

## Homework#3, Statistical Inference I, 2011 Fall

### 1. Skewness parameter

The gamma distribution has a density

$$f(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta} I(x > 0)$$

w.r.t. a Lebesgue measure. We observe iid samples  $(X_1, \dots, X_n)$  having the density  $f$ .

1) Find the cumulant generating function  $\kappa(t) = \log E[\exp(tX_1)]$

2) Let  $X$  be an arbitrary random variable having the m.g.f. in the neighborhood of zero. Show that  $\kappa'(0) = E(X)$ ,  $\kappa''(0) = E(X - E[X])^2$ ,  $\kappa'''(0) = E(X - E[X])^3$ .

(you can use properties of the m.g.f.)

3) For the Gamma distribution, find the skewness parameter  $\theta = \frac{\mu_3}{(\mu_2)^{3/2}}$  where

$\mu_n = E(X - E[X])^n$  is the central moment.

4) Find a consistent estimator of  $\theta$ , which is an explicit function of  $\bar{X}$  and

$S^2 = 1/(n-1) \sum_{i=1}^n (X_i - \bar{X})^2$  (answer if not unique). Prove the consistency.

### 2. From Exercise 63

Let  $(X_1, \dots, X_n)$  be iid from  $N(\mu, \sigma^2)$ . Consider the estimation of  $\sigma^2$  with the squared error loss.

(a) Show that  $(n-1)S^2/n$  is better than  $S^2 = 1/(n-1) \sum_{i=1}^n (X_i - \bar{X})^2$ .

(b) Can you find the best estimator of the form  $cS^2$ , where  $c > 0$  is a constant?

### 3. Give a detailed proof of Theorem 3.2 (i)

**Answer 1,**

$$1) M(t) = \frac{1}{(1-\beta t)^\alpha}, \quad \kappa(t) = -\alpha \log(1-\beta t).$$

$$3) \kappa'(t) = \frac{\alpha\beta}{1-\beta t}, \quad \kappa''(t) = \frac{\alpha\beta^2}{(1-\beta t)^2}, \quad \kappa'''(t) = \frac{2\alpha\beta^3}{(1-\beta t)^3},$$

$$\mu_2 = \alpha\beta^2, \quad \mu_3 = 2\alpha\beta^3, \quad \alpha_3 = \frac{\mu_3}{(\mu_2)^{3/2}} = \frac{2}{\sqrt{\alpha}}.$$

4): A consistent estimator of  $2/\sqrt{\alpha}$  is  $2S/\bar{X}$  (answer is not unique).

$\therefore$  Since  $E(X) = \alpha\beta$  and  $Var(X) = \alpha\beta^2$ , it follows that

$$\frac{2}{\sqrt{\alpha}} = \frac{\sqrt{Var(X)}}{E(X)}.$$

Since,  $\bar{X} \xrightarrow{P} E(X)$  and  $S^2 = 1/(n-1) \sum_{i=1}^n (X_i - \bar{X})^2 \xrightarrow{P} Var(X)$ , by Slutsky's

theorem,  $\frac{2\sqrt{S^2}}{\bar{X}} \xrightarrow{P} \frac{2}{\sqrt{\alpha}}$ .

Derivation based on the moment equations:

Since  $E(X) = \alpha\beta$ , and  $Var(X) = \alpha\beta^2$ , we set  $\bar{X} = \alpha\beta$  and  $S^2 = \alpha\beta^2$ . Hence,

$\hat{\alpha} = \bar{X}^2 / S^2 \rightarrow_p \alpha$  by Slutsky's theorem. Again, by Slutsky's theorem,

$$2S/\bar{X} = 2/\sqrt{\hat{\alpha}} \rightarrow_p 2/\sqrt{\alpha}.$$