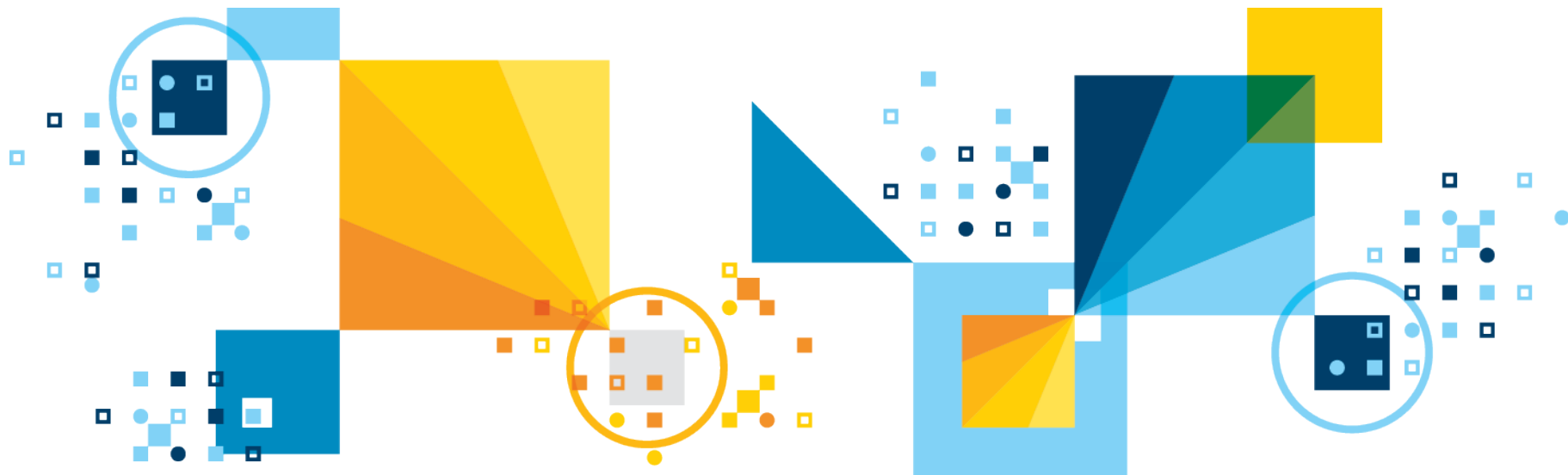


Solving Industrial Scheduling Problems with Constraint Programming

Philippe Laborie
IBM Analytics, Decision Optimization



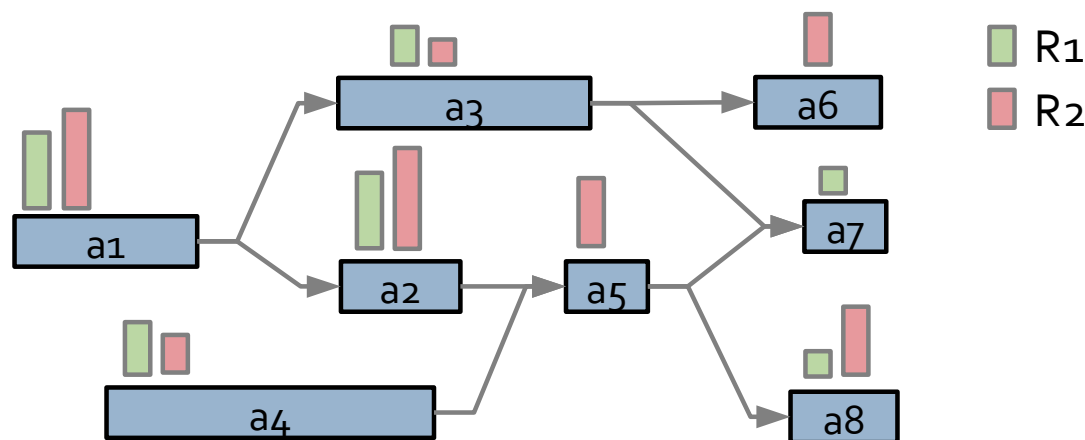
Sept. 23, 2015

Outline

- Scheduling problems: from theory to practice
- CP extensions for scheduling
- Simplifying model design & problem resolution

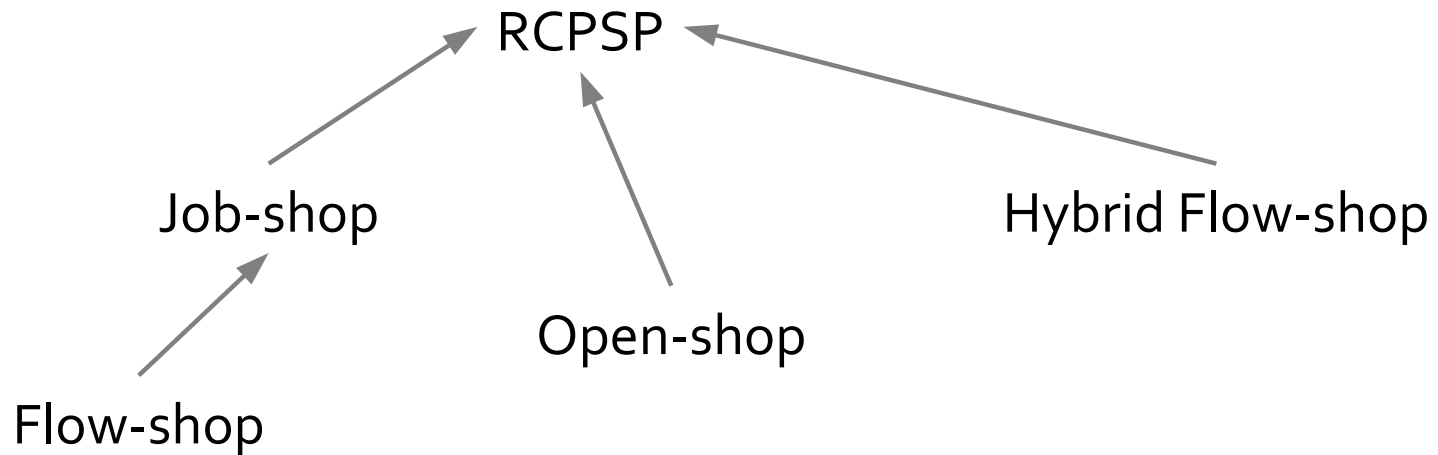
Scheduling problems: from **theory** to practice

- **Theory:** e.g. RCPSP (Resource Constrained Project Scheduling)
 - Problem definition:
 - A set of n tasks A_1, \dots, A_n with specified duration D_i for task A_i
 - A set of precedence constraints between tasks (precedence graph)
 - A set of m resources R_1, \dots, R_m with specified capacity C_j for resource R_j
 - Each task A_i requires some resources in a given quantity
 - Objective is to minimize project makespan
 - Classical benchmark: PSPLib = 30-120 tasks



Scheduling problems: from **theory** to practice

- **Theory:** e.g. RCPSP (Resource Constrained Project Scheduling)
 - From an academical point of view, this is a very generic problem that subsumes many other classical scheduling problems



Scheduling problems: from theory to practice

- Some real scheduling applications
 - Aerospace: Project scheduling, Aircraft assembly, Assembly line configuration, Satellite communication & observation, Rover activities ...
 - Energy & Utilities: Nuclear plant outage scheduling, Maintenance, Production planning ...
 - Mining: Open pit mining ...
 - Logistics, Supply Chain: Vehicle routing, Bikes and car sharing ...
 - Manufacturing: Production scheduling, Assembly lines, Factory configuration, Test scheduling ...
 - Media & Communications: Advertizing & program scheduling, Event & personnel scheduling ...
 - Travel & transportation: Airport scheduling (gates, landing, parking, ...), Port scheduling (quay cranes ...), Train scheduling ...
 - Health: Employee, Patient scheduling, Pharmaceutical products tests ...
 - Agriculture: Crop scheduling (harvesting ...)
 - General audience: Personal schedule organizer, Theme park planner ...

Scheduling problems: from theory to **practice**

- **Practice:** the problem looks like ...

this ...



or this ...



Scheduling problems: from theory to **practice**

- **Practice:** the problem is **large**
 - Typical problems are larger than 1000 activities
 - E.g. we solved a 1.000.000 tasks RCPSP-like problem for an aircraft manufacturer

Scheduling problems: from theory to **practice**

- **Practice:** the problem is **complex**
 - Heterogeneous types of **decisions**: start/end dates, resource allocation, mode selection, activity non-execution, activity durations, activity interruptions, resource quantities, producer/consumer matching, batch configuration, resource overconsumption, resource configuration, ...
 - Complex **resources**: inventories, setup times and costs, calendars/shifts, moving resources, spatial constraints, synchronization constraints (e.g. conveyer belt), batch constraints
 - Complex **activities**: optional, interruptible, setups, maintenance, work breakdown structure
 - Complex **objective function**: earliness/tardiness, temporal preferences, resource related costs (many types!), most of the problems are multi-objectives

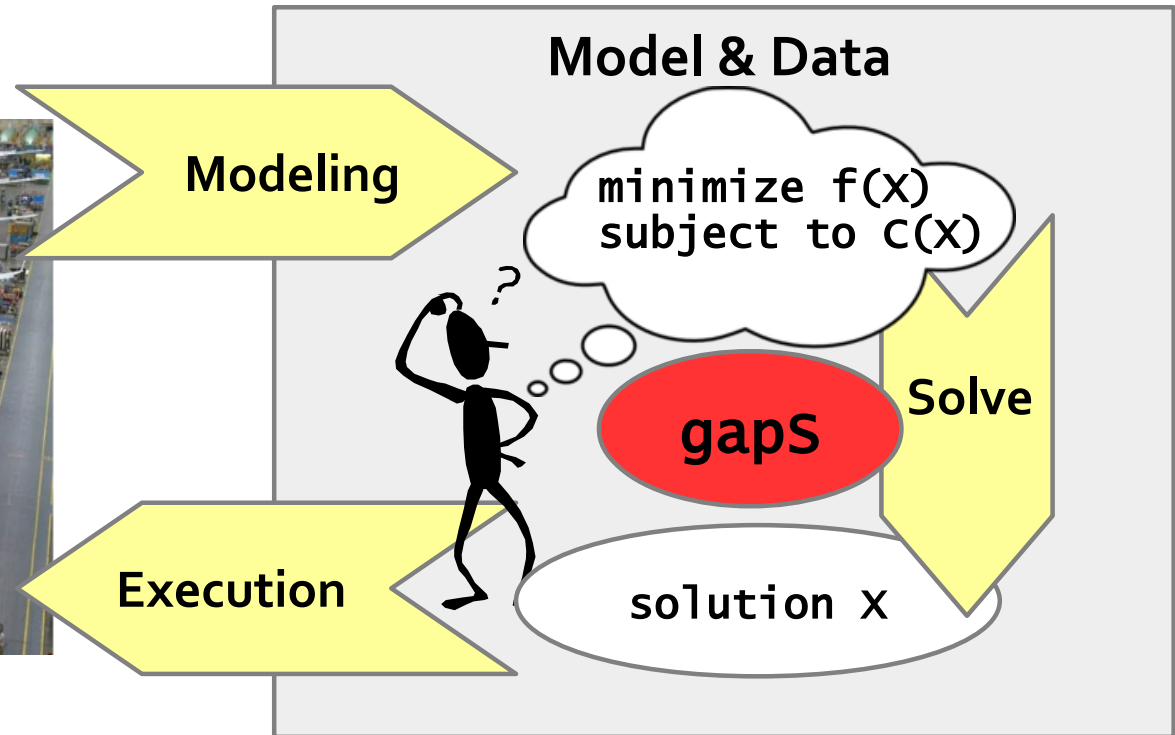
Scheduling problems: from theory to **practice**

- **Practice:** the problem is **not well defined**
 - Implicit constraints / objectives
 - Some aspects of the problem are critical, others can be:
 - Approximated
 - Relaxed
 - Over-constrained
 - Considered in a post-processing step
 - ...
- **Practice:** the data is:
 - Hard to get (different data sources, confidentiality, ...)
 - Uncertain / Imprecise / Incomplete / Incorrect
- Customers are expecting **good solutions** to **their problem**, they usually do not care about how the problem is modeled/solved

Scheduling problems: from theory to practice

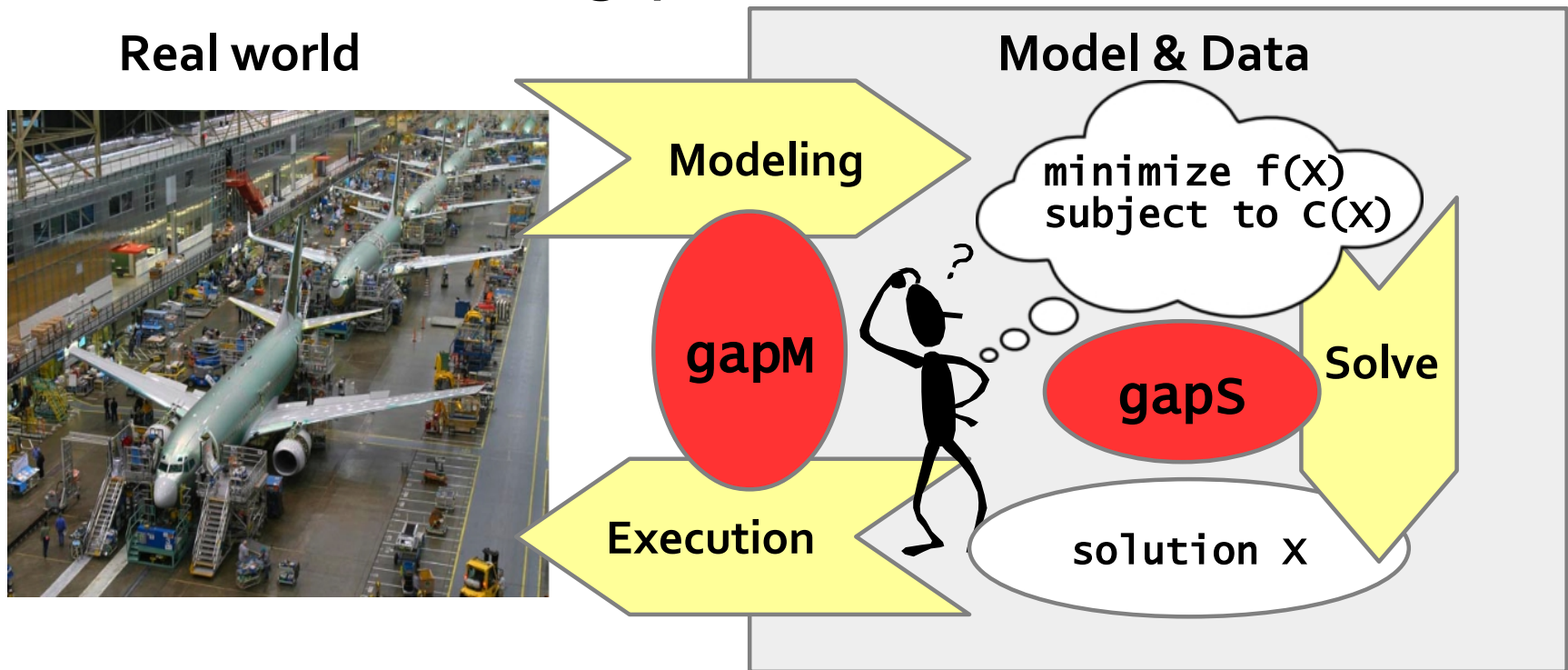
Practice: mind the real gap

Real world



Scheduling problems: from theory to practice

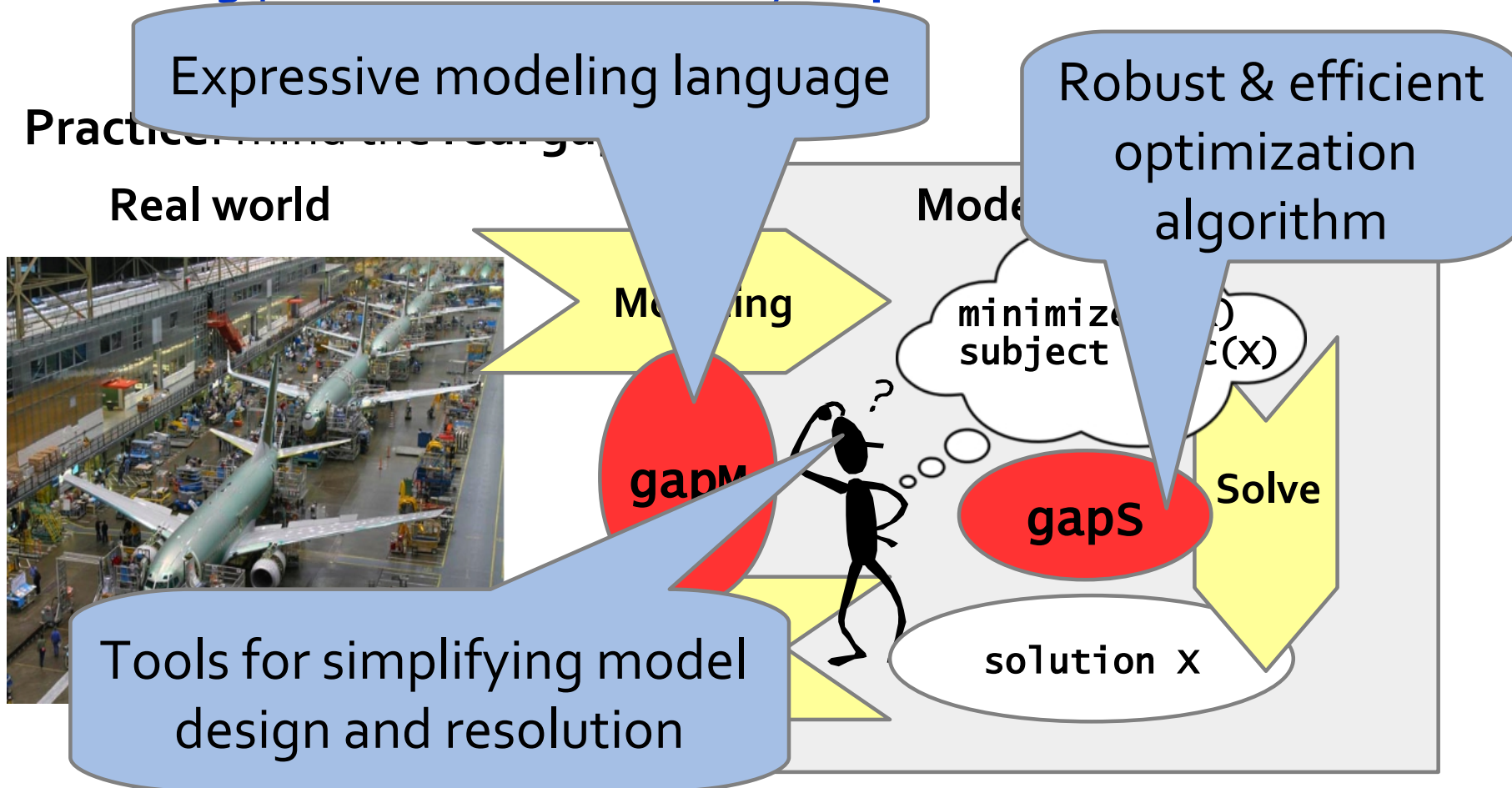
Practice: mind the real gap



- “Essentially, all models are wrong, but some are useful” – G. Box
- The real gap you must care about is **gapM \oplus gapS**

Scheduling problems: from theory to practice

■ Practical modeling and solving



- “Essentially, all models are wrong, but some are useful” – G. Box
- The real gap you must care about is $\text{gapM} \oplus \text{gapS}$

CP extensions for scheduling

- Extension of classical CSP with a new type of decision variable:
optional interval variable :

$$\text{Domain}(x) \subseteq \{\perp\} \cup \{[s,e) \mid s,e \in \mathbb{Z}, s \leq e\}$$

Absent interval

Interval of integers

- Introduction of mathematical notions such as **sequences** and **functions** to capture temporal dimension of scheduling problems

CP extensions for scheduling

- In scheduling models, **interval variables** usually represent an interval of time during which some property hold (e.g. an activity executes) and whose end-points (start/end) are decision variables of the problem.
- Examples:
 - A sub-project in a project, a task in a sub-project (Work Breakdown Structure)
 - A batch of operations
 - The setup of a tool on a machine
 - The moving of an item by a transportation device
 - The utilization interval of a resource
- Idea of the model (and search) is to avoid the enumeration of start/end values (continuous time)

CP extensions for scheduling

- An interval variable can be **optional** meaning that it is a decision to have it present or absent in a solution.
- Examples:
 - Unperformed tasks and optional sub-projects
 - Alternative resources, modes or recipes for processing an order, each mode specifying a particular combination of operational resources
 - Operations that can be processed in different temporal modes (e.g. series or parallel)
 - Activities that can be performed in an alternative set of batches or shifts

CP extensions for scheduling

- Example: RCPSP

```
dvar interval a[i in Tasks] size i.pt;

cumulFunction usage[r in Resources] =
    sum (i in Tasks: i.qty[r]>0) pulse(a[i], i.qty[r]);

minimize max(i in Tasks) endOf(a[i]);
subject to {
    forall (r in Resources)
        usage[r] <= Capacity[r];
    forall (i in Tasks, j in i.succs)
        endBeforeStart(a[i], a[<j>]);
}
```


CP extensions for scheduling

- Example: Job-shop scheduling problem

```
dvar interval op[j in Jobs][p in Pos] size Ops[j][p].pt;
dvar sequence mchs[m in Mchs] in
  all(j in Jobs, p in Pos : Ops[j][p].mch == m) op[j][p];

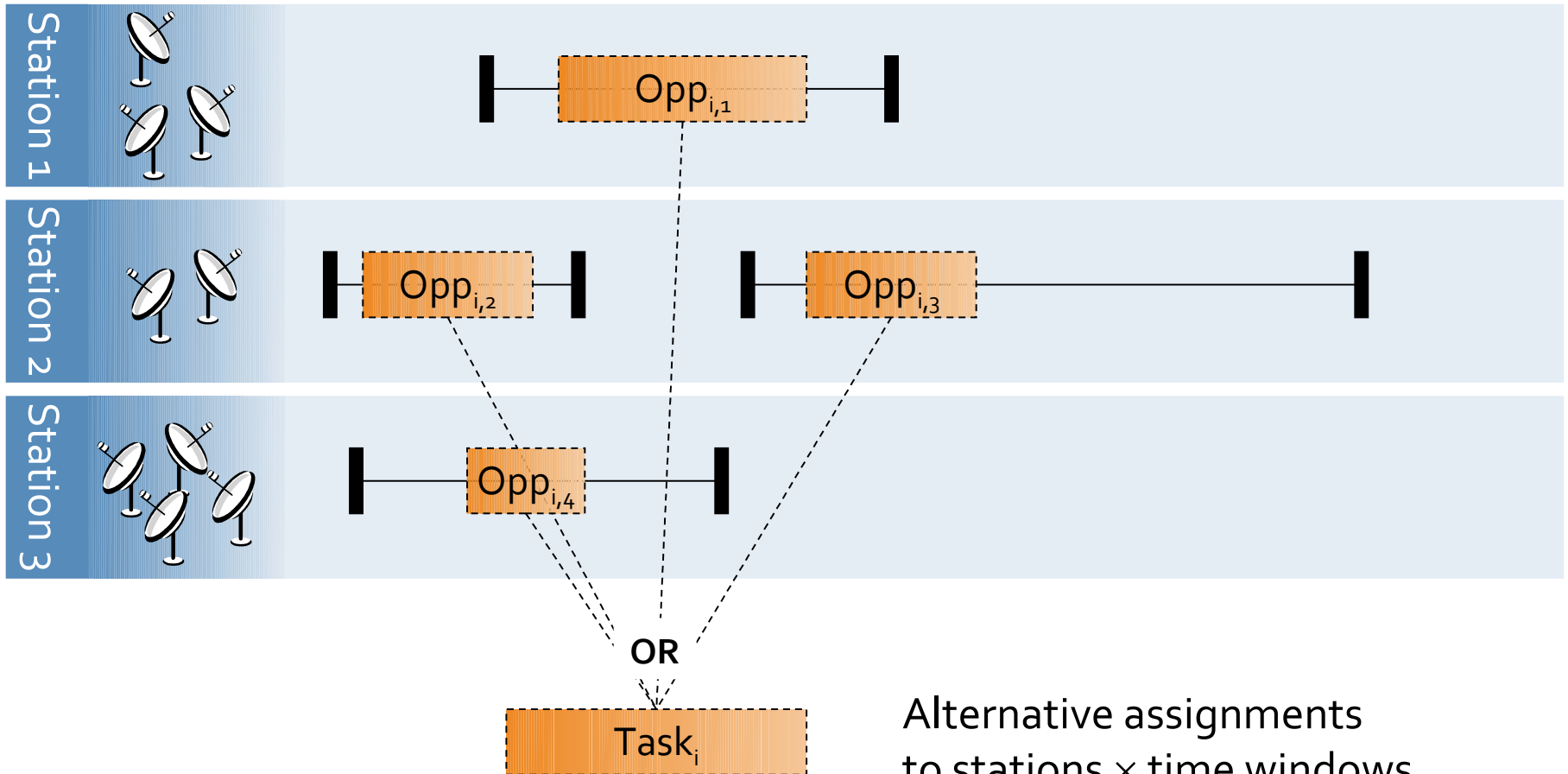
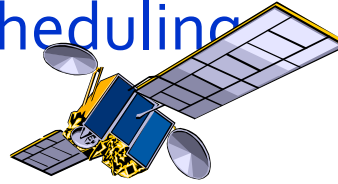
minimize max(j in Jobs) endOf(op[j][nbPos-1]);
subject to {
  forall (m in Mchs)
    noOverlap(mchs[m]);
  forall (j in Jobs, p in 1..nbPos-1)
    endBeforeStart(op[j][p-1], op[j][p]);
}
```

CP extensions for scheduling

- Satellite Control Network scheduling problem [1]
- n communication tasks for Earth orbiting satellites must be scheduled on a total of 32 antennas spread across 13 ground-based tracking stations
- In the instances, n ranges from 400 to 1300
- Objective: maximize the number of scheduled tasks

[1] Kramer & al.: Understanding Performance Trade-offs in Algorithms for Solving Oversubscribed Scheduling.

CP extensions for scheduling



Alternative assignments
to stations \times time windows
(opportunities)

CP extensions for scheduling

```
1 using CP;
2
3 tuple Station {
4     string name; // Ground station name
5     int id;      // Ground station identifier
6     int cap;     // Number of available antennas
7 }
8
9 tuple Opportunity {
10     string task; // Task
11     int station; // Ground station
12     int smin;    // Start of visibility window of opportunity
13     int dur;     // Task duration in this opportunity
14     int emax;    // End of visibility window of opportunity
15 }
16
17 {Station} Stations = ...;
18 {Opportunity} Opportunities = ...;
19 {string} Tasks = { o.task | o in Opportunities };
20
21 dvar interval task[t in Tasks] optional;
22 dvar interval opp[o in Opportunities] optional in o.smin..o.emax size o.dur;
23
24 maximize sum(t in Tasks) presenceOf(task[t]);
25 subject to {
26     forall(t in Tasks)
27         opportunitySelection: alternative(task[t], all(o in Opportunities: o.task==t) opp[o]);
28     forall(s in Stations)
29         numberOfAntennas: sum(o in Opportunities: o.station==s.id) pulse(opp[o],1) <= s.cap;
30 }
```

Automatic Search

- Search algorithm is **Complete**
- Core CP techniques used as a building block:
 - Tree search (Depth First)
 - Constraint propagation
- But also:
 - Deterministic multicore parallelism
 - Model presolve
 - Algorithms portfolios
 - Machine learning
 - Restarting techniques
 - Large Neighborhood Search
 - No-good learning
 - Impact-based branching
 - Opportunistic probing
 - Dominance rules
 - LP-assisted heuristics
 - Randomization
 - Evolutionary algorithms

Automatic Search

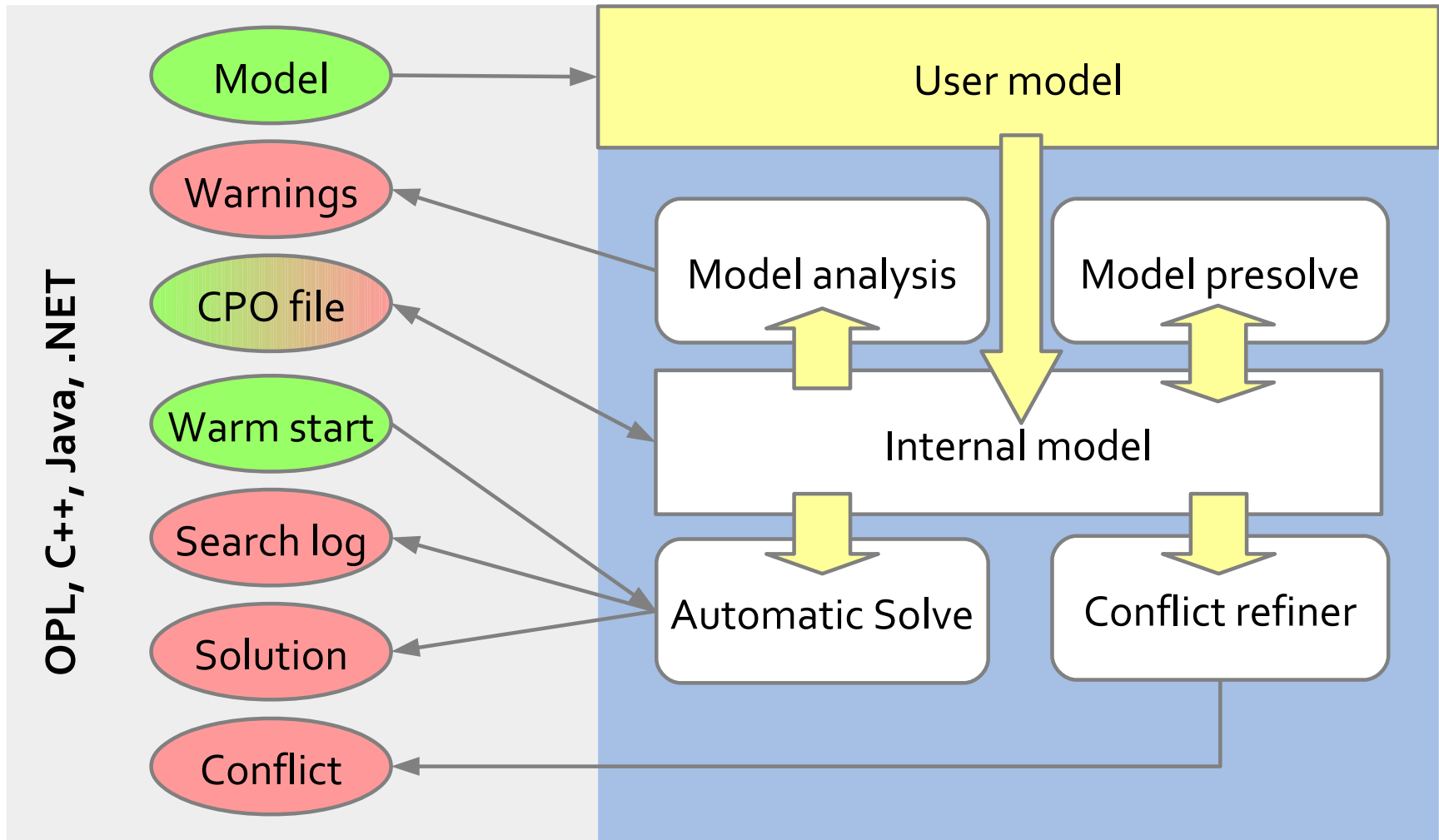
- Very good performance results on academical problems
 - CP-AI-OR 2015:

Benchmark set	Number of instances	Lower bound improvements	Upper bound improvements	Closed instances
JobShop	48	40	3	15
JobShopOperators	222	107	215	208
FlexibleJobShop	107	67	39	74
RCPSP	472	52	1	0
RCPSPMax	58	51	23	1
MultiModeRCPSP (j30)	552	No reference	3	535
MultiModeRCPSPMax	85	84	77	85

Table 1. Results summary

- Winner of the CP-2015 Industrial Modelling Challenge with a very simple model: csplib.org/Problems/probo73/models/model.mod.html
- More important: very good performance on industrial problems

Simplifying model design & problem resolution



Tools: I/O format

- Objective:
 - Make it easier to understand the content of a model
 - Communicate a model to IBM support team regardless of the API used to build it (OPL, C++, Java, .NET)
- Structure of a .cpo file
 - Human readable
 - Flat (no cycle, forall statements)
 - No user defined data types
 - Internal information such as CPO version or platform used
 - Includes search parameter values
- Facilities
 - Export model before/instead of solve
 - Export model during solve (with current domains)
 - Import model instead of normal modeling

Tools: I/O format

// Interval-related variables:

```
"task(1)" = intervalVar(optional);
"task(1A)" = intervalVar(optional);
...
"opp({1,1,62})" = intervalVar(optional, start=62..intervalmax, end=0..99, size=25);
"opp({1A,1,32})" = intervalVar(optional, start=32..intervalmax, end=0..69, size=33);
...
```

// Objective:

```
maximize(sum([presenceOf("task(1)"), presenceOf("task(1A)"), ...]));
```

...

// Constraints:

```
alternative("task(1)", ["opp({1,1,62})"], 1);
```

...

```
pulse("opp({3,1,58})", 1) + pulse("opp({1,1,62})", 1) + ... <= 4;
```

...

```
parameters {
    LogVerbosity = Quiet;
}
```

Tools: model warnings

- Like a compiler, CP Optimizer can analyze the model and print some warnings
 - When there is something suspicious in the model
 - Regardless how the model was created (C++, OPL, ...)
 - Including guilty part of the model in the cpo file format
 - Including source code line numbers (if known)
 - 3 levels of warnings, more than 50 types of warnings

satellite.cpo:2995:1: Warning: Constraint 'alternative': there is only one alternative interval variable.

alternative("task(1)", ["opp({1,1,62})"], 1)

satellite.cpo:2996:1: Warning: Constraint 'alternative': there is only one alternative interval variable.

alternative("task(1A)", ["opp({1A,1,32})"], 1)

Tools: model presolve

- Objective: **automatically** reformulate the model in order to speed-up its resolution
- Works on an internal representation of the model
- Different types of presolve:
 - Aggregation of basic constraints into global constraints
 - Constraint strengthening
 - Simplifications and factorizations

Tools: model presolve

- Examples of presolve rules
 - Common sub-expression elimination
 - Aggregation of $x \neq y$ cliques as **`allDifferent([x,y,...])`**
 - Precedence strengthening
 - If a and b cannot overlap and **`startsBeforeStart(a,b)`**
 - Then **`endsBeforeStart(a,b)`**
 - Precedence recognition
 - If **`endOf(a, - ∞) \leq startOf(b, $+\infty$)`**
 - Then **`endsBeforeStart(a,b)`**
 - Precedences are aggregated into a “time net” (STN) for faster and stronger propagation
 - 2-SAT clauses recognition
 - **`presenceOf(a) \leq presenceOf(b)`**
 - Such clauses are aggregated into a “logical net” for stronger propagation

Tools: search log

- Objective: understand what happens during the automatic search

```

! -----
! Maximization problem - 2980 variables, 853 constraints
! Workers                = 2
! TimeLimit              = 30
! Initial process time : 0.01s (0.00s extraction + 0.01s propagation)
!   . Log search space  : 4627.3 (before), 4627.3 (after)
!   . Memory usage      : 16.9 MB (before), 19.7 MB (after)
! Using parallel search with 2 workers.
! -----
!
!      Best Branches  Non-fixed   W      Branch decision
*      746      3945 0.79s      1      -
      746      4000      2924      1      on task("8")
      746      4000      2908      2      on opp({"186",2,66})
...
! Time = 1.37s, Explored branches = 35832, Memory usage = 55.5 MB
!
!      Best Branches  Non-fixed   W      Branch decision
      818      12000      2920      1      on task("184")
...
! -----
! Search terminated by limit, 6 solutions found.
! Best objective        : 826
! Number of branches    : 709092
! Number of fails       : 179648
! Total memory usage     : 54.5 MB (52.9 MB CP Optimizer + 1.6 MB Concert)
! Time spent in solve    : 30.03s (30.01s engine + 0.01s extraction)
! Search speed (br. / s) : 23625.4
! -----

```

Problem
characteristics

Tools: search log

- Objective: understand what happens during the automatic search

```

! -----
! Maximization problem - 2980 variables, 853 constraints
! Workers                = 2
! TimeLimit              = 30
! Initial process time : 0.01s (0.00s extraction + 0.01s propagation)
!   . Log search space  : 4627.3 (before), 4627.3 (after)
!   . Memory usage      : 16.9 MB (before), 19.7 MB (after)
! Using parallel search with 2 workers.
! -----
!
!      Best Branches  Non-fixed   W      Branch decision
*      746      3945 0.79s      1      -
!      746      4000      2924      1      on task("8")
!      746      4000      2908      2      on opp({"186",2,66})
!
...
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!
!      Best Branches  Non-fixed   W      Branch decision
!      818      12000      2920      1      on task("184")
!
...
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! Total memory usage     : 54.5 MB (52.9 MB CP Optimizer + 1.6 MB Concert)
! Time spent in solve    : 30.03s (30.01s engine + 0.01s extraction)
! Search speed (br. / s) : 23625.4
! -----

```

Modified
parameter values

Tools: search log

- Objective: understand what happens during the automatic search

```

! -----
! Maximization problem - 2980 variables, 853 constraints
! Workers                = 2
! TimeLimit              = 30
! Initial process time : 0.01s (0.00s extraction + 0.01s propagation)
!   . Log search space  : 4627.3 (before), 4627.3 (after)
!   . Memory usage      : 16.9 MB (before), 19.7 MB (after)
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!      Best Branches  Non-fixed   W      Branch decision
*      746      3945 0.79s        1          -
!      746      4000      2924      1      on task("8")
!      746      4000      2908      2      on opp({"186",2,66})
!
...
! Time = 1.37s, Explored branches = 35832, Memory usage = 55.5 MB
!
!      Best Branches  Non-fixed   W      Branch decision
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...
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! Search speed (br. / s) : 23625.4
! -----

```

Root node
information

Tools: search log

- Objective: understand what happens during the automatic search

```

! -----
! Maximization problem - 2980 variables, 853 constraints
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! Initial process time : 0.01s (0.00s extraction + 0.01s propagation)
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!      Best Branches  Non-fixed   W      Branch decision
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      746      4000      2924      1      on task("8")
      746      4000      2908      2      on opp({"186",2,66})
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! Search speed (br. / s) : 23625.4
! -----

```

New incumbent
solutions (time, worker)

Tools: search log

- Objective: understand what happens during the automatic search

```

! -----
! Maximization problem - 2980 variables, 853 constraints
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! Initial process time : 0.01s (0.00s extraction + 0.01s propagation)
!   . Log search space  : 4627.3 (before), 4627.3 (after)
!   . Memory usage      : 16.9 MB (before), 19.7 MB (after)
! Using parallel search with 2 workers.
! -----
!
!      Best Branches  Non-fixed   W      Branch decision
*      746      3945 0.79s        1        -
      746      4000      2924      1      on task("8")
      746      4000      2908      2      on opp({"186",2,66})
...
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!
!      Best Branches  Non-fixed   W      Branch decision
      818      12000      2920      1      on task("184")
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! -----
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! Total memory usage     : 54.5 MB (52.9 MB CP Optimizer + 1.6 MB Concert)
! Time spent in solve    : 30.03s (30.01s engine + 0.01s extraction)
! Search speed (br. / s) : 23625.4
! -----

```

Periodical log
with fail information,
number of unfixed
variables, current decision

Tools: search log

- Objective: understand what happens during the automatic search

```

! -----
! Maximization problem - 2980 variables, 853 constraints
! Workers                = 2
! TimeLimit              = 30
! Initial process time : 0.01s (0.00s extraction + 0.01s propagation)
!   . Log search space  : 4627.3 (before), 4627.3 (after)
!   . Memory usage      : 16.9 MB (before), 19.7 MB (after)
! Using parallel search with 2 workers.
! -----
!
!      Best Branches  Non-fixed   W      Branch decision
*      746      3945 0.79s        1          -
      746      4000      2924      1      on task("8")
      746      4000      2908      2      on opp({"186",2,66})
...
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!      Best Branches  Non-fixed   W      Branch decision
      818      12000      2920      1      on task("184")
...
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! Search terminated by limit, 6 solutions found.
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! Time spent in solve  : 30.03s (30.01s engine + 0.01s extraction)
! Search speed (br. / s) : 23625.4
! -----

```

Final information
with solution status
and search statistics

Tools: warm start

- Objective: Start search from a known (possibly incomplete) solution given by the user (warm start) in order to further improve it or to help to guide the engine towards a first feasible solution
- API: `IloCP::setStartingPoint(IloSolution warmstart)`
- Use cases:
 - Restart an **interrupted search** with the current incumbent
 - Start from an initial solution found by an available **heuristic**
 - Goal programming for **multi-objective** problems
 - When finding an initial solution is hard, solve an initial problem that **maximizes constraint satisfaction** and start from its solution
 - Successively solving **similar** problems (e.g. dynamic scheduling)
 - **Hierarchical** problem solving (e.g. planning → scheduling)

Tools: conflict refiner

- Objective: identify a reason for an inconsistency by providing a **minimal infeasible subset** of constraints for an infeasible model
- Use cases:
 - **Model debugging** (errors in model)
 - **Data debugging** (inconsistent data)
 - The model and data are correct, but the associated data represents a **real-world conflict** in the system being modeled
 - You create an infeasible model to test properties of (or extract information about) a similar model

Tools: conflict refiner

```

1  using CP;
2
3  tuple Station {
4      string name; // Ground station name
5      int id;      // Ground station identifier
6      int cap;     // Number of available antennas
7  }
8
9  tuple Opportunity {
10     string task; // Task
11     int station; // Ground station
12     int smin;    // Start of visibility window of opportunity
13     int dur;     // Task duration in this opportunity
14     int emax;    // End of visibility window of opportunity
15 }
16
17 {Station} Stations = ...;
18 {Opportunity} Opportunities = ...;
19 {string} Tasks = { o.task | o in Opportunities };
20
21 dvar interval task[t in Tasks];
22 dvar interval opp[o in Opportunities] optional in o.smin..o.emax size o.dur;
23
24
25 subject to {
26     forall(t in Tasks)
27         opportunitySelection: alternative(task[t], all(o in Opportunities: o.task==t) opp[o]);
28     forall(s in Stations)
29         numberOfAntennas: sum(o in Opportunities: o.station==s.id) pulse(opp[o],1) <= s.cap;
30 }

```

Tools: conflict refiner

```

!-----
! Satisfiability problem - 2,980 variables, 851 constraints
! Problem found infeasible at the root node
! -----
...
! -----
! Conflict refining - 851 constraints
! -----
!   Iteration      Number of constraints
*           1           851
*           2           426
...
*           58           5
*           59           5
! Conflict refining terminated
! -----
! Conflict status      : Terminated normally, conflict found
! Conflict size        : 5 constraints
! Number of iterations : 59
! Total memory usage   : 13.3 MB
! Conflict computation time : 0.51s
! -----

```

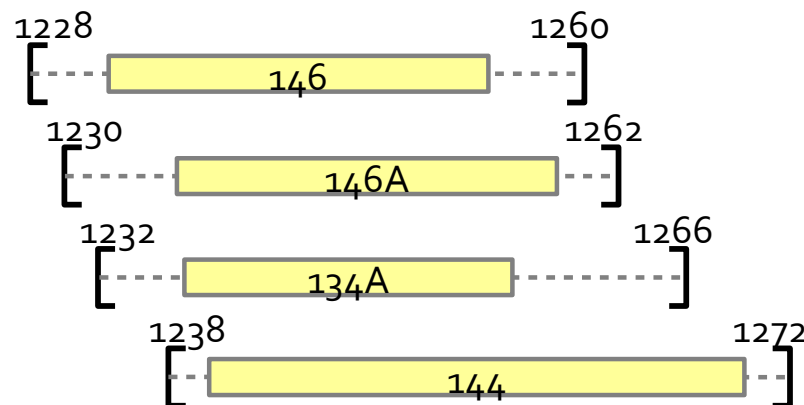
Tools: conflict refiner

■ Conflict:

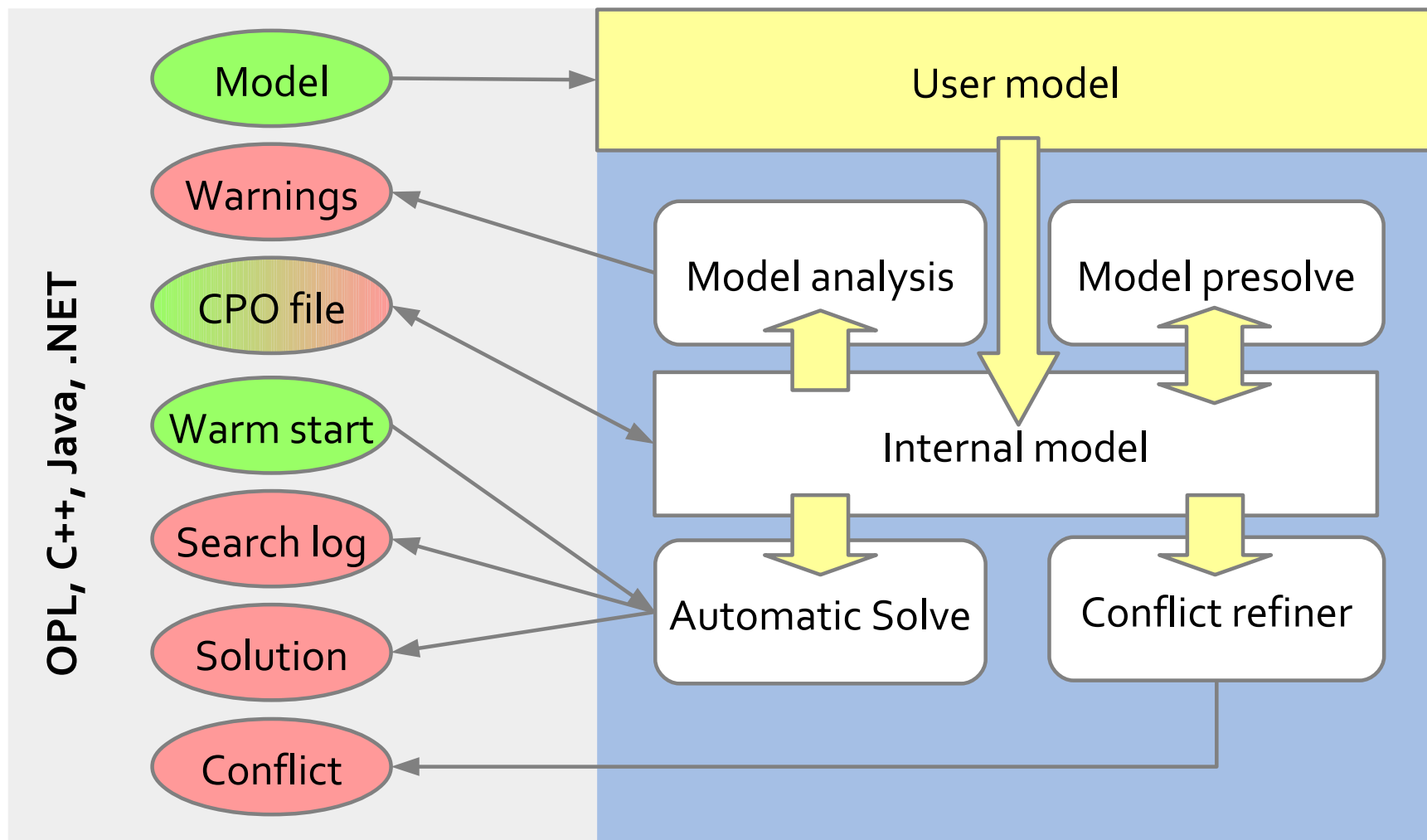
Line	In conflict	Element (5)
26	Yes	opportunitySelection["134A"]
26	Yes	opportunitySelection["144"]
26	Yes	opportunitySelection["146"]
26	Yes	opportunitySelection["146A"]
28	Yes	numberOfAntennas[<"LION",6,3>]

■ There is not enough antennas to accommodate all 4 tasks on their time-window on ground station "LION" (3 antennas):

- <134A, 6, 1232, 19, 1266>
- <144, 6, 1238, 31, 1272>
- <146, 6, 1228, 22, 1260>
- <146A, 6, 1230, 22, 1262>



Simplifying model design & problem resolution



From mathematical tools to real applications

