Math 104 Applied Matrix Theory Mock Final Solution

Problem 1. (10pts) Markov Process

Trump's moods tend to be variable. If he is in a good mood today, there is a 80% chance he will still be in a good mood tomorrow; but if he is grumpy today, there is only 60% chance that his mood will be good tomorrow.

(a) (5 pts.) Write down the Markov matrix M for Trump's moods.

$$M = \left[\begin{array}{cc} 0.8 & 0.6 \\ 0.2 & 0.4 \end{array} \right]$$

(b) (5 pts.) Over the long term, what percentage of the time is Trump in a good mood? M is a positive matrix, i.e. all its entries are positive. Hence by Perron-Frobenius, the equilibrium vector of M is the unique stationary vector of M. $\mathcal{N}(M-I)$ is spanned by $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$, so with probability $\frac{3}{3+1} = 75\%$, Trump will be in a good mood in the long term.

Problem 2. (30pts) Skew-Symmetric Matrices

Let A be an $n \times n$ real skew-symmetric matrix, i.e. $A^T = -A$, and all entries of A are real. Suppose $\det(A) \neq 0$ and n is even. Let

$$B = A^T A = -A^2.$$

(a) (10 pts.) Show that the eigenvalues of B are all positive. $\det(A) \neq 0$, A is invertible. Suppose $B\mathbf{v} = \mu\mathbf{v}$, where $\mathbf{v} \neq 0$. $A\mathbf{v} \neq \mathbf{0}$, since A is invertible.

$$0 < ||A\mathbf{v}||^2 = \mathbf{v}^T A^T A \mathbf{v} = \mathbf{v}^T B \mathbf{v} = \mu \mathbf{v}^T \mathbf{v} = \mu ||\mathbf{v}||^2, \qquad \mu > 0$$

- (b) (10 pts.) Show that the eigenvalues of A are purely imaginary. Since $\det(A) \neq 0$, A is invertible and its eigenvalues are all nonzero. A has eigenvalues $\lambda_1, \dots, \lambda_n$, some may be complex and/or repeated. $B = A^T A = -A^2$ has eigenvalues $-\lambda_k^2$. By the previous part, $\mu = -\lambda_k^2 > 0$, each λ_k must be purely imaginary.
- (c) (10 pts.) Show that $\det(A) > 0$. The complex eigenvalues of real A occur in complex conjugate pairs. So the eigenvalues are $\pm a_k i$ for some real $a \neq 0$. $\det(A)$, the product of all the eigenvalues are $\prod_k a_k^2 > 0$.

Problem 3. (35pts) Symmetric Matrices

Fix $a \in \mathbb{R}$, consider the symmetric matrix

$$A = \begin{bmatrix} 1 & 1 & a+1 \\ 1 & a+1 & 1 \\ a+1 & 1 & 1 \end{bmatrix} = J + aP$$

where J is the 3×3 matrix all of whose entries are 1 and P is the reverse-order identity matrix. Note that each row of A sums to a + 3.

(a) (10 pts.) Compute A^TA using the fact that A = J + aP. Express your result in the form $A^TA = \alpha I + \beta J$, where I is the 3×3 identity matrix.

$$A^{T}A = (J + aP)^{T}(J + aP) = (J + aP)(J + aP)$$
$$= J^{2} + aPJ + aJP + a^{2}P^{2} = 3J + 2aJ + a^{2}I = a^{2}I + (2a + 3)J$$

(b) (15 pts.) Find the singular values of A. Hint: You may use the fact that $A^TA = \alpha I + \beta J$ has eigenvector **1**. Compute $Tr(\alpha I + \beta J)$ and $\det(A^TA - \alpha I)$ to deduce the eigenvalues of A^TA ,

$$A^{T}A\mathbf{1} = (\alpha I + \beta J)\mathbf{1} = (\alpha + 3\beta)\mathbf{1} = (a^{2} + 3(2a + 3))\mathbf{1} = (a + 3)^{2}\mathbf{1}$$

 $Tr(\alpha I + \beta J) = 3\alpha + 3\beta = \alpha + 3\beta + \lambda_{2} + \lambda_{3}$
 $\det(A^{T}A - \alpha I) = \det(\beta J) = 0$

So α is an eigenvalue of A^TA . Since $\lambda_2 + \lambda_3 = 2\alpha$, we have $\lambda_2 = \lambda_3 = \alpha$. The singular values of A are the positive square roots of the eigenvalues of M:

$$|a+3|, |a|, |a|$$

$$||A||_F, ||A||_1, ||A||_2, ||A||_{\infty}$$

$$||A||_F = \sqrt{(a+3)^2 + 2a^2} = \sqrt{3}\sqrt{a^2 + 2a + 3}$$

$$||A||_1 = \max_{j=1}^n \sum_{i=1}^n |a_{ij}| = 2 + |a+1|$$

$$||A||_{\infty} = \max_{i=1}^n \sum_{j=1}^n |a_{ij}| = 2 + |a+1|$$

$$||A||_2 = \max\{|a+3|, |a|\}$$

Problem 4. (40pts.) SVD Let

$$A = \left[\begin{array}{rrrr} 1 & -1 & -1 \\ 1 & -1 & 1 \\ 1 & -1 & 0 \\ 2\sqrt{2} & 0 & 0 \end{array} \right]$$

(a) (20 pts.) Find the Reduced Singular Value Decomposition (RSVD) of A.

$$A^T A = \left[\begin{array}{rrr} 11 & -3 & 0 \\ -3 & 3 & 0 \\ 0 & 0 & 2 \end{array} \right]$$

which has characteristic polynomial

$$(\lambda - 2)(\lambda^{2} - 14\lambda + 24) = 0$$

$$\lambda_{1} = 12, \quad \mathbf{v_{1}} = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 & -1 & 0 \end{bmatrix}^{T}$$

$$\lambda_{2} = 2, \quad \mathbf{v_{2}} = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 & 3 & 0 \end{bmatrix}^{T}$$

$$\lambda_{3} = 2, \quad \mathbf{v_{3}} = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}^{T}$$

$$A^{T}A = VEV^{T}, \quad V = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 & 1 & 0 \\ -1 & 3 & 0 \\ 0 & 0 & \sqrt{10} \end{bmatrix}, \quad E = \begin{bmatrix} 12 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

$$A = UDV^{T}, \quad D = \begin{bmatrix} 2\sqrt{3} & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \sqrt{2} \end{bmatrix}$$

$$\mathbf{u_{1}} = \frac{A\mathbf{v_{1}}}{\sqrt{\lambda_{1}}} = \frac{1}{\sqrt{30}} \begin{bmatrix} 2 \\ 2 \\ 2 \\ 3\sqrt{2} \end{bmatrix}$$

$$\mathbf{u_{2}} = \frac{A\mathbf{v_{2}}}{\sqrt{\lambda_{2}}} = \frac{1}{\sqrt{5}} \begin{bmatrix} -1 \\ -1 \\ -1 \\ \sqrt{2} \end{bmatrix}$$

$$\mathbf{u_{3}} = \frac{A\mathbf{v_{3}}}{\sqrt{\lambda_{3}}} = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

$$U = \begin{bmatrix} \mathbf{u_{1}} & \mathbf{u_{2}} & \mathbf{u_{3}} \end{bmatrix} = \frac{1}{\sqrt{30}} \begin{bmatrix} 2 & -\sqrt{6} & -\sqrt{15} \\ 2 & -\sqrt{6} & \sqrt{15} \\ 2 & -\sqrt{6} & 0 \\ 3\sqrt{2} & 2\sqrt{2} & 0 \end{bmatrix}$$

(b) (10 pts.) Find the least squares solution to
$$A\mathbf{x} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ \sqrt{2} \end{bmatrix}$$
.

$$A^{T}A\vec{x} = A^{T} \begin{bmatrix} 1\\1\\1\\\sqrt{2} \end{bmatrix}, \quad \mathbf{x} = (A^{T}A)^{-1}A^{T} \begin{bmatrix} 1\\1\\1\\\sqrt{2} \end{bmatrix}$$
$$(A^{T}A)^{-1} = \frac{1}{24} \begin{bmatrix} 3 & 3 & 0\\3 & 11 & 0\\0 & 0 & 12 \end{bmatrix}, \quad (A^{T}A)^{-1}A^{T} = \frac{1}{24} \begin{bmatrix} 0 & 0 & 0 & 6\sqrt{2}\\-8 & -8 & -8 & 6\sqrt{2}\\-12 & 12 & 0 & 0 \end{bmatrix}$$

Alternatively,

$$A^{+} = VD^{-1}U^{T} = (A^{T}A)^{-1}A^{T} = \frac{1}{24} \begin{bmatrix} 0 & 0 & 0 & 6\sqrt{2} \\ -8 & -8 & -8 & 6\sqrt{2} \\ -12 & 12 & 0 & 0 \end{bmatrix}$$
$$\mathbf{x} = A^{+} \begin{bmatrix} 1 \\ 1 \\ 1 \\ \sqrt{2} \end{bmatrix} = \begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \\ 0 \end{bmatrix}$$

(c) (10 pts.) Find the least squares solution of minimal norm to $A^T \vec{y} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$. The pseudoinverse to A^T is

$$(A^{T})^{+} = (A^{+})^{T} = A(A^{T}A)^{-1}.$$

$$\mathbf{y} = (A^{T})^{+} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = (A^{+})^{T} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{24} \begin{bmatrix} 0 & -8 & -12 \\ 0 & -8 & 12 \\ 0 & -8 & 0 \\ 6\sqrt{2} & 6\sqrt{2} & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -\frac{5}{6} \\ \frac{1}{6} \\ -\frac{1}{1} \\ 1 \end{bmatrix}$$

Alternatively,

$$\mathbf{y} = A(A^T A)^{-1} \begin{bmatrix} 1\\1\\1 \end{bmatrix} = \frac{1}{24} \begin{bmatrix} -20\\4\\-8\\12\sqrt{2} \end{bmatrix}$$

Problem 5. (30 pts.) **System of Differential Equations** Consider the following system of differential equations

$$x'(t) = y(t)$$

 $y'(t) = -2x(t) + 3y(t) + z(t)$
 $z'(t) = x(t) - y(t)$

Given the initial conditions

$$x(0) = 1,$$
 $y(0) = 0,$ $z(0) = 1,$

solve for x(t), y(t) and z(t).

$$\frac{d}{dt} \begin{bmatrix} x(t) \\ y(t) \\ z(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ -2 & 3 & 1 \\ 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} x(t) \\ y(t) \\ z(t) \end{bmatrix} = A \begin{bmatrix} x(t) \\ y(t) \\ z(t) \end{bmatrix}$$
$$A = \begin{bmatrix} 0 & 1 & 0 \\ -2 & 3 & 1 \\ 1 & -1 & 0 \end{bmatrix}$$

has characteristic polynomial

$$\pi_A(\lambda) = \det(A - \lambda I) = (\lambda - 1)^3 = 0.$$

n(A-I)=1, so the Jordan canonical form of A has one Jordan block. A has eigenvector $\mathbf{v_1}=\begin{bmatrix}1\\1\\0\end{bmatrix}$,

$$(A-I)\mathbf{v_2} = \mathbf{v_1}, \qquad \mathbf{v_2} = \begin{bmatrix} 0\\1\\-1 \end{bmatrix}$$

 $(A-I)\mathbf{v_3} = \mathbf{v_2}, \qquad \mathbf{v_3} = \begin{bmatrix} 1\\1\\1 \end{bmatrix}$

A has 1-chain basis $\{\mathbf{v_1}, \mathbf{v_2}, \mathbf{v_3}\}$.

$$A\begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}, \quad AX = XJ$$

$$\exp(Jt) = \begin{bmatrix} e^t & te^t & \frac{t^2}{2}e^t \\ 0 & e^t & te^t \\ 0 & 0 & e^t \end{bmatrix}$$

$$\mathbf{x}(t) = e^{At}\mathbf{x}(0) = Xe^{Jt}X^{-1}\mathbf{x}(0)$$

$$= \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} e^t & te^t & \frac{t^2}{2}e^t \\ 0 & e^t & te^t \\ -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} e^t - te^t \\ -te^t \\ e^t \end{bmatrix}$$

You can use Cramer's Rule to find X^{-1} for a 3×3 matrix, or note that

$$\begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

implies

$$a+g=1, b=-h, c=-i, a+d+g=0, b+e+h=1, c+f+i=0, d=g, e=h, i-f=1$$

$$a=-2d, a+d=1, d=g=-1, a=2$$

$$b=-h=-e, b+2e=1, e=h=1, b=-1$$

$$c=-i, f=0, i=1=-c$$

For more practice with finding Jordan Normal Form and Jordan chain basis, you can try

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

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

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

To come up with more numerical examples, you can use reverse engineering. Come up with say 3 linearly independent vectors $\mathbf{v_1}, \mathbf{v_2}, \mathbf{v_3} \in \mathbb{R}^3$. Let X be a 3×3 matrix with $\mathbf{v_i}$'s as its columns. Suppose

$$(A-3I)\mathbf{v_1} = \mathbf{0}, \qquad (A-3I)\mathbf{v_2} = \mathbf{v_1}, \qquad (A-3I)\mathbf{v_3} = \mathbf{v_2}.$$

then compute

$$A = X \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} X^{-1}.$$

Now start with A, cover up the $\mathbf{v_i}$'s, see if you can recover the Jordan chain basis and the Jordan normal form.

You can be more ambitious and try with 4×4 matrix examples etc.

Problem 6. (35pts.) Magic Squares

A magic square is an $n \times n$ array of numbers such that each row, each column, and the two <u>main diagonals</u>, all have the same <u>sum</u>. Call this sum the magic sum. For example,

(a) (4 pts.) Let S be the set of all $n \times n$ magic squares. Show that S is a vector subspace of \mathbb{R}^{n^2} .

Just check that if A and B are magic squares with magic sums a and b respectively, then A+B is a magic square with magic sum a+b; and kA is a magic square with magic sum ka.

(b) (8 pts.) Find the dimension of the subspace of all 2×2 magic squares.

a	b	has magic sum
c	d	nas magic sun

$$a + b = c + d = a + c = b + d = a + d = b + c = s$$

which implies

$$a = b = c = d$$

So the dimension of the subspace of all 2×2 magic square is 1.

(c) (10 pts.) Show that the magic sum of a 3×3 magic square is three times its central entry. Let the central entry of a magic square be x, and the magic sum be s. Summing over all the entries in the square, we have 3s, $3 \times$ row sum s. Summing the middle row, middle column, and the two diagonals, we have

$$4s = 3s + 3x \Longrightarrow s = 3x$$

(d) (13 pts.) Find the dimension of the subspace of all 3×3 magic squares.

Hint: Denote the central entry by x, the top left entry by x + y, and the top right entry by x + z, and work out the rest of the entries in the magic square.

Denote the central entry by x, the top left entry by x + y, and the top right entry by x + z, s = 3x, we can fill out the magic square accordingly

x+y	x-y-z	x + z		1	1	1		1	-1	0		0	-1	1
x-y+z	x	x+y-z	$=x\cdot$	1	1	1	$+y\cdot$	-1	0	1	$+z\cdot$	1	0	-1
x-z	x + y + z	x - y		1	1	1		0	1	-1		-1	1	0

Hence the space of 3×3 magic squares is 3-dimensional with basis

(1	1	1		1	-1	0		0	-1	1)
₹	1	1	1	,	-1	0	1	,	1	0	-1	}
l	1	1	1		0	1	-1		-1	1	0	J