

HW 4

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Problem 1

Prove that the Euclidean norm $|x| := \sqrt{\langle x, x \rangle} = \sqrt{x_1^2 + \cdots + x_n^2}$ satisfies the triangle inequality. (Hint: use the Cauchy-Schwarz inequality $|\langle x, y \rangle|^2 \leq \langle x, x \rangle \cdot \langle y, y \rangle$.)

Answer

Triangle inequality is $\|x + y\| \leq \|x\| + \|y\|$.
Then, apply to Euclidean norm,

$$\begin{aligned} |x + y|^2 &= \langle x + y, x + y \rangle \\ &= \langle x, x \rangle + 2\langle x, y \rangle + \langle y, y \rangle \\ &= \langle x, x \rangle + 2|\langle x, y \rangle| + \langle y, y \rangle \end{aligned}$$

By Cauchy-Schwarz inequality, $|\langle x, y \rangle| \leq |x| \cdot |y|$, then

$$\begin{aligned} |x + y|^2 &\leq \langle x, x \rangle + 2|x| \cdot |y| + \langle y, y \rangle \\ &= |x|^2 + 2|x| \cdot |y| + |y|^2 \\ &= (|x| + |y|)^2 \end{aligned}$$

Therefore, $|x + y| \leq |x| + |y|$. \Rightarrow Triangle inequality holds.

Problem 2

Let A be a symmetric real-valued square matrix.

a) Show that if $\lambda + i\mu$ is an eigenvalue of A and $z = x + iy$ is a corresponding eigenvector, then $\mu = 0$ and x is an eigenvector. In other words, eigenvalues of symmetric matrices are always real and eigenvectors can always be chosen to be real. (Hint: show that $\bar{z}^T Az$ is real.)

Answer

$$\begin{aligned} Az &= (\lambda + i\mu)z \\ \Leftrightarrow A(x + iy) &= (\lambda + i\mu)(x + iy) \\ \Leftrightarrow Ax + iAy &= \lambda x - \mu y + i(\mu x + \lambda y) \end{aligned}$$

Devide the equation into real and imaginary parts,

Real parts : $Ax = \lambda x - \mu y$

imaginary parts : $Ay = \mu x + \lambda y$

Now, use the $\bar{z}^T Az$

$$\begin{aligned} \bar{z}^T Az &= (x^T - iy^T)A(x + iy) \\ &= x^T Ax + ix^T Ay - ix^T Ay + y^T Ay \end{aligned}$$

Since A is symmetric, $A = A^T$, so $x^T Ay = y^T Ax$

Then,

$$\begin{aligned} \bar{z}^T Az &= x^T Ax + ix^T Ay - ix^T Ay + y^T Ay \\ &= x^T Ax + y^T Ay \end{aligned}$$

Use this to the origin equation,

$$\begin{aligned} Az &= (\lambda + i\mu)z \Rightarrow \bar{z}^T Az = \bar{z}^T(\lambda + i\mu)z = (\lambda + i\mu)\bar{z}^T z \\ \text{and, } \bar{z}^T z &= x^T x + y^T y \geq 0 \end{aligned}$$

Thus, $\bar{z}^T Az$ is real.

Therefore, $\mu = 0$ and x is an eigenvector.

b) Show that eigenvectors of A corresponding to distinct eigenvalues are orthogonal.

Answer

Suppose that exist two distinct eigenvalues and corresponding eigenvectors.

$$\lambda_1 \rightarrow v_1$$

$$\lambda_2 \rightarrow v_2$$

Then, $Av_1 = \lambda_1 v_1$ and $Av_2 = \lambda_2 v_2$

Now,

$$v_2^T Av_1 = v_2^T (\lambda_1 v_1) = \lambda_1 (v_2^T v_1)$$

And A is symmetric, so $(Av_2)^T v_1 = v_2^T Av_1$

Thus,

$$\begin{aligned} v_2^T Av_1 &= (Av_2)^T v_1 = \lambda_1 (v_2^T v_1) \\ &\Leftrightarrow \lambda_2 (v_2^T v_1) = \lambda_1 (v_2^T v_1) \end{aligned}$$

Finally, we can get

$$(v_2^T v_1)(\lambda_2 - \lambda_1) = 0$$

Since λ_1 and λ_2 are distinct, $v_2^T v_1 = 0$

Therefore, eigenvectors of A corresponding to distinct eigenvalues are orthogonal.

Problem 3

Let M be a symmetric real-valued $n \times n$ matrix. Show that the following three statements are equivalent:

1. M is positive definite.
2. All eigenvalues of M are positive.
3. $M = N^T N$ for some nonsingular $n \times n$ matrix N .

(Hint: show $1 \Rightarrow 2 \Rightarrow 3 \Rightarrow 1$.)

Answer $1 \Rightarrow 2$

Suppose that M is positive definite. Then, $\forall x \neq 0, x^T M x > 0$

M is symmetric, so all eigenvalues are real ($\lambda_1, \lambda_2, \dots, \lambda_n$) and corresponding real eigenvectors (v_1, v_2, \dots, v_i)

Then, $Mv_i\lambda_i v_i$, and set $v_i = x$

$$\lambda_i = v_i^T M v_i \rightarrow x^T M x > 0$$

Therefore, all eigenvalues λ_i are **real number**.

Answer 2 \Rightarrow 3

Now, A 's all eigenvalues are real number and all eigenvectors are orthogonal (we proofed in problem 2).

Then, we can use eigenvalue decomposition (EVD) by Spectral Theorem.

Then,

$$M = Q\Sigma Q^T$$

where Q is $n \times n$ matrix composed by M 's eigenvectors and Σ is diagonal matrix composed by M 's eigenvalues.

Let $N = \Sigma^{1/2}Q$, then

$$N^T N = Q\Sigma Q^T = M$$

Answer 3 \Rightarrow 1

Suppose that $M = N^T N$ for some nonsingular $n \times n$ matrix N .

Then, $\forall x \neq 0$, $x^T M x = x^T N^T N x = (Nx)^T (Nx) = \|Nx\|^2 > 0$

Therefore, M is positive definite. Q.E.D

Problem 4

Let X and Y be linear vector spaces over \mathbb{R} equipped with inner products $\langle \cdot, \cdot \rangle_X$ and $\langle \cdot, \cdot \rangle_Y$, respectively. Let $L : X \rightarrow Y$ be a linear operator. We define the adjoint of L to be a linear operator $L^* : Y \rightarrow X$ with the property that

$$\langle y, Lx \rangle_Y = \langle L^*y, x \rangle_X \quad \forall x \in X, y \in Y$$

Assume that the map $LL^* : Y \rightarrow Y$ is invertible. Then the equation $Lx = y_0$ has a solution

$$x_0 = L^*(LL^*)^{-1}y_0$$

for each $y_0 \in Y$. Prove that if x_1 is any other solution of $Lx = y_0$, then $\langle x_1, x_1 \rangle \geq \langle x_0, x_0 \rangle$. (Hint: Let $y_1 := (LL^*)^{-1}y_0$. Using the definition of adjoint, show that $\langle y_1, Lx_0 \rangle = \langle x_0, x_0 \rangle$ and also that $\langle x_0, x_1 \rangle = \langle y_1, Lx_0 \rangle$. Complete the proof by using the fact that $\langle x_1 - x_0, x_1 - x_0 \rangle \geq 0$.)

Answer

Let $y_1 = (LL^*)^{-1}y_0$.

Then, $y_0 = LL^*y_1$ and $x_0 = L^*y_1$

To show that $\langle y_1, Lx_0 \rangle = \langle x_0, x_0 \rangle$,

$$\langle y_1, Lx_0 \rangle_Y = \langle y_1, L(L^*y_1) \rangle_Y = \langle y_1, (LL^*)y_1 \rangle_Y$$

by adjoint of L , $\langle y_1, LL^*y_1 \rangle_Y = \langle L^*y_1, L^*y_1 \rangle_X = \langle x_0, x_0 \rangle_X$

Also, $\langle x_0, x_1 \rangle_X = \langle y_1, Lx_1 \rangle_Y$

$$\langle x_0, x_1 \rangle_X = \langle L^*y_1, x_1 \rangle_X = \langle y_1, Lx_1 \rangle_Y$$

Now,

$$\begin{aligned} \langle x_1, x_1 \rangle &= \langle x_0 + (x_1 - x_0), x_0 + (x_1 - x_0) \rangle \\ &= \langle x_0, x_0 \rangle + 2(\langle x_0, x_1 - x_0 \rangle) + \langle x_1 - x_0, x_1 - x_0 \rangle \end{aligned}$$

Problem 5

Using the stability definitions given in class, determine if the systems below are stable, asymptotically stable, globally asymptotically stable, or neither. The first two systems are in \mathbb{R}^2 , the last is in \mathbb{R} .

a)

$$\dot{x}_1 = 0, \quad \dot{x}_2 = -x_2$$

b)

$$\dot{x}_1 = -x_2, \quad \dot{x}_2 = 0$$

c)

$$\dot{x} = 0 \text{ if } |x| > 1, \quad \dot{x} = -x \text{ if } |x| \leq 1$$

Justify your answers using only the definitions of stability (not eigenvalues or Lyapunov's method).

Answer

a) Have to AS, $x_1 = 0$ and $x_2 = 0$. But, $\dot{x}_1 = 0$, so x_1 is maintain the initial value. So this system is stable.

b) $\dot{x}_2 = 0$ is maintain the initial value, but $\dot{x}_1 = -x_2$ is linearly change the value. So this system is not stable, not AS, not GAS.

c) If $|x| > 1$, then the system is stable. Also, $|x| \leq 1$ is go to the equilibrium point, so when $|x| \leq 1$, the system is AS.

Problem 6

Consider the LTI system $\dot{x} = Ax$ where

$$A = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & -1 & 2 \end{pmatrix}$$

Identify the stable and unstable invariant subspaces by giving a real basis for each of them.

Answer

Stability of LTI system is determined by the eigenvalues of the matrix A .

The general solution of the system is $x(t) = e^{At}x(0)$, so negative eigenvalues are stable and positive eigenvalues are unstable.

```
>> A = [-1, 0, 0; 0, 2, 1; 0, -1, 2]
```

```
A =
```

```
   -1     0     0
    0     2     1
    0    -1     2
```

```
>> eig(A)
```

```
e =
```

```
   2.0000 + 1.0000i
   2.0000 - 1.0000i
  -1.0000 + 0.0000i
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The eigenvalues of the matrix A are $2 \pm i$ and -1 .
Therefore,

$$\text{Stable invariant subspace} = \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right\}$$

$$\text{Unstable invariant subspace} = \left\{ \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\}$$