USN: 1BM22CS253

LAB-3: Ant Colony Op miza on for the Traveling Salesman Problem:

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CODE:
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import numpy as np import
matplotlib.pyplot as plt
# 1. Define the Problem: Create a set of ci es with their coordinates ci
es = np.array([
  [0, 0], # City 0
  [1, 5], # City 1
  [5, 1], # City 2
  [6, 4], # City 3
[7, 8], # City 4
])
# Calculate the distance matrix between each pair of ci es
def calculate_distances(ci es):
                                   num_ci es = len(ci es)
distances = np.zeros((num_ci es, num_ci es))
  for i in range(num_ci es):
for j in range(num_ci es):
       distances[i][j] = np.linalg.norm(ci es[i] - ci es[j])
  return distances
distances = calculate_distances(ci es)
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# 2. Ini alize Parameters
num_ants = 10 num_ci es = len(ci es) alpha = 1.0 #
Influence of pheromone beta = 5.0 # Influence of
heuris c (inverse distance) rho = 0.5 # Evapora on
rate num itera ons = 30 ini al pheromone = 1.0
# Pheromone matrix ini aliza on pheromone = np.ones((num_ci es,
num_ci es)) * ini al_pheromone
# 3. Heuris c informa on (Inverse of distance) def heuris
                with np.errstate(divide='ignore'): # Ignore
c(distances):
division by zero
                     return 1 / distances
eta = heuris c(distances)
# 4. Choose next city probabilis cally based on pheromone and heuris c info def
choose next city(pheromone, eta, visited):
  probs = [] for j in
range(num_ci es):
    if j not in visited:
       pheromone_ij = pheromone[visited[-1], j] ** alpha
heuris c_ij = eta[visited[-1], j] ** beta
probs.append(pheromone_ij * heuris c_ij)
    else:
       probs.append(0) probs = np.array(probs)
np.random.choice(range(num_ci es), p=probs / probs.sum()) #
Construct solu on for a single ant def construct_solu on(pheromone,
eta): tour = [np.random.randint(0, num_ci es)] while len(tour) <
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num_ci es:
                 next_city = choose_next_city(pheromone, eta, tour)
tour.append(next_city) return tour
# 5. Update pheromones a er all ants have constructed their tours def
update_pheromones(pheromone, all_tours, distances, best_tour):
  pheromone *= (1 - rho) # Evaporate pheromones
  # Add pheromones for each ant's tour
for tour in all tours:
     tour_length = sum([distances[tour[i], tour[i + 1]] for i in range(-1, num_ci es - 1)])
for i in range(-1, num_ci es - 1):
       pheromone[tour[i], tour[i + 1]] += 1.0 / tour length
  # Increase pheromones on the best tour best_length = sum([distances[best_tour[i],
best_tour[i + 1]] for i in range(-1, num_ci es - 1)]) for i in range(-1, num_ci es - 1):
     pheromone[best_tour[i], best_tour[i + 1]] += 1.0 / best_length
# 6. Main ACO Loop: Iterate over mul ple itera ons to find the best solu on def
run_aco(distances, num_itera ons):
  pheromone = np.ones((num_ci es, num_ci es)) * ini al_pheromone
  best tour = None
best_length = float('inf')
  for itera on in range(num_itera ons):
     all_tours = [construct_solu on(pheromone, eta) for _ in range(num_ants)]
     all lengths = [\text{sum}([\text{distances}[\text{tour}[i], \text{tour}[i+1]] \text{ for } i \text{ in range}(-1, \text{ num } \text{ci es } -1)]) \text{ for tour in }
all tours]
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current_best_length = min(all_lengths)
                                                  current\_best\_tour =
all_tours[all_lengths.index(current_best_length)]
    if current_best_length < best_length:
best_length = current_best_length
                                          best_tour
= current_best_tour
    update_pheromones(pheromone, all_tours, distances, best_tour)
    print(f"Itera on {itera on + 1}, Best Length: {best_length}")
  return best_tour, best_length
# Run the ACO algorithm best_tour, best_length =
run_aco(distances, num_itera ons)
#7. Output the Best Solu on print(f"Best
Tour: {best_tour}") print(f"Best Tour
Length: {best_length}")
#8. Plot the Best Route def
plot_route(ci es, best_tour):
  plt.figure(figsize=(8, 6))
for i in range(len(ci es)):
    plt.sca er(ci es[i][0], ci es[i][1], color='red')
plt.text(ci es[i][0], ci es[i][1], f"City {i}", fontsize=12)
```

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Tteration 1, Best Length: 24.191626245470978
    Iteration 2, Best Length: 24.191626245470978
    Iteration 3, Best Length: 24.191626245470978
    Iteration 4, Best Length: 24.191626245470978
    Iteration 5, Best Length: 24.191626245478978
    Iteration 6, Best Length: 24.191626245470978
    Iteration 7, Best Length: 24.191626245470978
    Iteration 8, Best Length: 24.191626245470978
    Iteration 9, Best Length: 24.191626245470978
    Iteration 10, Best Length: 24.191626245470978
    Iteration 11, Best Length: 24.191626245470978
    Iteration 12, Best Length: 24.191626245470978
    Iteration 13, Best Length: 24.191626245470978
    Iteration 14, Best Length: 24.191626245470978
    Iteration 15, Best Length: 24.191626245470978
    Iteration 16, Best Length: 24.191626245470978
    Iteration 17, Best Length: 24.191626245470978
    Iteration 18, Best Length: 24.191626245470978
    Iteration 19, Best Length: 24.191626245470978
    Iteration 20, Best Length: 24.191626245470978
    Iteration 21, Best Length: 24.191626245470978
    Iteration 22, Best Length: 24.191626245470978
    Iteration 23, Best Length: 24.191626245470978
    Iteration 24, Best Length: 24.191626245470978
    Iteration 25, Best Length: 24.191626245470978
    Iteration 26, Best Length: 24.191626245470978
    Iteration 27, Best Length: 24.191626245470978
    Iteration 28, Best Length: 24.191626245470978
    Iteration 29, Best Length: 24.191626245470978
    Iteration 30, Best Length: 24.191626245470978
    Best Tour: [4, 3, 2, 0, 1]
    Best Tour Length: 24.191626245470978
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