## Quiz 3

## Select one problem!

Weighted Square Error Loss. We have seen that under squared error loss the Bayes rule [minimizer of Bayes risk  $r(\pi, \delta)$  or equivalently conditional on x minimizer of posterior expected loss  $E^{\theta|x}L(\theta, \delta(x))$  is the posterior expectation,

$$
\delta_{\pi}(x) = E^{\theta|x} \theta = \int_{\Theta} \theta \pi(\theta|x) d\theta = \frac{\int_{\Theta} \theta f(x|\theta) \pi(\theta) d\theta}{\int_{\Theta} f(x|\theta) \pi(\theta) d\theta}.
$$

If the loss is weighted square error

$$
L(\theta, a) = w(\theta)(\theta - a)^2,
$$

the Bayes rule is

$$
\delta_{\pi}(x) = \frac{E^{\theta|x}\theta w(\theta)}{E^{\theta|x}w(\theta)} = \frac{\int_{\Theta} \theta w(\theta) f(x|\theta) \pi(\theta) d\theta}{\int_{\Theta} w(\theta) f(x|\theta) \pi(\theta) d\theta}.
$$

To obtain a particular quality control score, 10 independent zero-mean half-normal attributes with the same scale are squared and summed up. The obtained score  $[X|\theta]$  is a scaled  $\chi^2_{10}$  with the likelihood from  $\mathcal{G}(5, 2\theta)$  family,

$$
f(x|\theta) = \frac{1}{\Gamma(5)(2\theta)^5} x^{5-1} e^{-x/(2\theta)}, \theta > 0.
$$

The parameter  $\theta$  is of interest.

From the past experience and experts' input you know that  $\theta$  is skewed to the right, most likely to cluster around 2 with an average close to 4. Use this information to elicit an inverse gamma prior on  $\theta$ ,  $[\theta] \sim \mathcal{IG}(\alpha, \beta).$ <sup>1</sup>

S  $\mathcal{L}\mathcal{G}(\alpha,\beta)$ .<br>Assume that the loss function is weighted square error loss  $L(\theta, a) = \left(\frac{a}{\theta} - 1\right)$  $\big)^2$ . This reduces penalty if the parameter  $\theta$  is large in magnitude (scale of attributes are large). If the statistician uses  $\delta(x) = 0$ , the loss is 1, irrespective of the value for  $\theta$ . On the other hand, if the parameter  $\theta$  is close 0 the penalty for an error is severe.

A single measurement of  $X = 16$  is available. What is the Bayes estimator of  $\theta$  under the weighted squared error loss?

## Solution.

If the prior is inverse-gamma  $\mathcal{IG}(\alpha, \beta)$  then (as in handout 0) the mean is  $\frac{1}{\beta(\alpha-1)}$  and the mode is  $\frac{1}{\beta(\alpha+1)}$ . Solving

$$
\frac{1}{\beta(\alpha - 1)} = 4
$$
  

$$
\frac{1}{\beta(\alpha + 1)} = 2
$$

<sup>1</sup>Recall that for a random variable with  $IG(\alpha, \beta)$  distribution the mean is  $\frac{1}{\beta(\alpha-1)}$  and the mode is  $\frac{1}{\beta(\alpha+1)}$ 

we get  $\alpha = 3$  and  $\beta = 1/8$ .

The weight function is  $w(\theta) = 1/\theta^2$ , from  $L(\theta, a) = \left(\frac{a}{\theta} - 1\right)$  $\big)^2 = \frac{1}{\theta^2} (\theta - a)^2.$ 

Since  $f(x|\theta)\pi(\theta) \propto \frac{1}{\theta^{(\alpha+5)+1}} \exp\left\{-\frac{1}{\theta}(\frac{x}{2}+\frac{1}{\beta})\right\}$ , then by completing to appropriate inverse gamma density that integrates to 1,

$$
\int \theta w(\theta) f(x|\theta) \pi(\theta) d\theta = \Gamma(\alpha + 6) \times (\frac{x}{2} + \frac{1}{\beta})^{-(\alpha + 6)}
$$

$$
\int w(\theta) f(x|\theta) \pi(\theta) d\theta = \Gamma(\alpha + 7) \times (\frac{x}{2} + \frac{1}{\beta})^{-(\alpha + 7)}.
$$

The Bayes rule is

$$
\delta_{\pi}(x) = \frac{\Gamma(\alpha+6) \times (\frac{x}{2} + \frac{1}{\beta})^{-(\alpha+6)}}{\Gamma(\alpha+7) \times (\frac{x}{2} + \frac{1}{\beta})^{-(\alpha+7)}} = \frac{x/2 + 1/\beta}{\alpha+6}.
$$

For  $X = 16, \alpha = 3$ , and  $\beta = 1/8$ , the Bayes rule is  $\frac{16}{9}$ .

Equivalently, show that posterior is  $IG(8, \frac{1}{x/2+8})$  and find the ratio

$$
\frac{E^{\theta|x| \frac{1}{\theta}}}{E^{\theta|x| \frac{1}{\theta^2}}}.
$$

Traffic. Marietta Traffic Authority is worried about the repeated accidents at the intersection of Canton and Piedmont Roads. As a city-engineer you would like to find the accident rate, even better, a credible set.

A well known model for modeling the number of road accidents in a particular space/time window is the Poisson distribution. Assume that  $X$  represents the number of accidents in a 2 month period at the intersection od Canton and Piedmont Roads. Assume that  $[X|\theta] \sim \mathcal{P}oi(\theta)$ . Nothing is known a priori about  $\theta$ , so it is reasonable to assume the noninformative prior

$$
\pi(\theta) = \frac{1}{\theta} \mathbf{1}(0 < \theta < \infty).
$$

In the six most recent two-month periods the following realizations for  $X$  are observed:  $3 \quad 2 \quad 1 \quad 1 \quad 3$ 2.

(a) Compare the Bayes estimator and the MLE (For Poisson, recall,  $\hat{\theta}_{MLE} = \bar{X}$ ).

(b) An  $(1 - \alpha)100\%$  HPD credible set has no finite form. Describe in principle how would you calculate this set approximately on a computer. Your description may be informal.

(c) The Mathematica code



gives that the median of the posterior is approximately 1.9447275.

If you test the hypotheses

$$
H_0: \theta \ge 2 \qquad \quad vs \qquad \quad H_1: \theta < 2,
$$

in Bayesian fashion, which one will be favored?

## Solution.

(a) The posterior is given by

$$
\pi(\theta|x) \propto e^{-n\theta} \theta^{\sum x_i - 1} \propto e^{-6\theta} \theta^{12-1}.
$$

The Bayes estimate  $\theta_{\text{Bayes}} = 12 \times \frac{1}{6} = 2$ . This coincides with the MLE,  $\bar{X} = 2$ . This is expected since the prior is non-informative.

(b) Numerically solve the equation  $\pi(\theta|x) = k$ . Select k in the range to obtain two real solutions  $\theta_1$  and  $\theta_2$  (the posterior density is unimodal). Now, try various k until  $\int_{\theta_1}^{\theta^2} \pi(\theta|x) d\theta$  is close to  $1 - \alpha$  $\int_{\theta_1}^{\theta} \pi(\theta|x) d\theta$  is close to  $1 - \alpha$ . This integral is a difference of two incomplete gammas and is available in many software packages.

(c) The median is less than 2, meaning that  $P(\theta < 2) > 1/2$ . That means  $H_1$  is favored.