

Econometrics-ps3

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a:

The qfunction we modify as below:

```
qfunction <- function(y,x,theta){  
  N=length(y) #Find sample size  
  #note that x should be N*K and theta should N*1  
  m=x %*% theta#Define the function m(), %*% matrix multiplication, transpose of theta  
  Q=(1/N)*sum((y-m)^2)# This is our objective function  
  out=Q  
}
```

b:

Since this question requires us to derive Hessian matrix: it's just the qderivfun we coded in class, and the analytic form as below:

First, set up the q:

$$q = \frac{(y - m(x, \theta_0))^2}{2} = \frac{(y - \theta_{01} - \theta_{02}priGPA - \theta_{03}ACT)^2}{2}$$

Thus, $\frac{\partial q}{\partial \theta_{01}} = (y - m(x, \theta_0)) \cdot (-1)$, $\frac{\partial q}{\partial \theta_{02}} = (\cdot)(-priGPA)$, and $\frac{\partial q}{\partial \theta_{03}} = (\cdot)(-ACT)$. Therefore, we can derive the Hessian Matrix as:

$$H_i = \begin{bmatrix} 1 & priGPA_i & ACT_i \\ priGPA_i & priGPA_i^2 & priGPA_i \cdot ACT_i \\ ACT_i & ACT \cdot priGPA_i & ACT_i^2 \end{bmatrix}$$

```
qderivfun<-function(y,x,thetahat){  
  N=length(y) #sample size  
  k1=dim(as.matrix(x)) #find number of x-variables  
  k=k1[2]+1 #add to x column to adjust for vector of ones  
  derivstore=matrix(0,N,k)  
  h=.Machine$double.eps^(1/3) #set approximation error  
  x=as.matrix(x) #declare as matrix  
  for (j in 1:k){  
    z=rep(0,k)  
    z[j]=h  
    for (i in 1:N){  
      Qph=qfunction(y[i],cbind(1,t(x[i,])),thetahat+z)  
      Qmh=qfunction(y[i],cbind(1,t(x[i,])),thetahat-z)  
      qderiv=(Qph-Qmh)/(2*h)  
      derivstore[i,j]=qderiv  
    }  
  }
```

```

}
B0hat=(1/N)*(t(derivstore)%*%derivstore) #B0hat from AVAR
ssum=colSums(derivstore)
return(list(derivstore=derivstore,ssum=ssum,B0hat=B0hat))
}

```

##c: For computing sum of hessian:it's the file of **qderivfun2.R**, and the analytical part as below:

$$\sum_{i=1}^N H_i = \begin{bmatrix} N & \sum_{i=1}^N priGPA_i & \sum_{i=1}^N ACT_i \\ \sum_{i=1}^N priGPA_i & \sum_{i=1}^N priGPA_i^2 & \sum_{i=1}^N priGPA_i \cdot ACT_i \\ \sum_{i=1}^N ACT_i & \sum_{i=1}^N ACT_i \cdot priGPA_i & \sum_{i=1}^N ACT_i^2 \end{bmatrix}$$

```

qderivfun2 <- function(y,x,thetahat){
  k=dim(as.matrix(x))#dimension of x-variable
  p=k[2]+1 #set dimension of theta space
  N=length(y)
  derivs=vector(,N)#pre-allocate space
  H=matrix(0,p,p)#Pre-allocate space of hessian matrix
  h=.Machine$double.eps^(1/3)
  for(i in 1:p){
    for (j in 1:p) {
      z=rep(0,p)#create scalar to approximate derivative at each p
      z[j]=h
      theta1=thetahat+z
      theta2=thetahat-z
      qout1=qderivfun(y,x,theta1)
      q1=qout1$derivstor[,i]
      qout2=qderivfun(y,x,theta2)
      q2=qout2$derivstor[,i]
      d=(q1-q2)/(2*h)# Central difference
      H[i,j]=sum(d)#store the hessian matrix
    }
  }
  A0hat=(1/N)*H
  return(list(H=H,A0hat=A0hat))
}

```

##d: Compute the analytical expression for A_0 : $A_0 = E[H(w, \hat{\theta})] = \frac{1}{N} \sum_{i=1}^N H(w_i, \hat{\theta})$.

##e: Estimate the populaton model under consideration:

```

library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.6      v dplyr  1.0.7
## v tidyr   1.1.4      v stringr 1.4.0
## v readr   2.1.1      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

attend<-read_csv("attend.csv")

## Rows: 680 Columns: 11

```

```
## -- Column specification -----
## Delimiter: ","
## dbl (11): attend, termgpa, priGPA, ACT, final, atndrte, hwrte, frosh, soph, ...

##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
attach(attend)

## The following object is masked _by_ .GlobalEnv:
##
##      attend

nlsout=nls(termgpa~(b0+b1*priGPA+b2*ACT),start=list(b0=1,b1=0.04321,b2=.9))
summary(nlsout)

##
## Formula: termgpa ~ (b0 + b1 * priGPA + b2 * ACT)
##
## Parameters:
##      Estimate Std. Error t value Pr(>|t|)
## b0 0.257753    0.150849   1.709    0.088 .
## b1 0.875274    0.042065  20.808   <2e-16 ***
## b2 0.003514    0.006564   0.535    0.593
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5585 on 677 degrees of freedom
##
## Number of iterations to convergence: 1
## Achieved convergence tolerance: 1.036e-07
```

f: Verify the result with the code qderivfun2.R:

```
source('~/.Dropbox/My Mac (Wei's MacBook Air)/Downloads/PhD-Coursework/22Spring/Econometrics/My_solution.
source('~/.Dropbox/My Mac (Wei's MacBook Air)/Downloads/PhD-Coursework/22Spring/Econometrics/My_solution.
source('~/.Dropbox/My Mac (Wei's MacBook Air)/Downloads/PhD-Coursework/22Spring/Econometrics/My_solution

thetahat=cbind(0.257753,0.875274,0.003514)
X=cbind(priGPA,ACT)
out=qderivfun(termgpa,X,t(thetahat))
out1=qderivfun2(termgpa,X,t(thetahat))
out1

## $H
##      [,1]      [,2]      [,3]
## [1,] 1360.000 3518.014 30614.00
## [2,] 3518.014 9503.247 80105.03
## [3,] 30614.000 80105.032 705678.00
##
## $A0hat
##      [,1]      [,2]      [,3]
## [1,] 2.00000 5.17355 45.02059
## [2,] 5.17355 13.97536 117.80152
## [3,] 45.02059 117.80152 1037.76176
```

##g: Calculate the estimated variance-covariance matrix $\hat{Avar}(\hat{\theta})$:

```
avartheta=solve(out1$A0hat)%*% out$B0hat)%*% solve(out1$A0hat)/length(termgpa)
avartheta
```

```
##           [,1]      [,2]      [,3]
## [1,]  0.0214344252 -0.0027355384 -5.955213e-04
## [2,] -0.0027355384  0.0022330729 -1.567609e-04
## [3,] -0.0005955213 -0.0001567609  4.598071e-05
```

##h: Test the null hypothesis:

```
vartheta=diag(avartheta)
se=vartheta^(1/2)
se
```

```
## [1] 0.146405004 0.047255400 0.006780908
```

```
t_stat=thetahat/se
t_stat
```

```
##           [,1]      [,2]      [,3]
## [1,]  1.760548 18.5222  0.5182197
```

Thus, from the results we derived above, we can't accept the alternative condition for $\theta_{01}, \theta_{03} \neq 0$, but another θ , we accept the alternative conditions that they are not equal to 0.