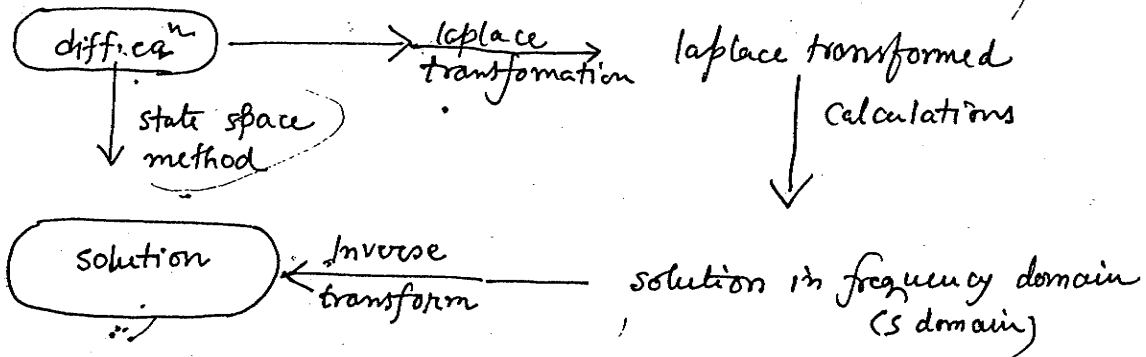


Prof. Mayer

Previous analysis of closed loops:



Earlier Exclusively frequency domain methods
 since the beginning of the 60's; also state space methods

- strategy in the case of state space methodology:

- * one remains in the time domain ✓
- * representation via systems of diff. eqⁿs ✓
- * descriptions through state variables from the state space ✓
- * the entirety of the state variables from the state space ✓

- Advantages of the state space representation:

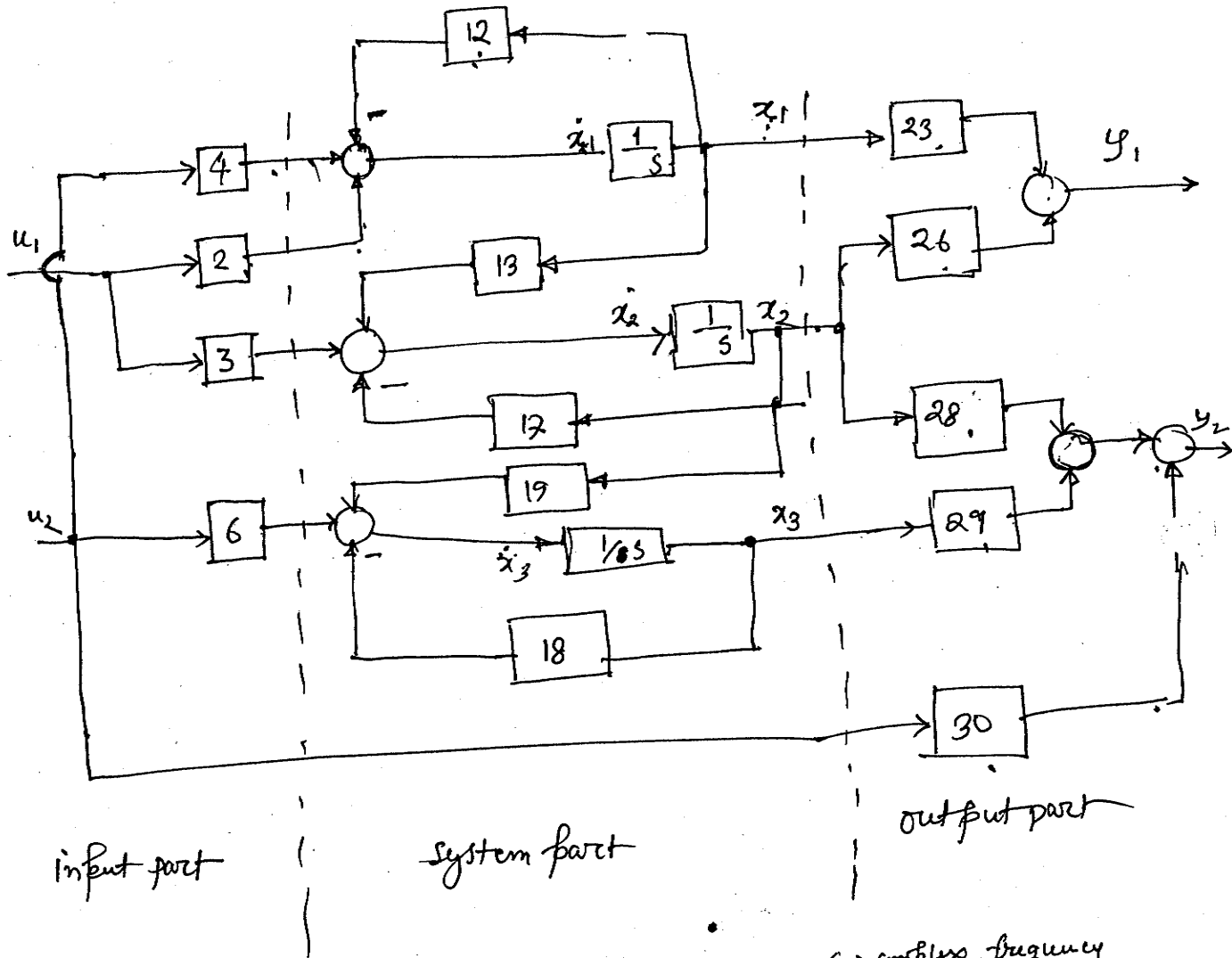
- * Applicability also on non linear and timevariant systems
- * further insights into the system behaviour are available
- * internal processes can be more precisely observed ✓
- * efficient treatment for multivariable systems ✓
- * easy transferability to numeric computation methods on digital computers ✓

It depends on the particular tasks, whether it is more advantageous to perform the analysis of control circuits either in the frequency domain or in the state space description.

Time domain
 state space description

Transformation into state space description

Example: linear time invariant multivariable system



$s \rightarrow$ complex frequency

Setting up the differential eqⁿs:

$$\begin{aligned} \dot{x}_1 &= -12x_1 + 2u_1 + 4u_2 \\ \dot{x}_2 &= 13x_1 - 17x_2 + 3u_1 \\ \dot{x}_3 &= 19x_2 - 18x_3 + 6u_2 \end{aligned}$$

Output equations:

$$\begin{aligned} y_1 &= 23x_1 + 26x_2 \\ y_2 &= 28x_2 + 29x_3 + 30u_2 \end{aligned}$$

\downarrow 2×3 \leftarrow column rows

System altogether is of 3rd order

General Notation:

square matrix

$$\begin{bmatrix} \dot{x}_1(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \underbrace{\begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}}_{\underline{A}} \cdot \underbrace{\begin{bmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{bmatrix}}_{\underline{x}(t)} + \underbrace{\begin{bmatrix} b_{11} & \dots & b_{1p} \\ \vdots & & \vdots \\ b_{m1} & \dots & b_{mp} \end{bmatrix}}_{\underline{B}} \cdot \underbrace{\begin{bmatrix} u_1(t) \\ \vdots \\ u_p(t) \end{bmatrix}}_{\underline{u}(t)}$$

$$\underbrace{\begin{bmatrix} y_1(t) \\ \vdots \\ y_2(t) \end{bmatrix}}_{\underline{y}(t)} = \underbrace{\begin{bmatrix} c_{11} & \dots & c_{1n} \\ \vdots & & \vdots \\ c_{21} & \dots & c_{2n} \end{bmatrix}}_{\underline{C}} \cdot \underbrace{\begin{bmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{bmatrix}}_{\underline{x}(t)} + \underbrace{\begin{bmatrix} d_{11} & \dots & d_{1p} \\ \vdots & & \vdots \\ d_{q1} & \dots & d_{qp} \end{bmatrix}}_{\underline{D}} \cdot \underbrace{\begin{bmatrix} u_1(t) \\ \vdots \\ u_p(t) \end{bmatrix}}_{\underline{u}(t)}$$

→ brief notation →

state Equations:

$$\begin{cases} \dot{\underline{x}}(t) = \underline{A} \cdot \underline{x}(t) + \underline{B} \cdot \underline{u}(t) \\ \underline{y}(t) = \underline{C} \cdot \underline{x}(t) + \underline{D} \cdot \underline{u}(t) \end{cases}$$

Representation in state space

Denotation:

- upper eqⁿ: "state differential equation"
- lower eqⁿ: "output eqⁿ"
- both eqⁿs together: "state eqⁿs"

Representation in state space

Denotation matrices:

- A "evolution matrix" $\dim[A] = n \times n \leftrightarrow$
The evolution matrix is a square matrix
- B "control matrix" $\dim[B] = n \times p \leftrightarrow$
- C "observation matrix" $\dim[C] = q \times n \leftrightarrow$
- D "Direct transmission matrix" $\dim[D] = q \times p \leftrightarrow$

In case of linear timeⁱⁿvariant systems the matrices A, B, C and D consist exclusively of constant elements.

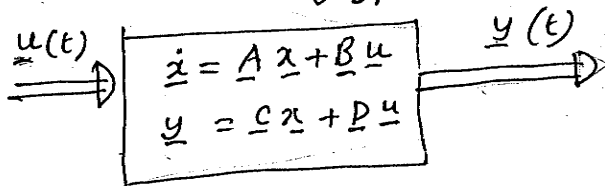
Important:

The input vector $u(t)$ consists only of the input variables, not of their derivatives.
 \Rightarrow the state variables in $x(t)$ have no irregularities

Annotation:

Thereby it is assumed that at the system entry step functions, but never δ -impulse functions may be assigned.

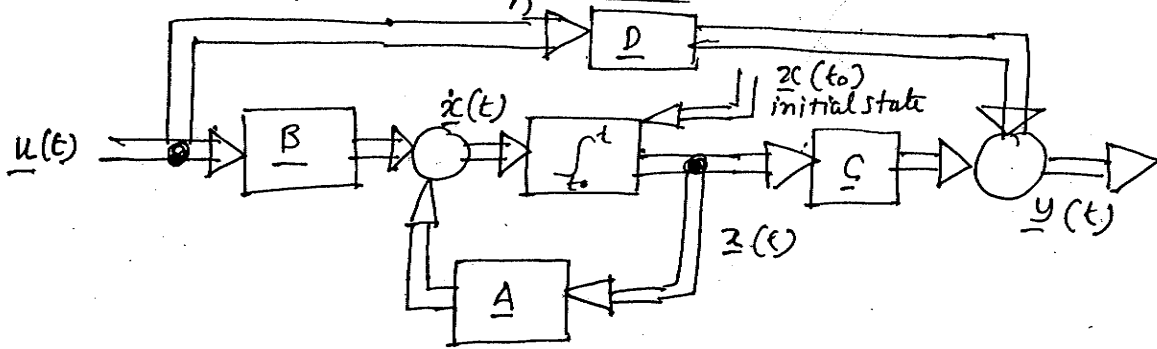
System definition: \Downarrow $\begin{matrix} x(t_0) \\ \text{Initial} \\ \text{state} \end{matrix}$



double lines clarify the transfer of vector variables.

special case: In state space description, which is pointed out here, time delay elements cannot be taken into consideration.

Flow diagrams of the state equations:



In practice the following applies:

- $p \leq n$: The number n of state variables in most cases is considerably higher than the number p of input variables.
- $q = p$: Furthermore, the number q of output variables in most cases is equal to the number p of input variables.
- $D = 0$: As input variables act normally only indirectly via the state vector x on the output variables, here the direct transmission matrix consists exclusively of zero elements.

- State variables
- Input variables
- Output variables

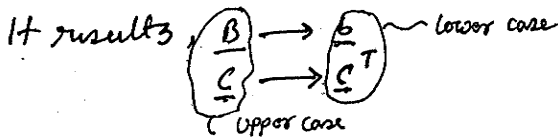
Contd.

Example: Single input/output system of a third order system without direct signal transmission.

If applies: $n > 1$ (Here $n=3$)
 $q=p=1$
 $D=0$

$$\dot{\underline{x}}(t) = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \underline{x}(t) + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} u(t)$$

$$y(t) = [c_1 \ c_2 \ c_3] \underline{x}(t)$$



$$c = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}, \quad c^T = [c_1 \ c_2 \ c_3]$$

(Capital letters for matrices; lower case for column vectors)
 This does not apply to signals!

Brief notation:

$$\begin{cases} \dot{\underline{x}} = \underline{A} \cdot \underline{x} + \underline{b} u \\ y = \underline{c}^T \cdot \underline{x} \end{cases}$$

in practice often occurring form of the state eq^{ns}

22 Derivation of the state equations from a linear differential equation of higher order:

Example: Differential equation of second order

$$\ddot{y}(t) + c_2 \dot{y}(t) + c_1 y(t) = c_0 u(t)$$

Introduction of state variables: $x_1(t) = y(t) \rightarrow \boxed{\dot{x}_1(t) = \dot{y}(t)}$
 $x_2(t) = \dot{y}(t) \rightarrow \dot{x}_2(t) = \ddot{y}(t)$

$$x_1(t) = x_2(t) \rightarrow$$

$$\dot{x}_2(t) = -c_2 x_2(t) - c_1 x_1(t) + c_0 u(t) \rightarrow$$

in vector notation (from the above two differential eq^{ns})

$$\dot{\underline{x}}(t) = \begin{bmatrix} 0 & 1 \\ -c_1 & -c_2 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 0 \\ c_0 \end{bmatrix} u(t)$$

Output eqⁿ:

$$y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} x(t)$$

A differential equation of second order is transformed into a system of 2 differential equations of first order and the output equation.

The transfer Matrix

Multi variable system: Transfer matrix

Single input/output system: Transfer function

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

$$y(t) = C x(t) + D u(t)$$

Transformation into the frequency domain

(for signals: capital letters for variables transformed in the frequency domain)

$$s X(s) = A X(s) + B U(s)$$

with initial values $x(t) = 0$

$$Y(s) = C X(s) + D U(s)$$

$$(sI - A) X(s) = B U(s)$$

$$X(s) = (sI - A)^{-1} B U(s)$$

$$Y(s) = \underbrace{[C (sI - A)^{-1} B + D]}_{G(s)} U(s)$$

The transfer matrix reads:

$$G(s) = C (sI - A)^{-1} B + D$$

Special case: single input/output system

$$G(s) = c^T (sI - A)^{-1} b + d$$

Example: single input/output system of 2nd order

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$c^T = [c_1 \quad c_2] \quad d = d$$

$$G(s) = [c_1 \quad c_2] \begin{bmatrix} s - a_{11} & -a_{12} \\ -a_{21} & s - a_{22} \end{bmatrix}^{-1} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} + d$$

$$= \frac{[c_1 \quad c_2] \begin{bmatrix} s - a_{22} & a_{12} \\ a_{21} & s - a_{11} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}}{s^2 - (a_{11} + a_{22})s + a_{11}a_{22} - a_{12}a_{21}} + d$$

Numerical values : $a_{11} = -1, a_{12} = 2, b_1 = 3, c_1 = 2$

$a_{21} = -4, a_{22} = -6, b_2 = 5, c_2 = 7$

$d = 0$

$$G(s) = \frac{4s + 7}{s^2 + 7s + 14}$$

$d = 3$

$$G(s) = \frac{3 \cdot s^3 + 62s + 49}{s^2 + 7s + 14}$$

(if $d > 0$, order of numerator = order of denominator)

Conclusion: $d \neq 0$ increases the order of the numerator by compared to $d = 0$

For multivariable systems, instead of the complex transfer function $G(s)$ the complex transfer matrix $G(s)$ exist.

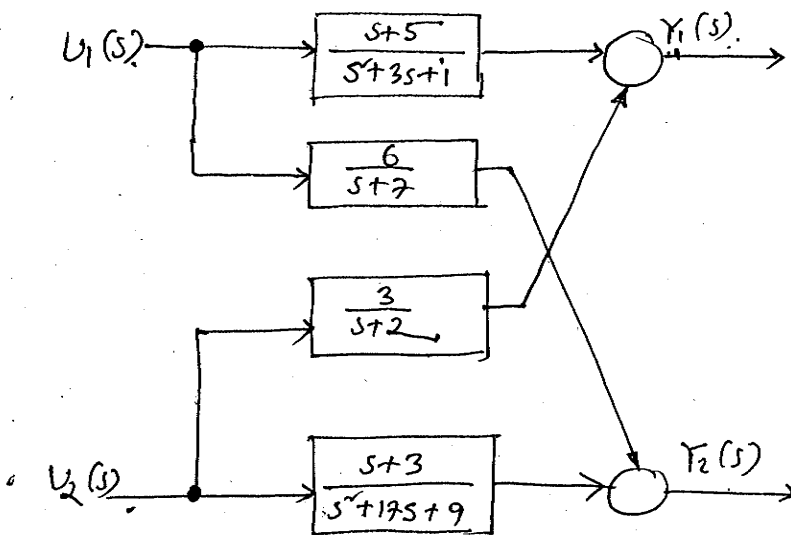
Example: system with 2 input and 2 output variables

$$G(s) = \begin{bmatrix} \frac{s+5}{s^2+3s+1} & \frac{3}{s+2} \\ \frac{6}{s+7} & \frac{s+3}{s^2+17s+9} \end{bmatrix}$$

$$Y(s) = G(s) \cdot U(s)$$

$$Y_1(s) = \frac{s+5}{s^2+3s+1} u_1(s) + \frac{3}{s+2} \cdot u_2(s)$$

$$Y_2(s) = \frac{6}{s+7} \cdot u_1(s) + \frac{s+3}{s^2+17s+9} \cdot u_2(s)$$



Controllability & observability

Controllability

The linear time-invariant system

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

is mentioned as controllable, if during a finite time period $t_0 \leq t \leq t_1$ each initial state x_0 at the moment t_0 can be moved to any other state x_1 at the moment t_1 by a suitable choice of the control vector, $u(t)$

Here are:

- * trajectories $x(t)$ not predetermined
- * no limitation of the control vector

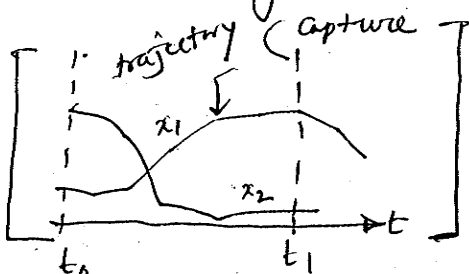
Determination of the controllability

Controllability matrix:

$$\underline{Q}_c = [\underline{B}, \underline{A}\underline{B}, \dots, \underline{A}^{n-1}\underline{B}] \text{ (This is a square matrix)}$$

Controllability condition:

A linear time-invariant system of n -th order is then and only then controllable, if the controllability matrix has the rank n .



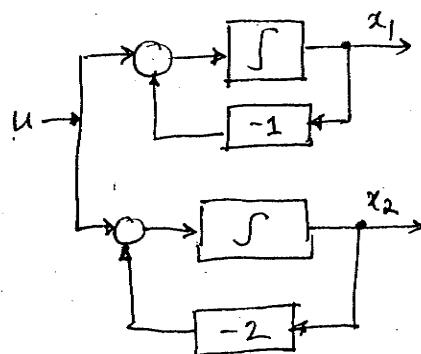
Example

1) $\underline{A} = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$, $\underline{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ where $n=2$

$$\underline{Q}_c = \begin{bmatrix} \underline{b} & \underline{A}\underline{b} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & -1 \\ 1 & -2 \end{bmatrix} \text{ rank } \underline{Q}_c = 2$$

System is controllable

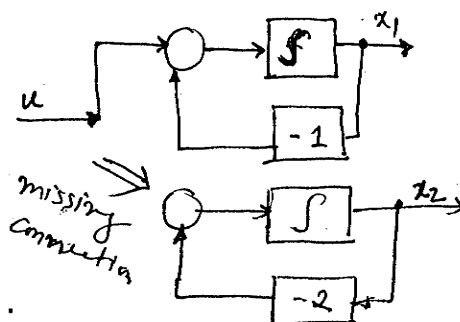


2) $\underline{A} = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$, $\underline{b} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$\underline{Q}_c = [\underline{b}, \underline{A}\underline{b}]$$

$$= \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix} \text{ rank } \underline{Q}_c = 1$$

System is not controllable.



3) $A = \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix}, b = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$Q_c = [b, Ab]$ rank $Q_c = 1$

\therefore System is not controllable

$$\begin{bmatrix} 1 & -2 \\ 1 & -2 \end{bmatrix}$$

Observability:

The linear time-invariant system

$$\begin{aligned} \dot{x}(t) &= A x(t) + B u(t) \\ y(t) &= C x(t) + D u(t) \end{aligned}$$

is mentioned as observable, if with known $u(t)$ from the measurement of $y(t)$ during a finite time period $t_0 \leq t \leq t_1$ the initial state x_0 at the moment t_0 can be clearly determined.

Here, it is important, where the initial state x_0 is located.

Determination of the observability:

observability matrix;

$$Q_o = \begin{bmatrix} C^T \\ C^T A \\ \vdots \\ C^T A^{n-1} \end{bmatrix}$$

A linear time-invariant system of n -th order is then and only then observable if the matrix Q_o has the rank n

Example:

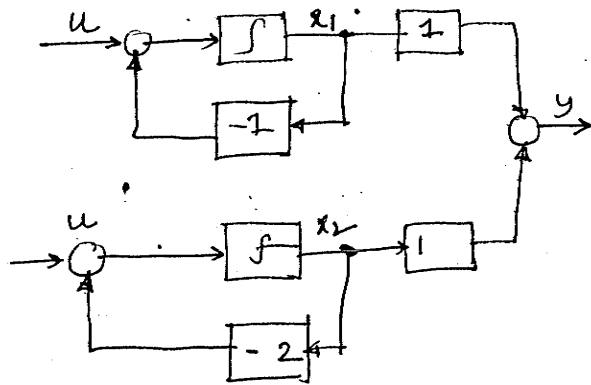
1) $A = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$

$C^T = [1 \quad 1]$

$Q_o = \begin{bmatrix} C^T \\ C^T A \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix}$

rank $Q_o = 2$

System is observable

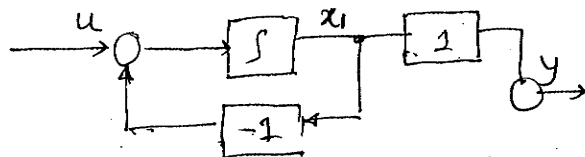


2) $A = \begin{bmatrix} -1 & 0 \\ 0 & -2 \end{bmatrix}$

$C^T = [1 \quad 0]$

$Q_o = \begin{bmatrix} C^T \\ C^T A \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix}$ rank $Q_o = 1$

System is not observable

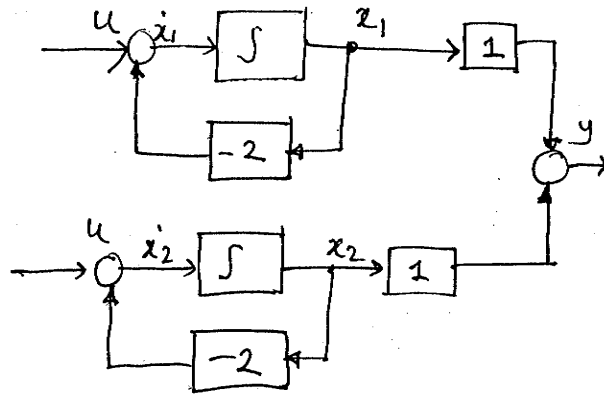


missing connection.

$$3) \quad A = \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix}$$

$$\underline{c}^T = \begin{bmatrix} 1 & 1 \end{bmatrix}$$

$$\underline{Q}_0 = \begin{bmatrix} 1 & 1 \\ -2 & -2 \end{bmatrix} \quad \text{rank } \underline{Q}_0 = 1$$



we will evaluate some examples

internet

institute control
Mayer's page
Exercises

The controllability canonical form:
linear time invariant system of n th order

Transfer function into the frequency domain:

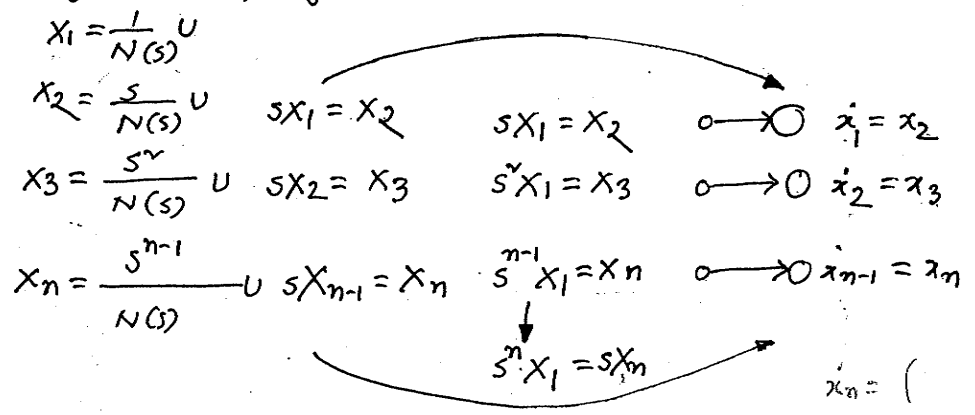
$$Y(s) = \frac{b_n s^n + b_{n-1} s^{n-1} + \dots + b_1 s + b_0}{s^n + a_{n-1} s^{n-1} + \dots + a_1 s + a_0} U(s) = \frac{Z(s)}{N(s)} U(s)$$

Corresponding differential eqⁿs

$$y^{(n)} + a_{n-1} y^{(n-1)} + \dots + a_1 y' + a_0 y = b_0 u + b_1 u' + \dots + b_{n-1} u^{(n-1)} + b_n u^{(n)}$$

disadvantage: Derivatives of $u(t)$ on the right hand side of the eqⁿ.

Substituting in the frequency domain



The initial values are assumed as zero

Calculation of \dot{x}_n

from $X_1 = \frac{1}{N(s)} U$ follows $\rightarrow N(s) X_1 = U$

$$s^n X_1 + a_{n-1} s^{n-1} X_1 + \dots + a_1 s X_1 + a_0 X_1 = U$$

$$s X_n = -a_{n-1} X_n - \dots - a_1 X_2 - a_0 X_1 + U$$



$$\dot{x}_n = -a_{n-1} x_n(t) - \dots - a_1 x_2(t) - a_0 x_1(t) + u(t) \quad (a)$$

$$Y(s) = Z(s) \cdot X_1$$

Determination of the output eqⁿ:

$$Y(s) = b_n s X_n + b_{n-1} X_n + \dots + b_1 X_2 + b_0 X_1$$



$$y(t) = b_n \dot{x}_n(t) + b_{n-1} x_n(t) + \dots + b_1 \dot{x}_2(t) + b_0 x_1(t)$$

$$= (b_n (-a_{n-1} x_n(t) - \dots - a_1 x_2(t) - a_0 x_1(t) + u(t)) + b_{n-1} x_n(t) + \dots + b_1 \dot{x}_2(t) + b_0 x_1(t))$$

$$y(t) = (b_{n-1} - b_n a_{n-1}) x_n(t) + \dots + (b_1 - b_n a_1) x_2(t) + (b_0 - b_n a_0) x_1(t) + b_n u(t)$$

The following eq^{ns} are available

$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = x_3(t)$$

...

$$\dot{x}_{n-1}(t) = x_n(t)$$

$$\dot{x}_n(t) = -a_{n-1} x_n(t) - \dots - a_1 x_2(t) - a_0 x_1(t) + u(t)$$

$$y(t) = (b_{n-1} - b_n a_{n-1}) x_n(t) + \dots + (b_1 - b_n a_1) x_2(t) + (b_0 - b_n a_0) x_1(t) + b_n u(t)$$

The conversion into matrix notation results in the controllability canonical form:

$$\dot{\underline{x}}(t) = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ \vdots & 0 & 1 & & \vdots \\ 0 & \vdots & & & -a_{n-1} \\ -a_0 & -a_1 & \dots & \dots & -a_{n-1} \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} u(t)$$

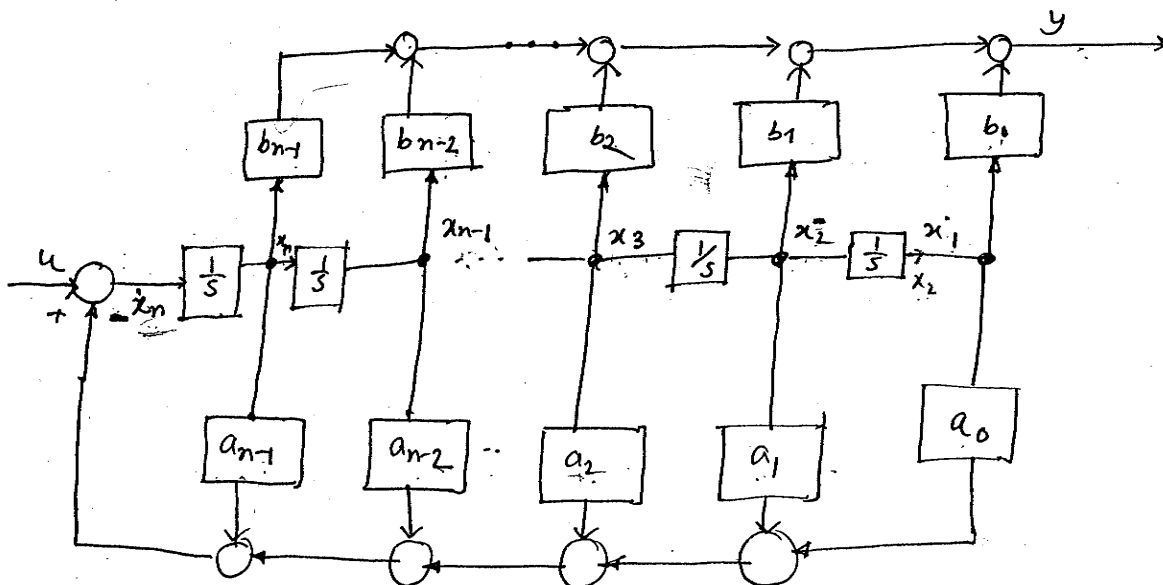
[b → coefficients of numerator]

$$y(t) = [b_0 - b_n a_0, b_1 - b_n a_1, \dots, b_{n-1} - b_n a_{n-1}] \underline{x}(t) + b_n u(t)$$

The controllability canonical form exists only, if the system is controllable.

Flow diagram of a system in the controllability canonical form.

Special case, $b_n = 0$



Observe

Example:

$$G(s) = \frac{s+3}{s^2+4s+1} \quad n=2$$

$$\begin{aligned} a_0 &= 1 & b_0 &= 3 \\ a_1 &= 4 (= a_{n-1}) & b_1 &= 1 (= b_{n-1}) \\ & & b_2 &= 0 (= b_n) \end{aligned}$$

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 \\ -1 & -4 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} 3 & 1 \end{bmatrix} x(t)$$

$$\begin{aligned} b_0 - b_n a_0 \\ = 3 - 0 \times 1 \end{aligned}$$

bq-

Test:

$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = -x_1(t) - 4x_2(t) + u(t)$$

$$\dot{x}_1(t) = -x_1(t) - 4x_1(t) + u(t)$$

$$\ddot{x}_1(t) + 4\dot{x}_1(t) + x_1 = u(t)$$



$$X_1(s)(s^2+4s+1) = U(s)$$

$$y(t) = 3x_1(t) + x_2(t)$$

$$y(t) = 3x_1(t) + \dot{x}_1(t)$$



$$Y(s) = (3+s)X_1(s) \rightarrow Y(s) = \frac{s+3}{s^2+4s+1} U(s)$$

$$G(s) = \frac{s+3}{s^2+4s+1}$$

Observability Canonical form:

Linear time-invariant system of n th order
simplified transfer function in the frequency domain

$$Y(s) = \frac{b_{n-1}s^{n-1} + \dots + b_1s + b_0}{s^n + a_{n-1}s^{n-1} + \dots + a_1s + a_0} U(s) = \frac{Z(s)}{N(s)} U(s)$$

Corresponding differential eqⁿ:

$$y^{(n)}(t) + a_{n-1}y^{(n-1)}(t) + \dots + a_1\dot{y} + a_0y = b_0u(t) + b_1\dot{u} + \dots + b_{n-1}u^{(n-1)}(t)$$

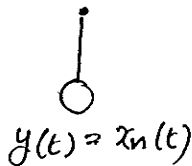
diradu: derivatives of $u(t)$ on the right-hand side of the eqⁿ.

Transformation into the frequency domain

$$s^n Y + a_{n-1} s^{n-1} Y + \dots + a_1 s Y + a_0 Y = b_0 U + b_1 s U + \dots + b_{n-1} s^{n-1} U$$

$$Y = \frac{1}{s} (b_{n-1} U - a_{n-1} Y + \frac{1}{s} (b_{n-2} U - a_{n-2} Y + \dots + \frac{1}{s} (b_1 U - a_1 Y + \frac{1}{s} (b_0 U - a_0 Y)) \dots))$$

$\underbrace{\hspace{10em}}_{X_1}$
 $\underbrace{\hspace{10em}}_{X_2}$
 $\underbrace{\hspace{10em}}_{X_{n-1}}$
 $X_n = Y$



Substitution: $X_1 = \frac{1}{s} (b_0 U - a_0 X_n)$

$$X_2 = \frac{1}{s} (b_1 U - a_1 X_n + X_1)$$

$$\vdots$$

$$X_n = \frac{1}{s} (b_{n-1} U - a_{n-1} X_n + X_{n-1})$$



$$\dot{x}_1(t) = -a_0 x_n(t) + b_0 u(t)$$

$$\dot{x}_2(t) = x_1(t) - a_1 x_n(t) + b_1 u(t)$$

$$\vdots$$

$$\dot{x}_n(t) = x_{n-1}(t) - a_{n-1} x_n(t) + b_{n-1} u(t)$$

$$y(t) = x_n(t)$$

observability canonical form:

$$\dot{x}(t) = \begin{bmatrix} 0 & 0 & \dots & -a_0 \\ 1 & 0 & & -a_1 \\ 0 & 1 & & -a_2 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & & 1 - a_{n-1} \end{bmatrix} x(t) + \begin{bmatrix} b_0 \\ b_1 \\ \vdots \\ b_{n-1} \end{bmatrix} u(t)$$

$$y(t) = [0 \ 0 \ \dots \ 0 \ 1] x(t)$$

The observability canonical form exists only, if the system is observable.

Next class diagram of observability

Exercise 1.

a) $\ddot{y}(t) - 6\dot{y}(t) + 9y(t) = 3u(t)$

introduction of state variables

$x_1(t) = y(t) \quad \therefore \dot{y}(t) = \dot{x}_1(t)$

$x_2(t) = \dot{y}(t) \quad \therefore \ddot{y}(t) = \dot{x}_2(t)$

results in $\dot{x}_2(t) - 6x_2(t) + 4x_1(t) = 3u(t)$

$\therefore \dot{x}_1(t) = x_2(t)$

$\dot{x}_2(t) = 6x_2(t) - 4x_1(t) + 3u(t)$

$\therefore \dot{\underline{x}}(t) = \begin{bmatrix} 0 & 1 \\ -4 & 6 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 0 \\ 3 \end{bmatrix} u(t)$

$\underline{y}(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} \underline{x}(t)$

b) $\ddot{y}(t) + 2y(t) = u(t)$

let, $y(t) = x_1(t) \quad \therefore \dot{y}(t) = \dot{x}_1(t)$

$\dot{y}(t) = x_2(t)$

$\therefore \ddot{y}(t) = \dot{x}_2(t)$

$\therefore \dot{x}_1(t) = x_2(t) \quad , \quad \square \quad \dot{x}_2(t) + 2x_1(t) = u(t)$

$\Rightarrow \dot{x}_2(t) = -2x_1(t) + u(t)$

$\therefore \dot{\underline{x}}(t) = \begin{bmatrix} 0 & 1 \\ -2 & 0 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(t)$

$\underline{\dot{y}}(t) = \begin{bmatrix} 0 & 1 \end{bmatrix} \underline{x}(t)$ ↖ $a_{22} = 0$ because of the absence of $\dot{y}(t)$

So, missing of 1st derivative $\dot{y}(t)$ results in in the zero element $a_{22} = 0$. in the evolution matrix.

c) $\ddot{y}(t) = u(t)$

let, $x_1 = y(t) \quad \therefore \dot{y}(t) = \dot{x}_1(t)$

$\dot{y}(t) = x_2(t)$	$\dot{x}_1(t) = x_2(t)$
$\therefore \ddot{y}(t) = \dot{x}_2(t)$	$\dot{x}_2(t) = x_3(t)$
$\ddot{y}(t) = x_3(t)$	$\dot{x}_3(t) = u(t)$
$\therefore \ddot{y}(t) = \dot{x}_3(t)$	$y(t) = x_1(t)$

$$\therefore \dot{\underline{x}}(t) = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u(t)$$

$$\underline{y}(t) = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \underline{x}(t)$$

d) $\ddot{y}(t) \dot{y}(t) + 4y(t) = 3u(t)$

The system is nonlinear (two derivatives are multiplied together)
so, we can not do it now.

2) a) The system is

- time invariant
- linear
- single input/output

b) system of 2nd order

c) special case $a_{11} = 0$, $b_1 = 0$, $c_2 = 0$, $d = 0$

$$\dot{x}_1(t) = 3x_2(t)$$

$$\dot{x}_2(t) = 4x_1(t) - 6x_2(t) + b_2 u(t)$$

$$y(t) = c_1 x_1(t)$$

Conversion of the second order eqⁿ:

$$\text{with, } \dot{x}_1(t) = 3x_2(t) \rightarrow x_2(t) = \frac{1}{3} \dot{x}_1(t)$$

$$\dot{x}_2(t) = 3\dot{x}_2(t) \rightarrow \dot{x}_2(t) = \frac{1}{3} \ddot{x}_1(t)$$

$$\frac{1}{3} \ddot{x}_1(t) = 4x_1(t) - 6 \times \frac{1}{3} \dot{x}_1(t) + b_2 u(t)$$

$$\Rightarrow \ddot{x}_1(t) = 12x_1(t) - 6\dot{x}_1(t) + 3b_2 u(t) \Rightarrow \ddot{x}_1(t) + 6\dot{x}_1(t) - 12x_1(t) = 3b_2 u(t)$$

conversion of output eqⁿ:

$$y(t) = c_1 x_1(t) \rightarrow x_1 = \frac{1}{c_1} y(t)$$

$$\therefore \dot{x}_1 = \frac{1}{c_1} \dot{y}(t)$$

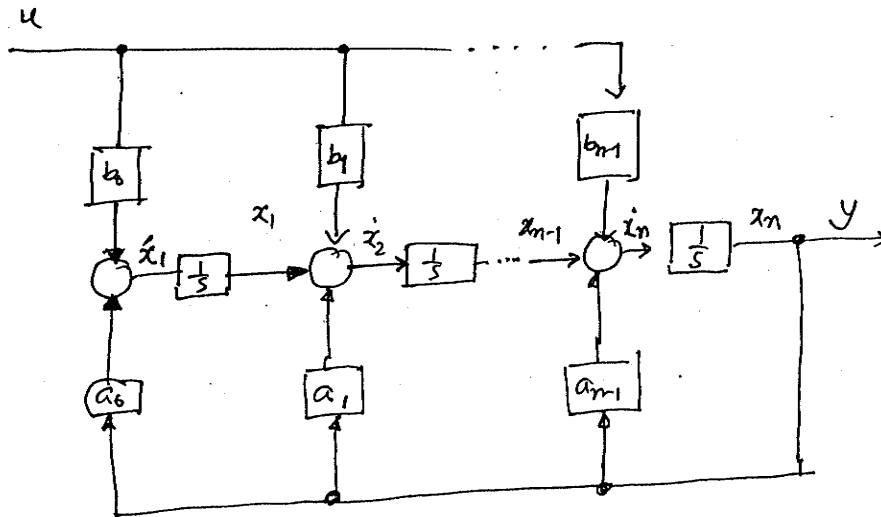
$$\ddot{x}_1(t) = \frac{1}{c_1} \ddot{y}(t)$$

$$\frac{1}{c_1} \ddot{y}(t) + 6 \times \frac{1}{c_1} \dot{y}(t) - 12 \times \frac{1}{c_1} y(t) = 3b_2 u(t)$$

$$\therefore \ddot{y}(t) + 6\dot{y}(t) - 12y(t) = 3b_2 c_1 u(t)$$

d) The equations are no more solvable in direct way as long as $a_{11} \neq 0$.

flow diagram of a system in the observability canonical form: ($b_n = 0$)



Example: Transfer function

$$G(s) = \frac{s+3}{s^2+4s+1}$$

$$a_0 = 1 \quad b_0 = 3$$

$$a_1 = 4 (= a_{n-1}) \quad b_1 = 1 (= b_{n-1})$$

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 \\ 1 & -4 \end{bmatrix} x(t) + \begin{bmatrix} 3 \\ 1 \end{bmatrix} u(t)$$

$$y(t) = [0 \quad 1] x(t)$$

Test: $y(t) = x_2(t)$

$$\frac{d}{dt} : \dot{y}(t) = \dot{x}_2(t)$$

$$y(t) = x_1(t) - 4x_2(t) + u(t)$$

$$\frac{d}{dt} : \ddot{y}(t) = \dot{x}_1(t) - 4\dot{x}_2(t) + \dot{u}(t)$$

$$\dot{y}(t) = -x_2(t) + 3u(t) - 4\dot{x}_2(t) + \dot{u}(t)$$

$$\ddot{y}(t) = -y(t) - 4\dot{y}(t) + 3u(t) + \dot{u}(t)$$

$$\Rightarrow \ddot{y}(t) + 4\dot{y}(t) + y(t) = 3u(t) + \dot{u}(t)$$

$$G(s) = \frac{Y(s)}{U(s)} = \frac{s+3}{s^2+4s+1}$$

The Jordan Canonical Form:

Controllability & Observability Canonical forms:

→ derivation from the coefficients of the numerator as well as the denominator polynomial of the complex transfer function.

Jordan Canonical form:

Derivation from the poles $\lambda_i = s_i$ of the transfer function. All poles must be known.

Development of the Jordan canonical form:

starting point: complex transfer function

$$Y(s) = \frac{b_n s^n + \dots + b_1 s + b_0}{s^n + a_{n-1} s^{n-1} + \dots + a_1 s + a_0} U(s)$$

By decomposition into partial fractions one receives

$$Y(s) = \left(\sum_{i=1}^n \frac{c_i}{s - \lambda_i} + d \right) U(s)$$

Definition of the states variables:

$$X_i(s) = \frac{1}{s - \lambda_i} U(s)$$

It results:

$$Y(s) = \sum_{i=1}^n c_i X_i(s) + d U(s)$$

↻ transformation into the time domain

$$\dot{x}_i(t) = \lambda_i x_i(t) + u(t)$$

$$y(t) = \sum_{i=1}^n c_i x_i(t) + d u(t)$$

Jordan Canonical form:

$$\dot{\underline{x}}(t) = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ \vdots & \lambda_2 & 0 & \dots \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & \dots & \lambda_n \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} u(t)$$

$$y(t) = [c_1 \ c_2 \ \dots \ c_n] \underline{x}(t) + d u(t)$$

Advantage: Diagonal matrix

- isolated system of differential equations
- Each differential equation can be solved by itself
- useful for pure investigation

Book

Franklin

Doble poles:

Multi (m-fold) pole at λ_1

it results from the decomposition into parallel fractions :

$$\begin{array}{l}
 X_1(s) = \frac{1}{s-\lambda_1} U(s) \\
 X_2(s) = \frac{1}{(s-\lambda_1)^2} U(s) \\
 \vdots \\
 X_m(s) = \frac{1}{(s-\lambda_1)^m} U(s) \\
 X_{m+1}(s) = \frac{1}{s-\lambda_{m+1}} U(s) \\
 \vdots \\
 X_n(s) = \frac{1}{s-\lambda_n} U(s)
 \end{array}
 \quad \left| \quad
 \begin{array}{l}
 x_1(s) = \frac{1}{s-\lambda_1} U(s) \\
 x_2(s) = \frac{1}{s-\lambda_1} x_1(s) \\
 \vdots \\
 x_m(s) = \frac{1}{s-\lambda_1} x_{m-1}(s) \\
 x_{m+1}(s) = \frac{1}{s-\lambda_{m+1}} U(s) \\
 \vdots \\
 x_n(s) = \frac{1}{s-\lambda_n} U(s)
 \end{array}$$

inverse Laplace transformation

$$\dot{x}_1(t) = \lambda_1 x_1(t) + u(t)$$

$$\dot{x}_2(t) = \lambda_1 x_2(t) + x_1(t)$$

$$\dot{x}_3(t) = \lambda_1 x_3(t) + x_2(t)$$

$$\dot{x}_m(t) = \lambda_1 x_m(t) + x_{m-1}(t)$$

$$\dot{x}_{m+1}(t) = \lambda_{m+1} x_{m+1}(t) + u(t)$$

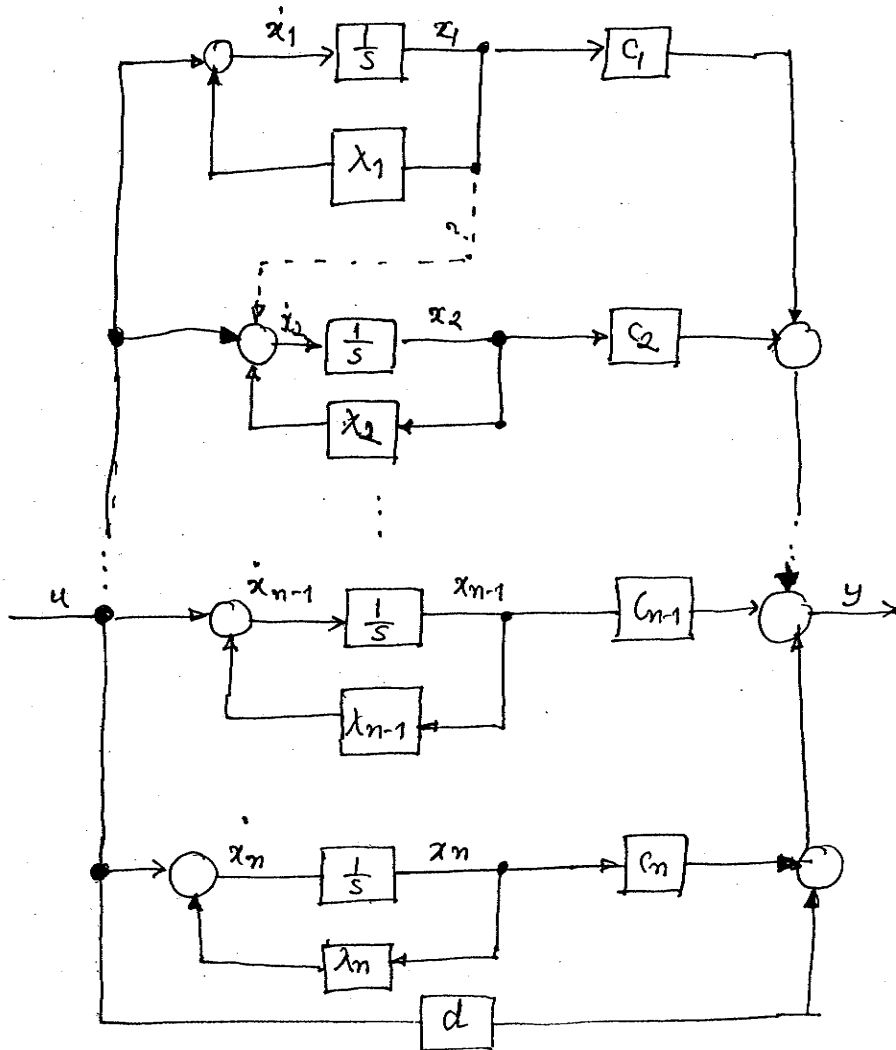
$$\dot{x}_n(t) = \lambda_n x_n(t) + u(t)$$

Jordan canonical form with multiple poles:

$$\dot{\underline{x}}(t) = \begin{bmatrix} \lambda_1 & 0 & 0 & 0 & 0 \\ 1 & \lambda_1 & 0 & 0 & 0 \\ 0 & 1 & \lambda_1 & 0 & 0 \\ 0 & 0 & 0 & \lambda_{m+1} & \vdots \\ 0 & 0 & 0 & \dots & \lambda_m \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} u(t)$$

$$y(t) = [c_1 \ c_2 \ \dots \ c_m] \underline{x}(t) + d u(t)$$

Flow diagram of a system in Jordan canonical form:



Conjugate complex pole pairs

Given is:

$$\begin{aligned} \dot{x}_1(t) &= \lambda_1 x_1(t) + u(t) \quad \text{with } \lambda_1 = a + jb \\ \vdots \\ \dot{x}_2(t) &= \lambda_2 x_2(t) + u(t) \quad \lambda_2 = a - jb \end{aligned}$$

Approach for conjugated complex state variables:

$$\begin{cases} x_1(t) = \xi(t) + j\eta(t) \\ x_2(t) = \xi(t) + j\eta(t) \end{cases}$$

$$\begin{aligned} \dot{\xi}(t) + j\dot{\eta}(t) &= (a + jb)(\xi(t) + j\eta(t)) + u(t) \\ \dot{\xi}(t) + j\dot{\eta}(t) &= a\xi(t) - b\eta(t) + j(b\xi(t) + a\eta(t)) + u(t) \end{aligned}$$

The comparison of the real and imaginary part provides

$$\begin{aligned} \dot{\xi}(t) &= a\xi(t) + b\eta(t) + u(t) \\ \dot{\eta}(t) &= b\xi(t) + a\eta(t) \end{aligned}$$

Thus the conjugated pair of poles was transferred into a real representation regarding Jordan Canonical form.

Disadvantages:

- Transformed state variables represent the system dynamics
- The differential equations of first order are no more decoupled
- In case of existence of further poles as well as ascertaining of the input and the output vector further transformations must be applied.

Consequence:

In the case of existence of conjugated complex pairs of poles the Jordan Canonical form should not be used if possible.

Eigen values & Eigen vectors

The eigen values of a linear system are identical to the poles of the corresponding complex transfer function.

Calculation of the Eigenvalues of the evolution matrix:

starting point: The transfer function for a single input/output system

$$G(s) = \underline{c}^T (sI - \underline{A})^{-1} \underline{b} + d$$

The poles of the system are determined by the denominator $N(s)$ of the polynomial $G(s)$

$$N(s) = 0$$

The denominator $N(s)$ results exclusively from the matrix inversion of $(sI - A)^{-1}$, consequently the following is valid,

$$\det(sI - \underline{A}) = 0$$

The breakup by $s_i, i = 1 \dots n$ yields the eigenvalues λ_i

Definition of the Eigenvalues:

$$\det(\lambda_i I - \underline{A}) = 0$$

characteristic equation of the matrix \underline{A}

Eigen vectors:

To each eigenvalue λ_i belongs an eigenvector \underline{v}_i , which is different from zero vector. It fulfils the vector equation

$$(\lambda_i I - \underline{A}) \underline{v}_i = \underline{0}$$

The eigen vectors are linear independent from each other,

example:

starting point is the evolution matrix

$$\underline{A} = \begin{bmatrix} -3 & 3 \\ 1 & -5 \end{bmatrix}$$

$$\det(\lambda_i I - \underline{A}) = \det \begin{bmatrix} \lambda_i + 3 & -3 \\ -1 & \lambda_i + 5 \end{bmatrix}$$

$$= (\lambda_i + 3)(\lambda_i + 5) - 3$$

$$= \lambda_i^2 + 8\lambda_i + 15 - 3$$

$$= \lambda_i^2 + 8\lambda_i + 12 = (\lambda_i + 2)(\lambda_i + 6)$$

The characteristic equation yields,

$$(\lambda_i + 2)(\lambda_i + 6) = 0$$

→ Eigenvalues $\lambda_1 = -2$, $\lambda_2 = -6$

Eigenvectors:

$$(\lambda_i I - A) \underline{v}_i = \underline{0}$$

$$\begin{bmatrix} \lambda_i + 3 & -3 \\ -1 & \lambda_i + 5 \end{bmatrix} \begin{bmatrix} v_{1i} \\ v_{2i} \end{bmatrix} = \underline{0}$$

results in

$$(\lambda_i + 3)v_{1i} - 3v_{2i} = 0$$

$$-v_{1i} + (\lambda_i + 5)v_{2i} = 0$$

one receives for:

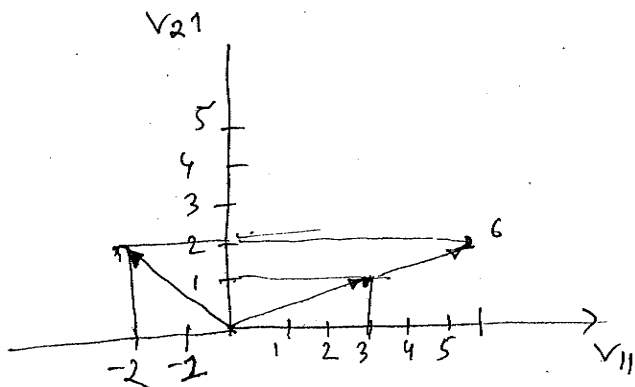
$$i = 1; \quad \lambda_1 = -2;$$

$$\left. \begin{array}{l} v_{11} + 3v_{21} = 0 \\ -v_{11} + 3v_{21} = 0 \end{array} \right\} v_{11} = 3v_{21}$$

$$i = 2; \quad \lambda_2 = -6;$$

$$\left. \begin{array}{l} -3v_{12} - 3v_{22} = 0 \\ -v_{12} - v_{22} = 0 \end{array} \right\} v_{12} = -v_{22}$$

only the direction is important for Eigen vector, not values.



$$v_{11} = 3v_{21}$$

↓
1
3

v_{21} & v_{22} can be freely chosen.

$v_{21} = 1$ & $v_{22} = 2$ results in

$$\underline{v}_1 = \begin{bmatrix} 3 \\ 1 \end{bmatrix} \quad \& \quad \underline{v}_2 = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

\underline{v}_1 & \underline{v}_2 are linear independent from each other.

The Equivalent System Transformation

System in the state space description:

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

$$\underline{y}(t) = \underline{C} \underline{x}(t) + \underline{D} u(t)$$

The new state vector:

(first define new state vector)

$$\bar{\underline{x}}(t) = \underline{T} \underline{x}(t)$$

\underline{T} is non-singular $n \times n$ matrix consisting of constant elements,
Besides, it applies

$$\underline{x}(t) = \underline{T}^{-1} \bar{\underline{x}}(t)$$

$$\Rightarrow \dot{\underline{x}}(t) = \underline{T}^{-1} \dot{\bar{\underline{x}}}(t)$$

By inserting into the state equations one receives

$$\underline{T}^{-1} \dot{\bar{\underline{x}}}(t) = \underline{A} \underline{T}^{-1} \bar{\underline{x}}(t) + \underline{B} u(t)$$

$$\underline{y}(t) = \underline{C} \underline{T}^{-1} \bar{\underline{x}}(t) + \underline{D} u(t)$$

Multiplication of the state differential equation by \underline{T} from the left shows:

$$\dot{\bar{\underline{x}}}(t) = \underbrace{\underline{T} \underline{A} \underline{T}^{-1}}_{\bar{\underline{A}}} \bar{\underline{x}}(t) + \underbrace{\underline{T} \underline{B}}_{\bar{\underline{B}}} u(t)$$

$$\underline{y}(t) = \underbrace{\underline{C} \underline{T}^{-1}}_{\bar{\underline{C}}} \bar{\underline{x}}(t) + \underbrace{\underline{D}}_{\bar{\underline{D}}} u(t)$$

Hence the transformation equations results in

$$\begin{array}{l} \bar{\underline{A}} = \underline{T} \underline{A} \underline{T}^{-1} \\ \bar{\underline{B}} = \underline{T} \underline{B} \\ \bar{\underline{C}} = \underline{C} \underline{T}^{-1} \\ \bar{\underline{D}} = \underline{D} \end{array}$$

The transformed state equations are

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

$$y(t) = \bar{C} \bar{x}(t) + \bar{D} u(t)$$

Eigenvectors don't belong to complex transfer functions

So, eigen values of these eq's are not same as before.
? vectors?

The initial values for the transformed system result in

$$\bar{x}(t_0) = T x(t_0)$$

Input/output behavior of the transformed system:

$$\begin{aligned} \bar{G}(s) &= \bar{C} (sI - \bar{A})^{-1} \bar{B} + \bar{D} \\ &= \bar{C} T^{-1} (sI - T^{-1} A T)^{-1} T B + D \end{aligned}$$

[double applying the rule for matrix calculation:
 $F^{-1} E^{-1} = (E F)^{-1}$]

$$\begin{aligned} \bar{G}(s) &= \bar{C} (sT^{-1} I T - T^{-1} A T)^{-1} T B + D \\ &= \bar{C} (sI - A)^{-1} B + D \\ \rightarrow \bar{G}(s) &= G(s) \end{aligned}$$

Conclusion:

The equivalent transformation causes

* a modified representation of system-internal dynamic from
caused by the transformed state vector $\bar{x}(t)$

* no modification regarding the dynamic input/output
behavior of the transformed system compared to the original
system, which was never transformed.

With the equivalent system transformation of the state vector $x(t)$ is
transformed. The input vector $u(t)$ as well as the output vector $y(t)$
remain unmodified.

The transformation of a system into the Jordan Canonical form:

System in the state space description

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

$$\underline{y}(t) = \underline{C} \underline{x}(t) + \underline{D} \underline{u}(t)$$

Transformation

$$\underline{x}(t) = \underline{V} \bar{\underline{x}}(t) \quad \text{with} \quad \underline{V} = \underline{T}^{-1}$$

The transformation matrix \underline{V} can be indicated in column vector notation as follows,

$$\underline{V} = [\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n] \quad \text{with} \quad \underline{v}_i = \begin{bmatrix} v_{1i} \\ v_{2i} \\ \vdots \\ v_{ni} \end{bmatrix} \quad i=1 \dots n$$

The system transformation into the Jordan Canonical form results in the following approach:

$$\dot{\bar{\underline{x}}}(t) = \underline{\underline{V}}^{-1} \underline{A} \underline{\underline{V}}(t) \bar{\underline{x}} + \underline{\underline{V}}^{-1} \underline{B} \underline{u}$$

Thus it is valid:

$$\underline{\underline{V}}^{-1} \underline{A} \underline{\underline{V}} = \underline{\underline{\Lambda}} \quad \text{with} \quad \underline{\underline{\Lambda}} = \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix}$$

$$\underline{A} \underline{V} = \underline{V} \underline{\Lambda}$$

left side:

$$\underline{A} \underline{V} = \underline{A} [\underline{v}_1 \dots \underline{v}_n]$$

$$= [\underline{A} \underline{v}_1 \dots \underline{A} \underline{v}_n]$$

right side:

$$\underline{V} \underline{\Lambda} = \underline{V} \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix}$$

$$= \begin{bmatrix} \underline{v}_1 \lambda_1 & \dots & \underline{v}_n \lambda_n \\ \vdots & & \vdots \\ \underline{v}_1 \lambda_1 & \dots & \underline{v}_n \lambda_n \end{bmatrix}$$

$$= [\lambda_1 \underline{v}_1 \dots \lambda_n \underline{v}_n]$$

$$\Rightarrow [A \underline{v}_1 \dots A \underline{v}_n] = [\lambda_1 \underline{v}_1 \dots \lambda_n \underline{v}_n]$$

It applies to every column:

$$A \underline{v}_1 = \lambda_1 \underline{v}_1$$

$$\Rightarrow (\lambda_1 I - A) \underline{v}_1 = 0$$

The operation $(\lambda_1 I - A) \underline{v}_1 = 0$ corresponds exactly to the determining equation of the Eigen vectors.

Consequently one receives

$$\underline{V} = \begin{bmatrix} \underline{v}_1 & \underline{v}_2 & \dots & \underline{v}_n \\ \vdots & \vdots & & \vdots \end{bmatrix}$$

The transformation matrix \underline{V} is composed columnwise from the Eigen vectors \underline{v}_i corresponding to the Eigen values λ_i .

Practical Procedure:

starting point:

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

$$\underline{g}(t) = \underline{C} \underline{x}(t) + \underline{D} \underline{u}(t)$$

Step 1: Find the Eigen values from the characteristic equation

$$\det(\lambda_i I - \underline{A}) = 0$$

Step 2:

Find the Eigen vectors \underline{v}_i , which belong to Eigen values λ_i respectively, according to

$$(\lambda_i I - \underline{A}) \underline{v}_i = 0$$

Here, one element of \underline{v}_i , which belong to eigen values λ_i normally can be chosen freely. Then, the remaining elements result from the element.

Step 3 put together the "the transformation matrix" \underline{V} .

$$\underline{V} = [\underline{v}_1, \underline{v}_2 \dots \underline{v}_n]$$

Step 4

Calculate the matrices

$$\begin{aligned} \underline{\bar{A}} &= \underline{V}^{-1} \underline{A} \underline{V} \\ \underline{\bar{B}} &= \underline{V}^{-1} \underline{B} \\ \underline{\bar{C}} &= \underline{C} \underline{V} \\ \underline{\bar{D}} &= \underline{D} \end{aligned}$$

Step 5:

set up the equations in the Jordan canonical form (first general form)

$$\begin{aligned} \dot{\underline{x}}(t) &= \underline{\bar{A}} \underline{x}(t) + \underline{\bar{B}} u(t) \\ \underline{y}(t) &= \underline{\bar{C}} \underline{x}(t) + \underline{\bar{D}} u(t) \end{aligned}$$

with $\underline{\bar{A}} = \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix}$ → By further standardization at last the Jordan canonical form come up.

Influence of the system transformation on the controllability and observability:

Controllability matrix for the transformed system:

$$\underline{\bar{Q}}_c = [\underline{\bar{B}}, \underline{\bar{A}} \underline{\bar{B}}, \dots, \underline{\bar{A}}^{n-1} \underline{\bar{B}}]$$

It applies to $i = 0, 1, \dots, n-1$

$$\begin{aligned} \underline{\bar{A}}^i \underline{\bar{B}} &= \underline{I} \underline{A} \underline{I}^{-1} \underline{I} \underline{A} \underline{I}^{-1} \dots \underline{I} \underline{A} \underline{I}^{-1} \underline{I} \underline{B} \\ &= \underline{I} \underline{A}^i \underline{B} \end{aligned}$$

$$\underline{\bar{Q}}_c = [\underline{I} \underline{B}, \underline{I} \underline{A} \underline{B}, \dots, \underline{I} \underline{A}^{n-1} \underline{B}]$$

$$= \underline{I} [\underline{B}, \underline{A} \underline{B}, \dots, \underline{A}^{n-1} \underline{B}]$$

$$\therefore \underline{\bar{Q}}_c = \underline{I} \underline{Q}_c$$

The corresponding applies to observability matrix

$$\underline{\bar{Q}}_o = \begin{bmatrix} \underline{\bar{C}} \\ \underline{\bar{C}} \underline{\bar{A}} \\ \vdots \\ \underline{\bar{C}} \underline{\bar{A}}^{n-1} \end{bmatrix} = \begin{bmatrix} \underline{C} \underline{I}^{-1} \\ \underline{C} \underline{A} \underline{I}^{-1} \\ \vdots \\ \underline{C} \underline{A}^{n-1} \underline{I}^{-1} \end{bmatrix} = \begin{bmatrix} \underline{C} \\ \underline{C} \underline{A} \\ \vdots \\ \underline{C} \underline{A}^{n-1} \end{bmatrix} \underline{I}^{-1}$$

$$\therefore \underline{\bar{Q}}_o = \underline{Q}_o \underline{I}^{-1}$$

As I is not singular, the rank of a matrix will not be altered during a multiplication with \bar{I}

→ controllability & observability of a linear time-invariant is transformation invariant.

Control with complete state feedback

starting point:

Time-invariant, linear, controllable, single input/output system:

$$\dot{\underline{x}}(t) = \underline{A} \underline{x}(t) + \underline{b} u(t)$$

$$y(t) = \underline{c}^T \underline{x}(t)$$

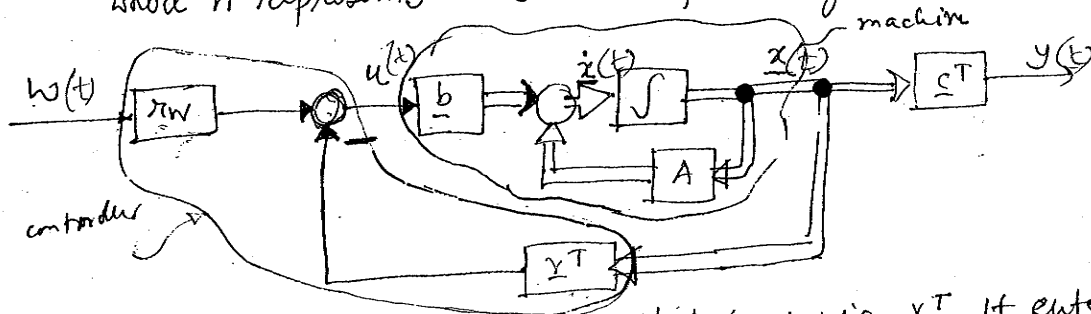
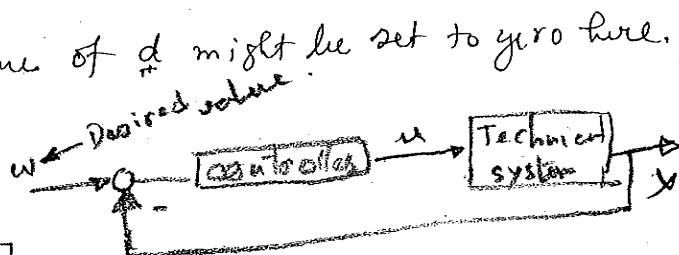
As usual in practice, the value of d might be set to zero here.

Feedback of the state vector

$$u(t) = r_w w(t) - \underline{z}^T \underline{x}(t)$$

$$\text{with } \underline{z}^T = [z_1, z_2, \dots, z_n]$$

where n represents the order of the system machine



The complete state vector $\underline{x}(t)$ is fed back via \underline{z}^T . It enters the control algorithm.

Behavior of the overall system

$$\dot{\underline{x}}(t) = \underline{A} \underline{x}(t) + \underline{b} u(t)$$

$$\dot{\underline{x}}(t) = \underline{A} \underline{x}(t) + \underline{b} (r_w w(t) - \underline{z}^T \underline{x}(t))$$

V.v.g.

$$\dot{\underline{x}}(t) = (\underline{A} - \underline{b} \underline{z}^T) \underline{x}(t) + \underline{b} r_w w(t)$$

overall eqn



$$s \underline{X}(s) = (\underline{A} - \underline{b} \underline{z}^T) \underline{X}(s) + \underline{b} r_w W(s)$$

$$\Rightarrow (s \underline{I} - \underline{A} + \underline{b} \underline{z}^T) \underline{X}(s) = \underline{b} r_w W(s)$$

$$\Rightarrow \underline{Y}(s) = (sI - \underline{A} + \underline{b} \underline{r}^T)^{-1} \underline{b} r_w W(s)$$

The poles of the transfer polynomial results from the characteristic equation:

$$\det(sI - \underline{A} + \underline{b} \underline{r}^T) = 0$$

Equivalent approach for the characteristic eqⁿ

$$(s - \lambda_1)(s - \lambda_2) \dots (s - \lambda_n) = 0$$

Comparing these two approaches yields:

$$\det(sI - \underline{A} + \underline{b} \underline{r}^T) = (s - \lambda_1)(s - \lambda_2) \dots (s - \lambda_n)$$

① Procedure for the concrete application here:

• step 1: Calculate the left side of the equation by inserting the elements for \underline{A} & \underline{b}

The demanded values for $\underline{r}^T = [r_1, r_2, \dots, r_n]$ still remain in general form

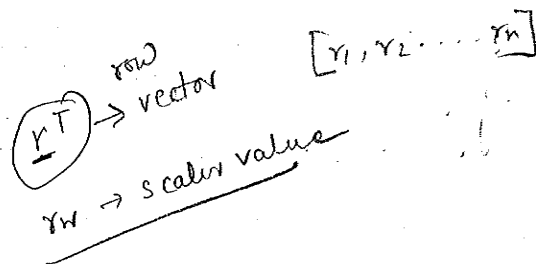
• step 2: Calculate the right hand side of the equation by selection of the eigen values (Poles) $\lambda_1, \lambda_2, \dots, \lambda_n$ for the overall system

• step 3: perform comparison of the coefficients for both sides of the equation with regard to all powers of the complex frequency s .

• step 4: find the solution of the n equations with n unknowns

Now the vector \underline{r}^T is found

Prerequisite for the derivation of a state feedback controller is the controllability of the system



Example: $\dot{x}(t) = \begin{bmatrix} -3 & 1 \\ 0 & -4 \end{bmatrix} x(t) + \begin{bmatrix} 1 \\ 2 \end{bmatrix} u(t)$

$$y(t) = \begin{bmatrix} 1 & 2 \\ 3 & 3 \end{bmatrix} x(t)$$

r^T, y^T

Assertion of the controllability:

$$\underline{Q}_c = \begin{bmatrix} b & -A b \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 2 & -8 \end{bmatrix} \rightarrow \text{rank } \underline{Q}_c = 2 = n$$

→ the system is controllable,
a state feed back controller can be developed

problem: The Eigen values of the closed loop should be located at $\lambda_1 = -1, \lambda_2 = -2$ [because, these roots make the system stable]

$$\det(sI - A + b r^T) = (s - \lambda_1)(s - \lambda_2)$$

1. diff side of the equation with $r^T = [r_1 \ r_2]$

$$\det(sI - A + b r^T) = \det \begin{bmatrix} s+3+r_1 & -1+r_2 \\ 2r_1 & s+4+2r_2 \end{bmatrix}$$

$$= s^2 + (r_1 + 2r_2 + 7)s + 6r_1 + 6r_2 + 12$$

2. Right hand side of the equation

$$(s - \lambda_1)(s - \lambda_2) = (s+1)(s+2) \\ = s^2 + 3s + 2$$

3. Comparison of eqⁿs:

$$r_1 + 2r_2 + 7 = 3 \quad \text{--- (1)}$$

$$6r_1 + 6r_2 + 12 = 2 \quad \text{--- (2)}$$

$$s: r_1 = -4 - 2r_2$$

∴

4. solution:

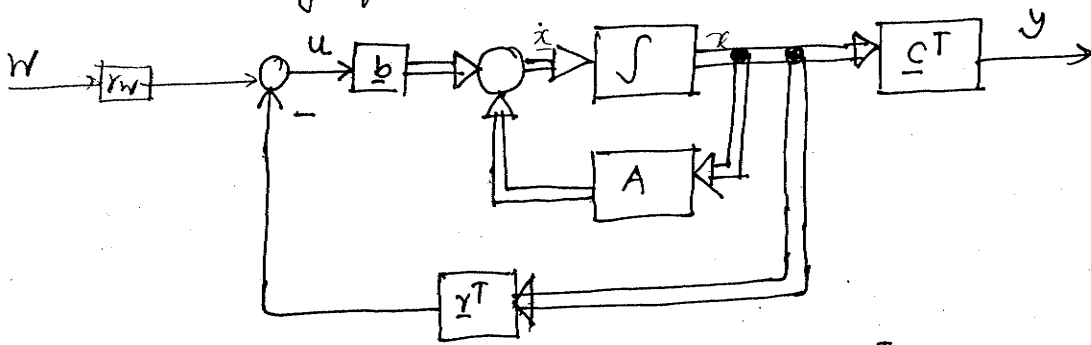
$$-24 - 12r_2 + 6r_2 + 12 = 2$$

$$6r_2 = -14 \rightarrow r_2 = -\frac{14}{6} = -\frac{7}{3}$$

$$r_1 = -4 + 2 \cdot \frac{7}{3} = \frac{2}{3}$$

$$\therefore \underline{r}^T = \begin{bmatrix} \frac{2}{3} & -\frac{7}{3} \end{bmatrix}$$

The following system structure is now available:



The feedback branch is implemented by y^T
 However, the nominal branch is not considered yet

Ascertainment of the gain of the desired value
 output equation:

$$y(t) = c^T x(t)$$



$$Y(s) = c^T X(s)$$

with $X(s) = (sI - A + b y^T)^{-1} b r_w W(s)$
 results in the complex transfer function,

$$Y(s) = \underbrace{c^T (sI - A + b y^T)^{-1} b r_w}_{F_w(s)} W(s)$$

It is required that in the steady state $y_{\infty} = w_{\infty}$ should be valid,
 i.e. the controlled value must adapt itself to the nominal value.
 Considering this, r_w must be calibrated

The following is valid:

$$Y(s) = F_w(s) \cdot W(s)$$

Final value theorem of the Laplace transform

$$y_{\infty} = \lim_{s \rightarrow 0} s \cdot F_w(s) \cdot W(s) \quad \text{with } W(s) = \frac{1}{s} w_{\infty}$$

$$= \lim_{s \rightarrow 0} F_w(s) \cdot s \cdot \frac{1}{s} w_{\infty}$$

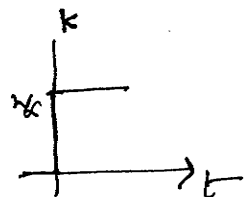
$$y_{\infty} = \lim_{s \rightarrow 0} F_w(s) \cdot w_{\infty}$$

$$\text{with } F_w(s) = c^T (sI - A + b y^T)^{-1} b r_w$$

$$\therefore y_{\infty} = c^T (b y^T - A)^{-1} b r_w w_{\infty}$$

with $y_{\infty} = w_{\infty}$

$$c^T (b y^T - A)^{-1} b r_w = 1$$



→ Equation for the calculation of the gain for the nominal value r_w Page 2

$$r_w = (\underline{c}^T (\underline{b} \underline{r}^T - \underline{A})^{-1} \underline{b})^{-1}$$

Supplement to the procedure listed above for the concrete application.

• step 5: Calculate the required gain r_w for the nominal value in accordance to the equation shown above.

Annotation:

The equation is based on the fact that in the steady-state case the value for the output y_{ss} is equal to the nominal value w_{ss} .

Continuation of the example:

5. Calculation of r_w

$$\underline{b} \underline{r}^T - \underline{A} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \begin{bmatrix} \frac{2}{3} & -\frac{7}{3} \end{bmatrix} - \begin{bmatrix} -3 & 1 \\ 6 & -4 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{2}{3} + 3 & -\frac{7}{3} - 1 \\ 4/3 & -\frac{14}{3} + 4 \end{bmatrix} = \begin{bmatrix} \frac{11}{3} & -\frac{10}{3} \\ \frac{4}{3} & -\frac{2}{3} \end{bmatrix}$$

$$(\underline{b} \underline{r}^T - \underline{A})^{-1} = \frac{\begin{bmatrix} -\frac{2}{3} & \frac{10}{3} \\ -4/3 & \frac{11}{3} \end{bmatrix}}{-\frac{22}{9} + \frac{40}{9}} = \begin{bmatrix} -\frac{1}{3} & \frac{5}{3} \\ -\frac{2}{3} & \frac{11}{6} \end{bmatrix}$$

$$(\underline{b} \underline{r}^T - \underline{A})^{-1} \underline{b} = \begin{bmatrix} -\frac{1}{3} & \frac{5}{3} \\ -\frac{2}{3} & \frac{11}{6} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$= \begin{bmatrix} 3 \\ 3 \end{bmatrix}$$

$$\underline{c}^T (\underline{b} \underline{r}^T - \underline{A})^{-1} \underline{b} = \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} 3 \\ 3 \end{bmatrix} = 3$$

$$(\underline{c}^T (\underline{b} \underline{r}^T - \underline{A})^{-1} \underline{b})^{-1} = \frac{1}{3}$$

$$\rightarrow r_w (\text{Input gain}) = \frac{1}{3}$$

for better understanding:

$$y_{ss} = \underbrace{\underline{c}^T (\underline{b} \underline{r}^T - \underline{A})^{-1} \underline{b}}_3 \cdot \underbrace{r_w}_{1/3} \cdot w_{ss}$$

$$\rightarrow y_{ss} = w_{ss}$$

Special case: The system exists already in the controllability canonical form

These are form transfer function

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \vdots & & 1 \\ -a_0 & -a_1 & -a_2 & & -a_{n-1} \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} u(t)$$

$$y(t) = [b_0 \ b_1 \ \dots \ b_{n-1}] x(t)$$

in general first is to consider: With $b_n=0$
introduction of a state feedback controller

$$u(t) = r_w w(t) - \gamma^T x(t) \quad \text{with } \gamma^T = [\gamma_1 \ \gamma_2 \ \dots \ \gamma_n]$$

state differential equation:

$$\dot{x}(t) = \underline{A} x(t) + \underline{b} u(t) \quad \rightarrow \text{(for the technical system)}$$

Overall system:

$$\dot{x}(t) = \underline{A} x(t) + \underline{b} (r_w w(t) - \gamma^T x(t))$$

$$\dot{x}(t) = \underbrace{(\underline{A} - \underline{b} \gamma^T)}_{\underline{A}_R} x(t) + \underline{b} r_w w(t) \dots \dots \textcircled{a}$$

With the special case of the controllability canonical form the evolution matrix \underline{A}_R of the overall system results in:

$$\underline{A}_R = \underline{A} - \underline{b} \gamma^T$$

$$= \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \vdots & & 1 \\ -a_0 & -a_1 & -a_2 & & -a_{n-1} \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} [\gamma_1, \gamma_2, \dots, \gamma_n]$$

$$= \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \underbrace{-(a_0 + \gamma_1)}_{a_{R,0}} & \underbrace{-(a_1 + \gamma_2)}_{a_{R,1}} & \vdots & & \underbrace{-(a_{n-1} + \gamma_n)}_{a_{R,n-1}} \end{bmatrix}$$

The following is valid,

$$a_{R,0} = a_0 + \gamma_1$$

$$a_{R,1} = a_1 + \gamma_2$$

$$\vdots$$

$$a_{R,n-1} = a_{n-1} + \gamma_n$$

$$\textcircled{a_{R,0}}$$

$$= a_0 + \gamma_1$$

$$a_0 + \gamma_1$$

$$\gamma_1 = \frac{a_{R,0}}{b_0}$$

controllability canonical matrix - Co control matrix $b = 0 \dots 1$

from this characteristic polynomial for the overall system can be calculated as follows:

$$s^n + a_{R,n-1}s^{n-1} + \dots + a_{R,1}s + a_{R,0} = 0$$

By a comparison of coefficients pole placement can be executed

Ascertainment of input gain r_w :

→ Considering the steady state case:

Here it should be valid: $y_{ss} = w$ (request!!!)

It applies to the steady state case: $\dot{x} = 0$

from the last line of the differential eqn results,

$$\text{from } \Rightarrow \dot{x}_n = -a_{R,0}x_1 - a_{R,1}x_2 - a_{R,2}x_3 - \dots - a_{R,n-1}x_{n-1} + r_w w$$

\uparrow \uparrow \uparrow \uparrow
 x_1 x_2 x_{n-1}

(also from matrix)
state \dot{x}

$$\leadsto 0 = -a_{R,0}x_1 + r_w w$$

$$x_1 = \frac{1}{a_{R,0}} r_w w$$

$$y = b_0 x_1 + b_1 x_2 + b_2 x_3 + \dots + b_{n-1} x_n$$

\uparrow \uparrow \uparrow
 x_1 x_2 x_{n-1}

$$\leadsto y = b_0 x_1$$

$$= b_0 \times \frac{1}{a_{R,0}} r_w w$$

on the other hand for $t \rightarrow \infty$ it is also valid

$$y = w \quad \leadsto \frac{b_0}{a_{R,0}} r_w = 1 \quad \leadsto \boxed{r_w = \frac{a_{R,0}}{b_0}}$$

$$a_{R,n-1} = a_{n-1} + m$$

next week exercise 2

$$s^m + (a_{R,n-1})s^{n-1} + \dots + a_{R,0} = (s-x_1)(s-x_2)\dots(s-x_n)$$

Exercise 2

1. Task:

$$a) \dot{x}_1(t) = a_{11}x_1(t) + 3x_2(t) + 0$$

$$\dot{x}_2(t) = 4x_1(t) - 6x_2(t) + b_2 u(t)$$

$$y(t) = c_1 x_1(t) + c_2 x_2(t) + d u(t)$$



$$sX_1(s) = a_{11}X_1(s) + 3X_2(s)$$

$$sX_2(s) = 4X_1(s) - 6X_2(s) + b_2 U(s)$$

$$Y(s) = c_1 X_1(s) + c_2 X_2(s) + d U(s)$$

conversion of the eqs in the frequency domain; from the 1st eqⁿ follows:

$$(s - a_{11})X_1 = 3X_2$$

$$X_1 = \frac{3X_2}{s - a_{11}}$$

inserted into the 2nd eqⁿ

$$sX_2 = 4 \cdot \frac{3X_2}{s - a_{11}} - 6X_2 + b_2 U$$

$$\Rightarrow s^2 X_2 - s a_{11} X_2 = 12 X_2 - 6 s X_2 + 6 a_{11} X_2 + b_2 s U - b_2 a_{11} U$$

$$X_2 = \frac{b_2 s - b_2 a_{11}}{s^2 + s(6 - a_{11}) - (12 + 6 a_{11})} U$$

$$\therefore X_1 = 3 \cdot \frac{b_2 s - b_2 a_{11}}{s^2 + s(6 - a_{11}) - (12 + 6 a_{11})} \cdot \frac{1}{s - a_{11}} U$$

$$\Rightarrow X_1 = \frac{3 b_2}{s^2 + s(6 - a_{11}) - (12 + 6 a_{11})} U$$

$$Y = c_1 X_1 + c_2 X_2 + d U$$

$$= \frac{d s^2 + (6d + c_2 b_2 - d a_{11})s + (3c_1 b_2 - c_2 b_2 a_{11} - 12d - 6a_{11}d)}{s^2 + s(6 - a_{11}) - (12 + 6a_{11})} \cdot U$$



inverse Laplace transformation

$$\ddot{y}(t) + (6 - a_{11})\dot{y}(t) - (12 + 6a_{11})y(t) = (3c_1 b_2 - c_2 b_2 a_{11} - 12d - 6a_{11}d)u(t) + (6d + c_2 b_2 - d a_{11})\dot{u}(t) + d\ddot{u}(t)$$

(b) with $d=0$ the 2nd derivative of the input variable u disappears,

c) result of a) with $a_{11} = 0$, $Q = 0$ and $d = 0$

$$\ddot{y}(t) + 6\dot{y}(t) - 12y(t) = 3G_1 b_2 u(t)$$

→ Conformity

2) a) Time invariant, linear, multivariable system

$$b) \underline{G}(s) = \underline{C}(sI - \underline{A})^{-1} \underline{B} + \underline{D}$$

$$G(s) = \begin{bmatrix} 3 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} s+2 & -1 \\ 0 & s+1 \end{bmatrix}^{-1} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & d_{22} \end{bmatrix}$$

$$= \begin{bmatrix} 3 & 0 \\ 0 & 0 \end{bmatrix} \frac{\begin{bmatrix} s+1 & 1 \\ 0 & s+2 \end{bmatrix}}{(s+2)(s+1)} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & d_{22} \end{bmatrix}$$

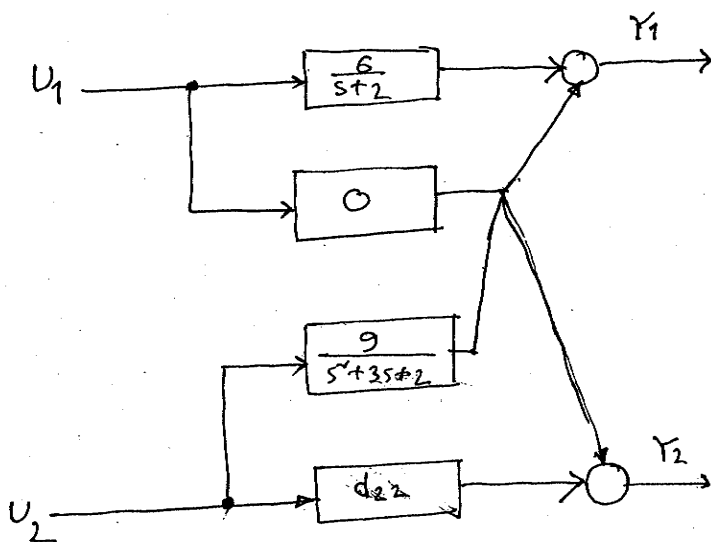
$$G(s) = \begin{bmatrix} \frac{6}{s+2} & \frac{9}{s^2+3s+2} \\ 0 & d_{22} \end{bmatrix}$$

$$c) \underline{Y}(s) = \underline{G}(s) \cdot \underline{U}(s)$$

$$Y_1(s) = \frac{6}{s+2} U_1(s) + \frac{9}{s^2+3s+2} U_2(s)$$

$$Y_2(s) = d_{22} U_2(s)$$

0 · U₁(s) +



c) Action of signals:

$$u_1(t) \rightarrow y_1(t):$$

effect via the control matrix, evolution matrix, observation matrix

$$u_1(t) \rightarrow y_2(t)$$

no effect.

$$u_2(t) \rightarrow y_1(t)$$

effect via the control matrix

Couplings in the evolution matrix, observation matrix

$$u_2(t) \rightarrow y_2(t):$$

Exclusively effect via the ~~to~~ direct transmission matrix

Exercise 3

1. (a) At both systems no couplings are present in each evolution matrix. That means, every state variable must have its own input through the system input (in case of controllability) as well as every state variable must be observable (in case of observability) through the system output.

- first example: $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ in ~~the~~ control matrix's element. So, not controllable
- 2nd example: $\begin{bmatrix} 0 & 4 \end{bmatrix}$ observation matrix, element 0. So, not observable.

1st system:

- not controllable (zero element in \underline{b})
- observable (no zero element in \underline{c}^T)

2nd system:

- controllable (no zero element in \underline{b})
- not observable (zero element in \underline{c}^T)

(b) Controllability matrix

$$\underline{Q}_c = [\underline{b}, A\underline{b}]$$

$$= \begin{bmatrix} 1, & \begin{bmatrix} -2 & 0 \\ 0 & -3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix}$$

$$\text{rank } \underline{Q}_c = 1 \neq 2$$

→ the matrix is not controllable

Observability matrix

$$\underline{Q}_o = \begin{bmatrix} \underline{c}^T \\ \underline{c}^T A \end{bmatrix} = \begin{bmatrix} 2 & 4 \\ \begin{bmatrix} -2 & 0 \\ 0 & -3 \end{bmatrix} \end{bmatrix}$$

$$\Rightarrow \underline{Q}_o = \begin{bmatrix} 2 & 4 \\ -4 & -12 \end{bmatrix}$$

$$\text{rank } \underline{Q}_o = 2$$

→ the system is observable.

2nd System:

Controllability matrix

$$Q_c = [b, A b] \\ = \begin{bmatrix} 1, & \begin{bmatrix} -2 & 0 \\ 0 & -3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 1 & -2 \\ 2 & -6 \end{bmatrix}$$

$$\text{rank } Q_c = 2$$

→ the system is controllable

Observability matrix

$$Q_o = \begin{bmatrix} c^T \\ c^T A \end{bmatrix} = \begin{bmatrix} 0 & 4 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} 1 & -2 \\ 0 & -3 \end{bmatrix}$$

$$Q_o = \begin{bmatrix} 0 & 4 \\ 0 & -12 \end{bmatrix}$$

$$\text{rank } Q_o = 1 \neq 2$$

→ the system is not observable

2. Co-efficients,

$$a_0 = 0 \quad b_0 = 2$$

$$a_1 = 1 \quad b_1 = 0$$

$$a_2 = 3 \quad b_2 = 4$$

$$(b_n =) b_3 = 1$$

controllability Canonical form:

$$\dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -a_0 & -a_1 & -a_2 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u$$

$$y = [b_0 - b_n a_0 \mid b_1 - b_n a_1 \mid b_2 - b_n a_2] x + b_n u$$

∴ for the present system

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & -3 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u(t)$$

$$y(t) = [2 \quad -1 \quad 1] x(t) + 1 \cdot u(t) \rightarrow \text{Controllable}$$

$$Q_c = [b \quad A b, A^2 b]$$

b) The system is controllable
If the controllability canonical form exists,
the controllability is principally present.

3. The shown system is neither controllable nor observable. As both state variables $x_1(t)$ and $x_2(t)$ show the same dynamic behavior, they can not be stimulated separately by the system input. Furthermore, taking the equality in the dynamics into account, at the system output no distinction betⁿ the influences of the state variable is possible.

Proof:

$$A = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$c^T = [1 \quad 1]$$

Controllability:

$$Q_c = [b \quad Ab] = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$$

Rank $Q_c = 1 \neq 2 \rightarrow$ not controllable

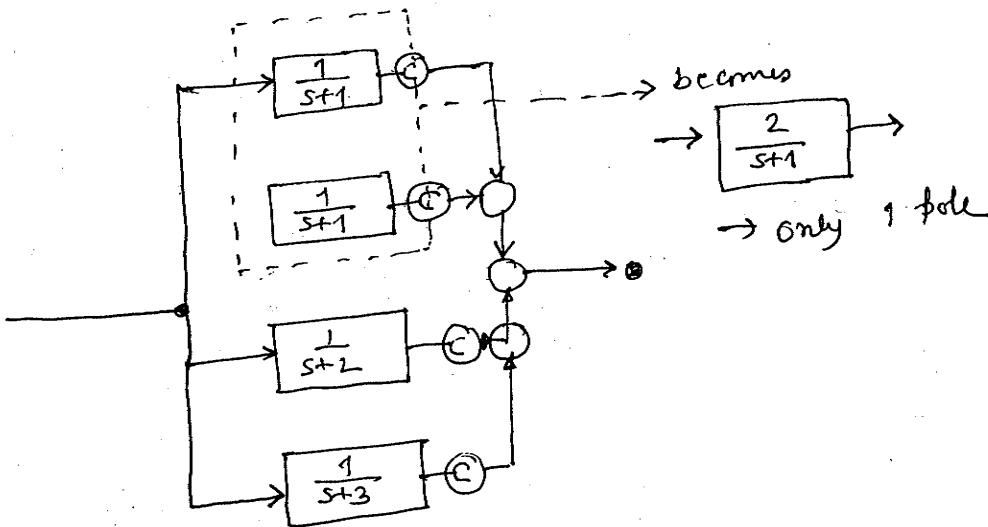
Observability:

$$Q_o = \begin{bmatrix} c^T \\ c^T A \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix}$$

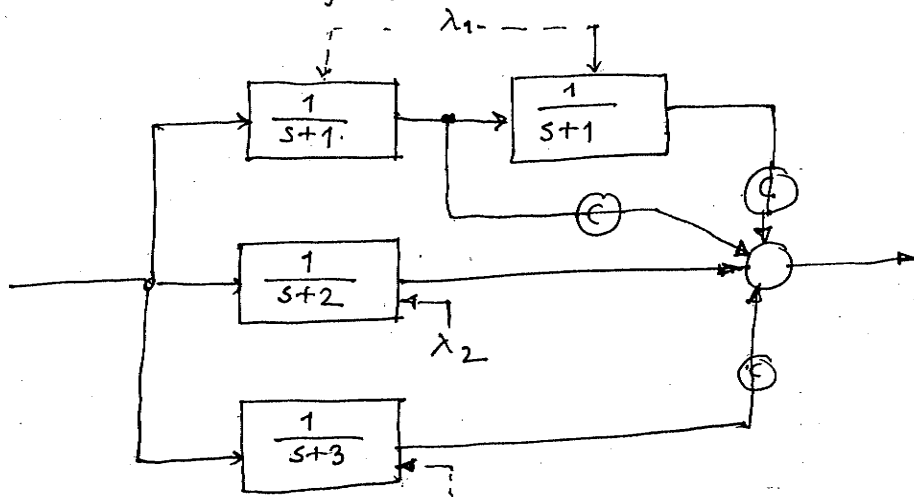
Rank $Q_o = 1 \neq 2 \rightarrow$ not observable

Additional considerations regarding the Jordan Canonical form (not component of the solution)

Parallel connection of 2 identical poles yields a 1 pole



w.r.t the rules of partial decomposition, the following is valid:



$$\dot{\underline{x}}(t) = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 1 & \lambda_1 & 0 \\ 0 & \lambda_2 & \lambda_3 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix} u(t)$$

the corresponds to the rules of decomposition...

2x2

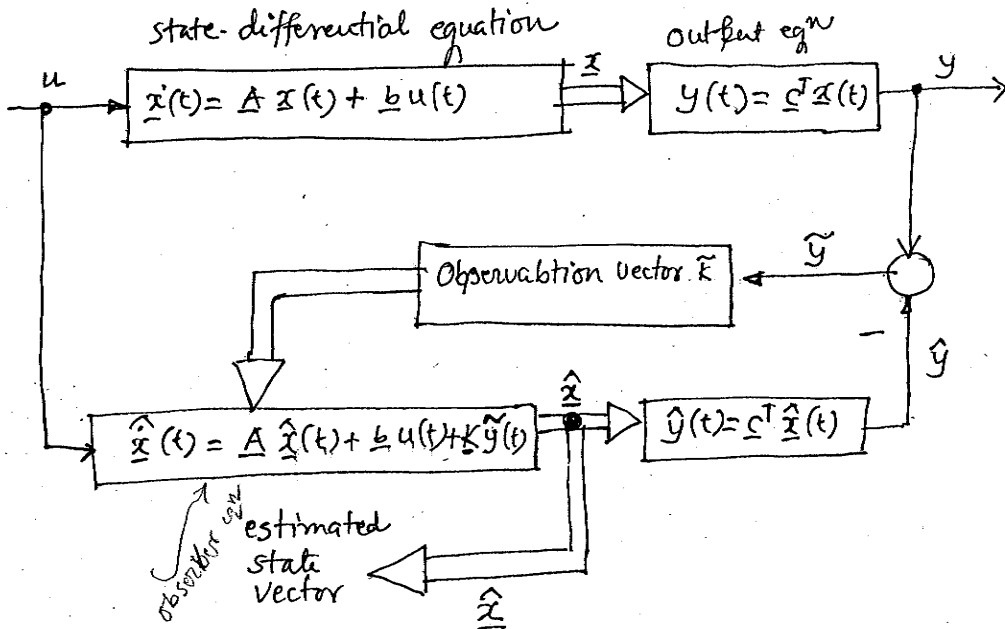
$$a_1 \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

State Estimation by an Observer

Problem:

for a system, whose only the input and the output variables can be measured, the prerequisite for the design of a state feedback controller shall be established. Here the system dynamics are well known.

→ Computational reconstruction of the progression of the state variables by an observer.



The feedback of the estimation error \tilde{y} between the measured output variable y and the estimated output variable \hat{y} into the observer equations prevents from diverging of \hat{y} from y . This is also valid in case of short-term disturbances into the controlled system.

state equations of the system: (single input/output system with $d=0$)

$$\dot{x}(t) = A x(t) + b u(t)$$

$$y(t) = c^T x(t)$$

Resulting from this the observer equations are

$$\dot{\hat{x}}(t) = A \hat{x}(t) + b u(t) + k (y(t) - \hat{y}(t))$$

$$\hat{y}(t) = c^T \hat{x}(t)$$

calculation of the estimation error $\tilde{x}(t) = x(t) - \hat{x}(t)$

$$\dot{\tilde{x}}(t) = A x(t) + b u(t)$$

$$- A \hat{x}(t) - b u(t) - k (y(t) - \hat{y}(t))$$

$$= A (x(t) - \hat{x}(t)) - k (c^T x(t) - c^T \hat{x}(t))$$

$$= A \tilde{x}(t) - k c^T \tilde{x}(t)$$

→ Estimation error differential equation

$$\dot{\tilde{x}}(t) = (\underline{A} - \underline{K}\underline{C}^T) \tilde{x}(t)$$

$$\rightarrow \tilde{A} = \underline{A} - \underline{K}\underline{C}^T \Rightarrow \dot{\tilde{x}}(t) = \tilde{A} \tilde{x}(t)$$

It is required that with any initial state \underline{x}_0 , $\hat{\underline{x}}_0$ the estimation error $\tilde{x}(t)$ for $t \rightarrow \infty$ converges to $\tilde{x}(t) \rightarrow 0$

This happens precisely in the case, in which the matrix \tilde{A} consists exclusively of eigenvalues with negative real parts.

- Calculations of the observation vector \underline{K} by pole placement for the observation error:

Transforming of the estimating error differential equation into the frequency domain.

$$s\tilde{x}(s) = (\underline{A} - \underline{K}\underline{C}^T) \tilde{x}(s) \Rightarrow (s\underline{I} - \underline{A} + \underline{K}\underline{C}^T) \tilde{x}(s) = 0$$
$$\rightarrow \tilde{x}(s) = (s\underline{I} - \underline{A} + \underline{K}\underline{C}^T)^{-1} \cdot 0 \Rightarrow \underline{x}(s) = (s\underline{I} - \underline{A} + \underline{K}\underline{C}^T)^{-1} \cdot 0$$

The Eigenvalues (poles) for the estimation error result from the characteristic equation

$$\det(s\underline{I} - \underline{A} + \underline{K}\underline{C}^T) = 0$$

Equivalent approach for the characteristic eq^s

$$(s - \lambda_1^B) (s - \lambda_2^B) \dots (s - \lambda_n^B) = 0$$

Setting equal these two approaches results in

$$\det(s\underline{I} - \underline{A} + \underline{K}\underline{C}^T) = (s - \lambda_1^B) (s - \lambda_2^B) \dots (s - \lambda_n^B)$$

The comparison of coefficients results in a system of n equations, that can be resolved after the n elements of the observation vector \underline{K} . $\lambda_1^B, \lambda_2^B, \dots, \lambda_n^B$ are the given eigenvalues for the estimation error differential equation of the observer. It is expedient to choose exclusively eigenvalues, which are located on the negative real axis in the complex s plane.

Procedure for the concrete application

- step 1:

Calculation of the left side of the equation by inserting the elements for \underline{A} and \underline{C}^T .

The unknown parameters for $\underline{K} = \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_n \end{bmatrix}$ remain inserted in general form.

• Step 2:

Calculation of the right side of the equation by the determination of the Eigenvalues (poles) $\lambda_1^B, \lambda_2^B, \dots, \lambda_n^B$

• Step 3:

Comparison of the co-efficients for both sides of the equation regarding all powers of the complex frequency s

• Step 4:

solving n equations according to n unknowns. Then, the observation vector \underline{K} is found.

Prerequisite for the design of an observer is the observability of the system.

Example:

For the following system \underline{K} should be found:

$$\dot{\underline{x}}(t) = \begin{bmatrix} -3 & 1 \\ 0 & -4 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 \\ 2 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} \frac{2}{3} & \frac{1}{3} \end{bmatrix} \underline{x}(t)$$

Determination of the observability

$$\underline{Q}_0 = \begin{bmatrix} \underline{c}^T \\ \underline{c}^T \underline{A} \end{bmatrix} = \begin{bmatrix} \frac{2}{3} & \frac{1}{3} \\ -\frac{6}{3} & -\frac{2}{3} \end{bmatrix} \rightarrow \text{rank } \underline{Q}_0 = 2$$

\rightarrow the system is observable
an observer can be designed.

Task: The Eigenvalues of the estimation error ^{differential} equation shall be placed at $\lambda_1^B = -10$, $\lambda_2^B = -20$.

$$\det(s\underline{I} - \underline{A} + \underline{K}\underline{c}^T) = (s - \lambda_1^B) \cdot (s - \lambda_2^B)$$

1. left side of the equation with $\underline{K} = \begin{bmatrix} k_1 \\ k_2 \end{bmatrix}$

$$\det(s\underline{I} - \underline{A} + \underline{K}\underline{c}^T) = \det \begin{bmatrix} s+3 + \frac{2}{3}k_1 & -1 + \frac{1}{3}k_1 \\ \frac{2}{3}k_2 & s+4 + \frac{1}{3}k_2 \end{bmatrix}$$

$$= s^2 + \left(\frac{2}{3}k_1 + \frac{1}{3}k_2 + 7\right)s + \frac{8}{3}k_1 + \frac{5}{3}k_2 + 12$$

2. right side of the equation 1

$$(s - \lambda_1^B)(s - \lambda_2^B) = (s+10)(s+20) \\ = s^2 + 30s + 200$$

3. Comparison of coefficients for both sides of the eqn.

$$s^2: 1 = 1$$

$$s: \frac{2}{3}k_1 + \frac{1}{3}k_2 + 7 = 30 \rightarrow k_1 = \frac{69 - k_2}{2}$$

$$\therefore: \frac{8}{3}k_1 + \frac{5}{3}k_2 + 12 = 200$$

4. solving the eqns:

$$\frac{8}{3} \cdot \frac{69 - k_2}{2} + \frac{5}{3}k_2 + 12 = 0$$

$$\rightarrow k_2 = 288$$

$$k_1 = (69 - 288)/2 = -\frac{219}{2} \rightarrow \underline{k} = \begin{bmatrix} -\frac{219}{2} \\ 288 \end{bmatrix}$$

Special case: The system already exists in the observability canonical form

$$\dot{\underline{x}}(t) = \begin{bmatrix} 0 & 0 & \dots & -a_0 \\ 1 & 0 & & -a_1 \\ 0 & 1 & & -a_2 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & -a_{n-1} \end{bmatrix} \underline{x}(t) + \begin{bmatrix} b_0 \\ b_1 \\ b_2 \\ \vdots \\ b_{n-1} \end{bmatrix} u(t)$$

$$y(t) = [0 \ 0 \ \dots \ 1] \underline{x}(t)$$

Introduction of an observer

$$\dot{\hat{\underline{x}}}(t) = \underline{A} \hat{\underline{x}}(t) + \underline{b} u(t) + \underline{k} (y(t) - \hat{y}(t))$$

$$\text{with } \underline{k} = \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_n \end{bmatrix}$$

Estimation error differential equation

$$\dot{\tilde{\underline{x}}}(t) = \underbrace{(\underline{A} - \underline{k} \underline{c}^T)}_{\tilde{\underline{A}}} \tilde{\underline{x}}(t)$$

The evolution matrix for the estimation error results in

$$\tilde{A} = A - KC^T$$

$$= \begin{bmatrix} 0 & \dots & -a_0 \\ 1 & 0 & -a_1 \\ 0 & 1 & -a_2 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 & -a_{n-1} \end{bmatrix} - \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_n \end{bmatrix} \begin{bmatrix} 0 & 0 & \dots & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 0 & \dots & (-a_0 - k_1) \\ 1 & 0 & & (-a_1 - k_2) \\ 0 & 1 & & \\ \vdots & \vdots & \ddots & \\ 0 & 0 & & 1 & (-a_{n-1} - k_n) \end{bmatrix}$$

It is valid:

$$a_{B,0}^k = a_0 + k_1$$

$$a_{B,1}^k = a_1 + k_2$$

⋮

$$a_{B,n-1}^k = a_{n-1} + k_n$$

The characteristic polynomial can be determined from this as follows

$$s^n + a_{B,n-1}^k s^{n-1} + \dots + a_{B,1}^k s + a_{B,0}^k = 0$$

Herewith, pole placement can be carried out by the comparison of co-efficients.

Next

- Eigen values
- Non linear Systems

Oleg Bauer
H-F 103

EX-4

Task 1.

$$\begin{aligned}
 F(s) &= \frac{(s^2 + 3s + 2)(s + 4)}{(s + 1)^2 (s + 3)^2} \\
 &= \frac{(s + 1)(s + 2)(s + 4)}{(s + 1)^2 (s + 3)^2} \\
 &= \frac{(s + 2)(s + 4)}{(s + 1)(s + 3)^2}
 \end{aligned}$$

Numerator and denominator of $F(s)$ now have no common divisor.

Double poles at $s = -3$

Partial fraction decomposition.

$$\begin{aligned}
 F(s) &= \frac{A}{s + 1} + \frac{B_1}{s + 3} + \frac{B_2}{(s + 3)^2} = \frac{A}{s + 1} + \frac{(s + 3)^2}{(s + 3)^2} \cdot \frac{B_1}{s + 3} + \frac{B_2}{(s + 3)^2} \\
 &= \frac{A}{s + 1} + \frac{B_1}{s + 3} + \frac{B_2}{(s + 3)^2} \cdot \frac{s + 1}{s + 1} \\
 &= \frac{As^2 + 6As + 9A + B_1s^2 + 4B_1s + 3B_1 + B_2s + B_2}{(s + 1)(s + 3)^2} \\
 &= \frac{(A + B_1)s^2 + (6A + 4B_1 + B_2)s + (9A + 3B_1 + B_2)}{(s + 1)(s + 3)^2}
 \end{aligned}$$

Comparison of coefficients:

$$\begin{aligned}
 \text{Coefficients in } &= \frac{s^2 + 6s + 8}{(s + 1)(s + 3)^2}
 \end{aligned}$$

$$s^2: A + B_1 = 1 \Rightarrow A = 1 - B_1$$

$$s: 6A + 4B_1 + B_2 = 6 \rightarrow -2B_1 + B_2 = 0$$

$$s^0: 9A + 3B_1 + B_2 = 8$$

$$\therefore B_2 = 2B_1$$

$$\Rightarrow 9 - 9B_1 + 3B_1 + 2B_1 = 8$$

$$B_1 = \frac{1}{4}, B_2 = \frac{1}{2}, A = \frac{3}{4}$$

from $Y(s) = F(s) \cdot U(s)$ it follows,

$$Y(s) = \underbrace{\frac{3}{4} \cdot \frac{1}{s + 1}}_{X_1(s)} \cdot U(s) + \underbrace{\frac{1}{4} \cdot \frac{1}{(s + 3)}}_{X_2(s)} \cdot U(s) + \underbrace{\frac{1}{2} \cdot \frac{1}{(s + 3)^2}}_{X_3(s)} \cdot U(s)$$

Introduction of state variables

$$sX_1(s) = -X_1(s) + U(s)$$

$$sX_2(s) = -3X_2(s) + U(s)$$

$$sX_3(s) = -3X_3(s) + X_2(s)$$

$$\left\{ \begin{array}{l} X_3(s) = \frac{1}{s+3} \cdot \frac{1}{s+3} \cdot U(s) \\ X_3(s) = \frac{1}{s+3} \cdot X_2(s) \end{array} \right.$$



Inverse Laplace Transformation.

$$\dot{\underline{x}}(t) = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 1 & -3 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} \frac{3}{4} & \frac{1}{4} & \frac{1}{2} \end{bmatrix} \underline{x}(t)$$

Task #2.

Transformation matrix,

$$T = \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix}$$

$$T^{-1} = \frac{\begin{bmatrix} 0 & 1 \\ -1 & 2 \end{bmatrix}}{-1}$$

$$= \begin{bmatrix} 0 & 1 \\ 1 & -2 \end{bmatrix}$$

$$\bar{A} = T \cdot A \cdot T^{-1}$$

$$= \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 \\ 0 & 2 & 3 \\ 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -2 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 8 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -2 \end{bmatrix} = \begin{bmatrix} 8 & -12 \\ 3 & -4 \end{bmatrix}$$

$$\bar{B} = T \cdot B = \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$$

$$\bar{C} = C \cdot T^{-1} = \begin{bmatrix} 3 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \end{bmatrix}$$

$$\boxed{\bar{D} = D = 0} \quad \underline{d} = \underline{d} = 0$$

$$\dot{\bar{x}}(t) = \begin{bmatrix} 3 & -12 \\ 3 & -4 \end{bmatrix} \cdot \bar{x}(t) + \begin{bmatrix} 4 \\ 1 \end{bmatrix} \cdot u(t)$$

$$y(t) = \begin{bmatrix} 1 & 1 \end{bmatrix} \cdot \bar{x}(t)$$

b) $\bar{x}(t) = I x(t)$

$$\bar{x}_1(t) = 2x_1(t) + x_2(t)$$

$$x_2(t) = x_1(t)$$

c) Calculation of the complex transfer functions

original system

$$F(s) = \underline{c}^T (sI - A)^{-1} \underline{b} + d$$

$$\begin{aligned} &= \begin{bmatrix} 3 & 1 \end{bmatrix} \begin{bmatrix} s-2 & -3 \\ 0 & s-2 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ 2 \end{bmatrix} \\ &= \frac{\begin{bmatrix} 3s-6 & 3+s-2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix}}{(s-2)(s-2)} \\ &= \frac{3s-6+18+2s-4}{s^2-4s+4} = \frac{5s+8}{s^2-4s+4} \end{aligned}$$

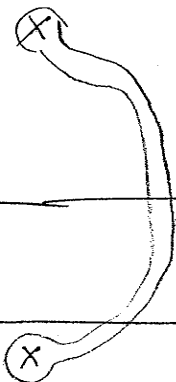
Transformed system:

$$\begin{aligned} \bar{F}(s) &= \bar{c}^T (sI - \bar{A})^{-1} \bar{b} + \bar{d} \\ &= \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} s-8 & 12 \\ -3 & s+4 \end{bmatrix}^{-1} \begin{bmatrix} 4 \\ 1 \end{bmatrix} \\ &= \frac{\begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} s+4 & -12 \\ 3 & s-8 \end{bmatrix} \begin{bmatrix} 4 \\ 1 \end{bmatrix}}{s^2-4s-32+36} \\ &= \frac{\begin{bmatrix} s+7 & s-20 \end{bmatrix} \begin{bmatrix} 4 \\ 1 \end{bmatrix}}{s^2-4s+4} \\ &= \frac{4s+28+s-20}{s^2-4s+4} = \frac{5s+8}{s^2-4s+4} \end{aligned}$$

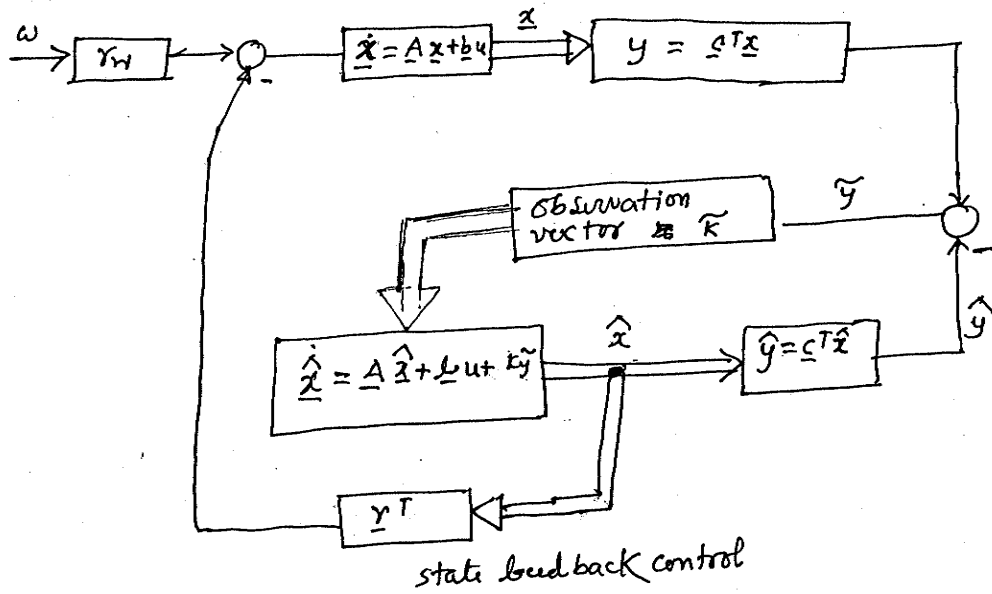
$$\therefore F(s) = \bar{F}(s)$$

\therefore both complex transfer functions, which describes the input/output behavior of the equivalent (transformed) systems are identical.

Consequence of the Equivalent system transformation:

	transformation Variant	transformation Invariant
A, B, C	X	
<u>D</u>		X
input/output behavior		X
Internal system behavior		
$u(t), y(t)$		X
$x(t)$		
controllability, observability		X
Eigen values		X
Eigen vectors	X	

The observer in the closed loop



Theorem of separation:

The Eigen values of the closed loop without the observer are not displaced by the insertion of the observer. The Eigen values of the observer are just added to them.

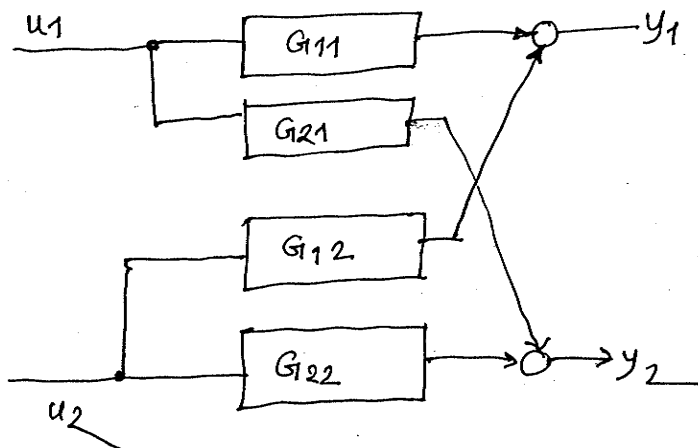
Procedure:

one develops first the automatic controller without regard of the observer. Then the observer is to be designed inserted into the closed loop.

The decoupling of multivariable systems

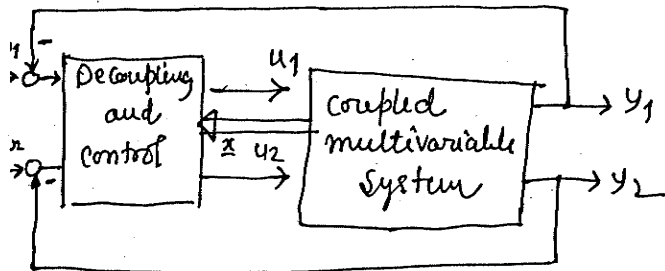
starting point: multivariable systems e.g. with 2 inputs and 2 output variables.

Task: A suitable control system must be found.



Tasks of the control system

- * Decoupling of the multivariable systems
- * Pole placement for the closed loop
- * Comparison between the desired and the controlled variables.



The multivariable system is given in general by

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

$$y(t) = C x(t) + D u(t)$$

The controlled variables are represented by the output variables $y_1(t)$, $y_2(t)$, ..., $y_n(t)$ of the system.

For the ~~out~~ i -th output variables it applies $\dot{x}(t) = A x(t) + B u(t)$

$$y_i(t) = c_i^T x(t) + d_i^T u(t)$$

c_i^T and d_i^T are the i -th lines of C and D each

Differential Order:

The differential order δ_i determines, on the which ^{derivative of the output variable} of the ~~outputs variable~~ y_i for the first time an ⁱⁿ output variable affects. Thereby, it is unimportant which input $u_1(t)$, $u_2(t)$, ..., $u_y(t)$ acts on it.

The following is valid:

$$\text{If } \delta_i = 0; y_i(t) = c_i^T x(t) + d_i^T u(t) \Rightarrow d_i^T \neq 0^T$$

$$\delta_i = 1 \quad y_i(t) = c_i^T x(t)$$

$$\dot{y}_i(t) = c_i^T \dot{x}(t)$$

$$= c_i^T (A x(t) + B u(t))$$

$$\rightarrow c_i^T B \neq 0^T$$

$$\delta_i = 2: y_i(t) = c_i^T \cdot x(t)$$

$$\dot{y}_i(t) = c_i^T \cdot A x(t)$$

$$\ddot{y}_i(t) = c_i^T A (A x(t) + B u(t))$$

From the evaluation of it follows,

$$\delta_i = 0 : \underline{d}_i^T \neq \underline{0}^T$$

$$\delta_i = 1 : \underline{d}_i^T = \underline{0}^T, \underline{c}_i^T \underline{B} \neq \underline{0}^T$$

$$\delta_i = 2 : \underline{d}_i^T = \underline{0}^T, \underline{c}_i^T \underline{B} = \underline{0}^T, \underline{c}_i^T \underline{A} \cdot \underline{B} \neq \underline{0}^T$$

$$\delta_i = 3 : \underline{d}_i^T = \underline{0}^T, \dots \underline{c}_i^T \underline{A} \cdot \underline{B} = \underline{0}^T, \underline{c}_i^T \underline{A}^2 \cdot \underline{B} \neq \underline{0}^T$$

$$\delta_i = 4 : \underline{d}_i^T = \underline{0}^T, \dots \underline{c}_i^T \underline{A}^2 \cdot \underline{B} = \underline{0}^T, \underline{c}_i^T \underline{A}^3 \cdot \underline{B} \neq \underline{0}^T$$

ie. $\delta_i = \min k$, to which applies,

$$\underline{c}_i^T \underline{A}^{k-1} \cdot \underline{B} \neq \underline{0}^T$$

However, if $\underline{d}_i^T \neq \underline{0}^T$ then

$\delta_i = 0$ is valid

consequently it applies

$$\underline{y}^{(i)}(t) = \underline{c}_i \underline{x}(t) + \underline{d}_i^T \underline{u}(t)$$

$$\underline{y}_i^{(i)}(t) = \underbrace{\underline{c}_i^T \underline{A}^{\delta_i}}_{\underline{c}_i^{*T}} \underline{x}(t) + \underbrace{\underline{c}_i^T \underline{A}^{\delta_i-1} \underline{B}}_{\underline{d}_i^{*T}} \underline{u}(t)$$

for $\underline{d}_i^T \neq \underline{0}^T$ prove

for $\underline{d}_i^T = \underline{0}^T$

Summarized, it applies

$$\underline{y}^*(t) = \underline{c}^* \underline{x}(t) + \underline{D}^* \underline{u}(t)$$

where

$$\underline{c}^* = \begin{bmatrix} \underline{c}_1^{*T} \\ \vdots \\ \underline{c}_q^{*T} \end{bmatrix} \quad \text{and} \quad \underline{D}^* = \begin{bmatrix} \underline{d}_1^{*T} \\ \vdots \\ \underline{d}_q^{*T} \end{bmatrix}$$

The elements of $\underline{y}^*(t)$

$$\underline{y}^*(t) = \begin{bmatrix} y_1^*(t) \\ \vdots \\ y_q^*(t) \end{bmatrix} = \begin{bmatrix} y_1(\delta_1) \\ \vdots \\ y_q(\delta_q) \end{bmatrix}$$

contain those derivatives of the output variables $y_1(t) \dots y_q(t)$, on which for the first time directly an input variable $u_1(t) \dots u_q(t)$ affects.

Design of the decoupling control system

The input/output of the system is described by $\underline{y}^*(t) = \underline{c}^* \underline{x}(t) + \underline{D}^* \underline{u}(t)$

Approach for the control algorithm,

$$u(t) = \underline{D}^{*-1} \left(-(\underline{C}^* + \underline{M}^*) \underline{x}(t) + \underline{L} \underline{w}(t) \right)$$

Prerequisite for decoupleability,

→ \underline{D}^* invertible

$$\Rightarrow \det(\underline{D}^*) \neq 0$$

One receives for the overall dynamics of the closed loop

$$\underline{y}^*(t) = \underline{C}^* \underline{x}(t) + \underline{D}^* \underline{D}^{*-1} \left(-(\underline{C}^* + \underline{M}^*) \underline{x}(t) + \underline{L} \underline{w}(t) \right)$$

$$\underline{y}^*(t) = -\underline{M}^* \underline{x}(t) + \underline{L} \underline{w}(t)$$

Conclusion: With decoupling control system a new dynamics can be impressed to the closed loop.

Therefore, the original system dynamics is compensated.

Calculation of the matrices M^* & L

starting point: Desired dynamics for the i -th isolated subsystem

$$y_i^{(\delta_i)}(t) = -\alpha_i \delta_{i-1} y_i^{(\delta_i-1)}(t) - \dots - \alpha_{i1} \dot{y}_i(t) - \alpha_{i0} y_i(t) + l_i \omega_i(t)$$

with α_{ij} and l_i constants co-efficients of the differential equation

Normally it is

$$\left. \begin{matrix} \alpha_{i0} = l_i \text{ controller gain} \\ \alpha_{i1} \\ \vdots \\ \alpha_i, \delta_{i-1} \end{matrix} \right\} \text{determination of the Eigenvalues of the differential eq}^n$$

For the evolution of the i -th line $M_i^{*T} \underline{x}(t)$ of $M^* \underline{x}(t)$ it is valid

$$M_i^{*T} \underline{x}(t) = \begin{cases} 0 & \text{for } \delta_i = 0 \\ \sum_{k=0}^{\delta_i-1} \alpha_{ik} \cdot y^{(k)}(t) & \text{for } \delta_i > 0 \end{cases}$$

With the derivatives of y_i it results

$$y_i(t) = \underline{c}_i^T \underline{x}(t)$$

$$\dot{y}_i(t) = \underline{c}_i^T \underline{A} \underline{x}(t)$$

$$\ddot{y}_i(t) = \underline{c}_i^T \underline{A}^2 \underline{x}(t)$$

$$y_i^{(\delta_i-1)}(t) = \underline{c}_i^T \underline{A}^{\delta_i-1} \underline{x}(t)$$

$$M_i^{*T} \underline{x}(t) = \begin{cases} 0 & \text{for } \delta_i = 0 \\ \sum_{k=0}^{\delta_i-1} \alpha_{ik} \underline{c}_i^T \underline{A}^k \underline{x}(t) & \text{for } \delta_i > 0 \end{cases}$$

From this one receives the conditional equation for M_i^{*T}

$$M_i^{*T} = \begin{cases} 0^T & \text{for } \delta_i = 0 \\ \sum_{k=0}^{\delta_i-1} \alpha_{ik} \underline{c}_i^T \underline{A}^k & \text{for } \delta_i > 0 \end{cases}$$

Determination of \underline{L} :

$$l_i = \alpha_i 0$$

$$\underline{L} = \begin{bmatrix} l_1 & & 0 \\ & \ddots & \\ 0 & & l_q \end{bmatrix}$$

The transfer matrix of overall system

$$\underline{G}(s) = \begin{bmatrix} G_1(s) & & 0 \\ & \ddots & \\ 0 & & G_q(s) \end{bmatrix}$$

Here with, $G_i(s)$, $i=1, \dots, q$ results in

$$G_i(s) = \frac{Y_i(s)}{W_i(s)}$$

$$G_i(s) = \begin{cases} l_i & \text{for } \delta_i = 0 \\ \frac{l_i}{s^{\delta_i} + \alpha_{i,\delta_i-1}s^{\delta_i-1} + \dots + \alpha_{i,1}s + \alpha_{i,0}} & \text{for } \delta_i > 0 \end{cases}$$

Procedure for the design of the decoupling control system:

Step 1:

Determine the differential order for every subsystem, (Every subsystem complies precisely with one output variable each.)

$$\delta_i = \begin{cases} 0, & \text{if } \underline{a}_i^T \neq \underline{0}^T \\ k, & \text{if } k \text{ is the smallest number} \\ & \text{in the range } 0 < k \leq n, \text{ for which} \\ & \text{the following is valid: } \underline{s}_i^T \underline{A}^{k-1} \underline{B} \neq \underline{0}^T \end{cases}$$

Step 2:

Calculate \underline{c}^* and \underline{d}^*

$$\underline{c}^* = \begin{bmatrix} \underline{s}_1^{*\top} \\ \vdots \\ \underline{s}_q^{*\top} \end{bmatrix}, \quad \underline{c}_i^{*\top} = \underline{c}_i^T \underline{A}^{\delta_i}$$

$$\underline{D}^* = \begin{bmatrix} \underline{d}_1^{*T} \\ \vdots \\ \underline{d}_q^{*T} \end{bmatrix}, \quad \underline{d}_i^{*T} = \begin{cases} \underline{d}_i^T & \text{for } \underline{d}_i^T \neq \underline{0}^T \text{ (means } \delta_i = 0) \\ \underline{c}_i^T \underline{A}^{\delta_i-1} \underline{B} & \text{for } \underline{d}_i^T = \underline{0}^T \end{cases}$$

• Step 3:

calculate the matrix \underline{D}^{*-1} by inverting \underline{D}^*

The system is only decouplable if
 $\det(\underline{D}^*) \neq 0$

• Step 4:

determination of the coefficients

$$l_1, \alpha_{10}, \dots, \alpha_{1, \delta_1-1} \quad \text{with } l_1 = \alpha_{10}$$

$$\vdots$$

$$l_q, \alpha_{q0}, \dots, \alpha_{q, \delta_q-1} \quad \text{with } l_q = \alpha_{q0}$$

due to the desired dynamics for the overall system.

• Step 5:

setup the matrices \underline{L} & \underline{M}^*

$$\underline{L} = \begin{bmatrix} l_1 & \dots & 0 \\ \vdots & & \vdots \\ 0 & & l_q \end{bmatrix}$$

$$\underline{M}^* = \begin{bmatrix} \underline{M}_1^{*T} \\ \vdots \\ \underline{M}_q^{*T} \end{bmatrix}, \quad \underline{M}_i^{*T} = \begin{cases} \underline{0}^T & \text{for } \delta_i = 0 \\ \sum_{k=0}^{\delta_i-1} \alpha_{ik} \underline{c}_i^T \underline{A}^k & \text{for } \delta_i > 0 \end{cases}$$

• Step 6:

Insert the calculated matrices into the control law:

$$\underline{u}(t) = \underline{D}^{*-1} \left(-\underline{c}^{*T} \underline{x}(t) + \underline{L} \underline{w}(t) - \underline{M}^* \underline{x}(t) \right)$$

Verbal Explanation:

Every output variable y_i is differentiated as long as one input variable will act in a direct manner on the appropriate derivative.

The resulting system of equations must be solved according to the occurring input variables u_i

The highest derivatives $y^{(\delta_i)}$ are replaced by the desired dynamics for the overall system.

Example:

for the following system, a decoupling controller is to design

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 2 & 4 \\ 1 & 2 & 1 \end{bmatrix} x(t) + \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 2 & 3 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} x(t) + \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} u(t)$$

$$u(t) = \begin{bmatrix} u_1(t) \\ u_2(t) \end{bmatrix}; \quad y(t) = \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix}$$

1. Differential order:

1. Subsystem: $\underline{d}_1^T = \underline{0}^T; \quad \underline{c}_1^T \underline{B} = \underline{0}^T;$

$$\underline{c}_1^T \underline{A} \underline{B} = \begin{bmatrix} 0 & 2 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 2 & 3 \end{bmatrix}$$

$$\Rightarrow \underline{c}_1^T \underline{A} \underline{B} = \begin{bmatrix} 2 & 0 \end{bmatrix} \neq \underline{0}^T$$

$$\leadsto \delta_1 = 2$$

2. Subsystem: $\underline{d}_2^T = \underline{0}^T; \quad \underline{c}_2^T \underline{B} = \begin{bmatrix} 2 & 3 \end{bmatrix} \neq \underline{0}^T$

$$\leadsto \delta_2 = 1$$

$$2. \quad \underline{c}^* = \begin{bmatrix} \underline{c}_1^T \underline{A}^2 \\ \underline{c}_2^T \underline{A} \end{bmatrix} = \begin{bmatrix} 0 & 4 & 8 \\ 1 & 2 & 1 \end{bmatrix}$$

$$\underline{D}^* = \begin{bmatrix} \underline{c}_1^T \underline{A} \underline{B} \\ \underline{c}_2^T \underline{B} \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 2 & 3 \end{bmatrix}$$

$$3. D^{*-1} = \frac{\begin{bmatrix} 3 & 0 \\ -2 & 2 \end{bmatrix}}{6} \quad \text{with } \det(D^*) = 6$$

$$= \begin{bmatrix} \frac{1}{2} & 0 \\ -\frac{1}{3} & \frac{1}{3} \end{bmatrix}$$

The system is decouplable

4. Determination of the co-efficients

↓ random

$$l_1 = 1 \rightarrow \alpha_{10} = l_1 = 1$$

↓ random

$$\alpha_{11} = 2$$

↓ random

$$l_2 = 1 \rightarrow \alpha_{20} = l_2 = 1$$

$$5. \underline{L} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\underline{M}^* = \begin{bmatrix} \alpha_{10} \underline{c}^T + \alpha_{11} \underline{c}^T \underline{A} \\ \alpha_{20} \underline{c}^T \end{bmatrix} = \begin{bmatrix} 2\alpha_{10} & 2\alpha_{11} & 0 \\ 0 & 0 & \alpha_{20} \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 4 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

6. Control law:

$$\underline{u}(t) = \underline{D}^{*-1} \left(-\underline{C}^* \underline{x}(t) + \underline{L} \underline{w}(t) - \underline{M}^* \underline{x}(t) \right)$$

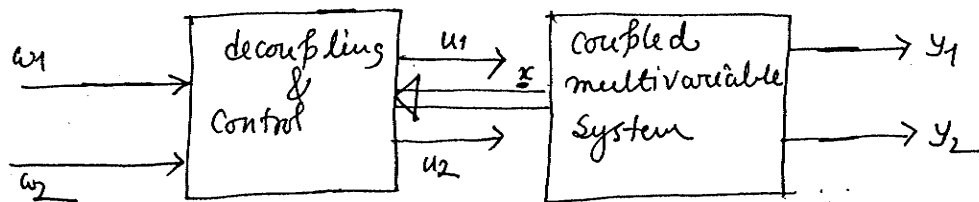
$$= \underline{D}^{*-1} \left(- \begin{bmatrix} 0 & 4 & 8 \\ 1 & 2 & 1 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \underline{w}(t) - \begin{bmatrix} 2 & 4 & 0 \\ 0 & 0 & 1 \end{bmatrix} \underline{x}(t) \right)$$

$$= \underline{D}^{*-1} \left(- \begin{bmatrix} 2 & 8 & 8 \\ 1 & 2 & 2 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \underline{w}(t) \right)$$

$$= \underline{D}^{*-1} \begin{bmatrix} w_1(t) - 2x_1(t) - 8x_2(t) - 8x_3(t) \\ w_2(t) - x_1(t) - 2x_2(t) - 2x_3(t) \end{bmatrix}$$

$$u_1(t) = \frac{1}{2} w_1(t) - x_1(t) - 4x_2(t) - 4x_3(t)$$

$$u_2(t) = \frac{1}{3} \left(-w_1(t) + w_2(t) + x_1(t) + 6x_2(t) + \frac{6x_3(t)}{3} \right)$$



Test:

$$y_1(t) = 2x_1(t)$$

$$\dot{y}_1(t) = 2\dot{x}_1(t) = 2x_2(t)$$

$$\dot{y}_1(t) = 2\dot{x}_2(t) = 4x_2(t) + 8x_3(t) + 2u_1(t)$$

$$= 4x_2(t) + 8x_3(t) + \omega_1(t) - 2x_1(t) - 8x_2(t) - 8x_3(t)$$

$$= \omega_1(t) - \underbrace{2x_1(t)}_{y_1} - \underbrace{4x_2(t)}_{2y_1}$$

$$\begin{aligned} \dot{y}_1(t) + \alpha_{11} \dot{y}_1(t) + \alpha_{10} y_1(t) &= \lambda_1 \omega_1(t) \\ \lambda_1 &= \frac{-2 \pm \sqrt{4-4}}{2} \\ &= -1, -1 \\ &\text{aperiodic limit} \\ &\text{double pole at } -1 \end{aligned}$$

$$y_2(t) = x_3(t)$$

$$\begin{aligned} \dot{y}_2(t) = \dot{x}_3(t) &= x_1(t) + 2x_2(t) + x_3(t) + 2u_1(t) + 3u_2(t) \\ &= x_1(t) + 2x_2(t) + x_3(t) + 2\left(\frac{1}{2}\omega_1(t) - x_1(t) - 4x_2(t) - 4x_3(t)\right) \\ &\quad + 3 \cdot \frac{1}{3}(-\omega_1(t) + \omega_2(t) + x_1(t) + 6x_2(t) + 6x_3(t)) \end{aligned}$$

$$\dot{y}_2(t) = \omega_2(t) - \underbrace{x_3(t)}_{y_2}$$

$$\dot{y}_2(t) + \alpha_{20} y_2(t) = \lambda_2 \omega_2(t)$$

Control of Non-linear System:

linear system \therefore

$$\begin{aligned} \dot{\underline{x}}(t) &= \underline{A} \underline{x}(t) + \underline{B} \underline{u}(t) \\ \underline{y}(t) &= \underline{C} \underline{x}(t) + \underline{D} \underline{u}(t) \end{aligned}$$

non-linearities:

$$\begin{aligned} \underline{A} &\rightarrow \underline{A}(\underline{x}(t)) & \underline{C} &\rightarrow \underline{C}(\underline{x}(t)) \\ \underline{B} &\rightarrow \underline{B}(\underline{x}(t)) & \underline{D} &\rightarrow \underline{D}(\underline{x}(t)) \end{aligned}$$

non linear system:

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

$$y(t) = C(x) + D(x) \cdot u(t)$$

Here is the special case that the control signals in $u(t)$ are still linear inputs to the system equations.

Example:

$$\dot{x}_1(t) = -x_1(t) \cdot x_2(t) + x_1(t) \cdot u_1(t) + u_2(t)$$

$$\dot{x}_2(t) = -\sqrt{x_1(t)} + \sqrt{x_1(t)} \cdot u_1(t) + (x_1(t) + x_2(t)) u_2(t)$$

$$y_1(t) = x_1(t) \cdot x_2(t)$$

$$y_2(t) = x_1(t) \cdot u_1(t)$$

→ non linear state space description

$$\dot{x}(t) = \begin{bmatrix} -x_1(t) \cdot x_2(t) \\ -\sqrt{x_1(t)} \end{bmatrix} + \begin{bmatrix} x_1(t) & 1 \\ \sqrt{x_1(t)} & x_1(t) + x_2(t) \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} x_1(t) \cdot x_2(t) \\ 0 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ x_1(t) & 0 \end{bmatrix} u(t)$$

shorter, more clear explanation: