

Question 2_GA algo

June 14, 2019

遗传算法解决 TSP 问题

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[212]: import pandas as pd;
import numpy as np;
import random;
import math;
import matplotlib.pyplot as plt;
import time;

[213]: # 基因序列类
class geneCode(object):
    def __init__(self, code = None):
        self.gene = code;
        self.ada_val = -1;

# 遗传算法类
class GA(object):
    def __init__(self, indiCount, aCrossRate, aChangeRate, codeLength, Fun):
        self.group = [];
        self.best = None;
        self.groupCnt = indiCount;
        self.generation = 1;
        self.crossRate = aCrossRate;
        self.changeRate = aChangeRate;
        self.crossCnt = 0;
        self.changeCnt = 0;
        self.num_city = codeLength;
        self.match = Fun;
        self.sum_ada = 0.0;
        self.initGroup();

    def initGroup(self):
        self.group = [];
        for i in range(self.groupCnt):
            gene = list(range(self.num_city));
            random.shuffle(gene);
            indi = geneCode(gene);
            self.group.append(indi);
```

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def cross(self, parent1, parent2):
    index1 = random.randint(0, self.num_city - 1)
    index2 = random.randint(index1, self.num_city - 1)
    tempGene = parent2.gene[index1:index2]
    newGene = []
    p1len = 0
    for g in parent1.gene:
        if p1len == index1:
            newGene.extend(tempGene)
            p1len += 1
        if g not in tempGene:
            newGene.append(g)
            p1len += 1
    self.crossCnt += 1
    return newGene

def change(self, gene):
    index1 = random.randint(0, self.num_city - 1)
    index2 = random.randint(0, self.num_city - 1)
    gene[index1], gene[index2] = gene[index2], gene[index1]
    self.changeCnt += 1
    return gene

# 估算个体的适应度
def eval(self):
    self.sum_ada = 0.0;
    self.best = self.group[0];
    for i in self.group:
        i.ada_val = self.match(i);
        self.sum_ada += i.ada_val;
        if self.best.ada_val < i.ada_val:
            self.best = i;

def choose(self):
    r = random.uniform(0, self.sum_ada)
    for i in self.group:
        r -= i.ada_val;
        if r <= 0:
            return i;

def child(self):
    parent1 = self.choose();
    rate = random.random();
    # 按概率交叉
    if rate < self.crossRate:
        parent2 = self.choose();

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        gene = self.cross(parent1, parent2);
    else:
        gene = parent1.gene;
    # 按概率突变
    rate = random.random();
    if rate < self.changeRate:
        gene = self.change(gene);
    return geneCode(gene);

def nextGeneration(self):
    self.eval();
    newGroup = [];
    newGroup.append(self.best);
    while len(newGroup) < self.groupCnt:
        newGroup.append(self.child());
    self.group = newGroup;
    self.generation += 1;

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[218]: class TSP(object):
    def __init__(self, aLifeCount = 100, N = 10):
        self.read_data(N);
        self.groupCount = aLifeCount;
        self.ga = GA(indiCount = self.groupCount, aCrossRate = 0.8,
                     aChangeRate = 0.01, codeLength = len(self.cities), Fun =
→self.match());

    def read_data(self, n):
        self.cities = [];
        if(n == 10):
            f=open('TSP10cities.tsp', 'r');
        if(n == 100):
            f=open('TSP100cities.tsp', 'r');
        while True:
            x = str(f.readline());
            if not x:
                break;
            x = x.replace("\n", "");
            x = x.split();
            self.cities.append((float(x[1]),float(x[2]),x[0]));

    def distance(self, order):
        total_dist = 0.0;
        for i in range(-1, len(self.cities) - 1):
            index1, index2 = order[i], order[i + 1];
            city1, city2 = self.cities[index1], self.cities[index2];
            total_dist += math.sqrt(pow((city1[0] - city2[0]), 2) +
→pow((city1[1] - city2[1]), 2));

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        return total_dist;

    def match(self):
        return lambda x: 1.0 / self.distance(x.gene);

    def run(self, iter_num = 0):
        for i in range(iter_num):
            self.ga.nextGeneration();
            distance = self.distance(self.ga.best.gene);
            if( (dist_record) and (distance == dist_record[-1]) ):
                end = time.clock();
                iter_record.append(i + 1);
                dist_record.append(distance);
            print("经过%d次迭代, 获得的总距离为: %f"%(self.ga.generation, distance))
            print("到收敛时经过时间为: %s" % (end - start));
            print("遍历城市顺序为: ")
            for i in self.ga.best.gene:
                print(self.cities[i][2], end = ">"),

```

问题规模为 $n = 10$

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[223]: iter_record = [];
        dist_record = [];
        start = time.clock();
        x = TSP(N = 10);
        x.run(1000);

```

经过 1001 次迭代, 获得的总距离为: 10127.552144

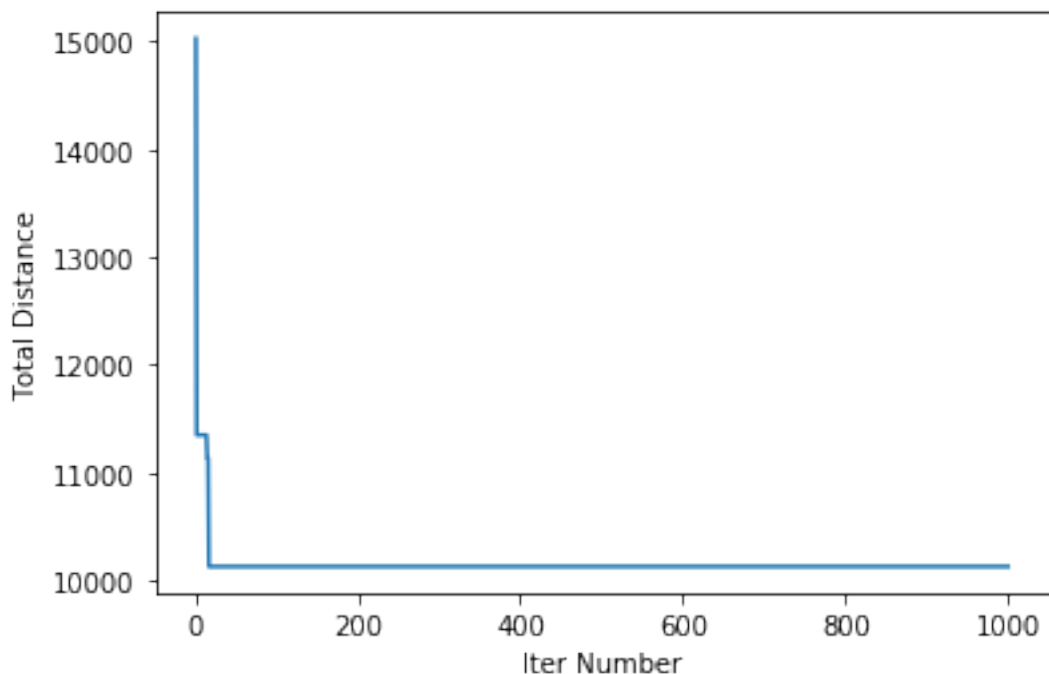
到收敛时经过时间为: 2.1557441538448074

遍历城市顺序为:

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[204]: plt.plot(iter_record, dist_record);
        plt.xlabel("Iter Number");
        plt.ylabel("Total Distance");

```



问题规模为 $n = 100$

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[221]: iter_record = [];
dist_record = [];
start = time.clock();
x = TSP(N = 100);
x.run(1000);
```

经过 1001 次迭代，获得的总距离为：105731.496890

到收敛时经过时间为：12.269688569149821

遍历城市顺序为：

44->1->81->9->19->18->22->26->32->29->46->8->25->35->54->33->39->50->36->13->15->58->60->61->53->12->70->56->48->38->64->24->62->17->34->89->42->41->7->91->37->31->30->49->65->45->68->66->79->67->52->74->69->96->16->14->23->59->6->40->76->10->43->11->20->97->4->21->57->75->63->88->47->71->72->99->82->86->95->90->55->5-

```
[222]: plt.plot(iter_record, dist_record);
plt.xlabel("Iter Number");
plt.ylabel("Total Distance");
```

