

1. (15 pts)

Let's think a domain consisting of three binary variables *Toothache*, *Cavity*, and *Catch*, and assume the full joint probability distribution is a table as follows.

	<i>toothache</i>		$\neg$ <i>toothache</i>	
	<i>catch</i>	$\neg$ <i>catch</i>	<i>catch</i>	$\neg$ <i>catch</i>
<i>cavity</i>	0.108	0.012	0.072	0.008
$\neg$ <i>cavity</i>	0.016	0.064	0.144	0.576

Please, compute the following probabilities.

- $P(\text{cavity})=?$
- $P(\text{cavity} \mid \text{toothache})=?$
- $P(\text{cavity} \mid \text{toothache}, \neg\text{catch})=?$
- $P(\text{toothache} \mid \text{cavity}, \neg\text{catch})=?$

2. (25 pts)

Let we have three components required to describe a HMM:  $P(X_0)$  a prior probability distribution over the states at time 0,  $P(X_t|X_{t-1})$  transition model, and  $P(E_t|X_t)$  observation model, where  $X_t$ ,  $E_t$ , and  $e_t$  describe hidden states, observed variables, and actual observations at time  $t$  respectively. Please, derive a recursive formula to estimate the belief state  $P(X_t|E_{1:t})$  at each time  $t$  based on a previous belief state and a new observation  $e_t$ . Make sure to indicate what property or probability rule is applied to at each procedural step to gain the full points.

3. (35pts)

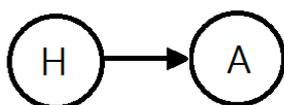


Figure 1.

Figure 1 shows a simple Bayesian network. H and A indicate hidden and observed binary variables, respectively, whose value is either 0 or 1. Throughout the simulations, we observed 100 times when A is 0, and 50 times when A is 1. We would like to use the EM algorithm to estimate probability distributions.

Set  $P(H=1)$ ,  $P(A=1|H=0)$ , and  $P(A=1|H=1)$  to  $\theta_1(t)$ ,  $\theta_2(t)$ , and  $\theta_3(t)$  respectively as parameters at  $t$ -th iteration. We assume the initial parameter values as follows.

$$\theta_1(0) = 0.4, \theta_2(0) = 0.25, \text{ and } \theta_3(0) = 0.54$$

Please, compute the three parameter values until  $t=2$ , that is to say,  $\theta_1(1)$ ,  $\theta_2(1)$ ,  $\theta_3(1)$ ,  $\theta_1(2)$ ,  $\theta_2(2)$ , and  $\theta_3(2)$ , while the EM algorithm operates.

4. (25pts)

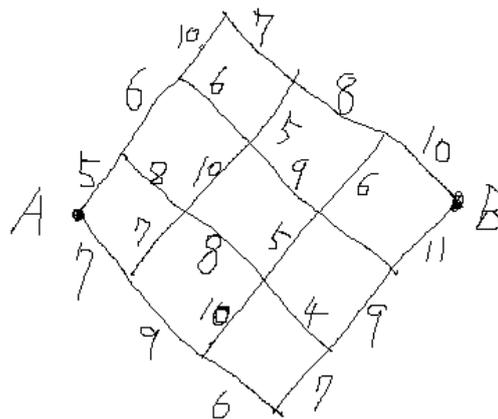


Figure 2.

This problem aims to solve a multistage optimization problem shown in Figure 2. We would find an optimal path from A to B. Each node indicates a state. Left-to-right move is only possible by choosing either upward or downward. A digit near each edge indicates a cost value caused while travelling the edge. Our goal is to find a path from A to B which causes a minimal total cost. A path means a sum of travelled edges. In the answer sheet, draw the figure 2 and indicates the optimal path. Also, explain how to find the optimal path simply.