## MATH36001 Solutions: Theory of Eigensystems 2015

Note: the solution to Example 5 in the handout is added at the end of this document.

**1**. The characteristic polynomial of A is given by  $p(t) = \det(tI - A) = (t - \lambda_1)(t - \lambda_2) \cdots (t - \lambda_n)$ so that  $p(0) = (-1)^n \det(A) = (-1)^n \prod_{i=1}^n \lambda_i$ .

One can show by induction that the coefficient of  $t^{n-1}$  in  $p(t) = \det(tI - A)$  is  $-\operatorname{trace}(A)$ and that of  $p(t) = (t - \lambda_1)(t - \lambda_2) \cdots (t - \lambda_n)$  is  $-\sum_{i=1}^n \lambda_i$ .

(A direct proof is obtained by using the Schur form  $A = UTU^*$ , and noting that trace(A) =  $\operatorname{trace}((UT)U^*) = \operatorname{trace}(U^*UT) = \operatorname{trace}(T)$ , where we have used the relation  $\operatorname{trace}(AB) =$ trace(BA) proven in exercise 9 (ii) of the basics handout.)

Since the eigenvalues of  $A^k$  are  $\lambda_1^k, \lambda_2^k, \ldots, \lambda_n^k$ , it follows that  $\operatorname{trace}(A^k) = \sum_{i=1}^n \lambda_i^k$ .

**2**. (a) Let  $x = \sum_{i=1}^{k} \alpha_i x_i \in S$ . Then  $Ax = \sum_{i=1}^{k} \alpha_i \lambda_i x_i \in S$  so that S is an invariant subspace under A.

(b) We have  $Ae_1 = 2e_1 \in S$  and  $Ae_2 = e_1 + 2e_2 \in S$  so S is an invariant subspace under A. Note that  $e_2$  is not an eigenvector.

**3**. Let  $U_k = \begin{bmatrix} u_1 & \dots & u_k \end{bmatrix}$ ,  $e_i$  is the *i*th column of the identity matrix and  $T_k$  denotes the  $k \times k$ leading principal submatrix of T. Rewrite  $A = UTU^*$  as AU = UT so that

$$AU_{k} = AU \begin{bmatrix} e_{1} & \dots & e_{k} \end{bmatrix} = UT \begin{bmatrix} e_{1} & \dots & e_{k} \end{bmatrix} = U \begin{bmatrix} T_{k} \\ O \end{bmatrix} = U_{k}T_{k}$$

and by Theorem 1,  $U_k$  is invariant under A.

4. (a) If  $A^*A = AA^*$  then  $B^*B = U^*A^*UU^*AU = U^*A^*AU = U^*AA^*U = U^*AUU^*A^*U = U^*AUU^*AU = U^*A$ 

 $BB^* \text{ so } B \text{ is normal. If } A^* = A \text{ then } B^* = U^*A^*U = U^*AU = B \text{ so } B \text{ is Hermitian.}$ (b) Take  $S = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$  and  $A = \begin{bmatrix} 0 & 2 \\ 2 & 0 \end{bmatrix}$  which is Hermitian and therefore normal. Then  $S^{-1}AS = \begin{bmatrix} 2 & 0 \\ 2 & -2 \end{bmatrix}$  is neither Hermitian nor normal.

5. Hermitian, skew-Hermitian and unitary matrices are normal. If A is normal, there exist a unitary matrix U and a diagonal matrix  $\Lambda = \text{diag}(\lambda_1, \ldots, \lambda_n)$  such that  $A = U\Lambda U^*$ .

(a) A Hermitian  $\Rightarrow A = UAU^* = (UAU^*)^* = U\overline{A}U^* \Rightarrow A = \overline{A}$ , i.e.,  $A \in \mathbb{R}^{n \times n}$ . Conversely, A normal with real eigenvalues  $\Rightarrow A^* = (UAU^*)^* = UAU^* = A$  so A is Hermitian.

(b) Proof similar to (a). Here,  $\Lambda = -\overline{\Lambda}$ , i.e.,  $\Lambda \in i\mathbb{R}^{n \times n}$ .

(c) A unitary  $\Rightarrow I = A^*A = U\overline{\Lambda}U^*U\Lambda U^* = U\overline{\Lambda}\Lambda U^* \Rightarrow \overline{\Lambda}\Lambda = I$ , i.e.,  $|\lambda_i| = 1, i = 1, \dots, n$ . Conversely, A normal with eigenvalues on the unit circle  $\Rightarrow A^*A = U\overline{A}U^*UAU^* = I \Rightarrow A$  is unitary.

**6**. (a) If N is nilpotent then  $N^k = O$  for some k > 0. Let  $\lambda$  be an eigenvalue of N with corresponding eigenvector x. Then  $N^2x = \lambda^2 x$ ,  $N^3x = \lambda^3 x$ , ...,  $N^k x = \lambda^k x = 0$ . Hence  $\lambda = 0$  since  $x \neq 0$ .

(b) Consider the generic  $4 \times 4$  strictly upper triangular matrix  $T = \begin{bmatrix} 0 & \times & \times & \times \\ 0 & 0 & \times & \times \\ 0 & 0 & 0 & \times \\ 0 & 0 & 0 & 0 \end{bmatrix}$ , where

 $\times$  denotes any entry that is not necessarily zero. Then,

In general, powering T causes the nonzero superdiagonals to move a diagonal at a time towards the top right-hand corner until there is none left. If T is  $n \times n$  then necessarily  $T^n = O$ .

(c) Let  $N = UTU^*$  be the Schur decomposition of N. If all eigenvalues of N are zero then T is strictly upper triangular and therefore nilpotent. Since  $N^k = UT^kU^*$  it follows that if T is nilpotent so is N.

7. When A is normal the spectral theorem says that there exist a unitary matrix  $U = [u_1 \ldots u_n]$  and a diagonal matrix  $\Lambda = \text{diag}(\lambda_1, \ldots, \lambda_n)$  such that

$$A = U\Lambda U^* = \begin{bmatrix} \lambda_1 u_1 & \dots & \lambda_n u_n \end{bmatrix} \begin{bmatrix} u_1^* \\ \vdots \\ u_n^* \end{bmatrix} = \sum_{i=1}^n \lambda_i u_i u_i^*.$$

A has two eigenvalues  $\lambda_1 = 1$  and  $\lambda_2 = 3$  with corresponding eigenvectors  $x_1 = (1/\sqrt{2}) \begin{bmatrix} -1 & 1 \end{bmatrix}^T$ and  $x_2 = (1/\sqrt{2}) \begin{bmatrix} 1 & 1 \end{bmatrix}^T$ . Therefore

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \begin{bmatrix} -1 & 1 \end{bmatrix} + \frac{3}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix}.$$

8. Let T be a normal upper triangular matrix. That T is diagonal is seen by equating the entries of  $T^*T$  and  $TT^*$  as follows.

 $e_1^T(T^*T)e_1 = e_1^TTT^*e_1 \Leftrightarrow t_{11}\bar{t}_{11} = t_{11}\bar{t}_{11} + \sum_{j=2}^n t_{1j}\bar{t}_{1j}$  which implies  $0 = \sum_{j=2}^n |t_{1j}|^2$ , i.e.,  $t_{1j} = 0$  for j = 2, ..., n. In a similar way we show that  $e_2^T(T^*T)e_2 = e_2^TTT^*e_2$  implies that  $t_{2j} = 0$  for j = 3, ..., n. Arguing for each diagonal entry in turn, we conclude that  $t_{ij} = 0$ , j > i, i = 1, ..., n. Since T is upper triangular we have shown that T is diagonal.

**9**. (a) 2 Jordan blocks so 2 linearly independent eigenvectors. (b) 3 linearly independent eigenvectors. (c) 4 linearly independent eigenvectors.

**10**.

(a) Symmetric matrix so diagonalizable.

(b)  $2 \times 2$  Jordan block so not diagonalizable.

(c) 2 distinct eigenvalues so diagonalizable.

(d)  $\lambda = 1$  with multiplicity 2 and only one eigenvector so not diagonalizable.

**11**. (a)

$$(i)\begin{bmatrix}2&&&\\&2&&\\&&2&\\&&&2\end{bmatrix}, (ii)\begin{bmatrix}2&1&&\\&2&&\\&&2&\\&&&2\end{bmatrix}, (iii)\begin{bmatrix}2&1&&\\&2&&\\&&2&1\\&&&&2\end{bmatrix}, (iv)\begin{bmatrix}2&1&&\\&2&1&\\&&2&\\&&&2\end{bmatrix}, (v)\begin{bmatrix}2&1&&\\&2&1&\\&&2&1\\&&&&2\end{bmatrix}$$

(b) (i): geometric multiplicity 4, (ii): geometric multiplicity 3, (iii): geometric multiplicity 2, (iv): geometric multiplicity 2, (v): geometric multiplicity 1.
(c)

(d) The minimal polynomials are (i)  $q(\lambda) = \lambda - 2$ , (ii)  $q(\lambda) = (\lambda - 2)^2$ , (iii)  $q(\lambda) = (\lambda - 2)^2$ , (iv)  $q(\lambda) = (\lambda - 2)^3$ , (v)  $q(\lambda) = (\lambda - 2)^4$ . Note that the Jordan forms in (ii) and (iii) have the same minimal polynomial.

12.  $\lambda = 3$  is an eigenvalue of algebraic multiplicity 3 and  $\operatorname{rank}(A - 3I) = 1$  so the eigenvalue 3 has two eigenvectors associated with it: these are solution of (A - 3I)x = 0. We find that  $x_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$  and  $x_2 = \begin{bmatrix} 0 \\ 1 \\ -2 \end{bmatrix}$  are eigenvectors (and are linearly independent). To find the generalized eigenvector associated with  $\lambda = 3$  solve  $(A - 3I)v = x_1$  to get  $v = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ . Now let  $X = \begin{bmatrix} x_1 & v & x_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & -2 \end{bmatrix}$ . Check that  $X^{-1}AX = \begin{bmatrix} 3 & 1 \\ -3 \\ -3 \end{bmatrix}$ .

**13**. The characteristic polynomial of A is  $p(\lambda) = \lambda^2 - 3\lambda + 2$  and  $p(A) = A^2 - 3A + 2I = O$ . Thus  $A^2 = 3A - 2I$ ,  $A^3 = A(A^2) = 3A^2 - 2A = 3(3A - 2I) - 2A = 7A - 6I$ ,  $A^4 = 7A^2 - 6A = 15A - 14I$ .

14. Let  $p(\lambda) = a_0 + a_1 \lambda + \dots + a_{n-1} \lambda^{n-1} + \lambda^n$  be the characteristic polynomial of A with  $a_0 \neq 0$  since A is nonsingular. From p(A) = O we obtain  $A^{-1}p(A) = a_0A^{-1} + a_1I + \dots + a_{n-1}A^{n-2} + A^{n-1} = O$ . Hence,  $A^{-1} = (-1/a_0)(a_1I + \dots + a_{n-1}A^{n-2} + A^{n-1})$ .

15. By polynomial long division any such p can be written  $p = q \psi + r$ , where the degree of the remainder r is less than that of q. But  $O = p(A) = q(A)\psi(A) + r(A) = r(A)$ , and this contradicts the minimality of the degree of q unless r = 0. Hence r = 0 and q divides p.

16. Let  $B = I - \frac{1}{3}A$ . From q(A) = O we obtain  $5I + 9B^2 - 15B = O$ . Premultiplying by  $B^{-1}$  and rearranging gives  $B^{-1} = (3A + 6I)/5$ , as required.

17. Note that  $(uv^*)u = (v^*u)u$  so  $\lambda = v^*u$  is an eigenvalue of  $uv^*$ . Let  $v_2, \ldots, v_n$  be n-1 linearly independent vectors such that  $v^*v_i = 0$ ,  $i = 2, \ldots, n$ . Then  $(uv^*)v_i = 0$  so  $\lambda = 0$  is an eigenvalue of multiplicity n-1. We deduce that the characteristic polynomial is  $p(\lambda) = \lambda^{n-1}(\lambda - v^*u)$  and the minimal polynomial is  $q(\lambda) = \lambda(\lambda - v^*u)$ . As a check, we have  $q(uv^*) = uv^*(uv^* - (v^*u)I) = v^*u(uv^* - uv^*) = O$ .

18. It is easy to see that for scalars x and y, p(x) - p(y) = q(x, y)(x - y) for some polynomial q of two variables. We can substitute tI for x and A for y to obtain

$$p(t)I - p(A) = q(tI, A)(tI - A).$$
 (1)

If p(A) = O then we have  $p(t)(tI - A)^{-1} = q(tI, A)$ , so that  $p(t)(tI - A)^{-1}$  is a polynomial in t. Conversely, if  $p(t)(tI - A)^{-1}$  is a polynomial in t then from (1) it follows that  $p(A)(tI - A)^{-1} = p(t)(tI - A)^{-1} - q(tI, A)$  is a polynomial. Since p(A) is a constant this implies that p(A) = O.

To obtain the Cayley–Hamilton theorem set  $p(t) = \det(tI - A)$ . From the formula  $B^{-1} = \operatorname{adj}(B)/\det(B)$ , where the adjugate adj is the transpose of the matrix of cofactors, we have  $p(t)(tI - A)^{-1} = \operatorname{adj}(tI - A)$  is a polynomial in t, so p(A) = O by the first part.

**19**.

(a)  $A^k X = XB^k$  clearly holds for k = 1. Assume that  $A^k X = XB^k$ . Then  $A^{k+1}X = A(A^kX) = AXB^k = XBB^k = XB^{k+1}$ .

(b) From (a) it follows that p(A)X = Xp(B) for any polynomial p(t). Choose p(t) to be the characteristic polynomial of A. By the Cayley–Hamilton theorem, p(A) = O and therefore p(A)X = O = Xp(B). Since  $p(B) = (B - \lambda_1 I) \cdots (B - \lambda_n I)$ , where  $\lambda_1, \ldots, \lambda_n$  are the eigenvalues of A, the matrix p(B) is nonsingular and p(B)X = O has only the solution X = O.

## Solution to **Example 5** from the handout:

 $\underline{\lambda = 1}$ : dim(null(A - I)) =  $n - \operatorname{rank}(A - I) = 14 - 11 = 3$  so there are 3 Jordan blocks with eigenvalue  $\lambda = 1$ . Also since  $\operatorname{rank}(A - I)^3 = \operatorname{rank}(A - I)^4$ , the index of  $\lambda = 1$  is 3 and therefore the size of the largest Jordan block with eigenvalue  $\lambda = 1$  is 3. The formula in (4) (see handout) gives

number of blocks of size 1:  $14 + 10 - 2 \times 11 = 2$ .

number of blocks of size 2:  $11 + 9 - 2 \times 10 = 0$ .

number of blocks of size 3:  $10 + 9 - 2 \times 9 = 1$ .

Hence  $\lambda = 1$  has algebraic multiplicity  $2 \times 1 + 1 \times 3 = 5$ .

 $\underline{\lambda = 2}$ : dim(null(A - 2I)) = n - rank(A - 2I) = 14 - 12 = 2 so there are 2 Jordan blocks with eigenvalue  $\lambda = 2$ . Also since rank(A - I)<sup>2</sup> = rank(A - I)<sup>3</sup>, the index of  $\lambda = 2$  is 2 and therefore the size of the largest Jordan block with eigenvalue  $\lambda = 2$  is 2. The formula in (4) gives

number of blocks of size 1:  $14 + 10 - 2 \times 12 = 0$ .

number of blocks of size 2:  $12 + 10 - 2 \times 10 = 2$ .

Hence  $\lambda = 2$  has algebraic multiplicity  $2 \times 2 = 4$ .

 $\underline{\lambda = 3}$ : dim(null(A - I)) =  $n - \operatorname{rank}(A - 3I) = 14 - 12 = 2$  so there are 3 Jordan blocks with eigenvalue  $\lambda = 3$ . Also since  $\operatorname{rank}(A - I)^4 = \operatorname{rank}(A - I)^5$ , the size of the largest Jordan block with eigenvalue  $\lambda = 3$  is 4. The formula in (4) gives

number of blocks of size 1:  $14 + 11 - 2 \times 12 = 1$ .

number of blocks of size 2:  $12 + 10 - 2 \times 11 = 0$ .

number of blocks of size 3:  $11 + 9 - 2 \times 10 = 0$ .

number of blocks of size 4:  $10 + 9 - 2 \times 9 = 1$ .

Hence  $\lambda = 1$  has algebraic multiplicity  $1 \times 1 + 1 \times 4 = 5$ . From this information we obtain

$$J = \operatorname{diag}\left( \begin{bmatrix} 1 & 1 & \\ & 1 & 1 \\ & & 1 \end{bmatrix}, \begin{bmatrix} 1 \end{bmatrix}, \begin{bmatrix} 1 & \\ & 2 \end{bmatrix}, \begin{bmatrix} 2 & 1 \\ & 2 \end{bmatrix}, \begin{bmatrix} 2 & 1 \\ & 2 \end{bmatrix}, \begin{bmatrix} 3 & 1 & \\ & 3 & 1 \\ & & 3 & 1 \\ & & & 3 \end{bmatrix}, \begin{bmatrix} 3 \end{bmatrix} \right).$$

What is the minimal polynomial of this matrix?