

1. Implement and demonstrate the FIND-S algorithm for finding the most specific hypothesis based on a given set of training data sample. Read the training data from a file.

Import random

import csv

```
attributes = [['sunny', 'rainy'], ['warm', 'cold'], ['windy', 'light'], ['strong', 'weak'], ['warm', 'cool'], ['same', 'change']]
```

print(attributes)

sun_attributes = len(attributes)

print(sun_attributes)

print("b the most general hypothesis: ['?', '?', '?', '?', '?', '?']")

print("b the most specific hypothesis: ['o', 'o', 'o', 'o', 'o', 'o']")

q = []

print("b the gives training dataset")

with open('c:\Users\Adm\desktop\Aids\data.csv', 'r') as csvfile:

reader = csv.reader(csvfile)

for row in reader:

q.append(row)

print(row)

print("the initial value of hypothesis: ")

h = ['o'] * sun_attributes

print(h)

for j in range(0, num_attributes):

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$$\neg c_{ij} = \neg c_j c_{ij}$$

for i is range(1, len(a));

if ($a[i]$ [sum-attributes] == 'yes');

for j is range(sum-attributes);

if ($c_{ijj} == '0'$ or $\neg c_{ij} == a[i][j]$):

$$\neg c_{ijj} = \neg c_{ij} c_{ij}$$

else:

$$\neg c_{ij} = '?'$$

Prior ("1s for training examples : 20% the hypothesis") formed ($i+1$, b)

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Dataset:

Sl. No	Sky	Airtemp	Humidity	Wind	water	Forecast	Enjoy Sport
1	Sunny	cooms	normal	strong	warm	same	Yes
2	Sunny	cooms	high	strong	warm	same	Yes
3	Rainy	cold	high	strong	warm	change	No
4.	Sunny	cooms	high	strong	cool	change	Yes

output:-

[['sunny', 'rainy'], ['cooms', 'cold'], ['normal', 'high'], ['strong', 'weak'], ['cooms', 'cool'], ['same', 'change']]

b.

the most general hypothesis. [?, ?, ?, ?, ?, ?, ?]

the most specific hypothesis: [0, 0, 0, 0, 0, 0, 0]

The given training dataset

['Sky', 'Airtemp', 'Humidity', 'Wind', 'water', 'Forecast', 'Enjoy Sport']

['sunny', 'cooms', 'normal', 'strong', 'warm', 'same', 'Yes']

['sunny', 'cooms', 'high', 'strong', 'warm', 'same', 'Yes']

['Rainy', 'cold', 'high', 'strong', 'warm', 'change', 'Yes']

The initial hypothesis is :

[0, 0, 0, 0, 0, 0]

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for training example : {0} the hypothesis is
['sunni', 'wants', '?', 'strange', '?', '?']

2. for a given set of set of training data example stored in a .csv file, implement and demonstrate the candidate-elimination algorithm to output a description of the set of all hypothesis consistent with the training example.

```
import csv
```

```
with open('c://users//Admisi//Desktop//41667//adult.csv')  
as f:
```

```
csv_file = csv.reader(f)
```

```
data = list(csv_file)
```

```
print(data)
```

```
s = data[1][:-1]
```

```
g = [[?] for i in range(len(s))] for j in range(  
len(s))]
```

```
for i in data:
```

```
if [-1] == "yes":
```

```
    for j in range(len(s)):
```

```
        if i[j] == s[j]:
```

```
            g[i][j] = '?'
```

```
        g[i][j] = '?'
```

```
elif [-1] == "No":
```

```
    for j in range(len(s)):
```

```
        if i[j] == s[j]:
```

```
            g[i][j] = s[j].
```

```
else:
```

```
    g[i][j] = "?"
```

```
print("10 steps of candidate elimination algorithm")  
data.index(i+1)
```

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printf(s)

printf(g)

gh = {};

for i is g:

for j is i:

if 'j' == '?' :

gb.append(i)

break

printf(") to find specific hypothesis: (", s)

printf(") to find general hypothesis: (", "igh")

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Dataset:

Sunny.	warm.	normal	strong	warm	same	Yes
Sunny	warm.	high.	strong	warm	same	No
Rainy.	cold.	high.	strong	warm	change	No
Sunny	warm	high.	strong	warm.	change	Yes

Output:

steps of candidate elimination algorithm 1

[['sunny', 'warm', '?', 'strong', '?', '?'],
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'],
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

steps of candidate elimination algorithm 2.

[['sunny', 'warm', '?', 'strong', '?', '?'],
[['sunny', '?', '?', '?', '?', '?'], ['?', 'wind', '?', '?', '?', '?'],
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

steps of candidate elimination algorithm 3.

[['sunny', 'warm', '?', 'strong', '?', '?'],
[['sunny', '?', '?', '?', '?', '?'], ['?', 'wind', '?', '?', '?', '?'],
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

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steps of candidate elimination algorithm &

['sunray', 'wants', '?', 'strong', '?', '?']

['sunray', '?', '?', '?', '?', '?'], ['?', 'wants', '?', '?', '?', '?'],

['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']

['?', '?', '?', '?', '?'],

find specific by postnysis

['sunray', 'wants', '?', 'strong', '?', '?']

find general hypothesis

['sunray', '?', '?', '?', '?', '?']

3. write a program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate data set for building the decision tree and apply this knowledge to classify a new sample.

```
import pandas as pd.
```

```
from pandas import DataFrame.
```

```
df_tennis = pd.read_csv('c://Users//Admin//Desktop//  
Aibin//playtennis.csv')
```

```
attribute_names = list(df_tennis.columns)
```

```
attribute_names.remove('playTennis').
```

```
print(attribute_names).
```

```
def entropy_of_list(lst):
```

```
from collections import Counter.
```

```
count = Counter(x for x in lst)
```

```
num_instances = len(lst)**1.
```

```
probs = [x/num_instances for x in count.values]
```

```
return entropy(probs).
```

```
def entropy(probs):
```

```
import math
```

```
return sum([-prob * math.log(prob) for  
prob in probs]).
```

```
Total_entropy = entropy_of_list(df_tennis['  
playTennis'])
```

```
def information_gains(df, split_attribute_name,  
target_attribute_name, tree_id=0):
```

```
df_split = df.groupby(split_attribute_name)
```

```
nobs = len(df.index)**1.
```

$\text{df_agg_ent} = \text{df split_agg} (\{\text{target_attribute_name:}$
 $\text{[entropy_of_list, lambda x: len(x)/len(df)}\}$
 $\text{df_agg_ent.columns} = [\text{'entropy'}, \text{'prop observations'}]$
 $\text{new_entropy} = \text{sum}(\text{df_agg_ent}[\text{'entropy'}] * \text{df_}$
 $\text{agg_ent}[\text{'prop observations'}])$
 $\text{old_entropy} = \text{entropy_of_list} (\text{df}[\text{target_attribute_name}])$
 $\text{print(split_attribute_name, 'Ig:', old_entropy_new_}$
 $\text{entropy})$
 returns old-entropy new-entropy

$\text{def id3} (\text{df, target_attribute_name, attribute_names, default_class} = \text{None}):$

$\text{from collections import Counter:}$

$\text{count} = \text{Counter} (\text{x for x in df[target_attribute_name]})$

$\text{if len(count)} = 1:$

$\text{return reset} (\text{i for count})$

$\text{elif df.empty or (not attribute_names):}$

$\text{return default_class}$

else:

$\text{default_class} = \max (\text{count}.values())$

$\text{gains} = [\text{Information_gain} (\text{df}, \text{attr}, \text{target_attribute_name}) \text{ for attr in attribute_names}]$

$\text{print} (")$

$\text{index_of_max} = \text{gains}.index (\max (\text{gains}))$

$\text{best_attr} = \text{attribute_names}[index_of_max]$

$\text{tree} = \{ \text{best_attr}: \{\}\}$

$\text{remaining_attribute_names} = [i \text{ for } i \text{ in attribute_names if } i \neq \text{best_attr}]$

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name of $\{ \} = \text{best-attr} \}$

for attr-val, data-subset is df. group by.
(best attr):

subtree = DB (data-subset, target-attribute-
name, remaining attribute-names-default-
class).

tree [best attr] [attr-val] = subtree

returns -

from print to print print.

tree = DB (df, team's, 'playtennis', attribute-
names).

print("b") The resultant decision tree is:
b") print(tree)

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Output

['outlook', 'temperature', 'humidity', 'wind'].

outlook 1G = 0.2467498197744391

temperature 1G = 0.029222565658554647

humidity 1G : 0.15183550136234136

wind 1G : 0.04818703040826927

temperature 1G : 0.01924309402199489

humidity 1G : 0.0199730400197489

wind 1G : 0.9209505944546686

temperature 2G : 0.570909505944546686

humidity 2G : 0.9709505944546686

wind 2G : 0.01997309402197489

The result of decision tree is

{'outlook': {'overcast': 'Yes',

'Rain': {'wind': {'strong': 'No', 'weak': 'Yes'}},

'Sun': {'Humidity': {'High': 'No', 'Normal': 'Yes'}}}

{}

4 Build an artificial neural network by implementing the back propagation algorithm & test the same using appropriate data sets.

from import math as

from random import seed,

from random import random.

def initialize_network (n_inputs, n_hidden,
n_outputs):

network = []

hidden_layer = [{}'weights': [random() for i in
range (n_inputs+1)]} for i in range (n_hidden)]

network.append (hidden_layer)

output_layer = [{}'weights': [random() for i in
range (n_hidden+1)]} for i in range (n_outputs)].

network.append (output_layer)

return network

def activate (weights, inputs):

activation = weights [-1].

for i in range (len(weights)-1):

activation += weights [i] * inputs [i].

return activation

def transfer (activation)

return log(1.0 + exp(-activation))

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def forward_propagate (network, row):

 Input = row

 For layer i in network:

 new_inputs = []

 for neuron in layer:

 activations = activate (neuron['weights'])

 inputs = neuron['outputs'] = transfer
 (activations)

 new_inputs.append (neuron['outputs'])

 Inputs = new_inputs.

 return inputs.

def transfer_derivative (output):

 return outputs (1.0 - output).

def backward_propagate_error (network,
 expected):

 for i in reversed (range (len (network))):

 layer = network [i].

 errors = []

 if i != len (network) - 1:

 for j in range (len (layer)):

 error = 0.0.

 For neuron i in network [i+1]:

 error += (neuron ['weights'][i])

 neuron ['delta']).

 errors.append (error).

 else:

 for j in range (len (layer)):

 neuron = layer [j]

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errors.append(excepted[j].neuron['output'])
 for j in range(len(layer)):
 zeros = layer[i].
 zeros['delta'] = error[j] * transfer-
 derivative(zeros['output'])

def update_weights(network, row, l-rate):

for i in range(len(network)):

inputs = row[i-1]

if i == 0:

inputs = [zeros['output']] for zeros
 in network[i-1].

for neuron in network[i]:

for j in range(len(inputs)):

zeros['weights'][i][j] += l-rate * neuron.
 ['delta'] * inputs[i].

zeros['weights'][i][j] += l-rate * neuron['
 delta'].

def train_network(network, train, l-rate, n-
 epoch, n_outputs):

for epoch in range(n-epoch):

sum_error = 0.

for row in train:

outputs = forward_propagate(network,
 row).

expected = [0 for i in range(n-
 outputs)]

expected[-1] = 1

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sum_error += sum([expected[i] - outputs[i] for i in range(len(expected))])

broadcast-propagate-error(network, expected)

update-weights(network, vars_l_rate)

print('> epoch = ', d, 'l_rate = ', l_rate, 'error = ', error)

∴ (epoch, l-rate, sum error)

seed(d)

dataset = [[2.4810850, 0.550537003, 0],
 [1.465489372, 0.862125076, 0],
 [3.326561688, 4.400293529, 0],
 [1.38807019, 1.8500000317, 0],
 [3.06407232, 3.00530599310],
 [7.627531014, 0.759262235, 1],
 [5.0330441048, 2.088626795, 1],
 [6.922596716, 1.77106367, 1],
 [8.675418631, -0.242068655, 1],
 [7.673756466, 3.505563011, 1]].

→ inputs = len(dataset) - 1

→ outputs = len([set([outputs[i] for i in dataset])])

→ network = initialize_network(inputs, 2, outputs)

→ train_network(network, dataset, 0.8, 20, outputs)

→ layer is network:

→ print(layer)

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epoch = 0	lrade = 0.500	error = 6.350
epoch = 1	lrade = 0.500	error = 5.531
epoch = 2	lrade = 0.500	error = 5.001
epoch = 3	lrade = 0.500	error = 4.998
epoch = 4	lrade = 0.500	error = 4.899
epoch = 5	lrade = 0.500	error = 4.173
epoch = 6	lrade = 0.500	error = 3.835
epoch = 7	lrade = 0.500	error = 3.806
epoch = 8	lrade = 0.500	error = 3.198
epoch = 9	lrade = 0.500	error = 0.898
epoch = 10	lrade = 0.500	error = 0.626
epoch = 11	lrade = 0.500	error = 0.347
epoch = 12	lrade = 0.500	error = 0.183
epoch = 13	lrade = 0.500	error = 1.953
epoch = 14	lrade = 0.500	error = 1.776
epoch = 15	lrade = 0.500	error = 1.614
epoch = 16	lrade = 0.500	error = 1.478
epoch = 17	lrade = 0.500	error = 1.366
epoch = 18	lrade = 0.500	error = 1.283
epoch = 19	lrade = 0.500	error = 1.152

{'weights': [-1.468875095432307, 1.950887325439314, 1.0858178629550299], 'output': 0.009980305604426185, 'delta': -0.059566604162323605},

{'weights': [0.37711098140462157, -0.0625909814552989, 0.027651237026427167], 'output': 0.94568890021123, 'delta': 0.00262796528508638379},

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[$\{w\}$ weights]: $[2.518394649397849, -0.339182750264,$
 $5.988, -0.9671565426390275],$ 'output': -
 $6.2564879202357587,$ 'delta': -
 $-0.092700592783645873,$
 1.00364221062

[$\{w\}$ weights]: $[-2.65891498484263, 0.423830864675827157]$
 $0.9202, 0.423830864675827157],$ 'output': $0.7790535202438367,$
'delta': $0.83803132564373543].$

Copy & type your handwritten notes
in the following boxes

s. create a program to implement the naive bayes classifier for a sample training data stored as .csv file. compute the accuracy of the classifier. Considering few test dataset.

```
import csv, random, math
import statistics as st
```

```
def loadCSV(filename):
    lines = csv.reader(open(filename, "r"))
    dataset = list(lines)
    for i in range(len(dataset)):
        dataset[i] = [float(x) for x in dataset[i]]
    return dataset
```

```
def splitDataset(dataset, splitRatio):
    testSize = int(len(dataset)*splitRatio)
    trainSet = list(dataset)
    testSet = []
    while len(testSet) < testSize:
        index = random.randrange(len(trainSet))
        testSet.append(trainSet.pop(index))
    return [trainSet, testSet]
```

```
def separateByClass(dataset):
    separated = {}
    for i in range(len(dataset)):
        x = dataset[i]
        if (x[-1] not in separated):
            separated[x[-1]] = []
            separated[x[-1]].append(x)
        else:
            separated[x[-1]].append(x)
```

separated[x[-1]] = []

separated[x[-1]].append(x)

return separated.

def compute_mean_std (dataset):

mean_std = [(ct - mean(attrs[attribute]) - st - stdv -
(attribute))]

for attribute in zip(*dataset) :

def mean_std(f):

return mean_std.

def summarize_byclass (dataset):

separated = separate_byclass (dataset):

summary = {}

for classvalue in instance in separated : items()

summary[classvalue] = compute_mean_std

(instances)

return summary.

def estimated_probability (x, mean, stdv):

exponent = math.exp(-(math.pow(x

mean, 2) / (2 * math.pow(stdv, 2)))

return (1 / math.sqrt(2 * math.pi) * stdv) *
exponent.

def calculate_class_probabilities (summary, testvector):

p = {}

for classvalue, classsummarized in summary:
items():

$P[\text{classvalues}] = 1$

for i in range(len(classsumaries)):

mean, std = classsumaries[i].

x = testvector[i].

$P[\text{classvalue} \neq \text{estimated probability}(x, \text{mean}, \text{std})]$:

return p.

def predict(classsumaries, testvector):

all_p = calculateclassprobabilities(classsumaries, testvector).

bestlabel, bestprob = None, -1.

for lbl, p in all_p.items():

bestprob = p

bestlabel = lbl.

return bestlabel.

def performclassification(classsumaries, testset):

predictions = []

for i in range(len(testset)):

results = predict(classsumaries, testset[i])

predictions.append(results)

return predictions.

def setAccuracy(testset, predictions):

correct = 0.

for i in range(len(testset)):

If testset[i][-1] == predictions[i]:

correct += 1.

return (correct / float(len(testset))) * 100.0.

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```
dataset = loadcsv('C:/Users/Admin/Desktop/16N1  
diabetes.csv')  
print('Pima Indian Diabetes dataset loaded..!')  
print('Total instances available :', len(dataset))  
print('Total attributes present :', len(dataset[0]))  
print('First five instances of dataset :')  
for i in range(5):  
    print(i+1, ':', dataset[i])  
splitRatio = 0.2  
trainset, testset = splitDataset(dataset,  
splitRatio)  
print('Dataset is split into training and  
test set.')  
print('Training example =', len(trainset), 'testing examples  
= ', len(testset))  
summarize = summarizeByClass(trainset)  
prediction = performClassification(summarize,  
testset)  
accuracy = getAccuracy(testset, prediction)  
print('Accuracy of the Naive Bayesian  
classifier is :', accuracy)
```

Teacher's Signature :

Output

pine indians diabetes dataset loaded.

Total instances available 768

Total attributes present - 8

first five rows ~~of dataset~~

1: [6.0, 148.0, 72.0, 35.0, 0.0, 33.6, 0.627, 80.0, 70]

2: [1.0, 85.0, 66.0, 0.0, 0.0, 26.6, 0.351, 31.0, 0]

3: [8.0, 183.0, 64.0, 0.0, 0.0, 23.3, 0.672, 32.0, 7.0]

4: [1.0, 89.0, 66.0, 23.0, 0.0, 28.1, 0.167, 21.0, 0.0]

5: [0.0, 133.0, 40.0, 35.0, 168.0, 43.1, 0.288, 33.0, 110]

~~dataset~~ is split into training and testing set

Training example - 615

Testing example = 163

Accuracy of the Naive Bayes classifier
is 75. ~~76.222809281046~~.

Assuming a set of documents that need to be classified, use the naive bayesian classifier model to perform this task. Built-in Java classes / API can be used to write the program, calculate the accuracy, precision and recalls for your dataset.

Import pandas as pd.

```
msg = pd.read_csv('data.csv', names=['message', 'label'])
```

```
print('Total instances in the dataset: ', msg.shape[0])
```

```
msg['labelnum'] = msg.label.map({'pos': 1, 'neg': 0})
```

x = msg.message

y = msg.labelnum

```
print('The message and its label of first 5 instances are listed below')
```

```
x5, y5 = x[0:5], msg.label[0:5]
```

for x, y in zip(x5, y5):

```
    print(x, ' ', y)
```

```
from sklearn.model_selection import train_test_split
```

```
print('Dataset is split into training and testing sample')
```

```
print('Total training instances: ', xtrain.shape[0])
```

```
print('Total testing instances: ', xtest.shape[0])
```

```
from sklearn.feature_extraction.text import TfidfVectorizer
```

Import CountVectorizer.

```
count_vect = CountVectorizer()
```

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```
xtrain-dtm = count-vec fit-transform (xtrain),  
xtest-dtm = count-vec + transform (xtest)  
print('Total features extracted using  
countvectorizer:', xtrain-dtm.shape[1])  
print('5 features for first 5 training instances  
listed below')  
df = pd.DataFrame(xtrain-dtm.toarray()), columns=  
count vec.get-feature_names()  
print(df[0:5])  
from sklearn.naive-bayes import MultinomialNB  
clf = MultinomialNB().fit(xtrain-dtm, ytrain)  
predicted = clf.predict(xtest-dtm)  
print('Classification results of testing samples  
are given below').  
for doc, pth in zip(xtest, predicted):  
    pred = 'pos' if q == 1 else 'neg'  
    print('{}s -> {}s'.format(doc, pred))  
from  
from sklearn import metrics  
print('Accuracy metrics')  
print('Accuracy of the classifier is', metrics.  
accuracy_score(ytest, predicted))  
print('Recall:', metrics.recall_score(ytest, predicted),  
'Precision:', metrics.precision_score(ytest,  
predicted))  
print('confusion matrix')  
print(metrics.confusion_matrix(ytest, predicted)).
```

Teacher's Signature :

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output

output.
total instance in the dataset : 18.

The message and its label of first 5 bytes
are listed below:

I love sandwich, too

This is an amazing place, too.

I feel very good about these beers, ps.

This is my best work, pos

What an awesome view, pos

Dataset is split into Training & Testing + Sepsis

Total training instances: 13

Total testing instance is

Total feature extracted using count vectorizer
feature for first 5 instances tracking are
listed below.

about. ans ans ~~sweat~~

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	tomorrow	very	view	we	want	what	will	with	work
0	0	0	0	0	0	0	0	0	0
1	0	0	0	1	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0

[5 rows x 46 columns]

classification result of testing sample are given below:

I love to dance \rightarrow pos

I am sick and tired of this place \rightarrow neg

This is an amazing place \rightarrow pos

What a great holiday \rightarrow pos

This is a bad locality to stay \rightarrow neg

accuracy metrics:

accuracy of the classifier is 1.0.

Recall : 1.0

precision : 1.0

confusion matrix.

$[[0, 0], [0, 3]]$

7. write a program to construct a Bayesian network considering medical data. Use this model to reconstruct the diagnosis of heart patient using standard heart disease dataset. You can use Java/python ML library classes.

```

import numpy as np
import pandas as pd
import csv
from pgmpy.estimators import MaximumLikelihoodEstimator
from pgmpy.models import BayesianModel
from pgmpy.inference import VariableElimination
heartdisease = pd.read_csv('heart.csv')
heartdisease = pd.read_csv('heart.csv').replace('?', np.nan)
print('sample instance from the dataset are given below')
print(heartdiseases.head())
print(heartdiseases.head())
print('The Attributes and datatypes')
print(heartdiseases.dtypes)
Model = BayesianModel([('age', 'heart disease'), ('sex', 'heart disease'),
                      ('heart disease'), ('exang', 'heart disease'), ('cp',
                      'heart disease'), ('heart disease', 'restecg'), ('heart disease',
                      'chol')])
print('In learning, CPD using maximum likelihood estimators')
model.fit(heartdiseases, estimator=MaximumLikelihoodEstimator)

```

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estimator).

```
print('I. Inference with Bayes network: ')
need disease test_infer = Variable_Eliminator(model)
print('In I. probability of heart disease given
evidence = resteq: 1')
q1 = heartDiseaseTestInfer.query(variables =
['heartdisease'])
evidence = { 'resteq': 1 }
print(q1)
print('In 2 probability of instances of given
evidence = CP: q2')
q2 = heartDiseaseTestInfer.query(variables =
['heartdisease'], evidence = { 'CP': 2 })
print(q2)
```

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output

few examples from the dataset given below

	age	sex	cp	restbps	chol	slope	ca	thal	heartdisease
0	63	1	1	145	233	2.3	3	0	0
1	67	1	4	160	286	1.5	2	3	1
2	67	1	4	120	229	2.6	2	2	0
3	37	1	3	130	200	3.5	3	6	0
4	41	0	2	130	204	1.4	1	0	1

[5 rows x 10 columns]

Attribute and datatypes

age	float64
sex.	float64
cp	float64
restbps	float64
chol	float64
fbs	float64
resteepg	float64
thalach.	float64
alcpalc.	float64
slope	float64
ca	float64
thal	float object
heartdisease	object
d_typeobject	float64

learning cbp

Inferencing using maximum likelihood estimated
Bayesian Network!

Expt. No.

1. probability of heart disease given evidence
 = resteq.

heart disease	$p_{bi}(\text{heart disease})$
heart disease (0)	0.1012
heart disease (1)	0.0000
heart disease (2)	0.2392
heart disease (3)	0.2015
heart disease (4)	0.458

2. probability of heart disease gives evidence
 = cf.

heart disease	$p_{bi}(\text{heart disease})$
heart disease (0)	0.3610
heart disease (1)	0.2519
heart disease (2)	0.1373
heart disease (3)	0.1537
heart disease (4)	0.1321