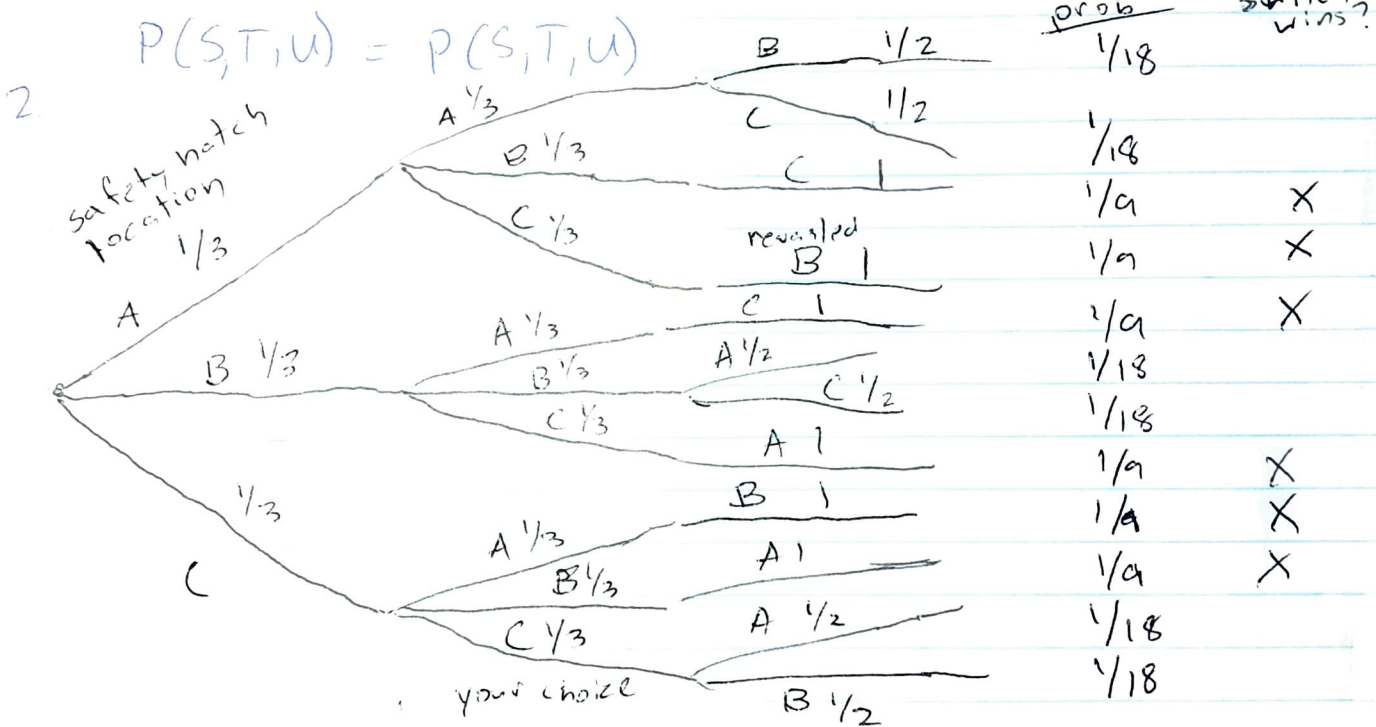


Gordon Ng CS505

1 $P(S|T) = \frac{P(S,T)}{P(T)}$ Conditional probability

$$P(S,T,U) = P(S|T,U) P(T|U) P(U)$$

$$\frac{P(S,T,U)}{P(T,U)} \times \frac{P(T,U)}{P(U)} \times \frac{P(U)}{1}$$



~~safety guess~~ | ~~door guessed~~ | ~~door that was requested~~ | ~~probability~~

$$P(\text{Safety}) = 1/3 \quad P(\text{HAL reveals}) = 1/2$$

$$P(\text{safety 1} \cap \text{HAL reveals}) = 1/3 \times 1/2 = 1/6$$

$$P(\text{safety 2} \cap \text{HAL reveals}) = 1/3 \times 1 = 1/3$$

$$P(\text{safety 3} \cap \text{HAL reveals}) = 1/3 \times 1 = 1/3$$

$$P(\text{stay}) = P(\text{safety 1} | \text{HAL reveals}) \rightarrow P(\text{safety 1} \cap \text{HAL}) / P(\text{HAL})$$

$$P(\text{stay}) = P(\text{safety 2} | \text{HAL reveals}) \rightarrow P(\text{safety 2} \cap \text{HAL}) / P(\text{HAL})$$

$$P(\text{stay} \cap \text{win}) = 1/3 \quad P(\text{stay} \cap \text{lose}) = 2/3$$

It is always smarter to switch given the possible events.

win
lose

3. Given $E[f(A)] = E[E(f(A)|B)]$

A & B are random variables

$$= \sum_b f(b) P_b(b) \quad \text{expected value rule}$$

$$= \sum_b E[f(A) | B=b] P_b(b) \quad \text{total expected theorem}$$

$$= E[X]$$

Law of iterated expectations

$$E[E(X|Y)] = E[X]$$

4. $P(S|B) = \prod_{i=1}^n \beta e^{-\beta s_i}$, $P(S|B) = X$

$$X = \frac{\beta^n}{e^{-\beta \sum_{i=1}^n s_i}}$$

$$\log X = n \log \beta - \beta \sum_{i=1}^n s_i$$

$$\frac{\partial \log L}{\partial \beta} = \frac{n}{\beta} - \sum_{i=1}^n s_i \quad (1)$$

$$\frac{n}{\beta} - \sum_{i=1}^n s_i = 0$$

$$\frac{n}{\beta} = \sum_{i=1}^n s_i$$

$$\hat{\beta} = \frac{n}{\sum_{i=1}^n s_i}$$

$$= \left. \frac{\partial^2 \log L}{\partial \beta^2} \right|_{\hat{\beta}} < 0$$

maximum likelihood

$$\boxed{\hat{\beta} = \frac{n}{\sum_{i=1}^n s_i}}$$

$$5. P(U|S) = 0.9$$

$$\frac{P(U \cap S)}{P(S)}$$

$$P(U|T) = 0.6$$

$$\frac{P(U \cap T)}{P(T)}$$

$$P(U|S \cap T)$$

$$\frac{P(U \cap S \cap T)}{P(S \cap T)}$$

$$\frac{P(S|U \cap T) P(U \cap T)}{P(S|T)}$$

$$P(U|S, T) = \frac{P(S|U, T) \cdot P(S|T)}{P(S|T)}$$

Not enough info.

We need $P(U|T)$, $P(S|T)$ and $P(S|U, T)$ but we are only given $P(U|T) = 0.6$ and $P(S|T)$ and $P(S|U, T)$ aren't given.

$$6. P(U|S) = 0.9 \quad P(S) = 0.5 \quad P(U|T) = 0.6 \quad P(T) = 0.5 \quad P(U|S \cap T)$$

$$P(S|T) = \frac{P(S \cap T)}{P(T)} = \frac{0.4 \cdot 0.5}{0.5} = 0.4$$

$$P(U \cap S) = 0.9 \times 0.5 \quad P(U \cap T) = 0.6 \times 0.5$$

$$P(U|S, T) = \frac{P(U \cap S, T)}{P(S, T)} \quad \text{not enough information}$$

$$7. P(U \cap S) = 0.3 \quad P(S) = 0.5 \quad P(T) = 1 \quad P(U|S \cap T) = \frac{P(U, S, T)}{P(S, T)}$$

$$\frac{P(U, S)}{P(S)} = \frac{0.3}{0.5} = \boxed{\frac{3}{5}}$$

$$8) P(G \cap L) = \frac{2}{3} \quad P(G) = \frac{1}{6} \quad P(\neg G \cap L) = \frac{1}{5} \quad P(G \cap L) = ?$$

$$\frac{1}{10} \times \frac{2}{3} = \boxed{\frac{1}{15}}$$

$$9. P(\text{fluffy}) = 2/3 \quad P(\text{cute} | \text{fluffy}) = 0.8 \quad P(\text{cute} | \neg \text{fluffy}) = 0.1$$

$$P(B|A') = 0.1 = \frac{P(A' \cap B)}{P(A')} = 0.1$$

$$P(A' \cap B) = 0.1 \times 1/3 = 1/30$$

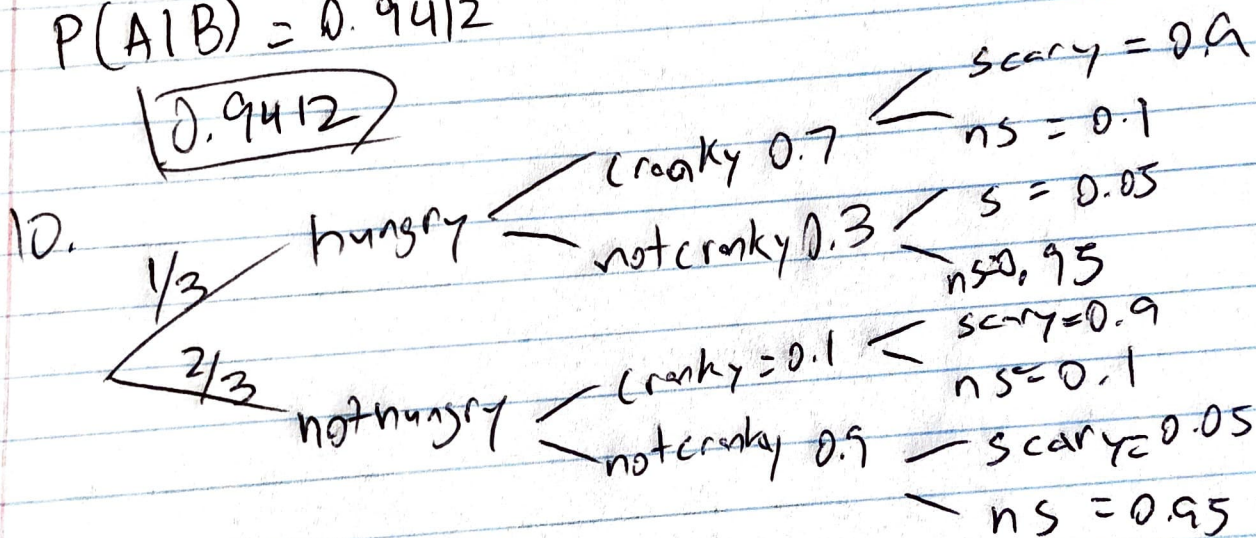
$$P(A' \cap B) = P(B) - P(A \cap B) \quad A$$

$$P(B) = P(A' \cap B) + P(A \cap B) = 17/30$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{8}{17} / \frac{17}{30} = \frac{16}{17}$$

$$P(A|B) = 0.9412$$

$$\boxed{0.9412}$$



$$\left(\frac{1}{3} \times 0.7 \times 0.9 \right) + \left(\frac{1}{3} \times 0.3 \times 0.05 \right) = \frac{0.21}{0.21 + 0.005}$$

$$\boxed{0.9767}$$

$$10. P(C|H,S) = P(C,H,S) / P(C,H,S) + P(\neg C,H,S)$$

$$P(C|H,S) = P(S|CH) P(C|H) P(H)$$

$$P(S|CH) = P(S|C)$$

$$P(S|\neg CH) = P(S|\neg C)$$

$$11. P(CD) = P(CDA) + P(CD\neg A)$$

$$P(CD) = 0 + P(CD\neg A)$$

$$P(C|A) = 0, P(CD) = 0 + P(CD\neg A)$$

$$P(CD\neg A) = P(C|D\neg A) P(D\neg A)$$

$$= P(C|\neg A) P(D\neg A)$$

$$P(D\neg A) = P(D\neg AB) + P(D\neg A\neg B) \text{ knowing } B$$

$$P(D\neg AB) = P(D|\neg AB) P(B|\neg A) P(\neg A)$$

$$= P(D|B) P(B|\neg A) P(\neg A)$$

$$P(D\neg AB) = P(D|\neg AB) P(B|\neg A) P(\neg A)$$

$$= P(D|\neg B) P(\neg B|\neg A) P(\neg A)$$

$$P(C,D,\neg A) = \frac{1}{2} \cdot$$

$$d. E(W) = E(E(W|M))$$

$$e. E(W^2) = E(E(W^2|M))$$

$$= E(\text{Var}(W|M) + E(W|M)^2)$$

$$= E(\text{Var}(W|M)) + E(E(W|M)^2)$$

12, vanilla = 1/6

$$\text{Var} = E(\text{Var})$$

AI prob = given

$$E(W) = \sum P(\text{blank} | \text{van})m + \sum P(\text{blank} | \text{AI})(1-m)$$

$$\text{Var}(W) = \text{Expected}(W^2) - \text{Expected}(W)^2$$

$$\text{Var}(W|M) = E(W^2|M) - E(E(W|M))^2$$