

Hints on 18.466 PS 5 due Friday, March 14, 2003

1 = 2.4-1 Clarification:  $0 < c < \infty$ . If  $X$  has density  $f_C$ ,  $\bar{E}X = \int_0^\infty xf_C(x)dx$  and  $\bar{E}(X^2) = \int_0^\infty x^2f_C(x)dx$  are known or can be found via integrating by parts. We have  $\bar{E}(\bar{X}) = \bar{E}X_1$  and  $\text{Var}(\bar{X}) = \text{Var}(X_1)/n$  for any  $X_1, \dots, X_n$  i.i.d. with finite variance.

2 = 2.4-3.  $\int_0^1 x^{a-1}(1-x)^{b-1}dx = B(a, b) = \Gamma(a)\Gamma(b)/\Gamma(a+b)$  for any  $a, b > 0$  where  $\Gamma(k) = (k-1)!$  for  $k=1, 2, \dots$ . Thus find  $\bar{E}X_{(1)}$  and  $\text{Var}(X_{(1)})$  for  $U[0, 1]$  and for  $U[\theta, \theta+b]$ . So find an unbiased estimator of  $\theta$  and its variance.

3 = 2.4-4 (a) For the binomial( $n, p$ ) distribution the mean is  $np$  and the variance is  $np(1-p)$ .  $f(A_j) = p^j(1-p)^{n-j}$ . Put the score function  $\frac{\partial}{\partial p} \log f(A_j)$  over a common denominator.

(b)  $X$  with this geometric distribution has  $\bar{E}X = 1/p$  and  $\text{Var}(X) = 1/p^2$ .

4 = 2.5-2. Use (2.5.1) in the extended form stated just after it,

$$f(t, x) = C(\theta(t)) h(x) e^{\sum_{j=1}^k \theta_j(t) T_j(x)}$$

$k$  can be reduced from 2 to 1 if  $\theta_1(p)$  and  $\theta_2(p)$  are affinely dependent, or if  $T_1(k)$  and  $T_2(k)$  are.