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FACULTY OF ENGINEERING & TECHNOLOGY
DEPARTMENT OF INFORMATION AND
COMMUNICATION ENGINEERING
COURSE NAME: ENGINEERING STATISTICS
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ASSIGNMENT TOPIC: MLE ILLUSTRATION FROM
POISSON AND NORMAL DISTRIBUTIONS

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WHAT IS MAXIMUM LIKELIHOOD ESTIMATION (MLE)

- MLE is a method to estimate parameters of a distribution based on observed data.
- It finds the values that maximize the likelihood of the data.
- Widely used in statistics, machine learning, and engineering.

GENERAL MLE STEPS

- Define the probability distribution of the data.
- Construct the likelihood function.
- Take log-likelihood for easier differentiation.
- Differentiate, set derivative = 0, and solve for the parameter(s).

POISSON DISTRIBUTION

Models count data (e.g., number of calls per hour)

PMF: $P(X = x) = (e^{-\lambda} * \lambda^x) / x!$, $x = 0, 1, 2, \dots$

Parameter: λ = average rate of events

MLE FOR POISSON DISTRIBUTION

Likelihood: $L(\lambda) = \prod [e^{(-\lambda)} * \lambda^{x_i} / x_i!]$

Log-likelihood: $\ln L(\lambda) = -n\lambda + \ln(\lambda) * \sum x_i + C$

SOLVING FOR MLE – POISSON

$$d/d\lambda \ln L = -n + (1/\lambda) * \sum x_i = 0$$

$$\rightarrow \hat{\lambda} = \sum x_i / n = \bar{x}$$

MLE of λ is the sample mean.

NORMAL DISTRIBUTION

Models continuous data (e.g., height, weight)

$$\text{PDF: } f(\mathbf{x}) = (1/\sqrt{2\pi\sigma^2}) * e^{-(\mathbf{x} - \mu)^2 / 2\sigma^2}$$

Parameters: μ = mean, σ^2 = variance

MLE FOR NORMAL DISTRIBUTION

Log-likelihood:

$$\ln L = -n/2 \ln(2\pi) - n/2 \ln(\sigma^2) - (1/2\sigma^2) \sum (\mathbf{x}_i - \mu)^2$$

MLE RESULTS – NORMAL DISTRIBUTION

Estimated Mean: $\hat{\mu} = \bar{x}$

Estimated Variance: $\hat{\sigma}^2 = (1/n) * \sum (x_i - \bar{x})^2$

MLE SUMMARY TABLE

Poisson: $\hat{\lambda} = \bar{\mathbf{x}}$

Normal: $\hat{\mu} = \bar{\mathbf{x}}$

Normal: $\hat{\sigma}^2 = (1/n) * \sum(\mathbf{x}_i - \bar{\mathbf{x}})^2$

APPLICATIONS OF MLE IN ENGINEERING

- Signal processing
- Quality control
- System reliability analysis
- Parameter estimation in communication systems
- Machine learning model fitting

CONCLUSION

- MLE finds parameters that make observed data most likely.
- Poisson and Normal MLEs are intuitive and based on sample stats.
- Widely applicable in engineering and data analysis.