## Likelihood Theory and Extensions (BIOS:7110) Breheny

## Assignment 6

Due: Monday, October 19

- 1. Completing the square.
  - (a) Show that if **A** is a symmetric, positive definite matrix, then any equation of the form

$$\mathbf{x}^{\mathsf{T}} \mathbf{A} \mathbf{x} + \mathbf{b}^{\mathsf{T}} \mathbf{x} + C = 0$$

can be rewritten as

$$(\mathbf{x} - \mathbf{m})^{\mathsf{T}} \mathbf{A} (\mathbf{x} - \mathbf{m}) = D;$$

provide expressions for  $\mathbf{m}$  and D in terms of  $\mathbf{A}$ ,  $\mathbf{b}$ , and C.

(b) Suppose we have the Taylor series expansion

$$\frac{1}{2}(\boldsymbol{\theta} - \tilde{\boldsymbol{\theta}})^{\top} \mathbf{H}(\boldsymbol{\theta} - \tilde{\boldsymbol{\theta}}) + \mathbf{u}^{\top}(\boldsymbol{\theta} - \tilde{\boldsymbol{\theta}}) + C = 0.$$

Show that (up to a constant not depending on  $\theta$ ) this is proportional to the log-likelihood from the model  $\mathbf{y} \sim N(\theta, \mathbf{V})$ ; find  $\mathbf{y}$  and  $\mathbf{V}$ .

2. Observed and Fisher information in the exponential family. Suppose that  $\mathbf{x}$  follows a distribution in the exponential family. Show that for all  $\boldsymbol{\theta}$ ,

$$\mathcal{J}(\boldsymbol{\theta}) = \mathcal{I}(\boldsymbol{\theta})$$

where  $\boldsymbol{\theta}$  is the natural parameter.

3. Quadratic approximation for the Gamma distribution. The pdf for the Gamma distribution with shape parameter  $\alpha$  and rate parameter  $\beta$  is given by

$$f(x|\alpha,\beta) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}.$$

Generate a random sample of size 50 from this distribution using the following code:

To do this problem, please take note of the following information. The derivative of the log of the gamma function is known as the digamma function, usually denoted  $\psi_0(\alpha)$ :

$$\psi_0(\alpha) = \frac{d}{d\alpha} \log \Gamma(\alpha).$$

The second derivative of the log of the gamma function is known as the *trigamma function*, and usually denoted  $\psi_1(\alpha)$ :

$$\psi_1(\alpha) = \frac{d^2}{d\alpha^2} \log \Gamma(\alpha).$$

These functions are available in R and called digamma() and trigamma(), respectively.

- (a) Derive the score vector for  $(\alpha, \beta)$ .
- (b) Solve for the MLE. This is only available in partially-closed form. To find the MLE, solve for  $\hat{\beta}$  in terms of  $\hat{\alpha}$  and use this to derive a one-dimensional score function for  $\alpha$ , then use the uniroot() function to obtain a numerical answer for  $\hat{\alpha}$ .
- (c) Derive the information matrix.
- (d) Plot the two-dimensional 95% confidence region based on the quadratic/normal approximation at the MLE. On the plot, label the true value of  $(\alpha, \beta)$ . You can do this however you'd like, but my suggestion is to use the ellipse package.
- (e) Overlay three gamma densities: the true gamma density for this problem, some other gamma density for parameter values inside the confidence interval, and a gamma density for a value outside the confidence interval. Comment briefly on the results.