

# 16831 Statistical Techniques, Fall 2014: Problem Set 1

Due: Thursday, September 11, beginning of the class

## Instructions

There are 2 questions on this assignment (3 pages). These are short, simple problems. Included is the maximum answer length.

## 1 Markov Assumption

*This is not meant to be a tricky question, just one to get you thinking about an important assumption that is often made. One paragraph each at most.*

### 1.1

Give one robotic example where the Markov assumption is used (correctly or not). Is the assumption valid or not? Explain.

### 1.2

Give one real-world example where the Markov assumption is used (correctly or not). Is the assumption valid or not? Explain.

## 2 Bayes Filter Derivation

Recall the derivation for the Bayes Filter in the slides:

$$Bel(x_t) = P(x_t | u_{1:t}, z_{1:t}) \tag{1}$$

$$\dots \tag{2}$$

$$\propto P(z_t | x_t) \int P(x_t | u_t, x_{t-1}) P(x_{t-1} | u_{1:t}, z_{1:t-1}) dx_{t-1} \tag{3}$$

$$\propto P(z_t | x_t) \int P(x_t | u_t, x_{t-1}) P(x_{t-1} | u_{1:t-1}, z_{1:t-1}) dx_{t-1} \tag{4}$$

$$\propto P(z_t | x_t) \int P(x_t | u_t, x_{t-1}) Bel(x_{t-1}) dx_{t-1} \tag{5}$$

### 2.1

In the slides, the Markov assumption is invoked between lines (3) and (4) to drop  $u_t$ . Why is this incorrect? (Why does the Markov assumption not enable you to drop  $u_t$ )

### 2.2

Provide a counter-example where knowing  $u_t$  gives you information about the state  $x_{t-1}$ .

### 2.3

What does the book assume about the controls  $u$  in order to drop  $u_t$  from the derivation of the Bayes Filter? Is this assumption reasonable? Why or why not?

### 2.4

There are weaker assumptions you can make about the controls to still drop  $u_t$  from the Bayes Filter derivation. Think conditional independence, and derive how  $u_t$  is dropped between lines (3) and (4).