

Q1

$$P(A \cup B | D)$$

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

(Using Conditional Probability)

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

(using set distribution)

$$= \frac{P(A \cap D) + P(B \cap D) - P(A \cap B \cap D)}{P(D)}$$

($P(X \cup Y) = P(X) + P(Y) - P(X \cap Y)$)

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

$$= P(A|D) + P(B|D) + P(A \cap B | D)$$

Proved

Q2 | Normal Case can be expressed as :

$$P_N(D_i) = 0.01 \quad \forall i \geq 1$$

where P_N is the probability in normal case.

Memory case can be expressed as :

$$P_M(D_1) = 0.01$$

$$P_M(D_i | D_{i-1}) = 2/5$$

$$\forall i > 1$$

$$P_M(D_i^c | D_{i-1}^c) = 1/165$$

and D_i is only dependent on D_{i-1} , independent of $D_1 - D_{i-2}$

$$(2) \quad P_M(D_1) = 0.01 \quad (\text{Given})$$

Let's assume $P_M(D_n) = 0.01$ for
some $n \geq 1$

$$\begin{aligned} P_M(D_{n+1}) &= P_M(D_{n+1} | D_n) P(D_n) \\ &\quad + P_M(D_{n+1} | D_n^c) P(D_n^c) \\ &= 2/5 \times 0.01 + (1/165) \times 0.99 \end{aligned}$$

$$\underline{P_M(D_{n+1}) = 0.01}$$

By induction

$$P_M(D_n) = 0.01 \quad \forall n \geq 1$$

(b) Let's first calculate the $P_8(B|B)$ & $P_8(E|B^c)$,
 later we can use this to compute the desired
 result.

Normal
 case

$$P_8(E|B) = P(D_1^c \cap D_2^c \cap D_3 \cap D_4 \cap D_5^c \cap D_6^c | B)$$

since in the normal case, all D_i are independent

$$= P_8(D_1^c | B) \times P_8(D_2^c | B) \times \dots$$

$$P_8(E|B) = (0.01)^2 (0.99)^4 \rightarrow \textcircled{1}$$

Memory case.

$$P_7(E|B^c) = P(D_1^c \cap D_2^c \cap D_3 \cap D_4 \cap D_5^c \cap D_6^c | B^c)$$

We can apply conditional probability & the fact
 that D_i is only dependent on D_{i-1} to reduce
 this to

$$= P(D_1^c | B^c) \times P(D_2^c | D_1^c \cap B^c) \times P(D_3 | D_2^c \cap B^c) \\
 \times P(D_4 | D_3 \cap B^c) \times P(D_5^c | D_4 \cap B^c) \\
 \times P(D_6^c | D_5^c \cap B^c)$$

$$= (0.99) \times \left(\frac{164}{165}\right)^2 \times \left(\frac{1}{165}\right) \times \left(\frac{2}{5}\right) \times \left(\frac{3}{5}\right) \rightarrow \textcircled{2}$$

Using Bayes Theorem

$$P_x(B|E) = \frac{P(E|B) P(B)}{P(E)}$$

$$= \frac{P(E|B) P(B)}{P(E|B) P(B) + P(E|B^c) P(B^c)}$$

Substituting values using (1) and (2) and solving gives

$$= ~~0.11898~~ 0.11898$$

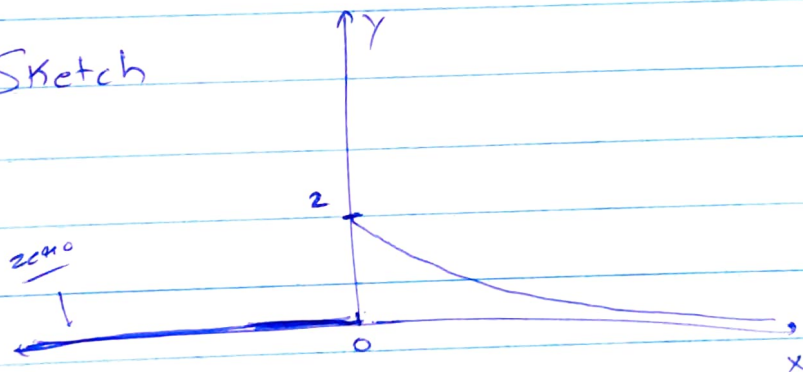
Q3] (a) $\int_{-\infty}^{\infty} f(x) = 1$

$$\int_0^{\infty} f(x) + \int_{-\infty}^0 f(x) = 1$$

$$\frac{c e^{-2x}}{(-2)} \Big|_0^{\infty} = 1$$

$$c = 2$$

Sketch



$$\begin{aligned} (b) \quad P_x(1 < x < 2) &= \int_1^2 2e^{-2x} \\ &= -e^{-2x} \Big|_1^2 \\ &= \frac{e^{-2} - e^{-4}}{e^4} \end{aligned}$$

Given:

Q4 $\log(x) \sim N(\mu, \sigma^2)$

$$\begin{aligned}\log(1/x) &= 1 - \log(x) \\ &\sim 1 - N(\mu, \sigma^2)\end{aligned}$$

Using $A \sim N(\mu, \sigma^2) \Rightarrow cA \sim N(c\mu, (c\sigma)^2)$
where c is a constant

and $c+A \sim N(c+\mu, \sigma^2)$

$$\log(1/x) \sim N(1-\mu, \sigma^2)$$

So $1/x$ is also log normally distributed,
but with a mean of $1-\mu$.