# 计算物理第一次作业

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## 1. 数据误差的避免

(a)

• 近考虑舍入误差计算机对两个数的相加可以表示为

$$a\oplus b=(a+b)(1+\delta), |\delta|<\epsilon_M/2$$

• 记机器计算的平均值为  $\overline{x}'$ 

$$\overline{x}' = rac{1}{N}(((x_1+x_2)(1+\delta_1)+x_3)(1+\delta_2)+...)$$

忽略高阶小量,有

$$egin{aligned} \overline{x}' &= rac{1}{N} (\Sigma_{i=1}^N x_i + (x_1 + x_2) \delta_1 + (x_1 + x_2 + x_3) \delta_2 + ... + \Sigma_{i=1}^N x_i \delta_N) \ &= \overline{x} + rac{1}{N} ((x_1 + x_2) \delta_1 + (x_1 + x_2 + x_3) \delta_2 + ... + \Sigma_{i=1}^N x_i \delta_N) \end{aligned}$$

• 利用 $|\delta_i| < \epsilon_M/2$ ,相对误差可以表达为

$$egin{aligned} rac{\overline{x}' - \overline{x}}{\overline{x}} & \leq rac{\epsilon_M}{2\Sigma_{i=1}^N x_i} ((N-1)(x_1 + x_2) + (N-2)x_3 + (N-3)x_4 + ...x_N) \ & = rac{\epsilon_M}{2\Sigma_{i=1}^N x_i} ((N-1)\Sigma_{i=1}^N x_i - x_3 - 2x_4 - (N-2)x_N) \ & pprox rac{1}{2} (N+1)\epsilon_M \end{aligned}$$

● 注:以上仅考虑无符号数,否则无上限

(b)

● 对第一种方法,乘法运算带来的舍入误差也可以表示为

$$a \bigstar b = ab(1+\delta)$$

先计算第一项,用与(a)中相同的方法,忽略一阶以上的高阶小量第一项最大的误差可以 估计为

$$\Delta_{11} = (N+1)\epsilon_M \Sigma_{i=1}^N x_i^2$$

○ 对第二项、舍入误差可估计为(a)的两倍

$$\Delta_{12} = (N+1)\epsilon_M(\Sigma_{i=1}^N x_i)^2/N^2 pprox \epsilon_M(\Sigma_{i=1}^N x_i)^2/N$$

。 考虑两项做差的舍入误差, 第一钟方法的舍入误差可以估计为

$$\Delta_1pprox \epsilon_M((\Sigma_{i=1}^Nx_i)^2+(\Sigma_{i=1}^Nx_i)^2/N+S^2)$$

- - 。 对于一次减法运算

$$x_i \nabla \overline{x} = (x_i - \overline{x}(1+\delta))(1+\delta_i) \approx (x_i - \overline{x})(1+\delta_i) - \delta \overline{x}$$

。 因此

$$(x_i-\overline{x})(x_i-\overline{x})=(x_i-\overline{x})(1+\delta_i)-\delta\overline{x})(x_i-\overline{x})(1+\delta_i)-\delta\overline{x})(1+\epsilon_i)\ =((x_i-\overline{x})^2(1+2\delta_i)-2\delta\overline{x}(x_i-\overline{x}))(1+\epsilon_i)\ =(x_i-\overline{x})^2(1+\epsilon_i+2\delta_i)-\delta\overline{x}(x_i-\overline{x})\leq (x_i-\overline{x})^2rac{3}{2}\epsilon_M-rac{1}{2}\epsilon_M\overline{x}(x_i-\overline{x})$$

 $\circ$  求和的运算与(a)相同, $\overline{x}$ 的效应被抵消了

$$\Delta_2 pprox rac{3}{2}(N+1)\epsilon_M S^2$$

• 可以看出来,第一种方法的误差上限可以非常大,因此第二种方法好。

(c)

● 首先显然有

$$\int_{0}^{1} \frac{1}{x+5} dx = \ln(\frac{6}{5})$$

● 其次

$$I_k + 5I_{k-1} = \int_0^1 rac{x^k + 5x^{k-1}}{x+5} = \int_0^1 x^{k-1} dx = rac{1}{k}$$

• 因此满足递推公式。考虑误差后,按照递推公式有

$$I_k + 5I_{k-1} + \epsilon_k + 5\epsilon_{k-1} = rac{1}{k}$$

$$\therefore \epsilon_k = -5\epsilon_{k-1}$$

$$|\epsilon_k| = |5|^k |\epsilon|$$

● 不稳定性指数扩大,因此该算法是不稳定的

## 2. 矩阵的模与条件数

(a)

● 只要将矩阵按第一列展开就可以发现,子矩阵的行列式不变,因此该矩阵的行列式与平凡矩阵 (1)的行列式相同,即

$$det = 1$$

(b)

• 用数学归纳法,首先

$$A_1^{-1} = (1)_{1 imes 1}$$

● 假设第n阶矩阵的可以写为

$$A_n^{-1}=egin{pmatrix} lpha & eta \ \gamma & A_{n-1}^{-1} \end{pmatrix} \ A_nA_n^{-1}=egin{pmatrix} lpha-\Sigma\gamma_i & eta+(-1,...,-1)A_{n-1}^{-1} \ A_{n-1}\gamma & I_{n-1} \end{pmatrix}=I_n$$

• 容易判断 $\gamma_i = 0$ 以及 $\alpha = 1$ ,右上角的等式可以写为

$$eta_i = \Sigma_j A_{n-1_{i,j}}^{-1}$$

• 于是最后的矩阵形式为

$$A_n^{-1} = egin{pmatrix} 1 & 2^0 & 2^1 & ... & 2^{n-2} \ 0 & 1 & 2^0 & ... & 2^{n-3} \ 0 & 0 & 1 & ... & 2^{n-4} \ ... & ... & ... & ... \ 0 & 0 & 0 & ... & 1 \end{pmatrix}$$

(c)

• 对于矢量,在 $p \to \infty$ 时给出的是最大模,因此

$$||A||_{\infty} = sup_{x=\emptyset}rac{max_i|\Sigma_{j=1}^na_{ij}x_j|}{max_n|x_n|} \ rac{|\Sigma_{j=1}^na_{ij}x_j|}{max_n|x_n|} \leq \Sigma_{j=1}^n|a_{ij}|rac{|x_j|}{max_n|x_n|} \leq \Sigma_{j=1}^n|a_{ij}|$$

• 注意到当 $x_j$ 与 $a_{ij}$ 符号相同时,等号是可以取到的

$$| \therefore ||A||_{\infty} = max_i \Sigma_{j=1}^n |a_{ij}|$$

(d)

● 对幺正矩阵

$$egin{aligned} UU^\dagger &= I \ ||Ux||_2 = (x^\dagger U^\dagger U x)^{1/2} = ||x||_2 \ dots \, ||U||_2 = sup_{x=\emptyset} rac{||Ux||_2}{||x||_2} = 1 \end{aligned}$$

同理

$$||U^\dagger||_2=1$$

• 对任意的A满足条件,同样地

$$||UA||_2 = sup_{x=\emptyset} rac{||UAx||_2}{||x||_2} = sup_{x=\emptyset} rac{||Ax||_2}{||x||_2} = ||A||_2$$

(e)

$$||A||_{\infty}=n$$
  $||A^{-1}||_{\infty}=2^{n-1}$   $K_{\infty}(A)=2^{n-1}n$ 

## 3.Hilbert矩阵

(a)

$$egin{aligned} rac{\partial D}{\partial c_i} &= \int_0^1 2(P_n(x) - f(x)) x^{i-1} dx = 0 \ &\therefore \int_0^1 x^{i-1} P_n dx = \int_0^1 x^{i-1} f(x) dx \ &P_n &= \sum_{j=1}^n c_j x^{j-1} \ &\sum_{j=1}^n \int_0^1 c_j x^{i+j-2} dx = \int_0^1 x^{i-1} f(x) dx \end{aligned}$$

$$\Sigma_{j=1}^n rac{1}{i+j-1} = \int_0^1 x^{i-1} f(x) dx$$

因此可以写成

$$H_nC=B, (H_n)_{ij}C_j=B_i$$
  $B_i=\int_0^1 x^{i-1}f(x)dx$   $(H_n)_{ij}=rac{1}{i+j-1}$ 

(b)

$$c^T H_n c = \int_0^1 (\Sigma_{ij} c_i x^{i-1} c_j x^{j-1}) dx \ = \int_0^1 (\Sigma_i c_i x^{i-1})^2 dx \geq = 0$$

当且仅当对所有 $\mathbf{i}$ , $c_i=0$ 时等号成立。从上述可以看出, $H_n$ 是对称正定的矩阵。

(c)

• 取对数后

$$egin{aligned} ln(det(H_n)) &= 4lnc_n - lnc_{2n} \ &= 4ln\prod_{i=1}^{n-1}i! - ln\prod_{i=1}^{2n-1}i! \ &= 4\Sigma_{i=1}^{n-1}(n-i)ln(i) - \Sigma_{i=1}^{2n-1}(2n-i)ln(i) \end{aligned}$$

• 编写python3程序 calculate.py, 进行计算计算结果为

n	$det(H_n)$
1	1.00000000000000
2	0.083333333333333
3	0.000462962962962
4	1.653439153439155e-07
5	3.749295132515107e-12
6	5.367299887358753e-18
7	4.835802623926094e-25

```
8 2.737050113791512e-33
```

9 9.720234311924803e-43

10 2.164179226431516e-53

(d)

● 编写python3程序 Solve.py, 取n=1-12的情况, 求解结果详见附件

### 附件

### Caculate.py

```
import numpy as np
def det(n):
    sum1 = 0
    sum2 = 0
    for i in range(1,n):
        sum1 = sum1 + (n-i)*np.log(i)
    for i in range(1,2*n):
        sum2 = sum2 + (2*n-i)*np.log(i)
    sum = 4.0 * sum1 - sum2
    det = np.exp(sum)
    return det
file1 = open(r'HW_1_source_code/det.txt','w')
for i in range(1,11):
    de = det(i)
    string = "|{}|".format(i)
    s = str(de)
    file1.write(string)
    file1.write(s)
    file1.write("| \n")
file1.close()
```

### Solve.py

```
import numpy as np
import sys
def GEM(A,b):
    rows = A.shape[0]
    if rows != len(b):
        print("incompatible matrix A and b")
        exit()
    for i in range(0,rows-1):
        # select the pivot by the max
        firstcol=[]
        for ii in range(i,rows):
            firstcol.append(A[ii][i])
        pivot = firstcol.index(max(firstcol)) + i
        A[[i,pivot],:] = A[[pivot,i],:]
        b[i],b[pivot] = b[pivot],b[i]
        for j in range(i+1,rows):
            if A[j,i]!=0.0:
                fro = A[j,i] / A[i,i]
                A[j,i] = 0.0
                A[j,i+1:rows] = - fro * A[i,i+1:rows] + A[j,i+1:rows]
                b[j] = - fro * b[i] + b[j]
    x = np.zeros(rows)
    for k in range(rows-1,-1,-1):
        x[k] = (b[k] - np.dot(A[k,k+1:rows],x[k+1:rows]))/A[k,k]
    return x
def choleskey(A,b):
    rows, cols = A. shape
    if rows != len(b):
        print("incompatible matrix A and b")
        exit(0)
    if rows != cols:
```

```
print("not correct matirx!")
        exit(0)
   H = np.zeros((rows,cols))
    H[0,0] = np.sqrt(A[0][0])
    for i in range(1, rows):
        for j in range (0,i):
            H[i,j] = (A[i,j] - np.dot(H[i,0:j],H[j,0:j]))/H[j,j]
        H[i,i] = np.sqrt(A[i,i] - np.dot(H[i,0:i],H[i,0:i]))
    x1 = np.zeros(rows)
    x2 = np.zeros(cols)
    for m in range(rows):
        x1[m] = (b[m] - np.dot(H[m,0:m],x1[0:m]))/H[m,m]
    for n in range(rows-1,-1,-1):
        x2[n] = (x1[n] - np.dot(H[n+1:rows,n],x2[n+1:rows]))/H[n,n]
    return x2
def create_hilbert(n):
    mat = []
    for i in range(1,n+1):
        line = []
        for j in range(1,n+1):
            h = 1.0 / (i+j-1)
            line.append(h)
        mat.append(line)
    hilbert = np.array(mat)
    return hilbert
def create_vector(n):
    return np.array([1 for i in range (1,n+1)],dtype=np.float)
file1 = open(r'HW_1_source_code/GEM.txt','w')
file2 = open(r'HW_1_source_code/Choleskey.txt','w')
for n in range (1,13):
    A1 = create_hilbert(n)
```

```
b1 = create_vector(n)
A2 = create_hilbert(n)
b2 = create_vector(n)

result_1 = GEM(A1,b1)
result_2 = choleskey(A2,b2)
string = "n = {} \n".format(n)

file1.write(string)
r1 = str(result_1) + "\n \n"
file1.write(str(r1))

file2.write(string)
r2 = str(result_2) + "\n \n"
file2.write(str(r2))
```

### 运行结果

#### Cholesky

```
n = 1
[1.]

n = 2
[-2. 6.]

n = 3
[ 3. -24. 30.]

n = 4
[ -4. 60. -180. 140.]

n = 5
[ 5. -120. 630. -1120. 630.]

n = 6
[ -6.000e+00 2.100e+02 -1.680e+03 5.040e+03 -6.300e+03 2.772e+03]

n = 7
[ 7.00000001e+00 -3.360000000e+02 3.78000000e+03 -1.68000000e+04 3.46500000e+04
```

```
-3.32640000e+04 1.20120000e+04]
```

n = 8

[-7.99999935e+00 5.03999963e+02 -7.55999950e+03 4.61999972e+04 -1.38599992e+05 2.16215989e+05 -1.68167992e+05 5.14799977e+04]

n = 9

[ 8.99992463e+00 -7.19994699e+02 1.38599090e+04 -1.10879343e+05 4.50447568e+05 -1.00900300e+06 1.26125422e+06 -8.23676489e+05 2.18789129e+05]

n = 10

[-9.99647239e+00 9.89693166e+02 -2.37534315e+04 2.40180050e+05 -1.26097314e+06 3.78298940e+06 -6.72542008e+06 7.00002152e+06 -3.93755829e+06 9.23634288e+05]

n = 11

[ 1.09013685e+01 -1.30959927e+03 3.83385824e+04 -4.77430486e+05 3.13490714e+06 -1.20437288e+07 2.84479818e+07 -4.18155209e+07 3.72530997e+07 -1.84014238e+07 3.86519637e+06]

n = 12

[-1.06586186e+01 1.54967606e+03 -5.49242901e+04 8.32010641e+05 -6.70895715e+06 3.21428166e+07 -9.69534472e+07 1.88838275e+08 -2.36982407e+08 1.84952571e+08 -8.16237546e+07 1.55564209e+07]

#### **GEM**

n = 1

[1,]

n = 2

[-2. 6.]

n = 3

[3. -24. 30.]

n = 4

[ -4. 60. -180. 140.]

n = 5

[5. -120. 630. -1120. 630.]

n = 6

[-6.000e+00 2.100e+02 -1.680e+03 5.040e+03 -6.300e+03 2.772e+03]

n = 7

[7.00000004e+00 -3.36000002e+02 3.78000002e+03 -1.68000001e+04 3.46500001e+04

-3.32640001e+04 1.20120000e+04]

n = 8

[-7.99999982e+00 5.03999988e+02 -7.55999982e+03 4.61999989e+04 -1.38599997e+05 2.16215995e+05 -1.68167996e+05 5.14799990e+04]

n = 9

[ 8.99994489e+00 -7.19996133e+02 1.38599337e+04 -1.10879522e+05 4.50448234e+05 -1.00900437e+06 1.26125581e+06 -8.23677458e+05 2.18789370e+05]

n = 10

[-9.99807268e+00 9.89833697e+02 -2.37564615e+04 2.40207859e+05 -1.26110680e+06 3.78335912e+06 -6.72602984e+06 7.00061336e+06 -3.93787017e+06 9.23703099e+05]

n = 11

[ 1.09475741e+01 -1.31447626e+03 3.84658805e+04 -4.78860520e+05 3.14345886e+06 -1.20738884e+07 2.85138169e+07 -4.19054647e+07 3.73279469e+07 -1.84361073e+07 3.87205691e+06]

n = 12

[-1.03948050e+01 1.51178142e+03 -5.36181293e+04 8.12917880e+05 -6.56096500e+06 3.14624032e+07 -9.49849286e+07 1.85160064e+08 -2.32551511e+08 1.81630379e+08 -8.02138627e+07 1.52977616e+07]