

# Problem Set 2

Jacob M. Montgomery

9/15/2017

## The gamma distribution

Let  $X$  be a random variable that is distributed according to the gamma distribution such that:

$$X \sim \text{Gamma}(\alpha, \beta) = \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)}$$

where  $\Gamma()$  is the gamma function (look it up).

1. Show that this distribution belongs in the exponential family. (Hint: Get *everything* into the exponent.)
2. Assume that  $\alpha$  is unknown and  $\beta$  is known. Find the sufficient statistic for  $\alpha$ . (Hint: Factorization theorem.)
3. Now let  $(X_1, X_2, \dots, X_n)$  be iid variables drawn from the gamma distribution. Find the Likelihood function.
4. Assume that  $\beta$  is unknown and  $\alpha$  is known. Find the Fisher's information for  $\beta$ .

## Let's try that again

Assume that  $(X_1, X_2, \dots, X_n)$  are drawn from the following pdf

$$f(X|\theta) = \frac{\Gamma(2\theta)}{\Gamma(\theta)^2} [x(1-x)]^{\theta-1}$$

5. Show that this pdf is a member of the exponential family.
6. Find the sufficient statistic for  $(X_1, X_2, \dots, X_n)$ .
7. How is your answer to (5) related to your answer to (6) (Hint: Look at the function of  $x$  in the exponent). Prove that this relationship will always hold for all members of the exponential family.

## Estimating the population variance

8. Let  $(X_1, X_2, \dots, X_n)$  be iid normally distributed data with mean  $\mu$  and variance  $\sigma^2$ . Show that the sample variance is an unbiased estimator for  $\sigma^2$ .
9. Show that it is a consistent estimator for  $\sigma^2$ .

## Bernoulli

10. Let  $X_1, \dots, X_n \sim \text{Bern}(p)$  and let  $\hat{p} = \frac{\sum X_i}{n}$ . Show that  $\hat{p}$  is a consistent estimator for  $p$  (ignoring the fact that we have already proved this more generically, but you can follow along using the same basic proof.).

### Normal variance

11. Let  $X_1, \dots, X_n$  be iid  $N(\mu, \sigma^2)$ . The statistic  $S^2$  is an unbiased estimator of  $\sigma^2$  (you showed this above). Further,

$$E(S^2 - \sigma^2) = \text{Var}(S^2) = \frac{2\sigma^4}{n-1}$$

(assertion). Show that  $S^2$  does not attain the information bound.

- Find the log-likelihood.
  - Take the second derivative in terms of  $\sigma^2$ .
  - Take the expected value and multiply by  $-1$ .
  - Show that is less than  $(\frac{2\sigma^4}{n-1})^{-1}$ .
12. Is this estimator efficient (using the finite sample definition)?

### And a few more things

13. Find the log likelihood for the normal distribution with unknown mean and variance.
14. Find the second derivative of the log-likelihood in terms of  $\mu$  and  $\sigma^2$ , noting that the result is going to be a Hessian (and it is going to be a bit ugly). You might want to check your answer with the TA before trying the next part.
15. Remember your definitions for  $\sigma^2$  and  $\mu$  in terms of expectations. Find the Fisher information for the normal distribution. Lots of things are going to cancel, and the answer is going to look pretty simple (it will be a diagonal matrix). Just do your algebra carefully and do your expectations right.