

# 中山大学数据科学与计算机学院

## 人工智能本科生实验报告

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教学班级	16级计科二班	专业（方向）	计算机科学与技术
学号	16337334	姓名	周启恒

## 变量消元

### (1) 原理

#### 变量消除原理：

- 变量消除法通过利用联合概率分布的分解来降低推理的复杂度。

给定一个联合概率分布:  $P(A, B, C, D) = P(A)P(B|A)P(C|B)P(D|C)$

此时需要求  $P(B)$ ，由概率定义可得到以下公式

$$P(B) = \sum_A \sum_C \sum_D P(A)P(B|A)P(C|B)P(D|C)$$

考虑到A,C,D之间的相关关系，将不相关的概率求和分解，得到以下公式：

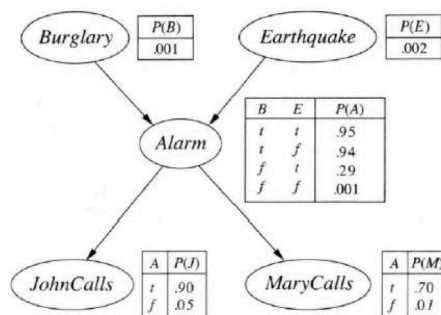
$$\begin{aligned} P(B) &= \sum_A P(A)P(B|A) \sum_C P(C|B) \sum_D P(D|C) \\ &= \sum_A P(A)P(B|A) \sum_C P(C|B) \end{aligned}$$

通过变量消除，计算复杂度大大降低，提升推断效率。

### (2) TASK关键代码（python带注释）

#### Task Burglary

- 网络图：



## 补全函数：

- **inference**：作出精确推断

- 先对evidence进行赋值(调用restrict函数)，替换成新的factor
- 对需要消除的变量依次操作
  - 对于含有要消除的变量(这里设为X)的factor，依次相乘(调用multiply函数)，并替换为新的factor
  - 将新的factor进行sum\_out操作，即把含有X的概率对应相加，消除X。

```
def inference(factor_list, query_variables,
              ordered_list_of_hidden_variables, evidence_list):
    #对证据进行限制
    for ev in evidence_list:
        # Your code here
        tmp_list = factor_list.copy()
        for i in range(len(factor_list)):
            if ev in factor_list[i].var_list:
                tmp_list.append(factor_list[i].restrict(ev,
evidence_list[ev]))
                tmp_list.remove(factor_list[i])
        factor_list = tmp_list
    #变量消除
    for var in ordered_list_of_hidden_variables:
        # Your code here
        first = False
        tmp_list = factor_list.copy()
        for i in range(len(factor_list)):
            if var in factor_list[i].var_list:
                tmp_list.remove(factor_list[i])
                if not first:
                    new_factor = factor_list[i]
                    first = True
                else: new_factor = new_factor.multiply(factor_list[i])
            tmp_list.append(new_factor.sum_out(var))
        factor_list = tmp_list

    print("RESULT: ")
```

```

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

```

- **same**: 判断两个factor的概率对应位置是否符合乘法的规则

```

def same(self, f1, f2, list1, list2):
    for i, j in zip(list1, list2):
        if f1[i] != f2[j]:
            return False
    return True

```

- **multiply**: 两个factor相乘

- 先求得两个factor含有的相同变量的列表
- 对于列表中的每个变量，在两个factor中找到概率表，对于每一项概率判断是否对应，进行相乘，生成新的概率。

$$f_3(A, B, C) = f_1(A, B) * f_2(B, C)$$

其中  $f_3(a, b, c) = f_1(a, b) * f_2(b, c)$ ，对应相乘。

```

def multiply(self, factor):
    '''function that multiplies with another factor'''
    # Your code here
    common = [i for i in self.var_list if i in factor.var_list]
    if not len(common): return self

    index1, index2 = [self.var_list.index(i) for i in common],
    [factor.var_list.index(i) for i in common]
    new_list = self.var_list + [i for i in factor.var_list if i not in
self.var_list]
    new_cpt = {}
    for i in self.cpt:
        for j in factor.cpt:
            if self.same(i, j, index1, index2):
                jStr = ''
                for s in range(len(j)):
                    if s not in index2: jStr += j[s]
                new_cpt[i + jStr] = self.cpt[i] * factor.cpt[j]
    new_node = Node('f' + str(new_list), new_list)
    new_node.set_cpt(new_cpt)
    return new_node

```

- **sum\_out**: 对factor中某一个变量进行求和，从而消去它。

- 假设需要消除的变量为A, 对 $f(A, B)$ :

$$f^*(B) = \sum_A f(a, B)$$

```
def sum_out(self, variable):
    '''function that sums out a variable given a factor'''
    # Your code here
    new_var_list = [j for j in self.var_list if j != variable]
    var_index = self.var_list.index(variable)
    new_cpt = {}
    for i in self.cpt:
        new_cpt[i[:var_index] + i[var_index+1:]] =
new_cpt.get(i[:var_index] + i[var_index+1:], 0) + self.cpt[i]
        new_node = Node('f' + str(new_var_list), new_var_list)
        new_node.set_cpt(new_cpt)
    return new_node
```

- **restrict**: 对变量进行赋值

- 假设需要消除的变量为A=a, 对 $f(A, B)$ :

$$f^*(B) = f(a, B)$$

```
def restrict(self, variable, value):
    '''function that restricts a variable to some value
    in a given factor'''
    # Your code here
    new_var_list = [j for j in self.var_list if j != variable]
    var_index = self.var_list.index(variable)
    new_cpt = {}
    for i in self.cpt:
        if i[var_index] == chr(value + ord('0')):
            new_cpt[i[:var_index] + i[var_index+1:]] = self.cpt[i]
    new_node = Node('f' + str(new_var_list), new_var_list)
    new_node.set_cpt(new_cpt.copy())
    return new_node
```

### (3) 结果展示

- 输出:

```

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})

```

- 结果:

```

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

```

```

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

```