# On Graph Complexity 

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#### Abstract

A circuit complexity of a graph is the minimum number of union and intersection operations needed to obtain the whole set of its edges starting from stars. Our main motivation to study this measure of graphs is that it is related to the circuit complexity of boolean functions.

We prove some lower bounds to the circuit complexity of explicitly given graphs. In particular, we use the graph theoretic frame to prove that some explicit subsets of $G F(2)^{n}$ cannot be covered by fewer than $2^{\Omega(n)}$ affine subspaces of $G F(2)^{n}$.

We conclude with several graph-theoretic problems whose solution would have intriguing consequences in computational complexity.


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## 1 Introduction

A major challenge in computational complexity is to exhibit an explicit boolean function $f:\{0,1\}^{m} \rightarrow\{0,1\}$ which has high computational complexity, i.e. cannot be computed using a small number of basic boolean operations such as OR $x \vee y$, AND $x \wedge y$ or Parity $x \oplus y=x+y(\bmod 2)$; inputs for such a circuit are literals, i.e. variables $x_{i}$ and their negations $\bar{x}_{i}$. Though this problem is intensively studied for more than fifty years there is no proof of a lower bound super-linear in the number of variables $m$. The problem of proving super-linear lower bounds is widely open even if we assume the additional restriction of circuit depth be ${ }^{1} O(\log m)$. The main difficulty here is that we want the function $f$ be explicitely constructed-easy counting shows that almost all functions require circuits of size $2^{\Omega(m)}$.

[^0]Pudlák, Rödl and Savický [32] observed that in oder to construct boolean functions requiring large circuits it would be enough to construct graphs that cannot be computed with a small number of union and intersection operations starting from some "simplest" graphs, like stars or cliques. In this paper we follow this approach in the case when input graphs are stars.

### 1.1 Graphs and circuits

Given a graph $G=(V, E)$ we associate to each its vertex $v$ a boolean variable $x_{v}$, and let $X=\left\{x_{v}: v \in V\right\}$. We say that a boolean function (or a circuit) $f(X)$ accepts a subset of vertices $S \subseteq V$ if $f\left(a_{S}\right)=1$ where $a_{S} \in\{0,1\}^{V}$ is a binary vector with 1's in positions $u$ for all $u \in S$, and 0 's elsewhere.

Definition 1.1. A boolean function $f(X)$ represents a graph $G$ if it accepts all edges and rejects all non-edges of $G$.

Hence, $f(X)$ represents the graph $G$ if for every input vector $a \in\{0,1\}^{X}$ with precisely two 1 's in, say, positions $u$ and $v, f(a)=1$ if $u v$ is an edge, and $f(a)=0$ if $u v$ is a non-edge of $G$. Note that if $u v$ is neither an edge nor a non-edge (in the bipartite case) or if $a$ contains more or less that two 1 's, then the value $f(a)$ may be arbitrary.

For example, the quadratic function $f_{G}(X)=\bigvee_{u v \in E} x_{u} x_{v}$ of a graph $G=(V, E)$ represents $G$ and can be computed by a trivial monotone depth-2 formula containing $|E|+1$ gate: $|E|$ fanin-2 AND gates and one OR gate of fanin $|E|$. Hence, no $n$ vertex graph $G$ of degree $d$ requires circuits of size larger than $d n$. However, this trivial upper bound may be exponentially far from the truth: a result of Alon [2] on the bipartite covering number of graphs implies (see Proposition 5.2 below) that every $n$-vertex graph $G$ of degree $d$ can be represented by a monotone CNF (conjunctive normal form) $f(X)=\bigwedge_{i=1}^{r} \bigvee_{v \in S_{i}} x_{v}$ with $r=O\left(d^{2} \log n\right)$. In particular, every graph of constant degree can be represented by a monotone depth- 2 formulas of logarithmic size.

In this paper we are interested in graphs which cannot be represented by small circuits. Although proving lower bounds on the circuit complexity of graphs may be of independent interest, we (just like the authors of [32]) consider the graph complexity mainly as a tool for proving lower bounds for boolean functions.

The translation of lower bounds for graphs to lower bounds for boolean functions is given by the following lemma (we give its proof in the Appendix). With every bipartite $n \times n$ graph $G \subseteq U \times W$ with $n=2^{m}$ and $U=W=\{0,1\}^{m}$ one may associate a boolean function $f$ in $2 m$ variables-the characteristic function of $G$ such that $f_{m}(u v)=1$ if and only if $u v \in G$.

Magnification Lemma: Given a circuit computing the characteristic function $f$ of a bipartite graph $G$, it is possible to replace its input literals by ORs of variables
so that the obtained monotone circuit (no negated inputs) represents the graph $G$. The same holds when Parity gates are used instead of OR gates.

This fact may be particularly useful in such circuit models where computing an OR (or a Parity) of input literals is "cheap." For example, if the circuit computing $f$ has unbounded fanin OR (or Parity) gates on the bottom (next to the inputs) level, then the obtained circuit represents $G$ and has the same number of gates! Hence, if we could prove that a bipartite $n \times n$ graph $G$ with $n=2^{m}$ cannot be represented using, say, fewer than $n^{\varepsilon}$ gates, this would immediately imply that the characteristic function $f$ of $G$ requires at least $n^{\varepsilon}=2^{\varepsilon m}$ gates, which is exponential in the number $2 m$ of variables of $f_{m}$ (this is where the term "magnification" comes from). That is, even moderate lower bounds on the computational complexity of graphs would yield high lower bounds on the computational complexity of boolean functions.

Note, however, that proving lower bounds for graphs may be more difficult task than for boolean functions. For example, the Parity function $x_{1} \oplus x_{2} \oplus \cdots \oplus x_{m}$ cannot be computed by a constant-depth circuit using a polynomial number of unbounded fanin AND and OR gates ([17]) whereas the corresponding to this function graph is just a union $(A \times \bar{B}) \cup(\bar{A} \times B)$ of two vertex-disjoint bipartite complete graphs, and can be represented by a circuit $\left(\bigvee_{u \in A \cup B} x_{u}\right) \wedge\left(\bigvee_{v \in \bar{A} \cup \bar{B}} x_{v}\right)$ using just three gates! This also demonstrates that the Magnification Lemma has no inverse: if a graph can be represented by a small circuit, this does not imply that its characteristic function can be computed by a small circuit.

On the other hand, non-trivial lower bounds on graph complexity even in the class of depth-3 circuits would resolve some old problems in the circuit complexity of boolean functions. Of particular interest is the case of $\Sigma_{3}$ circuits. These circuits consist of unbounded fanin AND and OR gates which are organised in three levels: the bottom (next to the inputs) level consists of OR gates, the middle level consists of AND gates, and the top level consists of a single OR gate. Inputs are variables and their negation. If there are no negated inputs then the circuit is monotone. The total number of gates is the size of a circuit. If all gates have fanout 1 , then the circuit is a formula.

Our motivation to study representation of graphs by depth-3 circuits comes from the following result due to Valiant [38]: if a boolean function $f$ in $m$ variables can be computed by a log-depth circuit of size $O(m)$ then $f$ can be computed by a $\Sigma_{3}$ formula of size $2^{O(m / \log \log m)}$; here a log-depth circuit is a circuit of depth $O(\log m)$ using any boolean functions in constant number of variables as gates. Together with the Magnification Lemma, this implies the following:

Valiant's Lemma: If a bipartite $n \times n$ graph cannot be represented by a monotone $\Sigma_{3}$ formula of size $n^{\varepsilon}$ with $\varepsilon=\omega(1 / \log \log \log n)$, then its characteristic function cannot be computed by a log-depth circuit of linear size.

In last two decades there was a considerable progress in proving lower bounds on the size of small-depth circuits $[1,12,39,17,34,37,13,19,28]$. However, for
$\Sigma_{3}$ circuits these lower bounds are of the form $2^{\Omega(\sqrt{m})}$, and hence, are too weak to imply lower bounds for log-depth circuits. The only known strongly exponential lower bounds were obtained in [29] under the restriction that the bottom OR gates have fanin 2, that is, when the circuit is just an OR of 2-CNFs. However, Valiant's reduction requires bottom fanin $m^{\varepsilon}$ and, as noted in [29], their argument fails already when bottom fanin is 4 .

### 1.2 Results

In this paper we are trying to obtain higher lower bounds for small depth circuits using the graph-theoretical frame. So far, we have not succeeded to do this for "pure" $\Sigma_{3}$ circuits but are able to do this for some of their variants. Among others we prove the following.

1. If a bipartite $n \times n$ graph $G$ has $M$ edges and contains no copies of $K_{a, b}$, then any $\Sigma_{3}$ circuit with Parity gates on the bottom needs at least $M /(a+b) n$ gates to represent $G$ (Theorem 3.2). This immediately yields strongly exponential lower bounds for many explicit boolean functions. In particular, this implies that any such circuit, detecting whether two given subsets of $\{1, \ldots, m\}$ are disjoint, needs $2^{\Omega(m)}$ gates (Corollary 3.5).
2. Any $\Sigma_{3}$ circuit which has Parity gates on the bottom and an arbitrary threshold gate on the top needs at least $\Omega(\sqrt{n})$ gates to represent an $n \times n$ Hadamard graph (Corollary 3.9). Again, this yields strongly exponential lower bounds for explicit boolean functions - including the Inner Product function-in this more general model.
3. There are explicit $n \times n$ graphs $G$ such that both $G$ and $\bar{G}$ can be represented by monotone $\Sigma_{3}$ circuits of size $O(\log n)$ but every monotone $\Pi_{3}$ circuit for $G$ must have size $\Omega(\sqrt{n})$; a $\Pi_{3}$ circuit is a dual version of $\Sigma_{3}$ circuits. Nothing similar holds in the context of boolean functions: the size of minimal a $\Pi_{3}$ circuit computing a function $f$ does not exceed the size of any $\Sigma_{3}$ circuit computing $\neg f$.
4. If a graph $G=(V, E)$ has no triangles and no 4-cycles, then any monotone formula (of arbitrary depth) computing the quadratic function $f_{G}(X)=$ $\bigvee_{u v \in E} x_{u} x_{v}$ has length $\Theta(|E|)$ (Theorem 7.3).

One of the aims of this paper is to draw once more readers attention to graph complexity. This concept has already led to interesting results $[7,36,32,35,30$, 26], and its potential seems to be far from being exhausted. As we will see, the circuit complexity of a graph is just a generalisation of its (edge) clique covering number, a well-known and widely studied measure of graphs. We show that nontrivial lower bounds on this generalised measure would have interesting consequences in computational complexity.

## Notation

We shall use standard graph theory notation. In particular, $K_{a, b}$ is a biclique (bipartite clique, complete bipartite graph) $A \times B, A \cap B=\emptyset$ with parts of size $a=|A|$ and $b=|B|$. We shall look at bipartite graphs $G$ with a fixed bipartition $V=U \cup W$ as sets $G \subseteq U \times W$ of their edges. By an $n$ to $n$ matching we will mean a bipartite graph $M_{n}$ consisting of $n$ vertex-disjoint edges. The (bipartite) complement of a bipartite graph $G \subseteq U \times W$ is a bipartite graph $\bar{G}=(U \times W) \backslash G$. A non-edge is a pair $u v$ of non-adjacent vertices. A non-edge in a bipartite graph with a bipartition $V=U \cup W$ is a pair $u v$ of non-adjacent vertices with $u \in U$ and $v \in W$; hence, pairs of vertices within one part of a bipartition are neither edges nor non-edges. The degree of a graph is the maximum degree of its vertices.

We will also use some well known graph covering measures of graphs. A fat matching is a union of vertex-disjoint bipartite cliques. A fat covering of a graph $G$ is a family of fat matchings such that each of these fat matchings is a (spanning) subgraph of $G$ and every edge of $G$ is an edge of at least one member of the family. Similarly, a bipartite clique covering of $G$ is a family $A_{1} \times B_{1}, \ldots, A_{t} \times B_{t}$ of complete bipartite subgraphs of $G$ such that every edge of $G$ is an edge of at least one member of the family. The number $t$ of subgraphs in such a covering is the size and the total number $\sum_{i=1}^{t}\left(\left|A_{i}\right|+\left|B_{i}\right|\right)$ of vertices is the weight of the covering.

Let $\mathrm{cc}(G)$ denote the minimum size and $\mathrm{cc}_{\mathrm{w}}(G)$ the minimum weight of a bipartite covering of $G$. These measures were first studied in [10], and now are the subject of extensive literature. In particular, it is known that $\mathrm{cc}_{\mathrm{w}}(G)=O\left(n^{2} / \log n\right)$ for every $n$-vertex graph, and there exist graphs matching this upper bound [8]. Let fat $(G)$ denote the minimum number of fat matchings in a fat covering of $G$. This measure was also considered by several authors, $[9,11,2,31]$ among others (here fat matchings are called "equivalence graphs"). In particular, it is known that fat $(G)=O(n / \log n)$ for every $n$-vertex graph [31]. In this paper we will use these measures to describe the circuit complexity of graphs.

## 2 Depth-2 circuits

To "warm-up" we start with the simplest model of circuits whose gates are arranged in two levels. Such circuits are easy to deal with, and the only goal of this section is to show that, even in this model, there may be a big discrepancy between the combinatorial and computational complexity of graphs: some "combinatorially complicated" (or "combinatorially interesting") graphs can be represented by very small circuits and some "combinatorially simple" graphs require large circuits (of the same type).
Example 2.1. The Kneser graph $K(r, k)(r>2 k \geq 4)$ has all $k$-element subsets $v$ of $\{1, \ldots, r\}$ as vertices, and two vertices are adjacent iff the corresponding $k$-subsets are disjoint. These graphs were introduced by Lovász [27] in his famous proof of Kneser's conjecture [23] that whenever the $k$-subsets of a $(2 k+s)$-set are divided into
$s+1$ classes, then two disjoint subsets end up in the same class. It is not difficult to see that $K(r, k)$ can be represented by the following depth- 2 circuit:

$$
\begin{equation*}
f(X)=\bigwedge_{i=1}^{r} \bigvee_{v \in S_{i}} x_{v} \tag{1}
\end{equation*}
$$

where $S_{i}=\{v: i \notin v\}$. Indeed, $u \neq v$ are non-adjacent in $K(r, k)$ iff $u \cap v \neq \emptyset$ iff $\exists i \in u \cap v$ iff $\{u, v\} \cap S_{i}=\emptyset$ for some $i$ iff $u v$ is rejected by some OR $\bigvee_{v \in S_{i}} x_{v}$. Note that with respect to the total number $n=\binom{r}{k}$ of vertices the representation is quite compact: the circuit has only $1+r=O\left(k n^{1 / k}\right)$ gates ( $r$ OR gates and one AND gate).

Example 2.2. An Hadamard matrix of oder $n$ is an $n \times n$ matrix with entries $\pm 1$ and with row vectors mutually orthogonal. A graph associated with an Hadamard matrix $M$ (or just an Hadamard graph) of oder $n$ is a bipartite $n \times n$ graph $H_{n}$ where two vertices $u$ and $v$ are adjacent if and only if $M(u, v)=+1$. An example of an Hadamard graph is the Sylvester graph $S(n)$. This is a bipartite $n \times n$ graph with $n=2^{r}$ vertices on each part identified with subsets of $\{1, \ldots, r\}$; two vertices $u$ and $v$ are adjacent iff $|u \cap v|$ is odd. It is easy to see that (for even $r$ ) this graph can be represented by a depth- 2 circuit

$$
\begin{equation*}
f(X)=\bigoplus_{i=1}^{r} \bigvee_{v \in S_{i}} x_{v} \tag{2}
\end{equation*}
$$

with $S_{i}=\{v: i \notin v\}$. Indeed, $u$ and $v$ are adjacent in $S(n)$ iff $|u \cap v|$ is odd iff $r-|u \cap v|$ is odd iff the number of sets $S_{i}$ containing at least one of $u$ and $v$ is odd. Again, the representation is quite compact: the circuit has only $r+1=\log (2 n)$ gates ( $r$ OR gates and one Parity gate of fanin $r+1$ ). On the other hand, each Hadamard graph (including the graph $S(n)$ ) is "combinatorially complicated" because, as shown in [32], it contains an induced subgraphs on $\sqrt{n}$ vertices which is Ramsey, meaning that it does not contain cliques or independent sets of size $\omega(\log n)$. By setting the corresponding variables in the circuit (2) to 0 , we obtain that this Ramsey graph can be represented by a depth- 2 circuit of $\operatorname{size} O(\log n)$. Razborov in [36] has shown that also some other "combinatorially complicated" graphs can be represented by small circuits of constant depth.

On the other hand, some "combinatorially simple" graphs-like an $n$ to $n$ matching $M_{n}$ or its complement-cannot be represented by depth- 2 circuits using fewer than $\Omega(n)$ gates. If a graph $G$ is represented by a depth- 2 circuit of the form (1) with top fanin $r$ then its complement is just a union of $r$ cliques. Hence, $r=\Omega(n)$ for any such circuit representing $\overline{M_{n}}$.

High lower bounds for depth-2 circuits of the form (2) can be obtained via simple rank argument: every such circuit representing a graph $G$ must have size at least
$\operatorname{rk}(G) / 2$ where $\operatorname{rk}(G)$ is the rank over $G F(2)$ of the adjacency matrix of $G$ (just because each graph represented by an OR gate is a complement of a clique, and hence, has rank at most 2). Hence, $r=\Omega(n)$ for any circuit of the form (2) representing $M_{n}$.

These examples show that proving high lower bounds for depth-2 circuits is a relatively easy task. However, the case of depth-3 circuits turns out to be much more difficult. And this is not surprising because, as we already mentioned above, high lower bounds on the size of such circuits (of the form $n^{\varepsilon}$ where $\varepsilon$ may even tend slowly to 0 ) would resolve some old problems in computational complexity of boolean functions, including the problem of proving a super-linear lower bound for log-depth circuits.

## 3 Depth-3 circuits with Parity gates on the bottom

By Valiant's Lemma, high lower bounds on the size of monotone $\Sigma_{3}$ circuits representing an explicit bipartite graphs would give us super-linear lower bounds for logarithmic depth circuits. Pudlák, Rödl and Savický asked in [32] whether $C_{4}$-free graphs are hard for such circuits.

It turns out that this question can be answered affirmatively if the OR gates on the bottom are replaced by Parity gates; we call such circuits $\Sigma_{3}^{\oplus}$ circuits. Such a circuit of top fanin $s$ and middle fanin $r$ is just an OR of $s$ boolean functions, each of which is a product of $r$ linear forms over $G F(2)$ :

$$
g\left(x_{1}, \ldots, x_{m}\right)=\prod_{i=1}^{r} \bigoplus_{j \in I_{i}} x_{i} \oplus \lambda_{i}
$$

where $\lambda_{i} \in\{0,1\}$ and $I_{i} \subseteq\{1, \ldots, m\}$. If all scalars $\lambda_{i}$ are equal 0 , the circuit is positive.

Remark 3.1. If we would require that the top gate of a circuit must also be a Parity gate (not an OR gate), then truly exponential lower bounds $2^{\Omega(m)}$ for such version of $\Sigma_{3}^{\oplus}$ circuits could be obtained using the algebraic (approximation by low-degree polynomials) techniques of [34, 37, 13]. However, these techniques seem incapable of proving such high lower bounds for $\Sigma_{3}^{\oplus}$ circuits themselves because, in this case, we would be forced to approximate the top OR gate as well, which would invariably result in the square root $2^{\Omega(\sqrt{m})}$ in the final bound.

Theorem 3.2. Let $G \subseteq U \times W$ be a bipartite $n \times n$ graph. If $G$ contains no copy of $K_{a, b}$, then any $\Sigma_{3}^{\oplus}$ circuit computing the characteristic function of $G$ has top fanin at least

$$
\frac{|G|}{(a+b) n}
$$

Theorem 3.2 follows directly from the Magnification Lemma and the following two lemmas.

Lemma 3.3. For every bipartite graph $G \subseteq U \times W$, $\operatorname{fat}(G)$ is the minimum top fanin of a $\Sigma_{3}^{\oplus}$ circuit representing $G$.

Proof. First note that in the case of graphs we can safely restrict ourselves to positive circuits, because $\oplus_{u \in A \cup \bar{B}} x_{u}$ represents the same graph as $1 \oplus \oplus_{u \in A \cup B} x_{u}$.

Let $g=\oplus_{v \in A \cup B} x_{v}$ with $A \subseteq U$ and $B \subseteq W$ be a gate on the bottom level of a $\Sigma_{3}^{\oplus}$ circuit representing $G$. Then $g$ represents a fat matching $(A \times \bar{B}) \cup(\bar{A} \times B)$ where $\bar{A}=U \backslash A$ and $\bar{B}=W \backslash B$. Since the intersection of any number of fat matchings is a fat matching, each AND gate on the middle level represents a fat matching. Hence, if the circuit has top fanin $s$, then the OR gate on the top represents a union of these $s$ fat matchings, implying that $s \geq \operatorname{fat}(G)$.

To show that $G$ can be represented by a $\Sigma_{3}^{\oplus}$ circuit of top fanin fat $(G)$, assume that $G$ is a union of $s$ fat matchings. Observe that every fat matching $M=A_{1} \times B_{1} \cup \cdots \cup$ $A_{r} \times B_{r}$ can be obtained as an intersection of $r$ fat matchings $A_{i} \times B_{i} \cup\left(A \backslash A_{i}\right) \times\left(B \backslash B_{i}\right)$ in $A \times B$ where $A(B)$ is the union of all $A_{j}$ 's ( $B_{j}$ 's). Since each fat matching $A \times B \cup \bar{A} \times \bar{B}$ can be represented by the Parity gate $g(X)=\oplus_{v \in A \cup \bar{B}} x_{v}$, we are done.

Lemma 3.4. Let $G \subseteq U \times W$ be a bipartite $n \times n$ graph. If $G$ contains no copy of $K_{a, b}$ then

$$
\operatorname{fat}(G) \geq \frac{|G|}{(a+b) n}
$$

Proof. Let $H=\bigcup_{i=1}^{t} A_{i} \times B_{i}$ be a fat matching, and suppose that $H \subseteq G$. By the definition of a fat matching, the sets $A_{1}, \ldots, A_{t}$, as well as the sets $B_{1}, \ldots, B_{t}$ are mutually disjoint. Moreover, since $G$ contains no copy of $K_{a, b}$, we have that $\left|A_{i}\right|<a$ or $\left|B_{i}\right|<b$ for all $i$. Hence, if we set $I=\left\{i:\left|A_{i}\right|<a\right\}$, then

$$
|H|=\sum_{i=1}^{t}\left|A_{i} \times B_{i}\right|=\sum_{i=1}^{t}\left|A_{i}\right| \cdot\left|B_{i}\right| \leq \sum_{i \in I} a \cdot\left|B_{i}\right|+\sum_{i \notin I}\left|A_{i}\right| \cdot b \leq(a+b) n
$$

Thus, no fat matching $H \subseteq G$ can cover more than $(a+b) n$ edges of $G$, implying that we need at least $|G| /(a+b) n$ fat matchings to cover all edges of $G$.

There are many explicit bipartite $n \times n$ graphs which are dense enough and do not have large bicliques. Theorem 3.2 immediately yields truly exponential lower bounds (i.e. lower bounds of the form $2^{\Omega(m)}$ ) on the top fanin of $\Sigma_{3}^{\oplus}$ circuits computing the characteristic functions of these graphs; recall that these functions have only $m=2 \log n$ variables. Here we restrict ourselves with few examples.

The disjointness function is a boolean function $D I S J_{2 m}$ in $2 m$ variables such that

$$
D I S J_{2 m}\left(y_{1}, \ldots, y_{m}, z_{1}, \ldots, z_{m}\right)=1 \text { if and only if } \sum_{i=1}^{m} y_{i} z_{i}=0
$$

Corollary 3.5. Every $\Sigma_{3}^{\oplus}$ circuit computing DISJ $J_{2 m}$ has top fanin $2^{\Omega(m)}$.

Proof. The function $D I S J_{2 m}$ is the characteristic function of the Kneser-type bipartite graph $K(r) \subseteq U \times V$ where $U$ and $W$ consist of all $n=2^{r}$ subsets of $[r]=\{1, \ldots, r\}$, and $u v \in K$ iff $u \cap v=\emptyset$. The graph $K(r)$ can contain a complete bipartite $a \times b$ subgraph $\emptyset \neq A \times B \subseteq K$ only if $a \leq 2^{k}$ and $b \leq 2^{r-k}$ for some $0 \leq k \leq r$, because $\left(\bigcup_{u \in A} x_{u}\right) \cap\left(\bigcup_{v \in B} x_{v}\right)=\emptyset$. In particular, $K(r)$ does not contain a copy of $K_{a, a}$ with $a>2^{r / 2}=\sqrt{n}$. Since this graph has

$$
|K(r)|=\sum_{u \in U} d(u)=\sum_{u \in U} 2^{r-|u|}=\sum_{i=0}^{r}\binom{r}{i} 2^{r-i}=3^{r} \geq n^{1.58}
$$

edges, Theorem 3.2 yields that any $\Sigma_{3}^{\oplus}$ circuit representing $K(r)$ must have top fanin at least $|K(r)| /(2 a n)=\Omega\left(n^{0.08}\right)$. It remains to apply the Magnification Lemma.

Remark 3.6. In the context of boolean functions, $\Sigma_{3}^{\oplus}$ circuits cannot be efficiently simulated by $\Sigma_{3}$ circuits: the Parity function $x_{1} \oplus x_{2} \oplus \cdots \oplus x_{m}$ has an obvious $\Sigma_{3}^{\oplus}$ circuit of size 1, whereas (as shown in [17]) this function requires $\Sigma_{3}$ circuits of size $2^{\Omega(\sqrt{m})}$. It may be, therefore, interesting to note that, in the context of graphs, the situation is entirely different: if a graph can be represented by a $\Sigma_{3}^{\oplus}$ circuit of size $L$ then $G$ can be represented by a monotone $\Sigma_{3}$ circuit of size at most $2 L$. This holds because we can just replace each parity gate $\bigoplus_{u \in S} x_{u}$ on the bottom level by an AND $\left(\bigvee_{u \in S} x_{u}\right) \wedge\left(\bigvee_{u \notin S} x_{u}\right)$ of two OR gates; the obtained monotone $\Sigma_{3}$ circuit will represent the same graph. Moreover, the graph $K(r)$ shows that $\Sigma_{3}^{\oplus}$ circuits may be even exponentially weaker: this graph can be represented by a monotone CNF of length $O(\log n)$ (see (1)) but requires $\Sigma_{3}^{\oplus}$ circuits of top fanin at least $\Omega\left(n^{\varepsilon}\right)$.

A prominent example of a dense bipartite graph without $K_{2,2}$ is the incidence $n \times n$ graph $P_{n} \subseteq U \times W$ of a projective plane $P G(2, q)$ of order $q\left(n=q^{2}+q+1\right)$. This graph is $k$-regular with $k=q+1$ and contains no copies of $K_{2,2}$. Hence, the graph has $\Omega\left(n^{3 / 2}\right)$ edges. This is almost optimal because, as shown in [24], no bipartite $n \times n$ graph without a copy of $K_{2,2}$ can have more than $(1+o(1)) n^{3 / 2}$ edges. By Singer's theorem ([16], p. 128) there exist $0 \leq a_{1}<a_{2}<\ldots<a_{q+1}<n$ such that $P_{n}$ is isomorphic to the bipartite graph will parts $U=W=\{0,1, \ldots, n-1\}$ in which $u \in U$ is joined to $v \in W$ iff $u=\left(v+a_{i}\right) \bmod n$ for some $1 \leq i \leq q+1$.

For the characteristic function $\pi_{2 m}$ of this graph, Theorem 3.2 yields
Corollary 3.7. Every $\Sigma_{3}^{\oplus}$ circuit computing $\pi_{2 m}$ has top fanin $\Omega\left(2^{m / 2}\right)$.
For every constant $a>1$ explicit constructions of $n \times n$ graphs (so-called normgraphs ) with $\Omega\left(n^{2-1 / a}\right)$ edges and no copies of $K_{a, a!+1}$ were found by Kollár, Rónyai and Szabó in [25]; explicit graphs without $K_{s, r}$ with larger values of $r$ and $s$ were earlier constructed by Andreev [4]. For the characteristic functions $f_{2 m}^{a}$ of these graphs, Theorem 3.2 yields

Corollary 3.8. For every constant $a>1$, every $\Sigma_{3}^{\oplus}$ circuit computing $f_{2 m}^{a}$ has top fanin $\Omega\left(2^{m-1 / a}\right)$.

The only previously known truly exponential lower bound for $\Sigma_{3}^{\oplus}$ circuits we are aware of was proved by Grolmusz [14] for the Inner Product function

$$
I P_{2 m}\left(y_{1}, \ldots, y_{m}, z_{1}, \ldots, z_{m}\right)=\sum_{i=1}^{m} y_{i} z_{i}(\bmod 2)
$$

Quite recently Pudlák and Rödl [33] have also proved such a lower bound for the characteristic functions of particular pseudorandom sets. Both proofs employ nontrivial facts - the probabilistic communication complexity of $I P$ in [14] and some properties of pseudorandom sets in [33].

Actually, the lower bounds in [14] and [33] were proved for a more general model of $\Sigma_{3}^{\oplus}$ circuits: instead of an OR gate they allow an arbitrary threshold gate on the top level. (Recall that a threshold- $k$ function accepts an input iff it contains at least $k$ 1's.) Let us show that lower bounds for this extended model can be proved in the context of graphs as well.

At this point, let us note that in some cases it can even make sense to reprove known lower bounds for boolean functions in the frame of graphs. For example, reproving known lower bound $2^{\Omega(\sqrt{m})}$ for $\Sigma_{3}$ circuits-or even proving a much weaker lower bound $2^{(\log m)^{\omega(1)}}$-in the graph-theoretic frame would give us a graph outside the second level of the communication complexity hierarchy introduced in [5].
Corollary 3.9. Any $\Sigma_{3}^{\oplus}$ circuit which has an arbitrary threshold gate on the top and represents an $n \times n$ Hadamard graph must have top fanin $\Omega(\sqrt{n})$.

Since the inner product function $I P_{2 m}$ is the characteristic function of an Hadamard $n \times n$ graph $H_{n}$ with $n=2^{m}$, Corollary 3.9 and the Magnification Lemma immediately yield a lower bound $\Omega\left(2^{m / 2}\right)$ for $I P_{2 m}$ in this class of circuits.

For the proof of Corollary 3.9 we need the so-called "discriminator lemma" for threshold gates. Let $\mathcal{F}$ be a family of subsets of a finite set $X$. For a subset $A \subseteq X$, let $\operatorname{thr}_{\mathcal{F}}(A)$ denote the minimum number $t$ for which there exist $t$ members $B_{1}, \ldots, B_{t}$ of $\mathcal{F}$ and a number $0 \leq k \leq t$ such that, for every $x \in X, x \in A$ if and only if $x$ belongs to at least $k$ of $B_{i}$ 's. A set $A$ is an $\varepsilon$-discriminator for a set $B$ if

$$
\left|\frac{|A \cap B|}{|A|}-\frac{|\bar{A} \cap B|}{|\bar{A}|}\right| \geq \varepsilon .
$$

Lemma 3.10. ([15]) If $\operatorname{thr}_{\mathcal{F}}(A) \leq t$ then $A$ is a $1 / t$-discriminator for some $B \in \mathcal{F}$.
Proof. Let $B_{1}, \ldots, B_{t} \in \mathcal{F}$ be a threshold- $k$ covering of $A$, i.e. $x \in A$ iff $x$ belongs to at least $k$ of $B_{i}$ 's. Our goal is to show that then $A$ is a $1 / t$-discriminator for at least one $B_{i}$. Since every element of $A$ belongs to at least $k$ of the sets $A \cap B_{i}$, the average size of these sets must be at least $k$. Since no element of $\bar{A}$ belongs to more than $k-1$ of the sets $\bar{A} \cap B_{i}$, the average size of these sets must be at most $k-1$. Hence,

$$
1 \leq \sum_{i=1}^{t} \frac{\left|A \cap B_{i}\right|}{|A|}-\sum_{i=1}^{t} \frac{\left|\bar{A} \cap B_{i}\right|}{|\bar{A}|}
$$

$$
\leq t \cdot \max _{1 \leq i \leq t}\left|\frac{\left|A \cap B_{i}\right|}{|A|}-\frac{\left|\bar{A} \cap B_{i}\right|}{|\bar{A}|}\right| .
$$

Proof of Corollary 3.9. Let $A$ be an $n \times n$ Hadamard graph. Lindsey's lemma (see, e.g. [3] or [5]) says that the absolute value of the difference between the number of $+1^{\prime}$ s and $-1^{\prime}$ s in any $a \times b$ submatrix of $A$ is at most $\sqrt{a b n}$. This, in particular, implies that both $A$ and $\bar{A}$ have $\Theta\left(n^{2}\right)$ edges. Hence, by Lemmas 3.3 and 3.10, it is enough to show that $\|A \cap B|-| \bar{A} \cap B\|=O\left(n^{3 / 2}\right)$ for every fat matching $B=\bigcup_{i=1}^{t} S_{i} \times R_{i}$. By Lindsey's lemma, the absolute value of the difference between $\left|A \cap\left(S_{i} \times R_{i}\right)\right|$ and $\left|\bar{A} \cap\left(S_{i} \times R_{i}\right)\right|$ does not exceed $\sqrt{s_{i} r_{i} n}$ where $s_{i}=\left|S_{i}\right|$ and $r_{i}=\left|R_{i}\right|$. Since, $\sum_{i=1}^{t} s_{i} \leq n$ and $\sum_{i=1}^{t} s_{i} \leq n$, we obtain

$$
\begin{aligned}
||A \cap B|-|\bar{A} \cap B|| & =\left|\sum_{i=1}^{t}\right| A \cap\left(S_{i} \times R_{i}\right)\left|-\sum_{i=1}^{t}\right| \bar{A} \cap\left(S_{i} \times R_{i}\right)| | \\
& \leq \sum_{i=1}^{t}| | A \cap\left(S_{i} \times R_{i}\right)\left|-\left|\bar{A} \cap\left(S_{i} \times R_{i}\right)\right|\right| \\
& \leq \sum_{i=1}^{t} \sqrt{s_{i} r_{i} n} \leq \sqrt{n} \sum_{i=1}^{t} \frac{s_{i}+r_{i}}{2} \leq n^{3 / 2} .
\end{aligned}
$$

## $4 \quad$ A tradeoff for $\Sigma_{3}$ circuits

We now use the graph theoretic frame to prove a trade-off between top and middle fanins in $\Sigma_{3}$ circuits, where middle fanin of a circuit is the maximum fanin of a gate in the middle level.

Theorem 4.1. If $I P_{2 m}$ is computed by a $\Sigma_{3}$ circuit with top fanin $s$ and middle fanin $r$, then both $s 2^{r}$ and $r^{s}$ must be at least $2^{\Omega(m)}$.

A trade-off $s r=\Omega\left(m^{3} /(\log m)^{5}\right)$ between these parameters for $I P_{2 m}$ was recently proved by Lokam [26] (also using the graph-theoretic frame). The trade-off in Theorem 4.1 is better only if one of the parameters $r$ or $s$ is at most $m^{\varepsilon}$ - the second parameter must then be at least $2^{\Omega\left(m^{1-\varepsilon}\right)}$.

Theorem 4.1 follows directly from Lemma 3.4 and the following
Lemma 4.2. If a bipartite graph $G$ can be represented by a monotone $\Sigma_{3}$ circuit of middle fanin $r$ and top fanin $s$, then $\operatorname{cc}(G) \leq s 2^{r}$ and $\operatorname{cc}(\bar{G}) \leq r^{s}$.

Proof. Take a monotone $\Sigma_{3}$ circuit of middle fanin at most $r$ and top fanin $s$, and let $G \subseteq U \times W$ be the bipartite graph represented by this circuit. Each gate $g=\bigvee_{i \in S} x_{i}$
on the bottom level represents a (bipartite) complement of a bipartite clique $A \times B$, where $A=U \backslash S$ and $B=W \backslash S$. Each such complement is a union of two bipartite cliques $A \times \bar{B}$ and $\bar{A} \times W$. Since the intersection of any number of bipartite cliques is a (possibly empty) bipartite clique, each AND gate on the middle level represents a union of at most $2^{r}$ bipartite cliques. Since $G$ is a union of $s$ such graphs, we have $\operatorname{cc}(G) \leq s 2^{r}$.

To prove $\operatorname{cc}(\bar{G}) \leq r^{s}$, observe that $\bar{G}$ is an intersection of $s$ graphs $H_{1}, \ldots, H_{s}$, each of which is a union of $r$ bipartite cliques. Since the intersection of any number of bipartite cliques is a bipartite clique, we have that $\operatorname{cc}(\bar{G}) \leq \prod_{i=1}^{s} \operatorname{cc}\left(H_{i}\right) \leq r^{s}$.

## 5 Combinatorics of $\Sigma_{3}$ circuits

In this section we give a combinatorial characterisation of graphs represented by monotone $\Sigma_{3}$ circuits. Recall that each such circuit is just an OR of CNFs (conjunctive normal forms), where a CNF of length $t$ is an AND

$$
\begin{equation*}
g(X)=\left(\bigvee_{u \in S_{1}} x_{u}\right) \wedge \cdots \wedge\left(\bigvee_{u \in S_{t}} x_{u}\right) \tag{3}
\end{equation*}
$$

of $t$ clauses, each of which is an OR of variables. Let $\operatorname{cnf}(G)$ denote the minimum length of a CNF representing $G$, and let $\Sigma_{3}(G)$ be the minimum number $t$ such that $G$ can be represented as a union of at most $t$ graphs $H$ such that $\operatorname{cnf}(H) \leq t$.

The length of CNFs can be described combinatorially in terms of the clique covering number as well as in terms of set-intersections. Say that a bipartite graph $G \subseteq U \times W$ admits an intersection representation of size $t$ if it is possible to associate with every vertex $u \in U \cup W$ a subset $A_{u}$ of $\{1, \ldots, t\}$ so that for every arc $u v \in U \times W, u v \in G$ iff $A_{u} \cap A_{v}=\emptyset$. Let $\operatorname{int}(G)$ denote the smallest $t$ for which $G$ admits such a representation.

Proposition 5.1. For every bipartite graph $G$ we have $\operatorname{cnf}(G)=\operatorname{cc}(\bar{G})=\operatorname{int}(G)$.
Proof. An OR of variables represents a complement of a biclique (and each complement of a biclique can be represented by an OR gate). Hence, a bipartite graph $G$ can be represented by a CNF of the form (3) iff $G$ is an intersection of complements of $t$ bicliques, or equivalently, iff the complement $\bar{G}$ can be represented as a union of $t$ bicliques, implying that $\operatorname{cnf}(G)=\operatorname{cc}(\bar{G})$.

The equality $\operatorname{cc}(\bar{G})=\operatorname{int}(G)$ is also easy to show. Given an intersection representation of $G \subseteq U \times W$ by subsets $A_{u}$ of $\{1, \ldots, t\}$ for $u \in U \cup W$, the $t$ sets $I_{i}=\left\{u: i \in A_{u}\right\}$ are independent and cover all non-edges of $G$. On the other hand, given a covering of the non-edges of $G$ by independent sets $I_{1}, \ldots, I_{t}$, one can take $A_{u}=\left\{i: u \in I_{i}\right\}$.

Proposition 5.1, together with an obvious observation that every bipartite clique $A \times B$ can be represented by a CNF consisting of two clauses $\bigvee_{u \in A} x_{u}$ and $\bigvee_{v \in B} x_{v}$,
gives a general upper bound

$$
\begin{equation*}
\Sigma_{3}(G) \leq \min \{\operatorname{cc}(G), \operatorname{cc}(\bar{G})\} \tag{4}
\end{equation*}
$$

By $(4), \Sigma_{3}(G) \leq \operatorname{cc}(G) \leq n$ is a trivial upper bound for every $n$-vertex graph $G$. For graphs of small degree we have a better upper bound.

Proposition 5.2. For every n-vertex graph $G$ of degree $d, \operatorname{cnf}(G)=O\left(d^{2} \log n\right)$ and $\Sigma_{3}(G)=O\left(d^{2 / 3} \log n\right)$

Proof. Alon [2] has proved (using a probabilistic argument) that cc $(\bar{G})=O\left(d^{2} \log n\right)$. Together with Proposition 5.1 this yields the first claim. To get the second claim, simply break $G$ into $d^{2 / 3}$ subgraphs of maximal degree $d^{1 / 3}$ each.

Proposition 5.3. If $H$ is a fat matching then both $\operatorname{cnf}(H)$ and $\Sigma_{3}(\bar{H})$ are at most $2 \log n$. Moreover, if $M_{n}$ is an $n$ to $n$ matching then $\Sigma_{3}\left(M_{n}\right)=\Omega(\log n)$.

Proof. Let us first look how an $n$ to $n$ matching $M_{n}$ can be represented by a monotone CNF of length $O(\log n)$. Let $t=2 \log n$ and associate with each vertex $u_{i}$ on the left side its own $t / 2$-element subset $A_{i}$ of $\{1, \ldots, t\}$, and assign to the unique matched vertex $v_{i}$ on the right side the complement $B_{i}=\overline{A_{i}}$ of this subset. It is clear that then $A_{i} \cap B_{j}=\emptyset$ iff $i=j$. Hence, $\operatorname{cnf}\left(M_{n}\right)=\operatorname{int}\left(M_{n}\right) \leq t=2 \log n$. The same argument clearly works for any fat matching $H$. The upper bound $\Sigma_{3}(\bar{H}) \leq 2 \log n$ follows from the upper bound (4).

To prove the lower bound $\Sigma_{3}\left(M_{n}\right)=\Omega(\log n)$, let $t=\Sigma_{3}\left(M_{n}\right)$. Then there exists a matching $H \subseteq M_{n}$ containing $|H| \geq n / t$ edges and admitting an intersection representation of size $t$. Since the sets of neighbours of any two vertices in $H$ are distinct, all the sets $A_{u}$ associated with vertices $u$ on the left (resp. on the right) part of the bipartition must be distinct. This implies $2^{t} \geq|H| \geq n / t$, and hence, $t=\Omega(\log n)$.

So far we do not know of any explicit $n$-vertex graphs $G$ with $\Sigma_{3}(G)$ substantially larger than $\log n$. The best what we know is the lower bound of the form $\Sigma_{3}\left(H_{n}\right) \geq$ $(\log n)^{3 / 2-o(1)}$ proved by Lokam in [26] for an Hadamard graph $H_{n}$.

## $6 \quad \Sigma_{3}$ versus $\Pi_{3}$ circuits

As mentioned above, no explicit $n$-vertex graphs requiring monotone $\Sigma_{3}$ circuits of size $(\log n)^{\omega(1)}$ are known. On the other hand, if we replace the ANDs by ORs and vice versa, then the situation is much easier. The obtained "dual" circuits are known as $\Pi_{3}$ circuits and have the form:

$$
f(X)=\bigwedge_{i=1}^{s} \bigvee_{j=1}^{r} \bigwedge_{v \in S_{i j}} x_{v} ;
$$

by the size of such a circuit we again mean $\max \{s, r\}$. Since ANDs of more than 3 variables do not contribute to the representation of graphs, such circuits are just a slight generalisation of CNFs. And indeed, the representation power of such circuits is much weaker than that of monotone $\Sigma_{3}$ circuits. To see this, take an $n$ to $n$ matching. We already know that both $M_{n}$ and $\bar{M}_{n}$ can be represented by monotone $\Sigma_{3}$ circuits of size $O(\log n)$. Moreover, $M_{n}$ has also a monotone $\Pi_{3}$ circuit of logarithmic size, because $\operatorname{cnf}\left(M_{n}\right)=O(\log n)$. On the other hand, we have

Theorem 6.1. Every monotone $\Pi_{3}$ circuit representing $\bar{M}_{n}$ must have size at least $\Omega(\sqrt{n})$.
Note that nothing similar holds in the case of boolean functions: every $\Sigma_{3}$ circuit for a boolean function $f$ can be tranformed to a $\Pi_{3}$ circuit of the same size for its complement $\neg f$.

A larger lower bound on the size of monotone $\Pi_{3}$ circuits can be obtained for Hadamard graphs.

Theorem 6.2. Every monotone $\Pi_{3}$ circuit representing an Hadamard graph of oder $n$ must have size at least $\Omega\left(n^{2 / 3}\right)$.

We derive both theorems from the following property of graphs represented by monotone $\Pi_{3}$ circuits.

Lemma 6.3. Suppose that a graph $G$ can be represented by a monotone $\Pi_{3}$ circuit of size $t$. Then it is possible to add to $\bar{G}$ a set $E$ of $|E| \leq t^{2}$ edges so that $\operatorname{cc}(\bar{G} \cup E) \leq t$.

Proof. Suppose that a graph $G \subseteq U \times W$ can be represented by a monotone $\Pi_{3}$ circuit of size $t$. Such a circuit is an AND of at most $t$ monotone DNFs $D_{1}, \ldots, D_{t}$, each containing at most $t$ monomials (ANDs of variables). Since we are interested in the behaviour of the circuit only on arcs (edges and non-edges), we may assume that none of these monomials contains more than two variables. Hence, each of the DNFs

$$
D_{i}=\bigvee_{u \in S_{i}} x_{u} \vee \bigvee_{u v \in F_{i}} x_{u} x_{v}
$$

accepts some set $S_{i} \subseteq U \cup W$ of vertices and some set $F_{i}$ of $\left|F_{i}\right| \leq t$ arcs. Let $E=\bigcup_{i=1}^{t} E_{i}$ where $E_{i}=F_{i} \cap G$ is the set of edges of $G$ accepted by the $i$-th DNF; hence, $|E| \leq t^{2}$. We may assume that the set $G \backslash H$ of remaining edges is non-empty, since otherwise we would have $H=G$, meaning that $\bar{G} \cup E$ is just a complete graph. By what was said, the CNF $\left(\bigvee_{u \in S_{1}} x_{u}\right) \wedge \cdots \wedge\left(\bigvee_{u \in S_{t}} x_{u}\right)$ must represent the graph $G \backslash E$. Hence, by Proposition 5.1, $\operatorname{cc}(\bar{G} \cup E)=\operatorname{cc}(\overline{G \backslash E})=\operatorname{cnf}(G \backslash E) \leq t$.

Proof of Theorem 6.1. Let now $t$ be the minimum size of a monotone $\Pi_{3}$ circuit representing $\bar{M}_{n}$. Then, by Lemma 6.3, it must be possible to add a set $E$ of $|E| \leq t^{2}$ edges to the matching $M_{n}$ so that the resulting graph $M_{n} \cup E$ can be covered by
at most $t$ cliques. At least one of these cliques, say $A \times B$, must contain at least $\left|M_{n}\right| / t=n / t$ edges of the matching $M_{n}$. But this means that

$$
|H \cap(A \times B)| \geq(n / t)^{2}-(n / t) .
$$

Together with $|E| \leq t^{2}$ this implies that $t$ must satisfy the inequality $(n / t)^{2}-(n / t) \leq$ $t^{2}$, that is, $t^{4} \geq n^{2}-t n$, which implies $t=\Omega(\sqrt{n})$.

Proof of Theorem 6.2. Let $t$ be the minimum size of a monotone $\Pi_{3}$ circuit representing a bipartite $n \times n$ Hadamard graph $H=H_{n}$. We may assume that $t \leq n / 16$, for otherwise there is nothing to prove. We will use the known fact that any Hadamard graph contains about the same number of edges and non-edges; in particular, both $|H|$ and $|\bar{H}|$ are at least $n^{2} / 4$.

By Lemma 6.3, there is a set $E$ of $|E| \leq t^{2}$ edges such that the graph $\bar{H} \cup E$ can be covered by at most $t$ cliques $R_{1}, \ldots, R_{t}$, that is, $\bar{H} \cup E=R_{1} \cup \cdots \cup R_{t}$. Let $N=|\bar{H}|$ be the total number of non-edges in $H$ (hence, $N \geq n^{2} / 4$ ) and take a clique $R \in\left\{R_{1}, \ldots, R_{t}\right\}$ containing the largest number of non-edges of $H$; hence, $N_{0}:=|R \cap \bar{H}| \geq N / t$. Let $N_{1}:=|R \cap H|$ be the number of edges of $H$ lying in $R$. Since $R \cap H$ can contain only edges from $E$, we have that $N_{1} \leq|E| \leq t^{2}$. On the other hand, by Lindsey's Lemma, $\left|N_{1}-N_{0}\right| \leq \sqrt{n|R|}$, implying that

$$
N_{1} \geq N_{0}-\sqrt{n|R|} .
$$

Remembering that

$$
N_{1}+N_{0}=|R| \geq \frac{N}{t} \geq \frac{n^{2}}{4 t} \geq 4 n
$$

we obtain

$$
2 N_{1} \geq|R|-\sqrt{n|R|}=|R|\left(1-\sqrt{\frac{n}{|R|}}\right) \geq \frac{N}{2 t}
$$

that is, $N_{1} \geq N /(4 t)$. Together with $N_{1} \leq t^{2}$, this implies that $t^{3} \geq N / 4$. Thus, $t$ must be at least $(N / 4)^{1 / 3} \geq\left(n^{2} / 16\right)^{1 / 3}=\Omega\left(n^{2 / 3}\right)$.

## 7 Quadratic functions of graphs

In oder to obtain high lower bounds on the non-monotone circuit complexity of boolean functions it would be enough, by the Magnification Lemma, to show that any monotone boolean function $f(X)$ representing a given graph $G$ requires large monotone circuits. That is, it is enough to deal with monotone circuits, but the lower bound must hold for all monotone boolean functions $f(X)$ representing $G$.

A natural monotone boolean function representing a given graph $G=(V, E)$ is the quadratic function $f_{G}$ defined by

$$
f_{G}(X)=\bigvee_{u v \in E} x_{u} x_{v}
$$

It is therefore useful (as the first step) to understand for what graphs these functions require large monotone circuits.

### 7.1 Quadratic functions and $\Sigma_{3}$ circuits

A complete star in a graph with $n$ vertices is a set of $n-1$ edges sharing one endpoint in common. If the graph is bipartite, then a complete star is a set of edges joining all vertices of one part with a fixed vertex of the other part. A graph is star-free if it contains no complete stars.

Theorem 7.1 ([20]). Let $G=(V, E)$ be a star-free graph of degree $d$, and let $F$ be a monotone $\Sigma_{3}$ circuit computing $f_{G}$. Then $F$ has at least $\sqrt{|E|} / d$ gates. If $F$ is a formula, then $F$ has at least $|E| / d^{2}$ gates.

Since every $\Sigma_{3}$ circuit is an OR of CNFs, the theorem is an easy consequence of the following lemma-we include its proof to demonstrate the kind of difficulties one faces when trying to obtain similar lower bounds for circuits recognising the graph $G$. Let $\operatorname{cnf}\left(f_{G}\right)$ denote the minimum length of (i.e., the number of clauses in) a monotone CNF computing $f_{G}$. A complete star in a graph with $n$ vertices is a set of $n-1$ edges sharing one endpoint in common.

Lemma 7.2. If $H$ is a star-free graph with $M$ edges and degree $d$, then $\operatorname{cnf}\left(f_{H}\right) \geq$ $M / d^{2}$.

Proof. Let $F$ be monotone CNF of length $t=\operatorname{cnf}\left(f_{H}\right)$ computing $f_{H}$. Since $H$ has no complete stars, this CNF must contain at least two clauses. Take any of these clauses $C=\bigvee_{u \in S} x_{u}$ and consider the shrinked CNF $F^{\prime}=F \backslash\{C\}$. Since $C$ must accept all edges of $H$, each of these edges must have at least one endpoint in $S$. But any one vertex in $S$ can be an endpoint of at most $d$ edges, implying that $|S| \geq M / d$.

Since $F$ is a shortest CNF computing $f_{H}$, the shrinked CNF $F^{\prime}$ must make an error, i.e. it must (wrongly) accept some independent set of $H$. That is, there must be an independent set $I$ such that every clause of $F^{\prime}$ contains a variable $x_{v}$ with $v \in I$. Since $F^{\prime}$ has only $t-1$ clauses, we may assume that $|I| \leq t-1$. This error must be corrected by the clause $C$, implying that every vertex $u \in S$ must be adjacent (in $H$ ) with at least one vertex in $I$, for otherwise $F$ would wrongly accept the independent set $I \cup\{u\}$ of $H$. Hence, at least one vertex $v \in I$ must have at least $|S| /|I| \geq M / t d$ neighbours in $H$. Since the degree of $v$ cannot exceed $d$, the desired lower bound $t \geq M / d^{2}$ follows.

Note that the main reason why such a simple argument does not work for $\operatorname{cnf}(H)$ is that in this last case the vertices of $S$ need not have neighbours in $I$ unless $I=\{v\}$ ( $F^{\prime}$ accepts a single vertex), in which case $v$ must have large degree (at least $|S|$ ). The only interesting errors made by $F^{\prime}$ are non-edges of $H$. Let $T$ be the set of endpoints of these non-edges. Observe that $T$ must be an independent set in $H$ : if
$u v$ and $u^{\prime} v^{\prime}$ are two non-edges accepted by $F^{\prime}$, then $u v^{\prime}$ cannot be an edge, because then we would be forced to include at least one of its endpoints $u$ and $v^{\prime}$ in $S$, and the original CNF $F$ would wrongly accept one of these two non-edges. Hence, $T$ is an independent set, and and it is enough that $S \cap T=\emptyset$ to correct all these errors. In particular, if $S=\bar{T}$ then the clause $C=\bigvee_{u \in S} x_{u}$ rejects all non-edges wrongly accepted by $F^{\prime}$, and accepts all edges.

### 7.2 Quadratic functions and boolean formulas

In this section we consider circuits of arbitrary depth with unbounded fanin AND and OR gates; inputs are literals (variables and their negations). The depth of a circuit is the length of a longest path from an input to the output gate. A formula is a circuit with all gates having fanout 1, i.e. the underlying graph in this case is just a tree. The length of a formula is the number of input literals.

Given a boolean function $f$ and a graph $G$, let $L(f)$ (resp., $L_{+}(f)$ ) be the minimum length of a formula (resp., monotone formula) computing $f$, and $L_{+}(G)$ the minimum length of a monotone formula representing $G$.

If $F$ is a formula computing the characteristic function $f$ (in $2 m$ variables) of a bipartite $n \times n$ graph $G$ (with $n=2^{m}$ ) then, by the Magnification Lemma, we can replace each input literal in $F$ by a monotone formula of length at most $2 n$ (computing the corresponding OR of variables) so that the resulting monotone formula recognises $G$. Thus,

$$
L(f) \geq L_{+}(G) /(2 n)
$$

Easy counting shows that $L_{+}(G)=\Omega\left(n^{2} / \log n\right)$ for most $n \times n$ graphs $G$. Pudlák, Rödl and Savický have proved in [32] that $L_{+}(G)=\Omega(n \log (n / a))$ for any $n \times n$ graph $G$ such that neither $G$ nor its complement contains a copy of $K_{a, a}$. But, so far, no explicit graph with $L_{+}(G)=\Omega\left(n \log ^{3} n\right)$ is known. Such a graph would improve the strongest currently known lower bound $\Omega\left(m^{3-o(1)}\right)$ on the (non-monotone) formula length of an explicit boolean function in $m$ variables [18].

The reason, why it is difficult to show that a given graph $G=(V, E)$ cannot be represented by a short (monotone!) formula $F$, is that we only know that the formula must behave correctly on the 2-element subsets of vertices: for all $S \subseteq V$ with $|S| \leq 2$

$$
\begin{equation*}
F(S)=0 \text { if and only if } S \text { is an independent set in } G . \tag{5}
\end{equation*}
$$

On larger sets the formula may output arbitrary values. In particular, it can accept independent sets of size $k \geq 3$.

In this section we look what happens if we require that the formula $F$ must reject independent sets only up to some size $k \geq 2$. That is, this time we require that (5) must hold for subsets $S \subseteq V$ of size $|S| \leq k$.

Note that the quadratic function $f_{G}(X)=\bigvee_{u v \in E} x_{u} x_{v}$ recognises all independent sets of $G$, but the corresponding formula has length $2|E|$. Can we essentially decrease
the length of the formula by relaxing this condition and requiring that it must reject only independent sets up to some size $k<n$ ? Using a rank-argument it can be shown that, for some graphs, this is not possible unless $k$ is smaller than two times the degree of $G$.

Theorem 7.3. Let $G=(V, E)$ be a triangle-free graph without 4-cycles and of degree $d$. Let $f$ be a monotone boolean function which accepts all edges and rejects all independent sets of $G$ of size at most $2 d$. Then any monotone circuit computing $f$ has depth at least $\log |E|-1$, and any monotone formula computing $f$ has length at least $|E| / 2$. In particular, $L_{+}\left(f_{G}\right)=\Theta(|E|)$.

Proof. We look at vertices as one element and edges as two element sets. For a vertex $y \in V$, let $I_{y}$ be the set of its neighbours. For an edge $y \in E$, let $I_{y}$ be the set of all its proper neighbours; that is, $v \in I_{y}$ precisely when $v \notin y$ and $v$ is adjacent with an endpoint of $y$. Since $G$ has no triangles and no 4 -cycles, the sets $I_{y}$ are independent sets of size at most $2 d$ and must be rejected by $f$; we will concentrate only on these independent sets.

Let $M$ be a matrix whose rows correspond to edges $x \in E$, columns to edges and vertices $y \in V \cup E$, and

$$
M_{x, y}=x \backslash I_{y} .
$$

A rectangle in $M$ is a submatrix $A \times B \subseteq M$ with the property that there is a vertex $v$ such that

$$
v \in x \backslash S_{y} \text { for all } x \in A \text { and } y \in B
$$

we call $v$ a common element of the rectangle. Let $\mathcal{R}$ be a smallest possible set of mutually disjoint rectangles covering the whole matrix $M$. It is well known that every monotone circuit computing $f$ has depth at least $\log |\mathcal{R}|$ (see [21]), and every monotone formula has length at least $|\mathcal{R}|$ (see [35]). Hence, it remains to prove that $|\mathcal{R}| \geq|G| / 2$.

To do this, re-fill the entries of $M$ with constants 0 and 1 by the following rule:

$$
\begin{equation*}
M_{x, y}=1 \text { if and only if } x \cap y \neq \emptyset \tag{6}
\end{equation*}
$$

Let $R=A^{\prime} \times B^{\prime}$ be a rectangle in $\mathcal{R}$, and let $v$ be its common element. Then $v \in x$ for all edges $x \in A^{\prime}$ and $v \notin I_{y}$ for all $y \in B^{\prime}$. Hence, for each $y \in B^{\prime}$, the corresponding column in $R$ is either the all- 1 column (if $v \in y$ ) or the all- 0 column (if $v \notin y$ ) because in this last case the second endpoint of $x$ cannot belong to $y$ (for otherwise, the first endpoint $v$ would belong to $I_{y}$ ). Thus, either the rectangle $R$ is monochromatic or we can split it into two monochromatic rectangles. This way we obtain a covering $\mathcal{R}^{\prime}$ of $M$ by at most $2|\mathcal{R}|$ mutually disjoint monochromatic rectangles. To estimate their number we use the rank argument. Let $\operatorname{rk}(M)$ stand for the rank of $M$ over $G F(2)$. Since the rectangles in $\mathcal{R}^{\prime}$ are mutually disjoint and have rank 1 , it follows that $\left|\mathcal{R}^{\prime}\right| \geq \operatorname{rk}(M)$. Hence, it remains to prove that $M$ has full row-rank over $G F(2)$.

Take an arbitrary subset $\emptyset \neq F \subseteq E$ of edges. We have to show that the rows of the submatrix $M_{F}$ of $M$ corresponding to the edges in $F$ cannot sum up to the all-0 row over $G F(2)$. If $F$ is not an even factor, that is, if the number of edges in $F$ containing some vertex $v$ is odd, then the column of $v$ in $M_{F}$ has an odd number of 1 's, and we are done. Hence, we may assume that $F$ is an even factor. Take an arbitrary edge $y=u v \in F$, and let $H \subseteq F$ be the set of edges in $F$ incident to at least one endpoint of $y$. Since both vertices $u$ and $v$ have even degree (in $F$ ), the edge $y$ has a nonempty intersection with an odd number of edges in $F$ : one intersection with itself and an even number of intersections with the edges in $H \backslash\{y\}$. Hence, the colum of $y$ in $M_{F}$ contains an odd number of 1's, as desired.

For the incidence $n \times n$ graph $P_{n}$ of a projective plane $P G(2, q)$ Theorem 7.3 yields
Corollary 7.4. $L_{+}\left(f_{P_{n}}\right)=\Theta\left(n^{3 / 2}\right)$.
Note that if we would only know that the formula must reject non-edges (independent sets of size 2) - the case interesting in the context of boolean functions - then the same rank argument with the matrix $M$ defined by the rule (6) would not work. In this case we would have that $M_{x, y}=1$ if and only if $|x \cap y|=1$ (edge and non-edge can share at most one vertex). That is, $M$ would be just a matrix of scalar products (over the reals) of the characteristic vectors of edges $x$ and non-edges $y$, and (even over the reals) the rank of $M$ would not exceed $n$.

### 7.3 Saturated graphs

If a circuit computes $f_{G}$ then it also represents the graph $G$. This holds for all graphs and all circuits. In general, however, the converse may not hold because the circuit representing a graph needs not to reject independent sets with more than two vertices. Hence, in general, lower bounds for circuits computing $f_{G}$ do not imply lower bounds for circuits representing $G$. Still, there are graphs for which this holds.

A natural way to force a circuit representing a graph $G$ to compute $f_{G}$ is to "kill of" all large independent subsets by including additional edges. This is a standard trick in boolean complexity to obtain so-called "slice functions."

For a bipartite graph $H \subseteq U \times W$, we define its saturated extension as a (nonbipartite) graph $G=(V, E)$ such that $V=U \cup W$, and $u v \in E$ iff either $u v \in H$ or both vertices $u, v$ lie in $U$ or in $W$. That is, the induced subgraphs of $G$ on $U$ as well as on $W$ are complete graphs.

Lemma 7.5. If $G$ is the saturated extension of a star-free bipartite graph, then every monotone boolean function representing $G$ coincides with $f_{G}$.

Proof. Let $G=(V, E)$ be a saturated extension of a bipartite graph $H \subseteq U \times W$. Suppose that $H$ has no complete stars, and let $f$ be a monotone boolean function representing $G$. Take an arbitrary subset $S \subseteq V$ of vertices. If $f_{G}(S)=1$ then $S$
contains both endpoints of some edge $u v \in E$. This edge must be accepted by $f$ and, since $f$ is monotone, $f(S)=1$. If $f_{G}(S)=0$ then $S$ is an independent set of $G$. But the only independent sets in $G$ are single vertices and non-edges of $H$. Hence, $f(S)=0$ because $f$ represents $H$ and $H$ contains no complete stars.

Note, however, that proving lower bounds on the monotone complexity of quadratic functions $f_{G}$ of saturated graphs is a difficult task. In particular, Theorem 7.1 and Theorem 7.3 both fail for such functions.

## 8 Open problems

In the context of this paper the most interesting problem remains to prove that some explicit bipartite $n \times n$ graph requires large monotone $\Sigma_{3}$ circuits. The difficult thing here is to prove this for an explicit graph-easy counting shows that $\Sigma_{3}(G)=\Omega(\sqrt{n})$ for almost all bipartite $n \times n$ graphs $G$. Explicit graphs with $\Sigma_{3}(G)=\Omega(\log n)$ are easy to find: we have already shown that such is, for example, the an $n$ to $n$ matching $M_{n}$. However, to obtain some important consequences in computational complexity, we need an explicit graph $G$ with $\Sigma_{3}(G) \geq n^{\varepsilon}$. If proved with $\varepsilon=\omega(1 / \sqrt{\log n})$, this would give the first explicit boolean function in $m$ variables requiring (non-monotone) $\Sigma_{3}$ circuit of size $2^{\omega(\sqrt{m})}$. If proved with $\varepsilon=(\log \log n)^{\omega(1)} / \log n$, this would give an explicit boolean function outside the second level of the communication complexity hierarchy introduced in [5]. If proved with $\varepsilon=\omega(1 / \log \log \log n)$, this would give the first super-linear lower bound for log-depth circuits, thus resolving a long-standing open question in computational complexity.

If a graph $G$ has a small monotone $\Sigma_{3}$ circuit, then some its dense subgraph $H$ must have a small CNF. We already know (see Proposition 5.1) that $\operatorname{cnf}(H)$ is the minimal number of independent sets covering all non-edges of $H$. Hence, in oder to show that $H$ cannot be represented by a small CNF one could try to show that no independent set of $H$ can cover too many non-edges. If the original graph is a good expander, then one could hope that also $H$ will have good enough expanding properties. As the initial graph with good expanding properties one could take, for example, the incidence graph $P_{n}$ of the projective plane $P G(2, q)$. As shown in [3], every set $X$ of vertices on one side of $P_{n}$ has at least $n-n^{3 / 2} /|X|$ neighbours on the other side. Hence, every independent set of $P_{n}$ can cover at most $O\left(n^{3 / 2}\right)$ nonedges, implying that $\operatorname{cnf}\left(P_{n}\right)=\Omega(\sqrt{n})$. However, removing edges may destroy the expansion property so that the remaining graph $H$ may contain large independent sets. For example, the adversary could remove all edges of $P_{n}$ lying in an $(n / c) \times(n / c)$ clique for a large enough constant $c>0$. Since no $m \times m C_{4}$-free graph can have more than $(1+o(1)) m^{3 / 2}$ edges [24], the resulting graph $H$ will still have a constant fraction of edges of $P_{n}$ but the corresponding independent set will already cover a constant fraction of all its non-edges. Hence, we need an argument allowing us to show that $\operatorname{cc}(\bar{H})$ is large even when some independent sets of $H$ are large.

Problem 8.1. Is there a constant $\varepsilon>0$ such that $\operatorname{cc}(\bar{H}) \geq D^{\varepsilon}$ for every subgraph $H$ of $P_{n}$ of average degree at least $D$ ?

We already know (Theorem 6.1) that the monotone $\Pi_{3}$ circuit size for an $n$ to $n$ matching $M_{n}$ is exponentially smaller than that for $\overline{M_{n}}$.

Problem 8.2. Exhibit an explicit bipartite graph $G$ for which $\Sigma_{3}(\bar{G})$ is exponentially larger than $\Sigma_{3}(G)$.

The following problem does not require a graph be explicitely given, and hence, may be apparently easier. Let $\mathcal{E}$ be the set of all bipartite $n \times n$ graphs $G$ with $\Sigma_{3}(G) \leq$ $\exp \left((\log \log n)^{c}\right)$ for some constant $c>0$. Let $\mathrm{co}-\mathcal{E}$ be the set of complements of graphs from $\mathcal{E}$.

Problem 8.3. Prove that $\mathcal{E} \neq \mathrm{co}-\mathcal{E}$.
This would separate the second level of the communication complexity hierarchy ([5]) and hence resolve a long-standing open question in communication complexity.

A somewhat more ambitious task is to prove non-trivial lower bounds on the minimum number of bits sent by the players on the worst case input in the following "edge/non-edge" game between two players, Alice and Bob: Alice gets an edge $x$ of $G$, Bob gets a non-edge $y$ of $G$, and their goal is to determine a vertex $v \in x \backslash y$; at the end of the game on input $(x, y)$ this vertex $v$ must be known to both players. The graph $G$ itself is known to players long before the game starts; hence, they may agree upon what characteristics of the graph they will use to encode the information about their inputs. It is clear that at least $\log _{2} n$ bits of communication are necessary (Bob must know the answer) and $2 \log _{2} n$ bits are enough (Alice can just send her entire edge).

As mentioned in Sect. 7.2, the number of communicated bits on the worst case input is at least the logarithm of the minimum number $R(G)$ of disjoint rectangles needed to cover the communication matrix $M$ of this game (see [21]). Rows of this matrix are labelled by edges and columns by non-edges of $G$; the $(x, y)$-th entry is $M_{x, y}=x \backslash y$. That is, each entry of $M$ is either a single vertex or a pair of adjacent vertices. As before, a rectangle in $M$ is a submatrix $A \times B \subseteq M$ with the property that there is a vertex $v$ such that $v \in x$ and $v \notin y$ for all $x \in A$ and $y \in B$.

Problem 8.4. Exhibit an $n$-vertex graph $G$ on which the edge/non-edge game needs at least $\log _{2} n+a \log \log n$ bits of communication for some $a \geq 2$, or equivalently, for which $R(G)=\Omega\left(n \log ^{a} n\right)$.

If proved with $a=2$ this would give a graph-theoretic proof of Khrapchenko's classical lower bound $\Omega\left(m^{2}\right)$ on the size of non-monotone formula on $m$ variables [22]. If proved with $a \geq 3$ this would improve the strongest currently known lower bound $\Omega\left(m^{3-o(1)}\right)$ due to Håstad [18].

Finally, let us recall yet another problem for graphs which would solve an old problem in circuit complexity. An $A C C$ circuit is a constant-depth circuit with unbounded fanin AND, OR and $\mathrm{MOD}_{k}$ gates for arbitrary positive integers $k$, where $\operatorname{MOD}_{k}\left(x_{1}, \ldots, x_{m}\right)=1$ iff $x_{1}+\ldots+x_{m}=0$ modulo $k$. Exponential lower bounds for such circuits are known only if one is allowed to use $\mathrm{MOD}_{k}$ gates for prime numbers $k$ $[36,37]$. However, the case of composite moduli $k$ - even the case of circuits with gates AND, OR and $\mathrm{MOD}_{6}$ - remains widely open.

Let $\alpha(G)$ denote the minimum number $t$ for which there exist $t$ bipartite complete graphs $H_{1}, \ldots, H_{t}$ and a subset $L \subseteq\{0,1, \ldots, t\}$ such that for every arc $u v$ of $G$, $u v \in E$ if and only if the number of the graphs $H_{i}$ that contain $u v$ is a member of $L$. Equivalently, $\alpha(G)$ is the minimal number $t$ for which there exists a subset $L \subseteq\{0,1, \ldots, t\}$ and an assignment $u \mapsto A_{u} \subseteq\{0,1, \ldots, t\}$ such that for every arc $u v$ of $G, u v \in E$ if and only if $\left|A_{u} \cap A_{v}\right| \in L$.

Problem 8.5. Exhibit a bipartite $n \times n$ graph $G$ with $\alpha(G)=\exp \left((\log \log n)^{\omega(1)}\right)$.
Together with the Magnification Lemma and the reduction [40, 6] of $A C C$ circuits to depth-2 circuits with a symmetric gate on the top, this would yield an exponential lower bound for $A C C$ circuits computing the characteristic function of $G$.

More combinatorial problems related to circuit complexity of boolean functions can be found in a survey paper of Pudlák and Rödl [31].

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## Appendix: Proof of the Magnification Lemma

In the lemma below, by a circuit we will mean an arbitrary computational model whose inputs are literals, i.e. variables $x_{i}^{1}=x_{i}$ and their negations $x_{i}^{0}=\bar{x}_{i}$. A boolean function $g$ is isolating if $g$ rejects the all- 0 vector $(0, \ldots, 0)$ and accepts all vectors containing precisely one 1 ; on other vectors the function can take arbitrary values. Since OR and Parity functions are isolating, the Magnification Lemma is a special case of Lemma 8.6 below.

Let $G \subseteq U \times W$ be a bipartite graph with $U=W=\{0,1\}^{m}$, and

$$
f\left(y_{1}, \ldots, y_{m}, z_{1}, \ldots, z_{m}\right)
$$

be its characteristic function; that is, $f(u v)=1$ iff $u v \in G$. Suppose we have a circuit $F$ computing $f$. A positive extension of $F$ has $2^{m+1}$ variables $\left\{x_{u}: u \in U\right\} \cup\left\{x_{v}: v \in\right.$ $V\}$, and is obtained from $F$ by replacing input literals $y_{i}^{a}$ and $z_{i}^{a}$ by functions

$$
Y_{i}^{a}=g\left(\left\{x_{u}: u \in U, u(i)=a\right\}\right) \text { and } Z_{i}^{a}=h\left(\left\{x_{v}: v \in V, v(i)=a\right\}\right)
$$

where $g$ and $h$ are arbitrary isolating functions, and $u(i)$ is the $i$-th bit of $u \in\{0,1\}^{m}$,

Lemma 8.6. Let $G \subseteq U \times W$ be a bipartite $n \times n$ graph. If a circuit $F$ computes the characteristic function of $G$, then every its positive extension $F^{+}$represents the graph $G$.

Proof. For an arc $u v \in U \times W$, let $X_{u, v}$ be the vector in $\{0,1\}^{U \cup V}$ with precisely two 1's in positions $u$ and $v$. Let $F$ be a circuit computing the characteristic function of $G$. Then $u v \in G$ iff $F(u v)=1$. Hence, it is enough to show that $F^{+}\left(X_{u, v}\right)=1$ iff $F(u v)=1$.

The only difference of the circuit $F^{+}$from $F$ is that instead of input literals it takes the corresponding isolating functions as inputs. Hence, it is enough to show that on an input vector $X_{u, v}$ these isolating functions output the same values as the corresponding literals do on the input vector $u v$. We show this only for $y$-literals (for $z$-literals the argument is the same).

Let $y_{i}^{a}$ be some input literal of $F$, and $u, v \in\{0,1\}^{m}$. By the definition, the function $Y_{i}^{a}=g\left(\left\{x_{u}: u \in U, u(i)=a\right\}\right)$ depends only on the variables $x_{u}$ corresponding to the left part $U$ of the bipartition such that $u(i)=a$. Each input of the form $X_{u, v}$ assigns precisely one 1 to these variables, and this 1 is in the position $x_{u}$. Hence, $Y_{i}^{a}\left(X_{u, v}\right)=1$ iff $Y_{i}^{a}$ depends on $x_{u}$ which can happen if and only if $u(i)=a$. On the other hand, we also have that $y_{i}^{a}(u v)=1$ if and only if $u(i)=a$. Thus, $Y_{i}^{a}\left(X_{u, v}\right)=y_{i}^{a}(u v)$, and we are done.

Note that Lemma 8.6 holds not only for bipartite graphs but also for arbitrary $k$-partite hypergraphs $G \subseteq V_{1} \times V_{2} \times \cdots \times V_{k}$ with $\left|V_{1}\right|=\left|V_{2}\right|=\ldots=\left|V_{k}\right|=n=2^{m}$. The only difference is that then the characteristic function $f$ of such a hypergraph has $k m$ instead of $2 m$ variables.

If we use OR gates as isolating functions in Lemma 8.6 then the only we need to obtain a positive extension $F^{+}$of $F$ is to simltineausly compute $4 m=4 \log n$ boolean sums. Pudlák, Rödl and Savický [32] have shown that, for any $k \geq 1$, any $r$ boolean sums built out of $n$ variables can be computed using at most $k n+k 2^{[r / k]+1}$ fanin- 2 AND and OR gates. Together with Lemma 8.6 this yields the following lower bound for fanin- 2 circuits.

Corollary 8.7. Let $f$ be the characteristic function of a bipartite $n \times n$ graph $G$. If $G$ cannot be represented by a monotone circuit of size $L$, then $f$ cannot be computed by a non-monotone circuit using fewer than $L-24 n$ gates.


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