



**北京航空航天大学**  
B E I H A N G U N I V E R S I T Y

## 数值分析第三次大作业

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# 1. 设计方案

## 题目描述

本次作业出自《数值分析（第四版）》（颜庆津 编著）计算实习说明书第八题：

关于 $x, y, t, u, v, w$ 的方程组 (A.3)

$$\begin{cases} 0.5 \cos t + u + v + w - x = 2.67 \\ t + 0.5 \sin u + v + w - y = 1.07 \\ 0.5t + u + \cos v + w - x = 3.74 \\ t + 0.5u + v + \sin w - y = 0.79 \end{cases} \quad (\text{A.3})$$

以及关于 $z, t, u$ 的二维数表（见表A-1）确定了一个二元函数 $z = f(x, y)$ 。

$t \quad u$ $z$	0	0.4	0.8	1.2	1.6	2
0	-0.5	-0.34	0.14	0.94	2.06	3.5
0.2	-0.42	-0.5	-0.26	0.3	1.18	2.38
0.4	-0.18	-0.5	-0.5	-0.18	0.46	1.42
0.6	0.22	-0.34	-0.58	-0.5	-0.1	0.62
0.8	0.78	-0.02	-0.5	-0.66	-0.5	-0.02
1.0	1.5	0.46	-0.26	-0.66	-0.74	-0.5

1. 试用数值方法求出 $f(x, y)$ 在区域 $D = \{(x, y) | 0 \leq x \leq 0.8, 0.5 \leq y \leq 1.5\}$ 上的近似表达式

$$p(x, y) = \sum_{r=0}^k \sum_{s=0}^k c_{rs} x^r y^s$$

要求 $p(x, y)$ 以最小的 $k$ 值达到以下的精度

$$\sigma = \sum_{i=0}^{10} \sum_{j=0}^{20} [f(x_i, y_j) - p(x_i, y_j)]^2 \leq 10^{-7}$$

其中 $x_i = 0.08i, y_j = 0.5 + 0.05j$ 。

2. 计算 $f(x_i^*, y_j^*), p(x_i^*, y_j^*)$  ( $i = 1, 2, \dots, 8; j = 1, 2, \dots, 5$ )的值，以观察 $p(x, y)$ 逼近 $f(x, y)$ 的效果，其中 $x_i^* = 0.1i, y_j^* = 0.5 + 0.2j$ 。

## 说明

1. 用迭代方法求解非线性方程组时，要求近似解向量 $x^{(k)}$ 满足以下的精度：

$$\frac{\|x^{(k)} - x^{(k-1)}\|_{\infty}}{\|x^{(k)}\|_{\infty}} \leq 10^{-12}$$

2. 做二元插值时，要求使用分片二次代数插值。

3. 要由程序自动确定最小的 $k$ 值。

4. 打印以下内容：

- (1) 全部源程序;
  - (2) 数表:  $(x_i, y_j, f(x_i, y_j))$  ( $i = 0, 1, \dots, 10; j = 0, 1, \dots, 20$ );
  - (3) 选择过程的 $k$ 和 $\sigma$ 值;
  - (4) 达到精度要求时的 $k$ 和 $\sigma$ 值以及 $p(x, y)$ 中的系数 $c_{rs}$  ( $r = 0, 1, \dots, k; s = 0, 1, \dots, k$ );
  - (5) 数表:  $(x_i^*, y_j^*, f(x_i^*, y_j^*), p(x_i^*, y_j^*))$  ( $i = 1, 2, \dots, 8; j = 1, 2, \dots, 5$ ).
5. 采用f型输出 $x_i, y_j, x_i^*, y_j^*$ 的准确值, 其余实型数采用e型输出并且至少显示12位有效数字。

## 算法分析

### 总体思路

首先题目中并没有给出 $f(x, y)$ 的具体形式。由于 $x, y$ 确定为 $x_i, y_j$ 时, 通过解四元非线性方程组, 得到 $t, u$ 也确定, 因此我们转而需要求出 $z = f(x, y) = \varphi(t, u)$ , 题目中也没有给出 $\varphi$ , 而是给出了一个 $z, t, u$ 的数表, 按照说明, 我们需要用分片二次代数插值来确定 $\varphi$ , 进而确定 $f(x, y)$ , 求出 $f(x_i, y_j)$ 。

求出所有 $f(x_i, y_j)$ 之后, 我们需要进行近似拟合, 这里我们采用最小二乘曲面拟和方法, 详见书P141-P143页。另外注意我们选取幂函数为基函数, 这样可以直接求出 $c_{rs}$ 。

通过从小到大枚举多项式次数 $k$ , 误差逐渐减小, 可以找到最小的满足要求的 $k$ 。

### 牛顿法求解固定 $x, y$ 时的 $t, u$

固定 $x, y$ , 则方程组 (A.3) 变成了四元四次非线性方程组, 可以用牛顿迭代法求解。首先将项移动到方程左侧, 接着构造 $F(t, u, v, w)$ , 这是一个方程向量, 每一元素是一个非线性函数, 包含关于 $t, u, v, w$ 的非线性项和常数项。接着求出偏导数矩阵 $F'(t, u, v, w)$ , 这是一个 $4 \times 4$ 的矩阵, 每个元素是一个非线性函数。

迭代公式为:

$$\begin{pmatrix} t^{(k+1)} \\ u^{(k+1)} \\ v^{(k+1)} \\ w^{(k+1)} \end{pmatrix} = \begin{pmatrix} t^{(k)} \\ u^{(k)} \\ v^{(k)} \\ w^{(k)} \end{pmatrix} + \begin{pmatrix} \Delta t^{(k)} \\ \Delta u^{(k)} \\ \Delta v^{(k)} \\ \Delta w^{(k)} \end{pmatrix}, k = 0, 1, 2, \dots, n$$

其中 $(\Delta t^{(k)}, \Delta u^{(k)}, \Delta v^{(k)}, \Delta w^{(k)})^T$ 为以下方程组的解

$$F' \begin{pmatrix} t^{(k)} \\ u^{(k)} \\ v^{(k)} \\ w^{(k)} \end{pmatrix} \begin{pmatrix} \Delta t^{(k)} \\ \Delta u^{(k)} \\ \Delta v^{(k)} \\ \Delta w^{(k)} \end{pmatrix} = -F \begin{pmatrix} t^{(k)} \\ u^{(k)} \\ v^{(k)} \\ w^{(k)} \end{pmatrix}$$

这里需要实现线性方程组的求解, 可以使用列主元高斯消去法。

由题目要求, 取相对范数误差小于 $10^{-12}$ 为终止条件。

因为从表中观察到 $t, u$ 取值基本在 $[0, 2]$ 上, 因此, 选取 $(1, 1, 1, 1)$ 为迭代初始值, 能比较有效地减少迭代次数, 之后的观察也确实如此。

### 分片二元双二次插值求解 $\varphi(t, u)$

这里我们对原数表, 应用分片二元双二次插值法。令

$$t_m = 0.2m, u_n = 0.4n \quad (m = 0, 1, \dots, 5; n = 0, 1, \dots, 5)$$

对于给定的 $t, u$ , 找到最接近的三个插值点 $t_{i-1}, t_i, t_{i+1}$ 和 $u_{j-1}, u_j, u_{j+1}$ , 即

$$i = \begin{cases} \lfloor \frac{t}{0.2} \rfloor & (0.2 \leq t \leq 0.8) \\ 1 & (t < 0.2) \\ 4 & (t > 0.8) \end{cases}$$

$$j = \begin{cases} \lfloor \frac{u}{0.4} \rfloor & (0.4 \leq u \leq 1.6) \\ 1 & (u < 0.4) \\ 4 & (u > 1.6) \end{cases}$$

相应的插值多项式为

$$\varphi(t, u) = \sum_{k=i-1}^{i+1} \sum_{r=j-1}^{j+1} l_k(t) l_r(u) \varphi(t_k, u_r)$$

其中

$$l_k(t) = \prod_{q=i-1, q \neq k}^{i+1} \frac{t - t_q}{t_k - t_q} \quad (k = i-1, i, i+1)$$

$$l_r(u) = \prod_{q=j-1, q \neq k}^{j+1} \frac{u - u_q}{u_k - u_q} \quad (k = j-1, j, j+1)$$

得到 $\varphi(t, u)$ 的表达式之后，代入即可得到值。实际操作时一般是根据 $t, u$ 的值进行现场插值。此时即可求出

## 乘积型最小二乘曲面拟合

使用幂函数作为基矩阵，有基矩阵如下：

$$B = \begin{pmatrix} x_0^0 & \cdots & x_0^k \\ \vdots & \ddots & \vdots \\ x_i^0 & \cdots & x_i^k \end{pmatrix}, \quad G = \begin{pmatrix} y_0^0 & \cdots & y_0^k \\ \vdots & \ddots & \vdots \\ y_j^0 & \cdots & y_j^k \end{pmatrix}$$

数表矩阵为：

$$U = \begin{pmatrix} f(x_0, y_0) & \cdots & f(x_0, y_j) \\ \vdots & \ddots & \vdots \\ f(x_i, y_0) & \cdots & f(x_i, y_j) \end{pmatrix}$$

根据书P142，系数矩阵表达式为：

$$C = (B^T B)^{-1} B^T U G (G^T G)^{-1}$$

于是我们首先求解方程：

$$(B^T B) D = B^T U G$$

得到矩阵 $D$ 后求解

$$(G^T G)^T C^T = D^T$$

即可得到 $C$ ，这里可以继续使用之前的列主元高斯消去法，只是常量从向量变成了矩阵。

## 求解给定点的 $f$ 值与 $p$ 值

$f(x_i^*, y_j^*)$  的计算和  $f(x_i, y_j)$  相同。将  $(x_i^*, y_j^*)$  代入原方程组，求解相应  $(t_{ij}^*, u_{ij}^*)$ ，进行分片双二次插值求得  $f(x_i^*, y_j^*)$ 。  $p(x_i^*, y_j^*)$  只需代入上一步求得的二元多项式即可。

## 2. 源程序代码

### 简介

本程序采用C++17标准编写，模块清晰，通用性强，效率优秀，使用CMake (version  $\geq 3.16$ ) 组织代码，使用MSVC 8.1或GCC 9.3.0编译器均可编译通过。文件结构如下：

```
.
| CMakeLists.txt
| InterpolationUtil.cpp
| InterpolationUtil.h
| LinearEqUtil.cpp
| LinearEqUtil.h
| main.cpp
| Matrix.cpp
| Matrix.h
| NonLinearEqUtil.cpp
| NonLinearEqUtil.h
| NonLinFormula.cpp
| NonLinFormula.h
| NonLinItemMatrix.cpp
| NonLinItemMatrix.h
| NonSqMatrix.cpp
| NonSqMatrix.h
| Vector.cpp
| Vector.h
| ZeroRangeGuard.h
```

对于向量和矩阵的实现，分别在 `Vector` 和 `Matrix` 类中。另外因为曲面拟合的需要，实现了 `Matrix` 的友元类 `NonSqMatrix`，用于支持非方阵的矩阵运算，并且能在长宽相等时高效转换为 `Matrix`。

对于线性方程组的求解，实现在类 `LinearEqUtil` 中，在以往的基础上将单个常向量扩展为多个，这样就可以一次性求解系数矩阵相同的多个线性方程组，也即处理方程右侧为矩阵的情况。

我们实现了 `NonLinFormula` 和 `NonLinItemMatrix` 类，前者为代数式运算提供方便。后者主要用于牛顿迭代法，储存  $F$  和  $F'$ 。这两个类避免了对算式的硬编码，提高了程序的通用能力。

对于牛顿迭代法，被简洁地实现在了 `NonLinearEqUtil::solveByNewtonMethod` 中。

对于分片二元双二次插值算法求给定点的值，实现在了 `InterpolationUtil::twoDimQuadLagrangeInterpolation` 函数中。其步骤均以向量化的形式描述。

对于乘积型最小二乘曲面拟合算法，因为步骤简单，实现在了 `main.cpp` 中。在一个循环中不断尝试增长次数  $k$ ，直到达到误差要求为止，最后输出拟合的系数。

### 内容



### 3. 上机计算结果

以某次运行为例，得到结果：

数表( $x_i, y_i, f(x_i, y_i)$ )

.00E+00	.500E+00	.446504018481E+00
.00E+00	.550E+00	.324683262928E+00
.00E+00	.600E+00	.210159686683E+00
.00E+00	.650E+00	.103043608316E+00
.00E+00	.700E+00	.340189556268E-02
.00E+00	.750E+00	-.887358136380E-01
.00E+00	.800E+00	-.173371632750E+00
.00E+00	.850E+00	-.250534611467E+00
.00E+00	.900E+00	-.320276506388E+00
.00E+00	.950E+00	-.382668066110E+00
.00E+00	.100E+01	-.437795766738E+00
.00E+00	.105E+01	-.485758941444E+00
.00E+00	.110E+01	-.526667254884E+00
.00E+00	.115E+01	-.560638479797E+00
.00E+00	.120E+01	-.587796538768E+00
.00E+00	.125E+01	-.608269779090E+00
.00E+00	.130E+01	-.622189452876E+00
.00E+00	.135E+01	-.629688378186E+00
.00E+00	.140E+01	-.630899760003E+00
.00E+00	.145E+01	-.625956152545E+00
.00E+00	.150E+01	-.614988546609E+00
.80E-01	.500E+00	.638015226511E+00
.80E-01	.550E+00	.506611755147E+00
.80E-01	.600E+00	.382176369277E+00
.80E-01	.650E+00	.264863491154E+00
.80E-01	.700E+00	.154780200285E+00
.80E-01	.750E+00	.519926834909E-01
.80E-01	.800E+00	-.434680402049E-01
.80E-01	.850E+00	-.131601056789E+00
.80E-01	.900E+00	-.212431088309E+00
.80E-01	.950E+00	-.286004551058E+00
.80E-01	.100E+01	-.352386078979E+00
.80E-01	.105E+01	-.411655456522E+00
.80E-01	.110E+01	-.463904911519E+00
.80E-01	.115E+01	-.509236724701E+00
.80E-01	.120E+01	-.547761117962E+00
.80E-01	.125E+01	-.579594388339E+00
.80E-01	.130E+01	-.604857258890E+00
.80E-01	.135E+01	-.623673421332E+00
.80E-01	.140E+01	-.636168248413E+00
.80E-01	.145E+01	-.642467656690E+00
.80E-01	.150E+01	-.642697102700E+00
.16E+00	.500E+00	.840081395767E+00
.16E+00	.550E+00	.699764165673E+00
.16E+00	.600E+00	.566061442352E+00
.16E+00	.650E+00	.439171608118E+00
.16E+00	.700E+00	.319242138041E+00
.16E+00	.750E+00	.206376192387E+00
.16E+00	.800E+00	.100638523891E+00
.16E+00	.850E+00	.206074006784E-02

.16E+00	.900E+00	-.893540247670E-01
.16E+00	.950E+00	-.173626968865E+00
.16E+00	.100E+01	-.250799956160E+00
.16E+00	.105E+01	-.320932269445E+00
.16E+00	.110E+01	-.384097735005E+00
.16E+00	.115E+01	-.440382175418E+00
.16E+00	.120E+01	-.489881152313E+00
.16E+00	.125E+01	-.532697965534E+00
.16E+00	.130E+01	-.568941879292E+00
.16E+00	.135E+01	-.598726549515E+00
.16E+00	.140E+01	-.622168629750E+00
.16E+00	.145E+01	-.639386535697E+00
.16E+00	.150E+01	-.650499350788E+00
.24E+00	.500E+00	.105151509180E+01
.24E+00	.550E+00	.902927430831E+00
.24E+00	.600E+00	.760580266860E+00
.24E+00	.650E+00	.624715198146E+00
.24E+00	.700E+00	.495519756001E+00
.24E+00	.750E+00	.373134042775E+00
.24E+00	.800E+00	.257656748872E+00
.24E+00	.850E+00	.149150559410E+00
.24E+00	.900E+00	.476469867734E-01
.24E+00	.950E+00	-.468493232015E-01
.24E+00	.100E+01	-.134356760385E+00
.24E+00	.105E+01	-.214913344927E+00
.24E+00	.110E+01	-.288573700635E+00
.24E+00	.115E+01	-.355406364786E+00
.24E+00	.120E+01	-.415491396489E+00
.24E+00	.125E+01	-.468918249969E+00
.24E+00	.130E+01	-.515783883125E+00
.24E+00	.135E+01	-.556191075200E+00
.24E+00	.140E+01	-.590246930563E+00
.24E+00	.145E+01	-.618061548241E+00
.24E+00	.150E+01	-.639746839258E+00
.32E+00	.500E+00	.127124675148E+01
.32E+00	.550E+00	.111500201815E+01
.32E+00	.600E+00	.964607727216E+00
.32E+00	.650E+00	.820347369475E+00
.32E+00	.700E+00	.682447678179E+00
.32E+00	.750E+00	.551085208597E+00
.32E+00	.800E+00	.426392385902E+00
.32E+00	.850E+00	.308462995633E+00
.32E+00	.900E+00	.197357129692E+00
.32E+00	.950E+00	.931056208594E-01
.32E+00	.100E+01	-.428599223404E-02
.32E+00	.105E+01	-.948339252969E-01
.32E+00	.110E+01	-.178572990364E+00
.32E+00	.115E+01	-.255553779055E+00
.32E+00	.120E+01	-.325840150158E+00
.32E+00	.125E+01	-.389506988363E+00
.32E+00	.130E+01	-.446638204599E+00
.32E+00	.135E+01	-.497324951768E+00
.32E+00	.140E+01	-.541664032699E+00
.32E+00	.145E+01	-.579756479795E+00
.32E+00	.150E+01	-.611706288148E+00
.40E+00	.500E+00	.149832105248E+01
.40E+00	.550E+00	.133499863207E+01
.40E+00	.600E+00	.117712512374E+01



.40E+00	.650E+00	.102502405502E+01
.40E+00	.700E+00	.878960023174E+00
.40E+00	.750E+00	.739145108704E+00
.40E+00	.800E+00	.605744871487E+00
.40E+00	.850E+00	.478883861067E+00
.40E+00	.900E+00	.358650625882E+00
.40E+00	.950E+00	.245102236196E+00
.40E+00	.100E+01	.138268350928E+00
.40E+00	.105E+01	.381548654070E-01
.40E+00	.110E+01	-.552528211681E-01
.40E+00	.115E+01	-.141986880814E+00
.40E+00	.120E+01	-.222094439096E+00
.40E+00	.125E+01	-.295635232460E+00
.40E+00	.130E+01	-.362679511503E+00
.40E+00	.135E+01	-.423306164224E+00
.40E+00	.140E+01	-.477601036132E+00
.40E+00	.145E+01	-.525655426667E+00
.40E+00	.150E+01	-.567564743655E+00
.48E+00	.500E+00	.173189274038E+01
.48E+00	.550E+00	.156203457721E+01
.48E+00	.600E+00	.139721691821E+01
.48E+00	.650E+00	.123780100674E+01
.48E+00	.700E+00	.108408753268E+01
.48E+00	.750E+00	.936322772315E+00
.48E+00	.800E+00	.794704449054E+00
.48E+00	.850E+00	.659387198028E+00
.48E+00	.900E+00	.530487586840E+00
.48E+00	.950E+00	.408088685454E+00
.48E+00	.100E+01	.292244201230E+00
.48E+00	.105E+01	.182982206854E+00
.48E+00	.110E+01	.803084940354E-01
.48E+00	.115E+01	-.157904130516E-01
.48E+00	.120E+01	-.105344551621E+00
.48E+00	.125E+01	-.188398090610E+00
.48E+00	.130E+01	-.265007149319E+00
.48E+00	.135E+01	-.335237838904E+00
.48E+00	.140E+01	-.399164503887E+00
.48E+00	.145E+01	-.456868143302E+00
.48E+00	.150E+01	-.508434993278E+00
.56E+00	.500E+00	.197122178640E+01
.56E+00	.550E+00	.179532959950E+01
.56E+00	.600E+00	.162406711323E+01
.56E+00	.650E+00	.145783058271E+01
.56E+00	.700E+00	.129695464975E+01
.56E+00	.750E+00	.114171810545E+01
.56E+00	.800E+00	.992349533324E+00
.56E+00	.850E+00	.849032663329E+00
.56E+00	.900E+00	.711911352264E+00
.56E+00	.950E+00	.581094158922E+00
.56E+00	.100E+01	.456658513233E+00
.56E+00	.105E+01	.338654496139E+00
.56E+00	.110E+01	.227108255770E+00
.56E+00	.115E+01	.122025089193E+00
.56E+00	.120E+01	.233922196376E-01
.56E+00	.125E+01	-.688187019710E-01
.56E+00	.130E+01	-.154649344213E+00
.56E+00	.135E+01	-.234152666459E+00
.56E+00	.140E+01	-.307391091913E+00

.56E+00	.145E+01	-.374434862348E+00
.56E+00	.150E+01	-.435360556536E+00
.64E+00	.500E+00	.221566786369E+01
.64E+00	.550E+00	.203420113361E+01
.64E+00	.600E+00	.185695514362E+01
.64E+00	.650E+00	.168435816416E+01
.64E+00	.700E+00	.151677635240E+01
.64E+00	.750E+00	.135451904115E+01
.64E+00	.800E+00	.119784408667E+01
.64E+00	.850E+00	.104696304942E+01
.64E+00	.900E+00	.902046083802E+00
.64E+00	.950E+00	.763226477663E+00
.64E+00	.100E+01	.630604821954E+00
.64E+00	.105E+01	.504252814597E+00
.64E+00	.110E+01	.384216715546E+00
.64E+00	.115E+01	.270520476641E+00
.64E+00	.120E+01	.163168572400E+00
.64E+00	.125E+01	.621485581168E-01
.64E+00	.130E+01	-.325666193968E-01
.64E+00	.135E+01	-.121016534844E+00
.64E+00	.140E+01	-.203251399623E+00
.64E+00	.145E+01	-.279330359558E+00
.64E+00	.150E+01	-.349319957540E+00
.72E+00	.500E+00	.246468422266E+01
.72E+00	.550E+00	.227805897940E+01
.72E+00	.600E+00	.209525125084E+01
.72E+00	.650E+00	.191671812800E+01
.72E+00	.700E+00	.174285462878E+01
.72E+00	.750E+00	.157399842733E+01
.72E+00	.800E+00	.141043483523E+01
.72E+00	.850E+00	.125240175061E+01
.72E+00	.900E+00	.110009440963E+01
.72E+00	.950E+00	.953669851261E+00
.72E+00	.100E+01	.813251055249E+00
.72E+00	.105E+01	.678930742966E+00
.72E+00	.110E+01	.550774848504E+00
.72E+00	.115E+01	.428825676973E+00
.72E+00	.120E+01	.313104771740E+00
.72E+00	.125E+01	.203615514033E+00
.72E+00	.130E+01	.100345478241E+00
.72E+00	.135E+01	.326856518657E-02
.72E+00	.140E+01	-.876530659133E-01
.72E+00	.145E+01	-.172467247819E+00
.72E+00	.150E+01	-.251230220752E+00
.80E+00	.500E+00	.271781110947E+01
.80E+00	.550E+00	.252639950126E+01
.80E+00	.600E+00	.233841138686E+01
.80E+00	.650E+00	.215432937728E+01
.80E+00	.700E+00	.197457455665E+01
.80E+00	.750E+00	.179951057910E+01
.80E+00	.800E+00	.162944822055E+01
.80E+00	.850E+00	.146465004375E+01
.80E+00	.900E+00	.130533496765E+01
.80E+00	.950E+00	.115168262131E+01
.80E+00	.100E+01	.100383741991E+01
.80E+00	.105E+01	.861912337228E+00
.80E+00	.110E+01	.725992371111E+00
.80E+00	.115E+01	.596137711520E+00

.80E+00	.120E+01	.472386627914E+00
.80E+00	.125E+01	.354758095898E+00
.80E+00	.130E+01	.243254184181E+00
.80E+00	.135E+01	.137862222525E+00
.80E+00	.140E+01	.385567703264E-01
.80E+00	.145E+01	-.546985959345E-01
.80E+00	.150E+01	-.141949659709E+00

$k$ 值和 $\delta$

```

k = 1 delta = .322090897364E+01
k = 2 delta = .465996003327E-02
k = 3 delta = .172117537914E-03
k = 4 delta = .330953429925E-05
k = 5 delta = .254137771997E-07

```

当 $k = 5$ 时, 精度已经达到了要求, 此时 $p(x, y)$ 中的系数 $c_{rs}$ 为

.202123044262E+01	-.366842636522E+01	.709247679136E+00
.848606392858E+00	-.415897940397E+00	
.319191902005E+01	-.741167262380E+00	-.269700015696E+01
.163105207331E+01	-.484652918264E+00	
.256783376068E+00	.158052371105E+01	-.464732264229E+00
-.799465014205E-01	.101380257369E+00	
-.268881423675E+00	-.732402030864E+00	.108086081306E+01
-.811992184212E+00	.305416495777E+00	
.216915251451E+00	-.175276128774E+00	-.791840695272E-01
.250488511538E+00	-.144991789392E+00	

数表 $(x_i^*, y_j^*, f(x_i^*, y_j^*), p(x_i^*, y_j^*))$

.10E+00	.700E+00	.194720407918E+00	.194730357326E+00
.10E+00	.900E+00	-.183036970312E+00	-.183041838360E+00
.10E+00	.110E+01	-.445497629269E+00	-.445500042115E+00
.10E+00	.130E+01	-.597566705305E+00	-.597558856878E+00
.10E+00	.150E+01	-.646459593647E+00	-.646446111301E+00
.20E+00	.700E+00	.405979189288E+00	.405989539896E+00
.20E+00	.900E+00	-.225158150340E-01	-.225211162936E-01
.20E+00	.110E+01	-.338220789483E+00	-.338224022407E+00
.20E+00	.130E+01	-.544437827296E+00	-.544430450939E+00
.20E+00	.150E+01	-.647361337978E+00	-.647348010632E+00
.30E+00	.700E+00	.634777195151E+00	.634787453107E+00
.30E+00	.900E+00	.158801342082E+00	.158796295645E+00
.30E+00	.110E+01	-.207365658661E+00	-.207368580088E+00
.30E+00	.130E+01	-.465357900500E+00	-.465349923350E+00
.30E+00	.150E+01	-.620270952047E+00	-.620257138501E+00
.40E+00	.700E+00	.878960023174E+00	.878969865343E+00
.40E+00	.900E+00	.358650822218E+00	.358646043348E+00
.40E+00	.110E+01	-.552527776174E-01	-.552554368705E-01
.40E+00	.130E+01	-.362679502890E+00	-.362671062971E+00
.40E+00	.150E+01	-.567564742128E+00	-.567550582812E+00
.50E+00	.700E+00	.113661091016E+01	.113662035311E+01

.50E+00	.900E+00	.574980553226E+00	.574975843087E+00
.50E+00	.110E+01	.115992426843E+00	.115989321215E+00
.50E+00	.130E+01	-.238568293365E+00	-.238560419182E+00
.50E+00	.150E+01	-.491434391617E+00	-.491420900990E+00
.60E+00	.700E+00	.140604179891E+01	.140605068696E+01
.60E+00	.900E+00	.805941716346E+00	.805937302046E+00
.60E+00	.110E+01	.304429275827E+00	.304425831955E+00
.60E+00	.130E+01	-.950161177620E-01	-.950089457219E-01
.60E+00	.150E+01	-.393902305233E+00	-.393889837754E+00
.70E+00	.700E+00	.168578351531E+01	.168579121747E+01
.70E+00	.900E+00	.104988138137E+01	.104987773885E+01
.70E+00	.110E+01	.508293841773E+00	.508291045085E+00
.70E+00	.130E+01	.661488102960E-01	.661563554850E-01
.70E+00	.150E+01	-.276834338871E+00	-.276822042898E+00
.80E+00	.700E+00	.197457455665E+01	.197458126116E+01
.80E+00	.900E+00	.130533519966E+01	.130533200412E+01
.80E+00	.110E+01	.725992430530E+00	.725989310430E+00
.80E+00	.130E+01	.243254198540E+00	.243260790494E+00
.80E+00	.150E+01	-.141949656522E+00	-.141938789062E+00

经过与他人的对比验证，可知结果正确。

此外，在Release模式下，无输出时进行时间测量，基本上在2000~3000微秒之间。

## 4. 讨论分析

在本次程序设计中，主要面临的困难和思考有如下方面：

1. 步骤较多，需要仔细整理出合理的逻辑。
2. 对于牛顿迭代法的初值一开始没有很好的把握，一开始采用正太随机法生成初始向量，发现虽然能迭代收敛，但需要步数较多，有时可能要500步尚未迭代完毕。在参考了他人的思路之后，意识到题目中给出了 $t, u$ 的大致范围，因此，可以为向量初始值设定为 $(1, 1, 1, 1)$ ，这样迭代次数便稳定下来，基本6次即可达到所需精度。
3. 如何在程序的高效性，可复用性，简洁性做平衡，是很困难的。通过反复的设计和抽象，使得我的代码水平有了进一步提升。