

Orthogonal = perpendicular

$$x^T y = 0 \quad (\vec{x} \cdot \vec{y} = 0)$$

$$\begin{aligned} \|x\|^2 + \|y\|^2 &= \|x+y\|^2 \\ x^T x + y^T y &= (x+y)^T (x+y) \\ x^T x + y^T y &= x^T x + y^T y + x^T y + y^T x \\ x^T y + y^T x &= 0 \quad (x^T y = y^T x) \\ x^T y + y^T x &= 0 \quad (x^T y = 0) \end{aligned}$$

$$\boxed{x^T y = 0}$$

Two subspaces are orthogonal if every vector in one is ortho. to every vector in the other.

Row space is ortho to nullspace.

$$\begin{aligned} A_1^T x &= 0 \\ A_2^T x &= 0 \\ A_3^T x &= 0 \end{aligned}$$

$$(1, 0, 0, \dots, 0)^T x = 0 \quad \text{(C-1)}$$

Column space is ortho. to left nullspace.

$A^T A$ is square, symmetric

Claim: $A^T A x = A^T b$ has the best soln.

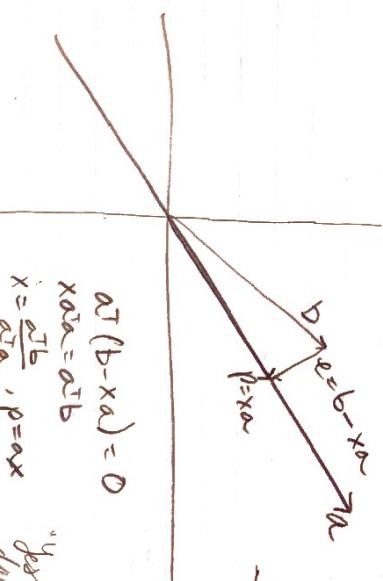
$$\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} x = b$$

rank $A^T A$ = rank A

$A^T A$ is invertible iff

A has ind. cols.

$$A^T A = \begin{bmatrix} 3 & 8 \\ 8 & 30 \end{bmatrix}$$



$$\begin{aligned} \text{Projection matrix } P &= \frac{aa^T}{a^T a} \\ \text{rank } P &= 1 \\ P^T &= P, \quad P^2 = P \end{aligned}$$

$$\begin{aligned} \text{Instead of } Ax = b \text{ solve } A^T x = p \\ P \text{ is proj of } L \text{ onto } C(A) \end{aligned}$$

$$\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$$

$$\begin{aligned} a_1^T (b - A\hat{x}) &= 0 \\ a_2^T (b - A\hat{x}) &= 0 \end{aligned}$$

$$\begin{aligned} \begin{bmatrix} a_1^T \\ a_2^T \end{bmatrix} &= A^T \\ A^T (b - A\hat{x}) &= 0 \\ A^T A \hat{x} &= A^T b \\ A^T A \hat{x} &= A^T b \\ \hat{x} &= (A^T A)^{-1} A^T b \\ P &= A\hat{x} = A(A^T A)^{-1} A^T b \\ P &= A(A^T A)^{-1} A^T b \end{aligned}$$

$P = \hat{x} a_1 + \hat{x}_2 a_2 + \dots$
 $P = A\hat{x}$
 $P = A(A^T A)^{-1} A^T b$
 $P = A(A^T A)^{-1} A^T b$
 A has ind. cols.
 $b - A\hat{x}$ is perp. to plane Note that if A is square then $P = I$.

Notes 14.15

2017-11-29

Lee 16

$$\begin{aligned} I &= P^T P \\ I &= P P^T \\ P^T P &= I \\ P P^T &= I \end{aligned}$$

$$\begin{aligned} (14.1) \quad x_1 - x_2 &= 1 \\ x_2 - x_3 &= 1 \\ -x_1 + x_3 &= -1 \\ 0 &= 1 \end{aligned}$$

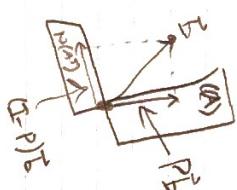
$$\begin{aligned} (14.2) \quad \left[\begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array} \right] &= \left[\begin{array}{c} 1 \\ 1 \\ -1 \end{array} \right] \\ A = \left[\begin{array}{cc} a & b \\ c & d \end{array} \right] & \quad c = xa, \quad d = xb \\ b = ya & \end{aligned}$$

$$\left[\begin{array}{c} 1 \\ -1 \end{array} \right]$$

$$\begin{aligned} (15.1) \quad & \text{Best line: } \\ & \text{basis for } C(A) : \left[\begin{array}{c} x_1 \\ x_2 \end{array} \right] \approx \left[\begin{array}{c} 1 \\ 2 \end{array} \right] \\ & \text{rank } r(A) : \left[\begin{array}{c} x_1 \\ x_2 \end{array} \right] \approx \left[\begin{array}{c} 1 \\ 2 \end{array} \right] \\ & \text{rank } r(A^*) : \left[\begin{array}{c} x_1 \\ x_2 \end{array} \right] \approx \left[\begin{array}{c} 1 \\ 2 \end{array} \right] \end{aligned}$$

Nullspace.

$$\begin{aligned} (15.2) \quad & = (I - P)^2 \\ & = (I - P)(I - P) \\ & = (I - P)I - (I - P)P \\ & = I^2 - PI - IP + P^2 \\ & = I - 2P + P \\ & = I - P, \quad Q.E.D. \end{aligned}$$



$$C + D = I$$

$$C + 2D = 2$$

$$C + 3D = 2$$

$$\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \\ 3 \end{bmatrix}$$

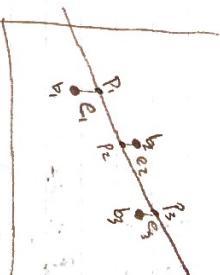
$$4 \quad x = \vec{b}$$

$$A^T A \hat{x} = A^T \vec{b}$$

$$\begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 3 & 6 \\ 6 & 14 \end{bmatrix}$$

$$\text{Minimize } \|e\|^2 = \|A\hat{x} - \vec{b}\|^2$$

$$= (C+D-1) + (C+2D-2)^2 + (C+3D-3)^2$$



Suppose $A^T A \hat{x} = 0$

$$x^T A^T A \hat{x} = 0$$

$$(A\hat{x})^T A \hat{x} = 0$$

$$A\hat{x} = 0$$

Q.E.D.

$$\begin{aligned} \text{Best line: } & y = \frac{2}{3}x + \frac{1}{2}t \\ \hat{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad & \vec{b} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}, \quad P = \begin{bmatrix} P_{x1} \\ P_{x2} \end{bmatrix}, \quad e = \begin{bmatrix} e_1 \\ e_2 \end{bmatrix} \\ & \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} (14.1) \quad & \left[\begin{array}{ccc} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{array} \right] \left[\begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array} \right] \approx \left[\begin{array}{c} 1 \\ 2 \\ 3 \end{array} \right] \\ & \left[\begin{array}{ccc} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{array} \right] \left[\begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array} \right] = \left[\begin{array}{c} 1 \\ 0 \\ 0 \end{array} \right] \end{aligned}$$

$$A^T A \hat{x} = A^T \vec{b}$$

$$\begin{aligned} I \hat{x} &= \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \\ \hat{x} &= \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \end{aligned}$$

Orthogonal
projection.

$$P = A(A^T A)^{-1} A^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I_3$$

proj square.

Pset 16
 (1, 7), (1, 7), (2, 21)

$$\begin{aligned} C - D &= 7 \\ C + D &= 7 \\ C + 2D &= 21 \end{aligned}$$

$$\begin{aligned} A^T A \vec{x} &= A^T b \\ \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x \\ b \end{bmatrix} &= \begin{bmatrix} 7 \\ 7 \\ 21 \end{bmatrix} \\ \begin{bmatrix} 3 & 2 \\ 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} \vec{x} \\ b \end{bmatrix} &= \begin{bmatrix} 3 \\ 2 \\ 6 \end{bmatrix} \begin{bmatrix} 38 \\ 42 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} \begin{bmatrix} 3 & 2 \\ 2 & 1 \\ 1 & 2 \end{bmatrix} \vec{x} &= \begin{bmatrix} 38 \\ 42 \end{bmatrix} \\ \begin{bmatrix} 1 & -1 \\ -1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 9 \\ 4 \\ 1 \end{bmatrix} &= \begin{bmatrix} 5 \\ 13 \\ 17 \end{bmatrix} = \vec{p} \end{aligned}$$

$$\begin{aligned} \vec{x} &= (A^T A)^{-1} A^T b \\ A \vec{x} &= A (A^T A)^{-1} A^T b \\ &= P_b \end{aligned}$$

$$| D = 9 + 4\sqrt{14} |$$

$$\begin{aligned} (6.2) \quad & \begin{bmatrix} 1 & -1 \\ -1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 9 \\ 4 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 13 \\ 17 \end{bmatrix} = \vec{p} \\ \begin{bmatrix} 7 \\ 1 \\ 2 \end{bmatrix} - \begin{bmatrix} 5 \\ 3 \\ 1 \end{bmatrix} &= \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix} \\ \begin{bmatrix} 2 \\ -2 \\ 1 \end{bmatrix} &= \begin{bmatrix} -6 \\ 4 \\ 1 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} P &= \begin{bmatrix} 1 & -1 & 1 \\ 1 & 1 & 2 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 9 & 4 & 1 \\ 4 & 1 & 2 \\ 1 & 2 & 1 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} 13 & 17 & 1 \\ 17 & 21 & 1 \\ 1 & 2 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 5 & -2 \\ 2 & 7 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 5 & 3 & 6 \\ 3 & 6 & -2 \\ 6 & -2 & 10 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} 5 & 3 & 6 \\ 3 & 6 & -2 \\ 6 & -2 & 10 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \end{aligned}$$

$$(6.3) \quad \begin{bmatrix} 5 & 3 & 6 \\ 3 & 6 & -2 \\ 6 & -2 & 10 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

2017-11-29

2017-11-29

2017-11-30

Lee 17 cont'd

$$A = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} - \frac{3}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$$

"Orthogonal matrix": matrix Q with OR. columns
 "Orthogonal matrix": Q^T . matrix, square
 ~~$Q^T = Q^{-1}$~~ if Q is real, normal
 $Q = \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & -1 \end{bmatrix}$ made-up word

$$Q = \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & -1 \end{bmatrix}$$

"Orthogonal matrix": \perp made of ± 1

All Orthogonal vectors are integr.

Q has OR. cols & rows

$$P = Q(Q^T Q)^{-1} Q^T = QQ^T \quad \{=I \text{ if } Q \text{ is square}\}$$

$$\begin{aligned} A^T A \mathbf{x} &= A^T b \\ Q^T Q \mathbf{x} &= Q^T b \\ \mathbf{x} &= Q^T b \end{aligned}$$

don't need inversion!

Gram-Schmidt Orthogonalization

Indep. vectors $a, b \rightarrow$ orthogonal $\vec{A}, \vec{B} \rightarrow$ orthonormal \vec{g}_1, \vec{g}_2

$$\vec{c} \xrightarrow{\vec{a} = A}$$

$$\vec{B} = \vec{b} - \frac{\vec{A}^T \vec{b}}{\vec{A}^T \vec{A}} \vec{A} = \vec{b} - \text{proj}_{\vec{A}}$$

$$\vec{C} = \vec{c} - \frac{\vec{A}^T \vec{c}}{\vec{A}^T \vec{A}} \vec{A} - \frac{\vec{B}^T \vec{c}}{\vec{B}^T \vec{B}} \vec{B}$$

2017-11-30

2017-11-30

Lee 19

Determinants
written $|A|$ or $\det A$

Only for square matrices

① $\det I = 1$

② when you exchange rows, you get reverse sign. $|0 1| = -1$

③ $|ta tb| = t |ab|$

④ $|a^t b^t| = |a b| + |a^t b^t|$

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

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

Part 17

$$\begin{aligned} Qx &= 0 \\ Q^T Qx &= Q^T 0 \\ x &= Q^{-1} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \end{aligned}$$

O.E.D.

$$\begin{aligned} (17.2) \quad & \begin{bmatrix} 1 & 0 & 0 \\ -1 & 0 & 0 \\ 0 & -1 & 0 \end{bmatrix} \\ & \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

$A =$

$$B = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} - \frac{A^T b}{A^T A} \quad a_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} - \frac{-1}{2} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} \\ 0 \end{bmatrix}$$

$$b_1 = \begin{bmatrix} \frac{\sqrt{3}}{2} \\ 0 \\ 0 \end{bmatrix}, b_2 = \begin{bmatrix} \frac{\sqrt{3}}{2} \\ 0 \\ 0 \end{bmatrix}, b_3 = \begin{bmatrix} 0 \\ \frac{\sqrt{3}}{2} \\ 0 \end{bmatrix} \rightarrow Q = \frac{1}{\sqrt{3}} \begin{bmatrix} \sqrt{3} & 1 & 1 \\ 0 & -\sqrt{3} & -1 \\ 0 & 0 & 2 \end{bmatrix}$$

$$C = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} - \frac{A^T C}{A^T A} \quad A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} - \frac{-1}{2} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & 0 \\ 0 & 0 & 1 \\ -\frac{1}{2} & 0 & 0 \end{bmatrix}$$

$$-\frac{B^T C}{B^T B} B$$

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

$\{a, b, c\}$ and $\{a, b, c\}$ are independent.

$$\text{rank}(\begin{bmatrix} 1 & 1 & 1 \end{bmatrix}) = 1 \Rightarrow \dim N(\begin{bmatrix} 1 & 1 & 1 \end{bmatrix}) = 3$$

$\{a, b, c\}$ and $\{a, b, c\}$ are bases. O.E.D.

Row of zeros $\Rightarrow \det A = 0$

$$\det A = \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = (-1)^{\# \text{ of rows}} (d)(d_2)(d_3) \dots (d_n)$$

$\det A = 0$ iff A is singular.

$$\det(AB) = (\det A)(\det B)$$

$$\det A^{-1} = \frac{1}{\det A}$$

$$\det(A^T) = \det A$$

(8.1)

If the entries of every row sum to 0, then 2 is a lin. comb of rows that equals 0. Then $\det A = 0$.

Subtracting 1 from 1 gives a matrix whose rows add to 0.

(8.2)

$$\begin{bmatrix} 1 & a & a^2 \\ 1 & b & b^2 \\ 1 & c & c^2 \end{bmatrix} = \begin{vmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{vmatrix} + \begin{vmatrix} 0 & a & a^2 \\ 0 & b & b^2 \\ 0 & c & c^2 \end{vmatrix} + \begin{vmatrix} 0 & 0 & a \\ 0 & 0 & b \\ 0 & 0 & c \end{vmatrix}$$

~~ad - bc~~

$$= (a+b+c) \begin{vmatrix} 1 & a & a^2 \\ 0 & b & b^2 \\ 0 & c & c^2 \end{vmatrix}$$

$$= (a+b+c) \begin{vmatrix} 1 & a & a^2 \\ 0 & 1 & b^2 \\ 0 & 0 & c^2 \end{vmatrix}$$

$$= (a+b+c)(a-b)(a-c)$$

$$= (a-b)(c-b)(c-a)$$

$$\text{See 19}$$

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

$$= \begin{vmatrix} a & 0 \\ 0 & d \end{vmatrix} + \begin{vmatrix} 0 & b \\ 0 & d \end{vmatrix} + \begin{vmatrix} 0 & b \\ c & 0 \end{vmatrix}$$

using ad - bc sing

$$\det A = \sum \pm a_{11}a_{22}\dots a_{nn}$$

Cofactor:

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$\det A = a_{11}C_{11} + a_{12}C_{12} + \dots + a_{1n}C_{1n}$$

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

Strang
"Dreaded cofactors."

"Oh, hell, oh I shouldn't have said 'hell' because it's all right."

2017-12-08

Lec 20

$$A^{-1} = \frac{1}{\det A} \begin{bmatrix} C_{11} & C_{12} & \dots & C_{1n} \\ C_{21} & C_{22} & \dots & C_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ C_{n1} & C_{n2} & \dots & C_{nn} \end{bmatrix}$$

cofactor matrix

$$Ax = b$$

$$x = A^{-1}b = \frac{1}{\det A} C^T b$$

$$x_1 = \frac{\det B_1}{\det A} \quad \begin{matrix} \leftarrow \text{Cramer's} \\ \text{rule} \end{matrix}$$

$$x_2 = \frac{\det B_2}{\det A} \quad \begin{matrix} \text{MUL stored} \\ \text{then elimination!} \end{matrix}$$

$$B_1 = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

B_j : a with col j replaced by b

Mathematics for the Matlab - Bell

det A = volume of a for parallelepiped



$$\text{if orthogonal: } Q^T Q = I$$

$$|Q|^2 = 1$$

$$Q = \pm 1$$

$$\Delta \text{ simplex} = \frac{1}{n!} \det A$$

Lec 21

Eigenvalues, Eigenvectors

$$\det [A - \lambda I] = 0$$

Eigenvector: x s.t. $Ax = \lambda x$
Eigenvalue: λ
If A is singular, $\lambda = 0$ is eval.

$$\text{if } A \text{ is a } 2 \times 2 \text{ proj matrix:}$$

$$P_x = x \text{ iff } x \text{ in the plane or } x \perp \text{plane}$$

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad x = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \lambda = 1$$

$$x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \lambda = -1$$

$\sum \lambda_{ii}$ is called the trace
the sum of entries is ~~called~~ trace.

$$Ax = \lambda x$$

$$(A - \lambda I)x = 0$$

i.e. $A - \lambda I$ is singular
 $\therefore \det(A - \lambda I) = 0$.
Find λ first.

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

$$\text{trace} = \lambda_1 + \lambda_2$$

$$\det = \lambda_1 \lambda_2$$

$$\begin{vmatrix} 3-\lambda & 1 \\ 1 & 2-\lambda \end{vmatrix} = 0$$

$$\text{In } 2 \times 2 \text{ case: } \lambda^2 - \text{tr} A + \det A = 0$$

$$\lambda^2 - 5\lambda + 8 = 0$$

$$\lambda_1 = 4, \lambda_2 = 2$$

$A - 4I = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \Rightarrow \lambda_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$
 $A - 2I = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \Rightarrow \lambda_2 = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$

Q proj mat for vec. sp?
Det trace

2017-12-29

2017-12-09

See 22

$$S^T A S = \Lambda$$

free 2 |
Autogressive matrices have one imaginary eigenvalue.

$$\lambda = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix} \rightarrow \det(\lambda - \lambda I) = \begin{vmatrix} 3-\lambda & 1 \\ 0 & 3-\lambda \end{vmatrix} = (3-\lambda)(3-\lambda)$$

$$\lambda_1 = 3, \quad \lambda_2 = 3$$

$$\begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix} \mathbf{x} = 0 \rightarrow x_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \text{ NO 2nd evec}$$

$$AS = SA$$

if n eigenvectors

$$S^{-1}AS = \Lambda$$

$$A = S\Lambda S^{-1}$$

A is a matrix times a diag times the inverse

If $\mathbf{x} = \lambda \mathbf{x}'$,
 $A^2 \mathbf{x} = \lambda^2 \mathbf{x}' \Rightarrow \mathbf{x}'$ are same eigenvectors
 $A^2 = S \Lambda S^{-1} S \Lambda S^{-1} = S \Lambda^2 S^{-1}$
 $A^k = S \Lambda^k S^{-1}$

$$A^k \rightarrow 0 \text{ as } k \rightarrow \infty \text{ if all } |\lambda_i| < 1$$

$$A^{k+1} = A^k A^1$$

interacts with us

$$U_{k+1} = A^k U_k$$

so

$$U_0 = c_1 x_1 + c_2 x_2 + \dots = S e$$

$$\begin{aligned} A^{100} U_0 &= c_1 A^{100} x_1 + c_2 A^{100} x_2 + \dots \\ A^{100} U_0 &= c_1 d_1^{100} x_1 + \dots \\ &= A^{100} S e \end{aligned}$$

$$\begin{aligned} F_{k+2} &= F_{k+1} + F_k \\ U_k &= \begin{bmatrix} F_{k+1} \\ F_k \end{bmatrix} \end{aligned}$$

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

$$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad \lambda^2 - \lambda - 1 = 0$$

$$\lambda = \frac{1 \pm \sqrt{5}}{2} \rightarrow \lambda_1 = \frac{1}{2}(1+\sqrt{5}) \approx 1.618 \dots$$

$$\lambda = \frac{1+\sqrt{5}}{2} \rightarrow \lambda_2 = \frac{1}{2}(1-\sqrt{5}) \approx -0.618$$

$$F_{100} \approx c_{100} (\frac{1+\sqrt{5}}{2})^{100}$$

2017-12-09
 D. direct sum
 D. not direct sum

$$\begin{aligned} AS &= A \begin{bmatrix} x_1 & x_2 & \dots \end{bmatrix} = \begin{bmatrix} \lambda_1 x_1 & \lambda_2 x_2 & \dots \end{bmatrix} \\ &= S \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \end{bmatrix} \end{aligned}$$

2017-12-09

Post. 21

$$(21) \quad (a) \quad S^T A S = S A S^{-1} S^{-T} A S^T = S A (S^T S)^{-1} A S^T$$

Δ Gérale
 Δ Décom

$$(19.1) \quad \det \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{vmatrix} = -1$$

(19.2) cofacteur accroissant.

$$\det P_n = 1$$

$$\det (P_n') = \det P_n + \begin{vmatrix} \text{top minor} \\ \dots \\ \dots \end{vmatrix}$$

$$\det P_n^{-1} = \det P_n - \det P_{n-1}$$

QED

$$(20.1) \quad C = \begin{bmatrix} 6 & -3 & 0 \\ -3 & 1 & -1 \\ -6 & 3 & 0 \end{bmatrix}$$

$$AC^T = \begin{bmatrix} 7 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & 7 \end{bmatrix}$$

$C = 0$.

$$(20.2) \quad \begin{aligned} \sin \phi \cos \theta & \quad P \cos \phi \cos \theta - P \sin \phi \sin \theta \\ \sin \phi \sin \theta & \quad P \cos \phi \sin \theta - P \sin \phi \cos \theta \\ \cos \phi & \quad 0 \end{aligned}$$

$$= P^2((\cos^2 \theta \sin^2 \phi)(\cos^2 \theta + \sin^2 \theta) + (\sin^2 \phi \sin^2 \theta)(\cos^2 \theta + \sin^2 \theta)) \\ = P^2 \sin^2 \phi$$

2017-12-09

Part 22

$$(4-\lambda)(2-\lambda) = 0 \\ \lambda_1 = 2, \lambda_2 = 4$$

$$S \in \left\{ \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \right\}$$

(ii)

$$\begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \begin{bmatrix} 6 & 9 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \frac{1}{10} \begin{bmatrix} 6 & 9 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \\ (6\alpha)(1-\alpha) - 36 = 0 \\ \alpha^2 - 7\alpha + 30 = (\alpha-10)(\alpha+2) = 0 \Rightarrow \alpha = 1, -2 \\ S = \begin{bmatrix} 6 & 1 \\ 4 & -3 \end{bmatrix}$$

$$A = \begin{bmatrix} -1 & 2 \\ 1 & -2 \end{bmatrix} \quad \frac{du}{dt} = Au \\ \lambda_1 = 0, \lambda_2 = -3 \\ \lambda_1 = 0 \\ \lambda_2 = -3 \\ \text{set } u = Sv$$

$$x = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, Ax = 0_x, \quad S_{\text{def}} = ASv \\ x_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, Ax_2 = -3x_2$$

$$u(t) = c_1 e^{\lambda_1 t} x_1 + c_2 e^{\lambda_2 t} x_2$$

$$\begin{bmatrix} u \\ v \end{bmatrix} = c_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + c_2 e^{-3t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$S = u(0)$$

$$\text{if } c_2 = 0:$$

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} + \frac{1}{3} e^{-3t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

- (i) Stability: If $\operatorname{Re}(\lambda) < 0$
 Steady state if $\lambda = 0$ and other $\lambda_i \leq 0$
 Otherwise if $\lambda > 0$

$$e^{At} = I + At + \frac{(At)^2}{2} + \frac{(At)^3}{6} + \dots + \frac{(At)^n}{n!} + \dots$$

$$= S \lambda_1^n S^{-1} e^{\lambda_1 t} + S \lambda_2^n S^{-1} e^{\lambda_2 t} + \dots + S \lambda_n^n S^{-1} e^{\lambda_n t} + \dots$$

$$= S \left(I + \lambda_1 t + \frac{\lambda_1^2 t^2}{2!} + \dots + \frac{\lambda_1^n t^n}{n!} \dots \right) S^{-1}$$

$$= S e^{\lambda_1 t} S^{-1} \quad (\text{assuming } A \text{ is diagonalizable})$$

$$e^{At} = \begin{bmatrix} e^{\lambda_1 t} & 0 \\ 0 & e^{\lambda_2 t} \end{bmatrix}$$

2017-12-11

Sec. 23

1st-order const coeff ODE system

$$\frac{du}{dt} = -u_1 + 2u_2 \\ \frac{dv}{dt} = u_1 - 2u_2$$

$$A = \begin{bmatrix} -1 & 2 \\ 1 & -2 \end{bmatrix} \quad \frac{du}{dt} = Au$$

$$u = Sv$$

$$x = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, Ax = 0_x, \quad S_{\text{def}} = ASv$$

2017-12-11

Sec 23

$$y'' + by' + ky = 0$$

~~u~~

$$u = \begin{bmatrix} y \\ y' \end{bmatrix}$$

$$\frac{du}{dt} = Au$$

~~free~~ ~~fixed~~

~~constant~~

$$u = S^{-1} A^{-1} v$$

the

2017-12-12

Sec 24

$$A = \begin{bmatrix} 1 & .1 & .3 \\ .2 & .4 & .3 \\ .7 & 0 & .1 \end{bmatrix}$$

Markov matrix

all rows sum to 1

all entries ≥ 0

all cols sum to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

all rows add to 1

all cols add to 1

$$\begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} = \begin{bmatrix} .9 & .1 & .0 \\ .1 & .2 & .8 \\ .0 & .8 & .1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}$$

populations after 10 years

projection matrix

projection is orthogonal

orthogonal basis

$v = x_1 g_1 + x_2 g_2 + \dots$

$g_1^T v = x_1$

$$v_k = c_1 P \begin{bmatrix} g_1 \\ g_2 \end{bmatrix} + c_2 P \begin{bmatrix} g_2 \\ g_1 \end{bmatrix}$$

$x_1 = c_1$

$x_2 = c_2$

2011-11-11

Part 23

See 24
Fourier Series

$$f(x) = a_0 + a_1 \cos nx + b_1 \sin nx + a_2 \cos 2nx + \dots$$

vectors
 $\sqrt{2}w = v_1 w_1 + v_2 w_2$

projections

$$f^T g = \int_D f(x)g(x)dx$$

$$a_1 = \frac{1}{\pi} \int_0^{2\pi} f(x) \cos nx dx$$

$$\frac{du}{dt} = \begin{bmatrix} 0 & c & -b \\ c & 0 & a \\ b & -a & 0 \end{bmatrix} u$$

$$\begin{bmatrix} -b & c & -b \\ c & -d & a \\ b & -a & -d \end{bmatrix} = -\lambda^3 + \lambda(\lambda^2 + b^2 + c^2) = 0$$

$$\lambda(\lambda^2 - (\lambda^2 + b^2 + c^2)) = 0$$

$$\lambda \in \mathbb{C}, \text{Im}\lambda \neq 0$$

$$\begin{bmatrix} e^{3t} & c e^{3t} & 1 \\ 0 & e^{3t} & 0 \\ 0 & 0 & \frac{1}{2} \end{bmatrix}$$

$$x_1 = \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} s & c & -b \\ -c & -s & a \\ b & a & -s \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\frac{du}{dt} = \begin{bmatrix} 0 & e^{3t} & ie^{-3t} \\ 0 & 0 & e^{3t} \\ 0 & 0 & e^{3t} \end{bmatrix} \begin{bmatrix} e^{3t} & ie^{-3t} & e^{3t} \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} e^{3t} & ie^{-3t} & e^{3t} \\ 0 & e^{3t} & 0 \\ 0 & 0 & e^{3t} \end{bmatrix} = \begin{bmatrix} e^{3t} & ie^{-3t} & e^{3t} \\ 0 & e^{3t} & 0 \\ 0 & 0 & e^{3t} \end{bmatrix}$$

Block

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} e^{3t} & ie^{-3t} & e^{3t} \\ 0 & e^{3t} & 0 \\ 0 & 0 & e^{3t} \end{bmatrix} = \begin{bmatrix} e^{6t} & ie^{-6t} & e^{6t} \\ 0 & e^{6t} & 0 \\ 0 & 0 & e^{6t} \end{bmatrix}$$

(3.1)

$$u = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \quad \frac{du}{dt} = \begin{bmatrix} 0 & c & -b \\ c & 0 & a \\ b & -a & 0 \end{bmatrix} u$$

$$2(a_1 u_1 - b_1 u_2 + c_1 u_3 + a_2 u_1 - c_2 u_2 + b_2 u_3 - a_3 u_1 + c_3 u_2 + b_3 u_3) = 0$$

$\|u\|$ is not changing.

where $a^2 + b^2 + c^2 = k$ for some k .

$$\frac{d}{dt} \|u(x)\|^2 = 2(a_1 u_1 - b_1 u_2 + c_1 u_3 + a_2 u_1 - c_2 u_2 + b_2 u_3 - a_3 u_1 + c_3 u_2 + b_3 u_3)$$

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

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

$$e^{At} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

2017-12-12

2017-12-23

D norm

Lee 25:

Symmetric matrices (real)

$$\text{Fact 24} \quad \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \quad \textcircled{a}$$

- (5) -1 is an eigenvalue. Also, if $A < 0$ but the trace > 0 .
- (6) If both eigenvalues were negative, the trace would be negative, but this is not.

24.2 $A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ invertible

permutation

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

projection

Markov

independents

Markov

$$\begin{aligned} A &= Q \Lambda Q^T \quad \text{Spectral Thm: } (\text{Spectrum is set of } \lambda_i) \\ Ax &= \lambda x \rightarrow \bar{A}\bar{x} = \bar{\lambda}\bar{x} \quad \text{If } \bar{a}\bar{b} = ab \quad \bar{b} = b \quad \bar{a} = a \\ A &\text{ is real} \\ \bar{A}x &= \bar{\lambda}x \\ \bar{x}^T A^T &= \bar{x}^T \bar{A} \\ A &\text{ is sym, so} \\ \bar{x}^T A &= \bar{x}^T \bar{A} \\ \text{and both sides by } x^T & \\ \bar{x}^T A x &= \bar{x}^T \bar{A} x \\ \bar{x}^T A x &= \bar{x}^T \lambda x \\ \bar{A} \bar{x}^T x &= \bar{\lambda} \bar{x}^T x \quad \text{and } \bar{x}^T x \neq 0, \text{ so} \\ \lambda &= \bar{\lambda}, \text{ so } J\text{eff}. \\ \text{For complex matrices, want } a^2 + b^2 > 0 & \\ \bar{A} &= A \end{aligned}$$

$$S^H = (S^H S)^{-1} S^H$$

$$\begin{aligned} Ax &= \sum_{i=1}^n \lambda_i x_i \\ \bar{A} \bar{x}^T x &= \bar{\lambda}_1 \bar{x}_1 + \bar{\lambda}_2 \bar{x}_2 + \dots \\ \lambda &= \bar{\lambda}_1, \text{ so } J\text{eff}. \\ \text{For complex matrices, want } a^2 + b^2 > 0 & \\ \bar{A} &= A \end{aligned}$$

$$A = Q \Lambda Q^T$$

Every symm matrix is the sum of rep. proj. matrices

For symm matrices:

of pivots = # pos. λ 's

positive definite matrices are great!

$A^T = A$ and all $\lambda > 0$

All subdeterminants are positive

2017-12-26

2017-12-26

Homework 26

$$z = \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix} \text{ in } \mathbb{C}^n$$

$$\text{length} = \sqrt{z^T z}$$

$$z^H = z^T z$$

inner product: $\langle z, w \rangle = z^H w$ (named after "correct") Hermitian

$$\text{Hermitian: } A^H = A$$

-real, $\perp \pi$

~~defn.~~

$$A = Q \Lambda Q^H = Q \Lambda Q^H$$

$Q^H Q = I \rightarrow Q$ is unitary

$$Q = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 0 & \omega & \dots & \omega^{n-1} \\ 0 & \omega^2 & \dots & \omega^{2(n-1)} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \omega^{(n-1)} & \dots & \omega^{(n-1)(n-1)} \end{bmatrix}, \quad q_i q_j = \begin{cases} 0 & i=j \\ \omega & i \neq j \end{cases}$$

Fourier Matrices:

$$F_n = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 0 & \omega & \dots & \omega^{n-1} \\ 0 & \omega^2 & \dots & \omega^{2(n-1)} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \omega^{(n-1)} & \dots & \omega^{(n-1)(n-1)} \end{bmatrix}, \quad \omega = e^{\frac{2\pi i}{n}}$$

$$BDT$$

$$F_4 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & \omega & \omega^2 & \omega^3 \\ 0 & \omega^2 & \omega & \omega^4 \\ 0 & \omega^3 & \omega^4 & \omega \end{bmatrix}, \quad \omega = e^{\frac{2\pi i}{4}}$$

$$F_8 = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & \omega & \omega^2 & \omega^4 & \omega^6 & \omega^8 & \omega^{16} & \omega^{32} \\ 0 & \omega^2 & \omega^4 & \omega^8 & \omega^{16} & \omega^{32} & \omega^{64} & \omega^{128} \\ \vdots & \vdots \\ 0 & \omega^{(n-1)} & \omega^{2(n-1)} & \omega^{4(n-1)} & \omega^{8(n-1)} & \omega^{16(n-1)} & \omega^{32(n-1)} & \omega^{64(n-1)} \end{bmatrix}$$

$$F_{64} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & \omega & \omega^2 & \omega^4 & \omega^8 & \omega^{16} & \omega^{32} & \omega^{64} & \omega^{128} & \omega^{256} & \omega^{512} & \omega^{1024} & \omega^{2048} & \omega^{4096} & \omega^{8192} & \omega^{16384} \\ 0 & \omega^2 & \omega^4 & \omega^8 & \omega^{16} & \omega^{32} & \omega^{64} & \omega^{128} & \omega^{256} & \omega^{512} & \omega^{1024} & \omega^{2048} & \omega^{4096} & \omega^{8192} & \omega^{16384} & \omega^{32768} \\ \vdots & \vdots \\ 0 & \omega^{(n-1)} & \omega^{2(n-1)} & \omega^{4(n-1)} & \omega^{8(n-1)} & \omega^{16(n-1)} & \omega^{32(n-1)} & \omega^{64(n-1)} & \omega^{128(n-1)} & \omega^{256(n-1)} & \omega^{512(n-1)} & \omega^{1024(n-1)} & \omega^{2048(n-1)} & \omega^{4096(n-1)} & \omega^{8192(n-1)} & \omega^{16384(n-1)} \end{bmatrix}$$

$$D = \begin{bmatrix} 1 & \omega & \omega^2 & \omega^3 \\ 0 & 1 & \omega & \omega^2 \\ 0 & 0 & 1 & \omega \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$F_4 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \omega & 0 & 0 \\ 0 & 0 & \omega & 0 \\ 0 & 0 & 0 & \omega \end{bmatrix}, \quad D_4 = \begin{bmatrix} 1 & 0 \\ 0 & \omega \end{bmatrix}$$

$$F_2 = \begin{bmatrix} 1 & 1 \\ 0 & -1 \end{bmatrix}, \quad D_2 = \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix}$$

$$(25) \quad \boxed{1} \quad \boxed{2} \quad \boxed{3} \quad \boxed{4}$$

$$\begin{bmatrix} I & P_1 \\ -P_1 & I \end{bmatrix}$$

$$\begin{bmatrix} F_2 & D_2 F_2 \\ F_2 & -P_1 F_2 \end{bmatrix}$$

$$P_2 = \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix}$$

$$\boxed{1} \quad \boxed{2} \quad \boxed{3} \quad \boxed{4}$$

$$\begin{bmatrix} I & C & A \\ B & I & D \\ C & D & I \end{bmatrix}$$

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} \bar{A}^T & \bar{B}^T \\ \bar{C}^T & \bar{D}^T \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\ 2 & 1 & 3 & 4 & 5 & 6 & 7 & 8 \\ 3 & 4 & 1 & 2 & 5 & 6 & 7 & 8 \\ 4 & 5 & 2 & 1 & 3 & 6 & 7 & 8 \\ 5 & 6 & 3 & 4 & 1 & 2 & 7 & 8 \\ 6 & 7 & 4 & 5 & 2 & 1 & 3 & 8 \\ 7 & 8 & 5 & 6 & 3 & 4 & 1 & 2 \\ 8 & 7 & 6 & 5 & 4 & 3 & 2 & 1 \end{bmatrix}$$

$$\begin{bmatrix} P_8 & P_8 \\ P_8 & P_8 \end{bmatrix}$$

2017-12-28

2017-12-29

See 27
Positive Definite Matrices
if $A = A^T$, then the following are cont:

- (1) all $a_{ii} > 0$
- (2) all subdets > 0
- (3) all pivots > 0
- (4) $x^T Ax \gg 0$ for any vector $x \neq 0$

$$x^T Ax = f(x_1, x_2) = 2x_1^2 + 12x_1x_2 + 18x_2^2 \\ ax^2 + 2bxy + cy^2$$

$$\text{Minimum: } f_{x_1} = 0, \quad f_{x_2} > 0 \\ \frac{\partial^2}{\partial x_i \partial x_i} (x^T Ax)$$

$$\frac{d}{dx}(Ax) = \lim_{h \rightarrow 0} \frac{A(x+h) - Ax}{h}$$

$$\lim_{h \rightarrow 0} \frac{Ax + Ah - Ax}{h} = A$$

\mathbb{Q} gives directions of axes
 Λ gives magnitudes

longest first then

$$A \text{ and } B \text{ are similar if } \exists M \text{ s.t. } B = MAM^{-1}$$

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \Rightarrow \lambda = \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -4 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 4 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & -15 \\ 1 & 6 \end{bmatrix}$$

similar

similar matrices

$$- B \neq M^{-1}AM \\ - same \lambda_i \\ - same det \\ - same trace \\ - same \# of \neq 0 x_i$$

BAD CASE:

$$\lambda_1 = \lambda_2 = 4$$

$$\begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} = 4I \text{ is in its own form}$$

big form (Jordan form)

$$J_i = \begin{bmatrix} 0 & & & \\ & \ddots & & \\ & & 0 & \\ & & & \ddots & 0 \end{bmatrix} \text{ One } J_i \text{ vs one } x_i$$

Every square A is sim to a Jordan matrix J :
Good case: $J=I$
One block per x_i
Proof
Don't need to compute J

See 28
Similar matrices

A is pos def for any rectangular A
 A is pos def, A^T is pos def. $(Q\Lambda Q^T)^T = Q\Lambda^T Q^T$

A is pos def, $A+B$ is pos def. $x^T(A+B)x = x^TAx + x^TBx = 0$
 $x^T A^T A x = (Ax)^T (Ax) \geq 0$ so at least has ≥ 0 -def

pos def iff rank n

(non negative)
 A and B are similar if $\exists M$ s.t. $B = MAM^{-1}$

$$SAS^{-1} = I \rightarrow A \text{ is similar to } I$$

"think of these matrices as people by their point in 1D"

2017-12-29

Lec 25

2017-12-29

Dec 29
Singular Value Decomposition (SVD)
 $A = U\Sigma V^T$ (U, V orthogonal, Σ diagonal)

SVD: an $n \times n$ basis for rowspace \rightarrow In basis
for col space

nullspace \rightarrow gives one diag of Σ

$$A \begin{bmatrix} y_1 & y_2 & \dots & y_r \end{bmatrix} = \tilde{U} \begin{bmatrix} \sigma_1 & & & \\ & \ddots & & \\ & & \sigma_r & \\ & & & 0 \end{bmatrix} \sum$$

$$A = \begin{bmatrix} u_1 & u_2 & \dots & u_r \end{bmatrix} \sum \begin{bmatrix} \sqrt{\lambda_1} & & & \\ & \ddots & & \\ & & \sqrt{\lambda_r} & \\ & & & 0 \end{bmatrix} V^T$$

$$\Sigma = \sqrt{\lambda_1} \begin{bmatrix} \sigma_1 & & & \\ & \ddots & & \\ & & \sigma_r & \\ & & & 0 \end{bmatrix} \sqrt{\lambda_1}$$

$$A A^T = U \Sigma V^T V \Sigma U^T = U \begin{bmatrix} \sigma_1^2 & & & \\ & \ddots & & \\ & & \sigma_r^2 & \\ & & & 0 \end{bmatrix} U^T$$

$$A \begin{bmatrix} u & v \end{bmatrix} \Rightarrow A^T A = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} u & u \\ v & v \end{bmatrix} = \begin{bmatrix} \frac{u}{\|u\|_2} & 0 \\ 0 & \frac{v}{\|v\|_2} \end{bmatrix}, \quad \text{S2}$$

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{u}{\|u\|_2} & 0 \\ 0 & \frac{v}{\|v\|_2} \end{bmatrix} \begin{bmatrix} u & v \end{bmatrix}$$

$$\therefore A A^T = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} u & v \end{bmatrix} = \begin{bmatrix} 3 & 2 & 0 \\ 3 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow S = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 3 & 2 & 0 \\ 3 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

 \equiv

$$A = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} \frac{u}{\|u\|_2} & 0 \\ 0 & \frac{v}{\|v\|_2} \end{bmatrix} = \begin{bmatrix} 25 & 50 \\ 50 & 100 \end{bmatrix} = \begin{bmatrix} 25 & 50 \\ 50 & 100 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$A A^T = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} u & v \end{bmatrix} = \begin{bmatrix} 25 & 50 \\ 50 & 100 \end{bmatrix} \Rightarrow 0, 125$$

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

u, v : orthonormal basis for row space
 u, v : " column space
 u, v : " nullspace
 u, v : " left nullspace

$$Ax = \lambda x \Rightarrow x^T Ax = d x^T x \Rightarrow d = \frac{x^T Ax}{x^T x}$$

$$[c] \begin{bmatrix} 0 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} c \\ -c \end{bmatrix} = i [c] \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$x^T x \text{ can be } 0.$$

$$\text{Q-R algorithm}$$

$$(Q^T A^T)^T = Q^T A^T = P^T N^T Q^T$$

$$P^T P = I, Q^T Q = I$$

$$P Q^T Q P^T = I$$

$$P Q^T = I, Q P^T = I$$

$$P Q^T Q P^T = I$$

$$\text{N/A}$$

$$\text{X}^T A X \neq 0$$

$$\text{X}^T A^T X > 0$$

$$\text{X}^T A^T X > 0$$

$$\text{P}^T P = I, Q^T Q = I$$

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

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

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

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

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

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

Lesson Notes 27, 28

2017-12-29

$$\begin{aligned} \textcircled{(27.1)} \quad & ABx = \lambda x \\ \textcircled{(27.2)} \quad & A^TBA^TBx = \lambda x^T Bx \\ \textcircled{(27.3)} \quad & x^T B^T A^T Bx = \lambda x^T Bx \\ \textcircled{(27.4)} \quad & Bx = \lambda x \quad A^T x \\ \textcircled{(27.5)} \quad & x^T Bx = \lambda x^T A^T x \\ \textcircled{(27.6)} \quad & \frac{x^T Bx}{x^T Ax} = \lambda > 0 \quad \text{QED} \end{aligned}$$

$$\textcircled{(27.7)} \quad x^2 + 10xy + 9y^2 = (x+6y)^2 - 27y^2$$

(28.1) sometimes +, sometimes -

$$M = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = MK$$

$$\begin{aligned} M &= \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = MK \\ &\text{where } M = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix}, K = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \\ &\text{and } B = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} \end{aligned}$$

$$\begin{aligned} &\text{where } M = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix}, K = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \\ &\text{and } B = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix} \end{aligned}$$

$$\begin{aligned} \textcircled{(28.2)} \quad & B = MAM^{-1} \Rightarrow B^2 = M^2 A^2 M^{-2} \quad \text{QED} \\ \textcircled{(28.3)} \quad & \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \text{ not sim.} \end{aligned}$$

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

$$\begin{aligned} \textcircled{(29.1)} \quad & A^2 = B^2 \\ \textcircled{(29.2)} \quad & \begin{bmatrix} 3 & 0 \\ 0 & 4 \end{bmatrix}, \begin{bmatrix} 3 & 1 \\ 0 & 4 \end{bmatrix} \text{ have the same eigenvalues} \\ \textcircled{(29.3)} \quad & \begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} \text{ can't be decomposed} \end{aligned}$$

$$\begin{aligned} \textcircled{(29.4)} \quad & \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} A \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} = B, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}^2 = I \end{aligned}$$

see 29

$$\begin{aligned} \textcircled{(29.1)} \quad & A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}, A^T A = \sqrt{2} \lambda V \sum^2 \sqrt{\lambda} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \\ \textcircled{(29.2)} \quad & \lambda_1 = 3, \lambda_2 = 1, \lambda_3 = -\frac{3-\sqrt{17}}{2}, \lambda_4 = \frac{3+\sqrt{17}}{2} \\ \textcircled{(29.3)} \quad & A^T A = \begin{bmatrix} \sigma_1^2 & 0 & 0 & 0 \\ 0 & \sigma_2^2 & 0 & 0 \\ 0 & 0 & \sigma_3^2 & 0 \\ 0 & 0 & 0 & \sigma_4^2 \end{bmatrix} = \frac{1}{4} \sum_{i=1}^4 \sigma_i^2, U, V = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \end{aligned}$$

2017-12-29

2017-12-30

Sec 3D

$$T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

Given target
Input: $T(v) = (a_1 v_1 + a_2 v_2 + \dots + a_m v_m)$

$$T = \frac{d}{dx}(1) \text{ Input: } c_1 + c_2 x + c_3 x^2$$

$$\text{Output basis: } \begin{bmatrix} 1 & x & x^2 \end{bmatrix}$$

Linear transformations

$$T(v+w) = T(v) + T(w)$$

$$T(cv) = cT(v)$$

$$T(v+dw) = cT(v) + dT(w)$$

$$\Rightarrow T(o) = 0$$

Proj. to aLT

Shift is not a aLT
Covariation is not L

$T(v) = Av$ for any matrix A (rect.)

$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ reflects over y-axis

Understands its by looking at matrix

$$T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

To gain understanding of LT , need only understand L
Can do it.
Coordinates w.r.t. basis v_1, v_2, v_3 .

$$T: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

Choose basis v_1, \dots, v_n for inputs

$\begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} + \sum \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ choose basis w_1, \dots, w_m for outputs.

$$\begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} v_1 & v_2 & \dots & v_n \\ w_1 & w_2 & \dots & w_m \end{bmatrix}$$

An eigenvector basis gives a diagonal matrix Λ

$$Tv = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$$

$$Tv = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} v$$

2017-12-30
that video signal
will go to the
bottom of the
bottom of the

$$\left[\begin{array}{c} \text{512 pixel} \\ 0 \leq x \leq 512 \\ n = 512^2 \end{array} \right]$$

$$\text{JPEG - Joint Photographic Experts Group}$$

$$\left[\begin{array}{c} \text{512} \\ 0 \leq x \leq 512 \\ \text{Standard basis: } \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \dots, \text{Any } T \text{ has similar A in all bases} \end{array} \right]$$

$$\text{Better basis: } \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \text{ whatever}$$

Break down into 8x8 blocks

$$\left[\begin{array}{c} \text{Fourier basis:} \\ \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & -1 & 1 & 1 & -1 & -1 & 1 & 1 \\ 1 & 1 & -1 & -1 & -1 & 1 & 1 & 1 \\ 1 & -1 & -1 & -1 & 1 & -1 & -1 & 1 \end{bmatrix} \text{ size: } 8 \times 8 \\ \dots \\ \text{64 vectors} \end{array} \right] \dots \left[\begin{array}{c} \text{64-dimensional} \end{array} \right]$$

Drop out all the vectors with really small values

"So Washington had a file of 3D million dimensions, cheaters on Congress, think sort of stuff..."

Wavelet basis: (RF)

$$\left[\begin{array}{c} \text{64 vectors} \\ \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \dots \end{array} \right]$$

orthogonal!

$$W^{-1} W = \text{diag. } D^T$$

Good basis:

(1) easy to multiply
- FFT or PLDF (fast wavelet T_P)

(2) a few vectors give enough info

D = new basis vectors in terms of old

$$X = Wc$$

c = old basis

2014-12-30

See 32 - Left, Right and pseudo-inverse 2014-12-30

30.1 Order:
 $T(x, y) = (2x, y)$ $\sqrt{(2x)^2 + (y)^2} = \sqrt{4x^2 + y^2} = 2\sqrt{x^2 + y^2}$
 $\frac{\partial T}{\partial x}(2x, y) = \frac{\partial(2x)}{\partial x} = 2$

30.2 $\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ Cartesian and $\begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$ polar

30.3 $T(x, y) = (\sqrt{x}, \sin(y))$

31.1 The dot product rule of any two vectors v_1 and v_2 is $v_1 \cdot v_2 = a_1 a_1 + a_1 a_2 + \dots + a_n a_1 + a_n a_2 + \dots + a_n a_n \geq 0$

where $a_i = v_i \cdot e_i$

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

The first basis is probably better for all three

31.2
 $\text{rank } A = n$
 $A^{-1} \text{ right} = A^T (A A^T)^{-1}$
 $A^{-1} \text{ left} = (A^T A)^{-1} A^T$
 If you put them in the wrong order you get $A(A^T A)^{-1} A^T$ or $A^T (A A^T)^{-1}$ (projection matrices)
 "statisticians are
 doing least-squares
 and they don't care if they're
 invertible! (but they do care about the condition number)"

A is a bijection from the row space to the column space;

pseudo-inverse denoted A^+ . For x in rowspace:

$$A^+(A x) = x \quad \text{for } x \neq 0$$

Then $x-y \in N(A)$, but $x-y \in C(A^+)$, so $(x-y) \cdot 0 = 0$, i.e. $x=y$, but $x \neq y$ so by contra. $A x \neq A y$

$$A = U \Sigma V^T \Rightarrow \Sigma = \begin{bmatrix} \sigma_1 & 0 \\ 0 & 0 \end{bmatrix} \Rightarrow \Sigma^+ = \begin{bmatrix} 0 & 0 \\ 0 & \sigma_1^{-1} \end{bmatrix} \Rightarrow \Sigma \Sigma^+ = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

$$A^+ = V \Sigma^+ U^T$$

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

"see you guys
 are in about great
 shape now if that
 means US is ending
 or the guy was
 good, but..."

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

"left inverse"

"right inverse"

Bent 32
B2:1

08-21-4107;

$$A^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad A_r^{-1} = A^T (AA^T)^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \left(\begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \right)^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1/2 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1/2 & 0 \\ 0 & 1 \end{bmatrix}$$

228

only a pseudowave, as $m \gg r$.

$$A = \begin{bmatrix} u & v \\ w & x \end{bmatrix}, \quad A^T A = \begin{bmatrix} u & w \\ v & x \end{bmatrix} \begin{bmatrix} u & v \\ w & x \end{bmatrix} = \begin{bmatrix} u^2 & uv \\ vw & x^2 \end{bmatrix} = \frac{1}{60} \begin{bmatrix} 25 & 50 \\ 50 & 100 \end{bmatrix} = \frac{1}{12} \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 25 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

$$A^+ = \begin{bmatrix} 1/5 & -1/5 & -1/5 \\ -1/5 & 1/5 & -1/5 \\ -1/5 & -1/5 & 1/5 \end{bmatrix} \begin{bmatrix} 5/5 & 0 & 0 \\ 0 & 5/5 & 0 \\ 0 & 0 & 5/5 \end{bmatrix} = \begin{bmatrix} 1/5 & -1/5 & -1/5 \\ -1/5 & 1/5 & -1/5 \\ -1/5 & -1/5 & 1/5 \end{bmatrix} \begin{bmatrix} 1/5 & 1/5 & 1/5 \\ 1/5 & 1/5 & 1/5 \\ 1/5 & 1/5 & 1/5 \end{bmatrix} = \begin{bmatrix} 1/25 & -1/25 & -1/25 \\ -1/25 & 1/25 & -1/25 \\ -1/25 & -1/25 & 1/25 \end{bmatrix}$$

卷之三

卷之三