

# E09 Variable Elimination

---

18364066 Yanzuo Lu

November 10, 2020

## Contents

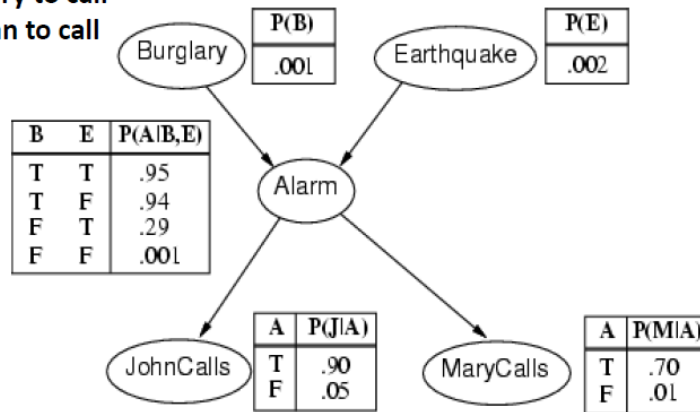
<b>1</b>	<b>VE</b>	<b>2</b>
<b>2</b>	<b>Task</b>	<b>5</b>
<b>3</b>	<b>Codes and Results</b>	<b>5</b>

# 1 VE

The burglary example is described as following:

- A burglary can set the alarm off
- An earthquake can set the alarm off
- The alarm can cause Mary to call
- The alarm can cause John to call

Note that these tables only provide the probability that  $X_i$  is true.  
(E.g.,  $\Pr(A \text{ is true} | B, E)$ )  
The probability that  $X_i$  is false is 1- these values



```
P(Alarm) =
0.002516442

P(J&&~M) =
0.050054875461

P(A | J&&~M) =
0.0135738893313

P(B | A) =
0.373551228282

P(B | J&&~M) =
0.0051298581334

P(J&&~M | ~B) =
0.049847949
```

Here is a VE template for you to solve the burglary example:

```
class VariableElimination:
    @staticmethod
    def inference(factorList, queryVariables,
        orderedListOfHiddenVariables, evidenceList):
        for ev in evidenceList:
            #Your code here
        for var in orderedListOfHiddenVariables:
            #Your code here
        print "RESULT:"
```

```

        res = factorList[0]
        for factor in factorList[1:]:
            res = res.multiply(factor)
        total = sum(res.cpt.values())
        res.cpt = {k: v/total for k, v in res.cpt.items()}
        res.printInf()

    @staticmethod
    def printFactors(factorList):
        for factor in factorList:
            factor.printInf()

class Util:
    @staticmethod
    def to_binary(num, len):
        return format(num, '0' + str(len) + 'b')

class Node:
    def __init__(self, name, var_list):
        self.name = name
        self.varList = var_list
        self.cpt = {}

    def setCpt(self, cpt):
        self.cpt = cpt

    def printInf(self):
        print "Name = " + self.name
        print " vars " + str(self.varList)
        for key in self.cpt:
            print "    key: " + key + " val : " + str(self.cpt[key])
        print ""

    def multiply(self, factor):
        """function that multiplies with another factor"""
        #Your code here
        new_node = Node("f" + str(newList), newList)
        new_node.setCpt(new_cpt)
        return new_node

```

```

def sumout(self, variable):
    """function that sums out a variable given a factor"""
    #Your code here
    new_node = Node("f" + str(new_var_list), new_var_list)
    new_node.setCpt(new_cpt)
    return new_node

def restrict(self, variable, value):
    """function that restricts a variable to some value
    in a given factor"""
    #Your code here
    new_node = Node("f" + str(new_var_list), new_var_list)
    new_node.setCpt(new_cpt)
    return new_node

# create nodes for Bayes Net
B = Node("B", ["B"])
E = Node("E", ["E"])
A = Node("A", ["A", "B", "E"])
J = Node("J", ["J", "A"])
M = Node("M", ["M", "A"])

# Generate cpt for each node
B.setCpt({'0': 0.999, '1': 0.001})
E.setCpt({'0': 0.998, '1': 0.002})
A.setCpt({'111': 0.95, '011': 0.05, '110':0.94, '010':0.06,
'101':0.29, '001':0.71, '100':0.001, '000':0.999})
J.setCpt({'11': 0.9, '01': 0.1, '10': 0.05, '00': 0.95})
M.setCpt({'11': 0.7, '01': 0.3, '10': 0.01, '00': 0.99})

print "P(A) *****"
VariableElimination.inference([B,E,A,J,M], ["A"], ["B", "E", "J", "M"],
    {}

print "P(B | J~M) *****"

```

```
VariableElimination.inference([B,E,A,J,M], ['B'], ['E','A'], {'J':1, 'M':0})
```

## 2 Task

- You should implement 4 functions: `inference`, `multiply`, `sumout` and `restrict`. You can turn to Figure 1 and Figure 2 for help.
- Please hand in a file named `E09_YourNumber.pdf`, and send it to `ai_2020@foxmail.com`

### The VE Algorithm

Given a Bayes Net with CPTs  $F$ , query variable  $Q$ , evidence variables  $E$  (observed to have values  $e$ ), and remaining variables  $Z$ . Compute  $\Pr(Q|E)$

- Replace each factor  $f \in F$  that mentions a variable(s) in  $E$  with its restriction  $f_{E=e}$  (this might yield a "constant" factor)
- For each  $Z_j$  in the order given –eliminate  $Z_j \in Z$  as follows:
  - Let  $f_1, f_2, \dots, f_k$  be the factors in  $F$  that include  $Z_j$
  - Compute new factor  $g_j = \sum_{Z_j} f_1 \times f_2 \times \dots \times f_k$
  - Remove the factors  $f_i$  from  $F$  and add new factor  $g_j$  to  $F$
- The remaining factors refer only to the query variable  $Q$ . Take their product and normalize to produce  $\Pr(Q|E)$ .

### The Product of Two Factors

- Let  $f(\mathbf{X}, \mathbf{Y})$  &  $g(\mathbf{Y}, \mathbf{Z})$  be two factors with variables  $\mathbf{Y}$  in common
- The **product** of  $f$  and  $g$ , denoted  $h = f \times g$  (or sometimes just  $h = fg$ ), is defined:

$$h(\mathbf{X}, \mathbf{Y}, \mathbf{Z}) = f(\mathbf{X}, \mathbf{Y}) \times g(\mathbf{Y}, \mathbf{Z})$$

f(A,B)		g(B,C)		h(A,B,C)			
ab	0.9	bc	0.7	abc	0.63	ab~c	0.27
a~b	0.1	b~c	0.3	a~bc	0.08	a~b~c	0.02
~ab	0.4	~bc	0.8	~abc	0.28	~ab~c	0.12
~a~b	0.6	~b~c	0.2	~a~bc	0.48	~a~b~c	0.12

Figure 1: VE and Product

### Summing a Variable Out of a Factor

- Let  $f(\mathbf{X}, \mathbf{Y})$  be a factor with variable  $\mathbf{X}$  ( $\mathbf{Y}$  is a set)
- We **sum out** variable  $\mathbf{X}$  from  $f$  to produce a new factor  $h = \sum_{\mathbf{X}} f$ , which is defined:

$$h(\mathbf{Y}) = \sum_{\mathbf{X} \in \text{Dom}(\mathbf{X})} f(\mathbf{X}, \mathbf{Y})$$

f(A,B)		h(B)	
ab	0.9	b	1.3
a~b	0.1	~b	0.7
~ab	0.4		
~a~b	0.6		

No error in the table. Here  $f(A,B)$  is not  $P(AB)_v$  but  $P(B|A)$ .

### Restricting a Factor

- Let  $f(\mathbf{X}, \mathbf{Y})$  be a factor with variable  $\mathbf{X}$  ( $\mathbf{Y}$  is a set)
- We **restrict** factor  $f$  to  $\mathbf{X}=a$  by setting  $\mathbf{X}$  to the value  $a$  and "deleting" incompatible elements of  $f$ 's domain. Define  $h = f_{\mathbf{X}=a}$  as:  $h(\mathbf{Y}) = f(a, \mathbf{Y})$

f(A,B)		h(B) = f_{A=a}	
ab	0.9	b	0.9
a~b	0.1	~b	0.1
~ab	0.4		
~a~b	0.6		

Figure 2: Sumout and Restrict

## 3 Codes and Results

```

1 from math import pow
2
3 class VariableElimination:
4     @staticmethod
5     def inference(factorList, queryVariables,
6                   orderedListOfHiddenVariables, evidenceList):
7         #Your code here
8         for i, factor in enumerate(factorList):
9             if ev not in factor.varList:
10                 continue
11             factorList[i] = factor.restrict(ev, evidenceList[ev])
12     for var in orderedListOfHiddenVariables:
13         #Your code here
14         index_list = []
15         for i, factor in enumerate(factorList):
16             if var in factor.varList:
17                 index_list.append(i)
18         new_factor_var = []
19         for i in index_list:
20             for factor_var in factorList[i].varList:
21                 if factor_var not in new_factor_var and factor_var
22                     != var:
23                     new_factor_var.append(factor_var)
24
25         cal_factor = Node("tmp", factorList[index_list[0]].varList.
26                             copy())
27         cal_factor.setCpt(factorList[index_list[0]].cpt.copy())
28         for i in index_list[1:]:
29             temp_factor = Node("tmp", factorList[i].varList.copy())
30             temp_factor.setCpt(factorList[i].cpt.copy())
31             cal_factor = cal_factor.multiply(temp_factor)

```

```

30
31         new_factor = cal_factor.sumout(var)
32         factorList.append(new_factor)
33         factorList = [factor for i, factor in enumerate(factorList)
34                        if i not in index_list]
35
36     print("RESULT:")
37     res = factorList[0]
38     for factor in factorList[1:]:
39         res = res.multiply(factor)
40     total = sum(res.cpt.values())
41     res.cpt = {k: v/total for k, v in res.cpt.items()}
42     res.printInf()
43
44     @staticmethod
45     def printFactors(factorList):
46         for factor in factorList:
47             factor.printInf()
48
49     class Util:
50         @staticmethod
51         def to_binary(num, len):
52             return format(num, '0' + str(len) + 'b')
53
54     class Node:
55         def __init__(self, name, var_list):
56             self.name = name
57             self.varList = var_list
58             self.cpt = {}
59
60         def setCpt(self, cpt):
61             self.cpt = cpt

```

```

62 def printInf(self):
63     print("Name = " + self.name)
64     print(" vars " + str(self.varList))
65     for key in self.cpt:
66         print("    key: " + key + " val : " + str(self.cpt[key]))
67     print("")
68
69 def multiply(self, factor):
70     """function that multiplies with another factor"""
71     #Your code here
72     newList = self.varList.copy()
73     not_same_index = []
74     same_index = []
75     for i, var in enumerate(factor.varList):
76         if var not in newList:
77             not_same_index.append(i)
78             newList.append(var)
79         else:
80             same_index.append((newList.index(var), i))
81
82     new_cpt = {}
83     for cpt1 in self.cpt:
84         for cpt2 in factor.cpt:
85             valid_flag = True
86             for si in same_index:
87                 if cpt1[si[0]] != cpt2[si[1]]:
88                     valid_flag = False
89                     break
90             if not valid_flag:
91                 continue
92
93             new_cpt_key = cpt1
94             for nsi in not_same_index:

```



```

95         new_cpt_key = new_cpt_key + cpt2[nsi]
96     if new_cpt_key in new_cpt:
97         new_cpt[new_cpt_key] += self.cpt[cpt1] * factor.cpt[
            cpt2]
98     else:
99         new_cpt[new_cpt_key] = self.cpt[cpt1] * factor.cpt[
            cpt2]
100
101     new_node = Node("f" + str(newList), newList)
102     new_node.setCpt(new_cpt)
103     return new_node
104
105 def sumout(self, variable):
106     """function that sums out a variable given a factor"""
107     #Your code here
108     var_index = self.varList.index(variable)
109     new_var_list = self.varList.copy()
110     new_var_list.remove(variable)
111
112     new_cpt = {}
113     for cpt1 in self.cpt:
114         new_cpt_key = ""
115         for i, ch in enumerate(cpt1):
116             if i != var_index:
117                 new_cpt_key = new_cpt_key + ch
118
119         if new_cpt_key in new_cpt:
120             new_cpt[new_cpt_key] += self.cpt[cpt1]
121         else:
122             new_cpt[new_cpt_key] = self.cpt[cpt1]
123
124     new_node = Node("f" + str(new_var_list), new_var_list)
125     new_node.setCpt(new_cpt)

```

```

126         return new_node
127
128     def restrict(self, variable, value):
129         """function that restricts a variable to some value
130         in a given factor"""
131         #Your code here
132         # print(self.varList, self.cpt)
133         var_index = self.varList.index(variable)
134         new_var_list = self.varList.copy()
135         new_var_list.remove(variable)
136
137         new_cpt = {}
138         for cpt1 in self.cpt:
139             if cpt1[var_index] != str(value):
140                 continue
141
142             new_cpt_key = ""
143             for i, ch in enumerate(cpt1):
144                 if i != var_index:
145                     new_cpt_key = new_cpt_key + ch
146
147             new_cpt[new_cpt_key] = self.cpt[cpt1]
148
149         # print(new_var_list, new_cpt)
150         new_node = Node("f" + str(new_var_list), new_var_list)
151         new_node.setCpt(new_cpt)
152         return new_node
153
154 # create nodes for Bayes Net
155 B = Node("B", ["B"])
156 E = Node("E", ["E"])
157 A = Node("A", ["A", "B", "E"])
158 J = Node("J", ["J", "A"])

```

```

159 M = Node("M", ["M", "A"])
160
161 # Generate cpt for each node
162 B.setCpt({'0': 0.999, '1': 0.001})
163 E.setCpt({'0': 0.998, '1': 0.002})
164 A.setCpt({'111': 0.95, '011': 0.05, '110':0.94, '010':0.06,
165 '101':0.29, '001':0.71, '100':0.001, '000':0.999})
166 J.setCpt({'11': 0.9, '01': 0.1, '10': 0.05, '00': 0.95})
167 M.setCpt({'11': 0.7, '01': 0.3, '10': 0.01, '00': 0.99})
168
169 print("P(A) *****")
170 VariableElimination.inference([B,E,A,J,M], ['A'], ['B', 'E', 'J', 'M'],
    {}))
171
172 print("P(B | J~M) *****")
173 VariableElimination.inference([B,E,A,J,M], ['B'], ['E', 'A'], {'J':1, 'M':0})

```

本次实验要求我们实现在 uncertainty 中提到的变量消除 VariableElimination 算法，用该算法解决贝叶斯网络中最基本的问题，给定一些已知的 Evidence 变量求 Query 变量的后验概率，而变量消除的算法核心就在于 Product 和 Sum 两个操作，操作具体在 PPT 中已经阐释地较为清楚。

在实现过程中，因子 factor 之间的三种操作 Product、Sum 和 Restrict 本质上是类似的，都是基于当前 factor 的变量列表 varlist 和条件概率表 cpt 计算一个新的 factor 的 varlist 和 cpt，计算新的 varlist 在 Product 中需要剔除掉重复的变量，在 Sum 和 Restrict 中则需要删掉给定的变量，计算新的 cpt 在 Product 中需要两两相乘并把相同的进行相加，在 Sum 中则不需要乘法直接将相同的相加即可，Restrict 中则在相加的基础上去除掉不符合给定 value 的条目。

在变量消除过程中依据这三种操作，注意维护 factorlist 的增删即可，每次消除一个变量都需要删除一些 factor 并把新的 factor 加进列表中，下面是运行结果。

```
PS C:\Users\YanzuoLu\OneDrive\人工智能\实验\E09_20201108_VE\src> python main.py
P(A) *****
RESULT:
Name = f['A']
vars ['A']
  key: 1 val : 0.0025164420000000002
  key: 0 val : 0.997483558

P(B | J~M) *****
RESULT:
Name = f['B']
vars ['B']
  key: 0 val : 0.9948701418665987
  key: 1 val : 0.0051298581334013015
```

Figure 3: result.png