

[◀ Return to "Self-Driving Car Engineer" in the classroom](#)[DISCUSS ON STUDENT HUB](#)

# Kidnapped Vehicle

## REVIEW

## CODE REVIEW 2

## HISTORY

### ▼ src/particle\_filter.cpp 2

```
1 /**
2  * particle_filter.cpp
3  *
4  * Created on: Dec 12, 2016
5  * Author: Tiffany Huang
6  */
7
8 #include "particle_filter.h"
9
10 #include <math.h>
11 #include <algorithm>
12 #include <iostream>
13 #include <iterator>
14 #include <numeric>
15 #include <random>
16 #include <string>
17 #include <vector>
18
19 #include "helper_functions.h"
20
21 using std::string;
22 using std::vector;
23
24
25 void ParticleFilter::init(double x, double y, double theta, double std[]) {
26     /**
```

```

27  * Set the number of particles. Initialize all particles's position and weig
28  */
29
30  // The number of particles was selected based on suggestion from project
31  // overview by Udacity:
32  // https://www.youtube.com/watch?v=-3HI3Iw3Z9g&feature=youtu.be
33  // I found, however, that as small as 10 and as large as 1000+ number would
34  // With smallest value, the error grows, but can still pass the acceptable t
35  // very small value of particles such as 5, some lucky initialization can pa
36  // With larger number of particles, such as 1000, the program becomes too sl
37  num_particles = 100;
38
39  // Random engine for particle pose noise generation.
40  std::default_random_engine gen;

```



#### SUGGESTION

This is actually pseudo random, we are always generating the same particles during every run of ParticleFi

we can create a truly random generator with something like below

```

random_device rd;
default_random_engine gen(rd());

```

```

41
42  // Sensor noise distributions and associated variables.
43  std::normal_distribution<double> x_dist(0, std[0]);
44  std::normal_distribution<double> y_dist(0, std[1]);
45  std::normal_distribution<double> theta_dist(0, std[2]);
46  double x_noise = 0.0, y_noise = 0.0, theta_noise = 0.0;
47
48  // Initializing all particles.
49  for (int i = 0; i < num_particles; i++) {
50      Particle p;
51      x_noise = x_dist(gen);
52      y_noise = y_dist(gen);
53      theta_noise = theta_dist(gen);
54
55      p.id = i;
56      p.x = x + x_noise;
57      p.y = y + y_noise;
58      p.theta = theta + theta_noise;
59      // All initial weights are one per assignment requirement.
60      p.weight = 1.0;
61
62      particles.push_back(p);
63  }
64
65  is_initialized = true;
66 }
67
68 void ParticleFilter::prediction(double delta_t, double std_pos[],
69                                double velocity, double yaw_rate) {
70  /**
71   * Use measurement to calculate prediction for particular state.
72   * Adds random Gaussian noise to each particle state.
73   */

```

```

75 std::default_random_engine gen;
76 std::normal_distribution<double> x_dist(0, std_pos[0]);
77 std::normal_distribution<double> y_dist(0, std_pos[1]);
78 std::normal_distribution<double> theta_dist(0, std_pos[2]);
79 double x_noise, y_noise, theta_noise = 0;
80
81 for (int i = 0; i < num_particles; i++) {
82     x_noise = x_dist(gen);
83     y_noise = y_dist(gen);
84     theta_noise = theta_dist(gen);
85
86     // Considering two cases: when the car drives straight (yaw_rate is close
87     // and car is turning. Similarly to Kalman filter project, we do not direc
88     // yaw rate to zero angle of steering, but rather check zero's vicinity.
89     // Interestingly, I found the noise to be very important. Without
90     // adding noise the main particle will steer off the trajectory of the car
91     if (fabs(yaw_rate) < 0.001) {
92         particles[i].x += velocity * delta_t * cos(particles[i].theta) + x_noise
93         particles[i].y += velocity * delta_t * sin(particles[i].theta) + y_noise
94         particles[i].theta += theta_noise;
95     }
96     else {
97         particles[i].x += velocity / yaw_rate
98             * (sin(particles[i].theta + yaw_rate*delta_t) - sin(particles[i].the
99         particles[i].y += velocity / yaw_rate
100             * (cos(particles[i].theta) - cos(particles[i].theta + yaw_rate*delta
101         particles[i].theta += yaw_rate * delta_t + theta_noise;
102     }
103 }
104 }
105
106 void ParticleFilter::dataAssociation(vector<LandmarkObs> predicted,
107                                     vector<LandmarkObs>& observations) {
108     /**
109     * Find the predicted measurement that is closest to each observed measureme
110     * and assign the observed measurement to this particular landmark.
111     */
112
113     for (unsigned int i = 0; i < observations.size(); i++) {
114
115         // Observations are already in global coordinate system, thus can be dir
116         LandmarkObs curr_observation = observations[i];
117
118         // Initializing the minimum distance and matching index of the landmark.
119         double min_distance = std::numeric_limits<double>::max();
120         int match_index = std::numeric_limits<int>::min();
121
122         // Going through each observable landmark and getting the best match to
123         // Best match is found through nearest neighbor algorithm. Assuming the
124         // observable landmarks is greater than the actual number of observation
125         for (unsigned int j = 0; j < predicted.size(); j++) {
126             LandmarkObs curr_landmark = predicted[j];
127
128             double curr_distance = dist(curr_observation.x, curr_observation.y,
129                                         curr_landmark.x, curr_landmark.y);
130
131             if (curr_distance < min_distance) {
132                 min_distance = curr_distance;
133                 match_index = curr_landmark.id;
134             }
135         }

```

```

136
137     // Recording the match between observation and landmark.
138     observations[i].id = match_index;
139 }
140 }
141
142 void ParticleFilter::updateWeights(double sensor_range, double std_landmark[],
143                                   const vector<LandmarkObs> &observations,
144                                   const Map &map_landmarks) {
145     /**
146     * Update the weights of each particle using a multi-variate Gaussian
147     * distribution.
148     *
149     * The observations are given in the VEHICLE'S coordinate system.
150     * The particles are located according to the MAP'S coordinate system.
151     * Thus, the transform between the two systems is performed.
152     */
153     for (int i = 0; i < num_particles; i++) {
154         Particle p = particles[i];
155         // Product of multi-variate Gaussian.
156         double multivar_prod = 1;
157
158         // The subset of landmarks that fall within the sensor range.
159         vector<LandmarkObs> landmarks_p;
160
161         // Populating the observable subset of landmarks.
162         for (unsigned int j = 0; j < map_landmarks.landmark_list.size(); j++) {
163             Map::single_landmark_s l = map_landmarks.landmark_list[j];
164
165             //The subset is defined by radial distance between observation and a lan
166             if (dist(l.x_f, l.y_f, p.x, p.y) <= sensor_range){
167                 landmarks_p.push_back(LandmarkObs{l.id_i, l.x_f, l.y_f});
168             }
169         }
170
171         // Transforming observations into global coordinate system.
172         vector<LandmarkObs> transformed_os;
173         for (unsigned int j = 0; j < observations.size(); j++) {
174             LandmarkObs o = observations[j], t;
175
176             t.id = o.id;
177             t.x = cos(p.theta)*o.x - sin(p.theta)*o.y + p.x;
178             t.y = sin(p.theta)*o.x + cos(p.theta)*o.y + p.y;
179             transformed_os.push_back(t);
180         }
181
182         // Match subset of landmarks to their observations.
183         dataAssociation(landmarks_p, transformed_os);
184
185         for (unsigned int j = 0; j < transformed_os.size(); j++) {
186             LandmarkObs o = transformed_os[j];
187             // Recording the matched x, y, and id in the following var.
188             LandmarkObs m;
189             m.id = o.id;
190
191             // Scanning the subset to get the donor for coordinates.
192             for (unsigned int k = 0; k < landmarks_p.size(); k++) {
193                 LandmarkObs p = landmarks_p[k];
194                 if (p.id == m.id) {
195                     m.x = p.x;
196                     m.y = p.y;

```

```

197     }
198 }
199
200 // Obtaining new weight for this observation with multivariate Gaussian.
201 double std_x = std_landmark[0];
202 double std_y = std_landmark[1];
203 double new_weight = ( 1/(2*M_PI*std_x*std_y)) * exp( -( pow(m.x-o.x,2)/(
204     + (pow(m.y-o.y,2)/(2*pow(std_y, 2))) ) );
205
206 // Updating multivariate product.
207 multivar_prod *= new_weight;
208 }
209 // Weight is updated on actual particle from the list as it is used downst
210 particles[i].weight = multivar_prod;
211 }
212 }
213
214 void ParticleFilter::resample() {
215     /**
216      * Re-sample particles with replacement with probability proportional
217      * to their weight.
218      */
219     vector<Particle> new_particles;
220     // Using library implementation of discrete distribution generator
221     // with weights. Replaces the re-sample wheel algorithm.
222     std::default_random_engine gen;
223
224     // Need to create iterable for required distribution parameter.
225     vector<double> weights;
226     for (int i = 0; i < num_particles; i++) {
227         weights.push_back(particles[i].weight);
228     }
229
230     for (int i = 0; i < num_particles; ++i) {
231         std::discrete_distribution<> discrete_dist(weights.begin(), weights.end(
232         new_particles.push_back(particles[discrete_dist(gen)]);
233     }
234
235     particles = new_particles;

```

#### SUGGESTION

We can avoid the deep copy of vector data by using move semantics like below. See [here](#) for more details.

```
particles = std::move(new_particles);
```

```

236 }
237
238 void ParticleFilter::SetAssociations(Particle& particle,
239     const vector<int>& associations,
240     const vector<double>& sense_x,
241     const vector<double>& sense_y) {
242     // particle: the particle to which assign each listed association,
243     // and association's (x,y) world coordinates mapping
244     // associations: The landmark id that goes along with each listed associatio
245     // sense_x: the associations x mapping already converted to world coordinate
246     // sense_y: the associations y mapping already converted to world coordinate
247     particle.associations= associations;

```

```
248 particle.sense_x = sense_x;
249 particle.sense_y = sense_y;
250 }
251
252 string ParticleFilter::getAssociations(Particle best) {
253     vector<int> v = best.associations;
254     std::stringstream ss;
255     copy(v.begin(), v.end(), std::ostream_iterator<int>(ss, " "));
256     string s = ss.str();
257     s = s.substr(0, s.length()-1); // get rid of the trailing space
258     return s;
259 }
260
261 string ParticleFilter::getSenseCoord(Particle best, string coord) {
262     vector<double> v;
263
264     if (coord == "X") {
265         v = best.sense_x;
266     } else {
267         v = best.sense_y;
268     }
269
270     std::stringstream ss;
271     copy(v.begin(), v.end(), std::ostream_iterator<float>(ss, " "));
272     string s = ss.str();
273     s = s.substr(0, s.length()-1); // get rid of the trailing space
274     return s;
275 }
276
```

► src/particle\_filter.h

► src/map.h

► src/main.cpp

► src/helper\_functions.h

► data/map\_data.txt

► cmakepatch.txt

► README.md

► CMakeLists.txt

RETURN TO PATH

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