2.适应度函数: 计算每个个体的冲突数, 冲突越少, 适应度越高。 3. 初始化种群: 随机生成一定数量的个体作为初始种群。 4. 选择操作:根据适应度选择优秀的个体,使用轮盘赌法进行选择。 5. 交叉操作:以一定的概率对选中的父代进行交叉,产生新的子代。 6. 变异操作:以一定的概率对新个体进行变异,改变个体的基因,增加种群多样性。 7. 迭代进化: 重复选择、交叉和变异操作,生成新一代种群。 8. 解的判定:在每一代中检查是否存在满足条件的个体(冲突数为0),如果找到则记录解。 → 交叉操作 — 是否找到解? 计算适应度 产生新种群 开始 ── 初始化种群 ─ 达到最大代数? RESET Part 2:基本代码实现 2.1 导入相关库并初始化优化参数 In [269... import random import multiprocessing # 参数设置 N = 8 # 皇后数量 POP_SIZE = 500 # 每个种群的规模 MAX GEN = 5000 # 最大代数 NUM_POPULATIONS = 4 # 并行种群数量 CROSS_RATE = 0.9 # 交叉概率 MUTATION_RATE = 0.3 # 变异概率 ELITE_SIZE = 20 # 精英个体数量 2.2 定义相关操作和函数 2.2.1 种群初始化 In [270... # 初始化种群 def init_population(): population = [] for _ in range(POP_SIZE): individual = random.sample(range(N), N) # 不重复的排列,增加初始多样性 population.append(individual) return population 2.2.2 适应度函数 In [271... # 计算适应度函数,冲突越少适应度越高 def fitness(individual): conflicts = 0 for i in range(N): for j in range(i + 1, N): if abs(individual[i] - individual[j]) == j - i: conflicts += 1 return max_fitness - conflicts $max_fitness = N * (N - 1) // 2 # 最大冲突数为28$ 2.2.3 选择操作 In [272... # 选择操作 (锦标赛选择) def selection(population): selected = [] for _ in range(POP_SIZE): ind1 = random.choice(population) ind2 = random.choice(population) fitter = ind1 if fitness(ind1) > fitness(ind2) else ind2 selected.append(fitter) return selected 2.2.4 交叉操作 In [273... # 排序交叉 (Order Crossover, OX) def crossover(parent1, parent2): if random.random() < CROSS_RATE:</pre> point1, point2 = sorted(random.sample(range(N), 2)) child = [None]*N child[point1:point2+1] = parent1[point1:point2+1] ptr = 0 for i in range(N): if parent2[i] not in child: while child[ptr] is not None: ptr **+=**1 child[ptr] = parent2[i] return child else: return parent1[:] 2.2.5 变异操作 In [274... # 变异操作(逆序变异) def mutation(individual): if random.random() < MUTATION_RATE:</pre> point1, point2 = sorted(random.sample(range(N), 2)) individual[point1:point2+1] = reversed(individual[point1:point2+1]) return individual 2.2.6 进化操作 In [275... # 检查是否为正确解 def is_valid(individual): return fitness(individual) == max_fitness 2.3 主程序 In [282... # 主函数 def genetic_algorithm(printif = True): population = init_population() solutions = [] gen = 0number = 0 while gen < MAX_GEN:</pre> gen += 1 # 计算种群适应度 population_fitness = [(individual, fitness(individual)) for individual in population] population_fitness.sort(key=lambda x: x[1], reverse=True) # 精英保留 elites = [individual for individual, fit in population_fitness[:ELITE_SIZE]] # 检查是否有新解 for individual, fit in population_fitness: if fit == max_fitness and individual not in solutions: solutions.append(individual) #print(f"在第 {gen} 代找到新解: {individual}") # 如果找到所有92个解,提前退出 if len(solutions) >= 92: if printif: print("\n已找到所有可能的解。") gen = MAX_GEN break # 选择 selected = selection(population) # 生成新种群 new_population = elites.copy() while len(new_population) < POP_SIZE:</pre> parent1 = random.choice(selected) parent2 = random.choice(selected) child = crossover(parent1, parent2) child = mutation(child) new_population.append(child) population = new_population # 输出所有解 unique_solutions = [] for solution in solutions: if solution not in unique_solutions: unique_solutions.append(solution) number += 1 if printif: print(f"\n共找到 {len(unique_solutions)} 个独特的解: ") for solution in unique_solutions: print(solution) return number if __name__ == "__main__": print(genetic_algorithm(False)) 92 Part 3:测试程序效率及准确性 In [283... import matplotlib.pyplot as plt result = [] for i in range(1, 51): result.append(genetic_algorithm(False)) plt.plot(result) plt.xlabel('times') plt.ylabel('number of solutions') plt.show() 96 94 number of solutions 88 10 20 30 40 50 times

Homework 2: 八皇后问题

该程序使用遗传算法来解决8皇后问题。主要思路如下:

Part 1: 问题描述

八皇后问题

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1. 编码方式:采用顺序编码,每个个体由长度为8的数组表示,数组的每个元素代表皇后在对应列上的行位置。

把8个皇后放在8×8的棋盘上,使其不能互相攻击,即任意两个皇后都不能处于同一行、同一列或同一斜线上。请编写一个程序,输出所有的解。

要求:采用顺序编码遗传算法;源代码一份,50次运行测试程序效率及准确性;电子版作业报告一份:包含问题描述、编程思路及方法、源代码(添加适当的注释)