

221310332 周立成的人工智能作业 3

1. 设有3个独立的结论 H_1, H_2, H_3 ,以及两个独立的证据 E_1, E_2 ,他们的先验概率和条件概率分别为:

$$\begin{array}{lll} P(H_1) = 0.4 & P(H_2) = 0.3 & P(H_3) = 0.3 \\ P(E_1|H_1) = 0.5 & P(E_1|H_2) = 0.3 & P(E_1|H_3) = 0.5 \\ P(E_2|H_1) = 0.7 & P(E_2|H_2) = 0.9 & P(E_2|H_3) = 0.1 \end{array}$$

试利用概率方法分别求出:

- (1):已知证据 E_1 出现时 $P(H_1|E_1), P(H_2|E_1), P(H_3|E_1)$ 的概率值;说明 E_1 的出现对结论 H_1, H_2, H_3 的影响;

解:

对于 $P(H_i|E_1)$,根据 贝叶斯公式: $P(H_i|E_1) \times P(E_1) = P(E_1|H_i) \times P(H_i)$

于是会有: $P(H_i|E_1) = \frac{P(E_1|H_i) \times P(H_i)}{P(E_1)}$

题目中已经给出 $P(E_1|H_i)$ 与 $P(H_i)$ 了,下面来求解 $P(E_1)$

根据 全概率公式,应该有 $P(E_1) = \sum_{i=1}^3 P(E_1|H_i) \times P(H_i)$

带入会有: $P(H_i|E_1) = \frac{P(E_1|H_i) \times P(H_i)}{\sum_{i=1}^3 P(E_1|H_i) \times P(H_i)}$

带入数值:
$$\begin{cases} P(H_1|E_1) = \frac{P(E_1|H_1) \times P(H_1)}{P(E_1|H_1) \times P(H_1) + P(E_1|H_2) \times P(H_2) + P(E_1|H_3) \times P(H_3)} \\ \quad = \frac{0.5 \times 0.4}{0.5 \times 0.4 + 0.3 \times 0.3 + 0.5 \times 0.3} = \frac{5}{11} = 0.454545 \\ P(H_2|E_1) = \frac{P(E_1|H_2) \times P(H_2)}{P(E_1|H_1) \times P(H_1) + P(E_1|H_2) \times P(H_2) + P(E_1|H_3) \times P(H_3)} \\ \quad = \frac{0.3 \times 0.3}{0.5 \times 0.4 + 0.3 \times 0.3 + 0.5 \times 0.3} = \frac{9}{44} = 0.204545 \\ P(H_3|E_1) = \frac{P(E_1|H_3) \times P(H_3)}{P(E_1|H_1) \times P(H_1) + P(E_1|H_2) \times P(H_2) + P(E_1|H_3) \times P(H_3)} \\ \quad = \frac{0.5 \times 0.3}{0.5 \times 0.4 + 0.3 \times 0.3 + 0.5 \times 0.3} = \frac{15}{44} = 0.340909 \end{cases}$$

观察结果,可知 E_1 对结论 H_1 的影响较大,对 H_3 的影响其次,对 H_2 的影响较小

- (2):已知 E_1, E_2 同时出现时 $P(H_1|E_1E_2), P(H_2|E_1E_2), P(H_3|E_1E_2)$ 的概率值,试说明 E_1, E_2 同时出现对结论 H_1, H_2, H_3 的影响;

解:

对于 $P(H_i|E_1E_2)$,有贝叶斯公式: $P(H_i|E_1E_2) \times P(E_1E_2) = P(E_1E_2|H_i)P(H_i)$

于是有: $P(H_i|E_1E_2) = \frac{P(E_1E_2|H_i)P(H_i)}{P(E_1E_2)}$

其中有: $P(E_1E_2|H_i) = P(E_1|H_i) \times P(E_2|H_i)$

以及有: $P(E_1E_2) = P(E_1E_2|H_i)P(H_i) + P(E_1E_2|\neg H_i)P(\neg H_i) = \sum_{i=1}^3 (P(E_1|H_i) \times P(E_2|H_i) \times P(H_i))$

带入有: $P(H_i|E_1E_2) = \frac{P(E_1|H_i) \times P(E_2|H_i) \times P(H_i)}{\sum_{i=1}^3 (P(E_1|H_i) \times P(E_2|H_i) \times P(H_i))}$

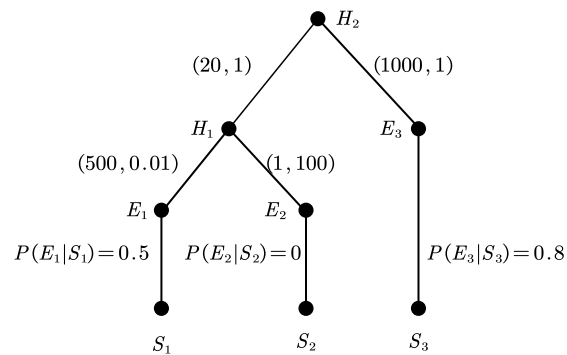
$$\text{带入数据,有:} \left\{ \begin{aligned} P(H_1|E_1E_2) &= \frac{P(E_1|H_1)P(E_2|H_1)P(H_1)}{P(E_1|H_1)P(E_2|H_1)P(H_1) + P(E_1|H_2)P(E_2|H_2)P(H_2) + P(E_1|H_3)P(E_2|H_3)P(H_3)} \\ &= \frac{0.5 \times 0.7 \times 0.4}{0.5 \times 0.7 \times 0.4 + 0.3 \times 0.9 \times 0.3 + 0.5 \times 0.1 \times 0.3} = \frac{35}{59} = 0.59322 \\ P(H_2|E_1E_2) &= \frac{P(E_1|H_2)P(E_2|H_2)P(H_2)}{P(E_1|H_1)P(E_2|H_1)P(H_1) + P(E_1|H_2)P(E_2|H_2)P(H_2) + P(E_1|H_3)P(E_2|H_3)P(H_3)} \\ &= \frac{0.3 \times 0.9 \times 0.3}{0.5 \times 0.7 \times 0.4 + 0.3 \times 0.9 \times 0.3 + 0.5 \times 0.1 \times 0.3} = \frac{81}{236} = 0.34322 \\ P(H_3|E_1E_2) &= \frac{P(E_1|H_3)P(E_2|H_3)P(H_3)}{P(E_1|H_1)P(E_2|H_1)P(H_1) + P(E_1|H_2)P(E_2|H_2)P(H_2) + P(E_1|H_3)P(E_2|H_3)P(H_3)} \\ &= \frac{0.5 \times 0.1 \times 0.3}{0.5 \times 0.7 \times 0.4 + 0.3 \times 0.9 \times 0.3 + 0.5 \times 0.1 \times 0.3} = \frac{15}{236} = 0.0635593 \end{aligned} \right.$$

可见 E_1E_2 同时出现会使得 H_1, H_2 成立的概率提升,使得 E_3 成立的概率大幅下降

2. 设有如下推理规则:

$$\begin{aligned} R_1: & \text{ IF } E_1 \text{ THEN } (500, 0.01) \quad H_1 \\ R_2: & \text{ IF } E_2 \text{ THEN } (1, 100) \quad H_1 \\ R_3: & \text{ IF } E_3 \text{ THEN } (1000, 1) \quad H_2 \\ R_4: & \text{ IF } H_1 \text{ THEN } (20, 1) \quad H_2 \end{aligned}$$

且已知 $P(H_1)=0.1, P(H_2)=0.1, P(H_3)=0.1$, 初始证据的概率为 $P(E_1|S_1)=0.5, P(E_2|S_2)=0, P(E_3|S_3)=0.8$ 。用主观贝叶斯方法求 H_2 的后验概率 $P(H_2|S_1, S_2, S_3)$ 。



解:

先推导一个数学结论,这在利用 LS, LN 计算时会方便些☺

$$\text{方程组: } \begin{cases} \frac{x}{y} = a \\ \frac{1-x}{1-y} = b \end{cases} \rightarrow x = ay \rightarrow \frac{1-ay}{1-y} = b \rightarrow 1-ay = b-by \rightarrow (b-a)y = b-1 \rightarrow y = \frac{b-1}{b-a} \rightarrow x = a \frac{b-1}{b-a}$$

做出如图所示的推理网络,便可采用自下向上的方式算出 $O(H_2|S_1, S_2, S_3)$,从而得到 $P(H_2|S_1 S_2 S_3)$

1. 求出 $O(H_1|S_1)$, (利用 $P(H_1|S_1)$)

求出 $P(H_1|S_1)$, 在这一条分支上, H_1 的证据是 E_1 , 且 $P(E_1|S_1) = 0.5$ 可以用 EH 公式来计算, 而 EH 公式中要用到 $P(E_1)$:

1.1. 利用 LS 公式求出 $P(E_1), P(H_1|E_1)$ 等

LS 公式: $LS = \frac{P(E_1|H_1)}{P(E_1|\neg H_1)} = 500$, 且又有公式 $O(H_1|E_1) = LS \times O(H_1)$, 题目中给到了 $P(H_1) = 0.1$,

不难做出转化: $O(H_1) = \frac{P(H_1)}{1-P(H_1)} = \frac{0.1}{1-0.1} = \frac{1}{9} = 0.111111$,

带入有 $O(H_1|E_1) = LS \times O(H_1) = 500 \times \frac{1}{9} = 55.5556$

再利用一步公式: $P(X) = \frac{O(X)}{1+O(X)}$ 令 $x = H_1|E_1$, 有 $P(H_1|E_1) = \frac{O(H_1|E_1)}{1+O(H_1|E_1)} = \frac{\frac{500}{9}}{1+\frac{500}{9}} = \frac{500}{509} = 0.982318$

再根据 $\begin{cases} LS_1 = \frac{P(E_1|H_1)}{P(E_1|\neg H_1)} = 500 \\ LN_1 = \frac{1-P(E_1|H_1)}{1-P(E_1|\neg H_1)} = 0.01 \end{cases}$

$\rightarrow P(E_1|H_1) = LS_1 \frac{LN_1 - 1}{LN_1 - LS_1} = 500 \times \frac{0.01 - 1}{0.01 - 500} = \frac{49500}{49999} = 0.99002$

利用一次贝叶斯公式: $P(E_1|H_1) \times P(H_1) = P(H_1|E_1) \times P(E_1)$

$\rightarrow P(E_1) = \frac{P(E_1|H_1) \times P(H_1)}{P(H_1|E_1)} = \frac{\frac{49500}{49999} \times 0.1}{\frac{500}{509}} = \frac{53091}{499990} = 0.100784$

1.2. $P(E_1) = 0.10 < P(E_1|S_1)$, 那么利用 EH 公式后半部分进行计算得到 $P(H_1|S_1)$

EH 公式: $P(H_1|S_1) = P(H_1) + \frac{P(H_1|E_1) - P(H_1)}{1 - P(E_1)} \times [P(E_1|S_1) - P(E_1)]$ [$P(E_1) \leq P(E_1|S_1) \leq 1$]

把 $\begin{cases} P(H_1) = 0.1 \\ P(H_1|E_1) = \frac{500}{509} \\ P(E_1|S_1) = 0.5 \\ P(E_1) = \frac{53091}{499990} \end{cases}$ 带入上式:

$\rightarrow P(H_1|S_1) = 0.1 + \frac{\frac{500}{509} - 0.1}{1 - \frac{53091}{499990}} \times \left[0.5 - \frac{53091}{499990} \right] = \frac{222353491}{454943182} = 0.48875$

1.3.再次利用 $O(X) = \frac{P(X)}{1-P(X)}$ 得到 $O(H_1|S_1)$

$$\text{令 } x = H_1|S_1, \text{ 会有 } \boxed{O(H_1|S_1)} = \frac{P(H_1|S_1)}{1-P(H_1|S_1)} = \frac{\frac{222353491}{454943182}}{1 - \frac{222353491}{454943182}} = \frac{\boxed{222353491}}{\boxed{232589691}} = 0.95599$$

2. 求出 $O(H_1|S_2)$, (利用 $P(H_1|S_2)$) [此时 H_1 的证据是 E_2 了]

2.1. 利用 EH 公式求出 $P(H_1|S_2)$

我们理应使用 EH 公式, 但观察题目条件, 发现 $P(E_2|S_2) = 0$, 故可以直接使用 Duda 的公式计算:

$$\text{当 } P(E_2|S_2) = 0 \text{ 时, 有 } \begin{cases} P(\neg E_2|S_2) = 1 \\ \boxed{P(H_1|S_2) = P(H_1|\neg E_2)} \end{cases}$$

2.2. 利用 修改的贝叶斯公式, 求出 $O(H_1|\neg E_2)$, 进一步得到 $P(H_1|\neg E_2)$

根据公式 $O(H_1|\neg E_2) = LN_2 \times O(H_1)$, 其中根据上一问提供的 $\boxed{O(H_1) = \frac{1}{9}}$, 可以得到:

$$\boxed{O(H_1|\neg E_2)} = 100 \times \frac{1}{9} = \frac{\boxed{100}}{\boxed{9}} = 11.1111$$

$$\text{进一步根据公式: } \boxed{P(H_1|\neg E_2)} = \frac{O(H_1|\neg E_2)}{1 + O(H_1|\neg E_2)} = \frac{\frac{100}{9}}{1 + \frac{100}{9}} = \frac{\boxed{100}}{\boxed{109}} = 0.917431$$

2.3. 得到 $O(H_1|S_2)$

那么再次根据 $\boxed{P(H_1|S_2) = P(H_1|\neg E_2)}$ 会有 $P(H_1|S_2) = \frac{100}{109}$

$$\text{根据公式 } \boxed{O(H_1|S_2)} = \frac{P(H_1|S_2)}{1 - P(H_1|S_2)} = \frac{\frac{100}{109}}{1 - \frac{100}{109}} = \frac{\boxed{100}}{\boxed{9}} \text{ (好 } ba \text{ 绕回去了...)}$$

3. 计算 $O(H_1|S_1S_2)$

根据 所有观察下的后验概率 公式: $O(H_1|S_1S_2) = \frac{O(H_1|S_1)}{O(H_1)} \times \frac{O(H_1|S_2)}{O(H_1)} \times O(H_1)$

$$\text{带入有: } O(H_1|S_1S_2) = \frac{\frac{222353491}{232589691}}{\frac{1}{9}} \times \frac{\frac{100}{9}}{\frac{1}{9}} \times \frac{1}{9} = \frac{\boxed{22235349100}}{\boxed{232589691}} = 95.599$$

4. 根据推理网络, 下面应该求 $\boxed{O(E_3|S_3)}$, 利用 $P(E_3|S_3) = 0.8$

$$\text{根据公式, 有 } O(E_3|S_3) = \frac{P(E_3|S_3)}{1 - P(E_3|S_3)} = \frac{0.8}{1 - 0.8} = 4$$

5. 求出 $O(H_2|S_1S_2)$, [需要注意此时 H_1 是 H_2 的证据 (根据 R_3)]

5.1. 计算 $P(H_1|S_1S_2)$

既然证据是 H_1 , 那么根据 EH 公式, 比较 $\boxed{P(H_1)}$, 与 $\boxed{P(H_1|S_1S_2)}$

$$\text{那么 } \boxed{P(H_1|S_1S_2)} = \frac{O(H_1|S_1S_2)}{1 + O(H_1|S_1S_2)} = \frac{\frac{22235349100}{232589691}}{1 + \frac{22235349100}{232589691}} = \frac{22235349100}{22467938791} = 0.989648$$

显然的, $P(H_1) < P(H_1|S_1S_2)$, 需要根据 EH 公式后半半来计算

$$\boxed{P(H_2|S_1S_2) = P(H_2) + \frac{P(H_2|H_1) - P(H_2)}{1 - P(H_1)} \times [P(H_1|S_1S_2) - P(H_1)]}$$

5.2. 计算暂时不知道的 $P(H_2|H_1)$

根据修改的贝叶斯公式 $O(H|E) = LS \times O(H)$, 这里 E 是 H_1 , H 是 H_2 , 带入有:

$$O(H_2|H_1) = LS_{\text{4}} \times O(H_2), \text{ 其中 } O(H_2) = \frac{P(H_2)}{1 - P(H_2)} = \frac{0.1}{1 - 0.1} = \frac{1}{9} = 0.111111$$

$$\text{有 } O(H_2|H_1) = 20 \times \frac{1}{9} = \frac{20}{9} = 2.22222$$

$$\text{根据公式 } P(H_2|H_1) = \frac{O(H_2|H_1)}{1 + O(H_2|H_1)} = \frac{\frac{20}{9}}{1 + \frac{20}{9}} = \frac{20}{29} = 0.689655$$

$$5.3. \text{数据} \begin{cases} P(H_1|S_1S_2) = \frac{22235349100}{22467938791} = 0.989648 \\ P(H_1) = P(H_2) = 0.1 \\ P(H_2|H_1) = \frac{20}{29} \end{cases} \quad \text{带入EH公式:}$$

$$P(H_2|S_1S_2) = P(H_2) + \frac{P(H_2|H_1) - P(H_2)}{1 - P(H_1)} \times [P(H_1|S_1S_2) - P(H_1)]$$

$$P(H_2|S_1S_2) = 0.1 + \frac{\frac{20}{29} - 0.1}{1 - 0.1} \times \left[\frac{22235349100}{22467938791} - 0.1 \right] = \frac{444939571691}{651570224939} = 0.682873$$

5.4. 根据 $P(H_2|S_1S_2)$ 得到 $O(H_2|S_1S_2)$

$$\text{那么容易知道 } O(H_2|S_1S_2) = \frac{P(H_2|S_1S_2)}{1 - P(H_2|S_1S_2)} = \frac{\frac{444939571691}{651570224939}}{1 - \frac{444939571691}{651570224939}} = \frac{444939571691}{206630653248} = 2.15331$$

6. 计算 $O(H_2|S_3)$ [利用 $P(H_2|S_3)$], 根据推理网络, 这里 H_2 对应的证据是 E_3

6.1 肯定还需要利用EH公式, 那么一定要先求出 $P(E_3)$

可以利用贝叶斯公式 $P(E_3|H_2) \times P(H_2) = P(H_2|E_3) \times P(E_3)$ 求

$$\text{根据修改的贝叶斯公式: } O(H_2|E_3) = LS_3 \times O(H_2) = 1000 \times \frac{1}{9} = \frac{1000}{9} = 111.111$$

$$P(H_2|E_3) = \frac{O(H_2|E_3)}{1 + O(H_2|E_3)} = \frac{\frac{1000}{9}}{1 + \frac{1000}{9}} = \frac{1000}{1009} = 0.99108$$

$$\text{根据LS, LN公式: } \begin{cases} LS_3 = \frac{P(E_3|H_2)}{P(E_3|\neg H_2)} = 1000 \\ LN_3 = \frac{1 - P(E_3|H_2)}{1 - P(E_3|\neg H_2)} = 1 \end{cases}, \text{ 有 } P(E_3|H_2) = LS_3 \times \frac{LN_3 - 1}{LN_3 - LS_3} = 0$$

那么带入会发现 $P(E_3) = 0$

6.2. 使用EH公式计算

$$\text{由于 } P(E_3) = 0 < P(E_3|S_3) = 0.8, \text{ 采用EH公式第二部分计算, 带入数据 } \begin{cases} P(H_2) = 0.1 \\ P(E_3|S_3) = 0.8 \\ P(E_3) = 0 \\ P(H_2|E_3) = \frac{1000}{1009} \end{cases}$$

$$P(H_2|S_3) = P(H_2) + \frac{P(H_2|E_3) - P(H_2)}{1 - P(E_3)} \times (P(E_3|S_3) - P(E_3))$$

$$P(H_2|S_3) = 0.1 + \frac{\frac{1000}{1009} - 0.1}{1 - 0} \times (0.8 - 0) = \frac{41009}{50450} = 0.812864$$

6.3.使用公式得到 $O(H_2|S_3)$

$$O(H_2|S_3) = \frac{P(H_2|S_3)}{1 - P(H_2|S_3)} = \frac{\frac{41009}{50450}}{1 - \frac{41009}{50450}} = \frac{41009}{9441} = 4.34371$$

7.最终合成出 $O(H_2|S_1S_2S_3)$,得到 $P(H_2|S_1S_2S_3)$

7.1.得到 $O(H_2|S_1S_2S_3)$

根据所有观察下 H_2 的后验概率:
$$O(H_2|S_1S_2S_3) = \frac{O(H_2|S_1S_2)}{O(H_2)} \times \frac{O(H_2|S_3)}{O(H_2)} \times O(H_2)$$

把数据:
$$\begin{cases} O(H_2|S_1S_2) = \frac{444939571691}{206630653248} = 2.15331 \\ O(H_2) = \frac{1}{9} \\ O(H_2|S_3) = \frac{41009}{9441} \end{cases} \quad \text{带入}$$

$$O(H_2|S_1S_2S_3) = \frac{\frac{444939571691}{206630653248}}{\frac{1}{9}} \times \frac{\frac{41009}{9441}}{\frac{1}{9}} \times \frac{1}{9} = \frac{18246526895476219}{21675555257152} = 84.1802$$

7.2.得到 $P(H_2|S_1S_2S_3)$

根据公式:
$$P(H_2|S_1S_2S_3) = \frac{O(H_2|S_1S_2S_3)}{1 + O(H_2|S_1S_2S_3)} = \frac{\frac{18246526895476219}{21675555257152}}{1 + \frac{18246526895476219}{21675555257152}} = \frac{18246526895476219}{18463282450733371} = 0.98826$$