

Mobile robot short Project

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Shared Link with the teacher: <https://drive.matlab.com/sharing/bfe7e5c6-3b20-43d3-9f90-27cd7a658e06>

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Given the known workspace: Sensors_Data.mat, Laser_Data_CV__d_b.mat and Environment.png

Answer the following question:

Pose estimation (10%)

Before start:

- Open the Simulink model EKF_Pose_estimation.slx and get familiar with: 'Where2Find_Code.pdf'.
- Have a look to section See an animation of the file: 3_MR_SP_support.mlx

Every thing is done, the exercise consists in compile in a mlx file all the concepts.

1.- Implement in this mlx file:

```
clear
close all
load('Sensors_Data.mat')
load('Laser_Data_CV__d_b.mat')
sim("EKF_Pose_estimation_1.slx");
```

a) Pose_theoric, assume no noise. (review Mobile Robot Kinematics folder)

```
Tf = 60.08;
```

```

Ts = 0.02;
W = 0.52;
Gear_ratio = 100;
r = 0.1;

R_inc = right_angular_speed(:, 2) / Gear_ratio * r * Ts;
L_inc = left_angular_speed(:, 2) / Gear_ratio * r * Ts;
R_acu = [right_angular_speed(:, 1), cumsum(R_inc)];
L_acu = [left_angular_speed(:, 1), cumsum(L_inc)];

t=linspace(0,Tf,length(R_acu(:,2)))';
delta_d=(R_inc+L_inc)/2;
delta_t=(R_inc-L_inc)/W;

Initial_pose=transl(8.95,17.2,0)*trotz(-pi/2);
Initial_position=transl(Initial_pose);
Initial_orientation=-pi/2;

% Pose theoretic
x(1) = Initial_position(1);
y(1) = Initial_position(2);
o(1) = Initial_orientation;

```

b) Pose_estimation by adding noise in odometry. review Pose uncertainty folder)

```

% Pose estimation
V = [3.61e-4 0; 0 9e-6]; % Noise
noise_d = V(1, 1);
noise_t = V(2, 2);
x_est(1) = Initial_position(1);
y_est(1) = Initial_position(2);
o_est(1) = Initial_orientation;
for i=1:(length(t)-1)

    x(i+1) = x(i) + delta_d(i) * cos(o(i));
    y(i+1) = y(i) + delta_d(i) * sin(o(i));
    o(i+1) = o(i) + delta_t(i);

    x_est(i+1) = x_est(i) + (delta_d(i) + randn*sqrt(noise_d)) * cos(o(i) + randn*sqrt(noise_t));
    y_est(i+1) = y_est(i) + (delta_d(i) + randn*sqrt(noise_d)) * sin(o(i) + randn*sqrt(noise_t));
    o_est(i+1) = o(i) + delta_t(i) + randn*sqrt(noise_t);
end

Pose_theoric = [x; y; o];
% Here we use the data provided by the slx in order to have the same Pose_estimation
% than the teacher (the random numbers were generating a different path for
% us and it is preferable to have the same)
Pose_estimation = [Pose_est.Data(:,1)';Pose_est.Data(:,2)';Pose_est.Data(:,3)']

```

```

Pose_estimation = 3x3001
    8.9501    8.9500    8.9501    8.9500    8.9503    8.9504    8.9504    8.9504 ...

```

17.1652	17.1488	17.1736	17.1671	17.1145	17.0568	17.0580	17.0619
4.7140	4.7072	4.7082	4.7069	4.7176	4.7136	4.7157	4.7179

c) Ricatti equation for estimating the covariance matrix representing the uncertainty in the robot pose.

```
% Calculating the Jacobians

F_x = zeros([3,3,3004]);
F_x(1,1,:) = 1; F_x(2,2,:) = 1; F_x(3,3,:) = 1;
F_x(1,3,:) = -delta_d .* sin(o(:));
F_x(2,3,:) = delta_d .* cos(o(:));

F_v = zeros([3 2 3004]);
F_v(3,2,:) = 1;
F_v(1,1,:) = cos(o(:));
F_v(2,1,:) = sin(o(:));

Pk = zeros([3 3 3004]);
Pk(:, :, 1) = eye(3) * 0.0001;

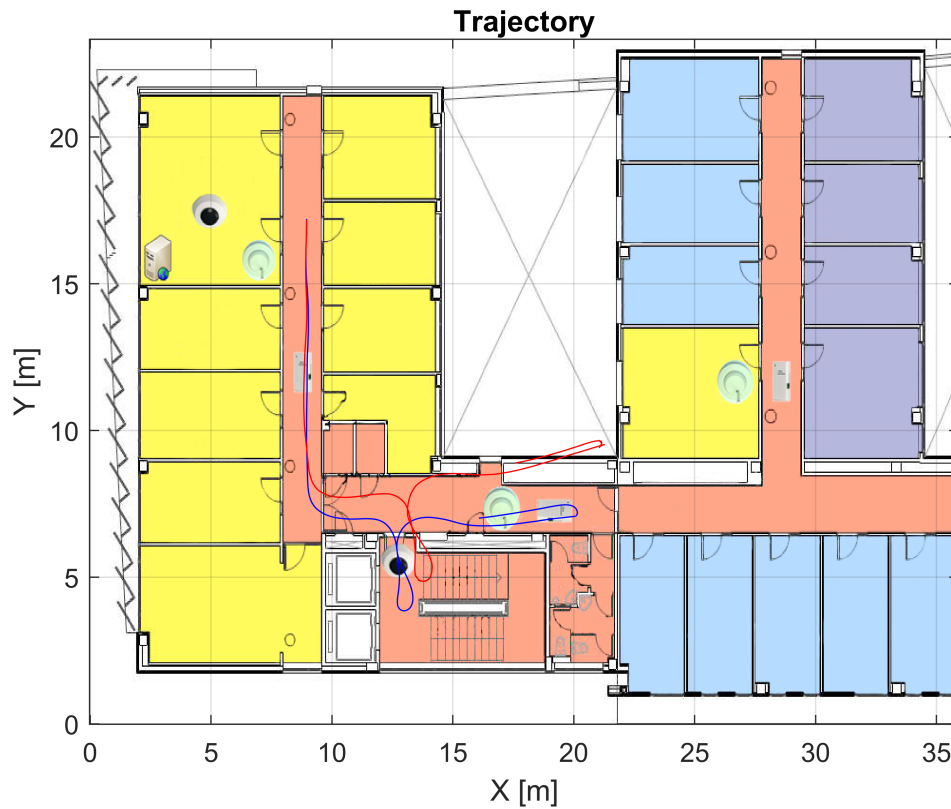
for i=2:3004
    Pk(:, :, i) = F_x(:, :, i)*Pk(:, :, i-1)*F_x(:, :, i)' + F_v(:, :, i)*V*F_v(:, :, i)';
end
```

2.- Display in a figure:

a) x,y poses for both trajectories: theoric and estimated

```
figure
I=imread('Enviroment.png'); % Read environment
x_ima=[0 35.9];
y_ima=[23.31 0];
image(I, 'XData', x_ima, 'YData', y_ima);
axis xy
grid on
title('Trajectory')
xlabel ('X [m]')
ylabel ('Y [m]')
hold on

plot(Pose_theoric(1,:), Pose_theoric(2,:), 'b') % Display a) Pose_theoric
plot(Pose_estimation(1,:), Pose_estimation(2,:), 'r') % Display b) Pose_estimation
hold off
```



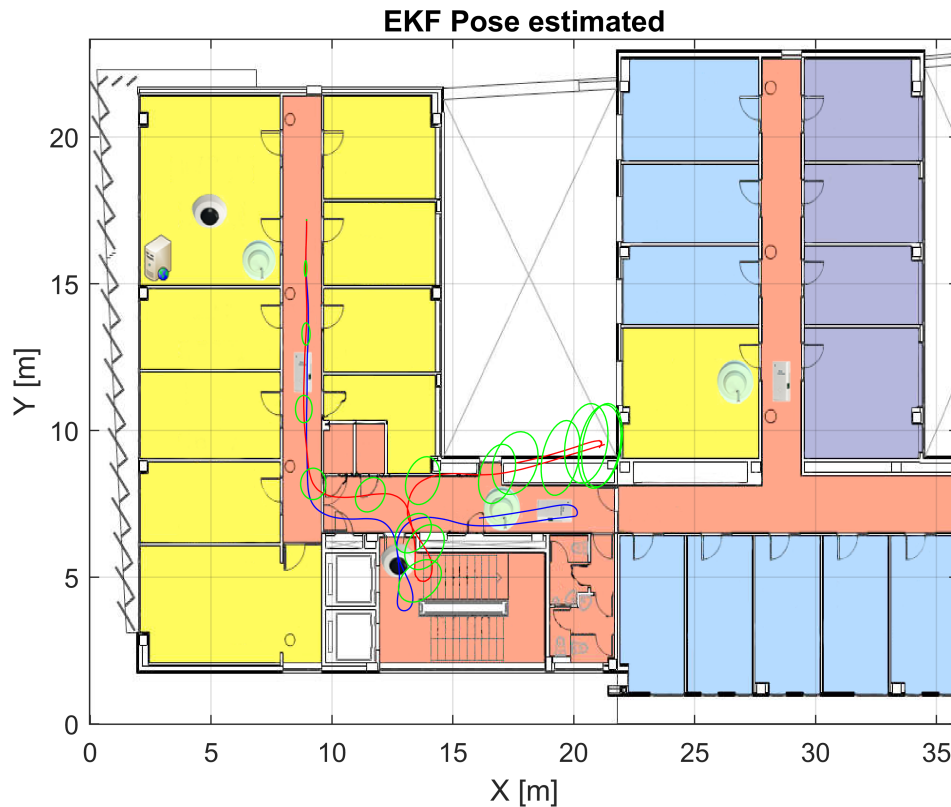
b) ellipses representing x,y uncertainty. Do it every 15-20 poses.

```
figure
image(I,'XData',x_ima,'YData',y_ima);
title('Trajectory')
axis xy
grid on
xlabel ('X [m]')
ylabel ('Y [m]')
hold on
plot(Pose_theoric(1,:), Pose_theoric(2,:), 'b')           % Display a) Pose_theoric
plot(Pose_estimation(1,:), Pose_estimation(2,:), 'r')      % Display b) Pose_estimation

for i=1:round(length(Pose_t.Data)/15):length(Pose_t.Data)

    plot_ellipse(Pk(1:2,1:2,i),[Pose_estimation(1,i), Pose_estimation(2,i)],'g');
    title('EKF Pose estimated');
end

hold off
```



c) represent uncertainty in orientation by adding a isosceles triangle in front of the robot the base

```
video = VideoWriter('uncertainty_theoric.avi');
video.FrameRate = 10;
video.Quality = 100;
open(video);
Robot = [0 -0.2 0 1;0.4 0 0 1;0 0.2 0 1];
Triangle = [0 -0.2 0 1;0.4 0 0 1;0 0.2 0 1];
for i=1:20:(length(t)-1)
    clf
    image(I,'XData',x_ima,'YData',y_ima);           % Display environment
    axis xy
    grid on
    title('Trajectory')
    xlabel ('X [m]')
    ylabel ('Y [m]')
    hold on
    plot(Pose_theoric(1,:), Pose_theoric(2,:), 'b')   % Display a) Pose_theoric
    plot(Pose_estimation(1,:), Pose_estimation(2,:), 'r') % Display b) Pose_estimation

    Triangle_uncertainty = Triangle;
    e = Pk(:, :, i);
    aux1 = Triangle_uncertainty(3, 2) * sqrt(det(e(1:2, 1:2)));
    aux2 = Triangle_uncertainty(1, 2) * sqrt(det(e(1:2, 1:2)));
    Triangle_uncertainty(3, 2) = aux1;
```

```

Triangle_uncertainty(1, 2) = aux2;

Robot_pose = transl(x(i), y(i), 0) * trotx(o(i));
Robot_final = Robot_pose * Robot';
Triangle_final = Robot_pose * transl(0.5, 0, 0) * trotx(pi) * Triangle_uncertainty';

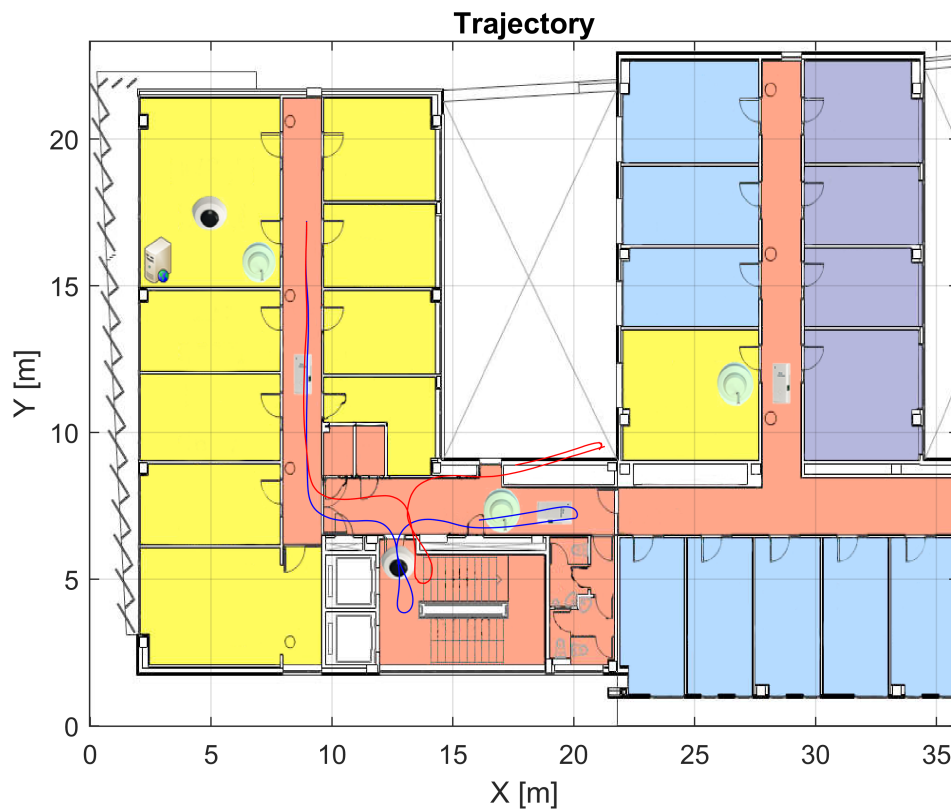
rob = patch(Robot_final(1, :), Robot_final(2, :), 'r');
tri = patch(Triangle_final(1, :), Triangle_final(2, :), 'yellow');

pause(0.2);

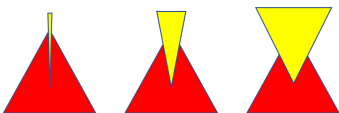
frame = getframe(gcf);
writeVideo(video, frame);
delete(rob);
delete(tri);

end

```



```
close(video)
```



The robot will be represented by a triangle: Robot= [0 -0.2 0 1;0.4 0 0 1;0 0.2 0 1]

Record a movie of the robot moving along the corridor displaying both trajectories.

See: help VideoWriter to get familiar with the matlab functions

Laser data (10%)

Make a movie of laser data in Robot reference frame.

copy here the code of the last TODO Lab about laser data. See the video of ATENEA: [Movie of the Laser Data seen in Robot Reference FrameURL](#)

```
video = VideoWriter('mapping.avi');
video.FrameRate = 20;
video.Quality = 100;
open(video);
Robot = [0 -0.2 0 1;0.4 0 0 1;0 0.2 0 1];

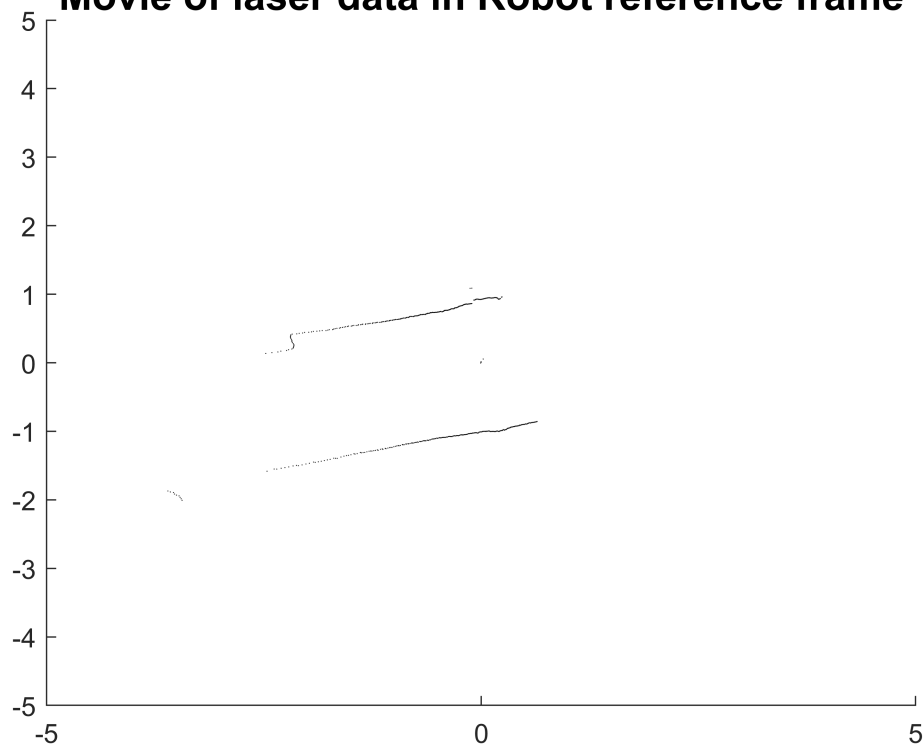
for k=1:6:(length(t)-1)
    clf
    Robot_pose = transl(x(k), y(k), 0) * trozt(o(k));
    Robot_final = Robot_pose * Robot';

    if (mod(k, 6) == 1)
        clear x2 y2
        count = 1;
        m = size(polar_laser_data);
        i = ceil(k/20.1611);
        for j=1:(m(2)-1)
            if polar_laser_data(i, j) / 1000 > 0
                alpha = (j-1) * 0.3515;
                alpha = o(20 * (i-1) + 1) - 120 * pi/180 + alpha * pi/180;
                x2(count) = (polar_laser_data(i, j) / 1000) * cos(alpha);
                y2(count) = (polar_laser_data(i, j) / 1000) * sin(alpha);
                count = count + 1;
            end
        end
    end

    scatter(x2, y2, 1, '.', 'black');
    axis([-5 5 -5 5]);
    title("Movie of laser data in Robot reference frame", 'FontSize',15);
    pause(0.2)

    frame = getframe(gcf);
    writeVideo(video, frame);
end
```

Movie of laser data in Robot reference frame



```
close(video)
```

Localization (40%)

Localize the Robot by using the Similarity Transform.

There are two assumptions :

No correction of the trajectory is done.

1.- Make a 'log' with the following columns: time stamp, estimated noisy pose, number of Land-marks seen and the errors. **The log that we have made is named Log_without_correction**

```
landmarks = [...  
    7.934 16.431 0 1;...  
    9.583 16.431 0 1;...  
    9.584 13.444 0 1;...  
    9.584 10.461 0 1;...  
    7.973 10.534 0 1;...  
    7.934 7.547 0 1;...  
    9.584 6.654 0 1;...  
    13.001 6.525 0 1;...  
    17.007 8.136 0 1 ...  
]';
```

```
Detected_landmarks = [];  
Reference_landmarks = [];
```



```

Log_without_correction = [];

image(I, 'XData', x_ima, 'YData', y_ima);
axis xy
hold on

plot(Pose_theoric(1,:), Pose_theoric(2,:), 'b')           % Display Pose_theoric
plot(Pose_estimation(1,:), Pose_estimation(2,:), 'r')     % Display Pose_estimation
title('trajectories');
xlabel ('x [meters]');
ylabel ('y [meters]');

for time=3:length(l_s_b(:,1))

    Robot_pose = transl(Pose_estimation(1,time),Pose_estimation(2,time),0)*trotz(Pose_estim

for landmark_index=1:length(landmarks(1,:))
    if l_s_b(time,landmark_index) ~= 0

        d=l_s_d(time,landmark_index);
        b=l_s_b(time,landmark_index);

        [X,Y]=pol2cart(b(1,1),d(1,1));

        L_w=Robot_pose*[X Y 0 1]';

        Detected_landmarks = [Detected_landmarks L_w];
        Reference_landmarks = [Reference_landmarks landmarks(1:2,landmark_index)];
    end
end

if ~isempty(Detected_landmarks)

    scatter(Reference_landmarks(1,:),Reference_landmarks(2,:), 50, 'r','filled');
    scatter(Detected_landmarks(1,:),Detected_landmarks(2,:), 50, 'g','filled');

    A = [];
    for i=1:size(Reference_landmarks, 2)
        A = [A;[Reference_landmarks(1,i),Reference_landmarks(2,i),1,0]];
        A = [A;[Reference_landmarks(2,i),-Reference_landmarks(1,i),0,1]];
    end

    B = [];
    for i=1:size(Detected_landmarks, 2)
        B = [B; Detected_landmarks(1,i); Detected_landmarks(2,i)];
    end

    Similarity_transform = linsolve(A,B);
    tx_ST = Similarity_transform(3);
    ty_ST = Similarity_transform(4);

```

```

alpha_ST = atan2(Similarity_transform(2),Similarity_transform(1));

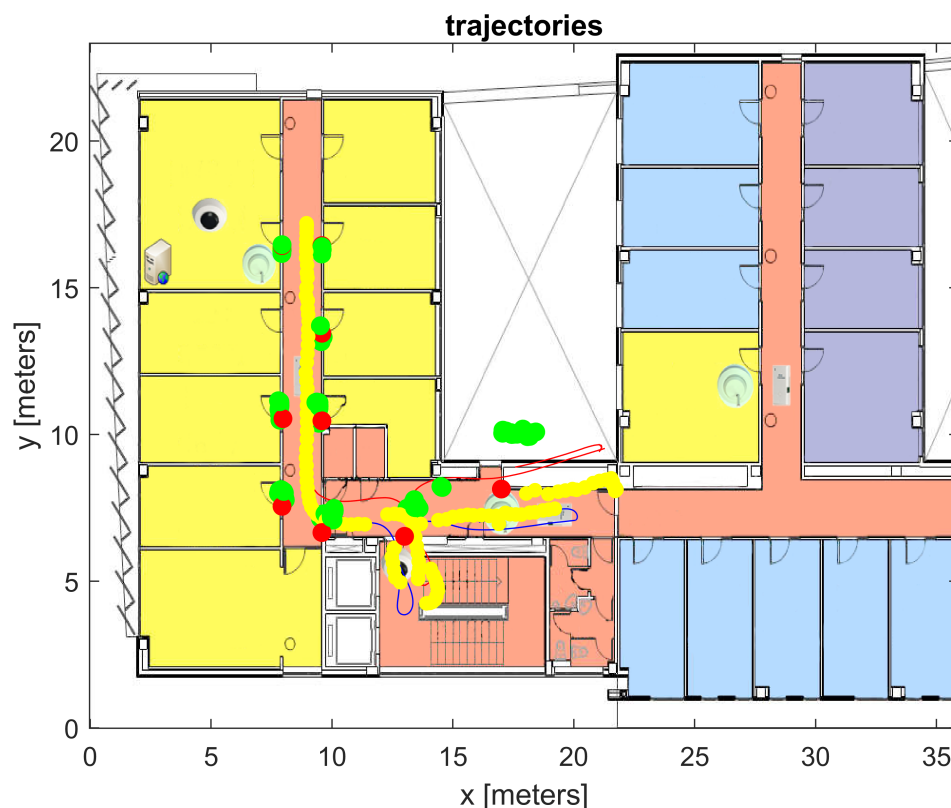
Error = transl(-tx_ST,-ty_ST,0)*trotz(alpha_ST);
Corrected_robot = Error * [transl(Robot_pose); 1];
scatter(Corrected_robot(1),Corrected_robot(2), 30, 'yellow','filled');

% time stamp, estimated noisy pose, corrected position, number of Land-marks, and t
Single_log.Time_stamp = time; % timestamp is
Single_log.Robot_pose = Robot_pose; % estimated no
Single_log.Corrected_robot = Corrected_robot; % corrected pos
Single_log.Number_of_landmarks = length(Detected_landmarks(1,:)); % number of lan
Single_log.Error = Error; % errors

Log_without_correction = [Log_without_correction Single_log];
end

Detected_landmarks = [];
Reference_landmarks = [];
end
hold off

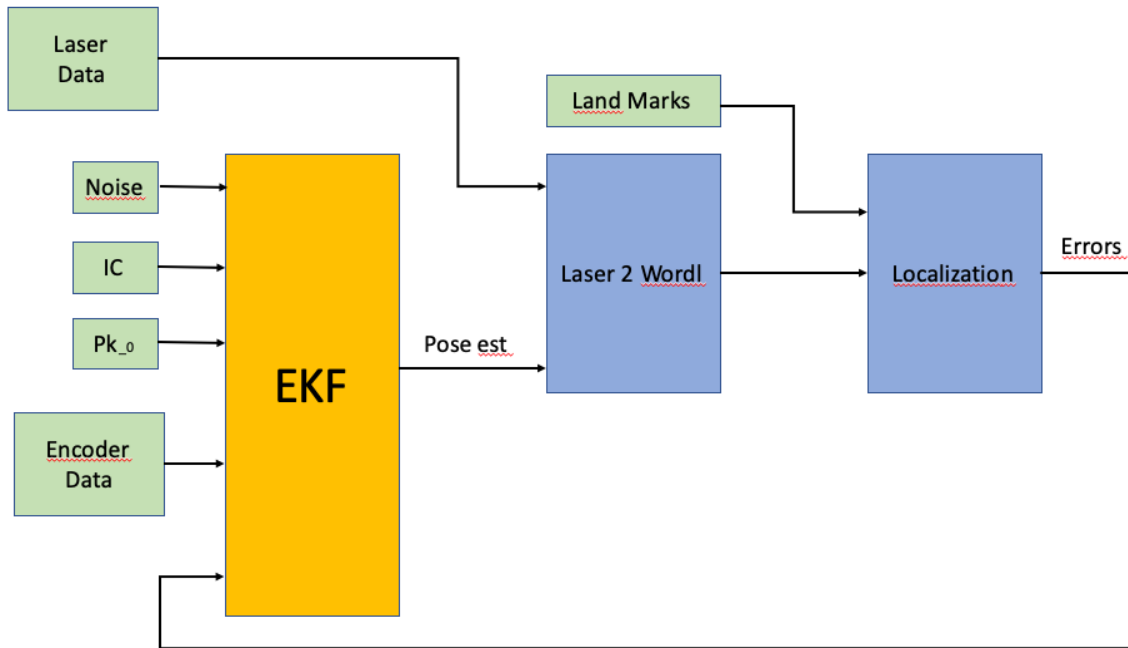
```



Correct the noisy trajectory.

To get an inspiration get familiar MR_SP_support.mlx

Pay attention to the relationship of the variables



Update the estimated covariance matrix by the sensor, knowing that the Laser scanner has an accuracy of 4 mm with a standard deviation of 0.2 mm.

- 1.- Display in a figure: the map, theoric trajectory (no noise) the noisy trajectory and the corrected trajectory.
- 2.- Make a 'log' with the following columns: time stamp, estimated noisy pose, corrected position, number of Land-marks, and the errors. **The log that we have done is named Log_with_correction**

```

Detected_landmarks = [];
Reference_landmarks = [];
Log_with_correction = [];

Position_with_correction(:,1) = zeros(size(Pose_theoric(:,1)));
Position_with_correction(:,2) = Pose_theoric(:,2);

Orientation_with_correction = zeros(3004,1);
Orientation_with_correction(2) = Pose_theoric(3,2);
mu_noise = 0;
mu_laser = 0;
mu_fused = 0;

image(I, 'XData', x_ima, 'YData', y_ima);
axis xy
hold on

plot(Pose_theoric(1,:), Pose_theoric(2,:), 'b')           % Display Pose_theoric
plot(Pose_estimation(1,:), Pose_estimation(2,:), 'r')     % Display Pose_estimation
title('trajectories'); xlabel ('x [meters]'); ylabel ('y [meters]')

for time=3:length(l_s_b(:,1))

```

```

Position_with_correction(1,time) = Position_with_correction(1,time-1) + (delta_d(time)
Position_with_correction(2,time) = Position_with_correction(2,time-1) + (delta_d(time)
Orientation_with_correction(time) = mod(Orientation_with_correction(time-1) + delta_t(t

Robot_pose = transl(Position_with_correction(1,time),Position_with_correction(2,time),0

for landmark_index=1:length(landmarks(1,:))
    if l_s_b(time,landmark_index) > 0

        d=l_s_d(time,landmark_index);
        b=l_s_b(time,landmark_index);

        [X,Y]=pol2cart(b(1,1),d(1,1));

        L_w=Robot_pose*[X Y 0 1]';

        Detected_landmarks = [Detected_landmarks L_w];
        Reference_landmarks = [Reference_landmarks landmarks(1:2,landmark_index)];
    end
end

if ~isempty(Detected_landmarks)

    scatter(Reference_landmarks(1,:),Reference_landmarks(2,:), 50, 'r','filled');
    scatter(Detected_landmarks(1,:),Detected_landmarks(2,:), 50, 'g','filled');

    A = [];
    for i=1:size(Reference_landmarks, 2)
        A = [A;[Reference_landmarks(1,i),Reference_landmarks(2,i),1,0]];
        A = [A;[Reference_landmarks(2,i),-Reference_landmarks(1,i),0,1]];
    end

    B = [];
    for i=1:size(Detected_landmarks, 2)
        B = [B; Detected_landmarks(1,i); Detected_landmarks(2,i)];
    end

    Similarity_transform = linsolve(A,B);
    tx_ST = Similarity_transform(3);
    ty_ST = Similarity_transform(4);
    alpha_ST = atan2(Similarity_transform(2),Similarity_transform(1));

    Error = transl(-tx_ST,-ty_ST,0)*trotz(alpha_ST);
    Corrected_robot = Error * [transl(Robot_pose); 1];
    scatter(Corrected_robot(1),Corrected_robot(2), 30, 'yellow','filled');
    Position_with_correction(1,time) = Corrected_robot(1);
    Position_with_correction(2,time) = Corrected_robot(2);

    %time stamp, estimated noisy pose, corrected position, number of Land-marks, and th

```

```

Single_log.Time_stamp = time; % timestamp is
Single_log.Robot_pose = Robot_pose; % estimated no
Single_log.Corrected_robot = Corrected_robot; % corrected pos
Single_log.Number_of_landmarks = length(Detected_landmarks(1,:)); % number of lan
Single_log.Error = Error; % errors

Log_with_correction = [Log_with_correction Single_log];

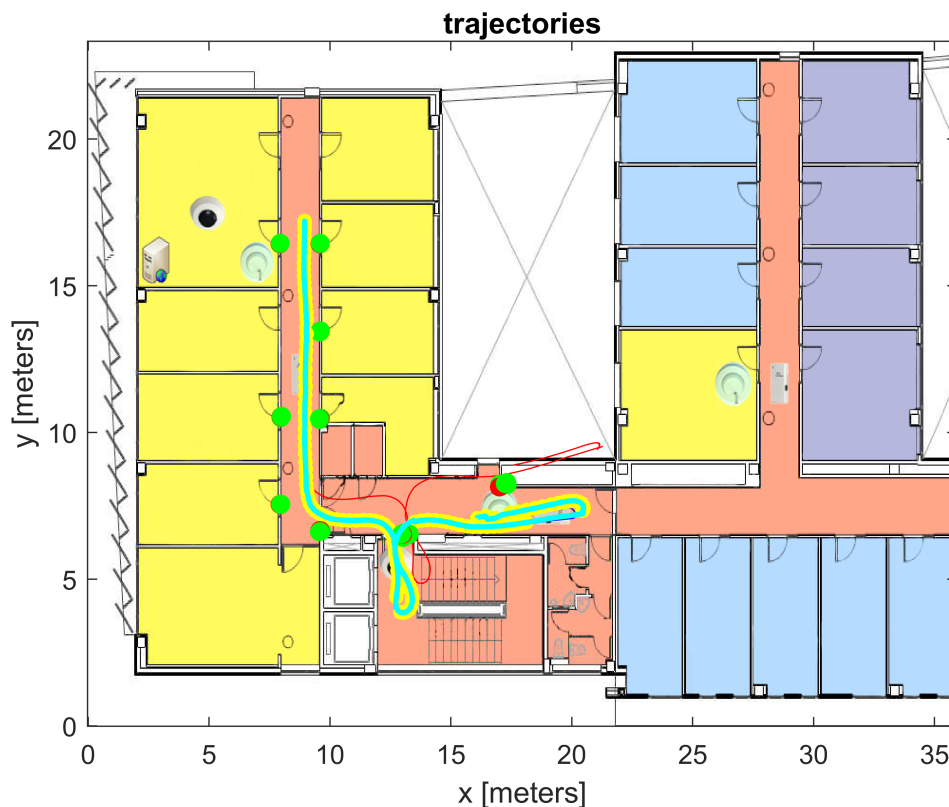
end

Detected_landmarks = [];
Reference_landmarks = [];

end

plot(Position_with_correction(1,2:end), Position_with_correction(2,2:end),'k','Color','c',
hold off

```



Mapping (40%)

This is part of the last TODO.

Walls

Draw the 'walls/obstacles' *seeing by* the laser data. Do it only when Land Marks are available and the trajectory have been corrected.

Make a movie erasing the previous walls after 200ms for better understanding.

See the video: '3_mapping.mp4' for inspiration

```
close all

video = VideoWriter('mapping2.avi');
video.FrameRate = 20;
video.Quality = 100;
open(video);
Robot = [0 -0.2 0 1; 0.4 0 0 1; 0 0.2 0 1];
visible = 0;
for k=1:3:(length(t)-1)
    clf
    plot(Pose_theoric(1,:), Pose_theoric(2,:), 'b--');
    xlim([5.6 24.0])
    ylim([0.8 18.0])
    hold on

    Robot_pose = transl(x(k), y(k), 0) * trotx(o(k));
    Robot_final = Robot_pose * Robot';

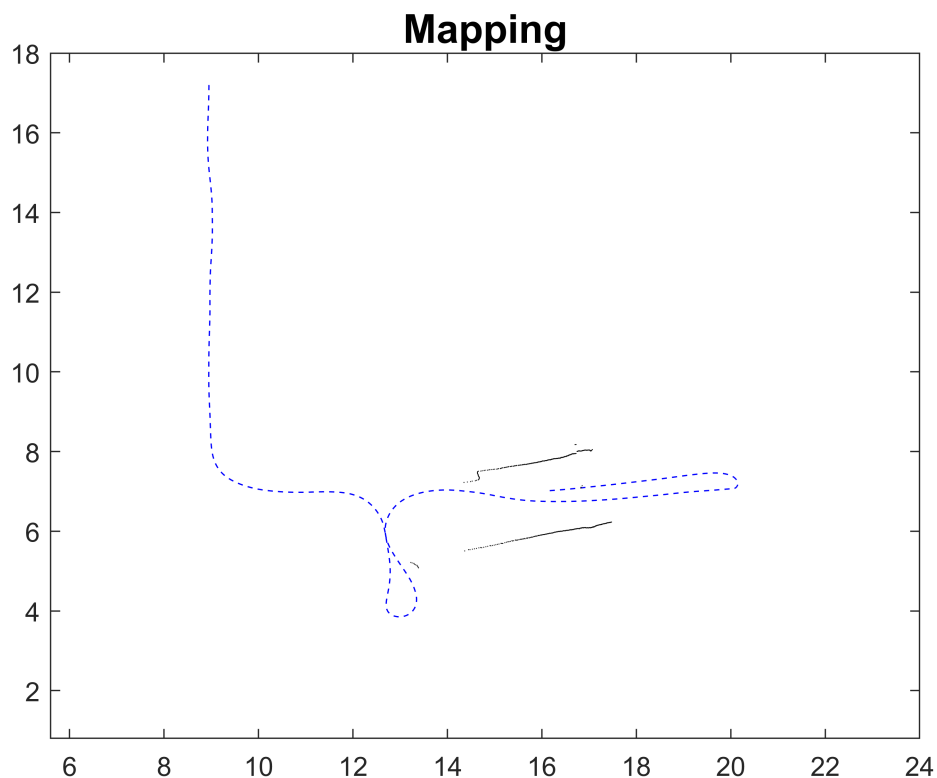
    rob = patch(Robot_final(1, :), Robot_final(2, :), 'r');

    if (mod(k, 6) == 1)
        clear x2 y2
        count = 1;
        m = size(polar_laser_data);
        i = ceil(k/20.1611);
        for j=1:(m(2)-1)
            if polar_laser_data(i, j) / 1000 > 0
                alpha = (j-1) * 0.3515;
                alpha = o(20 * (i-1) + 1) - 120 * pi/180 + alpha * pi/180;
                x2(count) = x(20 * (i-1) + 1) + (polar_laser_data(i, j) / 1000) * cos(alpha);
                y2(count) = y(20 * (i-1) + 1) + (polar_laser_data(i, j) / 1000) * sin(alpha);
                count = count + 1;
            end
        end
    end

    if mod(k, 20) == 0
        visible = 0;
        for i = 1:width(l_s_b)
            if visible == 0 && l_s_b(k, i) ~= 0
                P = landmarks(:, i);
                dist = sqrt( (x(k) - P(1))^2 + (y(k) - P(2))^2 );
                if dist <= 4
                    visible = 1;
                end
            end
        end
    end
end
```

```
end
end

if visible
    scatter(x2, y2, 1, '.', 'black');
end
title("Mapping", 'FontSize', 15);
pause(0.2)
frame = getframe(gcf);
writeVideo(video, frame);
delete(rob);
end
```



```
close(video)
```

Occupancy grid

Use Breshehan algorithm to build the map. Do it only when Land Marks are available and the trajectory have been corrected.

Use the idea behind the line tracing: Visit: https://es.wikipedia.org/wiki/Algoritmo_de_Bresenham

See: Mapping.mlx and '4_occupancy.mp4' for inspiration

```
video = VideoWriter('Occupancy.avi');
video.FrameRate = 5;
video.Quality = 100;
```

```

open(video);

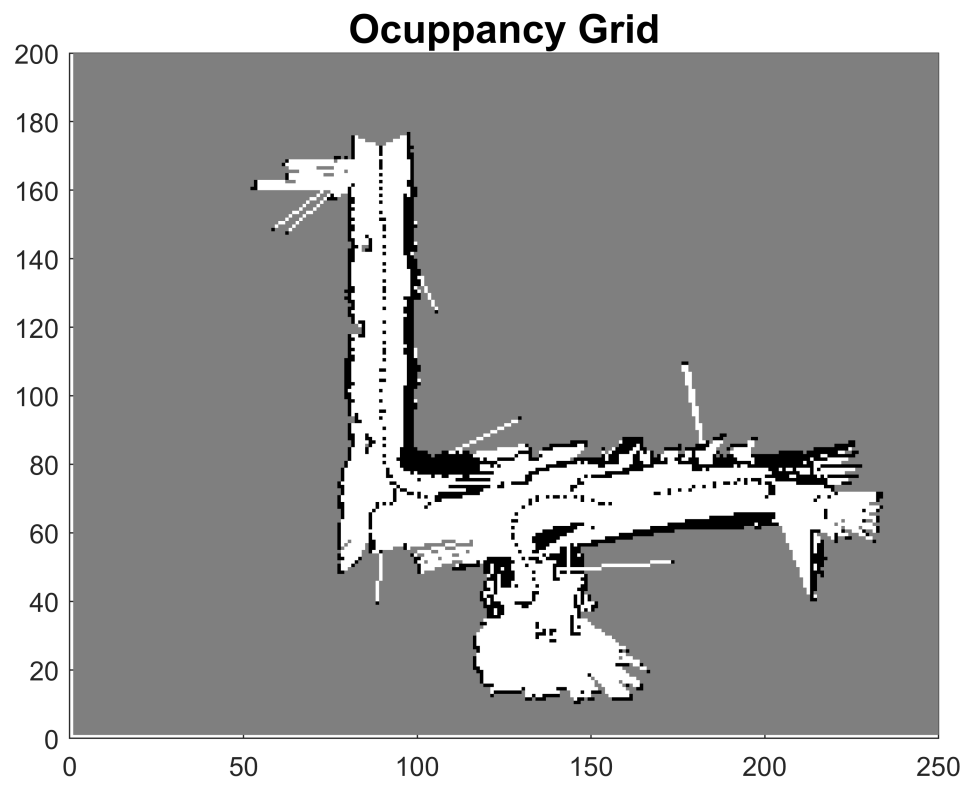
resolution = 10;
A = ones(20*resolution, 25*resolution);
for p=20:20:(length(t)-1)
    clear x2 y2 o2
    i = ceil(p/20.1611);
    count = 1;
    m = size(polar_laser_data);
    for j=1:(m(2)-1)
        if polar_laser_data(i, j) / 1000 > 0
            alpha = (j-1) * 0.3515;
            alpha = o(20 * (i-1) + 1) - 120 * pi/180 + alpha * pi/180;
            x2(count) = x(20 * (i-1) + 1) + (polar_laser_data(i, j) / 1000) * cos(alpha);
            y2(count) = y(20 * (i-1) + 1) + (polar_laser_data(i, j) / 1000) * sin(alpha);
            o2(count) = alpha;
            count = count + 1;
        end
    end

    for j = 1:(count-1)
        P_w = transl(0, 0, 0) * trotz(0) * [x2(j) y2(j) 0 1]';
        [X Y] = bresenham(x(p)*resolution, y(p)*resolution, P_w(1)*resolution, P_w(2)*resolution);
        l_xy = [X Y];
        L=height(l_xy);
        for k=1:L
            if l_xy(k, 1) > 0 && l_xy(k, 2) > 0
                if k == L
                    A(l_xy(k, 2),l_xy(k, 1)) = 0;
                else
                    A(l_xy(k, 2),l_xy(k, 1)) = 2;
                end
            end
        end
    end

    h = pcolor(A);
    h.EdgeColor = 'none';
    axis xy
    ylim([0 20*resolution]);
    xlim([0 25*resolution]);
    title("Occupancy Grid", 'FontSize',15);
    colormap(gray(3))
    pause(0.2)

    frame = getframe(gcf);
    writeVideo(video, frame);
end

```

```
close(video);
```