

A Compact Representation for Syntactic Dependencies in QBFs

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TNF

QBF: Quantified Boolean Formulae.

QDPLL: DPLL-like QBF solvers, formulae in prenex CNF.

- decision order must respect “quantification order”.

Example (Dependencies in QBF)

$\forall x \exists y. (x \vee \neg y) \wedge (\neg x \vee y)$ is satisfiable. Value of y depends on value of x .
→ erroneously conclude unsatisfiability if y is assigned before x .

Our Results:

- given: syntactic dependency relation D .
- static and compact dependency graph (DAG) representing D .
- graph is applicable to QBF solvers of any kind.
- in QDPLL: find assignable variables before decision-making.
- experiments: structured formulae from QBFEVAL 2005 - 2008.

QBFs in Prenex CNF: $S_1 \dots S_n. \phi$

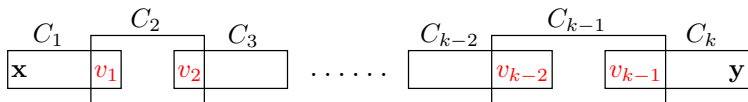
- ϕ in CNF and quantifier-free, quantified variables $V = V_{\exists} \cup V_{\forall}$.
- scopes $S_1 < \dots < S_n$, ordered by nesting $\delta(S_i) = i$, type $q(S_i) \in \{\forall, \exists\}$.

Dependency Schemes [SamerSzeider-JAR'09]:

- relation $D \subseteq (V_{\exists} \times V_{\forall}) \cup (V_{\forall} \times V_{\exists})$.
- $y \in D(x)$: “ y depends on x ”, i.e. assign x before y in QDPLL.
- $|D_1| < |D_2|$: D_1 less restrictive, i.e. more freedom for decisions in QDPLL.

Example (Trivial Dependency Scheme)

$D^{\text{triv}}: y \in D^{\text{triv}}(x) \Leftrightarrow \delta(x) < \delta(y) \text{ and } q(x) \neq q(y).$



Definition (X-path)

For $x, y \in V$, $X \subseteq V$, an X -path between x and y is a sequence C_1, \dots, C_k of clauses where $x \in C_1$, $y \in C_k$ and $C_i \cap C_{i+1} \cap X \neq \emptyset$ for $1 \leq i < k$.

- For $x \in V$: if $q(x) = \exists$ then $\overline{q(x)} := \forall$ and $\overline{q(x)} := \exists$ otherwise.
- For a QBF and $q \in \{\exists, \forall\}$: $V_{q,i} := \{y \in V_q \mid i \leq \delta(y)\}$.

Definition (Standard Dependency Scheme)

For $x \in V$, $i = \delta(x) + 1$:

$D^{\text{std}}(x) = \{y \in V_{\overline{q(x)},i} \mid \text{there is an } X\text{-path between } x \text{ and } y \text{ for } X = V_{\exists,i}\}$.

- $D^{\text{std}}(x)$ contains all differently quantified, *larger* y which are connected to x over *existential* variables *larger* than x .
- Observe: $|D^{\text{std}}| < |D^{\text{triv}}|$.

In Practice: computing *full* D^{std} in $O(|V| \cdot |\phi|)$ time.

- traversing clauses in ϕ for *each* $x \in V$.
- too expensive to be done dynamically at decision points within QDPLL.

Our Work:

- *static* and *compact* graph representation (DAG) for D^{std} .
- graph is built once, serves as over-approximation for exact D^{std} .
- classes of variables represent connection information.
- connection information is shared between variables.

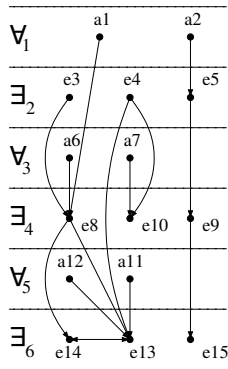
Definition (Variable Connection)

For $x, y \in V$, x is *connected* to y wrt. scope S_i ($x \rightarrow_i y$) iff. $y \in V_{\exists, i \leq \delta(y)}$ and $x, y \in C$ for $C \in \phi$. Relation \rightarrow_i^* is the refl. trans. closure of \rightarrow_i .

Example (ongoing)

i	$q(S_i)$	S_i	
1	\forall	$a1, a2$	$(a2, e5, e9)$
2	\exists	$e3, e4, e5$	$(e5, e9, e15)$
3	\forall	$a6, a7$	$(e3, e8, e13)$
4	\exists	$e8, e9, e10$	$(e3, e8, e13)$
5	\forall	$a11, a12$	$(e4, e13, e14)$
6	\exists	$e13, e14, e15$	$(a1, a6, e8, e14)$ $(a11, a12, e13)$

- trans. edges not shown.
- $e3 \rightarrow_4 e8$ but $e3 \not\rightarrow_5 e8$.
- $e3 \rightarrow_2^* e14$ and also $e14 \rightarrow_2^* e3$.

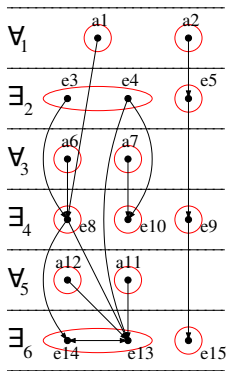


Definition (Equivalence)

For $x, y \in V$, x is *equivalent* to y ($x \approx y$) iff. either (1) $x = y$ or (2) $q(x) = q(y) = \exists$, $\delta(x) = \delta(y) = i$ and $x \rightarrow_i^* y$.

Example (continued)

- $e3 \approx e4$ since $q(e3) = q(e4) = \exists$,
 $\delta(e3) = \delta(e4) = 2$ and $e3 \rightarrow_2^* e4$.
 - $e5 \not\approx e4$ because $e5 \not\rightarrow_2^* e4$.
 - trivially $a11 \approx a11$ and $e3 \not\approx e14$.
-
- \rightarrow_i^* on \approx potentially more compact.
 - $[x] \rightarrow_i^* [y]$: connection between classes.



Goal: compact representation of *all* existential class connections for D^{std} .

Problem: with \rightarrow_i^* on \approx still need to search for connected classes.

Definition (Directed Connection)

For $x \in V, y \in V_{\exists}$, $[x] \rightsquigarrow^* [y]$ iff. $\delta(x) \leq \delta(y)$ and $x \rightarrow_i^* y$ for $i = \delta(x)$.
Relation \rightsquigarrow is the refl. trans. reduction of \rightsquigarrow^* .

- $[x] \rightsquigarrow^* [y]$ respects scope ordering, excludes variables smaller than x .

Lemma (Connection Forest, C-Forest)

For V_{\exists} , \rightsquigarrow can be represented as a forest.

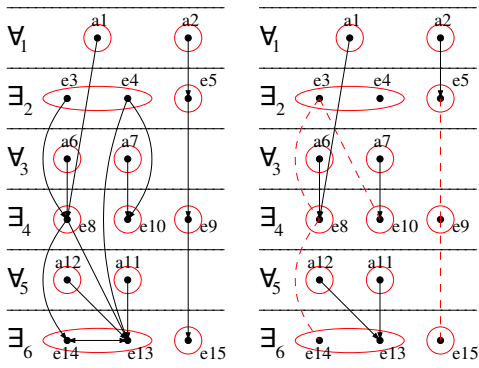
C-Forest: compact representation of all *existential* connections.

- no more searching: classes are connected to all of their descendants.

Example (continued)

Right figure:

- \rightsquigarrow on V_{\exists} : c-forest.
- $[e3] \rightsquigarrow [e8]$
- $[e8] \rightsquigarrow [e14]$
- $[e3] \rightsquigarrow^* [e14]$
- also $[e4] \rightsquigarrow^* [e14]$
since $[e4] = [e3]$



Practical Problem: how to find connected descendants of x in c-forest?

Essentially: need set of “root classes” for each $x \in V$.

- descendants of root classes *exactly* represent all connections of x .
- computing D^{std} : c-forest + root classes.

Finding Root Classes: finding smallest ancestors in c-forest.

Definition (Smallest Ancestor)

For $y \in V_{\exists}$, $i \leq \delta(y)$ and the c-forest, $h(i, [y]) = [y']$ denotes the smallest ancestor of $[y]$ in the c-forest such that $i \leq \delta(y')$.

Definition (Root Classes)

For $x \in V$, $H_i(x) := \{[z] \mid [z] = h(i, [y]) \text{ for } [y] \text{ where } x \rightarrow_i y\}$ is the set of root classes of x with respect to scope S_i .

- Finding root classes $H_i(x)$ starting from clauses containing x .

Definition (Root Class Descendants)

For $x \in V$, $H_i^*(x) := \{[y] \mid [z] \rightsquigarrow^* [y] \text{ for } [z] \in H_i(x)\}$ is the set of root class descendants of x with respect to scope S_i .

- Sets H_i^* are sufficient for computing D^{std} from c-forest.

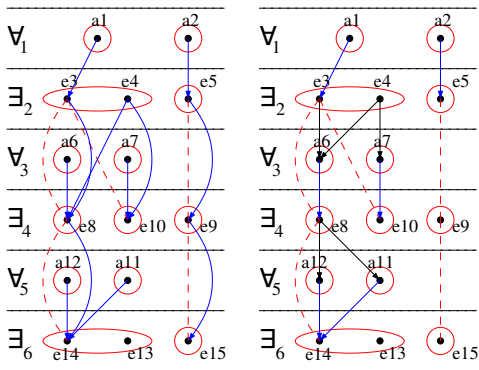
Theorem (D^{std} by Checking Root Class Decendants)

For $x \in V$, $i = \delta(x) + 1$:

$$D^{\text{std}}(x) = \{y \in V_{\frac{q(x),i}{q(x),i}} \mid H_i^*(x) \cap H_j^*(y) \neq \emptyset \text{ for } j = \delta(y)\}.$$

Example (continued)

- left figure: blue edges \rightarrow as sets $H_i(x)$ for $i = \delta(x) + 1$, $x \in V$.
- right figure: D^{std} graph
- implicitly:
 - $e_{15} \in D^{\text{std}}(a_2)$
 - $e_{13} \in D^{\text{std}}(a_1)$
 - $e_{13} \in D^{\text{std}}(a_{11})$



Graph for D^{std} : c-forest as core

- already present: for $x \in V_V$, $i = \delta(x) + 1$: $D^{\text{std}}(x) = H_i^*(x)$
- for $x \in V_E$: insert edges to represent $D^{\text{std}}(x)$

Experimental Results

	QBFEVAL'05	QBFEVAL'06	QBFEVAL'07	QBFEVAL'08
<i>size</i>	211	216	1136	3328
<i>total time</i>	7.94	1.35	227.05	300.31
<i>max. time</i>	0.58	0.03	7.96	8.11
<i>avg. time</i>	0.04	0.01	0.2	0.09
$x \in V_{\forall}$				
<i>max.</i> $ D^{\text{std}}(x) $	256535	9993	2177280	2177280
<i>avg.</i> $ D^{\text{std}}(x) $	82055.87	4794.60	33447.6	19807
<i>max.</i> $ H_i(x) $	256	1	518	518
<i>avg.</i> $ H_i(x) $	3.26	0.98	2.02	1.14
<i>max.</i> $ H_i^*(x) $	797	5	797	1872
<i>avg.</i> $ H_i^*(x) $	19.51	1.12	39.06	8.24
<i>avg.</i> $\frac{ \{y \in D^{\text{std}}(x)\} }{ \{y \in D^{\text{std}}(x)\} }$	3.44%	0.04%	6.42%	1.21%
$x \in V_{\exists}$				
<i>max.</i> $ D^{\text{std}}(x) $	5040	440	5040	22696
<i>avg.</i> $ D^{\text{std}}(x) $	12.76	2.98	3.24	4
<i>max.</i> $ H_i(x) $	24	7	490	490
<i>avg.</i> $ H_i(x) $	0.14	0.13	0.17	0.13
<i>max.</i> $ H_i^*(x) $	797	7	797	1872
<i>avg.</i> $ H_i^*(x) $	5.16	0.16	1.32	1.31
<i>avg.</i> $\frac{ \{y \in D^{\text{std}}(x)\} }{ \{y \in D^{\text{std}}(x)\} }$	2.37%	0.4%	2.76%	2.09%
<i>classes per variables</i>	10.96%	4.99%	11.45%	7.11%

QDPLL for QBF: quantification order matters.

- limited freedom for decisions.

Dependency Schemes:

- dependency relations $D \subseteq (V_{\exists} \times V_{\forall}) \cup (V_{\forall} \times V_{\exists})$.
- $y \in D(x)$: assign x before y in QDPLL.
- Standard Dependency Scheme D^{std} : based on variable connections.

Achievements: static, compact graph representation D^{std} for QBF in PCNF.

- compactness: connection relation on equivalence classes.
- c-forest: sharing connection information.
- two orders of magnitude more compact than simple graph.

Ongoing and Future Work:

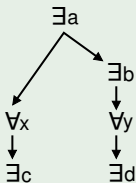
- integration into QDPLL: maintaining top-down “decision frontier”.
- comparing dependency schemes in QDPLL.

Example

$$\exists a, b \forall x, y \exists c, d. (a \vee x \vee c) \wedge (a \vee b) \wedge (b \vee d) \wedge (y \vee d)$$

After minimizing $\exists c, \exists d, \forall x$ and $\forall y$, *non-deterministic* choice:

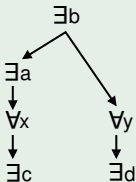
1. minimize $\exists b$ before $\exists a$



Extract D^{tree} from parse tree (descendants):

$$1. D^{\text{tree}} = \{(a, x), (x, c), (a, y), (b, y), (y, d)\}$$

2. minimize $\exists a$ before $\exists b$



$$2. D^{\text{tree}} = \{(b, x), (a, x), (x, c), (b, y), (y, d)\}$$

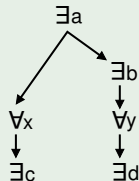
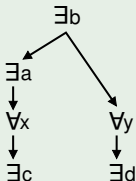
Example (continued)

$$\exists a, b \forall x, y \exists c, d. (a \vee x \vee c) \wedge (a \vee b) \wedge (b \vee d) \wedge (y \vee d)$$

By D^{std} we get:

$$\begin{array}{c} \exists a \\ \downarrow \\ \forall x \\ \downarrow \\ \exists c \end{array}$$

$$\begin{array}{c} \exists b \\ \downarrow \\ \forall y \\ \downarrow \\ \exists d \end{array}$$



Either $(a, y) \in D^{\text{tree}}$ or $(b, x) \in D^{\text{tree}}$ but *both* $(a, y) \notin D^{\text{std}}$ and $(b, x) \notin D^{\text{std}}$.

Theorem

For $x \in V, i = \delta(x) + 1$:

$$D^{\text{std}}(x) = \{y \in V_{\overline{q(x)},i} \mid \exists w \in V_{\exists,i} : x \rightarrow_i^* w \text{ and } y \rightarrow_i^* w\} \quad (1)$$

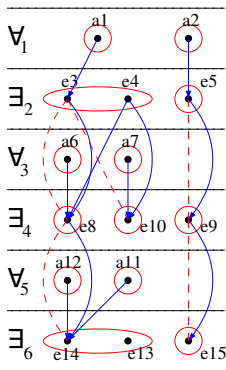
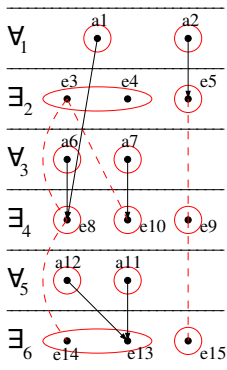
$$= \{y \in V_{\overline{q(x)},i} \mid \exists w \in V_{\exists,i} : x \rightarrow_i^* [w] \text{ and } [y] \rightarrow_i^* [w]\} \quad (2)$$

$$= \{y \in V_{\overline{q(x)},i} \mid \exists w \in V_{\exists,i} : x \rightarrow_i^* [w] \text{ and } [y] \rightsquigarrow_i^* [w]\} \quad (3)$$

Example (continued)

Blue edges \rightarrow : sets $H_i(x)$ for $i = \delta(x) + 1$, $x \in V$.

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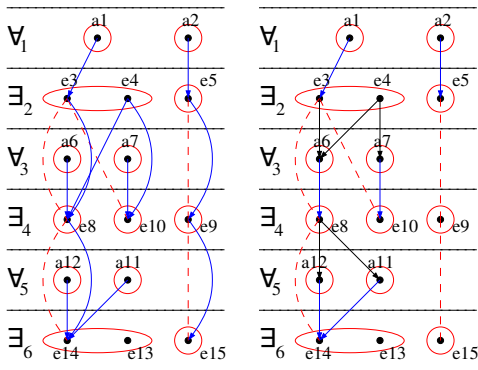
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Example (continued)

- right figure: D^{std} graph
- implicitly:
 - $e15 \in D^{\text{std}}(a2)$
 - $e13 \in D^{\text{std}}(a1)$
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Graph for D^{std} :

- already present: for $x \in V_V, i = \delta(x) + 1$: $D^{\text{std}}(x) = H_i^*(x)$
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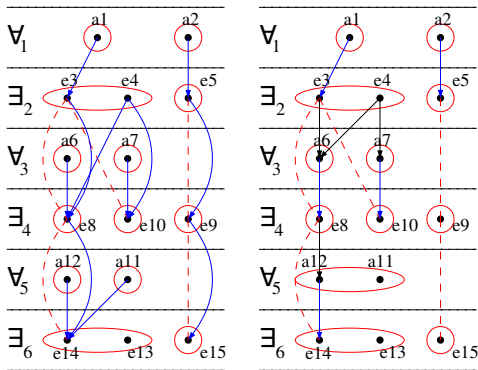
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 - $e13 \in D^{\text{std}}(a11)$
 - $a11 \in D^{\text{std}}(e8)$

Optimization: merging universal classes [a11], [a12] with same H_i .

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