## Math 362, Problem set 5

## Due 3/16/10

1. (3.7.6) Another estimating chi-square: Let the result of a random experiment be classifies as one of the mutually exclusive and exhaustive ways  $A_1, A_2, A_3$  and also as one of the mutually exclusive and exhaustive ways  $B_1, B_2, B_3, B_4$ . Two hundred independent trials of the experiment result in the following data

	$B_1$	$B_2$	$B_3$	$B_4$
$A_1$	10	21	15	6
$A_2$	11	27	21	13
$A_3$	6	19	27	24

Test, at the 0.05 significance level the hypothesis of independence of the A and B attribute, namely  $H_0 : \mathbb{P}(A_i \cap B_i) = \mathbb{P}(A_i)\mathbb{P}(B_i)$ , i = 1, 2, 3 and j = 1, 2, 3, 4 against the alternative of dependence.

- 2. (5.8.5) Determine a method to generate random observations for the following pdf:  $f(x) = 4x^3$  for 0 < x < 1, zero elsewhere.
- 3. (5.8.18) For  $\alpha>0$  and  $\beta>0,$  consider the following accept/reject algorithm:
  - (1) Generate  $U_1$  and  $U_2$  iid uniform(0,1) random variables. Set  $V_1 = U_1^{1/\alpha}$  and  $V_2 = U_2^{1/\beta}$ .
  - (2) Set  $W = V_1 + V_2$ . If  $W \le 1$ , set  $X = V_1/W$ , else goto Step(1).
  - (3) Deliver X.

Show that X has a beta distribution with parameters  $\alpha$  and  $\beta$ .

*Note/Hint:* The analysis is quite similar to the analysis of the algorithm we did in class. That is, look for the cdf  $\mathbb{P}(X \leq x)$  and note that it is some conditional probability.

4. (5.9.1) Let  $x_1, \ldots, x_n$  be the values of a random sample. A bootstrap sample  $\mathbf{x}^{*'} = (x_1^*, \ldots, x_n^*)$  is a random sample of  $x_1, \ldots, x_n$  drawn with replacement.

- (a) Show that  $\mathbb{E}[x_i^*] = \bar{x}$
- (b) If n is odd, show that median  $\{x_i^*\} = x_{((n+1)/2)}$
- (c) Show that  $\operatorname{Var}(x_i^*) = \frac{1}{n} \sum_{i=1}^n (x_i \bar{x})^2$

*Notes:* There was a part (a) on the book problem, which I cut as it's obvious but hard to write up nicely. It asks you to show that the  $x_i^*$  are independent (which they are because I chose with replacement) and have the CDF  $\hat{F}_n$  which is  $\hat{F}_n(x) = \frac{1}{n} \times (\# \text{ of } x_i < x)$ , which is obvious because I select the  $x_i^*$  uniformly from  $x_1, \ldots, x_n$ . Note that in (c), you have  $\frac{1}{n}$  and not  $\frac{1}{n-1}$  because  $\bar{x}$  is the *real* mean of the  $x_i^*$  as opposed to the sample mean of the  $x_i^*$ .

- 5. (5.9.2) Let  $X_1, \ldots, X_n$  be a random sample from a  $\Gamma(1, \beta)$  distribution.
  - (a.) Show that the confidence interval  $(2n\bar{X}/(\chi_{2n}^2)^{(1-(\alpha/2))}, 2n\bar{X}/(\chi_{2n}^2)^{\alpha/2})$ is an exact  $(1-\alpha)100\%$  confidence interval for  $\beta$ . Notation note:  $(\chi_{2n}^2)^{(1-(\alpha/2))}$  is the books (rather confusing) way of saying 'the number such that  $\mathbb{P}(\chi^2(2n) < (\chi_{2n}^2)^{(1-(\alpha/2))}) = 1 - \alpha/2$ ; for instance for  $\alpha = .5$  and n = 4, so that 2n = 8, we have (by our chart)  $(\chi_8^2)^{(1-(\alpha/2))} = 17.535$ . Unfortunately the notation makes the problem look harder than it is.
  - (b.) Using part (a), show that the 90% confidence interval discussed in Example 5.9.1 is (64.99, 136.69).
- 6. (5.9.10) Let  $z^*$  be drawn at random from the discrete distribution which has mass  $n^{-1}$  at each point  $z_i = x_i - \bar{x} + \mu_0$ , where  $(x_1, \ldots, x_n)$  is the realization of a random sample. Determine  $\mathbb{E}[z^*]$  and  $\operatorname{Var}(z^*)$ .