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Outline

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►Task planning

- ✓ Statement
- ✓ Difficulty
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Robotics-oriented agent architectures

- ✓ Statement
- ✓ Difficulty
- ✓ Some architectures
- Conclusion & References

≻Conclusion



Part I ----

Task planning



« Given generic operators, a state and goals,
find a sequence of instantiated operators, which lead the initial state to a state which includes the goals. »



> « Planner » : computer software which solve this problem.



➤The crane domain:

✓ 1 crane, *a* locations, *b* trucks, *c* stacks, *d* containers.



If a = 5, b = 3, c = 3, d = 100, then ~ 10^{277} states. Classical planning is in NP.

You cannot enumerate all states!



><u>A 1 :</u> The agent is the sole cause of change.

✓ No other agent, artificial or human.

 \geq <u>A 2 :</u> The environment is totally observable, the agent has perfect knowledge of the environment.

- The agent cannot reason (e.g., plan) on things he does not know.
- ►<u>A 3 :</u> The environment is static.
 - ✓ The environment does not change spontaneously.



- Representation language to define:
 - ✓ a domain : operators
 - \checkmark a problem : initial state and goals.
- >An operator is composed of :
 - Pre-condition : term which must hold for the action to be executable.
 - Effect / post-condition : term the truth value of which is changed by the action, when compared to the incoming situation.

➤A term might be sometimes true, sometimes false, depending on the time at which it is considered.

- Logical operator « not »
 - Example : (not (ON MOUSE PAD))
- ✓ « Fluent » (term)
 - Example : (ON MOUSE PAD)



State: a symbolic description of the agent / environment.

Changes over time, because of the agent actions.

➤Goal : term which has to be satisfied. (= pre-condition).

✓ Might be conjunctive, e.g. : Rich && Handsome && Famous.

Sub-goal : goal obtained by regressing another goal. (= precondition).

Support : post-conditions which unify with a given posterior precondition.

Causal link : relation between a post-condition and a posterior pre-condition, which is satisfied by this post-condition.

E.g. : operator WIN-LOTO et goal (OWN \$1,000,000,000)

➢Mutual exclusion (*mutex*) : 2 pre- or post-conditions which conflict together.

 E. g. : pre-condition (ON MOUSE PAD) and the post-condition (not (ON MOUSE PAD)) in parallel.

Mbat about the table?

?t

➤What about the table? And the arm? What if several arms? What about colored blocks? Or with a notch? Or of different sizes?

puton

Conditionals ? Universal quantification?

Example of PDDL operator: the blocks world

>Operator :

?b

?11

(:action puton :parameters (?b ?u ?t - block) :precondition (and (clear ?b) (on ?b ?u)) (clear ?t)) :effect (and (not (on ?b ?u)) (clear ?u) (on ?b ?t) (not (clear ?t))))



?b

?u



Example : a planning problem in PDDL



(define (problem blocks-24-1) (:domain blocks) (:objects X W V U T S R Q P O N M L K J I H G F E D C A B) (:init (CLEAR K) (CLEAR I) (ONTABLE C) (ONTABLE O) (ON K F) (ON F T) (ON T B) (ON B G) (ON G R) (ON R M) (ON M E) (ON E J) (ON J V) (ON V N) (ON N U) (ON U H) (ON H C) (ON I A) (ON A P) (ON P Q) (ON Q D) (ON D W) (ON W X) (ON X S) (ON S L) (ON L O) (HANDEMPTY))

(:goal (and

(ON L C) (ON C P) (ON P Q) (ON Q M) (ON M B) (ON B G) (ON G F) (ON F K) (ON K E) (ON E R) (ON R A) (ON A W) (ON W T) (ON T N) (ON N J) (ON J U) (ON U S) (ON S D) (ON D H) (ON H V) (ON V O) (ON O I) (ON I X))))

The anomaly of Gerald Jay Sussman (1/16)

with :

puton ?b ?u ?t (clear ?b) (not (on ?b ?u)) (on ?b ?u) (=> (<> ?u table) (clear ?t) (clear ?u)) (on ?b ?t) (=> (<> ?t table) (not (clear ?t)))



The anomaly of Gerald Jay Sussman (2/16)





The anomaly of Gerald Jay Sussman (3/16)









The anomaly of Gerald Jay Sussman (6/16)





The anomaly of Gerald Jay Sussman (7/16)





The anomaly of Gerald Jay Sussman (8/16)





The anomaly of Gerald Jay Sussman (9/16)





The anomaly of Gerald Jay Sussman (10/16)



The anomaly of Gerald Jay Sussman (11/16)



The anomaly of Gerald Jay Sussman (12/16)



The anomaly of Gerald Jay Sussman (13/16)



The anomaly of Gerald Jay Sussman (14/16)





The anomaly of Gerald Jay Sussman (15/16)





The anomaly of Gerald Jay Sussman (16/16)







Algorithms (1/3) : intuitively ...



Plan = (Templates *T*, Operators *Op*, PartialOrder *O*, Unification *U*).
 Definitions :

- Conflict : a post-condition might destroy a causal link (threat).
- ✓ Satisfying a pre-condition p: adding an operator before p, which a post-condition q unifies with p (*i.e.*, q = p).

▶ PLANNER1(T, Op., O, U) :

WHILE(conflict || at least 1 unsatisfied pre-condition)

- 1. Solve conflicts :
 - Add a unification / non unification constraint;
 - Add a precedence constraint.
- 2. Choose an unsatisfied pre-condition p;
- 3. Satisfy p :
 - Add a unification / non-unification constraint;
 - Add a precedence constraint ;
 - Add an operator.



Algorithms (2/3) : in the state space



A state S_i = a state of the environment.

- ✓ S_0 = initial state, composed of the fluents $f_1, ..., f_n$.
- ✓ Example : a given configuration of blocks.

Successors(S_i) = n states S_{i+1} , ..., S_{i+n} which can be reached by an instantiated operator applicable in S_i . Solution(S_i) iff S_i includes the goals g_1 , ..., g_i .

► PLANNER2(Alg., S₀, Successors, Solution) :

- ✓ Alg. = any state-space search algorithm.
- ✓ Heuristics ?

Example :

- ✓ Algorithm A* of *Heuristic Search Planner* (HSP) [Bonet 98]
- ✓ Heuristics: a graph plan without negative pre-conditions.



>A state = a partially-ordered partially instantiated plan.

▶ PLANNER3(T, Op., O, U) :

Search algorithm in a space of states (e.g., A*) :

- A state = a partial plan (T, Op., O', U').
- A successor = obtained by solving a conflict, or satisfying a pre-condition.
 - Adding a unification / non-unification constraint to U
 - Adding a precedence constraint to O
 - Adding an operator from T to Op
- A solution function = no conflict && all pre-conditions are satisfied.

≻Heuristics?

Example:

- Universally quantified Conditional Partial-Order Planner (UCPOP) [Penberthy 92]
- http://www.cs.washington.edu/homes/weld/ucpop.html



Levels include mutual exclusions (*mutex*): a level is not a state!

Forward development of a plan of graph, backward search.

Hierachical Tasks Network (HTN)

Additional knowledge:

✓ A task can be decomposed into sub-tasks.



Search algorithm by refining plans.

✓ Simple Task Network (STN).

≻Example :

- ✓ The HTN planner used by Jason Wolfe at Willow Garage (CA).
- ✓ SHOP and SHOP2, by Dana Nau from Univ. Maryland.

SAT-based planning [Kautz 92]

≻The SAT problem:

- Find a truth value for each proposition, which together satisfy a given logical formula, using AND, OR and NOT.
- Proposition logics.
- ✓ 1st problem which has been proved NP-complete (1971).
- ✓ International Conference SAT'12. http://www.satisfiability.org/

>Principle of SAT-based planning:

- 1. n = 1 // Length of the solution-plan.
- 2. Turn a plan of length n into a formula in proposition logic.
- **3**. Attempt to prove this formula using a SAT solver.
- 4. IF it fails, THEN (i) increment *n*, (ii) GOTO 2.

Example of formulas, in the blocks world :

- ✓ For all x, y, z, i:
 - on(x, y, i) && clear(x, i) && clear(z, i) && puton(x, y, z, i) => clear(y, i+1) && on(x, z, i+1)
- ✓ For all x, x', y, y', z, z', i:
 - x <> x' && y <> y' && z <> z' => not puton(x, y, z, i) || not puton(x', y', z', i)





>Constraint programming:

- <u>Statement:</u> for each variable, search for a value from the variable's domain, so that all values satisfy the constraints.
- ✓ <u>Representation</u>: Variables / Domains / Constraints.
- ✓ <u>Algorithm:</u>
 - Choose a variable
 - Choose a value from the variable's domain
 - Propagate this assignment through constraints
 - If a domain becomes empty, backtrack on previous choices.

Principle of CSP-based planning:

- **1.** Heuristically estimate the length *n* of a solution-plan;
- 2. Turn the planning problem into a dynamic CSP;
- **3.** Attempt at finding a solution plan of length *n* using a CSP solver;
- **4.** IF failure, THEN increment n ; GOTO 2.

Examples :

- ✓ IxTeT planner from LAAS-CNRS in Toulouse [Laborie 95].
 - http://spiderman-2.laas.fr/RIA/lxTeT/ixtet-planner.html
- Constraint Programming Temporal planner (CPT) from ONERA Toulouse [Vidal 06].
 - http://v.vidal.free.fr/onera/#cpt

Conditional planning (1 / 2)



- >The agent may not know the output of its own actions.
- Plans have branches:
 - ✓ IF < test > THEN Plan_A ELSE Plan_B
 - ✓ Obtain a plan in every case
- ➢ Full observability: the agent knows its state.
 - ✓ No need for an operator « observe ».
- >Actions might fail: disjunctive effects.
- ≻Conditional effects.
 - ✓ EFFECTS : IF To_Left, CleanL ; IF To_Right, CleanD.
- ➤Conditional planning is harder than NP.
- Example (double Murphy) : a vacuum-cleaner agent must cleam all rooms.
 - <u>Rule 1:</u> the vacuum cleaner sometimes drops dust when it moves to a clean room.
 - <u>Rule 2</u>: the vacuum cleaner sometimes drops dust if CLEAN is executed in a clean room.



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History



- >1971: STRIPS from Richard Fikes.
- ▶1977: NOAH from Earl Sacerdoti
- ▶1981: MOLGEN from Mark Stefik
- ▶1986: IxTeT from Malik Ghallab.
- ▶1986: SIPE from David Wilkins.
- >1987: TWEAK from David Chapman.
- ➤1991: SNLP from Mac Allister & Rosenblitt.
- >1992: UCPOP from Anthony Barrett & Daniel Weld.
- >1992: BLACKBOX/SATPLAN from Henry Kautz & Bart Selman.
- >1997: GRAPHPLAN from Avrim Blum & Merrick Furst.
- >2000: HSP from Hector Geffner.
- >2000: YAHSP from Vincent Vidal.
- >2001: FF from Jörg Hoffmann,
- ≥2005: CPT from Vincent Vidal.
- ▶2007: DAE from Marc Schoenauer.

References (1 / 2) - Part I



[Weld 94] Daniel Weld, An Introduction to Least Commitment Planning, A.
 I. Magazine, 15(4), pages 27-61, Winter 1994.

[Russel 2010] Stuart Russell, Peter Norvig. Artificial Intelligence: A Modern Approach. Prentice Hall, 2010, 3rd edition. Chapitre 10.

 [Ghallab et al. 04] Malik Ghallab, Dana Nau, Paolo Traverso. Automated Planning: Theory and Practice. Morgan Kaufmann, San Mateo, CA, May 04, 635 pages.

▶ PDDL 3.1.

http://en.wikipedia.org/wiki/Planning_Domain_Definition_Language

Conférences :

- International Conference on Automated Planning and Scheduling (ICAPS). http://www.icaps.org
- ✓ International Joint Conference on A.I. (IJCAI). http://www.ijcai.org
- European Conference on A.I. (ECAI). http://www.ecai.org
- ✓ National Conference on A.I. (AAAI). http://www.aaai.org

► Journals :

✓ A. I. Journal (AIJ).

http://www.elsevier.com/wps/find/journaldescription.cws_home/505601/description#description

✓ Journal of A.I. Research (JAIR). http://www.jair.org/

References (2 / 2) – Part II



- [Blum 97] A. Blum, M. Furst. Fast Planning through Planning Graph Analysis. Artificial Intelligence, 90:281-300, 1997.
- [Bonet 98] B. Bonet, H. Geffner. HSP: Heuristic Search Planner. In Proceedings of Artificial Intelligence Planning Systems (AIPS), 1998.
- [Kautz 92] H. Kautz, B. Selman. Planning as Satisfiability. In Proceedings of ECAI'92.
- [Laborie 95] P. Laborie, M. Ghallab. Planning with Sharable Resource Constraints. In Proceedings of the International Joint Conference on Artificial Intelligence (IJCAI), 1995, pages 1643 – 1651.
- [Penberthy 92] J. S. Penberthy, D. Weld. UCPOP: A Sound, Complete, Partial-Order Planner for ADL. In Proceedings of 3rd International Conference on Knowledge Representation and Reasoning (KR'92), Cambridge, MA, 1992.
- [Vidal 06] V. Vidal, H. Geffner. Branching and Pruning: An Optimal Temporal POCL Planner based on Constraint Programming. Artificial Intelligence, 170(3): 298-335, 2006.

Conclusion of part I (1/2)

- Domain explored for 40 years
- Some planners now are available:
 - ✓ CPT, FF, SATPLAN, ...
 - ✓ International Planning Competition (IPC).
 - http://ipc.icaps-conference.org/
- Conditional planning: more difficult...
- ➢ Properties of a planner:
 - ✓ Correctness
 - ✓ Completeness
 - Optimality
 - Canonicity
 - ✓ Efficiency

≻Hint:

 Merge probabilistic planning (MDP) and symbolic planning (STRIPS).





Demo of the CPT task planner...



≻When to plan?

✓ Before executing (off-line planning).

✓ While executing (on-line planning).



Part II ----

Robotics-oriented agent architectures

Statement



An agent is a system including reasoning (e.g., temporal), perceiving its environment, acting on it and interacting with other agents (artificial or human). »





Examples of robotic agents





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Architecture Sense-Plan-Act [Nilsson 80]





No symbol [Brooks 91].



2-level architecture [Hayes-Roth et al. 95] (1/3)













2-level++ architecture [Baltié et al. 07]



3-level architecture [Gat 98]



LAAS-CNRS architecture [Alami et al. 98] **Robotic agent** Procedural Action Reasoning planning Deliberative System (*PRS*) (IxTeT) Functional Executive Behavior 1 Behavior *n* Sensors Actuators **Environment**

Architectures for robots



>Open RObotic COntrol Software, from H. Bruyninckx (Belgium).

✓ Ontology, no task planning.

Robotic Operating System (ROS), from Willow Garage (CA).
 An HTN planner [Wolfe 10], no ontology.



Conclusion of Part II



Domain explored since [Nilsson 80], i.e., ~30 years.
 No unique architecture makes consensus!
 Critical properties:

- ✓ Real-time
 - How to get immediately a good reaction?
- ✓ Safety
 - How to get a good reaction in the worst case?
- ≻Hints:
 - ✓ Ontology to analyze the context.
 - Multi-Agent Systems : intelligence emerges from interaction among agents.

References (1 / 2) – Part II



▶ [Alami et al. 98] R. Alami, R. Chatila, S. Fleury, M. Ghallab, F. Ingrand. *An Architecture for Autonomy*. In *International Journal of Robotics Research* (Special Issue on ``Integrated Architectures for Robot Control and Programming"), Vol 17, N° 4, April 1998. LAAS Report N°97352.

▶ [Baille et al. 99] Gérard Baille & al, *Le CyCab de l'INRIA Rhône-Alpes*. Rapport de recherche de l'INRIA Rhône-Alpes n°0229, April 1999 (in French).

[Baltie et al. 07] J. Baltié, E. Bensana, P. Fabiani, J. – L. Farges, S. Millet, P. Morignot, B. Patin, G. Petitjean, G. Pitois, J. – C. Poncet. *Multi-Vehicle Missions: Architecture and Algorithms for Distributed On Line Planning*. In Dimitri Vrakas and Ioannis Vlahavas (eds.), Artificial Intelligence for Advanced Problem Solving Techniques, Information Science Reference. December 2007.

▶ [Beetz et al. 10] M. Beetz, D. Jain, L. Mösenlechner, M. Tenorth. Towards Performing Everyday Manipulation Activities. Robotics and Autonomous Systems, April 2010.

[Brooks 85] Brooks, R. A. A Robust Layered Control System for a Mobile Robot. In IEEE Journal of Robotics and Automation, Vol. 2, No. 1, March 1986, pp. 14–23.

▶ [Brooks 91] R. Brooks. *Intelligence without reason*. Proceedings of 12th Int. Joint Conf. on Artificial Intelligence (IJCAI'91), Sydney, Australia, August 1991, pp. 569–595.

[Campa et al. 96] Giampiero Campa, Mario Innocenti, Jacqueline Wilkie. Model-Based Robust Control for a Towed Underwater Vehicle. AIAA Guidance, Navigation and Control Conference, San Diego, California, July 29-31, 1996.

► [Gat 98] Gat, E. *Three-layer architectures*. In D. Kortenkamp et al. Eds. A.I. and mobile robots. AAAI Press, 1998.

▶ [Hayes-Roth et al. 95] Hayes-Roth, B.; Pfleger, K.; Morignot, P.; & Lalanda, P. *Plans and Behavior in Intelligent Agents*. Knowledge Systems Laboratory, KSL-95-35, Stanford Univ., CA, March, 1995.

References (2 / 2) - Part II



> [Nilsson 80] Nils J. Nilsson. *Principles of ArtificialIntelligence*. Palo Alto: Tioga. 1980.

▶ [Muscettola et al. 98] N. Muscettota, P. Pandurag Nayak, Barney Pell and B.C. Williams. *Remote Agent: to Boldly go where no AI System Has Gone Before. Artificial Intelligence*, Elsevier, 103, pp.5-47, 1998.

▶ [Russel 2010] Stuart Russell, Peter Norvig. *Artificial Intelligence: A Modern Approach*. Prentice Hall, 2010, 3rd edition. Chapitre 11.

▶ [Schoppers 95] Schoppers, M. The use of dynamics in an intelligent controller for a space faring rescue robot. In Artificial Intelligence Journal, 73 (1995):175-230.

▶ [Teichteil et al. 11] F. Teichteil-Königsburg, C. Lesire, G. Infantes. A Generic Framework for Anytime Execution-Driven Planning in Robotics. In Proceedings of the International Conference on Robotics and Automation, Shanghai, China, May 2011, pages 299-304.

▶ [Wolfe et al. 10] J. Wolfe, B. Marthi, S. Russell. *Combining Task and Motion Planning for Mobile Manipulation*. In Proceedings of the International Conference on Automated Planning and Scheduling, Toronto, Canada, 2010.





$$f(x) = e^x$$

If P is different than NP, then we are fighting against the exponential function in the worst case!



Thank you for your attention!