

Computational Techniques 2009-2010

- 1 a) The range space, column space or the range of matrix A , where $A \in \mathbb{R}^{m \times n}$, is defined as the set of all possible linear combinations of the column vectors of matrix A . This is also referred to as the span of A .

The null space, or kernel, of a matrix $A \in \mathbb{R}^{m \times n}$, is defined as the set of solutions to the homogeneous equation:

$$A\bar{v} = 0 \quad (\text{where } \bar{v} \text{ is a } n \times 1 \text{ vector})$$

In order to prove that a subset is a subspace, it needs to satisfy the following three properties:

If W is a set of one or more vectors from a vector space V , then W is a subspace of V if and only if the following 3 conditions hold:

- the subspace W contains the zero vector
- if \bar{u} and \bar{v} are vectors in W , then $\bar{u} + \bar{v}$ is in W
→ we say that the subspace is closed under addition.
- if $k \in \mathbb{R}$, i.e. it is a real number scalar, and \bar{v} is any vector in W , then $k\bar{u}$ is also in the subspace W .
→ we say that the subspace is closed under scalar multiplication

EXTRA STUFF

$$A = \begin{bmatrix} \bar{v}_1 & \bar{v}_2 & \dots & \bar{v}_n \end{bmatrix}$$

\swarrow rows \downarrow columns

$$\bar{v}_1, \bar{v}_2, \dots, \bar{v}_n \in \mathbb{R}^m$$

The range of A , $R(A)$, is a subspace of \mathbb{R}^m

THEOREM: If A is an $m \times n$ matrix, then $R(A)$ is a subspace of \mathbb{R}^m

PROOF: First of all, notice that if \bar{y} is in $R(A)$, then $A\bar{x} = \bar{y}$ for some $\bar{x} \in \mathbb{R}^n$. Since A is $m \times n$ and \bar{x} is $n \times 1$, $A\bar{x} = \bar{y}$ will be $m \times 1$. That is \bar{y} will be in \mathbb{R}^m . This shows that $R(A)$ is a subset of \mathbb{R}^m .

$${}_m \begin{bmatrix} A \end{bmatrix} {}_n \begin{bmatrix} \bar{x} \end{bmatrix} = \begin{bmatrix} \bar{y} \end{bmatrix} {}_m$$

Verifying that the range space, $R(A)$ is a subspace.

Let $W = R(A)$

$$m \begin{bmatrix} A \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix}_n = \begin{bmatrix} \vec{y} \end{bmatrix}_m$$

1) Let 0_m denote the zero vector in \mathbb{R}^m and 0_n denote the zero vector in \mathbb{R}^n . Notice that $A 0_n = 0_m$. Hence $A \vec{x} = 0_m$ is satisfied by at least one $\vec{x} \in \mathbb{R}^n$, namely $\vec{x} = 0_n$. Thus, $0_m \in W$. ■

2) Suppose $\vec{y}, \vec{z} \in R(A)$. Let $\vec{x}_1, \vec{x}_2 \in \mathbb{R}^n$ be such that $A \vec{x}_1 = \vec{y}$ and $A \vec{x}_2 = \vec{z}$. Then $\vec{y} + \vec{z} = A \vec{x}_1 + A \vec{x}_2 = A(\vec{x}_1 + \vec{x}_2) \in R(A)$ ■

[This means that the vector $(\vec{x}_1 + \vec{x}_2)$ is sent by A to $(\vec{y} + \vec{z})$ and we wanted to prove that there is such a vector].

3) Suppose $\vec{y} \in R(A)$. Let $c \in \mathbb{R}$ (c is a real number scalar). Since $\vec{y} \in R(A)$, there exists a vector $\vec{v} \in \mathbb{R}^n$ such that $A \vec{v} = \vec{y}$. Consider $A \vec{x} = c \vec{y}$. Then let $\vec{x} = c \vec{v}$, a vector in \mathbb{R}^n . Then $A \vec{x} = A(c \vec{v}) = c(A \vec{v}) = c \vec{y}$. Therefore $c \vec{y}$ is in $R(A)$. ■

Verifying that the null space, $N(A)$ is a subspace.

Let $W = N(A)$

1) Notice that $A 0_n = 0_m$. Hence the equation $A \vec{x} = 0_m$ is satisfied by $\vec{x} = 0_n$. It follows that $0_n \in W$. ■

2) Suppose that $\vec{x}_1, \vec{x}_2 \in W$. This means that $A \vec{x}_1 = 0_m$ and $A \vec{x}_2 = 0_m$. Let $\vec{x} = \vec{x}_1 + \vec{x}_2$. Then $A \vec{x} = A(\vec{x}_1 + \vec{x}_2) = A \vec{x}_1 + A \vec{x}_2 = 0_m + 0_m = 0_m$. Therefore $\vec{x} = \vec{x}_1 + \vec{x}_2 \in W$. ■

3) Suppose that $\vec{x}_1 \in W$. Let $c \in \mathbb{R}$ (a real scalar). Since $\vec{x}_1 \in W$, we have $A \vec{x}_1 = 0_m$. Let $\vec{x} = c \vec{x}_1$. Then: $A \vec{x} = A(c \vec{x}_1) = c(A \vec{x}_1) = c 0_m = 0_m$. Therefore $\vec{x} = c \vec{x}_1 \in W$. ■

EXTRA STUFF

THEOREM: If A is an $m \times n$ matrix then $N(A)$ is a subspace of \mathbb{R}^n

PROOF: First of all notice that if \vec{x} is in $N(A)$, then $A \vec{x} = 0_m$.

Since A is $m \times n$ and $A \vec{x}$ is $m \times 1$, it follows that \vec{x} must be $n \times 1$. That is $\vec{x} \in \mathbb{R}^n$. Therefore $N(A)$ is a subset of \mathbb{R}^n .

b) i) To find the range space of A , perform Gaussian Elimination to achieve echelon form first.

$$A = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ -1 & 0 & 1 \\ 2 & 1 & -3 \end{bmatrix} \xrightarrow{\substack{R_1 := R_4 \\ R_2 := R_1 \\ R_4 := R_2}} \begin{bmatrix} 2 & 1 & -3 \\ 1 & -1 & 0 \\ -1 & 0 & 1 \\ 0 & 1 & -1 \end{bmatrix} \xrightarrow{R_3 := R_2 + R_3} \begin{bmatrix} 2 & 1 & -3 \\ 1 & -1 & 0 \\ 0 & -1 & 1 \\ 0 & 1 & -1 \end{bmatrix}$$

$$\xrightarrow{R_4 := R_4 + R_3} \begin{bmatrix} 2 & 1 & -3 \\ 1 & -1 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{R_1 := R_1 - R_2} \begin{bmatrix} 1 & 2 & -3 \\ 1 & -1 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{R_2 := R_2 - R_1}$$

$$\begin{bmatrix} 1 & 2 & -3 \\ 0 & -3 & 3 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{R_2 := R_2/3} \begin{bmatrix} 1 & 2 & -3 \\ 0 & -1 & 1 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{R_3 := R_3 - R_2} \begin{bmatrix} 1 & 2 & -3 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\xrightarrow{R_2 := -1 \times R_2} \begin{bmatrix} 1 & 2 & -3 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \begin{array}{l} \text{Those columns which are linearly} \\ \text{independent have leading 1's} \\ \therefore \text{First two columns are linearly independent.} \end{array}$$

$$\text{range}(A) = \left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \\ 2 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \\ 1 \end{bmatrix} \right\} \quad \text{dimension} = 2.$$

$$A^T = \begin{bmatrix} 1 & 0 & -1 & 2 \\ -1 & 1 & 0 & 1 \\ 0 & -1 & 1 & -3 \end{bmatrix} \xrightarrow{R_2 := R_2 + R_1} \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & 1 & -1 & 3 \\ 0 & -1 & 1 & -3 \end{bmatrix} \xrightarrow{R_3 := R_2 + R_3}$$

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

$$\text{null}(A^T) = \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & 1 & -1 & 3 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = 0 \quad \therefore \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -2 \\ -3 \\ 0 \\ 1 \end{bmatrix}$$

$$\left. \begin{array}{l} x_1 - x_3 + 2x_4 = 0 \\ x_2 - x_3 + 3x_4 = 0 \end{array} \right\} \quad \begin{array}{l} x_1 = x_3 - 2x_4 \\ x_2 = x_3 - 3x_4 \end{array}$$

$$\therefore \text{null}(A^T) = \text{span} \left(\begin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ -3 \\ 0 \\ 1 \end{bmatrix} \right)$$

dimension = 2.

$$\begin{aligned}
 \text{ii) } \text{range}(A) &\Rightarrow \text{dimension} = 2 \\
 \text{range}(A^T) &\Rightarrow \text{dimension} = 2 \\
 \text{null}(A) &\Rightarrow \text{dimension} = n - r = 3 - 2 = 1 \\
 \text{null}(A^T) &\Rightarrow \text{dimension} = m - r = 4 - 2 = 2
 \end{aligned}
 \left. \vphantom{\begin{aligned} \text{range}(A) \\ \text{range}(A^T) \\ \text{null}(A) \\ \text{null}(A^T) \end{aligned}} \right\} \text{ - always the same}$$

[where m is the row, n is the column or a $m \times n$ matrix]

Not exactly sure what it means by "geometric shape".

Attempted answer below:

Suppose I have 2 planes in \mathbb{R}^3 and they form a system $A\bar{x} = \bar{b}$.

The nullspace of A represents geometrically the vectors that form the intersection between the 2 planes shifted to the origin.

The ~~range~~ range of A represents the span of the normal vectors to the 2nd plane.

$$\begin{aligned}
 \text{iii) } \text{null}(A) &\perp \text{range}(A^T) \\
 \text{null}(A^T) &\perp \text{range}(A)
 \end{aligned}$$

$$\text{range}(A) = \left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \\ 2 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

$$\text{nullspace}(A) = \text{span} \left(\begin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ -3 \\ 0 \\ 1 \end{bmatrix} \right)$$

Two vectors are orthogonal if their dot product is zero.

$$\begin{bmatrix} 1 \\ 0 \\ -1 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} -2 \\ -3 \\ 0 \\ 1 \end{bmatrix} = (1 \times -2) + (0 \times -3) + (-1 \times 0) + (2 \times 1) = -2 + 0 + 0 + 2 = 0$$

(can also do for other combinations)

$$\text{c i) } A\bar{x} = \bar{b}$$

$$\begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ -1 & 0 & 1 \\ 2 & 1 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 0 \\ 2 \end{bmatrix}$$

$$\left. \begin{aligned} x_1 - x_2 &= 2 \\ x_2 - x_3 &= 1 \\ -x_1 + x_3 &= 0 \\ 2x_1 + x_2 - 3x_3 &= 2 \end{aligned} \right\} \begin{aligned} x_1 &= 2 + x_2 \\ x_2 &= 1 + x_3 \\ x_1 &= x_3 \end{aligned} \left\{ \begin{aligned} x_1 &= 2 + 1 + x_3 \\ x_1 &= x_3 \end{aligned} \right\} \begin{aligned} x_1 &= 3 + x_3 \\ x_1 &= x_3 \end{aligned} \right\}$$

inconsistent equations. \therefore no solution.

- ii) Notes on page 73/138 or 70 ^{of the background notes} go some way to explaining what to do.

This is a least squares problem, so

$$\|Ax - b\|_2 \text{ is minimised when } A^T(b - Ax) = 0$$

which in turn is equivalent to :

x minimises $\|Ax - b\|_2$ if and only if $Ax = b_R$ ^{the range of A} and

(Need some confirmation on this)

$b - Ax = b_N$ ^{the null space of A}

- 2 a i) $\text{trace}(A)$ is the sum of the diagonal entries of a matrix A .

$$\text{trace}(A) = 7 + 6 + 5 = 18$$

$$\begin{aligned} \det(A) &= 7 \begin{vmatrix} 6 & -2 \\ -2 & 5 \end{vmatrix} - -2 \begin{vmatrix} -2 & -2 \\ 0 & 5 \end{vmatrix} + 0 \begin{vmatrix} -2 & 6 \\ 0 & -2 \end{vmatrix} \\ &= 7(30 - 4) - -2(-10 - 0) + 0(4 - 0) \\ &= 7(26) + 2(-10) + 0 \\ &= 182 - 20 \\ &= 162 \end{aligned}$$

Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the eigenvalues of a n by n matrix A .

Then:

$$\det(A) = \lambda_1 \times \lambda_2 \times \dots \times \lambda_n$$

$$\text{trace}(A) = \lambda_1 + \lambda_2 + \dots + \lambda_n$$

ii) Characteristic equation $\det(A - \lambda I) = 0$

$$\left| \begin{pmatrix} 7 & -2 & 0 \\ -2 & 6 & -2 \\ 0 & -2 & 5 \end{pmatrix} - \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{pmatrix} \right| = 0$$

$$\begin{vmatrix} 7-\lambda & -2 & 0 \\ -2 & 6-\lambda & -2 \\ 0 & -2 & 5-\lambda \end{vmatrix} = 0$$

$$(7-\lambda) \begin{vmatrix} 6-\lambda & -2 \\ -2 & 5-\lambda \end{vmatrix} - (-2) \begin{vmatrix} -2 & -2 \\ 0 & 5-\lambda \end{vmatrix} + 0 \begin{vmatrix} -2 & -2 \\ 0 & 5-\lambda \end{vmatrix} = 0$$

$$(7-\lambda)((6-\lambda)(5-\lambda)-4) + 2(-2(5-\lambda)-0) + 0 = 0$$

$$(7-\lambda)(30-6\lambda-5\lambda+\lambda^2-4) + 2(-10+2\lambda) = 0$$

$$(7-\lambda)(\lambda^2-11\lambda+26) + 4\lambda-20 = 0$$

$$7\lambda^2-77\lambda+182-\lambda^3+11\lambda^2-26\lambda+4\lambda-20=0$$

$$-\lambda^3+18\lambda^2-99\lambda+162=0$$

$$\begin{array}{r} \lambda - 9 \sqrt{\frac{-\lambda^2 + 9\lambda - 18}{-\lambda^3 + 18\lambda^2 - 99\lambda + 162}} \\ \hline -\lambda^3 + 9\lambda^2 \\ \hline 9\lambda^2 - 99\lambda \\ 9\lambda^2 - 81\lambda \\ \hline -18\lambda + 162 \\ -18\lambda + 162 \\ \hline 0 \end{array}$$

$$(\lambda-9)(-\lambda^2+9\lambda-18)$$

$$(\lambda-9)(-\lambda+6)(\lambda-3)$$

\therefore eigenvalues $\lambda = 9, 6, 3$

iii) There exists a basis of orthonormal eigenvectors if and only if the matrix A is symmetric (spectral theorem).

$$\lambda = 9, 6, 3$$

For $\lambda = 9$:

$$\left. \begin{array}{l} -2x_1 - 2x_2 = 0 \\ -2x_1 - 3x_2 - 2x_3 = 0 \\ -2x_2 - 4x_3 = 0 \end{array} \right\} \begin{array}{l} -2x_1 = 2x_2 \\ -2x_2 = 4x_3 \end{array}$$

$$\begin{pmatrix} -2 & -2 & 0 \\ -2 & -3 & -2 \\ 0 & -2 & -4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\begin{array}{l} -x_1 = x_2 \\ -x_2 = 2x_3 \end{array} \therefore \bar{v}_1 = \begin{pmatrix} -2 \\ 2 \\ -1 \end{pmatrix}$$

For $\lambda = 6$

$$\begin{pmatrix} 1 & -2 & 0 \\ -2 & 0 & -2 \\ 0 & -2 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\left. \begin{array}{l} x_1 - 2x_2 = 0 \\ -2x_1 - 2x_3 = 0 \\ -2x_2 - x_3 = 0 \end{array} \right\} \begin{array}{l} x_1 = 2x_2 \\ -x_1 = x_3 \\ -2x_2 = x_3 \end{array}$$

$$\therefore \bar{V}_2 = \begin{pmatrix} 2 \\ 1 \\ -2 \end{pmatrix}$$

For $\lambda = 3$

$$\begin{pmatrix} 4 & -2 & 0 \\ -2 & 3 & -2 \\ 0 & -2 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\left. \begin{array}{l} 4x_1 - 2x_2 = 0 \\ -2x_1 + 3x_2 - 2x_3 = 0 \\ -2x_2 + 2x_3 = 0 \end{array} \right\} \begin{array}{l} 2x_1 = x_2 \\ x_3 = x_2 \end{array}$$

$$\therefore \bar{V}_3 = \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix}$$

The eigen vectors are orthogonal to each other since A is symmetric.

Normalize each vector to make it 'orthonormal':

$$\|V_1\| = \sqrt{(-2)^2 + (2)^2 + (1)^2} = 3 \quad \therefore \bar{V}_1 = \frac{1}{3} \begin{pmatrix} -2 \\ 2 \\ 1 \end{pmatrix} \quad \bar{V}_2 = \frac{1}{3} \begin{pmatrix} 2 \\ 1 \\ -2 \end{pmatrix} \quad \bar{V}_3 = \frac{1}{3} \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix}$$

$$\|V_2\| = \sqrt{(2)^2 + (1)^2 + (-2)^2} = 3$$

$$\|V_3\| = \sqrt{(1)^2 + (2)^2 + (2)^2} = 3$$

$$\text{iv) } Q = \begin{pmatrix} -2 & 2 & 1 \\ 2 & 1 & 2 \\ -1 & -2 & 2 \end{pmatrix} \quad (\text{combine eigenvectors above})$$

(We can also check that Q is non-singular by calculating determinant)

$$Q^{-1} \Rightarrow \textcircled{1} \text{ find } \det(Q) \Rightarrow$$

$$\begin{aligned} \det(Q) &= -2 \begin{vmatrix} 1 & 2 \\ -2 & 2 \end{vmatrix} - 2 \begin{vmatrix} 2 & 2 \\ -1 & 2 \end{vmatrix} + 1 \begin{vmatrix} 2 & 1 \\ -1 & -2 \end{vmatrix} \\ &= -2(2 - -4) - 2(4 + 2) + 1(-4 - -1) \\ &= -2(6) - 2(6) + (-3) \\ &= -12 - 12 - 3 \\ &= -27 \end{aligned}$$

$$\textcircled{2} Q^T = \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix}$$

$\det(Q) \neq 0$
 \therefore non-singular

③ Find determinant of minor matrices

$$Q_{11}^T = \begin{vmatrix} 1 & -2 \\ 2 & 2 \end{vmatrix} = 2 - (-4) \quad Q_{12}^T = \begin{vmatrix} 2 & -2 \\ 1 & 2 \end{vmatrix} = 4 - (-2) \quad Q_{13}^T = \begin{vmatrix} 2 & 1 \\ 1 & 2 \end{vmatrix} = 4 - 1$$

$$Q_{21}^T = \begin{vmatrix} 2 & -1 \\ 2 & 2 \end{vmatrix} = 4 - (-2) \quad Q_{22}^T = \begin{vmatrix} -2 & -1 \\ 1 & 2 \end{vmatrix} = -4 - (-1) \quad Q_{23}^T = \begin{vmatrix} -2 & 2 \\ 1 & 2 \end{vmatrix} = -4 - 2$$

$$Q_{31}^T = \begin{vmatrix} 2 & -1 \\ 1 & -2 \end{vmatrix} = -4 - (-1) \quad Q_{32}^T = \begin{vmatrix} -2 & -1 \\ 2 & -2 \end{vmatrix} = 4 - (-2) \quad Q_{33}^T = \begin{vmatrix} -2 & 2 \\ 2 & 1 \end{vmatrix} = -2 - 4$$

$$\Rightarrow \begin{pmatrix} 6 & 6 & 3 \\ 6 & -3 & -6 \\ -3 & 6 & -6 \end{pmatrix}$$

④ Apply matrix of cofactors.

$$\begin{pmatrix} 6 & 6 & 3 \\ 6 & -3 & -6 \\ -3 & 6 & -6 \end{pmatrix} \times \begin{pmatrix} + & - & + \\ - & + & - \\ + & - & + \end{pmatrix} = \begin{pmatrix} 6 & -6 & 3 \\ -6 & -3 & 6 \\ -3 & -6 & -6 \end{pmatrix}$$

$$\therefore Q^{-1} = \frac{1}{-27} \begin{pmatrix} 6 & -6 & 3 \\ -6 & -3 & 6 \\ -3 & -6 & -6 \end{pmatrix}$$

$$= \frac{-3}{-27} \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix} = \frac{1}{9} \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix}$$

v) This diagonal matrix $\Lambda = Q^{-1} A Q$

$$\Lambda = \frac{1}{9} \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix} \begin{pmatrix} 7 & -2 & 0 \\ -2 & 6 & -2 \\ 0 & -2 & 5 \end{pmatrix} \begin{pmatrix} -2 & 2 & 1 \\ 2 & 1 & -2 \\ -1 & -2 & 2 \end{pmatrix}$$

$$\Lambda = \begin{pmatrix} 9 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

Which is a matrix containing the eigenvalues corresponding to the eigenvectors in Q .

vi) A positive definite matrix is a symmetric matrix with all positive eigenvalues.

A is such a matrix since its eigen values are 9, 6, 3 which are all > 0 .

vii) Yes there exists a lower triangular matrix $L \in \mathbb{R}^{3 \times 3}$ with $A = LL^T$ since A is symmetric.

b i)

$$A = U S V^T$$

The columns of U are orthonormal eigenvectors of AA^T

S is a diagonal matrix containing the square roots of eigenvalues from U or V in descending order
the columns of V are orthonormal eigenvectors of $A^T A$

Essentially we are computing: $U^T A V = S$

~~For V~~

For V

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

$$\det(A^T A - \lambda I) = 0$$

$$\begin{vmatrix} 2-\lambda & 0 \\ 0 & 1-\lambda \end{vmatrix} = 0$$

$$(2-\lambda)(1-\lambda) = 0$$

$$\lambda = 2, 1$$

Finding eigenvectors:

$$\lambda = 2$$

$$\begin{pmatrix} 0 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$-x_2 = 0$$

$$\bar{v}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \leftarrow \text{any number}$$

$$\lambda = 1$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$x_1 = 0$$

$$\bar{v}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \leftarrow \text{any number}$$

$$\therefore \bar{V} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

ordered by descending value of eigen values.

For U

$$AA^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

$$\det(AA^T - \lambda I) = 0$$

$$\begin{vmatrix} 1-\lambda & 0 & 1 \\ 0 & 1-\lambda & 0 \\ 1 & 0 & 1-\lambda \end{vmatrix} = 0 \quad 0 = 1-\lambda \begin{vmatrix} 1-\lambda & 0 \\ 0 & 1-\lambda \end{vmatrix} - 0 \begin{vmatrix} 0 & 0 \\ 1 & 1-\lambda \end{vmatrix} + 1 \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix}$$

$$(1-\lambda)(1-\lambda)(1-\lambda) + -(1-\lambda) = 0$$

$$(1-\lambda)(1-\lambda-\lambda+\lambda^2) - 1 + \lambda = 0$$

$$(1-\lambda)(1-2\lambda+\lambda^2) - 1 + \lambda = 0$$

$$1 - 2\lambda + \lambda^2 - \lambda + 2\lambda^2 - \lambda^3 - 1 + \lambda = 0$$

$$-2\lambda - \lambda^3 + 3\lambda^2 = 0$$

$$-\lambda^3 + 3\lambda^2 - 2\lambda = 0$$

$$\lambda(-\lambda^2 + 3\lambda - 2) = 0$$

$$\lambda(-\lambda + 2)(\lambda - 1)$$

Eigenvalues: 0, 1, 2

For $\lambda = 0$

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad \left. \begin{array}{l} x_1 + x_3 = 0 \\ x_2 = 0 \\ x_1 + x_3 = 0 \end{array} \right\} \quad x_1 = -x_3 \quad \bar{u}_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

For $\lambda = 1$

$$\begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad \left. \begin{array}{l} x_3 = 0 \\ x_1 = 0 \end{array} \right\} \quad \bar{u}_2 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

For $\lambda = 2$

$$\begin{pmatrix} -1 & 0 & 1 \\ 0 & -1 & 0 \\ 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad \left. \begin{array}{l} -x_1 + x_3 = 0 \\ -x_2 = 0 \\ x_1 - x_3 = 0 \end{array} \right\} \quad \bar{u}_3 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

$$\therefore \bar{U} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix} \quad \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & -1 \end{pmatrix}$$

ordered by descending
value of eigen value.

Finally we have to convert \bar{V}_I and \bar{U}_I into ~~an~~ orthogonal matrices which we do by applying the Gram-Schmidt orthonormalization process to the column vectors.

For \bar{V}_I

$$\textcircled{1} \quad \bar{U}_1 = \frac{\bar{V}_1}{|\bar{V}_1|} \quad \bar{U}_1 = \frac{[\Phi, 0]}{\sqrt{\Phi^2 + 0^2}} = [\Phi, 0] \quad \text{(not strictly correct} \\ \rightarrow \text{dividing by } \Phi)$$

$$\textcircled{2} \quad \bar{\omega}_2 = \bar{V}_2 - \bar{U}_1 \cdot \bar{V}_2 \times \bar{U}_1 \quad \bar{\omega}_2 = \begin{bmatrix} 0 \\ \Phi \\ 0 \end{bmatrix} - \begin{bmatrix} \Phi \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ \Phi \end{bmatrix} \times \begin{bmatrix} \Phi \\ 0 \end{bmatrix} \\ \bar{U}_2 = \frac{\bar{\omega}_2}{|\bar{\omega}_2|} = \begin{bmatrix} 0 \\ \Phi \\ 0 \end{bmatrix} - 0 \times \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ \Phi \\ 0 \end{bmatrix}$$

$$\bar{U}_2 = \frac{[0, \Phi]}{\sqrt{0^2 + \Phi^2}} = [0, \Phi]$$

$$\therefore \bar{V} = \begin{pmatrix} \Phi & 0 \\ 0 & \Phi \end{pmatrix}$$

For \bar{U}_I

$$\textcircled{1} \quad \bar{U}_1 = \frac{\bar{V}_1}{|\bar{V}_1|} \quad \bar{U}_1 = \frac{[1, 0, 1]}{\sqrt{1^2 + 0^2 + 1^2}} = \frac{[1, 0, 1]}{\sqrt{2}} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$\textcircled{2} \quad \bar{\omega}_2 = \bar{V}_2 - \bar{U}_1 \cdot \bar{V}_2 \times \bar{U}_1 \quad \bar{\omega}_2 = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \frac{1}{\sqrt{2}} \cdot \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \times \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \\ \bar{U}_2 = \frac{\bar{\omega}_2}{|\bar{\omega}_2|} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} - 0 \times \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\textcircled{3} \quad \bar{\omega}_3 = \bar{V}_3 - \bar{U}_1 \cdot \bar{V}_3 \times \bar{U}_1 - \bar{U}_2 \cdot \bar{V}_3 \times \bar{U}_2 \\ \bar{U}_3 = \frac{\bar{\omega}_3}{|\bar{\omega}_3|} \quad \bar{\omega}_3 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} - \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \times \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \frac{1}{\sqrt{2}} - \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \times \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{bmatrix} \times \begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \end{bmatrix} = \begin{bmatrix} 0 \\ -\frac{1}{2} \times \frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}} \times \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}$$

$$\bar{\omega}_3 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} - \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \times \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \frac{1}{\sqrt{2}} - 0 \times \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \\ \bar{\omega}_3 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} - 0 = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$$

$$\bar{U}_3 = \frac{[1, 1, -1]}{\sqrt{1^2 + 1^2 + (-1)^2}} = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

$$\therefore U = \begin{pmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{3}} \\ 0 & 0 & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{3}} \end{pmatrix}$$

ii) The singular values of A are the square roots of the non-zero eigenvalues which then populate the diagonal.

Dimensions of A are $m \times n$ so $\Rightarrow \sqrt{2}, \sqrt{1}$

corresponding matrix:

$$\begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{1} \\ 0 & 0 \end{bmatrix} = S \leftarrow \text{Augmented!}$$

$$l_2 \text{ norm of } \|A\|_2 = \sqrt{1^2 + 0^2 + 0^2 + 1^2 + 1^2 + 0^2} \\ = \sqrt{3}$$

iii) Not sure about this one.

We have essentially computed $U^T A V = S$

$$\text{so } A = U S V^T$$

$$\text{so } \mathbb{R}^3 \text{ basis} = U = \begin{pmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{3}} \\ 0 & 0 & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{3}} \end{pmatrix}$$

$$\mathbb{R}^2 \text{ basis} = V^T = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

Not convinced that all of my answers are correct since.

$$U^T A V \neq S \text{ for some reason } \ddot{\smile}$$

3 a i) The p-norm is defined for an m-component vector \vec{v} as follows:

$$\|x\|_p = \left(\sum_{i=1}^m |x_i|^p \right)^{1/p}$$

The subordinate matrix norm $\|A\|$ is defined as

$$\|A\| = \max_{x \neq 0} \frac{\|Ax\|}{\|x\|}$$

ii) Condition number of a matrix A is defined as.

$$\text{cond}(A) = \|A\| \cdot \|A^{-1}\|$$

By convention $\text{cond}(A) = \infty$ if A singular.

can check that A is non-singular $\Rightarrow \det(A) \neq 0$

$$\begin{aligned} \|A\|_1 &= \max(1+1+1, 1+1+1, 1+1+1) \\ &= \max(3, 4, 4) \\ &= 4 \end{aligned}$$

$$\begin{aligned} A^{-1} \Rightarrow \det(A) &= 1 \begin{vmatrix} 3 & -1 \\ -1 & 1 \end{vmatrix} - 0 \begin{vmatrix} 0 & -1 \\ 2 & 1 \end{vmatrix} + 2 \begin{vmatrix} 0 & 3 \\ 2 & -1 \end{vmatrix} \\ &= 3 - 1 + 2(-6) \\ &= 2 - 12 \\ &= -10 \end{aligned}$$

$$A^T \Rightarrow \begin{bmatrix} 1 & 0 & 2 \\ 0 & 3 & -1 \\ 2 & -1 & 1 \end{bmatrix} \quad \begin{array}{l} \text{determinants} \\ \text{of minor} \\ \text{matrices} \end{array} \Rightarrow \begin{bmatrix} 2 & 2 & -6 \\ 2 & -3 & -1 \\ -6 & -1 & 3 \end{bmatrix} \quad \begin{array}{l} \text{matrix} \\ \Rightarrow \text{of} \\ \text{cofactors} \end{array}$$

$$A^{-1} = \frac{1}{-10} \begin{bmatrix} 2 & -2 & -6 \\ -2 & -3 & 1 \\ -6 & 1 & 3 \end{bmatrix}$$

$$\begin{aligned} \|A^{-1}\| &= \max\left(\left|\frac{2}{-10}\right| + \left|\frac{-2}{-10}\right| + \left|\frac{-6}{-10}\right|, \left|\frac{-2}{-10}\right| + \left|\frac{-3}{-10}\right| + \left|\frac{1}{-10}\right|, \left|\frac{-6}{-10}\right| + \left|\frac{1}{-10}\right| + \left|\frac{3}{-10}\right|\right) \\ &= \max\left(1, \frac{1}{2}, 1\right) \\ &= 1 \end{aligned}$$

$$\therefore \text{cond}(A) = 4 \times 1 = 4$$

b i) A metric space is a non-empty set S of objects (called points) together with a function $d: S \times S \rightarrow \mathbb{R}^+$ (called the metric of the space) satisfying the following four properties for all points $x, y, z \in S$:

- 1) $d(x, x) = 0$ ($d \rightarrow$ distance)
- 2) $d(x, y) > 0$ if $x \neq y$
- 3) $d(x, y) = d(y, x)$ (symmetry)
- 4) $d(x, y) \leq d(x, z) + d(z, y)$ (triangle inequality)

The non-negative number $d(x, y)$ is to be thought of as the distance from x to y

Let $S \subseteq \mathbb{R}^k$ and $d(x, y) = \|x - y\|_2 \Rightarrow$ called the euclidean metric.

If we take x, y, z to be vectors/matrices we can show that the difference is a metric by satisfying the 4 conditions above.

$$\text{ii) } (I - M)G_n = I - M^{n+1}$$

$$IG_n - MG_n = I - M^{n+1}$$

$$I(I + M + M^2 + \dots + M^n) - M(I + M + M^2 + \dots + M^n) = I - M^{n+1}$$

$$I + \cancel{M} + \cancel{M^2} + \dots + \cancel{M^n} - \{ \cancel{M} - \cancel{M^2} - \cancel{M^3} + \dots + M^{n+1} \} = I - M^{n+1}$$

$$I - M^{n+1} = I - M^{n+1} \quad \square$$

iii) If $\|M\| < 1$ then all vectors \vec{v} of M have components, m , such that each component < 1 . If $m < 1$, then let $m = \frac{1}{c}$. M^n means that each component will be multiplied by itself n times, i.e. $(\frac{1}{c})^n$.

Hence we can see that as $n \rightarrow \infty$, $(\frac{1}{c})^n \rightarrow 0$, and hence since this will happen to every component, m , of matrix M , we can see that as $n \rightarrow \infty$, $M^n \rightarrow 0$.

$$(I - M)G_n = I - M^{n+1}$$

$$(I - M)G_\infty = I - 0$$

$$(I - M)G_\infty = I$$

$$G_\infty = \frac{I}{(I - M)}$$

$$G_\infty = (I - M)^{-1}$$

as $n \rightarrow \infty$

$M^{n+1} \rightarrow 0$ since $M^n \rightarrow 0$

\Downarrow is this a valid deduction?

A a) NOT COVERED

$$b i) \frac{s-1}{(s+1)(s-2)} = \frac{A}{(s+1)} + \frac{B}{(s-2)}$$

$$= \frac{A(s-2) + B(s+1)}{(s+1)(s-2)}$$

$$s-1 = A(s-2) + B(s+1)$$

$$s-1 = As - 2A + Bs + B$$

$$s-1 = (A+B)s - 2A + B$$

Equating coefficients:

$$\left. \begin{array}{l} 1 = A+B \\ -1 = -2A+B \end{array} \right\} \begin{array}{l} A = 1-B \\ A = \frac{-1-B}{-2} \end{array}$$

$$-2(1-B) = -1-B$$

$$-2 + 2B = -1 - B$$

$$2B + B = -1 + 2$$

$$3B = 1$$

$$B = \frac{1}{3} \quad \therefore A = 1 - \frac{1}{3} = \frac{2}{3}$$

$$\therefore \frac{s-1}{(s+1)(s-2)} = \frac{2}{3(s+1)} + \frac{1}{s(s-2)}$$

$$ii) \int \dot{y}_1 = \int 4y_2 + 2e^{-t}$$

$$\int \dot{y}_1 = 4y_2 t + -2e^{-t} + C$$

$$y_1 = 4y_2 t - 2e^{-t} + C$$

$$0 = 4 \times 0 \times 1 - 2e^{-1} + C$$

$$0 = 0 - 2e^{-1} + C$$

$$2e^{-1} = C$$

$$\therefore y_1 = 4y_2 t - 2e^{-t} + 2/e$$

$$y_2 = 2y_2 t^2 + 2e^{-t} + 2/e t + \left(\frac{1}{e} - \frac{2}{e^{1/2}} \right)$$

$$\int \dot{y}_2 = \int y_1$$

$$\int \dot{y}_2 = \int 4y_2 t - 2e^{-t} + C$$

$$\int \dot{y}_2 = 2y_2 t^2 + 2e^{-t} + Ct + k$$

$$y_2 = 2y_2 t^2 + 2e^{-t} + Ct + k$$

$$0 = 2 \cdot 0 \cdot -0.5^2 + 2e^{-(0.5)} + \frac{2}{e} \times -0.5 + k$$

$$0 = \frac{2}{e^{1/2}} - \frac{1}{e} + k$$

$$\frac{1}{e} - \frac{2}{e^{1/2}} = k$$

$$k = \frac{1}{e} - \frac{2e^{1/2}}{e}$$

$$= \frac{1 - 2e^{1/2}}{e}$$