CMPSC 448: Machine Learning. Take-home Final. May 7, 12:20pm – May 8, 12:20pm

1. Academic Integrity

- You may not discuss the final with anybody
- You may not post any questions related to the final on the internet. The ONLY exception is the course Piazza page
- You cannot look at anyone else's code. You may not let anyone look at your code.
- You may not import anything except for the fractions library

2. How the test works

- Submit a blank final py file to the gradescope assignment titled "Download Your Final From Here".
- Autograder for that assignment will print out the function definitions and exact questions that you must answer.
- Copy and paste this into your final.py, then fill in your code.
- Submit your final code to the gradescope assignment titled "Submit Your Final Here"

3. Coding Style

- All code (except import statements and constants) should be inside functions.
- Code must have comments and any constants should be stored in a variable defined near the top of your file.

4. Bayesian Networks

4.1. **Numeric Stability.** In this assignment, to avoid rounding issues, you will need to use Fraction datatypes in python. Here is an example of their usage:

```
1 from fractions import Fraction as frac
2 half = frac(1,2)
```

If you <u>choose</u> not to use fraction datatypes, you might get a correct answer marked as incorrect due to rounding issues.

4.2. **Bayesian Network Parameters.** The parameters of a Bayesian network will be passed to your code through a parameter **bn**. It will be a class and its usage is described as follows: suppose the Bayesian network has directed edges (a,b), (a,c), (b,d), (c,d). The parameters of this network are the conditional probabilities P(a), P(b|a), P(c|a), P(d|b,c). The corresponding **bn** variable will be defined something like this:

```
1 import numpy as np
2 from fractions import Fraction as frac
  class BayesNet1:
     def __init__(self, seed, k=10):
         prng = np.random.RandomState(seed)
         self.prob_a = frac(prng.randint(1, 2**k), 2**k) # P(a=1)
          self.prob_b = {(1,): frac(prng.randint(1, 2**k), 2**k), # P(b=1 | a=1)
                    (0,): frac(prng.randint(1, 2**k), 2**k) # P(b=1 | a=0)
          self.prob_c = \{(1,): frac(prng.randint(1, 2**k), 2**k), # P(c=1 | a=1)\}
12
                    (0,): frac(prng.randint(1, 2**k), 2**k) # P(c=1 | a=0)
                   }
          self.prob_d = {
                    (0, 0): : frac(prng.randint(1, 2**k), 2**k), # P(d=1 | b=0, c=0)
                    (0, 1): : frac(prng.randint(1, 2**k), 2**k), # P(d=1 | b=0, c=1)
16
                    (1, 0): : frac(prng.randint(1, 2**k), 2**k), # P(d=1 | b=1, c=0)
17
                    (1, 1): : frac(prng.randint(1, 2**k), 2**k), # P(d=1 | b=1, c=1)
18
19
20
      # the * forces you to name the parameters after the *
      # so bn.a(value=1) is a valid function call
      # but bn.a(1) will return an error that
      # looks like: TypeError: a() takes 1 positional argument but 2 were given
24
      def a(self, *, value): #returns P(a=value)
          if value == 1:
26
               return self.prob_a
          else:
              return 1-self.prob_a
29
30
      def b(self, *, value, a): #returns P(b=value | a)
31
          tmp = self.prob_b[(a,)]
32
          if value == 1:
               return tmp
35
          else:
36
               return 1-tmp
37
38
       def c(self, *, value, a): #returns P(c=value | a)
39
          tmp = self.prob_c[(a,)]
          if value == 1:
               return tmp
42
          else:
43
              return 1-tmp
44
45
      def d(self, *, value, b, c): #returns P(d=value | b, c)
46
          tmp = self.prob_d[(b,c)]
47
          if value == 1:
48
               return tmp
49
          else:
              return 1-tmp
  # example usage
54 \text{ bn} = BayesNet1()
```

```
# get parameter p(a=1)

bn.a(value=1) # must call with arg names, bn.a(1) is incorrect

bn.a(value=0) # get parameter p(a=0)

bn.d(value=0, b=1, c=0) # get parameter P(d=0 | b=1, c=0)

bd.a(value=1, d=2) #throws error because P(a|d) is not a parameter
```

If a parameter is not needed for a particular problem, using it may throw an exception. For example, for the Bayesian network in the code above, P(a,b) can be computed without using the parameter $P(d \mid b,c)$. That is, after you write P(a,b) in terms of the network parameters and simplify, you will notice that P(d|b,c) is not used at all. Thus our implementation of **bn** might not define the function **bn.d(value, b, c)**. This is used to test that you simplified the expression correctly.

4.3. **Types of Questions.** There will be two types of questions:

- (1) Probability calculations: given network parameters, compute probabilities such as $P(A = 1 \mid B = 0, C = 1)$.
- (2) D-separation. The questions might ask you if A is conditionally independent of D given E. You will write a function that returns **result**, **pathverdict** where **result** is the answer (are they conditionally independent?) and **pathverdict** looks like:

```
[
  (('a', 'b', 'd'), False)
  (('a', 'e', 'd'), False)
]
```

each element in the list is a tuple. The first part of the tuple describes the path (e.g., a,e,d is an undirected path from a to d) and the second part of the tuple tells us if that path is blocked or not. Make sure **pathverdict** contains <u>all</u> of the appropriate undirected paths. A path <u>cannot</u> repeat nodes (so a e a e d is not a path). You have to hard-code the appropriate paths, and results. In other words, for d-separations questions, your functions should look like:

```
def question0():
    parthverdict = [
        (('a', 'b', 'd'), False)
        (('a', 'e', 'd'), False)
    ]
    result = False
    return result, pathverdict
```

In this case, this answer indicates that the first path is not blocked and the second path is not blocked.

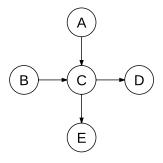


FIGURE 1. Bayesian Network 1: All nodes are upper-case

Question 1 (5 pts). This is a d-separation question using the Bayesian Network in Figure 1. In final.py, fill in the function question1 as instructed so that it returns result, pathverdict.

Question 2 (5 pts). This is a probability question using the Bayesian Network in Figure 1. In final.py, fill in the function question 2 as instructed. Make sure it returns a frac (not float).

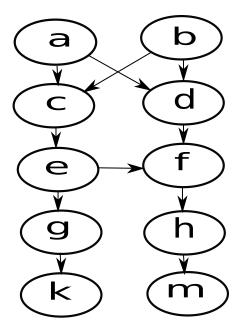


FIGURE 2. Bayesian Network 2: All nodes are lower-case

Question 3 (10 pts). This is a d-separation question using the Bayesian Network in Figure 2. In final.py, fill in the function question3 as instructed so that it returns result, pathverdict.

Question 4 (10 pts). This is a d-separation question using the Bayesian Network in Figure 2. In final.py, fill in the function question4 as instructed so that it returns result, pathverdict.

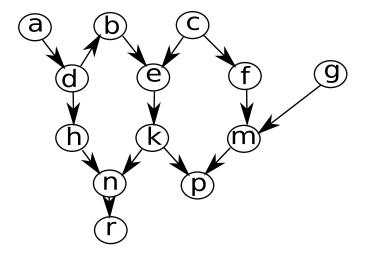


FIGURE 3. Bayesian Network 3, All nodes are lower-case

Question 5 (10 pts). This is a d-separation question using the Bayesian Network in Figure 3. In final.py, fill in the function question5 as instructed so that it returns result, pathverdict.

Question 6 (10 pts). This is a d-separation question using the Bayesian Network in Figure 3. In final.py, fill in the function question6 as instructed so that it returns result, pathverdict.

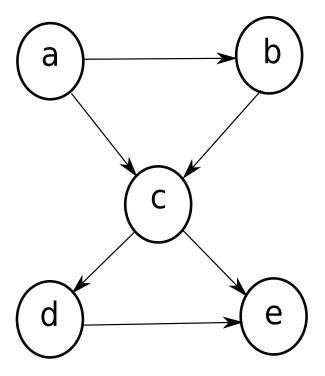


FIGURE 4. Bayesian Network 4, All nodes are lower-case

Question 7 (10 pts). This is a probability question using the Bayesian Network in Figure 4. In final.py, fill in the function question 7 as instructed. Make sure it returns a frac (not float) and does not use any unnecessary network parameters.

Question 8 (10 pts). This is a probability question using the Bayesian Network in Figure 4. In final.py, fill in the function question8 as instructed. Make sure it returns a frac (not float) and does not use any unnecessary network parameters.

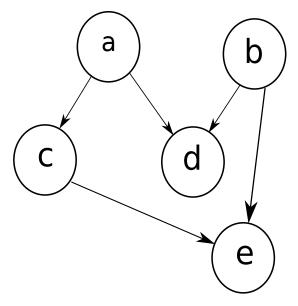


FIGURE 5. Bayesian Network 5, All nodes are lower-case

Question 9 (10 pts). This is a probability question using the Bayesian Network in Figure 5. In final.py, fill in the function question 9 as instructed. Make sure it returns a frac (not float) and does not use any unnecessary network parameters.

Question 10 (10 pts). This is a probability question using the Bayesian Network in Figure 5. In final.py, fill in the function question10 as instructed. Make sure it returns a frac (not float) and does not use any unnecessary network parameters.