

Homework 3

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An ECON - 8070 Homework Assignment

October 9, 2024

10.1

Problem

Let X be distributed Poisson: $\pi(k) = \frac{\exp(-\theta)\theta^k}{k!}$ for nonnegative integer k and $\theta > 0$.

- Find the log-likelihood function $l_n(\theta)$.
- Find the MLE $\hat{\theta}$ for θ

Solution

- We can obtain log likelihood by first finding the likelihood function

$$L(\theta) = \prod_{i=1}^n \left[\frac{(\exp(-\theta))^{\theta^{k_i}}}{k_i!} \right]$$

Then, we take the natural log

$$\begin{aligned} l_n(\theta) &= \log(L(\theta)) = \sum_{i=1}^n [\log(\exp(-\theta)) + k_i \log(\theta) - \log(k_i!)] \\ &= \sum_{i=1}^n [-\theta + k_i \log(\theta) - \log(k_i!)] \\ &= -n\theta + \log(\theta) \sum_{i=1}^n k_i - \sum_{i=1}^n \log(k_i!) \end{aligned}$$

- Now we need to find $\hat{\theta} = \arg \max_{\theta \in \Theta} l_n(\theta)$. We can do this by taking the derivative of l_n w.r.t θ

$$\frac{dl_n(\theta)}{d\theta} = -n + \frac{1}{\theta} \sum_{i=1}^n k_i = 0$$

Then, we solve for $\hat{\theta}$

$$\begin{aligned} \theta n &= \sum_{i=1}^n k_i \\ \hat{\theta} &= \frac{1}{n} \sum_{i=1}^n k_i \end{aligned}$$

The MLE is the sample mean.

10.6

Problem

Let X be Bernoulli $\pi(X|p) = p^x(1-p)^{1-x}$.

- Calculate the information for p by taking the variance of the score.
- Calculate the information for p by taking the expectation of (minus) the second derivative. Did you obtain the same answer?

Solution

- Let's start with log-likelihood:

$$l(p) = x \log(p) + (1-x) \log(1-p)$$

To find the score, we differentiate w.r.t. p

$$S(p) = \frac{x}{p} - \frac{(1-x)}{(1-p)}$$

Next, we will find the variance such that $\text{var}[S(p)] = \mathbb{E}(S(p)^2)$

$$\mathbb{E}(s(p)^2) = \mathbb{E}\left(\frac{x^2}{p^2}\right) + \mathbb{E}\left(\frac{(1-x)^2}{(1-p)^2}\right) = \frac{1}{p} + \frac{1}{(1-p)}$$

$$\therefore \mathcal{I} = \frac{1}{p(1-p)}$$

(b) Second derivative

$$\frac{d^2 l(p)}{dp^2} = \frac{-x}{p^2} - \frac{(1-x)}{(1-p)^2}$$

Negative expectation

$$\begin{aligned} \mathcal{I}(p) &= -\mathbb{E}\left(\frac{d^2 l(p)}{dp^2}\right) = \frac{-x}{p^2} - \frac{(1-x)}{(1-p)^2} \\ &= \mathbb{E}\left(\frac{x}{p^2}\right) + \mathbb{E}\left(\frac{1-x}{(1-p)^2}\right) \\ &= \frac{p}{p^2} + \frac{(1-p)}{(1-p)^2} \\ &= \frac{1}{p} + \frac{1}{(1-p)} \\ &= \frac{1}{p(1-p)} \end{aligned}$$

11.3

Problem

A Bernoulli random variable X is

$$\begin{aligned} \mathbb{P}[X = 0] &= 1 - p \\ \mathbb{P}[X = 1] &= p \end{aligned}$$

- (a) Propose a moment estimator \hat{p} for p .
- (b) Find the variance of the asymptotic distribution of $\sqrt{n}(\hat{p} - p)$

Solution

(a) The first moment of a Bernoulli is the mean

$$\mathbb{E}[X] = p$$

The method of moments for p , here, is the sample mean

$$\hat{p} = \frac{1}{n} \sum_{i=1}^n X_i$$

(b) For a Bernoulli distribution $\text{var}(X) = p(1-p)$. We can use central limit theorem to deduce $\sqrt{n}(\hat{p} - p) \rightarrow N(0, \sigma^2)$. We can put these together since $\sigma^2 = \text{var}(X)$. So, the asymptotic distribution is $\sqrt{n}(\hat{p} - p) \rightarrow N(0, p(1-p))$

11.4

Problem

Propose a moment estimator $\hat{\lambda}$ for the parameter λ of a Poisson distribution.

Solution

A Poisson distribution has distribution $\mathbb{E}[X] = \lambda$. As seen in the Bernoulli case, the moment estimator for λ is the sample mean

$$\hat{\lambda} = \frac{1}{n} \sum_{i=1}^n X_i$$

13.3

Problem

Take the exponential model with parameter λ . We want a test for $\mathbb{H}_0 : \lambda = 1$ against $\mathbb{H}_1 : \lambda \neq 1$.

(a) Develop a test based on the sample mean \bar{X}_n .

Solution

For an exponential distribution, $\mathbb{E}[X] = \frac{1}{\lambda}$ and $\text{var}(X) = \frac{1}{\lambda^2}$. H_0 :

$$\lambda = 1 \Rightarrow \mathbb{E}[X] = 1 \text{ and } \text{var}(X) = 1$$

By central limit theorem,

$$\begin{aligned} \sqrt{n}(\bar{X}_n - \mathbb{E}[X]) &\rightarrow N(0, \text{var}(X)) \\ \sqrt{n}(\bar{X}_n - 1) &\rightarrow N(0, 1) \end{aligned}$$

Thus, we can use $Z = \sqrt{n}(\bar{X}_n - 1)$ due to its following the normal distribution. We would set the test to reject H_0 if $|Z| > z_{(\frac{\alpha}{2})}$ such that z is the $(1 - \frac{\alpha}{2})$ quantile of the standard normal.

13.5

Problem

Take the model $X \sim N(\mu, \sigma^2)$. Propose a test for $\mathbb{H}_0 : \mu = 1$ against $\mathbb{H}_1 : \mu \neq 1$.

Solution

There are two main options. If σ^2 is known, a Z test can be used. If not, a t-test should be used. The t-test would look like

$$t = \frac{\bar{X}_n - \mu_0}{\frac{s}{\sqrt{n}}}$$

The Z test would look like

$$Z = \frac{\bar{X}_n - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$

We would reject H_0 if $T > t$ such that t is the $\frac{1-\alpha}{2}$ quantile with $n - 1$ degrees of freedom. Similarly, for the Z test, we reject H_0 if $Z > z$ such that z is the $\frac{1-\alpha}{2}$.