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Build a Forward Planning Agent

REVIEW

CODE REVIEW 4

HISTORY

▼ my_planning_graph.py 4

```

1 from itertools import chain, combinations
2 from aimacode.planning import Action
3 from aimacode.utils import expr
4
5 from layers import BaseActionLayer, BaseLiteralLayer, makeNoOp, make_node
6
7
8 class ActionLayer(BaseActionLayer):
9
10     def inconsistent_effects(self, actionA, actionB):
11         """ Return True if an effect of one action negates an effect of the other
12
13         Hints:
14             (1) '~Literal' can be used to logically negate a literal
15             (2) 'self.children' contains a map from actions to effects
16
17         See Also
18         -----
19         layers.ActionNode
20         """
21         # TODO: implement this function
22         # raise NotImplementedError
23         return (self.children[actionA] & set([~l for l in self.children[actionB]]))!=set()

```

AWESOME

great, you are the first person -I grade- to use this trick (and I always suggest in my review that they use it) .. good work

```

24
25
26     def interference(self, actionA, actionB):
27         """ Return True if the effects of either action negate the preconditions of the other
28
29         Hints:
30             (1) '~Literal' can be used to logically negate a literal
31             (2) 'self.parents' contains a map from actions to preconditions
32
33         See Also
34         -----
35         layers.ActionNode
36         """
37         # TODO: implement this function
38         # raise NotImplementedError
39         return (self.parents[actionA] & set([~i for i in self.children[actionB]]))!=set() or (self.parents[actionB] & set
40
41     def competing_needs(self, actionA, actionB):
42         """ Return True if any preconditions of the two actions are pairwise mutex in the parent layer
43
44         Hints:
45             (1) 'self.parent_layer' contains a reference to the previous literal layer
46             (2) 'self.parents' contains a map from actions to preconditions
47
48         See Also
49         -----
50         layers.ActionNode
51         layers.BaseLayer.parent_layer
52         """
53         # TODO: implement this function
54         # raise NotImplementedError
55         for action_a_pc in actionA.preconditions:
56             for action_b_pc in actionB.preconditions:
57                 if self.parent_layer.is_mutex(action_a_pc, action_b_pc):
58                     return True # cond are mutex in parent_layer
59         return False
60
61 class LiteralLayer(BaseLiteralLayer):
62
63     def inconsistent_support(self, literalA, literalB):
64         """ Return True if all ways to achieve both literals are pairwise mutex in the parent layer
65
66         Hints:
67             (1) 'self.parent_layer' contains a reference to the previous action layer
68             (2) 'self.parents' contains a map from literals to actions in the parent layer
69
70         See Also
71         -----
72         layers.BaseLayer.parent_layer
73         """
74         # TODO: implement this function
75         # raise NotImplementedError
76         for lit_a_act in self.parents[literalA]:
77             for lit_b_act in self.parents[literalB]:
78                 if not self.parent_layer.is_mutex(lit_a_act, lit_b_act):
79                     return False
80

```

AWESOME

well done

```

80         # if both literals are mutex in self.parent_layer
81         return True
82
83     def negation(self, literalA, literalB):
84         """ Return True if two literals are negations of each other """
85         # TODO: implement this function
86         # raise NotImplementedError
87         if literalA == ~literalB:
88             return True
89         return False
90
91 class PlanningGraph:
92     def __init__(self, problem, state, serialize=True, ignore_mutexes=False):
93         """
94         Parameters
95         -----
96         problem : PlanningProblem
97             An instance of the PlanningProblem class
98
99         state : tuple(bool)
100             An ordered sequence of True/False values indicating the literal value
101             of the corresponding fluent in problem.state map

```

```

102         of the corresponding fluent in problem.state_map
103
104     serialize : bool
105         Flag indicating whether to serialize non-persistence actions. Actions
106         should NOT be serialized for regression search (e.g., GraphPlan), and
107         should be serialized if the planning graph is being used to estimate
108         a heuristic
109
110     """
111     self.serialize = serialize
112     self._is_levelled = False
113     self.ignore_mutexes = ignore_mutexes
114     self.goal = set(problem.goal)
115
116     # make no-op actions that persist every literal to the next layer
117     no_ops = [make_node(n, no_op=True) for n in chain(*(makeNoOp(s) for s in problem.state_map))]
118     self.actionNodes = no_ops + [make_node(a) for a in problem.actions_list]
119
120     # initialize the planning graph by finding the literals that are in the
121     # first layer and finding the actions they should be connected to
122     literals = [s if f else ~s for f, s in zip(state, problem.state_map)]
123     layer = LiteralLayer(literals, ActionLayer(), self.ignore_mutexes)
124     layer.update_mutexes()
125     self.literal_layers = [layer]
126     self.action_layers = []
127
128     def find_level_costs(self): # find goals and layer costs
129         level_num = 0
130         costs = [0]
131         goals_finished = set()
132
133         for goals in self.goal:
134             if goals in self.literal_layers[0]:
135                 goals_finished.add(goals)
136
137         if not self.goal - goals_finished:
138             return costs
139
140         while not self._is_levelled:
141             self._extend()

```

AWESOME

you nailed this part :D (y)

```

141         level_num += 1
142
143         for goal in self.goal - goals_finished:
144             if goal in self.literal_layers[level_num]:
145                 costs.append(level_num)
146                 goals_finished.add(goal)
147
148         if not self.goal - goals_finished:
149             return costs
150
151     def h_levelsum(self):
152         """ Calculate the level sum heuristic for the planning graph
153
154         The level sum is the sum of the level costs of all the goal literals
155         combined. The "level cost" to achieve any single goal literal is the
156         level at which the literal first appears in the planning graph. Note
157         that the level cost is **NOT** the minimum number of actions to
158         achieve a single goal literal.
159
160         For example, if Goal_1 first appears in level 0 of the graph (i.e.,
161         it is satisfied at the root of the planning graph) and Goal_2 first
162         appears in level 3, then the levelsum is 0 + 3 = 3.
163
164         Hints
165         ----
166         (1) See the pseudocode folder for help on a simple implementation
167         (2) You can implement this function more efficiently than the
168             sample pseudocode if you expand the graph one level at a time
169             and accumulate the level cost of each goal rather than filling
170             the whole graph at the start.
171
172         See Also
173         -----
174         Russell-Norvig 10.3.1 (3rd Edition)
175
176         # TODO: implement this function
177         # raise NotImplementedError
178         return sum(self.find_level_costs())
179
180     def h_maxlevel(self):
181         """ Calculate the max level heuristic for the planning graph
182
183         The max level is the largest level cost of any single goal literal.
184         The "level cost" to achieve any single goal literal is the level at
185         which the literal first appears in the planning graph. Note that
186         the level cost is **NOT** the minimum number of actions to achieve
187         a single goal literal.
188
189         For example, if Goal1 first appears in level 1 of the graph and
190         Goal2 first appears in level 3, then the levelsum is max(1, 3) = 3.
191
192         Hints
193         ----
194         (1) See the pseudocode folder for help on a simple implementation
195         (2) You can implement this function more efficiently if you expand
196             the graph one level at a time until the last goal is met rather
197             than filling the whole graph at the start.
198
199         See Also
200         -----
201         Russell-Norvig 10.3.1 (3rd Edition)
202
203         Notes
204         ----
205         WARNING: you should expect long runtimes using this heuristic with A*
206
207         # TODO: implement maxlevel heuristic
208         # raise NotImplementedError
209         return max(self.find_level_costs())
210
211     def h_setlevel(self):
212         """ Calculate the set level heuristic for the planning graph
213
214         The set level of a planning graph is the first level where all goals
215         appear such that no pair of goal literals are mutex in the last
216         layer of the planning graph.
217
218         Hints
219         ----
220         (1) See the pseudocode folder for help on a simple implementation
221         (2) You can implement this function more efficiently if you expand
222             the graph one level at a time until you find the set level rather
223             than filling the whole graph at the start.
224
225         See Also
226         -----
227         Russell-Norvig 10.3.1 (3rd Edition)
228
229         Notes
230         ----
231         WARNING: you should expect long runtimes using this heuristic on complex problems
232
233         # TODO: implement setlevel heuristic
234         # raise NotImplementedError
235         level_num = 0
236         while not self._is_levelled:

```

SUGGESTION

there is a hidden (and common) mistake here which doesn't affect your results because we test each heuristic separately, there is a hidden assumption in your code that the only function modifying the state (self._is_levelled and self.literal_layers) is the current heuristic function. this is of course true for the way we are testing

but in general you can expect that multiple heuristics might be used simultaneously and so you should take care of that (e.g. loop through `self.literal_layers` before entering this loop, that works in this case because it only checks layer 0 and in the general case it will check all layers that were expanded -including those expanded by other heuristics)

```
237         finished_goals = True
238         extended_layers = self.literal_layers[-1]
239
240         for goals in self.goal:
241             if goals not in extended_layers:
242                 finished_goals = False
243
244         if finished_goals == False:
245             level_num += 1
246             self._extend()
247             continue
248
249         mutex_all = False
250         for a_goal in self.goal:
251             for b_goal in self.goal:
252                 if extended_layers.is_mutex(a_goal, b_goal):
253                     mutex_all = True
254
255         if mutex_all == False:
256             return (level_num)
257
258         level_num += 1
259         self._extend()
260
261     return (level_num)
262
263     #####
264     # DO NOT MODIFY CODE BELOW THIS LINE #
265     #####
266
267     def fill(self, maxlevels=-1):
268         """ Extend the planning graph until it is leveled, or until a specified number of
269             levels have been added
270
271         Parameters
272         -----
273         maxlevels : int
274             The maximum number of levels to extend before breaking the loop. (Starting with
275             a negative value will never interrupt the loop.)
276
277         Notes
278         ----
279         YOU SHOULD NOT THIS FUNCTION TO COMPLETE THE PROJECT, BUT IT MAY BE USEFUL FOR TESTING
280         """
281         while not self._is_leveled:
282             if maxlevels == 0: break
283             self._extend()
284             maxlevels -= 1
285         return self
286
287     def _extend(self):
288         """ Extend the planning graph by adding both a new action layer and a new literal layer
289
290         The new action layer contains all actions that could be taken given the positive AND
291         negative literals in the leaf nodes of the parent literal level.
292
293         The new literal layer contains all literals that could result from taking each possible
294         action in the NEW action layer.
295         """
296         if self._is_leveled: return
297
298         parent_literals = self.literal_layers[-1]
299         parent_actions = parent_literals.parent_layer
300         action_layer = ActionLayer(parent_actions, parent_literals, self._serialize, self._ignore_mutexes)
301         literal_layer = LiteralLayer(parent_literals, action_layer, self._ignore_mutexes)
302
303         for action in self.actionNodes:
304             # actions in the parent layer are skipped because are added monotonically to planning graphs,
305             # which is performed automatically in the ActionLayer and LiteralLayer constructors
306             if action not in parent_actions and action.preconditions <= parent_literals:
307                 action_layer.add(action)
308                 literal_layer |= action.effects
309
310             # add two-way edges in the graph connecting the parent layer with the new action
311             parent_literals.add_outbound_edges(action, action.preconditions)
312             action_layer.add_inbound_edges(action, action.preconditions)
313
314             # add two-way edges in the graph connecting the new literal layer with the new action
315             action_layer.add_outbound_edges(action, action.effects)
316             literal_layer.add_inbound_edges(action, action.effects)
317
318         action_layer.update_mutexes()
319         literal_layer.update_mutexes()
320         self.action_layers.append(action_layer)
321         self.literal_layers.append(literal_layer)
322         self._is_leveled = literal_layer == action_layer.parent_layer
323
```

[RETURN TO PATH](#)

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