

b) Define uncertainty of the platform by “sampling” algorithms platform Platform model: Deadreckoning.

A GUI has been designed for deploying both Deadreckoning and Odometry models. It can be seen in Figure 1.

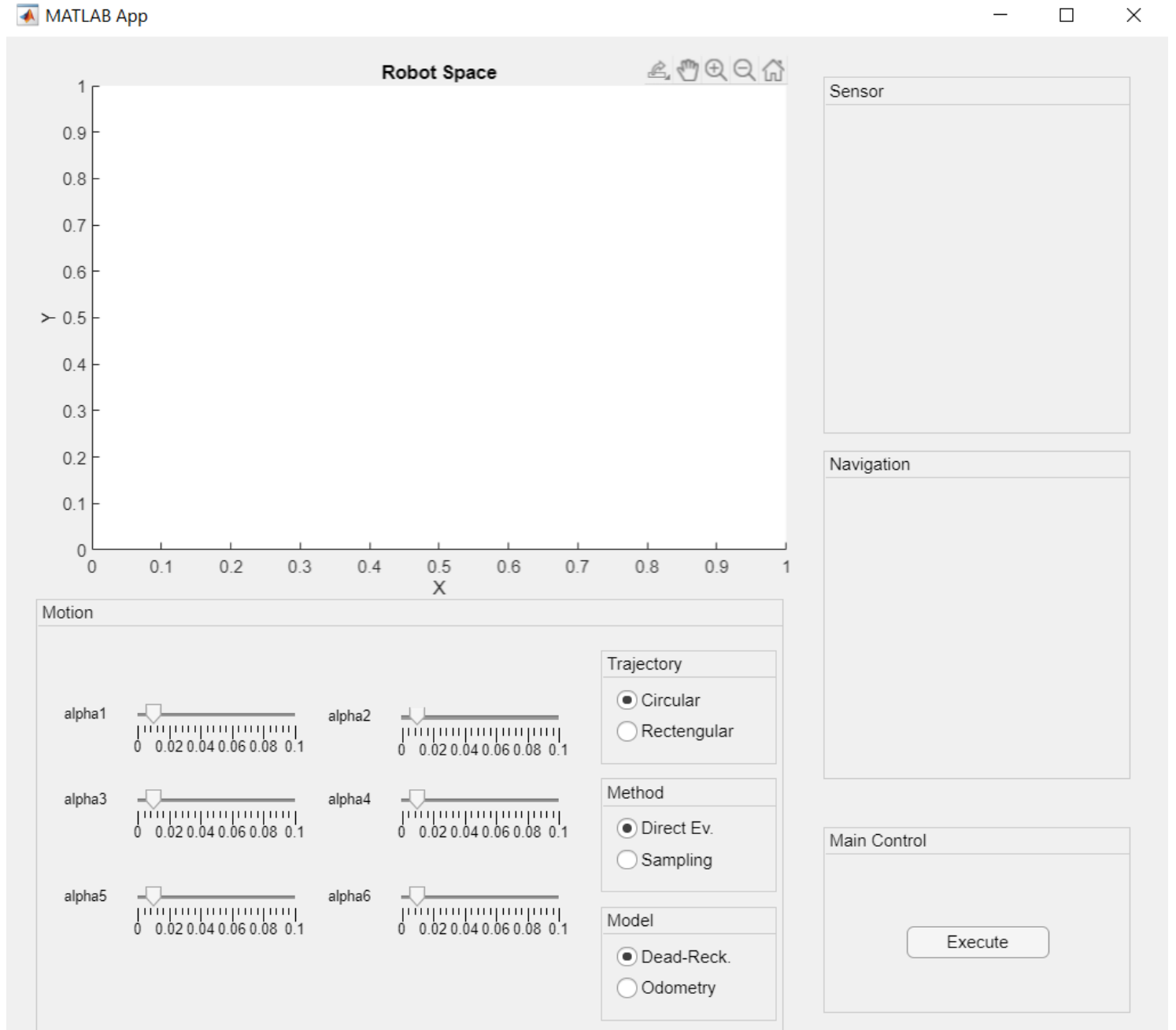


Figure 1 GUI

Common “sample” function for generating gaussian random variables are as below.

```
function sample = normal_dist(app,a1, a2, v,w)
    b = (a1 * abs(v) + a2 * abs(w));
    ran_num = -b + (b+b).*rand(12,1);
    sum = 0;
    for i=1:12
        sum = sum + ran_num(i);
    end
    sample = 0.5 * sum;
end
```

The MATLAB code of sampling algorithm for Deadreckoning Circular Trajectory model has been given below.

```

if(app.CircularButton.Value == 1)

    sample_n = 150;
    x=zeros(sample_n,8); y=zeros(sample_n,8); tet = zeros(sample_n,8);
    x(:,1) = 1; tet(:,1) = pi/2;
    r = 1;
    t = 1; %time of traviling for 45 degrees
    w = -pi/4; %Assumed 45 degrees travel in one sec.
    v = w * r;
    plot(app.UIAxes,x(1,1),y(1,1),'k.','LineWidth',2);
    hold(app.UIAxes,"on");
    for i=1:7
        for j=1:sample_n
            v_hat = v + normal_dist(app,a1, a2, v,w);
            w_hat = w + normal_dist(app,a3, a4, v,w);
            gama_hat = normal_dist(app,a5, a6, v,w);
            x(j,i+1) = x(j,i) - v_hat/w_hat * sin(tet(j,i)) + v_hat/w_hat * sin(tet(j,i)
+ w_hat*t);
            y(j,i+1) = y(j,i) + v_hat/w_hat * cos(tet(j,i)) - v_hat/w_hat * cos(tet(j,i)
+ w_hat*t);

            tet(j,i+1) = tet(j,i) + w_hat*t + gama_hat*t;
            plot(app.UIAxes,x(j,i+1),y(j,i+1),'k.','LineWidth',2);
            hold(app.UIAxes,"on");
        end
    end
    hold(app.UIAxes,'off');
end

if(app.RectengularButton.Value == 1)
sample_n = 100;
x=zeros(sample_n,13); y=zeros(sample_n,13); tet = zeros(sample_n,13);
x(:,1) = -4; y(:,1) = -4; tet(:,1) = 0; %Initial position.
t = 1; %time of traviling for 2 meters
v = 2; %Assumed traveled 2 meters in 1 sec.
w = 0.001; %Assumed no rotation.
plot(app.UIAxes,x(1,1),y(1,1),'k.','LineWidth',2);
hold(app.UIAxes,"on");
for i=1:12
    if(i == 4 || i == 8)
        w = pi/2;
    else
        w = 0.001;
    end
    for j=1:sample_n
        v_hat = v + normal_dist(app,a1, a2, v,w);
        w_hat = w + normal_dist(app,a3, a4, v,w);
        gama_hat = normal_dist(app,a5, a6, v,w);
        x(j,i+1) = x(j,i) - v_hat/w_hat * sin(tet(j,i)) + v_hat/w_hat * sin(tet(j,i) +
w_hat*t);
        y(j,i+1) = y(j,i) + v_hat/w_hat * cos(tet(j,i)) - v_hat/w_hat * cos(tet(j,i) +
w_hat*t);

        tet(j,i+1) = tet(j,i) + w_hat*t + gama_hat*t;
        plot(app.UIAxes,x(j,i+1),y(j,i+1),'k.','LineWidth',2);
        hold(app.UIAxes,'on');
    end
end
hold(app.UIAxes,'off');

```

```
end  
end
```

The results for this algorithm for value of 0.01 for alpha1 to alpha6 have been given in Figure 2.

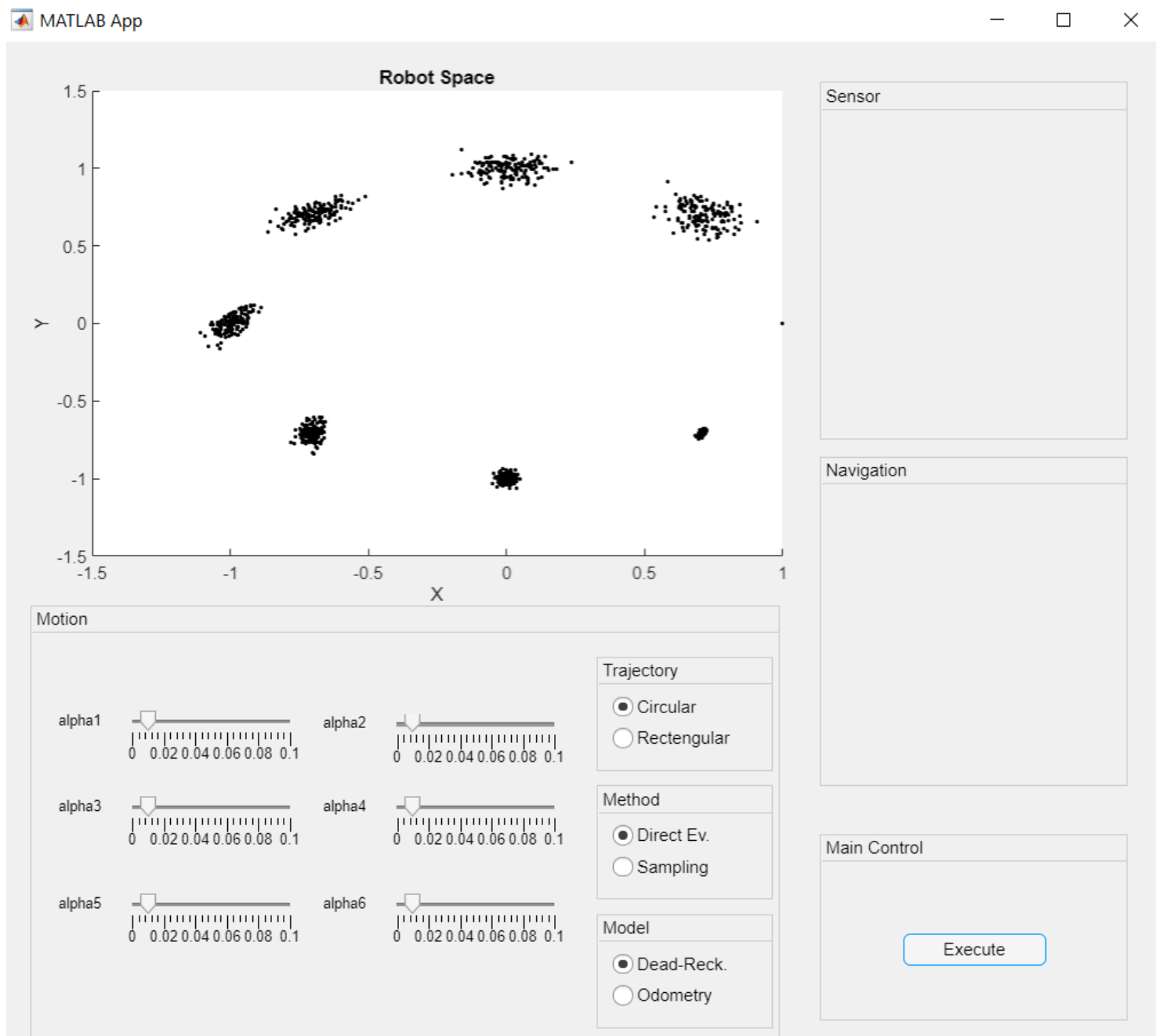


Figure 2 Deadreckoning model with sampling algorithm for $\alpha_N = 0.01$

Increasing alpha values causes more uncertainty in the model. This can be seen in Figure 3.

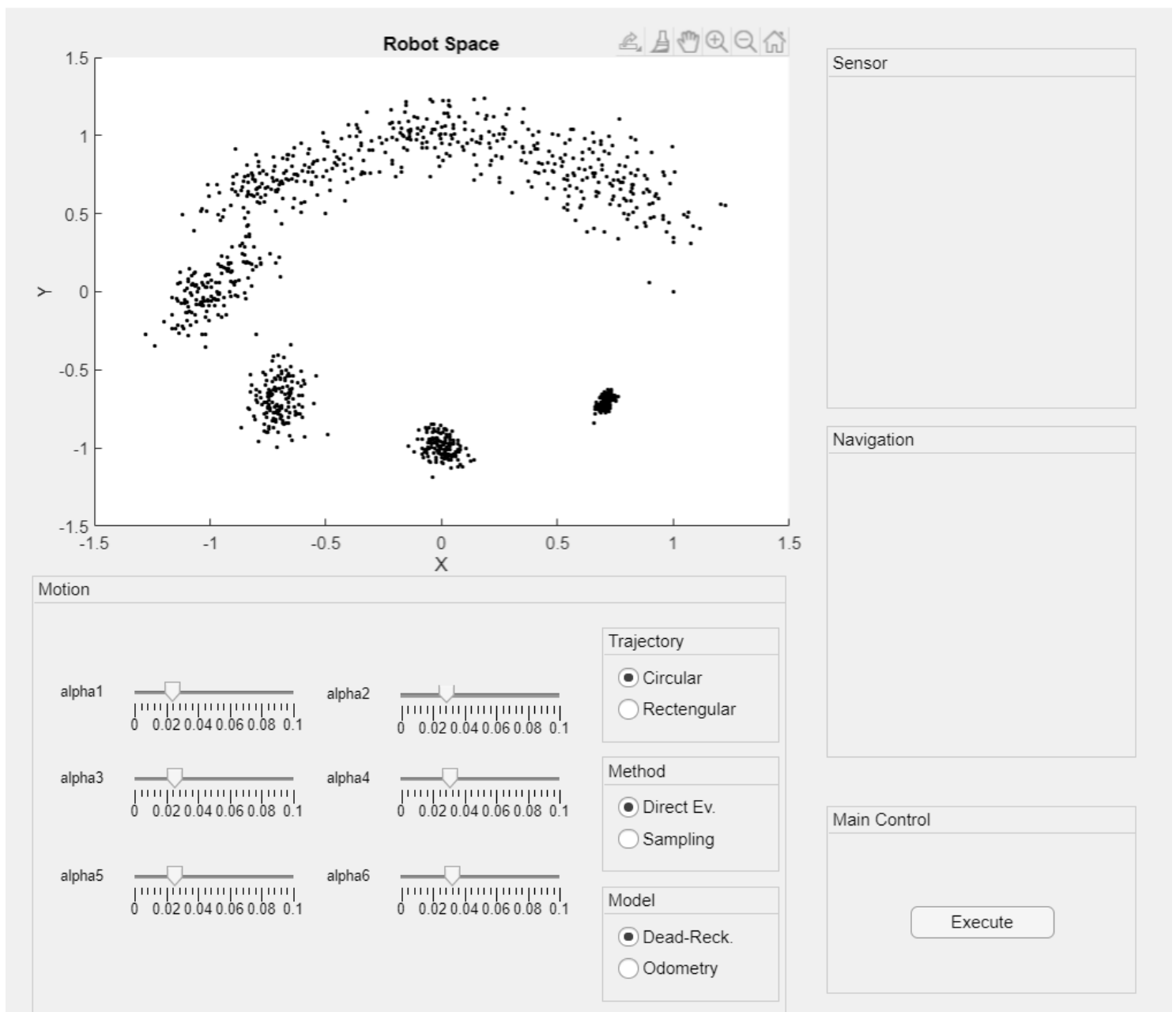


Figure 3 Deadreckoning model with sampling algorithm for $\alpha_N > 0.01$

c) Do the same for (a) & (b) using “odometry model”.

The MATLAB code of sampling algorithm for Odometry Circular Trajectory model has been given below.

```
function Odometry(app)
    a1 = app.alpha1Slider.Value; a2 = app.alpha2Slider.Value;
    a3 = app.alpha3Slider.Value; a4 = app.alpha4Slider.Value;
    a5 = app.alpha5Slider.Value; a6 = app.alpha6Slider.Value;
    if(app.CircularButton.Value == 1)
        sample_n = 150;
        xc = 0; yc = 0;
        x=zeros(sample_n,8); y=zeros(sample_n,8); tet = zeros(sample_n,8);
        xp=zeros(sample_n,8); yp=zeros(sample_n,8); tetp = zeros(sample_n,8);
        xp(:,1) = 1; yp(:,1) = 0; tetp(:,1) = -pi/2;
        tet(:,1) = -pi/2;
        r = 1;
        t = 1; %time of traviling for 45 degrees
```

```

w = -pi/4;
for i=1:8
    x(:,i) = xc - r*sin(tet(1,i));
    y(:,i) = yc + r*cos(tet(1,i));
    tet(1,i+1) = tet(1,i) + w*t;
end
for i=1:7
    for j=1:sample_n
        delta_rot1 = atan2((y(j,i+1)-y(j,i)),(x(j,i+1)-x(j,i))) - tet(j,i);
        delta_trans = sqrt((y(j,i+1)-y(j,i))^2 + (x(j,i+1)-x(j,i))^2);
        delta_rot2 = tet(j,i+1) - tet(j,i) - delta_rot1;

        delta_rot1_hat = delta_rot1 -
normal_dist(app,a1,a2,delta_rot1,delta_trans);
        delta_trans_hat = delta_trans -
normal_dist(app,a3,a4,delta_trans,delta_rot1+delta_rot2);
        delta_rot2_hat = delta_rot2 -
normal_dist(app,a1,a2,delta_rot2,delta_trans);

        xp(j,i+1) = xp(j,i) + delta_trans_hat * cos(tet(j,i)+delta_rot1_hat);
        yp(j,i+1) = yp(j,i) + delta_trans_hat * sin(tet(j,i)+delta_rot1_hat);
        tetp(j,i+1) = tetp(j,i) + delta_rot1_hat + delta_rot2_hat;
    end
end
plot(app.UIAxes,xp,yp, 'k. ');
end
end

```

The results for this algorithm for value of 0.01 for alpha1 to alpha6 and values greater then 0.001 have been given in Figure 4.

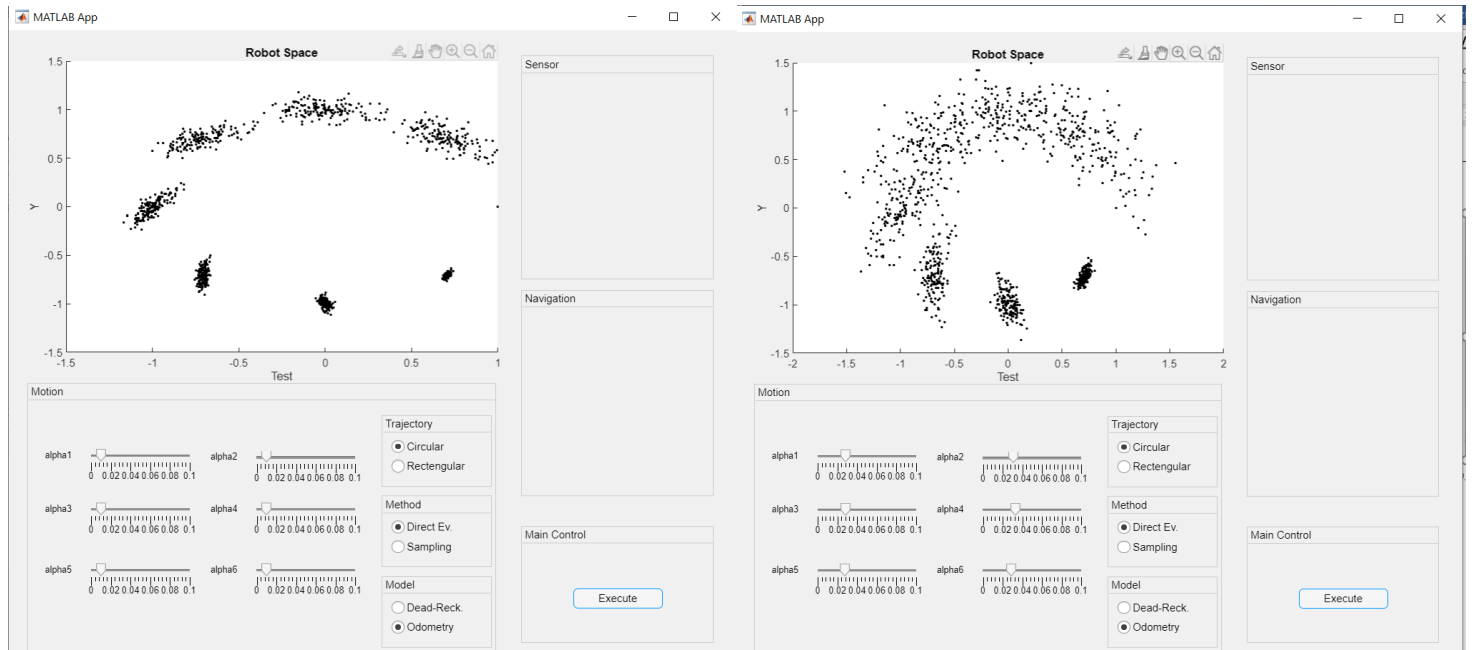


Figure 4 Odometry model with sampling algorithm for $\alpha_N = 0.01$ and $\alpha_N > 0.01$

II) Motion model for a Rectangular trajectory

a) Use odometry model and plot the uncertainty using sampling method.

MATLAB code for odometry model and rectangular trajectory is as below.

```
if(app.RectangularButton.Value == 1)
    sample_n = 150;
    xc = 0; yc = 0;
    x=zeros(sample_n,13); y=zeros(sample_n,13); tet = zeros(sample_n,13);
    xp=zeros(sample_n,13); yp=zeros(sample_n,13); tetp = zeros(sample_n,13);
    x(:,1) = -4; y(:,1) = -4; tet(:,1) = 0;
    xp(:,1) = -4; yp(:,1) = -4; tetp(:,1) = 0;
    tet(:,1) = 0;
    r = 1;
    t = 1; %time of traveling for 45 degrees
    d2r = pi/180;
    w = 0.001; %Assumed 45 degrees travel in one sec.
    v = 2;
    for i=2:5
        x(:,i) = x(:,i-1) + 2;
        y(:,i) = y(1,1);
        tet(:,i) = 0;
    end

    for i=6:9
        x(:,i) = x(1,5);
        y(:,i) = y(:,i-1) + 2;
        tet(:,i) = pi/2;
    end

    for i=10:13
        x(:,i) = x(:,i-1) - 2;
        y(:,i) = 4;
        tet(:,i) = pi;
    end

    for i=1:12
        for j=1:sample_n
            delta_rot1 = atan2((y(j,i+1)-y(j,i)),(x(j,i+1)-x(j,i))) - tet(j,i);
            delta_trans = sqrt((y(j,i+1)-y(j,i))^2 + (x(j,i+1)-x(j,i))^2);
            delta_rot2 = tet(j,i+1) - tet(j,i) - delta_rot1;

            delta_rot1_hat = delta_rot1 -
normal_dist(app,a1,a2,delta_rot1,delta_trans);
            delta_trans_hat = delta_trans -
normal_dist(app,a3,a4,delta_trans,delta_rot1+delta_rot2);
            delta_rot2_hat = delta_rot2 -
normal_dist(app,a1,a2,delta_rot2,delta_trans);

            xp(j,i+1) = xp(j,i) + delta_trans_hat * cos(tetp(j,i)+delta_rot1_hat);
            yp(j,i+1) = yp(j,i) + delta_trans_hat * sin(tetp(j,i)+delta_rot1_hat);
            tetp(j,i+1) = tetp(j,i) + delta_rot1_hat + delta_rot2_hat;
        end
    end
    plot(app.UIAxes,xp,yp, 'k. ');
end
end
```

Results can be seen in Figure 5 for both $\alpha_N = 0.01$ and $\alpha_N > 0.01$.

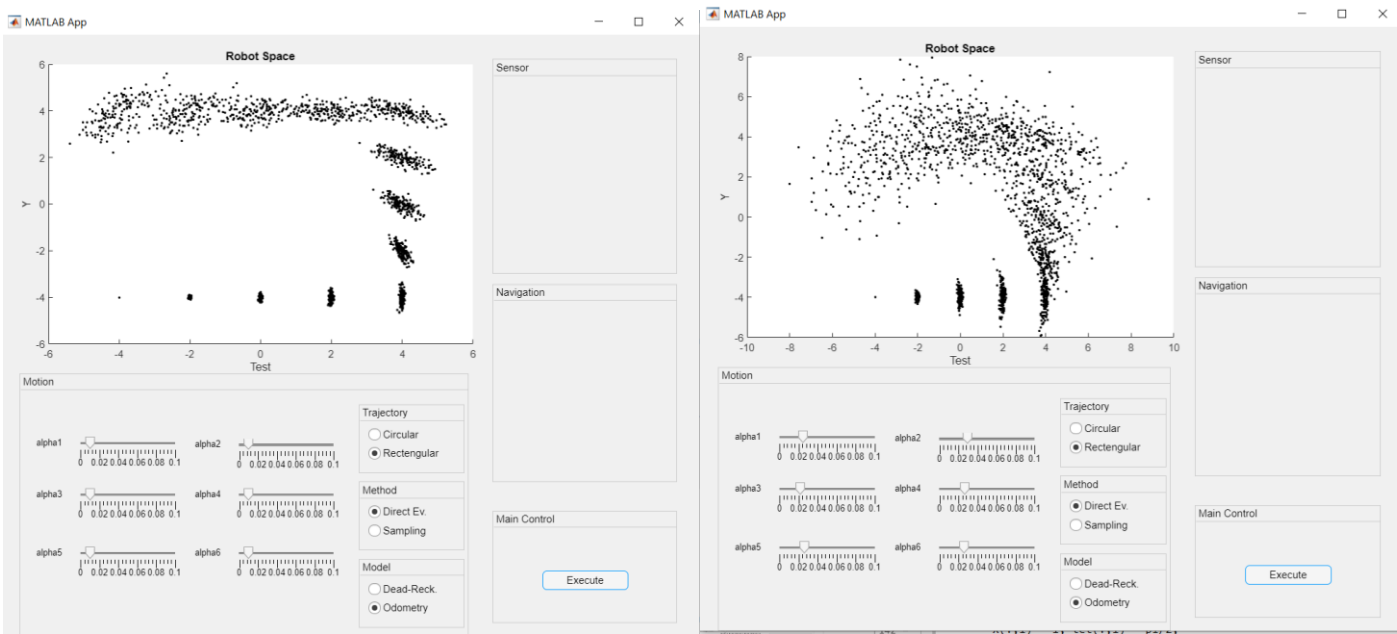


Figure 5 Odometry Rectengular Trajectory with Sampling Algorithm

The results for Deadreckoning algorithm is in Figure 6.

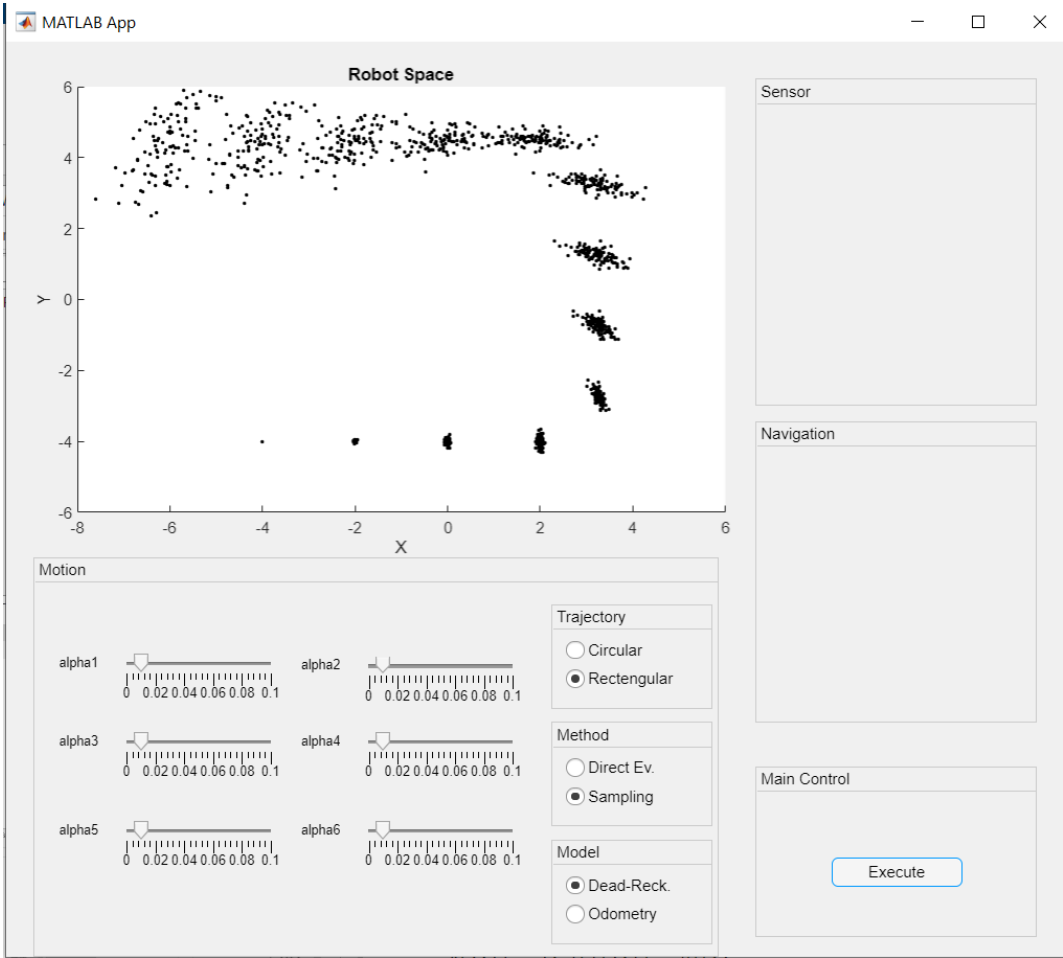


Figure 6 Deadreckoning Rectengular Trajectory with Sampling Algorithm