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## **Solutions to Sample Midterm Questions**

1. Parameter fitting.

(a) Density and likelihood function

(i) The density is 
$$f_x(x) = \frac{d}{dx} F_x(x) = \frac{a}{b} \left(\frac{x}{b}\right)^{-(a+1)} = ab^a x^{-(a+1)}$$
 for  $x \ge b$  and

$$f_X(x) = 0$$
 for  $x < b$ 

(ii) The likelihood function is

$$L_n(a,b;X_1,...,X_n) = \prod_{i=1}^n f_X(X_i)$$
  
=  $\prod_{i=1}^n I(b \le X_i) \times a^n b^{na} \prod_{i=1}^n X_i^{-(a+1)}$ 

where I is an indicator function.

(b) Since  $L_n$  is initially increasing in b, choose b as large as possible:  $\hat{b} = X_{(1)} = \min_{1 \le i \le n} X_i$ .

To choose  $\hat{a}$ , maximize the function  $h(a) = a^n \hat{b}^{na} \prod_{i=1}^n X_i^{-(a+1)}$ . Equivalently, maximize  $\tilde{h}(a) = \log h(a) = n \log a + na \log \hat{b} - (a+1) \sum_{i=1}^n \log X_i$ .  $\frac{d}{da} \tilde{h}(a) = \frac{n}{a} + n \log \hat{b} - \sum_{i=1}^n \log X_i = 0$  implies  $\hat{a} = \left[\frac{1}{n} \sum_{i=1}^n \log\left(\frac{X_i}{X_{(1)}}\right)\right]^{-1}$ .

(c) Given  $\hat{b}$  we have

$$E[X] = \int_{b}^{\infty} x f_{X}(x) dx = \int_{b}^{\infty} a\hat{b}^{a} x^{-a} dx = \frac{a\hat{b}^{a} x^{1-a}}{1-a} \bigg|_{b}^{\infty} = \frac{a\hat{b}}{a-1} .$$

Substitute the sample mean  $\overline{X}_n$  for E[X] and solve the equation  $\overline{X}_n = \frac{ab}{a-1}$  to get  $\hat{a} = \frac{X_n}{\overline{X}_n - \hat{b}}$ .

- 2. Random variate generation
  - (a) The picture is as follows.



Using the basic version of the acceptance-rejection method, we obtain the following algorithm:

<u>Algorithm:</u>	
1.	Generate uniform random numbers $U_1, U_2$
2.	Set $Z_1 = bU_1$ and $Z_2 = U_2 / (1 - e^{-b})$ .
3.	If $Z_2 \le \frac{e^{-Z_1}}{1 - e^{-b}}$ , then return $Z_1$ , else go to Step 1.

(b) A much more efficient inversion algorithm is as follows, based on the fact that  $F_{\chi}(x) = \frac{1 - e^{-x}}{1 - e^{-b}}$ :

Algorithm:

- 1. Generate a uniform random number U
- 2. Return  $X = -\ln(1 (1 e^{-b})U)$
- 3. Linear congruential generators
  - a) Starting with  $x_0 = 1$ , we have the following results

i	$x_{i}$	$3^i \mod 7$
1	3	3
2	2	2
3	6	6
4	4	4
5	5	5
6	1	1

By inspection of the second column, we see that all values between 1 and 6 occur, so the generator has full period. Inspection of the third column shows that 3 is a primitive element modulo 7, which also proves that the generator has full cycle since 7 is prime.

- b) From Lecture 6, Slide 8, the RANDU generator is an MCG with modulus  $2^{31}$ . By the Claim on Slide 10, the period is at most  $2^{k-2} = 2^{6-2} = 2^4 = 16$ .
- 4. Bayesian estimation. Write  $X_{(n)} = \max_{1 \le i \le n} X_i$ , and note that  $X_i \ge 0$ ,  $\forall i$ . Then the likelihood is given by  $f(X_{1:n} | a) = \prod_{i=1}^{n} [I(X_i \le a) / a] = I(X_{(n)} \le a) / a^n$ , where I(A) denotes the indicator of event Aand  $X_{1:n} = X_1, \dots, X_n$ . The prior is  $f(a) = I(0 \le a \le b) / b$ . By Bayes' Theorem, the posterior density of a is

$$f(a \mid X_{1:n}) = \frac{f(X_{1:n} \mid a)f(a)}{\int f(X_{1:n} \mid a)f(a)da} = \frac{I(X_{(n)} \le a)I(0 \le a \le b)a^{-n}b^{-1}}{\int I(X_{(n)} \le a)I(a \le b)a^{-n}b^{-1}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{\int I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le a \le b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le b \ge b)a^{-n}da}{I(X_{(n)} \le a \le b)a^{-n}da} = \frac{I(X_{(n)} \le b \ge b)a^{-n}da}{I(X_{(n)} \le b \ge b)a^{-n}da} = \frac{I(X_{(n)} \le b \ge b)a^{-n}da}{I(X_{(n)} \le b \ge b)a^{-n}da} = \frac{I(X_{(n)} \le b \ge b)a^{-n}da}{I(X_{(n)} \le b \ge b)a^{-n}da} = \frac{I(X_{(n)} \le b \ge b)a^{-n}da}{I(X_{(n)} \le b \ge b)a^{-n}da} = \frac{I(X_{(n)} \le b \ge b)a^{-n}da}{I(X_{(n)} \le b \ge b)a^{-n}da} = \frac{I(X_{(n)} \ge b)a^{-n}da}{I(X_{(n)} \ge b)a^{-n}da}$$

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Thus the posterior mean is 
$$\hat{a} = \int_{a}^{a} af(a \mid X_{1:n}) da = \frac{\int_{X_{(n)}}^{b} a^{1-n} da}{(X_{(n)}^{1-n} - b^{1-n})/(n-1)} = \frac{(X_{(n)}^{2-n} - b^{2-n})/(n-2)}{(X_{(n)}^{1-n} - b^{1-n})/(n-1)}.$$