

CS221 Problem Session

Week 8

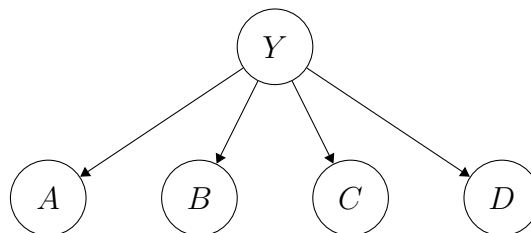
1) [CA session] Problem 1: The Bayesian Bag of Candies Model (Again)

You have a lot of candy left over from Halloween, and you decide to give them away to your friends. You have four types of candy: **A**pple, **B**anana, **C**aramel, **D**ark-Chocolate. You decide to prepare candy bags using the following process.

- For each candy bag, you first flip a (biased) coin Y which comes up heads ($Y = H$) with probability λ and tails ($Y = T$) with probability $1 - \lambda$.
- If Y comes up heads ($Y = H$), you make a **H**ealthy bag, where you:
 - (a) Add one **A**pple candy with probability p_1 or nothing with probability $1 - p_1$;
 - (b) Add one **B**anana candy with probability p_1 or nothing with probability $1 - p_1$;
 - (c) Add one **C**aramel candy with probability $1 - p_1$ or nothing with probability p_1 ;
 - (d) Add one **D**ark-Chocolate candy with probability $1 - p_1$ or nothing with probability p_1 .
- If Y comes up tails ($Y = T$), you make a **T**asty bag, where you:
 - (a) Add one **A**pple candy with probability p_2 or nothing with probability $1 - p_2$;
 - (b) Add one **B**anana candy with probability p_2 or nothing with probability $1 - p_2$;
 - (c) Add one **C**aramel candy with probability $1 - p_2$ or nothing with probability p_2 ;
 - (d) Add one **D**ark-Chocolate candy with probability $1 - p_2$ or nothing with probability p_2 .

For example, if $p_1 = 1$ and $p_2 = 0$, you would deterministically generate: **H**ealthy bags with one **A**pple and one **B**anana; and **T**asty bags with one **C**aramel and one **D**ark-Chocolate. For general values of p_1 and p_2 , bags can contain anywhere between 0 and 4 pieces of candy.

Denote A, B, C, D random variables indicating whether or not the bag contains candy of type **A**pple, **B**anana, **C**aramel, and **D**ark-Chocolate, respectively.



- (a) You realize you need to make more candy bags, but you've forgotten the probabilities you used to generate them. So you try to estimate them looking at the 5 bags you've already made:

<i>bag</i> 1 :	(H ealthy, { A pple, B anana})
<i>bag</i> 2 :	(T asty, { C aramel, D ark-Chocolate})
<i>bag</i> 3 :	(H ealthy, { A pple, B anana})
<i>bag</i> 4 :	(T asty, { C aramel, D ark-Chocolate})
<i>bag</i> 5 :	(H ealthy, { A pple, B anana})

Estimate λ, p_1, p_2 by maximum likelihood.

- (b) That was too easy, let's try again:

<i>bag</i> 1 :	(H ealthy, { A pple, B anana, C aramel})
<i>bag</i> 2 :	(T asty, { A pple, C aramel, D ark-Chocolate})
<i>bag</i> 3 :	(H ealthy, { B anana, C aramel})
<i>bag</i> 4 :	(T asty, { A pple, B anana, D ark-Chocolate})
<i>bag</i> 5 :	(H ealthy, { A pple, B anana})

Estimate λ, p_1, p_2 by maximum likelihood (i.e. counting and normalizing). Hint: Estimate $p_{1/2}$ or $1 - p_{1/2}$ but not both.

- (c) You find out your little brother had been playing with your candy bags, and had mixed them up (in a uniformly random way). Now you don't even know which ones were **H**ealthy and which ones were **T**asty. So you need to re-estimate λ, p_1, p_2 , but now without knowing whether the bags were **H**ealthy or **T**asty.

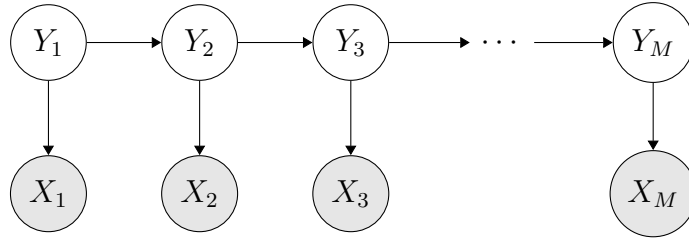
<i>bag</i> 1 :	(? , { A pple, B anana, C aramel})
<i>bag</i> 2 :	(? , { C aramel, D ark-Chocolate})
<i>bag</i> 3 :	(? , { A pple, B anana, C aramel})
<i>bag</i> 4 :	(? , { C aramel, D ark-Chocolate})
<i>bag</i> 5 :	(? , { A pple, B anana, C aramel})

You remember the EM algorithm is just what you need. Initialize with $\lambda = 0.5, p_1 = 0.5, p_2 = 0$, and run one step of the EM algorithm. Hint: You might use conditional probabilities found in last week's problem session (2d) for this problem.

(i) E-step:

(ii) M-step:

- (d) You decide to make candy bags according to a new process. You create the first one as described above. Then with probability μ , you create a second bag of the same type as the first one (Healthy or Tasty), and of different type with probability $1 - \mu$. Given this type, the bag is filled with candy as before. Then with probability μ , you create a third bag of the same type as the second one (Healthy or Tasty), and of different type with probability $1 - \mu$. And so on, you repeat the process M times. Denote Y_i, A_i, B_i, C_i, D_i the variables at each time step, for $i = 0, \dots, M$. Let $X_i = (A_i, B_i, C_i, D_i)$. Note that in the figure below, each X_i represents four separate nodes A_i, B_i, C_i, D_i , each with parent Y_i (just like the earlier figure).



Now you want to compute:

$$\mathbb{P}(Y_i = \mathbf{Healthy} \mid X_0 = (1, 1, 1, 0), \dots, X_i = (1, 1, 1, 0))$$

exactly for all $i = 0, \dots, M$, and you decide to use the forward-backward algorithm.

Suppose you have already computed the marginals:

$$f_i = \mathbb{P}(Y_i = \mathbf{Healthy} \mid X_0 = (1, 1, 1, 0), \dots, X_i = (1, 1, 1, 0))$$

for some $i \geq 0$. Recall the first step of the algorithm is to compute an intermediate result *proportional* to

$$\mathbb{P}(Y_{i+1} \mid X_0 = (1, 1, 1, 0), \dots, X_i = (1, 1, 1, 0), X_{i+1} = (1, 1, 1, 0))$$

- (i) Write an expression that is **proportional** to

$$\mathbb{P}(Y_{i+1} = \mathbf{Healthy} \mid X_0 = (1, 1, 1, 0), \dots, X_i = (1, 1, 1, 0), X_{i+1} = (1, 1, 1, 0))$$

in terms of f_i and the parameters p_1, p_2, λ, μ .

(ii) Write an expression that is **proportional** to

$$\mathbb{P}(Y_{i+1} = \mathbf{Tasty} \mid X_0 = (1, 1, 1, 0), \dots, X_i = (1, 1, 1, 0), X_{i+1} = (1, 1, 1, 0))$$

in terms of f_i and the parameters of the model p_1, p_2, λ, μ . The proportionality constant should be the same as in (i) (you don't need to find it).

(iii) Let h be the answer for part (i), and t for part (ii). Write an expression for

$$\mathbb{P}(Y_{i+1} = \mathbf{Healthy} \mid X_0 = (1, 1, 1, 0), \dots, X_i = (1, 1, 1, 0), X_{i+1} = (1, 1, 1, 0))$$

in terms of h, t .