

# Improving Greedy Best-First Search by Removing Unintended Search Bias

University of Tokyo Masataro Asai, Alex Fukunaga

## 1. Backgrounds

### Greedy Best First Search

- Best-first search using goal estimate (h-value): expand the node w/ the smallest h-value in the OPEN list
- Search is solely guided by the heuristic estimate h
- May be misguided by incorrect heuristics

### Diversified Greedy Best First Search

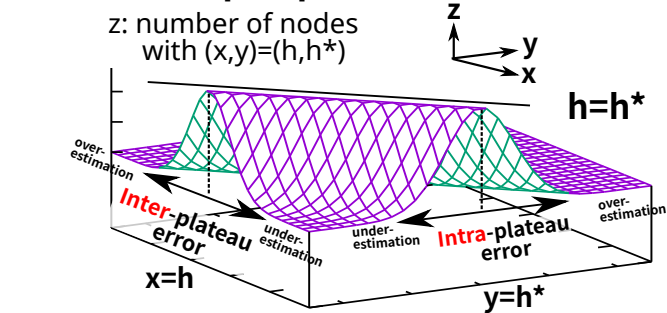
- Strategy to avoid the pathological behavior
- Occasionally ignore the best-h-first ordering

### Plateau (also called Uninformative Heuristic Region)

- A set of nodes with the same h-value
- Indistinguishable in terms of priority
- Large plateau slows down the search

## 2. Intra/Inter Plateau Diversification

Address two perspectives of "heuristic error"



Nodes w/ same  $h^*$ -value have different  $h$ -value  
→ Naive GBFS always selects minimum  $h$   
→ low- $h^*$  nodes w/ high- $h$  may not be expanded

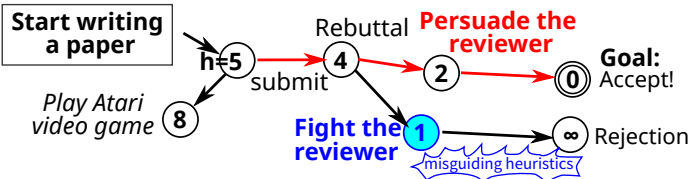
**Inter-plateau Diversification:**  
Randomized h-selection

Nodes w/ same  $h$ -value have different  $h^*$ -value  
→ Naive tiebreaking may keep expanding high- $h^*$  nodes

**Intra-plateau Diversification:**  
Randomized tiebreaking  
(=node selection in a certain  $h$ -plateau)

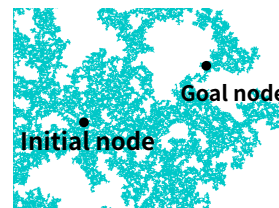
- h-selection strategy and tiebreaking strategy after h-selection do not interfere each other
- Therefore, their effects are orthogonal

note: GBFS-LE(Local Exploration) (Xie 14) for UHR does not restrict LE in a plateau, so it cannot be used for showing the orthogonality of Intra/Inter-plateau diversification.



## 3. Invasion Percolation for Search Diversification

Invasion Percolation (IP) (Wilkinson and Willemsen '83)



- Physical model for the distribution of fluid slowly invading porous media (e.g. water replacing the oil in a porous rock for retrieving oil)
- Theoretical characteristics are well studied in physics; Fractal dimension, Isotropic (rotational invariance)
- Node marking algorithm for IP is shown to equivalent to Prim's method for Minimum Spanning Tree (Barbasi, '96)

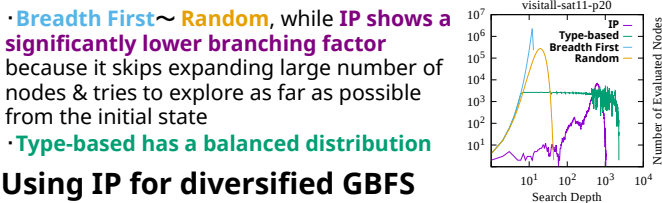
Using IP for diversified blind search

$$r_{BIP}(n) = \min_{e \in \text{parents}(n)} \text{memo}(e, \text{random}())$$

- Sort the search nodes according to  $r_{BIP}$  function and run best-first search
- $r_{BIP}$ : minimum of the random values memoised on each search edge
- **Embankment effect**: high- $r_{BIP}$  value nodes surround a certain region and prevent / delay the exploration inside it
- **Basic random search does not have this effect.**  
random selection  
== pick the first node of the results of random sort  
== allocate new random values to the nodes without memo  
→ new value may allow for exploration inside embankment



Distribution of the nodes per depth



Using IP for diversified GBFS

- Can be used as both Intra- and Inter-plateau diversification
- Intra-plateau: use  $r_{BIP}$  for breaking the tie in GBFS
- Inter-plateau: Alternate normal GBFS (best-h-first) expansion with expansion from an additional queue sorted by  $r_{BIP}$

## 4. Evaluation

Exp1: The performance of the same algorithm as Intra/Inter-plateau div.

hd (random depth tiebreaking, Asai '16) vs hD (=Type-based, Xie'14) --- both are Type-based):  
→ **orthogonal improvements were observed**

Exp2: Combining Inter+Intra (hdD=hd+hD):  
→ **Better performance than either of single variants (hd,hD)**

→ **If you have a diversification method, always consider using them for BOTH intra-plateau AND inter-plateau**

5 min, 4GB, 10 runs, all based on FD. Number of solved instances

	h <sup>OG</sup>				h <sup>FF</sup>				h <sup>CC</sup>				h <sup>FF</sup>			
	h	hd	hD	hdD	h	hd	hD	hdD	h	hb	hB	hbB	h	hb	hB	hbB
total	187	194.2	206.1	215.8	192	208	207.4	223.9	187	187.2	206.8	208.7	192	207.8	232.9	237.7
elevators	9	8	8.7	9.7	19	14	15.9	13.7	9	9.2	12.6	13.3	9.7	19	18.2	18.5
nomystery	7	6	15.4	15.1	9	7	16.6	17	7	6.4	5.5	5.6	15.1	9	6.6	7.6
parcprinter	20	20	19.4	18.7	20	20	20	20	20	19.6	13.7	12.4	18.7	20	20	19.9
pegsol	20	20	20	20	20	20	20	20	20	20	19.7	19.8	20	20	20	20
scanalyzer	20	20	19.9	20	15	15.1	18	18.6	20	20	20	20	20	15	16.6	19.1
sokoban	16	16	16.9	17	19	19	17.4	17.4	16	15.9	15.8	15.2	17	19	18.6	18.5
tidybot	16	16	16.7	16.6	16	16	16.7	16.7	16	17.3	17.5	17.5	18.6	16	15	16.4
woodwork	2	2	2.7	7.7	2	2	4	7.2	2	1.8	14	12.8	7.7	2	1.5	14.8
barman	0	0	0	0	0	0	1.5	1	0	0	0	0	0	0	7.6	6.5
cavendish	7	7	7	7	7	7	7	7.2	7	7.1	7	6.9	7	7	7	7.2
childsack	1	6	0.1	1.5	0	4	0	0.3	1	0	0.1	0	1.5	0	0	0.1
citycar	0	0	7.8	4.7	0	0	7.2	7.1	0	0.2	1.1	0.4	4.7	0	3	3.8
f oortile	0	0	2	2	2	2	2	2.1	0	0	0.5	0.2	2	2	2	2.1
ged	0	0	9.6	9.7	19	19	14	13.8	0	0	4.8	4.6	9.7	19	19.2	12.8
hiking	18	16.9	19.5	19.7	20	20	19.8	20	18	15.9	18.7	18.8	19.7	20	17.6	19.9
maintenance	16	16	16.1	15.8	11	8	10.7	11.1	16	14.6	14.9	14.1	15.8	11	6.7	10
openstacks	0	3.5	0	0.5	0	12.6	0	7	0	0.1	2.5	2.4	0.5	0	15.7	11.7
parking	7	9.7	1.2	4.1	4	7.5	1.4	5.7	7	10.4	7.6	10.9	4.1	4	5.4	2.3
tetris	18	17.1	12.4	14.3	1	5.8	3.2	4.9	18	19.7	17.6	19.4	14.3	1	8.6	7
thoughtful	5	5	5	5	8	9	12.7	13.1	5	4.9	5.2	5.2	5	8	9.1	11.2
transport	5	3	3.7	4.7	0	0	0	0	5	4.1	6	7.1	4.7	0	0	0
visital	0	0	0	0	0	0	0	0	0	0	2	2.1	0	0	3.4	3.8

Exp3: IP diversification (hb,hB,hbB=hb+hB):  
→ Better performance in different domains than hd/hD/hdD

Type-based and IP showed different improvements  
→ Combining Type-based and IP also improves performance.

Exp4: LAMA + d + D + b + B : LAMAd<sup>2</sup>DB (combining inter/intra Type-based and inter/intra IP)  
→ **State-of-the-Art performance**

LAMA	+d	+D	+dD	+b	+B	+bB	+db <sup>2</sup> DB
293.2	296.5	294.3	295.4	293.3	287.6	297.6	<b>304.5</b>

## 5. Future directions

- Type-GBFS, ε-GBFS, DBFS, IP... many approaches
- IP shows impressive performance and is based on fractals
- How they differ? And why?
- Understand them as algorithms creating random fractals
- They can be quantitatively analysed by fractal dimension: Representing how "sparse" the searched space is