Statistics for Computer Science

Assignment 2

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Exercise 3

1. Write down the formula for likelihood function of Poisson distribution.

The formula for the Poisson probability mass function is:

$$P(x,\lambda) = \frac{e^{-\lambda}\lambda^x}{x!}$$

In Poisson distribution, the parameter of interest is λ . Having sequence of X_n , the probability of observing the sequence X_n will be the product of probabilities of each of them.

Therefore, the kernel of likelihood function of Poisson distribution is:

$$L(\lambda|X) = \prod_{i=1}^{N} \frac{\lambda^{X_i} e^{-\lambda}}{X_i!}$$

2. Write down the formula for log-likelihood function of Poisson distribution.

The formula for log-likelihood function of Poisson distribution is obtained by using natural logarithm on the likelihood function of Poisson distribution.

Therefore, the kernel of log-likelihood function of Poisson distribution is:

$$l(\lambda|X) = \ln\left(\prod_{i=1}^{N} \frac{\lambda^{X_i} e^{-\lambda}}{X_i!}\right)$$
$$l(\lambda|X) = \sum_{i=1}^{N} X_i \ln \lambda - N\lambda$$

3. Write down the likelihood equation and work out the exact formula for $\hat{\lambda}$.

$$\begin{split} L(\lambda) &= \prod_{i=1}^N \frac{e^{-\lambda} \lambda^{x_i}}{x_i!} = e^{-N\lambda} \frac{\lambda^{\sum_1^N x_i}}{\prod_{i=1}^N x_i} \\ lnL(\lambda) &= -N\lambda + \sum_1^N x_i ln(\lambda) - ln\left(\prod_{i=1}^N x_i\right) \\ \frac{dlnL(\lambda)}{dp} &= -N + \sum_1^N x_i \frac{1}{\lambda} \\ \hat{\lambda} &= \frac{\sum_{i=1}^N x_i}{N} \end{split}$$

4. Create your own R-function for calculating the value of log-likelihood function of Poisson distribution.

The value of log-likelihood function of Poisson distribution is -7612.856.

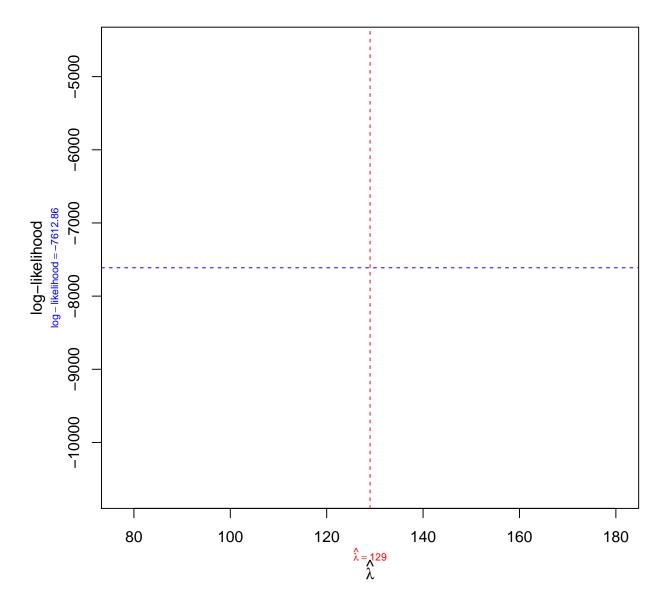
5. Using function optimize() find $\hat{\lambda}$. Compare it to the estimate you get from the exact formula.

```
15 #this function will take 2 parameters x and n
16 #x is the sequence of observed values
17 #n is number of obervation
18 #finally it will return lambda hat which is the mean
19 lambda.hat <- function(x, n){
20
     return(sum(x)/n)
21 }
22 ans.lambda.hat <- lambda.hat(x, n)
23
24 #using optimize function to obtain lambda hat
25 #the optimize function take the log-likelihood of Poisson distribution
      to optimize with the given interval
26 #optimize function will return maximum and objective value
27 #in this case we interest in the value of maximum
28 lambda.hat.est <- optimize(f = poi.log.likelihood, interval = c(88, 129)
      , maximum = T, x = x, n=n) maximum
```

The exact value of $\hat{\lambda}$ is 104.35 and the estimate value of $\hat{\lambda}$ is 128.9999.

6. Plot the log-likelihood function, highlight the maximum and denote the maximum likelihood estimate in plot margin.

log-likelihood function of Poisson Distribution



Exercise 4

1. Write down the null and the alternative hypotheses in mathematical form.

Null hypothesis— $H_0: \rho = 0$

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Alternative hypothesis— $H_1: \rho \neq 0$

2. Calculate the value of test statistic.

```
39 setwd(getwd())
40 body <- read.table(file = 'body-measurements.txt', header = T)
41 body <- na.omit(body)
42
43 body.f.height <- body[body$sex == 'f', 'body.H']
44 body.f.neck <- body[body$sex == 'f', 'neck.C']
45
46 n.x <- length(body.f.height)
47 n.y <- length(body.f.neck)
48 mu.x <- mean(body.f.height)
49 mu.y <- mean(body.f.neck)
50 sigma2.x <- sd(body.f.height)
51 sigma2.y <- sd(body.f.neck)</pre>
     correlation <- cor(body.f.height, body.f.neck, method = c("pearson", "</pre>
52
        kendall", "spearman"))
53
54
  # Test Statistic Comparing Two Population Means: 978.969544271
55
56 z <-(1/2)*log((1+r)/(1-r))
```

Error in eval(expr, envir, enclos): object 'r' not found

Exercise 5

1. Write down the null and the alternative hypotheses in mathematical form.

```
Null hypothesis— H_0: p_1 - p_2 = 0
```

Alternative hypothesis— $H_1: p_1 - p_2 \neq 0$

2. Calculate the value of test statistic.

```
58  n1  <- 200
59  x1  <- 32
60  p1  <- x1 / n1
61
62  n2  <- 230
63  x2  <- 21
64  p2  <- x2 /n2
65
66  p0  <- 0.5
67
68  alpha  <- 0.05
```