Planning (Ch. 10)



Forward search Action(GoTo(x, y, z), z)Precondition: $At(x, y) \wedge Mobile(x)$, Effect: $\neg At(x, y) \land At(x, z)$) Last time... Initial: At(Truck, UPSD) ^ Package(UPSD, P1) ^ Package(UPSD, P2) ^ Mobile(Truck) Goal: Package(H1, P1) ^ Package(H2, P2) Action(Load(m, x, y), Precondition: $At(m, y) \wedge Package(y, x)$, Effect: $\neg Package(y, x) \land Package(m, x) \land At(m, y))$ Action(Deliver(m, x, y), Precondition: $At(m, y) \wedge Package(m, x)$, Effect: $\neg Package(m, x) \land Package(y, x) \land At(m, y))$

Heuristics for planning

Backwards search has a smaller branching factor in general, but it is hard to use heuristics

This is due to it looking at sets of states, and not a single state for the next action

For this reason, it is often better to apply a good heuristic to the dumb forward search

Heuristics for planning

Reformulate our grilling problem as actions: Action(MakeSandwitch(x), Precondition: $Meat(x) \land Make(Bread, x, Bread)$) Effect: $\neg Meat(x) \land Sandwitch(Bread)$

If our goal is just "Sandwitch(Bread)", backtracking search would try to solve: $Meat(x) \land Make(Bread, x, Bread))$... but since "x" is still a variable, this represents a set of states rather than one

Heuristics for planning

In "search" we had no generalize-able heuristics as each problem could be different

Heuristics in planning are found the same way, we (1) relax the problem (2) solve it optimally

Two generic ways to always do this are:1. Add more actions2. Reduce number of states

Heuristics: add actions

Multiple ways to add actions (to goal faster):

1. Ignore preconditions completely - also ignore any effects not related to goals

This becomes set-covering problem, which is NP-hard but has P approximations

2. Ignore any deletions in effects (i.e. anything with ¬), also NP-hard but P approximation

Ignore preconditions

By simply removing preconditions, we allow every action to happen at every state

Action(GoTo(x, y, z), Precondition: $At(x, y) \land Mobile(x)$, Effect: $\neg At(x, y) \land At(x, z)$)

Action(GoTo(x, y, z), Precondition: Effect: $\neg At(x, y) \land At(x, z)$)

Ignore preconditions

More importantly for the solution is how the Delivery action changes

The USPD can now just directly deliver to houses, so goal is: Deliver(USPD, P1, H1) and then Deliver(USPD, P2, H2) Action(Deliver(m, x, y),



Precondition:

Effect: $\neg Package(m, x) \land Package(y, x))$

Ignore negative effects

To use this heuristic, the goal & preconditions cannot have negatives (i.e. $\neg At(Truck, H2)$)

This can always be rewritten to something else

Action(GoTo(x, y, z), Precondition: $At(x, y) \land Mobile(x)$, Effect: $\neg At(x, y) \land At(x, z)$) Action(GoTo(x, y, z), Precondition: $At(x, y) \land Mobile(x)$, Effect: At(x, z))

Ignore negative effects

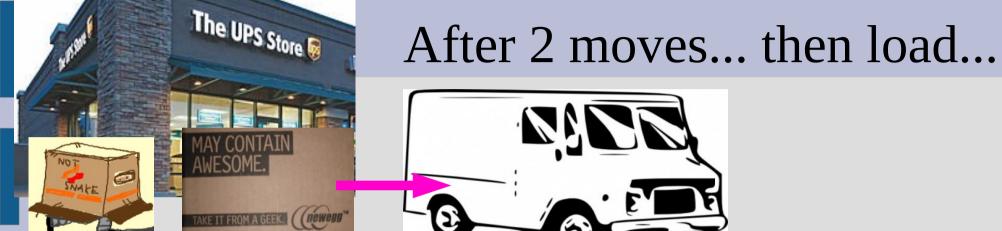
For the UPS delivery example, it does not help us find a solution faster (min is 6 still)

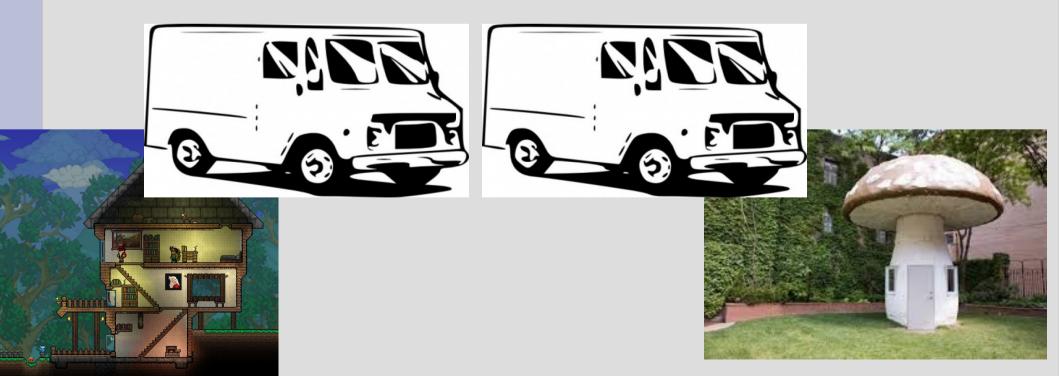
However, there are many more solutions as every action "copies" instead of "moves"

For example, a solution could be: Move, Move, Load, Load, Deliver, Deliver

This is possible as truck exists at all 3 spots!

$At(Truck, UPSD) \land At(Truck, H1) \land At(Truck, H2) \\ \land Package(UPSD, P1) \land Package(UPSD, P2) \land Mobile(Truck)$





 $At(Truck, UPSD) \land At(Truck, H1) \land At(Truck, H2) \land Package(Truck, P2) \land Package(UPSD, P1) \land Package(UPSD, P2) \land Mobile(Truck)$





Group similar states together into "super states" and solve the problem within "super states" separately (divide & conquer)

An admissible but bad heuristic would be the maximum of all "super states" individual solutions (but this is often poor)

A possibly non-admissible would be the sum of all "super states" (need independence)

These "super states" can created in many ways

Delete relations/fluents (e.g. no more "At")
 Merge objects/literals (e.g. merge UPSD and Truck)



You then need to solve two problems:1. Between the abstract "super states"2. Within each "super state"

For example, if we were instead delivering 3 packages, 1 to H1 and 2 to H2...

We combine the two packages for H2 into a single "super package" with only one load and deliver (overall "super state" solution)

We then can simply see that each load/deliver corresponds to two individual loads/delivers (within super state solution)

Or you could split up the problem into two sub-parts:

Part 1: delivering packages to house 1 Part 2: delivering packages to house 2

As trucks don't have capacity limits, the total actions could be the number of sub-actions (this is not true in general)



A heuristic we will go over in detail is <u>graph</u> <u>plan</u>ning, which tries to do all possible actions at each step

The graph plan heuristic is nice because it is always admissible and computable in P time

The basic idea of graph plan is to track all the statements that could be true at any time (and allow multiple actions simultaneously)

Graph plan is an underestimate because once a relation/literal is added, it is never removed

Unlike the "remove negative effects" heuristic, we allow both negative and positive effects

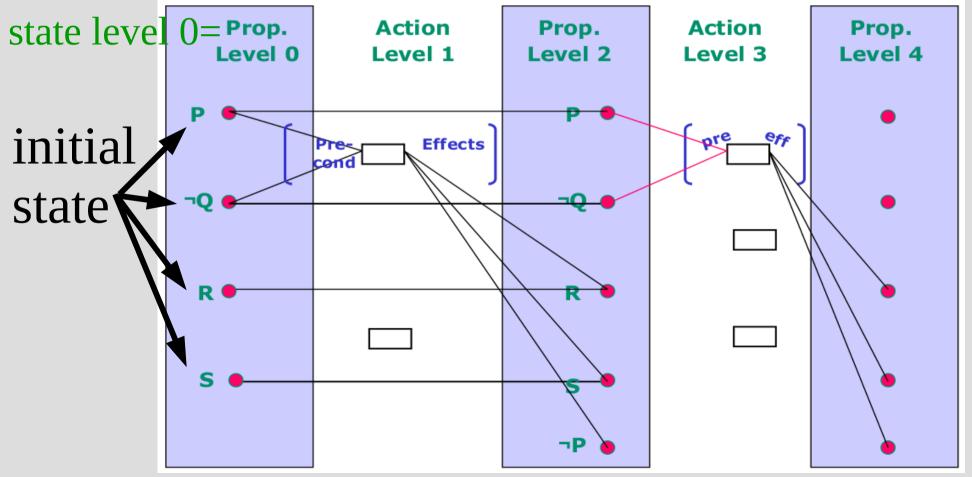
But we can also use any preconditions that have been found anytime before (not quite as open as completely removing them)

These simplifications/relaxations probably make the problem too easy

So we also track pairs of both actions and literals that are in conflict (called <u>mutex</u>es)

First, let's go over how to convert actions and relations into graph plan, then later we will add in the mutexes

Graph plan will alternate between possible facts ("state level") and actions ("action level")



You start with the relations of the initial state on the left (now explicitly stating negatives)

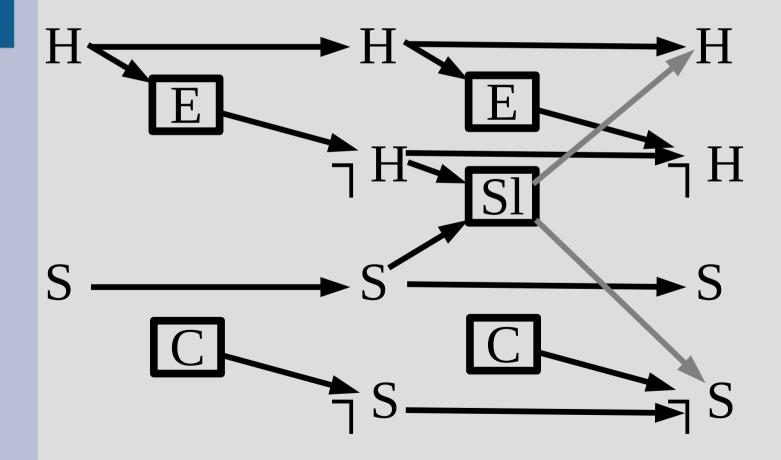
Then you add "no actions" which simply keep all the relationships the same but move them to the right

Then you add actions, which you do by linking preconditions on the left to resulting effects on the right (adding any new ones)

Consider this problem:
Initial: $Sleepy(me) \wedge Hungry(me)$
Goal: $\neg Sleepy(me) \wedge \neg Hungry(me)$ Action(Eat(x),
Precondition: Hungry(x),
Effect: $\neg Hungry(x)$)Action(Coffee(x),
Precondition: ,
Effect: $\neg Sleepy(x)$)

Action(Sleep(x), Precondition: $Sleepy(x) \land \neg Hungr(x)$, Effect: $\neg Sleepy(x) \land Hungry(x)$)

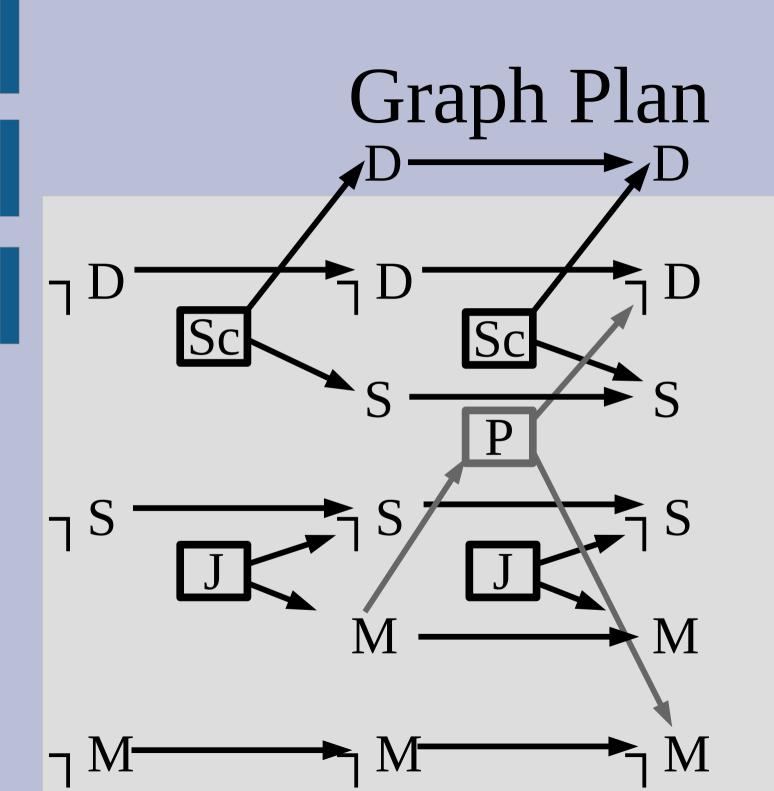
Consider this problem:



Each set of relations/literals are what we call <u>levels</u> of the graph plan, S = states, A = actions

State level 0 is $S_0 = \{H, S\}$ $A_0 = \{C, E, all "no ops"\}$ $S_1 = \{H, \neg H, S, \neg S\}$ $A_1 = \{C, E, Sl, all "no ops"\}$ $S_2 = \{H, \neg H, S, \neg S\}$

You do it! (show 3 state and 2 action levels) Initial: $\neg Money(me) \land \neg Smart(me) \land \neg Debt(me)$ Goal: $Money(me) \land Smart(me) \land \neg Debt(me)$ Action(School(x), Action(Job(x), Precondition: , Precondition: , Effect: $Debt(x) \wedge Smart(x)$) Effect: $Money(x) \land \neg Smart(x)$) Action(Pay(x), Precondition: Money(x), Effect: $\neg Money(x) \land \neg Debt(x)$)



The graph plan allows multiple actions to be done in a single turn, which is why S_1 has both γ Sleepy(me) and γ Hungry(me)

You keep building the graph until either:(1) You find your goal (more on this later)(2) The graph converges (i.e. states, actions and mutexes stop changing)

Mutexes

A <u>mutex</u> are two things that cannot be together (i.e. cannot happen or be true simultaneously)

You can put mutexes:

- 1. Between two relationships/literals
- 2. Between actions

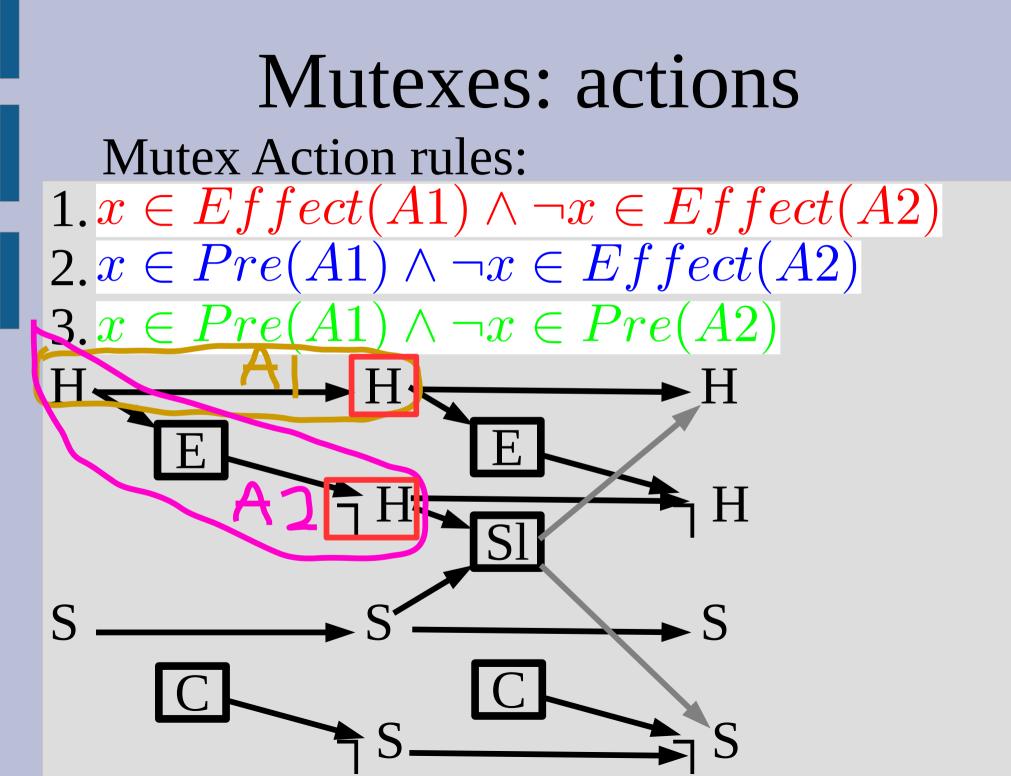
There are different rules for doing mutexes between actions vs. relations

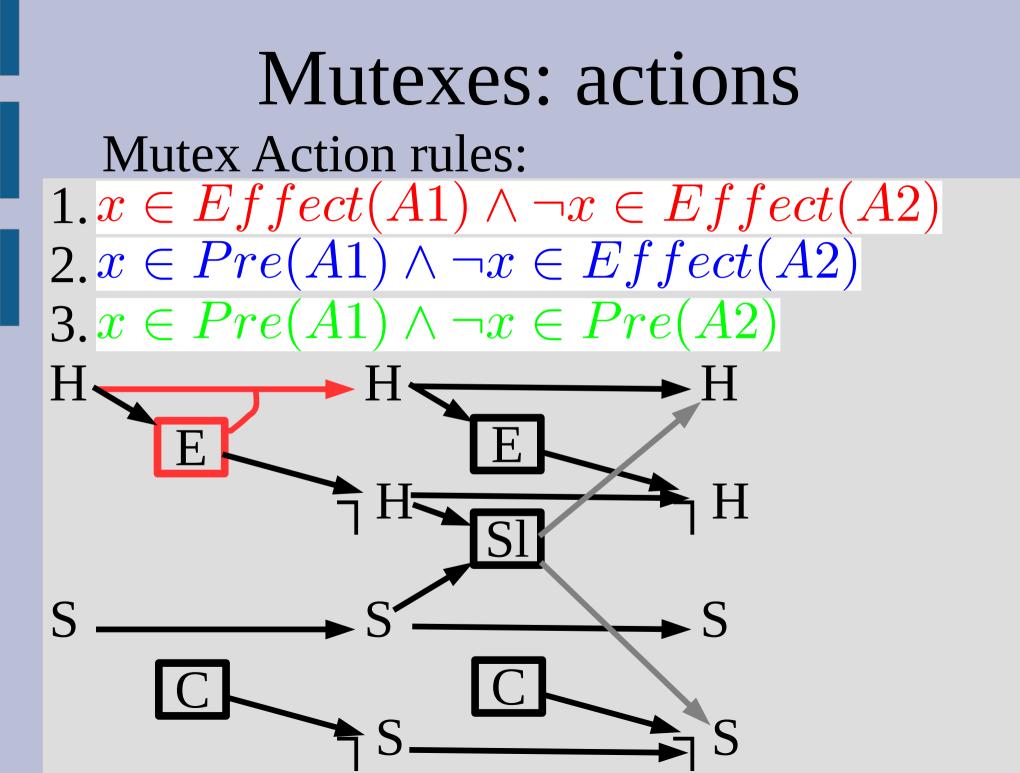
Mutexes: actions

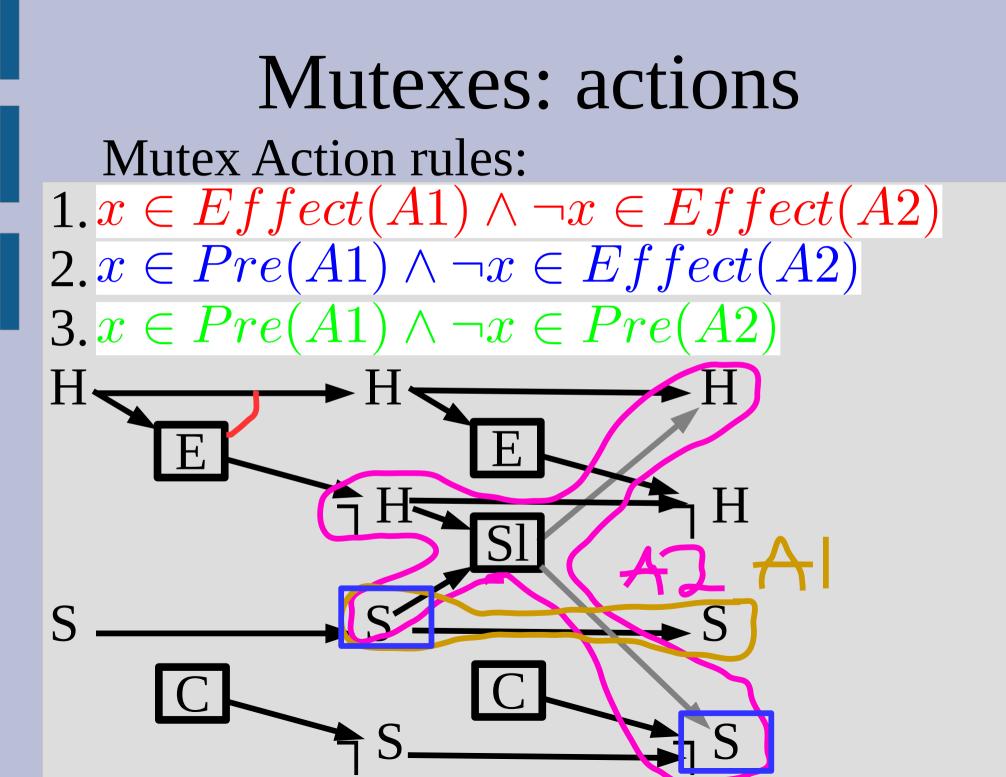
For all of these cases I will assume actions two actions: A1 and A2

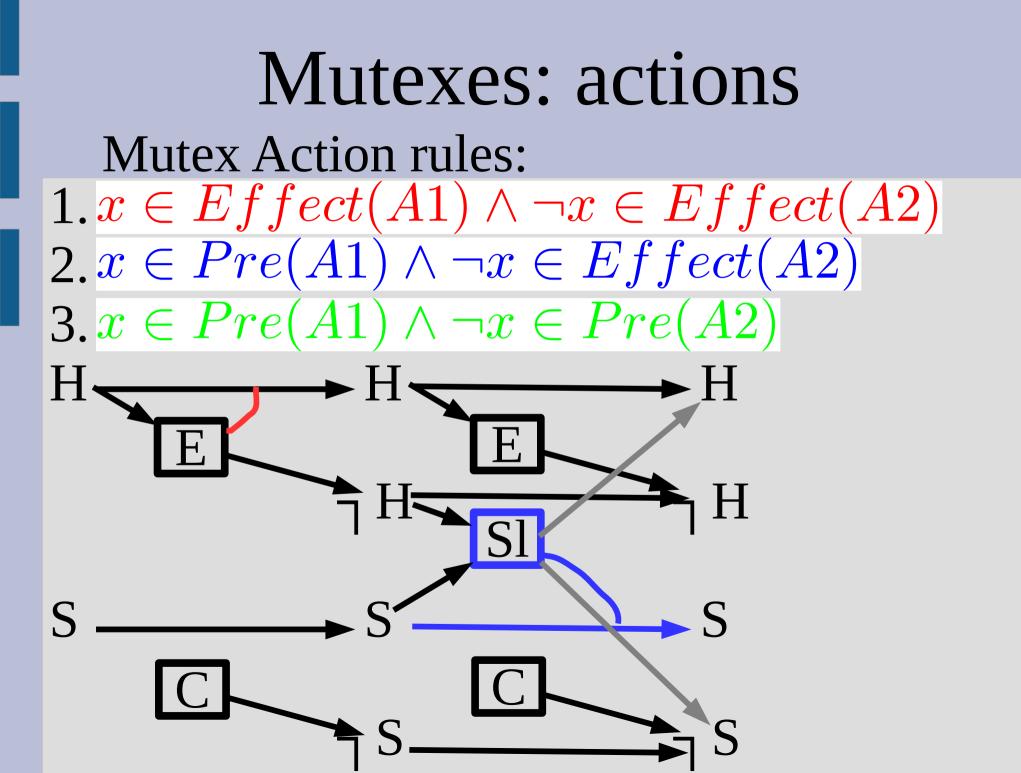
These actions have preconditions and effects: Pre(A1) and Effect(A1), respectively

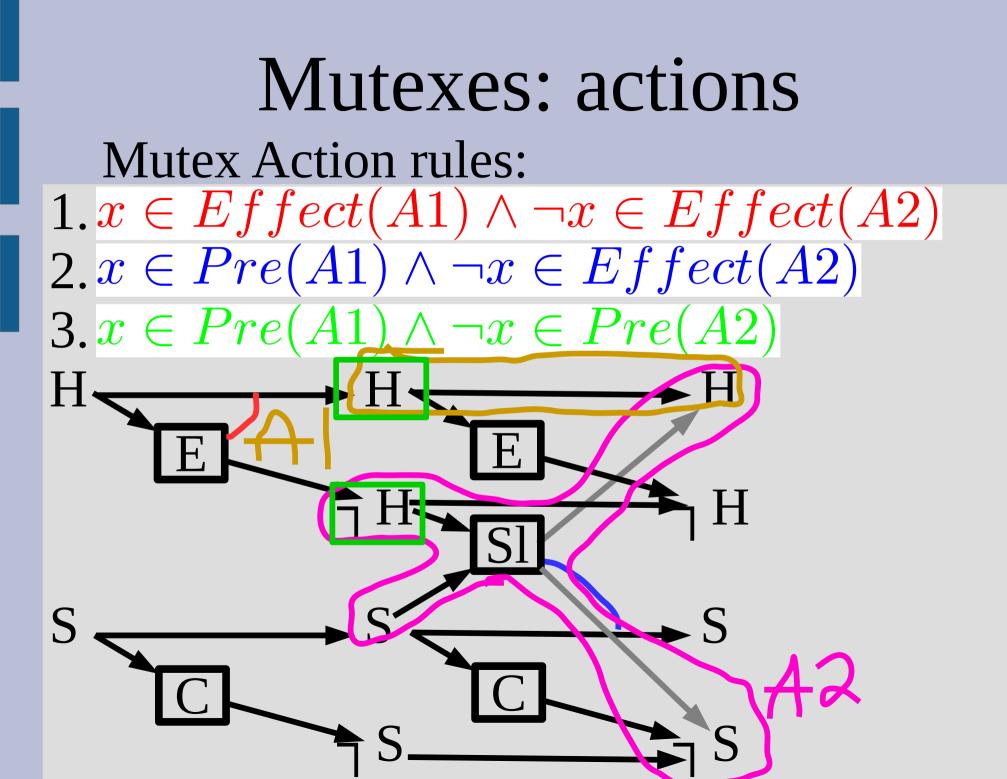
For example, I will abbreviate below as: Action(Eat(x), Precondition: Hungry(x), Effect: $\neg Hungry(x)$) A1 = Eat $H \in Pre(A1)$ $\neg H \in Effect(A1)$

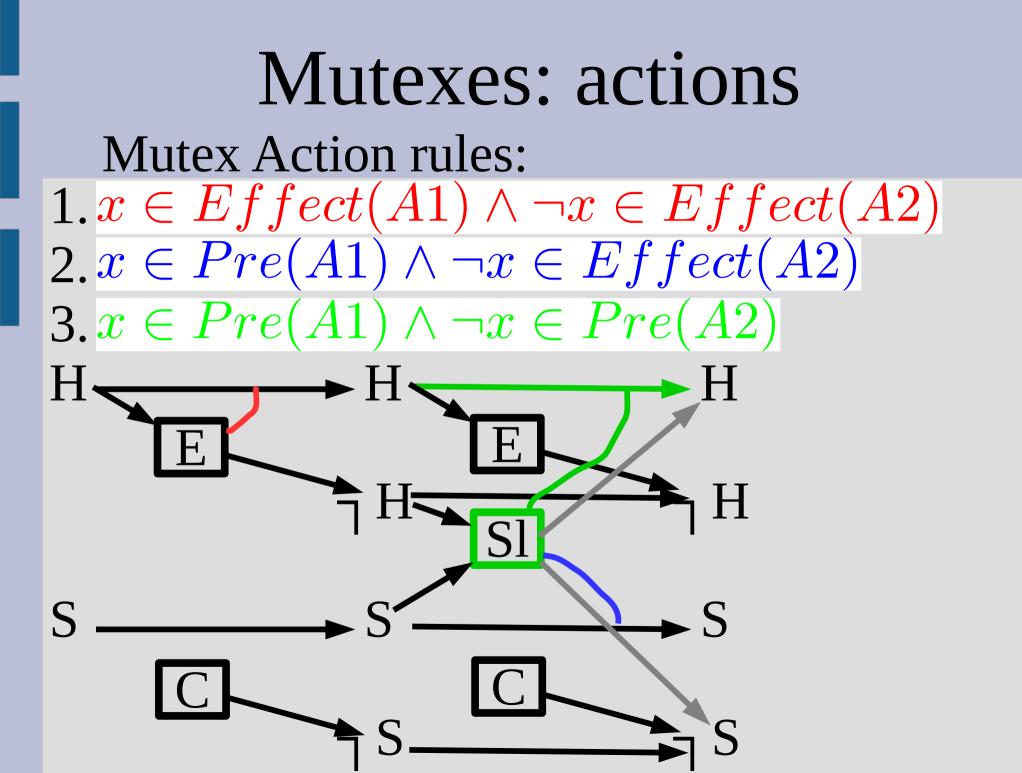


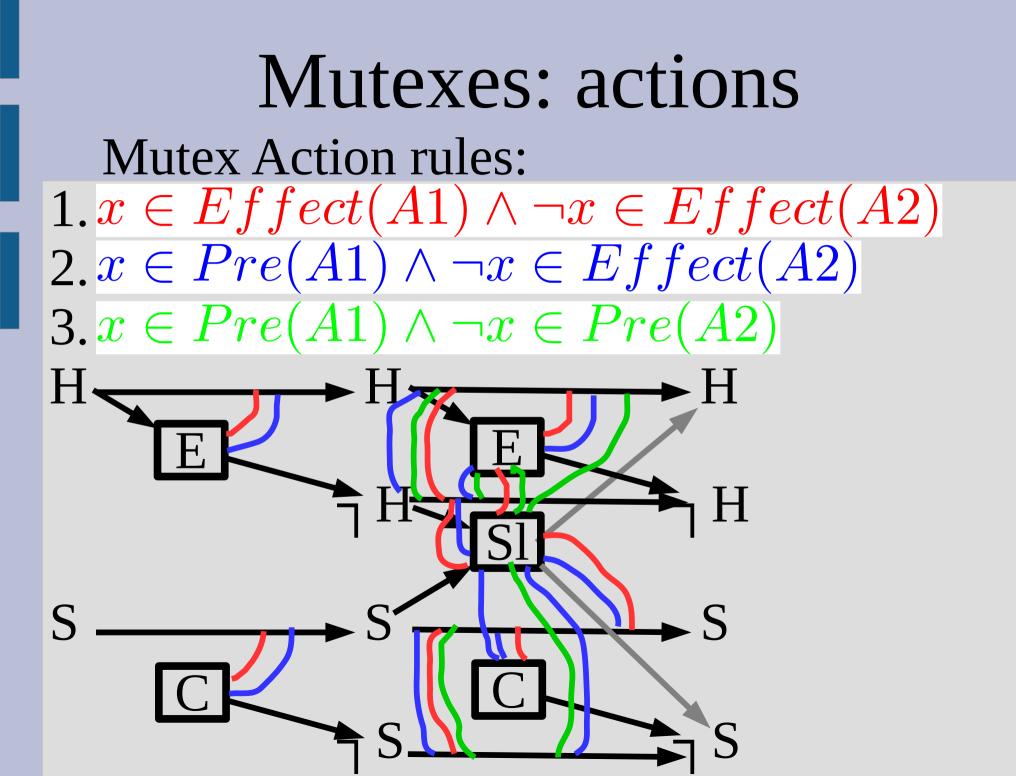












There are 2 rules for states, but unlike action-mutexes they can change across levels

Opposite relations are mutexes (x and ¬ x)
 If there are mutexes between all possible actions that "lead" to a pair of states...

Two ways that "leading" can be in mutex:1. Actions are in mutex2. Preconditions of action pair are in mutex

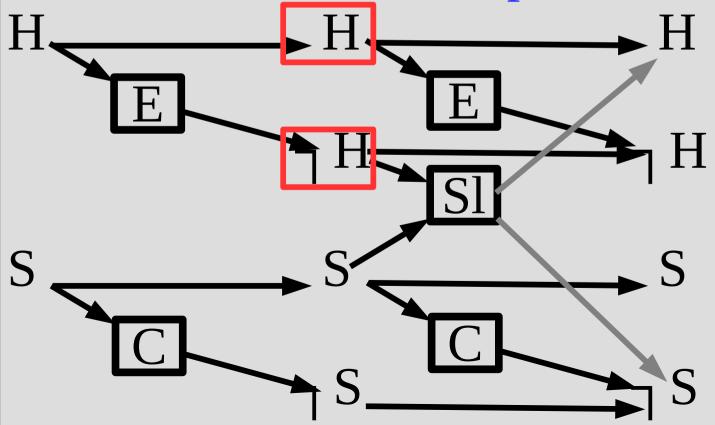
Another way to compute state mutexes:

(1) Add mutexes between all pairs in state(2) If any pair of actions can lead to this pair of relationships, un-mutex them

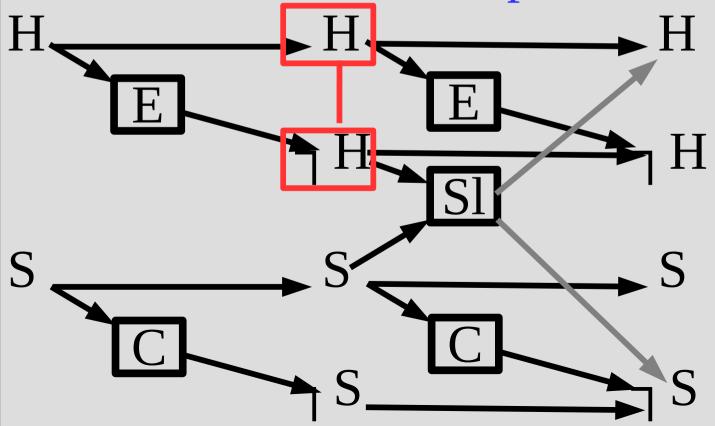
Recap:

If any valid pair of actions = no mutex All ways of reaching invalid = mutex

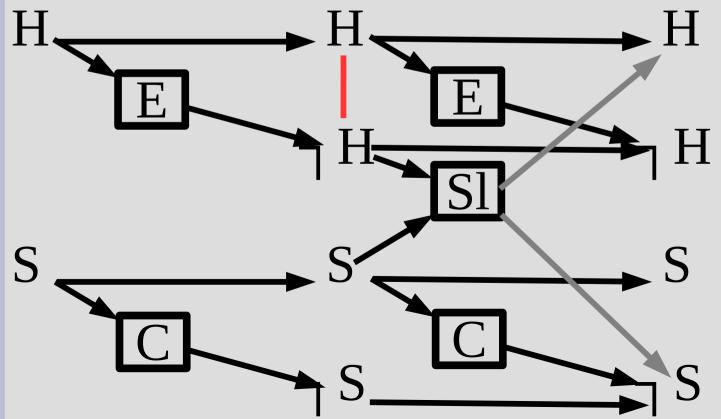
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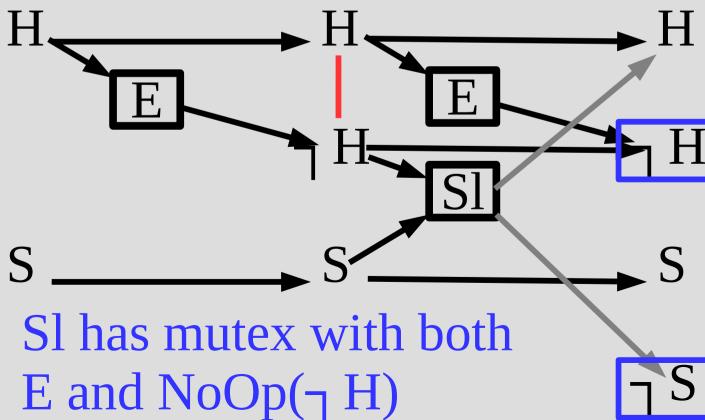


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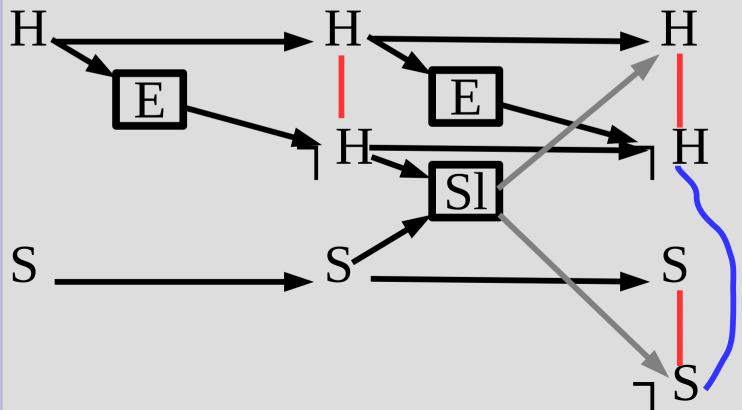
None... but if we remove coffee...

Opposite relations are mutexes (x and ¬ x)
 If there are mutexes between all possible actions that lead to a pair of states



This mutex will be gone on the next level (as you can eat again)

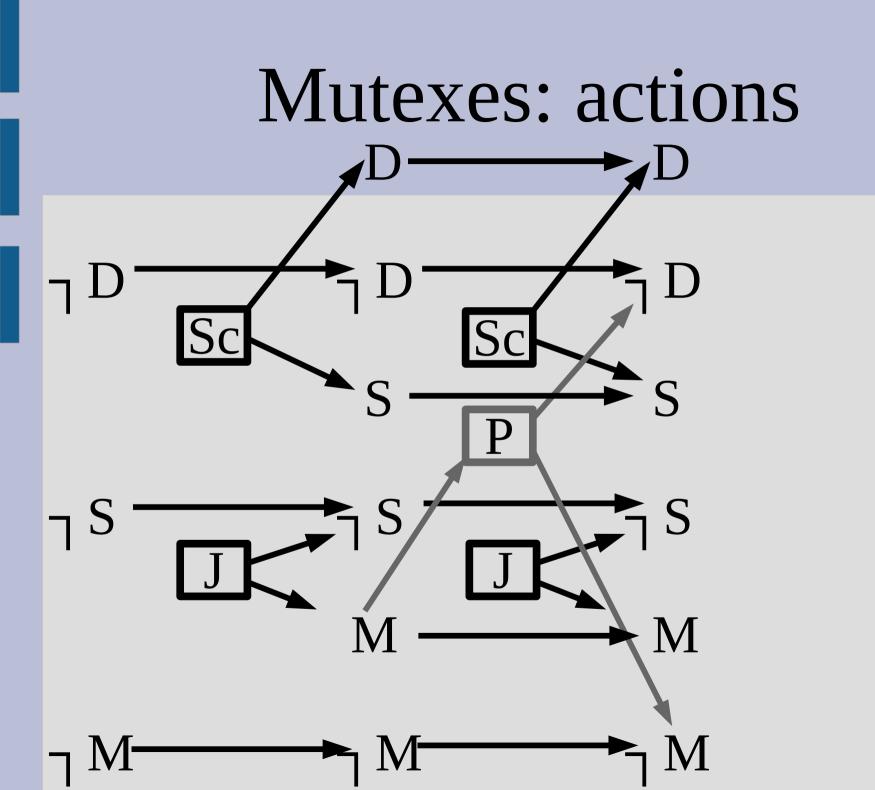
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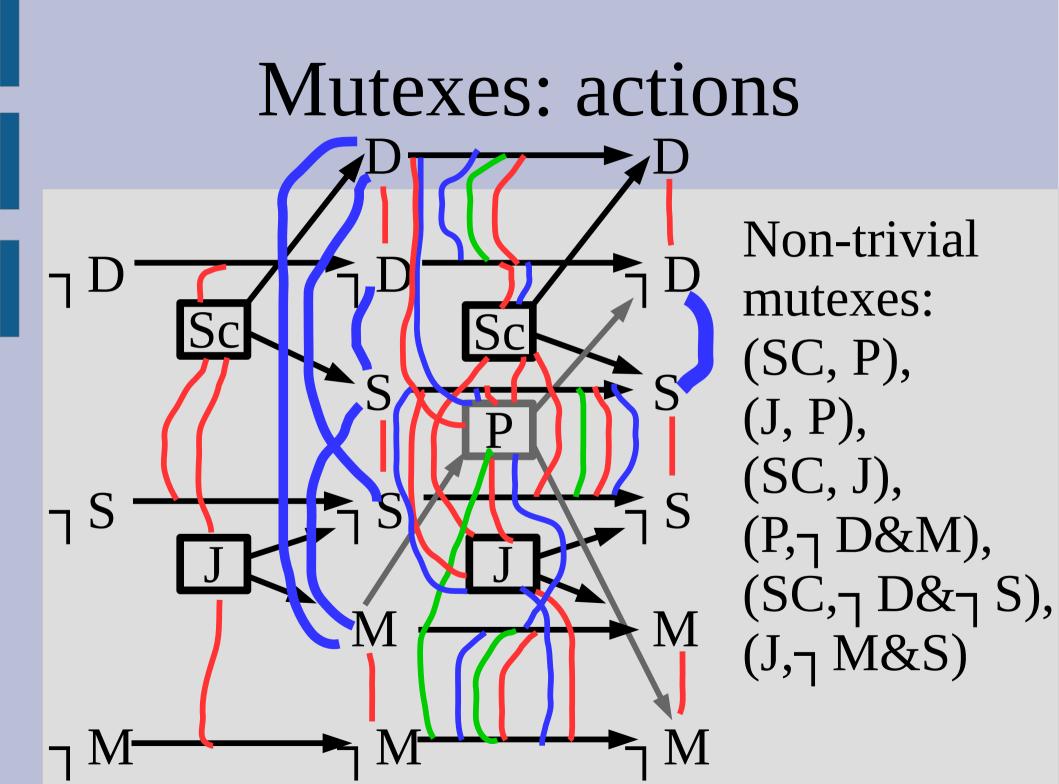


Mutexes: actions

You do it! Initial: $\neg Money \land \neg Smart \land \neg Debt$ Goal: $\neg Money \land Smart \land \neg Debt$ Action(School, Action (Job, Precondition:, Precondition:, Effect: $Debt \wedge Smart$) Effect: $Money \wedge \neg Smart$) Action(Pay, Precondition: Money,

Effect: $\neg Money \land \neg Debt$)





GraphPlan

GraphPlan can be computed in $O(n(a+l)^2)$, where n = levels before convergence a = number of actions l = number of relations/literals/states (square is due to needing to check all pairs)

The original planning problem is PSPACE, which is known to be harder than NP

GraphPlan: states

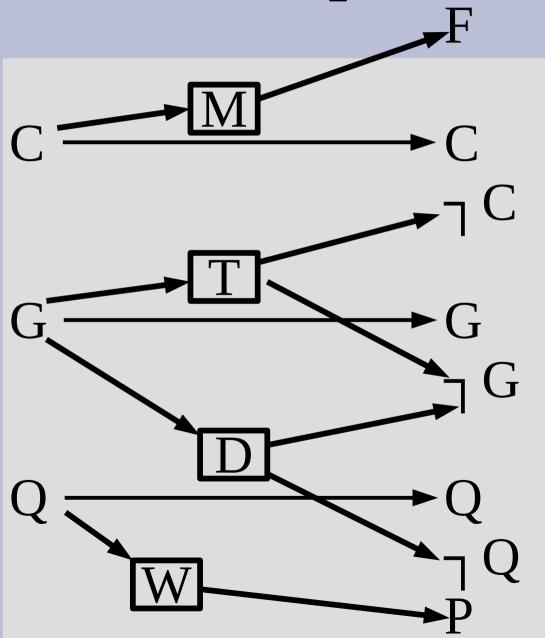
Let's consider this problem: Initial: $Clean \land Garbage \land Quiet$ Goal: $Food \land \neg Garbage \land Present$

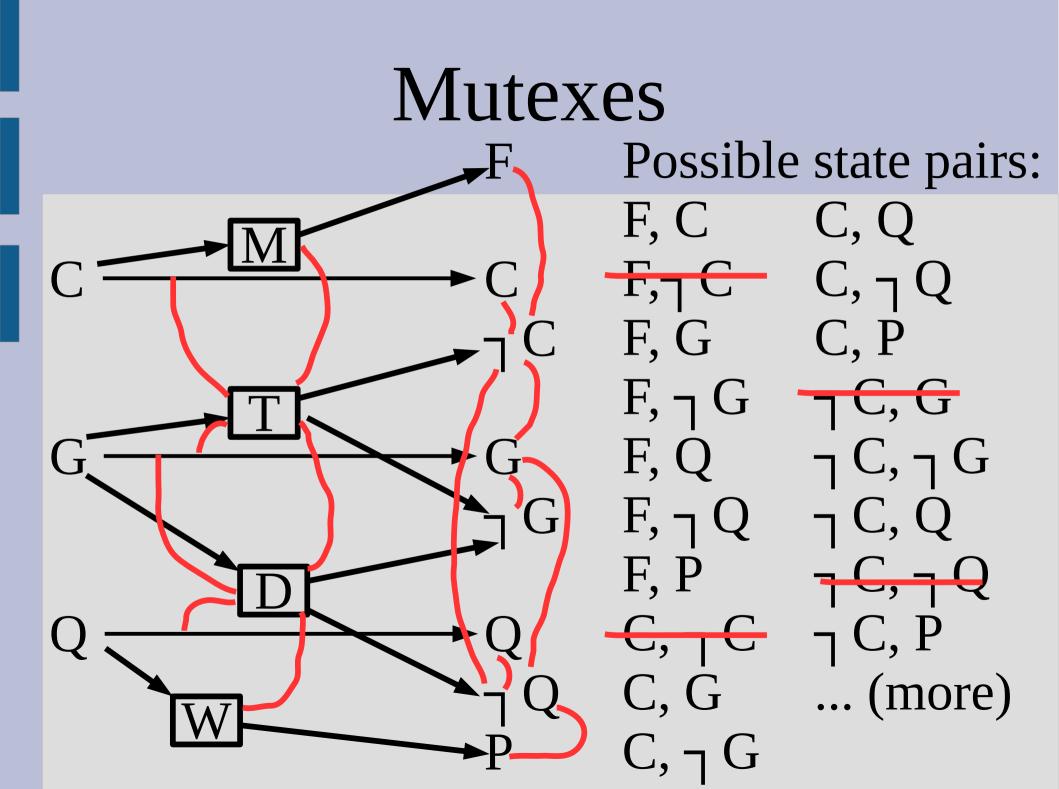
Action: (*MakeFood*, Precondition: *Clean*, Effects: *Food*)

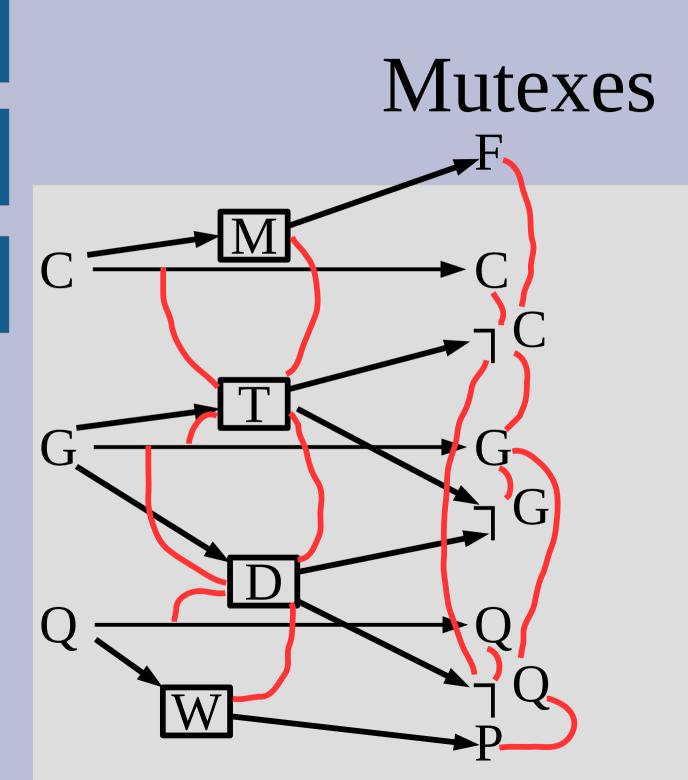
Action: (*Takeout*, Precondition: *Garbage*, Effects: $\neg Garbage \land \neg Clean$)

Action: (*Wrap*, Precondition: *Quiet*, Effects: *Present*) Action: (*Dolly*, Precondition: *Garbage*, Effects: $\neg Garbage \land \neg Quiet$)

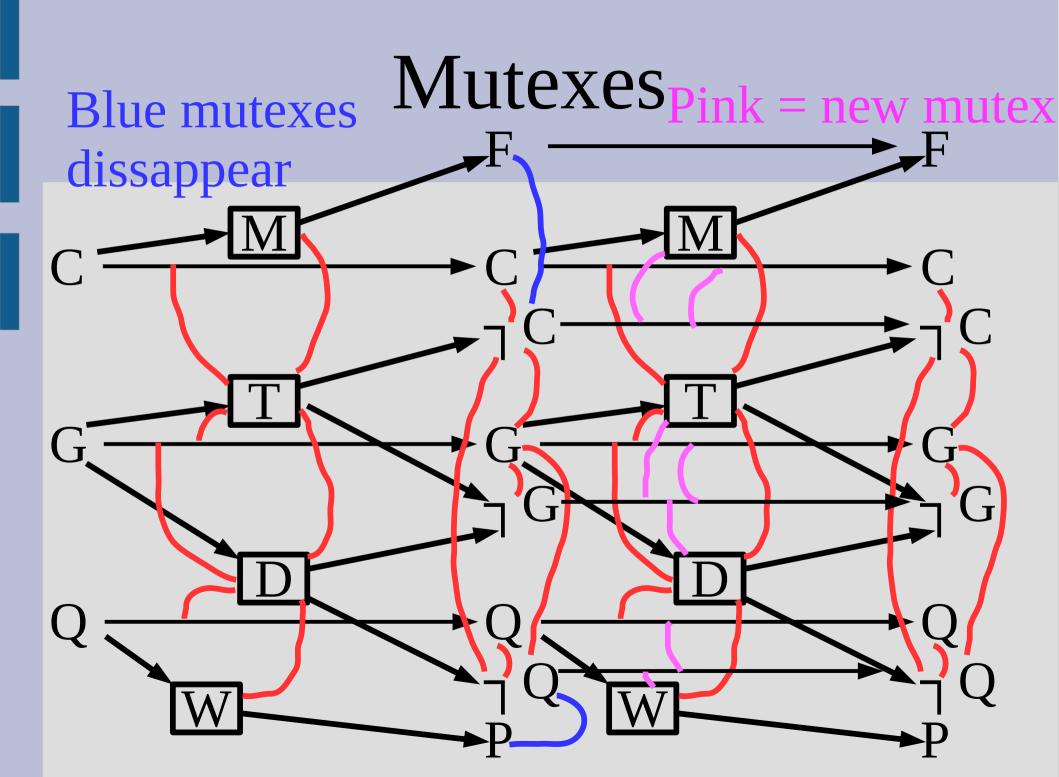
GraphPlan: states







Make one more level here!



GraphPlan is optimistic, so if any pair of goal states are in mutex, the goal is impossible

3 basic ways to use GraphPlan as heuristic:(1) Maximum level of all goals(2) Sum of level of all goals (not admissible)(3) Level where no pair of goals is in mutex

(1) and (2) do not require any mutexes, but are less accurate (quick 'n' dirty)

For heuristics (1) and (2), we relax as such:

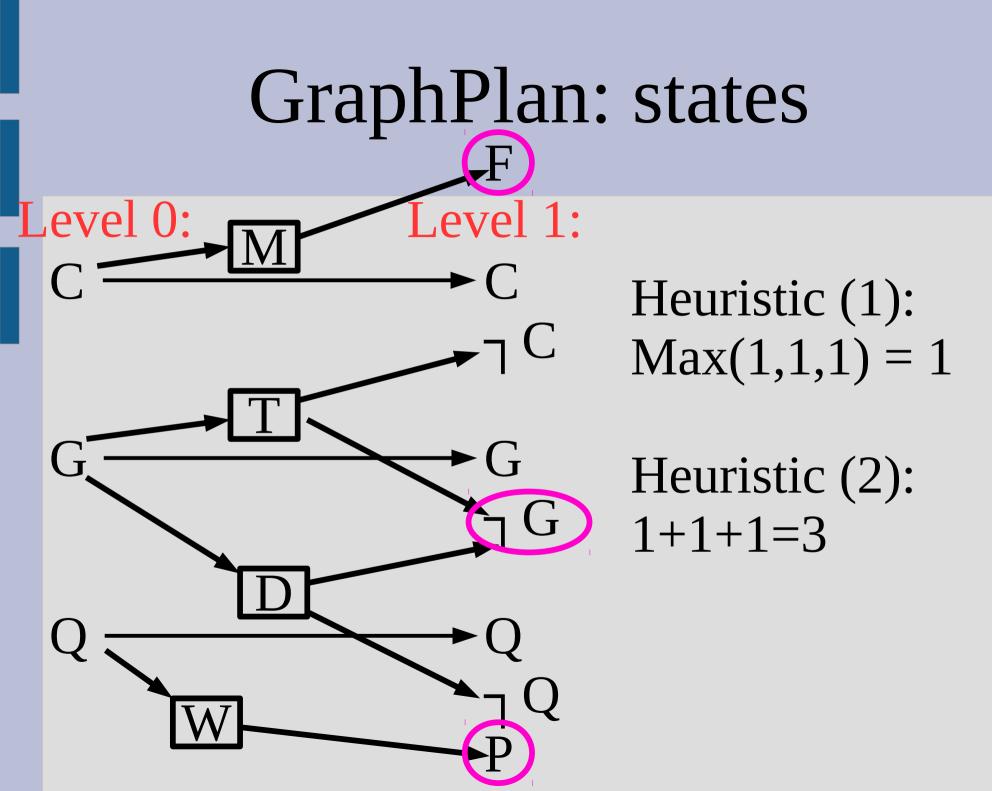
- 1. Multiple actions per step, so can only take fewer steps to reach same result
- 2. Never remove any states, so the number of possible states only increases

This is a valid simplification of the problem, but it is often too simplistic directly

Heuristic (1) directly uses this relaxation and finds the first time when all 3 goals appear at a state level

(2) tries to sum the levels of each individual first appearance, which is not admissible(but works well if they are independent parts)

Our problem: goal={Food, \neg Garbage, Present} First appearance: F=1, \neg G=1, P=1



Often the problem is too trivial with just those two simplifications

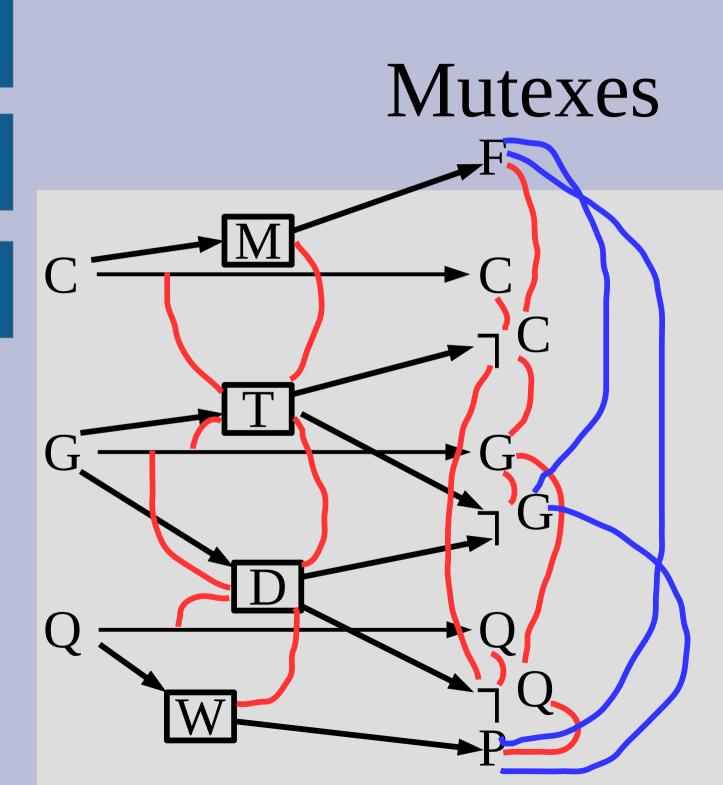
So we add in mutexes to keep track of invalid pairs of states/actions

This is still a simplification, as only impossible state/action pairs in the original problem are in mutex in the relaxation

Heuristic (3) looks to find the first time none of the goal pairs are in mutex

For our problem, the goal states are: (Food, ¬ Garbage, Present)

So all pairs that need to have no mutex: (F, \neg G), (F, P), (\neg G, P)



None of the pairs are in mutex at level 1

This is our heuristic estimate

Finding a solution

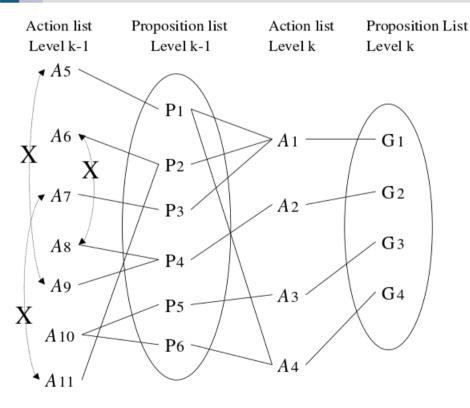
GraphPlan can also be used to find a solution:(1) Converting to a Constraint Sat. Problem(2) Backwards search

Both of these ways can be run once GraphPlan has all goal pairs not in mutex (or converges)

Additionally, you might need to extend it out a few more levels further to find a solution (as GraphPlan underestimates)

GraphPlan as CSP

Variables = states, Domains = actions out of Constraints = mutexes & preconditions



Variables: $G_1, \dots, G_4, P_1 \dots P_6$

Domains: $G_1: \{A_1\}, G_2: \{A_2\}G_3: \{A_3\}G_4: \{A_4\}$ $P_1: \{A_5\}P_2: \{A_6, A_{11}\}P_3: \{A_7\}P_4: \{A_8, A_9\}$ $P_5: \{A_{10}\}P_6: \{A_{10}\}$

Constraints (normal):
$$P_1 = A_5 \Rightarrow P_4 \neq A_9$$

 $P_2 = A_6 \Rightarrow P_4 \neq A_8$
 $P_2 = A_{11} \Rightarrow P_3 \neq A_7$

Constraints (Activity): $G_1 = A_1 \Rightarrow Active\{P_1, P_2, P_3\}$ $G_2 = A_2 \Rightarrow Active\{P_4\}$ $G_3 = A_3 \Rightarrow Active\{P_5\}$ $G_4 = A_4 \Rightarrow Active\{P_1, P_6\}$

Init State: $Active \{G_1, G_2, G_3, G_4\}$

(b) DCSP from Do & Kambhampati

(a) Planning Graph

Finding a solution

For backward search, attempt to find arrows back to the initial state(without conflict/mutex)

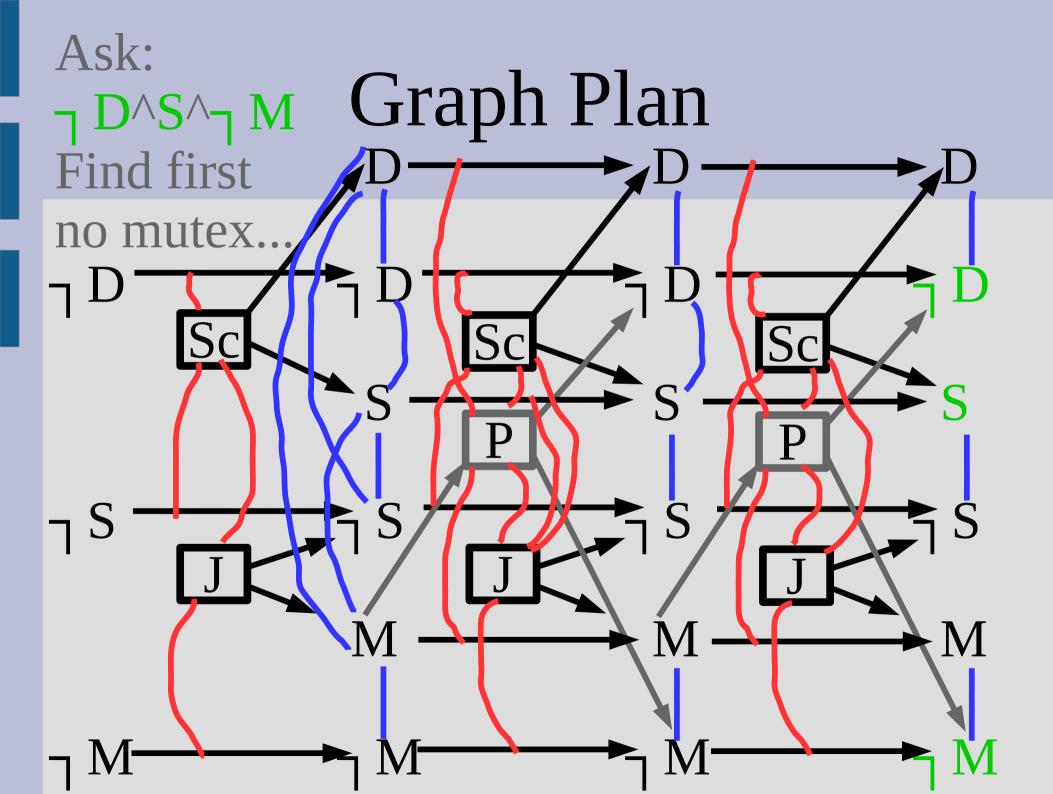
Start by finding actions that satisfy all goal conditions, then recursively try to satisfy all of the selected actions' preconditions

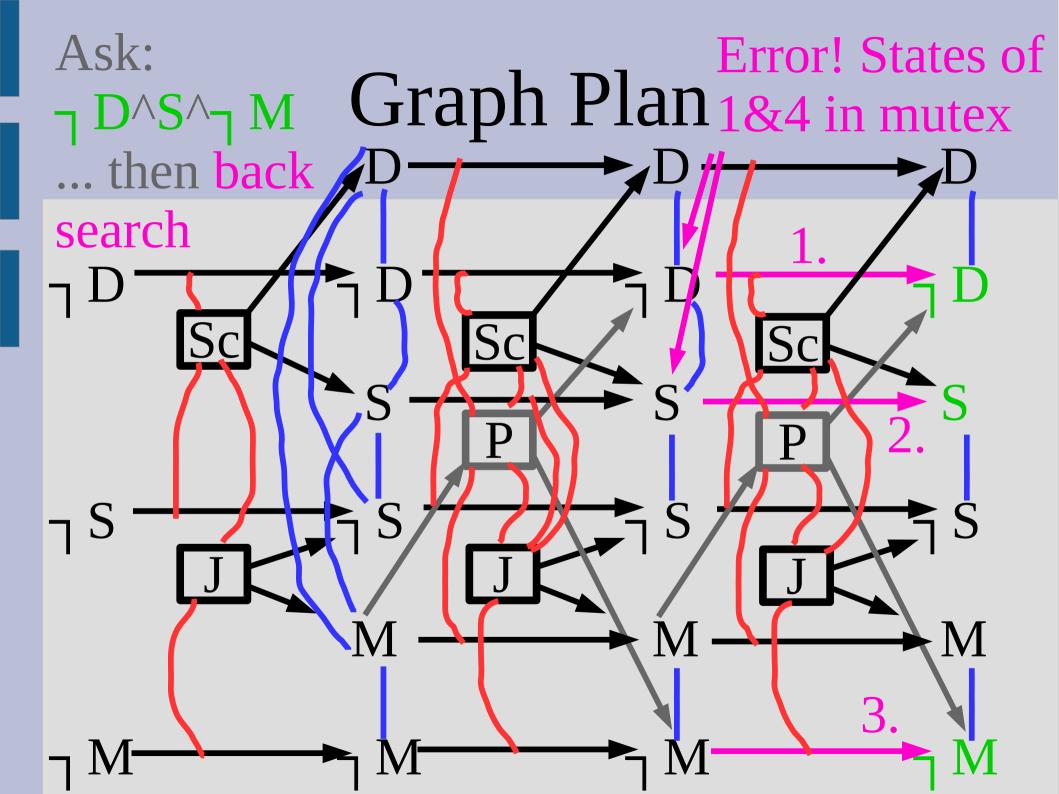
If this fails to find a solution, mark this level and all the goals not satisfied as: (level, goals) (level, goals) stops changing, no solution

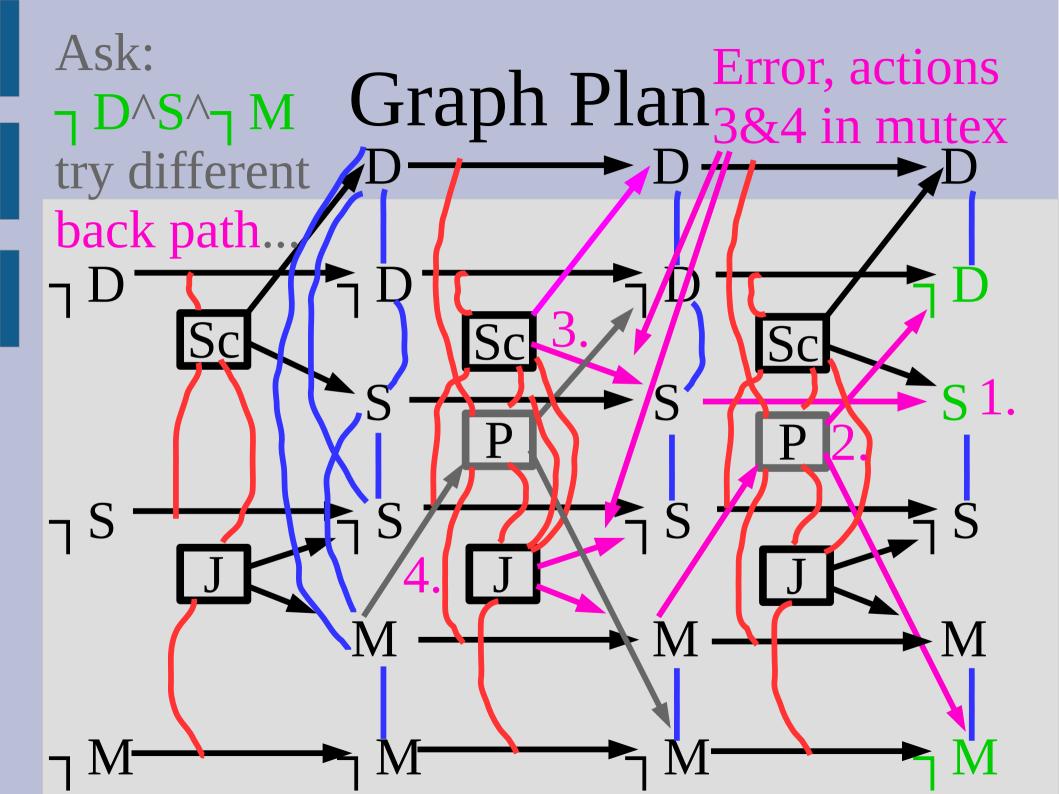
Graph Plan

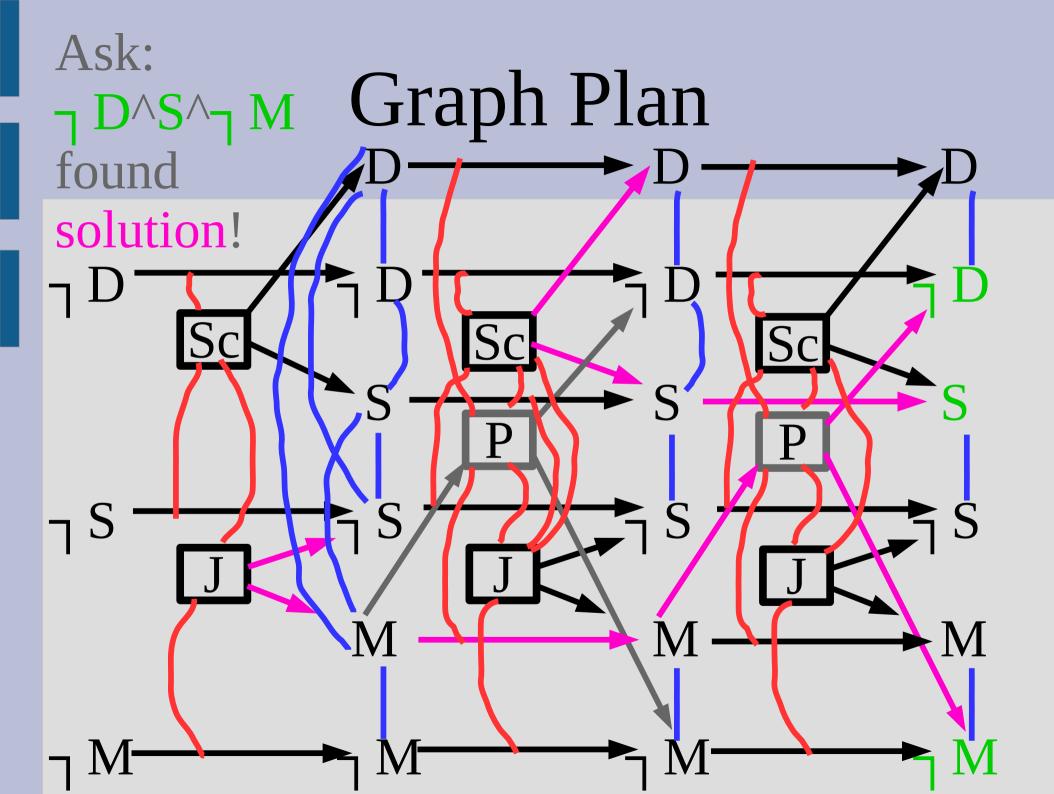
Remember this... Initial: $\neg Money \land \neg Smart \land \neg Debt$ Goal: $\neg Money \land Smart \land \neg Debt$ Action(School, Action (Job, Precondition:, Precondition: , Effect: $Debt \land Smart$) Effect: $Money \land \neg Smart$) Action(Pay, Precondition: Money,

Effect: $\neg Money \land \neg Debt$)









Finding a solution

Formally, the algorithm is:

graph = initial noGoods = empty table (hash) for level = 0 to infinity if all goal pairs not in mutex solution = recursive search with noGoods if success, return paths if graph & noGoods converged, return fail graaph = expand graph

