

# Practice Midterm w/ Solutions

*QPM II*

*October 3, 2018*

1. We observe iid data  $X_1, \dots, X_n$  that represents the number of faculty in a department who leave each year due to retirement, failed retentions, and tenure denials. Your dean tells you that you should model this using a Poisson distribution.

$$X_i \sim \frac{\lambda^x e^{-\lambda}}{x!}$$

- a. Show that this data is in the exponential family of distributions of the form  $h(x) \exp(\eta(\theta)T(x) - A(\eta))$

$$\begin{aligned} h(x) &= x^0 \\ \eta(\theta) &= \log(\lambda) \\ T(x) &= x \\ A(\eta) &= \exp(\eta) \end{aligned}$$

- b. Calculate the likelihood.

$$\frac{\lambda^{\sum^n x_i} \exp(-n\lambda)}{\prod^n x_i!}$$

- c. Calculate the log-likelihood.

$$\log(\lambda) \sum^n x_i - n\lambda - \sum^n x_i!$$

- d. Find the MLE for  $\lambda$ .

$$\bar{x}$$

- e. Show that the MLE is unbiased and consistent.

$$\begin{aligned} E\left[\frac{\sum x}{n}\right] &= \lambda \\ \text{Var}\left[\frac{\sum x}{n}\right] &= \frac{\lambda}{n} \end{aligned}$$

- f. Prove (directly) that the MLE is a sufficient statistic for  $\lambda$ , given that  $\text{Pois}(\alpha) + \text{Pois}(\beta) \sim \text{Pois}(\alpha + \beta)$ .  
**Do not use the factorization theorem, or your knowledge of the exponential family form.**

$$\frac{\prod^n (\sum x_i)!}{\prod^n x_i!} n^{\sum x_i} = f(x)$$

- g. Find the asymptotic distribution of the MLE for  $\lambda$  and show that it is asymptotically efficient.

$$N\left(\lambda, \frac{\lambda}{n}\right)$$

$$CRLB = Var(\hat{\lambda})$$

- h. Calculate a 95% confidence interval for the MLE.

$$\bar{x} \pm 1.96 \left( \sqrt{\frac{\bar{x}}{n}} \right)$$

- i. Using the delta method, find the asymptotic distribution of  $2\sqrt{\lambda}$ .

$$N\left(2\sqrt{\lambda}, \frac{1}{n}\right)$$

2. Your dean points out that the asymptotic properties of the MLE are not relevant because the sample size is so small. Your data consists of the following:

$$\mathbf{x} = (2, 2, 1, 3, 2, 1, 0, 4)$$

- a. Estimate the standard error of the asymptotic distribution using the non-parametric bootstrap. Set your number of bootstraps to  $10^4$ .

```
## $theta_hat
## [1] 1.875
##
## $se_boot
## [1] 0.4105978
##
## $n_boot
## [1] 10000
```

- b. Unrelated to the dean's request, you're interested in running a computational check for your answer to Problem 1-i. Estimate the standard error of the asymptotic distribution of  $2\sqrt{\lambda}$  using the parametric bootstrap, given  $\lambda = 4$ . Set your number of bootstraps to  $10^4$ . Use  $n = 100$  for each bootstrap.

```
## $theta_hat
## [1] 4
##
## $se_boot
## [1] 0.09983653
##
## $n_boot
## [1] 10000
```

3. No luck. The dean's office remains skeptical of your ability to estimate the asymptotic distribution using such a small sample. You decide to pull out your BBG (big Bayesian guns). You decide to use a Bayesian prior with

$$\pi(\lambda) = \frac{\beta^\alpha}{\Gamma(\alpha)} \lambda^{\alpha-1} e^{-\beta\lambda}$$

- a. Given  $\alpha = 10$  and  $\beta = 5$ , find the posterior distribution of  $\lambda$

$$p(\lambda|x) = \text{Gamma}(\lambda; \alpha = 25, \beta = 13)$$

- b. Using this posterior, find the  $E(\lambda|\mathbf{x})$  and the 95% highest posterior density. (Partial credit for the 95% credible interval.)

$$E[p(\lambda|x)] = \frac{25}{13}$$

```
## [1] "95% CI"
##      2.5%      97.5%
## 1.239392 2.735195
## [1] "95% highest posterior density"
##      lower      upper
## 1.190282 2.672481
## attr(,"credMass")
## [1] 0.95
```