

HOMework 4

GRAPHICAL MODELS AND LEARNING THEORY

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Problem 1: Graphical Models and MCMC

Problem 1.1: A model for diagnosing illness

(a) Conditional independence structure questions

Let:

- W = weeks traveling
- S = has stomach flu
- E = has ebola
- R = rash
- A = abdominal pain

- i. $vars \perp\!\!\!\perp S = \{\}$ (None)
- ii. $vars \perp\!\!\!\perp S|W, A = \{\}$ (None)
- iii. $vars \perp\!\!\!\perp R|E = \{A, S, W\}$
- iv. $S \perp\!\!\!\perp E|A, R$? False.

(b) Computing quantities from the model:

i.

$$\begin{aligned} p(E|A = t, R = t, W = 0) &= \frac{1}{Z_1} \sum_S p(E, S = s, A = t, R = t, W = 0) \\ &= \frac{1}{Z_1} \sum_S p(W = 0)p(S = s|W = 0)p(E|W = 0)p(A = t|S = s, E)p(R = t|E) \end{aligned}$$

Eliminate S: (See tables 1 and 2)

$$\begin{aligned} &\frac{1}{Z_1} \sum_S p(W = 0)p(S = s|W = 0)p(E|W = 0)p(A = t|S = s, E)p(R = t|E) \\ &= \frac{1}{Z_1} p(W = 0)p(E|W = 0)p(R = t|E) \sum_S p(S = s|W = 0)p(A = t|S = s, E) \end{aligned}$$

*Note $p(W = 0)$ is constant so it's combined with the normalizing constant.

$$\begin{aligned} & \frac{1}{Z_1} \sum_S p(W=0)p(S=s|W=0)p(E|W=0)p(A=t|S=s,E)p(R=t|E) \\ &= \frac{1}{Z_1} p(W=0)p(E|W=0)p(R=t|E) \sum_S p(S=s|W=0)p(A=t|S=s,E) \end{aligned}$$

$$\begin{aligned} & \frac{1}{Z_1} p(W=0)p(E|W=0)p(R=t|E) \sum_S p(S=s|W=0)p(A=t|S=s,E) \\ &= \frac{1}{Z_2} p(E|W=0)p(R=t|E) \sum_S p(S=s|W=0)p(A=t|S=s,E) \end{aligned}$$

Table 1: $p(S=s|W=0)$

| S | $p(S=s W=0)$ |
|---|--------------|
| F | 0.9 |
| T | 0.1 |

Table 2: $p(A=t|S,E)$

| S | E | $p(A=t S,E)$ |
|---|---|--------------|
| F | F | 0.01 |
| F | T | 0.95 |
| T | F | 0.95 |
| T | T | 0.99 |

• **For E=t:**

$$\begin{aligned} & p(E=t|W=0)p(R=t|E=t)* \\ & [p(S=f|W=0)p(A=t|S=f,E=t) + p(S=t|W=0)p(A=t|S=t,E=t)] \\ &= (0.001)(0.9)[(0.9)(0.95) + (0.1)(0.99)] = 0.0008586 \end{aligned}$$

• **For E=f:**

$$\begin{aligned} & p(E=f|W=0)p(R=t|E=f)* \\ & [p(S=f|W=0)p(A=t|S=f,E=f) + p(S=t|W=0)p(A=t|S=t,E=f)] \\ &= (0.999)(0.2)[(0.9)(0.01) + (0.1)(0.95)] = 0.0207792 \end{aligned}$$

• **Final:**

$$p(E=t|A=t, R=t, W=0) = \frac{0.0008586}{0.0008586+0.0207792} = 0.0397$$

$$p(E=f|A=t, R=t, W=0) = \frac{0.0207792}{0.0008586+0.0207792} = 0.9603$$

ii. When $W=2$

• **For E=t:**

$$\begin{aligned} & p(E=t|W=2)p(R=t|E=t) \\ & [p(S=f|W=2)p(A=t|S=f,E=t) + p(S=t|W=2)p(A=t|S=t,E=t)] \\ &= (0.04)(0.9)[(0.8)(0.95) + (0.2)(0.99)] = 0.034488 \end{aligned}$$

• **For E=f:**

$$\begin{aligned}
& p(E = f|W = 2)p(R = t|E = f) \\
& [p(S = f|W = 2)p(A = t|S = f, E = f) + p(S = t|W = 2)p(A = t|S = t, E = f)] \\
& = (0.96)(0.2)[(0.8)(0.01) + (0.2)(0.95)] = 0.038016
\end{aligned}$$

• **Final:**

$$\begin{aligned}
p(E = t|A = t, R = t, W = 2) &= \frac{0.034488}{0.034488+0.038016} = 0.4757 \\
p(E = f|A = t, R = t, W = 2) &= \frac{0.038016}{0.034488+0.038016} = 0.5243
\end{aligned}$$

iii. $p(W|R = t) = \frac{1}{Z} \sum_{A,S,E} p(W)p(S|W)p(E|W)p(A|S, E)p(R = t|E)$

Let:

- $p(W) = \Phi_W$
- $p(S|W) = \Phi_{SW}$
- $p(E|W) = \Phi_{EW}$
- $p(A|S, E) = \Phi_{ASE}$
- $p(R = t|E) = \Phi_{RE}$

a. Eliminate A:

- $P_{initial} = \frac{1}{Z} \sum_{A,S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{ASE} \Phi_{RE} = \frac{1}{Z} \sum_{S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{RE} \sum_A \Phi_{ASE}$
- Tables: Φ_{ASE}
- Message: $\mu_{S,E}$ (See table 3)
- $P_{final} = \frac{1}{Z} \sum_{S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{RE} \mu_{S,E}$
- **Note:** $\mu_{S,E}$ is left out of the equation after this since the marginal probability is 1.

Table 3: $\mu_{S,E}$

| S | E | $\mu_{S,E}$ |
|---|---|-------------|
| F | F | 1 |
| F | T | 1 |
| T | F | 1 |
| T | T | 1 |

b. Eliminate S:

- $P_{initial} = \frac{1}{Z} \sum_{S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{RE} = \frac{1}{Z} \sum_E \Phi_W \Phi_{EW} \Phi_{RE} \sum_S \Phi_{SW}$
- Tables: Φ_{SW}
- Message: μ_{W1} (See table 4)
- $P_{final} = \frac{1}{Z} \sum_E \Phi_W \Phi_{EW} \Phi_{RE} \mu_{W1}$
- **Note:** μ_{W1} is left out of the equation after this since the marginal probability is 1.

Table 4: μ_{W1}

| W | μ_{W1} |
|---|------------|
| F | 1 |
| T | 1 |

c. Eliminate E:

- $P_{initial} = \frac{1}{Z} \sum_E \Phi_W \Phi_{EW} \Phi_{RE} = \frac{1}{Z} \Phi_W \sum_E \Phi_{EW} \Phi_{RE}$
- Tables: $\Phi_{EW} \Phi_{RE}$
- Message: μ_{W2} (See tables 5, 6, 7)
- **Note:** We can evaluate $R = t$ here since R is observed.
- $P_{final} = \frac{1}{Z} \Phi_W \mu_{W2}$

Table 5: Intermediate table (1) - evaluate $p(R = t|E)$

| E | $p(R = t E)$ |
|---|--------------|
| F | 0.2 |
| T | 0.9 |

Table 6: Intermediate table (2) - evaluate $\mu_{W,E}$

| W | E | $\Phi_{EW} * p(R = t E) = \mu_{W,E}$ |
|---|---|--------------------------------------|
| 0 | F | $0.999 * 0.2 = 0.1998$ |
| 0 | T | $0.001 * 0.9 = 0.0009$ |
| 1 | F | $0.980 * 0.2 = 0.196$ |
| 1 | T | $0.020 * 0.9 = 0.018$ |
| 2 | F | $0.960 * 0.2 = 0.192$ |
| 2 | T | $0.040 * 0.9 = 0.036$ |

Table 7: μ_{W2}

| W | μ_{W2} |
|---|----------------------------|
| 0 | $0.1998 + 0.0009 = 0.2007$ |
| 1 | $0.196 + 0.018 = 0.214$ |
| 2 | $0.192 + 0.036 = 0.228$ |

d. **Final:**

- Combine tables: Φ_W and μ_{W2} (See table 8)
- Normalize:

$$p(W = 0|R = t) = 0.19836/0.20934 = 0.947$$

$$p(W = 1|R = t) = 0.00642/0.20934 = 0.031$$

$$p(W = 2|R = t) = 0.00456/0.20934 = 0.022$$

iv. $p(W|R = f) = \frac{1}{Z} \sum_{A,S,E} p(W)p(S|W)p(E|W)p(A|S,E)p(R = f|E)$

Let:

- $p(W) = \Phi_W$
- $p(S|W) = \Phi_{SW}$
- $p(E|W) = \Phi_{EW}$
- $p(A|S,E) = \Phi_{ASE}$

Table 8: Combine tables Φ_W and μ_{W2}

| W | Φ_W | μ_{W2} | $\Phi_W * \mu_{W2}$ |
|---|----------|------------|---------------------|
| 0 | 0.95 | 0.2088 | 0.1936 |
| 1 | 0.03 | 0.214 | 0.00642 |
| 2 | 0.02 | 0.228 | 0.00456 |

- $p(R = t|E) = \Phi_{RE}$

a. Eliminate A:

- $P_{initial} = \frac{1}{Z} \sum_{A,S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{ASE} \Phi_{RE} = \frac{1}{Z} \sum_{S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{RE} \sum_A \Phi_{ASE}$
- Tables: Φ_{ASE}
- Message: $\mu_{S,E}$ (See table 9)
- $P_{final} = \frac{1}{Z} \sum_{S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{RE} \mu_{S,E}$
- **Note:** $\mu_{S,E}$ is left out of the equation after this since the marginal probability is 1.

Table 9: $\mu_{S,E}$

| S | E | $\mu_{S,E}$ |
|---|---|-------------|
| F | F | 1 |
| F | T | 1 |
| T | F | 1 |
| T | T | 1 |

b. Eliminate S:

- $P_{initial} = \frac{1}{Z} \sum_{S,E} \Phi_W \Phi_{SW} \Phi_{EW} \Phi_{RE} = \frac{1}{Z} \sum_E \Phi_W \Phi_{EW} \Phi_{RE} \sum_S \Phi_{SW}$
- Tables: Φ_{SW}
- Message: μ_{W1} (See table 10)
- $P_{final} = \frac{1}{Z} \sum_E \Phi_W \Phi_{EW} \Phi_{RE} \mu_{W1}$
- **Note:** μ_{W1} is left out of the equation after this since the marginal probability is 1.

Table 10: μ_{W1}

| W | μ_{W1} |
|---|------------|
| F | 1 |
| T | 1 |

c. Eliminate E:

- $P_{initial} = \frac{1}{Z} \sum_E \Phi_W \Phi_{EW} \Phi_{RE} = \frac{1}{Z} \Phi_W \sum_E \Phi_{EW} \Phi_{RE}$
- Tables: $\Phi_{EW} \Phi_{RE}$
- Message: μ_{W2} (See tables 11, 12, 13)
- **Note:** We can evaluate $R = t$ here since R is observed.
- $P_{final} = \frac{1}{Z} \Phi_W \mu_{W2}$

d. Final:

Table 11: Intermediate table (1) - evaluate $p(R = f|E)$

| E | $p(R = f E)$ |
|---|--------------|
| F | 0.8 |
| T | 0.1 |

Table 12: Intermediate table (2) - evaluate $\mu_{W,E}$

| W | E | $\Phi_{EW} * p(R = f E) = \mu_{W,E}$ |
|---|---|--------------------------------------|
| 0 | F | $0.999 * 0.8 = 0.7992$ |
| 0 | T | $0.001 * 0.1 = 0.0001$ |
| 1 | F | $0.980 * 0.8 = 0.784$ |
| 1 | T | $0.020 * 0.1 = 0.002$ |
| 2 | F | $0.960 * 0.8 = 0.768$ |
| 2 | T | $0.040 * 0.1 = 0.004$ |

- Combine tables: Φ_W and μ_{W2} (See table 14)

$$p(W = 0|R = f) = 0.75933/0.79835 = 0.951$$

$$p(W = 1|R = f) = 0.02358/0.79835 = 0.030$$

$$p(W = 2|R = f) = 0.01544/0.79835 = 0.019$$

$$E[p(W|R = f)] = (0)(0.951) + (1)(0.030) + (2)(.019) = .068$$

$$(c) \ p(\theta_i|\theta_1, \dots, \theta_{i-1}, \theta_{i+1}, \dots, \theta_n) = \frac{p(\theta_1, \dots, \theta_n)}{\theta_1, \dots, \theta_{i-1}, \theta_{i+1}, \dots, \theta_n} = \frac{1}{Z_1} \prod_{j=1}^n p(\theta_j|parents(\theta_j))$$

We consider: (1) when θ_i is the child, (2) when θ_i is a parent, and (3) the co-parents of θ_i (included in the term $p(\theta_k|parents(\theta_k))$ shown in the equation below).

$$= \frac{1}{Z_2} p(\theta_i|parents(\theta_i)) \prod_{\theta_k \in children(\theta_i)} p(\theta_k|parents(\theta_k))$$

Problem 1.2: Markov chain Monte Carlo

- (d) I tested the convergence rate 10 times. On average, it took 4874 iterations to converge to within 1% of the exact quantities.

Table 13: μ_{W2}

| W | μ_{W2} |
|---|----------------------------|
| 0 | $0.7992 + 0.0001 = 0.7993$ |
| 1 | $0.784 + 0.002 = 0.786$ |
| 2 | $0.768 + 0.004 = 0.772$ |

Table 14: Combine tables Φ_W and μ_{W2}

| W | Φ_W | μ_{W2} | $\Phi_W * \mu_{W2}$ |
|---|----------|------------|---------------------|
| 0 | 0.95 | 0.7993 | 0.75933 |
| 1 | 0.03 | 0.786 | 0.02358 |
| 2 | 0.02 | 0.772 | 0.01544 |