

# EMTH211 — Exam Questions

1. Consider the matrix

$$A = \begin{bmatrix} 2 & 1 & 3 \\ 4 & -1 & 3 \\ -2 & 5 & 5 \end{bmatrix}$$

- (a) Compute the  $LU$  decomposition of  $A$ .
- (b) Use the  $LU$  decomposition from above to solve

$$A\mathbf{x} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

- (a) As usual, we use the following sequence of row operations to get  $A$  into upper row-echelon form:

$$\begin{aligned} A &= \begin{bmatrix} 2 & 1 & 3 \\ 4 & -1 & 3 \\ -2 & 5 & 5 \end{bmatrix} \\ &\xrightarrow{R_2 - 2R_1} \begin{bmatrix} 2 & 1 & 3 \\ 0 & -3 & -3 \\ -2 & 5 & 5 \end{bmatrix} \\ &\xrightarrow{R_3 + R_1} \begin{bmatrix} 2 & 1 & 3 \\ 0 & -3 & -3 \\ 0 & 6 & 8 \end{bmatrix} \\ &\xrightarrow{R_3 + 2R_2} \begin{bmatrix} 2 & 1 & 3 \\ 0 & -3 & -3 \\ 0 & 0 & 2 \end{bmatrix} = U \end{aligned}$$

We can write this as

$$E_1 E_2 E_3 A = U,$$

where

$$E_1 = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \quad E_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 1 \end{bmatrix}$$

Thus we have  $A = LU$  where (beware the order!)

$$\begin{aligned} L &= E_3^{-1} E_2^{-1} E_1^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & -2 & 1 \end{bmatrix} \end{aligned}$$

(b) We first solve  $L\mathbf{y} = \mathbf{b}$  by forward substitution, and get

$$\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ 4 \end{bmatrix}.$$

Secondly we solve  $U\mathbf{x} = \mathbf{y}$  by backsubstitution and get

$$\mathbf{x} = \begin{bmatrix} -1.5 \\ -2 \\ 2 \end{bmatrix}.$$

2. The set of all  $2 \times 2$  matrices  $M_{2 \times 2}$  with real entries is vector space with the usual matrix addition and matrix scalar multiplication. Are the following subsets  $W$  and  $U$  subspaces? (Give reasons)

•

$$W = \left\{ A \in M_{2 \times 2} : A = \begin{bmatrix} a & b \\ 0 & c \end{bmatrix} \right\}$$

•

$$U = \left\{ \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in M_{2 \times 2} : a + c = 1 \right\}$$

$W$  is a subspace, since it is closed under addition:

$$\begin{bmatrix} a & b \\ 0 & c \end{bmatrix} + \begin{bmatrix} d & e \\ 0 & f \end{bmatrix} = \begin{bmatrix} a+d & b+e \\ 0 & c+f \end{bmatrix} \in W$$

and under scalar multiplication:

$$k \begin{bmatrix} a & b \\ 0 & c \end{bmatrix} = \begin{bmatrix} ka & kb \\ 0 & kc \end{bmatrix} \in W.$$

$U$  isn't a subspace, since it is neither closed under addition nor scalar multiplication. As a specific example:

$$\begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} \in U,$$

but

$$2 \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 4 & 0 \\ 0 & -2 \end{bmatrix}$$

isn't, since  $4 - 2 \neq 1$ .

3. Consider the following matrix

$$A = \begin{bmatrix} 1 & 3 & 1 & 4 \\ 2 & 3 & 0 & 1 \\ 1 & 0 & 4 & 0 \\ 0 & 0 & 5 & 3 \end{bmatrix}$$

- (a) Determine the row rank (that is the dimension of the row space) of the matrix  $A$
- (b) Find a basis for the null-space

$$\text{null}(A) = \{\mathbf{x} \in V : A\mathbf{x} = \mathbf{0}\}$$

of  $A$ .

- (c) How are the row rank and the nullity (the dimension of the null space) of a matrix related in general?

- (a) To find the row rank, we get  $A$  into row reduced form

$$\begin{bmatrix} 1 & 3 & 1 & 4 \\ 2 & 3 & 0 & 1 \\ 1 & 0 & 4 & 0 \\ 0 & 0 & 5 & 3 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 3 & 1 & 4 \\ 0 & -3 & -2 & -7 \\ 0 & -3 & 3 & -4 \\ 0 & 0 & 5 & 3 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 3 & 1 & 4 \\ 0 & -3 & -2 & -7 \\ 0 & 0 & 5 & 3 \\ 0 & 0 & 5 & 3 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 3 & 1 & 4 \\ 0 & -3 & -2 & -7 \\ 0 & 0 & 5 & 3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

The first three rows of this reduced matrix form a basis. Thus the (row rank here is 3)

- (b) To find a basis for the null space we would like to find a general solution to  $A\mathbf{x} = \mathbf{0}$ , so we would also try and find the row reduced form of  $A$ , but we have already done this above. So we can solve by back-substitution:  $x_4 = t$ , since  $x_4$  is a free variable. Then  $5x_3 = -3t$  or  $x_3 = -\frac{3}{5}t$ .  $-3x_2 - 2x_3 - 7x_4 = 0$ , so  $-3x_2 = 2x_3 + 7x_4 = -\frac{6}{5}t + 7t = \frac{29}{5}t$  and therefore  $x_2 = -\frac{29}{15}t$ . Finally  $x_1 = -3x_2 - x_3 - 4x_4 = \frac{29}{5}t + \frac{3}{5}t - 4t = \frac{12}{5}t$  So the general solution can be written as

$$\mathbf{x} = \begin{bmatrix} \frac{12}{5} \\ -\frac{29}{15} \\ -\frac{3}{5} \\ 1 \end{bmatrix} t$$

That vector, or any multiple (such as 5 here) form a basis.

- (c) If  $A$  is a  $n \times m$  matrix then  $(\text{row})\text{rank}(A) + \text{nullity}(A) = n$



#### 4a) Eigenvector and eigenvalue

$$A\underline{x} = \lambda \underline{x} \quad \underline{x} \neq 0$$

if  $A$   $n \times n$ .

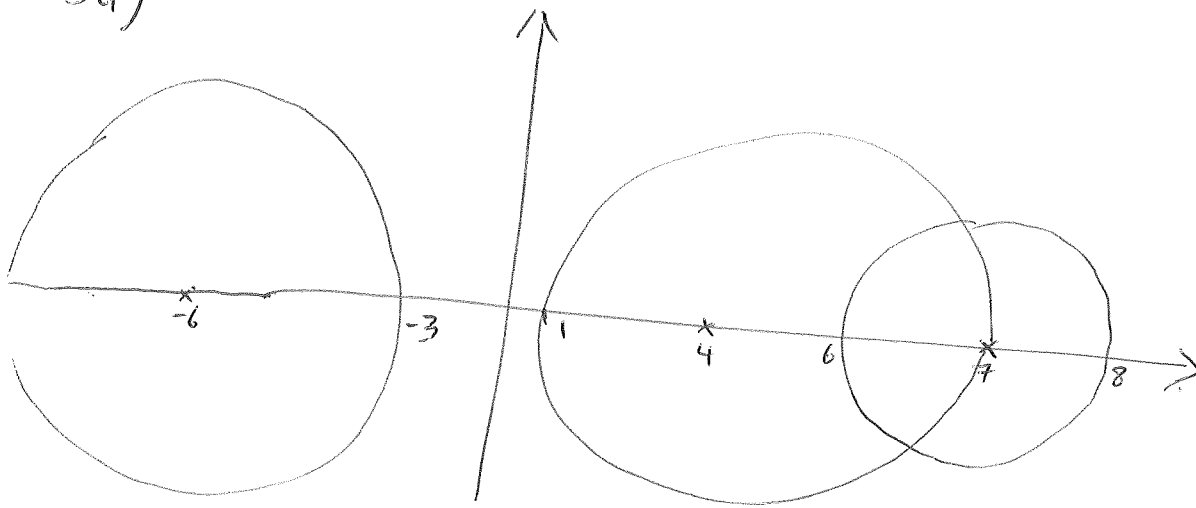
Those directions which remain unchanged under a 'transformation' by  $A$ . &c. ...

b) ~~AVA~~  $\underline{x} = \underline{x}_1 + 3\underline{x}_2$

$$\begin{aligned} A^4 \underline{x} &= A^4 (\underline{x}_1 + 3\underline{x}_2) \\ &= A^4 \underline{x}_1 + 3A^4 \underline{x}_2 \\ &= \lambda_1^4 \underline{x}_1 + 3\lambda_2^4 \underline{x}_2 \\ &= 1^4 \underline{x}_1 + 3 \cdot 4^4 \underline{x}_2 \\ &= \underline{x}_1 + 768 \underline{x}_2 \\ &= \begin{bmatrix} 768 \\ 1 \end{bmatrix} \end{aligned}$$

c) Yes it is possible. The remaining eigenspace could be one-dimensional. The union of the bases of eigenspaces <sup>must</sup> have 7 vectors in order for the matrix to be diagonalisable.

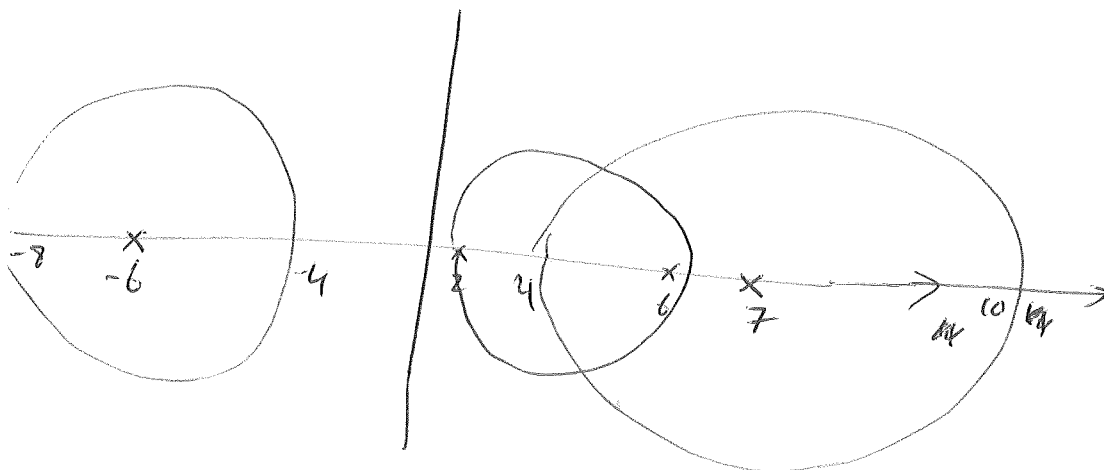
5a)

Rows

$$r=3$$

$$r=3$$

$$r=1$$

Columns

$$r=2$$

$$r=2$$

$$r=3$$

Two circles overlapping  
 One circle disjoint - if single disk is disjoint from the other disks, then this must contain exactly one eigenvalue and as  $A$  is real, this  $e$ -value must be real.

other eigenvalues will be located in region

$$(D_2 \cup D_3) \cap (C_2 \cup C_3)$$

b)

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

$$\underline{x}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

(1)

$$\lambda_1 = 2$$

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

$$(A - 2I)\underline{x}_0 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

(2)

$$\underline{x}_1 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

scale  $\underline{y}_1 = \begin{bmatrix} -1/2 \\ 1 \end{bmatrix}$

$$(A - 2I)\underline{y}_1 = \begin{bmatrix} 1.5 \\ -3 \end{bmatrix} \quad \text{scale: } \begin{bmatrix} -1/2 \\ 1 \end{bmatrix} \quad (-3)$$

Produced eigenvector -3

$$\lambda_2 - \lambda_1 = -3$$

$$\lambda_2 - 2 = -3 \Rightarrow \lambda_2 = -1$$

ii) Solve  $A \underline{x}_1 = y_0$  where  $\underline{x}_0 = y_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

Row reduce to determine  $\underline{x}_1$

$$\text{So } y_1 = \frac{\underline{x}_1}{\|\underline{x}_1\|}$$

$$\underline{x}_2 \text{ from } A \underline{x}_2 = y_1$$

Unit Smallest eigenvalue of  $A$  is the reciprocal of the "scalar"

"If  $A$  is invertible with eigenvalue  $\lambda$ , then  $A^{-1}$  has eigenvalue  $1/\lambda$ ."

→ Apply power method to  $A^{-1}$ , its dominant e-value will be the reciprocal of the smallest (in magnitude) e-value of  $A$ .

&c....



- a) Gram-Schmidt process converts a basis  
 for a subspace  $U$  into an orthogonal basis  
 for  $U$ .  
 It does this by "subtracting" projections  
 set  $\underline{v}_1 = \underline{u}_1$   
 $\underline{v}_2 = \underline{u}_2 - (\text{its proj on } \underline{u}_1)$   
 $\underline{v}_3 = \underline{u}_3 - (\text{the proj's of } \underline{u}_3 \text{ on } \underline{v}_1 \text{ \& } \underline{v}_2)$

$$\begin{aligned}
 \text{b) } R = Q^T A &= \begin{bmatrix} & & & \\ & & & \\ & & & \\ & & & \end{bmatrix} \begin{bmatrix} 5 & -9 \\ 1 & -3 \\ -3 & 7 \\ 1 & -3 \end{bmatrix} \\
 &= \begin{bmatrix} 6 & -12 \\ 0 & 2 \end{bmatrix}
 \end{aligned}$$

c) check  $QR = A$  ✓.

d)

$$QR_{\underline{x}} = \begin{bmatrix} 11 \\ 1 \\ -5 \\ 1 \end{bmatrix}$$

$$R_{\underline{x}} = Q^T \begin{bmatrix} 11 \\ 1 \\ -5 \\ 1 \end{bmatrix}$$

$$\text{or } \begin{bmatrix} 6 & -12 \\ 0 & 2 \end{bmatrix} \underline{x} = \begin{bmatrix} 12 \\ 2 \end{bmatrix}$$

$$\cancel{x_2 = 2}$$

$$x_1 = 4$$

$$\underline{x} = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$$

check!

$$\begin{bmatrix} 5 & -9 \\ 1 & -3 \\ -3 & 7 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 1 \end{bmatrix} = \begin{bmatrix} 11 \\ 1 \\ -5 \\ 1 \end{bmatrix} \checkmark$$

d) as only RHS has changed, expect new system  $A\underline{x} = \underline{b}$  to be inconsistent.

QR still compute  $\underline{x}$ , but it is the least squares solution to

$$A\underline{x} = \underline{b}, \text{ i.e., the value of } \underline{x}$$

which makes  $\|A\underline{x} - \underline{b}\|_2$

as small as possible.

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Given

$$\underline{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \text{ and } \underline{x}_2 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad (\text{from } E_1)$$

$$\underline{x}_3 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \quad (\text{from } E_2)$$

Want orthogonal basis vectors  $\underline{x}_1, \underline{x}_2, \& \underline{x}_3$

$$\underline{x}_1 = \underline{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

$$\underline{x}_2 = \underline{x}_2 - \left( \frac{\underline{x}_1 \cdot \underline{x}_2}{\underline{x}_1 \cdot \underline{x}_1} \right) \underline{x}_1$$

$$= \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \left( \frac{2}{2} \right) \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\underline{x}_3 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

Normalise

$$\underline{q}_1 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix} \quad \underline{q}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad \underline{q}_3 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}$$

$$\begin{aligned} A &= \lambda_1 \underline{q}_1 \underline{q}_1^T + \lambda_2 \underline{q}_2 \underline{q}_2^T + \lambda_3 \underline{q}_3 \underline{q}_3^T \\ &= \begin{bmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 1/2 & 0 \\ 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} + (-2) \begin{bmatrix} 1/2 & -1/2 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 0 \end{bmatrix} \\ &= \begin{bmatrix} -1/2 & 3/2 & 0 \\ 1/2 & -1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

b)  $B = \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix}$

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

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

eigenvalues are  $(1-\lambda)^2 - 1 = 0$

(P)

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

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

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

eigenvalues  $\lambda = 0, 2$

~~eigenvectors are~~

(V)

eigenvectors are:  $\begin{pmatrix} 2 \end{pmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  &  $\begin{pmatrix} 0 \end{pmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

normalise :

$$\begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \quad \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

(U)

d.  $u_1 = \frac{1}{\sigma_1} = \frac{1}{\sqrt{2}} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$   
 $= \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

where  $\sigma_1 = \sqrt{2}$  &  $\sigma_2 = 0$

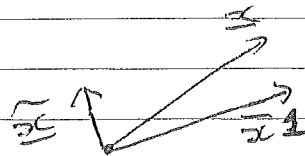
$\sigma_2 = 0$  :  $u_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  (by inspection)

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



1(a) i) To center a variable  $\underline{x}$ , the mean  $\bar{x}$  is subtracted from each element of  $\underline{x}$ . The centered vector  $\tilde{\underline{x}}$  will have a mean of zero. Centering decomposes  $\underline{x}$  into two orthogonal components.

$$\Rightarrow \underline{x} = \tilde{\underline{x}} + \bar{x} \underline{1}$$



$$\begin{aligned} \text{ii)} \quad (\underline{x} - \bar{x} \underline{1}) \cdot \underline{1} &= \sum_{i=1}^n (x_i - \bar{x}) \\ &= \sum_{i=1}^n x_i - n\bar{x} \\ &= n\bar{x} - n\bar{x} \\ &= 0. \end{aligned}$$

$$\text{iii)} \quad s_x = \frac{\|\tilde{\underline{x}}\|}{\sqrt{n-1}} = \frac{\|\underline{x} - \bar{x} \underline{1}\|}{\sqrt{n-1}}$$

$$\begin{aligned} s_{\tilde{x}} &= \frac{\|\tilde{\underline{x}} - \bar{\tilde{x}} \underline{1}\|}{\sqrt{n-1}} \\ &= \frac{\|\underline{x} - \bar{x} \underline{1} - 0\|}{\sqrt{n-1}} \end{aligned}$$

$$= s_x$$

$$(b) \quad s_x = \frac{\|\tilde{x}\|}{\sqrt{n_x-1}} = \frac{25}{\sqrt{36}} = 4.17$$

$$s_y = \frac{\|\tilde{y}\|}{\sqrt{n_y-1}} = \frac{36}{\sqrt{81}} = 4$$

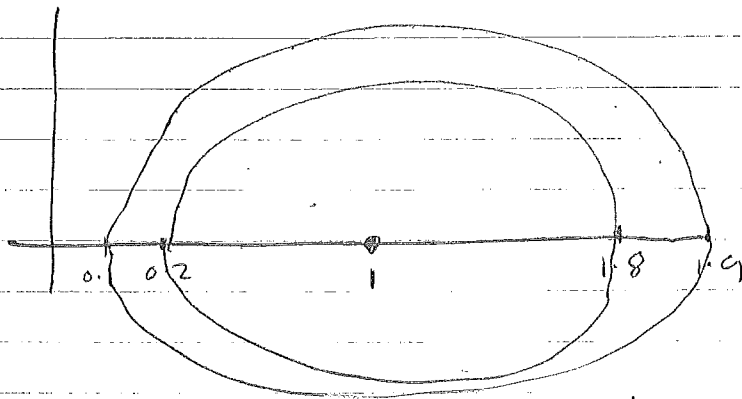
Hence,  $x$  has greater spread.

(c) Let

$$C = \begin{pmatrix} r_{xx} & r_{xy} & r_{xz} \\ r_{xy} & r_{yy} & r_{yz} \\ r_{xz} & r_{yz} & r_{zz} \end{pmatrix} = \begin{pmatrix} 1 & 0.7 & 0.2 \\ 0.7 & 1 & -0.1 \\ 0.2 & -0.1 & 1 \end{pmatrix}$$

If  $C$  is a correlation matrix, the eigenvalues of  $C$  must be positive.

Gerschgorin's Disks:



Hence, the three eigenvalues are positive and so  $C$  is a correlation matrix and the given correlations are valid.



$$(d) i) s_y = \frac{\|y\|}{\sqrt{n-1}} = 3$$

$$\Rightarrow \sqrt{n-1} = \frac{15}{3} = 5$$

$$n = 26$$

$$ii) s_x = \frac{\|\tilde{x}\|}{5} = \frac{20}{5} = 4$$

$$s_z = \frac{\|\tilde{z}\|}{5} = \frac{5}{5} = 1$$

$$iii) r_{xy} = \cos \angle (\tilde{x}, \tilde{y}) = \cos(30) \\ = 0.866$$

$$r_{xz} = \cos \angle (\tilde{x}, \tilde{z}) = \cos(110) \\ = -0.342$$

$$r_{yz} = \cos \angle (\tilde{y}, \tilde{z}) = \cos(80) \\ = 0.1736$$

$$2 (a) \quad \tilde{x} \cdot \hat{e} = 0$$

$$\tilde{x} \cdot (\hat{y} - \hat{b}_1 \tilde{x}) = 0$$

$$\tilde{x} \cdot \hat{y} - \hat{b}_1 \tilde{x} \cdot \tilde{x} = 0$$

$$\hat{b}_1 = \frac{\tilde{x} \cdot \hat{y}}{\|\tilde{x}\|^2}$$

$$(b) \quad r_{xy} = \cos \angle (\tilde{x}, \hat{y})$$

$$= \frac{\tilde{x} \cdot \hat{y}}{\|\tilde{x}\| \|\hat{y}\|}$$

Now.

$$r_{xy} \frac{\|\hat{y}\|}{\|\tilde{x}\|} = \frac{\tilde{x} \cdot \hat{y}}{\|\tilde{x}\| \|\hat{y}\|} \cdot \frac{\|\hat{y}\|}{\|\tilde{x}\|}$$

$$= \frac{\tilde{x} \cdot \hat{y}}{\|\tilde{x}\|^2}$$

$$= \hat{b}_1$$

c) ① Independent

② Common Variance

③ Normally distributed with mean 0.

$$d) \quad F = \frac{(n-2)R^2}{1-R^2} = \frac{30(0.64)}{0.36}$$

$$\approx 53.3$$

The 95% percentile of  $F(1, 30)$  is

$$F_{0.95}(1, 30) = 4.17.$$

Since  $F = 53.3 > F_{0.95}(1, 30) = 4.17$ ,

we conclude at level 0.05 that the SLR model is a good fit.

(e) The type 1 error in this case is concluding that the model is a good fit, when it is actually a bad fit (false positive).

(f) The type 2 error in this case is concluding that the model is not a good fit, when it is actually a good fit (false negative).

(g) If  $\underline{x}' = -2\underline{x}$ , then

$$\begin{aligned}\tilde{\underline{x}}' &= -2\underline{x} + 2\bar{x}\underline{1} \\ &= -2\tilde{\underline{x}}\end{aligned}$$

Hence,

$$r_{\underline{x}'y} = \frac{-2\tilde{\underline{x}} \cdot \tilde{\underline{y}}}{\| -2\tilde{\underline{x}} \| \| \tilde{\underline{y}} \|}$$

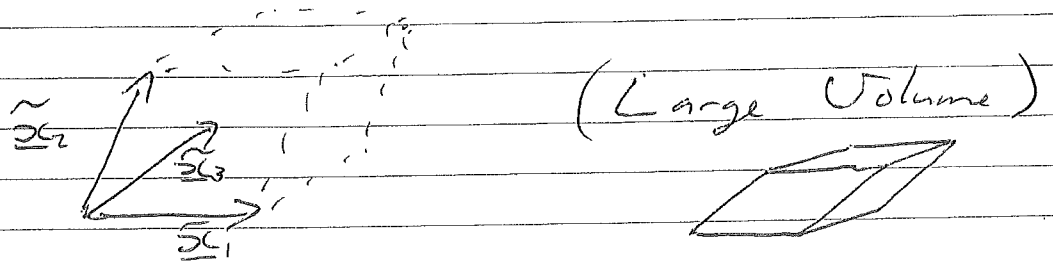
$$= \frac{-2\tilde{\underline{x}} \cdot \tilde{\underline{y}}}{2\|\tilde{\underline{x}}\| \|\tilde{\underline{y}}\|}$$

$$= -r_{\underline{x}y}$$

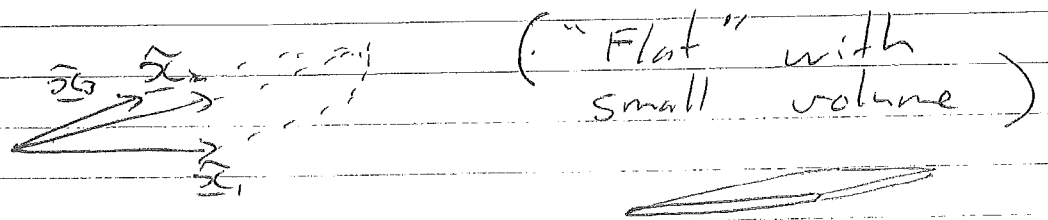
$$= -10.$$

3 (a) When the three predictor vectors almost fall into a subspace of dimension less than three.

(b) If the standardized vectors are well-conditioned, the volume of the parallelepiped formed by the three vectors will be close to one.



If the standardized vectors are ill-conditioned, the volume of the parallelepiped formed by the three vectors will be close to zero.



(c) The normalized volume is given by

$$\begin{aligned} A_{\text{norm}} &= \sqrt{\det(c)} \\ &= \sqrt{0.1 \times 1 \times 3} \\ &= 0.548 \end{aligned}$$

$$\begin{aligned}
 (d) \quad H &= \left(1 + \frac{2p+5}{6} - n\right) \ln(1 - \det(c)) \\
 &= \left(1 + \frac{11}{6} - 34\right) \ln(1 - 0.3) \\
 &= 11.1164
 \end{aligned}$$

The 95% percentile of  $\chi^2(3)$  is

$$\chi^2_{0.95}(3) = 7.81$$

Since  $H = 11.1164 > \chi^2_{0.95}(3) = 7.81$ ,

we conclude at level 0.05 that the MLR model is stable.

(e) The 95% percentile of  $F(3, 30)$  is

$$F_{0.95}(3, 30) = 2.92.$$

Since  $F = 3 > F_{0.95}(3, 30) = 2.92$ ,

we conclude at level 0.05 that the MLR model is a good fit.

(f) The decision in part (e) would not change because

$$F_{0.90}(3, 30) < F_{0.95}(3, 30).$$

