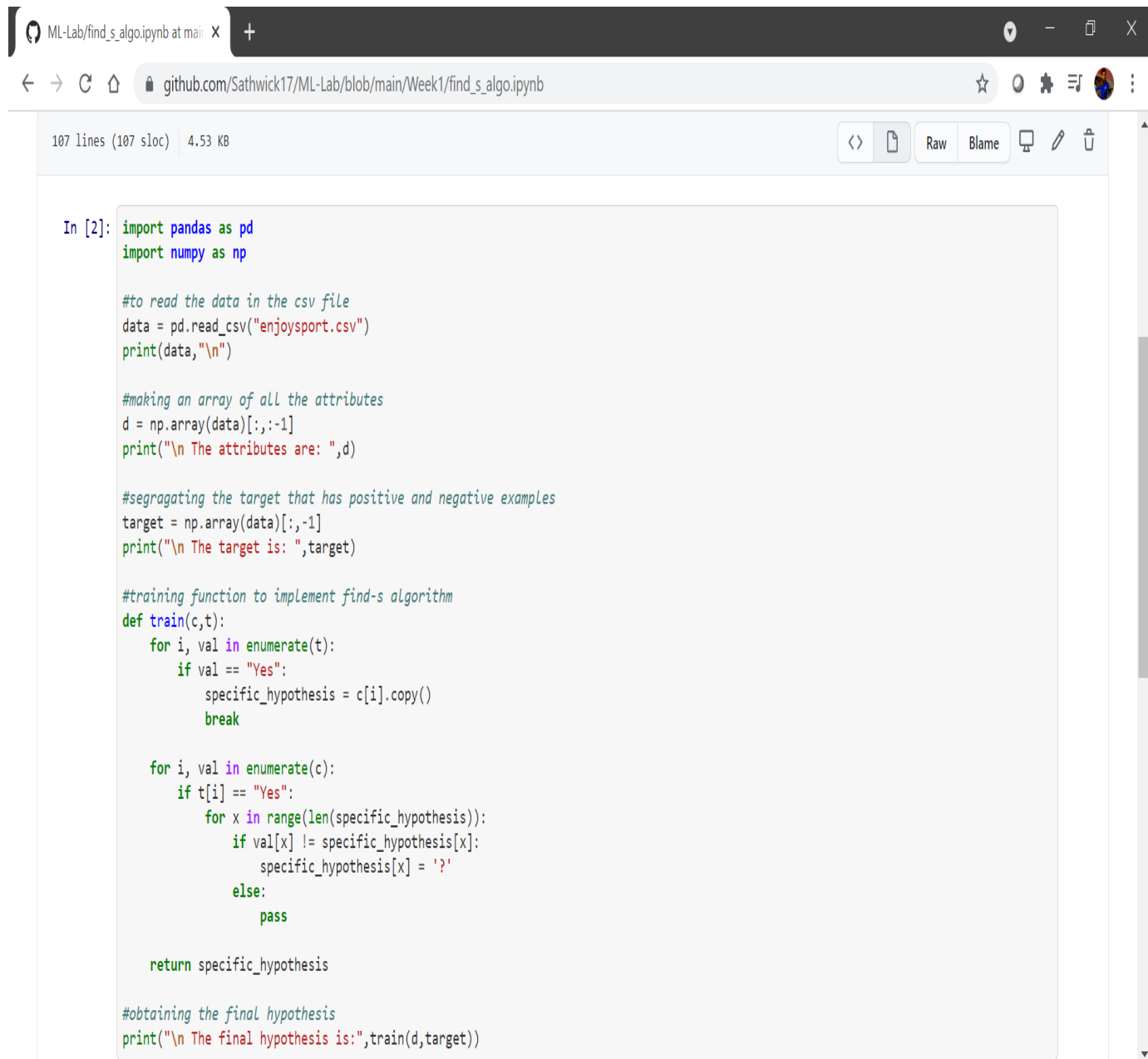


ML-Lab-01 Report

Lab-01

Program:-



```
In [2]: import pandas as pd
import numpy as np

#to read the data in the csv file
data = pd.read_csv("enjoysport.csv")
print(data, "\n")

#making an array of all the attributes
d = np.array(data)[:,:-1]
print("\n The attributes are: ",d)

#segragating the target that has positive and negative examples
target = np.array(data)[:,-1]
print("\n The target is: ",target)

#training function to implement find-s algorithm
def train(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)):
                if val[x] != specific_hypothesis[x]:
                    specific_hypothesis[x] = '?'
            else:
                pass

    return specific_hypothesis

#obtaining the final hypothesis
print("\n The final hypothesis is:",train(d,target))
```

```
ML-Lab/find_s_algo.ipynb at mai x +
github.com/Sathwick17/ML-Lab/blob/main/Week1/find_s_algo.ipynb

#obtaining the final hypothesis
print("\n The final hypothesis is:",train(d,target))

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']

-----
UnboundLocalError                                Traceback (most recent call last)
<ipython-input-2-567811ca203e> in <module>
    32
    33 #obtaining the final hypothesis
--> 34 print("\n The final hypothesis is:",train(d,target))

<ipython-input-2-567811ca203e> in train(c, t)
    29         pass
    30
--> 31     return specific_hypothesis
    32
    33 #obtaining the final hypothesis

UnboundLocalError: local variable 'specific_hypothesis' referenced before assignment

In [ ]:
```

Lab-02:-

Program:-

```
import
numpy
as np

import pandas as pd

data = pd.read_csv('../input/lab2elimination/enjoysport.csv')
concepts = np.array(data.iloc[:,0:-1])
print(concepts)
target = np.array(data.iloc[:,-1])
print(target)
def learn(concepts, target):
    specific_h = concepts[0].copy()
```

```

print("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):
    print("For Loop Starts")
    if target[i] == "yes":
        print("If instance is Positive ")
        for x in range(len(specific_h)):
            if h[x] != specific_h[x]:
                specific_h[x] = '?'
                general_h[x][x] = '?'

    if target[i] == "no":
        print("If instance is Negative ")
        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(" steps of Candidate Elimination Algorithm",i+1)
    print(specific_h)
    print(general_h)
    print("\n")
    print("\n")

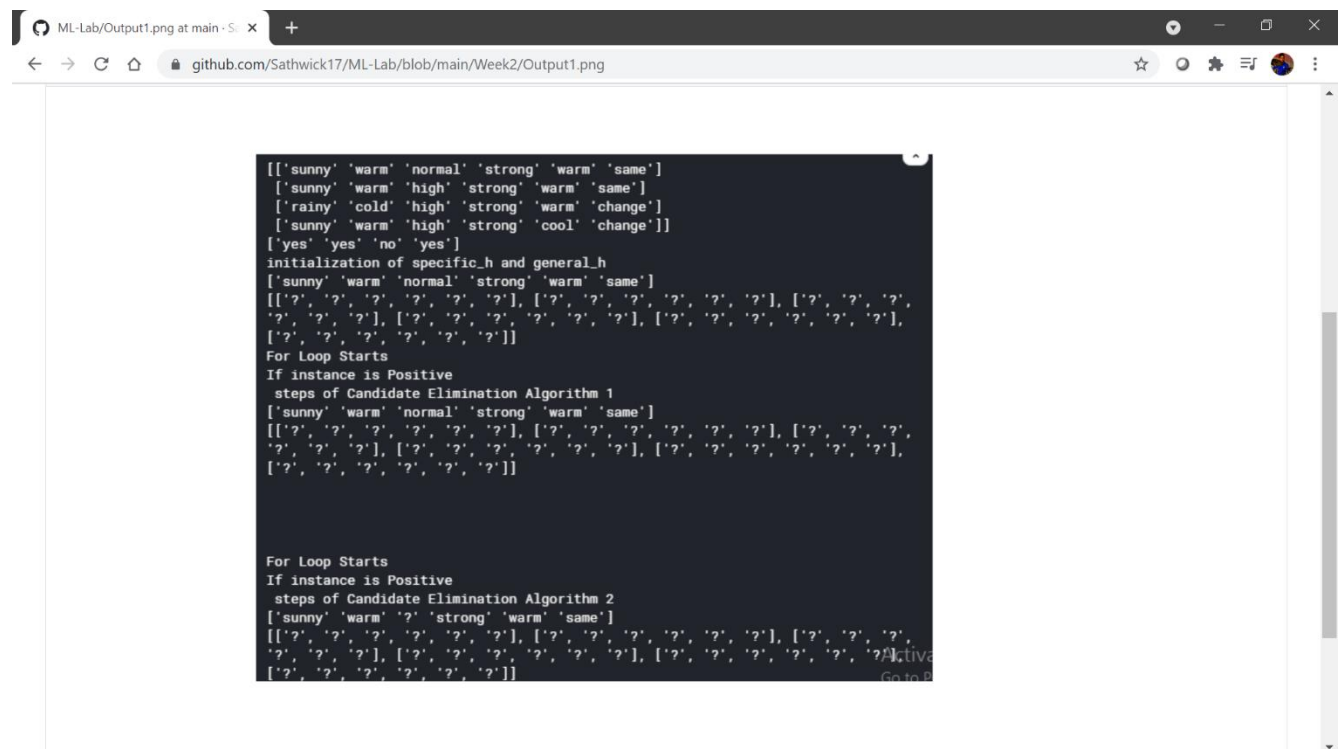
    indices = [i for i, val in enumerate(general_h) if val == ['?',
'?', '?', '?', '?']]
    for i in indices:
        general_h.remove(['?', '?', '?', '?', '?', '?'])
    return specific_h, general_h

s_final, g_final = learn(concepts, target)

print("Final Specific_h:", s_final, sep="\n")
print("Final General_h:", g_final, sep="\n")

```

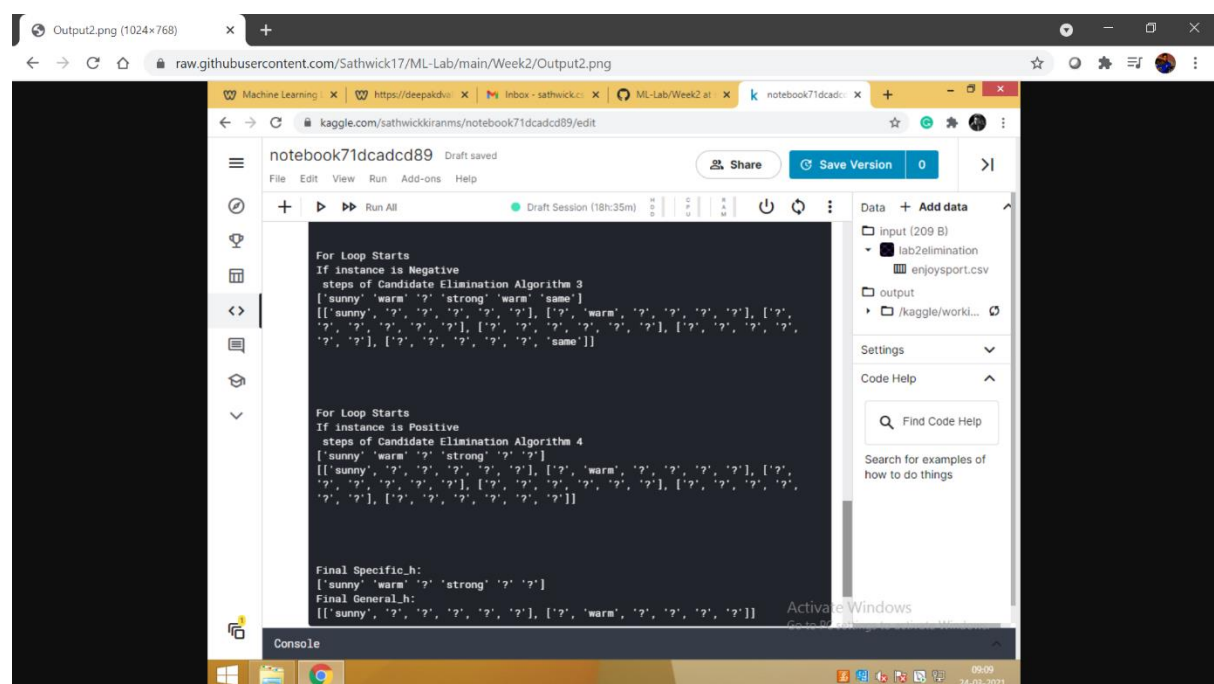
Output:-



The screenshot shows a web browser window with the address bar displaying 'github.com/Sathwick17/ML-Lab/blob/main/Week2/Output1.png'. The main content area shows a code file named 'Output1.png' with the following text:

```
[[['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'high' 'strong' 'warm' 'same']
['rainy' 'cold' 'high' 'strong' 'warm' 'change']
['sunny' 'warm' 'high' 'strong' 'cool' 'change']]
['yes' 'yes' 'no' 'yes']
initialization of specific_h and general_h
[['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?',
 '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
 ['?', '?', '?', '?', '?', '?']]
For Loop Starts
If instance is Positive
steps of Candidate Elimination Algorithm 1
[['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?',
 '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
 ['?', '?', '?', '?', '?', '?']]

For Loop Starts
If instance is Positive
steps of Candidate Elimination Algorithm 2
[['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?',
 '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
 ['?', '?', '?', '?', '?', '?']]
```



The screenshot shows a Kaggle notebook interface with the address bar displaying 'raw.githubusercontent.com/Sathwick17/ML-Lab/main/Week2/Output2.png'. The notebook is titled 'notebook71dcadcd89' and is in 'Draft saved' state. The code file 'Output2.png' contains the following text:

```
For Loop Starts
If instance is Negative
steps of Candidate Elimination Algorithm 3
[['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['sunny' 'warm' '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?',
 '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?',
 '?', '?'], ['?', '?', '?', '?', '?', '?', 'same']]

For Loop Starts
If instance is Positive
steps of Candidate Elimination Algorithm 4
[['sunny' 'warm' '?' 'strong' '?' '?']
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?',
 '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?',
 '?', '?'], ['?', '?', '?', '?', '?', '?']]

Final Specific_h:
[['sunny' 'warm' '?' 'strong' '?' '?']
Final General_h:
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]
```

Lab-03:-

Program and Output:-

```
{
  "cells": [
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      "execution_count": 3,
      "metadata": {},
      "outputs": [
        {
          "name": "stdout",
          "output_type": "stream",
          "text": [
            "\n",
            "Input Data Set is:\n",
            "      Outlook Temperature Humidity    Wind Answer\n",
            "0      sunny      hot      high     weak   no\n",
            "1      sunny      hot      high  strong   no\n",
            "2  overcast      hot      high     weak   yes\n",
            "3       rain     mild     high     weak   yes\n",
            "4       rain     cool    normal     weak   yes\n",
            "5       rain     cool    normal  strong   no\n",
            "6  overcast     cool    normal  strong   yes\n",
            "7      sunny     mild     high     weak   no\n",
            "8      sunny     cool    normal     weak   yes\n",
            "9       rain     mild    normal     weak   yes\n",
            "10     sunny     mild    normal  strong   yes\n",
            "11  overcast     mild     high  strong   yes\n",
            "12  overcast      hot    normal     weak   yes\n",
            "13     rain     mild     high  strong   no\n"
          ]
        }
      ],
      "source": [
        "\n",
        "import pandas as pd\n",
        "df = pd.read_csv('id3.csv')\n",
        "print(\"\\n\\n Input Data Set is:\\n\\n\", df)"
      ]
    },
    {
      "cell_type": "code",
      "execution_count": 4,
      "metadata": {},
      "outputs": [
        {
          "name": "stdout",
          "output_type": "stream",
          "text": [
            "Target Attribute is: Answer\n",
            "Predicting Attributes: ['Outlook', 'Temperature', 'Humidity',\n",
            "'Wind']\n"
          ]
        }
      ]
    }
  ]
}
```

```

    ]
}
],
"source": [
    "\n",
    "t = df.keys() [-1]\n",
    "print('Target Attribute is: ', t)\n",
    "# Get the attribute names from input dataset\n",
    "attribute_names = list(df.keys())\n",
    "##Remove the target attribute from the attribute names list\n",
    "attribute_names.remove(t) \n",
    "print('Predicting Attributes: ', attribute_names)"
]
},
{
    "cell_type": "code",
    "execution_count": 5,
    "metadata": {},
    "outputs": [],
    "source": [
        "#Function to calculate the entropy of collection S\n",
        "import math\n",
        "def entropy(probs): \n",
        "    return sum( [-prob*math.log(prob, 2) for prob in probs])\n",
        "\n",
        "#Function to calculate the entropy of the given Data Sets/List with
\n",
        "#respect to target attributes\n",
        "def entropy_of_list(ls,value): \n",
        "    from collections import Counter\n",
        "    cnt = Counter(x for x in ls)# Counter calculates the propotion of
class\n",
        "    print('Target attribute class count(Yes/No)=' ,dict(cnt))\n",
        "    total_instances = len(ls) \n",
        "    print(\n\"Total no of instances/records associated with {0} is:
{1}\n\".format(value,total_instances ))\n",
        "    probs = [x / total_instances for x in cnt.values()] # x means no
of YES/NO\n",
        "    print(\n\"Probability of Class {0} is:
{1:.4f}\n\".format(min(cnt),min(probs)))\n",
        "    print(\n\"Probability of Class {0} is:
{1:.4f}\n\".format(max(cnt),max(probs)))\n",
        "    return entropy(probs) # Call Entropy"
    ]
},
{
    "cell_type": "code",
    "execution_count": 6,
    "metadata": {},
    "outputs": [],
    "source": [
        "def information_gain(df, split_attribute, target_attribute,battr):\n",
        "    print(\n\"\\n\\n\\n-----Information Gain Calculation of
\n\",split_attribute, \n\" -----\\n\") \n",
        "    df_split = df.groupby(split_attribute) # group the data based on
attribute values\n",
        "    glist=[]\n",
        "    for gname,group in df_split:\n",
        "        print('Grouped Attribute Values \\n',group)\n",
        "        glist.append(gname) \n",
        "    \n",

```

```

        glist.reverse()\n",
        "    nobs = len(df.index) * 1.0    \n",
        "    df_agg1=df_split.agg({target_attribute:lambda x:entropy_of_list(x,\n",
glist.pop())})\n",
        "    df_agg2=df_split.agg({target_attribute :lambda x:len(x)/nobs})\n",
        "    \n",
        "    df_agg1.columns=['Entropy']\n",
        "    df_agg2.columns=['Proportion']\n",
        "    \n",
        "    # Calculate Information Gain:\n",
        "    new_entropy = sum( df_agg1['Entropy'] * df_agg2['Proportion'])\n",
        "    if battr != 'S':\n",
        "        old_entropy = entropy_of_list(df[target_attribute], 'S-\n",
'+df.iloc[0][df.columns.get_loc(battr)])\n",
        "    else:\n",
        "        old_entropy = entropy_of_list(df[target_attribute], battr)\n",
        "    return old_entropy - new_entropy"
    ]
},
{
    "cell_type": "code",
    "execution_count": 7,
    "metadata": {},
    "outputs": [],
    "source": [
        "def id3(df, target_attribute, attribute_names,\n",
default_class=None, default_attr='S'):\n",
        "    \n",
        "    from collections import Counter\n",
        "    cnt = Counter(x for x in df[target_attribute]) # class of YES\n",
/NO\n",
        "    \n",
        "    ## First check: Is this split of the dataset homogeneous?\n",
        "    if len(cnt) == 1:\n",
        "        return next(iter(cnt)) # next input data set, or raises\n",
StopIteration when EOF is hit.\n",
        "    \n",
        "    ## Second check: Is this split of the dataset empty? if yes,\n",
return a default value\n",
        "    elif df.empty or (not attribute_names):\n",
        "        return default_class # Return None for Empty Data Set\n",
        "    \n",
        "    ## Otherwise: This dataset is ready to be devied up!\n",
        "    else:\n",
        "        # Get Default Value for next recursive call of this\n",
function:\n",
        "        default_class = max(cnt.keys()) #No of YES and NO Class\n",
        "        # Compute the Information Gain of the attributes:\n",
        "        gainz=[]\n",
        "        for attr in attribute_names:\n",
        "            ig= information_gain(df, attr,\n",
target_attribute, default_attr)\n",
        "            gainz.append(ig)\n",
        "            print('Information gain of ', attr, ' is : ', ig)\n",
        "        \n",
        "        index_of_max = gainz.index(max(gainz)) # Index\n",
of Best Attribute\n",
        "        best_attr = attribute_names[index_of_max] # Choose\n",
Best Attribute to split on\n",
        "        print("\\n\\nAttribute with the maximum gain is: ",\n",
best_attr)\n",

```

```

        "        # Create an empty tree, to be populated in a moment\n",
        "        tree = {best_attr:{}} # Initiate the tree with best attribute
as a node \n",
        "        remaining_attribute_names =[i for i in attribute_names if i !=
best_attr]\n",
        "        \n",
        "        # Split dataset-On each split, recursively call this
algorithm.Populate the empty tree with subtrees, which\n",
        "        # are the result of the recursive call\n",
        "        for attr_val, data_subset in df.groupby(best_attr):\n",
        "            subtree = id3(data_subset,target_attribute,
remaining_attribute_names,default_class,best_attr)\n",
        "            tree[best_attr][attr_val] = subtree\n",
        "        return tree"
    ]
},
{
    "cell_type": "code",
    "execution_count": 8,
    "metadata": {},
    "outputs": [
        {
            "name": "stdout",
            "output_type": "stream",
            "text": [
                "\n",
                "\n",
                "-----Information Gain Calculation of Outlook ----- \n",
                "Grouped Attribute Values \n",
                "    Outlook Temperature Humidity    Wind Answer\n",
                "2    overcast      hot      high    weak    yes\n",
                "6    overcast      cool    normal    strong   yes\n",
                "11   overcast      mild     high    strong   yes\n",
                "12   overcast      hot     normal    weak     yes\n",
                "Grouped Attribute Values \n",
                "    Outlook Temperature Humidity    Wind Answer\n",
                "3     rain      mild     high    weak     yes\n",
                "4     rain      cool     normal    weak     yes\n",
                "5     rain      cool     normal    strong    no\n",
                "9     rain      mild     normal    weak     yes\n",
                "13    rain      mild     high    strong    no\n",
                "Grouped Attribute Values \n",
                "    Outlook Temperature Humidity    Wind Answer\n",
                "0     sunny      hot      high    weak     no\n",
                "1     sunny      hot      high    strong    no\n",
                "7     sunny      mild     high    weak     no\n",
                "8     sunny      cool     normal    weak     yes\n",
                "10    sunny      mild     normal    strong   yes\n",
                "Target attribute class count(Yes/No)= {'yes': 4}\n",
                "Total no of instances/records associated with overcast is: 4\n",
                "Probability of Class yes is: 1.0000\n",
                "Probability of Class yes is: 1.0000\n",
                "Target attribute class count(Yes/No)= {'yes': 3, 'no': 2}\n",
                "Total no of instances/records associated with rain is: 5\n",
                "Probability of Class no is: 0.4000\n",
                "Probability of Class yes is: 0.6000\n",
                "Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}\n",
                "Total no of instances/records associated with sunny is: 5\n",
                "Probability of Class no is: 0.4000\n",
                "Probability of Class yes is: 0.6000\n",
                "Target attribute class count(Yes/No)= {'no': 5, 'yes': 9}\n",
            ]
        }
    ]
}

```



```

"Total no of instances/records associated with S is: 14\n",
"Probability of Class no is: 0.3571\n",
"Probability of Class yes is: 0.6429\n",
"Information gain of Outlook is : 0.2467498197744391\n",
"\n",
"\n",
"-----Information Gain Calculation of Temperature -----\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"4        rain          cool    normal    weak    yes\n",
"5        rain          cool    normal    strong   no\n",
"6 overcast          cool    normal    strong   yes\n",
"8        sunny          cool    normal    weak    yes\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"0        sunny          hot     high     weak    no\n",
"1        sunny          hot     high     strong   no\n",
"2 overcast          hot     high     weak    yes\n",
"12 overcast          hot    normal    weak    yes\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"3        rain          mild     high     weak    yes\n",
"7        sunny          mild     high     weak    no\n",
"9        rain          mild     normal    weak    yes\n",
"10       sunny          mild     normal    strong   yes\n",
"11 overcast          mild     high     strong   yes\n",
"13       rain          mild     high     strong   no\n",
"Target attribute class count(Yes/No)= {'yes': 3, 'no': 1}\n",
"Total no of instances/records associated with cool is: 4\n",
"Probability of Class no is: 0.2500\n",
"Probability of Class yes is: 0.7500\n",
"Target attribute class count(Yes/No)= {'no': 2, 'yes': 2}\n",
"Total no of instances/records associated with hot is: 4\n",
"Probability of Class no is: 0.5000\n",
"Probability of Class yes is: 0.5000\n",
"Target attribute class count(Yes/No)= {'yes': 4, 'no': 2}\n",
"Total no of instances/records associated with mild is: 6\n",
"Probability of Class no is: 0.3333\n",
"Probability of Class yes is: 0.6667\n",
"Target attribute class count(Yes/No)= {'no': 5, 'yes': 9}\n",
"Total no of instances/records associated with S is: 14\n",
"Probability of Class no is: 0.3571\n",
"Probability of Class yes is: 0.6429\n",
"Information gain of Temperature is : 0.029222565658954647\n",
"\n",
"\n",
"-----Information Gain Calculation of Humidity -----\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"0        sunny          hot     high     weak    no\n",
"1        sunny          hot     high     strong   no\n",
"2 overcast          hot     high     weak    yes\n",
"3        rain          mild     high     weak    yes\n",
"7        sunny          mild     high     weak    no\n",
"11 overcast          mild     high     strong   yes\n",
"13       rain          mild     high     strong   no\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"4        rain          cool    normal    weak    yes\n",
"5        rain          cool    normal    strong   no\n",
"6 overcast          cool    normal    strong   yes\n",

```

```

"8      sunny      cool   normal   weak   yes\n",
"9      rain       mild   normal   weak   yes\n",
"10     sunny      mild   normal   strong  yes\n",
"12     overcast   hot    normal   weak   yes\n",
"Target attribute class count(Yes/No)= {'no': 4, 'yes': 3}\n",
"Total no of instances/records associated with high is: 7\n",
"Probability of Class no is: 0.4286\n",
"Probability of Class yes is: 0.5714\n",
"Target attribute class count(Yes/No)= {'yes': 6, 'no': 1}\n",
"Total no of instances/records associated with normal is: 7\n",
"Probability of Class no is: 0.1429\n",
"Probability of Class yes is: 0.8571\n",
"Target attribute class count(Yes/No)= {'no': 5, 'yes': 9}\n",
"Total no of instances/records associated with S is: 14\n",
"Probability of Class no is: 0.3571\n",
"Probability of Class yes is: 0.6429\n",
"Information gain of Humidity is : 0.15183550136234136\n",
"\n",
"\n",
"-----Information Gain Calculation of Wind -----\n",
"Grouped Attribute Values \n",
"      Outlook Temperature Humidity      Wind Answer\n",
"1      sunny      hot      high   strong   no\n",
"5      rain       cool     normal strong   no\n",
"6      overcast   cool     normal strong   yes\n",
"10     sunny      mild     normal strong   yes\n",
"11     overcast   mild     high   strong   yes\n",
"13     rain       mild     high   strong   no\n",
"Grouped Attribute Values \n",
"      Outlook Temperature Humidity      Wind Answer\n",
"0      sunny      hot      high   weak    no\n",
"2      overcast   hot      high   weak    yes\n",
"3      rain       mild     high   weak    yes\n",
"4      rain       cool     normal weak     yes\n",
"7      sunny      mild     high   weak    no\n",
"8      sunny      cool     normal weak     yes\n",
"9      rain       mild     normal weak     yes\n",
"12     overcast   hot      normal weak     yes\n",
"Target attribute class count(Yes/No)= {'no': 3, 'yes': 3}\n",
"Total no of instances/records associated with strong is: 6\n",
"Probability of Class no is: 0.5000\n",
"Probability of Class yes is: 0.5000\n",
"Target attribute class count(Yes/No)= {'no': 2, 'yes': 6}\n",
"Total no of instances/records associated with weak is: 8\n",
"Probability of Class no is: 0.2500\n",
"Probability of Class yes is: 0.7500\n",
"Target attribute class count(Yes/No)= {'no': 5, 'yes': 9}\n",
"Total no of instances/records associated with S is: 14\n",
"Probability of Class no is: 0.3571\n",
"Probability of Class yes is: 0.6429\n",
"Information gain of Wind is : 0.04812703040826927\n",
"\n",
"Attribute with the maximum gain is: Outlook\n",
"\n",
"\n",
"-----Information Gain Calculation of Temperature -----\n",
"Grouped Attribute Values \n",
"      Outlook Temperature Humidity      Wind Answer\n",
"4      rain       cool     normal   weak   yes\n",
"5      rain       cool     normal strong  no\n",
"Grouped Attribute Values \n",

```

```

"    Outlook Temperature Humidity    Wind Answer\n",
"3    rain      mild      high      weak      yes\n",
"9    rain      mild      normal     weak      yes\n",
"13   rain      mild      high      strong     no\n",
"Target attribute class count(Yes/No)= {'yes': 1, 'no': 1}\n",
"Total no of instances/records associated with cool is: 2\n",
"Probability of Class no is: 0.5000\n",
"Probability of Class yes is: 0.5000\n",
"Target attribute class count(Yes/No)= {'yes': 2, 'no': 1}\n",
"Total no of instances/records associated with mild is: 3\n",
"Probability of Class no is: 0.3333\n",
"Probability of Class yes is: 0.6667\n",
"Target attribute class count(Yes/No)= {'yes': 3, 'no': 2}\n",
"Total no of instances/records associated with S-rain is: 5\n",
"Probability of Class no is: 0.4000\n",
"Probability of Class yes is: 0.6000\n",
"Information gain of Temperature is : 0.01997309402197489\n",
"\n",
"\n",
"-----Information Gain Calculation of Humidity -----\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"3    rain      mild      high      weak      yes\n",
"13   rain      mild      high      strong     no\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"4    rain      cool      normal     weak      yes\n",
"5    rain      cool      normal     strong     no\n",
"9    rain      mild      normal     weak      yes\n",
"Target attribute class count(Yes/No)= {'yes': 1, 'no': 1}\n",
"Total no of instances/records associated with high is: 2\n",
"Probability of Class no is: 0.5000\n",
"Probability of Class yes is: 0.5000\n",
"Target attribute class count(Yes/No)= {'yes': 2, 'no': 1}\n",
"Total no of instances/records associated with normal is: 3\n",
"Probability of Class no is: 0.3333\n",
"Probability of Class yes is: 0.6667\n",
"Target attribute class count(Yes/No)= {'yes': 3, 'no': 2}\n",
"Total no of instances/records associated with S-rain is: 5\n",
"Probability of Class no is: 0.4000\n",
"Probability of Class yes is: 0.6000\n",
"Information gain of Humidity is : 0.01997309402197489\n",
"\n",
"\n",
"-----Information Gain Calculation of Wind -----\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"5    rain      cool      normal     strong     no\n",
"13   rain      mild      high      strong     no\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"3    rain      mild      high      weak      yes\n",
"4    rain      cool      normal     weak      yes\n",
"9    rain      mild      normal     weak      yes\n",
"Target attribute class count(Yes/No)= {'no': 2}\n",
"Total no of instances/records associated with strong is: 2\n",
"Probability of Class no is: 1.0000\n",
"Probability of Class no is: 1.0000\n",
"Target attribute class count(Yes/No)= {'yes': 3}\n",
"Total no of instances/records associated with weak is: 3\n",
"Probability of Class yes is: 1.0000\n",

```

```

"Probability of Class yes is: 1.0000\n",
"Target attribute class count(Yes/No)= {'yes': 3, 'no': 2}\n",
"Total no of instances/records associated with S-rain is: 5\n",
"Probability of Class no is: 0.4000\n",
"Probability of Class yes is: 0.6000\n",
"Information gain of Wind is : 0.9709505944546686\n",
"\n",
"Attribute with the maximum gain is: Wind\n",
"\n",
"\n",
"-----Information Gain Calculation of Temperature -----\n",
"Grouped Attribute Values \n",
"  Outlook Temperature Humidity Wind Answer\n",
"8 sunny cool normal weak yes\n",
"Grouped Attribute Values \n",
"  Outlook Temperature Humidity Wind Answer\n",
"0 sunny hot high weak no\n",
"1 sunny hot high strong no\n",
"Grouped Attribute Values \n",
"  Outlook Temperature Humidity Wind Answer\n",
"7 sunny mild high weak no\n",
"10 sunny mild normal strong yes\n",
"Target attribute class count(Yes/No)= {'yes': 1}\n",
"Total no of instances/records associated with cool is: 1\n",
"Probability of Class yes is: 1.0000\n",
"Probability of Class yes is: 1.0000\n",
"Target attribute class count(Yes/No)= {'no': 2}\n",
"Total no of instances/records associated with hot is: 2\n",
"Probability of Class no is: 1.0000\n",
"Probability of Class no is: 1.0000\n",
"Target attribute class count(Yes/No)= {'no': 1, 'yes': 1}\n",
"Total no of instances/records associated with mild is: 2\n",
"Probability of Class no is: 0.5000\n",
"Probability of Class yes is: 0.5000\n",
"Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}\n",
"Total no of instances/records associated with S-sunny is: 5\n",
"Probability of Class no is: 0.4000\n",
"Probability of Class yes is: 0.6000\n",
"Information gain of Temperature is : 0.5709505944546686\n",
"\n",
"\n",
"-----Information Gain Calculation of Humidity -----\n",
"Grouped Attribute Values \n",
"  Outlook Temperature Humidity Wind Answer\n",
"0 sunny hot high weak no\n",
"1 sunny hot high strong no\n",
"7 sunny mild high weak no\n",
"Grouped Attribute Values \n",
"  Outlook Temperature Humidity Wind Answer\n",
"8 sunny cool normal weak yes\n",
"10 sunny mild normal strong yes\n",
"Target attribute class count(Yes/No)= {'no': 3}\n",
"Total no of instances/records associated with high is: 3\n",
"Probability of Class no is: 1.0000\n",
"Probability of Class no is: 1.0000\n",
"Target attribute class count(Yes/No)= {'yes': 2}\n",
"Total no of instances/records associated with normal is: 2\n",
"Probability of Class yes is: 1.0000\n",
"Probability of Class yes is: 1.0000\n",
"Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}\n",
"Total no of instances/records associated with S-sunny is: 5\n",

```

```

"Probability of Class no is: 0.4000\n",
"Probability of Class yes is: 0.6000\n",
"Information gain of Humidity is : 0.9709505944546686\n",
"\n",
"\n",
"-----Information Gain Calculation of Wind -----\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"1    sunny        hot        high    strong    no\n",
"10   sunny        mild       normal    strong    yes\n",
"Grouped Attribute Values \n",
"    Outlook Temperature Humidity    Wind Answer\n",
"0    sunny        hot        high    weak     no\n",
"7    sunny        mild       high    weak     no\n",
"8    sunny        cool       normal   weak     yes\n",
"Target attribute class count(Yes/No)= {'no': 1, 'yes': 1}\n",
"Total no of instances/records associated with strong is: 2\n",
"Probability of Class no is: 0.5000\n",
"Probability of Class yes is: 0.5000\n",
"Target attribute class count(Yes/No)= {'no': 2, 'yes': 1}\n",
"Total no of instances/records associated with weak is: 3\n",
"Probability of Class no is: 0.3333\n",
"Probability of Class yes is: 0.6667\n",
"Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}\n",
"Total no of instances/records associated with S-sunny is: 5\n",
"Probability of Class no is: 0.4000\n",
"Probability of Class yes is: 0.6000\n",
"Information gain of Wind is : 0.01997309402197489\n",
"\n",
"Attribute with the maximum gain is: Humidity\n",
"\n",
"The Resultant Decision Tree is:\n",
"{'Outlook': {'overcast': 'yes',\n",
"            'rain': {'Wind': {'strong': 'no', 'weak': 'yes'}},\n",
"            'sunny': {'Humidity': {'high': 'no', 'normal':\n",
"yes'}}}}}\n"
]
}
],
"source": [
"\n",
"from pprint import pprint\n",
"tree = id3(df,t,attribute_names)\n",
"print("\n\nThe Resultant Decision Tree is:")\n",
"pprint(tree)"
]
},
{
"cell_type": "code",
"execution_count": null,
"metadata": {},
"outputs": [],
"source": []
}
],
"metadata": {
"kernel_spec": {
"display_name": "Python 3",
"language": "python",
"name": "python3"
},

```

```

"language_info": {
  "codemirror_mode": {
    "name": "ipython",
    "version": 3
  },
  "file_extension": ".py",
  "mimetype": "text/x-python",
  "name": "python",
  "nbconvert_exporter": "python",
  "pygments_lexer": "ipython3",
  "version": "3.8.3"
}
},
"nbformat": 4,
"nbformat_minor": 4
}

```

Lab-04:-

ML-Lab/bayesian_classifier.ipynb

github.com/Sathwick17/ML-Lab/blob/main/Week4/bayesian_classifier.ipynb

In [21]: `import pandas as pd`
`data = pd.read_csv('PlayTennis.csv')`
`data.head()`

Out[21]:

	PlayTennis	Outlook	Temperature	Humidity	Wind
0	No	Sunny	Hot	High	Weak
1	No	Sunny	Hot	High	Strong
2	Yes	Overcast	Hot	High	Weak
3	Yes	Rain	Mild	High	Weak
4	Yes	Rain	Cool	Normal	Weak

In [22]: `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']]

```
ML-Lab/bayesian_classifier.ipynb x +
github.com/Sathwick17/ML-Lab/blob/main/Week4/bayesian_classifier.ipynb

In [23]: 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 [24]: class NaiveBayesClassifier:

         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:
```

```
ML-Lab/bayesian_classifier.ipynb x +
github.com/Sathwick17/ML-Lab/blob/main/Week4/bayesian_classifier.ipynb

         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

In [25]: 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
```

```
ML-Lab/bayesian_classifier.ipynb x +
github.com/Sathwick17/ML-Lab/blob/main/Week4/bayesian_classifier.ipynb
In [25]: 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)

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
```

Lab-05:-

```
ML-Lab/Bayesian_network_heart1 x +
github.com/Sathwick17/ML-Lab/blob/main/Week5/Bayesian_network_heartDataSet%20.ipynb
In [1]: !pip install bayespy

Collecting bayespy
  Downloading https://files.pythonhosted.org/packages/84/8b/0fd9136cb49faf11606a52abf5e2598586f264c18e91fa559af35790ad24/bayespy-0.5.2
  2.tar.gz (490kB)
    |#####| 491kB 18.9MB/s
Requirement already satisfied: numpy>=1.10.0 in /usr/local/lib/python3.7/dist-packages (from bayespy) (1.19.5)
Requirement already satisfied: scipy>=0.13.0 in /usr/local/lib/python3.7/dist-packages (from bayespy) (1.4.1)
Requirement already satisfied: h5py in /usr/local/lib/python3.7/dist-packages (from bayespy) (2.10.0)
Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from h5py->bayespy) (1.15.0)
Building wheels for collected packages: bayespy
  Building wheel for bayespy (setup.py) ... done
  Created wheel for bayespy: filename=bayespy-0.5.22-cp37-none-any.whl size=379429 sha256=abfa01e5f44dc25ce5bd43dc6b6047b16ef71d00bb82
  43c186528126b814508
  Stored in directory: /root/.cache/pip/wheels/e8/32/77/9d0787cae2c5483aafada817a65e84cb086d52079ab9692be7
Successfully built bayespy
Installing collected packages: bayespy
Successfully installed bayespy-0.5.22

In [12]: import bayespy as bp
import numpy as np
import csv
!pip3 install colorama
!pip3 install colorama
from colorama import init
from colorama import Fore, Back, Style
init()

# Define Parameter Enum values
# Age
ageEnum = {'SuperSeniorCitizen': 0, 'SeniorCitizen': 1,
           'MiddleAged': 2, 'Youth': 3, 'Teen': 4}

# Gender
genderEnum = {'Male': 0, 'Female': 1}

# FamilyHistory
familyHistoryEnum = {'Yes': 0, 'No': 1}

# Diet (Calorie Intake)
dietEnum = {'High': 0, 'Medium': 1, 'Low': 2}
```



```
ML-Lab/Bayesian_network_heart | X +
github.com/Sathwick17/ML-Lab/blob/main/Week5/Bayesian_network_heartDataSet%20.ipynb
diseaseEnum = {'High': 0, 'Medium': 1, 'Low': 2}
# LifeStyle
lifeStyleEnum = {'Athlete': 0, 'Active': 1, 'Moderate': 2, 'Sedetary': 3}
# Cholesterol
cholesterolEnum = {'High': 0, 'BorderLine': 1, 'Normal': 2}
# HeartDisease
heartDiseaseEnum = {'Yes': 0, 'No': 1}

Requirement already satisfied: colorama in /usr/local/lib/python3.7/dist-packages (0.4.4)
Requirement already satisfied: colorama in /usr/local/lib/python3.7/dist-packages (0.4.4)

In [4]: import pandas as pd

In [5]: data = pd.read_csv("/content/heart_disease_data.csv")

In [7]: data = np.array(data, dtype='int8')
N = len(data)

In [8]: # Input data column assignment
p_age = bp.nodes.Dirichlet(1.0*np.ones(5))
age = bp.nodes.Categorical(p_age, plates=(N,))
age.observe(data[:, 0])

p_gender = bp.nodes.Dirichlet(1.0*np.ones(2))
gender = bp.nodes.Categorical(p_gender, plates=(N,))
gender.observe(data[:, 1])

p_familyhistory = bp.nodes.Dirichlet(1.0*np.ones(2))
familyhistory = bp.nodes.Categorical(p_familyhistory, plates=(N,))
familyhistory.observe(data[:, 2])

p_diet = bp.nodes.Dirichlet(1.0*np.ones(3))
diet = bp.nodes.Categorical(p_diet, plates=(N,))
diet.observe(data[:, 3])

p_lifestyle = bp.nodes.Dirichlet(1.0*np.ones(4))
lifestyle = bp.nodes.Categorical(p_lifestyle, plates=(N,))

p_lifestyle = bp.nodes.Dirichlet(1.0*np.ones(4))
lifestyle = bp.nodes.Categorical(p_lifestyle, plates=(N,))
lifestyle.observe(data[:, 4])

p_cholesterol = bp.nodes.Dirichlet(1.0*np.ones(3))
cholesterol = bp.nodes.Categorical(p_cholesterol, plates=(N,))
cholesterol.observe(data[:, 5])

In [9]: # Prepare nodes and establish edges
# np.ones(2) -> HeartDisease has 2 options Yes/No
# plates(5, 2, 2, 3, 4, 3) -> corresponds to options present for domain values
p_heartdisease = bp.nodes.Dirichlet(np.ones(2), plates=(5, 2, 2, 3, 4, 3))
heartdisease = bp.nodes.MultiMixture(
    [age, gender, familyhistory, diet, lifestyle, cholesterol], bp.nodes.Categorical, p_heartdisease)
heartdisease.observe(data[:, 6])
p_heartdisease.update()

In [11]: #print("Sample Probability")
#print("Probability(HeartDisease|Age=SuperSeniorCitizen, Gender=Female, FamilyHistory=Yes, DietIntake=Medium, LifeStyle=Sedetary, Cholesterol=High)")
#print(bp.nodes.MultiMixture([ageEnum['SuperSeniorCitizen'], genderEnum['Female'], familyHistoryEnum['Yes'], dietEnum['Medium'], LifeStyleEnum['Sedetary'], cholesterolEnum['High']], bp.nodes.Categorical, p_heartdisease).get_moments()[0][heartDiseaseEnum['Yes']])

# Interactive Test
m = 0
while m == 0:
    print("\n")
    res = bp.nodes.MultiMixture([int(input('Enter Age: ' + str(ageEnum))), int(input('Enter Gender: ' + str(genderEnum))), int(input('Enter FamilyHistory: ' + str(familyHistoryEnum))), int(input('Enter dietEnum: ' + str(dietEnum))), int(input('Enter LifeStyle: ' + str(lifeStyleEnum))), int(input('Enter Cholesterol: ' + str(cholesterolEnum))), bp.nodes.Categorical, p_heartdisease).get_moments()[0][heartDiseaseEnum['Yes']]
    print("Probability(HeartDisease) = " + str(res))

# print(Style.RESET_ALL)
m = int(input("Enter for Continue:0, Exit :1 "))
```

ML-Lab/Bayesian_network_heart

github.com/Sathwick17/ML-Lab/blob/main/Week5/Bayesian_network_heartDataSet%20.ipynb

```
print("\n")
res = bp.nodes.MultiMixture([int(input('Enter Age: ' + str(ageEnum))), int(input('Enter Gender: ' + str(genderEnum))), int(input('Enter FamilyHistory: ' + str(familyHistoryEnum))), int(input('Enter dietEnum: ' + str(dietEnum))), int(input('Enter LifeStyle: ' + str(lifeStyleEnum))), int(input('Enter Cholesterol: ' + str(cholesterolEnum))), b
p.nodes.Categorical, p_heartdisease).get_moments()[0][heartDiseaseEnum['Yes']]
print("Probability(HeartDisease) = " + str(res))

# print(Style.RESET_ALL)
m = int(input("Enter for Continue:0, Exit :1 "))

Enter Age: {'SuperSeniorCitizen': 0, 'SeniorCitizen': 1, 'MiddleAged': 2, 'Youth': 3, 'Teen': 4}4
Enter Gender: {'Male': 0, 'Female': 1}0
Enter FamilyHistory: {'Yes': 0, 'No': 1}0
Enter dietEnum: {'High': 0, 'Medium': 1, 'Low': 2}1
Enter LifeStyle: {'Athlete': 0, 'Active': 1, 'Moderate': 2, 'Sedetary': 3}1
Enter Cholesterol: {'High': 0, 'BorderLine': 1, 'Normal': 2}2
Probability(HeartDisease) = 0.5
Enter for Continue:0, Exit :1 0

Enter Age: {'SuperSeniorCitizen': 0, 'SeniorCitizen': 1, 'MiddleAged': 2, 'Youth': 3, 'Teen': 4}1
Enter Gender: {'Male': 0, 'Female': 1}0
Enter FamilyHistory: {'Yes': 0, 'No': 1}1
Enter dietEnum: {'High': 0, 'Medium': 1, 'Low': 2}0
Enter LifeStyle: {'Athlete': 0, 'Active': 1, 'Moderate': 2, 'Sedetary': 3}2
Enter Cholesterol: {'High': 0, 'BorderLine': 1, 'Normal': 2}0
Probability(HeartDisease) = 0.5
Enter for Continue:0, Exit :1 1
```

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