

## Ant Colony Optimization for the Traveling Salesman Problem

```
import numpy as np

import random

class ACO:

    def __init__(self, n_ants, n_iterations, alpha, beta, rho, pheromone_deposit, cities):

        self.n_ants = n_ants

        self.n_iterations = n_iterations

        self.alpha = alpha # importance of pheromone

        self.beta = beta # importance of heuristic information

        self.rho = rho # pheromone evaporation rate

        self.pheromone_deposit = pheromone_deposit # pheromone deposit for the best path

        self.cities = cities

        self.num_cities = len(cities)

        self.distances = self.calculate_distances(cities)

        self.pheromone_matrix = np.ones((self.num_cities, self.num_cities))

    def calculate_distances(self, cities):

        distances = np.zeros((len(cities), len(cities)))

        for i, (x1, y1) in enumerate(cities):

            for j, (x2, y2) in enumerate(cities):

                distances[i, j] = np.sqrt((x1 - x2) ** 2 + (y1 - y2) ** 2)

        return distances

    def run(self):

        best_distance = float('inf')

        best_path = None

        for iteration in range(self.n_iterations):

            all_paths = []
```

```
all_distances = []
```

```
for ant in range(self.n_ants):
```

```
    path = self.construct_solution()
```

```
    path_distance = self.calculate_path_distance(path)
```

```
    all_paths.append(path)
```

```
    all_distances.append(path_distance)
```

```
    if path_distance < best_distance:
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```
        best_distance = path_distance
```

```
        best_path = path
```

```
self.update_pheromones(all_paths, all_distances)
```

```
print(f"Iteration {iteration+1}: Best Distance = {best_distance}")
```

```
return best_path, best_distance
```

```
def construct_solution(self):
```

```
    path = [random.randint(0, self.num_cities - 1)]
```

```
    while len(path) < self.num_cities:
```

```
        current_city = path[-1]
```

```
        next_city = self.choose_next_city(current_city, path)
```

```
        path.append(next_city)
```

```
    return path
```

```
def choose_next_city(self, current_city, path):
```

```
    probabilities = []
```

```
    for next_city in range(self.num_cities):
```

```
        if next_city not in path:
```

```
            pheromone = self.pheromone_matrix[current_city, next_city] ** self.alpha
```

```
            heuristic = (1.0 / self.distances[current_city, next_city]) ** self.beta
```

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        probabilities.append((next_city, pheromone * heuristic))
    else:
        probabilities.append((next_city, 0))

total = sum(prob for _, prob in probabilities)
probabilities = [(city, prob / total if total > 0 else 0) for city, prob in probabilities]

selected_city = random.choices(
    [city for city, _ in probabilities],
    weights=[prob for _, prob in probabilities],
    k=1
)[0]
return selected_city

def calculate_path_distance(self, path):
    distance = 0
    for i in range(len(path) - 1):
        distance += self.distances[path[i], path[i + 1]]
    distance += self.distances[path[-1], path[0]] # return to start
    return distance

def update_pheromones(self, paths, distances):
    self.pheromone_matrix *= (1 - self.rho) # evaporate pheromones

    # deposit pheromones based on path quality
    for path, distance in zip(paths, distances):
        for i in range(len(path) - 1):
            self.pheromone_matrix[path[i], path[i + 1]] += self.pheromone_deposit / distance
        self.pheromone_matrix[path[-1], path[0]] += self.pheromone_deposit / distance # return to
start

```

```
# Example usage with random cities

cities = [(random.randint(0, 100), random.randint(0, 100)) for _ in range(10)]

print(cities)

aco = ACO(n_ants=10, n_iterations=100, alpha=1, beta=2, rho=0.5, pheromone_deposit=10,
cities=cities)

best_path, best_distance = aco.run()

print("\nBest Path:", best_path)

print("Best Distance:", best_distance)
```

```
Iteration 66: Best Distance = 326.99882689334635
Iteration 67: Best Distance = 326.99882689334635
Iteration 68: Best Distance = 326.99882689334635
Iteration 69: Best Distance = 326.99882689334635
Iteration 70: Best Distance = 326.99882689334635
Iteration 71: Best Distance = 326.99882689334635
Iteration 72: Best Distance = 326.99882689334635
Iteration 73: Best Distance = 326.99882689334635
Iteration 74: Best Distance = 326.99882689334635
Iteration 75: Best Distance = 326.99882689334635
Iteration 76: Best Distance = 326.99882689334635
Iteration 77: Best Distance = 326.99882689334635
Iteration 78: Best Distance = 326.99882689334635
Iteration 79: Best Distance = 326.99882689334635
Iteration 80: Best Distance = 326.99882689334635
Iteration 81: Best Distance = 326.99882689334635
Iteration 82: Best Distance = 326.99882689334635
Iteration 83: Best Distance = 326.99882689334635
Iteration 84: Best Distance = 326.99882689334635
Iteration 85: Best Distance = 326.99882689334635
Iteration 86: Best Distance = 326.99882689334635
Iteration 87: Best Distance = 326.99882689334635
Iteration 88: Best Distance = 326.99882689334635
Iteration 89: Best Distance = 326.99882689334635
Iteration 90: Best Distance = 326.99882689334635
Iteration 91: Best Distance = 326.99882689334635
Iteration 92: Best Distance = 326.99882689334635
Iteration 93: Best Distance = 326.99882689334635
Iteration 94: Best Distance = 326.99882689334635
Iteration 95: Best Distance = 326.99882689334635
Iteration 96: Best Distance = 326.99882689334635
Iteration 97: Best Distance = 326.99882689334635
Iteration 98: Best Distance = 326.99882689334635
Iteration 99: Best Distance = 326.99882689334635
Iteration 100: Best Distance = 326.99882689334635

Best Path: [7, 9, 5, 4, 1, 3, 8, 0, 6, 2]
Best Distance: 326.99882689334635
```