

ML REPORT

Name:- M S Sathwick Kiran

USN:- 1BM18CS050

Week 1:

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
import numpy as np # linear algebra import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# Input data files are available in the read-only "../input/" directory #
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory
import os for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session

/kaggle/input/datasetcsv/data.csv
```

```
data =
pd.read_csv("/kaggle/input/datasetcsv/data.csv")
print("The entered data is \n") print(data,"\n") d =
np.array(data)[:, :-1]
print("\n The attributes are:
\n",d) target = np.array(data)[:, -1]
print("\n The target is:
",target) def training(c,t):
for i, val in enumerate(t):
if val == "Yes":
    specific_hypothesis = c[i].copy()
break
```

```

        for i, val in enumerate(c):            if t[i] ==
"Yes":                for x in
range(len(specific_hypothesis)):In                if
val[x] != specific_hypothesis[x]:
                    specific_hypothesis[x] = '?'
else:                pass            return
specific_hypothesis
print("\n The final hypothesis is:",training(d,target))

```

Output:

The entered data is

	Weather	Temperature	Humidity	Goes
0	Sunny	Warm	Mild	Yes
1	Rainy	Cold	Mild	No
2	Sunny	Moderate	Nomal	Yes
3	Sunny	Cold	High	Yes

The attributes are:

```

[['Sunny ' 'Warm ' 'Mild']
['Rainy' 'Cold' 'Mild']
['Sunny ' 'Moderate' 'Nomal']
['Sunny ' 'Cold' 'High ']]

```

The target is: ['Yes' 'No' 'Yes' 'Yes']

The final hypothesis is: ['Sunny ' '?' '?']

Week 2:

This Python 3 environment comes with many helpful analytics libraries installed

It is defined by the kaggle/python Docker image: <https://github.com/kaggle/docker-python>

For example, here's several helpful packages to load

```

import numpy as np # linear algebra import pandas as pd # data
processing, CSV file I/O (e.g. pd.read_csv)

```

*# Input data files are available in the read-only "../input/" directory #
For example, running this (by clicking run or pressing Shift+Enter) will
list all files under the input directory*

```

import
os for
dirname,

```

```

_,
filenames
in
os.walk('
/kaggle/i
nput'):
for
filename
in
filenames
:
    print(os.path.join(dirname, filename))

```

You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"
You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session

```

data =
pd.read_csv("/kaggle/input/candidatecsv/candidate.csv")
print("Entered data is") print(data) concepts =
np.array(data)[:,-1] print("\n The attributes are: \n",d)
target = np.array(data)[:,-1] print("\n The target is:
",target)

```

```

Entered data is      sky airtemp humidity      wind water
forecast enjoysport 0  sunny      warm   normal   strong  warm
same                yes
1  sunny      warm      high  strong   warm      same      yes
2  rainy      cold      high  strong   warm      change      no 3  sunny
   warm      high  strong  cool    change      yes

```

The attributes are:

```

[['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'high' 'strong' 'warm' 'same']
['rainy' 'cold' 'high' 'strong' 'warm' 'change']
['sunny' 'warm' 'high' 'strong' 'cool' 'change']]

```

The target is: ['yes' 'yes' 'no' 'yes']

```

#training function to implement candidate_elimination algorithm
def learn(concepts, target):
    specific_h = concepts[0].copy()
    print("\n Initialization of specific_h and general_h")
    print(specific_h)
    general_h = [["?" for i in range(len(specific_h))] for i in range(len(specific_h))]
    print(general_h)
    for i, h in enumerate(concepts):
        if target[i] == "yes":
            for x in range(len(specific_h)):
                if h[x] != specific_h[x]:
                    specific_h[x] = '?'
                    general_h[x][x] = '?'
            print(specific_h)
            print(general_h)
        if target[i] == "no":
            for x in range(len(specific_h)):
                if h[x] != specific_h[x]:
                    general_h[x][x] = specific_h[x]
    else:
        general_h[x][x] = '?'
    print("\n Steps of Candidate Elimination Algorithm", i+1)
    print(specific_h)
    print(general_h)
    indices = [i for i, val in enumerate(general_h) if val == ['?', '?', '?', '?', '?', '?']]
    for i in indices:
        general_h.remove(['?', '?', '?', '?', '?', '?'])
    return specific_h, general_h
s_final, g_final = learn(concepts, target)

#obtaining the final hypothesis
print("\nFinal Specific_h:", s_final, sep="\n")
print("\nFinal General_h:", g_final, sep="\n")

```

Output:

```

Initialization of specific_h and general_h
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']

Steps of Candidate Elimination Algorithm 1
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']

```

Steps of Candidate Elimination Algorithm 2

Steps of Candidate Elimination Algorithm 3

Steps of Candidate Elimination Algorithm 4

Final Specific h:

Final General h:

Week 3:

```
import  
math
```

```
import csv  
def load_csv(filename):  
    lines=csv.reader(open(filename,"r"))  
    dataset = list(lines)  
    headers = dataset.pop(0)  
    return dataset,headers
```

```
class Node:  
    def  __init__ (self,attribute):  
        self.attribute=attribute  
        self.children=[]  
        self.answer=""
```

```
def subtables(data,col,delete):  
    dic={}  
    coldata=[row[col] for row in data]
```

```

        attr=list(set(coldata))

        counts=[0]*len(attr)
r=len(data)        c=len(data[0])        for x
in range(len(attr)):        for y in
range(r):        if
data[y][col]==attr[x]:

                                counts[x]+=1

                                for x in range(len(attr)):
dic[attr[x]]=[[0 for i in range(c)] for j in range(counts[x])]
pos=0        for y in range(r):
    if data[y][col]==attr[x]:
        if delete:
            del data[y][col]
            dic[attr[x]][pos]=data[y]
            pos+=1        return
attr,dic

def entropy(S):
    attr=list(set(S))
    if len(attr)==1:
        return 0

    counts=[0,0]
for i in range(2):

                                counts[i]=sum([1 for x in S
if attr[i]==x])/(len(S)*1.0)

    sums=0        for
cnt in counts:
        sums+=-1*cnt*math.log(cnt,2)
return sums

def compute_gain(data,col):
    attr,dic =
subtables(data,col,delete=False)

    total_size=len(data)
entropies=[0]*len(attr)        ratio=[0]*len(attr)

    total_entropy=entropy([row[-1] for row in data])        for x
in range(len(attr)):

```

```
ratio[x]=len(dic[attr[x]])/(total_size*1.0)
```



```

        entropies[x]=entropy([row[-1] for row in dic[attr[x]]])
total_entropy-=ratio[x]*entropies[x]        return total_entropy

```

```

def build_tree(data,features):
    lastcol=[row[-1] for row in data]
    if(len(set(lastcol))==1:
node=Node("")
node.answer=lastcol[0]        return
node

```

```

        n=len(data[0])-1
        gains=[0]*n
for col in range(n):
        gains[col]=compute_gain(data,col)
split=gains.index(max(gains))
node=Node(features[split])        fea =
features[:split]+features[split+1:]

```

```

        attr,dic=subtables(data,split,delete=True)

        for x in range(len(attr)):
            child=build_tree(dic[attr[x]],fea)
node.children.append((attr[x],child))        return node

```

```

def print_tree(node,level):
if node.answer!="":
        print("    "*level,node.answer)
return

```

```

        print("    "*level,node.attribute)        for
value,n in node.children:        print("
"*(level+1),value)
print_tree(n,level+2)

```

```

def classify(node,x_test,features):
    if node.answer!="":
print(node.answer)        return

```

```

        pos=features.index(node.attribute)
        for value, n in node.children:
if x_test[pos]==value:
            classify(n,x_test,features)

'''Main program'''    dataset,features=load_csv("id3.csv")
node1=build_tree(dataset,features)

print("The decision tree for the dataset using ID3 algorithm is")
print_tree(node1,0)    testdata,features=load_csv("id3_test.csv")

for xtest in testdata:
    print("The test instance:",xtest)
print("The label for test instance:")
classify(node1,xtest,features)

```

Output:

```

bmsce@bmsce-Precision-T1700:~/Documents/LAB - 3 - DECISION TREE$ python ml3.py
The decision tree for the dataset using ID3 algorithm is
(' ', 'Outlook')
(' ', 'overcast')
(' ', 'yes')
(' ', 'sunny')
(' ', 'Humidity')
(' ', 'high')
(' ', 'no')
(' ', 'normal')
(' ', 'yes')
(' ', 'rain')
(' ', 'Wind')
(' ', 'strong')
(' ', 'no')
(' ', 'weak')
(' ', 'yes')
('The test instance:', ['rain', 'cool', 'normal', 'strong'])
The label for test instance:
no
('The test instance:', ['sunny', 'mild', 'normal', 'strong'])
The label for test instance:
yes

```

Week 4:

```
import pandas as pd
```

```
data = pd.read_csv('PlayTennis.csv') data.head()
```

```
y = list(data['PlayTennis'].values) X
```

```
= data.iloc[:,1:].values
```

```
print(f'Target Values: {y}')
```

```
print(f'Features: \n{X}')
```

```
Target Values: ['No', 'No', 'Yes', 'Yes', 'Yes', 'No', 'Yes', 'No', 'Yes', 'Yes', 'Yes', 'Yes', 'Yes', 'No'] Features:
```

```
[['Sunny' 'Hot' 'High' 'Weak']  
 ['Sunny' 'Hot' 'High' 'Strong']  
 ['Overcast' 'Hot' 'High' 'Weak']  
 ['Rain' 'Mild' 'High' 'Weak']  
 ['Rain' 'Cool' 'Normal' 'Weak']  
 ['Rain' 'Cool' 'Normal' 'Strong']  
 ['Overcast' 'Cool' 'Normal' 'Strong']  
 ['Sunny' 'Mild' 'High' 'Weak']  
 ['Sunny' 'Cool' 'Normal' 'Weak']  
 ['Rain' 'Mild' 'Normal' 'Weak']  
 ['Sunny' 'Mild' 'Normal' 'Strong']  
 ['Overcast' 'Mild' 'High' 'Strong']  
 ['Overcast' 'Hot' 'Normal' 'Weak']  
 ['Rain' 'Mild' 'High' 'Strong']]
```

```
y_train = y[:8]
```

```
X_train = X[:8]
```

```
X_val = X[8:]
```

```
print(f"Number of instances in training set: {len(X_train)}")
```

```
print(f"Number of instances in testing set: {len(X_val)}")
```

```
Number of instances in training set: 8
```

```
Number of instances in testing set: 6
```

```
class NaiveBayesClassifier:
```

```
    y_val = y[8:]
```

```
    def __init__(self, X, y):
```

```
        self.X, self.y = X, y
```

```
        self.N = len(self.X)
```



```

        self.dim = len(self.X[0])

        self.attrs = [[] for _ in range(self.dim)]

        self.output_dom = {}

        self.data = []
        for i in range(len(self.X)):
            for j in range(self.dim):
                if not self.X[i][j] in self.attrs[j]:
                    self.attrs[j].append(self.X[i][j])
                    if not self.y[i] in self.output_dom.keys():
                        self.output_dom[self.y[i]] = 1
            else:
                self.output_dom[self.y[i]] += 1

        self.data.append([self.X[i], self.y[i]])
    def classify(self, entry):
        solve = None
        max_arg = -1
        for y in self.output_dom.keys():
            prob = self.output_dom[y]/self.N
            for i in range(self.dim):
                cases = [x for x in self.data if x[0][i] == entry[i]
                        and x[1] == y]
                n = len(cases)
                prob *= n/self.N

            if prob > max_arg:
                max_arg = prob
                solve = y

        return solve

nbc = NaiveBayesClassifier(X_train, y_train)

total_cases = len(y_val)
good = 0
bad = 0
predictions = []
for i in range(total_cases):
    predict = nbc.classify(X_val[i])
    predictions.append(predict)
    if y_val[i] == predict:
        good += 1
    else:
        bad += 1

```

```
print('Predicted values:', predictions) print('Actual values:', y_val)
print() print('Total number of testing instances in the dataset:',
total_cases) print('Number of correct predictions:', good)
print('Number of wrong predictions:', bad) print()
print('Accuracy of Bayes Classifier:', good/total_cases)
```

Output:

Predicted values: ['No', 'Yes', 'No', 'Yes', 'Yes', 'No']
Actual values: ['Yes', 'Yes', 'Yes', 'Yes', 'Yes', 'No']

Total number of testing instances in the dataset: 6
Number of correct predictions: 4
Number of wrong predictions: 2

Accuracy of Bayes Classifier: 0.6666666666666666

Week 5:

This Python 3 environment comes with many helpful analytics libraries installed # It is defined by the kaggle/python Docker image: <https://github.com/kaggle/docker-python> # For example, here's several helpful packages to load **import numpy as np**

linear algebra

import pandas as pd import pgmpy as pgmpy from pgmpy.estimators

import MaximumLikelihoodEstimator **from pgmpy.models import**

BayesianModel **from pgmpy.inference import** VariableElimination

import os for dirname, _, filenames **in** os.walk('/kaggle/input'):

for filename **in** filenames: **print**(os.path.join(dirname, filename)) *# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All" # You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session*

#read Cleveland Heart Disease data

heartDisease = pd.read_csv("/kaggle/input/bayesiannetwork/heart.csv")

heartDisease = heartDisease.replace('?', np.nan)

#display the data

print('Sample instances from the dataset are given below')

print(heartDisease.head())

```

#display the Attributes names and
datatypes print('\n Attributes and
datatypes') print(heartDisease.dtypes)
#Creat Model- Bayesian Network
model = BayesianModel([('age', 'heartdisease'), ('sex', 'heartdisease'), ('
exang', 'heartdisease'), ('cp', 'heartdisease'), ('heartdisease', 'restecg')
, ('heartdisease', 'chol')])
#Learning CPDs using Maximum Likelihood Estimators print('\n
Learning CPD using Maximum likelihood estimators')
model.fit(heartDisease, estimator=MaximumLikelihoodEstimator)
# Inferencing with Bayesian Network
print('\n Inferencing with Bayesian Network:')
HeartDiseasetest_infer = VariableElimination(model) #computing
the Probability of HeartDisease given restecg
print('\n 1.Probability of HeartDisease given evidence= restecg :1')
q1=HeartDiseasetest_infer.query(variables=['heartdisease'], evidence={'r
estecg':1}) print(q1)
#computing the Probability of HeartDisease given cp
print('\n 2.Probability of HeartDisease given evidence= cp:2 ')
q2=HeartDiseasetest_infer.query(variables=['heartdisease'], evidence={'c
p':2}) print(q2)

```

Output:

Sample instances from the dataset are given below

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak
0	63	1	1	145	233	1	2	150	0	2.3
1	67	1	4	160	286	0	2	108	1	1.5
2	67	1	4	120	229	0	2	129	1	2.6
3	37	1	3	130	250	0	0	187	0	3.5
4	41	0	2	130	204	0	2	172	0	1.4

	ca	thal
heartdisease	0	0
6	0	
1	3	2
2	7	1
3	3	0
4	3	0

```

Attributes and datatypes
age                int64
sex                int64 cp
int64 trestbps
int64 chol
int64 fbs
int64 restecg
int64 thalach

```



```
int64 exang
int64 oldpeak
float64 slope
int64 ca
object thal
object heartdisease
int64 dtype: object
```

Learning CPD using Maximum likelihood estimators

```
Finding Elimination Order: : 0%|          | 0/5 [00:00<?, ?it/s]
0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: age: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: chol: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: cp: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: sex: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: exang: 100%|██████████| 5/5 [00:00<00:00, 189.65it/s]
Finding Elimination Order: : 100%|██████████| 5/5 [00:00<00:00, 132.81it/s]
Finding Elimination Order: : 0%|          | 0/5 [00:00<?, ?it/s]
0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: age: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: chol: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: restecg: 0%|          | 0/5 [00:00<?, ?it/s]
Eliminating: sex: 0%|          | 0/5 [00:00<?, ?it/s]      Eliminating:
exang: 100%|██████████| 5/5 [00:00<00:00, 230.00it/s]
```

Inferencing with Bayesian Network:

1.Probability of HeartDisease given evidence= restecg :1

```
+-----+-----+
| heartdisease | phi(heartdisease) |
+=====+=====+
| heartdisease(0) | 0.1012 |
+-----+-----+
| heartdisease(1) | 0.0000 |
+-----+-----+
| heartdisease(2) | 0.2392 |
+-----+-----+
| heartdisease(3) | 0.2015 |
+-----+-----+
| heartdisease(4) | 0.4581 |
+-----+-----+
```

2.Probability of HeartDisease given evidence= cp:2

```
+-----+-----+
| heartdisease | phi(heartdisease) |
+=====+=====+
| heartdisease(0) | 0.3610 |
+-----+-----+
| heartdisease(1) | 0.2159 |
+-----+-----+
| heartdisease(2) | 0.1373 |
+-----+-----+
| heartdisease(3) | 0.1537 |
+-----+-----+
```

```
| heartdisease(4) | 0.1321 +-----+-----+
-----+
```

WEEK 6:

```
from pgmpy.models import BayesianModel from
pgmpy.factors.discrete import TabularCPD from
pgmpy.inference import VariableElimination

cancer_model = BayesianModel([('Pollution', 'Cancer'),
                              ('Smoker', 'Cancer'),
                              ('Cancer', 'Xray'),
                              ('Cancer', 'Dyspnoea')]) print('Bayesian network
nodes are:') print("\t", cancer_model.nodes())
print('Bayesian network edges are:')
print("\t", cancer_model.edges())

cpd_poll = TabularCPD(variable='Pollution', variable_card=2, values=[[0.9
], [0.1]])
cpd_smoke = TabularCPD(variable='Smoker', variable_card=2, values=[[0.3],
[0.7]])
cpd_cancer = TabularCPD(variable='Cancer', variable_card=2, values=[[0.03
, 0.05, 0.001, 0.02],
                                                                    [0.97,
0.95, 0.999, 0.98]],
                        evidence=['Smoker', 'Pollution'],
                        evidence_card=[2, 2])
cpd_xray = TabularCPD(variable='Xray', variable_card=2, values=[[0.9, 0.2]
, [0.1, 0.8]],
                      evidence=['Cancer'], evidence_card=[2])
cpd_dysp = TabularCPD(variable='Dyspnoea', variable_card=2, values=[[0.65
, 0.3], [0.35, 0.7]],
                      evidence=['Cancer'], evidence_card=[2])

Bayesian network nodes are:
['Pollution', 'Cancer', 'Smoker', 'Xray', 'Dyspnoea'] Bayesian
network edges are:
[('Pollution', 'Cancer'), ('Cancer', 'Xray'), ('Cancer', 'Dyspn oea'),
('Smoker', 'Cancer')]

cancer_model.add_cpds(cpd_poll, cpd_smoke, cpd_cancer, cpd_xray, cpd_dysp)
print('Model generated by adding cpts(cpds)') print('Checking
correctness of model:', end='') print(cancer_model.check_model())

Model generated by adding cpts(cpds)
Checking correctness of model:True

print('All local dependencies are as follows')
cancer_model.get_independencies()
All local dependencies are as follows
```

Out[10]:

```
(Pollution ⊥ Smoker)
(Pollution ⊥ Dyspnoea, Xray | Cancer)
(Pollution ⊥ Xray | Dyspnoea, Cancer)
(Pollution ⊥ Dyspnoea | Cancer, Xray)
(Pollution ⊥ Dyspnoea, Xray | Cancer, Smoker)
(Pollution ⊥ Xray | Dyspnoea, Cancer, Smoker)
(Pollution ⊥ Dyspnoea | Cancer, Xray, Smoker)
(Smoker ⊥ Pollution)
(Smoker ⊥ Dyspnoea, Xray | Cancer)
(Smoker ⊥ Xray | Dyspnoea, Cancer)
(Smoker ⊥ Dyspnoea, Xray | Pollution, Cancer)
(Smoker ⊥ Dyspnoea | Cancer, Xray)
(Smoker ⊥ Xray | Dyspnoea, Pollution, Cancer)
(Smoker ⊥ Dyspnoea | Pollution, Cancer, Xray)
(Xray ⊥ Dyspnoea, Pollution, Smoker | Cancer)
(Xray ⊥ Pollution, Smoker | Dyspnoea, Cancer)
(Xray ⊥ Dyspnoea, Smoker | Pollution, Cancer)
(Xray ⊥ Dyspnoea, Pollution | Cancer, Smoker)
(Xray ⊥ Smoker | Dyspnoea, Pollution, Cancer)
(Xray ⊥ Pollution | Dyspnoea, Cancer, Smoker)
(Xray ⊥ Dyspnoea | Pollution, Cancer, Smoker)
(Dyspnoea ⊥ Pollution, Xray, Smoker | Cancer)
(Dyspnoea ⊥ Xray, Smoker | Pollution, Cancer)
(Dyspnoea ⊥ Pollution, Smoker | Cancer, Xray)
(Dyspnoea ⊥ Pollution, Xray | Cancer, Smoker)
(Dyspnoea ⊥ Smoker | Pollution, Cancer, Xray)
(Dyspnoea ⊥ Xray | Pollution, Cancer, Smoker)
(Dyspnoea ⊥ Pollution | Cancer, Xray, Smoker) print('Displaying
CPDs')
print(cancer_model.get_cpds('Pollution'))
print(cancer_model.get_cpds('Smoker'))
print(cancer_model.get_cpds('Cancer'))
print(cancer_model.get_cpds('Xray'))
print(cancer_model.get_cpds('Dyspnoea'))
Displaying CPDs
+-----+-----+
| Pollution(0) | 0.9 |
+-----+-----+
| Pollution(1) | 0.1 |
+-----+-----+
+-----+-----+
| Smoker(0) | 0.3 |
+-----+-----+
| Smoker(1) | 0.7 |
+-----+-----+
+-----+-----+-----+-----+-----+-----+
| Smoker      | Smoker(0)      | Smoker(0)      | Smoker(1)      | Smoker(1)
|
+-----+-----+-----+-----+-----+-----+
```

	Pollution	Pollution(0)	Pollution(1)	Pollution(0)	Pollution(1)
Cancer(0)	0.03	0.05	0.001	0.02	
Cancer(1)	0.97	0.95	0.999	0.98	

	Cancer	Cancer(0)	Cancer(1)
Xray(0)	0.9	0.2	
Xray(1)	0.1	0.8	

	Cancer	Cancer(0)	Cancer(1)
Dyspnoea(0)	0.65	0.3	
Dyspnoea(1)	0.35	0.7	

```

cancer_infer=VariableElimination(cancer_model)
print('\n Inferencing with bayesian network') print("\n
Probability of Cancer given smoker")
q=cancer_infer.query(variables=['Cancer'],evidence={'Smoker':1})
print(q)

print("\n Probability of Cancer given smoker,pollution")
q=cancer_infer.query(variables=['Cancer'],evidence={'Smoker':1,'Polluti
on':1}) print(q)

```

Output:

```

Finding Elimination Order: : 0%|          | 0/3 [00:00<?, ?it/s]
0%|          | 0/3 [00:00<?, ?it/s]
Eliminating: Dyspnoea: 0%|          | 0/3 [00:00<?, ?it/s]
Eliminating: Pollution: 0%|          | 0/3 [00:00<?, ?it/s]
Eliminating: Xray: 100%|██████████| 3/3 [00:00<00:00, 359.52it/s]

0%|          | 0/2 [00:00<?, ?it/s]
Finding Elimination Order: : 0%|          | 0/2 [00:00<?, ?it/s]

0%|          | 0/2 [00:00<?, ?it/s]

Eliminating: Dyspnoea: 0%|          | 0/2 [00:00<?, ?it/s]

Eliminating: Xray: 100%|██████████| 2/2 [00:00<00:00, 333.49it/s]A

```

Inferencing with bayesian network

Probability of Cancer given smoker

+-----+-----+	
Cancer	phi(Cancer)
+=====+	
Cancer(0)	0.0029
+-----+-----+	
Cancer(1)	0.9971
+-----+-----+	

Probability of Cancer given smoker,pollution

+-----+-----+	
Cancer	phi(Cancer)
+=====+	
Cancer(0)	0.0200
+-----+-----+	
Cancer(1)	0.9800
+-----+-----+	