

# STRANDS

Task scheduling and execution  
for long-term autonomy

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# Long-Term Autonomy in Everyday Environments



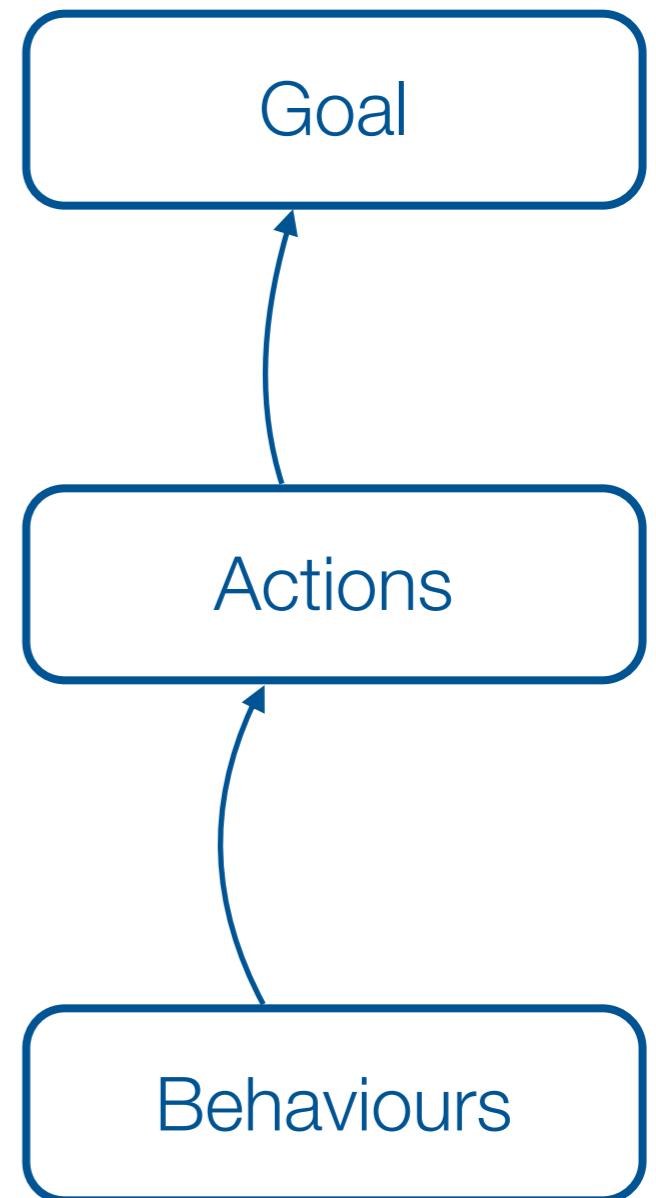


## Goal-Oriented Behaviour

We assume our robot has *goals* which are either provided by a user or generated by an internal system

Some system will provide a sequence of *actions* which achieve a given goal

Each *action* maps to an underlying *behaviour* which is the implementation of the *action* on the robot



Planning gives us an **ordering** of actions to achieve a goal

Real-world problems also need **time** and **resources**

**Scheduling** assigns time and resources to **jobs**

A **job** is a collection of **actions** with ordering constraints. Each action has a **duration**.

The aim is to make an **assignment of times to actions** (a **schedule**) in order to achieve some criterion, e.g. **makespan**.

*Jobs( {AddEngine1 < AddWheels1 < Inspect1},  
      {AddEngine2 < AddWheels2 < Inspect2} )*

*Action( AddEngine1, DURATION:30)*

*Action( AddEngine2, DURATION:60)*

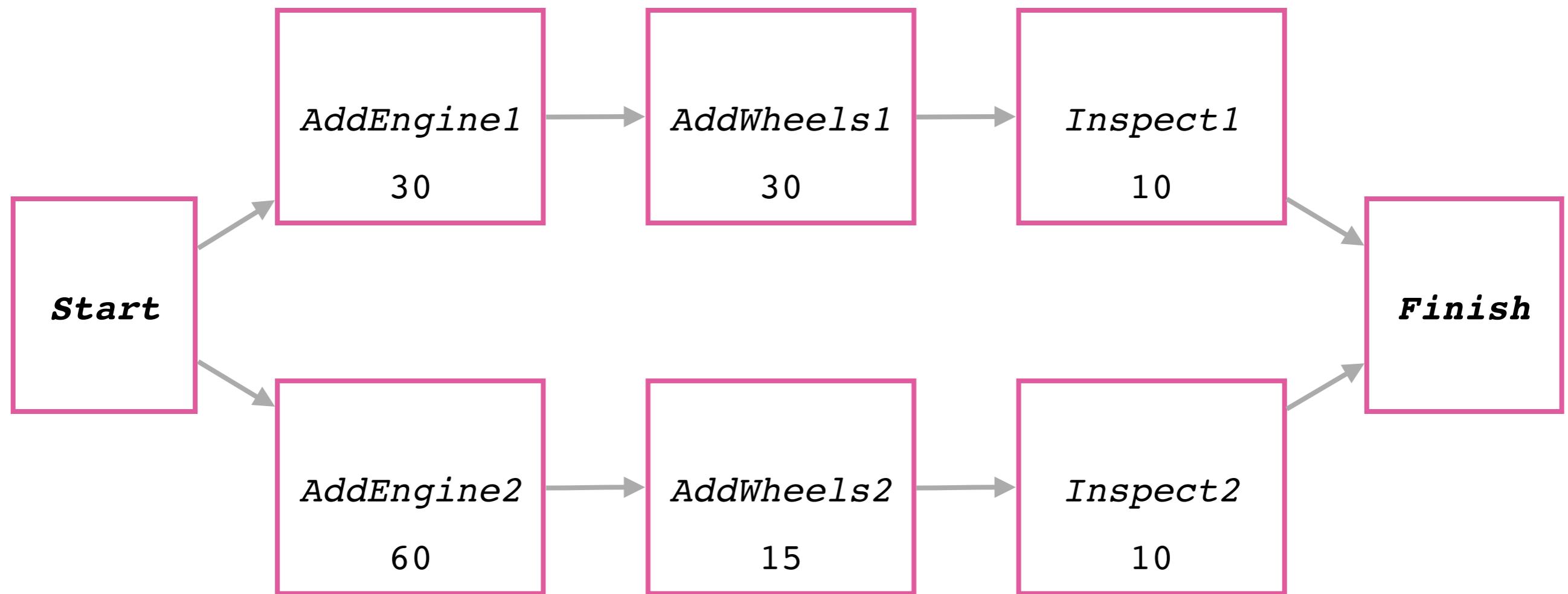
*Action( AddWheels1, DURATION:30)*

*Action( AddWheels1, DURATION:15)*

*Action( Inspect, DURATION:10)*

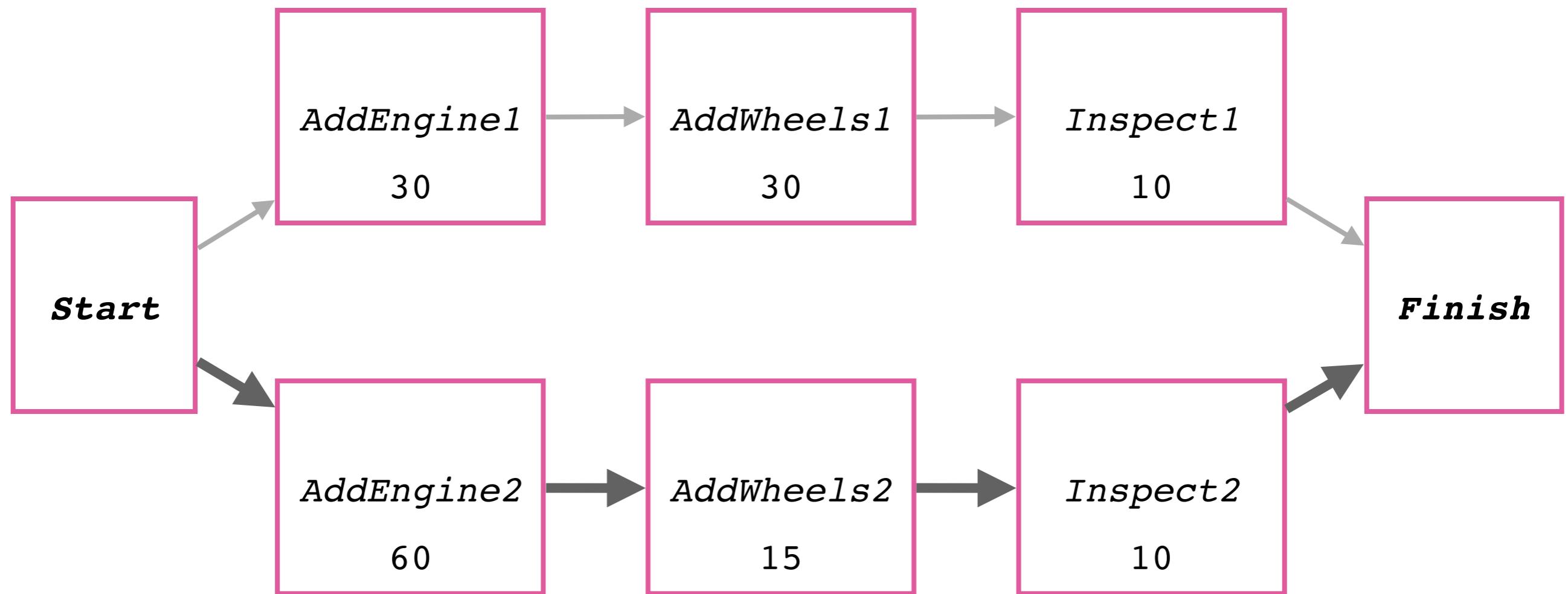
For each action assign  
**earliest start** time  $ES$  and **latest start** time  $LS$

The **critical path** method: define the path through the action graph with longest duration

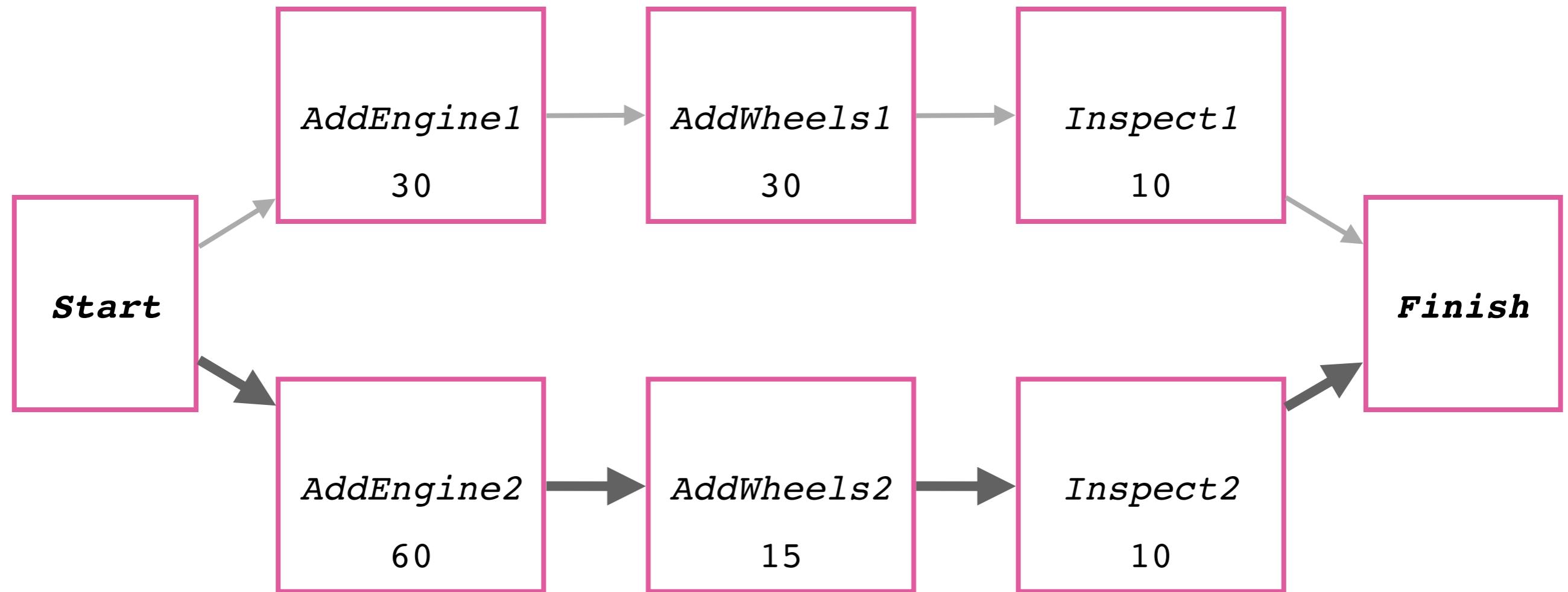


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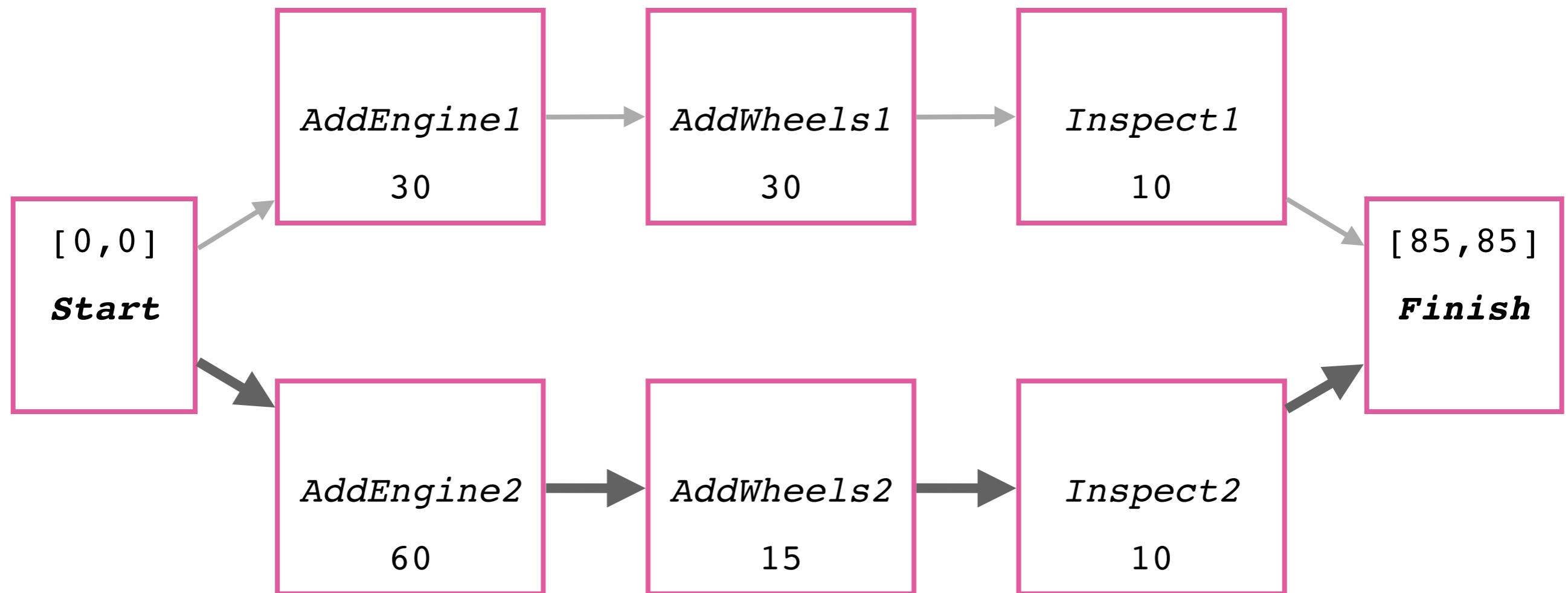


Use the **critical path** to define the overall length of the schedule,  
i.e. [ *ES*, *LS* ] for *Start* and *Finish*



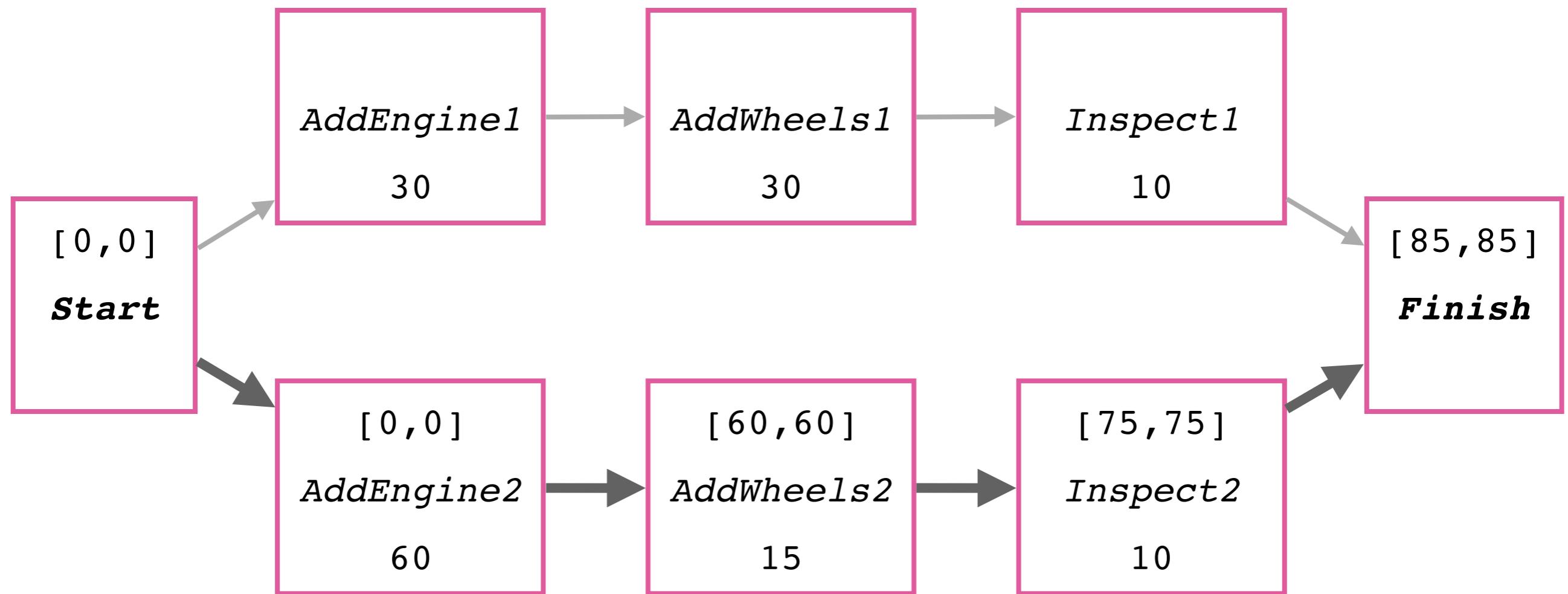
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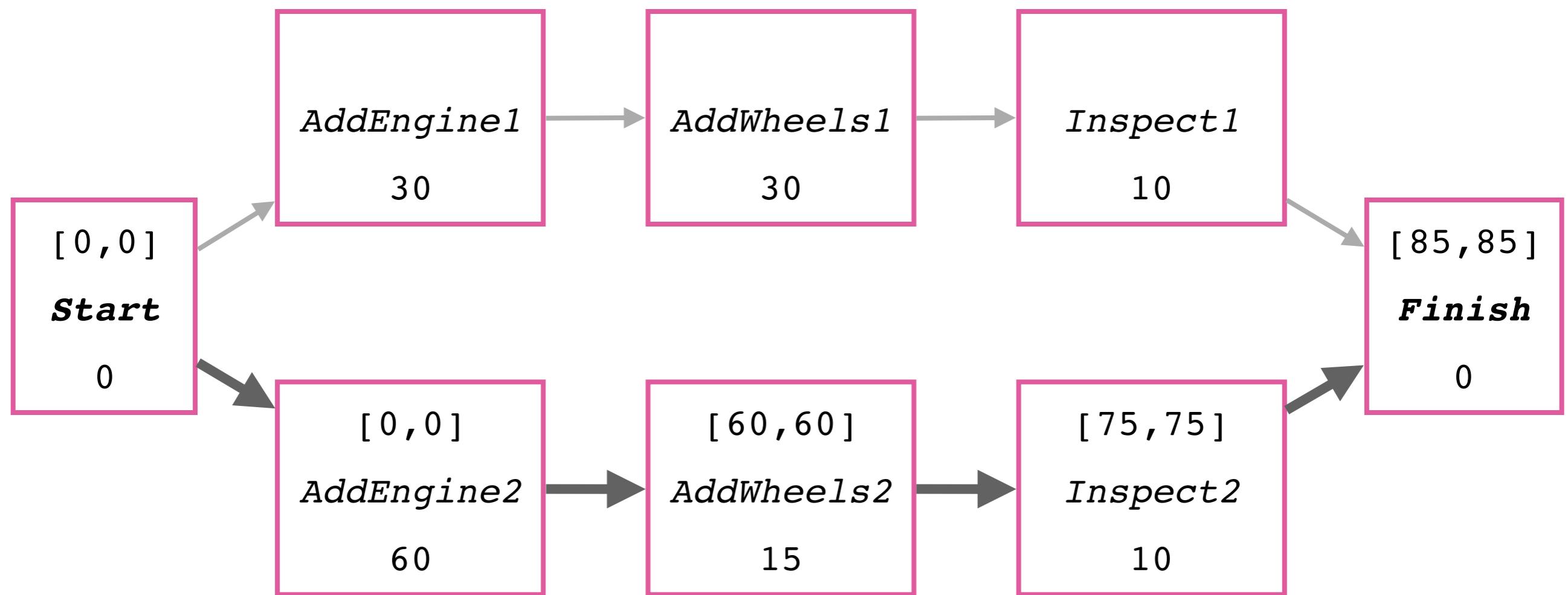
and for the actions on the critical path



Use these constraints to complete [ *ES*, *LS* ]  
for the remaining actions

$$ES(B) = \max_{A < B} ES(A) + Duration(A)$$

$$LS(A) = \min_{B > A} LS(B) - Duration(A)$$

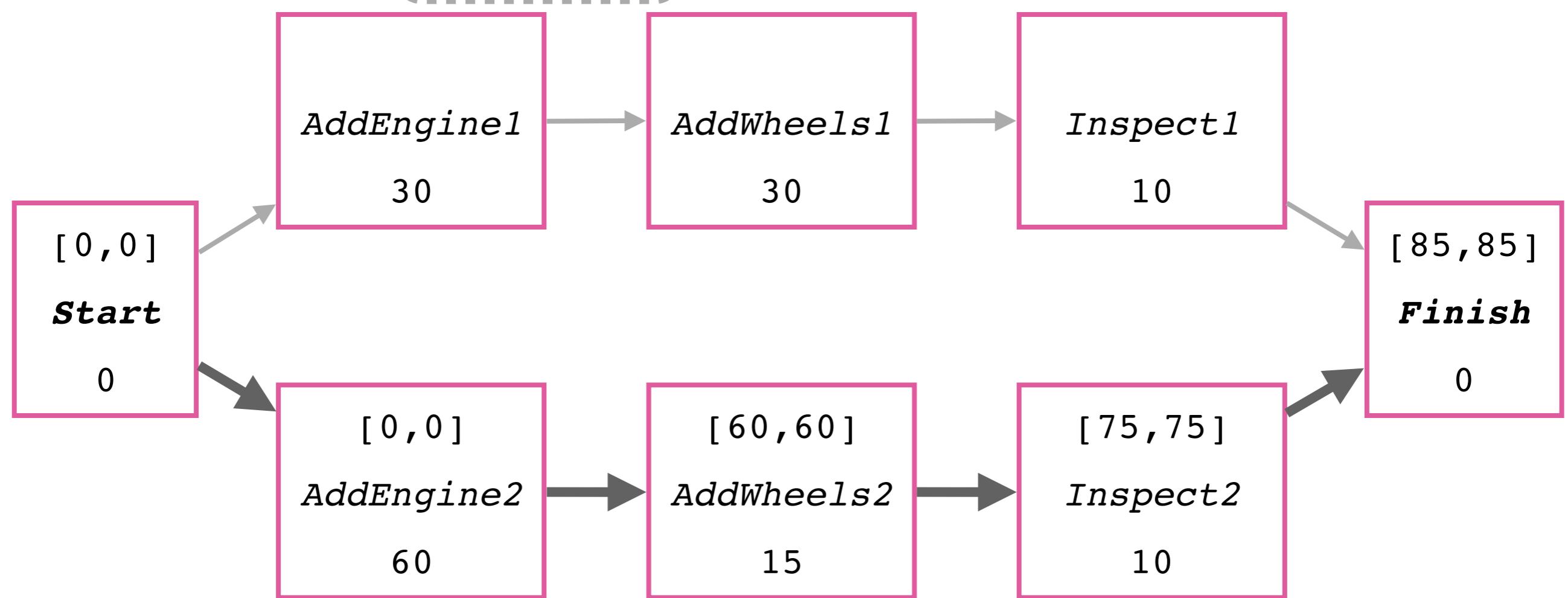


Action with *latest* earliest  
start preceding  $B$

Action with *earliest* latest  
start preceding  $A$

$$ES(B) = \max_{A < B} ES(A) + Duration(A)$$

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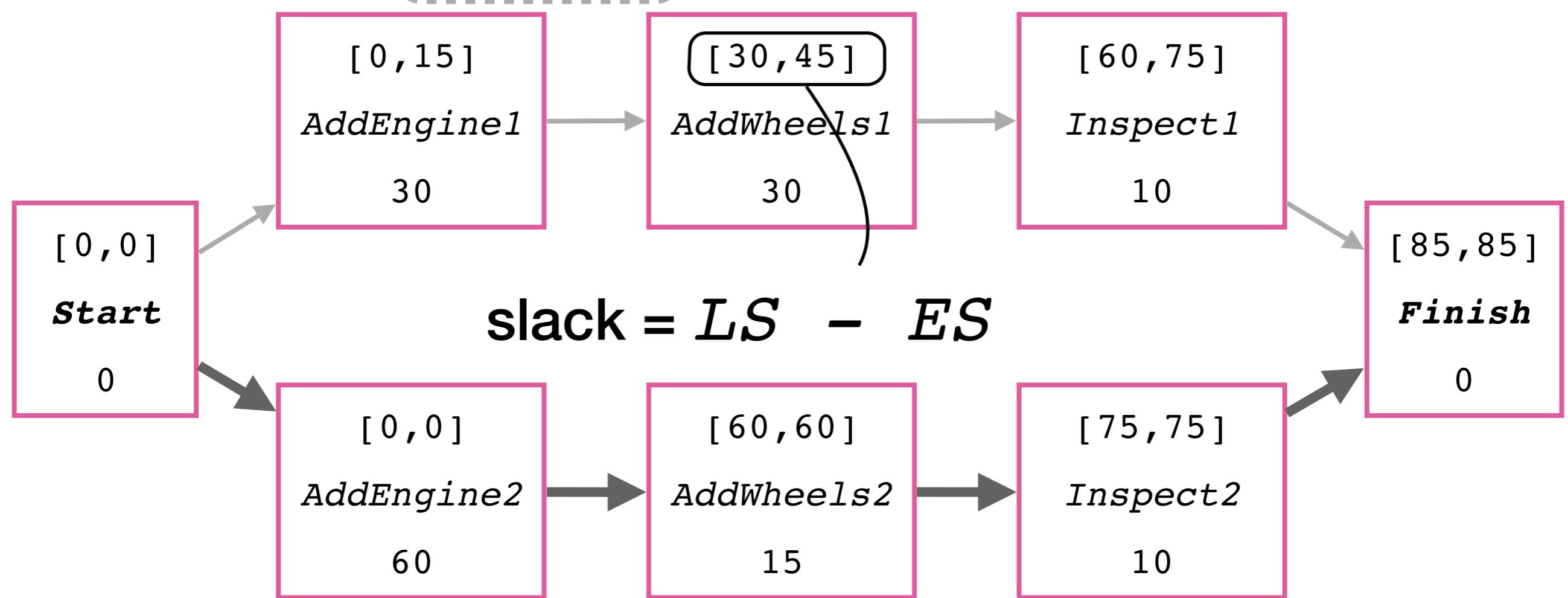


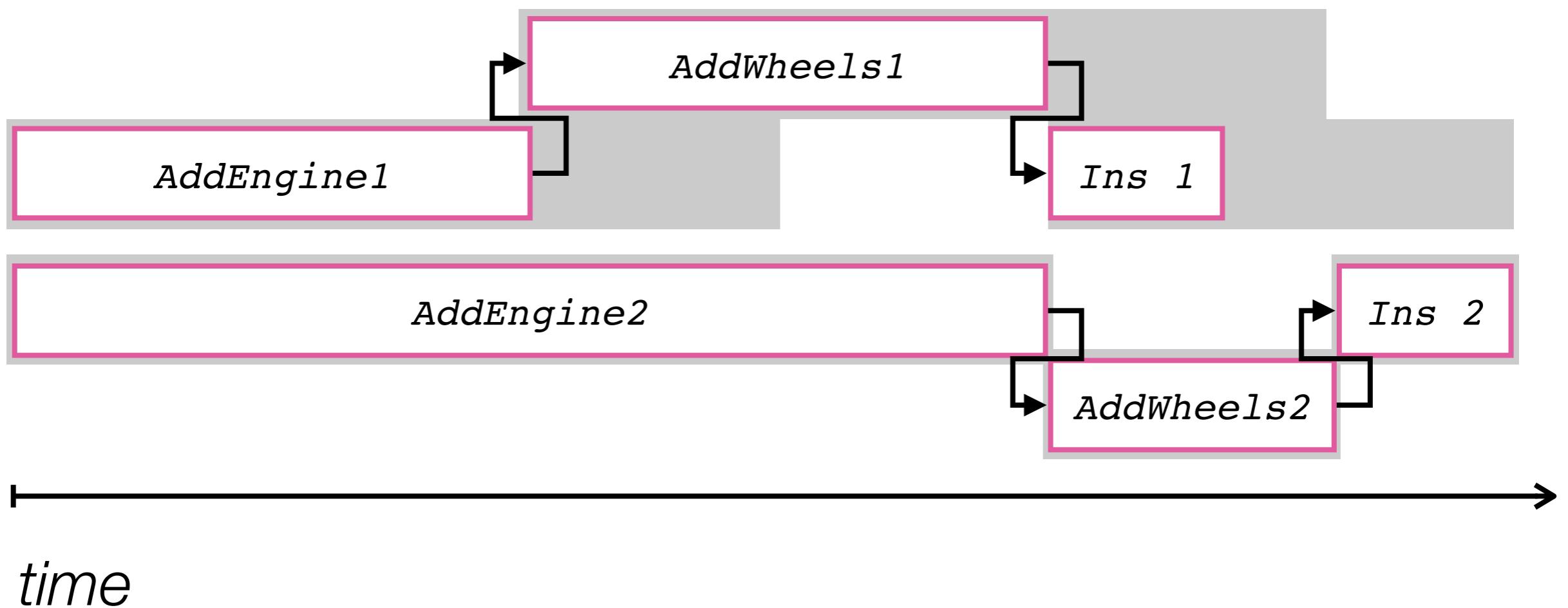
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$$ES(Start) = 0$$

$$ES(B) = \max_{A < B} ES(A) + Duration(A)$$

$$LS(Finish) = ES(Finish)$$

$$LS(A) = \min_{B > A} LS(B) - Duration(A)$$

Scheduling ordered tasks with no additional constraints is pretty easy: a **conjunction of linear constraints**

Solve with **dynamic programming, integer programming** etc.

*Jobs( {AddEngine1 < AddWheels1 < Inspect1},  
      {AddEngine2 < AddWheels2 < Inspect2} )*

*Action( AddEngine1, DURATION:30)*

*Action( AddEngine2, DURATION:60)*

*Action( AddWheels1, DURATION:30)*

*Action( AddWheels1, DURATION:15)*

*Action( Inspect, DURATION:10)*

*Jobs( {AddEngine1 < AddWheels1 < Inspect1},  
{AddEngine2 < AddWheels2 < Inspect2} )*

*Resources( EngineHoists(1), WheelStations(1),  
Inspectors(2), LugNuts(500) )*

*Action( AddEngine1, DURATION:30)*

*Action( AddEngine2, DURATION:60)*

*Action( AddWheels1, DURATION:30)*

*Action( AddWheels1, DURATION:15)*

*Action( Inspect, DURATION:10)*

*Jobs( {AddEngine1 < AddWheels1 < Inspect1},  
{AddEngine2 < AddWheels2 < Inspect2} )*

*Resources( EngineHoists(1), WheelStations(1),  
Inspectors(2), LugNuts(500) )*

*Action( AddEngine1, DURATION:30,  
**USE**: EngineHoists(1))*

*Action( AddEngine2, DURATION:60,  
**USE**: EngineHoists(1)))*

*Action( AddWheels1, DURATION:30,  
**CONSUME**: LugNuts(20), **USE**: WheelStations(1)))*

*Action( AddWheels1, DURATION:15,  
**CONSUME**: LugNuts(20), **USE**: WheelStations(1)))*

*Action( Inspect, DURATION:10,  
**USE**: Inspectors(1)))*

$$ES(Start) = 0$$

$$ES(B) = \max_{A < B} ES(A) + Duration(A)$$

$$LS(Finish) = ES(Finish)$$

$$LS(A) = \min_{B > A} LS(B) - Duration(A)$$

Now we have to include **disjunctions** so we're back to an **NP-hard** problem.

Scheduling for a **mobile robot** introduces both challenges and simplifications.

A **task** is a single, indivisible unit of behaviour to achieve a goal (often implicit)

Actions

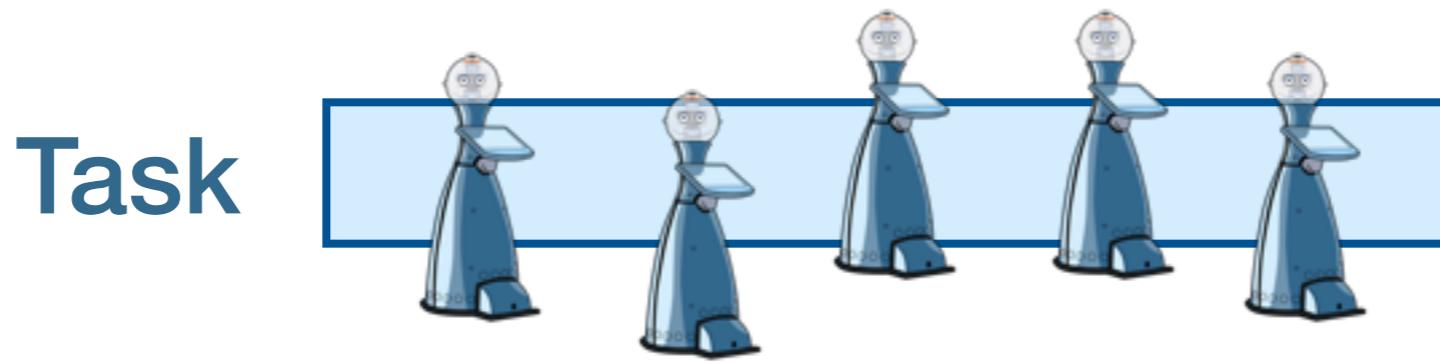
Goal

*Task Scheduling for Mobile Robots Using Interval Algebra*  
Mudrová and Hawes. In, ICRA '15.

How to tell a robot **what time** to do something?

Not just order, but precise starting times (e.g. 14:02)

Considering up to 100 tasks



Task

*task<sub>i</sub>*

*s<sub>i</sub>*

*d<sub>i</sub>*

*e<sub>i</sub>*

*t<sub>i</sub>*

*action*

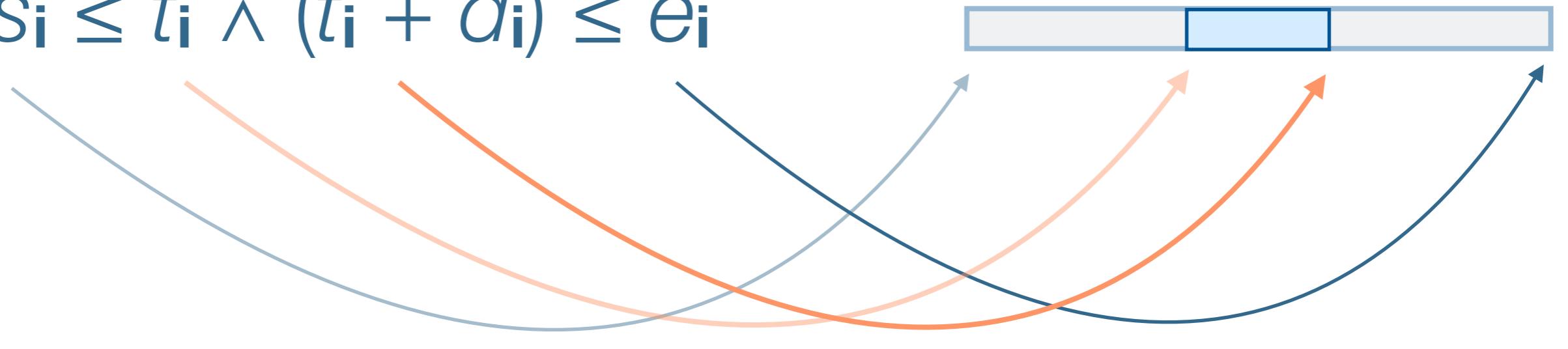
*p<sup>s</sup><sub>i</sub>*

*p<sup>e</sup><sub>i</sub>*



$$\forall i : \min \sum (t_i - s_i)$$

$$s_i \leq t_i \wedge (t_i + d_i) \leq e_i$$



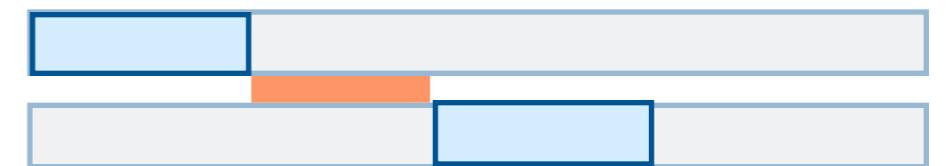
\* e.g. Brian Coltin, Manuela Veloso, and Rodrigo Ventura. *Dynamic User Task Scheduling for Mobile Robots*. In Proceedings of the AAAI Workshop on Automated Action Planning for Autonomous Mobile Robots at AAAI. 2011.

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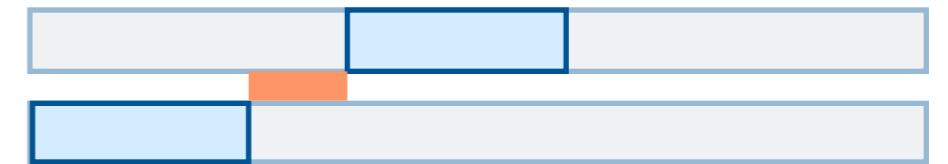
$$\forall i, j : t_i + d_i + time(p^e_i, p^s_j) \leq t_j$$

or

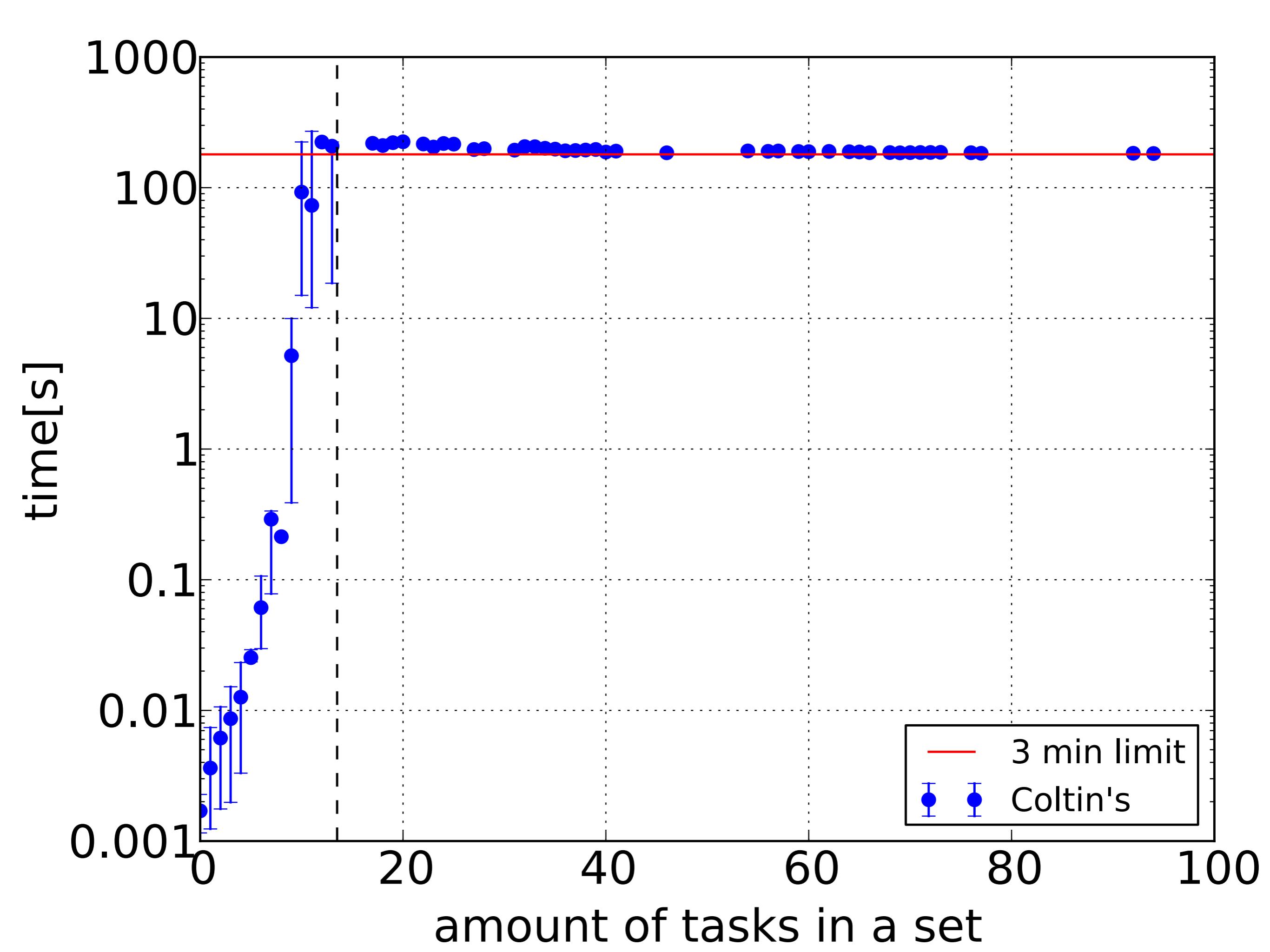


$$\forall i, j : t_j + d_j + time(p^e_j, p^s_i) \leq t_i$$

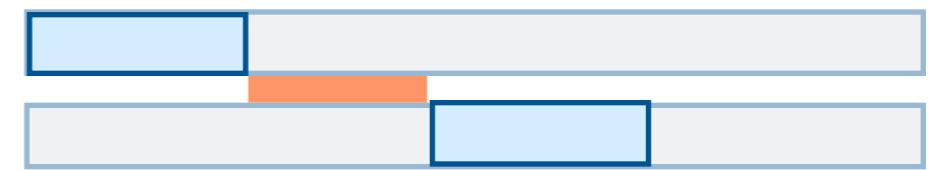
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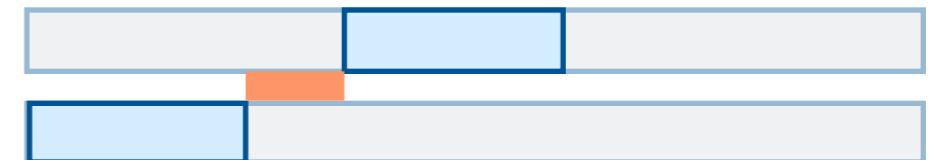
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$$\forall i,j : t_i + d_i + \text{time}(p^e_i, p^s_j) \leq t_j$$



$$\forall i,j : t_j + d_j + \text{time}(p^e_j, p^s_i) \leq t_i$$



# Reasoning about tasks' time windows: $S_i \rightarrowtail S_j$ vs $S_j \rightarrowtail S_i$

no constraint

$i$  precedes  $j$

$j$  precedes  $i$

both

$i$  before  $j$

$i$  overlaps  $j$

$j$  overlaps  $i$

$i$  during  $j$

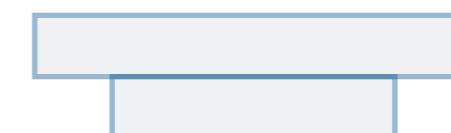
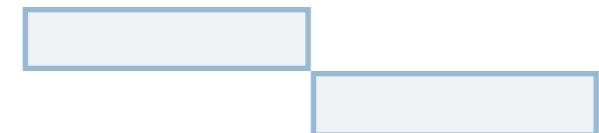


$i$  meets  $j$

$i$  starts  $j$

$j$  starts  $i$

$j$  during  $i$



$j$  before  $i$

$j$  finishes  $i$

$i$  finishes  $j$

$i$  equals  $j$



$j$  meets  $i$

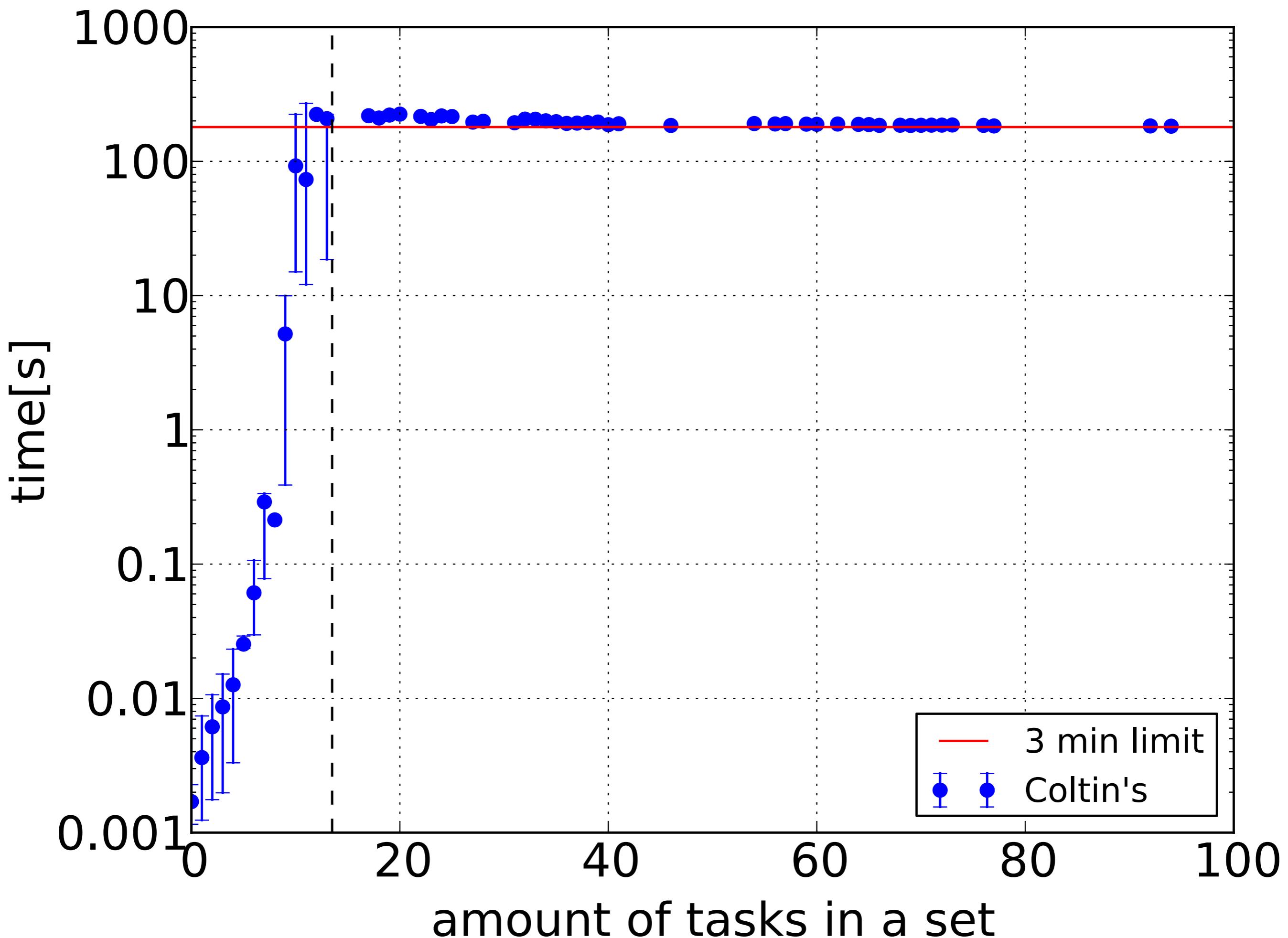


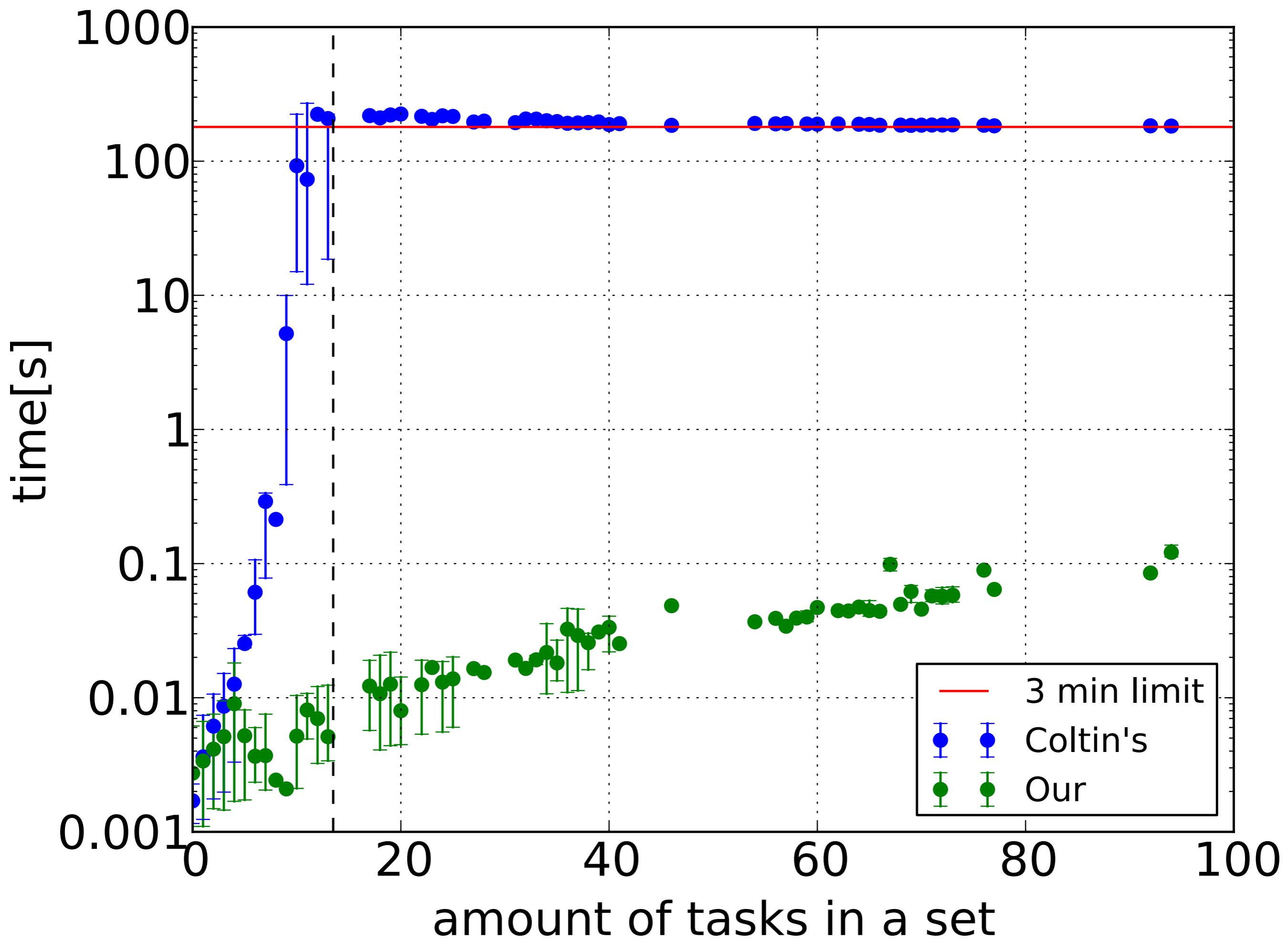
**Allen's Interval Algebra**

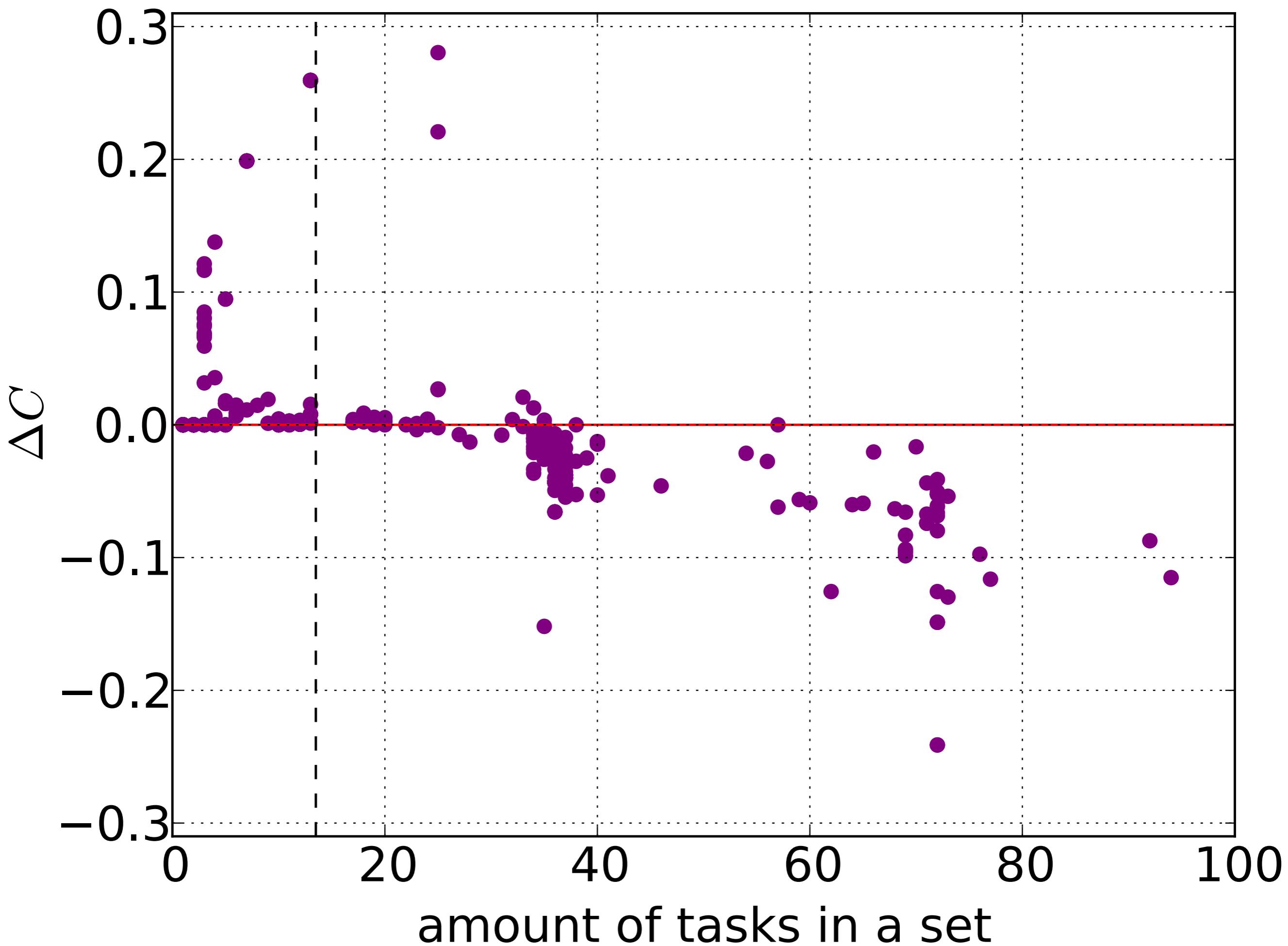
J. F. Allen. *Maintaining knowledge about temporal intervals.* Communications of the ACM, 26(11):832– 843, 1983.

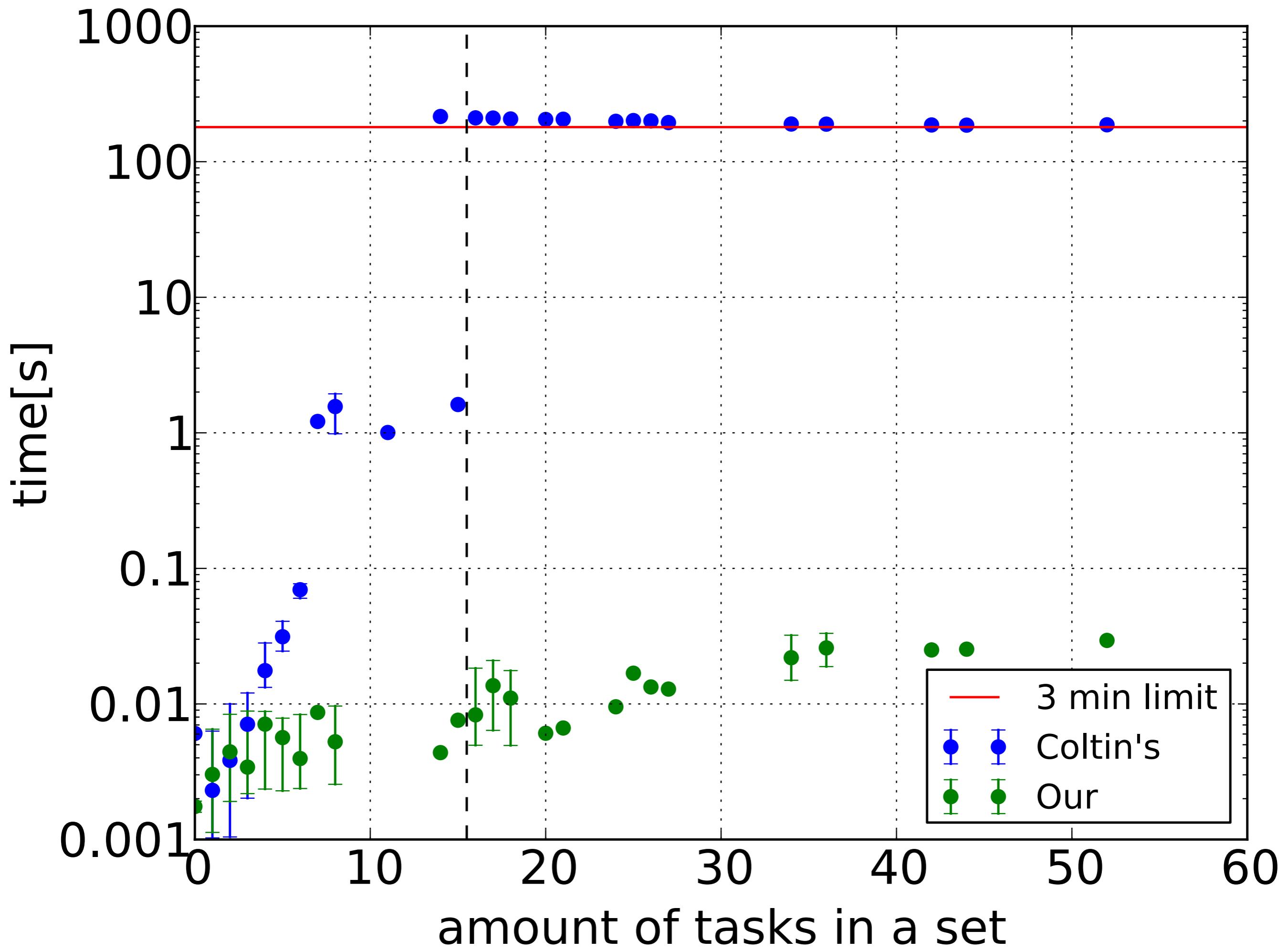


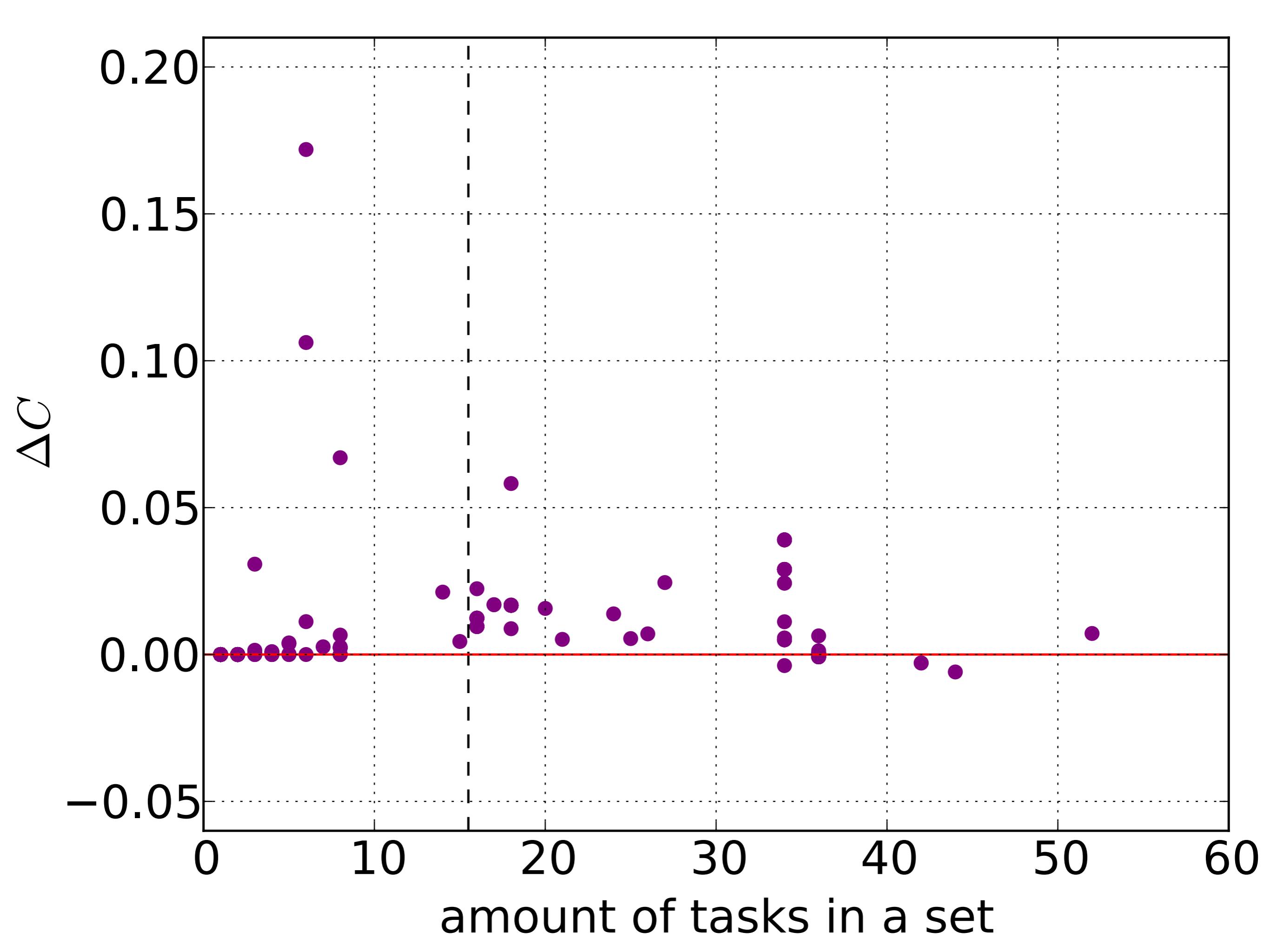
	Care	Security
# Problems	606	358
Smallest Problem	1	1
Largest Problem	135	71
Mean Problem Size	28.88 ( $\sigma$ 26.28)	9.59 ( $\sigma$ 12.97)
# Problems >15	349 (58%)	106 (30%)

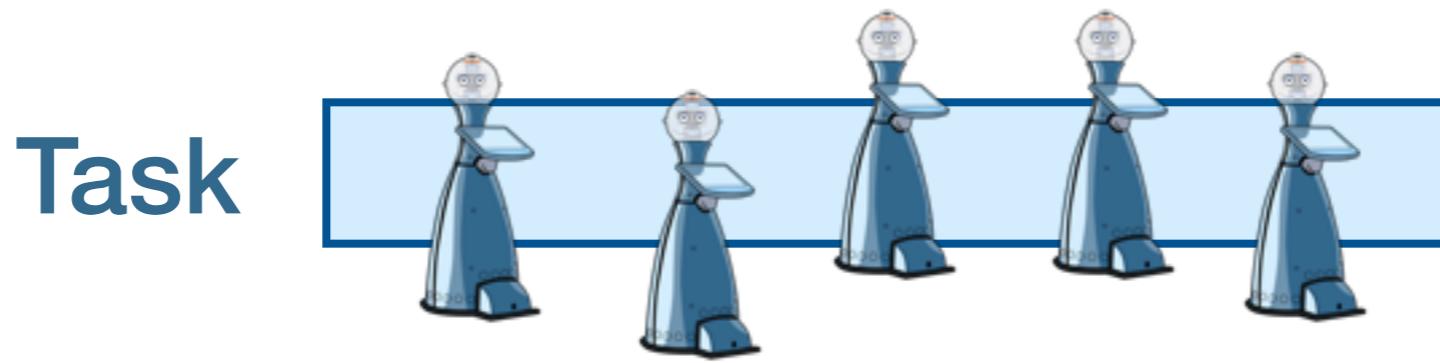




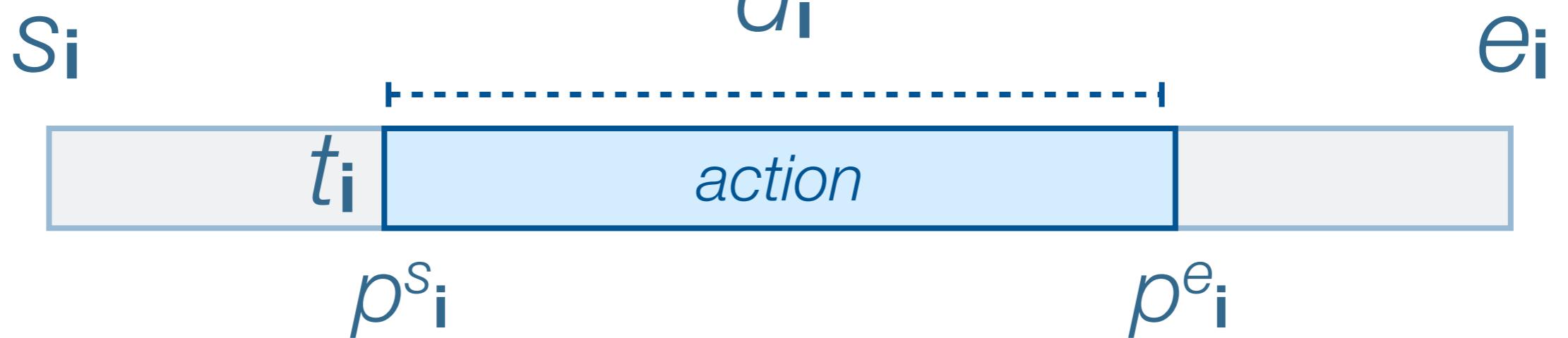


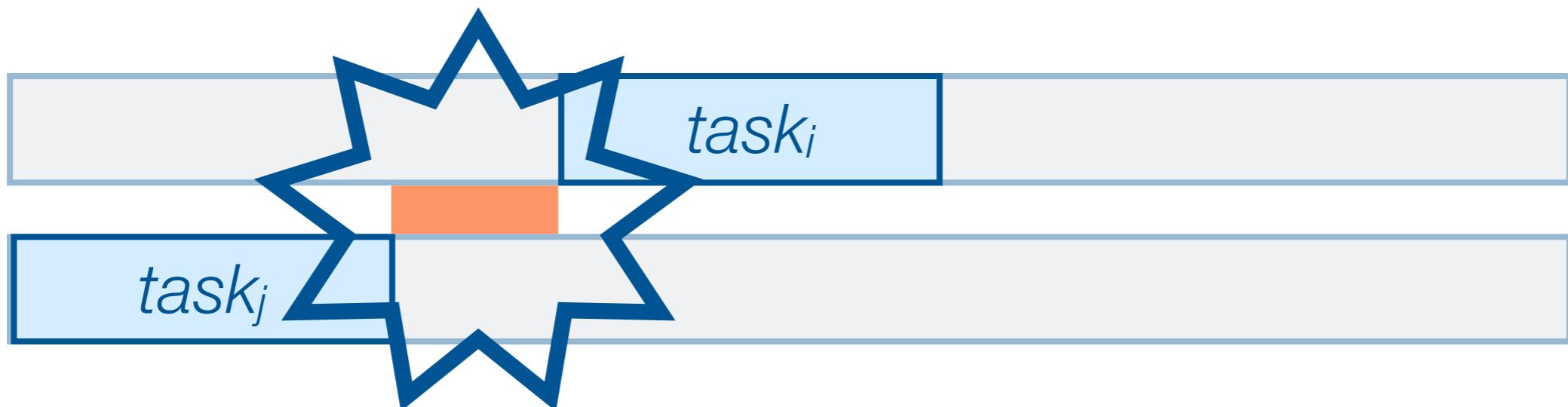
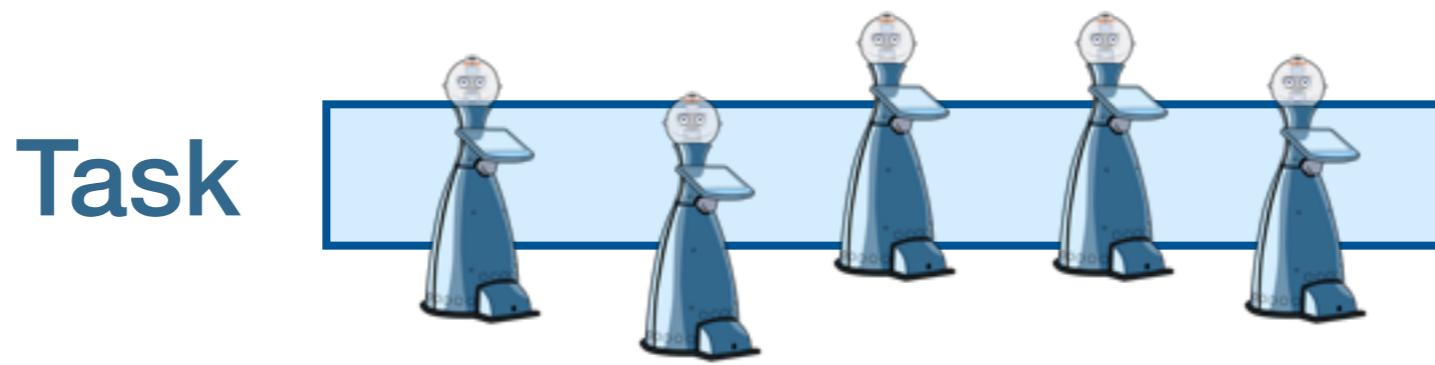






*task<sub>i</sub>*

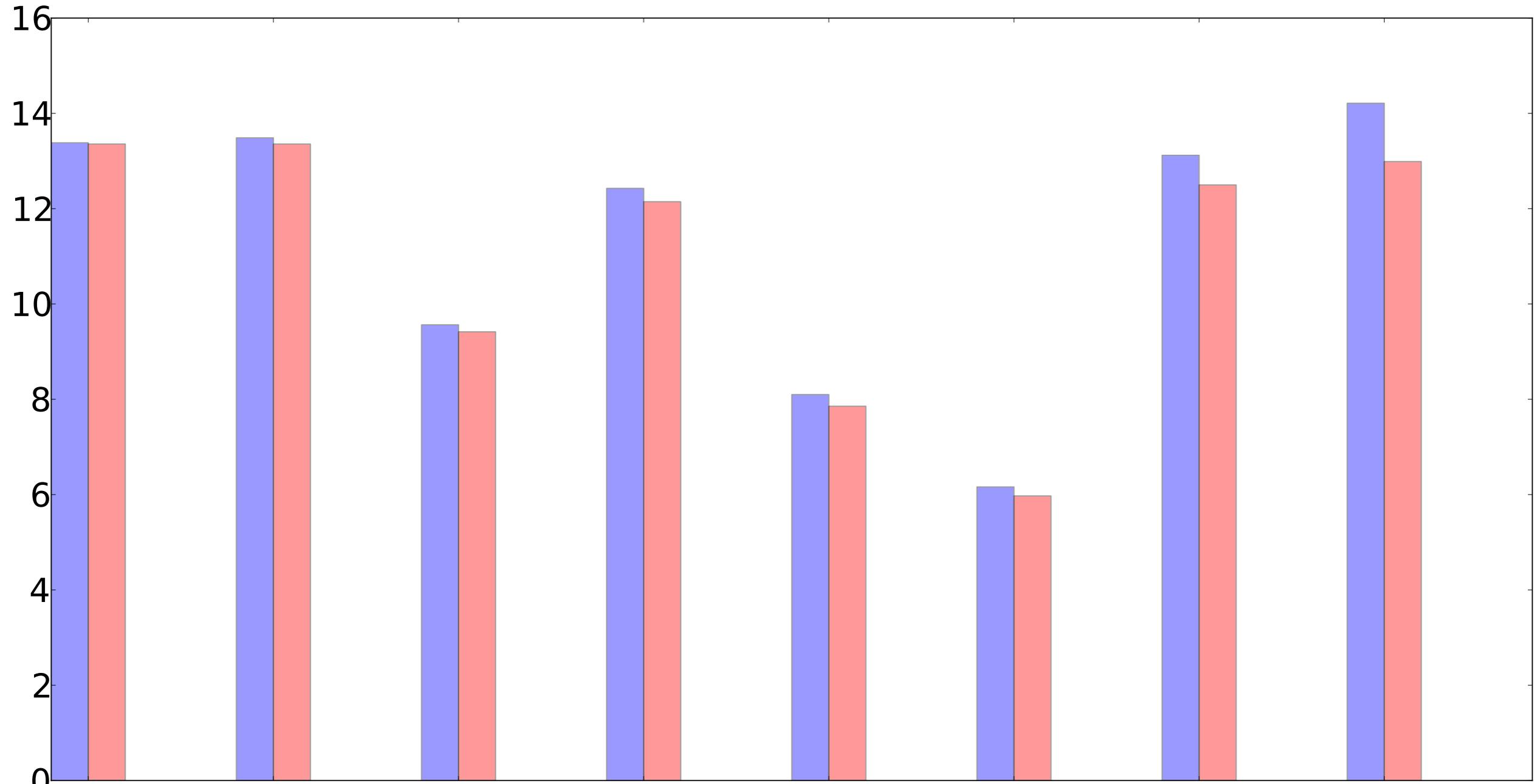




*Optimal and Dynamic Planning for  
Markov Decision Processes with Co-Safe LTL Specifications*  
Lacerda, Parker and Hawes. In, IROS'14.



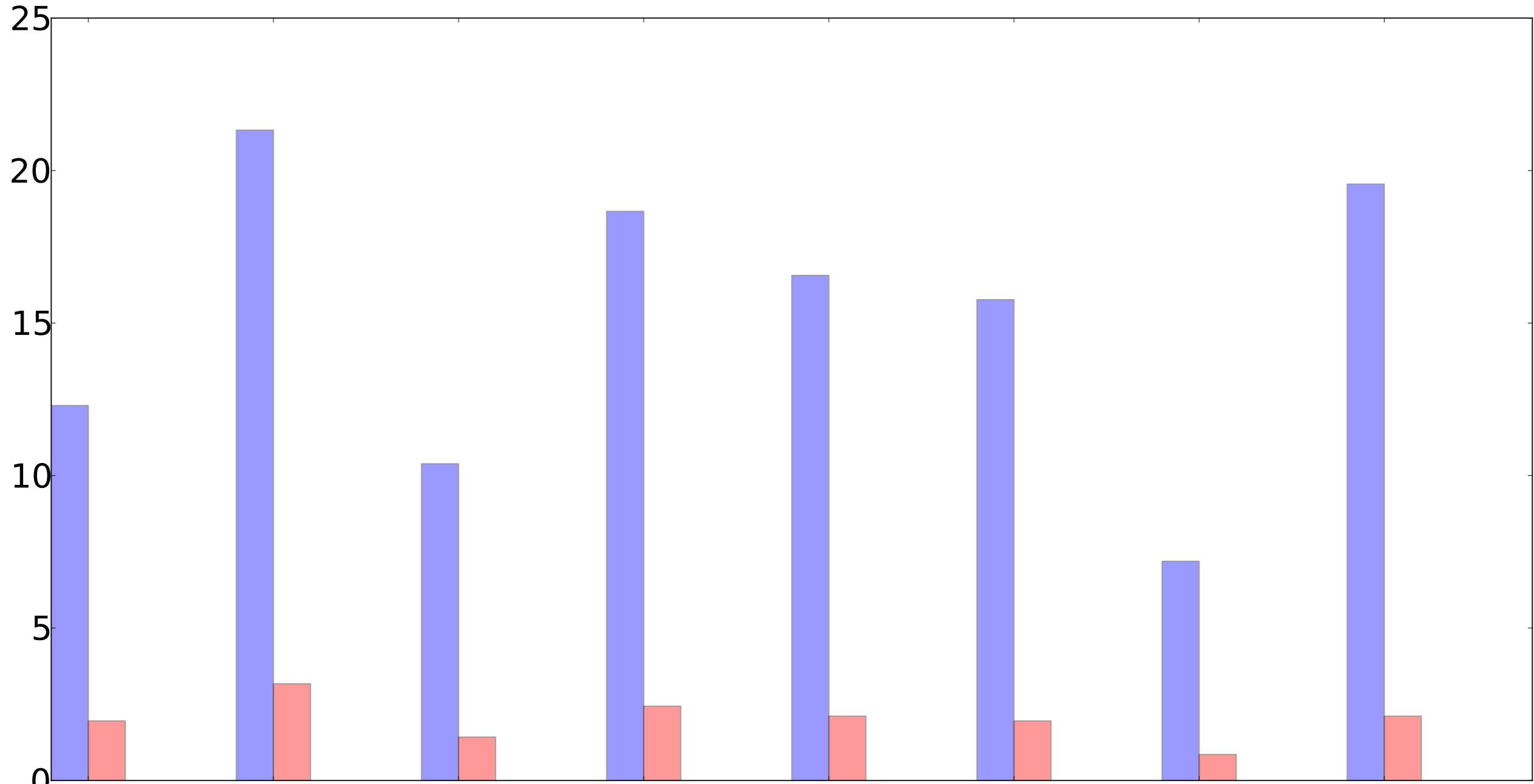
## Best 8 matches between straight-line and recorded times



mean time from  
robot

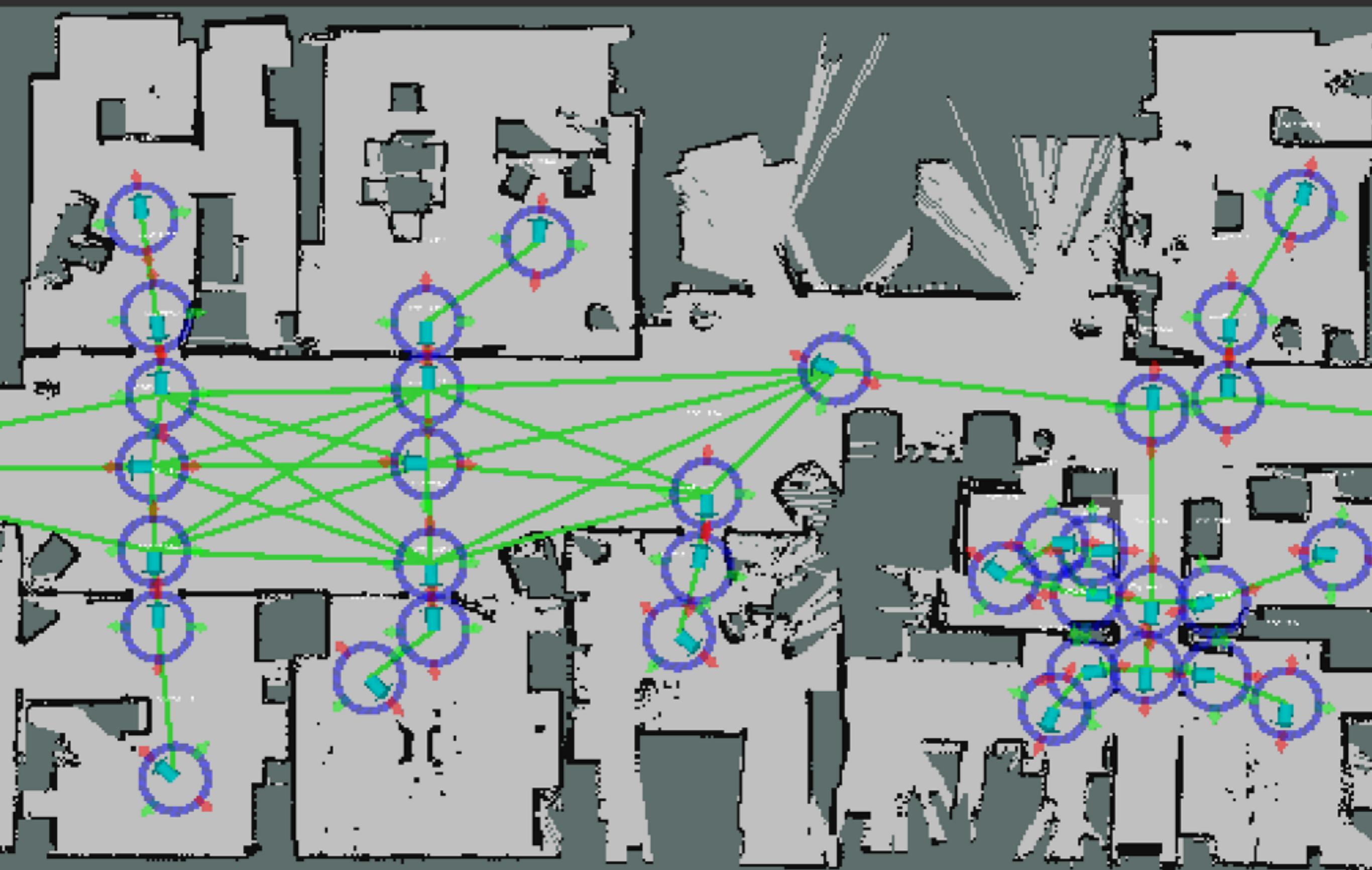
straight line time

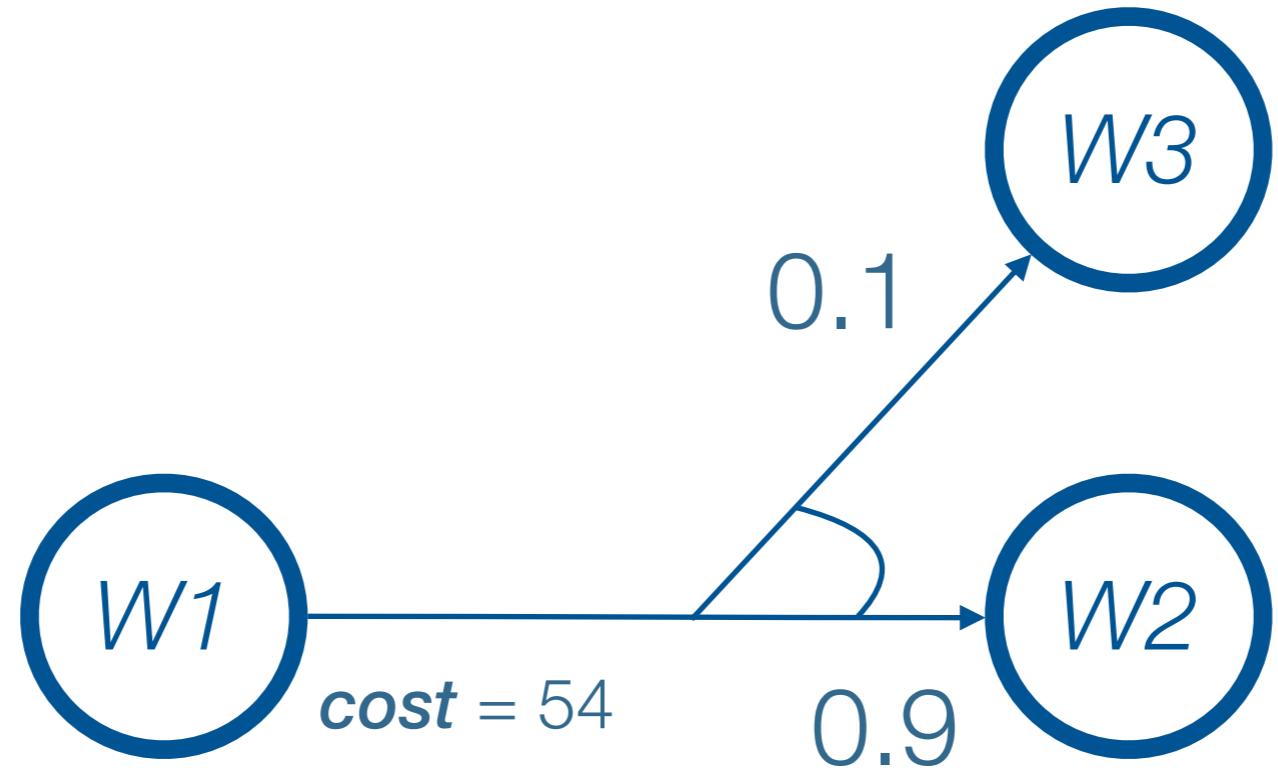
## Worst 8 matches between straight-line and recorded times



█ mean time from  
robot

█ straight line time



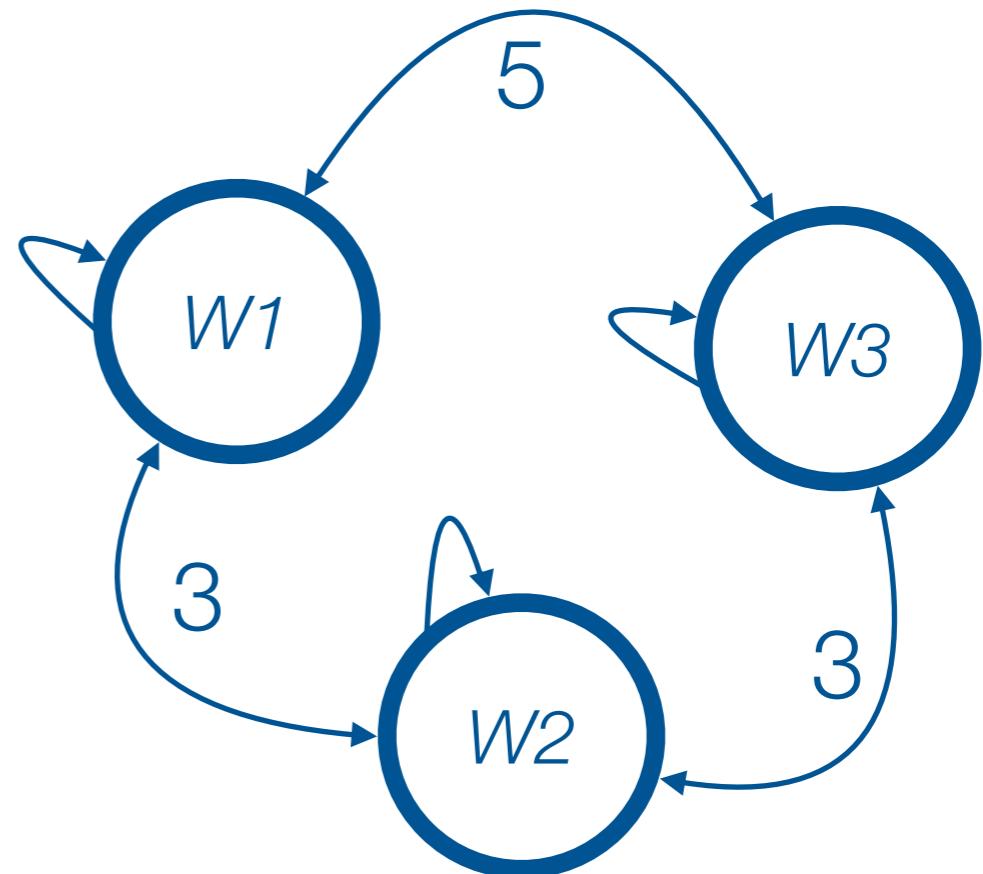
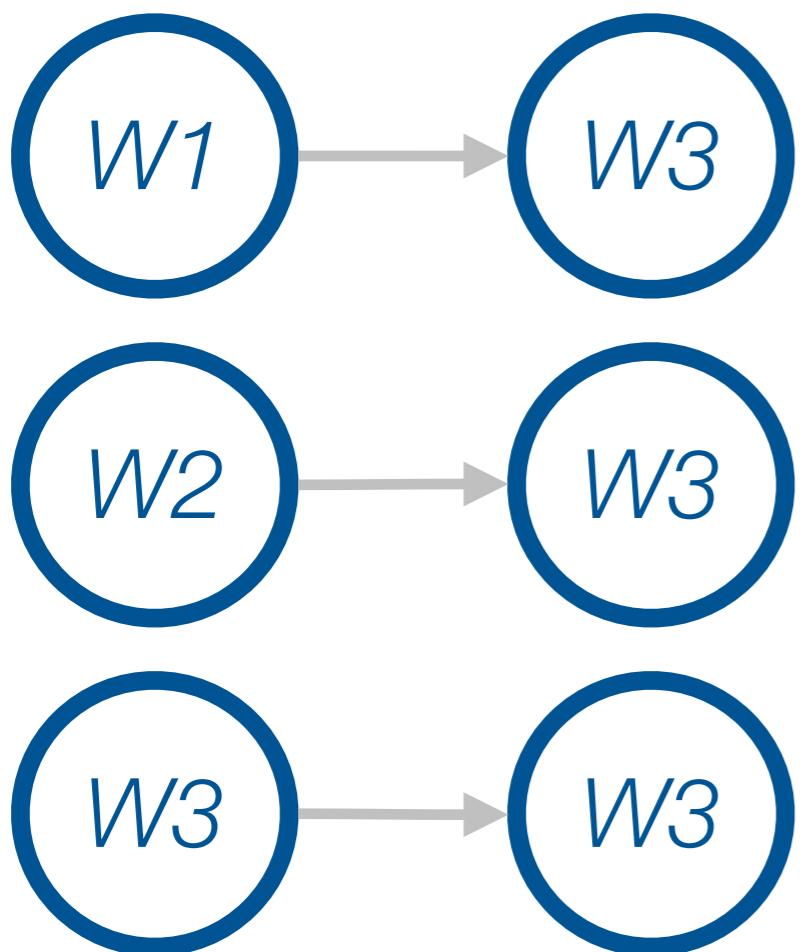


*action* goto  $W_2$  from  $W_1$

Why use an MDP?

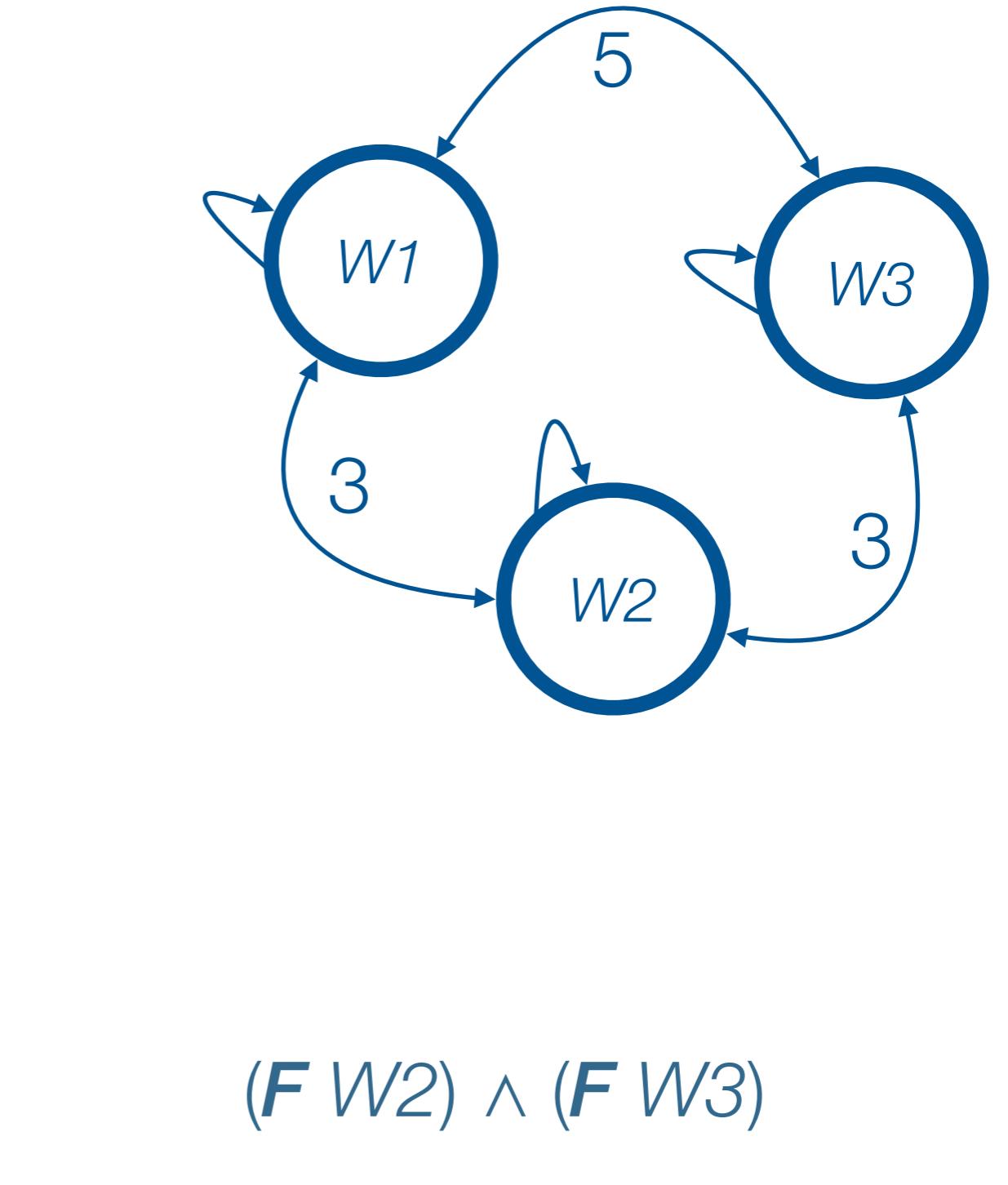
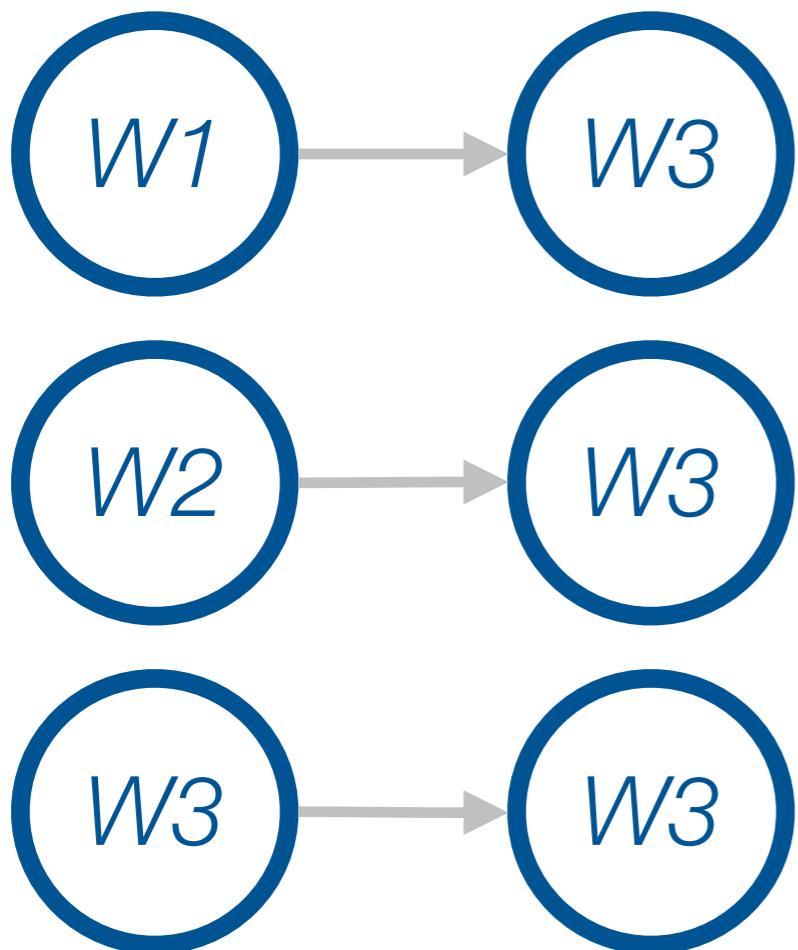
Goal is to be in state  $W_3$

**Policy:**



$(F W2)$   
eventually reach  $W2$

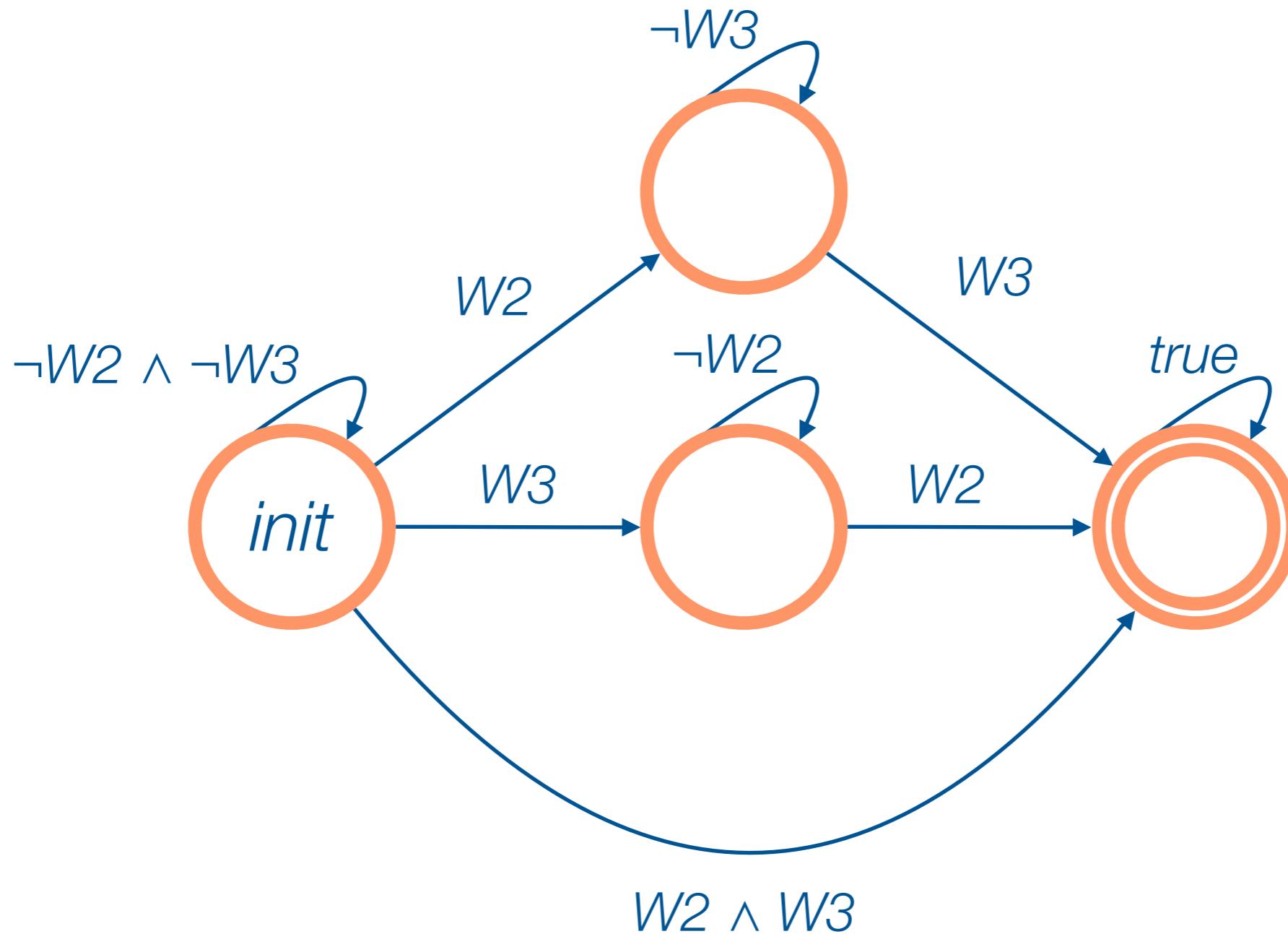
Policy:

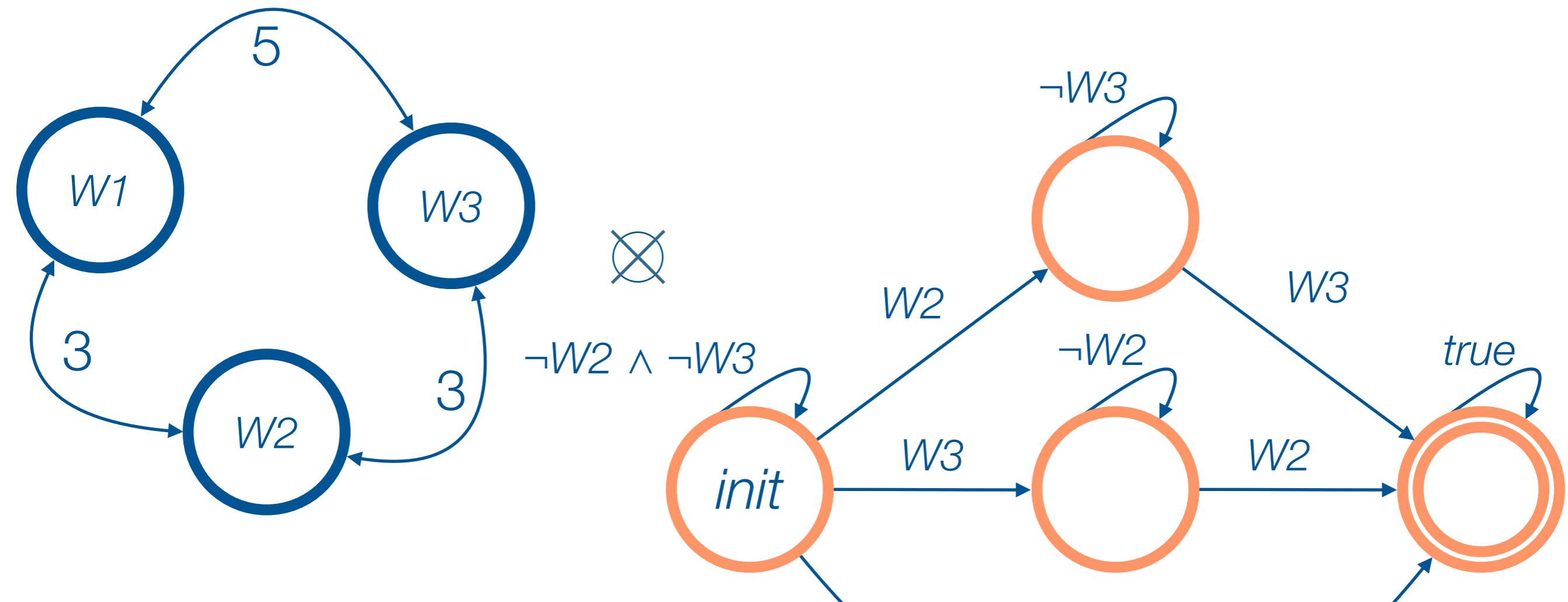
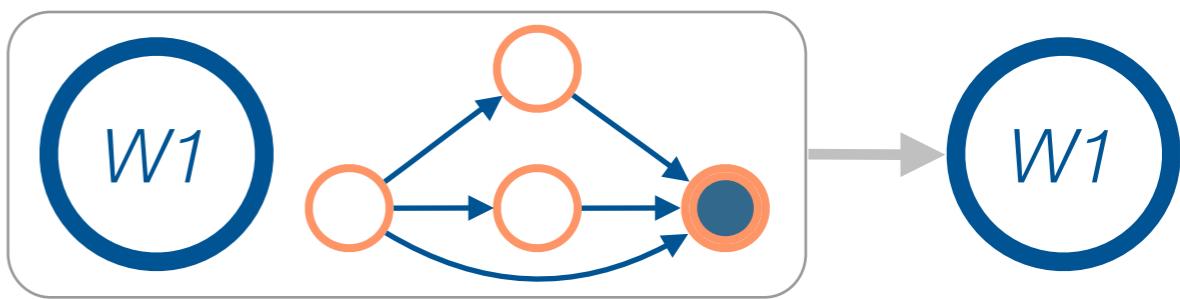
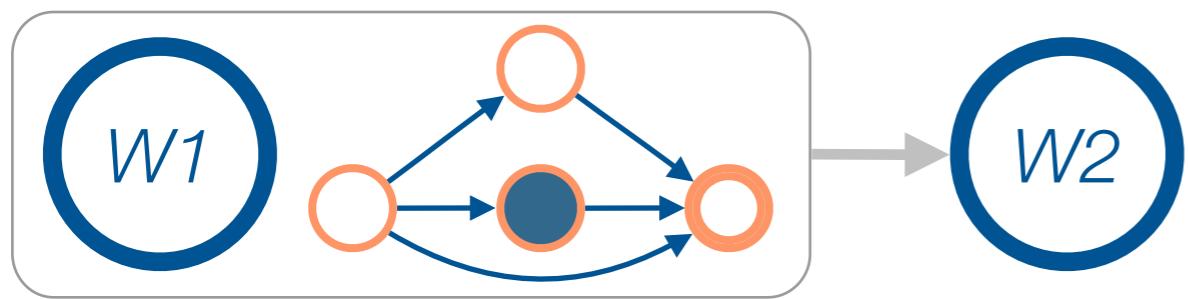
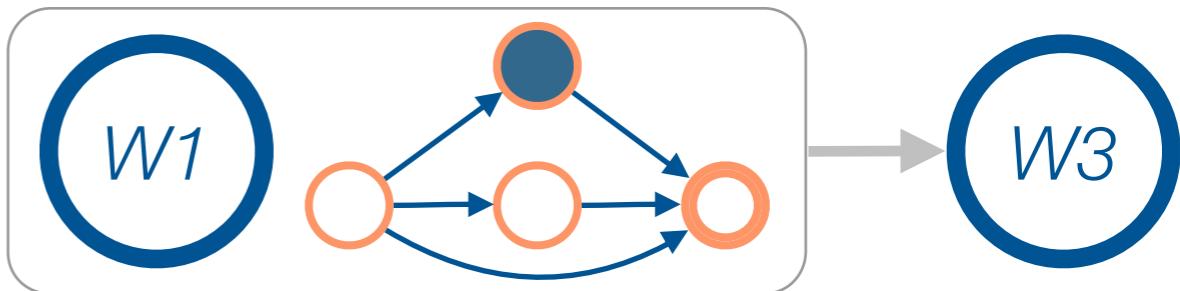
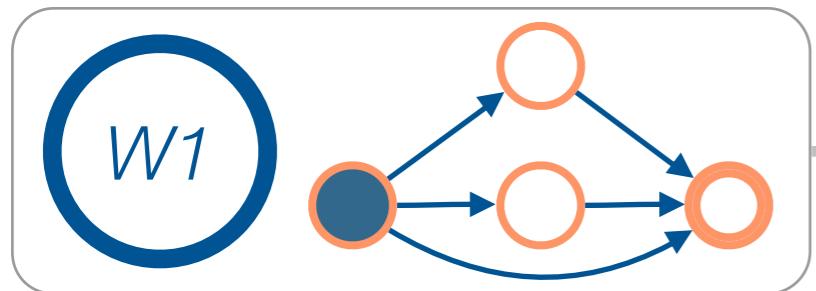


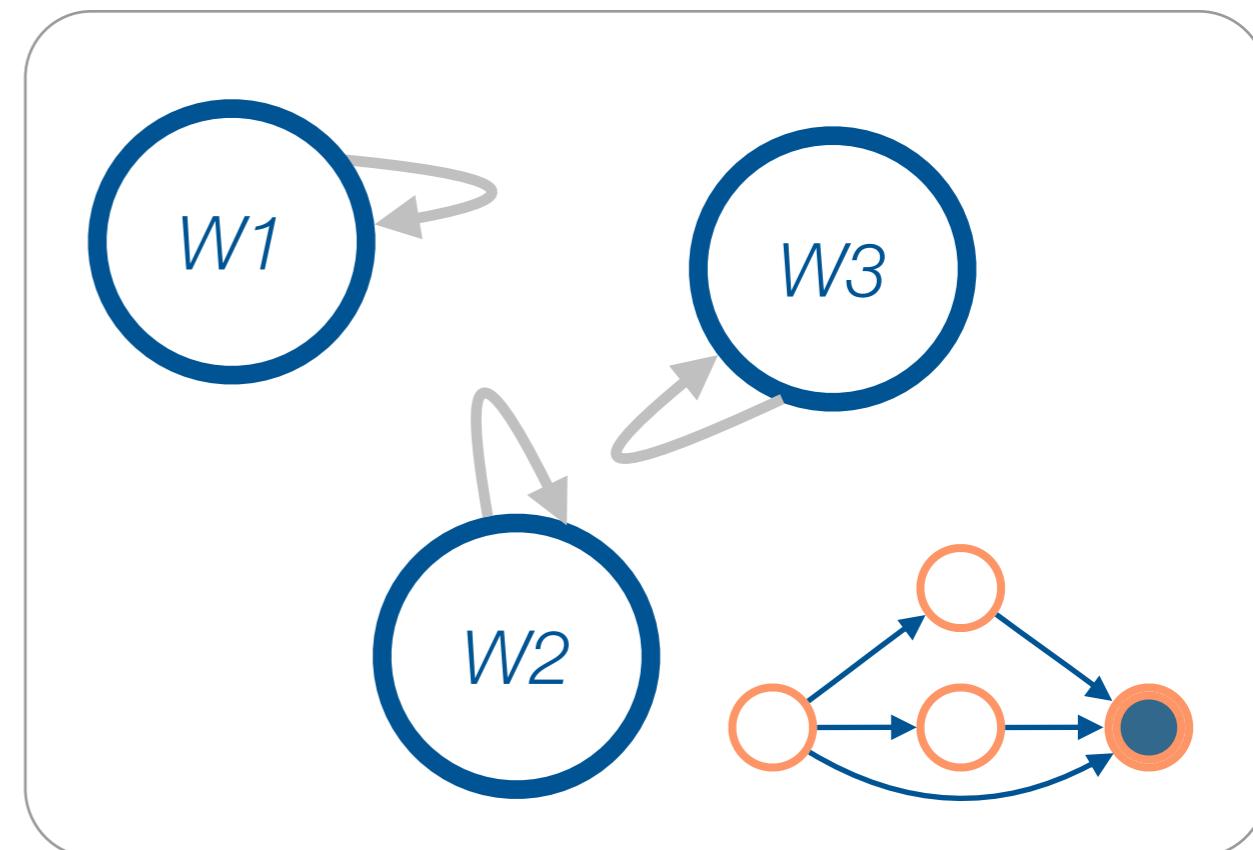
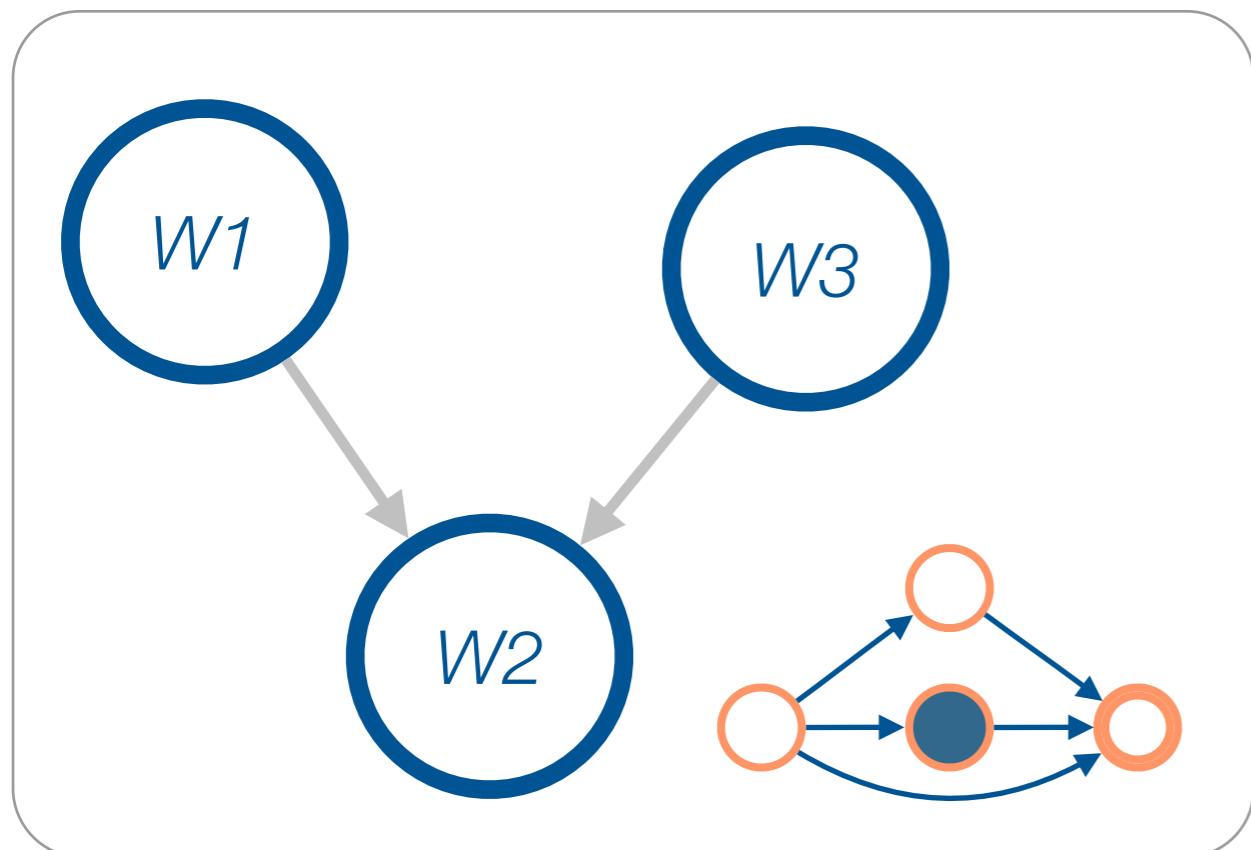
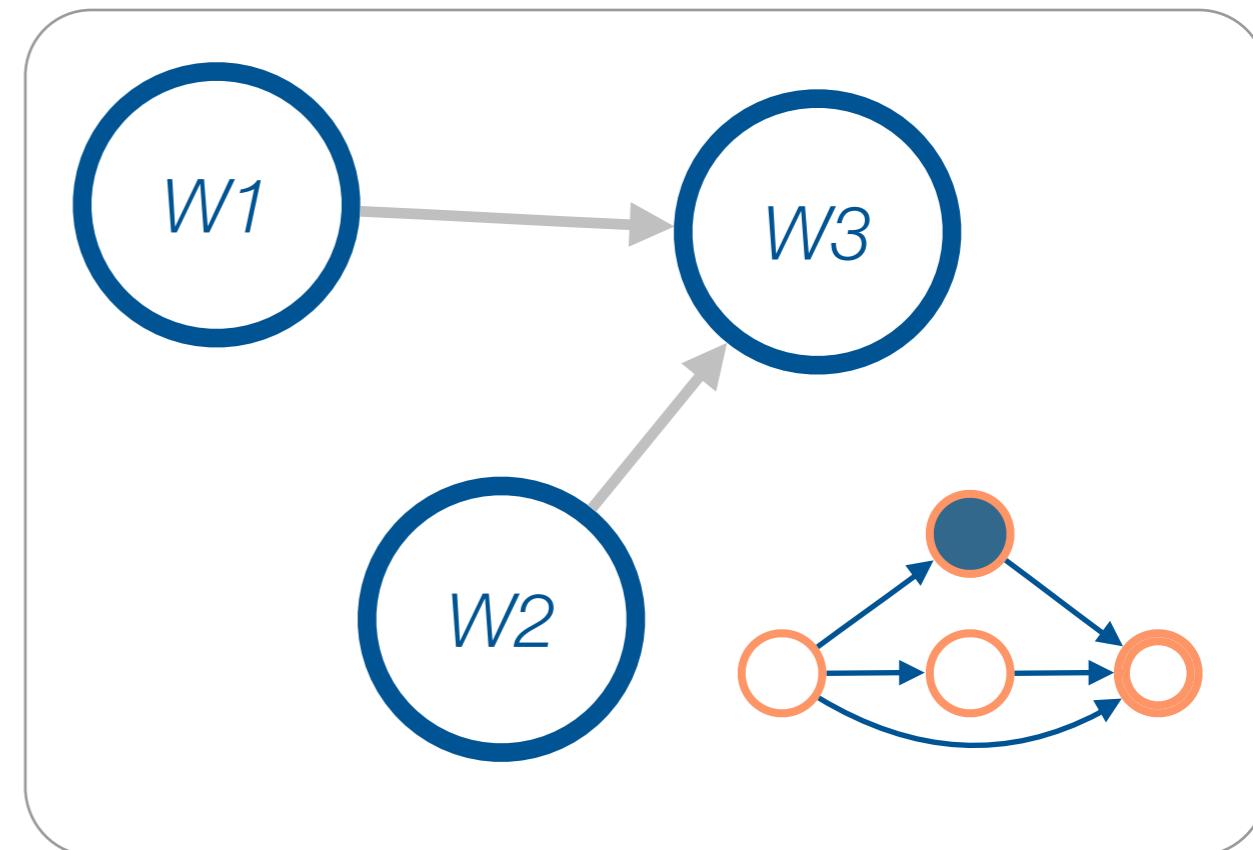
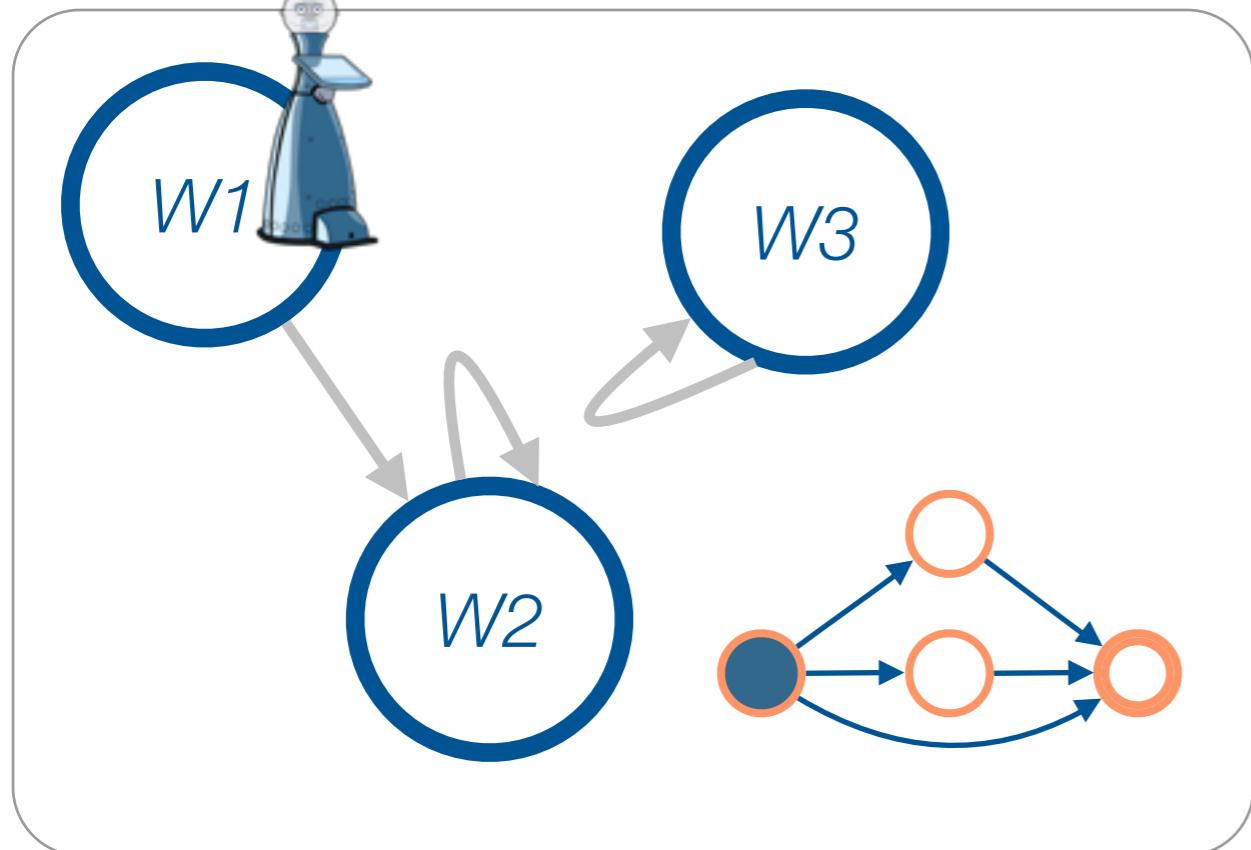
$(F W2) \wedge (F W3)$   
eventually reach  $W2$  and  $W3$

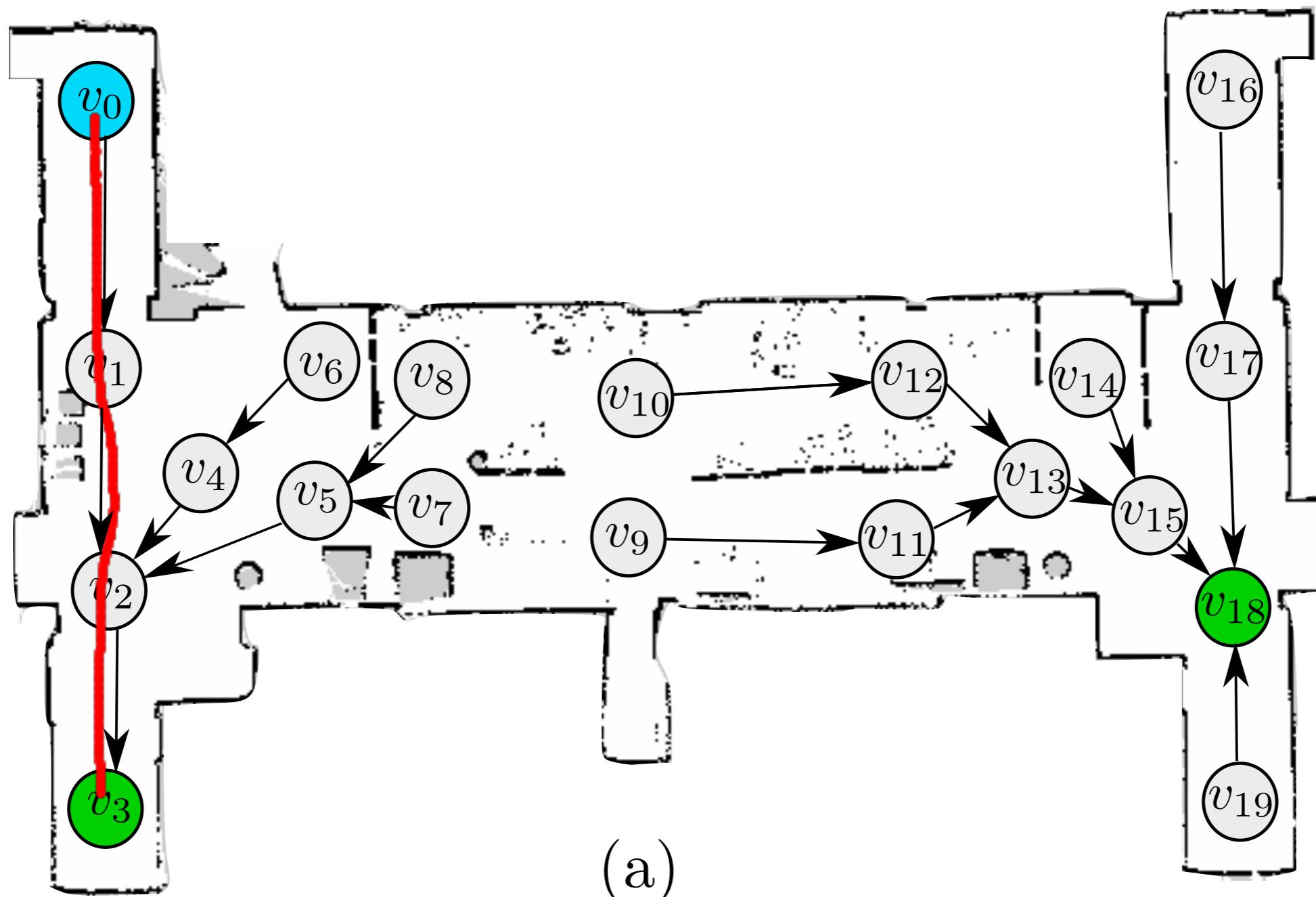
$$(F W2) \wedge (F W3)$$

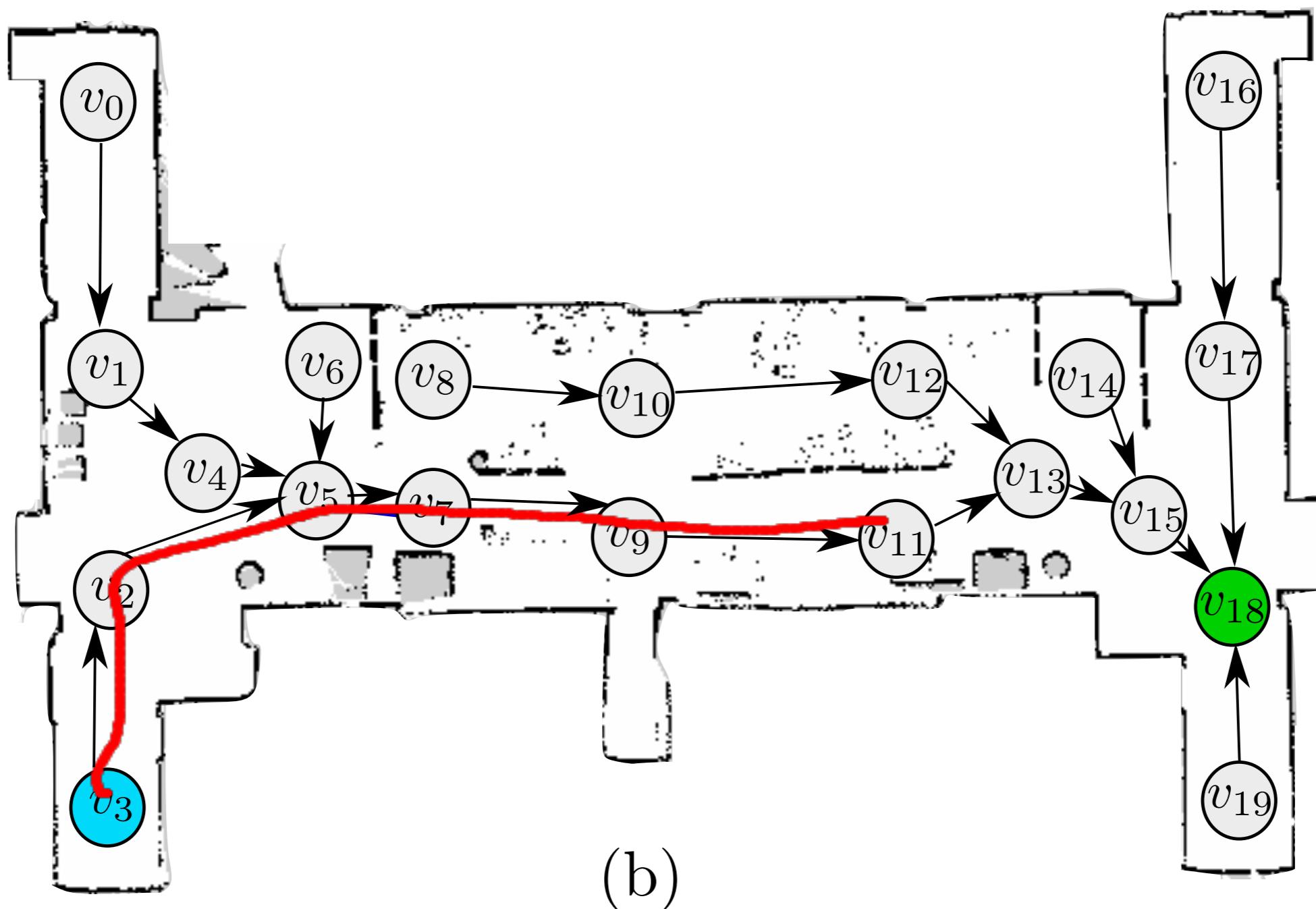
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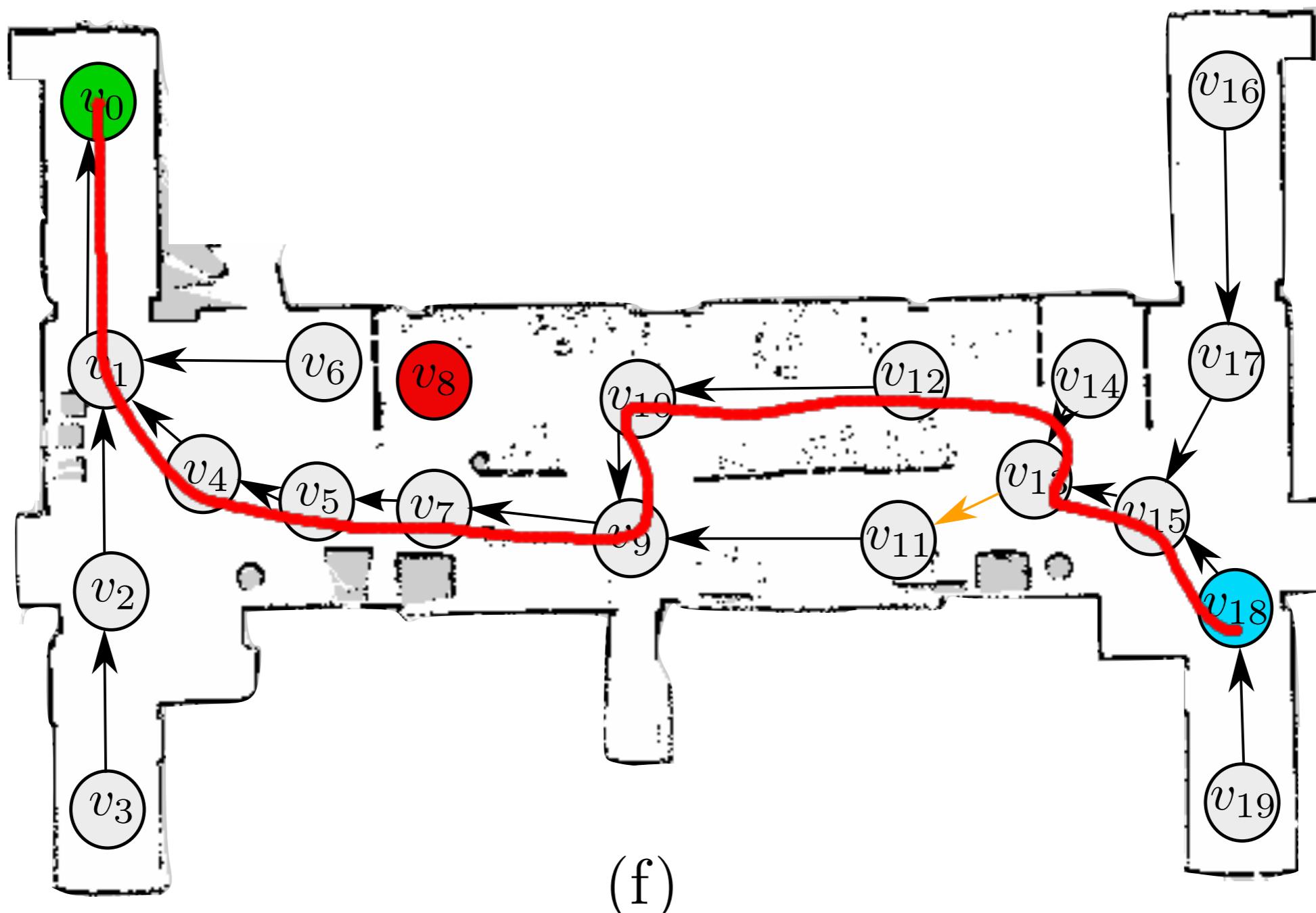


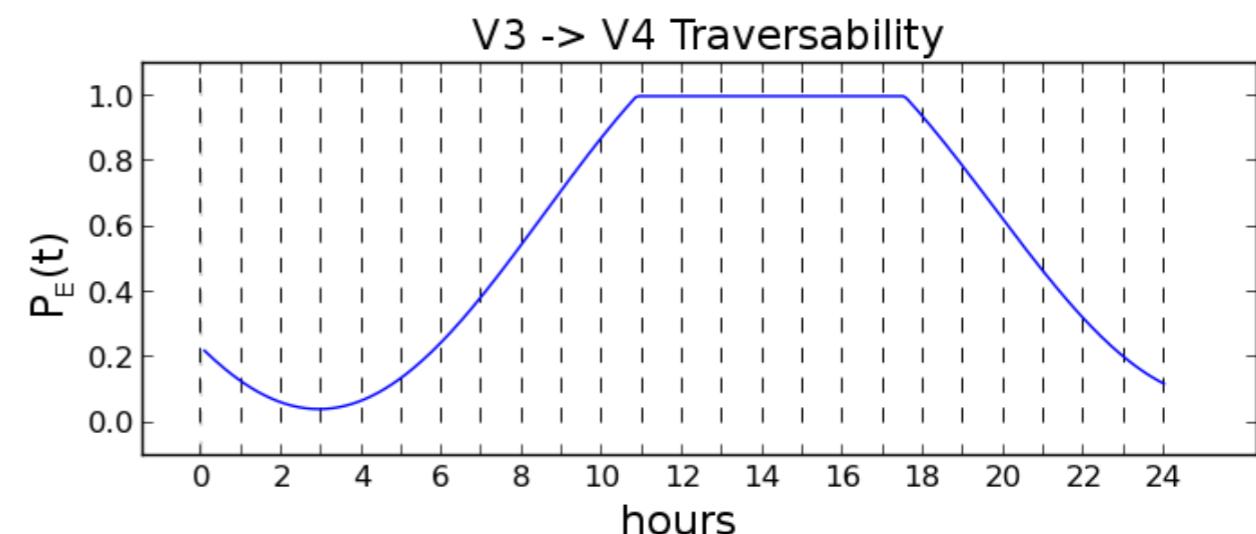
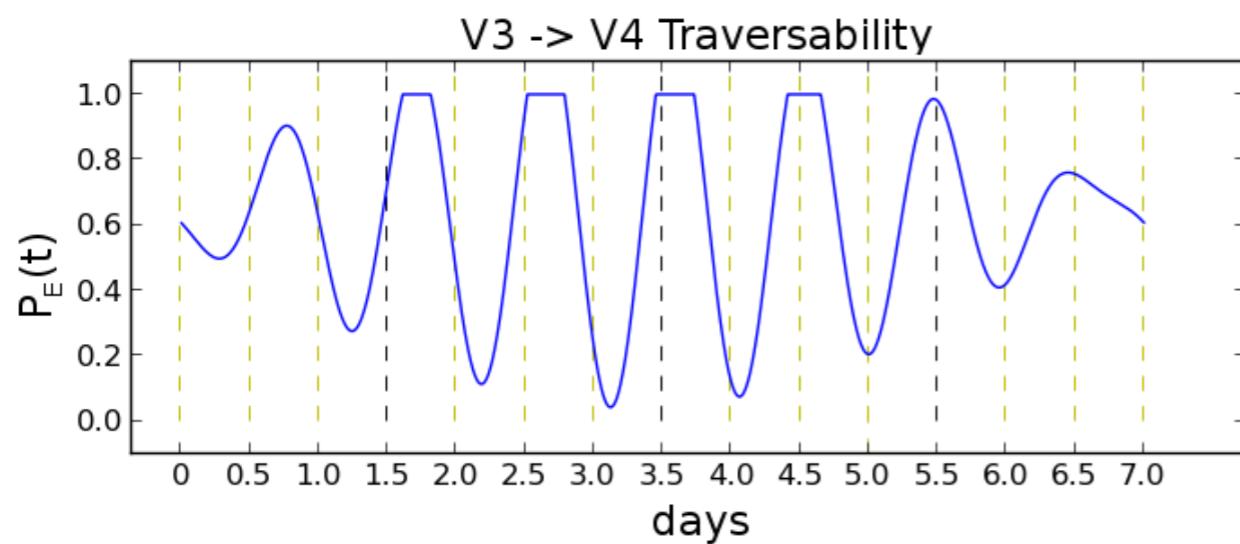
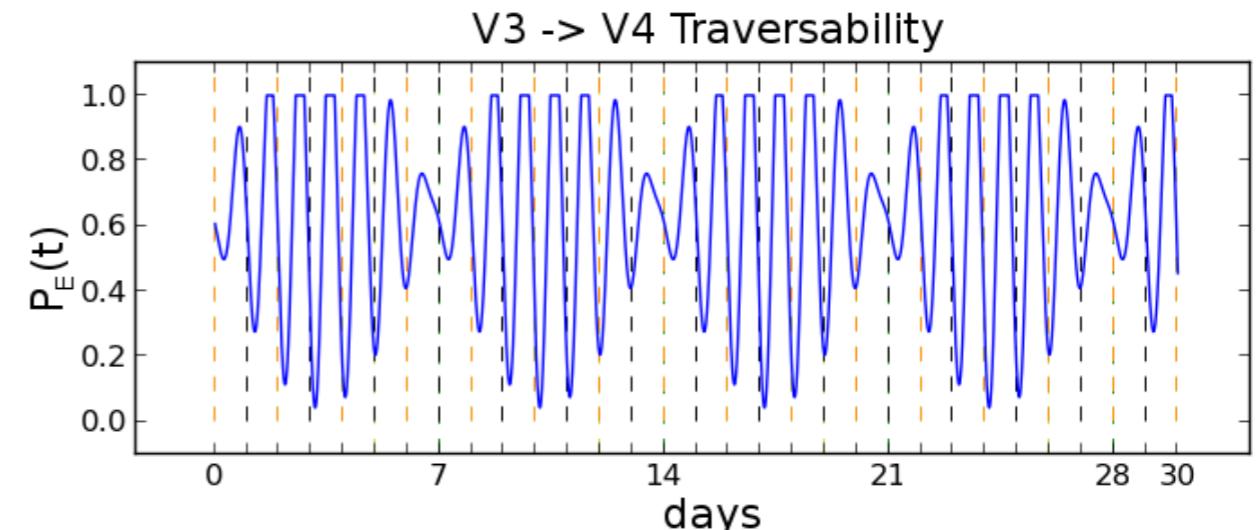
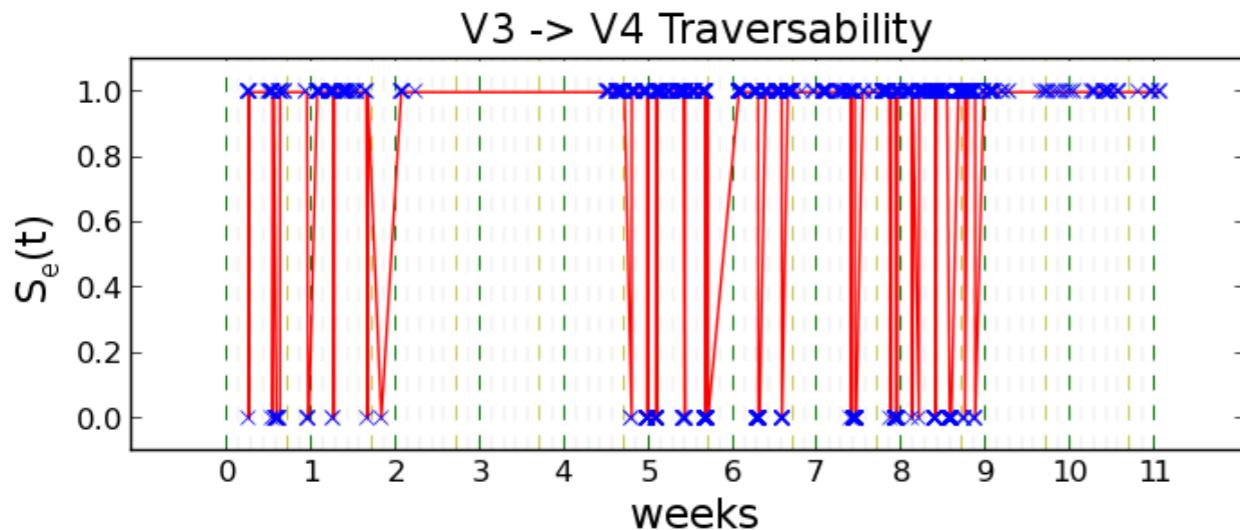






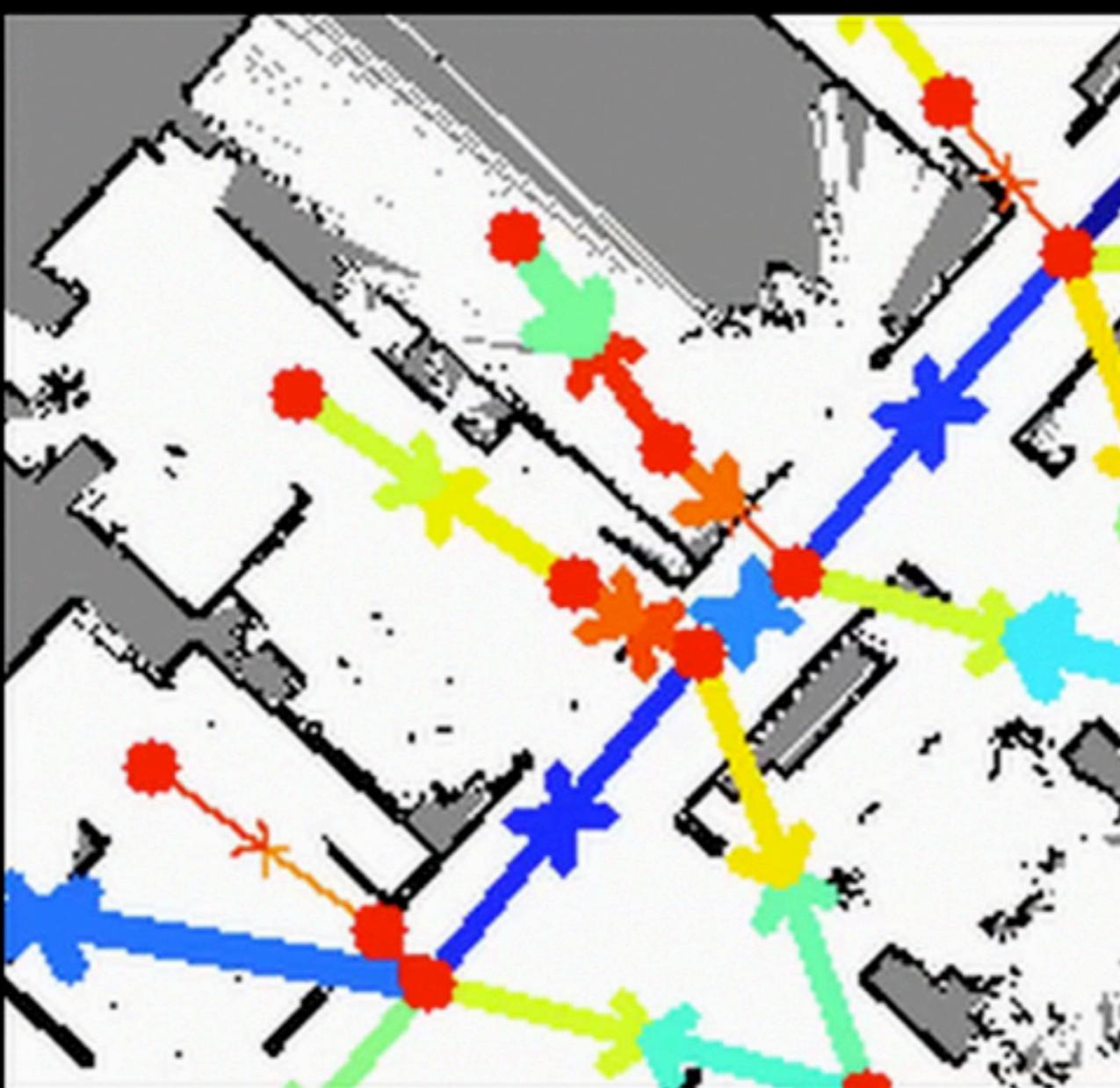






J. Pulido Fentanes, B. Lacerda, T. Krajiník, N. Hawes, and M. Hanheide.  
Now or later? predicting and maximising success of navigation actions  
from long-term experience. In ICRA, 2015.

Sun 07 Jun 2015 01:00:00 (BST)



Predicted Speed (m/s)

0

0.5



# Task framework

