

## Probability Assignment - CSU44062

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Q1

	gw	!gw
ps	140	10
!ps	700	150

(a) According to conditional probability from lectures:

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

Therefore, we can calculate  $P(gw|ps)$  with:

$$P(gw|ps) = \frac{P(gw \cap ps)}{P(ps)}$$

From the table above, we know that

$$P(gw \cap ps) = \frac{140}{1000}$$

(Since 140 on the table is  
 (when gw AND ps, and all the  
 numbers in the table add to 1000))

$$P(ps) = \frac{150}{1000}$$

(Since the row of ps equates  
 to  $140 + 10 = 150$ . Also, all numbers  
 in the table add to 1000)

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$$\therefore P(gw|ps) = \frac{\frac{140}{1000}}{\frac{150}{1000}} = \frac{14}{15} \approx 0.93$$

Which counts are irrelevant to this calculation:

- $7ps \wedge gw = 700$
- $7ps \wedge \neg gw = 150$

Both these counts (i.e. the entire  $7ps$  row) are irrelevant to this calculation.

(b) According to Conditional probability

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

Therefore, we can calculate  $P(ps|gw)$  with:

$$P(ps|gw) = \frac{P(ps \wedge gw)}{P(gw)}$$

From the table, we know that:

$$P(ps \wedge gw) = \frac{140}{1000}$$

(since 140 on the table is when ps  $\wedge$  gw, and all the numbers on the table add to 1000)

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$$P(gw) = \frac{840}{1000}$$

Since the row or column of gw equates to  $140 + 700 = 840$ . Again, all numbers in the table add to 1000)

$$\therefore P(ps|gw) = \frac{\left(\frac{140}{1000}\right)}{\left(\frac{840}{1000}\right)} = \frac{1}{6} \approx 0.16$$

Which counts are irrelevant to this calculation:

- $\neg gw \wedge ps = 10$
- $\neg gw \wedge \neg ps = 150$

Both of these counts (i.e. the entire  $\neg gw$  column) are irrelevant to this calculation)

Q2

$$(i) P(A \wedge B) = P(A) * P(B)$$

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

To prove: (i) and (ii) are equivalent

Step ① : Prove (i)  $\rightarrow$  (ii) [ (i) implies (ii) ]

$$(1) P(A \wedge B) = P(A) \times P(B) \quad (i)$$

(2) Recall from conditional probability that:

$$P(A|B) = \frac{P(A \wedge B)}{P(B)}$$

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Now substitute our equation for  $P(A \cap B)$  from (1) into the equation for  $P(A|B)$ .

$$\therefore P(A|B) = \frac{P(A) \times P(B)}{P(B)}$$

$$(3) \quad P(A|B) = P(A) \times \frac{P(B)}{P(B)} = P(A) \times 1$$

$$(4) \quad P(A|B) = P(A) \quad (\text{ii})$$

Hence proved  $(\text{i}) \rightarrow (\text{ii})$

Step ②: Prove  $(\text{ii}) \rightarrow (\text{i})$  [ $(\text{ii})$  implies  $(\text{i})$ ]

$$(1) \quad P(A|B) = P(A) \quad (\text{ii})$$

(2) We know from conditional probability that:

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

Now we have two equations for  $P(A|B)$  from (1) and (2). Let's combine them equal to each other.

$$\therefore \frac{P(A \cap B)}{P(B)} = P(A)$$

$$(3) \quad P(A \cap B) = P(A) \times P(B) \quad (\text{ii})$$

Hence proved  $(\text{ii}) \rightarrow (\text{i})$

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Q3

yng : Age = "smaller than 30"

$\neg$ yng : Age = "not smaller than 30"

pb : PB = true

(a)

Supposing

$$P(\text{yng}) = 0.25$$

$$P(\text{pb}|\text{yng}) = 0.95$$

$$P(\text{pb}|\neg\text{yng}) = 0.01$$

Recall from lectures that the Bayesian rule states:

$$\text{Choose } \arg_{\omega} \max P(x|\omega) P(\omega)$$

(i.e. picking the  $\omega$  which makes the combination you are looking at as likely as possible)

In this case:

$$\text{yng: } P(\text{pb}|\text{yng}) P(\text{yng}) = 0.95 \times 0.25 = 0.2375$$

$$\neg\text{yng: } P(\text{pb}|\neg\text{yng}) P(\neg\text{yng}) = 0.01 \times (1 - 0.25) = 0.01 \times 0.75 = 0.0075$$

∴ The best guess is yng since  $0.2375 > 0.0075$

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(b)

Supposing

$$p(\text{yng}) = 0.01$$

$$p(\text{pb}|\text{yng}) = 0.95$$

$$p(\text{pb}|\neg\text{yng}) = 0.01$$

The bayesian rule from (a) applied:

$$\text{yng: } p(\text{pb}|\text{yng}) \times p(\text{yng}) = 0.95 \times 0.01 = 0.0095$$

$$\neg\text{yng: } p(\text{pb}|\neg\text{yng}) \times p(\neg\text{yng}) = 0.01 \times (1 - 0.01) = 0.01 \times 0.99 = 0.0099$$

$\therefore$  The best guess is  $\neg\text{yng}$  since  $0.0099 > 0.0095$

(c)

Supposing

$$p(\text{yng}) = 0.01$$

$$p(\text{pb}|\text{yng}) = 0.95$$

$$p(\text{pb}|\neg\text{yng}) = 0.0001$$

The bayesian rule from (a) applied:

$$\text{yng: } p(\text{pb}|\text{yng}) \times p(\text{yng}) = 0.95 \times 0.01 = 0.0095$$

$$\begin{aligned} \neg\text{yng: } p(\text{pb}|\neg\text{yng}) \times p(\neg\text{yng}) &= 0.0001 \times (1 - 0.01) \\ &= 0.0001 \times 0.99 \\ &= 0.000099 \end{aligned}$$

$\therefore$  The best guess is yng since  $0.0095 > 0.000099$

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Q4

	noisy: +	noisy: -
cool: +	62	108
cool: -	38	292

$$p(\text{cool: +}) = \frac{62+108}{500} = 0.34$$

$$p(\text{cool: +} | \text{noisy: +}) = \frac{62}{62+38} = 0.62$$

From lectures, we know that ~~the~~ one of the formulations of independence can be written as:

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

We can apply this to cool and noisy, in fact to the "+" settings of these variables. This is done above with  $p(\text{cool: +})$  and  $p(\text{cool: +} | \text{noisy: +})$ . However we realize that:

$$P(\text{cool: +} | \text{noisy: +}) \neq p(\text{cool: +})$$

Since

$$0.62 \neq 0.34$$

therefore

$$P(X|Y) \neq P(X)$$

Thus, we can say that cool: + is NOT independent of noisy: +.

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Q5

<u>open: +</u>	<u>noisy: +</u>	<u>noisy: -</u>
<u>Cool: +</u>	54	36
<u>Cool: -</u>	6	4

<u>open: +</u>	<u>noisy: +</u>	<u>noisy: -</u>
<u>Cool: +</u>	8	72
<u>Cool: -</u>	32	288

$$P(\text{Cool: +} | \text{open: +}) = \frac{54 + 36}{100} = \frac{9}{10}$$

$$P(\text{Cool: +} | \text{open: +}, \text{noisy: +}) = \frac{54}{54 + 6} = \frac{9}{10}$$

From lectures, we discussed that an alternative definition for conditional independence is:

$$P(X|Y, Z) = P(X|Z) \quad (1)$$

We can rewrite this equation using some basic commutativity:

$$P(X|Y, Z) = P(X|Z, Y) \left[ \begin{array}{l} \text{probability of } X \text{ given } Y \text{ AND } Z \\ \text{probability of } X \text{ given } Z \text{ AND } Y \end{array} \right]$$

$$P(X|Z, Y) = P(X|Y) \quad [\text{using (1)}]$$

$$\therefore P(X|Y, Z) = P(X|Y) \quad [\text{proves conditional independence}]$$

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Therefore, if we can prove the equation  ~~$P(X|YZ)$~~ ,  
 $P(X|Y, Z) = P(X|Y)$ , then we can prove that  
conditional independence is true.

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

$$P(\text{cool:} + | \text{open:} +, \text{noisy:} +) = P(\text{cool:} + | \text{open:} +)$$

$$\frac{9}{10} = \frac{9}{10}$$

∴ Cool: + is conditionally independent  
of noisy: + given open: +.

Q6

4 tosses of a coin : H H H T

where H = heads

T = tails

Let's say  $\theta_h$  = probability of a toss of the coin giving H.

For  $\theta_h = 0.1$

$$\begin{aligned} P(H, H, H, T) &= (0.1) \times (0.1) \times (0.1) \times (1 - 0.1) \\ &= 0.0009 \end{aligned}$$

For  $\theta_h = 0.5$

$$\begin{aligned} P(H, H, H, T) &= (0.5) \times (0.5) \times (0.5) \times (1 - 0.5) \\ &= 0.0625 \end{aligned}$$

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For  $\theta_h = 0.75$

$$\begin{aligned} P(H, H, H, T) &= (0.75) \times (0.75) \times (0.75) \times (1 - 0.75) \\ &= 0.10546875 \end{aligned}$$

For  $\theta_h = 0.9$

$$\begin{aligned} P(H, H, H, T) &= (0.9) \times (0.9) \times (0.9) \times (1 - 0.9) \\ &= 0.0729 \end{aligned}$$

N.B. For ~~this~~ sequence, the general formula for the probability of the sequence using  $\theta_h$  would be:

$$P(H, H, H, T) = (\theta_h) \times (\theta_h) \times (\theta_h) \times (1 - \theta_h)$$

where  $\theta_h$  = probability of H

Since coin ~~flips~~ flips are independent, the probability of the sequence H HHT can be found by multiplying the probability of each event in the sequence.