

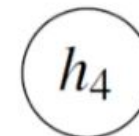
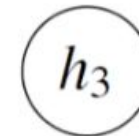
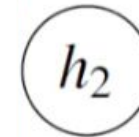
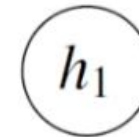
Learning Domain-Independent Policies for Open List Selection

André Biedenkapp, David Speck, Silvan Sievers,
Frank Hutter, Marius Lindauer, Jendrik Seipp

What state should
I expand next?

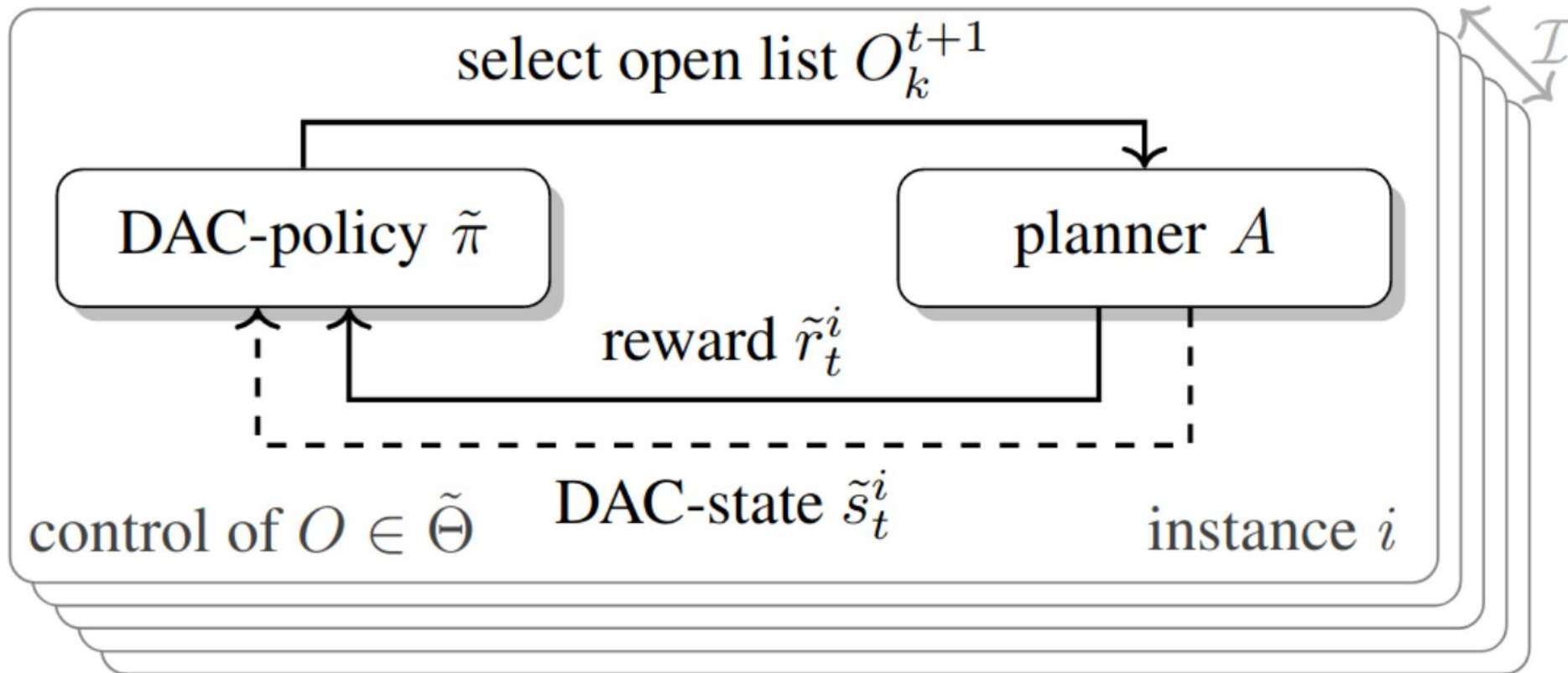


Planner

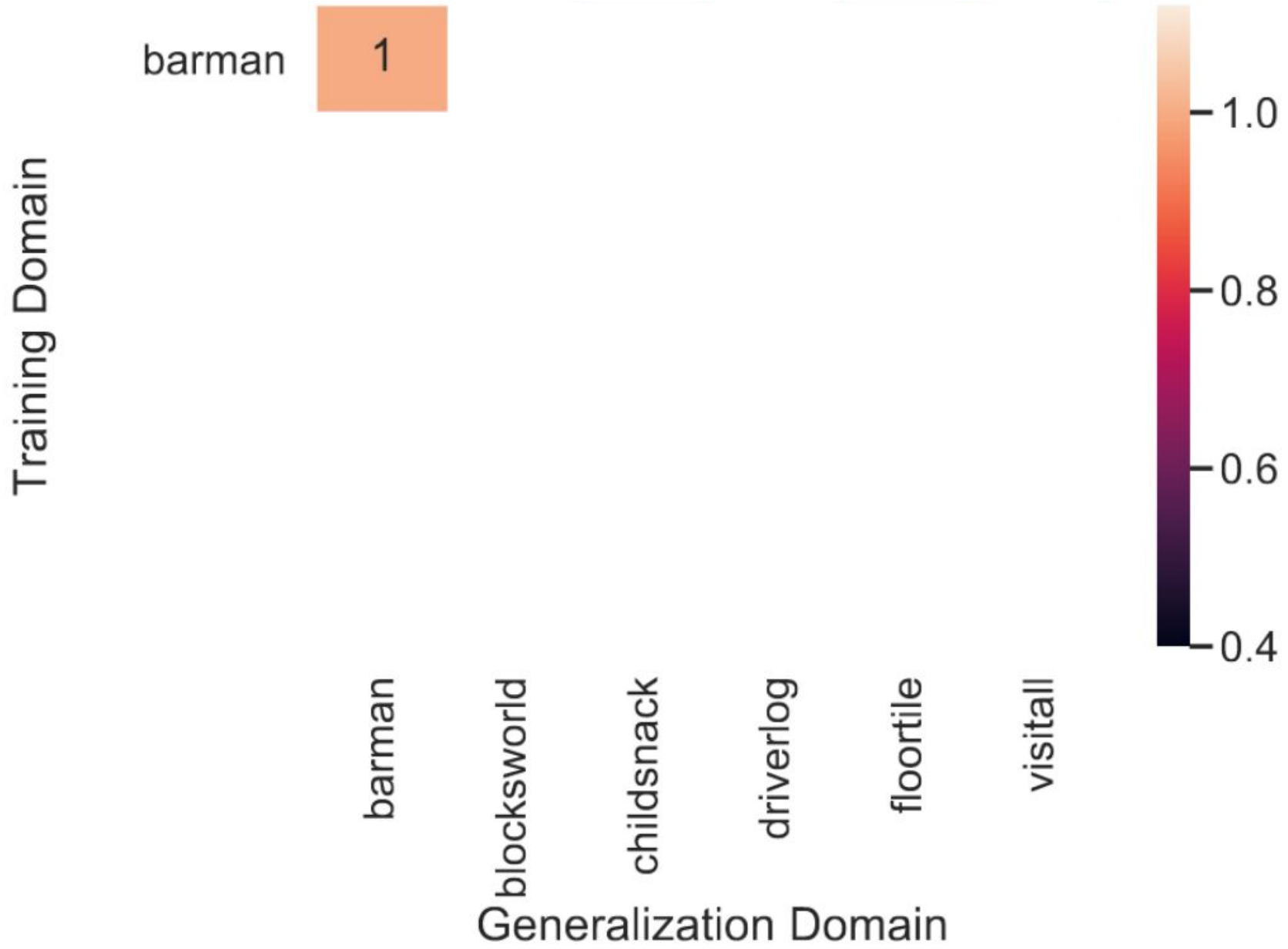


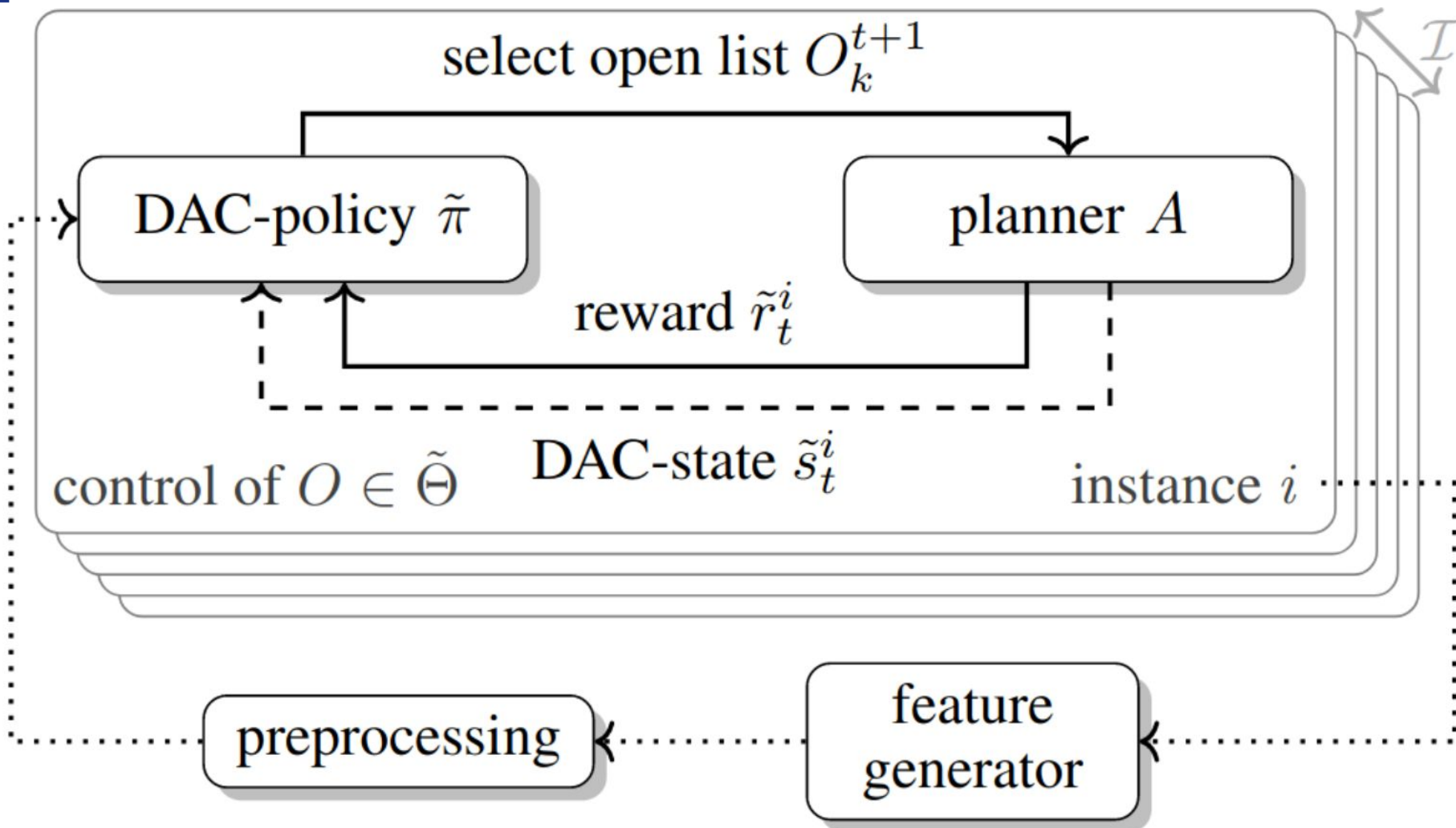
Satisficing planning with multiple heuristics

- Search for a good plan
- Inadmissible heuristics are difficult to combine
- States evaluated with each heuristic
- One separate open list for each heuristic



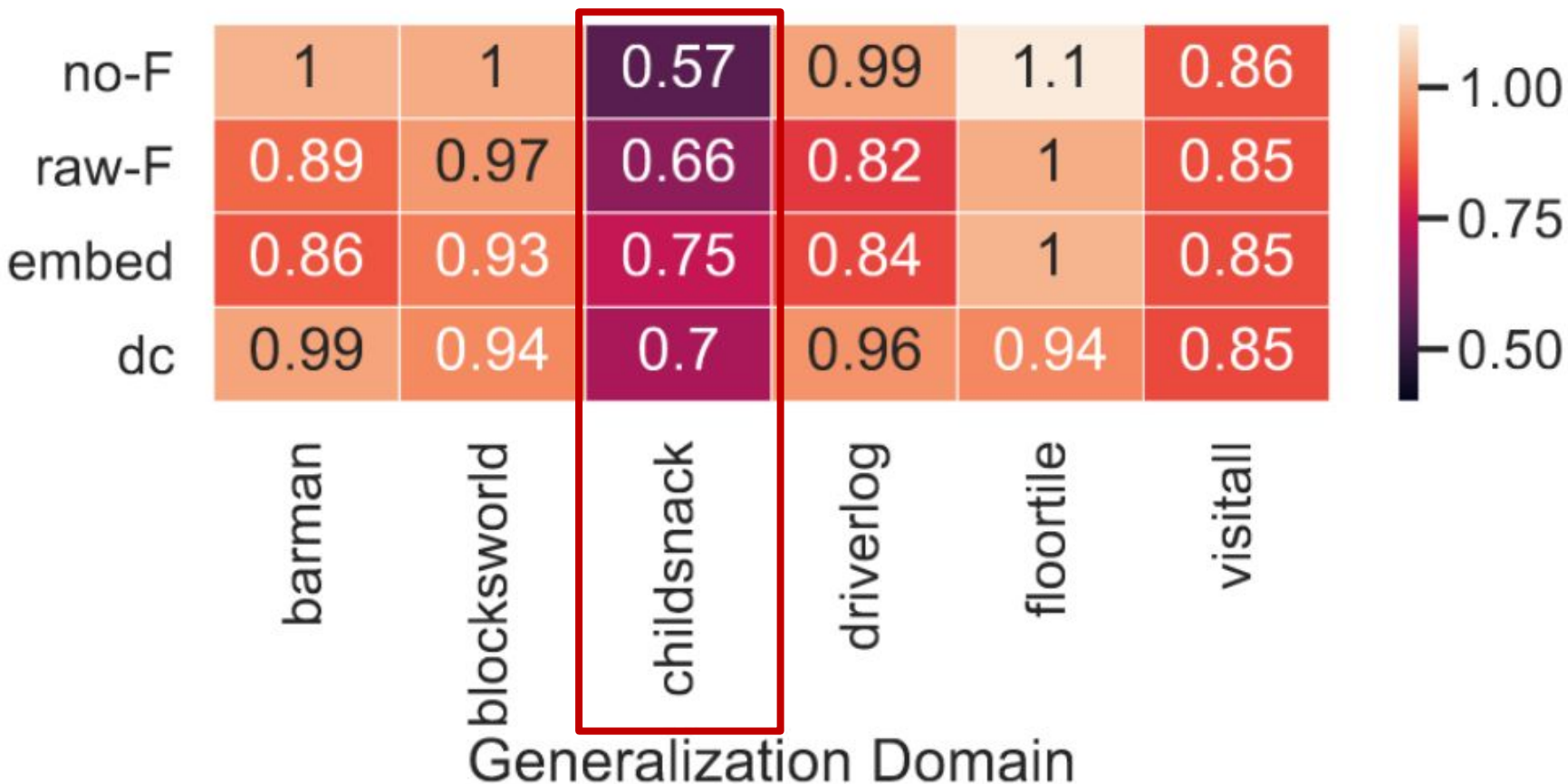
Speck, D.; Biedenkapp, A.; Hutter, F.; Mattmüller, R.; and Lindauer, M. 2021. Learning Heuristic Selection with Dynamic Algorithm Configuration. In Proc. ICAPS 2021, 597–605.





- Using Only Dynamic Features (no-F)
- Concatenating Instance and Dynamic Features (raw-F)
- Learning Separate Representations (embed)
- Decoupling Instance and Dynamic Features (dc)

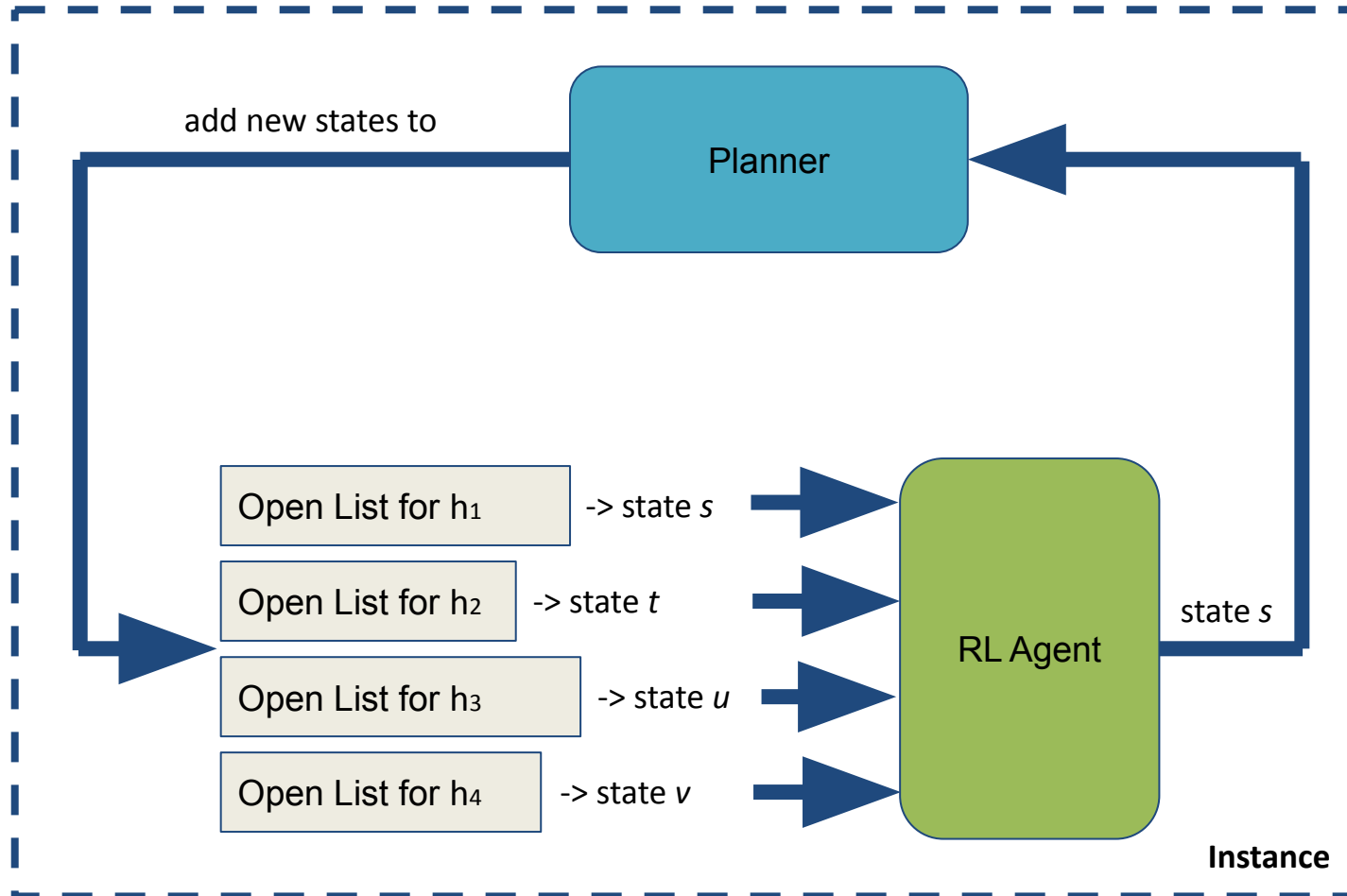
Training Method





Looking forward to meeting you at the poster!

Goal: Select configuration for the problem
at hand and adapt while planning





Configuration -> Planner -> Performance

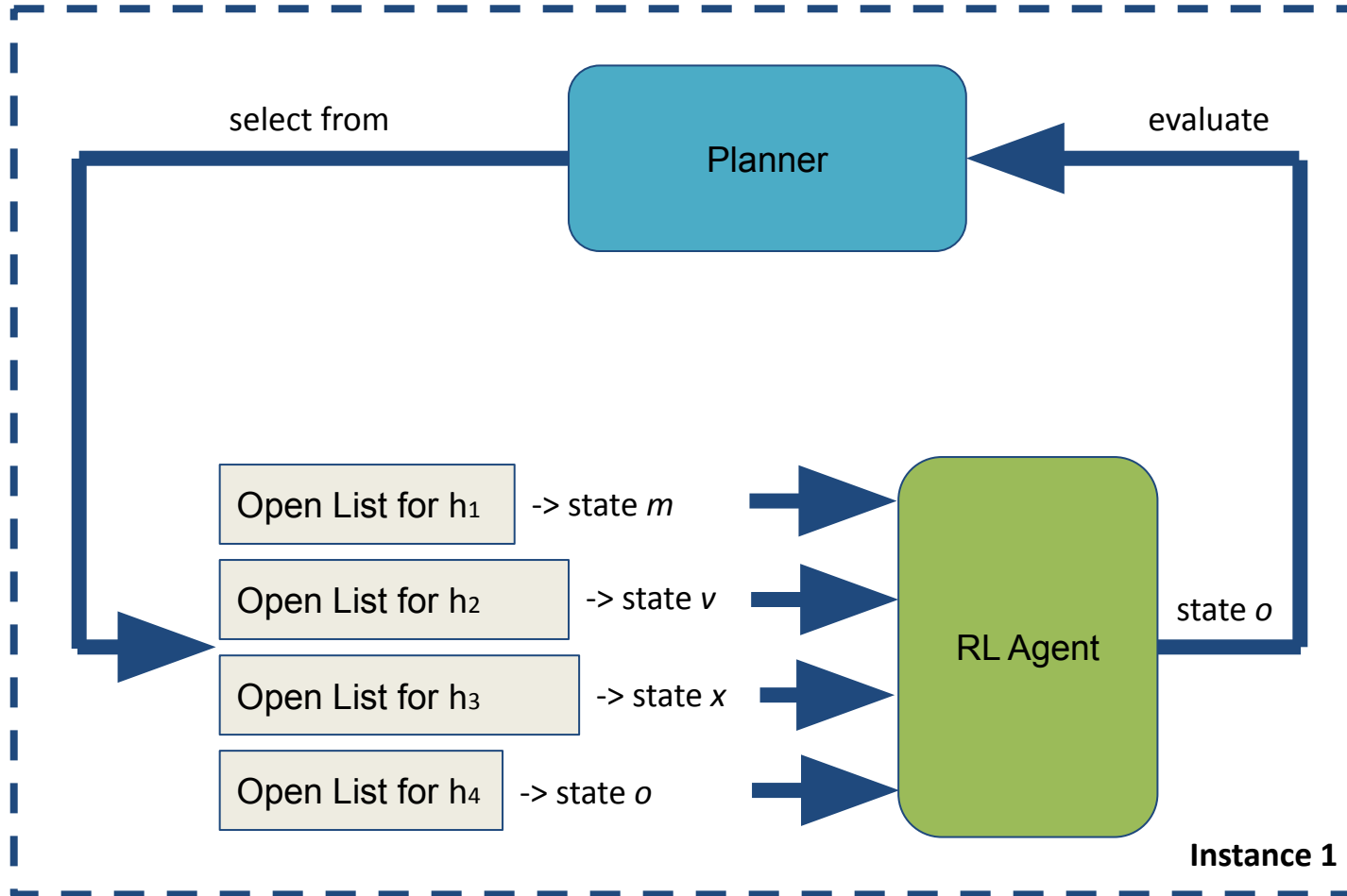
Goal: Find single best configuration

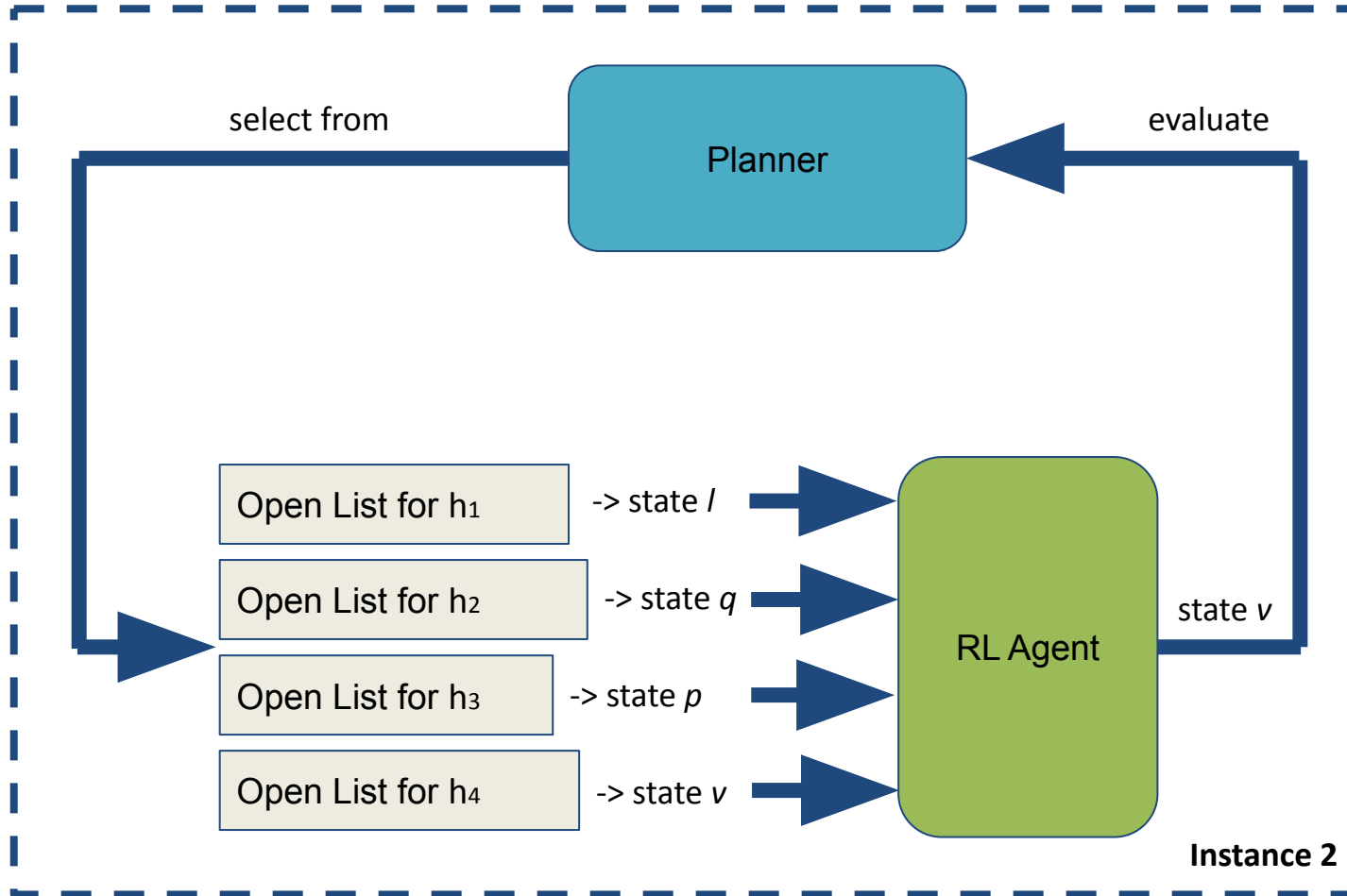
Configuration -> Planner -> Performance

Goal: Select best configuration for the problem
at hand



Configuration
+
Instance Features -> Planner -> Performance





1. We generalize previous DAC approaches to learn domain-independent open list selection policies
2. We present novel ways to learn from instance specific features jointly with dynamic features
3. Our learned policies reduce the required number of node expansions on several domains
4. We use DAC as a tool to gain insights on why LAMA's policy has such strong performance