# Binary Constraint Trees and Structured Decomposability 

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#### Abstract

A binary constraint tree (BCT, Wang and Yap 2022) is a normalized binary CSP whose constraint graph is a tree. A BCT constraint is a constraint represented with a BCT where some of the variables may be hidden (i.e. existentially quantified and used only for internal representation). Structured decomposable negation normal forms (SDNNF) were introduced by Pipatsrisawat and Darwiche (2008) as a restriction of decomposable negation normal forms (DNNF). Both DNNFs and SDNNFs were studied in the area of knowledge compilation. In this paper we show that the BCT constraints are polynomially equivalent to SDNNFs. In particular, a BCT constraint can be represented with an SDNNF of polynomial size and, on the other hand, a constraint that can be represented with an SDNNF, can be represented as a BCT constraint of polynomial size. This generalizes the result of Wang and Yap (2022) that shows that a multivalued decision diagram (MDD) can be represented with a BCT. Moreover, our result provides a full characterization of binary constraint trees using a language that is well studied in the area of knowledge compilation. It was shown by Wang and Yap (2023) that a CSP on $n$ variables of domain sizes bounded by $d$ that has treewidth $k$ can be encoded as a BCT on $O(n)$ variables with domain sizes $O\left(d^{k+1}\right)$. We provide an alternative reduction for the case of binary CSPs. This allows us to compile any binary CSP to an SDNNF of size that is parameterized by $d$ and $k$.


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## 1 Introduction

Constraint satisfaction problems (CSPs) offer an expressive and natural way of formulating problems. A CSP is a problem of checking satisfiablity of a conjunction of constraints on variables with finite domains. Constraints can be represented in various ways, which include tables (see e.g. [2, 17, 18]) or multivalued decision diagrams (MDD, see e.g. [1, 5, 6]).

The representation using binary constraint trees was introduced in [27]. A BCT constraint is a constraint $c$ defined on a set of variables $\mathbf{x}$ that is represented with a normalized binary CSP $P$ whose constraint graph is a tree. The CSP $P$ itself is defined on a set of variables $\mathbf{z}$ which may include some hidden variables in addition to all the original variables from $\mathbf{x}$. BCTs have a nice property that an arc consistency propagator can be used to check their consistency [12]. Any CSP can be turned into a binary one with an encoding such as dual encoding [11], hidden variable encoding [22], double encoding [24], or bipartite encoding [25].

Decomposable negation normal forms (DNNFs) were introduced in [7] as a tractable language for knowledge representation. Structured DNNFs (SDNNF) were introduced in [20]. The definition of SDNNFs is based on the notion of a $v$-tree which is a rooted binary tree whose leaves are in one-to-one correspondence with the constraint variables (both original and hidden). The conjunction gates in an SDNNF are then required to respect a particular v-tree (see definitions 5 and 6 in Section 2.3 for more details). The structural requirements imposed

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on SDNNFs allow for instance a polynomial time construction of an SDNNF representing the conjunction of two SDNNFs that respect the same v-tree - something that is not possible for DNNFs without any structural requirements. Although DNNFs were introduced as a representation of functions on boolean variables, they were also considered as a representation of constraints on variables with finite domains [1, 14].

Both BCT constraints and SDNNFs have a structure based on a tree. We use this similarity to show the main result of this paper: BCT constraints and SDNNF constraints are polynomially equivalent. In particular, we show a polynomial time transformation of a BCT constraint into an SDNNF and also a polynomial time transformation of an SDNNF into a BCT constraint. The polynomial equivalence of BCTs with SDNNFs offers a characterization of BCTs by a language of SDNNFs which has been extensively studied in the area of knowledge compilation [3, 20, 21, 23]. Our result also generalizes the previous construction of a BCT constraint for an MDD described in [27]. It was shown in [26] that BCTs are strictly more succinct than MDDs. This also follows from a combination of our result with the fact that SDNNFs are strictly more succinct than MDDs by [20].

Recently, [28] studied BCTs from the perspective of knowledge compilation together with several other languages that are being used to represent ad-hoc constraints. The authors studied BCTs with respect to the queries and transformations considered in the knowledge compilation map [10] and showed that BCTs allow answering consistency, clausal entailment and model enumeration queries in polynomial time which is (unsurprisingly) the same as in the case of structured DNNFs [20]. The authors of [28] also studied BCTs with respect to transformations. If the input BCTs or SDNNFs are required to have the same tree structure, then they allow polynomial-time bounded conjunction, unbounded disjunction, forgetting any number of variables, and conditioning [20, 28]. Interestingly, [20] only considers the case of combining SDNNFs that respect the same v-tree while [28] also considers the case of combining BCTs that are not required to have the same tree structure. In this case BCTs do not allow polynomial time bounded conjunction, they do not allow an unbounded disjunction, and the case of bounded disjunction is unresolved in [28]. We believe that our result might help to resolve the case of bounded disjunction for BCTs, because it might be easier to reason about a disjunction of two SDNNFs than BCTs. It is also worth mentioning that according to [20], AOMDDs introduced in [19] are strictly less succinct than SDNNFs and thus also strictly less succinct than BCTs. This already answers one of the questions posed in [28].

Our transformation of a BCT constraint into an SDNNF leads to a smooth SDNNF. It is thus possible to use a domain consistency propagator for smooth DNNFs described in [14] as a domain consistency propagator for BCT constraints. An encoding of BCT constraints with propagation complete CNF formulas was described in [26]. Various CNF encodings of DNNF theories were considered in [1] and a propagation complete encoding of smooth DNNFs was introduced in [16]. Our result thus offers an alternative way of reducing BCT constraints to a CNF encoding.

If CNF $\varphi$ has treewidth $k$, then it can be compiled to an SDNNF of size that is parameterized by $k$ by the construction described in [21]. In particular, if $\varphi$ has $n$ variables and $m$ clauses, then an equivalent SDNNF can be constructed in time $O\left(n m 2^{k}\right)$. We can obtain a similar result also for binary CSPs, but we have to take into account also the domain sizes. It was shown in [28] that if $P$ is a CSP on $n$ variables of domain size $d$ that has treewidth $k$, then it can be encoded as a BCT with $O(n)$ variables with domain size $d^{k+1}$. The construction in [28] uses the encoding described in [11]. In addition, if $P$ is a binary CSP, its consistency can be checked in time $O\left(n d^{k+1}\right)$ by [13]. To have a complete compilation procedure of a binary CSP into an SDNNF, we provide a direct reduction of a binary CSP
to a BCT parameterized by the treewidth and the domain size. The bound we obtain is the same as in [28], so our result is not really new in this sense, but we obtain a slightly better bound on the size of an SDNNF constructed for the given binary CSP than if we would simply combine the bound of [28] with our construction of an SDNNF.

The paper is organized as follows. We introduce the necessary notation in Section 2, including the definitions of BCTs and structured DNNFs. In Section 3, we show that a BCT constraint can be represented with an SDNNF. The transformation of an SDNNF to a BCT is described in Section 4. A tranformation of a binary CSP with bounded treewidth into an SDNNF is described in Section 5. Section 6 closes the paper with a few concluding remarks.

## 2 Definitions

In this section, we shall recall the necessary notation and notions used in the paper.

### 2.1 BCT Constraint

We use a notation adapted from [27] where binary constraint trees were introduced.
A CSP $P$ is a pair $(\mathbf{x}, C)$ where $\mathbf{x}$ is a set of variables and $C$ is a set of constraints. Each variable $x$ has a finite domain denoted $\operatorname{dom}(x)$. A literal on a variable $x$ is a variable value assignment $(x, a)$. A tuple over a set of variables $\left\{x_{i_{1}}, x_{i_{2}}, \ldots, x_{i_{r}}\right\}$ is a set of literals $\left\{\left(x_{i_{1}}, a_{1}\right),\left(x_{i_{2}}, a_{2}\right), \ldots,\left(x_{i_{r}}, a_{r}\right)\right\}$. Each constraint $c_{j}$ has a constraint scope $\operatorname{scp}\left(c_{j}\right) \subseteq \mathbf{x}$ and a relation $\operatorname{rel}\left(c_{j}\right)$ defined by a set of tuples over $\operatorname{scp}\left(c_{j}\right)$. A constraint $c$ is a binary constraint if $|\operatorname{scp}(c)|=2$ and it is a unary constraint if $|\operatorname{scp}(c)|=1$. A CSP $P$ is called a binary CSP if it consists of binary and unary constraints. A binary CSP is normalized if its constraints have pairwise different scopes. Given any set of variables $\mathbf{z}$ and literals $\tau$, we use $\tau[\mathbf{z}]=\{(x, a) \in \tau \mid x \in \mathbf{z}\}$ to denote a subset of $\tau$, while $T[\mathbf{z}]=\{\tau[\mathbf{z}] \mid \tau \in T\}$ is the projection of a set of tuples $T$ on $\mathbf{z}$. A tuple $\tau$ over $\mathbf{x}$ is a solution of $P$ if $\tau[\operatorname{scp}(c)] \in \operatorname{rel}(c)$ for all constraints $c \in C$ and $a \in \operatorname{dom}(x)$ for all $(x, a) \in \tau$. We use $\operatorname{sol}(\mathbf{x}, C)$ (or $\operatorname{sol}(P))$ to denote the set of all solutions of $P$. We also say that $P$ is satisfied by its solution and that a solution of $P$ satisfies all constraints in $C$. A support of a value $a \in \operatorname{dom}(x)$ in a constraint $c$ is a tuple $\tau \in \operatorname{rel}(c)$ such that $(x, a) \in \tau$ and $b \in \operatorname{dom}(y)$ for all $(y, b) \in \tau$.

- Definition 1 ([27]). A Binary Constraint Tree (BCT) is a normalized binary CSP whose constraint graph is a tree. A BCT constraint $c$ is a pair $(\mathbf{x}, P)$ such that $P=(\mathbf{z}, C)$ is a BCT, $\operatorname{scp}(c)=\mathbf{x} \subseteq \mathbf{z}$, and $\operatorname{rel}(c)=\operatorname{sol}(\mathbf{z}, C)[\mathbf{x}]$. A tree binary encoding (TBE) of a constraint $c^{*}$ is a BCT $P=(\mathbf{z}, C)$ such that the BCT constraint $\left(\operatorname{scp}\left(c^{*}\right), P\right)$ has the same constraint relation as $c^{*}$ where the variables in scp $\left(c^{*}\right)$ and $\mathbf{z} \backslash \operatorname{scp}\left(c^{*}\right)$ are called the original and hidden variables, respectively.
- Example 2. Let us consider a BCT constraint $c^{*}=(\mathbf{x}, P)$ on three variables $\mathbf{x}=$ $\left\{x_{1}, x_{2}, x_{3}\right\}$ where $P=(\mathbf{z}, C)$ is a BCT described as follows. We have one hidden variable $y$ in $P$, i.e. $\mathbf{z}=\left\{x_{1}, x_{2}, x_{3}, y\right\}$, and three constraints $C=\left\{c_{1}, c_{2}, c_{3}\right\}$ with $\operatorname{scp}\left(c_{i}\right)=\left\{x_{i}, y\right\}$, $i=1,2,3$. The domain of all variables (original and hidden) is $\{1,2,3\}$. For $i=1,2,3$, we set $\operatorname{rel}\left(c_{i}\right)=\left\{\left(\left(x_{i}, a\right),(y, b)\right) \mid a \neq b\right\}$. That is, $c_{i}$ enforces that $y$ has a different value from $x_{i}$ in any solution to $c^{*}$. Altogether, $c^{*}$ is equal to the negation of the alldifferent constraint over the variables $x_{1}, x_{2}, x_{3}$.

A general construction of a BCT representing the negation of the alldifferent constraint over variables $x_{1}, \ldots, x_{r}$ with domains $D=\{1, \ldots, r\}$ was described in [27] where the authors also noted that the size of the MDD representing the constraint is exponential in $r$.


Figure 1 A binary constraint tree $P=(\mathbf{z}, C)$ from Example 3.

To demonstrate the techniques described in the paper, we shall consider a simple constraint that is a bit less symmetrical than the negation of the alldifferent constraint on three variables.

- Example 3. Figure 1 shows a BCT $P=(\mathbf{z}, C)$ that is defined on variables $\mathbf{z}=\left(z_{1}, z_{2}, z_{3}, z_{4}\right)$ with domains $\operatorname{dom}\left(z_{i}\right)=\{0,1,2\}$ for all $i=1, \ldots, 4$. $C$ consists of three constraints. Constraints $c_{1}$ and $c_{2}$ represent inequalities $z_{1} \leq z_{3}$ and $z_{2} \leq z_{3}$ respectively and its relation is shown in figures 1 b and 1 c . Constraint $c_{3}$ represents inequality $z_{3}<z_{4}$ and its relation is shown in Figure 1d. Note that literals $\left(z_{4}, 0\right)$ and $\left(z_{3}, 2\right)$ do not have support in $c_{3}$.

Let us now consider a constraint $c^{*}$ with scope $\operatorname{scp}\left(c^{*}\right)=\left\{x_{1}, x_{2}, x_{3}\right\}$ where $\operatorname{dom}\left(x_{i}\right)=$ $\{0,1,2\}$ for $i=1,2,3$ and the set of tuples $\operatorname{rel}\left(c^{*}\right)$ represents inequality $\max \left(x_{1}, x_{2}\right)<x_{3}$. If we identify variables $z_{1}, z_{2}$, and $z_{4}$ with $x_{1}, x_{2}$, and $x_{3}$ respectively, then $P$ is a tree binary encoding of $c^{*}$ in which $z_{1}, z_{2}$, and $z_{4}$ are original variables and $z_{3}$ is a hidden variable.

Note that the hidden variable $z_{3}$ is not actually needed for the constraint representation. We keep it to demonstrate how a hidden variable can be later forgotten in an SDNNF. We may also observe that literals $\left(z_{4}, 0\right)$ and $\left(z_{3}, 2\right)$ do not have a support in constraint $c_{3}$. We shall see later how this situation is dealt with during the construction of an SDNNF representing $c^{*}$.

### 2.2 DNNF

The notion of a DNNF was introduced in [7] as a restriction of NNF. We consider a multivalued variant that was used for instance in $[14,16]$. This form is suitable for using DNNFs to represent constraints.

Consider a set of variables $\mathbf{x}=\left\{x_{1}, \ldots, x_{n}\right\}$ with a finite domain $\operatorname{dom}\left(x_{i}\right)$ for each $x_{i} \in \mathbf{x}$. A sentence in negation normal form (NNF) $D$ is a rooted DAG with vertices $V$, root $\rho \in V$, the set of edges $E$, and the set of leaves $L \subseteq V$. The inner vertices (also called gates) are labeled with logical connectives $\wedge$ or $\vee$. Each edge $(v, u)$ in $D$ connects an inner vertex $v$ labeled $\wedge$ or $\vee$ with one of its inputs $u$. The edge is directed from $v$ to $u$, so the inputs of a
vertex are its successors (or child vertices). The leaves are labeled with literals of variables $\mathbf{x}$, i.e. each leaf is labeled with a literal $\left(x_{i}, a\right)$ on a variable $x_{i} \in \mathbf{x}$ and a value $a \in \operatorname{dom}\left(x_{i}\right)$. We assume that each literal $\left(x_{i}, a\right)$ is used as a label of at most one leaf. Some of the literals may be missing in $D$, however, we assume that for each $i=1, \ldots, n$ at least one leaf of $D$ is labeled with a literal on variable $x_{i}$. For technical reasons, we also allow to label leaves with constants 0 or 1 . If a NNF is nonempty, constants can always be propagated and these constants are thus needed only on a NNF without any variables.

If all the constraint variables $x_{i} \in \mathbf{x}$ are boolean (i.e. $\left.\operatorname{dom}\left(x_{i}\right)=\{0,1\}\right)$ and we identify literal $\left(x_{i}, 1\right)$ with the propositional literal $x_{i}$ and literal $\left(x_{i}, 0\right)$ with the propositional literal $\neg x_{i}$, then we obtain the usual definition of a NNF for representing a boolean function.

Assume that $c$ is a constraint with the scope $s c p(c)=\mathbf{x}$ and that $D$ is a NNF defined on the variables $\mathbf{x}$. We say that $D$ represents constraint $c$ if for every tuple $\tau$ over variables $\mathbf{x}$ we have that $\tau \in \operatorname{rel}(c)$ if and only if $D$ evaluates to true on the tuple $\tau$. Evaluating $D$ on $\tau$ is done in a straightforward manner, we simply set the leaves $\left(x_{i}, a\right) \in \tau$ to true and the remaining leaves to false, then we use the usual semantic of the circuit $D$ to get the value on this assignment.

Following [14], we define the decomposability and smoothness properties with respect to constraint variables $x_{1}, \ldots, x_{n}$. For a vertex $v \in V$, let us denote $\operatorname{var}(v) \subseteq \mathbf{x}$ the set of variables in the subcircuit of $D$ rooted at $v$. More precisely, a variable $x_{i} \in \mathbf{x}$ belongs to $\operatorname{var}(v)$ if and only if there is a directed path from $v$ to a leaf labeled with a literal $\left(x_{i}, a\right)$ for a value $a \in \operatorname{dom}\left(x_{i}\right)$. We have by assumption that $\operatorname{var}(\rho)=\mathbf{x}$.

- Definition 4. We define the following structural restrictions of NNFs.
- We say that NNF $D$ is decomposable (DNNF), if for every vertex $v=v_{1} \wedge \cdots \wedge v_{k}$ the sets of variables $\operatorname{var}\left(v_{1}\right), \ldots, \operatorname{var}\left(v_{k}\right)$ are pairwise disjoint.
- We say that DNNF $D$ is smooth if for every vertex $v=v_{1} \vee \cdots \vee v_{k}$ we have $\operatorname{var}(v)=$ $\operatorname{var}\left(v_{1}\right)=\cdots=\operatorname{var}\left(v_{k}\right)$.

Assume that $D$ is a DNNF representing constraint $c$ with scope $s c p(c)=\mathbf{z}$. Let $\mathbf{x} \subseteq \mathbf{z}$ and $\mathbf{y}=\mathbf{z} \backslash \mathbf{x}$. By forgetting variables $\mathbf{y}$ in $D$ we mean the construction of a DNNF $D^{\prime}$ that represents the constraint $c^{\prime}$ which is a projection of $c$ on variables $\mathbf{x}$. In particular, $\operatorname{scp}\left(c^{\prime}\right)=\mathbf{x}$ and $\operatorname{rel}\left(c^{\prime}\right)=\operatorname{rel}(c)[\mathbf{x}]$. Forgetting can be done efficiently on a DNNF, we simply replace every literal ( $y, a$ ) with constant 1 for all $y \in \mathbf{y}$ and $a \in \operatorname{dom}(y)[8]$.

### 2.3 Structured DNNF

Structured DNNFs were introduced in [20]. Structured decomposability is based on the notion of a v-tree defined as follows.

- Definition 5 ([20]). A v-tree for a set of variables $\mathbf{x}$ is a full, rooted binary tree whose leaves are in one-to-one correspondence with the variables in $\mathbf{x}$.

Given a node $t$ of a v-tree $T$, we denote $\operatorname{var}(t)$ the set of variables associated with the leaves in the subtree of $T$ rooted at $t$. We also denote $\operatorname{var}(T)=\operatorname{var}(\sigma)$ where $\sigma$ is the root of $T$. For a non-leaf node $t$, we use $t^{l}\left(t^{r}\right)$ to denote the left (right) child node of $t$. For the rest of this paper, we will assume that each conjunction in a DNNF has exactly two non-constant inputs, while a disjunction can have any number of inputs. This is a technical assumption used in [20] mainly to simplify the definition of a SDNNF, since with this assumption, it is enough to consider only binary v-trees. Note also that we can make this assumption without loss of generality due to the associativity and commutativity of conjunction.

(a) SDNNF $D_{P}$.

(b) v-tree $T$.

Figure 2 An example of an SDNNF $D_{P}$ and the corresponding v-tree $T$. This particular SDNNF is the result of our construction on the BCT $P$ from Example 3. The labels of the nodes mark the steps of our construction.

- Definition 6 ([20]). A DNNF $D$ respects a $v$-tree $T$ if for every conjunction $v=v_{1} \wedge v_{2}$ in $D$, there is a node $t$ in $T$, such that $\operatorname{var}\left(v_{1}\right) \subseteq \operatorname{var}\left(t^{l}\right)$ and $\operatorname{var}\left(v_{2}\right) \subseteq \operatorname{var}\left(t^{r}\right)$.

Let $v$ be a node in a DNNF $D$ that respects a v-tree $T$. The decomposition node ( $d$-node) of $v$ is defined as the deepest node $d$ in $T$ such that $\operatorname{var}(v) \subseteq \operatorname{var}(d)$.

- Definition 7 ([20]). A DNNF that respects a given v-tree $T$ is denoted as $D N N F_{T}$. Moreover, the language of structured DNNFs (SDNNF) consists of all $D N N F_{T}$ for any v-tree $T$.

Given a $\mathrm{DNNF}_{T} D$, we can construct an equivalent smooth $\mathrm{DNNF}_{T} D^{\prime}$ in quadratic time [23]. It means that we can always assume that the input $\mathrm{DNNF}_{T}$ is smooth.

- Example 8. Figure 2 shows an example of a smooth $\mathrm{DNNF}_{T}$. In particular, $\mathrm{DNNF}_{T} D_{P}$ on Figure 2a respects v-tree $T$ on Figure 2b. $\mathrm{DNNF}_{T} D_{P}$ represents the BCT constraint $(\mathbf{z}, P)$ where $P$ is the BCT from Example 3.

For the construction of an SDNNF representing a BCT constraint, we will need the following operation for composing two v-trees. Assume $T_{1}$ and $T_{2}$ are two v-trees on disjoint sets of variables, i.e. $\operatorname{var}\left(T_{1}\right) \cap \operatorname{var}\left(T_{2}\right)=\emptyset$. Then $T=T_{1} \circ T_{2}$ denotes the v-tree with a newly added root $\sigma$ whose left child node $\sigma^{l}$ is set to the root of $T_{1}$ and the right child node $\sigma^{r}$ is set to the root of $T_{2}$. It follows that $\operatorname{var}(T)=\operatorname{var}\left(T_{1}\right) \cup \operatorname{var}\left(T_{2}\right)$.

## 3 Compiling a BCT Constraint into a Structured DNNF

We shall show in this section that we can construct an SDNNF representing a given BCT constraint in polynomial time.

- Theorem 9. Let $c^{*}=(\mathbf{x}, P)$ be a BCT constraint where $P=(\mathbf{z}, C)$ is a BCT. Then there is a smooth SDNNF $D$ representing $c^{*}$ with $O(m d)$ nodes and $O\left(m d^{2}\right)$ edges where $m=|\mathbf{z}|$ and $d=\max _{z_{i} \in \mathbf{z}}\left|\operatorname{dom}\left(z_{i}\right)\right|$.

We will describe the construction of $D$ in the rest of this section, thus proving Theorem 9 . Assume that $n=|\mathbf{x}|$ and $\mathbf{z}=\left\{z_{1}, \ldots, z_{m}\right\}$ where $\mathbf{x} \subseteq \mathbf{z}$ and $m \geq n$ is the number of all variables. We assume that $\left|\operatorname{dom}\left(z_{i}\right)\right| \leq d$ for every $i=1, \ldots, m$.

The construction proceeds in two steps. First, we describe a construction of a SDNNF $D_{P}$ representing $P$. A SDNNF $D$ that represents $c^{*}$ then originates from $D_{P}$ by forgetting the hidden variables $\mathbf{y}=\mathbf{z} \backslash \mathbf{x}$. This step can be done efficiently by [20].

Let $G$ be the constraint graph of $P . G$ is a tree with the set of nodes $\mathbf{z}$, each edge corresponds to a single constraint from $C$. Let $G^{+}$denote a directed tree that originates from $G$ by picking an arbitrary node as a root and directing all edges from the root towards the leaves. Let us assume that the nodes $z_{1}, \ldots, z_{m}$ are ordered in a reverse topological order with respect to $G^{+}$. It means that $z_{m}$ is the root and if $\left(z_{i}, z_{j}\right)$ is an edge in $G^{+}$, then $i>j$. See Figure 1a for an example.

For every $i=1, \ldots, m$, let us consider the subtree $G_{i}$ of $G^{+}$rooted at $z_{i}$. Let $C_{i} \subseteq C$ denote the set of constraints corresponding to the edges of $G_{i}$. Denote $\mathbf{z}_{i}=\bigcup_{c \in C_{i}} s c p(c)$. In this way, we have defined BCT $P_{i}=\left(\mathbf{z}_{i}, C_{i}\right)$. For every value $a \in \operatorname{dom}\left(z_{i}\right)$, we also define BCT $P_{i, a}$ as a restriction of $P_{i}$ to the solutions that contain literal $\left(z_{i}, a\right)$. This can be best understood as adding a unary constraint with scope $z_{i}$ and a single relation $\left(z_{i}, a\right)$ to $C_{i}$. Another way of looking at it is restricting the relation of every constraint $c \in C_{i}$ with $z_{i} \in \operatorname{scp}(c)$ to the tuples containing $\left(z_{i}, a\right)$ and setting $\operatorname{dom}\left(z_{i}\right)$ to $\{a\}$.

The algorithm proceeds for every $i=1, \ldots m$ in order and constructs for every $a \in \operatorname{dom}\left(z_{i}\right)$ a SDNNF $D_{i, a}$ representing BCT constraint $\left(\mathbf{z}_{i}, P_{i, a}\right)$ and a v-tree $T_{i}$ that is respected by $D_{i, a}$ using the following steps:
(A1) If $z_{i}$ is a leaf (no edges leave $z_{i}$ in $G^{+}$), then $P_{i}$ has no constraints and $P_{i, a}$ has the domain of $z_{i}$ restricted to the single value $a$. DNNF $D_{i, a}$ is a single node labeled with literal $\left(z_{i}, a\right)$ and v-tree $T_{i}$ is a single node labeled with variable $z_{i}$.
(A2) Assume that $z_{i}$ is not a leaf and it has $k$ outgoing edges $\left(z_{i}, z_{i_{1}}\right), \ldots,\left(z_{i}, z_{i_{k}}\right)$ associated with constraints $c_{1}, \ldots, c_{k} \in C$. For every $p=1, \ldots, k$ we have that $i_{p}<i$, because the nodes are processed in a reverse topological order, and thus we have already constructed $D_{i_{p}, b}$ and $T_{i_{p}}$ for each $b \in \operatorname{dom}\left(z_{i_{p}}\right)$. Let us now construct $D_{i, a}$ for $a \in \operatorname{dom}\left(z_{i}\right)$ in the following two steps.
(A2a) For every $p=1, \ldots, k$, define $E_{i, a}^{i_{p}}$ as follows:

$$
E_{i, a}^{i_{p}}=\bigvee_{\left\{\left(z_{i}, a\right),\left(z_{i_{p}}, b\right)\right\} \in \operatorname{rel}\left(c_{p}\right)} D_{i_{p}, b}
$$

(A2b) We construct $D_{i, j}$ as a conjunction of literal $\left(z_{i}, a\right)$ with all DNNFs $E_{i, a}^{i_{p}}$ for $p=1, \ldots, k$. However, since we only allow conjunctions with two inputs, we construct $D_{i, a}$ in $k+1$ steps as follows.

$$
\begin{align*}
D_{i, a}^{(1)} & =E_{i, a}^{i_{1}}  \tag{1}\\
D_{i, a}^{(p)} & =D_{i, a}^{(p-1)} \wedge E_{i, a}^{i_{p}} \quad \text { for } p=2, \ldots, k  \tag{2}\\
D_{i, a} & =\left(z_{i}, a\right) \wedge D_{i, a}^{(k)} \tag{3}
\end{align*}
$$

In addition, we define $T_{i}$ as follows:

$$
\begin{align*}
T_{i}^{(1)} & =T_{i_{1}}  \tag{4}\\
T_{i}^{(p)} & =T_{i}^{(p-1)} \circ T_{i_{p}} \quad \text { for } p=2, \ldots, k  \tag{5}\\
T_{i} & =z_{i} \circ T_{i}^{(k)} \tag{6}
\end{align*}
$$

Variable $z_{i}$ is identified with a tree consisting of a single leaf labeled with $z_{i}$ in step (6).

Once we have constructed $D_{m, a}$ for every $a \in \operatorname{dom}\left(z_{m}\right)$, we compose them to obtain $D_{P}$ that respects v-tree $T=T_{m}$ as follows:

$$
\begin{equation*}
D_{P}=\bigvee_{a \in \operatorname{dom}\left(z_{m}\right)} D_{m, a} \tag{7}
\end{equation*}
$$

Intuitively, $E_{i, a}^{i_{p}}$ represents the fact that $\left(z_{i}, a\right)$ has a support $\left(\left(z_{i}, a\right),\left(z_{i_{p}}, b\right)\right)$ in $c_{p}$. Moreover, the constraints below $z_{i_{p}}$ in $G^{+}$can be satisfied with the value of $z_{i_{p}}$ set to $b$. SDNNF $D_{i, a}$ represents the models of all constraints that correspond to the edges in the subtree of $G^{+}$rooted at $z_{i}$ and that contain literal $\left(z_{i}, a\right)$. If $\left(z_{i}, a\right)$ does not have a support in $c_{p}$ for some $p=1, \ldots, k$, then the empty disjunction $E_{i, a}^{i_{p}}$ is equal to constant 0 and so is $D_{i, a}$. However, it is possible that $E_{i, a}^{i_{p}}$ is inconsistent even if $\left(z_{i}, a\right)$ has a support $\left(\left(z_{i}, a\right),\left(z_{i_{p}}, b\right)\right)$ in $c_{p}$, but $D_{i_{p}, b}$ is inconsistent for every such $b$.

- Example 10. Figure 2 shows the result of the construction when applied to the BCT $P$ from Example 3. Note that there is no leaf labeled with literal $\left(z_{4}, 0\right)$, because $\left(z_{4}, 0\right)$ has no support in constraint $c_{3}$. For the same reason, there is no leaf labeled with literal $\left(z_{3}, 2\right)$. Consequently, there are no leaves labeled with literals $\left(z_{1}, 2\right)$ and $\left(z_{2}, 2\right)$. It is worth noting that value 2 would be removed from the domains of variables $z_{1}, z_{2}$, and $z_{3}$ and value 0 would be removed from the domain of $z_{4}$ when enforcing arc consistency. In this way, enforcing arc consistency is part of the construction.

If variable $z_{3}$ would be forgotten from $D_{P}$, we would obtain a SDNNF representing the constraint $c^{*}$ from Example 3. This would amount to replacing leaves labeled with literals $\left(z_{3}, 0\right)$ and $\left(z_{3}, 1\right)$ with constant 1 . In this case, it just means removing these leaves altogether. Afterwards, we could simplify the SDNNF by removing the trivial gates with a single input, the result of this simplification can be seen in Figure 3.

We will now show that $D_{P}$ is a smooth SDNNF of polynomial size that represents $P$. We will start by showing that $D_{P}$ is a smooth SDNNF.

- Lemma 11. $D_{P}$ is a smooth $S D N N F$ that respects $v$-tree $T_{m}$.

Proof. Let us first show that $D_{P}$ is an SDNNF that respects v-tree $T_{m}$. We will proceed by induction on $i=1, \ldots, m$. If $z_{i}$ is a leaf of $G^{+}$, then by step (A1), $D_{i, a}$ consists of a single leaf node for every $a \in \operatorname{dom}\left(z_{i}\right)$. It follows that $D_{i, a}$ respects $T_{i}$ which also consists of a single leaf node.

(a) Simplified SDNNF.

(b) Simplified v-tree $T$.

Figure 3 Simplified SDNNF for the BCT constraint $c^{*}$ from Example 3 representing constraint $\max \left(x_{1}, x_{2}\right)<x_{3}$. The SDNNF originated from the SDNNF in Figure 2a by forgetting $z_{3}$, identifying $x_{1}=z_{1}, x_{2}=z_{2}, x_{3}=z_{4}$, and removing gates with a single input.

Let us now assume that $z_{i}$ is not a leaf, in particular $i>1$. Let us assume that $z_{i}$ has $k$ outgoing edges to nodes $z_{i_{1}}$ to $z_{i_{k}}$. Assume a value $a \in \operatorname{dom}\left(z_{i}\right)$. By induction hypothesis, for every $p=1, \ldots, k$ and every $b \in \operatorname{dom}\left(z_{i_{p}}\right)$ we have that $D_{i_{p}, b}$ is a SDNNF respecting v-tree $T_{i_{p}}$. It follows that $E_{i, a}^{i_{p}}$ constructed in step (A2a) is a SDNNF respecting $T_{i_{p}}$. We have that $\operatorname{var}\left(E_{i, a}^{i_{p}}\right)=\mathbf{z}_{i_{p}}$. Since the subtrees rooted at nodes $z_{i_{1}}, \ldots, z_{i_{k}}$ are pairwise disjoint, the same is true for sets $\mathbf{z}_{i_{1}}, \ldots, \mathbf{z}_{i_{p}}$. Moreover, variable $z_{i}$ is not in any of these sets. Therefore, $D_{i, a}$ constructed in step ( A 2 b ) is a DNNF. The construction of the tree $T_{i}$ proceeds in a way similar to the construction of $D_{i, a}$, and thus $D_{i, a}$ respects $T_{i}$. In particular, the node of $T_{i}$ introduced in (5) is the d-node of the conjunction (2) for the same value of $p$, and the node introduced in (6) is the d-node of the conjunction (3).
$D_{P}$ is constructed in step (7) as a disjunction of $D_{m, a}, a \in \operatorname{dom}\left(z_{m}\right)$. As each of these SDNNFs respects $T_{m}$, the same is true for $D_{P}$.

Let us now show the smoothness. Assume that $D_{i, a}$ is nontrivial, i.e. it is not just a single leaf labeled with 0 . We show by induction that then $D_{i, a}$ is a SDNNF that depends on all variables in $\mathbf{z}_{i}$. This is true for the leaves. If $z_{i}$ is not a leaf, $D_{i, a}$ is constructed in step (A2b). By induction hypothesis used on each $D_{i_{p}, b}$ we get that $E_{i, a}^{i_{p}}$ is a smooth disjunction that depends on all variables in $\mathbf{z}_{i_{p}}$. Thus also $D_{i, a}$ depends on all variables in $\mathbf{z}_{i}=\left\{z_{i}\right\} \cup \bigcup_{p=1}^{k} \mathbf{z}_{i_{p}}$. It follows that also the disjunction introduced in the final step (7) is smooth.

Now, let us estimate the size of $D_{P}$.

- Lemma 12. SDNNF $D_{P}$ has $O(m d)$ nodes and $O(m d+s)$ edges where $s=\sum_{c \in C}|r e l(c)|$.

Proof. For every $i=1, \ldots, m$, we add one disjunction for each value $a \in \operatorname{dom}\left(z_{i}\right)$ in step (A2a) and $k$ conjunctions in step (A2b) assuming $z_{i}$ is not a leaf. One more disjunction gate is added in the final step (7). Altogether, we thus add $O(m d)$ gates to $D_{P}$. Each conjunction gate has at most two inputs, thus $O(m d)$ edges are leaving the conjunction gates. For every constraint $c_{p}$ with scope $\left\{z_{i}, z_{i_{p}}\right\}$, every tuple $\left(\left(z_{i}, a\right),\left(z_{i_{p}}, b\right)\right)$ adds one edge leaving disjunction gate $E_{i, a}^{i_{p}}$. The total number of edges leaving the disjunction gates added in step (A2a) is thus at most $s$. The disjunction gate added in the final step has at most $\left|\operatorname{dom}\left(z_{i}\right)\right| \leq d$ inputs. Altogether, we have $O(m d+s)$ edges in $D_{P}$.

It remains to show that $D_{P}$ represents $P$.

- Lemma 13. $S D N N F D_{P}$ represents $P$.

Proof. We shall first show by induction on $i$ that $D_{i, a}$ represents $P_{i, a}$ for every $i=1, \ldots, m$ and $a \in \operatorname{dom}\left(z_{i}\right)$. This is true for leaves (and thus also for $i=1$ ), because $P_{i, a}$ does not have any constraints, it depends only on $z_{i}$, and the domain of $z_{i}$ is restricted to the value $a$. A single node labeled with literal $\left(z_{i}, a\right)$ added in step (A1) is thus a correct representation of $P_{i, a}$ in this case.

Let us now consider a variable $z_{i}$ with outgoing edges $\left(z_{i}, z_{i_{1}}\right), \ldots,\left(z_{i}, z_{i_{k}}\right)$ associated with constraints $c_{1}, \ldots, c_{k} \in C$ where $k \geq 1 . C_{i}$ is thus a disjoint union of $C_{i_{p}}, p=1, \ldots, k$ with $\left\{c_{1}, \ldots, c_{k}\right\}$. Let us also consider a value $a \in \operatorname{dom}\left(z_{i}\right)$. Let us assume by induction hypothesis that each $D_{i_{p}, b}$ represents $P_{i_{p}, b}$ for every $b \in \operatorname{dom}\left(z_{i_{p}}\right)$. Recall that the scope of $P_{i, a}$ is the set $\mathbf{z}_{i}$ of variables in the subtree of $G^{+}$rooted at $z_{i}$. Let $\tau$ be a tuple of variables $\mathbf{z}_{i}$ and let us fix some $a \in \operatorname{dom}\left(z_{i}\right)$.

Let us first assume that $\tau \in \operatorname{sol}\left(P_{i, a}\right)$. It follows that $\left(z_{i}, a\right) \in \tau$. We will show that $\tau$ satisfies $D_{i, a}$. Let us consider literals $\left(z_{i_{1}}, b_{1}\right), \ldots,\left(z_{i_{k}}, b_{k}\right) \in \tau$. For every $p=1, \ldots, k$ we have that $\tau$ satisfies $P_{i_{p}}$. It satisfies $P_{i_{p}, b_{p}}$ as well since $\left(z_{i_{p}}, b_{p}\right) \in \tau$. By induction hypothesis, circuit $D_{i_{p}, b_{p}}$ represents $P_{i_{p}, b_{p}}$ and thus it evaluates to true on $\tau$. By definition of $P_{i, a}$, $\tau$ satisfies $P_{i}$ and thus also constraint $c_{p}$. It follows that $\left\{\left(z_{i_{p}}, b_{p}\right),\left(z_{i}, a\right)\right\} \in \operatorname{rel}\left(c_{p}\right)$ and thus, by step (A2a), also $E_{i, a}^{i_{p}}$ evaluates to true. Since this holds for every $p=1, \ldots, k$ and $\left(z_{i}, a\right) \in \tau$, we have by step (A2b) that $D_{i, a}$ evaluates to true on $\tau$.

Let us now assume that $D_{i, a}$ evaluates to true on $\tau$ and let us show that $\tau \in \operatorname{sol}\left(P_{i, a}\right)$. We have $\left(z_{i}, a\right) \in \tau$ by (3), it remains to show that $\tau \in \operatorname{sol}\left(P_{i}\right)$. Let $p \in\{1, \ldots, k\}$ be arbitrary. We have by (1) to (3) that $E_{i, a}^{i_{p}}$ evaluates to true. By (A2a), we have that $D_{i_{p}, b_{p}}$ evaluates to true on $\tau$ for some $\left\{\left(z_{i}, a\right),\left(z_{i_{p}}, b_{p}\right)\right\} \in \operatorname{rel}\left(c_{p}\right)$. Induction hypothesis implies $\tau\left[\mathbf{z}_{i_{p}}\right] \in \operatorname{sol}\left(P_{i_{p}, b_{p}}\right)$ and thus also $\left(z_{i_{p}}, b_{p}\right) \in \tau$. Together with $\left(z_{i}, a\right) \in \tau$ we obtain that $c_{p}$ is satisfied by $\tau$. In addition, all constraints in $P_{i_{p}}$ are satisfied by $\tau$. Since this holds for every $p=1, \ldots, k$, we get that all constraints of $P_{i}$ are satisfied and thus $\tau \in \operatorname{sol}\left(P_{i}\right)$.

Let us now show that $D_{P}$ represents $P$. If $\tau$ is a tuple satisfying $P$ and $\left(z_{m}, a\right) \in \tau$, then $\tau$ satisfies $P_{m, a}$. It follows that $D_{m, a}$ evaluates to true and that $D_{P}$ evaluates to true as well. If, on the other hand, $D_{P}$ evaluates to true on $\tau$, then $D_{m, a}$ evaluates to true for some $a \in \operatorname{dom}\left(z_{m}\right)$. Thus $\tau$ is a solution of both $P_{m, a}$ and $P=P_{m}$.

Theorem 9 now follows from the above propositions.
Proof of Theorem 9. The SDNNF $D$ representing $c^{*}$ originates from $D_{P}$ by forgetting variables $\mathbf{y}=\mathbf{z} \backslash \mathbf{x}$. This step can be performed in polynomial time by [20] by replacing the literals on variables from $\mathbf{y}$ with constants 1 and then propagating these constants. Note that in $D_{P}$, this is equivalent to removing the corresponding leaves which were added in (3). In particular, this step preserves smoothness which is ensured for $D_{P}$ by Lemma 11. The size bound on $D$ follows from Lemma 12 using the fact that $|C| \leq m$ and $|\operatorname{rel}(c)| \leq d^{2}$ for every $c \in C$.

## 4 Compiling an SDNNF to a BCT Constraint

In this section, we shall show the following theorem.

- Theorem 14. Let $c^{*}$ be a constraint represented by a smooth SDNNF. Then there is a tree binary encoding of $c^{*}$ of polynomial size.

It is useful to look at a DNNF in terms of certificates [4], also called minimal satisfying subtrees in [16]. A certificate for a satisfying assignment is simply a minimal satisfied subDNNF that contains the output gate. Due to decomposability, the certificates of a DNNF are trees whose leaves are some of the leaves of the DNNF. In addition, no two leaves of a certificate are labeled with a literal of the same variable. Assume $D$ is a smooth DNNF on variables $\mathbf{x}=\left(x_{1}, \ldots, x_{n}\right)$ and $S$ its certificate. We shall also assume that $D$ has no leaves labeled with constants 0 or 1 (as these can always be propagated). Then for each $i=1, \ldots, n$, we have that $S$ contains exactly one leaf associated with a literal of variable $x_{i}$ (see also [16] for more details). The leaves of $S$ thus determine a tuple $\tau$ on which $D$ evaluates to true. We say in this case that the leaves of $S$ are associated with the literals in tuple $\tau$. The certificates are thus in one-to-one correspondence with the satisfying assignments of $D$.

Let us now fix a constraint $c^{*}$ with $s c p\left(c^{*}\right)=\mathbf{x}$. Let us assume that $c^{*}$ is represented by a smooth SDNNF $D$ that respects a v-tree $T$ and let $\rho$ denote the root of $D$. In particular, $\operatorname{var}(T)=\operatorname{var}(\rho)=\mathbf{x}$. Then the certificates of $D$ also respect $T$. In fact, we will show that if $S$ is a certificate of $D$, then the conjunction gates of $S$ are in one-to-one correspondence with the inner nodes of $T$. This property lies at the basis of our construction of a tree binary encoding $P=(\mathbf{z}, C)$ of $c^{*}$. The idea is to introduce a hidden variable for each inner node $t$ of $T$ with the domain being the $\wedge$-gates whose d-node is $t$. The constraints make sure that the models of $P$ are in one-to-one correspondence with the certificates of $D$.

We will construct a BCT $P=(\mathbf{z}, C)$ satisfying $\mathbf{x} \subseteq \mathbf{z}$ and $\operatorname{rel}\left(c^{*}\right)=\operatorname{sol}(P)[\mathbf{x}]$. For each inner node $t$ of $T$, we introduce a hidden variable $y_{t}$. The set of all these hidden variables will be denoted as $\mathbf{y}$. We then define $\mathbf{z}=\mathbf{x} \cup \mathbf{y}$. The domain of an original variable $x_{i} \in \mathbf{x}$ is $\operatorname{dom}\left(x_{i}\right)$ as given by the constraint $c^{*}$. For a hidden variable $y_{t} \in \mathbf{y}$, we set $\operatorname{dom}\left(y_{t}\right)=\Lambda(t)$ where $\Lambda(t)$ denotes the set of conjunction gates of $D$ that have d-node $t$.

The constraints of $C$ correspond to the edges of $T$. In particular, for each edge $\left(t, t^{\prime}\right)$ of $T$ where $t^{\prime}$ is a child node of $t$, we add a constraint $c_{t, t^{\prime}}$ to $C$ whose definition differs slightly depending on whether $t^{\prime}$ is a leaf or an inner node of $T$. A sequence of vertices $v_{0}, \ldots, v_{k}$ of $D$ is called an $\vee$-path if it is a path, nodes $v_{1}, \ldots, v_{k-1}$ are $\vee$-gates, $v_{0}$ is a conjunction gate and $v_{k}$ is either a conjunction gate, or a leaf node.
(C1) Assume $t^{\prime}$ is a leaf labeled with variable $x_{i}$. Then $\operatorname{scp}\left(c_{t, t^{\prime}}\right)=\left\{y_{t}, x_{i}\right\}$ and $\operatorname{rel}\left(c_{t, t^{\prime}}\right)$ is defined as a set of tuples $\left\{\left(y_{t}, v\right),\left(x_{i}, a\right)\right\}$ such that $D$ contains a $\vee$-path from $v$ to the leaf labeled with literal $\left(x_{i}, a\right)$.
(C2) Assume $t^{\prime}$ is an inner node of $T$. Then $\operatorname{scp}\left(c_{t, t^{\prime}}\right)=\left\{y_{t}, y_{t^{\prime}}\right\}$ and $\operatorname{rel}\left(c_{t, t^{\prime}}\right)$ is defined as a set of tuples $\left\{\left(y_{t}, v\right),\left(y_{t^{\prime}}, v^{\prime}\right)\right\}$ such that $D$ contains a $\vee$-path from $v$ to $v^{\prime}$.

Example 15. Figure 4 shows the result of the application of the construction to the SDNNF from Figure 2. Recall that the SDNNF itself was constructed as a representation of the BCT $P$ from Example 3. BCT in Figure 4 differs from $P$ in that it has three hidden variables $y_{1}$, $y_{2}$, and $y_{3}$. Note that $y_{1}$ and $y_{2}$ are basically equivalent to $z_{3}$ and $y_{3}$ is equivalent to $z_{4}$. The auxiliary variables can thus be easily eliminated by which we obtain the constraints of $P$.


Figure 4 Example of the construction of the tree binary encoding of a constraint represented by the SDNNF $D$ in Figure 2.

The size of $P$ defined in this way is clearly polynomial in the size of $D$. The rest of this section is devoted to showing the correctness of the construction. We will start with a few technical propositions on the structure of the certificates of $D$.

- Lemma 16. Assume $v$ is a node of $D$ with $d$-node $t$ in $T$. Then $\operatorname{var}(v)=\operatorname{var}(t)$. Moreover, 1. if $v=v_{1} \wedge v_{2}$, then $t^{l}$ is the d-node of $v_{1}$ and $t^{r}$ is the $d$-node of $v_{2}$, and

2. if $v=v_{1} \vee \cdots \vee v_{k}$, then $t$ is the $d$-node of all input nodes $v_{1}, \ldots, v_{k}$.

Proof. We will proceed by the induction on the structure of $D$. If $v=\rho$ is the root of $D$, then $\operatorname{var}(v)=\mathbf{x}$. It follows that its d-node $t$ is the root of $T$ and thus $\operatorname{var}(t)=\mathbf{x}=\operatorname{var}(v)$.

Let us assume that $v=v_{1} \wedge v_{2}$ and that $\operatorname{var}(v)=\operatorname{var}(t)$. Since $D$ does not contain leaves labeled with constants, we have that both $\operatorname{var}\left(v_{1}\right)$ and $\operatorname{var}\left(v_{2}\right)$ are nonempty and thus $\operatorname{var}\left(v_{i}\right) \subsetneq \operatorname{var}(v)$ for $i=1,2$ and $\operatorname{var}(v)=\operatorname{var}\left(v_{1}\right) \cup \operatorname{var}\left(v_{2}\right)$. Let $t_{i}$ be the d-node of $v_{i}, i=1,2$. By the definition of structured DNNFs, both $t_{1}$ and $t_{2}$ are descendants of $t$ in $T$ and thus $\operatorname{var}\left(t_{i}\right) \subseteq \operatorname{var}(t), i=1,2$. By the definition of d-nodes, we also have that $\operatorname{var}\left(v_{i}\right) \subseteq \operatorname{var}\left(t_{i}\right)$, $i=1,2$. It follows that $\operatorname{var}(t)=\operatorname{var}(v)=\operatorname{var}\left(v_{1}\right) \cup \operatorname{var}\left(v_{2}\right) \subseteq \operatorname{var}\left(t_{1}\right) \cup \operatorname{var}\left(t_{2}\right) \subseteq \operatorname{var}(t)$ and thus $\operatorname{var}(t)=\operatorname{var}\left(t_{1}\right) \cup \operatorname{var}\left(t_{2}\right)$. The only possibility is that both $t_{1}$ and $t_{2}$ are the child nodes of $t$ and thus $t_{1}=t^{l}, t_{2}=t^{r}$, and $\operatorname{var}\left(v_{i}\right)=\operatorname{var}\left(t_{i}\right), i=1,2$.

Assume that $v=v_{1} \vee \cdots \vee v_{k}$ and $\operatorname{var}(v)=\operatorname{var}(t)$. By smoothness we get that $\operatorname{var}(t)=\operatorname{var}(v)=\operatorname{var}\left(v_{1}\right)=\cdots=\operatorname{var}\left(v_{k}\right)$. It follows that $t$ is the d-node of all input nodes $v_{1}, \ldots, v_{k}$.

We have shown that if $\operatorname{var}(v)=\operatorname{var}(t)$ and $v^{\prime}$ is a child node of $v$ with d-node $t^{\prime}$, then $\operatorname{var}\left(v^{\prime}\right)=\operatorname{var}\left(t^{\prime}\right)$ which also holds for the leaves.

- Lemma 17. Assume that $v_{0}, \ldots, v_{k}$ is a $\vee$-path with $k>0$. Assume that $t$ is the $d$-node of $v_{0}$ and $t^{\prime}$ is the d-node of $v_{k}$. Then $v_{1}, \ldots, v_{k-1}$ have $d$-node $t^{\prime}$ and $t^{\prime}$ is a child node of $t$.

Proof. By smoothness, $\operatorname{var}\left(v_{1}\right)=\operatorname{var}\left(v_{2}\right)=\cdots=\operatorname{var}\left(v_{k}\right)$ and thus all gates $v_{1}, \ldots, v_{k-1}$ have the same d-node as $v_{k}$. In particular, $t^{\prime}$ is the d-node of $v_{1}$ which is an input to the conjunction gate $v_{0}$. By Lemma 16 we have that $t^{\prime}$ is a child node of $t$.

Based on Lemma 17, we can show the following proposition.

- Lemma 18. Assume $S$ is a certificate and $t$ is an inner node of $T$. Then $S$ contains exactly one conjunction gate $v$ from $\Lambda(t)$.

Proof. Consider a variable $x_{i} \in \operatorname{var}(t)$ and the path $v_{0}, \ldots, v_{k}$ in $S$ that leads from the root $v_{0}=\rho$ to the leaf $v_{k}$ of $S$ labeled with a literal on $x_{i}$. It follows that $\operatorname{var}\left(v_{j}\right) \subseteq \operatorname{var}\left(v_{j-1}\right)$ for every $j=1, \ldots, k$. Moreover $\operatorname{var}\left(v_{0}\right)=\mathbf{x}$ and $\operatorname{var}\left(v_{k}\right)=\left\{x_{i}\right\}$. Let $v_{i_{1}}, \ldots, v_{i_{p}}$ be the subsequence of $v_{0}, \ldots, v_{k}$ formed only by conjunction nodes. Then by Lemma 17 we get that $t_{i_{1}}, \ldots, t_{i_{p}}, t_{k}$ form a path in $T$ from the root to the leaf labeled with $x_{i}$. For some index $i_{j}$ we thus have that $t_{i_{j}}=t$ and it follows that $v_{i_{j}}$ is a conjunction gate in $S$ with d-node $t$.

Let us now assume that $S$ contains two $\wedge$-gates $v_{1}$ and $v_{2}$ with the same d-node $t$, thus $\operatorname{var}\left(v_{1}\right)=\operatorname{var}\left(v_{2}\right)=\operatorname{var}(t)$ by Lemma 16. However, Lemma 16 also implies that there is no path from $v_{1}$ to $v_{2}$ or from $v_{2}$ to $v_{1}$. If we take the paths from the root $\rho$ to $v_{1}$ and $v_{2}$ in $S$, they have to split in a $\wedge$-gate $v$ (by minimality of $S$ ), but then $v$ is not decomposable.

It follows that $v_{i_{j}}$ is the only conjunction gate in $S$ that belongs to $\Lambda(t)$.
Note that each literal on a variable from $\mathbf{z}=\mathbf{x} \cup \mathbf{y}$ is associated with a node of $D$. In particular, for $x_{i} \in \mathbf{x}$, literal $\left(x_{i}, a\right)$ is associated with the leaf of $D$ labeled with $\left(x_{i}, a\right)$. To this end, we need to assume that every such literal has a leaf labeled with it. However, if $D$ does not contain any leaf associated with literal $\left(x_{i}, a\right)$, then this literal does not have a support in $c^{*}$ and $a$ can be removed from $\operatorname{dom}\left(x_{i}\right)$. We may thus assume that no such value is in $\operatorname{dom}\left(x_{i}\right)$. For an inner node $t$ of $T$, a literal $\left(y_{t}, v\right)$ is associated with the node $v \in \Lambda(t)$.

- Lemma 19. Let $\tau \in \operatorname{sol}(P)$ be a tuple that is a solution to $P$. Then $\tau[\mathbf{x}] \in \operatorname{rel}\left(c^{*}\right)$.

Proof. Since $D$ represents $c^{*}$, it is enough to show that there is a certificate $S$ of $D$ whose leaves are associated with the literals in $\tau[\mathbf{x}]$.

Tuple $\tau$ associates a node $v$ of $D$ with every node $t$ of $T$. We proceed by induction on the structure of $T$ to describe a certificate $S_{t}$ for the sub-DNNF of $D$ rooted at $v$.

Assume first that $t$ is a leaf of $T$ labeled with variable $x_{i}$. Consider the literal $\left(x_{i}, a\right) \in \tau$ and set the certificate $S_{t}$ to a single node labeled with this literal.

Assume now that $t$ is an inner node of $T$. Since $t$ is an inner node of $T$, we have that $\left(y_{t}, v\right) \in \tau$ for some $v \in \Lambda(t)$. Tuple $\tau$ also contains literals associated with $t^{l}$ and $t^{r}$. These literals associate a nodes $v_{l}$ and $v_{r}$ of $D$ with $t^{l}$ and $t^{r}$ respectively. By induction hypothesis, we have constructed certificate $S_{l}$ and $S_{r}$ for the sub-DNNFs rooted at $v_{l}$ and $v_{r}$ respectively. Since $\tau$ satisfies constraints $c_{t, t^{l}}$ and $c_{t, t^{r}}, D$ contains a $\vee$-paths from $v$ to $v_{l}$ and from $v$ to $v^{r}$. Certificate $S_{t}$ for the the sub-DNNF rooted at $v$ is then constructed as a union of $S_{l}, S_{r}$, node $v$ and the $\vee$-paths from $v$ to $v_{l}$ and $v_{r}$.

Let $\sigma$ be the root of $T$ and let us assume that $v$ is the node of $D$ associated with $\sigma$ by $\tau$. Let $S_{\sigma}$ be the certificate of the sub-DNNF rooted at $v$. If $v=\rho$ is the root of $D$, then $S=S_{\sigma}$ is a certificate of $D$. Otherwise, $D$ contains a path from $\rho$ to $v$ that consists only of $\vee$-gates and we construct $S$ by combining this path with $S_{\sigma}$.

- Lemma 20. For every $\tau^{*} \in \operatorname{rel}\left(c^{*}\right)$, there is $\tau \in \operatorname{sol}(P)$ satisfying $\tau^{*}=\tau[\mathbf{x}]$.

Proof. Since $\tau^{*} \in \operatorname{rel}\left(c^{*}\right)$, there is a certificate $S$ of $D$ whose leaves are associated with the literals from $\tau^{*}$. By Lemma 18, the certificate $S$ contains exactly one conjunction gate $v_{t} \in \Lambda(t)$ for each inner node $t$. We form $\tau$ by adding literals ( $y_{t}, v_{t}$ ) to $\tau^{*}$ for all internal nodes $t$ of $T$. Let us check that $\tau$ satisfies all constraints of $P$. Let $c_{t, t^{\prime}}$ be a constraint of $P$ where $t^{\prime}$ is a child node of $t$ in $T$. By Lemma 16, one of the child nodes of $v_{t}$ in $D$ has d-node $t^{\prime}$, let us denote it $v_{1}$. Since $v_{t}$ is a conjunction gate, $v_{1}$ must belong to $S$. If $v_{1}$ is a disjunction, then by Lemma 16, its child nodes have d-node $t^{\prime}$, too. If we follow the path in $S$ from $v_{1}$ to a leaf or to the next conjunction gate, we get a $\vee$-path that ends in the node $v_{k}$ whose d-node is still $t^{\prime}$ and $v_{k}$ is either a leaf or a conjunction gate.

If $t^{\prime}$ is a leaf of $T$ labeled with variable $x_{i}$, we must have that $v_{k}$ is a leaf of $S$ labeled with a literal $\left(x_{i}, a\right)$ for some $a \in \operatorname{dom}\left(x_{i}\right)$, it follows that $\left(x_{i}, a\right) \in \tau^{*} \subseteq \tau$ and $\left\{\left(y_{t}, v_{t}\right),\left(x_{i}, a\right)\right\} \in$ $\operatorname{rel}\left(c_{t, t^{\prime}}\right)$, constraint $c_{t, t^{\prime}}$ is thus satisfied by $\tau$.

If $t^{\prime}$ is an inner node, then $v_{k}$ is a conjunction gate $v_{t^{\prime}}$ associated with $t^{\prime}$ in $S$. It follows that $\left(y_{t^{\prime}}, v_{t^{\prime}}\right) \in \tau$ and $\left\{\left(y_{t}, v_{t}\right),\left(y_{t^{\prime}}, v_{t^{\prime}}\right)\right\} \in \operatorname{rel}\left(c_{t, t^{\prime}}\right)$, constraint $c_{t, t^{\prime}}$ is thus satisfied by $\tau$.

Theorem 14 now follows by the above construction from the following proposition.

- Theorem 21. $P=(\mathbf{z}, C)$ is a tree binary encoding of $c^{*}$ of polynomial size.

Proof. $C$ consists of $O(n)$ constraints and the total size of the domains of variables in $\mathbf{z}$ is bounded by the number of the nodes in $D$. Lemmas 19 and 20 imply that $P$ is a TBE of $c^{*}$.

## 5 Binary Constraint Graphs With Bounded Treewidth

In this section, we shall extend the construction from Section 3 to BCG constraints that naturally generalize BCT constraints.

- Definition 22. A BCG constraint $c$ is a pair $(\mathbf{x}, P)$ such that $P=(\mathbf{z}, C)$ is a normalized binary $C S P, \operatorname{scp}(c)=\mathbf{x} \subseteq \mathbf{z}$ and $\operatorname{rel}(c)=\operatorname{sol}(\mathbf{z}, C)[\mathbf{x}]$.

The construction we describe is parameterized by the treewidth of the underlying constraint graph and the domain size. The treewidth of a graph is defined using a tree decomposition.

Given an undirected graph $G=(V, E)$, its tree decomposition is defined as a pair $(T, \chi)$ where $T=\left(V_{T}, E_{T}\right)$ is a tree and $\chi: V_{T} \rightarrow \mathcal{P}(V)$ is a function that assigns each vertex $t \in V_{T}$ a subset of $V$ called a bag that satisfies the following conditions:
(d1) $V=\bigcup_{t \in V_{T}} \chi(t)$.
(d2) For each edge $\{u, v\} \in E$ there is a node $t \in V_{T}$ such that $\{u, v\} \subseteq \chi(t)$.
(d3) If a node $v$ is contained in two bags $\chi\left(t_{1}\right)$ and $\chi\left(t_{2}\right)$, then $v \in \chi(t)$ for every node $t$ on the path connecting $t_{1}$ with $t_{2}$.
The width of the tree decomposition is defined as $\max _{t \in V_{T}}|\chi(t)|-1$. The treewidth $t w(G)$ of $G$ is the minimum width among all possible tree decompositions of $G$. It should be noted that any tree decomposition of a graph $G$ on $n$ vertices can be transformed into a tree decomposition with the same width and $O(n)$ nodes [15].

We are now ready to formulate the main result of this section.

- Theorem 23. Assume that $c^{*}=(\mathbf{x}, P)$ is a $B C G$ constraint defined by a normalized binary CSP $P=(\mathbf{z}, C)$. Denote $G$ the constraint graph of $P$. Denote $m=|\mathbf{z}|$ and $d=\max _{z_{i} \in \mathbf{z}}\left|\operatorname{dom}\left(z_{i}\right)\right|$. Then there is an SDNNF $D$ representing $c^{*}$ with $O\left(m d^{t w(G)+1}\right)$ nodes and $O\left(m d^{2 t w(G)+1}\right)$ edges.

The proof of Theorem 23 is based on the following proposition.

- Theorem 24. Assume that $c^{*}=(\mathbf{x}, P)$ is a $B C G$ constraint defined by a normalized binary CSP $P=(\mathbf{z}, C)$. Denote $G$ the constraint graph of $P$. Denote $m=|\mathbf{z}|$ and $d=\max _{z_{i} \in \mathbf{z}}\left|\operatorname{dom}\left(z_{i}\right)\right|$. Then $c^{*}$ has a tree binary encoding $P^{\prime}=\left(\mathbf{z}^{\prime}, C^{\prime}\right)$ with $\left|\mathbf{z}^{\prime}\right|=O(m)$ and $\left|\operatorname{dom}\left(z_{i}^{\prime}\right)\right| \leq d^{t w(G)+1}$ for every $z_{i}^{\prime} \in \mathbf{z}^{\prime}$.

Note that Theorem 24 actually follows from Proposition 4 in [28] which is based on the encoding described in [11]. If we would simply combine the bound given by Theorem 24 with the bound given by Theorem 9 , we would get an SDNNF $D$ representing $c^{*}$ with

(a) Binary constraint graph

(b) Tree decomposition.

(c) Tree binary encoding.

Figure 5 Example of the construction, see the description in Example 25.
$O\left(m d^{t w(G)+1}\right)$ nodes and $O\left(m d^{2(t w(G)+1)}\right)$ edges. We provide a specific construction that proves Theorem 24 and that can be combined with Lemma 12 to prove a slightly better bound stated in Theorem 23. The construction we describe below is similar to the dual encoding described in [11].

Let us fix a BCG constraint $c^{*}=(\mathbf{x}, P)$ where $P=(\mathbf{z}, C)$ is a normalized binary CSP with constraint graph $G$. Let us assume that $\mathbf{z}=\left(z_{1}, \ldots, z_{m}\right)$ and $d=\max _{i=1}^{m}\left|\operatorname{dom}\left(z_{i}\right)\right|$. Let us also fix a tree decomposition $(T, \chi)$ of $G$. Let us assume that $V_{T}=\left(t_{1}, \ldots, t_{N}\right)$ for some $N=O(m)$. We shall describe a BCT $P^{\prime}=\left(\mathbf{z}^{\prime}, C^{\prime}\right)$ which is a TBE of $c^{*}$. First, let us define the variables in $\mathbf{z}^{\prime}$. We associate a new variable $v_{i}$ with every $t_{i}, i=1, \ldots, N$. Then we set $\mathbf{z}^{\prime}=\mathbf{z} \cup \mathbf{v}$ where $\mathbf{v}=\left(v_{1}, \ldots, v_{N}\right)$. The domains of variables in $\mathbf{z}$ are given by $c^{*}$. For every $v_{i}, i=1, \ldots, N$, we set the domain as follows. Let us consider the set of constraints defined on variables from $\chi\left(t_{i}\right)$ as $C_{i}=\left\{c \in C \mid \operatorname{scp}(c) \subseteq \chi\left(t_{i}\right)\right\}$. Then the domain of $v_{i}$ is defined as the set of solutions to $\operatorname{CSP}\left(\chi\left(t_{i}\right), C_{i}\right)$, i.e. $\operatorname{dom}\left(v_{i}\right)=\operatorname{sol}\left(\chi\left(t_{i}\right), C_{i}\right)$.

Let us now define the constraints in $C^{\prime}$.
(T1) For every edge $\left\{t_{i}, t_{j}\right\} \in E_{T}$ we add a constraint $c_{i, j}^{\prime}$ into $C^{\prime}$ with $\operatorname{scp}\left(c_{i, j}^{\prime}\right)=\left\{v_{i}, v_{j}\right\}$. The constraint relation $\operatorname{rel}\left(c_{i, j}^{\prime}\right)$ consists of pairs $\left\{\left(v_{i}, \tau_{1}\right),\left(v_{j}, \tau_{2}\right)\right\}$ where $\tau_{1} \in \operatorname{dom}\left(v_{i}\right)$, $\tau_{2} \in \operatorname{dom}\left(v_{j}\right)$, and $\tau_{1}\left[\chi\left(t_{i}\right) \cap \chi\left(t_{j}\right)\right]=\tau_{2}\left[\chi\left(t_{i}\right) \cap \chi\left(t_{j}\right)\right]$.
(T2) For every $z_{i}, i=1, \ldots, m$, we pick a representative node $t_{r_{i}} \in V_{T}$ satisfying $z_{i} \in \chi\left(t_{r_{i}}\right)$. We then add a constraint $c_{i}^{\prime}$ into $C^{\prime}$ with $\operatorname{scp}\left(c_{i}^{\prime}\right)=\left\{z_{i}, t_{r_{i}}\right\}$. The set of tuples $\operatorname{rel}\left(c_{i}^{\prime}\right)$ consists of pairs $\left\{\left(z_{i}, a\right),\left(v_{r_{i}}, \tau\right)\right\}$ where $a \in \operatorname{dom}\left(z_{i}\right), \tau \in \operatorname{dom}\left(v_{r_{i}}\right)$, and $\left(z_{i}, a\right) \in \tau$.

- Example 25. Let us consider a binary CSP $P=(\mathbf{z}, C)$ with $\mathbf{z}=\left\{z_{1}, \ldots, z_{5}\right\}$ whose constraint graph $G$ is depicted in Figure 5a. We shall use $c_{i, j} \in C$ to denote the constraint with scope $\left\{z_{i}, z_{j}\right\}$. Figure 5 b shows a tree decomposition $T$ of the graph with the contents of the bags inside the rectangles. The structure of the tree binary encoding $P^{\prime}$ of $P$ is then shown in Figure 5c. The domain of variable $v_{1}$ consists of tuples $\tau$ on variables $z_{1}, z_{2}$, and $z_{3}$ satisfying constraints $c_{1,2}, c_{2,3}$, and $c_{1,3}$. Assume a tuple $\sigma^{\prime} \in \operatorname{sol}\left(P^{\prime}\right)$. Constraint $c_{2}^{\prime}$ makes sure that if $\left(z_{2}, a\right) \in \sigma^{\prime}$, then $\sigma^{\prime}$ contains $\left(v_{2}, \tau\right)$ satisfying $\left(z_{2}, a\right) \in \tau$. Similarly for other variables. Constraints $c_{1,2}^{\prime}$ and $c_{2,3}^{\prime}$ extend this property also to nodes $v_{2}$ and $v_{3}$. The tuples assigned to variables $v_{1}, v_{2}$, and $v_{3}$ are thus consistent with each other and also with constraints $C$. We thus have that $\sigma^{\prime}[\mathbf{z}] \in \operatorname{sol}(P)$.

The proof of the correctness of the above construction and thus also the proof of Theorem 24 is moved to the appendix. Here, we will describe its application for proving the main result of this section.

Proof of Theorem 23. Using Theorem 24, we obtain a TBE encoding $P^{\prime}=\left(\mathbf{z}^{\prime}, C^{\prime}\right)$ with $O(m)$ variables and the domain sizes bounded by $d^{t w(G)+1}$. We can then apply Theorem 9 to obtain an SDNNF $D$ that represents $c^{*}$. $D$ has $O\left(m d^{t w(G)+1}\right)$ nodes. By Lemma 12 , the number of edges of $D$ is bounded by $O\left(m d^{t w(G)+1}+s\right)$ where $s=\sum_{c^{\prime} \in C^{\prime}}\left|\operatorname{rel}\left(c^{\prime}\right)\right|$. Since $\left|C^{\prime}\right|=O(m)$, it is enough to show that $\left|\operatorname{rel}\left(c^{\prime}\right)\right| \leq d^{2 t w(G)+1}$ for every $c^{\prime} \in C^{\prime}$.

Assume first a constraint $c_{i, j}^{\prime}$ added in step (T1). We may assume that $G$ is connected (otherwise we process each connected component of $G$ separately) and therefore $\chi\left(t_{i}\right) \cap \chi\left(t_{j}\right) \neq$ $\emptyset$. The number of pairs of tuples $\tau_{1}$ and $\tau_{2}$ that satisfy $\tau_{1}\left[\chi\left(t_{i}\right) \cap \chi\left(t_{j}\right)\right]=\tau_{2}\left[\chi\left(t_{i}\right) \cap \chi\left(t_{j}\right)\right]$ is thus at most $d^{t w(G)+1} \cdot d^{t w(G)}=d^{2 t w(G)+1}$. Therefore $\left|\operatorname{rel}\left(c_{i, j}^{\prime}\right)\right| \leq d^{2 t w(G)+1}$.

Assume now a constraint $c_{i}^{\prime}$ added in step (T2). The number of tuples $\tau$ satisfying that $\left(z_{i}, a\right) \in \tau$ for one particular $a \in \operatorname{dom}\left(z_{i}\right)$ is at most $d^{t w(G)}$ and thus $\left|\operatorname{rel}\left(c_{i, j}^{\prime}\right)\right| \leq d^{t w(G)+1} \leq$ $d^{2 t w(G)+1}$.

Note that the size estimate in Theorem 23 is only an upper bound and the real size of $P^{\prime}$ and the SDNNF $D$ depends on the particular tree decomposition and, in particular, on how much the bags intersect. Therefore, there is a space for optimization in a practical setting.

## 6 Conclusion

As the main result of our paper, we have shown that binary constraint trees are polynomially equivalent to structured DNNF circuits. We would like to note that for a given BCT $P$ the construction in Section 3 leads to a deterministic SDNNF $D_{P}$ (thanks to rule 3 in step (A2b)). This means that for every pair of distinct children $v_{1}$ and $v_{2}$ of a disjunction node, the sub-NNFs rooted at $v_{1}$ and $v_{2}$ do not share any models (see [9, 20]). This property allows for instance model counting on $D_{P}$. However, forgetting the hidden variables from $D_{P}$ does not preserve determinism in general [10] and thus the actual result of the construction is not a deterministic SDNNF. Introducing hidden variables is thus an important part of the construction described in Section 4 since SDNNFs are strictly more succinct than deterministic SDNNFs [20].

Several rules for reducing the number of hidden variables in a BCT constraint were described in [27], it would be interesting to investigate the effect of these rules on a SDNNF that is compiled into a BCT constraint, reduction rules are applied to it and then it is compiled back to a SDNNF. When compiling the BCT constraint back to a SDNNF, we can pick an arbitrary node of the constraint tree as a root which allows us to change the structure of the SDNNF to a different orientation of the v-tree. This, for instance, extends the applicability of the conjoin operation described in [20] to conjoining two SDNNFs whose v -trees differ, but their undirected versions are the same.

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## A Proof of Theorem 24

In this section, we shall prove the correctness of the construction described in Section 5. We shall also prove Theorem 24 that states the properties of the construction. We use the same notation that was used in Section 5. In particular, we assume a fixed BCG constraint $c^{*}=(\mathbf{x}, P)$ where $P=(\mathbf{z}, C)$ is a normalized binary CSP with constraint graph $G$. We assume that that BCT $P^{\prime}=\left(\mathbf{z}^{\prime}, C^{\prime}\right)$ is the result of the construction from Section 5 . We shall show that $P^{\prime}$ is a TBE of $c^{*}$. The construction of $C^{\prime}$ implies the following property.

- Lemma 26. Let $p \in\{1, \ldots, m\}$ be arbitrary and let $t_{i} \in V_{T}$ be such that $z_{p} \in \chi(t)$. Assume that $\sigma^{\prime} \in \operatorname{sol}\left(P^{\prime}\right)$ and assume that $\left(z_{p}, a\right),\left(v_{i}, \tau\right) \in \sigma^{\prime}$. Then $\left(z_{p}, a\right) \in \tau$.

Proof. Let $t_{r_{p}}$ be the representative node picked for $z_{p}$ in step (T2) and consider the path $t_{r_{p}}=t_{j_{1}}, t_{j_{2}}, \ldots, t_{j_{k}}=t_{i}$ in $T$ connecting $t_{r_{p}}$ with $t_{i}$. For every $q=1, \ldots, k$ we have that $z_{p} \in \chi\left(t_{r_{p}}\right) \cap \chi\left(t_{i}\right)$ and thus $z_{p} \in \chi\left(t_{j_{q}}\right)$ by condition (d3). Denote $\tau_{q} \in \operatorname{dom}\left(v_{j_{q}}\right)$ the tuple for which $\left(v_{j_{q}}, \tau_{q}\right) \in \sigma^{\prime}$. We shall show by induction on $q$ that $\left(z_{p}, a\right) \in \tau_{q}$ for every $q=1, \ldots, k$. Since $\tau_{k}=\tau$, we then have that $\left(z_{p}, a\right) \in \tau$.

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Since $t_{1}=t_{r_{p}}$, we have that $\left(z_{p}, a\right) \in \tau_{1}$ by constraint $c_{p}^{\prime}$ added to $C^{\prime}$ in step (T2). Assume now that $q>1$ and consider constraint $c_{j_{q-1}, j_{q}}^{\prime}$ added in step (T1). The induction hypothesis implies that $\left(z_{p}, a\right) \in \tau_{q-1}$, and by definition of $\operatorname{rel}\left(c_{j_{q-1}, j_{q}}^{\prime}\right)$ we also have that $\left(z_{p}, a\right) \in \tau_{q}$.

We are now ready to prove the correctness of the construction.
Lemma 27. ( $\left.\mathbf{x}, P^{\prime}\right)$ is a tree binary encoding of $B C G$ constraint $c^{*}=(\mathbf{x}, P)$.
Proof. The constraint graph of $P^{\prime}$ is a tree that originates from $T$ by adding leaves corresponding to the constraints $c_{i}^{\prime}$ added in step (T2). We shall show that $\operatorname{sol}\left(P^{\prime}\right)[\mathbf{z}]=\operatorname{sol}(P)$. Then $\operatorname{rel}\left(c^{*}\right)=\operatorname{sol}(\mathbf{z}, C)[\mathbf{x}]=\operatorname{sol}\left(\mathbf{z}^{\prime}, C^{\prime}\right)[\mathbf{x}]$ and the proposition follows.

Assume first that we have a solution $\sigma^{\prime} \in \operatorname{sol}\left(P^{\prime}\right)$. Denote $\sigma=\sigma^{\prime}[\mathbf{z}]$ and let us show that $\sigma$ satisfies all constraints of $P$. Let $c \in C$ be a constraint with $\operatorname{scp}(c)=\left\{z_{p}, z_{q}\right\}$. We have $\left(z_{p}, a\right) \in \sigma$ and $\left(z_{q}, b\right) \in \sigma$ for some $a \in \operatorname{dom}\left(z_{p}\right)$ and $b \in \operatorname{dom}\left(z_{q}\right)$. By condition (d2) we have that $\operatorname{scp}(c) \subseteq \chi\left(t_{i}\right)$ for some $t_{i} \in V_{T}$. Consider literal $\left(v_{i}, \tau\right) \in \sigma^{\prime}$. By Lemma 26, we have that $\left(z_{p}, a\right) \in \tau$ and $\left(z_{q}, b\right) \in \tau$. Since $\tau \in \operatorname{dom}\left(v_{i}\right)$, we have that $\left\{\left(z_{p}, a\right),\left(z_{q}, b\right)\right\}=$ $\tau[\operatorname{scp}(c)] \in \operatorname{rel}(c)$. Since this holds for every constraint $c \in C$, we get that $\sigma \in \operatorname{sol}(P)$.

Assume now that we have a solution $\sigma \in \operatorname{sol}(P)$. Let us now define a tuple $\sigma^{\prime}=$ $\sigma \cup\left\{\left(v_{i}, \sigma\left[\chi\left(t_{i}\right)\right]\right) \mid i=1, \ldots, N\right\}$. Since $\sigma \in \operatorname{sol}(P)$, we have that $\sigma\left[\chi\left(t_{i}\right)\right] \in \operatorname{rel}\left(C_{i}\right)$ and thus $\sigma\left[\chi\left(t_{i}\right)\right] \in \operatorname{dom}\left(v_{i}\right)$. Tuple $\sigma^{\prime}$ is thus correctly defined. It also satisfies all constraints (T1) and (T2) and thus $\sigma^{\prime} \in \operatorname{sol}\left(P^{\prime}\right)$.

Proof of Theorem 24. Assume that $P^{\prime}=\left(\mathbf{z}^{\prime}, C^{\prime}\right)$ is constructed as above. Then $\left(\mathbf{x}, P^{\prime}\right)$ is a tree binary encoding of $c^{*}=(\mathbf{x}, P)$ by Lemma 27. We may assume by [15] that $\left|V_{T}\right|=O(m)$ and thus $\left|\mathbf{z}^{\prime}\right|=m+\left|V_{T}\right|=O(m)$. For every variable $z_{i} \in \mathbf{z}$ we have $\left|\operatorname{dom}\left(z_{i}\right)\right| \leq d$ by assumption. For every variable $v_{i} \in \mathbf{z}^{\prime} \backslash \mathbf{z}$ we have that $\left|\chi\left(t_{i}\right)\right| \leq t w(G)+1$ and thus $\left|\operatorname{dom}\left(v_{i}\right)\right| \leq d^{t w(G)+1}$ by the definition of $\operatorname{dom}\left(v_{i}\right)$.

