

P03 Planning and Uncertainty

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1 STRIPS planner

In this part, you will implement a simple STRIPS planner. The input of your planner is a PDDL domain file and a problem file in the STRIPS restriction, that is, preconditions of actions and the goal are conjunctions of atoms, and effects of actions are conjunctions of literals. The output of your planner is a sequence of actions to achieve the goal.

1. Describe with sentences the main ideas behind computing the heuristic for a state using reachability analysis from lecture notes. (10 points)
 - 可达性分析启发式函数的核心思想是考虑这样一个松弛问题：规划的所有行动都不考虑它们效果中的删除部分，每一层执行所有能执行的动作（默认不考虑已经执行过的任务，因为他们不会对知识库有影响），这样问题的知识库只会增加，不会减少。若在这样的松弛问题中都达不到目标，那该状态大概率无法到达目标，其启发式函数值较大；若能达到，再使用 Count-Actions 的思路，每次用行动层的效果覆盖上一层的新前提来尽可能模拟到达目标的过程，启发式函数值为所有层行动的总数。
2. Implement a STRIPS planner by using A* search and the heuristic function you implemented.(20 points)

```
1  import re
2  import copy
3  from timeit import default_timer as timer
4
5  # 测试时使用的，实际上搜索树很浅
6  MAX_DEPTH = 50
7
8  def parser(file_name):
9      str = open(file_name, 'r').read()
10
11     # https://github.com/pucrs-automated-planning/pddl-parser/blob/master/
12     # PDDL.py
13     stack = []
14     list = []
15     for t in re.findall(r'[(\s)]|[\s(]+', str):
16         if t == '(':
17             stack.append(list)
18             list = []
```

```

18         elif t == ')':
19             if stack:
20                 l = list
21                 list = stack.pop()
22                 list.append(l)
23             else:
24                 raise Exception('Missing open parentheses')
25         else:
26             list.append(t)
27     return list[0]
28
29 class Action:
30     def __init__(self, init_list, types):
31         self.name = init_list[1]
32         self.parameters = {}
33
34         if types != ['AnyType']:
35             for i in range(0, len(init_list[3]), 3):
36                 if init_list[3][i+2] in types:
37                     self.parameters[init_list[3][i][1:]] = init_list[3][i +
38                                     2]
39             else:
40                 for i in init_list[3]:
41                     self.parameters[i[1:]] = 'AnyType'
42         self.precondition = []
43         for i in init_list[5]:
44             if type(i) is list:
45                 if i[0] != 'not':
46                     self.precondition.append([True, [i[0]] + [j[1:] for j in
47                                     i[1:]]])
48                 else:
49                     self.precondition.append([False, [i[1][0]] + [j[1:] for
50                                     j in i[1][1:]]])
51         self.effect = []
52         for i in init_list[7]:
53             if type(i) is list:
54                 if i[0] != 'not':

```

```

52         self.effect.append([True, [i[0]] + [j[1:] for j in i
53                                     [1:]]])
54     else:
55         self.effect.append([False, [i[1][0]] + [j[1:] for j in i
56                                     [1][1:]]])
57
58     # 寻找可能的动作实例, 返回 [[ 动作名 , add_list , del_list ]...]
59     # 一种动作也有多种可能, 这个 bug 找了一年
60     def do(self, objects, kb):
61         rst = []
62         mapping_list = []
63         self.create_mapping(objects, {}, 0, mapping_list)
64         for i in mapping_list:
65             precondition = self.assign_precondition(i)
66             flag = True
67             for j in precondition:
68                 if (j[0] and j[1] not in kb) or (not j[0] and j[1] in kb):
69                     flag = False
70                     break
71             if flag:
72                 rst.append(self.assign_effect(i))
73         return rst
74
75     # 启发式函数的辅助, 返回 [[ 肯定的 precondition , add ]...]
76     def relax_do(self, objects, kb):
77         rst = []
78         mapping_list = []
79         self.create_mapping(objects, {}, 0, mapping_list)
80         for i in mapping_list:
81             precondition = self.assign_precondition(i)
82             flag = True
83             for j in precondition:
84                 if (j[0] and j[1] not in kb) or (not j[0] and j[1] in kb):
85                     flag = False
86                     break
87             if flag:
88                 pre = []

```

```

87         for j in range(len(precondition)):
88             if precondition[j][0]:
89                 pre.append(precondition[j][1])
90                 rst.append([pre, self.assign_effect(i)[1]])
91         return rst
92
93     # 创建对象间的映射以实现实例化，返回映射的列表
94     def create_mapping(self, objects, curr, depth, rst):
95         if depth == len(self.parameters):
96             rst.append(copy.deepcopy(curr))
97             return
98         ind, val = list(self.parameters.items())[depth]
99         for i in objects[val]:
100             if i not in curr.values(): # 这里默认一个动作参数中不能出现相同对象
101                 curr[ind] = i
102                 self.create_mapping(objects, curr, depth+1, rst)
103                 curr.pop(ind)
104
105     # 实例化条件
106     def assign_precondition(self, mapping):
107         precondition = copy.deepcopy(self.precondition)
108         for i in range(len(precondition)):
109             for j in range(1, len(precondition[i][1])):
110                 precondition[i][1][j] = mapping[precondition[i][1][j]]
111         return precondition
112
113     # 实例化效果
114     def assign_effect(self, mapping):
115         effect = copy.deepcopy(self.effect)
116         add = []
117         delete = []
118         for i in range(len(effect)):
119             for j in range(1, len(effect[i][1])):
120                 effect[i][1][j] = mapping[effect[i][1][j]]
121             if effect[i][0]:
122                 add.append(effect[i][1])
123             else:

```

```

124         delete.append(effect[i][1])
125     return [[self.name]+list(mapping.values()), add, delete]
126
127     def printInfo(self):
128         print("\t{" + self.name + "}")
129         print("\tparameters:", self.parameters)
130         print("\tprecondition:", self.precondition)
131         print("\teffect:", self.effect)
132
133
134     class Strips:
135         def __init__(self, domain_filename, problem_filename):
136             domain = parser(domain_filename)
137             problem = parser(problem_filename)
138             self.types = ['AnyType'] # 没有类型时
139             self.actions = []
140             for i in domain:
141                 if type(i) is list:
142                     if i[0] == ':types':
143                         self.types = i[1:]
144                     if i[0] == ':action':
145                         self.actions.append(Action(i, self.types))
146             self.objects = {}
147             object = problem[3][1:]
148             tail = 0
149             if self.types != ['AnyType']:
150                 for ind, i in enumerate(object):
151                     if i == '-':
152                         self.objects[object[ind+1]] = object[tail:ind]
153                         tail = ind + 2
154             else:
155                 self.objects['AnyType'] = object
156             self.kb = problem[4][1:]
157             self.goal = []
158             goal = problem[5][1][1:]
159             for i in goal:
160                 if i[0] != 'not':

```

```

161         self.goal.append([True, i])
162     else:
163         self.goal.append([False, i[1:]])
164     self.path = []
165
166     # 使用 A* 搜索解决问题, 返回拓展结点数和路径
167     def search(succ):
168         if len(succ) == 0:
169             return 1, []
170         state, action, effect = min(succ, key=lambda x: x[2][3] + x[2][4])
171         if effect[3] > MAX_DEPTH: # 限制深度, 在所给测例中其实没用
172             return 1, []
173         succ.remove((state, action, effect))
174         state.step(effect)
175         if state.achive():
176             return 1, state.path
177         for i in state.observe():
178             if not Strips.inverse(effect, i[2]): # 阻止搜索在两点间反复踏步
179                 succ.append(i)
180             # succ.append(i)
181
182         rst = Strips.search(succ)
183         return rst[0] + 1, rst[1]
184
185     # 观察周围, 获取动作信息, 返回 [( 状态 , 动作 , [ 动作
186     # 名 , add_list , del_list , g, h]) ...]
187     def observe(self):
188         tem = []
189         for i in self.actions:
190             for can in i.do(self.objects, self.kb):
191                 can.append(len(self.path)) # g
192                 h = self.heuristic(can)
193                 can.append(self.heuristic(can)) # h
194                 tem.append((copy.deepcopy(self), i, can))
195         return tem
196
197     # 应用 reachability_analysis 和 count_actions , 返回h
198     def heuristic(self, action):

```

```

198     curr_state = copy.deepcopy(self)
199     curr_state.step(action)
200     state_layer = [copy.deepcopy(curr_state.kb)]
201     action_layer = []
202     new_to_pre = {}
203     while(not curr_state.achive()):
204         actions = []
205         for i in curr_state.actions:
206             for can in i.relax_do(curr_state.objects, curr_state.kb):
207                 can.append(i)
208                 actions.append(can) # (pre, add, action)
209     action_layer.append(copy.deepcopy(actions))
210     for i in actions:
211         for j in i[1]:
212             if j not in curr_state.kb:
213                 new_to_pre[str(tuple(j))] = i
214                 curr_state.kb.append(j)
215             if curr_state.kb == state_layer[-1]: # kb 不再变化
216                 return 100
217     state_layer.append(copy.deepcopy(curr_state.kb))
218     state_layer = [ [tuple(j) for j in i] for i in state_layer]
219
220     return Strips.count_acts(new_to_pre, state_layer, action_layer,
221                               curr_state.goal, state_layer[-1])
222
223 # 利用递归来计算 h
224 def count_acts(new_to_pre, state_layer, action_layer, G, S):
225     if len(state_layer) == 1:
226         return 0
227     state_layer.remove(S)
228     s1 = set(S)
229     s2 = set(state_layer[-1])
230     Gp = s1 & s2
231     Gn = s1 - s2
232     A = []
233     for i in Gn:
234         action = new_to_pre[str(i)]

```



```

234         if action[2] not in A:
235             A.append(action[2])
236             Gp = Gp | set([tuple(j) for j in action[0]])
237         return Strips.count_acts(new_to_pre, state_layer, action_layer, Gp,
238                                state_layer[-1]) + len(A)
239
240 # 行动
241 def step(self, action):
242     self.path.append(action[0])
243     for i in action[1]: # add
244         self.kb.append(i)
245     for i in action[2]: # del
246         self.kb.remove(i)
247
248 # 判断是否达到目标
249 def achive(self):
250     for i in self.goal:
251         if not i[0]: # 否定的目标
252             for j in self.kb:
253                 if j == i[1]:
254                     return False
255         else: # 肯定的目标
256             for j in self.kb:
257                 find = False
258                 if j == i[1]:
259                     find = True
260                 break
261             if not find:
262                 return False
263     return True
264
265 # 检查互逆动作以提升搜索效率
266 def inverse(effect1, effect2):
267     if effect1[0][0] != effect2[0][0]:
268         return False
269     for i in effect1[1]:
270         if i not in effect2[2]:

```

```

270         return False
271     for i in effect2[1]:
272         if i not in effect1[2]:
273             return False
274     return True
275
276 def printInfo(self):
277     print("{domain}")
278     print("types:", self.types)
279     print("actions:␣")
280     for curr in self.actions:
281         curr.printInfo()
282     print("{problem}")
283     print("objects:", self.objects)
284     print("kb:", self.kb)
285     print("goal:", self.goal)
286     print()
287
288 def test(num):
289     prefix = "pddl/test" + str(num) + "/test" + str(num)
290     strips = Strips(prefix + "_domain.pddl", prefix + "_problem.pddl")
291     begin = timer()
292     count, path = Strips.search(strips.observe())
293     end = timer()
294     length = len(path)
295     print("{test␣"+str(num)+"}")
296     print()
297     print("ANSWER:",end="␣")
298     for i in range(length):
299         print(path[i],end="->" if i != length-1 else "")
300         if (i+1) % 3 == 0 and i != length-1:
301             print("\n\t",end="")
302     print()
303     print("Time␣Cost:␣"+str(end-begin)+"s")
304     print("Node␣Expended:", count)
305     print()
306

```

```
307     def main():
308         for i in range(5):
309             test(i)
310
311     if __name__ == "__main__":
312         main()
```

- parser 是自己写的。

3. Explain any ideas you use to speed up the implementation. (10 points)

- 在使用 A* 搜索时，不仅是选出 $g+h$ 最小的动作，也会对动作进行筛选，不选和上次动作完全互逆的动作以防止“反复徘徊”。这是因为，可达性分析启发式函数不会删除知识，这导致搜索前期计算启发式函数时一些具有否定前提的动作永远不会被执行，这使得边界集合的启发式 h 值都很大，经常出现“反复徘徊”问题。而检查相邻动作是否完全互逆，可以解决这个问题。

4. Run your planner on the 5 test cases, and report the returned plans and the running times. Analyse the experimental results. (10 points)

- 结果如下：针对每个测例，显示了行动序列，解决问题的时间，搜索时拓展的节点数。测例 0,1,3,4 的问题规模都比较小，测例 0, 1 问题规模太小了，分别花了 2 步和 3 步直接搜到了结果，没有多走任何弯路，测例 3,4 也几乎是直接搜到了结果，都只多拓展了一个节点。测例 2 的目标路径有 10 步，拓展结点为 16 个，表现也很好，这里如果不使用“反复徘徊”检测，拓展结点数会稍微多一点，为 18 个。时间上，全部都是瞬间出结果，只有测例 2 稍微多一点，接近 1s。
- 总得说所给测例都较小，但也能看出 A* 搜索和可达性分析启发式效果拔群。

```
{test 0}

ANSWER: ['move', 'npc', 'town', 'field']->['move', 'npc', 'field', 'castle']
Time Cost: 0.0016088000000000005s
Node Expended: 2

{test 1}

ANSWER: ['move', 'npc', 'town', 'tunnel']->['move', 'npc', 'tunnel', 'river']->['move', 'npc', 'river', 'castle']
Time Cost: 0.0065931000000000045s
Node Expended: 3
```

```
{test 2}

ANSWER: ['move', 'npc', 'town', 'field']->['attack', 'npc', 'ogre', 'field', 'river']->['move', 'npc', 'field', 'river']->
        ['attack', 'npc', 'dragon', 'river', 'cave']->['open', 'npc', 'box1', 'river']->['collect-fire', 'npc', 'box1', 'river', 'reddust']->
        ['move', 'npc', 'river', 'cave']->['open', 'npc', 'box2', 'cave']->['collect-earth', 'npc', 'box2', 'cave', 'browndust']->
        ['build-fireball', 'npc']
Time Cost: 0.7458731999999999s
Node Expended: 16
```

```
{test 3}

ANSWER: ['unstack', 'b', 'a', 'x']->['move', 'b', 'x', 'y']->['stack', 'a', 'b', 'x']
Time Cost: 0.034036999999999984s
Node Expended: 4

{test 4}

ANSWER: ['unstack', 'b', 'a', 't1', 't3']->['stack', 'a', 't1', 'b', 't3']
Time Cost: 0.3862719000000001s
Node Expended: 3
```

2 Diagnosing by Bayesian Networks

2.1 Variables and their domains

- (1) PatientAge: ['0 – 30 ', '31 – 65 ', '65 + ']
- (2) CTScanResult: ['Ischemic Stroke ', 'Hemorrhagic Stroke ']
- (3) MRIScanResult: ['Ischemic Stroke ', 'Hemorrhagic Stroke ']
- (4) StrokeType: ['Ischemic Stroke ', 'Hemorrhagic Stroke ', 'Stroke Mimic ']
- (5) Anticoagulants: ['Used ', 'Not used ']
- (6) Mortality: ['True ', 'False ']
- (7) Disability: ['Negligible ', 'Moderate ', 'Severe ']

2.2 CPTs

Note: [CTScanResult, MRIScanResult, StrokeType] means:

$P(\text{StrokeType} = \dots \mid \text{CTScanResult} = \dots \wedge \text{MRIScanResult} = \dots)$

(1)

[PatientAge]

['0 – 30 ', 0.10],

['31 – 65 ', 0.30],

['65+', 0.60]

(2)

[CTScanResult]

['Ischemic Stroke ', 0.7],

['Hemorrhagic Stroke ', 0.3]

(3)

[MRIScanResult]

['Ischemic Stroke ', 0.7],

['Hemorrhagic Stroke ', 0.3]

(4)

[Anticoagulants]

[Used ', 0.5],

['Not used ', 0.5]

(5)

[CTScanResult, MRIScanResult, StrokeType])

['Ischemic Stroke ', 'Ischemic Stroke ', 'Ischemic Stroke ', 0.8],

['Ischemic Stroke ', 'Hemorrhagic Stroke ', 'Ischemic Stroke ', 0.5],

['Hemorrhagic Stroke ', 'Ischemic Stroke ', 'Ischemic Stroke ', 0.5],

['Hemorrhagic Stroke ', 'Hemorrhagic Stroke ', 'Ischemic Stroke ', 0],

['Ischemic Stroke ', 'Ischemic Stroke ', 'Hemorrhagic Stroke ', 0],

['Ischemic Stroke ', 'Hemorrhagic Stroke ', 'Hemorrhagic Stroke ', 0.4],

['Hemorrhagic Stroke ', 'Ischemic Stroke ', 'Hemorrhagic Stroke ', 0.4],

['Hemorrhagic Stroke ', 'Hemorrhagic Stroke ', 'Hemorrhagic Stroke ', 0.9],

```
[ 'Ischemic Stroke ', 'Ischemic Stroke ', 'Stroke Mimic ', 0.2 ],
[ 'Ischemic Stroke ', 'Hemorrhagic Stroke ', 'Stroke Mimic ', 0.1 ],
[ 'Hemorrhagic Stroke ', 'Ischemic Stroke ', 'Stroke Mimic ', 0.1 ],
[ 'Hemorrhagic Stroke ', 'Hemorrhagic Stroke ', 'Stroke Mimic ', 0.1 ],
```

(6)

```
[StrokeType, Anticoagulants, Mortality]
```

```
[ 'Ischemic Stroke ', 'Used ', 'False ', 0.28 ],
[ 'Hemorrhagic Stroke ', 'Used ', 'False ', 0.99 ],
[ 'Stroke Mimic ', 'Used ', 'False ', 0.1 ],
[ 'Ischemic Stroke ', 'Not used ', 'False ', 0.56 ],
[ 'Hemorrhagic Stroke ', 'Not used ', 'False ', 0.58 ],
[ 'Stroke Mimic ', 'Not used ', 'False ', 0.05 ],
```

```
[ 'Ischemic Stroke ', 'Used ', 'True ', 0.72 ],
[ 'Hemorrhagic Stroke ', 'Used ', 'True ', 0.01 ],
[ 'Stroke Mimic ', 'Used ', 'True ', 0.9 ],
[ 'Ischemic Stroke ', 'Not used ', 'True ', 0.44 ],
[ 'Hemorrhagic Stroke ', 'Not used ', 'True ', 0.42 ],
[ 'Stroke Mimic ', 'Not used ', 'True ', 0.95 ]
```

(7)

```
[StrokeType, PatientAge, Disability]
```

```
[ 'Ischemic Stroke ', '0-30 ', 'Negligible ', 0.80 ],
[ 'Hemorrhagic Stroke ', '0-30 ', 'Negligible ', 0.70 ],
[ 'Stroke Mimic ', '0-30 ', 'Negligible ', 0.9 ],
[ 'Ischemic Stroke ', '31-65 ', 'Negligible ', 0.60 ],
[ 'Hemorrhagic Stroke ', '31-65 ', 'Negligible ', 0.50 ],
[ 'Stroke Mimic ', '31-65 ', 'Negligible ', 0.4 ],
[ 'Ischemic Stroke ', '65+ ', 'Negligible ', 0.30 ],
[ 'Hemorrhagic Stroke ', '65+ ', 'Negligible ', 0.20 ],
```

```
[ 'Stroke Mimic' ,           '65+'   , 'Negligible' , 0.1] ,

[ 'Ischemic Stroke' ,       '0-30'   , 'Moderate' , 0.1] ,
[ 'Hemorrhagic Stroke' ,   '0-30'   , 'Moderate' , 0.2] ,
[ 'Stroke Mimic' ,         '0-30'   , 'Moderate' , 0.05] ,
[ 'Ischemic Stroke' ,       '31-65'  , 'Moderate' , 0.3] ,
[ 'Hemorrhagic Stroke' ,   '31-65'  , 'Moderate' , 0.4] ,
[ 'Stroke Mimic' ,         '31-65'  , 'Moderate' , 0.3] ,
[ 'Ischemic Stroke' ,       '65+'   , 'Moderate' , 0.4] ,
[ 'Hemorrhagic Stroke' ,   '65+'   , 'Moderate' , 0.2] ,
[ 'Stroke Mimic' ,         '65+'   , 'Moderate' , 0.1] ,

[ 'Ischemic Stroke' ,       '0-30'   , 'Severe' , 0.1] ,
[ 'Hemorrhagic Stroke' ,   '0-30'   , 'Severe' , 0.1] ,
[ 'Stroke Mimic' ,         '0-30'   , 'Severe' , 0.05] ,
[ 'Ischemic Stroke' ,       '31-65'  , 'Severe' , 0.1] ,
[ 'Hemorrhagic Stroke' ,   '31-65'  , 'Severe' , 0.1] ,
[ 'Stroke Mimic' ,         '31-65'  , 'Severe' , 0.3] ,
[ 'Ischemic Stroke' ,       '65+'   , 'Severe' , 0.3] ,
[ 'Hemorrhagic Stroke' ,   '65+'   , 'Severe' , 0.6] ,
[ 'Stroke Mimic' ,         '65+'   , 'Severe' , 0.8]
```

2.3 Tasks

1. Briefly describe with sentences the main ideas of the VE algorithm. (10 points)

- 为了解决计算条件概率时复杂度 $O(2^{n-k})$ 问题 (n 为总变量数, k 为所求变量数), 使用 DP 想法, 逐层计算条件概率, 每层消除一个变量, 将问题复杂度降至 $O(2^{O(w)})$ (w 为贝叶斯网络的树宽, $w \ll n$)

2. Implement the VE algorithm (C++ or Python) to calculate the following probability values: (10 points)

- $p1 = P(\text{Mortality} = \text{'True'} \wedge \text{CTScanResult} = \text{'Ischemic Stroke'} \mid \text{PatientAge} = \text{'31-65'})$
- $p2 = P(\text{Disability} = \text{'Moderate'} \wedge \text{CTScanResult} = \text{'Hemorrhagic Stroke'} \mid \text{PatientAge} = \text{'65+'} \wedge \text{MRIScanResult} = \text{'Hemorrhagic Stroke'})$

(c) $p3 = P(\text{StrokeType}=\text{'Hemorrhagic Stroke'} \mid \text{PatientAge}=\text{'65+'} \wedge \text{CTScanResult}=\text{'Hemorrhagic Stroke'} \wedge \text{MRIScanResult}=\text{'Ischemic Stroke'})$

(d) $p4 = P(\text{Anticoagulants}=\text{'Used'} \mid \text{PatientAge}=\text{'31-65'})$

(e) $p5 = P(\text{Disability}=\text{'Negligible'})$

```

1  import copy
2  import random
3  import numpy as np
4  from timeit import default_timer as timer
5  num = 0
6
7  class VariableElimination:
8      def printFactors(factorList):
9          for factor in factorList:
10             factor.printInf()
11
12     def default_order(list, factorList):
13         hold = []
14         for i in list:
15             hold.append(i)
16             yield i
17         print("default_order:", hold)
18
19     def random_order(list, factorList):
20         length = len(list)
21         hold = []
22         for i in range(length):
23             pick = random.choice(list)
24             list.remove(pick)
25             hold.append(pick)
26             yield pick
27         print("random_order:", hold)
28
29     def min_fill(list, factorList):
30         pass
31
32     def min_neighbor(list, factorList):

```



```

33         length = len(list)
34         hold = []
35         for i in range(length):
36             var_list = [j.varList for j in factorList if len(j.varList)]
37             pick = random.choice(min(var_list, key=len))
38             hold.append(pick)
39             yield pick
40         print("min-neighbor:", hold)
41
42     def min_weight(list, factorList):
43         length = len(list)
44         hold = []
45         for i in range(length):
46             var_map = dict(zip(list, [0]*(length-i)))
47             for node in factorList:
48                 for var in node.varList:
49                     if var in list:
50                         var_map[var] += 1
51             pick = min(var_map.items(), key=lambda x: x[1])[0]
52             hold.append(pick)
53             list.remove(pick)
54             yield pick
55         print("min-weight:", hold)
56
57     def inference(factorList, queryVariables,
58                 orderedListOfHiddenVariables, evidenceList, queryVal, f):
59         for ev in evidenceList:
60             #Your code here
61             for pos, factor in enumerate(factorList):
62                 factorList[pos] = factor.restrict(ev, str(evidenceList[ev]))
63                 # 注意这里需要转
64                 # 成 str
65             print()
66         width = 0
67         for var in f(orderedListOfHiddenVariables, factorList):
68             #Your code here
69             temList= []
70             for pos, factor in enumerate(factorList):

```

```

69         if var in factor.varList:
70             temList.append(factor)
71         for factor in temList:
72             factorList.remove(factor)
73         curr = temList[0]
74         for i in temList[1:]:
75             curr = curr.multiply(i)
76         curr = curr.sumout(var)
77         factorList.append(curr)
78         if factorList != []:
79             # 消除宽为最大超图大小
80             max_width = max([len(i.varList) for i in factorList])
81             width = width if max_width <= width else max_width
82         print("WIDTH:", width)
83         print("RESULT:")
84         res = factorList[0]
85         for factor in factorList[1:]:
86             res = res.multiply(factor)
87         total = sum(res.cpt.values())
88         res.cpt = {k: v/total for k, v in res.cpt.items()}
89         # res.printInf()
90         print(res.cpt[queryVal])
91
92     class Util:
93         def lfind(l, val):
94             for pos, i in enumerate(l):
95                 if i==val:
96                     return pos
97             return -1
98
99     class Node:
100         def __init__(self, name, var_list):
101             self.name = name
102             self.varList = var_list
103             self.cpt = {}
104         def setCpt(self, cpt):
105             self.cpt = cpt

```

```

106     def printInf(self):
107         print("Name_=" + self.name)
108         print("_vars_" + str(self.varList))
109         for key in self.cpt:
110             print("___key:_" + str(key) + "_val_:_" + str(self.cpt[key]))
111         print()
112     def multiply(self, factor):
113         """function that multiplies with another factor"""
114         #Your code here
115         newList = list(set(self.varList+factor.varList))
116         length = len(newList)
117         new_cpt = {}
118         if self.varList == []:
119             for key, val in factor.cpt.items():
120                 new_cpt[key] = val*self.cpt['']
121         elif factor.varList == []:
122             for key, val in self.cpt.items():
123                 new_cpt[key] = val*factor.cpt['']
124         else:
125             map1 = {}
126             map2 = {}
127             for pos, i in enumerate(self.varList):
128                 map1[pos] = Util.lfind(newList, i)
129             for pos, i in enumerate(factor.varList):
130                 map2[pos] = Util.lfind(newList, i)
131             convex = list(set(self.varList) & set(factor.varList))
132             test = convex != []
133             if test:
134                 repeated_var = convex[0]
135                 pos1 = Util.lfind(self.varList, repeated_var)
136                 pos2 = Util.lfind(factor.varList, repeated_var)
137
138             for key1, val1 in self.cpt.items():
139                 if test:
140                     repeated_ins = key1[pos1]
141                 for key2, val2 in factor.cpt.items():
142                     if test and repeated_ins == key2[pos2]:

```

```

143         key = np.zeros(length, dtype=int)
144         for ind, i in map1.items():
145             key[i] = key1[ind]
146         for ind, i in map2.items():
147             key[i] = key2[ind]
148         new_cpt["".join([str(num) for num in key])] = val1 *
            val2
149     new_node = Node("f" + str(newList), newList)
150     new_node.setCpt(new_cpt)
151     return new_node
152 def sumout(self, variable):
153     """function that sums out a variable given a factor"""
154     #Your code here
155     if variable not in self.varList:
156         return self
157     pos = Util.lfind(self.varList, variable)
158     node_list = []
159     for i in set([j[pos] for j in self.cpt]):
160         node_list.append(self.restrict(variable, i))
161     new_var_list = copy.deepcopy(node_list[0].varList)
162     new_cpt = copy.deepcopy(node_list[0].cpt)
163     for node in node_list[1:]:
164         for key, p in node.cpt.items():
165             new_cpt[key] += node.cpt[key]
166
167     new_node = Node("f" + str(new_var_list), new_var_list)
168     new_node.setCpt(new_cpt)
169     return new_node
170 def restrict(self, variable, value):
171     """function that restricts a variable to some value
172     in a given factor"""
173     #Your code here
174     if variable not in self.varList:
175         return self
176     new_var_list = copy.deepcopy(self.varList) # 这里必须要用深复制
177     new_var_list.remove(variable)
178     length = len(new_var_list)

```

```

179         new_cpt = {}
180         pos = Util.lfind(self.varList, variable)
181         for key, p in self.cpt.items():
182             if key[pos] == value:
183                 key = key[0:pos]+key[pos+1:]
184                 new_cpt[key] = p
185         new_node = Node("f" + str(new_var_list), new_var_list)
186         new_node.setCpt(new_cpt)
187         return new_node
188
189     def initBN():
190         # create nodes for Bayes Net
191         P = Node("PatientAge", ["P"])
192         C = Node("CTScanResult", ["C"])
193         MR = Node("MRIScanResult", ["MR"])
194         A = Node("Anticoagulants", ["A"])
195         S = Node("StrokeType", ["S", "C", "MR"])
196         MO = Node("Mortality", ["MO", "S", "A"])
197         D = Node("Disability", ["D", "S", "P"])
198
199         # Generate cpt for each node
200         P.setCpt({'0': 0.1, '1': 0.3, '2': 0.6})
201         C.setCpt({'0': 0.7, '1': 0.3})
202         MR.setCpt({'0': 0.7, '1': 0.3})
203         A.setCpt({'0': 0.5, '1': 0.5})
204         S.setCpt({'000': 0.8, '001': 0.5, '010': 0.5, '011': 0.0, \
205                 '100': 0.0, '101': 0.4, '110': 0.4, '111': 0.9, \
206                 '200': 0.2, '201': 0.1, '210': 0.1, '211': 0.1})
207         MO.setCpt({'000': 0.28, '010': 0.99, '020': 0.1, '001': 0.56, '011':
208                 0.58, '021': 0.05, \
209                 '100': 0.72, '110': 0.01, '120': 0.9, '101': 0.44, '111':
210                 0.42, '121': 0.95})
211         D.setCpt({'000': 0.80, '010': 0.70, '020': 0.90, '001': 0.60, '011':
212                 0.50, '021': 0.40, '002': 0.30, '012': 0.20,
213                 '022': 0.10, \
214                 '100': 0.10, '110': 0.20, '120': 0.05, '101': 0.30, '111':
215                 0.40, '121': 0.30, '102': 0.40, '112': 0.20,

```

```

212         '122': 0.10, \
213         '200': 0.10, '210': 0.10, '220': 0.05, '201': 0.10, '211':
            0.10, '221': 0.30, '202': 0.30, '212': 0.60,
214         '222': 0.80})
215     return [P, C, MR, A, S, MO, D]
216
217 def test(factor_list, f):
218     global num
219     print("ORDER_"+str(num)+":")
220     num += 1
221     begin1 = timer()
222     VariableElimination.inference(copy.deepcopy(factor_list), ['A'], ['C', '
        MR', 'S', 'MO', 'D'], {'P': 1}, '1', f)
223     end1 = timer()
224     print("time1:", (end1 - begin1)*1000, "ms")
225     begin2 = timer()
226     VariableElimination.inference(copy.deepcopy(factor_list), ['D'], ['P', '
        C', 'MR', 'A', 'S', 'MO'], {}, '0', f)
227     end2 = timer()
228     print("time2:", (end2 - begin2)*1000, "ms")
229     print()
230     print()
231
232 def main():
233     factor_list = initBN()
234
235     # test
236     VariableElimination.inference(copy.deepcopy(factor_list), ['MO', 'C'], [
        'MR', 'A', 'S', 'D'], {'P': 1}, '10', VariableElimination.
        default_order)
237     VariableElimination.inference(copy.deepcopy(factor_list), ['D', 'C'], ['
        A', 'S', 'MO'], {'P': 2, 'MR': 1}, '11', VariableElimination.
        default_order)
238     VariableElimination.inference(copy.deepcopy(factor_list), ['S'], ['A', '
        MO', 'D'], {'P': 2, 'C': 1, 'MR': 0}, '1', VariableElimination.
        default_order)
239     VariableElimination.inference(copy.deepcopy(factor_list), ['A'], ['C', '

```

```

        'MR', 'S', 'MO', 'D'], {'P': 1}, '1', VariableElimination.
        default_order)
240 VariableElimination.inference(copy.deepcopy(factor_list), ['D'], ['P', '
        C', 'MR', 'A', 'S', 'MO'], {}, '0', VariableElimination.default_order
        )
241 # test(factor_list, VariableElimination.default_order)
242 # test(factor_list, VariableElimination.random_order)
243 # test(factor_list, VariableElimination.random_order)
244 # test(factor_list, VariableElimination.random_order)
245 # test(factor_list, VariableElimination.min_neighbor)
246 # test(factor_list, VariableElimination.min_weight)
247
248
249 if __name__ == "__main__":
250     main()

```

```

default order: ['MR', 'A', 'S', 'D']
WIDTH: 3
RESULT:
0.17587499999999995

default order: ['A', 'S', 'MO']
WIDTH: 3
RESULT:
0.057

default order: ['A', 'MO', 'D']
WIDTH: 2
RESULT:
0.39999999999999997

default order: ['C', 'MR', 'S', 'MO', 'D']
WIDTH: 3
RESULT:
0.5000000000000001

default order: ['P', 'C', 'MR', 'A', 'S', 'MO']
WIDTH: 3
RESULT:
0.38977

```

- 该代码基于实验九的框架。上图是在人为给定变量消除顺序下的测试，计算出了正确结果。

3. Implement an algorithm to select a good order of variable elimination. (10 points)

- 见上方代码 VariableElimination 类内的 4 个 order 函数，同时需要把 main 内的原测试注释，将原注释取消注释进行 order 相关的测试。

4. Compare the running times of the VE algorithm for different orders of variable elimination, and fill out the following table: For test cases p4 and p5, for each of the order selected by your algorithm and 5 other orders, report the elimination with, and the total running time of the VE algorithm. For each case, the first order of elimination should be the one chosen by your algorithm. Analyze the results. (20 points)

- 测试结果如下：

```
ORDER 0:

default order: ['C', 'MR', 'S', 'MO', 'D']
WIDTH: 3
RESULT:
0.5
time1: 6.702700000000061 ms

default order: ['P', 'C', 'MR', 'A', 'S', 'MO']
WIDTH: 3
RESULT:
0.38977
time2: 5.2714999999999845 ms

ORDER 1:

random order: ['MR', 'D', 'S', 'MO', 'C']
WIDTH: 3
RESULT:
0.5
time1: 5.010199999999965 ms

random order: ['MR', 'A', 'S', 'C', 'MO', 'P']
WIDTH: 4
RESULT:
0.3897700000000006
time2: 12.750799999999952 ms
```

```
ORDER 2:

random order: ['D', 'MO', 'MR', 'S', 'C']
WIDTH: 3
RESULT:
0.5
time1: 3.326199999999946 ms

random order: ['S', 'P', 'MO', 'C', 'A', 'MR']
WIDTH: 6
RESULT:
0.38977
time2: 28.335099999999947 ms

ORDER 3:

random order: ['D', 'S', 'MO', 'C', 'MR']
WIDTH: 4
RESULT:
0.5
time1: 5.089499999999969 ms

random order: ['MO', 'A', 'P', 'S', 'MR', 'C']
WIDTH: 3
RESULT:
0.38977
time2: 3.269999999999995 ms
```



```

ORDER 4:

min-neighbor: ['C', 'MR', 'A', 'S', 'MO']
WIDTH: 3
RESULT:
0.3249
time1: 2.4287999999998977 ms

min-neighbor: ['P', 'C', 'MR', 'A', 'S', 'MO']
WIDTH: 3
RESULT:
0.38977
time2: 2.9660999999999715 ms

ORDER 5:

min-weight: ['MO', 'D', 'C', 'MR', 'S']
WIDTH: 3
RESULT:
0.5
time1: 1.795499999999995 ms

min-weight: ['MO', 'P', 'C', 'MR', 'A', 'S']
WIDTH: 3
RESULT:
0.38977
time2: 2.3180999999999896 ms

```

Test case	Elimination order	Elimination width	Total time(ms)
p4	[C,MR,S,MO,D]	3	6.7
p4	[MR,D,S,MO,C]	3	5.0
p4	[D,MO,MR,S,C]	3	3.2
p4	[D,S,MO,C,MR]	4	5.1
p4	[C,MR,A,S,MR]	3	2.4
p4	[MO,D,C,MR,S]	3	1.8
p5	[P,C,MR,A,S,MO]	3	5.3
p5	[MR,S,A,C,MO,P]	4	12.8
p5	[S,P,MO,C,A,MR]	6	28.3
p5	[MO,A,P,S,MR,C]	3	3.3
p5	[P,C,MR,A,S,MO]	3	3.0
p5	[MO,P,C,MR,A,S]	3	2.3

- 如图和表所示，对测例 4,5 使用了 6 种变量消除顺序，第一种是默认给出的顺序，第二至第四种顺序是随机顺序，第五种顺序是优先消除邻居数小的节点的变量，第六种顺序是优

先消除在图中节点所占权重小的变量。

- 总体上，消除宽度越大，运行时间越久。
- 两种算法（最少邻居，最小权重）都取得了不错的效果，最小权重算法获得了最小的时间。

3 Due: 11:59pm, Saturday, Nov. 28, 2020

Please hand in a file named P03_YourNumber.pdf, and send it to ai_2020@foxmail.com