Knowledge Compilation und #SAT

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• The SAT Problem (SAT).

SAT

- Given a Boolean formula φ of n variables.
- ? Find an assignment that satisfies φ .

- The SAT Problem (SAT).
- Counting SAT Problem (**#SAT**).

#SAT

- Given a Boolean formula φ of n variables.
- ? How many assignments in $2^{\mathrm{Var}(\varphi)}$ satisfy φ ?

- The SAT Problem (SAT).
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Notation

Let $SAT(\chi) \subseteq 2^{VAR(\chi)}$ be the set of all satisfying assignments of χ $SAT(\chi) = \{\rho : VAR(\chi) \to \{0,1\} : \rho(\chi) = 1\}.$

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SAT: Is $SAT(\varphi) = \emptyset$. #SAT: Find $|SAT(\varphi)|$.

- The SAT Problem (SAT).
- Counting SAT Problem (#SAT).

Example

$$\varphi = X_1 \wedge (X_2 \vee \neg X_3)$$

Clearly, $\#SAT(\varphi) = 3$.

- The SAT Problem (SAT).
- Counting SAT Problem (#SAT).
- Negation Normal Form (NNF).

Negation Normal Form

A Boolean formula φ is in NNF form, if it contains only disjunctions and conjunctions over a set of positive and negative literals.

Example. $\varphi = X_1 \vee \neg X_2$.

- The SAT Problem (SAT).
- Counting SAT Problem (#SAT).
- Negation Normal Form (NNF).
- Conjunctive Normal Form (CNF).

Conjunctive Normal Form

A Boolean formula φ is in CNF, if it is a conjunction of one or more clauses, where each clauses is a disjunction of one or more literals. Note that each CNF formula is an NNF formula as well.

- The SAT Problem (SAT).
- Counting SAT Problem (#SAT).
- Negation Normal Form (NNF).
- Conjunctive Normal Form (CNF).
- Decomposable Negation Normal Form (DNNF).

Decomposable Negation Normal Form

A Boolean formula φ is in DNNF, if it is in NNF and for each conjunction subformula $phi' := \psi_1 \wedge \psi_2$ we have $VAR(\psi_1) \cap VAR(\psi_2) = \emptyset$.

• Satisfying each subformula is independent.

- The SAT Problem (SAT).
- Counting SAT Problem (#SAT).
- Negation Normal Form (NNF).
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- deterministic Decomposable Negation Normal Form (d-DNNF).

deterministic Decomposable Negation Normal Form

A Boolean formula φ is in d-DNNF, if it is in DNNF and for each disjunction subformula $\varphi' = \psi_1 \vee \psi_2$ we have $SAT(\psi_1) \cap SAT(\psi_2) = \emptyset$.

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- decision Decomposable Negation Normal Form (dec-DNNF).

decision Decomposable Negation Normal Form

A Boolean formula φ is in dec-DNNF, if it is in DNNF and each disjunction subformula φ' is of the form $\varphi' = (X \wedge \psi_1) \vee (\neg X \wedge \psi_2)$ for some variable $X \in \mathrm{VAR}(\varphi)$.

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Note. Each dec-DNNF is a d-DNNF.

Examples

[On the blackboard..]

Goals for today

Build a dec-DNNF of polynomial size for β -acyclic graphs.

Show a subclass of β -acyclic formulas, where each *structured* d-DNNF has exponential size.

Assignments

• Given a CNF Formula φ , an **assignment** for C is a function $\tau: VAR(C) \rightarrow \{0,1\}.$

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- A partial assignment $\tau_{|V'}$ satisfies a CNF-formula φ $(\tau_{|V'} \models \varphi)$, if for each clause $C \in \varphi$ there is a variable $v \in VAR(C) \cap V'$ such that $\tau_{|V'}(v) = 1$ if and only if v appears in C as a positive literal.

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Example

$$\varphi := (v_1 \vee \neg v_2 \vee v_3) \wedge (v_1 \vee v_2) \wedge (\neg v_2 \vee \neg v_3)$$

For $V' = \{v_1, v_2\}, \tau_{|V'}(v_1) = 1, \tau_{|V'}(v_2) = 0$, the partial assignment $\tau_{|V'}$ satisfies φ .

- Hypergraph \mathcal{H} .
 - A set of vertices $V(\mathcal{H})$.
 - Edges $E(\mathcal{H})$, defined as subsets over $V(\mathcal{H})$.

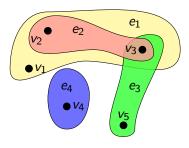


Figure: https://tex.stackexchange.com/a/11

https://tex.stackexchange.com/a/1195/163902

- Hypergraph \mathcal{H} .
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- A walk is sequence $(e_1, x_1, \dots, x_n, e_{n+1})$, $e_i \in \mathcal{H}, x_i \in V(\mathcal{H})$ and $x_i \in e_i \cap e_{i-1}$ for all $i \in [n]$.

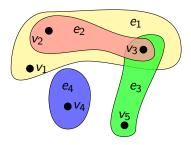


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- A path is a walk that never goes twice through the same vertex nor the same edge.

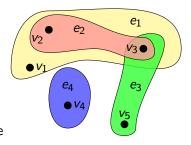


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- Different ways to translate acyclicity to hypergraphs.

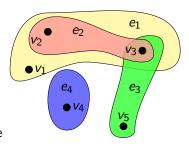
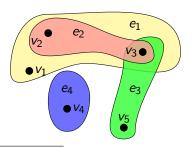


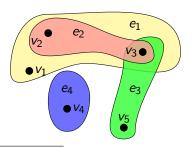
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• Let $\rho := v_1, \dots v_n$ be an enumeration of the vertices.



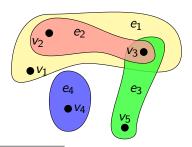
 $^{^{1}}e_{i>i}:=e\cap\{v_{i},\ldots v_{n}\}.$

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- ρ is a β -elimination, if for all¹



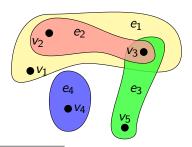
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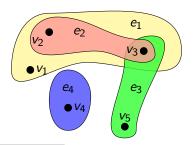
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- ullet A hypergraph is eta-acyclic, if it admits a eta-elimination.



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- A hypergraph is β -acyclic, if it admits a β -elimination.
- We define $V_{\leq v_i} := \{v_j; j \leq i\}$.

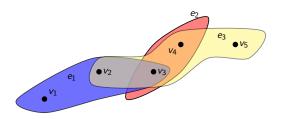


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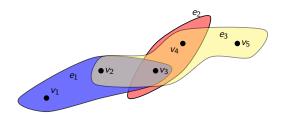


Polynomial upper-bound on the practical method

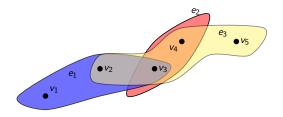
• Let \mathcal{H} be a β -acyclic graph and $v_1, \ldots v_n$ a β -elimination.



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- \mathcal{H}_e^{\times} denotes the subgraph of \mathcal{H} , that contain the edges f, such that there is a walk from f to e that goes only through edges smaller than e and vertices smaller than (or equal to) x.



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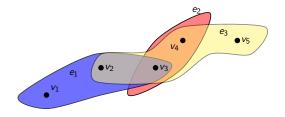


Figure: Note that $e_2 \notin \mathcal{H}_{e_3}^{v_2}$ meanwhile $e_2 \in \mathcal{H}_{e_3}^{v_3}$

Lemma (lemma 2)

For $x, y \in V(\mathcal{H}), x \leq y$ and for $e, f \in \mathcal{H}, e \leq f$,

$$\text{if } V(\mathcal{H}_e^x) \cap V(\mathcal{H}_f^y) \cap V_{\leq x} \neq \emptyset, \text{ then } \mathcal{H}_e^x \subseteq \mathcal{H}_f^y.$$

In particular, for all $y \in V(\mathcal{H})$,

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Proof sketch. For $g \in \mathcal{H}_e^x$, there is a path from g to e using edges smaller than e and vertices smaller than x.

There is also a path from e to f. Concatenate both paths to get a path from g to f.

Lemma (lemma 4)

For $e, f \in \mathcal{H}, e \leq f$, If there exists a vertex $x \in V(\mathcal{H})$, such that $x \in e \cap f$, then $e \cap V_{>x} \subseteq f$.

Lemma (lemma 4)

For $e, f \in \mathcal{H}, e \leq f$, If there exists a vertex $x \in V(\mathcal{H})$, such that $x \in e \cap f$, then $e \cap V_{>x} \subseteq f$.

Proof sketch. If $y \in e \setminus f$ such that y > x, then \mathcal{H} is not β -acyclic.

A path $(e_1, x_1, \dots e_{n+1})$ is called decreasing, if $e_i > e_{i+1}$ and $x_i > x_{i+1}$ for all i.

Lemma (lemma 5)

For $x \in V(\mathcal{H})$, $e \in \mathcal{H}$ and $f \in \mathcal{H}_e^x$, there exists a decreasing path from e to f going through vertices smaller than x.

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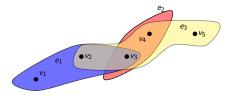
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Proof sketch. Any shortest path from e to f is decreasing. A path exists by definition.

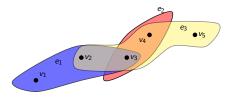
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For every $x \in V(\mathcal{H})$ and $e \in \mathcal{H}, V(\mathcal{H}_e^x) \cap V_{\geq x} \subseteq e$



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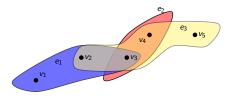
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Intuitively, this allows us to use dynamic programming, since all variables in \mathcal{H}_e^{\times} not contained in e are smaller than x.

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- We define $\tau_C^{\times} := \tau_{C| \geq x}$,

 $F[\tau_C^x]$ results from F by removing all variables greater than x from each clause.

Lemma (lemma 6)

Let $x \neq x_1 \in \mathrm{VAR}(F)$ and let y be the predecessor of x for <. Let $e \in \mathcal{H}$ and $\tau : (e \cap V_{\geq x}) \to \{0,1\}$. Then either $F_e^x[\tau] \equiv 1$ or there exists $U \subseteq \mathcal{H}_e^x$ such that

$$F_e^{\times}[\tau] \equiv \bigwedge_{g \in U} F_g^{y}[\tau_{C_g}^{y}],$$

where C_g is some clause in F_e^{\times} such that $\mathrm{VAR}(C_g) = g$. Moreover, all and-gates are decomposable and U can be computed in polynomial time.

$$F_e^{\mathsf{x}}[\tau] \equiv \bigwedge_{g \in U} F_g^{\mathsf{y}}[\tau_{C_g}^{\mathsf{y}}],$$

Proof sketch.

ullet Let A be the set edges g, where $au \not\models C_g$ for some $\it corresponding C_g$.

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i.e.
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- Choose U as the set of "maximal" edges $g \in A$,
- For each $f \in A$, there is $g \in U$ such that $f \in \mathcal{H}_g^y$. i.e. $g \not\subseteq \mathcal{H}_f^y$ for all $f \in A$, $f \neq g$.
- *U* can be computed in polynomial time.

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- Let A be the set edges g, where $\tau \not\models C_g$ for some corresponding C_g .
- For each clause C such that $VAR(C) \notin A$ we have $\tau \models C$.
- Choose U as the set of "maximal" edges $g \in A$,
- For each $f \in A$, there is $g \in U$ such that $f \in \mathcal{H}_g^y$. i.e. $g \not\subseteq \mathcal{H}_f^y$ for all $f \in A$, $f \neq g$.
- *U* can be computed in polynomial time.
- The edges in *U* are pairwise disjoint. Hence, the and-gate is decomposable.

Corollary (corollary 7)

Let $x \neq x_1 \in VAR(F)$ and let y be the predecessor of x for <. For every $C \in \mathcal{H}$, there exist $U_0, U_1 \subseteq \mathcal{H}^x_{VAR(C)}$ such that

$$F_{\mathrm{VAR}(C)}^{x}[\tau_{C}^{x}] \equiv (x \wedge \bigwedge_{g \in U_{1}} F_{g}^{y}[\tau_{C_{g}}^{y}]) \vee (\neg x \wedge \bigwedge_{g \in U_{2}} F_{g}^{y}[\tau_{C_{g}}^{y}]).$$

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Moreover, all conjunctions are decomposable and U_0 , U_1 can be computed in polynomial time.

Proof sketch. Let $\tau_1 := \tau_C^x \cup \{x \mapsto 1\}$ and $\tau_0 := \tau_C^x \cup \{x \mapsto 0\}$.

$$F_{\mathrm{VAR}(C)}^{\mathsf{x}}[\tau_C^{\mathsf{x}}] = (\mathsf{x} \wedge F_{\mathrm{VAR}(C)}^{\mathsf{x}}[\tau_1]) \vee (\neg \mathsf{x} \wedge F_{\mathrm{VAR}(C)}^{\mathsf{x}}[\tau_0])$$

Apply lemma 6 on each of the terms.

Theorem (theorem 8)

Let F be a β -acyclic CNF-formula. One can construct in polynomial time in $\operatorname{size}(F)$ a dec-DNNF D of size $O((\operatorname{size}(F)))$ and fanin at most $|\mathcal{H}|$ computing F.

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Proof sketch. Let D_i be a dec-DNNF of fanin $|\mathcal{H}|$ at most such that for each $e \in \mathcal{H}$, $C \in F$ such that VAR(C) = e and $j \leq i$, there exists a gate in D_i computing $F_e^{x_j}[\tau_c^{x_j}]$.

• Construct D_i inductively over i.

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• For $e = \max \mathcal{H}$ and a clause C such that $\mathrm{VAR}(C) = e$, we have $\mathcal{H}_e^{v_n} = \mathcal{H}$ and $\tau_{\mathrm{VAR}(C)}^{x_n} = \emptyset$, hence there is a gate in D_n computing $F_e^{x_n}[\tau_C] = F$.

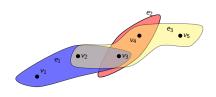
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- We add at most 7 gates per edges per vertex.

Example



$$F=\{\{\overline{v_1},v_2,v_3\},\{\overline{v_3},v_4\},\{v_2,v_3,\overline{v_4},\overline{v_5}\}\}$$

The rest on the blackboard..

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conclusion

Exhaustive DPLL can yield efficient algorithms "theoretically", if we can find a good order to choose the variable (such an ordering must be computable in polynomial time) and a good method of cashing.

Lower-bound on the theoretical method

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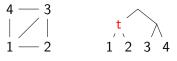




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- A formula φ is structured, if there is a vtree T over the vertices of φ , such that φ respects T.

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Incidence graphs and structure of formulas

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- The incidence graph of a CNF-Formula is the incidence graph of its hyper graph.
- The MIM-width of a CNF-formula is the MIM-width of its incidence graph.

Theorem (theorem 9)

There exists an infinite family \mathcal{F} of β -acyclic CNF-formulas such that for every $F \in \mathcal{F}$ having n variables, there is no structured DNNF of size less than $2^{\Omega(\sqrt{n})}$ computing F.

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Theorem (theorem $1)^1$

There exists an infinite family of β -acyclic hypergraphs of incidence MIM-width $\Omega(n)$ where n is the number of vertices of the hypergraph.

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- Let r be a boolean function over X and let (Y,Z) be a partition of X. We call r a (Y,Z)-rectangle if and only if for every $\tau,\tau'\in\{0,1\}^X$ such that $\tau\models r$ and $\tau'\models r$, we have $\tau|Y\cup\tau'|Z)\models r$.
- A (Y, Z)-rectangle cover of a boolean function f is a set $R = \{r_1, \ldots, r_q\}$ of (Y, Z)-rectangles such that $\operatorname{sat}(f) = \bigcup_{i=1}^q \operatorname{sat}(r_i)$.

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Theorem (theorem $11)^{2,3}$

Let D be a DNNF on variables X respecting the vtree T. For every vertex t of T, there exists a $(X_t, X \setminus X_t)$ -rectangle cover of D of size at most |D|, where $X_t = \operatorname{VAR}(T_t)$.

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Let F be a CNF-formula. Let $\hat{F} := \{K \cup \{c_K\} | K \in F\}$ where we add a fresh variable to each clause.

Theorem (theorem 12)

Let F be a monotone formula of incidence MIM-width k. Any structured DNNF computing \hat{F} is of size at least $2^{k/2}$.

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Proof sketch (lemma 12). Find an assignment τ of \hat{F} such that

$$\hat{F}[\tau] \equiv \bigwedge_{e \in N} (x_e \vee c_e).$$

Conclusion

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• Building a structured d-DNNF is not always the best choice we have.

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- Building a structured d-DNNF is not always the best choice we have.
- If the structure implies a good elimination ordering, exhaustive DPLL might be a better shot.