(0)$ is 2-compatible with p , then $g^0(-b) = g^1(-b) = \dots = g^m(-b) = 0$. Hence we get that if a leaf l is 1-compatible (2-compatible) with a (b) and l contains a variable x_i , then $a_i = 0$ ($-b_i = 1$). Hence the protocol strongly computes the Q_2 -communication game for MAJ_{2n+1} . ◀

Proof of Theorem 3. We will realize the protocol from the proof of Corollary 25 in such a way that it will give us $O(\log^2 n)$ -depth polynomial-size dag-like protocol with polynomial-time computable light form, strongly computing Q_k -communication game for THR_{n+1}^{kn+1} . By Theorem 27 this means that there is a polynomial-time computable $O(\log^2 n)$ -depth polynomial-size circuit $C \leq \text{THR}_{n+1}^{kn+1}$, consisting only of THR_2^{k+1} gates and variables. With the same argument as in Corollary 25 one can show that C coincides with THR_{n+1}^{kn+1} .

We will use the same tree T as in the proof of Corollary 25. Let us specify the underlying dag G of our protocol π . For a node v of T let \mathcal{S}_v be the set of all tuples $(s_1, s_2, \dots, s_k) \in \{0, 1, \dots, kn+1\}^k$ such that $s_1 + s_2 + \dots + s_k < |T_v|$. For every node v of T and for every $(s_1, s_2, \dots, s_k) \in \mathcal{S}_v$ the dag G will contain a node identified with a tuple $(v, s_1, s_2, \dots, s_k)$. These nodes of G will be called the *main nodes* (there will be some other nodes too). The starting node of G will be (r, n, \dots, n) , where r is the root of T . Note that if l is a leaf of T , then $|T_l| = 1$. Hence the only main node having l as the first coordinate is $(l, 0, \dots, 0)$. The set of terminals of π will coincide with the set of all main nodes of the form $(l, 0, \dots, 0)$, where l is a leaf of T . The output of π in $(l, 0, \dots, 0)$ is l .

For an integer $s \leq kn + 1$ let $W(s)$ be a binary tree of depth $O(\log n)$ with $|\{(a, b) \mid a, b \in \{0, 1, \dots, s\}, a + b = s\}|$ leaves. We assume that leaves of $W(s)$ are identified with elements of $\{(a, b) \mid a, b \in \{0, 1, \dots, s\}, a + b = s\}$. We use $W(s)$ in the construction of G . Namely, take any main node $(v, s_1, s_2, \dots, s_k)$ with a non-leaf v . Attach $W(s_1)$ to it. Then attach to every leaf of $W(s_1)$ a copy of $W(s_2)$. Next, to every leaf of the resulting tree attach a copy of $W(s_3)$ and so on. In this way we obtain a binary tree $W(v, s_1, \dots, s_k)$ of depth $O(\log n)$ growing at (v, s_1, \dots, s_k) . Its leaves can be identified with tuples of integers $(a_1, b_1, \dots, a_k, b_k)$ satisfying $a_1, b_1, \dots, a_k, b_k \geq 0, a_1 + b_1 = s_1, \dots, a_k + b_k = s_k$. We will merge every leaf of $W(v, s_1, \dots, s_k)$ with some main node. Namely, take a leaf $(a_1, b_1, \dots, a_k, b_k)$. If $a_1 + \dots + a_k < |T_{v_0}|$, then we merge $(a_1, b_1, \dots, a_k, b_k)$ with the main node (v_0, a_1, \dots, a_k) . Otherwise it should hold that $b_1 + \dots + b_k < |T_{v_1}|$. In this case we merge $(a_1, b_1, \dots, a_k, b_k)$ with the main node (v_1, b_1, \dots, b_k) .

Description of the dag of π is finished. Since k is constant, there are $n^{O(1)}$ main nodes and to each we attach a tree of depth $O(\log n)$. Hence π is $O(\log^2 n)$ -depth and $n^{O(1)}$ -size. Let us define a partition of non-terminal nodes between parties. Take a main node (v, s_1, \dots, s_k) , where v is not a leaf of T . The tree $W(v, s_1, \dots, s_k)$, growing from (v, s_1, \dots, s_k) consists of copies of $W(s_1), \dots, W(s_k)$. We simply say that the i th party communicates in copies of $W(s_i)$. After that we conclude that the light form of π is polynomial-time computable.

Now let us specify how the i th party communicates inside $W(s_i)$. Assume that $x \in \{0, 1\}^{kn+1}$ is the input to the i th party. If $|T_v \cap \text{supp}(x)| \neq s_i$, then the i th party communicates arbitrarily. Now, assume that $|T_v \cap \text{supp}(x)| = s_i$. Then the i th party communicates in such a way that the resulting path descends from the root of $W(s_i)$ to the leaf identified with a pair of integers $(|T_{v_0} \cap \text{supp}(x)|, |T_{v_1} \cap \text{supp}(x)|)$.

From this we immediately get the following observation. Let p be a path from the root of $W(v, s_1, \dots, s_k)$ to a leaf identified with a tuple $(a_1, b_1, \dots, a_k, b_k)$. Further, assume that $x \in (\text{THR}_{n+1}^{kn+1})^{-1}(0)$, satisfying $|T_v \cap \text{supp}(x)| = s_i$, is i -compatible with p . Then $a_i = |T_{v_0} \cap \text{supp}(x)|$ and $b_i = |T_{v_1} \cap \text{supp}(x)|$. Indeed, any such p passes through a copy $W(s_i)$ and leaves $W(s_i)$ in a leaf identified with $(|T_{v_0} \cap \text{supp}(x)|, |T_{v_1} \cap \text{supp}(x)|)$.

From this observation one can easily deduce that if $x \in (\text{THR}_{n+1}^{kn+1})^{-1}(0)$ is i -compatible with a main node (v, s_1, \dots, s_k) , then $|T_v \cap \text{supp}(x)| = s_i$. Indeed, we can obtain this by induction on the depth of v . Induction step easily follows from the previous paragraph. As for induction base we notice that $|T_r \cap \text{supp}(x)| = n$ for the root r of T (as in the proof of Corollary 25 we assume that $|\text{supp}(x)| = n$ as party can always add missing 1's).

In particular, this means that π strongly computes Q_k -communication game for THR_{n+1}^{kn+1} . Indeed, any terminal of π is of the form $(l, 0, \dots, 0)$, where l is a leaf of T . If $x \in (\text{THR}_{n+1}^{kn+1})^{-1}(0)$ is i -compatible with $(l, 0, \dots, 0)$, then, as shown in the previous paragraph, $|T_l \cap \text{supp}(x)| = |\{l\} \cap \text{supp}(x)| = 0$. This means that $x_l = 0$ and hence the output of the protocol is correct. ◀

8 Open problems

- Can Q_k -communication game for THR_{n+1}^{kn+1} be solved in $O(\log n)$ bits of communication for $k \geq 3$? Equivalently, can THR_{n+1}^{kn+1} be computed by $O(\log n)$ -depth circuit, consisting only of THR_2^{k+1} and variables? Can a deeper look into the construction of AKS sorting network help here (note that we only use this sorting network as a black-box)?
- Can at least R_k -communication game for THR_{n+1}^{kn+1} be solved in $O(\log n)$ bits of communication for $k \geq 3$? Again, this is equivalent to asking whether THR_{n+1}^{kn+1} can be computed by $O(\log n)$ -depth circuit, consisting only of THR_2^{k+1} and *literals*. Note that if we allow

literals (along with \wedge and \vee gates), then there are much simpler constructions of a $O(\log n)$ -depth formula for MAJ_n and, in fact, for every symmetric Boolean function [16]. Moreover, this can be done in terms of communication complexity [2]. A natural approach would be to apply ideas of [2] to R_k -communication games.

- Are there any other interesting functions in Q_k and R_k which can be analyzed with our technique?

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