

## Statistics: Homework 3

10.5 Given  $X_1, \dots, X_n \sim \text{Uniform}(0, \theta)$  and  $Y = \max\{X_1, \dots, X_n\}$ , we have the cdf of  $Y$  to be  $F_Y(y) = (y/\theta)^n$  for  $y \in [0, 1/2]$ .

(a) When we choose to reject  $H_0$  when  $Y > c$ , the power function is  $\beta(\theta) = 1 - (c/\theta)^n$ ,  $c \in [0, 1/2]$ .

(b) Given size of the test to be .05, we need to solve,

$$1 - (2c)^n = .05$$

which gives us a solution of  $c = 1/2(.95)^{1/n}$

(c) The size,  $\alpha = \beta(1/2) = 1 - (2c)^n$ ,  $c \in [0, 1/2]$ . Thus, when  $n = 20$ ,  $Y = .48$ , the p-value is

$$\inf\{\alpha : X^n \in R_\alpha\} = 1 - (2 \times .48)^{20} = 0.557997566$$

We would conclude that we do not reject  $H_0$  with an approximate probability of 0.56, which does not give a strong evidence to reject  $H_0$

(d) When  $n = 20$ ,  $Y = .52$ , using the  $\alpha$  formula in (c) gives us  $1 - (2 \times .52)^{20} = -1.19112314$ . But the given  $Y = .52 > 1/2$  which is out of the defined boundaries of the size, i.e.  $F_Y(0.52; \theta = 1/2) = 0$ . Hence the p-value is 0. This allows us to conclude that  $H_0$  is to be rejected as the p-value always lies in the criteria region; a very strong reason to reject  $H_0$ .

10.7b Let  $H_0 : F_T = F_S$  and  $H_1 : F_T \neq F_S$ , where the subscripts denote Twain and Snodgrass respectively. The observed value of the test statistic given by the absolute difference of their means,  $|\bar{T} - \bar{S}|$  is

$$|0.231875 - 0.2097| = 0.022175$$

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Have to do some simulation here.

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Under this p-value, do we reject  $H_0$  at a 5 percent level? How about 2.5 percent level?

10.8 (a) The size of this test with rejection region  $R$  is

$$\begin{aligned} \mathbb{P}(T(X^n) > c | \theta = 0) &= \mathbb{P}(\bar{X}_n > c) \\ &= \mathbb{P}(Z > \sqrt{nc}), \quad Z \text{ is the standard normal distribution} \\ &= 1 - \Phi(\sqrt{nc}), \quad \Phi \text{ is the cdf of the standard normal} \end{aligned}$$

where by Central Limit Theorem,  $\bar{X}_n \sim N(0, 1/\sqrt{n})$ . Thus given size  $\alpha$ , the  $c$  is  $\Phi^{-1}(1 - \alpha)/\sqrt{n}$

(b) Under  $H_1 : \theta = 1$ , the power is  $\beta(1) = \mathbb{P}(T(X^n) > c | \theta = 1) = 1 - \Phi(\sqrt{n}(c - 1))$ . Thus when  $n \rightarrow \infty$ ,  $\sqrt{n}(c - 1) \rightarrow \infty$  for  $c \neq 1$  which then  $1 - \Phi(\sqrt{n}(c - 1)) \rightarrow 1$ .

(c)

10.12 (a) We known that the MLE for  $\lambda$  is  $\bar{X}_n = n^{-1} \sum_{i=1}^n X_i$ . The Fisher information  $I_n(\lambda)$  is

$$I_n(\lambda) = nI(\lambda) = -n\mathbb{E}_\lambda \left( \frac{\partial^2 f_X(X; \lambda)}{\partial \lambda^2} \right) = -n\mathbb{E}_\lambda \left( -\frac{X}{\lambda^2} \right) = \frac{n}{\lambda}$$

thus by the property of MLE,

$$\frac{\bar{X}_n - \lambda}{\hat{\text{se}}} \rightsquigarrow N(0, 1)$$

thus the size of the Wald test

$$\mathbb{P} \left( \left| \frac{\bar{X}_n - \lambda_0}{\sqrt{\lambda_0/n}} \right| > z_{\alpha/2} \right)$$

(b)

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import numpy as np
from scipy.stats import norm
def poisson_sample(l, n):
    """
    Generates n Poisson distributed samples with parameter l.
    """
    return np.random.poisson(lam = l, size = n)
def wald_test(sample, n = 20, alpha = .05, null_lambda = 1):
    """
    Performs Wald test and returns p-value.
    """
    xbar = np.mean(sample)
    test_statistic = np.absolute((xbar - null_lambda) / (null_lambda / n) ** 0.5)
    return 2 * (1 - norm.cdf(test_statistic))
def multwald(l = 1, n = 20, alpha = .05, null_lambda = 1, B = 10000):
    """
    Performs Wald test B times and return proportion of test where null hypothesis is
    rejected.
    """
    count = 0
    for i in np.arange(B):
        sample = poisson_sample(l, n)
        if wald_test(sample) < alpha:
            count += 1

    return count/B
multwald()
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From performing the simulation of Wald 10000 times, the proportion of null rejected is 0.0564 which is very close to the type I error rate of  $\alpha$ .

11.3

11.4