

# Solution Keys of MATH 270A

Question 1:

2.3 (a) Since  $A = A^*$   $\Rightarrow x^*Ax$  and  $x^*x$  are hermitian, where  $x$  is eigenvector of eigenvalue  $\lambda$ .

Consider  $x^*Ax = (x^*Ax)^* = \lambda x^*x$ , both sides are real.

(b) Use the fact from (a)

Consider  $y^*Ax = \lambda_1 y^*x$   
 $x^*Ay = \lambda_2 x^*y$ , actually they are equal.

2.5 (a) Similarly as 2.3 (a).

(b) If  $\lambda=0$  is an eigenvalue of  $I-S$

$$\Rightarrow (I-S)x = \lambda x \Rightarrow S^*x = (1-\lambda)x$$

$\Rightarrow 1-\lambda$  i.e 1 is an eigenvalue of  $S$ .

Contradiction.

(c) Try to show  $[(I-S)^T(I+S)]^{-1} = [(I-S)^T(I+S)]^*$   
 by using  $S^* = -S$ .

2.6 (a) Let  $AA^{-1} = I \Rightarrow I + (1+\alpha + \alpha^* u)uv^*$   
 $\alpha$  must be zero.

(b) Consider  $\text{Null}(A)$  that contains non-zero element.

[3.2] Use the definition of Matrix induced norm

[3.3] (c) and (d) can be derived by applying (a) and (b).

[3.4] (b) Hint: To show that the "deletion matrix" has norm less than "1".

[4.4] Yes. Let  $A = U_1 \Sigma V_1^*$ , notice that for two unitary matrices ~~and~~  $U_1$  and  $U_2$ ,  $U_1 \cdot U_2$  is still unitary.

[4.5] Hints: Since  $A^* = A^T$ , notice that  $A^* \cdot A$  is a square matrix. and  $A^* \cdot A = V \Sigma^* \Sigma V^*$ , then apply ~~the~~ the facts

$$= V \Sigma^* \Sigma V^T$$

that real square matrix has real singular values, and looks like the product column by column.

Question 2.

[6.1] Use The 6.1.

[6.3] Use the fact that  $\text{Rank}(A) = \text{Rank}(A^* \cdot A)$ .

[6.5] Pick  $V \in \text{Range}(P)$ , then  $Pv = V \Rightarrow \|P\|_2 \geq 1$ .  
For equality, use facts  $P^* = P$  and definition of  $\| \cdot \|_2$ .

7.5

Use facts:

i)  $\text{Rank}(AB) \leq \min(\text{Rank } A, \text{Rank } B)$

ii) if  $B \in \mathbb{R}^{n \times k}$  has full rank

$$\text{Rank}(AB) = \text{rank}(A).$$

11.1

Let  $A = U\Sigma V^*$ , try to simplify  $A^+$ 's SVD

s.t we can see  $\|A^+\|_2 = \frac{1}{\sigma_n}$ , where  $\sigma_n$  is the smallest singular value of  $A$ .

Question 3.

20.1

To show the uniqueness, you can look at the eigenvalues of  $A$  by ~~assuming~~ assuming there  $\exists$  another pair of  $\bar{U}$  and  $\bar{V}$ .

20.2

Can be shown by ~~observing~~ observing the way we construct  $L$ .

20.3

cb7. Same as 20.2, try to show it by construct the matrix  $L$ .

21.2 Same as 20.2, try to construct P and L.

21.6 Show that the resulting matrix after each step of Gaussian elimination is still diagonally dominant.

22.1 Since  $V = L_1 \cdots L_{n-1} \cdot PA$ , where  $\|P\|_\infty = 1$   
Consider  $U_i = a_i - \sum_{j=1}^{i-1} l_{ij} u_j$ , where  $|l_{ij}| \leq 1$ .

To show the result by induction.

23.1 ~~No~~ Yes, because  $A^T A$  is a Hermitian with positive eigenvalues.

Question 4.

12.1 Notice that  $\|\cdot\|_F = \sqrt{\sum \delta_i^2}$ , where  $\delta_i$  are singular values.  
and  $\|\cdot\|_2 = \delta_{\max}$ .

14.1 For example:

(a)  $\sin x = O(1) \Leftrightarrow \exists$  a constant  $C$   
s.t.  $|\sin x| \leq C$  with sufficient  
large  $x$ .

which is ~~not~~ True.

14.2 (a)  $(1 + O(\varepsilon))(1 + O(\varepsilon))$

$$= 1 + 2O(\varepsilon) + O(\varepsilon)^2 = 1 + O(\varepsilon).$$

16.1 To show it is backward stable, that is to show  
for a  $\tilde{A}$  with  $\frac{\|A - \tilde{A}\|}{\|A\|} = O(\varepsilon)$ , we have.

$$\tilde{B} = Q_K \cdots Q_1 \tilde{A}, \text{ i.e. } \cancel{Q_1 \cdots Q_K (A + \Delta A)}$$

from right to left

i.e. for  $\tilde{A} = A + \Delta A$

$$Q_1(A + \Delta A) = Q_1 A + \cancel{Q_1 Q_2 A}$$

$$Q_2 \cdot Q_1(A + \Delta A) = Q_2 Q_1 A + Q_2 Q_1 \Delta A$$

;

$$\tilde{B} = Q_K \cdots Q_1 A + Q_K \cdots Q_1 \Delta A = Q_K \cdots Q_1 (A + \Delta A)$$

17.1 (a)  $\| \Delta R \|_1 \leq \max_i \sum_j | \Delta r_{ij} | \leq m \max_i \sum_j | r_{ij} | \cancel{\leq} \epsilon + O(\epsilon^2)$

17.2  $\| \tilde{x} - x \|_1$ , where  $\begin{cases} (R + \Delta R) \tilde{x} = b \\ Rx = b \end{cases} \Rightarrow R\tilde{x} - Rx = -\Delta R\tilde{x}$   
 $\Rightarrow R(x - \tilde{x}) = \Delta R\tilde{x} = \Delta R(R + \Delta R)^{-1}b.$

18.4 For square case,  $y$  can be obtained exactly independent of  $A$ .

19.1 (19.4) is  $\begin{cases} R^T + Ax = b \\ A^T R = 0 \end{cases}$ , since least square solution solves  $A^T A x = A^T b$ .

One can show  $x$  from the system is the same in  $A^T A x = A^T b$ .

Question 5.

24.4 (a)  $A^n = U[\lambda^n] U^*$

25.1 (a) By hint, ~~then~~ then look at the dimension of eigenspace.

25.2 (a) To reach accuracy  $O(\epsilon)$   
 we need iterate  $m$  times where  $C^m = O(\epsilon)$   
 $\Rightarrow m = \frac{\log \epsilon}{\log C}$ , totally  $m \cdot O(1) = O(\log \epsilon)$ .

26.3 (b) Use the fact  $A$  is normal  
 $\Leftrightarrow A$  is diagonalizable by unitary matrix.  
then apply Bauer-Fike result.

27.1 For a  $X$ , consider  $\frac{X}{\|X\|}$  be the  $i$ -th column of  $Q$ .