

Exercises I5-I6

#statistics

Example 5.1

$$1 = \sum_{x=1}^{\infty} c_{\theta} \theta^x = c_{\theta} \left(\frac{1}{1-\theta} - 1 \right)$$

and solving the equation we get $f(x; \theta) = (1 - \theta)\theta^{x-1}$ a geometric rv with parameter $1 - \theta$.

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(a) Wald confidence interval for θ of level $1 - \alpha$, $\alpha = 0.9$ $\bar{x} = 10$ and $n = 30$.

$$|W| = \left| \frac{\hat{\theta} - \theta_0}{\widehat{se}} \right|$$

$$L(\theta) = \frac{(1-\theta)^n}{\theta^n} \theta^{\sum_{i=1}^n x_i} = (1 - \theta)^n \theta^{n(\bar{x}-1)}$$

$$\ell(\theta) = n \log(1 - \theta) + n(\bar{x} - 1) \log \theta$$

$$J(\theta) = \frac{n(\bar{x}-1)}{\theta^2} + \frac{n}{(1-\theta)^2}$$

$$\hat{\theta} = \frac{\bar{x}-1}{\bar{x}} = 0.9$$

$$J(\hat{\theta}) = 3333.3333$$

$$\widehat{se} = 1/\sqrt{\hat{J}} = 0.01732$$

$$R = \{\mathbf{X} : |W| \geq z_{1-\frac{\alpha}{2}}\} = \{\mathbf{X} : \theta_0 \leq \hat{\theta} - z_{1-\frac{\alpha}{2}} \widehat{se}\} \vee \theta_0 \geq \hat{\theta} + z_{1-\frac{\alpha}{2}} \widehat{se}\}$$

$$R^c = \{\mathbf{X} : \hat{\theta} - z_{1-\frac{\alpha}{2}} \widehat{se} \leq \theta_0 \leq \hat{\theta} + z_{1-\frac{\alpha}{2}} \widehat{se}\}$$

$$z_{1-\frac{0.1}{2}} = 1.644854$$

confidence interval of level $1 - \alpha = 0.9$:

$$CI = [0.871511, 0.928489]$$

TODO CONTROLLA

(b) I would reject H_0 since $\theta_0 = 0.5$ is outside the confidence interval.

(c) The likelihood ratio test statistic is $\lambda(\mathbf{x}) = \frac{L(\theta_0)}{L(\hat{\theta})} = \frac{L(\theta_0)}{4.42 \cdot 10^{-43}}$

$$R_\alpha(\theta_0) = \{\mathbf{x} : -2 \log(\lambda(\mathbf{x})) > \chi_{1,1-\alpha}^2\}$$

We can compute $-2 \log(\lambda(\mathbf{x}))$ and $\chi_{1,1-\alpha}^2$ and plot them, then we take the values of θ for which the sample is in the acceptance region, this is done with the following R script:

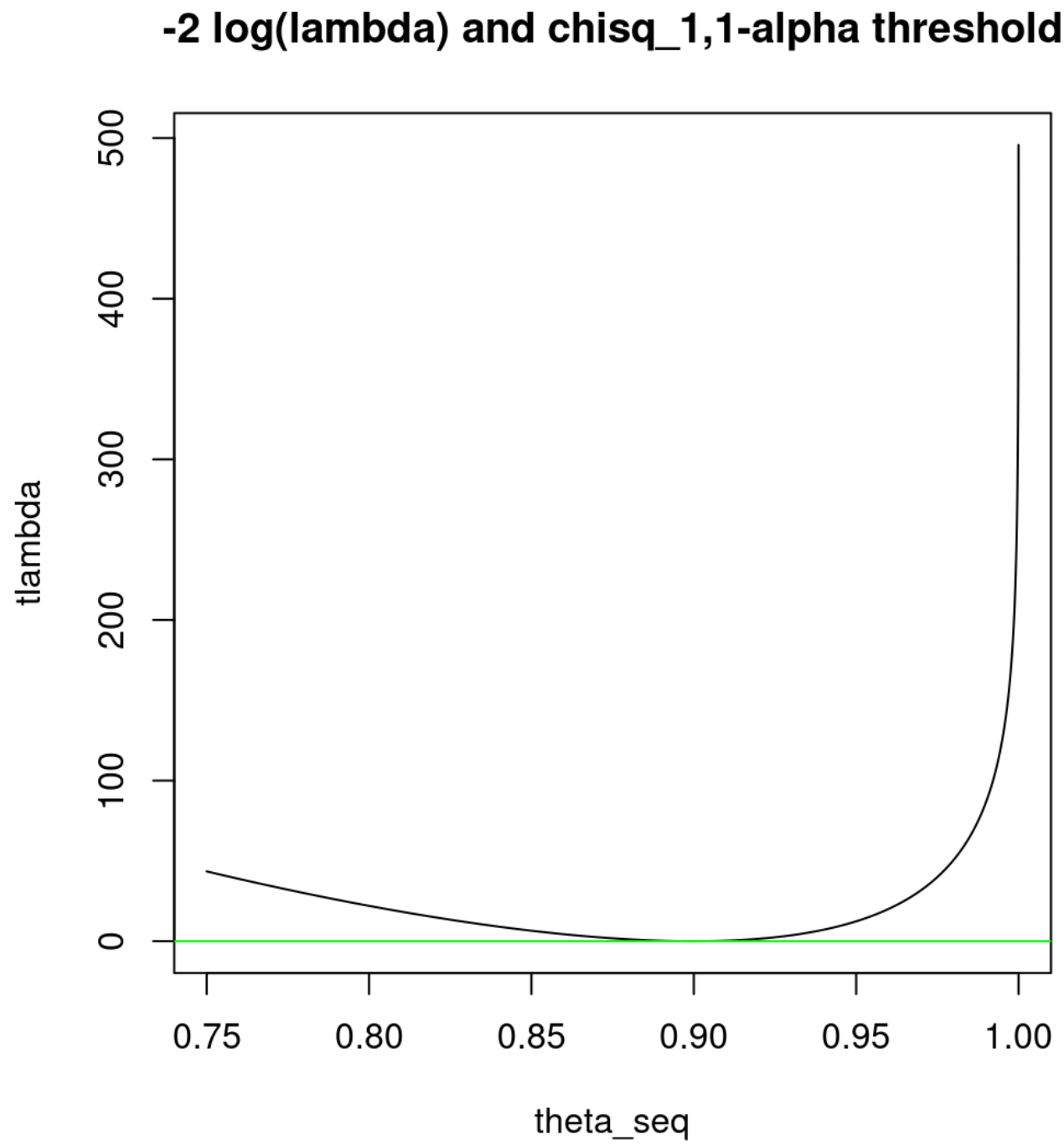
```
ell_f <-function(theta){  
  n=30  
  bar_x = 10  
  n*log(1-theta)+n*(bar_x-1)*log(theta)  
}  
theta_seq = seq(0.75,1,0.00001)  
  
alpha = 0.9  
threshold = qchisq(p = 1-alpha,df = 1)  
MLE = 0.9  
tlambda = -2*ell_f(theta_seq)+2*ell_f(MLE)  
plot(theta_seq,tlambda,main = '-2 log(lambda) and  
chisq_1,1-alpha threshold',lwd=1,lty=1,type = 'l')  
abline(h=threshold,col='green')
```

```
confidence_interval = theta_seq[tlambda<threshold]
sprintf('[ %f , %f
]',confidence_interval[1],confidence_interval[length
h(confidence_interval)])
```

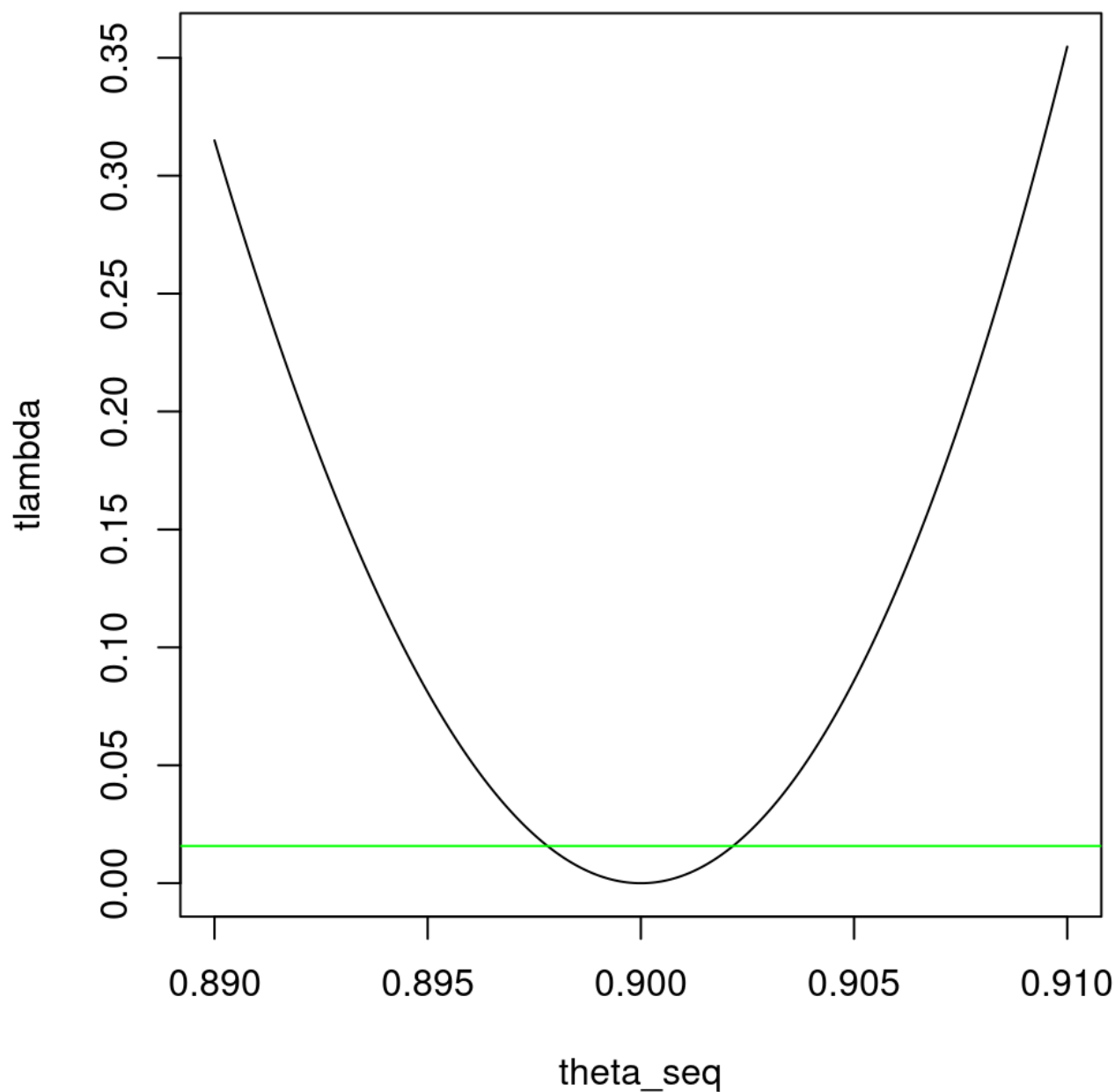
that outputs:

```
"[ 0.897810 , 0.902160 ]"
```

which is the confidence interval for $\alpha = 0.9$.



-2 log(lambda) and chisq_1,1-alpha threshold



(d) p-value:

$$\text{p-value} = P_{\theta}(-2 \log(\lambda) \geq -2 \log(\lambda_{obs}))$$

$$\lambda_{obs} = 1$$

Take the log on both sides and multiply by -2 .

$$\text{p-value} = P_{\theta}(\chi_1^2 \geq 0) = 1.$$

TODO controlla

Example 5.2

Let $X_i \sim \text{Poi}(\theta_1), i = 1, \dots, m$ and $Y_j \sim \text{Poi}(\theta_2), j = 1, \dots, n$, with X_i, Y_j being independent for all i, j .

(a) Log-likelihood ratio test for $H_0 : \theta_1 = \theta_2$ against $H_1 : \theta_1 \neq \theta_2$ at the level α . $\bar{y} = 6, \bar{x} = 2, m = 15, n = 10$, compute the test and get the p-value.

$$T_n = \frac{\bar{X} - \bar{Y}}{\sqrt{\frac{S_x^2}{m} + \frac{S_y^2}{n}}}$$

And we reject H_0 if $|T_n| > t_{\nu, 1 - \frac{\alpha}{2}}$, with:

$$\nu = \frac{\left(\frac{S_x^2}{m} + \frac{S_y^2}{n}\right)^2}{\frac{S_x^4}{m^2(m-1)} + \frac{S_y^4}{n^2(n-1)}}$$

TODO come si fa senza sample variance?????

Example 5.3

$X_i \sim \mathcal{N}(\mu, \sigma^2), i = 1, \dots, n$ both parameters are unknown

(a) $H_0 : \mu = 1$ against $H_1 : \mu \neq 1$ and observed sample with $\bar{x} = 2.1$ and $s^2 = 1.2$ determine n needed for $\beta(2) = 0.01$, (since $\mu = 2$ we are under H_1)

$$\beta(2) = P_{\mu=2}(\mathbf{x} \notin R)$$

if R is defined with a LRT we are in the case of a t-test,

$$R = \{\mathbf{x} : \left| \frac{\sqrt{n}(\bar{x} - \mu_0)}{s} \right| \geq t_{n-1, 1-\frac{\alpha}{2}}\}$$

$$\beta(2) = 1 - P_{\mu=2}(\text{reject } H_0) = P_{\mu=2}(\mathbf{x} \in R) = 0.01$$

$$\bar{x} \sim \mathcal{N}(\mu, \frac{\sigma^2}{n})$$

$$T_n \sim \mathcal{N}(\frac{\sqrt{n}}{s}, \frac{\sigma^2}{s^2})$$

$$\beta(2) = 1 - 2\Phi\left(\left(t_{n-1, 1-\frac{\alpha}{2}} - \frac{\sqrt{n}}{s}\right) \frac{s}{\sigma}\right)$$

TODO come lo calcolo senza sapere sigma?????

Example 5.4

X_1, \dots, X_n iid random sample from $\text{Unif}(0, \theta)$, $\theta > 0$. Construct a $1 - \alpha$ confidence interval for θ .

We use a LRT, we have that $L(\theta) = \frac{1}{\theta^n}$ if $x_{(n)} \leq \theta$ and 0 otherwise.

$$R_\alpha(\theta_0) = \{\mathbf{x} : -2 \log(\lambda(\mathbf{x})) > \chi_{1, 1-\alpha}^2\}$$

$$\hat{\theta} = x_{(n)}$$

$$\lambda(\mathbf{x}) = x_{(n)}^n \cdot \frac{1}{\theta^n}$$

$$R_\alpha(\theta_0) = \{\mathbf{x} : e^{-2 \log(\lambda(\mathbf{x}))} < e^{\chi_{1, 1-\alpha}^2}\}$$