

# Improving Hierarchical Planning Performance by the Use of Landmarks

Mohamed Elkawkagy, Pascal Bercher,  
Bernd Schattenberg, and Susanne Biundo

*Institute of Artificial Intelligence*

*Ulm University*

*Germany*



ulm university universität  
**uulm**

# Agenda

- Introduction
- Motivation
- Landmark-Aware Strategies
- Evaluation
- Conclusion

# HTN Planning- Problem Formalization

$\Pi = \langle \mathbf{S}_{init}, \mathbf{P}_{init}, \mathbf{D} \rangle$  is an HTN planning problem with

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

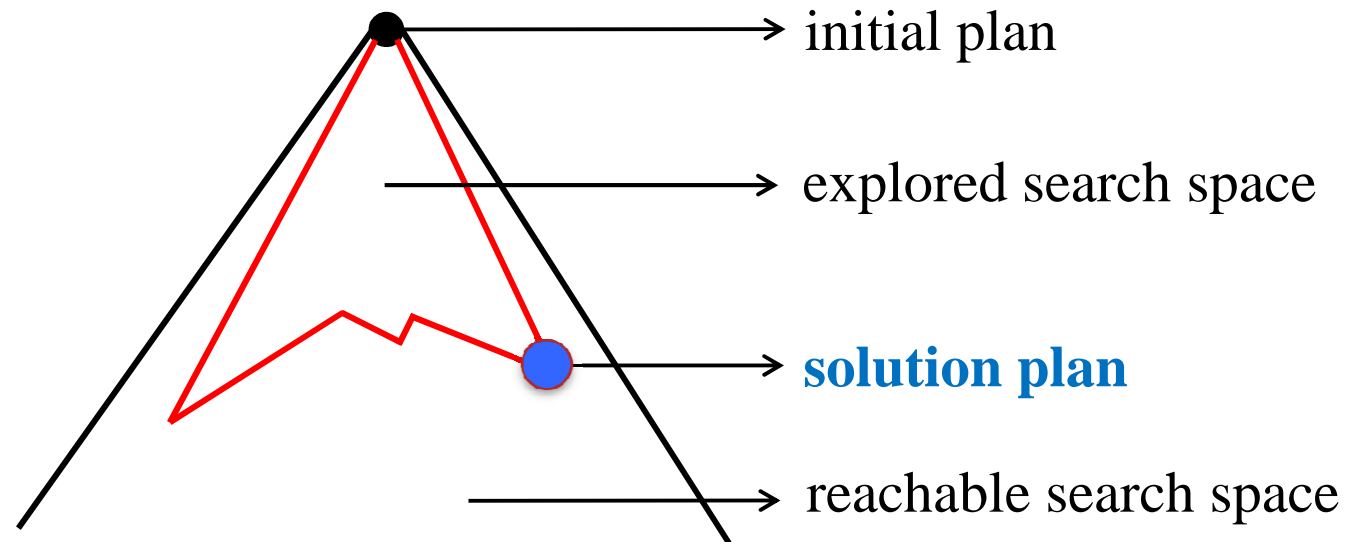
# HTN Planning - Solution Formalization

A **Solution** of a planning problem  $\Pi = \langle S_{init}, P_{init}, D \rangle$  is obtained by refining the initial plan  $P_{init}$  stepwise into a plan  $P_{sol}$ , s.t.

- $P_{sol}$  contains only primitive tasks,
- $P_{sol}$  is obtained from  $P_{init}$  by decomposing non-primitive tasks,
- $P_{sol}$  is executable in  $S_{init}$

# Motivation

❖ Hierarchical planning is based on search techniques in the space of plans.

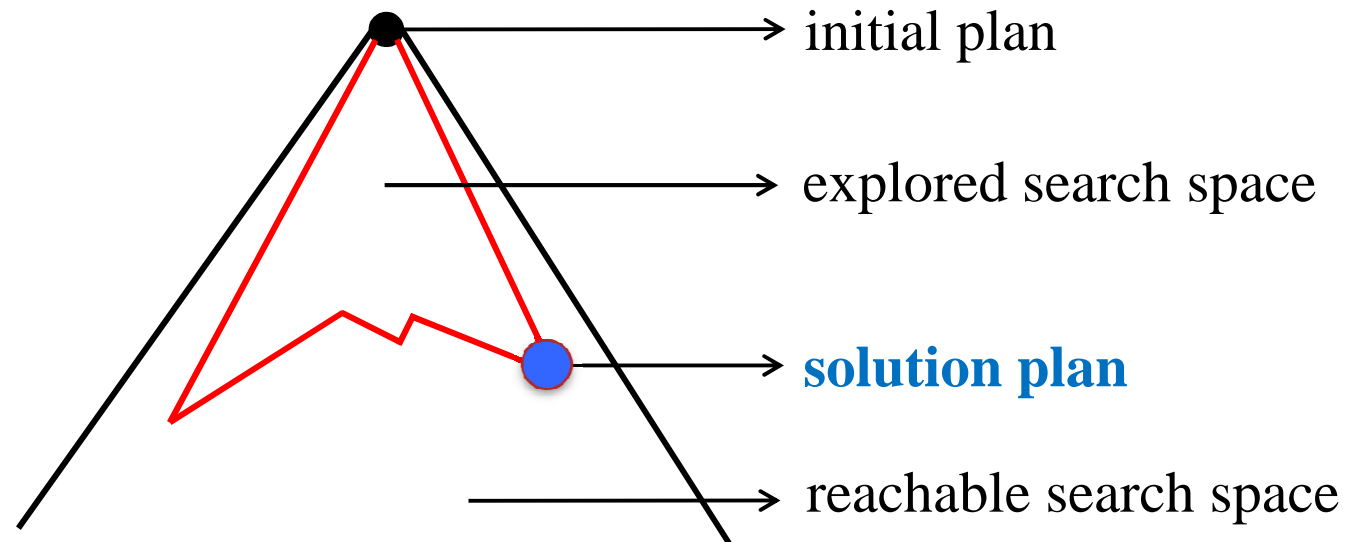


❖ Well-known HTN planners and their strategies:

- UMCP [*Erol et al., 1995*]: (make primitive, then resolve conflicts)
- EMS [*McCluskey, 2000*]: (expand, then make sound)
- SHOP Family [*Nau et al., 1999, Nau et al., 2003*]: (progression)

# Motivation

- ❖ Hierarchical planning is based on search techniques in the space of plans.



- ❖ Development of novel domain-independent search strategies for HTN-based planning :
  - Strategies estimate future branching factor based on landmark information.

# Search Procedure

---

## Algorithm: Refinement Planning Algorithm

---

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

**Output** : A solution or Fail.

```
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
```

orders plan modifications according to a given strategy

Orders the updated fringe

# Search Procedure

---

## Algorithm: Refinement Planning Algorithm

---

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

**Output** : A solution or Fail.

```
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
```

Use landmarks to reduce branching factor by deciding which non-primitive task to decompose first.



# Landmarks

- ❖ **Landmarks** are tasks that occur in the plan sequences leading from the problem's initial plan to its solution [Elkawkagy et al, 2010].

**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:

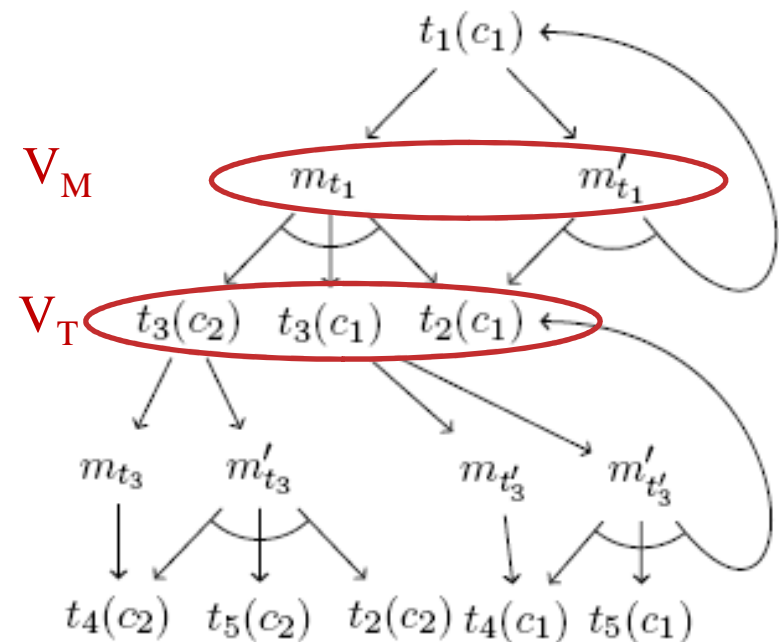
Intersection of method's plans

## Optional Tasks:

Remaining tasks

**Figure:** Task Decomposition Graph

$$\text{TDG} = \langle V_T, V_M, E \rangle$$



# Landmark-Aware Strategies – $lm_2$

$lm_2$

Given a partial plan  $P$ , let  $t_i, t_j \in P$  be abstract tasks.

Let  $LT$  be a landmark table and

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

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

$f_{lm_2}^{ModOrd}$  decomposes task  $t_i$  before task  $t_j$  if and only if

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

# Landmark-Aware Strategies – $lm_1$

$lm_1$

Given a partial plan  $P$ , let  $t_i, t_j \in P$  be abstract tasks.

Let  $LT$  be a landmark table and

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

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

$f_{lm_1}^{ModOrd}$  decomposes task  $t_i$  before task  $t_j$  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

## Example

Let a plan P contain two abstract tasks  $t_1(c_1)$  and  $t_3(c_2)$ .

Let

$$\langle t_1(c_1), \{t_2(c_1)\}, \{\{t_3(c_2), t_3(c_1)\}, \{t_1(c_1)\}\} \rangle \in \text{LT}$$

$$\langle t_3(c_1), \{t_4(c_1)\}, \{\{t_5(c_1), t_2(c_1)\}\} \rangle \in \text{LT}$$

→ Primitive tasks

$$f_{\text{lm}_1}^{\text{ModOrd}} : \sum_{o \in O(t_3)} |o|_{\text{LT}} \prec \sum_{o \in O(t_1)} |o|_{\text{LT}} \Leftrightarrow 0 \prec 2 + 1$$

**Decompose  $t_3$**

$$f_{\text{lm}_2}^{\text{ModOrd}} : \sum_{o \in O(t_3)} |o| \prec \sum_{o \in O(t_1)} |o| \Leftrightarrow 2 \prec 2 + 1$$

**Decompose  $t_3$**

# Landmark-Aware Strategies – $Im_1^*$ , $Im_2^*$

## Closure of the optional set

The closure of the optional set for a given task  $t$  and landmark table  $LT$  is the transitive closure of the optional sets on a recursive traversal of the table entries, beginning in task  $t$ .

$$\mathbf{O}^*(t) = \mathbf{O}(t) \cup \bigcup_{o \in \mathbf{O}(t)} \left( \bigcup_{\bar{t} \in o} \mathbf{O}^*(\bar{t}) \right)$$

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

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

# Evaluation Benchmark

- ❖ We ran our evaluations over four distinguished benchmark domains:

Domain Name	Methods	Abstract tasks	Primitive tasks
UM-Translog	51	21	48
SmartPhone	94	50	87
Satellite	8	3	5
WoodWorking	14	6	13

- ❖ *UM-Translog Domain* describes scenarios of transporting various types of goods by various means (trucks, trains,...) via appropriate infrastructure (roads, transport centers,...).
- ❖ *SmartPhone* manages the operation of a smart phone by a human user such as sending messages and creating contacts or appointments.

# Evaluation Benchmark

- ❖ We ran our evaluations over four distinguished benchmark domains:

Domain Name	Methods	Abstract tasks	Primitive tasks
UM-Translog	51	21	48
SmartPhone	94	50	87
Satellite	8	3	5
WoodWorking	14	6	13

- ❖ *Satellite Domain* manages scientific stellar observations by earth orbiting instrument platforms.
- ❖ *WoodWorking* specifies the processing of raw wood into smooth and varnished product parts.



# Evaluation - Comparison

- ❖ 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  $f^{\text{ModOrd}}$  and plan ordering  $f^{\text{PlanOrd}}$  functions.
- ❖ Org columns show the time needed to solve the *original* (unreduced) problem.
- ❖ Red columns show the time in seconds needed to solve the problem if our domain *reduction* is used.



# Evaluation

UM-Translog domain							SmartPhone domain					
<i>f<sub>ModOrd</sub></i>	#1		#3		#5		#1		#2		#5	
	org	red	org	red	org	red	org	red	org	red	org	red
UMCP	952	244	994	229	187	122	80	<b>30</b>	256	<b>115</b>	-	-
EMS	2056	1048	2199	1806	969	259	107	52	235	148	-	-
SHOP	1735	353	1911	274	5874	4012	95	73	-	-	-	-
Lm <sub>1</sub>	<b>243</b>	<b>180</b>	<b>447</b>	<b>184</b>	<b>121</b>	<b>111</b>	<b>50</b>	<b>30</b>	<b>134</b>	<b>53</b>	-	<b>465</b>
Lm <sub>1</sub> <sup>*</sup>	1772	212	<b>370</b>	<b>205</b>	657	<b>109</b>	65	50	392	173	-	-
Lm <sub>2</sub>	3311	255	1670	248	716	162	60	50	<b>181</b>	<b>53</b>	-	<b>680</b>
Lm <sub>2</sub> <sup>*</sup>	846	226	991	238	583	340	98	76	1632	327	-	697
Lcf	1878	225	3020	209	187	118	63	40	-	159	<b>8455</b>	6827
Da-HotSpot	2414	1958	-	2030	589	352	<b>45</b>	43	-	203	<b>1747</b>	1041
Du-HotSpot	1319	775	987	1090	692	224	52	46	638	166	-	3421
HotZone	<b>473</b>	<b>196</b>	498	224	<b>149</b>	121	65	<b>33</b>	490	212	-	-

❖ **red** == best result / **blue** == second-best result

❖ Dashes indicate that no solution was found within a limit of 150 minutes.



# Evaluation

WoodWorking domain							Satellite domain					
<i>f</i> ModOrd	#1		#3		#5		#1		#2		#3	
	org	red	org	red	org	red	org	red	org	red	org	red
UMCP	228	133	259	125	892	218	<b>51</b>	<b>41</b>	2035	1336	4150	1894
EMS	415	298	-	2457	-	512	62	53	2608	2856	-	-
SHOP	-	-	-	-	-	3578	113	111	<b>270</b>	<b>264</b>	-	-
Lm <sub>1</sub>	<b>96</b>	<b>55</b>	<b>171</b>	159	<b>564</b>	<b>197</b>	209	208	<b>767</b>	652	<b>458</b>	<b>400</b>
Lm <sub>1</sub> <sup>*</sup>	<b>82</b>	<b>50</b>	614	<b>98</b>	2109	1245	<b>54</b>	<b>43</b>	1024	969	<b>2617</b>	2569
Lm <sub>2</sub>	881	433	-	362	-	-	151	140	-	5804	2816	251
Lm <sub>2</sub> <sup>*</sup>	1359	403	-	367	-	893	191	99	-	-	2636	2553
Lcf	2067	350	-	-	-	-	154	77	1551	1338	-	4069
Da-HotSpot	113	85	<b>355</b>	<b>110</b>	<b>573</b>	<b>201</b>	85	78	2136	1131	-	1131
Du-HotSpot	-	-	-	-	-	-	270	150	-	-	-	-
HotZone	-	-	-	-	-	-	142	62	-	4764	-	-

❖ We solved problems for which some established strategies do not find solutions within the given resource bounds.



# Conclusion

- ❖ Partitioning of method's tasks in mandatory and optional sets.
- ❖ Landmark strategies use size of (closure of) optional sets to estimate branching factor.
- ❖ Landmark-aware strategies outperform the established ones on almost the evaluated problems .