

Lec 8 Existence Thm of linear system [BN 2.2]

Warm up.

Existence?	f cts
Uniqueness?	f Lip & bdd
Extension of soln?	A priori estimate. (needs RHS to be bdd)

Thm. $\frac{dy}{dt} = A(t)y + g(t) \quad y(t_0) = y_0$

If $A(t), g(t)$ are cts on some interval $[a, b]$, and $t_0 \in [a, b]$, $y_0 < \infty$.

Then system has a unique soln $\phi(t)$ satisfying $\phi(t_0) = y_0$ and existing on the interval $[a, b]$.

Only need to show $|\phi(t)| \leq M$ indep of t . (What. $\phi(t)$ is a vector, what does $|\phi(t)|$ mean?)

Tools (norm) [BN 14, 2.1]

recall when $x \in \mathbb{R}$, we use $|x|$ to measure the "length" of x .

Idea of norm: want to present "length"

1. vector \mathbb{R}^n n-dim linear (vector) sp over \mathbb{R} . (E_n in book)

$$y \in \mathbb{R}^n \quad \bar{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

- Euclidean norm (2 norm)

$$\|\bar{y}\|_2 = \left[|y_1|^2 + \dots + |y_n|^2 \right]^{\frac{1}{2}} = \left[\sum_i |y_i|^2 \right]^{\frac{1}{2}} \quad \|\cdot\|_2 \quad \text{natural / physical way.}$$

- 1-norm (more convenient for us)

$$\|\bar{y}\|_1 = |y_1| + \dots + |y_n| = \sum_i |y_i| \quad \|\cdot\|_1$$

Q: If Norm is not unique. Generalize? What is a norm, generally speaking?

Norm $\forall x \in X$.

- $\|x\| \geq 0$ and $\|x\|=0$ if $x=0$

- $\|cx\| = |c|\|x\|$ $\forall c$ const-

- $\|x_1+x_2\| \leq \|x_1\| + \|x_2\| \quad \forall x_1, x_2 \in X$ (triangle Ineq)

Easy to check: $\|\cdot\|_2, \|\cdot\|_1$ are norms.

Rmk: 1. In fact, one can def any p -norm $\|\cdot\|_p = \left(\sum_i |y_i|^p\right)^{1/p}$ ($p \geq 1$)
 Even when p is not integer.

In particular, $p=\infty$, $\|\cdot\|_\infty$ is defined as $\max_{1 \leq i \leq n} |y_i|$
 (one could check the norm definition)

2. For finite-D vector sp., norms are equivalent!

In the sense, if a, b .

$$C_1\|x\|_b \leq \|x\|_a \leq C_2\|x\|_b \quad \text{for some const } C_1, C_2.$$

ex: $\|\cdot\|_\infty$ and $\|\cdot\|_1$ are equivalent.

$$\max_{1 \leq i \leq n} |y_i| \leq |y_1| + \dots + |y_n| \leq n \max_{1 \leq i \leq n} |y_i|$$

pf $\|\cdot\|_2$ and $\|\cdot\|_1$ equivalent (HW)

Why is this useful? It tells us we could work w/ any norm we want in a priori estimate.

3. In fact, def of norms can also be generalized to ∞ -D vector sp.

namely, sp of seqs. $y = (y_1, y_2, \dots)$

Then one can define p -norm. $\|\cdot\|_p = \left(\sum_{i \in \mathbb{N}} |y_i|^p\right)^{1/p}$

But note that for " ∞ -D", we do NOT have the norm equivalence.

4. Metric sp?

A norm can be turned into a metric.

(X, d) $d(x, y)$ is metric "induced metric"

One could def $d(x, y) = \|x - y\|$ to construct a metric from norm.

\Rightarrow normed sp is a metric sp.

The converse is not true. In general, a metric can not be turned into a norm.

Metric $d(x, y)$

- $d(x, y) > 0$ and $d(x, y) = 0 \iff x = y$

- $d(x, y) = d(y, x)$

- $d(x, y) \leq d(x, z) + d(z, y)$

Ex Discrete metric on \mathbb{R}

$$d(x, y) = \begin{cases} 1 & x \neq y \\ 0 & x = y \end{cases}$$

[Metric satisfying 3 more conditions can have induced norm]

$d(x, y) = d(x+z, y+z)$ translation invr.

$d(cx, cy) = |c| d(x, y)$. homogeneous.

can define $\|x\| \triangleq d(x, 0)$

2. matrix. $\mathbb{R}^{n \times n}$ or $M_n(\mathbb{R})$

Intuitively, we could reshape a matrix to a vector, s.t. all the $\| \cdot \|_p$ can still be defined.

For us, the most useful one:

$$[A] = \sum_{j=1}^n \sum_{i=1}^m |a_{ij}|$$

Properties: $\forall A, B \in \mathbb{R}^{n \times m}, b \in \mathbb{R}^n$

$$\text{i)} |A+B| \leq |A| + |B|$$

$$\text{ii)} |AB| \leq |A||B| \quad (\text{HW})$$

$$\star \text{iii)} |Ab| \leq |A| \cdot |b| \quad (\text{will be used later})$$

$$\text{pf of iii)} Ab = \begin{pmatrix} \sum_{k=1}^n a_{1k} b_k \\ \vdots \\ \sum_{k=1}^n a_{nk} b_k \end{pmatrix}$$

$$|Ab| = \left| \sum_{k=1}^n a_{1k} b_k \right| + \dots + \left| \sum_{k=1}^n a_{nk} b_k \right| \leq \sum_{j=1}^n \sum_{k=1}^n |a_{jk}| |b_k| = \sum_{j=1}^n |a_{j*}| \sum_{k=1}^n |b_k|$$

$$= |A| |b| = \left(\sum_i \sum_j |a_{ij}| \right) \left(\sum_k |b_k| \right) \quad \text{more terms, } b/c$$

fixed i, j , LHS is $|a_{ij}| |b_j|$. RHS is $|a_{ij}| \left(\sum_k |b_k| \right)$.

Time to prove a priori estimate. \checkmark 1-norm

$$\text{WTS: } |\phi(t)| \leq M \quad \text{for some const } M \text{ indep of } t.$$

$$\text{Pf: } \dot{\phi}(t) = A(t)\phi(t) + g(t)$$

$$\phi(t) = y_0 + \int_{t_0}^t A(s)\phi(s) ds + \int_{t_0}^t g(s) ds \quad t \in [a, b]$$

$$\Rightarrow |\phi(t)| \leq |y_0| + \left| \int_{t_0}^t A(s)\phi(s) ds \right| + \left| \int_{t_0}^t g(s) ds \right|$$

$$\leq |y_0| + \int_{t_0}^t |A(s)| |\phi(s)| ds + \int_{t_0}^t |g(s)| ds$$

triangle Ineq (def of norm)

properties of matrix norm.

$|g(s)|, s \in [a, b]$ has max, Denote $\max_{s \in [a, b]} |g(s)| = K_0$ continuous fn on closed interval has max.

$$|y_0| + \left| \int_{t_0}^t g(s) ds \right| \leq |y_0| + K_0 (t - t_0)$$

$$\leq |y_0| + K_0(b-a) \triangleq K_1 \text{ const.}$$

Scalar (1-D)

$$|A(s)| \quad s \in [a, b] \quad \text{has max, } \max_{s \in [a, b]} |A(s)| = K_2$$

$$\int_a^t |A(s)| |\phi(s)| ds \leq K_2 \int_a^t |\phi(s)| ds$$

Hence, we have

$$|\phi(t)| \leq K_1 + K_2 \int_a^t |\phi(s)| ds.$$

By Gronwall Ineq.

$$|\phi(t)| \leq K_1 e^{K_2(t-a)} \leq K_1 e^{K_2(b-a)} \quad \text{const indep of time!} \quad \square.$$

Pmk. The result is true for $I = (0, b), [a, b], (-\infty, b), (a, +\infty), (-\infty, +\infty)$.

namely, needs $A(t), g(t)$ to be defined for I .

IVP admits global soln. if \forall open interval $I \subseteq \mathbb{R}$ s.t.

$$t_0 \in I, \quad \{x \in \mathbb{R}^n : (t, x) \in D\} \neq \emptyset \quad \forall t \in I.$$

\exists a fcn $y : I \rightarrow \mathbb{R}^n$ which is a soln of (IVP) on I .

Now know \exists 1 soln, next Q: what is the structure of soln sp. (is a vector sp) Start w/ homo linear ODE.

"Dictionary" of vector sp. / linear alg.

- linear subspace

Let S be a set of vectors in \mathbb{R}^n , then S is a linear subsp of \mathbb{R}^n if.

1. $\forall v_1, v_2 \in S, v_1 + v_2 \in S$

2. $\forall v \in S$ and any number c , vector $cv \in S$.

(closed under the formation of "+" and "c·")

- To Span

Vectors v_1, \dots, v_k span a linear subsp S of \mathbb{R}^n if every vector in S is a linear combo of vectors v_1, \dots, v_k .

- Linear dep. of vectors

v_1, \dots, v_k are linearly dep if $\exists c_1, \dots, c_k$ not all zero s.t. $c_1 v_1 + c_2 v_2 + \dots + c_k v_k = 0$.

linear indep. v_1, \dots, v_k are linearly indep. if the eq $c_1 v_1 + \dots + c_k v_k = 0$ has only soln $c_1 = \dots = c_k = 0$.

- Basis

Vectors v_1, \dots, v_k form a basis for a linear subspace S if

(i) they are indep.

(ii) they span S .

- Dim

Dim of linear subspace $S \subseteq \mathbb{R}^n$ is the number of vectors in a basis of S .

- Eval / E-vec

If \exists vector $\vec{v} \neq 0$ s.t. $A\vec{v} = \lambda\vec{v}$ for some scalar λ .

then λ is called eigenvalue, \vec{v} is called evec associated w/ eigenval. λ .

Note that eigenvalues are roots of characteristic poly $\det(A - \lambda I)$.

Thm. $A \in \mathbb{R}^{n \times n}, 1 \leq k \leq n$

- If v_1, \dots, v_k are e-vcs for a matrix A whose corresponding eval $\lambda_1, \dots, \lambda_k$ are distinct.
then $\{v_1, \dots, v_k\}$ is linearly indep.
- If $n \times n$ matrix A has n distinct eval, then any set of corresponding e-vectors $\{v_1, \dots, v_n\}$ is a basis of \mathbb{R}^n .

Note that $C(\mathbb{R})$ sp of pts fcn on \mathbb{R} is ∞ -dim. since $1, t, t^2, \dots$ is an infinite set in $C(\mathbb{R})$ and linearly indep. on any interval.

- det (Review \rightarrow HW)

If $A = [a_1 \dots a_n]$

column vectors are linearly indep. $\det A \neq 0$.

HW review def. AB , Ax .