

1. Testing binomial distributions I

[4 points]

Imagine we want to test whether a given coin is fair.

- (a) Suppose that there is *no* additional information. What would be a reasonable test for at level $\alpha = 0.05$ to determine whether the coin is fair, when we flip it $n = 200$ times? Formulate H_0 and H_1 and determine the critical region.

$$(a) H_0 = \{ \text{the coin is fair} \} = \{ p=0.5 \}$$

$$H_1 = \{ \text{the coin is not fair} \} = \{ p \neq 0.5 \}.$$

Let X_i be the Bernoulli Random Variable with parameter p .

$$X_i = \begin{cases} 0 & \text{Head} \\ 1 & \text{Tail.} \end{cases}$$

$$\Rightarrow S(X) = \sum_{i=1}^{n=200} X_i \sim \text{Bin}(200, p)$$

$S(X)$ takes values in $\{0, \dots, 200\}$.

Let $[100-k, 100+k] \cap \mathbb{N}$ be C_2

$$\text{let } \phi(S(x)) = \begin{cases} 0 & \text{if } S(x) \in C_2 \\ 1 & \text{if } S(x) \notin C_2 \end{cases}$$

$$\text{Type I error } \mathbb{P}_{p=0.5} [\phi(S(x)) = 1] = \mathbb{P} [S(x) \geq 100+k] + \mathbb{P} [S(x) \leq 100-k]$$

$$= \sum_{i=100+k}^{200} \binom{200}{i} \left(\frac{1}{2}\right)^{200}$$

$$+ \sum_{j=0}^{100-k} \binom{200}{j} \left(\frac{1}{2}\right)^{200} \leq 0.05$$

$$\Rightarrow \sum_{j=0}^{100-k} \binom{200}{100-j} \left(\frac{1}{2}\right)^{200} = 0.025$$

Thanks to Online Calculator, we find $k=15$

Critical region: $[0, 85] \cup [115, 200]$

Note that we know
these two parts are symmetric

- (b) Suppose now that we suspect that the coin has a higher chance to land on heads. How would the test in (a) change?

We should make the test like this

$$\phi(x) = \begin{cases} 1 & \sum x_i \geq k_2 \\ 0 & \sum x_i < k_2 \end{cases}$$

$$\Rightarrow P_{\alpha=0.05} (\sum x_i \geq k_2) \leq 0.05$$

The biggest k_2 satisfies the condition is

$$\Rightarrow k_2 = 113$$

- (c) Consider both situations in (a) and (b), but use an approximation of the binomial distribution coming from the central limit theorem. How do the critical values change?

For (a), If we use CLT.

$$\text{Var } X_j = 200 \cdot \frac{1}{2} \cdot \frac{1}{2} = 50$$

We are to consider

$$\Pr_{0.5} \left[\sum_{j=1}^n X_j \leq 100 - k \right] = 0.025$$

Approximate it by Central limit Thm.

$$\frac{\sum_{j=1}^n X_j - 100}{\sqrt{200 \cdot \frac{1}{4}}} \leq \frac{100 - k - 100}{\sqrt{200 \cdot \frac{1}{4}}} \approx \Phi \left(\frac{-k}{\sqrt{50}} \right) \leq 0.025$$

$$\Rightarrow \frac{-k}{\sqrt{50}} = -1.96 \Rightarrow k = 13.86 \approx k = 14$$

We could use the trick to substitute $100 - k$ by $100 - k - \frac{1}{2}$

Then $k = 13.86 + 0.5 = 14.36$ which is more accurate.

for (b)

$$\Pr_{0.5} \left[\sum_{j=1}^n X_j \geq k_2 \right] \leq 0.05$$

$$\Rightarrow \frac{\sum_{j=1}^n X_j - 100}{\sqrt{50}} \geq \frac{k_2 - 100}{\sqrt{50}} \approx \Phi \left(\frac{k_2 - 100}{\sqrt{50}} \right) \leq 0.05$$

$$\Phi \left(\frac{k_2 - 100}{\sqrt{50}} \right) \geq 0.95$$

$$\Rightarrow \frac{k_2 - 100}{\sqrt{50}} = 1.644$$

$$\Rightarrow k_2 = 11.62 \Rightarrow k_2 = 112$$

2. Testing binomial distributions II

[4 Points]

The first digit of various numerical data sets (such as electricity bills, street addresses, stock prices,...) typically follows the Benford law. This is a distribution on $(\{1, \dots, 9\}, \mathcal{P}(\{1, \dots, 9\}))$ with probability mass function

$$p(k) = \log_{10} \left(1 + \frac{1}{k} \right), \quad 1 \leq k \leq 9.$$

Here $\log_{10} = \frac{\log}{\log(10)}$ is the logarithm with base 10.

- (a) Show that $\underbrace{(p(k))}_{k \in \{1, \dots, 9\}}$ defines a probability mass function on $\{1, \dots, 9\}$.

By Benford's law, 1 should be the first digit in roughly 30% of the numbers in a valid statistical data set. Suppose we suspect a given sample of 100 independent data points to be fraudulent, since the first digit 1 shows up only 17 times in the sample.

(a) To show $(p(k))$ defines a probability mass function

We show it is

- Nonnegative: $p(k) = \log_{10}(1 + \frac{1}{k}) \geq \log(1) = 0$

- Sum up to 1:

$$\sum_{k=1}^9 p(k) = \log_{10} \prod_{i=1}^{n=9} \left(1 + \frac{1}{i} \right) = \log_{10} \frac{2}{1} \cdot \frac{3}{2} \cdot \frac{4}{3} \cdots \frac{9}{8} \cdot \frac{10}{9} = \log_{10} 10 = 1$$

(b)

- (b) Formulate a testing problem and construct the Neyman-Pearson test ϕ^* at level $\alpha = 0.05$ for the hypotheses

H_0 : the probability for 1 as first digit is 30%, against

H_1 : the probability for 1 as first digit is smaller than 30%.

Find the critical region and determine whether H_0 can be rejected with our observation.

Here we don't know the distribution of 1 as first digit.

Therefore, we model each observation as a Bernoulli Random Variable $X_i \sim \text{Ber}(0)$

In that case, $H_0: \theta = 0.3$

$H_1: \theta < 0.3$

We assume the # of observation is $n=100$

Suppose $\theta_0 > \theta_1$, which is the case here.

$$\begin{aligned} L(\theta_0, \theta_1; (x_1, \dots, x_n)) &= \frac{\theta_1^{\sum_{i=1}^n x_i} (1-\theta_1)^{n-\sum_{i=1}^n x_i}}{\theta_0^{\sum_{i=1}^n x_i} (1-\theta_0)^{n-\sum_{i=1}^n x_i}} \\ &= \left(\frac{\theta_1}{\theta_0}\right)^{\sum_{i=1}^n x_i} \left(\frac{1-\theta_1}{1-\theta_0}\right)^{n-\sum_{i=1}^n x_i} \end{aligned}$$

$$\theta_1^{x_i} (1-\theta_1)^{n-x_i}$$

Note that $\theta_0 = 0.3$
 $\theta_1 < \theta_0$

where we can find a monotone decreasing function for $T(x) = \sum_{i=1}^n x_i$

By Theorem 7.11, we have $\phi^*(x) = \mathbb{1}_{\{T(x) \leq k_2^*\}}$ is a UMP test at level $\alpha \in (0, 1)$ where $\alpha = P_{\theta_0} [T(x) \leq k_2^*]$ for

$$H_0 = \Theta_0 = [\theta_0, +\infty) \cap \Theta = \theta_0$$

$$H_1 = \Theta_0 = (-\infty, \theta_0) \cap \Theta$$

It remains to decide k_2^* for $n=100$

$$P_{\theta_0=0.3} \left[\sum_{i=1}^n x_i \leq k_2^* \right] \leq 0.05$$

$$k_2^* = 22$$

Therefore, the critical region of ϕ^* is $\{0, \dots, 22\}$

and the test can be formulated as

$$\phi^*(x) = \begin{cases} 0 & \sum_{i=1}^n x_i > 22 \\ 1 & \sum_{i=1}^n x_i \leq 22 \end{cases}$$

- (c) Determine the p -value of the test ϕ^* from the previous subexercise. Explain its interpretation in the context of the problem.

$$p\text{-value} = \inf \{ \alpha \in \mathcal{A} : T(x) \in C_\alpha \}$$

In this case, the p -value: $P_{\theta_0} \left\{ \sum_{i=1}^n x_i \leq 22 \right\} = 0.0478$

In this case, it means if we set $k_2^* = 22$, the probability that a true H_0 get rejected is 0.0478. Alternatively, we could set the α to be 0.0478, and the test is still valid at that significance.

3. Neyman-Pearson test for Poisson distributions [4 Points]

The number of claims reported to an insurance company during a year is Poisson-distributed with some parameter $\lambda > 0$. From previous years, the insurance company uses $\lambda = \lambda_0$ as a model parameter. The company notices that the number of claims has increased, and therefore wants to test

$$H_0 : \lambda = \lambda_0, \text{ against}$$

$$H_1 : \lambda = \lambda_1,$$

where $\lambda_1 > \lambda_0$.

(a) Find a Neyman-Pearson test ϕ^* at level $\alpha \in (0, 1)$ for this testing problem.

(a) $H_0 : \lambda = \lambda_0 \quad \text{against}$

$H_1 : \lambda = \lambda_1$

$$f(k) = \frac{\lambda^k e^{-\lambda}}{k!}$$

$$f_{\lambda_1, \lambda_0}(x) = \frac{f_{\lambda_1}(x_1, \dots, x_n)}{f_{\lambda_0}(x_1, \dots, x_n)} =$$

$$\frac{\frac{\lambda_1^{\sum x_i} e^{-n\lambda_1}}{x_1! \dots x_n!}}{\frac{\lambda_0^{\sum x_i} e^{-n\lambda_0}}{x_1! \dots x_n!}}$$

$$= \left(\frac{\lambda_1}{\lambda_0}\right)^{\sum x_i} \cdot e^{-n(\lambda_1 - \lambda_0)}$$

$$\phi^*(x) = \begin{cases} 1 & f_{\lambda_1, \lambda_0}(x) \geq c \\ 0 & f_{\lambda_1, \lambda_0}(x) < c \end{cases}$$

and $\mathbb{P}_{\lambda_0}[\phi^*(x) = 1] \leq \alpha$.

(b) Is the test from the previous subexercise a uniformly most powerful test for H_0 against

$$\tilde{H}_1: \lambda \in \{\lambda' ; \lambda' > \lambda_0\}?$$

The Neyman-Pearson test we got in last subexercise is only guaranteed to be the most powerful test for $H_0: \{\lambda_0\}$ against $H_1: \{\lambda_1\}$.

- (c) Suppose now that $\lambda_0 = 9000$ and the company observed 9876 claims. Decide whether H_0 is rejected at a level $\alpha = 0.05$.

Hint: Use that $\sum_{k=0}^{9155} \frac{(9000)^k}{k!} e^{-9000} \approx 0.9491$ and $\sum_{k=0}^{9156} \frac{(9000)^k}{k!} e^{-9000} \approx 0.9502$.

(c)

Here we could see the determinant factor is $\sum_{i=1}^k x_i$,

where, in this case, we only have the observation for 1 year.

$$\Rightarrow P_{\lambda_0} \{ \phi_0(x) \geq c \} \leq \alpha$$

$$\Leftrightarrow P_{\lambda_0} \{ X \geq k_2^* \} \leq \alpha = 0.05$$

$$\Rightarrow P_{\lambda_0} \{ X < k_2^* \} > 0.95$$

$$\Rightarrow \text{we should set } k_2^* = 9157$$

Clearly, $X = 9876$ clearly reject the hypothesis H_0

4. Testing for the variance in normal distributions

[4 Points]

Suppose that $X_1, \dots, X_n \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(\mu, \sigma^2)$ with known $\mu \in \mathbb{R}$ and unknown $\sigma^2 > 0$. Assume furthermore that $0 < \sigma_0^2 < \sigma_1^2$. Show that the test

$$\phi^*(\mathbf{X}) = \begin{cases} 1, & \text{if } \sum_{i=1}^n (X_i - \mu)^2 > \underline{\sigma_0^2 \chi_{n,1-\alpha}^2}, \\ 0, & \text{if } \sum_{i=1}^n (X_i - \mu)^2 \leq \underline{\sigma_0^2 \chi_{n,1-\alpha}^2}, \end{cases}$$

is the Neyman-Pearson test at level $\alpha \in (0, 1)$ for $H_0 : \sigma^2 = \sigma_0^2$ against $H_1 : \sigma^2 = \sigma_1^2$.

Proof:

$$L_{\sigma_0, \sigma_1} = \frac{\int_{\sigma_1^2} f_{\sigma_1^2}(x_1, \dots, x_n)}{\int_{\sigma_0^2} f_{\sigma_0^2}(x_1, \dots, x_n)} = \frac{\left(\frac{1}{\sigma_1 \sqrt{n}}\right)^n e^{-\frac{1}{2} \frac{\sum (x_i - \mu)^2}{\sigma_1^2}}}{\left(\frac{1}{\sigma_0 \sqrt{n}}\right)^n e^{-\frac{1}{2} \frac{\sum (x_i - \mu)^2}{\sigma_0^2}}}$$

$$= K(\sigma_1, \sigma_0) \cdot \exp\left(\frac{1}{2} \left(\frac{1}{\sigma_0^2} - \frac{1}{\sigma_1^2}\right) \sum_{i=1}^n (x_i - \mu)^2\right) > 0$$

Here we can see, L_{σ_0, σ_1} increases as $\sum_{i=1}^n (x_i - \mu)^2$ increases.

$$\text{Therefore: } L_{\sigma_0, \sigma_1} \geq c_\alpha^* \Leftrightarrow \sum_{i=1}^n (x_i - \mu)^2 \geq k_\alpha^*$$

It remains to decide k_α^*

$$\text{We want: } \mathbb{P}_{\sigma_0} \left(\sum_{i=1}^n (x_i - \mu)^2 \geq k_\alpha^* \right) = \alpha$$

$\Rightarrow k_\alpha^*$ should be the α quantile of $\sum_{i=1}^n (x_i - \mu)^2$

$$\sum_{i=1}^n (x_i - \mu)^2 \sim \sum_{i=1}^n N(0, \sigma_0^2)$$

$$\times \frac{1}{\sigma_0^2} \sum_{i=1}^n (x_i - \mu)^2 \sim \sum_{i=1}^n \mathcal{N}^2(0, 1) = \chi_n^2$$

Therefore,

$$\Rightarrow \mathbb{P} \left(\sum_{i=1}^n (x_i - \mu)^2 > k_d^* \right) = \alpha$$

$$\Leftrightarrow \mathbb{P} \left(\frac{\sum_{i=1}^n (x_i - \mu)^2}{\sigma_0^2} \leq \frac{k_d^*}{\sigma_0^2} \right) = 1 - \alpha$$

$$\mathbb{P} \left(\chi_n^2 \leq \frac{k_d^*}{\sigma_0^2} \right) = 1 - \alpha$$

$$\Rightarrow \frac{k_d^*}{\sigma_0^2} = \chi_n^2, 1 - \alpha \Rightarrow k_d^* = \sigma_0^2 \chi_n^2, 1 - \alpha$$

Hence, the UMP test is

$$\phi_{0X}^* = \begin{cases} 1 & \sum_{i=1}^n (x_i - \mu)^2 > \sigma_0^2 \chi_n^2, 1 - \alpha \\ 0 & \sum_{i=1}^n (x_i - \mu)^2 \leq \sigma_0^2 \chi_n^2, 1 - \alpha. \end{cases}$$