

# MS-E1651 - Numerical Matrix Computations

## Exercise 3

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1. (P40)

**Solution:**

(a) Given:  $(A + uv^T)\hat{x} = b$  (1)  $Ax = b$  (2)

We want to show that:

$(A + uv^T)w = uv^T x$  (3)

Now, let's evaluate  $(A + uv^T)w$  on the left side of equation (3) using equation (4):

$$\begin{aligned}(A + uv^T)w &= (A + uv^T)x - (A + uv^T)\hat{x} \\ &= Ax + uv^T x - (A + uv^T)\hat{x} \\ &= b + uv^T x - b \\ &= uv^T x\end{aligned}$$

(b) We know that:

$$\begin{aligned}(A + uv^T)w &= uv^T x \\ Aw + uv^T w &= uv^T x \\ A^{-1}Aw + A^{-1}uv^T w &= A^{-1}uv^T x \\ \rightarrow w &= A^{-1}uv^T x - A^{-1}uv^T w \\ w &= A^{-1}u(v^T x) - A^{-1}u(v^T w)\end{aligned}$$

Both  $(v^T w)$  and  $(v^T x)$  are in  $\mathbb{R}$  and they are scalar. Thus we can have above equation as follows:

$$w = (v^T w - v^T x)A^{-1}u$$

and if we put  $\alpha = (v^T w - v^T x)$  we would have the result that  $w = \alpha A^{-1}u$  for some  $\alpha \in \mathbb{R}$ .

(c) Now, let's find the value of  $\alpha$  using these results.

Starting from the equation in (a):

$$(A + uv^T)w = uv^T x$$

.

Substituting  $w = \alpha A^{-1}u$  from (b):

$$(A + uv^T)(\alpha A^{-1}u) = uv^T x$$

Now, let's simplify this equation:

$$\alpha(A + uv^T)A^{-1}u = uv^T A^{-1}b$$

Both terms  $\alpha(A + uv^T)A^{-1}$  and  $v^T A^{-1}b$  are scalar and both are coefficient of  $u$  in both side of equation, so they must be equal. Thus we have

$$\alpha(A + uv^T)A^{-1} = v^T A^{-1}b$$

$$\alpha = \frac{v^T A^{-1}b}{(A + uv^T)A^{-1}}$$

$$\alpha = \frac{v^T A^{-1}b}{(1 + uv^T A^{-1})}$$

Now, let's find  $\hat{x}$  using the result from **(b)**:

$$w = \alpha A^{-1}u$$

$$w = \frac{v^T A^{-1}b}{1 + v^T A^{-1}u} A^{-1}u$$

Now, let's express  $\hat{x}$  in terms of  $w$ :

$$\hat{x} = A^{-1}b - w$$

$$\hat{x} = A^{-1}b - \frac{v^T A^{-1}b}{1 + v^T A^{-1}u} A^{-1}u$$

and since  $v^T A^{-1}b$  is a scalar we can write as follows:

$$\hat{x} = A^{-1}b - \frac{A^{-1}uv^T A^{-1}b}{1 + v^T A^{-1}u}$$

## 2. (P41) **Solution:**

(a) Based on the definition we have

$$(U\Sigma V^T - U\delta\Sigma V^T)e = b$$

Multiplying both side of equation by  $U^T$  from gives us

$$(\Sigma V^T - \delta\Sigma V^T)e = U^T b$$

$$(\Sigma - \delta\Sigma)V^T e = U^T b$$

Then we would have  $(\Sigma - \delta\Sigma) = \text{diag}(\sigma_1 - \delta\sigma_1, \dots, \sigma_n - \delta\sigma_n)$  and if we define matrix  $\hat{\Sigma} = \text{diag}(\frac{\sigma_1^{-1}}{1-\sigma_1^{-1}\delta\sigma_1}, \dots, \frac{\sigma_n^{-1}}{1-\sigma_n^{-1}\delta\sigma_n})$ , we will have

$$\hat{\Sigma}(\Sigma - \delta\Sigma) = (\Sigma - \delta\Sigma)\hat{\Sigma} = I$$

So if multiply both side of equation by  $\hat{\Sigma}$ , we will have

$$V^T e = \hat{\Sigma} U^T b$$

Moreover it is enough to multiply by  $V$  from right in both side of equation, so we will have

$$e = V \hat{\Sigma} U^T b$$

(b) First of all, we know that inverse of  $A$  is

$$A^{-1} = V \Sigma^{-1} U^T$$

In this problem we are going to use unitarily invariant property, which is

$$\|UA\|_2 = \|A\|_2$$

if  $U$  is unitary matrix. Since norm-2 of  $b$  is 1, we would have

$$e \leq \frac{\|A^{-1}\|_2}{1 - \|A^{-1}\|_2 \|\delta A\|_2} \quad (1)$$

$$e \leq \frac{\|V \Sigma^{-1} U^T\|_2}{1 - \|V \Sigma^{-1} U^T\|_2 \|U \delta \hat{\Sigma} V^T\|_2} \quad (2)$$

$$e \leq \frac{\|\Sigma^{-1} U^T\|_2}{1 - \|\Sigma^{-1} U^T\|_2 \|\delta \hat{\Sigma} V^T\|_2} \quad (3)$$

$$e \leq \frac{\|\Sigma^{-1}\|_2}{1 - \|\Sigma^{-1}\|_2 \|\delta \hat{\Sigma}\|_2} \quad (4)$$

So we would have

$$e \leq \frac{\max_i |\sigma_i|}{1 - (\max_i |\frac{1}{\sigma_i}|)(\max_i |\frac{\sigma_i^{-1}}{1-\sigma_i^{-1}\delta\sigma_i}|)}$$

### 3. (P43) **Solution:**

(a)

$$(345)_{10} = (101011001)_2 \approx (1.010)_2 \times 2^8$$

$$(\frac{1}{3})_{10} = (0.0101010101\dots)_2 \approx (1.011)_2 \times 2^{-2}$$

(b)

$$(1.010)_2 \times 2^8 = 320 \rightarrow |\hat{x}_1 - x_1| = 25$$

$$(1.011)_2 \times 2^{-2} = \frac{11}{32} \rightarrow |\hat{x}_2 - x_2| = \frac{1}{96}$$

(c)

$$u = \frac{1}{2^3} = \frac{1}{8}$$

(d)

$$\hat{x}_1 + \hat{x}_2 = (10100000000.00)_2 \times 2^{-2} + (1.011)_2 \times 2^{-2} = (10100000001.011)_2 \times 2^{-2} \quad (5)$$

$$= (1.010)_2 \times 2^8 \quad (6)$$

#### 4. (P47) Solution:

(a) In this part, we use the previous problem P46. Then, for all  $i = 2, \dots, n$ , we write  $1 + u$  instead of  $1 + \delta_i$ , and for all  $k = 1, \dots, n$ , we write  $1 + u$  instead of  $1 + \hat{\delta}$ . In the other word, we are trying to make  $|fl(\sum x_i y_i) - \sum x_i y_i|$  greater, and also for each  $i$ , we consider the inner multiplication like  $\sum_{k=1}^n (1 + \delta_i)$ .

$$|fl(\sum x_i y_i) - \sum x_i y_i| \leq |\sum x_i y_i (1 + u)^n - x_i y_i| \quad (7)$$

Based on Lemma (1.5), we have:

$$|fl(\sum x_i y_i) - \sum x_i y_i| \leq |\sum x_i y_i (1 + u)^n - x_i y_i| \leq |\sum (1 + \frac{nu}{1 - nu}) x_i y_i - \sum x_i y_i| = \frac{nu}{1 - nu} |x^T y| \quad (8)$$

$$\rightarrow \frac{nu}{1 - nu} |x^T y| \leq \frac{nu}{1 - nu} \|x\| \|y\| \leq \frac{nu}{1 - nu} \quad (9)$$

In the last part, we used Cauchy–Schwarz inequality to conclude the inequality.

(b) Note that since  $U$  and  $V$  are unitary matrix, each row and column of them has length exactly 1. Thus, if we name row of  $U$ ,  $R_i^T$  and column of  $V$ ,  $C_j$ , for all  $i$  and  $j$ ,

$$\|R_i\|_2 = \|C_j\|_2 = 1$$

So conditions of part (a) hold here, and we can use part (a). Also note that  $(UV)_{ij} = R_i^T C_j$ . So  $|E_{ij}| = |fl(R_i^T C_j) - (UV)_{ij}| = |fl(R_i^T C_j) - R_i^T C_j|$ . Moreover, based on part (a), we know that  $|fl(R_i^T C_j) - R_i^T C_j| \leq \frac{nu}{1 - nu}$ . So we can write

$$fl(UV) = UV - E$$