# STATE VECTOR MODEL

An example of finding the state vector model of a system given the transfer function in the s-domain.

# The Problem:

Given the transfer function:

$$G(s) = \frac{100}{s^2 + 3s + 1}$$

find **A**, **b**, and **D** of the state vector model.

## **State Vector Model:**

$\dot{X}(t) = AX(t) + bu(t)$	X(t) = state vector, consisting of the output signal and its derivatives $\dot{x}(t)$ = first derivative of the state vector A = a square matrix b = a vector
	u(t) = system input signal

1) Write the transfer function as a function of output divided by the input:

$$G(s) = \frac{C(s)}{U(s)} = \frac{100}{s^2 + 3s + 1}$$

2) Cross multiply:

$$s^{2}C(s)+3sC(s)+C(s)=100U(s)$$

3) Convert to the time domain (inverse Laplace transform):

$$\ddot{c}(t) + 3\dot{c}(t) + c(t) = 100u(t)$$

# 4) Pick a solution:

Let 
$$x_1(t) = c(t), x_2(t) = \dot{c}(t)$$

The solution is not unique, but we just always use this one. Note that  $x_1(t)$  and  $x_2(t)$  are elements of the state vector X(t) and that there are two elements in this case because of the 2<sup>nd</sup> order polynomial in the denominator of the transfer function. If the polynomial was 3<sup>rd</sup> order, we would include  $x_3(t) = \ddot{c}(t)$  and so on.

# **5)** Solve for $\dot{X}(t)$ :

We want to get  $\dot{x}_{(t)}$  in terms of x and u. It can be seen from step 4 that  $\dot{x}_1(t) = x_2(t)$ . It can also be seen from step 4 that  $\dot{x}_2(t) = \ddot{c}(t)$ . Solving the expression in step 3 for  $\ddot{c}(t)$  we have  $\ddot{c}(t) = 100u(t) - 3\dot{c}(t) - c(t)$ . We can express this in terms of x and u to get  $\dot{x}_2(t) = 100u(t) - 3x_2(t) - x_1(t)$ 

## 6) Back to the state vector model:

State vector model:  $\dot{X}(t) = AX(t) + bu(t)$ 

State vector model showing matrices: 
$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} u(t).$$

Carrying out the operations: 
$$\hat{x}_{1}(t) = a_{11}x_{1}(t) + a_{12}x_{2}(t) + b_{1}u(t) \hat{x}_{2}(t) = a_{21}x_{1}(t) + a_{22}x_{2}(t) + b_{2}u(t)$$

## 7) Plugging in to the state vector model for matrices A and b:

Using the results of steps 5 and 6 we can find the values for the matrices A and b.

$$\dot{x}_{1}(t) = x_{2}(t): \quad \dot{x}_{1}(t) = \underbrace{a_{11}}_{0} x_{1}(t) + \underbrace{a_{12}}_{1} x_{2}(t) + \underbrace{b_{1}}_{0} u(t)$$
$$\dot{x}_{2}(t) = 100u(t) - 3x_{2}(t) - x_{1}(t): \quad \dot{x}_{2}(t) = \underbrace{a_{21}}_{-1} x_{1}(t) + \underbrace{a_{22}}_{-3} x_{2}(t) + \underbrace{b_{2}}_{100} u(t)$$
So matrices **A** and **b** are: 
$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & -3 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} b_{1} \\ b_{2} \end{bmatrix} = \begin{bmatrix} 0 \\ 100 \end{bmatrix}$$

# 8) Finding matrix D from the output equation:

The output equation is  $c(t) = \mathbf{D}X(t)$ .

The output equation in matrix form is  $c(t) = \begin{bmatrix} d_1 & d_2 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$ .

Carrying out the multiplication we have  $c(t) = d_1x_1(t) + d_2x_2(t)$ .

We already know from step 4 that  $x_1(t) = c(t)$ .

So we can determine the matrix values  $c(t) = \underbrace{d_1 x_1(t)}_{1} + \underbrace{d_2 x_2(t)}_{0}$ 

Therefore  $\mathbf{D} = \begin{bmatrix} 1 & 0 \end{bmatrix}$ .

As it turns out, matrix  $\mathbf{D}$  is predictable. The first element is always 1 and the remaining elements are zeros. The number of elements is equal to the order of the polynomial in the denominator of the transfer function.