

Module 5 Homework

1. (20 points)

A random sample of size 6 from the $\exp(\lambda)$ distribution results in observations: 1.433, 0.524, 0.384, 4.515, 1.852, 0.429. Find the MLE on this data set in two ways:

a) by numerical optimization of the likelihood (please include R code) and

Answer)

```
\label{lik-function} $$\lim <-function(lam) \ prod(dexp(c(1.433, 0.524, 0.384, 4.515, 1.852, 0.429), rate=lam))$$ $$nlik<-function(lam) -lik(lam)$$ $$optim(par=1, nlik)$par
```

Output)

[1] 0.6566895

b) by the analytic formula.

Answer)

n<- 6

Observations<- sum(1.433, 0.524, 0.384, 4.515, 1.852, 0.429) analytic <- n/Observations

analytic

Output)

[1] 0.6566707

2. (15 points)

A random sample X_1, X_2, \dots, X_{75} follows chi-square distribution with **m** degree of freedom, has sample mean $\bar{X} = 98.6$ and sample standard deviation s = 9.4. (a) Find the point estimator of **m** using the method of moments. 2) (a) is (very) easy.

Answer)

- a)
 #In chi-square Distribution the population mean(m) is equal to sample mean(X)
 m=98.6
- **(b)** Find a one-sided 90% lower confidence interval of **m**.

Answer)

Output)

[1] 100.013

Please provide the formulas and the derivations together with your numerical answer.

Problem 3 (35 points)

On the Golub et al. (1999) data set, analyze the Zyxin gene expression data separately for the ALL and AML groups.

(a) Find the bootstrap 95% CIs for the mean and for the variance of the gene expression in each group separately.

Answer)

```
#mean and variance for ALL
```

```
data(golub, package = "multtest")

ZyxinALL<-golub[2124,1:27]
n<-length(ZyxinALL)
nboot<-1000
boot.xbarmALL<-rep(NA,nboot)
boot.xbarvALL<-rep(NA,nboot)
for(i in 1:nboot) {
   data.star<-ZyxinALL[sample(1:n,replace=TRUE)]
   boot.xbarmALL[i]<-mean(data.star)
   boot.xbarvALL[i]<-var(data.star)
}
quantile(boot.xbarmALL,c(0.025,0.975))
quantile(boot.xbarvALL,c(0.025,0.975))</pre>
```

#mean and variance for AML

```
ZyxinAML<-golub[2124,28:38]
n<-length(ZyxinAML)
nboot<-1000
boot.xbarmAML<-rep(NA,nboot)
boot.xbarvAML<-rep(NA,nboot)
for(i in 1:nboot) {
   data.star<-ZyxinAML[sample(1:n,replace=TRUE)]
   boot.xbarmAML[i]<-mean(data.star)
   boot.xbarvAML[i]<-var(data.star)
}
quantile(boot.xbarmAML,c(0.025,0.975))
quantile(boot.xbarvAML,c(0.025,0.975))
```

```
> quantile(boot.xbarmAML,c(0.025,0.975))
    2.5%    97.5%
1.384754  1.797642
> quantile(boot.xbarvAML,c(0.025,0.975))
    2.5%    97.5%
0.05193054  0.20493144
```

(b) Find the parametric 95% CIs for the mean and for the variance of the gene expression in each group separately. (You need to choose the appropriate approximate formula to use: z-interval, t-interval or chi-square interval.)

Answer)

parametric mean for ALL and AML

```
gol.fac <- factor(golub.cl,levels=0:1,labels=c("ALL","AML"))
xmALL <- golub[2124, gol.fac=="ALL"]
nmALL<-length(xmALL)
ci.mALL <- mean(xmALL)+qt(c(0.025,0.957),df=nmALL-
1)*sd(xmALL)/sqrt(nmALL)

gol.fac <- factor(golub.cl,levels=0:1,labels=c("ALL","AML"))
xmAML <- golub[2124, gol.fac=="AML"]
nmAML<-length(xMAML)
ci.mAML <- mean(xmAML)+qt(c(0.025,0.957),df=nmAML-
1)*sd(xmAML)/sqrt(nmAML)
print(ci.mALL)
print(ci.mAML)
```

```
> print(ci.mALL)
[1] -0.58073875 -0.04653493
> print(ci.mAML)
[1] 1.339698 1.797752
```

Answer)

parametric varaiance for ALL and AML

```
gol.fac <- factor(golub.cl,levels=0:1,labels=c("ALL","AML"))
xvALL <- golub[2124, gol.fac=="ALL"]
nvALL<-length(xvALL) #sample size n
ci.vALL <- var(xvALL)+qt(c(0.025,0.957),df=nvALL-
1)*sd(xvALL)/sqrt(nvALL)

gol.fac <- factor(golub.cl,levels=0:1,labels=c("ALL","AML"))
xvAML <- golub[2124, gol.fac=="AML"]
nvAML<-length(xvAML) #sample size n
ci.vAML <- var(xvAML)+qt(c(0.025,0.957),df=nvAML-
1)*sd(x)/sqrt(nvAML)
print(ci.vALL)
print(ci.vALL)
```

```
> print(ci.vALL)
[1] 0.2365522 0.7707560
> print(ci.vAML)
[1] -0.1118260 0.3462277
```



(c) Find the bootstrap 95% CI for the <u>median</u> gene expression in both groups separately.

Answer)

#median for ALL

```
ZyxinmdALL<-golub[2124,1:27]
n<-length(ZyxinmdALL)
nboot<-1000
boot.xbarmdALL <- rep(NA, nboot)
for (i in 1:nboot) {
   data.star <- ZyxinmdALL[sample(1:n,replace=TRUE)]
   boot.xbarmdALL[i]<-median(data.star)
}
quantile(boot.xbarmdALL,c(0.025,0.975))
```

Output)

Answer)

median for AML

```
ZyxinmdAML<-golub[2124,28:38]
n<-length(ZyxinmdAML)
nboot<-1000
boot.xbarmdAML <- rep(NA, nboot)
for (i in 1:nboot) {
   data.star <- ZyxinmdAML[sample(1:n,replace=TRUE)]
   boot.xbarmdAML[i]<-median(data.star)
```

quantile(boot.xbarmdAML,c(0.025,0.975))

Output)

```
> quantile(boot.xbarmdAML,c(0.025,0.975))
    2.5%    97.5%
1.22814  1.82829
```

Please provide numerical answers for each part. Please also submit your R codes used for the calculations (the R code should be clearly labeled and separated for each part).

4. (30 points)

For a random sample of 50 observations from Poisson distribution, we have two ways to construct a 90% CI for the parameter λ .

- (1) Since the Poisson mean is λ , we can use the interval for the sample mean $(\bar{X} + t_{0.05,49} \sqrt{\frac{\bar{X}}{50}}, \bar{X} + t_{0.95,49} \sqrt{\frac{\bar{X}}{50}})$.
- (2) Since the Poisson variance is also λ , we can use the interval for the sample variance directly: $(\frac{49s^2}{\chi_{0.95.49}^2}, \frac{49s^2}{\chi_{0.05.49}^2})$.
 - (a) Write a R-script to conduct a Monte Carlo study for the coverage probabilities of the two CIs. That is, to generate nsim=1000 such data sets from the Poisson distribution. Check the proportion of the CIs that contains the true parameter λ.

Answer)

```
nsim <- 1000
MCsim<- function(nsim, lambda) {
  cov1<-cov2<-rep(NA,nsim) # create empty matrices to store data
  for (i in 1:nsim) {
    x<- rpois(50, lambda) # The question says Poisson
    xbar<- mean(x) #find mean of x
```

```
Xsd<- sd(x)# find sd of x CI1<-c(xbar+(qt(0.05, 49)*sqrt(xbar/50)), xbar+qt(0.95, 49)*sqrt(xbar/50)) #use the formula for conf interval for mean given in the question CI2<-c(49*(Xsd^2)/qchisq(0.95, 49), 49*(Xsd^2)/qchisq(0.05, 49)) cov1[i]<-(CI1[1]<lambda)&(lambda<CI1[2]) cov2[i]<-(CI2[1]<lambda)&(lambda<CI2[2]) } print(paste("When lambda=", lambda, ": coverage for first CI is", mean(cov1), ", coverage for second CI is", mean(cov2), ".")) # Just to keep your output presentable }
```

(b) Run the Monte Carlo simulation for nsim=1000 runs, at three different parameter values: λ =0.1, λ =1 and λ =10. Report the coverage probabilities of these two CIs at each of the three parameter values.

Answer)

```
MCsim(1000, 0.1)
MCsim(1000, 1)
MCsim(1000, 10)
```

```
> MCsim(1000, 0.1)
[1] "When lambda= 0.1 : coverage for first CI is 0.867 , c
overage for second CI is 0.552 ."
> MCsim(1000, 1)
[1] "When lambda= 1 : coverage for first CI is 0.908 , cov
erage for second CI is 0.849 ."
> MCsim(1000, 10)
[1] "When lambda= 10 : coverage for first CI is 0.906 , co
verage for second CI is 0.899 ."
```