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1  #!/usr/bin/env python
2  # -*- coding: UTF-8 -*-
3  """
4  @Project : Study
5  @File : 遗传代码实现.py
6  @Author : 少年又远方
7  @Date : 2022/4/28 17:17
8  @LastDate:
9  @Description:
10 """
11 import matplotlib.pyplot as plt
12 import numpy as np
13 import math
14 import random
15 import time
16
17 start = time.time()
18
19 # 31个城市的坐标
20 city_condition = [[106.54, 29.59], [91.11, 29.97], [87.68, 43.77], [106.27, 38.47], [111.65, 40.82],
21                  [108.33, 22.84], [126.63, 45.75], [125.35, 43.88], [123.38, 41.8], [114.48, 38.03], [112.53, 37.87],
22                  [101.74, 36.56], [117.0, 36.65], [113.6, 34.76], [118.78, 32.04], [117.27, 31.86],
23                  [120.19, 30.26], [119.3, 26.08], [115.89, 28.68], [113.0, 28.21], [114.31, 30.52],
24                  [113.23, 23.16], [121.5, 25.05], [110.35, 20.02], [103.73, 36.03], [108.95, 34.27],
25                  [104.06, 30.67], [106.71, 26.57], [102.73, 25.04], [114.1, 22.2], [113.33, 22.13]]
26
27 # 距离矩阵
28 # 使用numpy计算生成距离矩阵:
29 city_count = 31
30 Distance = np.zeros([city_count, city_count])
31 for i in range(city_count):
32     for j in range(city_count):
33         Distance[i][j] = math.sqrt(
34             (city_condition[i][0] - city_condition[j][0]) ** 2 + (city_condition[i][1] - city_condition[j][1]) ** 2)
35
36 # 种群数
37 count = 200
38 # 改良次数
39 improve_count = 500
40 # 进化次数
41 iteration = 2000
42 # 设置强者的定义概率, 即种群前20%为强者
43 retain_rate = 0.2
44 # 变异率
45 mutation_rate = 0.1
46 # 设置起点
47 index = [i for i in range(city_count)]
48
49 # 总距离
50 def get_total_distance(path_new):
51     distance = 0
52     for i in range(city_count - 1):
53         # count为30, 意味着回到了开始的点, 此时的值应该为0.
54         distance += Distance[int(path_new[i])][int(path_new[i + 1])]
55     distance += Distance[int(path_new[-1])][int(path_new[0])]
56     return distance
57
58
59 # 改良
60 # 思想: 随机生成两个城市, 任意交换两个城市的位置, 如果总距离减少, 就改变染色体。
61 # 此处不必关心, 此函数用于种群的初始化
62 def improve(x):
63     i = 0
64     distance = get_total_distance(x)
65     while i < improve_count:
66         u = random.randint(0, len(x) - 1)
67         v = random.randint(0, len(x) - 1)
68         if u != v:
69             new_x = x.copy()
70             # 随机交叉两个点, t为中间数
71             t = new_x[u]
72             new_x[u] = new_x[v]
73             new_x[v] = t
74             new_distance = get_total_distance(new_x)
75             if new_distance < distance:
76                 distance = new_distance
77                 x = new_x.copy()
78         else:
79             continue
80         i += 1
81
82
83 # 适应度评估, 选择, 迭代一次选择一次
84 def selection(population):
85     # 对总距离从小到大进行排序
86     graded = [[get_total_distance(x), x] for x in population]
87     graded = [x[1] for x in sorted(graded, key=lambda x: x[0])]
88     '''for item in graded:
89         item0 = sorted(item)
90         check = list(range(31))
91         print(item0==check)
92     '''
93     # 选出适应性强的染色体
94     # 注记:此处的x为列表, 列表的第一个元素为总距离, 第二个元素为染色体
95     # ***** Begin *****
96     retain_length = math.floor(len(graded)*retain_rate)
97     # 适应度强的集合,直接加入选择中
98     parents = graded[:retain_length+1]
99     # 轮盘赌算法选出K个适应性不强的个体, 保证种群的多样性
100    s = graded[retain_length:]
101    # 挑选的不强的个数
102    k = count * 0.2

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103 # 存储适应度
104 a = []
105 for i in range(0, len(s)):
106     a.append(get_total_distance(s[i]))
107 sum = np.sum(a)
108 b = np.cumsum(a / sum)
109 while k > 0: # 迭代一次选择k条染色体
110     t = random.random()
111     for h in range(1, len(b)):
112         if b[h - 1] < t <= b[h]:
113             parents.append(s[h])
114             k -= 1
115         break
116     return parents
117
118 # 交叉繁殖
119 def crossover(parents):
120     # 生成子代的个数,以此保证种群稳定
121     target_count = count - len(parents)
122     # 孩子列表
123     children = []
124     while len(children) < target_count:
125         male_index = random.randint(0, len(parents) - 1)
126         female_index = random.randint(0, len(parents) - 1)
127         # 在适应度强的中间选择父母染色体
128         if male_index != female_index:
129             male = parents[male_index]
130             female = parents[female_index]
131             left = random.randint(0, len(male) - 2)
132             right = random.randint(left + 1, len(male) - 1)
133             # print(female)
134             # 交叉片段
135             gene1 = male[left:right]
136             gene2 = female[left:right]
137             # 得到原序列通过改变序列的染色体,并复制出来备用。
138             child1_c = male[right:] + male[:right] # P3
139             child2_c = female[right:] + female[:right] # P4
140             child1 = child1_c.copy()
141             child2 = child2_c.copy()
142
143             # 已经改变的序列=>去掉交叉片段后的序列
144             for o in gene2:
145                 child1_c.remove(o)
146             for o in gene1:
147                 child2_c.remove(o)
148             # 交换交叉片段
149             seg = len(male)-1-right
150             new_child1 = child1_c[seg:]+gene2+child1_c[:seg]
151             new_child2 = child2_c[seg:]+gene1+child2_c[:seg]
152             children.append(new_child1)
153             children.append(new_child2)
154     return children
155
156 # 变异
157 def mutation(children):
158     # children现在包括交叉和优质的染色体
159     for i in range(len(children)):
160         if random.random() < mutation_rate:
161             child = children[i]
162             # 产生随机数
163             u = random.randint(0, len(child) - 4)
164             v = random.randint(u + 1, len(child) - 3)
165             w = random.randint(v + 1, len(child) - 2)
166             # 采用切片操作实现变异,将下标在u, v之间的数值插入下标为w,元素后面
167             child = child[0:u] + child[v:w] + child[u:v] + child[w:]
168             children[i] = child
169     return children
170
171 # 得到最佳纯输出结果
172 def get_result(population):
173     graded = [[get_total_distance(x), x] for x in population]
174     graded = sorted(graded)
175     return graded[0][0], graded[0][1]
176
177 if __name__ == '__main__':
178     # 使用改良圈算法初始化种群
179     population = []
180     for i in range(count):
181         # 随机生成个体
182         x = index.copy()
183         # 随机排序
184         random.shuffle(x)
185         improve(x)
186         population.append(x)
187     # 主函数:
188     register = []
189     i = 0
190     distance, result_path = get_result(population)
191     register.append(distance)
192     while i < iteration:
193         # 选择繁殖个体群
194         parents = selection(population)
195         # 交叉繁殖
196         children = crossover(parents)
197         # 变异操作
198         children = mutation(children)
199         # 更新种群
200         population = parents + children

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205     distance, result_path = get_result(population)
206     register.append(distance)
207     if i%50 == 0:
208         print("{}次迭代: {}".format(i))
209         print("当前距离: ", distance)
210         print("当前路径: ", result_path)
211     i = i + 1
212     print("迭代次数: ", iteration)
213     print("最优值是: ", distance)
214     print("最优路径: ", result_path)
215     if distance < 170:
216         print("路径长度符合要求")
217     else:
218         print("路径长度太长")
219     plt.rcParams['font.sans-serif'] = 'SimHei' # 设置中文显示
220     plt.rcParams['axes.unicode_minus'] = False
221     plt.figure(1)
222     X = []
223     Y = []
224     for item in result_path:
225         X.append(city_condition[item][0])
226         Y.append(city_condition[item][1])
227     plt.plot(X, Y, '-o')
228     for i in range(len(X)):
229         plt.text(X[i] + 0.05, Y[i] + 0.05, str(result_path[i]), color='red')
230     plt.xlabel('横坐标')
231     plt.ylabel('纵坐标')
232     plt.title('轨迹图')
233     plt.show()
234     plt.figure(2)
235     plt.plot(np.array(register))
236     plt.title('优化过程')
237     plt.ylabel('最优值')
238     plt.xlabel('代数({})->{}'.format(0, iteration))

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