



Robust Execution of Temporal Plans

Slide Contributions:
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16.410 / 16.413
October 6th, 2010



Model-based Embedded & Robotic Systems



Assignments

- **Remember:**
 - Problem Set #5 due today, Wed, Oct. 6th, 2010.
 - Problem Set #6 out today.
- **Reading:**
 - Today: Dechter, R., I. Meiri, J. Pearl, “Temporal Constraint Networks,” *Artificial Intelligence*, 49, pp. 61-95, 1991.
 - Wednesday: Logic [AIMA] Ch. 7, 8
- **Exam:**
 - Mid-Term - October 20th.

Executing Time Critical Missions

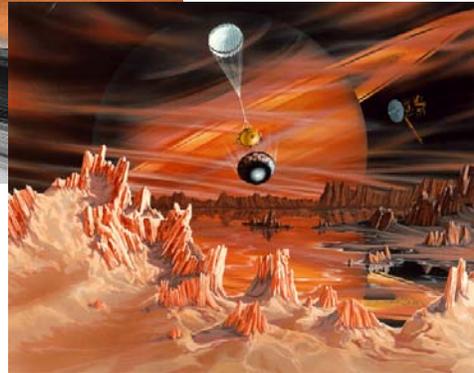
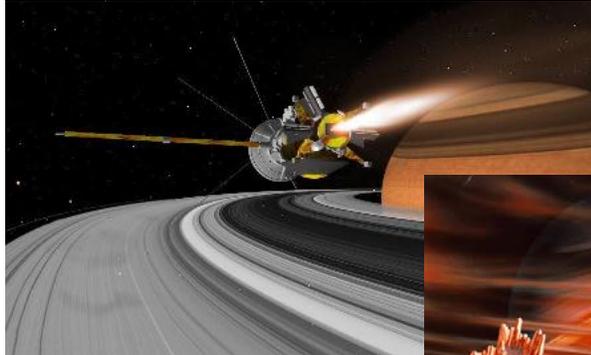


Image credit: NASA.



Team Coordination under Time Pressure



Images of scrub nurses and surgeons removed due to copyright restrictions.

An effective Scrub Nurse:

- works hand-to-hand, face-to-face with surgeon,
- assesses and anticipates needs of surgeon,
- provides assistance and tools in order of need,
- responds quickly to changing circumstances,
- responds quickly to surgeon's cues and requests.



Human-Robot Teaming



Images of human-robot teaming (in surgical, space, and rescue settings) removed due to copyright restrictions.



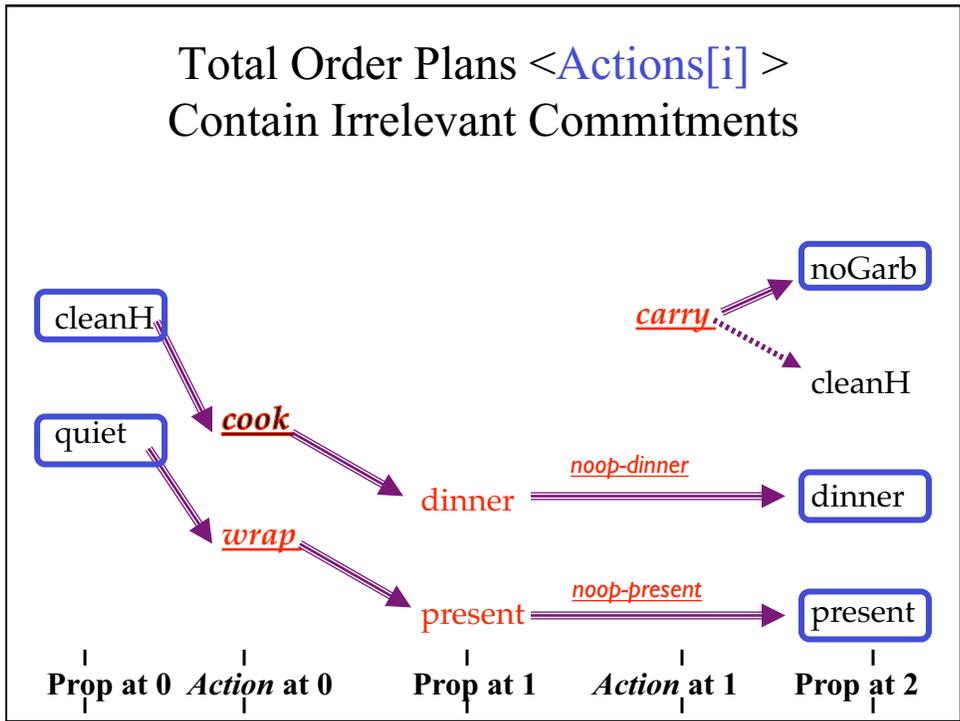
Robust Execution of Time-critical Tasks



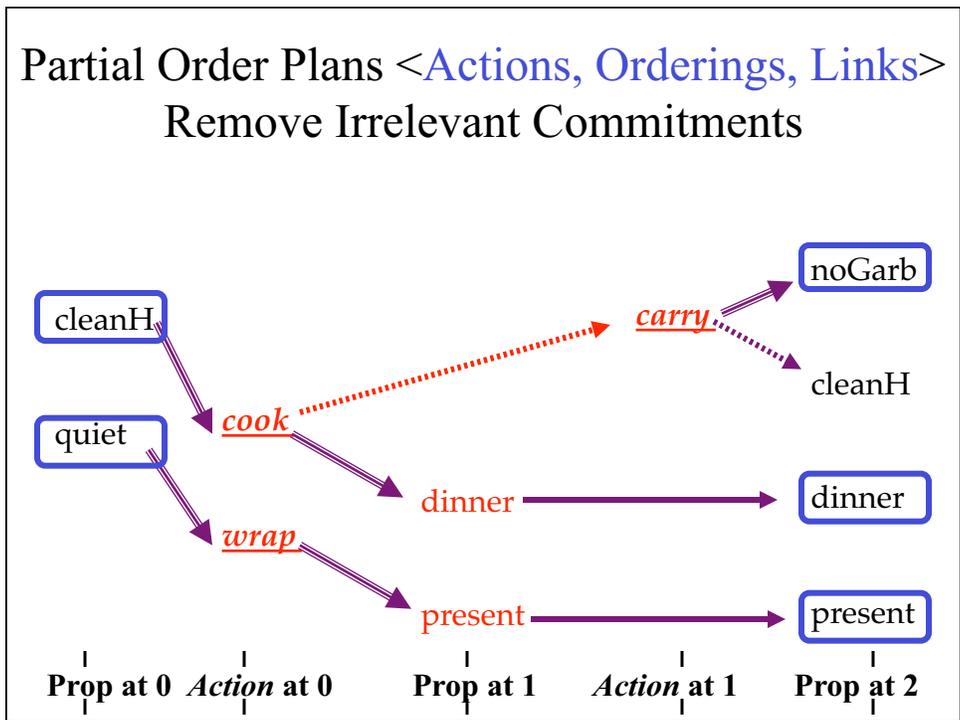
- Executing Simple Plans
- Robust Execution
 - Describing Temporal Plans
 - Checking Temporal Plan Consistency
 - Scheduling Plans
 - Robust, Dynamic Scheduling



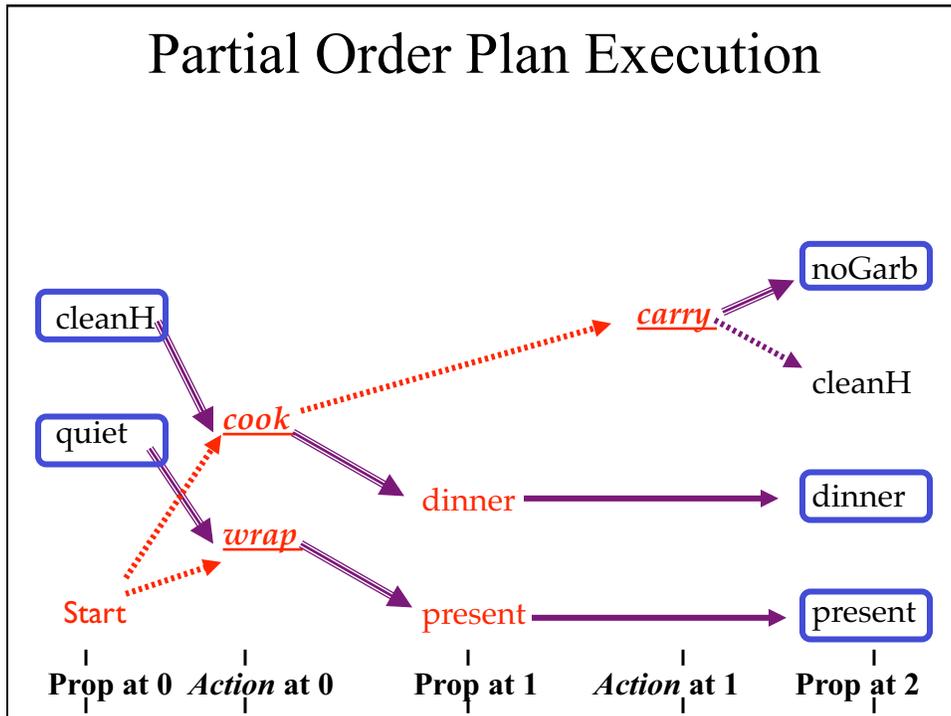
Total Order Plans $\langle \text{Actions}[i] \rangle$
 Contain Irrelevant Commitments



Partial Order Plans $\langle \text{Actions, Orderings, Links} \rangle$
 Remove Irrelevant Commitments



Partial Order Plan Execution



Partial Order Plan Execution

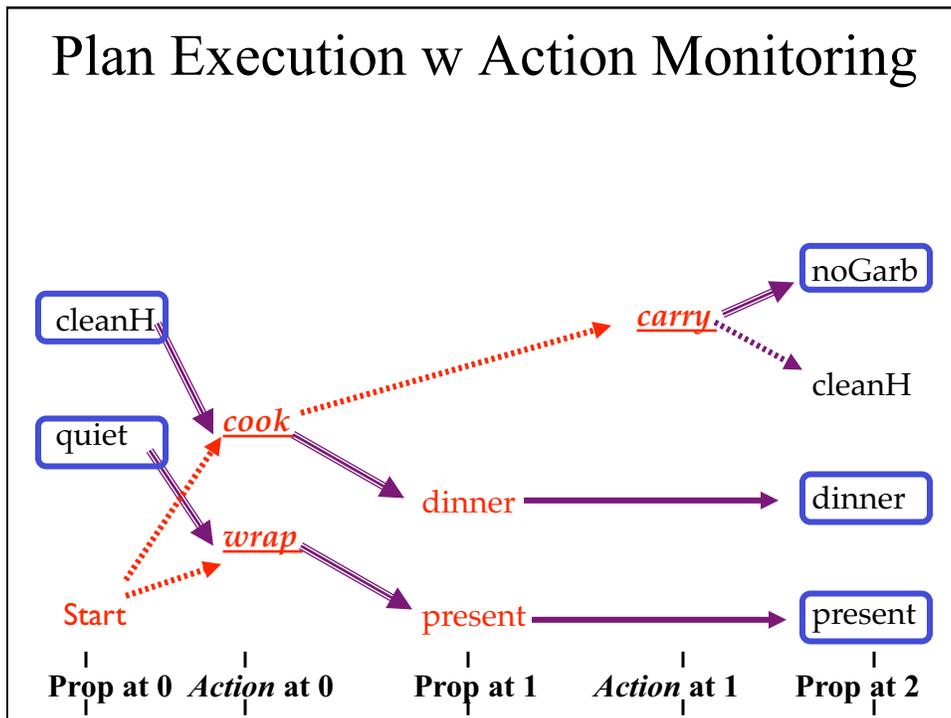
Initialize queue **Ready**, with action **Start**.

Mark all **actions** as “not executed.”

Loop

- If **Ready** is **empty**, Then **terminate**.
- Dequeue action **a** from **Ready** and **execute**.
- When completed, mark **a** as **executed**.
- For each succeeding action **b** such that $a < b$ or $\text{linked}(a,b,p)$,
 - If every preceding action **c** is marked “executed,” such that $c < b$ or $\text{linked}(c,b,p)$,
 - Then **queue b** on **Ready**.

Plan Execution w Action Monitoring



Plan Execution w Action Monitoring

Initialize queue `Ready`, with action `Start`.

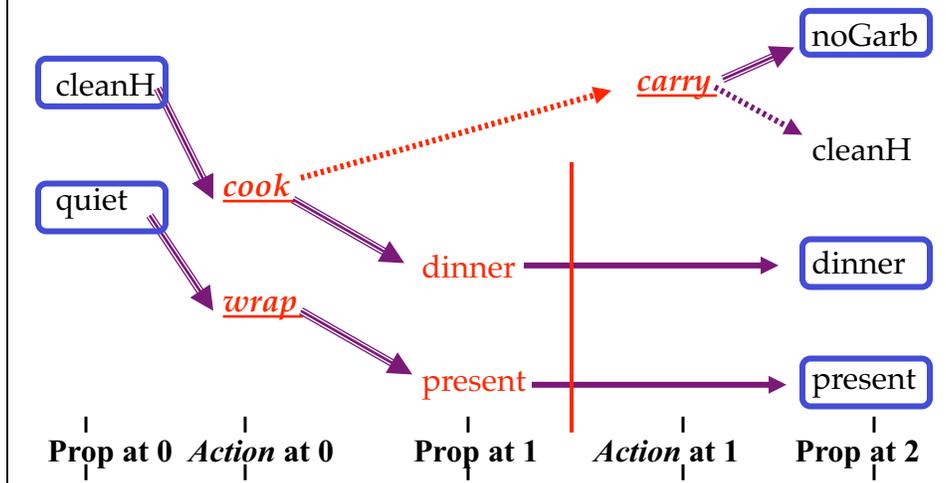
Mark all actions as “not executed.”

Loop

- If `Ready` is empty, Then terminate.
- Dequeue action `a` from `Ready`.
- If `a`'s preconditions satisfied, then execute, else fail.
- When completed, mark `a` as executed.
- For each succeeding action `b` such that $a < b$ or $\text{linked}(a,b,p)$,
 - If every preceding action `c` is marked “executed,” such that $c < b$ or $\text{linked}(c,b,p)$,
 - Then queue `b` on `Ready`.

Execution Monitoring

- Check if any preconditions of unexecuted actions are violated.
- ⇒ Check if a causal link that crosses the current time is violated.



Plan Execution w Execution Monitoring

Initialize agenda **Ready** with action **Start**

Initialize agenda **ActiveLinks** to empty

Mark all actions as “not executed.”

Loop

- If **Ready** is empty then terminate.
- For each link on **ActiveLinks**
 - If the proposition for link doesn't hold,
Then return failure
- Dequeue action **a** from **Ready**
- If preconditions of action are satisfied
 - Then execute
 - Else return failure
- ... (continued on next slide)

Plan Execution w Execution Monitoring (cont)

Loop

- ... (continued from previous slide)
- Mark **a** as “executed.”
- For each action **c** such that $\text{linked}(c,a,p)$.
 - dequeue $\langle c,a,p \rangle$ from **ActiveLinks**.
- For each action **d** such that $\text{linked}(a,d,p)$.
 - queue $\langle a,d,p \rangle$ on **ActiveLinks**.
- For each action **b** such that $a < b$ or $\text{linked}(a,b,p)$.
 - If every action **c** has been executed, such that $c < b$ or $\text{linked}(c,b,p)$
 - Then queue **b** on **Ready**.

Robust Execution of Time-critical Tasks



Model-based Embedded & Robotic Systems

- Executing Simple Plans
- **Robust Execution**



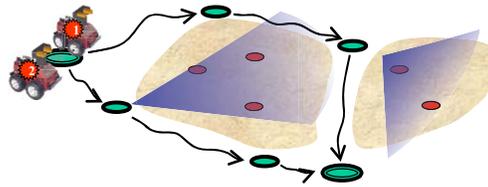
Executing Timed Programs and Plans Robustly



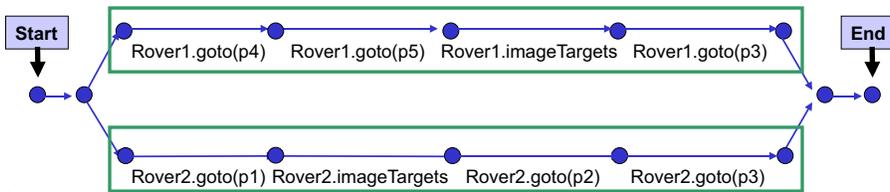
Model-based Embedded & Robotic Systems

```

imageScienceTargets(Rover1, Rover2)
(Parallel
  (Sequence
    [5,10] Rover1.goto(p4);
    [5,10] Rover1.goto(p5);
    [2,5] Rover1.imageTargets();
    [5,10] Rover1.goto(p3);
  )
  (Sequence
    [5,10] Rover2.goto(p1);
    [5,10] Rover2.imageTargets();
    [2,5] Rover2.goto(p2);
    [5,10] Rover2.goto(p3);
  )
)
    
```



in RMPL [williams et al]



Agents adapt to temporal disturbances in a coordinated manner by scheduling the start of activities on the fly.

In general, categorize durations into controllable and uncontrollable (STNUs).



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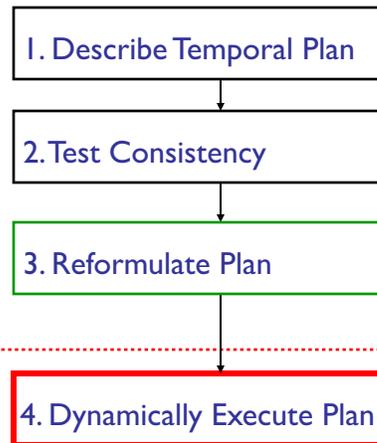
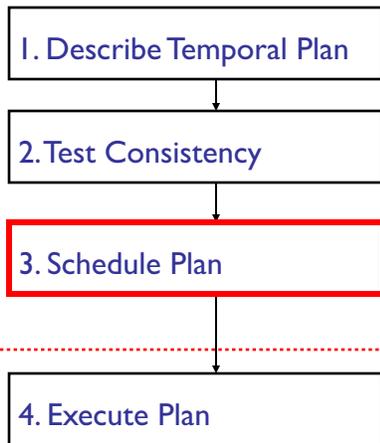
To Execute a Temporal Plan



Model-based Embedded & Robotic Systems

Part I: Scheduling Off-line

Part II: Scheduling Online



offline
online



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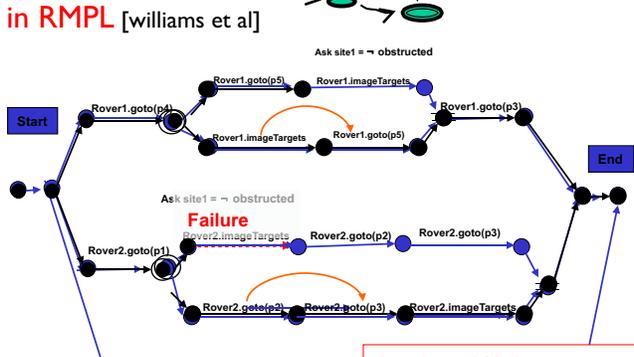
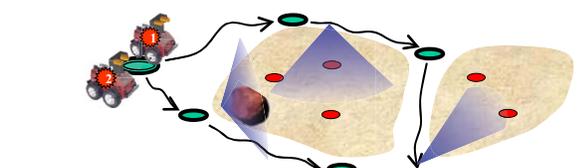


Expanding Robustness by Dynamically Choosing Methods

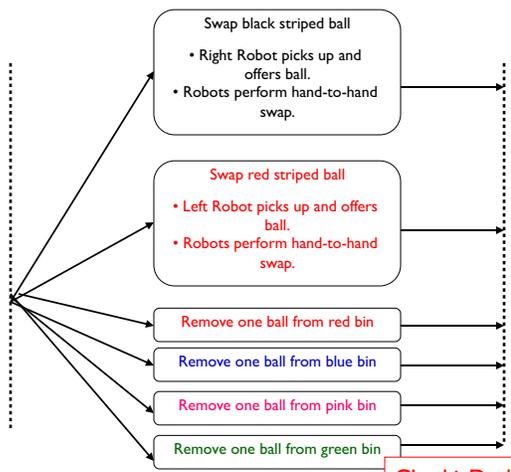
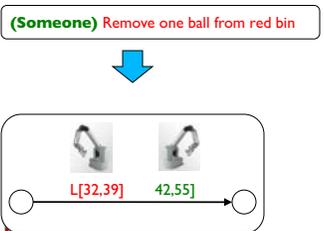


```

{
  [5,10] Rover1.goto(p4);
  choose {
    {
      [5,10] Rover1.goto(p5);
      maintaining( site1 = ~ obstructed);
      [2,5] Rover1.imageTargets();
    }
    {
      [2,5] Rover1.imageTargets();
      [5,10] Rover1.goto(p5);
    }
  };
  [5,10] Rover1.goto(p3);
},
[5,10] Rover2.goto(p1);
choose {
  {
    [2,5] Rover2.imageTargets();
    maintaining( site1 = ~ obstructed);
    [5,10] Rover2.goto(p2);
    [5,10] Rover2.goto(p3);
  }
  {
    [5,10] Rover2.goto(p2);
    [5,10] Rover2.goto(p3);
    [2,5] Rover2.imageTargets();
  }
}
}
    
```



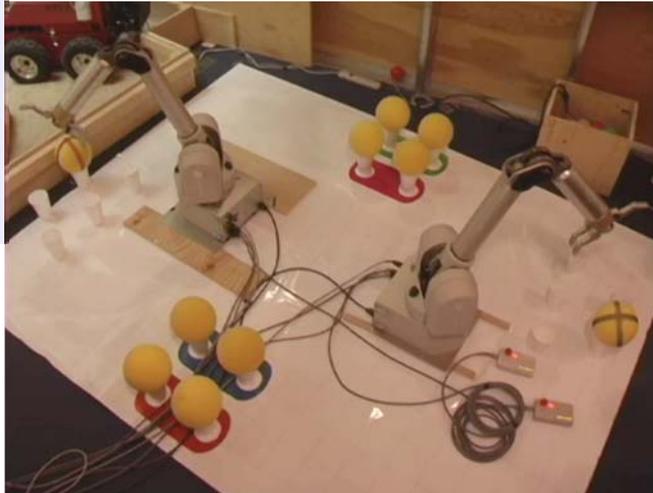
Expanding Robustness by Dynamically Assigning Tasks



in RMPL [williams et al]

Chaski, Drake, Kirk [Kim; Shah; Conrad]

Expanding Robustness by Dynamically Assigning Tasks



- Off-nominal
- Partner adapts in response to teammate's failure.

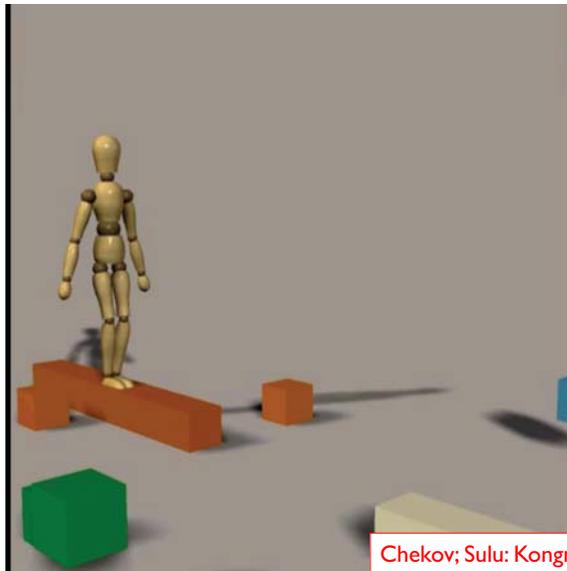


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Expanding Robustness by Coordinating Underactuated Systems



Chekov; Sulu; Kongming
[Hofmann; Leaute; Blackmore; Ono; Li]



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Expanding Robustness by Coordinating Underactuated Systems

MRAS
Model-based Robustly Embedded & Adaptive Systems



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Chekov; Sulu; Kongming
[Hofmann; Leaute; Blackmore; Ono; Li]

Robust Execution of Time-critical Tasks

MRAS
Model-based Robustly Embedded & Adaptive Systems

- Executing Simple Plans
- Robust Execution
 - Describing Temporal Plans
 - Checking Temporal Plan Consistency
 - Scheduling Plans
 - Robust, Dynamic Scheduling

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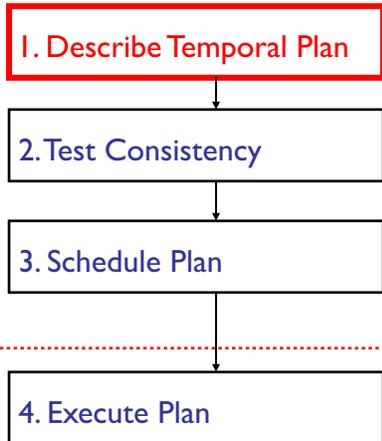
24



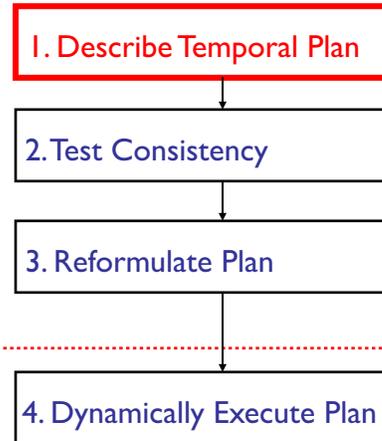
To Execute a Temporal Plan



Part I: Schedule Off-line



Part II: Schedule Online



offline
online

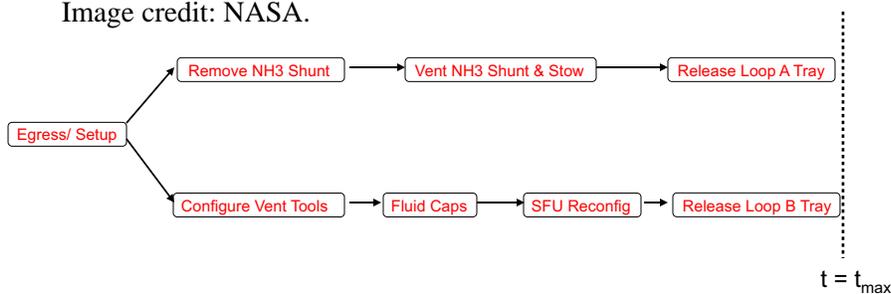


Describing Temporal Plans



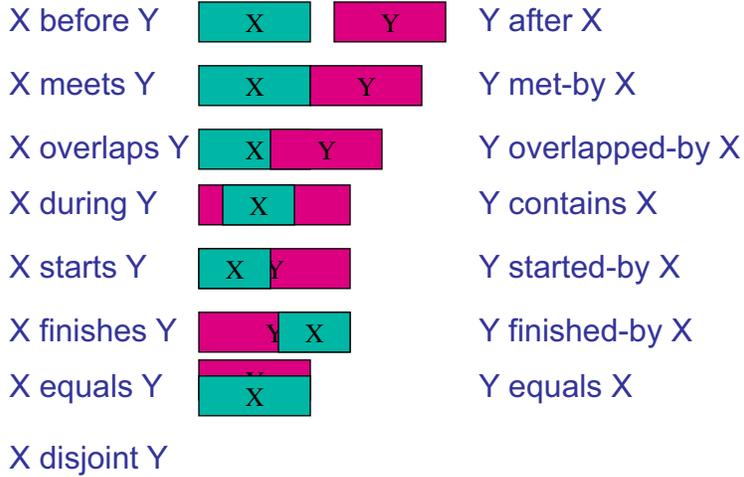
Image credit: NASA.

- Activities to perform
- Relationships among activities



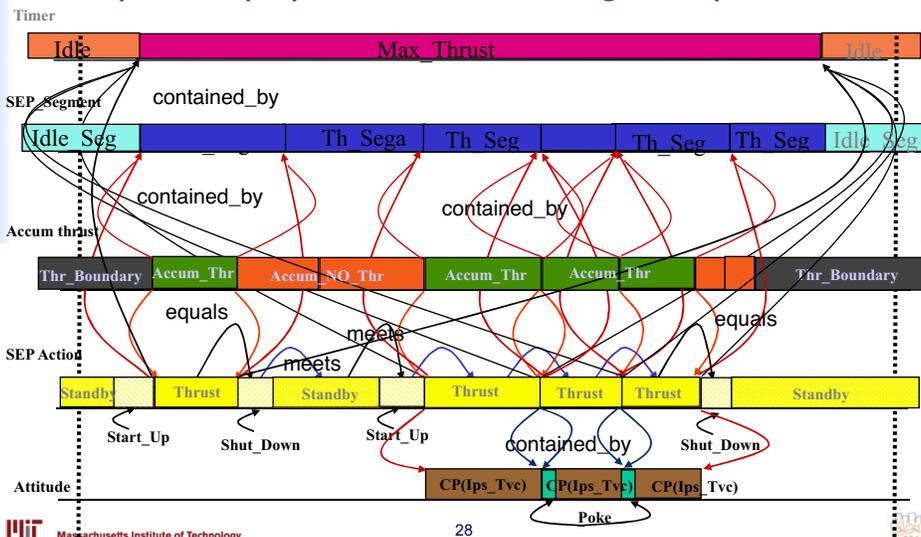
Describing Temporal Plans

Qualitative Temporal Relationships (Allen 83)



Describing Temporal Plans

Example: Deep Space One Remote Agent Experiment



Describing Temporal Plans

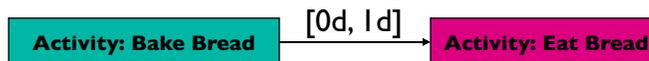
Adding Metric Information

- Going to the store takes at least 10 min and at most 30 min.

[10min, 30min]

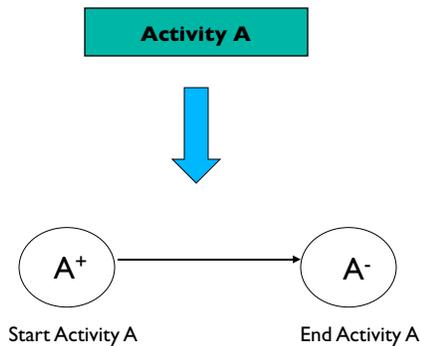
Activity: Going to the store

- Bread should be eaten within one day of baking.



Describing Temporal Plans

Simplify by reducing interval relations to relations on timepoints.



Describing Temporal Plans



Model-based Scheduling & Robotic Systems

Qualitative Temporal Relationships as timepoint inequalities

X before Y		$X^+ < Y^-$	
X meets Y		$X^+ = Y^-$	
X overlaps Y		$Y^- < X^+ \text{ and } X^- < Y^+$	
X during Y		$Y^- < X^- \text{ and } X^+ < Y^+$	
X starts Y		$X^- = Y^- \text{ and } X^+ < Y^+$	
X finishes Y		$X^- < Y^- \text{ and } X^+ = Y^+$	
X equals Y		$X^- = Y^- \text{ and } X^+ = Y^+$	
X disjoint Y		$X^+ < Y^- \text{ or } Y^+ < X^-$	



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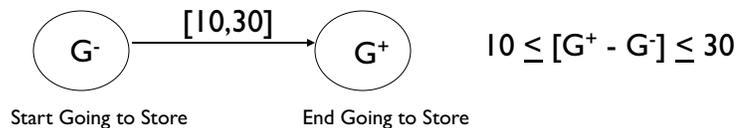
Describing Temporal Plans



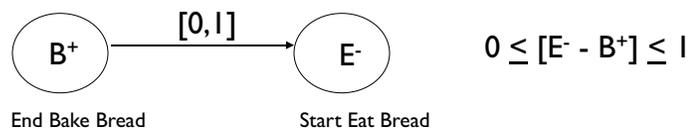
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Encode metric Information by generalizing inequalities to interval constraints.

- Going to the store takes at least 10 min and at most 30 min.



- Bread should be eaten within one day of baking.



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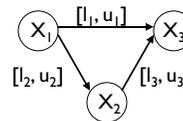
Temporal Relations Described as an STP



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• Simple Temporal Problem (STP)

- variables X_1, \dots, X_n , representing time points with real-valued domains,
- binary constraints of the form:



$$(X_k - X_i) \in [a_{ik}, b_{ik}].$$

Sufficient to represent:

- all Allen relations but 1...
- simple metric constraints

Can't represent:

- Disjoint activities



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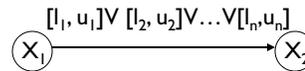
Temporal Relations Described as a TCSP



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• Temporal Constraint Satisfaction Problem (TCSP)

- Extends STP by allowing multiple intervals for each binary constraints:



$$(X_k - X_i) \in P\left(\{[a_{ik}, b_{ik}] \mid a_{ik} \leq b_{ik}\}\right).$$

Supports:

- Multiple time windows for accomplishing an activity.
- Different methods of accomplishing an activity.



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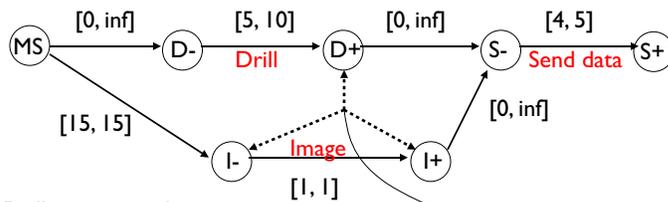
Temporal Relations Described as a DTP



- Disjunctive Temporal Problem (DTP)

- Extends TCSP by allowing non-binary constraints.

Activities of Mars Rover: Drill (D) , Image (I), Send Data (S)



Drilling causes vibration.

Image cannot occur

- during the last two minutes before drilling, or
- during the first minute after drilling ends.

$$2 \leq D^+ - I^+ \leq \text{inf}$$

OR

$$1 \leq I^- - D^+ \leq \text{inf}$$

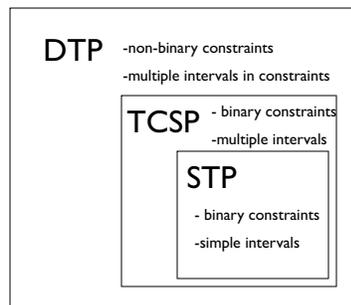


A Temporal Plan Described as a DTP



- Disjunctive Temporal Problem (DTP)

- extends a TCSP by allowing non-binary constraints.

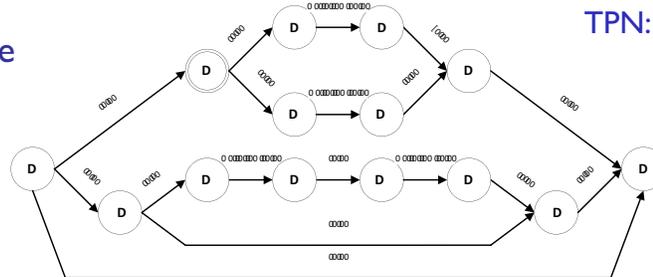
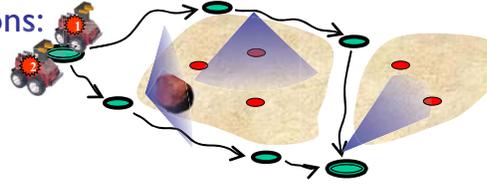


Temporal Plan Networks and Conditional STPs



RMPL - Nested Compositions:

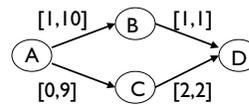
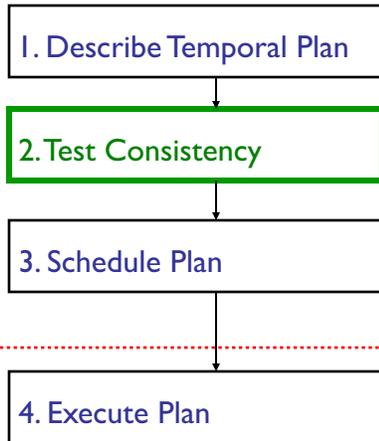
- Activity
- Sequence
- Parallel
- Choice
- With Time



To Execute a Temporal Plan



Part I : Schedule Off-line



offline
online

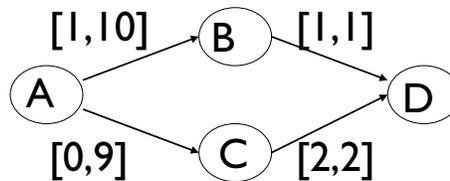


Consistency of an STP



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Input: An STP $\langle X, C \rangle$ where $C_j = \langle X_k, X_i, l_j, u_j \rangle$



Output: True iff there exists an X satisfying C



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Map STP to Equivalent Distance Graph



Model-based Embedded & Robotic Systems

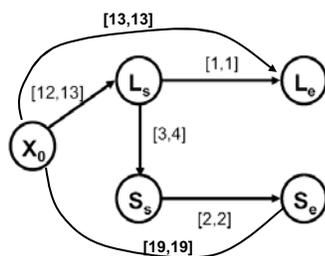
Idea: Map STN to distance (weighted) graph and check for negative cycles.

- Map upper bound to outgoing, non-negative arc.
- Map lower bound to incoming, negative arc.

$$l \leq X_j - X_i \leq u$$

$$X_j - X_i \leq u$$

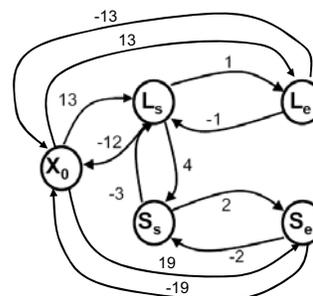
$$X_i - X_j \leq -l$$



Constraint Graph



For efficient inference



Distance Graph



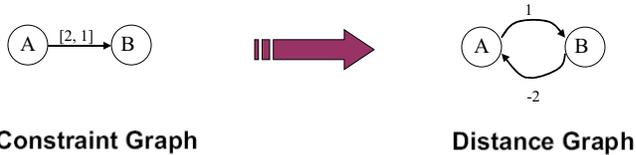
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STP Consistency

- Example of inconsistent constraint:

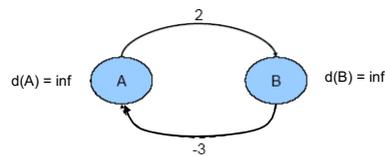


- An STP is consistent iff its distance graph has no negative cycles.
- Detect by computing shortest path from one node to all other nodes.
 - Single Source Shortest Path (SSSP)

STP Consistency: Generic Labeling Algorithm

Detect negative cycles by computing the shortest-path from a single node to all other nodes (Single Source Shortest Path).

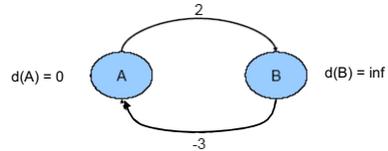
1. For all nodes s in graph G
2. $d(s) = \text{inf}$
3. $d(s_{\text{start}}) = 0$
4. while some $\text{arc}(i,j)$ is violating,
5. $d(j) = d(i) + c(i,j)$



STP Consistency: Generic Labeling Algorithm



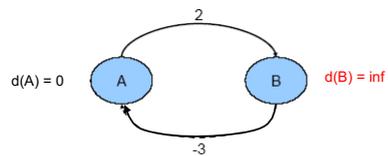
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STP Consistency: Generic Labeling Algorithm



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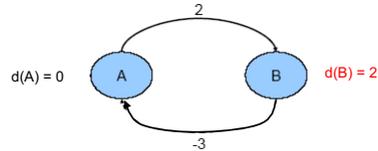
$\text{arc}(i,j)$ is violating if,
 $d(j) > d(i) + c(i,j)$



STP Consistency: Generic Labeling Algorithm



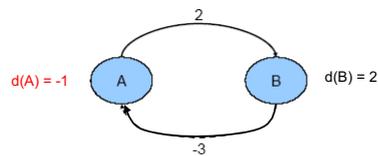
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STP Consistency: Generic Labeling Algorithm



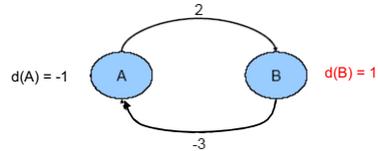
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STP Consistency: Generic Labeling Algorithm



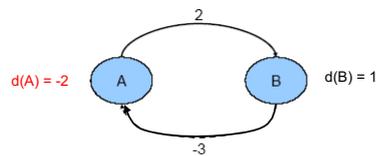
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STP Consistency: Generic Labeling Algorithm



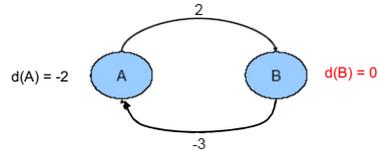
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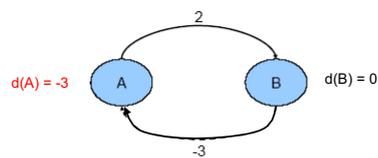
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STP Consistency: Generic Labeling Algorithm



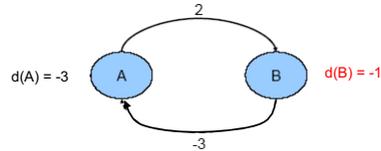
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5. $d(j) = d(i) + c(i,j)$



STP Consistency: Generic Labeling Algorithm



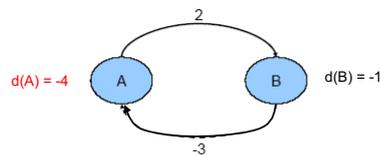
1. For all nodes s in graph G
2. $d(s) = \text{inf}$
3. $d(s_{\text{start}}) = 0$
4. while some $\text{arc}(i,j)$ is violating,
5. $d(j) = d(i) + c(i,j)$



STP Consistency: Generic Labeling Algorithm



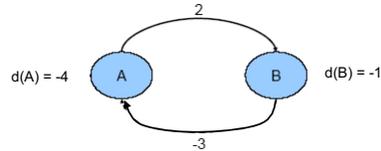
1. For all nodes s in graph G
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STP Consistency: Generic Labeling Algorithm



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5. $d(j) = d(i) + c(i,j)$



How do we detect inconsistency?

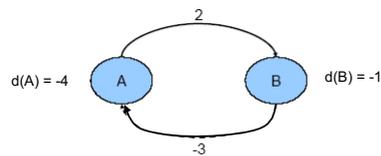
1. One way: Check for any d -value to drop below $-nC$



STP Consistency: FIFO Labeling Algorithm



1. For all nodes s in graph G
2. $d(s) = \text{inf}$
3. $d(s_{\text{start}}) = 0$
4. while some $\text{arc}(i,j)$ is violating,
5. $d(j) = d(i) + c(i,j)$

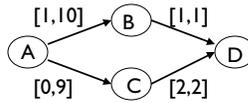
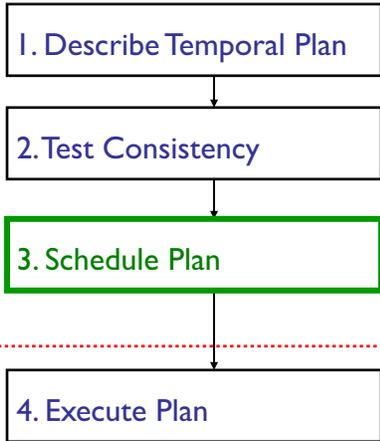


- Maintain queue of updated nodes.
- For each node on queue, check for outgoing arcs that may be potentially violating.



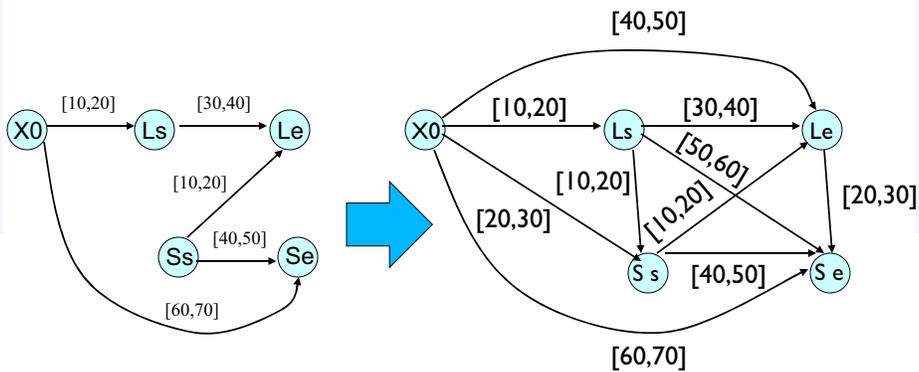
To Execute a Temporal Plan

Part I : Schedule Off-line



offline
online

Scheduling

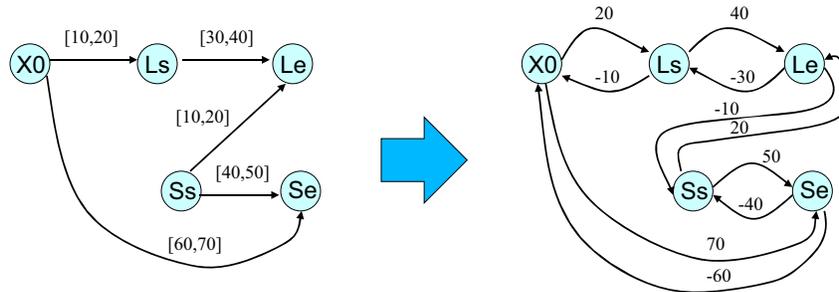


- Idea: Expose Implicit Constraints in STP

Scheduling with All Pairs Shortest Path Graph



Model-based Scheduling & Robotic Systems



- Idea: Expose Implicit Constraints in STP
- Compute All-Pairs-Shortest-Path (APSP) of d-graph (Floyd-Warshall).



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Distance Graph G_d implies Constraints



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- Path constraint: $i_0 = i, i_1 = \dots, i_k = j$

$$X_j - X_i \leq \sum_{j=1}^k u_{i_{j-1}, i_j}$$

→ Conjoined path constraints result in the shortest path as bound:

$$X_j - X_i \leq d_{ij}$$

where d_{ij} is the shortest path from i to j



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All Pairs Shortest Path

Floyd-Warshall (alternatively Johnson)

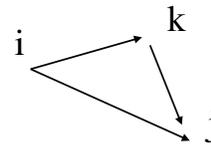


1. for $i := 1$ to n do $d_{ij} \leftarrow 0$;
2. for $i, j := 1$ to n do $d_{ij} \leftarrow w(i,j)$;

3. for $k := 1$ to n do
4. for $i, j := 1$ to n do
5. $d_{ij} \leftarrow \min\{d_{ij}, d_{ik} + d_{kj}\}$;

Initialize distances

Take minimum distance over all triangles



Complexity $O(n^3)$



APSP

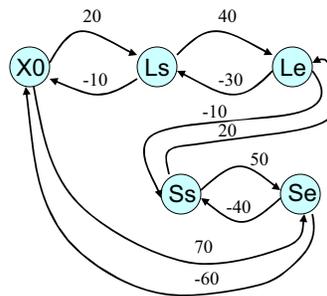


$$i = X_0, k = L_s, j = L_e$$

	X_0	L_s	L_e	S_s	S_e
X_0	0	20	inf 60	inf	70
L_s	-10	0	40	inf	inf
L_e	inf	-30	0	-10	inf
S_s	inf	inf	20	0	50
S_e	-60	inf	inf	40	0

Initial d-graph

3. for $k := 1$ to n do
4. for $i, j := 1$ to n do
5. $d_{ij} \leftarrow \min\{d_{ij}, d_{ik} + d_{kj}\}$;



$$S_{\text{latest}} = (d_{01}, \dots, d_{0n})$$



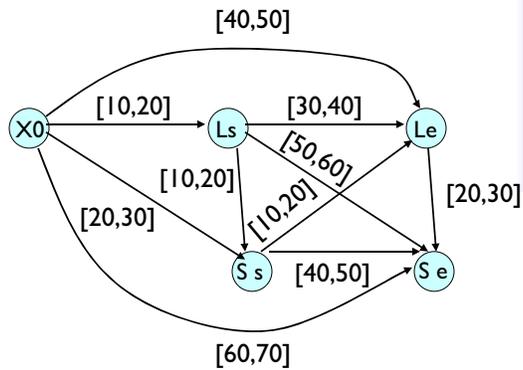
Scheduling with All Pairs Shortest Path Graph



Model-based Scheduling & Robotic Systems

	X_0	Ls	Le	Ss	Se
X_0	0	20	50	30	70
Ls	-10	0	40	20	60
Le	-40	-30	0	-10	30
Ss	-20	-10	20	0	50
Se	-60	-50	-20	40	0

APSP d-graph



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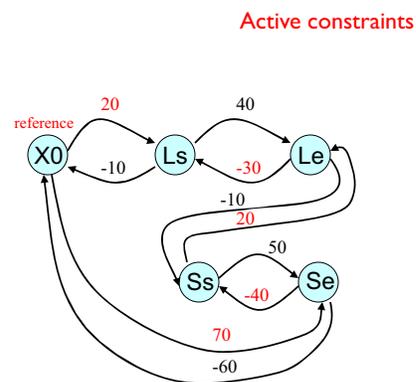
Scheduling: Latest Solution



Model-based Scheduling & Robotic Systems

	X_0	Ls	Le	Ss	Se
X_0	0	20	50	30	70
Ls	-10	0	40	20	60
Le	-40	-30	0	-10	30
Ss	-20	-10	20	0	50
Se	-60	-50	-20	40	0

APSP d-graph



$$S_{\text{latest}} = (d_{01}, \dots, d_{0n})$$



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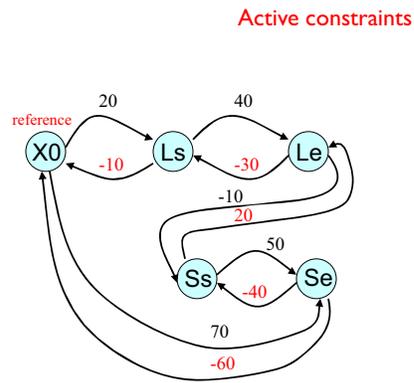
62



Scheduling: Earliest Solution

	X_0	Ls	Le	Ss	Se
X_0	0	20	50	30	70
Ls	-10	0	40	20	60
Le	-40	-30	0	-10	30
Ss	-20	-10	20	0	50
Se	-60	-50	-20	40	0

APSP d-graph



$$S_{\text{earliest}} = (-d_{10}, \dots, d_{n0})$$

Scheduling: Window of Feasible Values

	X_0	Ls	Le	Ss	Se
X_0	0	20	50	30	70
Ls	-10	0	40	20	60
Le	-40	-30	0	-10	30
Ss	-20	-10	20	0	50
Se	-60	-50	-20	40	0

Latest Times

Earliest Times

APSP d-graph

- Ls in [10, 20]
- Le in [40, 50]
- Ss in [20, 30]
- Se in [60, 70]

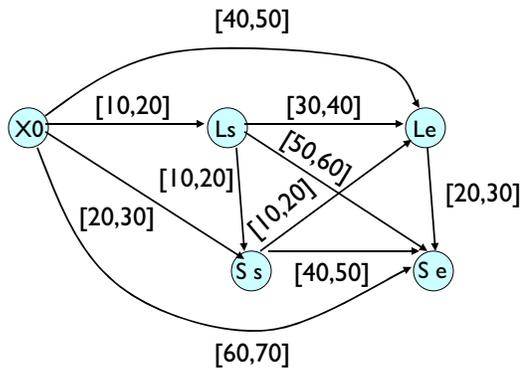
Scheduling without Search: Solution by Decomposition



- Can assign variables in any order, without backtracking.

Key ideas

- Incrementally tighten feasible intervals, as commitments are made.
- Perform on demand.



Scheduling without Search: Solution by Decomposition

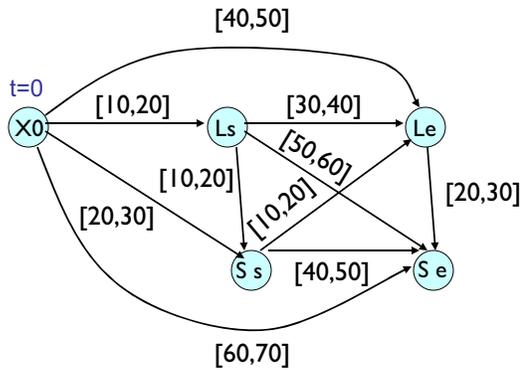


- Can assign variables in any order, without backtracking.

Key ideas

- Incrementally tighten feasible intervals, as commitments are made.
- Perform on demand.

- Select value for X0



Scheduling without Search: Solution by Decomposition

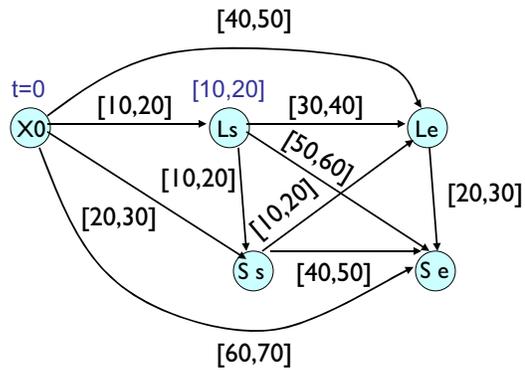


- Can assign variables in any order, without backtracking.

Key ideas

- Incrementally tighten feasible intervals, as commitments are made.
- Perform on demand.

- Select value for X_0
- Select value for L_s , consistent with X_0



Scheduling without Search: Solution by Decomposition

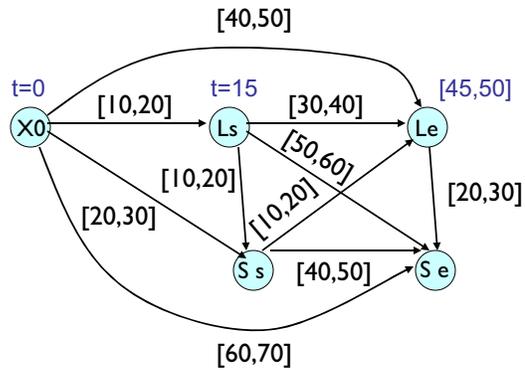


- Can assign variables in any order, without backtracking.

Key ideas

- Incrementally tighten feasible intervals, as commitments are made.
- Perform on demand.

- Select value for X_0
- Select value for L_s , consistent with X_0
- Select value for L_e , consistent with X_0, L_s



Scheduling without Search: Solution by Decomposition

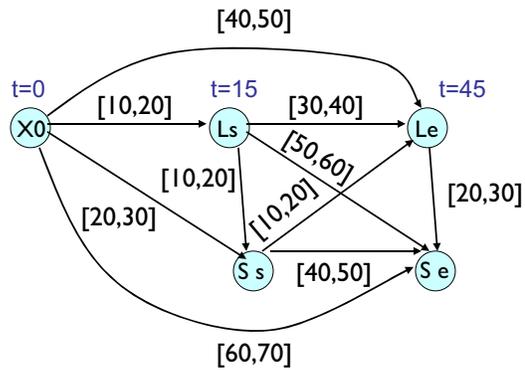


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Scheduling without Search: Solution by Decomposition

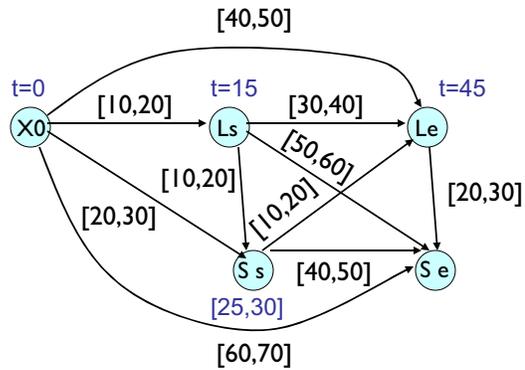


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- Select value for X_0
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- Select value for L_e , consistent with X_0, L_s
- Select value for S_s , consistent with X_0, L_s, L_e



Scheduling without Search: Solution by Decomposition

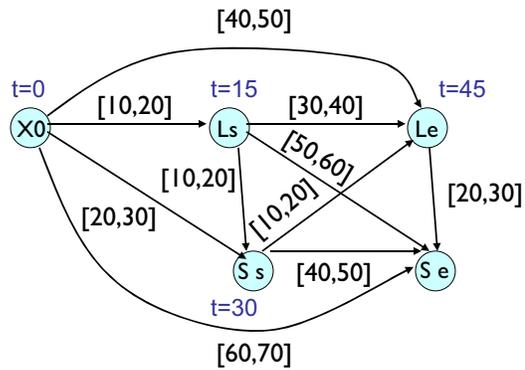


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Key ideas

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- Select value for S_s , consistent with X_0, L_s, L_e



Scheduling without Search: Solution by Decomposition

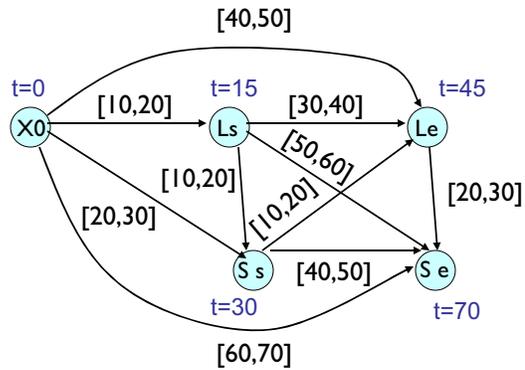


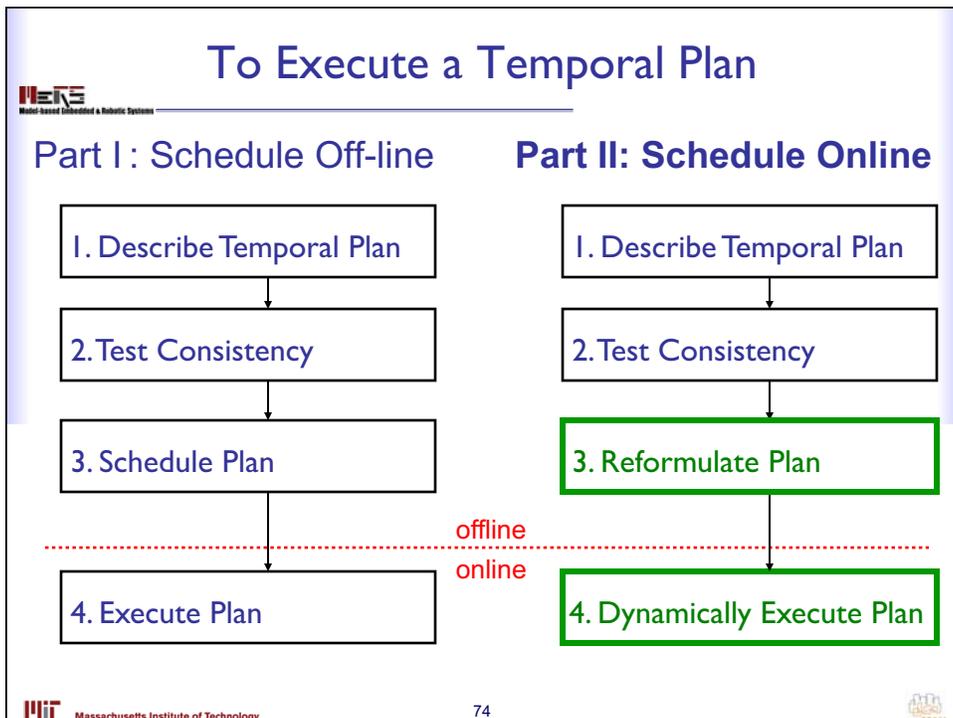
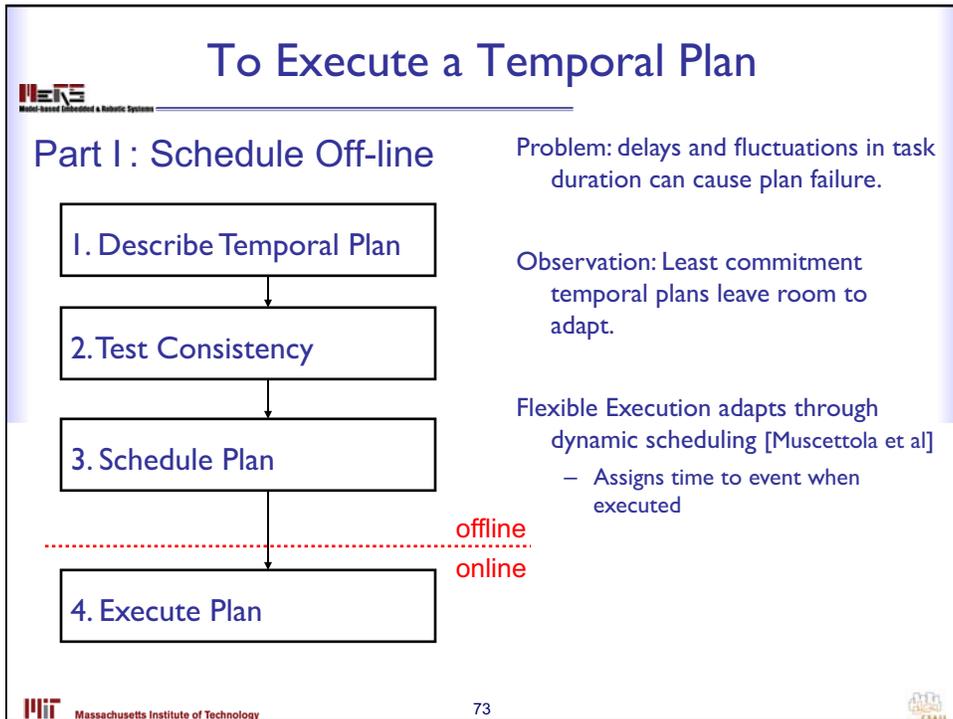
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Key ideas

- Incrementally tighten feasible intervals, as commitments are made.
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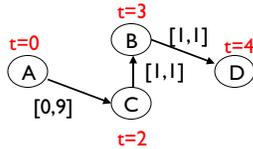
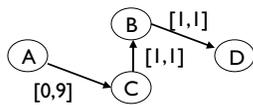
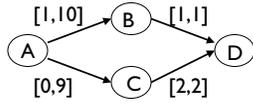
- Select value for X_0
- Select value for L_s , consistent with X_0
- Select value for L_e , consistent with X_0, L_s
- Select value for S_s , consistent with X_0, L_s, L_e
- Select value for S_e ...





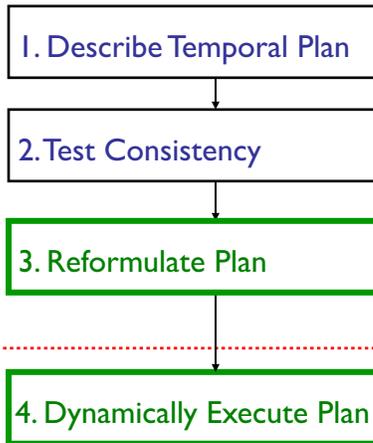
To Execute a Temporal Plan

How do we schedule on line?



offline
online

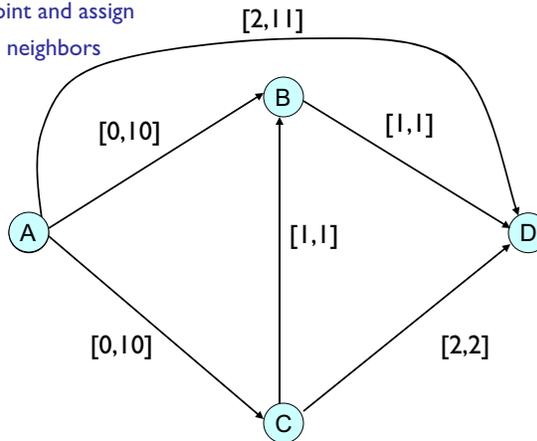
Part II: Schedule Online



Dynamic Scheduling by Decomposition?

Consider a Simple Example

- Select executable timepoint and assign
- Propagate assignment to neighbors

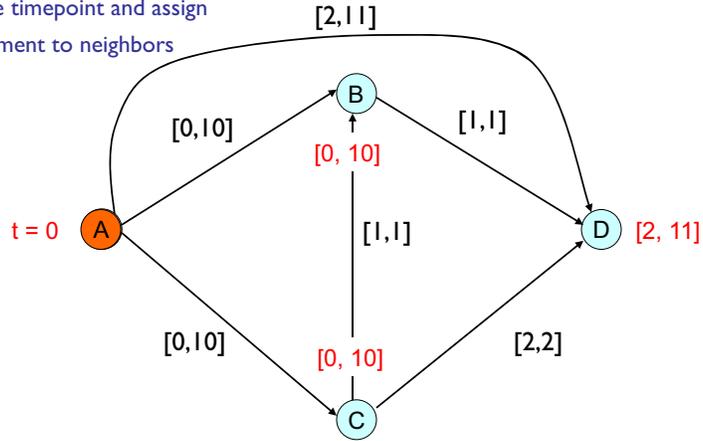


Dynamic Scheduling by Decomposition?



Consider a Simple Example

- Select executable timepoint and assign
- Propagate assignment to neighbors



Dynamic Scheduling by Decomposition?

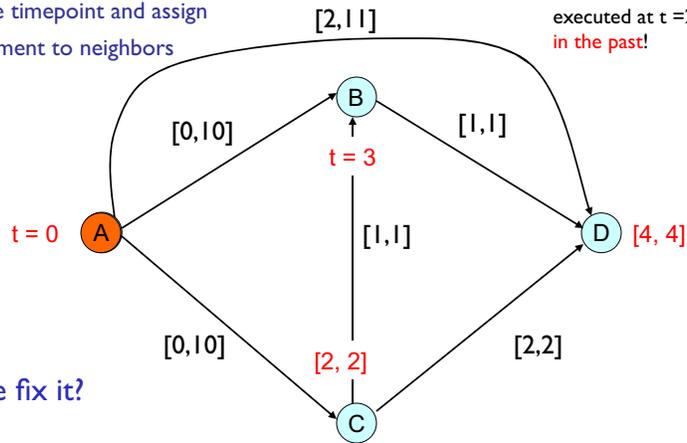


Consider a Simple Example

- Select executable timepoint and assign
- Propagate assignment to neighbors

Uh oh!

C must be executed at $t=2$ in the past!



How can we fix it?

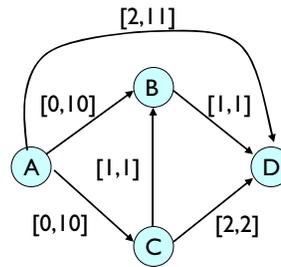


Dispatching Execution Controller



Model-based Embedded & Robotic Systems

- How can we fix it?
 - Assignments must monotonically increase in value.
 - Respect induced orderings.
- Execute an event when **enabled** and **alive**
 - Enabled – Predecessors are completed
 - Alive – Current time within bound of task



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Dispatching Execution Controller



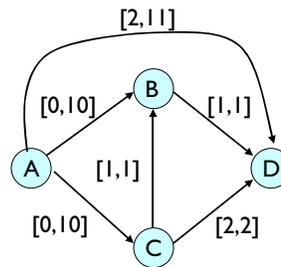
Model-based Embedded & Robotic Systems

Initially:

- $E =$ Time points w/o predecessors
- $S = \{ \}$

Repeat:

1. Wait until current time has advanced such that some TP in E is active
2. Set TP's execution time to current time.
3. Add TP to S .
4. Propagate time of execution to TP's immediate neighbors
5. Add to E , all immediate neighbors that become enabled
 - TP enabled if all +lb edges starting at TP have their destination in S .



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Dynamic Scheduling through Dispatchable Execution

MESE
Model-based Embedded & Robotic Systems

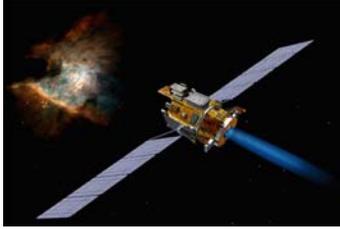
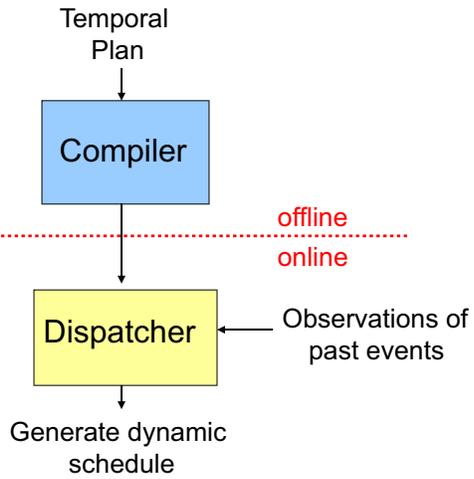


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