

Homework 3. (due March 25)

- (1) Suppose Y_1, \dots, Y_n are i.i.d. with density

$$f_{Y_i}(y; \mu) = \frac{1}{\mu} \exp\left(-\frac{y}{\mu}\right), y > 0, \mu > 0.$$

Show that the leading term in the saddlepoint approximation to the density of $\bar{Y} = \hat{\mu}$ reproduces the gamma density, with $\Gamma(n)$ replaced by Stirling's approximation to it. Deduce that the renormalized saddlepoint approximation is exact.

- (2) (Calibration) Suppose we have n observations y_1, \dots, y_n from a linear regression with known x 's:

$$Y_i = \alpha + \beta x_i + \sigma e_i$$

where e_1, \dots, e_n are i.i.d. $N(0, 1)$. We will take a further observation y_0 , say, from the model

$$Y_0 = \alpha + \beta \psi + \sigma e_0$$

where e_0 is also $N(0, 1)$ and independent of e_1, \dots, e_n . The parameter of interest is ψ , the value of x corresponding to the new observation y_0 . Derive the likelihood ratio statistic for testing ψ , and show that it is a function of a suitably defined t -statistic.

- (3) (Severini, Ch.4). Let Y_1, \dots, Y_n be independent and identically distributed from a density $f(y | \theta)$, $\theta \in \mathbb{R}$, and let $\pi(\theta)$ be a probability density function for θ . Define the estimator $\tilde{\theta}$ by

$$\tilde{\theta}(y) = \frac{\int \theta \pi(\theta | y) d\theta}{\int \pi(\theta | y) d\theta}.$$

Using the Laplace approximation for integrals, show that $\tilde{\theta} - \hat{\theta} = O_p(1/n)$, where $\hat{\theta} = \hat{\theta}(y)$ is the maximum likelihood estimator of θ .

- (4) (Wasserman, Ch.11; Berger & Wolpert, 1984). Let X_1, X_2 be independent Bernoulli random variables with $\Pr(X_i = 1) = \Pr(X_i = -1) = 1/2$, and define $Y_i = X_i + \theta$, $-\infty < \theta < \infty$, and suppose Y_1 and Y_2 are observed.

- (a) Show that

$$C = \begin{cases} \{Y_1 - 1\} & \text{if } Y_1 = Y_2, \\ \{(Y_1 + Y_2)/2\} & \text{if } Y_1 \neq Y_2, \end{cases}$$

is a confidence set for θ with confidence 0.75. Suppose we observe $y_1 = 15$ and $y_2 = 17$. What interpretation should we give to the statement that C is a 0.75-confidence set, since we are certain $\theta = 16$?

- (b) Suppose θ is an integer, and let $\pi(\theta)$ be a probability mass function that is positive for every θ . Compute the posterior distribution for θ when $y_1 = 15$ and $y_2 = 17$, and contrast the inference statement with the confidence statement.
- (5) (A Neyman-Scott problem). A class of problems where maximum likelihood estimators are not consistent are those in which the number of nuisance parameters increases with the sample size. These are often called Neyman-Scott problems.¹ For example, if $Y_{ij}, j = 1, \dots, m_i; i = 1, \dots, n$ follow a $N(\mu_i, \sigma^2)$ distribution, the maximum likelihood estimator of σ^2 is inconsistent as $n \rightarrow \infty$; in particular if $m_i \equiv 2$, then $\hat{\sigma}^2 \rightarrow \sigma^2/2$; see CH Example 9.24.
- (a) Suppose that Y_{i1} and Y_{i2} are independent observations from exponential distributions with means $\psi\lambda_i$ and ψ/λ_i , respectively, $i = 1, \dots, n$. Show that the maximum likelihood estimator of ψ is not consistent, but converges in probability to $(\pi/4)\psi$.
- (b) A modification to the profile likelihood to account for estimation of nuisance parameters was proposed in Cox & Reid (1987, JRSS B):

$$\ell_m(\psi) = \ell(\psi, \hat{\lambda}_\psi) - \frac{1}{2} \log |j_{\lambda\lambda}(\psi, \hat{\lambda}_\psi)|,$$

where $\lambda = (\lambda_1, \dots, \lambda_n)$ and $\hat{\lambda}_\psi$ is the constrained maximum likelihood estimator of λ . This is to be computed using a parametrization of the nuisance parameter that is *orthogonal* to the parameter of interest ψ , with respect to expected Fisher information. (The correction term $\frac{1}{2} \log |j_{\lambda\lambda}(\psi, \hat{\lambda}_\psi)|$ is not invariant to reparameterizations.) Show for the exponential case that λ is orthogonal to ψ , and that the value of ψ that solves $\ell'_m(\psi) = 0$, $\hat{\psi}_m$, say, converges to $(\pi/3)\psi$.

¹Neyman, J. and Scott, E.L. (1948). Consistent estimates based on partially consistent observations. *Econometrica* **16**, 1–32.