



KIET Group of Institutions, Delhi-NCR, Gzb

Department of Computer Applications (NBA Accredited)

(An ISO – 9001: 2015 Certified & 'A+' Grade accredited Institution by NAAC)



Experiment-No.8

Objective: Write a program to demonstrate the working of Bayesian network for the following graph:

Calculate the probability of a burglary if John and Mary calls (0: True, 1: False)

Calculate the probability of alarm starting if there is a burglary and an earthquake (0: True, 1: False)

Calculate the probability of alarm starting if there is a burglary and an earthquake (0: True, 1: False)

Scheduled Date:	Compiled Date:	Submitted Date:
27 Sep 2023	28 Sep 2023	30- Sep 2023

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Python code:

```
pip install pgmpy

import pgmpy. inference
import pgmpy. models
import networkx as nx
import pylab as plt

model = pgmpy. models . BayesianNetwork ([ ( ' Burglary ' , 'Alarm ' ) , ( ' Ea
rthquake ' , 'Alarm ' ) ,
                                     ( 'Alarm ' , ' JohnCalls ' ) , ( 'Al
arm ' , 'MaryCalls ' ) ] )

# Define conditional probability distributions (CPD)
# Probability of burglary (True , False )
```



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```
cpd burglary = pgmpy. factors . discrete .TabularCPD(" Burglary " , 2 , [[0.01] , [0.999]])

# Probability of earthquake (True , False )

cpd earthquake = pgmpy. factors . discrete .TabularCPD("Earthquake" , 2 , [[0.002] , [0.998]])

# Probability of alarm going of (True , False ) given a burglary and/or earthquake

cpd alarm = pgmpy. factors . discrete .TabularCPD( 'Alarm ' , 2 , [[0.95 , 0.94 , 0.29 , 0.001] , [0.05 , 0.06 , 0.71 , 0.999]] ,

                                                                                               evidence =['Burglary ' , 'Earthquake '] , evidence card =[2, 2])

# Probability that John calls (True , False ) given that the alarm has sounded

cpd john = pgmpy. factors . discrete .TabularCPD( ' JohnCalls ' , 2 , [[0.90 , 0.05] , [0.10 , 0.95]] , evidence =['Alarm '] , evidence card =[2])

# Probability that Mary calls (True , False ) given that the alarm has sounded

cpd mary = pgmpy. factors . discrete .TabularCPD( ' MaryCalls ' , 2 , [[0.70 , 0.01] , [0.30 , 0.99]] , evidence =['Alarm '] , evidence card =[2])

# Add CPDs to the network structure

model . add cpds ( cpd burglary , cpd earthquake , cpd alarm , cpd john , cpd mary)

# Check if the model is valid , throw an exception otherwise

model . check model ()

# Print probability distributions

print ( ' Probability distribution , P( Burglary ) ')

print ( cpd burglary )

print ()

print ( ' Probability distribution , P(Earthquake ) ')

print ( cpd earthquake )
```



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```
print ()
print ( ' Joint probability distribution , P(Alarm | Burglary , Earthquake )
')
print ( cpd alarm )
print ()
print ( ' Joint probability distribution , P( JohnCalls | Alarm) ' )
print ( cpd john )
print ()
print ( ' Joint probability distribution , P(MaryCalls | Alarm) ' )
print (cpd mary) print () # Plot the model
nx. draw(model , with labels=True) plt . savefig ( ' alarm1 .png ' ) plt . clo
se ()

# Perform variable elimination for inference
# Variable elimination (VE) is a an exact inference algorithm in bayesian net
works
infer = pgmpy. inference . VariableElimination (model)

# Calculate the probability of a burglary i f John and Mary calls (0: True ,
1: False )
posterior probability = infer . query ([ ' Burglary ' ] , evidence={'JohnCalls
' : 0 , 'MaryCalls ' : 0})

# Print posterior probability
print ( ' Posterior probability of Burglary i f JohnCalls (True) and MaryCall
s(True) ' )

print ( posterior probability )

print ()

# Calculate the probability of alarm starting i f there is a burglary and an
earthquake (0: True , 1: False )
posterior probability = infer . query ([ ' Alarm ' ] , evidence= { ' Burglary '
: 0 , 'Earthquake ' : 0})
```



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```
# Print posterior probability

print ( ' Posterior probability of Alarm sounding i f Burglary (True) and Ear
thquake(True) ' )

print ( posterior probability )

print ()
```

Output:

Probability distribution, P(Burglary)

```
+-----+
| Burglary(0) | 0.001 |
+-----+
| Burglary(1) | 0.999 |
+-----+
```

Probability distribution, P(Earthquake)

```
+-----+
| Earthquake(0) | 0.002 |
+-----+
| Earthquake(1) | 0.998 |
+-----+
```

Joint probability distribution, P(Alarm | Burglary, Earthquake)

```
+-----+-----+-----+-----+-----+
| Burglary | Burglary(0) | Burglary(0) | Burglary(1) | Burglary(1) |
+-----+-----+-----+-----+-----+
| Earthquake | Earthquake(0) | Earthquake(1) | Earthquake(0) | Earthquake(1) |
+-----+-----+-----+-----+-----+
| Alarm(0) | 0.95 | 0.94 | 0.29 | 0.001 |
+-----+-----+-----+-----+-----+
| Alarm(1) | 0.05 | 0.06 | 0.71 | 0.999 |
+-----+-----+-----+-----+-----+
```

Joint probability distribution, P(JohnCalls | Alarm)

```
+-----+-----+
| Alarm | Alarm(0) | Alarm(1) |
+-----+-----+
| JohnCalls(0) | 0.9 | 0.05 |
+-----+-----+
| JohnCalls(1) | 0.1 | 0.95 |
+-----+-----+
```

Joint probability distribution, P(MaryCalls | Alarm)

```
+-----+-----+
| Alarm | Alarm(0) | Alarm(1) |
+-----+-----+
| MaryCalls(0) | 0.7 | 0.01 |
+-----+-----+
| MaryCalls(1) | 0.3 | 0.99 |
+-----+-----+
```

Posterior probability of Burglary if JohnCalls(True) and MaryCalls(True)

```
+-----+-----+
| Burglary | phi(Burglary) |
+-----+-----+
| Burglary(0) | 0.2842 |
+-----+-----+
| Burglary(1) | 0.7158 |
+-----+-----+
```

Finding Elimination Order: 0/0 [00:00<?, 70%]

0/0 [00:00<?, 70%]

Posterior probability of Alarm sounding if Burglary(True) and Earthquake(True)

```
+-----+-----+
| Alarm | phi(Alarm) |
+-----+-----+
| Alarm(0) | 0.9500 |
+-----+-----+
| Alarm(1) | 0.0500 |
+-----+-----+
```