

CENTRAL UNIVERSITY OF SOUTH BIHAR

DATA SCIENCE AND APPLIED STATISTICS

(SIDS ASSIGNMENT) SEMESTER - II

[SESSION: 2023-2025]

SUBMITTED TO

SUBMITTED BY

Dr. SANDEEP KUMAR

SURESH Kr. PRAJAPATI Enroll. no. :CUSB2302222008

ASSISTANT PROFESSOR

DEPARTMENT OF STATISTICS
SCHOOL OF MATHEMATICS, STATISTICS AND COMPUTER
SCIENCE CENTRAL UNIVERSITY OF SOUTH BIHAR

Suresh Kumar Prajapati

CUSB2302222008

2024-04-30

Lab Problem-1

Let x1, x2.....xn be random sample from $N(\mu, \sigma^2)$ where n=30,50,60 and μ =-2,0.2, σ^2 =4 then show T1=sample mean is MVUE, In comparison with T2=sample median.

```
M=0
Med=0
for(i in 1:1000){
  x = rnorm(30, -2, 4)
  M[i]=mean(x)
  Med[i]=median(x)
 Mean=c(M)
  Median=c(Med)
}
mean(Mean)
## [1] -1.990801
mean(Median)
## [1] -1.99167
V mean=var(Mean)
V_median=var(Median)
V_mean
## [1] 0.5700001
V median
## [1] 0.8533776
M=0
Med=0
for(i in 1:1000){
  x = rnorm(50, -2, 4)
  M[i]=mean(x)
  Med[i]=median(x)
  Mean=c(M)
  Median=c(Med)
}
mean(Mean)
```

```
## [1] -2.010509
mean(Median)
## [1] -2.016698
V_mean=var(Mean)
V_median=var(Median)
V mean
## [1] 0.3292908
V_median
## [1] 0.4868359
M=0
Med=0
for(i in 1:1000){
  x = rnorm(60, -2, 4)
  M[i]=mean(x)
  Med[i]=median(x)
  Mean=c(M)
  Median=c(Med)
}
mean(Mean)
## [1] -2.011126
mean(Median)
## [1] -1.996596
V_mean=var(Mean)
V_median=var(Median)
V_mean
## [1] 0.2493991
V_median
## [1] 0.3899442
```

Conclusion

Since the mean and median of observed sample is approx -2, conclude that mean and median is the unbiased estimator of generated sample. Also, the variance of mean is less than the variance of median, so mean is the minimum variance unbiased estimator.

Suresh Kumar Prajapat

CUSB2302222008

2024-05-01

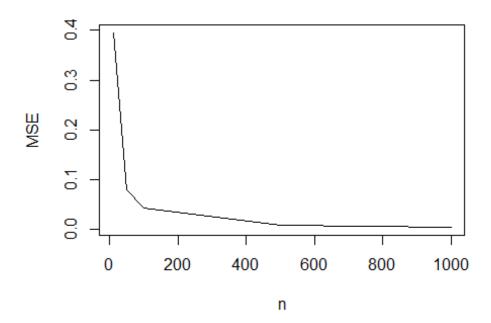
Lab problem-2

Let $X{\exp()}$ then show that \dot{x} is parameter θ , Also plot the graph of MSE for varies n.

```
n=c(10,50,100,500,1000)
Theta=0.5
M1=c(0)
for(i in 1:1000){
  Y1=rexp(n[1],0.5)
  M1[i]=mean(Y1)
 M1=c(M1)
}
Mean_M1=mean(M1)
Var_M1=var(M1)
Mean_M1
## [1] 1.998824
Var_M1
## [1] 0.3949734
M2=c(0)
for(i in 1:1000){
  Y2=rexp(n[2],0.5)
  M2[i]=mean(Y2)
 M2=c(M2)
Mean_M2=mean(M2)
Var_M2=var(M2)
Mean M2
## [1] 2.018911
Var_M2
## [1] 0.07873411
M3=c(0)
for(i in 1:1000){
  Y3 = rexp(n[3], 0.5)
  M3[i]=mean(Y3)
 M3=c(M3)
```

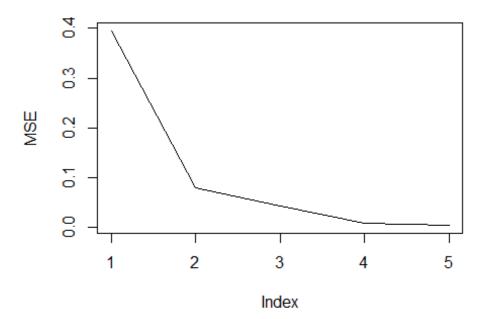
```
Mean_M3=mean(M3)
Var_M3=var(M3)
Mean_M3
## [1] 1.997803
Var_M3
## [1] 0.04175563
M4=c(0)
for(i in 1:1000){
 Y4=rexp(n[4],0.5)
  M4[i]=mean(Y4)
 M4=c(M4)
}
Mean M4=mean(M4)
Var_M4=var(M4)
Mean_M4
## [1] 1.994475
Var_M4
## [1] 0.008115649
M5=c(0)
for(i in 1:1000){
  Y5 = rexp(n[5], 0.5)
  M5[i]=mean(Y5)
  M5=c(M5)
}
Mean_M5=mean(M5)
Var_M5=var(M5)
Mean_M5
## [1] 1.999166
Var_M5
## [1] 0.004207102
#ploting graph for MSE
MSE=c(Var_M1, Var_M2, Var_M3, Var_M4, Var_M5)
plot(n,MSE,main="MSE VS n",type="1")
```

MSE VS n



plot(MSE,main="Graph of MSE",type="1")

Graph of MSE



Suresh Kumar Prajapati

CUSB2302222008

2024-04-30

Lab problem-3

Generate a random sample of size 40 from exponential distribution

$$f(x;\theta) = \theta e^{-\theta x}$$

 θ = 0.5,0.2 estimate the population parameter by using fisherscoring method for the value ϵ = 0.002 also calculate number of iteration.

```
rm(list=ls())
sam_x = rexp(40, 0.5)
theta g1=0.01
x_bar1=mean(sam_x)
x_bar1
## [1] 2.120643
theta_f1=theta_g1+theta_g1^2*(1/theta_g1-x_bar1)
theta f1
## [1] 0.01978794
count=1
while(abs(theta f1-theta g1)>0.002){
  count=count+1
  theta_g1=theta_f1
  theta_f1=theta_g1+theta_g1^2*(1/theta_g1-x_bar1)
print(count)
## [1] 9
print(theta f1)
## [1] 0.471547
sam_y = rexp(40, 0.2)
theta_g2=0.01
x_bar2=mean(sam_y)
x_bar2
## [1] 5.368385
theta_f2=theta_g2+theta_g2^2*(1/theta_g2-x_bar2)
theta f2
```

```
## [1] 0.01946316

count=1
while(abs(theta_f2-theta_g2)>0.002){
   count=count+1
   theta_g2=theta_f2
   theta_f2=theta_g2+theta_g2^2*(1/theta_g2-x_bar2)
}
print(count)
## [1] 8
print(theta_f2)
## [1] 0.1862756
```

Conclusion

1. For θ = 0.5,the number of iteration is 9 For θ = 0.2,the number of iteration is 8

Suresh Kumar Prajapati

CUSB2302222008

2024-05-01

Lab Problem-4

Let x1, x2....xn be random sample from $N(\theta, \sigma^2)$ where $\sigma^2 = 4, 16, 25$ then compute the minimum sample size for $\epsilon = 0.01, 0.001$ and $\delta = 0.002, 0.004$

```
epsilon=c(0.01,0.001)
delta=c(0.002,0.004)
sigma_sq=c(4,16,25)
P1=round((sigma_sq[1]/(delta[1]*epsilon[1]^2))+1)
options(scipen=P1)
P1
## [1] 20000001
P2=round((sigma_sq[1]/(delta[1]*epsilon[2]^2))+1)
options(scipen=P2)
P2
## [1] 2000000001
P3=round((sigma_sq[1]/(delta[2]*epsilon[1]^2))+1)
options(scipen=P3)
Р3
## [1] 10000001
P4=round((sigma_sq[1]/(delta[2]*epsilon[2]^2))+1)
options(scipen=P4)
P4
## [1] 1000000001
P5=round((sigma_sq[2]/(delta[1]*epsilon[1]^2))+1)
options(scipen=P5)
P5
```

```
## [1] 80000001
P6=round((sigma_sq[2]/(delta[1]*epsilon[2]^2))+1)
options(scipen=P6)
P6
## Warning in print.default(x): NAs introduced by coercion to integer range
## [1] 8e+09
## Warning in print.default(x): NAs introduced by coercion to integer range
P7=round((sigma sq[2]/(delta[2]*epsilon[1]^2))+1)
options(scipen=P7)
P7
## [1] 40000001
P8=round((sigma_sq[2]/(delta[2]*epsilon[2]^2))+1)
options(scipen=P8)
P8
## Warning in print.default(x): NAs introduced by coercion to integer range
## [1] 4e+09
## Warning in print.default(x): NAs introduced by coercion to integer range
P9=round((sigma_sq[3]/(delta[1]*epsilon[1]^2))+1)
options(scipen=P9)
P9
## [1] 125000001
P10=round((sigma_sq[3]/(delta[1]*epsilon[2]^2))+1)
options(scipen=P10)
P10
## Warning in print.default(x): NAs introduced by coercion to integer range
## [1] 1.25e+10
## Warning in print.default(x): NAs introduced by coercion to integer range
P11=round((sigma_sq[3]/(delta[2]*epsilon[1]^2))+1)
options(scipen=P11)
P11
## [1] 62500001
P12=round((sigma_sq[3]/(delta[2]*epsilon[2]^2))+1)
options(scipen=P12)
P12
```

```
## Warning in print.default(x): NAs introduced by coercion to integer range
## [1] 6.25e+09
## Warning in print.default(x): NAs introduced by coercion to integer range
10
```