## Controls Notes

#### M516

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This document is an ongoing collection of symbols, theorems, tools, and terms I have found useful for studying control theory, available as PDF, HTML, and LATEX source code.

## Variables and Symbols

x • state vector (state space) (W)

Type:  $\mathbb{R}^n$ 

H • Hamiltonian matrix

 $\bullet \ \ \mathsf{Hamiltonian} \ (\mathsf{Hamiltonian} \ \mathsf{mechanics}) \ (\mathsf{W})$ 

type:  $\mathbb{R}^n o \mathbb{R}$ 

- Assuming discrete time linear system:

$$H_k = L(x_k, u_k, k) + p_{k+1}^T f(x_k, u_k, k)$$

 $\mathcal{L}$  • the Lagrangian

type:  $\mathbb{R}^n o \mathbb{R}$ 

A ball, defined as

$$B\left(x_{0},\epsilon\right)=\left\{ x\in\mathbb{R}^{n}:\left\Vert x-x_{0}\right\Vert \leq\epsilon\right\}$$

C •  $C^1$  = Continuously differentiable, i.e. the first derivative is continuous.

•  $C^n$  = The  $n^{\text{th}}$  derivative is continuous.

ullet C: the set of all complex numbers a+bi where a and b are real and  $i=\sqrt{-1}$ 

 $e_{\#}$  • The #th unit vector

$$e_{1} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \qquad e_{2} = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \qquad e_{3} = \begin{pmatrix} 0 \\ 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \qquad \dots$$

 $\nabla$  The del operator, which represents one of many long but similar operators on a vector field  $v \in \mathbb{R}^n$ .

•  $\nabla f$ : Gradient of a function  $f: \mathbb{R}^n \to \mathbb{R}$ , returning an n-dimensional vector. (W) This vector points in the direction of the greatest increase, and its magnitude is the slope.

For example, a mountain climber could approximate the shape of a convex mountain as a function  $f_{mountain}$  that computes the altitude given some latitude and longitude (assuming a very small mountain very far from the poles). In other words,  $f_{mountain}: \mathbb{R}^2 \to \mathbb{R}$ . The climber could know which direction to climb to summit the peak: it's the direction  $\nabla f$ , and the grade or slope of the mountain is  $|\nabla f|$ 

Note that if n = 1,  $\nabla f$  is the standard derivative of f. Formally speaking:

$$\nabla f = \sum_{i=1}^{n} \frac{\partial f}{\partial x_i} e_i$$

- ullet  $\nabla \cdot \vec{v}$ : The divergence of a vector field  $\vec{v}$
- $\nabla imes \vec{v}$ : The curl of a vector field  $\vec{v}$
- $\Delta f$ : the Laplace operator on a function  $f: \mathbb{R}^n \to \mathbb{R}$ , equivalent to the divergence of the gradient of f, i.e.

$$\delta f = \nabla^2 f = \nabla \cdot \nabla f$$

*J* • Cost to go function type:  $\mathbb{R}^n \to \mathbb{R}$ 

p • Lagrange multiplier (W)

 $\langle \mathsf{expr} \rangle$  • Lie bracket notation (W)  $\langle a, b \rangle = b^T a$ 

 $\|expr\|$  • Vector norm (TODO)

Matrix norm (TODO)
 Topodianal name (TODO)

• Functional norm (TODO)

• Norm of a system y(u,t) = h(u(t)), where y is an n-dimensional the output of the system, u is an m-dimensional control vector (TODO)

## **Named Theorems and Conditions**

Poincaré-Bendixson theo TODO orem

## Small-gain Theorem

Given

- ullet  $H_1$ : an Input-Output System with input  $e_1$  and output  $y_1$  that is finite-gain  $\mathcal{L}_p$ -stable
- $H_2$ : an Input-Output System with input  $e_2$  and output  $y_2$  that is finite-gain  $\mathcal{L}_p$ -stable
- $y_1 = H_1 e_1$
- $\bullet \ y_2 = H_2 e_2$
- $e_1 = u_1 y_2$
- $e_2 = u_2 y_1$

By the definition of finite-gain  $\mathcal{L}_v$ -stable,

$$||y_{1_{\tau}}||_{\mathcal{L}_{v}} \leq \gamma_{1}||e_{1_{\tau}}||_{\mathcal{L}_{v}} + \beta_{1}$$

(The  $\mathcal{L}_p$  norm of the  $y_1$  is truncated by  $\tau$ , i.e. the system response is zero when  $t > \tau$ . This is less than or equal to The  $\mathcal{L}_p$  norm of the  $e_1$  truncated by  $t < \tau$ , multiplied by some gain value  $\gamma_1$ , plus some bias  $\beta_1$ )

As long as a system does not have a finite escape time, we can compute the  $\mathcal{L}_p$  norm of the system.

Likewise,

$$||y_{2\tau}||_{\mathcal{L}_p} \le \gamma_2 ||e_{2\tau}||_{\mathcal{L}_p} + \beta_2$$

The Small-gain Theorem tells us,

$$\left\| \begin{matrix} y_{1_{\tau}} \\ y_{2_{\tau}} \end{matrix} \right\|_{\mathcal{L}_{p}} \leq \frac{1}{1 - \gamma_{1} \gamma_{2}} \left( \|u_{1_{\tau}}\|_{\mathcal{L}_{p}} + \gamma_{2} \|u_{2_{\tau}}\|_{\mathcal{L}_{p}} + \gamma_{2} \beta_{1} + \beta_{2} \right) = \gamma_{3} \left( \text{some } \mathcal{L}_{p}\text{-stable system} \right) \quad (1)$$

Therefore, if  $\gamma_1$  and  $\gamma_2$  are less than one, the feedback connection is input/output stable (finite-gain  $\mathcal{L}_p$ -stable)

## **Terms**

Classes Systems f Given a dynamic system

• Dynamic system

$$\dot{x} = f(x, u, t)$$

• Time-invariant system is a dynamic system

$$\dot{x} = f(x, u)$$

• Autonomous system is a dynamic system

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

• Linear system (W) is a dynamic system

$$\dot{x} = f(x, u, t) = A(t)x + B(t)u$$

• Linear time-invariant system is a linear system and a time-invariant system

$$\dot{x} = f(x, u) = Ax + Bu$$

Lipschitz Continuity Lipschitz continuous functions are continuous and differentiable almost anywhere in a domain.

(W,) (UC

Given a domain D and a function  $f:D\to\mathbb{R}$ ,  $D\in\mathbb{R}^n$ ,

Berkley)

f is Lipschitz continuous if  $\exists L > 0$  such that  $|f(x) - f(y)| < L||(x - y)||\forall x, y \in D$ 

Hessian (W), (Kahn Academy), (Wolfram) • A  $n \times n$  matrix of all 2nd order partial derivatives of some function  $f : \mathbb{R}^n \to \mathbb{R}$ 

$$Hf(\vec{x}) = f''(\vec{x}) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_1 \partial x_3} & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2 \partial x_2} & \frac{\partial^2 f}{\partial x_2 \partial x_3} & \cdots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \frac{\partial^2 f}{\partial x_3 \partial x_1} & \frac{\partial^2 f}{\partial x_3 \partial x_2} & \frac{\partial^2 f}{\partial x_3 \partial x_3} & \cdots & \frac{\partial^2 f}{\partial x_3 \partial x_n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \frac{\partial^2 f}{\partial x_n \partial x_3} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_n} \end{pmatrix}$$

$$(2)$$

• The determinant of a Hessian matrix

definite (W) Warning: this definition does not appear to be common outside of controls

Given a real-valued, continuously differentiable function  $V(x): \mathbb{R} \to \mathbb{R}$  V(x) can be classified as

• (globally) positive semidefinite if

$$V(x) \ge 0 \qquad \forall x \in \mathbb{R}$$

(v is greater than or equal to 0 regardless of x)

• (globally) positive definite if positive semidefinite AND

$$V(x) = 0 \iff x = 0$$

(V(x) is zero if and only if x is zero)

• (globally) negative semidefinite if

$$V(x) \le 0 \quad \forall x \in \mathbb{R}$$

(v is less than or equal to 0 regardless of x)

• (globally) negative definite if negative semidefinite AND

$$V(x) = 0 \iff x = 0$$

(V(x) is zero if and only if x is zero)

• locally positive definite (l.p.d) if

$$V(x) \ge 0 \quad \forall x \in N$$

where N is a small open neighborhood containing  $\vec{0}$ 

(v is greater than or equal to 0 regardless of x in some small open neighborhood N that contains the zero vector)

#### AND

$$V(x) = 0 \iff x = 0$$

(V(x) is zero if and only if x is zero)

Note that the criteria for a function to be locally positive definite are similar, but more relaxed than, those for globally positive definite functions.

• positive definite on some domain  $D \in \mathbb{R}^n$  if we only care if the conditions for positive definite functions hold for all x in D.

Stability (MIT)

Given an autonomous system

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

and some open connected region  ${\cal D}$  containing  $\vec{0}$ 

Stability is usually used to describe trajectories around the origin of a system.

## Stability

The equilibrium point x=0 is stable if  $\forall \epsilon>0,\ \exists \delta(\epsilon)>0$  such that  $\|x(0)\|<\delta \implies \|x(t)\|<\epsilon$ 

## - In the sense of Lyapunov

If there exists a scalar, continuously-differentiable function V(x) such that

$$V(x) > 0 \quad \forall x \in \mathcal{D} \setminus \left\{ \vec{0} \right\}, \qquad V(\vec{0}) = 0$$

(V(x) is a locally positive definite function)

#### AND

$$\dot{V}(x) = \frac{\partial V}{\partial x} f(x) \le 0 \quad \forall x \in \mathcal{D} \setminus \left\{ \vec{0} \right\}, \qquad V(\vec{0}) = 0$$

 $(\dot{V}(x))$  is a locally negative semidefinite function)

then the origin is stable in the sense of Lyapunov, and V(x) is a Lyapunov function of f(x).

## Instability

The equilibrium point x = 0 is unstable if it is not stable

## · Asymptotic stability

The equilibrium point x=0 is asymptotically stable if it is stable and  $\exists \delta_1$  such that  $\|x(0)\| < \delta_1 \implies \lim_{t \to \infty} x(t) = 0$ 

# $t \to \infty$ – In the sense of Lyapunov

The origin is asymptotically stable in the sense of Lyapunov if stable AND

$$\dot{V}(x) = \frac{\partial V}{\partial x} f(x) < 0 \quad \forall x \in \mathcal{D} \setminus \left\{ \vec{0} \right\}$$

 $(\dot{V}(x))$  is a locally negative definite function)

## Exponential stability

#### - In the sense of Lyapunov

The origin is exponentially stable in the sense of Lyapunov if stable AND

$$\dot{V}(x) = \frac{\partial V}{\partial x} f(x) \le -\alpha V(x) \quad \forall x \in \mathcal{D} \setminus \left\{ \vec{0} \right\}$$

## • Uniform stability

The equilibrium point x=0 is uniformly stable if it is stable and, for each epsilon>0, there exists a  $\delta(\epsilon)>0$ , independent of  $t_0$ .

# Stability (continued)

- Global asymptotic stability
  - In the sense of Lyapunov

If the origin is globally asymptotically stable in the sense of Lyapunov if is asymptotically stable and

$$||x|| \to \infty \implies V(x) \to \infty$$

(V(x) is radially unbounded)

- L-stability (TODO)
- I/O L-stability (TODO)
- Finite-gain L-stability (TODO)
- Small-signal I/O L-stability (TODO)
- Small-signal finite-gain L-stability (TODO)

## Class function

A continuous scalar function on  $\mathbf{R}^+$  is

- class  $\kappa$  if it is:
  - zero at zero
  - strictly increasing
  - continuous
- class  $\kappa_{\infty}$  if it is:
  - zero at zero
  - strictly increasing
  - continuous
  - $\infty$  at  $\infty$

Radially Unbounded function

A function V(x) is radially unbounded if

$$||x|| \to \infty \implies ||V(x)|| \to \infty$$

sup (supremum) Like a maximum of a functions, but includes limits that aren't necessarily a part of the domain of the function. (TODO)

#### Hurwitz

Hurwitz (polynomial):

A polynomial whose roots that are all in the left-half plane. (In other words, the real part of every root is strictly negative)

• Hurwitz (matrix) (W):

A square matrix whose characteristic polynomial is Hurwitz, meaning all eigenvalues are in the left-half plane. (In other words, the real part of every eigenvalue is strictly negative)

 Routh-Hurwitz stability criterion (IEEE): TODO

Any hyperbolic fixed point (or equilibrium point) of a continuous dynamical system is locally asymptotically stable if and only if the Jacobian of the dynamical system is Hurwitz stable at the fixed point.

A system is stable if its control matrix is a Hurwitz matrix.

The negative real components of the eigenvalues of the matrix represent negative feedback. Similarly, a system is inherently unstable if any of the eigenvalues have positive real components, representing positive feedback.

## Zero-state observable

A time-invariant system of the form

$$\begin{cases} \dot{x} = f(x, u) \\ y = h(x, u) \end{cases}$$

is zero-state observable if

$$\begin{cases} y \equiv 0 \\ u \equiv 0 \end{cases} \implies x \equiv 0$$

In other words, when u=0, any nonzero state behavior will be observed at the output  $(y\neq 0)$ 

### Sets

#### Invariant Set

A set of vectors M is invariant with respect to  $\dot{x} = f(x)$  if

$$x(0) \in M \implies x(t) \in M, \quad \forall t \in \mathbb{R}$$

(if a solution belongs to M at some time instant, then it belongs to M for all future and past time)

## • Positively Invariant Set

A set of vectors M is positively invariant with respect to  $\dot{x} = f(x)$  if

$$x(0) \in M \implies x(t) \in M, \quad \forall t \ge 0$$

(if a solution belongs to M at some time instant, then it belongs to M for all future time)

## • Open Set

A set  $D \subset \mathbb{R}^n$  (D, which is a set of real vectors) is an **open set** if

$$\forall x \subset D, \quad \exists \epsilon > 0 \quad \text{ such that } \quad B(x, \epsilon) \subset D$$

(for all vectors x in the domain D, there exists a real scalar  $\epsilon$  such that we can create a ball around x with radius  $\epsilon$ , and that whole ball is in D)

#### Closed Set

A set  $D \subset \mathbb{R}^n$  (D, which is a set of real vectors) is a **closed set** if

$$\mathbb{R}^n \setminus D$$
 is an open set

(everywhere outside of D is open)

### • Bounded Set

A set  $D \subset \mathbb{R}^n$  (D, which is a set of real vectors) is a **bounded set** if

$$\exists \epsilon > 0 \quad \text{ such that } \quad D \subset B\left(0,\epsilon\right)$$

(D fits in a ball with a finite, constant radius  $\epsilon$ )

## Compact Set

A set  $D \subset \mathbb{R}^n$  (D, which is a set of real vectors) is a **compact set** if it is closed and bounded.

Passivity

For a system y = h(u,t),  $h: \mathbb{R}^m \times [0,\infty) \to \mathbb{R}^n$  (output state y (an n-dimensional vector) is a function of the input state u (an m-dimensional vector) and time t)

• Invariant Set

A set of vectors M is invariant with respect to  $\dot{x} = f(x)$  if

$$x(0) \in M \implies x(t) \in M, \quad \forall t \in \mathbb{R}$$

(if a solution belongs to M at some time instant, then it belongs to M for all future and past time)

Adjoint

• The {adjoint or Hermitian transpose} of a matrix A (Wolfram) is its conjugate transpose, denoted as A',  $A^*$ ,  $A^H$ , or  $A^{\dagger}$  i.e.

$$A^H = \overline{A}^T$$

Interesting properties of ajoint matrices:

$$-A^H = \overline{A}^{\dot{T}} = \overline{A}^T$$

- If a matrix is its own conjugate transpose, that matrix is called **self-adjoint** or **Hermetian**
- If A is a real matrix,  $A^H = A^T$

Warning: In some older literature, the "adjoint of a matrix" may mean the adjunct matrix of a square matrix (W)

- The adjoint representation of a vector space (Wolfram) (TODO)
- The adjoint equation (W) (TODO)