

# Planning and Optimization

## E10. Merge-and-Shrink: Algorithm

Malte Helmert and Gabriele Röger

Universität Basel

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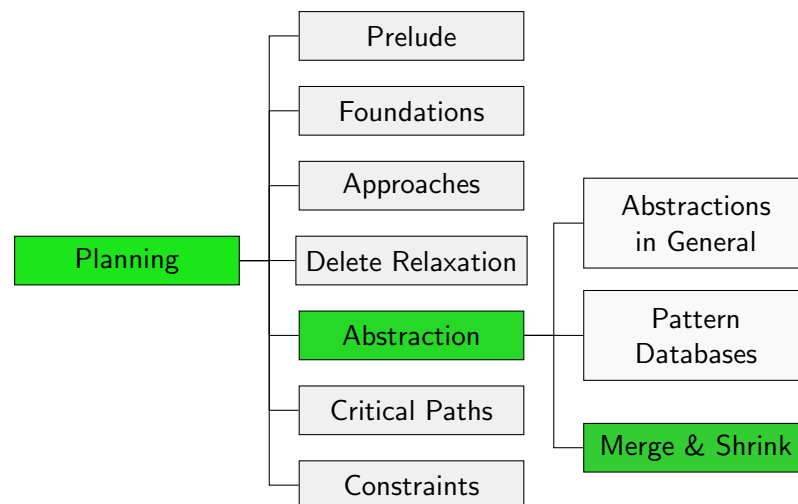
E10.1 Generic Algorithm

E10.2 Example

E10.3 Maintaining the Abstraction

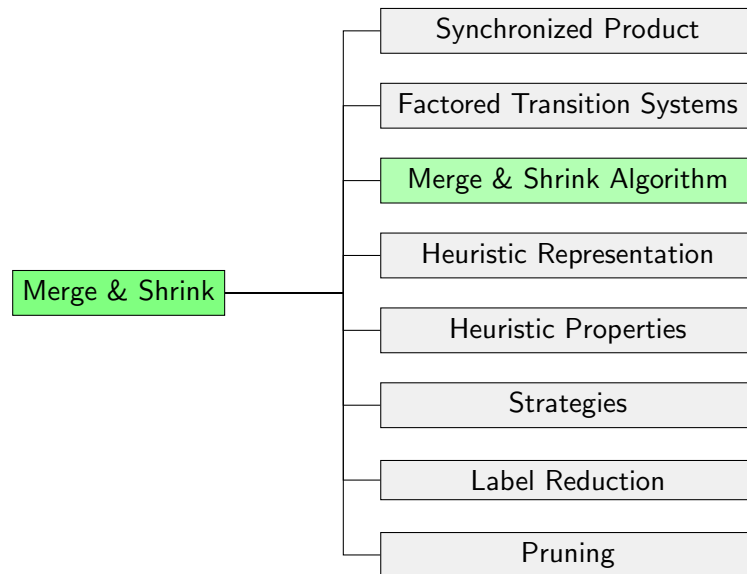
E10.4 Summary

## Content of this Course



## E10.1 Generic Algorithm

## Merge-and-Shrink



## Generic Merge-and-shrink Abstractions: Outline

Using the results of the previous chapter, we can develop a **generic abstraction computation procedure** that **takes all state variables into account**.

- ▶ **Initialization:** Compute the FTS consisting of all atomic projections.
- ▶ **Loop:** Repeatedly apply a transformation to the FTS.
  - ▶ **Merging:** Combine two factors by replacing them with their synchronized product.
  - ▶ **Shrinking:** If the factors are too large, make one of them smaller by abstracting it further (applying an arbitrary abstraction to it).
- ▶ **Termination:** Stop when only one factor is left.

The final factor is then used for an abstraction heuristic.

## Generic Algorithm Template

Generic Merge & Shrink Algorithm for planning task  $\Pi$ 

```

F := F(Π)
while |F| > 1:
  select type ∈ {merge, shrink}
  if type = merge:
    select T1, T2 ∈ F
    F := (F \ {T1, T2}) ∪ {T1 ⊗ T2}
  if type = shrink:
    select T ∈ F
    choose an abstraction mapping β on T
    F := (F \ {T}) ∪ {Tβ}
return the remaining factor Tα in F
  
```

In Ch. E12 and E13, we will include more transformation types (label reduction and pruning)

## Merge-and-Shrink Strategies

Choices to resolve to instantiate the template:

- ▶ When to merge, when to shrink?
  - ↪ **general strategy**
- ▶ Which abstractions to merge?
  - ↪ **merge strategy**
- ▶ Which abstraction to shrink, and how to shrink it (which  $\beta$ )?
  - ↪ **shrink strategy**

merge and shrink strategies ↪ Ch. E11/E12

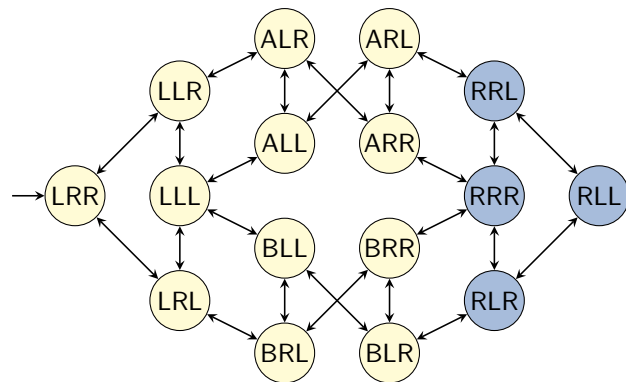
## General Strategy

A typical **general strategy**:

- ▶ define a **limit  $N$**  on the number of states allowed in each factor
- ▶ in each iteration, select two factors we would like to merge
- ▶ merge them if this does not exhaust the state number limit
- ▶ otherwise shrink one or both factors just enough to make a subsequent merge possible

## E10.2 Example

## Back to the Running Example

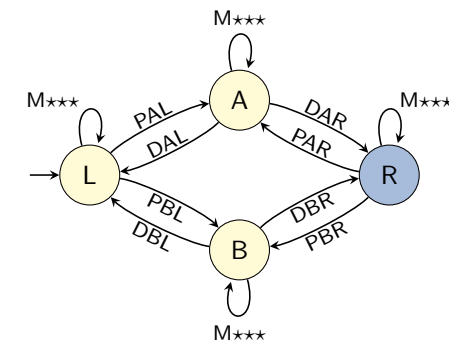


Logistics problem with one package, two trucks, two locations:

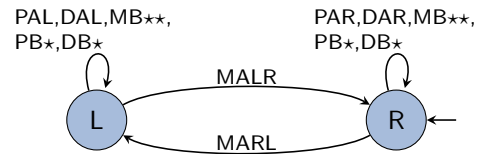
- ▶ state variable **package**:  $\{L, R, A, B\}$
- ▶ state variable **truck A**:  $\{L, R\}$
- ▶ state variable **truck B**:  $\{L, R\}$

## Initialization Step: Atomic Projection for Package

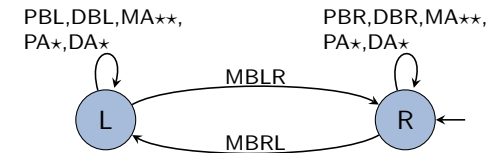
$\mathcal{T}^{\pi}_{\{\text{package}\}}$ :



## Initialization Step: Atomic Projection for Truck A

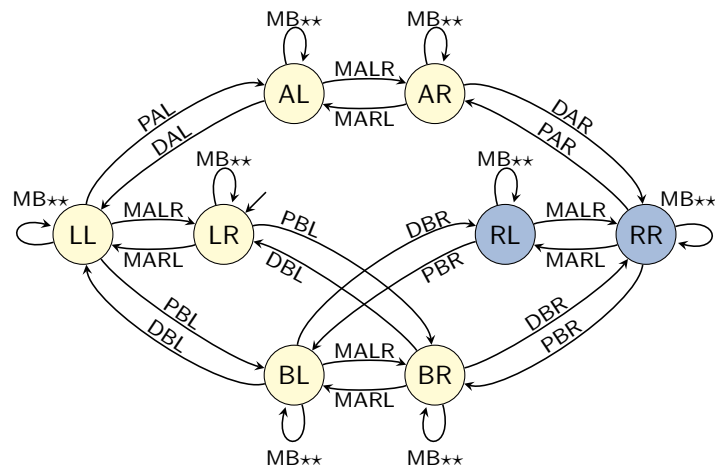
 $\mathcal{T}^\pi\{\text{truck A}\}:$ 


## Initialization Step: Atomic Projection for Truck B

 $\mathcal{T}^\pi\{\text{truck B}\}:$ 


current FTS:  $\{\mathcal{T}^\pi\{\text{package}\}, \mathcal{T}^\pi\{\text{truck A}\}, \mathcal{T}^\pi\{\text{truck B}\}\}$

## First Merge Step

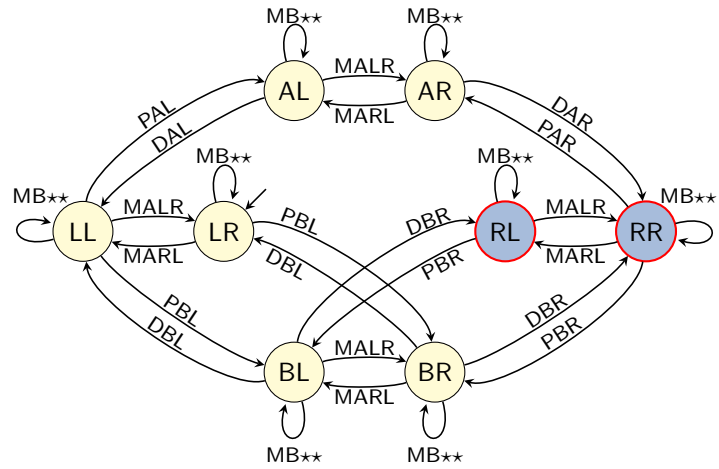
 $\mathcal{T}_1 := \mathcal{T}^\pi\{\text{package}\} \otimes \mathcal{T}^\pi\{\text{truck A}\}:$ 


current FTS:  $\{\mathcal{T}_1, \mathcal{T}^\pi\{\text{truck B}\}\}$

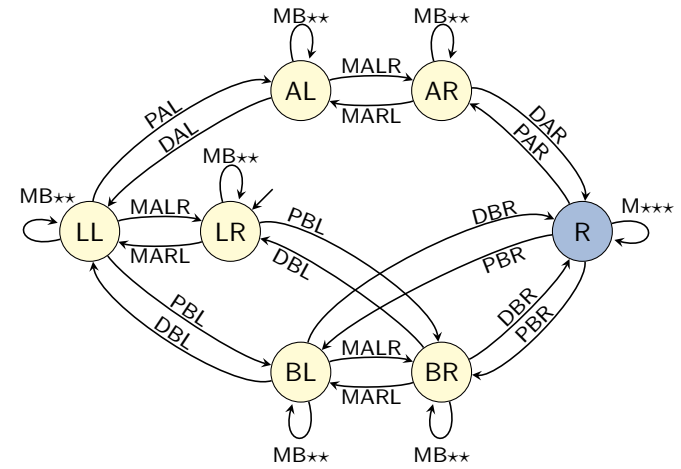
## Need to Shrink?

- ▶ With sufficient memory, we could now compute  $\mathcal{T}_1 \otimes \mathcal{T}^\pi\{\text{truck B}\}$  and recover the full transition system of the task.
- ▶ However, to illustrate the general idea, we assume that memory is too restricted: we may never create a factor with more than **8 states**.
- ▶ To make the product fit the bound, we shrink  $\mathcal{T}_1$  to 4 states. We can decide freely **how exactly** to abstract  $\mathcal{T}_1$ .
- ▶ In this example, we manually choose an abstraction that leads to a good result in the end. Making good shrinking decisions algorithmically is the job of the **shrink strategy**.

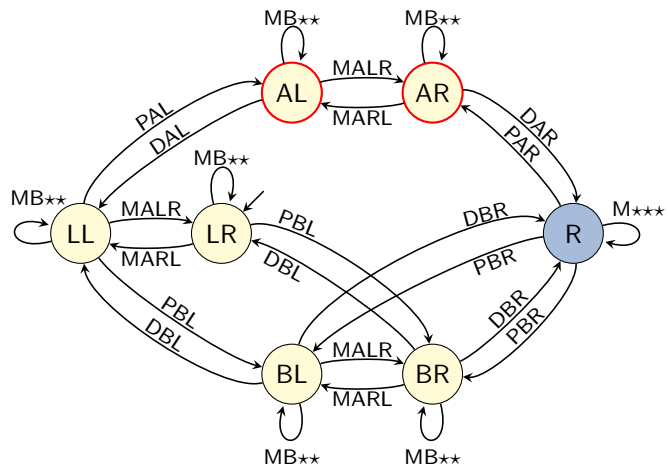
## First Shrink Step

 $\mathcal{T}_2 := \text{some abstraction of } \mathcal{T}_1$ 

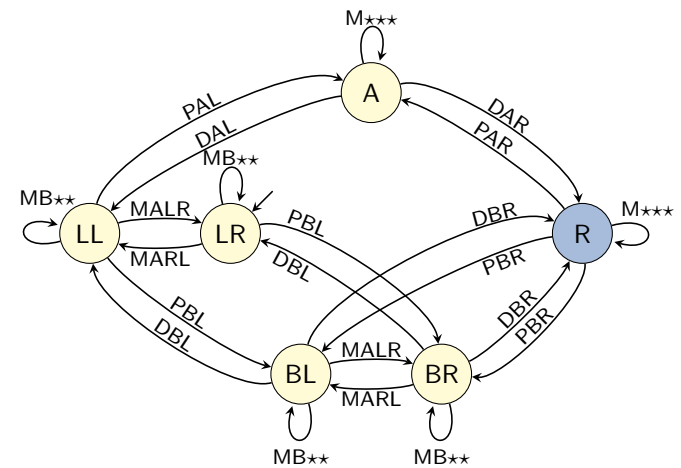
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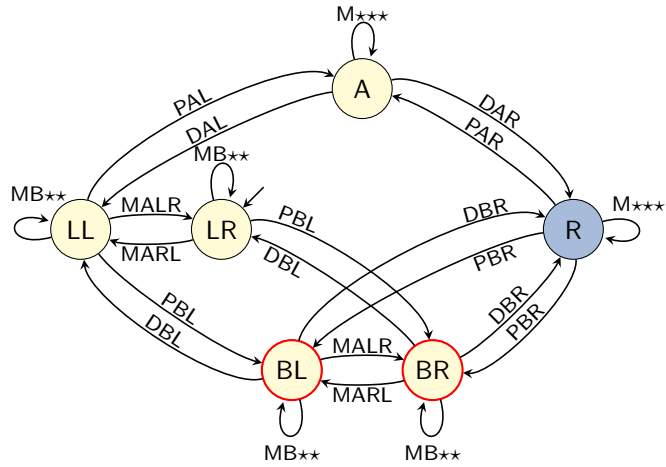
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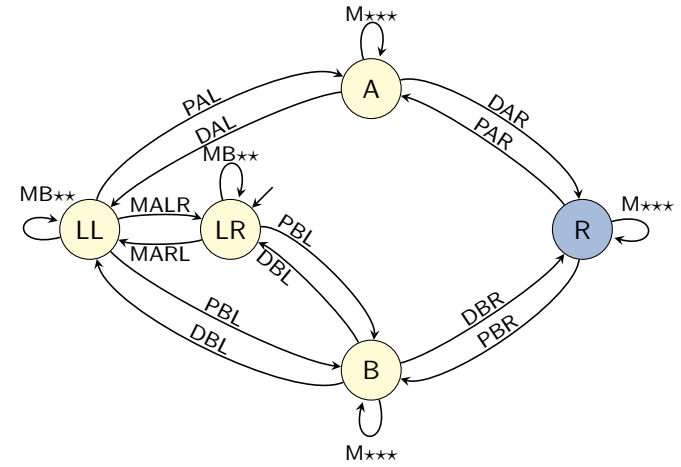
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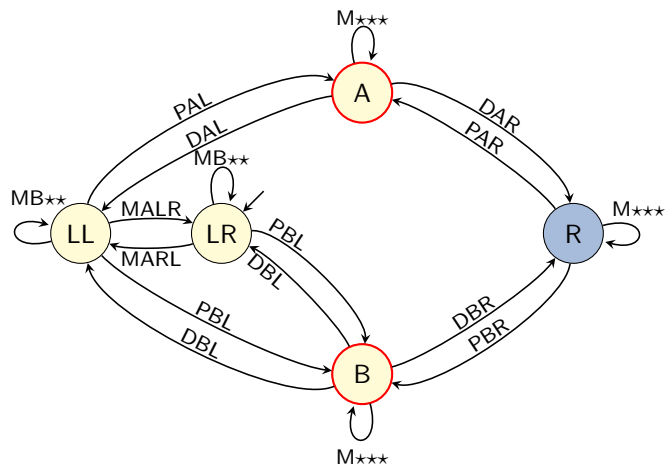
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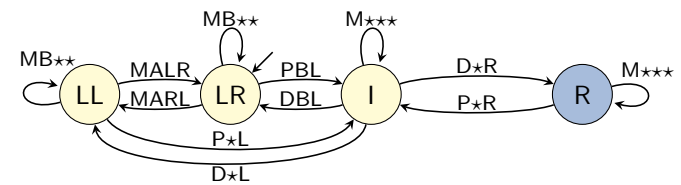
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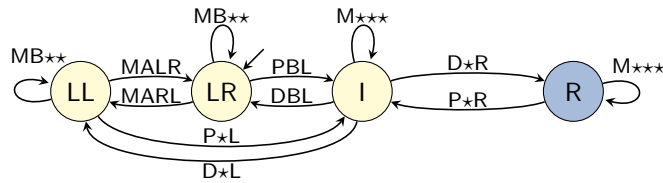
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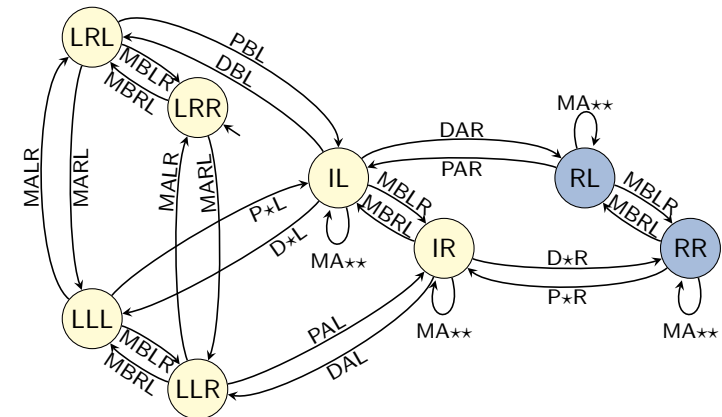
$\mathcal{T}_2 :=$  some abstraction of  $\mathcal{T}_1$



current FTS:  $\{\mathcal{T}_2, \mathcal{T}^\pi\{\text{truck B}\}\}$

## Second Merge Step

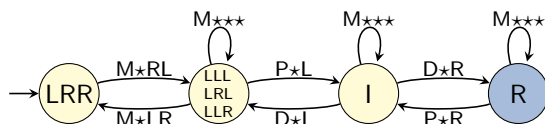
$\mathcal{T}_3 := \mathcal{T}_2 \otimes \mathcal{T}^\pi\{\text{truck B}\}$ :



current FTS:  $\{\mathcal{T}_3\}$

## Another Shrink Step?

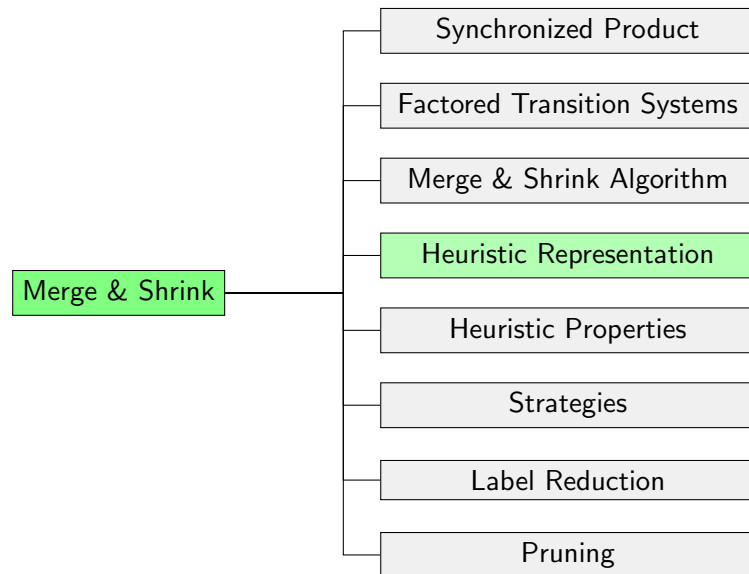
- ▶ At this point, merge-and-shrink construction stops. The distances in the final factor define the heuristic function.
- ▶ If there were further state variables to integrate, we would shrink again, e.g., leading to the following abstraction (again with four states):



- ▶ We get a heuristic value of 3 for the initial state, **better than any PDB heuristic** that is a proper abstraction.
- ▶ The example generalizes to arbitrarily many trucks, even if we stick to the fixed size limit of 8.

## E10.3 Maintaining the Abstraction

## Merge-and-Shrink



## Generic Algorithm Template

Generic Merge & Shrink Algorithm for planning task  $\Pi$

$F := F(\Pi)$

**while**  $|F| > 1$ :

**select**  $type \in \{\text{merge}, \text{shrink}\}$

**if**  $type = \text{merge}$ :

**select**  $\mathcal{T}_1, \mathcal{T}_2 \in F$

$F := (F \setminus \{\mathcal{T}_1, \mathcal{T}_2\}) \cup \{\mathcal{T}_1 \otimes \mathcal{T}_2\}$

**if**  $type = \text{shrink}$ :

**select**  $\mathcal{T} \in F$

**choose** an abstraction mapping  $\beta$  on  $\mathcal{T}$

$F := (F \setminus \{\mathcal{T}\}) \cup \{\mathcal{T}^\beta\}$

**return** the remaining factor  $\mathcal{T}^\alpha$  in  $F$

- ▶ The algorithm computes an abstract transition system.
- ▶ For the heuristic evaluation, we need an abstraction.
- ▶ How to maintain and represent the corresponding abstraction?

## The Need for Succinct Abstractions

- ▶ One major difficulty for non-PDB abstraction heuristics is to **succinctly represent the abstraction**.
- ▶ For pattern databases, this is easy because the abstractions – projections – are very **structured**.
- ▶ For less rigidly structured abstractions, we need another idea.

## How to Represent the Abstraction? (1)

**Idea:** the computation of the abstraction follows the sequence of product computations

- ▶ For the **atomic abstractions**  $\pi_{\{v\}}$ , we generate a **one-dimensional table** that denotes which value in  $\text{dom}(v)$  corresponds to which abstract state in  $\mathcal{T}^{\pi_{\{v\}}}$ .
- ▶ During the **merge** (product) step  $\mathcal{A} := \mathcal{A}_1 \otimes \mathcal{A}_2$ , we generate a **two-dimensional table** that denotes which pair of states of  $\mathcal{A}_1$  and  $\mathcal{A}_2$  corresponds to which state of  $\mathcal{A}$ .
- ▶ During the **shrink** (abstraction) steps, we make sure to keep the table in sync with the abstraction choices.



## How to Represent the Abstraction? (2)

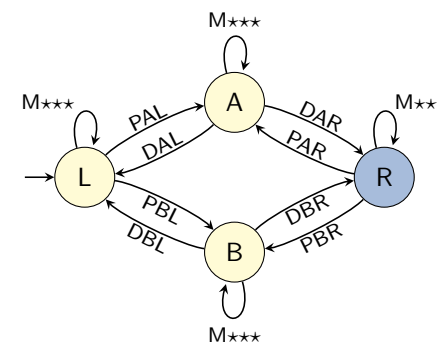
**Idea:** the computation of the abstraction mapping follows the sequence of product computations

- ▶ Once we have computed the final abstract transition system, we compute all **abstract goal distances** and store them in a **one-dimensional table**.
- ▶ At this point, we can **throw away** all the abstract transition systems – we just need to keep the tables.
- ▶ During **search**, we do a sequence of table lookups to navigate from the atomic abstraction states to the final abstract state and heuristic value  
 $\rightsquigarrow 2|V|$  lookups,  $O(|V|)$  time

Again, we illustrate the process with our running example.

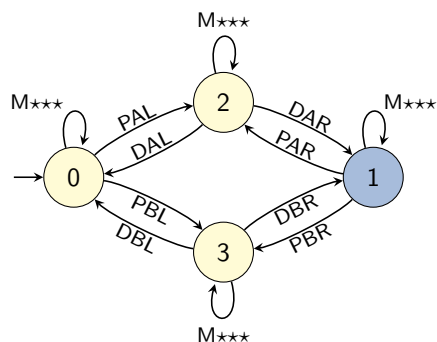
## Abstraction Example: Atomic Abstractions

Computing abstractions for the transition systems of atomic abstractions is simple. Just number the states (domain values) consecutively and generate a table of references to the states:



## Abstraction Example: Atomic Abstractions

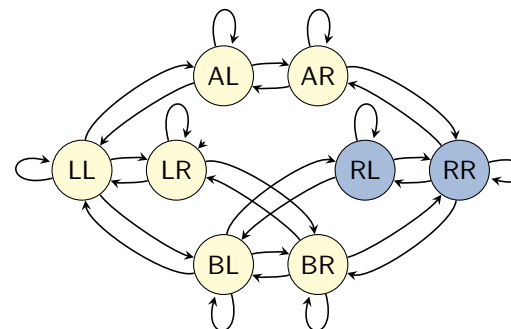
Computing abstractions for the transition systems of atomic abstractions is simple. Just number the states (domain values) consecutively and generate a table of references to the states:



L	R	A	B
0	1	2	3

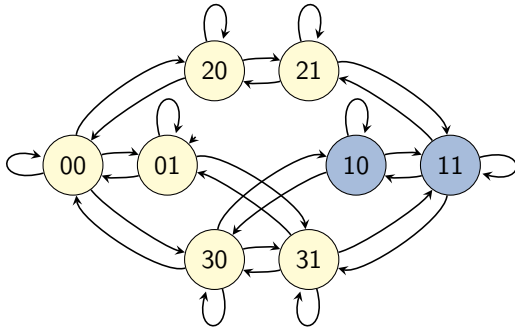
## Abstraction Example: Merge Step

For product transition systems  $\mathcal{A}_1 \otimes \mathcal{A}_2$ , we again number the product states consecutively and generate a table that links state pairs of  $\mathcal{A}_1$  and  $\mathcal{A}_2$  to states of  $\mathcal{A}$ :



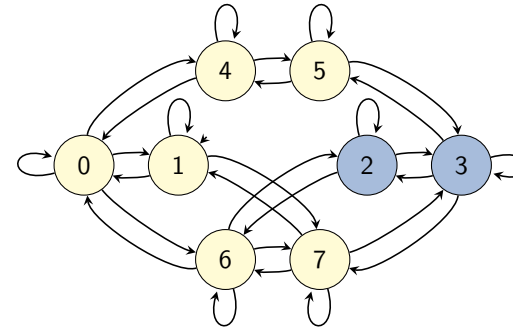
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	$s_2 = 0$	$s_2 = 1$
$s_1 = 0$	0	1
$s_1 = 1$	2	3
$s_1 = 2$	4	5
$s_1 = 3$	6	7

## Maintaining the Abstraction when Shrinking

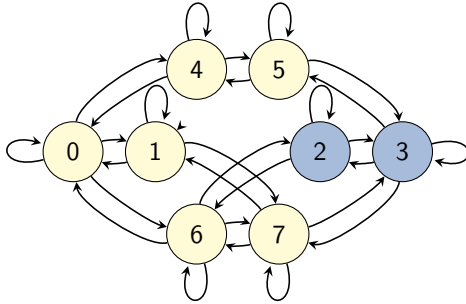
- ▶ The hard part in representing the abstraction is to keep it consistent when shrinking.
- ▶ In theory, this is easy to do:
  - ▶ When combining states  $i$  and  $j$ , arbitrarily use one of them (say  $i$ ) as the number of the new state.
  - ▶ Find all table entries in the table for this abstraction which map to the other state  $j$  and change them to  $i$ .
- ▶ However, doing a table scan each time two states are combined is very inefficient.
- ▶ Fortunately, there also is an efficient implementation which takes constant time per combination.

## Maintaining the Abstraction Efficiently

- ▶ Associate each abstract state with a linked list, representing **all table entries that map to this state**.
- ▶ Before starting the shrink operation, initialize the lists by scanning through the table, then **discard the table**.
- ▶ While shrinking, when combining  $i$  and  $j$ , **splice the list elements of  $j$  into the list elements of  $i$** .
  - ▶ For linked lists, this is a **constant-time operation**.
- ▶ Once shrinking is completed, renumber all abstract states so that there are no gaps in the numbering.
- ▶ Finally, regenerate the mapping table from the linked list information.

## Abstraction Example: Shrink Step

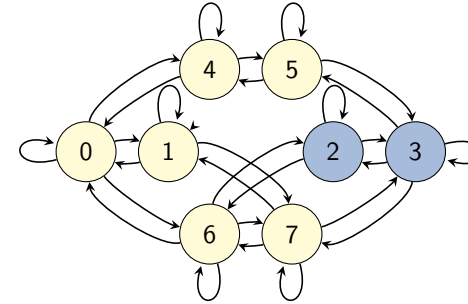
Representation before shrinking:



	$s_2 = 0$	$s_2 = 1$
$s_1 = 0$	0	1
$s_1 = 1$	2	3
$s_1 = 2$	4	5
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## Abstraction Example: Shrink Step

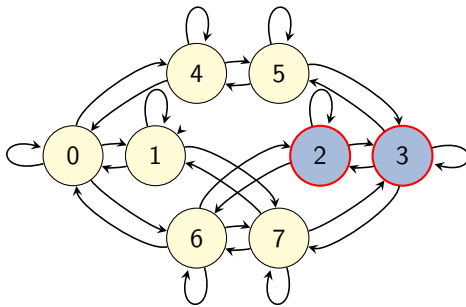
1. Convert table to linked lists and discard it.



$list_0 = \{(0, 0)\}$   
 $list_1 = \{(0, 1)\}$   
 $list_2 = \{(1, 0)\}$   
 $list_3 = \{(1, 1)\}$   
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 $list_5 = \{(2, 1)\}$   
 $list_6 = \{(3, 0)\}$   
 $list_7 = \{(3, 1)\}$

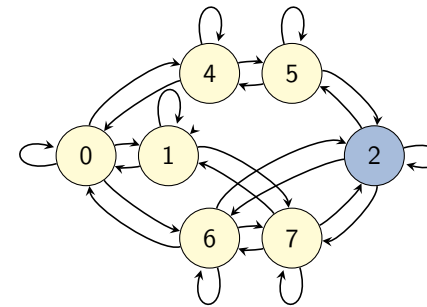
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## Abstraction Example: Shrink Step

2. When combining  $i$  and  $j$ , splice  $list_j$  into  $list_i$ .

$list_0 = \{(0, 0)\}$   
 $list_1 = \{(0, 1)\}$   
 $list_2 = \{(1, 0), (1, 1)\}$   
 $list_3 = \emptyset$   
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 $list_5 = \{(2, 1)\}$   
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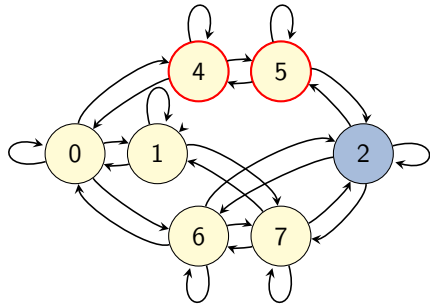
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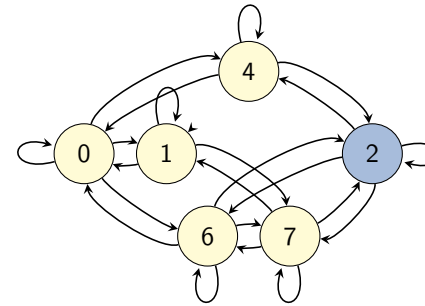
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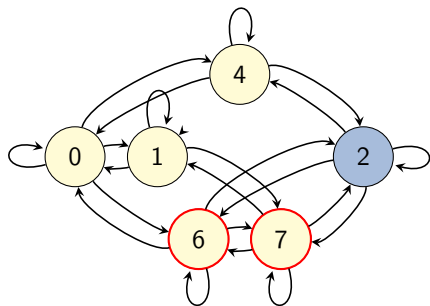
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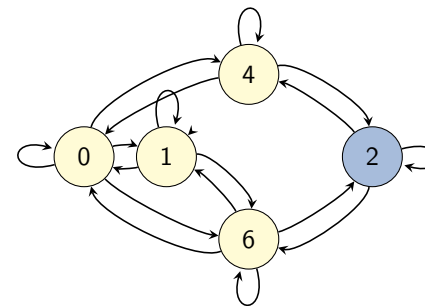
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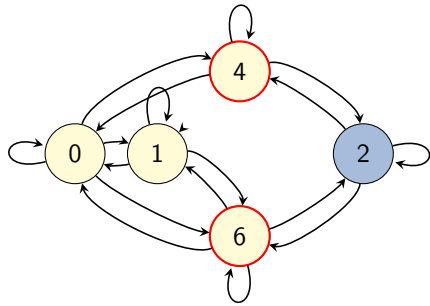
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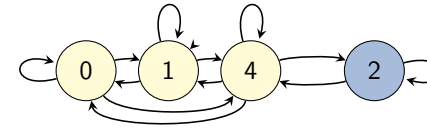
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## Abstraction Example: Shrink Step

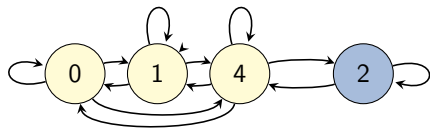
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## Abstraction Example: Shrink Step

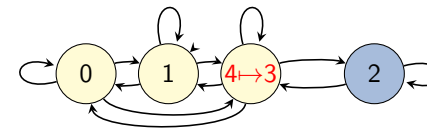
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## Abstraction Example: Shrink Step

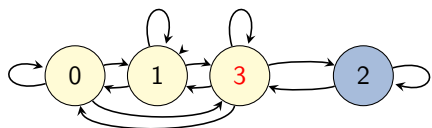
3. Renumber abstract states consecutively.



$list_0 = \{(0, 0)\}$   
 $list_1 = \{(0, 1)\}$   
 $list_2 = \{(1, 0), (1, 1)\}$   
 $list_3 = \emptyset$   
 $list_4 = \{(2, 0), (2, 1), (3, 0), (3, 1)\}$   
 $list_5 = \emptyset$   
 $list_6 = \emptyset$   
 $list_7 = \emptyset$

## Abstraction Example: Shrink Step

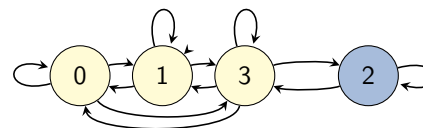
## 3. Renumber abstract states consecutively.



$list_0 = \{(0, 0)\}$   
 $list_1 = \{(0, 1)\}$   
 $list_2 = \{(1, 0), (1, 1)\}$   
 $list_3 = \{(2, 0), (2, 1), (3, 0), (3, 1)\}$   
 $list_4 = \emptyset$   
 $list_5 = \emptyset$   
 $list_6 = \emptyset$   
 $list_7 = \emptyset$

## Abstraction Example: Shrink Step

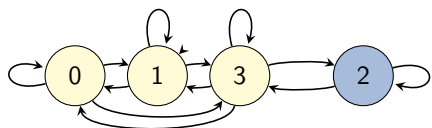
## 4. Regenerate the mapping table from the linked lists.



$list_0 = \{(0, 0)\}$   
 $list_1 = \{(0, 1)\}$   
 $list_2 = \{(1, 0), (1, 1)\}$   
 $list_3 = \{(2, 0), (2, 1), (3, 0), (3, 1)\}$   
 $list_4 = \emptyset$   
 $list_5 = \emptyset$   
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## Abstraction Example: Shrink Step

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 $list_5 = \emptyset$   
 $list_6 = \emptyset$   
 $list_7 = \emptyset$

	$s_2 = 0$	$s_2 = 1$
$s_1 = 0$	0	1
$s_1 = 1$	2	2
$s_1 = 2$	3	3
$s_1 = 3$	3	3

## The Final Heuristic Representation

At the end, our heuristic is represented by six tables:

- three one-dimensional tables for the atomic abstractions:

$T_{\text{package}}$	L	R	A	B	$T_{\text{truck A}}$	L	R	$T_{\text{truck B}}$	L	R
	0	1	2	3		0	1		0	1

- two tables for the two merge and subsequent shrink steps:

$T_{m\&s}^1$	$s_2 = 0$	$s_2 = 1$	$T_{m\&s}^2$	$s_2 = 0$	$s_2 = 1$
$s_1 = 0$	0	1	$s_1 = 0$	1	1
$s_1 = 1$	2	2	$s_1 = 1$	1	0
$s_1 = 2$	3	3	$s_1 = 2$	2	2
$s_1 = 3$	3	3	$s_1 = 3$	3	3

- one table with goal distances for the final transition system:

$T_h$	$s = 0$	$s = 1$	$s = 2$	$s = 3$
$h(s)$	3	2	0	1

Given a state  $s = \{\text{package} \mapsto L, \text{truck A} \mapsto L, \text{truck B} \mapsto R\}$ , its heuristic value is then looked up as:

- $h(s) = T_h[T_{m\&s}^2[T_{m\&s}^1[T_{\text{package}}[L], T_{\text{truck A}}[L]], T_{\text{truck B}}[R]]]$

## E10.4 Summary

## Summary (1)

- ▶ Merge-and-shrink abstractions are constructed by iteratively **transforming** the factored transition system of a planning task.
- ▶ **Merge** transformations combine two factors into their synchronized product.
- ▶ **Shrink** transformations reduce the size of a factor by abstracting it.
- ▶ Merge-and-shrink abstractions are **represented by a set of reference tables**, one for each atomic abstraction and one for each merge-and-shrink step.
- ▶ The heuristic representation uses an additional table for the goal distances in the final abstract transition system.

## Summary (2)

- ▶ Projections of  $SAS^+$  tasks correspond to merges of atomic factors.
- ▶ By also including shrinking, merge-and-shrink abstractions **generalize** projections: they can reflect **all** state variables, but in a potentially **lossy** way.