# **Homework 5: Car Tracking Report**

## Part I. Implementation (20%):

#### Part 1:

```
def observe(self, agentX: int, agentY: int, observedDist: float) -> None:

# BEGIN_YOUR_CODE (our solution is 9 lines of code, but don't worry if you deviate from this)

for row in range(self.belief.getNumRows()):

for col in range(self.belief.getNumCols()):

    u = math.sqrt((agentY - rowToY(row)) ** 2 + (agentX - colToX(col)) ** 2)

    p = util.pdf(u, Const.SONAR_STD, observedDist)

    self.belief.setProb(row, col, p * self.belief.getProb(row, col))

self.belief.normalize()

return

raise Exception("Not implemented yet")

# END_YOUR_CODE
```

## Explanation of part 1:

For every tile, we first compute the true distance from agent to the tile as our mean of Gaussian distribution. Then, with the mean(u), Const.SONAR\_STD, and observed distance, we can get the probability by calling util.pdf(). Thus, we are able to update  $P(H_t|E_{1:t})$  with previous belief and product of  $P(E_t|H_t)$  and  $P(H_t|E_{1:t-1})$  by calling belief.setProb(). Finally, normalize the probability we just updated for every tile because sum of them may exceed 1.

#### Part 2:

```
def elapseTime(self) -> None:
    if self.skipElapse: # ONLY FOR THE GRADER TO USE IN Part 1
        return
    # BEGIN_YOUR_CODE (our solution is 10 lines of code, but don't worry if you deviate from this)
    new_belief = util.Belief(self.belief.getNumRows(), self.belief.getNumCols(), value=0)
    for oldTile, newTile in self.transProb:
        new_belief.addProb(newTile[0], newTile[1], self.belief.getProb(*oldTile) * self.transProb[(oldTile, newTile)])
    new_belief.normalize()
    self.belief = new_belief
    return
    raise Exception("Not implemented yet")
# END_YOUR_CODE
```

## Explanation of part 2:

First, create a new belief which will be assigned to self.belief after finish computing the  $P(H_{t+1}|E_{1:t})$  to new belief by transition probability of every (old\_tile, new\_tile) ( $P(H_{t+1}|H_t)$  and  $P(H_t|E_{1:t})$ . And then, like part 1, we should do normalization of probability to ensure sum of probability would not exceed 1.

### Part 3-1:

```
def observe(self, agentX: int, agentY: int, observedDist: float) -> None:
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              observe distribution = collections.defaultdict(float)
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              for row, col in self.particles:
                 u = math.sqrt((agentY - rowToY(row)) ** 2 + (agentX - colToX(col)) ** 2)
                 p = util.pdf(u, Const.SONAR_STD, observedDist)
                 observe_distribution[(row, col)] = self.particles[(row, col)] * p
              new_particles = collections.defaultdict(int)
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              for _ in range(self.NUM_PARTICLES):
                 particle = util.weightedRandomChoice(observe_distribution)
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                 new_particles[particle] += 1
              self.particles = new_particles
              self.updateBelief()
```

## Explanation of part 3-1:

In the first half of part 3-1, we do same thing as part 1, the only thing different is that we replace self.belief with self.particles[tile] (descrbes the number of particles on particular tile) and observe distribution (record the new distribution of particles).

Then, create new\_particle which will be assigned to self.particles after we resample the particles in next half. In the second half of part 3-1, we call util.weightedRandomChoice() with observe\_distribution to get the tile where the number of particles will increase.

#### Part 3-2:

## Explanation of part 3-2:

This part is also similar to part 2 roughly, and we call function util.weightedRandomChoice() with transition probability for every particle to get new particles distribution. Finally, assign new particles distribution to self.particles and finish part 3-2.