
```
# Q1 consistent estimator : c(mu,1) t1=mean,t2=median

# Que. Based on the random samples generated from cauchy( $\mu$ , 1) check the
consistency of the estimators T1 = X and T2= sample median for the
parameter  $\mu$  . Construct the following table.

#           Sample size (n), P(|T1-  $\mu$ |<0.1), P(|T2-  $\mu$ |<0.1)


# Program
n=5
mu=1.3
sigma=1
eps=.1
x=rcauchy(n*n,mu,sigma)
data=matrix(x,n,n)
t1=apply(data,1,mean)
t2=apply(data,1,median)
prob_1=mean(abs(t1-mu)<eps)
prob_2=mean(abs(t2-mu)<eps)


#simulation study:
rm(list = ls(all=TRUE))
n=c(10,100,1000)
mu=2.1
sd=1
eps=.1
est_t1_prob=0
est_t2_prob=0
for (i in 1:length(n))
{
  x=rcauchy(n[i]*n[i],mu,sd)
  a=matrix(x,n[i],n[i])
  t1=apply(a,1,mean)
  t2=apply(a,1,median)
  est_t1_prob[i]=mean(abs(t1-mu)<eps)
```

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    est_t2_prob[i]=mean(abs(t2-mu)<eps)
  }
cbind(n,est_t1_prob,est_t2_prob)

##          n est_t1_prob est_t2_prob
## [1,]    10          0.100          0.400
## [2,]   100          0.070          0.450
## [3,]  1000          0.057          0.964

cat("Estimated prob of consistent estimator corresponding to mean",est_t1_p
rob,"\n")

## Estimated prob of consistent estimator corresponding to mean 0.1 0.07 0.
057

cat("Estimated prob of consistent estimator corresponding to mean",est_t2_p
rob,"\n")

## Estimated prob of consistent estimator corresponding to mean 0.4 0.45 0.
964

#Q2 Check the consistency of  $T_1 = X(n)$ ,  $T_2 = \bar{X}$  and  $T_3 = 2\bar{X}$  in the
# case of  $Uniform(0,\theta)$  based on the random sample. Suggest one more consist
ent
# estimator for  $\theta$ . Compare  $T_1, T_2$ , and  $T_3$  in terms of MSE.

n=5
a=0
b=6
eps=.1
x=runif(n*n,a,b)
data=matrix(x,n,n)
t1=apply(data,1,max)
t2=apply(data,1,mean)
t3=2*t2
est_t1_prob=mean(abs(t1-b)<eps)
est_t2_prob=mean(abs(t2-b)<eps)
est_t3_prob=mean(abs(t3-b)<eps)
mse_1=mean((t1-b)^2)
mse_2=mean((t2-b)^2)
mse_3=mean((t3-b)^2)
cbind(n,est_t1_prob,est_t2_prob,est_t3_prob,mse_1,mse_2,mse_3)

##          n est_t1_prob est_t2_prob est_t3_prob    mse_1    mse_2    mse_3
## [1,]    5           0           0           0 1.418695 11.33218 3.051098

#simulation:
n=c(50,100,500,1000)

```

```

a=0
b=6
eps=0.1
est_t1_prob=0;est_t2_prob=0;est_t3_prob=0;est_t4_prob=0
mse_1=0;mse_2=0;mse_3=0;mse_4=0
for (i in 1:length(n))
{
  x=runif(n[i]*n[i],a,b)
  data=matrix(x,n[i],n[i])
  t1=apply(data,1,max)
  t2=apply(data,1,mean)
  t3=2*t2
  est_t1_prob[i]=mean(abs(t1-b)<eps)
  est_t2_prob[i]=mean(abs(t2-b)<eps)
  est_t3_prob[i]=mean(abs(t3-b)<eps)
  mse_1[i]=mean((t1-b)^2)
  mse_2[i]=mean((t2-b)^2)
  mse_3[i]=mean((t3-b)^2)
}
cbind(n,est_t1_prob,est_t2_prob,est_t3_prob,mse_1,mse_2,mse_3,mse_4)
##          n est_t1_prob est_t2_prob est_t3_prob      mse_1      mse_2
mse_3
## [1,]    50         0.580          0          0.080 0.0269905620 9.024411 0.25
220601
## [2,]   100         0.800          0          0.170 0.0063386671 8.902444 0.12
105806
## [3,]   500         0.998          0          0.470 0.0003019584 9.034023 0.02
496796
## [4,]  1000         1.000          0          0.612 0.0000832190 8.995953 0.01
281578
##          mse_4
## [1,]          0
## [2,]          0
## [3,]          0
## [4,]          0

# Q.3 Based on the random samples, check the consistency of empirical
# distribution function at  $t=2$  for the FX ( $t$ ) , where  $F(.)$  is cdf of
# Exponential distribution with parameter  $\theta$  .
# Program:

```

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n=5
t=2
theta=1
eps=.05
est_t1_prob=0
n=c(100,200,500)
f_x=1-exp(-t*theta)
for (i in 1:length(n))
{
  x=rexp(n[i]*n[i],theta)
  y=matrix(x,n[i],n[i])
  y1=y<=t
  t1=apply(y1,1,mean)
  est_t1_prob[i]=mean(abs(t1-f_x)<eps)
}
cbind(n,est_t1_prob)

```

```

##      n est_t1_prob
## [1,] 100      0.83
## [2,] 200      0.96
## [3,] 500      1.00

```

*#4 Let X be a random variable having poisson distribution with parameter λ .
 # Suggest one consistent estimator for $e^{(-\lambda)}$. Based on the random samples,
 # demonstrate the consistency of the estimator suggested by you.*

Program:

```

rm(list = ls(all=T))
n=5
eps=.06
lambda=2
est_t1_prob=0
est_t2_prob=0
n=c(50,100,200,500)
for (i in 1:length(n))
{
  x=matrix(rpois(n[i]*100,lambda = lambda),100,n[i])
  t1=apply(x,1,mean)
  t2=exp(-t1)

```

```

t3=apply(x,1,var)
t4=exp(-t3)
est_t1_prob[i]=mean(abs(t2-exp(-lambda))<eps)
est_t2_prob[i]=mean(abs(t4-exp(-lambda))<eps)

}
cbind(n,est_t1_prob,est_t2_prob)
##           n est_t1_prob est_t2_prob
## [1,]  50           0.94           0.71
## [2,] 100           1.00           0.91
## [3,] 200           1.00           0.96
## [4,] 500           1.00           1.00

```

AC

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# Q.1 let X be a random variable having Poisson( $\lambda$ ). Based on the random
# sample of size n= 20 from Poisson( $\lambda$ ), obtain 95% ACI for  $\lambda$ ,
# using: a)VST b) Pivotal method.
# Program:

rm(list=ls(all=T))
n=100; lambda=2;
x=rpois(n,lambda)
m=mean(x)
z=qnorm(0.025,lower.tail=F)
L_1=((2*m+z^2/n)-sqrt((2*m+z^2/n)^2-4*m^2))/2
U_1=((2*m+z^2/n)+sqrt((2*m+z^2/n)^2-4*m^2))/2
cbind(L_1,U_1)
##           L_1      U_1
## [1,] 1.657655 2.20076

length_1=U_1-L_1
L_2=(sqrt(m)-z/(2*sqrt(n)))^2
U_2=(sqrt(m)+z/(2*sqrt(n)))^2
cbind(L_2,U_2)
##           L_2      U_2

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## [1,] 1.648731 2.190476
```

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length_2=U_2-L_2
```

```
# Q.2 let X be a random variable having  $B(1, \theta)$ . Based on the random sample  
of size  $n= 25$  from  $B(1,0.4)$ , obtain 95% ACI for  $\theta$ ,
```

```
# using: a) Pivotal method.
```

```
# Program:
```

```
rm(list=ls(all=T))
```

```
n=25
```

```
th=0.4
```

```
x=rbinom(n,1,th)
```

```
m=mean(x)
```

```
z=qnorm(0.025,lower.tail = F)
```

```
L_1=((2*n*m+z^2)-sqrt((4*z^2*n*m*(1-m)+z^4)))/(2*(z^2+n))
```

```
U_1=((2*n*m+z^2)+sqrt((4*z^2*n*m*(1-m)+z^4)))/(2*(z^2+n))
```

```
cbind(L_1,U_1,length_1=U_1-L_1)
```

```
##           L_1           U_1 length_1
```

```
## [1,] 0.3003129 0.6650148 0.364702
```

```
# Q.3 . let X be a random variable having  $\text{Cauchy}(\mu,1)$ . Based on the random  
sample of size  $n= 25$  from  $\text{Cauchy}(4,1)$ , obtain 95% ACI for  $\mu$ ,
```

```
# using: a) Pivotal method.
```

```
# Program:
```

```
rm(list=ls(all=T))
```

```
n=25
```

```
mu=4
```

```
x=rcauchy(n,mu,1)
```

```
u=c(median(x),rep(0,10))
```

```
v=c(median(x),rep(0,10))
```

```
I=1/2
```

```
z=qnorm(0.025,lower.tail = F)
```

```
for(i in 1:10)
```

```
{
```

```
  y=x-u[i]
```

```
  d_1=2*sum((y)/(1+y^2))
```

```
  d2_1=2*sum((2*y^2-1-y^2)/(1+y^2)^2)
```

```

    u[i+1]=u[i]-d_1/d2_1
    v[i+1]=u[i]+d_1/(n*I)
}
u
## [1] 3.510699 3.609152 3.605713 3.605710 3.605710 3.605710 3.605710 3.60
5710
## [9] 3.605710 3.605710 3.605710
v
## [1] 3.510699 3.592675 3.606081 3.605710 3.605710 3.605710 3.605710 3.60
5710
## [9] 3.605710 3.605710 3.605710
L_1=v-(z*sqrt(2/n))
U_1=v+(z*sqrt(2/n))
cbind(L_1,U_1)
##           L_1           U_1
## [1,] 2.956337 4.065060
## [2,] 3.038313 4.147036
## [3,] 3.051720 4.160443
## [4,] 3.051348 4.160072
## [5,] 3.051348 4.160071
## [6,] 3.051348 4.160071
## [7,] 3.051348 4.160071
## [8,] 3.051348 4.160071
## [9,] 3.051348 4.160071
## [10,] 3.051348 4.160071
## [11,] 3.051348 4.160071

# Q.4 Let X be a random variable having Exp(mean=?).
#      Based on the random sample of size n= 25 from Exp(??), obtain 95% A
CI      for ? using: a)VST b) Pivotal method.
# Program:
{
rm(list=ls(all=T))
n=25
theta=3
x=rexp(n,theta)
z=qnorm(0.025,lower.tail = F)
m=mean(x)
m1=log(m)

```

```

L_1=exp(m1-z/sqrt(n))
U_1=exp(m1+z/sqrt(n))}
cbind(L_1,U_1,length_1=U_1-L_1)

##           L_1           U_1   length_1
## [1,] 0.2259846 0.494948 0.2689633

```

```

s1=1+(z/sqrt(n))
s2=1-(z/sqrt(n))
L_2=mean(x)/s1
U_2=mean(x)/s2
cbind(L_2,U_2,length_2=U_2-L_2)

##           L_2           U_2   length_2
## [1,] 0.2402604 0.5500606 0.3098001

```

Q.5 Let X be a random variable having Laplace(θ , 1). Based on the random sample of size $n= 25$ from Laplace(2.5,1), obtain 95% ACI for θ , using Pivotal method.

```

#   Program:
rm(list=ls(all=T))
n=25
t=2.5
z=qnorm(0.025,lower.tail = F)
Y=runif(n,0,1)
X=t+log(2*Y)*(Y<=0.5)-log(2*(1-Y))*(Y<=0.5)
L1=median(X)-z/sqrt(n)
U1=median(X)+z/sqrt(n)
cbind(L1,U1,length1=U1-L1)

##           L1           U1   length1
## [1,] 2.108007 2.891993 0.7839856

```