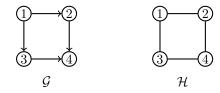


## Probabilistic Graphical Models: Problem Set 5

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1. What is the treewidth of each of the following graphs?



- 2. Let  $\mathcal{G} = ([m], E)$  be a DAG and let  $\Phi = \{f_{X_i}(X_i \mid X_{\operatorname{pa}_{\mathcal{G}}(i)})\}_{i=1}^m$  be its set of factors. What is the relationship between  $\mathcal{G}$  and  $\mathcal{H}_{\Phi}$ ?
- 3. Show that every induced graph  $\mathcal{G}_{\Phi,\prec}$  is chordal.
- 4. Let  $\mathcal{G}$  be a connected, undirected graph. Show that  $\mathcal{G}$  is a tree if and only if its treewidth is  $\omega_{\mathcal{G}} = 1$ .
- 5. (Medical diagnosis again.) Suppose in our medical diagnosis example we assume that the distribution  $\mathbf{X} = [Con, M, F, H, CC, C19, S]^T$  is Markov to the DAG presented in the lecture. We assume  $\mathbf{X}$  abides by the following hierarchy:

```
S \sim \text{Uniform}(\{0, 1, 2, 3\})
                                                     C19 \sim \text{Ber}(\theta_9)
 H|S=0\sim \mathrm{Ber}(\theta_1)
                                    Con|H = 0, CC = 0, C19 = 0 \sim Ber(\theta_{10})
                                                                                               M|CC = 0, C19 = 0 \sim Ber(\theta_{18})
 H|S=1\sim \mathrm{Ber}(\theta_2)
                                    Con|H = 0, CC = 0, C19 = 1 \sim Ber(\theta_{11})
                                                                                               M|CC = 0, C19 = 1 \sim Ber(\theta_{19})
 H|S=2\sim \mathrm{Ber}(\theta_3)
                                    Con|H = 0, CC = 1, C19 = 0 \sim Ber(\theta_{12})
                                                                                               M|CC = 1, C19 = 0 \sim Ber(\theta_{20})
 H|S=3\sim \mathrm{Ber}(\theta_4)
                                    Con|H = 1, CC = 0, C19 = 0 \sim Ber(\theta_{13})
                                                                                               M|CC = 1, C19 = 1 \sim Ber(\theta_{21})
CC|S = 0 \sim Ber(\theta_5)
                                    Con|H = 0, CC = 1, C19 = 1 \sim Ber(\theta_{14})
                                                                                                F|CC = 0, C19 = 0 \sim Ber(\theta_{22})
CC|S = 1 \sim Ber(\theta_6)
                                    Con|H = 1, CC = 0, C19 = 1 \sim Ber(\theta_{15})
                                                                                                F|CC = 0, C19 = 1 \sim Ber(\theta_{23})
CC|S = 2 \sim Ber(\theta_7)
                                    Con|H = 1, CC = 1, C19 = 0 \sim Ber(\theta_{16})
                                                                                                F|CC = 1, C19 = 0 \sim Ber(\theta_{24})
CC|S = 3 \sim Ber(\theta_8)
                                    Con|H = 1, CC = 1, C19 = 1 \sim Ber(\theta_{17})
                                                                                                F|CC = 1, C19 = 1 \sim Ber(\theta_{25})
```

We assume that, based on historical data, we have fit the model parameters as

$$[\theta_1, \dots, \theta_{25}]^T = [0.03, 0.70, 0.48, 0.65, 0.73, \\ 0.16, 0.22, 0.96, 0.81, 0.57, \\ 0.56, 0.36, 0.01, 0.42, 0.32, \\ 0.64, 0.12, 0.53, 0.57, 0.52, \\ 0.11, 0.53, 0.59, 0.54, 0.06]^T.$$

Let  $\mathbf{Y} = [S, Con, M, F]^T$ . We observe for a new patient with the data  $\mathbf{y} = [0, 1, 0, 1]^T$ . Use variable elimination to compute the posterior expectations

$$E[H|Y = y],$$
  $E[CC|Y = y],$  and  $E[C19|Y = y].$