

PART 3: FEEDBACK LINEARIZATION

Consider a control-affine nonlinear sys

$$(*) \quad \dot{x} = f(x) + \sum_{i=1}^m g_i(x) u_i$$

and x_0 s.t. $f(x_0) = 0$

Control problem: find a feedback law $u(t) = \gamma(x(t))$ s.t. x_0 is (locally) asympt. stable

$u(t) = \gamma(x(t))$ = static state feedback

What kind of approach can we follow.

1) Jacobian linearization around x_0

$$A = \frac{\partial f(x)}{\partial x} \quad B = [g_1(x_0) \dots g_m(x_0)]$$

• If $\dot{x} = Ax + Bu$ is controllable

\Rightarrow choosing $u = Kx$ we can get

$\text{Re}[\lambda(A+BK)] < 0$ i.e. locally stabilize the nonlinear system

• If $\dot{x} = Ax + Bu$ stabilizable (i.e.

non-controllable modes have real part < 0) $\Rightarrow u = Kx$ can locally

stabilize nonlinear system also

in this case

$$A \rightarrow \begin{bmatrix} A_{11} & 0 \\ A_{21} & A_{22} \end{bmatrix}$$

$$B \rightarrow \begin{bmatrix} 0 \\ B_2 \end{bmatrix}$$

$$\text{Re}[\lambda(A_{11})] < 0$$

- advantage : simple!
- disadvantage : local result
different but not necessary
- mathematically not interesting ...

2) Lyapunov-based design

- use Lyapunov theory to construct control Lyapunov functions
- disadvantage : must find Lyapunov function!

3) today and next time : techniques based on exact linearization - 3 versions :

- i) state linearization = exact lineariz. by change of state variables
- ii) feedback linearization = exact linearization by change of state var and change of input (-> feedback equiv.)
- iii) input-output linearization : partial linearizat. of the dynamics via feedback equivalence

State Linearization

= linearization by means of a change of state

def (x) is linearizable by means of a change of state if \exists diffeomorphism $z = \Phi(x)$ s.t. $\Phi(0) = 0$ that transforms (x) into $\dot{z} = Az + Bu$

• Condition for this to happen

$$z = \Phi(x) \Rightarrow \dot{z} = \frac{\partial \Phi}{\partial x} \dot{x} = \frac{\partial \Phi}{\partial x} (f(x) + \sum g_i(x) u_i)$$

$$= Az + \sum b_i u_i \quad (B = [b_1 \dots b_m])$$

$$\Leftrightarrow \begin{cases} \frac{\partial \Phi(x)}{\partial x} f(x) = A \Phi(x) \\ \frac{\partial \Phi(x)}{\partial x} g_i(x) = b_i \end{cases}$$

set of PDEs that must be solved

• Either we use the notation

$$\left. \frac{\partial \Phi}{\partial x} f(x) \right|_{x = \Phi^{-1}(z)} = Az \quad (\text{and also } Ad_{\Phi^{-1}} f)$$

which is the same thing -

• Necessary and sufficient "geometric" conditions for state linearization

then the system (x) can be linearized to a controllable linear system at x_0 by means of a change of state iff near x_0 we have:

- 1) $\dim(\text{span}\{ad_f^k g_i(x), i=1, \dots, m, k=0, \dots, n-1\}) = n$
 $\forall x \in B(x_0)$
- 2) $[ad_f^k g_i, ad_f^l g_j] = 0 \quad \forall i, j=1, \dots, m, \forall k, l \geq 0$
 $\forall x \in B(x_0)$

Conditions are similar to STLC we saw earlier - (case in which LARC implies STLC)

Proof (idea)

Assume $\exists \Phi$ s.t.

$$\left. \frac{\partial \Phi}{\partial x} f \right|_{x=\Phi^{-1}(z)} = Az$$

$$\left. \frac{\partial \Phi}{\partial x} g_i \right|_{x=\Phi^{-1}(z)} = b_i$$

for some (A, B) controllable - then $[f, g_i]$ is transformed to

$$\left. \frac{\partial \Phi}{\partial x} [f, g_i] \right|_{x=\Phi^{-1}(z)} = \left[\left. \frac{\partial \Phi}{\partial x} f, \frac{\partial \Phi}{\partial x} g_i \right] \right|_{x=\Phi^{-1}(z)} = [Az, b_i]$$

$$= -Ab_i \quad (\text{we computed this earlier})$$

and similarly: $\frac{\partial \Phi}{\partial x} \left. \frac{\partial f^k}{\partial x} g_i \right|_{x=\Phi^{-1}(y)} = (-1)^k A^k b_i$ (1169)

\Rightarrow in order to have (A, B) controllable, cond 1) must hold

As for cond. 2):

$$\frac{\partial \Phi}{\partial x} \left[\frac{\partial f^k}{\partial x} g_i, \frac{\partial f^p}{\partial x} g_j \right] \Big|_{x=\Phi^{-1}(z)} = [(-1)^k A^k b_i, (-1)^p A^p b_j] = 0$$

since they are all const. v.f. //

example $x \in \mathbb{R}^2$

$$\dot{x}_1 = x_1 \ln x_2$$

$$\dot{x}_2 = -x_2 \ln x_1 + x_2 u$$

$$x_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$f(x) = \begin{bmatrix} x_1 \ln x_2 \\ -x_2 \ln x_1 \end{bmatrix} \quad g(x) = \begin{bmatrix} 0 \\ x_2 \end{bmatrix}$$

$$f(x_0) = 0 \quad (\Rightarrow \dim(\text{span}\{f(x_0), g(x_0)\}) = 1)$$

$$\begin{aligned} [f, g] &= \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \ln x_2 \\ -x_2 \ln x_1 \end{bmatrix} - \begin{bmatrix} \ln x_2 & \frac{x_1}{x_2} \\ -\frac{x_2}{x_1} & \ln x_1 \end{bmatrix} \begin{bmatrix} 0 \\ x_2 \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ -x_2 \ln x_1 \end{bmatrix} - \begin{bmatrix} x_1 \\ -x_2 \ln x_1 \end{bmatrix} = \begin{bmatrix} -x_1 \\ 0 \end{bmatrix} \end{aligned}$$

$$\text{span}\{g, \frac{\partial f}{\partial x} g\} \Big|_{x_0} = \text{span}\left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \end{bmatrix} \right\} \text{ has dim } 2$$

\Rightarrow cond 1) is satisfied

cond 2) is automatically satisfied since we have a scalar control. $m=1$.

\rightarrow system can be linearized by change of state

in this case:

$$z = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \Phi(x) = \begin{bmatrix} \ln x_1 \\ \ln x_2 \end{bmatrix}$$

(1170)

to verify it (and compute A, b)

$$\bullet \frac{\partial \Phi(x)}{\partial x} f(x) = A \Phi(x) \quad \text{i.e.}$$

$$\begin{bmatrix} \frac{1}{x_1} & 0 \\ 0 & \frac{1}{x_2} \end{bmatrix} \begin{bmatrix} x_1 \ln x_2 \\ -x_2 \ln x_1 \end{bmatrix} = \begin{bmatrix} \ln x_2 \\ -\ln x_1 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}}_A \underbrace{\begin{bmatrix} \ln x_1 \\ \ln x_2 \end{bmatrix}}_{\Phi(x)}$$

$$\bullet \frac{\partial \Phi}{\partial x} g(x) = b$$

$$\begin{bmatrix} \frac{1}{x_1} & 0 \\ 0 & \frac{1}{x_2} \end{bmatrix} \begin{bmatrix} 0 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} = b$$

alternatively: compute $\Phi^{-1}(z) = \begin{bmatrix} e^{z_1} \\ e^{z_2} \end{bmatrix}$

and compute

$$\begin{aligned} \dot{z} &= \frac{\partial \Phi}{\partial x} \dot{x} = \frac{\partial \Phi}{\partial x} (f(x) + g(x)u) \Big|_{x=\Phi^{-1}(z)} \\ &= \begin{bmatrix} \frac{1}{x_1} & 0 \\ 0 & \frac{1}{x_2} \end{bmatrix} \left(\begin{bmatrix} x_1 \ln x_2 \\ -x_2 \ln x_1 \end{bmatrix} + \begin{bmatrix} 0 \\ x_2 \end{bmatrix} u \right) \Big|_{x=\begin{bmatrix} e^{z_1} \\ e^{z_2} \end{bmatrix}} \end{aligned}$$

resulting linear system is

$$\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = -z_1 + u \end{cases}$$

Remark: state linearization (exact) \neq Jacobian linearization
(in which you neglect higher order terms)

Actually, in this example exact lineariz.

and Jacobian lineariz. at x_0 coincide!

$$A_0 = \frac{\partial f}{\partial x} \Big|_{x_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}} = \begin{bmatrix} \ln x_2 & \frac{x_1}{x_2} \\ -\frac{x_2}{x_1} & -\ln x_1 \end{bmatrix} \Big|_{x_0} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

$$B_0 = g(b) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

however if you compute Jacobian lin its value is only local, while the exact lineariz. ~~has~~ is valid where $\bar{\Phi}(0)$ is a diffeomorphism i.e. on the entire

$\mathcal{H} = \mathbb{R}_+^2$ (open) where the system is defined

Feedback linearization

(1171)

motivating example :

mechanical system (e.g. robotic arm)

$$M(q) \ddot{q} + C(q, \dot{q}) \dot{q} + g(q) = u$$

where $M(q) > 0$ matrix of inertia / masses

$C(q, \dot{q})$ = matrix of Coriolis forces

$g(q)$ = gravitational forces

rewrite it as 1st order order ODEs :

$$\frac{d}{dt} \begin{bmatrix} q \\ \dot{q} \end{bmatrix} = \begin{bmatrix} \dot{q} \\ M^{-1}(q) (-C(q, \dot{q}) \dot{q} - g(q) + u) \end{bmatrix}$$

If I choose a change of feedback

$$u = C(q, \dot{q}) \dot{q} + g(q) + M(q) v \quad \text{where } v = \text{new input}$$

→ all dynamics are canceled and I get

$$\frac{d}{dt} \begin{bmatrix} q \\ \dot{q} \end{bmatrix} = \begin{bmatrix} \dot{q} \\ v \end{bmatrix} \quad \rightarrow \text{linear system of 2n eq.}$$

→ also a change of feedback can linearize the dynamics!

change of feedback is a feedback (i.e. requires to know the state!) It is also called prefeedback or feedforward control

what kind of nonlinearities can I eliminate in this way?

1) input-multiplicative invertible nonlinearities

$$\begin{aligned} \dot{x} &= Ax + f(x)u && \text{with } f(x) = B\theta(x) \\ &= Ax + \sum_i \beta_i(x)u_i \end{aligned}$$

B n x m matrix
 $\theta(x)$ m x m matrix
invertible

then choosing

$$u = \beta(x)v \quad \text{with } \beta(x) = \theta^{-1}(x)$$

leads to

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

2) additive nonlinearities

$$\dot{x} = Ax + B(\gamma(x) + u) \quad \gamma(x) \text{ n x 1 vector f.}$$

then $u = -\gamma(x) + v$ leads to

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

change of feedback alone can be used to linearize systems of the form

$$\dot{x} = Ax + B\theta(x)(\gamma(x) + u)$$

by choosing $u(x) = \alpha(x) + \beta(x)v$
with $\alpha(x) = -\gamma(x)$
 $\beta(x) = \theta^{-1}(x)$

$$\begin{aligned} \Rightarrow \dot{x} &= Ax + B\theta(x)(\gamma(x) - \gamma(x) + \theta^{-1}(x)v) \\ &= Ax + Bv \end{aligned}$$

Combining change of state $\Phi(x)$ and change of feedback $u = \alpha(x) + \beta(x)v$ we get a full feedback equivalence transform

(1173)

Consider $\dot{x} = f(x) + \sum_{i=1}^m g_i(x) u_i \quad (*)$

use • change of state $z = \Phi(x)$ diffeom.

• change of input $u = \alpha(x) + \beta(x)v$

$\beta(x)$ invertible

we get

$$\dot{z} = \left(\frac{\partial \Phi}{\partial x} f(x) + \frac{\partial \Phi}{\partial x} \sum g_i(x) u_i \right) \Bigg|_{\substack{x = \Phi^{-1}(z) \\ u_i = \alpha_i(\Phi^{-1}(z)) + \beta_i(\Phi^{-1}(z))v}}$$

(from which $v_i = \frac{u_i - \alpha_i(x)}{\beta_i(x)}$)

plug in the input u_i

$$= \frac{\partial \Phi}{\partial x} f(x) + \sum \frac{\partial \Phi}{\partial x} g_i(x) (\alpha_i(x) + \beta_i(x) v_i)$$

$$= Az + Bv$$

which implies (taking $z = \Phi(x)$ in this last expr.)

$$\left. \begin{aligned} & \frac{\partial \Phi}{\partial x} f(x) + \sum \frac{\partial \Phi}{\partial x} g_i(x) \alpha_i(x) = A \Phi(x) \\ & \frac{\partial \Phi}{\partial x} g_i(x) \beta_i(x) = b_i \end{aligned} \right\}$$

→ system of PDE (difficult to solve!)

Also in this case \exists geometric character. (1174)

From (*) construct the chain of distrib.

$$D_1 = \text{span} \{ g_1, \dots, g_m \}$$

$$D_2 = \text{span} \{ g_i, \text{ad}_f g_i, i=1, \dots, m \}$$

$$D_3 = \text{span} \{ g_i, \text{ad}_f g_i, \text{ad}_f^2 g_i, i=1, \dots, m \}$$

$$\vdots$$
$$D_k = \text{span} \{ \text{ad}_f^j g_i, i=1, \dots, m, j=0, 1, \dots, k-1 \}$$

this is clearly growing $D_1 \subset D_2 \subset \dots \subset D_k$

thm Consider (*) with $f(x_0) = 0$ and assume local accessibility at x_0 . Then (*) is feedback ~~equivalent~~ linearizable around x_0 iff the distributions D_1, D_2, \dots, D_k are all involutive and constant dimensional near x_0 . The resulting linear system is controllable.

Proof (idea) Similar to thm for state linearization, in spirit: the input v.f. g_i and $\text{ad}_f^k g_i$ have to "behave" like constant v.f.

For the linear system $\dot{z} = Az + Bv$, the dist

D_k^{linear} and $D_k^{linear} = \text{Im} [B AB \dots A^{k-1} B]$

which are all involutive (because they are const. v. f.)

Since $D_k = \frac{\partial \Phi}{\partial x} D_k^{linear}$ with $\frac{\partial \Phi}{\partial x}$ full rank then so must be also the D_k

When we apply also the feedback $u = \alpha(x) + \beta(x)v$ we pass from f, g_i to

$\tilde{f}(x) = f(x) + \sum g_i(x) \alpha_i(x)$
 $\tilde{g}_i(x) = g_i(x) \beta_i(x)$

One should show that $D_k^{(x)} = \tilde{D}_k(x)$ for the new distribution (tedious...)

meaning: D_k involutive are feedback invariants (i.e. invariant to the feedback transformations $u = \alpha(x) + \beta(x)v$, but only when D_k are themselves involutive, not in general //

$D_k \leftrightarrow \tilde{D}_k \leftrightarrow \frac{\partial \Phi}{\partial x} \tilde{D}_k = D_k^{linear}$

example

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

$$\dot{x}_2 = -x_1^2 + u$$

$$x_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

this cannot be linearized by simple change of feedback, like $u = x_1^2 + v$ which leads to

$$\begin{cases} \dot{x}_1 = \delta \sin x_2 \\ \dot{x}_2 = v \end{cases}$$

Let us verify that the system is feedback linearizable

$D_1 = \text{span}\{g\} = \text{span}\left\{\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right\}$ has dim = 1 everywhere and is involutive by const.

$$[f, g] = \text{ad}_f g = 0 - \begin{bmatrix} 0 & \delta \cos x_2 \\ -2x_1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} \delta \cos x_2 \\ 0 \end{bmatrix}$$

$D_2 = \text{span}\{g, \text{ad}_f g\} = \text{span}\left\{\begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} \delta \cos x_2 \\ 0 \end{bmatrix}\right\}$ has dim = 2 and const. rank around

$$x_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$\Rightarrow D_2$ involutive

\Rightarrow we also have LARC cond

\Rightarrow the syst. is feedback linearizable

In fact, we can continue with the state change

$$z = \Phi(x) = \begin{bmatrix} x_1 \\ \gamma \sin x_2 \end{bmatrix} \Rightarrow \frac{\partial \Phi}{\partial x} = \begin{bmatrix} 1 & 0 \\ 0 & \gamma \cos x_2 \end{bmatrix}$$

$$\Rightarrow \dot{z} = \frac{\partial \Phi}{\partial x} \dot{x} = \begin{bmatrix} 1 & 0 \\ 0 & \gamma \cos x_2 \end{bmatrix} \begin{bmatrix} \gamma \sin x_2 \\ -x_1^2 + u \end{bmatrix} \Big|_{x = \Phi^{-1}(z)}$$

$$= \begin{bmatrix} \gamma \sin(x_2) \\ \gamma \cos(x_2) (-x_1^2 + u) \end{bmatrix} \Big|_{x = \Phi^{-1}(z)}$$

If we choose now

$$u = x_1^2 + \frac{v}{\gamma \cos x_2}$$

and use $x = \Phi^{-1}(z) = \begin{bmatrix} z_1 \\ \arcsin(\frac{z_2}{\gamma}) \end{bmatrix}$ (easy to compute in this case...)

$$\Rightarrow \dot{z} = \begin{bmatrix} \gamma \sin x_2 \\ v \end{bmatrix} \Big|_{x = \Phi^{-1}(z)} = \begin{bmatrix} \gamma \sin(\arcsin(\frac{z_2}{\gamma})) \\ v \end{bmatrix} = \begin{bmatrix} z_2 \\ v \end{bmatrix}$$

↑
simpler after
change of feedback

⇒ linear system

(Here I do not need $\Phi^{-1}(z)$. Enough to observe that $z_2 = \gamma \sin x_2$ and plug it into $\dot{z} = \begin{bmatrix} \gamma \sin x_2 \\ v \end{bmatrix}$)

For $\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = v \end{cases}$ I can now design a stabilizing linear controller $v = Kz = K\Phi(x)$

⇒ $u = \alpha(x) + \beta(x) \frac{K\Phi(x)}{\gamma \cos x_2} = x_1^2 + \frac{K\Phi(x)}{\gamma \cos x_2}$ is the complete feedback that stabilizes the original system.

Input-Output feedback Linearization

motivating example: consider system previous example, but with output

$$\begin{cases} \dot{x}_1 = \gamma \sin x_2 \\ \dot{x}_2 = -x_1^2 + u \\ y = x_2 \end{cases} \leftarrow \text{linear output eq.}$$

If we consider ~~only~~ the feedback ^{linearizing} ~~change~~ just mentioned

$$z = \Phi(x) = \begin{bmatrix} x_1 \\ \gamma \sin x_2 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \Phi^{-1}(z) = \begin{bmatrix} z_1 \\ \arcsin\left(\frac{z_2}{\gamma}\right) \end{bmatrix}$$

$$u = x_1^2 + \frac{v}{\gamma \cos x_2}$$

then the system + output in the new basis and input is

$$\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = v \\ y = \arcsin\left(\frac{z_2}{\gamma}\right) \end{cases}$$

→ nonlinear output eq.

Sometimes it might be useful to linearize only the input-output dependence not the entire state vector.

This is particularly true when the residual nonlinear part is not influencing the output.

If I just use

$$u = x_1^2 + v \quad \text{then}$$

$$\left. \begin{cases} \dot{x}_1 = \gamma \sin x_2 \\ \dot{x}_2 = v \\ y = x_2 \end{cases} \right\} \begin{array}{l} \text{this part of the dynamics} \\ \text{does not affect the output.} \\ \text{input-output part is} \\ \text{linear} \end{array}$$

→ the change of input has decoupled x_1 from the output i.e. it has made it unobservable (we did not discuss this notion in class but it corresponds to its linear counterpart) - //

Next: we try to build a general theory for input-output feedback linearization, based on these principles.

Assumption : SISO system

$$(*) \begin{cases} \dot{x} = f(x) + g(x)u \\ y = h(x) \end{cases} \quad \begin{array}{l} f, g, h \in C^\infty \\ x_0 \text{ s.t. } f(x_0) = 0 \end{array}$$

y does not depend on u : let us differentiate it until an explicit dependence on u appears -

$$\begin{aligned} \dot{y} &= \frac{\partial h}{\partial x} \dot{x} = \frac{\partial h}{\partial x} (f(x) + g(x)u) \\ &= L_f h(x) + L_g h(x)u \end{aligned} \quad \text{using Lie deriv.}$$

• if $L_g h(x) \neq 0 \quad \forall x \in B(x_0)$

$$\text{then } u(x) = \frac{1}{L_g h(x)} (-L_f h(x) + v)$$

yields the first order linear system

$$\dot{y} = v$$

⇒ we have achieved input-output linearization : part of the state vector ~~space~~ can be changed to get

$$\begin{cases} \dot{z}_1 = v \\ y = z_1 \end{cases} \quad \left. \begin{array}{l} \end{array} \right\} \text{1st order integrator} \\ \text{+ rest of the system...}$$

• If instead $L_g h(x) = 0 \quad \forall x \in B(x_0)$
then differentiate again

$$\begin{aligned} \ddot{y} &= \frac{\partial}{\partial x} (L_f h(x)) \dot{x} \\ &= \frac{\partial}{\partial x} (L_f h) f + \frac{\partial}{\partial x} (L_f h) g u \\ &= L_f^2 h + L_g L_f h u \end{aligned}$$

• If $L_g L_f h \neq 0$ then choose input

$$u = \frac{1}{L_g L_f h} (-L_f^2 h + v)$$

$\Rightarrow \ddot{y} = v$ double integrator

\Rightarrow part of the state vector can become

$$\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = v \\ y = z_1 \end{cases}$$

• If $L_g L_f h = 0$ then differentiate again

• Call $r =$ smallest integer for which
 $L_g L_f^{r-1} h(x) \neq 0 \quad \forall x \in B(x_0)$

~~Let the system (*) is said to have relative degree r~~

def the system (*) is said to have relative degree r at x_0 if

$$L_g L_f^k h(x) = 0 \quad \forall x \in B(x_0), k=0, 1, \dots, r-2$$

$$L_g L_f^{r-1} h(x) \neq 0 \quad \forall x \in B(x_0)$$

the input-output system is then

$$y^{(r)} = L_f^r h(x) + L_g L_f^{r-1} h(x) u$$

or written in state space form (r -dimensional)

$$\begin{cases} \dot{\xi}_1 = \xi_2 \\ \dot{\xi}_2 = \xi_3 \\ \vdots \\ \dot{\xi}_{r-1} = \xi_r \\ \dot{\xi}_r = L_f^r h(x) + L_g L_f^{r-1} h(x) u \\ y = \xi_1 \end{cases}$$

Using the feedback $u = \frac{1}{L_g L_f^{r-1} h} (-L_f^r h(x) + v)$

It becomes

$$y^{(r)} = v$$

or, in state space form (r -dimensional)

$$\dot{\xi} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix} \xi + \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 1 \end{bmatrix} v$$

$$y = [1 \ 0 \ \dots \ 0] \xi$$

this is a linear system of rel. degree = r
 (this is a chain of integrators)

^{SISO}
In linear systems: relative degree = 1183
= excess of poles w.r.t. zeros

From $y^{(r)} = v$

in the Laplace domain, the transfer function is

$$Y(s) = \frac{1}{s^r} U(s)$$

\Rightarrow rel. degree is r -

What is the partial change of state variables that accomplish this?

By construction we can take

$$\phi_1(x) = h(x)$$

$$\phi_2(x) = L_f h(x)$$

\vdots

$$\phi_r(x) = L_f^{r-1} h(x)$$

(i.e. y and its first r derivatives)

These are "good" candidates for a change of basis

Prop The differentials

$$dh(x) = \frac{\partial h}{\partial x}, \quad dL_f h(x) = \frac{\partial L_f h(x)}{\partial x}, \dots$$

$$\dots dL_f^{r-1} h(x) = \frac{\partial L_f^{r-1} h(x)}{\partial x} \text{ are linearly}$$

independent ("like rows of a Jacobian $n \times n$ ")

To complete the change of basis we must choose other $\phi_{r+1}(x) \dots \phi_n(x)$ so that

$$\begin{bmatrix} \eta \\ \eta \end{bmatrix} = \Phi(x) = \begin{bmatrix} \phi_1(x) \\ \vdots \\ \phi_r(x) \\ \phi_{r+1}(x) \\ \vdots \\ \phi_n(x) \end{bmatrix} \quad \text{the new state}$$

the Jacobian $\frac{\partial \Phi}{\partial x}$ must be full rank near x_0
 $\forall x \in B(x_0)$

One way (choice is not unique) to choose $\phi_{r+1}(x) \dots \phi_n(x)$ is such that the input u does not appear in the ODEs for η

Prop $\exists \phi_{r+1}(x), \dots, \phi_n(x)$ s.t. $\Phi(x)$ is ^{diffeomorphism} nonsingular at x_0 and $L_g \phi_i(x) = 0 \forall i = r+1, \dots, n$
 $\forall x \in B(x_0)$

meaning: $\Delta_g(x) = \text{span}\{g(x)\}$ is a regular distribution at $x \in B(x_0)$, of dim 1
 \Rightarrow it is involutive $\Rightarrow \exists$ ~~max~~ integral ~~function~~ manifold of dim 1
 $\Rightarrow \exists$ $n-1$ functions $\lambda_1, \dots, \lambda_{n-1}$ s.t.

$\frac{\partial \lambda_i}{\partial x}$ $i=1, \dots, n-1$ linearly indep.

$$L_g \lambda_i(x) = \frac{\partial \lambda_i}{\partial x} g = 0 \quad i=1, \dots, n-1$$

→ enough to choose for $\phi_i(x)$ any $n-r$ of these functions $\lambda_i(x)$, s.t. $\Phi(x) = \begin{bmatrix} \lambda(x) \\ L_f \phi(x) \\ L_f^{i-1} \phi(x) \\ \vdots \\ \phi(x) \end{bmatrix}$ is differentiable

Then $\frac{\partial \phi_i(x)}{\partial x} \perp g(x)$ for all $i=r+1, \dots, n$

→ when we apply the change of basis the input u does not appear

In fact, if $\eta_i = \phi_i(x)$ $i=r+1, \dots, n$

$$\dot{\eta}_i = \frac{d}{dt} \phi_i(x) = \frac{\partial \phi_i}{\partial x} \dot{x} = \frac{\partial \phi_i}{\partial x} (f(x) + g(x)u)$$

$$= \frac{\partial \phi_i}{\partial x} f(x) + \frac{\partial \phi_i}{\partial x} g(x) u$$

= 0 by construction

$$\left(= L_f \phi_i(x) + \underbrace{L_g \phi_i(x)}_{=0} u \right)$$

⇒ in the new basis $\begin{bmatrix} z \\ \eta \end{bmatrix} = \Phi(x)$

we have the system

$$\left\{ \begin{array}{l} \dot{z}_1 = z_2 \\ \vdots \\ \dot{z}_{r-1} = z_r \\ \dot{z}_r = L_f^r h(x) + L_g L_f^{r-1} h(x) u = u \\ \dot{\eta} = q(z, \eta) \\ y = z_1 \end{array} \right. \quad \left. \begin{array}{l} \\ \\ \\ \\ \end{array} \right\} \begin{array}{l} \\ \\ \\ n-r \text{ dimensional} \\ \end{array}$$

this is called the normal form of a nonlinear siso system

Notice how the η part of the state vector does not enter into the output eq. ⇒ the η -subsystem is unobservable from the given output.

Change of basis: first part (known)

$$\xi_1 = \phi_1(x) = h(x) = x_3$$

$$\xi_2 = \phi_2(x) = L_f h(x) = x_2$$

while for the second part $g \cdot \phi_3(x)$ we must find function $\phi_3(x)$ s.t. $L_g \phi_3(x) = 0$

$$\begin{aligned} L_g \phi_3 &= \frac{\partial \phi_3}{\partial x} g(x) = \frac{\partial \phi_3}{\partial x} \begin{bmatrix} \exp(x_2) \\ 1 \\ 0 \end{bmatrix} = \\ &= \begin{bmatrix} \frac{\partial \phi_3}{\partial x_1} & \frac{\partial \phi_3}{\partial x_2} & \frac{\partial \phi_3}{\partial x_3} \end{bmatrix} \begin{bmatrix} \exp(x_2) \\ 1 \\ 0 \end{bmatrix} = \frac{\partial \phi_3}{\partial x_1} \exp(x_2) + \frac{\partial \phi_3}{\partial x_2} = 0 \end{aligned}$$

\Rightarrow must solve a PDE (may be difficult in general...)

In this case it is easy:

$$\phi_3(x) = 1 + x_1 - \exp(x_2)$$

In fact

$$L_g \phi_3 = \frac{\partial \phi_3}{\partial x} g = \begin{bmatrix} 1 & -\exp(x_2) & 0 \end{bmatrix} \begin{bmatrix} \exp(x_2) \\ 1 \\ 0 \end{bmatrix} = 0$$

the whole change of basis is then

$$\begin{bmatrix} \xi_1 \\ \xi_2 \\ \eta \end{bmatrix} = \underline{\Phi}(x) = \begin{bmatrix} \phi_1(x) \\ \phi_2(x) \\ \phi_3(x) \end{bmatrix} = \begin{bmatrix} x_3 \\ x_2 \\ 1 + x_1 - \exp(x_2) \end{bmatrix}$$

Its Jacobian $\frac{\partial \Phi}{\partial x} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & -\exp(x_2) & 0 \end{bmatrix}$ has full rank $\forall x \Rightarrow$ global transf.

Its inverse is also easy to compute $\Phi^{-1}(z)$

$$\begin{cases} x_1 = -1 + \eta + \exp(\xi_2) \\ x_2 = \xi_2 \\ x_3 = \xi_1 \end{cases}$$

System in normal form is

$$\begin{cases} \dot{\xi}_1 = \xi_2 \\ \dot{\xi}_2 = L_f^2 h(x) + L_g L_f h(x) u = x_1 x_2 + u \\ \dot{\eta} = L_f \phi_3(x) = \frac{\partial \phi_3}{\partial x} f(x) = [1 - \exp(x_2) \ 0] \begin{bmatrix} -x_1 \\ x_1 x_2 \\ x_2 \end{bmatrix} \\ \quad = -x_1 - \exp(x_2) x_1 x_2 \end{cases}$$

on the r.h.s. I must now replace x with (ξ_1, η)

$$\begin{cases} \dot{\xi}_1 = \xi_2 \\ \dot{\xi}_2 = (-1 + \eta + \exp(\xi_2)) \xi_2 + u = v \end{cases}$$

this I can skip computing if I apply first the change of input!

$$\begin{cases} \dot{\eta} = +1 - \eta - \exp(\xi_2) - \exp(\xi_2) (-1 + \eta + \exp(\xi_2)) \xi_2 \\ \quad = (1 - \eta - \exp(\xi_2)) (1 + \xi_2 \exp(\xi_2)) \end{cases}$$