

# Potential Heuristics: Weakening Consistency Constraints

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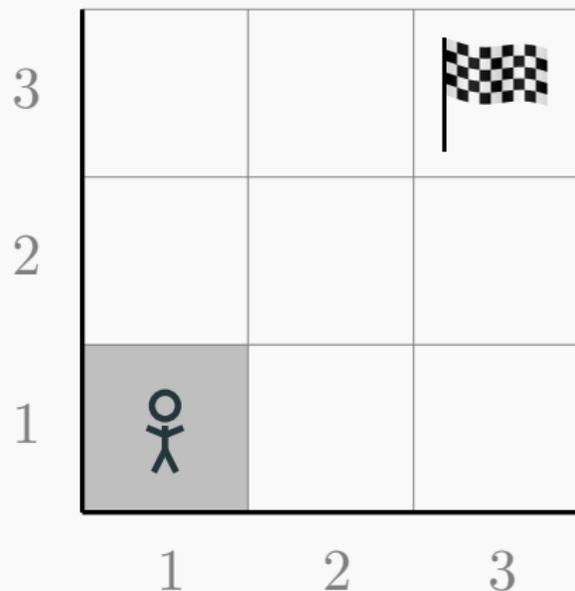
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# FDR Encoding Visitone



$$V = \{at\} \cup \{vis_{i,j} \mid (i,j) \in \{1, \dots, 3\}^2\}$$

$$\text{dom}(at) = \{p_{i,j} \mid (i,j) \in \{1, \dots, 3\}^2\}$$

$$\text{dom}(vis_{i,j}) = \{\top, \perp\}$$

$$O = \{move_{i_1, j_1, i_2, j_2} \mid (i_1, j_1, i_2, j_2) \in \{1, \dots, 3\}^4 \\ |i_1 - j_1| + |i_2 - j_2| = 1\}$$

$$c = \{o \mapsto 1 \mid o \in O\}$$

$$\mathcal{I} = \{at \leftarrow p_{1,1}, vis_{1,1} \leftarrow \top\}$$

$$\cup \{vis_{i,j} \leftarrow \perp \mid (i,j) \in \{1, \dots, n\}^2 \setminus \{(1,1)\}\}$$

$$\mathcal{G} = \{vis_{n,n} \leftarrow \top\}$$

$$move_{i_1, j_1, i_2, j_2} :$$

$$pre : \{at \leftarrow p_{i_1, j_1}\}$$

$$eff : \{at \leftarrow p_{i_2, j_2}, vis_{i_2, j_2} \leftarrow \top\}$$

# Admissible Heuristics

Pro forma:

- $v \leftarrow x$  is a fact with  $v \in \mathcal{V}, x \in \text{dom}(v)$ .
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A heuristic  $h : S \mapsto \mathbb{R} \cup \{\infty\}$  is:

- **goal-aware:**  $h(s) \leq 0$  for every  $s \in S$  with  $G \subseteq S$
- **consistent:**  $h(s) \leq h(s') + c(o)$  for all  $s \in S$  and  $s'$  reached by  $o \in O$  applicable in  $s$
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*Strictly speaking the above definitions should consider forward reachable state and define each property  $p$  as "forward  $p$ ".*

## (One Dimensional) Potential Heuristics

The potential heuristic  $h^P : S \rightarrow \mathbb{R} \cup \{\infty\}$  sums the potential values of a potential function  $P : \mathcal{F} \rightarrow \mathbb{R}$  over the facts in the given state  $s$ :

$$h^P(s) = \sum_{f \in s} p(f)$$

## Consistence Constraints for Opeator $o$

We ensure: “Potentials of facts deleted by  $o$ ” - “Potentials of facts added by  $o$ ”  $\leq c(o)$

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If life was easy ( $\mathcal{V}(\text{eff}(o)) \subseteq \mathcal{V}(\text{pre}(o))$ ):

$$\sum_{v \in \mathcal{V}(\text{eff}(o)), v \leftarrow x \in \text{pre}(o)} \text{P}(v \leftarrow x) - \sum_{f \in \text{eff}(o)} \text{P}(f) \leq c(o)$$

$\mathcal{V}(\cdot)$  denotes variables occuring in a set of facts.

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$\mathcal{V}(\cdot)$  denotes variables occuring in a set of facts.

Extended (always correct):

$$\sum_{v \in \mathcal{V}(\text{eff}(o)) \setminus \mathcal{V}(\text{pre}(o))} \max_{x \in \text{dom}(v)} \text{P}(v \leftarrow x) + \sum_{v \in \mathcal{V}(\text{eff}(o)), v \leftarrow x \in \text{pre}(o)} \text{P}(v \leftarrow x) - \sum_{f \in \text{eff}(o)} \text{P}(f) \leq c(o)$$

# Disambiguations

$\mathcal{D}_o(v)$  denotes a subset of  $\{v \leftarrow x \mid x \in \text{dom}(v)\}$ , restricted by the precondition of  $o$ .  
(Similarly,  $\mathcal{D}_G(v)$  for the goal.)

**Goal-Awareness:**

$$\sum_{v \in \mathcal{V}} \max_{f \in \mathcal{D}_G(v)} P(f) \leq 0$$

**Consistency:**

$$\sum_{v \in \mathcal{V}(\text{eff}(o))} \max_{f \in \mathcal{D}_o(v)} P(f) - \sum_{f \in \text{eff}(o)} P(f) \leq c(o)$$

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*Besides the concise notation, they make an important difference for experiments.*

## P for (Almost) Perfect Heuristic

$at \leftarrow p_{x,y}$

2	1	0
3	2	1
4	3	2

(Manhattan Distance)

$vis_{x,y} \leftarrow \top$

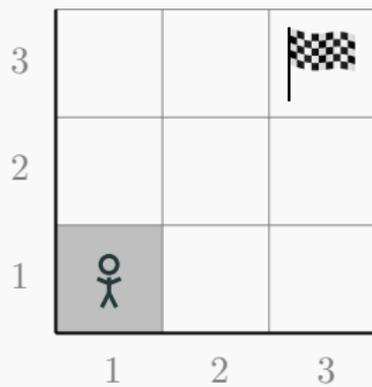
0	0	-4
0	0	0
0	0	0

(Cancel Out Positions)

$h^P(s) = h^*(s)$  for all states  $s$  with  $h^*(s) > 0$ . Generalizes to  $n \times n$ .

# Investigating the heuristic

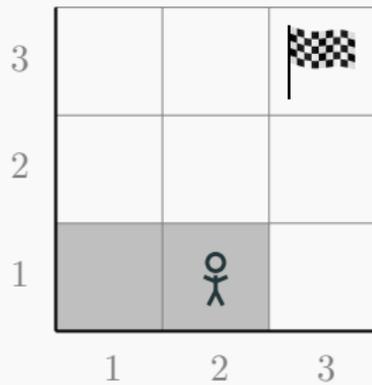
$s :$



$$h^P(s) = 4$$

# Investigating the heuristic

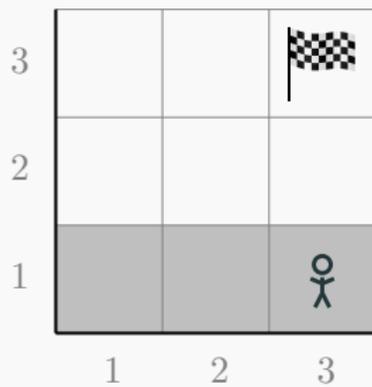
$s :$



$$h^P(s) = 3$$

# Investigating the heuristic

$s$  :



$$h^P(s) = 2$$

# Investigating the heuristic

$s$  :

3			
2			
1			
	1	2	3

$$h^P(s) = 1$$

## Investigating the heuristic: It is not consistent!



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$$h^P(s) - h^P(s') = 1 - (-4) = 5 \not\leq 1 = c(\text{move}_{3,2,3,3})$$

## Just to confirm: It violates our LP constraints

Consider the consistency constraint of  $h^P$  for  $move_{3,2,3,3}$ :

$$\begin{aligned} & p(at \leftarrow p_{3,2}) + \max_{x \in \{\top, \perp\}} p(vis_{3,3} \leftarrow x) - p(at \leftarrow p_{3,3}) - p(vis_{3,3} \leftarrow \top) \\ = & \quad 1 \quad + \quad \quad \quad 0 \quad \quad \quad - \quad \quad 0 \quad \quad - \quad (-4) \quad = \quad 5 \\ & \not\leq 1 = c(move_{3,2,3,3}) \end{aligned}$$

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*In fact one can prove any heuristic value to be bounded by  $\leq 2$  under our constraint encoding.*

## When is this a problem?

Intuitively prevents us from capturing distances of any kind. We can, e.g., also observe this in:

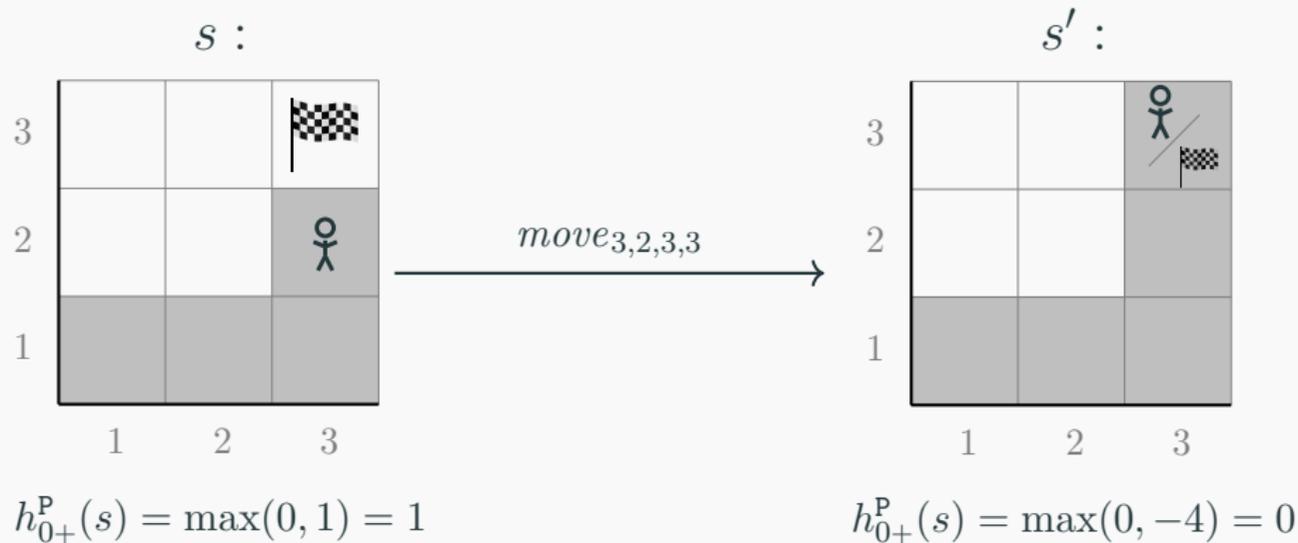
- Logistics (In the paper)
- Elevators (In the experiments)

But do we use an inconsistent heuristic (in search)?

$$h_{0+}^P(s) := \max(h^P(s), 0)$$

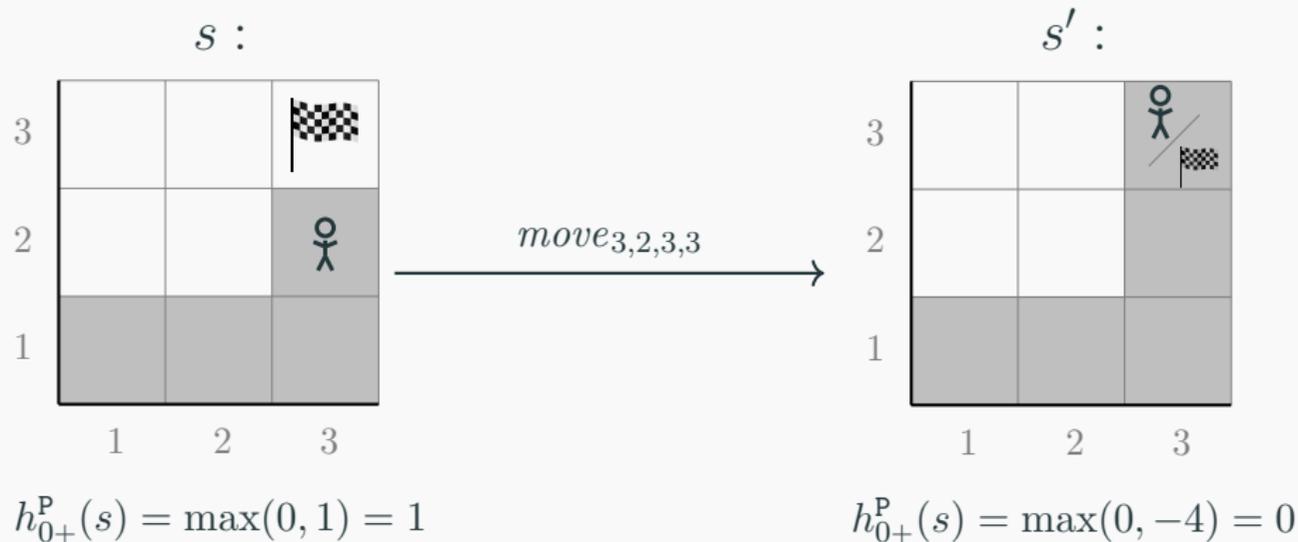
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$$h_{0+}^P(s) - h_{0+}^P(s') = 1 - 0 = 1 \leq c(move_{3,2,3,3})$$

## Deriving consistency constraints for $h_{0+}^P$

Consistency for  $h_{0+}^P$  in transition from  $s$  to  $s'$  via operator  $o$ :

$$\begin{aligned} h_{0+}^P(s) - h_{0+}^P(s') &\leq c(o) \\ \Leftrightarrow \max(h^P(s), 0) - \max(h^P(s'), 0) &\leq c(o) \end{aligned}$$

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$$\Leftrightarrow \max(h^P(s), 0) - \max(h^P(s'), 0) \leq c(o)$$

$\Leftrightarrow$  **do the math**

$$\min(h^P(s) - h^P(s'), h^P(s)) \leq c(o)$$

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$$\min(\underbrace{h^P(s) - h^P(s')}_{\text{old constraint}}, \underbrace{h^P(s)}_{?}) \leq c(o)$$

## Estimating $h^P$ s

Estimate on  $h^P(s)$  for all  $s$  where the precondition of  $o$  is fulfilled:

$$\sum_{v \in \mathcal{V}} \max_{f \in \mathcal{D}_o(v)} P(f)$$

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Revisiting the example:

$$\begin{aligned} & p(at \leftarrow p_{3,2}) + \sum_{(i,j) \in \{1,\dots,3\}^2} \max_{x \in \{\top, \perp\}} p(vis_{i,j} \leftarrow x) \\ &= 1 + 0 \\ &\leq 1 = c(move_{3,2,3,3}) \end{aligned}$$

## Putting it together with indicator constraints

$$\min(h^P(s) - h^P(s'), h^P(s)) \leq c(o)$$

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We want to encode an or:

$$ind = 0 \Rightarrow \sum_{v \in \mathcal{V}(\text{eff}(o))} \max_{f \in \mathcal{D}_o(v)} P(f) - \sum_{f \in \text{eff}(o)} P(f) \leq c(o)$$

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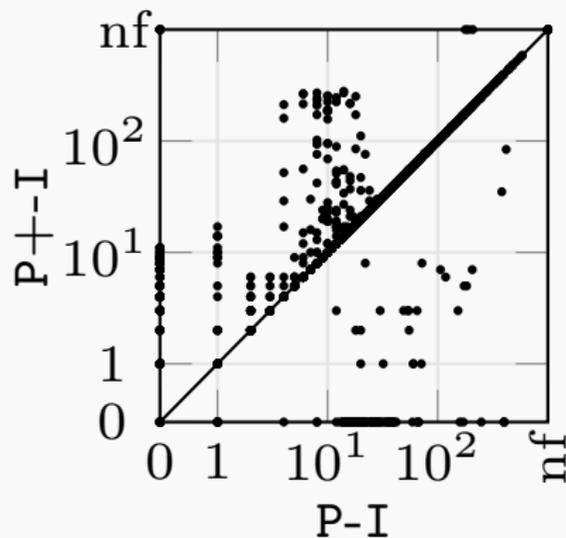
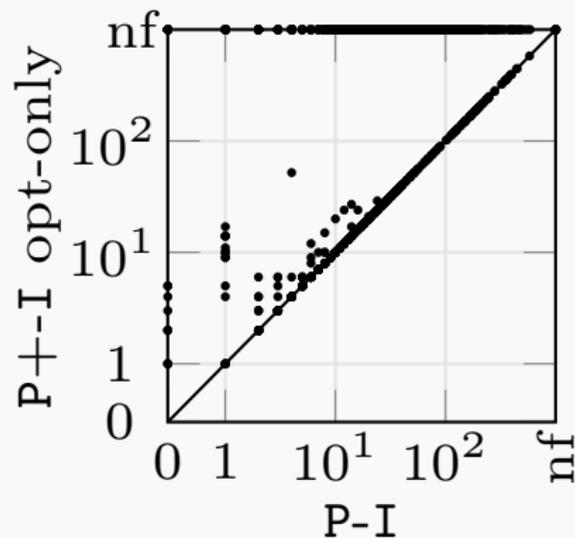
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$$ind = 1 \Rightarrow \sum_{v \in \mathcal{V}} \max_{f \in \mathcal{D}_o(v)} P(f)$$

**Pro:** Optimal solution will be strictly better

**Con:** We need an ILP

## Experiments (Initial h-value)



P+ is our ILP variant. opt-only requires to find an optimal solution. (Other configuration takes best after timeout.)

# Experiments (Coverage)

Domain	I			A			AI		
	P	P+	por	P	P+	por	P	P+	por
blocks (35)	21	<b>28</b>	<b>28</b>	28	28	28	28	28	28
caldera (20)	10	<b>12</b>	10	12	12	12	12	12	12
driverlog (20)	9	<b>13</b>	<b>13</b>	<b>13</b>	11	<b>13</b>	13	13	13
elevators (50)	31	31	31	31	<b>35</b>	<b>35</b>	31	<b>33</b>	<b>33</b>
freecell (80)	48	<b>60</b>	<b>60</b>	37	37	37	<b>74</b>	69	<b>74</b>
ged (20)	16	<b>18</b>	<b>18</b>	15	15	15	15	<b>18</b>	<b>18</b>
logistics (63)	13	<b>23</b>	<b>23</b>	24	24	24	24	24	24
mprime (35)	18	<b>20</b>	19	25	25	25	25	25	25
nomystery (20)	10	<b>14</b>	<b>14</b>	<b>14</b>	13	<b>14</b>	14	14	14
parking (40)	1	<b>7</b>	<b>7</b>	<b>16</b>	6	<b>16</b>	<b>16</b>	7	<b>16</b>
pipesw-notank (50)	25	<b>26</b>	<b>26</b>	25	25	25	<b>30</b>	28	<b>30</b>
recharging-robots (20)	<b>13</b>	12	<b>13</b>	<b>13</b>	12	<b>13</b>	<b>13</b>	11	<b>13</b>
scanalyzer (50)	25	25	25	23	23	23	<b>27</b>	25	<b>27</b>
snake (20)	<b>15</b>	14	<b>15</b>	<b>14</b>	13	<b>14</b>	<b>16</b>	13	<b>16</b>
spider (20)	<b>14</b>	11	<b>14</b>	<b>14</b>	11	<b>14</b>	<b>16</b>	12	<b>16</b>
tidybot (40)	<b>32</b>	30	<b>32</b>	<b>32</b>	30	<b>32</b>	<b>32</b>	30	<b>32</b>
tpp (30)	6	<b>8</b>	<b>8</b>	6	6	6	8	8	8
trucks (30)	9	<b>14</b>	<b>14</b>	<b>14</b>	11	<b>14</b>	14	14	14
visitall (40)	25	<b>30</b>	<b>30</b>	25	<b>28</b>	<b>28</b>	27	<b>30</b>	<b>30</b>
woodworking (50)	19	<b>27</b>	<b>27</b>	<b>27</b>	25	<b>27</b>	31	31	31
zenotravel (20)	8	<b>11</b>	<b>11</b>	11	11	11	11	11	11
others (1053)	558	<b>566</b>	565	569	566	<b>570</b>	<b>584</b>	582	<b>584</b>
Σ (1806)	926	1000	<b>1003</b>	988	967	<b>996</b>	1061	1038	<b>1069</b>

Thank you :)