

Landmark-Aware Strategies for Hierarchical Planning

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HTN Planning – Problem Formalization

$\pi = \langle s_{init}, P_{init}, D \rangle$ is an HTN planning problem with

- s_{init} is the initial state
- P_{init} is the initial partial plan that needs to be decomposed
- $D = \langle T, M \rangle$ is the domain model with
 - T is a set of task schemata of the form
 $t(\bar{\tau}) = \langle pre, eff \rangle$, $\bar{\tau}$ is the parameter list of t
 - M is a set of decomposition methods of the form
 $m = \langle t(\bar{\tau}), P \rangle$, P is a partial plan



HTN Planning – Solution Formalization

Let $\pi = \langle s_{init}, P_{init}, D \rangle$ an HTN planning problem.

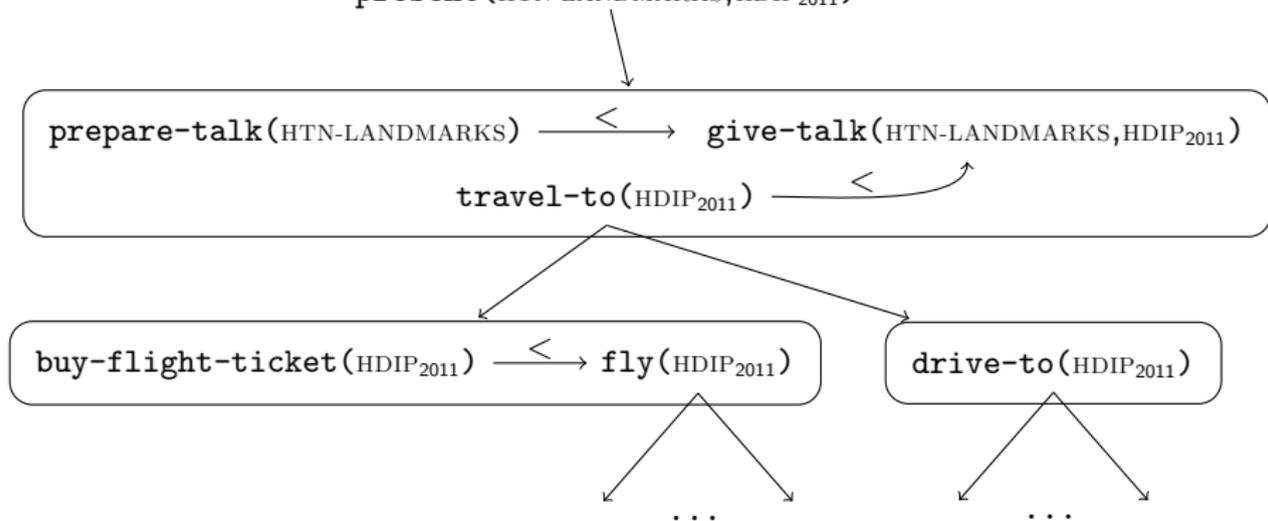
The partial plan P_{sol} is a solution to π if and only if

- P_{sol} contains only primitive tasks
- P_{sol} is obtained from P_{init} by decomposing non-primitive tasks
- Every linearization of P_{sol} is executable in s_{init}



HTN Planning – Example

`present(HTN-LANDMARKS,HDIP2011)`



Motivation

- Development of novel HTN planning search strategies other than progression:
 - Strategies estimate future branching factor based on landmark information.
- Other well-known HTN planners and their strategies:
 - UMCP (make primitive, then resolve conflicts)
 - EMS (expand, then make sound)
 - SHOP (progression)
 - SHOP2 (progression)



Algorithm: Generic Refinement Planning Algorithm

Input : The sequence $\text{Fringe} = \langle P_{\text{init}} \rangle$.

Output : A solution or fail.

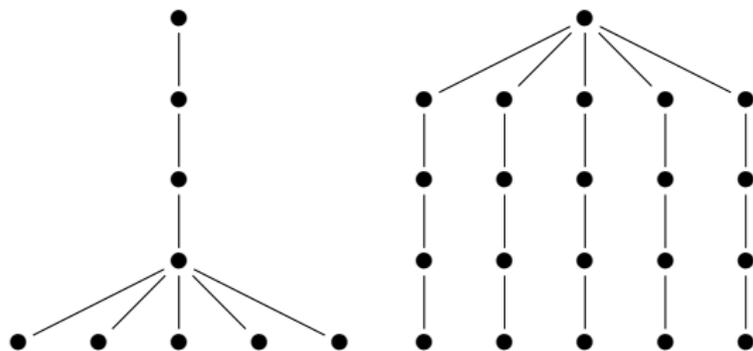
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1 while  $\text{Fringe} = \langle P_1 \dots P_n \rangle \neq \varepsilon$  do
2    $F \leftarrow f^{\text{FlawDet}}(P_1)$ 
3   if  $F = \emptyset$  then return  $P_1$ 
4    $\langle m_1 \dots m_k \rangle \leftarrow f^{\text{ModOrd}} \left( \bigcup_{f \in F} f^{\text{ModGen}}(f) \right)$ 
5    $\text{succ} \leftarrow \langle \text{app}(m_1, P_1) \dots \text{app}(m_k, P_1) \rangle$ 
6    $\text{Fringe} \leftarrow f^{\text{PlanOrd}}(\text{succ} \circ \langle P_2 \dots P_n \rangle)$ 
7 return fail
```



Landmark-Aware Strategies – Motivation

Chosen ordering functions:

- Plan ordering function $f^{PlanOrd}$: *Fewer Modifications First*
- Modification ordering function f^{ModOrd} : *Use landmarks to reduce branching factor by deciding which non-primitive task to decompose first.*



Decomposition Graph & Landmark Table

Figure: Task Decomposition Graph

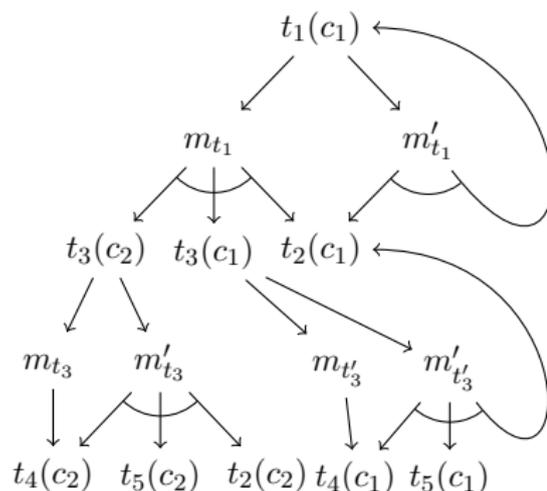


Table: Landmark Table

t	$M(t)$	$O(t)$
$t_1(c_1)$	$\{t_2(c_1)\}$	$\{\{t_3(c_2), t_3(c_1)\}, \{t_1(c_1)\}\}$
$t_3(c_2)$	$\{t_4(c_2)\}$	$\{\emptyset, \{t_5(c_2), t_2(c_2)\}\}$
$t_3(c_1)$	$\{t_4(c_1)\}$	$\{\emptyset, \{t_5(c_1), t_2(c_1)\}\}$

Mandatory Tasks:

Intersection of method's plans

Optional Tasks:

Remaining tasks



Landmark-Aware Strategies – lm_1, lm_2

Let P be a partial plan, $t_i, t_j \in P$ be ground non-primitive tasks.
Let LT be a landmark table and

- $\langle t_i, M(t_i), O(t_i) \rangle \in LT,$
- $\langle t_j, M(t_j), O(t_j) \rangle \in LT$

Definition (Landmark-aware strategy lm_2)

f^{ModOrd} decomposes t_i before t_j if and only if

$$\sum_{o \in O(t_i)} |o| < \sum_{o \in O(t_j)} |o|$$



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Definition (Landmark-aware strategy lm_1)

f^{ModOrd} decomposes t_i before t_j if and only if

$$\sum_{o \in O(t_i)} |o|_{LT} < \sum_{o \in O(t_j)} |o|_{LT}$$

for $|o|_{LT} := |\{t \in o \mid \langle t, M(t), O(t) \rangle \in LT\}|$.



Landmark-Aware Strategies – lm_1, lm_2 (Examples)

Let a plan P contain $t_1(c_1)$ and $t_3(c_1)$. It holds:

- $\langle t_1(c_1), \{t_2(c_1)\}, \{ \{t_3(c_2), t_3(c_1)\}, \{t_1(c_1)\} \} \rangle \in LT$,
- $\langle t_3(c_1), \{t_4(c_1)\}, \{ \emptyset, \{t_5(c_1), t_2(c_1)\} \} \rangle \in LT$

Definition (Landmark-aware strategy lm_2)

f^{ModOrd} decomposes t_i before t_j if and only if

$$\sum_{o \in O(t_i)} |o| < \sum_{o \in O(t_j)} |o|$$

$$\sum_{o \in O(t_3(c_1))} |o| < \sum_{o \in O(t_1(c_1))} |o| \Leftrightarrow 0 + 2 < 2 + 1 \Rightarrow \text{decompose } t_3!$$



Landmark-Aware Strategies – lm_1, lm_2 (Examples)

Let a plan P contain $t_1(c_1)$ and $t_3(c_1)$. It holds:

- $\langle t_1(c_1), \{t_2(c_1)\}, \{ \{t_3(c_2), t_3(c_1)\}, \{t_1(c_1)\} \} \rangle \in LT$,
- $\langle t_3(c_1), \{t_4(c_1)\}, \{ \emptyset, \{t_5(c_1), t_2(c_1)\} \} \rangle \in LT$

Definition (Landmark-aware strategy lm_1)

f^{ModOrd} decomposes t_i before t_j if and only if

$$\sum_{o \in O(t_i)} |o|_{LT} < \sum_{o \in O(t_j)} |o|_{LT}$$

for $|o|_{LT} := |\{t \in o \mid \langle t, M(t), O(t) \rangle \in LT\}|$.

$$\sum_{o \in O(t_3(c_1))} |o|_{LT} < \sum_{o \in O(t_1(c_1))} |o|_{LT} \Leftrightarrow 0 + 0 < 2 + 1 \Rightarrow \text{decompose } t_3!$$



Landmark-Aware Strategies – lm_1^* , lm_2^*

Closure of the strategies traverse the graph to the very last level:

$$O^*(t) = O(t) \cup \bigcup_{o \in O(t)} \left(\bigcup_{t' \in o} O^*(t') \right)$$

with $\langle t, M(t), O(t) \rangle \in LT$

t	$M(t)$	$O(t)$	$O^*(t)$
$t_1(c_1)$	$\{t_2(c_1)\}$	$\{\{t_3(c_2), t_3(c_1)\}, \{t_1(c_1)\}\}$	$O(t_1(c_1)) \cup O(t_3(c_2)) \cup O(t_3(c_1))$
$t_3(c_2)$	$\{t_4(c_2)\}$	$\{\emptyset, \{t_5(c_2), t_2(c_2)\}\}$	$O(t_3(c_2))$
$t_3(c_1)$	$\{t_4(c_1)\}$	$\{\emptyset, \{t_5(c_1), t_2(c_1)\}\}$	$O(t_3(c_1))$



Evaluation – Domains

Conducted experiments on two different HTN planning domains:

- UM Translog
 - Logistics domain – originally formulated for UMCP.
 - 13 qualitatively different problem instances.
 - Deep hierarchy! $\rightarrow lm_i/lm_i^*$ strategies well-informed.
- Satellite
 - Satellite observation tasks, originally designed for classical planning, hierarchy put on top of it.
 - Problem structure: $x - y - z$:
x observations, y satellites, z modes.
 - Shallow hierarchy! $\rightarrow lm_i/lm_i^*$ strategies less informed.



Evaluation – Some Notes

- Comparisons with:
 - standard HTN strategies like *lcf* (least commitment first) and *HotSpot/HotZone* (modifications that don't effect many plan fragments).
 - planning systems mentioned in motivation: our generic refinement algorithm can simulate behavior of (almost) any system when using the according modification and plan ordering functions f^{ModOrd} and $f^{PlanOrd}$.



Evaluation – UM Translog

f^{ModOrd}	1	2	3	4	5	6	7	8	9	10	11	12	13
	#n	#n	#n	#n	#n	#n	#n	#n	#n	#n	#n	#n	#n
UMCP	58	156	177	55	57	308	63	75	220	161	92	90	70
EMS	147	405	211	127	114	–	1571	113	–	2558	879	500	784
SHOP	89	164	146	106	83	926	98	95	–	247	121	173	150
lm ₁	52	133	145	62	53	291	63	71	158	183	75	72	142
lm ₁ [*]	51	135	154	52	65	266	61	61	304	158	78	89	189
lm ₂	62	135	141	53	55	339	109	73	420	211	84	91	104
lm ₂ [*]	124	146	137	57	51	305	110	81	367	142	87	84	114
lcf	55	155	162	78	127	327	62	86	–	227	79	90	247
da-HotSpot	144	644	239	114	148	723	99	120	–	184	641	588	172
du-HotSpot	101	459	1508	160	117	–	1047	75	–	1390	424	307	643
HotZone	55	197	191	55	55	–	159	122	–	701	81	76	345

blue → best result **bold** → second-best result

#n → number of search nodes



Evaluation – Satellite

f^{ModOrd}	1-1-1	1-2-1	2-1-1	2-1-2	2-2-1	2-2-2
	#n	#n	#n	#n	#n	#n
UMCP	83	36	883	1558	278	1062
EMS	65	47	1586	–	1219	–
SHOP	62	105	138	–	1406	–
lm_1	73	194	560	352	693	295
lm_1^*	78	34	847	1803	739	619
lm_2	78	128	4890	200	–	483
lm_2^*	73	91	–	1905	–	146
lcf	86	71	1120	3022	407	–
da-HotSpot	56	68	782	832	2186	142
du-HotSpot	100	139	–	–	–	–
HotZone	61	57	1281	–	1094	871

blue → best result **bold** → second-best result

#n → number of search nodes



Summary

- Partitioning of method's tasks in mandatory and optional sets.
- Landmark strategies use size of (closure of) optional sets to estimate branching factor.
- Evaluation looks promising.

