# Assignment #8 – Machine Learning – Professor Haugh

## **Jaime Gacitua**

## **jg3499**

## **Due Monday May 2, 2016**

## Question 1

1. Computing is solving the likelihood problem. This can be solved by using the relationship

We have , and

To get there we have to iterate over the.

For,

Moving to,

Finally, on,

Finally, using the relationship presented before,

1. To find we need to solve the smoothing problem – inferring the past. To get this value, we have to solve for the following relationship:

We already calculated. We have to do the backward recursion to get the:

For,

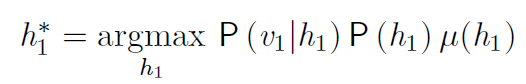
For,

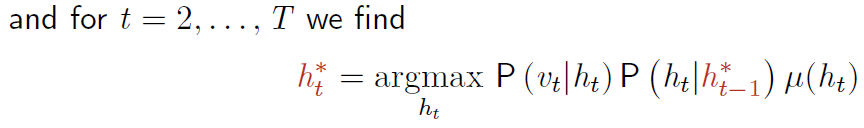
For,

Finally, using the above formula,

1. The most likely path problem can be solved using the Viterbi Algorithm,

Once we have solved all the way to we backtrack to obtain the most likely hidden path. We get





Solving the sequence, we get the most probable hidden state sequence is

The calculations can be found below.

|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
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| |  |  |  |  | | --- | --- | --- | --- | | t=3 |  |  |  | |  | h\_(t-1) |  |  | | h\_(t) | 1 | 2 | 3 | | 1 | 0.350 | 0.000 | 0.000 | | 2 | 0.120 | 0.240 | 0.000 | | **3** | 0.160 | **0.320** | 0.800 | | max | 0.3500 | 0.3200 | 0.8000 | |  |  |  |  | | h3 = |  | 3 |  | | |  |  |  |  | | --- | --- | --- | --- | | t=2 |  |  |  | |  | h\_(t-1) |  |  | | h\_(t) | 1 | 2 | 3 | | 1 | 0.053 | 0.000 | 0.000 | | **2** | **0.058** | 0.115 | 0.000 | | 3 | 0.032 | 0.064 | 0.160 | | max | 0.058 | 0.115 | 0.160 | |  |  |  |  | | h2 = | 2 |  |  | |
| |  |  | | --- | --- | | t=1 |  | |  | a | | h\_(t) | 1 | | **1** | **0.036** | | 2 | 0.005 | | 3 | 0.000 | | max | 0.036 | |  |  | | h1 = | 1 | |  |

## Question 2

1. First we use the definition of conditional probability, ,

Again, we use the same definition, but in the form,

By the structure of the HMM model, is independent from, therefore,

On the denominator, we use the law of total probabilities, conditioning over,

On the denominator, we use the definition of conditional probability, as, yielding,

By the structure of the HMM model, is independent from, therefore, we finally arrive to

With this expression, we see that

Represents the transmission matrix. Known.

Represents the solution to the filtering problem:

Assuming we have already solved for,

Finally,

1. Re-writing the smoothing problem by using total probabilities over ,

Using the definition of conditional probability,

Because of the structure of the HMM, is independent of, therefore,

If we define, then we can rewrite the last relationship as

We can then solve iteratively for, working backwards.

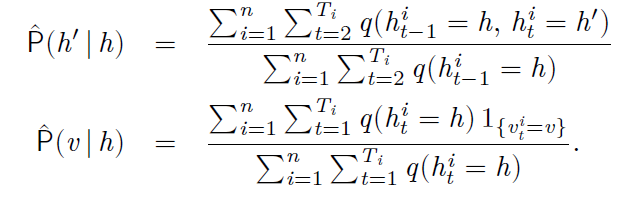
In addition, we must be clear on how to define the gamma of period, which is the same as the solution to the filtering problem,

## Question 3

Suppose the state space of the hidden sequence is of cardinality, and the visible state space has cardinality.

If we initialize the transition matrix, and emission matrix to uniformly constant values, we will have

When we update the parameters in the M-step, we have to apply the following equations,



Therefore,

If we start with uniformly constant values,

* The update of the matrix will be meaningless, because all of its elements will stay unchanged as.
* The update of the matrix will be meaningless, because all of its elements will have the same value, for all time periods and hidden states, with a value equal to,

## Question 4

(a)

Having,