

Planning and Optimization

E13. Merge-and-Shrink: Pruning and Usage in Practise

Malte Helmert and Gabriele Röger

Universität Basel

November 22, 2023

Planning and Optimization

November 22, 2023 — E13. Merge-and-Shrink: Pruning and Usage in Practise

E13.1 Pruning

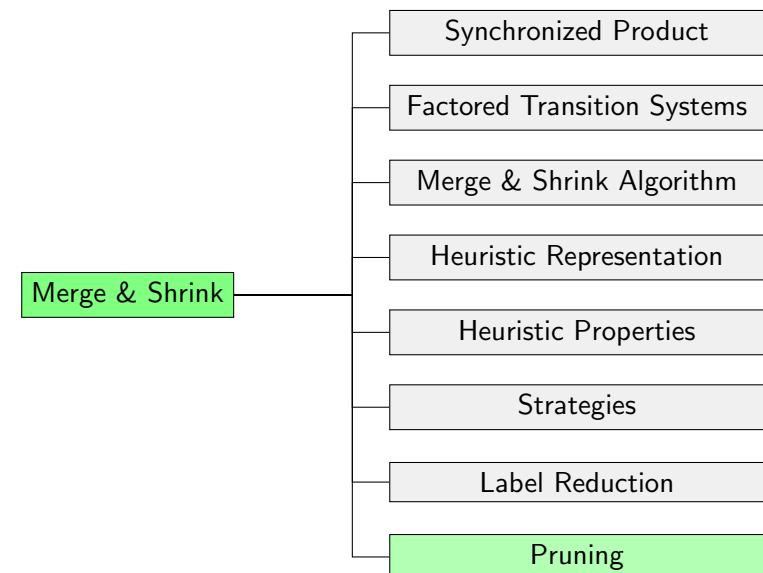
E13.2 Merge-and-Shrink in Practise

E13.3 Literature

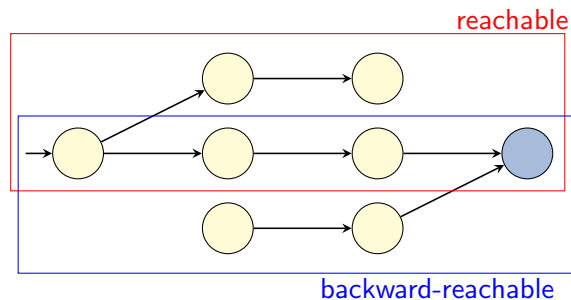
E13.4 Summary

E13.1 Pruning

Merge-and-Shrink



Alive States



- ▶ state s is reachable if we can reach it from the initial state
- ▶ state s is backward-reachable if we can reach the goal from s
- ▶ state s is alive if it is reachable and backward-reachable
→ only alive states can be traversed by a solution
- ▶ a state s is dead if it is not alive.

Pruning States (1)

- ▶ If in a factor, state s is dead/not backward-reachable then all states that “cover” s in a synchronized product are dead/not backward-reachable in the synchronized product.
- ▶ Removing such states and all adjacent transitions in a factor does not remove any solutions from the synchronized product.
- ▶ This pruning leads to states in the original state space for which the merge-and-shrink abstraction does not define an abstract state.
→ use heuristic estimate ∞

Pruning States (2)

- ▶ Keeping **exactly all backward-reachable states** we still obtain safe, consistent, goal-aware and admissible (with conservative transformations) or perfect heuristics (with exact transformations).
- ▶ Pruning unreachable, backward-reachable states can render the heuristic unsafe because pruned states lead to infinite estimates.
- ▶ However, all reachable states in the original state space will have admissible estimates, so we can use the heuristic like an admissible one in a forward state-space search such as A^* (but not in other contexts like such as orbit search).

We usually prune all dead states to keep the factors small.

E13.2 Merge-and-Shrink in Practise

Merge-and-Shrink

- ▶ Merge-and-Shrink is a general framework.
- ▶ The full framework also covers label reduction and pruning.
- ▶ For all transformations, we need to select a strategy.
merge, shrink, label reduction, pruning strategy
- ▶ The **general strategy** orchestrates the transformations.
How can this look like in practise?

Merge-and-Shrink in Fast Downward

Input: Factored transition system F , merge strategy MS, shrink strategy SS, prune strategy PS, label reduction strategy LRS, size limit $N \in \mathbb{N}$.

Output: Trans. system \mathcal{T} and mapping σ from states of $\otimes F$ to states of \mathcal{T} .

▷ Copy input factored transition system, compute Σ to represent the identity state mapping on $\otimes F'$, set λ to the identity label mapping.
 $\langle F', \Sigma, \lambda \rangle \leftarrow \langle F, \{\pi_{\mathcal{T}} \mid \mathcal{T} \in F'\}, \text{id} \rangle$

for $\mathcal{T} \in F$ **do**

▷ Prune atomic factor \mathcal{T} with PS.

$\langle F', \Sigma, \lambda \rangle \leftarrow \text{COMPOSETRANSFORMATION}(\text{PRUNE}(F', \mathcal{T}))$

end for

...

Merge-and-Shrink in Fast Downward (cont'd)

while $|F'| > 1$ **do**

▷ With MS, select two factors from F to be merged in this iteration.
 $\mathcal{T}_1, \mathcal{T}_2 \leftarrow \text{SELECT}(F')$

▷ With LRS, apply a label reduction to F .

$\langle F', \Sigma, \lambda \rangle \leftarrow \text{COMPOSETRANSFORMATION}(\text{LABELREDUCTION}(F'))$

▷ With SS, shrink \mathcal{T}_1 and \mathcal{T}_2 so that the size of their product respects N .

$\langle F', \Sigma, \lambda \rangle \leftarrow \text{COMPOSETRANSFORMATION}(\text{SHRINK}(F', \mathcal{T}_1, \mathcal{T}_2, N))$

▷ With LRS, apply a label reduction to F .

$\langle F', \Sigma, \lambda \rangle \leftarrow \text{COMPOSETRANSFORMATION}(\text{LABELREDUCTION}(F'))$

▷ Apply the merge transformation.

$\langle F', \Sigma, \lambda \rangle \leftarrow \text{COMPOSETRANSFORMATION}(\text{MERGE}(F', \mathcal{T}_1, \mathcal{T}_2))$

▷ With PS, prune the product factor \mathcal{T}^{\otimes} of \mathcal{T}_1 and \mathcal{T}_2 .

$\langle F', \Sigma, \lambda \rangle \leftarrow \text{COMPOSETRANSFORMATION}(\text{PRUNE}(F', \mathcal{T}^{\otimes}))$

end while

return single elements $\mathcal{T} \in F$ and $\sigma \in \Sigma$

Stopping Early



- ▶ Merge-and-shrink has significant precomputation time before we can start the search.
- ▶ We typically stop the algorithm after a preset time (e.g. half of the time that is overall available).
- ▶ The factored transition system then still contains several factors. Each of them induces an individual heuristic.
- ▶ We can combine them by taking the maximum or use a generalization of operator cost partitioning (cf. Ch. G7/8) to labels to obtain better estimates.
- ▶ Cost partitioning benefits from additional snapshots of factors from several iterations of merge-and-shrink.

State of the art: snapshots and saturated cost partitioning (Ch.G8)

E13.3 Literature

Literature (1)



References on merge-and-shrink abstractions:

-  Klaus Dräger, Bernd Finkbeiner and Andreas Podelski.
Directed Model Checking with Distance-Preserving Abstractions.
Proc. SPIN 2006, pp. 19–34, 2006.
Introduces merge-and-shrink abstractions (for model checking) and DFP merging strategy.
-  Malte Helmert, Patrik Haslum and Jörg Hoffmann.
Flexible Abstraction Heuristics for Optimal Sequential Planning.
Proc. ICAPS 2007, pp. 176–183, 2007.
Introduces merge-and-shrink abstractions for planning.



Literature (2)

-  Raz Nissim, Jörg Hoffmann and Malte Helmert.
Computing Perfect Heuristics in Polynomial Time:
On Bisimulation and Merge-and-Shrink Abstractions
in Optimal Planning.
Proc. IJCAI 2011, pp. 1983–1990, 2011.
Introduces bisimulation-based shrinking.
-  Malte Helmert, Patrik Haslum, Jörg Hoffmann
and Raz Nissim.
Merge-and-Shrink Abstraction: A Method
for Generating Lower Bounds in Factored State Spaces.
Journal of the ACM 61 (3), pp. 16:1–63, 2014.
Detailed journal version of the previous two publications.


Literature (3)

-  Silvan Sievers, Martin Wehrle and Malte Helmert.
Generalized Label Reduction for Merge-and-Shrink Heuristics.
Proc. AAAI 2014, pp. 2358–2366, 2014.
Introduces modern version of label reduction.
(There was a more complicated version before.)
-  Gaojian Fan, Martin Müller and Robert Holte.
Non-linear merging strategies for merge-and-shrink
based on variable interactions.
Proc. SoCS 2014, pp. 53–61, 2014.
Introduces UMC and MIASM merging strategies

Literature (4)

-  Malte Helmert, Gabriele Röger and Silvan Sievers.
On the Expressive Power of Non-Linear Merge-and-Shrink Representations.
Proc. ICAPS 2015, pp. 106–1014, 2015.
Shows that **linear merging can require a super-polynomial blow-up** in representation size.
-  Silvan Sievers and Malte Helmert.
Merge-and-Shrink: A Compositional Theory of Transformations of Factored Transition Systems.
JAIR 71, pp. 781–883, 2021.
Detailed theoretical analysis of task transformations as **sequence of transformations**.

Literature (5)

-  Silvan Sievers, Florian Pommerening, Thomas Keller and Malte Helmert.
Cost-Partitioned Merge-and-Shrink Heuristics for Optimal Classical Planning.
Proc. IJCAI 2020, pp. 4152–4160, 2020.
Extends **saturated cost partitioning** to merge-and-shrink.

E13.4 Summary

Summary

- ▶ **Pruning** is a transformation that is used to keep the size of the factors small. It depends on the intended application how aggressive the pruning can be.
- ▶ In practise, it is beneficial to set a **time limit** for merge-and-shrink. The factors can be considered as individual admissible heuristics.