HMM

(All the screenshots for all questions are from our terminal)

The state transition matrix A is

	Н	С
Н	0.7	0.3
С	0.4	0.6

The observation matrix B is

	S	М	L
Н	0.1	0.4	0.5
С	0.7	0.2	0.1

The initial state distribution π is

	Н	С
π	0.6	0.4

This observation sequence is (S,M,S,L) = (0,1,0,2)

An HMM is defined by A, B and π (and, implicitly, by the dimensions N and M). The HMM is denoted by λ = (A, B, π).

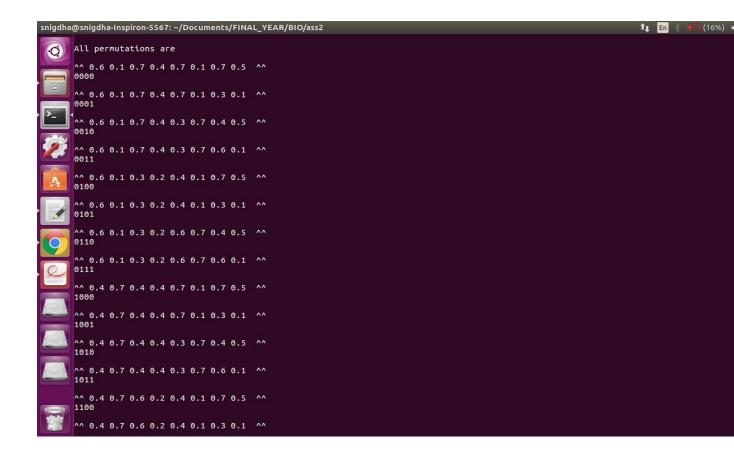
Consider a generic state sequence of length four X = (x0, x1, x2, x3) with corresponding observations O = (O0, O1, O2, O3).

Then $\pi x0$ is the probability of starting in state x0. Also, bx0 (O0) is the probability of initially observing O0 and ax0,x1 is the probability of transiting from state x0 to state x1. Continuing, we see that the probability of the state sequence X is given by

 $P(X, O) = \pi x_0 bx_0 (O_0)ax_0, x_1 bx_1 (O_1)ax_1, x_2 bx_2 (O_2)ax_2, x_3 bx_3 (O_3).$

NOTE: In our code, '0' stands for H & '1' stands for 'C'

So for some of the example sequences, probabilities are in the image below:



Using the above formula we output Table 1 which calculates the probability & normalized probability of all possible state sequences.

'printAllKLength' calls the recursive function 'printAllKLengthRec' which in turn calls 'table1_probability'. 'k' is the size of the observation sequence. In our e.g. it is 4. So we have to calculate probabilities of all possible state sequences from HHHH to CCCC i.e. in our code from 0000 to 1111.

So table-1 is displayed as follows:

```
TABLE-1 is
probability
                  normalized probability
0.0004121000
                   0.0427431000
0.0000351000
                   0.0036641000
0.0007061000
                   0.0732741000
0.0002121000
                   0.0219821000
0.0000501000
                   0.0052341000
0.0000041000
                   0.0004491000
                   0.0314031000
0.0003021000
0.0000911000
                   0.0094211000
0.0010981000
                   0.1139821000
0.0000941000
                   0.0097701000
0.0018821000
                   0.1953981000
0.0005641000
                   0.0586191000
0.0004701000
                   0.0488491000
0.0000401000
                   0.0041871000
0.0028221000
                   0.2930961000
0.0008471000
                   0.0879291000
```

Then for table 2:

Let $\lambda = (A, B, \pi)$ be a given model and let $O = (O0, O1, \dots, OT - 1)$ be a series of observations. We want to find $P(O \mid \lambda)$. Let $X = (x0, x1, \dots, xT - 1)$ be a state sequence.

We perform the forward algorithm, or α -pass. Alpha is defined as $\alpha t(i) = P(O0, O1, ..., Ot, xt = qi \mid \lambda)$ & calculate using,

$$\alpha_t(i) = \left[\sum_{j=0}^{N-1} \alpha_{t-1}(j) a_{ji}\right] b_i(\mathcal{O}_t)$$

The 'get_alpha' computes the ' α ' matrix. It returns the sum of the last row.

Then, we perform the backward algorithm, or β -pass. Beta is defined as $\beta t(i) = P(Ot+1, Ot+2, ..., OT -1 | xt = qi, \lambda)$ & calculate using,

$$\beta_t(i) = \sum_{j=0}^{N-1} a_{ij}b_j(\mathcal{O}_{t+1})\beta_{t+1}(j).$$

In our code, 'get_beta' function calculates 'β' matrix.

Now, for Gamma t = 0, 1, ..., T - 1 and $i = 0, 1, ..., N - 1, & is defined as, <math>yt(i) = P(xt = qi \mid O, \lambda)$

Since $\alpha t(i)$ measures the relevant probability up to time t and $\beta t(i)$ measures the relevant probability after time t, it is calculated as,

$$\gamma_t(i) = \frac{\alpha_t(i)\beta_t(i)}{P(\mathcal{O} \mid \lambda)}.$$

In our code, 'get_gamma' function first calculates ' γ -dash' matrix & then ' γ ' matrix by dividing each element of ' γ -dash' by the last row sum of ' α ' matrix.

Below is the screenshot of the alpha, beta, gamma-dash & the final gamma matrix.

```
ALPHA is
0.0600001000 0.2800001000
0.0616001000 0.0372001000
0.0058001000 0.0285601000
0.0077421000 0.0018881000
Last row sum of ALPHA is 0.0096301000
BETA is
0.0302001000 0.0279201000
0.0812001000 0.1244001000
0.3800001000 0.2600001000
1.0000001000 1.0000001000
GAMMA DASH is
0.0018121000 0.0078181000
0.0050021000 0.0046281000
0.0022041000 0.0074261000
0.0077421000 0.0018881000
GAMMA is
0.1881701000 0.8118301000
0.5194321000 0.4805681000
0.2288781000 0.7711221000
0.8039791000 0.1960211000
```