

STATISTICS DEPARTMENT
M.S. EXAMINATION

PART I
CLOSED BOOK

Friday, May 25, 2001

9:00 a.m. - 1:00 p.m.

School of Education Conference Room, AE 123

Instructions: Complete *all five* problems. Each problem counts 20 points. Unless otherwise noted, points are allocated approximately equally to lettered parts of a problem. Spend your time accordingly.

Begin each problem on a new page. Write the problem number and the page number in the specified locations at the top of each page. Also write your chosen ID code number on every page. Please write only within the black borderlines, leaving at least 1" margins on both sides, top and bottom of each page. Write on one side of the page only.

At the end of this part of the exam you will turn in your answers sheets, but you will keep the question sheets and your scratch paper.

1. (a) Suppose that we have 2 different groups of patients with a certain disease, 101 patients in each group. In one group, they take diagnostic test #1 and 95 patients test positive for the disease; in the 2nd group, they take diagnostic test #2 and 88 test positive for the disease. Assume that the total of 202 patients are randomly chosen from some population and are randomly assigned to the two groups. Can one conclude that the probability of a testing positive differs between the two diagnostic tests? Show your work.
 - (b) Suppose in another scenario that 101 patients randomly chosen from some population took both diagnostic tests. 82 tested positive for both tests, 13 tested positive for diagnostic test #1 but not for #2, a (+, -) result, and 6 tested positive for #2 but not for #1, a (-, +) result.
 - (i) Suppose we focus our attention on only those 19 patients where there was a lack of agreement between the two tests. Can one conclude that the probability of a (+, -) result (versus a (-, +) result) is not equal to .5? Show your work.
 - (ii) Explain why a "YES" answer to (i) would allow one to establish that it is not the case that the probability of a positive result is the same for both diagnostic tests.
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2. A professional gambler expects to have a 'winning day' 4 out of 5 days when he plays a certain gambling game. He wins \$5000 on his winning days; on his losing days, he loses \$10,000. Assume independence between different plays of the game.
 - (a) What is the probability that his 5th loss occurs on the 15th play of the game?
 - (b) Assuming that his 5th loss occurs on the 15th play, how much will his gain be after 15 plays?
 - (c) Given that the gambler has just suffered his 5th loss, how many total games would we have expected him to play?
 - (d) Given that the gambler has just suffered his 5th loss, what is his expected gain?
 - (e) Suppose X = number of plays that it takes to achieve the gambler's w^{th} loss. Explain why for w , a large integer, X is approximately normally distributed.

3. Sam installs TV cables. He is about to install a cable in an apartment and a cable in a house. Let X be the time that he requires to install a cable in the apartment and let Y be the time that he requires to install a cable in the house. Suppose that $0 < a < b < \infty$ and that X has a uniform distribution on the interval (a, b) . Suppose that $0 < \tau < \infty$ and that Y has the probability density function f_Y defined by

$$f_Y(y) = \begin{cases} 2\tau^2 y^{-3} & \text{for } \tau < y < \infty \\ 0 & \text{elsewhere.} \end{cases}$$

5 (a) Find $P(Y > b)$.

5 (b) Find $E(Y - X)$.

(c) Suppose that X and Y are independent and suppose that $0 < \tau < a$.

5 (i) Find $E(\frac{1}{X})$.

5 (ii) Let $U = \frac{Y}{X}$. Find the probability density function of U .

4. Let X_1, X_2, \dots, X_n be independent and identically distributed two-parameter exponential random variables with p.d.f.

$$f(x|\theta, \eta) = \frac{1}{\theta} e^{-(x-\eta)/\theta} \quad (1)$$

where $x > \eta$ and $\theta > 0$.

- Calculate the method of moments estimators of θ and η . (Hint: Recall the exponential and gamma densities when computing $E[X]$. Also recall that a location parameter does not change the variance of a random variable. Therefore, the variance of the two-parameter exponential is the same as the one-parameter exponential.)
- Calculate the maximum likelihood estimator, $\hat{\theta}$, of θ .
- Calculate the maximum likelihood estimator, $\hat{\eta}$, of η .
- Find the sufficient statistics for θ and η .
- Explain why the maximum likelihood estimators are functions of the sufficient statistics.

For the rest of the problem, assume that $\eta = 0$.

- Compute the Cramer-Rao Lower Bound for θ . Show that the MLE of θ , $\hat{\theta}$, achieves the CRLB.
- Give an asymptotic $100(1 - \alpha)\%$ confidence interval for θ .
- Suppose the sample size is $n = 5$. Explain why the confidence interval given above may not be accurate in terms of its coverage probability. As an alternative method of calculating a confidence interval for θ , describe how you would implement the parametric bootstrap to produce the bootstrap estimate of θ , $\tilde{\theta}^*$, a bootstrap estimate of the standard error, $s_{\tilde{\theta}}$, and a 95% Bootstrap confidence interval for θ .

5. Suppose that n is a positive integer and that $0 < p < 1$. Let N be a binomial random variable with parameters n and p . For each positive integer i , let X_i be a random variable whose probability density function is

$$f_X(x) = \begin{cases} \frac{1}{2}x^2 \exp(-x) & \text{if } x > 0 \\ 0 & \text{elsewhere.} \end{cases}$$

Suppose that N, X_1, X_2, \dots are independent. Let

$$Y_k = \begin{cases} 1 & \text{if } k = 0 \\ \prod_{i=1}^k X_i & \text{if } k = 1, 2, \dots \end{cases}$$

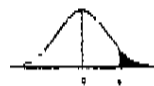
7 (a) Find $P(Y_1 > 2)$.

7 (b) Find $E(Y_n)$ in terms of $\mu = E(X_1)$, n , and p . You need not determine the value of μ .

6 (c) For $k = 1, 2, \dots$ let $Z_k = \frac{Y_k}{E(Y_k)}$. Show that $Z_k, k = 1, 2, \dots$ is a Martingale.

CUMULATIVE DISTRIBUTION OF CHI-SQUARE*

Degree of Freedom	Probability of a Greater Value												
	0.995	0.990	0.975	0.950	0.900	0.750	0.500	0.250	0.100	0.050	0.025	0.010	0.005
1					0.02	0.10	0.45	1.32	2.71	3.84	5.02	6.63	7.88
2	0.01	0.02	0.05	0.10	0.21	0.58	1.39	2.77	4.61	5.99	7.38	9.21	10.60
3	0.07	0.11	0.22	0.35	0.58	1.21	2.37	4.11	6.25	7.81	9.35	11.34	12.84
4	0.21	0.30	0.48	0.71	1.06	1.92	3.36	5.39	7.78	9.49	11.14	13.28	14.86
5	0.41	0.55	0.83	1.15	1.61	2.67	4.35	6.63	9.24	11.07	12.83	15.09	16.75
6	0.68	0.87	1.24	1.64	2.20	3.45	5.35	7.84	10.64	12.59	14.45	16.81	18.55
7	0.99	1.24	1.69	2.17	2.83	4.25	6.35	9.04	12.02	14.07	16.01	18.48	20.28
8	1.34	1.65	2.18	2.73	3.49	5.07	7.34	10.22	13.36	15.51	17.53	20.09	21.96
9	1.73	2.09	2.70	3.33	4.17	5.90	8.34	11.39	14.68	16.92	19.02	21.67	23.59
10	2.16	2.56	3.25	3.94	4.87	6.74	9.34	12.55	15.99	18.31	20.48	23.21	25.19
11	2.60	3.05	3.82	4.57	5.58	7.58	10.34	13.70	17.28	19.68	21.92	24.72	26.76
12	3.07	3.57	4.40	5.23	6.30	8.44	11.34	14.85	18.55	21.03	23.34	26.22	28.30
13	3.57	4.11	5.01	5.89	7.04	9.30	12.34	15.98	19.81	22.36	24.74	27.69	29.82
14	4.07	4.66	5.63	6.57	7.79	10.17	13.34	17.12	21.06	23.68	26.12	29.14	31.32
15	4.60	5.23	6.27	7.26	8.55	11.04	14.34	18.25	22.31	25.00	27.49	30.58	32.80
16	5.14	5.81	6.91	7.96	9.31	11.91	15.34	19.37	23.54	26.30	28.85	32.00	34.27
17	5.70	6.41	7.56	8.67	10.09	12.79	16.34	20.49	24.77	27.59	30.19	33.41	35.72
18	6.26	7.01	8.23	9.39	10.86	13.68	17.34	21.60	25.99	28.87	31.53	34.81	37.16
19	6.84	7.63	8.91	10.12	11.65	14.56	18.34	22.72	27.20	30.14	32.85	36.19	38.58
20	7.43	8.26	9.59	10.85	12.44	15.45	19.34	23.83	28.41	31.41	34.17	37.57	40.00
21	8.03	8.90	10.28	11.59	13.24	16.34	20.34	24.93	29.62	32.67	35.48	38.93	41.40
22	8.64	9.54	10.98	12.34	14.04	17.24	21.34	26.04	30.81	33.92	36.78	40.29	42.80
23	9.26	10.20	11.69	13.09	14.85	18.14	22.34	27.14	32.01	35.17	38.08	41.64	44.18
24	9.89	10.86	12.40	13.85	15.66	19.04	23.34	28.24	33.20	36.42	39.36	42.98	45.56
25	10.52	11.52	13.12	14.61	16.47	19.94	24.34	29.34	34.38	37.65	40.65	44.31	46.93
26	11.16	12.20	13.84	15.38	17.29	20.84	25.34	30.43	35.56	38.89	41.92	45.64	48.29
27	11.81	12.88	14.57	16.15	18.11	21.75	26.34	31.53	36.74	40.11	43.19	46.96	49.64
28	12.46	13.56	15.31	16.93	18.94	22.66	27.34	32.62	37.92	41.34	44.46	48.28	50.99
29	13.12	14.26	16.05	17.71	19.77	23.57	28.34	33.71	39.09	42.56	45.72	49.59	52.34
30	13.79	14.95	16.79	18.49	20.60	24.48	29.34	34.80	40.26	43.77	46.98	50.89	53.67
40	20.71	22.16	24.43	26.51	29.05	33.66	39.34	45.62	51.80	55.76	59.34	63.69	66.77
50	27.99	29.71	32.36	34.76	37.69	42.94	49.33	56.33	63.17	67.50	71.42	76.15	79.49
60	35.53	37.48	40.48	43.19	46.46	52.29	59.33	66.98	74.40	79.08	83.30	88.38	91.95
70	43.28	45.44	48.76	51.74	55.33	61.70	69.33	77.58	85.53	90.53	95.02	100.42	104.22
80	51.17	53.54	57.15	60.39	64.28	71.14	79.33	88.13	96.58	101.88	106.63	112.33	116.32
90	59.20	61.75	65.65	69.13	73.29	80.62	89.33	98.64	107.56	113.14	118.14	124.12	128.30
100	67.33	70.06	74.22	77.93	82.36	90.13	99.33	109.14	118.50	124.34	129.56	135.81	140.17



Areas for the Normal Curve

	.02	.03	.04	.05	.06	.07	.08	.09
920	.4880	.4840	.4801	.4761	.4721	.4681	.4641	
522	.4483	.4443	.4404	.4364	.4325	.4286	.4247	
129	.4090	.4052	.4013	.3974	.3936	.3897	.3859	
745	.3707	.3669	.3632	.3594	.3557	.3520	.3483	
372	.3336	.3300	.3264	.3228	.3192	.3156	.3121	
015	.2981	.2946	.2912	.2877	.2843	.2810	.2776	
676	.2643	.2611	.2578	.2546	.2514	.2483	.2451	
358	.2327	.2296	.2266	.2236	.2206	.2177	.2148	
061	.2033	.2005	.1977	.1949	.1922	.1894	.1867	
788	.1762	.1736	.1711	.1685	.1660	.1635	.1611	
539	.1515	.1492	.1469	.1446	.1423	.1401	.1379	
314	.1292	.1271	.1251	.1230	.1210	.1190	.1170	
112	.1093	.1075	.1056	.1038	.1020	.1003	.0985	
934	.0918	.0901	.0885	.0869	.0853	.0838	.0823	
778	.0764	.0749	.0735	.0721	.0708	.0694	.0681	
643	.0630	.0618	.0606	.0594	.0582	.0571	.0559	
526	.0516	.0505	.0495	.0485	.0475	.0465	.0455	
427	.0418	.0409	.0401	.0392	.0384	.0375	.0367	
3	.0336	.0329	.0322	.0314	.0307	.0301	.0294	
227	.0268	.0262	.0256	.0250	.0244	.0239	.0233	
217	.0212	.0207	.0202	.0197	.0192	.0188	.0183	
170	.0166	.0162	.0158	.0154	.0150	.0146	.0143	
132	.0129	.0125	.0122	.0119	.0116	.0113	.0110	
102	.0099	.0096	.0094	.0091	.0089	.0087	.0084	
078	.0075	.0073	.0071	.0069	.0068	.0066	.0064	
059	.0057	.0055	.0054	.0052	.0051	.0049	.0048	
044	.0043	.0041	.0040	.0039	.0038	.0037	.0036	
033	.0032	.0031	.0030	.0029	.0028	.0027	.0026	
024	.0023	.0023	.0022	.0021	.0021	.0020	.0019	
018	.0017	.0016	.0016	.0015	.0015	.0014	.0014	
013	.0012	.0012	.0011	.0011	.0011	.0010	.0010	

Percentage Points of the t Distribution

df	$\alpha = .1$	$\alpha = .05$	$\alpha = .025$	$\alpha = .01$	$\alpha = .005$
1	3.078	6.314	12.706	31.821	63
2	1.886	2.920	4.303	6.965	9
3	1.638	2.353	3.182	4.541	5
4	1.533	2.132	2.776	3.747	4
5	1.476	2.015	2.571	3.365	4
6	1.440	1.943	2.447	3.143	3
7	1.415	1.895	2.365	2.998	3
8	1.397	1.860	2.306	2.896	3
9	1.383	1.833	2.262	2.821	3
10	1.372	1.812	2.228	2.764	3
11	1.363	1.796	2.201	2.718	3
12	1.356	1.782	2.179	2.681	3
13	1.350	1.771	2.160	2.650	3
14	1.345	1.761	2.145	2.624	2
15	1.341	1.753	2.131	2.602	2
16	1.337	1.746	2.120	2.583	2
17	1.333	1.740	2.110	2.567	2
18	1.330	1.734	2.101	2.552	2
19	1.328	1.729	2.093	2.539	2
20	1.325	1.725	2.086	2.528	2
21	1.323	1.721	2.080	2.518	2
22	1.321	1.717	2.074	2.508	2
23	1.319	1.714	2.069	2.500	2
24	1.318	1.711	2.064	2.492	2
25	1.316	1.708	2.060	2.485	2
26	1.315	1.706	2.056	2.479	2
27	1.314	1.703	2.052	2.473	2
28	1.313	1.701	2.048	2.467	2
29	1.311	1.699	2.045	2.462	2
30	1.310	1.697	2.042	2.457	2
40	1.303	1.684	2.021	2.423	2
60	1.296	1.671	2.000	2.390	2
120	1.289	1.658	1.980	2.358	2
240	1.285	1.651	1.970	2.342	2
	1.282	1.645	1.960	2.326	2

closed book #1

Answer

(a) Suppose that we have 2 different groups of patients with a certain disease, 101 patients in each group. In one group, they take diagnostic test #1 and 95 patients test positive for the disease and in the 2nd group, they take diagnostic test #2 and 88 test positive for the disease. Assume that the total of 202 patients are randomly chosen from some population and are randomly assigned to the two groups. Can one conclude that it is not the case that the probability of a positive result is the same for both diagnostic tests? Show your work.

(b) Suppose in another scenario that 101 patients randomly chosen from some population took both diagnostic tests. 82 of them tested positive for both tests, 13 of them tested positive for diagnostic test #1 but not for #2, a (+, -), result, and 6 of them tested positive for #2 but not for #1, a (-, +) result.

- (i) Supposing we focused our attention on only those 19 patients where there was a lack of agreement between the two tests. Can one conclude that the probability of a (+, -) result (versus a (-, +) result) is not equal to .5? Show your work.
 - (ii) Explain why if your answer to (i) is "YES", then you have established that it is not the case that the probability of a positive result is the same for both diagnostic tests.
-

>>> DA

! Datafile: C:\backup\E Drive\MSSOLUT\msp2s012nd.cy3

! Table 1 of 1

Table1	+	-	TOTAL
test #1	95 (91.5)	6 (9.5)	101
test #2	88 (91.5)	13 (9.5)	101
TOTAL	183	19	202

$$\begin{aligned} \text{Pearson's } \chi^2 &= \sum \frac{(O-E)^2}{E} \\ &= 2 \left(\frac{(3.5)^2}{91.5} + \frac{(3.5)^2}{9.5} \right) \\ &= 2.847 \end{aligned}$$

>>> RC CH/EX

Datafile: C:\backup\E Drive\MSSOLUT\msp2s012nd.cy3

PEARSON CHI-SQUARE TEST

Statistic based on the observed 2 by 2 table(x):

CH(X): Pearson Chi-Square Statistic = 2.847

P-value = 0.0916
 Thus at $\alpha = 0.05$
 one cannot conclude
 that P(+) is different
 2.091 with Yates CC)
 for the 2 tests.

Asymptotic p-value: (based on Chi-Square distribution with 1 df)

Two-sided : Pr { CH(X) .GE. 2.847 } = 0.0916

One-sided : 0.5 * Two-sided = 0.0458

Exact p-value and point probabilities:

Two-sided : Pr { CH(X) .GE. 2.847 } = 0.1466

Pr { CH(X) .EQ. 2.847 } = 0.0952

One-sided : Let y be the value in Row 1 and Column 1

y = 95 min(Y) = 82 max(Y) = 101 mean(Y) = 91.50 std(Y) = 2.080

Pr { Y .GE. 95 } = 0.0733

Pr { Y .EQ. 95 } = 0.0476

>>> DA

! Datafile: C:\backup\E Drive\MSSOLUT\msp2s01.cy3

! Table 1 of 1 diagnostic test 2

Table1	+	-	TOTAL
diagnostic test 1	82	13	95
-	6	0	6
TOTAL	88	13	101

>>> PS MC/EX

Datafile: C:\backup\E Drive\MSSOLUT\msp2s01.cy3

MCNEMAR'S TEST

Statistic based on the observed 2 by 2 table(x) :

Min	Max	Mean	Std-dev	Observed	Standardized
-19.00	19.00	0.0000	4.359	7.000	1.606

Asymptotic Inference:

One-sided p-value: Pr { Test Statistic .GE. Observed }	=	0.0541
Two-sided p-value: 2 * One-sided	=	<u>0.1083</u>

Exact Inference:

One-sided p-value: Pr { Test Statistic .GE. Observed }	=	0.0835
Pr { Test Statistic .EQ. Observed }	=	0.0518
Two-sided p-value: 2*One-Sided	=	<u>0.1671</u>

Asymptotic

Newton $\chi^2 = \frac{(13-6)^2}{13+6} = \frac{49}{19} = (1.606)^2$

p-value = 0.1083 (2-sided)

Cannot conclude at $\alpha = .05$

∴ We can also
use the
chi-square goodness
of fit test

Exact

$X = \# \text{ of } (+, -) = 13$

$n = 19$

$\therefore 2P(X \geq 13) = 2(0.0835) = 0.1671$
(from binomial table, $n=19, x=13$) ; Cannot conclude at $\alpha = .05$

ii) Suppose you conclude that $P(+, -) > P(-, +)$
then $P(+, \text{test 1}) = P(+, +) + P(+, -)$
 $> P(+, +) + P(-, +) = P(+, \text{test 2}).$

CBX
Open book #2

Answer

A professional gambler figures to have 'winning day' 4 out of 5 days that he plays a certain gambling game. He wins \$5000 on his winning days; otherwise, on his losing days, he loses \$10,000. Assume independence between different plays of the game.

- What is the probability that his 5th loss occurs on the 15th play of the game?
- Assuming that his 5th loss occurs on the 15th play, how much will be his gain after 15 plays?
- After the gambler suffers his 5th loss, how many total games would we expect to have been played by then?
- What is the gambler's expected gain after suffering his 5th loss?
- Suppose X = number of plays that it takes to achieve the gambler's w^{th} loss. Explain why for w , a large integer, X is approximately normally distributed.

(a) X = # of games until τ and including)
the 5th loss is hypergeometric.

$$P(X=15) = \binom{14}{4} (.8)^{10} (.2)^5$$

(b) $10(5000) - 5(10000) = 0$ (even, Steven)

(c) $\frac{5}{(\frac{1}{5})} = 25 = E(X)$

(d) $G = (X - 5) \cdot 5000 - 5(10,000)$

$$E(G) = (20) 5000 - 5(10,000) = 50,000$$

(e) Let Y_1 = # of plays thru 1st loss

$i \geq 1$; Y_{i+1} = # of plays after i^{th} loss thru $(i+1)^{\text{st}}$ loss,

Y_1, Y_2, \dots are iid geometrics

$X = Y_1 + Y_2 + \dots + Y_w$ is Normal

by the central limit theorem.

Openbook #3

3. Sam installs TV cables. He is about to install a cable in an apartment and a cable in a house. Let X be the time that he requires to install a cable in the apartment and let Y be the time that he requires to install a cable in the house. Suppose that $0 < a < b < \infty$ and that X has a uniform distribution on the interval (a, b) . Suppose that $0 < \tau < \infty$ and that Y has the probability density function f_Y defined by

$$f_Y(y) = \begin{cases} 2\tau^2 y^{-3} & \text{for } \tau < y < \infty \\ 0 & \text{elsewhere.} \end{cases}$$

5 # (a) Find $P(Y > b)$. $- P(Y > b) = \left(\int_b^{\infty} 2\tau^2 y^{-3} dy = -\frac{2\tau^2 y^{-2}}{2} \Big|_{y=b}^{y=\infty} = \frac{\tau^2}{b^2} \right) \text{ if } b \geq \tau$
 $\left\{ 1 \text{ if } b < \tau \right.$

5 # (b) Find $E(Y - X)$. $E(Y - X) = E(Y) - E(X)$. $E(X) = \frac{a+b}{2}$. $E(Y) = \int_{\tau}^{\infty} y \cdot 2\tau^2 y^{-3} dy = 2\tau^2 \left(-\frac{1}{2} y^{-2} \right) \Big|_{y=\tau}^{y=\infty} = \tau^2$

(c) Suppose that X and Y are independent and suppose that $0 < \tau < a$.

4 (i) Find $P(Y > X)$. $P(Y > X) = \int_a^b \frac{P(Y > x)}{b-a} dx = \int_a^b \frac{\tau^2}{x^2} dx$. Hence $E(Y) - E(X) = 2\tau - \frac{a+b}{2}$

5 # (i) Find $E\left(\frac{Y}{X}\right)$. $E\left(\frac{Y}{X}\right) = E(Y)E\left(\frac{1}{X}\right) = \tau^2 \left(\frac{1}{x} \right) \Big|_{x=a}^{x=b} = \tau^2 \left(\frac{1}{a} - \frac{1}{b} \right) = \frac{\tau^2}{ab}$

5 # (ii) Let $U = \frac{Y}{X}$. Find the probability density function of U .

$P(U > u) = P(Y > ux)$

Note U always exceeds $\frac{\tau}{b}$:
 $U = \frac{Y}{X} > \frac{\tau}{b} > \frac{\tau}{a}$

$$= \int_a^b \frac{P(Y > ux)}{b-a} dx = \int_a^b \frac{\tau^2}{(ux)^2} dx = \frac{1}{u^2} \frac{\tau^2}{ab} \text{ if } ua \geq \tau$$

for $\frac{\tau}{b} \leq u < \frac{\tau}{a}$,

$$P(U > u) = \int_a^b P(Y > ux) dx$$

$$= \int_a^b \frac{1}{b-a} \frac{P(Y > ux)}{1} dx + \int_a^b \frac{1}{b-a} \frac{P(Y > ux)}{1} dx$$

$$= \frac{\tau^2}{a^2} - a + \left(\frac{\tau}{u} \right)^2 \left(\frac{1}{x} \right) \Big|_{x=\frac{\tau}{u}}^{x=b}$$

$$= \frac{2\tau}{u^2} - a - \left(\frac{\tau}{u} \right)^2 \frac{1}{b} \therefore f_U(u) = \frac{2}{u^3} \frac{\tau^2}{ab} I\left(\frac{\tau}{b}, \infty\right)$$

$$u = \frac{x}{y}, \quad v = \frac{y}{x}$$

$$x = \frac{v}{u}, \quad y = v$$

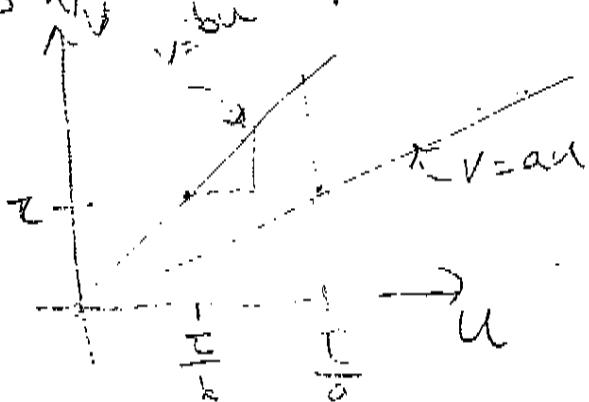
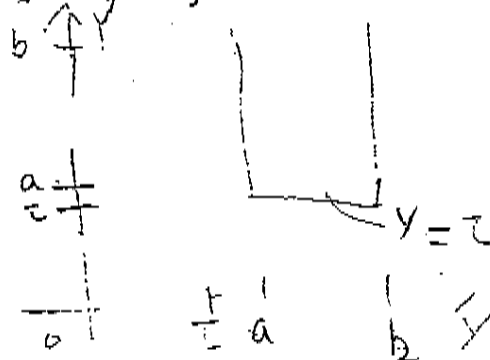
$$J = \begin{vmatrix} -\frac{v}{u^2} & \frac{1}{u} \\ 0 & 1 \end{vmatrix}$$

$$= -\frac{v}{u^2}$$

$$u, v(u, v) = f_{x, y}(x(u, v), y(u, v)) |J|$$

$$= \frac{1}{b-a} I(a, b) \left(\frac{v}{u}\right) 2\tau^2 v^{-2} I\left(\frac{v}{u}, \infty\right) \frac{v}{u^2}$$

$$= \frac{1}{u^2(b-a)} \int_{au}^{bu} v^{-2} I\left(\frac{v}{u}, \infty\right) dv + \frac{1}{u^2(b-a)} \int_{\frac{\tau}{b}}^{\frac{\tau}{a}} v^{-2} I\left(\frac{v}{u}, \frac{\tau}{a}\right) dv$$



$$v = 0 \Rightarrow \frac{v}{u} = 0 \quad v = au$$

$$f_{u, v}(u) = \frac{2\tau^2}{u^2(b-a)} \left[\int_{au}^{bu} v^{-2} dv I\left(\frac{v}{u}, \infty\right) + \int_{\frac{\tau}{b}}^{\frac{\tau}{a}} v^{-2} dv I\left(\frac{v}{u}, \frac{\tau}{a}\right) \right]$$

$$= \frac{2\tau^2}{u^2(b-a)} \left[-\frac{1}{v} \Big|_{v=au}^{v=bu} I\left(\frac{v}{u}, \infty\right) + -\frac{1}{v} \Big|_{v=\tau}^{v=b\tau} I\left(\frac{v}{u}, \frac{\tau}{a}\right) \right]$$

$$= 2\tau^2 \left[\frac{1}{au} - \frac{1}{bu} \right] I\left(\frac{1}{u}, \infty\right) + \left(\frac{1}{\tau} - \frac{1}{bu} \right) I\left(\frac{1}{u}, \frac{\tau}{a}\right)$$

$$E(u) = \frac{2\tau^2}{ab} \int_{\frac{\tau}{a}}^{\frac{\tau}{b}} u^{-2} du + \frac{2\tau}{b-a} \int_{\frac{\tau}{b}}^{\frac{\tau}{a}} \frac{1}{u} - \frac{\tau}{b} u^{-2} du$$

$$= \frac{2\tau^2}{ab} \left[-u^{-1} \right]_{u=\frac{\tau}{a}}^{u=\frac{\tau}{b}} + \frac{2\tau}{b-a} \left[\ln(u) + \frac{\tau}{b} u^{-1} \right] \Big|_{u=\frac{\tau}{b}}^{u=\frac{\tau}{a}}$$

$$= \frac{2\tau^2}{ab} \left[\frac{a}{\tau} \right] + \frac{2\tau}{b-a} \left[\ln\left(\frac{\tau}{a}\right) - \ln\left(\frac{\tau}{b}\right) + \frac{\tau}{b} \left(\frac{a}{\tau} - \frac{b}{\tau}\right) \right]$$

$$= \frac{2\tau}{b} + \frac{2\tau}{b-a} \left[\ln(b) - \ln(a) \right] - \frac{2\tau}{b}$$

— closed book #4

$$a) E[X] = \int_{\eta}^{\infty} x \frac{1}{t} e^{-\frac{x-\eta}{t}} dx = \frac{1}{t} \int_{\eta}^{\infty} x e^{-\frac{x-\eta}{t}} dx$$

$$= \frac{1}{t} \int_{\eta}^{\infty} (x - \eta + \eta) e^{-\frac{x-\eta}{t}} dx$$

$$= \frac{1}{t} \int_{\eta}^{\infty} (x - \eta) e^{-\frac{x-\eta}{t}} dx + \frac{\eta}{t} \int_{\eta}^{\infty} e^{-\frac{x-\eta}{t}} dx$$

$$\text{let } y = x - \eta, dy = dx$$

$$\text{let } z = x - \eta, dz = dx$$

$$= \frac{1}{t} \int_0^{\infty} y e^{-\frac{y}{t}} dy + \frac{\eta}{t} \int_0^{\infty} e^{-\frac{z}{t}} dz$$

$$= \underbrace{t \Gamma(2) \int_0^{\infty} \frac{1}{t^2 \Gamma(2)} y^{2-1} e^{-\frac{y}{t}} dy}_{\text{Gamma}(2, 1)} + \underbrace{\eta \int_0^{\infty} \frac{1}{t} e^{-\frac{z}{t}} dz}_{\text{Exp}(t)}$$

$$\Gamma(2) = 1$$

Gamma(2, 1)

Exp(t)

$$= t + \eta$$

Using the hint.

$$E[X^2] = \text{Var}(X) + (E[X])^2 = t^2 + (t + \eta)^2$$

To find the MME's solve

$$E[X] = \theta + \eta = \frac{1}{n} \sum x_i$$

$$E[X^2] = \theta^2 + (\theta + \eta)^2 = \frac{1}{n} \sum x_i^2$$

$$(1) \quad \tilde{\eta} = \frac{1}{n} \sum x_i - \tilde{\theta} = \bar{x} - \tilde{\theta}$$

$$(2) \quad \theta^2 + (\theta + \eta)^2 = \frac{1}{n} \sum x_i^2$$

$$2\theta^2 + 2\theta\eta + \eta^2 = \frac{1}{n} \sum x_i^2$$

$$2\theta^2 + 2\theta(\bar{x} - \theta) + (\bar{x} - \theta)^2 = \frac{1}{n} \sum x_i^2$$

$$\cancel{2\theta^2} + \cancel{2\theta\bar{x}} - \cancel{2\theta^2} + \bar{x}^2 - \cancel{2\bar{x}\theta} + \theta^2 = \frac{1}{n} \sum x_i^2$$

$$\theta^2 = \frac{1}{n} \sum x_i^2 - \bar{x}^2$$

$$\theta^2 = \frac{\sum x_i^2 - \frac{(\sum x_i)^2}{n}}{n}$$

$$\tilde{\theta} = \sqrt{\frac{\sum x_i^2 - \frac{(\sum x_i)^2}{n}}{n}}$$

$$b) \quad L(\theta, \eta) = \prod_{i=1}^n \frac{1}{\theta} e^{-\frac{x_i - \eta}{\theta}} \quad x_{(1)} \leq x_i < \infty$$

$$= \theta^{-n} e^{-\frac{1}{\theta} \sum (x_i - \eta)} \quad x_{(1)} < \infty.$$

$$\ell(\theta, \eta) = -n \log \theta - \frac{1}{\theta} \sum (x_i - \eta)$$

$$\frac{d}{d\theta} \ell(\theta, \eta) = -\frac{n}{\theta} + \frac{1}{\theta^2} \sum (x_i - \eta) = 0$$

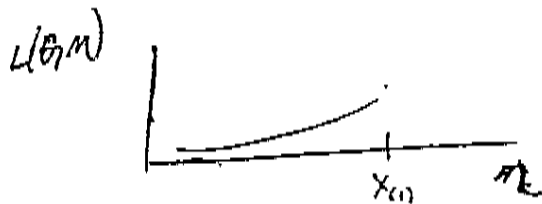
$$\frac{n}{\theta} = \frac{1}{\theta^2} \sum (x_i - \eta)$$

$$\hat{\theta} = \frac{\sum (x_i - \hat{\eta})}{n} //$$

c) irregular estimation

$$L(\theta, \eta) = \theta^{-n} e^{-\frac{1}{\theta} \sum x_i} e^{+\frac{\eta}{\theta}}$$

$$L(\theta, \eta) \propto e^{+\eta} \quad \eta < x_{(1)}$$



$$\hat{\eta} = x_{(1)}$$

$$d) f(x_1, \dots, x_n | \theta, \eta) = \theta^{-n} e^{-\frac{1}{\theta} \sum (x_i - \eta)} I_{(\eta, \infty)}(x_{(1)})$$

$$= \theta^{-n} e^{+\frac{\eta}{\theta}} e^{-\frac{\sum x_i}{\theta}} I_{(\eta, \infty)}(x_{(1)})$$

$\therefore (\sum x_i, x_{(1)})$ are suff. for (θ, η)

e) From the factorization theorem, the likelihood is $g(T, \theta) h(\underline{x})$, which depends on θ only through T .
Therefore, to maximize this quantity we need only maximize $g(T, \theta)$.

$$f) f(x|\theta) = \frac{1}{\theta} e^{-\frac{x}{\theta}}, \quad \log f(x|\theta) = -\log \theta - \frac{x}{\theta}$$

$$\frac{d}{d\theta} \log f(x|\theta) = -\frac{1}{\theta} - \frac{x}{\theta^2}$$

$$\frac{d^2}{d\theta^2} \log f(x|\theta) = \frac{1}{\theta^2} - \frac{2x}{\theta^3}$$

$$I(\theta) = -E\left[\frac{d^2}{d\theta^2} \log f(x|\theta)\right] = -E\left[\frac{1}{\theta^2} - \frac{2x}{\theta^3}\right]$$

$$= -\frac{1}{\theta^2} + \frac{2}{\theta^3} E[x] = -\frac{1}{\theta^2} + \frac{2}{\theta^3} \theta = \frac{1}{\theta^2}$$

$$CRLB = \frac{1}{n I(\theta)} = \frac{\theta^2}{n}$$

For the $\text{Exp}(\theta)$, the MLE is $\hat{\theta} = \bar{X}$.

Since $\text{Var}(\hat{\theta}) = \text{Var}(\bar{X}) = \frac{\theta^2}{n}$, $\hat{\theta}$ achieves the CRLB

$$g) \hat{\theta} \pm z_{\alpha/2} \frac{1}{\sqrt{n I(\hat{\theta})}} \Rightarrow \hat{\theta} \pm z_{\alpha/2} \frac{\hat{\theta}}{\sqrt{n}} \quad \text{where } \hat{\theta} = \bar{x}$$

- e) The asymptotic CI computed in g may not be accurate since it requires a large sample. The coverage probability may be low since the CI would be too narrow as a result of the small sample size.

To implement the parametric bootstrap:

- ① Using the MLE $\hat{\theta} = \bar{x}$ generate B samples from $\text{Exp}(\hat{\theta})$. ② Compute

$\hat{\theta}_1^*, \dots, \hat{\theta}_B^*$ as the sample mean of

each bootstrap sample. ③ The bootstrap estimate of θ is

$$\bar{\hat{\theta}}^* = \frac{1}{B} \sum_{i=1}^B \hat{\theta}_i^*$$

④ The bootstrap estimate of $S_{\hat{\theta}}$ is $S_{\hat{\theta}} = \sqrt{\frac{\sum (\hat{\theta}_i^* - \bar{\hat{\theta}}^*)^2}{B-1}}$

- ⑤ An approximate 95% bootstrap CI can be obtained from the 0.025 and 0.975 percentiles of the $\hat{\theta}_1^*, \dots, \hat{\theta}_B^*$.

closed book #3

Suppose that n is a positive integer and that $0 < p < 1$. Let N be a binomial random variable with parameters n and p . For each positive integer i , let X_i be a random variable whose probability density function is

$$f_X(x) = \begin{cases} \frac{1}{2} x^2 \exp(-x) & \text{if } x > 0 \\ 0 & \text{elsewhere.} \end{cases}$$

Suppose that N, X_1, X_2, \dots are independent. Let

$$Y_k = \begin{cases} 1 & \text{if } k = 0 \\ \prod_{i=1}^k X_i & \text{if } k = 1, 2, \dots \end{cases}$$

(a) Find $P(Y_1 > 2)$.
 $P(Y_1 > 2) = P(X_1 > 2) = \int_2^{\infty} \frac{1}{2} x^2 e^{-x} dx = -\frac{1}{2} x^2 e^{-x} \Big|_2^{\infty} + \int_2^{\infty} x e^{-x} dx$
 $= 2e^{-2} - x e^{-x} \Big|_2^{\infty} + \int_2^{\infty} e^{-x} dx = 2e^{-2} - 2e^{-2} + e^{-2} = e^{-2}$ or $P(Y_1 > 2) = P(N(2) < 3) = e^{-2} + 2e^{-2} + \frac{2^2}{2!} e^{-2}$

(b) Find $E(Y_N | N=10)$ in terms of $\mu = E(X_1)$. You need not determine the value of μ .

$$E(Y_N | N=10) = E\left(\prod_{i=1}^{10} X_i\right) = \prod_{i=1}^{10} E(X_i) = \mu^{10}$$

(c) Find $E(Y_N)$ in terms of $\mu = E(X_1)$, n , and p . You need not determine the value of

$$\mu \cdot E(Y_N) = \sum_{k=0}^n E(Y_k | N=k) \binom{n}{k} p^k (1-p)^{n-k} = \sum_{k=0}^n \mu^k \binom{n}{k} p^k (1-p)^{n-k} = (1-p + \mu p)^n$$

For $k=1, 2, \dots$ let $Z_k = \frac{Y_k}{E(Y_k)}$. Show that $Z_k, k=1, 2, \dots$ is a Martingale.

$$\begin{aligned} E(Z_{k+1} | Z_1, \dots, Z_k) &= E\left(Z_k \frac{Y_{k+1}}{E(Y_{k+1})} \mid Z_1, \dots, Z_k\right) \\ &= Z_k E\left(\frac{Y_{k+1}}{E(Y_{k+1})} \mid Z_1, \dots, Z_k\right) \\ &= Z_k E\left(\frac{Y_{k+1}}{E(Y_{k+1})}\right) \\ &= Z_k \end{aligned}$$