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1 (33 pts.)

Suppose an $n \times n$ matrix A has n independent eigenvectors x_1, \ldots, x_n . Then you could write the solution to $\frac{du}{dt} = Au$ in three ways:

$$u(t) = e^{At}u(0), \quad \text{or}$$
$$u(t) = Se^{\Lambda t}S^{-1}u(0), \quad \text{or}$$
$$u(t) = c_1e^{\lambda_1 t}x_1 + \ldots + c_ne^{\lambda_n t}x_n.$$

Here, $S = [x_1 | x_2 | \dots | x_n].$

(a) From the definition of the exponential of a matrix, show why e^{At} is the same as $Se^{\Lambda t}S^{-1}$.

Solution. Recall that $A = S\Lambda S^{-1}$, and $A^k t^k = S\Lambda^k t^k S^{-1}$. Then, definition of the exponential:

$$\exp(At) = \sum_{k=0}^{\infty} \frac{A^k t^k}{k!} = S\left(\sum_{k=0}^{\infty} \frac{\Lambda^k t^k}{k!}\right) S^{-1} = Se^{\Lambda t} S^{-1}.$$

(b) How (do you	find	c_1,\ldots,c_n	from <i>u</i>	u(0)) and S ?
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Solution. Since $e^0 = 1$, we see that

$$u(0) = c_1 x_1 + \ldots + c_n x_n = S \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix},$$

where we used the definition of the matrix product. Thus the answer is:

$$\begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = S^{-1}u(0).$$

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(c) For this specific equation, write u(t) in any one of the (added: latter two of the) three forms, using *numbers* not symbols: You can choose which form.

$$\frac{du}{dt} = \begin{bmatrix} 1 & 2\\ -1 & 4 \end{bmatrix} u, \text{ starting from } u(0) = \begin{bmatrix} 4\\ 3 \end{bmatrix}.$$

Solution. We diagonalize A and get:

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix}$$

.

Thus $c = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, so for the second form

$a_i(t) =$	2	1	2	0	1	-1	4	_	2	1	e^{2t}	0	1	
a(t) =	1	1	0	3	1	2	3		1	1	0	e^{3t}	2	,

while in the third form:

$$u(t) = e^{2t} \begin{bmatrix} 2\\1 \end{bmatrix} + 2e^{3t} \begin{bmatrix} 1\\1 \end{bmatrix}.$$

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2 (30 pts.)

This question is about the real matrix

$$A = \begin{bmatrix} 1 & c \\ 1 & -1 \end{bmatrix}, \quad \text{for} \quad c \in \mathbb{R}.$$

(a) - Find the eigenvalues of A, depending on c.

- For which values of c does A have real eigenvalues?

Solution. Since $0 = trA = \lambda_1 + \lambda_2$, we see that $\lambda_2 = -\lambda_1$.

Also, $-1 - c = \det A = -\lambda_1^2$. Thus,

$$\lambda = \pm \sqrt{1+c}.$$

Therefore,

the eigenvalues are real precisely when $c \ge -1$.

(b) - For one particular value of c, convince me that A is similar to both the matrix

$$B = \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix},$$

and to the matrix

$$C = \begin{bmatrix} 2 & 2 \\ 0 & -2 \end{bmatrix}.$$

- Don't forget to say which value \boldsymbol{c} this happens for.

Solution. If two matrices are similar, then they do have the same eigenvalues (those are 2, -2 for both B and C). Here we must therefore have 0 = trA and $-1 - c = \det A = -4$. We see that this happens precisely when c = 3, where we check that indeed the eigenvalues are 2, -2. However, this does not guarantee that they are similar - and hence is not convincing.

Convincing: The eigenvalues 2, -2 are different, so both A, B and C are diagonalizable,

with the same diagonal matrix (for example to $\Lambda = B$!). Therefore A, B and C are all similar when c = 3.

- (c) For one particular value of c, convince me that A cannot be diagonalized. It is not similar to a diagonal matrix Λ , when c has that value.
 - Which value c?
 - Why not?

Solution. As we saw above, trA = 0, so regardless of c the eigenvalues come in pairs $\lambda_2 = -\lambda_1$. This means that whenever $\lambda_1 \neq 0$, we have two different eigenvalues, and hence A is diagonalizable (not what we're after).

Thus we need $\lambda_1 = \lambda_2 = 0$, a repeated eigenvalue, which happens when c = -1 (so det A = 0) as the only suspect – does it work?

Convincing: For
$$c = -1$$
, we have $N(A - 0 \cdot I) = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$

With only a 1-dimensional space of eigenvectors for the matrix, we are convinced that A is not diagonalizable for c = -1.

3 (37 pts.)

(a) Suppose A is an n × n symmetric matrix with eigenvalues λ₁ ≤ λ₂ ≤ ... ≤ λ_n.
- What is the largest number real number c that can be subtracted from the diagonal entries of A, so that A - cI is positive semidefinite?
- Why?

Solution. - We first realize that: If A is symmetric, then A - cI is also symmetric, since in general $(A + B)^T = A^T + B^T$ (simple, but very important to check!).

- Then we realize that the eigenvalues of A - cI are $\lambda_1 - c \leq \lambda_2 - c \leq \ldots \leq \lambda_n - c$. Therefore:

 $c = \lambda_1$ is the largest that can ensure positive semidefiniteness (and it does).

(b) Suppose B is a matrix with independent columns.

- What is the nullspace N(B)?

- Show that $A = B^T B$ is positive definite. Start by saying what that means about $x^T A x$.

Solution. - Then Bx = 0 only has the zero solution, so $N(B) = \{0\}$.

- Again, we start by observing that $A^T = A$ is symmetric. Then we recall what positive definite means (the "energy" test):

$$x^T A x > 0$$
 whenever $x \neq 0$.

Thus, we see here (by definition the inner product property of the transpose of a matrix):

$$x^{T}Ax = x^{T}(B^{T}(Bx)) = (Bx)^{T}(Bx) = ||Bx||^{2} \ge 0.$$

So $A = B^T B$ is positive semidefinite. But finally, the equality $||Bx||^2 = 0$, only happens when Bx = 0 which by $N(B) = \{0\}$ means x = 0. (c) This matrix A has rank r = 1:

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}.$$

- Find its largest singular value σ from $A^T A$.

- From its column space and row space, respectively, find unit vectors u and v so that

$$Av = \sigma u$$
, and $A = u\sigma v^T$.

- From the nullspaces of A and A^T put numbers into the full SVD (Singular Value Decomposition) of A:

$$A = \begin{bmatrix} | & | \\ u & \dots \\ | & | \end{bmatrix} \begin{bmatrix} \sigma & 0 \\ 0 & \dots \end{bmatrix} \begin{bmatrix} | & | \\ v & \dots \\ | & | \end{bmatrix}^T.$$

Solution. We compute:

$$A^T A = \begin{bmatrix} 5 & 5\\ 5 & 5 \end{bmatrix}$$

Thus the two eigenvalues are $\lambda_1 = 0$ and $\lambda_2 = 10$, and $\sigma = \sqrt{10}$. For v, we find a vector in $N(A^T A - 10I)$, and normalize to unit length:

$$v = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}.$$

Then we find u using

$$u = \frac{Av}{\sigma} = \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}$$

Since we have the orthogonal sums of subspaces $\mathbb{R}^2 = \mathbb{R}^m = c(A) \oplus N(A^T)$ and also $\mathbb{R}^2 = \mathbb{R}^n = c(A^T) \oplus N(A)$, we need to find one unit vector from each of N(A) and $N(A^T)$ and augment to v and u, respectively:

$$v_2 = \begin{bmatrix} 1\sqrt{2} \\ -1\sqrt{2} \end{bmatrix} \in N(A),$$
$$u_2 = \begin{bmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix} \in N(A^T),$$

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Thus, we finally see the full SVD:

$$A = U\Sigma V^{T} = \begin{bmatrix} 1/\sqrt{5} & -2\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix} \begin{bmatrix} \sqrt{10} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}^{T}.$$

We remember, as a final check, to verify that the square matrices U and V both contain orthonormal bases of \mathbb{R}^2 as they should:

$$UU^T = I_2,$$
$$VV^T = I_2.$$