

1.1 Probabilistic reasoning

Variable assumption:

- M represents my children's situation, $M = 1$ means meltdown, $M = 0$ means not;
- L represents if I will be late, $L = 1$ means late, $L = 0$ means not;

According to the question:

$$P(M = 1) = 0.01$$

$$P(L = 1|M = 1) = 0.98$$

$$P(L = 1|M = 0) = 0.03$$

And we have to get $P(M = 1|L = 1)$:

$$\begin{aligned} P(M = 1|L = 1) &= \frac{P(L = 1|M = 1)P(M = 1)}{P(L = 1|M = 1)P(M = 1) + P(L = 1|M = 0)P(M = 0)} \\ &= \frac{0.98 \times 0.01}{0.98 \times 0.01 + 0.03 \times (1 - 0.01)} \\ &\approx 0.2481 \end{aligned}$$

1.2 Conditioning on background evidence

(a)

$$\begin{aligned} P(X|Y, E) &= \frac{P(X, Y, E)}{P(Y, E)} \\ &= \frac{P(Y|X, E)P(X|E)P(E)}{P(Y|E)P(E)} \\ &= \frac{P(Y|X, E)P(X|E)}{P(Y|E)} \end{aligned}$$

(b)

$$\begin{aligned} P(X|E) &= \frac{P(X, E)}{P(E)} \\ &= \frac{\sum_y P(X, Y = y, E)}{P(E)} \\ &= \frac{\sum_y P(X, Y = y|E)P(E)}{P(E)} \\ &= \sum_y P(X, Y = y|E) \end{aligned}$$

1.3 Conditional independence

(i) \Rightarrow (ii) :

$$\begin{aligned} P(X|Y, E) &= \frac{P(X, Y|E)}{P(Y|E)} \\ &\stackrel{(i)}{=} \frac{P(X|E)P(Y|E)}{P(Y|E)} \\ &= P(X|E) \end{aligned}$$

(ii) \Rightarrow (iii) :

$$\begin{aligned} P(Y|X, E) &= \frac{P(X|Y, E)P(Y|E)}{P(X|E)} \\ &\stackrel{(ii)}{=} \frac{P(X|Y, E)P(Y|E)}{P(X|Y, E)} \\ &= P(Y|E) \end{aligned}$$

(iii) \Rightarrow (i) :

$$\begin{aligned} P(X, Y|E) &= P(Y|X, E)P(X|E) \\ &\stackrel{(iii)}{=} P(X|E)P(Y|E) \end{aligned}$$

In conclusion, these three statements are equivalent.

1.4 Creative writing

(a)

- X represents if a student can pass the test, $X = 1$ means pass, $X = 0$ means not;
- Y represents if the test is difficult, $Y = 1$ means difficult, $Y = 0$ means not;
- Z represents if the student is smart, $Z = 1$ means smart, $Z = 0$ means not;

If the student is smart, it will raise the probability of passing the test. Therefore:

$$P(X = 1|Z = 1) > P(X = 1)$$

If the test is difficult, it will decrease the probability that a smart student can pass the test. Therefore:

$$P(X = 1|Y = 1, Z = 1) < P(X = 1|Z = 1)$$

(b)

- X represents if a student can pass the test, $X = 1$ means pass, $X = 0$ means not;
- Y represents if the test is easy, $Y = 1$ means easy, $Y = 0$ means not;
- Z represents if the student is smart, $Z = 1$ means smart, $Z = 0$ means not;

If the test is easy, it will raise the probability of passing the test. Therefore:

$$P(X = 1) < P(X = 1|Y = 1)$$

If the student is smart, it will raise the probability passing the easy test. Therefore:

$$P(X = 1|Y = 1) < P(X = 1|Z = 1, Y = 1)$$

In conclusion:

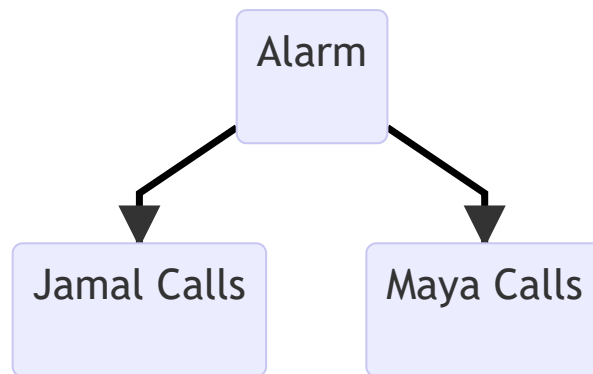
$$P(X = 1) < P(X = 1|Y = 1) < P(X = 1|Z = 1, Y = 1)$$

(c)

Actually it's hard for me to create proper example fitting the problem. We can use the example in class and set proper numerical value to fit it.

- Z represents if the alarm is ringing, $Z = 1$ means ringing, $Z = 0$ means not;
- X represents if Jamal will call, $X = 1$ means calling, $X = 0$ means not;
- Y represents if Maya will call, $Y = 1$ means calling, $Y = 0$ means not;

And the relevance of X , Y and Z are as follows:



Suppose that:

$$P(Z = 1) = 0.9$$

$$P(X = 1|Z = 0) = 0.8$$

$$P(X = 1|Z = 1) = 0.1$$

$$P(Y = 1|Z = 0) = 0.1$$

$$P(Y = 0|Z = 1) = 0.2$$

In the class we have know that X and Y are conditionally independent given Z .
Therefore:

$$P(X, Y|Z) = P(X|Z)P(Y|Z)$$

Then:

$$\begin{aligned}
P(X, Y) &= \sum_z P(X, Y, Z = z) \\
&= \sum_z P(X|Y, Z = z)P(Y|Z = z)P(Z = z) \\
&= \sum_z P(X|Z = z)P(Y|Z = z)P(Z = z)
\end{aligned}$$

And:

$$\begin{aligned}
P(X) &= \sum_z P(X|Z = z)P(Z = z) \\
P(Y) &= \sum_z P(Y|Z = z)P(Z = z)
\end{aligned}$$

Using the concluded formula, we can get:

$$\begin{aligned}
P(X = 1, Y = 1) &= \sum_{i=0}^1 P(X = 1|Z = i)P(Y = 1, Z = i)P(Z = i) \\
&= 0.8 \times 0.1 \times (1 - 0.9) + 0.1 \times 0.2 \times 0.9 \\
&= 0.026
\end{aligned}$$

And:

$$\begin{aligned}
P(X = 1) &= P(X = 1|Z = 0)P(Z = 0) + P(X = 1|Z = 1)P(Z = 1) \\
&= 0.8 \times (1 - 0.9) + 0.1 \times 0.9 \\
&= 0.17 \\
P(Y = 1) &= P(Y = 1|Z = 0)P(Z = 0) + P(Y = 1|Z = 1)P(Z = 1) \\
&= 0.1 \times (1 - 0.9) + 0.2 \times 0.9 \\
&= 0.19
\end{aligned}$$

Therefore:

$$\begin{aligned}
P(X = 1)P(Y = 1) &= 0.17 \times 0.19 \\
&= 0.0323
\end{aligned}$$

In conclusion:

$$P(X = 1, Y = 1) < P(X = 1)P(Y = 1)$$

1.5 Probabilistic inference

(a)

$$\begin{aligned}
P(B = 1|A = 1) &= P(B = 1, E = 0|A = 1) + P(B = 1, E = 1|A = 1) \\
&= \frac{P(A = 1|B = 1, E = 0)P(B = 1, E = 0)}{P(A = 1)} + \\
&\quad \frac{P(A = 1|B = 1, E = 1)P(B = 1, E = 1)}{P(A = 1)} \\
&= \frac{P(A = 1|B = 1, E = 0)P(B = 1)P(E = 0)}{P(A = 1)} + \\
&\quad \frac{P(A = 1|B = 1, E = 1)P(B = 1)P(E = 1)}{P(A = 1)}
\end{aligned}$$

Then $P(A = 1)$ need to be calculated:

$$\begin{aligned}
P(A = 1) &= \sum_{b,e} P(A = 1, B = b, E = e) \\
&= \sum_{b,e} P(A = 1|B = b, E = e)P(B = b, E = e) \\
&= 0.002 \times 0.995 \times 0.999 + 0.96 \times 0.995 \times 0.001 + \\
&\quad 0.35 \times 0.005 \times 0.999 + 0.98 \times 0.005 \times 0.001 \\
&= 0.00469636
\end{aligned}$$

In conclusion:

$$\begin{aligned}
P(B = 1|A = 1) &= \frac{0.96 \times 0.001 \times 0.995}{0.00469636} + \frac{0.98 \times 0.001 \times 0.005}{0.00469636} \\
&\approx 0.204435
\end{aligned}$$

(b)

$$\begin{aligned}
P(B = 1|A = 1, E = 0) &= \frac{P(A = 1|B = 1, E = 0)P(B = 1|E = 0)}{P(A = 1|E = 0)} \\
&= \frac{P(A = 1|B = 1, E = 0)P(B = 1)}{\sum_b P(A = 1|B = b, E = 0)P(B = b|E = 0)} \\
&= \frac{P(A = 1|B = 1, E = 0)P(B = 1)}{\sum_b P(A = 1|B = b, E = 0)P(B = b)} \\
&= \frac{0.96 \times 0.001}{0.002 \times (1 - 0.001) + 0.96 \times 0.001} \\
&\approx 0.32454
\end{aligned}$$

(c)

$$\begin{aligned}
P(A = 1|J = 1) &= \frac{P(J = 1|A = 1)P(A = 1)}{P(J = 1)} \\
&= \frac{P(J = 1|A = 1)P(A = 1)}{P(J = 1, A = 0) + P(J = 1, A = 1)} \\
&= \frac{P(J = 1|A = 1)P(A = 1)}{P(J = 1|A = 0)P(A = 0) + P(J = 1|A = 1)P(A = 1)} \\
&= \frac{0.93 \times 0.00469636}{0.1 \times (1 - 0.00469636) + 0.93 \times 0.00469636} \\
&\approx 0.04204
\end{aligned}$$

(d)

$$\begin{aligned}
&P(A = 1|J = 1, M = 1) \\
&= \frac{P(J = 1|M = 1, A = 1)P(M = 1|A = 1)P(A = 1)}{P(J = 1, M = 1)} \\
&= \frac{P(J = 1|A = 1)P(M = 1|A = 1)P(A = 1)}{P(J = 1, M = 1, A = 0) + P(J = 1, M = 1, A = 1)} \\
&= \frac{P(J = 1|A = 1)P(M = 1|A = 1)P(A = 1)}{P(J = 1|A = 0)P(M = 1|A = 0)P(A = 0) + P(J = 1|A = 1)P(M = 1|A = 1)P(A = 1)} \\
&= \frac{0.93 \times 0.65 \times 0.00469636}{0.1 \times 0.01 \times (1 - 0.00469636) + 0.93 \times 0.65 \times 0.00469636} \\
&\approx 0.740418
\end{aligned}$$

(e)

$$\begin{aligned}
P(A = 1|M = 0) &= \frac{P(M = 0|A = 1)P(A = 1)}{P(M = 0)} \\
&= \frac{P(M = 0|A = 1)P(A = 1)}{P(M = 0|A = 0)P(A = 0) + P(M = 0|A = 1)P(A = 1)} \\
&= \frac{(1 - 0.65) \times 0.00469636}{(1 - 0.01) \times (1 - 0.00469636) + (1 - 0.65) \times 0.00469636} \\
&\approx 0.0016654
\end{aligned}$$

(f)

$$\begin{aligned}
P(A = 1|M = 0, E = 1) &= \frac{P(M = 0|A = 1, E = 1)P(A = 1|E = 1)}{P(M = 0|E = 1)} \\
&= \frac{P(M = 0|A = 1)P(A = 1|E = 1)}{P(M = 0|E = 1)}
\end{aligned}$$

Calculate $P(A = 1|E = 1)$:

$$\begin{aligned}
P(A = 1|E = 1) &= P(A = 1, B = 0|E = 1) + P(A = 1, B = 1|E = 1) \\
&= P(A = 1|B = 0, E = 1)P(B = 0|E = 1) + \\
&\quad P(A = 1|B = 1, E = 1)P(B = 1|E = 1) \\
&= P(A = 1|B = 0, E = 1)P(B = 0) + \\
&\quad P(A = 1|B = 1, E = 1)P(B = 1) \\
&= 0.35 \times (1 - 0.001) + 0.98 \times 0.001 \\
&= 0.35063
\end{aligned}$$

Calculate $P(M = 0|E = 1)$:

$$\begin{aligned}
P(M = 0|E = 1) &= \sum_a P(M = 0, A = a|E = 1) \\
&= \sum_a P(M = 0|A = a, E = 1)P(A = a|E = 1) \\
&= \sum_a [P(M = 0|A = a) \sum_b P(A = a, B = b|E = 1)] \\
&= \sum_a [P(M = 0|A = a) \sum_b P(A = a|B = b, E = 1)P(B = b|E = 1)] \\
&= \sum_a [P(M = 0|A = a) \sum_b P(A = a|B = b, E = 1)P(B = b)] \\
&= (1 - 0.01) \times [(1 - 0.35) \times (1 - 0.001) + (1 - 0.98) \times 0.001] + \\
&\quad (1 - 0.65) \times [0.35 \times (1 - 0.001) + 0.98 \times 0.001] \\
&= 0.7655968
\end{aligned}$$

In conclusion:

$$\begin{aligned}
P(A = 1|M = 0, E = 1) &= \frac{P(M = 0|A = 1)P(A = 1|E = 1)}{P(M = 0|E = 1)} \\
&= \frac{(1 - 0.65) \times 0.35063}{0.7655968} \\
&= 0.16029
\end{aligned}$$

The results of (a) and (b), (c) and (d), (e) and (f) are all different. They are consistent with commonsense patterns of reasoning.

1.6 Kullback-Leibler distance

(a)

Assume that:

$$f(z) = \log z - (z - 1)$$

Then:

$$\frac{d}{dz} f(z) = \frac{1}{z} - 1 = \frac{1 - z}{z}$$

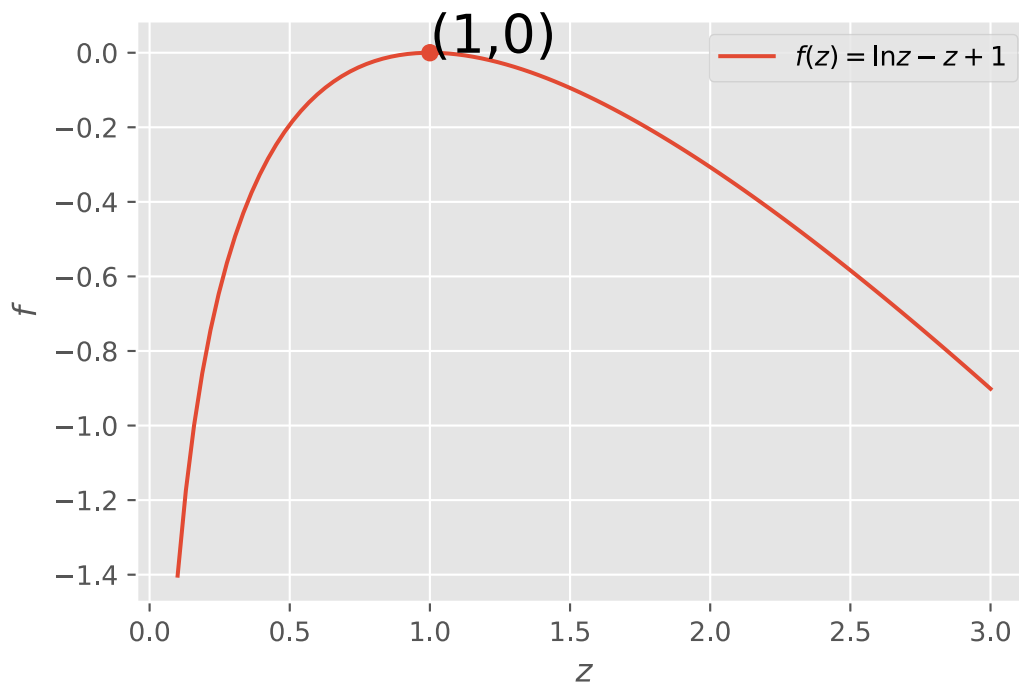
The maximum of $f(z)$ is $f(1) = 0$. Therefore:

$$f(z) \leq 0$$

$$\Leftrightarrow \log z \leq z - 1$$

Only if $z = 1$, then $\log z = z - 1$.

The figure of $f(z)$ is as follows:



(b)

Using the conclusion of (a):

$$\begin{aligned} \text{KL}(p, q) &= \sum_i p_i \log\left(\frac{p_i}{q_i}\right) \\ &= - \sum_i p_i \log\left(\frac{q_i}{p_i}\right) \\ &\geq - \sum_i p_i \left(\frac{q_i}{p_i} - 1\right) \\ &= \sum_i p_i - \sum_i q_i \\ &= 0 \end{aligned}$$

In conclusion:

$$\text{KL}(p, q) \geq 0$$

Only if $\frac{p_i}{q_i} = 1$ for all i , then $\text{KL}(p, q) = 0$.

(c)

Assumption:

- X represents if a student pass the exam, $X = 1$ means pass, $X = 0$ means not;
- E represents the test is easy;

- E' represents the student is smart;

Then assume that:

$$\begin{aligned} p &: P(X = 1|E) = 0.8, P(X = 0|E) = 0.2; \\ q &: P(X = 1|E') = 0.6, P(X = 0|E') = 0.4; \end{aligned}$$

Using the formula:

$$\begin{aligned} \text{KL}(p, q) &= \sum_i p_i \log\left(\frac{p_i}{q_i}\right) \\ &= 0.8 \log \frac{0.8}{0.6} + 0.2 \log \frac{0.2}{0.4} \\ &\approx 0.091516 \\ \text{KL}(q, p) &= \sum_i q_i \log\left(\frac{q_i}{p_i}\right) \\ &= 0.6 \log \frac{0.6}{0.8} + 0.4 \log \frac{0.4}{0.2} \\ &\approx 0.104650 \end{aligned}$$

Therefore, KL distance is not a symmetric function, which means that $\text{KL}(p, q) \neq \text{KL}(q, p)$.