

## Automated reasoning for first-order logic Theory, Practice and Challenges

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Modular instantiation-based reasoning

The problem: Show that a given formula is a theorem.

Ground (SAT/SMT)

 $P(a) \lor Q(c,d)$  $\neg P(a) \lor Q(d,c)$ 

Very efficient Not very expressive DPLL Industry

#### First-Order

 $\forall x \exists y \ Q(x, y) \lor \neg Q(y, f(x))$  $P(a) \lor Q(d, c)$ 

Very expressive Ground: not as efficient Resolution/Superposition Academia  $\rightarrow$  Industry

From Ground to First-Order: Efficient at gound + Expressive?

## Traditional Methods: Resolution

#### **Reasoning Problem**

Given a set of first order clauses S, prove S is unsatisfiable.





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#### Reasoning Problem

Given a set of first order clauses S, prove S is unsatisfiable.

Resolution :

 $\frac{C \lor L \quad \overline{L'} \lor D}{(C \lor D)\sigma}$ 

Example :  $Q(x) \lor P(x) \quad \neg P(a) \lor R(y)$ 

$$\frac{Q(a) \vee R(y)}{Q(a) \vee R(y)}$$

Weaknesses:



- Inefficient in propositional case
- Length of clauses can grow fast
- Recombination of clauses
- No effective model representation

## Basic idea behind instantiation proving

Can we approximate first-order by ground reasoning?

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Theorem (Herbrand). For a quantifier free formula  $\varphi(\bar{x})$ ;  $\forall \bar{x}\varphi(\bar{x})$  is unsatisfiable iff  $\bigwedge_i \varphi(\bar{t}_i)$  is unsatisfiable, for some ground terms  $\bar{t}_1, \ldots, \bar{t}_n$ .

Basic idea: Interleave instantiation with propositional reasoning.

Main issues:

- How to restrict instantiations.
- ▶ How to interleave instantiation with propositional reasoning.

## Different approaches

Gilmore (1960): generation of ground instances Robinson (1965): resolution Plaisted et al (1992): hyper-linking Plaisted & Zhu (2000): semantics-based instance generation Letz & Stenz (2000): disconnection tableaux-type calculus Hooker et al (2002): generation of instances with sem. selection Baumgartner & Tinelli (2003): ME: Lifting of DPLL Ganzinger & Korovin (2003): Inst-Gen calculus, modular ground reasoning

Claessen (2005): Equinox

... many instantiation based methods for different fragments/logics

First-Order Clauses











 $p(f(x), b) \lor q(x, y)$  $\neg p(f(f(x)), y)$  $\neg q(f(x), x)$ 

 $p(f(x), b) \lor q(x, y)$  $\neg p(f(f(x)), y)$  $\neg q(f(x), x)$ 

 $p(f(\perp), b) \lor q(\perp, \perp)$   $\neg p(f(f(\perp)), \perp)$  $\neg q(f(\perp), \perp)$   $p(f(x), b) \lor q(x, y)$  $\neg p(f(f(x)), y)$  $\neg q(f(x), x)$ 

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 $\neg p(f(f(x)), b)$ 

 $p(f(x), b) \lor q(x, y)$ 

 $\neg p(f(f(x)), y)$ 

 $\neg q(f(x), x)$ 

 $p(f(\bot), b) \lor q(\bot, \bot)$  $\neg p(f(f(\bot)), \bot)$  $\neg q(f(\bot), \bot)$ 

 $p(f(f(\perp)), b) \lor q(f(\perp), \perp)$ 

 $\neg p(f(f(\perp)), b)$ 

 $p(f(\perp), b) \lor q(\perp, \perp)$ 

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 $p(f(x), b) \lor q(x, y)$  $\neg p(f(f(x)), y)$  $\neg q(f(x), x)$ 

 $p(f(f(x)), b) \vee q(f(x), y)$ 

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 $p(f(x), b) \vee q(x, y)$ 

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 $p(f(\perp), b) \vee q(\perp, \perp)$  $\neg p(f(f(\perp)), \perp)$  $\neg q(f(\perp), \perp)$  $p(f(f(\perp)), b) \lor q(f(\perp), \perp)$  $\neg p(f(f(\perp)), b)$  $p(f(\perp), b) \vee q(\perp, \perp)$  $\neg p(f(f(\perp)), \perp)$  $\neg q(f(\perp), \perp)$ 

 $p(f(x), b) \lor q(x, y)$  $\neg p(f(f(x)), y)$  $\neg q(f(x), x)$ 

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 $p(f(x), b) \vee q(x, y)$ 

 $\neg p(f(f(x)), y)$ 

 $\neg q(f(x), x)$ 

The final set is propositionally unsatisfiable.

Resolution :

$$\frac{(C \lor L) \quad (\overline{L'} \lor D)}{(C \lor D)\sigma}$$
$$\sigma = \mathrm{mgu}(L, L')$$

#### Weaknesses of resolution:

Inefficient in the ground/EPR case Length of clauses can grow fast Recombination of clauses No explicit model representation Instantiation :

$$\frac{(C \lor L) \qquad (\overline{L'} \lor D)}{(C \lor L)\sigma \qquad (\overline{L'} \lor D)\sigma}$$
$$\sigma = \mathrm{mgu}(L, L')$$

Strengths of instantiation: Modular ground reasoning Length of clauses is fixed Decision procedure for EPR No recombination Semantic selection Redundancy elimination Effective model presentation

## Redundancy Elimination

The key to efficiency is redundancy elimination.

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Ground clause C is redundant if

- $C_1,\ldots,C_n\models C$
- $\blacktriangleright C_1,\ldots,C_n\prec C$

Where  $\prec$  is a well-founded ordering.

- $\blacktriangleright P(a) \models Q(b) \lor P(a)$
- ►  $P(a) \prec Q(b) \lor P(a)$

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Where  $\prec$  is a well-founded ordering.

►  $P(a) \prec Q(b) \lor P(a)$ 

Theorem [Ganzinger, Korovin]. Redundant clauses/closures can be eliminated.

Consequences:

- many usual redundancy elimination techniques
- redundancy for inferences
- new instantiation-specific redundancies

Simplifications by SAT/SMT solver [Korovin IJCAR'08]

Can off-the-shelf ground solver be used to simplify ground clauses?

Abstract redundancy:

 $C_1,\ldots,C_n\models C$  $C_1,\ldots,C_n\prec C$ 

 $S_{gr} \models C$  — ground solver follows from smaller ?

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Basic idea:

- split  $D \subset C$
- check  $S_{gr} \models D$
- add D to S and remove C

 $S_{gr} \models C$  — ground solver follows from smaller ?

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- split  $D \subset C$
- check  $S_{gr} \models D$
- ▶ add *D* to *S* and remove *C*

 $S_{gr} \models C$  — ground solver follows from smaller ? Global ground subsumption:

 $\frac{D \vee C}{D}$ 

where  $S_{gr} \models D$  and  $C' \neq \emptyset$ 

### Global Ground Subsumption [Korovin IJCAR'08]

S<sub>gr</sub>

 $egreen Q(a, b) \lor P(a) \lor P(b)$   $P(a) \lor Q(a, b)$ egreen P(b) С

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A minimal  $D \subset C$  such that  $S_{gr} \models D$  can be found in a linear number of implication checks. Sgr

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Global Ground Subsumption generalises:

- strict subsumption
- subsumption resolution

▶ ...

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The main idea:

 $S_{gr} \models \forall \bar{x} C(\bar{x})$  $C_1(\bar{x}), \dots, C_n(\bar{x}) \in S$   $S_{gr} \models C(\bar{d})$  for fresh  $\bar{d}$  $C_1(\bar{d}), \dots, C_n(\bar{d}) \models C(\bar{d})$  Off-the-shelf ground solver can be used to simplify ground clauses.

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 $S_{gr} \models \forall \bar{x} C(\bar{x})$   $C_1(\bar{x}), \dots, C_n(\bar{x}) \in S$  $C_1(\bar{x}), \dots, C_n(\bar{x}) \prec C(\bar{x})$ 

Non-Ground Global Subsumption

 $S_{gr} \models C(\overline{d})$  for fresh  $\overline{d}$  $C_1(\overline{d}), \dots, C_n(\overline{d}) \models C(\overline{d})$  as in Global Subsumption

S

 $egreen P(x) \lor Q(x)$   $egreen Q(x) \lor S(x, y)$  $P(x) \lor S(x, y)$  С











# Finer-grained control: closure orderings

Finer-grained control: replace ground clauses with ground closures. Closure, a closure is a pair  $C \cdot \sigma$ , where C is a clause and  $\sigma$  a grounding substitution

 $(A(a) \lor B(x)) \cdot [b/x]$ 

Represents: ground clause  $C\sigma$ 

 $A(a) \vee B(b)$ 

Closure ordering: any total, well-founded ordering such that  $C\theta\cdot \tau\prec C\cdot \sigma$  if

- $C\sigma = C\theta\tau$ , and
- $\theta$  properly instantiates C

Slogan: more specific representations take priority over less specific ones Ex:  $(p(a) \lor q(z)) \cdot [b/z] \prec (p(y) \lor q(z)) \cdot [a/y, b/z]$ 

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# Closure-based redundancy elimination

Definition call  $C \cdot \sigma$  redundant in S if

- $C_1 \cdot \sigma_1, \ldots, C_n \cdot \sigma_n \models C \cdot \sigma$  and
- $\blacktriangleright C_1 \cdot \sigma_1, \ldots, C_n \cdot \sigma_n \prec C \cdot \sigma$

Theorem. [Ganzinger, Korovin]

Redundant closures (and clauses) can be eliminated.

Consequences:

- generalises usual redundancy
- new instantiation specific redundancies
  - blocking non-proper instances (merging variables) can be eliminated
  - dismatching constraints
- redundancy for inferences

Example:

$$p(x) \lor \underline{\neg q(f(x))}$$
(1)  
$$\underline{p(f(x))} \lor \neg q(f(f(x)))$$
(2)  
$$\underline{q(f(f(a)))}$$
(3)

Then the inference between (1) and (2) is redundant!

Why? the conclusion is represented twice  $p(f(a)) \lor \neg q(f(f(a)))$  $p(f(x)) \lor \neg q(f(f(x))) \cdot [a/x] \prec p(x) \lor \neg q(f(x)) \cdot [f(a)/x]$ 

This can be represented as a dismatching constraint.

 $p(x) \lor \underline{\neg q(f(x))} \mid x \triangleleft_{ds} f(x)$ 

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How to make closures redundant? Instantiate!

Every proper instantiation inference makes closures redundant in the premise.

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#### Example

 $A(f(y)) \lor D_1 \qquad \neg A(x) \lor C$  $A(f^3(y)) \lor D_2$  $A(f^5(y)) \lor D_3$  $\dots$  $A(f^{i_n}(y)) \lor D_n$ 

All other inferences with  $\neg A(x) \lor C$  are blocked!

Premises inherit the constraints during instantiation inferences.

#### Example

 $\begin{array}{ll} A(f(y)) \lor D_1 & \neg A(x) \lor C \mid x \triangleleft_{ds} f(y) \\ A(f^3(y)) \lor D_2 & \neg A(f(y)) \lor C \\ A(f^5(y)) \lor D_3 & \\ \cdots & \\ A(f^{i_n}(y)) \lor D_n \end{array}$ 

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Premises inherit the constraints during instantiation inferences.

- Inst-Gen is sound and complete for first-order logic
- combines efficient ground reasoning with first-order reasoning
- decision procedure for effectively propositional logic (EPR)
- redundancy elimination
  - usual: tautology elimination, strict subsumption
  - global subsumption:
    - non-ground simplifications using SAT/SMT reasoning
  - closure-based redundancies:

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Equational instantiation-based reasoning

Superposition calculus:

 $\frac{C \lor s \simeq t \quad L[s'] \lor D}{(C \lor D \lor L[t])\theta}$ 

where (i)  $\theta = mgu(s, s')$ , (ii) s' is not a variable, (iii)  $s\theta\sigma \succ t\theta\sigma$ , (iv) ... The same weaknesses as resolution has:

- ▶ Inefficient in the ground/EPR case
- Length of clauses can grow fast
- Recombination of clauses
- No explicit model representation





#### Incomplete !

 $\begin{array}{rcl} f(h(x)) &\simeq & c \\ h(x) &\simeq & x \\ f(a) & \not\simeq & c \end{array}$ 

This set is inconsistent but the contradiction is not deducible by the inference system above.

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The idea is to consider proofs generated by unit superposition:

$$\frac{h(x) \simeq x \quad f(h(y)) \simeq c}{\frac{f(x) \simeq c}{\frac{c \neq c}{\Box}}}$$
$\begin{array}{rcl} f(h(x)) &\simeq & c \\ h(x) &\simeq & x \\ f(a) & \not\simeq & c \end{array}$ 

This set is inconsistent but the contradiction is not deducible by the inference system above.

The idea is to consider proofs generated by unit superposition:

$$\frac{\frac{h(x) \simeq x - f(h(y)) \simeq c}{f(x) \simeq c} [x/y]}{\frac{c \neq c}{\Box}} [a/x]$$

 $\begin{array}{rcl} f(h(x)) &\simeq & c \\ h(x) &\simeq & x \\ f(a) & \not\simeq & c \end{array}$ 

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The idea is to consider proofs generated by unit superposition:

$$\frac{h(x) \simeq x \quad f(h(y)) \simeq c}{\frac{f(x) \simeq c}{\Box}} \begin{bmatrix} x/y \end{bmatrix} \quad f(a) \neq c}{\begin{bmatrix} a/x \end{bmatrix}}$$

Propagating substitutions:  $\{h(a) \simeq a; f(h(a)) \simeq c; f(a) \neq c\}$ ground unsatisfiable. Superposition+Instantiation

 $\begin{array}{rcl} f(h(x)) &\simeq & c & \lor & C_1(x,y) \\ h(x) &\simeq & x & \lor & C_2(x,y) \\ f(a) & \not\simeq & c & \lor & C_3(x,y) \end{array}$ 

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The idea is to consider proofs generated by unit superposition:

$$\frac{h(x) \simeq x \quad f(h(y)) \simeq c}{\frac{f(x) \simeq c}{\Box}} \begin{bmatrix} x/y \end{bmatrix} \quad f(a) \neq c} \begin{bmatrix} a/x \end{bmatrix}$$

Propagating substitutions:  $\{h(a) \simeq a; f(h(a)) \simeq c; f(a) \neq c\}$ ground unsatisfiable.

## Superposition+Instantiation

f(h(x))	$\simeq$	С	V	$C_1(x,y)$	f(h(a))	$\simeq$	С	$\vee$	$C_1(a, y)$
h(x)	$\simeq$	x	$\vee$	$C_2(x,y)$	h(a)	$\simeq$	а	V	$C_2(a, y)$
f(a)	≄	С	$\vee$	$C_3(x,y)$	f(a)	≄	с	V	$C_3(a, y)$

This set is inconsistent but the contradiction is not deducible by the inference system above.

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f.-o. clauses *S* 

Theorem.[Ganzinger, Korovin CSL'04] Inst-Gen-Eq is sound and complete. 77/144



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- sound and complete for first-order logic with equality
- combines SMT for ground reasoning and superposition-based unit reasoning
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### Labelled Unit Superposition [Korovin, Sticksel LPAR'10]

General idea: Dismatching constraints can be used to block already derived proofs!

Unit superposition with dismatching constraints:

 $\frac{(l\simeq r)\mid [\ D_1\ ]\quad L[l']\mid [\ D_2\ ]}{L[r]\theta\mid [\ (D_1\wedge D_2)\theta\ ]} \ (\theta)$ 



where (i)  $\theta = \text{mgu}(I, I')$ ; (ii) I' is not a variable; (iii) for some grounding substitution  $\sigma$  satisfying  $(D_1 \land D_2)\theta$   $|\sigma \succ r\sigma$  (iv)  $\mu = \text{mgu}(s, t)$ ; (v)  $D\mu$  is satisfiable

Next technical issue: The same unit literal can

- correspond to different clauses,
- have different dismatching constraints
- be represented many times in the same proof search

#### Solution: labelled approach

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## Tree Labelled Unit Superposition

- Preserve Boolean structure of proofs
- Closure is a propositional variable in an AND/OR tree
- Conjunction  $\land$  in superposition, disjunction  $\lor$  in merging



# **OBDD** Labelled Unit Superposition



#### Disadvantages of trees

- Not produced in normal form
- Sequence of inferences determines shape
- Potential growth ad infinitum
- OBDD as normal form
- Maintenance effort
- Reordering required

### Labels: Sets vs. Trees vs. OBDDs

iProver-Eq – CVC3 as a background solver on pure equational problems. (developed with Christoph Sticksel)



f.-o. clauses  $\boldsymbol{S}$ 

theory T

#### Theory instantiation [Ganzinger, Korovin LPAR'06]



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#### Conditions on completeness:

- complete ground reasoning modulo T
- answer completeness of unit reasoning modulo T
- T is universal

Answer completeness: If  $L_1 \tau \land \ldots \land L_n \tau \models_{\mathcal{T}} \Box$  for ground  $\tau$ . Then

$$\frac{L_1,\ldots,L_n}{L_1\theta,\ldots,L_n\theta} \ UC$$

such that  $\theta$  is a genralization of  $\tau$  and  $L_1\theta\bot,\ldots,L_n\theta\bot\vdash_{\mathcal{T}}\Box$ 

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#### Evaluation





## CASC 2013 results

#### General first-order (FOF) 300 problems

	Vampire	Е	iProver	E-KRHyper	Prover9
prob	281	249	167	122	119
time	12	29	12	8	12

#### Effectively propositional 100 problems

#### First-order satisfiability (FNT) 150 problems

Non-cyclic sorts for first-order satisfiability [Korovin FroCoS'13]
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#### First-order satisfiability (FNT) 150 problems

	iProver	Paradox	CVC4	Е	Nitrox	Vampire	
prob	122	99	96	79	79	78	
time	52	2	25	20	29	30	

Non-cyclic sorts for first-order satisfiability [Korovin FroCoS'13]

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EPR: No functions except constants:  $P(x, y) \lor \neg Q(c, y)$ Transitivity:  $\neg P(x, y) \lor \neg P(y, z) \lor P(x, z)$ Symmetry:  $P(x, y) \lor \neg P(y, x)$ Verification:

 $\forall A(\texttt{wren}_{h1} \land A = \texttt{wraddrFunc} \rightarrow \\ \forall B(\texttt{range}_{[35,0]}(B) \rightarrow (\texttt{imem}'(A, B) \leftrightarrow \texttt{iwrite}(B)))).$ 

### Applications:

- Hardware Verification (Intel)
- Planning/Scheduling
- Finite model reasoning

EPR is hard for resolution, but decidable by instantiation methods.

Direct reduction to SAT — exponential blow-up. Satisfiability for EPR is NEXPTIME-complete. More succinct but harder to solve.... Any gain?

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General lemmas:

 $egar{alpha} \neg a(x) \lor b(x) \quad \neg b(x) \lor \texttt{mem}(x, y)$  $a(x) \lor \texttt{mem}(x, y)$ 

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More expressive logics can speed up calculations!

# Hardware verification



### Functional Equivalence Checking

- > The same functional behaviour can be implemented in different ways
- Optimised for:
  - Timing better performance
  - Power longer battery life
  - Area smaller chips

Verification: optimisations do not change functional behaviour
Method of choice: Bounded Model Checking (BMC) used at Intel, IBM

EPR encoding:

- $s_0, \ldots, s_k$  constants denote unrolling bounds
- first-order formulas I(S), P(S), T(S, S')
- next state predicate Next(S, S')

BMC can be encoded  $I(s_0); \neg P(s_k);$  initial and final states  $\forall S, S'(Next(S, S') \rightarrow T(S, S'));$  transition relation  $Next(s_0, s_1); Next(s_1, s_2); \dots Next(s_{k-1}, s_k);$  next state relation

- EPR encoding provides succinct representation
- avoids copying transition relation
- reasoning can be done at higher level

BMC with bit-vectors, memories:

[M. Emmer, Z. Khasidashvili, K. Korovin, C. Sticksel, A. Voronkov IJCAR'12

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# Experiments: iProver vs Intel BMC

Problem	# Memories	# Transient BVs	Intel BMC	iProver BMC
ROB2	2 (4704 bits)	255 (3479 bits)	50	8
DCC2	4 (8960 bits)	426 (1844 bits)	8	11
DCC1	4 (8960 bits)	1827 (5294 bits)	7	8
DCI1	32 (9216 bits)	3625 (6496 bits)	6	4
BPB2	4 (10240 bits)	550 (4955 bits)	50	11
SCD2	2 (16384 bits)	80 (756 bits)	4	14
SCD1	2 (16384 bits)	556 (1923 bits)	4	12
PMS1	8 (46080 bits)	1486 (6109 bits)	2	10

Large memories:

iProver outperforms highly optimised Intel SAT-based model checker.

Implementation

- Inst-Gen also uses SAT solver and resolution for simplifications
- Query answering: using answer substitutions
- ► Finite model finding: based on EPR/sort inference/non-cyclic sorts
- Bounded model checking mode: (Intel format)
- Proof representation: non-trivial due to SAT solver simplifications
- Model representation: using formulas in term algebra; special model representation for hardware BMC

# iProver implementation features

iProver is implemented in OCaml, around 50,000 LOC Core:

- Inst-Gen Given clause algorithm
- ► SAT solvers for ground reasoning: MiniSAT, PicoSAT, Lingeling
- strategy scheduling
- preprocessing
- splitting with naming

Simplifications:

- Literal selection
- Subsumption (forward/backward)
- Subsumption resolution (forward/backward)
- Dismatching constraints
- Blocking non-proper instantiators
- Global subsumption: SAT solver is used for non-ground simplifications

# Inst-Gen given clause algorithm

Passive: clauses that are waiting to participate in inferences

- priority queues based on lexicographic combinations of parameters
  - -- inst\_pass\_queue1 [-conj\_dist; +conj\_symb; -num\_var]
  - -- inst\_pass\_queue2 [+age; -num\_symb]

Active: clauses between which all inferences are done

- unification index on selected literals
  - Non-perfect discrimination trees

Given clause: C

- 1. C next clause from the top of Passive
- 2. simplify C: compressed feature indexes
- 3. perform all inferences between C and Active
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# Inst-Gen Loop



[Korovin (Essays in Memory of Harald Ganzinger 2013])

### Why indexing:

- Single subsumption is NP-hard.
- ▶ We can have 100,000 clauses in our search space
- Applying naively between all pairs of clauses we need 10,000,000,000 subsumption checks !

### Indexes in iProver:

- non-perfect discrimination trees for unification, matching
- compressed feature vector indexes for subsumption, subsumption resolution, dismatching constraints.

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### Discrimination trees



Efficient filtering unification, matching and generalisation candidates

- clause C can not subsume any clause with number of literals strictly less than C
- clause C can not subsume any clause with number of positive literals strictly less than C
- clause C can not subsume any clause with the number of occurrences of a symbol f less than in C



Design efficient filters based on "features of clauses":

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Fix: a list of features:

- 1. number of literals
- 2. number of occurrences of *f*
- 3. number of occurrences of g

With each clause associate a feature vector: numeric vector of feature values Example: feature vector of  $C = p(f(f(x))) \lor \neg p(g(y))$  is fv(C) = [2, 2, 1]

### Arrange feature vectors in a trie data structure.

For retrieving all candidates which can be subsumed by C we need to traverse only vectors which are component-wise greater or equal to  $f_V(C)$ .

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**Example**: is signature contains 1000 symbols and we use all symbols as features then feature vector for every clause will be 1000 in length.

Basic idea: for each clause most features will be 0.

Compress feature vector: use list of pairs  $[(p_1, v_1), \ldots, (p_n, v_1)]$  where  $p_i$  are non-zero positions and  $v_i$  are values that start from this position. Sequential positions with the same value are combined.

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iProver has solid performance over the whole range of TPTP.

iProver excels on EPR problems and in turn on satisfiability, bounded model checking and other encodings into EPR.

### PhD opportunities at the University of Manchester

### PhD opportunities in reasoning, logic and verification, please contact: korovin@cs.man.ac.uk