

MIT 18.06 Exam 3 **Solutions**, Spring 2022  
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**Problem 1 (10+10+10 points):**

The matrix

$$A = \begin{pmatrix} 3 & 1 \\ 2 & 2 \end{pmatrix}$$

has an eigenvalue  $\lambda_1 = 1$  and corresponding eigenvector  $x_1 = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$ .

- (a) What is the other eigenvalue  $\lambda_2$  and a corresponding eigenvector  $x_2 = \begin{pmatrix} 1 \\ ?? \end{pmatrix}$ ?
- (b)  $B$  is a  $2 \times 2$  matrix such that  $Bx_k = (1 - \lambda_k + \lambda_k^2)x_k$  for the two eigenvectors ( $k = 1, 2$ ). What is  $B$ ?
- (c) What is  $A^{3/2} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ ?

**Solution:**

- (a)  $\text{trace}(A) = 3 + 2 = 5 = \lambda_1 + \lambda_2$ , so the other eigenvalue is  $\lambda_2 = 5 - \lambda_1 = \boxed{4}$ .  
To find a corresponding eigenvector, we need to solve

$$(A - 4I)x_2 = \begin{pmatrix} -1 & 1 \\ 2 & -2 \end{pmatrix} x_2 = 0.$$

By inspection, the second column is minus the first, so a solution is

$x_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  or any multiple thereof (but you were requested to scale  $x_2$  so that the first component = 1).

- (b)  $Bx_k = (1 - \lambda_k + \lambda_k^2)x_k$  is an eigen-equation:  $B$  has the same eigenvectors as  $A$  but with the eigenvalues replaced by  $1 - \lambda_k + \lambda_k^2$ . That means that

$$B = I - A + A^2 = \begin{pmatrix} 1 & \\ & 1 \end{pmatrix} - \begin{pmatrix} 3 & 1 \\ 2 & 2 \end{pmatrix} + \underbrace{\begin{pmatrix} 3 & 1 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ 2 & 2 \end{pmatrix}}_{\begin{pmatrix} 11 & 5 \\ 10 & 6 \end{pmatrix}} = \boxed{\begin{pmatrix} 9 & 4 \\ 8 & 5 \end{pmatrix}}.$$

You could have also solved this by diagonalization:  $B = X \begin{pmatrix} 1 - \lambda_2 + \lambda_2^2 & \\ & 1 - \lambda_2 + \lambda_2^2 \end{pmatrix} X^{-1}$

where  $X = \begin{pmatrix} x_1 & x_2 \end{pmatrix}$  is the matrix of eigenvectors, but this may be more work since you have to compute  $X^{-1}$ , unless you happen to remember the formula for the inverse of a  $2 \times 2$  matrix.

- (c) The key trick, as usual, is that  $A^{3/2}$  multiplies an *eigenvector* (where  $A$  acts like a scalar) by  $\lambda^{3/2}$ . So, to apply  $A^{3/2}$  to an arbitrary vector, we just expand that vector in the basis of the eigenvectors and then multiply each term by  $\lambda^{3/2}$ . Here,

$$\begin{pmatrix} 1 \\ -1 \end{pmatrix} = c_1 \underbrace{\begin{pmatrix} 1 \\ -2 \end{pmatrix}}_{x_1} + c_2 \underbrace{\begin{pmatrix} 1 \\ 1 \end{pmatrix}}_{x_2} = \underbrace{\begin{pmatrix} 1 & 1 \\ -2 & 1 \end{pmatrix}}_X \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}.$$

Proceeding by Gaussian elimination, we add twice the first row to the second row to obtain:

$$\underbrace{\begin{pmatrix} 1 & 1 \\ 0 & 3 \end{pmatrix}}_U \begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \implies c_2 = 1/3, c_1 = 1 - 1/3 = 2/3.$$

(Yes, the answer requires the dread “fractions.” Sorry!) Hence

$$A^{3/2} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \frac{2}{3} \lambda_1^{3/2} x_1 + \frac{1}{3} \lambda_2^{3/2} x_2 = \frac{2}{3} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + \frac{8}{3} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \boxed{\begin{pmatrix} 10/3 \\ 4/3 \end{pmatrix}}.$$

**Problem 2 (7+7+7 points):**

$A$  is a square matrix such that  $N(A - I)$  is spanned by  $\begin{pmatrix} 1 \\ 2 \end{pmatrix}$  and  $N(A - 5I)$  is spanned by  $\begin{pmatrix} 1 \\ -2 \end{pmatrix}$

- (a) Without much calculation, you can tell that  $A$  **is** / **is not** (choose 1) Hermitian because \_\_\_\_\_.
- (b) What is  $A$ ? You can leave your answer as a **product of matrices and/or matrix inverses** without multiplying/inverting them.
- (c) What is  $e^{A+I}$ ? You can leave your answer as a **product of matrices and/or matrix inverses** without multiplying/inverting them, but your answer should not have exponentials of matrices or infinite series.

**Solution:**

- (a) The two nullspace vectors are eigenvectors of  $A$  with  $\lambda = 1$  and  $5$ , respectively, but they are clearly **not orthogonal**, so  $A$  is **not** Hermitian.
- (b) From the dimensions of the vectors,  $A$  must be a  $2 \times 2$  matrix, and we are given two eigenvectors for two eigenvalues. Hence, it is diagonalizable and

$$A = X\Lambda X^{-1} = \boxed{\begin{pmatrix} 1 & 1 \\ 2 & -2 \end{pmatrix} \begin{pmatrix} 1 & \\ & 5 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 2 & -2 \end{pmatrix}^{-1}}$$

You **weren't required** to simplify it further, but it turns out that  $A = \begin{pmatrix} 3 & -1 \\ -4 & 3 \end{pmatrix}$  if you work it all out.

- (c)  $e^{A+I}$  has the same eigenvectors as  $A$ , with the eigenvalues replaced by  $\lambda \rightarrow e^{\lambda+1}$ . So, we can again use the diagonalization

$$e^{A+I} = \boxed{\begin{pmatrix} 1 & 1 \\ 2 & -2 \end{pmatrix} \begin{pmatrix} e^2 & \\ & e^6 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 2 & -2 \end{pmatrix}^{-1}}$$

**Problem 3 (4+4+4+4+4+4 points):**

For each of the following, say whether it **must** be true, it **may** be true, or it **cannot** be true. No justification needed.

- (a) If a matrix is diagonalizable, it **must/may/cannot** have orthogonal eigenvectors.
- (b)  $M$  is a Markov matrix. If  $M^n x$  converges to a steady state as  $n \rightarrow \infty$  for *any* vector  $x$ , the  $M$  **must/may/cannot** be a positive Markov matrix (i.e. have all entries  $> 0$ ).
- (c) If a matrix  $A$  is *not* diagonalizable, then  $\det(A - \lambda I)$  **must/may/cannot** have repeated roots.
- (d) If  $A^n x$  goes to zero as  $n \rightarrow \infty$  for *some*  $x$ , then  $A$  **must/may/cannot** have an eigenvalue  $\lambda$  with  $|\lambda| > 1$ .
- (e) If  $e^{At} x$  goes to zero as  $t \rightarrow \infty$  for *every*  $x$ , then  $A$  **must/may/cannot** have an eigenvalue  $\lambda$  with  $|\lambda| > 1$ .
- (f) If  $A$  has an eigenvector  $\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ , then it **must/may/cannot** have an eigenvector  $\begin{pmatrix} -3 \\ -6 \\ -9 \end{pmatrix}$ .

**Solution:**

- (a) **May.** (All “normal” matrices  $AA^H = AA^H$ , such as Hermitian matrices, are diagonalizable with orthogonal eigenvectors, but the converse is not true: not all diagonalizable matrices are normal. On the other hand, all diagonalizable matrices are *similar* to normal matrices, so there is *some* change of basis in which their eigenvectors are orthogonal.)
- (b) **May.** (All *positive* Markov matrices *must* yield a steady state—they have a single  $\lambda = 1$  eigenvalue and all others have  $|\lambda| < 1$ , but the converse is not true: a Markov matrix with zero entries *may* still have a single  $|\lambda| = 1$  eigenvalue. On the other hand, although *any* Markov matrix *must* have a  $\lambda = 1$  eigenvalue, it *may* also have other eigenvalues like  $\lambda = -1$  that can cause  $M^n x$  to oscillate forever without converging.)
- (c) **Must.** Non-diagonalizable (defective) matrices can only arise when the characteristic polynomial has repeated roots. (The converse is not true, however: a matrix with repeated eigenvalues *may* still be diagonalizable.)
- (d) **May.** Even if there is some  $|\lambda_k| > 1$ , you can still get decaying  $A^n x$  if  $x$  is chosen to be an eigenvector  $x_j$  of a different eigenvalue with  $|\lambda_j| < 1$ , or to be a linear combination of such eigenvectors.

(e) **May.** For  $e^{At}x$  to decay, all of its eigenvalues must have *negative real parts*. This is unrelated to the magnitude  $|\lambda|$ . For example, it could have an eigenvalue  $\lambda = -2$ .

(f) **Must.**  $\begin{pmatrix} -3 \\ -6 \\ -9 \end{pmatrix} = -3 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ , and all nonzero multiples of an eigenvector are also eigenvectors (of the same eigenvalue).

#### Problem 4 (25 points):

Suppose  $A$  is a real-symmetric matrix with eigenvalues  $\lambda_1 = 1$ ,  $\lambda_2 = 3$ ,  $\lambda_3 = 0$ , and  $\lambda_4 = 7$ , with corresponding eigenvectors:

$$x_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, x_2 = \begin{pmatrix} 1 \\ -1 \\ 1 \\ -1 \end{pmatrix}, x_3 = \begin{pmatrix} 1 \\ 1 \\ -1 \\ -1 \end{pmatrix}, x_4 = \begin{pmatrix} 1 \\ -1 \\ -1 \\ 1 \end{pmatrix}.$$

Now, we construct a sequence of vectors  $y_0, y_1, y_2, \dots$  where each vector  $y_{k+1}$  in the sequence is computed from the previous vector  $y_k$  by solving

$$(A - 2I)y_{k+1} = y_k$$

for  $y_{k+1}$ . If  $y_0 = \begin{pmatrix} 4 \\ 3 \\ 2 \\ 1 \end{pmatrix}$ , **give a good approximation** for  $y_{100}$ .

#### Solution:

Rearranging, we have  $y_{k+1} = (A - 2I)^{-1}y_k$ , so

$$y_k = (A - 2I)^{-k}y_0.$$

For  $k = 100$ , this will be dominated by the largest  $|\lambda|$  eigenvalues of  $(A - 2I)^{-1}$ , but this matrix has the **same eigenvectors** as  $A$  with its eigenvalues  $\lambda$  replaced by  $\frac{1}{\lambda - 2}$ . So, the eigenvalues of  $(A - 2I)^{-1}$  are

$$\frac{1}{\lambda_1 - 2} = -1, \frac{1}{\lambda_2 - 2} = 1, \frac{1}{\lambda_3 - 2} = -\frac{1}{2}, \text{ and } \frac{1}{\lambda_4 - 2} = \frac{1}{5}.$$

Of these, the largest magnitudes are  $-1$  and  $+1$ , which both have magnitude 1, so  $y_{100}$  will be dominated by the  $x_1$  and  $x_2$  terms in the expansion of  $y_0$ . More explicitly, if we expand  $y_0$  in the basis of eigenvectors:

$$y_0 = c_1x_1 + c_2x_2 + c_3x_3 + c_4x_4,$$

then

$$y_{100} = (A - 2I)^{-100}y_0 = (-1)^{100}c_1x_1 + 1^{100}c_2x_2 + \left(-\frac{1}{2}\right)^{100}c_3x_3 + \left(\frac{1}{5}\right)^{100}c_4x_4 \approx c_1x_1 + c_2x_2.$$

To compute this explicitly, we merely need to compute  $c_1$  and  $c_2$ . But  $A$  is Hermitian and hence the eigenvectors must be (and are) **orthogonal**, so we just

need **orthogonal projection** to compute the coefficients of the basis expansion:

$$c_1 = \frac{x_1^T}{x_1^T x_1} y_0 = \frac{1}{4} (1 \ 1 \ 1 \ 1) \begin{pmatrix} 4 \\ 3 \\ 2 \\ 1 \end{pmatrix} = \frac{5}{2},$$

$$c_2 = \frac{x_2^T}{x_2^T x_2} y_0 = \frac{1}{4} (1 \ -1 \ 1 \ -1) \begin{pmatrix} 4 \\ 3 \\ 2 \\ 1 \end{pmatrix} = \frac{1}{2}.$$

Therefore,

$$y_{100} \approx \frac{5}{2} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ -1 \\ 1 \\ -1 \end{pmatrix} = \boxed{\begin{pmatrix} 3 \\ 2 \\ 3 \\ 2 \end{pmatrix}}.$$

Note that the next biggest term is on the order of  $\frac{1}{2^{100}} \approx 7.9 \times 10^{-31}$ , so this approximation is pretty darn good! Actually, the  $c_3 = 1$  term is the only correction, since  $c_4 = 0$  ( $x_4^T y_0 = 0$ ).