Mobile robot short Project

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Link: https://drive.matlab.com/sharing/49eb5fa6-0641-425c-a130-9cf41b2e1058

Link Videos: https://drive.google.com/drive/folders/12hRJl6bwnyQormSw4NueEOiBilSblFel?usp=sharing

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clear
close all
clf
```

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

Answer the following question:

Pose estimation (10%)

Base on the Simulink model EKF Pose estimation.slx

(1) Implement in this mlx file:

```
load('Sensors_Data.mat')
load('Encoder_Data.mat')
load('Laser_Data_CV__d_b.mat')
```

a) Pose theoric, asumme no noise.

```
L_speed = left_angular_speed; L_speed(:,2) = L_speed(:,2) .* r_w ./ 100;
R_speed = right_angular_speed; R_speed(:,2) = R_speed(:,2) .* r_w ./ 100;
```

```
L_displ = L_speed; L_displ(:,2) = L_displ(:,2) .* Ts;
R_displ = R_speed; R_displ(:,2) = R_displ(:,2) .* Ts;

S = 0.26;
delta_d = (L_displ(:,2) + R_displ(:,2)) / 2;
delta_th = (R_displ(:,2) - L_displ(:,2)) / (2*S);

tx = 8.65; ty = 17.2; alpha = -pi/2;
Pose_theoric(:,:,1) = transl(tx,ty,0) * trotz(alpha); % initial pose
for i = 2:length(delta_th)
    Pose_theoric(:,:,i) = Pose_theoric(:,:,i-1) * transl(delta_d(i-1),0,0) * trotz(delta_ton_theoric(:,i)) = transl(Pose_theoric(:,:,i-1));
    Orientation_theoric(:,i) = tr2rpy(Pose_theoric(:,:,i-1));
end
```

b) Pose estimation by adding noise in odometry.

```
= 0.025^2; % testing with some values to have more difference between theoric a
ProcNoiseD
ProcNoiseTheta = 0.009^2; % testing with some values to have more difference between theoric a
V = ...
    [
       ProcNoiseD 0;
                    ProcNoiseThetal;
noiseD = rand*V(1,1);
noiseTheta = rand*V(2,2);
Position_estimation(:,1) = zeros(size(Position_theoric(:,1)));
Position_estimation(:,2) = Position_theoric(:,2);
Orientation_estimation = zeros(3004,1);
Orientation_estimation(2) = Orientation_theoric(3,2);
for i=3:length(delta_th)
    Position_estimation(1,i) = Position_estimation(1,i-1) + (delta_d(i)+noiseD)*cos(Orientation)
    Position estimation(2,i) = Position estimation(2,i-1) + (delta d(i)+noiseD)*sin(Orientation
    Orientation_estimation(i) = mod(Orientation_estimation(i-1) + delta_th(i) + noiseTheta , 2°
end
```

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

```
% Calcultaing the Jacobians

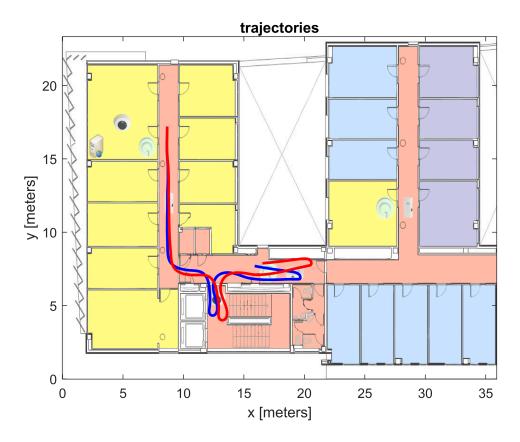
F_x = zeros([3,3,3004]);
F_x(1,1,:) = 1; F_x(2,2,:) = 1; F_x(3,3,:) = 1;
F_x(1,3,:) = -delta_d .* sin(Orientation_theoric(3,:))';
F_x(2,3,:) = delta_d .* cos(Orientation_theoric(3,:))';

F_v = zeros([3 2 3004]);
F_v(3,2,:) = 1;
F_v(1,1,:) = cos(Orientation_theoric(3,:))';
F_v(2,1,:) = sin(Orientation_theoric(3,:))';
Pk1 = zeros([3 3 3004]);
```

(2) Display in a figure

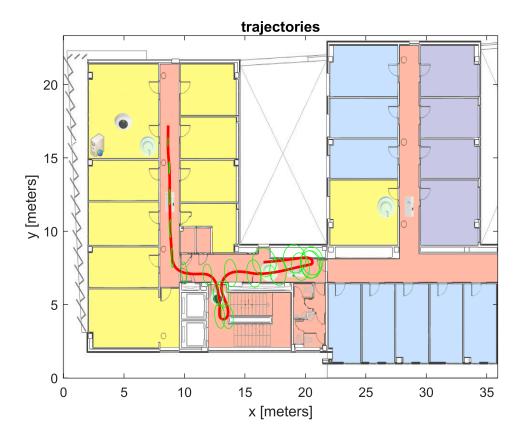
a) x,y poses for both trajectories: theoric and estimated

```
I=imread('Environment.png');
x_{ima=[0 35.9]};
y_ima=[23.31 0];
image(I, 'XData', x_ima, 'YData', y_ima, "AlphaData", 0.75);
axis xy
hold on
    plot( ...
        Position_theoric(1,2:end), ...
        Position_theoric(2,2:end), ...
        'k', ...
        'Color','b', ...
        "LineWidth",2 ...
    title('trajectories'); xlabel ('x [meters]'); ylabel ('y [meters]')
    plot( ...
        Position_estimation(1,2:end), ...
        Position_estimation(2,2:end), ...
        'k', ...
        'Color','r', ...
        "LineWidth",2 ...
hold off
```



b) ellipses representing x,y uncertanty. don't do it for all poses (every 15-20 poses could be nice).

```
image(I,'XData',x_ima,'YData',y_ima,"AlphaData",0.75);
axis xy
hold on
    plot( ...
        Position_estimation(1,2:end), ...
        Position_estimation(2,2:end), ...
        'k', ...
        'Color','r', ...
        "LineWidth",2 ...
   title('trajectories'); xlabel ('x [meters]'); ylabel ('y [meters]')
   for i=1:length(delta_th)
        if (mod(i,150)==0)
            plot_ellipse( ...
                Pk1(1:2,1:2,i), ...
                [Position_estimation(1,i), Position_estimation(2,i)], ...
                 'g'...
            );
        end
    end
hold off
```



c) represent uncertanty in orientation by adding a isosceles triange in front of the robot the base







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

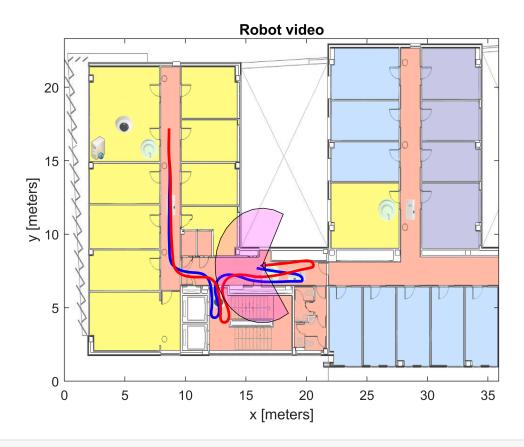
Record a movie of the robot moving along the corridor displaying both trajectories.

See: help VideoWriter to get familiar with the matlab functions

```
Robot= [ ...
     0 -0.2 0 1; ...
     0.4     0 0 1; ...
     0 0.2 0 1 ...
]';
Laser_Range=3.9; % meters
w=linspace(-30*pi/180,210*pi/180,20); %Laser foot print
s_x=Laser_Range*sin(w); s_y=Laser_Range*cos(w);
Lfp.v=[[s_x 0];[s_y 0]]';
Lfp.f= [1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 1];
r_patch=[];s_patch=[];
speed = 5;

Robot_video = VideoWriter('Robot_Video.avi');
Robot_video.FrameRate = 45;
Robot_video.Quality = 100;
```

```
open(Robot video);
image(I, 'XData', x_ima, 'YData', y_ima, "AlphaData", 0.75);
axis xy
hold on
    plot( ...
        Position_theoric(1,2:end), ...
        Position_theoric(2,2:end), ...
        'k', ...
        'Color', 'b', ...
        "LineWidth",2 ...
    title('Robot video'); xlabel ('x [meters]'); ylabel ('y [meters]')
    plot( ...
        Position_estimation(1,2:end), ...
        Position_estimation(2,2:end), ...
        'k', ...
        'Color','r', ...
        "LineWidth",2 ...
    )
    for i=2:length(delta_th)
        k_m=mod(i,speed);
        if k_m==0
            delete(r_patch);
            delete(s_patch);
            Robot_pose=transl(Position_estimation(1,i),Position_estimation(2,i),0)*trotz(Orient
            Robot_tr=Robot_pose*Robot; % moving the robot
            Laser_fp=Robot_pose*transl(0.25,0,0)*[Lfp.v zeros(21,1) ones(21,1)]';
            axis([0 35.9 0 23.31]);
            r_patch = patch(Robot_tr(1,:), Robot_tr(2,:),'m');
            s_patch = patch('Faces',Lfp.f,'Vertices',Laser_fp(1:2,:)','FaceColor','m','FaceAlph
            %pause(0.01);
            frame=getframe(gcf);
            writeVideo(Robot_video, frame);
        end
    end
hold off
```



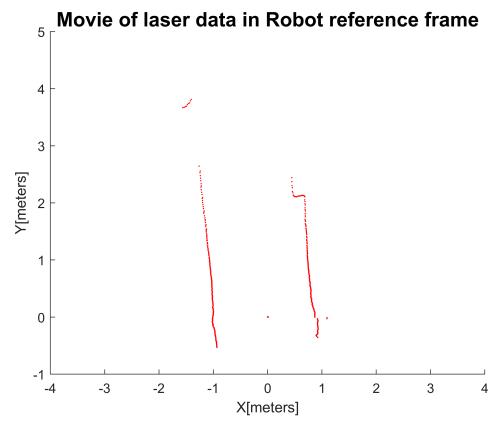
close(Robot_video)

Laser data (10%)

Make a movie of laser data in Robot reference frame.

```
Ts_laser=0.4;
laser_data = polar_laser_data(:,2:683) / 1000;
n_data_by_sampling = size(laser_data,2);
n_sampling = size(laser_data,1);
ploted_barrier = [];
Laser_video = VideoWriter('Laser_Data_Video.avi');
Laser_video.FrameRate = 10;
Laser_video.Quality = 100;
open(Laser video);
laser_points_L=[];
laser_points_R=[];
figure
hold on
title("Movie of laser data in Robot reference frame", 'FontSize',15);
axis([-4 4 -1 5]);
xlabel('X[meters]');ylabel('Y[meters]');
ang_res = 0.3515;
ang = [1:length(laser_data)]*ang_res - 30;
ang = ang * pi/180;
```

```
[x,y] = pol2cart(ang,laser_data);
T_R_L = transl(0.6,0,0)*trotz(deg2rad(-90));
for i = 1:n_sampling
    delete(ploted_barrier);
    laser_points_R(1,:,i) = x(i,:);
    laser_points_R(2,:,i) = y(i,:);
    laser_points_R(3,:,i) = zeros(1,n_data_by_sampling);
    laser_points_R(4,:,i) = ones(1,n_data_by_sampling);
    ploted_barrier=scatter( ...
        laser_points_R(1,:,i), ...
        laser_points_R(2,:,i), ...
        1, ...
        'red', ...
        'filled' ...
    );
    frame=getframe(gcf);
   writeVideo(Laser_video,frame);
end
hold off
```



close(Laser_video)

Localization (40%)

Localize the Robot by using the Similarity Transform.

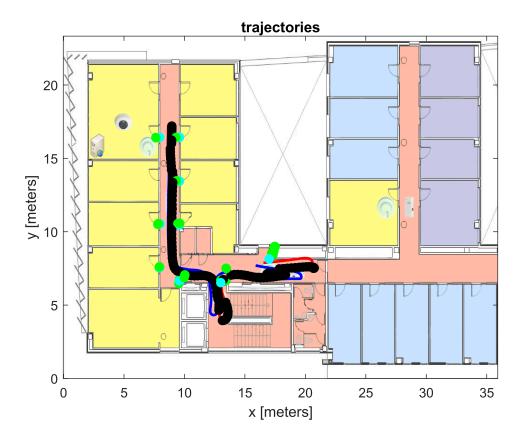
There are two asummtions:

(1) No correction of the trajectory is done.

- 1.- Make a 'log' with the following columns: time stamp, estimated noisy pose, corrected position, number of Land-marks, and the errors.
- 2.- Display in a figure: the map, the noisy trajectory with a visible 'mark' of the corrected position when Land-Mark are avalaible.

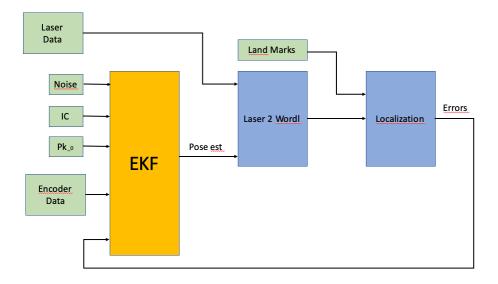
```
landmarks = [...
      7.934 16.431 0 1;...
      9.583 16.431 0 1;...
     9.584 13.444 0 1;...
     9.584 10.461 0 1;...
     7.973 10.534 0 1;...
     7.934 7.547 0 1;...
     9.584 6.654 0 1;...
     13.001 6.525 0 1;...
     17.007 8.136 0 1 ...
]';
Detected landmarks = [];
Reference_landmarks = [];
Log_no_correction = [];
image(I, 'XData', x_ima, 'YData', y_ima, "AlphaData", 0.75);
axis xy
hold on
    plot( ...
        Position_theoric(1,2:end), ...
        Position_theoric(2,2:end), ...
        'k', ...
        'Color', 'b', ...
        "LineWidth",2 ...
    )
    plot( ...
        Position estimation(1,2:end), ...
        Position_estimation(2,2:end), ...
        'k', ...
        'Color', 'r', ...
        "LineWidth",2 ...
    title('trajectories'); xlabel ('x [meters]'); ylabel ('y [meters]')
    for time=3:length(l_s_b(:,1))
        Robot_pose = transl(Position_estimation(1,time),Position_estimation(2,time),0)*trotz(On
```

```
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, 'c','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
        \mathsf{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), 50, 'black','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 not
        Single_log.Corrected_robot = Corrected_robot;
                                                                             % corrected pos
        Single_log.Number_of_landmarks = length(Detected_landmarks(1,:));
                                                                             % number of lar
        Single_log.Error = Error;
                                                                             % errors
        Log_no_correction = [Log_no_correction Single_log];
   end
   Detected_landmarks = [];
    Reference_landmarks = [];
end
```



(2) Correct the noisy trajectory.

Pay attention to the relationship of the variables



Make the data fusion of the PDF's, knowing that the Laser scanner has and acuraccy 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.

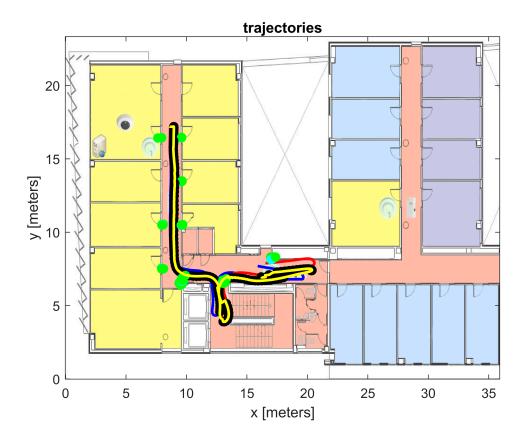
2.- Make a 'log' with the following columns: time stamp, estimated noisy pose, corrected position, number of Land-marks, and the errors.

```
Detected_landmarks = [];
Reference_landmarks = [];
Log_with_correction = [];
Position with correction(:,1) = zeros(size(Position theoric(:,1)));
Position_with_correction(:,2) = Position_theoric(:,2);
Orientation_with_correction = zeros(3004,1);
Orientation_with_correction(2) = Orientation_theoric(3,2);
Orientation estimation;
mu_noise = 0;
mu laser = 0;
mu_fused = 0;
image(I, 'XData', x_ima, 'YData', y_ima, "AlphaData", 0.75);
axis xy
hold on
    plot( ...
        Position_theoric(1,2:end), ...
        Position_theoric(2,2:end), ...
        'k', ...
        'Color', 'b', ...
        "LineWidth",2 ...
    )
    plot( ...
        Position_estimation(1,2:end), ...
        Position_estimation(2,2:end), ...
        'k', ...
        'Color','r', ...
        "LineWidth",2 ...
   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_the
        Robot pose = transl(Position_with_correction(1,time),Position_with_correction(2,time),@
        for landmark_index=1:length(landmarks(1,:))
            if l_s_b(time,landmark_index) > 0
                d=1 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, 'c','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));
   % ----- %
   % ---- trying to apply kalman's algorithm ---- %
   % NOTA: no es la primera prueba que hacemos, pero lo hemos
   % dejado así porque el código se entendía mejor de esta manera.
   % Somos conscientes de que no funciona la parte fused.
        % sigma_noise_x = ProcNoiseD;
        % sigma_noise_y = ProcNoiseTheta;
        % sigma noise theta = ProcNoiseTheta;
        % sigma_laser = (0.2 * 10^{-3})^2;
        % sigma_fused = (sigma_noise_x*sigma_laser)/(sigma_noise_x+sigma_laser);
        % sigma_fused_t = (sigma_noise_theta*sigma_laser)/(sigma_noise_theta+sigma_lase
        % error_fused_x = ((sqrt(2*pi*sigma_fused))^-1) * (exp(-((tx_ST-mu_fused)^2))
       % error_fused_y = ((sqrt(2*pi*sigma_fused))^-1) * (exp(-((ty_ST-mu_fused)^2) //

        % error fused theta = ((sqrt(2*pi*sigma fused t))^-1) * (exp(-((alpha ST-mu function))
        %
        % RotzTxy = [ ...
        % cos(error fused theta) sin(error fused theta) 0 error fused x; ...
```

```
% -sin(error fused theta) cos(error fused theta)
                                                                         0 error fused y; ...
                %
                           0
                                                   0
                                                            1
                                                                           0; ...
                %
                           0
                                                   0
                                                            0
                                                                           1 ...
                % ];
                %
                % cp(:,time) = RotzTxy * [Position_estimation(1,time) Position_estimation(2:time)
                % Position estimation(1,time)
                % Position_estimation(2,time)
                % Position_with_correction(1,time) = cp(1,time);
                % Position_with_correction(2,time) = cp(2,time);
            Error = transl(-tx_ST,-ty_ST,0)*trotz(alpha_ST);
            Corrected_robot = Error * [transl(Robot_pose); 1];
            scatter(Corrected_robot(1),Corrected_robot(2), 50, 'black','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 the
            Single_log.Time_stamp = time;
                                                                                  % timestamp is
            Single_log.Robot_pose = Robot_pose;
                                                                                  % estimated noi
            Single_log.Corrected_robot = Corrected_robot;
                                                                                  % corrected pos
            Single_log.Number_of_landmarks = length(Detected_landmarks(1,:));
                                                                                  % number of lar
            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','y', ...
        "LineWidth",2 ...
    )
hold off
```



Mapping (40%)

(1) Walls

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

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

```
Mapping_video = VideoWriter('Mapping_Video.avi');
Mapping_video.FrameRate = 10;
Mapping_video.Quality = 100;
open(Mapping_video);

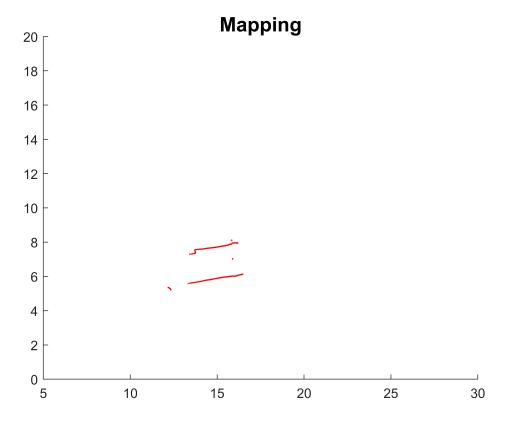
figure
hold on
title("Mapping", 'FontSize',15);

Ts_laser=0.4;
laser_data = log_urg(:,2:681) / 1000;

n_data_by_sampling = size(laser_data,2);
n_sampling = size(laser_data,1);

ploted_barrier = [];
ang_res = 0.3515;
```

```
ang = [1:length(laser data)] * ang res-30;
ang = ang * pi/180;
[x,y] = pol2cart(ang,laser data);
T_R_L = transl(0.6,0,0)*trotz(deg2rad(-90));
laser_points_L = [];
for landmark_i = 1:length(Log_with_correction)
    log = Log_with_correction(landmark_i);
    i = log.Time_stamp / 20;
    if i < 150
        delete(ploted barrier);
        mapping = [];
        laser_points_L(1,:,i) = x(i,:);
        laser_points_L(2,:,i) = y(i,:);
        laser_points_L(3,:,i) = zeros(1,n_data_by_sampling);
        laser_points_L(4,:,i) = ones(1,n_data_by_sampling);
        for j = 1:length(laser_points_L(1,:,i))
            final_pose= log.Robot_pose * T_R_L * laser_points_L(:,j,i);
            mapping = [mapping; final_pose(1) final_pose(2)];
        end
        ploted_barrier = scatter( ...
            mapping(:,1), ...
            mapping(:,2), ...
            1, ...
            'red', ...
            'filled' ...
        );
        axis ([5 30 0 20])
        frame=getframe(gcf);
        writeVideo(Mapping_video,frame);
    end
end
% end
hold off
```



close(Mapping_video);

(2) Occupancy grid

Use Breshehan algorithm to build the map. Do it only when Land Marks are avalaible 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 for inspiration

```
figure
hold on
title("Ocuppancy Grid", 'FontSize',15);

A = ones(100,100);
s=pcolor (A);
colormap(gray(2))
s.EdgeColor = [0.9 0.9 1];
axis ([0 30 0 30]);
drawnow;

for landmark_i = 1:length(Log_with_correction)
    log = Log_with_correction(landmark_i);
    i = log.Time_stamp / 20;

if i < 150
    robot_position = transl(log.Robot_pose);</pre>
```

```
for j = 1:length(laser_points_L(1,:,i))
    final_pose= log.Robot_pose * T_R_L * laser_points_L(:,j,i);
    l_xy=bresenham([robot_position(1) robot_position(2)], [final_pose(1) final_pose(2)]

    for j = 1:length(l_xy(:,1))
        A(l_xy(j,2),l_xy(j,1))=0; % Hemos tenido que poner las x e y rotadas para que send
    end
end
s = pcolor (A);
s.EdgeColor = [0.9 0.9 1];
colormap(gray(2));
drawnow;

end
end
hold off
```

