Bipartite Perfect Matching is in quasi-NC

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November 9, 2015

Abstract

We show that the bipartite perfect matching problem is in quasi-NC$^2$. That is, it has uniform circuits of quasi-polynomial size and $O(\log^2 n)$ depth. Previously, only an exponential upper bound was known on the size of such circuits with poly-logarithmic depth.

We obtain our result by an almost complete derandomization of the famous Isolation Lemma when applied to yield an efficient randomized parallel algorithm for the bipartite perfect matching problem.

1 Introduction

The perfect matching problem has been widely studied in complexity theory. It has been of particular interest in the study of derandomization and parallelization. The perfect matching problem, PM, asks whether a given graph contains a perfect matching.

The problem has a polynomial-time algorithm due to Edmonds [Edm65]. However, its parallel complexity is still not completely resolved as of today. The problem can be solved by randomized efficient parallel algorithms due to Lovász [Lov79], i.e., it is in RNC, but it is not known whether randomness is necessary, i.e., whether it is in NC. The class NC represents the problems which have efficient parallel algorithms, i.e., they have uniform circuits of polynomial size and poly-logarithmic depth. For the perfect matching problem, nothing better than an exponential-size circuit was known, in the case of poly-logarithmic bounded depth.

The construction version of the problem, Search-PM, asks to construct a perfect matching in a graph if one exists. It is in RNC due to Karp et al. [Kuw86] and Mulmuley et al. [MVV87]. The latter algorithm applies the celebrated Isolation Lemma. Both algorithms work with a weight assignment on the edges of the graph. A weight assignment is called isolating for a graph $G$ if the minimum weight perfect matching in $G$ is unique, if one exists. Mulmuley et al. [MVV87] showed that given an isolating weight assignment with polynomially bounded integer weights for a graph $G$, then a perfect matching in $G$ can be constructed in NC. To get an isolating weight assignment they use randomization. This is where the Isolation Lemma comes into play.

∗Supported by DFG grant TH 472/4
Lemma 1.1 (Isolation Lemma [MVV87]). For a graph $G(V,E)$, let $w$ be a random weight assignment, where edges are assigned weights chosen uniformly and independently at random from $\{1,2,\ldots,2|E|\}$. Then $w$ is isolating with probability $\geq 1/2$.

Derandomizing this lemma means to construct such a weight assignment deterministically in NC. This remains a challenging open question. A general version of this lemma, which considers a family of sets and requires a unique minimum weight set, has also been studied. The general version is related to the polynomial identity testing problem and circuit lower bounds [AM08].

The Isolation Lemma has been derandomized for some special classes of graphs, e.g., planar bipartite graphs [DKR10 TV12], strongly chordal graphs [DK98], graphs with a small number of perfect matchings [GKS7 AHT07]. In this work, we make a significant step towards the derandomization of the Isolation Lemma for bipartite graphs. In Section 3 we construct an isolating weight assignment for these graphs with quasi-polynomially large weights. Previously, the only known deterministic construction was the trivial one that used exponentially large weights. As a consequence we get that for bipartite graphs, PM and Search-PM are in quasi-NC$^2$. That is, they can be solved by uniform Boolean circuits of quasi-polynomial size and quadratic logarithmic depth.

In Section 4 we show that our approach also gives an alternate NC-algorithm for Search-PM in bipartite planar graphs. This case already has known NC-algorithms [MN95 MV00 DKR10]. Our algorithm is in NC$^3$, while the previous best known upper bound is already NC$^2$ [MN95 DKR10].

Our result also gives an RNC-algorithm for PM in bipartite graphs which uses very few random bits. The original RNC-algorithm of Lovász [Lov79] uses $O(m \log n)$ random bits. This has been improved by Chari, Rohatgi, and Srinivasan [CRS95] to $O(n \log m/n)$ random bits. They actually construct an isolating weight assignment using these many random bits. To the best of our knowledge, the best upper bound today on the number of random bits is $(n + n \log m/n)$ by Chen and Kao [CK97], that is, the improvement to [CRS95] was to bring down the constant factor to two. In Section 5 we achieve an exponential step down to $O(\log^3 n)$ random bits. Note however, that we do not get an isolating weight assignment with polynomially bounded weights.

We give a short outline of the main ideas of our approach. For any two perfect matchings, the edges where they differ form disjoint cycles. For a cycle $C$, its circulation is defined to be the difference of weights of two perfect matchings which differ exactly on the edges of $C$. Datta et al. [DKR10] showed that a weight assignment which ensures nonzero circulation for every cycle is isolating. It is not clear if there exists such a weight assignment with small weights. Instead, we give nonzero circulations only to small cycles. Then, we consider the union of minimum weight perfect matchings. In the bipartite case, this union is significantly smaller than the original graph. In particular, at least half of its nodes have degree $\leq 2$. We keep repeating this procedure with weight assignments that give nonzero circulation to every cycle with a small number of degree $> 2$ nodes. In each round, the number of degree $> 2$ nodes decreases by half. In $\log n$ rounds, we get a unique minimum weight perfect matching.
2 Preliminaries

2.1 Matchings and Complexity

By \( G(V,E) \) we denote a graph with vertex set \( V \) and edge set \( E \). We consider only undirected graphs in this paper. A graph is bipartite if there exists a partition \( V = L \cup R \) of the vertices such that all edges are between vertices of \( L \) and \( R \). The girth of \( G \) is the length of the shortest cycle in \( G \). If \( G \) has no cycle, then the girth is defined to be infinite.

In a graph \( G(V,E) \), a matching \( M \subseteq E \) is a subset of edges with no two edges sharing an endpoint. A matching which covers every vertex is called a perfect matching. For any weight assignment \( w: E \to \mathbb{Z} \) on the edges of a graph, the weight of a matching \( M \) is defined to be the sum of weights of all the edges in \( M \), i.e., \( w(M) = \sum_{e \in M} w(e) \).

A weight function \( w \) is called isolating for \( G \) if there is a unique perfect matching of minimum weight in \( G \).

A graph \( G \) is matching-covered if each edge in \( G \) participates in some perfect matching. In the literature, matching-covered is also called 1-extendable and these notions require \( G \) to be connected. Note: in this paper, we use matching-covered also for non-connected graphs!

The perfect matching problem \( \text{PM} \) is to decide whether a given graph has a perfect matching. Its construction version \( \text{Search-PM} \) is to compute a perfect matching of a given graph, or to determine that no perfect matching exists.

Analogous to NC, Barrington [Bar92] defined the class quasi-NC\(^k \) as the class of problems which have uniform circuits of quasi-polynomial size \( 2^{O(1)n} \) and poly-logarithmic depth \( O(\log^k n) \). Here, uniformity means that local queries about the circuit can be answered in poly-logarithmic time (see [Bar92] for details). The class quasi-NC is the union of classes quasi-NC\(^k \), over all \( k \geq 0 \).

2.2 An RNC algorithm for Search-PM

Let us first recall the RNC algorithm of Mulmuley, Vazirani & Vazirani [MVV87] for the construction of a perfect matching (SEARCH-PM). Though the algorithm works for any graph, we will only consider bipartite graphs here.

Let \( G \) be a bipartite graph with vertex partitions \( L = \{u_1, u_2, \ldots, u_n\} \) and \( R = \{v_1, v_2, \ldots, v_n\} \), and weight function \( w \). Consider the following \( n \times n \) matrix \( A \) associated with \( G \),

\[
A(i,j) = \begin{cases} 
2^{w(e)}, & \text{if } e = (u_i, v_j) \in E, \\
0, & \text{otherwise.}
\end{cases}
\]

The algorithm in [MVV87] computes the determinant of \( A \). An easy argument shows that this determinant is the signed sum over all perfect matchings in \( G \):

\[
\det(A) = \sum_{\pi \in S_n} \text{sgn}(\pi) \prod_{i=1}^{n} A(i, \pi(i)) = \sum_{M \text{ perfect matching in } G} \text{sgn}(M) 2^{w(M)}
\]

Equation (2) holds because the product \( \prod_{i=1}^{n} A(i, \pi(i)) \) is nonzero if and only if the permutation \( \pi \) corresponds to a perfect matching. Here sgn(\( M \)) is the sign of the corresponding
permutation. If the graph \( G \) does not have a perfect matching, then clearly \( \det(A) = 0 \). However, even when the graph has perfect matchings, there can be cancellations due to \( \text{sgn}(M) \), and \( \det(A) \) may become zero. To avoid such cancellations, one needs to design the weight function \( w \) cleverly. In particular, if \( G \) has a perfect matching and \( w \) is isolating, then \( \det(A) \neq 0 \). This is because the term \( 2^{w(M)} \) corresponding to the minimum weight perfect matching cannot be canceled with other terms, which are strictly higher powers of 2.

Given an isolating weight assignment for \( G \), one can easily construct the minimum weight perfect matching in \( \text{NC} \). Let \( M^* \) be the unique minimum weight perfect matching in \( G \). For every edge \( e \in E \), compute the determinant of the matrix \( A_e \) associated with \( G - e \). If the term \( 2^{w(M^*)} \) disappears from \( \det(A_e) \), then \( e \in M^* \). Doing this in parallel for each edge, we can find all the edges in \( M^* \).

As already explained in the introduction, the Isolation Lemma delivers the isolating weight assignment with high probability. Moreover, the weights chosen by the Isolation Lemma are polynomially bounded. Therefore, the entries in matrix \( A \) have polynomially many bits. This suffices to compute the determinant in \( \text{NC}^2 \) [Ber84]. Hence, also the construction is in \( \text{NC}^2 \). Put together, this yields an \( \text{RNC} \)-algorithm for \text{SEARCH-PM}.

### 2.3 Matching Polytope

Matchings are also one of the well-studied objects in polyhedral combinatorics. Matchings have an associated polytope, called the \textit{perfect matching polytope}. We use some properties of this polytope to construct an isolating weight assignment. The perfect matching polytope also forms the basis of one of the \( \text{NC} \)-algorithms for bipartite planar matching [MV00].

The perfect matching polytope \( \text{PM}(G) \) of a graph \( G(V,E) \) with \( |E| = m \) edges is a polytope in the edge space, i.e., \( \text{PM}(G) \subseteq \mathbb{R}^m \). For any perfect matching \( M \) of \( G \), consider its incidence vector \( x^M = (x^M_e)_e \in \mathbb{R}^m \) given by

\[
x^M_e = \begin{cases} 
1, & \text{if } e \in M, \\
0, & \text{otherwise}.
\end{cases}
\]

This vector is referred as a \textit{perfect matching point} for any perfect matching \( M \). The \textit{perfect matching polytope} of a graph \( G \) is defined to be the convex hull of all its perfect matching points,

\[
\text{PM}(G) = \text{conv}\{ x^M | M \text{ is a perfect matching in } G \}.
\]

Any weight function \( w : E \to \mathbb{R} \) on the edges of a graph \( G \) can be naturally extended to \( \mathbb{R}^m \) as follows: for any \( x = (x_e)_e \in \mathbb{R}^m \), define

\[
w(x) = \sum_{e \in E} w(e) x_e.
\]

Clearly, for any matching \( M \), we have \( w(M) = w(x^M) \). In particular, let \( M^* \) be a perfect matching in \( G \) of minimum weight. Then

\[
w(M^*) = \min\{ w(x) | x \in \text{PM}(G) \}.
\]

The following lemma gives a simple description of the perfect matching polytope of a bipartite graph \( G \) which is well known, see for example [LP86].
Lemma 2.1. Let $G$ be a bipartite graph and $x = (x_e)_e \in \mathbb{R}^m$. Then $x \in \text{PM}(G)$ if and only if

\begin{align*}
\sum_{e \in \delta(v)} x_e &= 1 & v \in V, (3) \\
x_e &\geq 0 & e \in E, (4)
\end{align*}

where $\delta(v)$ denotes the set of edges incident on the vertex $v$.

It is easy to see that any perfect matching point will satisfy these two conditions. In fact, all perfect matching points are vertices of this polytope. The non-trivial part is to show that any point satisfying these two conditions is in the perfect matching polytope [LP86, Chapter 7]. For general graphs, the polytope described by (3) and (4) can have vertices which are not perfect matchings. Thus, the description does not capture the perfect matching polytope for general graphs.

2.4 Nice Cycles and Circulation

Let $G(V, E)$ be a graph with a perfect matching. A cycle $C$ in $G$ is a nice cycle, if the subgraph $G - C$ still has a perfect matching. In other words, a nice cycle can be obtained from the symmetric difference of two perfect matchings. Note that a nice cycle is always an even cycle.

For a weight assignment $w$ on the edges, the circulation $c_w(C)$ of an even length cycle $C = (v_1, v_2, \ldots, v_k)$ is defined as the alternating sum of the edge weights of $C$,

$$ c_w(C) = |w(v_1, v_2) - w(v_2, v_3) + w(v_3, v_4) - \cdots - w(v_k, v_1)|. $$

The definition is independent of the edge we start with because we take the absolute value of the alternating sum.

The circulation of nice cycles was one crucial ingredient of the isolation in bipartite planar graphs given by Datta et al. [DKR10]. An important property is as follows: Let $M_1, M_2$ be two perfect matchings such that their symmetric difference is precisely the nice cycle $C$. Then we have $c_w(C) = |w(M_1) - w(M_2)|$. That is, $M_1$ and $M_2$ have the same weight, if and only if $c_w(C) = 0$. Consequently, when all nice cycles have nonzero circulation, then the minimum perfect matching is unique.

Lemma 2.2 ([DKR10]). Let $G$ be a graph with a perfect matching, and let $w$ be a weight function such that all nice cycles in $G$ have nonzero circulation. Then the minimum perfect matching is unique. That is, $w$ is isolating.

We will construct an isolating weight function for bipartite graphs. However, our weight function will not necessarily have nonzero circulation on all nice cycles. We start out with a weight assignment which ensures nonzero circulations for a small set of cycles in a black-box way, i.e., without being able to compute the set efficiently. The following lemma describes a standard trick for this.

Lemma 2.3 ([CRS95]). Let $G$ be a graph with $n$ nodes. Then, for any number $s$, one can construct a set of $O(n^2 s)$ weight assignments with weights bounded by $O(n^2 s)$, such that for any set of $s$ cycles, one of the weight assignments gives nonzero circulation to each of the $s$ cycles.
Proof. Let us first assign exponentially large weights. Let $e_1, e_2, \ldots, e_m$ be some enumeration of the edges of $G$. Define a weight function $w$ by $w(e_i) = 2^{i-1}$, for $i = 1, 2, \ldots, m$. Then clearly every cycle has a nonzero circulation. However, we want to achieve this with small weights.

We consider the weight assignment modulo small numbers, i.e., the weight functions $\{w \mod j \mid 2 \leq j \leq t\}$ for some appropriately chosen $t$. We want to show that for any fixed set of $s$ cycles $\{C_1, C_2, \ldots, C_s\}$, one of these assignments will work, when $t$ is chosen large enough. That is, we want

$$\exists j \leq t \forall i \leq s : c_{w \mod j}(C_i) \neq 0.$$ 

This will be true provided

$$\exists j \leq t : \prod_{i=1}^{s} c_w(C_i) \neq 0 \pmod{j}.$$ 

In other words,

$$\lcm(2, 3, \ldots, t) \mid \prod_{i=1}^{s} c_w(C_i).$$ 

This can be achieved by setting $\lcm(2, 3, \ldots, t) > \prod_{i=1}^{s} c_w(C_i)$. The product $\prod_{i=1}^{s} c_w(C_i)$ is upper bounded by $2^{n^2 s}$. Furthermore, we have $\lcm(2, 3, \ldots, t) > 2^t$ for $t \geq 7$ (see [Nai82]). Thus, choosing $t = n^2 s$ suffices. Clearly, the weights are bounded by $t = n^2 s$. \qed

3 Isolation in Bipartite Graphs

In this section we present our main result, an almost efficient parallel algorithm for the perfect matching problem.

**Theorem 3.1.** For bipartite graphs, PM and Search-PM are in quasi-NC$^2$.

Our starting point is Lemma 2.2 which requires nonzero circulations for all nice cycles. Recall that the construction algorithm requires the weights to be polynomially bounded. As the number of nice cycles can be exponential in the number of nodes, even the existence of such a weight assignment is not immediately clear. Nonetheless, Datta et al. [DKR10] give a construction of such a weight assignment for bipartite planar graphs. For general bipartite graphs, this is still an open question.

Our approach is to work with a weight function which gives nonzero circulation to only small cycles. Lemma 2.3 describes a way to find such weights. The cost of this weight assignment is proportional to the number of small cycles. Further, it is a black-box construction in the sense that one does not need to know the set of cycles. It just gives a set of weight assignments such that one of them has the desired property.

3.1 Union of Minimum Weight Perfect Matchings

Let us assign a weight function for bipartite graph $G$ which gives nonzero circulation to all small cycles. Consider a new graph $G_1$ obtained by the union of minimum weight perfect matchings in $G$. Our hope is that $G_1$ is significantly smaller than the original graph $G$. Note
that it is not clear if one can efficiently construct $G_1$ from $G$. This is because the determinant of the bi-adjacency matrix with weights in equation (1) from Section 2.2 can still be zero. As we will see, we do not need to construct $G_1$, it is just used in the argument. Our final weight assignment will be completely black-box in this sense.

The following lemma is the main reason why our technique is restricted to bipartite graphs: it shows that the graph $G_1$ constructed from the minimum weight perfect matchings in $G$ contains no other perfect matchings than these. In Figure 1, we give an example that this does not hold in general graphs.

**Lemma 3.2.** Let $G(V,E)$ be a bipartite graph with weight function $w$. Let $E_1$ be the union of all minimum weight perfect matchings in $G$. Then every perfect matching in the graph $G_1(V,E_1)$ has the same weight – the minimum weight of any perfect matching in $G$.

**Proof.** We use the description of the perfect matching polytope for bipartite graphs from Lemma 2.1. Let $q$ be the weight of a minimum weight perfect matching in $G$. Recall that

$$
\min\{w(x) \mid x \in \text{PM}(G)\} = q.
$$

The intersection of $\text{PM}(G)$ with the hyperplane $H(x) = \{x \mid w(x) = q\}$ is a face $F$ of the polytope.

One can describe a face of a polytope by replacing some of the inequalities in the description of the polytope by equalities. For the perfect matching polytope, the inequalities are given by (4). Thus, for the face $F$ there exists a set $S \subseteq E$ such that for any $x = (x_e)_e$, we have $x \in F$ if and only if

$$
\begin{align*}
\sum_{e \in \delta(v)} x_e &= 1 \quad v \in V, \quad (5) \\
x_e &\geq 0 \quad e \in E \setminus S, \quad (6) \\
x_e &= 0 \quad e \in S. \quad (7)
\end{align*}
$$

Clearly, for any minimum weight perfect matching $M$, its matching point $x^M$ satisfies the above three conditions as it lies on the face $F$. In particular, equation (7) implies that $S \cap M = \emptyset$. It follows that

$$E_1 \cap S = \emptyset. \quad (8)$$

Now, consider any perfect matching $M'$ in the graph $G_1(V,E_1)$. By equation (8), we have $M' \cap S = \emptyset$. Hence its matching point $x^{M'}$ satisfies equation (7). Since $x^{M'}$ represents a perfect matching, it also satisfies (6) and (5). Thus, we have $x^{M'} \in F$. Hence, $w(M') = w(x^{M'}) = q$.

Next, we show that $G_1$ is significantly smaller than $G$. As all the perfect matchings in $G_1$ have the same weight, every nice cycle in $G_1$ has zero circulation. The following lemma proves an even stronger statement: every cycle in $G_1$ has zero circulation. It essentially comes from the fact that all cycles are in the linear span of nice cycles. Note that $G_1$ is matching-covered, as it is a union of perfect matchings.

**Lemma 3.3.** Let $H$ be a matching-covered bipartite graph. Let $w$ be a weight function such that every perfect matching in $H$ has the same weight. Then for each cycle $C$ in $H$, we have $c_w(C) = 0$. 

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Figure 1: A non-bipartite weighted graph where every edge is contained in a minimum perfect matching of weight 1. However, the graph also has a perfect matching of weight 3. That is, Lemma 3.2 does not hold for non-bipartite graphs.

Proof. Let the weight of each perfect matching be $q$. Any point $x$ in the perfect matching polytope $PM(H)$ is a convex combination of perfect matchings. Therefore its weight $w(x)$ must also be $q$.

Let $x_1, x_2, \ldots, x_t$ be all the perfect matching points of $H$, i.e., the corners of $PM(H)$. Consider the average point $x = \frac{x_1 + x_2 + \cdots + x_t}{t}$. Because each edge participates in a perfect matching, every coordinate of $x = (x_e)_e$ is nonzero.

Now, consider a cycle $C$ in $H$ with edges $(e_1, e_2, \ldots, e_p)$ in cyclic order. We show that when we move from point $x$ along the cycle $C$, we remain inside the perfect matching polytope. This technique of moving along the cycle has been used by Mahajan and Varadarajan [MV00].

To elaborate, consider a new point $y = (y_e)_e \in \mathbb{R}^m$ such that for all $e \in E$,

$$y_e = \begin{cases} 
  x_e + (-1)^i \varepsilon, & \text{if } e = e_i, \text{ for some } 1 \leq i \leq p, \\
  x_e, & \text{otherwise},
\end{cases}$$

for some $\varepsilon > 0$. We choose $\varepsilon$ to be small enough so that $y_e \geq 0$ for all $e \in E$. Clearly, the vector $x - y$ has nonzero coordinates only on cycle $C$ with alternating $\varepsilon$ and $-\varepsilon$. Hence,

$$|w(x - y)| = \varepsilon \cdot c_w(C).$$

We argue that $y$ fulfills the conditions of Lemma 2.1 and therefore also lies in the perfect matching polytope. Because $y_e \geq 0$ for all $e \in E$, it satisfies inequality (4) from Lemma 2.1. It remains to show that $y$ also satisfies

$$\sum_{e \in \delta(v)} y_e = 1 \quad v \in V. \quad (10)$$

To see this, let $v \in V$. We consider two cases:

1. $v \notin C$. Then $y_e = x_e$ for each edge $e \in \delta(v)$. Thus, we get (10) from equation (3) for $x$.

2. $v \in C$. Let $e_j$ and $e_{j+1}$ be the two edges from $C$ which are incident on $v$. By definition, $y_{e_j} = x_{e_j} + (-1)^j \varepsilon$ and $y_{e_{j+1}} = x_{e_{j+1}} + (-1)^{j+1} \varepsilon$. For any other edge $e \in \delta(v)$, we have $y_e = x_e$. Combining this with equation (3) for $x$, we get that $y$ satisfies (10) for $v$.

We conclude that $y$ lies in the polytope $PM(H)$. Hence, $w(y) = q$. It follows that $w(x - y) = w(x) - w(y) = q - q = 0$. From equation (9), we conclude that $c_w(C) = 0$. \qed
We apply Lemma 3.3 to graph $G_1$. Recall that by our weight function, each small cycle in $G$ has a nonzero circulation. Therefore $G_1$ has no small cycles. The following lemma states this in a slightly more general form which will be required later.

**Lemma 3.4.** Let $G$ be a bipartite graph with weight assignment $w$ such that some cycles $C_1, C_2, \ldots, C_t$ in $G$ have nonzero circulations. Then the graph $G_1$, obtained by the union of minimum weight perfect matchings in $G$, does not have any of the cycles $C_1, C_2, \ldots, C_t$.

We want to argue that the number of edges in $G_1$ has significantly decreased from that in $G$. We use the following result of Alon, Hoory, & Linial [AHL02] which states that the nodes of graphs with no small cycles have a small average degree.

**Theorem 3.5** ([AHL02]). Let $H$ be a graph with $n$ nodes, average degree $d \geq 2$ and girth $g$. Then,

$$n \geq 2(d - 1)^{g/2 - 1}.$$

The proof is fairly easy for graphs where each node has degree $\geq d$: do a breadth-first search of the graph starting from an arbitrary node until depth $g/2 - 1$. When one reaches a node $v$ via an edge $e$, there are $\geq d - 1$ edges incident on $v$ other than $e$. So, the search looks like a $(d - 1)$-ary tree of depth $g/2 - 1$. As there are no cycles of length $< g$, all the nodes in the tree should be distinct, which are $\geq 2(d - 1)^{g/2 - 1}$. This gives us the desired bound. Alon et al. generalized the argument to average degree $\geq d$.

If we put $g = 4 \log n - 2$ in Theorem 3.5 and take logarithms on both sides, we get

$$
\begin{align*}
\log n &\geq 2(\log n - 1) \log(d - 1) + 1 \\
1 &\geq 2\log(d - 1) \\
2 &\geq (d - 1)^2 \\
1 + \sqrt{2} &\geq d.
\end{align*}
$$

Hence, the length of small cycles in our weighting scheme is chosen to be $4 \log n - 2$. Then $G_1$ has average degree $< 2.5$.

**Corollary 3.6.** Let $H$ be a graph with girth $g \geq 4 \log n - 2$. Then $H$ has average degree $< 2.5$.

It follows that at least a constant fraction of the nodes in $G_1$ have degree $\leq 2$. Hence, one can say that the graph $G_1$ has significantly fewer edges than $G$. Now, the plan is to repeat this procedure for $G_1$ with a new weight function. However, $G_1$ has no small cycles. The standard cycle length counts the number of nodes in a cycle. The idea now is to count instead the number of degree $> 2$ nodes in a cycle. This can also be viewed as contracting degree 2 nodes with their neighbors and then considering again small cycles. When we give nonzero circulations to these cycles in $G_1$ and again take the union of minimum weight perfect matchings, these cycles disappear from $G_1$. This, in turn, means that the number of degree $> 2$ nodes further reduces by a constant fraction. We continue this for $O(\log n)$ rounds until all cycles disappear and a single perfect matching remains. Now, we put all the ingredients together and formally define our weight assignment.

### 3.2 Constructing the Weight Assignment

Let $G(V, E) = G_0$ be the given bipartite graph with $n$ nodes. Define $\ell = 4 \log n - 2$. Let $w_0$ be a weight function that gives nonzero circulation to all cycles containing $< \ell$ nodes of
degree > 2 in $G_0$. Let $k$ be the number of rounds for which we need to construct weight assignments. We will define subgraphs $G_i$ and weight assignments $w_i$, for $i = 1, 2, \ldots, k$, which will be obtained in successive rounds.

$G_i$: the union of minimum weight perfect matchings in $G_{i-1}$ according to weight $w_{i-1}$.

$w_i$: a weight function such that any cycle in $G_i$ with < $\ell$ nodes of degree > 2 has nonzero circulation.

Our final weight function $w$ will be a combination of $w_0, w_1, \ldots, w_k$. We combine them in a way that the weight assignment in a later round does not interfere with the order of perfect matchings given by earlier round weights. Let $B$ be a number greater than the weight of any edge under any of these weight assignments. Then, define

\[ w = w_0B^k + w_1B^{k-1} + \cdots + w_kB^0. \tag{11} \]

Clearly, the precedence decreases from $w_0$ to $w_k$. That is, for any two perfect matchings $M_1$ and $M_2$ in $G_0$, we have $w(M_1) < w(M_2)$, if and only if there exists an $0 \leq i \leq k$ such that

\[
\begin{align*}
    w_j(M_1) &= w_j(M_2), \quad \text{for } j < i, \\
    w_i(M_1) &< w_i(M_2).
\end{align*}
\]

The number $k$ of rounds is chosen to be the minimum number such that the number of degree > 2 nodes in the graph $G_k$ is less than $\ell$. We will see later that $k = O(\log n)$. First, we prove that the weight assignment $w$ is isolating in $G_0$.

**Lemma 3.7.** The weight assignment $w$ defined in (11) is isolating for $G_0$.

**Proof.** We split the proof in a sequence of claims. By the definition of the graphs $G_i$ and the weight functions $w_i$, we have:

**Claim 1.** For any $1 \leq i \leq k$, any two perfect matchings in $G_i$ have the same weight according to $w_j$, for all $j < i$.

We first prove the isolation in the last round.

**Claim 2.** $w$ is isolating in $G_k$.

**Proof.** As the number of nodes in $G_k$ with degree > 2 is < $\ell$, the weight assignment $w_k$ gives a nonzero circulation to each cycle in $G_k$. From Lemma 3.4, the union of minimum weight perfect matchings in $G_k$ has no cycles. In other words, it has a unique minimum weight perfect matching according to $w_k$.

From Claim 1, all perfect matchings in $G_k$ have equal weight according to $w_j$, for all $j < k$. Thus, $G_k$ has a unique minimum weight perfect matching according to $w$. This proves Claim 2.

Next we claim that the final weight $w$ of the perfect matchings is decreasing in subsequent rounds.

**Claim 3.** For any $1 \leq i \leq k$, let $M_1$ be a perfect matching in $G_i$ and $M_2$ be a perfect matching in $G_{i-1}$ which is not in $G_i$. Then $w(M_1) < w(M_2)$.
Proof. From the construction of $G_i$ and Lemma 3.2 it follows that

$$w_{i-1}(M_1) < w_{i-1}(M_2).$$

From Claim 1, we have $w_j(M_1) = w_j(M_2)$ for all $j < i - 1$. Due to the precedence of the weight assignments, we get that $w(M_1) < w(M_2)$. This proves Claim 3.

From Claim 3 we get that all the minimum weight perfect matchings in $G_0$ according to $w$ come from $G_k$. Putting this together with Claim 2, we get that $w$ is isolating in $G_0$. This proves the lemma.

Next we want to give a bound on the number $k$ of rounds we need. We show that in each round the number of nodes with degree $> 2$ decreases by half. For some $i \geq 1$, let $U \subseteq V$ be the set of degree $> 2$ nodes in $G_{i-1}$. Now, $w_{i-1}$ assigns nonzero circulation to every cycle in $G_{i-1}$ which has less than $\ell$ nodes from $U$. From Lemma 3.4, the union of minimum weight perfect matchings (i.e., $G_i$) has none of these cycles. Now, we want to show that at least half of the nodes in $U$ have degree $\leq 2$ in $G_i$.

**Lemma 3.8.** Let $G(V, E)$ be a graph with $n$ nodes. Let $U \subseteq V$ be its set of degree $> 2$ nodes. Let $G_1(V, E_1)$ be a matching-covered subgraph of $G$ such that any cycle in $G_1$ contains at least $\ell = 4\log n - 2$ nodes from $U$. Then at least half of the nodes in $U$ have degree $\leq 2$ in $G_1$.

**Proof.** Let $T = V - U$ be the set of nodes of degree $\leq 2$ in $G$. In the following, by *degree of a node* we mean its degree in $G_1$.

Observe first that any node of degree 1 in $G_1$ can only be connected to a node of degree 1. That is, they form a connected component in $G_1$ that consists of a single edge. This is because $G_1$ is matching-covered. We delete all degree 1 nodes from $G_1$.

For the degree 2 nodes in $T$, let us identify them with one of their two neighbors. In more detail, let $(u_0, u_1, u_2, \ldots, u_{p+1})$ be a path in $G_1$, for some $p \geq 1$, such that the nodes $u_0, u_{p+1}$ are in $U$ and the nodes $u_1, u_2, \ldots, u_p$ are from $T$ and have degree 2. Delete the nodes $u_1, u_2, \ldots, u_p$ and add an edge $(u_0, u_{p+1})$ as shown in Figure 2. Note that $u_0$ and $u_{p+1}$ cannot have another path with nodes only coming from $T$, because by our assumption, any cycle in $G_1$ has many nodes from $U$.

![Figure 2: A degree 2 node contracted with its neighbor.](image)

In summary, we deleted all nodes in $T$ and the nodes of degree 1 in $U$ from $G_1$. Let $G_1'$ be the resulting graph. Hence, the nodes in $G_1'$ are exactly those nodes in $U$ whose degree is $> 1$ in $G_1$. Note that the degree of any node in $G_1'$ is the same as in $G_1$, as one can see in Figure 2.

By the assumption of the lemma, any cycle in $G_1'$ has length $\geq \ell$. From Corollary 3.6 we have that $G_1'$ has average degree $< 2.5$. As $G_1'$ does not have any degree 1 nodes, it follows that at least half of its nodes have degree 2. This, in turn, means that at least half the nodes in $U$ have degree $\leq 2$ in $G_1$. 

\[\square\]
We apply Lemma 3.8 to $G_{i-1}$ and $G_i$.

**Corollary 3.9.** For each $1 \leq i \leq k$, the number of degree $> 2$ nodes in $G_i$ is at most half of that in $G_{i-1}$.

We want a value of $k$ such that $G_k$ has fewer than $\ell$ nodes of degree $> 2$. Clearly, $k = \log n - 1$ suffices. Now, the only thing that remains is to bound the values of the weights assigned. First we bound the number of cycles which have a small number of degree $> 2$ nodes.

**Lemma 3.10.** In a graph with $n$ nodes, the number of cycles with $< \ell$ nodes of degree $> 2$ is $\leq n \ell$.

**Proof.** We have $n$ choices for the first node of the cycle. If the first node has degree 2 then the cycle has to contain both its neighbors. Hence, there is a unique choice for the second node. Otherwise there are at most $n$ choices for the second node. In general, if the recently chosen node has degree 2, then there is just one choice for the next node. Otherwise there are at most $n$ choices. As there are fewer than $\ell$ nodes of degree $> 2$ in the cycle, the number of choices made will be bounded by $n \ell - 1$. Combined with the choices for the first node we get the bound of $n \ell$.

It remains to bound the number of weight assignments we need to try and the values of the weights. From Lemma 3.10 we get that $w_i$ needs to give nonzero circulation to at most $n^{4 \log n - 2}$ cycles for each $0 \leq i \leq k$. Hence, the set of weight assignments we get from Lemma 2.3 has size $O(n^{4 \log n})$ and the weights involved are bounded by $O(n^{4 \log n})$. For each weight assignment $w_i$, one needs to try all choices given by Lemma 2.3 as we do not know which one would work. This means, in total, we need to try $O(n^{4(k+1) \log n}) = O(n^{4 \log^2 n})$ weight assignments. Recall that the number $B$ is the highest weight assigned by any $w_i$. Hence, we have $B = O(n^{4 \log n})$. Therefore the weights in the assignment $\nu$ in equation (11) are bounded by $n^{O(\log^2 n)}$. In other words, the weights have $O(\log n)$ bits. Moreover, it is straightforward to see that these weight assignments can be constructed in quasi-NC$^1$ with circuit size $2^{O(\log^3 n)}$.

**Lemma 3.11.** In quasi-NC$^1$, one can construct a set of $O(n^{4 \log^2 n})$ weight functions $[n] \times [n] \to \mathbb{N}$, where the weights have $O(\log^3 n)$ bits, such that for any given bipartite graph with $2n$ nodes, one of the weight functions is isolating.

Up to this point, we can conclude that the existence of a perfect matching in bipartite graphs can be decided in quasi-NC$^2$. But we can even construct a perfect matching: recall the construction algorithm of Mulmuley et al. [MVV87] from Section 2.2. They compute the determinant of the bi-adjacency matrix that has entry $2^{w(e)}$ for edge $e$. As our weights have $O(\log^3 n)$ bits, the determinant entries have quasi-polynomial bits. This determinant can still be computed in parallel, but with quasi-polynomial ($2^{O(\log^3 n)}$) size circuits. In particular, it can be computed in quasi-NC$^2$ by the algorithm of Berkowitz [Ber84]. Thus, our construction algorithm is in quasi-NC$^2$ with circuit size $2^{O(\log^3 n)}$. This finishes the proof of Theorem 3.1.
4 Bipartite Planar Graphs

The Search-PM problem already has some known NC-algorithms in the case of bipartite planar graphs [MN95, MV00, DKR10]. The one by Mahajan and Varadarajan [MV00] is in $NC^3$, while the other two are in $NC^2$. Our approach from the previous section can be modified to give an alternate NC-algorithm for this case, in $NC^3$ however.

The weights in our scheme in Section 3.2 become quasi-polynomial for two reasons: (i) we want nonzero circulation for all cycles of length $< 4 \log n - 2$, and (ii) we need to combine the different weight functions from $\log n$ rounds using a different scale.

To solve the first problem we observe that for the planar case, it suffices to give nonzero circulation to constant length cycles. To see this, recall that Euler’s polyhedron formula gives a linear bound on the number of edges of a planar graph (see for example [Har69, Corollary 11.1(a)]).

Theorem 4.1. Let $H$ be a planar graph with $n \geq 3$ nodes, $m$ edges, and girth $g \leq n$. Then

$$m \leq \frac{g}{g-2}(n-2)$$

The case not covered by Theorem 4.1 is when $H$ has no cycles. Then we have $m \leq n - 1$. Recall that the average degree of a graph is $d = 2m/n$. By Theorem 4.1, we get

$$d = \frac{2m}{n} \leq \frac{2g}{g-2} \frac{n-2}{n} < \frac{2g}{g-2} \quad (12)$$

We want $d < 2.5$. By equation (12) this holds for $g \geq 10$. Note also that if a graph has no cycles, then the average degree is $< 2$. Hence, we get the following consequence for planar graphs, which is analogous to Corollary 3.6.

Corollary 4.2. Let $H$ be a planar graph with girth $g \geq 10$. Then $H$ has average degree $< 2.5$.

Since there are only polynomially many cycles of length $< 10$, one can ensure nonzero circulations for them with polynomially bounded weights by Lemma 2.3.

To solve the second problem, we use the fact that in planar graphs, one can count the number of perfect matchings of a given weight in $NC^2$ by the Pfaffian orientation technique [Kas67, Vaz89]. As a consequence, we can actually construct the graphs $G_i$ in each round in $NC^2$. Thereby we avoid having to combine the weight functions from different rounds.

In more detail, in the $i$-th round, we need to compute the union of minimum weight perfect matchings in $G_{i-1}$ according to $w_{i-1}$. For each edge $e$, we decide in parallel if deleting $e$ reduces the count of minimum weight perfect matchings. If yes, then edge $e$ should be present in $G_i$. As it takes $\log n$ rounds to reach a single perfect matching, the algorithm is in $NC^3$.

5 An RNC-Algorithm with Few Random Bits

We can also present our result for bipartite perfect matching in an alternate way. For PM, instead of quasi-NC, we can get an RNC circuit but with only poly-logarithmically many random bits.

Theorem 5.1. For bipartite graphs, there is an RNC-algorithm for PM which uses $O(\log^3 n)$ random bits.
The rest of this section is devoted to prove Theorem 5.1. In our algorithm from Section 3, there are two reasons that we need quasi-polynomially large circuits: (i) we need to try quasi-polynomially many different weight assignments and (ii) each weight assignment has quasi-polynomially large weights.

To solve the first problem, we modify Lemma 2.3 to get a random weight assignment which works with high probability.

**Lemma 5.2 ([CRS95, KS01]).** Let \( G \) be a graph with \( n \) nodes. there is a random weight assignment on the edges which uses \( O(\log ns) \) random bits and assigns weights with \( O(\log ns) \) bits such that for any set of \( s \) cycles, it gives nonzero circulation to each of the \( s \) cycles, with probability at least \( 1 - 1/n \).

**Proof.** We follow the construction of Lemma 2.3 and give exponential weights modulo small numbers. Here, we use only prime numbers as moduli. Recall the weight function \( w \) defined by \( w(e_i) = 2^{i-1} \). Let us choose a random number \( p \) among first \( t \) prime numbers. We take our random weight assignment to be \( w \mod p \). We want to show that with high probability this weight function gives nonzero circulation to every cycle in \( \{C_1, C_2, \ldots, C_s\} \). In other words, \( \prod_{i=1}^{s} c_w(C_i) \equiv 0 \pmod{p} \). As the product is bounded by \( 2^n ns \), it has at most \( n^2 s \) prime factors. Let us choose \( t = n^3 s \). This would mean that a random prime works with probability at least \( 1 - 1/n \). As the \( t \)-th prime can only be as large as \( 2^{t \log t} \), the weights are bounded by \( 2^{t \log t} = O(n^3 s \log ns) \), and hence have \( O(\log ns) \) bits. A random prime with \( O(\log ns) \) bits can be constructed using \( O(\log ns) \) random bits (see [KS01]).

Recall from Section 3.2 that for a bipartite graph \( G \) with \( n \) nodes, we had \( k = \log n \) rounds and constructed one weight function in each round. We do the same here, however, we use the random scheme from Lemma 5.2 to choose the weight functions \( w_0, w_1, \ldots, w_k \). The probability that all of them provide nonzero circulation on their respective set of cycles \( \geq 1 - (k + 1)/n \geq 1 - \log n/n \) using the union bound.

Now, instead of combining them to form a single weight assignment, we use a different variable for each weight assignment. We modify the construction of matrix \( A \) from Section 2.2. For variables \( x_0, x_1, \ldots, x_k \), define

\[
A(i,j) = \begin{cases} x_0^{w_0(e)} x_1^{w_1(e)} \cdots x_k^{w_k(e)}, & \text{if } e = (u_i, v_j) \in E, \\ 0, & \text{otherwise.} \end{cases}
\]

From arguments similar to those in Section 2.2 one can write

\[
\det(A) = \sum_{M \text{ perfect matching in } G} \text{sgn}(M) x_0^{w_0(M)} x_1^{w_1(M)} \cdots x_k^{w_k(M)},
\]

where \( \text{sgn}(M) \) is the sign of the corresponding permutation. From the construction of the weight assignments it follows that if the graph has a perfect matching then the lexicographically minimum term in \( \det(A) \) comes from a unique perfect matching. Thus, we get the following lemma.

**Lemma 5.3.** \( \det(A) \neq 0 \iff G \text{ has a perfect matching}. \)

Recall that by Lemma 3.10 each \( w_i \) needs to give nonzero circulation to \( O(n^{4\log n - 2}) \) cycles. Thus, the weights obtained by the scheme of Lemma 5.2 will use \( O(\log^2 n) \) bits.
Hence, the polynomial computed by \( \det(A) \) will have quasi-polynomially degree. To avoid this, we split the weights into \( \log n \) bit chunks of length \( r = O(\log n) \) each. For \( 0 \leq i \leq k \) and \( 0 \leq j \leq r \), define \( w_{ij} \)'s to be the unique weight functions which satisfy the following two conditions for each edge \( e \):

\[
\begin{align*}
w_{ij}(e) &< n, \\
w_i(e) &= \sum_{j=0}^{r} w_{ij}(e) n^j.
\end{align*}
\]

For variables \( \{y_{ij}\}_{i,j} \), define the matrix \( A' \) by

\[
A'(i,j) = \begin{cases} 
\prod_{i=0}^{k} \prod_{j=0}^{r} y_{ij}^{w_{ij}(e)}, & \text{if } e = (u_i, v_j) \in E, \\
0, & \text{otherwise}.
\end{cases}
\]

Clearly, substituting \( y_{ij} = x_i^{n^j} \) in \( A' \) gives us \( A \). Hence, we get the following.

**Lemma 5.4.** \( \det(A') \neq 0 \iff \det(A) \neq 0 \).

Now, \( \det(A') \) is a polynomial of individual degree \( < n^2 \) with \( O(\log^2 n) \) variables, as both \( k \) and \( r \) are \( O(\log n) \). To test if \( \det(A') \) is nonzero one can apply the standard randomized polynomial identity test [Sch80, Zip79, DL78]. That is, to plug-in random values for variables \( y_{ij} \), independently from \( \{1, 2, \ldots, n^3\} \). If \( \det(A') \neq 0 \), then the evaluation is nonzero with high probability.

**Number of random bits:** For a weight assignment \( w_i \), we need \( O(\log n s) \) random bits from Lemma 5.2 where \( s = O(n \log n) \) by Lemma 3.10. Thus, the number of random bits required for all \( w_i \)'s together is \( O(k \log^2 n) = O(\log^3 n) \). Finally, we need to plug-in \( \log n \) random bits for each \( y_{ij} \). This again requires \( O(\log^3 n) \) random bits.

**Complexity:** The weight construction involves taking exponential weights modulo small primes by Lemma 5.2. Primality testing can be done in \( \text{NC} \), as the numbers involved have \( O(\log^2 n) \) bits [Rab80]. Thus, the weight assignments can be constructed in \( \text{NC} \). Moreover, the determinant with polynomially bounded entries can be computed in \( \text{NC} \).

In summary, we get an \( \text{RNC} \)-algorithm that uses \( O(\log^3 n) \) random bits as claimed in Theorem 5.1.

**Discussion**

The major open question remains whether one can do isolation with polynomialsally bounded weights. Our construction requires quasi-polynomial weights for two reasons: (i) we work with \( O(\log n) \)-length cycles and (ii) it takes \( \log n \) rounds to reach a unique perfect matching and the graphs obtained in the successive rounds cannot be constructed. To get polynomially bounded weights one needs to circumvent both these problems.

For non-bipartite graphs, the isolation question is open even in the planar case. For this case, our approach fails in its first step: Lemma 3.2 no longer holds as demonstrated in Figure 1. Can one assign weights in a way which ensures that the union of minimum weight perfect matchings is significantly smaller than the original graph?

It needs to be investigated if our ideas can lead to isolation in other objects. For example, isolation of paths in a directed graph, which is related to the \( \text{NL} \) versus \( \text{UL} \) question.
Acknowledgements

We would like to thank Manindra Agrawal and Nitin Saxena for their constant encouragement and very helpful discussions. We thank Arpita Korwar for discussions on some techniques used in Section 5.

References


