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5-E

LAB PROGRAM-6

Write a program to implement Simulated Annealing algorithm:

N-queens problem

```
import mlrose_hiive as mlrose
import numpy as np

def queens_max(position):
    n = len(position)
    attacks = 0

    for i in range(n):
        for j in range(i + 1, n):
            # Check if queens attack each other
            if position[i] == position[j] or abs(position[i] -
position[j]) == j - i:
                attacks += 1

    # The fitness is the total number of pairs of queens minus the
number of attacks
    return (n * (n - 1) // 2) - attacks

# Define the custom fitness function
objective = mlrose.CustomFitness(queens_max)

# Set up the optimization problem
problem = mlrose.DiscreteOpt(length=8, fitness_fn=objective,
maximize=True, max_val=8)
T = mlrose.ExpDecay()

# Define the initial position
initial_position = np.array([4, 6, 1, 5, 2, 0, 3, 7])

# Run the simulated annealing algorithm
best_state, best_fitness, _ = mlrose.simulated_annealing(problem,
schedule=T, max_attempts=500, max_iters=5000,
init_state=initial_position)

print('The best position found is:', best_state)
print('The number of queens that are not attacking each other is:',
best_fitness)
```

OUTPUT:

```
➡ The best position found is: [2 5 7 1 3 0 6 4]
   The number of queens that are not attacking each other is: 28.0
```

Another Application of SA- Job Scheduling Problem

```
import mlrose_hiive as mlrose
import numpy as np

# Define the processing times for each job
job_times = [2, 14, 4, 16, 6, 5, 3, 12] # Processing times for each job
num_machines = 3 # Number of machines available

# Define a fitness function to calculate the makespan (total time taken
# by the slowest machine)
def job_scheduling_fitness(state):
    machine_times = [0] * num_machines # Initialize machine times

    for i, job in enumerate(state):
        machine_times[job] += job_times[i] # Add job time to the
        assigned machine

    makespan = max(machine_times) # Makespan is the max time of all
    machines
    return -makespan # We negate because we want to minimize the
    makespan

# Define a custom fitness function
objective = mlrose.CustomFitness(job_scheduling_fitness)

# Define the optimization problem
problem = mlrose.DiscreteOpt(length=len(job_times),
    fitness_fn=objective, maximize=True, max_val=num_machines)

# Define the simulated annealing schedule
schedule = mlrose.ExpDecay() # Exponential decay schedule for
    simulated annealing

# Initial state: assign each job to a random machine
initial_state = np.random.randint(0, num_machines, size=len(job_times))

# Perform the simulated annealing algorithm
best_state = mlrose.simulated_annealing(
    problem=problem,
    schedule=schedule,
    max_attempts=100,
```

```

        max_iters=1000,
        init_state=initial_state
    )

    # Since best_state is an object with both the best fitness and state,
    # extract them
    best_position = best_state[0]
    best_fitness = best_state[1]

    print("Best job-to-machine assignment:", best_position)
    print("Minimum makespan:", -best_fitness)

    # Display machine assignments
    machine_assignments = [[] for _ in range(num_machines)]
    for job, machine in enumerate(best_position):
        machine_assignments[machine].append((job, job_times[job]))

    for i, jobs in enumerate(machine_assignments):
        print(f"Machine {i+1} jobs:", jobs, "Total time:", sum(job[1] for
job in jobs))

```

```

➡ Best job-to-machine assignment: [1 2 0 1 2 0 1 0]
Minimum makespan: 21.0
Machine 1 jobs: [(2, 4), (5, 5), (7, 12)] Total time: 21
Machine 2 jobs: [(0, 2), (3, 16), (6, 3)] Total time: 21
Machine 3 jobs: [(1, 14), (4, 6)] Total time: 20

```