

Linear

n-th order ODE
 $y^{(n)} + a_1 y^{(n-1)} + \dots + a_{n-1} y' + a_n y = b_0 u^{(n)} + b_1 u^{(n-1)} + \dots + b_{n-1} u' + b_n u$
 TF $\frac{Y(s)}{U(s)} = \frac{b_0 s^n + b_1 s^{n-1} + \dots + b_{n-1} s + b_n}{s^n + a_1 s^{n-1} + \dots + a_{n-1} s + a_n}$
 $\dot{x} = Ax + Bu$
 $y = Cx + Du$
 solution $x(t) = ?$ $x(t_0) = x_0$
 if $\dot{x}(t) = Ax(t) + Bu(t)$ — 0

$e^{At} (e^{-At} x(t))'$
 $= e^{At} (-A) e^{-At} x(t) + e^{At} e^{-At} \dot{x}(t)$
 $= -Ax(t) + \dot{x}(t)$ — 0
 given $0 \in \mathbb{D}$
 $-Ax(t) + \dot{x}(t) = Bu(t)$
 $e^{At} (e^{-At} x(t))' = e^{At} Bu(t)$
 $(e^{-At} x(t))' = e^{-At} Bu(t)$

from $t_0 \rightarrow t$
 $\int_{t_0}^t (e^{-At} x(t))' dt = \int_{t_0}^t e^{-At} Bu(t) dt$
 $e^{-At} x(t) - e^{-At_0} x(t_0) = \int_{t_0}^t e^{-At} Bu(t) dt$
 $x(t) = e^{A(t-t_0)} x_0 + \int_{t_0}^t e^{A(t-\tau)} Bu(\tau) d\tau$

transfer function
 $\mathbb{D}(x(t_0))$
 $x_0 = e^{A(t-t_0)} x_0 + \int_{t_0}^t e^{A(t-\tau)} Bu(\tau) d\tau$
 $\frac{dx(t)}{dt} = \frac{d}{dt} [e^{A(t-t_0)} x_0 + \int_{t_0}^t e^{A(t-\tau)} Bu(\tau) d\tau]$
 $= A [e^{A(t-t_0)} x_0 + \int_{t_0}^t e^{A(t-\tau)} Bu(\tau) d\tau] + e^{A(t-t_0)} x_0 + \int_{t_0}^t e^{A(t-\tau)} Bu(\tau) d\tau$
 $= Ax(t) + Bu(t)$

Definition
 $A \in \mathbb{R}^{n \times n}$
 $e^{At} = I_n + At + \frac{A^2 t^2}{2!} + \dots + \frac{A^k t^k}{k!} + \dots$

Linear
 $f: X \rightarrow Y$ is a "linear function"
 i.e. f is $\textcircled{1}$ additive $f(x_1+x_2) = f(x_1) + f(x_2)$
 $\textcircled{2}$ homogeneous $f(\alpha x) = \alpha f(x)$
 in general robotic systems aren't linear.
 Euler-Lagrange systems
 $M(q)\ddot{q} + C(q,\dot{q})\dot{q} + G(q) = u$
 Adv. q. spring-mass-damper (+) nonlinear damping
 $ky = m\ddot{y} + b\dot{y}$

Linear
 $m\ddot{y} + c\dot{y} + ky = u$
 non-linear
 $m\ddot{y} + c\dot{y} + ky + k_2 y^2 = u$
 det $x_1 = \dot{y}$
 $x_2 = y$
 $\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = \frac{1}{m}(-Cx_2 - kx_1 - k_2 x_1^2) + \frac{1}{m}u \end{cases}$

Autonomous system not explicitly dependent on t
 $\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ \frac{1}{m}(-Cx_2 - kx_1 - k_2 x_1^2) \end{bmatrix}$ (n=2)
 Non-autonomous system explicitly dependent on t
 $\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ \frac{1}{m}(-Cx_2 - kx_1 - k_2 x_1^2 + \cos(\omega t)) \end{bmatrix}$

nonlinear system
 $\dot{x} = f(x, u)$
 $y = h(x, u)$
 w/ control law
 $u = k(x, u)$
 $\dot{x} = f(x, k(x, u))$ $\dot{x} = f(x, k(x))$
 $y = h(x, k(x, u))$ $y = h(x, k(x))$
 solution of the nonlinear system
 $\dot{x} = f(x, \tau)$
 $x(t_0) = x_0$
 $x(t) = ?$
 a may not be a solution
 solutions don't have internal equilibria: not unique
 addition $\rightarrow \infty$ (t $\rightarrow \infty$) when $\tau \rightarrow Y$

equilibrium points
 $\dot{x} = 2\sqrt{x} \rightarrow x(t) = \begin{cases} 0 & 0 \leq t < \infty \\ t^2 & 0 \leq t < \infty \end{cases}$
 $x(0) = 0$
 $\dot{x} = 2\sqrt{x}$
 $x(0) = 0$
 eq. don't change time
 $\dot{x} = -x + x^2 \rightarrow x(t) = \frac{x_0}{x_0 + (1-x_0)e^t}$
 $x(0) = x_0$

$\forall t \in [t_0, \frac{t_0 + \infty}{2}]$, $x_0 > 1$
 $\Rightarrow x_0 + (1-x_0)e^{\frac{t_0 + \infty}{2}} = \frac{x_0}{x_0 + (1-x_0)e^{\frac{t_0 + \infty}{2}}}$
 $= x_0 + (1-x_0) \frac{x_0}{x_0 - 1}$
 $= 0$
 $\Rightarrow x(t)$ not defined
 Theorem: Local Existence Global Uniqueness
 $\forall x, y \in B = \{x \in \mathbb{R}^n \mid \|x - x_0\| \leq r\}$
 $\forall x, y \in B \subset \mathbb{R}^n$ (compact) (for local)
 $\forall t \in [t_0, t_1]$
 $\exists \delta > 0$ s.t.
 $\dot{x} = f(x, t)$ w/ $x(t_0) = x_0$
 has a unique solution over $[t_0, t_0 + \delta]$ local
 $\dot{x} = f(x, t)$ w/ $x(t_0) = x_0$
 has a unique solution over $[t_0, t_1]$ global

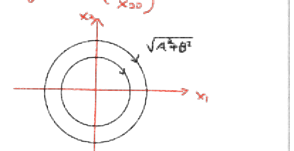
Global Existence & Uniqueness
 $\dot{x} = Ax$
 $\|Ax - Ay\| \leq L \|x - y\|$
 $\Rightarrow \|Ax - Ay\| \leq \|A\| \|x - y\|$ Cauchy-Schwarz
 $\Rightarrow \text{let } \|A\| = L$
 $\therefore \|Ax - Ay\| \leq L \|x - y\|$
 e.g. multiple equilibrium pts
 $\dot{x} = f(x)$, $x(0) = x_0$
 def: an equilibrium point is x^* if $f(x^*) = 0$
 nonlinear system have "multiple isolated equilibrium points"
 $\dot{x} = -x + x^3$, $x(0) = x_0$
 $x^* = 0$ or ± 1

Limit cycles
 explain the periodic behavior (closed loop in the phase space)
 - limit cycle is isolated
 - limit cycle is not dependent to initial condition
 some polynomial systems will converge to a closed curve. e.g. Van der Pol
 e.g. bifurcations e.g. chaos
 different parameters sensitive to different equilibrium internal condition pts.

Equilibrium
 $x^* = x^*$ is an equilibrium
 the system starts @ x^*
 it will remain x^* for the rest of time
 i.e. $\dot{x} = f(x) = 0$, x^* is the set of all possible eq. pts.
 在平衡点附近的系统总是稳定的。
 但！不代表它是稳定的。
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 但！不代表它是稳定的。

Controlling nonlinear systems
 nonlinear systems:
 complexity: peculiar behaviors
 difficulty: closed form solutions are unavailable
 phase plane methods!
 describing functions!
 Lyapunov theory!

Phase Plane
 consider a 2-dimensional autonomous sys.
 $\dot{x}_1 = f_1(x_1, x_2)$ $x_1(0) = x_{10}$
 $\dot{x}_2 = f_2(x_1, x_2)$ $x_2(0) = x_{20}$
 $x(t, x_0) = \begin{bmatrix} x_1(t, x_{10}) \\ x_2(t, x_{20}) \end{bmatrix}$
 def: plane w/ x_1, x_2 as coordinates is called phase plane

e.g. LC circuit
 $\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$
 $\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} A \\ B \end{bmatrix}$
 soln
 $x_1(t) = \sqrt{A^2 + B^2} \sin(\omega t + \theta)$
 $x_2(t) = \sqrt{A^2 + B^2} \cos(\omega t + \theta)$
 $\theta = \tan^{-1}(\frac{x_{10}}{x_{20}})$


it is unique if $f_1 \neq 0$ or $f_2 \neq 0$
 phase trajectories will not intersect
 it is NOT unique if $f_1 = 0$, $f_2 = 0$
 phase trajectories intersect

For linear systems, the stability is uniquely determined by the nature of their singular points.
 For nonlinear systems, the stability behavior may be more complex. The system may have multiple isolated singular points some of which are stable and some of which are not. The stable singular point may be globally asymptotically stable or just locally asymptotically stable.
 Phase Plane Analysis (linear system)
 - classification of a singular point
 - visualize system trajectories
 $\dot{x} = Ax$
 $\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$
 soln $x_{1,2}(t) = \begin{cases} k_1 e^{\lambda_1 t} e^{i\omega t} & \lambda_1 = \lambda_2 \\ (k_1 + k_2 t) e^{\lambda t} & \lambda_1 = \lambda_2 \end{cases}$

Case 1 $\text{sign}(\lambda_1) = \text{sign}(\lambda_2) \in \mathbb{R}$
 stable/unstable node
 $\lambda_1 < 0$ $x(t) \rightarrow 0$ exponentially
 $\lambda_2 < 0$ $t \rightarrow \infty$
 $\lambda_1 > 0$ $x(t) \rightarrow \infty$ exponentially
 $\lambda_2 > 0$ $t \rightarrow \infty$

Case 2 - saddle point
 $\text{sign}(\lambda_1) \neq \text{sign}(\lambda_2) \in \mathbb{R}$
 assume $\lambda_1 < 0$ $\lambda_2 > 0$
 $x_{1,2}(t) = \begin{cases} k_1 e^{\lambda_1 t} + k_2 e^{\lambda_2 t} & \lambda_1 \neq \lambda_2 \\ (k_1 + k_2 t) e^{\lambda t} & \lambda_1 = \lambda_2 \end{cases}$
 $\Delta \text{ dir } k_2 \neq 0$
 $x(t) \rightarrow \infty$, $t \rightarrow \infty$
 $\Delta \text{ dir } k_2 = 0$
 $\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} k_{11} \\ k_{12} \end{bmatrix} e^{\lambda t}$ $\frac{x_1}{x_2} = \frac{k_{11}}{k_{12}}$

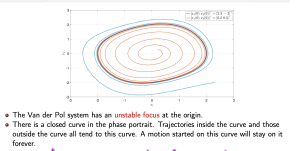
Case 3 $\lambda_1 \lambda_2 \rightarrow$ complex w/ conjugate
 stable/unstable focus
 $\lambda_{1,2} = \sigma \pm i\omega$
 $\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} A \sin(\omega t + \theta) \\ B \cos(\omega t + \theta) \end{bmatrix} e^{\sigma t}$
 $\Delta \text{ if } \sigma < 0$ $x(t) \rightarrow 0$, $t \rightarrow \infty$
 $\Delta \text{ if } \sigma > 0$ $x(t) \rightarrow \infty$, $t \rightarrow \infty$

Case 4 $\lambda_1 \lambda_2$ complex conjugate w/ zero real part
 center point
 $\lambda_{1,2} = \pm i\omega$
 $\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} A \sin(\omega t + \theta) \\ B \cos(\omega t + \theta) \end{bmatrix}$
 $\frac{x_1^2}{A^2} + \frac{x_2^2}{B^2} = 1$
 unstable saddle point
 unstable
 marginally stable

Phase plane analysis (nonlinear system)
 strategy: Jacobian Linearization
 recall: $\begin{cases} \dot{x}_1 = f_1(x_1, x_2) & x_1(0) = x_{10} \\ \dot{x}_2 = f_2(x_1, x_2) & x_2(0) = x_{20} \\ x(t, x_0) \end{cases}$
 $\Rightarrow \dot{x}_1 = Ax_1 + bx_2 + f_1(x_1, x_2)$
 $\dot{x}_2 = Cx_1 + dx_2 + f_2(x_1, x_2)$
 $\textcircled{1}$ singular pt
 $\Rightarrow \dot{x}_1 = Ax_1 + bx_2$
 $\dot{x}_2 = Cx_1 + dx_2$

Jacobian Linearization
 $\begin{cases} \dot{x}_1 = f_1(x_1, x_2) \\ \dot{x}_2 = f_2(x_1, x_2) \end{cases}$
 assume $f_1(x_1^*, x_2^*) = f_2(x_1^*, x_2^*) = 0$
 $x^* = [x_1^*, x_2^*]^T$ is singular pt.
 $A_{x^*} \equiv \begin{bmatrix} a & b \\ c & d \end{bmatrix}$
 $= \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(x_1^*, x_2^*) & \frac{\partial f_1}{\partial x_2}(x_1^*, x_2^*) \\ \frac{\partial f_2}{\partial x_1}(x_1^*, x_2^*) & \frac{\partial f_2}{\partial x_2}(x_1^*, x_2^*) \end{bmatrix}$

$f(x, x_0) = f(x_1^*, x_2^*) + \frac{\partial f}{\partial x_1}(x_1^*, x_2^*) (x_1 - x_1^*) + \frac{\partial f}{\partial x_2}(x_1^*, x_2^*) (x_2 - x_2^*) + \text{O}(\|x - x^*\|^2)$
 $\dot{x} = A_{x^*} (x - x^*)$
 $\dot{x} = Ax - Ax^*$
 we hv
 $\begin{cases} \dot{x}_1 = ax_1 + bx_2 + \text{O} \\ \dot{x}_2 = cx_1 + dx_2 + \text{O} \end{cases}$
 Jacobian Linearization

Limit cycle
 $\textcircled{1}$ Limit Cycles


The Van der Pol system has an unstable focus at the origin.
 There is a closed curve in the phase portrait. Trajectories inside the curve and those outside the curve all tend to this curve. A motion started on this curve will stay on it forever.
 closed: traj. is closed, periodic nature
 isolated: traj. is isolated, limiting nature
 $\hookrightarrow L$ is isolated if
 \exists region $B = \{x: \text{dist}(x, L) < \epsilon, \epsilon > 0\}$
 $\in B$
 $\text{dist}(x, L) = \inf_{y \in L} \|x - y\|$
 stable unstable semi-stable
 (for $\text{dist}(x, L) < \epsilon$) (for $\text{dist}(x, L) > \epsilon$) (near traj. don't have change)

Existence of Limit Cycles
 refer to slides.

Diagram showing various phase portraits: stable node, unstable node, saddle point, stable focus, unstable focus, center point, and marginally stable.

Diagram showing various phase portraits: stable node, unstable node, saddle point, stable focus, unstable focus, center point, and marginally stable.

Autonomous System

$\dot{x} = f(x)$
 $x \in \mathbb{R}^n$
 $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$
 W.L.O.G.
 if $f(x^*) = 0$ & $x^* \neq 0$
 then let $\xi = x - x^*$
 $\dot{\xi} = \frac{df}{dx}(x^*) \xi$
 $= A \xi$
 $= f(\xi + x^*) \equiv \tilde{f}(\xi)$
 $\xi = 0$ is the equilibrium point of $\tilde{f}(\xi)$

objective: determine the stability of $\dot{x} = f(x)$
 w.o. getting the solution

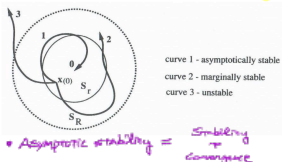
Stable
 x^* is said that
 equilibrium pt $x^* = 0$ of $\dot{x} = f(x)$
 is stable if
 if $\forall \epsilon > 0, \exists \delta > 0$ s.t. $\|x(0)\| < \delta \implies \|x(t)\| < \epsilon$
 $\forall t \geq 0$

if $\forall \epsilon > 0, \exists \delta > 0, \forall x(0) \in B_\delta(0) \implies \|x(t)\| < \epsilon$
 $\forall t \geq 0$
 then x^* is **locally asymptotically stable**

if $\forall \epsilon > 0, \exists \delta > 0, \forall x(0) \in B_\delta(0) \implies \|x(t)\| < \epsilon$
 $\forall t \geq 0$
 then x^* is **globally asymptotically stable**

Asymptotic Stable
 it is said that
 equilibrium point $x=0$ of $\dot{x}=f(x)$
 is asymptotically stable if
 1. it is stable
 2. $\exists \epsilon > 0$ s.t.
 $\|x(0)\| < \epsilon \implies \lim_{t \rightarrow \infty} x(t) = 0$
 (convergence condition)
 ($B_\epsilon = \{x: \|x\| < \epsilon\}$ is "domain of attraction")

$\dot{x} = Ax$
 • is stable $\iff x=0$ is stable
 • is marginally stable $\iff x=0$ is stable
 • is unstable $\iff x=0$ is unstable



Exponential Stability
 $x=0$ of $\dot{x}=f(x)$
 is exponentially stable if $\exists \alpha > 0, \lambda > 0$ s.t.
 $\|x(t)\| \leq \alpha \|x(0)\| e^{-\lambda t} \quad \forall t \geq 0$
 exponential stability \implies asymptotic stability

Local/global stability
 if asymptotic stability holds for any x_0
 the equilibrium point is asymptotically stable in the large.
 globally asymptotically stable

$\dot{x} = Ax$
 $\lambda(A)$ has (-) real parts
 \implies origin is globally exponentially stable
 \implies stability for linear system is "global" & "exponential"

Ljapunov indirect method

• perform linearization locally
 • local stability
 • Jacobian Linearization
 $\dot{x} = \frac{\partial f}{\partial x} \Big|_{x=0} x + O(\|x\|^2)$
 $\hat{=} Ax + O(\|x\|^2)$
 $A = \frac{\partial f}{\partial x} \Big|_{x=0} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$

$\frac{\partial f}{\partial x} \Big|_{x=0} = J$ of $f(x) @ 0$
 $\dot{x} = Ax$
 is the "Jacobian Linearization" of $\dot{x} = f(x) @ x=0$

Ljapunov's Indirect Method
 $\dot{x} = Ax \implies \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} \lambda_1 x_1 \\ \lambda_2 x_2 \\ \vdots \\ \lambda_n x_n \end{bmatrix}$
 • $\forall \lambda_i, \text{Im}(\lambda_i) < 0$
 then the point is **locally asymptotically stable**

• if any $\lambda_i, \text{Re}(\lambda_i) > 0$
 then the point is **unstable**
 • if $\forall \lambda_i, \text{Re}(\lambda_i) \leq 0$
 $\text{Re}(\lambda_i) = 0$
 for at least 1, no conclusion

ex. $\dot{x} = f(x, u)$
 $\dot{y} = h(x, u)$
 $A = \frac{\partial f}{\partial x} \Big|_{x=0, u=0}$
 $B = \frac{\partial f}{\partial u} \Big|_{x=0, u=0}$
 $C = \frac{\partial h}{\partial x} \Big|_{x=0, u=0}$
 $\dot{x} = Ax + Bu + \tilde{C}$
 $\dot{y} = Cx + \tilde{D}$

Let feedback control law $u = -Kx$
 $\dot{x} = f(x, -Kx) \equiv f_c(x)$
 $\frac{\partial f_c(x)}{\partial x} \Big|_{x=0} = \frac{\partial f(x, u)}{\partial x} \Big|_{x=0, u=-Kx}$
 $= \frac{\partial f(x, u)}{\partial x} \Big|_{x=0, u=0} + \frac{\partial f(x, u)}{\partial u} \Big|_{x=0, u=0} (-K)$
 $= A - BK$

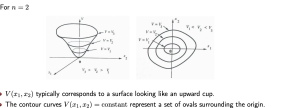
$\hat{=} A - BK$
 $\hat{=} \text{stable}$ i.e. $\forall \lambda(A-BK); < 0$

• (A, B) is controllable if $\text{rank}(BAB \dots A^m B) = n$
 $\exists K \in \mathbb{R}^{m \times n}$ s.t. $(A-BK)$ is stable
 • $\dot{x} = Ax + Bu$
 is a small perturbation and valid when x, u are small
 • $u = -Kx$ only guarantees "local" asymptotic stability

Ljapunov direct method

consider nonlinear damped system
 $m\dot{x} + b\dot{x}(1+x) + k_1x + k_2x^3 = 0$
 \implies total energy:
 $V(x) = \frac{1}{2}m\dot{x}^2 + \int_0^x (k_1x + k_2x^3) dx$
 $= \frac{1}{2}m\dot{x}^2 + \frac{1}{2}k_1x^2 + \frac{1}{4}k_2x^4$
 • zero energy when $\dot{x}=0, x=0$
 • equilibrium point
 • asymptotic stability implies mechanical energy $\rightarrow 0$
 • instability implies mechanical energy $\rightarrow \infty$

Locally positive energy
 • $V(x)$ is locally P.D.
 if $V(0) = 0$
 $\& V(x) \in B_{R_0}$
 $\forall x \in B_{R_0} \implies V(x) > 0$
 • $V(x)$ is globally P.D. if $B_{R_0} = \mathbb{R}^n$



Ljapunov Function
 $\frac{\partial V}{\partial x} = \begin{bmatrix} \frac{\partial V}{\partial x_1} & \dots & \frac{\partial V}{\partial x_n} \end{bmatrix}$
 $\dot{V}(x) = \frac{\partial V}{\partial x} \frac{dx}{dt} = \frac{\partial V}{\partial x} f(x) \leq 0$
 $\implies V(x)$ is a Ljapunov function (local)
 if $V(x)$ PD $\forall x \in \mathbb{R}^n$
 $\dot{V}(x) \leq 0$
 $V(x) \implies$ global Ljapunov function

Ljapunov Stability Theorem (Local)
 • $V(x)$ PD
 $\dot{V}(x)$ NSD } equilibrium point @ origin is stable
 • if $\dot{V}(x)$ is ND } equilibrium point @ origin is asymptotically stable

Ljapunov Stability Theorem (Global)
 • $V(x)$ PD } @ all point
 $\dot{V}(x)$ ND } i.e. globally
 $V(x) \implies \infty$ as $\|x\| \implies \infty$ (radially unbounded)
 then equilibrium point @ origin is globally asymptotically stable

Invariant Set Theorem

recall:
 $\dot{V}(x)$ needs to be ND
 \implies asymptotic stability
 (non-trivial in practice!)
Invariant set
 a set $G \in \mathbb{R}^n$ is an invariant set if \forall traj. starting from $p \in G$ remains in G for all time

Invariant Set Theorem (local)
 $V(x)$
 • for some $\epsilon > 0, S_\epsilon = \{x: V(x) < \epsilon\}$ is bounded
 • $\dot{V}(x) \leq 0 \quad \forall x \in S_\epsilon$
 • let R be the set of all pts in S_ϵ where $\dot{V}(x) = 0$
 • M is the largest invariant set in R (union of all invariant sets)
 Then $x(t)$ originating in S_ϵ tends to M as $t \rightarrow \infty$ when $\dot{V} < 0$ i.e., ND

$R = M \cup \beta$
 β Ljapunov stability theorem (local) is a special case here
Using it
 • $V(x)$ is PD
 • $\dot{V}(x)$ is NSD
 • set R
 $S_\epsilon \setminus R = \{x: -\epsilon < V(x) < \epsilon \text{ and } \dot{V}(x) = 0\}$
 • if set R contains no traj. other than $x=0$
 \implies equilibrium point @ origin is asymptotically stable
 S_ϵ is the domain of attraction

Invariant Set Theorem (global)
 • $V(x) \implies \infty$ as $\|x\| \implies \infty$
 • $\dot{V}(x) \leq 0 \quad \forall x \in \mathbb{R}^n$
 • set R
 $\dot{V}(x) = 0$
 M largest invariant set in R
 Then all $x(t; x_0)$ that are globally asymptotically converge to M as $t \rightarrow \infty$

易言之:
 - 上述条件成立, 任何 $x(t; x_0)$ 会 globally asymptotically 收敛至 invariant set M
 - 这个 M 非空, 他在 R i.e., $M \in R$
 R 为什么? $R = \{x: \dot{V}(x) = 0\}$

use it?
 show that
 • $V(x)$ PD
 • $\dot{V}(x)$ NSD
 • $\dot{V}(x) = 0 \implies x = 0$

INSUM!
 • $V(x)$ PD, $\dot{V}(x)$ NSD
 • radially unbounded $V(x) \implies \infty$ as $\|x\| \implies \infty$
 • $\dot{V}(x) = 0$ only when $x = 0$
 \implies asymptotically stable

• $V(x)$ PD
 $\dot{V}(x)$ NSD
 \implies stable