

Stat 543 Assignment 10
Asymptotics of LRTs and Wald and Score Tests

1. (More Estimation in a Zero-Inflated Poisson Model) Consider the situation of Problem 1 of Assignment 9, and in particular, inference based on the $n = 20$ observations Vardeman simulated from the distribution.

- (a) Find a large sample 90% joint confidence region for (p, λ) based on the loglikelihood function (based on inverting LRT's). Plot this in the (p, λ) -plane and to the extent possible, compare it to the elliptical region you found in Assignment 9.
- (b) Note that for a fixed value of λ ,

$$p = \frac{\frac{n_0}{n} - 1}{\exp(-\lambda) - 1}$$

maximizes the likelihood. Use this fact to find and plot the profile loglikelihood for λ . Use this plot and make an approximate 90% confidence interval for λ . How does this interval compare to the one you found in part e) of Problem 1 from Assignment 9?

2. Consider again the model of Problem 2 of Assignment 9. Below are $n = 20$ observations that Vardeman simulated from this model.

1.36, 1.35, 0.78, 1.85, 2.32, 0.55, 1.07, -0.57, -0.38, 0.25,
 -0.36, 1.71, 1.40, 0.46, 3.16, -0.78, 0.69, -0.03, 1.26, 0.44

- (a) Plot the loglikelihood for this sample. What, approximately, is the maximum likelihood estimate for α ?
 - (b) If you wished to test the hypothesis $H_0: \alpha = .4$ with Type I error probability .1, what would be your decision here? Carefully explain. (Use a likelihood ratio test).
 - (c) Use the plot from a) and make an approximate 90% confidence interval for α based on the likelihood function (based on inverting LRT's). Use the method of f) of Problem 2 on Assignment 9 and make another approximate 90% interval. How do these 2 intervals compare?
3. Suppose that X, Y and Z are independent binomial variables, $X \sim \text{bin}(n, p_1)$, $Y \sim \text{bin}(n, p_2)$ and $Z \sim \text{bin}(n, p_3)$. For the parameter space (for (p_1, p_2, p_3)) $\Theta = [0, 1]^3$, we will consider testing $H_0: p_1 = p_2 = p_3$ based on (X, Y, Z) .
- (a) Find the general forms of the likelihood ratio tests, the Wald tests and the score tests of this hypothesis.

- (b) Use the fact that the parameter space here is basically 3-dimensional while Θ_0 is basically 1-dimensional so that there are 2 independent constraints involved and the limiting χ^2 distributions of the test statistics thus have $\nu = 2$ associated degrees of freedom to actually carry out these tests with $\alpha \approx .05$ if $X = 33$, $Y = 53$ and $Z = 59$, all based on $n = 100$.

The Following Are Problems About the Large Sample Behavior of Posterior Distributions and Likely Will NOT be Covered on the Stat 543 Final Exam.

1. Problems 1.2.4 and 1.2.5 of B&D.
2. Consider Bayesian inference for the binomial parameter p . In particular, for sake of convenience, consider the Uniform $(0, 1)$ (Beta(α, β) for $\alpha = \beta = 1$) prior distribution.
 - (a) It is possible to argue from reasonably elementary principles that in this binomial context, where $\Theta = (0, 1)$, the Beta posteriors have a consistency property. That is, simple arguments can be used to show that for any fixed p_0 and any $\epsilon > 0$, for $X_n \sim \text{binomial}(n, p_0)$, the random variable

$$Y_n = \int_{p_0 - \epsilon}^{p_0 + \epsilon} \frac{1}{B(\alpha + X_n, \beta + (n - X_n))} p^{\alpha + X_n - 1} (1 - p)^{\beta + (n - X_n) - 1} dp$$

(which is the posterior probability assigned to the interval $(p_0 - \epsilon, p_0 + \epsilon)$) converges in p_0 probability to 1 as $n \rightarrow \infty$. This part of the problem is meant to lead you through this argument. Let $\epsilon > 0$ and $\delta > 0$.

- i) Argue that there exists m such that if $n \geq m$, $\left| \frac{x_n}{n} - \frac{\alpha + x_n}{\alpha + \beta + n} \right| < \frac{\epsilon}{3} \quad \forall x_n = 0, 1, \dots, n$.
- ii) Note that the posterior variance is $\frac{(\alpha + x_n)(\beta + n - x_n)}{(\alpha + \beta + n)^2(\alpha + \beta + n + 1)}$. Argue there is an m' such that if $n \geq m'$ the probability that the posterior assigns to $\left(\frac{\alpha + x_n}{\alpha + \beta + n} - \frac{\epsilon}{3}, \frac{\alpha + x_n}{\alpha + \beta + n} + \frac{\epsilon}{3} \right)$ is at least $1 - \delta \quad \forall x_n = 0, 1, \dots, n$.
- iii) Argue that there is an m'' such that if $n \geq m''$ the p_0 probability that $\left| \frac{X_n}{n} - p_0 \right| < \frac{\epsilon}{3}$ is at least $1 - \delta$.

Then note that if $n \geq \max(m, m', m'')$ i) and ii) together imply that the posterior probability assigned to $\left(\frac{x_n}{n} - \frac{2\epsilon}{3}, \frac{x_n}{n} + \frac{2\epsilon}{3} \right)$ is at least $1 - \delta$ for any realization x_n . Then provided $\left| \frac{x_n}{n} - p_0 \right| < \frac{\epsilon}{3}$ the posterior probability assigned to $(p_0 - \epsilon, p_0 + \epsilon)$ is also at least $1 - \delta$. But iii) says this happens with p_0 probability at least $1 - \delta$. That is, for large n , with p_0 probability at least $1 - \delta$, $Y_n \geq 1 - \delta$. Since δ is arbitrary, (and $Y_n \leq 1$) we have the convergence of Y_n to 1 in p_0 probability.

- (b) Vardeman intends to argue in class that posterior densities for large n tend to look normal (with means and variances related to the likelihood material). The posteriors in this binomial problem are Beta $(\alpha + x_n, \beta + (n - x_n))$ (and we can think of $X_n \sim \text{Bi}(n, p_0)$ as derived as the sum of n iid Bernoulli (p_0) variables). So we ought to expect Beta distributions for large parameter values to look roughly normal. To illustrate this do the following. For $\rho = .3$ (for example ... any

other value would do as well), consider the Beta $(\alpha + n\rho, \beta + n(1 - \rho))$ (posterior) distributions for $n = 10, 20, 40$ and 100 . For $p_n \sim \text{Beta}(\alpha + n\rho, \beta + n(1 - \rho))$ plot the probability densities for the variables

$$\sqrt{\frac{n}{\rho(1 - \rho)}} (p_n - \rho)$$

on a single set of axes along with the standard normal density. Note that if W has pdf $f(\cdot)$, then $aW + b$ has pdf $g(\cdot) = \frac{1}{a} f\left(\frac{\cdot - b}{a}\right)$. (Your plots are translated and rescaled posterior densities of p based on possible observed values $x_n = .3n$.) If this is any help in doing this plotting, Vardeman tried to calculate values of the Beta function using MathCad and got the following: $(B(4, 8))^{-1} = 1.32 \times 10^3$, $(B(7, 15))^{-1} = 8.14 \times 10^5$, $(B(13, 29))^{-1} = 2.291 \times 10^{11}$ and $(B(31, 71))^{-1} = 2.967 \times 10^{27}$.