

Artificial Intelligence

Roman Barták

Department of Theoretical Computer Science and Mathematical Logic

Automated Planning

Today we will explore techniques for **action planning** – how to find a sequence of actions to reach a given goal.

- **problem representation**

- situation calculus (pure logical representation)
- using sets of predicates (instead of formulas)
- planning domain vs. planning problem

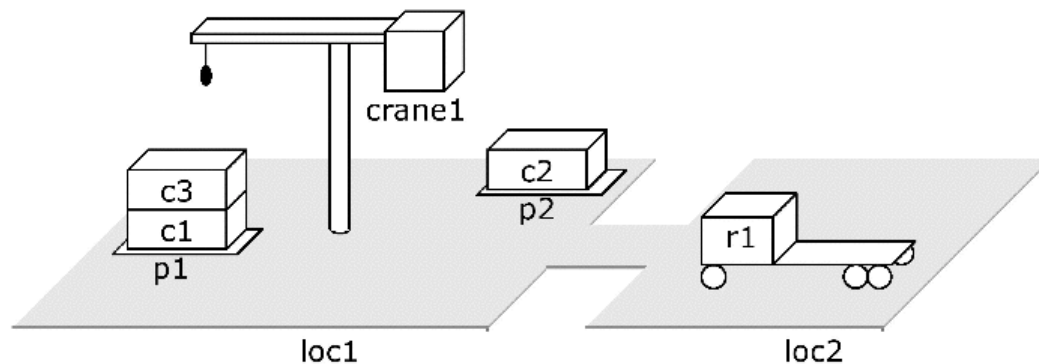
- **planning techniques**

- state-space planning
 - forward and backward
- plan-space planning
 - partially ordered plans



We can simplify the full FOL model into a so called **classical representation** of planning problems.

State is a set of instantiated atoms (no variables). There is a finite number of states!



{attached(p1,loc1), in(c1,p1), in(c3,p1), top(c3,p1), on(c3,c1), on(c1,pallet), attached(p2,loc1), in(c2,p2), top(c2,p2), on(c2,pallet), belong(crane1,loc1), empty(crane1), adjacent(loc1,loc2), adjacent(loc2,loc1), at(r1,loc2), occupied(loc2), unloaded(r1)}.

– The truth value of some atoms is changing in states:

- **fluents**

- *example: at(r1,loc2)*

– The truth value of some state is the same in all states

- **rigid atoms**

- *example: adjacent(loc1,loc2)*

We will use a classical **closed world assumption**.

An atom, that is not included in the state, does not hold at that state!

operator o is a triple $(\text{name}(o), \text{precond}(o), \text{effects}(o))$

– **name(o): name of the operator** in the form $n(x_1, \dots, x_k)$

- n : a symbol of the operator (a unique name for each operator)
- x_1, \dots, x_k : symbols for variables (operator parameters)
 - Must contain all variables appearing in the operator definition!

– **precond(o):**

- literals that must hold in the state so the operator is applicable on it

– **effects(o):**

- literals that will become true after operator application (only fluents can be there!)

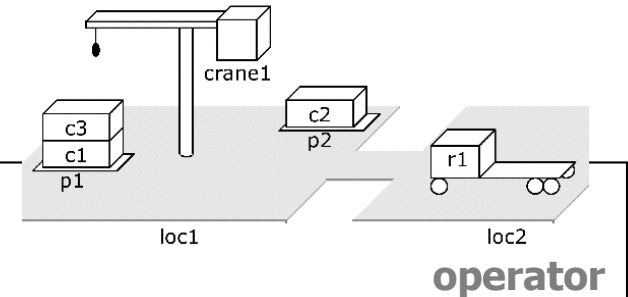
`take(k, l, c, d, p)`

`:: crane k at location l takes c off of d in pile p`

`precond: belong(k, l), attached(p, l), empty(k), top(c, p), on(c, d)`

`effects: holding(k, c), \neg empty(k), \neg in(c, p), \neg top(c, p), \neg on(c, d), top(d, p)`

An action is a fully instantiated operator – substitute constants to variables



`take(k, l, c, d, p)`

`:: crane k at location l takes c off of d in pile p`

`precond: belong(k, l), attached(p, l), empty(k), top(c, p), on(c, d)`

`effects: holding(k, c), \neg empty(k), \neg in(c, p), \neg top(c, p), \neg on(c, d), top(d, p)`

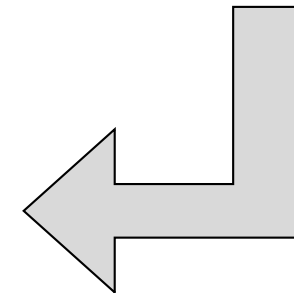
`take(crane1,loc1,c3,c1,p1)`

action

`:: crane crane1 at location loc1 takes c3 off c1 in pile p1`

`precond: belong(crane1,loc1), attached(p1,loc1),
empty(crane1), top(c3,p1), on(c3,c1)`

`effects: holding(crane1,c3), \neg empty(crane1), \neg in(c3,p1),
 \neg top(c3,p1), \neg on(c3,c1), top(c1,p1)`



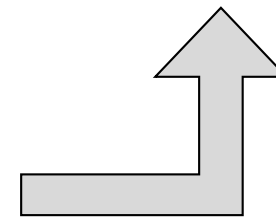
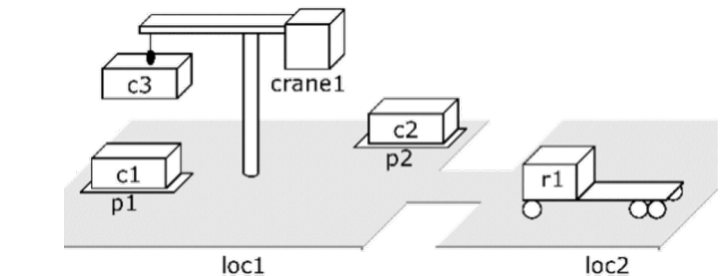
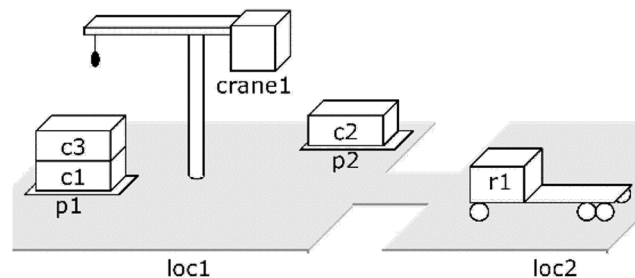
Notation (let S be a set of literals):

- $S^+ = \{\text{positive atoms in } S\}$
- $S^- = \{\text{atoms, whose negation is in } S\}$

Action a is **applicable** to state s if and only if
 $\text{precond}^+(a) \subseteq s \wedge \text{precond}^-(a) \cap s = \emptyset$

The result of application of action a to s is
 $\gamma(s,a) = (s - \text{effects}^-(a)) \cup \text{effects}^+(a)$

```
take(crane1,loc1,c3,c1,p1)
;; crane crane1 at location loc1 takes c3 off c1 in pile p1
precond: belong(crane1,loc1), attached(p1,loc1),
         empty(crane1), top(c3,p1), on(c3,c1)
effects:  holding(crane1,c3),  $\neg$ empty(crane1),  $\neg$ in(c3,p1),
          $\neg$ top(c3,p1),  $\neg$ on(c3,c1), top(c1,p1)
```



Let L be a language and O be a set of operators.

Planning domain Σ over language L with operators O is a triple (S, A, γ) :

- **states** $S \subseteq P(\{\text{all instantiated atoms from } L\})$
- **actions** $A = \{\text{all instantiated operators from } O \text{ over } L\}$
 - action \mathbf{a} is **applicable** to state \mathbf{s} if
 $\text{precond}^+(\mathbf{a}) \subseteq \mathbf{s} \wedge \text{precond}^-(\mathbf{a}) \cap \mathbf{s} = \emptyset$
- **transition function** γ :
 - $\gamma(\mathbf{s}, \mathbf{a}) = (\mathbf{s} - \text{effects}^-(\mathbf{a})) \cup \text{effects}^+(\mathbf{a})$, if \mathbf{a} is applicable on \mathbf{s}
 - S is closed with respect to γ (if $\mathbf{s} \in S$, then for every action \mathbf{a} applicable to \mathbf{s} it holds $\gamma(\mathbf{s}, \mathbf{a}) \in S$)

Planning problem P is a triple (Σ, s_0, g) :

- $\Sigma = (S, A, \gamma)$ is a planning domain
- s_0 is an initial state, $s_0 \in S$
- g is a set of instantiated literals
 - state s satisfies the goal condition g if and only if $g^+ \subseteq s \wedge g^- \cap s = \emptyset$
 - $S_g = \{s \in S \mid s \text{ satisfies } g\}$ – a set of goal states

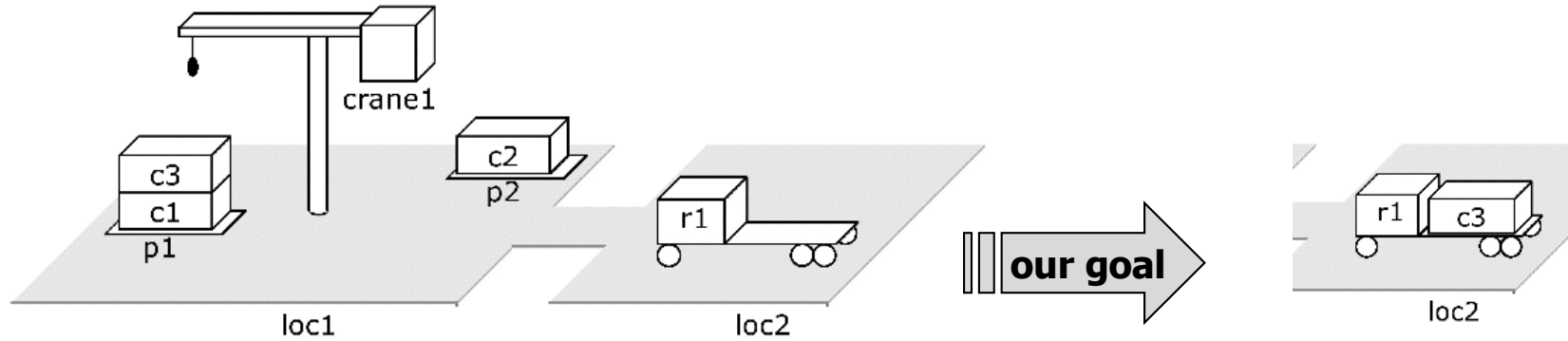
Plan is a sequence of actions $\langle a_1, a_2, \dots, a_k \rangle$.

Plan $\pi = \langle a_1, a_2, \dots, a_k \rangle$ is a **solution plan** for problem P iff $\gamma^*(s_0, \pi)$ satisfies the goal condition g .

Usually the planning problem is given by a triple (O, s_0, g) .

- O defines the the operators and predicates used (domain model)
- s_0 provides the particular constants (objects)

Classical representation: example plan

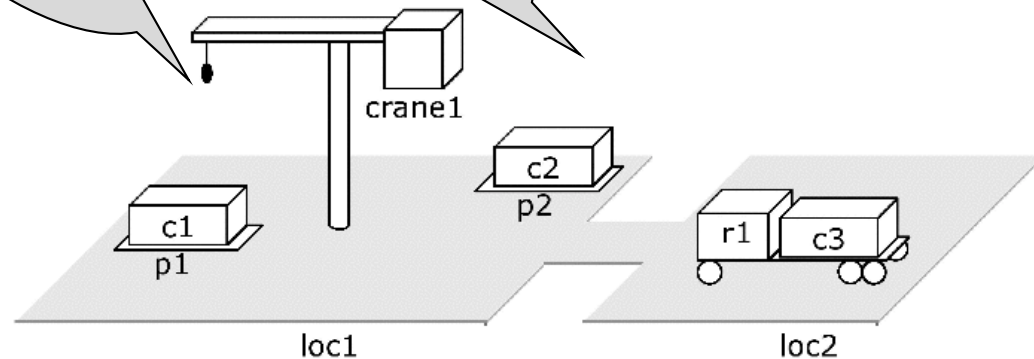


$s_1 = \{ \text{attached}(p1, loc1), \text{in}(c1, p1), \text{in}(c3, p1), \text{top}(c3, p1), \text{on}(c3, c1), \text{on}(c1, \text{pallet}), \text{attached}(p2, loc1), \text{in}(c2, p2), \text{top}(c2, p2), \text{on}(c2, \text{pallet}), \text{belong}(\text{crane1}, loc1), \text{empty}(\text{crane1}), \text{adjacent}(loc1, loc2), \text{adjacent}(loc2, loc1), \text{at}(r1, loc2), \text{occupied}(loc2), \text{unloaded}(r1) \}.$

$g = \{ \text{loaded}(r1, c3), \text{at}(r1, loc2) \}$

$\langle \text{move}(r1, loc2, loc1), \text{take}(\text{crane1}, loc1, c3, c1, p1), \text{load}(\text{crane1}, loc1, c3, r1), \text{move}(r1, loc1, loc2) \rangle$

$\langle \text{take}(\text{crane1}, loc1, c3, c1, p1), \text{move}(r1, loc2, loc1), \text{load}(\text{crane1}, loc1, c3, r1), \text{move}(r1, loc1, loc2) \rangle$



The search space corresponds to the state space of the planning problem.

- search nodes correspond to world states
- arcs correspond to state transitions by means of actions
- the task is to find a path from the initial state to some goal state

Basic approaches

- forward search (progression)
 - start in the initial state and apply actions until reaching a goal state
- backward search (regression)
 - start with the goal and apply actions in the reverse order until a subgoal satisfying the initial state is reached
 - lifting (actions are only partially instantiated)

Forward-search(O, s_0, g)

$s \leftarrow s_0$

$\pi \leftarrow$ the empty plan

loop

if s satisfies g then return π

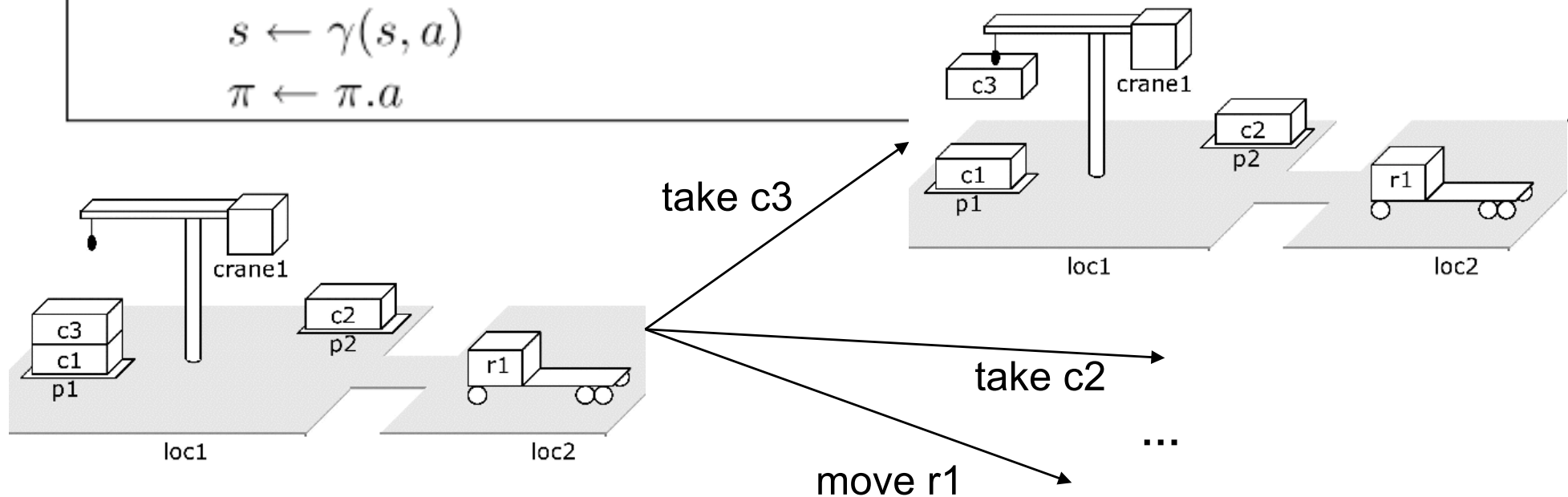
$E \leftarrow \{a \mid a \text{ is a ground instance of an operator in } O,$
and $\text{precond}(a)$ is true in $s\}$

if $E = \emptyset$ then return failure

nondeterministically choose an action $a \in E$

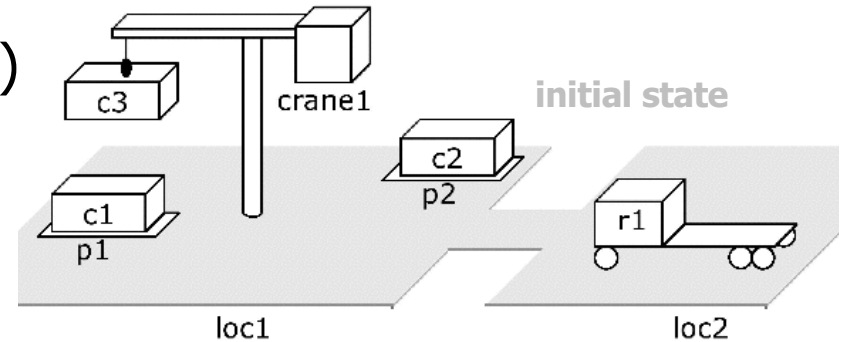
$s \leftarrow \gamma(s, a)$

$\pi \leftarrow \pi.a$



Forward planning: an example

{belong(crane1,loc1), adjacent(loc2,loc1)
holding(crane1,c3), unloaded(r1),
at(r1,loc2), \neg occupied(loc1),
occupied(loc2),...}



move(r1,loc2,loc1)

move(r, l, m)
;; robot r moves from location l to location m
precond: adjacent(l, m), at(r, l), \neg occupied(m)
effects: at(r, m), occupied(m), \neg occupied(l), \neg at(r, l)

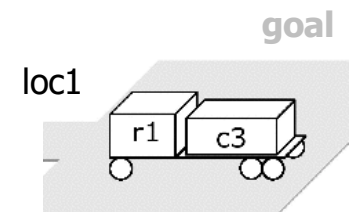
{belong(crane1,loc1),
adjacent(loc2,loc1), holding(crane1,c3), unloaded(r1),
at(r1,loc1), occupied(loc1), ...}

load(crane1,loc1,c3,r1)

load(k, l, c, r)
;; crane k at location l loads container c onto robot r
precond: belong(k, l), holding(k, c), at(r, l), unloaded(r)
effects: empty(k), \neg holding(k, c), loaded(r, c), \neg unloaded(r)

{belong(crane1,loc1), adjacent(loc2,loc1),
empty(crane1), loaded(r1,c3),
at(r1,loc1), occupied(loc1), ...}

Goal = {at(r1,loc1),loaded(r1,c3)}



Start with a goal (not a goal state as there might be more goal states) and through sub-goals try to reach the initial state.

Action a is relevant for a goal g if and only if:

- action a contributes to goal g: $g \cap \text{effects}(a) \neq \emptyset$
- effects of action a are not conflicting goal g:
 - $g^- \cap \text{effects}^+(a) = \emptyset$
 - $g^+ \cap \text{effects}^-(a) = \emptyset$

A regression set of the goal g for (relevant) action a is
 $\gamma^{-1}(g,a) = (g - \text{effects}(a)) \cup \text{precond}(a)$

Example:

goal: **{on(a,b), on(b,c)}**

action **stack(a,b)** is relevant

by backward application of the action we get a new goal:

{holding(a), clear(b), on(b,c)}

stack(x,y)

Precond: holding(x), clear(y)

Effects: \neg holding(x), \neg clear(y),

on(x,y), clear(x), handempty

Backward planning: algorithm

Backward-search(O, s_0, g)

$\pi \leftarrow$ the empty plan

loop

if s_0 satisfies g then return π

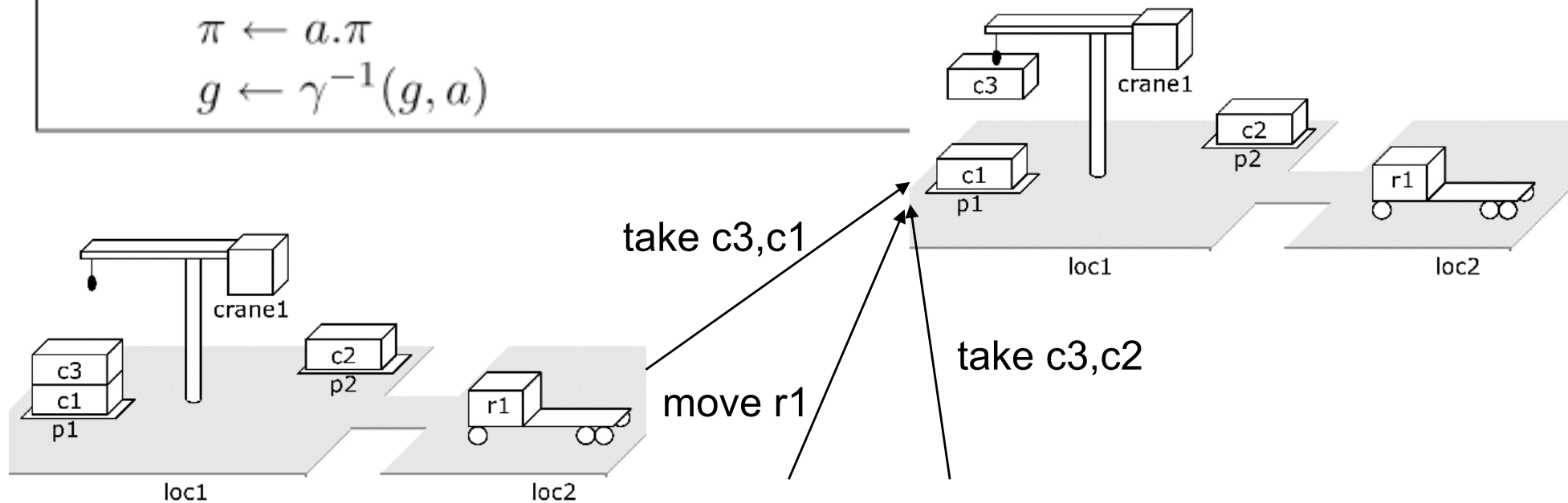
$A \leftarrow \{a \mid a \text{ is a ground instance of an operator in } O$
and $\gamma^{-1}(g, a)$ is defined}

if $A = \emptyset$ then return failure

nondeterministically choose an action $a \in A$

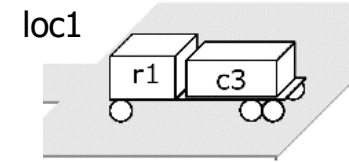
$\pi \leftarrow a.\pi$

$g \leftarrow \gamma^{-1}(g, a)$



Backward planning: an example

Goal = {at(r1,loc1),loaded(r1,c3)}



load(crane1,loc1,c3,r1)

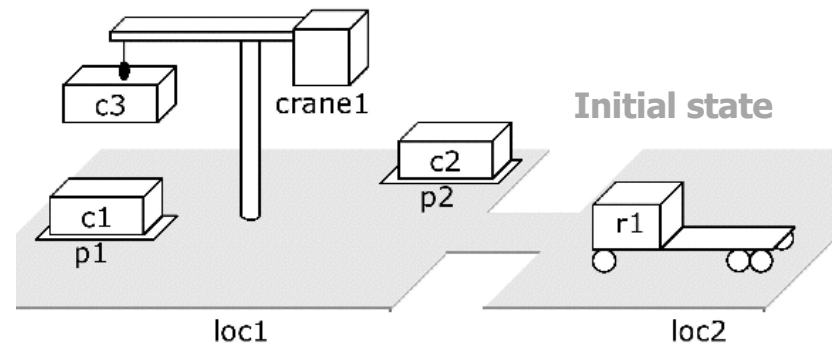
load(k, l, c, r)
;; crane k at location l loads container c onto robot r
precond: belong(k, l), holding(k, c), at(r, l), unloaded(r)
effects: empty(k), \neg holding(k, c), loaded(r, c), \neg unloaded(r)

{at(r1,loc1), belong(crane1,loc1),
holding(crane1,c3), unloaded(r1)}

move(r1,loc2,loc1)

move(r, l, m)
;; robot r moves from location l to location m
precond: adjacent(l, m), at(r, l), \neg occupied(m)
effects: at(r, m), occupied(m), \neg occupied(l), \neg at(r, l)

{belong(crane1,loc1), holding(crane1,c3),
unloaded(r1),
adjacent(loc2,loc1),
at(r1,loc2),
 \neg occupied(loc1)}



Lifted-backward-search(O, s_0, g)

$\pi \leftarrow$ the empty plan

loop

if s_0 satisfies g then return π

$A \leftarrow \{(o, \theta) \mid o \text{ is a standardization of an operator in } O,$
 $\theta \text{ is an mgu for an atom of } g \text{ and an atom of effects } (o),$
 $\text{and } \gamma^{-1}(\theta(g), \theta(o)) \text{ is defined}\}$

if $A = \emptyset$ then return failure

nondeterministically choose a pair $(o, \theta) \in A$

$\pi \leftarrow$ the concatenation of $\theta(o)$ and $\theta(\pi)$

$g \leftarrow \gamma^{-1}(\theta(g), \theta(o))$

Notes:

- standardization = a copy with fresh variables
- mgu = most general unifier
- by using the variables we can decrease the branching factor but the trade off is more complicated loop check

The principle of plan space planning is similar to backward planning:

- start from an **“empty” plan** containing just the description of initial state and goal
- **add other actions** to satisfy not yet covered (open) goals
- if necessary **add other relations** between actions in the plan

Planning is realised as **repairing flaws in a partial plan**

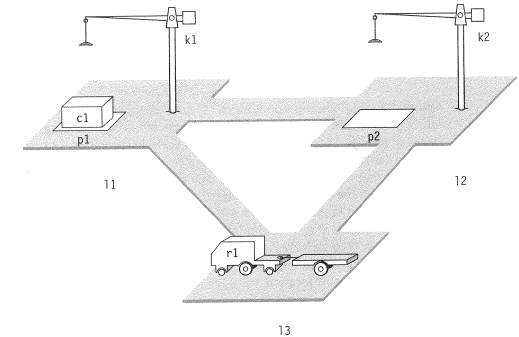
- go from one partial plan to another partial plan until a complete plan is found

Assume a partial plan with the following two actions:

- take(k1,c1,p1,l1)
- load(k1,c1,r1,l1)

Possible modifications of the plan:

- **adding a new action**
 - to apply action **load**, robot r1 must be at location l1
 - action **move**(r1,l,l1) moves robot r1 to location l1 from some location l
- **binding the variables**
 - action **move** is used for the right robot and the right location
- **ordering some actions**
 - the robot must **move** to the location before the action **load** can be used
 - the order with respect to action **take** is not relevant
- **adding a causal relation**
 - new action is added to move the robot to a given location that is a precondition of another action
 - the causal relation between **move** and **load** ensures that no other action between them moves the robot to another location



The initial state and the goal are encoded using two **special actions** in the initial partial plan:

- **Action a_0 represents the initial state** in such a way that atoms from the initial state define effects of the action and there are no preconditions. This action will be before all other actions in the partial plan.
- **Action a_∞ represents the goal** in a similar way – atoms from the goal define the precondition of that action and there is no effect. This action will be after all other actions.

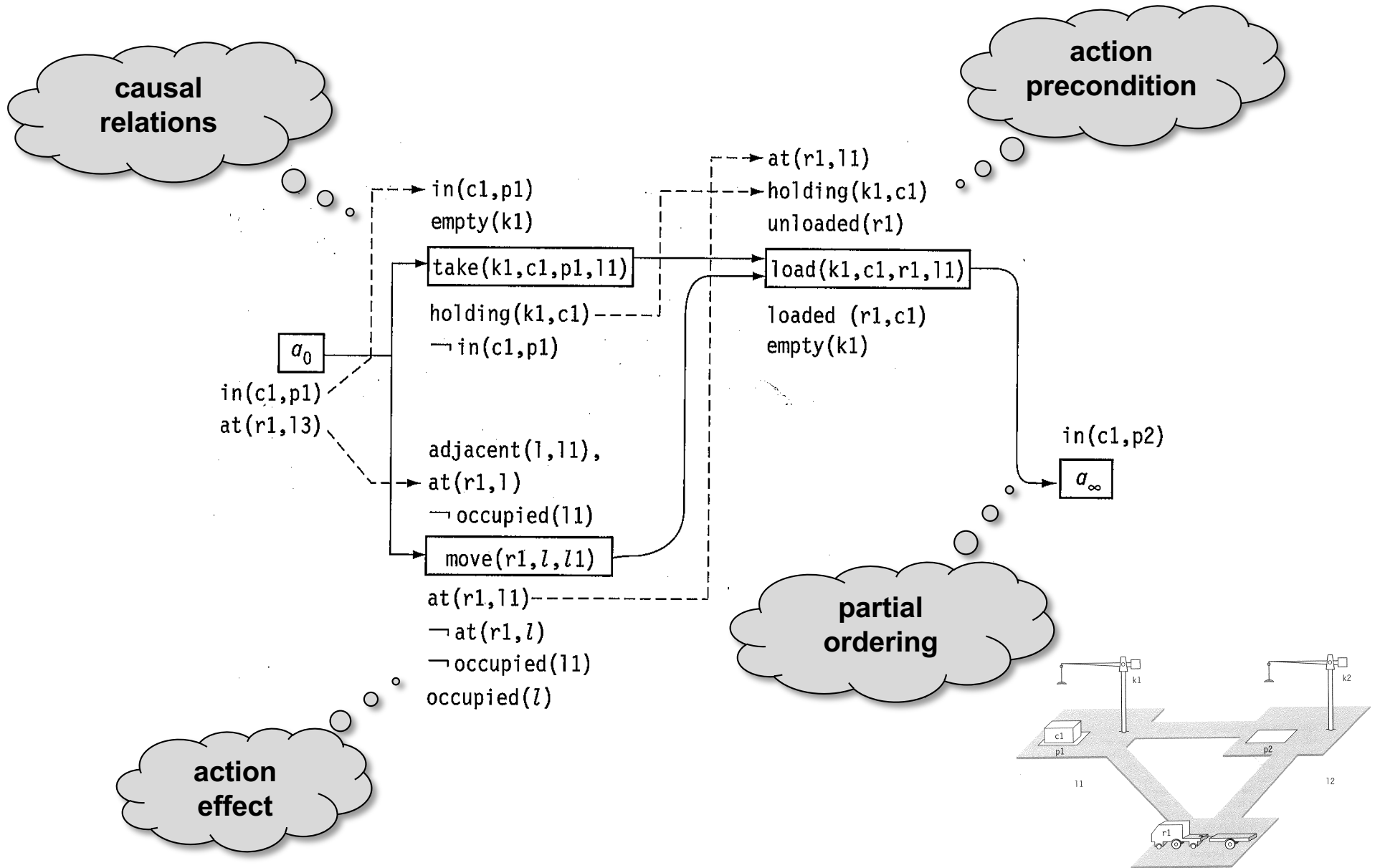
Planning is realised by **repairing flaws** in the partial plan.

The search nodes correspond to partial plans.

A partial plan Π is a tuple $(A, <, B, L)$, where

- A is a set of partially instantiated planning operators $\{a_1, \dots, a_k\}$
- $<$ is a partial order on A ($a_i < a_j$)
- B is set of constraints in the form $x=y$, $x \neq y$ or $x \in D_i$
- L is a set of causal relations $(a_i \rightarrow^p a_j)$
 - a_i, a_j are ordered actions $a_i < a_j$
 - p is a literal that is effect of a_i and precondition of a_j
 - B contains relations that bind the corresponding variables in p

Partial plan: an example



Open goal is an example of a **flaw**.

This is a precondition **p** of some operator **b** in the partial plan such that no action was decided to satisfy this precondition (there is no causal relation $a_i \rightarrow^p b$).

The open goal p of action b can be resolved by:

- finding an operator **a** (either present in the partial plan or a new one) that can give **p** (**p** is among the effects of **a** and **a** can be before **b**)
- binding the variables from **p**
- adding a causal relation $a \rightarrow^p b$

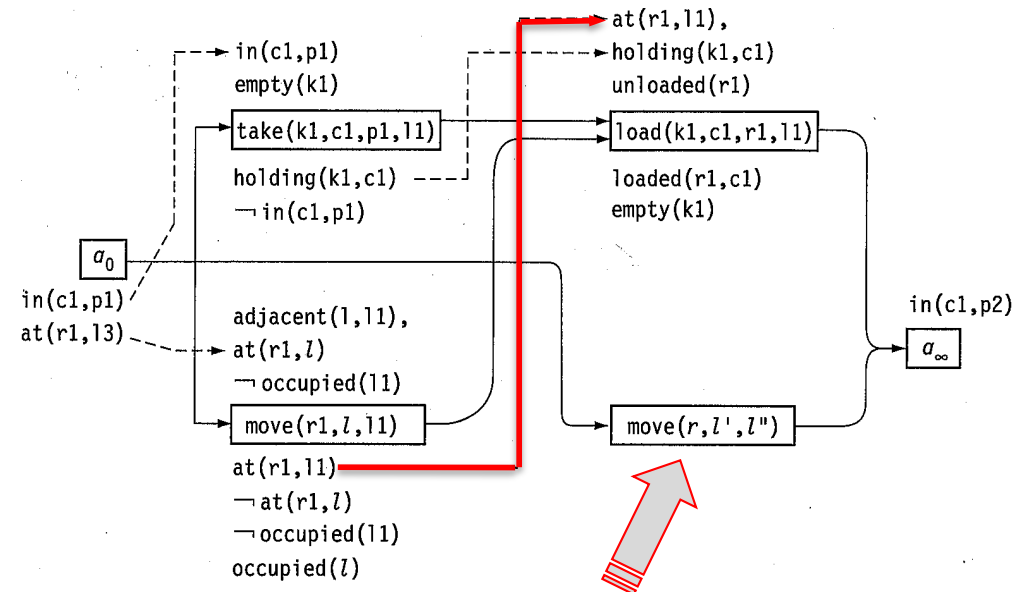
Threat is another example of **flaw**.

It is an action that can influence existing causal relation.

- Let $a_i \rightarrow^p a_j$ be a causal relation and action **b** has among its effects a literal unifiable with the negation of **p** and action **b** can be between actions a_i and a_j . Then **b** is threat for that causal relation.

We can **remove the threat** by one of the following ways:

- ordering **b** before a_i
- ordering **b** after a_j
- binding variables in **b** in such a way that **p** does not bind with the negation of **p**



Partial plan $\Pi = (A, <, B, L)$ is a **solution plan** for the problem $P = (\Sigma, s_0, g)$ if:

- partial ordering $<$ and constraints B are globally consistent
 - there are no cycles in the partial ordering
 - we can assign variables in such a way that constraints from B hold
- Any linearly ordered sequence of fully instantiated actions from A satisfying $<$ and B goes from s_0 to a state satisfying g .

Hmm, but this definition **does not say how** to verify that a partial plan is a solution plan!

Claim: Partial plan $\Pi = (A, <, B, L)$ is a solution plan if:

- there are no flaws (no open goals and no threats)
- partial ordering $<$ and constraints B are globally consistent

PSP = Plan-Space Planning

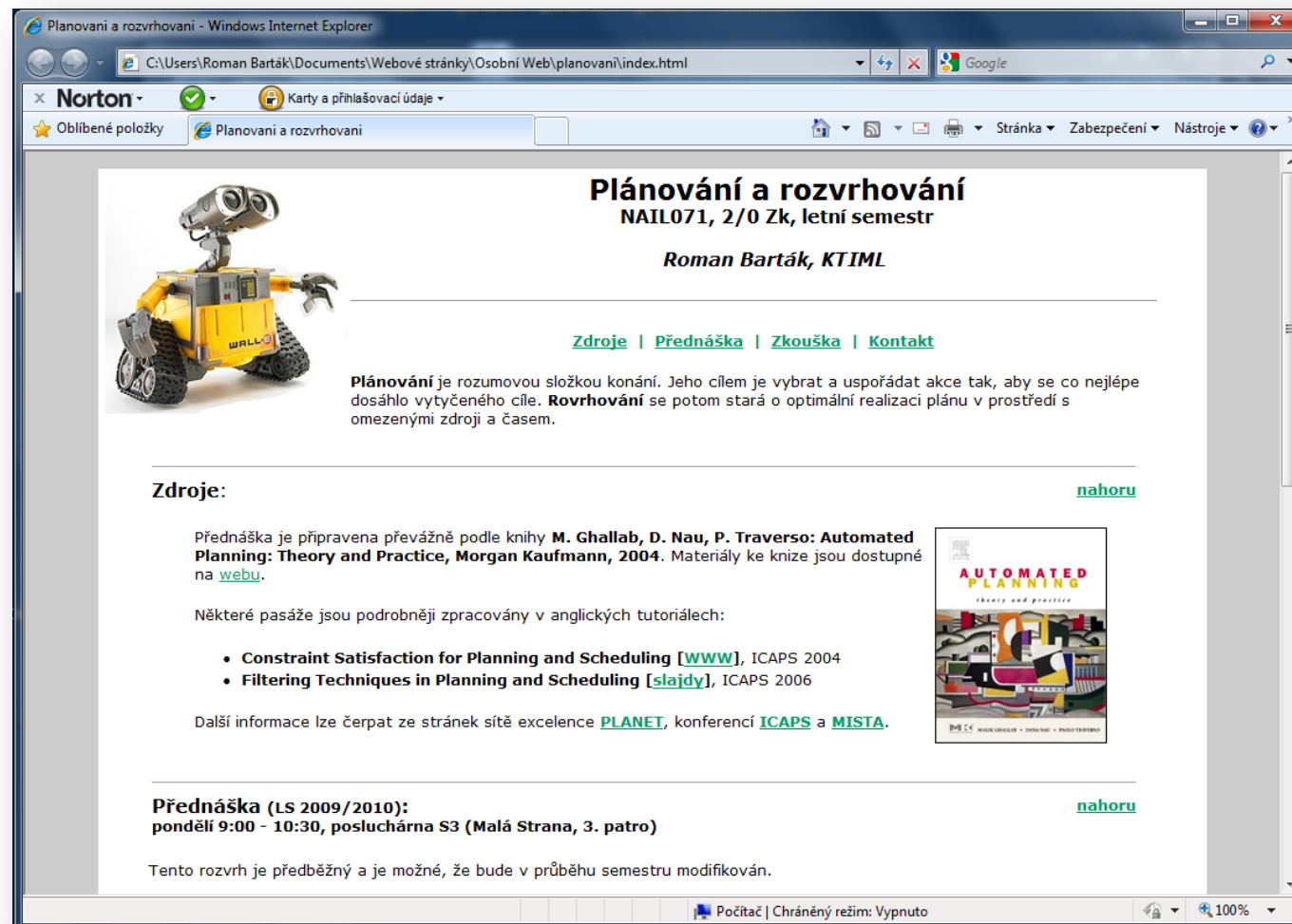
```
PSP( $\pi$ )
   $flaws \leftarrow \text{OpenGoals}(\pi) \cup \text{Threats}(\pi)$ 
  if  $flaws = \emptyset$  then return( $\pi$ )
  select any flaw  $\phi \in flaws$ 
   $resolvers \leftarrow \text{Resolve}(\phi, \pi)$ 
  if  $resolvers = \emptyset$  then return(failure)
  nondeterministically choose a resolver  $\rho \in resolvers$ 
   $\pi' \leftarrow \text{Refine}(\rho, \pi)$ 
  return(PSP( $\pi'$ ))
end
```

Notes:

- The selection of flaw is deterministic (all flaws must be resolved).
- The resolvent is selected non-deterministically (search in case of failure).

Course Planning and scheduling

– <http://ktiml.mff.cuni.cz/~bartak/planovani/>



Planovani a rozvrhování - Windows Internet Explorer

C:\Users\Roman Barták\Documents\Webové stránky\Osobní Web\planovani\index.html

Norton Karty a přihlašovací údaje

Oblíbené položky Planovani a rozvrhování

Plánování a rozvrhování

NAIL071, 2/0 Zk, letní semestr

Roman Barták, KTIML

[Zdroje](#) | [Přednáška](#) | [Zkouška](#) | [Kontakt](#)

Plánování je rozumovou složkou konání. Jeho cílem je vybrat a uspořádat akce tak, aby se co nejlépe dosáhlo vytyčeného cíle. **Rozvrhování** se potom stará o optimální realizaci plánu v prostředí s omezenými zdroji a časem.

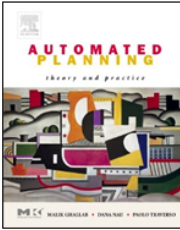
Zdroje: [nahoru](#)

Přednáška je připravena převážně podle knihy **M. Ghallab, D. Nau, P. Traverso: Automated Planning: Theory and Practice, Morgan Kaufmann, 2004**. Materiály ke knize jsou dostupné na [weby](#).

Některé pasáže jsou podrobněji zpracovány v anglických tutoriálech:

- **Constraint Satisfaction for Planning and Scheduling** [[www](#)], ICAPS 2004
- **Filtering Techniques in Planning and Scheduling** [[slajdy](#)], ICAPS 2006

Další informace lze čerpat ze stránek sítě excelence [PLANET](#), konferencí [ICAPS](#) a [MISTA](#).



Přednáška (LS 2009/2010): [nahoru](#)
pondělí 9:00 - 10:30, posluchárna S3 (Malá Strana, 3. patro)

Tento rozvrh je předběžný a je možné, že bude v průběhu semestru modifikován.

Počítač | Chráněný režim: Vypnuto 100%

An **agent view** of Artificial Intelligence

- an agent is an entity perceiving environment and acting upon it
- a **rational agent** maximizes expected performance

Problem solving with simple state space

- **search** techniques
- exploiting extra information → heuristic search **A***
- factored states → **constraint satisfaction**
- more agents → **adversarial search** (games)

Knowledge representation

- propositional and first-order **logic**
- **inference** procedures

Automated planning

- situation calculus
- state-space and plan-space planning





© 2013 Roman Barták

Department of Theoretical Computer Science and Mathematical Logic

bartak@ktiml.mff.cuni.cz