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Information Theory 2018, Final Exam
Solutions

Q1

- a) ~~0.02~~
 b) ~~0.95~~
 c) ~~0.01~~

d) $P(\text{Attentive dog} \mid \text{Test Positive})$

Work
 $P(A \mid T) P(T) = P(T \mid A) P(A)$

$$\Rightarrow P(A \mid T) = \frac{P(T \mid A) P(A)}{P(T)}.$$

Now $P(T \mid A) = \cancel{0.98} \cdot 0.95$

$P(A) = \cancel{0.02} \cdot 0.02$

$$\begin{aligned} P(T) &= P(T \mid \neg A) P(\neg A) \\ &\quad + P(T \mid A) P(A) \\ &= 0.01 \times 0.98 + 0.95 \times 0.02 \\ &= \cancel{0.0098} + \cancel{0.019} \\ &= 0.0288 \end{aligned}$$

∴

$$P(A \mid T) = \frac{0.95 \times 0.02}{0.0288} = 0.6597$$

$$= 66\%.$$

0382

5. a) No Regen $P(X = \text{max } Y = y)$
- $P(X = n) P(Y = y)$

$$5) P(X=1) = P(X=1 | Y=1) P(Y=1) + P(X=1 | Y=0) P(Y=0)$$

$$x \leq 0.2 P(Y=1) + 0.6 P(Y=0)$$

Do not know $P(Y=1)$.
will use known
is $P(X=1) \leq 0.6$.
thus $P(X=0) \geq 0.4$

$$\begin{aligned}
 \text{c) } P(X=1 | Y=0) &= P(Y=0 | X=1) P(X=1) \\
 P(Y=0 | X=1) &= \frac{P(X=1 | Y=0) P(Y=0)}{P(X=1)} \\
 &= \frac{0.6 \cdot P(Y=0)}{0.21}
 \end{aligned}$$

$$c) \frac{P(Y=0|X=0)}{P(Y=1|X=0)} = \frac{P(X=0, Y=0)}{P(X=0, Y=1)} \cdot \frac{P(X=0)}{P(X=1)}$$

~~Davidson's socks~~

$$(a) \Rightarrow \frac{P(Y=0|X=0)}{P(X=0)} \cdot P(X=0)$$

$$(b) \Rightarrow \frac{P(Y=1|X=0)}{P(X=0, Y=0)} ; P(X=0)$$

$$\frac{P(x_0, y_0) P(x=1 | x=0)}{P(x_0, y_0)} \rightarrow P(x=1 | y=0)$$

(4)

$$\hookrightarrow \frac{P(X=0 | X=0)}{P(X=0, Y=0)} = \frac{P(Y=1 | X=0)}{P(X=0, Y=1)}$$

$$\Rightarrow \frac{P(X=0, Y=1)}{P(X=0, Y=0)} = \frac{P(Y=1 | X=0)}{P(Y=0 | X=0)}$$

flip both sides & are we done.

d)

- 1) No, does not separate
- 2) Yes, because $P_{\theta} \cdot p = q \Rightarrow q = p$
- 3) No, just from $p = q$
- 4) No, just from $p = q$, looks like the
- 5) No, same as 3
- 6) No, $I(X; Y)$ requires $X \neq Y$
- 7) Yes
- 8) Yes $I(X; X) = H(X)$, \therefore Data from future does not change much

Q3 a). $H(X) = \log n = H(Y) = \boxed{\text{Haus}}$.
 $H(Z) = \log_2 2n = H(Y) + 1$

b) Use decomposition:

$$\begin{aligned} H(p) &= H_2(p_1) + p_1 p_1 H\left(\frac{p_2}{1-p_1}, \dots, \frac{p_4}{1-p_1}\right) \\ &= H_2(p_1) + (1-p_1) \left[H_2\left(\frac{p_2}{1-p_1}\right) + (1-\frac{p_2}{1-p_1}) H_2\left(\frac{\frac{p_3}{1-p_1}}{1-\frac{p_2}{1-p_1}}\right) \right] \end{aligned}$$

\leftarrow

- d)
- 1) No, Due not symmetric
 - 2) Yes, because then $p=q \Rightarrow q=p$
 - 3) No, just more $p=q$
 - 4) No, just more $p=q$. Could be dep.
 - 5) No. Same as 3
 - 6) No, $I(X; Y)$ requires $X \perp Y$. See 4
 - 7) Yes.
 - 8) Yes $I(X; X) = H(X)$. Determine further when not always result.

3e)

3d) Jensen using concave user

$$F(f(x)) \leq f(F(x))$$

$$\text{Thus } F(g(x)) \leq 1.$$

3e)

No sole party wins this competition is always non-negative
the Markov's \neq

$$P(X \geq \lambda) \leq \frac{E(X)}{\lambda}.$$

$$\lambda = 2E(X) \text{ Thus probability is } \leq \frac{1}{2}.$$

3f)

$$P(|X - E(X)| \geq 2\sqrt{V(X)}) \leq \frac{1}{2^2}$$

$$V(X) = 10 \Rightarrow \text{Need } 2\sqrt{10} = 10 \Leftrightarrow 2 = \sqrt{10}$$

$$\text{Thus } p \leq \frac{1}{10}$$

$$f_{ab} = \frac{1}{2} \log \left(\frac{1}{2} \right) + \frac{1}{2} \log \left(\frac{1}{2} \right) = -\frac{1}{2} \log 2$$

$$f_{bab} = f_{bab} = \left(\frac{1}{2} \right)^2 \cdot \frac{1}{2} =$$

$$-\frac{1}{2} \log \left(\frac{1}{2} \right) = 0.918$$

$$H(X) = H \left(\frac{1}{2} \right) = 0.917 \quad \checkmark$$

last table as it was not
that last part is not, also, the last
part (3) - 2 is not part of the

$$\text{QF. a) } p(x) = \left(\frac{1}{8}, \frac{7}{8}\right)$$

$$p(y) = (3/4, 1/4)$$

$$p(x|y=1) = (0, 1)$$

$$p(x|y=2) = \left(\frac{1}{2}, \frac{1}{2}\right)$$

$$\text{b) } H(x) = H_2\left(\frac{1}{8}\right) = -\frac{1}{8} \log_2 \frac{1}{8} - \frac{7}{8} \log_2 \frac{7}{8}$$

$$= \frac{1}{8} \cdot 3 - \frac{7}{8} \frac{\log_{10}(\frac{7}{8})}{\log_{10}(2)}$$

$$= \frac{1}{8} \cdot 3 + \frac{7}{8} \cdot 0.1926$$

$$= -375 + 0.69$$

$$= 0.544 \text{ bits}$$

$$H(x|y=1) = H_2(0) = 0$$

$$\text{b) } H(x|y=2) = H_2\left(\frac{1}{2}\right) = 1 \geq 0.544$$

$$\text{c) } H(x|Y) = \sum_{y \in \{1, 2\}} p(y) H(x|Y=y) = 0 + \frac{1}{2} \cdot 1 = \frac{1}{2} < H(x)$$

$$\text{QF. a) } \{aaa, aab, aba, ahh, baa, bab, bba, bbb\}$$

$$\text{b) } ahh, bba, \text{ or } bab$$

$$T_{NP} = \{x \in A^3 \mid |1 - \frac{1}{3} \log_2 p(x) - H(X)| < 0.2\}$$

$$p(aah) = p(bab) = \left(\frac{2}{3}\right)^2 \cdot \frac{1}{3} = \frac{4}{27}$$

$$-\frac{1}{3} \log_2 \left(\frac{4}{27}\right) = 0.918$$

$$H(x) = H_2\left(\frac{1}{3}\right) = 0.917 \quad \checkmark$$

$$\text{c) } \text{Least probable seq is aac at freq } \frac{1}{27}$$

$$\text{so } \frac{1}{27} = A \setminus \{a, a, a\}$$

Next least prob is aab, aba, baa. at

$$\text{prob } \left(\frac{1}{3}\right)^2 \cdot \frac{2}{3} = \frac{2}{27} \cdot \text{ Need 2 of these} > \frac{1}{10}$$

A

d) $|X^{1000}| = 2^{1000}$. $|T_{Npl}| = 2^{\frac{N\bar{H}(x) + p}{2}} = 2^{\frac{1000 \times 0.908 + 2}{2}} = 2^{918.2}$

Ratio is $\approx 2^{-82}$. - Rat is the fraction of sequences that are typical!

e) By the asymptotic equipartition property
The probability is close to 1. (!?)

Q6. a) $4 \cdot \frac{1}{2} + 2 \cdot \frac{1}{4} + \frac{1}{8} \cdot 2 + \frac{1}{8} \cdot 1 = 2 \frac{7}{8}$

b) No, 0 is a prefix of 0000

c) No, 0000 can be deleted as a

d) $H(x) = \frac{1}{2} \log_2 2 + \frac{1}{2} \log_2 4 + \frac{3}{8} \log_2 8$
 $\sim \frac{1}{2} + \frac{1}{2} + \frac{3}{8}$
 $= 1.75$

e) the Kraft inequality: Require $\sum_i 2^{-l_i} \leq 1$

We have $\frac{1}{2} + \frac{1}{4} + \frac{1}{4} + \frac{1}{8} \geq 1$, so no.

f) Shannon code word lengths are $l_i = \lceil \log_2 \frac{1}{p_i} \rceil$

So $l_1 = 1$

ex. 0

$l_2 = 2$

10

$l_3 = 3$

110

$l_4 = 4$

111

g) $2 \cdot \frac{1}{2} \quad \frac{1}{2} \quad \frac{1}{2} - 1$

$\frac{1}{2} \cdot \frac{1}{4} \quad \frac{1}{4} \quad \frac{1}{4} \quad \frac{1}{2} - 1$

$\frac{1}{8} \cdot \frac{1}{4} \quad \frac{1}{4} \quad \frac{1}{4} - 1$

$\Rightarrow \{0, 10, 110, 111\}$

$\frac{1}{8} \cdot 1$

h) Extended ensemble. Non-symbol codes

- 7) a) $\text{length}_S =$, not prefix free chain, then
 b) $\text{length}_S =$
 c) $\text{length}_S = f(x) + f(y)$
 d) Code not added even b/c
 e) Looking for key + odds $(x_1 y_2)$

$$= f(z) + f(y_2) + f(x_1 y)$$

Since $x_1 z_1 y_2$, then (z, y_2) is
 a packed chain

d) The data processing inequality implies that

$$I(x; y) \geq I(x; z)$$

and among a), we also have

$$I(z; y) \geq I(z; x)$$

But $I(z; x) = I(x; z)$ and so

$$I(x; z) \leq I(z; y)$$

and $I(x; z) \leq I(x; y)$.

The result follows.

i) $\phi(x, y, z) = \phi(z(y)) \phi(y(x)) \phi(x)$
 Since $x \perp \!\!\! \perp z \mid y$

ii) Yes, by i)

iii) Yes, by the data processing

iv) ~~Part b)~~ $I(x; y) \leq I(x; z) + I(y; z)$

so $I(x; y) \leq I(x; z) + I(y; z)$

min to both sides $I(x; y) \leq I(x; z) + I(y; z)$

so $I(x; y) \leq I(x; z) + I(y; z)$

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a)

If (X, Y, Z) is a Markov chain, then
 $X \perp Z | Y$ and thus
 $p(X, Z | Y) = p(X | Y) p(Z | Y)$

Thus $p(Z, Y, X) = p(Z) p(Y | Z) p(X | Y, Z)$

$$= p(Z) p(Y | Z) p(X | Y)$$

Since $X \perp Z | Y$. Thus (Z, Y, X) is
a Markov chain

5) The data processing inequality implies that

$$I(X; Y) \geq I(X; Z)$$

and using a), we also have

$$I(Z; Y) \geq I(Z; X)$$

But $I(Z; X) = I(X; Z)$ and so

$$I(X; Z) \leq I(Z; Y)$$

$$\text{and } I(X; Z) \leq I(X; Y).$$

The result follows.

c) i) $p(x, y, z) = p(z | g) p(g | x) p(x)$
Since $X \perp Z | Y$

ii) Yes, by i)

iii) Yes, by the data processing \neq
Part (i) showed $I(X, Z) \leq \min(I(X, Y), I(Y, Z))$

to capacity of C^{∞}

$$= \min_I I(X, Z) \leq \min \min(I(X, Y), I(Y, Z))$$

$$\leq \min(C_{xy}, C_{yz}) \quad (\text{The argument for } \text{ii) might not be the other in capacity.})$$

$$(v) \quad C^{xz} = \min_p I(x; z)$$

$$\text{then } f \leq \min_p \min(I(x; y), I(y; z))$$

The minimization with over p will also minimize $I(x; y)$, $I(y; z)$ or both.

If it minimize $I(x; y)$, then $I_{p_y}(y; z)$ must to greater that $C^{xy} = \min_p I(x; y)$.

Here p_y is the distribution of Y induced by X .
This is true because otherwise the user would have minimized $I_{p_y}(y; z)$.

In this case we get

$$C^{xz} \leq \min_p (C^{xy}, I_{p_y}(y; z))$$

The other case (same agent) implies

$$C^{xz} \leq \min_p (I_p(x; y), C^{yz}).$$

If \min_p minimizes both

$$\text{then } C^{xz} \leq \min \left[\min_p (C^{xy}, I_{p_y}(y; z)), \min_p (I_p(x; y), C^{yz}) \right]$$

$$= \min (C^{xy}, C^{yz}).$$

Intuition is simple send container - thinnest-fibre limit capacity of the whole.

Q9 a) By symmetry there is a binary symmetric channel $Q = \begin{bmatrix} 1-f & f \\ f & 1-f \end{bmatrix}$

where $f = p(\hat{X}=1 | X=1) = p(\hat{X}=-1 | X=1)$

$$= \int_{S-B}^0 \frac{1}{2B} dx \quad \text{where } B > 5$$

$$= \frac{B-5}{2B} (B > 5) \quad \begin{matrix} & & 1 \\ & & \text{---} \\ & & | \\ & & 5 \\ S-B & & \text{---} & & 1 \\ & & | & & \\ & & 5 & & \\ & & \text{---} & & \\ & & & & \end{matrix}$$

$$0 \quad 0 \quad (B \leq 5)$$

for the \hat{X}
when $B \leq 5$, $f=0$ and $Q^{XY} = I$.

Now let us just determine the capacity of the binary symmetric channel.

By we require $\max_p I(X; Y)$

where p is the distribution of X . Since Q^{XY} is symmetric, the optimal distribution is uniform. We thus just need to compute $I(X; Y)$ when $p_X = (\frac{1}{2}, \frac{1}{2})$.

the $I(X; Y) = H(Y) - H(Y|X)$

$$H(Y) = H_2(f p_0 + (1-f) p_1)$$

$$= H_2(f \cdot \frac{1}{2} + (1-f) \cdot \frac{1}{2}) = H_2(\frac{1}{2}) = 1.$$

$$H(Y|X) = \sum_x p(X=x) H(f) = H_2(f).$$

Thus $C^{XY} = H(Y) - H(Y|X)$
 $= 1 - H_2(f)$

So when $B > 5$, $C^{XY} = 1 - H_2\left(\frac{5-5}{2B}\right)$

- c) Yes, channel coding theorem
- d) No. CCT would require N . Large block length to get this close to the capacity.
Looking for 10 bits (all joint typically),
 $N=7$ all channels look like noisy
channels on uses large enough block sizes.
- e). Rate will be $\frac{1}{1024} (2^{-10})$!