Problem 1: Suppose $X \sim \mathcal{N}(\mu, 1)$ and that we define two estimators by

$$\hat{\mu}_1 = X$$
 and $\hat{\mu}_2 = 3$.

Compute the mean squared errors of $\hat{\mu}_1$ and $\hat{\mu}_2$.

Since $\hat{\mu}_2 = 3$ is constant, we have

$$R(\mu, \hat{\mu}_2) = E((\mu - 3)^2) = (\mu - 3)^2.$$

By linearity of expectation, we have

$$R(\mu, \hat{\mu}_1) = E((\mu - \hat{\mu}_1)^2) = \mu^2 - 2\mu E(\hat{\mu}_1) + E(\hat{\mu}_1^2).$$

But we have $E(\hat{\mu}_1) = \mu$, and from the short-cut formula for variance, we have

$$E(\widehat{\mu}_1^2) = 1 + E(\widehat{\mu}_1)^2 = 1 + \mu^2.$$

Thus,

$$R(\mu, \hat{\mu}_1) = \mu^2 - 2\mu^2 + 1 + \mu^2 = 1.$$

Problem 2: Suppose that the number X of defects on a roll of magnetic recording tape has a Poisson distribution for which the mean μ is either 1.0 or 1.5. Supposing that the prior probability function on μ is given by

$$p(\mu) = \begin{cases} 0.4 &: \mu = 1.0, \\ 0.6 &: \mu = 1.5, \\ 0 &: \text{otherwise,} \end{cases}$$

and that a selected roll of tape has X=3 defects, what is the posterior probability function of μ ? What is $\hat{\mu}_{MAP}$?

From Bayes' Theorem and the Law of Total Probability, we have

$$p(\mu|x=3) = \frac{p(x=3|\mu)p(\mu)}{p(x=3|\mu=1.0)p(\mu=1.0) + p(x=3|\mu=1.5)p(\mu=1.5)}$$

$$= \frac{\frac{1}{3!}\mu^3e^{-\mu}p(\mu)}{\frac{1}{3!}e^{-1}\cdot 0.4 + \frac{1}{3!}(1.5)^3e^{-1.5}\cdot 0.6}$$

$$\approx \begin{cases} 0.246 & : \mu=1.0, \\ 0.754 & : \mu=1.5, \\ 0 & : \text{otherwise.} \end{cases}$$

From this, we see that $\hat{\mu}_{MAP} = 1.5$.