

# Ph.D. Qualifying Examination in Statistics

Department of Mathematics  
University of Louisville  
August 12, 2016, 9:00am–12:30pm

Do 3 problems from each section.

## SECTION 1

**Problem 1.** Suppose  $Y$  is a random variable with probability density function (pdf)

$$f(y|\alpha) = \begin{cases} \frac{\alpha - 1}{y^\alpha} & \text{for } y > 1 \\ 0 & \text{otherwise} \end{cases}$$

where  $\alpha > 1$ .

- (a) A family of probability density functions is called an exponential family if it can be expressed as  $f(x|\theta) = h(x)C(\theta) \exp\{W(\theta)t(x)\}$ . Is  $\{f_Y(y|\alpha)\}$  an exponential family? If yes, define  $\theta$  and find  $h(y)$ ,  $C(\alpha)$ ,  $W(\alpha)$ , and  $t(y)$ ? If not, justify your answer.
- (b) Find  $E[\ln Y]$ .

**Problem 2.** Suppose  $X_1, \dots, X_n$  is a random sample from a normal distribution with mean 0 and variance  $\theta^2$  having probability density function

$$f(x|\theta) = \frac{1}{\sqrt{2\pi\theta^2}} \exp\left\{-\frac{1}{2\theta^2}x^2\right\}, \theta > 0.$$

- (a) Find the maximum likelihood estimator of  $\theta^2$ .
- (b) Is the estimator in part (a) sufficient for  $\theta^2$ ? Justify your answer.
- (c) What is the maximum likelihood estimator of  $P(X_1 > \theta^2)$ ? Write your answer in terms of the standard normal cdf  $\Phi(t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2}u^2\right\} du$ ?

**Problem 3.** Let  $X$  be a Poisson( $\lambda$ ) random variable having probability mass function

$$P(X = x) = \begin{cases} \frac{1}{x!} \lambda^x e^{-\lambda} & \text{if } x = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

Suppose that  $X_1, \dots, X_n$  are independent Poisson( $\lambda$ ) random variables.

- (a) Show that  $\bar{X} = \sum_{i=1}^n X_i/n$  is an unbiased estimator of  $\lambda$ .
- (b) What is the variance of  $\bar{X}$ ?
- (c) Show that  $\bar{X}$  satisfies the Cramér-Rao Lower Bound for an unbiased estimator of  $\lambda$ .

**Problem 4.** Suppose that  $X_1, \dots, X_n$  are independently and identically distributed according to the uniform(0,  $\theta$ ) distribution. Let  $M_n = \max(X_1, \dots, X_n)$ . Let

$$\delta_c(X_1, \dots, X_n) = \begin{cases} 1 & \text{if } M_n \geq c \\ 0 & \text{otherwise} \end{cases}$$

be the test function for which  $\delta_c = 1$  indicates that the null hypothesis should be rejected and  $\delta_c = 0$  indicates that it should not be rejected.

- (a) For  $0 < c < \theta$ , compute the power function  $\beta(\theta, \delta_c) = P_\theta(\delta_c = 1)$ .
- (b) In testing  $H_0 : \theta \leq \frac{1}{2}$  versus  $H_A : \theta > \frac{1}{2}$ , what choice of  $c$  would make the test based on the test function  $\delta_c$  have size 0.05? Your choice of  $c$  should depend on  $n$ .
- (c) How large should  $n$  be so that the test specified in (b) has power 0.98 for  $\theta = \frac{3}{4}$ ?

## SECTION 2

**Problem 5.** Let  $\mathbf{y}$  be an  $n$ -dimensional column vector,  $\mathbf{X}$  be an  $n \times k$  matrix,  $\boldsymbol{\beta}$  be a  $k$ -dimensional column vector, and  $\mathbf{0}_k$  is a  $k$ -dimensional vector of zeros.

- (a) Let  $Q(\boldsymbol{\beta}) = \|\mathbf{y} - \mathbf{X}\boldsymbol{\beta}\|^2$ . Compute  $\frac{\partial Q}{\partial \boldsymbol{\beta}}$ .
- (b) Derive the solution to the score equation  $\frac{\partial Q}{\partial \boldsymbol{\beta}} = \mathbf{0}_k$ . Let  $\hat{\boldsymbol{\beta}}$  denote the solution.
- (c) Show that  $\|\mathbf{y} - \mathbf{X}\boldsymbol{\beta}\|^2 = \|\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}\|^2 + \|\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})\|^2$ .

**Problem 6.** Suppose that  $V_1, V_2, V_3,$  and  $V_4$  are independent standard normal random variables. Let  $\bar{V} = \frac{V_1 + V_2 + V_3 + V_4}{4}$ . Using the tables attached to this exam, compute (approximately) the following probabilities.

- (a)  $P\left(\bar{V} > 1 \text{ and } \sum_{i=1}^4 (V_i - \bar{V})^2 < 1\right)$
- (b)  $P\left(\bar{V} > \sqrt{\sum_{i=1}^4 (V_i - \bar{V})^2}\right)$

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**FORMULAS (for problems 7 and 8):**

Suppose  $\mathbf{y} \sim \text{Normal}_n(\mathbf{X}\boldsymbol{\beta}, \sigma^2\mathbf{I}_n)$ ,  $\mathbf{X}$  is a  $n \times p$  full rank matrix,  $n > p$ , and  $\mathbf{X}^\top \mathbf{X}$  is invertible. Let  $\hat{\boldsymbol{\beta}}$  be the MLE of  $\boldsymbol{\beta}$  and  $\hat{\sigma}^2$  be the MLE of  $\sigma^2$ . Then  $\frac{\|\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})\|^2/p}{\|\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}\|^2/(n-p)} \sim f_{p, n-p}$ . If, in addition,  $\mathbf{K}$  is a  $m \times p$  full rank matrix (with  $m \leq p$ ) such that  $\mathbf{K}^\top \boldsymbol{\beta} = \mathbf{k}$ , then

$$\begin{aligned} F &= \frac{(RSS_H - RSS)/m}{RSS/(n-p)} \\ &= \frac{\|\mathbf{X}(\hat{\boldsymbol{\beta}} - \tilde{\boldsymbol{\beta}})\|^2/m}{\|\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}\|^2/(n-p)} \\ &= \frac{(\mathbf{K}^\top \hat{\boldsymbol{\beta}} - \mathbf{k})^\top (\mathbf{K}^\top (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{K})^{-1} (\mathbf{K}^\top \hat{\boldsymbol{\beta}} - \mathbf{k})/m}{(\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}})^\top (\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}})/(n-p)} \sim f_{m, n-p} \end{aligned}$$

where  $\tilde{\boldsymbol{\beta}}$  is the restricted MLE of  $\boldsymbol{\beta}$  satisfying  $\mathbf{K}^\top \tilde{\boldsymbol{\beta}} = \mathbf{k}$ ,  $RSS_H = \|\mathbf{y} - \mathbf{X}\tilde{\boldsymbol{\beta}}\|^2$ , and  $RSS = \|\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}}\|^2 = n\hat{\sigma}^2$ .

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**Problem 7.** Suppose that  $\mathbf{y}$  is a 4-dimensional  $\text{Normal}(\mathbf{X}\boldsymbol{\beta}, \sigma^2\mathbf{I})$  random vector where  $\mathbf{X}$  is a  $4 \times 2$  matrix with columns  $\mathbf{J}$  and  $\mathbf{x}$ ,  $\mathbf{J}$  is an 4-dimensional vector of ones,  $\mathbf{x}$  is a non-random 4-dimensional vector,  $\boldsymbol{\beta} = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$  is a unknown vector of intercept and slope parameters, and  $\sigma^2$  is the unknown variance parameter. Let  $\hat{\beta}_0$ ,  $\hat{\beta}_1$ , and  $\hat{\sigma}^2$  denote the maximum likelihood estimators of  $\beta_0$ ,  $\beta_1$  and  $\sigma^2$ , respectively.

(a) If  $\mathbf{J}^T \mathbf{x} = 1$  and  $\mathbf{x}^T \mathbf{x} = 3$ , find values  $A$ ,  $B$ , and  $C$  such that

$$A \left( \frac{\hat{\beta}_0 - \beta_0}{\hat{\sigma}} \right)^2 + B \left( \frac{\hat{\beta}_0 - \beta_0}{\hat{\sigma}} \right) \left( \frac{\hat{\beta}_1 - \beta_1}{\hat{\sigma}} \right) + C \left( \frac{\hat{\beta}_1 - \beta_1}{\hat{\sigma}} \right)^2 \leq 1$$

is a 95% confidence ellipse for  $\boldsymbol{\beta}$ .

(b) Now, also suppose that  $\mathbf{J}^T \mathbf{y} = 2$ ,  $\mathbf{y}^T \mathbf{y} = 38$ , and  $\mathbf{x}^T \mathbf{y} = 10$ . Compute the maximum likelihood estimates  $\hat{\beta}_0$ ,  $\hat{\beta}_1$ , and  $\hat{\sigma}^2$ . Is  $\boldsymbol{\beta} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$  inside the confidence ellipse in part (a)?

**Problem 8.** Suppose that

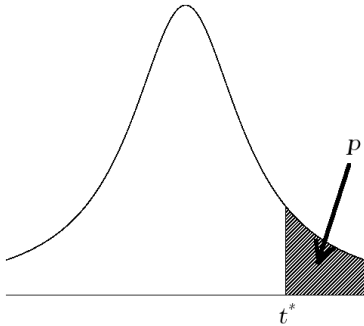
$$y_{ijk} = \mu_i + \alpha_{ij} + e_{ijk} \text{ for } i, j, k = 1, 2$$

where  $\alpha_{11} + \alpha_{12} = \alpha_{21} + \alpha_{22} = 0$  and  $e_{ijk}$  are independent  $\text{Normal}(0, \sigma^2)$  random variables. Given the data

		$i$	
		1	2
$j$	1	7,9	6,4
	2	1,3	2,4

test  $H_0 : \alpha_{11} = 0$  versus  $H_0 : \alpha_{11} \neq 0$  at level 0.05. Make sure to state your conclusion and to show sufficient details (such as your test statistic and the corresponding critical value or bound for the  $P$ -value) to justify your conclusion.

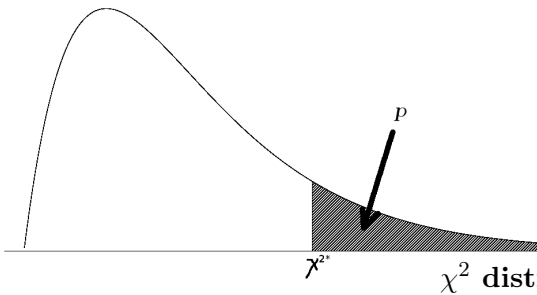




The critical value  $t^*$  is the value such that the area under the density curve of a  $t$  distribution with  $df$  degrees of freedom to the right of  $t^*$  is equal to  $p$ . It is also the value such that the area under the curve between  $-t^*$  and  $t^*$  is equal to  $C$ .

**$t$  distribution critical values**

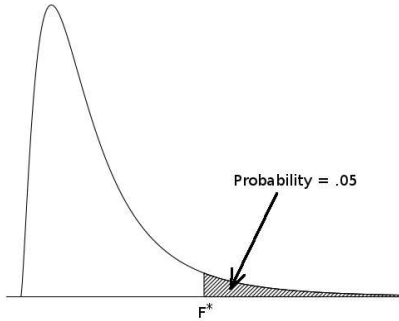
df	Upper-tail probability $p$									
	.05	.045	.04	.035	.03	.025	.02	.015	.01	.005
1	6.3	7.0	7.9	9.1	10.6	12.7	15.9	21.2	31.8	63.7
2	2.9	3.1	3.3	3.6	3.9	4.3	4.8	5.6	7.0	9.9
3	2.4	2.5	2.6	2.8	3.0	3.2	3.5	3.9	4.5	5.8
4	2.1	2.2	2.3	2.5	2.6	2.8	3.0	3.3	3.7	4.6
5	2.0	2.1	2.2	2.3	2.4	2.6	2.8	3.0	3.4	4.0
6	1.9	2.0	2.1	2.2	2.3	2.4	2.6	2.8	3.1	3.7
7	1.9	2.0	2.0	2.1	2.2	2.4	2.5	2.7	3.0	3.5
	90%	91%	92%	93%	94%	95%	96%	97%	98%	99%
	Confidence level $C$									



The critical value  $\chi^{2*}$  is the value such that the area under the density curve of a  $\chi^2$  distribution with  $df$  degrees of freedom to the right of  $\chi^{2*}$  is equal to  $p$ .

**$\chi^2$  distribution critical values**

df	Upper-tail probability $p$									
	.90	.80	.70	.60	.50	.40	.30	.20	.10	.05
1	0.02	0.06	0.1	0.3	0.5	0.7	1.1	1.6	2.7	3.8
2	0.2	0.4	0.7	1.0	1.4	1.8	2.4	3.2	4.6	6.0
3	0.6	1.0	1.4	1.9	2.4	2.9	3.7	4.6	6.3	7.8
4	1.1	1.6	2.2	2.8	3.4	4.0	4.9	6.0	7.8	9.5
5	1.6	2.3	3.0	3.7	4.4	5.1	6.1	7.3	9.2	11.1
6	2.2	3.1	3.8	4.6	5.3	6.2	7.2	8.6	10.6	12.6
7	2.8	3.8	4.7	5.5	6.3	7.3	8.4	9.8	12.0	14.1



The critical value  $F^*$  is the value such that the area under the density curve of an  $F$  distribution with  $df1$  degrees of freedom in the numerator and  $df2$  degrees of freedom in the denominator to the right of  $F^*$  is equal to .05.

**F distribution critical values**

		df1 = degrees of freedom in the numerator										
		1	2	3	4	5	10	20	30	40	50	100
df2 = degrees of freedom in the denominator	1	161.45	199.50	215.71	224.58	230.16	241.88	248.01	250.10	251.14	251.77	253.04
	2	18.51	19.00	19.16	19.25	19.30	19.40	19.45	19.46	19.47	19.48	19.49
	3	10.13	9.55	9.28	9.12	9.01	8.79	8.66	8.62	8.59	8.58	8.55
	4	7.71	6.94	6.59	6.39	6.26	5.96	5.80	5.75	5.72	5.70	5.66
	5	6.61	5.79	5.41	5.19	5.05	4.74	4.56	4.50	4.46	4.44	4.41
	6	5.99	5.14	4.76	4.53	4.39	4.06	3.87	3.81	3.77	3.75	3.71
	7	5.59	4.74	4.35	4.12	3.97	3.64	3.44	3.38	3.34	3.32	3.27
	8	5.32	4.46	4.07	3.84	3.69	3.35	3.15	3.08	3.04	3.02	2.97
	9	5.12	4.26	3.86	3.63	3.48	3.14	2.94	2.86	2.83	2.80	2.76
	10	4.96	4.10	3.71	3.48	3.33	2.98	2.77	2.70	2.66	2.64	2.59
	11	4.84	3.98	3.59	3.36	3.20	2.85	2.65	2.57	2.53	2.51	2.46
	12	4.75	3.89	3.49	3.26	3.11	2.75	2.54	2.47	2.43	2.40	2.35
	13	4.67	3.81	3.41	3.18	3.03	2.67	2.46	2.38	2.34	2.31	2.26
	14	4.60	3.74	3.34	3.11	2.96	2.60	2.39	2.31	2.27	2.24	2.19
	15	4.54	3.68	3.29	3.06	2.90	2.54	2.33	2.25	2.20	2.18	2.12
	16	4.49	3.63	3.24	3.01	2.85	2.49	2.28	2.19	2.15	2.12	2.07
	17	4.45	3.59	3.20	2.96	2.81	2.45	2.23	2.15	2.10	2.08	2.02
	18	4.41	3.55	3.16	2.93	2.77	2.41	2.19	2.11	2.06	2.04	1.98
	19	4.38	3.52	3.13	2.90	2.74	2.38	2.16	2.07	2.03	2.00	1.94
	20	4.35	3.49	3.10	2.87	2.71	2.35	2.12	2.04	1.99	1.97	1.91
	21	4.32	3.47	3.07	2.84	2.68	2.32	2.10	2.01	1.96	1.94	1.88
	22	4.30	3.44	3.05	2.82	2.66	2.30	2.07	1.98	1.94	1.91	1.85
	23	4.28	3.42	3.03	2.80	2.64	2.27	2.05	1.96	1.91	1.88	1.82
	24	4.26	3.40	3.01	2.78	2.62	2.25	2.03	1.94	1.89	1.86	1.80
	25	4.24	3.39	2.99	2.76	2.60	2.24	2.01	1.92	1.87	1.84	1.78
	26	4.23	3.37	2.98	2.74	2.59	2.22	1.99	1.90	1.85	1.82	1.76
	27	4.21	3.35	2.96	2.73	2.57	2.20	1.97	1.88	1.84	1.81	1.74
	28	4.20	3.34	2.95	2.71	2.56	2.19	1.96	1.87	1.82	1.79	1.73
	29	4.18	3.33	2.93	2.70	2.55	2.18	1.94	1.85	1.81	1.77	1.71
	30	4.17	3.32	2.92	2.69	2.53	2.16	1.93	1.84	1.79	1.76	1.70
40	4.08	3.23	2.84	2.61	2.45	2.08	1.84	1.74	1.69	1.66	1.59	
50	4.03	3.18	2.79	2.56	2.40	2.03	1.78	1.69	1.63	1.60	1.52	
100	3.94	3.09	2.70	2.46	2.31	1.93	1.68	1.57	1.52	1.48	1.39	
1000	3.85	3.00	2.61	2.38	2.22	1.84	1.58	1.47	1.41	1.36	1.26	