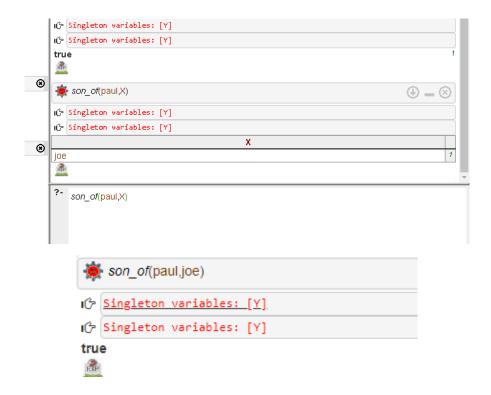
| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

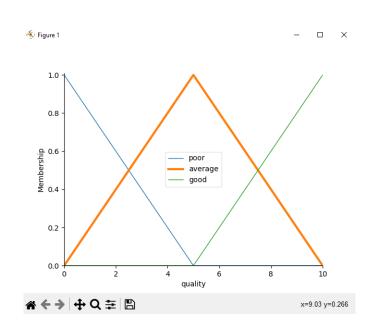
father(joe,paul) father(joe,mary) father(joe,hope) mother(jane,paul) mother(jane,mary) mother(jane,hope) male(paul) male(joe) male(raphl) male(X):-father(X,Y) female(mary) female(jane) female(hope) female(X):-mother(X,Y) son_of(X,Y):- father(Y,X),male(X) son_of(X,Y):- mother(Y,X),male(X) daughter_of(X,Y):- father(Y,X),female(X) daughter_of(X,Y):- mother(Y,X),female(X) $sibling_of(X,Y):-father(Z,X),father(Z,Y),X\=Y$ $sibling_of(X,Y):=mother(Z,X),mother(Z,Y),X\=Y$

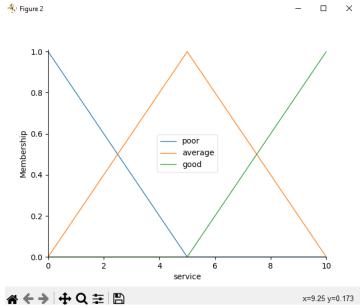


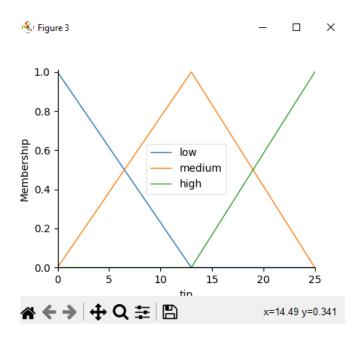
| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

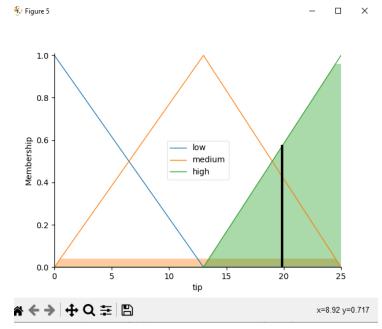
| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
import numpy as np
import skfuzzy as fuzz
from skfuzzy import control as ctrl
quality = ctrl.Antecedent(np.arange(0, 11, 1), 'quality')
service = ctrl.Antecedent(np.arange(0, 11, 1), 'service')
tip = ctrl.Consequent(np.arange(0, 26, 1), 'tip')
quality.automf(3)
service.automf(3)
tip['low'] = fuzz.trimf(tip.universe, [0, 0, 13])
tip['medium'] = fuzz.trimf(tip.universe, [0, 13, 25])
tip['high'] = fuzz.trimf(tip.universe, [13, 25, 25])
quality['average'].view()
service.view()
tip.view()
rule1 = ctrl.Rule(quality['poor'] | service['poor'], tip['low'])
rule2 = ctrl.Rule(service['average'], tip['medium'])
rule3 = ctrl.Rule(service['good'] | quality['good'], tip['high'])
rule1.view()
tipping ctrl = ctrl.ControlSystem([rule1, rule2, rule3])
tipping = ctrl.ControlSystemSimulation(tipping ctrl)
tipping.input['quality'] = 6.5
tipping.input['service'] = 9.8
tipping.compute()
print(tipping.output['tip'])
tip.view(sim=tipping)
```









| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

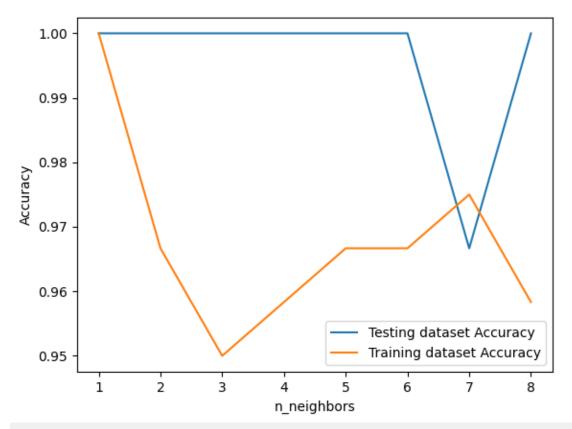
| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
# Import necessary modules
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model selection import train test split
from sklearn.datasets import load_iris
import numpy as np
import matplotlib.pyplot as plt
irisData = load iris()
# Create feature and target arrays
X = irisData.data
y = irisData.target
# Split into training and test set
X_train, X_test, y_train, y_test = train_test_split(
X, y, test_size = 0.2, random_state=42)
neighbors = np.arange(1, 9)
train accuracy = np.empty(len(neighbors))
test_accuracy = np.empty(len(neighbors))
# Loop over K values
for i, k in enumerate(neighbors):
knn = KNeighborsClassifier(n neighbors=k)
knn.fit(X_train, y_train)
# Compute training and test data accuracy
train_accuracy[i] = knn.score(X_train, y_train)
test accuracy[i] = knn.score(X test, y test)
# Generate plot
plt.plot(neighbors, test_accuracy, label = 'Testing dataset Accuracy')
plt.plot(neighbors, train_accuracy, label = 'Training dataset Accuracy')
plt.legend()
plt.xlabel('n neighbors')
```

plt.ylabel('Accuracy')

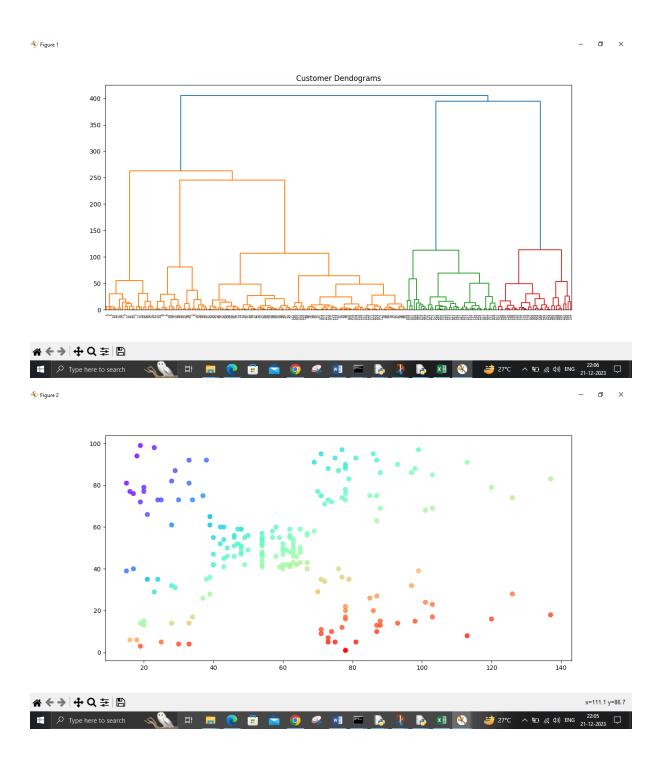
plt.show()





| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
customer_data = pd.read_csv("C:\Users\Sakib\Downloads\Mall_Customers.csv")
customer_data.shape
customer data.head()
data = customer data.iloc[:, 3:5].values
import scipy.cluster.hierarchy as shc
plt.figure(figsize=(10, 7))
plt.title("Customer Dendograms")
dend = shc.dendrogram(shc.linkage(data, method='ward'))
from sklearn.cluster import AgglomerativeClustering
cluster = AgglomerativeClustering()
cluster.fit_predict(data)
plt.figure(figsize=(10, 7))
plt.scatter(data[:,0], data[:,1], c=np.arctan2(data[:,0], data[:,1]), cmap='rainbow', s=50,
alpha=0.8)
plt.show()
```



| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

main.py

from numpy import where

from sklearn.datasets import make_classification

from matplotlib import pyplot

x,y = make_classification(n_samples=1000, n_features=2, n_informative=2, n_redundant=0,
n_clusters_per_class=1, random_state=4)

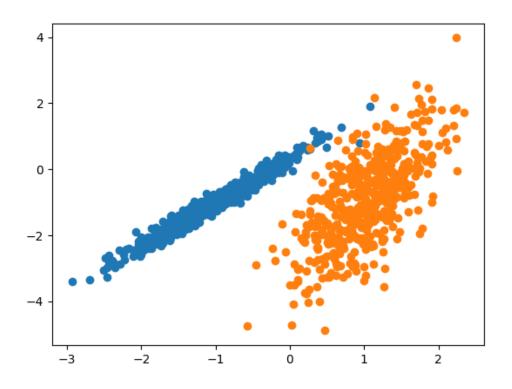
for class_value in range(2):

row_ix = where(y == class_value)

pyplot.scatter(x[row_ix, 0], x[row_ix, 1])

pyplot.show()





| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
#Import scikit-learn dataset library
from sklearn import datasets
#Import svm model
from sklearn import svm
# Import train_test_split function
from sklearn.model selection import train test split
#Import scikit-learn metrics module for accuracy calculation
from sklearn import metrics
#Load dataset
cancer = datasets.load_breast_cancer()
# print the names of the 13 features
print("Features: ", cancer.feature names)
# print the label type of cancer('malignant' 'benign')
print("Labels: ", cancer.target_names)
# print data(feature)shape
cancer.data.shape
# print the cancer data features (top 5 records)
print(cancer.data[0:5])
# print the cancer labels (0:malignant, 1:benign)
print(cancer.target)
# Split dataset into training set and test set
X_train, X_test, y_train, y_test = train_test_split(cancer.data, cancer.target,
test size=0.3,random state=109) # 70% training and 30% test
#Create a sym Classifier
clf = svm.SVC(kernel='linear') # Linear Kernel
#Train the model using the training sets
clf.fit(X train, y train)
#Predict the response for test dataset
```

```
y_pred = clf.predict(X_test)
# Model Accuracy: how often is the classifier correct?
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
# Model Precision: what percentage of positive tuples are labeled as such?
print("Precision:",metrics.precision_score(y_test, y_pred))
# Model Recall: what percentage of positive tuples are labelled as such?
print("Recall:",metrics.recall_score(y_test, y_pred))
print("______done by Muskan Momin_")
```

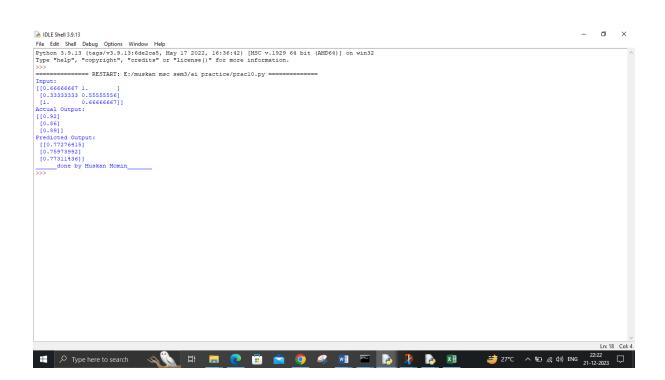


| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
import numpy as np
X = np.array(([2, 9], [1, 5], [3, 6]), dtype=float) # two inputs [sleep,study]
y = np.array(([92], [86], [89]), dtype=float) # one output [Expected % in Exams]
X = X / np.amax(X, axis=0) # maximum of X array longitudinally
y = y / 100
# Sigmoid Function
def sigmoid(x):
  return 1/(1 + np.exp(-x))
# Derivative of Sigmoid Function
def derivatives sigmoid(x):
  return x * (1 - x)
# Variable initialization
epoch = 5000 # Setting training iterations
Ir = 0.1 # Setting learning rate
inputlayer neurons = 2 # number of features in data set
hiddenlayer_neurons = 3 # number of hidden layers neurons
output neurons = 1 # number of neurons at output layer
# weight and bias initialization
wh = np.random.uniform(size=(inputlayer neurons, hiddenlayer neurons)) # weight of the
link from input node to hidden node
bh = np.random.uniform(size=(1, hiddenlayer neurons)) # bias of the link from input node
to hidden node
wout = np.random.uniform(size=(hiddenlayer neurons, output neurons)) # weight of the
link from hidden node to output node
bout = np.random.uniform(size=(1, output neurons)) # bias of the link from hidden node to
output node
# draws a random range of numbers uniformly of dim x*y
for i in range(epoch):
  # Forward Propogation
```

```
hinp1 = np.dot(X, wh)
  hinp = hinp1 + bh
 hlayer_act = sigmoid(hinp)
  outinp1 = np.dot(hlayer act, wout)
  outinp = outinp1 + bout
  output = sigmoid(outinp)
  # Backpropagation
  EO = y - output
  outgrad = derivatives sigmoid(output)
  d output = EO * outgrad
  EH = d_output.dot(wout.T)
  # how much hidden layer weights contributed to error
  hiddengrad = derivatives_sigmoid(hlayer_act)
  d hiddenlayer = EH * hiddengrad
# dotproduct of nextlayererror and currentlayerop
wout += hlayer_act.T.dot(d_output) * Ir
wh += X.T.dot(d hiddenlayer) * Ir
print("Input: \n" + str(X))
print("Actual Output: \n" + str(y))
print("Predicted Output: \n", output)
print("_____done by Muskan Momin_____")
```



| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
import random
# Number of individuals in each generation
POPULATION SIZE = 100
# Valid genes
GENES = "'abcdefghijklmnopqrstuvwxyzABCDEFGHIJKLMNOP
QRSTUVWXYZ 1234567890, .-;: !"#%&/()=?@${[]}""
# Target string to be generated
TARGET = "Muskan Momin"
class Individual(object):
       111
       Class representing individual in population
       111
       def __init__(self, chromosome):
              self.chromosome = chromosome
              self.fitness = self.cal fitness()
       @classmethod
       def mutated genes(self):
              create random genes for mutation
              global GENES
              gene = random.choice(GENES)
              return gene
       @classmethod
       def create_gnome(self):
              111
              create chromosome or string of genes
              111
```

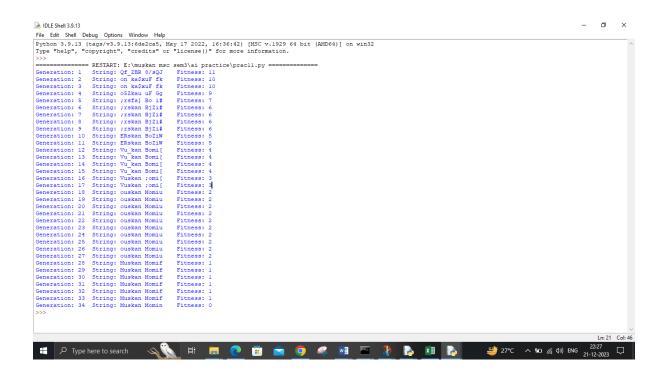
```
global TARGET
      gnome len = len(TARGET)
      return [self.mutated_genes() for _ in range(gnome_len)]
def mate(self, par2):
      ш
      Perform mating and produce new offspring
      # chromosome for offspring
      child chromosome = []
      for gp1, gp2 in zip(self.chromosome, par2.chromosome):
              # random probability
              prob = random.random()
              # if prob is less than 0.45, insert gene
              # from parent 1
              if prob < 0.45:
                     child_chromosome.append(gp1)
              # if prob is between 0.45 and 0.90, insert
              # gene from parent 2
              elif prob < 0.90:
                     child_chromosome.append(gp2)
              # otherwise insert random gene(mutate),
              # for maintaining diversity
              else:
                     child_chromosome.append(self.mutated_genes())
      # create new Individual(offspring) using
      # generated chromosome for offspring
      return Individual(child chromosome)
def cal_fitness(self):
```

111

```
Calculate fittness score, it is the number of
              characters in string which differ from target
              string.
              111
              global TARGET
              fitness = 0
              for gs, gt in zip(self.chromosome, TARGET):
                      if gs != gt: fitness+= 1
              return fitness
# Driver code
def main():
       global POPULATION_SIZE
       #current generation
       generation = 1
       found = False
       population = []
       # create initial population
       for _ in range(POPULATION_SIZE):
                             gnome = Individual.create gnome()
                             population.append(Individual(gnome))
       while not found:
              # sort the population in increasing order of fitness score
              population = sorted(population, key = lambda x:x.fitness)
              # if the individual having lowest fitness score ie.
              # 0 then we know that we have reached to the target
              # and break the loop
              if population[0].fitness <= 0:
                      found = True
                      break
```

```
new generation = []
              # Perform Elitism, that mean 10% of fittest population
              # goes to the next generation
              s = int((10*POPULATION SIZE)/100)
              new_generation.extend(population[:s])
              # From 50% of fittest population, Individuals
              # will mate to produce offspring
              s = int((90*POPULATION SIZE)/100)
              for _ in range(s):
                     parent1 = random.choice(population[:50])
                     parent2 = random.choice(population[:50])
                     child = parent1.mate(parent2)
                     new generation.append(child)
              population = new_generation
              print("Generation: {}\tString: {}\tFitness: {}".\
                     format(generation,
                     "".join(population[0].chromosome),
                     population[0].fitness))
              generation += 1
       print("Generation: {}\tString: {}\tFitness: {}".\
              format(generation,
              "".join(population[0].chromosome),
              population[0].fitness))
if name == ' main ':
       main()
```

Otherwise generate new offsprings for new generation



| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
import numpy as np
class QLearningAgent:
  def init (self, num states, num actions, learning rate=0.1, discount factor=0.9,
exploration rate=0.1):
    # Initialize Q-table with zeros
    self.q table = np.zeros((num states, num actions))
    self.learning rate = learning rate
    self.discount factor = discount factor
    self.exploration_rate = exploration_rate
  def select action(self, state):
    # Choose the action with the highest Q-value, with exploration
    return np.argmax(self.q table[state, :]) if np.random.rand() > self.exploration rate else
np.random.choice(len(self.q_table[state, :]))
  def update q table(self, state, action, reward, next state):
    # Update Q-value using the Q-learning update rule
    best_next_action = np.argmax(self.q_table[next_state, :])
    self.q table[state, action] += self.learning rate * (reward + self.discount factor *
self.q table[next state, best next action] - self.q table[state, action])
class ParkingEnvironment:
  def init (self):
    # Initialize parking environment with 3 states and 2 actions
    self.num_states = 3
    self.num actions = 2
    self.goal state = 2
    self.agent state = 0
    self.done = False
  def reset(self):
    # Reset the environment for a new episode
    self.agent_state = 0
```

```
self.done = False
  def step(self, action):
    if self.done:
      return self.agent state, 0, self.done
    # Update agent's position based on the action
    if action == 0: # Move left
      self.agent_state = max(0, self.agent_state - 1)
    else: # Move right
      self.agent state = min(self.num states - 1, self.agent state + 1)
    # Provide reward based on the agent reaching the goal state
    reward = 1 if self.agent_state == self.goal_state else 0
    self.done = (self.agent_state == self.goal_state)
    return self.agent_state, reward, self.done
# Main loop
num states = 3
num actions = 2
num episodes = 100
# Create Q-learning agent and parking environment
agent = QLearningAgent(num states, num actions)
environment = ParkingEnvironment()
# Training loop
for episode in range(num episodes):
  environment.reset()
  # Episode loop
  while not environment.done:
    state = environment.agent state
    action = agent.select action(state)
    next_state, reward, done = environment.step(action)
    agent.update_q_table(state, action, reward, next_state)
```

```
# Test the trained agent

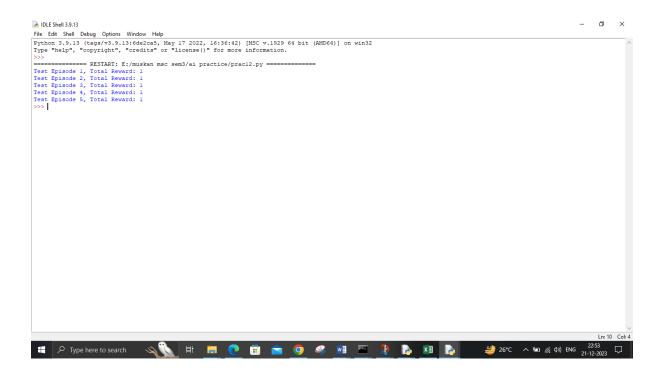
test_episodes = 5

for episode in range(test_episodes):
    environment.reset()

# Episode loop for testing

while not environment.done:
    state = environment.agent_state
    action = agent.select_action(state)
    next_state, reward, done = environment.step(action)

print(f"Test Episode {episode + 1}, Total Reward: {reward}")
```



| Writeup: | |
|----------|--|
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |
| | |

| Aim: |
|--------------|
| Description: |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| |
| - <u> </u> |
| |
| |
| |
| |
| |
| |
| |

```
# Importing the required libraries
import nltk
from nltk import CFG
# Defining the grammar rules
grammar = CFG.fromstring("""
  S -> NP VP
  NP -> Det N | Det N PP
  VP -> V NP | V NP PP
  PP -> P NP
  Det -> 'The' | 'a' | 'the'
  N -> 'dog' | 'cat' | 'house' | 'car'
  V -> 'chased' | 'ate' | 'drove'
  P -> 'in' | 'on' | 'at'
# Creating the parser
parser = nltk.ChartParser(grammar)
# Parsing a sentence
sentence = "The dog chased the cat"
for tree in parser.parse(sentence.split()):print(tree)
print("_____done by Muskan Momin____")
```

