

Practice3 Heteroschedasticity

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November 19, 2012

1. Set M , numbers of Montecarlo simulation equal to 100 and N , numbers of random draws equal to 100. Define

$$y_i = \beta_0 + \beta_1 x_i + u_i$$

where

- $\{y_i, x_i\}$ are i.i.d;
 - $E[x_i, u_i] = 0$;
 - $E[x_i' x_i]$ is full rank;
 - $Var(x_i u_i) = f(x_i) < \infty$.
- (a) Estimate β_1 assuming that $Var(x_i u_i) = f(x_i) = \sigma^2 E(x_i' x_i)$ where σ^2 is the variance of u_i using heteroschedastic and homoschedastic variance estimator. Build a 95% and a 90% c.i. and count the number of times that β is within the interval ;
 - (b) Do the same under the assumption that $Var(x_i u_i) = f(x_i)$. (*HINT: Write $u_i = u_i * \sqrt{x_i^2}$*). What does it change? What do you learn?

Solution:

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\begin{footnotesize}
clear;
clc;
%% Second Practice
%% Answer to the questions:
%
g=1; % Define the numbers of random draws and Montecarlo simulations
test=5; % This define the C.I.
s='homo'; % Define homoschedasticity or heteroschedasticity
if strcmp('heteroschedastic', s)
    display('Heteroschedasticity')
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else
    display('Homoschedasticity')
end
if g==1;
N=1000; % Number of sample draw
M=1000; % Number of time we do the loop- Number of MonteCarlo
else
    N=100;
    M=100;
end
k=1; % Number of coefficient we want to estimate excluding the constant

%% Initialize the Vector and Matrices
beta=NaN(2,1);
Sigma=NaN(2,2);

%% IF:
% We do not want to loose time changing everywhere the name of
% distribution.

    distribution='norm';
    Parameters = [0,1];

%% Beta
% We decide which are going to be the exact value of beta. The first is the
% constant while the second is the one that multiply x-vector.
    Real_Beta= [1 ; 5];

%% Initialize the Confidence Interval

    Intervarl_hom=NaN(M,2);
    Intervarl_het =NaN(M,2);

%%

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    if test==5;
    Number=1.96;
    else
        Number=1.64;
    end

    %% Loop
    for m=1:M
        %% Matrix of x:
        % We create a matrix of x where in the first column there will be all
        % ones while in the second column there will be data distributed as a
        % Normal with mean 5 and variance equal to 1.

        x=[ones(N,1) normrnd(0,1,N,1)]; % X-Data-Vector
        %% u-vector:
        % U vector should be a vector N*1 with zero mean.
        % We want that  $E[u|x]=0$  then  $E[u]=0$  too. Indeed we can apply Law of
        % Iterated Mean to prove that  $E_x[E[u|x]]=E[u]$ .
        % We know that if  $E[u|x]=0 \rightarrow E[u]=0$  that is not equivalent to say if
        %  $E[u]=0 \rightarrow E[u|x]=0$ . But we know that if  $E[u]$  different from 0
        %  $\rightarrow E[u|x]$  different from zero too.
        u= randdraw(distribution, Parameters, [N 1]);
        if strcmp('heteroschedastic', s)
            u= u .*sqrt(x(:,2).^2); % Heteroschedastic Variance
        else
            u=u; % Homoschedastic Case
        end
    end

    %% Y
    % We build y as  $y=x * \beta + u$  where u is orthogonal to x (independent).

    y = x* Real_Beta+ u;

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% We estimate beta as :
% \hat{\beta} = (x'x)^{-1} * x'y
beta= inv(transpose(x)*x)*transpose(x)*y;

%% Estimation of u_hat
u_hat= y - x *beta;

%% Variance
%% Estimation of u_hat variance
variance=(u_hat'*u_hat)/(N-k-1);

%% Variance
% Under Homoschedasticity we estimate the variance matrix of variance-covariance as:
% VAR(\hat{\beta}) = (\frac{1}{N-k-1} \sum_{i=0}^N \hat{u}_i^2) * (x'x)^{-1}
Sigma_hom(:,:)=variance*inv(transpose(x)*x);
%% Heteroschedastic case
XE=x.*[u_hat u_hat];% TEHERE IS a BETTER WAY TO WRITE THIS- UP TO YOU!
Sigma_het (:,:)=(x'*x)\(XE'*XE)/(x'*x);

%% Build the C.I.
Intervarl_hom(m,:)=[(beta(2,1)-Number*sqrt(Sigma_hom(2,2)))...
(beta(2,1)+Number*sqrt(Sigma_hom(2,2)))];
Intervarl_het(m,:)=[(beta(2,1)-Number*sqrt(Sigma_het(2,2)))...
(beta(2,1)+Number*sqrt(Sigma_het(2,2)))];

end

%% Count the Numbers of time that Real Beta is in the C.I.
%i=0;

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%j=0;

Hom=find(Intervarl_hom(:,2)<Real_Beta(k+1,1) | Intervarl_hom(:,1)>(Real_Beta(k+1,1)))
Hom_Prob=1-length(Hom)/ size(Intervarl_hom,1);

Het=find(Intervarl_het(:,2)<Real_Beta(k+1,1) | Intervarl_het(:,1)>(Real_Beta(k+1,1)))
Het_Prob=1-length(Het)/ size(Intervarl_het,1);

%% Display the Result
display('Homoschedastic C. I.')
display(Hom_Prob)
display('Heteroschedastic C. I.')
display(Het_Prob)
\end{footnotesize}

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1 Problem Set 3

2. A standard stata output gives to you the following informations: 1

Linear regression

Number of obs = 706
F(1, 704) = 65.69
Prob > F = 0.0000
R-squared = 0.1033
Root MSE = 421.14

		Robust				
sleep		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
totwrk		-.1507458	.0185992	-8.10	0.000	-.1872623 -.1142294
_cons		3586.377	41.98156	85.43	0.000	3503.953 3668.801

- .
- (a) Write a function in Matlab with the same output (except for F , $Prob > F$ and $Root \text{ } MSE$). (Remember that it should be as flexible as possible).
3. (a) Show analytically and graphically that if n , observations is equal to $k + 1$, numbers of coefficients, then $R^2 = 1$.