

homework3_Ortu

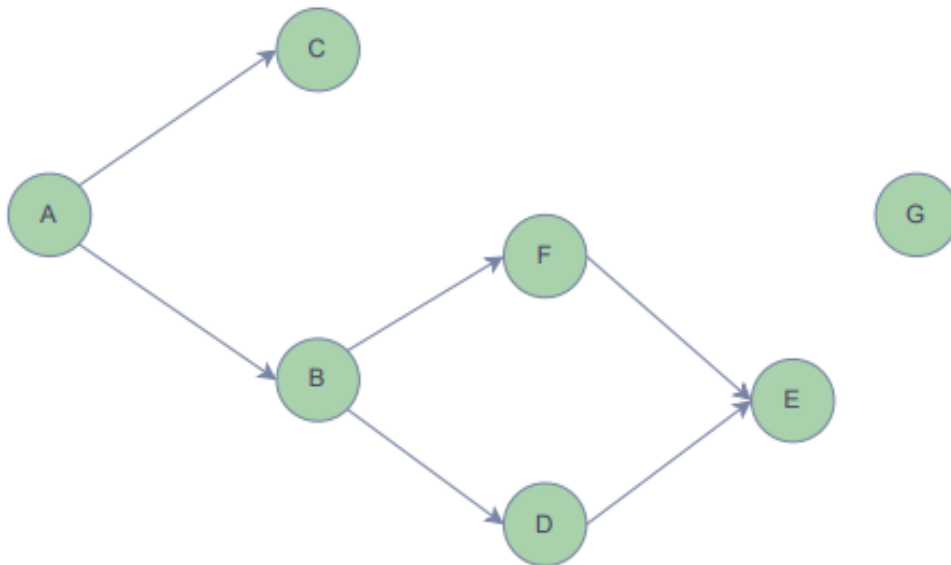
May 2, 2022

1 Homework 3 - Francesco Ortu

1.1 Ex 1

1. Draw the Bayesian Network representing the joint distribution

$$P(A, B, C, D, E, F, G) = P(A)P(B|A)P(F|B)P(C|A)P(D|B)P(E|D, F)P(G).$$



2. Indicate whether the following statements on (conditional) independence are True or False and motivate your answer.
 - a. $A \perp\!\!\!\perp D$: FALSE
D is a descendant of A
 - b. $F \perp\!\!\!\perp D$: FALSE
F and D have the same parents

c. $A \perp\!\!\!\perp B|C$: FALSE

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

d. $A \perp\!\!\!\perp D|B$: TRUE

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

e. $D \perp\!\!\!\perp F|E$: FALSE

head to head

$$P(D, F|E) = \frac{P(D, F, E)}{P(E)} = \frac{P(D, F)P(E|F, D)}{P(E)} \neq P(D|E)P(F|E)$$

f. $B \perp\!\!\!\perp F|E$: FALSE

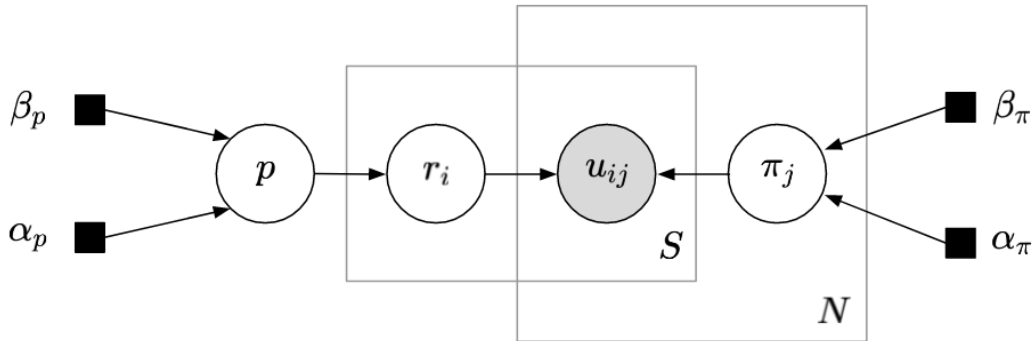
$$P(B, F|E) = \frac{P(B, F, E)}{P(E)} = \frac{P(B)P(F|B)P(E|F)}{P(E)} \neq P(B|E)P(F|E)$$

g. $A \perp\!\!\!\perp D|\{B, F\}$: TRUE

$$\begin{aligned} P(A, D|B, F) &= \frac{P(A, D, B, F)}{P(B, F)} = \frac{P(A)P(B|A)P(F|B)P(D|B)}{P(B)P(F|B)} \\ &= P(A|B)P(D|B) = \frac{P(B, A)P(A)P(F|B)}{P(A)P(B)P(F|B)} \cdot P(D|B) \\ &= \frac{P(B|A)P(A)P(F|B)}{P(B)P(F|B)} \cdot \frac{P(D|B)P(F|B)P(B)}{P(B)P(F|B)} \\ &= \frac{P(A, B, F)}{P(B, F)} \cdot \frac{P(D, B, F)}{P(B, F)} = P(A|B, F)P(D|B, F) \end{aligned}$$

1.2 EX 2

- Write the generative model represented by the following directed graph, knowing that:
 - p and π_j are sampled from Beta distributions;
 - r_i is sampled from a Bernoulli distribution;
 - u_{ij} is sampled from a Bernoulli distribution with parameter $r_i(1 - \pi_j) + (1 - r_i)\pi_j$.



2. Implement the generative model using `pyro`.

Set the hyperparameters to $\alpha_p = 1, \beta_p = 1, \alpha_\pi = 1, \beta_\pi = 5$ and evaluate your model on the observations `data = dist.Bernoulli(0.6).sample((12,6))`.

Remember to use plate notation and to condition on the observed data!

The joint distribution is:

$$P(p, r, \mu, \pi | \alpha_p, \alpha_\pi, \beta_p, \beta_\pi) = P(p | \alpha_p, \beta_p) \prod_{j=1}^N P(\pi_j | \beta_\pi, \alpha_\pi) \prod_{i=1}^S P(r_i | p) \prod_{\substack{i=1 \dots N \\ j=1 \dots S}} P(\mu_{ij} | r_i, p_j)$$

Given the observations, we can compute the joint distribution:

$$P(p, r, \pi | \mu, \alpha_p, \alpha_\pi, \beta_p, \beta_\pi) = \frac{P(p, r, \mu, \pi | \alpha_p, \alpha_\pi, \beta_p, \beta_\pi)}{P(\mu | \alpha_p, \alpha_\pi, \beta_p, \beta_\pi)}$$

```
[ ]: import pyro
import torch
import pyro.distributions as dist
```

```
[ ]: def model(data):
    N = data.size()[0]
    S = data.size()[1]
    #set hyperparameters
    alpha_p = 1
    beta_p = 1
    alpha_pi = 1
    beta_pi = 5

    p = pyro.sample('p', dist.Beta(alpha_p, beta_p))

    r_plate = pyro.plate('r_plate', S, dim = -1)
    pi_plate = pyro.plate("pi_plate", N, dim = -2)

    with pi_plate:
        pi = pyro.sample('pi', dist.Beta(alpha_pi, beta_pi))

    with r_plate:
        r = pyro.sample('r', dist.Bernoulli(probs=p))

    with pi_plate, r_plate:
        u = pyro.sample("u", dist.Bernoulli( r * ( 1 - pi) + ( 1 - r ) * pi),
        ↪obs=data )

    return p, r, pi, u
```

```
[ ]: data = dist.Bernoulli(0.6).sample((12,6))

p,r,pi ,u = model(data)

print(p,r,pi,u)

pyro.render_model(model, model_args=(data,),render_distributions=True) #plates_
↳that are overlapped are rendered as two separate plates
```

```
tensor(0.6268) tensor([1., 1., 0., 0., 1., 1.]) tensor([[0.1436],
[0.0748],
[0.0337],
[0.0497],
[0.0330],
[0.2033],
[0.0222],
[0.0875],
[0.1386],
[0.0146],
[0.0477],
[0.2430]]) tensor([[0., 0., 0., 1., 0., 1.],
[0., 1., 1., 1., 1., 1.],
[0., 1., 1., 0., 1., 1.],
[1., 1., 0., 0., 0., 0.],
[0., 1., 0., 1., 0., 1.],
[0., 0., 1., 1., 1., 0.],
[1., 1., 1., 0., 1., 0.],
[1., 1., 1., 1., 1., 1.],
[1., 0., 1., 0., 0., 1.],
[0., 1., 0., 0., 0., 0.],
[0., 0., 0., 1., 0., 1.],
[1., 0., 0., 1., 0., 1.]])
```

```
[ ]:
```

