

```
A = [1.4000 -0.7100 0.1540 -0.0120
      1.0000 0 0 0
      0 1.0000 0 0
      0 0 1.0000 0];
```

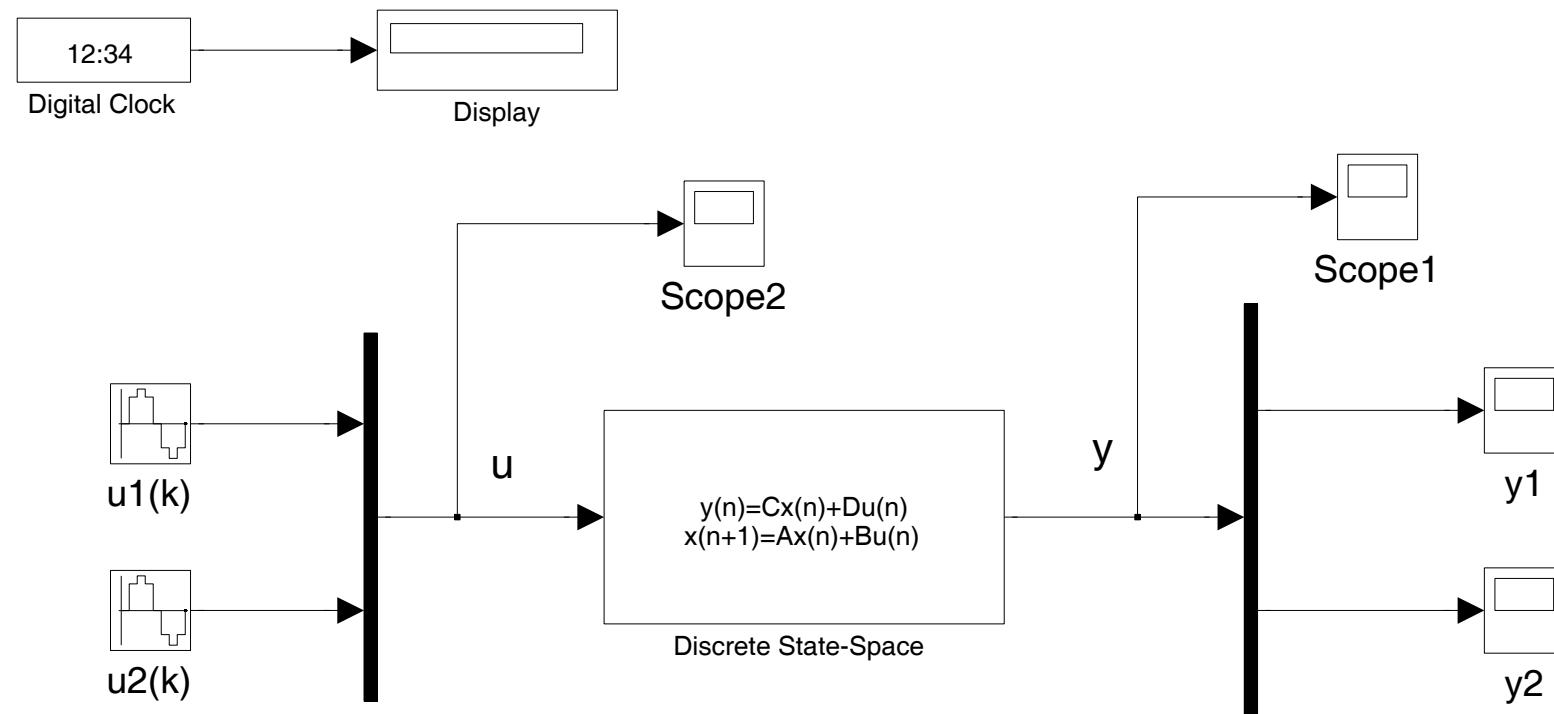
```
B = [0 0
      0.9734 0
      0 0
      -0.8299 1.0000];
```

```
C = [1 0 0 0
      0 1 0 1];
```

```
D = [0 0
      0 0];
```

```
x0 = [0.2 0 0.55 0]
```

```
Ts = 0.01;
```



Esercizio

Dato il sistema nello spazio degli stati a tempo discreto descritto dalle matrici assegnate A , B , C e D , con tempo di campionamento $T_s = 0.01s$, e per una durata della simulazione di $10s$, determinare:

1. Il banco di osservatori con ingressi non noti (UIO) per l'isolamento dei guasti a gradino di ampiezza unitaria che iniziano a $5s$ sui sensori di ingresso; si esegua il progetto degli UIO in Matlab e lo schema della simulazione in Simulink.
2. Il banco di filtri di Kalman per l'isolamento dei guasti a gradino di ampiezza unitaria che iniziano a $5s$ sui sensori d'uscita; i rumori sui sensori sono descritti da processi gaussiani incorrelati a media nulla e varianze secondo i valori:

```
q1 = (0.05)^2; % Rumore su u1  
q2 = (0.03)^2; % Rumore su u2  
r1 = (0.15)^2; % Rumore su y1  
r2 = (0.2)^2; % Rumore su y2
```

Si esegua il progetto dei filtri di Kalman in Matlab e lo schema della simulazione in Simulink.

3. Il banco di reti neurali e sistemi fuzzy con 1 ingresso ed 1 uscita in grado di isolare i guasti sui sensori di ingresso e di uscita. Si utilizzi un tempo di campionamento $T_s = 0.05s$ per una durata della simulazione di almeno $300s$. Si esegua il training in Matlab e gli schemi per le simulazioni in Simulink.

```
%%%  
%%% KF conditions  
%%%  
  
rank(ctrb(A,B(:,1)))  
rank(ctrb(A,B(:,2)))  
rank(obsv(A,C(1,:)))  
rank(obsv(A,C(2,:)))  
  
%%%  
%%% UIO conditions  
%%%  
  
E1 = B(:,1);  
H1 = E1*pinv(C*E1);  
rank(E1)  
rank(C*E1)  
rank(obsv(A-H1*C*A,C))  
  
E2 = B(:,2);  
H2 = E2*pinv(C*E2);  
rank(E2)  
rank(C*E2)  
rank(obsv(A-H2*C*A,C))
```

```
%%%
%%% UIO design for isolation of input faults u1 & u2
%%%

v = [0.1 0.2 0.3 0.4];

%%% UIO decoupled from the 1st input: see matrix Buiol!

E1 = B(:,1);

H1 = E1*pinv(C*E1);

T1 = eye(size(H1*C)) - H1*C;

K11 = place( (A-H1*C*A)' , C' , v )';

F1 = A-H1*C*A - K11*C;

K21 = F1*H1;

K1 = K11 + K21;

Auiol = F1;
Buiol = [T1*B K1];
Cuiol = eye(4);      % Dimension of the matrix A: 4 states!
Duiol = [zeros(4,2) H1]; % 4 outputs and 4 inputs

%%% UIO decoupled from the 2nd input: see matrix Buio2!

E2 = B(:,2);

H2 = E2*pinv(C*E2);

T2 = eye(size(H2*C)) - H2*C;

K12 = place( (A-H2*C*A)' , C' , v )';

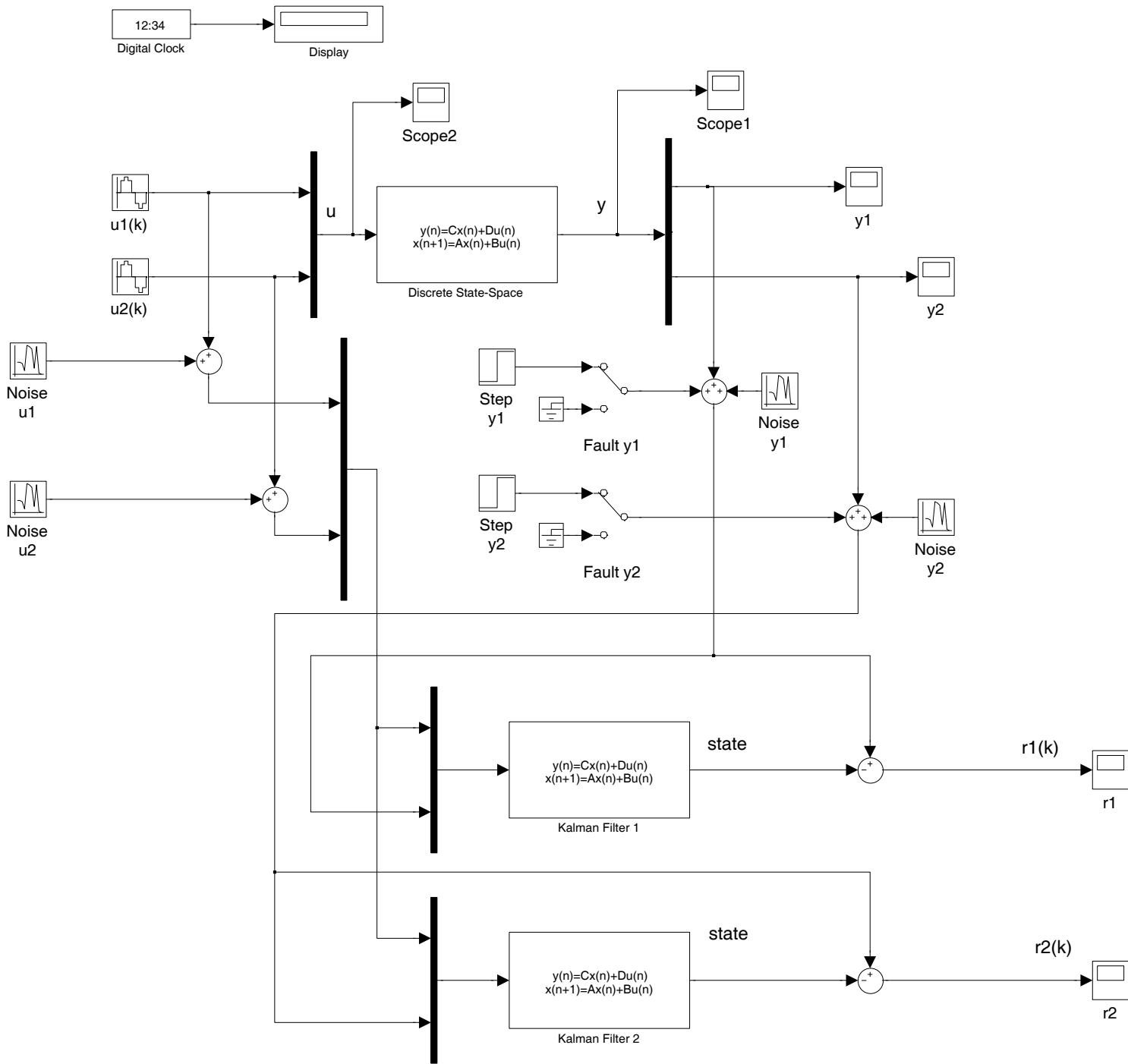
F2 = A-H2*C*A - K12*C;

K22 = F2*H2;

K2 = K12 + K22;

Auio2 = F2;
```

```
Buio2 = [T2*B K2];  
Cuio2 = eye(4); % Dimension of the matrix A: 4 states!  
Duio2 = [zeros(4,2) H2]; % 4 outputs and 4 inputs  
  
return
```



```
%%%
%%% Kalman Filter design example for FDI
%%%

q1 = (0.05)^2; % Noise on the first input u1
q2 = (0.03)^2; % Noise on the second input u2
Q = diag([q1 q2]);

r1 = (0.15)^2; % Noise on the first output y1
r2 = (0.2)^2; % Noise on the second output y2

%%%
%%% Check the controllability of (A,B) and the
%%% observability of (A,C), which must be equal
%%% to the dimension of A
%%%

c1 = C(1,:); % KF for the first output!

[Pkf1,Ekf1,Kkft1] = dare(A',c1',B'*B',r1); % Dual CARE
% for KF1
Kkf1 = Kkft1';

%%% Kalman filter matrices: apart from the Kalman gain,
%%% it is an output observer!

Akf1 = A - Kkf1 * c1;
Bkf1 = [B Kkf1];
Ckf1 = c1;
Dkf1 = zeros(1,3); % 1 output and 3 inputs

c2 = C(2,:); % KF for the second output!

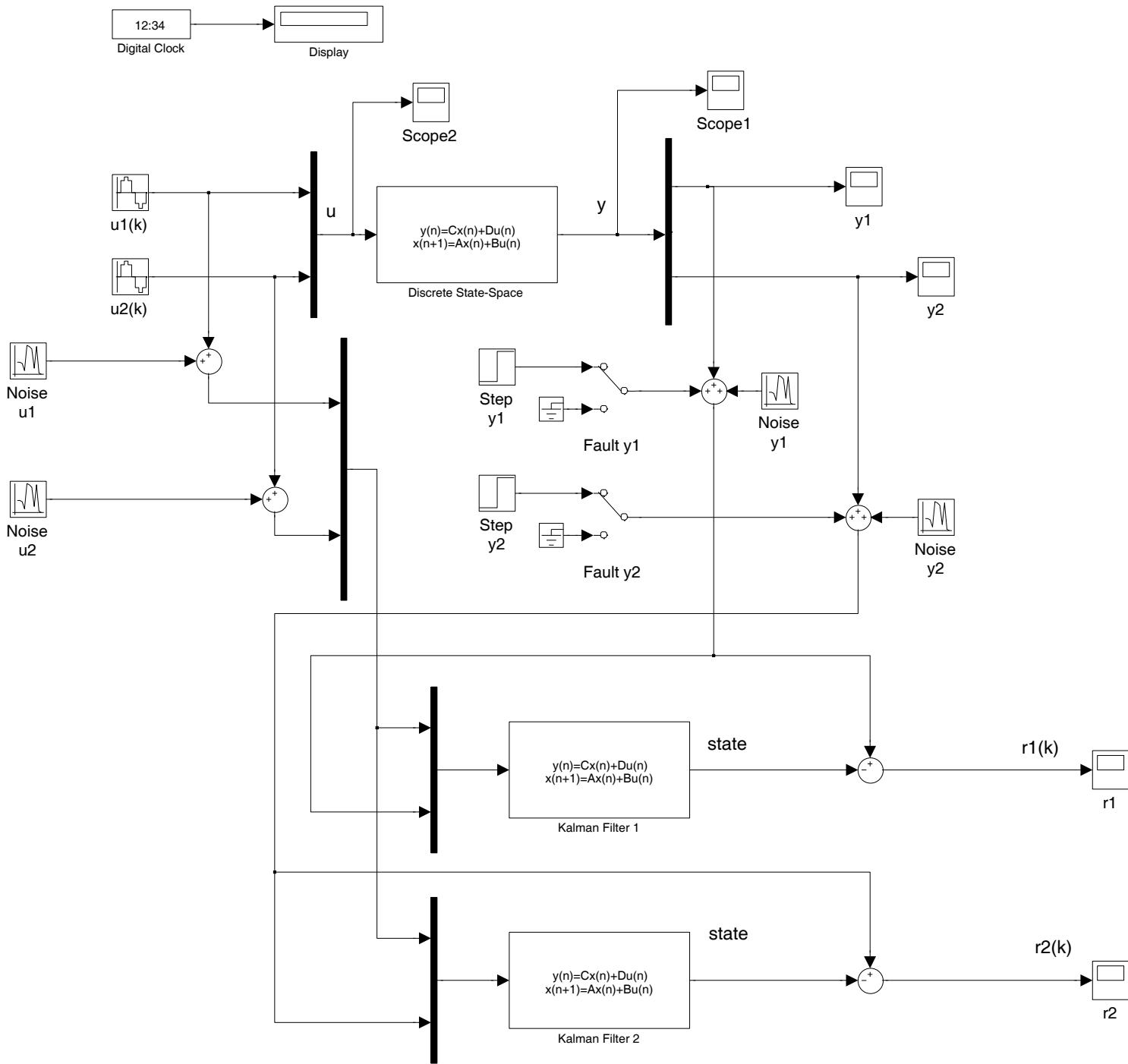
[Pkf2,Ekf2,Kkft2] = dare(A',c2',B'*B',r2); % Dual CARE
% for KF2
Kkf2 = Kkft2';

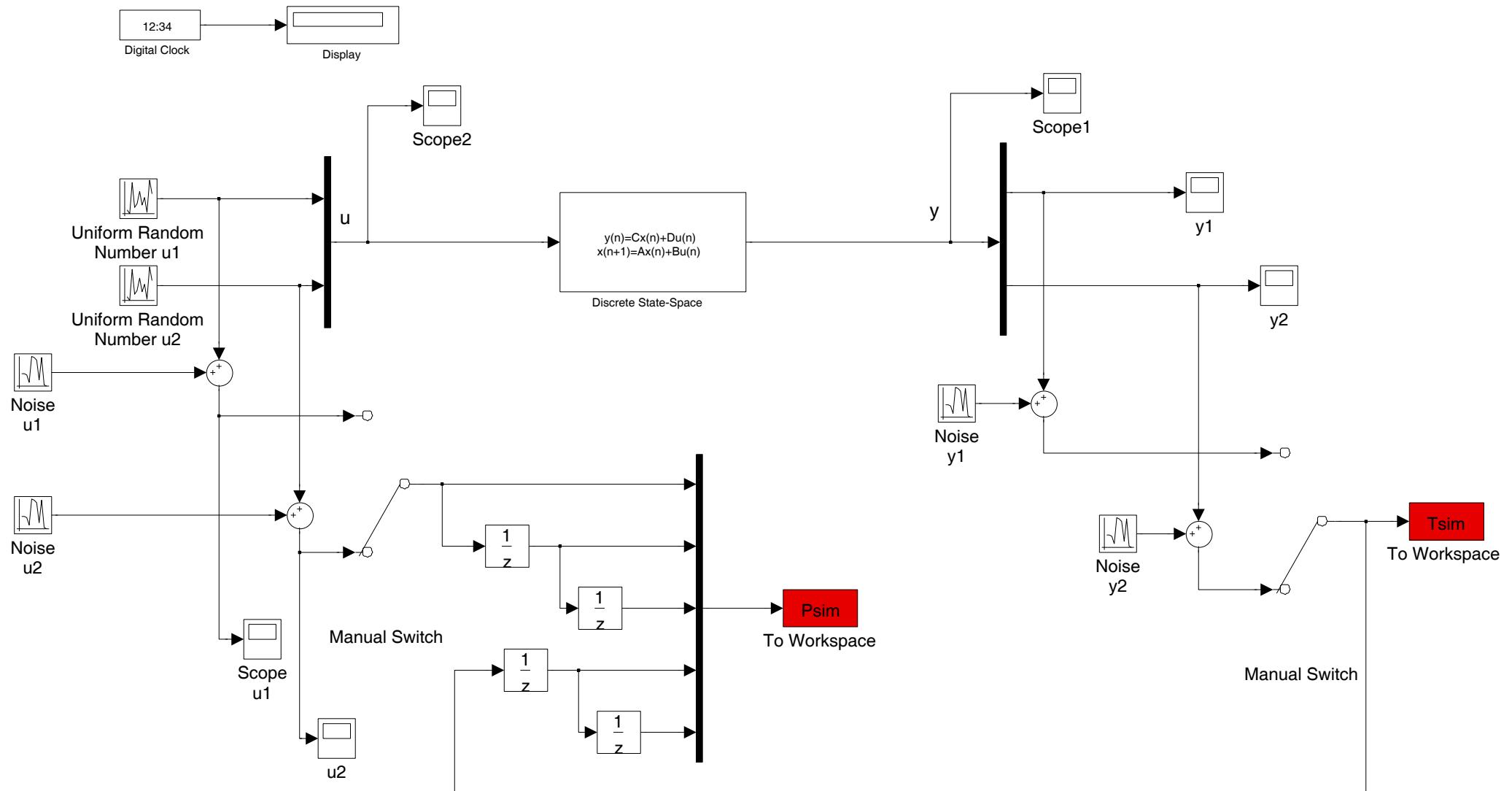
%%% Kalman filter matrices: apart from the Kalman gain,
%%% it is an output observer!

Akf2 = A - Kkf2 * c2;
Bkf2 = [B Kkf2];
Ckf2 = c2;
```

```
Dkf2 = zeros(1,3); % 1 output and 3 inputs
```

```
return
```





```

%%%
%%% File "train_net3.m": neural network training and generation
%%%

% Caricare Psim e Tsim;

P = Psim(1:round(size(Psim,1)/3),:)';
T = Tsim(1:round(size(Psim,1)/3),1)';

%%%
%%% Neural network parameters
%%%

Si = 4; % Number of neurons in the input layer
Sh = 8; % Number of neurons in the hidden layer
So = 1; % Number of neurons in the output layer
           % It is equal to the rows of the matrix T

TFi = 'tansig'; % Sigmoidal tangent activation function
TFh = 'tansig';
TFO = 'purelin'; % Linear activation function

%BTF = 'traingdx'; % Function for the training of the
                   % backpropagation NN, default
BTF = 'trainlm'; % Levenberg-Marquardt backpropagation

BLF = 'learngdm'; % Backpropagation function
                   % weight/bias, default

PF = 'mse'; % Performance function Mean Square Error
             % default

PR = minmax(P); % Equal to: [min(P)' , max(P)' ], it
                  % determines minimal and maximal values of
                  % inputs and output

val.P = Psim(round(size(Psim,1)/3)+1:2*round(size(Psim,1)/3),:)';
           % validation data
val.T = Tsim(round(size(Psim,1)/3)+1:2*round(size(Psim,1)/3),:)';
test.P = Psim(2*round(size(Psim,1)/3)+1:end,:); % test data
test.T = Tsim(2*round(size(Psim,1)/3)+1:end,:);

%net = newff(P,T,[Si Sh So],{TFi TFh TFO},BTF,BLF,PF);
           % Note: it generates a NN

```

```
% with 4 layers!!!
net = newff(P,T,[Si Sh],{TFi TFh TFO},BTF,BLF,PF);

%%% Parameters for the NN training
%%%

net.trainParam.epochs = 300;      % Number of epochs
net.trainParam.goal    = 1e-4;    % Value of the final error
net.trainParam.show    = 1;       % Show the plot after 1 epoch
net.trainParam.lr     = 0.05;    % Learning rate for trainlm function
net.trainParam.mc     = 0.9;     % Momentum constant: gradient value
                                % during the training phase: if 0 ->
                                % weights are changed only on the basis
                                % of the gradient; if 1 -> gradient
                                % function is completely neglected

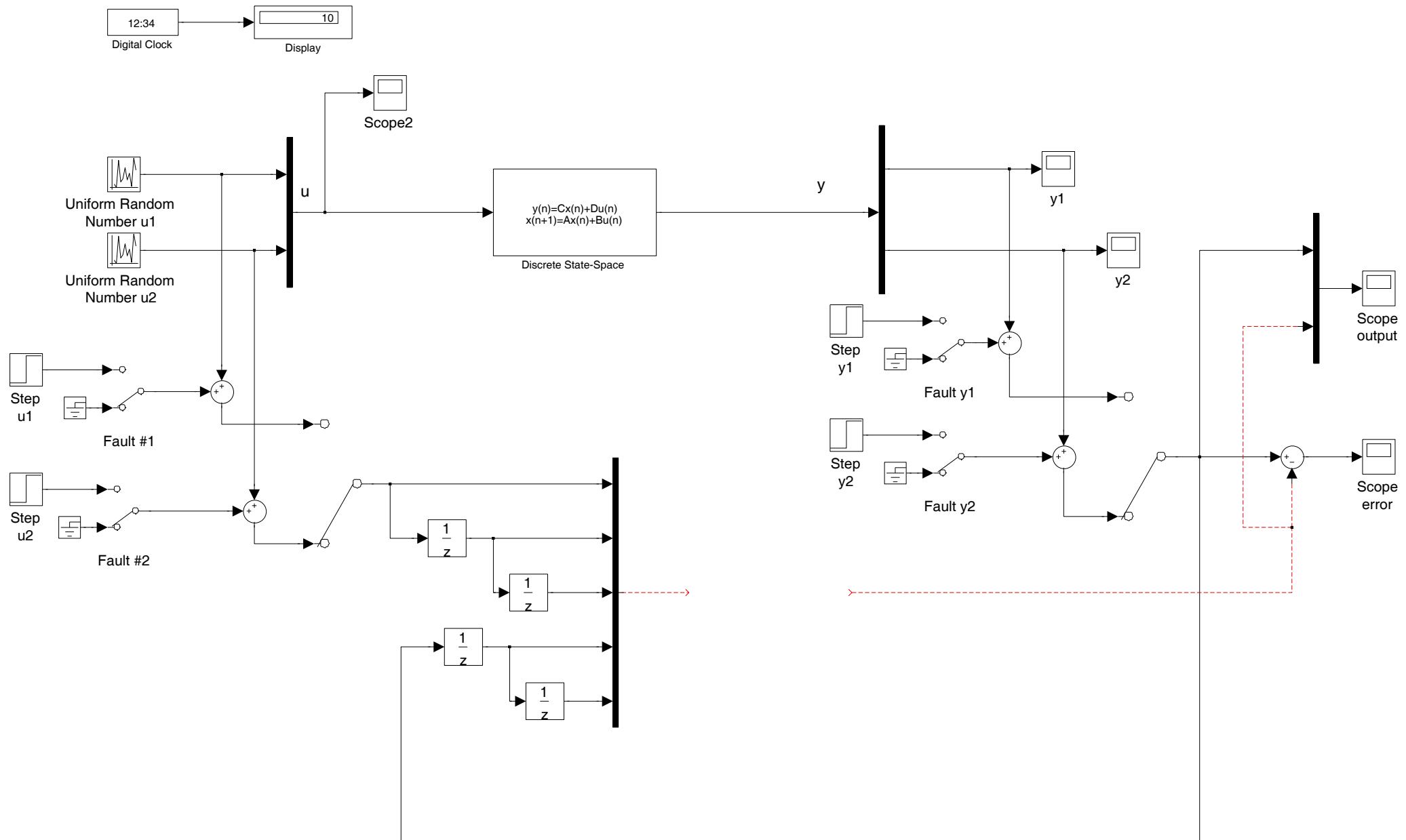
net = train(net,P,T,[],[],val,test); % training function

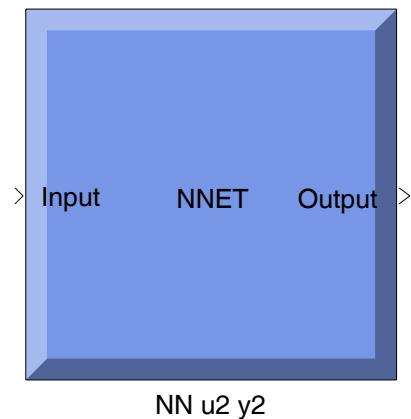
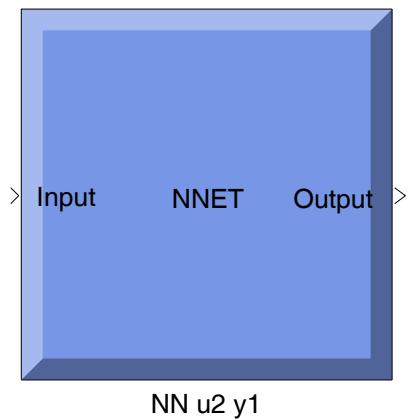
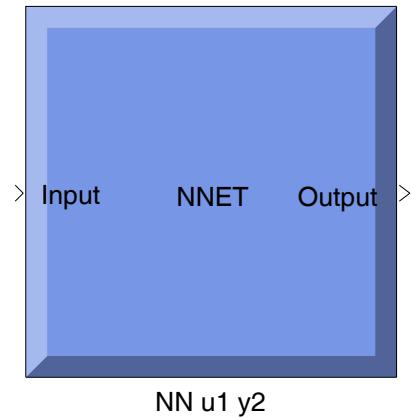
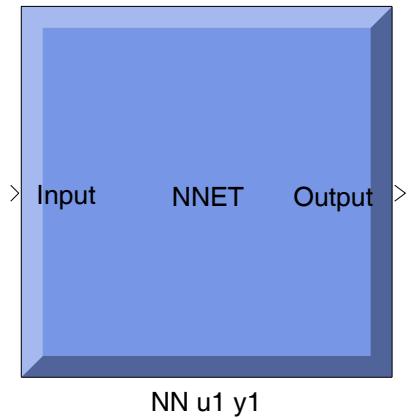
%Ts = ...; % Sampling time
% NOTE: it should be equal to the sampling time used
% for collecting the matrices Tsim and Psim!!!

net.sampleTime = Ts;

gensim(net,Ts); % It creates the neural network in Simulink

return
```





```
%%%  
%%% File "gen_data_train_fuzzy.m": partiziona opportunamente i dati  
%%% per il training del sistema fuzzy  
%%%  
  
% Caricare P e T che devono essere nel workspace.  
% UYtrain: dati di training  
% UYval : dati di validazione  
% UYtest : dati di test  
  
UYtrain = [Psim(1:round(size(Psim,1)/3),:)...  
          Tsim(1:round(size(Psim,1)/3),1)];  
  
UYval = [Psim(round(size(Psim,1)/3)+1:2*round(size(Psim,1)/3),:)...  
          Tsim(round(size(Psim,1)/3)+1:2*round(size(Psim,1)/3),:)];  
  
UYtest = [Psim(2*round(size(Psim,1)/3)+1:end,:)...  
          Tsim(2*round(size(Psim,1)/3)+1:end,:)];  
  
return
```

