

1]

1.1] (a) Sample space =  $\{HH, HT, TH, TT\}$

(b) Event space =  $\{\phi, \{HH\}, \{HT\}, \{TH\}, \{TT\}, \{HH, HT\}, \{HH, TH\}, \{HH, TT\}, \{HT, TH\}, \{HT, TT\}, \{TH, TT\}, \{HH, HT, TH\}, \{HH, HT, TT\}, \{HH, TH, TT\}, \{HT, TH, TT\}, \{HH, HT, TH, TT\}\}$

Event space is power set of sample space.

(c) (i) probability of each

$$\Omega = \{HH, TH, HT, TT\}$$

$$P(\Omega) = 1$$

$$P(HH \cup TH \cup HT \cup TT) = 1$$

Since  $\{HH\}, \{TH\}, \{HT\}, \{TT\}$  are mutually exclusive

$$\& P(HH) = P(TH) = P(HT) = P(TT)$$

$$4 \cdot P(HH) = 1$$

probability of each elementary event =  $\frac{1}{4}$

(ii) probability of at least one head =  $P(\{HH\} \cup \{HT\} \cup \{TH\})$

$$= \frac{1}{4} + \frac{1}{4} + \frac{1}{4}$$

since they

$$= \frac{3}{4}$$

(iii) probability of exactly one head =  $P(\{HT\} \cup \{TH\})$

$$= \frac{1}{4} + \frac{1}{4}$$

$$= \frac{1}{2}$$

2]

2.1)

$$f(k, n, p) = \frac{n!}{k! (n-k)!} p^k (1-p)^{n-k}$$

probability of correctly recognizing a word ( $p$ ) = 0.9

$$p(\text{recognizing 45 out of 50 words}) = f(45, 50, 0.9)$$

$$= \frac{50!}{45! 5!} (0.9)^{45} (0.1)^5$$

2.2)

$$f(k, \lambda) = \frac{\lambda^k e^{-\lambda}}{k!}; \lambda = 10$$

(a)

probability that zero road accidents in that day

$$f(0, \lambda) = \frac{10^0 e^{-10}}{0!} = e^{-10}$$

(b)

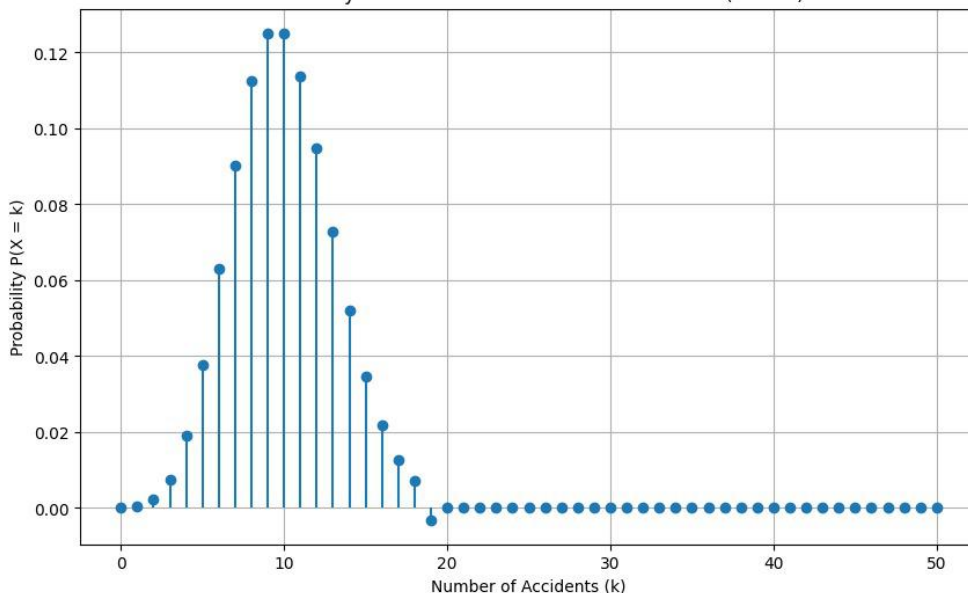
probability of occurrence of 8 or 9 road accidents in a day  $\Rightarrow f(8, \lambda) + f(9, \lambda)$

$$= \frac{10^8 e^{-10}}{8!} + \frac{10^9 e^{-10}}{9!}$$

$$= e^{-10} \left[ \frac{10^8}{8!} + \frac{10^9}{9!} \right]$$

(c)

Probability Mass Function of Poisson Distribution ( $\lambda = 10$ )



3.1] P.D.F for normal distribution  $f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

(a) P.D.F at  $x=0$ ;  $\mu=1$ ,  $\sigma=1$

$$p = f(0) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}}$$

(b) P.D.F at  $x=1$ ,  $\mu=0$ ,  $\sigma=1$

$$p = f(1) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}}$$

(c)  $P(x_1 \leq x \leq x_2) = 0.3$

$$P(x_1 \leq x \leq x_3) = 0.45$$

$$P(x_2 \leq x \leq x_3) = ?$$

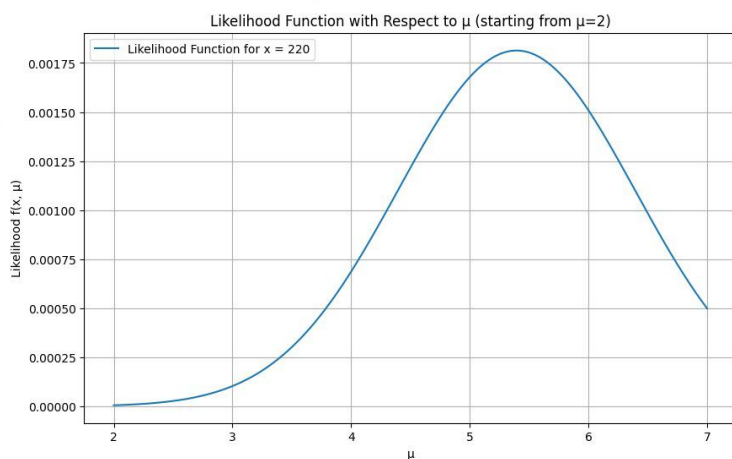
$$P(x_1 \leq x \leq x_3) = P(x_1 \leq x \leq x_2) + P(x_2 \leq x \leq x_3)$$

$$0.45 = 0.3 + P(x_2 \leq x \leq x_3)$$

$$P(x_2 \leq x \leq x_3) = 0.15 //$$

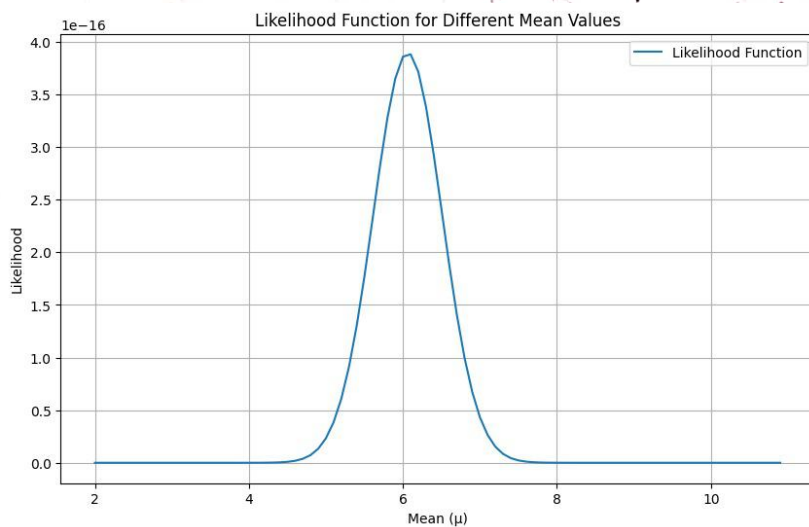
4]  
4.1]  $f(x, \mu) = \frac{1}{x\sqrt{2\pi}} e^{-\frac{(\log x - \mu)^2}{2}}$

(a)  $L(\mu|x) = f(x, \mu) = f(220, \mu) = \frac{1}{220\sqrt{2\pi}} e^{-\frac{(\log 220 - \mu)^2}{2}}$



(b)  $x = [3303.25, 443, 220, 560, 880]$

$$L(\mu | x) = f(x, \mu) = \frac{1}{(\prod_{i=1}^n x_i) (2\pi)^{n/2}} e^{-\frac{1}{2} \sum_{i=1}^n (\log x_i - \mu)^2}$$



(c) from the above graph, ~~the~~ Likelihood is max

(at  $\mu = 6$ )

$$L(\mu = 6 | x) = \frac{1}{(\prod_{i=1}^n x_i) (2\pi)^{n/2}} e^{-\frac{1}{2} \sum_{i=1}^n (\log x_i - 6)^2}$$

$$L(\mu = 6 | x) = 0.12$$

$$\frac{\partial}{\partial \mu} L(\mu | x) = \frac{\partial}{\partial \mu} \left[ \frac{1}{(\prod_{i=1}^n x_i) (2\pi)^{n/2}} e^{-\frac{1}{2} \sum_{i=1}^n (\log x_i - \mu)^2} \right]$$

$$= \frac{1}{(\prod_{i=1}^n x_i) (2\pi)^{n/2}} e^{-\frac{1}{2} \sum_{i=1}^n (\log x_i - \mu)^2} \cdot \sum_{i=1}^n (\log x_i - \mu)$$