

H 310 Day

~~① If $B = M^{-1}AM + C = N^{-1}BN$~~

~~so that $C = N^{-1}(M^{-1}AM)N$~~

~~= (M \cdot N)^{-1}A(M \cdot N)~~

~~so if $T = M \cdot N$ we have $C = T^{-1}AT$. If B is similar to A~~
~~and C is similar to B then ~~T is similar to A~~~~
~~then C is similar to A~~

~~② If $C = F^{-1}AF$ and also $C = G^{-1}BG$ then~~

$$F^{-1}AF = G^{-1}BG$$

$$\text{so } B = G F^{-1} A F G^{-1} = (FG^{-1})^{-1} A (FG^{-1})$$

~~Joining $M = FG^{-1}$ we have $B = M^{-1}AM$~~

~~So if C is similar to A and C is similar to B then
 A is similar to B .~~

~~③ see or look for a M such that~~

$$A = M^{-1}BM \quad \text{or} \quad MA = BM$$

~~Let $M = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ then $MA = BM$ is given by~~

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} a & 0 \\ c & 0 \end{bmatrix} = \begin{bmatrix} c & d \\ c & d \end{bmatrix} \quad \text{so pick } d=0 \\ + a=c. \quad \text{so let } a=1 + b=2$$

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

For the next pair of $A + B$ we have

$$MA = BM$$

$$\Rightarrow \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} a+b & a+b \\ c+d & c+d \end{bmatrix} = \begin{bmatrix} a-c & b-d \\ -a+c & -b+d \end{bmatrix}$$

$$\Rightarrow \begin{array}{l} a+b = a-c \\ a+b = b-d \\ c+d = -a+c \\ c+d = -b+d \end{array} \quad \begin{array}{l} b = -c \\ a = -d \\ d = -a \\ c = -b \end{array}$$

Thus we have the relation that $b = -c$ + $a = -d$.

$$\text{Pick } a=1 + b=2 \text{ to get } M = \begin{bmatrix} 1 & 2 \\ -2 & -1 \end{bmatrix}$$

For the next pair of $A + B$ we have $MA = BM$ gives

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} a+3b & 2c+4d \\ c+3d & 2c+4d \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} a+3b & 2a+4b \\ c+3d & 2c+4d \end{bmatrix} = \begin{bmatrix} 4a+3c & 4b+3d \\ 2a+c & 2b+d \end{bmatrix}$$

$$\Rightarrow a+3b = 4a+3c \quad -3a+3b-3c=0$$

$$2a+4b = 4b+3d \quad \Rightarrow \quad 2a-3d=0$$

$$c+3d = 2a+c \quad 2a-3d=0$$

$$2c+4d = 2b+d \quad \cancel{2a} + \cancel{2c} - 2b - 2c - 3d = 0$$

Thus we have the system

$$\begin{bmatrix} -3 & 3 & -3 & 0 \\ 2 & 0 & 0 & -3 \\ 0 & 2 & -2 & -3 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = 0 \Rightarrow \begin{bmatrix} 2 & 0 & 0 & -3 \\ -3 & 3 & -3 & 0 \\ 0 & 2 & -2 & -3 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 2 & 0 & 0 & -3 \\ 0 & 3 & -3 & -\frac{9}{2} \\ 0 & 2 & -2 & -3 \end{bmatrix} \Rightarrow \begin{bmatrix} 2 & 0 & 0 & -3 \\ 0 & 1 & -1 & -\frac{3}{2} \\ 0 & 0 & 0 & -6 \end{bmatrix} \xrightarrow{\text{Row operations}} \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$(\frac{3}{2})(-3) - (\frac{3}{2})(2) + (-3) = -6 \quad \Rightarrow d=0, a=0, c=b \quad \text{so taking } b=1$$

$$\text{we have } M = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

(4) A has eigenvalues $0 + \frac{1}{2}i$ is a ~~vector~~

If A has eigenvalues $0 + \frac{1}{2}i$ it has 2 linearly ~~independent~~ independent eigenvectors and therefore can be factored into $A = S\Lambda S^{-1}$ which says that A is similar. Thus from problem 2 since every matrix with eigenvalues $0 + \frac{1}{2}i$ or similar to $\Lambda = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ then they themselves are similar.

(5) $A_1 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ has $\lambda = 1$ only

$A_2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ has $\lambda = -1 + 1$

$A_3 = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ has $\lambda = 1 + 1 = 0$

$A_4 = \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}$ has $\lambda = 1 + 1 = 0$

$A_5 = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$ has $\lambda = 1 + 1 = 0$

$A_6 = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}$ has $\lambda = 1 + 1 = 0$

Thus $A_3, A_4, A_5 + A_6$ are similar

(b) Two similar matrices will have the same eigenvalues. For the 2×2 matrices described in the book we can have eigenvalues given by

$$M = \begin{bmatrix} j^k & l^m \\ k^l & m^j \end{bmatrix} \quad \text{w/ } j, l, k, m = 0, 1$$

$$\lambda^2 - \text{Tr}(M)\lambda + \det(M) = 0$$

so $\text{free}(M) = \cancel{j+k+m} j+m$ only 4 choices

$$+ \det(M) = \cancel{j+k+l} jm - kl$$

$$\lambda = T \pm \sqrt{T^2 - 4D}$$

$j \ l \ k \ m$	$\text{Tr}(M)$	$\det(M)$	λ 's
0 0 0 0	2	0	0
0 0 0 1	1	0	$\frac{1 \pm i}{2} = 1, 0$
0 0 1 0	0	0	0
0 0 1 1	1	0	1, 0
0 1 0 0	0	0	0
0 1 0 1	1	0	1, 0
0 1 1 0	0	-1	$\frac{0 \pm \sqrt{-4}}{2} = \pm i$
0 1 1 1	1	-1	$\frac{1 \pm \sqrt{1+4}}{2} = \frac{1 \pm \sqrt{5}}{2}$
1 0 0 0	1	0	1, 0
1 0 0 1	2	1	$\frac{2 \pm \sqrt{4-4}}{2} = 1$
1 0 1 0	1	0	1, 0
1 0 1 1	2	1	$\frac{2 \pm \sqrt{4-4}}{2} = 1$
1 1 0 0	1	0	1, 0
1 1 0 1	2	1	1
1 1 1 0	1	-1	$\frac{1 \pm \sqrt{5}}{2}$
1 1 1 1	2	0	$\frac{2 \pm \sqrt{4}}{2} = \frac{2 \pm 2}{2} = 0, 2$

For each set of repeated eigenvalues we can have a matrix that is diagonalizable or not. These ~~represent~~ two matrices denote two different "families" since the diagonalizable matrices can be ^{factored} ~~decomposed~~ into a diagonal matrix, while the non-diagonal matrices can only be factored into a Jordan block structure like $\begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix}$. Thus we have the following

sets of similar matrices w/ (λ_1, λ_2) given by

(λ_1, λ_2)	Property	Number of matrices in each family
$(0, 0)$	diagonalizable	3 shared
$(0, 0)$	non-diagonalizable	3 shared
$(2, 1)$ $(0, 0)$	diagonalizable	6
$(-1, 1)$	diagonalizable	1
$(\frac{1-\sqrt{5}}{2}, \frac{1+\sqrt{5}}{2})$	non-diagonalizable	2
$(1, 1)$	diagonalizable	3 3 shared
$(1, 1)$	non-diagonalizable	3 shared
$(0, 2)$	diagonalizable	1

bring 8 families of similar matrices. One should check that in the cases when $(\lambda_1, \lambda_2) = (0, 0) + (1, 1) = (1, 1)$ that both types of families (i.e. diagonal + non-diagonal actually occur

- ⑦ (a) If $x \in$ is in Nullspace of A , then
- $Ax=0$ so $M^T x$ when multiplied ~~is~~ on the left by $M^T A M$ ~~is~~ gives
- $$M^T A M (M^T x) = M^T A x = M^T \cdot 0 = 0$$
- $\therefore M^T x$ is in the Nullspace of $M^T A M$.

- (b) Since \forall vector $x \in$ Nullspace of A there exists a vector $M^T x$ in the Nullspace of $M^T A M$ and for every vector x in the Nullspace of $M^T A x$ there exists a vector $M x$ in the Nullspace of A (since $M^T A M x$ must then ~~be~~ equal zero.) Thus the Nullspace of $A + M^T A M$ has the same number of elements \therefore the dimension of the nullspace is the same.

- ⑧ No, the order of association of eigenvectors to eigenvalues could be different among the two matrices could be different.
~~the order~~ ^{associations} It the ~~order~~ ^{associations} is the same then I would think that $A=B$

With n independent eigenvectors again the answer is NO. to the question of $A = B$. The logic from the prior discussion still holds. If A has eigenvalues $0, 0$ w/ a single eigenvector $\propto (1, 0)$ Then

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} = M^{-1} A M \quad \text{or}$$

$$A = M \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} M^{-1} \quad \text{with } M \text{ a matrix with the 1st column of}$$

which ~~is~~ ^{is} the vector $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ + the 2nd column must be ~~linearly independent~~

to the 1st column. This gives many possible A 's. Consider

$$M_1 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \cancel{M_2 = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}} \quad M_2 = \begin{bmatrix} 1 & a \\ 0 & b \end{bmatrix} \quad b \neq 0$$

$$\text{then } M_1^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \cancel{M_2^{-1} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}} = \frac{1}{b} \begin{bmatrix} b & -a \\ 0 & 1 \end{bmatrix}$$

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \cancel{\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}}$$

$$= \cancel{\begin{bmatrix} 1 & 1 & 0 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix}}$$

$$= \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \end{bmatrix}$$

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Then $A_2 = M \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}^M$

$$A_2 = M \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}^{M^2}$$

$$= \begin{bmatrix} 1 & a \\ 0 & b \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} b & -a \\ 0 & 1 \end{bmatrix}$$

$$= \frac{1}{b} \begin{bmatrix} 1 & a \\ 0 & b \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$= \frac{1}{b} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \neq A \quad \text{unless } b = 1.$$

Thus in this case also there is the possibility of ~~more than two~~ different ~~similar~~ matrices with this property

⑨ If $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ then $A^2 = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix}$

$$\therefore A^3 = \begin{bmatrix} 1 & 3 \\ 0 & 1 \end{bmatrix}$$

We guess that $A^k = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$

to check by induction $A^{k+1} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$

$$= \begin{bmatrix} 1 & k+1 \\ 0 & 1 \end{bmatrix} \neq$$

If $k=0$ $A^0 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$ + $k=-1$ gives

$A^{-1} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$. We can check that $A^{-1}A = I$

$$\begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ yes}$$

⑦ If $J = \begin{bmatrix} c & 1 \\ 0 & c \end{bmatrix}$ then

$$J^2 = \begin{bmatrix} c & 1 \\ 0 & c \end{bmatrix} \begin{bmatrix} c & 1 \\ 0 & c \end{bmatrix} = \begin{bmatrix} c^2 & 2c \\ 0 & c^2 \end{bmatrix}$$

$$+ J^3 = \begin{bmatrix} c & 1 \\ 0 & c \end{bmatrix} \begin{bmatrix} c^2 & 2c \\ 0 & c^2 \end{bmatrix} = \begin{bmatrix} c^3 & 3c^2 \\ 0 & c^3 \end{bmatrix}$$

We guess that

$$J^k = \begin{bmatrix} c^k & kc^{k-1} \\ 0 & c^k \end{bmatrix}$$

$$\text{The } J^0 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$+ J^1 = \begin{bmatrix} c^1 & -c^0 \\ 0 & c^1 \end{bmatrix}$$

We can check $J^{-1} = I$ to get

$$\begin{bmatrix} c^{-1} - c^{-2} \\ 0 \end{bmatrix} \begin{bmatrix} c & 1 \\ 0 & c \end{bmatrix} = \begin{bmatrix} 1 & c^{-1} - c^{-2} \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ Yes}$$

The first sheet

(11) $\frac{du}{dt} = Ju = \begin{bmatrix} 5 & 1 & 0 \\ 0 & 5 & 1 \\ 0 & 0 & 5 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ where the unknowns are (x, y, z)

Add a 4th unknown w and the equation $\frac{dw}{dt} = 5w + x$ would

result in the following system to $u = \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix}$

$$\frac{du}{dt} = \begin{bmatrix} 5 & 1 & 0 & 0 \\ 0 & 5 & 1 & 0 \\ 0 & 0 & 5 & 0 \\ 1 & 0 & 0 & 5 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} \quad \text{Following the sum logic as}$$

later we can solve for $z(t)$ then $y(t)$, then $x(t)$ giving exactly the solutions found in the text namely

$$z = z(0) e^{5t}$$

$$y = (y(0) + t z(0)) e^{5t}$$

$$x = (x(0) + t y(0) + \frac{1}{2} t^2 z(0)) e^{5t}$$

When ~~the~~ these function forms are put back into the equation for $w(t)$ the following results

$$\frac{dw}{dt} = (x(0) + ty(0) + \frac{1}{2}t^2 z(0)) e^{st} + S(wt)$$

or

$$\frac{dw}{dt} - S(wt) = (x(0) + ty(0) + \frac{1}{2}t^2 z(0)) e^{st}$$

which has a solution given by

$$w(t) = (w(0) + tx(0) + \frac{1}{2}t^2 y(0) + \frac{1}{6}t^3 z(0)) e^{st}$$

check

$$\begin{aligned} w'(t) &= (x(0) + ty(0) + \frac{1}{2}t^2 z(0)) e^{st} + S(\dots) e^{st} \\ &= (\dots) e^{st} + S(wt) \end{aligned}$$

(12) Let $M = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ m_{41} & m_{42} & m_{43} & m_{44} \end{bmatrix}$

Then $JM = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ m_{41} & m_{42} & m_{43} & m_{44} \end{bmatrix}$

$$= \begin{bmatrix} m_{21} & m_{22} & m_{23} & m_{24} \\ 0 & 0 & 0 & 0 \\ m_{41} & m_{42} & m_{43} & m_{44} \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$+ M_k = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ m_{41} & m_{42} & m_{43} & m_{44} \end{bmatrix} \quad \left[\begin{array}{cccc} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$= \begin{bmatrix} 0 & m_{11} & m_{12} & 0 \\ 0 & m_{22} & m_{23} & 0 \\ 0 & m_{32} & m_{33} & 0 \\ 0 & m_{42} & m_{43} & 0 \end{bmatrix}$$

So if $JM = M_k$

$$\Rightarrow \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ 0 & 0 & 0 & 0 \\ m_{21} & m_{22} & m_{23} & m_{24} \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & m_{11} & m_{12} & 0 \\ 0 & m_{22} & m_{23} & 0 \\ 0 & m_{32} & m_{33} & 0 \\ 0 & m_{42} & m_{43} & 0 \end{bmatrix}$$

$$\text{So } m_{21} = 0$$

$$m_{44} = 0$$

$$m_{22} = m_{11}$$

$$m_{42} = 0$$

$$m_{23} = m_{12}$$

$$m_{43} = 0$$

$$m_{24} = 0$$

Thus ~~we~~ we can deduce that

~~$m_{22} = 0$~~

$$m_{11} = 0$$

~~$m_{23} = 0$~~

$$m_{12} = 0$$

~~$m_{41} = 0$~~

$$m_{33} = 0$$

~~$m_{42} = m_{32}$~~

$$m_{32} = 0$$

~~$m_{43} = m_{33}$~~

giving for M

$$M = \begin{bmatrix} 0 & 0 & m_{13} & m_{14} \\ m_{21} & 0 & 0 & 0 \\ m_{31} & 0 & 0 & m_{34} \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

~~Ex.~~ Since this matrix has a column of all zeros it is not invertible

+ : $M^{-1}JM = J$ is impossible

(B) When A is a zero block J_i it looks like

$$J_i = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}$$

we have $J_i^T = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}$

Dropping the "i" subscript. For this part of the problem we are looking for
that ~~path~~ ~~path~~ $M^{-1}JM = J^T$ equivalently a M

$$M^{-1}JM = J^T \text{ equivalently}$$

~~then~~ $M_{ij} =$

$$M_i = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

~~then~~ $M_i^T = \begin{bmatrix} 1 & -1 & & \\ 0 & 1 & -1 & \\ & & \ddots & \\ 0 & 0 & \dots & 1 \end{bmatrix}$

To try to find such a matrix M that makes $J + JT$ similar, let's consider a ~~very simple~~ very simple case to begin with

$$J = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad J^T = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \quad \text{then}$$

$$\text{If } M = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$JM = MJ^T$ is equivalent to

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} a+1 & b+1 \\ c+1 & d+1 \end{bmatrix} = \begin{bmatrix} a+b & b+d \\ c+d & d+d \end{bmatrix}$$

$$\Rightarrow \begin{array}{l} c=b \\ d=0 \end{array} \quad \text{everything else arbitrary.}$$

pick $a=1, b=1, c=1, d=0$. Then $M = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$

consider $JM = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 1+1 & 1 \\ 1 & 0 \end{bmatrix}$

+ $MJ^T = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1+1 & 1 \\ 1 & 0 \end{bmatrix}$ the sum

For a 3×3 case we would have

$$\mathbf{J} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{so} \quad \mathbf{J}^T = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

Then with $\mathbf{M} = \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix}$ we have

$$\mathbf{JM} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix}$$

$$= \begin{bmatrix} m_{11} + m_{21} & m_{12} + m_{22} & m_{13} + m_{23} \\ m_{21} + m_{31} & m_{22} + m_{32} & m_{23} + m_{33} \\ m_{31} & m_{32} & m_{33} \end{bmatrix}$$

$$+ \mathbf{M} \mathbf{J}^T = \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} m_{11} + m_{12} & m_{12} + m_{13} & m_{13} \\ m_{21} + m_{22} & m_{22} + m_{23} & m_{23} \\ m_{31} + m_{32} & m_{32} + m_{33} & m_{33} \end{bmatrix}$$

$$\text{So } m_{31} = m_{22}$$

$$m_{32} = 0$$

$$m_{22} = m_{13}$$

$$m_{32} = m_{23} \Rightarrow m_{23} = 0$$

$$m_{33} = 0$$

$$m_{23} = 0$$

$$m_{33} = 0$$

with the other variables arbitrary. Let remaining all non-specified variables be 1
Thus

$$M = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

Thus I hypothesize that $M = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 0 \\ 1 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \end{bmatrix}$
for general n

$$M_{ij} = 1 \quad \cancel{\text{if}} \quad i+j \leq n+1 \quad \cancel{n+1 - n+1}$$

$$M_{ij} = 0 \quad i+j > n+1$$

Let's check our requirement that $JM = MJ^T$ in this more general case

the (i,j) th entry of $\mathbf{JM} = M\mathbf{J}^T$ gives in this case

$$\sum_{k=1}^n J_{ik} M_{kj} = \sum_{k=1}^n M_{ik} (\mathbf{J}^T)_{kj} = \sum_{k=1}^n M_{ik} J_{jk} \quad 1 \leq i, j \leq n$$

Since ~~J_{ij}~~ this becomes due to the banded structure of \mathbf{J} the following

$$J_{ii} M_{ij} + J_{i,i+1} M_{i+1,j} = M_{i,j} J_{jj} + M_{i,j+1} J_{j,j+1}$$

Since $J_{i,i} = J_{ii}$ this simplifies to

$$J_{i,i+1} M_{i+1,j} = M_{i,j+1} J_{j,j+1}$$

Since $J_{i,i+1} = 1$ this becomes

$$M_{i+1,j} = M_{i,j+1}$$

which will be true given the banded form on M or described above

$M_{i,j} = M(i,j)$. Thus the matrix M needs \mathbf{J} & \mathbf{J}^T similar

Now if \mathbf{J} is a Jordan matrix it is composed of Jordan Blocks J_i as

$$\mathbf{J} = \begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{bmatrix} \quad \text{using } M \text{ constructed of blocks is clear we have}$$

that $M = \begin{bmatrix} M_1 & & \\ & \ddots & \\ & & M_s \end{bmatrix}$

so $JM = MJ^T$ shows (w/ M a block matrix of M_i 's)

$$\begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{bmatrix} \begin{bmatrix} M_1 & & \\ & \ddots & \\ & & M_s \end{bmatrix} = \begin{bmatrix} M_1 & & \\ & \ddots & \\ & & M_s \end{bmatrix} \begin{bmatrix} J_1^T & & \\ & \ddots & \\ & & J_s^T \end{bmatrix}$$

$$= \begin{bmatrix} J_1 M_1 & & \\ & J_2 M_2 & \\ & & \ddots \\ & & & J_s M_s \end{bmatrix} = \begin{bmatrix} M_1 J_1^T & & \\ & \ddots & \\ & & M_s J_s^T \end{bmatrix}$$

Since we have equality of the sub-blocks involving $J_i + M_i$ we have

~~to~~ find a matrix M (\approx block matrix) that diagonalizes J .
(w/ s @ Jordan Blocks)

Let

Third let A be any matrix. Then by the Jordan form for A we can decompose A into its Jordan blocks as

$$M^T A M = J \quad \text{or} \quad A = M J M^{-1} \quad \text{so} \quad A^T = (M^{-1})^T J^T M^T$$

Further find the M_0 that makes $J^T + J$ similar ~~to~~ from part
the second part, i.e. ~~$J^T = M_0^{-1} J M_0$~~ the

$$\begin{aligned} A^T &= (M^{-1})^T M_0^{-1} J M_0 M^T \\ &= (M^T)^{-1} M_0^{-1} M^{-1} A M M_0 M^T \\ &= (M \cdot M_0 \cdot M^T)^{-1} A (M \cdot M_0 \cdot M^T) \end{aligned}$$

So $A^T + A$ are similar by other relationship above

(14) We can generate an ~~#~~ of matrices similar to a given matrix A by constructing an invertible M such that one computing

$$B = M^{-1}AM.$$

(15)

$$\begin{aligned}\det(A - \lambda I) &= \det(A - \lambda M^{-1}M) \\ &= \det(M^{-1}(M^{-1}AM - \lambda I)M) \\ &= \det(M^{-1}(M^{-1}AM - \lambda I)M^{-1}) \\ &= \det(A - \lambda M \cdot M^{-1}) \\ &= \det(M^{-1}(M^{-1}AM - \lambda I)M^{-1}) \\ &= \det(M) \cdot \det(M^{-1}AM - \lambda I) \cdot \det(M^{-1}) \\ &= \det(M) \det(M^{-1}AM - \lambda I) \det(M)^{-1} \\ &= \det(M^{-1}AM - \lambda I)\end{aligned}$$

Thus A & $M^{-1}AM$ have the same characteristic polynomial
 \therefore the same eigenvalues

(16) $\begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} d & c \\ b & a \end{bmatrix}$ are similar since they have the same eigenvalues.

$$\begin{bmatrix} b & a \\ d & c \end{bmatrix} + \begin{bmatrix} c & d \\ a & b \end{bmatrix}$$
 are similar for the same reason

to show that $\begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} b & a \\ d & c \end{bmatrix}$ are not similar

let $a=1, b=c=d=0$ Then the two matrices are given by

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$
 which have different eigenvalues $\pm i$.

are still not similar

(17) (a) True if it uses ~~is~~ ~~exists~~ both A + B
~~is invertible~~ but you would have to have the sum eigenvalues but the invertible matrix cannot have 0 as an eigenvalue while the singular matrix must have 0 as an eigenvalue

(b) False. From example 3 the projection matrix

$$A = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \text{has eig-decomposition } A = P^{-1} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} P$$

~~but it is~~ \therefore ~~similar to it.~~ \Rightarrow ~~The matrix of~~ is symmetric

but with $M = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$ the matrix is similar to

$$M^T A M = \begin{bmatrix} 1 & .5 \\ 0 & 0 \end{bmatrix} \text{ is not symmetric}$$

(c) If A is similar to $-A$ ~~then~~ the $A + -A$

~~must have the same eigenvalues thus~~

~~the eigenvalues must all be zero. But we will not have~~

$$\text{a counter } A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad \text{and } M = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

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

then

$$AM = M(-A)$$

~~$\begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix}$~~

$$\Rightarrow \begin{bmatrix} c & d \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & -a \\ 0 & -c \end{bmatrix}$$

$$\Rightarrow c=0 \quad + \quad d=-a \quad b \text{ is arbitrary}$$

Thus $A + -A$ is similar w/ a M given

by $M = \begin{bmatrix} a & b \\ 0 & -a \end{bmatrix}$

and $A \neq 0$.

(d) True, the eigenvalues of $A + A + A + I$ will be different
 \therefore they cannot be similar

(18) If B is invertible, AB has the same eigenvalues

As BA . The eigenvalues of AB are given by

$$\det(AB - \lambda I) = 0$$

$$\Rightarrow \det(B^{-1}BAB - \lambda I) = 0$$

$$\Rightarrow \det(B^{-1}(BA - \lambda B \cdot B^{-1})B) = 0$$

$$\Rightarrow \det(B^{-1}) \det(BA - I) \det(B) = 0$$

$\Rightarrow \det(BA - I) = 0 \therefore$ since $AB + BA$ have the same characteristic equation \therefore the same eigenvalues

(19)

$$\begin{bmatrix} I & -A \\ 0 & I \end{bmatrix} \begin{bmatrix} AB & 0 \\ B & 0 \end{bmatrix} \begin{bmatrix} I & A \\ 0 & I \end{bmatrix} = \begin{bmatrix} I & -A \\ 0 & I \end{bmatrix} \begin{bmatrix} AB & ABA \\ B & \cancel{ABA} \end{bmatrix}$$

$(m+n) \times$

$$= \begin{bmatrix} 0 & 0 \\ B & BA \end{bmatrix}$$

(a) A is $m \times n$ & B is $n \times m$, so AB is $m \times m$

& BA is $n \times n$. The two I 's in the 1st matrix are

of size $m \times$

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- (19) (a) ~~\rightarrow~~ If A is $m \times n$ & B is $n \times m$ then
 AB is $m \times m$ & BA is $n \times n$, consistent sizes are given below

$$\begin{bmatrix} I_{m \times m} & -A_{m \times n} \\ 0_{n \times m} & I_{n \times n} \end{bmatrix} \begin{bmatrix} AB_{m \times m} & 0_{m \times n} \\ B_{n \times m} & 0_{n \times n} \end{bmatrix} \begin{bmatrix} I_{m \times m} & A_{m \times n} \\ 0_{n \times m} & I_{n \times n} \end{bmatrix}$$

(b) Since

$$\begin{bmatrix} I_{m \times m} & -A_{m \times n} \\ 0_{n \times m} & I_{n \times n} \end{bmatrix} \begin{bmatrix} I_m & A \\ 0 & I_n \end{bmatrix} = \begin{bmatrix} I_m & 0 \\ 0 & I_n \end{bmatrix}$$

$$\begin{bmatrix} I_m & -A \\ 0 & I_n \end{bmatrix} \begin{bmatrix} I_m & A \\ 0 & I_n \end{bmatrix} = \begin{bmatrix} I_m & 0 \\ 0 & I_n \end{bmatrix}$$

This block equation is $M^{-1}FM = G$

~~the~~ Based on the eigenvalue concept given in the text, when ~~$M \rightarrow A$~~

AB has m eigenvalues plus n zeros

BA has n eigenvalues plus m zeros

Since when $m \geq n$ AB has the same eigenvalues as BA
 but with $m-n$ zero eigenvalues

(20)

(a) If A is similar to B then

$$\cancel{\text{if } A \text{ is similar to } B} \quad A = M^{-1}BM$$

$$\text{Then } A^2 = M^{-1}BM \cdot M^{-1}BM = M^{-1}B^2M$$

so A^2 is similar to B^2

$$(b) \cancel{\text{if } A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}} \quad A^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$+ B = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

~~but~~ ~~not always~~

Let A be ~~diagonalizable~~ with eigenvalues -1 and 2

B be diagonalizable with eigenvalues $+1$

Then A^2 is ~~diagonalizable~~ has eigenvalues $+1$ & B^2 has
 eigenvalues $+1$. They can be made similar while $A + B$
 cannot (since they don't have the same eigenvalues)

$$\text{For example, let } A = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} + B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

then since $A + B$ is diagonal

the eigenvalues are the values found on the diagonal. Also

$$A^2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + B^2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ so } A^2 + B^2 \text{ is}$$

similar with $A + B$ or not.

- (c) $\begin{bmatrix} 3 & 0 \\ 0 & 4 \end{bmatrix} + \begin{bmatrix} 3 & 1 \\ 0 & 4 \end{bmatrix}$ are both diagonalizable with the same eigenvalues, therefore they are similar.

- (d) $\begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix}$ is diagonalizable but $\begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix}$ is not
the simplest matrix it can be reduced to is a Jordan form matrix.

- (e) Considering the ~~the~~ requested transformation

~~we~~ a sign introduced
 $|A - \lambda I| = 0$ is exchanged by the 1st row exchange i.e.

$$- |A' - \lambda I'| = 0$$

The column exchange introduces another sign and ~~another~~ a negative sign and ~~another~~ a permutation of I' . This permutation of I' results in the identity matrix back again so we have $(-1)^2 |A'' - \lambda I| = 0$
 shows that $A + A''$ have the same eigenvalues.

S

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① If $A = U\Sigma V^T$ then $A^T A = V \Sigma^2 V^T$.

For this problem $A^T A$ is given by

~~for~~

$$A^T A = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 & 4 \\ 2 & 3 \end{bmatrix} = \begin{bmatrix} 5 & 20 \\ 20 & 80 \end{bmatrix}$$

which has eigenvalues given by

~~$$\begin{vmatrix} 5-\lambda & 20 \\ 20 & 80-\lambda \end{vmatrix} = 0$$~~

$$\Rightarrow (5-\lambda)(80-\lambda) - 400 = 0$$

$$\Rightarrow 400 - 5\lambda - 80\lambda + \lambda^2 - 400 = 0$$

$$\Rightarrow \lambda^2 - 85\lambda = 0 \Rightarrow \lambda = 0 + \lambda = 85.$$

$A^T A$ has eigenvectors given by (for $\lambda=0$) thus ~~$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$~~

$$v_1 = \begin{bmatrix} -4 \\ 1 \end{bmatrix} = v_1 = \frac{1}{\sqrt{16+1}} \begin{bmatrix} -4 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{17}} \begin{bmatrix} -4 \\ 1 \end{bmatrix}$$

for $\lambda = 85$

$$A^T A - 85I = \begin{bmatrix} -80 & 20 \\ 20 & -5 \end{bmatrix} \Rightarrow v_2 = \begin{bmatrix} 1 \\ 4 \end{bmatrix} \Rightarrow v_2 = \frac{1}{\sqrt{17}} \begin{bmatrix} 1 \\ 4 \end{bmatrix}$$

Then Since the book takes $\begin{bmatrix} 1 \\ 4 \end{bmatrix}$ to be the nonzero eigenvalue of $A^T A$

we have $\begin{bmatrix} 1 \\ 4 \end{bmatrix} = \cancel{\frac{1}{\sqrt{17}}} \begin{bmatrix} 1 \\ 4 \end{bmatrix} +$

$$v_2 = \frac{1}{\sqrt{17}} \begin{bmatrix} -4 \\ 1 \end{bmatrix}$$

(2) (a) computing AA^T gives

$$\begin{bmatrix} 1 & 4 \\ 2 & 8 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 4 & 8 \end{bmatrix} = \begin{bmatrix} 17 & 34 \\ 34 & 68 \end{bmatrix}$$

which has an eigenvalue equal to 0 + a second eigenvalue given by

~~$\lambda_1 = 0 + \sqrt{85}^2 = 17 + 68 = 85$~~ as expected from problem 1.

The eigenvectors of AA^T are then given by

off

$$AA^T - 85I = \begin{bmatrix} -68 & 34 \\ 34 & -17 \end{bmatrix} \Rightarrow v_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$\Rightarrow v_1 = \frac{1}{\sqrt{15}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

For ~~v_2~~ v_2 we have

$$v_2 = \begin{bmatrix} -2 \\ 1 \end{bmatrix} = v_2 = \frac{1}{\sqrt{15}} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

$$(b) Av_1 = \begin{bmatrix} 1 & 4 \\ 2 & 8 \end{bmatrix} \frac{1}{\sqrt{15}} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \frac{1}{\sqrt{15}} \begin{bmatrix} 17 \\ 34 \end{bmatrix} = \sqrt{17} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$\text{w/ } \beta_1^2 = 85 \Rightarrow \beta_1 = \sqrt{85} = \sqrt{17} \cdot \sqrt{5} \text{ so } Av_1 \text{ done}$$

~~for~~

belongs

$$Av_1 = \sqrt{17} \cdot \sqrt{5} \cdot \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \sqrt{85} \cdot v_1, \text{ as requested}$$

All entries of the SVD are given by

$$\begin{bmatrix} 1 & 4 \\ 2 & 8 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \\ \frac{3}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \sqrt{85} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{17}} & -\frac{4}{\sqrt{17}} \\ \frac{4}{\sqrt{17}} & \frac{1}{\sqrt{17}} \end{bmatrix}^T$$

lets check this by computing the product of the matrices on the right hand side. We do this remembering that $\sqrt{85} = \sqrt{5} \cdot \sqrt{17}$

$$\begin{bmatrix} \cancel{\frac{1}{\sqrt{5}}} & 0 & \cancel{0} & \cancel{\sqrt{5}} & = \cancel{1/\sqrt{17}} \\ \cancel{2/\sqrt{5}} & 0 & 0 & 0 \end{bmatrix} =$$

$$\begin{bmatrix} \cancel{\frac{1}{\sqrt{5}}} & \cancel{\frac{3}{\sqrt{5}}} & \cancel{0} & \cancel{-4/\sqrt{5}} \\ \cancel{-2/\sqrt{5}} & \cancel{1/\sqrt{5}} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 4 \\ 2 & 8 \end{bmatrix}$$

$$\begin{bmatrix} \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \\ \frac{3}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \sqrt{85} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{17}} & -\frac{4}{\sqrt{17}} \\ \frac{4}{\sqrt{17}} & \frac{1}{\sqrt{17}} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \\ \frac{3}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \sqrt{5} & 4\sqrt{5} \\ \cancel{4\sqrt{5}} & 0 \end{bmatrix} = \begin{bmatrix} 1 & 4 \\ 2 & 8 \end{bmatrix} \quad \text{yes}$$

- ③ Since this A has $\text{rank} = 1$ we have from the fact that
 The matrices $U + V$ contain orthogonal basis for all far subspaces
- The 1st $r=1$ columns of V span the rowspace of A $\left\{ \begin{matrix} A = U \Sigma V^T \\ \sim \quad \sim \quad \sim \end{matrix} \right.$
 this is $v_1 = \frac{1}{\sqrt{17}} \begin{bmatrix} 1 \\ 4 \end{bmatrix}$
 - The Rest $n-r = 2-1=1$ columns of $U + V$ span the nullspace of A
 this is $v_2 = \frac{1}{\sqrt{17}} \begin{bmatrix} -4 \\ 1 \end{bmatrix}$
 - The 1st $r=1$ columns of U span the column space of A
 this is $u_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$
 - The rest $m-r = 2-1=1$ columns of U span the nullspace of A^T . This is $u_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$

④ (a) $A^T A$ is given by

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

which has ~~two~~ eigen eigenvectors give by

$$\begin{vmatrix} 2-\lambda & 1 \\ 1 & 1-\lambda \end{vmatrix} = 0 \quad \Rightarrow \quad (2-\lambda)(1-\lambda) - 1 = 0$$

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

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

$$\text{So } \lambda = \frac{3 \pm \sqrt{9 - 4(1)(1)}}{2} = \frac{3 \pm \sqrt{5}}{2}, \text{ the eigenvectors of } A^T A$$

are given by the null space of

$$\begin{aligned} A^T A - \lambda I &= \begin{bmatrix} 2 - \left(\frac{3 \pm \sqrt{5}}{2}\right) & 1 \\ 1 & 1 - \left(\frac{3 \pm \sqrt{5}}{2}\right) \end{bmatrix} \\ &= \begin{bmatrix} \frac{1 \mp \sqrt{5}}{2} & 1 \\ 1 & -\frac{2 \mp \sqrt{5}}{2} \end{bmatrix} \end{aligned}$$

~~After~~ One could do row reduction on this matrix to show that the 2nd row is a multiple of the 1st but if I assume I have done everything correctly, I know after is free and obtain for the eigenvectors of $A^T A - \lambda I$ the following

$$\text{if } b_1^2 = \frac{3 - \sqrt{5}}{2} \text{ and } b_2^2 = \frac{3 + \sqrt{5}}{2}$$

$$v_1 \propto \begin{bmatrix} 2 \\ 1+\sqrt{5} \\ -1 \end{bmatrix} + v_2 \propto \begin{bmatrix} 2 \\ 1-\sqrt{5} \\ -1 \end{bmatrix}$$

$$\text{or } v_1 \propto \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix} + v_2 \propto \begin{bmatrix} 2 \\ -(1-\sqrt{5}) \end{bmatrix}$$

$$\text{So that } v_1 = \frac{1}{\sqrt{4+1+2\sqrt{5}+5}} \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix}$$

$$= \frac{1}{\sqrt{10+2\sqrt{5}}} \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix}$$

$$+ v_2 = \frac{1}{\sqrt{4+1-2\sqrt{5}+5}} \begin{bmatrix} 2 \\ -(1-\sqrt{5}) \end{bmatrix} = \frac{1}{\sqrt{10-2\sqrt{5}}} \begin{bmatrix} 2 \\ -(1-\sqrt{5}) \end{bmatrix}$$

For the matrix AAT^T we have the same matrix!! ($A^T = A$ in this case)

Thus $\mathbf{B} v_1 = v_1 + v_2 = v_2$. The singular value decomposition of A is then given by $A = U\Sigma V^T$ or

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} \frac{2}{\sqrt{10+2\sqrt{5}}} & \frac{2}{\sqrt{10-2\sqrt{5}}} \\ \frac{-(1+\sqrt{5})}{\sqrt{10+2\sqrt{5}}} & \frac{-(1-\sqrt{5})}{\sqrt{10-2\sqrt{5}}} \end{bmatrix} \begin{bmatrix} \frac{3-\sqrt{5}}{2} & 0 \\ 0 & \frac{3+\sqrt{5}}{2} \end{bmatrix} \begin{bmatrix} \frac{2}{\sqrt{10+2\sqrt{5}}} & \frac{2}{\sqrt{10-2\sqrt{5}}} \\ \frac{-(1+\sqrt{5})}{\sqrt{10+2\sqrt{5}}} & \frac{-(1-\sqrt{5})}{\sqrt{10-2\sqrt{5}}} \end{bmatrix}^T$$

which will be checked by direct calculation of the products of the matrices on the right hand side.

⑤ Consider

$$AV_1 = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix} \frac{1}{\sqrt{10+2\sqrt{5}}}$$

$$= \begin{bmatrix} 2-1-\sqrt{5} \\ 2 \end{bmatrix} \frac{1}{\sqrt{10+2\sqrt{5}}} = \begin{bmatrix} 1-\sqrt{5} \\ 2 \end{bmatrix} \frac{1}{\sqrt{10+2\sqrt{5}}}$$

$$= +\frac{(1-\sqrt{5})}{2} \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix} \frac{1}{\sqrt{10+2\sqrt{5}}}$$

check $\frac{-(1-\sqrt{5})(1+\sqrt{5})}{2} = -\frac{(1-\sqrt{5})}{2} = 2$ yes ✓

Thus the choice becomes

$$\begin{aligned} A_{V_1} &= \frac{(1-\sqrt{5})}{2\sqrt{2}\sqrt{5+\sqrt{5}}} \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix} \\ &= \cancel{\frac{(1-\sqrt{5})}{2\sqrt{2}\sqrt{5+\sqrt{5}}}} \quad \cancel{\begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix}} \end{aligned}$$

$$= \frac{(1-\sqrt{5})}{2\sqrt{2}\sqrt[4]{5}\sqrt{1+\sqrt{5}}} \begin{bmatrix} 2 \\ -(1+\sqrt{5}) \end{bmatrix}$$

↙

~~check~~

$$\frac{1-\sqrt{5}}{2\sqrt{2}\sqrt[4]{5}\sqrt{(1+\sqrt{5})(1-\sqrt{5})}}$$

... Finsl ... check algebra to this point.

⑦ For $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$ we have

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

Then this is the centered distance matrix which has eigenvalues given by

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

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

$$\Rightarrow (1-\lambda)[\cancel{\lambda^2 - 3\lambda + 2^2} - \cancel{\lambda}] = 0$$

$$\Rightarrow (1-\lambda)(\cancel{\lambda - 3}) = 0$$

~~which gives $\lambda = 1 + \sqrt{13+9-14} = \sqrt{3+5}$~~

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

so $\lambda = 1, \lambda = 0, \lambda = 3$

Since the SVD orders them with positive values first
let $\sigma_1^2 = 3$, $\sigma_2^2 = 1$, Then the corresponding eigenvectors are
given by For $\sigma_1^2 = 3$ we have

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

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

~~Pivot vars x_1, x_2 ,~~
~~Free vars x_3 .~~

\therefore the basis to get

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

pivot vars $x_1 + x_2$, free vars x_3
let $x_3 = 1$ then

$$x_1 = 1 + x_2 \quad \therefore v_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

For $\sigma_2^2 = 1$ we have

$$\begin{bmatrix} 0 & 1 & 0 \\ -1 & 1 & 1 \\ 0 & 1 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

pivot vars are $x_1 + x_2 +$ free vars are x_3

~~∴~~ : $v_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$. For the 7th eigenvalue we have

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

$$\Rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \therefore v_3 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

Normalizing we have

$$v_1 = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, v_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}, v_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

Computing $A^T A$ we have

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

which has eigenvalues given by $b_1^2 = 3 + b_2^2 = 1$

The eigenvectors for $b_2^2 = 3$ are given by the nullspace

$$\begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \text{ so } v_1 \propto \begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow v_1 = \frac{(\pm 1)}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

The eigenvectors to $\sigma_2^2 = 1$ are given by the nullspace of

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \text{ so } v_2 \times \begin{bmatrix} 1 \\ -1 \end{bmatrix} = v_2 = \frac{(\pm 1)}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Thus the SVD of A is given by $A = \cancel{V} \Sigma \cancel{U}^T$

consider \cancel{A}^{SVD} given by

~~$$A = \Sigma V^T$$~~

equivalently

~~$$AV = \cancel{\Sigma} V^T$$~~

~~$$A = \Sigma V^T$$~~

~~$2 \times 2 \cdot 2 \times 3 \cdot 3 \cdot 3$~~

~~$$\begin{bmatrix} Y_{12} & Y_{12} \\ Y_{12} & -Y_{12} \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{6}} & Y_{12} & Y_{13} \\ \frac{1}{\sqrt{6}} & 0 & -Y_{13} \\ \frac{1}{\sqrt{6}} & Y_{12} & Y_{13} \end{bmatrix}^T$$~~

which we will check by computing the product of the matrices on the right hand side. We have

~~$$= \begin{bmatrix} \frac{3}{\sqrt{2}} & Y_{12} & 0 \\ \frac{3}{\sqrt{2}} & -Y_{12} & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{2}{\sqrt{6}} & Y_{16} \\ -Y_{12} & 0 & Y_{12} \\ Y_{13} & -Y_{13} & Y_{13} \end{bmatrix}$$~~

To determine the signs of v_1 & v_2 consider

$$Av_i = b_i v_i \text{ for } i=1 \text{ we have}$$

$$Av_1 = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} \frac{\sqrt{6}}{6} \\ \frac{3\sqrt{6}}{6} \\ \frac{3\sqrt{6}}{6} \end{bmatrix} = \begin{bmatrix} \frac{3\sqrt{6}}{6} \\ \frac{3\sqrt{6}}{6} \end{bmatrix} = \sqrt{3} \begin{bmatrix} \frac{\sqrt{3}\sqrt{6}}{6} \\ \frac{\sqrt{3}\sqrt{6}}{6} \end{bmatrix} = \sqrt{3} \begin{bmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{bmatrix}$$

$$\text{so } v_1 = \begin{bmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{bmatrix}$$

For $i=2$ we have

$$Av_2 = b_2 v_2$$

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

$$\text{so } v_2 = \begin{bmatrix} 0 \\ \frac{-\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{bmatrix}$$

Then the SVD of A is given by

$$A = U \Sigma V^T \quad U \text{ is } 2 \times 2 \\ \Sigma \text{ is } 2 \times 3 \\ + V \text{ is } 3 \times 3$$

This gives

$$A = \begin{bmatrix} Y_R & -Y_R \\ Y_R & Y_R \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} Y_R & -Y_R & Y_B \\ Y_B & 0 & -Y_B \\ Y_R & Y_R & Y_B \end{bmatrix}^T$$

which we can check by computing the product of these matrices

$$= \begin{bmatrix} \frac{\sqrt{3}}{2} & -\frac{1}{2} \\ \frac{\sqrt{3}}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} Y_R & \frac{3}{2}Y_B & Y_B \\ -Y_R & 0 & Y_R \\ Y_B & -Y_B & Y_B \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2} + \frac{1}{2} & 1 & \frac{1}{2} - \frac{1}{2} \\ \frac{1}{2} - \frac{1}{2} & 1 & \frac{1}{2} + \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \checkmark$$

(B) We desire $A\mathbf{v}_1 = \mathbf{v}_1, A\mathbf{v}_2 = \mathbf{v}_2, \dots, A\mathbf{v}_n = \mathbf{v}_n$

so the singular values of A are all 1. Thus $\Sigma = \text{diag}(1, 1, \dots, 1)$

~~and~~ the $n \times n$ identity matrix. With this we have $= \mathbf{I}$

$$A = \mathbf{U}\Sigma\mathbf{V}^T = \mathbf{U}\mathbf{I}\mathbf{V}^T = \mathbf{U}\mathbf{V}^T$$

(9) $A\mathbf{v} = 12\mathbf{u}$ for $\mathbf{V} = \frac{1}{2}([1, 1, 1, 1]) + \mathbf{U} = \frac{1}{3}(2, z_1)$.

Its single value must be 12 so

$$A = \mathbf{U}\Sigma\mathbf{V}^T = \frac{1}{3} \begin{bmatrix} 2 \\ 2 \\ 2 \\ 1 \end{bmatrix} (12) \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$= \frac{12}{3} \begin{bmatrix} 2 \\ 2 \\ 2 \\ 1 \end{bmatrix} [1, 1, 1, 1]$$

$$= 2 \begin{bmatrix} 2 & 2 & 2 & 2 \\ 2 & 2 & 2 & 2 \\ 1 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 4 & 4 & 4 & 4 \\ 4 & 4 & 4 & 4 \\ 2 & 2 & 2 & 2 \end{bmatrix}$$

(10) If A has orthogonal columns with length b_1, \dots, b_n

Then $A\mathbf{e}_i = b_i \mathbf{w}_i$

so the matrix \mathbf{V} is the identity $\mathbf{V} = \mathbf{I}$

$$\Sigma = \text{diag}(b_1, b_2, \dots, b_n)$$

$$\mathbf{U} = [\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n]$$

(11) The SVD expresses A as

$$A = \Sigma V^T$$

The product ΣV^T scales each column of V by the corresponding magnitude σ_i in Σ . By considering a block matrix factorization of ΣV^T into columns of V^T into rows (which are the columns of V) we have

$$A = [\sigma_1 u_1 \ \sigma_2 u_2 \ \cdots \ \sigma_n u_n] \begin{bmatrix} v_1^T \\ v_2^T \\ \vdots \\ v_n^T \end{bmatrix}$$

which when we compute this block matrix multiply gives

$$= \sigma_1 u_1 v_1^T + \sigma_2 u_2 v_2^T + \cdots + \sigma_n u_n v_n^T$$

(12) Since A is symmetric it has the following decomposition

$$A = \Sigma \Lambda \Sigma^T = [u_1 \ u_2] \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} v_1^T \\ v_2^T \end{bmatrix}$$

Its SVD would then be given by

$$\Sigma = [u_1 \ u_2] \quad \Sigma = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \quad V = [v_1 \ -v_2]$$

$$\text{so that } \Sigma V^T = [u_1 \ u_2] \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} v_1^T \\ -v_2^T \end{bmatrix} = [u_1 \ u_2] \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} v_1^T \\ v_2^T \end{bmatrix}$$

- (13) If $A = QR$ then Q is an ~~orthogonal~~ basis for the columns of A
~~that is $R \in \mathbb{R}^{n \times n}$~~

also if the SVD of A requires it to have a decomposition given by $A = U\Sigma V^T$ w/ V an orthogonal basis for the columns of A

~~so $QR = U\Sigma V^T$~~ I would then guess that the matrix Q simply ~~changes the basis~~ changes the basis ~~given by the matrix~~ given by the matrix U in the SVD. so given ~~that~~ in other words given the SVD of ~~R~~ $R = U\Sigma V^T$ A is then given by

$$A = (Q\hat{U}\hat{\Sigma}\hat{V}^T) = (\cancel{Q\hat{U}})\hat{\Sigma}(Q\hat{U})\hat{\Sigma}\hat{V}^T$$

which is the SVD of A : only the matrix ~~Q~~ "U" changes between the SVD of A and that of R .

- (14) To make A singular Add the matrix $-B_2 \cdot I$ to A

or $-B_2 T V^T$ since for invertible matrices $V = T$.

$$\therefore A' = \cancel{A - B_2 I} = A - B_2 T V^T$$

$$= A - B_2 T V^T$$

$$= T \begin{bmatrix} B_1 & \\ B_2 & \end{bmatrix} V^T - T \begin{bmatrix} -B_2 & \\ & -B_2 \end{bmatrix} V^T$$

$$= T \begin{bmatrix} B_1 - B_2 & 0 \\ 0 & 0 \end{bmatrix} V^T$$

Note σ_2 was chosen since it ~~is~~ is the smallest ~~biggest~~ singular value.

(15) (a) If $A = \Phi A$ then singular values get multiplied by ~~$\sqrt{\Phi^T \Phi}$~~ Φ i.e. $\Sigma \Rightarrow \Phi \Sigma$

(b) If $A = \Sigma V^T$ from the SVD

$$A^T = V \Sigma^T \Phi^T = V \Sigma^T$$

If $A = \Sigma V^T$ from the SVD then

$$A^{-1} = (\Phi^T)^{-1} \Sigma^{-1} V^T = V \Sigma^{-1} \Phi^T$$

(16) If A is square & invertible I believe this is correct.

In other cases $I \neq \Phi \Sigma^T \Phi$ which would be required for this to be true. Another way to see this is that the square of the singular values of $A+I$ should be the eigenvalues of $(A+I)^T(A+I) = A^T A + I^T I + A^T I + I^T A$

$$= A^T A + I + A^T + A$$

$$\neq A^T A + I$$

which would give ~~singular values~~ & ~~eigenvalues~~ $\Phi \Sigma^T \Phi$ ~~of~~ equivalent to $\Sigma + I$.

In short, it is the inequality between $(A+I)^T(A+I) + A^T A + I$ that is