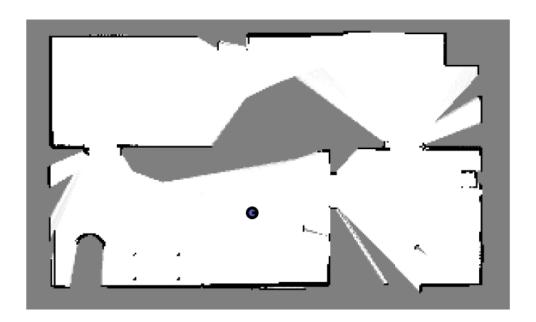
Rob521 - Assignment 2 Morgan Tran Tranmorg 1006331159

## Question #1 -build an occupancy grid map

In this part, we had to implement an occupancy grip mapping algorithm that builds the map from perfect ground-truth localization. This is completely identical to the ROB521 Lab3 mapping that we had to complete and thus I implemented the method that my team used in that lab.

By iterating through all of the lidar scans, and proceeding if the lidar measurement is within the laser range limits (r\_min and r\_max) and if it wasn't a nan value. We converted the robot pose and the lidar end position to the map frame. Ray traced to see where the lidar did not find an obstacle with the end of the lidar scan being an obstacle. We then updated the log map with the beta and alpha values for free and occupied space (0.5 for beta and 3 for alpha). And then put that through the log-odds to recover the map at that time frame. Do that for all time frames.

In the below map, black is occupied (alpha) and white is free (beta), while grey are unknown.

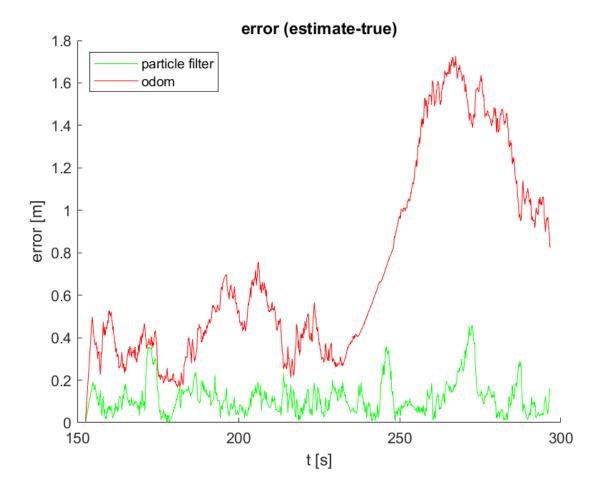


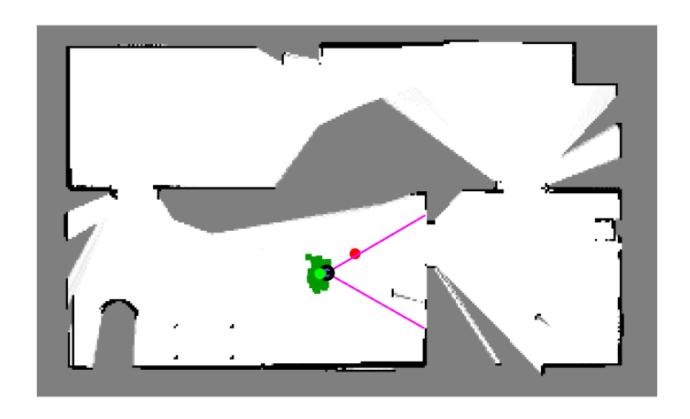
## Question #2 - localization from an occupancy grid map using particle filter

In this part, we had to implement a particle filter localization algorithm to localize the robot. We only used two scans as this was outlined in the doc and was sufficient to localize. Noise was added to the wheel odometry and to the laser scan. We first estimated the y\_laser given the particle pose and information on the laser and occupancy map. To do so we just propagated and found the first occupied cell from the ray.

The next thing was to calculate the weight of the particle given a normal distribution with the mean of the predicted and a given variance. This was done using normpdf.

The below graph shows the error of the particle filter and error of odom.





CODE BELOW

```
% ass3 q1.m
% =======
% This assignment will introduce you to the idea of first building an
% occupancy grid then using that grid to estimate a robot's motion using a
% particle filter.
% There are two questions to complete (5 marks each):
%
    Question 1: code occupancy mapping algorithm
응
    Question 2: see ass3_q2.m
% Fill in the required sections of this script with your code, run it to
% generate the requested plot/movie, then paste the plots into a short report
% that includes a few comments about what you've observed. Append your
% version of this script to the report. Hand in the report as a PDF file
% and the two resulting AVI files from Questions 1 and 2.
% requires: basic Matlab, 'gazebo.mat'
% T D Barfoot, January 2016
clear all;
% set random seed for repeatability
rng(1);
% =============
% load the dataset from file
ground truth poses: t true x true y true theta true
% odometry measurements: t_odom v_odom omega_odom
           laser scans: t laser y laser
2
    laser range limits: r_min_laser r_max_laser
    laser angle limits: phi_min_laser phi_max_laser
load gazebo.mat;
% Question 1: build an occupancy grid map
% Write an occupancy grid mapping algorithm that builds the map from the
% perfect ground-truth localization. Some of the setup is done for you
% below. The resulting map should look like "ass2_q1_soln.png". You can
% watch the movie ass2_q1_soln.mp4 to see what the entire mapping process
% should look like. At the end you will save your occupancy grid map to
% the file "occmap.mat" for use in Question 2 of this assignment.
% allocate a big 2D array for the occupancy grid
```

**%** =======

1

```
% resolution of occ grid
ogres = 0.05;
ogxmin = -7;
                                % minimum x value
                                % maximum x value
ogxmax = 8;
ogymin = -3;
                                % minimum y value
                                % maximum y value
ogymax = 6;
ognx = (ogxmax-ogxmin)/ogres;
                                % number of cells in x direction
ogny = (ogymax-ogymin)/ogres; % number of cells in y direction
                                % occupancy grid in log-odds format
oglo = zeros(ogny,ognx);
ogp = zeros(ogny,ognx);
                                % occupancy grid in probability format
% precalculate some quantities
numodom = size(t_odom,1);
npoints = size(y laser,2);
angles = linspace(phi_min_laser, phi_max_laser, npoints);
dx = ogres*cos(angles);
dy = ogres*sin(angles);
% interpolate the noise-free ground-truth at the laser timestamps
t_interp = linspace(t_true(1),t_true(numodom),numodom);
x_interp = interp1(t_interp,x_true,t_laser);
y_interp = interp1(t_interp,y_true,t_laser);
theta_interp = interp1(t_interp,theta_true,t_laser);
omega_interp = interp1(t_interp,omega_odom,t_laser);
% set up the plotting/movie recording
vid = VideoWriter('ass2 q1.avi');
open(vid);
figure(1);
clf;
pcolor(oqp);
colormap(1-gray);
shading('flat');
axis equal;
axis off;
M = getframe;
writeVideo(vid,M);
BETA = 0.5;
ALPHA = 3;
% loop over laser scans (every fifth)
for t=1:5:size(t_laser,1)
    % ----insert your occupancy grid mapping algorithm here-----
    for i = 1:npoints
        if ~isnan(y_laser(t,i)) && r_min_laser < y_laser(t,i) && y_laser(t,i)</pre>
 < r_max_laser
```

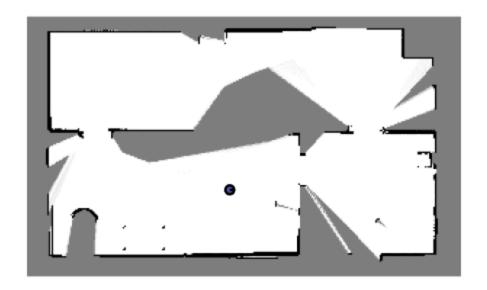
```
x  start = (x  interp(t)-ogxmin);
        y_start = (y_interp(t)-ogymin);
        range_mes = y_laser(t,i);
        cur_angle = angles(i) + theta_interp(t);
        % x_end and y_end calculation
        x_end = x_start + range_mes * cos(cur_angle);
        y_end = y_start + range_mes * sin(cur_angle);
        % convert to map coordinates
        x_start = round(x_start/ogres);
        y_start = round(y_start/ogres);
        x_end = round(x_end/ogres);
        y_end = round(y_end/ogres);
        % ray trace
        [rr, cc] = ray_trace(x_start, y_start, x_end, y_end);
        [row_bound, col_bound] = size(oglo);
                % iterate over each cell index pair
        for j=1:size(rr, 1)
            % skip over any instances where index is outside map
            if rr(j) > row\_bound \mid \mid cc(j) > col\_bound \mid \mid rr(j) <= 0 \mid \mid ...
                    cc(j) \ll 0
                continue
            end
            % update log-odds
            if j < (size(rr,1))
                % all cells except last cell are set as free
                oglo(rr(j), cc(j)) = oglo(rr(j), cc(j)) - BETA;
            else
                % last cell is set as occupied
                oglo(rr(j), cc(j)) = oglo(rr(j), cc(j)) + ALPHA;
            end
        end
    end
end
oglo;
% recovering probabilities from log-odds
ogp = exp(oglo)./(1 + exp(oglo));
```

%convert to map frame

```
% ----end of your occupancy grid mapping algorithm-----
    % draw the map
    clf;
   pcolor(ogp);
    colormap(1-gray);
    shading('flat');
   axis equal;
   axis off;
    % draw the robot
   hold on;
   x = (x_{interp(t)-ogxmin)/ogres;
    y = (y_interp(t)-ogymin)/ogres;
    th = theta_interp(t);
    r = 0.15/ogres;
    set(rectangle( 'Position', [x-r y-r 2*r 2*r], 'Curvature', [1
 1]), 'LineWidth', 2, 'FaceColor', [0.35 0.35 0.75]);
    set(plot([x x+r*cos(th)]', [y y+r*sin(th)]', 'k-'), 'LineWidth',2);
    % save the video frame
   M = getframe;
   writeVideo(vid,M);
   pause(0.1);
end
close(vid);
print -dpng ass2_q1.png
save occmap.mat ogres ogxmin ogxmax ogymin ogymax ognx ogny oglo ogp;
% given two points in a matrix, returns the diagonal indices
function [rr, cc] = ray_trace(x1, y1, x2, y2)
    % Initialize the return arrays
   rr = [];
   cc = [];
    % Calculate the x and y distances between the two points
   dx = abs(x2 - x1);
   dy = abs(y2 - y1);
    % Calculate the step size for the x and y axes
    if x1 < x2
        sx = 1;
    else
```

```
sx = -1;
end
if y1 < y2
   sy = 1;
else
   sy = -1;
end
% Initialize the error term
err = dx - dy;
% Traverse the line using Bresenham's algorithm
while true
   % Add the current point to the return arrays
   rr = [rr; y1];
   cc = [cc; x1];
    % Check if we've reached the end point
    if x1 == x2 && y1 == y2
        break;
    end
    % Update the error term
    e2 = 2*err;
    if e2 > -dy
        err = err - dy;
        x1 = x1 + sx;
    end
    if e2 < dx
        err = err + dx;
        y1 = y1 + sy;
    end
end
```

end



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```
% =======
% ass3 q2.m
% =======
% This assignment will introduce you to the idea of first building an
% occupancy grid then using that grid to estimate a robot's motion using a
% particle filter.
% There are three questions to complete (5 marks each):
%
    Question 1: see ass3 q1.m
응
    Question 2: code particle filter to localize from known map
% Fill in the required sections of this script with your code, run it to
% generate the requested plot/movie, then paste the plots into a short report
% that includes a few comments about what you've observed. Append your
% version of this script to the report. Hand in the report as a PDF file
% and the two resulting AVI files from Questions 1 and 2.
% requires: basic Matlab, 'gazebo.mat', 'occmap.mat'
% T D Barfoot, January 2016
clear all;
% set random seed for repeatability
rng(1);
% load the dataset from file
ground truth poses: t true x true y true theta true
% odometry measurements: t_odom v_odom omega_odom
          laser scans: t laser y laser
    laser range limits: r_min_laser r_max_laser
    laser angle limits: phi_min_laser phi_max_laser
load gazebo.mat;
% -----
% load the occupancy map from question 1 from file
% ogres: resolution of occ grid
% ogxmin: minimum x value
% ogxmax: maximum x value
% ogymin: minimum y value
% ogymax: maximum y value
  ognx: number of cells in x direction
   ogny: number of cells in y direction
   oglo: occupancy grid in log-odds format
   ogp: occupancy grid in probability format
```

```
load occmap.mat;
% Question 2: localization from an occupancy grid map using particle filter
% Write a particle filter localization algorithm to localize from the laser
% rangefinder readings, wheel odometry, and the occupancy grid map you
% built in Question 1. We will only use two laser scan lines at the
% extreme left and right of the field of view, to demonstrate that the
% algorithm does not need a lot of information to localize fairly well. To
% make the problem harder, the below lines add noise to the wheel odometry
% and to the laser scans. You can watch the movie "ass2 q2 soln.mp4" to
% see what the results should look like. The plot "ass2_q2_soln.png" shows
% the errors in the estimates produced by wheel odometry alone and by the
% particle filter look like as compared to ground truth; we can see that
% the errors are much lower when we use the particle filter.
% interpolate the noise-free ground-truth at the laser timestamps
numodom = size(t odom,1);
t_interp = linspace(t_true(1),t_true(numodom),numodom);
x_interp = interp1(t_interp,x_true,t_laser);
y_interp = interp1(t_interp,y_true,t_laser);
theta interp = interp1(t interp, theta true, t laser);
omega_interp = interp1(t_interp,omega_odom,t_laser);
% interpolate the wheel odometry at the laser timestamps and
% add noise to measurements (yes, on purpose to see effect)
v_interp = interp1(t_interp,v_odom,t_laser) + 0.2*randn(size(t_laser,1),1);
omega_interp = interp1(t_interp,omega_odom,t_laser) +
 0.04*randn(size(t_laser,1),1);
% add noise to the laser range measurements (yes, on purpose to see effect)
% and precompute some quantities useful to the laser
y laser = y laser + 0.1*randn(size(y laser));
npoints = size(y_laser,2);
angles = linspace(phi_min_laser, phi_max_laser,npoints);
dx = ogres*cos(angles);
dy = ogres*sin(angles);
y_laser_max = 5; % don't use laser measurements beyond this distance
% particle filter tuning parameters (yours may be different)
% noise on longitudinal speed for propagating particle
v_noise = 0.2;
u_noise = 0.2;
                     % noise on lateral speed for propagating particle
omega noise = 0.04;
                     % noise on rotational speed for propagating particle
laser_var = 0.5^2; % variance on laser range distribution
% generate an initial cloud of particles
x_particle = x_true(1) + 0.5*randn(nparticles,1);
y particle = y true(1) + 0.3*randn(nparticles,1);
theta_particle = theta_true(1) + 0.1*randn(nparticles,1);
```

```
% compute a wheel odometry only estimate for comparison to particle
% filter
x_{odom_only} = x_{true}(1);
y odom only = y true(1);
theta_odom_only = theta_true(1);
% error variables for final error plots - set the errors to zero at the start
pf err(1) = 0;
wo_err(1) = 0;
% set up the plotting/movie recording
vid = VideoWriter('ass2_q2.avi');
open(vid);
figure(2);
clf;
hold on;
pcolor(oqp);
set(plot( (x_particle-ogxmin)/ogres, (y_particle-ogymin)/ogres, 'g.'
),'MarkerSize',10,'Color',[0 0.6 0]);
set(plot( (x_odom_only-ogxmin)/ogres, (y_odom_only-ogymin)/ogres, 'r.'
),'MarkerSize',20);
x = (x_{interp}(1) - ogxmin) / ogres;
y = (y_interp(1)-ogymin)/ogres;
th = theta interp(1);
r = 0.15/ogres;
set(rectangle( 'Position', [x-r y-r 2*r 2*r], 'Curvature', [1
1]), 'LineWidth', 2, 'FaceColor', [0.35 0.35 0.75]);
set(plot([x x+r*cos(th)]', [y y+r*sin(th)]', 'k-'),'LineWidth',2);
set(plot( (mean(x_particle)-ogxmin)/ogres, (mean(y_particle)-ogymin)/
ogres, 'g.' ), 'MarkerSize', 20);
colormap(1-gray);
shading('flat');
axis equal;
axis off;
M = getframe;
writeVideo(vid,M);
% loop over laser scans
for i=2:size(t laser,1)
    % update the wheel-odometry-only algorithm
    dt = t_laser(i) - t_laser(i-1);
    v = v_{interp(i)};
    omega = omega_interp(i);
    x_{odom_only} = x_{odom_only} + dt*v*cos(theta_odom_only);
    y odom only = y odom only + dt*v*sin( theta odom only );
    phi = theta_odom_only + dt*omega;
    while phi > pi
        phi = phi - 2*pi;
    while phi < -pi</pre>
        phi = phi + 2*pi;
    end
    theta_odom_only = phi;
```

```
% loop over the particles
   for n=1:nparticles
       % propagate the particle forward in time using wheel odometry
       % (remember to add some unique noise to each particle so they
       % spread out over time)
       v = v interp(i) + v noise*randn(1);
       u = u noise*randn(1);
       omega = omega_interp(i) + omega_noise*randn(1);
       x_particle(n) = x_particle(n) + dt*(v*cos( theta_particle(n) ) -
u*sin( theta_particle(n) ));
       y particle(n) = y particle(n) + dt*(v*sin(theta particle(n)) +
u*cos( theta_particle(n) ));
       phi = theta_particle(n) + dt*omega;
       while phi > pi
           phi = phi - 2*pi;
       end
       while phi < -pi
           phi = phi + 2*pi;
       end
       theta_particle(n) = phi;
       % pose of particle in initial frame
       T = [cos(theta_particle(n)) -sin(theta_particle(n)) x_particle(n); ...
            sin(theta_particle(n)) cos(theta_particle(n)) y_particle(n); ...
       % compute the weight for each particle using only 2 laser rays
       % (right=beam 1 and left=beam 640)
       w particle(n) = 1.0;
       for beam=1:2
           % we will only use the first and last laser ray for
           % localization
           if beam==1 % rightmost beam
           elseif beam==2 % leftmost beam
               j = 640;
           end
           % -----insert your particle filter weight calculation here -----
           if ~isnan(y_laser(i,j)) && y_laser(i,j) < y_laser_max</pre>
               row pos = max(1,round((y particle(n)-ogymin)/ogres));
               col_pos = max(1,round((x_particle(n)-ogxmin)/ogres));
               threshold = 0.5;
               %get predicted range: y_laser_pred
               y_laser_pred = predict_laser_range(row_pos, col_pos, ...
                   theta_particle(n), angles(j), ogp, ogres, threshold);
               % calculate weight using gaussian pdf function
```

```
w_particle(n) = w_particle(n)*w_gain*normpdf(y_laser(i,
  j), ...
                                                y_laser_pred, sqrt(laser_var));
                             end
                             % ----end of your particle filter weight calculation-----
                   end
         end
         % resample the particles using Madow systematic resampling
         w_bounds = cumsum(w_particle)/sum(w_particle);
         w target = rand(1);
         j = 1;
         for n=1:nparticles
                while w_bounds(j) < w_target</pre>
                           j = mod(j,nparticles) + 1;
                end
                x_particle_new(n) = x_particle(j);
                y_particle_new(n) = y_particle(j);
                theta_particle_new(n) = theta_particle(j);
                w_target = w_target + 1/nparticles;
                if w_target > 1
                          w_target = w_target - 1.0;
                           j = 1;
                end
         end
         x_particle = x_particle_new;
         y_particle = y_particle_new;
         theta_particle = theta_particle_new;
         % save the translational error for later plotting
         pf_err(i) = sqrt( (mean(x_particle) - x_interp(i))^2 + (mean(y_particle) -
  y_interp(i))^2 );
         wo_err(i) = sqrt((x_odom_only - x_interp(i))^2 + (y_odom_only - y_odom_only - y_odom
  y_interp(i))^2 );
         % plotting
         figure(2);
         clf;
         hold on;
         pcolor(ogp);
         set(plot( (x_particle-ogxmin)/ogres, (y_particle-ogymin)/ogres, 'g.'
  ), 'MarkerSize', 10, 'Color', [0 0.6 0]);
         set(plot( (x_odom_only-ogxmin)/ogres, (y_odom_only-ogymin)/ogres, 'r.'
  ), 'MarkerSize', 20);
         x = (x_{interp(i)-ogxmin)/ogres;
         y = (y_interp(i)-ogymin)/ogres;
         th = theta_interp(i);
         if ~isnan(y_laser(i,1)) & y_laser(i,1) <= y_laser_max</pre>
                 set(plot([x x+y_laser(i,1)/ogres*cos(th+angles(1))]', [y y
+y_laser(i,1)/ogres*sin(th+angles(1))]', 'm-'),'LineWidth',1);
```

```
end
    if ~isnan(y laser(i,640)) & y laser(i,640) <= y laser max</pre>
       set(plot([x x+y_laser(i,640)/ogres*cos(th+angles(640))]', [y y
+y_laser(i,640)/ogres*sin(th+angles(640))]', 'm-'),'LineWidth',1);
    end
    r = 0.15/ogres;
    set(rectangle( 'Position', [x-r y-r 2*r 2*r], 'Curvature', [1
 1]), 'LineWidth', 2, 'FaceColor', [0.35 0.35 0.75]);
    set(plot([x x+r*cos(th)]', [y y+r*sin(th)]', 'k-'), 'LineWidth',2);
    set(plot( (mean(x_particle)-ogxmin)/ogres, (mean(y_particle)-ogymin)/
ogres, 'g.' ), 'MarkerSize', 20);
    colormap(1-gray);
    shading('flat');
    axis equal;
    axis off;
    % save the video frame
    M = getframe;
    writeVideo(vid,M);
    pause(0.01);
end
close(vid);
% final error plots
figure(3);
clf;
hold on;
plot( t_laser, pf_err, 'g-' );
plot( t_laser, wo_err, 'r-' );
xlabel('t [s]');
ylabel('error [m]');
legend('particle filter', 'odom', 'Location', 'NorthWest');
title('error (estimate-true)');
print -dpng ass2_q2.png
% returns the expected laser measurement given the particle's
% position and map
function y_exp = predict_laser_range(row, col, theta, beam_angle, map, ogres,
 thresh)
    new_angle = atan2(sin(theta+beam_angle), cos(theta+beam_angle));
    incr = 0;
    r p = row;
    c_p = col;
    % angle is between -pi/4 and pi/4
    if -pi/4<=new_angle && new_angle<=pi/4</pre>
```

```
c_{inc} = 1;
       r inc = tan(new angle);
       a = cos(new_angle);
   % angle is between -3pi/4 and 3pi/4
   elseif 3*pi/4<=new_angle || new_angle<=-3*pi/4</pre>
       c_{inc} = -1;
       r_inc = -tan(new_angle);
       a = cos(new_angle);
   % angle is between pi/4 and 3pi/4
   elseif pi/4<new_angle && new_angle<3*pi/4</pre>
       c_inc = 1/tan(new_angle);
       r_{inc} = 1;
       a = sin(new_angle);
   % angle is between -pi/4 and -3pi/4
   else
       c_inc = -1/tan(new_angle);
       r_{inc} = -1;
       a = sin(new_angle);
   end
   [row_bound, col_bound] = size(map);
   while r_p > 0 && c_p>0 && r_p<=row_bound && c_p<=col_bound && map(r_p,
c_p) < thresh</pre>
       incr = incr +1;
       r_p = row + round(incr * r_inc);
       c_p = col + round(incr * c_inc);
   end
   y_exp = abs((incr/a)*ogres);
```

end

