

Plan-Space Search

•Searching for a Solution Plan in a Graph of Partial Plans



Literature

•Malik Ghallab, Dana Nau, and Paolo Traverso. *Automated Planning – Theory and Practice*, chapter 2 and 5. Elsevier/Morgan Kaufmann, 2004.

•J. Penberthy and D. S. Weld. UCPOP: A sound, complete, partialorder for ADL. In *Proceeding s of the International Conference on Knowledge Representation and Reasoning*, pages 103-114, 1992.



State-Space vs. Plan-Space Search

•state-space search: search through graph of nodes representing world states

•search space directly corresponds to graph representation of state-transition system

plan-space search: search through graph of partial plans

nodes: partially specified plans

arcs: plan refinement operations

•least commitment principle: do not add constraints to the plan that are not strictly needed

solutions: partial-order plans

•partial-order plan: set of actions + set of orderings; not necessarily total order

•state-space algorithms also maintain partial plan – but always in total order



Overview

The Search Space of Partial Plans

now: defining the search space: partial plans + plan refinement operations

•Plan-Space Search Algorithms

•Extensions of the STRIPS Representation



Partial Plans

plan: set of actions organized into some structure

•organization e.g. sequence

partial plan:

subset of the actions

subset of the organizational structure

temporal ordering of actions

rationale: what the action achieves in the plan

•refers only to subset of actions

subset of variable bindings

•plan refinement operators accordingly: add actions, add ordering constraints, add causal links, add variable bindings



Adding Actions

partial plan contains actions

initial state

goal conditions

 can be represented as two actions with only effects or preconditions

set of operators with different variables

•least commitment principle: introduce actions only for a reason

reason for adding new actions

to achieve unsatisfied preconditions

to achieve unsatisfied goal conditions

•note: new actions can be added anywhere in the current partial plan



Adding Actions: Example

•empty plan:

•initial state: all initially satisfied conditions (green)

•goal: conditions that need to be satisfied (red)

```
•add operator: 1:move(r_1, l_1, m_1)
```

•number (1) to provide unique reference to this operator instance

•also used as variable index for unique variables

 least commitment principle: choose values for variables only when necessary

```
•add operator:2:load(k_2, l_2, c_2, r_2)
```



Adding Causal Links

partial plan contains causal links

·links from the provider

•an effect of an action or

•an atom that holds in the initial state

to the consumer

•a precondition of an action or

a goal condition

causal link implies ordering constraint

•but: provider need not come directly before consumer

reasons for adding causal links

prevent interference with other actions

 keeping track of rationale: any action inserted between provider and consumer must not clobber conditions in causal link

•preconditions without a causal link pointing to them are open sub-gaols



Adding Causal Links: Example

- •add link from 1:move to goal
 - changes colour of goal to green now satisfied
- •add link from 2:load to goal
- •add link from initial state to 1:move



Adding Variable Bindings

partial plan contains variable bindings

•new operators introduce new (copies of) variables into the plan

•each copy of an operator has its own set of variables that are different from variables in other operators instances

solution plan must contain actions

variable binding constraints keep track of possible values for variables and co-designation

•convention (here): give number to operator instances to distinguish them; let variables have index of operator they belong to least commitment principle:

•least commitment principle: add only necessary variable binding constraints

reasons for adding variable bindings

to turn operators into actions

to unify and effect with the precondition it supports

Adding Examp	Variable Bindings: le	
initial state attached(pile,loc) in(cont,pile) top(cont,pile) on(cont,pallet) belong(crane,loc1)	1:move(r ₁ ,I ₁ ,m ₁) precontitions a((r,I _k)) coccupied(m,)) adjacent(I ₁ ,m ₁) boccupied(I ₁) roccupied(I ₁)	
adjacent(loc2,loc1)	variable bindings:	
at(robot,loc2)	r, robot	
occupied(loc2)		
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Adding Variable Bindings: Example

•bind variables due to causal link:

- •bind r_1 to robot
- •bind m_1 to loc2
- note: variables in operator no longer red to indicate they are bound
- •clobbering: move may also destroy goal condition
- •introduce variable inequality: $I_1 \neq \text{loc2}$
- •clobbering now impossible
- •introduce causal link from initial state
- •bind I_1 to loc1

•note consistency with inequality



Adding Ordering Constraints

partial plan contains ordering constraints

•binary relation specifying the temporal order between actions in the plan

•temporal relation: qualitative, not quantitative (at this stage)

reasons for adding ordering constraints

•all actions after initial state

•all actions before goal

causal link implies ordering constraint

to avoid possible interference

•interference can be avoided by ordering the potentially interfering action before the provider or after the consumer of a causal link

•least commitment principle: introduce ordering constraints only if necessary

•result: solution plan not necessarily totally ordered



Adding Ordering Constraints: Example

•ordering constraints

due to causal links

•also: all actions before goal

•ordering: all actions after initial state

•orderings may occur between actions



Definition of Partial Plans

•A partial plan is a tuple $\pi = (A, \prec, B, L)$, where:

• $A = \{a_1, ..., a_k\}$ is a set of partially instantiated planning operators;

• \prec is a set of ordering constraints on A of the form $(a_i \prec a_j)$;

•*B* is a set of binding constraints on the variables of actions in *A* of the form x=y, $x\neq y$, or $x\in D_x$;

•*L* is a set of causal links of the form $\langle a_i - [p] \rightarrow a_j \rangle$ such that:

• a_i and a_i are actions in A;

•the constraint $(a_i \prec a_j)$ is in \prec ;

•proposition p is an effect of a_i and a precondition of a_j ; and

•the binding constraints for variables in a_i and a_j appearing in p are in B.

•sub-gaols in a partial plan: preconditions without causal links

•different view: partial plan as set of (sequential) plans

•those that meet the specified constraints and can be refined to a total order plan by adding constraints

•note: partial plans with two types of additional flexibility:

•actions only partially ordered and

•not all variables need to be instantiated



Plan-Space Search: Initial Search State

•problem: plan space representation does not maintain states, but need to give initial state and goal description

represent initial state and goal as dummy actions

init: no preconditions, initial state as effects

goal: goal conditions as preconditions, no effects

•empty plan π_0 = ({init, goal},{(init < goal)},{},{}):

two dummy actions init and goal;

one ordering constraint: init before goal;

no variable bindings; and

•no causal links.



Plan-Space Search: Initial Search State Example

 note empty box for preconditions in init and empty box for effects in goal



Plan-Space Search: Successor Function

states are partial plans

generate successor through plan refinement operators (one or more):

•more required to keep partial plans consistent, e.g. adding a causal link implies adding an ordering constraint

•adding an action to A

•adding an ordering constraint to ≺

adding a binding constraint to B

•adding a causal link to L

•successors must be consistent: constraints in a partial plan must be satisfiable

•plan-space planning decouple two sub-problems:

•which actions need to be performed

•how to organize these actions

•partial plan as set of plans: refinement operation reduces the set to smaller subset

•next: to define planning as plan-space search problem: need to define goal state



Total vs. Partial Order

•Let $\mathcal{P}=(\Sigma, s_i, g)$ be a planning problem. A plan π is a <u>solution</u> for \mathcal{P} if $p(s_i, \pi)$ satisfies g.

•solution defined for state transition system

•problem: $p(s_i, \pi)$ only defined for sequence of ground actions

partial order corresponds to total order in which all partial order constraints are respected

•partial ordering is consistent iff it is free of loops

•note: there may be an exponential number of total ordering consistent with a given partial ordering

partial instantiation corresponds to grounding in which variables are assigned values consistent with binding constraints

•note: exponential combinatorics of assigning values to variables



Partial Order Solutions

•Let $\mathcal{P}=(\Sigma, s_i, g)$ be a planning problem. A plan $\pi = (A, \prec, B, L)$ is a (partial order) <u>solution</u> for \mathcal{P} if:

 •its ordering constraints ≺ and binding constraints *B* are consistent; and

•for every sequence $\langle a_1, \dots, a_k \rangle$ of all the actions in *A*-{init, goal} that is

•totally ordered and grounded and respects \prec and B

• $\gamma(s_i, \langle a_1, \dots, a_k \rangle)$ must satisfy g.

•note: causal links do not play a role in the definition of a solution

•with exponential number of sequences to check, definition is not very useful (as computational procedure for goal test)

•idea: use causal links to verify that every precondition of every action is supported by some other action

•problem: condition not strong enough



Threat: Example

•start with partial plan from previous example (grounded; initial state not shown due to limited space on slide)

•introduce new 3:move action to achieve at(robot,loc1) precondition of 2:load action

 note: still many unachieved preconditions – not a solution yet

•add causal link to maintain rationale

•add ordering to be consistent with causal link

•new: label causal link with condition it protects

•threat: effect of 1:move is negation of condition protected by causal link

•if 1:move is executed between 3:move and 2:load the plan is no longer valid

•possible solution: additional ordering constraint



Threats

•An action a_k in a partial plan $\pi = (A, \prec, B, L)$ is a threat to a causal link

 $\langle a_i - [p] \rightarrow a_j \rangle$ iff:

• a_k has an effect $\neg q$ that is possibly inconsistent with p, i.e. q and p are unifiable;

•the ordering constraints $(a_i \prec a_k)$ and $(a_k \prec a_j)$ are consistent with \prec ; and

•the binding constraints for the unification of *q* and *p* are consistent with *B*.



Flaws

•A flaw in a plan $\pi = (A, \prec, B, L)$ is either:

•an unsatisfied sub-goal, i.e. a precondition of an action in *A* without a causal link that supports it; or

•a threat, i.e. an action that may interfere with a causal link.



Flawless Plans and Solutions

•Proposition: A partial plan $\pi = (A, \prec, B, L)$ is a solution to the planning problem $\mathcal{P}=(\Sigma, s_i, g)$ if:

• π has no flaw;

•the ordering constraints ≺ are not circular; and

•the variable bindings *B* are consistent.

•computation:

•let partial plans in the search space only violate the first condition (have flaws)

•partial plans that violate either of the last two conditions cannot be refined into a solution and need not be generated

•Proof: by induction on number of actions in A

base case: empty plan

•no flaws – every goal condition is supported by causal link from initial state

•induction step: totally ordered plan minus first step is solution implies plan including first step is a solution:

 $\gamma(s_i, \langle a_1, \dots, a_k \rangle) = \gamma(\gamma(s_i, a_1), \langle a_2, \dots, a_k \rangle)$

•truncated plan is solution to different problem



Overview

The Search Space of Partial Plans

just done: defining the search space: partial plans + plan refinement operations

•Plan-Space Search Algorithms

•now: an algorithm that performs the search through the space of partial plans

•Extensions of the STRIPS Representation



Plan-Space Planning as a Search Problem

•given: statement of a planning problem *P*=(*O*,*s_i*,*g*)
•define the search problem as follows:

•initial state: $\pi_0 = (\{\text{init}, \text{goal}\}, \{(\text{init} \prec \text{goal})\}, \{\}, \{\})$

goal test for plan state p: p has no flaws

•path cost function for plan π : $|\pi|$

•successor function for plan state p: refinements of p that maintain \prec and B

•note: plan space may be infinite even when state space is finite



PSP Procedure: Basic Operations

•PSP: Plan-Space Planner

•main principle: refine partial π plan while maintaining \prec and *B* consistent until π has no more flaws

basic operations:

•find the flaws of π , i.e. its sub-goals and its threats

•simple for empty plan – all goal conditions are unachieved sub-goals and no threats

select one of the flaws

- •find ways to resolve the chosen flaw
- choose one of the resolvers for the flaw

•refine π according to the chosen resolver

•modify the plan in such a way that \prec and *B* are in a consistent state for the generated successor

•aim: no need to verify consistency of \prec and *B* for goal test



•PSP: Pseudo Code

•function PSP(plan)

-refines the given partial plan into a solution plan; start with initial plan $\pi_{\rm 0}$

•allFlaws ← plan.openGoals() + plan.threats()

•if allFlaws.empty() then return plan

•see proposition in previous section: no flaws implies solution

•flaw ← allFlaws.selectOne()

•allResolvers ← flaw.getResolvers(plan)

•represents all possible ways of removing the selected flaw from the partial plan

•if allResolvers.empty() then return failure

•no resolvers means plan cannot be made flawless

resolver ← allResolvers.chooseOne()

•newPlan ← plan.refine(resolver)

 must maintain consistency of ≺ and B; new plan may contain new flaws

•return PSP(newPlan)



PSP: Choice Points

resolver ← allResolvers.chooseOne()

non-deterministic choice

•flaw ← allFlaws.selectOne()

deterministic selection

•all flaws need to be resolved before a plan becomes a solution

order not important for completeness

order is important for efficiency

•for finding first plan, not so for finding all plans

•deterministic implementation: using IDA*, for example



Implementing *plan*.openGoals()

•finding unachieved sub-goals (incrementally):

•in π_0 : goal conditions

•when adding an action: all preconditions are unachieved sub-goals

 when adding a causal link: protected proposition is no longer unachieved



Implementing plan.threats()

•finding threats (incrementally):

•in π_0 : no threats

•when adding an action a_{new} to $\pi = (A, \prec, B, L)$:

•for every causal link $\langle a_i - [p] \rightarrow a_i \rangle \in L$

•if $(a_{new} \prec a_i)$ or $(a_i \prec a_{new})$ then next link

•if the new action must occur before the provider or after the consumer of the link

•else for every effect q of a_{new}

•if $(\exists \sigma: \sigma(p) = \sigma(\neg q))$ then q of a_{new} threatens $\langle a_i - [p] \rightarrow a_i \rangle$

• $\exists \sigma$: test whether there is a substitution consistent with *B*!

•when adding a causal link $\langle a_i - [p] \rightarrow a_j \rangle$ to $\pi = (A, \prec, B, L)$:

•for every action *a_{old}*∈A

•if $(a_{old} \prec a_i)$ or $(a_j = a_{old})$ or $(a_j \prec a_{old})$ then next action

•else for every effect q of a_{old}

•if $(\exists \sigma: \sigma(p) = \sigma(\neg q))$ then q of a_{old} threatens $\langle a_i - [p] \rightarrow a_i \rangle$



Implementing *flaw*.getResolvers(*plan*) •for unachieved precondition *p* of a_g : •add causal links to an existing action: •for every action $a_{old} \in A$ •if $(a_g = a_{old})$ or $(a_g \prec a_{old})$ then next action •else for every effect *q* of a_{old} •if $(\exists \sigma: \sigma(p) = \sigma(q))$ then adding $\langle a_{old} - [\sigma(p)] \rightarrow a_g \rangle$ is a resolver •add a new action and a causal link: •for every effect *q* of every operator *o* •if $(\exists \sigma: \sigma(p) = \sigma(q))$ then adding $a_{new} = 0$.newInstance() and $\langle a_{new} - [\sigma(p)] \rightarrow a_g \rangle$ is a resolver



Implementing *flaw*.getResolvers(*plan*) •for effect *q* of action *a_t* threatening $\langle a_i - [p] \rightarrow a_j \rangle$: •order action before threatened link: •if $(a_t = a_i)$ or $(a_j \prec a_t)$ then not a resolver •else adding $(a_t \prec a_i)$ is a resolver •order threatened link before action: •if $(a_t = a_i)$ or $(a_t \prec a_i)$ then not a resolver •else adding $(a_j \prec a_t)$ is a resolver •else adding $(a_j \prec a_t)$ is a resolver •ethend variable bindings such that unification fails: •for every variable *v* in *p* or *q* •if $v \neq \sigma(v)$ is consistent with *B* then adding $v \neq \sigma(v)$ is a resolver



Implementing *plan*.refine(*resolver*)

•refines partial plan with elements in resolver by adding:

an ordering constraint;

- one or more binding constraints;
- •a causal link; and/or
- •a new action.

no testing required

•all testing already done in *flaw*.getResolvers(*plan*)

•must update flaws:

unachieved preconditions (see: plan.openGoals())

•threats (see: plan.threats())



Maintaining Ordering Constraints

•required operations:

•query whether $(a_i \prec a_j)$

•adding (*a_i≺a_j*)

•without consistency testing

•possible internal representations:

 maintain set of predecessors/successors for each action as given

maintain only direct predecessors/successors for each action

maintain transitive closure of ≺ relation

•operations have different time and space complexity

•note: query performed more often than addition



Maintaining Variable Binding Constraints •types of constraints:

```
•unary constraints: x \in D_x
```

```
•equality constraints: x = y
```

 unary and equality constraints can be solved in linear time

```
•inequalities: x ≠ y
```

•inequalities give rise to general CSP problem

note: general CSP problem is NP-complete

PSP: Data Flow	
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PSP: Data Flow

•deterministic step: selecting a a flaw

- •no backtracking required
- •selection important for efficiency
- •heuristic guidance required

•non-deterministic step: choosing a resolver for a flaw

•implemented as backtracking

•order in which resolvers are tried important for efficiency

•heuristic guidance required

•note: admissible heuristics (A*) must have step cost greater than zero



PSP: Sound and Complete

•Proposition: The PSP procedure is sound and complete: whenever π_0 can be refined into a solution plan, PSP(π_0) returns such a plan.

•Proof:

•soundness: ≺ and *B* are consistent at every stage of the refinement

•completeness: induction on the number of actions in the solution plan

•note: non-deterministic version is complete, deterministic implementation must avoid infinite branches



PSP Implementation: PoP

based on UCPOP

•extended input:

partial plan (as before)

•agenda: set of pairs (*a*,*p*) where *a* is an action an *p* is one of its preconditions

•initial agenda: one pair for each precondition of the goal step

search control by flaw type

unachieved sub-goal (on agenda): as before

 threats: resolved as part of the successor generation process



PoP: Pseudo Code (1) •function PoP(*plan*, *agenda*) •if *agenda*.empty() then return *plan* •(a_g, p_g) \leftarrow *agenda*.selectOne() •deterministic choice point •agenda \leftarrow *agenda* - (a_g, p_g) •relevant \leftarrow *plan*.getProviders(p_g) •finds all actions •either from within the *plan* or •from new instances of an operator •that have an effect that unifies with *condition* •if *relevant*.empty() then return failure •(a_p, p_p, σ) \leftarrow *relevant*.chooseOne() •non-deterministic choice point •plan.L \leftarrow plan.L $\cup \langle a_p - [p] \rightarrow a_g \rangle$

•plan.B ← plan.B ∪ σ

•must succeed for elements of relevant



PoP: Pseudo Code (2)

•if $a_p \notin plan.A$ then

if the action is new and needs to be added to the plan *plan.add(a_p)*

•involves updating set of actions and ordering constraints •agenda \leftarrow agenda + a_p .preconditions

•all preconditions of the new action are new sub-goals

•newPlan ← plan

•for each *threat* on $\langle a_p - [p] \rightarrow a_g \rangle$ or due to a_p do

•note: two sources of threats are treated identically

•allResolvers ← threat.getResolvers(newPlan)

•if allResolvers.empty() then return failure

resolver ← allResolvers.chooseOne()

•second non-deterministic choice point

note: loop does not add to agenda

•return PSP(newPlan,agenda)



State-Space vs. Plan-Space Planning

•state-space planning VS. plan-space planning

•finite search space vs. finite search space

•important: portion of search space explored/generated

•explicit representation of intermediate states vs. no intermediate states

•explicit representation allows for efficient domain specific heuristics and control knowledge

action ordering reflects control strategy vs. choice of actions and organization independent

 causal structure only implicit vs. explicit representation of rationale

•important for plan execution

•search nodes relatively simple and successors easy to compute vs. search nodes are complex and successors expensive to compute



Using Partial-Order Plans: Main Advantages

more flexible during execution

 saved rationale facilitates execution monitoring and replanning

using constraint managers facilitates extensions such as:

temporal constraints

resource constraints

•both very important for realistic planning

distributed and multi-agent planning fit naturally into the framework



Overview

The Search Space of Partial Plans

•Plan-Space Search Algorithms

•just done: an algorithm that performs the search through the space of partial plans

•Extensions of the STRIPS Representation

•now: extensions to the restricted STRIPS representation and approaches to deal with them



Existential Quantification in Goals

•allow existentially quantified conjunction of literals as goal:

 $\bullet g = \exists x_1, \dots, x_n : I_1 \land \dots \land I_m$

•but still no free variables in the goal literals

•rewrite into equivalent planning problem:

•new goal g' = {p} where p is an unused proposition
symbol

•introduce additional operator $o = (op-g(x_1,...,x_n),\{I_1,...,I_m\},\{p\})$

•solution plans will be one step longer, with last step being an instantiation of op-g

•no increase in expressive power (can rewrite every problem), but sometimes makes representation appear more natural

•in plan-space search: no change needed

•partial plans do not require all variables to be instantiated



Example: Existential Quantification in Goals

•goal: $\exists x, y$: on(x,c1) \land on(y,c2)

•rewritten goal: *p*

```
•new operator: o = (op-g(x,y),{on(x,c1),on(y,c2)},{p})
```

```
•plan-space search goal: on(x,c1) \land on(y,c2)
```

•variables in goals are implicitly existentially quantified



Typed Variables

•allow typed variables in operators:

•name(o) = $n(x_1:t_1,...,x_k:t_k)$ where t_i is the type of variable x_i

•types usually given with problem specification

•exact syntax for typing not important

•rewrite into equivalent planning problem:

•add preconditions $\{t_1(x_1), \dots, t_k(x_k)\}$ to o

•types are unary predicates here

•if constant c_i is of type t_j , add $t_j(c_i)$ to the initial state •remove types from operator names

•similarly: typed relations

•advantages: readability, reduced number of actions (operator instances)



DWR Example: Typed Variables

```
•operator: move(r:robot,l:location,m:location)
```

```
•precond: adjacent(I,m), at(r,I), ¬occupied(m)
```

```
•effects: at(r,m), occupied(m), ¬occupied(l), ¬at(r,l)
```

```
•rewritten operator:move(r,I,m)
```

```
•precond: adjacent(l,m), at(r,l), ¬occupied(m), robot(r),
loaction(l), location(m)
```

```
•effects: at(r,m), occupied(m), ¬occupied(l), ¬at(r,l)
```

```
•rewritten initial state:
```

```
•s_i \cup \{robot(r1), container(c1), container(c2), ...\}
```

•note: dealing with typed variables directly in the planner is far more efficient



Conditional Operators

•conditional planning operators:

•*n* = $o(x_1,...,x_n)$ as before,

•(precond₀,effects₀) are the unconditional preconditions and effects of the operator, and

•not strictly necessary, could be added to every conditional pair

•(precond_{*i*}, effects_{*i*}) for $i \ge 1$ are the conditional preconditions and effects of the operator.

•a ground instance *a* of *o* is applicable in state *s* if *s* satisfies precond₀

•let I={i∈[0,n] | s satisfies precond_i(a)}; then:

• $\gamma(s,a)$ =(s - $\bigcup_{i \in \Lambda}$ effects⁻(a)) \cup ($\bigcup_{i \in \Lambda}$ effects⁺(a))



DWR Example: Conditional Operators

•relation at(o,l): object o is at location l

•at previously only used for location of robot

conditional move operator:

move(*r*,*l*,*m*,*c*)

•new parameter *c*, the potentially loaded container

```
•precond<sub>0</sub>: adjacent(I,m), at(r,I), ¬occupied(m)
```

•effects₀: at(r,m), occupied(m), ¬occupied(l), ¬at(r,l)

•as before

```
•precond<sub>1</sub>: loaded(r,c)
```

```
•effects<sub>1</sub>: at(c,m), ¬at(c,l)
```

•if the container is loaded it will move with the robot



Extending PoP to handle Conditional Operators •modifying *plan*.getProviders(*p*_g):

new action with matching conditional effect

add precondition of conditional effect to agenda

•along with unconditional preconditions

managing conditional threats:

•when the threatening effect is a conditional effect

new alternative resolver: add negated precondition of threatening conditional effect to agenda

•other alternatives: ordering and binding constraints



Quantified Expressions

 allow universally quantified variables in conditional preconditions and effects:

•for-all x₁,...,x_n: (precond_i,effects_i)

•*a* is applicable in state *s* if *s* satisfies precond₀

•applicability depends only on unconditional preconditions

•Let σ be a substitution for x_1, \dots, x_n such that $\sigma(\text{precond}_i(a))$ and $\sigma(\text{effects}_i(a))$ are ground.

If s satisfies σ(precond_i(a)) then

• σ (effects_{*i*}(*a*)) are effects of the action.

•conditional effect occurs for all possible substitutions where the state satisfies the conditional preconditions

•note: cannot use binding constraints to resolve conditional threats here



DWR Example: Quantified Expressions

•extension: robots can carry multiple containers

•extended move operator: move(r,l,m)

```
•container is no longer a parameter
•precond<sub>0</sub>: adjacent(I,m), at(r,I), ¬occupied(m)
•effects<sub>0</sub>: at(r,m), occupied(m), ¬occupied(I), ¬at(r,I)
•for-all x: (the containers)
•precond<sub>1</sub>: loaded(r,x)
•effects<sub>1</sub>: at(x,m), ¬at(x,I))
```



Disjunctive Preconditions

•allow alternatives (disjunctions) in preconditions:

```
•precond = precond<sub>1</sub> \vee ... \vee precond<sub>n</sub>
```

•*a* is applicable in state *s* if *s* satisfies at least one of precond₁ ... precond_n

•effects remain unchanged

•rewrite:

replace operator with *n* disjunctive preconditions by *n* operators with precond_i as precondition

•leads to exponentially larger search space; more efficiently handled in the planner



DWR Example: Disjunctive Preconditions

 robot can move between locations if there is a road between them or the robot has all-wheel drive

extended move operator:

move(*r*,*l*,*m*)

```
•precond: (road(I,m), at(r,I), ¬occupied(m)) ∨ (all-wheel-drive(r), at(r,I), ¬occupied(m))
```

•more complex formulae can be transformed into DNF

```
•effects: at(r,m), occupied(m), ¬occupied(l), ¬at(r,l)
```



Axiomatic Inference: Static Case

•idea: represent knowledge that is not explicit in the state of the world; derived knowledge

•axioms over rigid relations:

•example: $\forall I_1, I_2$: adjacent $(I_1, I_2) \leftrightarrow adjacent(I_2, I_1)$

•adjacent is a symmetric relationship

state-specific axioms:

•example: $\forall c$: container(c) \leftrightarrow at(c,loc1) holds in s_i

•in the initial state, all containers are at location loc1

approach: pre-compute

•rigid relations: cannot appear in effects, truth value does not change from state to state; hence, can be pre-computed

•state-specific axioms: expansion into ground atoms possible because of finite domains

•same technique for quantified effects



Axiomatic Inference: Dynamic Case

axioms over flexible relations:

•example: $\forall k, x$: ¬holding(k, x) \leftrightarrow empty(k)

•a crane is empty iff it is not holding anything

•approach:

•divide relations into primary and secondary where secondary relations do not appear in effects

•primary relations may appear in preconditions and effects, secondary relations only in preconditions

transform axioms into implications where primary relations must not appear in right-hand side

•truth value of secondary relations depends on primary relations which are modified by operators

•example:

•primary: holding / secondary: empty

•must remove empty-relation from effects of all operators

• $\forall k \neg \exists x$: holding(k, x) \rightarrow empty(k)

• $\forall k \exists x$: holding(k, x) $\rightarrow \neg$ empty(k)



Extended Goals

not part of classical planning formalisms

•some problems can be translated into equivalent classical problems, e.g.

 states to be avoided: add corresponding preconditions to operators

 states to be visited twice: introduce visited relation and maintain in operators

•constraints on solution length: introduce count relation that is increased with each step

•requires function symbols and integer arithmetic (attached procedures)



Other Extensions

Function Symbols

infinite domains, undecidable in general

•expansion into ground atoms no longer possible, reasoning within one state is undecidable

Attached Procedures

•evaluate relations using special code rather than general inference

•efficiency may be necessary in real-world domains

•example: numeric computations

variables must usually be bound to evaluate relations

•difficult with least commitment approach taken by many planners

semantics of such relations



Overview

The Search Space of Partial Plans

•Plan-Space Search Algorithms

•Extensions of the STRIPS Representation

•just done: extensions to the restricted STRIPS representation and approaches to deal with them