**DATE:29.08.25**

**TASK:5**

**Implementation of Ant Colony Optimization to find the shortest and most efficient route for ride-sharing or delivery services using Python**

**AIM**

To implementation of Ant Colony Optimization to find the shortest and most efficient route for school route planning using Python

**ALGORITHM**

1. Start with the distance matrix of the school and all student stops.
2. Initialize parameters: number of ants, iterations, pheromone influence (α), visibility influence (β), and evaporation rate.
3. Initialize pheromone trails on all paths equally.
4. For each iteration, place all ants at the school (starting point).
5. For each ant, select the next stop using a probability rule based on pheromone level and visibility (1/distance).
6. Repeat the selection until the ant has visited all stops exactly once.
7. Make the ant return to the school to complete the tour.
8. Calculate the total distance (cost) of the route for each ant.
9. Update pheromone trails:

* Evaporate old pheromone.
* Add new pheromone inversely proportional to route length (shorter routes get more).

1. After all iterations, choose the route with the minimum distance as the optimal school bus route.

**PROGRAM**

**School Bus Route Planning**

import numpy as np

from numpy import inf

# Distance matrix (School + 4 Stops)

# 1 = School, 2–5 = Stops

d = np.array([[0, 10, 12, 11, 14], # distances from School

[10, 0, 13, 15, 8], # Stop 1

[12, 13, 0, 9, 14], # Stop 2

[11, 15, 9, 0, 16], # Stop 3

[14, 8, 14, 16, 0]]) # Stop 4

iteration = 100 # number of iterations

n\_ants = 5 # number of ants

n\_citys = 5 # number of cities (School + Stops)

# Parameters

e = 0.5 # evaporation rate

alpha = 1 # pheromone influence

beta = 2 # visibility influence

# Visibility = 1/distance

visibility = 1 / d

visibility[visibility == inf] = 0

# Initial pheromone levels

pheromone = 0.1 \* np.ones((n\_citys, n\_citys))

# Routes for ants

routes = np.ones((n\_ants, n\_citys + 1))

for ite in range(iteration):

routes[:, 0] = 1 # all ants start at the School (city 1)

for i in range(n\_ants):

temp\_visibility = np.array(visibility)

for j in range(n\_citys - 1):

cur\_loc = int(routes[i, j] - 1)

temp\_visibility[:, cur\_loc] = 0

# pheromone and visibility contributions

pheromone\_feat = np.power(pheromone[cur\_loc, :], alpha)

vis\_feat = np.power(temp\_visibility[cur\_loc, :], beta)

prob = pheromone\_feat \* vis\_feat

prob = prob / prob.sum() # normalize

cum\_prob = np.cumsum(prob)

r = np.random.random()

city = np.nonzero(cum\_prob > r)[0][0] + 1

routes[i, j + 1] = city

# last unvisited city

left = list(set([i for i in range(1, n\_citys + 1)]) - set(routes[i, :-2]))[0]

routes[i, -2] = left

route\_opt = np.array(routes)

dist\_cost = np.zeros((n\_ants, 1))

for i in range(n\_ants):

s = 0

for j in range(n\_citys - 1):

s += d[int(route\_opt[i, j]) - 1, int(route\_opt[i, j + 1]) - 1]

dist\_cost[i] = s

dist\_min\_loc = np.argmin(dist\_cost)

dist\_min\_cost = dist\_cost[dist\_min\_loc]

best\_route = routes[dist\_min\_loc, :]

pheromone = (1 - e) \* pheromone

for i in range(n\_ants):

for j in range(n\_citys - 1):

dt = 1 / dist\_cost[i]

pheromone[int(route\_opt[i, j]) - 1, int(route\_opt[i, j + 1]) - 1] += dt

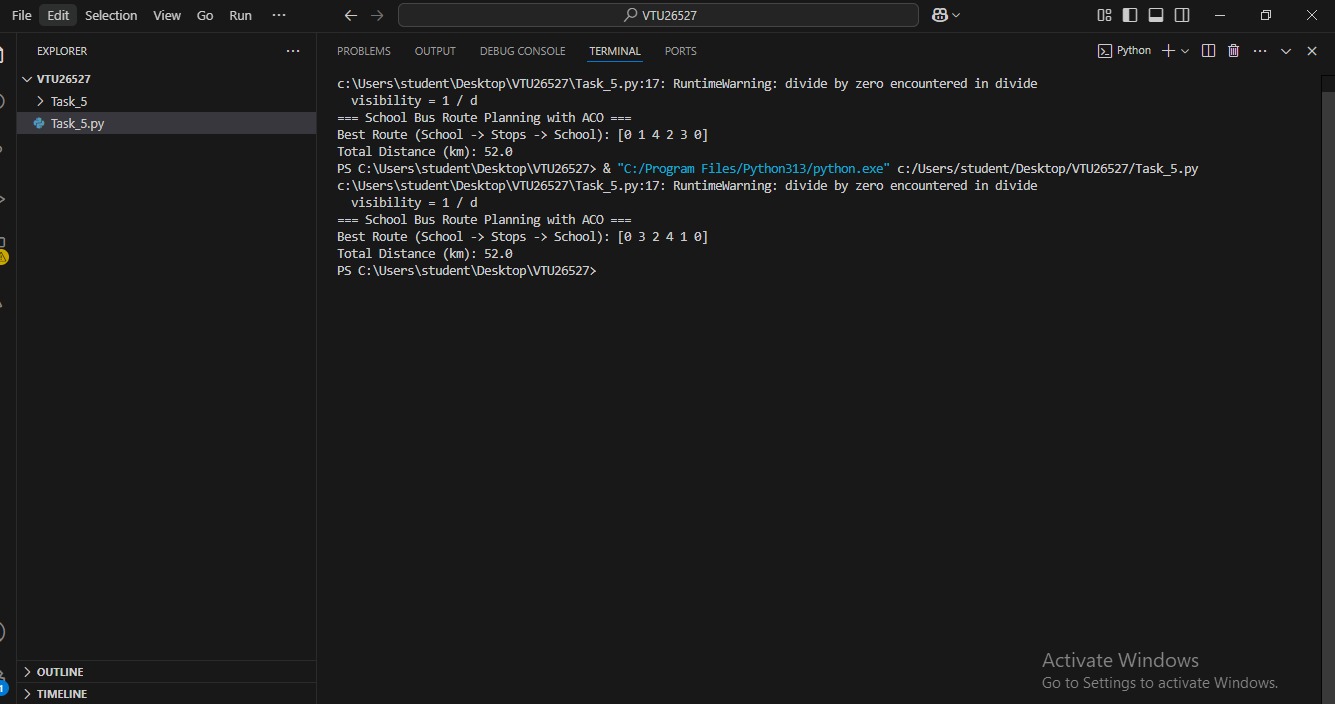
# Display final result

print("=== School Bus Route Planning with ACO ===")

print("Best Route (School -> Stops -> School):", best\_route.astype(int))

print("Total Distance (km):", int(dist\_min\_cost[0]) + d[int(best\_route[-2]) - 1, 0])

**OUTPUT**



**RESULT**

Thus, the implementation of Ant Colony Optimization to find the shortest and most efficient route for school route planning using Python was successfully executed and output was verified