

Output :

data							
city	airtemp	humidity	wind	water	forecast	enjoyed	
0 rainy	warm	normal	strong	warm	None	yes	
1 rainy	warm	high	strong	warm	None	yes	
2 rainy	cold	high	strong	warm	change	no	
3 rainy	warm	high	strong	cold	change	yes	

```

concepts
array(['rainy', 'warm', 'normal', 'strong', 'warm', 'None'],
      ['rainy', 'warm', 'high', 'strong', 'warm', 'None'],
      ['rainy', 'cold', 'high', 'strong', 'warm', 'change'],
      ['rainy', 'warm', 'high', 'strong', 'cold', 'change']),
dtype = object)

```

```

target
array(['yes', 'yes', 'no', 'yes'], dtype = object)

```

```

print(train(concepts, target))

```

```

['rainy', 'warm', '?', 'strong', '?', '?']

```

Implement and demonstrate the FIND-S algorithm for finding the most specific hypothesis based on a given set of training data samples. Read the data from a .csv file (enjoySport Dataset)

```

import pandas as pd
import numpy as np
data = pd.read_csv('D:\Users\enjoySport.csv')
data
concepts = np.array(data)[0:-1]
concepts
target = np.array(data)[-1]
target
def train(con, tar):
    for i, val in enumerate(con):
        if val == 'yes':
            specific_h = con[i].copy()
            break
    for i, val in enumerate(con):
        if con[i] == 'yes':
            for x in range(len(specific_h)):
                if val[x] != specific_h[x]:
                    specific_h[x] = '?'
            else:
                pass
    return specific_h
print(train(concepts, target))

```

Pushed:

initialization of specific- t

['strong', 'warm', 'normal', 'warm', 'heavy', 'same']

initialization of general-h

Investigation of general-
 $[C^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ}]$, $[C^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ}]$, $[C^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ}]$,
 $[C^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ}]$, $[C^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ}]$, $[C^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ} \dot{P}^{\circ}]$

7 instance is positive

Step 1

['sɜ:ni', 'wɔ:ɪn', 'nɒlɪd', 'ʌlɒŋ', 'wɔ:ɪn', 'beɪm']

[illegible]

[P P P P P P] [P P P P P P] [P P P P P P]

If instance is feasible

Auf 2

['heavy', 'warm', ('P', 'strong', 'warm', 'hand')]

[illegible][illegible]

If instance is Negative

2540 3

['Runny', 'warm', (p), 'strong', 'warm', 'same']

[illegible]

0
[p' p' p' p' p' p' p'] [p p p p p p p] [p p p p p p p] [p p p p p p p]

If instance is Positive

4. Self

['ju:ni', 'we:n', 'p', 'strɒŋ', 'p', 'p', 'p']

[illegible]

$\frac{P}{Q}$, $\left[\frac{P}{Q}\right]$

८

For a given set of training data examples stored in a .csv file (e.g. sports dataset), implement a descriptive the candidate - Elimination algorithm the output a description of the set of all hypotheses consistent with the training examples.

infant pandas as pd
infant nursery as np

```
data = pd.read_csv("c:\users\enjoysoft.csv")
```

concept = np.array (data.iloc[:, 0:-1])

$$\text{target} = \text{np.array}(\text{data_iloc}[:, -1])$$

def learn_concepts, target):

Specific $h = \text{concepts}[0].\text{category}$

print "initialization of specific h \ n", specific h)

general $\mathbf{h} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ for i in range (last species h))

for i in range(len(specific)):

point (initialization of general run, general run)

12. b. in enumerate (concepts):

if target == "yes" :

point (If instance is false")

for x in $\text{range}(\text{len}(\text{phrase} - th))$ is

$$h[x] := \Delta \text{ hedge} - h[x]$$
$$\Delta H_{\text{ref}} - nRT = -j$$
$$g_{\text{general}}(x)(x) = i$$

Final specific k:
['strong', 'warm', 'p', 'strong', 'p', 'p']

Final general k:
[['strong', 'p', 'p', 'p', 'p', 'p'], ['p', 'warm', 'p', 'p', 'p', 'p']]



if target[1] == "no":

print("If instance is Negative")

for i in range(len(specific_k)):

if h[i] != specific_k[i]:

general_k[i] = specific_k[i]

else:

general_h[i][2] = 'p'

print("Step 1", print(i+1))

print(specific_k)

print(general_k)

print("\n")

indices = [i for i, val in enumerate(general_k) if val ==

['p', 'p', 'p', 'p', 'p', 'p']]

for i in indices:

general = remove(['p', 'p', 'p', 'p', 'p', 'p'])

return specific_k, general_k

def final, general = learn(concepts, target)

print("Final specific k:", specific_k, hyp = "\n")

print("Final general k:", general, sep = "\n")

outlook	Temperature	Humidity	Windy	Play-Tennis
0 Sunny	Hot	High	False	No
1 Sunny	Hot	High	True	No
2 overcast	Hot	High	False	Yes
3 Rainy	Mild	High	False	Yes
4 Rainy	Cool	Normal	True	No
5 Rainy	Cool	Normal	True	Yes
6 overcast	Cool	Normal	True	No
7 Sunny	Mild	High	False	Yes
8 Sunny	Cool	Normal	False	Yes
9 Rainy	Mild	Normal	True	Yes
10 Sunny	Mild	High	True	Yes
11 overcast	Mild	Normal	False	Yes
12 overcast	Hot	High	True	No
13 Rainy	Mild	High	True	No

outlook	Temperature	Humidity	Windy	Play-Tennis
0 Sunny	Hot	High	False	No
1 Sunny	Hot	High	True	No
2 overcast	Hot	High	False	Yes
3 Rainy	Mild	High	False	Yes
4 Rainy	Cool	Normal	False	Yes

outlook	Temperature	Humidity	Windy	Play-Tennis
9 Rainy	Mild	Normal	False	Yes
10 Sunny	Mild	Normal	True	Yes
11 overcast	Mild	High	True	Yes
12 overcast	Hot	Normal	False	Yes
13 Rainy	Mild	High	True	No

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 & apply the knowledge to classify a new sample (Play Tennis Dataset)

import numpy as np
import pandas as pd
from sklearn import metrics

df = pd.read_csv("C:\Users\PlayTennis.csv")
X = df[['outlook', 'Temperature', 'Humidity', 'Windy']]
y = df['Play-Tennis']

X.shape
X.head()
X.tail()
X.describe()

from sklearn import preprocessing
X = preprocessing.LabelEncoder()
X = X.fit_transform(X)

feature_cols = ['outlook', 'Temperature', 'Humidity', 'Windy']
X = df[feature_cols]
y = df['Play-Tennis']

outlook	Temperature	Humidity	windy	Play-Tennis
cool	14	14	14	14
warm	3	2	2	2
top	Sunny	mild	False	Yes
big	5	6	8	9

outlook	Temperature	Humidity	windy	Play-Tennis
0	2	1	0	0
1	2	1	1	0
2	0	0	0	1
3	1	2	0	1
4	1	0	0	1
5	1	0	1	0
6	0	1	1	1
7	2	0	0	0
8	2	1	0	1
9	1	2	0	1
10	2	1	1	1
11	0	2	1	1
12	0	1	0	1
13	1	2	1	0

Accuracy = 0.8

Actual	Predicted
13	0
0	0
4	1
9	1
2	1

from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.30)

from sklearn.tree import DecisionTreeClassifier
classifier = DecisionTreeClassifier(criterion = "entropy", random_state=0)
classifier.fit(x_train, y_train)

y_pred = classifier.predict(x_test)

from sklearn.metrics import accuracy_score
print("Accuracy:", accuracy_score(y_test, y_pred))

data = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})

from sklearn.metrics import classification_report, confusion_matrix
print("Confusion matrix (y_test, y_pred)")

print(classification_report(y_test, y_pred))

[11]
[03]

	freis on	recall	fl-note	support
0	1.00	0.50	0.67	2
1	0.75	1.00	0.86	3
avg / total	0.85	0.80	0.78	5

	Outlook	Temperature	Humidity	Windy	Play-Tennis
0	Sunny	Hot	High	False	No
1	Sunny	Hot	High	True	Yes
2	Overcast	Hot	High	False	Yes
3	Rainy	Mild	High	False	Yes
4	Rainy	Cool	Normal	True	No
5	Rainy	Cool	Normal	True	Yes
6	Overcast	Cool	Normal	True	No
7	Sunny	Mild	High	False	Yes
8	Sunny	Cool	Normal	False	Yes
9	Rainy	Mild	Normal	False	Yes
10	Sunny	Mild	Normal	True	Yes
11	Overcast	Mild	High	True	Yes
12	Overcast	Hot	Normal	False	Yes
13	Rainy	Mild	High	True	No

14

[14, 5]

	Outlook	Temperature	Humidity	Windy	Play-Tennis
0	Sunny	Hot	High	False	No
1	Sunny	Hot	High	True	No
2	Overcast	Hot	High	False	Yes
3	Rainy	Mild	High	False	Yes
4	Rainy	Cool	Normal	False	Yes

	Outlook	Temperature	Humidity	Windy	Play-Tennis
9	Rainy	Mild	Normal	False	Yes
10	Sunny	Mild	Normal	True	Yes
11	Overcast	Mild	High	True	Yes
12	Overcast	Hot	Normal	False	Yes
13	Rainy	Mild	High	True	No

4

Write a program to demonstrate the working of the decision tree based CART algorithm. (PlayTennis dataset)

Import numpy as np
import pandas as pd
from sklearn import metrics

dt = pd.read_csv("C:\Users\PlayTennis.csv")

value = ['Outlook', 'Temperature', 'Humidity', 'Windy']

dt

dt.head()

dt.shape

dt.head()

dt.tail()

dt.describe()

from sklearn import preprocessing

String to int = preprocessing.LabelEncoder()

dt = dt.apply(String to int.fit_transform)

dt

feature_cols = ['Outlook', 'Temperature', 'Humidity', 'Windy']

x = dt[feature_cols]

y = dt['Play-Tennis']

from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.3)

outlook	temperature	humidity	windy	play-tennis
cool	14	14	14	14
warm	3	2	2	2
hot	sunny	high	false	yes
few	5	7	8	9

outlook	temperature	humidity	windy	play-tennis
0	2	1	0	0
1	2	1	0	1
2	0	1	0	1
3	1	2	0	1
4	1	0	1	1
5	1	0	1	0
6	0	1	1	1
7	2	2	0	0
8	2	0	0	1
9	1	2	0	1
10	2	2	1	1
11	0	2	1	1
12	0	1	0	1
13	1	2	1	0

accuracy : 4.0

	Actual	Predicted
10	1	1
5	0	0
9	1	1
6	1	1
12	1	1

from sklearn.tree import DecisionTreeClassifier
classifier = DecisionTreeClassifier(criterion = 'gini', random_state = 100)
classifier.fit(x_train, y_train)

y_pred = classifier.predict(x_test)

from sklearn.metrics import accuracy_score
print("accuracy is", metrics.accuracy_score(y_test, y_pred))

data = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})

from sklearn.metrics import classification_report, confusion_matrix
print(confusion_matrix(y_test, y_pred))

print(classification_report(y_test, y_pred))

[C1 0
0 4 7]

	fraction	ac cell	fluoride	hydro
0	1.00	1.00	1.00	1
1	1.00	1.00	1.00	4
avg/total	1.00	1.00	1.00	5

14

(14, 5)

	outlook	temperature	humidity	windy
0	Sunny	Hot	High	True
1	Sunny	Hot	High	False
2	Overcast	Hot	High	False
3	Rainy	Mild	High	False
4	Rainy	Cool	Normal	True
5	Rainy	Cool	Normal	True
6	Overcast	Cool	Normal	False
7	Sunny	Mild	High	False
8	Sunny	Cool	Normal	False
9	Rainy	Mild	Normal	True
10	Sunny	Mild	Normal	True
11	Overcast	Mild	High	True
12	Overcast	Hot	Normal	False
13	Rainy	Mild	High	True

0 25
1 30
2 46
3 45
4 52
5 23
6 43
7 35
8 38
9 46
10 48
11 52
12 44
13 30

Name: Gof Players, dtype: int64

5

Write a program to demonstrate Decision tree algorithm on a given dataset.

import numpy as np
import pandas as pd
from sklearn import metrics

df = pd.read_csv("C:\Users\harish\dataReg.csv")

len(df)

df.shape

X = df.drop("Gof Players", axis = 1)

y = df["Gof Players"]

X

y

from sklearn.preprocessing import LabelEncoder
from sklearn import preprocessing, LabelEncoder

Xstring = X.tostring() # preprocessing, LabelEncoder()

X = X.apply(Xstring.tostring(), fit = transform)

X

from sklearn.tree import DecisionTreeRegressor

reg = DecisionTreeRegressor()

reg = reg.fit(X, y)

	Outlook	Temperature	Humidity	Windy
0	2	1	0	1
1	2	1	0	0
2	0	1	0	0
3	1	2	1	0
4	1	0	1	1
5	1	0	1	1
6	0	0	1	0
7	2	0	1	0
8	2	0	1	0
9	1	2	1	1
10	2	2	0	1
11	0	2	0	0
12	0	1	1	0
13	1	2	0	1

Result is : [30.]

Result is : [25.]

```

y_pred = reg.predict([[2,1,0,1]])
print("Result is :", y_pred)

y_pred = reg.predict([[2,1,0,0]])
print("Result is :", y_pred)

```

training completed in 1 epoch, bias = -1.0
Final weights: $w_1 = 1.2$, $w_2 = 1.1$, bias = -1.0
Testing function for AND gate:
Input: [0 0], output: 0, Expected: 0
Input: [0 1], output: 0, Expected: 0
Input: [1 0], output: 0, Expected: 0
Input: [1 1], output: 1, Expected: 1

6

Implement a Perceptron algorithm for AND logic gate with 2-bit binary input. Test for different input parameters

Input mapping as np

Inputs = np.array([

[0, 0],

[1, 0],

[1, 1]

])

$w_1, w_2 = 1.2, 0.6$

bias = -1.0

threshold = 1

learning_rate = 0.5

def activation_function(net_input):

return 1 if net_input >= threshold else 0

epochs = 0

while True:

error_count = 0

for i in range(len(inputs)):

net_input = $w_1 \times \text{inputs}[i][0] + w_2 \times \text{inputs}[i][1] + \text{bias}$

output = activation_function(net_input)

error = expected_output[i] - output

if error != 0:

$w_1 += \text{learning_rate} \times \text{error} \times \text{inputs}[i][0]$

$w_2 += \text{learning_rate} \times \text{error} \times \text{inputs}[i][1]$

bias = learning rate * error

error count = 1

epochs + 1

if error count == 0 :

break

print ("Training completed in {epochs} epochs")

print ("Final weights : w1 = {w1}, w2 = {w2}, bias = {bias}")

print ("Testing function for AND gate :")

for i in range(len(inputs)) :

net_input = w1 * inputs[i][0] + w2 * inputs[i][1] + bias

output = activation_function(net_input)

print ("input : {inputs[i][0]}, output : {output}")

Expected : {expected_output[i]}")