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1 Problem

1.1 General Problem

We formulate our problem with the following five concepts:

- State(S) including a initial state (s_0) the agent begins with
- Actions(A) available to agent at each State
- Transition function $(F: \mathbb{S} \times \mathbb{A} \to \mathbb{S})$ that takes a (state, action) pair and return a new state
- Goal Test($GT: S \to \{True, False\}$) that takes a state s and return true if $s \in S_{goal}$, the Goal State of the problem.
- Path Cost $(C: \mathbb{S} \times \mathbb{S} \to \mathbb{R})$ That takes two states and return cost moving from one to another

1.2 Single-player Chexers

In this section, we formally describe how the above framework fits the Chexers game.

• Denote set of pieces with $\mathbb{P} = \{(r, q, t)\}.$ where

 $r, q, -(r+q) \in [-3, 3]$ denotes the location on board. $t \in \{red, blue, green, block\}$ stands for type of piece.

- $\mathbb{S} := \{p_i \in \mathbb{P}, t_b | i \leq n\}$ where n is number of pieces on board, where t_b is the searching type.*
- $\mathbb{A} := \{(Move, p_i), (Jump, p_i), (Exit, p_i)\}, \forall p_i \text{ where } t(p_i) = t_b. |\mathbb{A}_s| \leq 6n_p \ \forall s.$
- f(s,a) := s' where s' differs exactly by one piece $r(p_{i,s}), q(p_{i,s}) \neq r(p_{i,s'}), q(p_{i,s'})$ or $p_i \notin s'$
- $c(s, s') := 1, \forall s, s' \text{ if } \exists a \text{ such that } f(s, a) = s'$
- gt(s) = True if $\nexists p_i$ such that $t(p_i) = t_b$

2 Search

2.1 preliminary

We use the following Algorithm 1, to search for our goal. Based on our problem specification, we have all required components except for $\mathbf{H}(node)$, so we will propose one type of \mathbf{H} we found to be most effective of all and compare it against $\mathbf{Null}\ H = 0$ and $\mathbf{bad}\ \mathbf{H}\ H_{bad}(node)$.

2.2 Heuristic

As a Problem We propose the following problem definition for mapping a good heuristic value for each pieces and we define $H(s) = \sum_{p_i} h(p_i), \forall p_i$ that $t(p_i) = t_b$ after relaxing the problem to be: a piece can choose freely between **move**, **jump** regardless normal constrains and **cannot** move on blocks.

- $S := \{(cost_i, position_i) | location_i \in board\}^*,$
- A: for pos_i with $cost_i = min(cost_n)$, $\forall pos_{j\neq i}$ reachable from pos_i and $cost_j > cost_i + 1$. Update $(pos_j, cost_i + 1)$. If $\nexists pos_j$, remove $(pos_i, cost_i)$ from s and put $h(pos_i) = cost_i$
- c(s, s') = 0 and f(s, a) follows definition in A
- $gt(s') = True \text{ if } s' = \emptyset$

Heuristic problem solution Finding heuristic is straight forward with the given above problem definition. Supporting all the given operations to 1, and you should end up with a complete heuristic map.

^{*}Initial State s_0 given by problem specification.

 $[*]s_0 = \{(1, pos_i) | can_exit(pos_i) = True\} | \{(\infty, pos_i) | can_exit(pos_i) = False\} |$

Algorithm 1 General A* algorithm PRIORITY-QUEUE ▶ Min Priority Queue (min key) ADD(q, key, value),> Add (key, value) to q. If value exists, update the key POP(q) \triangleright Pop the value with **least** key GET(q, value)▷ Get the key associated with a value \triangleright All above Queue operations can operate in $\Theta(1)$ NODE > stores associated **state** and its **parent** Require: EXPAND(node)⊳ expand a **node** to get it's **children** Require: G(node)⊳ get **total cost** arriving this node Require: H(node)▷ Computes an **admissible estimation** of cost to goal state Require: C(node1, node2)▷ Give Path cost arriving node 2 from node 1 1: **procedure** A*(problem, initial) $openSet \leftarrow PRIORITY-QUEUE(H(initial), initial)$ 2: $closedSet \leftarrow \{\}$ 3: while openSet is not empty do 4: $node \leftarrow POP(openSet)$ 5: 6: ADD(closedSet, node)7: if GOAL-TEST(node) is True then 8: ${\bf return}\ node$ 9: end if 10: for child in EXPAND(node) do cost = G(node) + C(node, child) + H(child)11: if child in closedSet then continue 12: else if child not in openSet or cost < GET(openSet, child) then 13: ADD(openSet, cost, child) 14: end if 15: 16: end for end while 17: return no solution 18:

Admissible? The algorithm is guaranteed to provide us with a admissible cost estimation for each game state for the following reason:

- The relaxed rule let pieces always able to jump. This will in all cases reduce the number of action taken.

2.3 The search

19: end procedure

20: All operations O(1) unless explicitly stated