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

## Assignment 6

Due: Monday, October 19

1. Completing the square.
(a) Show that if $\mathbf{A}$ is a symmetric, positive definite matrix, then any equation of the form

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

can be rewritten as

$$
(\mathbf{x}-\mathbf{m})^{\top} \mathbf{A}(\mathbf{x}-\mathbf{m})=D
$$

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 $\boldsymbol{\theta}$ ) this is proportional to the log-likelihood from the model $\mathbf{y} \sim \mathrm{N}(\boldsymbol{\theta}, \mathbf{V})$; 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 $\boldsymbol{\theta}$,

$$
\mathscr{I}(\theta)=\mathcal{I}(\theta)
$$

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

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

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

```
set.seed(1)
x <- rgamma(50, shape=2, rate = 1)
```

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 $\widehat{\beta}$ 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.

