

# Supplement for Nonlinear Systems and Control

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Work In Progress

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Some exercises are mentioned in the textbook's mainbody. So I organize some solutions here for reference. Only a little solutions is presented in this supplement. They are 1.1 3.24 5.6 .

# Math Review

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**Note:** One implication of this definition is that  $y = 0$  is a constant solution to a linear homogeneous differential equation, but not for the non-homogeneous case.

## 2.2 First Order Linear Differential Equations

Given a first order non-homogeneous linear differential equation

$$y' + p(t)y = f(t) \quad (5)$$

using variation of parameters the general solution is given by

$$y(t) = v(t)e^{P(t)} + Ae^{P(t)} \quad (6)$$

where  $v'(t) = e^{-P(t)}f(t)$  and  $P(t)$  is an antiderivative of  $-p(t)$

## 2.3 Numerical methods for ode

The closed-loop control system is usually written as

$$\dot{x} = f(t, x). \quad (7)$$

To verify the control performance, several numerical method is important.

- <https://www.math.hkust.edu.hk/~machas/numerical-methods-for-engineers.pdf>

### 2.3.1 Euler method – First Order

$$x_{n+1} = x_n + \Delta t f(t_n, x_n) \quad (8)$$

For small enough  $\Delta t$ , the numerical solution should converge to the exact solution of the ode, when such a solution exists. The Euler Method has a local error, that is, the error incurred over a single time step, of  $O(\Delta t^2)$ . The global error, however, comes from integrating out to a time  $T$ . If this integration takes  $N$  time steps, then the global error is the sum of  $N$  local errors. Since  $N = \frac{T}{\Delta t}$ , the global error is given by  $O(\Delta t)$ , and it is customary to call the Euler Method a first-order method.

### 2.3.2 Modified Euler, Heun's method, predictor-corrector method – Second Order

$$\begin{aligned} k_1 &= \Delta t f(t_n, x_n) & k_2 &= \Delta t f(t_n + \Delta t, x_n + k_1) \\ x_{n+1} &= x_n + \frac{1}{2}(k_1 + k_2) \end{aligned} \quad (9)$$

### 2.3.3 Runge-Kutta methods

First, we compute the Taylor series for  $x_{n+1}$  directly:

$$x_{n+1} = x(t_n + \Delta t) = x(t_n) + \Delta t \dot{x}(t_n) + \frac{1}{2}(\Delta t)^2 \ddot{x}(t_n) + O(\Delta t^3) \quad (10)$$

Now,  $\dot{x}(t_n) = f(t_n, x_n)$ . The second derivative is more tricky and requires partial derivatives. We have

$$\ddot{x}(t_n) = \left. \frac{d}{dt} f(t, x(t)) \right|_{t=t_n} = f_t(t_n, x_n) + \dot{x}(t_n) f_x(t_n, x_n) = f_t(t_n, x_n) + f(t_n, x_n) f_x(t_n, x_n) \quad (11)$$

Putting all the terms together, we obtain

$$x_{n+1} = x_n + \Delta t f(t_n, x_n) + \frac{1}{2}(\Delta t)^2 (f_t(t_n, x_n) + f(t_n, x_n) f_x(t_n, x_n)) + O(\Delta t^3) \quad (12)$$

Second, we compute the Taylor series for  $x_{n+1}$  from the Runge-Kutta formula. We start with

$$x_{n+1} = x_n + a\Delta t f(t_n, x_n) + b\Delta t f(t_n + \alpha\Delta t, x_n + \beta\Delta t f(t_n, x_n)) + O(\Delta t^3) \quad (13)$$

and the Taylor series that we need is

$$\begin{aligned}
& f(t_n + \alpha\Delta t, x_n + \beta\Delta t f(t_n, x_n)) \\
&= f(t_n, x_n) + \alpha\Delta t f_{t(t_n, x_n)} + \beta\Delta t f(t_n, x_n) f_{x(t_n, x_n)} + O(\Delta t^2)
\end{aligned} \tag{14}$$

The Taylor-series for  $x_{n+1}$  from the Runge-Kutta method is therefore given by

$$x_{n+1} = x_n + (a + b)\Delta t f(t_n, x_n) + (\Delta t)^2 (\alpha b f_{t(t_n, x_n)} + \beta b f(t_n, x_n) f_{x(t_n, x_n)}) + O(\Delta t^3) \tag{15}$$

Comparing (12) and (15), we find three constraints for the four constants.

$$a + b = 1, \alpha b = 1/2, \beta b = 1/2 \tag{16}$$

### 2.3.4 Second-order Runge-Kutta methods

The family of second-order Runge-Kutta methods that solve  $\dot{x} = f(t, x)$  is given by

$$\begin{aligned}
k_1 &= \Delta t f(t_n, x_n), \quad k_2 = \Delta t (f_n + \alpha\Delta t, x_n + \beta k_1), \\
x_{n+1} &= x_n + a k_1 + b k_2
\end{aligned} \tag{17}$$

where we have derived three constraints for the four constants  $\alpha, \beta, a$  and  $b$ :

$$a + b = 1, \alpha b = \frac{1}{2}, \beta b = \frac{1}{2} \tag{18}$$

The modified Euler method corresponds to  $\alpha = \beta = 1$  and  $a = b = \frac{1}{2}$ . The function  $f(t, x)$  is evaluated at the times  $t = t_n$  and  $t = t_n + \Delta t$ .

The midpoint method corresponds to  $\alpha = \beta = \frac{1}{2}$ ,  $a = 0$  and  $b = 1$ . In this method, the function  $f(t, x)$  is evaluated at the times  $t = t_n$  and  $t = t_n + \Delta t/2$  and we have

$$\begin{aligned}
k_1 &= \Delta t f(t_n, x_n) \quad k_2 = \Delta t f\left(t_n + \frac{1}{2}\Delta t, x_n + \frac{1}{2}k_1\right), \\
x_{n+1} &= x_n + k_2
\end{aligned} \tag{19}$$

### 2.3.5 Higher Order Runge-Kutta methods

Higher-order Runge-Kutta methods can also be derived, but require substantially more algebra. For example, the general form of the third-order method is given by

$$\begin{aligned}
k_1 &= \Delta t f(t_n, x_n), \\
k_2 &= \Delta t f(t_n + \alpha\Delta t, x_n + \beta k_1), \\
k_3 &= \Delta t f(t_n + \gamma\Delta t, x_n + \delta k_1 + \epsilon k_2), \\
x_{n+1} &= x_n + a k_1 + b k_2 + c k_3
\end{aligned} \tag{20}$$

with constraints  $\alpha, \beta, \gamma, \delta, \epsilon, a, b$  and  $c$ . The fourth-order method has stages  $k_1, k_2, k_3$  and  $k_4$ . The fifth-order method requires at least six stages. The table below gives the order of the method and the minimum number of stages required.

order	2	3	4	5	6	7	8
minimum #stage	2	3	4	6	7	9	11

Because the fifth-order method requires two more stages than the fourth-order method, the fourth-order method has found some popularity. The general fourth-order method with four stages has 13 constants and 11 constraints. A particularly simple fourth-order method that has been widely used in the past by physicists is given by

$$\begin{aligned}
 k_1 &= \Delta t f(t_n, x_n), & k_2 &= \Delta t f\left(t_n + \frac{1}{2}\Delta t, x_n + \frac{1}{2}k_1\right), \\
 k_3 &= \Delta t f\left(t_n + \frac{1}{2}\Delta t, x_n + \frac{1}{2}k_2\right), & k_4 &= \Delta t f(t_n + \Delta t, x_n + k_3);
 \end{aligned}
 \tag{21}$$

$$x_{n+1} = x_n + \frac{1}{6}(k_1 + 2k_2 + 2k_3 + k_4) \tag{22}$$

### 2.3.6 Adaptive Runge-Kutta methods

An adaptive ode solver automatically finds the best integration step-size  $\Delta t$  at each time step. The Dormand-Prince method, which is implemented in MATLAB's most widely used solver, `[t,y,te,ye,ie] = ode45(odefun,tspan,y0,options)`, determines the step size by comparing the results of fourth- and fifth- order Runge-Kutta methods. This solver requires six function evaluations per time step, and saves computational time by constructing both fourth- and fifth-order methods using the same function evaluation's.

### 2.3.7 stiff ODE

- <https://ww2.mathworks.cn/help/matlab/math/solve-stiff-odes.html?lang=en>

For some ODE problems, the step size taken by the solver is forced down to an unreasonably small level in comparison to the interval of integration, even in a region where the solution curve is smooth. **These step sizes can be so small that traversing a short time interval might require millions of evaluations.** This can lead to the solver failing the integration, but even if it succeeds it will take a very long time to do so.

Equations that cause this behavior in ODE solvers are said to be stiff. The problem that stiff ODEs pose is that explicit solvers (such as `ode45`) are untenably slow in achieving a solution. This is why `ode45` is classified as a nonstiff solver along with `ode23`, `ode78`, `ode89`, and `ode113`.

Solvers that are designed for stiff ODEs, known as stiff solvers, typically do more work per step. The pay-off is that they are able to take much larger steps, and have improved numerical stability compared to the nonstiff solvers.

### 3 Topological Space

- <http://staff.ustc.edu.cn/~wangzuoq/Courses/22S-Topology/>

#### 3.1 Metric Space

**Definition 3.1.1:** A **metric space**  $(M, d)$  consists of a set  $M$  and a mapping, called distance,  $d : M \times M \rightarrow \mathbb{R}$ , which satisfies the following:

1.  $0 \leq d(x, y) < \infty, \forall x, y \in M$
2.  $d(x, y) = 0$ , if and only if  $x = y$
3.  $d(x, y) = d(y, x)$
4. Triangle Inequality:  $d(x, z) \leq d(x, y) + d(y, z), x, y, z \in M$

**Definition 3.1.2:** Let  $V$  be a vector space, if there exists a mapping  $\|\cdot\| : V \rightarrow \mathbb{R}$ , satisfying

1.  $\|x\| \geq 0, \forall x \in V$  and  $\|x\| = 0$ , if and only if  $x = 0$
2.  $\|rx\| = |r|\|x\|, r \in \mathbb{R}, x \in V$
3. Triangle Inequality:  $\|x + y\| \leq \|x\| + \|y\|, x, y \in V$

Then  $(V, \|\cdot\|)$  is called a **normed space**, and  $\|x\|$  is called the norm of  $x$

**Definition 3.1.3:** Given a vector space  $V$ . If there exists a mapping  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{C}$ , satisfies the following requirements:

1.  $\langle x, x \rangle \geq 0, \forall x \in V$ . Moreover,  $\langle x, x \rangle = 0$ , if and only if  $x = 0$
2.  $\langle x, y \rangle = \overline{\langle y, x \rangle}, x, y \in V$
3.  $\langle ax + by, z \rangle = a\langle x, z \rangle + b\langle y, z \rangle, x, y, z \in V, a, b \in \mathbb{C}$

Then  $(V, \langle \cdot, \cdot \rangle)$  is called an **inner product space**, and  $\langle x, y \rangle$  is called the inner product of  $x, y$

**Proposition 3.1.1:**

1. Let  $V$  be an inner product space. Define a norm on  $V$  as  $\|x\| = \sqrt{\langle x, x \rangle}$ . Then  $V$  becomes a normed space. Such a norm is called the norm induced by the inner product
2. Let  $M$  be a normed space. Define a distance  $d$  as  $d(x, y) = \|x - y\|$ . Then  $M$  becomes a metric space.

**Definition 3.1.4:** A metric space  $M$  is complete if each Cauchy sequence  $\{x_n\}$  converges to a point  $x \in M$ .

#### 3.2 Topological Spaces

**Definition 3.2.1:** Given a set  $X$  and a set of its subsets  $\mathcal{T}$ .

1.  $(X, \mathcal{T})$  is called a topological space, if  $\mathcal{T}$  satisfies the following
  1.  $X \in \mathcal{T}, \emptyset \in \mathcal{T}$ ;
  2. If  $U_\lambda \in \mathcal{T}, \forall \lambda \in \Lambda \subset \mathbb{R}$ , then  $\cup_{\lambda \in \Lambda} U_\lambda \in \mathcal{T}$ ;
  3. If  $U_i \in \mathcal{T}, i = 1, \dots, n$ , then  $\cap_{i=1}^n U_i \in \mathcal{T}$
2. An element in  $U \in \mathcal{T}$  is called an open set. Its complement, denoted by  $U^c$  is called a closed set.
3. For a point  $x \in X$ , a subset  $N$  is called a neighborhood of  $x$  if there exists an open set  $U$  such that  $x \in U \subset N$
4. Let  $\mathcal{B} \subset \mathcal{T}$ .  $\mathcal{B}$  is a topological basis if any element  $U \in \mathcal{T}$  can be expressed as a union of some elements in  $\mathcal{B}$
5. A set of neighborhoods  $\mathcal{N} \subset \mathcal{T}$  is called a neighborhood basis of  $x$  if for any neighborhood  $N$  of  $x$ , there is an  $N_0 \in \mathcal{N}$ , such that  $N_0 \subset N$

### 3.3 Continuous Mapping

**Definition 3.3.1:** Let  $M, N$  be two topological spaces. A mapping  $\pi : M \rightarrow N$  is continuous, if one of the following two equivalent conditions holds:

- For any  $U \subset N$  open, its inverse image

$$\pi^{-1}(U) := \{x \in M \mid \pi(x) \in U\} \quad (23)$$

is open

- For any  $C \subset N$  closed, its inverse image  $\pi^{-1}(C)$  is closed

**Note:** the two conditions are equivalent.

**Definition 3.3.2:** Let  $M, N$  be two topological spaces.  $M$  and  $N$  are said to be homeomorphic (iff) if there exists a mapping  $\pi : M \rightarrow N$ , which is

1. one-to-one (iff injective),
2. onto (iff surjective)
3. and continuous (both  $\pi$  and  $\pi^{-1}$  are continuous).

$\pi$  is called a homeomorphism.

**Note:** If a mapping is both injective and surjective it is said to be bijective (iff).

**Definition 3.3.3:** Given a topological space  $M$ .

1. A set  $U \subset M$  is said to be clopen if it is both closed and open. A topological space (iff),  $M$ , is said to be **connected** if the only two clopen sets are  $M$  and  $\emptyset$
2. A continuous mapping  $\pi : I = [0, 1] \rightarrow M$  is called a path on  $M$ .  $M$  is said to be **pathwise (or arcwise) connected** if for any two points  $x, y \in M$  there exists a path,  $\pi$ , such that  $\pi(0) = x$  and  $\pi(1) = y$

**Tip on open and closed set:** As described by topologist James Munkres, unlike a door, “a set can be open, or closed, or both, or neither!” [https://en.wikipedia.org/wiki/Clopen\\_set](https://en.wikipedia.org/wiki/Clopen_set) A set is closed if its complement is open. But A set can be closed or open if its complement is closed.

[https://en.wikipedia.org/wiki/Open\\_set](https://en.wikipedia.org/wiki/Open_set)

A subset  $U$  of a metric space  $(M, d)$  is called open if, for any point  $x$  in  $U$ , there exists a real number



$\varepsilon$  such that any point  $y \in M$  satisfying  $d(x, y) < \varepsilon$  belongs to  $U$ . Equivalently,  $U$  is open if every point in  $U$  has a neighborhood contained in  $U$ .

**Note:**  $\mathbb{R}$  is connected and A pathwise connected space  $M$  is connected while the converse is incorrect.

**Definition 3.3.4:** A topological space  $M$  is said to be *locally connected* at  $x \in M$  if every neighborhood  $N_x$  of  $x$  contains a connected neighborhood  $U_x$ , i.e.,  $x \in U_x \subset N_x$ .  $M$  is said to be locally connected if it is locally connected at each  $x \in M$

local connectedness does not imply connectedness (hence no pathwise connectedness); conversely, pathwise connectedness (connectedness) does not imply local connectedness.

**Definition 3.3.5:** Let  $\{U_\lambda | \lambda \in \Lambda\}$  be a set of open sets in  $M$ . The set is called an open covering of  $M$  if

$$\bigcup_{\lambda \in \Lambda} U_\lambda \supset M. \quad (24)$$

$M$  is said to be a compact space if every open covering has a finite sub-covering, i.e., there exists a finite subset  $\{U_{\lambda_i} | i = 1, 2, \dots, k\}$  such that

$$\bigcup_{i=1}^k U_{\lambda_i} \supset M \quad (25)$$

From Calculus we know that with the conventional topology, a set,  $U \subset \mathbb{R}^n$  is compact, if and only if it is bounded and closed. Unfortunately, it is not true for general metric spaces.

**Definition 3.3.6:** In a topological space,  $M$ , a sequence  $\{x_k\}$  is said to converge to  $x$ , if for any neighborhood  $U \ni x$  there exists a positive integer  $N > 0$  such that when  $n > N$ ,  $x_n \in U$ .

1. Impose the discrete topology on  $\mathbb{R}^1$ . i.e., each point is an open set. Then  $x_k$  converges to nowhere. Because it can never get into any  $\{r\}$ , which is a neighborhood of  $r$

**Proposition 3.3.1:** Let  $M, N$  be two topological spaces.  $M$  is first countable.  $f : M \rightarrow N$  is continuous, if and only if for each  $x_k \rightarrow x$ ,  $f(x_k) \rightarrow f(x)$

**Definition 3.3.7:** A topological space is called sequentially compact if every sequence contains a convergent subsequence.

**Definition 3.3.8: (Bolzano-Weierstrass)** Let  $M$  be a first countable topological space. if  $M$  is compact, it is sequentially compact.

### 3.4 Quotient Spaces

**Definition 3.4.1:** Let  $S$  be any set and  $\sim$  be a relation between two elements of  $S$ .  $\sim$  is said to be an equivalent relation if

1.  $x \sim x$
2. If  $x \sim y$ , then  $y \sim x$
3. If  $x \sim y$ , and  $y \sim z$ , then  $x \sim z$

**Definition 3.4.2:** Let  $M$  be a topological space, " $\sim$ " an equivalent relation on  $M$

## 4 Differentiable Manifold

### 4.1 Structure of Manifolds

**Definition 4.1.1:** Let  $(M, \mathcal{T})$  be a second countable,  $T_2$ (Hausdorff) topological space.  $M$  is called an  $n$  dimensional topological manifold if there exists a subset  $\mathcal{A} = \{A_\lambda \mid \lambda \in \Lambda\} \subset \mathcal{T}$ , such that

1.  $\bigcup_{\lambda \in \Lambda} A_\lambda \supset M$ ;
2. For each  $U \in \mathcal{A}$  there exists a homeomorphism  $\varphi : U \rightarrow \varphi(U) \subset \mathbb{R}^n$ , which is called a coordinate chart, denoted by  $(U, \varphi)$ .
3. Moreover, if for two coordinate charts:  $(U, \varphi)$  and  $(V, \Psi)$ , if  $U \cap V$  is not empty, then both  $\Psi \circ \varphi^{-1} : \varphi(U \cap V) \rightarrow \Psi(U \cap V)$  and  $\varphi \circ \Psi^{-1} : \Psi(U \cap V) \rightarrow \varphi(U \cap V)$  are  $C^r(C^\infty, C^\omega)$ . such two coordinate charts are said to be consistent.
4. If a coordinate chart,  $W$ , is consistent with all charts in  $\mathcal{A}$ , then  $W \in \mathcal{A}$ .

Then  $(M, \mathcal{T})$  is called a  $C^r(C^\infty, \text{analytic, respectively})$  differentiable manifold.

**Definition 4.1.2:** Let  $M, N$  be two  $C^r$  manifolds with dimensions  $m, n$  respectively.  $F : M \rightarrow N$  is called a  $C^r$  mapping, if for each  $x \in M$  and  $y = F(x) \in N$  there are coordinate charts  $(U, \varphi)$  about  $x$  and  $(V, \psi)$  about  $y$ , such that

$$\tilde{F} = \psi \circ F \circ \varphi^{-1} \quad (26)$$

### 4.2 Fiber Bundle

### 4.3 Vector Field

### 4.4 One Parameter Group

### 4.5 Lie Algebra of Vector Fields

### 4.6 Co-tangent Space

### 4.7 Lie Derivatives

### 4.8 Frobenius' Theory

### 4.9 Lie Series, Chow's Theorem

### 4.10 Tensor Field

### 4.11 Riemannian Geometry

### 4.12 Symplectic Geometry

# **Solutions**

MLCT  
(Working In Progress)

## chapter 1 Introduction

**Exercise 1.1:** A mathematical model that describes a wide variety of physical nonlinear systems is the  $n$ th-order differential equation

$$y^{(n)} = g(t, y, \dot{y}, \dots, y^{(n-1)}, u) \quad (27)$$

where  $u$  and  $y$  are scalar variables. With  $u$  as input and  $y$  as output, find a state model.

Solution:

Let  $x_1 = y, x_2 = \dot{y}, \dots, x_n = y^{(n-1)}$

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_{n-1} &= x_n \\ \dot{x}_n &= g(t, x_1, \dots, x_n, u) \\ y &= x_1 \end{aligned} \quad (28)$$

## chapter 2 Second Order Systems

## chapter 3 Fundamental Properties

**Exercise 3.24 :** Let  $V : \mathbb{R} \times \mathbb{R}^n \rightarrow \mathbb{R}$  be continuously differentiable. Suppose that  $V(t, 0) = 0$  for all  $t \geq 0$  and

$$V(t, x) \geq c_1 \|x\|^2; \left\| \frac{\partial V}{\partial x}(t, x) \right\| \leq c_4 \|x\|, \forall (t, x) \in [0, \infty) \times D \quad (29)$$

where  $c_1$  and  $c_4$  are positive constants and  $D \subset \mathbb{R}^n$  is a convex domain that contains the origin  $x = 0$

1. Show that  $V(t, x) \leq \frac{1}{2}c_4 \|x\|^2$  for all  $x \in D$ .  
Hint: Use the representation  $V(t, x) = \int_0^1 \frac{\partial V}{\partial x}(t, \sigma x) d\sigma x$
2. Show that the constants  $c_1$  and  $c_4$  must satisfy  $2c_1 \leq c_4$
3. Show that  $W(t, x) = \sqrt{V(t, x)}$  satisfies the Lipschitz condition

$$|W(t, x_2) - W(t, x_1)| \leq \frac{c_4}{2\sqrt{c_1}} \|x_2 - x_1\|, \forall t \geq 0, \forall x_1, x_2 \in D \quad (30)$$

**Solution to 1**

$$V(t, x) = \int_0^1 \frac{\partial V}{\partial x}(t, \sigma x) d\sigma x \leq \int_0^1 \left\| \frac{\partial V}{\partial x}(t, \sigma x) \right\| \|x\| d\sigma \leq \int_0^1 c_4 \sigma d\sigma \|x\|^2 \leq \frac{1}{2}c_4 \|x\|^2 \quad (31)$$

**Solution to 2**

Since

$$c_1 \|x\|^2 \leq V(t, x) \leq \frac{1}{2}c_4 \|x\|^2, \forall x \in D \quad (32)$$

we must have  $c_1 \leq \frac{1}{2}c_4$

**Solution to 3**

Consider two points  $x_1$  and  $x_2$  such that  $\alpha x_1 + (1 - \alpha)x_2 \neq 0$  for all  $0 \leq \alpha \leq 1$ ; that is, the origin does not lie on the line connecting  $x_1$  and  $x_2$ . The Jacobian  $[\partial W / \partial x]$  is defined for every  $x = \alpha x_1 + (1 - \alpha)x_2$  and given by

$$\frac{\partial W}{\partial x}(t, x) = \frac{1}{2\sqrt{V(t, x)}} \frac{\partial V}{\partial x}(t, x) \quad (33)$$

By the mean value theorem, there is  $\alpha^* \in (0, 1)$  such that, with  $z = \alpha^* x_1 + (1 - \alpha^*)x_2$

$$W(t, x_2) - W(t, x_1) = \frac{\partial W}{\partial x}(t, z)(x_2 - x_1) = \frac{1}{2\sqrt{V(t, z)}} \frac{\partial V}{\partial x}(t, z)(x_2 - x_1) \quad (34)$$

Hence

$$|W(t, x_2) - W(t, x_1)| \leq \frac{1}{2\sqrt{c_1}\|z\|} \quad (35)$$

Consider now the case when the origin lies on the line connecting  $x_1$  and  $x_2$ ; that is,  $0 = \alpha_0 x_1 + (1 - \alpha_0)x_2$  for some  $\alpha_0 \in [0, 1]$ . We have

$$\begin{aligned} |W(t, x_2) - W(t, 0)| &= |W(t, x_2)| = \sqrt{V(t, x_2)} \leq \sqrt{\frac{c_4}{2}}\|x_2\| \\ |W(t, x_1) - W(t, 0)| &= |W(t, x_1)| = \sqrt{V(t, x_1)} \leq \sqrt{\frac{c_4}{2}}\|x_1\| \end{aligned} \quad (36)$$

$$|W(t, x_2) - W(t, x_1)| = |W(t, x_2) - W(t, 0) + W(t, 0) - W(t, x_1)| \leq \sqrt{\frac{c_4}{2}}(\|x_1\| + \|x_2\|)$$

Since the origin lies on the line connecting  $x_1$  and  $x_2$ , we have  $\|x_2\| + \|x_1\| = \|x_2 - x_1\|$ . We also have  $1 \leq \sqrt{c_4/2c_1}$ . Therefore,

$$|W(t, x_2) - W(t, x_1)| \leq \frac{c_4}{2\sqrt{c_1}}\|x_2 - x_1\| \quad (37)$$

## chapter 4 Lyapunov Stability

## chapter 5 Input-Output Stability

**Exercise 5.6:** Verify that  $D_+ W(t)$  satisfies (38)(5.12 in textbook) when  $V(t, x(t)) = 0$ .

$$D_+ W \leq \frac{c_4 L}{2\sqrt{c_1}}\|u(t)\| \quad (38)$$

Hint: Using Exercise 3.24, show that

$$V(t + h, x(t + h)) \leq c_4 h^2 L^2 \|u\|^2 / 2 + ho(h) \quad (39)$$

where  $\frac{o(h)}{h} \rightarrow 0$  as  $h \rightarrow 0$ . Then apply  $c_4 \geq 2c_1$

From textbook, the system is

$$\begin{aligned} \dot{x} &= f(t, x, u), x(0) = x_0 \\ y &= h(t, x, u) \end{aligned} \quad (40)$$

From  $V(t, x(t)) = 0$  and (32), we have  $x(t) = 0$

Let  $V(t, x(t)) = 0$ .

$$\begin{aligned} D_+ W &= \limsup_{h \rightarrow 0^+} \frac{1}{h} [W(t+h, x(t+h)) - W(t, x(t))] \\ &= \limsup_{h \rightarrow 0^+} \frac{1}{h} \sqrt{V(t+h, x(t+h))} \end{aligned} \quad (41)$$

From (32), We have

$$V(t+h, x(t+h)) \leq \frac{c_4}{2} \|x(t+h)\|^2 \quad (42)$$

From textbook 5.9, we have

$$\|f(t, x(t), u) - f(t, x(t), 0)\| \leq L\|u\| \quad (43)$$

Use Taylor Series:

$$\begin{aligned} x(t+h) &= f(t, x, u)h + o(h) \\ \Rightarrow \|x(t+h)\|^2 &\leq (\|f(t, x, u)\|h + \|o(h)\|)^2 \end{aligned} \quad (44)$$

$$\frac{1}{h^2} V(t+h, x(t+h)) \leq \frac{c_4}{2} \left( \frac{\|x(t+h)\|}{h} \right)^2 \leq \frac{c_4}{2} \left( \|f(t, x, u)\| + \frac{\|o(h)\|}{h} \right)^2 \quad (45)$$

$$\limsup_{h \rightarrow 0^+} \frac{1}{h} \sqrt{V(t+h, x(t+h))} \leq \sqrt{\frac{c_4}{2}} \|f(t, x, u)\| \leq \sqrt{\frac{c_4}{2}} L\|u\| \quad (46)$$

since  $\sqrt{c_4/(2c_1)} \geq 1$ . Thus

$$D_+ W \leq \sqrt{\frac{c_4}{2}} L\|u\| \sqrt{c_4/(2c_1)} = \frac{c_4 L}{2\sqrt{c_1}} \|u(t)\| \quad (47)$$

which agrees with the right hand side of (38)

## chapter 6 Passivity

## chapter 7 Frequency Domain analysis of Feedback Systems

## chapter 8 Advanced Stability Analysis

## chapter 9 Stability of Perturbed Systems

## chapter 10 Perturbation Theory and Averaging

## chapter 11 Singular Perturbations

## chapter 12 Feedback Control

## chapter 13 Feedback Linearization

## chapter 14 Nonlinear Design Tools