

An Admissible HTN Planning Heuristic

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- How to solve these problems based on well-informed domain-independent heuristics?
- How to provide optimality guarantees?



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We base our formalization upon *hybrid planning*, which fuses

- hierarchical task network (HTN) planning with
- partial order causal link (POCL) planning.



Related Work:



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- HTN planners using control knowledge (e.g., SHOP2).



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- HTN planners using control knowledge (e.g., SHOP2).
- Hierarchical heuristic search planners for different problem classes (e.g., hybrid, HGN, GTN, HTN+preferences).



A hybrid planning problem is given in terms of:

- a planning domain stating the tasks (primitive and abstract) and methods (how to refine an abstract action?), and
- an initial partial plan, which needs to be refined into a solution.



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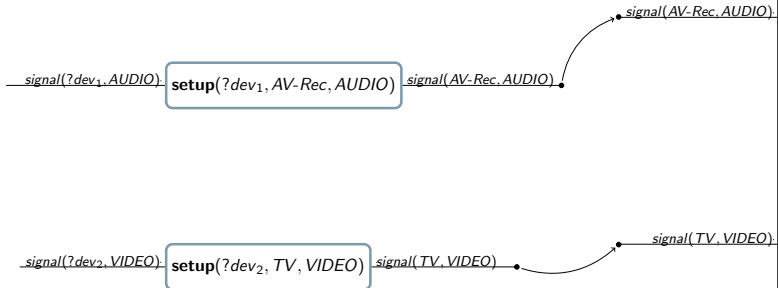
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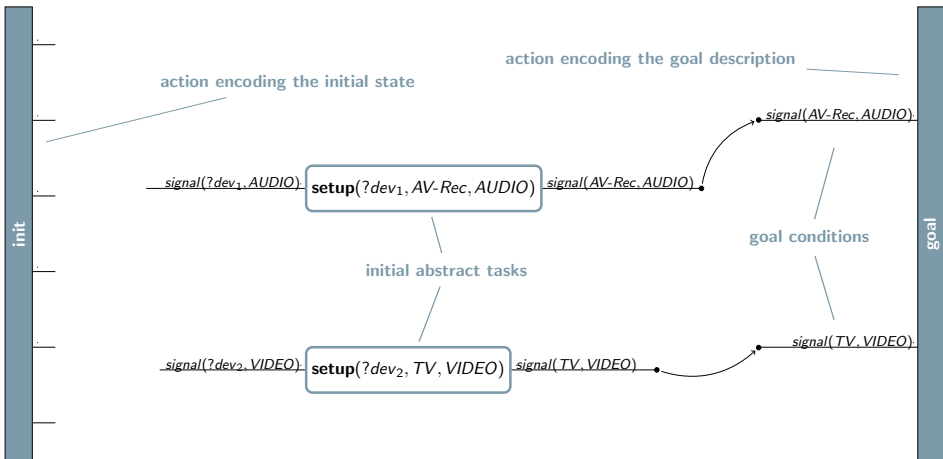
A solution is a partially ordered plan, such that:

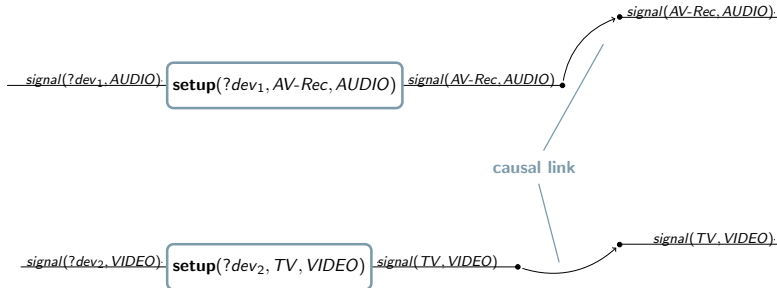
- it is a primitive refinement of the initial partial plan,
- all its linearizations are executable in the initial state, and
- all its linearization satisfy the goal description.

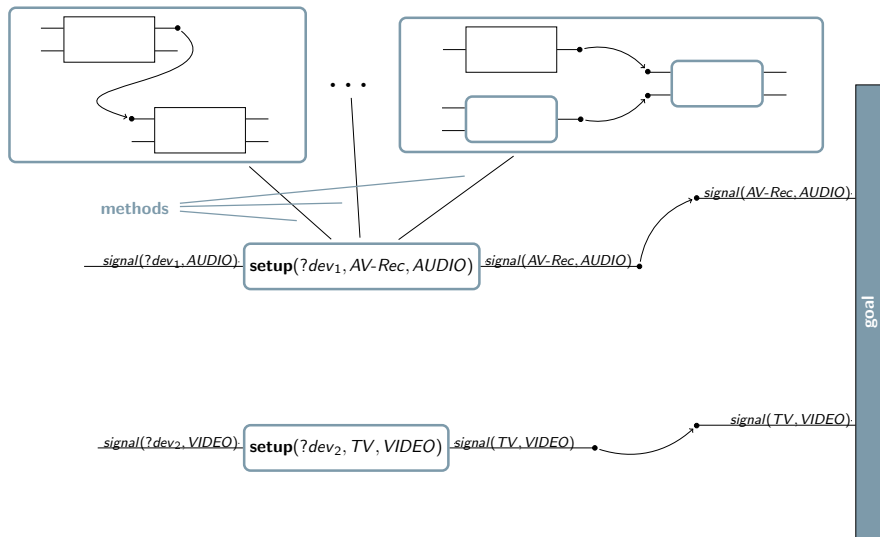
↪ *see example given next.*

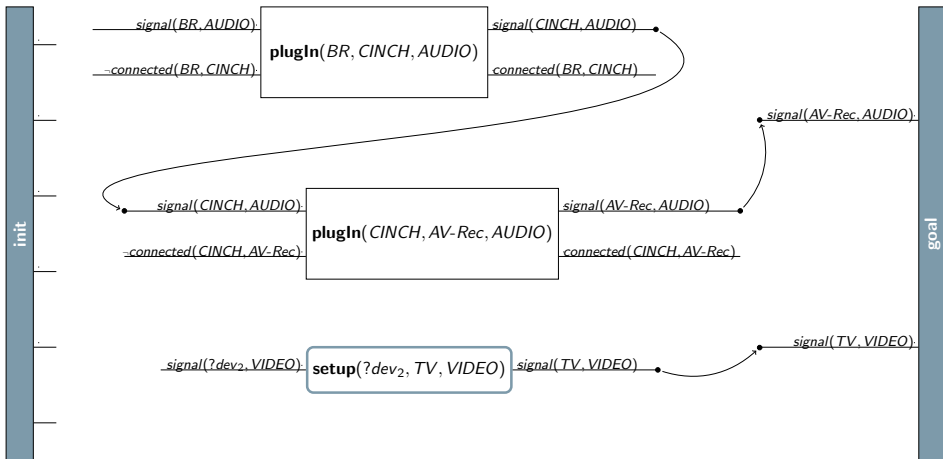


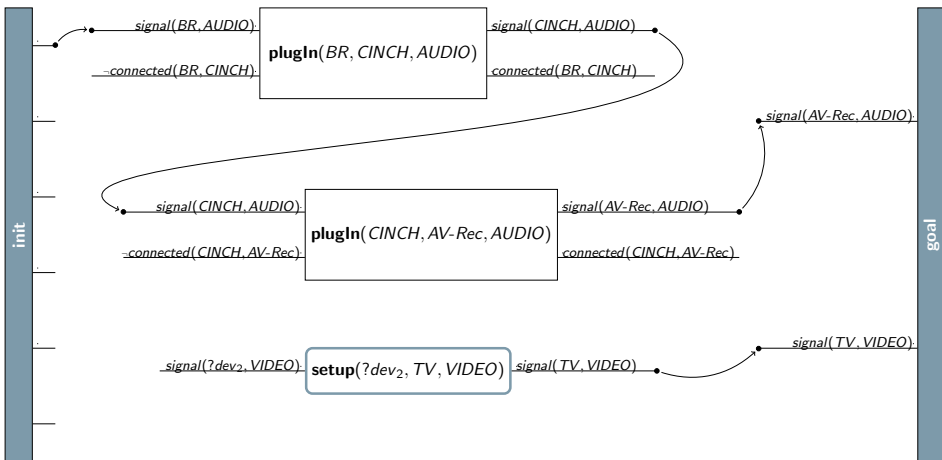


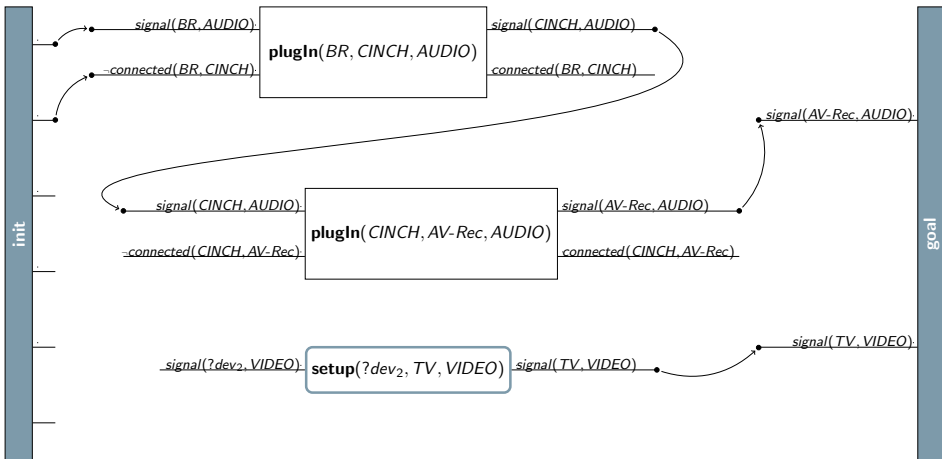


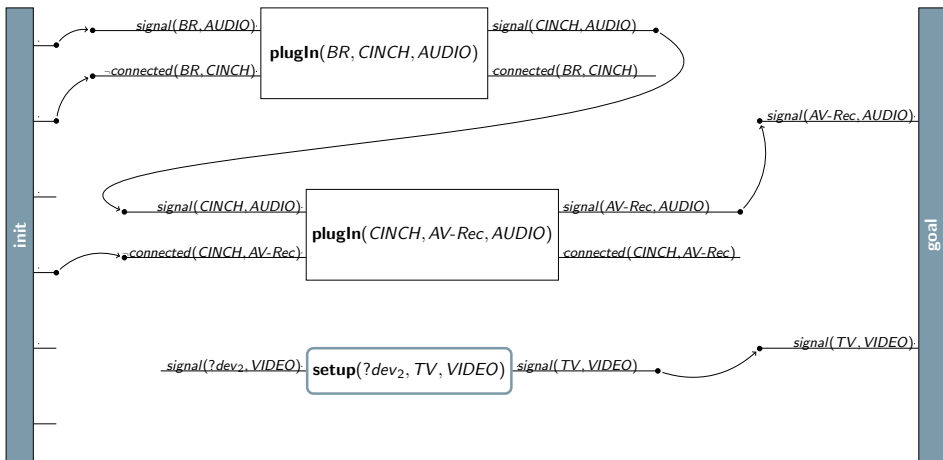


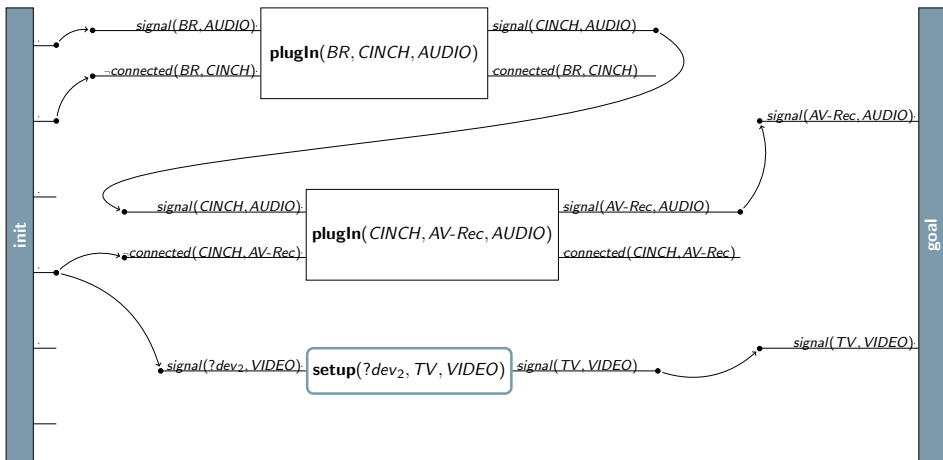


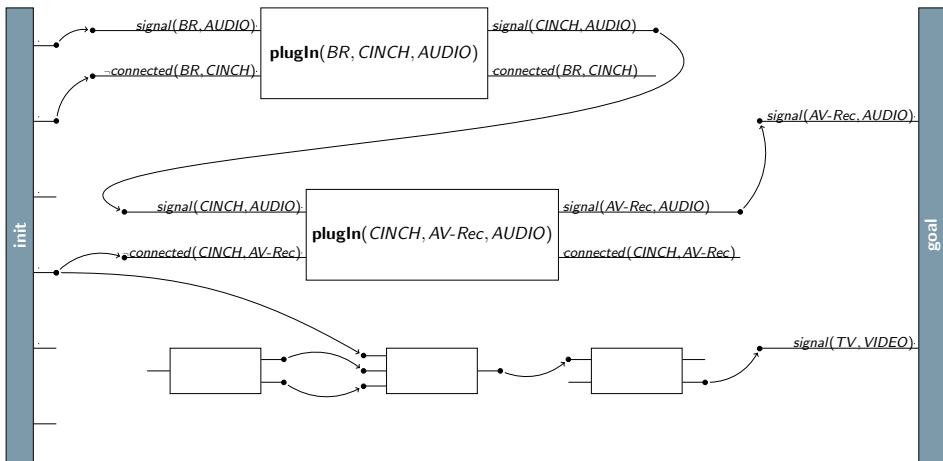


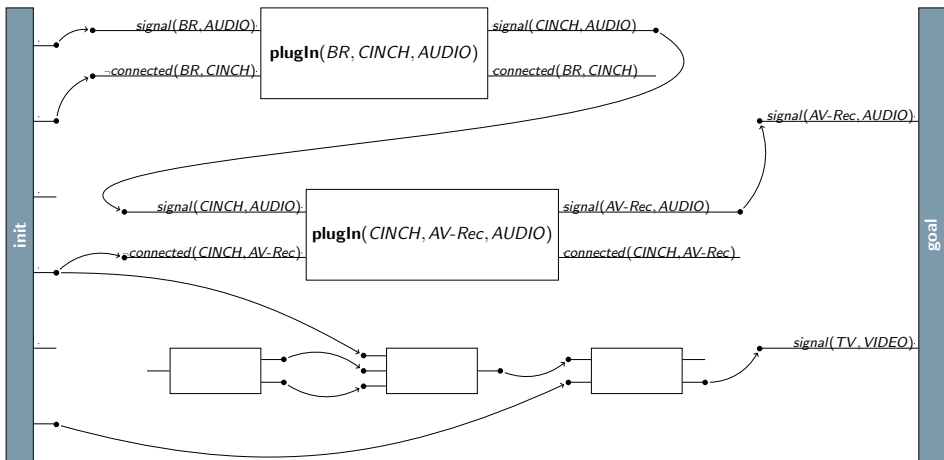


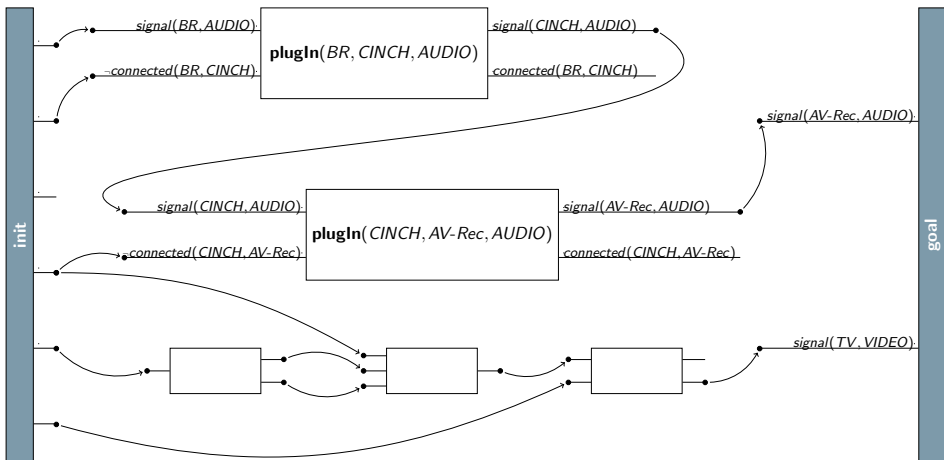












How to come up with heuristic functions?

- Both HTN and hybrid problems are *undecidable*, we hence need severe problem relaxations that them *decidable* and *tractable*.
- \rightsquigarrow We allow task insertion.
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How does the heuristic work?

Step 1:

- Build a so-called task-decomposition graph (TDG), which represents the hierarchical problem structure.

Step 2:

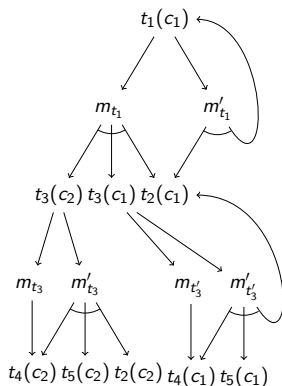
- Calculate heuristic estimates for all tasks in that TDG.

Step 3:

- For a given plan, retrieve the TDG's estimates for all abstract tasks in that plan.



Task decomposition Graph (TDG), example:



A TDG is a ground representation of the task hierarchy.

It is a (possibly cyclic) bipartite graph $\langle V_T, V_M, E_{T \rightarrow M}, E_{M \rightarrow T} \rangle$ consisting of:

- task vertices V_T (abstract task vertices are OR nodes)
- method vertices V_M (which are AND nodes)



Let $\langle V_T, V_M, E_{T \rightarrow M}, E_{M \rightarrow T} \rangle$ be a ground TDG. Then, we can calculate a heuristics by exploiting its AND/OR structure:

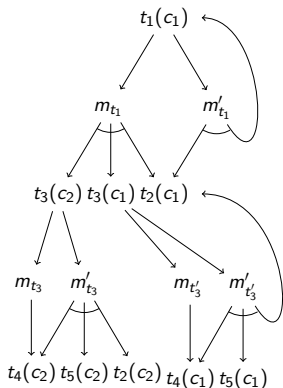
$$h_T(v_t) := \begin{cases} \text{cost}(v_t) & \text{if } v_t \text{ is primitive} \\ \min_{(v_t, v_m) \in E_{T \rightarrow M}} h_M(v_m) & \text{else} \end{cases}$$

For a method vertex $v_m = \langle PS, \prec, CL, VC \rangle$, we set:

$$h_M(v_m) := \sum_{(v_m, v_t) \in E_{M \rightarrow T}} h_T(v_t)$$



Heuristic computation, example:



$$h(m'_{t_3}) = \text{cost}(t_4(c_1))$$

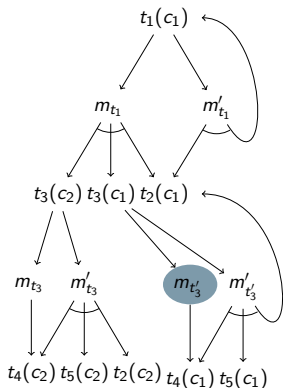
$$h(m'_{t'_3}) = \text{cost}(t_4(c_1)) \\ + \text{cost}(t_5(c_1)) \\ + \text{cost}(t_2(c_1))$$

$$h(t_3(c_1)) = \min\{h(m'_{t_3}), h(m'_{t'_3})\}$$

...



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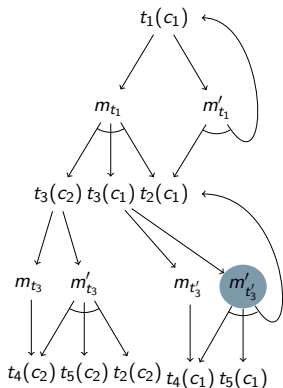
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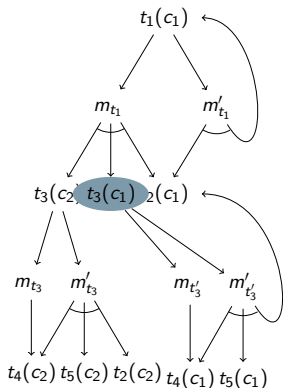
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Improving Heuristic estimates:

- So far, the TDG is computed only *once*.
 \rightsquigarrow It is a preprocessing heuristic.
- Re-computing the TDG during planning is expensive, but increases heuristic estimates. (We show in which situations TDG recomputation might improve these estimates to avoid unnecessary recomputations).



Space and time increase and decrease due to TGD recomputation:

Qualitative:

Strategy	space	time	space	time	space	time	space	time	
	< = >	< = >	< = >	< = >	< = >	< = >	< = >	< = >	
	UM-Translog		SmartPhone		Satellite		Woodworking		
A^*	TDG _m	2 19 0	1 15 5	1 4 0	1 4 0	22 0 0	0 17 5	7 2 0	1 6 2
	TDG _c	2 19 0	0 13 8	1 4 0	0 4 1	18 0 0	0 10 8	6 2 0	4 3 1
A_2^*	TDG _m	2 19 0	3 16 2	1 4 0	0 4 1	22 0 0	0 13 9	4 5 0	1 6 2
	TDG _c	2 19 0	4 11 6	1 4 0	1 4 0	20 0 0	0 11 9	5 5 0	4 3 3

Quantitative:

Best results (for Woodworking):

TDG_m: 83% search space reduction with 45% runtime reduction

TDG_c: 92% search space reduction with 40% runtime reduction

Worst results (for Satellite):

TDG_c: 5.4% search space reduction, but 100% runtime *increase*



Strategy		UM-Tr. (21 inst.)			SmartPh. (5 inst.)			Satellite (22 inst.)			Woodw. (11 inst.)			Summary (59 inst.)		
		#s	#o	cost	#s	#o	cost	#s	#o	cost	#s	#o	cost	#s	#o	cost
blind	Uniform	21	21	1.00	4	4	1.00	17	17	1.00	8	8	1.00	50	50	1.00
	BF	21	21	1.00	4	4	1.00	15	15	1.00	7	7	1.00	47	47	1.00
	DF	21	21	1.00	5	1	1.60	19	7	2.09	8	4	1.44	53	33	2.09
systems	UMCP _{BF}	21	21	1.00	4	4	1.00	15	15	1.00	7	7	1.00	47	47	1.00
	UMCP _{DF}	21	21	1.00	4	1	1.60	17	6	2.09	6	4	1.29	48	32	2.09
	UMCP _h	21	21	1.00	5	4	1.40	19	11	1.50	7	7	1.00	52	43	1.50
	Compile	18	18	1.00	5	5	1.00	21	18	1.10	5	5	1.00	49	46	1.10
	Compile _{opt}	16	16	1.00	5	5	1.00	9	9	1.00	5	5	1.00	35	35	1.00
A*	ADD	21	21	1.00	4	1	1.20	21	21	1.00	10	9	1.17	56	52	1.20
	ADD-r	21	21	1.00	5	5	1.00	19	18	1.08	9	4	1.25	54	48	1.25
	Relax	21	21	1.00	5	5	1.00	18	18	1.00	10	8	1.17	54	52	1.17
	OC	21	21	1.00	4	4	1.00	21	21	1.00	10	7	1.17	56	53	1.17
	TDG _m /-rec	21	21	1.00	5	5	1.00	22	21	1.31	9	9	1.00	57	56	1.31
	TDG _c /-rec	21	21	1.00	5	5	1.00	18	18	1.00	8	8	1.00	52	52	1.00
Greedy-A*	ADD	21	21	1.00	4	0	1.20	21	20	1.09	10	9	1.17	56	50	1.20
	ADD-r	21	21	1.00	5	5	1.00	20	17	1.10	10	4	1.25	56	47	1.25
	Relax	21	21	1.00	5	5	1.00	18	15	1.10	10	4	1.25	54	45	1.25
	OC	21	21	1.00	4	4	1.00	22	21	1.09	10	7	1.22	57	53	1.22
	TDG _m /-rec	21	21	1.00	5	5	1.00	22	17	1.31	9	8	1.08	57	51	1.31
	TDG _c	21	21	1.00	5	5	1.00	20	20	1.00	10	10	1.00	56	56	1.00
	TDG _c -rec	21	21	1.00	5	5	1.00	20	20	1.00	11	11	1.00	57	57	1.00



Summary:

- Introduced the first admissible heuristic for standard HTN and hybrid planning.
- The proposed heuristic(s) perform best both in terms of coverage and plan quality.

Also in the paper and poster:

- A variant of the heuristic tailored to hybrid planning systems.
- Investigation of TDG recomputation to improve heuristic accuracy.

