

Math 390

Friday, March 12

Normal, Positive Semi-Definite, SVD

Proof from last time, need to assume $\underline{x} \in \mathbb{R}^n$ (not \mathbb{C}^n)

Theorem Given \underline{x} , define $\underline{v} = \underline{x} \pm \|\underline{x}\| \underline{e}_1$, P Householder for \underline{v} ,
 then $P\underline{x} = \mp \|\underline{x}\| \underline{e}_1$ (P "zeros out" \underline{x})

Proof

$$\begin{aligned} \langle \underline{v}, \underline{v} \rangle &= (\underline{x} \pm \|\underline{x}\| \underline{e}_1)^* (\underline{x} \pm \|\underline{x}\| \underline{e}_1) \\ &= (\underline{x}^* \pm \|\underline{x}\| \underline{e}_1^*) (\underline{x} \pm \|\underline{x}\| \underline{e}_1) \\ &= \underline{x}^* \underline{x} \pm \|\underline{x}\| \underline{x}^* \underline{e}_1 \pm \|\underline{x}\| \underline{e}_1^* \underline{x} + \|\underline{x}\|^2 \underline{e}_1^* \underline{e}_1 \\ &= \underline{x}^* \underline{x} \pm \|\underline{x}\| \underline{x}^* \underline{e}_1 \pm \|\underline{x}\| \underline{e}_1^* \underline{x} + \underline{x}^* \underline{x} \quad (\dagger) \end{aligned}$$

$$= 2 \underline{x}^* \underline{x} \pm \|\underline{x}\| \overline{\rho_1^* \underline{x}} \pm \|\underline{x}\| \underline{e}_1^* \underline{x}$$

$$= 2 \underline{x}^* \underline{x} \pm 2 \|\underline{x}\| \underline{e}_1^* \underline{x}$$

$$= 2 (\underline{x}^* \pm \|\underline{x}\| \underline{e}_1^*) \underline{x} = 2 (\underline{x} \pm \|\underline{x}\| \underline{e}_1)^* \underline{x} = \underline{2 \underline{v}^* \underline{x}}$$

1.7 Normal Matrices

Defn A square, we say A is normal if $A^*A = AA^*$.

Examples diagonal matrix, Hermitian (self-adjoint) (symmetric),
Unitary ($Q^*Q = I$, so $QQ^* = I$), skew-symmetric ($A^t = -A$)

Example 1.7.2 $A = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ normal, not one of \rightarrow

Theorem OD (FCLA)

$$A = U^* D U \iff A \text{ normal}$$

orthonormal
diagonalization \Rightarrow

\uparrow diagonal
 \uparrow unitary

1.8 Positive Semi-Definite Matrices

Defn A p.s.d., if A Hermitian, and for all \underline{x} $\langle \underline{x}, A\underline{x} \rangle \geq 0$

positive definite $>$

negative semi-definite \leq

negative definite $<$

Theorem Suppose A is $m \times n$ matrix then A^*A $n \times n$ & AA^* $m \times m$ are p.s.d.

Proof Check AA^* Hermitian,
 $(AA^*)^* = (A^*)^* A^* = AA^*$.

Check $\langle \underline{x}, (AA^*)\underline{x} \rangle = \langle \underline{x}, A(A^*\underline{x}) \rangle = \langle A^*\underline{x}, A^*\underline{x} \rangle \geq 0$ for all $\underline{x} \in \mathbb{C}^m$

$\langle \underline{u}, \underline{u} \rangle \in \mathbb{R}$ $\alpha \in \mathbb{C}, \alpha = a+bi$
 $\underline{u}^* \underline{u}$ $\bar{2}\alpha = a^2 + b^2$

inner-product property

Theorem PIP

$$w = f(x, y, z) \quad \vec{x} = (x, y, z)^t$$

$$= f(\vec{x})$$

$$f(\vec{x}) = f(\vec{x}_0) + \vec{\nabla} f(\vec{x}_0) \cdot \vec{x} + \frac{1}{2} \vec{x}^t H \vec{x} \quad \langle \vec{x}, H \vec{x} \rangle$$

max/min, critical point

$$\vec{\nabla} f(\vec{x}_0) = \vec{0}$$

$$H = \begin{bmatrix} f_{xx} & f_{xy} & f_{xz} \\ f_{yx} & f_{yy} & f_{yz} \\ f_{zx} & f_{zy} & f_{zz} \end{bmatrix}$$

x_0 is a min, want to be always positive

$$\left[\begin{array}{l} D = f_{xx} f_{yy} - f_{xy}^2 \\ f_{xx} > 0 \end{array} \right]$$