

Lecture – 35

*Stability Analysis of Nonlinear Systems
Using Lyapunov Theory – III*

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Outline

- Review of Lyapunov Theorems
- LaSalle's Theorem
- Domain of Attraction

Review of Lyapunov Theorems

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Definitions

System Dynamics

$$\dot{X} = f(X) \quad f : D \rightarrow \mathbb{R}^n \text{ (a locally Lipschitz map)}$$

D : an open and connected subset of \mathbb{R}^n

Equilibrium Point (X_e)

$$\dot{X}_e = f(X_e) = 0$$

Definitions

Stable Equilibrium

X_e is stable, provided for each $\varepsilon > 0$, $\exists \delta(\varepsilon) > 0$:

$$\|X(0) - X_e\| < \delta(\varepsilon) \Rightarrow \|X(t) - X_e\| < \varepsilon \quad \forall t \geq t_0$$

Unstable Equilibrium

If the above condition is not satisfied, then the equilibrium point is said to be unstable

Definitions

Convergent Equilibrium

$$\text{If } \exists \delta : \|X(0) - X_e\| < \delta \implies \lim_{t \rightarrow \infty} X(t) = X_e$$

Asymptotically Stable

If an equilibrium point is both stable and convergent, then it is said to be asymptotically stable.

Definitions

Exponentially Stable

$$\exists \alpha, \lambda > 0: \quad \|X(t) - X_e\| \leq \alpha \|X(0) - X_e\| e^{-\lambda t} \quad \forall t > 0$$

whenever $\|X(0) - X_e\| < \delta$

Convention

The equilibrium point $X_e = 0$

(without loss of generality)

Lyapunov Stability Theorems

Theorem – 1 (Stability)

Let $X = 0$ be an equilibrium point of $\dot{X} = f(X)$, $f : D \rightarrow \mathbb{R}^n$.

Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable function such that:

- (i) $V(0) = 0$
- (ii) $V(X) > 0$, in $D - \{0\}$
- (iii) $\dot{V}(X) \leq 0$, in $D - \{0\}$

Then $X = 0$ is "stable".

Lyapunov Stability Theorems

Theorem – 2 (Asymptotically stable)

Let $X = 0$ be an equilibrium point of $\dot{X} = f(X)$, $f : D \rightarrow \mathbb{R}^n$.

Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable function such that:

- (i) $V(0) = 0$
- (ii) $V(X) > 0$, in $D - \{0\}$
- (iii) $\dot{V}(X) < 0$, in $D - \{0\}$

Then $X = 0$ is "asymptotically stable".

Lyapunov Stability Theorems

Theorem – 3 (Globally asymptotically stable)

Let $X = 0$ be an equilibrium point of $\dot{X} = f(X)$, $f : D \rightarrow \mathbb{R}^n$.

Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable function such that:

- (i) $V(0) = 0$
- (ii) $V(X) > 0$, in $D - \{0\}$
- (iii) $V(X)$ is "radially unbounded"
- (iv) $\dot{V}(X) < 0$, in $D - \{0\}$

Then $X = 0$ is "globally asymptotically stable".

Lyapunov Stability Theorems

Theorem – 3 (Exponentially stable)

Suppose all conditions for asymptotic stability are satisfied.

In addition to it, suppose \exists constants k_1, k_2, k_3, p :

$$(i) \quad k_1 \|X\|^p \leq V(X) \leq k_2 \|X\|^p$$

$$(ii) \quad \dot{V}(X) \leq -k_3 \|X\|^p$$

Then the origin $X = 0$ is "exponentially stable".

Moreover, if these conditions hold globally, then the origin $X = 0$ is "globally exponentially stable".

Analysis of Linear Time Invariant System

System dynamics: $\dot{X} = AX, \quad A \in \mathbb{R}^{n \times n}$

Lyapunov function: $V(X) = X^T P X, \quad P > 0$ (pdf)

Derivative analysis:
$$\begin{aligned}\dot{V} &= \dot{X}^T P X + X^T P \dot{X} \\ &= X^T A^T P X + X^T P A X \\ &= X^T (A^T P + P A) X\end{aligned}$$

Analysis of Linear Time Invariant System

For stability, we aim for $\dot{V} = -X^T Q X$ ($Q > 0$)

By comparing $X^T (A^T P + P A) X = -X^T Q X$

For a non-trivial solution

$$PA + A^T P + Q = 0$$

(Lyapunov Equation)

Analysis of Linear Time Invariant Systems

- Choose an arbitrary symmetric positive definite matrix Q ($Q = I$)
- Solve for the matrix P from the *Lyapunov equation* and verify whether it is positive definite
- Result: If P is positive definite, then $\dot{V}(X) < 0$ and hence the origin is “asymptotically stable”.

Lyapunov's Indirect Theorem

Let the linearized system about $X = 0$ be $\Delta\dot{X} = A(\Delta X)$.

The theorem says that if all the eigenvalues λ_i ($i = 1, \dots, n$) of the matrix A satisfy $\text{Re}(\lambda_i) < 0$ (i.e. the linearized system is exponentially stable), then for the nonlinear system the origin is locally exponentially stable.

Instability theorem

Consider the autonomous dynamical system and assume $X=0$ is an equilibrium point. Let $V : D \rightarrow \mathbb{R}$ have the following properties:

(i) $V(0) = 0$

(ii) $\exists X_0 \in \mathbb{R}^n$, arbitrarily close to $X = 0$, such that $V(X_0) > 0$

(iii) $\dot{V} > 0 \quad \forall X \in U$, where the set U is defined as follows

$$U = \{X \in D : \|X\| \leq \varepsilon \text{ and } V(X) > 0\}$$

Under these conditions, $X=0$ is unstable

Construction of Lyapunov Functions

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Variable Gradient Method:

* Select a $\nabla V = \frac{\partial V}{\partial X} = g(X)$ that contains some adjustable parameters

* Then $dV(X) = \left(\frac{\partial V}{\partial X}\right)^T dX$

$$\int_{\tilde{X}=0}^X dV(\tilde{X}) = \int_{\tilde{X}=0}^X \left(\frac{\partial V}{\partial \tilde{X}}\right)^T d\tilde{X}$$

$$V(X) - V(0) = \int_{\tilde{X}=0}^X g(\tilde{X}) d\tilde{X}$$

Note:

To recover a unique V ,
 $\nabla V = g(X)$ must satisfy
the "Curl Condition":

$$i.e. \quad \frac{\partial g_i}{\partial x_j} = \frac{\partial g_j}{\partial x_i}$$

However, note that the integral value depends on the initial and final states (not on the path followed). Hence, integration can be conveniently done along each of the co-ordinate axes in turn; i.e.

Variable Gradient Method:

$$\begin{aligned} V(X) &= \int_0^{x_1} g_1(\tilde{x}_1, 0, \dots, 0) d\tilde{x}_1 \\ &+ \int_0^{x_2} g_2(x_1, \tilde{x}_2, 0, \dots, 0) d\tilde{x}_2 \\ &\vdots \\ &+ \int_0^{x_n} g_n(x_1, \dots, x_{n-1}, \tilde{x}_n) d\tilde{x}_n \end{aligned}$$

Note: The free parameter of $g(X)$ are constrained to satisfy the symmetric condition, which is satisfied by all gradients of a scalar functions.

Variable Gradient Method:

Theorem: A function $g(X)$ is the gradient of a scalar

function $V(X)$ if and only if the matrix $\left[\frac{\partial g(X)}{\partial X} \right]$

is symmetric; where

$$\left[\frac{\partial g(X)}{\partial X} \right] \triangleq \begin{bmatrix} \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial g_n}{\partial x_1} & \dots & \frac{\partial g_n}{\partial x_n} \end{bmatrix}$$

Krasovskii's Method

Let us consider the system $\dot{X} = f(X)$

Let $A(X) \triangleq \left[\frac{\partial f}{\partial X} \right]$: Jacobian matrix

Theorem :

If the matrix $F(X) \triangleq A(X) + A^T(X)$ is ndf for all $X \in D$ ($0 \in D$), then the equilibrium point is locally asymptotically stable and a Lyapunov function for the system is

$$V(X) = f^T(X) f(X)$$

Note: If $D = \mathbb{R}^n$ and $V(X)$ is radially unbounded,

then the equilibrium point is globally asymptotically stable.

Krasovskii's Method

$$\begin{aligned}\dot{V}(X) &= f^T \dot{f} + \dot{f}^T f \\ &= f^T \left[\frac{\partial f}{\partial X} \right]^T \dot{X} + \dot{X}^T \left[\frac{\partial f}{\partial X} \right] f \\ &= f^T (A^T + A) f \\ &= f^T F f\end{aligned}$$

Hence, if $F(X)$ is negative definite, $\dot{V}(X)$ is ndf.

So, by Lyapunov's theorem, $X = 0$ is asymptotically stable.

Generalized Krasovskii's Theorem

Theorem :

Let
$$A(X) \triangleq \left[\frac{\partial f(X)}{\partial X} \right]$$

A sufficient condition for the origin to be asymptotically stable is that

\exists two pdf matrices P and Q : $\forall X \neq 0$, the matrix

$$F(X) = A^T P + PA + Q$$

is negative semi-definite in some neighbourhood D of the origin.

In addition, if $D = \mathbb{R}^n$ and $V(X) \triangleq f^T(X) P f(X)$ is radially unbounded, then the system is globally asymptotically stable.

Generalized Krasovskii's Theorem

Proof : $V(X) = f^T(X) P f(X)$

$$\dot{V}(X) = \left[f^T P \dot{f} + \dot{f}^T P f \right]$$

$$= f^T P \left(\frac{\partial f}{\partial X} \right)^T \dot{X} + \left[\left(\frac{\partial f}{\partial X} \right)^T \dot{X} \right]^T P f$$

$$= f^T P A^T f + f^T A P f$$

$$= f^T (P A^T + A P + Q - Q) f$$

$$= \underbrace{f^T (P A^T + A P + Q) f}_{nsdf} - \underbrace{f^T Q f}_{ndf}$$

$$< 0 \text{ (ndf)} \quad \text{Hence, the result.}$$

Invariant and Limit Sets

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Invariant Set

A set M is said to be an "invariant set" with respect to the system $\dot{X} = f(X)$ if:

$$\boxed{X(0) \in M} \Rightarrow \boxed{X(t) \in M, \forall t > 0}$$

Examples:

- (i) An equilibrium point ($M = X_e$)
- (ii) Any trajectory of an autonomous system ($M = X(t)$)

Limit Set

Definition:

Let $X(t)$ be a trajectory of the dynamical system $\dot{X} = f(X)$. Then the set N is called the limit set (or positive limit set) of $X(t)$ if for any $p \in N$, \exists a sequence of times $\{t_n\} \in [0, \infty]$ such that $X(t_n) \rightarrow p$ as $t_n \rightarrow \infty$.

Note: Roughly, the limit set N of $X(t)$ is whatever $X(t)$ tends to in the limit.

Limit Set

Example:

- (i) An asymptotically stable equilibrium point is the limit set of any solution starting from a close neighbourhood of the equilibrium point.

- (ii) A stable limit cycle is the limit set for any solution starting sufficiently close to it

LaSalle's Theorem

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A Useful Theorem (Subset of LaSalle's Theorem)

Theorem : The equilibrium point $X = 0$ of the autonomous system $\dot{X} = f(X)$ is asymptotically stable if:

- (i) $V(X) > 0$ (pdf) $\forall X \in D$ [$0 \in D$]
- (ii) $\dot{V}(X) \leq 0$ (nsdf) in a bounded region $R \subset D$
- (iii) $\dot{V}(X)$ does not vanish along any trajectory in R
other than the null solution $X = 0$

Moreover,

If the above conditions hold good for $R = \mathbb{R}^n$ and $V(X)$ is radially unbounded, then $X = 0$ is globally asymptotically stable.

Example

Example:

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -x_2 - \alpha x_1 - (x_1 + x_2)^2 x_2$$

Solution: Let $V(X) = \alpha x_1^2 + x_2^2$, $\alpha > 0$

$$\dot{V}(X) = \left(\frac{\partial V}{\partial X} \right)^T f(X)$$

$$= [2\alpha x_1 \quad 2x_2] \begin{bmatrix} x_2 \\ -x_2 - \alpha x_1 - (x_1 + x_2)^2 x_2 \end{bmatrix}$$

$$= 2\alpha x_1 x_2 - 2x_2^2 - 2\alpha x_1 x_2 - 2(x_1 + x_2)^2 x_2^2$$

Example

$$\dot{V}(X) = -2x_2^2 \left[1 + (x_1 + x_2)^2 \right]$$
$$\leq 0 \quad (\text{nsdf})$$

Now $\dot{V}(X) = 0 \quad \forall t$

$$\Leftrightarrow x_2(t) = 0 \quad \forall t$$

$$\Rightarrow \dot{x}_2 = 0$$

$$-x_2 - \alpha x_1 - (x_1 + x_2)^2 x_2 = 0 \quad (\text{However, } x_2 = 0)$$

$$\therefore x_1 = 0 \quad \text{i.e. } X = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Example

Here we have :

- (i) $\dot{V}(X)$ does not vanish along any trajectory other than $X = 0$
- (ii) $\dot{V} \leq 0$ in \mathbb{R}^n
- (iii) $V(X)$ is radially unbounded,

Hence, the origin is **Globally asymptotically stable.**

LaSalle's Theorem

Let $V : D \rightarrow \mathbb{R}$ be a continuously differentiable (not necessarily pdf) function

and (i) $M \subset D$ be a compact set, which is

invariant with respect to the solution of $\dot{X} = f(X)$

(ii) $\dot{V} \leq 0$ in M

(iii) $E = \{X : X \in M \text{ and } \dot{V}(X) = 0\}$

i.e. E is the set of all points of $M : \dot{V} = 0$

(iv) N is the largest invariant set in E

Then Every solution starting in M approaches N as $t \rightarrow \infty$.

Lasalle's Theorem

Remarks:

- (i) $V(X)$ is required only to be continuously differentiable
It need not be positive definite.
- (ii) LaSalle's Theorem applies not only to equilibrium points, but also to more general dynamic behaviours such as limit cycles.
- (iii) The earlier theorems (on asymptotic stability) can be derived as a corollary of this theorem.

Stability Analysis of a Limit Cycle Using LaSalle's theorem

Example:

$$\begin{aligned}\dot{x}_1 &= x_2 + x_1(\beta^2 - x_1^2 - x_2^2) \\ \dot{x}_2 &= -x_1 + x_2(\beta^2 - x_1^2 - x_2^2), \quad \beta > 0\end{aligned}$$

Solution:

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Moreover,

$$\begin{aligned}\frac{d}{dt}(x_1^2 + x_2^2 - \beta^2) &= 2x_1\dot{x}_1 + 2x_2\dot{x}_2 \\ &= 2x_1[x_2 + x_1(\beta^2 - x_1^2 - x_2^2)] \\ &\quad + 2x_2[-x_1 + x_2(\beta^2 - x_1^2 - x_2^2)]\end{aligned}$$

Stability Analysis of a Limit Cycle Using LaSalle's theorem

$$= 2(x_1^2 + x_2^2) (\beta^2 - x_1^2 - x_2^2)$$

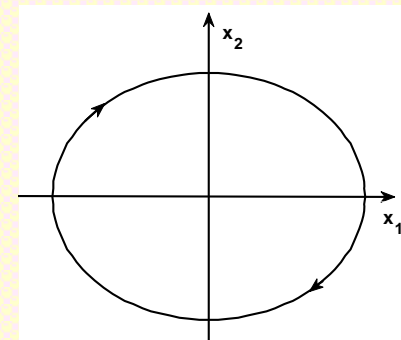
$$= 0 \quad \text{if} \quad x_1^2 + x_2^2 = \beta^2$$

\therefore The set of points defined by $x_1^2 + x_2^2 = \beta^2$ is an invariant set ; i.e any trajectory starting on this circle at t_0 stays on the circle $\forall t \geq t_0$

The trajectories on this invariant set are the solution of :

$$\dot{X} = f(X) \Big|_{(x_1^2 + x_2^2 = \beta^2)}$$

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ -x_1 \end{bmatrix} \Rightarrow \text{A clock-wise motion}$$



Stability Analysis of a Limit Cycle Using LaSalle's theorem

$$\text{Let } V(X) = \frac{1}{4}(x_1^2 + x_2^2 - \beta^2)^2 \quad [\text{Note: } V(X) \geq 0 \text{ in } \mathbb{R}^2]$$

$$\dot{V}(X) = \begin{bmatrix} \frac{\partial V}{\partial x_1} & \frac{\partial V}{\partial x_2} \end{bmatrix} \begin{bmatrix} f_1(X) \\ f_2(X) \end{bmatrix}$$

$$= (x_1^2 + x_2^2 - \beta^2) \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} x_2 + x_1(\beta^2 - x_1^2 - x_2^2) \\ -x_1 + x_2(\beta^2 - x_1^2 - x_2^2) \end{bmatrix}$$

$$= (x_1^2 + x_2^2 - \beta^2)(x_1^2 + x_2^2)(\beta^2 - x_1^2 - x_2^2)$$

$$= -(x_1^2 + x_2^2)(x_1^2 + x_2^2 - \beta^2)^2$$

$$\leq 0 \quad \text{Note: } \dot{V}(X) = -4(x_1^2 + x_2^2)V(X)$$

Stability Analysis of a Limit Cycle Using LaSalle's theorem

Moreover $\dot{V}(X) = 0$

\Leftrightarrow Either $(x_1^2 + x_2^2) = 0$ or $x_1^2 + x_2^2 = \beta^2$

i.e. Either $\underbrace{\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}}_{\text{origin}} = \underbrace{\begin{bmatrix} 0 \\ 0 \end{bmatrix}}_{\text{Here, } \dot{X} = 0}$ or $\underbrace{x_1^2 + x_2^2 = \beta^2}_{\text{Circle of radius } \beta}$
 (i.e it is an equilibrium point) It is an invariant set (i.e it is a limit cycle)

LaSalle's Theorem :

Step-1: For any $c > \beta$, let us define

$$M = \{X \in \mathbb{R}^2 : V(X) \leq c\}$$

In this set, $\dot{V}(X) \leq 0$
 (and this is true $\forall X \in M$)
 $\therefore M$ is an invariant set

By construction, M is closed and bounded

Stability Analysis of a Limit Cycle Using LaSalle's theorem

Step-2 [To find $E = \{X \in M : \dot{V}(X) = 0\}$]

It is already shown that

$$E = (0,0) \cup \{X \in \mathbb{R}^2 : x_1^2 + x_2^2 = \beta^2\}$$

Step-3 [To find N : The largest invariant set in E]

Since both the subsets that constitute E are invariant,

$$N = E$$

Hence, By Lasalle's Theorem, every motion starting

in M converges either to the origin or to the limit cycle, $x_1^2 + x_2^2 = \beta^2$

Stability Analysis (of limit cycle)

Further analysis:

Note that $V(X) = \frac{1}{4}(x_1^2 + x_2^2 - \beta^2)^2$ is a measure of

distance of a point $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ to the limit cycle, since:

$$V(X) = 0 \quad , \quad \text{if } x_1^2 + x_2^2 = \beta^2$$

$$\text{Also } V(X) = \left(\frac{\beta^4}{4}\right) \text{, if } \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Stability Analysis of a Limit Cycle Using LaSalle's theorem

Selecting: (i) $\beta : \beta < (\beta^4 / 4)$, (i.e. $\beta > \sqrt[3]{4}$)

(ii) $c : \beta < c < (\beta^4 / 4)$

(iii) $M = \{X \in \mathbb{R}^2 : V(X) \leq c\}$ (this excludes origin)

Then applying LaSalle's theorem, it follows that

any trajectory in M will converge to the limit cycle

\Rightarrow The limit cycle is Convergent / Attractive.

Corollary:

Letting $\varepsilon \rightarrow 0^+$, this also shows that the origin is unstable!

Domain of Attraction

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Domain of Attraction

Definition: Let $\psi(X, t)$ be trajectories of $\dot{X} = f(X)$ with initial condition X at $t = 0$. Then the Domain of attraction is defined as

$$D_A \triangleq \{X \in D : \psi(X, t) \rightarrow X_e \text{ as } t \rightarrow \infty\}$$

Philosophy : Around any asymptotically stable equilibrium point, there is a domain of attraction.

Question : Can we estimate a domain of attraction ?

Ans: Yes!

Domain of Attraction

Example: $\dot{x}_1 = 3x_2$

$$\dot{x}_2 = -5x_1 + x_1^3 - 2x_2$$

Eq. point: $x_2 = 0$

$$x_1(-5 + x_1^2) = 0 \Rightarrow x_1 = 0, \pm\sqrt{5}$$

\therefore This system has three eq. points $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$, $\begin{bmatrix} \sqrt{5} \\ 0 \end{bmatrix}$, $\begin{bmatrix} -\sqrt{5} \\ 0 \end{bmatrix}$

Let us study the stability of $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Define $V(X) = ax_1^2 - bx_1^4 + cx_1x_2 + dx_2^2$

Domain of Attraction

where, a, b, c, d need to be chosen "appropriately".

$$\begin{aligned}\dot{V}(X) &= \begin{bmatrix} \frac{\partial V}{\partial x_1} & \frac{\partial V}{\partial x_2} \end{bmatrix} \begin{bmatrix} 3x_2 \\ -5x_1 + x_1^3 - 2x_2 \end{bmatrix} \\ &= (3c - 4d)x_2^2 + (2d - 12b)x_1^3x_2 \\ &\quad + (6a - 10d - 2c)x_1x_2 + cx_1^4 - 5cx_1^2\end{aligned}$$

Choose:

$$\begin{bmatrix} 2d - 12b = 0 \\ 6a - 10d - 2c = 0 \end{bmatrix} \Rightarrow (a = 12, b = 1, c = d = 6) \text{ (one choice)}$$

Domain of Attraction

With this choice,

$$V(X) = 3(x_1 + 2x_2)^2 + 9x_1^2 + 3x_2^2 - x_1^4 \quad (\text{locally } pdf)$$

$$\dot{V}(X) = -6x_2^2 - 30x_1^2 + 6x_1^4 \quad (\text{locally } ndf)$$

Hence, the system is locally asymptotically stable.

Note: Here, $V(X) > 0$ and $\dot{V}(X) < 0$ as long as $-1.6 < x_1 < 1.6$

We may be tempted to conclude that $D = \{X \in \mathbb{R}^2 : -1.6 < x_1 < 1.6\}$

is a region of attraction .

Surprise : The conclusion is incorrect!

This is because D is NOT an invariant set

Theorem: Domain of Attraction

Theorem:

Let (i) X_e be an equilibrium point of the system $\dot{X} = f(X)$

(ii) $V(X): D \rightarrow \mathbb{R}$ be a continuously differentiable function

(iii) $M \subset D$ be a compact set containing X_e such that " M is invariant with respect to the solution of the system"

(iv) \dot{V} is such that
$$\dot{V} < 0 \quad \forall X \neq X_e \text{ in } M$$
$$= 0 \quad \text{if } X = X_e$$

Under these assumption, M is a subset of the domain of attraction, i.e. M is an estimate of domain of attraction.

Proof: In LaSalle's theorem, $E = \{X : X \in M \ \& \ \dot{V} = 0\} = X_e$. Hence the result !

Example....Contd.

$$V(X) = 12x_1^2 - x_1^4 + 6x_1x_2 + 6x_2^2$$

$$\dot{V}(X) = -6x_2^2 - 30x_1^2 + 6x_1^4$$

We already know that

$V(X) > 0$ and $\dot{V}(X) < 0$ happens in

$$D = \left\{ X \in \mathbb{R}^2 : -1.6 < x_1 < 1.6 \right\}$$

$$\left[\begin{array}{l} \text{Note:} \\ V(0) = 0 \\ \dot{V}(0) = 0 \end{array} \right]$$

Domain of Attraction

Let us find the minimum of $V(X)$ along the very edge of this set (to restrict this set further).

Then

$$V|_{x_1=1.6} = 24.16 + 9.6x_2 + 6x_2^2$$

$$\frac{\partial}{\partial x_2} \left(V|_{x_1=1.6} \right) = 9.6 + 12x_2 = 0$$

$$\Rightarrow x_2 = \frac{-9.6}{12} = -0.8$$

Domain of Attraction

Similarly

$$\begin{aligned}\frac{\partial}{\partial x_2} \left(V \Big|_{x_1 = -1.6} \right) &= \frac{\partial}{\partial x_2} \left(24.16 - 9.6x_2 + 6x_2^2 \right) \\ &= -9.6 + 12x_2 = 0 \\ \Rightarrow x_2 &= 0.8\end{aligned}$$

Also $\frac{\partial^2}{\partial x_2^2} \left(V \Big|_{x_1 = \pm 1.6} \right) = 12 > 0$

$\therefore V(X)$ has local minima when $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.6 \\ -0.8 \end{bmatrix}, \begin{bmatrix} -1.6 \\ 0.8 \end{bmatrix}$

Domain of Attraction

Moreover, $V(1.6, -0.8) = V(-1.6, -0.8) = 20.32$

(i.e. both the minimums are equal)

[Else, we need to choose the minimum of the two minimums.]

$\therefore M = \{X \in D : V(X) \leq 20.32 - \varepsilon\} \subset D$ is an invariant set,

and hence, M is an estimate of the domain of attraction

Note: As long as $\varepsilon > 0$, the local minimums are excluded.

Hence $X(t) \rightarrow 0$ as long as it starts in M

An Interesting Result

Lemma

If a real function $V(t)$ satisfies the

in equality $\dot{V}(t) \leq -\alpha V(t)$, $\alpha \in \mathbb{R}$

Then $V(t) \leq e^{-\alpha t} V(0)$

Proof:

Let $Z(t) = \dot{V} + \alpha V$

then $\dot{V} + \alpha V = Z(t)$ (Note: $Z(t) \leq 0$)

An Interesting Result

Let us consider $Z(t)$ as an "external input"
to this "linear system"

Then

$$V(t) = e^{-\alpha t} V(0) + \underbrace{\int_0^t \underbrace{e^{-\alpha(t-\tau)}}_{\geq 0} \cdot \underbrace{1 \cdot Z(\tau)}_{\leq 0} d\tau}_{\leq 0}$$

$$\therefore \boxed{V(t) \leq e^{-\alpha t} V(0)}$$

References

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Thanks for the Attention...!

