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ภาคผนวก ก

โปรแกรม การจำลองบนคอมพิวเตอร์

การสร้างอินพุทของระบบ

1. กรณีที่มี 2 observers

```
% DIRECTION FINDING PROJECT
% 2 PASSIVE STATION SIMULATION MODULE
% CREATE BEARING ANGLES
```

```
% initialize all variables
```

```
% Station location
```

```
source=[15 15];
```

```
station_i=[0 0;0 0]; % stations' location matrix
```

```
    a=[0 0 ]; % angle matrix
```

```
    row=2; % define number of row
```

```
for j=1:row
```

```
    a(1,j)=input('angle ');
```

```
    station_i(j,1)=input('x position ');
```

```
    station_i(j,2)=input('y position ');
```

```
end
```

```
% start 1 to 40 observations
```

```
for obsv =1:40 % 40 observations
```

```

% A 100 montecarlo run start point

reff=1;      % first point for shifting
for monte=1:100    % 100 monte carlo runs

    N=randn(observ,2) ; % Normalized Gaussian Random Variable

% create matrix b,n and theta

for i=1:obsv
    for j=1:2
        b(i,j)=0; % matrix for bearing angle
        n(i,j)=0; % matrix for normalized Gaussian noise
        theta(i,j)=0; % matrix for real angle
    end
end

for i=1:obsv
    theta(i,1)=a(1,1); % copy real angles
    theta(i,2)=a(1,2);
end

n=N;      % Gaussian noise
b=theta+n; % add noise

% store n into Zi
z=n;      % noise

if observ ~=1
    z=(sum(z))/obsv;
end

```

% Noise Storage Matrix

ZI(obsv,1)=z(1,1); % noise bearing matrix for all

ZI(obsv,2)=z(1,2);

if obsv ~= 1

 sZI=sum(ZI)/obsv;

else

 sZI=ZI;

end

Zi(obsv,reff)=sZI(1,1);

Zi(obsv,reff+1)=sZI(1,2);

if obsv ~=1

 b=(sum(b))/obsv;

end

% BEARING ANGLE Storage Matrix

% store b into Bi

Bi(obsv,reff) =b(1,1); % bearing matrix for all

Bi(obsv,reff+1)=b(1,2); % next processes.

reff=reff+2; % 2steps shift

end % end a 100 monte carlo runs

end % end of 40 observations loop

%=====end of bearing random=====



2. กรณีที่มี 3 observers

```

% DIRECTION FINDING PROJECT
% 3 PASSIVE STATION SIMULATION MODULE
% CREATE BEARING ANGLES

% initialize all variables

% Station location
source=[15 15];

station_i=[0 0;0 0]; % stations' location matrix
a=[0 0 ]; % angle matrix
row=3; % define number of row

for j=1:row
a(1,j)=input('angle ');
station_i(j,1)=input('x position ');
station_i(j,2)=input('y position ');
end

% start 1 to 40 observations

for obsv =1:40 % 40 observations

% A 100 montecarlo run start point

reff=1; % first point for shifting
for monte=1:100 % 100 monte carlo runs

N=randn(obsv,3) ; % Normalized Gaussian Random Variable

```

```

% create matrix b,n and theta

for i=1:obsv
    for j=1:3
        b(i,j)=0; % matrix for bearing angle
        n(i,j)=0; % matrix for normalized Gaussian noise
        theta(i,j)=0; % matrix for real angle
    end
end

for i=1:obsv
    theta(i,1)=a(1,1); % copy real angles
    theta(i,2)=a(1,2);
    theta(i,3)=a(1,3);
end

n=N; % Gaussian noise
b=theta+n; % add noise

% store n into Zi
z=n; % noise

if obsv ~=1
    z=(sum(z))/obsv;
end

% Noise Storage Matrix
ZI(obsv,1)=z(1,1); % noise bearing matrix for all
ZI(obsv,2)=z(1,2);
ZI(obsv,3)=z(1,3);

```

```

if obsv ~= 1
    sZI=sum(ZI)/obsv;
else
    sZI=ZI;
end

Zi(obsv,reff)=sZI(1,1);
Zi(obsv,reff+1)=sZI(1,2);
Zi(obsv,reff+2)=sZI(1,3);

if obsv ~=1
    b=(sum(b))/obsv;
end

% BEARING ANGLE Storage Matrix
% store b into Bi
Bi(obsv,reff) =b(1,1); % bearing matrix for all
Bi(obsv,reff+1)=b(1,2); % next processes.
Bi(obsv,reff+2)=b(1,3);

reff=reff+3;    % 2steps shift

end % end a 100 montecarlo run

end % end of observation loop

%=====end of bearing random=====

```


3. กรณีที่มี 4 observers

```

% DIRECTION FINDING PROJECT
% 4 PASSIVE STATION SIMULATION MODULE
% CREATE BEARING ANGLES

% initialize all variables

% Station location
source=[15 15];

station_i=[0 0;0 0]; % stations' location matrix
a=[0 0 ]; % angle matrix
row=4; % define number of row

for j=1:row
a(1,j)=input('angle ');
station_i(j,1)=input('x position ');
station_i(j,2)=input('y position ');
end

% start 1 to 40 observations

for obsv =1:40 % 40 observations

% A 100 montecarlo run start point

reff=1; % first point for shifting
for monte=1:100 % 100 monte carlo runs

N=randn(obsv,4) ; % Normalized Gaussian Random Variable

```

```

% create matrix b,n and theta

for i=1:obsv
    for j=1:4
        b(i,j)=0; % matrix for bearing angle
        n(i,j)=0; % matrix for normalized Gaussian noise
        theta(i,j)=0; % matrix for real angle
    end
end

for i=1:obsv
    theta(i,1)=a(1,1); % copy real angles
    theta(i,2)=a(1,2);
    theta(i,3)=a(1,3);
    theta(i,4)=a(1,4);
end

n=N; % Gaussian noise
b=theta+n; % add noise

% store n into Zi
z=n; % noise

if obsv ~=1
    z=(sum(z))/obsv;
end

% Noise Storage Matrix
ZI(obsv,1)=z(1,1); % noise bearing matrix for all
ZI(obsv,2)=z(1,2);
ZI(obsv,3)=z(1,3);
ZI(obsv,4)=z(1,4);

```

```

if obsv ~= 1
    sZI=sum(ZI)/obsv;
else
    sZI=ZI;
end

Zi(obsv,reff)=sZI(1,1);
Zi(obsv,reff+1)=sZI(1,2);
Zi(obsv,reff+2)=sZI(1,3);
Zi(obsv,reff+3)=sZI(1,4);

if obsv ~=1
    b=(sum(b))/obsv;
end

% BEARING ANGLE Storage Matrix
% store b into Bi
Bi(obsv,reff) =b(1,1); % bearing matrix for all
Bi(obsv,reff+1)=b(1,2); % next processes.
Bi(obsv,reff+2)=b(1,3);
Bi(obsv,reff+3)=b(1,4);

reff=reff+2;    % 2steps shift

end % end a 100 montecarlo run

end % end of observation loop

%=====end of bearing random=====

```

ภาคผนวก ข

โปรแกรม การจำลองบนคอมพิวเตอร์

การประมาณหาที่ตั้งแหล่งกำเนิดสัญญาณ โดยวิธี EXTENDED KALMAN FILTER

1. กรณีที่มี 2 observers

% Direction Finding Project

% 2 passive station simulation module

% estimator: EXTENDED KALMAN FILTER

% find priori estimate position

% Station Location

source=[15 15];

% find EXTENDED KALMAN FILTER

for obsv =1:40

for monte =1:2:200

X00=[x; y]; % x,y initial

Xe=[0;0]; % x,y at emitter

I=[1 0;0 1]; % identity matrix

Xk1=[0;0]; % location of sensor1

Xk2=[station_i(2,1);0]; % location of sensor2

% transfer bearing to new variables

Zn(1,1)=Zi(obsv,monete)*pi/180;

```

Zn(2,1)=Zi(observ,monte+1)*pi/180;
% find nonlinear function

u1=(X00(2,1)-Xk1(2,1))/(X00(1,1)-Xk1(1,1));
u2=(X00(2,1)-Xk2(2,1))/(X00(1,1)-Xk2(1,1));

h11=(-u1/(1+u1^2))*(1/(X00(1,1)-Xk1(1,1)));
h12=(1/(1+u1^2))*(1/(X00(1,1)-Xk1(1,1)));

h21=(-u2/(1+u2^2))*(1/(X00(1,1)-Xk2(1,1)));
h22=(1/(1+u2^2))*(1/(X00(1,1)-Xk2(1,1)));

An(1,1)=h11;
An(1,2)=h12;

An(2,1)=h21;
An(2,2)=h22;

Wn=inv((An)'*An)*(An)'*Zn;

% find optimum estimate position
Xe=I*X00+Wn;
X00=Xe;

% save squar error to stack
stack((monte+1)/2,1)=((Xe(1,1)-source(1,1)))^2; % x error
stack((monte+1)/2,2)=((Xe(2,1)-source(1,2)))^2; % y error

end % end of fetching loop

% find mean square error of 100 monte
msekf=sum(stack)/100;

```

```
% mean sq.error of 100runs, 40 observations
```

```
ekf(obsv,1)=msekf(1,1);
```

```
ekf(obsv,2)=msekf(1,2);
```

```
if obsv == 1
```

```
    EKF2(obsv,1)=ekf(obsv,1);
```

```
    EKF2(obsv,2)=ekf(obsv,2);
```

```
else
```

```
    buff=sum(ekf)/obsv ;
```

```
    EKF2(obsv,1)=buff(1,1);
```

```
    EKF2(obsv,2)=buff(1,2);
```

```
end
```

```
end % end of 40 observations
```

```
%=====end of Extended Kalman Filter=====
```

2. กรณีที่มี 3 observers

```

% Direction Finding Project
% 3 passive station simulation module
% estimator: EXTENDED KALMAN FILTER

% find priori estimate position

% Station Location
source=[15 15];

% find EXTENDED KALMAN FILTER

for obsv =1:40
    for monte =1:3:300

        X00=[x; y];    % x,y initial
        %Xe=[0;0];    % x,y optimum
        I=[1 0;0 1];  % identity matrix

        Xk1=[0;0];    % location of sensor1
        Xk2=[station_i(2,1);0]; % location of sensor2
        Xk3=[station_i(3,1);0]; % location of sensor3

% transfer bearing to new variables

Zn(1,1)=Zi(obsv,monte)*pi/180;
Zn(2,1)=Zi(obsv,monte+1)*pi/180;
Zn(3,1)=Zi(obsv,monte+2)*pi/180;

```

% find nonlinear function

$$u1=(X00(2,1)-Xk1(2,1))/(X00(1,1)-Xk1(1,1));$$

$$u2=(X00(2,1)-Xk2(2,1))/(X00(1,1)-Xk2(1,1));$$

$$u3=(X00(2,1)-Xk3(2,1))/(X00(1,1)-Xk3(1,1));$$

$$h11=(-u1/(1+u1^2))*(1/(X00(1,1)-Xk1(1,1)));$$

$$h12=(1/(1+u1^2))*(1/(X00(1,1)-Xk1(1,1)));$$

$$h21=(-u2/(1+u2^2))*(1/(X00(1,1)-Xk2(1,1)));$$

$$h22=(1/(1+u2^2))*(1/(X00(1,1)-Xk2(1,1)));$$

$$h31=(-u3/(1+u3^2))*(1/(X00(1,1)-Xk3(1,1)));$$

$$h32=(1/(1+u3^2))*(1/(X00(1,1)-Xk3(1,1)));$$

$$An(1,1)=h11;$$

$$An(1,2)=h12;$$

$$An(2,1)=h21;$$

$$An(2,2)=h22;$$

$$An(3,1)=h31;$$

$$An(3,2)=h32;$$

$$Wn=inv((An)'*An)*(An)'*Zn;$$

% find optimum estimate position

$$Xe=I*X00+Wn;$$

$$X00=Xe;$$


```

% save squar error to stack
    stack((monte+1)/2,1)=((Xe(1,1)-source(1,1)))^2; % x error
    stack((monte+1)/2,2)=((Xe(2,1)-source(1,2)))^2; % y error

    end % end of fetching loop

% find mean square error of 100 monte

    msekf=sum(stack)/100;

% mean sq.error of 100runs, 40 observations
    ekf(obsv,1)=msekf(1,1);
    ekf(obsv,2)=msekf(1,2);

    if obsv == 1
        EKF3(obsv,1)=ekf(obsv,1);
        EKF3(obsv,2)=ekf(obsv,2);
    else
        buff=sum(ekf)/obsv ;
        EKF3(obsv,1)=buff(1,1);
        EKF3(obsv,2)=buff(1,2);
    end

end % end of 40 observations

%=====end of Extended Kalman Filter=====

```

3. กรณีที่มี 4 observers

```

% Direction Finding Project
% 4 passive station simulation module
% estimator: EXTENDED KALMAN FILTER

% find priori estimate position

% Station Location
source=[15 15];

% find EXTENDED KALMAN FILTER

for obsv =1:40
    for monte =1:4:400

        X00=[x; y];    % x,y initial
        Xe=[0;0];     % x,y optimum
        I=[1 0;0 1];  % identity matrix

        Xk1=[0;0];    % location of sensor1
        Xk2=[station_i(2,1);0]; % location of sensor2
        Xk3=[station_i(3,1);0]; % location of sensor3
        Xk4=[station_i(4,1);0]; % location of sensor4

% transfer bearing to new variables

Zn(1,1)=Zi(obsv,monte)*pi/180;
Zn(2,1)=Zi(obsv,monte+1)*pi/180;
Zn(3,1)=Zi(obsv,monte+2)*pi/180;
Zn(4,1)=Zi(obsv,monte+3)*pi/180;

```

% find nonlinear function

$$u1=(X00(2,1)-Xk1(2,1))/(X00(1,1)-Xk1(1,1));$$

$$u2=(X00(2,1)-Xk2(2,1))/(X00(1,1)-Xk2(1,1));$$

$$u3=(X00(2,1)-Xk3(2,1))/(X00(1,1)-Xk3(1,1));$$

$$u4=(X00(2,1)-Xk4(2,1))/(X00(1,1)-Xk4(1,1));$$

$$h11=(-u1/(1+u1^2))*(1/(X00(1,1)-Xk1(1,1)));$$

$$h12=(1/(1+u1^2))*(1/(X00(1,1)-Xk1(1,1)));$$

$$h21=(-u2/(1+u2^2))*(1/(X00(1,1)-Xk2(1,1)));$$

$$h22=(1/(1+u2^2))*(1/(X00(1,1)-Xk2(1,1)));$$

$$h31=(-u3/(1+u3^2))*(1/(X00(1,1)-Xk3(1,1)));$$

$$h32=(1/(1+u3^2))*(1/(X00(1,1)-Xk3(1,1)));$$

$$h41=(-u4/(1+u4^2))*(1/(X00(1,1)-Xk4(1,1)));$$

$$h42=(1/(1+u4^2))*(1/(X00(1,1)-Xk4(1,1)));$$

$$An(1,1)=h11;$$

$$An(1,2)=h12;$$

$$An(2,1)=h21;$$

$$An(2,2)=h22;$$

$$An(3,1)=h31;$$

$$An(3,2)=h32;$$

$$An(4,1)=h41;$$

$$An(4,2)=h42;$$

$$Wn=inv((An)*An)*(An)^*Zn;$$

```

% find optimum estimate position
    Xe=I*X00+Wn;
    X00=Xe;

% save squar error to stack
    stack((monte+1)/2,1)=((Xe(1,1)-source(1,1)))^2; % x error
    stack((monte+1)/2,2)=((Xe(2,1)-source(1,2)))^2; % y error

end % end of fetching loop

% find mean square error of 100 monte

    msekf=sum(stack)/100;

% mean sq.error of 100runs, 40 observations
    ekf(obsv,1)=msekf(1,1);
    ekf(obsv,2)=msekf(1,2);

if obsv == 1
    EKF4(obsv,1)=ekf(obsv,1);
    EKF4(obsv,2)=ekf(obsv,2);
else

    buff=sum(ekf)/obsv ;
    EKF4(obsv,1)=buff(1,1);
    EKF4(obsv,2)=buff(1,2);
end

end % end of 40 observations

%=====end of Extended Kalman Filter=====

```



ประวัติผู้เขียน

ร้อยเอก สุรเดช เคารพครู เกิดวันที่ 26 มิถุนายน พ.ศ.2511 ที่จังหวัด กรุงเทพมหานคร สำเร็จการศึกษาปริญญาวิทยาศาสตรบัณฑิต (สาขา วิศวกรรมไฟฟ้าสื่อสาร) จากโรงเรียน นายร้อย พระจุลจอมเกล้า รุ่นที่ 38 เมื่อปี พ.ศ.2534 และเข้าศึกษาต่อในหลักสูตร วิศวกรรมศาสตรมหาบัณฑิต สาขาวิศวกรรมไฟฟ้า ภาควิชา วิศวกรรมไฟฟ้า จุฬาลงกรณ์มหาวิทยาลัย เมื่อปี พ.ศ. 2536 ปัจจุบัน ปฏิบัติราชการเป็น อาจารย์ประจำแผนกวิชาสงครามอิเล็กทรอนิกส์ โรงเรียนทหารสื่อสาร กรมการทหารสื่อสาร เขต ดุสิต กรุงเทพมหานคร