# CSEP 590A Assignment 4 Hidden Markov Model Viterbi Algorithm Name: Dipak C Boyed

# 1. Introduction

In this project I have implemented the Hidden Markov Model (HMM) Viterbi algorithm in a console executable that traverses a given DNA sequence to find high GC content patches and output the following as the result to the console output:

Number and size of each high GC content patch in the DNA sequence and log probability (ln) of the Viterbi path.

In addition, I also implemented an E-M based Viterbi training model that iterates 10 times over the data to converge on the log probability of the Viterbi path and the GC content patches.

#### 1.1 Supported Environment

**NOTE**: All development was done on the Windows platform (OS: Vista) in Visual Studio 2005 and the programs were compiled and linked against .NET 2.0 framework in C#.

I was able to successfully build and use the program in the Windows platform.

#### 1.2 Source Code

The following major classes define the source code for my sequence alignment application:

File Name	Description
Program.cs	Container class for the console application
HMMViterbiAlgorithm.cs	Class representing the Viterbi algorithm to perform and common functionality.

Table 1 List of source files

#### 1.3 Program Input

The following figure of the Help menu describes the program input:

```
C:\Users\diboyed\Desktop\MLE\HMMUiterbi>DiBoyed_HMMUiterbi.exe

Help Menu: 'DiBoyed_HMMUiterbi.exe'

Performs HMM Viterbi algorithm on a given DNA sequence to find high GC patches.
Outputs the number and size of high GC content patches, log probability of the Viterbi path.
Also implements Viterbi training over 10 iterations.

Usage:
DiBoyed_HMMUiterbi.exe \( \sequence \> [-e \left\) \( \sequence \) [-t \( \tansition \) file \right\)]

Options:
\( \sequence1 \> FASTA file \) representing the DNA sequence.
\( \sequence2 \> (-e \) \( \sequence \) enission file \right\)]

Options:
\( \sequence1 \> FASTA file \) containing initial emission values.
\( \sequence1 \> (-e \) \( \sequence \) enission file \right\)]

Optional file containing initial transition values.

Notes:

By Default, uses the following emissions and transition values:

Emissions:

\( \text{A} \) C \( \text{G} \) T

\( \sequence2 \) LowGCState \( \sequence3 \) d.25, \( \text{0.25}, \text{0.25}, \) d.25

\( \text{HighGCState} \) \( \sequence3 \) d.20, \( \text{0.30}, \text{0.30}, \) d.30, \( \text{0.20} \)

Transitions:

\( \sequence3 \) \( \sequence3 \) d.2001

\( \text{HighGCState} \) \( \sequence3 \) d.2001

\( \sequence3 \) d.2001

\( \text{HighGCState} \) d.20 and (\text{Fansition file} \) must contain a ',' separated list of values only with one line for each row of values and the same order shown above.
```

Figure 1 Program Help Menu

#### 1.4 Program Output

The following figure shows an example of the results/output displayed by the application:

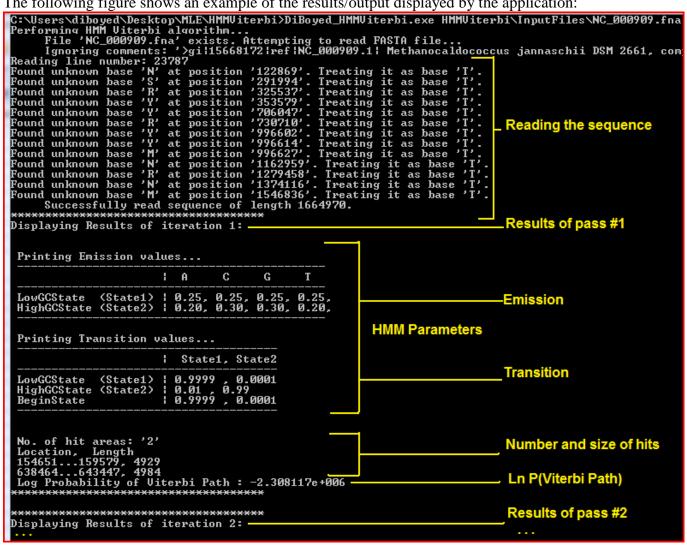


Figure 2 Sample program output

Besides writing current operations to console, the program writes a Results section at the end of each iteration (1 through 10):

(i) The HMM parameters (Emission and Transition values recalculated using Viterbi training).

NOTE: in order to make indexing easier, I list the BeginState transition values in the end of my transition matrix.

- (ii) The number of hits (patches with high GC content) and size of each such patch.
- (iii) The log probability of the Viterbi path.

# 2. Test Results

For all Dice examples I tested, I assumed the following:

- Rolls of 1-5 were treated as A
- Roll of 6 was treated as C
- State1 => Fair dice, State2 => Loaded dice.
- Hit area represents rolls where Viterbi predicts a 'Loaded dice' was used.
- Emissions and Transitions were supplied using the -e/-t optional switches in my application.

### 2.1 Simple Test Case: Casino Dice rolling '316664':

First, I ran the simple casino dice example presented in the lec06-casino-hmm.xls spreadsheet.

Input	DiBoyed_HMMViterbi.exe HMMViterbi\InputFiles\hw4diceSimple.fasta -e HMMViterbi\InputFiles\hw4DiceEmissions.txt -t HMMViterbi\InputFiles\hw4DiceSimpleTransitions.txt	
Output	OutputLogs\hw4DiceSimple.output.txt	

The results (number and size of hit areas as well as log probability of Viterbi path) matched the results provided in the spreadsheet

(http://www.cs.washington.edu/education/courses/csep590a/08au/slides/lec06-casino-hmm.xls)

## 2.2 Test Case: Casino Dice 300 rolls (Durbin fig 3.5)

Next, I ran the casino dice example of 300 rolls from Durbin fig 3.5.

Input	DiBoyed_HMMViterbi.exe HMMViterbi\InputFiles\hw4dice.fasta -e HMMViterbi\InputFiles\hw4DiceEmissions.txt -t HMMViterbi\InputFiles\hw4DiceTransitions.txt
Output	OutputLogs\hw4Dice.output.txt

The results (number and size of hit areas) exactly matched the results provided in the book and slides. The hit areas were found correctly in the very first iteration while the log probability of the Viterbi path converged after the 2<sup>nd</sup> iteration.

Results of 1<sup>st</sup> iteration:

No. of hit areas: '4' (Hit area is Loaded dice)

Location, Length

49...66, 18 79...112, 34 180...192, 13 271...289, 19

Log Probability of Viterbi Path: -5.351774e+002

#### 2.3 Test Results on NC 000909.fna

Finally, I ran the application on the NC\_0009-9.fna (FASTA file) which is the genome sequence data for *Methanocaldococcus jannaschii*.

```
Input DiBoyed_HMMViterbi.exe HMMViterbi\InputFiles\NC_000909.fna
Output OutputLogs\NC_000909.output.txt
```

The two figures below summarize the results of the 1<sup>st</sup> iteration and the 10<sup>th</sup> iteration:

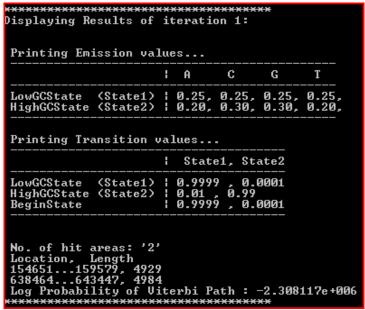


Figure 3 Results of 1st iteration

As mentioned in the homework description, I saw 2 subsequences each of length around ~5000.

Viterbi training was used to re-compute the emission and transition values and the computation was re-run for 10 iterations. The below table displays the no. of hit areas and log probability of the Viterbi path in each iteration:

Iteration	No. of Hits	Ln P(Viterbi Path)		
1	2	-2.308117E+06		
2	23	-2.188057E+06		
3	34	-2.187965E+06		
4	35	-2.187960E+06		
5	36	-2.187960E+06		
6	36	-2.187960E+06		
7	36	-2.187960E+06		
8	36	-2.187960E+06		
9	36	-2.187960E+06		
10	36	-2.187960E+06		

Table 2 Result summary for each iteration

Finally, the figure below shows the result of the 10<sup>th</sup> iteration:

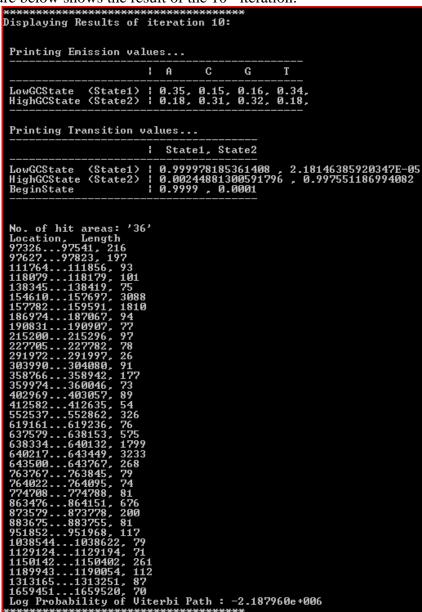


Figure 4 Results of 10th iteration

The results heavily overlapped the RNA location mentioned in the NC\_000909.rnt file.

I also traced the below graphs to confirm the convergence of the hit areas and log probability over the iterations:

