

STAT 513 fa 2018 Final Exam

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Do not open this test until told to do so; no calculators allowed; no notes allowed; no books allowed; show your work so that partial credit may be given.

pmf/pdf	\mathcal{X}	$M_X(t)$	$\mathbb{E}X$	$\text{Var } X$
$p_X(x; p) = p^x(1-p)^{1-x}$,	$x = 0, 1$	$pe^t + (1-p)$	p	$p(1-p)$
$p_X(x; n, p) = \binom{n}{x}p^x(1-p)^{n-x}$,	$x = 0, 1, \dots, n$	$[pe^t + (1-p)]^n$	np	$np(1-p)$
$p_X(x; p) = (1-p)^{x-1}p$,	$x = 1, 2, \dots$	$\frac{pe^t}{1-(1-p)e^t}$	p^{-1}	$(1-p)p^{-2}$
$p_X(x; p, r) = \binom{x-1}{r-1}(1-p)^{x-r}p^r$,	$x = r, r+1, \dots$	$\left[\frac{pe^t}{1-(1-p)e^t}\right]^r$	rp^{-1}	$r(1-p)p^{-2}$
$p_X(x; \lambda) = e^{-\lambda}\lambda^x/x!$	$x = 0, 1, \dots$	$e^{\lambda(e^t-1)}$	λ	λ
$p_X(x; N, M, K) = \binom{M}{x}\binom{N-M}{K-x}/\binom{N}{K}$	$x = 0, 1, \dots, K$	¡complicadísimo!	$\frac{KM}{N}$	$\frac{KM}{N} \frac{(N-K)(N-M)}{N(N-1)}$
$p_X(x; K) = \frac{1}{K}$	$x = 1, \dots, K$	$\frac{1}{K} \sum_{x=1}^K e^{tx}$	$\frac{K+1}{2}$	$\frac{(K+1)(K-1)}{12}$
$p_X(x; x_1, \dots, x_n) = \frac{1}{n}$	$x = x_1, \dots, x_n$	$\frac{1}{n} \sum_{i=1}^n e^{tx_i}$	$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$	$\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$
$f_X(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$	$-\infty < x < \infty$	$e^{\mu t + \sigma^2 t^2/2}$	μ	σ^2
$f_X(x; \alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} \exp\left(-\frac{x}{\beta}\right)$	$0 < x < \infty$	$(1-\beta t)^{-\alpha}$	$\alpha\beta$	$\alpha\beta^2$
$f_X(x; \lambda) = \frac{1}{\lambda} \exp\left(-\frac{x}{\lambda}\right)$	$0 < x < \infty$	$(1-\lambda t)^{-1}$	λ	λ^2
$f_X(x; \nu) = \frac{1}{\Gamma(\nu/2)2^{\nu/2}} x^{\nu/2-1} \exp\left(-\frac{x}{2}\right)$	$0 < x < \infty$	$(1-2t)^{-\nu/2}$	ν	2ν
$f_X(x; \alpha, \beta) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1}(1-x)^{\beta-1}$	$0 < x < 1$	$1 + \sum_{k=1}^{\infty} \frac{t^k}{k!} \left(\prod_{r=0}^k \frac{\alpha+r}{\alpha+\beta+r}\right)$	$\frac{\alpha}{\alpha+\beta}$	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$

Let X_1, \dots, X_n be a random sample with likelihood function $L(\theta; X_1, \dots, X_n)$. Then for hypotheses of the form $H_0: \theta = \theta_0$ versus $H_1: \theta \neq \theta_0$ the likelihood ratio takes the form

$$\text{LR}(X_1, \dots, X_n) = \frac{L(\theta_0; X_1, \dots, X_n)}{L(\hat{\theta}; X_1, \dots, X_n)},$$

where $\hat{\theta}$ is the maximum likelihood estimator of θ .

1. Given a set of outcomes from n independent Bernoulli trials with unknown success probability p , Agresti and Coull in [1] suggested the estimator

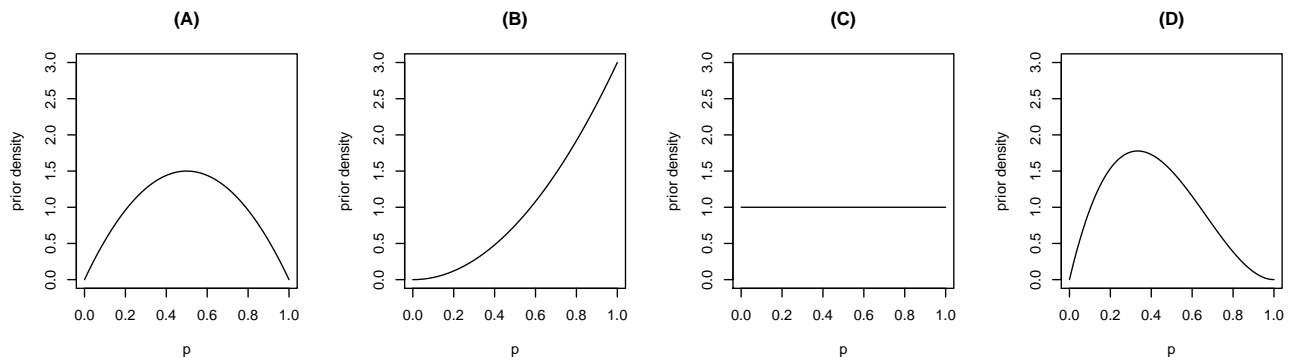
$$\tilde{p} = \frac{Y + 2}{n + 4}$$

for p , where Y is the number of success in the n trials. This can be interpreted as the proportion of successes after adding two successes and two failures to the observed set of outcomes. This question is designed to make you understand why this is a reasonable thing to do.

Suppose we have the Bayesian hierarchical model

$$\begin{aligned} Y|p &\sim \text{Binomial}(n, p) \\ p &\sim \text{Beta}(\alpha, \beta). \end{aligned}$$

- Find the posterior distribution of $p|Y$.
- Give a formula for the posterior mean of $p|Y$.
- Describe how you would construct a 95% credible interval for p after observing a value of Y .
- Find choices of α and β such that the posterior mean of $p|Y$ is equal to the estimator \tilde{p} .
- The pdfs of four Beta distributions are plotted below. Select the distribution which, if used as a prior for p , results in a posterior mean of $p|Y$ which is given by adding *one* success and *one* failure to the observed set of outcomes.



- Suppose two researchers have different prior beliefs about p such that they make different choices of the prior parameters α and β . Then, suppose the two researchers are given the same data set and each of them computes the posterior mean of p . Consider the following scenarios:

- The data set contains 10 observations.
- The data set contains 100 observations.

In which scenario will the researchers' posterior means be closer to one another?

2. Suppose Y_1, \dots, Y_n are independent random variables having pdf

$$f(y; \beta) = \frac{y}{\beta^2} \exp\left(-\frac{y}{\beta}\right) \mathbf{1}(y > 0).$$

- (a) Write down the likelihood function $L(\beta; Y_1, \dots, Y_n)$ of the parameter β based on the data Y_1, \dots, Y_n .
- (b) Give the log-likelihood function $\ell(\beta; Y_1, \dots, Y_n)$.
- (c) Find the maximum likelihood estimator of β .
- (d) Suppose it is of interest to test

$$H_0: \beta = \beta_0 \text{ versus } H_1: \beta \neq \beta_0$$

for some β_0 . Write down the likelihood ratio corresponding to these hypotheses and simplify it so that it is a function of $\bar{Y}_n = n^{-1} \sum_{i=1}^n Y_i$, β_0 , and n .

- (e) For any $\alpha \in (0, 1)$, give a test which will have size approximately equal to α when n is large. Your answer should be of the form, “Reject H_0 if and only if ...”
3. Consider the 2×2 table below, which contains counts from a fictional study in which it was recorded whether subjects in a control and a treatment group experienced an adverse event (like a headache or nausea, etc.).

	Adverse event	No adverse event
Control	18	382
Treatment	2	98

- (a) Compute the table of expected values under the hypothesis that there is no association between the control and treatment groups and experiencing an adverse event (no calculators allowed; you shouldn't need one).
- (b) Which of the tests that we learned in class do you recommend for testing the null hypothesis of no association? Explain your answer.
- (c) Let p_1 and p_2 be the probabilities of experiencing an adverse event in the control and treatment groups, respectively, and let Y_1 and Y_2 be the observed numbers of subjects experiencing adverse events in the control and treatment groups, respectively, and suppose we adopt the Bayesian setup

$$Y_1|p_1 \sim \text{Binomial}(400, p_1), \quad p_1 \sim \text{Beta}(2, 2)$$

$$Y_2|p_2 \sim \text{Binomial}(100, p_2), \quad p_2 \sim \text{Beta}(2, 2),$$

where p_1 and p_2 are independent and $Y_1|p_1$ and $Y_2|p_2$ are independent.

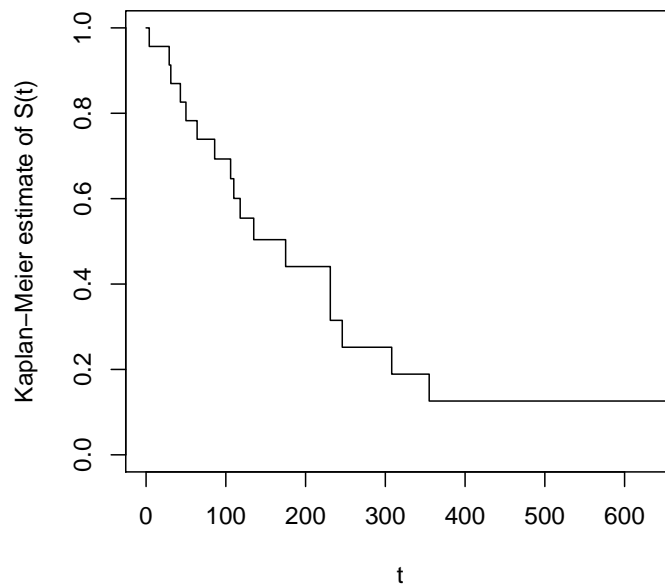
Under this Bayesian setup, give an expression for the posterior mean of $p_2 - p_1|Y_1, Y_2$, and give its value based on observing $Y_1 = 18$ and $Y_2 = 2$ under the sample sizes $n_1 = 400$ and $n_2 = 100$.

4. The following data is a subset of a larger data set from [2] containing the waiting times of patients waiting to receive a liver transplant between 1990 and 1999. The data below are waiting times for females with blood type ‘O’ who are 40 years old or younger.

4	29	31	43	50	64	83+	86	106	110	118	129+
135	144+	146+	175	231	231	246	308	355	406+	637+	

A ‘+’ marker indicates a censoring time (death, withdrawal from the study, etc.) prior to the patient’s receiving a transplant.

- (a) A researcher who wishes to estimate the mean waiting time for patients in this group to receive a liver transplant proposes taking the mean of the above 23 numbers, ignoring the censoring information. Give a critique of this approach, explaining your reasoning.
- (b) The Kaplan-Meier estimate of the survival function based on these data is plotted below.

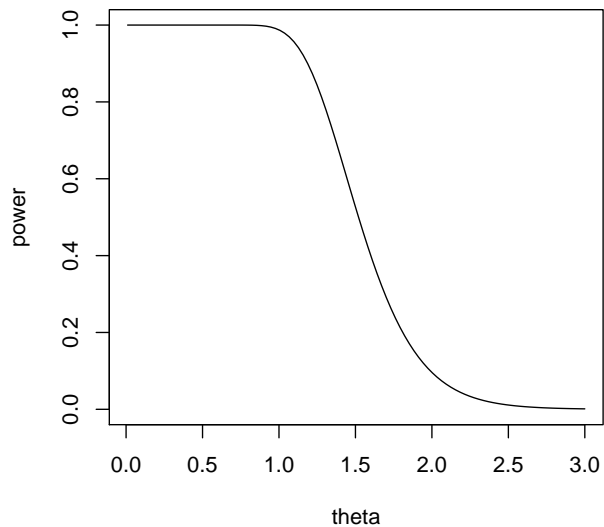


According to this estimate of the survival function, what is the probability that a patient from this group will receive a liver transplant in *fewer than* 300 months?

5. Below is a plot of the power curve of a test of the hypotheses

$$H_0: \theta \geq 2 \text{ versus } H_1: \theta < 2$$

for some parameter θ .



Use the plot to find the following as accurately as you can:

- (a) The size of the test.
- (b) The probability of making an incorrect decision when $\theta = 1.5$.
- (c) The probability of making a correct decision when $\theta = 2.5$.

6. Let Y_1, \dots, Y_n be a random sample from the $\text{Normal}(\mu, \sigma^2)$ distribution, where σ^2 is known.

- (a) Given a real number μ_0 and a value of $\alpha \in (0, 1)$, give a size- α test of the hypotheses

$$H_0: \mu \leq \mu_0 \text{ versus } H_1: \mu > \mu_0.$$

- (b) Give an expression for the power function $\gamma(\mu)$ of your test.
- (c) For any value $\mu_1 > \mu_0$, interpret the value of $\gamma(\mu_1)$.
- (d) For any value $\mu_1 > \mu_0$, describe in detail how the value of $\gamma(\mu_1)$ is affected by
 - i. the sample size n .
 - ii. the value of σ^2 .
 - iii. the choice of α .

References

- [1] Alan Agresti and Brent A. Coull. Approximate is better than “exact” for interval estimation of binomial proportions. *The American Statistician*, 52(2):119–126, 1998.
- [2] W. Ray Kim, Terry M. Therneau, Joanne T. Benson, Walter K. Kremers, Charles B. Rosen, Gregory J. Gores, and E. Rolland Dickson. Deaths on the liver transplant waiting list: an analysis of competing risks. *Hepatology*, 43(2):345–351, 2006.