# Incremental Network Design with Minimum Spanning Trees 

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#### Abstract

Given an edge-weighted graph $G=(V, E)$ and a set $E_{0} \subset E$, the incremental network design problem with minimum spanning trees asks for a sequence of edges $e_{1}^{\prime}, \ldots, e_{r}^{\prime} \in E \backslash E_{0}$ minimizing $\sum_{t=1}^{T} w\left(X_{t}\right)$ where $w\left(X_{t}\right)$ is the weight of a minimum spanning tree $X_{t}$ for the subgraph $\left(V, E_{0} \cup\left\{e_{1}^{\prime}, \ldots, e_{t}^{\prime}\right\}\right)$ and $T=\left|E \backslash E_{0}\right|$. We prove that this problem can be solved by a greedy algorithm.


## 1 Introduction

Network planning involves two stages. First, the structure of the network needs to be decided. Second, the construction of the network needs to be scheduled. The first stage, the network design stage, has received considerable attention in the operations research literature (see the survey papers [8, 9] and the references therein). The second stage, the network construction stage, has received far less attention. However, because the construction of a network often stretches over a long period of time, the sequence in which the network is constructed is important as it defines when specific parts of the network become operational. It may even be beneficial to construct temporary links, i.e., links that are not part of the ultimate network, in order for parts of the network to become operational.

The class of incremental network design problems introduced by Baxter et al. [1 integrates the two stages of network planning and captures the trade-offs between construction cost and operational benefit. More specifically, an incremental network design problem can be associated with any network optimization problem $P$, e.g., finding a shortest path, finding a maximum flow, etc. In the most basic version, in addition to the network optimization problem $P$, an instance is given by a network $G=(V, E)$ with vertex set $V$ and edge set $E$ and an existing edge set $E_{0}$. The edge set $E \backslash E_{0}$ is referred to as the potential edge set and its cardinality $T=\left|E \backslash E_{0}\right|$ as the planning horizon. Let $\varphi_{P}(G)$ denote the value of an optimal solution to network optimization problem $P$ on network $G$. We are seeking a sequence $E_{0} \subset E_{1} \subset \cdots \subset E_{T}=E$ with $\left|E_{i} \backslash E_{i-1}\right|=1$ giving rise to networks $G_{0}, G_{1}, \ldots, G_{T}=G$, such that $\sum_{t=1}^{T} \varphi_{P}\left(G_{t}\right)$ is minimum (assuming that $P$ is a minimization problem). That is, in the basic version, a single edge can be built in each period of the planning horizon and we are seeking to minimize the operational costs over the planning horizon.

In more elaborate versions, a construction cost may be associated with building a potential edge and a budget may be available in each period, and the objective is to minimize the operational costs over the planning horizon subject to the constraint that the construction cost of the set of potential edges built in a period does not exceed the budget in that period.

Two natural heuristics for incremental network design problems, quickest-improvement and quickest-to-ultimate, are also of interest. Quickest-improvement always seeks to improve the value of the solution to the network optimization as quickly as possible, i.e., by adding as few potential edges to the network as possible. A description of quickest-improvement can be found in Algorithm 1.

```
Algorithm 1 quickest-improvement
\(i \leftarrow 0 ; E^{\prime} \leftarrow E_{0}\)
while \(\varphi_{P}\left(G_{E^{\prime}}\right)>\varphi_{P}\left(G_{E}\right)\) do
    \(k \leftarrow \min \left\{\left|E^{\prime \prime}\right| \quad: \quad E^{\prime \prime} \subseteq E \backslash E^{\prime}, \varphi_{P}\left(G_{E^{\prime}}\right)-\varphi_{P}\left(G_{E^{\prime} \cup E^{\prime \prime}}\right)>0\right\}\)
    \(i \leftarrow i+1 ; E_{i} \leftarrow \arg \max \left\{\varphi_{P}\left(G_{E^{\prime}}\right)-\varphi_{P}\left(G_{E^{\prime} \cup E^{\prime \prime}}\right) \quad: \quad E^{\prime \prime} \in E \backslash E^{\prime},\left|E^{\prime \prime}\right|=k\right\}\)
    \(E^{\prime} \leftarrow E^{\prime} \cup E_{i}\)
return \(\left(E_{1}, \ldots, E_{i}, E \backslash \bigcup_{j=0}^{i} E_{j}\right)\)
```

Quickest-to-ultimate first finds an optimal solution to the network optimization on the complete network, referred to as an ultimate solution, and then always seeks to improve the value of the solution to the network optimization as quickly as possible, but choosing only potential edges that are part of the ultimate solution. A description of quickest-to-ultimate can be found in Algorithm 2 .

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Algorithm 2 quickest-to-ultimate
\(\bar{E}=\left\{\right.\) edges in an optimal solution to \(P\) on \(\left.G_{E}\right\}\)
\(i \leftarrow 0 ; E^{\prime} \leftarrow E_{0}\)
while \(\varphi_{P}\left(G_{E^{\prime}}\right)>\varphi_{P}\left(G_{E}\right)\) do
    \(k \leftarrow \min \left\{\left|E^{\prime \prime}\right| \quad: \quad E^{\prime \prime} \in \bar{E} \backslash E^{\prime}, \varphi_{P}\left(G_{E^{\prime}}\right)-\varphi_{P}\left(G_{E^{\prime} \cup E^{\prime \prime}}\right)>0\right\}\)
    \(i \leftarrow i+1 ; E_{i} \leftarrow \arg \max \left\{\varphi_{P}\left(G_{E^{\prime}}\right)-\varphi_{P}\left(G_{E^{\prime} \cup E^{\prime \prime}}\right) \quad: \quad E^{\prime \prime} \in \bar{E} \backslash E^{\prime},\left|E^{\prime \prime}\right|=k\right\}\)
    \(E^{\prime} \leftarrow E^{\prime} \cup E_{i}\)
return \(\left(E_{1}, \ldots, E_{i}, E \backslash \bigcup_{j=0}^{i} E_{j}\right)\)
```

Incremental network design problems have been studied for the shortest path problem 1 and for the maximum flow problem [7]. In both cases, it was found that even the basic version of the incremental network design problem is NP-complete. For the natural heuristics described above it has been shown that for the shortest path problem, neither yields a constant factor approximation algorithm, but that for the maximum flow problem with the additional restriction that all arcs have unit capacity, quickest-to-ultimate yields a 2 -approximation algorithm and quickest-improvement yields a $3 / 2$-approximation algorithm.

These results have raised two questions: (1) Does there exist a network optimization problem for which the incremental design problem is polynomially solvable? (2) Does there exist a network optimization problem for which either quickest-to-ultimate or quickest-improvement solves the incremental design problem optimally?

In this paper, we answer both questions in the affirmative. We show that the basic version of the incremental network design problem with minimum spanning trees is solved by both quickest-improvement and quickest-to-ultimate.

The incremental network design problem with minimum spanning trees (IND-MST) is defined as follows. For a given graph $G=(V, E)$, a weight function $w: E \rightarrow \mathbb{R}$, and a set of existing edges $E_{0}$, such that the subgraph $G=\left(V, E_{0}\right)$ is connected, find a sequence $X_{0}, X_{1}, \ldots, X_{T}$ of spanning trees which minimizes the sum of the weights $w\left(X_{0}\right)+\cdots+w\left(X_{T}\right)$ subject to the condition that $X_{0} \subseteq E_{0}$ and $\left|X_{i} \cap\left(E \backslash\left(E_{0} \cup X_{1} \cup \cdots \cup X_{i-1}\right)\right)\right| \leqslant 1$ for $1 \leqslant i \leqslant T$, i.e., at most one edge from $E \backslash E_{0}$ might be added in each step. This has some similarity with the problem of maintaining a dynamic minimum spanning tree while the network data changes [3, 4, 5, 6, 11. In contrast to these dynamic minimum spanning tree problems, in our setting the network changes are not given as input, but are part of the decisions to be made. We will show that IND-MST can be solved by a greedy algorithm. This is a consequence of the corresponding result for the incremental matroid design problem with minimum weight matroid bases, which is stated in Section 2 and proved in Section 3

## 2 Incremental matroid design

Let $M=(E, \mathcal{I})$ be a matroid of rank $r$, where $E$ is the ground set, and $\mathcal{I} \subseteq 2^{E}$ is the collection of independent sets. We follow the notation of Schrijver [10]: the rank of a matroid $M$ is denoted by $\operatorname{rk}(M)$, minimal dependent sets are called circuits, for $A \subset E$ and $e \in E$ we write $A+e=A \cup\{e\}$ and $A-e=A \backslash\{e\}$, and we denote the closure of a set $A \subseteq E$ by $\operatorname{span}(A)$ :

$$
\operatorname{span}(A)=\{e \in E: \operatorname{rk}(A+e)=\operatorname{rk}(A)\}
$$

An important tool in our arguments is the following strong exchange property which was first proved by Brualdi [2].

Strong exchange property. If $X$ and $Y$ are bases of a matroid $M$ and $e \in X$, then there exists an element $e^{\prime} \in Y$ such that $X-e+e^{\prime}$ and $Y-e^{\prime}+e$ are bases of $M$.

As additional input, we are given a weight function $w: E \rightarrow \mathbb{R}$ and a subset $E_{0} \subset E$ such that $E_{0}$ contains a basis of $M$. We define a function $f: 2^{E \backslash E_{0}} \rightarrow \mathbb{R}$ by

$$
f(A)=\min \left\{w(X): X \subseteq E_{0} \cup A \text { is a basis of } M\right\} \quad \text { for } A \subseteq E \backslash E_{0}
$$

The incremental matroid design problem with minimum weight bases (IMD-MWB) problem for the time horizon $T=\left|E \backslash E_{0}\right|$ is the following optimization problem:

$$
\begin{equation*}
\min \left\{\sum_{i=0}^{T} f\left(A_{i}\right): A_{0}=\emptyset,\left|A_{i} \backslash A_{i-1}\right|=1 \text { for } i \in[T]\right\} \tag{1}
\end{equation*}
$$

For a basis $X$, a pair $\left(e, e^{\prime}\right) \in X \times(E \backslash X)$ is called an exchange pair for $X$ if $X-e+e^{\prime}$ is another basis. It is called an optimal exchange pair if $w(e)-w\left(e^{\prime}\right)$ is maximum. Algorithm 3 is a natural greedy strategy for solving IMD-MWB, where the output defines the sets $A_{i}$ in (1) via $A_{i}=\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ for $i \leqslant k$ and $A_{i}=A_{i-1}+e^{\prime}$ for arbitrary $e^{\prime} \in E \backslash\left(E_{0} \cup A_{i-1}\right)$ for $k+1 \leqslant i \leqslant T$. This corresponds to using quickest-improvement.

```
Algorithm 3 Greedy algorithm for the incremental minimum weight basis problem
\(k \leftarrow 0\)
\(w^{*} \leftarrow\) minimum weight of a basis of \(M\)
\(X \leftarrow\) any minimum weight basis of the submatroid induced by \(E_{0}\) (which is also a basis of \(M\) )
while \(w(X)>w^{*}\) do
    \(k \leftarrow k+1\)
    \(\left(e, e^{\prime}\right) \leftarrow\) an optimal exchange pair for \(X\)
    \(X \leftarrow X-e+e^{\prime}\)
    \(e_{k}^{\prime} \leftarrow e^{\prime}\)
return \(\left(e_{1}^{\prime}, \ldots, e_{k}^{\prime}\right)\)
```

Our main result is a consequence of the following theorem.
Theorem 1. Algorithm 3 finds an optimal solution for the problem IMD-MWB.
If the second component $e^{\prime}$ of an exchange pair $\left(e, e^{\prime}\right)$ for $X$ belongs to $Y \backslash X$, where $Y \subseteq E$, then we call such a pair an exchange pair for $(X, Y)$. Before proving Theorem 1 we observe that the search for an optimal exchange pair can be restricted to exchange pairs $\left(e, e^{\prime}\right)$ for $(X, Y)$, where $Y$ is a fixed minimum weight basis of $M$. This corresponds to using quickest-to-ultimate and leads to Algorithm 4.

```
Algorithm 4 Simplified greedy algorithm
\(X \leftarrow\) any minimum weight basis of the submatroid induced by \(E_{0}\)
\(Y \leftarrow\) any minimum weight basis of the matroid \(M\)
for \(k=1, \ldots,|Y \backslash X|\) do
    \(\left(e, e^{\prime}\right) \leftarrow\) an optimal exchange pair for \((X, Y)\)
    \(X \leftarrow X-e+e^{\prime}\)
    \(e_{k}^{\prime} \leftarrow e^{\prime}\)
return \(\left(e_{1}^{\prime}, \ldots, e_{k}^{\prime}\right)\)
```

Corollary 1. Algorithm 4 finds an optimal solution for the problem IMD-MWB.
Proof. This follows from the claim that for any basis $X$ of $M$, and any minimum weight basis $Y$, there is an optimal exchange pair $\left(e, e^{\prime}\right)$ for $X$ with $e^{\prime} \in Y$. Suppose the claim is false and let $\left(e, e^{\prime}\right)$ be an optimal exchange pair for $X$. By the strong exchange property applied to the bases $X^{\prime}=X-e+e^{\prime}$ and $Y$ and the element $e^{\prime} \in X^{\prime}$, there exists an $e^{\prime \prime} \in Y$ such that $X^{\prime}-e^{\prime}+e^{\prime \prime}=X-e+e^{\prime \prime}$ and $Y-e^{\prime \prime}+e^{\prime}$ are bases. Our assumption implies $w\left(e^{\prime \prime}\right)>w\left(e^{\prime}\right)$, while from the minimality of $Y$ it follows that $w\left(e^{\prime \prime}\right) \leqslant w\left(e^{\prime}\right)$.

## 3 Proof of Theorem 1

The following lemma is stated in 4 for graphical matroids. The argument works in general, and in order to make our presentation self-contained we include the short proof.

Lemma 1. Let $M=(E, \mathcal{I})$ be a matroid, $E_{0} \subseteq E, A \subseteq E \backslash E_{0}$. In addition, let $M_{0}=\left(E_{0}, \mathcal{I}_{0}\right)$ and $M_{A}=\left(E_{0} \cup A, \mathcal{I}_{A}\right)$ be the matroids induced by $E_{0}$ and $E_{0} \cup A$, respectively, i.e. $\mathcal{I}_{0}=\left\{X \cap E_{0}: X \in \mathcal{I}\right\}$ and $\mathcal{I}_{A}=\left\{X \cap\left(E_{0} \cup A\right): X \in \mathcal{I}\right\}$, and let $X$ be a minimum weight basis for the matroid $M_{0}$. Then there is a minimum basis $Y$ of $M_{A}$ such that $Y \subseteq X \cup A$.

Lemma 1 is proved by iterating the next lemma which states that a single element exchange is sufficient in order to update the minimum weight basis after one potential element is added.

Lemma 2. Let $A \subseteq E \backslash E_{0}$, and let $X_{A}$ be a minimum weight basis for the matroid $M_{A}$ induced by $E_{0} \cup A$. For every $e \in E \backslash\left(E_{0} \cup A\right)$, the set $X_{A}+e-e^{\prime}$ is a minimum weight basis for the matroid $M_{A+e}$ induced by $E_{0} \cup A+e$, where $e^{\prime}$ is an element of maximum weight in the circuit of $X+e$.

Proof. Suppose the statement is false and let $Y$ be a basis of $M_{A+e}$ with $w(Y)<w(X)+w(e)-w\left(e^{\prime}\right)$. Then $e \in Y$, and by the strong exchange property, there exists $e^{\prime \prime} \in X$ such that $Y+e^{\prime \prime}-e$ and $X+e-e^{\prime \prime}$ are bases. The choice of $e^{\prime}$ implies $w\left(e^{\prime \prime}\right) \leqslant w\left(e^{\prime}\right)$, while minimality of $X$ and our assumption on $Y$ imply that

$$
\begin{aligned}
w(X) \leqslant w\left(Y-e+e^{\prime \prime}\right)=w(Y)-w(e) & +w\left(e^{\prime \prime}\right) \\
& <w(X)+w(e)-w\left(e^{\prime}\right)-w(e)+w\left(e^{\prime \prime}\right)=w(X)-w\left(e^{\prime}\right)+w\left(e^{\prime \prime}\right),
\end{aligned}
$$

hence $w\left(e^{\prime \prime}\right)>w\left(e^{\prime}\right)$, and this contradiction concludes the proof.
Proof of Lemma 1. Let $A=\left\{e_{1}, \ldots, e_{k}\right\}$, set $A_{0}=\emptyset$ and $A_{i}=\left\{e_{1}, \ldots, e_{i}\right\}$ for $i=1, \ldots, k$, and apply Lemma 2 with $A=A_{i}, e=e_{i+1}$ for $i=0, \ldots, k-1$.

For $t=0,1, \ldots, T$, let

$$
F_{t}=\min \left\{f(A): A \subseteq E \backslash E_{0},|A|=t\right\} .
$$

Note that $F_{0}>F_{1}>\cdots>F_{t}=F_{t+1}=\cdots=F_{T}$ for some $t \leqslant \min \{r, T\}$ and Algorithm 3 terminates with $k \geqslant t$. Clearly, $F_{0}+F_{1}+\cdots+F_{T}$ is a lower bound for (11). The correctness of Algorithm 3 follows from the fact that it achieves this lower bound, which in turn is a consequence of the following extension property.

Lemma 3. Let $k<t$ be arbitrary, and let $A, B \subseteq E \backslash E_{0}$ with $|A|=k,|B|=k+1, f(A)=F_{k}$ and $f(B)=F_{k+1}$. Then there exists $e \in B \backslash A$ such that $f(A+e)=F_{k+1}$.

Proof. Let $M_{0}, M_{A}$ and $M_{B}$ denote the submatroids induced by $E_{0}, E_{0} \cup A$ and $E_{0} \cup B$, respectively. We have $\operatorname{rk}\left(M_{0}\right)=\operatorname{rk}\left(M_{A}\right)=\operatorname{rk}\left(M_{B}\right)=r$ because $E_{0}$ contains a basis of $M$. By the optimality of $A$ and $B$ and since $F_{k+1}<F_{k}$, the sets $A$ and $B$ are contained in every minimum weight basis of $M_{A}$ and $M_{B}$, respectively, and by Lemma 1, there are minimum weight bases $X, X_{A}$ and $X_{B}$ for these submatroids with $X_{A} \backslash X=A$ and $X_{B} \backslash X=\stackrel{\rightharpoonup}{B}$. We define a bipartite digraph $(\mathcal{U} \cup \mathcal{V}, \mathcal{A})$ with parts

$$
\mathcal{U}=X_{A} \backslash \operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right), \quad \mathcal{V}=X_{B} \backslash \operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right)
$$

Note that $\operatorname{rk}\left(\left(X_{B} \cap X\right) \cup A\right) \leqslant\left|\left(X_{B} \cap X\right) \cup A\right|=r-1$, hence $\mathcal{U}, \mathcal{V} \neq \emptyset$. Also, $\mathcal{U} \subseteq X \subseteq E_{0}$ and $\mathcal{V} \subseteq B \subseteq E \backslash E_{0}$, hence $\mathcal{U} \cap \mathcal{V}=\emptyset$. The element set is defined by

$$
\begin{aligned}
& \mathcal{A}=\left\{\left(e, e^{\prime}\right) \in \mathcal{U} \times \mathcal{V}:\left(e, e^{\prime}\right) \text { is an exchange pair for } X_{A}\right\} \\
& \cup\left\{\left(e, e^{\prime}\right) \in \mathcal{V} \times \mathcal{U}:\left(e, e^{\prime}\right) \text { is an exchange pair for } X_{B}\right\} .
\end{aligned}
$$

Let $e^{\prime} \in \mathcal{V}$. Then $e^{\prime} \in E \backslash\left(E_{0} \cup A\right)$, hence $e^{\prime} \notin X_{A}$ and $X_{A}+e^{\prime}$ contains a circuit $C$. We claim that $C-e^{\prime} \nsubseteq \operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right)$, and this implies that there exist $e \in C \cap \mathcal{U}$, and consequently $\left(e, e^{\prime}\right) \in \mathcal{A}$. For the sake of contradiction, suppose the claim is false and $C-e^{\prime} \subseteq \operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right)$. From the fact that $C$ is a circuit, it follows that

$$
e^{\prime} \in \operatorname{span}\left(C-e^{\prime}\right) \subseteq \operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right)
$$

which is a contradiction to $e^{\prime} \in \mathcal{V}$. Similarly, if $e^{\prime} \in \mathcal{U}$, then $e^{\prime} \in X_{A} \backslash A \subseteq E_{0}$ and $e^{\prime} \notin X_{B} \cap X$, which implies $e^{\prime} \notin X_{B}$, hence there exists a circuit $C$ in $X_{B}+e^{\prime}$. As before, the assumption that $C-e^{\prime}$ is contained in $\operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right)$ leads to the contradiction $e^{\prime} \in \operatorname{span}\left(\left(X_{B} \cap X\right) \cup A\right)$. By this argument, for every $e^{\prime} \in \mathcal{U}$ there exists an $e \in \mathcal{V}$ with $\left(e, e^{\prime}\right) \in \mathcal{A}$. We conclude that every node in the digraph $(\mathcal{U} \cup \mathcal{V}, \mathcal{A})$ has positive indegree, thus the digraph contains a directed cycle, and this implies that there are $e^{\prime}, e^{\prime \prime} \in \mathcal{U}$ and $e \in \mathcal{V}$ such that $\left(e^{\prime}, e\right) \in \mathcal{A},\left(e, e^{\prime \prime}\right) \in \mathcal{A}$, and $w\left(e^{\prime}\right) \geqslant w\left(e^{\prime \prime}\right)$. From this we derive

$$
\begin{aligned}
& f(A+e) \leqslant w\left(X_{A}+e-e^{\prime}\right)=f(A)+w(e)-w\left(e^{\prime}\right) \leqslant F_{k}+w(e)-w\left(e^{\prime \prime}\right) \\
& \leqslant f(B-e)+w(e)-w\left(e^{\prime \prime}\right) \leqslant w\left(X_{B}-e+e^{\prime \prime}\right)+w(e)-w\left(e^{\prime \prime}\right)=w\left(X_{B}\right)=F_{k+1} .
\end{aligned}
$$

The converse inequality $f(A+e) \geqslant F_{k+1}$ is obvious and this concludes the proof.

## 4 Run-time analysis

In order to bound the time complexities of the problems IMD-MWB and IND-MST, we argue that the initial basis $X$, the ultimate basis $Y$ and the list of exchange pairs $\mathcal{E}$ can be determined simultaneously. The idea is to consider the elements of $E$ in order of nondecreasing weights and to construct and maintain three independent sets $X, Y$ and $Z$ using the following update rules:

1. An element $e \in E$ is added to $X$ if and only if $e \in E_{0}$ and the addition of $e$ does not create a circuit in $X$. Hence $X$ is an initial minimum weight basis when the algorithm terminates.
2. An element $e \in E$ is added to $Y$ if and only if the addition of $e$ does not create a circuit in $Y$. Hence $Y$ is an ultimate minimum weight basis when the algorithm terminates.
3. An element $e \in E$ is added to $Z$ if and only if it is added to $X$ or $Y$. We will show that because an edge added to $Y$ does not create a cycle in $Y$, it does not create a cycle in $Z$. Therefore, if an edge $e$ added to $Z$ creates a cycle $C$, then $e \in E_{0}$ and $C$ must contain an element of $E \backslash E_{0}$ and a maximum weight element $e^{\prime}$ of $C \backslash E_{0}$ is removed from $Z$. The pair $\left(e, e^{\prime}\right)$ is added to the set $\mathcal{E}$ of exchange pairs.

To finish up, the set $\mathcal{E}$ of exchange pairs $\left(e, e^{\prime}\right)$ is ordered such that $w(e)-w\left(e^{\prime}\right)$ is nonincreasing, and ties are broken in favor of the pair that was added to $\mathcal{E}$ last. More precisely, we index $\mathcal{E}=\left\{\left(e_{1}, e_{1}^{\prime}\right), \ldots,\left(e_{k}, e_{k}^{\prime}\right)\right\}$, such that for every $i \in\{1, \ldots, k-1\}$, we have either $w\left(e_{i}\right)-w\left(e_{i}^{\prime}\right)>w\left(e_{i+1}\right)-w\left(e_{i+1}^{\prime}\right)$, or $w\left(e_{i+1}\right)-w\left(e_{i+1}^{\prime}\right)=$ $w\left(e_{i}\right)-w\left(e_{i}^{\prime}\right)$ and the pair $\left(e_{i}, e_{i}^{\prime}\right)$ is added to $\mathcal{E}$ after $\left(e_{i+1}, e_{i+1}^{\prime}\right)$. A formal description can be found in Algorithm 5. For the remainder of the section, let $k$ be the size of the set $\mathcal{E}$ returned by the algorithm.

```
Algorithm 5 Efficient solution of the problem IMD-MWB
\(X \leftarrow \emptyset ; \quad Y \leftarrow \emptyset ; \quad Z \leftarrow \emptyset \quad / /\) initialize three independent sets
\(\mathcal{E} \leftarrow \emptyset \quad / /\) initialize set of exchange pairs
for \(e \in E\) do // in nondecreasing order of weight
    if \(e \in E_{0}\) then
        if \(X+e \in \mathcal{I}\) then
            \(X \leftarrow X+e ; \quad Z \leftarrow Z+e\)
            if \(Z\) contains a circuit \(C\) then
                \(e^{\prime}=\arg \max \left\{w\left(e^{\prime \prime}\right): e^{\prime \prime} \in C \backslash E_{0}\right\}\)
                \(Z \leftarrow Z-e^{\prime} ; \quad \mathcal{E} \leftarrow \mathcal{E} \cup\left\{\left(e, e^{\prime}\right)\right\}\)
    if \(Y+e \in \mathcal{I}\) then
                \(Y \leftarrow Y+e ; \quad Z \leftarrow Z+e\)
sort \(\mathcal{E}\)
return \(X\) and \(\mathcal{E}=\left\{\left(e_{1}, e_{1}^{\prime}\right), \ldots,\left(e_{k}, e_{k}^{\prime}\right)\right\}\)
```

In order to show the correctness of Algorithm 5, we introduce some additional notation. Let $X_{0}=X$ and $X_{i}=X_{i-1}-e_{i}+e_{i}^{\prime}$ for $\left(e_{i}, e_{i}^{\prime}\right) \in \mathcal{E}$ for $i=1,2, \ldots, k$. Let $m=|E|$. We say that $e$ has position $p$, denoted by $\operatorname{pos}(e)=p$, if $e$ is the element that is handled in the $p$-th iteration of the for-loop, $1 \leqslant p \leqslant m$. Note that $\operatorname{pos}(e)<\operatorname{pos}\left(e^{\prime}\right)$ implies $w(e) \leqslant w\left(e^{\prime}\right)$.

Let $X^{p}, Y^{p}, Z^{p}$ denote the sets $X, Y$ and $Z$ after the $p$-th iteration of the for-loop has been completed, $0 \leqslant p \leqslant m$. It is obvious, that if $p \leqslant l$, then

$$
X^{p} \subseteq X^{l} \quad \text { and } \quad Y^{p} \subseteq Y^{l}
$$

For $i=1, \ldots, k$, we denote the circuit that leads to the deletion of $e_{i}^{\prime}$ from $Z$ by $C_{i}$. In other words, if $\operatorname{pos}\left(e_{i}\right)=p$ then $C_{i}$ is the unique circuit in $Z^{p-1}+e_{i}$ and we have $Z^{p}=Z^{p-1}+e_{i}-e_{i}^{\prime}$. If $e$ is any element of $C_{i}$, then $\operatorname{pos}(e) \leqslant \operatorname{pos}\left(e_{i}\right)$ and hence $w(e) \leqslant w\left(e_{i}\right)$. By the choice of $e_{i}^{\prime}$ we have

$$
\begin{equation*}
w(e) \leqslant w\left(e_{i}^{\prime}\right) \text { for all } e \in C_{i} \backslash E_{0} \tag{2}
\end{equation*}
$$

Let

$$
H^{p}=\left\{e_{i}: i \in[k] \text { and } \operatorname{pos}\left(e_{i}\right) \leqslant p\right\}, \quad \text { and } \quad H^{\prime p}=\left\{e_{i}^{\prime}: i \in[k] \text { and } \operatorname{pos}\left(e_{i}\right) \leqslant p\right\}
$$

That is, the sets $H^{p}$ and $H^{\prime p}$ contain the edges involved in exchanges occurring either before or when the edge in position $p$ is examined. More precisely, we have

$$
\left(X^{p} \times Y^{p}\right) \cap \mathcal{E}=\left\{\left(e_{i}, e_{i}^{\prime}\right): e_{i} \in H^{p} \text { and } e_{i}^{\prime} \in H^{\prime p}\right\}
$$

Lemma 4. We have for $0 \leqslant p \leqslant m$

1. $X^{p} \subseteq Z^{p} \subseteq X^{p} \cup Y^{p}$,
2. $\operatorname{span}\left(Y^{p}\right)=\operatorname{span}\left(Z^{p}\right)$,
3. $Y^{p} \backslash X^{p} \subseteq E \backslash E_{0}$,
4. $X^{p}, Y^{p}, Z^{p} \in \mathcal{I}$,
5. $X^{p} \backslash Y^{p}=H^{p}$,
6. $Y^{p} \backslash Z^{p}=H^{\prime p}$.

Proof. We proceed by induction on $p$. The case $p=0$ is trivial. Now consider the step $p-1 \rightarrow p$. From $X^{p-1} \subseteq Z^{p-1} \subseteq X^{p-1} \cup Y^{p-1}$ we obtain directly $X^{p} \subseteq Z^{p} \subseteq X^{p} \cup Y^{p}$ since an element is added to $Z$ iff it is added to $X$ or to $Y$ and an element is deleted from $Z$ only if this element does not belong to $E_{0}$, i.e. not to $X$. Now we prove the other assertions. Let $e$ be the element with $\operatorname{pos}(e)=p$.

If $X^{p}=X^{p-1}$ and $Y^{p}=Y^{p-1}$ then also $Z^{p}=Z^{p-1}, H^{p}=H^{p-1}$ and $H^{\prime p}=H^{p-1}$ and the assertion follows from the induction hypothesis. So there are three main cases:

Case 1. $X^{p}=X^{p-1}+e$ and $Y^{p}=Y^{p-1}$. Then $e \in \operatorname{span}\left(Y^{p-1}\right)$. By the induction hypothesis, $e \in$ $\operatorname{span}\left(Z^{p-1}\right)$ and consequently $Z^{p}=\left(Z^{p-1}+e\right)-e^{\prime}$ for some $e^{\prime}$ in the unique circuit of $Z^{p-1}+e$ where $e^{\prime} \subseteq E \backslash E_{0}$ and $e^{\prime} \neq e$. Obviously, $Z^{p} \in \mathcal{I}$ and $\operatorname{span}\left(Z^{p}\right)=\operatorname{span}\left(Z^{p-1}+e\right)=\operatorname{span}\left(Z^{p-1}\right)=$ $\operatorname{span}\left(Y^{p-1}\right)=\operatorname{span}\left(Y^{p}\right)$. Obviously, $X^{p}, Y^{p} \in \mathcal{I}$ and $Y^{p} \backslash X^{p} \subseteq Y^{p-1} \backslash X^{p-1} \subseteq E \backslash E_{0}$. Finally, $X^{p} \backslash Y^{p}=\left(X^{p-1}+e\right) \backslash Y^{p-1}=\left(X^{p-1} \backslash Y^{p-1}\right)+e=H^{p-1}+e=H^{p}$ and $Y^{p} \backslash Z^{p}=Y^{p-1} \backslash\left(\left(Z^{p-1}+\right.\right.$ $\left.e)-e^{\prime}\right)=\left(Y^{p-1} \backslash Z^{p-1}\right)+e^{\prime}=H^{p-1}+e^{\prime}=H^{\prime p}$.

Case 2. $X^{p}=X^{p-1}$ and $Y^{p}=Y^{p-1}+e$. Then $Z^{p}=Z^{p-1}+e$ and hence $\operatorname{span}\left(Y^{p}\right)=\operatorname{span}\left(Z^{p}\right)$. Obviously, $X^{p}, Y^{p} \in \mathcal{I}$ and $e \notin \operatorname{span}\left(Y^{p-1}\right)$, i.e. $e \notin \operatorname{span}\left(Z^{p-1}\right)$. Thus $Z^{p} \in \mathcal{I}$.
If $e \in E_{0}$ then $e \in \operatorname{span}\left(X^{p-1}\right) \subseteq \operatorname{span}\left(Z^{p-1}\right)=\operatorname{span}\left(Y^{p-1}\right)$, which contradicts $Y^{p-1}+e \in \mathcal{I}$. Hence $e \notin E_{0}$. Then $Y^{p} \backslash X^{p}=\left(Y^{p-1}+e\right) \backslash X^{p-1} \subseteq E \backslash E_{0}$. Finally, $X^{p} \backslash Y^{p}=X^{p-1} \backslash\left(Y^{p-1}+e\right)=$ $X^{p-1} \backslash Y^{p-1}=H^{p-1}=H^{p}$ and $Y^{p} \backslash Z^{p}=Y^{p-1} \backslash Z^{p-1}=H^{\prime p-1}=H^{\prime p}$ 。

Case 3. $X^{p}=X^{p-1}+e$ and $Y^{p}=Y^{p-1}+e$. Then $X^{p}, Y^{p} \in \mathcal{I}$ and $e \notin \operatorname{span}\left(Y^{p-1}\right)$, i.e. $e \notin \operatorname{span}\left(Z^{p-1}\right)$. Thus $Z^{p-1}+e \in \mathcal{I}$. This implies $Z^{p}=Z^{p-1}+e$ and consequently $\operatorname{span}\left(Y^{p}\right)=\operatorname{span}\left(Z^{p}\right)$ as well as $Y^{p} \backslash X^{p}=Y^{p-1} \backslash X^{p-1} \subseteq E \backslash E_{0}$. Finally, $X^{p} \backslash Y^{p}=X^{p-1} \backslash Y^{p-1}=H^{p-1}=H^{p}$ and $Y^{p} \backslash Z^{p}=$ $Y^{p-1} \backslash Z^{p-1}=H^{\prime p-1}=\overline{H^{\prime p}}$.

In the following, we denote the three independent sets that the algorithm terminates with by $X, Y$, and $z$, i.e., $X=X^{m}, Y=Y^{m}$ and $Z=Z^{m}$.

Corollary 2. We have

1. $X=Z$,
2. $X \backslash Y=\left\{e_{1}, \ldots, e_{k}\right\}$,
3. $Y \backslash X=\left\{e_{1}^{\prime}, \ldots, e_{k}^{\prime}\right\}$.

Proof. By our general supposition that $E_{0}$ contains a basis, the set $X$ is a basis for $M$. Since, by Lemma 4 $Z$ is independent and $X \subseteq Z$, we have $X=Z$. Again, by Lemma $4 X X Y=X^{m} \backslash Y^{m}=H^{m}=\left\{e_{1}, \ldots, e_{k}\right\}$ and $Y \backslash X=Y \backslash Z=Y^{m} \backslash Z^{m}=H^{\prime m}=\left\{e_{1}^{\prime}, \ldots, e_{k}^{\prime}\right\}$.
Corollary 3. If $\operatorname{pos}\left(e_{i}\right)=p, i \in[k]$, then $C_{i} \cap H^{\prime p}=\left\{e_{i}^{\prime}\right\}$.
Proof. We have $C_{i} \subseteq Z^{p-1}+e_{i}=Z^{p}+e_{i}^{\prime}$ and thus $C_{i} \cap H^{\prime p}=C_{i} \cap\left(Y^{p} \backslash Z^{p}\right) \subseteq\left(Z^{p}+e_{i}^{\prime}\right) \cap\left(Y^{p} \backslash Z^{p}\right)=\left\{e_{i}^{\prime}\right\}$. Clearly, $e_{i}^{\prime} \in C_{i} \cap H^{\prime p}$.

Lemma 5. Let $C$ be a circuit in $X \cup Y$. Let

$$
e^{*}=\arg \max \{\operatorname{pos}(e): e \in C\}
$$

Then $e^{*}=e_{l}$ for some $l \in[k]$.
Proof. Let $\operatorname{pos}\left(e^{*}\right)=q$. Then $C-e^{*} \subseteq X^{q-1} \cup Y^{q-1}$. This implies $e^{*} \in \operatorname{span}\left(X^{q-1} \cup Y^{q-1}\right)$ and from Lemma 4 we obtain $e^{*} \in \operatorname{span}\left(X^{q-1} \cup \overline{Z^{q-1}}\right)=\operatorname{span}\left(Z^{q-1}\right)$. Since $e^{*} \in X^{q} \cup Y^{q}$, but $e^{*} \notin X^{q-1} \cup Y^{q-1}$ we must have $X^{q}=X^{q-1}+e^{*}$ and $Z^{q}=Z^{q-1}+e^{*}-e^{\prime}$ for some $e^{\prime} \in E \backslash E_{0}$. Hence $e^{*}=e_{l}$ for some $l \in[k]$.

Lemma 6. We have $C_{i} \subseteq X \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ for every $i \in[k]$.
Proof. Using Corollary 2, we obtain $C_{i} \subseteq X \cup Y=X \cup(Y \backslash X)=X \cup\left\{e_{1}^{\prime}, \ldots, e_{k}^{\prime}\right\}$. Assume that there is some $j>i$ such that $e_{j}^{\prime} \in C_{i}$. Then $\operatorname{pos}\left(e_{j}\right)>\operatorname{pos}\left(e_{i}\right)$ and thus $w\left(e_{j}\right) \geqslant w\left(e_{i}\right)$. From (2) and $e_{j}^{\prime}$ in $C_{i} \backslash E_{0}$, we obtain $w\left(e_{i}^{\prime}\right) \geqslant w\left(e_{j}^{\prime}\right)$. Consequently, $w\left(e_{j}\right)-w\left(e_{j}^{\prime}\right) \geqslant w\left(e_{i}\right)-w\left(e_{i}^{\prime}\right)$, a contradiction to the ordering of $\mathcal{E}$.

Lemma 7. For $i \in[k]$ the pair $\left(e_{i}, e_{i}^{\prime}\right)$ is an exchange pair for $\left(X_{i-1}, Y\right)$.

Proof. Clearly, $e_{i}^{\prime} \in Y \backslash\left(X \cup\left\{e_{1}^{\prime}, \ldots, e_{i-1}^{\prime}\right\}\right) \subseteq Y \backslash X_{i-1}$. We have to show that $e_{i}$ lies in the unique circuit of $X_{i-1}+e_{i}^{\prime}=\left(X \backslash\left\{e_{1}, \ldots, e_{i-1}\right\}\right) \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$. An equivalent statement is that there is a circuit in $\left(X \backslash\left\{e_{1}, \ldots, e_{i-1}\right\}\right) \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ containing $e_{i}$. From Lemma we know that there is at least a circuit in $X \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ containing $e_{i}$, namely $C_{i}$. For a circuit $C$, let

$$
\mu_{C}^{i}=\max \left\{\operatorname{pos}\left(e_{j}\right): j \in[i-1] \text { and } e_{j} \in C_{i}\right\}
$$

where the maximum extended over an empty set is defined to be $-\infty$.
Assume that there is no circuit in $\left(X \backslash\left\{e_{1}, \ldots, e_{i-1}\right\}\right) \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ containing $e_{i}$. Then we choose a circuit $C$ in $X \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ containing $e_{i}$ such that $\mu_{C}^{i}$ is minimal. By our assumption, $\mu_{C}^{i}$ is finite and there exists an integer $l \in[i-1]$ such that $e_{l} \in C_{i}$ and $\operatorname{pos}\left(e_{l}\right)=\mu_{C}^{i}$. The circuit $C_{l}$ also contains $e_{l}$ and is contained in $X \cup\left\{e_{1}^{\prime}, \ldots, e_{\dot{C}}^{\prime}\right\}$ in view of Lemma 6 and $l<i$. Note that $\operatorname{pos}(e)<\operatorname{pos}\left(e_{l}\right)$ for all $e \in C_{l}-e_{l}$. But there is also a circuit $\widetilde{C} \subseteq C \cup C_{l}-e_{l}$. Obviously, $\widetilde{C} \subseteq X \cup\left\{e_{1}^{\prime}, \ldots, e_{i}^{\prime}\right\}$ and $\mu_{\widetilde{C}}^{i}<\mu_{C}^{i}$, a contradiction to the choice of $C$.

Lemma 8. For $i \in[k]$ the pair $\left(e_{i}, e_{i}^{\prime}\right)$ is an optimal exchange pair for $\left(X_{i-1}, Y\right)$.
Proof. Assume that $\left(e_{i}, e_{i}^{\prime}\right)$ is not optimal. Then there is a better exchange pair $\left(\hat{e}, \hat{e}^{\prime}\right)$ for $\left(X_{i-1}, Y\right)$. We have $\hat{e}^{\prime} \in Y \backslash X_{i-1}=\left(Y \backslash\left(X \cup\left\{e_{1}^{\prime}, \ldots, e_{i-1}^{\prime}\right\}\right)\right) \cup\left(Y \cap\left\{e_{1}, \ldots, e_{i-1}\right\}\right)$. From Corollary 2 it follows that $\hat{e}^{\prime}=e_{j}^{\prime}$ for some $j \in[k]$. Since $e_{1}^{\prime}, \ldots, e_{i-1}^{\prime} \notin Y \backslash X_{i-1}$,

$$
\begin{equation*}
j \geqslant i \tag{3}
\end{equation*}
$$

Let $\widehat{C}$ be the unique circuit in $X_{i-1}+e_{j}^{\prime}$ (containing $\hat{e}$ ) and let $e^{*}=\arg \max \{\operatorname{pos}(e): e \in \widehat{C}\}$. Then $w\left(e^{*}\right) \geqslant w(\hat{e})$. By Lemma 5. $e^{*}=e_{l}$ for some $l \in[k]$. In particular, $e^{*} \neq e_{j}^{\prime}$ and thus $\left(e^{*}, e_{j}^{\prime}\right)$ is also an exchange pair for $\left(X_{i-1}, Y\right)$. Since $e_{1}, \ldots, e_{i-1} \notin X_{i-1}$,

$$
\begin{equation*}
l \geqslant i \tag{4}
\end{equation*}
$$

We have (with $\left.e^{*}=e_{l}\right) w\left(e_{l}\right)-w\left(e_{j}^{\prime}\right) \geqslant w(\hat{e})-w\left(e_{j}^{\prime}\right)$ and hence

$$
\begin{equation*}
w\left(e_{l}\right)-w\left(e_{j}^{\prime}\right)>w\left(e_{i}\right)-w\left(e_{i}^{\prime}\right) \tag{5}
\end{equation*}
$$

By the ordering of $\mathcal{E}$ and by (4),

$$
\begin{equation*}
w\left(e_{i}\right)-w\left(e_{i}^{\prime}\right) \geqslant w\left(e_{l}\right)-w\left(e_{l}^{\prime}\right) . \tag{6}
\end{equation*}
$$

The inequalities (5) and (6) imply

$$
\begin{equation*}
w\left(e_{j}^{\prime}\right)<w\left(e_{l}^{\prime}\right) \tag{7}
\end{equation*}
$$

Let $p=\operatorname{pos}\left(e_{l}\right)$. Then $\operatorname{pos}\left(e_{j}\right)<p$, because otherwise $w\left(e_{j}\right) \geqslant w\left(e_{l}\right)$ and hence, in view of (3) and the ordering of $\mathcal{E}$,

$$
w\left(e_{l}\right)-w\left(e_{j}^{\prime}\right) \leqslant w\left(e_{j}\right)-w\left(e_{j}^{\prime}\right) \leqslant w\left(e_{i}\right)-w\left(e_{i}^{\prime}\right)
$$

a contradiction to 5 . Moreover, $\widehat{C} \subseteq X^{p} \cup Y^{p}$ since $e_{l}$ has maximal position in $\widehat{C}$. For a circuit $C$ let

$$
\begin{aligned}
\alpha_{C}^{p} & =\max \left\{w\left(e_{h}^{\prime}\right): h \in[k], e_{h}^{\prime} \in C \text { and } \operatorname{pos}\left(e_{h}\right) \leqslant p\right\}, \\
\nu_{C} & =\min \left\{\operatorname{pos}\left(e_{h}\right): h \in[k] \text { and } e_{h}^{\prime} \in C\right\},
\end{aligned}
$$

where the maximum (resp. minimum) extended over an empty set is defined to be $-\infty$ (resp. $\infty$ ). For $\widehat{C}$ these values are finite since $e_{j}^{\prime} \in \widehat{C}$ and $\operatorname{pos}\left(e_{j}\right)<p$. Hence there is an integer $g \in[k]$ such that $e_{g}^{\prime} \in \widehat{C}, \operatorname{pos}\left(e_{g}\right) \leqslant p$ and $w\left(e_{g}^{\prime}\right)=\alpha_{\widehat{C}}^{p}$. Note that

$$
\begin{equation*}
g \leqslant i-1 \text { or } g=j \tag{8}
\end{equation*}
$$

since $\widehat{C} \subseteq X \cup\left\{e_{1}^{\prime}, \ldots, e_{i-1}^{\prime}, e_{j}^{\prime}\right\}$. We say that a circuit $C$ is $p$-majorized by a number $\alpha$ if

$$
w\left(e_{h}^{\prime}\right) \leqslant \alpha \quad \text { for all } h \in[k] \text { with } e_{h}^{\prime} \in C \text { and } \operatorname{pos}\left(e_{h}\right) \leqslant p
$$

Note that $\widehat{C}$ is $p$-majorized by $\alpha_{\widehat{C}}^{p}$. Now choose a circuit $C^{*}$ in $X^{p} \cup Y^{p}$ that contains $e_{l}$, is $p$-majorized by $\alpha_{\widehat{C}}^{p}$ and has maximal $\nu$-value.

Note that, in view of Lemma 4, $C^{*} \subseteq X^{p} \cup Y^{p} \subseteq Z^{p} \cup Y^{p}=Z^{p} \cup\left(Y^{p} \backslash Z^{p}\right)=Z^{p} \cup H^{\prime p}$. We have $\nu_{C^{*}} \leqslant p$ because otherwise $C^{*} \cap H^{\prime p}=\emptyset$ and $C^{*}$ would be a circuit in $Z^{p}$.

Assume that $\nu_{C^{*}}<p$. Then choose $q \in[k]$ such that $e_{q}^{\prime} \in C^{*}$ and $\operatorname{pos}\left(e_{q}\right)=\nu_{C^{*}}$ and consider the circuit $C_{q}$. We have $C_{q} \subseteq X^{p} \cup Y^{p}$ because of $\operatorname{pos}\left(e_{q}\right)<p$. Moreover, $w\left(e_{h}^{\prime}\right) \leqslant w\left(e_{q}^{\prime}\right)$ and $\operatorname{pos}\left(e_{h}\right) \geqslant \operatorname{pos}\left(e_{q}\right)$ for all $h \in[k]$ with $e_{h}^{\prime} \in C_{q}$ by the choice of $e_{q}^{\prime}$ in $C_{q}$ and Corollary 3 . Since $C^{*}$ is $p$-majorized by $\alpha_{\widehat{C}}^{p}$, in particular $w\left(e_{q}^{\prime}\right) \leqslant \alpha_{\widehat{C}}^{p}$ and thus also $C_{q}$ is $p$-majorized by $\alpha_{\widehat{C}}^{p}$. From $e_{q}^{\prime} \in C^{*} \cap C_{q}$ and $e_{l} \in C^{*} \backslash C_{q}$ it follows that there is a circuit $\widetilde{C} \subseteq C^{*} \cup C_{q}-e_{q}^{\prime}$ containing $e_{l}$. Clearly, $\widetilde{C} \subseteq X^{p} \cup Y^{p}$ and $\widetilde{C}$ is $p$-majorized by $\alpha_{\widehat{C}}^{p}$. Obviously, $\min \left\{\operatorname{pos}\left(e_{h}\right): h \in[k]\right.$ and $\left.e_{h}^{\prime} \in C_{q} \cup C^{*}\right\}=\operatorname{pos}\left(e_{q}\right)$. Thus $\nu_{\widetilde{C}}>\nu_{C^{*}}$, a contradiction to the choice of $C^{*}$.

Consequently, $\nu_{C^{*}}=p$. Since $e_{l}$ is the (unique) element of position $p$, necessarily $e_{l}^{\prime} \in C^{*}$. Because $C^{*}$ is $p$-majorized by $\alpha_{\widehat{C}}^{p}$, in particular $w\left(e_{l}^{\prime}\right) \leqslant \alpha_{\widehat{C}}^{p}=w\left(e_{g}^{\prime}\right)$. From $\sqrt{7}$ we obtain $g \neq j$, hence by (8) and (4),

$$
\begin{equation*}
g \leqslant i-1<l . \tag{9}
\end{equation*}
$$

On the other hand, the relation $\operatorname{pos}\left(e_{g}\right) \leqslant \operatorname{pos}\left(e_{l}\right)$ implies $w\left(e_{g}\right) \leqslant w\left(e_{l}\right)$. Consequently,

$$
w\left(e_{l}\right)-w\left(e_{l}^{\prime}\right) \geqslant w\left(e_{g}\right)-w\left(e_{g}^{\prime}\right),
$$

and then the ordering of $\mathcal{E}$ implies $l \leqslant g$, which contradicts (9). Thus our first assumption was false, and the proof is complete.

Let us assume that Algorithm 5 gets as input the elements of $E$ in nondecreasing order of weights. Then in each iteration of the for-loop there are three independence tests, and at most one search for the element $e^{\prime} \in Z$ to be removed. This latter step is necessary at most $r$ times and can be done in time $O(|Z|)=O(r)$, where $r=\operatorname{rk}(M)$ is the rank of the matroid. Assuming the availability of an independence oracle for the matroid, the for-loop terminates in time $O\left(|E|+r^{2}\right)$. The final sorting of $\mathcal{E}$ takes time $O(r \log r)$. To summarize, we have proved the following bound for the time complexity of the problem IMD-MWB.

Theorem 2. Relative to an independence oracle, the problem IMD-MWB for a matroid $M=(E, \mathcal{I})$ of rank $r$ can be solved in time

$$
O\left(\max \left\{|E|+r^{2},|E| \log |E|\right\}\right)
$$

Finally, we consider the special case where $M$ is a graphical matroid to solve our original problem INDMST.

Theorem 3. The problem IND-MST for a graph with $n$ vertices and $m$ edges can be solved in time

$$
O\left(\max \left\{n^{2}, m \log m\right\}\right) .
$$

Proof. Constructing minimum spanning trees $X$ and $Y$ takes $m \log m$ time. Maintaining the set $Z$ takes $m \log m$ time, except for the time taken to search for exchange pairs. It can happen at most $n-1$ times that an edge $e$ added to $Z$ creates a cycle, and when that happens, the cycle itself and the edge $e^{\prime} \in E \backslash E_{0}$ on the cycle with the largest weight can be found in $O(n)$ time by DFS. This gives a run-time bound of

$$
O\left(\max \left\{m \log n+n^{2}, m \log m\right\}\right)=O\left(\max \left\{n^{2}, m \log m\right\}\right) .
$$

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