

Travelling Salesman Problem

By Genetic Algorithm

Overview

This report shows the pseudocodes and flowcharts of genetic algorithm and simulated annealing to solve the travelling salesman problem.

Genetic Algorithm

Part 1: code

best_route=0

n = desired population size

#generate n solutions as the first generation

for i **in** range(n):

 randomly generate a route (sequence of cities)

endfor

while fitness(best_route) is sufficiently large **or** maximum generation number is met:

this is the end condition

for i **in** range(generation):

 fitness[i] = 1/(total distance of route i)

use the reciprocal of total distance as fitness function

evaluate the fitness of every route in this generation

if fitness[i]>best_route:

 best_route = route i

#"best" represents the best existing route

endif

endfor

elite = first k routes of this generation(sort fitness from largest to smallest)

elite is a group of good individuals in this generation

k is the target size of the elite group

rest = randomly selected (n-k) routes according to their fitness weight

```
# rest is a group of parents that are not elite  
# since the number of total parents should be n but only k elite individuals are selected, "rest"  
parents should be selected
```

```
mating_pool = rest + elite  
# mating_pool is a group of parents for next generation, which are either from "elite" group  
or from "rest" group
```

```
child1 = elite  
# child1: "elite" group  
# child2: children generated by crossover of two parents
```

```
for i in range(n-k):  
    parent1, parent2 = randomly pick 2 parents from mating pool  
    child2[i] = crossover(parent1,parent2)  
    #for (n-k) times, randomly choose two parents, and breed a child by crossover  
    #the logic of crossover: keep part of route from parent 1 and pick the sequence of other  
cities from parent 2
```

```
endfor
```

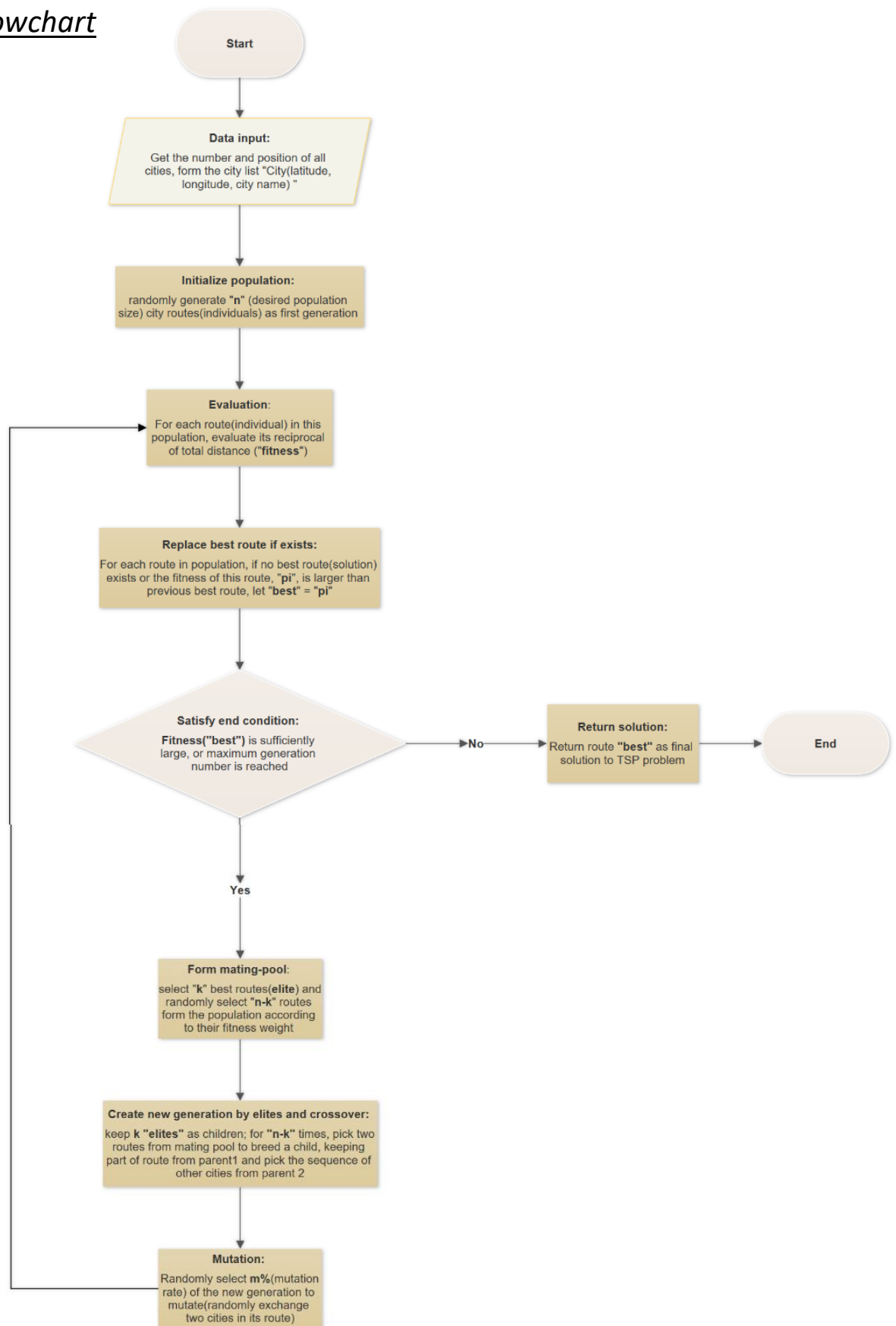
```
child = child1 + child2  
# the next generation are either from "elite" group or from children generated by crossover of  
two parents
```

```
generation = mutate(child)  
#mutate is to randomly select m%(a desired mutation rate) of the child, which are the routes,  
and randomly switch two cities of the selected routes
```

```
endwhile
```

```
solution = best_route  
#best is the best route (solution)
```

Part 2: flowchart



Simulated Annealing

Part 1: code

#At first, randomly generate a solution

initial_solution = randomly generate a route (sequence of cities)

current_distance = the total distance of this route

current_route = initial_solution

best_distance = 0

#ending condition: the temperature is sufficiently low

while temperature > temperature_end:

 new_route = switch(current_route)

generate a new solution by switching two random cities on previous solution

 new_distance = distance(new_route)

the total distance of this new route

 probability = $\exp(-(\text{new_distance} - \text{current_distance}) / \text{temperature})$

#this is the probability defined by simulated annealing algorithm

 a = a number randomly generated from 0 to 1

if new_distance < current_distance **or** a < probability:

if the new route is shorter or the random number is less than the replacement

probability, replace the old solution by current solution

 current_distance = new_distance

 current_route = new_route

endif

if current_distance < best_distance:

if the new route is shorter than the best existing route, replace the best route by this route

 best_distance = current_distance

 best_route = current_route

endif

 decrease temperature

endwhile

solution = best_route

best_route is the best solution to this problem

Part 2: flowchart

