

On Automating the Calculus of Relations

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Relations

- one of the most ubiquitous concepts in mathematics and computing
- origins in the late 19th century
- 1941: The calculus of (binary) relations (A.Tarski)
- first-order, equational axioms

Relation Calculi

Applications

- program semantics (Dijkstra, Hoare, . . .)
- refinement calculus (Back, Scott, . . .)
- verification

Relation-based Formal Methods

- Alloy (Jackson)
- B (Abrial)
- Z (Spivey)
- algebraic approach to functional programming (Bird, de Moor)

Further Applications

- data bases, graphs, preference modelling, modal reasoning, linguistics, hardware verification, design of algorithms, . . .

Relations

- a **binary relation** R on a set A is a subset of $A \times A$ (a set of ordered pairs)
- operations
 - union $R \cup S$
 - intersection $R \cap S$
 - complement \overline{R}
 - relative product $R; S$
 $(a, b) \in R; S \Leftrightarrow \exists c. (a, c) \in R \text{ and } (c, b) \in S$
 - converse \check{R} $(a, b) \in \check{R} \Leftrightarrow (b, a) \in R$
- $(2^{A^2}, \cup, \cap, \overline{}, \check{}, ;, 1_A)$ is called **proper relation algebra** of all binary relations
- expressiveness of the calculus of binary relations is that of the three-variable fragment of first-order logic

Relation algebra

Definition

A **relation algebra** is a structure $(A, +, ;, \bar{}, \check{}, 1)$ satisfying the axioms

$$\begin{aligned} (x + y) + z &= x + (y + z) , & x + y &= y + x , & x &= \overline{\overline{x} + \overline{y}} + \overline{\overline{x} + \overline{y}} , \\ (x; y); z &= x; (y; z) , & (x + y); z &= x; z + y; z , & x; 1 &= x , \\ \check{\check{x}} &= x , & (x + y)^\check{} &= \check{x} + \check{y} , & \check{x}; \overline{\overline{x; \overline{y} + \overline{y}}} &= \overline{\overline{y}} . \end{aligned}$$

- meet can be defined as $x \cdot y = \overline{\overline{x} + \overline{y}}$
- a partial order is given by $x \leq y \Leftrightarrow x + y = y$
- a relation algebra is representable iff it is isomorphic to a proper one
- too weak to prove some truths about binary relations
- but: translation into logic can introduce quite complex expressions with nested quantifiers and destroy the inherent algebraic structure
- equational theory is undecidable

On Automating the Calculus of Relations

- interactive proof-checkers (von Oheimb, Kahl)
- special-purpose proof systems, e.g.,
 - tableaux calculi (Maddux)
 - Rasiowa-Sikorski calculus (Orlowska)
- translation into the (undecidable) fragment of predicate logic (SPASS 3.0)

Why not use off-the-shelf theorem provers combined with Tarski's equational axioms?

Results and Experience

- more than 100 theorems proved as base library
- most of them without difficulties
- some needed restriction of axioms or additional hypothesis
Axiom selection systems seem necessary (e.g., SRASS)
- Prover9/Waldmeister perform best
(evaluation of more than 10 ATP systems)
- a comparison between our approach and translation into predicate logic is still missing

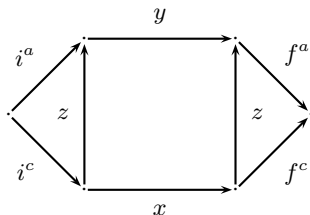
Simulation Laws for Data Refinement

- **program refinement** investigates the stepwise transformation of abstract specifications to executable code
- **data refinement** is a variant that considers the transformation of *abstract* data types (ADTs) into *concrete* ADTs

Abstract ADTs

- observed through the effects of their operations on states
- operations are usually modelled as binary relations
- further operations model the initialisation and finalisation of ADTs

Simulations



Definition (de Roeper, Engelhardt)

Let x , y and z be elements of some relation algebra.

- x U -simulates y with respect to z ($x \subseteq_U^z y$) if $\check{z}; x; z \leq y$,
- x L -simulates y with respect to z ($x \subseteq_L^z y$) if $\check{z}; x \leq y; \check{z}$,
- x \check{U} -simulates y with respect to z ($x \subseteq_{\check{U}}^z y$) if $x \leq z; y; \check{z}$,
- x \check{L} -simulates y with respect to z ($x \subseteq_{\check{L}}^z y$) if $x; z \leq z; y$.

(z is the abstraction relation; \subseteq the simulation relation)

Data Refinement

Theorem (soundness of simulations)

- L - and \check{L} -simulations are sound for data refinement
- U -simulations are sound if the simulation relation is total ($1 \leq x; \check{x}$)
- \check{U} -simulations are sound if the simulation relation is a function ($\check{x}; x \leq 1$)

Remarks

- the proof uses structural induction
- the entire induction cannot be treated by ATP systems
- but: all base cases and induction steps can be proven fully automatically

Stepwise Proof for L -simulation

base cases

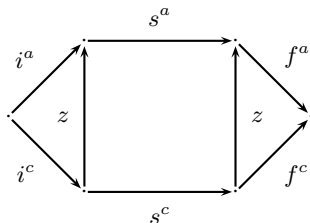
- $0 \subseteq_L^z 0$ and $1 \subseteq_L^z 1$
(Prover9: $< 10 s$)
- the case of atomic operations holds by assumption

induction step

Let $s_1^c \subseteq_L^z s_1^a$ and $s_2^c \subseteq_L^z s_2^a$.

- composition: $s_1^c; s_2^c \subseteq_L^z s_1^a; s_2^a$
(Prover9: $< 3 s$)
- choice: $s_1^c + s_2^c \subseteq_L^z s_1^a + s_2^a$
(Prover9: $< 2 s$ using an additional distributivity law)
- iteration: $(s_1^c)^* \subseteq_L^z (s_1^a)^*$
(Prover9: $< 1 s$)
(* is the reflexive, transitive closure and can be axiomatised in first-order logic)

Stepwise Proof for L -simulation



final step Let $i^c \leq i^a; \check{z}, \check{z}; f^c \leq f^a$ and $s^c \subseteq_L^z s^a$

- $i^c; s^c; f^c \leq i^a; s^a; f^a$
(Prover9: < 1 s)

Conclusion

- combination of relation algebra and ATP systems is feasible
- ATP systems can speed up finding proofs / counterexamples (We found flaws in the soundness proof for U and \check{U} -simulations)
- alternative higher-order, special-purpose, translational and finitist approaches
- examples suggest that formal methods become more automatic
- practical verification tasks often require the integration of algebraic techniques into a wider context:
Most induction proofs require higher-order reasoning, but the base case and the induction step can often be discharged algebraically.

Outlook

- results hopefully pave the way for interesting applications in relational software development methods like B, Z or Alloy
- relations are not only used for ADTs, e.g.,
weakest liberal precondition ($wlp(x, p) = \overline{x; \overline{p}}$) or
weakest precondition
- find ways of combining the abstract pointfree level with the concrete data
- integration of ordered chaining techniques (Bachmair, Ganzinger) into modern ATP systems would make relational reasoning more efficiently
- a combination with hypothesis learning techniques seems indispensable for tackling more complex applications and larger specifications