## UNIVERSITY OF SUSSEX

Scientific Computing
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Problem Sheet 2 (Problem 2 will be assessed)

Deadline: 12pm on Wednesday, October 30th, 2013.

Penalties will be imposed for submissions beyond this date.

Final submission date: Thursday, October 31st, 2013.

No submissions will be accepted beyond this date.

1. Consider the *n* simultaneous equations  $\mathbf{A}\mathbf{x} = \mathbf{b}$ , where

$$A_{ij} = (i+j)^2, \ b_i = \sum_{j=0}^{n-1} A_{ij}, \ i = 0, 1, ..., n-1, \ j = 0, 1, ..., n-1$$

Clearly, the solution is  $\mathbf{x} = (1, 1, \dots, 1)^T$ . Write a program that solves these equations for any given n using both the provided code GaussPivot and the internal solve routine. Run your program with n = 2, 3, and 4 and comment on the results.

**Solution:** We first import the required modules and start the main for loop over the matrix sizes to be done:

```
from numpy import *
from numpy.linalg import *
from gaussPivot import *
import pylab
for n in range(2,5):
```

Next, we have to setup the matrix A and right hand side vector b, e.g. as follows:

```
A=zeros((n,n))

for i in range(n):
    for j in range(n):
        A[i][j]=(1.0*i+j)**2

print h
x0=ones(n)
b=dot(h,x0)
```

## Important: make sure arrays are filled with floats, not integers!

Then, the solution is given simply by

```
x1=solve(A,b)
x2=gaussPivot(A,b)
error1[n-2]=max(x1-x0)
error2[n-2]=max(x2-x0)
print(n,error1[n-2], error2[n-2])
```

The result is:

```
[[ 0. 1.]
 [1.4.]]
(2, 0.0, 0.0)
              4.]
[[ 0.
         1.
 [ 1.
         4.
              9.]
         9.
             16.]]
(3, 6.6613381477509392e-16, 0.0)
[[ 0.
              4.
                   9.]
 [ 1.
         4.
              9.
                  16.]
 Γ 4.
         9.
                  25.]
             16.
 [ 9.
        16.
             25.
                  36.]]
Matrix is singular
```

Clearly, the errors for n=2 and 3 are quite small, but for n=4 the matrix becomes singular, i.e. the 4 equations are not all independent.

2. Use the built-in NumPy linalg function vander(v) to create the  $n \times n$  Vandermonde matrix:

$$V_{m,n} = \begin{pmatrix} v_0^n & v_0^{n-1} & \cdots & v_0 & 1\\ v_1^n & v_1^{n-1} & \cdots & v_1 & 1\\ \vdots & \vdots & \ddots & \vdots & \vdots\\ v_n^n & v_n^{n-1} & \cdots & v_n & 1 \end{pmatrix}$$
(1)

where  $v = (v_0, v_1, \dots, v_n)$  is a vector.

- (a) Use this to create an  $n \times n$  matrix A with v = (1, 2, 3, ..., n) for n=5 to 27.
- (b) For this matrix A construct a vector b so that the solution to Ax = b is  $x = (1, 1, ..., 1)^T$ .
- (c) Test your construction by applying the internal Numpy function solve(A,b) to compute x. What is the largest element-by-element error for each n? Plot the decimal log of the max element-by-element error vs. n. Comment on the results. Why do you think this happens?

[40]

## Solution:

We first import the required modules, numpy, and numpy.linalg:

```
from numpy import *
from numpy.linalg import *
import pylab

xx=zeros(23) # we will store here the errors for plotting
error=zeros(23)
```

Next, we setup a for loop for the required range of values and calculate the Vandermonde matrix, the b vector which corresponds to a solution of ones and then call the solution routines:

```
pylab.semilogy(xx,error) #plot errors
pylab.show()
```

## Result is

- 5 1.73860925656e-13
- 6 2.97051272469e-12
- 7 5.20652410074e-11
- 8 7.55688844833e-11
- 9 4.83710294041e-08
- 10 2.71211799774e-06
- 11 4.40419572507e-05
- 12 0.000300584294233
- 13 0.0179439816366
- 14 6.98573835649
- 15 86.7835673439
- 16 59212.9703732
- 17 14488.7032158
- 18 91767.8664893
- 19 821035.517948
- 20 156770.4008
- 21 43018.0330413
- 22 552355.387253
- 23 790748.359786
- 24 2957109.1522
- 25 76330146.2547
- 26 173012589.13
- 27 19356424864.0

plotted in the Figure below. Errors rise very fast with n and for n > 13 become larger than the solution itself, i.e. results become meaningless.

