CONSTRUCTING A PERFECT MATCHING IS IN RANDOM NC

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Received 15 April 1985

We show that the problem of constructing a perfect matching in a graph is in the complexity class Random NC; i.e., the problem is solvable in polylog time by a randomized parallel algorithm using a polynomial-bounded number of processors. We also show that several related problems lie in Random NC. These include:

(i) Constructing a perfect matching of maximum weight in a graph whose edge weights are given in unary notation;

(ii) Constructing a maximum-cardinality matching;

(iii) Constructing a matching covering a set of vertices of maximum weight in a graph whose vertex weights are given in binary;

(iv) Constructing a maximum s-t flow in a directed graph whose edge weights are given in linear

1. Introduction

In this paper we show that the problem of constructing a perfect matching in a graph is in the complexity class Random NC; i.e., the problem is solvable in polylog time by a randomized parallel algorithm using a polynomial-bounded number of processors. We also show that several related problems lie in Random NC. These include:

(i) Constructing a perfect matching of maximum weight in a graph whose edge weights are given in unary notation;

(ii) Constructing a maximum-cardinality matching;

(iii) Constructing a matching covering a set of vertices of maximum weight in a graph whose vertex weights are given in binary;

(iv) Constructing a maximum s-t flow in a directed graph whose edge weights are given in unary.

Our results are based on a theorem of Tutte [15] showing that a graph has a perfect matching if and only if the determinant of a certain skew-symmetric matrix with indeterminates as elements is not identically zero. Around 1979 Lovász [11]

[†] Research supported by NSF Grant # DCR-8411954.

[‡] Research supported by a Weizmann Post-Doctoral fellowship, and by DARPA Grant N00039-83-C-1036.

^{††} Research supported in part by DARPA Grant N00039-82-C-0235.

AMS subject classification (1980): 68 E 10.

suggested that Tutte's Theorem, combined with a fundamental randomized technique for testing whether a matrix with polynomial entries has a nonzero determinant [13], provides a simple polynomial-time randomized algorithm for testing whether a graph has a perfect matching. In 1982 Borodin, von zur Gathen and Hopcroft [3] observed that, since the problem of computing the determinant of a numerical matrix lies in NC, a parallel algorithm can be constructed based on Lovász's approach. This algorithm establishes that the problem of deciding whether a graph contains a perfect matching lies in the complexity class Random NC, but leaves open the question of whether a parallel algorithm of comparable efficiency exists for constructing a perfect matching in a graph that is known to have such a matching. In 1984 Rabin and Vazirani [12], using clever algebraic techniques related to Tutte's Theorem, gave an attractive randomized polynomial-time sequential algorithm for constructing a maximum matching. They also observed that if a graph has a unique perfect matching then the problem of finding it lies in NC. The special case of finding a perfect matching in an interval graph was shown to be in Random-NC by Kozen, Vazirani and Vazirani [10].

The additional ingredient that allowed us to obtain a Random-NC parallel algorithm for constructing a perfect matching in an arbitrary graph was the introduction of a useful set function called Rank. Our result is based on a Random-NC reduction of the matching problem to the problem of computing the Rank function, and on a Random-NC algorithm for computing the Rank of sets.

2. The perfect matching algorithm

We present the perfect matching algorithm in three stages. First we reduce the problem of finding a perfect matching to the problem of identifying a large set of redundant edges in a graph that has a perfect matching and is not very sparse. We say that a set of edges in a graph that has a perfect matching is redundant if the removal of these edges results in a graph that still has a perfect matching. We then show that a random-NC procedure can construct a large set of redundant edges provided that a certain integer-valued function, called the rank, which is defined over all subsets of edges in the graph, is computable in random-NC. Finally we give a random-NC algorithm for computing the rank function.

2.1. A high level description of the algorithm

Let IG = (IV, IE) denote the input graph. The procedure Find-Perfect-Matching uses three main variables M, V and E.

The set variable M collects the edges that the procedure designates for the output matching. In contrast with the classical matching algorithms, our procedure never retreats from partial solutions. Once an edge is added to the set M it remains there until the procedure terminates, and it appears in the final output.

Throughout the execution of the procedure the variable V stores the set of vertices that are not yet covered by the matching M. The set E stores a subset of the set of edges connecting vertices in V; these are the edges that are still candidates for inclusion in the matching.

We start the procedure with *M* IG has a perfect matching then with holds throughout the execution of the

The graph G=(V,

If this condition is ever violated the pr in IG. On the other hand, if the cond rithm to consider edges in the set E, si in G is clearly a perfect matching in the set E.

Procedure Find-Perfect-Matching (V, while $|E| \ge 0$ do

if |E| < (3/4)|V| then [G is spars] find a set NM, $|NM| \ge (1/4)|V|$ $M \leftarrow M \cup NM$;

 $V \leftarrow V - \{\text{vertices covered by } E \leftarrow \{(v, u) | (v, u) \in E, v, u \in V\}$ else [G is not sparse]

Find-Redundant-Set(RE); [with probability $\beta > 0$, |RE|] $E \leftarrow E - RE$;

end.

The main while loop of the property Assuming that the graph (V, E) always set V is also empty, thus, that the variation

Each iteration of the while loo constant factor. The execution disting

Case a. |E| < (3/4)|V|, (the grafect matching, the degrees of all its vertices at least (1/2)|V| vertices have degree in this edge is included in can identify at least (1/4)|V| edges the graph (V', E'), defined by the vertices edges, has a perfect matching. Furthe computation can be done in O(1)

Case b. $|E| \ge (3/4)|V|$. This can is solved by the probabilistic procedure $1-o(1/|E|^2)$ the procedure produces has a perfect matching. With probability $\beta > 0$ the procedure factor.

The analysis of the number of a perfect matching requires a fact from

Fact 1. [1] Let X be a random variatindependent Bernoulli trials with proba $\leq (1-a) \, np = \exp(-1/2 \, a^2 \, np)$.

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edure the variable V stores the set of ing M. The set E stores a subset of the are the edges that are still candidates

We start the procedure with $M=\emptyset$, V=IV, and E=IE. If the input graph IG has a perfect matching then with probability 1-o(1) the following property holds throughout the execution of the procedure:

The graph G=(V,E) has a perfect matching.

If this condition is ever violated the procedure will fail to produce a perfect matching in IG. On the other hand, if the condition remains true then it suffices for the algorithm to consider edges in the set E, since the set M together with a perfect matching in G is clearly a perfect matching in the input graph IG.

```
Procedure Find-Perfect-Matching (V, E, M); while |E| \ge 0 do if |E| < (3/4)|V| then [G \text{ is sparse}] find a set NM, |NM| \ge (1/4)|V|, of edges that lie in every perfect matching; M \leftarrow M \cup NM; V \leftarrow V - \{\text{vertices covered by } NM\}; E \leftarrow \{(v, u)|(v, u) \in E, v, u \in V\}; else [G \text{ is not sparse}] Find-Redundant-Set(RE); [with probability \beta > 0, |RE| \ge \alpha |E|] E \leftarrow E - RE; end.
```

The main while loop of the procedure is executed until the set E is empty. Assuming that the graph (V, E) always has perfect matching, this implies that the set V is also empty, thus, that the variable M stores a perfect matching in IG.

Each iteration of the while loop tries to decrease the size of the set E by a constant factor. The execution distinguishes between two cases:

Case a. |E| < (3/4)|V|, (the graph G = (V, E) is sparse). Since G has a perfect matching, the degrees of all its vertices are at least 1. However, |E| < (3/4)|V|, thus at least (1/2)|V| vertices have degree exactly 1. If an edge is incident to a vertex with degree 1 this edge is included in any perfect matching in this graph. Thus we can identify at least (1/4)|V| edges that can be added to the set M and the remaining graph (V', E'), defined by the vertices in V that are not covered by the new matching edges, has a perfect matching. Furthermore, $|E'| \le |E| - (1/4)|V| \le (2/3)|E|$, and the computation can be done in O(1) steps using O(|E|) processors.

Case b. $|E| \ge (3/4)|V|$. This case, which is the novel part of the algorithm, is solved by the probabilistic procedure **Find-Redundant-Set**. With probability $1-o(1/|E|^2)$ the procedure produces a set RE such that the graph (V, E-RE) has a perfect matching. With probability $\beta > 0$, $|RE| \ge \alpha |E|$ for some $\alpha > 0$. Thus with probability $\beta > 0$ the procedure reduces the size of our problem by a constant factor.

The analysis of the number of iterations of the while loop required to create a perfect matching requires a fact from probability theory.

Fact 1. [1] Let X be a random variable distributed as the number of successes in n independent Bernoulli trials with probability of success p. If 0 < a < 1 then $P[X \le \le (1-a) \ np] < \exp(-1/2 \ a^2 \ np)$.

Lemma 2.1. If there are constants α , $\beta > 0$ such that with probability at least β a call to the procedure Find-Redundant-Set identifies at least a fraction α of the edges in E as redundant, then, uniformly for all problem instances, the expected number of iterations of the while loop within this procedure is $O(\log E)$, and, moreover, for some c>0 and d>0 the number of iterations of the while loop is bounded above by $d\log E$ with probability $1-O(E^{-c})$.

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Proof. We may assume without loss of generality that $\alpha \le 1/3$. Call an iteration of the while loop a success if it disposes of at least $\alpha |E|$ edges, either by determining that they lie in every perfect matching (Case a) or determining that they are redundant (Case b). Then every iteration in which Case a occurs is necessarily a success, and every iteration in which Case b occurs is a success with probability at least β . Let $\eta = (1-\alpha)^{-1}$. The number of successes required to reduce the number of edges to 1, starting from an initial edge set E, is at most $\log_n |E|$, and hence the number of successes required to complete the entire computation is at most $1 + \log_n |E|$. Since each iteration has probability of success at least β , independently of the outcomes of all previous iterations, the expected number of iterations is at most $\beta^{-1}(1+\log_n|E|)$. Moreover, the number of successes in the first $2\beta^{-1}\log_n|E|$ iterations is a random variable that stochastically dominates the number of successes in $2\beta^{-1}\log_{\eta}|E|$ independent Bornoulli trials with probability of success β . Applying Fact 1, we find that the probability of having $\log_n |E|$ or fewer successes in the first $2\beta^{-1}\log_{\eta}|E|$ iterations is at most $\exp(-(1/8)\log_{\eta}|E|) = |E|^{-c}$, where c == $(8 \ln \eta)^{-1} > 0$. Hence, with probability $1 - |E|^{-c}$ procedure Find-Perfect-Matching terminates within $2\beta^{-1} \log_{\eta} |E|$ iterations.

As we have mentioned before, the probability that a call to the procedure Find-Perfect-Matching fails to find a perfect matching is o(1). Since this bound on the failure probability is independent of the input, and sice we can check at the end of the procedure whether M is a perfect matching, we can simply repeat the procedure until a perfect matching is found. This leads to the following randomized algorithm, which finds a perfect matching in polylog expected time in any graph in which such a matching exists.

Perfect Matching Algorithm

```
M \leftarrow \emptyset;

while |M| < |IV|/2

V \leftarrow IV;

E \leftarrow IE;

M \leftarrow \emptyset

Find-Perfect-Matching (V, E, M);

end.
```

2.2. Identifying a large set of redundant edges

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Given a graph G=(V, E) such that 1. G has a perfect matching,
```

2. $E \ge 3/4|V|$,

and

we need to find a set $RE \subseteq E$ such that $1. |RE| \ge \alpha |E|$ for some $\alpha > 0$ and

2. (V, E-RE) has a perfect m

Our algorithm uses a powerful sets of edges in G.

Definition 2.1. Let PM denote the set $S \subseteq E$ define

Rank(S)

In words Rank(S) is an intege S that occur together in a perfect mate. We use the Rank function in the

Lemma 2.2. For a fixed $S \subseteq E$, let G = (V, E) has a perfect matching then

Proof. If Rank(S) = k then there is a If $Rank(S \cup \{e\}) = Rank(S)$ then (argument holds simultaneously for all *RE* leaves the graph with at least one

Example. Suppose the graph G is a c 4, 5, 6. Then PM consists of the two so $Rank(S) = Max(|S \cap \{1, 3, 5\}|, |S \cap \{2, Rank(S) = 2 \text{ and } RE = \{4, 6\}.$ Note destroy the perfect matching $\{1, 3, 5\}$.

To guarantee a uniform lower large enough, we use the following pro

Procedure Find-Redundant-Set;

begin

choose a random number i fr $\{1, ..., 5/6|E|\}$; choose a $S \subseteq E$ from the unif of E; $RE \leftarrow \{e \in E - S | Rank(S \cup \{e\}) =$

end.

Lemma 2.3. There are constants α , β >

Proof. Let |V|=n, |E|=m, and let p_i at random from the *i*—element subsets E-S, Rank $(S \cup \{e\})=$ Rank(S)+1.

We can think of a random set a dom choices (without repetitions) from

h that with probability at least β a call es at least a fraction α of the edges voblem instances, the expected number edure is $O(\log E)$, and, moreover, for s of the while loop is bounded above by

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obability that a call to the procedure t matching is o(1). Since this bound he input, and sice we can check at the transfer matching, we can simply repeat the this leads to the following randomized polylog expected time in any graph

we need to find a set $RE \subseteq E$ such that

1. $|RE| \ge \alpha |E|$ for some $\alpha > 0$

and

2. (V, E-RE) has a perfect matching.

Our algorithm uses a powerful integer-valued function defined over all subsets of edges in G.

Definition 2.1. Let PM denote the set of perfect matchings in the graph G. For any set $S \subseteq E$ define

$$\operatorname{Rank}(S) = \operatorname{Max}_{A \subset PM} |S \cap A|.$$

In words Rank(S) is an integer giving the maximum number of edges from S that occur together in a perfect matching of G.

We use the Rank function in the following way.

Lemma 2.2. For a fixed $S \subseteq E$, let $RE = \{e \in E - S | \text{Rank}(S \cup \{e\}) = \text{Rank}(S)\}$. If G = (V, E) has a perfect matching then so does G' = (V, E - RE).

Proof. If $\operatorname{Rank}(S) = k$ then there is a perfect matching A in G such that $|S \cap A| = k$. If $\operatorname{Rank}(S \cup \{e\}) = \operatorname{Rank}(S)$ then $(S \cup \{e\}) \cap A = S \cap A$ and e is not in A. This argument holds simultaneously for all the edges in RE; therefore removing the set RE leaves the graph with at least one perfect matching, the set A.

Example. Suppose the graph G is a cycle consisting of the successive edges 1, 2, 3, 4, 5, 6. Then PM consists of the two sets $\{1, 3, 5\}$ and $\{2, 4, 6\}$. Hence, for any set S, $Rank(S) = Max(|S \cap \{1, 3, 5\}|, |S \cap \{2, 4, 6\}|)$. Consider the set $S = \{1, 2, 3\}$. Then Rank(S) = 2 and $RE = \{4, 6\}$. Note that the deletion of edges 4 and 6 does not destroy the perfect matching $\{1, 3, 5\}$.

To guarantee a uniform lower bound on the probability that the set RE is large enough, we use the following procedure:

Procedure Find-Redundant-Set;

begin

choose a random number i from the uniform distribution over the range $\{1, ..., 5/6|E|\}$;

choose a $S \subseteq E$ from the uniform distribution over the *i*—element subsets of E;

 $RE \leftarrow \{e \in E - S | \operatorname{Rank}(S \cup \{e\}) = \operatorname{Rank}(S)\}$

end.

Lemma 2.3. There are constants α , $\beta > 0$ such that $Prob\{|RE| \ge \alpha |E|\} \ge \beta$.

Proof. Let |V| = n, |E| = m, and let p_i denote the probability that for a set S drawn at random from the i—element subsets of E, and for an edge drawn at random from E-S, $Rank(S \cup \{e\}) = Rank(S) + 1$.

We can think of a random set S as being constructed by a sequence of i random choices (without repetitions) from the set E. Rank(S) is then equal to the num-

of redundant edges

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new random edge was added to the set

$$=\sum_{j=0}^{i-1}p_j.$$

raph has a perfect matching, so we can

$$S(E) = \frac{n}{2} \le \frac{4}{6} m.$$

n compute a lower bound for the aveedure. The size i of the set S is chosen en a random set S of size i is chosen, is added to the set RE is $1-p_i$. There fore

$$\frac{1}{5} \left(\frac{5}{6} m - \sum_{i=1}^{(5/6)m} p_i \right) \ge \frac{1}{5} \frac{1}{6} m.$$

$$\mathsf{rob}\{|RE|=j\}) \le$$

 $n(\operatorname{Prob}\{|RE| \geq \mu m\}),$

$$\{|RE| \ge \mu m\},$$

$$\geq \frac{1}{60}$$
.

(|E|) steps using $h_1(|E|)$ processors' g functions, then Find-Redundant-Set $h_1(|E|)$ processors.

 $\operatorname{nk}(S)$ and of $\operatorname{Rank}(S \cup \{e\})$ for all $\operatorname{nk}(|E|)$ processors. The determination been done requires O(1) steps using

nk function

ive a solution for the case where the the general case is handled. raph with edge set E and n vertices ex set of one part and $V = \{v_1, v_2, ...\}$

..., v_n }, the vertex set of the other part. Associate with each edge $\{u_i, v_j\}$ an indeterminate x_{ij} , and let $B=(b_{ij})$ be an $n\times n$ matrix of indeterminates defined by the following rule:

 $b_{ij} = \begin{cases} x_{ij} & \text{if} \quad \{u_i, v_j\} \in E \\ 0 & \text{if} \quad \{u_i, v_i\} \notin E. \end{cases}$

Edmonds [5] has observed that G has a perfect matching if and only if $det(B) \not\equiv 0$. This is true because each of the n! terms in the sum

$$\sum_{\sigma \in S_{-}} \operatorname{sign}(\sigma) \Pi b_{i,\sigma(i)}$$

is a product of n entries in B that correspond to the edges of one of the n! perfect matchings in the complete bipartite graph. If all the edges of a perfect matching exist in G than the corresponding term in $\det(B)$ is not identically zero; otherwise one of the entries is zero and $\Pi b_{i,\sigma(i)}=0$. It is easy to verify that monomials can not cancel each other in this summation, and therefore $\det(B)\not\equiv 0$ if and only if G has a perfect matching.

To compute the Rank of a given set $S \in E$ we refine this argument. Define the matrix $B[S] = (b_{ij}^S)$ as follows:

$$b_{ij}^{S} = \begin{cases} yx_{ij} & \text{if } \{u_i, v_j\} \in S \\ x_{ij} & \text{if } \{u_i, v_j\} \in E - S \\ 0 & \text{if } \{u_i, v_j\} \notin E. \end{cases}$$

I.e. B[S] is derived from the matrix B by tagging all entries that correspond to edges in S with a new variable y.

Each non-zero monomial in det(B[S]) corresponds to a perfect matching in G. The degree in y of a monomial is equal to the number of elements from S that participate in the corresponding perfect matching. Again monomials can not cancel each other, hence the degree in y of det(B[S]) is equal to that of the monomial with the maximum degree in y, which in turn is equal to the maximum number of elements from S that participate in a perfect matching in G, i.e., to Rank(S).

So in order to compute Rank(S) we have to compute the degree in y of the multivariate polynomial det(B[S]). Rewriting det(B[S]) as a polynomial in y, we get

$$\det (B[S]) = \sum_{t=0}^{n} Q_{t} [\{x_{ij}\}] y^{t},$$

and $\operatorname{Rank}(S) = \operatorname{Max}\{t | Q_t \neq 0\}$. Unfortunately, we can not test directly whether Q_t , which is a polynomial in up to |E| indeterminates, is identically zero. Instead, we use a well-known probabilistic method.

Theorem 2.1. (Schwartz) [13] Let \tilde{Q}_t denote the value of the polynomial Q_t when each indeterminate in Q_t is replaced by a random integer in the range 1, ..., J. If $Q_t \neq 0$ then $\text{Prob}\{\tilde{Q}_t = 0\} \leq |E|/J$.

Let $\tilde{B}[S]$ denote the matrix B[S] after all its indeterminates, except y, have been replaced by random integers. To compute the \tilde{Q}_t 's we have to compute $\det(\tilde{B}[S])$ as a polynomial in one variable y.

Theorem 2.2. (Borodin, Cook, Pippenger) [2]. The determinant of an $n \times n$ matrix of polynomials with a constant number of variables and the degree of each matrix element bounded by n can be computed in $O(\log^2 n)$ steps using $O(n^{4.5})$ processors.

We can now summarize the algorithm for computing the Rank function in the bipartite case.

Procedure Rank(S);

begin

construct the matrix $B[S] = [b_{ij}^S]$; replace each indeterminate x_{ij} by a random integer in the range $1, ..., |IE|^4$; compute $\det(\tilde{B}[S]) = \sum_{i} \tilde{Q}_i y^i$;

Rank(S) = Max
$$\{t | \tilde{Q}_t \neq 0\}$$
;

In the case of a general graph our method of computing the rank function relies on the following theorem of Tutte.

Theorem 2.3. (Tutte) [15]. Let G=(V,E) denote a general graph with vertex set $V=\{1,2,...,n\}$ and define the skew-symmetric matrix $B=(b_{ij})$ as follows:

$$b_{ij} = \begin{cases} x_{ij} & \text{if} \quad \{i, j\} \in E \quad and \quad i < j \\ -x_{ij} & \text{if} \quad \{i, j\} \in E \quad and \quad i > j \\ 0 & \text{if} \quad \{i, j\} \notin E. \end{cases}$$

Then G has a perfect if and only if $det(B) \not\equiv 0$.

Tutte's theorem can be extended to yield an algorithm for computing the Rank function in general graphs.

Theorem 2.4. Let G = (V, E), $S \subseteq E$, and define the matrix $B[S] = [b_{ij}^s]$ as follows:

$$b_{ij}^{S} = \begin{cases} yx_{ij} & \text{if } \{i,j\} \in S & \text{and } i < j \\ -yx_{ij} & \text{if } \{i,j\} \in S & \text{and } i > j \\ x_{ij} & \text{if } \{i,j\} \in E - S & \text{and } i < j \\ -x_{ij} & \text{if } \{i,j\} \in E - S & \text{and } i > j \\ 0 & \text{if } \{i,j\} \notin E. \end{cases}$$

Then Rank(S) is equal to half of the degree in y of det(B[S]).

Proof. Let P be the set of all permutations σ of $\{1, 2, ..., n\}$ such that, for all i, $b_{i,\sigma(i)} \neq 0$. We classify these permutations according to their cycle structures. Since all elements on the main diagonal are 0, no permutation in P contains a cycle of length 1. Let OP be the set of all permutations in P which contain at least one odd cycle and let EP be the set of all permutations in S in which all cycles are of even length. Then P is the disjoint union of OP and EP. Let M be the set of all permutations in P such that every cycle is of length 2. Then P is the disjoint union of P and the permutations in P are in one-to-one correspondence with the perfect matchings in P.

We have

$$\det (B[S]) = \sum_{\sigma \in P} \operatorname{sign}(\sigma) \prod_{i} b_{i,\sigma(i)}.$$

We shall show that the contribution of (B[S]) is equal to zero; i.e., that

$$\sum_{\sigma \in OP} sign ($$

We shall partition the permutations co two permutations in each pair make a Let σ be a permutation containing at le of $\{1, 2, ..., n\}$ occurring in an odd cycle Then σ is paired with a permutation $\bar{\sigma}$ C is reversed. Thus $\bar{\sigma}$ is defined as folk $\bar{\sigma}(j) = \sigma^{-1}(j)$. It is easy to verify that each pair makes a net contribution

Call an ordered pair $\langle i, j \rangle$ an S—pairs correspond to the entries in Associated with each permutation σ in

is a nonzero monomial in the variables degree of y in this monomial. Then the in σ , i.e. the number of S—pairs $\langle i, \sigma(i) \rangle$

We shall show that, for every per $\tau \in M$ such that the y—degree of τ is good pairs and the even pairs, in such a if $\langle \sigma(i), \sigma(\sigma(i)) \rangle$ is an even pair. In on the traversal of any cycle of σ , odd pairs possible because all the cycles in σ that the set of odd pairs in σ contains even pairs in σ . If the y—degree of σ is d/2 S—pairs. Now define τ by the foll $\tau(i) = \sigma(i)$ and $\tau(\sigma(i)) = i$. Then τ lies

Each permutation in M correspondegree of this permutation is twice the ring. Hence, using the result proven in of any permutation is just twice the radet (B[S]) is twice the rank of S, we nee y—degree make a nonzero net contributer exists a permutation in M among mial is not cancelled by the monomial of

Thus, using the probabilistic me din, Cook and Pippenger, we have a Ra of a set of vertices in a general graph.

Lemma 2.5. 1. The procedure Rank is e

2. The probability that the procedure \mathbf{Ran} is bounded by $O(|IE|^{-3})$, and this even

]. The determinant of an $n \times n$ matrix riables and the degree of each matrix $\log^2 n$ steps using $O(n^{4,5})$ processors.

for computing the Rank function in

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lenote a general graph with vertex set is matrix $B = (b_{ij})$ as follows:

 $\in E$ and i < j $\in E$ and i > j

€E.

0.

yield an algorithm for computing the

the matrix $B[S] = [b_{ij}^s]$ as follows:

and i < j

and i > j

E-S and i < j

E-S and i>j

₹.

in y of det(B[S]).

or of $\{1, 2, ..., n\}$ such that, for all *i*, coording to their cycle structures. Since permutation in *P* contains a cycle of as in *P* which contain at least one odd as in *S* in which all cycles are of even d EP. Let M be the set of all permutations the perfect matchings in G.

$$(\sigma) \prod_{i} b_{i,\sigma(i)}.$$

We shall show that the contribution of the permutations containing odd cycles to det(B[S]) is equal to zero; i.e., that

$$\sum_{\sigma \in OP} \operatorname{sign}(\sigma) \prod_{i} b_{i,\sigma(i)} = 0.$$

We shall partition the permutations containing odd cycles into pairs, such that the two permutations in each pair make a net contribution of zero to the determinant. Let σ be a permutation containing at least one odd cycle. Let i be the least element of $\{1, 2, ..., n\}$ occurring in an odd cycle of σ , and let C be the odd cycle containing i. Then σ is paired with a permutation $\bar{\sigma}$ which is the same as σ except that the cycle C is reversed. Thus $\bar{\sigma}$ is defined as follows: if $j \notin C$ then $\bar{\sigma}(j) = \sigma(j)$; if $j \in C$ then $\bar{\sigma}(j) = \sigma^{-1}(j)$. It is easy to verify that this rule partitions OP into pairs, and that each pair makes a net contribution of zero to the determinant.

Call an ordered pair $\langle i,j \rangle$ an S—pair if $\{i,j\}$ is an edge in S. Then the S—pairs correspond to the entries in the matrix B[S] which involve variable y. Associated with each permutation σ in EP is the term sign $(\sigma) \prod b_{i,\sigma(i)}$. This term

is a nonzero monomial in the variables $\{x_{ij}\}$ and y. Define the y—degree of σ as the degree of y in this monomial. Then the y—degree of σ is just the number of S—pairs in σ , i.e. the number of S—pairs $\langle i, \sigma(i) \rangle$.

We shall show that, for every permutation $\sigma \in EP$ there exists a permutation $\tau \in M$ such that the y—degree of τ is greater than or equal to the y—degree of σ . To construct τ , partition the pairs $\langle i, \sigma(i) \rangle$, $i=1,2,\ldots,n$ into two sets, called the odd pairs and the even pairs, in such a way that, $\langle i, \sigma(i) \rangle$ is an odd pair if and only if $\langle \sigma(i), \sigma(\sigma(i)) \rangle$ is an even pair. In other words, the partition is chosen so that, in the traversal of any cycle of σ , odd pairs and even pairs alternate. Such a partition is possible because all the cycles in σ are even. Assume without loss of generality that the set of odd pairs in σ contains at least as many S—pairs as does the set of even pairs in σ . If the y—degree of σ is d, then the set of odd pairs contains at least d/2 S—pairs. Now define τ by the following rule: if $\langle i, \sigma(i) \rangle$ is an odd pair then $\tau(i) = \sigma(i)$ and $\tau(\sigma(i)) = i$. Then τ lies in M and the y—degree of τ is at least d.

Each permutation in M corresponds to a perfect matching in G, and the y—degree of this permutation is twice the number of edges from S in this perfect matching. Hence, using the result proven in the last paragraph, the maximum y—degree of any permutation is just twice the rank of S. To show that the degree of y in det(B[S]) is twice the rank of S, we need to show that the permutations of maximum y—degree make a nonzero net contribution to det(B[S]). But this is clear, since there exists a permutation in M among those of maximum y—degree, and its monomial is not cancelled by the monomial of any other permutation.

Thus, using the probabilistic method of Schwartz and the algorithm of Borodin, Cook and Pippenger, we have a Random-NC algorithm for computing the rank of a set of vertices in a general graph.

Lemma 2.5. 1. The procedure Rank is executed in $O(\log^2 |V|)$ steps using $O(|V|^{4.5})$ processors.

2. The probability that the procedure **Rank** fails to compute the correct value of Rank (S) is bounded by $O(|IE|^{-3})$, and this event does not depend on the input.

Combining now the results of Lemmas 2.1—2.5 we have

Theorem 2.5. For any input graph IG = (IV, IE):

- 1. The procedure Find-Perfect-Matching uses $O(|IV|^{6,5})$ processors and terminates within $O(\log^3 |IE|)$ steps with probability $1-|IE|^{-c}$ for some c>0.
- 2. The probability that the procedure fails to produce a perfect matching when applied to a graph that possesses one is bounded by $|IE|^{-1}$.

3. Further results

In this section we derive Random-NC algorithms for several further problems related to matching and network flows. We begin by giving such an algorithm for finding a perfect matching of maximum weight in an edge-weighted graph G = $=(IV, IE, \omega)$, when the edge-weights $\omega(\{i, j\})$ are given in unary.

Definition 3.1. Let MW denote the set of perfect matchings of maximum weight in the weighted graph $G=(V, E, \omega)$. For any set $S\subseteq E$ define

$$\operatorname{Rank}_{W}(S) = \operatorname{Max}_{A \in MW} |S \cap A|.$$

In words $Rank_{W}(S)$ is an integer giving the maximum number of edges from S that participate in a perfect matching of maximum weight in G.

It is easy to verify that running the procedure **Find-Perfect-Matching** with the new rank function computes the desired perfect matching in an expected number of iterations of order $\log |E|$. The only difficulty is to show that the new Rank_W function is computable in Random NC. The following theorem, in combination with Theorem 2.2, establishes this fact.

Theorem 3.1. Let $G = (V, E, \omega)$, $S \subseteq E$, and define the matrix B[S] as follows:

$$b_{ij}^{S} = \begin{cases} yz^{w}x_{ij} & \text{if } \{i,j\} \in S, & i < j, \text{ and } \omega(\{i,j\}) = w \\ -yz^{w}x_{ij} & \text{if } \{i,j\} \in S, & i > j, \text{ and } \omega(\{i,j\}) = w \\ z^{w}x_{ij} & \text{if } \{i,j\} \in E - S, & i < j, \text{ and } \omega(\{i,j\}) = w \\ -z^{w}x_{ij} & \text{if } \{i,j\} \in E - S, & i > j, \text{ and } \omega(\{i,j\}) = w \\ 0 & \text{if } \{i,j\} \notin E. \end{cases}$$

Let $det(B[S]) = \sum_{i} Q_i z^i$ and let $L = Max \{t | Q_t \neq 0\}$.

- 1. The the maximum weight of a perfect matching in G is L/2.
- 2. Rank_W(S) is equal to half the degree in y of Q_L , the coefficient of z^L .

The proof of Theorem 3.1 is quite similar to that of Theorem 2.4. First it is shown that the permutations containing odd cycles make a net contribution of zero to det (B[S]). Then attention is restricted to EP, the set of permutations with all cycles even that make a nonzero contribution to det(B[S]). Let L be the highest

degree to which z occurs in the monodet d be the highest degree of y that of It is shown that, among the permuta and of degree d in y, there is at least of mutation is shown to correspond to a from S; moreover, it is shown that this matchings of maximum weight, has a it is shown that the monomials of detations in EP make a nonzero net condet (B[S]) is of degree L in z, and the d in y.

In the following paragraphs we further matching and flow problems lie below can be performed in logspace.

- 1. Maximum cardinality matching cardinality matching in a graph G with to the problem of constructing a period G' with vertex set V in which each e that |V| is even. In this reduction G' $\{i,j\}$ of the complete graph receives otherwise. This reduction shows that lies in Random-NC. (A different reduction
- 2. Vertex weighted matching. The Random NC even when the weights In this problem we are given a graph of a positive weight w(v). We seek a material weight.

This problem can be approache from matroid theory that we require [16]. Call a set of vertices S independent vertices in S; then our goal is to constant I be the family of all independent matroid, the rank of a set of vertices from S that are covered by a matchinal matroid can be constructed by the foin order of decreasing weight; then, if Rank $\{\{v_1, v_2, ..., v_i\}\} > \text{Rank} (\{v_1, v_2, ..., v_i\}) > \text{Rank}$ matching is constructed by finding a part I by I, using the main algorithm of this

Thus, a Random-NC algorithm is at hand provided we can give a Ra of a set of vertices in this matroid. But reduced to that of determining a max with vertex set V whose edges are of w of generality that the given graph G graph K is the complete graph on vert

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Calgorithms for several further prob-We begin by giving such an algorithm weight in an edge-weighted graph G=}) are given in unary.

rfect matchings of maximum weight in et $S \subseteq E$ define

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the procedure Find-Perfect-Matching sired perfect matching in an expected only difficulty is to show that the new NC. The following theorem, in combi-

define the matrix B[S] as follows:

$$i < j$$
, and $\omega(\{i, j\}) = w$

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, and $\omega(\{i, j\}) = w$

$$i < j$$
, and $\omega(\{i, j\}) = w$
 $i > j$, and $\omega(\{i, j\}) = w$

 $Q_t \not\equiv 0$.

sing in
$$G$$
 is $L/2$.

y of Q_L , the coefficient of z^L .

ilar to that of Theorem 2.4. First it is eycles make a net contribution of zero EP, the set of permutations with all on to det(B[S]). Let L be the highest

degree to which z occurs in the monomial associated with such a permutation, and let d be the highest degree of y that occurs in a monomial that is of degree L in z. It is shown that, among the permutations whose monomials are of degree L in z and of degree d in y, there is at least one whose cycles are all of length 2. Such a permutation is shown to correspond to a matching of weight L/2 containing d/2 edges from S; moreover, it is shown that this matching is of maximum weight and, among matchings of maximum weight, has a maximum number of edges from S. Finally, it is shown that the monomials of degree L in z and d in y associated with permutations in EP make a nonzero net contribution to det(B[S]), so that the polynomial det(B[S]) is of degree L in z, and the coefficient of z in this polynomial is of degree d in y.

In the following paragraphs we use the technique of reducibility to show that further matching and flow problems lie in Random NC. All the reductions mentioned below can be performed in logspace.

- 1. Maximum cardinality matching. The problem of constructing a maximum cardinality matching in a graph G with vertex set V and edge set E is easily reduced to the problem of constructing a perfect matching of maximum weight in a graph G' with vertex set V in which each edge has weight zero or one. We can assume that |V| is even. In this reduction G' is the complete graph on vertex set V; edge $\{i,j\}$ of the complete graph receives weight one if $\{i,j\}$ lies in E, and weight zero otherwise. This reduction shows that the maximum cardinality matching problem lies in Random-NC. (A different reduction for this problem is given in [12]).
- **2. Vertex weighted matching.** The vertex-weighted matching problem is in Random NC even when the weights of the vertices are given in binary notation. In this problem we are given a graph G with vertex set V in which each vertex v has a positive weight w(v). We seek a matching that covers a set of vertices of maximum total weight.

This problem can be approached with the help of matroid theory. The results from matroid theory that we require can be found in the comprehensive reference [16]. Call a set of vertices S independent if there is a matching that covers all the vertices in S; then our goal is to construct an independent set of maximum weight. Let I be the family of all independent sets. The structure (V, I) is a matroid. In this matroid, the rank of a set of vertices S is just the maximum number of vertices from S that are covered by a matching. A maximum-weight independent set T in a matroid can be constructed by the following rule: let the elements be $v_1, v_2, \ldots, v_{|V|}$ in order of decreasing weight; then, for $i=1, 2, \ldots, |V|$, v_i lies in T if and only if Rank $(\{v_1, v_2, \ldots, v_i\}) > \text{Rank} (\{v_1, v_2, \ldots, v_{i-1}\})$. Once T is known, the desired matching is constructed by finding a perfect matching in the subgraph of G induced by T, using the main algorithm of this paper.

Thus, a Random-NC algorithm for the vertex-weighted matching problem is at hand provided we can give a Random-NC algorithm for computing the rank of a set of vertices in this matroid. But the problem of computing Rank(S) is easily reduced to that of determining a maximum-weight perfect matching in a graph K with vertex set V whose edges are of weight 0, 1 and 2. We can assume without loss of generality that the given graph G = (V, E) has an even number of vertices. The graph K is the complete graph on vertex set V. If edge $\{i, j\}$ does not lie in E then

 $\{i, j\}$ is given weight 0; if $\{i, j\}$ lies in E, then the weight of $\{i, j\}$ is $|S \cap \{i, j\}|$. Clearly, the maximum weight of a perfect matching in K is the rank of S. Thus we have shown that the vertex-weighted matching problem lies in Random NC.

3. Network flow. First, consider the problem of constructing a maximum s-tflow in a directed graph in which each edge has capacity 1. There is a classical reduction of this problem to the problem of constructing a maximum matching in a bipartite graph, as follows. Let the flow network be G = (V, E), with source s and sink t. We may assume that s has in-degree 0 and t has out-degree 0. The reduction constructs a bipartite graph H with bipartition (V_1, V_2) . Each part of the bipartition is a copy of the edge set of G. Thus, $V_1 = \{(e, 1) | e \in E\}$ and $V_2 = \{(e, 2) | e \in E\}$. If the head of edge e is also the tail of edge f (i.e., e=(i,j) and f=(j,k) for some i, j and k) then H contains and edge between (e, 1) and (f, 2). If an edge e in G is incident with neither s nor t, then H has an edge from (e,1) to (e,2). Then a maximum matching in H yields a maximum flow in G according to the following rule: edge e carries a flow of 1 if and only if (e, 1) is matched with some vertex (f, 2), where $e \neq f$, or (e, 2) is matched with some vertex (f, 1), where $e \neq f$. The reduction just given extends easily to the case in which the flow network has edges with integer capacities, provided these capacities are given in unary. The idea is to replace each edge $\{i, j\}$, of capacity c, with c parallel edges from i to j, each of capacity 1. All capacities in the resulting network are 1, and thus the reduction to bipartite matching applies. Thus, we have shown that the following problem is in Random NC: construct a maximum s-t flow in a directed flow network whose edge capacities are given in unary.

We summarize the constructions and reductions given in this section by a

theorem.

Theorem 3.2. The following problems lie in Random-NC:

(i) Constructing a perfect matching of maximum weight in a graph whose edge weights are given in unary;

(ii) Constructing a maximum matching;

(iii) Constructing a matching covering a set of vertices of maximum weight in a graph whose vertex weights are given in binary;

(iv) Constructing a maximum s-t flow in a directed graph whose edge weights are given in unary.

Our result about network flows stands in interesting contrast to the following result due to Goldschlager, Shaw and Staples [7]: the problem of constructing a maximum s-t flow in a directed flow network with edge capacities given in binary is complete in P with respect to logspace reductions. Since it is generally believed that such complete problems do not lie in Random NC, it appears that the parallel complexity of the max-flow problem depends critically on whether the capacities are given in unary or in binary. Nevertheless, the following result can be given:

Theorem 3.3. There is a randomized parallel algorithm to construct a maximum s-t flow in a directed network whose edge weights are given in binary, such that the number of processors used is bounded by a polynomial in the number of vertices, and the time used is $O((\log V)^k \log C)$, where C is the largest capacity of any edge and k is a constant.

This result is proved by combin the Edmonds—Karp scaling technique problem in Random NC, since log C is tion of the number of bits needed to exp

4. Di

Each of the randomized algorithm of giving an erroneous result; for e may fail to produce a perfect matching it its failure does not indicate with certain the algorithm many times in parallel ca ponentially low level, but can never ergiven a Random NC algorithm for the of the matching problem. As Karloff pour algorithms to yield algorithms for the mum matching problem which run in procrect result; i.e., Las Vegas algorithms

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algorithm to construct a maximum ights are given in binary, such that the vnomial in the number of vertices, and is the largest capacity of any edge and γ

This result is proved by combining the methods of the present paper with the Edmonds-Karp scaling technique [6]. Of course the result does not place the problem in Random NC, since log C is a linear, rather than polylogarithmic, function of the number of bits needed to express C in binary.

4. Discussion

Each of the randomized algorithms given in this paper has a small probability of giving an erroneous result; for example, procedure Find-Perfect-Matching may fail to produce a perfect matching in a graph that possesses one, and therefore its failure does not indicate with certainty that no perfect matching exists. Running the algorithm many times in parallel can reduce the probability of error to an exponentially low level, but can never eradicate it entirely. Howard Karloff [8] has given a Random NC algorithm for the odd-set cover problem, which is the "dual" of the matching problem. As Karloff points out, this result can be combined with our algorithms to yield algorithms for the perfect matching problem and the maximum matching problem which run in polylog expected time and always give the correct result; i.e., Las Vegas algorithms rather than Monte Carlo algorithms.

It remains an open question whether randomization can be dispensed with entirely in these problems. It would be very nice to show that the problem of deciding whether a graph has a perfect matching lies in NC, and even nicer to show that the

problem of constructing a perfect matching lies in NC.

Finally, the investigations reported here have led us into a broader study of the relation between decision problems and search problems; the results of that study are reported in the companion paper [9].

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> The following computational problem Conference, 1981): Given a graph G = (V, E)tex coloring. Decide whether f is a proper verte linear comparisons.

> Lower bounds on the complexity of thi of G. It is shown how geometric parameters complexity of this problem.

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The element uniqueness problem v [1] is the following: Given n reals $f_1, ...$ distinct. The elementary steps are linear legal coloring problem where the graph article [1] serves as the starting point for consider the legal coloring problem. Fo problem into an equivalent geometric p metric objects involved can be used to d legal coloring problem. Letting V = Vwith G a subset of \mathbb{R}^n as follows:

$$S(G) = \{(x_1, ..., x_n) \in \mathbb{R}^n \}$$

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