# Worksheet 4

# 1 Norms Equivalency

Two norms in a finite-dimensional linear space X (e.g.:  $\mathbb{R}^n$ ),  $\|\cdot\|_a$  and  $\|\cdot\|_b$  are called equivalent if there is a constant c such that for all x in X,

$$\|x\|_a \le c\|x\|_b, \qquad \|x\|_b \le c\|x\|_a.$$
 (1)

#### 1.1

Suppose  $\|\cdot\|_a$  and  $\|\cdot\|_b$  are equivalent, and we know that an algorithm produces a sequence of vectors  $\{x_n\}_{n\geq 1}$ ,  $\|x_n\|_a \to 0$  as  $n\to\infty$ . What could we conclude about  $\|x_n\|_b$ 's behavior for  $n\to\infty$ ?

#### 1.2

Suppose  $\|\cdot\|_a$  and  $\|\cdot\|_b$  are equivalent,  $\|\cdot\|_b$  and  $\|\cdot\|_c$  are equivalent. Are  $\|\cdot\|_a$  and  $\|\cdot\|_c$  equivalent?

#### 1.3

We first show that the vector norms on  $\mathbb{R}^n$ ,  $\|\cdot\|_2$  and  $\|\cdot\|_{\infty}$ , are equivalent. To do this prove the inequality:

$$\|\boldsymbol{x}\|_{\infty} \leq \|\boldsymbol{x}\|_{2} \leq \sqrt{n} \|\boldsymbol{x}\|_{\infty}.$$

#### 1.4

The induced matrix norm on  $\mathbb{R}^{n\times n}$ :  $\|\cdot\|_2$  and  $\|\cdot\|_{\infty}$  are equivalent as well. Prove the inequality

$$||A||_{\infty} \le \sqrt{n} ||A||_2,$$

$$||A||_2 \le \sqrt{n} ||A||_{\infty}.$$

#### 1.5

(optional) Prove that the norms  $\|\cdot\|_1$  and  $\|\cdot\|_2$  are equivalent on  $\mathbb{R}^n$ .

## 2 matrix 2-Norms

We are going to show that, if U is a unitary matrix  $(U^TU = I)$ , then

$$||AU||_2 = ||A||_2, \quad ||UA||_2 = ||A||_2$$

# 2.1

Suppose U is a unitary matrix, show that  $||Ux||_2 = ||x||_2$  for any vector x.

#### 2.2

recall that the definition of matrix 2-norm is  $||A||_2 = \max_{x\neq 0} \frac{||Ax||_2}{||x||_2}$ , show that  $||UA||_2 = ||A||_2$  and  $||AU||_2 = ||A||_2$  using the definition and above result.

# 2.3 connection with SVD

For any matrix  $A \in \mathbb{R}^{m \times n}$ , we can express it by Singular Value Decomposition (SVD)

$$A = U\Sigma V^T$$

where  $U \in \mathbb{R}^{m \times m}$  and  $V \in \mathbb{R}^{n \times n}$  are orthogonal matrices, and  $\Sigma \in \mathbb{R}^{m \times n}$  is a diagonal matrix with non-negative entries  $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_r \geq 0$  on the diagonal. Here r = rank(A). Based the uni-Invariance of the frobenius norm and 2-norm, we have,

$$||A||_2 = ||\Sigma||_2$$

Show that  $\|\Sigma\|_2 = \sigma_1$ .

#### 2.4

show that  $||A^{-1}|| = 1/\sigma_r$ 

## 3 An ill-conditioned matrix

#### 3.1 review of condition number

$$\kappa(A) = ||A||_2 ||A^{-1}||_2$$

$$A = \begin{bmatrix} 1 & 0.9 \\ 0.9 & 1 \end{bmatrix}$$

This matrix is real and symmetric, so it can be diagonalized as

$$A = Q\Lambda Q^T$$

Solve the eigenvalues of A and using the eigenvalues to compute the condition number of A.

## 3.2

Now we considet solve the linear system Ax = b with  $b = [1, 1]^T$  and  $Ax = \tilde{b}$  with  $\tilde{b} = [1.1, 0.9]^T$ . You can solve the linear system by LU decomposition or directly compute the inverse of A.

#### 3.3

Compute the relative error of the solution x and input b. What is the theoretical upper bound of ratio of them? Compare the actual error with the theoretical bound.

#### 3.4

Try  $b = [1, -1]^T$  and  $\tilde{b} = [1.1, -0.9]^T$ . Also compute the ratio of errors.

## 4 Conditional number for the Hilbert matrix

The Hilbert matrix  $H \in \mathbb{R}^{n \times n}$  is a matrix with entries

$$h_{ij} = \frac{1}{i+j-1}.$$

### 4.1

Using MATLAB, compute the 2-norm-based condition numbers for n = 3, 5, 10, 20, 25. You can use MATLAB's built-in function cond() to compute the condition number of a matrix.

#### 4.2

Let's consider a relative right hand side perturbation  $\delta \boldsymbol{b}$  of a linear system with  $\|\delta \boldsymbol{b}\|_2 / \|\boldsymbol{b}\|_2 \approx 10^{-15}$ . Write down the corresponding bounds  $\|\delta \boldsymbol{x}\|_2 / \|\boldsymbol{x}\|_2$  from the theory we discussed in class.

#### 4.3

Now, let's compute the actual error. Use the right-hand side vector with entries  $b_i = \sum_{j=1}^n (j/(i+j-1))$  chosen such that the solution vector has entries  $x_i = i$ . Now, Compute the numerical solutions  $x_i = i$  then re-compute  $x_i = i$  and compare the relative right-hand side error and the relative error in the solutions. How much are these better than the estimates you got from the condition number?

<sup>&</sup>lt;sup>1</sup>Note that all these computations contain tiny errors due to the final precision of computer computations.