# APPROXIMATING CLIQUE-WIDTH AND BRANCH-WIDTH 

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#### Abstract

We construct a polynomial-time algorithm to approximate the branch-width of certain symmetric submodular functions, and give two applications.

The first is to graph "clique-width". Clique-width is a measure of the difficulty of decomposing a graph in a kind of tree-structure, and if a graph has clique-width at most $k$ then the corresponding decomposition of the graph is called a " $k$-expression". We find (for fixed $k$ ) an $O\left(n^{9} \log n\right)$-time algorithm that, with input an $n$-vertex graph, outputs either a $\left(2^{3 k+2}-1\right)$ expression for the graph, or a true statement that the graph has clique-width at least $k+1$. (The best earlier algorithm algorithm, by Johansson [13], constructed a $k \log n$-expression for graphs of clique-width at most $k$.) It was already known that several graph problems, NPhard on general graphs, are solvable in polynomial time if the input graph comes equipped with a $k$-expression (for fixed $k$ ). As a consequence of our algorithm, the same conclusion follows under the weaker hypothesis that the input graph has clique-width at most $k$ (thus, we no longer need to be provided with an explicit $k$-expression).

Another application is to the area of matroid branch-width. For fixed $k$, we find an $O\left(n^{4}\right)$ time algorithm that, with input an $n$-element matroid in terms of its rank oracle, either outputs a branch-decomposition of width at most $3 k-1$ or a true statement that the matroid has branch-width at least $k+1$. The previous algorithm by Hliněný [11] was only for representable matroids.


## 1. Introduction

Some algorithmic problems, NP-hard on general graphs, are known to be solvable in polynomial time when the input graph admits a decomposition into trivial pieces by means of a tree-structure of cutsets of bounded order. However, it makes a difference whether the input graph is presented together with the corresponding tree-structure of cutsets or not. We have in mind two kinds of decompositions, "tree-width" and "clique-width" decompositions. These are similar graph invariants, and while the results of this paper concern clique-width, we begin with tree-width for purposes of comparison.

Having bounded clique-width is more general than having bounded tree-width, in the following sense. Every graph $G$ of tree-width at most $k$ has clique-width at most $O\left(2^{k}\right)[5,7]$, and for such graphs (for $k$ fixed) the clique-width of $G$ can be determined in linear time [9]. No bound in the reverse direction holds, for there are graphs of arbitrary large tree-width with clique-width at most $k$. (But, for fixed $t$, if $G$ does not contain $K_{t, t}$ as a subgraph, then the tree-width is at most $3 k(t-1)-1$ [10].)

The algorithmic situation with tree-width is as follows:

- Numerous problems have been shown to be solvable in polynomial time when the input graph is presented together with a decomposition of bounded tree-width. Indeed, every

[^0]graph property expressible in monadic second order logic with quantifications over vertices, vertex sets, edges, and edge sets ( $\mathrm{MSO}_{2}$-logic) can be solved in polynomial time (see [6]).

- For fixed $k$ there is a polynomial time algorithm [18] that either decides that an input graph has tree-width at least $k+1$, or outputs a decomposition of tree-width at most $4 k$.
- Consequently, by combining these algorithms, it follows that the same class of problems mentioned above can be solved on inputs of bounded tree-width; the input does not need to come equipped with the corresponding decomposition.
- In particular, one of these problems is the problem of deciding whether a graph has tree-width at most $k$. Consequently, for fixed $k$ there is a polynomial (indeed, linear) time algorithm [1] to test whether an input graph has tree-width at most $k$, and if so to output the corresponding decomposition.
For inputs of bounded clique-width, less progress has so far been made. (We will define clique-width properly later.)
- Some problems have been shown to be solvable in polynomial time when the input graph is presented together with a decomposition of bounded clique-width. This class of problems is smaller than the corresponding set for tree-width, but still of interest. For instance, deciding whether the graph is Hamiltonian [22], finding the chromatic number [14], and various partition problems [8] are solvable in polynomial time; and so is any problem that can be expressed in monadic second order logic with quantifications over vertices and vertex sets ( $\mathrm{MSO}_{1}$-logic; see $[3,6]$ ).
- For fixed (general) $k$ there was so far no known polynomial time algorithm that either decides that an input graph has clique-width at least $k+1$, or outputs a decomposition of clique-width bounded by any function of $k$. The best hitherto was an algorithm of Johansson [13], that with input an $n$-vertex graph $G$, either decides that $G$ has cliquewidth at least $k+1$ or outputs a decomposition of clique-width at most $k \log n$. Our main result fills this gap.
- Consequently, it follows that the same class of problems mentioned above can be solved on inputs of bounded clique-width; the input does not need to come equipped with the corresponding decomposition.
- However, the problem of deciding whether a graph has clique-width at most $k$ is not known to belong to this class. There is still no polynomial time algorithm to test whether $G$ has clique-width at most $k$, for fixed general $k$.
We shall prove the following.
Theorem 1.1. For fixed $k$, there is an algorithm that with input an $n$-vertex graph $G$, either decides that $G$ has clique-width at least $k+1$, or outputs a decomposition of $G$ with clique-width at most $2^{3 k+2}-1$. Its running time is $O\left(n^{9} \log n\right)$.

The main tool for this algorithm is branch-width, which is closely related to tree-width, and was introduced in [19]. We develop a general algorithm to approximate the branch-width of certain symmetric submodular functions. Then we define the "rank-width" of a graph to be the branch-width of a symmetric submodular function determined by a graph; and since
our algorithm applies to this submodular function, we can approximate the rank-width of a graph in polynomial time. But we also prove that if clique-width is bounded, then rank-width is bounded, and vice versa; and consequently we can approximate clique-width in polynomial time.

We also apply this algorithm to matroids, and obtain an algorithm to approximate the branch-width of matroids, which was known before only for representable matroids by Hliněný [11]. We prove:
Theorem 1.2. For fixed $k$ there is an algorithm which, with input an n-element matroid $\mathcal{M}$ in terms of its rank oracle, either decides that $\mathcal{M}$ has branch-width at least $k+1$, or outputs a branch-decomposition for $\mathcal{M}$ of width at most $3 k-1$. Its running time and number of oracle calls is at most $O\left(n^{4}\right)$.

## 2. Branch-width

Let $V$ be a finite set and $f: 2^{V} \rightarrow \mathbb{Z}$ be a function. If

$$
f(X)+f(Y) \geq f(X \cap Y)+f(X \cup Y)
$$

for all $X, Y \subseteq V$, then $f$ is said to be submodular. If $f$ satisfies $f(X)=f(V \backslash X)$ for all $X \subseteq V$, then $f$ is said to be symmetric.

A subcubic tree is a tree with at least two vertices such that every vertex is incident with at most three edges. A leaf of a tree is a vertex incident with exactly one edge. We call $(T, L)$ a partial branch-decomposition of a symmetric submodular function $f$ if $T$ is a subcubic tree and $L: V \rightarrow\{v: v$ is a leaf of $T\}$ is a surjective function. (If $|V| \leq 1$ then $f$ admits no partial branch-decomposition.) If in addition $L$ is bijective, we call ( $T, L$ ) a branch-decomposition of $f$. If $L(v)=t$, then we say $t$ is labeled by $v$ and $v$ is a label of $t$.

For an edge $e$ of $T$, the connected components of $T \backslash e$ induce a partition $(X, Y)$ of the set of leaves of $T$. The width of an edge $e$ of a partial branch-decomposition $(T, L)$ is $f\left(L^{-1}(X)\right)$. The width of $(T, L)$ is the maximum width of all edges of $T$. The branch-width $\mathrm{bw}(f)$ of $f$ is the minimum width of a branch-decomposition of $f$. (If $|V| \leq 1$, we define $\operatorname{bw}(f)=f(\emptyset)$.)

For the application to matroids, we assume that the reader is familiar with the basic notions of matroid theory (see [16]). Let us review matroid theory briefly for the purpose of this paper.

A matroid $\mathcal{M}=(E, r)$ is a pair formed by a finite set $E$ of elements and a rank function $r: 2^{E} \rightarrow \mathbb{Z}$ satisfying the following axioms:
i) $0 \leq r(X) \leq|X|$ for all $X \subseteq E$.
ii) If $X \subseteq Y \subseteq E$, then $r(X) \leq r(Y)$.
iii) $r$ is submodular.

We write $E(\mathcal{M})=E$. For $Y \subseteq E(\mathcal{M}), \mathcal{M} \backslash Y$ is the matroid $\left(E(\mathcal{M}) \backslash Y, r^{\prime}\right)$ where $r^{\prime}(X)=$ $r(X)$. For $X \subseteq E(\mathcal{M}), \mathcal{M} / Y$ is the matroid $\left(E(\mathcal{M}) \backslash Y, r^{\prime}\right)$ where $r^{\prime}(X)=r(X \cup Y)-r(Y)$. If $Y=\{e\}$, we denote $\mathcal{M} \backslash e=\mathcal{M} \backslash\{e\}$ and $\mathcal{M} / e=\mathcal{M} /\{e\}$. It is routine to prove that $\mathcal{M} \backslash Y$ and $\mathcal{M} / Y$ are matroids.

For $X \subseteq E, \lambda(X)=r(X)+r(E(\mathcal{M}) \backslash X)-r(\mathcal{M})+1$ is the connectivity function of $\mathcal{M}$. A branch-decomposition and the branch-width of a matroid $\mathcal{M}$ are defined as a branchdecomposition and the branch-width of $\lambda$.

## 3. Clique-width

The notion of clique-width was first introduced by Courcelle and Olariu [7]. Let $k$ be a positive integer. We call $(G, l a b)$ a $k$-graph if $G$ is a graph and lab is a mapping from its vertex set to $\{1,2, \ldots, k\}$. (In this paper, all graphs are finite and have no loops or parallel edges.) We call $l a b(v)$ the label of a vertex $v$.

We need the following definitions and operations on $k$-graphs.
(1) For $i \in\{1, \ldots, k\}$, let $\cdot_{i}$ denote an isolated vertex labeled by $i$.
(2) For $i, j \in\{1,2, \ldots, k\}$ with $i \neq j$, we define a unary operator $\eta_{i, j}$ such that

$$
\eta_{i, j}(G, l a b)=\left(G^{\prime}, l a b\right)
$$

where $V\left(G^{\prime}\right)=V(G)$, and $E\left(G^{\prime}\right)=E(G) \cup\{v w: v, w \in V, l a b(v)=i, l a b(w)=j\}$. This adds edges between vertices of label $i$ and vertices of label $j$.
(3) We let $\rho_{i \rightarrow j}$ be the unary operator such that

$$
\rho_{i \rightarrow j}(G, l a b)=\left(G, l a b^{\prime}\right)
$$

where

$$
l a b^{\prime}(v)= \begin{cases}j & \text { if } \operatorname{lab}(v)=i \\ \operatorname{lab}(v) & \text { otherwise }\end{cases}
$$

This mapping relabels every vertex labeled by $i$ into $j$.
(4) Finally, $\oplus$ is a binary operation that makes the disjoint union. Note that $G \oplus G \neq G$.

A well-formed expression $t$ in these symbols is called a $k$-expression. The $k$-graph produced by performing these operations in order therefore has vertex set the set of occurrences of the constant symbols in $t$; and this $k$-graph (and any $k$-graph isomorphic to it) is called the value $\operatorname{val}(t)$ of $t$. If a $k$-expression $t$ has value $(G, l a b)$, we say that $t$ is a $k$-expression of $G$. The clique-width of a graph $G$, denoted by $\operatorname{cwd}(G)$, is the minimum $k$ such that there is a $k$-expression of $G$.

For instance, $K_{4}$ (the complete graph with four vertices) can be constructed by

$$
\rho_{2 \rightarrow 1}\left(\eta_{1,2}\left(\rho_{2 \rightarrow 1}\left(\eta_{1,2}\left(\rho_{2 \rightarrow 1}\left(\eta_{1,2}\left({ }_{1} \oplus \cdot \cdot_{2}\right)\right) \oplus \cdot \cdot_{2}\right)\right) \oplus \cdot \cdot_{2}\right)\right) .
$$

Therefore, $K_{4}$ has a 2-expression, and $\operatorname{cwd}\left(K_{4}\right) \leq 2$. It is easy to see that $\operatorname{cwd}\left(K_{4}\right)>1$, and therefore $\operatorname{cwd}\left(K_{4}\right)=2$.

Some other examples: cographs, which are graphs with no induced path of length 3, are exactly the graphs of clique-width at most 2 ; the complete graph $K_{n}(n>1)$ has clique-width 2 ; and trees have clique-width at most 3 [7].

For some classes of graphs, it is known that clique-width is bounded and algorithms to construct a $k$-expression have been found. For example, cographs [4], graphs of clique-width at most 3 [2], and $P_{4}$-sparse graphs (every five vertices have at most one induced subgraph isomorphic to a path of length 3) [3] have such algorithms.

## 4. Extension of a submodular function

In this section, we define an "interpolation" of an integer-valued submodular function. Later we will use it to prove the main theorem.

For a finite set $V$, we define (with a slight abuse of terminology) $3^{V}$ to be $\{(X, Y): X, Y \subseteq$ $V, X \cap Y=\emptyset\}$.

Definition 4.1. Let $f: 2^{V} \rightarrow \mathbb{Z}$ be an integer-valued submodular function such that $f(\emptyset) \leq$ $f(X)$ for all $X \subseteq V$. We call $f^{*}: 3^{V} \rightarrow \mathbb{Z}$ an interpolation of $f$ if
i) $f^{*}(X, V \backslash X)=f(X)$ for all $X \subseteq V$,
ii) (uniform) if $C \cap D=\emptyset, A \subseteq C$, and $B \subseteq D$, then $f^{*}(A, B) \leq f^{*}(C, D)$,
iii) (submodular) $f^{*}(A, B)+f^{*}(C, D) \geq f^{*}(A \cap C, B \cup D)+f^{*}(A \cup C, B \cap D)$ for all $(A, B),(C, D) \in 3^{V}$.
iv) $f^{*}(\emptyset, \emptyset)=f(\emptyset)$.

Assuming that $\emptyset$ is a minimizer of $f$ is not a serious restriction, because first of all it is true for all symmetric submodular functions, and secondly if we let

$$
g(X)= \begin{cases}f(X) & \text { if } X \neq \emptyset \\ \min _{Z} f(Z) & \text { otherwise }\end{cases}
$$

then $g$ is also submodular.
Proposition 4.1. Let $f: 2^{V} \rightarrow \mathbb{Z}$ be a submodular function such that $f(\emptyset) \leq f(X)$ for all $X \subseteq V$, and let $f^{*}: 3^{V} \rightarrow \mathbb{Z}$ be an interpolation of $f$. Then:
(1) for all $(X, Y) \in 3^{V} \rightarrow, f^{*}(X, Y) \leq \min _{X \subseteq Z \subseteq V \backslash Y} f(Z)$.
(2) $f^{*}(\emptyset, Y)=f(\emptyset)$ for all $Y \subseteq V$.
(3) If $f(\{v\})-f(\emptyset) \leq 1$ for every $v \in V$, then for every fixed $B \subseteq V, f^{*}(X, B)-f(\emptyset)$ is the rank function of a matroid on $V \backslash B$.
Proof.
(1) If $X \subseteq Z \subseteq V \backslash Y$, then $f^{*}(X, Y) \leq f^{*}(Z, V \backslash Z)=f(Z)$.
(2) $f(\emptyset)=f^{*}(\emptyset, \emptyset) \leq f^{*}(\emptyset, Y) \leq f^{*}(\emptyset, V)=f(\emptyset)$.
(3) Let $r(X)=f^{*}(X, B)-f(\emptyset)$. It is trivial that $r$ is submodular and nondecreasing. Moreover,

$$
0 \leq r(X)=f^{*}(X, B)-f(\emptyset) \leq f(X)-f(\emptyset) \leq|X|
$$

and therefore $r$ is the rank function of a matroid on $V \backslash B$.
We define $f_{\min }(X, Y)=\min f(Z)$, the minimum being taken over all $Z$ satisfying $X \subseteq Z \subseteq$ $V \backslash Y$.

Proposition 4.2. Let $f: 2^{V} \rightarrow \mathbb{Z}$ be a submodular function such that $f(\emptyset) \leq f(X)$ for all $X \subseteq V$. Then $f_{\min }$ is an interpolation of $f$.

Proof. The first, second, and last conditions are trivial. Let us prove submodularity. Let $X$, $Y$ be subsets of $V$ such that $A \subseteq X \subseteq V \backslash B, C \subseteq Y \subseteq V \backslash D, f_{\min }(A, B)=f(X)$, and $f_{\text {min }}(C, D)=f(Y)$. Then

$$
\begin{aligned}
f(X)+f(Y) & \geq f(X \cap Y)+f(X \cup Y) \\
& \geq f_{\min }(A \cap C, B \cup D)+f_{\min }(A \cup C, B \cap D)
\end{aligned}
$$

Thus, $f_{\text {min }}$ is an interpolation.

In general $f_{\text {min }}$ is not the only interpolation of a function $f$, and sometimes it is better for us to work with other interpolations that can be evaluated more quickly.

We remark that if $f^{*}: 3^{V} \rightarrow \mathbb{Z}$ is a uniform submodular function satisfying $f^{*}(\emptyset, \emptyset)=$ $f^{*}(\emptyset, V)$, then there is a submodular function $f: 2^{V} \rightarrow \mathbb{Z}$ such that $f(\emptyset) \leq f(X)$ for all $X \subseteq V$ and $f^{*}$ is an interpolation of $f$.

## 5. Branch-Width and Well-Linkedness

Definition 5.1. Let $V$ be a finite set and let $f: 2^{V} \rightarrow \mathbb{Z}$ be a symmetric submodular function satisfying $f(\emptyset)=0$. We say that $W \subseteq V$ is well-linked with respect to $f$ if for every partition $(X, Y)$ of $W$ and every $Z$ with $X \subseteq Z \subseteq V \backslash Y$, we have

$$
f(Z) \geq \min (|X|,|Y|)
$$

This notion is analogous to the notion of well-linkedness [17] related to tree-width of graphs.
Theorem 5.1. Let $V$ be a finite set with $|V| \geq 2$, and let $f: 2^{V} \rightarrow \mathbb{Z}$ be a symmetric submodular function such that $f(\emptyset)=0$. If with respect to $f$ there is a well-linked set of size $k$, then $\operatorname{bw}(f) \geq k / 3$.
Proof. Let $W$ be a well-linked set of size $k$, and suppose that $(T, L)$ is a branch decomposition of $f$. We will show that $(T, L)$ has width at least $k / 3$. We may assume that $T$ does not have a vertex of degree 2 , by suppressing any such vertices. For each edge $e=u v$ of $T$, let $A_{u v}$ be the set of elements of $V$ that are mapped by $L$ into the connected component of $T \backslash e$ containing $u$, and let $B_{u v}=V \backslash A_{u v}$.

We may assume that $W \neq \emptyset$; choose $w \in W$. Since $W$ is well-linked with respect to $f$, $f(\{w\}) \geq 1$, and therefore the width of $(T, L)$ is at least 1 . Consequently we may assume that assume $k>3$.

Suppose first that $\min \left(\left|A_{u v} \cap W\right|,\left|B_{u v} \cap W\right|\right)<k / 3$ for every edge $u v$ of $T$. Direct every edge $u v$ from $u$ to $v$ if $\left|A_{u v} \cap W\right|<k / 3$ and $\left|B_{u v} \cap W\right| \geq k / 3$. By the assumption, each edge is given a unique direction. Since the number of vertices is more than the number of edges in $T$, there is a vertex $t \in V(T)$ such that every edge incident with $t$ has head $t$.

If $t$ is a leaf of $T$, let $s$ be the neighbour of $t$. Since $t s$ has head $t$, it follows that $\left|B_{s t} \cap W\right| \geq$ $k / 3$. But $\left|B_{s t}\right|=1<k / 3$, a contradiction.

So, $t$ has three neighbours $x, y, z$ in $T$ such that $\left|A_{x t} \cap W\right|<k / 3,\left|A_{y t} \cap W\right|<k / 3$, and $\left|A_{z t} \cap W\right|<k / 3$. But $|W|=\left|A_{x t} \cap W\right|+\left|A_{y t} \cap W\right|+\left|A_{z t} \cap W\right|<k=|W|$, a contradiction.

We deduce that there exists $u v \in E(T)$ such that $\left|A_{u v} \cap W\right| \geq k / 3$ and $\left|B_{u v} \cap W\right| \geq k / 3$. Hence $f\left(A_{u v}\right) \geq \min \left(\left|A_{u v} \cap W\right|,\left|B_{u v} \cap W\right|\right) \geq k / 3$, and the width of $(T, L)$ is at least $k / 3$.

Theorem 5.2. Let $V$ be a finite set, let $f: 2^{V} \rightarrow \mathbb{Z}$ be a symmetric submodular function such that $f(\{v\}) \leq 1$ for all $v \in V$ and $f(\emptyset)=0$, and let $k \geq 0$ be an integer. If with respect to $f$, there is no well-linked set of size $k$, then $\mathrm{bw}(f) \leq k$.
Proof. We may assume that $\mathrm{bw}(f)>0$, and so $|V| \geq 2$. We may assume that $k>0$. For two partial branch-decompositions $(T, L)$ and $\left(T^{\prime}, L^{\prime}\right)$ of $f$, we say that $(T, L)$ extends $\left(T^{\prime}, L^{\prime}\right)$ if $T^{\prime}$ is obtained by contracting some edges of $T$ and for every $v \in V, L^{\prime}(v)$ is the vertex of $T^{\prime}$ that corresponds to $L(v)$ under the contraction.

We will prove that, if there is no well-linked set of size $k$ with respect to $f$, then for every partial branch-decomposition $\left(T_{s}, L_{s}\right)$ of $f$ with width at most $k$, there is a branchdecomposition of $f$ of width at most $k$ extending $\left(T_{s}, L_{s}\right)$. Since $k \geq 1$ and $f$ trivially admits a partial branch-decomposition of width 1 (using the two-vertex tree with vertices $u$, $v$, and mapping all vertices of $V$ except one to $u$, and the last to $v$ ), this implies the statement of the theorem.

Pick a partial branch-decomposition $(T, L)$ of $f$ extending $\left(T_{s}, L_{s}\right)$ such that the width of $(T, L)$ is at most $k$ and the number of leaves of $T$ is maximum.

We claim that $(T, L)$ is a branch-decomposition of $f$, that is, $L$ is a bijection. Suppose therefore that there is a leaf $t$ of $T$ such that $B=L^{-1}(\{t\})$ has more than one element.
(1) $f(B)=k$.

Suppose that $f(B)<k$. Let $v \in B$. Construct a subcubic tree $T^{\prime}$ by adding two vertices $t_{1}$ and $t_{2}$ and edges $t_{1} t, t_{2} t$ to $T$. Let $L^{\prime}(v)=t_{1}$ and $L^{\prime}(w)=t_{2}$ for all $w \in B \backslash\{v\}$ and $L^{\prime}(x)=L(x)$ for all $x \in V \backslash B$. Then $\left(T^{\prime}, L^{\prime}\right)$ is a partial branch-decomposition extending $(T, L)$. Moreover $f(\{v\}) \leq 1 \leq k$ and $f(B \backslash\{v\}) \leq f(B)+f(\{v\}) \leq k$, and so the width of $\left(T^{\prime}, L^{\prime}\right)$ is at most $k$. But the number of leaves of $T^{\prime}$ is greater than that of $T$, a contradiction.

Let $f^{*}$ be an interpolation of $f$. By Proposition 4.1, $f^{*}(X, B)$ is the rank function of a matroid on $V \backslash B$. Let $X$ be a base of this matroid. Then $|X|=f^{*}(V \backslash B, B)=f(B)=k$.

Since $X$ is not well-linked, there exists $Z \subseteq V$ such that

$$
f(Z)<\min (|Z \cap X|,|(V \backslash Z) \cap X|)
$$

Since $f(Z \backslash B)=f^{*}(Z \backslash B, B \cup(V \backslash Z)) \geq f^{*}(Z \cap X, B)=|Z \cap X|>f(Z)$, it follows that $Z \cap B \neq \emptyset$. Similarly $B \backslash Z=(V \backslash Z) \cap B \neq \emptyset$.

Construct a subcubic tree $T^{\prime}$ by adding two vertices $t_{1}$ and $t_{2}$ and edges $t_{1} t, t_{2} t$ to $T$. Let $L^{\prime}(x)=t_{1}$ if $x \in B \cap Z, L^{\prime}(x)=t_{2}$ if $x \in B \backslash Z$ and $L^{\prime}(x)=L(x)$ otherwise.

By submodularity,

$$
\begin{aligned}
|(V \backslash Z) \cap X|+f(B) & >f(Z)+f(B) \geq f(Z \cup B)+f(Z \cap B) \\
& =f((V \backslash Z) \backslash B)+f(Z \cap B) \\
& \geq f^{*}((V \backslash Z) \cap X, B)+f(Z \cap B) \\
& =|(V \backslash Z) \cap X|+f(Z \cap B),
\end{aligned}
$$

and so $f(Z \cap B)<f(B) \leq k$ and similarly $f(B \backslash Z)<f(B) \leq k$. Therefore $\left(T^{\prime}, L^{\prime}\right)$ is a partial branch-decomposition extending $(T, L)$ of width at most $k$. But the number of leaves of $T^{\prime}$ is greater than that of $T$, a contradiction.

Corollary 5.3. For all $k \geq 0$, there is a polynomial-time algorithm that, with input a set $V$ with $|V| \geq 2$ and a symmetric submodular function $f: 2^{V} \rightarrow \mathbb{Z}$ with $f(\{v\}) \leq 1$ for all $v \in V$ and $f(\emptyset)=0$, outputs either a well-linked set of size $k$ or a branch-decomposition of width at most $k$.

The proof of Theorem 5.2 shows an algorithm that either finds a well-linked set of size $k$, or constructs a branch-decomposition of $f$ of width at most $k$. By combining with Theorem 5.1, we get an algorithm that either concludes that $\mathrm{bw}(f)>k$ or finds a branch-decomposition of width at most $3 k+1$.

Let us analyze the running time of the algorithm of Theorem 5.2. To do so, we must be more precise about how the input function $f$ and $f^{*}$ are accessed. We consider two different situations, as follows:

- In the first case, we assume that only $f$ is given as input, and in the sense that we can compute $f(X)$ for a set $X$; and we need to compute values of $f^{*}$ from this input.
- In the second case, we assume that an interpolation $f^{*}$ of $f$ is given as input (in the same sense, that for any pair $(X, Y)$ we can compute $f^{*}(X, Y)$ ), and we need to compute $f$ from $f^{*}$.
For the first analysis, let $\gamma$ be the time to compute $f(X)$ for any set $X$. In this case we shall use $f^{*}=f_{\min }$. To calculate $f_{\min }$, we use the submodular function minimization algorithm [12], whose running time is $O\left(n^{5} \gamma \log M\right)$ where $M$ is the maximum value of $f$ and $n=|V|$. Thus, we can calculate $f_{\min }$ in $O\left(n^{5} \gamma \log n\right)$ time. Finding a base $X$ can be done by calculating $f^{*}$ at most $O(n)$ times, and therefore takes time $O\left(n^{6} \gamma \log n\right)$. To check whether $X$ is well-linked, we try all partitions of $X ; 2^{k-1}$ tries (a constant). And finding the set $Z$ for a given partition of $X$ can be done in time $O\left(n^{5} \gamma \log n\right)$ by submodular function minimization algorithms. Since the process is cycled through at most $O(n)$ times (because if $(T, L)$ is a partital branch-decomposition then $|V(T)| \leq 2 n-2)$, it follows that in this case the time complexity is $O\left(n^{7} \gamma \log n\right)$.

For the second analysis, let $\delta$ be the time to compute $f^{*}(X)$ for any set $X$. Finding a base $X$ can be done in time $O(n \delta)$. Finding $Z$ to show that $X$ is not well-linked can be done in time $O\left(n^{5} \delta \log n\right)$. Thus, the time complexity in this case is $O\left(n^{6} \delta \log n\right)$.

In summary, then, we have shown the following two statements.
Corollary 5.4. For given $k$, there is an algorithm as follows. It takes as input a finite set $V$ with $|V| \geq 2$ and a symmetric submodular function $f: 2^{V} \rightarrow \mathbb{Z}$, such that $f(\{v\}) \leq 1$ for all $v \in V$ and $f(\emptyset)=0$. It either concludes that $\mathrm{bw}(f)>k$ or outputs a branch-decomposition of $f$ of width at most $3 k+1$; and its running time (excluding evaluating $f$ ) and number of evaluations of $f$ are both $O\left(|V|^{7} \log |V|\right)$.

Corollary 5.5. For given $k$, there is an algorithm as follows. It takes as input a finite set $V$ with $|V| \geq 2$ and a function $f^{*}$ which is an interpolation of some symmetric submodular function $f: 2^{V} \rightarrow \mathbb{Z}$, such that $f(\{v\}) \leq 1$ for all $v \in V$ and $f(\emptyset)=0$. It either concludes that $\operatorname{bw}(f)>k$ or outputs a branch-decomposition of $f$ of width at most $3 k+1$; and its running time is $O\left(|V|^{6} \delta \log |V|\right)$, where $\delta$ is the time for each evaluation of $f^{*}$.

## 6. Application to Clique-width

Definition 6.1. Let $G$ be a graph and let $A, B \subseteq V(G)$ be disjoint. Let $M_{A}^{B}(G)$ be the matrix ( $m_{i j}: i \in A, j \in B$ ) over the 2-element field $\operatorname{GF}(2)$, where $m_{i j}=1$ if $i, j$ are adjacent in $G$, and $m_{i j}=0$ otherwise. We define $\operatorname{cutrk}_{G}^{*}(A, B)=\operatorname{rk}\left(M_{A}^{B}(G)\right)$ where rk is the matrix rank function; and we define the cut-rank function cutrk ${ }_{G}$ of $G$ by $\operatorname{cutrk}_{G}(X)=\operatorname{cutrk}_{G}^{*}(X, V(G) \backslash X)$ for
$X \subseteq V(G)$. We will show that $\operatorname{cutrk}_{G}$ is symmetric submodular and cutrk ${ }_{G}^{*}$ is an interpolation of $\operatorname{cutrk}_{G}$.

Proposition 6.1. Let $M=\left(m_{i j}: i \in C, j \in R\right)$ be a matrix over a field $F$. For $X \subseteq R$ and $Y \subseteq C$, let $M[X, Y]$ denote the submatrix $\left(m_{i j} ; i \in X, j \in Y\right)$. Then for all $X_{1}, X_{2} \subseteq R$ and $Y_{1}, Y_{2} \subseteq C$, we have

$$
\operatorname{rk}\left(M\left[X_{1}, Y_{1}\right]\right)+\operatorname{rk}\left(M\left[X_{2}, Y_{2}\right]\right) \geq \operatorname{rk}\left(M\left[X_{1} \cup X_{2}, Y_{1} \cap Y_{2}\right]\right)+\operatorname{rk}\left(M\left[X_{1} \cap X_{2}, Y_{1} \cup Y_{2}\right]\right)
$$

Proof. See [15, Proposition 2.1.9], [21, Lemma 2.3.11], or [20].
Corollary 6.2. Let $G$ be a graph. If $\left(X_{1}, Y_{1}\right),\left(X_{2}, Y_{2}\right) \in 3^{V(G)}$ then

$$
\operatorname{cutrk}_{G}^{*}\left(X_{1}, Y_{1}\right)+\operatorname{cutrk}_{G}^{*}\left(X_{2}, Y_{2}\right) \geq \operatorname{cutrk}_{G}^{*}\left(X_{1} \cap X_{2}, Y_{1} \cup Y_{2}\right)+\operatorname{cutrk}_{G}^{*}\left(X_{1} \cup X_{2}, Y_{1} \cap Y_{2}\right)
$$

Moreover, if $X_{1}, X_{2} \subseteq V(G)$, then

$$
\operatorname{cutrk}_{G}\left(X_{1}\right)+\operatorname{cutrk}_{G}\left(X_{2}\right) \geq \operatorname{cutrk}_{G}\left(X_{1} \cap X_{2}\right)+\operatorname{cutrk}_{G}\left(X_{1} \cup X_{2}\right)
$$

Proof. Let $M$ be the $V(G) \times V(G)$ adjacency matrix of $G$ over $\mathrm{GF}(2)$. The first statement follows from 6.1 applied to $M$. The second follows from the first by setting $Y_{i}=V(G) \backslash X_{i}(i=$ $1,2)$.

A rank-decomposition of $G$ is a branch-decomposition of cutrk $_{G}$, and the rank-width $\operatorname{rwd}(G)$ of $G$ is the branch-width of cutrk $_{G}$.

The following proposition shows a relation between clique-width and rank-width.
Proposition 6.3. For any graph $G, \operatorname{rwd}(G) \leq \operatorname{cwd}(G) \leq 2^{\mathrm{rwd}(G)+1}-1$.
Proof. We may assume that $|V(G)| \geq 2$, because if $|V(G)| \leq 1$, then $\operatorname{rwd}(G)=0$ and $\operatorname{cwd}(G) \leq 1$.

A rooted binary tree is a subcubic tree with a specified vertex, called the root, such that every non-root vertex has one, two or three incident edges and the root has at most two incident edges. A vertex $u$ of a rooted binary tree is called a descendant of a vertex $v$ if $v$ belongs to the path from the root to $u$; and $u$ is called a child of $v$ if $u, v$ are adjacent in $T$ and $u$ is a descendant of $v$.

First we show that $\operatorname{rwd}(G) \leq \operatorname{cwd}(G)$. Let $k=\operatorname{cwd}(G)$. Let $t$ be a $k$-expression with value ( $G, l a b$ ) for some choice of lab. We recall that a $k$-expression is a well-formed expression with four types of symbols; the constants, two unary operators, and the binary operator forming disjoint union. The parentheses of the expression form a tree structure. Thus there is a rooted binary tree $T$, each vertex $v$ of which corresponds to a $k$-expression say $N(v)$; and letting $V_{0}, V_{1}, V_{2}$ denote the sets of vertices in $T$ with zero, one and two children respectively, we have for each vertex $v \in V(T)$ :

- if $v \in V_{0}$ then $N(v)$ is a 1-term expression consisting just of a constant term
- if $v \in V_{1}$ with child $u$, then $N(v)$ is obtained from $N(u)$ by applying one of the two unary operators
- if $v \in V_{2}$ with children $u_{1}, u_{2}$, then $N(v)$ is obtained from $N\left(u_{1}\right), N\left(u_{2}\right)$ by applying $\oplus$
- if $v$ is the root then $N(v)=(G, l a b)$.

In particular, each vertex $v \in V_{0}$ gives rise to a unique vertex of $G$; let us call this $L(v)$. Then $L$ is a bijection between $V(G)$ and the set of leaves of $T$. Consequently $(T, L)$ is a branch-decomposition of cutrk $_{G}$. Let us study its width. Let $u, v \in V(T)$, where $u$ is a child of $v$, and let $T_{1}, T_{2}$ be the components of $T \backslash e$, where $e$ is the edge $u v$ and $u \in V\left(T_{1}\right)$. Let $X_{i}=\left\{L(t): t \in V_{0} \cap V\left(T_{i}\right)\right\}$ for $i=1,2$. Thus $\left(X_{1}, X_{2}\right)$ is a partition of $V(G)$, and we need to investigate $\operatorname{cutrk}_{G}\left(X_{1}\right)$. Let $N(u)=\left(G_{1}, l a b_{1}\right)$. Thus $V\left(G_{1}\right)=X_{1}$. If $x, y \in X_{1}$, and $l a b_{1}(x)=l a b_{1}(y)$, then $x, y$ are adjacent in $G$ to the same members of $X_{2}$, from the properties of the iterative construction of ( $G, l a b$ ); and since the function $l a b_{1}$ has at most $k$ different values, it follows that $X_{1}$ can be partitioned into $k$ subsets so that the members of each subset have the same neighbours in $X_{2}$. Consequently $\operatorname{cutr}_{G}\left(X_{1}\right) \leq k$. Since this applies for every edge of $T$, we deduce that $(T, L)$ is a branch-decomposition of cutr $_{G}$ with width at most $k$. Hence $\operatorname{rwd}(G) \leq k=\operatorname{cwd}(G)$.

Now we show the second statement of the theorem, that $\operatorname{cwd}(G) \leq 2^{\mathrm{rwd}(G)+1}-1$. Let $k=\operatorname{rwd}(G)$ and $(T, L)$ be a rank-decomposition of $G$ of width $k$. By subdividng one edge of $T$, and suppressing all other vertices of $T$ with degree 2 , we may assume that $T$ is a rooted binary tree; its root has degree 2 , and all other vertices have degree 1 or 3 .

For $v \in V(T)$, let $D_{v}=\{x \in V(G): L(x)$ is a descendant of $v$ in $T\}$, and let $G_{v}$ denote the subgraph of $G$ induced on $D_{v}$. We claim that for every $v \in V(T)$, there is a map $l a b_{v}$ and a $\left(2^{k+1}-1\right)$-expression $t_{v}$ with value $\left(G_{v}, l a b_{v}\right)$, such that
(i) if $l a b_{v}(x)=1$ then $x \in D_{v}$ is nonadjacent to every vertex of $G \backslash D_{v}$,
(ii) if $x, y \in D_{v}$ and there exists $z \in V(G) \backslash D_{v}$ such that $x$ is adjacent to $z$ but $y$ is not, then $l a b_{v}(x) \neq l a b_{v}(y)$,
(iii) for each $x \in D_{v}, l a b_{v}(x) \in\left\{1,2, \ldots, 2^{k}\right\}$.

We prove this by induction on the number of vertices of $T$ that are descendants of $v$. If $v$ is a leaf, let $t_{v}=\cdot_{1}$. Then $t_{v}$ satisfies the above conditions. Thus we may assume that $v$ has exactly two children $v_{1}, v_{2}$.

By the inductive hypothesis, there are $\left(2^{k+1}-1\right)$-expressions $t_{1}, t_{2}$ with values $\left(G_{v_{i}}, l a b_{v_{i}}\right)$ for $i=1,2$, satisfying the statements above. Let $F$ be the set of pairs $(i, j)$ with $i, j \in$ $\left\{1,2, \ldots, 2^{k}\right\}$, such that there is an edge $x y$ of $G$, with $x \in D_{v_{1}}, l a b_{v_{1}}(x)=i, y \in D_{v_{2}}$ and $l a b_{v_{2}}(y)=j$. It follows from the second condition above that if $(i, j) \in F$ then every vertex $x \in D_{v_{1}}$ with $\operatorname{lab}_{v_{1}}(x)=i$ is adjacent in $G$ to every vertex $y \in D_{v_{2}}$ with $\operatorname{lab}_{v_{2}}(y)=j$. Let

$$
t^{*}=\left(\begin{array}{cc}
0 \\
(i, j) \in F
\end{array} \eta_{i, j+2^{k}-1}\right)\left(t_{v_{1}} \oplus\left(\begin{array}{c}
2^{k} \\
i=2
\end{array} \rho_{i \rightarrow i+2^{k}-1}\right)\left(t_{v_{2}}\right)\right) .
$$

Then $t^{*}$ is a $\left(2^{k+1}-1\right)$-expression with value $\left(G_{v}, l a b^{*}\right)$ say, and it satisfies the first two displayed conditions above. However, it need not yet satisfy the third. Let us choose a $\left(2^{k+1}-1\right)$-expression $t_{v}$ with value $\left(G_{v}, l a b_{v}\right)$ say, satisfying the first two conditions above, and satisfying the following:

- $\left\{l a b_{v}(x): x \in D_{v}\right\}$ is minimal
- subject to this condition, $\max \left(l a b_{v}(x): x \in D_{v}\right)(=r$ say $)$ is as small as possible.
(We call these the "first and second optimizations".) For $i=1, \ldots, r$ let $X_{i}=\left\{x \in D_{v}\right.$ : $\left.l a b_{v}(x)=i\right\}$. The definition of $r$ implies that $X_{r} \neq \emptyset$. If there exists $i$ with $2 \leq i<r$ such that $X_{i}=\emptyset$, then applying the function $\rho_{r \rightarrow i}$ to $t_{v}$ produces a $k$-expression contradicting the
second optimization. Thus, $X_{2}, \ldots, X_{r}$ are all nonempty. For $1 \leq i \leq r$ let $Y_{i}$ be the set of vertices of $V(G) \backslash D_{v}$ with a neighbour in $X_{i}$. From the first condition above, $Y_{1}=\emptyset$. From the second condition above, every vertex in $X_{i}$ is adjacent to every member of $Y_{i}$ for all $i$ with $1 \leq i \leq r$. If there exist $i, j$ with $1 \leq i<j \leq r$ such that $Y_{i}=Y_{j}$, then applying $\rho_{j \rightarrow i}$ to $t_{v}$ produces a $k$-expression contradicting the first optimization. Thus $Y_{1}, \ldots, Y_{r}$ are all distinct.

Let $M$ be the matrix $\left(m_{i j}: i \in D_{v}, j \in V(G) \backslash D_{v}\right)$, where $m_{i j}=1$ if $i, j$ are adjacent and 0 otherwise. Then $M$ has $r-1$ distinct nonzero rows. Since $(T, L)$ has width $k$, it follows that $M$ has rank at most $k$, and therefore $M$ has at most $2^{k}-1$ distinct nonzero rows (this is an easy fact about any matrix over $\mathrm{GF}(2))$. We deduce that $r \leq 2^{k}$, and therefore $t_{v}$ satisfies the third condition above.

This completes the proof that the $k$-expressions $t_{v}$ exist as described above. In particular, if $v$ is the root of $T$ then $G_{v}=G$, and so $t_{v}$ is a $2^{k+1}-1$-expression of $G$. We deduce that $\operatorname{cwd}(G) \leq 2^{k+1}-1$.

The above proof gives an algorithm that converts a rank-decomposition of order $k$ into a $\left(2^{k+1}-1\right)$-expression. Let $n=|V(G)|$, and let $(T, L)$ be the input rank-decomposition. At each non-leaf vertex $v$ of $T$, we first construct $F$, in $O\left(\left(2^{k}\right)^{2}\right)=O(1)$ time. Then merging sets with the same neighbours outside $D_{v}$ will take time $O\left(2^{2 k} n\right)=O(n)$. The number of non-leaf vertices $v$ of $T$ is $O(n)$. Therefore, the time complexity is $O\left(n^{2}\right)$. Note that we may assume that checking the adjacency of two vertices can be done in constant time, because we preprocess the input to construct an adjacency matrix in time $O\left(n^{2}\right)$.
Corollary 6.4. For given $k$, there is an algorithm that, with input an n-vertex graph $G$, either concludes that $\operatorname{rwd}(G)>k$ or outputs a rank-decomposition of width at most $3 k+1$. Its running time is $O\left(n^{9} \log n\right)$.
Proof. cutrk ${ }_{G}^{*}$ can be calculated in time $O\left(n^{3}\right)$, so the claim follows from 5.5.
Corollary 6.5. For given $k$, there is an algorithm that, with input a graph $G$, either concludes that $\operatorname{cwd}(G)>k$ or outputs a $\left(2^{3 k+2}-1\right)$-expression of $G$. Its running time is $O\left(n^{9} \log n\right)$.

Proof. This is immediate from 6.4 and 6.3.

## 7. Application to the Branch-width of a Matroid

In this section, we will show an interpolation of the connectivity function $\lambda$ of a matroid that can be evaluated faster than $\lambda_{\text {min }}$, and a method to apply the matroid intersection theorem to avoid the general submodular function minimization algorithms. So, approximating the branch-width of matroids can be done much faster than that of general symmetric submodular functions.

The following proposition is due to Jim Geelen (private communication).
Proposition 7.1. Let $\mathcal{M}$ be a matroid with rank function r. Let $\lambda(X)=r(X)+r(E(\mathcal{M}) \backslash$ $X)-r(\mathcal{M})+1$ be the connectivity function of $\mathcal{M}$. Let $B$ be a base of $\mathcal{M}$. Then

$$
\lambda_{B}(X, Y)=r(X \cup(B \backslash Y))+r(Y \cup(B \backslash X))-|B \backslash X|-|B \backslash Y|+1
$$

is an interpolation of $\lambda$.

Proof. We verify the three conditions of the definition of an interpolation.

1) If $Y=E(\mathcal{M}) \backslash X$, then

$$
\lambda_{B}(X, Y)=r(X)+r(Y)-r(B \cap X)-r(B \cap Y)+1=r(X)+r(Y)-r(\mathcal{M})+1=\lambda(X)
$$

2) Let $X_{1} \subseteq X_{2}$ and $Y_{1} \subseteq Y_{2}$. Then

$$
r\left(X_{2} \cup\left(B \backslash Y_{2}\right)\right) \geq r\left(X_{1} \cup\left(B \backslash Y_{2}\right)\right) \geq r\left(X_{1} \cup\left(B \backslash Y_{1}\right)\right)-\left(\left|B \backslash Y_{1}\right|-\left|B \backslash Y_{2}\right|\right) .
$$

Therefore,

$$
r\left(X_{2} \cup\left(B \backslash Y_{2}\right)\right)-\left|B \backslash Y_{2}\right| \geq r\left(X_{1} \cup\left(B \backslash Y_{1}\right)\right)-\left|B \backslash Y_{1}\right|
$$

Similarly,

$$
r\left(Y_{2} \cup\left(B \backslash X_{2}\right)\right)-\left|B \backslash X_{2}\right| \geq r\left(Y_{1} \cup\left(B \backslash X_{1}\right)\right)-\left|B \backslash X_{1}\right| .
$$

By adding both inequalities, we deduce that $\lambda_{B}\left(X_{2}, Y_{2}\right) \geq \lambda_{B}\left(X_{1}, Y_{1}\right)$.
3) Let $X_{1} \cap Y_{1}=\emptyset$ and $X_{2} \cap Y_{2}=\emptyset$. It is easy to show that

$$
(P \cap R) \cup(Q \cap S) \subseteq(P \cup Q) \cap(R \cup S)
$$

for any choice of sets $P, Q, R, S$. Since $r$ is submodular and increasing,

$$
\begin{aligned}
& r\left(X_{1} \cup\left(B \backslash Y_{1}\right)\right)+r\left(X_{2} \cup\left(B \backslash Y_{2}\right)\right) \\
\geq & r\left(\left(X_{1} \cup\left(B \backslash Y_{1}\right)\right) \cup\left(X_{2} \cup\left(B \backslash Y_{2}\right)\right)\right)+r\left(\left(X_{1} \cup\left(B \backslash Y_{1}\right)\right) \cap\left(X_{2} \cup\left(B \backslash Y_{2}\right)\right)\right) \\
\geq & r\left(\left(X_{1} \cup X_{2}\right) \cup\left(B \backslash\left(Y_{1} \cap Y_{2}\right)\right)\right)+r\left(\left(X_{1} \cap X_{2}\right) \cup\left(B \backslash\left(Y_{1} \cup Y_{2}\right)\right)\right) .
\end{aligned}
$$

Similarly

$$
r\left(Y_{1} \cup\left(B \backslash X_{1}\right)\right)+r\left(Y_{2} \cup\left(B \backslash X_{2}\right)\right) \geq r\left(\left(Y_{1} \cup Y_{2}\right) \cup\left(B \backslash\left(X_{1} \cap X_{2}\right)\right)\right)+r\left(\left(Y_{1} \cap Y_{2}\right) \cup\left(B \backslash\left(X_{1} \cup X_{2}\right)\right)\right)
$$

But also

$$
\left|B \backslash X_{1}\right|+\left|B \backslash X_{2}\right|=\left|B \backslash\left(X_{1} \cap X_{2}\right)\right|+\left|B \backslash\left(X_{1} \cup X_{2}\right)\right| .
$$

Adding, we deduce that

$$
\lambda_{B}\left(X_{1}, Y_{1}\right)+\lambda_{B}\left(X_{2}, Y_{2}\right) \geq \lambda_{B}\left(X_{1} \cap X_{2}, Y_{1} \cup Y_{2}\right)+\lambda\left(X_{1} \cup X_{2}, Y_{1} \cap Y_{2}\right)
$$

To apply 5.5 to matroid branch-width, we needed a submodular function minimization algorithm that, given a matroid $\mathcal{M}$ and two disjoint subsets $X$ and $Y$, will output $Z \subseteq E(\mathcal{M})$ such that $X \subseteq Z \subseteq E(\mathcal{M}) \backslash Y$ and $\lambda(Z)$ is minimum. We would like to show that this can be done by the matroid intersection algorithm. Let $\mathcal{M}_{1}=\mathcal{M} / X \backslash Y, \mathcal{M}_{2}=\mathcal{M} \backslash X / Y$. Let $r_{1}$, $r_{2}$ be the rank function of $\mathcal{M}_{1}, \mathcal{M}_{2}$, respectively. Then by the matroid intersection algorithm, we can find $U \subseteq E(\mathcal{M}) \backslash X \backslash Y$ minimizing $r_{1}(U)+r_{2}(E(\mathcal{M}) \backslash X \backslash Y \backslash U)$. Using the fact $r_{1}(U)=r(U \cup X)-r(X), r_{2}(U)=r(U \cup Y)-r(Y)$, we get $Z, X \subseteq Z \subseteq E(\mathcal{M}) \backslash Y$ minimizing $\lambda(Z)$. And this can be done in $O\left(n^{3}\right)$ time (in terms of the rank oracle), where $n=|E(\mathcal{M})|$.

We deduce:

Corollary 7.2. For given $k$, there is an algorithm that, with input an n-element matroid $\mathcal{M}$, given by its rank oracle, either concludes that $\mathrm{bw}(\mathcal{M})>k$ or outputs a branch-decomposition of $\mathcal{M}$ of width at most $3 k-1$. Its running time and number of oracle calls is at most $O\left(n^{4}\right)$.

Proof. Pick a base $B$ of $\mathcal{M}$ arbitrarily. We use $\lambda_{B}$ as an interpolation of $\lambda$. For a given partition $(A, B)$, finding a base $X$ can be done in time $O(n)$. Finding $Z$ to prove that $X$ is not welllinked can be done in $O\left(2^{3 k-2} n^{3}\right)$. Therefore, the time complexity is $O\left(n+n\left(n+2^{3 k-2} n^{3}\right)\right)=$ $O\left(8^{k} n^{4}\right)$.

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