

UCDHSC — Dept. of Mathematical Sciences
Applied Linear Algebra Preliminary Exam
(Solutions) May 25, 2007

1. For each of the following statements, prove that it is true or give an example to show that it is false. Throughout, $A \in \mathcal{C}^{m \times m}$ unless otherwise indicated.
- (a) If λ is an eigenvalue of A and $\mu \in \mathcal{C}$, then $\lambda - \mu$ is an eigenvalue of $A - \mu I$.
 - (b) If A is real and λ is an eigenvalue of A , then so is $-\lambda$.
 - (c) If A is real and λ is an eigenvalue of A , then so is $\bar{\lambda}$.
 - (d) If λ is an eigenvalue of A and A is nonsingular, then λ^{-1} is an eigenvalue of A^{-1} .
 - (e) If all the eigenvalues of A are zero, then $A = 0$.
 - (f) If A is Hermitian and λ is an eigenvalue of A , then $|\lambda|$ is a singular value of A .
 - (g) If A is diagonalizable and its eigenvalues are equal, then A is diagonal.

Solution.

- (a) True. If λ is an eigenvalue of A , then there exists x different from 0 such that $Ax = x\lambda$ which implies that $(A - \mu I)x = x(\lambda - \mu)$ and so $\lambda - \mu$ is an eigenvalue of $A - \mu I$.
- (b) False. Take the identity matrix as a counter-example. All the eigenvalues are 1s.
- (c) True. If A is real then the characteristic polynomial of A ($p_A(x) = \det(xI - A)$) has real coefficients. If λ is an eigenvalue of A then it is a root of the characteristic polynomial. Since the coefficients of $p_A(x)$ are all real, its roots are complex conjugates, that is to say if λ is a root of $p_A(x)$, then so is $\bar{\lambda}$ (possibly the same root if λ is real). Since the roots of $p_A(x)$ are eigenvalues of A , we conclude that $\bar{\lambda}$ is an eigenvalue of A .
- (d) True. First note that if A is nonsingular then A^{-1} exists and we also have that all the eigenvalues of A are nonzero so that λ^{-1} is well defined. If λ is an eigenvalue of A , then there exists x different from 0 such that $Ax = x\lambda$ which implies that $A^{-1}x = x\lambda^{-1}$ so λ^{-1} is an eigenvalue of A^{-1} .
- (e) False. Take

$$A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix},$$

the only eigenvalue of A is 0 but A is not 0.

- (f) True. There is two ways to go.
 - i. If λ is an eigenvalue of A , then, first, note that λ is real (since A is Hermitian), second, there exists x of norm 1 such that $Ax = x\lambda$ which implies that

$$Ax = (x \operatorname{sign}(\lambda)) |\lambda|,$$

where $\operatorname{sign}(\lambda) = 1$ if $\lambda \geq 0$ and $\operatorname{sign}(\lambda) = -1$ otherwise.

Let us call $v = x$, $\sigma = |\lambda|$, and $u = (x \operatorname{sign}(\lambda))$, then this rewrites

$$Av = u\sigma. \quad (*)$$

Since A is Hermitian, $A^*x = x\lambda$ which implies that

$$A^*(x \operatorname{sign}(\lambda)) = x|\lambda|.$$

So,

$$A^*u = v\sigma. \quad (**)$$

Equalities (*) and (**) imply that (u, σ, v) is a singular triplet for A . We conclude that, if A is Hermitian and λ is an eigenvalue of A , then $|\lambda|$ is a singular value of A .

- ii. If A is Hermitian then A is unitarily diagonalizable; that is to say: $A = V\Lambda V^*$, where Λ is a diagonal real matrix with the eigenvalues of A on the diagonal and V is unitary ($VV^* = I$). Consider the diagonal matrix, D , made of $+1$ and -1 on the diagonal defined as $d_{ii} = \operatorname{sign}(\lambda_i)$, then note that:

$$D\Lambda = |D| \quad \text{and} \quad D^2 = I$$

Defining $\Sigma = |D| = D\Lambda$ and $U = VD$, we can write

$$A = V\Lambda V^* = (VD)(D\Lambda)V^* = U|\Lambda|V^* = U\Sigma V^*.$$

Note that U and V are unitary, Σ is diagonal with positive real elements on the diagonal so that $A = U\Sigma V^*$ is a singular value decomposition of A . Since we have defined $\Sigma = |\Lambda|$, we conclude that, if A is Hermitian and λ is an eigenvalue of A , then $|\lambda|$ is a singular value of A .

- (g) True. If A is diagonalizable and its eigenvalues are equal, then there exists a nonsingular matrix V such that $A = VDV^{-1}$ with $D = \lambda I$, so $A = V(\lambda I)V^{-1} = \lambda I$, thus A is diagonal. (A is called a homothety.)

2. Suppose $A, B \in \mathcal{C}^{m \times m}$ such that $\operatorname{Spectrum}(A) \cap \operatorname{Spectrum}(B) = \emptyset$.

- (a) Let $p_A(x)$ denote the characteristic polynomial of A . Prove that $p_A(B)$ is nonsingular.
 (b) Consider $X \in \mathcal{C}^{m \times m}$, such that $AX = XB$. Prove that for any polynomial P , we have $P(A)X = XP(B)$.
 (c) Prove that

$$\forall X \in \mathcal{C}^{m \times m}, (AX = XB \implies X = 0).$$

- (d) Prove that $\forall C \in \mathcal{C}^{m \times m}$, there exists a unique $X \in \mathcal{C}^{m \times m}$, such that $AX - XB = C$.

Solution.

- (a) Using the Fundamental Theorem of Algebra, there exists $\lambda_1, \lambda_2, \dots, \lambda_m$ in \mathcal{C} such that

$$p_A(x) = \prod_{i=1}^m (\lambda_i - x),$$

therefore

$$p_A(B) = \prod_{i=1}^m (\lambda_i I - B).$$

$$\begin{aligned}
p_A(B) \text{ is nonsingular} &\iff (\forall i \in \{1, \dots, n\}, \lambda_i I - B \text{ is nonsingular}), \\
&\iff (\forall i \in \{1, \dots, n\}, \lambda_i \notin \text{Spectrum}(B)), \\
&\iff \text{Spectrum}(A) \cap \text{Spectrum}(B) = \emptyset.
\end{aligned}$$

(b) Let us consider $X \in \mathcal{C}^{m \times m}$, such that $AX = XB$, then

$$A^2X = A(AX) = A(XB) = (AX)B = (XB)B = XB^2.$$

We can show (by induction) that, for any positive integer k , we have

$$A^k X = XB^k.$$

Therefore, for any polynomial P , we have

$$P(A)X = XP(B).$$

(c) If we consider the previous equation with the characteristic polynomial of A , this gives us

$$p_A(A)X = Xp_A(B).$$

But $p_A(A) = 0$ (Cayley-Hamilton Theorem), and thus

$$Xp_A(B) = 0.$$

We have proved in (a) that $p_A(B)$ is nonsingular, therefore

$$X = 0.$$

(d) The mapping

$$\begin{aligned}
\phi: \mathcal{C}^{m \times m} &\longrightarrow \mathcal{C}^{m \times m} \\
X &\longmapsto AX - XB
\end{aligned}$$

is linear. It is injective (see (c.)). Since $\mathcal{C}^{m \times m}$ is of finite dimension, ϕ is bijective.

Remark: the equation $AX - XB = C$ is called the (continuous) Sylvester equation.

3. A matrix $A \in \mathcal{C}^{n \times n}$ is tripotent if $A^3 = A$. Show that every tripotent matrix is diagonalizable. (Hint: Observe that A is annihilated by the polynomial $p(t) = t^3 - t$.)

Solution: Observe that A is annihilated by the polynomial $p(t) = t^3 - t = t(t-1)(t+1)$. The minimal polynomial divides this polynomial, so is of the form $m(t) = t^{d_0}(t-1)^{d_1}(t+1)^{d_2}$, with $0 \leq d_i \leq 1$ for $i = 0, 1, 2$. Thus, each Jordan block in the Jordan canonical form has order 1, which means that the Jordan canonical form is diagonal, so A is diagonalizable.

4. Let V be a finite-dimensional vector space over the field \mathbb{F} . Suppose $T \in \mathcal{L}(V)$ is self-adjoint, $\lambda \in \mathbb{F}$, and $\epsilon > 0$. Suppose there exists $v \in V$ such that $\|v\| = 1$ and

$$\|Tv - \lambda v\| < \epsilon.$$

Prove that T has an eigenvalue λ' such that $|\lambda - \lambda'| < \epsilon$.

Solution: Since T is self-adjoint, it has (by the spectral theorem) a set of orthonormal eigenvectors (e_1, \dots, e_n) that form a basis for V . Let $\lambda_1, \dots, \lambda_n$ be the corresponding eigenvalues. Let v be as described above, and let (v_1, \dots, v_n) be its coordinates with respect to the basis. That is,

$$v = v_1 e_1 + \dots + v_n e_n.$$

Then $Tv = \lambda_1 v_1 e_1 + \dots + \lambda_n v_n e_n$. Since $\|Tv - \lambda v\| < \epsilon$, we have

$$\begin{aligned} \epsilon^2 &> \|Tv - \lambda v\|^2 \\ &= \|(\lambda_1 - \lambda)v_1 e_1 + \dots + (\lambda_n - \lambda)v_n e_n\|^2 \\ &= |\lambda_1 - \lambda|^2 |v_1|^2 + \dots + |\lambda_n - \lambda|^2 |v_n|^2 \\ &\geq (\min\{|\lambda_1 - \lambda|^2, \dots, |\lambda_n - \lambda|^2\}) (|v_1|^2 + \dots + |v_n|^2) \\ &= (\min\{|\lambda_1 - \lambda|^2, \dots, |\lambda_n - \lambda|^2\}) \|v\|^2 \\ &= (\min\{|\lambda_1 - \lambda|^2, \dots, |\lambda_n - \lambda|^2\}). \\ &= |\lambda_j - \lambda|^2 \text{ for some } j. \end{aligned}$$

Thus, $\epsilon > |\lambda_j - \lambda|$ for some eigenvalue λ_j .

5. Compute the singular value decomposition of $A = \begin{pmatrix} 1 & 0 & i \\ 0 & 1 & -i \end{pmatrix}$.

Solution: $AA^* = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}$. The eigenvalues of this matrix are $\lambda_1 = 3$ and $\lambda_2 = 1$. The corresponding unit eigenvectors are $u_1 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$, and $u_2 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$. The singular values of A are $s_1 = \sqrt{\lambda_1} = \sqrt{3}$ and $s_2 = \sqrt{\lambda_2} = 1$. Thus,

$$A = U\Sigma V^*,$$

where $U = (u_1 \ u_2)$, $\Sigma = \begin{pmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$, and V^* is a 3×3 unitary matrix, with columns v_1, v_2 and v_3 . Then

$$v_1 = \frac{1}{s_1} A^* u_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -i & i \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{6}} \\ -\frac{1}{\sqrt{6}} \\ -\frac{2i}{\sqrt{6}} \end{pmatrix}.$$

and

$$v_2 = \frac{1}{s_2} A^* u_2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -i & i \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}.$$

v_3 must be a unit vector orthogonal to both v_1 and v_2 . Thus, $\langle (1, -1, -2i)^T, v_3 \rangle = 0$, and $\langle (1, 1, 0)^T, v_3 \rangle = 0$. It follows that $v_3 = \left(\frac{1}{\sqrt{3}}, \frac{-1}{\sqrt{3}}, \frac{i}{\sqrt{3}} \right)^T$, so

$$V = \begin{pmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{3}} \\ -\frac{2i}{\sqrt{6}} & 0 & \frac{i}{\sqrt{3}} \end{pmatrix},$$

and the singular value decomposition of A is

$$A = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{6}} & \frac{2i}{\sqrt{6}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{3}} & -\frac{i}{\sqrt{3}} \end{pmatrix}.$$

6. Let V be a vector space over the field \mathbb{F} , and let T be a linear map from V to \mathbb{F} . Prove that if $u \in V$ is not in $\text{null } T$, then

$$V = \text{null } T \oplus \{au : a \in \mathbb{F}\}.$$

Solution: Suppose $u \in V$ is not in T . If $a \in \mathbb{F}$ and $au \in \text{null } T$, then $0 = T(au) = aTu$, which implies that $a = 0$ (because $Tu \neq 0$). Thus

$$\text{null } T \cap \{au : a \in \mathbb{F}\} = \{0\}.$$

If $v \in V$, then find a such that $v = (v - au) + au$ and $v - au \in \text{null } T$. To find such an a , note that $T(v - au) = 0 \Leftrightarrow a = \frac{Tv}{Tu}$. Thus,

$$v = \left(v - \frac{Tv}{Tu}u \right) + \frac{Tv}{Tu}u,$$

which expresses v as the sum of a vector in $\text{null } T$ and a scalar multiple of u . Hence $V = \text{null } T + \{au : a \in \mathbb{F}\}$. Since $\text{null } T \cap \{au : a \in \mathbb{F}\} = \{0\}$, we have $V = \text{null } T \oplus \{au : a \in \mathbb{F}\}$.

7. Prove that if there exists a linear map on V whose null space and range are both finite dimensional, then V is finite dimensional.

Solution: Suppose there is a linear map $T : V \rightarrow W$ such that $\text{null } T$ and $\text{range } T$ are both finite dimensional. Then there exist vectors $u_1, \dots, u_m \in V$ and $w_1, \dots, w_n \in \text{range } T$ such that (u_1, \dots, u_m) spans $\text{null } T$ and (w_1, \dots, w_n) spans $\text{range } T$. Because each $w_j \in \text{range } T$, there exists $v_j \in V$ such that $w_j = Tv_j$.

Suppose $v \in V$. Then $Tv \in \text{range } T$, so there exist $b_1, \dots, b_n \in \mathbb{F}$ such that

$$\begin{aligned} Tv &= b_1w_1 + \dots + b_nw_n \\ &= b_1Tv_1 + \dots + b_nTv_n \\ &= T(b_1v_1 + \dots + b_nv_n) \end{aligned}$$

It follows that $v - (b_1v_1 + \cdots + b_nv_n) \in \text{null } T$. Thus, there exist a_1, \dots, a_m such that

$$v - (b_1v_1 + \cdots + b_nv_n) = a_1u_1 + \cdots + a_mu_m,$$

or

$$v = b_1v_1 + \cdots + b_nv_n + a_1u_1 + \cdots + a_mu_m.$$

So, $(v_1, \dots, v_n, u_1, \dots, u_m)$ spans V , so V is finite-dimensional.

8. Let $\beta = \{v_1, \dots, v_5\}$ be an arbitrary ordered basis for \mathfrak{R}^5 . Let $A = \begin{pmatrix} 1 & 1 & 1 & 0 & 0 \\ 0 & 2 & 1 & 3 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$.

Define $T : \mathfrak{R}^5 \rightarrow \mathfrak{R}^5$ by $[T(v)]_\beta = A[v]_\beta$ for all $v \in \mathfrak{R}^5$. (Note: the notation $[w]_\beta$ denotes the coordinate representation of w relative to the basis β .)

- Compute the eigenvalues of T and both the minimal and characteristic polynomials of T .
- Find the Jordan form for T .

Solution:

- Since A is an upper triangular matrix, the eigenvalues of T lie on the diagonal of A . Thus, T has eigenvalues $\lambda = 1$ with algebraic multiplicity 3 and $\lambda = 2$ with algebraic multiplicity 2. The characteristic polynomial is

$$p(x) = (x - 1)^3(x - 2)^2.$$

The minimum polynomial $m(x)$ must divide this polynomial, so

$$m(x) = (x - 1)^{d_1}(x - 2)^{d_2},$$

where $1 \leq d_1 \leq 3$ and $1 \leq d_2 \leq 2$. Now,

$$A - I = \begin{pmatrix} 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 3 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix},$$

which has a null space of dimension 2. Thus, $\lambda = 1$ has two ordinary eigenvectors corresponding to $\lambda = 1$, which forces $d_1 = 2$. Similarly,

$$A - 2I = \begin{pmatrix} -1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & -1 \end{pmatrix},$$

which has a null space of dimension 1. Thus, $\lambda = 2$ has one ordinary eigenvalue, which forces $d_2 = 2$. Thus, the minimal polynomial is

$$m(x) = (x - 1)^2(x - 2)^2.$$

(b) Using the information from above, the Jordan form is

$$J = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 2 \end{pmatrix}.$$