



STRANDS

Task scheduling and execution
for long-term autonomy

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@hawesie



Long-Term Autonomy in Everyday Environments

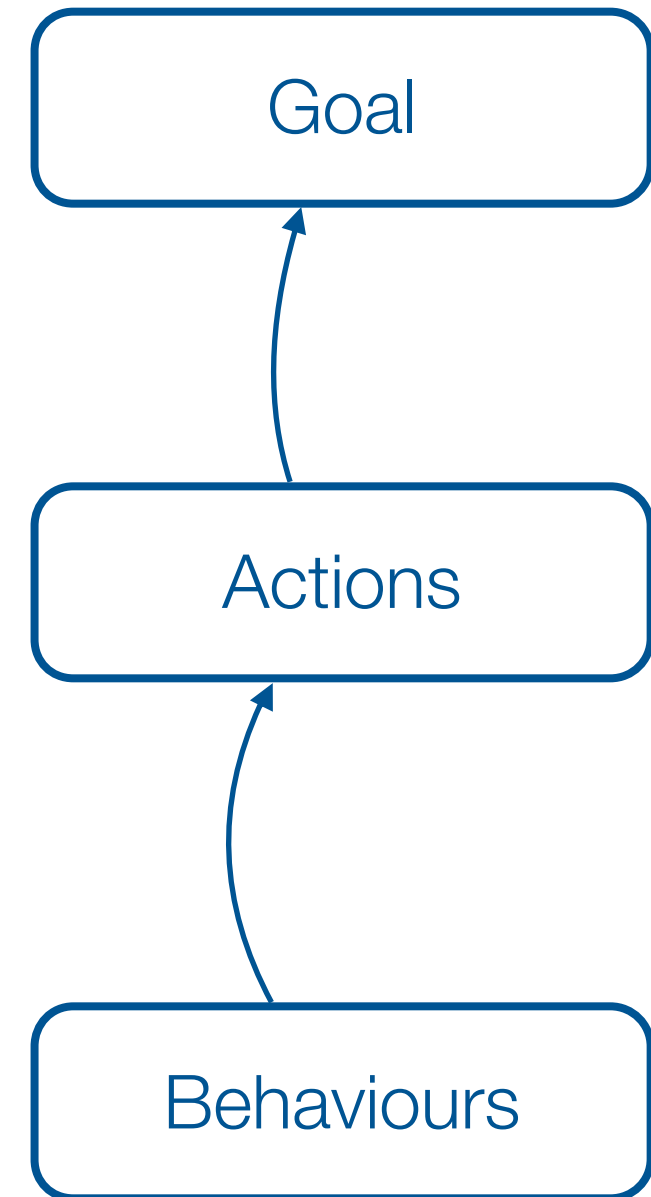


Goal-Oriented Behaviour

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 )
```

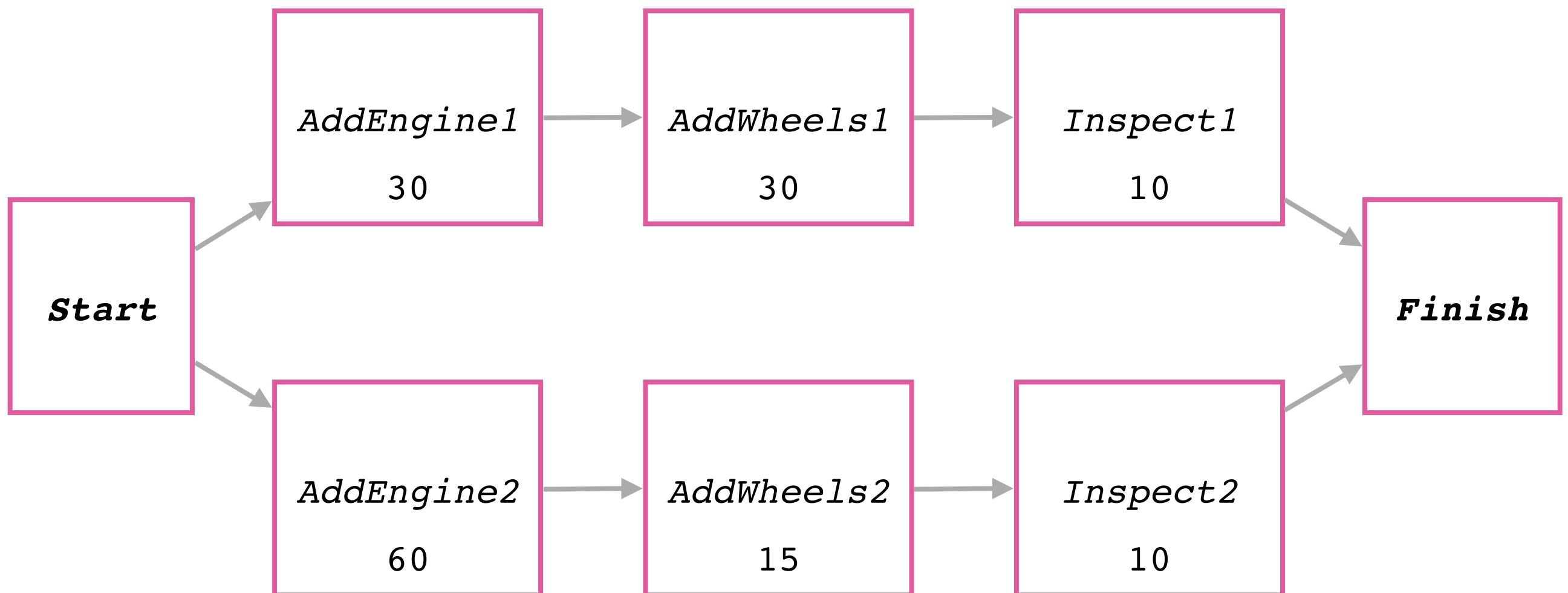
```
Action( AddWheels1, DURATION:30 )
```

```
Action( AddWheels1, DURATION:15 )
```

```
Action( Inspect, DURATION:10 )
```

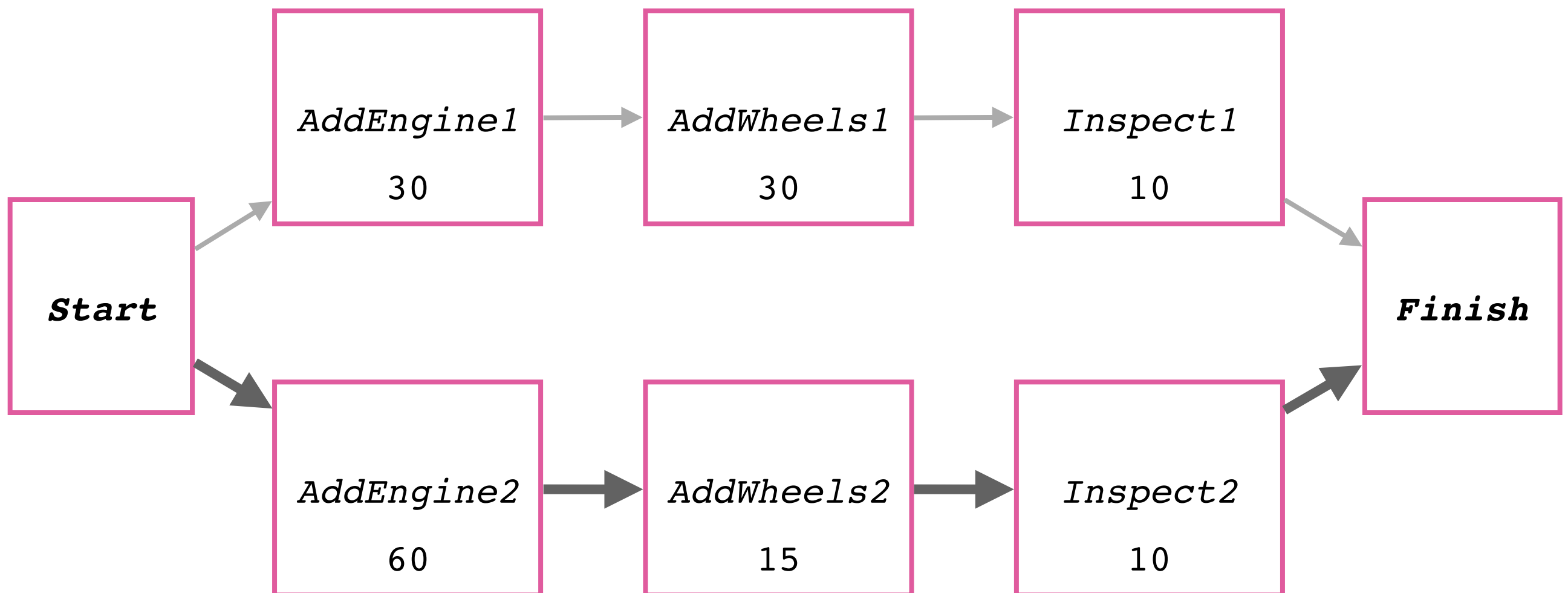
For each action assign
earliest start time *ES* and **latest start** time *LS*

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

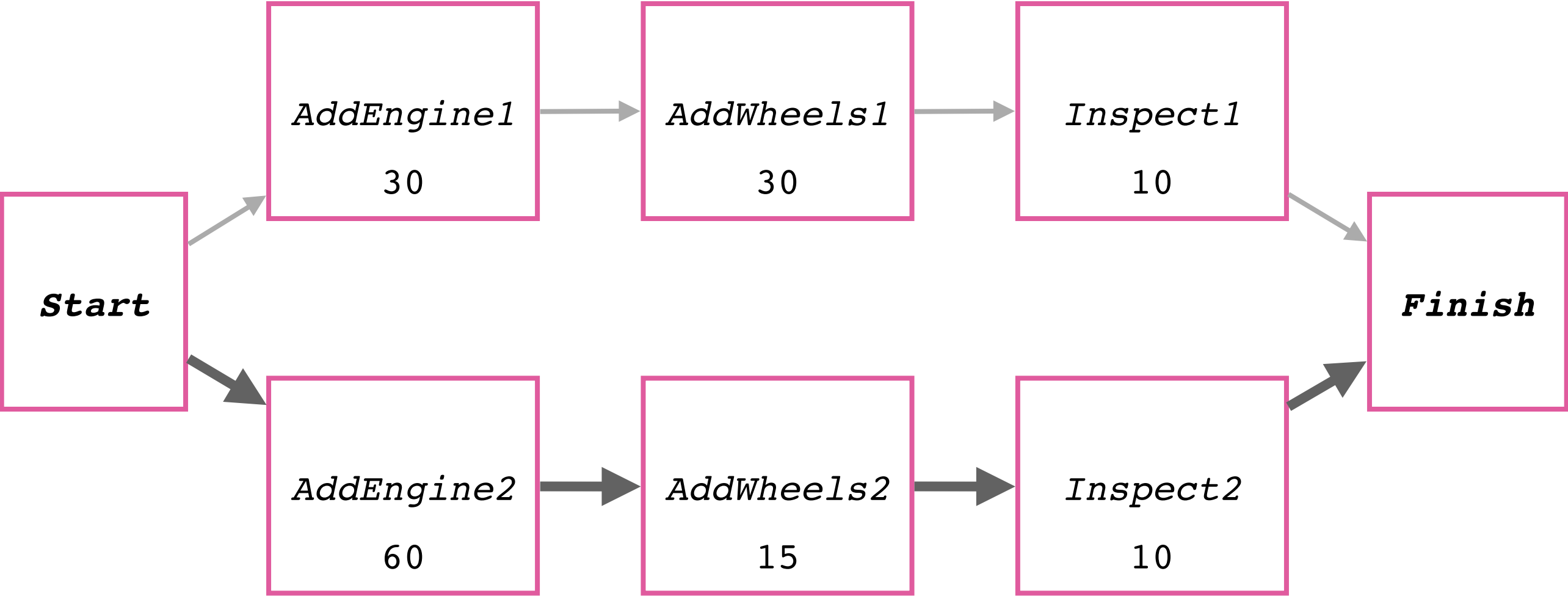


For each action assign
earliest start time *ES* and **latest start** time *LS*

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

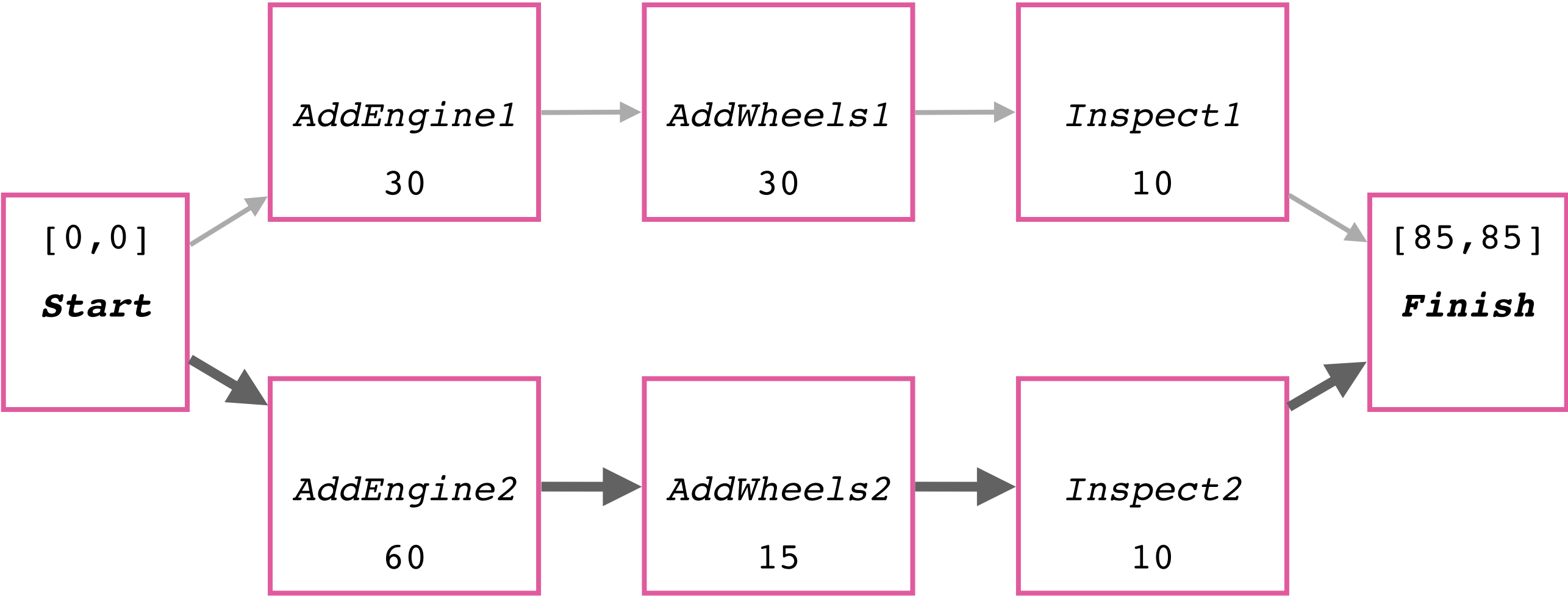


Use the **critical path** to define the overall length of the schedule,
i.e. [*ES*, *LS*] for *Start* and *Finish*



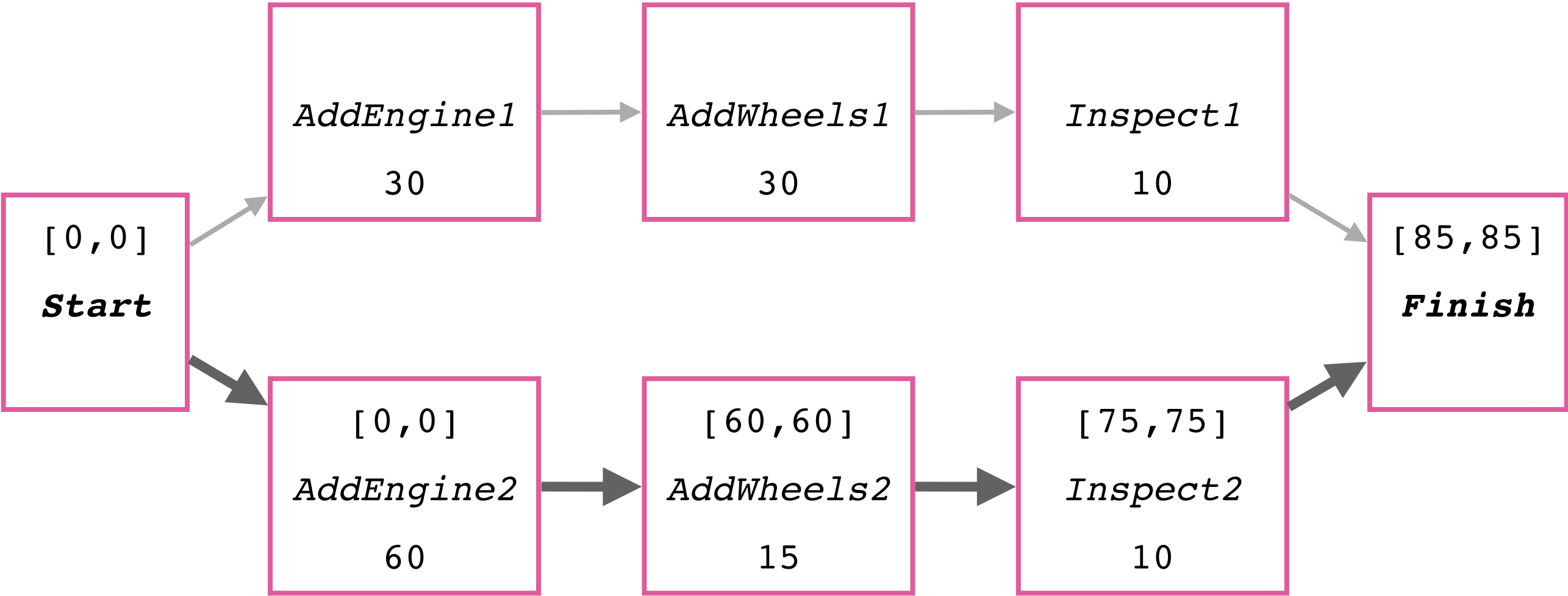
Use the **critical path** to define the overall length of the schedule,
i.e. $[ES, LS]$ for *Start* and *Finish*

and for the actions on the critical path



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

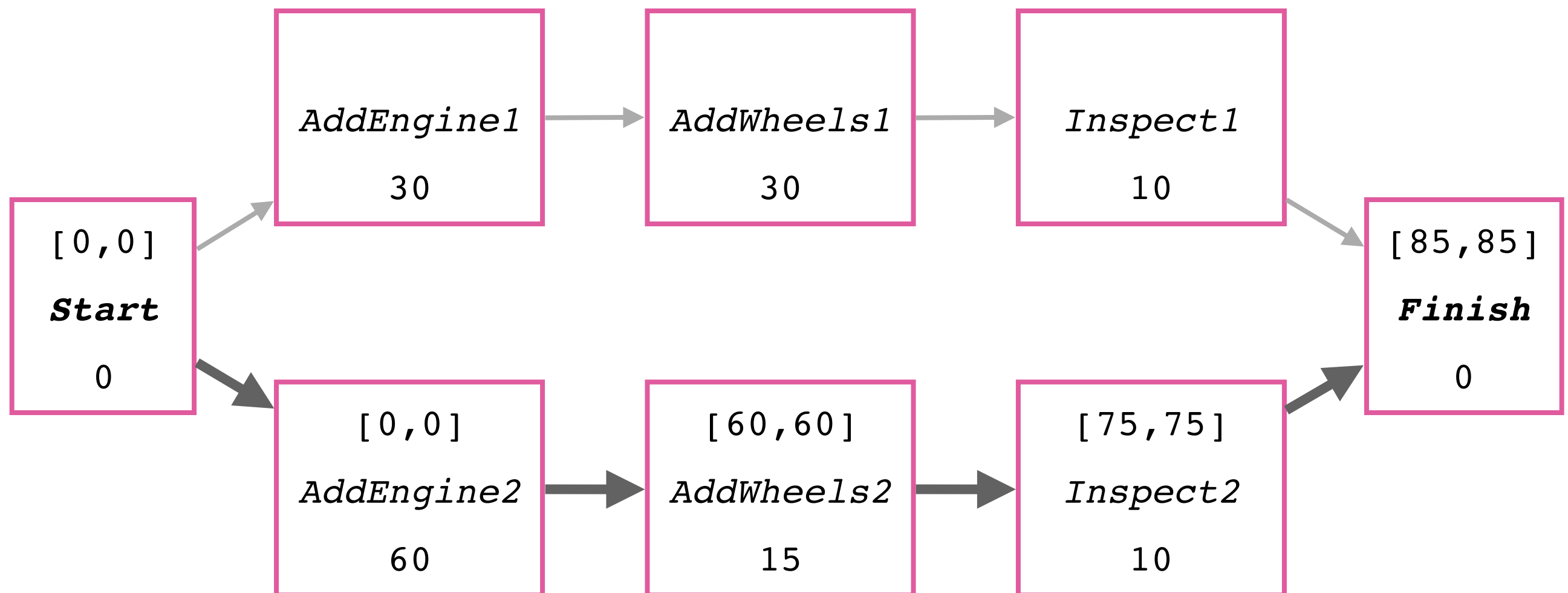
and for the actions on the critical path



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

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

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

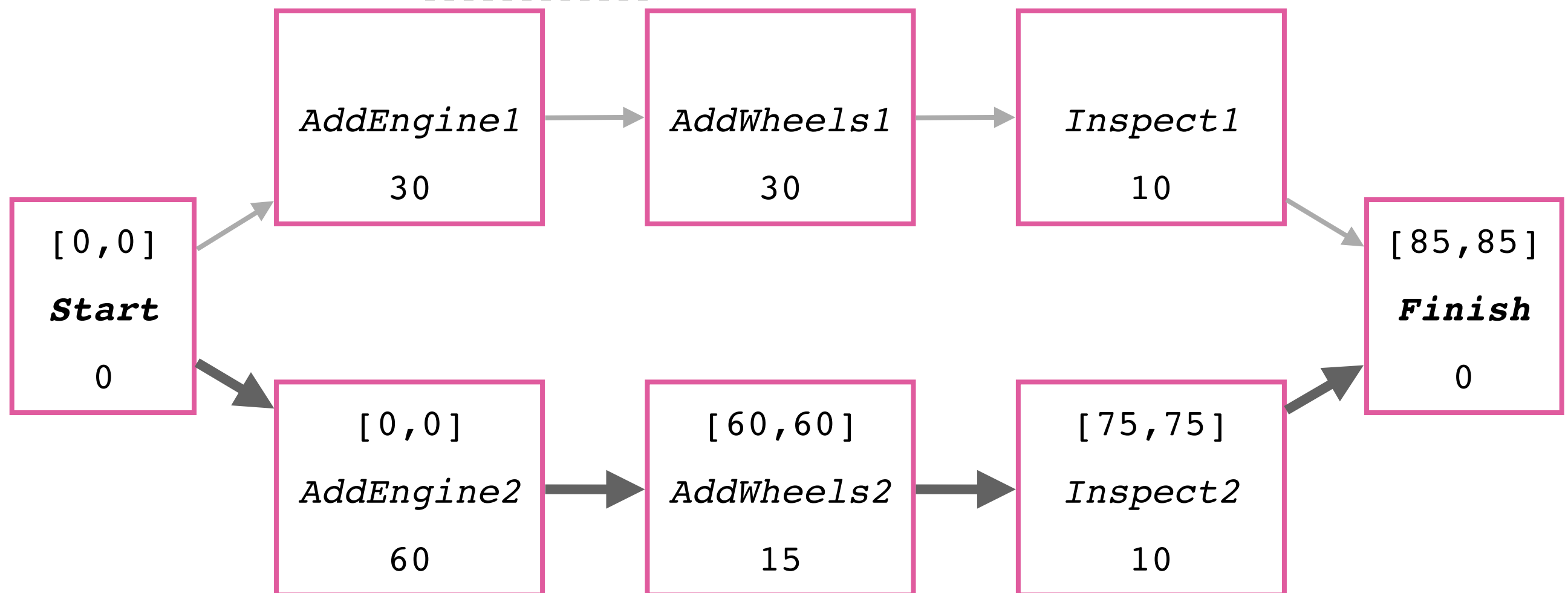


Action with *latest* earliest start preceding *B*

Action with *earliest* latest start preceding *A*

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

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

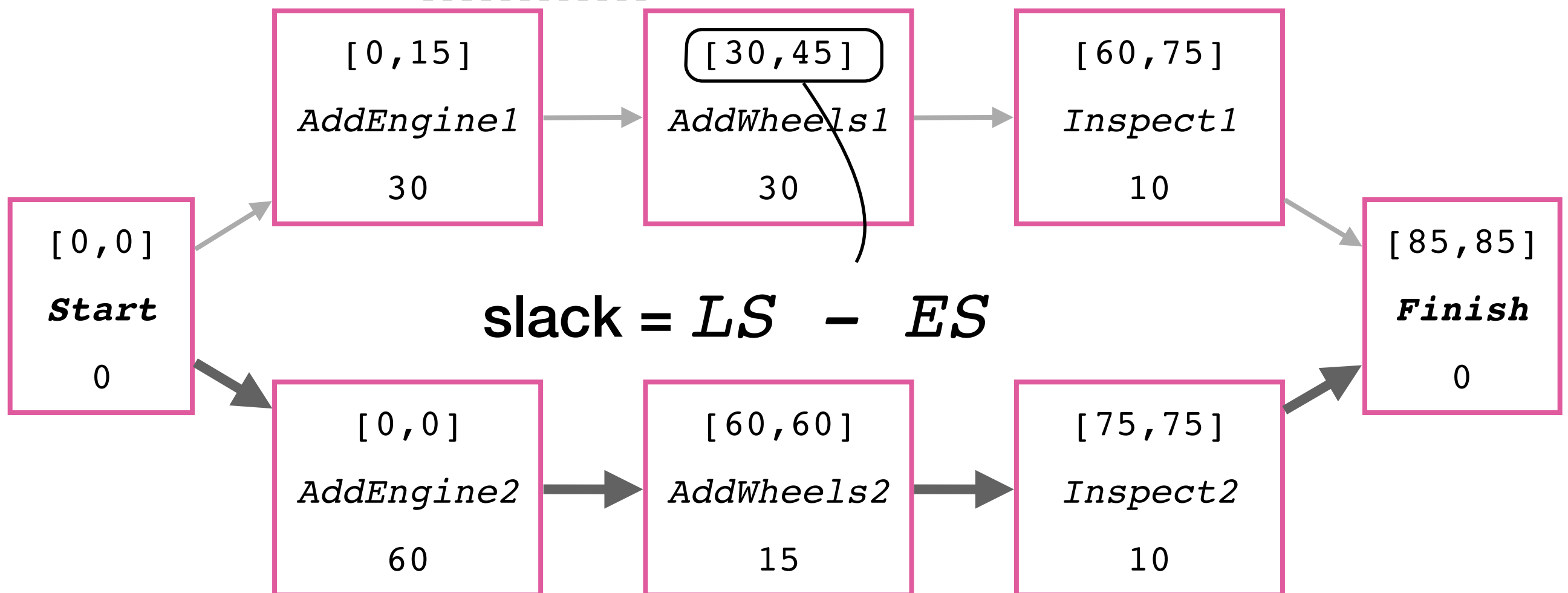


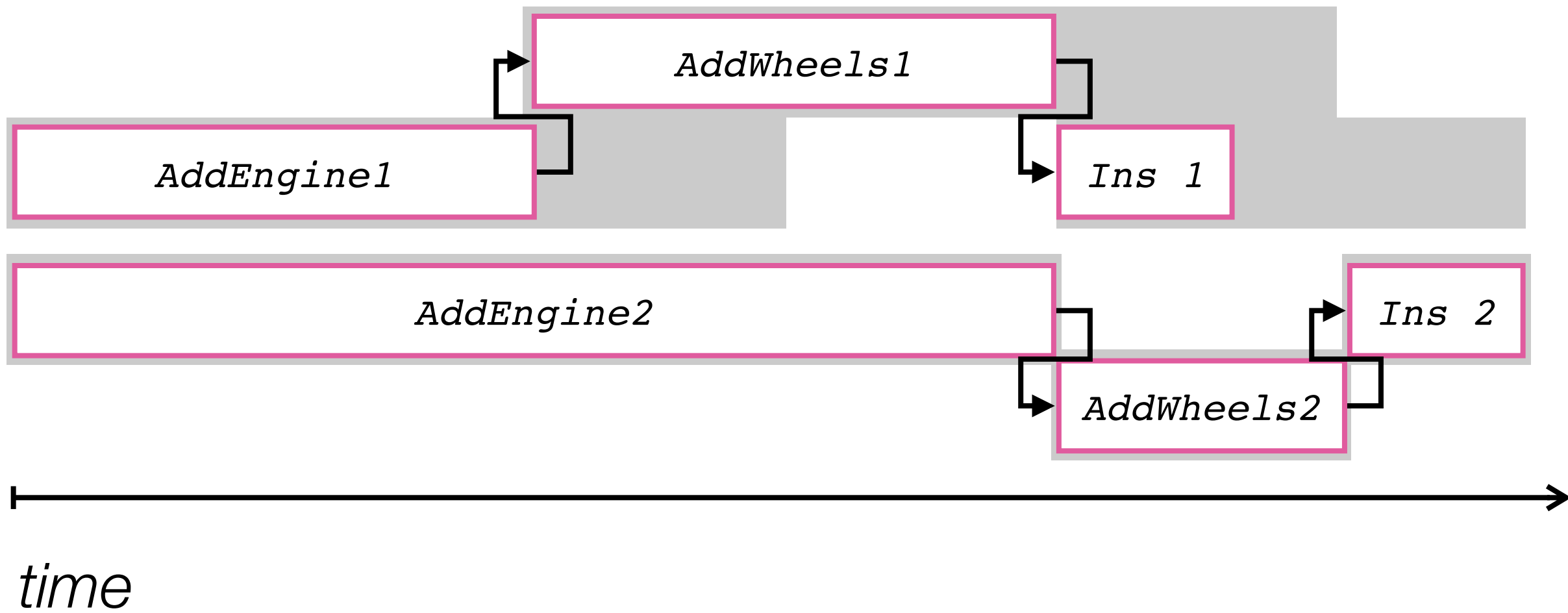
Action with *latest* earliest start preceding *B*

Action with *earliest* latest start preceding *A*

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

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





$$ES(Start) = 0$$

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

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

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

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)



Task Scheduling for Mobile Robots Using Interval Algebra

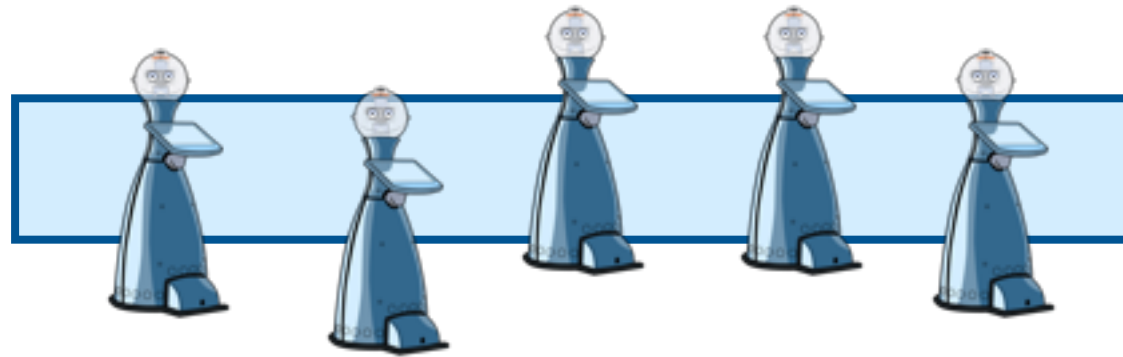
Mudrová and Hawes. In, *ICRA '15*.

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



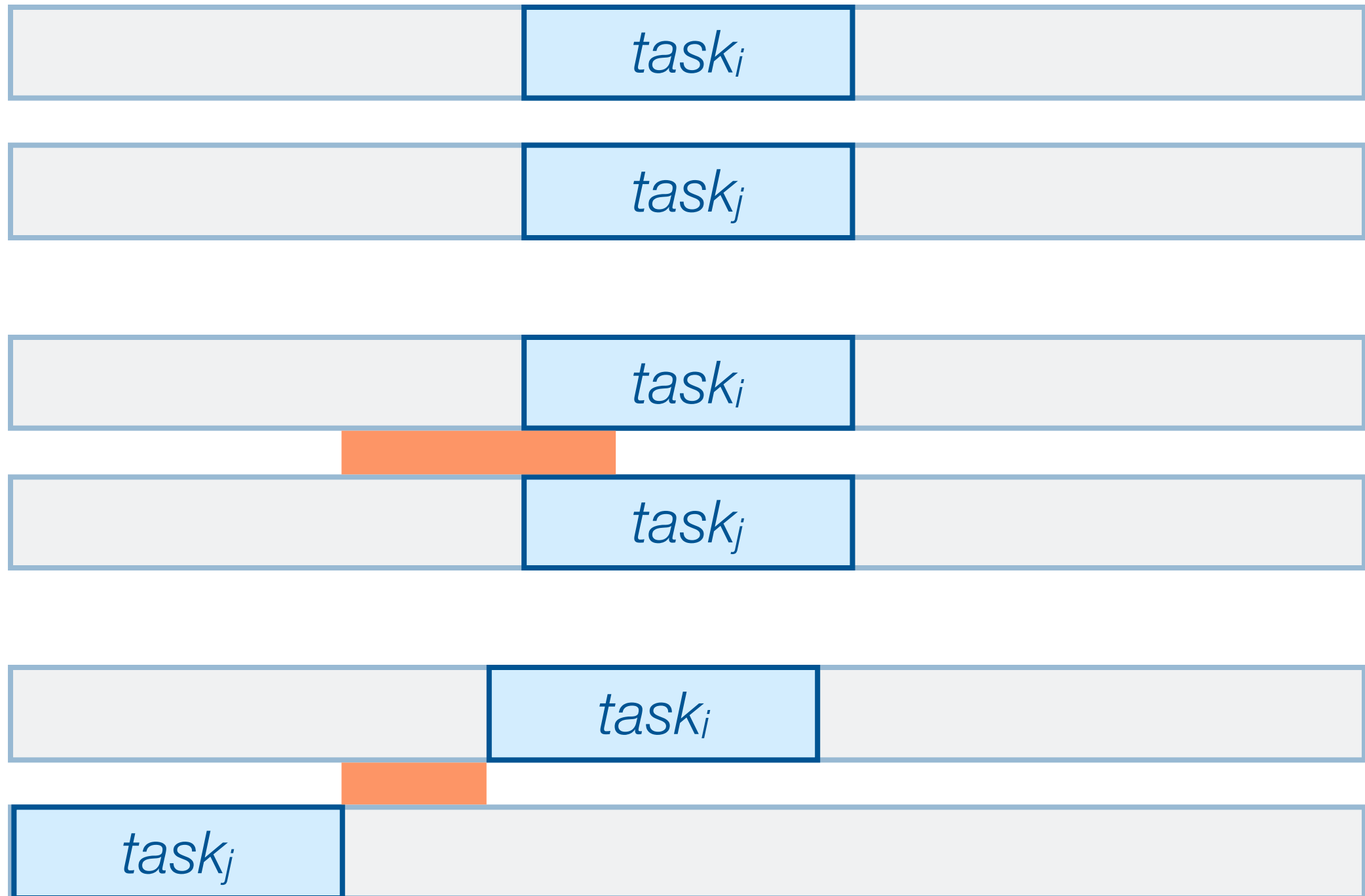
$task_i$

d_i

S_i

e_i



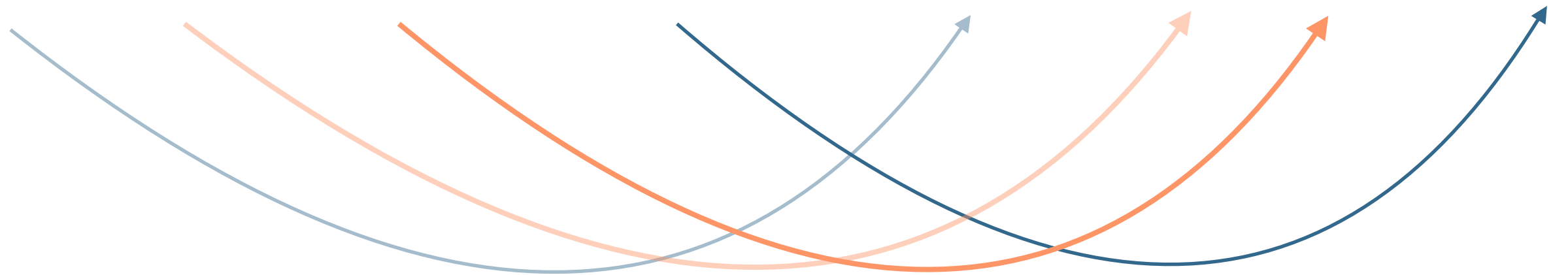


$$\forall i : \min \sum (t_i - s_i)$$

Coltin et al.*

Scheduling using mixed-integer programming

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

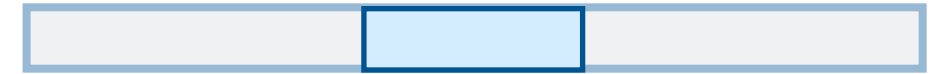


* e.g. Brian Coltin, Manuela Veloso, and Rodrigo Ventura. *Dynamic User Task Scheduling for Mobile Robots*. In Proceedings of the AAAI Workshop on Automated Action Planning for Autonomous Mobile Robots at AAAI. 2011.

Coltin et al.*

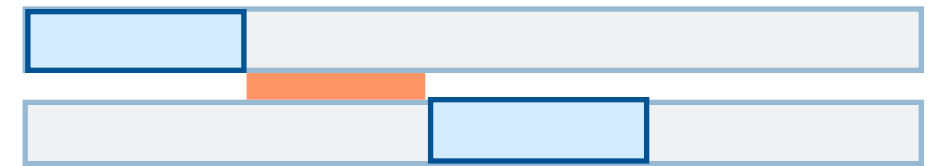
Scheduling using mixed-integer programming

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



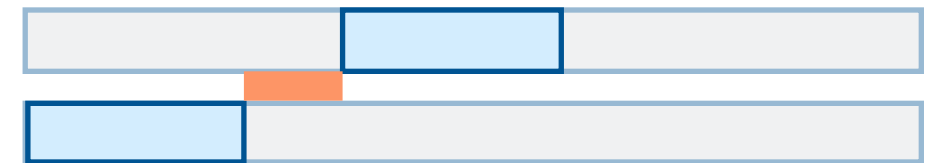
$$\forall i, j : t_i + d_i + \text{time}(p^{e_i}, p^{s_j}) \leq t_j$$

or

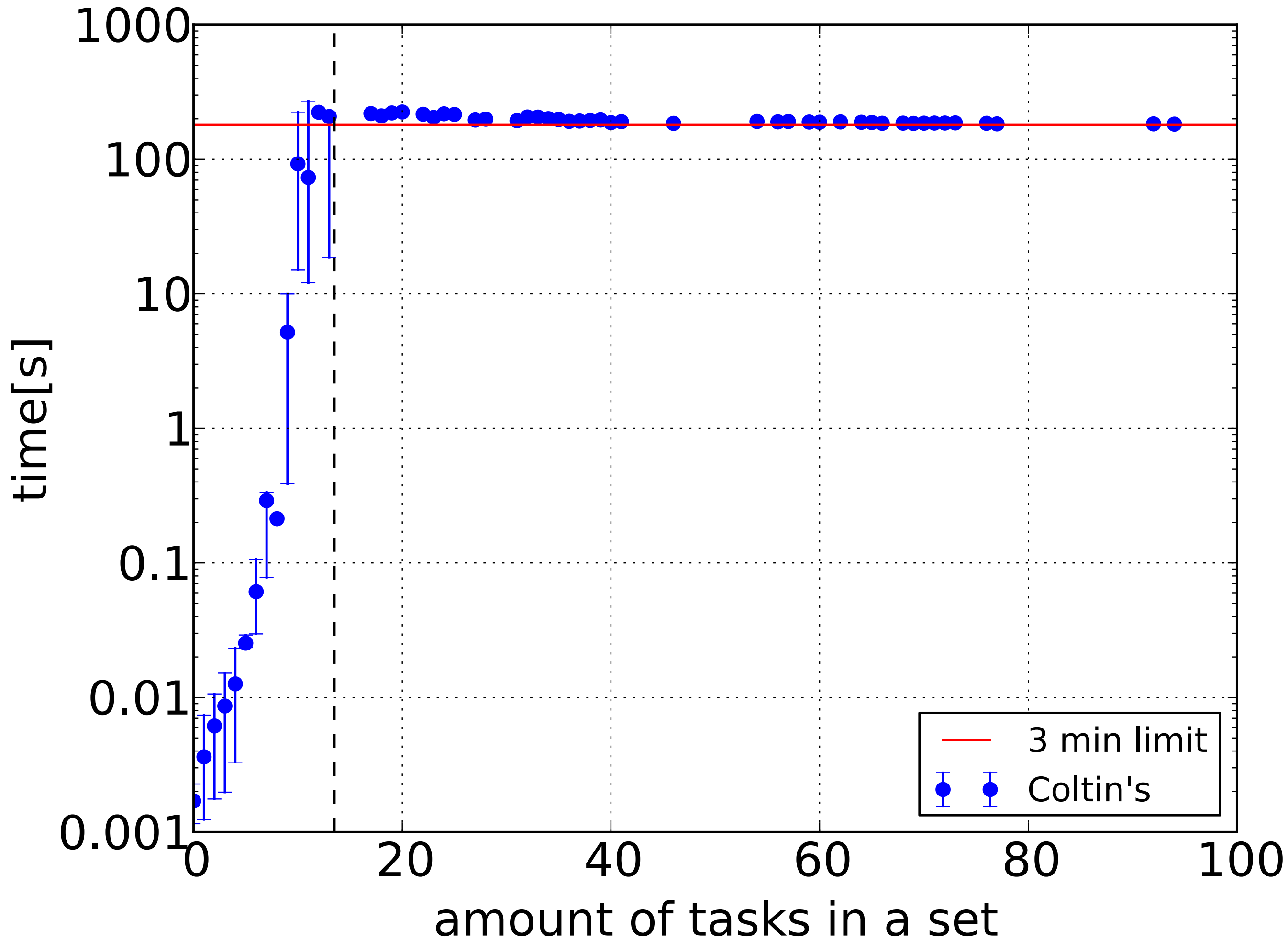


$$\forall i, j : t_j + d_j + \text{time}(p^{e_j}, p^{s_i}) \leq t_i$$

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



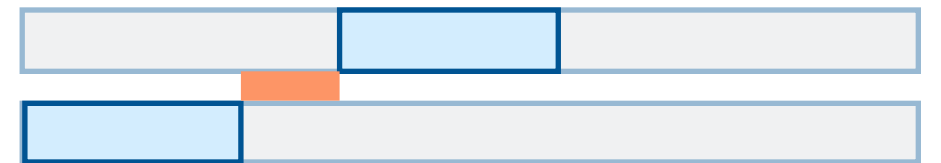
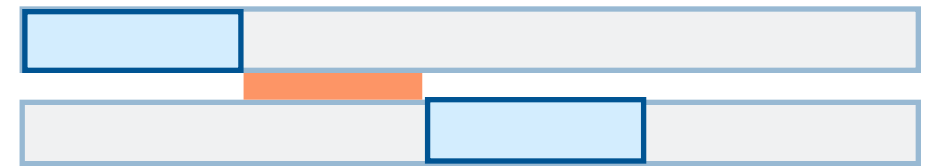
* e.g. Brian Coltin, Manuela Veloso, and Rodrigo Ventura. *Dynamic User Task Scheduling for Mobile Robots*. In Proceedings of the AAAI Workshop on Automated Action Planning for Autonomous Mobile Robots at AAAI. 2011.



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

or

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



Reasoning about **tasks'** time windows: $S_i \in_i$ vs $S_j \in_j$

no constraint

i precedes **j**

j precedes **i**

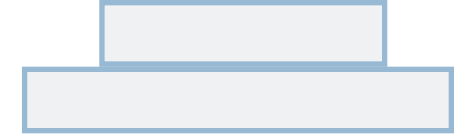
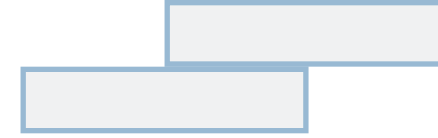
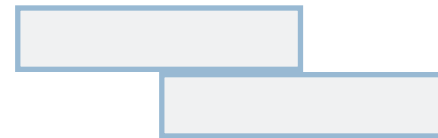
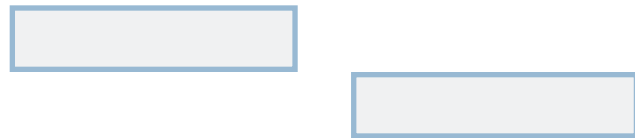
both

i before j

i overlaps j

j overlaps i

i during j

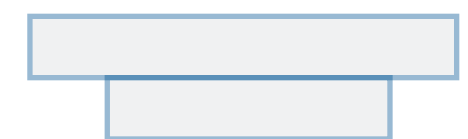
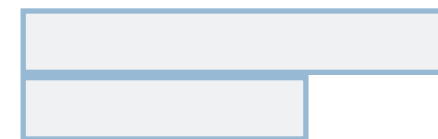
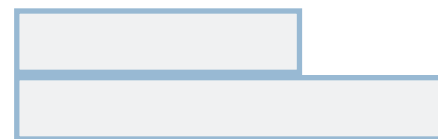
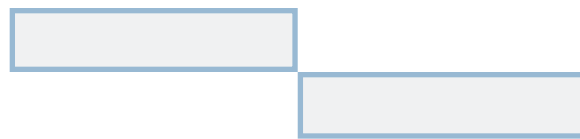


i meets j

i starts j

j starts i

j during i

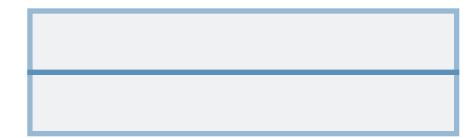


j before i

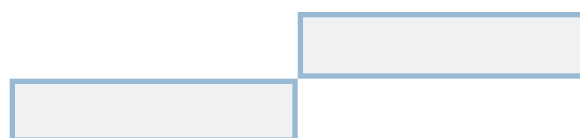
j finishes i

i finishes j

i equals j

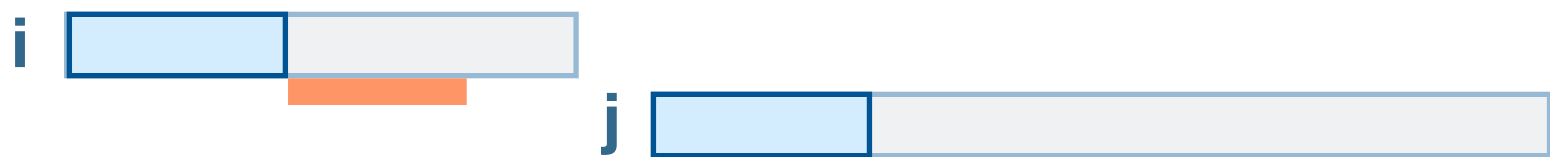


j meets i



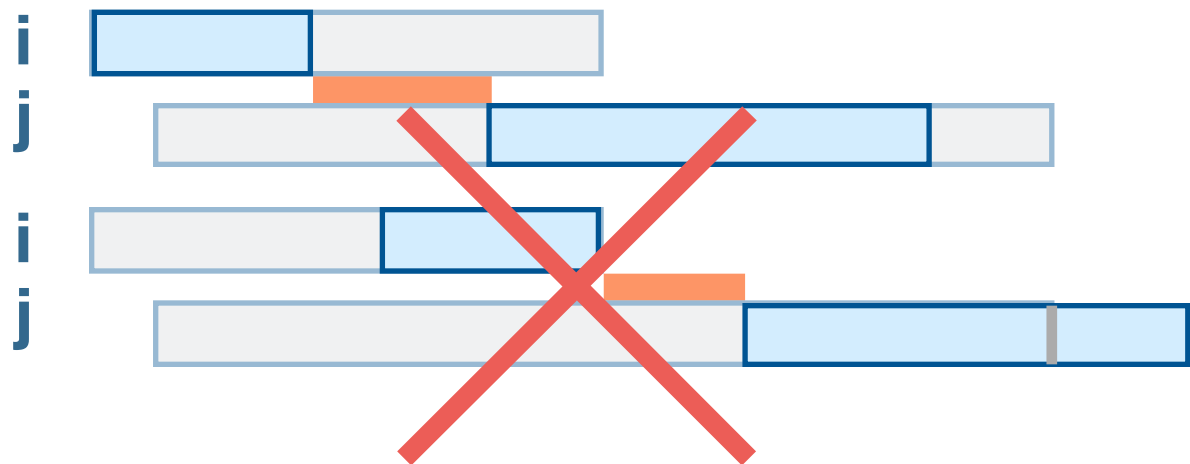
Allen's Interval Algebra

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



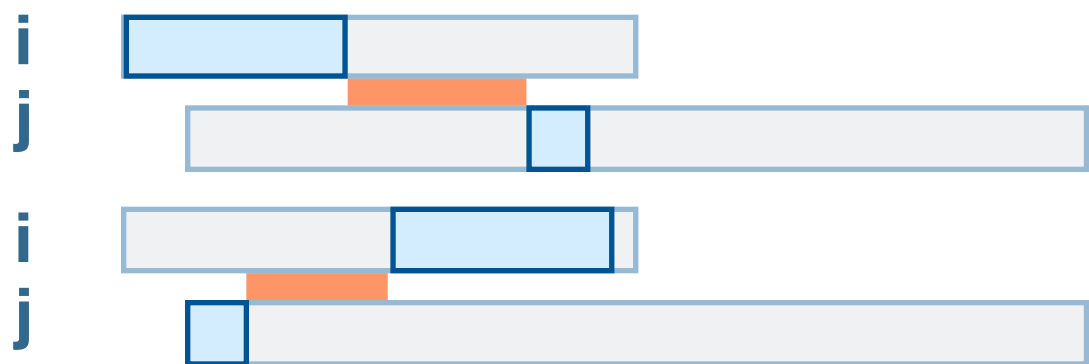
i before j

no order constraint



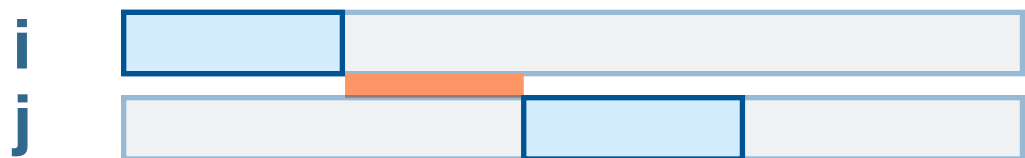
i overlaps j

choose only possible order constraint



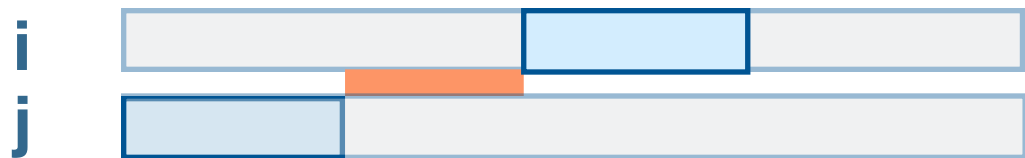
i overlaps j

choose order to satisfy $\forall i : \min \sum (t_i - s_i)$

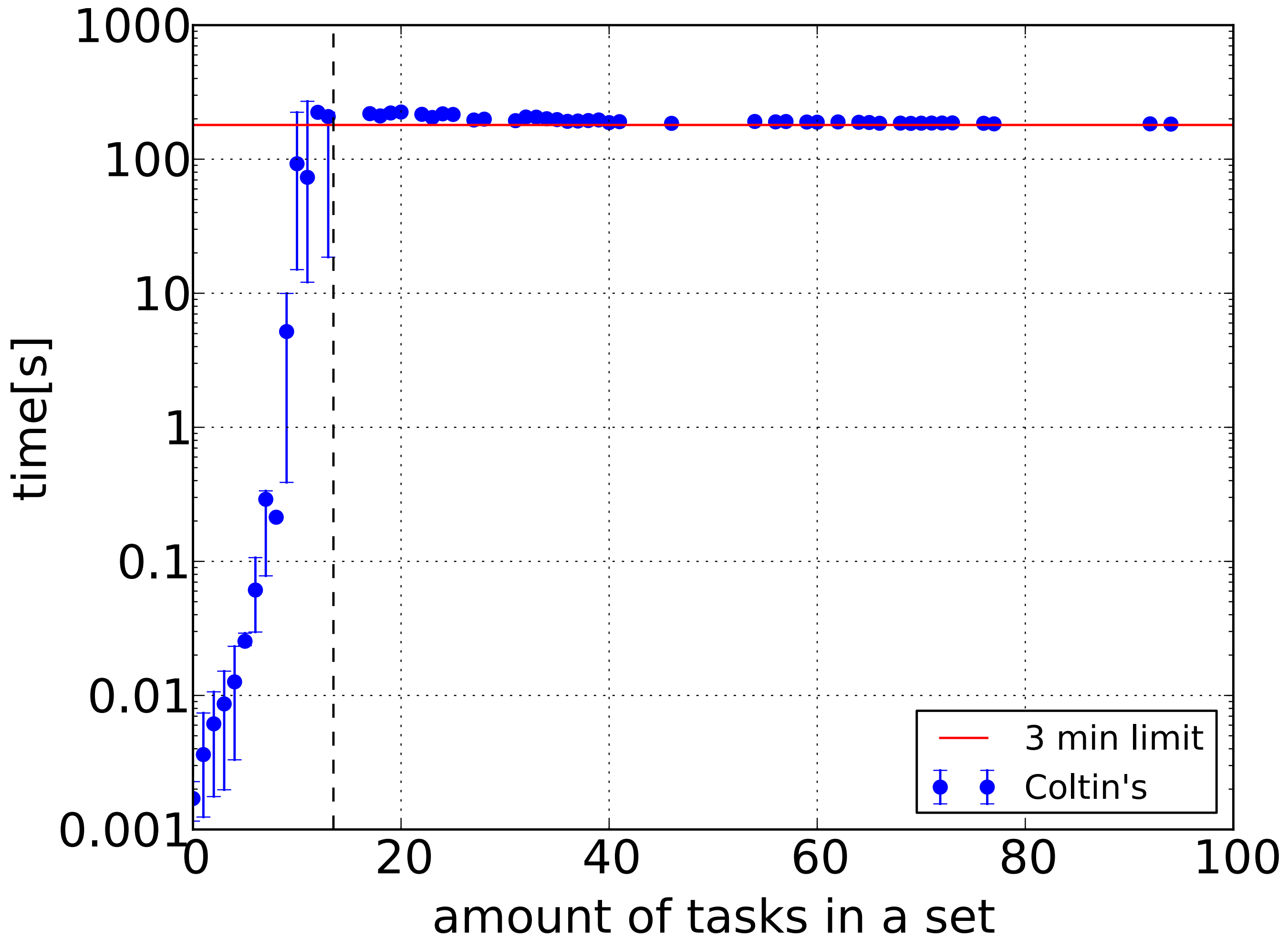


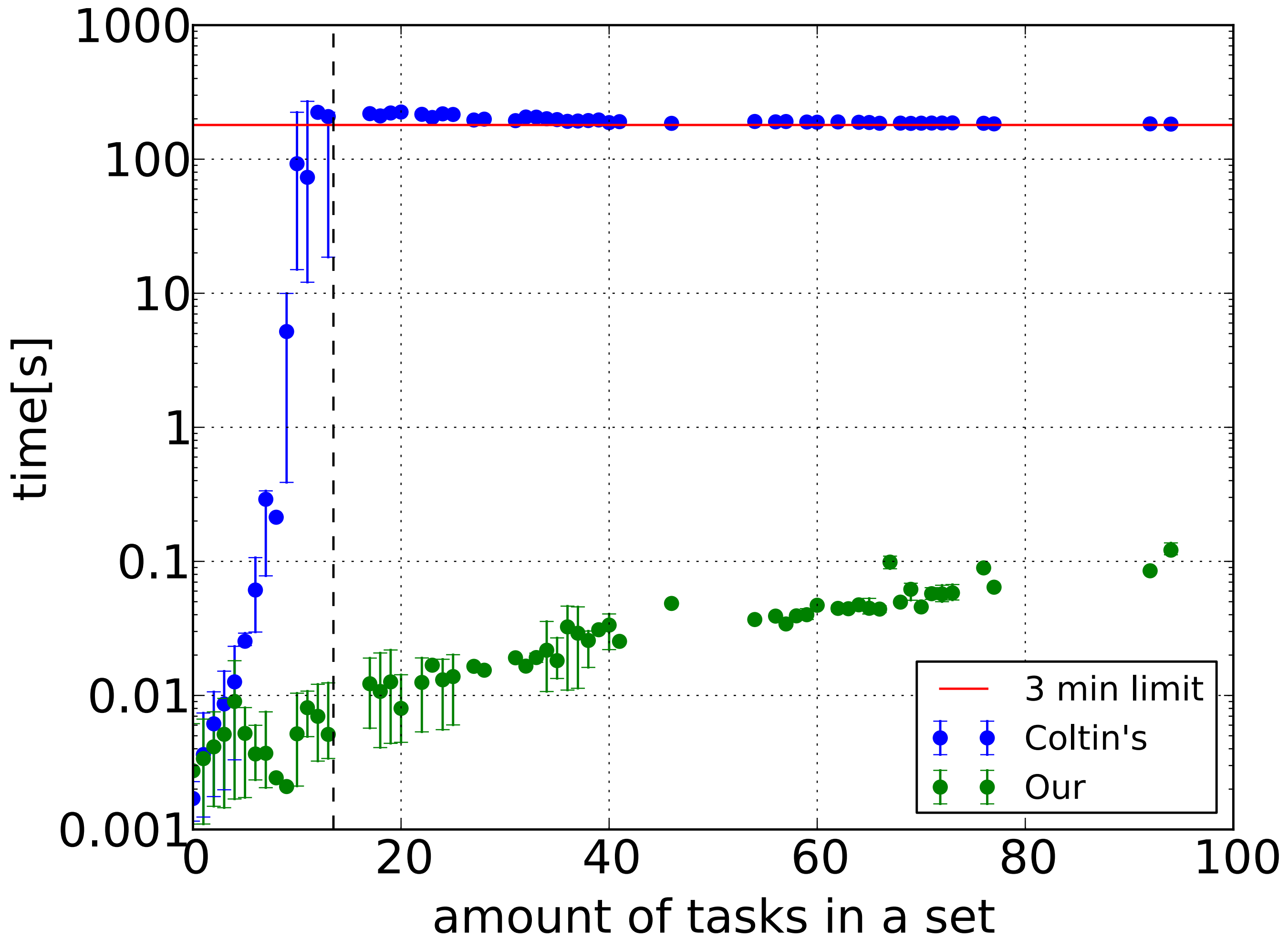
i equals j

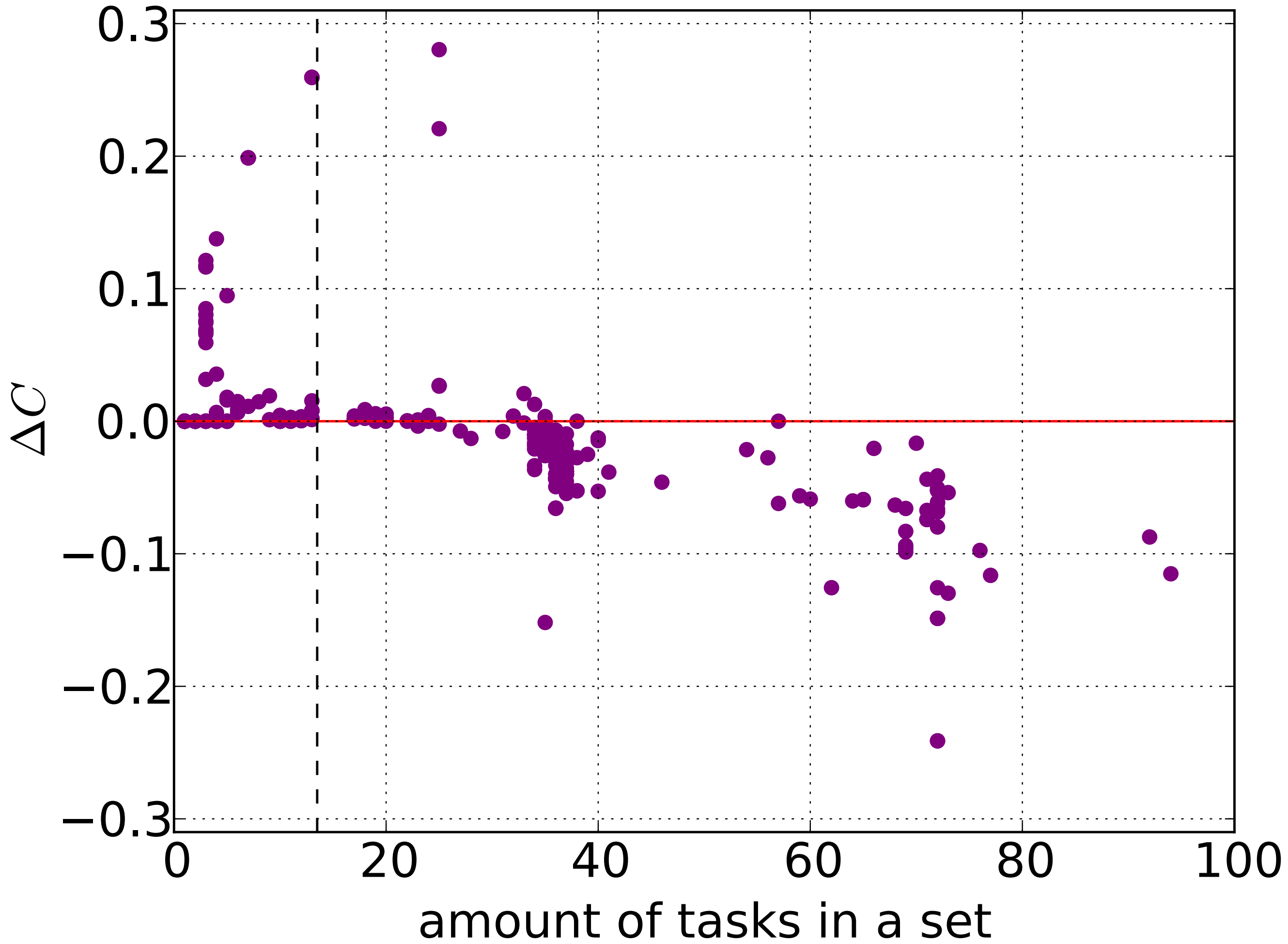
pick first one seen

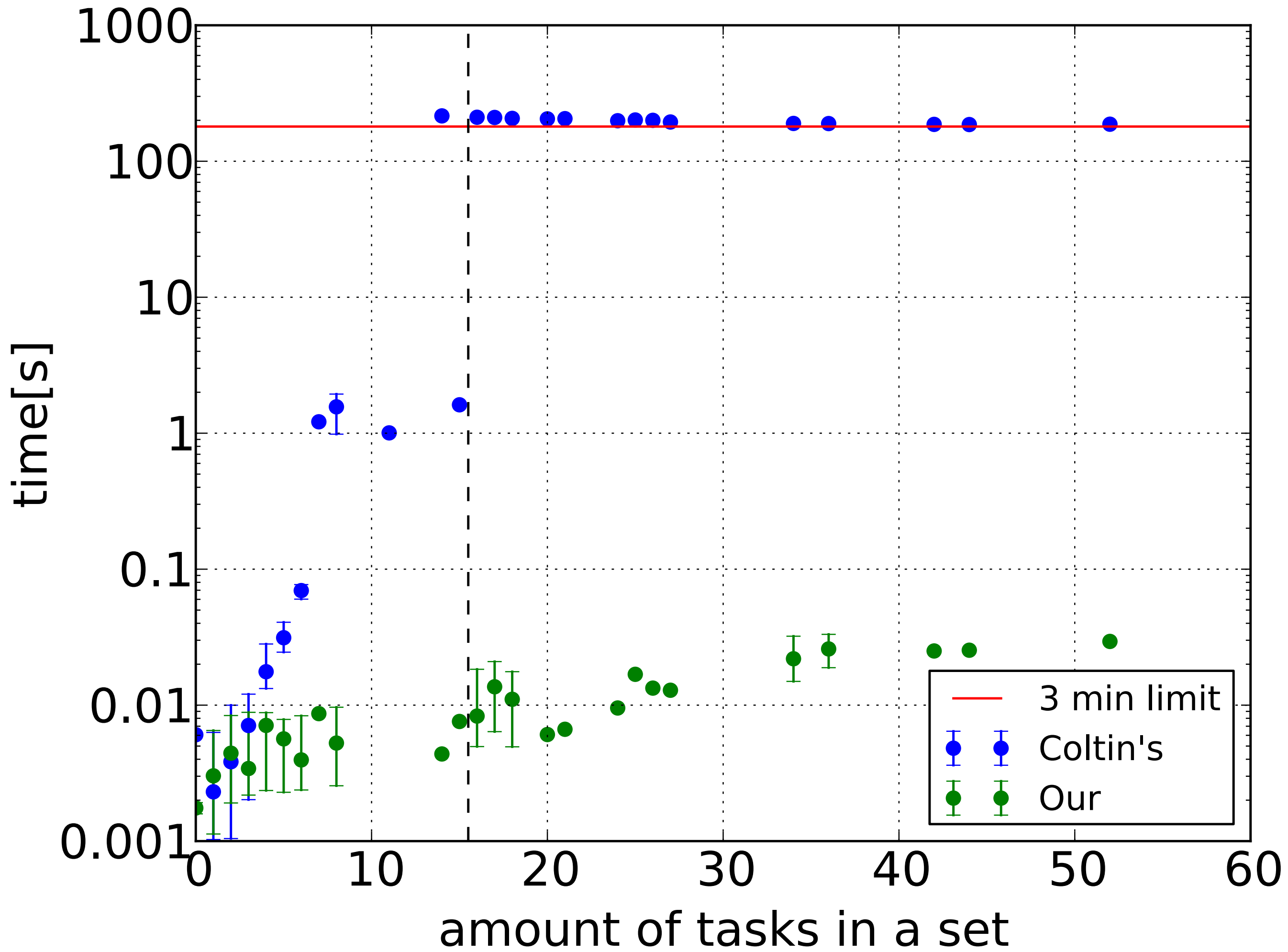


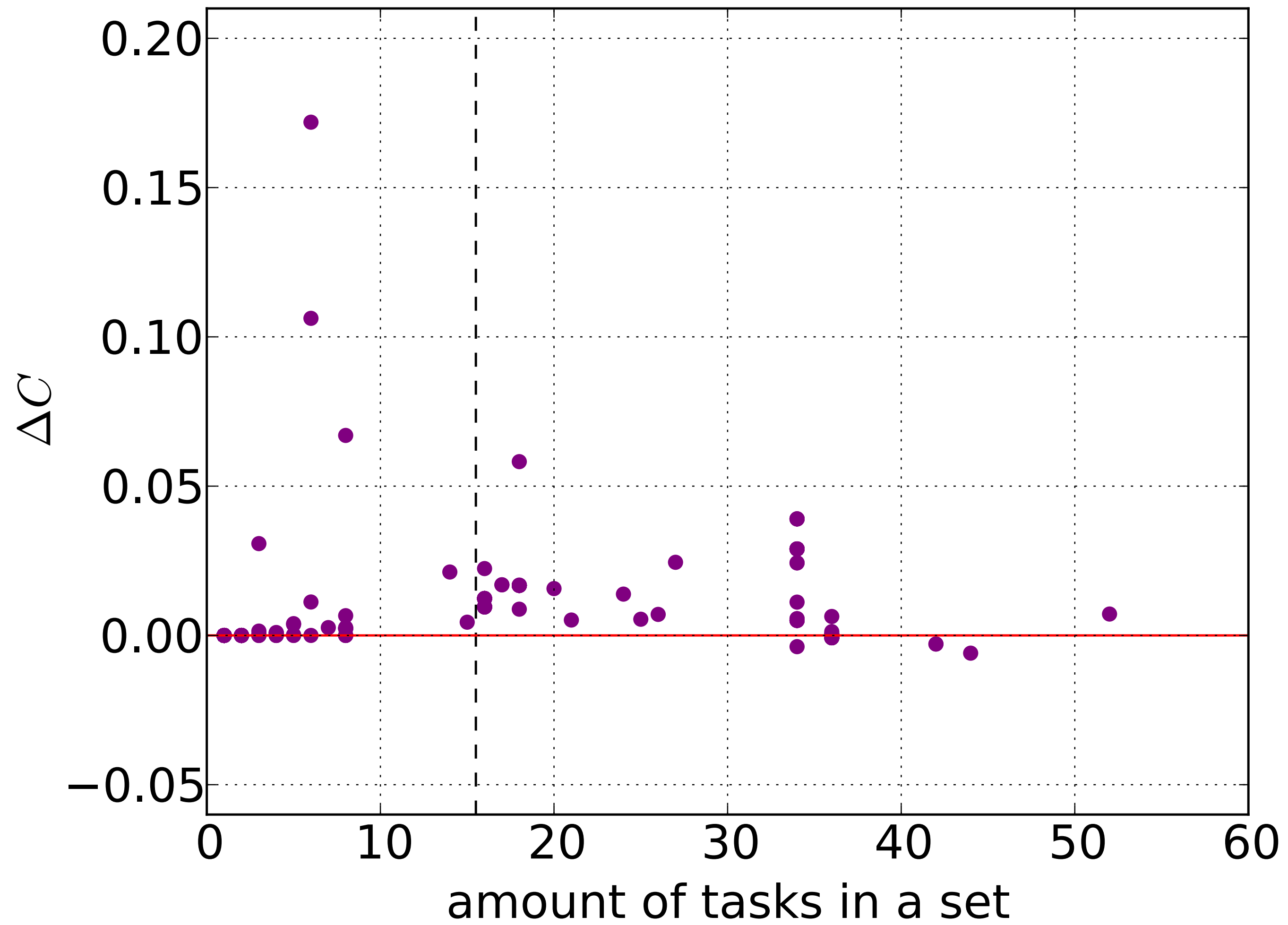
	Care	Security
# Problems	606	358
Smallest Problem	1	1
Largest Problem	135	71
Mean Problem Size	28.88 (σ 26.28)	9.59 (σ 12.97)
# Problems >15	349 (58%)	106 (30%)



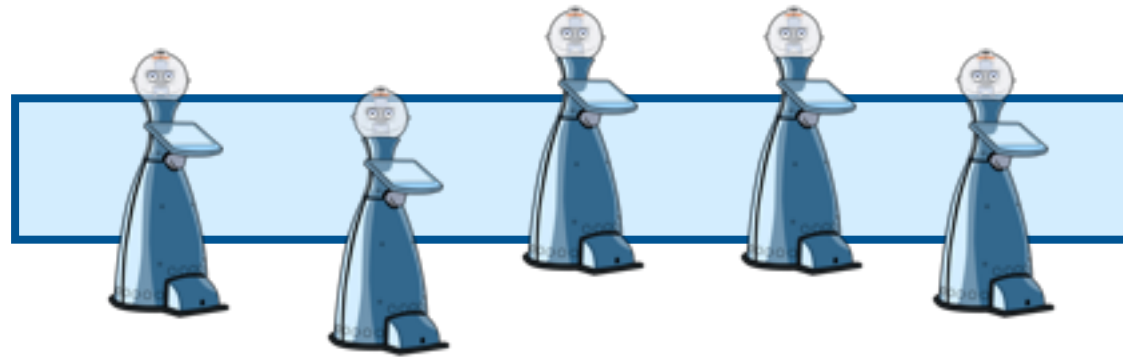








Task



$task_i$

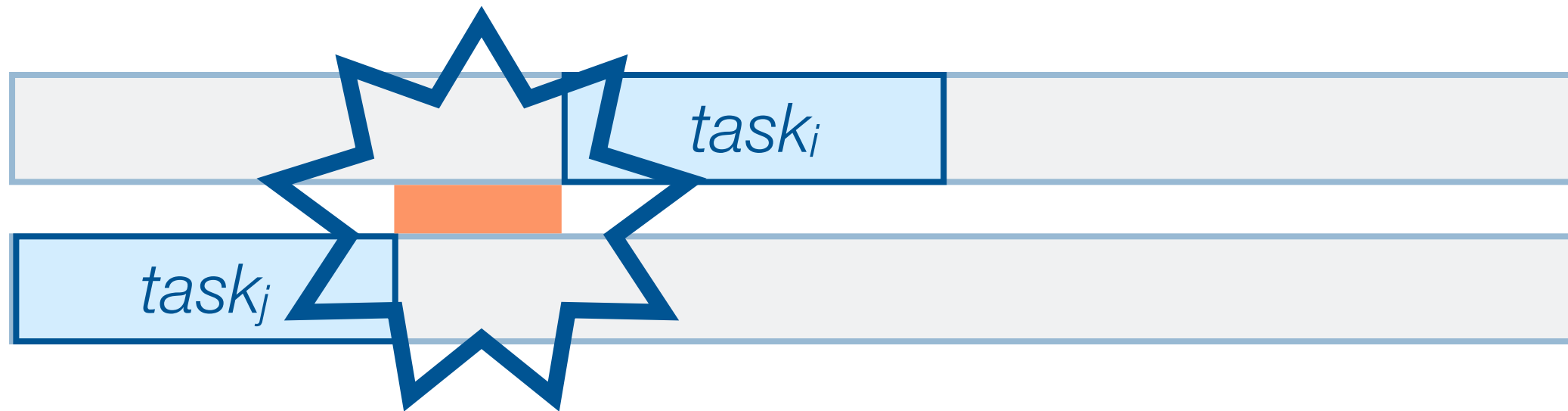
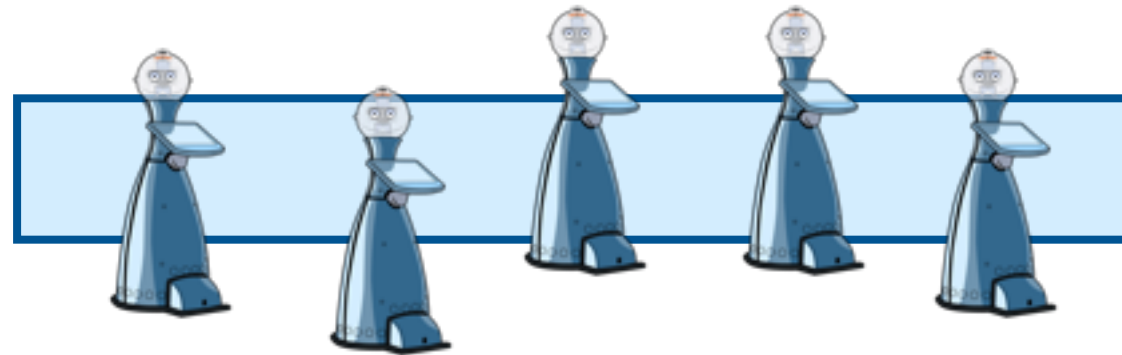
d_i

S_i

e_i



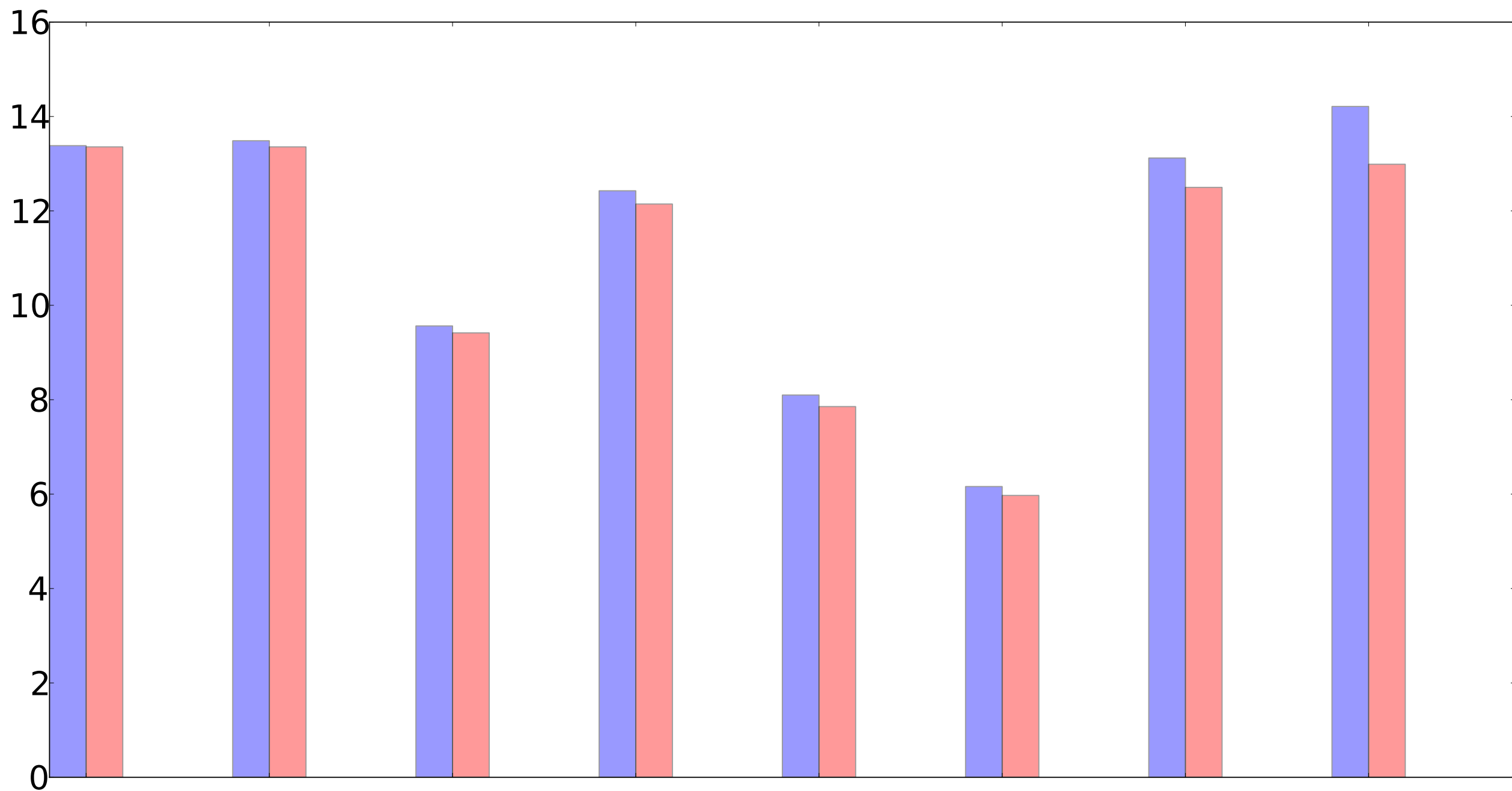
Task



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



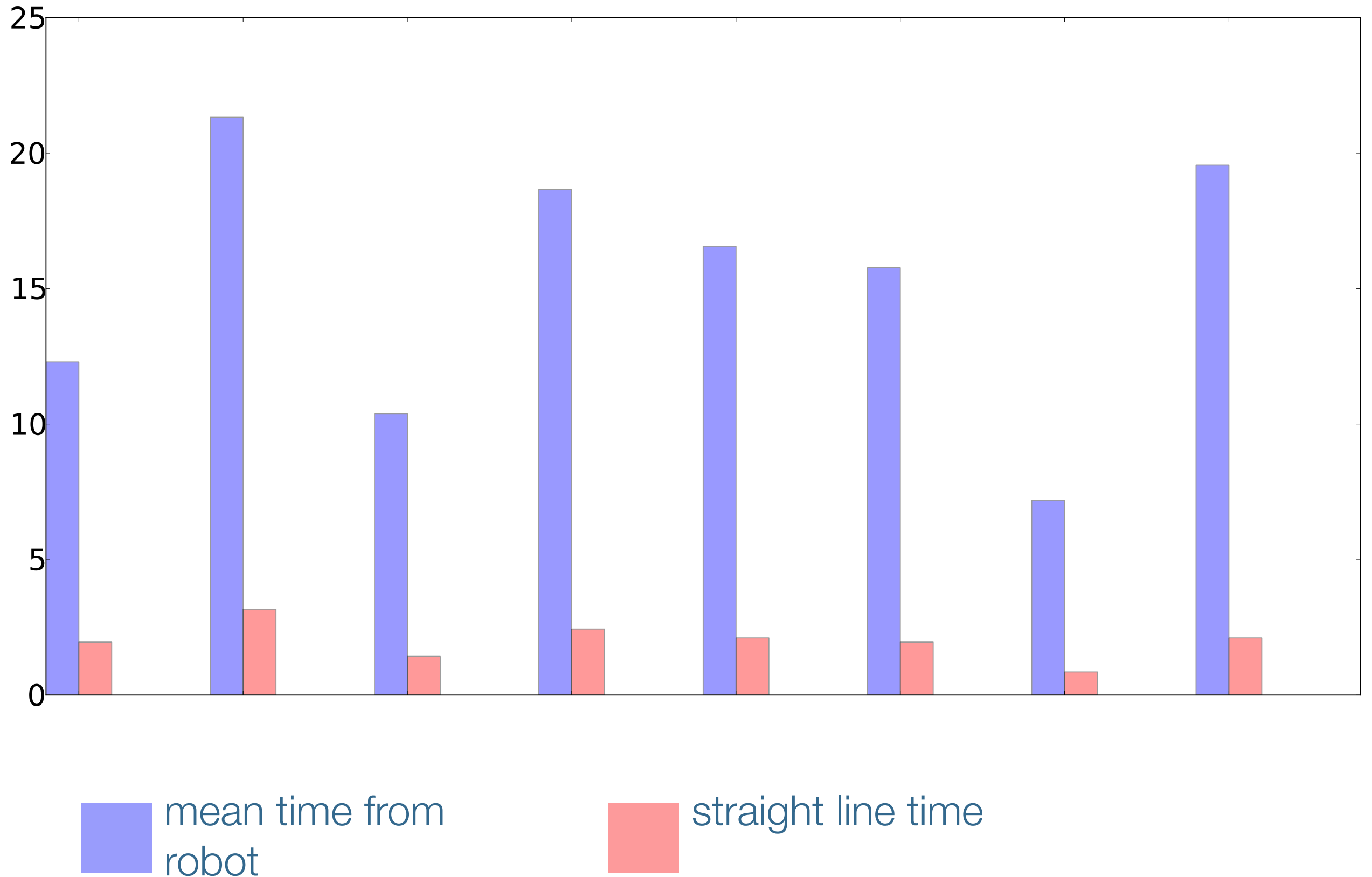
Best 8 matches between straight-line and recorded times

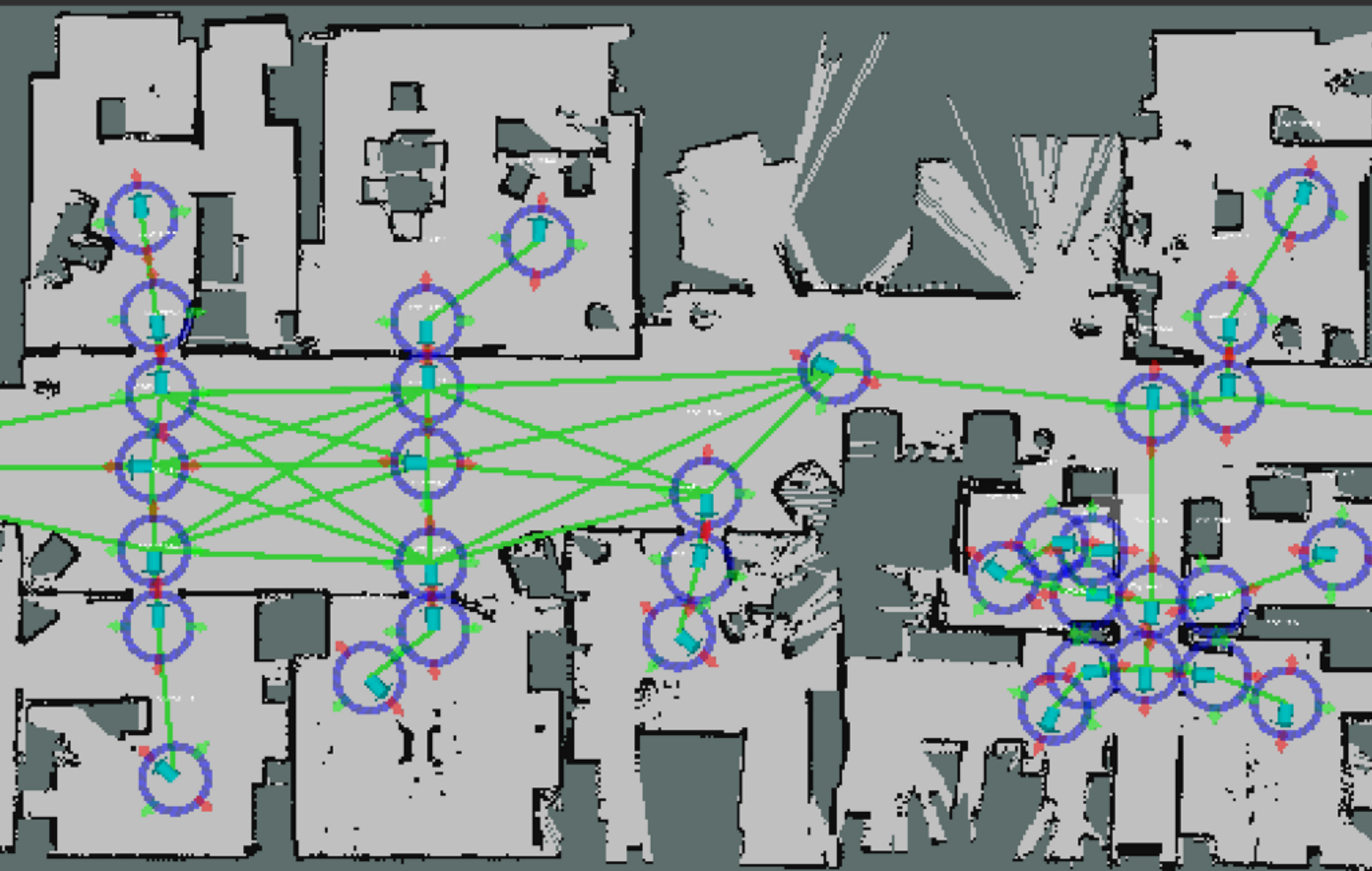


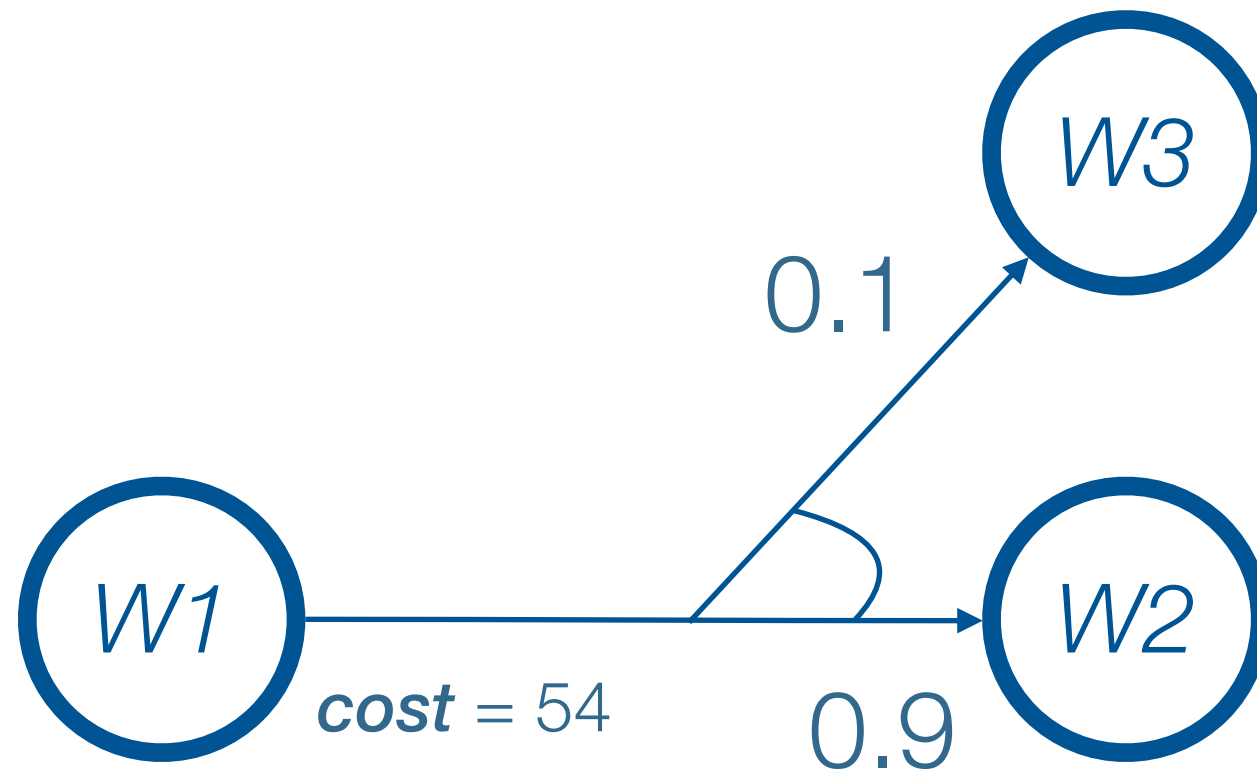
mean time from robot

straight line time

Worst 8 matches between straight-line and recorded times





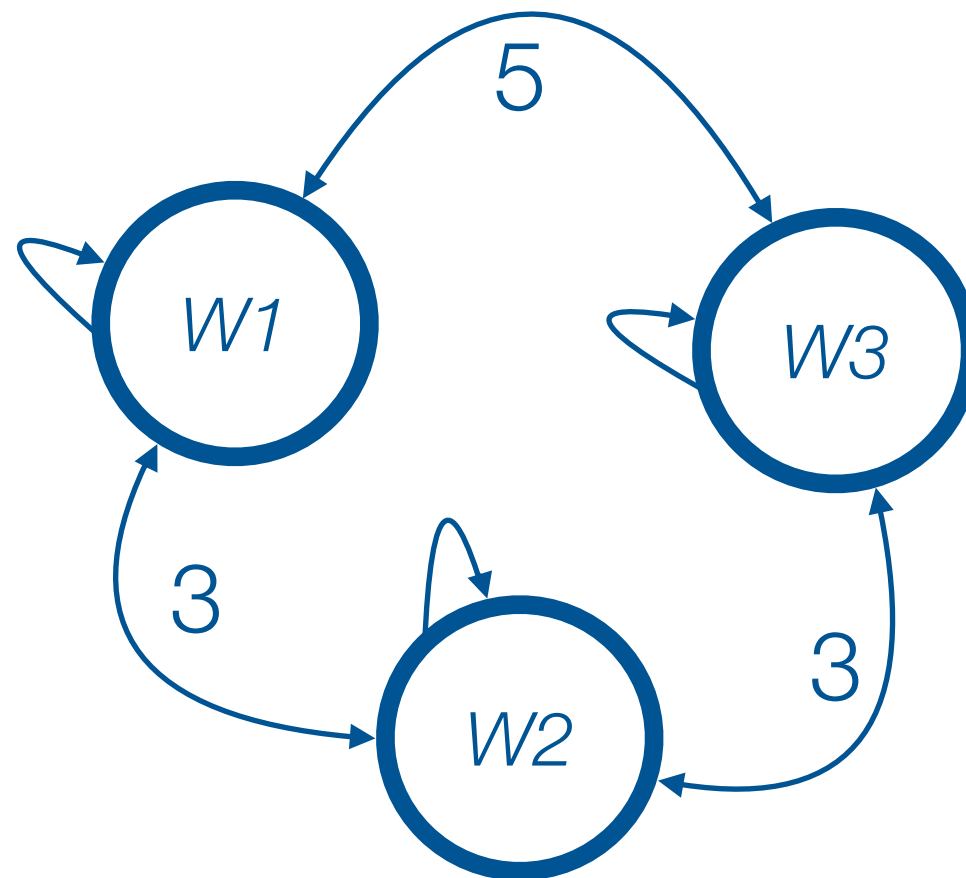
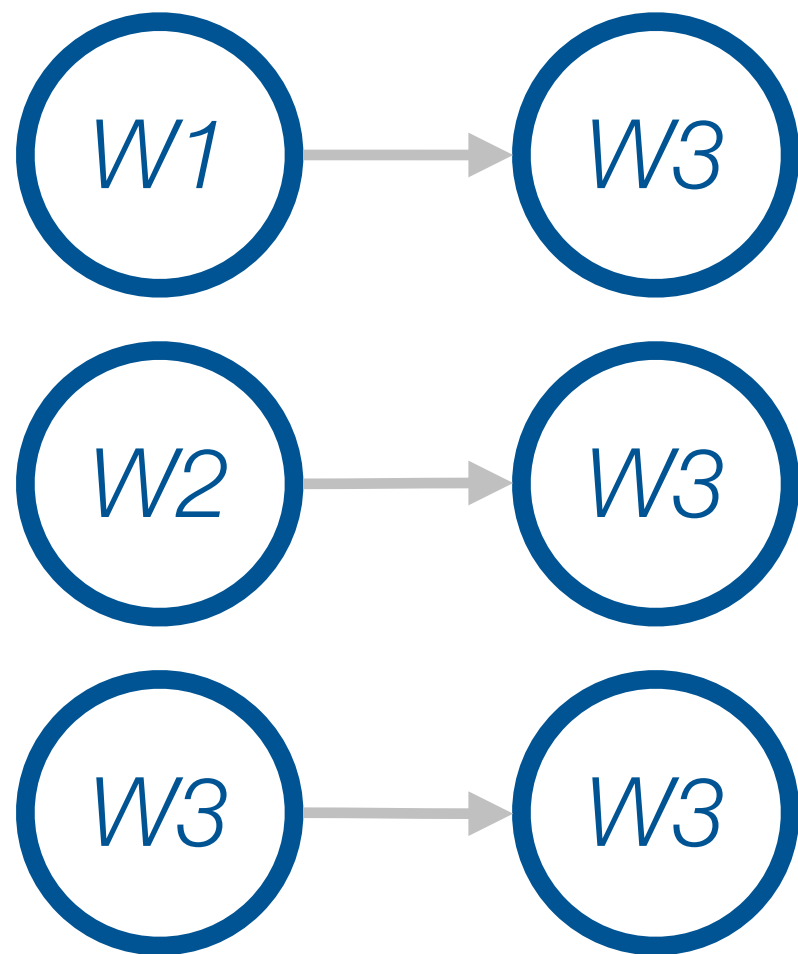


action goto $W2$ from $W1$

Why use an MDP?

Goal is to be in state $W3$

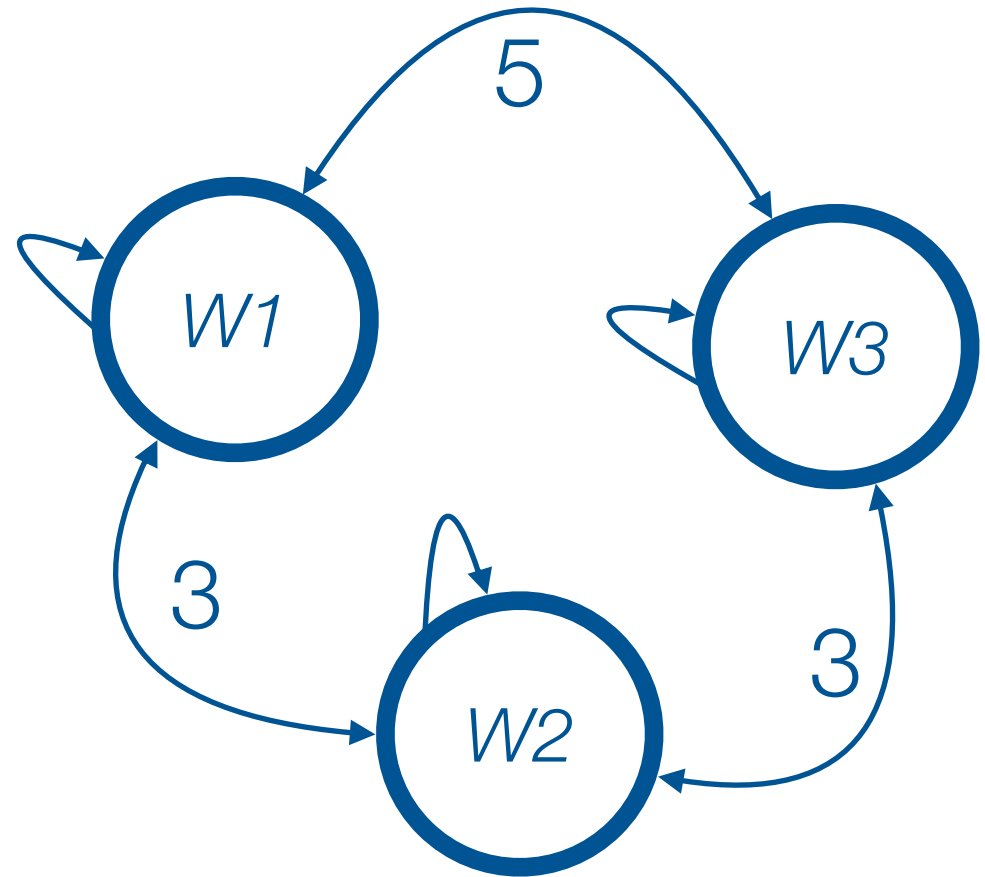
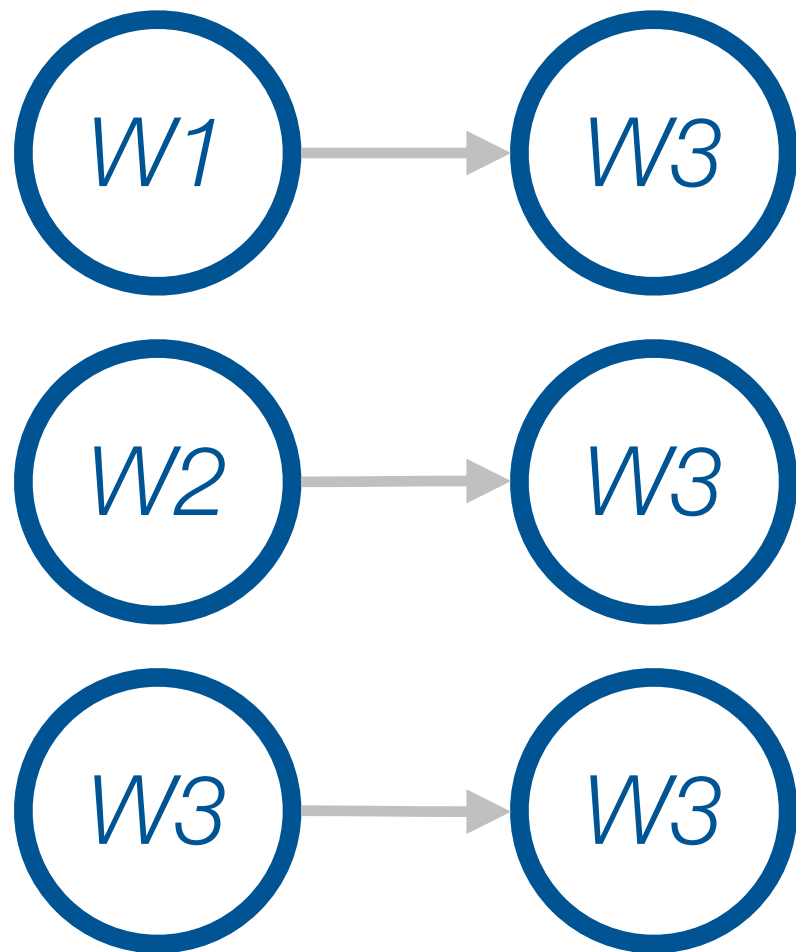
Policy:



$(F W2)$

eventually reach $W2$

Policy:

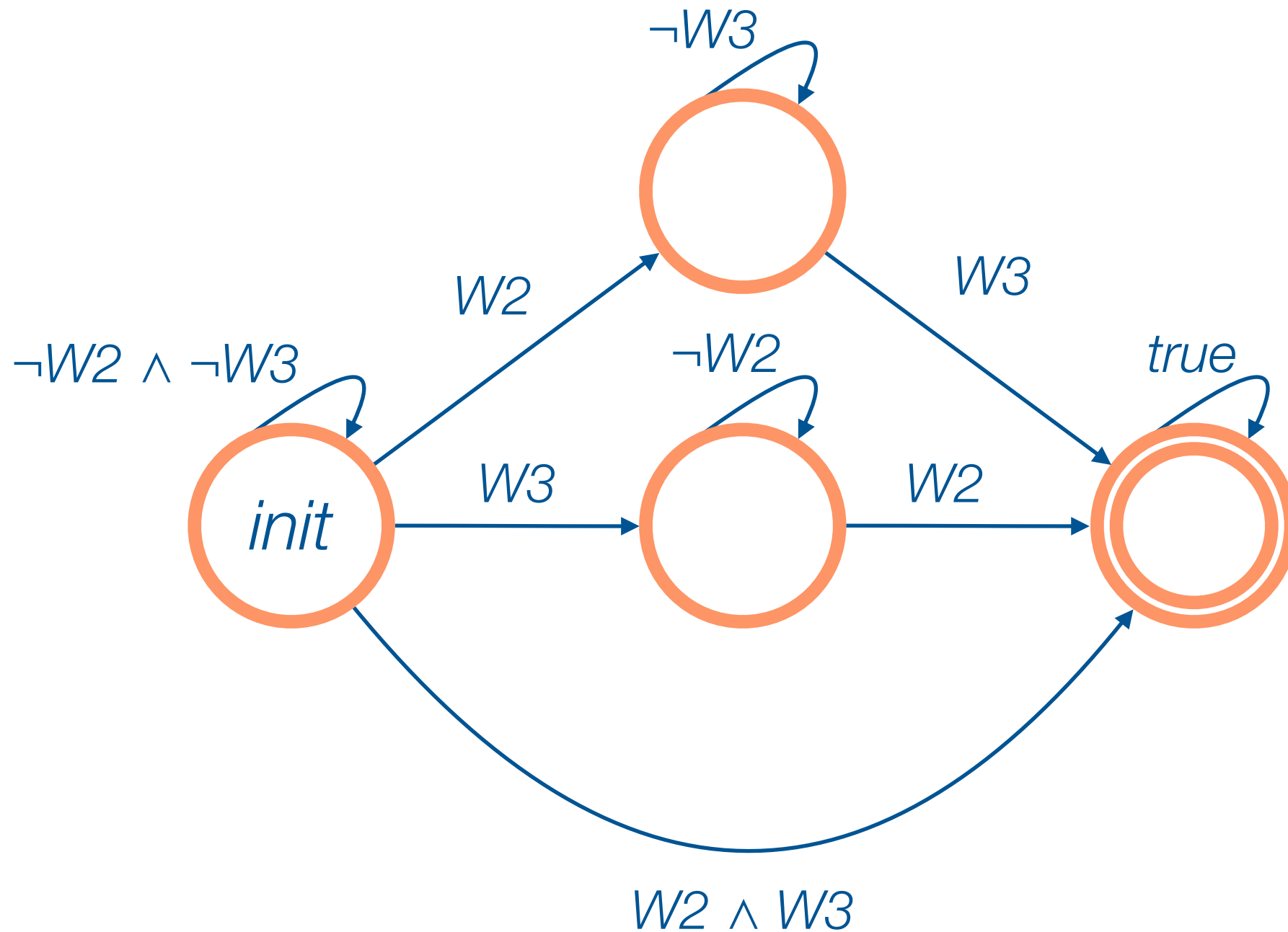


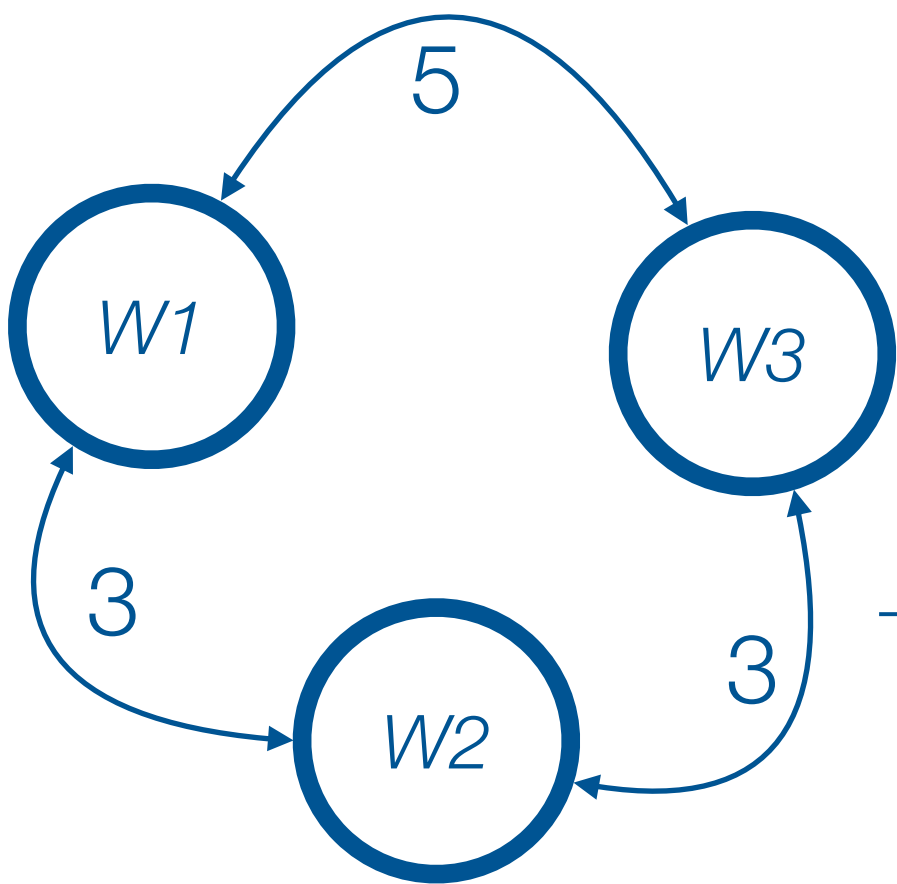
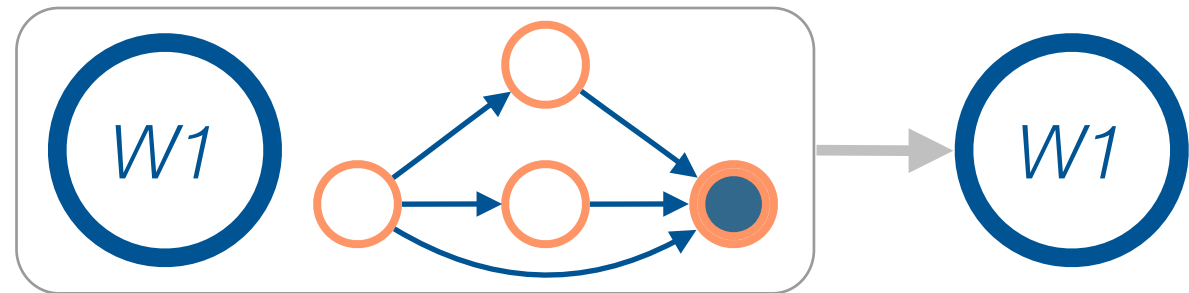
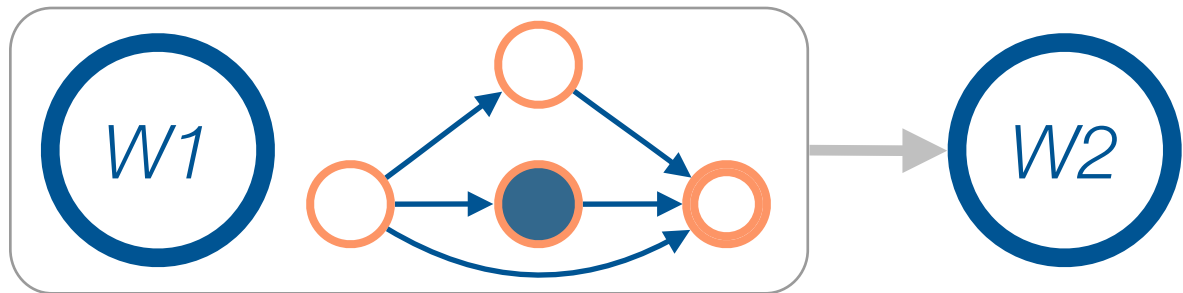
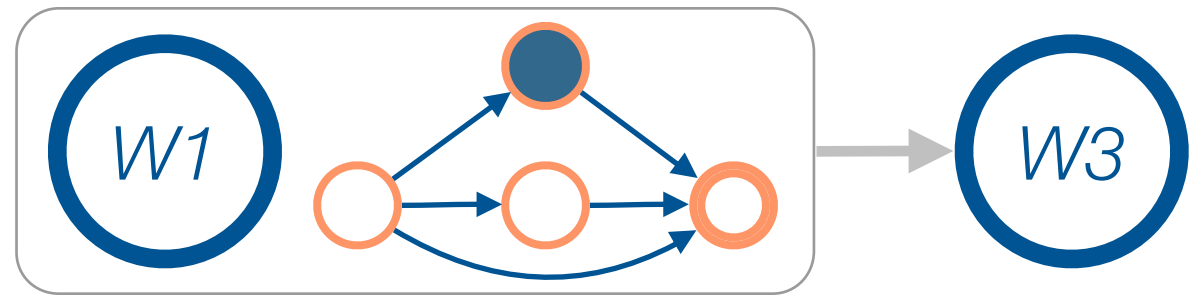
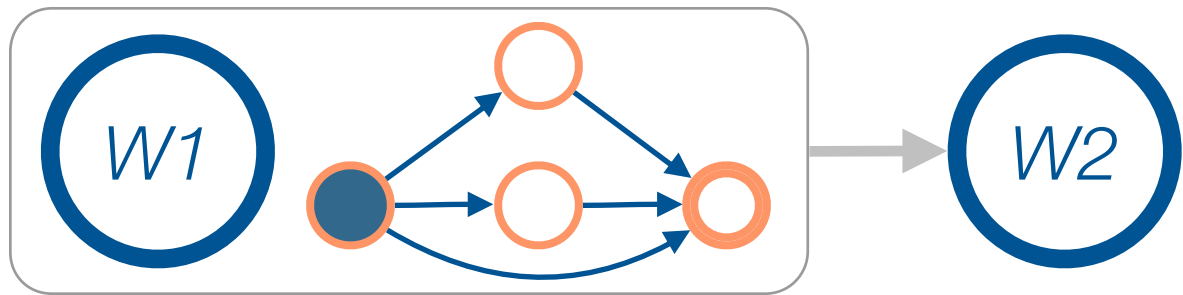
$(F W2) \wedge (F W3)$

eventually reach $W2$ and $W3$

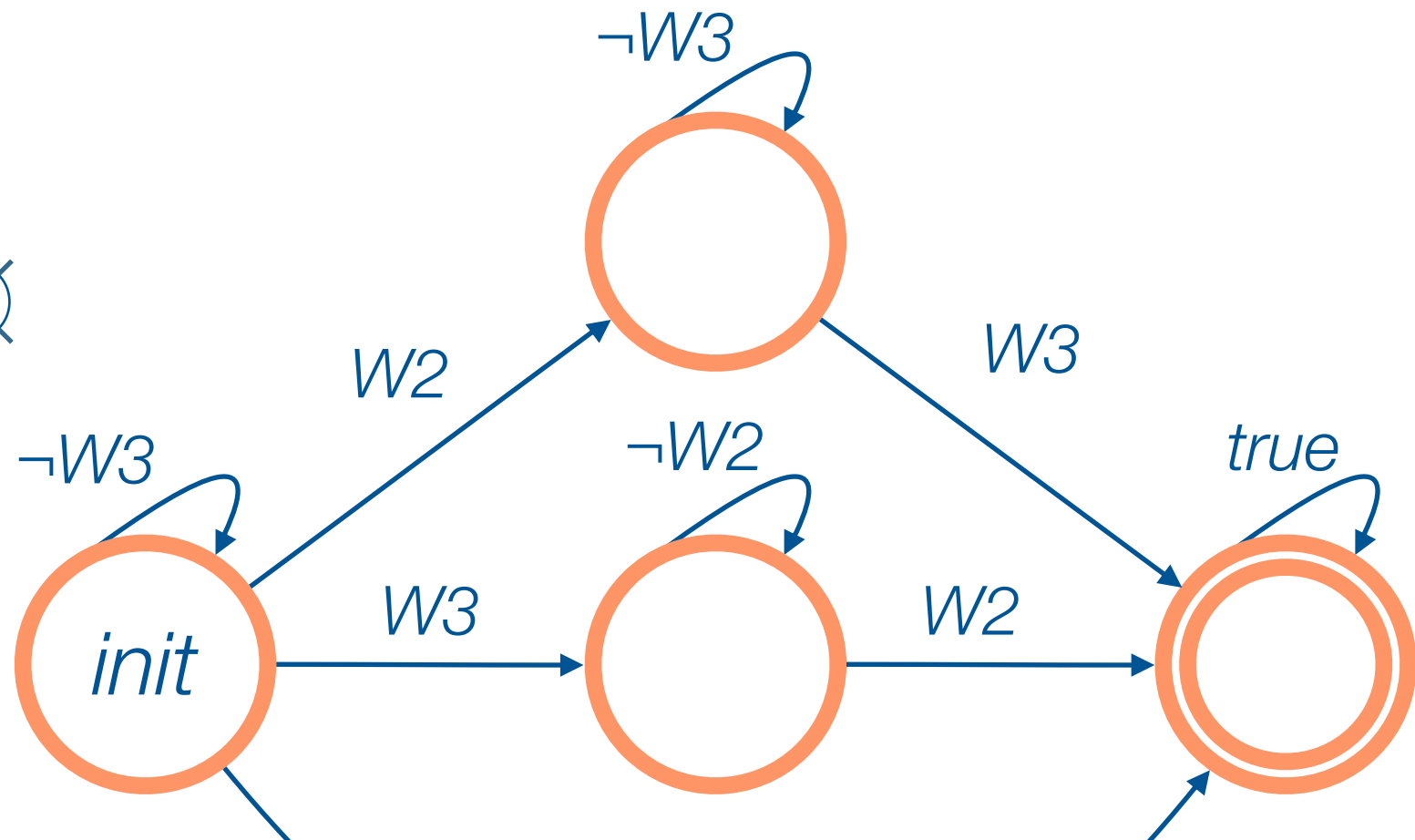
$$(F W2) \wedge (F W3)$$

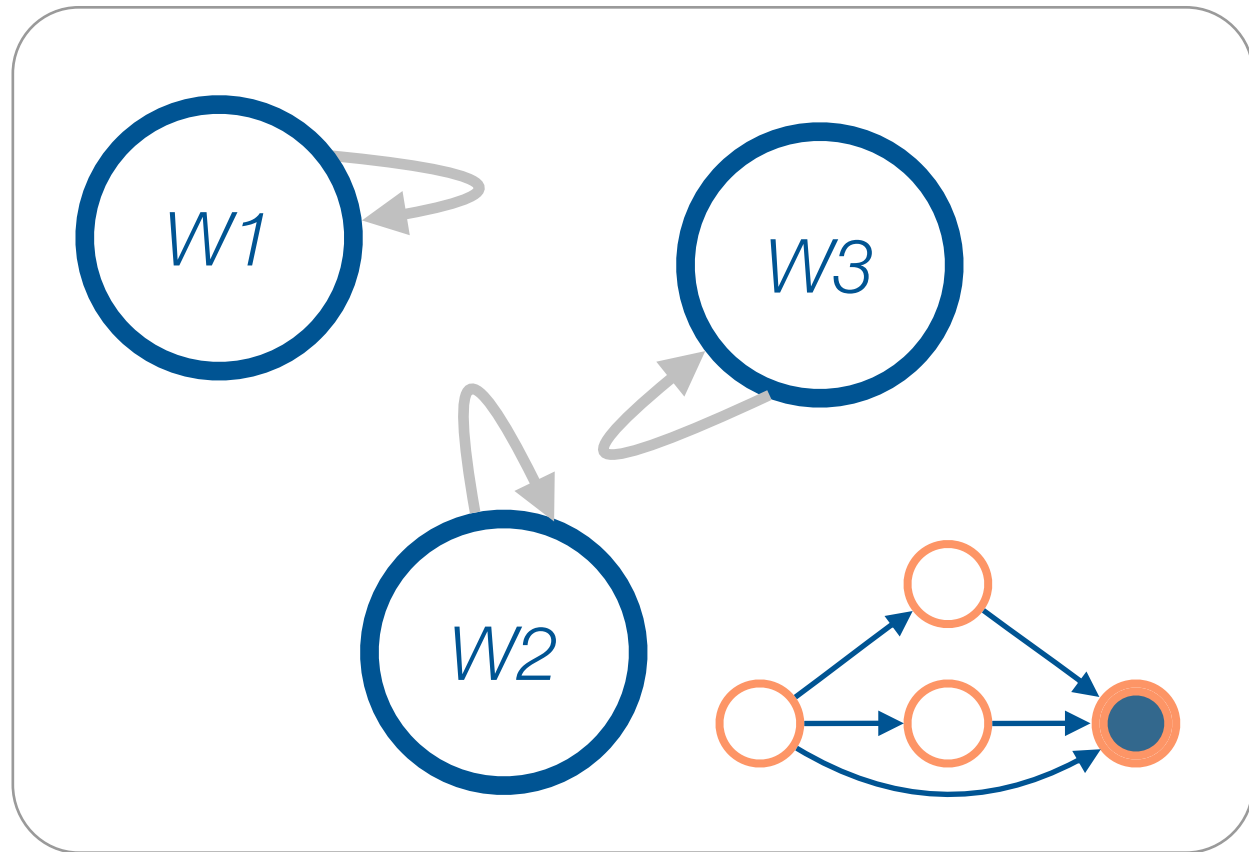
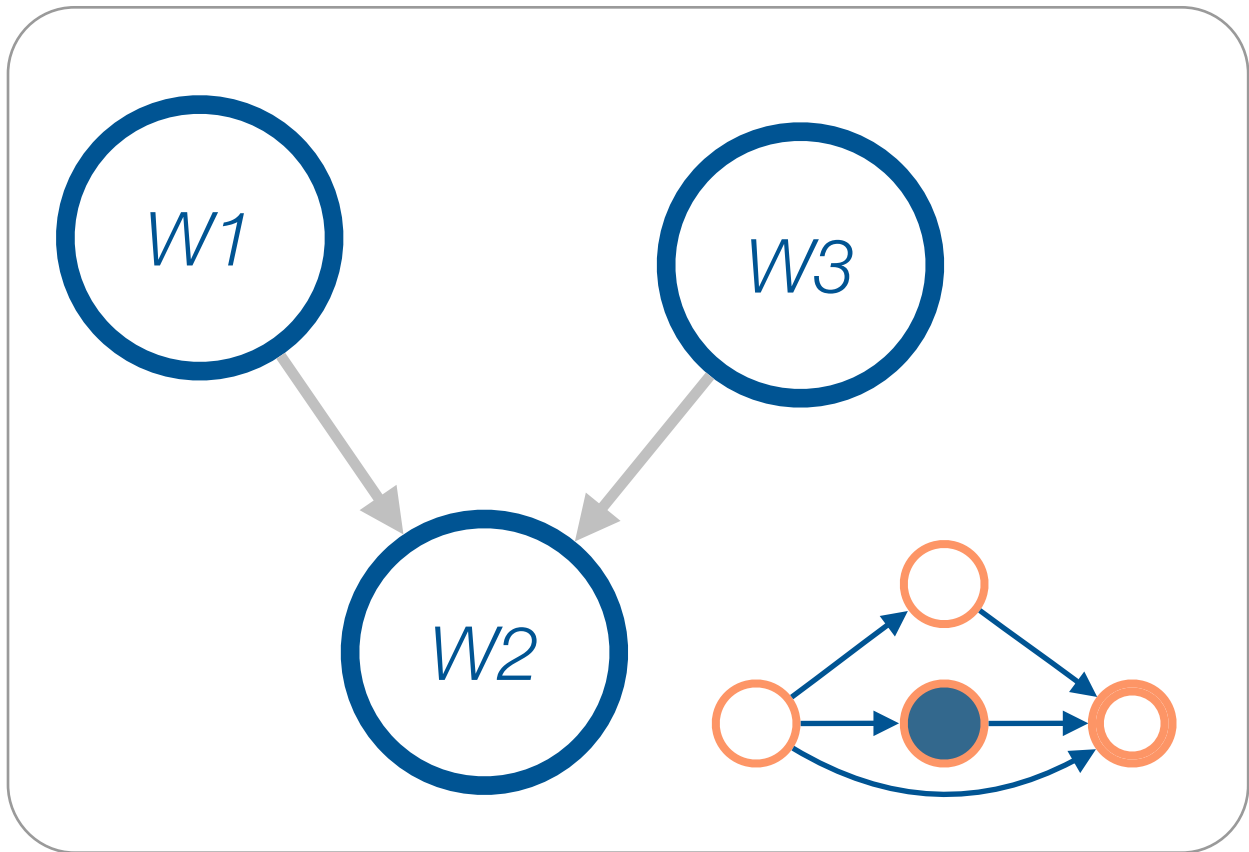
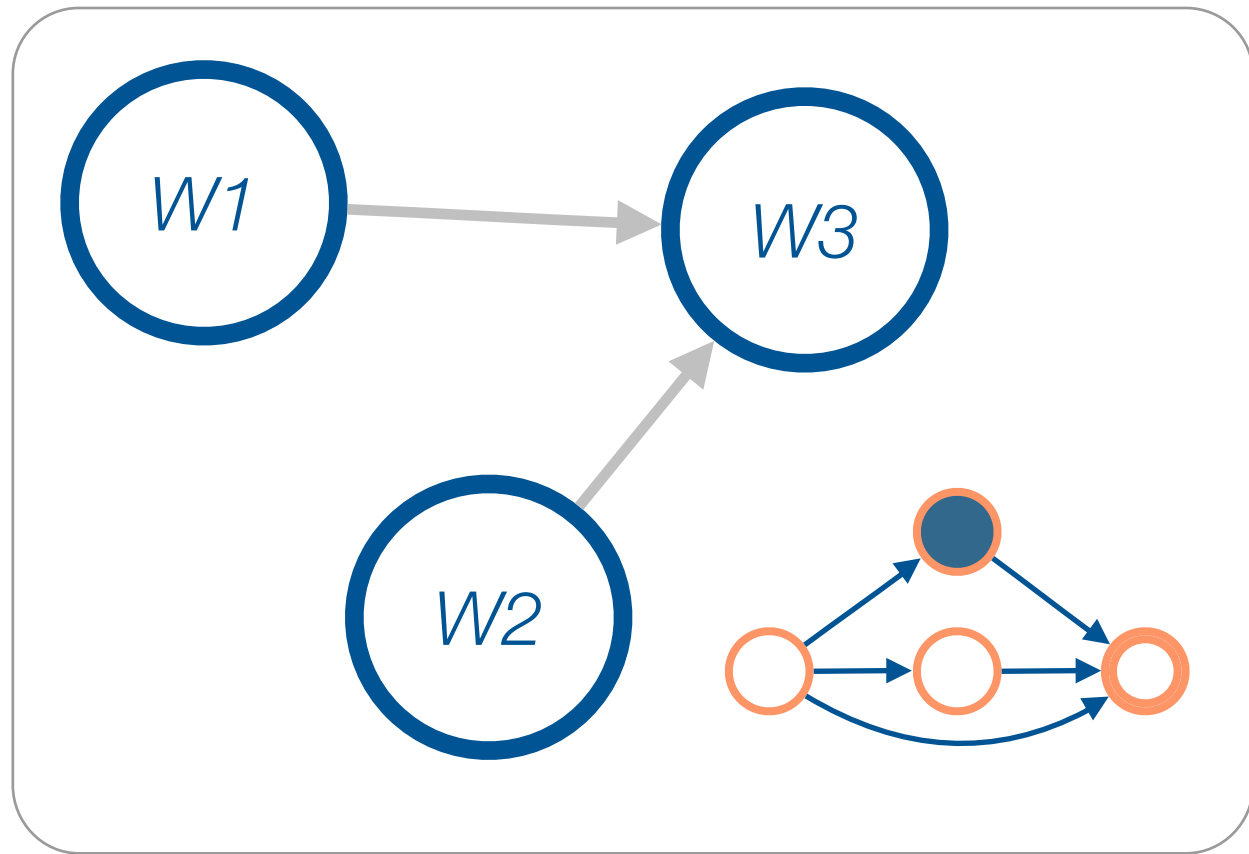
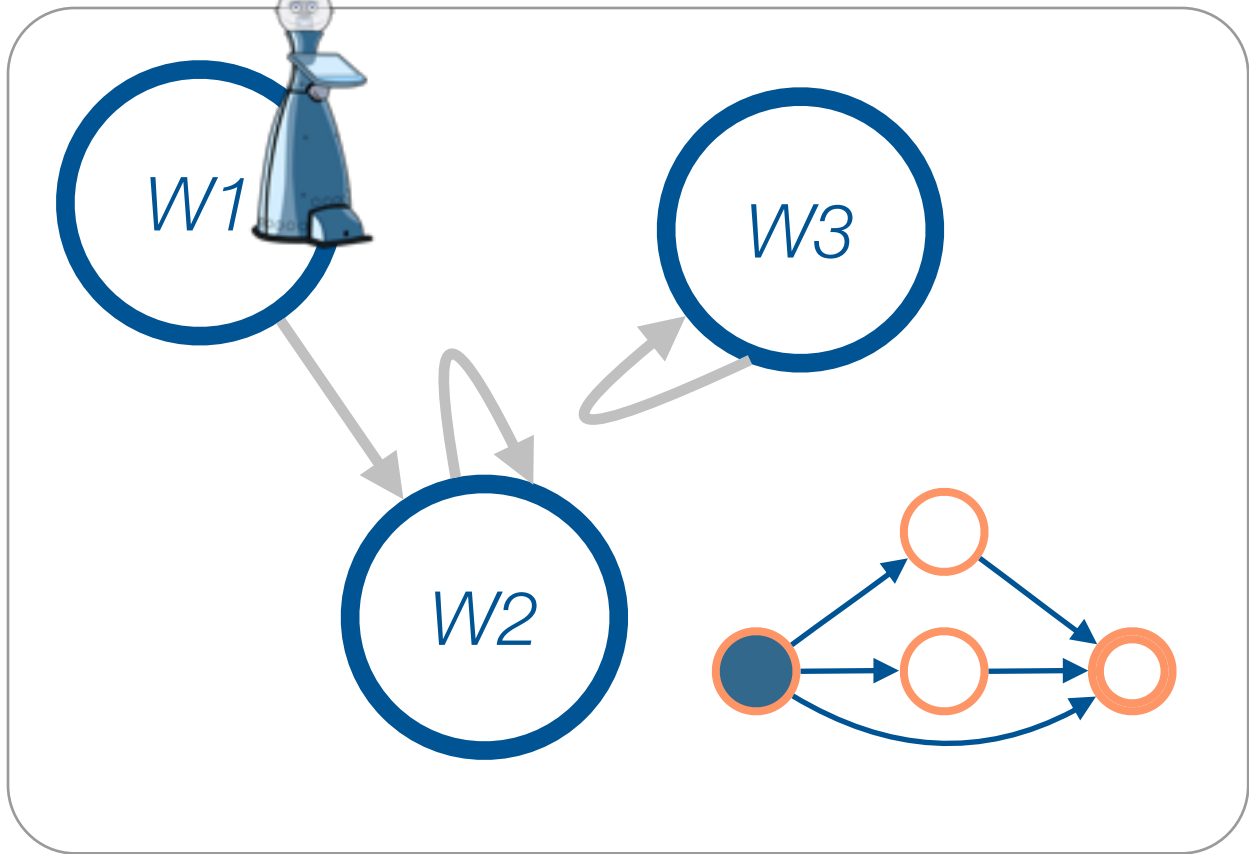
eventually reach $W2$ and $W3$

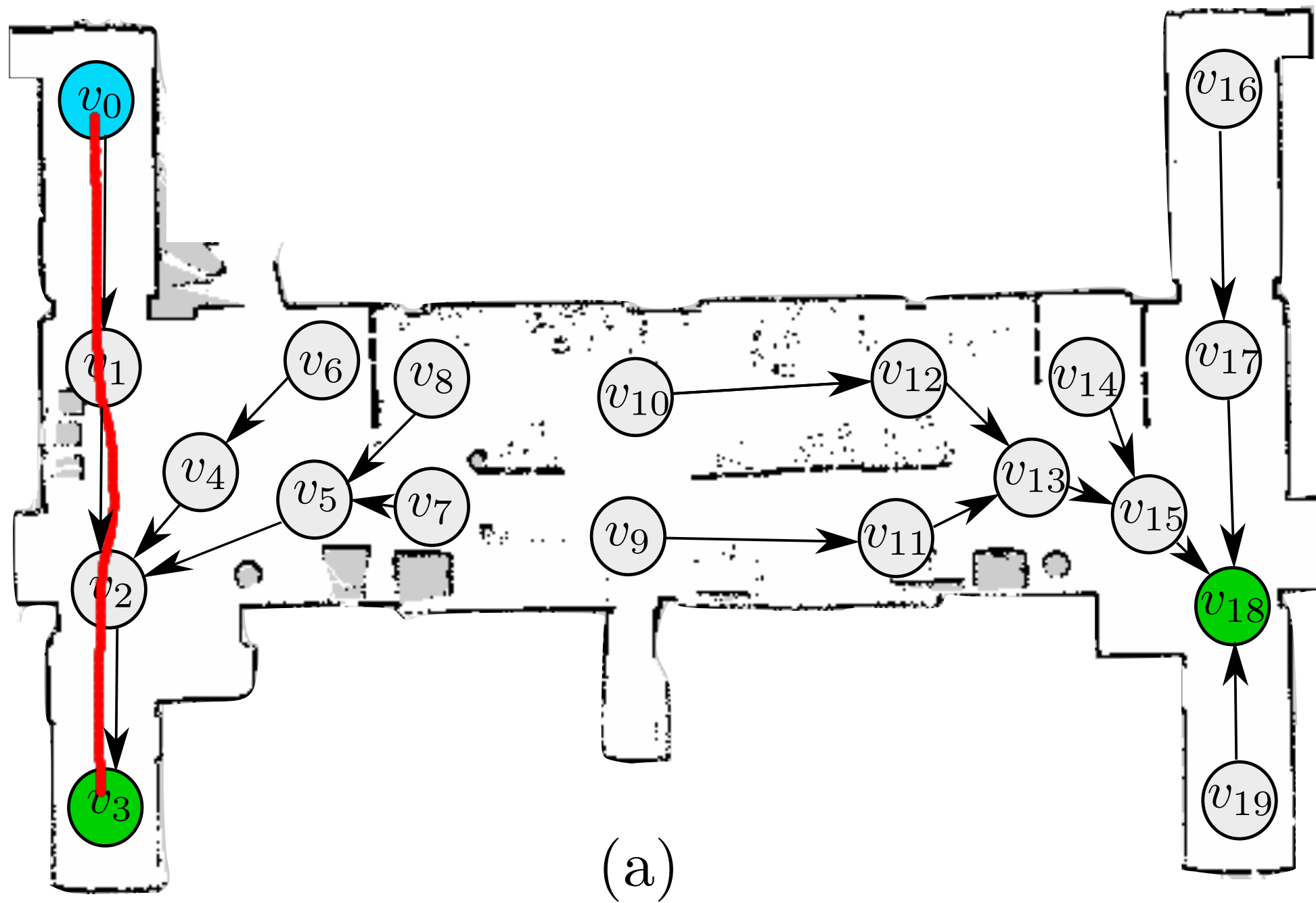


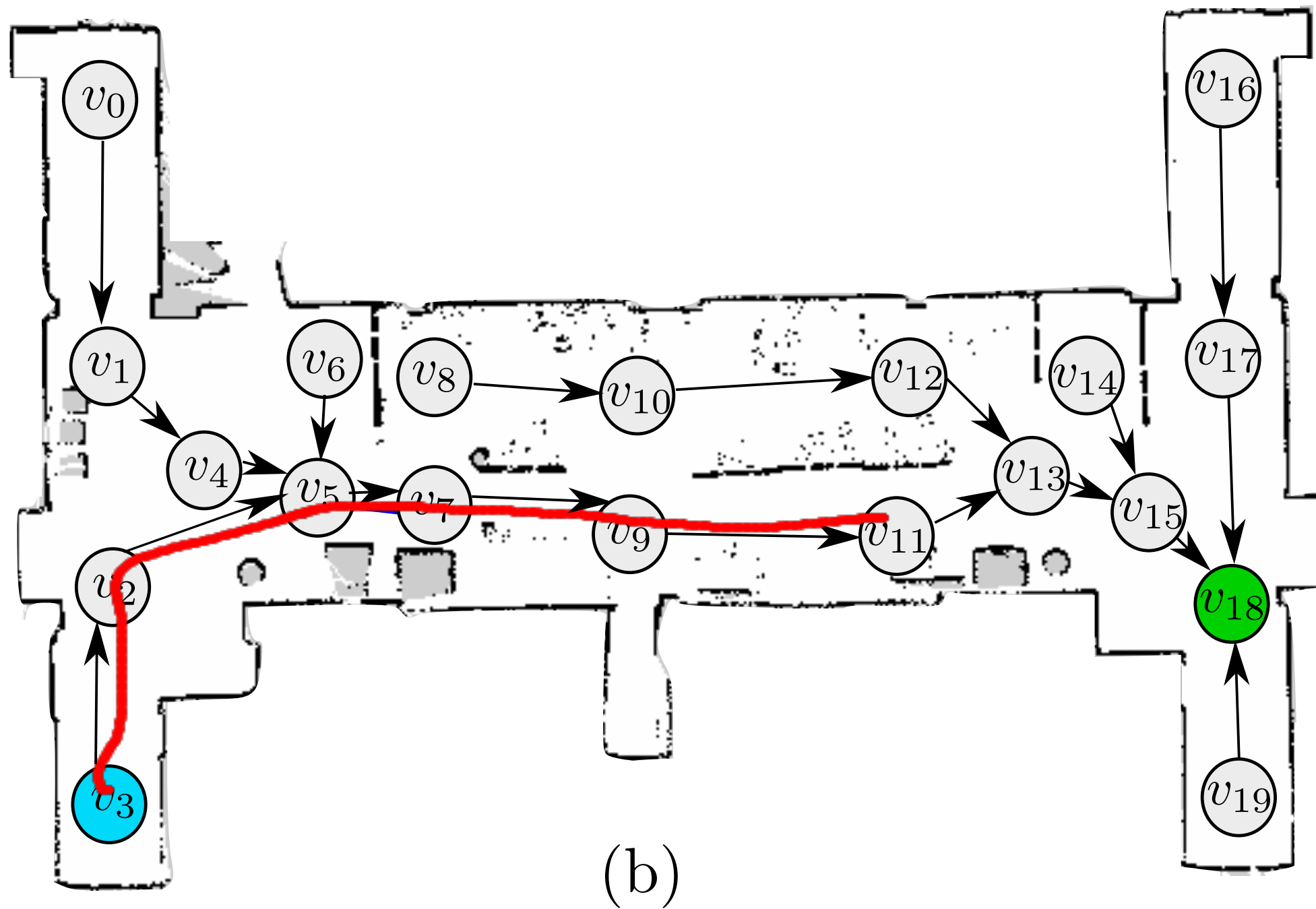


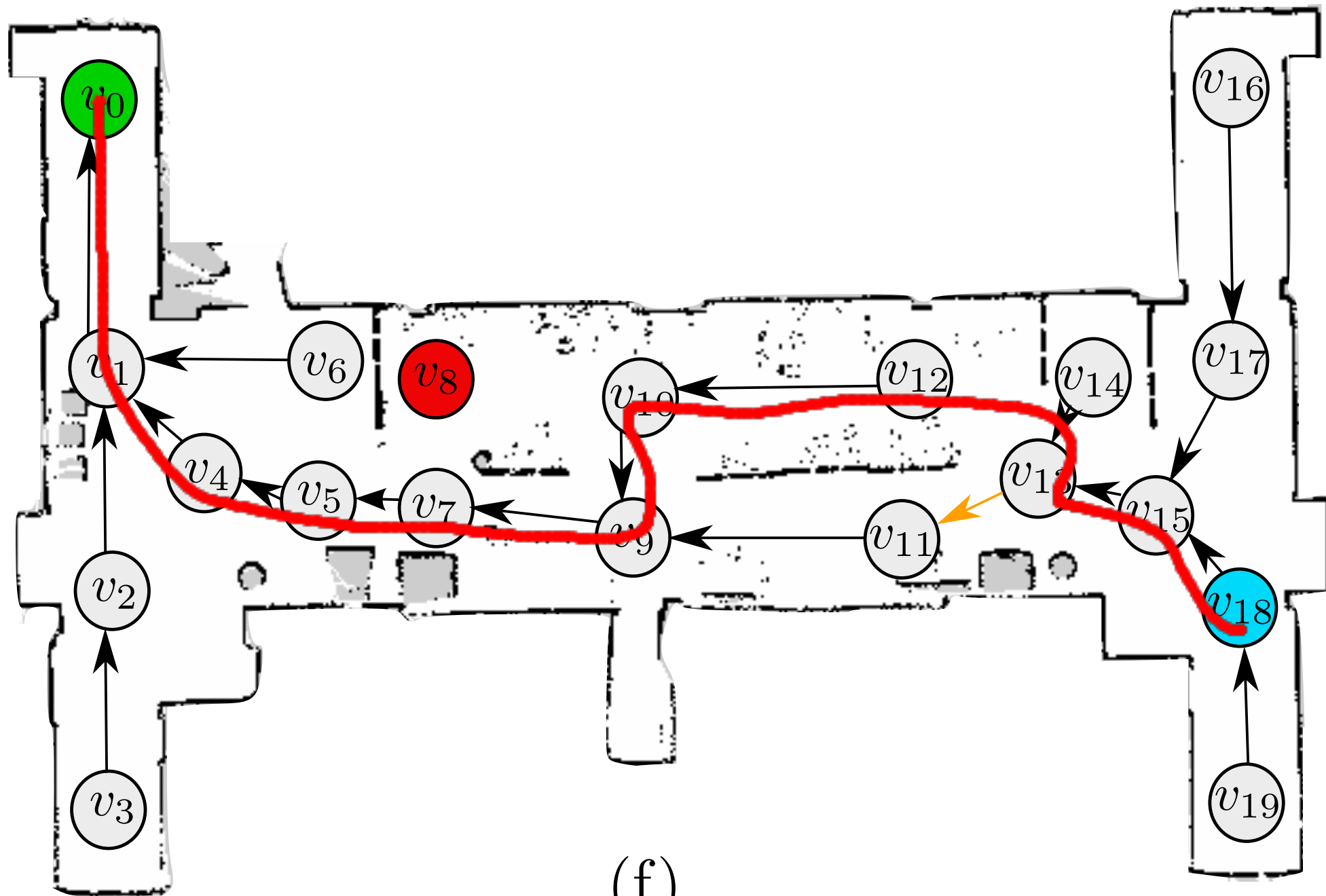
\otimes
 $\neg W2 \wedge \neg W3$



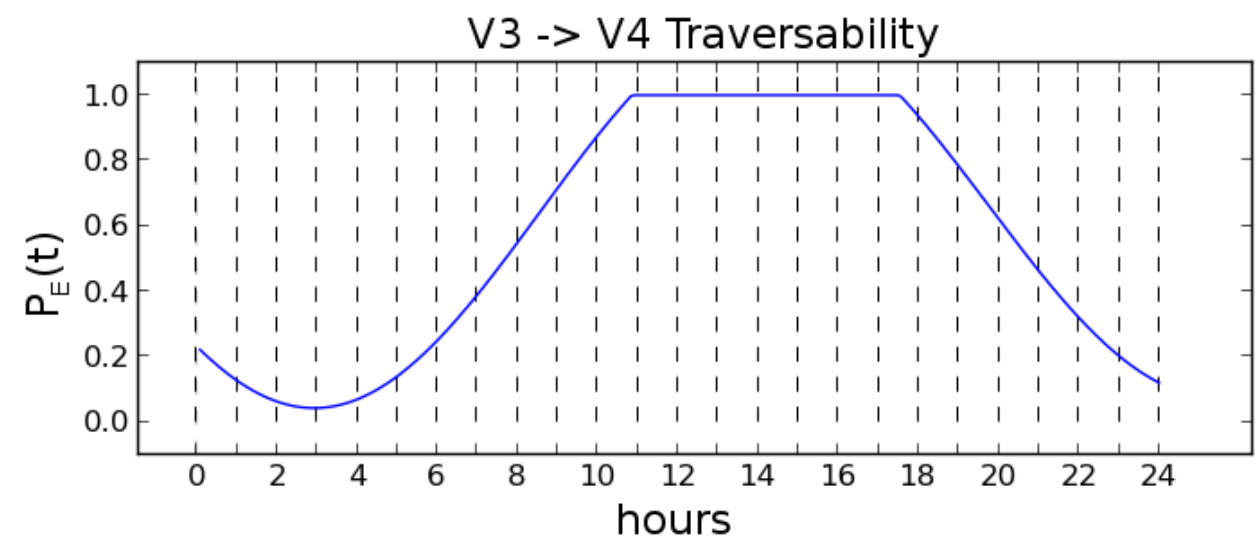
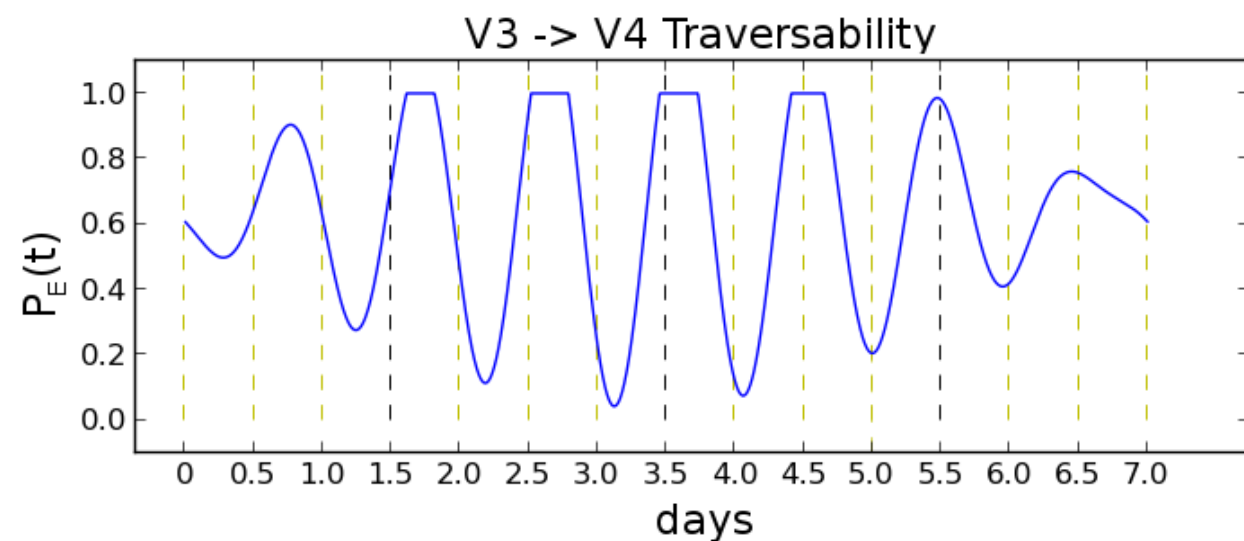
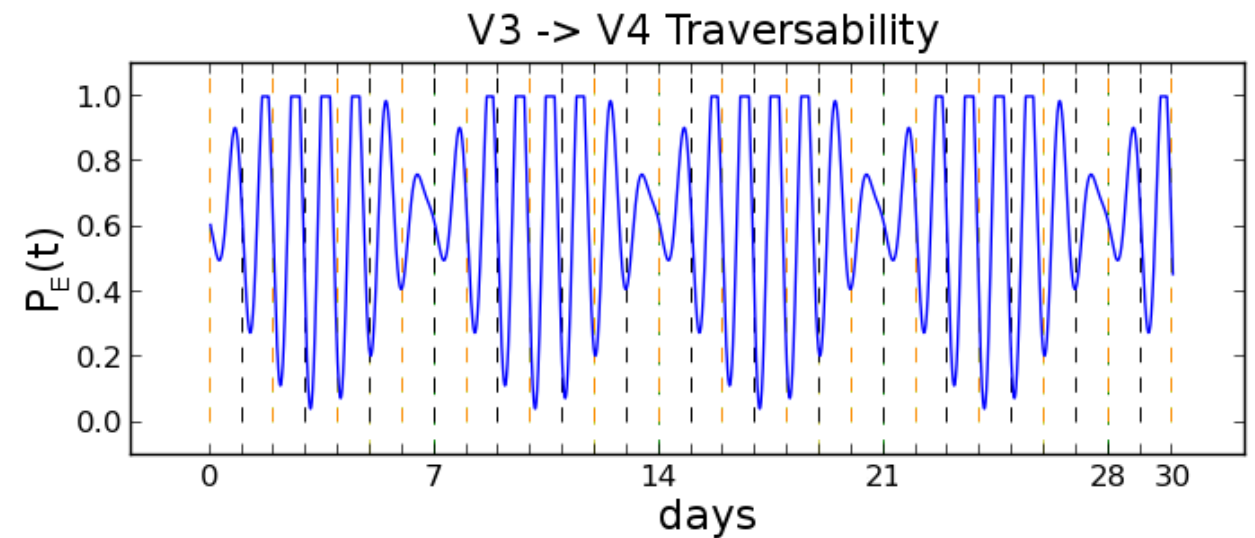
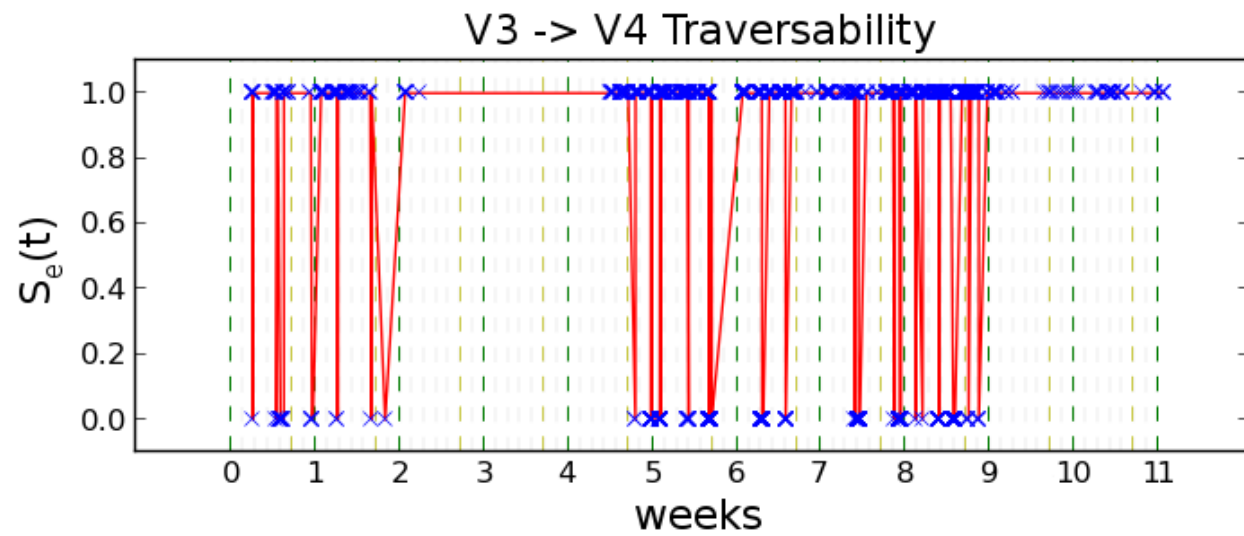






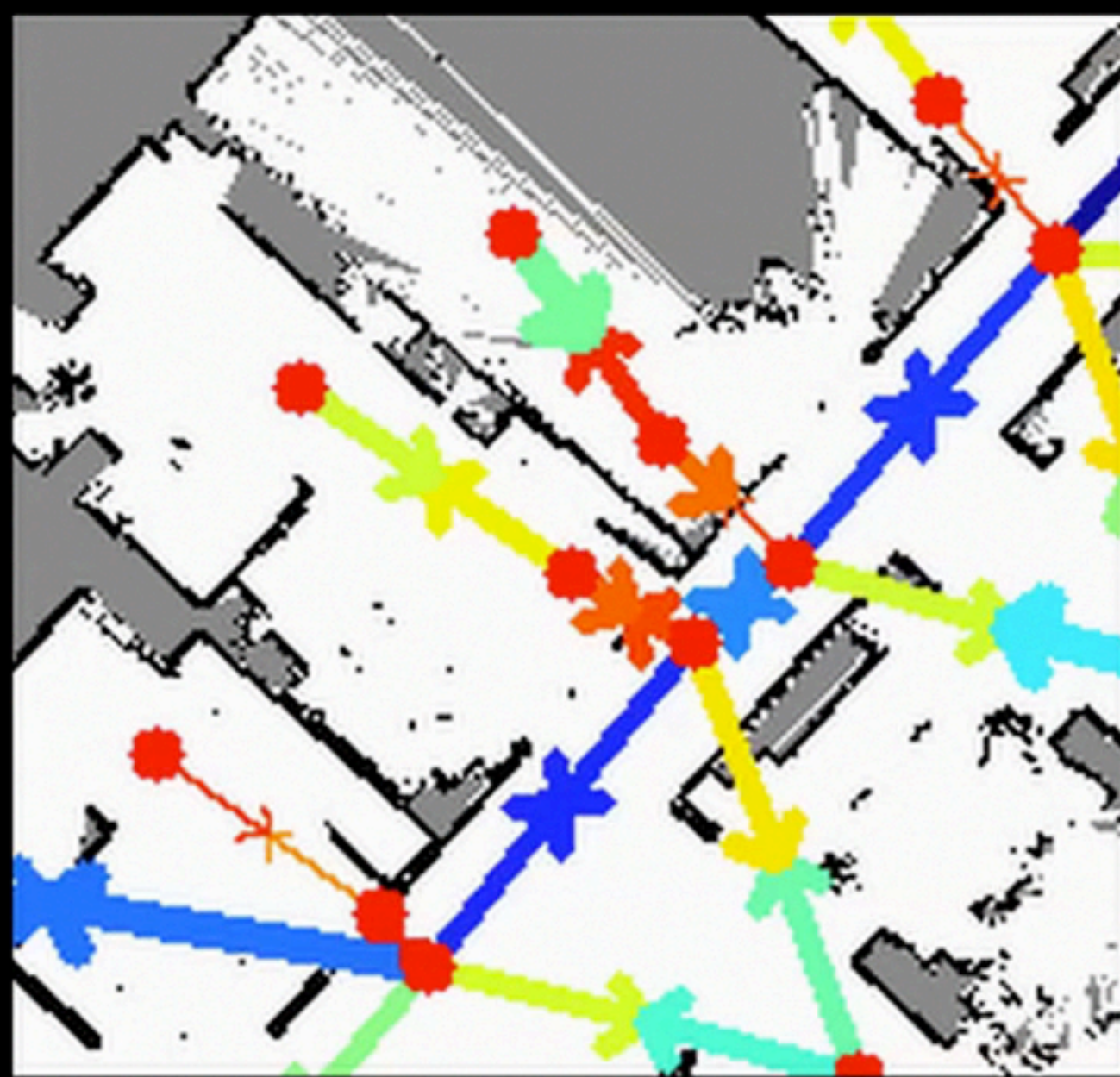


(f)



J. Pulido Fentanes, B. Lacerda, T. Krajník, 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

