

### Basic Equations

#### Network Flows

1. the flow in an arc is only in one directions
2. flow into a node = flow out of a node
3. flow into the network = flow out of the network

#### Balancing Chemical Equations

1. add x's before each combo and both side
2. carbo =  $x_1 + 2(x_3)$ , set as system, solve

#### Matrix

augmented matrix	variables and solution(rhs)
coefficient matrix	coefficients only, no rhs

### Vectors, Norm, Dot Product

magnitude (norm) of vector  $v$  is  $\|v\|$ ;  $\|v\| \geq 0$

if  $k > 0$ ,  $kv$  same direction as  $v$       magnitude =  $k\|v\|$

if  $k < 0$ ,  $kv$  opposite direction to  $v$       magnitude =  $|k|\|v\|$

vectors in  $R^n$  ( $n =$  dimension)       $v = (v_1, v_2, \dots, v_n)$

$v = P_1P_2 = OP_2 - OP_1$       displacement vector

norm/magnitude of vector  $\|v\|$        $\sqrt{(v_1)^2 + (v_2)^2 + \dots}$

$\|v\| = 0$  iff  $v = 0$        $\|kv\| = |k|\|v\|$

unit vector  $u$  in same direct as  $v$        $u = (1/\|v\|)v$

$e_1 = (1, 0, \dots)$  ...  $e_n = (0, \dots, 1)$  in  $R^n$       standard unit vector

$d(u, v) = \sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2 + \dots + (u_n - v_n)^2} = \|u - v\|$

$d(u, v) = 0$  iff  $u = v$

### Vectors, Norm, Dot Product (cont)

$u \cdot v = u_1v_1 + u_2v_2 + \dots + u_nv_n$       dot product

$\|u\| \|v\| \cos(\theta)$

$u$  and  $v$  are orthogonal if  $u \cdot v = 0$  ( $\cos(\theta) = 0$ )

a set of vectors is an orthogonal set iff  $v_i \cdot v_j = 0$ , if  $i \neq j$

a set of vectors is an orthonormal set iff  $v_i \cdot v_j = 0$ , if  $i \neq j$ , and  $\|v_i\| = 1$  for all  $i$

$(u \cdot v)^2 \leq \|u\|^2 \|v\|^2$       Cauchy-Schwarz Inequality

or  $|u \cdot v| \leq \|u\| \|v\|$

$d(uv, w) \leq d(u, w) + d(v, w)$       Triangle Inequality

$\|u + v\| \leq \|u\| + \|v\|$

$\|v_1 + v_2 + \dots + v_k\| \leq \|v_1\| + \|v_2\| + \dots + \|v_k\|$

### Lines and Planes

a vector equation with parameter  $t$        $x = x_0 + tv$ ,  $-\infty < t < +\infty$

solution set for 3 dimension linear equation is a plane

if  $x$  is a point on this plane       $n \cdot (x - x_0) = 0$   
(point-normal equation)

$A(x - x_0) + B(y - y_0) + C(z - z_0) = 0$        $x_0 = (x_0, y_0, z_0)$ ,  $n = (A, B, C)$

general/algebraic equation       $Ax + By + Cz = D$

two planes are parallel if  $n_1 = kn_2$ , orthogonal if  $n_1 \cdot n_2 = 0$

### Matrix Algebra, Identity and Inverse Matrix

$(A + B)_{ij} = (A)_{ij} + (B)_{ij}$        $(A - B)_{ij} = (A)_{ij} - (B)_{ij}$

$(cA)_{ij} = c(A)_{ij}$        $(A^T)_{ij} = (A)_{ji}$

$(AB)_{ij} = a_i b_{1j} + a_i b_{2j} + \dots + a_i b_{kj}$

Inner Product (number) is  $u^T v = u \cdot v$ ,  $u$  and  $v$  same size

Outer Product (matrix) is  $uv^T$ ,  $u$  and  $v$  can be any size

$(A^T)^T = A$        $(kA)^T = k(A)^T$

$(A+B)^T = A^T + B^T$        $(AB)^T = B^T A^T$

$\text{tr}(A^T) = \text{tr}(A)$        $\text{tr}(AB) = \text{tr}(BA)$

$u^T v = \text{tr}(uv^T)$        $\text{tr}(uv^T) = \text{tr}(vu^T)$

$\text{tr}(A) = a_{11} + a_{22} + \dots + a_{nn}$        $(A^T)_{ij} = A_{ji}$

Identity matrix is square matrix with 1 along diagonals

If  $A$  is  $m \times n$ ,  $A^T$  is  $n \times m$  and  $mA = A$

a square matrix is invertible(nonsingular)       $AB = BA$

if:

$B$  is the inverse of  $A$        $B = A^{-1}$

if  $A$  has no inverse,  $A$  is not invertible (singular)

$\det(A) = ad - bc \neq 0$  is invertible

if  $A$  is invertible:       $(AB)^{-1} = B^{-1}A^{-1}$

$(A^n)^{-1} = A^{-n} = (A^{-1})^n$        $(A^T)^{-1} = (A^{-1})^T$

$(kA)^{-1} = 1/k(A^{-1})$ ,  $k \neq 0$

### Elementary Matrix and Unifying Theorem

elementary matrices are invertible

$A^{-1} = E_k E_{k-1} \dots E_2 E_1$

$[A | I] \rightarrow [I | A^{-1}]$

(how to find inverse of  $A$ )

$Ax = b$ ;  $x = A^{-1}b$



By fionaw

[cheatography.com/fionaw/](https://cheatography.com/fionaw/)

Published 16th July, 2020.

Last updated 10th August, 2020.

Page 1 of 4.

Sponsored by **Readable.com**

Measure your website readability!

<https://readable.com>

### Elementary Matrix and Unifying Theorem (cont)

- A → RREF =
  - A can be express as a product of E
  - A is invertible
  - $Ax = 0$  has only the trivial solution
  - $Ax = b$  is consistent for every vector b in  $\mathbb{R}^n$
  - $Ax = b$  has exactly 1 solution for every b in  $\mathbb{R}^n$
  - column and rowvectors of A are linealy independent
  - $\det(A) \neq 0$
  - $\lambda = 0$  is not an eigenvalue of A
  - TA is one to one and onto
- If not, then all no.

### Consistency

$$EAx = Eb \rightarrow Rx = b', \text{ where } b' = Eb$$

$(Ax=b) [A | b] \rightarrow [EA | Eb]$  ( $Rx = b'$ )  
(but treat b as unknown:  $b_1, b_2, \dots$ )

**For it to be consistent, if R has zero rows at the bottom, b' that row must equal to zero**

### Homogeneous Systems

Linear Combination of the vectors:

$$v = c_1v_1 + c_2v_2 \dots + c_nv_n$$

(use matrix to find c)

$$Ax = 0 \quad \text{Homogeneous}$$

$$Ax = b \quad \text{Non-homogeneous}$$

$$x = x_0 + t_1v_1 + t_2v_2 \dots + t_kv_k \quad \text{Homogeneous}$$

$$x = t_1v_1 + t_2v_2 \dots + t_kv_k \quad \text{Non-homogeneous}$$

xp is any solution of NH system  
and xh is a solution of H system

$$x = xp + xh$$

### Examples of Subspaces

IF:  $w_1, w_2$  are within S then  $w_1+w_2$  are within S and  $kw_1$  is within S

- the zero vector 0 it self is a subspace

-  $\mathbb{R}^n$  is a subspace of all vectors

- Lines and planes through the origin are subspaces

- The set of all vectors b such that  $Ax = b$  is consistent, is a subspace

- If  $\{v_1, v_2, \dots, v_k\}$  is any set of vectors in  $\mathbb{R}^n$ , then the set W of all linear combinations of these vector is a subspace

$$W = \{c_1v_1 + c_2v_2 + \dots + c_kv_k\}; c \text{ are within real numbers}$$

### Span

- the span of a set of vectors  $\{v_1, v_2, \dots, v_k\}$  is the set of all linear combinations of these vectors

$$\text{span} \{v_1, v_2, \dots, v_k\} = \{t_1v_1, t_2v_2, \dots, t_kv_k\}$$

If  $S = \{v_1, v_2, \dots, v_k\}$ , then  $W = \text{span}(S)$  is a subspace

$Ax = b$  is consistent if and only if b is a linear combination of  $\text{col}(A)$

### Linear Independent

- if unique solution for a set of vectors, then it is linearly independent

$$c_1v_1 + c_2v_2 \dots + c_nv_n = 0; \text{ all the } c = 0$$

- for dependent, not all the  $c = 0$

Dependent if:

- a linear combination of the other vectors
- a scalar multiple of the other
- a set of more than n vectors in  $\mathbb{R}^n$

Independent if:

- the span of these two vectors form a plane

### Linear Independent (cont)

- list the vectors as the columns of a matrix, row reduce it, if many solution, then it is dependent

- after RREF, the columns with leading 1's are a maxmially linearly independent subset according to Pivot Theorem

### Diagonal, Triangular, Symmetric Matrices

Diagonal Matrices	all zeros along the diagonal
Lower Triangular	zeros above diagonal
Upper Triangular	zeros below the diagonal
Symmetric if:	$A^T = A$
Skew-Symmetric if:	$A^T = -A$

### Determinants

$\det(A) = a_{1j}C_{1j} + a_{2j}C_{2j} \dots + a_{nj}C_{nj}$  expansion along jth column

$\det(A) = a_{i1}C_{i1} + a_{i2}C_{i2} \dots + a_{in}C_{in}$  expansion along the ith row

$$C_{ij} = (-1)^{i+j} M_{ij}$$

$M_{ij}$  = deleted ith row and jth column matrix

- pick the one with most zeros to calculate easier

$$\det(A^T) = \det(A) \quad \det(A^{-1}) = 1/\det(A)$$

$$\det(AB) = \det(A)\det(B) \quad \det(kA) = k^n \det(A)$$

- A is invertible iff  $\det(A)$  not equal 0

- det of triangular or diagonal matrix is the product of the diagonal entries

$$\det(A) \text{ for } 2 \times 2 \text{ matrix} \quad \mathbf{ad - bc}$$



### Adjoint and Cramer's Rule

$$\text{adj}(A) = C^T \quad C^T = \text{matrix cofactor of } A$$

$$A^{-1} = (1/\det(A)) \text{adj}(A) \quad \text{adj}(A)A = \det(A) I$$

$$x_1 = \det(A_1) / \det(A) \quad x_2 = \det(A_2) / \det(A)$$

$$x_n = \det(A_n) / \det(A) \quad \det(A) \text{ not equal } 0$$

An is the matrix when the nth column is replaced by b

### Hyperplane, Area/Volume

$$\text{a hyperplane in } \mathbb{R}^n \quad \mathbf{a}_1x_1 + \mathbf{a}_2x_2 + \dots + \mathbf{a}_nx_n = \mathbf{b}$$

- can also written as  $ax = b$

to find  $a^{\text{perp}}$   $ax = 0$ , find the span

if A is 2x2 matrix:

-  $|\det(A)|$  is the **area** of parallelogram

if A is 3x3 matrix:

-  $|\det(A)|$  is the **volume** of parallelepiped

- subtract points to get three vectors, then make it to a matrix to find the area/volume

### Cross Product

$$\mathbf{u} \times \mathbf{v} = (\mathbf{u}_2\mathbf{v}_3 - \mathbf{u}_3\mathbf{v}_2, \mathbf{u}_3\mathbf{v}_1 - \mathbf{u}_1\mathbf{v}_3, \mathbf{u}_1\mathbf{v}_2 - \mathbf{u}_2\mathbf{v}_1)$$

$$\mathbf{u} \times \mathbf{v} = -\mathbf{v} \times \mathbf{u} \quad k(\mathbf{u} \times \mathbf{v}) = (k\mathbf{u}) \times \mathbf{v} = \mathbf{u} \times (k\mathbf{v})$$

$$\mathbf{u} \times \mathbf{u} = \mathbf{0} \quad \text{parallel vectors has } 0 \text{ for c.p.}$$

$$\mathbf{u} \cdot (\mathbf{u} \times \mathbf{v}) = 0 \quad \mathbf{v} \cdot (\mathbf{u} \times \mathbf{v}) = 0$$

$\mathbf{u} \times \mathbf{v}$  is perpendicular to span  $\{\mathbf{u}, \mathbf{v}\}$

$\|\mathbf{u} \times \mathbf{v}\| = \|\mathbf{u}\| \|\mathbf{v}\| \sin(\theta)$ , where  $\theta$  is the angle between vectors

### Complex Number

complex number  $\mathbf{a} + \mathbf{ib}$

$$(\mathbf{a} + \mathbf{ib}) + (\mathbf{c} + \mathbf{id}) = (\mathbf{a} + \mathbf{c}) + \mathbf{i}(\mathbf{b} + \mathbf{d})$$

$$(\mathbf{a} + \mathbf{ib}) - (\mathbf{c} + \mathbf{id}) = (\mathbf{a} - \mathbf{c}) + \mathbf{i}(\mathbf{b} - \mathbf{d})$$

$$(\mathbf{a} + \mathbf{ib})(\mathbf{c} + \mathbf{id}) = (\mathbf{ac} + \mathbf{bd}) + \mathbf{i}(\mathbf{ad} + \mathbf{bc})$$

$$(\mathbf{a} + \mathbf{ib})(\mathbf{c} + \mathbf{id}) = (\mathbf{ac} + \mathbf{bd}x^2) + \mathbf{i}(\mathbf{ad} + \mathbf{bc})$$

$$\mathbf{i}^2 = -1$$

$$\mathbf{z} = \mathbf{a} + \mathbf{ib} \quad \mathbf{z} \text{ bar} = \mathbf{a} - \mathbf{ib}$$

$$\begin{aligned} \text{the length(magnitude) of vector } \mathbf{z} & \quad |\mathbf{z}| = \sqrt{\mathbf{z} \times \mathbf{z} \text{ bar}} \\ & = \sqrt{\mathbf{a}^2 + \mathbf{b}^2} \end{aligned}$$

$$\mathbf{z}^{-1} = 1/\mathbf{z} = \mathbf{z} \text{ bar} / |\mathbf{z}|^2$$

$$\mathbf{z}_1 / \mathbf{z}_2 = \mathbf{z}_1 \mathbf{z}_2^{-1}$$

$$\mathbf{z} = |\mathbf{z}| (\cos(\theta) + \mathbf{i}(\sin(\theta))) \quad \text{polar form (r = } |\mathbf{z}|)$$

$$\mathbf{z}_1 \mathbf{z}_2 = |\mathbf{z}_1| |\mathbf{z}_2| (\cos(\theta_1 + \theta_2) + \mathbf{i}(\sin(\theta_1 + \theta_2)))$$

$$\mathbf{z}_1 / \mathbf{z}_2 = |\mathbf{z}_1| / |\mathbf{z}_2| (\cos(\theta_1 - \theta_2) + \mathbf{i}(\sin(\theta_1 - \theta_2)))$$

$$\mathbf{z}^n = r^n (\cos(n\theta) + \mathbf{i} \sin(n\theta)) \quad r = |\mathbf{z}|$$

$$e^{i\theta} = \cos(\theta) + \mathbf{i} \sin(\theta)$$

$$e^{i\pi} = -1 \quad e^{i\pi} + 1 = 0$$

$$\mathbf{z}_1 \mathbf{z}_2 = r_1 r_2 e^{i(\theta_1 + \theta_2)} \quad \mathbf{z}^n = r^n e^{in\theta}$$

$$\mathbf{z}_1 / \mathbf{z}_2 = r_1 / r_2 e^{i(\theta_1 - \theta_2)}$$

### Eigenvalues and Eigenvectors

$$A\mathbf{x} = \lambda\mathbf{x}$$

$$\det(\lambda I - A) = (-1)^n \det(A - \lambda I)$$

$$p_A(\lambda) = 3x3: \det(A - \lambda I); 2x2: \det(\lambda I - A)$$

- solve for  $(\lambda I - A)\mathbf{x} = 0$  for eigenvectors

#### Work Flow:

- form matrix
- compute  $p_A(\lambda) = \det(\lambda I - A)$
- find roots of  $p_A(\lambda)$  -> eigenvalues of A
- plug in roots then solve for the equation

### Linear Transformation

$\mathbf{f}: \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  $n = \text{domain}$ ,  $m = \text{co-domain}$

$$\mathbf{f}(x_1, x_2, \dots, x_n) = (y_1, \dots, y_m)$$

$T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation if

$$1. T(c\mathbf{u}) = cT(\mathbf{u})$$

$$2. T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$$

for any linear transformation,  $T(\mathbf{0}) = \mathbf{0}$

$$\mathbf{R}\theta = [T(\mathbf{e}_1) \ T(\mathbf{e}_2)] = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \quad \text{matrix for rotation}$$

$\cos\theta$

reflection across y-axis:  $T(x, y) = (-x, y)$

reflection across x-axis:  $T(x, y) = (x, -y)$

reflection across diagonal  $y = x$ ,  $T(x, y) = (y, x)$

orthogonal projection onto the x-axis:  $T(x, y) = (x, 0)$

orthogonal projection onto the y-axis:  $T(x, y) = (0, y)$

$\mathbf{u} = (1/\|\mathbf{v}\|)\mathbf{v}$ ; express it vertically as  $\mathbf{u}_1$  and  $\mathbf{u}_2$

$$\mathbf{A} = \begin{bmatrix} (\mathbf{u}_1)^2 & \mathbf{u}_1\mathbf{u}_2 \\ \mathbf{u}_1\mathbf{u}_2 & (\mathbf{u}_2)^2 \end{bmatrix} \quad \text{projection matrix}$$

contraction with  $0 \leq k < 1$  (shrink),  $k > 1$  (stretch)

$$[\mathbf{x}, \mathbf{y}] \rightarrow [k\mathbf{x}, k\mathbf{y}]$$

compression in x-direction  $[\mathbf{x}, \mathbf{y}] \rightarrow [k\mathbf{x}, \mathbf{y}]$

compression in y-direction  $[\mathbf{x}, \mathbf{y}] \rightarrow [\mathbf{x}, k\mathbf{y}]$

shear in x-direction  $T(x, y) = (x + ky, y)$ ;

$$[\mathbf{x} + k\mathbf{y} \ (1, k), \mathbf{y} \ (0, 1)]$$

shear in y-direction  $T(x, y) = (x, y + kx)$ ;

$$[\mathbf{x} \ (1, 0), \mathbf{y} \ (k, 1)]$$

orthogonal projection on the xy-plane:  $[\mathbf{x}, \mathbf{y}, \mathbf{0}]$

orthogonal projection on the xz-plane:  $[\mathbf{x}, \mathbf{0}, \mathbf{z}]$

orthogonal projection on the yz-plane:  $[\mathbf{0}, \mathbf{y}, \mathbf{z}]$

reflection about the xy-plane:  $[\mathbf{x}, \mathbf{y}, -\mathbf{z}]$

reflection about the xz-plane:  $[\mathbf{x}, -\mathbf{y}, \mathbf{z}]$

reflection about the yz-plane:  $[-\mathbf{x}, \mathbf{y}, \mathbf{z}]$

### Orthogonal Transformation

an orthogonal transformation is a linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  that preserves lengths;  $\|T(\mathbf{u})\| = \|\mathbf{u}\|$

$\|T(\mathbf{u})\| = \|\mathbf{u}\| \iff T(\mathbf{x}) \cdot T(\mathbf{y}) = \mathbf{x} \cdot \mathbf{y}$  for all  $\mathbf{x}, \mathbf{y}$  in  $\mathbb{R}^n$

orthogonal matrix is square matrix  $A$  such that  $A^T = A^{-1}$

1. if  $A$  is orthogonal, then so is  $A^T$  and  $A^{-1}$
2. a product of orthogonal matrices is orthogonal
3. if  $A$  is orthogonal, then  $\det(A) = 1$  or  $-1$
4. if  $A$  is orthogonal, then rows and columns of  $A$  are each orthonormal sets of vectors

### Kernel, Range, Composition

$\ker(T)$  is the set of all vectors  $\mathbf{x}$  such that  $T(\mathbf{x}) = \mathbf{0}$ , RREF matrix, find the vector,  **$\ker(T) = \text{span}\{\mathbf{v}\}$**

the solution space of  $A\mathbf{x} = \mathbf{0}$  is the null space;

**$\text{null}(A) = \ker(A)$**

range of  $T$ ,  $\text{ran}(T)$  is the set of vectors  $\mathbf{y}$  such that  $\mathbf{y} = T(\mathbf{x})$  for some  $\mathbf{x}$

$\text{ran}(T) = \text{col}(T) = \text{span}\{[\text{col}1], [\text{col}2] \dots\}$ ;  $A\mathbf{x} = \mathbf{b}$

Important Facts:

1.  $T$  is one to one iff  $\ker(T) = \{\mathbf{0}\}$
2.  $A\mathbf{x} = \mathbf{b}$ , if consistent, has a unique solution

iff  $\text{null}(A) = \{\mathbf{0}\}$ ;  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution iff  $\text{null}(A) = \{\mathbf{0}\}$

Important facts 2:

1.  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is onto iff the system  $T\mathbf{x} = \mathbf{y}$  has a solution  $\mathbf{x}$  in  $\mathbb{R}^n$  for every  $\mathbf{y}$  in  $\mathbb{R}^m$
2.  $A\mathbf{x} = \mathbf{b}$  is consistent for every  $\mathbf{b}$  in  $\mathbb{R}^m$  ( $A$  is onto) iff  $\text{col}(A) = \mathbb{R}^m$

The composition of  $T_2$  with  $T_1$  is:  $T_2 \circ T_1$

$(T_2 \circ T_1)(\mathbf{x}) = T_2(T_1(\mathbf{x}))$ ;  $T_2 \circ T_1: \mathbb{R}^n \rightarrow \mathbb{R}^m$

composition of linear transformations corresponds to matrix application:  **$[T_2 \circ T_1] = [T_1][T_2]$**

### Kernel, Range, Composition (cont)

$[T(\theta_1 + \theta_2)] = [T\theta_2] \circ [T\theta_1]$ ;

rotate then shear  $\neq$  shear then rotate

linear trans  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  has an inverse iff  $T$  is one to one,  $T^{-1}: \mathbb{R}^m \rightarrow \mathbb{R}^n$ ,  **$T\mathbf{x} = \mathbf{y} \iff \mathbf{x} = T^{-1}\mathbf{y}$**

for  $\mathbb{R}^n$  to  $\mathbb{R}^n$ ,  $[T^{-1}] = [T]^{-1}$ ;  $[T]^{-1} \circ T = 1_n \iff$

$[T^{-1}][T] = 1_n$

$1_n$  is identity transformation;  $1_n$  is identity matrix

### Basis, Dimension, Rank

$S$  is a basis for the subspace  $V$  of  $\mathbb{R}^n$  if:

$S$  is linearly independent and  $\text{span}(S) = V$

$\dim(V) = k$ ,  $k$  is the # of vectors

$\text{row}(A)$  = rows with leading ones after RREF

$\text{col}(A)$  = columns with leading ones from original  $A$

$\text{null}(A)$  = free variable's vectors

$\text{null}(A^T)$  = after transform, the free variable vector

The Rank Theorem:  $\text{rank}(A) = \text{rank}(A^T)$  for any matrix have the same dimension

$\text{rank}(A)$  = # of free vectors in span

$\dim(\text{row}(A)) = \dim(\text{col}(A)) = \text{rank}(A)$

$\dim(\text{null}(A)) = \text{nullity}(A)$

### Orthogonal Complement, Dimension Theorem

$S^\perp = \{\mathbf{v} \in \mathbb{R}^n \mid \mathbf{v} \cdot \mathbf{w} = 0 \text{ for all } \mathbf{w} \in S\}$

$S^\perp$  is a subspace of  $\mathbb{R}^n$ ;  $S^\perp = \text{span}(S)^\perp = W^\perp$

$\text{row}(A)^\perp = \text{null}(A)$        $\text{null}(A)^\perp = \text{row}(A)$   
 $((S^\perp)^\perp = S$  iff  $S$  is subspace

$\text{col}(A)^\perp = \text{null}(A^T)$        $\text{null}(A^T)^\perp = \text{col}(A)$

The Dimension Theorem       $\text{rank}(A) + \text{nullity}(A) = n$

$A$  is  $m \times n$  matrix       $(k + (n-k) = n)$

if  $W$  is a subspace of  $\mathbb{R}^n$        $\dim(W) + \dim(W^\perp) = n$

