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210205

CBS 698C

# Assignment - 1

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## Part - 1

i) a) Sample space  $\Omega = \{HH, HT, TH, TT\}$

b) Event space  $\Sigma = \{\emptyset, \{HH\}, \{HT\}, \{TH\}, \{TT\}, \{HH, TH\}, \{HH, HT\}, \{HH, TT\}, \{HT, TH\}, \{HT, TT\}, \{TH, TT\}, \{HH, HT, TH\}, \{HT, TH, TT\}, \{HH, HT, TT\}, \{HH, TH, TT\}, \{HH, HT, TH, TT\}\}$

c) i)  $P(\{HH\}) = P(\{HT\}) = P(\{TH\}) = P(\{TT\}) = p$   
 also  $p(\{HH\}) + p(\{HT\}) + p(\{TH\}) + p(\{TT\}) = 1$

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

ii) Probability of at least 1 head  
 $P(\{HH, TH, HT\}) = 3/4$

(iii) Probability of exact 1 head :-  
 $P(\{HT, TH\}) = 1/2$

## Part - 2

2.1)  $p$  = probability of recognizing correct word = 0.9.

$k$  = number of correctly recognized words

$n$  = total number of words.

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

$$p = 0.9 \quad n = 50 \quad k = 45$$

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

$$f(45, 50, 0.9) = 0.184 \\ = 0.18$$

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

$k$  road accidents in a day.

(a) Probability of zero accidents.

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

(b) Probability of more than 7 & less than 10.

$$f(8, 10) + f(9, 10) = \frac{(10)^8 e^{-10}}{8!} + \frac{(10)^9 e^{-10}}{9!}$$

$$= 0.237$$

(c) graph

Part - 3

(3.1)  $f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

$X$  random variable, normally distributed with above probability density function.

(a)

Probability density of obtaining  $x=0$   
given that  $\mu=1, \sigma=1$

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

$$= \frac{1}{\sqrt{2\pi}e} = 0.24$$

(b)

Probability density  $x=0, \mu=0, \sigma=1$

$$f(1) = \frac{1}{\sqrt{2\pi}} e^{\frac{-(1-0)^2}{2}} = \frac{1}{\sqrt{2\pi}e} = 0.24$$

(c)

$$P(x_1 \leq X \leq x_2) = 0.3$$

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

$$P(x_2 \leq X \leq x_3) = P(x_1 \leq X \leq x_3) - P(x_1 \leq X \leq x_2)$$

$$= 0.45 - 0.3$$

$$= 0.15$$

Part-4

(4.1)

Graph.

$$(c) \quad \mu = 6.04 \quad \mu \approx 6$$

```
In [ ]: import numpy as np
import matplotlib.pyplot as plt
from math import factorial, exp

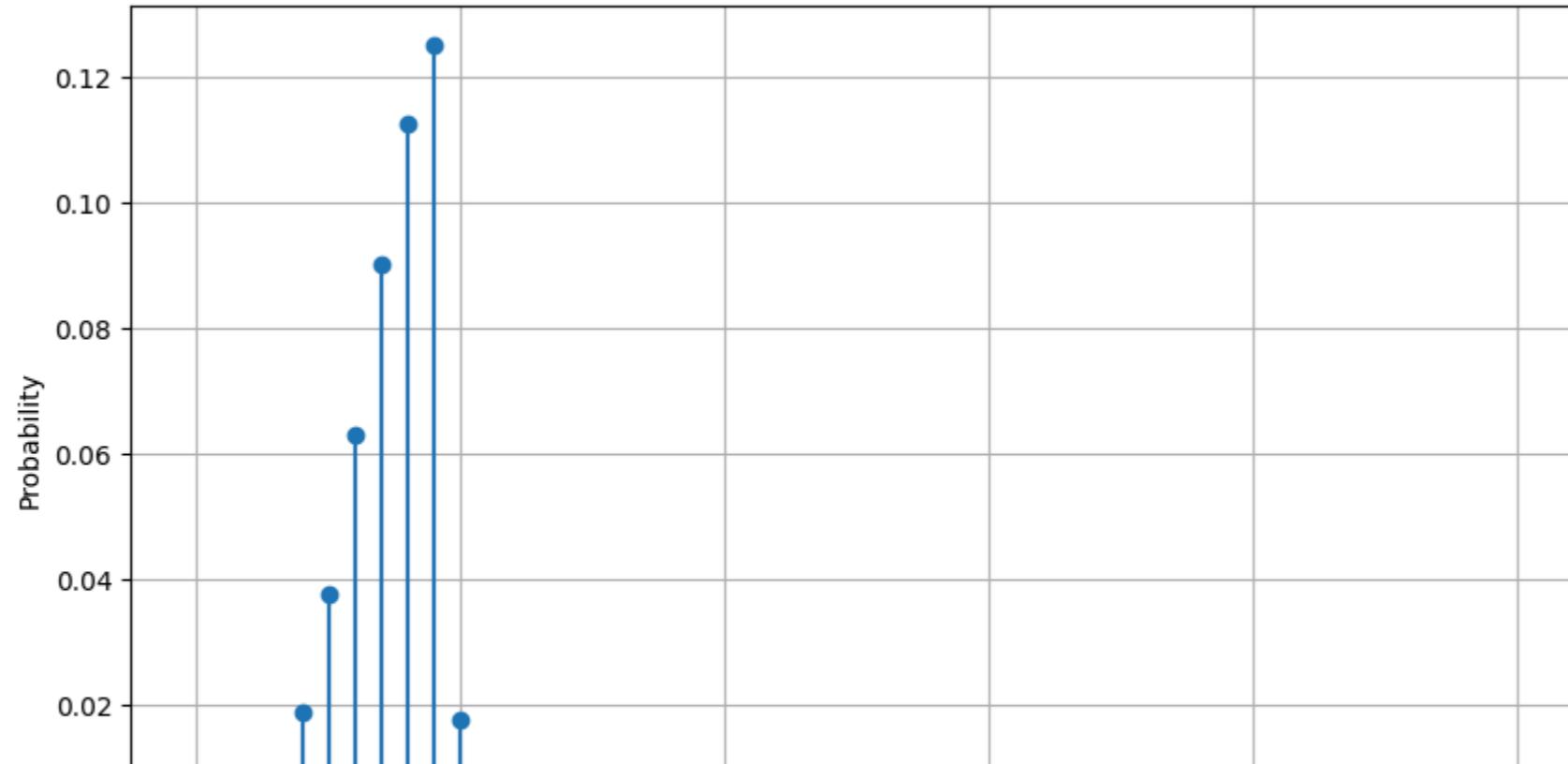
def poisson_pmf(k, lambd):
    return (lambd**k * exp(-lambd)) / factorial(k)

lambd = 10
k_values = np.arange(0, 51)
probabilities = [poisson_pmf(k, lambd) for k in k_values]

plt.figure(figsize=(10, 6))
plt.stem(k_values, probabilities, basefmt=" ", use_line_collection=True)

plt.xlabel('Number of Road Accidents (k)')
plt.ylabel('Probability')
plt.title('Poisson Distribution PMF ( $\lambda = 10$ )')
plt.grid(True)
plt.show()

C:\Users\aryan\AppData\Local\Temp\ipykernel_18896\4063919027.py:16: MatplotlibDeprecationWarning: The 'use_line_collection' parameter of stem() was deprecated in Matplotlib 3.6 and will be removed two minor releases later. If any parameter follows 'use_line_collection', they should be passed as keyword, not positionally.
plt.stem(k_values, probabilities, basefmt=" ", use_line_collection=True)
```



```
In [ ]: import numpy as np
import matplotlib.pyplot as plt

def f(x, mu):
    return (1 / (x * np.sqrt(2 * np.pi))) * np.exp(-(np.log(x) - mu)**2 / 2)

x = 220

mu_values = np.linspace(-10, 10, 400)
f_values = f(x, mu_values)

plt.figure(figsize=(10, 6))
plt.plot(mu_values, f_values, label='Likelihood function')
plt.xlabel('mu')
plt.ylabel('f(x, mu)')
plt.title('Likelihood Function with x = 220')
plt.legend()
plt.grid(True)
plt.show()
```



```
In [ ]: x_values = np.array([303, 443, 220, 560, 880])

def combined_likelihood(mu, x_values):
    likelihood = 1
    for x in x_values:
        likelihood *= f(x, mu)
    return likelihood

combined_f_values = [combined_likelihood(mu, x_values) for mu in mu_values]

plt.figure(figsize=(10, 6))
plt.plot(mu_values, combined_f_values, label='Combined Likelihood function', color='red')
plt.xlabel('mu')
plt.ylabel('Likelihood')
plt.title('Combined Likelihood Function for Observed Recognition Times')
plt.legend()
plt.grid(True)
# plt.show()

max_likelihood_mu = mu_values[np.argmax(combined_f_values)]
print(f"The value of  $\mu$  that maximizes the likelihood is approximately: {max_likelihood_mu}")

plt.scatter(max_likelihood_mu, np.max(combined_f_values), color='green', label='Maximum Likelihood Point')

plt.legend()
plt.show()
```

The value of  $\mu$  that maximizes the likelihood is approximately: 6.0401002506265655

