ML LAB RECORD

```
In [1]: import pandas as pd
import numpy as np
Out[3]: Weather AirTemperature Humidity WaterTemperature Wind Goes out
          0 Sunny Moderate
                                        Normal
                                                   Warm
                                                                       Strong Yes
          1 Sunny
                      High
                                        Normal
                                                   Too Warm
                                                                       Strong No
          2 Rainy
                      Low
                                        High
                                                                       Breezy No
          3 Rainy High
                                        High
                                                                       Breezy Yes
           4 Snowy Moderate
                                        Normal Cold
                                                                       Strong Yes
In [4]: data = np.array(weatherinfo)[:,:-1]
In [6]: values = np.array(weatherinfo)[:,-1]
Out[6]: array(['Yes', 'No', 'No', 'Yes', 'Yes'], dtype=object)
In [6]: values = np.array(weatherinfo)[:,-1]
Out[6]: array(['Yes', 'No', 'No', 'Yes', 'Yes'], dtype=object)
          Find S Algorithm
Initial hypothesis: ['NULL', 'NULL', 'NULL', 'NULL']

After 0 iteration in dataset, the hypothesis is: ['Sunny', 'Moderate', 'Normal', 'Warm', 'Strong']

After 1 iteration in dataset, the hypothesis is: ['Sunny', 'Moderate', 'Normal', 'Warm', 'Strong']

After 2 iteration in dataset, the hypothesis is: ['Sunny', 'Moderate', 'Normal', 'Warm', 'Strong']

After 3 iteration in dataset, the hypothesis is: ['?', '?', '?', 'Warm', '?']

After 4 iteration in dataset, the hypothesis is: ['?', '?', '?', '?', '?']

Final hypothesis: ['?', '?', '?', '?', '?']
In [ ]:
```

Program 2:

Candidate Elimination Algorithm

```
In [1]: import pandas as pd import numpy as np
In [7]: data = pd.read_csv(r'D:\6th Sem\Machine Learning\Program 02\enjoysport.csv')
data = pd.DataFrame(data)
 Out[7]: sky airtemp humidity wind water forcast 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
 In [9]: concepts = np.array(data)[:,:-1]
concepts
In [16]: target = np.array(data)[:,-1]
target
Out[16]: array(['yes', 'yes', 'no', 'yes'], dtype=object)
In [21]: specific_h = concepts[0].copy()
  general_h = [['?' for i in range(len(specific_h))] for i in range(len(specific_h))]
  print('Initially, ')
  print('Specific Hypothesis : ')
  print(specific_h)
  print('General Hypothesis : ')
  print(general_h)
  print('general_h)
  print('general_h)
                  for i, h in enumerate(concepts):
    if target[i] == "yes": # FIND S ALGO
        for x in range(len(specific_h)):
        if h[x] != specific_h[x]:
            specific_h[x] = '?'
            general_h[x][x] = '?'
        print(specific_h)
    print(specific_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')
                   else:
    general_h[x][x] = '?'
    print(" 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(['?','?','?','?'])
                   print("\n")
print("Final Specific Hypothesis :\n", specific_h)
print("\n")
```

```
print("Final General Hypothesis :\n", general_h )
                    ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
                    steps of Candidate Elimination Algorithm 1
                        steps of Candidate Elimination Algorithm 2
                    steps of Candidate Elimination Algorithm 3

[['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?', '?'], ['?'], ['?', '?'], ['?'], ['?', '?'], ['?', '?'], ['?'], ['?', '?'], ['?'], ['?', '?'], ['?'], ['?', '?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'], ['?'],
                    Final Specific Hypothesis :
                                                                          'strong' '?' '?']
                        ['sunny' 'warm'
                     Final General Hypothesis :
                     Final General hypothesis :
[['sunny', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?']]
                                                                                                                                                                                               מיני מיני מיני מיני מיני] עריני מיני מיני מיני מינין עריני<mark>.</mark>
In [ ]:
```

Program 3:

Decision Tree ID3 Algorithm

```
In [262]: import numpy as np import pandas as pd import math
In [263]: data = pd.read_csv("id3.csv")
    data
Out[263]: Outlook Temperature Humidity Wind Answer
          0 sunny
                                  high
                      hot
                                           weak no
           1 sunny
                      hot
                                  high
                                           strong no
           2 overcast hot
                                  high
                                           weak yes
           3 rain
                                  high
           4 rain
                                           weak yes
           5 rain
                      cool
                                  normal
                                           strong no
           6 overcast cool
                                  normal
                                           strong yes
           7 sunny
                      mild
                                  high
                                           weak no
           8 sunny
                                           weak yes
                      cool
                                  normal
           9 rain
                      mild
                                           weak yes
                                  normal
           10 sunny
                                            strong yes
                      mild
                                  normal
           11 overcast mild
                                  high
                                            strong yes
           12 overcast hot
                                  normal
                                            weak yes
           13 rain
                                           strong no
```

```
In [264]: input = data.drop('Answer',axis=1)
input.head(3)

Out[264]: Outlook Temperature Humidity Wind

0 sunny hot high weak
1 sunny hot high strong
2 overcast hot high weak

*

In [265]: target = data['Answer']
target.head(3)

Out[265]: 0 no
2 yes

Name: Answer, dtype: object

In [266]: def attr_list(df):
    attributes = df.columns.tolist()[:-1]
    return attributes
    print(attr_list(data))
    ['outlook', 'Temperature', 'Humidity', 'Wind']

In [267]: def attr.values(attributes,data):
    values = {}
    for i in attributes:
        values[i] = data[i].unique()
        return values
    print(attr_values(attr_list(data),data))
```

```
In [267]: def attr_values(attributes,data):
                     values = {}
for i in attributes:
                          values[i] = data[i].unique()
               print(attr_values(attr_list(data),data))
               {'Outlook': array(['sunny', 'overcast', 'rain'], dtype=object), 'Temperature': array(['hot', 'mild', 'cool'], dtype=object), 'Humidit y': array(['high', 'normal'], dtype=object), 'Wind': array(['weak', 'strong'], dtype=object)}
In [268]: def entropy(data_n):
                     entropy(data_n):
p=len(data_n.where(data_n['Answer']=='yes').dropna())
n=len(data_n.where(data_n['Answer']=='no').dropna())
                     if(n==p):
                           return 1
                     if(n==0 or p==0):
    return 0
                     return \theta ==(((-1)*p*math.log(p/(p+n),2)/(p+n)) + ((-1)*n*math.log(n/(p+n),2)/(p+n))) return float(e)
for x in att_values:
                           A In att_values.
p_val = len(data_n.where((data_n[attribute]==x) & (data_n['Answer']=='yes')).dropna())
n_val = len(data_n.where((data_n[attribute]==x) & (data_n['Answer']=='no')).dropna())
i += ((p_val+n_val)/(p+n))*(entropy(data_n.where(data_n[attribute]==x).dropna()))
                      return i
In [270]: class Node:

def init (calf attributa).
```

```
In [270]: class Node:
                def __init__(self,attribute):
                    self.attribute=attribute
                     self.children=[]
                     self.answer = {}
In [271]: NG = ['NG','NG','NG']
root = Node(NG)
            print(root.attribute)
            ['NG', 'NG', 'NG']
In [272]: def Compute_tree(df,root):
                if(len(set(df[df.columns.tolist()[-1]])) == 1):
                return set(df[df.columns.tolist()[-1]])
e = entropy(df)
                attributes = attr_list(df)
                 values = attr_values(attributes,df)
                 gain_of_attr = {}
                gain = 0
                 best_attr = attributes[0]
                for a in attributes:
    gain_of_attr[a] = e - information(a,values[a],df)
    gain,best_attr = (gain_of_attr[a],a) if(gain_of_attr[a] > gain) else (gain,best_attr)
                node = Node(best_attr)
                node.children = values[best attr]
                for x in node.children:
                    node.answer[x]=Compute_tree(df.where(df[best_attr]==x).dropna(),root)
                return node
            r=Compute_tree(data,root)
In [273]: #print(r.attribute, r.children, r.answer['sunny'].attribute)
```

```
In [273]: #print(r.attribute,r.children,r.answer['sunny'].attribute)
print(((" ")*(len(r.children[@])+5),r.attribute)
for c in r.children:
    if(r.answer[c]=={'yes'} or r.answer[c]=={'no'}):
        print(c,r.answer[c],end=" ")

    Outlook
sunny    overcast {'yes'} rain

In [250]: def Display_tree(node,lable):
    if(node=={'yes'} or node==('no'}):
        return
    print(" ")*(24),"",str(node.attribute).upper(),"(",lable,")")
    print(" "Broches of :",node.attribute,"\"")
    for c in node.children:
        if(node.answer[c]!={'yes'} and node.answer[c]!={'no'}):
        print(c,"\n",node.answer[c],end="\n")
        print()
    for c in node.children:
        if(node.answer[c]!={'yes'} and node.answer[c]!={'no'}):
        Display_tree(node.answer[c],"child of "*str(node.attribute).upper()*" through the branch : "*c)

In [251]: Display_tree(r,"root")

    OutLook ( root )

Brnches of : Outlook
sunny
```

```
n [251]: Display_tree(r,"root")
                                   OUTLOOK ( root )
         Brnches of : Outlook
         sunnv
         overcast
          {'yes'}
         rain
                                   HUMIDITY ( child of OUTLOOK through the branch : sunny )
         Brnches of : Humidity
         high
          {'no'}
         normal
          {'yes'}
                                   WIND ( child of OUTLOOK through the branch : rain )
         Brnches of : Wind
          {'yes'}
         strong
{'no'}
```

Program 4:

Bayesian Classifier

Naive Bayes Classifier

```
In [36]: import pandas as pd
                  data = pd.read_csv('PlayTennis.csv')
data.head()
Out[36]: PlayTennis Outlook Temperature Humidity Wind
                   0 No
                                           Sunny Hot
                   1 No
                                            Sunny
                                                                                                    Strong
                   2 Yes
                                           Overcast Hot
                                                                                  High
                                                                                                   Weak
                   3 Yes
                                           Rain
                                                           Mild
                                                                                  High
                                                                                                   Weak
                                                          Cool
                   4 Yes
                                           Rain
                                                                                  Normal Weak
In [40]: y = list(data['PlayTennis'].values)
X = data.iloc[:,1:].values
                  print(f'Target Values: {y}')
print(f'Features: \n{X}')
                   Tanget Values: ['No', 'No', 'Yes', 'Yes', 'Yes', 'No', 'Yes', 'No', 'Yes', 'Yes', 'Yes', 'Yes', 'Yes', 'No']
                  Target Values: [No, No, Yes, Features:
[['Sunny' 'Hot' 'High' 'Weak']
['Overcast' 'Hot' 'High' 'Weak']
['Rain' 'Mild' 'High' 'Weak']
                   ['Sunny' 'Hot' 'High' 'Strong']
['Overcast' 'Hot' 'High' 'Weak']
['Rain' 'Mild' 'High' 'Weak']
['Rain' 'Cool' 'Normal' 'Weak']
['Overcast' 'Cool' 'Normal' 'Strong']
['Overcast' 'Cool' 'Normal' 'Weak']
['Sunny' 'Mild' 'High' 'Weak']
['Sunny' 'Mild' 'Normal' 'Weak']
['Sunny' 'Mild' 'Normal' 'Weak']
['Sunny' 'Mild' 'Normal' 'Strong']
['Overcast' 'Mild' 'High' 'Strong']
['Overcast' 'Hot' 'Normal' 'Weak']
['Rain' 'Mild' 'High' 'Strong']]
In [41]: y_train = y[:8]
y_val = 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
In [42]: class NaiveBayesClassifier:
                         def __init__(self, X, y):
                               self.X, self.y = X, y
```

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

self.N = 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[1][j] in self.attrs[j]:
            self.attrs[j].append(self.X[1][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():
```

```
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
```

```
[46]: 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)
```

Program 5:

Bayesian Network

Baysian Network

```
In [2]: import numpy as np
import pandas as pd
import csv
          Import CSV
from pgmpy.estimators import MaximumLikelihoodEstimator
from pgmpy.models import BayesianModel
from pgmpy.inference import VariableElimination
In [13]: heartDisease = pd.read_csv('heart.csv')
heartDisease.head(10)
Out[13]:
         age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal heartdisease
          0 63 1 1 145
                                 233 1 2
                                                                 2.3
                                                                               0 6
          1 67 1
                    4 160
                                                                               3 3
                                                                 1.5
          2 67 1
                    4 120
                                 229 0 2
                                                                 2.6
                                                                               2 7
          3 37 1
                    3 130
                                 250 0 0
                                                                 3.5
                                                                               0 3
          4 41 0
                    2 130
                                 204 0 2
                                                                               0 3
                                                                                       0
          5 56 1
                    2 120
                                 236 0 0
                                                                 0.8
                                                                               0 3
                                                                                       0
          6 62 0 4 140
                                 268 0 2
                                                                 3.6
                                                                               2 3
                                                                                       3
          7 57 0 4 120
                                 354 0 0
                                                   163
                                                                 0.6
                                                                               0 3
                                                                                       0
          8 63 1 4 130
                                 254 0 2
                                                   147
                                                                 1.4
                                                                               1 7
                                                                                       2
          9 53 1 4 140
                                 203 1 2
                                                   155
                                                                 3.1
                                                                               0 7
          mean | 54.438944 | 0.679868
                                      3.158416
                                                  131.689769 | 246.693069 | 0.148515
                                                                                   0.990099
                                                                                               149.607261 | 0.326733
                                                                                                                    1.039604
                                                                                                                               1.600660
                                                                                                                                           0.683168
               9.038662
                          0.467299
                                      0.960126
                                                  17.599748 51.776918 0.356198
                                                                                   0.994971
                                                                                               22.875003 0.469794
                                                                                                                     1.161075
                                                                                                                                0.616226
                                                                                                                                           0.944808
                29.000000 0.000000
                                      1.000000
                                                  94.000000 126.000000 0.000000
                                                                                   0.000000
                                                                                               71.000000 0.000000
                                                                                                                     0.000000
                                                                                                                                 1.000000
                                                                                                                                           0.000000
               48.000000 0.000000
                                      3.000000
                                                  120.000000 211.000000 0.000000
                                                                                   0.000000
                                                                                               133.500000 0.000000
                                                                                                                     0.000000
                                                                                                                                 1.000000
                                                                                                                                           0.000000
```

	min	29.000000	0.000000	1.000000	94.000000	126.000000	0.000000	0.000000	/1.000000	0.000000	0.000000	1.000000	0.000000	4	
	25%	48.000000	0.000000	3.000000	120.000000	211.000000	0.000000	0.000000	133.500000	0.000000	0.000000	1.000000	0.000000	3	
	50%	56.000000	1.000000	3.000000	130.000000	241.000000	0.000000	1.000000	153.000000	0.000000	0.800000	2.000000	0.000000	3	
	75%	61.000000	1.000000	4.000000	140.000000	275.000000	0.000000	2.000000	166.000000	1.000000	1.600000	2.000000	1.000000	7	
	max	77.000000	1.000000	4.000000	200.000000	564.000000	1.000000	2.000000	202.000000	1.000000	6.200000	3.000000	3.000000	7	
	4)	•	
In [7]:	<pre>model = BayesianModel([('age','heartdisease'),('sex','heartdisease'),('exang','heartdisease'),('cp','hear</pre>														
	Learning CPD using Maximum likelihood estimators														
In [8]:	3]: model.fit(heartDisease, estimator=MaximumLikelihoodEstimator) 9]: print('Inferencing with Bayesian Network:') HeartDiseasetest_infer = VariableElimination(model)														
In [9]:															
	Infere	Inferencing with Bayesian Network:													
In [24]:	<pre>print('1. Probability of HeartDisease given evidence = restecg : 1') q1 = HeartDiseasetest_infer.query(variables=['heartdisease'],evidence={'restecg':1}) print(q1)</pre>														
	0% Elimin Elimin Elimin Elimin	ating: chol ating: age: ating: cp: ating: exan	.on Order: : 0/5 [00:00< .: 0% 0% 0% 100%	?, ?it/s] 0/s 0/5 0/5	5 [00:00 ,<br [00:00 ,<br [00:00 , ?</th <th>?it/s] it/s] , ?it/s]</th> <th></th> <th></th> <th></th> <th></th> <th></th> <th></th> <th></th> <th></th>	?it/s] it/s] , ?it/s]									

```
Finding Elimination Order: : 0% | | 0/5 [00:00<?, ?it/s] | 0% | | 0/5 [00:00<?, ?it/s] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:00<0.00] | 0/5 [00:0
                                                                                                                                                | 0/5 [00:00<?, ?it/s]
             1. Probability of HeartDisease given evidence = restecg : 1
              | heartdisease | phi(heartdisease)
+-----
             | heartdisease(1) |
                                                                                                               0.0000
              heartdisease(2)
                                                                                                               0.2361
                                                                                                               0.2017
              heartdisease(3)
              heartdisease(4)
8]: print('Tuples with \'restecg = 1\' in the database are:')
heartDisease[heartDisease['restecg'] == 1]
             Tuples with 'restecg = 1' in the database are:
                          age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal heartdisease
              231 55 0 4 180
                                                                                       327 0 1
                                                                                                                                                                                   3 4
                                                                                                                                                                                                           2
                                                                                                                                                                                                                           0 3
                                                                                                                                                                                                                                                  2
              257 76 0 3 140
                                                                                        197 0 1
                                                                                                                                                                                                                           0
                                                                                                                                                                                                                                     3
                                                                                                                                                                                                                                                  0
             282 55 0 4 128
                                                                                                                                                                                                                          1 7
                                                                                     205 0 1
                                                                                                                                      130
                                                                                                                                                                                 20
  Out[18]:
                                         age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal heartdisease
                               231 55 0 4 180
                                                                                                    327 0
                                                                                                                                                                                                                                    0
                                                                                                                                                                                                                                             3
                                                                                                                                                      117
                                                                                                                                                                                            3.4
                                                                                                                                                                                                                                                          2
                               257 76 0
                                                                 3 140
                                                                                                                                                                                                                                     0 3
                                                                                                      197 0
                                                                                                                                                                                              1.1
                                                                                                                                                                                                                   2
                                                                                                                                                                                                                                                          0
                                                                                                                                                      116
                                                                                                                                                                          0
                               282 55 0
                                                                 4 128
                                                                                                                                                                                                                                     1 7
                                                                                                                                                                                                                                                          3
                                                                                                    205 0
                                                                                                                                                                                                                   2
                                                                                                                                                      130
                                                                                                                                                                                            2.0
                               285 58 1 4 114
                                                                                                                                                                                                                                    3 6
                                                                                                    318 0
                                                                                                                                                                         0
                                                                                                                                                                                            4.4
                                                                                                                                                                                                                   3
                                                                                                                                                                                                                                                          4
                                                                                                                                                      140
  In [20]: print('2. Probability of HeartDisease given evidence = cp : 2')
q2=HeartDiseasetest_infer.query(variables=['heartdisease'],evidence={'cp':2})
                             Finding Elimination Order: : 100% | 1/5 [00:37<00:00, 7 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/5 | 1/
                             2. Probability of HeartDisease given evidence = cp : 2
                              | heartdisease | phi(heartdisease) |
                              | heartdisease(0) |
                                                                                                                          0.3742
                                  heartdisease(1)
                              heartdisease(2)
                                                                                                                           0.1375
                                | heartdisease(2) |
                                heartdisease(3)
                                                                                                                            0.1541
                                | heartdisease(4) |
                                     heartdisease(4) | 0.1323
   In [23]: print('Tuples with \'cp = 2\' in the database are:') heartDisease[heartDisease['cp'] == 2].head(10)
                               Tuples with 'cp = 2' in the database are:
    Out[231:
                                        age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal heartdisease
                                                                 2
                                 4 41
                                                     0
                                                                           130
                                                                                                    204 0
                                                                                                                          2
                                                                                                                                                    172
                                                                                                                                                                                                                                   0
                                                                                                                                                                                                                                            3
                                                                                                                                                                                             1.4
                                                                                                                                                                                                                                                        0
                                5 56
                                                                  2
                                                                            120
                                                                                                    236 0
                                                                                                                                                     178
                                                                                                                                                                         0
                                                                                                                                                                                            0.8
                                                                                                                                                                                                                                   0
                                                                                                                                                                                                                                            3
                                                                                                                                                                                                                                                        0
                                                                                                                            0
                                 11 56
                                                                  2
                                                                                                    294 0
                                                                                                                                                                                                                                            3
                                                      0
                                                                            140
                                                                                                                            2
                                                                                                                                                     153
                                                                                                                                                                         0
                                                                                                                                                                                            1.3
                                                                                                                                                                                                                                   0
                                                                                                                                                                                                                                                        0
                                 13 44
                                                                 2
                                                                            120
                                                                                                    263 0
                                                                                                                                                    173
                                                                                                                                                                                                                                   0
                                                                                                                                                                                                                                                        0
                                                                                                                            0
                                                                                                                                                                         0
                                                                                                                                                                                            0.0
                                  16 48
                                                                 2
                                                                            110
                                                                                                    229 0
                                                                                                                                                                                            1.0
                                                                                                                                                                                                                                   0
                                                                                                                            0
                                                                                                                                                     168
                                 19 49
                                                                 2
                                                                            130
                                                                                                    266 0
                                                                                                                                                    171
                                                                                                                                                                                            0.6
                                                                                                                                                                                                                                   0
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                                22 58
                                                                 2
                                                                            120
                                                                                                    284 0
                                                                                                                                                    160
                                                                                                                                                                                            1.8
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                                                                                                                                                                                                                                           3
                                 42 71 0
                                                                 2
                                                                            160
                                                                                                    302 0
                                                                                                                            0
                                                                                                                                                    162
                                                                                                                                                                                           0.4
                                                                                                                                                                                                                                   2
                                                                                                                                                                                                                                           3
                                                                                                                                                                                                                                                        0
                                 50 41 0
                                                                 2
                                                                            105
                                                                                                    198 0
                                                                                                                            0
                                                                                                                                                    168
                                                                                                                                                                                           0.0
                                                                                                                                                                                                                                   1
                                                                                                                                                                                                                                            3
                                                                                                                                                                                                                                                        0
                                53 44
                                                     1
                                                                 2
                                                                            130
                                                                                                   219 0 2
                                                                                                                                                    188
                                                                                                                                                                         0
                                                                                                                                                                                           0.0
                                                                                                                                                                                                                                   0 3
                                                                                                                                                                                                                                                        0
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