

ภาคผนวก

ภาคผนวก ก

โปรแกรมที่ใช้ในการวิจัย

ในการวิจัยครั้งนี้ ใช้โปรแกรมสำเร็จรูปทางสถิติ โปรแกรม R version 2.9.1 ในการจำลองข้อมูล คำนวณค่าขีดจำกัดล่างและขีดจำกัดบน และเกณฑ์ที่ใช้ในการวิจัย

A comparison on confidence intervals estimation methods for the poisson population mean.

```
function (n,M,alpha,alpha1,beta1,alpha2,beta2,alpha3,beta3,alpha4,beta4)
{
  ####Written by Rungtawan , 15/06/09
  ####n is a sample size.
  ####M is a number of simulation runs.
  ####alpha is a significance level.
  ####alpha1,beta1,alpha2,beta2,alpha3,beta3,alpha4 and beta4 are parameter of
  ####gamma distribution

  L.WALD=rep(0,M)
  U.WALD=rep(0,M)
  L.SCORE=rep(0,M)
  U.SCORE=rep(0,M)
  L.ADJUST=rep(0,M)
  U.ADJUST=rep(0,M)
  L.BAYE1=rep(0,M)
  U.BAYE1=rep(0,M)
  L.BAYE2=rep(0,M)
  U.BAYE2=rep(0,M)
  L.BAYE3=rep(0,M)
  U.BAYE3=rep(0,M)
```

```
L.BAYE4=rep(0,M)
U.BAYE4=rep(0,M)
length.WALD=rep(0,M)
length.SCORE=rep(0,M)
length.ADJUST=rep(0,M)
length.BAYE1=rep(0,M)
length.BAYE2=rep(0,M)
length.BAYE3=rep(0,M)
length.BAYE4=rep(0,M)

lambda<-c(0.2,0.4,0.6,0.8,1)
y1=rep(0,5)
y2=rep(0,5)
y3=rep(0,5)
y4=rep(0,5)
y5=rep(0,5)
y6=rep(0,5)
y7=rep(0,5)

for(k in 1:length(lambda))
{
    count1=rep(0,M)
    count2=rep(0,M)
    count3=rep(0,M)
    count4=rep(0,M)
    count5=rep(0,M)
    count6=rep(0,M)
    count7=rep(0,M)
```

```

for(i in 1:M)

{
    x=rpois(n,lambd[k])
    z=qnorm(1-alpha/2)

##### compute upper and lower bound

    lambda_hat=mean(x)

    L.WALD[i]=lambda_hat-z*sqrt(lambda_hat/n)
    U.WALD[i]=lambda_hat+z*sqrt(lambda_hat/n)

    L.SCORE[i]=(lambda_hat+(z^2/(2*n)))-
z*sqrt(lambda_hat/n+(z^2/(4*n*n)))

    U.SCORE[i]=(lambda_hat+(z^2/(2*n)))+
z*sqrt(lambda_hat/n+(z^2/(4*n*n)))

    lambda_adjust=(sum(x)+(z^2)/2)/n
    L.ADJUST[i]=lambda_adjust-z*sqrt(lambda_adjust/n)
    U.ADJUST[i]=lambda_adjust+z*sqrt(lambda_adjust/n)
    df1=2*(sum(x)+alpha1)
    L.BAYE1[i]=beta1*qchisq(alpha/2,df1)/(2*(beta1*n+1))
    U.BAYE1[i]=beta1*qchisq(1-alpha/2,df1)/(2*(beta1*n+1))

    df2=2*(sum(x)+alpha2)
    L.BAYE2[i]=beta2*qchisq(alpha/2,df2)/(2*(beta2*n+1))
    U.BAYE2[i]=beta2*qchisq(1-alpha/2,df2)/(2*(beta2*n+1))

    df3=2*(sum(x)+alpha3)
    L.BAYE3[i]=beta3*qchisq(alpha/2,df3)/(2*(beta3*n+1))
    U.BAYE3[i]=beta3*qchisq(1-alpha/2,df3)/(2*(beta3*n+1))

    df4=2*(sum(x)+alpha4)
    L.BAYE4[i]=beta4*qchisq(alpha/2,df4)/(2*(beta4*n+1))
    U.BAYE4[i]=beta4*qchisq(1-alpha/2,df4)/(2*(beta4*n+1))

```

```
##### compute coverage probability and average length
```

```

    if((lambda[k]>=L.WALD[i])&(lambda[k]<=U.WALD[i]))
        {count1[i]=1}

    if((lambda[k]>=L.SCORE[i])&(lambda[k]<=U.SCORE[i]))
        {count2[i]=1}

    if((lambda[k]>=L.ADJUST[i])&(lambda[k]<=U.ADJUST[i]))
        {count3[i]=1}

    if((lambda[k]>=L.BAYE1[i])&(lambda[k]<=U.BAYE1[i]))
        {count4[i]=1}

    if((lambda[k]>=L.BAYE2[i])&(lambda[k]<=U.BAYE2[i]))
        {count5[i]=1}

    if((lambda[k]>=L.BAYE3[i])&(lambda[k]<=U.BAYE3[i]))
        {count6[i]=1}

    if((lambda[k]>=L.BAYE4[i])&(lambda[k]<=U.BAYE4[i]))
        {count7[i]=1}

    length.WALD[i]=U.WALD[i]-L.WALD[i]
    length.SCORE[i]=U.SCORE[i]-L.SCORE[i]
    length.ADJUST[i]=U.ADJUST[i]-L.ADJUST[i]
    length.BAYE1[i]=U.BAYE1[i]-L.BAYE1[i]
    length.BAYE2[i]=U.BAYE2[i]-L.BAYE2[i]
    length.BAYE3[i]=U.BAYE3[i]-L.BAYE3[i]
    length.BAYE4[i]=U.BAYE4[i]-L.BAYE4[i]

}

y1[k]=mean(count1)
y2[k]=mean(count2)

```

```

y3[k]=mean(count3)
y4[k]=mean(count4)
y5[k]=mean(count5)
y6[k]=mean(count6)
y7[k]=mean(count7)

cat("\n", "n = ", n, " Lambda = ", lambda[k], "\n")
cat(" Baye1=(", alpha1, ",", beta1, ") Baye2=(", alpha2, ",", beta2, ")
Baye3=(", alpha3, ",", beta3, ") Baye4=(", alpha4, ",", beta4, ") ", "\n")
cat(" Coverage Prob Wald = ", mean(count1), " Length Wald = ",
mean(length.WALD), "\n")
cat(" Coverage Prob Score = ", mean(count2), " Length Score = ",
mean(length.SCORE), "\n")
cat(" Coverage Prob Adjust = ", mean(count3), " Length Adjust = ",
mean(length.ADJUST), "\n")
cat(" Coverage Prob Baye1 = ", mean(count4), " Length Baye1 = ",
mean(length.BAYE1), "\n")
cat(" Coverage Prob Baye2 = ", mean(count5), " Length Baye2 = ",
mean(length.BAYE2), "\n")
cat(" Coverage Prob Baye3 = ", mean(count6), " Length Baye3 = ",
mean(length.BAYE3), "\n")
cat(" Coverage Prob Baye4 = ", mean(count7), " Length Baye4 = ",
mean(length.BAYE4), "\n")

}

main_title=paste("n = ", n)
plot(lambda, y1, type="o", lwd=2, col="black", main=main_title, ylab="Covera
ge Probability : CP", xlim=c(0.2, 1.2), ylim=c(0.98, 0.995), pch=20, lty=1)
lines(lambda, y2, type="o", lwd=2, col="red", pch=20, lty=1)
lines(lambda, y3, type="o", lwd=2, col="green", pch=20, lty=1)

```

```

lines(lambda,y4,type="o",lwd=2,col="blue",pch=20,lty=1)
lines(lambda,y5,type="o",lwd=2,col="dark red",pch=20,lty=1)
lines(lambda,y6,type="o",lwd=2,col="yellow",pch=20,lty=1)
lines(lambda,y7,type="o",lwd=2,col="pink",pch=20,lty=1)

legend("bottomright",c("Wald","Score","Adjust","Baye(1,1)","Baye(1,2)","Ba
ye(2,1)","Baye(2,2)"),
      lty=c(1,1,1,1,1,1,1),col=c("black","red","green","blue","dark
red","yellow","pink"),lwd=2,cex=0.93)
}

```

A comparison on confidence intervals estimation methods for the geometric population mean.

```
function (n,M,alpha,alpha1,beta1,alpha2,beta2,alpha3,beta3,alpha4,beta4)
{
  ####Written by Rungtawan , 17/06/09
  ####n is a sample size.
  ####M is a number of simulation runs.
  ####alpha is a significance level.
  ####alpha1,beta1,alpha2,beta2,alpha3,beta3,alpha4 and beta4 are parameter of
  ####beta distribution

  L.WALD=rep(0,M)
  U.WALD=rep(0,M)
  L.SCORE=rep(0,M)
  U.SCORE=rep(0,M)
  L.ADJUST=rep(0,M)
  U.ADJUST=rep(0,M)
  L.BAYE1=rep(0,M)
  U.BAYE1=rep(0,M)
  L.BAYE2=rep(0,M)
  U.BAYE2=rep(0,M)
  L.BAYE3=rep(0,M)
  U.BAYE3=rep(0,M)
  L.BAYE4=rep(0,M)
  U.BAYE4=rep(0,M)
  length.WALD=rep(0,M)
  length.SCORE=rep(0,M)
  length.ADJUST=rep(0,M)
  length.BAYE1=rep(0,M)
  length.BAYE2=rep(0,M)
```

```

length.BAYE3=rep(0,M)
length.BAYE4=rep(0,M)

p<-c(0.1,0.3,0.5,0.7,0.9)
y1=rep(0,5)
y2=rep(0,5)
y3=rep(0,5)
y4=rep(0,5)
y5=rep(0,5)
y6=rep(0,5)
y7=rep(0,5)

for(k in 1:length(p))
{
    count1=rep(0,M)
    count2=rep(0,M)
    count3=rep(0,M)
    count4=rep(0,M)
    count5=rep(0,M)
    count6=rep(0,M)
    count7=rep(0,M)

    for(i in 1:M)
    {
        y=rgeom(n,p[k])
        x=y+1
        z=qnorm(1-alpha/2)

##### compute upper and lower bound

        p_hat=1/mean(x)
        L.WALD[i]=(1/p_hat)-z*sqrt((1-p_hat)/(n*p_hat^2))
        U.WALD[i]=(1/p_hat)+z*sqrt((1-p_hat)/(n*p_hat^2))
    }
}

```

$$L.SCORE[i] = ((n/p_hat) - ((z^2)/2) - z * \sqrt{((z^2)/4 - (n/p_hat) * (1 - 1/p_hat))}) / (n - z^2)$$

$$U.SCORE[i] = ((n/p_hat) - ((z^2)/2) + z * \sqrt{((z^2)/4 - (n/p_hat) * (1 - 1/p_hat))}) / (n - z^2)$$

$$p_adjust = (n - z^2) / (\sum(x) - (z^2)/2)$$

$$L.ADJUST[i] = 1/p_adjust - z * \sqrt{(1 - p_adjust) / ((n - z^2) * (p_adjust^2))}$$

$$U.ADJUST[i] = 1/p_adjust + z * \sqrt{(1 - p_adjust) / ((n - z^2) * (p_adjust^2))}$$

$$a1 = 2 * (\beta_1 + \sum(x) - n)$$

$$b1 = 2 * (\alpha_1 + n)$$

$$L.BAYE1[i] = (2 * (\alpha_1 + n) + 2 * (\beta_1 + \sum(x) - n) * qf(\alpha/2, a1, b1)) / (2 * (\alpha_1 + n))$$

$$U.BAYE1[i] = (2 * (\alpha_1 + n) + 2 * (\beta_1 + \sum(x) - n) * qf(1 - \alpha/2, a1, b1)) / (2 * (\alpha_1 + n))$$

$$a2 = 2 * (\beta_2 + \sum(x) - n)$$

$$b2 = 2 * (\alpha_2 + n)$$

$$L.BAYE2[i] = (2 * (\alpha_2 + n) + 2 * (\beta_2 + \sum(x) - n) * qf(\alpha/2, a2, b2)) / (2 * (\alpha_2 + n))$$

$$U.BAYE2[i] = (2 * (\alpha_2 + n) + 2 * (\beta_2 + \sum(x) - n) * qf(1 - \alpha/2, a2, b2)) / (2 * (\alpha_2 + n))$$

$$a3 = 2 * (\beta_3 + \sum(x) - n)$$

$$b3 = 2 * (\alpha_3 + n)$$

$$L.BAYE3[i] = (2 * (\alpha_3 + n) + 2 * (\beta_3 + \sum(x) - n) * qf(\alpha/2, a3, b3)) / (2 * (\alpha_3 + n))$$

$$U.BAYE3[i] = (2 * (\alpha_3 + n) + 2 * (\beta_3 + \sum(x) - n) * qf(1 - \alpha/2, a3, b3)) / (2 * (\alpha_3 + n))$$

```

a4=2*(beta4+sum(x)-n)
b4=2*(alpha4+n)
L.BAYE4[i]=(2*(alpha4+n)+2*(beta4+sum(x)-
n)*qf(alpha/2,a4,b4))/(2*(alpha4+n))
U.BAYE4[i]=(2*(alpha4+n)+2*(beta4+sum(x)-n)*qf(1-
alpha/2,a4,b4))/(2*(alpha4+n))

```

```
##### compute coverage probability and average length
```

```

if((1/p[k]>=L.WALD[i])&(1/p[k]<=U.WALD[i]))
  {count1[i]=1}

if((1/p[k]>=L.SCORE[i])&(1/p[k]<=U.SCORE[i]))
  {count2[i]=1}

if((1/p[k]>=L.ADJUST[i])&(1/p[k]<=U.ADJUST[i]))
  {count3[i]=1}

if((1/p[k]>=L.BAYE1[i])&(1/p[k]<=U.BAYE1[i]))
  {count4[i]=1}

if((1/p[k]>=L.BAYE2[i])&(1/p[k]<=U.BAYE2[i]))
  {count5[i]=1}

if((1/p[k]>=L.BAYE3[i])&(1/p[k]<=U.BAYE3[i]))
  {count6[i]=1}

if((1/p[k]>=L.BAYE4[i])&(1/p[k]<=U.BAYE4[i]))
  {count7[i]=1}

length.WALD[i]=U.WALD[i]-L.WALD[i]
length.SCORE[i]=U.SCORE[i]-L.SCORE[i]
length.ADJUST[i]=U.ADJUST[i]-L.ADJUST[i]
length.BAYE1[i]=U.BAYE1[i]-L.BAYE1[i]

```

```

length.BAYE2[i]=U.BAYE2[i]-L.BAYE2[i]
length.BAYE3[i]=U.BAYE3[i]-L.BAYE3[i]
length.BAYE4[i]=U.BAYE4[i]-L.BAYE4[i]
}
y1[k]=mean(count1)
y2[k]=mean(count2)
y3[k]=mean(count3)
y4[k]=mean(count4)
y5[k]=mean(count5)
y6[k]=mean(count6)
y7[k]=mean(count7)

cat("\n","n = ",n," E(X) = 1/p = ",1/p[k],"\n")
cat(" Baye1=(,alpha1,,",beta1,) Baye2=(,alpha2,,",beta2,)
Baye3=(,alpha3,,",beta3,) Baye4=(,alpha4, ",",beta4,)" ", "\n")
cat(" Coverage Prob Wald = ", mean(count1)," Length Wald = ",
mean(length.WALD), "\n")
cat(" Coverage Prob Score = ", mean(count2)," Length Score = ",
mean(length.SCORE), "\n")
cat(" Coverage Prob Adjust = ", mean(count3)," Length Adjust = ",
mean(length.ADJUST), "\n")
cat(" Coverage Prob Baye1 = ", mean(count4)," Length Baye1 = ",
mean(length.BAYE1), "\n")
cat(" Coverage Prob Baye2 = ", mean(count5)," Length Baye2 = ",
mean(length.BAYE2), "\n")
cat(" Coverage Prob Baye3 = ", mean(count6)," Length Baye3 = ",
mean(length.BAYE3), "\n")
cat(" Coverage Prob Baye4 = ", mean(count7)," Length Baye4 = ",
mean(length.BAYE4), "\n")
}

```

```

main_title=paste("n = ",n)

plot(p,y1,type="o",lwd=2,col="black",main=main_title,ylab="Coverage
Probability : CP",xlim=c(0.1,1.15),ylim=c(0.8,1),pch=20,lty=1)

lines(p,y2,type="o",lwd=2,col="red",pch=20,lty=1)
lines(p,y3,type="o",lwd=2,col="green",pch=20,lty=1)
lines(p,y4,type="o",lwd=2,col="blue",pch=20,lty=1)
lines(p,y5,type="o",lwd=2,col="dark red",pch=20,lty=1)
lines(p,y6,type="o",lwd=2,col="yellow",pch=20,lty=1)
lines(p,y7,type="o",lwd=2,col="pink",pch=20,lty=1)

legend("bottomright",c("Wald","Score","Adjust","Baye(0.5,2)","Baye(1,1)","B
aye(1,2)","Baye(2,1)"),

lty=c(1,1,1,1,1,1,1),col=c("black","red","green","blue","dark
red","yellow","pink"),lwd=2,cex=0.93)

}

```

A comparison on confidence intervals estimation methods for the exponential population mean.

```
function (n,M,alpha,alpha1,beta1,alpha2,beta2,alpha3,beta3,alpha4,beta4)

{
  ####Written by Rungtawan , 20/06/09
  ####n is a sample size.
  ####M is a number of simulation runs.
  ####alpha is a significance level.
  ####alpha1,beta1,alpha2,beta2,alpha3,beta3,alpha4 and beta4 are parameter of
  ####gamma distribution

  beta_hat=rep(0,M)
  z=rep(0,M)
  beta_adjust=rep(0,M)
  L.WALD=rep(0,M)
  U.WALD=rep(0,M)
  L.SCORE=rep(0,M)
  U.SCORE=rep(0,M)
  L.ADJUST=rep(0,M)
  U.ADJUST=rep(0,M)
  L.BAYE1=rep(0,M)
  U.BAYE1=rep(0,M)
  L.BAYE2=rep(0,M)
  U.BAYE2=rep(0,M)
  L.BAYE3=rep(0,M)
  U.BAYE3=rep(0,M)
  L.BAYE4=rep(0,M)
  U.BAYE4=rep(0,M)
  L.B_SCORE=rep(0,M)
  U.B_SCORE=rep(0,M)
```

```
L.B_BAYE1=rep(0,M)
U.B_BAYE1=rep(0,M)
L.B_BAYE2=rep(0,M)
U.B_BAYE2=rep(0,M)
L.B_BAYE3=rep(0,M)
U.B_BAYE3=rep(0,M)
L.B_BAYE4=rep(0,M)
U.B_BAYE4=rep(0,M)
length.WALD=rep(0,M)
length.SCORE=rep(0,M)
length.ADJUST=rep(0,M)
length.BAYE1=rep(0,M)
length.BAYE2=rep(0,M)
length.BAYE3=rep(0,M)
length.BAYE4=rep(0,M)
length.B_SCORE=rep(0,M)
length.B_BAYE1=rep(0,M)
length.B_BAYE2=rep(0,M)
length.B_BAYE3=rep(0,M)
length.B_BAYE4=rep(0,M)

beta<-c(1,5,10,20,30)
y1=rep(0,5)
y2=rep(0,5)
y3=rep(0,5)
y4=rep(0,5)
y5=rep(0,5)
y6=rep(0,5)
y7=rep(0,5)
y8=rep(0,5)
```

```

y9=rep(0,5)
y10=rep(0,5)
y11=rep(0,5)
y12=rep(0,5)

for(k in 1:length(beta))
{
  count1=rep(0,M)
  count2=rep(0,M)
  count3=rep(0,M)
  count4=rep(0,M)
  count5=rep(0,M)
  count6=rep(0,M)
  count7=rep(0,M)
  count8=rep(0,M)
  count9=rep(0,M)
  count10=rep(0,M)
  count11=rep(0,M)
  count12=rep(0,M)

  for(i in 1:M)
  {
    x=rexp(n,1/beta[k])
    z=qnorm(1-alpha/2)

##### compute upper and lower bound

    beta_hat=mean(x)
    L.WALD[i]=beta_hat-z*sqrt((beta_hat^2)/n)
    U.WALD[i]=beta_hat+z*sqrt((beta_hat^2)/n)

    L.SCORE[i]=(n*beta_hat-z*sqrt(n*beta_hat^2))/(n-z^2)
    U.SCORE[i]=(n*beta_hat+z*sqrt(n*beta_hat^2))/(n-z^2)
  }
}

```

```

beta_adjust=sum(x)/(n-(z^2))
L.ADJUST[i]=beta_adjust-z*sqrt(beta_adjust^2/(n-z^2))
U.ADJUST[i]=beta_adjust+z*sqrt(beta_adjust^2/(n-z^2))

df1=2*(alpha1+n)
L.BAYE1[i]=2*(beta1*sum(x)+1)/(beta1*qchisq(1-alpha/2,df1))
U.BAYE1[i]=2*(beta1*sum(x)+1)/(beta1*qchisq(alpha/2,df1))

df2=2*(alpha2+n)
L.BAYE2[i]=2*(beta2*sum(x)+1)/(beta2*qchisq(1-alpha/2,df2))
U.BAYE2[i]=2*(beta2*sum(x)+1)/(beta2*qchisq(alpha/2,df2))

df3=2*(alpha3+n)
L.BAYE3[i]=2*(beta3*sum(x)+1)/(beta3*qchisq(1-alpha/2,df3))
U.BAYE3[i]=2*(beta3*sum(x)+1)/(beta3*qchisq(alpha/2,df3))

df4=2*(alpha4+n)
L.BAYE4[i]=2*(beta4*sum(x)+1)/(beta4*qchisq(1-alpha/2,df4))
U.BAYE4[i]=2*(beta4*sum(x)+1)/(beta4*qchisq(alpha/2,df4))

```

```
##### compute coverage probability and average length
```

```

if((beta[k]>=L.WALD[i])&(beta[k]<=U.WALD[i]))
  {count1[i]=1}

if((beta[k]>=L.SCORE[i])&(beta[k]<=U.SCORE[i]))
  {count2[i]=1}

if((beta[k]>=L.ADJUST[i])&(beta[k]<=U.ADJUST[i]))
  {count3[i]=1}

if((beta[k]>=L.BAYE1[i])&(beta[k]<=U.BAYE1[i]))
  {count4[i]=1}

```

```

if((beta[k]>=L.BAYE2[i])&(beta[k]<=U.BAYE2[i]))
  {count5[i]=1}

if((beta[k]>=L.BAYE3[i])&(beta[k]<=U.BAYE3[i]))
  {count6[i]=1}

if((beta[k]>=L.BAYE4[i])&(beta[k]<=U.BAYE4[i]))
  {count7[i]=1}

length.WALD[i]=U.WALD[i]-L.WALD[i]
length.SCORE[i]=U.SCORE[i]-L.SCORE[i]
length.ADJUST[i]=U.ADJUST[i]-L.ADJUST[i]
length.BAYE1[i]=U.BAYE1[i]-L.BAYE1[i]
length.BAYE2[i]=U.BAYE2[i]-L.BAYE2[i]
length.BAYE3[i]=U.BAYE3[i]-L.BAYE3[i]
length.BAYE4[i]=U.BAYE4[i]-L.BAYE4[i]

#---Bootstrap Estimation -----#
L.b_score=rep(0,200); U.b_score=rep(0,200)
L.b_baye1=rep(0,200); U.b_baye1=rep(0,200)
L.b_baye2=rep(0,200); U.b_baye2=rep(0,200)
L.b_baye3=rep(0,200); U.b_baye3=rep(0,200)
L.b_baye4=rep(0,200); U.b_baye4=rep(0,200)

for(j in 1:200)
{
##### compute upper and lower bound of bootstrap

  xBoots=sample(x,n,replace=TRUE)
  beta_hat_boot=mean(xBoots)

```

```

        L.b_score[j]=(n*beta_hat_boot-
z*sqrt(n*beta_hat_boot^2))/(n-z^2)
        U.b_score[j]=(n*beta_hat_boot+
z*sqrt(n*beta_hat_boot^2))/(n-z^2)

        df1=2*(alpha1+n)
        L.b_baye1[j]=2*(beta1*sum(xBoots)+1)/
(beta1*qchisq(1-alpha/2,df1))
        U.b_baye1[j]=2*(beta1*sum(xBoots)+1)/
(beta1*qchisq(alpha/2,df1))

        df2=2*(alpha2+n)
        L.b_baye2[j]=2*(beta2*sum(xBoots)+1)/
(beta2*qchisq(1-alpha/2,df2))
        U.b_baye2[j]=2*(beta2*sum(xBoots)+1)/
(beta2*qchisq(alpha/2,df2))

        df3=2*(alpha3+n)
        L.b_baye3[j]=2*(beta3*sum(xBoots)+1)/
(beta3*qchisq(1-alpha/2,df3))
        U.b_baye3[j]=2*(beta3*sum(xBoots)+1)/
(beta3*qchisq(alpha/2,df3))

        df4=2*(alpha4+n)
        L.b_baye4[j]=2*(beta4*sum(xBoots)+1)/
(beta4*qchisq(1-alpha/2,df4))
        U.b_baye4[j]=2*(beta4*sum(xBoots)+1)/
(beta4*qchisq(alpha/2,df4))

    }

```

```

##### compute coverage probability and average length of bootstrap

    L.B_SCORE[i]=mean(L.b_score);
U.B_SCORE[i]=mean(U.b_score)
    L.B_BAYE1[i]=mean(L.b_baye1);
U.B_BAYE1[i]=mean(U.b_baye1)
    L.B_BAYE2[i]=mean(L.b_baye2);
U.B_BAYE2[i]=mean(U.b_baye2)
    L.B_BAYE3[i]=mean(L.b_baye3);
U.B_BAYE3[i]=mean(U.b_baye3)
    L.B_BAYE4[i]=mean(L.b_baye4);
U.B_BAYE4[i]=mean(U.b_baye4)
    if((beta[k]>=L.B_SCORE[i])&(beta[k]<=U.B_SCORE[i]))
        {count8[i]=1}

    if((beta[k]>=L.B_BAYE1[i])&(beta[k]<=U.B_BAYE1[i]))
        {count9[i]=1}

    if((beta[k]>=L.B_BAYE2[i])&(beta[k]<=U.B_BAYE2[i]))
        {count10[i]=1}

    if((beta[k]>=L.B_BAYE3[i])&(beta[k]<=U.B_BAYE3[i]))
        {count11[i]=1}

    if((beta[k]>=L.B_BAYE4[i])&(beta[k]<=U.B_BAYE4[i]))
        {count12[i]=1}

    length.B_SCORE[i]=U.B_SCORE[i]-L.B_SCORE[i]
    length.B_BAYE1[i]=U.B_BAYE1[i]-L.B_BAYE1[i]
    length.B_BAYE2[i]=U.B_BAYE2[i]-L.B_BAYE2[i]
    length.B_BAYE3[i]=U.B_BAYE3[i]-L.B_BAYE3[i]
    length.B_BAYE4[i]=U.B_BAYE4[i]-L.B_BAYE4[i]

}

```

```

y1[k]=mean(count1)
y2[k]=mean(count2)
y3[k]=mean(count3)
y4[k]=mean(count4)
y5[k]=mean(count5)
y6[k]=mean(count6)
y7[k]=mean(count7)
y8[k]=mean(count8)
y9[k]=mean(count9)
y10[k]=mean(count10)
y11[k]=mean(count11)
y12[k]=mean(count12)

cat("\n"," n = ",n," Beta = ",beta[k],"n")
cat(" Baye1=(",alpha1,",",beta1,")    Baye2=(",alpha2,",",beta2,")
Baye3=(",alpha3,",",beta3,")    Baye4=(",alpha4, ",",beta4,")" ,"n")
cat(" Coverage Prob Wald  = ", mean(count1)," Length Wald  = ",
mean(length.WALD), "n")
cat(" Coverage Prob Score  = ", mean(count2)," Length Score = ",
mean(length.SCORE), "n")
cat(" Coverage Prob Adjust = ", mean(count3)," Length Adjust = ",
mean(length.ADJUST), "n")
cat(" Coverage Prob Baye1  = ", mean(count4)," Length Baye1  = ",
mean(length.BAYE1), "n")
cat(" Coverage Prob Baye2  = ", mean(count5)," Length Baye2  = ",
mean(length.BAYE2), "n")
cat(" Coverage Prob Baye3  = ", mean(count6)," Length Baye3  = ",
mean(length.BAYE3), "n")
cat(" Coverage Prob Baye4  = ", mean(count7)," Length Baye4  = ",
mean(length.BAYE4), "n")

```

```

      cat(" Coverage Prob Bootstrap_Score  = ", mean(count8), " Length
Bootstrap_Score  = ", mean(length.B_SCORE), "\n")

      cat(" Coverage Prob Bootstrap_Baye1  = ", mean(count9), " Length
Bootstrap_Baye1  = ", mean(length.B_BAYE1), "\n")

      cat(" Coverage Prob Bootstrap_Baye2  = ", mean(count10)," Length
Bootstrap_Baye2  = ", mean(length.B_BAYE2), "\n")

      cat(" Coverage Prob Bootstrap_Baye3  = ", mean(count11)," Length
Bootstrap_Baye3  = ", mean(length.B_BAYE3), "\n")

      cat(" Coverage Prob Bootstrap_Baye4  = ", mean(count12)," Length
Bootstrap_Baye4  = ", mean(length.B_BAYE4), "\n")
    }

    main_title=paste("n = ",n)

    plot(beta,y1,type="o",lwd=2,col="black",main=main_title,ylab="Coverage
Probability : CP",xlim=c(1,37),ylim=c(0.885,0.92),pch=20,lty=1)

    lines(beta,y2,type="o",lwd=2,col="red",pch=20,lty=1)
    lines(beta,y3,type="o",lwd=2,col="green",pch=20,lty=1)
    lines(beta,y4,type="o",lwd=2,col="blue",pch=20,lty=1)
    lines(beta,y5,type="o",lwd=2,col="dark red",pch=20,lty=1)
    lines(beta,y6,type="o",lwd=2,col="yellow",pch=20,lty=1)
    lines(beta,y7,type="o",lwd=2,col="pink",pch=20,lty=1)
    lines(beta,y8,type="o",lwd=2,col="purple",pch=20,lty=1)
    lines(beta,y9,type="o",lwd=2,col="brown",pch=20,lty=1)
    lines(beta,y10,type="o",lwd=2,col="dark blue",pch=20,lty=1)
    lines(beta,y11,type="o",lwd=2,col="orange",pch=20,lty=1)
    lines(beta,y12,type="o",lwd=2,col="gray",pch=20,lty=1)

    legend("bottomright",c("Wald","Score","Adjust","Baye(0.5,2)","Baye(0.5,3)","Baye
(1,1)","Baye(1,2)","B_Score","B_Baye(0.5,2)","B_Baye(0.5,3)","B_Baye(1,1)","B_Baye(1,2)"),lty
=c(1,1,1,1,1,1,1,1,1,1,1,1),col=c("black","red","green","blue","dark
red","yellow","pink","purple","brown","dark blue","orange","gray"),lwd=2,cex=0.76)
  }

```