

# Notes in ECEN 5448

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## 1 Some matrix Algebra

$A_{n \times m}$   $\text{Range}(A) = \text{Im}(A) := \{y : y = Ax \text{ for some } x \in \mathbb{R}^m\}$

$$A = \begin{bmatrix} | & | & & | \\ a_1 & a_2 & \dots & a_m \\ | & | & & | \end{bmatrix}$$

$a_i \in \mathbb{R}^n$  and  $Ax = x_1 a_1 + x_2 a_2 + x_3 a_3 + \dots + x_m a_m$ .

$$A = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 2 \\ 2 & 2 & 4 \\ 3 & 3 & 6 \end{pmatrix}$$

$$A_{\text{reduced}} = \begin{pmatrix} 1 & 0 \\ 1 & 1 \\ 2 & 2 \\ 3 & 3 \end{pmatrix}$$

$$\text{Im}(A) = \text{Im}(A_{\text{reduced}})$$

Reminder about a subspace,  $V$  is a subspace of  $\mathbb{R}^n$  if  $\forall v_1, v_2, \dots \in V$  and any  $\alpha \in \mathbb{R}, v_1 + \alpha v_2 \in V$ .

For a given subspace  $V$  find a matrix such that the image of  $A$  is  $V$ .

For arbitrary subspace  $V \subseteq \mathbb{R}^n$  we know there exists a basis  $\{v_1, \dots, v_k\} \subseteq \mathbb{R}^n$  for  $V$ . If we let  $A = [v_1 v_2 \dots v_k]$ , then  $\text{Im}(A) = V$ .

$\text{rank}(A) :=$  minimum number of columns of  $A$  that span  $\text{Im}(A) = \dim(\text{Im}(A))$ .

Basis: We say that  $\{b_1, \dots, b_k\}$  is a basis for a subspace  $S$  if the bs are linearly independent and  $\forall s \in S, \exists c \in \mathbb{R}^d, s = Bc$  where  $B$  is the matrix made up of the  $b$  vectors (linear combinations of  $b$  form every point in  $S$ ).

Fact: Suppose that columns of  $B_{n \times n}$  are a basis for  $S$ .  $\forall$  invertible matrices  $T_{n \times m}$ ; the columns of  $\tilde{B} := BT$ , are a basis for  $S$ .

Proof: columns of  $\tilde{B}$  are independent iff  $\tilde{B}x = 0$  implies that  $x = 0$ .

if:  $0 = \tilde{B}x = B(Tx) \implies Tx = 0 \implies x = 0$  because  $T$  is invertible.

Second part: any vector in  $S$  can be constructed with  $\tilde{B}$  just like  $B$ .

Let  $s \in S, \exists c \in \mathbb{R}^m; s = Bc$ . If we let  $\tilde{c} = T^{-1}c$ . Then,  $\tilde{B}\tilde{c} = B\tilde{c} = BT\tilde{c} = BTT^{-1}c = Bc = s$ .

The columns of  $B$  are an orthonormal basis for  $S$  if in addition to being a basis for  $S$ ,  $B^T B = I$ . Note that  $B^T B = I$  iff  $b_i^T b_j = 0, i \neq j$  and  $b_i^T b_j = 1, i = j$ .

Important fact: For a symmetric (or Hermitian) matrix  $A$ , there always exists  $M$  and diagonal matrix  $\Gamma$ ;

$$A = M\Gamma M^T, MM^T = I$$

If  $Av = \lambda_1 v$ ,  $Aw = \lambda w$ , then  $w^T Av = \lambda_1 w^T v = \lambda_2 w^T v$  because column eigenvectors are also row eigenvectors for symmetric matrices.

FACT: For symmetric positive definite (p.d.) matrix  $P$ ,  $\exists$  a symmetric matrix  $R$ ;  $P = R^2$ .

Proof: for a p.d.  $P$ ,  $P = M\Gamma M^T = M\sqrt{\Gamma}M^T M\sqrt{\Gamma}M^T = R * R$ .

$R = M\sqrt{\Gamma}M^T$ .

For a basis matrix  $B$ ,  $B^T B$  is p.d.  $\implies$

$$B^T B = R * R$$

for some invertible  $R$ . Now define  $\tilde{B} = BR^{-1}$ . Then  $\tilde{B}$  is an orthonormal basis for  $S$ .

$$\tilde{B}^T \tilde{B} = R^{-T} B^T B R^{-1} = R^{-T} R R R^{-1} = I$$

so there is always an orthonormal basis for any basis.

Let  $S$  be a subspace of  $\mathbb{R}^n$ . Then we define the orthogonal compliment of  $S$  by:

$$S^\perp := \{x : x^T s = 0 \forall s \in S\}$$

Suppose the columns of  $B$  form an orthonormal basis for  $S$ .  $\forall x \in \mathbb{R}^n$ ,

$$x = x_s + x_{s^\perp}$$

where  $x_s \in S$  and  $x_{s^\perp} \in S^\perp$ .

Proof: Let  $x_s = BB^T x$  and  $x_{s^\perp} = (I - B^T B)x$ .

can show that this projects  $x$  onto  $S$  and  $S^\perp$ .

$x_s$  has to be in  $S$  because it is constructed by  $B$  times something which is spanned by the image of  $B$ .

Then,  $x_s \in S$ . show that  $x_{s^\perp} \in S^\perp$ , we need to show that  $\forall c (Bc \perp x_s)$ .

$$c^T B^T (I - BB^T)x = c^T B^T x - c^T B^T x = 0$$

Corollary 1:  $S^\perp = IM(I - B^T B)$

Corollary 2:  $n = \dim(S) + \dim(S^\perp)$ .

## 2 BACK TO CONTROLLABILITY

$\dot{x} = Ax + Bu$  is controllable if  $\leftrightarrow$  of  $e^{-A\tau}$  are independent.

$\leftrightarrow \Omega(0, T) = \int_0^T e^{-A\tau} B B^T e^{-A^T \tau} d\tau$  is non-singular

$\leftrightarrow \text{rank}(BAB A^2 B \dots A^{n-1} B) = n$ .

Proof: Use Cayley Hamilton,

$$\begin{aligned} e^{-A\tau} &= \alpha_0(\tau)I + \alpha_1(\tau)A + \dots + \alpha_{n-1}(\tau)A^{n-1} \\ \implies e^{-A\tau} &= \alpha_0(\tau)B + \alpha_1(\tau)AB + \dots + \alpha_{n-1}(\tau)A^{n-1}B \\ &\implies \exists x, x(BAB \dots A^{n-1}B) = 0, \\ \implies x^T e^{-A\tau} B &= \alpha_0(\tau)x^T B + \alpha_1(\tau)x^T AB + \dots + \alpha_{n-1}(\tau)x^T A^{n-1}B = 0 \end{aligned}$$

Question: Show that  $\text{rank}(A) < n$ ,  $\exists x \neq 0, x^T A = 0$ .

Controllability  $\implies \text{rank}(BAB \dots A^{n-1}B) = n$ .

Suppose that,  $\dot{x} = Ax + Bu$  is not controllable:

$$\begin{aligned} &\implies \exists y; y^T e^{-A\tau} B = 0 \implies y^T B = 0 \\ &\implies \frac{d}{d\tau} (y^T e^{-A\tau} B = -y^T e^{-A\tau} AB) \Big|_{\tau=0} = -y^T AB \end{aligned}$$

$$\implies \frac{d^{n-1}}{d\tau^{n-1}} y^T e^{-A\tau} \implies (\pm) y^T A^{n-1} B = 0$$