

Bipartite Perfect Matching is in quasi-NC

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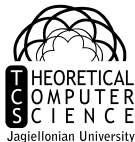


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Definition

Uniform circuit - The circuit where local queries about the circuit can be answered in poly-logarithmic time.

Definition

NC^k - is the class of decision problems decidable by uniform boolean circuits with a polynomial number of gates of at most two inputs and depth $O(\log^i n)$, or the class of decision problems solvable in time $O(\log^i n)$ on a parallel computer with a polynomial number of processors.

Definition

NC - the union of classes NC^k , over all $k \geq 0$.

Definition

quasi-NC^k - is the class of decision problems decidable by uniform boolean circuits with a quasi-polynomial $2^{\log^{O(1)} n}$ number of gates of at most two inputs and depth $O(\log^k n)$.

Definition

quasi-NC - the union of classes *quasi-NC^k*, over all $k \geq 0$.

Definition

Isolating weight function - function w is called isolating for G , if there is a unique perfect matching of minimum weight in G .

Lemma

Lemma 1.1 (*Isolation Lemma*) 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 $\{1, 2, \dots, 2|E|\}$. Then w is isolating with probability $\geq \frac{1}{2}$.

An RNC algorithm for Search-PM

Let L and R be vertex partitions of G , let w be a weight function of G . Consider the following $\frac{n}{2} \times \frac{n}{2}$ matrix A associated with G ,

$$A(i, j) = \begin{cases} 2^{w(e)}, & \text{if } e = (l_i, r_j) \in E, \\ 0, & \text{otherwise.} \end{cases}$$

The algorithm for SEARCH-PM computes the determinant of A . This determinant is the signed sum over all perfect matchings in G :

$$\begin{aligned} \det(A) &= \sum_{\pi \in S_{\frac{n}{2}}} \operatorname{sgn}(\pi) \prod_{i=1}^{\frac{n}{2}} A(i, \pi(i)) \\ &= \sum_{M \text{ pm in } G} \operatorname{sgn}(M) 2^{w(M)} \end{aligned}$$

If G does not have a perfect matching, then $\det(A) = 0$. But such result can be also an outcome of cancellations due to $\operatorname{sgn}(M)$. To avoid this situation, w needs to be designed correctly. In particular if G has a perfect matching and w is isolating, then $\det(a) \neq 0$ (since the term corresponding to the minimum weight perfect matching cannot be canceled).

Given an insulating weight assignment for G , it is possible to construct the minimum weight perfect matching in NC.

Let M^* be the unique minimum weight perfect matching in G .
 $w(M^*)$ is equal to the highest power of 2 dividing $\det(A)$.
For every edge $e \in E$ we can compute $\det(A_e)$, where A_e is matrix associated with $G - e$. If the highest power of 2 that divides $\det(A_e)$ is larger than $2^{w(M^*)}$, then $e \in M^*$. It can be done in parallel to find all edges of M^* .

The Matching Polytope

Definition

The perfect matching point of a graph G is a point in the edge space $(\mathbb{R}^{|E|})$. For any perfect matching M of G , consider its incidence vector $x^M = (x_e^M)_e \in \mathbb{R}^{|E|}$ given by

$$x_e^M = \begin{cases} 1, & \text{if } e \in M, \\ 0, & \text{otherwise.} \end{cases}$$

Definition

The perfect matching polytope $PM(G)$ of a graph G is a polytope in the edge space $(\mathbb{R}^{|E|})$. It is defined to be the convex hull of all its perfect matching points.

For $w: E \rightarrow R$ and $x = (x_e)_e \in \mathbb{R}^{|E|}$:

$$w(x) = \sum_{e \in E} w(e)x_e$$

$$w(M) = w(x^M)$$

Lemma

Lemma 2.1 *Let G be a bipartite graph and $x = (x_e)_e \in \mathbb{R}^{|E|}$. Then $x \in PM(G)$ if and only if*

$$\sum_{e \in \delta(v)} x_e = 1 \quad v \in V$$

and

$$x_e \geq 0 \quad e \in E$$

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

For general graphs, the polytope described by such conditions can have vertices which are not perfect matchings.

Definition

A cycle C in G is nice, if the subgraph $G - C$ still has a perfect matching. In other words, it can be obtained from the symmetric difference of two perfect matching. It is always an even cycle.

Definition

The circulation $c_w(C)$ of an even cycle C is the alternating sum of the edge weights of C ,

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

Lemma

Lemma 2.2 *Let G be a graph with a perfect matching, and let w be a weight function that all nice cycles in G have nonzero circulation. The minimum perfect matching is unique. That is, w is isolating.*

Lemma

Lemma 2.3 *Let G be a graph with n nodes. Then, for any number s , one can construct a set of $O(n^2s)$ weight assignments with weights bounded by $O(n^2s)$, such that for any set of s cycles, one of the assignments gives nonzero circulation to each of the s cycles.*

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- For any two $M \in PM(G)$ the edges where they differ form disjoint cycles.
- For a cycle C , $c_w(C)$ is defined to be the difference of weights of two perfect matchings which differ exactly on the edges of C .
- Lemma 2.2, but it is not clear if there exists such a weight assignment with small weights.
- We use a weight function that has nonzero circulations only for small cycles.

- We consider the subgraph G' , which is the union of minimum weight perfect matching in G .
- In bipartite case it is not only smaller, but also does not contain any small cycles.
- We show that if graph has no cycles of length $< r$, then the number of cycles of length $< 2r$ is polynomially bounded.
- For $\log n$ rounds: in the i -th round, assign weight which ensure nonzero circulations for all cycles with length $< 2^i$. Since the graph obtained after $(i - 1)$ -th rounds has no cycles of length $< 2^{i-1}$, the number of cycles of length $< 2^i$ is small.
- In $\log n$ rounds, we get a unique minimum weight perfect matching.

The union of Minimum Weight Perfect Matchings

Lemma

Lemma 3.2 *Let $G(V, E)$ be a bipartite graph with weight function w . Let C be a cycle in G such that $c_w(C) \neq 0$. Let E_1 be the union of all minimum weight perfect matchings in G . Then graph $G_1(V, E_1)$ does not contain cycle C .*

Proof. Let the weight of the minimum weight perfect matchings in G be q . Let $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_t$ be all the minimum weight perfect matching points of G , i.e., the corners of $\text{PM}(G)$ corresponding to the weight q . Consider the average point $\mathbf{x} \in \text{PM}(G)$ of these matching points,

$$\mathbf{x} = \frac{\mathbf{x}_1 + \mathbf{x}_2 + \dots + \mathbf{x}_t}{t}.$$

Clearly, $w(\mathbf{x}) = q$. Since each edge in E_1 participates in a minimum weight perfect matching, for $\mathbf{x} = (x_e)_e$, we have that $x_e \neq 0$ for all $e \in E_1$. Now, consider a cycle C with $c_w(C) \neq 0$. Let the edges of cycle C be (e_1, e_2, \dots, e_p) in cyclic order. For the sake of contradiction let us assume that all the edges of C lie in E_1 . We show that when we move from point \mathbf{x} along the cycle C , we reach a point in the perfect matching polytope with a weight smaller than q . This technique of moving along the cycle has been used by Mahajan and Varadarajan [MV00]. To elaborate, consider a new point $\mathbf{y} = (y_e)_e$ 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 $\varepsilon \neq 0$. Clearly, the vector $\mathbf{x} - \mathbf{y}$ has nonzero coordinates only on cycle C , where its entries are alternating ε and $-\varepsilon$. Hence,

$$w(\mathbf{x} - \mathbf{y}) = \pm \varepsilon \cdot c_w(C). \quad (5)$$

As $c_w(C) \neq 0$, we get $w(\mathbf{x} - \mathbf{y}) = w(\mathbf{x}) - w(\mathbf{y}) \neq 0$. We choose $\varepsilon \neq 0$ such that

- its sign is such that $w(\mathbf{y}) < w(\mathbf{x}) = q$, and

We argue that \mathbf{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 \mathbf{y} also satisfies

$$\sum_{e \in \delta(v)} y_e = 1 \quad v \in V. \quad (6)$$

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 (6) from equation (3) for \mathbf{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 \mathbf{x} , we get that \mathbf{y} satisfies (6) for v .

We conclude that \mathbf{y} lies in the polytope $\text{PM}(G)$. Since $w(\mathbf{y}) < q$, there must be a corner point of the polytope, which corresponds to a perfect matching in G with weight $< q$. This gives a contradiction. \square

Corollary

Corollary 3.3 *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 .*

Lemma

Lemma 3.4 *Let H be a graph with n nodes that has no cycles of length $\leq r$. Let $r' = 2r$ when r is even, and $r' = 2r - 2$ otherwise. Then H has $\leq n^4$ cycles of length $\leq r'$.*

Proof. Let $C = (v_0, v_1, \dots, v_{\ell-1})$ be a cycle of length $\ell \leq r'$ in G . Let $f = \ell/4$. We successively choose four nodes on C with distance $\leq \lceil f \rceil \leq r/2$ and *associate* them with C . We start with $u_0 = v_0$ and define $u_i = v_{\lceil if \rceil}$, for $i = 1, 2, 3$. Note that the distance between u_3 and u_0 is also $\leq \lceil f \rceil$. Since we could choose any node of C as starting point u_0 , the four nodes (u_0, u_1, u_2, u_3) associated with C are not uniquely defined. However, they uniquely describe C .

Claim 1. *Cycle C is the only cycle in H of length $\leq r'$ that is associated with (u_0, u_1, u_2, u_3) .*

Proof. Suppose $C' \neq C$ would be another such cycle. Let $p \neq p'$ be paths of C and C' , respectively, that connect the same u -nodes. Note that p and p' create a cycle in H of length at most

$$|p| + |p'| \leq \frac{r}{2} + \frac{r}{2} \leq r,$$

which is a contradiction. This proves the claim. □

There are $\leq n^4$ ways to choose 4 nodes and their order. By Claim 1, this gives a bound on the number of cycles of length $\leq r'$. □

Constructing the Weight Assignment

Let $G(V, E) = G_0$ be bipartite graph with n nodes that has perfect matchings. Define $k = \log n - 1$. Note that the shortest cycles have length 4. Define

w_i : a weight function such that all cycles in G_i of length $\leq 2^{i+2}$ have nonzero circulations.

G_{i+1} : the union of minimum weight perfect matchings in G_i according to weight w_i .

By the definition of G_i , any two perfect matchings in G_i have the same weight, not only according to w_i , but also to w_j for all $j < i$, for any $1 \leq i \leq k$.

By Lemma 3.2, graph G_i does not have any cycles of length $\leq 2^{i+1}$ for each $1 \leq i \leq k$. In particular, G_k does not have any cycles, since $2^{k+1} \geq n$. Therefore G_k has a unique perfect matching.

Final weight function w will be a combination of w_0, w_1, \dots, w_{k-1} . 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

$$w = w_0 B^{k-1} + w_1 B^{k-2} + \dots + w_{k-1} B^0.$$

In the definition of w , the precedence decreases from w_0 to w_{k-1} .

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 $0 \leq i \leq k-1$ such that

$$w_j(M_1) = w_j(M_2), \quad j < i,$$

$$w_i(M_1) < w_i(M_2).$$

The perfect matchings left in G_i have a strictly smaller weight with respect to w than the ones in G_{i-1} that did not make G_i .

Lemma

Lemma 3.5. *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. Since M_1 and M_2 are perfect matching in G_{i-1} , we have $w_j(M_1) = w_j(M_2)$, for all $j < i - 1$. From the definition of G_i and Corollary 3.3, it follows that $w_{i-1}(M_1) < w_{i-1}(M_2)$. Hence we get that $w(M_1) < w(M_2)$.

It follows that the unique perfect matching in G_k has a strictly smaller weight with respect to w than all other perfect matchings.

Corollary

Corollary 3.6. *The weight assignment*

$$w = w_0 B^{k-1} + w_1 B^{k-2} + \dots + w_{k-1} B^0$$

is isolating for G_0 .

It remains to bound the values of the weights assigned. In the first round we give nonzero circulation to all cycles of length 4. The number of such cycles is $\leq n^4$. In the i -th round, we have graph G_i that does not have any cycles of length $\leq 2^{i+1}$. For G_i , we give nonzero circulation to all cycles of length $\leq 2^{i+2}$. By Lemma 3.4, the number of such cycles is $\leq n^4$. Therefore, each w_i needs to give nonzero circulations to $\leq n^4$ cycles, for $0 \leq i < k$.

Now we apply Lemma 2.3 with $s = n^4$. This yields a set of $O(n^6)$ weight assignments with weights bounded by $O(n^6)$. Recall that the number B used in definition of w is the highest weight assigned by any w_i , so $B = O(n^6)$. Therefore the weights in the assignment w are bounded by $B^k = O(n^{6 \log n})$. That is, the weights have $O(\log^2 n)$ bits.

For each w_i we have $O(n^6)$ possibilities and we need to try all of them. In total, we need to try $O(n^{6k}) = O(n^{6 \log n})$ weight assignments in parallel. Every weight assignment can be constructed in quasi- NC^1 with circuit size $2^{O(\log^2 n)}$.

Lemma

Lemma 3.7. *In quasi-NC¹, one can construct a set of $O(n^{6 \log n})$ integer weight functions on $[\frac{n}{2}] \times [\frac{n}{2}]$, where the weights have $O(\log^2 n)$ bits, such that for any given bipartite graph with n nodes, one of the weight functions is isolating.*

With this construction of weight functions, we can decide the existence of a perfect matching in a bipartite graph in quasi- NC^2 as follows:

- Recall the bi-adjacency matrix A which has entry $2^w(e)$ for edge e .
- Compute $\det(A)$ for each of the constructed weight functions in parallel.
- If the given graph has a perfect matching, then one of the weight functions isolates a perfect matching (for this $\det(A)$ will be nonzero).
- When there is no perfect matching, then $\det(A)$ will be zero for any weight function.

Weights constructed in this way have $O(\log^2 n)$ bits, so the determinant entries have quasi-polynomial bits. The determinant can be computed in parallel, with circuits of quasi-polynomial size $2^{O(\log^2 n)}$. We need to compute $2^{O(\log^2 n)}$ -many determinants in parallel, so the algorithm is in quasi- NC^2 with circuit size $2^{O(\log^2 n)}$.

To construct a perfect matching we want to follow algorithm presented at the beginning with each of weight functions.

For a weight function w which is isolating, the algorithm outputs the unique minimum weight perfect matching M . If we have a weight function w' which is not isolating, still $\det(A)$ might be non-zero with respect to w' . Then the algorithm computes a set of edges M' that might or might not be a perfect matching. We can verify if M' is perfect matching, and in this case, we will output M' . As the algorithm involves computation of similar determinants as in the decision algorithm, it is in quasi- NC^2 with circuit size $2^{O(\log^2 n)}$.

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Theorem 4.1. *For bipartite graphs, there is an RNC^2 -algorithm for PM which uses $O(\log^2 n)$ random bits.*

To prove Theorem 4.1, consider 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. We show how to come down to polynomial bounds in both cases by using randomization.

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

Lemma 4.2 ([CRS95, KS01]). Let G be a graph with n nodes and $s \geq 1$. There is a random weight assignment w which uses $O(\log ns)$ random bits and assigns weights bounded by $O(n^3 s \log ns)$, i.e., with $O(\log ns)$ bits, such that for any set of s cycles, w 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 the first t prime numbers. We take our random weight assignment to be $w \bmod p$. We want to show that with high probability this weight function gives nonzero circulation to every cycle in $\{C_1, C_2, \dots, C_s\}$. In other words, $\prod_{i=1}^s c_w(C_i) \not\equiv 0 \pmod{p}$. As the product is bounded by $2^{n^2 s}$, 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 $2t \log t$, the weights are bounded by $2t \log t = O(n^3 s \log ns)$, 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]). \square

Recall from Section 3.2 that for a bipartite graph G with n nodes, we had $k = \lceil \log n \rceil - 1$ rounds and constructed one weight function in each round. We do the same here, however, we use the random scheme from Lemma 4.2 to choose each of the weight functions w_0, w_1, \dots, w_{k-1} independently. The probability that all of them provide nonzero circulation on their respective set of cycles $\geq 1 - k/n \geq 1 - \log n/n$ 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.

Let $L = \{u_1, u_2, \dots, u_{n/2}\}$ and $R = \{v_1, v_2, \dots, v_{n/2}\}$ be the vertex partition of G . For variables x_0, x_1, \dots, x_{k-1} , define an $n/2 \times n/2$ matrix A by

$$A(i, j) = \begin{cases} x_0^{w_0(e)} x_1^{w_1(e)} \dots x_{k-1}^{w_{k-1}(e)}, & \text{if } e = (u_i, v_j) \in E, \\ 0, & \text{otherwise.} \end{cases}$$

From arguments similar to those in Section [2.2](#), one can write

$$\det(A) = \sum_{M \text{ perfect matching in } G} \text{sgn}(M) x_0^{w_0(M)} x_1^{w_1(M)} \dots x_{k-1}^{w_{k-1}(M)},$$

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 $\det(A)$, with respect to the exponents of variables x_0, x_1, \dots, x_{k-1} in this precedence order, comes from a unique perfect matching. Thus, we get the following lemma.

Lemma 4.3. $\det(A) \neq 0 \iff G$ has a perfect matching.

Recall that each w_i needs to give nonzero circulations to n^4 cycles. Thus, the weights obtained by the scheme of Lemma 4.2 will be bounded by $O(n^7 \log n)$. This means the weight of a matching will be bounded by $O(n^8 \log n)$. Hence $\det(A)$ is a polynomial of individual degree $O(n^8 \log n)$ with $\log n$ variables. 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 x_i , independently from $\{1, 2, \dots, n^9\}$. 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 $O(\log ns)$ random bits from Lemma 4.2, where $s = n^4$. Thus, the number of random bits required for all w_i 's together is $O(k \log n) = O(\log^2 n)$. Finally, we need to plug in $O(\log n)$ random bits for each x_i . This again requires $O(\log^2 n)$ random bits.

Complexity: The weight construction involves taking exponential weights modulo small primes by Lemma 4.2. Primality testing can be done by the brute force algorithm in NC^2 , as the numbers involved have $O(\log n)$ bits. Thus, the weight assignments can be constructed in NC^2 . Moreover, the determinant with polynomially bounded entries can be computed in NC^2 [Ber84].

In summary, we get an RNC^2 -algorithm that uses $O(\log^2 n)$ random bits as claimed in Theorem 4.1.

Theorem 4.4. *For bipartite graphs, there is an RNC^3 -algorithm for SEARCH-PM which uses $O(\log^2 n)$ random bits.*

Let again $G(V, E)$ be the given bipartite graph with vertex partition $L = \{u_1, u_2, \dots, u_{n/2}\}$ and $R = \{v_1, v_2, \dots, v_{n/2}\}$. We construct the weight assignments w_0, w_1, \dots, w_{k-1} as in Lemma 4.2 in the randomized decision version. Let M^* be the unique minimum weight perfect matching in G with respect to the combined weight function w . Let $w_r(M^*) = w_r^*$, for $0 \leq r < k$.

Recall from Section 3.2 the sequence of subgraphs G_1, G_2, \dots, G_k of $G = G_0$, where G_{r+1} consists of the minimum perfect matchings of G_r according to weight w_r . In order to compute M^* , we would like to actually construct all the graphs G_1, G_2, \dots, G_k . However, it is not clear how to achieve this with $O(\log^2 n)$ random bits. Instead, we will construct a sequence of graphs H_1, H_2, \dots, H_k such that H_r will be a subgraph of G_r , for each $1 \leq r \leq k$. Furthermore, each H_r will contain the matching M^* . Recall that G_k consists of the unique perfect matching M^* . Hence, once we have $H_k = G_k$, we are done.

Let $H_0 = G$ and $0 \leq r < k$. We describe the r -th round. Suppose we have constructed the graph $H_r(V, E_r)$ and want to compute H_{r+1} . An edge will appear in H_{r+1} only if it participates in a matching M with $w_r(M) = w_r^*$. Thus, we will have that H_{r+1} is a subgraph of G_{r+1} . For an edge e , let $\mathbf{X}_r^{w(e)}$ denote the product

$$\mathbf{X}_r^{w(e)} = x_r^{w_r(e)} x_{r+1}^{w_{r+1}(e)} \cdots x_{k-1}^{w_{k-1}(e)}.$$

For a matching M , the term $\mathbf{X}_r^{w(M)}$ is defined similarly. Let $N(e)$ denote the set of edges which are neighbors of an edge e in G_r , i.e. all edges $e' \neq e$ that share an endpoint with e . For an edge $e \in E_r$, define the $n/2 \times n/2$ matrix A_e as

$$A_e(i, j) = \begin{cases} \mathbf{X}_r^{w(e')}, & \text{if } e' = (u_i, v_j) \in E_r - N(e), \\ 0, & \text{otherwise.} \end{cases}$$

Note that the matrix A_e has a zero entry for each neighboring edge of e . Thus, its determinant is a sum over all perfect matchings which contain e . That is,

$$\det(A_e) = \sum_{\substack{M \text{ pm in } H_r \\ e \in M}} \text{sgn}(M) \mathbf{X}_r^{w(M)}.$$

Consider the coefficient c_e of $x_r^{w_r^*}$ in $\det(A_e)$,

$$c_e = \sum_{\substack{M \text{ pm in } H_r \\ w_r(M) = w_r^*, e \in M}} \text{sgn}(M) \mathbf{X}_{r+1}^{w(M)}.$$

Define the graph H_{r+1} to be the union of all the edges e for which the polynomial $c_e \neq 0$. We claim that each edge of M^* appears in H_{r+1} . For any edge $e \in M^*$, the polynomial c_e will contain the term $\mathbf{X}_{r+1}^{w(M^*)}$. As the matching M^* is isolated in H_r with respect to the weight vector $(w_{r+1}, \dots, w_{k-1})$, the polynomial c_e is nonzero.

For the construction of H_{r+1} , we need to test if c_e is nonzero, for each edge e in H_r . As argued above in the decision part, the degree of c_e is $O(n^8 \log^2 n)$. We apply the standard zero-test, i.e., we plug in random values for the variables x_{r+1}, \dots, x_{k-1} independently from $\{1, 2, \dots, n^{11}\}$. The probability that the evaluation will be nonzero is at least $1 - O(\log^2 n/n^3)$. To compute this evaluation, we plug in values of x_{r+1}, \dots, x_{k-1} in $\det(A_e)$ and find the coefficient of $x_r^{w_r^*}$. This can be done in NC^2 [BCP84, Corollary 4.4]. For all the edges, we use the same random values for variables x_{r+1}, \dots, x_{k-1} in each identity test. The probability that the test works successfully for each edge is at least $1 - O(\log^2 n/n)$ by the union bound. We continue this for k rounds to find H_k , which is a perfect matching.

We need again $O(\log^2 n)$ random bits for the weight assignments w_0, w_1, \dots, w_{k-1} and the values for the x_i 's. Note that we use the same random bits for x_i in all k rounds. This decreases the success probability, which is now at least $1 - O(\log^3 n)/n$ by the union bound.

In NC^2 , we can construct the weight assignments and compute the determinants in each round. As we have $k = O(\log n)$ rounds, the overall complexity becomes NC^3 .

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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^3 -algorithm for this case.

The weights in our scheme in Section 3.2 become quasi-polynomial because we need to combine the different weight functions from $\log n$ rounds using a different scale. To solve this 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 .

Weighted perfect matchings and maximum matchings

A generalization of the perfect matching problem is the *weighted perfect matching problem* (WEIGHT-PM), where we are given a weighted graph, and we want to compute a perfect matching of minimum weight. There is no NC-reduction known from WEIGHT-PM to the perfect matching problem. However, the isolation technique works for this problem as well, when the weights are small integers. We put the given weights on a higher scale and put the weights constructed by our scheme in Section 3 on a lower scale. This ensures that a minimum weight perfect matching according to the combined weight function also has minimum weight according to the given weight assignment. Our scheme ensures that there is a unique minimum weight perfect matching. One can construct this perfect matching following the algorithm of Mulmuley et al. [MVV87] (Section 2.2).

Corollary 5.1. *For bipartite graphs, WEIGHT-PM with quasi-polynomially bounded integer weights is in quasi-NC².*

The maximum matching problem asks to find a maximum size matching in a given graph. It is well known that the maximum matching problem (MM) is NC-equivalent to the perfect matching problem (see for example [GKM13]). The equivalence holds for both decision versions and the construction versions. The reductions also preserve bipartiteness of the graph. Thus, we get the following corollary.

Corollary 5.2. *For bipartite graphs, MM is in quasi-NC².*