

# TTK4215 System Identification and Adaptive Control

## Solution 3

### Instrumental Variables Method

#### Problem 1 (Instrumental Variables Method)

a) From

$$\frac{1}{N} \sum_{t=1}^N \xi(t) \left[ -\varphi^T(t) \tilde{\theta}_N^{IV} + v_0(t) \right] = 0, \quad (1)$$

we get

$$\tilde{\theta}_N^{IV} = \left( \frac{1}{N} \sum_{t=1}^N \xi(t) \varphi^T(t) \right)^{-1} \frac{1}{N} \sum_{t=1}^N \xi(t) v_0(t). \quad (2)$$

Letting  $N \rightarrow \infty$ , we have

$$\lim_{N \rightarrow \infty} \tilde{\theta}_N^{IV} = \left( \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N \xi(t) \varphi^T(t) \right)^{-1} \left( \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N \xi(t) v_0(t) \right). \quad (3)$$

As usual we are assuming quasi-stationary processes, so the limits above can be replaced by the expectation operator to obtain

$$\lim_{N \rightarrow \infty} \tilde{\theta}_N^{IV} = (\bar{E} \xi(t) \varphi^T(t))^{-1} \bar{E} \xi(t) v_0(t). \quad (4)$$

b) Starting from the equation that gives the least-squares estimate

$$\frac{1}{N} \sum_{t=1}^N \varphi(t) \left[ y(t) - \varphi^T(t) \hat{\theta}_N^{LS} \right] = 0, \quad (5)$$

the IV-method simply replaces the first instance of the regression vector with the vector of instrumental variables, to obtain

$$\frac{1}{N} \sum_{t=1}^N \xi(t) \left[ y(t) - \varphi^T(t) \hat{\theta}_N^{IV} \right] = 0. \quad (6)$$

The result is obtained by solving (6) for the IV-estimate  $\hat{\theta}_N^{IV}$ .

c) See solution to Assignment 2.

d) The following MATLAB code generates the data, and solves this assignment (you were not supposed to generate data, of course).

```

% Set true parameters
a=[0.8;0.15];
b=[0.5;-0.3;0.2];
c=[0.1];

% Generate noise, and input signal.
e=random('Normal',0,1,10000,1);
U=random('Normal',0,1,10000,1);

% Set previous data (u and y).
prevu=[0;0;0];
prevy=[0;0];
preve=[0];
N=length(e);
Y=zeros(N,1);

% Generate output.
for t=1:N,
% Compute current y
    y=-a'*prevy+b'*prevu+e(t)+c'*preve;
% Store it.
    Y(t)=y;
% Set previous data:
    prevu=[U(t);prevu(1:2)];
    prevy=[y;prevy(1)];
    preve=[e(t)];
end

% Store data
save Data_Assignment3 U Y;

%Compute LS-estimate from U and Y.

% Initialize the sums of the LS equation
SUM1=zeros(5,5);
SUM2=zeros(5,1);

% Initilize storage
THETALS=[];
THETAIV=[];

% Time loop.
for t=1:N,
    % Regressor
    if (t==1)
        phi=[0;0;0;0;0;0];

```

```

elseif(t==2)
    phi=[-Y(1);0;U(1);0;0];
elseif(t==3)
    phi=[-Y(2);-Y(1);U(2);U(1);0];
else
    phi=[-Y(t-1);-Y(t-2);U(t-1);U(t-2);U(t-3)];
end

% Compute sums
SUM1=SUM1+phi*phi';
SUM2=SUM2+phi*Y(t);

% If j is larger or equal to 10, compute and store estimate.
if (t>=10)
    theta=inv(SUM1)*SUM2;
    THETALS=[THETALS,theta];
end
end

% Plot estimate as function of N.
plot(THETALS');
legend('a1','a2','b1','b2','b3');
grid;
% Display the final estimate
theta
pause;

% Repeat for IV
% First produce x sequence
X=zeros(N,1);

for t=1:N,
% Compute current x
    if (t==1)
        phi=[0;0;0;0;0];
    elseif(t==2)
        phi=[-X(1);0;U(1);0;0];
    elseif(t==3)
        phi=[-X(2);-X(1);U(2);U(1);0];
    else
        phi=[-X(t-1);-X(t-2);U(t-1);U(t-2);U(t-3)];
    end

    if (t>=10)
        x=phi'*THETALS(:,t-9);
    else

```

```

        x=0;
    end
    % Store it.
    X(t)=x;
end

% Time loop.
for t=1:N,
    % Regressor and instrumental variables.
    if (t==1)
        phi=[0;0;0;0;0];
        zeta=[0;0;0;0;0];
    elseif(t==2)
        phi=[-Y(1);0;U(1);0;0];
        zeta=[-X(1);0;U(1);0;0];
    elseif(t==3)
        phi=[-Y(2);-Y(1);U(2);U(1);0];
        zeta=[-X(2);X(1);U(2);U(1);0];
    else
        phi=[-Y(t-1);-Y(t-2);U(t-1);U(t-2);U(t-3)];
        zeta=[-X(t-1);-X(t-2);U(t-1);U(t-2);U(t-3)];
    end

    % Compute sums
    SUM1=SUM1+zeta*phi';
    SUM2=SUM2+zeta*Y(t);

    % If j is larger or equal to 10, compute and store estimate.
    if (t>=10)
        theta=inv(SUM1)*SUM2;
        THETAIV=[THETAIV,theta];
    end
end

% Plot estimate as function of N.
plot(THETAIV');
legend('a1','a2','b1','b2','b3');
grid;
% Display the final estimate
theta

```

## Problem 2 (Extremum Seeking)

The following MATLAB code implements the peak seeking algorithm.

```

function [thetaout,etaout,xiout]=ext_seek(y,thetastart,dt)

persistent t theta eta xi omega omegal omegah k a

if isempty(t)
    % Initialize time
    t=0;

    % Initialize states of extremum seeking controller
    theta=thetastart;
    eta=y;
    xi=0;

    % Define perturbation frequency and amplitude
    omega=0.5;
    a=0.02;

    % Define cutoff frequencies for filters:
    omegal=0.05;
    omegah=0.05;

    % Define gain for update of theta.
    k=0.5;
end

% Extremum seeking:
theta_dot=k*xi;
xi_dot=-omegal*xi+omegal*(y-eta)*a*sin(omega*t);
eta_dot=-omegah*eta+omegah*y;

% Euler integration
theta=theta+dt*theta_dot;
xi=xi+dt*xi_dot;
eta=eta+dt*eta_dot;

% Saturation, if reasonable
%theta=max(theta,0);
%theta=min(theta,1);

% Return variables
thetaout=theta+a*sin(omega*t);
etaout=eta;
xiout=xi;

% Update time variable for next call to ext_seek:
t=t+dt;

```