

ROB521

Assignment 2: Wheel Odometry

Jonathan Spraggett
Student number: 1003958737

March 26th 2023

Abstract

This assignment involves working with a dataset from a mobile robot simulation environment to estimate the motion of a mobile robot using wheel odometry and building a simple map. The task consists of three questions: write code for a noise-free wheel odometry algorithm, add noise to the data, re-run the wheel odometry algorithm, and build a map from ground truth and noisy wheel odometry data.

1 Question 1

Noise-free wheel odometry algorithm

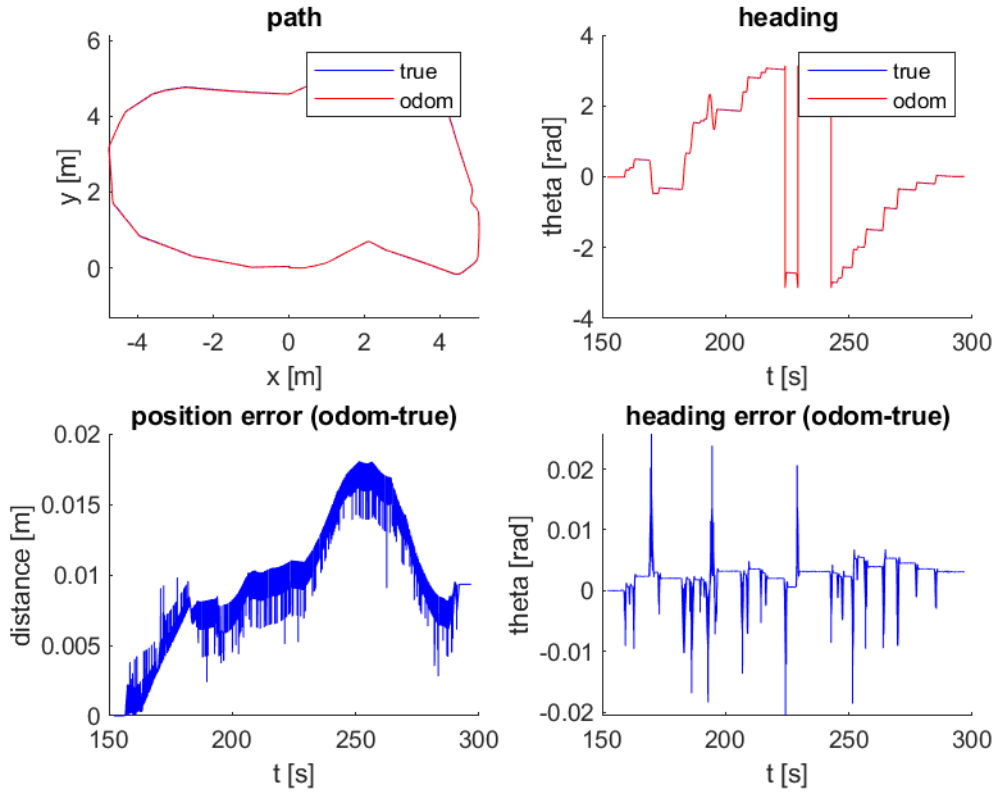


Figure 1

An algorithm for wheel odometry was utilized to calculate the current position of the robot based on its linear velocity and angular velocity from the previous estimation. Figure 1 depicts the position, orientation, and deviation of the robot's pose from the actual position. The plots indicate that since there is no noise in the linear velocity and angular velocity, the odometry estimation closely resembles the ground truth, and the errors are negligible.

2 Question 2

Noise wheel odometry algorithm

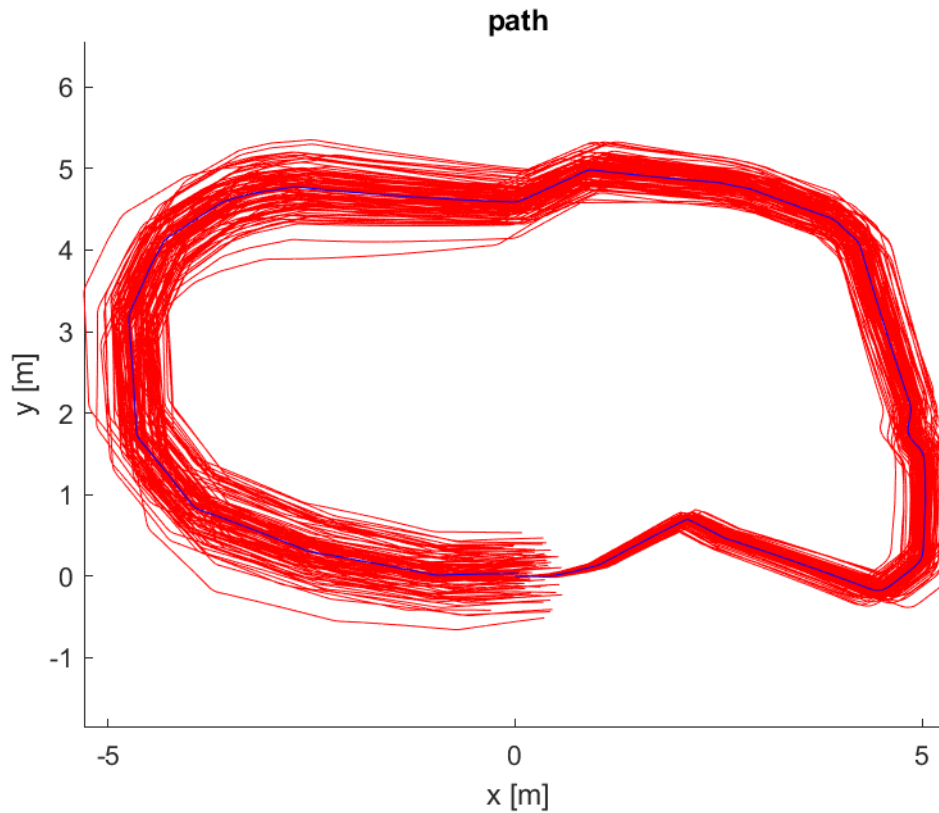


Figure 2

The algorithm for wheel odometry was rerun with 100 simulations, incorporating random noise in the linear and angular velocities. Figure 2 illustrates that as time progresses, the error from the actual path increases due to unbounded uncertainty in dead reckoning. Hence, it can be inferred that the accumulation of errors over time is due to the impact of random noise on the algorithm's estimation.

3 Question 3

Build a map from noisy and noise-free wheel odometry

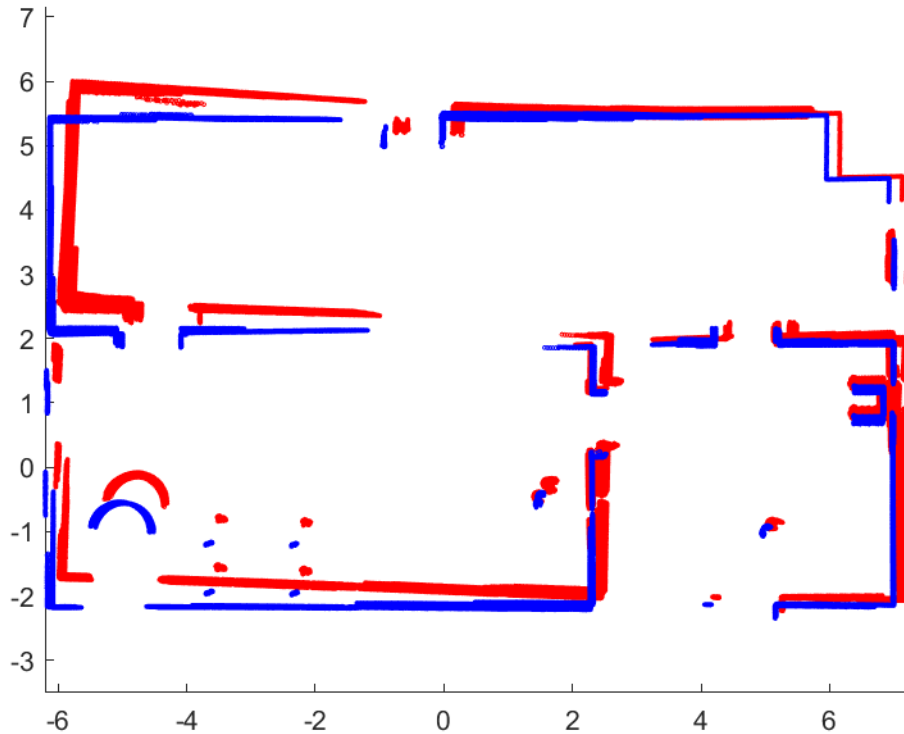


Figure 3

Laser scans were utilized to construct a comprehensive map of the surrounding environment. However, to achieve this, the laser scans needed to undergo multiple transformations from the sensor frame to the vehicle frame, and then to the inertial frame, via a transformation matrix. The generated map was depicted in Figure 3, where the blue map was obtained using error-free odometry, while the red map utilized noisy odometry. As demonstrated in Figure 3, the robot's movements around the map led to the accumulation of noise-induced errors in the odometry, ultimately resulting in an increasing misalignment of the laser scans as the robot continued to move.

Matlab Code

```

1  % =====
2  % ROB521_assignment2.m
3  % =====
4  %
5  % This assignment will introduce you to the idea of estimating the motion
6  % of a mobile robot using wheel odometry, and then also using that wheel
7  % odometry to make a simple map. It uses a dataset previously gathered in
8  % a mobile robot simulation environment called Gazebo. Watch the video,
9  % 'gazebo.mp4' to visualize what the robot did, what its environment
10 % looks like, and what its sensor stream looks like.
11 %
12 % There are three questions to complete (5 marks each):
13 %
14 %     Question 1: code (noise-free) wheel odometry algorithm
15 %     Question 2: add noise to data and re-run wheel odometry algorithm
16 %     Question 3: build a map from ground truth and noisy wheel odometry
17 %
18 % Fill in the required sections of this script with your code, run it to
19 % generate the requested plots, then paste the plots into a short report
20 % that includes a few comments about what you've observed. Append your
21 % version of this script to the report. Hand in the report as a PDF file.
22 %
23 % requires: basic Matlab, 'ROB521_assignment2_gazebo_data.mat'
24 %
25 % T D Barfoot, December 2015
26 %
27 clear all;
28
29 % set random seed for repeatability
30 rng(1);
31
32 % =====
33 % load the dataset from file
34 % =====
35 %
36 %     ground truth poses: t_true x_true y_true theta_true
37 %     odometry measurements: t_odom v_odom omega_odom
38 %     laser scans: t_laser y_laser
39 %     laser range limits: r_min_laser r_max_laser
40 %     laser angle limits: phi_min_laser phi_max_laser
41 %
42 load ROB521_assignment2_gazebo_data.mat;
43
44 % =====

```

```

45 % Question 1: code (noise-free) wheel odometry algorithm
46 % =====
47 %
48 % Write an algorithm to estimate the pose of the robot throughout motion
49 % using the wheel odometry data (t_odom, v_odom, omega_odom) and assuming
50 % a differential-drive robot model. Save your estimate in the variables
51 % (x_odom y_odom theta_odom) so that the comparison plots can be generated
52 % below. See the plot 'ass2-q1-soln.png' for what your results should look
53 % like.
54
55 % variables to store wheel odometry pose estimates
56 numodom = size(t_odom,1);
57 x_odom = zeros(numodom,1);
58 y_odom = zeros(numodom,1);
59 theta_odom = zeros(numodom,1);
60
61 % set the initial wheel odometry pose to ground truth
62 x_odom(1) = x_true(1);
63 y_odom(1) = y_true(1);
64 theta_odom(1) = theta_true(1);
65
66 % -----insert your wheel odometry algorithm here-----
67 for i=2:numodom
68     dt = t_odom(i) - t_odom(i-1);
69     x_odom(i) = x_odom(i-1) + v_odom(i-1)*cos(theta_odom(i-1))*dt;
70     y_odom(i) = y_odom(i-1) + v_odom(i-1)*sin(theta_odom(i-1))*dt;
71     theta_odom(i) = theta_odom(i-1) + omega_odom(i-1)*dt;
72
73     theta_odom(i) = wrapToPi(theta_odom(i));
74
75 end
76 % -----end of your wheel odometry algorithm-----
77
78 % plot the results for verification
79 figure(1)
80 clf;
81
82 subplot(2,2,1);
83 hold on;
84 plot(x_true, y_true, 'b');
85 plot(x_odom, y_odom, 'r');
86 legend('true', 'odom');
87 xlabel('x [m]');
88 ylabel('y [m]');
89 title('path');
90 axis equal;

```

```

91
92 subplot(2,2,2);
93 hold on;
94 plot(t_true,theta_true,'b');
95 plot(t_odom,theta_odom,'r');
96 legend('true','odom');
97 xlabel('t [s]');
98 ylabel('theta [rad]');
99 title('heading');
100
101 subplot(2,2,3);
102 hold on;
103 pos_err = zeros(numodom,1);
104 for i=1:numodom
105     pos_err(i) = sqrt((x_odom(i)-x_true(i))^2 + (y_odom(i)-y_true(i))^2);
106 end
107 plot(t_odom,pos_err,'b');
108 xlabel('t [s]');
109 ylabel('distance [m]');
110 title('position error (odom-true)');
111
112 subplot(2,2,4);
113 hold on;
114 theta_err = zeros(numodom,1);
115 for i=1:numodom
116     phi = theta_odom(i) - theta_true(i);
117     while phi > pi
118         phi = phi - 2*pi;
119     end
120     while phi < -pi
121         phi = phi + 2*pi;
122     end
123     theta_err(i) = phi;
124 end
125 plot(t_odom,theta_err,'b');
126 xlabel('t [s]');
127 ylabel('theta [rad]');
128 title('heading error (odom-true)');
129 print -dpng ass2_q1.png
130
131
132 % =====
133 % Question 2: add noise to data and re-run wheel odometry algorithm
134 % =====
135 %
136 % Now we're going to deliberately add some noise to the linear and

```

```

137 % angular velocities to simulate what real wheel odometry is like. Copy
138 % your wheel odometry algorithm from above into the indicated place below
139 % to see what this does. The below loops 100 times with different random
140 % noise. See the plot 'ass2_q2_soln.pdf' for what your results should look
141 % like.
142
143 % save the original odometry variables for later use
144 v_odom_noisefree = v_odom;
145 omega_odom_noisefree = omega_odom;
146
147 % set up plot
148 figure(2);
149 clf;
150 hold on;
151
152 % loop over random trials
153 for n=1:100
154
155     % add noise to wheel odometry measurements (yes, on purpose to see
156     % effect)
157     v_odom = v_odom_noisefree + 0.2*randn(numodom,1);
158     omega_odom = omega_odom_noisefree + 0.04*randn(numodom,1);
159
160     % -----insert your wheel odometry algorithm here-----
161     for i=2:numodom
162         dt = t_odom(i) - t_odom(i-1);
163         x_odom(i) = x_odom(i-1) + v_odom(i-1)*cos(theta_odom(i-1))*dt;
164         y_odom(i) = y_odom(i-1) + v_odom(i-1)*sin(theta_odom(i-1))*dt;
165         theta_odom(i) = theta_odom(i-1) + omega_odom(i-1)*dt;
166
167         theta_odom(i) = wrapToPi(theta_odom(i));
168     end
169     % -----end of your wheel odometry algorithm-----
170
171     % add the results to the plot
172     plot(x_odom, y_odom, 'r');
173 end
174
175 % plot ground truth on top and label
176 plot(x_true, y_true, 'b');
177 xlabel('x [m]');
178 ylabel('y [m]');
179 title('path');
180 axis equal;
181 print -dpng ass2_q2.png

```



```

182
183
184 % =====
185 % Question 3: build a map from noisy and noise-free wheel odometry
186 % =====
187 %
188 % Now we're going to try to plot all the points from our laser scans in the
189 % robot's initial reference frame. This will involve first figuring out
190 % how to plot the points in the current frame, then transforming them back
191 % to the initial frame and plotting them. Do this for both the ground
192 % truth pose (blue) and also the last noisy odometry that you calculated in
193 % Question 2 (red). At first even the map based on the ground truth may
194 % not look too good. This is because the laser timestamps and odometry
195 % timestamps do not line up perfectly and you'll need to interpolate. Even
196 % after this, two additional patches will make your map based on ground
197 % truth look as crisp as the one in 'ass2_q3_soln.png'. The first patch is
198 % to only plot the laser scans if the angular velocity is less than
199 % 0.1 rad/s; this is because the timestamp interpolation errors have more
200 % of an effect when the robot is turning quickly. The second patch is to
201 % account for the fact that the origin of the laser scans is about 10 cm
202 % behind the origin of the robot. Once your ground truth map looks crisp,
203 % compare it to the one based on the odometry poses, which should be far
204 % less crisp, even with the two patches applied.
205
206 % set up plot
207 figure(3);
208 clf;
209 hold on;
210
211 % precalculate some quantities
212 npoints = size(y_laser,2);
213 angles = linspace(phi_min_laser, phi_max_laser, npoints);
214 cos_angles = cos(angles);
215 sin_angles = sin(angles);
216
217 for n=1:2
218
219     if n==1
220         % interpolate the noisy odometry at the laser timestamps
221         t_interp = linspace(t_odom(1), t_odom(numodom), numodom);
222         x_interp = interp1(t_odom, x_odom, t_interp);
223         y_interp = interp1(t_odom, y_odom, t_interp);
224         theta_interp = interp1(t_odom, theta_odom, t_interp);
225         omega_interp = interp1(t_odom, omega_odom, t_interp);
226     else
227         % interpolate the noise-free odometry at the laser timestamps

```

```

228     t_interp = linspace(t_true(1),t_true(numodom),numodom);
229     x_interp = interp1(t_interp,x_true,t_laser);
230     y_interp = interp1(t_interp,y_true,t_laser);
231     theta_interp = interp1(t_interp,theta_true,t_laser);
232     omega_interp = interp1(t_interp,omega_odom,t_laser);
233 end
234
235 % loop over laser scans
236 for i=1:size(t_laser,1)
237
238     % —————insert your point transformation algorithm here—————
239
240     if abs(omega_interp(i)) >= 0.1
241         continue;
242     end
243
244     TIV = [cos(theta_interp(i)), -sin(theta_interp(i)), 0, x_interp(i);
245           sin(theta_interp(i)), cos(theta_interp(i)), 0, y_interp(i);
246           0,0,1,0;
247           0,0,0,1];
248
249     rpv = [(y_laser(i,:) - 0.1).*cos_angles; (y_laser(i,:) - 0.1).*
250           sin_angles; zeros(1, npoints); ones(1, npoints)];
251
252     rpi = TIV * rpv;
253
254     if n==1
255         scatter(rpi(1,:), rpi(2,:), 2, 'r');
256
257     else
258         scatter(rpi(1,:), rpi(2,:), 2, 'b');
259     end
260
261     % —————end of your point transformation algorithm—————
262 end
263 end
264
265 axis equal;
266 print -dpng ass2_q3.png

```