

# Communications and Networks: Communications Option

## Probability: Sum of Independent Gaussian Variables

Let  $Z = X + Y$  where  $X$  and  $Y$  are two independent Gaussian r.v.'s with means  $\mu_X$  and  $\mu_Y$ , and variances  $\sigma_X^2$  and  $\sigma_Y^2$ , respectively, i.e.,  $X \sim \mathcal{N}(\mu_X, \sigma_X^2)$  and  $Y \sim \mathcal{N}(\mu_Y, \sigma_Y^2)$ .

1. Determine the distribution of  $Z$ .
2. Determine the mean  $\mu_Z$  and the variance  $\sigma_Z^2$  of  $Z$ .

If you want to use characteristic functions, then the characteristic function of a Gaussian r.v.  $\theta$  with mean  $\mu_\theta$  and variance  $\sigma_\theta^2$  is  $\phi_\theta(w) = \mathbb{E}\{e^{jw\theta}\} = \exp(+jw\mu_\theta - \frac{1}{2}w^2\sigma_\theta^2)$ .

## Communications: Photon Bucket Array Reception

Consider a deep-space optical communications link employing array of  $N$  telescopes. The photodetector outputs  $X_1, X_2, \dots, X_N$  can be characterized by independent, identically distributed (i.i.d.) random variables (r.v.'s) with Poisson probability mass function (pmf), i.e.,

$$p_X(x|\lambda) = \frac{\lambda^x e^{-\lambda}}{x!}, \quad x = 0, 1, 2, \dots$$

where  $\lambda > 0$ . Recall that the mean and variance of  $X_1$  are, respectively,

$$\begin{aligned} \mathbb{E}\{X_1\} &= \lambda \\ \mathbb{V}\{X_1\} &= \lambda. \end{aligned}$$

In many subsystems, it is desirable to use the estimator  $\hat{\lambda}(\mathbf{X})$  of  $\lambda$  based on the observation  $\mathbf{X}$ .

1. Find a sufficient statistic for  $\lambda$  based on observation  $\mathbf{X} = [X_1 \ X_2 \ \dots \ X_N]$ .
2. Find a minimal sufficient statistic for  $\lambda$  based on observation  $\mathbf{X}$ .
3. Derive the maximum likelihood estimator  $\hat{\lambda}_{\text{ML}}(\mathbf{X})$  of  $\lambda$  based on the observation  $\mathbf{X}$ .

4. Assume that your answer in part (b) is also a complete sufficient statistic for  $\lambda$ . Find a uniformly minimum variance unbiased (UMVU) estimator  $\hat{\lambda}_{\text{UMVU}}(\mathbf{X})$  of  $\lambda$  based on the observation  $\mathbf{X}$ .

**Bonus Questions** (attempt only if you completed all the above questions)

5. Prove that your answer in part (b) is also a complete sufficient statistic for  $\lambda$  based on observation  $\mathbf{X}$ .
6. Derive the Fisher information  $I_{\mathbf{X}}(\lambda)$  in  $\mathbf{X}$  about  $\lambda$ . Does  $\hat{\lambda}_{\text{UMVU}}(\mathbf{X})$  satisfy the information inequality (or Cramer-Rao lower bound) with an equality?