

1. (10 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. (30 pts)

For each of the following statements, determine whether it is TRUE or FALSE. Make sure you give a sufficient but brief justification of your answers. Notice that each statement is independent of the others, so you cannot use the assumptions of one in the others.

- (1) If random vectors x and y are joint Gaussian and $E[x|y]=E[x]$, then x and y are statistically independent.
- (2) Given two random variables x and y , if $E[x|y]=E[x]$, then for all functions $f(\cdot)$, $E[xf(y)]=E[x]E[f(y)]$ (whenever these expectation are defined).
- (3) Given two random variables x and y , if $E[xf(y)]=E[x]E[f(y)]$ for all functions $f(\cdot)$, then $E[x|y]=E[x]$.

3. (30pts)

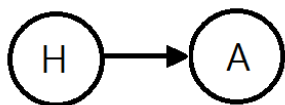


Figure 1.

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 30 times when A is 0, and 70 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.

$$\theta_1(0) = 0.3, \theta_2(0) = 0.2, \text{ and } \theta_3(0) = 0.6$$

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. (30pts)

Consider a linear quadratic regulator problem that minimizes

$$J = \frac{1}{2}x(t_f)^T Q_f x(t_f) + \frac{1}{2} \int_{t_0}^{t_f} x(t)^T Q(t)x(t) + u(t)^T R(t)u(t) dt$$

subject to $\dot{x}(t) = A(t)x(t) + b(t)u(t)$.

The optimal control should be a feedback term.

$$u^*(t) = -K(t)x(t).$$

Using the Hamilton-Jacobi-Bellman (HJB) equation, show that a possible optimal value function is of the form

$$V^*(x(t), t) = \frac{1}{2}x(t)^T K(t)x(t)$$

In the process demonstrate that the following condition must be satisfied.

$$-K = Q(t) + K(t)A(t) + A(t)^T K(t) - K(t)B(t)R(t)^{-1}B(t)^T K(t)$$

with a boundary condition: $K(t_f) = Q_f$.