

```
In [1]: %matplotlib notebook
from numpy import *
from matplotlib.pyplot import *
```

## Homework 10 : Conditioning and stability of linear least squares

The least squares problem

$$A\mathbf{x} = \mathbf{b} \quad (1)$$

where  $A \in \mathcal{R}^{m \times n}$ ,  $m \geq n$  has four associated "conditioning" problems, described in the table in Theorem 18.1 of TB (page 131). These are

1. Sensitivity of  $\mathbf{y} = A\mathbf{x}$  to right hand side vector  $\mathbf{b}$ ,
2. Sensitivity of the solution  $\mathbf{x}$  to right hand side vector  $\mathbf{b}$ ,
3. Sensitivity of  $\mathbf{y} = A\mathbf{x}$  to the coefficient matrix  $A$ , and
4. Sensitivity of the solution  $\mathbf{x}$  to the coefficient matrix  $A$ .

### Problem 1

**Sensitivity of  $\mathbf{y}$  to a perturbation in  $\mathbf{b}$ .**

In TB Lecture 12, the relative condition number is defined as

$$\kappa = \sup_{\delta x} \left( \frac{\|\delta f\|}{\|f(x)\|} \bigg/ \frac{\|\delta x\|}{\|x\|} \right) \quad (2)$$

#### Problem 1(a)

Arguing directly from this definition, establish the condition number of  $\mathbf{y}$  with respect to perturbations in  $\mathbf{b}$  given by TB Lecture 18

$$\kappa = \frac{1}{\cos \theta} \quad (3)$$

**Hint:** The input " $x$ " in this problem is  $\mathbf{b}$  and the output (or model) " $f$ " is  $\mathbf{y}$ . Show geometrically that the supremum is attained with  $P\delta b = \delta b$ .

#### Problem 1(b)

For  $\theta = \pi/2$ , the condition number is  $\infty$ . Illustrate what this means by considering the least squares problem

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} [x] = \begin{bmatrix} -1 \\ 2 \end{bmatrix} \quad (4)$$

Use the results in TB 11.11 and 11.12 (page 82) to determine the projection operator  $P$  for this problem. Then compute  $\mathbf{y} = P\mathbf{b}$  and show that  $P\mathbf{b} = 0$ . Find a perturbation  $\delta\mathbf{b}$  so that  $P\delta\mathbf{b} = \delta\mathbf{b} = \delta\mathbf{y} \neq 0$ . Explain what a condition number  $\kappa = \infty$  might mean here. Illustrate your argument graphically.

### Problem 1(c)

Now consider the problem

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} [x] = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \quad (5)$$

For this problem, show that  $\kappa = 1$ . What is qualitatively different about this problem than the problem in which  $\kappa = \infty$ ?

## Problem 2

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Problem 18.1 in TB (page 136)

## Problem 3

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Show that if  $(\lambda, \mathbf{v})$  is an eigenvalue/eigenvector pair for matrix  $A$ , then  $((\lambda - \mu)^{-1}, \mathbf{v})$  is an eigenvalue/eigenvector pair for the matrix  $(A - \mu I)^{-1}$ .

Why is this observation useful when using the power iteration to find an eigenvalue close to  $\mu$ ?

## Problem 4

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Exercise 29.1 (Lecture 29, TB page 223). This is a five part problem that asks you to code an eigenvalue solver for a real, symmetric matrix using the shifted  $QR$  algorithm. Do your code in Python, using the Numpy `qr` algorithm where needed.

The basic steps are :

1. Reduce your matrix  $A$  to tridiagonal form. You may use the hessenberg code we wrote in class.

2. Implement the unshifted  $QR$  code (also done in class). Use the Numpy routine `qr`. Your iteration should stop when the off diagonal elements are smaller (in absolute value) than  $\tau \approx 10^{-12}$ .
3. Find all eigenvalues of a matrix  $A$  using the "deflation" idea described in Algorithm 28.2.
4. Introduce the Wilkinson shift, described in Lecture 29.

## Notes

- Your code should work for a real, symmetric matrix
- Your code does not have to be efficient in the sense of optimizing the cost of matrix/vector multiplies and so on.
- Apply your algorithm to the Hilbert matrix `scipy.linalg.hilbert`. The entries of the  $m \times m$  Hilbert matrix are given by

$$H_{ij} = \frac{1}{i+j-1}, \quad i, j = 1, 2, \dots, m \quad (6)$$

```
In [11]: def display_mat(msg,A):
          print(msg)
          fstr = {'float' : "{:>10.6f}".format}
          with printoptions(formatter=fstr):
              display(A)
          print("")
```

```
In [12]: from scipy.linalg import hilbert

H = hilbert(5)
display_mat("Hilbert matrix : ", H)
```

```
Hilbert matrix :
array([[ 1.000000,  0.500000,  0.333333,  0.250000,  0.200000],
       [ 0.500000,  0.333333,  0.250000,  0.200000,  0.166667],
       [ 0.333333,  0.250000,  0.200000,  0.166667,  0.142857],
       [ 0.250000,  0.200000,  0.166667,  0.142857,  0.125000],
       [ 0.200000,  0.166667,  0.142857,  0.125000,  0.111111]])
```

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In [ ]:
```