STAC51 TUT02

Week 3: Jan 28, 2021

Binomial Theorem

Theorem 3.2.2 (Binomial Theorem) For any real numbers x and y and integer $n \geq 0$,

$$(x+y)^n = \sum_{i=0}^n \binom{n}{i} x^i y^{n-i}.$$

$$\sum_{i=0}^{n} \binom{n}{i} =$$

Mgf of $X \sim binomial(n,p)$

$$M_X(t)$$
=

Multinomial Theorem

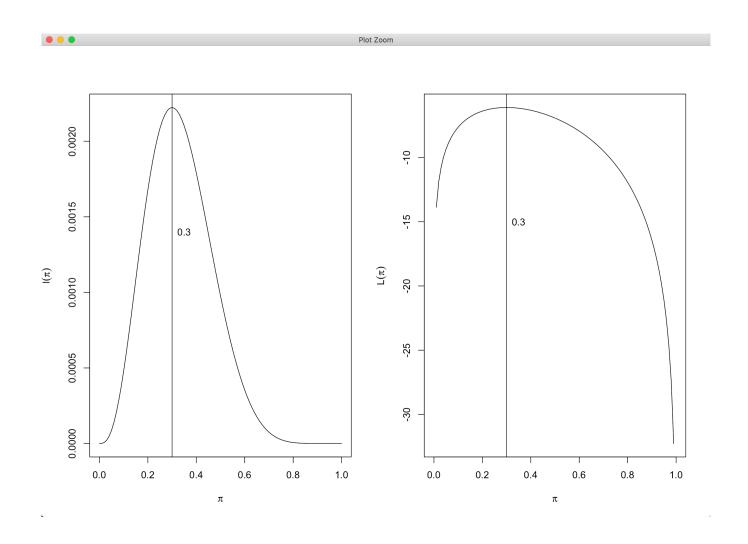
Theorem 4.6.4 (Multinomial Theorem) Let m and n be positive integers. Let A be the set of vectors $\mathbf{x} = (x_1, \ldots, x_n)$ such that each x_i is a nonnegative integer and $\sum_{i=1}^{n} x_i = m$. Then, for any real numbers p_1, \ldots, p_n ,

$$(p_1+\cdots+p_n)^m=\sum_{\mathbf{x}\in\mathcal{A}}\frac{m!}{x_1!\cdots x_n!}p_1^{x_1}\cdots p_n^{x_n}.$$

Binomial Example

- Let's assume that we want to test that a coin is fair
- We toss the coin 10 (n) times and we got 3 (Y) heads
- $\ell(\pi) = \pi^3 (1 \pi)^7$ that is $L(\pi) = 3 \log(\pi) + 7 \log(1 \pi)$. Thus $\hat{\pi} = 3/10$
- Plot the likelihood function and find the value of π that maximizes $\ell(\pi)$.

R Tutorial: The likelihood plots



EXAMPLE

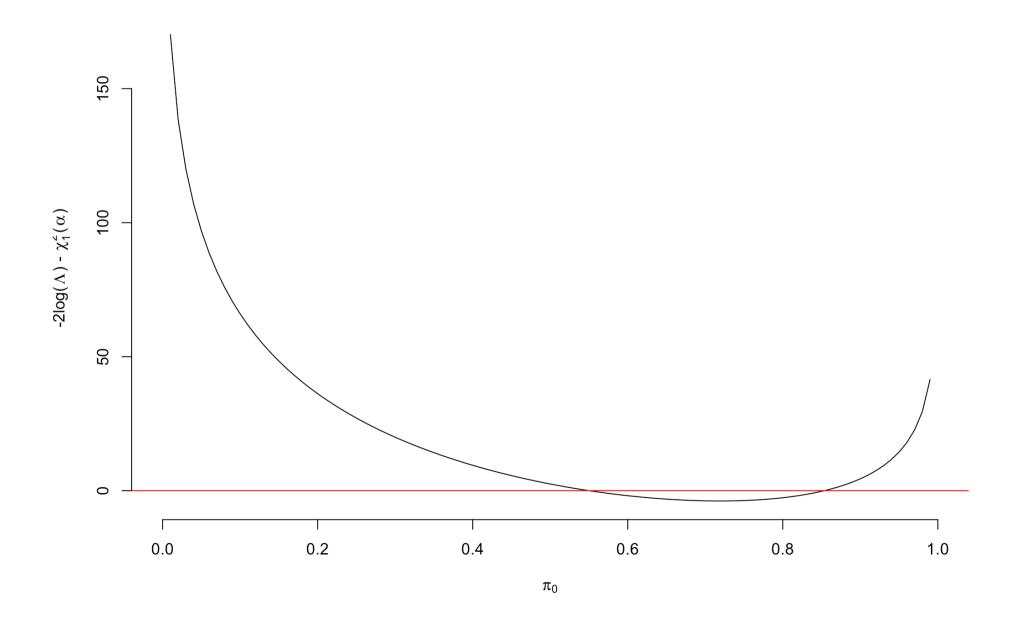
A coin was tossed 32 times and observed 23 heads. Use the likelihood ratio test to test H_0 : $\pi=0.5$ and H_a : $\pi\neq0.5$

R Tutorial: LR based Confidence Interval

- Likelihood based confidence interval for π is the set of values of π_0 for which, $-2\log(\Lambda) < \chi_1^2(\alpha)$
- We can find the boundaries of the interval by solving the equation $-2(L_0-L_1)=\chi_1^2(\alpha)$
- The solution may not be obtainable analytically but we can use numerical methods
- For the previous example we need to solve the following equation,

$$2\left[23\log\left(\frac{23}{32\pi_0}\right) + (32-23)\log\left(\frac{32-23}{32-32\pi_0}\right)\right] - 3.84 = 0$$

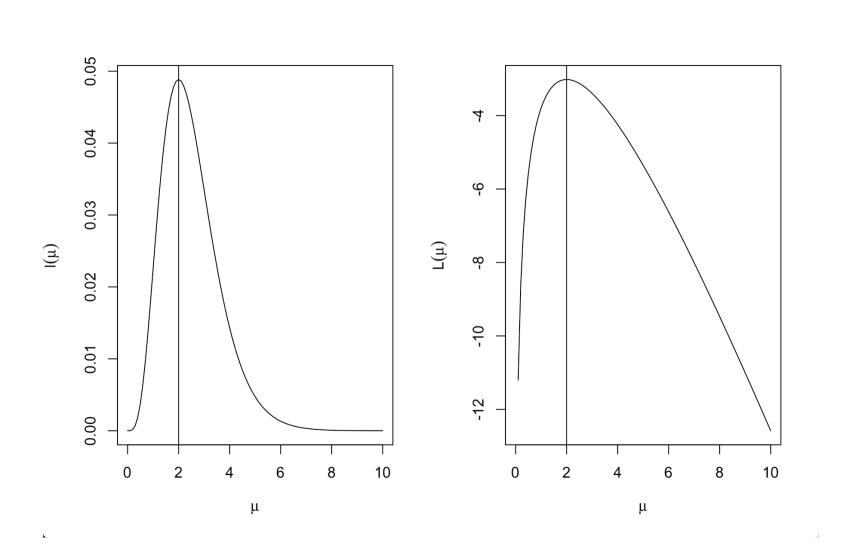
We will solve it with 'R' in the following slide



Assume that the number of cases of tetanus reported in the United States during a single month in 2005 has a Poisson distribution with parameter μ . The number of cases reported in January and February are 1 and 3 respectively.

- (a) Find and plot the likelihood function over the space of potential values for μ .
- (b) What is the maximum likelihood estimate (MLE) of μ ?
- (c) Give an estimate of the probability that there is no case of tetanus reported for a given month.

R: plot Poisson likelihood



For $X \sim bin(n, p)$ with $0 , <math>n \in \mathbb{N}$ we know that

$$\frac{X - np}{\sqrt{npq}} \stackrel{d}{\to} Z \sim N(0, 1)$$
 as $n \to \infty$

a) Use the above approximation in distribution plus the fact that for $Z \sim N(0,1)$ we get $P(|Z| \le 2) \approx .95$ to derive an approximate 95% p-symmetric confidence interval for p and note how its width, W_n , converges to 0 as $n \to \infty$.

b) As $n \to \infty$ we also know that $\widehat{p} = X/n \stackrel{d}{\to} p$, in which case, by slutsky's theorem, we may amend the denominator of the above approximation, thus to obtain, with much greater ease than in a), a slutsky-amended approximate 95% p-symmetric confidence interval for p. Verify that as $n \to \infty$, the width, V_n , of the amended interval also converges to 0

Example 8.3.2 (Binomial power function) Let $X \sim \text{binomial}(5, \theta)$. Consider testing $H_0: \theta \leq \frac{1}{2}$ versus $H_1: \theta > \frac{1}{2}$. Consider first the test that rejects H_0 if and only if all "successes" are observed. The power function for this test is

$$\beta_1(\theta) = P_{\theta}(X \in R) = P_{\theta}(X = 5) = \theta^5.$$

The graph of $\beta_1(\theta)$ is in Figure 8.3.1. In examining this power function, we might decide that although the probability of a Type I Error is acceptably low $(\beta_1(\theta) \le (\frac{1}{2})^5 = .0312)$ for all $\theta \le \frac{1}{2}$, the probability of a Type II Error is too high $(\beta_1(\theta))$ is too small) for most $\theta > \frac{1}{2}$. The probability of a Type II Error is less than $\frac{1}{2}$ only if $\theta > (\frac{1}{2})^{1/5} = .87$. To achieve smaller Type II Error probabilities, we might consider using the test that rejects H_0 if X = 3, 4, or 5. The power function for this test is

$$\beta_2(\theta) = P_{\theta}(X = 3, 4, \text{ or } 5) = {5 \choose 3} \theta^3 (1 - \theta)^2 + {5 \choose 4} \theta^4 (1 - \theta)^1 + {5 \choose 5} \theta^5 (1 - \theta)^0.$$

The graph of $\beta_2(\theta)$ is also in Figure 8.3.1. It can be seen in Figure 8.3.1 that the second test has achieved a smaller Type II Error probability in that $\beta_2(\theta)$ is larger for $\theta > \frac{1}{2}$. But the Type I Error probability is larger for the second test; $\beta_2(\theta)$ is larger for $\theta \leq \frac{1}{2}$. If a choice is to be made between these two tests, the researcher must decide which error structure, that described by $\beta_1(\theta)$ or that described by $\beta_2(\theta)$, is more acceptable.

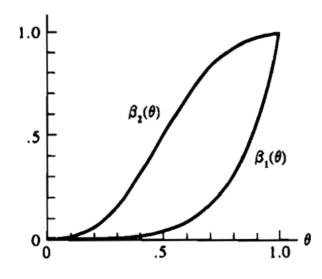


Figure 8.3.1. Power functions for Example 8.3.2