Lecture II - part 2 the knowledge base system IDP

June 7, 2016

Inference of the KBS: progress report

Model expansion and visualisation Imperative + Declarative Programming (IDP)

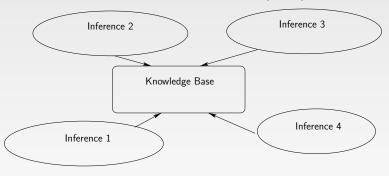
Inference of the KBS: progress report

Model expansion and visualisation Imperative + Declarative Programming (IDP)

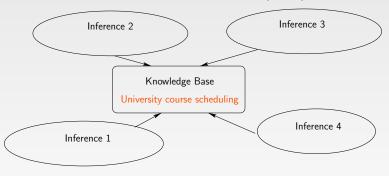
How to use Logic for problem solving

A logic theory is a bag of (descriptive) information

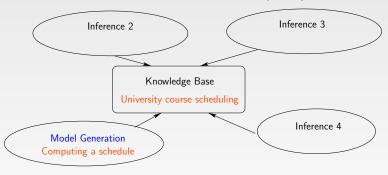
- A logic theory cannot be executed
- A logic theory is not a program
- A logic theory is not a representation of a problem
- So how can we use a logic theory to solve problems?



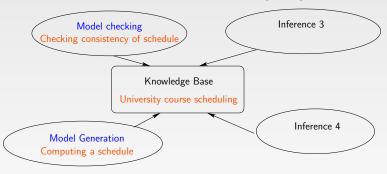
- Manages a declarative Knowledge Base (KB): a theory
- Equiped with different forms of inference:



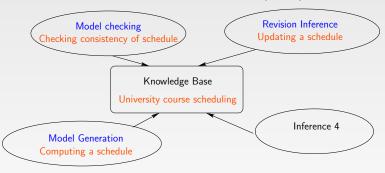
- ► Manages a declarative Knowledge Base (KB): a theory
- Equiped with different forms of inference:



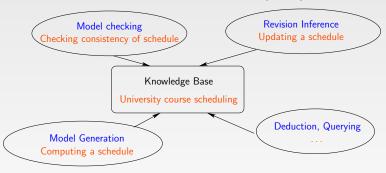
- ► Manages a declarative Knowledge Base (KB): a theory
- Equiped with different forms of inference:
 - Model generation: Computing a schedule



- Manages a declarative Knowledge Base (KB): a theory
- Equiped with different forms of inference:
 - Model generation: Computing a schedule
 - Model checking: Verifying consistency of a schedule



- Manages a declarative Knowledge Base (KB): a theory
- Equiped with different forms of inference:
 - Model generation: Computing a schedule
 - Model checking: Verifying consistency of a schedule
 - Update and Revision: Updating a given schedule



- Manages a declarative Knowledge Base (KB): a theory
- Equiped with different forms of inference:
 - Model generation: Computing a schedule
 - Model checking: Verifying consistency of a schedule
 - Update and Revision: Updating a given schedule
 - Deduction for verification of the KB Querying of defined predicates, ...

A KBS demo

The course selection demo: interactive configuration http://krr.bitbucket.org/courses/ 5 Used forms of inference:

- Model Checking (P)
- Propagation (P)
- Model Generation (NP)
- Model Generation+Optimization (NP^{NP})
- Explanation (P)

A KBS demo

Demonstrating a principle that procedural programming languages can't do:

Reusing the same specification/theory/knowledge base to solve different types of problems.

Implementation of KBS

IDP3:

- A KBS system
- Programming environment
 - Programming with theories, structures, inference methods
 - In an extension of procedural language Lua:

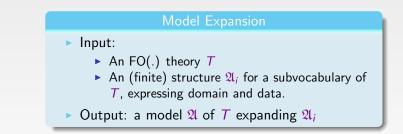
Built by KRR-members Broes De Cat, Bart Bogaerts, Joachim Jansen, Pieter Van Hertum, Jo Devriendt, Ingmar Dasseville (and ex-members Johan Wittocx, Maarten Mariën, Stef De Pooter)

Implementation of KBS

Forms of inference currently under development:

- (Finite) Model expansion (the core component of IDP3)
- Optimisation
- Propagation
- Querying structures
- Δ-model generation and revision:
 - \blacktriangleright $\sim\!\!view$ materialisation and update in databases
 - computing & updating defined predicates
- Progression of temporal FO(.) theories.
- Model revision
- Debugging, Explanation.

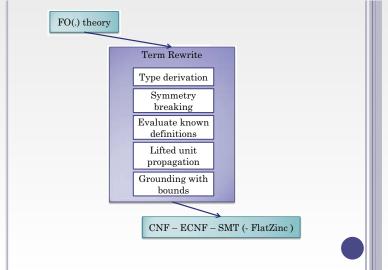
Model generation/expansion



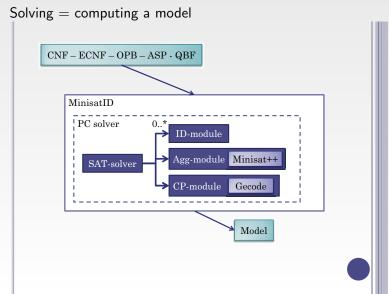
Special case: Herbrand Model Generation

The grounder

$Grounding = Eliminating \ quantification$



MinisatID: SMT solver



Technology

Technologies from different computational logic areas integrated:

- Constraint Programming technology
- Sat Modulo Theory (SMT)
- Logic Programming
- Answer Set Programming
- MIP: Mixed Integer Programming

| Solver | AST (sec.) | PSI (%) |
|-----------|------------|---------|
| minisatid | 950.91 | 51.62 |
| g12cpx | 1126.98 | 41.68 |
| fzn2smt | 1143.47 | 38.13 |
| ortools | 1316.25 | 30.65 |
| g12lazyfd | 1306.10 | 30.31 |
| gecode | 1354.65 | 29.51 |
| izplus | 1350.42 | 28.05 |
| bprolog | 1423.45 | 24.73 |
| jacop | 1435.123 | 24.67 |
| g12fd | 1424.80 | 23.57 |
| mistral | 1525.83 | 16.91 |
| g12mip | 1597.54 | 12.58 |

 $\label{eq:table} \begin{array}{l} {\sf Table}: \ {\sf Experimental evaluation of MiniZinc solvers on the CSPs in Benchmark Set B AmadiniGM13.} \end{array}$

| Benchmark | # solved IDP | # solved Gringo-Clasp |
|---------------------------------|--------------|-----------------------|
| Perm. P. Matching | 10 | 10 |
| Valves Location $*$ | 7 | 4 |
| Still-Life * | 2 | 3 |
| Graceful Graphs | 3 | 9 |
| Bottle Filling | 10 | 10 |
| NoMystery | 9 | 6 |
| Sokoban | 7 | 5 |
| Ricochet Robots | 7 | 10 |
| Crossing Minim. * | 0 | 9 |
| Solitaire | 8 | 9 |
| Weighted Sequence | 10 | 10 |
| Stable Marriage | 10 | 10 |
| Incremental Sched. | 6 | 5 |
| Visit All core | 6 | 7 |
| Knight's Tour _{core} | 1 | 0 |
| Maximal Clique $*_{core}$ | 0 | 1 |
| Graph Colouring _{core} | 7 | 4 |

 $\ensuremath{\mathsf{Table}}$: Experimental results for benchmarks of the 2013 ASP competition.

Other demos

- Model expansion with logic-based visualisation
- Programming with logical inference
- Temporal Reasoning: planning
- Temporal reasoning: execution and optimisation

Demos and examples can be accessed from https://dtai.cs.kuleuven.be/software/idp/

IDPd3: logic based visualisation

dtai.cs.kuleuven.be/krr/idp-ide/?present=
SudokuVisualisatie

- This application serves to solve sudoku puzzles and to visualise the outcome. The theory T expresses the 3 laws of sudokus: one occurrence of each number in each row, column and block. Model expansion inference solves a puzzle specified in the input structure by returning a model \mathfrak{A} . De solution is sudoku^{\mathfrak{A}}. The user theory T_D3 (see next slide) is mainly a definition of IDPd3 graphical predicate and function symbols defined in terms of symbols interpreted in \mathfrak{A} . Δ -model expansion on T_D3 and input structure \mathfrak{A} computes a value for these predicates which is then transferred to the graphical program d3.
- This is a simple illustration of logic to transform one sort of datastructure in another.

```
d3_type(1, Text(r, k)) = text <-.
d3_x(1, Text(r, k)) = 5*k <-.
d3_y(1, Text(r, k)) = 5*r + 1 <-.
d3_text_size(1, Text(r, k)) = 3 <-.
d3_text_label(1, Text(r, k)) = t <-
sudoku(r, k) = c & toString(c) = t.
d3_color(1, Text(r, k)) = "black" <-.</pre>
```

```
d3_order(1, Cell(r, k)) = 0 <-.
d3_order(1, Text(r,k)) = 1 <-.</pre>
```

This definition defines slide 1 (argument 1) with:

- for each cell (r, c) a rectangular object denoted Cell(r,c), and a text object Text(r,k).
- It defines type, width, height, color, x and y position of the rectangular object.
- It defines type, text size and label, color, x and y position of the text object
- The sudoku number at cell (r, c) is the label of the text object..
- that text objects are in front of rectangular objects

 Δ -model expansion expands a structure interpreting sudoku into a structure interpreting all these graphical symbols. This is fed into d3.

IDPd3: an example

Input structure

$$\textit{sudoku} = \{1, 1 \rightarrow 1; \ldots; 9, 9 \rightarrow 4\}$$

. . .

$\downarrow \Delta \text{-model generation}$

$$\begin{aligned} d3_type &= \{1, Cell(1, 1) \rightarrow rect; 1, Text(1, 1) \rightarrow text; \dots \} \\ d3_color &= \{1, Cell(1, 1) \rightarrow white; 1, Text(1, 1) \rightarrow black; \dots \} \\ d3_x &= \{1, Cell(1, 1) \rightarrow 5; 1, Cell(1, 2) \rightarrow 10; \dots \} \end{aligned}$$

 \downarrow translation to d3 input + d3

A prototype of a knowledge-based programming environment

- IDP3: A programming environment
- High level objects: vocabularies, theories, structures
- Functionalities for manipulation and inference
- Implemented in the language Lua
- A new way of mixing Declarative and Procedural knowledge

A demo: generating Sudoku-puzzles

Sudoku-puzzle requirements'

- (consistency) It should allow one unique solution
- (minimality) If we delete any value of the puzzle, it has at least two solutions.

Background knowledge base in IDP

Vocabulary

```
vocabulary sudokuVoc {
    extern vocabulary grid::simpleGridVoc
    type Num isa nat
    type Block isa nat
    Sudoku(Row,Col) : Num
    InBlock(Block,Row,Col)
}
```

Theory

A demo: generating Sudoku-puzzles

Puzzle := emptyGenerate at most 2 solutions for Puzzle While 2 solutions were found do{ Select a random position where the two solutions differ Extend Puzzle with the value of the first solution at this position Generate at most 2 solutions for Puzzle For each position of Puzzle that contains a value do { Delete the value at this position Generate at most 2 solutions for Puzzle If there are two solutions, undo the deletion of the value. Visualize the puzzle and its unique solution

Procedures

```
procedure createSudoku() {
   math.randomseed(os.time())
   local puzzle = grid::makeEmptyGrid(9)
   stdoptions.nrmodels = 2
   local currsols = modelExpand(sudokuTheory,puzzle)
   while \#currsols > 1 do
      repeat
         col = math.random(1,9)
         row = math.random(1,9)
         num = currsols[1][sudokuVoc::Sudoku](row,col)
      until num ~= currsols[2][sudokuVoc::Sudoku](row,col)
      makeTrue(puzzle[sudokuVoc::Sudoku].graph,{row,col,num})
      currsols = modelExpand(sudokuTheory,puzzle)
   end
```

```
printSudoku(puzzle)
```

```
}
```

Discussion

Two sorts of inferences:

- generating solutions to puzzles: model expansion
- Visualizing through Δ-model expansion
 - computing a model of a definition Δ
 - A special case of model expansion
 - No search
 - Can be implemented very differently
 - View materialisation in deductive databases.

Acces and manipulation of structures.

- Structures are objects in the environment
- Puzzle and its solutions are structures
- Checking and updating values at positions of puzzle

Reasoning on Temporal theories

- Temporal theory T : Linear Time Calculus
- Use Model expansion for planning with optimisation
 - In the lecture I show a little video showing a idpd3-generated video with the optimal plan to remove all gold.
- Use Progression for interactive execution.
 - Input: T, structure \mathfrak{A} representing state at time i
 - Output: structure \mathfrak{A}' representing possible state at time i + 1.
- Illustration: pacman.

Reasoning on Temporal theories

- Temporal theory T : Linear Time Calculus
- Use Model expansion for planning with optimisation
 - In the lecture I show a little video showing a idpd3-generated video with the optimal plan to remove all gold.
- Use Progression for interactive execution.
 - Input: T, structure \mathfrak{A} representing state at time i
 - Output: structure \mathfrak{A}' representing possible state at time i + 1.
- Illustration: pacman.

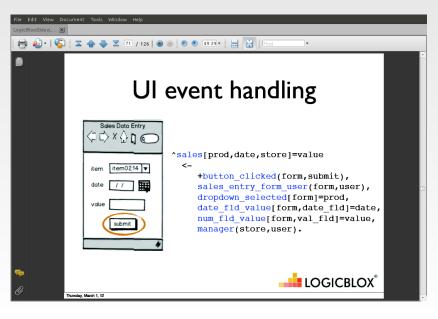
IDP is the only system that we know of that can use the same formal specification to solve both tasks.

The pacman demo:

- Go to the IDP-web page https://dtai.cs.kuleuven.be/software/idp/
- Let column: select "Demos"
- Next page contains different visualised demos of IDP. Select bottom middle "More visualised demos.
- Next page, select link below "Pacman".
- Next page, press Run
- ► A labyrinth with above, a green and black rectangles appear, indicating possible move actions van the pacman. Click the black rectangles and pacman moves in the right direction.
- Due to memory limitations on the server, only a few moves can be played.
- To avoid memory limitations, download the system.

A company running standard software systems on this principle:

LogicBlox - Datalog : http://www.logicblox.com/





Generic interactive configuration

http://krr.bitbucket.org/autoconfig/ Another application, selectable from the demos webpage (left middle box) is interactive decision enactment = business logic terminology for generic interactive configuration.

- Automatic generation of a window with values for all ground literals of a configuration problem
 - Three-valued structure underneath
- User can select values
 - System propagates user choices
 - Optimal propagation versus approximate propagation
- System expands current partial structure on demand
- System searches model minimizing a cost term on demand
- Generic, incremental:
 - Adding symbols
 - Other vocabularies

The application is a benchmark problem of business rules

- USERV
- Deciding car insurance policy for client.

To be submitted to RuleML.

Advantages compared to business rule systems

- Clear declarative sematincs: formal and informal
- Reasoning backward from desired outcome.
- Increased functionalities
 - computing solutions under uncertainty
 - propagation
 - optimisation
 - explanation (is not implemented)
 - **۱**...

Integrating logic in imperative languages

Joost Vennekens: Lowering the learning curve for declarative programming: a Python API for the IDP system. International Workshop on User-Oriented Logic Programming (IULP 2015) 31st August 2015, Cork (Ireland) CoRR abs/1511.00916

```
idp.Type("Number", range(10))
idp.Function("Given(Square): Number", d)
idp.Function("Sol(Square): Number")
idp.Define("Diff(Square, Square)", "lambda x,y: x !=
y and (SameRow(x,y) or SameCol(x,y) or
SameSmallSq(x,y))")
idp.Constraint("all(Sol(x) == Given(x) for x in
Square if Given(x) != 0)")
idp.Constraint("all(Sol(x) != 0 for x in Square)")
idp.Constraint("all(Sol(x) != Sol(y) for (x,y) in
Diff)")
show(idp.Sol)
```

See webpages:

- https://dtai.cs.kuleuven.be/software/idp/
- Online IDE (edit, safe, download, run, many examples)
- Demos
 - map coloring
 - interactive course selection
 - more visualised demos
 - pacman interactive execution
 - Science Week example : visualisation of errors
- tutorial, manual
- ICLP-contest 2015 solutions

Future

Knowledge-based software engineering

- Important gains to be made:
 - development time
 - compactness
 - correctness
 - reuse
 - maintainability
- Great scientific and practical challenges

We are in the process of searching niches with industry where our technology could already make a difference

A KRR-team with Ingmar Dasseville, Jo Devriendt and Matthias van der Hallen won the International Logic Programming and Constraint Programming competition with IDP.