

# BST169: Course Work Project answer

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## 1 BST169: Course Work Project

### 1.1 Question 1:

1. Consider the model:

$$y_i = \beta_0 + \beta_1 * x_{1,i} + \beta_2 * x_{2,i} + \epsilon_i \quad (1)$$

What is the requirement for  $\epsilon_i$  such that the following test statistics will be valid to test  $H_0: \beta_1 + \beta_2 = 1$ ?

- $W = N * (SSR_R - SSR_U) / SSR_U$  (Wald).
- $LM = N * (SSR_R - SSR_U) / SSR_R$  (Lagrange Multiplier),
- $LR = N * \ln(SSR_R / SSR_U)$  (Likelihood Ratio)

where  $SSR_R$  is the sum of squared residuals obtained from the restricted model, while  $SSR_U$  is from the unrestricted model.

#### 1.1.1 answer

$$\beta_1 + \beta_2 = 1$$

$\Leftrightarrow$

$$R * \beta = 1,$$

where  $R = \begin{bmatrix} 1 & 1 \end{bmatrix}$   $\beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$

1. Wald test

$$H_0: \beta_1 + \beta_2 = 1$$

$$y_i = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon_i$$

$$y_i = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon_i \text{ with } \beta_1 + \beta_2 = 1$$

0).

$$E(x_i \epsilon_i) = 0, i = 1, 2, \dots, N;$$

$$E(|x_i \epsilon_i|^{2+\delta}) < \Delta < 1, \text{ for } \exists \delta > 0, k = 1, \dots, K + 1 \text{ and } i = 1, 2, \dots, N$$

- 1). chi-sq distribution

$$\sqrt{N}(R * \tilde{\beta} - 1) \sim N(0, RM_N^{-1}U_N M_N^{-1}R'),$$

where  $\tilde{\beta} = (X'X)^{-1}X'y$

$$(1/N)(R\tilde{\beta} - 1)(R(X'X)^{-1}\tilde{U}_N(X'X)^{-1}R')^{-1}(R\tilde{\beta} - 1)' \sim \chi^2,$$

where  $X = \begin{bmatrix} X_1 & X_2 \end{bmatrix}$ ,  $\tilde{\beta} = \begin{bmatrix} \tilde{\beta}_1 \\ \tilde{\beta}_2 \end{bmatrix}$

2). homoscedasity

$$\tilde{U}_N = (SSR_U / (N - K - 1)) X'X / N$$

is a symmetrical positive definite matrix computed from the constrained regresson such that  $\tilde{U}_N - U_N \rightarrow 0$

2. Lagrange Multiplier

$$y_i = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon_i + \lambda(\beta_1 + \beta_2 - 1)$$

=>

$$y = X\beta + \epsilon_i + \lambda(R * \beta - 1)$$

1). chi-sq distribution

$$(1/N)(R\tilde{\beta} - 1)(R(X'X)^{-1}\tilde{U}_N(X'X)^{-1}R')^{-1}(R\tilde{\beta} - 1)' \sim \chi^2$$

2). homoscedasity

$\tilde{U}_N$  is a symmetrical positive definite matrix computed from the constrained regresson such that  $\tilde{U}_N - U_N \rightarrow 0$

3. Likelihood Ratio

$$\epsilon_i \sim i.i.d.N(0, \sigma^2)$$

## 1.2 Question 2

2. For the data set **pbp.csv**, can we use the **three test statistics** mentioned in the previous question to test  $H_0 : \beta_1 + \beta_2 = 1$ ? Why? If W and LM are not valid, how can one modify them for the test? What is your conclusion from the valid test?

No, the Wald test and LM test may be invalid. Because there may have heteroscedasticity, the requirements of Wald and LM test (homoscedasticity) are not satisfied. Thus, the external test-heteroscedasticity test should be used before proceeding with the Wald and LM test.

There are two heteroscedasticity tests: White test and Breusch-Pagan-Godfrey Test. But the White test is more general for this linear regression model.

~~~

$$\begin{aligned} \text{equ1:} \quad y_i &= \beta_0 + \beta_1 * x_{1,i} + \beta_2 * x_{2,i} + \epsilon_i \\ \text{equ2:} \quad y_i - x_2 &= \beta_0 + \beta_1(x_1 - x_2) + \epsilon_i \end{aligned}$$

```
pbp=read.csv("/Users/sn0wfree/Dropbox/PhD_1st_study/BST169_Econometrics/Crousework_Project/pbp.csv")
#head(pbp)
#str(pbp)
```

```
require(lmtest)
```

```
## Loading required package: lmtest
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      as.Date, as.Date.numeric
```

```
equ1<-lm(y~x1 + x2,data=pbp)
```

```
equ2<-lm((y-x2)~(x1-x2), data=pbp)
```

```
#Breusch-Pagan-Godfrey Test
```

```
bptest(equ1)
```

```
##  
## studentized Breusch-Pagan test  
##  
## data: equ1  
## BP = 93, df = 2, p-value < 2.2e-16
```

```
bptest(equ2)
```

```
##  
## studentized Breusch-Pagan test  
##  
## data: equ2  
## BP = 0.81949, df = 1, p-value = 0.3653
```

```
#White test
```

```
bptest(residuals(equ1)~x1+x2+x1*x2+x1^2+x2^2,data=pbp)
```

```
##  
## studentized Breusch-Pagan test  
##  
## data: residuals(equ1) ~ x1 + x2 + x1 * x2 + x1^2 + x2^2  
## BP = 96.036, df = 3, p-value < 2.2e-16
```

```
bptest(residuals(equ2)~(x1-x2)+(x1-x2)^2,data=pbp)
```

```
##  
## studentized Breusch-Pagan test  
##  
## data: residuals(equ2) ~ (x1 - x2) + (x1 - x2)^2  
## BP = 0.81949, df = 1, p-value = 0.3653
```

From White test and Breusch-Pagan-Godfrey Test, the **equ1** results reject the NULL hypothesis: Homoscedasticity, Which means the heteroscedasticity exist. And **equ2** do not reject the NULL hypothesis. thus there exist Homoscedasticity

Overall, Wald and LM test is invalid. The original equation: equ1 exist the heteroscedasticity.

Solutaion: Using Weighted Least Squared method to estimated the targeted regression rather than OLS.

### 1.3 Question 3

3. Generate  $y_i$  from the following model,

$$y_i = \beta_0 + \beta_1 * x_{1,i} + (1 - \beta_1) * x_{2,i} + \sqrt{x_{1,i}} * \epsilon_1 \quad (2)$$

where  $x_{1,i}$  follows chi-squared distribution with **2** degrees of freedom. Generate  $\epsilon_1$  from student t distribution with 6 degrees of freedom and  $x_{2,i} \sim U(0, 10)$ . Check whether Wald, LR and LM in Question 1 follow chi-squared distribution by Monte Carlo. (The R command: `ks.test(, 'pchisq', 2)` can be used.) If W and LM are not valid, calculate the correct test statistics and also verify them by Monte Carlo. Please consider different sample sizes.

In text,  $x_1 \sim \chi^2(2)$  and, Generate  $x_2 \sim U(0, 10)$  and  $\epsilon_i \sim T(6)$ .

Here set the sample size with 10 + loop and loop 10000 times. and the equation 2 transform to

$$y_i = \beta_0 + \beta_1 * x_{1,i} + \beta_2 * x_{2,i} + e_i \quad (2.1)$$

where

$$\beta_2 = 1 - \beta_1$$

$$e_i = \sqrt{x_{1,i}} * \epsilon_1$$

And I assume that  $\beta_1 = 0.5$ ,  $\beta_0 = 24$ , and given a fixed value set  $x_1$

from mc1.r mc2.r

lecture Monte Carlo

sample size; estimation:power of test `ks.test(x,"pchisq",2)`

```
require(lmtest)
require(MASS)
```

```
## Loading required package: MASS
```

```
require(stats)
##boost up: translate programme language code into Byte-code.
require(compiler)
```

```
## Loading required package: compiler
```

```
enableJIT(3)
```

```
## [1] 0
```

```
##boost up-end for continues
```

```
#assumption part
```

```
loop=100
```

```
#Warning: the loop time cannot be larger any more;please forgive me, this all my Macbook fault. And the
```

```
beta_1=0.4
```

```
beta_0=1
```

```
x1_store=rchisq(80+20, 2)
```

```
#initial valueset
```

```
original_N=10
```

```
signlevel=0.05
```

```
W_count=rep(0,loop)
```

```
LM_count=rep(0,loop)
```

```
LR_count=rep(0,loop)
```

```
#
```

```
# for loop start:Monte Carlo
```

```
for(j in 1:loop){
```

```
W=rep(0,loop)
```

```
LM=rep(0,loop)
```

```
LR=rep(0,loop)
```

```
N=original_N+j
```

```
for (i in 1:loop){
```

```
#generation part:data
```

```
x1=rchisq(N, 2)
```

```
x2=runif(N,0,10)
```

```
epsilon=rt(N,6)
```

```
y=beta_0+beta_1*x1+(1-beta_1)*x2+sqrt(x1)*epsilon
```

```
#generation part:regression
```

```
equ1<-lm(y~x1+x2)
```

```
equ2<-lm((y-x2)~(x1-x2))
```

```

#calculation
##pre-cal:heteroscedasticity
if (bptest(equ1)$p.value<signlevel){
  equ1<-lm(y~x1+x2,weights=(1/x1^0.5))
  equ2<-lm((y-x2)~(x1-x2),weights=(1/x1^0.5))
}
#calc SSR and Wald,LM, and LR
SSRu=sum(residuals(equ1)^2)
SSRr=sum(residuals(equ2)^2)

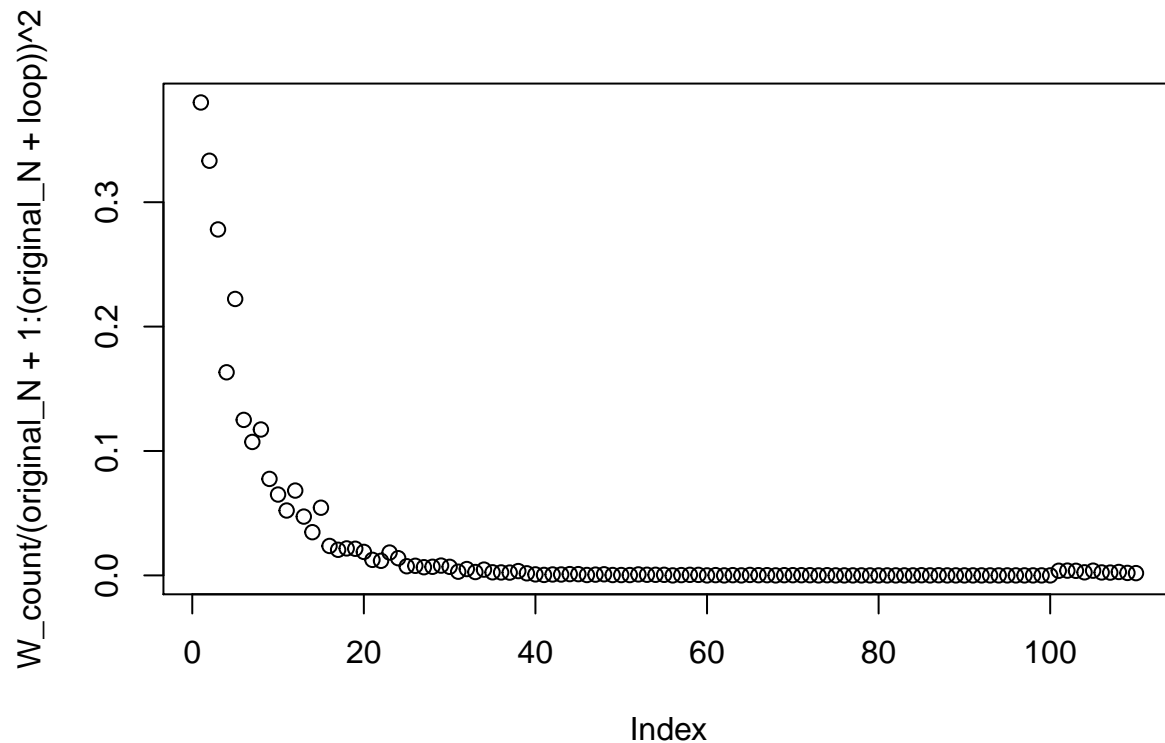
W=N*((SSRr-SSRu)/(SSRu))
LM=N*((SSRr-SSRu)/(SSRr))
LR=N*(log(SSRr/SSRu))

if(ks.test(W,'pchisq',2)$p.value>signlevel){W_count[j]=W_count[j]+1}
if(ks.test(LM,'pchisq',2)$p.value>signlevel){LM_count[j]=LM_count[j]+1}
if(ks.test(LR,'pchisq',2)$p.value>signlevel){LR_count[j]=LR_count[j]+1}
}
}

plot(W_count/(original_N+1:(original_N+loop))^2)

## Warning in W_count/(original_N + 1:(original_N + loop))^2: longer object
## length is not a multiple of shorter object length

```

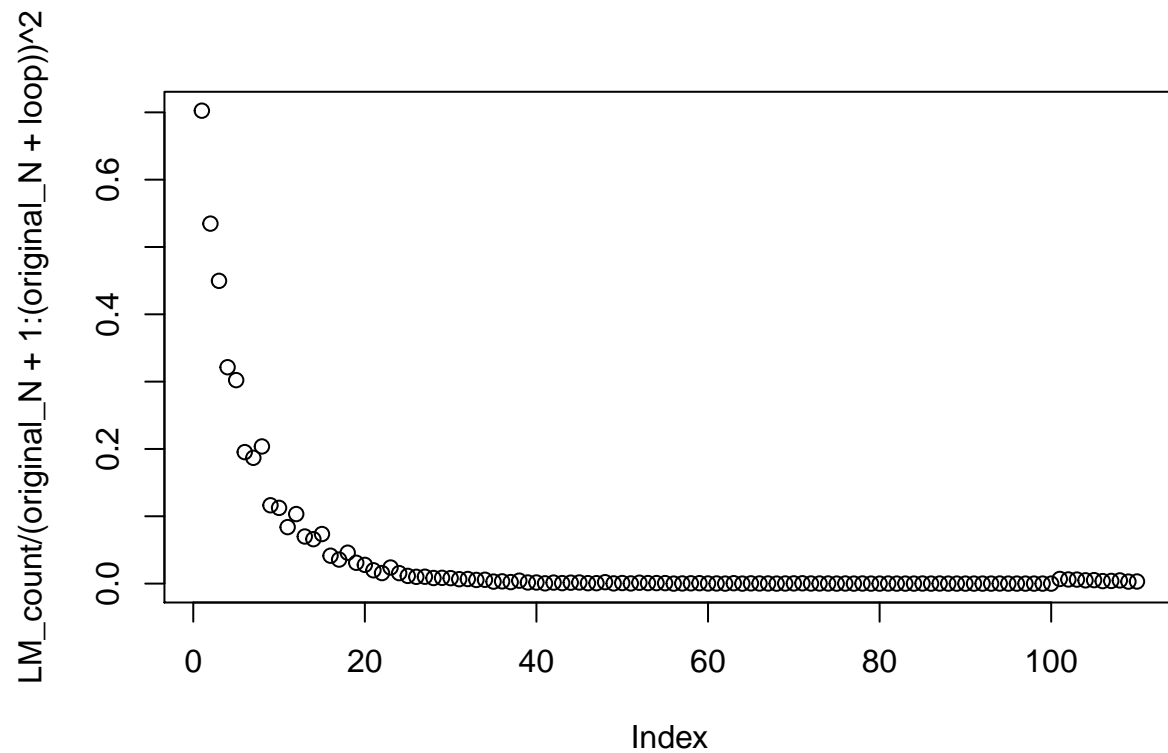


```

plot(LM_count/(original_N+1:(original_N+loop))^2)

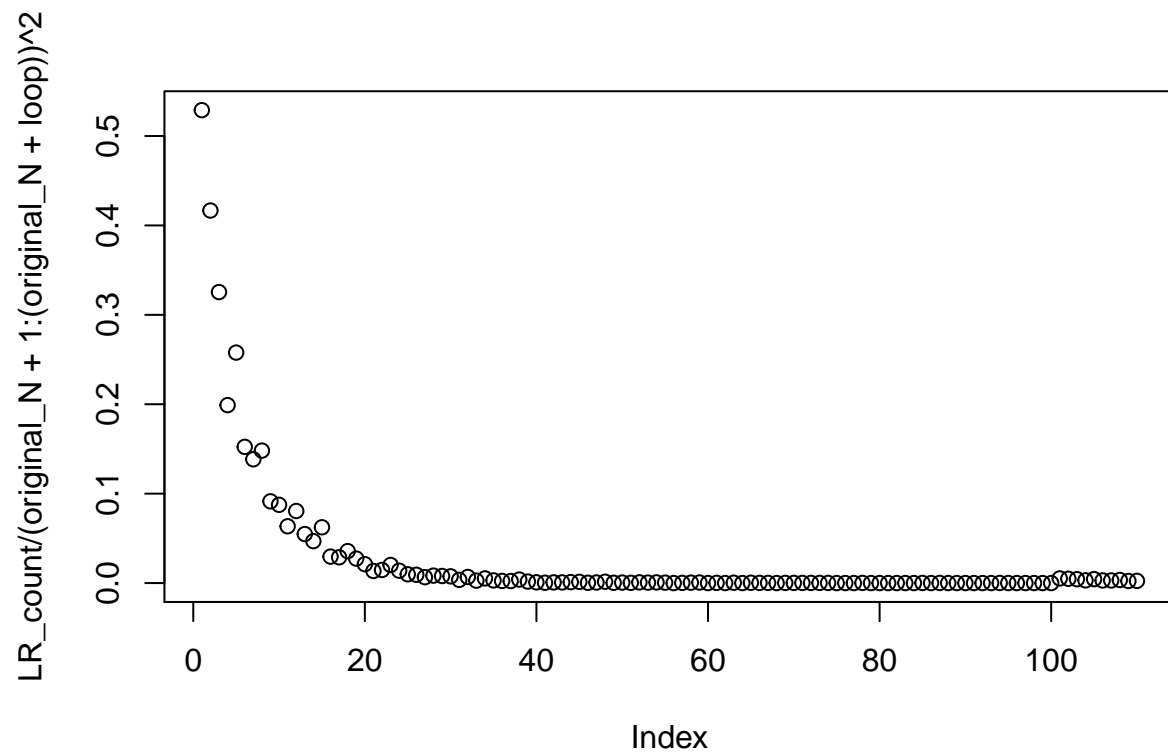
## Warning in LM_count/(original_N + 1:(original_N + loop))^2: longer object
## length is not a multiple of shorter object length

```



```
plot(LR_count/(original_N+1:(original_N+loop))^2)
```

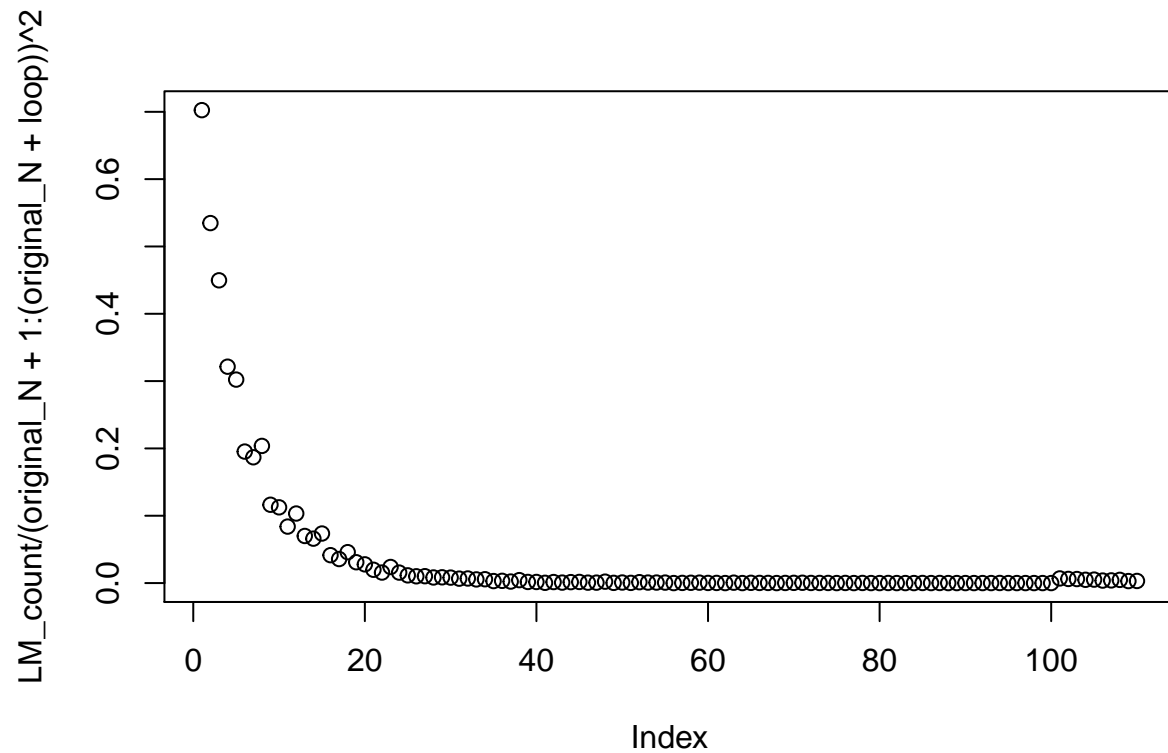
```
## Warning in LR_count/(original_N + 1:(original_N + loop))^2: longer object
## length is not a multiple of shorter object length
```



```
if(ks.test(W,'pchisq',2)p.value < signlevel)print(Walddonot followchi - squiredistribution)elseprint(Walddofollowchi -
squiredistribution)if(ks.test(LM,'pchisq',2)p.value
```

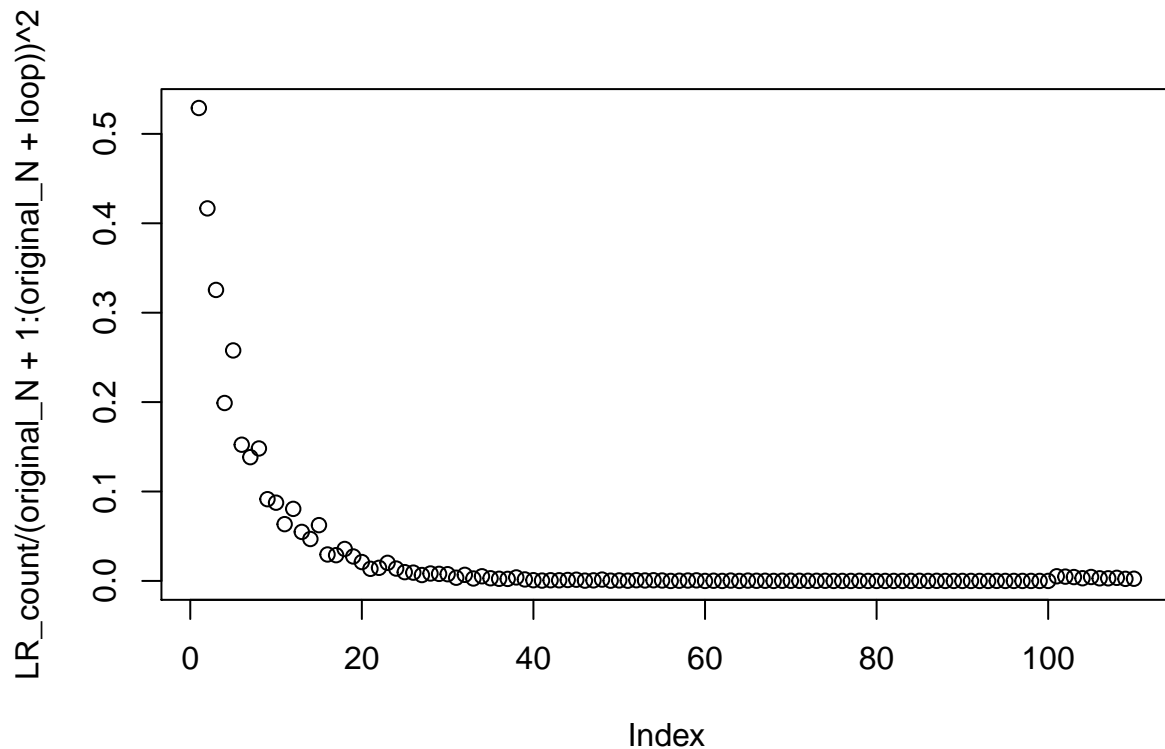
```
plot(LM_count/(original_N+1:(original_N+loop))^2)
```

```
## Warning in LM_count/(original_N + 1:(original_N + loop))^2: longer object  
## length is not a multiple of shorter object length
```



```
plot(LR_count/(original_N+1:(original_N+loop))^2)
```

```
## Warning in LR_count/(original_N + 1:(original_N + loop))^2: longer object  
## length is not a multiple of shorter object length
```



## 1.4 Question 5

For the data set pbp.csv, suppose Equation (2) is the true model. Use proper bootstrapped errors from the true model to study whether different test statistics for  $H_0 : \beta_1 + \beta_2 = 1$  in the previous questions follow chi-squared distribution. Explain your results.

---

reject null, make type I error

bootstrap,

homoscedacity two types of bootstrap robust test bootstrap for  $\epsilon_i$  different performance

t-distribution  $\hat{\epsilon}^2 \Rightarrow$  F distribution single  $\beta \rightarrow$  t-test

$y_i = \beta_0 + \beta_1(x_1 - x_2) + \theta x_2 + \epsilon_i$   $H_0 : \theta = \beta_1 + \beta_2$  use t test for  $\theta$  is as same with f-test with  $\beta_1, \beta_2$

```
require(lmtest)
require(MASS)
require(stats)
##boost up: translate programme language code into Byte-code.
require(compiler)
enableJIT(3)
```

```
## [1] 3
```

```
##boost up-end for continues
```

```
#assumption part
```

```
loop=50
```

```
#Warning: the loop time cannot be larger any more;please forgive me, this all my Macbook fault. And the
```



```

#initial valueset
original_N=10
signlevel=0.05

theta_count=rep(0,loop)

#
# for loop start:Monte Carlo
for(j in 1:loop){
  theta=rep(0,loop)
  N=original_N+j
  for (i in 1:loop){
    #generation part:data
    s=sample(1:length(pbp$y),N,replace = 1)
    question5.y<-pbp$y[s]
    question5.x1<-pbp$x1[s]
    question5.x2<-pbp$x2[s]
    ee=rnorm(N,mean=0,sd=sqrt(var(question5.y)))
    #epsilon=ee/sqrt(x1)
    #y=beta_0+beta_1*x1+(1-beta_1)*x2+sqrt(x1)*epsilon
    #generation part:regression
    question5.equ1<-lm(question5.y~question5.x1+question5.x2)
    #question5.equ2<-lm((question5.y-question5.x2)~(question5.x1-question5.x2))
    #calculation
    ##pre-cal:heteroscedasticity
    if (bptest(question5.equ1)$p.value<signlevel){
      question5.equ1<-lm(question5.y~question5.x1+question5.x2,weights=(1/question5.x1^2))
      #question5.equ2<-lm((question5.y-question5.x2)~(question5.x1-question5.x2),weights=(1/question5.x1^0.
    }
    #calc

    theta[i]=question5.equ1$coefficients[2]+question5.equ1$coefficients[3]

  }
  if(t.test(theta,rep(1,loop))$p.value<signlevel){theta_count[j]=theta_count[j]+1}
}

plot(theta_count/(N+1:(N+loop)^2))

```

