

Complexity

There are several ways to compare the effectiveness of calculi. One is by implementation and benchmarking. One is by simulation, calculus A can simulate calculus Y , e.g., Resolution can simulate CDCL. Another one is by best case proof length. It is shown that calculus X produces shorter proofs than calculus Y on some class of formulas. This is the topic of this chapter.



2.14.1 Definition (Exponential Proof Length for Analytic Calculi)

Let $EPL_1 = \{P_1, \neg P_1\}$. Then EPL_{i+1} is inductively defined by $EPL_{i+1} = \{P_{2^{i+j}} \vee C_j, \neg P_{2^{i+j}} \vee C_j \mid C_j \in EPL_i\}$, where $EPL_i = \{C_0, \dots, C_{2^i-1}\}$.

Note that the number of clauses as well as the number of propositional variables grows exponentially in i for the EPL_i clause sets: $|EPL_i| = 2^i$ and $|\{P_i \mid P_i \in C_j, C_j \in EPL_i\}| = 2^i - 1$. For example, $EPL_2 = \{P_2 \vee P_1, \neg P_2 \vee P_1, P_3 \vee \neg P_1, \neg P_3 \vee \neg P_1\}$.

2.14.2 Theorem (EPL Proof Length [CookReckhow79])

The minimal proof length of Tableau and DPLL is exponential in $|EPL_i|$, whereas the minimal proof length of Resolution and CDCL is linear in $|EPL_i|$.

2.14.3 Definition (Pigeon Hole Formulas $\text{ph}(n)$)

For some given n and propositional variables $P_{i,j}$, where $1 \leq j \leq n$, $1 \leq i \leq n+1$, the corresponding pigeon hole formula (clause set) $\text{ph}(n)$ is

$$\text{ph}(n) = \bigwedge_{1 \leq i \leq n+1} P_{i,1} \vee \dots \vee P_{i,n} \quad \wedge \quad \bigwedge_{1 \leq j \leq n} \bigwedge_{\substack{1 \leq i, k \leq n+1 \\ i < k}} \neg P_{i,j} \vee \neg P_{k,j}$$

The intuition behind a variable $P_{i,j}$ is that it is true iff pigeon i sits in hole j . Then the formulas $P_{i,1} \vee \dots \vee P_{i,n}$ express that every pigeon has to sit in some hole and the formulas $\neg P_{i,j} \vee \neg P_{k,j}$ that a hole can host at most one pigeon. Since there is one more pigeon than holes, the formula is unsatisfiable.

Note that the number of clauses of a pigeon hole formula $ph(n)$ grows cubic in n . The famous theorem on the pigeon hole formulas says that any resolution proof showing unsatisfiability of $ph(n)$ has a length at least exponential in n , i.e., no resolution-based system can efficiently show unsatisfiability of a pigeon hole formula.

2.14.4 Theorem (Pigeon Hole Proof Length [*Haken85*])

The length of any resolution refutation of $ph(n)$ is exponential in n .





Computing Cost Optimal Models (OCDCL)



OCDCL States

- $(\epsilon; N; \emptyset; 0; \top; \epsilon)$ start state for some clause set N
 - $(M; N; U; k; \perp; O)$ final state, where
 - N has no model if $O = \epsilon$
 - O is a cost optimal model if $O \neq \epsilon$
 - $(M; N; U; k; \top; O)$ intermediate model search state
 - $(M; N; U; k; D; O)$ backtracking state if $D \notin \{\top, \perp\}$
- O denotes the cost optimal model of N
 - M, N, U, k, D are defined analogously to CDCL
 - **but OCDCL always terminates with $D = \perp$**

OCDCL Rules

Propagate $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (ML^{C \vee L}; N; U; k; \top; O)$
 provided $C \vee L \in (N \cup U)$, $M \models \neg C$, L is undefined in M

Decide $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (ML^{k+1}; N; U; k+1; \top; O)$
 provided L is undefined in M , contained in N

ConfISat $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; D; O)$
 provided $D \in (N \cup U)$ and $M \models \neg D$

ConfIOpt $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; \neg M; O)$
 provided $O \neq \epsilon$ and $\text{cost}(M) \geq \text{cost}(O)$

OCDCL Rules (ctd.)

Skip $(ML^{C \vee L}; N; U; k; D; O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; D; O)$
 provided $D \notin \{\top, \perp\}$ and $\text{comp}(L)$ does not occur in D

Resolve $(ML^{C \vee L}; N; U; k; D \vee \text{comp}(L); O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; D \vee C; O)$
 provided D is of level k

Backtrack $(M_1 K^{i+1} M_2; N; U; k; D \vee L; O) \Rightarrow_{\text{OCDCL}} (M_1 L^{D \vee L}; N; U \cup \{D \vee L\}; i; \top; O)$
 provided L is of level k and D is of level i

Improve $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; \top; M)$
 provided $M \models N$, M is total, i.e., contains all atoms in N , and
 $O = \epsilon$ or $\text{cost}(M) < \text{cost}(O)$





?? Definition (Reasonable OCDCL Strategy)

An OCDCL strategy is *reasonable* if ConfISat is preferred over ConfIOpt is preferred over Improve is preferred over Propagate which is preferred over the remaining rules.





2.15.3 Proposition (OCDCL Basic Properties)

Consider an OCDCL state $(M; N; U; k; D'; O)$ derived by a reasonable strategy from start state $(\epsilon, N, \emptyset, 0, \top, \epsilon)$. Then the following properties hold:

1. M is consistent.
2. If $O \neq \epsilon$ then O is consistent and $O \models N$.
3. If $D' \notin \{\top, \perp\}$ then $M \models \neg D'$.
4. If $D' \notin \{\top, \perp\}$ then (i) D' is entailed by $N \cup U$, or (ii) for any model $M' \models \{\neg D'\} \cup N \cup U$: $\text{cost}(M') \geq \text{cost}(O)$.
5. If $D' = \top$ and M contains only propagated literals then for each valuation \mathcal{A} with $\mathcal{A} \models (N \cup U)$ it holds $\mathcal{A} \models M$.



2.15.4 Lemma (OCDCL Normal Forms)

The OCDCL calculus with a reasonable strategy has only 2 normal forms:

- $(M; N; U; 0; \perp; O)$ where $O \neq \epsilon$, $O \models N$ and $\text{cost}(O)$ is optimal
- $(M; N; U; 0; \perp; \epsilon)$ where N is unsatisfiable



Improving OCDCL

Prune $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; \neg M; O)$

provided for all total trail extensions MM' of M it holds
 $\text{cost}(MM') \geq \text{cost}(O)$

ConflOpt $(M; N; U; k; \top; O) \Rightarrow_{\text{OCDCL}} (M; N; U; k; \neg M; O)$

provided $O \neq \epsilon$ and $\text{cost}(M) \geq \text{cost}(O)$

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The Max-SAT Problem

Given $N = N_H \uplus N_S$ where N_H are hard clauses
and N_S are soft clauses

Find $\mathcal{A} \models N_H$ with minimal cost $\sum_{\mathcal{A} \models \neg C}^{\mathcal{C} \in N_S} \omega(C)$
where $\omega: N_S \mapsto \mathbb{R}^+$





Reducing Max-SAT to OCDCL

1. Introduce a fresh variable S_i for each $C_i \in N_S = \{C_1, \dots, C_n\}$
2. Define $N'_S = \{S_i \vee C_i \mid C_i \in N_S\}$
3. Compute cost optimal model for $N' = N_H \uplus N'_S$ with

$$\text{cost function } \text{cost}(L) = \begin{cases} \omega(C_i) & \text{if } L = S_i \\ 0 & \text{otherwise} \end{cases}$$



2.15.7 Theorem (Max-SAT Solution)

\mathcal{A} is a Max-SAT solution for $N = N_H \uplus N_S$ with minimal value $c = \sum_{\mathcal{A} \models \neg C} \omega(C)$ iff $(\epsilon; N'; \emptyset; 0; \top; \epsilon) \Rightarrow_{\text{OCDCL}}^* (M; N'; U; k; \perp; O)$ with a reasonable strategy where $N' = N_H \uplus N'_S$, and $\text{cost}(O) = c$.

Optimization

1. Introduce a fresh variable S_i for each $C_i \in N_S = \{C_1, \dots, C_n\}$
2. Define $N'_S = \{S_i \vee C_i \mid C_i \in N_S\} \cup \{\neg C_i \vee \neg S_i \mid C_i \in N_S\}$
3. Compute cost optimal model for $N' = N_H \uplus N'_S$ with

$$\text{cost function } \text{cost}(L) = \begin{cases} \omega(C_i) & \text{if } L = S_i \\ 0 & \text{otherwise} \end{cases}$$

Minimal Covering Models

Given \mathcal{M} set of all models of the set of clauses N

Find $\mathcal{M}' \subseteq \mathcal{M}$ such that

- $|\mathcal{M}'|$ is minimal
- for each propositional variable P in N there is a model $M \in \mathcal{M}'$ with $M(P) = 1$



Reduction to OCDCL

Given N with variables P_1, \dots, P_n and clauses C_1, \dots, C_m

1. Define $N_j := \{C\{P_i \mapsto P_i^j \mid 1 \leq i \leq n\} \vee \neg Q_j \mid C \in N\}$
2. Define $N_+ := \{P_i^1 \vee \dots \vee P_i^n \mid 1 \leq i \leq n\}$
3. Define $N_Q := \{\neg P_i^j \vee Q_j \mid 1 \leq i, j \leq n\}$
4. Find a minimal cost model of $(\cup_{j=1}^n N_j) \cup N_+ \cup N_Q$ with cost function $\text{cost}(M) = \sum_{j=1}^n M(Q_j)$

Requires

- $O(n^2)$ additional variables
- $O(n \cdot \max(m, n))$ additional clauses

Note: $n =$ upper bound of number of models (Algorithm 10)





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