

Existence of a CQLF is a sufficient but not necessary condition for uniform asympt. stabs. of a switched system -

example

$$A_1 = \begin{bmatrix} -1 & -1 \\ 1 & -1 \end{bmatrix} \quad A_2 = \begin{bmatrix} -1 & -10 \\ 0.1 & -1 \end{bmatrix} \quad P = \{e_1, e_2\}$$

the two systems do not share any QLF
(show it!)

However the switching system $\dot{x} = A_0 x$ is
globally uniformly asympt. stab. for all 5 switch. signals

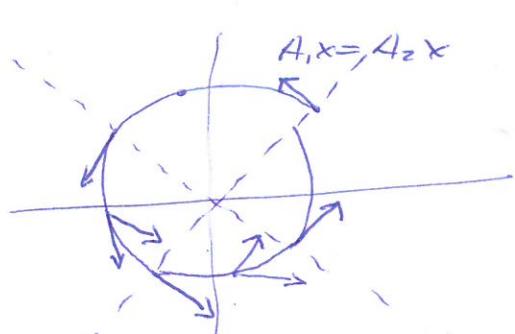
To see it: look at worst-case switching

The two systems $A_1 x$ and $A_2 x$ are collinear
on the dashed lines

In between one of the two
points more outside than
the other (and they switch
roles on the dashed lines)

\Rightarrow worst-case consists of following always the
one pointing most outwards

Even doing that the trajectory must end converge
to the origin -



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Another way to show $\text{outf. as stab.} \Rightarrow \text{use higher order homogeneous polynomials in place of quadratic polynomials}$

$$\text{ex } V(x) = x^T P x = x^T \begin{bmatrix} p_1 & p_2 \\ p_2 & p_3 \end{bmatrix} x = [x_1 \ x_2]^T \begin{bmatrix} p_1 & p_2 \\ p_2 & p_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = p_1 x_1^2 + 2p_2 x_1 x_2 + p_3 x_2^2 \text{ homogeneous polyn. of order 2}$$

If I can find higher order homogeneous polyn.

$$V(x) = \sum_j p_j x_1^{k_{j1}} x_2^{k_{j2}} \quad \text{s.t. } k_{j1} + k_{j2} = 4 \quad (\text{for instance})$$

s.t. $V(x)$ is pos. def., then I can try to use it as Lyapunov function.

ex $V(x) = (x^T P x)^2$ 4th order homog. polyn. but not useful because it "represents" twice the same Lyapunov funct.

A systematic way to compute higher order homog. polyn. funct. which are automatically pos. def.

Use Kronecker products. \otimes

For matrices $A, B \in \mathbb{R}^{n \times n}$

$$A \otimes B = \begin{bmatrix} a_{11}B & a_{12}B & \dots & a_{1n}B \\ a_{21}B & \vdots & & \\ a_{n1}B & & \ddots & a_{nn}B \end{bmatrix}$$

For a vector $x \in \mathbb{R}^n$

$$\tilde{x} = x \otimes x = \begin{bmatrix} x_1 x \\ \vdots \\ x_n x \end{bmatrix} = \begin{bmatrix} x_1^2 \\ x_1 x_2 \\ \vdots \\ x_1 x_n \\ x_n x_1 \\ \vdots \\ x_n^2 \end{bmatrix} \text{ all monomials having power } \in \mathbb{R}^{n^2}$$

$$A \otimes B \in \mathbb{R}^{n^2 \times n^2} \quad x \otimes x \in \mathbb{R}^{n^2}$$

$$\dot{\tilde{x}} = Ax$$

For \tilde{x} a linear dynamics becomes:

$$\begin{aligned} \frac{d}{dt} \tilde{x} &= \frac{d}{dt}(x \otimes x) = \dot{x} \otimes x + x \otimes \dot{x} = Ax \otimes x + x \otimes Ax \\ &= (A \otimes I)\tilde{x} + (I \otimes A)\tilde{x} = \underbrace{(A \otimes I + I \otimes A)}_{\text{Kronecker sum}} \tilde{x} \\ &\leq A \oplus A \end{aligned}$$

\Rightarrow A quadratic function can be constructed in \mathbb{C}

$$V(\tilde{x}) = \tilde{x}^T P \tilde{x} \quad \tilde{x} \in \mathbb{R}^{n^2} \quad P \in \mathbb{R}^{n^2 \times n^2}$$

$$\begin{aligned} \dot{V}(\tilde{x}) &= \dot{\tilde{x}}^T P \tilde{x} + \tilde{x}^T P \dot{\tilde{x}} = \tilde{x}^T (A \oplus A)^T P \tilde{x} + \tilde{x}^T P (A \oplus A) \tilde{x} \\ &= \tilde{x}^T ((A \oplus A)^T P + P (A \oplus A)) \tilde{x} \end{aligned}$$

If I call $\tilde{A} = A \oplus A$ I get a Lyapunov eq in $\mathbb{R}^{n^2 \times n^2}$

$$\tilde{A}^T P + P \tilde{A}$$

For a single linear system there is no advantage in using polynomials of order higher than 2

($\tilde{A}^T P + P \tilde{A} < 0$ for $P > 0$ is an iff condition for \tilde{A} stab)

However when searching for a CLF, a

Homog. polyn. Lyap. funct is less conservative than a quadratic Lyap. funct.

$$\left\{ \begin{array}{l} \tilde{A}_i^T P + P \tilde{A}_i < 0 \\ P > 0 \end{array} \right. \quad \begin{array}{l} \text{is a less conservative} \\ \text{soft-cond.} \end{array}$$

For the example: in 8-th order homog. poly exists
(not a QCLF, not a 4-HPLF or 6-HPLF)

Center Manifold theory (Ch 8, khalil)

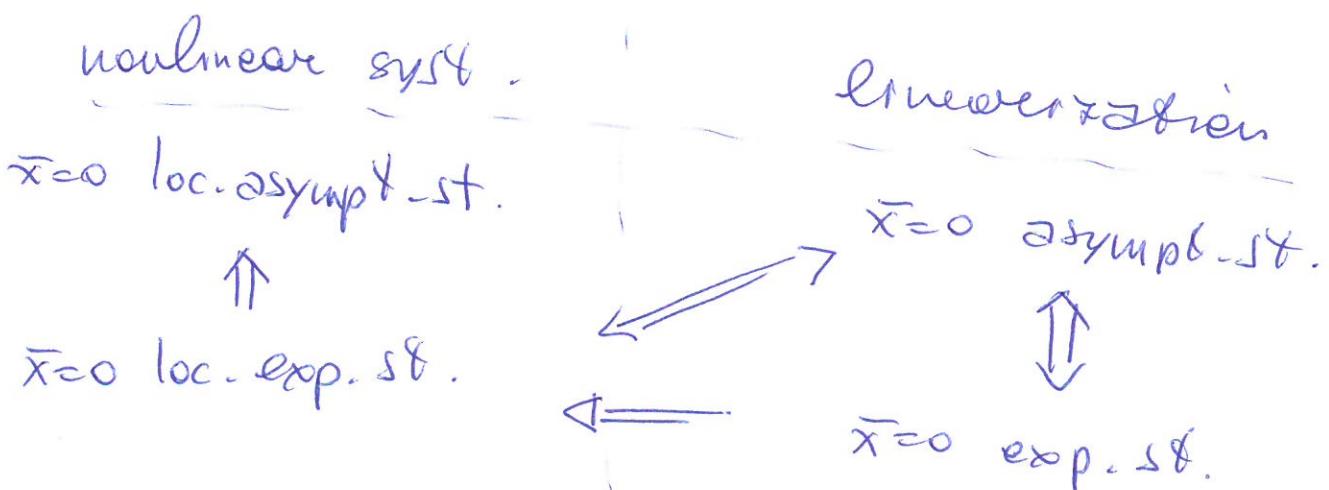
Recall that for an autonomous linear system

$$\dot{x} = f(x) \quad f(0) = 0$$

we can investigate stability of $\bar{x}=0$ by looking at the linearization

$$\dot{x} = Ax \quad A = \frac{\partial f}{\partial x} \Big|_{\bar{x}=0}$$

we had:



If $\text{Re}[\lambda_i] \leq 0$ and $\text{Re}[\lambda_i] = 0$ for some i , then linearization is inconclusive.

Center manifold theory studies these inconclusive cases. It does so by "projecting" the dynamics on the manifold which is the "ptolong" the eigenspace of the λ_i with $\text{Re}[\lambda_i] = 0$

manifold (more details next time): it is
 a k -dimensional ^{smooth} hypersurface in \mathbb{R}^n
 represented by an implicit eq. like $\eta(x) = 0$

where $\eta: \mathbb{R}^n \rightarrow \mathbb{R}^{n-k}$

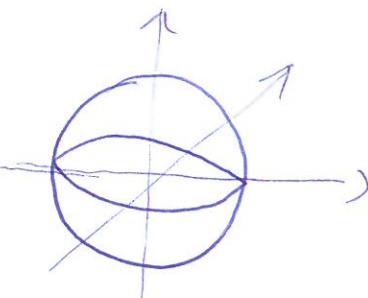
$$\mathcal{M} = \text{manifold} = \{x \in \mathbb{R}^n \mid \eta(x) = 0\}$$

example: sphere in \mathbb{R}^n : $S^n = \left\{ x \in \mathbb{R}^n \mid \|x\|_2^2 = 1 \right\}$
 $= \left\{ x \in \mathbb{R}^n \mid \sum_{i=1}^n x_i^2 = 1 \right\}$

is a $(n-1)$ -dimensional manifold

ex: S^2 in \mathbb{R}^3

2-dim manf. in \mathbb{R}^3



manifold is invariant (for $x \circ f(t)$) if

$$\eta(x(0)) = 0 \Rightarrow \eta(x(t)) = 0 \quad \forall t \geq 0$$

Consider nonlinear autonomous system

$$\dot{x} = f(x) \quad 0 = f(0) \quad f \in C^2$$

Compute linearization at $\bar{x} = 0$

$$\text{Jacob. } A = \left. \frac{\partial f}{\partial x} \right|_{\bar{x}=0}$$

rewrite system as

$$\dot{x} = Ax + \underbrace{\left(f(x) - \left. \frac{\partial f}{\partial x} \right|_{\bar{x}} (0)x \right)}_A = Ax + \tilde{f}(x)$$

$$\text{where by construction} \quad \begin{cases} \tilde{f}(0) = 0 \\ \left. \frac{\partial \tilde{f}}{\partial x} \right|_{\bar{x}} (0) = 0 \end{cases}$$

Since we are interested in the unstable case,

- A has
 - k eigenvalues with $\operatorname{Re}[\lambda_i] = 0$
 - $n-k$ eigenval. with $\operatorname{Re}[\lambda_i] < 0$

\Rightarrow make a change of basis T s.t.

$$TAT^{-1} = \begin{bmatrix} A_1 & 0 \\ 0 & A_2 \end{bmatrix} \quad \begin{cases} \operatorname{Re}[\lambda_i] = 0 & \text{k-dim} \\ \operatorname{Re}[\lambda_i] < 0 & (n-k)-\text{dim} \end{cases}$$

$$\Rightarrow \begin{bmatrix} y \\ z \end{bmatrix} = Tx$$

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system becomes:

$$(*) \quad \begin{cases} \dot{y} = A_1 y + g_1(y, z) \\ \dot{z} = A_2 z + g_2(y, z) \end{cases}$$

$$\text{with } \begin{cases} g_1(0, 0) = 0 \\ \frac{\partial g_1}{\partial y}(0, 0) = 0 \\ \frac{\partial g_1}{\partial z}(0, 0) = 0 \end{cases}$$

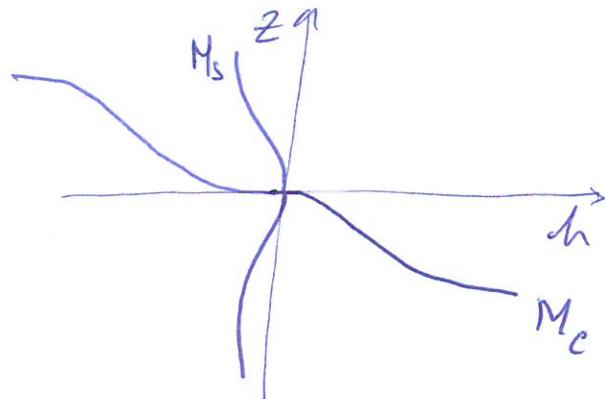
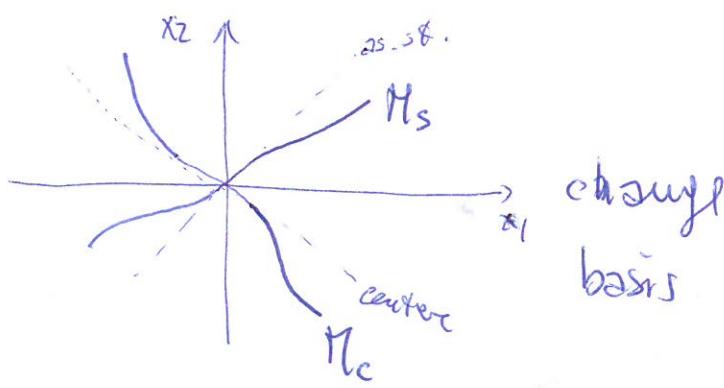
y is the "center" part

z is the asympt. st. part.

properties are valid
by construction from
 $\tilde{f}(0) = 0 \quad \frac{\partial \tilde{f}}{\partial x}(0) = 0$

def An invariant manifold $z = h(y)$ $h \in C^\infty$
for the system (*) is called a center manifold
if $h(0) = 0, \frac{\partial h}{\partial y}(0) = 0$

Meaning: ^{locally} I split \mathbb{R}^n into 2 vector subspaces, one for
the eigenvalues of A_1 (\rightarrow center), the other
for the eigenvalues of A_2 (\rightarrow as. st.) - then
I "project" these eigenvalues to manifolds



center manifold $\mathcal{M}_c = \{(y, z) \text{ s.t. } z = h(y)\}$

\mathcal{M}_c is tangent to $y=0$ at 0 (and passes through origin)
 $\Rightarrow h(0)=0$

$$\left\{ \begin{array}{l} \frac{\partial h}{\partial y}(0)=0 \\ \end{array} \right. \Rightarrow h(y) \text{ has terms on } y \text{ of power } 2 \text{ or higher (by construction)} \quad \text{by construction}$$

Measuring \mathcal{M}_s = stable manifold, is less important since it contains as. stable directions \Rightarrow what matters is what happens on the center manifold existence thm:

thm For the system (*) with $f_i \in C^2$ and satisfying (**) $\exists \delta > 0$ and $h \in C^\infty$ defined in $\|y\| < \delta$ s.t. $z = h(y)$ is a center manifold for the system.

Center manifold is invariant \Rightarrow
 $(y(0), z(0)) \in \mathcal{M}_c \quad \text{i.e. } z(0) = h(y(0)) \Rightarrow$
 $z(t) = h(y(t)) \quad \forall t \geq 0 \quad ((y(t), z(t)) \in \mathcal{M}_c \quad \forall t)$

\Rightarrow we can replace $z(t)$ with $h(y(t))$ in the equation

\Rightarrow reduced system (of dim $n - k$ aus drin in \mathcal{H}_c)

$$\ddot{y} = A_1 y + g_1(y, h(y))$$

this reduced system decides as. stability of the entire system. \rightarrow reduction principle

Thm If $\bar{y} = 0$ of the reduced system is asymptotically st. (unstable) then $\bar{x} = 0$ of the original system is as. st. (unstable)

Proof Idea: "outside \mathcal{H}_c " asymptotic convex. rules (even exponential)

i.e. if on a ball B_r $z^{(0)} \neq h(y^{(0)})$
then $(y(t), z(t))$ is not in \mathcal{H}_c

Denote $w(t) = z - h(y)$ the deviation from \mathcal{H}_c

then $\|w(t)\| \leq e^{\gamma t} \|w(0)\|$

$\Rightarrow \mathcal{H}_c$ is locally attracting as a set

\Rightarrow only what happens on \mathcal{H}_c matters //

the derivation w helps in finding an expression for M_c -

Change basis : $\begin{bmatrix} y \\ z \end{bmatrix} \rightarrow \begin{bmatrix} y \\ w \end{bmatrix}$ $z = w + h(y)$

$$\dot{y} = A_1 y + g_1(y, w + h(y))$$

$$\begin{aligned} \dot{w} &= \dot{z} - \frac{\partial h}{\partial y} \dot{y} = A_2(w + h(y)) + g_2(y, w + h(y)) \\ &\quad - \frac{\partial h}{\partial y} (A_1 y + g_1(y, w + h(y))) \end{aligned}$$

the center manifold M_c in the coord. $\begin{bmatrix} y \\ w \end{bmatrix}$ is $w=0$ - Hence being in M_c and staying in M_c means $w(t) = 0 \Rightarrow \dot{w}(t) = 0$ i.e.

$$0 = \dot{w}(t) = A_2 w(y) + g_2(y, h(y)) - \frac{\partial h}{\partial y} (A_1 y + g_1(y, h(y)))$$

this is a PDE that must be solved w.r.t. $h(y)$ in order to find the M_c , with boundary conditions $\begin{cases} h(0) = 0 \\ \frac{\partial h}{\partial y}(0) = 0 \end{cases}$

\rightarrow difficult!

Very often an explicit expression for $h(y)$ cannot be found. However $h(y)$ can be approximated via a Taylor expansion -

Thm If a function $\phi(y)$, with $\phi(0) = 0$
 $\frac{d\phi}{dy}(0) = 0$, C^2 can be found such that

$$\Psi(\phi(y)) = \Theta(\|y\|^p) \quad (\text{infinite small of order } p)$$

for some $p \geq 1$, then for $\|y\|$ sufficiently

small it is $h(y) - \phi(y) = \Theta(\|y\|^p)$

and the reduced system can be represented as

$$\dot{y} = A_1 y + g_1(y, \phi(y)) + \Theta(\|y\|^{p+1})$$

Asympt. stab (instab.) can be deduced from it

$p > 1$ because Taylor expansion of $h(y)$ starts at $p=2$ by construction

$$h(y) = k_1 y^2 + k_2 y^3 + \dots$$

or $h(y) = 0$, also possible

example

$$\overset{\circ}{x}_1 = x_1 x_2$$

$$\overset{\circ}{x}_2 = -x_2 + 2x_1^2$$

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$$A = \left. \frac{\partial f}{\partial x} \right|_0 = \begin{bmatrix} x_2 & x_1 \\ 2x_1 & -1 \end{bmatrix} \Big|_0 = \begin{bmatrix} 0 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} A_1 & 0 \\ 0 & A_2 \end{bmatrix}$$

$$\Rightarrow \begin{cases} y = x_1 \\ z = x_2 \end{cases} \quad \begin{cases} \dot{y} = 0 + yz \\ \dot{z} = -z + 2y^2 \end{cases}$$

$M_c = \{(y, z) \text{ s.t. } z = h(y)\}$ = center manifold
on M_c , reduced system is $\dot{y} = yh(y)$

Must find $z = h(y)$ that define center manifolds
use the PDE

$$\Phi(h(y)) = A_z h(y) + g_z(y, h(y)) - \frac{\partial h}{\partial y} (A_1 y + g_1(y, h(y))) = 0$$

$$-h(y) + 2y^2 - \frac{\partial h}{\partial y}(y h(y)) = 0 \quad \text{s.t. } h(0)=0 \\ \frac{\partial h}{\partial y}(0)=0$$

this PDE is in general impossible to solve in $h(y)$

\Rightarrow use Taylor expansion as approximation,

Simplest approxim: $\phi(y) = 0$

$$\Rightarrow \partial y^2 = 0$$

Ok only if $\alpha = 0$, not in general
sys is $\dot{y} = 0$ i.e. $\dot{y} = 0$ stable $\Rightarrow x = 0$ stable
but not asympt. st.

next simplest approximation: power 2 in h

$$\phi(y) = k_2 y^2 + \mathcal{O}(|y|^3)$$

k_2 unknown coeff. $\phi(y) = \phi(y) + \mathcal{O}(|y|^3)$

plug in into the PDE:

$$\Psi(\phi(y)) = -k_2 y^2 + \partial y^2 - 2k_2 y (y k_2 y^2) = 0$$

$$\text{If } k_2 = \alpha \Rightarrow \Psi(\phi(y)) = -2k_2^2 y^4 = -2\alpha^2 y^4 = \mathcal{O}(|y|^4)$$

\Rightarrow in the reduced system

approximation

$$\ddot{y} = y h(y) \approx \alpha y^3 + \mathcal{O}(|y|^4)$$

• when $\alpha > 0$ reduced system is unstable
 \Rightarrow orig syst. is unstable

• when $\alpha < 0$ reduced syst. is as. stab
 \Rightarrow orig syst. is as. st.

• when $\alpha = 0$

~~the~~ PDE is $-h(y) - \frac{\partial h(y)}{\partial y} (y h(y)) = 0$
 which is solved by $h(y) = 0$

\Rightarrow reduced syst. is $\dot{y} = 0$ which is stable
 (but not asympt. st.)

\Rightarrow original syst. is stable

Other concepts of stability

absolute, robust, practical, diagonal, semiglobal, ...

Systems given with inputs and outputs

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

$$y = h(x, u)$$

i) Input-state stability

For linear systems: BIBO (Bounded Input
Bounded Output) stability

$$\dot{x} = Ax + Bu$$

$$y = Cx$$

$$y(t) = \underbrace{C e^{At} x(0)}_{\text{free eval}} + \underbrace{\int_0^t e^{A(t-\tau)} B u(\tau) d\tau}_{\text{forced eval.}}$$

BIBO stab:

$$\|u\| < k_u \Rightarrow \|y\| < k_y$$

asympt. stab. of A \Rightarrow BIBO stability
(for minimal realiz: $A \in \mathbb{R}$)

to show it:

$$\|y(t)\| \leq \|c\| \underbrace{e^{-\lambda t}}_{\text{expon. stab.}} \|x(0)\| + \|c\| \frac{k}{\lambda} \underbrace{\|B\| \sup_u \|u\|}_{\text{boundedness}}$$

If nonlinear syst. with $u=0$ is expon. stab. then the same type of input-to-state stability behavior can be expected also for the nonlinear system

However you need to have also a global Lipschitz condition, which is difficult to get in practice

example $\dot{x} = -x + (x^2 + 1)u$

If $u=0$ then $x=0$ is exp. st. for $x^2 = -x$. However if $u=1$ then $\dot{x} = -x + x^2 + 1$ diverges to ∞ \Rightarrow unstable

"small signals" stability: for $\|x(0)\| < r_1$, $\|u(t)\| < r_2$

$$\|y\|_\infty \leq c_1 \|u\|_\infty + c_2 \|x(0)\|$$

2) Dissipative / passive systems

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

$$y = h(x, u)$$

associated with a supply rate function

$$w: \mathcal{U} \times \mathcal{Y} \rightarrow \mathbb{R}$$

Let system is dissipative w.r.t. the supply rate $w(u, y)$ if \exists a positive def. storage function $S: X \rightarrow \mathbb{R}_+$ s.t. the following dissipation inequality holds:

$$S(x(t)) \leq S(x(0)) + \int_0^t w(u(\tau), y(\tau)) d\tau$$

or, differentiating

$$\dot{S}(x) \leq w(u, y)$$

Measuring

- supply rate = "infinitesimal energy" supplied to the system (power flow into the syst.)
- storage function = energy stored in the syst.

- dissipation : stored energy is bounded by the total energy supplied externally
i.e. dissipative system cannot internally create energy -
- storage function is related to Lyapunov function

Thm In a dissipative system
 \checkmark If $w(x, y) \leq 0 \forall y$ and $x^* = 0$ is a minimum of S , then $x^* = 0$ is a locally stable eq. of $\dot{x} = f(x, 0)$ and $V(x) = S(x) - S(0) \geq 0$ is a Lyapunov func.

Proof \checkmark $\dot{V}(x) = \dot{S}(x) \leq w(x, y) \leq 0$ when $y = 0$

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21) Passive systems are a subclass of dissipative systems - They have supply rate function

$$w(u, y) = u^T y \quad (\text{implies: } n^{\circ} \text{ of inputs} = n^{\circ} + \text{outputs})$$

def the system $\dot{x} = f(x, u)$,
 $y = h(x, u)$ is

i) passive if it is dissipative w.r.t. the supply rate $w(u, y) = u^T y$
 i.e. $u^T y \geq \dot{S}(x)$

ii) lossless if $u^T y = \dot{S}(x)$

iii) strictly passive if $u^T y \geq \dot{S}(x) + \psi(x)$
 with $\psi(x)$ pos. def.

Meaning of passivity: energy that flows into the system $^{(u^T y)}$ is more than the increase of stored energy (\dot{S})

example: in a circuit with a resistor the power that flows in "changes" the capacitors and inductors, but gets lost in the resistors

example integrator $\begin{cases} \dot{x} = u \\ y = x \end{cases}$ $G(s) = \frac{1}{s}$

Supply rate $w(u, y) = uy$

Storage funct. $S(x) = \frac{1}{2}x^2 > 0$

$$\Rightarrow S(x) = \frac{1}{2}x\dot{x} = uy$$

\Rightarrow lossless system

example low-pass filter $\begin{cases} \dot{x} = -x + u \\ y = x \end{cases}$ $G(s) = \frac{1}{s+1}$

Supply rate $w(u, y) = uy$

Storage funct. $S(x) = \frac{1}{2}x^2$

$$\Rightarrow S(x) = x\dot{x} = -x^2 + uy$$

$\Rightarrow uy > S(x) \Rightarrow$ strictly passive $\cancel{\Rightarrow}$

$\Rightarrow \dot{x} = 0$ is globally as. st. when $u = 0$

thm System is passive $\left\{ \begin{array}{l} S(x) \geq 0 \\ \Rightarrow \dot{x} = 0 \text{ is } \cancel{\text{stab.}} \text{ when } u = 0 \end{array} \right.$

System is strictly passive $\left\{ \begin{array}{l} S(x) > 0 \\ \Rightarrow \dot{x} = 0 \text{ is } \cancel{\text{asympt. stab.}} \text{ when } u = 0 \end{array} \right.$