

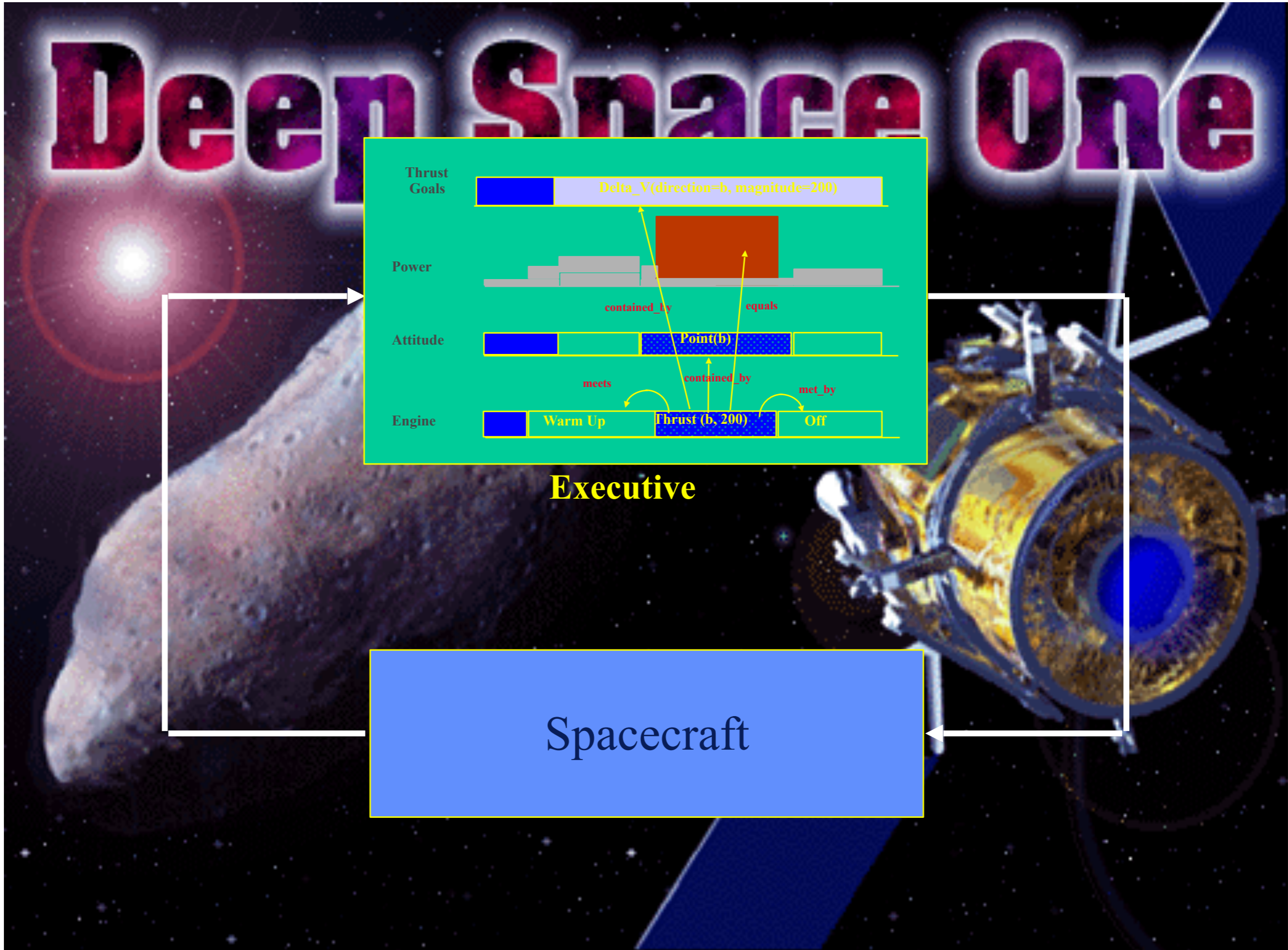
Deliberation with Temporal Models



Automated Planning and Acting

Malik Ghallab, Dana Nau
and Paolo Traverso

Planning and Acting Success Stories

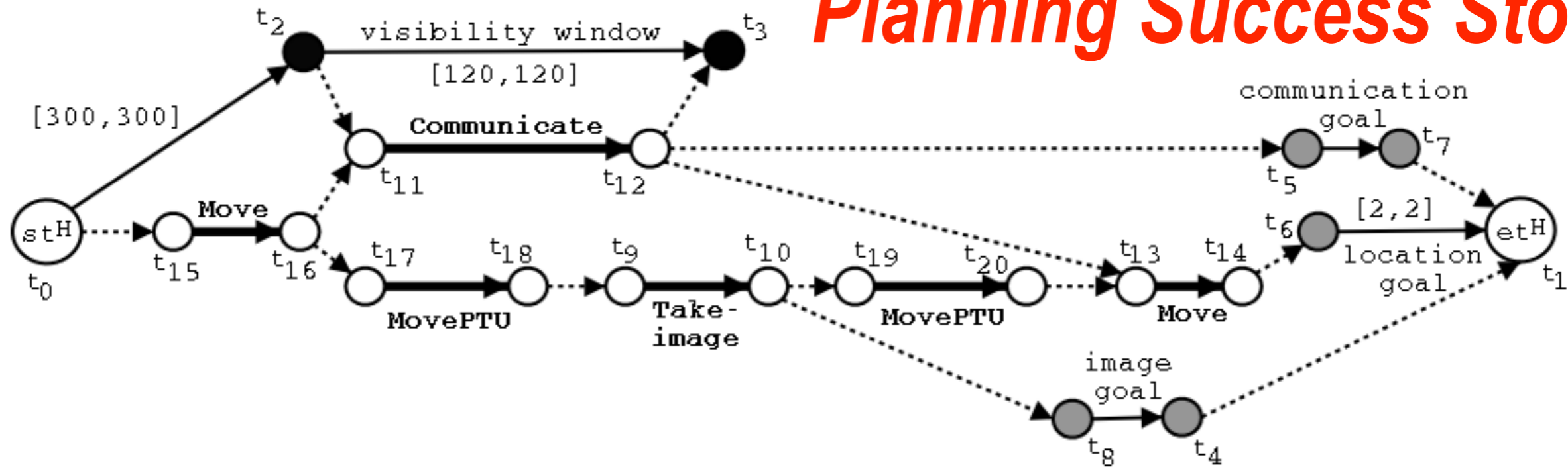


Planning Success Stories



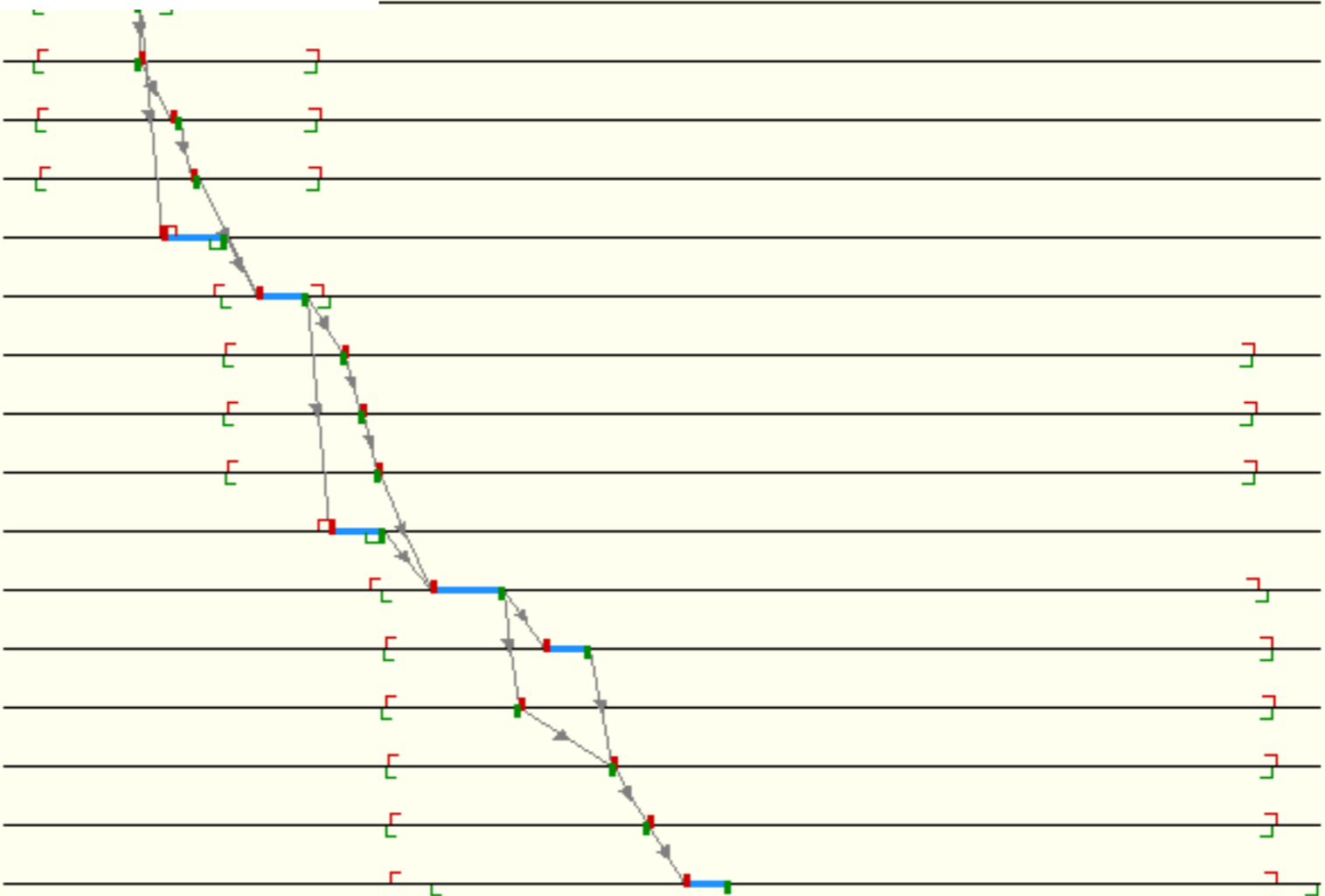
[IxTeT, LAAS]

Planning Success Stories

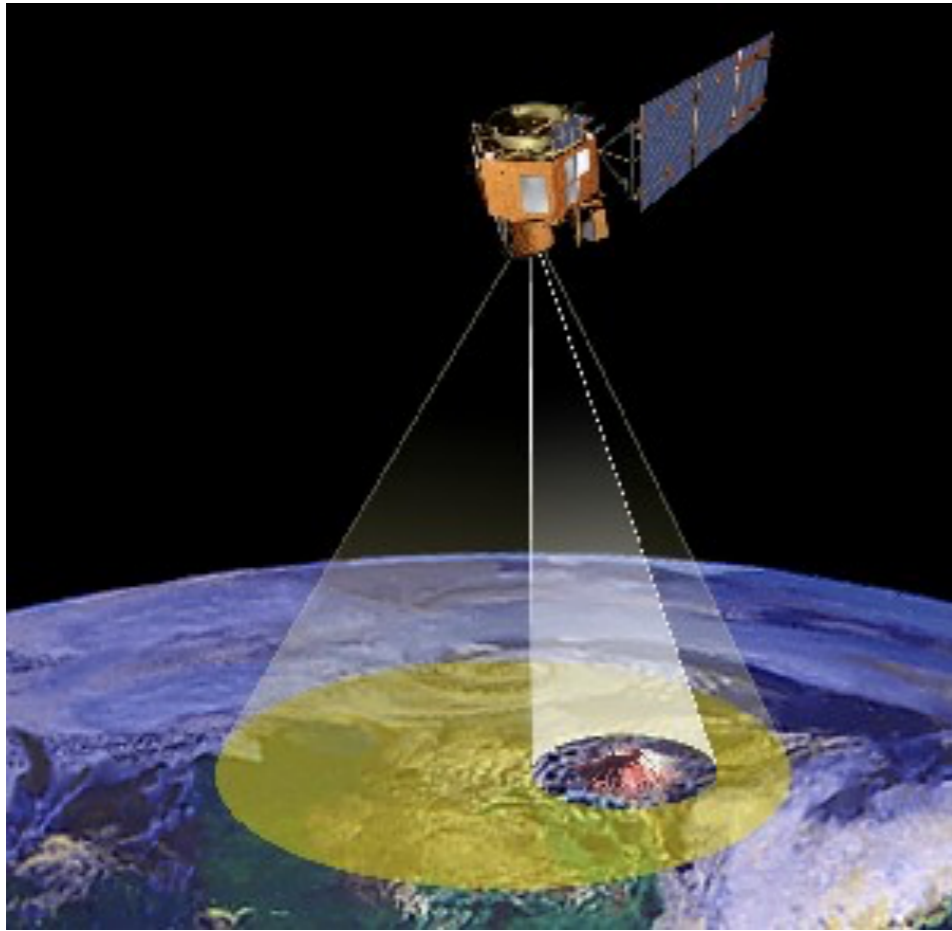


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MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ2, 9.000000, -0.500000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
COMMUNICATE (W1)
MOVE (9.000000, -0.500000, 10.000000, -3.000000)
MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ4, 10.000000, -3.000000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
COMMUNICATE (W2)
MOVE (10.000000, -3.000000, 8.000000, -5.000000)
DOWNLOAD_IMAGES ()
MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ3, 8.000000, -5.000000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
MOVE (8.000000, -5.000000, 0.500000, -0.500000)
    
```



Planning Success Stories



[*Casper, JPL*]



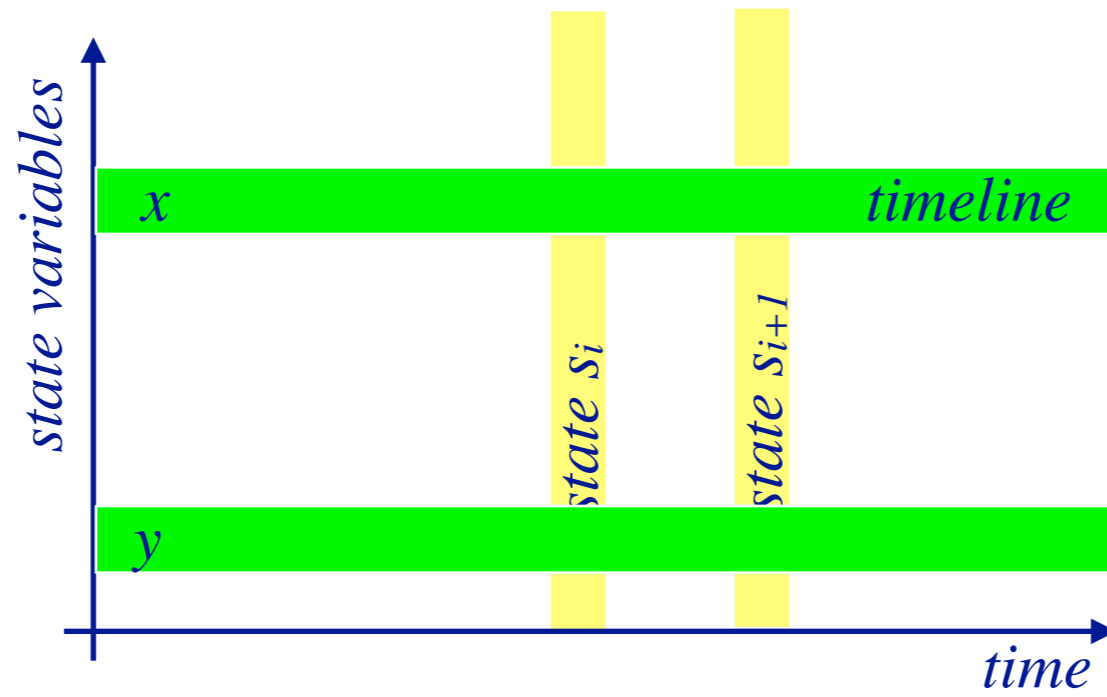
[*T-ReX, MBARI*]

Common point to these success stories:
explicit representation of time

Motivations for Temporal Models

- ▶ *Duration* of actions
- ▶ *Delayed effects*, conditions, and resources borrowed or consumed at various moments along an action duration
- ▶ *Timed goals* with relative or absolute temporal constraints
- ▶ *Exogenous events* expected to occur in the future time
- ▶ *Maintenance actions*: maintain a property (\neq changing a value), e.g., tracking a moving target, keeping a spring latch in position
- ▶ *Concurrency* of actions with interacting and joint effects
- ▶ *Delayed commitment*: instantiation at acting time

Temporal Models

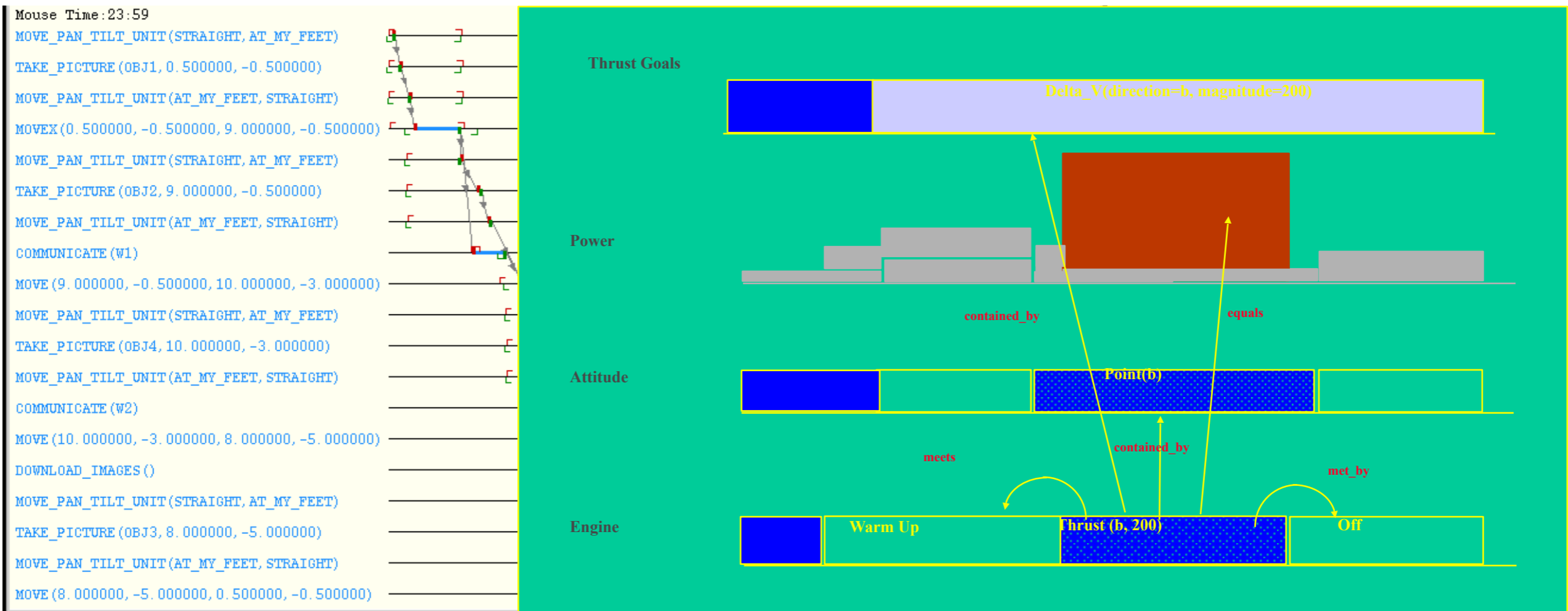


Motivations for Temporal Models

	States	Timelines
Duration of actions	✓	✓
Delayed effects	✓	✓
Timed goals	✓	✓
Exogenous events	~	✓
Maintenance actions	✓	✓
Concurrency	—	✓
Delayed commitment	—	✓

Timeline

- A set of constraints on state variables and events
- Reflects *predicted* actions and events
- Timeline planning akin to constraint-based planning



- ✓ Introduction
- ▶ Representation
 - Timelines
 - Actions and tasks
 - Chronicles
- ▶ Temporal planning
- ▶ Consistency and controllability
- ▶ Acting with executable primitives
- ▶ Acting with atemporal refinement
- ▶ Conclusion

Representation

- ▶ Quantitative discrete model of time
 - variables referring to time points
 - simple constraints
- ▶ Temporal assertions
 - *persistance* over an interval
 - *change* over an interval

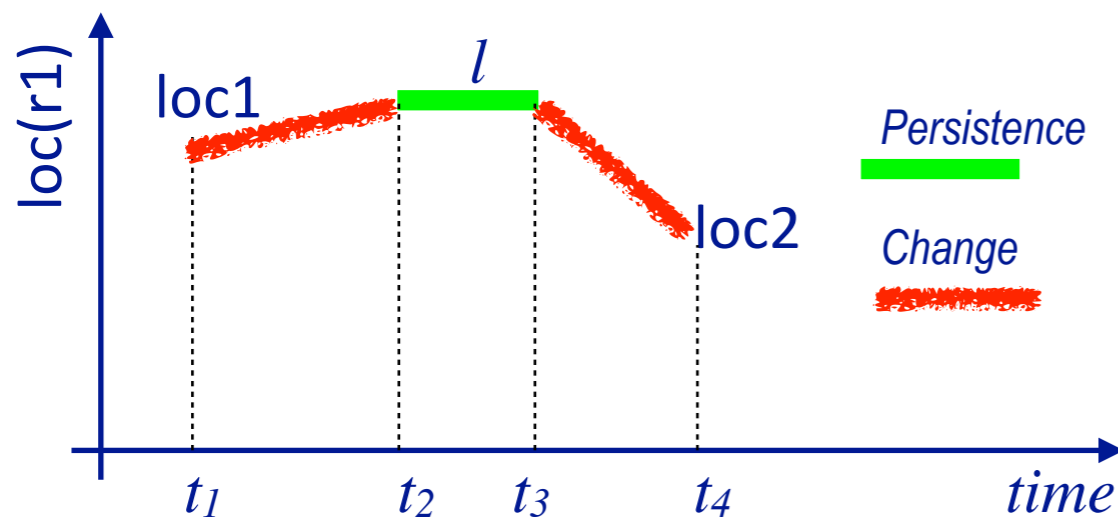
$$d \leq t' - t \leq d'$$

$$[t_1, t_2]x = v$$

$$[t_1, t_2]x:(v_1, v_2)$$

Partially predicted evolution of a state variable: a pair $(\mathcal{T}, \mathcal{C})$

- \mathcal{T} : temporal assertions
- \mathcal{C} : constraints



$$[t_1, t_2] \text{loc}(r1) : (\text{loc1}, l)$$

$$[t_2, t_3] \text{loc}(r1) = l$$

$$[t_3, t_4] \text{loc}(r1) : (l, \text{loc2})$$

$$t_1 < t_2 < t_3 < t_4$$

$$l \neq \text{loc1}$$

$$l \neq \text{loc2}$$

To restrict the value of $\text{loc}(r1)$ in $[t_1, t_2]$

$$[t_1, t_1 + 1] \text{loc}(r1) : (\text{loc1}, \text{route})$$

$$[t_2 - 1, t_2] \text{loc}(r1) : (\text{route}, l)$$

$$[t_1 + 1, t_2 - 1] \text{loc}(r1) = \text{route}$$

Consistent and Secure Timeline

- ▶ A ground instance of $(\mathcal{T}, \mathcal{C})$ is consistent if it satisfies \mathcal{C} and no state variable in \mathcal{T} has more than one value at the same time
- ▶ $(\mathcal{T}, \mathcal{C})$ is *consistent* if it has a consistent ground instance
- ▶ $(\mathcal{T}, \mathcal{C})$ is *secure* if it is consistent and every ground instance that satisfies \mathcal{C} is consistent

$$[t_1, t_2] \text{loc}(r) = \text{loc1}$$

$$[t_2, t_3] \text{loc}(r) : (\text{loc1}, \text{loc2})$$

$$t_1 < t_2 < t_3$$

secure timeline

Consistent and Secure Timeline

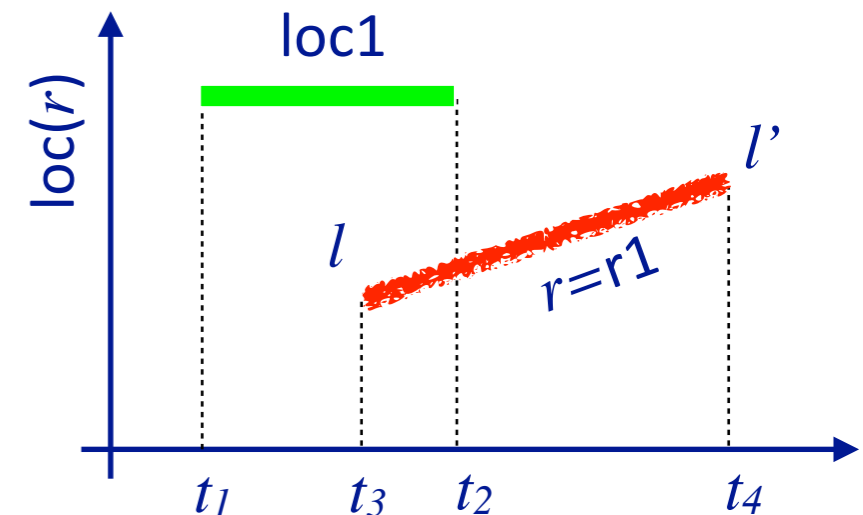
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- ▶ $(\mathcal{T}, \mathcal{C})$ is *secure* if it is consistent and every ground instance that satisfies \mathcal{C} is consistent

$$[t_1, t_2] \text{loc}(r) = \text{loc1}$$

$$[t_3, t_4] \text{loc}(r1) : (l, l')$$

$$t_1 < t_2, t_3 < t_4$$

timeline consistent
but not secure



Conflicting assertions

=> separation constraints

$$r \neq r1$$

$$t_2 < t_3$$

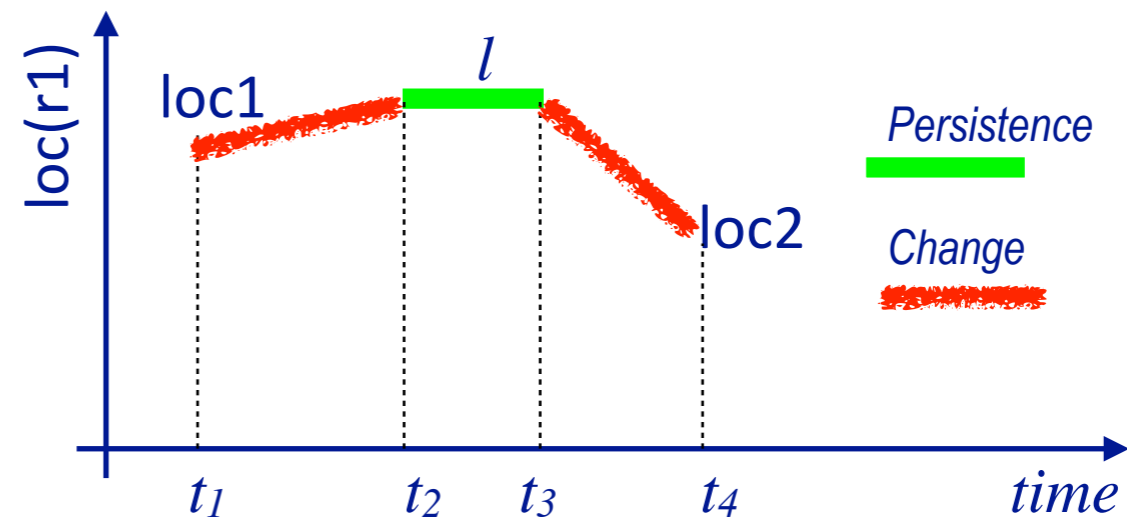
$$t_4 < t_1$$

$$t_2 = t_3, r = r1, l = \text{loc1}$$

$$t_4 = t_1, r = r1, l' = \text{loc1}$$

Causally supported timeline

- ▶ *Causal support* of the value of x : reasons that substantiate it
 - Prior knowledge about current state or dynamics of environment
 - Observation
 - Prediction of actions effects



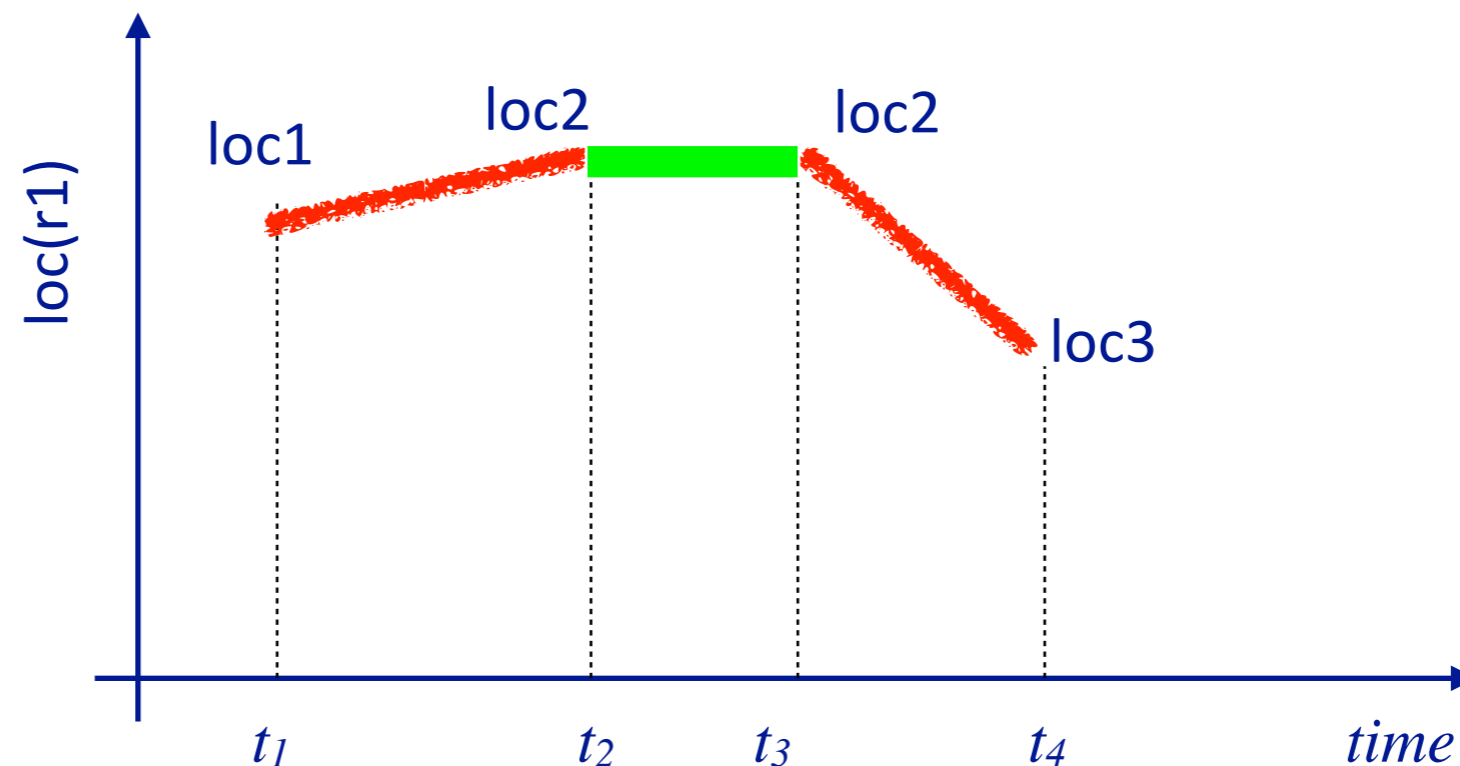
- ▶ Causally supported timeline:
all its assertions have a causal support

Finding causal support

► Adding a persistence assertion

$[t_1, t_2] \text{loc}(r1):(\text{loc1}, \text{loc2}), [t_3, t_4] \text{loc}(r1):(\text{loc2}, \text{loc3})$
 $t_1 < t_2 < t_3 < t_4$

$[t_2, t_3] \text{loc}(r1) = \text{loc2}$

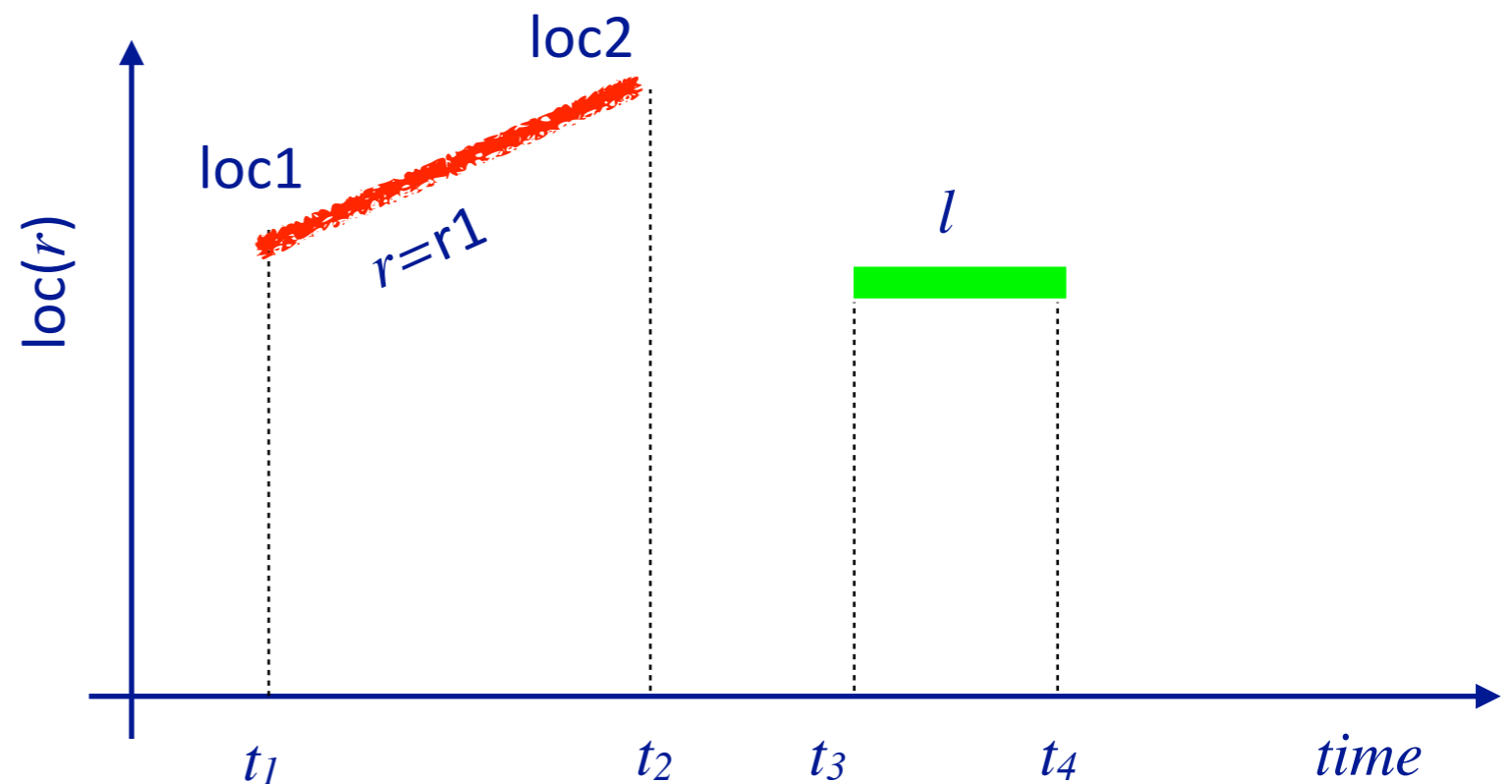


Finding causal support

- ▶ Adding a persistence assertion
- ▶ Adding constraints

$[t_1, t_2] \text{loc}(r1):(\text{loc1}, \text{loc2}), [t_3, t_4] \text{loc}(r) = l$
 $t_1 < t_2 < t_3 < t_4$

$t_2 = t_3, r = r1, l = \text{loc2}$

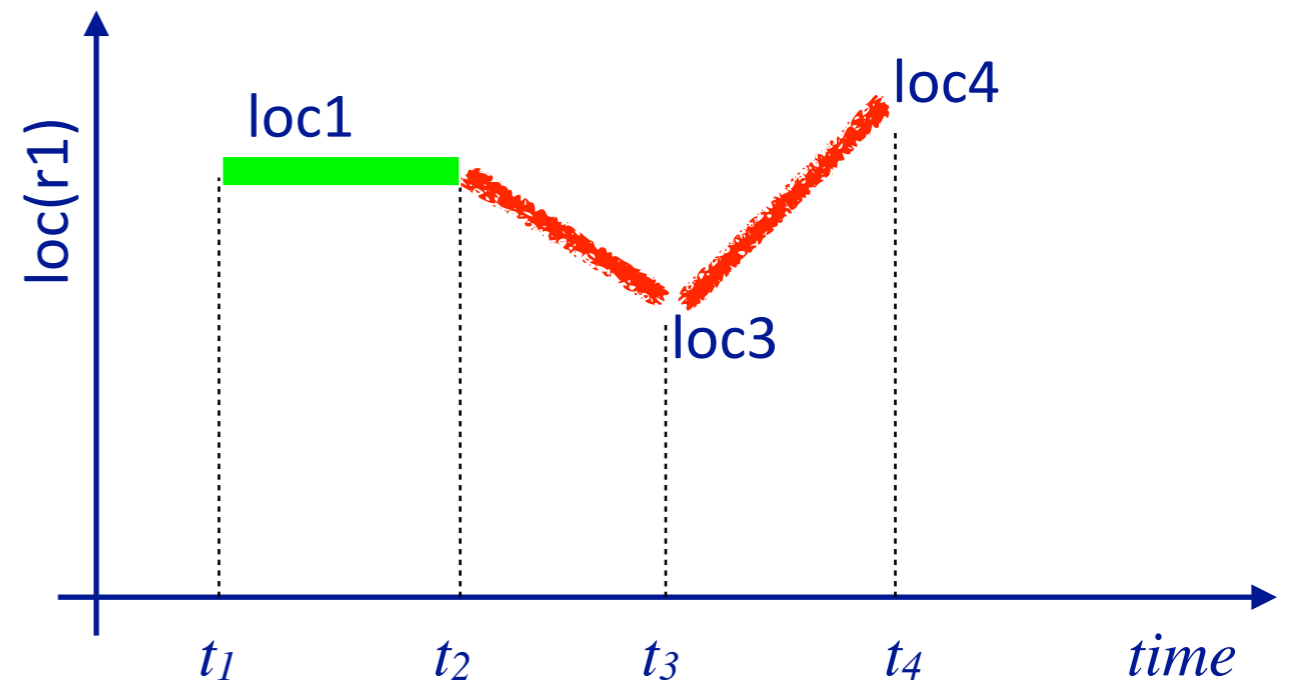


Finding causal support

- ▶ Adding a persistence assertion
- ▶ Adding constraints
- ▶ Adding a change assertion → corresponds to an additional action

$[t_1, t_2] \text{loc}(r1) = \text{loc1}$, $[t_3, t_4] \text{loc}(r1) : (\text{loc3}, \text{loc4})$
 $t_1 < t_2 < t_3 < t_4$

$[t_2, t_3] \text{loc}(r1) : (\text{loc1}, \text{loc3})$



Example

▶ Domain objects

$r \in Robots, k \in Cranes, c \in Containers$
 $p \in Piles, d \in Docks, w \in Waypoints$

▶ State variables

$loc(r) \in Docks \cup Waypoints$ for $r \in Robots$

$freight(r) \in Containers \cup \{empty\}$ for $r \in Robots$

$grip(k) \in Containers \cup \{empty\}$ for $k \in Cranes$

$pos(c) \in Robots \cup Cranes \cup Piles$ for $c \in Containers$

$stacked-on(c) \in Containers \cup \{empty\}$ for $c \in Containers$

$top(p) \in Containers \cup \{empty\}$ for $p \in Piles$

$occupant(d) \in Robots \cup \{empty\}$ for $d \in Docks$

▶ Rigid relations

$attached \subseteq (Cranes \cup Piles) \times Docks$

$adjacent \subseteq Docks \times Waypoints$

$connected \subseteq Waypoints \times Waypoints$

► Primitive actions

$\text{leave}(r, d, w)$: robot r leaves dock d to an adjacent waypoint w

$\text{enter}(r, d, w)$: r enters d from an adjacent waypoint w

$\text{navigate}(r, w, w')$: r navigates from waypoint w to a connected one w'

$\text{stack}(k, c, p)$: crane k holding container c stacks it on top of pile p

$\text{unstack}(k, c, p)$: crane k unstacks a container c from the top of pile p

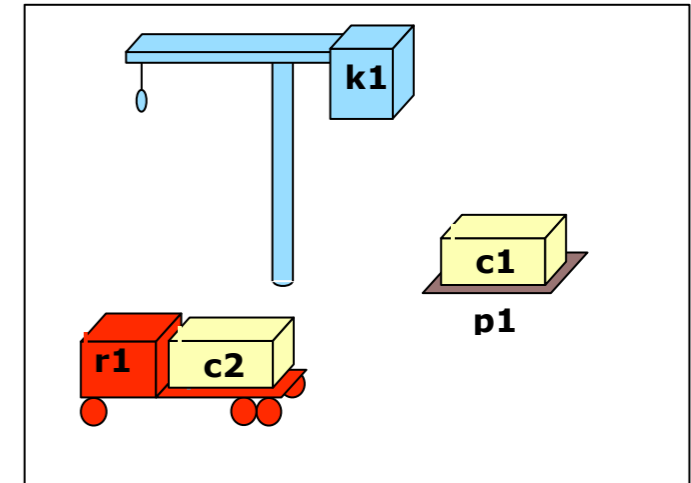
$\text{put}(k, c, r)$: crane k holding a container c and puts it onto r

$\text{take}(k, c, r)$: crane k takes container c from robot r

take(k, c, r)

assertions: $[t_s, t_e] \text{pos}(c):(r, k)$
 $[t_s, t_e] \text{grip}(k):(\text{empty}, c)$
 $[t_s, t_e] \text{freight}(r):(c, \text{empty})$
 $[t_s, t_e] \text{loc}(r)=d$

constraints: $\text{attached}(k, d)$



leave(r, d, w)

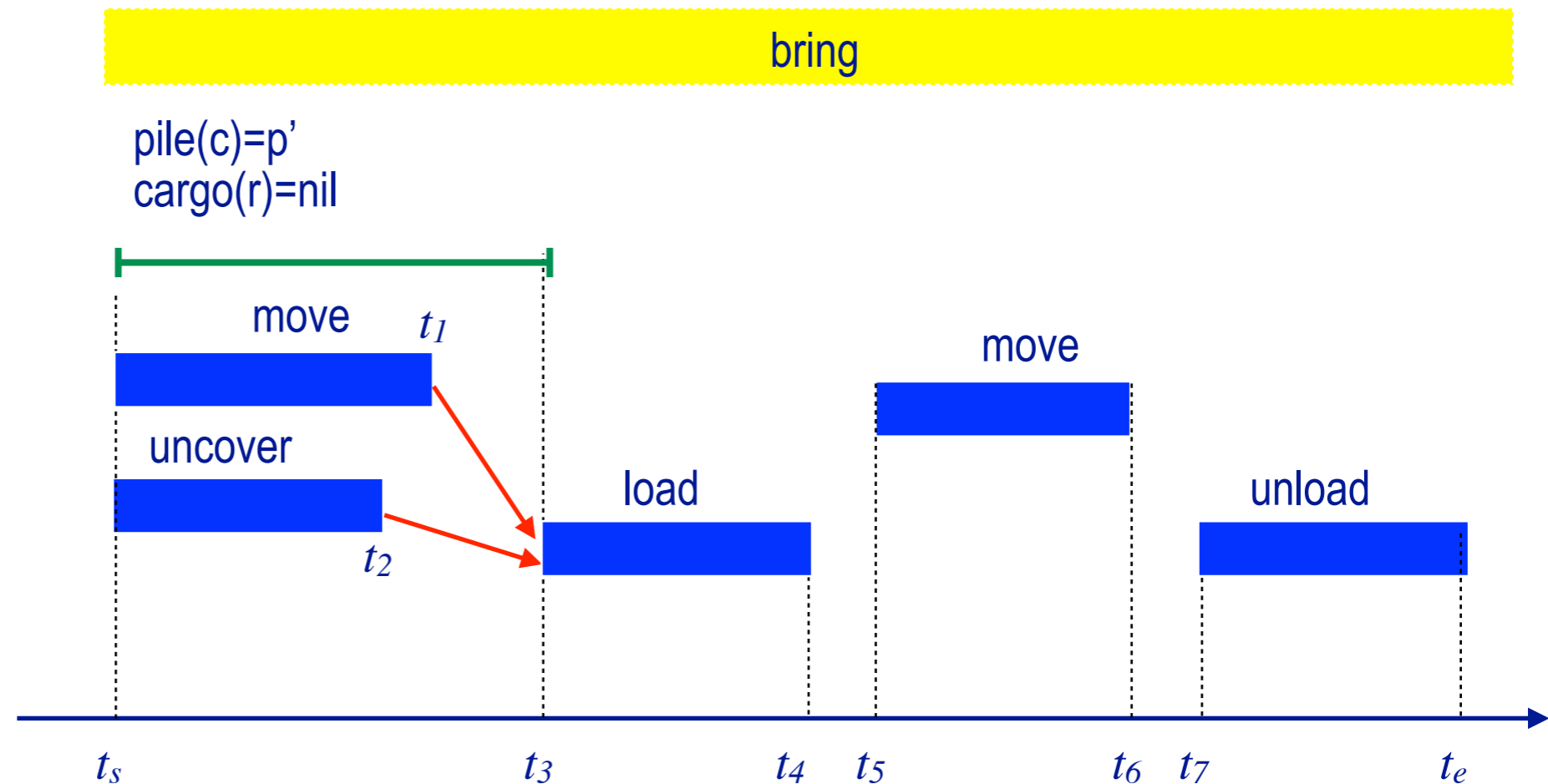
assertions: $[t_s, t_e] \text{loc}(r):(d, w)$
 $[t_s, t_e] \text{occupant}(d):(r, \text{empty})$

constraints: $t_e \leq t_s + \delta_1$
 $\text{adjacent}(d, w)$

Tasks and methods

► Tasks

$[t_s, t_e]$ bring(r, c, p)



$[t_s, t_1]$ move(r, d)

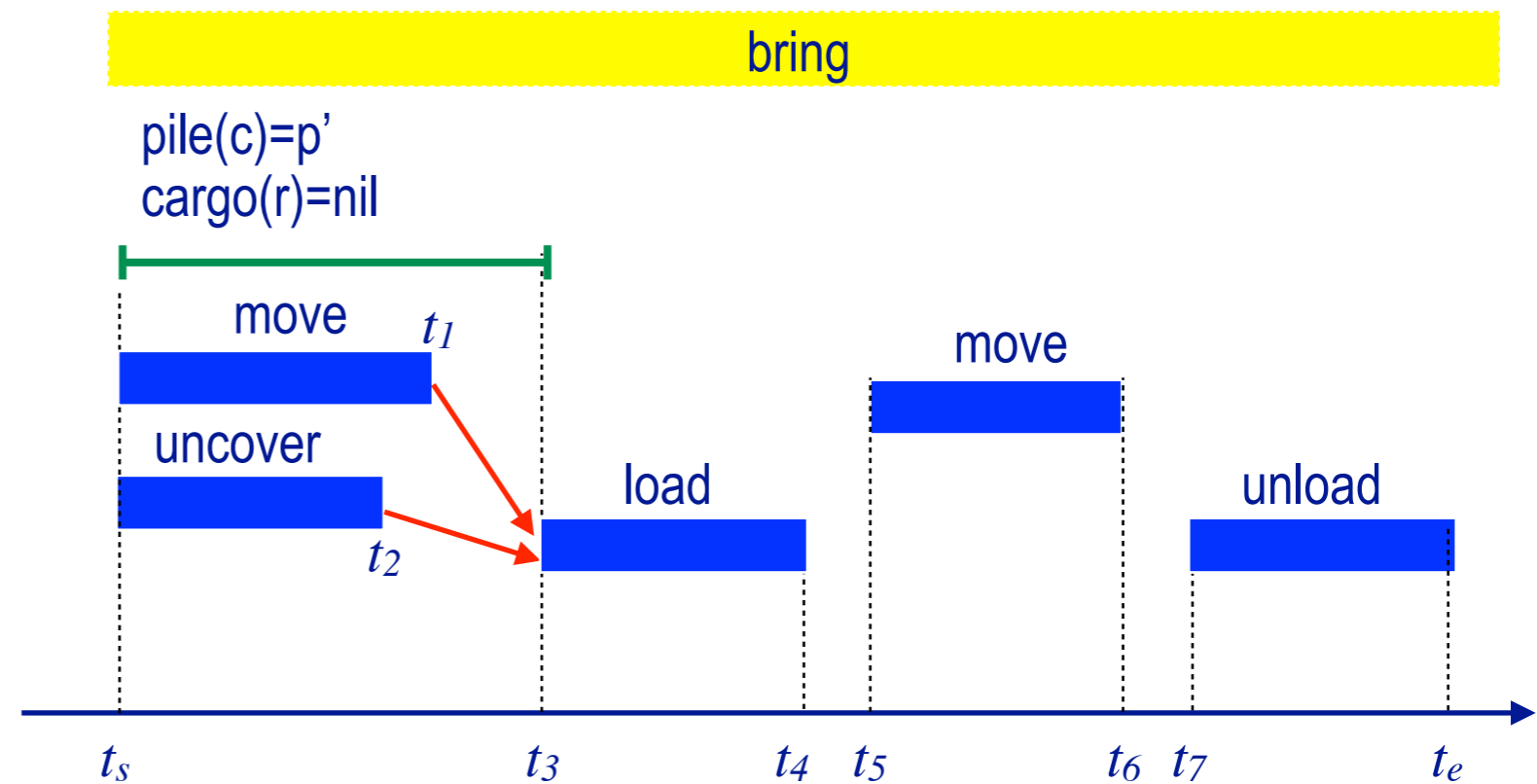
$[t_s, t_2]$ uncover(k, c, p)

$[t_3, t_4]$ load(k, r, c, p)

$[t_7, t_e]$ unload(k, r, c, p)

Tasks and methods

► Methods



m-bring($r, c, p, p', d, d', k, k'$)

task: bring(r, c, p)

refinement: $[t_s, t_1]$ move(r, d')

$[t_s, t_2]$ uncover(c, p')

$[t_3, t_4]$ load(k', r, c, p')

$[t_5, t_6]$ move(r, d)

$[t_7, t_e]$ unload(k, r, c, p)

assertions: $[t_s, t_3]$ pile(c)= p'

$[t_s, t_3]$ freight(r)=empty

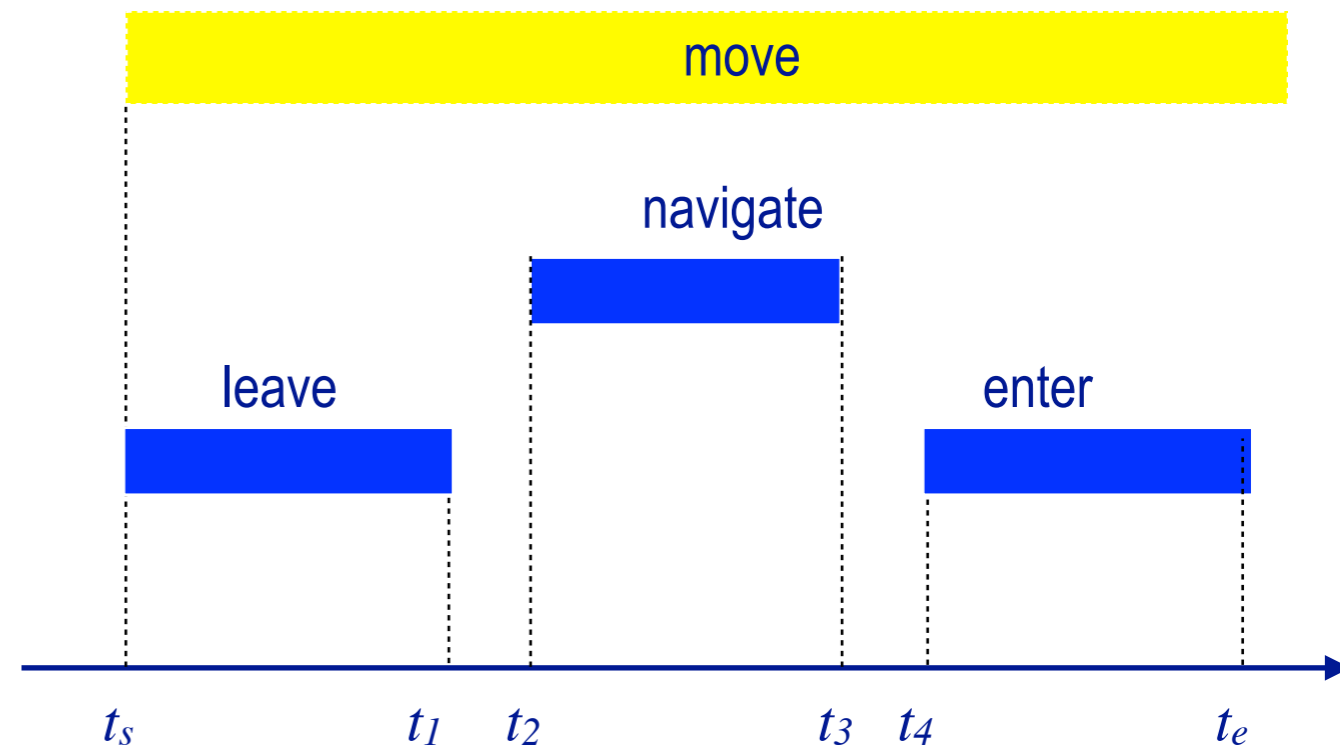
constraints: attached(p', d'), attached(p, d), $d \neq d'$

attached(k', d'), attached(k, d)

$t_1 \leq t_3$, $t_2 \leq t_3$, $t_4 \leq t_5$, $t_6 \leq t_7$

Tasks and methods

► Methods



$m\text{-move1}(r, d, d', w, w')$

task: $\text{move}(r, d)$

refinement: $[t_s, t_1]\text{leave}(r, d', w')$

$[t_2, t_3]\text{navigate}(w', w)$

$[t_4, t_e]\text{enter}(r, d, w)$

assertions: $[t_s, t_s + 1]\text{loc}(r) = d'$

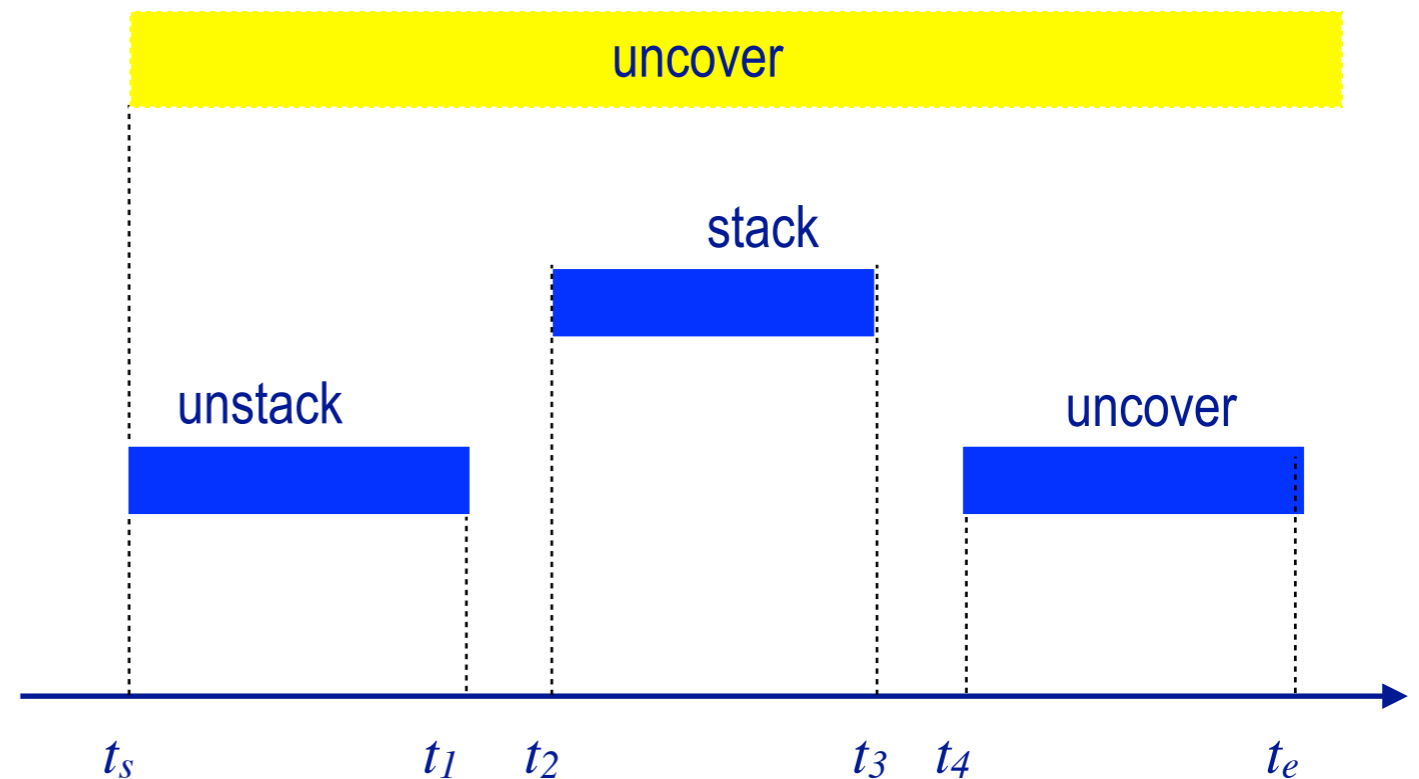
constraints: $\text{adjacent}(d, w)$, $\text{adjacent}(d', w')$, $d \neq d'$

$\text{connected}(w, w')$

$t_1 \leq t_2$, $t_3 \leq t_4$

Tasks and methods

► Methods



$m\text{-uncover}(c, p, k, d, p')$

task: $\text{uncover}(c, p)$

refinement: $[t_s, t_1]\text{unstack}(k, c', p)$

$[t_2, t_3]\text{stack}(k, c', p')$

$[t_4, t_e]\text{uncover}(c, p)$

assertions: $[t_s, t_s + 1]\text{pile}(c) = p$

$[t_s, t_s + 1]\text{top}(p) = c'$

$[t_s, t_s + 1]\text{grip}(k) = \text{empty}$

constraints: $\text{attached}(k, d), \text{attached}(p, d)$

$\text{attached}(p', d), p \neq p', c' \neq c$

$t_1 \leq t_2, t_3 \leq t_4$

- ▶ Chronicle $\phi = (\mathcal{A}, \mathcal{S}_{\mathcal{T}}, \mathcal{T}, \mathcal{C})$
 - \mathcal{A} : temporally qualified actions and tasks
 - $\mathcal{S}_{\mathcal{T}}$: a priori supported assertions
 - \mathcal{T} : temporally qualified assertions
 - \mathcal{C} : constraints
- ▶ ϕ represents
 - Current state and future predicted events
 - Tasks to be performed
 - Assertions and constraints to be satisfied
 - => planning problems and (partial) plans

Chronicles

ϕ_0 :

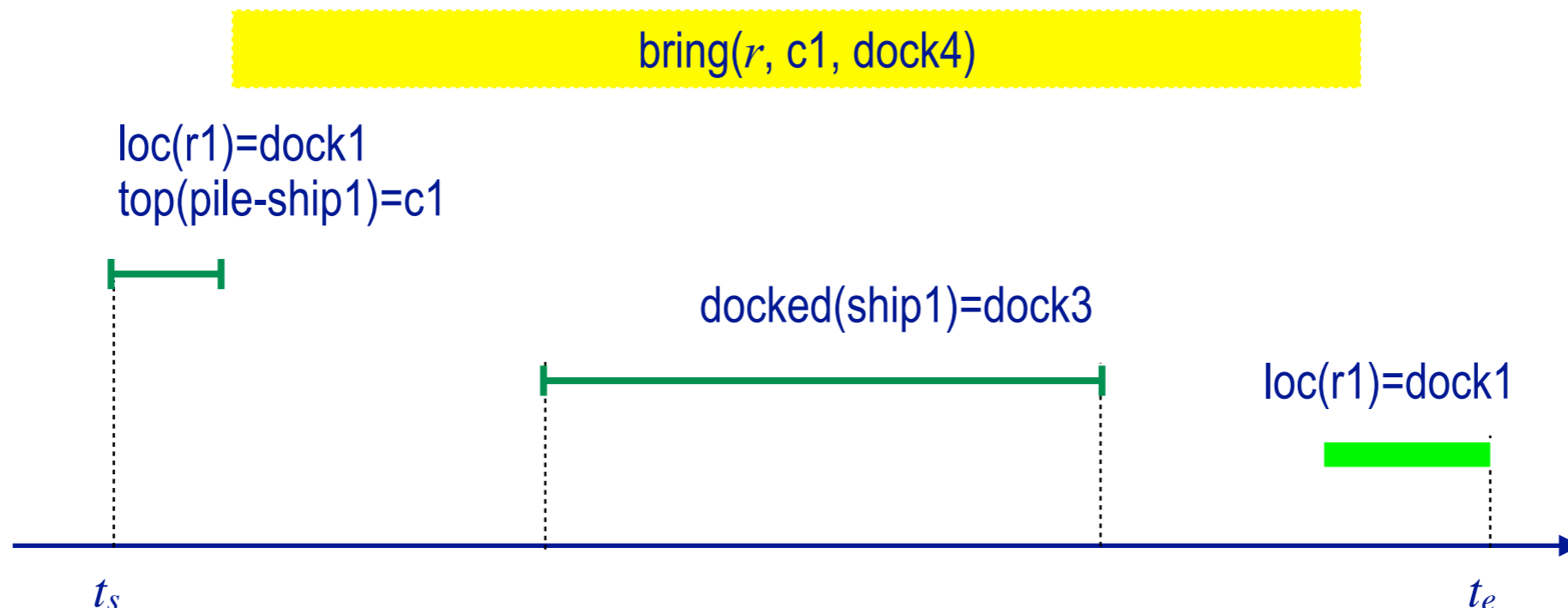
tasks: $[t, t']$ bring($r, c1, dock4$)

supported: $[t_s]$ loc($r1$)= $dock1$
 $[t_s]$ loc($r2$)= $dock2$
 $[t_s + 10, t_s + \delta]$ docked($ship1$)= $dock3$
 $[t_s]$ top($pile1$)= $c1$
 $[t_s]$ pos($c1$)= $pallet$

assertions: $[t_e]$ loc($r1$) = $dock1$
 $[t_e]$ loc($r2$) = $dock2$

constraints: attached($pile1, ship1$)
 $t_s < t < t' < t_e$, $20 \leq \delta \leq 30$, $t_s = 0$

Initial chronicle



► Partial plan

ϕ :

tasks: $[t_0, t_1]$ leave(r1,dock1,w1)
 $[t_1, t_2]$ navigate(r1,w1,w2)
 $[t_3, t_4]$ enter(r1,dock2,w2)
 $[t'_0, t'_1]$ leave(r2,dock2,w2)
 $[t'_1, t'_2]$ navigate(r2,w2,w1)
 $[t'_3, t'_4]$ enter(r2,dock1,w1)

supported: $\mathcal{S}_{\mathcal{T}}$

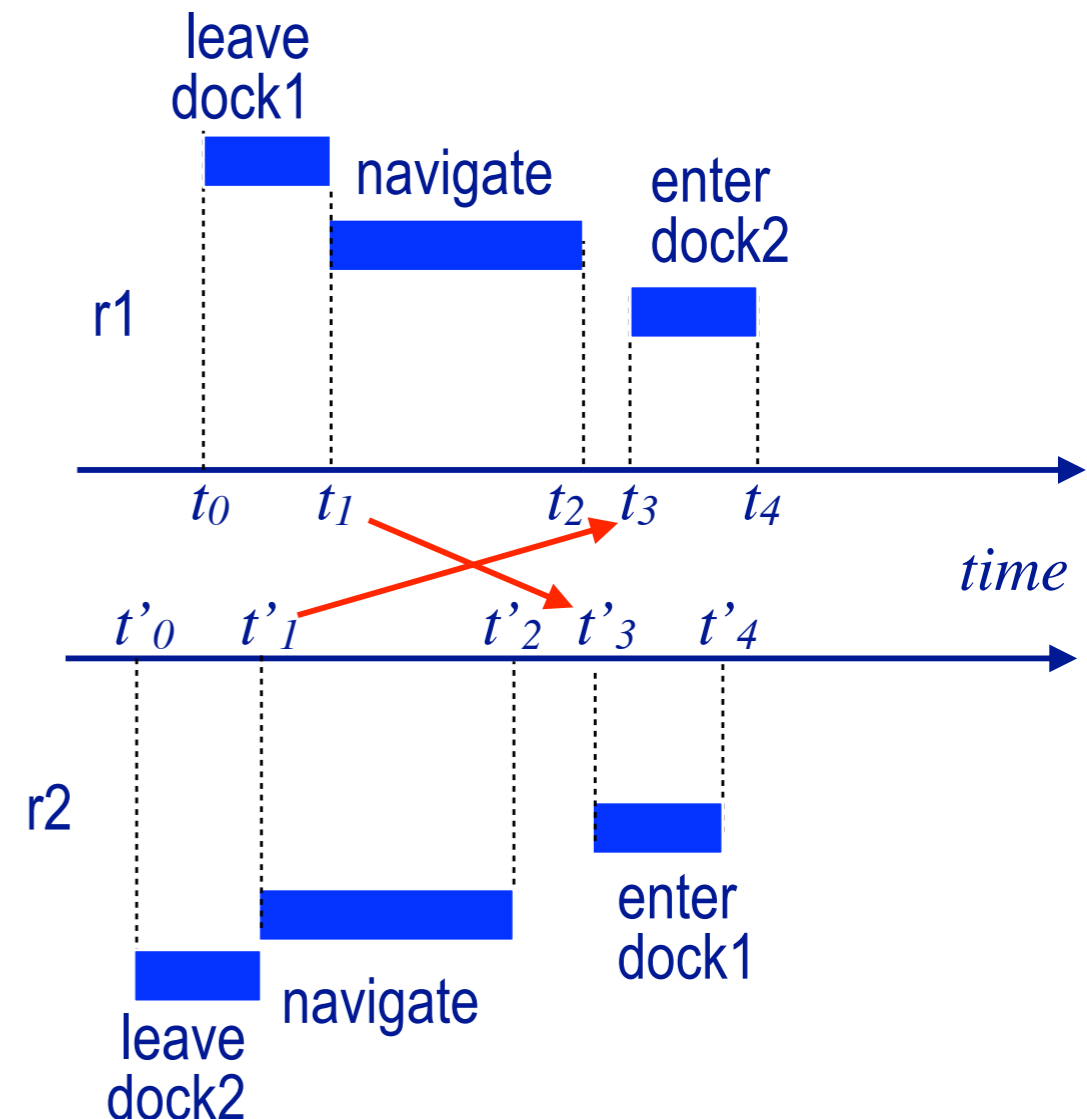
assertions: \mathcal{T}

constraints: $t'_1 < t_3$, $t_1 < t'_3$, $t_s < t_0$

$t_s < t'_0$, $t_4 < t_e$, $t'_4 < t_e$

adjacent(dock1,w1), adjacent(dock2,w2)

connected(w1,w2)



- ✓ Introduction
- ✓ Representation
- ▶ Temporal planning
 - Resolvers and flaws
 - Search space
- ▶ Consistency and controllability
- ▶ Acting with executable primitives
- ▶ Acting with atemporal refinement
- ▶ Conclusion

Temporal Planning

Starting from an *initial chronicle*:

- ▶ Refine into primitive actions
- ▶ Add causal supports to
- ▶ Add separation constraints for

nonrefined tasks

nonsupported assertion

conflicting assertions

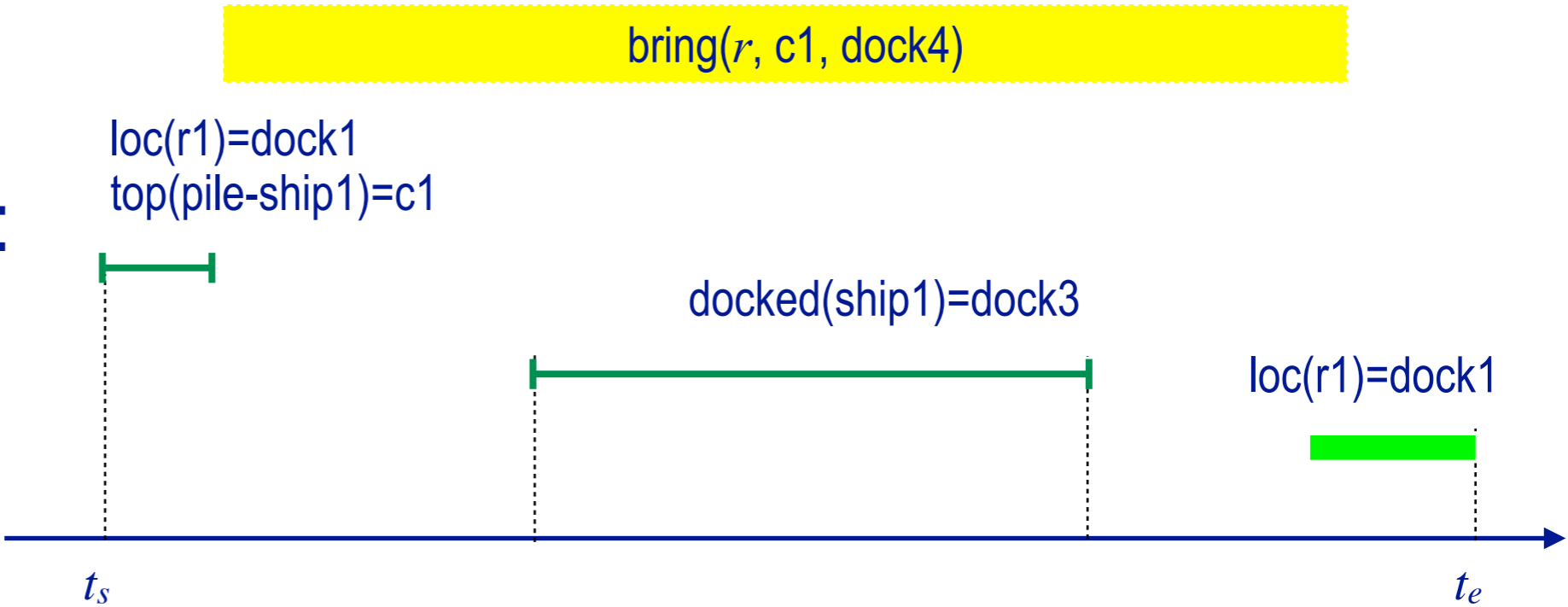
flaws

Resolvers of flaws:

▶ Methods

▶ Actions

▶ Persistence assertions and constraints



Temporal Planning

A chronicle ϕ is a valid solution plan iff

- ϕ does not contain nonrefined tasks
- all assertions in ϕ are causally supported, either by supported assertions initially in ϕ_0 or by assertions from methods and primitives in the plan
- the chronicle ϕ is secure

*no conflicting assertions in
consistant instances*

TemPlan(ϕ, Σ)

$Flaws \leftarrow$ set of flaws of ϕ

if $Flaws = \emptyset$ then return ϕ

arbitrarily select $f \in Flaws$

$Resolvers \leftarrow$ set of resolvers of f

if $Resolvers = \emptyset$ then return failure

nondeterministically choose $\rho \in Resolvers$

$\phi \leftarrow \text{Transform}(\phi, \rho)$

Templan(ϕ, Σ)

Combines in CSP-based approach

- ▶ *task decomposition planning*
- ▶ *plan-space planning*
- ▶ *temporal planning*

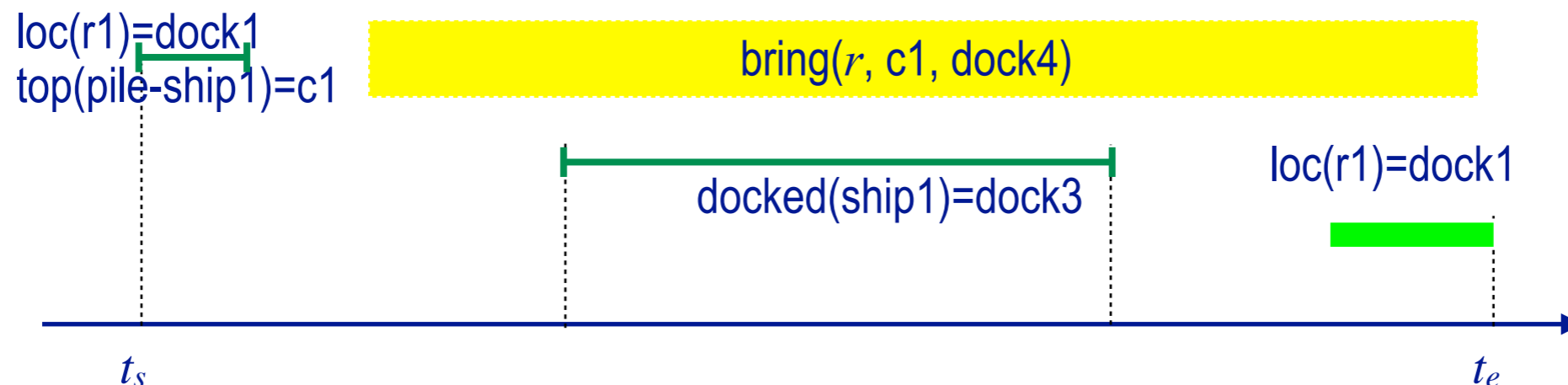
Resolvers for flaws

Resolver for a **nonrefined task** in ϕ :

an instance m of a method applicable to the *task* s.t. all the constraints of m are consistent with those of ϕ .

Transforming $\phi = (\mathcal{A}, \mathcal{S}_{\mathcal{T}}, \mathcal{T}, \mathcal{C})$ with resolver m :

- replace in \mathcal{A} the task by the subtasks and actions of m
- add the assertions of m and those of the primitives in m either to $\mathcal{S}_{\mathcal{T}}$ if these assertions are causally supported or to \mathcal{T}
- add to \mathcal{C} the constraints of m and those of its actions.

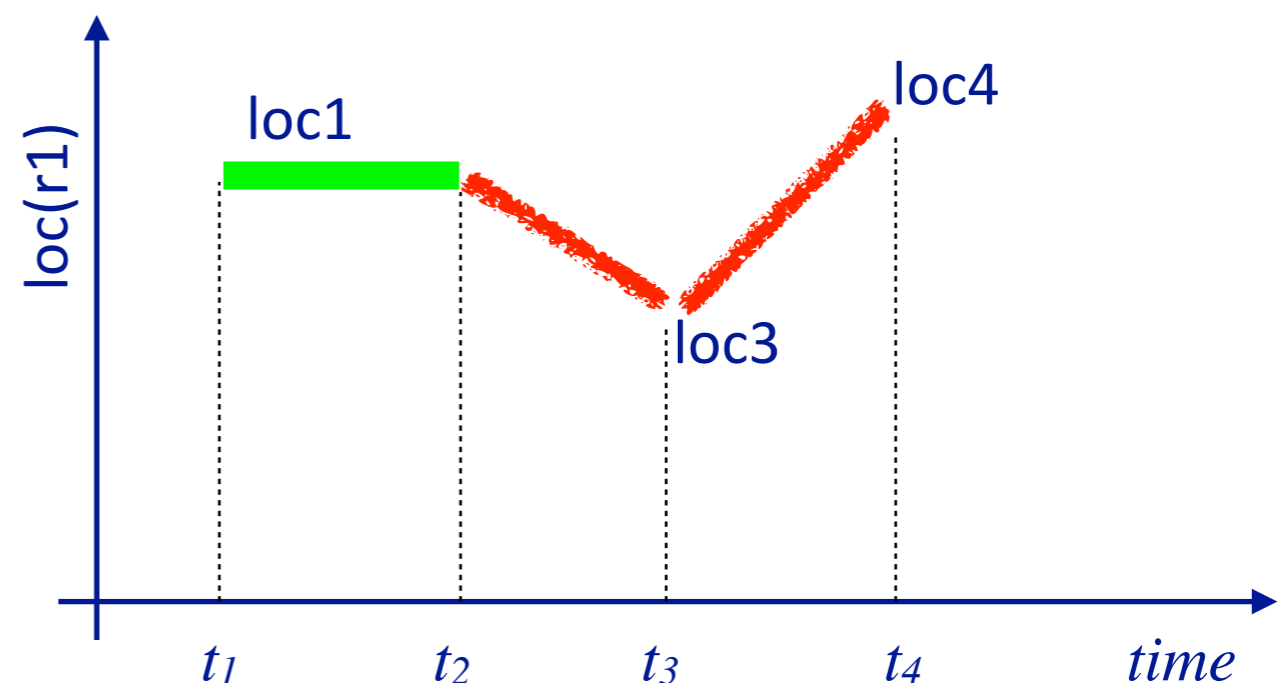


Resolvers for flaws

Nonsupported assertions in $\phi = (\mathcal{A}, \mathcal{S}_{\mathcal{T}}, \mathcal{T}, \mathcal{C})$: those initially in ϕ_0 plus those from the refinement of tasks and the insertion of actions

Different ways to support an assertion $\alpha \in \mathcal{T}$:

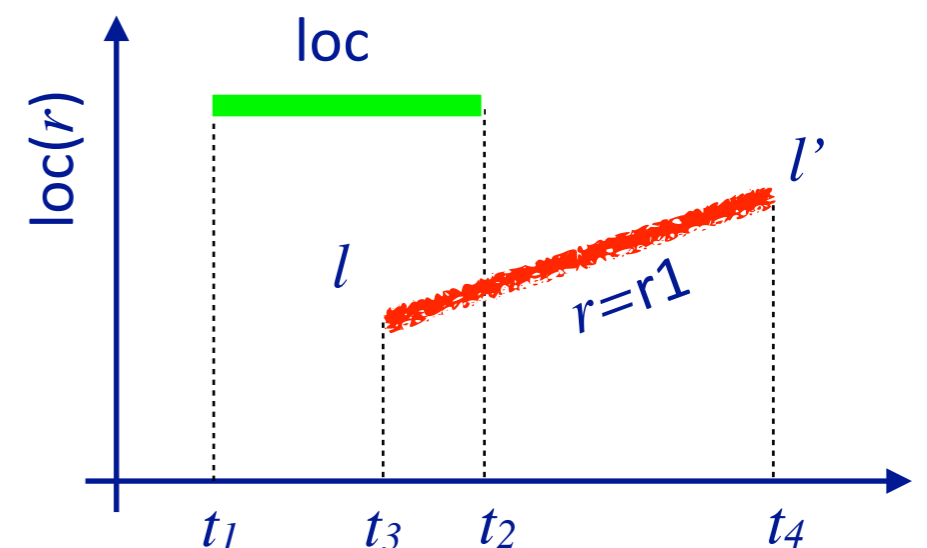
- add in \mathcal{C} constraints on object and temporal variables
- add a persistence assertion in $\mathcal{S}_{\mathcal{T}}$
- add in \mathcal{A} a task or an action that brings an assertion supporting α



Resolvers for flaws

Flaws due to **conflicting assertions** handled incrementally by maintaining ϕ a secure chronicle:

- Detect possible conflicts for each new assertion in ϕ
- Find sets of separation constraints consistent with the constraints in current ϕ
- Add separation constraints to ϕ



Search Space

- ▶ Planning search space: a directed graph
 - Node: a chronicle ϕ
 - Edge (ϕ, ϕ') : $\phi' = \text{Transform}(\phi, \varrho)$, ϱ resolver for flaw in ϕ
 - *Acyclic graph*
 - *Not necessarily finite*
- ▶ Heuristics
 - Flaw with smallest numbers of resolvers (as in *variable-ordering*)
 - Resolver the least constraining to current ϕ (as in *value-ordering*)
 - Elaborate heuristics based on domain transition graphs and reachability graphs
- ▶ Critical operation: maintain consistency of C

- ✓ Introduction
- ✓ Representation
- ✓ Temporal planning
- ▶ Consistency and controllability
 - Object constraints
 - Temporal constraint
 - Controllability of an STNU
- ▶ Acting with executable primitives
- ▶ Acting with atemporal refinement
- ▶ Conclusion

- ▶ Object constraints maintained by Templan

$$l \neq \text{loc}2, l \in \{\text{loc}3, \text{loc}4\}$$

$$r = r1, o \neq o'$$

$$\text{loc}(r) \neq l'$$

- ▶ Temporal constraints maintained by Templan

$$a < t$$

$$t < t'$$

$$a \leq t - t \leq b$$

- ▶ Possibly coupled constraints

$$t < f(l, r)$$

=> Assume no coupled constraint

Object Constraint

- ▶ Unary and binary constraints on object variables
due to binding and separation constraints and rigid relations
- ▶ Corresponds to maintaining the consistency of a general CSP over finite domains => NP-complete problem
- ▶ Incremental arc or path consistency algorithms
 - Not complete, but efficient trade-off for filtering inconsistent instances
 - Do not reduce the completeness of the algorithm, just prune fewer nodes in search tree
- ▶ Combined with complete algorithms, e.g., forward-checking on the free variables remaining in the final plan

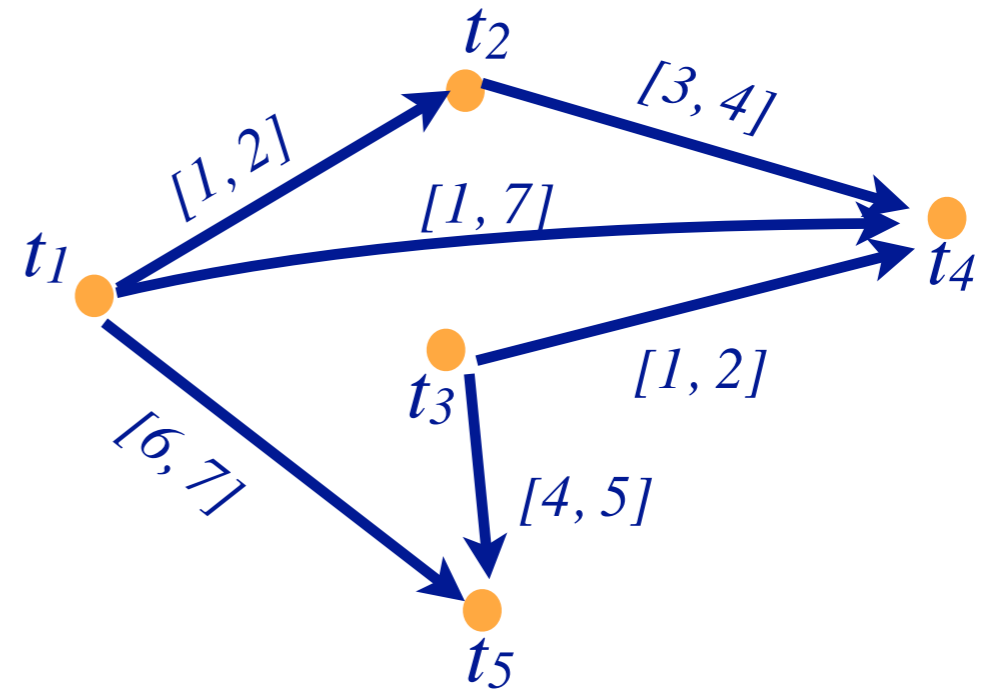
Temporal Constraints

- ▶ Simple temporal networks (STN)

$$a \leq t_j - t_i \leq b$$

notation $r_{ij} = [a, b]$

entails $r_{ji} = [-b, -a]$



STN

- ▶ Incrementally *synthesized* by Templan starting from ϕ_0
- ▶ Incrementally *instantiated* at acting time
- ▶ *Maintained consistent throughout planning and acting*

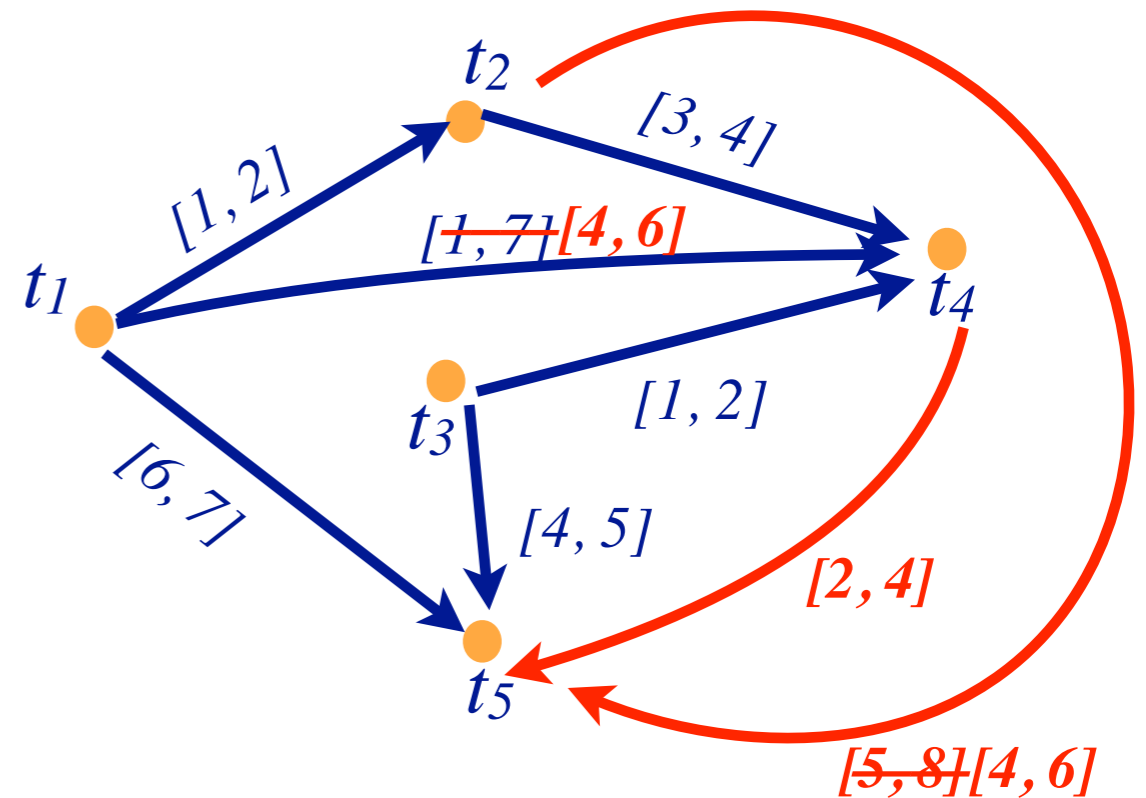
Temporal Constraints

Simple temporal networks (STN)

$$a \leq t_j - t_i \leq b$$

notation $r_{ij} = [a, b]$

entails $r_{ji} = [-b, -a]$



Constraint propagation rules

Conditions	Propagated constraint
$t_1 \xrightarrow{[a,b]} t_2, t_2 \xrightarrow{[a',b']} t_3$	$t_1 \xrightarrow{[a+a',b+b']} t_3$
$t_1 \xrightarrow{[a,b]} t_2, t_1 \xrightarrow{[a',b']} t_2$	$t_1 \xrightarrow{[\max\{a,a'\}, \min\{b,b'\}]} t_2$

$$r_{12} \bullet r_{23}$$

$$r_{12} \cap r'_{12}$$

Path consistency algorithm

$PC(\mathcal{V}, \mathcal{E})$

for each $k : 1 \leq k \leq n$ do

for each pair $i, j : 1 \leq i < j \leq n, i \neq k, j \neq k$ do

$r_{ij} \leftarrow r_{ij} \cap [r_{ik} \bullet r_{kj}]$

if $r_{ij} = \emptyset$ then return *inconsistent*

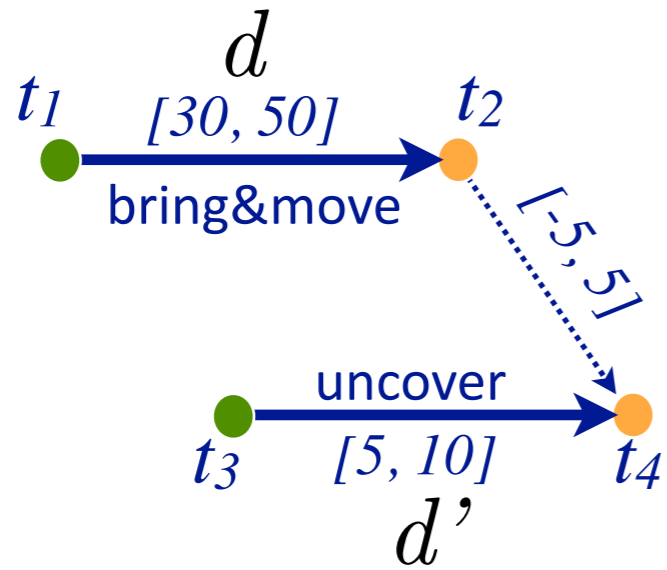
- ▶ Algorithm complete
- ▶ Returns a *minimal* network when consistent
- ▶ Complexity in time $O(n^3)$, incremental update in $O(n^2)$

Controllability

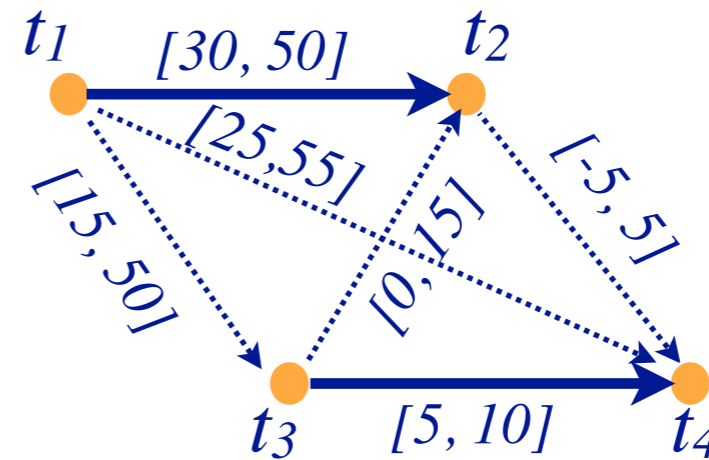
- ▶ Controllable vs contingent time points
 - t_1 and t_3 : controllable
 - t_2 and t_4 : contingent
random variables that are known to satisfy some constraints
- ▶ PC cannot be allowed to constrain a contingent time point
- ▶ Even if minimal network does not constrain any contingent time point the corresponding plan may not be feasible



Controllability



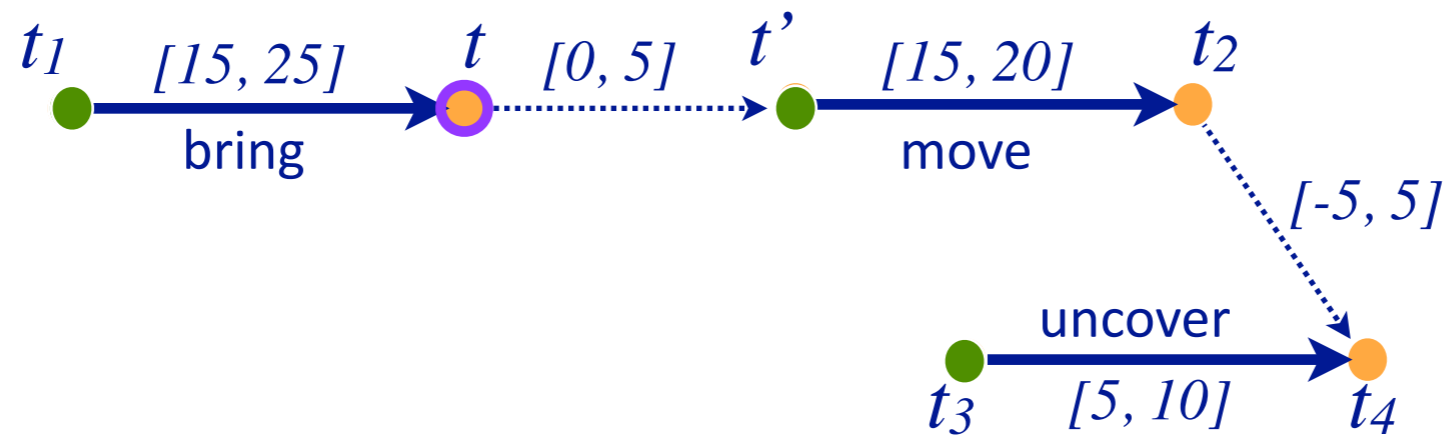
minimal network



$$\begin{aligned} d - d' - 5 &\leq t_3 \\ t_3 &\leq d - d' + 5 \end{aligned}$$

$$\Rightarrow 40 \leq t_3 \leq 25 !!$$

Dynamic Controllability



observe t

assign t' at any moment after t in $[0, 5]$

assign t_3 10 units after t'

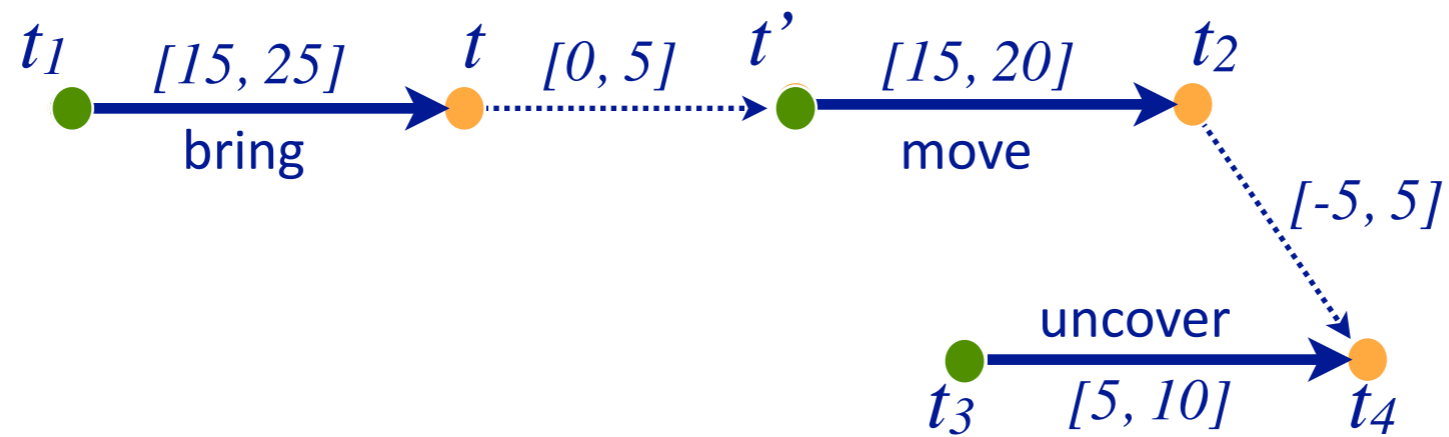
Dynamic Controllability

- ▶ Simple temporal network with uncertainty (STNU)
 - Controllable and contingent time points
 - Constraints, as in STN, controllable and contingent
- ▶ Controllable STNU: there exist values for controllable points that meets all constraints
 - *Strong controllability*: solution works for *all* possible values of contingent points in their predicted intervals
 - *Weak controllability*: solution as a function of the values of contingent points, if *known* in advance
 - *Dynamic controllability*: solution that is built dynamically, for each controllable point given the *observation* of past contingent points

Dynamic Controllability

- ▶ A *dynamic execution strategy* for an STNU:
 - online procedure for assigning, in some order, a value to each controllable point t , (i.e., triggering commands at right moment)
 - such that all controllable constraints are met, and
 - given that the values of all contingent variables *preceding* t are known and fit their assumed constraints
- ▶ An STNU is *dynamically controllable* if there exists a dynamic execution strategy for it
- ▶ Assigning a value to controllable t = triggering a command

Dynamic Controllability



observe t

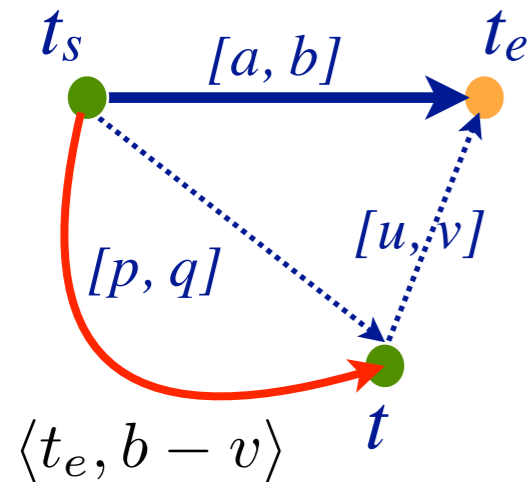
assign t' at any moment after t in $[0, 5]$

assign t_3 10 units after t'

Checking dynamic controllability of an STNU

- ▶ For a chronicle $\phi = (A, S_T, T, C)$ temporal constraints in C correspond to an STNU
- ▶ TemPlan: maintains incrementally STNU dynamically controllable
 - If Path Consistency reduces a contingent constraint
=> not dynamically controllable
 - Otherwise: test of dynamic controllability as an extension of Path Consistency with additional constraint propagation rules

Dynamic Controllability Checking



if $u < 0$ and $v \geq 0$ then
 t should wait until either $t_s + b - v$
 or t_e occurs

Constraint propagation rules

Conditions	Propagated constraint
$t_s \xrightarrow{[a,b]} t_e, t \xrightarrow{[u,v]} t_e, u \geq 0$	$t_s \xrightarrow{[b',a']} t$
$t_s \xrightarrow{[a,b]} t_e, t \xrightarrow{[u,v]} t_e, u < 0, v \geq 0$	$t_s \xrightarrow{\langle t_e, b' \rangle} t$
$t_s \xrightarrow{[a,b]} t_e, t_s \xrightarrow{\langle t_e, u \rangle} t$	$t_s \xrightarrow{[\min\{a, u\}, \infty]} t$
$t_s \xrightarrow{\langle t_e, b \rangle} t, t' \xrightarrow{[u,v]} t$	$t_s \xrightarrow{\langle t_e, b' \rangle} t'$
$t_s \xrightarrow{\langle t_e, b \rangle} t, t' \xrightarrow{[u,v]} t, t_e \neq t$	$t_s \xrightarrow{\langle t_e, b - u \rangle} t'$

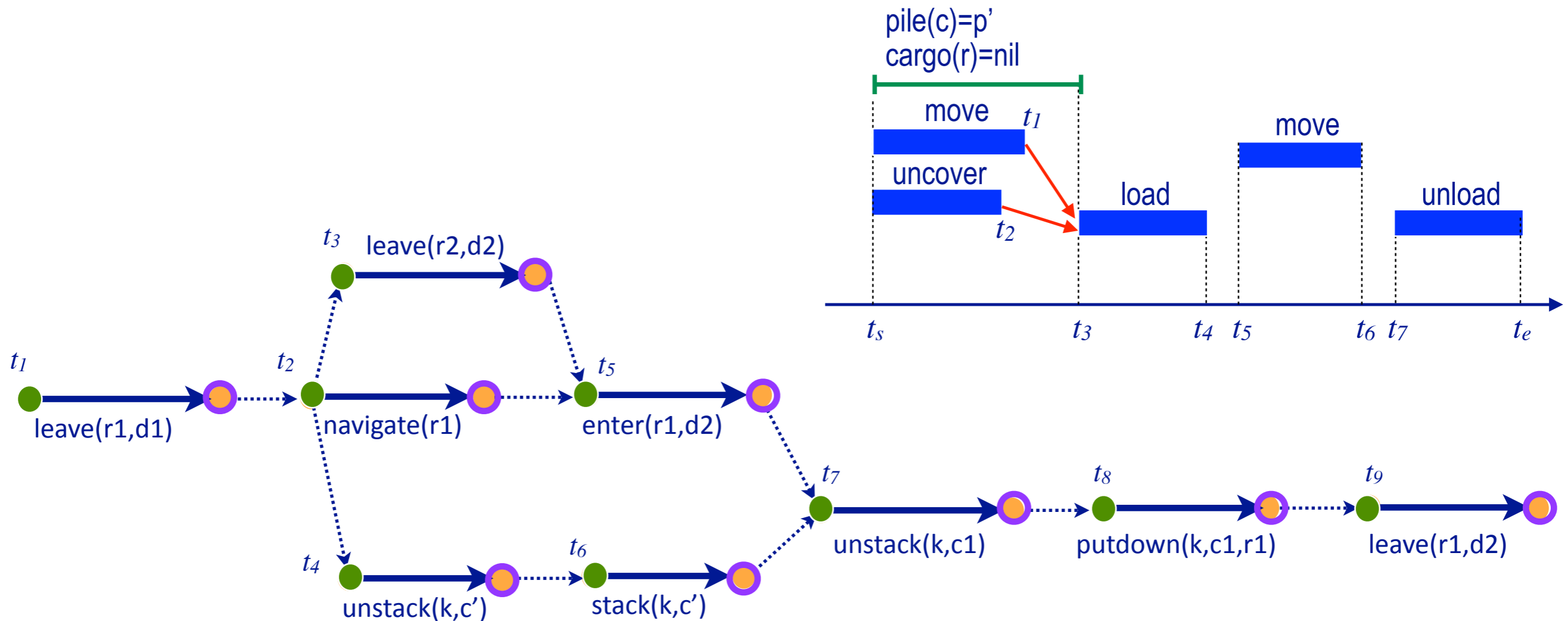
$$a' = a - u, b' = b - v$$

- ✓ Introduction
- ✓ Representation
- ✓ Temporal planning
- ✓ Consistency and controllability
- ▶ Acting with executable primitives
 - Dispatching
 - Observation actions
- ▶ Acting with atemporal refinement
- ▶ Conclusion

Dispatching Algorithm

Problem

- ▶ Given a dynamically controllable plan with executable primitives
- ▶ Trigger corresponding commands from online observations



Dispatching Algorithm

Plan grounded in realtime: when constrained w.r.t. absolute bounds or when execution starts

- Future point t is bounded with absolute bounds $[l_t, u_t]$
- Past point is instantiated

A controllable time point t that remains in the future

- t is **alive** if the current time $now \in [l_t, u_t]$
- t is **enabled** if
 - t is alive,
 - for every precedence constraint $t' < t$, t' has occurred, and
 - for every wait constraint $\langle t_e, \alpha \rangle$, either t_e has occurred or α has expired

Dispatching Algorithm

Dispatch(*plan*)

initialize the network

while there are controllable points that have not occurred do

 update *now*

 update contingent points that have been observed

enabled \leftarrow set of enabled points

 for every $t \in \textit{enabled}$ such that $\textit{now} = u_t$ do

 trigger t

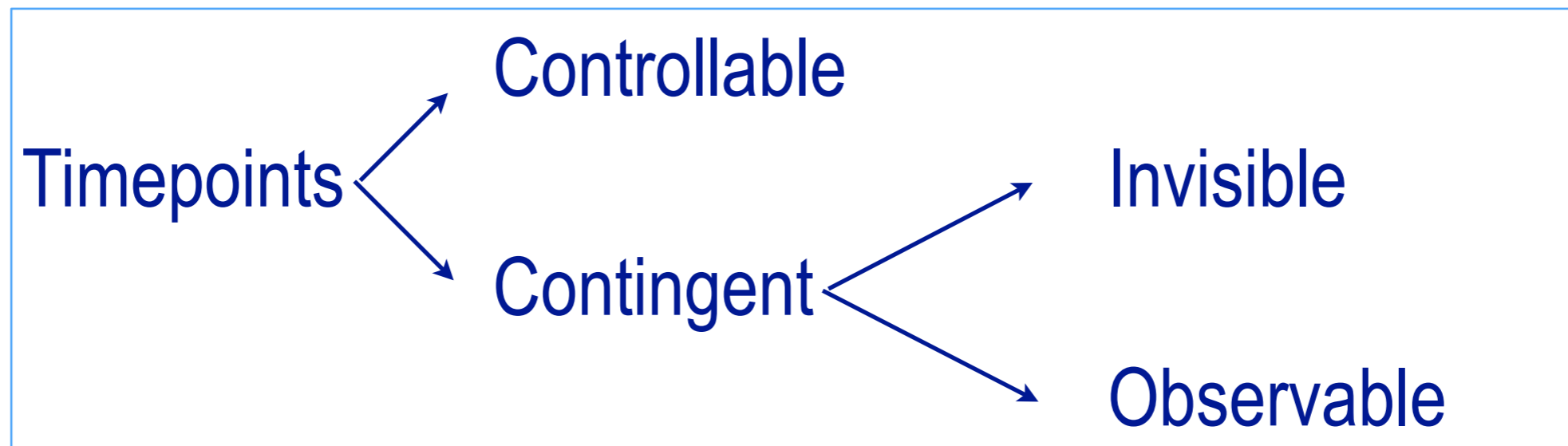
 arbitrarily choose other points in *enabled*; trigger them

 propagate in the network the values of triggered points

► Temporal monitoring

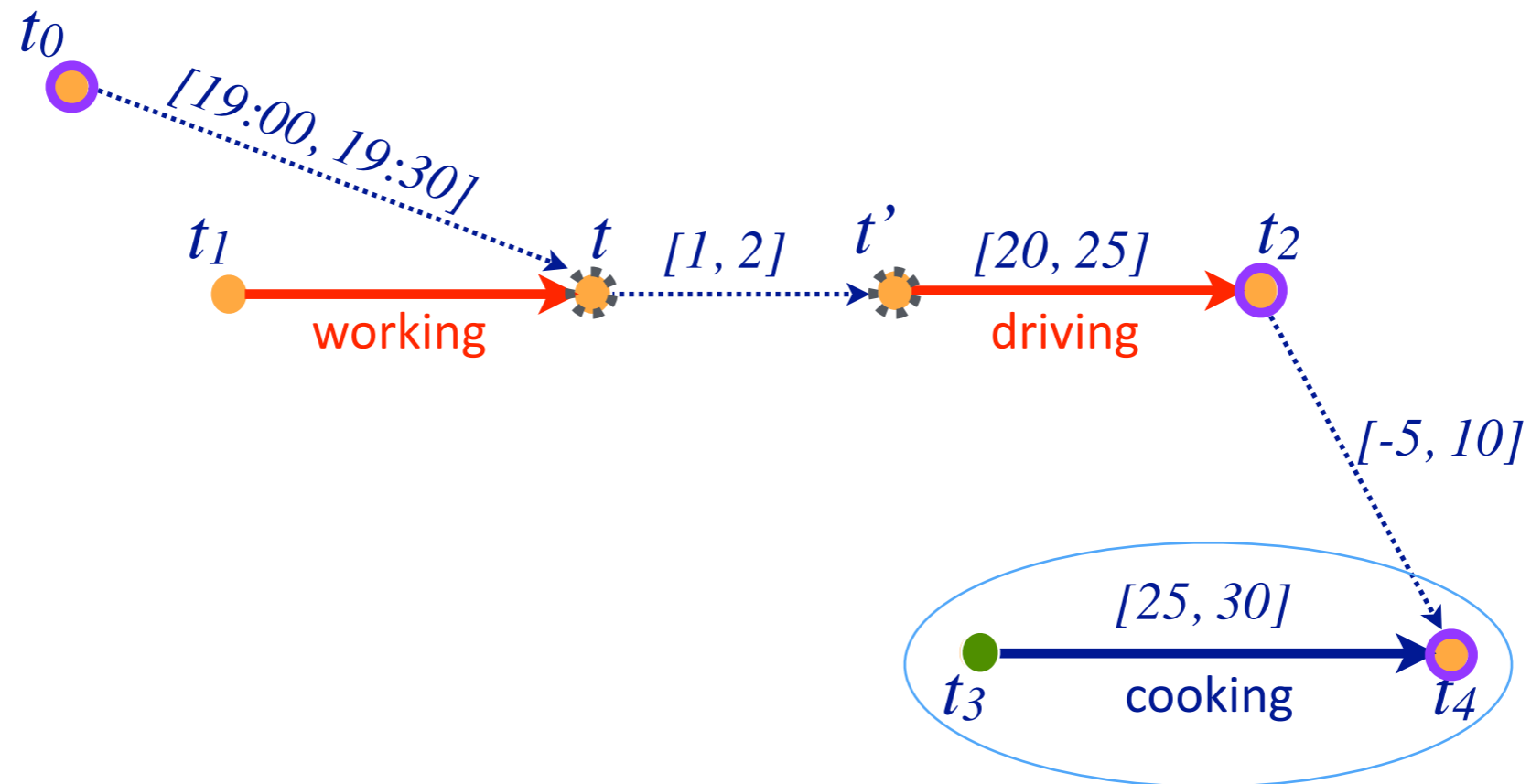
Observation Actions

- ▶ Assumption: all occurrences of contingent events are *observable*
 - Observation needed for dynamic controllability
 - In general not all events are observable
- ▶ Refining STNU into POSTNU



- ▶ Is POSTNU dynamically controllable ?

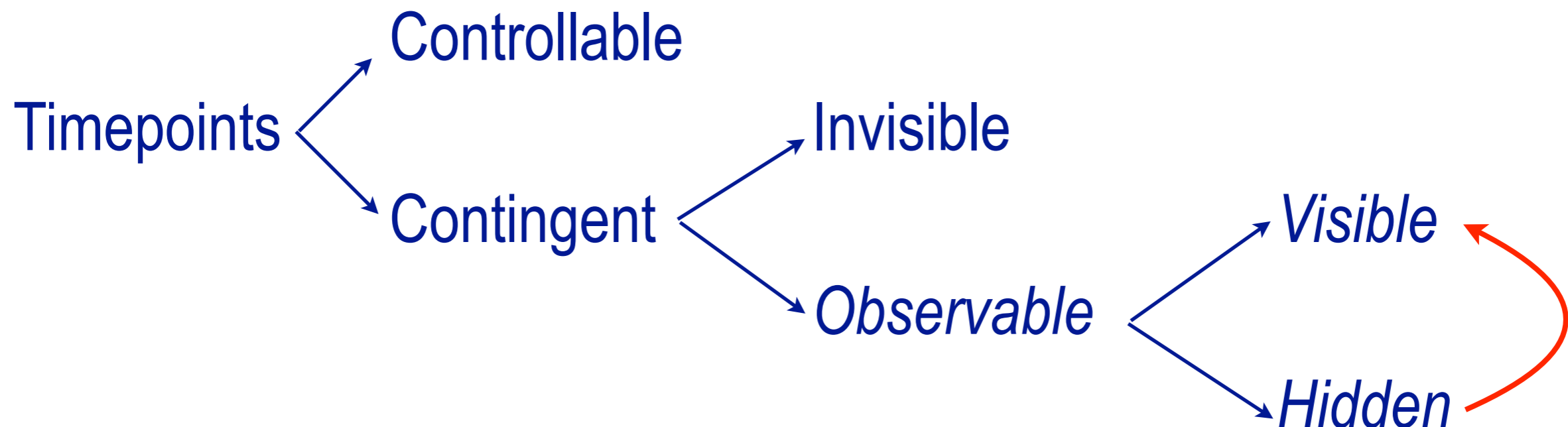
Observation Actions



- Controllable
- Contingent {
 - Invisible
 - observable

Observation Actions

- ▶ POSTNU dynamically controllable if there exists an execution strategy that chooses future controllable points to meet all the constraints, given the observation of past *visible* points
- ▶ Observable \neq visible
 - Observable means *it will be known when observed*
 - It can be temporarily hidden



- ✓ Introduction
- ✓ Representation
- ✓ Temporal planning
- ✓ Consistency and controllability
- ✓ Acting with executable primitives
- ▶ Acting with atemporal refinement
- ▶ Conclusion

Atemporal Refinement of Primitive Actions

- ▶ Planning primitives are compound tasks at the acting level
- ▶ Refined into commands with refinement methods in RAE

m-leave(r, d, w, e)

task: leave(r, d, w)

pre: $\text{loc}(r)=d, \text{adjacent}(d, w), \text{exit}(e, d, w)$

body: until empty(e) wait(1)

goto(r, e)

Operational
model

Descriptive
model

leave(r, d, w)

assertions: $[t_s, t_e] \text{loc}(r):(d, w)$

$[t_s, t_e] \text{occupant}(d):(r, \text{empty})$

constraints: $t_e \leq t_s + \delta_1$

adjacent(d, w)

Atemporal Refinement of Primitive Actions

- ▶ Planning primitives are compound tasks at the acting level
- ▶ Refined into commands with refinement methods in RAE

m-unstack(k, c, p)

task: unstack(k, c, p)

pre: $\text{pos}(c)=p, \text{top}(p)=c, \text{grip}(k)=\text{empty}$
 $\text{attached}(k, d), \text{attached}(p, d)$

body: locate-grasp-position(k, c, p)

move-to-grasp-position(k, c, p)

grasp(k, c, p)

until firm-grasp(k, c, p) ensure-grasp(k, c, p)

lift-vertically(k, c, p)

move-to-neutral-position(k, c, p)

Operational
model

Atemporal Refinement

▶ Pros

- Simple online refinement with RAE
- Avoids breaking down uncertainty of contingent duration
- Can be augmented with temporal monitoring functions in RAE
e.g., watchdogs, methods with duration preferences

▶ Cons

- Does not handle temporal requirements at the command level,
e.g., concurrency synchronization

- ▶ Rich chronicle representation with temporal refinement
- ▶ Planning with chronicle refinement
- ▶ Consistency and controllability
- ▶ Acting with chronicle dispatching and refinement
 - *ANML modeling language*
 - *FAPE Acting and Planning Environment*

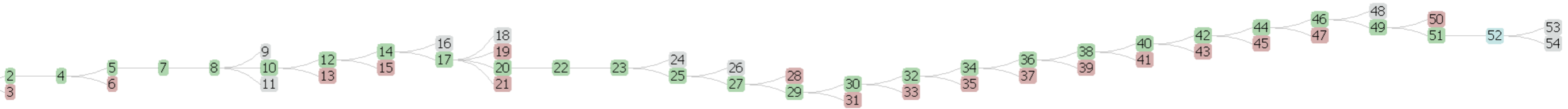
- ▶ Temporal models enrich descriptive and operational models of actions
- ▶ Chronicle-based approach very flexible for integrating generative and task decomposition techniques
- ▶ Acting refinement methods can be extended to integrate temporal construct of chronicles

Example: dwr

```
action uncover (containers c, piles p){
  [start] c.in == p ;
  :decomposition {[all] p.top == c ; };
  :decomposition {
    constant (...);
    [start] p.top == prevtop ;
    p != otherp; c != prevtop ;
    k.attached == d; p.ondock == d ;
    otherp.ondock ==d ;
    [all] p.available == true ;
    [all] otherp.available == true ;
    [all] contains {
      s1 : unstack(k,prevtop,p) ;
      s2: stack(k,prevtop,otherp);
      s3: uncover(c,p) ;};
    end(s1) <= start(s2);
    end(s2) <= start(s3); };
};
```

```
action goto (robots r, docks to){
  constant docks from;
  constant waypoints wa, wb ;
  [start] r.loc == from ;
  :decomposition {from == to ; };
  :decomposition {from != to ;
    adjacent(from, wa) ;
    adjacent(to, wb) ;
    [all] contains {
      s1 : leave(r, from, wa) ;
      s2: navigate(r, wa, wb) ;
      s3 : enter(r, to, wb) ;};
    end(s1) <= start(s2);
    end(s2) <= start(s3);};
};
```


Example: search tree



Example: plan found

