Problem 1. Conditional independence

Proof:

$$P(A|B) = \frac{P(A,B)}{P(B)}$$

Conditioning on C, we get $P(A|B,C) = \frac{P(A,B|C)}{P(B|C)}$

We already know that P(A,B|C) = P(A|C)P(B|C),

Thus we obtain that
$$P(A|B,C) = \frac{P(A,B|C)}{P(B|C)}$$

$$= \frac{P(A|C)P(B|C)}{P(B|C)}$$

$$= P(A|C) QED$$

Problem 2. Bayesian belief networks

Part a

$$P(E = F, B = F, A = F, J = T, M = F)$$

= $P(E=F)P(B=F)P(A=F|E=F,B=F)P(J=T|A=F)P(M=F|A=F)$

Part b

Substitute the probabilities, we get

$$\begin{split} &P(E=F,\,B=F,\,A=F,\,J=T,\,M=F)\\ &=P(E=F)P(B=F)P(A=F|E=F,B=F)P(J=T|A=F)P(M=F|A=F)\\ &=(0.998)(0.999)(0.999)(0.05)(0.99)\\ &\approx 0.0493 \end{split}$$

Problem 3. Bayesian belief networks

Part a (blind solution)

$$P(B=T, E=T) = \sum_{a \in T, F} \sum_{c \in T, F} \sum_{d \in T, F, X} \sum_{f \in T, F} P(A = a, B = T, C = c, D = d, E = T, F = f)$$

$$= \sum_{a \in T, F} \sum_{c \in T, F} \sum_{d \in T, F, X} \sum_{f \in T, F} P(A = a) P(B = T) P(C = c) P(D = d | A = a, B = T, C = c) P(E = T | C = c) P(F = f | D = d)$$

Computational Cost:

Number of additions = $2^3(3) - 1 = 23$

Number of multiplications: $2^3(3)(5) = 120$

Part b (interleaving sums and products)

$$P(B=T, E=T)$$

$$= \sum_{a \in T, F} \sum_{c \in T, F} \sum_{d \in T, F, X} \sum_{f \in T, F} P(A = a) P(B = T) P(C = c) P(D = d | A = a, B = T, C = c) P(E = T | C = c) P(F = f | D = d)$$

$$= \sum_{a \in T, F} \sum_{c \in T, F} \sum_{d \in T, F, X} P(A = a) P(B = T) P(C = c) P(D = d | A = a, B = T, C = c) P(E = T | C = c) \sum_{f \in T, F} P(F = f | D = d)$$

$$= \sum_{a \in T, F} \sum_{c \in T, F} P(A = a) P(B = T) P(C = c) P(E = T | C = c) \sum_{d \in T, F, X} P(D = d | A = a, B = T, C = c) \sum_{f \in T, F} P(F = f | D = d)$$

$$= \sum_{a \in T.F} P(A = a) \sum_{c \in T.F} P(B = T) P(C = c) P(E = T | C = c) \sum_{d \in T.F.X} P(D = d | A = a, B = T, C = c) \sum_{f \in T.F} P(F = f | D = d)$$

$$= P(B = T) \sum_{a \in T, F} P(A = a) \sum_{c \in T, F} P(C = c) P(E = T | C = c) \sum_{d \in T, F, X} P(D = d | A = a, B = T, C = c) \sum_{f \in T, F} P(F = f | D = d)$$

Computational Cost:

Number of additions = 9

Number of multiplications = 16

Comparing the number of additions and multiplications computational cost in these two methods, we can see that the interleaving one decreases the cost to a great extent thus making the computation more efficient.

Problem 4. Pneumonia diagnosis

Part a

When pneumonia is true:

	True	False
Fever	0.9	0.1
Paleness	0.7	0.3
Cough	0.9	0.1
High WBC	0.8	0.2

When pneumonia is false:

	True	False
Fever	0.6	0.4
Paleness	0.5	0.5
Cough	0.1	0.9
High WBC	0.5	0.5

Part b

$$= \frac{P(\text{Pneumonia} = \text{T, Fever} = \text{T, Paleness} = \text{F, Cough} = \text{T, HighWBC} = \text{F})}{P(\text{Fever} = \text{T, Paleness} = \text{F, Cough} = \text{T, HighWBC} = \text{F})}$$

 $=\frac{P(Pneumonia=T)P(\text{ Fever}=T|\text{Pneumonia}=T)P(\text{Paleness}=F|\text{Pneumonia}=T)P(\text{Cough}=T|\text{Pneumonia}=T)P(\text{HighWBC}=F|\text{Pneumonia}=T)}{\sum_{p\in T,F}P(Pneumonia=p)P(Fever=T|Pneumonia=p)P(Paleness=F|Pneumonia=p)P(Cough=T|Pneumonia=p)P(HighWBC=F|Pneumonia=p)}$

$$=\frac{(0.02)(0.9)(0.3)(0.9)(0.2)}{(0.02)(0.9)(0.3)(0.9)(0.2)+(0.98)(0.6)(0.5)(0.1)(0.5)}$$

 ≈ 0.062

Part c

$$P(Pneumonia = T | Fever = T, Cough = T)$$

$$= \frac{P(\text{Pneumonia} = \text{T, Fever} = \text{T, Cough} = \text{T})}{\sum_{p \in T, F} P(\text{Pneumonia} = \text{p, Fever} = \text{T, Cough} = \text{T})}$$

$$-\frac{{}^{P(Pneumonia=T)P(Fever=T\,|Pneumonia=T)P(Cough=T|Pneumonia=T)}}{\sum_{p\in T,F\,P(Pneumonia=p)P(Fever=T|Pneumonia=p)}P(Fever=T|Pneumonia=p)P(Cough=T|Pneumonia=p)}$$

$$=\frac{(0.02)(0.9)(0.9)}{(0.02)(0.9)(0.9)+(0.98)(0.6)(0.1)}$$

=0.216

Part d

Computed probabilities:

Fever	Paleness	Cough	HighWBC	P(pneumonia symptoms)
1	0	1	0	0.06202143950995407
1	0	1	-1	0.1418563922942207
0	1	-1	0	0.0028490028490028487
1	-1	-1	0	0.012096774193548388
-1	0	1	-1	0.09926470588235294