

## Chapter 1. The State Space Modelling Approach

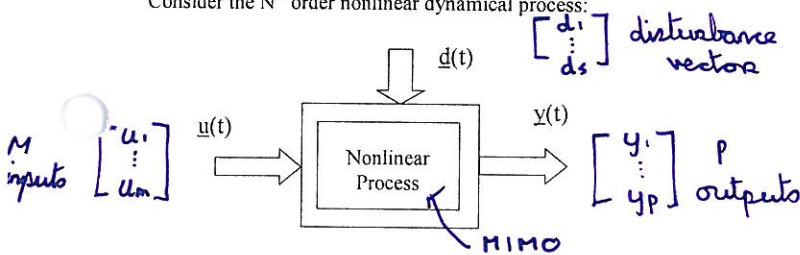
## 1.1 The State Space Model

Transformation of an  $N^{\text{th}}$  order multi-input-multi-output system to  $N$  first order differential equations: **TIME DOMAIN MODEL**

## Benefits

- Robust method of simulating high order differential equations
  - Easy analysis of dynamics
  - Allows for use of model reduction methods
  - Can easily apply advanced control  $\rightarrow$  **OPTIMAL CONTROL**
  - Used for estimation  $\rightarrow$  **KALMAN FILTERING**
- LINEAR QUADRATIC REGULATOR**  
**MULTIVARIATE**

Consider the  $N^{\text{th}}$  order nonlinear dynamical process:



In general this process could be represented by a model consisting of:  $p$  differential equations (1 each o/p) possibly high order + non linear

THIS MEANS PROCESS IS  $n^{\text{th}}$  order

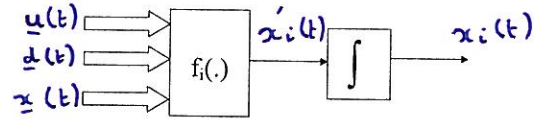
It is possible to transform this set of coupled nonlinear differential equations to yield a set of  $N$  first order differential equations with states,  $\{x_1(t), x_2(t), \dots, x_N(t)\}$ .

STATE VECTOR  $\underline{x}(t) = \begin{bmatrix} x_1(t) \\ \vdots \\ x_N(t) \end{bmatrix}$  REPRESENTS THE DYNAMIC STATE OF PROCESS

The differential equation describing the dynamics of the  $i^{\text{th}}$  state  $x_i(t)$  can be written as:

$$\frac{d}{dt} x_i(t) = f_i(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \quad \text{general}$$

Which could be represented by the following simulation diagram:



The  $N$  first order differential equations could be written as:

$$\frac{d}{dt} \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_N(t) \end{bmatrix} = \underline{\dot{x}}(t) = \begin{bmatrix} f_1(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \\ f_2(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \\ \vdots \\ f_N(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \end{bmatrix} \quad \left. \vphantom{\frac{d}{dt}} \right\} N \text{ equations}$$

$\Rightarrow$  This represents the internal dynamics of the process

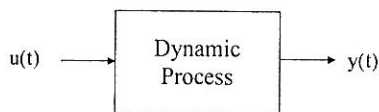
The output  $\underline{y}(t)$  is then generated by:

$$\underline{y}(t) = \begin{bmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_p(t) \end{bmatrix} = \begin{bmatrix} h_1(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \\ h_2(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \\ \vdots \\ h_p(\underline{x}(t), \underline{u}(t), \underline{d}(t)) \end{bmatrix} \quad \left. \vphantom{\underline{y}(t)} \right\} \begin{array}{l} p \text{ outputs} \\ \therefore p \text{ eqs} \end{array}$$

THIS IS STATIC

## 1.1.1 Some Example State-Space Systems:

## i) Third Order Linear Process



Modelled by the following differential equation:

$$\frac{d^3}{dt^3} y(t) + 5 \frac{d^2}{dt^2} y(t) + 3 \frac{d}{dt} y(t) + y(t) = u(t)$$

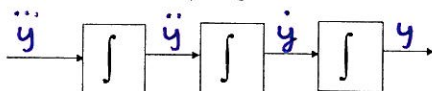
Which could of course be represented by the transfer function model:

$$U(s) \rightarrow \frac{1}{s^3 + 5s^2 + 3s + 1} \rightarrow Y(s)$$

Rearrange to yield an expression for the highest derivative:

$$\ddot{\ddot{y}} = u(t) - y - 3\dot{y} - 5\ddot{y}$$

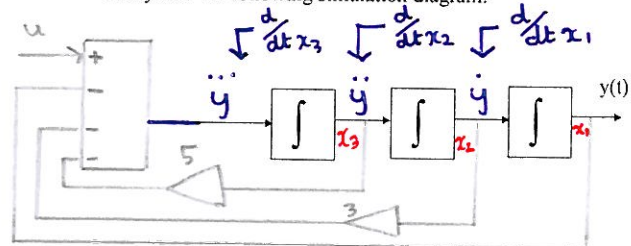
Form lower derivatives then by integration:



But we know that:

$$\ddot{\ddot{y}}(t) = u(t) - 5\ddot{y}(t) - 3\dot{y}(t) - y(t)$$

This yields the following simulation diagram:



Assign the state variables as the outputs of integrators:

$$\underline{x}_1(t), \underline{x}_2(t), \underline{x}_3(t)$$

We can now specify the state vector for the process as:

$$\underline{x}(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix}$$

Now this could be written in matrix form as:

$$\frac{d}{dt} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & -3 & -5 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u(t)$$

$$\underline{y}(t) = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix}$$

ASIDE

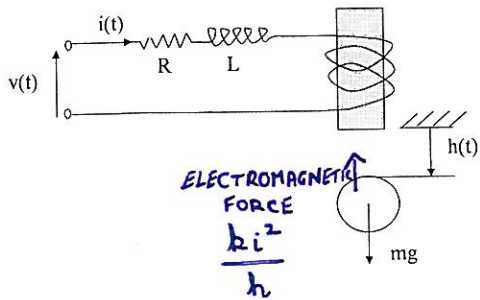
$$\frac{d}{dt} x_1 = \dot{x}_1 = x_2$$

$$\frac{d}{dt} x_2 = \dot{x}_2 = x_3$$

$$\frac{d}{dt} x_3 = \dot{x}_3 = u - y - 3\dot{y} - 5\ddot{y} = u - x_1 - 3x_2 - 5x_3$$

### ii) Third Order Nonlinear Process

Consider the following magnetic suspension system:



Can be modelled by:

coupled  
non linear  
3rd order  
model

$m \frac{d^2 h(t)}{dt^2} = mg - \frac{K i^2(t)}{h(t)}$  2nd order (FORCES NEWTON'S LAWS)  
 $L \frac{di(t)}{dt} = v(t) - Ri(t)$  1st order (electrical)

Rewritten as:

$$\frac{d^2 h(t)}{dt^2} = g - \frac{Ki^2(t)}{mh(t)}$$

$$\frac{di(t)}{dt} = \frac{1}{L}(v(t) - Ri(t))$$

## 1.2 Derivation of the Linear State Space Model - by Linearisation

Consider first the linearisation of a multivariate function:

$$z = g(\underline{w}) \quad \text{where} \quad \underline{w} = \begin{bmatrix} w_1 \\ \vdots \\ w_N \end{bmatrix} = g(w_1, w_2, \dots, w_N)$$

about the operating point vector:

$$\underline{w}_0 = \begin{bmatrix} w'_1 \\ \vdots \\ w'_N \end{bmatrix}$$

Use the first order Taylor's series approximation :

$$z \approx g(w_0) + \left. \frac{\partial g}{\partial w_1} \right|_{w=w_0} (w_1 - w'_1) + \left. \frac{\partial g}{\partial w_2} \right|_{w=w_0} (w_2 - w'_2) + \dots + \left. \frac{\partial g}{\partial w_N} \right|_{w=w_0} (w_N - w'_N)$$

which could be written as:

could be written as:

$$\Delta \underline{z} = \begin{bmatrix} \frac{\partial g}{\partial w_1} & \frac{\partial g}{\partial w_2} & \dots & \frac{\partial g}{\partial w_N} \end{bmatrix}_{w_0} \left( \underline{w} - \underline{w}_0 \right)$$

$\Delta \underline{w}$

[

$w_1 - w'_1$

$w_2 - w'_2$

$\vdots$

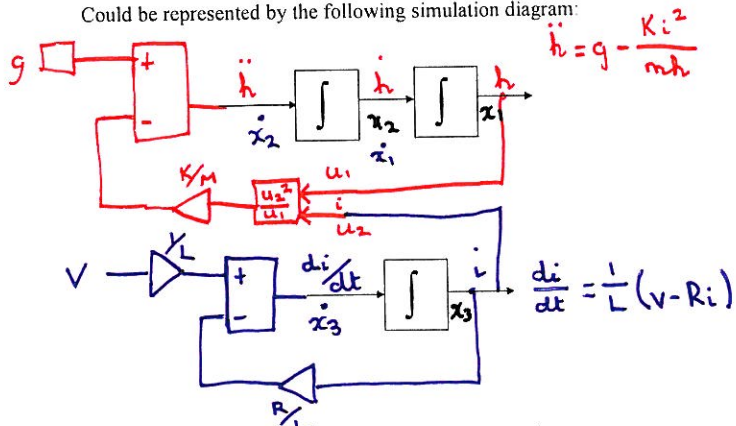
$w_N - w'_N$ ]

Now for simplicity first consider a linearisation of the  $i^{\text{th}}$  state equation:

$$\dot{x} = \frac{dx_i(t)}{dt} = f_i(x(t))$$

let  $z = \frac{dx_i}{dt}$   $\therefore z_0 = \frac{dx_i}{dt}$  at operating point

Could be represented by the following simulation diagram:



Note that three integrators are required – system is 3<sup>rd</sup> order

Now can arbitrarily assign the three states  $\{x_1(t), x_2(t), x_3(t)\}$

And define the state vector:

$$\underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

The state-equations are now:

$$\begin{aligned} \dot{x}_1 &= x_2(t) && \text{or} \\ \dot{x}_2 &= g - \frac{Kx_3^2(t)}{mx_1(t)} && \frac{d}{dt} \underline{x} = f(\underline{x}, u) \\ \dot{x}_3 &= \frac{1}{L}(u(t) - Rx_3(t)) && = \begin{bmatrix} f_1(\underline{x}, u) \\ f_2(\underline{x}, u) \\ f_3(\underline{x}, u) \end{bmatrix} \\ n: &&& \\ y(t) &= x_1(t) && \end{aligned}$$

with output equation:

↑  
the gap  $h(t)$

Could be linearised to yield:  $y = f_i(x)$

$$\Delta z = \left[ \frac{\partial f}{\partial x_1} \bigg|_{x_0} \quad \frac{\partial f}{\partial x_2} \bigg|_{x_0} \quad \dots \quad \frac{\partial f}{\partial x_N} \bigg|_{x_0} \right] \Delta \underline{x}$$

But:  $\ddot{x}_0 = \frac{dx_i}{dt} \bigg|_0$   
 $\ddot{x} = \frac{dx_i}{dt}$   
 Hence the linearised equation becomes.

Hence the linearised equation becomes.

$$\mathbf{z} - \mathbf{z}_0 = \left[ \frac{\partial f_1}{\partial x_1} \bigg|_{x_0} \quad \frac{\partial f_1}{\partial x_2} \bigg|_{x_0} \quad \dots \quad \frac{\partial f_1}{\partial x_N} \bigg|_{x_0} \right] \Delta \mathbf{x}(t) = \frac{d\mathbf{x}_1}{dt} = \frac{d}{dt} \Delta \mathbf{x}_1$$

LINEAR EQUATION IN STATE VECTOR

Now let us expand to include input and disturbance vector:

Now let us expand to include input and disturbance vector:

$$\frac{dx_i(t)}{dt} = f_i(\underline{x}(t), \underline{u}(t), \underline{d}(t))$$

Which will have the linearisation about the operating point:

$$\frac{d}{dt} \Delta x_i(t) = \left[ \frac{\partial f_i}{\partial x_1} \bigg|_{op} \quad \frac{\partial f_i}{\partial x_2} \bigg|_{op} \quad \dots \quad \frac{\partial f_i}{\partial x_n} \bigg|_{op} \right] \Delta x(t) + \left[ \frac{\partial f_i}{\partial u_1} \bigg|_{op} \quad \frac{\partial f_i}{\partial u_2} \bigg|_{op} \quad \dots \quad \frac{\partial f_i}{\partial u_m} \bigg|_{op} \right] \Delta u(t) + \left[ \frac{\partial f_i}{\partial d_1} \bigg|_{op} \quad \frac{\partial f_i}{\partial d_2} \bigg|_{op} \quad \dots \quad \frac{\partial f_i}{\partial d_l} \bigg|_{op} \right] \Delta d(t)$$

which could further be written as:

$$\frac{d}{dt} \Delta x_i(t) = \underbrace{[a_{i1} \quad a_{i2} \quad \cdots \quad a_{iN}]}_{\text{state matrix}} \Delta \underline{x}(t) + \underbrace{[b_{i1} \quad b_{i2} \quad \cdots \quad b_{im}]}_{\text{input matrix}} \Delta \underline{u}(t) + \underbrace{[e_{i1} \quad e_{i2} \quad \cdots \quad e_e]}_{\text{disturbance matrix}} \Delta \underline{d}(t)$$

CONSTANTS CALCULATED  
AT OPERATING POINT



This could of course be repeated for all N state equations:

$$N \left\{ \frac{d}{dt} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \vdots \\ \Delta x_N \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ a_{N1} & a_{N2} & \dots & a_{NN} \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \vdots \\ \Delta x_N \end{bmatrix} + \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1m} \\ b_{21} & b_{22} & \dots & b_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ b_{N1} & b_{N2} & \dots & b_{Nm} \end{bmatrix} \begin{bmatrix} \Delta u_1 \\ \Delta u_2 \\ \vdots \\ \Delta u_m \end{bmatrix} + \begin{bmatrix} e_{11} & e_{12} & \dots & e_{1s} \\ e_{21} & e_{22} & \dots & e_{2s} \\ \vdots & \vdots & \ddots & \vdots \\ e_{N1} & e_{N2} & \dots & e_{Ns} \end{bmatrix} \begin{bmatrix} \Delta d_1 \\ \Delta d_2 \\ \vdots \\ \Delta d_s \end{bmatrix} \right.$$

Or in more compact form as:

$$\frac{d}{dt} \Delta \underline{x} = A \Delta \underline{x} + B \Delta \underline{u} + E \Delta \underline{d} \quad \text{INTERNAL DYNAMICS}$$

Now consider the static output equation for the i<sup>th</sup> output:

$$y_i(t) = h_i(\underline{x}(t), \underline{u}(t), \underline{d}(t))$$

The linearisation about the operating point  $\{\underline{x}_0, \underline{u}_0, \underline{d}_0\}$  is:

$$\Delta y_i(t) = \left[ \frac{\partial h_i}{\partial \underline{x}_1} \bigg|_{op} \quad \frac{\partial h_i}{\partial \underline{x}_2} \bigg|_{op} \quad \dots \quad \frac{\partial h_i}{\partial \underline{x}_N} \bigg|_{op} \right] \Delta \underline{x}(t) + \left[ \frac{\partial h_i}{\partial \underline{u}_1} \bigg|_{op} \quad \frac{\partial h_i}{\partial \underline{u}_2} \bigg|_{op} \quad \dots \quad \frac{\partial h_i}{\partial \underline{u}_m} \bigg|_{op} \right] \Delta \underline{u}(t) + \left[ \frac{\partial h_i}{\partial \underline{d}_1} \bigg|_{op} \quad \frac{\partial h_i}{\partial \underline{d}_2} \bigg|_{op} \quad \dots \quad \frac{\partial h_i}{\partial \underline{d}_s} \bigg|_{op} \right] \Delta \underline{d}(t)$$

### 1.2.1 Linearisation Examples:

#### i) The magnetic suspension system

$$m \frac{d^2 h(t)}{dt^2} = mg - \frac{Ki^2(t)}{h(t)}$$

$$L \frac{di(t)}{dt} = v(t) - Ri(t)$$

Find a linear model about the desired operating airgap  $h=0.01m$

The process parameters are:

$$L=10mH \quad M=0.05 \text{ Kg} \quad g=10ms^{-2} \quad R=1 \text{ ohm} \quad K=0.01Nm/A^2$$

First find the operating point, **EQUILIBRIUM POINT**

From the force equation:

$$m \frac{d^2 h(t)}{dt^2} = mg - \frac{Ki^2(t)}{h(t)} \quad \begin{matrix} 0.01 \\ \frac{Ki_0^2}{h_0} = mg - 10 \\ i_0 = 0.707 \text{ A} \end{matrix}$$

From the electrical equation:

$$L \frac{di(t)}{dt} = v(t) - Ri(t) \quad \begin{matrix} 10 \\ 0.707 \text{ V} = V_0 \end{matrix}$$

The operating point vector will now be defined as:

$$\underline{x}_0 = \begin{bmatrix} h_0 \\ \dot{h}_0 \\ i_0 \end{bmatrix} = \begin{bmatrix} 0.01 \\ 0 \\ 0.707 \end{bmatrix} \quad \text{AND} \quad \underline{u}_0 = V_0 = 0.707$$

**OPERATING POINT STATE VECTOR**

$$\frac{\partial h_i}{\partial x_i} \bigg|_{op} \quad \frac{\partial h_i}{\partial d_i} \bigg|_{op}$$

which could be written as:

$$\Delta y_i(t) = [c_{i1} \ c_{i2} \ \dots \ c_{iN}] \Delta \underline{x}(t) + [d_{i1} \ d_{i2} \ \dots \ d_{im}] \Delta \underline{u}(t) + [f_{i1} \ f_{i2} \ \dots \ f_{is}] \Delta \underline{d}(t)$$

This could be repeated for all the P outputs:

$$P \left\{ \begin{bmatrix} \Delta y_1 \\ \Delta y_2 \\ \vdots \\ \Delta y_P \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1N} \\ c_{21} & c_{22} & \dots & c_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ c_{P1} & c_{P2} & \dots & c_{PN} \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \vdots \\ \Delta x_N \end{bmatrix} + \begin{bmatrix} d_{11} & d_{12} & \dots & d_{1m} \\ d_{21} & d_{22} & \dots & d_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ d_{P1} & d_{P2} & \dots & d_{Pm} \end{bmatrix} \begin{bmatrix} \Delta u_1 \\ \Delta u_2 \\ \vdots \\ \Delta u_m \end{bmatrix} + \begin{bmatrix} f_{11} & f_{12} & \dots & f_{1s} \\ f_{21} & f_{22} & \dots & f_{2s} \\ \vdots & \vdots & \ddots & \vdots \\ f_{P1} & f_{P2} & \dots & f_{Ps} \end{bmatrix} \begin{bmatrix} \Delta d_1 \\ \Delta d_2 \\ \vdots \\ \Delta d_s \end{bmatrix} \right\} M$$

which again could be written in more compact form as:

$$\Delta \underline{y} = C \Delta \underline{x} + D \Delta \underline{u} + F \Delta \underline{d}$$

This yields the following linear state-space model for a process about a particular operating point:

$$\begin{aligned} \frac{d}{dt} \underline{x}(t) &= A \underline{x}(t) + B \underline{u}(t) + E \underline{d}(t) \quad \text{DESCRIBES INTERNAL DYNAMICS} \\ \underline{y}(t) &= C \underline{x}(t) + D \underline{u}(t) + F \underline{d}(t) \quad \text{MAPS STATES TO OUTPUTS} \end{aligned}$$

drop the  $\Delta$  notation

Now linearise about this operating point vector:

The state equations can be written as:

$$\begin{aligned} \dot{x}_1 &= x_2(t) = f_1(x_2) \\ \dot{x}_2 &= g - \frac{Kx_3^2(t)}{mx_1(t)} = f_2(x_1, x_3) \\ \dot{x}_3 &= \frac{1}{L}(u(t) - Rx_3(t)) = f_3(x_3, u) \end{aligned}$$

Now define deviations of the states from their operating point values as:

$$\Delta \underline{x} = \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \end{bmatrix} = \underline{x} - \underline{x}_0 \quad \Delta u = u - u_0$$

Hence the linearised model is:

$$\begin{aligned} \frac{d}{dt} \Delta x_1(t) &= \frac{\partial f_1}{\partial x_2} \bigg|_{op} \Delta x_2(t) \\ \frac{d}{dt} \Delta x_2(t) &= \frac{\partial f_2}{\partial x_1} \bigg|_{op} \Delta x_1(t) + \frac{\partial f_2}{\partial x_3} \bigg|_{op} \Delta x_3(t) \\ \frac{d}{dt} \Delta x_3(t) &= \frac{\partial f_3}{\partial x_3} \bigg|_{op} \Delta x_3(t) + \frac{\partial f_3}{\partial u} \bigg|_{op} \Delta u(t) \end{aligned}$$

But we know that:  $op \Rightarrow x_1 = 0.01 \quad x_2 = 0$

$$\begin{aligned} f_1(x_2) &= x_2 \quad \frac{\partial f_1}{\partial x_2} = 1 \text{ everywhere} \\ f_2(x_1, x_3) &= g - \frac{Kx_3^2}{Mx_1} \quad \frac{\partial f_2}{\partial x_1} = \frac{Kx_3^2}{Mx_1^2} = \frac{K(0.707)^2}{M(0.01)^2} = 999.69 \\ f_3(x_3, u) &= \frac{u}{L} - \frac{R}{L}x_3 \quad \frac{\partial f_3}{\partial x_3} = -\frac{R}{L} = -100 \text{ everywhere!} \end{aligned}$$

This yields the following linearised model about the operating point:

$$\begin{aligned}\frac{d}{dt} \Delta x_1(t) &= \Delta x_2(t) \\ \frac{d}{dt} \Delta x_2(t) &= 999.69 \Delta x_1(t) - 28.28 \Delta x_3(t) \\ \frac{d}{dt} \Delta x_3(t) &= -100 \Delta x_3(t) + 100 \Delta u(t)\end{aligned}$$

And of course:

$$\Delta y(t) = h(t) - h_0 = \Delta x_1(t)$$

This then could be written in matrix form as:

$$\begin{aligned}\frac{d}{dt} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \end{bmatrix} &= \begin{bmatrix} 0 & 1 & 0 \\ 999.69 & 0 & -28.28 \\ 0 & 0 & -100 \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 100 \end{bmatrix} \Delta u(t) \\ \Delta y(t) &= \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \end{bmatrix}\end{aligned}$$

## ii) Permanent magnet DC motor

Could be modelled by the following coupled equations:

$$\begin{aligned}\frac{di}{dt} &= \frac{1}{L} (v(t) - Ri(t) - K_m \omega(t)) \\ \frac{d\omega}{dt} &= \frac{1}{J} (K_M i(t) - B\omega(t) - T_L(t))\end{aligned}$$

*2 states, 1 input, 1 dist → T<sub>L</sub>*

Now from the electrical equation:

$$\frac{di}{dt} = \frac{1}{L} (v(t) - Ri(t) - K_m \omega(t)) \Rightarrow V_0 - Ri_0 - K_m \omega_0 = 0$$

$$V_0 = (K_m + \frac{RB}{K_m}) \omega_0 + \frac{RK_f}{K_m} \omega_0^2$$

Assign the states:  $\underline{x}(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} i(t) \\ \omega(t) \end{bmatrix}$

and input and output:  $\underline{u}(t) = v(t)$  and  $\underline{y}(t) = v_T(t) = K_T \omega(t)$

**INPUT**                      **OUTPUT**

hence:

**LINEAR**  $\dot{x}_1(t) = \frac{1}{L} (u(t) - Rx_1(t) - K_m x_2(t)) = f_1(x_1, x_2, u)$

**NON LINEAR**  $\dot{x}_2(t) = \frac{1}{J} (K_M x_1(t) - Bx_2(t) - K_f x_2^2(t)) = f_2(x_1, x_2)$

**LINEAR**  $y(t) = K_T x_2(t) = h(x_2)$

The linearised model about the operating point is:

$$\begin{aligned}\frac{d}{dt} \Delta x_1(t) &= \left. \frac{\partial f_1}{\partial x_1} \right|_{op} \Delta x_1(t) + \left. \frac{\partial f_1}{\partial x_2} \right|_{op} \Delta x_2(t) + \left. \frac{\partial f_1}{\partial u} \right|_{op} \Delta u(t) \\ \frac{d}{dt} \Delta x_2(t) &= \left. \frac{\partial f_2}{\partial x_1} \right|_{op} \Delta x_1(t) + \left. \frac{\partial f_2}{\partial x_2} \right|_{op} \Delta x_2(t) \\ \Delta y(t) &= \left. \frac{\partial h}{\partial x_2} \right|_{op} \Delta x_2(t)\end{aligned}$$

Obviously the electrical equation is linear:

$$\dot{x}_1(t) = \frac{1}{L} (u(t) - Rx_1(t) - K_m x_2(t)) = f_1(x_1, x_2, u)$$

$$\frac{\partial f_1}{\partial x_1} = -\frac{R}{L} \quad \frac{\partial f_1}{\partial x_2} = -\frac{K_m}{L} \quad \frac{\partial f_1}{\partial u} = \frac{1}{L} \text{ everywhere!}$$

As you can see these are linear differential equations:

$$\frac{d}{dt} \begin{bmatrix} i \\ \omega \end{bmatrix} = \begin{bmatrix} -\frac{R}{L} & -\frac{K_m}{L} \\ \frac{K_M}{J} & -\frac{B}{J} \end{bmatrix} \begin{bmatrix} i \\ \omega \end{bmatrix} + \begin{bmatrix} \frac{1}{L} \\ 0 \end{bmatrix} v + \begin{bmatrix} 0 \\ -\frac{1}{J} \end{bmatrix} T_L$$

$2 \times 2$                        $2 \times 1$                        $2 \times 1$

$$\dot{\underline{x}} = A\underline{x} + B\underline{u} + E\underline{d}$$

If a tachometer of gain  $K_T$  V/rads<sup>-1</sup> is used to measure the speed, then the output equation could be written as:

$$v_T(t) = \begin{bmatrix} 0 & K_T \end{bmatrix} \begin{bmatrix} i \\ \omega \end{bmatrix}$$

Consider now that the motor is driving a nonlinear fan load:

$$T_L(t) = K_f \omega^2(t)$$

The process then would be modelled as:

$$\begin{aligned}\frac{di}{dt} &= \frac{1}{L} (v(t) - Ri(t) - K_m \omega(t)) \\ \frac{d\omega}{dt} &= \frac{1}{J} (K_M i(t) - B\omega(t) - K_f \omega^2(t))\end{aligned}$$

**THIS IS NOW NON LINEAR** → **NO LONGER AN INDEPENDENT INPUT** (T<sub>L</sub>)

Generate a linear state-space model which describes the dynamics of this process close to the operating speed:

First find the operating point:  $i_0, \omega_0, V_0$

$$\begin{aligned}\text{EQUILIB} \quad \frac{d}{dt} &= \frac{1}{J} (K_M i(t) - B\omega(t) - K_f \omega^2(t)) = 0 \\ K_M i_0 &= B\omega_0 + K_f \omega_0^2 \\ i_0 &= \frac{B\omega_0 + K_f \omega_0^2}{K_M}\end{aligned}$$

*OP CURVE*

Now concentrating on the mechanical equation:

$$\dot{x}_2(t) = \frac{1}{J} (K_M x_1(t) - Bx_2(t) - K_f x_2^2(t)) = f_2(x_1, x_2)$$

$$\frac{\partial f_2}{\partial x_1} = \frac{K_M}{J}$$

$$\frac{\partial f_2}{\partial x_2} = \frac{-B}{J} - \frac{2K_f x_2}{J} \Rightarrow \left. \frac{\partial f_2}{\partial x_2} \right|_{op} = \frac{-B}{J} - \frac{2K_f \omega_0}{J}$$

And simply for the output equation:

$$y(t) = K_T x_2(t) = h(x_2) \quad \frac{\partial h}{\partial x_2} = K_T$$

This yields the following linear state-space model:

$$\Delta \underline{\dot{x}} = \begin{bmatrix} i(t) - i_0 \\ \omega(t) - \omega_0 \end{bmatrix} \frac{d}{dt} \Delta \underline{x} = \begin{bmatrix} -\frac{R}{L} & -\frac{K_m}{L} \\ \frac{K_M}{J} & -\frac{B}{J} - \frac{2K_f \omega_0}{J} \end{bmatrix} \Delta \underline{x}(t) + \begin{bmatrix} \frac{1}{L} \\ 0 \end{bmatrix} \Delta u(t)$$

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

$$\Delta y = v_T(t) - K_T \omega_0 \rightarrow \text{OPERATING POINT TACHO O/P}$$

$$\Delta u = v(t) - V_0$$