Marc Vinyals

DPLL

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```
Algorithm 1: DPLL
while not solved do
if conflict then backtrack()
else if unit then propagate()
else branch()
```

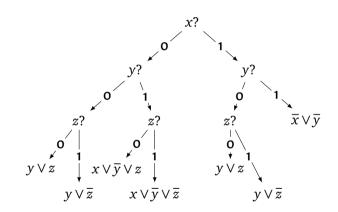
State: partial assignment

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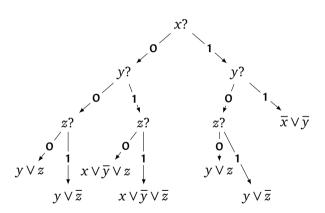
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Resolution

Interpret DPLL run as resolution proof

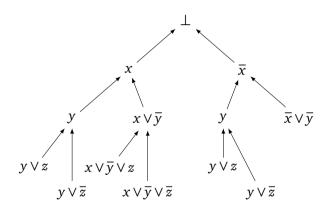
$$\frac{C \vee v \qquad D \vee \overline{v}}{C \vee D}$$



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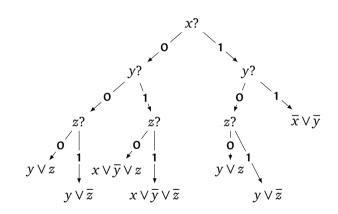


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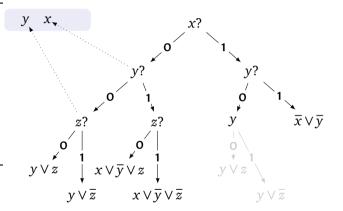


CDCL

Algorithm 2: CDCL
while not solved do
if conflict then learn()
else if unit then propagate()
else
maybe forget()
maybe restart()
branch()

State: partial assignment & learned clauses

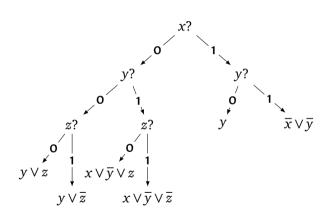
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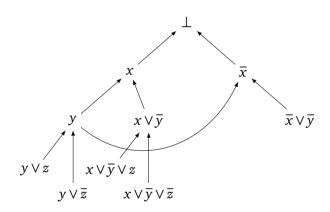
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- CDCL implicit proofs are in resolution form
- DPLL proofs only in weaker "tree-like" resolution form
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Partial results in 2000s

```
[Beame, Kautz, Sabharwal'04]
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[Beame, Kautz, Sabharwal '04] [Van Gelder '05] [Hertel, Bacchus, Pitassi, Van Gelder '08] [Buss, Hoffmann, Johannsen '08]

Yes (under natural model)

[Pipatsrisawat, Darwiche '09] [Atserias, Fichte, Thurley '09]

CDCL equivalent to Resolution: Results

Theorem [Pipatsrisawat, Darwiche '09] With non-deterministic variable decisions, CDCL can efficiently find resolution proofs

Theorem [Atserias, Fichte, Thurley'09]
With random variable decisions,
CDCL can efficiently find bounded-width resolution proofs

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CDCL can efficiently find reproduce resolution proofs

Theorem [Pipatsrisawat, Darwiche '09] With non-deterministic variable decisions,

Theorem [Atserias, Fichte, Thurley'09]

With **random** variable decisions, CDCL can efficiently find **bounded-width** resolution proofs

- ▶ Derivation $\pi = C_1, ..., C_t$.
- ▶ Goal: learn every clause $C_i \in \pi$.

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 $x \lor y$ is absorbed:

- if x = 0 then propagate z = 1 and y = 1;
- ightharpoonup if y=0 then propagate z=1 and x=1.

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Algorithm 3: Simulation

for C_i \in \pi do

while C_i not absorbed do

if conflict then

learn()

restart()

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- Clauses not thrown away
- Frequent restarts
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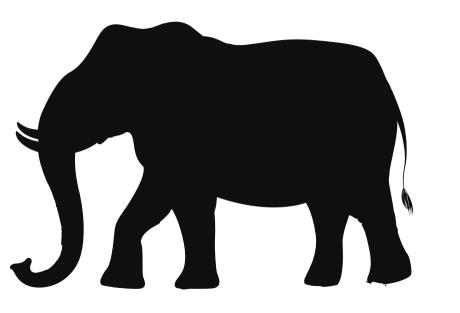
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Optimal variable choices are needed

No deterministic algorithm simulates resolution unless FPT hierarchy collapses.

[Alekhnovich, Razborov'01]

► No deterministic algorithm simulates resolution unless P = NP.

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CDCL with any static order exponentially worse than resolution.

[Mull, Pang, Razborov'19]

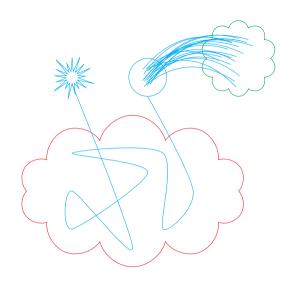
CDCL with VSIDS and similar heuristics exponentially worse than resolution.

[V'20]

- Give a score q = q(x) to variable x.
- At each conflict
 - ▶ Bump q' = q + 1 if x involved.
 - ▶ Decay $q' = 0.95 \cdot q$ all variables.
- Pick variable with largest score

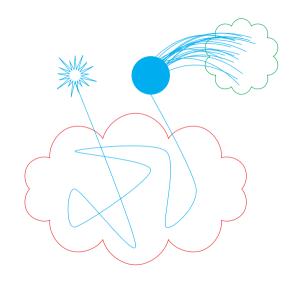
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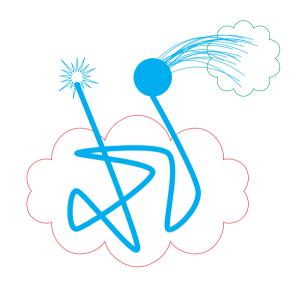
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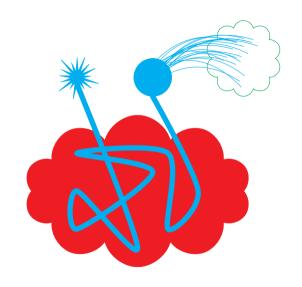
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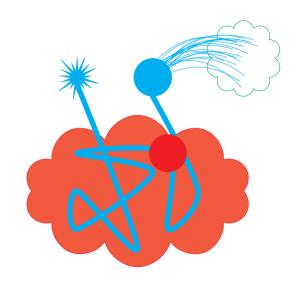
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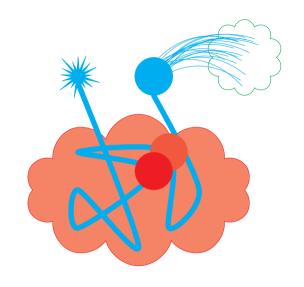
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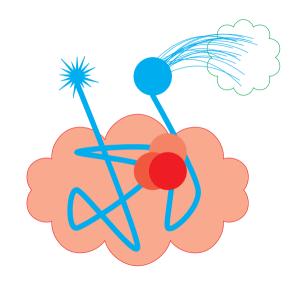
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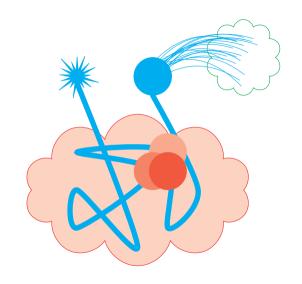
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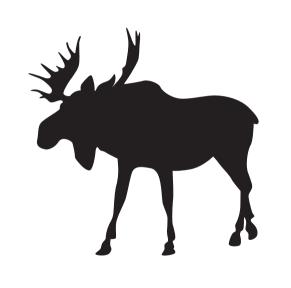


Hard Formulas for VSIDS

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CDCL equivalent to Resolution: Assumptions

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Throwing Clauses Away

No great theoretical framework

▶ With nondeterministic erasures enough to keep only $n \ll L$ clauses in memory.

[Esteban, Torán '01]

- But more are needed to simulate resolution:
- ► Keeping $\ll n$ clauses can exponentially blow-up runtime.

[Ben Sasson, Nordström '11]

► Keeping $\ll n^k$ clauses can superpolynomially blow-up runtime.

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[Thapen '16]

What about clauses with low LBD?



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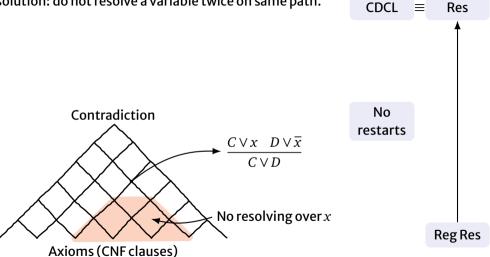
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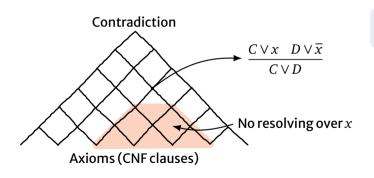
[Hertel, Bacchus, Pitassi, Van Gelder '08]

CDCL without restarts between regular and standard resolution.

Regular resolution: do not resolve a variable twice on same path.



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- Regular resolution exponentially weaker than general.
 (Exist formulas with short proofs but exponentially long regular proofs)

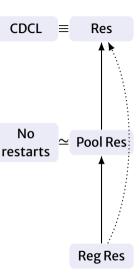




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Pool res ≥ Regular res ⇒ Formulas that separate general and regular are good candidates to separate general and pool.

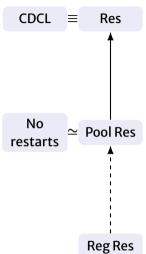


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Formula with CDCL proof of length L but requires L + 1 w/o restarts?

CDCL Res No → Pool Res restarts **Reg Res**



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- C asserting if unit after backtracking.
- 1UIP is asserting.

- Less overhead with decision learning scheme.
- Is decision faster than 1UIP?
- What overhead is needed?

Merge Resolution

► A resolution step is a merge if *C* and *D* share a literal.

$$\frac{x \lor y \lor z \quad x \lor y \lor \overline{z}}{x \lor y} \qquad \frac{x \lor z \quad y \lor \overline{z}}{x \lor y}$$
Not a merge
$$\frac{x \lor z \quad y \lor \overline{z}}{x \lor y}$$

Merge resolution: at least one premise either axiom or merge.

Marc Vinvals

[Andrews '68]

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Merge resolution: at least one premise either axiom or merge.

[Andrews '68]

- Merge resolution 2.0: only reuse merges.
- 1UIP produces merge resolution proofs.
- ightharpoonup Merge resolution can simulate standard resolution with O(n) overhead.
- And $\Omega(n)$ overhead sometimes needed.

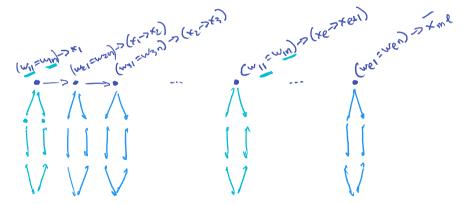
[Fleming, Ganesh, Kolokolova, Li, V]

Tricky Formulas for Merge Resolution

$$\mathcal{W}: \qquad \qquad w_j^k = w_j^{k+1} \qquad \qquad \text{for } j \in [\ell], \text{for } k \in [n-1]$$

$$\mathcal{X}: \qquad \qquad (w_{\hat{\imath},1} = w_{\hat{\imath},n}) \to (x_{i-1} \to x_i) \qquad \qquad \text{for } i \in [m\ell]$$

where $\hat{\imath} = i \pmod{\ell}$, $x_0 := 1$, $x_{n\ell} := 0$, $m \simeq n$, $\ell \simeq \log n$.



Beyond Resolution

- How much should we focus on resolution anyway?
- Preprocessing ⇒ introduce new variables ⇒ extended resolution.

[Kullmann'99]

ER as powerful as extended Frege ⇒ no hope to analyse with current tools.

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► ER as powerful as extended Frege ⇒ no hope to analyse with current tools.

- Modern solvers use inprocessing, this is now a pressing issue.
- Can still study DRAT without new variables as a proof system (DRAT⁻).
- ► Many hard principles for resolution easy in DRAT.

[Buss, Thapen'19]

Take Home

- ► CDCL equivalent to Resolution
- ► But only under assumptions, not all reasonable

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Open Problems

- How to model space?
- Are restarts important?
- ► How much overhead do we need?

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Images: Vecteezy.cor

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Thanks!