## Machine Learning for Robotics: Assignment 1

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# Contents

1	$\mathbf{Ass}$	ignment 1	2
	1.1	Exercise 1: GMM Parameters	2
	1.2	Exercise 2	2
	1.3	Exercise 3	4
		1.3.1 Task 1	4
		1.3.2 Task 2	4
		1.3.3 Task 3	2

## Chapter 1

## Assignment 1

#### Exercise 1: GMM Parameters 1.1

$$\pi = \begin{pmatrix} 0.2404 \\ 0.2615 \\ 0.2971 \\ 0.201 \end{pmatrix}$$

$$\mu = \begin{pmatrix} -0.04318 & 0.02622 & -0.01932 & -0.01473 \\ 0.04459 & 0.06172 & -0.01668 & -0.07965 \end{pmatrix}$$

$$\Sigma_1 = \begin{pmatrix} 0.0001751 & 0.000262 \\ 0.000262 & 0.0003983 \end{pmatrix}$$

$$\Sigma_2 = \begin{pmatrix} 0.001082 & -0.0004242 \\ -0.0004242 & 0.0002431 \end{pmatrix}$$

$$\Sigma_3 = \begin{pmatrix} 0.0007435 & -0.0005916 \\ -0.0005916 & 0.0006103 \end{pmatrix}$$

$$(1.5)$$

$$\mu = \begin{pmatrix} -0.04318 & 0.02622 & -0.01932 & -0.01473 \\ 0.04459 & 0.06172 & -0.01668 & -0.07965 \end{pmatrix}$$
 (1.2)

$$\Sigma_1 = \begin{pmatrix} 0.0001751 & 0.000262\\ 0.000262 & 0.0003983 \end{pmatrix}$$
 (1.3)

$$\Sigma_2 = \begin{pmatrix} 0.001082 & -0.0004242 \\ -0.0004242 & 0.0002431 \end{pmatrix}$$
 (1.4)

$$\Sigma_3 = \begin{pmatrix} 0.0007435 & -0.0005916 \\ -0.0005916 & 0.0006103 \end{pmatrix}$$
 (1.5)

$$\Sigma_4 = \begin{pmatrix} 0.0003941 & 0.0002164 \\ 0.0002164 & 0.0001274 \end{pmatrix}$$
 (1.6)

#### Exercise 2 1.2

The Loglikelihoods in 1.1 show that none of the sequences are larger than -120, which classifies them all to sequence 2.

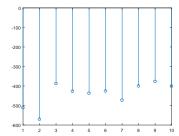


Figure 1.1: Loglikelihoods of the sequences

## 1.3 Exercise 3

### 1.3.1 Task 1

### 1.3.2 Task 2

- 1.) In this Task  $\gamma$  was set to 0.99.
- 2.) A high  $\gamma$  value (close to 1) puts more weight on long term gains whereas low  $\gamma$  values focus on short term gain when it comes to the value.
- 3.) For this setting a average of 6 iterations were necessary for the algorithm to converge.
  - 4.) 1.2, 1.3

### 1.3.3 Task 3

- 1.) The parameters in this task were  $\epsilon = 0.01$  and  $\alpha = 0.1$ .
  - 2.) When testing a pure greedy policy the algorithm converges very slowly.
  - 3.) QLearning needs about 30000 steps to find a optimal policy.
  - 4.) 1.4, 1.5

AAAAAAAAAAAAAA

Figure 1.2: Resulting state sequence for initial state 10

AAAAAAAAAAAAAAA

Figure 1.3: Resulting state sequence for initial state 3

PARAQQARAQQARA

Figure 1.4: Resulting state sequence for initial state 5

Figure 1.5: Resulting state sequence for initial state 12