4.4.3简化模型的求解及结果分析

微粒群算法,又称粒子群优化(particle swarm optimization, PSO)是一类基于群体智能的随机优化算法。是由J. Kennedy和R. C. Eberhart等于1995年开发的一种演化计算技术。现已广泛应用于函数优化、神经网络训练、模式分类、模糊系统控制以及其他遗传算法的应用领域[曾建潮,介婧,崔志华.微粒群算法[M].北京:科学出版社,2004:9- 13]

Particle Swarm Optimization is a [random](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%99%BA%E8%83%BD%E7%9A%84%E9%9A%8F%E6%9C%BA%E4%BC%98%E5%8C%96%E7%AE%97%E6%B3%95) [optimization](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%99%BA%E8%83%BD%E7%9A%84%E9%9A%8F%E6%9C%BA%E4%BC%98%E5%8C%96%E7%AE%97%E6%B3%95) [algorithm](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%99%BA%E8%83%BD%E7%9A%84%E9%9A%8F%E6%9C%BA%E4%BC%98%E5%8C%96%E7%AE%97%E6%B3%95" \t "_blank) based on **swarm intelligence, which** **is ev**[olutionary](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%BC%94%E5%8C%96%E8%AE%A1%E7%AE%97%E6%8A%80%E6%9C%AF) computation [technology](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%BC%94%E5%8C%96%E8%AE%A1%E7%AE%97%E6%8A%80%E6%9C%AF" \t "_blank)[,](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%BC%94%E5%8C%96%E8%AE%A1%E7%AE%97%E6%8A%80%E6%9C%AF" \t "_blank) **developed by** J. Kennedy and R. C. Eberhart **in 1995. Now, it is widely used in function, optimization neural network training and fuzzy control system.**

In the D –dimension search space, a group consists of n particles and i th particle refers to D-[dimensional](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E4%B8%89%E7%BB%B4%E5%90%91%E9%87%8F) [vectors](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E4%B8%89%E7%BB%B4%E5%90%91%E9%87%8F" \t "_blank), $x\_i=(x\_{i**1**},x\_{i**2**},\**cdots**,x\_{iD})^T,i=**1**,**2**,\**cots**,n$. Then we plug x\_i back into a objective function to get its [adaptive](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E9%80%82%E5%BA%94%E5%80%BC" \t "_blank) [value](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E9%80%82%E5%BA%94%E5%80%BC" \t "_blank). By comparing the size of value ,we can measure the strength and weakness.

The flight speed of i th [particle](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E7%B2%92%E5%AD%90) should also be n [dimensional](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E4%B8%89%E7%BB%B4%E5%90%91%E9%87%8F) [vectors](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E4%B8%89%E7%BB%B4%E5%90%91%E9%87%8F):

$$v\_i=(v\_{i**1**},v\_{i**2**},\**cdots**,v\_{iD})^T,i=**1**,**2**,\**cdots**,n$$

But the speed of each dimension is limited the area $[-v\_{max},v\_{max}]$,$v\_{max}$ determines the accuracy between current position and the best position. If $v\_{max}$ is high, the particle may miss the optimal solution, on the contrary, if $v\_{max}$ is too small, the particle will not explore insufficiently so as to be trapped in [local](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%B1%80%E9%83%A8%E6%9C%80%E4%BC%98%E8%A7%A3" \t "_blank) [optima](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%B1%80%E9%83%A8%E6%9C%80%E4%BC%98%E8%A7%A3" \t "_blank)[.](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%B1%80%E9%83%A8%E6%9C%80%E4%BC%98%E8%A7%A3" \t "_blank)

The best position i th particle searched to date is:

$$p\_i=(p\_{i1},p\_{i2},\cdots,p\_{iD})^T,i=1,2,\cdots,n$$

At the n ith iteration, the [optimal](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%90%9C%E7%B4%A2%E5%88%B0%E6%9C%80%E4%BC%98%E7%9A%84%E4%BD%8D%E7%BD%AE%E4%B8%BA) [location](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%90%9C%E7%B4%A2%E5%88%B0%E6%9C%80%E4%BC%98%E7%9A%84%E4%BD%8D%E7%BD%AE%E4%B8%BA" \t "_blank) the particle swarm searched is:

$$p\_g=(p\_{g1},p\_{g2},\cdots,p\_{gD})^T,g=1,2,\cdots,m$$

Before finding two optimal values, the particle swarm updates speed and position in the following equations;

$$**\begin{*cases*}**

v\_{id}^{k+**1**}=w\**timesv**\_{id}^k+c\_**1**r\_**1**(p\_{id}-x\_{id}^k)+c\_**2**r\_**2**(p\_{gd}-x\_{gd}^k)\\

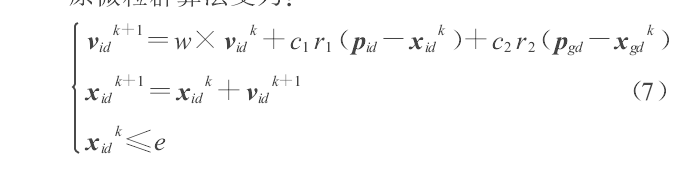
x\_{id}^{k+**1**}=x\_{id}^k+v\_{id}^{k+**1**}

**\end{*cases*}$$**

$i=1,2,\cdots,n$;$d=1,2,\cdots,D$;where $c\_1$ and $c\_2$, are learning factor and **nonnegative number, which belong to $[0,2]$;$r\_1$and $r\_2$** are random numbers between 0 and 1;k refers to **iteration.**

在利用粒子群算法时，为了简化问题，增加对粒子扩散的约束，主观因素引入一个参数e。则粒子群算法变为：

When using particle swarm [algorithm](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%99%BA%E8%83%BD%E7%9A%84%E9%9A%8F%E6%9C%BA%E4%BC%98%E5%8C%96%E7%AE%97%E6%B3%95),in order to simplify the problem and increase the confinement on the particle, so we introduce a subjective factor e.



$$**\begin{*cases*}**

v\_{id}^{k+**1**}=w\**times** v\_{id}^k+c\_**1**r\_**1**(p\_{kl}-x\_{kl}^k)+c\_**2**r\_**2**(p\_{gd}-x\_{gl}^k)\\

x\_{id}^{k+**1**}=x\_{id}^k+v\_{id}^{k+**1**}\\

x\_{id}^{k}\**leqslant** e

**\end{*cases*}$$**

参数设定如下：w的取值范围[0.8,1.0]。常数c1,c2取值范围为[0,2]。权重变化速度的界限常数vmax取值范围为[0,1]。 e作为约束条件分别设置为(5,10,15,20,25,30,35,40,45,50),本实验结束条件为最大迭代次数600次。

P[arameters](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%8F%82%E6%95%B0%E8%AE%BE%E5%AE%9A%E5%A6%82%E4%B8%8B) [are](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%8F%82%E6%95%B0%E8%AE%BE%E5%AE%9A%E5%A6%82%E4%B8%8B) [set](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%8F%82%E6%95%B0%E8%AE%BE%E5%AE%9A%E5%A6%82%E4%B8%8B) as follows:

The range of w is 0.8 to1.0;The constant range is 0 to 2.where $v\_max$, is the limit of speed and range from 0 to 1.e as [condition](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E7%BA%A6%E6%9D%9F%E6%9D%A1%E4%BB%B6) [of](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E7%BA%A6%E6%9D%9F%E6%9D%A1%E4%BB%B6) [constraint](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E7%BA%A6%E6%9D%9F%E6%9D%A1%E4%BB%B6) is set $(5,10,15,20,25,30,35,40,45,50)$respectively. The number of iterations just only reach 600 times.

结果分析（待定）

5.灵敏性分析

[Sensitivity](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E7%81%B5%E6%95%8F%E6%80%A7%E5%88%86%E6%9E%90" \t "_blank) [Analysis](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E7%81%B5%E6%95%8F%E6%80%A7%E5%88%86%E6%9E%90" \t "_blank)

5.1 ROI定义式的灵敏性分析

The sensitivity analysis of the definition of ROI

在实际情况中，很多因素我们是有很大不确定性的。我们计算ROI对毕业生的工资（题目所给数据中的一项指标）的灵敏性程度。我们记ROI为r，则得到:

公式

[In](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%9C%A8%E5%AE%9E%E9%99%85%E6%83%85%E5%86%B5%E4%B8%AD" \t "_blank) the [actual](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%9C%A8%E5%AE%9E%E9%99%85%E6%83%85%E5%86%B5%E4%B8%AD) [situation](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%9C%A8%E5%AE%9E%E9%99%85%E6%83%85%E5%86%B5%E4%B8%AD" \t "_blank)[,](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%9C%A8%E5%AE%9E%E9%99%85%E6%83%85%E5%86%B5%E4%B8%AD" \t "_blank)many factors exist [considerable](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%BE%88%E5%A4%A7%E4%B8%8D%E7%A1%AE%E5%AE%9A%E6%80%A7" \t "_blank) [uncertainty](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%BE%88%E5%A4%A7%E4%B8%8D%E7%A1%AE%E5%AE%9A%E6%80%A7" \t "_blank)[.](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%BE%88%E5%A4%A7%E4%B8%8D%E7%A1%AE%E5%AE%9A%E6%80%A7" \t "_blank) We can get the degree of sensitivity between graduated students’ income and ROI by calculating and define ROI as r, then we get

$$S(r,m)=\frac{dr}{dm}\cdot\frac{m}{r}=1.6$$

由公式可以看出来，工资每增加10%, 学校的ROI提高16%。ROI定义式对工资具有较好的稳定度。因为我们所定义的ROI是毕业学生的社会贡献程度。而工资水平放映了他一定的贡献度。

From the equation ,we can explicitly see that the ROI improves16% when the income increases 10% in per, which reflect the [good](http://cn.bing.com/dict/clientsentence?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%85%B7%E6%9C%89%E7%A8%B3%E5%AE%9A%E5%BA%A6" \t "_blank) [stability](http://cn.bing.com/dict/clientsentence?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%85%B7%E6%9C%89%E7%A8%B3%E5%AE%9A%E5%BA%A6" \t "_blank) of income in terms of

ROI [definition](http://cn.bing.com/dict/clientsentence?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E5%AE%9A%E4%B9%89%E5%BC%8F" \t "_blank), because the ROI we define is the degree of contributions to society for students. Coincidently, wage can show contributions to some extent.

5.2简化模型的灵敏性分析

实际投资中，我们面临的投资学校数目将是巨大的。我们分析投资结果对可投资数目的灵敏性。通过增加数目，粒子群算法收敛性越来越差。这个情况说明的我们的模型稳定性很差。需要进一步分析约束条件的设定。

In the actual investment, we will have to solve more schools to invest. We analyze the sensitivity of the investment results about the numbers of schools. The convergence of PSO becomes worse and worse with the numbers increasing. This result indicates that the stability of our model is poor. We need future work to improve our model.

6.模型评价

Model analysis

优点

Strength

我们对数据分析的较为透彻。我们仔细分析了数据类型及性质。理解了“null”值出现的原因，有效的解决了问题。

We have a clear understanding of data by analyzing data types and properties and know the reason why ‘’NULL’’ appears, solving the problem effectively.

我们将问题分解为几个子问题求解，将问题细化。我们一共建立了三个模型，分别用于求解投资学校、投资金额以及持续时间。模型之间相互联系，共同构成一个整体。

We further classify a problem into some small problems and build three models, which are used in identifying investment schools, investment amount and duration time. The models interconnect mutually and [constitute](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%9E%84%E6%88%90%E4%B8%80%E4%B8%AA%E6%95%B4%E4%BD%93) [a](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%9E%84%E6%88%90%E4%B8%80%E4%B8%AA%E6%95%B4%E4%BD%93" \t "_blank) [whole](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%9E%84%E6%88%90%E4%B8%80%E4%B8%AA%E6%95%B4%E4%BD%93" \t "_blank)[.](http://cn.bing.com/dict/clientsearch?mkt=zh-CN&setLang=zh&form=BDVEHC&ClientVer=BDDTV3.5.0.4311&q=%E6%9E%84%E6%88%90%E4%B8%80%E4%B8%AA%E6%95%B4%E4%BD%93" \t "_blank)

合理提出假设，并初步细化模型。我们首先考虑简化问题，通过提出合理的假设，简化问题的分析。然后逐步去掉假设，建立复杂模型，逼近实际情况。

Reasonable assumptions are proposed properly and the models are break down. We firstly put forward

缺点

Weakness

定义的ROI考虑不全面。我们参考金融领域对ROI的定义方式。我们只考虑了与ROI最直接相关的因素。所以，对于ROI的定义不够准确，在一定程度上无法反映各个大学真是的回报率。

数据处理主观判断影响较大。通过对题目给的数据以及收集的数据的整理，我们主观舍弃了一些没有价值的指标。使我们最后分析的结果有一定影响。

AHP评价结果受主观因素影响较大，不够准确。AHP的判断矩阵是通过查阅文献资料、询问专家后确定下来的。所以结果受主观因素影响很大。

持续时间估算的不够准确。由于学校不同年份的数据较少，每个学校只有4到5年的数据。根据图像走势作出的判断不够准确。

Future work

完善对ROI的定义。进一步分析数据，找出与ROI呈负相关的因素。建立更能够反映学校回报率的ROI定义。优化投资策略。

结合模糊综合评价，降低AHP的主观性的影响。模糊综合评价更依赖于数据。整合两种评价方法，选取恰当的权值比例。能够有效地减少AHP的主观因素影响。这样可以进一步优化候选名单。

完善动态的多期的投资组合模型。多期多目标动态投资组合模型，考虑情况更加接近实际情况。通过这个模型，可以在统一分析投资资金、投资学校和投资时间的情况下求解各值。我们会继续思考合适的算法。我们尽力将动态多期模型完善并求解出结果。

7.参考文献