

Feedback on the use of PDDL solvers in an industrial R&D context

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Feedback on the use of **PDDL** solvers in an industrial R&D context

Johann Dreo

Message



Happy planning solvers users



Not happy with:

No on-demand binding between model and action costs

Debugging of domains too hard

Team

- 2007 2013 2020
 - 1) develop solvers
 - 2) applications
- Research & Technology
 - Decision & Optimization lab.,
 - Johann Dreo,
 - Pierre Savéant.

Others

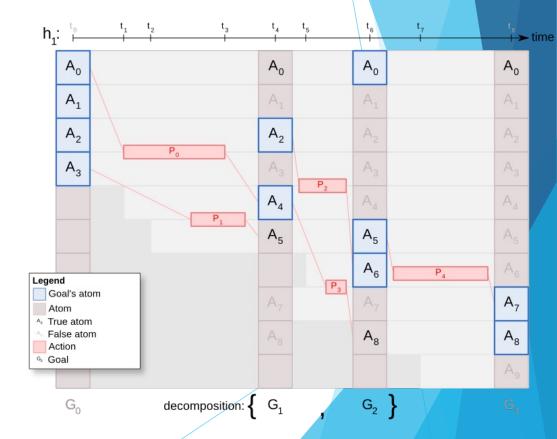
- INRIA : March Schoenauer
- ONERA : Vincent Vidal

- Former students/postdocs
 - Jacques Bibai
 - Caner Candan
 - Mátyás Brendel
 - 🕨 Mostepha R. Khouadjia
 - Alexandre Quemy

Divide-and-Evolve (DAE)

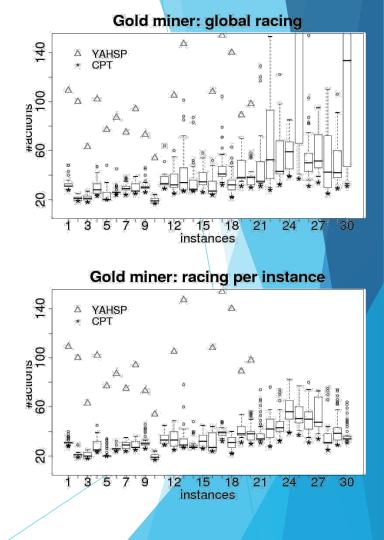
Divide And Evolve

- Black-box heuristic for problem decomposition
- YAHSP for subproblem solving
- 2010 : silver medal at GECCO Hummies awards
 - Competing against human planner
- 2011 : (one of the) winner in temporal satisficing track at IPC



Research: 2007 - 2013

- DAE-CPT
- DAE-YAHSP2
- DAE-YAHSP3
- Automatic configuration of DAE
 - Using racing
 - Using generic optimization
- Multi-objective optimization
 - And PDDL (Zeno) benchmark

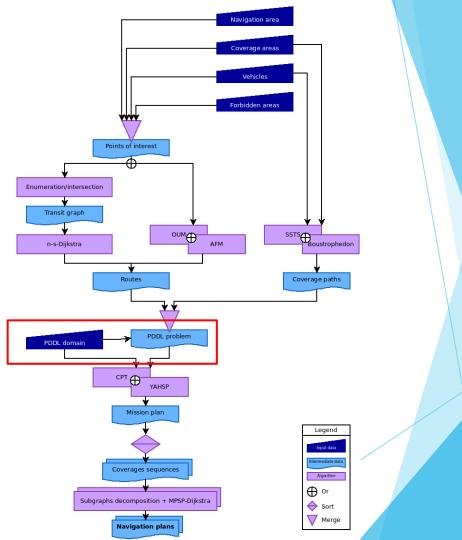


Applications: 2013 - 2020

Multi-vehicles mission planning (~ resources displacement on geographic graphs)

- Emergency vehicles in heavy traffic
- Delivery during resource-scare crisis
- Aerial drone mission planning
- Multi-aerial drones surveillance system
- Underwater uncrewed multi-vehicles mission planning
- Multi-types uncrewed multi-vehicles mission planning
- Always various actions (more than just "move", e.g. scan, pick, …)
- Not always required concurrency

Multi-vehicles planning



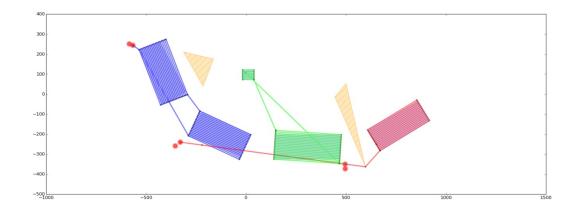
Well known Pro & Cons of Planning Modelling

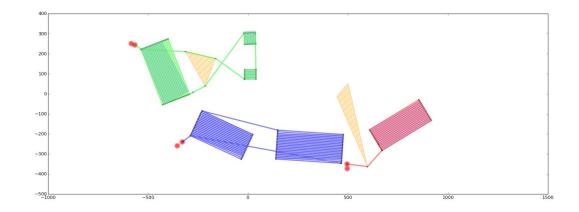
- Advantages
 - Optimal is always better than humans
 - Satisficing is often good enough
 - In practice, fast enough to compute many real-life instance sizes if the model is coarse enough

- Drawbacks
 - Poor correlation in decision space

Drawbacks of planning model

 Poor correlation in decision space





Well known Pro & Cons of Planning Modelling

Advantages

- Optimal is always better than humans
- Satisficing is often good enough
- In practice, fast enough to compute many real-life instance sizes if the model is coarse enough

Drawbacks

- Poor correlation in decision space
- Optimal runtime is unpredictable
- Satisficing quality is unpredictable
- Do not scale well with instance size

User expectations VS reality

Expectations

- Model once, solve many instances
- Model by domain experts, of-the-shelf solvers
- Atomic costs computation is the bottleneck
- Separation of model and solver

Reality

- Problem landscapes diversity is HUGE
- Domain experts is a VERY scarce resource
- Hard to embbed in solvers/models for ondemand computation.
- PDDL itself is not scalable enough

Example PDDL domain

(define (domain domain surveillance v0) (:requirements :typing :durative-actions) (:types Vehicle - Object Type A - Vehicle Type B - Vehicle Location - Object Point Area - Location Start Wait - Point) (:functions ; Time to move from x to y (move time ?x - Location ?y - Location) ; Time to scan area m

(scan time ?m - Area)

)

(:predicates

; Vehicle d is located at location x (located ?d - Vehicle ?x - Location) ; Vehicle d is not already busy (available ?d - Vehicle) ; Sensor activation (sensor_on ?d - Vehicle) (sensor_off ?d - Vehicle) ; Vehicle position relatively to surface. (underwater ?d - Vehicle)

(abovewater ?d - Vehicle)

; Area m is scanned / not scanned

(revealed ?m - Area) ; scanned

(concealed ?m - Area) ; not scanned

; NOTE: At init an goal, we cannot use the "not" operator,

thus we must have two predicates for the state of areas

; in order to specify a switch of status:

areas are concealed in the init and revealed in the goal.

2

;

Example PDDL domain

```
; dive
```

(:durative-action dive

```
:parameters (?d - Vehicle)
```

```
:duration (= ?duration 10)
```

:condition (and

(at start (available ?d)) ; the vehicle is not doing something else (at start (abovewater ?d)) ; the vehicle is above surface (at start (sensor_off ?d)) ; not allowed to use sensor while diving ; NOTE: you cannot use the "not" operator in a condition.

)

)

```
:effect (and
```

(at start (not (available ?d))) ; the vehicle is doing something from the very beginning (over all (not (available ?d))) ; the vehicle is doing something during the action

```
(at end (available ?d) ) ; the vehicle is no more doing something
(at end (underwater ?d) ) ; the vehicle is under water
(at end (not (abovewater ?d))) ; the vehicle is not above surface
```

) ; dive

; [...]

۱.

Example PDDL instance

(define (problem {{NAME}}) (:domain surveillance_v0)

(:objects [...])

(:init

{ { **#VEHICLES** } }

(located {{NAME}} {{TAKEOFF}})

(available {{NAME}})

{{/VEHICLES}}

{ { **#MISSIONS** } }

```
(concealed {{NAME}})
```

{{/MISSIONS}}

{ { **#MOVES** } }

(= (move_time {{FROM}} {{TO}})
{{TIME}})

{ { /MOVES } }

{ { **#MISSIONS** } }

(= (scan_time {{NAME}})
{{SCAN_TIME}})

{{/MISSIONS}}

) ; init

(:goal

(and

```
{ { #MISSIONS } }
```

(revealed {{NAME}})

{{/MISSIONS}}

)

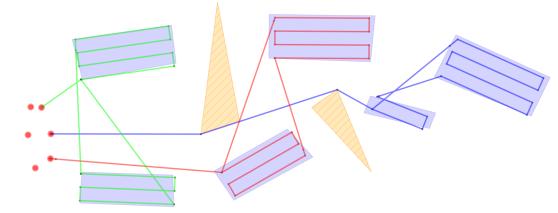
) ; goal

)

(:metric minimize (total-time))

Example PDDL plan

- ; Time 0.024
- ; Length 11
- ; Makespan 41.81
- 0 : (move_drone d1 s1 m3) [11.95]
- 0 : (move_drone d2 s2 m1) [10]
- 0 : (move_drone d0 s0 m2) [7.28]
- 7.28 : (scan_mission d0 m2) [24]
- 10 : (scan_mission d2 m1) [8]
- 11.95 : (scan_mission d1 m3) [32]
- 18 : (move_drone d2 m1 m0) [7.81]





Classical modeller story line

- Start from a close-enough existing PDDL domain
- Rename objects and predicates
- Try to add new actions and constraints
 - Parse error in solver, but error message does not help
 - Try other solvers, various error messages (or none)
 - Random guesses
- Make an instance template
 - Does not scale because cost function is static
 - And pre-computations leads to combinatorial explosion
- Call me to complain

The problem with PDDL

Modelling itself is hard

- Generic modelling problems
- Fidelity/Rapidity compromise
- Scalability of search space
- Modelling in PDDL is even harder
 - (The language syntax is not a problem)
 - (Plan validation is good enough)
 - The lack of some kind of "compiler" is the main problem
 - Using solvers to gain feedback is almost ineffective

Conclusion

Happy with:

- Algorithmics
- Performances
- Modelling features
- Not happy (to the best of our knowledge) with:
 - No lazy binding between model and action costs computation functions
 - Debugging of domains is too hard
 - Ad-hoc model conversion within solvers?

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