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# STAT 245

## HOMEWORK 0

## 1. Moments of Poisson Distribution

Let  $X$  be a random variable with a Poisson distribution. Find  $E(X^4)$ .

One way to this using the formula

$$E(X^4) = \sum_k x^4 \Pr\{X = k\}.$$

Observe

Stirling numbers of the second kind

$$x^4 = x + 7x(x-1) + 6x(x-1)(x-2) + x(x-1)(x-2)(x-3).$$

We can derive each summation term individually.

$$\begin{aligned} E(X^4) &= \sum_x x^4 \Pr\{X = x\} \\ &= \sum_x x \Pr\{X = x\} + \sum_x 7x(x-1) \Pr\{X = x\} \\ &\quad + \sum_x 6x(x-1)(x-2) \Pr\{X = x\} \\ &\quad + \sum_x x(x-1)(x-2)(x-3) \Pr\{X = x\}. \end{aligned}$$

$$\sum_x x \Pr\{X = x\} = \lambda$$

$$\begin{aligned} \sum_x 7x(x-1) \Pr\{X = x\} &= \sum_x 7x(x-1) e^{-\lambda} \frac{\lambda^x}{x!} \\ &= 7e^{-\lambda} \sum_x x(x-1) \frac{\lambda^x}{x(x-1)(x-2)!} \\ &= 7e^{-\lambda} \lambda^2 \sum_x \frac{\lambda^{x-2}}{(x-2)!} \\ &= 7e^{-\lambda} \lambda^2 e^{\lambda} \\ &= 7\lambda^2 \end{aligned}$$

Using a similar method, find

$$\sum_x 6x(x-1)(x-2) \Pr\{X = x\} = 6\lambda^3$$

and

$$\sum_x x(x-1)(x-2)(x-3) \Pr\{X = x\} = \lambda^4.$$

Combining the results

$$E(X^4) = \lambda + 7\lambda^2 + 6\lambda^3 + \lambda^4.$$

## 2. Poisson- and $\chi^2$ -tails

For  $\lambda > 0$ , let  $X_\lambda$  be a discrete random variable with a Poisson distribution with expected value  $\lambda$ . For (integer)  $d \in \mathbb{N}$ , let  $Y_d$  be a continuous random variable with a  $\chi^2$ -distribution with  $d$  degrees of freedom. In other words, the distribution of  $Y_d$  has the probability density function

$$f_{\chi_d^2}(y) = \frac{1}{2^{\frac{d}{2}} \Gamma(\frac{d}{2})} y^{\frac{d}{2}-1} e^{-\frac{y}{2}}, / y \geq 0$$

where  $\Gamma(\cdot)$  is the Gamma-function which satisfies  $\Gamma(n) = (n-1)!$  for  $n \in \mathbb{N}$ . Show that for all  $\lambda > 0$  and all (integer)  $c \in \mathbb{N}$ ,

$$\Pr\{X_\lambda \geq c+1\} = \Pr\{0 \leq Y_{2(c+1)} \leq 2\lambda\} = \int_0^{2\lambda} f_{\chi_{2(c+1)}^2}(y) dy.$$

First observe

$$\Pr\{X_\lambda \geq c+1\} = 1 - \sum_{k=0}^c e^{-\lambda} \frac{\lambda^k}{k!}$$

whose derivative is just Poisson density

$$\frac{d}{d\lambda} \Pr\{X_\lambda \geq c+1\} = \Pr\{X = c\} = e^{-\lambda} \frac{\lambda^c}{c!}.$$

Next observe

$$\Pr\{0 \leq Y_{2(c+1)} \leq 2\lambda\} = \int_0^{2\lambda} \frac{1}{2^{c+1} c!} y^c e^{-\frac{y}{2}} dy$$

whose derivative is

$$\begin{aligned} \frac{\partial}{\partial \lambda} \Pr\{0 \leq Y_{2(c+1)} \leq 2\lambda\} &= \frac{\partial}{\partial \lambda} \frac{\Gamma(c+1) - \Gamma(c+1, \lambda)}{c!} \\ &= e^{-\lambda} \frac{\lambda^c}{c!} \end{aligned}$$

Hence  $\frac{\partial}{\partial \lambda} \Pr\{0 \leq Y_{2(c+1)} \leq 2\lambda\} = \frac{\partial}{\partial \lambda} \Pr\{X_\lambda \geq c+1\}$  which implies

$$\Pr\{X_\lambda \geq c+1\} = \Pr\{0 \leq Y_{2(c+1)} \leq 2\lambda\}.$$

## 3. Approximation to Binomial probabilities

Let  $X$  be distributed according to a Binomial( $n, p$ ) distribution. We are interested in the probability  $\Pr(X = k)$  for

1.  $n = 7, p = 0.3, k = 3$ ;

The binomial probability

$$\Pr\{X = 3\} = \binom{7}{3} .3^3 (1 - .3)^4 = .2268.$$

The Normal Distribution approximation with  $E(X) = np$  and  $\text{Var}(X) = np(1 - p)$ .

P.187 in Rice has a nice explanation if this

$$\begin{aligned} \Pr\{X \geq 3\} &= \Pr\left\{\frac{X - 2.1}{\sqrt{1.47}} \leq \frac{3 - 2.1}{\sqrt{1.47}}\right\} \\ &\approx 1 - \Phi(0.742) \\ &= .249 \end{aligned}$$

The Poisson approximation  $\lambda = np$ .

$$\Pr\{X = 3\} = e^{-2.1} \frac{2.1^3}{3!} = .189$$

2.  $n = 40$ ,  $p = 0.4$ ,  $k = 11$ ;

Binomial = .0357, Normal approximation = 0.035, and Poisson = 0.0495.

3.  $n = 400$ ,  $p = .0025$ ,  $k = 2$ ;

Binomial = .18417, Normal approximation = .2419, and Poisson = .1839.

The Poisson is a good approximation when  $p$  is small and  $n$  is large. The Normal is a good approximation when  $n$  is large and  $p$  is close to  $\frac{1}{2}$ , i.e. the binomial distribution is symmetric.

#### 4. Conditional distributions in Poisson process

Let  $(X_t)_{t \geq 0}$  be a Poisson process, and let

$$T_1 = \min\{t > 0 : X_t \geq 1\}$$

be the time to the first event.

- Find the conditional distribution of  $X_s$  given  $X_t = n$  for fixed time points  $t > s > 0$  and integer  $n \in \mathbb{N}$ .

The conditional distribution is given

$$\Pr(X_s | X_t = n) = \frac{\Pr(X_s \cap X_t)}{\Pr(X_t = n)}.$$

Note that

$$\Pr(X_s \cap X_t) = \Pr(X_s = x \cap X_t - X_s = n - x)$$

where  $X_s$  and  $X_t - X_s$  are independent of each other.

Then

$$\begin{aligned}
 \Pr(X_s \mid X_t = n) &= \frac{e^{-\lambda s} \frac{(\lambda s)^x}{x!} \cdot e^{-\lambda(t-s)} \frac{[\lambda(t-s)]^{n-x}}{(n-x)!}}{e^{-\lambda t} \frac{(\lambda t)^n}{n!}} \\
 &= \frac{n!}{x!(n-x)!} \cdot \frac{e^{-\lambda s} e^{-\lambda(t-s)} (\lambda s)^x [(t-s)\lambda]^{n-x}}{e^{-\lambda t} (\lambda t)^n} \\
 &= \binom{n}{x} \frac{s^x (t-s)^{n-x}}{t^x \cdot t^{n-x}} \\
 &= \binom{n}{x} \left(\frac{s}{t}\right)^x \left(1 - \frac{s}{t}\right)^{n-x}
 \end{aligned}$$

2. Show that the conditional distribution of  $T_1$  given  $X_t = 1$  is the uniform distribution on the interval  $(0, t]$ .

Consider  $\Pr\{T_1 > s \mid X_t = 1\}$  for  $0 < s < t$ .

$$\begin{aligned}
 \Pr\{T_1 > s \mid X_t = 1\} &= \frac{\Pr\{0 \text{ events in } (0, s] \cap 1 \text{ event in } (s, t]\}}{\Pr\{X_t = 1\}} \\
 &= \frac{e^{-\lambda s} \cdot e^{-\lambda(t-s)} \lambda(t-s)}{e^{-\lambda t} \lambda t} \\
 &= \frac{t-s}{t}
 \end{aligned}$$

Then

$$\Pr\{T_1 \leq s \mid X_t = 1\} = 1 - \Pr\{T_1 > s \mid X_t = 1\} = \frac{s}{t}.$$

Taking the derivative of the above equation results in

$$\Pr\{T_1 = s \mid X_t = 1\} = \frac{1}{t}.$$

### 5. Data from Poisson process

A detector counts the number of particles emitted from a radioactive source over the course of 10-second intervals. For 180 such 10-second intervals, the following counts were observed:

Count	# Intervals
0	23
1	77
2	34
3	26
4	13
5	7

This table states, for example, that in 34 of the 10-second intervals a count of 2 was recorded. Sometimes, however, the detector did not function properly and recorded counts over intervals of length 20 seconds. This happened 20 times and recorded counts are

Count	# Intervals
0	2
1	4
2	9
3	5

Assume a Poisson process model for the particle emission process. Let  $\lambda > 0$  (time unit = 1 sec.) be the unknown rate of the Poisson process.

1. Formulate an appropriate likelihood function for the described scenario and derive the maximum likelihood estimator for  $\hat{\lambda}$  of the rate  $\lambda$ . Compute  $\hat{\lambda}$  for the above data.

First observe

$$\Pr\{X = x\} = e^{-\lambda} \frac{\lambda^x}{x!}$$

then

$$\text{lik}(\lambda) = \prod_{i=1}^n e^{-\lambda} \frac{\lambda^{x_i}}{x_i!}$$

and the log is

$$\begin{aligned} l(\lambda) &= \sum_{i=1}^n (-\lambda + X_i \log \lambda - \log(X_i!)) \\ &= -n\lambda + \log \lambda \sum_{i=1}^n X_i - \sum_{i=1}^n \log X_i!. \end{aligned}$$

Setting  $l'(\lambda)$  to zero gives

$$l'(\lambda) = -n + \frac{1}{\lambda} \sum_{i=1}^n X_i = 0$$

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

That data observes 347 occurrences over 2200 seconds which can be use to compute

$$\hat{\lambda} = \frac{347}{2200} = .1577.$$

The above method works, but can be consicely stated and found by deriving  $\hat{\lambda}$  beginging with

$$\text{Lik}(\theta) = \prod_{i=1}^{180} e^{-10\lambda} \frac{(10\lambda)^{y_i}}{y_i!} \cdot \prod_{j=1}^{20} e^{-20\lambda} \frac{(20\lambda)^{z_j}}{z_j!}$$

2. What approximation to the distribution of  $\hat{\lambda}$  does the central limit theorem suggest?

Explanation of this in Rice p.262

Let

$$S = X_1 + X_2$$

and  $\hat{\lambda} = \frac{s}{n}$  is a random variable.

$$\Pr\{\hat{\lambda} = v\} = \Pr\{s = nv\} = e^{-n\lambda_0} \frac{(n\lambda_0)^{nv}}{(nv)!}$$

for  $v$  such that  $nv$  is a nonnegative integer.

Since  $S \sim \text{Pois}(n\lambda_0)$

$$E(\hat{\lambda}) = \frac{1}{n} E(S) = \lambda_0$$

and

$$\text{Var}(\hat{\lambda}) = \frac{1}{n^2} \text{Var}(S) = \frac{\lambda_0}{n}.$$

Since  $E(\hat{\lambda}) = \lambda_0$ ,  $\hat{\lambda}$  is unbiased and centered at  $\lambda_0$  with standard error

$$\sigma_{\hat{\lambda}} = \sqrt{\frac{\lambda_0}{n}}.$$

The standard error can be estimated

$$s_{\hat{\lambda}} = \sqrt{\frac{\hat{\lambda}}{n}}$$

therefore  $\hat{\lambda} \sim N(\lambda_0, s_{\hat{\lambda}})$ .