

Model Based Engineering

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Contents

1	State Feedback	2
2	Overview	2
3	Reachability	4
3.1	Testing for reachability with impulse functions	5
3.2	Testing for reachability with the convolution equation	6
3.3	Reachable Canonical Form	7
4	Stabilization by State Feedback	9
5	Example: Vehicle Steering	10
6	Exercises	11
6.0.1	Exercise 1	11
6.0.2	Exercise 2	12
6.0.3	Exercise 3	12
6.0.4	Exercise 4	12
6.0.5	Exercise 5	14
6.0.6	Exercise 6	14
6.0.7	Exercise 7	15
6.0.8	Exercise 8	16
6.0.9	Exercise 9	17
6.0.10	Exercise 10	18
6.0.11	Exercise 11	19

1 State Feedback

The state, $x(t)$, of a dynamical system is a collection of variables that permits prediction of the future development of a system. In this chapter we want to explore the idea of designing the dynamics of a system through feedback of the state. A typical feedback control system with state feedback is shown in figure 1

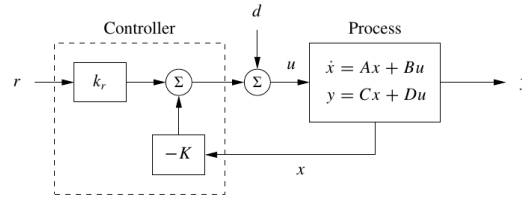


Fig. 1: A feedback control system with state feedback. The controller uses the system state x and the reference input r to command the process through its input u . We model disturbances via the additive input d .

The goal of the feedback controller is to regulate the output of the system y such that it tracks the reference input in the presence of disturbances and also uncertainty in the process dynamics.

2 Overview

We have already mentioned the general form of the state-space model is

$$\dot{x} = f(x, u, d) \quad (1)$$

$$y = h(x, u, d) \quad (2)$$

both f and h can be nonlinear functions of their arguments however in this section we will assume that we are dealing with a linear system. Our aim is to design a state feedback controller. Thus, let's consider again the system

$$\dot{x} = Ax + Bu \quad (3)$$

$$y = Cx + Du \quad (4)$$

The poles of the system are given by the eigenvalues of the matrix A which are the roots of the characteristic polynomial

$$\det(sI - A) = 0 \quad (5)$$

Feedback control is a method for shaping the dynamics of the system. This is done by modifying the eigenvalues of the system using the input u . Indeed, let's assume that u can be written as

$$u = -Kx + K_r r \quad (6)$$

with $K \in \mathbb{R}^{p \times n}$, $k_r \in \mathbb{R}^{p \times r}$ and $r \in \mathbb{R}^r$ is the reference input. We can shape the dynamics of the system via the feedback gain K . K_r is used to change the steady state level of the system. The system dynamics becomes

$$\dot{x} = Ax + B(-Kx + K_r r) \quad (7)$$

$$= (A - BK)x + BK_r r \quad (8)$$

The control objective now is to choose K such that the closed loop dynamics $A - BK$ get the desired properties.

Remark 2.1. Steady state reference gain

The steady-state reference gain K_r does not affect the stability of the system but it does affect the steady state solution.

The gain K_r is chosen such that

$$y(t) \approx r(t), \text{ as } t \rightarrow \infty \quad (9)$$

At steady state

$$\dot{x} = 0 \quad (10)$$

thus we can write

$$0 = (A - BK)x + BK_r r \quad (11)$$

$$y = Cx \quad (12)$$

Hence, we can write for the output y

$$y = -C(A - Bk)^{-1}BK_r r \quad (13)$$

If we want $y(t) \approx r(t)$, as $t \rightarrow \infty$, then K_r should be chosen as

$$K_r = -(C(A - BK)^{-1}B)^{-1} \quad (14)$$

The design of the controller becomes a two step process.

- shape the dynamics by choosing K
- set the steady-state level by choosing K_r

This way the reference gain depends on the feedback gain. Using the steady-state feedback gain K_r we can achieve zero steady-state error but this depends on the model parameters also. Integration can be used in order to remove the steady-state error. This can be done by introducing a new variable to remove the steady-state error:

We will illustrate how this is done assuming a SISO system.

$$\dot{z}(t) = y(t) - r(t) \quad (15)$$

The new state-space model becomes

$$\begin{bmatrix} \dot{x} \\ \dot{z} \end{bmatrix} = \begin{bmatrix} Ax - Bu \\ y - r \end{bmatrix} = \begin{bmatrix} Ax - Bu \\ Cx - r \end{bmatrix} \quad (16)$$

the above can be written as

$$\begin{bmatrix} A & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u + \begin{bmatrix} 0 \\ -1 \end{bmatrix} r \quad (17)$$

Thus, the new controller now becomes

$$u(t) = -Kx(t) - K_I z(t) + K_r r(t) \quad (18)$$

3 Reachability

One of the fundamental properties of a control system is what set of points in the state space can be reached through the choice of a control input. It turns out that the property of reachability is also fundamental in understanding the extent to which feedback can be used to design the dynamics of a system.

Definition 3.1. Reachability

A linear system is reachable if for any $x_0, x_f \in R^n$ there exists a $T > 0$ and $u : [0, T] \rightarrow R$ such that the corresponding solution satisfies $x(0) = x_0$ and $x(T) = x_f$.

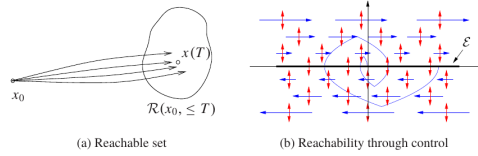


Fig. 2: The reachable set for a control system. The set $R(x_0, \leq T)$ shown in (a) is the set of points reachable from x_0 in time less than T . The phase portrait in (b) shows the dynamics for a double integrator, with the natural dynamics drawn as horizontal arrows and the control inputs drawn as vertical arrows. The set of achievable equilibrium points is the x axis. By setting the control inputs as a function of the state, it is possible to steer the system to the origin, as shown on the sample path.

The definition of reachability addresses whether it is possible to reach all points in the state space in a transient fashion. In many applications, the set of points that we are most interested in reaching is the set of equilibrium points of the system (since we can remain at those points once we get there). The set of all possible equilibria for constant controls is given by:

$$E = \{x_e : Ax_e + Bu_e = 0 \text{ for some } u_e \in R\} \quad (19)$$

To find general conditions under which a linear system is reachable, we will first give a heuristic argument based on formal calculations with impulse functions.

Remark 3.1. Equilibrium points and reachability

We note that if we can reach all points in the state space through some choice of input, then we can also reach all equilibrium points.

3.1 Testing for reachability with impulse functions

When the initial state is zero, the response of the system to an input $u(t)$ is given by

$$x(t) = \int_0^t e^{A(t-\tau)} Bu(\tau) d\tau \quad (20)$$

Let's choose the input to be an impulse function $\delta(t)$. The state then becomes

Remark 3.2. Impulse function $\delta(t)$

$$x_\delta = \int_0^t e^{A(t-\tau)} B u(\tau) d\tau = \frac{dx_s}{dt} = e^{At} B \quad (21)$$

(Note that the state changes instantaneously in response to the impulse.) We can find the response to the derivative of an impulse function by taking the derivative of the impulse response (Exercise 5.1):

$$\frac{dx_\delta}{dt} = A e^{At} B \quad (22)$$

Continuing this process and using the linearity of the system, the input

$$u(t) = \alpha_1 \delta(t) + \alpha_2 \frac{d\delta(t)}{dt} + \cdots + \alpha_n \frac{d^{n-1}\delta(t)}{dt^{n-1}} \quad (23)$$

gives the state

$$x(t) = \alpha_1 e^{At} B + \alpha_2 A e^{At} B + \cdots + \alpha_n A^{n-1} e^{At} B \quad (24)$$

On the right is a linear combination of the columns of the matrix

$$W_r = [B \quad AB \quad \cdots \quad A^{n-1}B] \quad (25)$$

To reach an arbitrary point in the state space, we thus require that there are n linear independent columns of the matrix W_r .

Remark 3.3. The reachability matrix W_r

The matrix W_r is called the reachability matrix.

3.2 Testing for reachability with the convolution equation

However, an input $u(t)$ consisting of a sum of impulse functions and their derivatives is a very violent signal. To see that an arbitrary point can be reached with smoother signals we can make use of the convolution equation. Assuming that the initial condition is zero, the state of a linear system is given by

$$x(t) = \int_0^t e^{A(t-\tau)} B u(\tau) d\tau = \int_0^t e^{A(\tau)} B u(t-\tau) d\tau \quad (26)$$

It follows from the theory of matrix functions, specifically the Cayley-Hamilton theorem (see Exercise 6.10), that

Remark 3.4. The Cayley-Hamilton theorem

$$e^{A\tau} = I\alpha_0(\tau) + A\alpha_1(\tau) + \cdots + A^{n-1}\alpha_{n-1}(\tau) \quad (27)$$

where α_i are scalar functions, and we find that

$$x(t) = B \int_0^t \alpha_0(\tau)u(t-\tau)d\tau + AB \int_0^t \alpha_1(\tau)u(t-\tau)d\tau + \cdots + A^{n-1}B \int_0^t \alpha_{n-1}(\tau)u(t-\tau)d\tau \quad (28)$$

Again we observe that the right-hand side is a linear combination of the columns of the reachability matrix W_r given by equation (6.3). This basic approach leads to the following theorem.

Theorem 3.1. *Reachability rank condition*

A linear system is reachable if and only if the reachability matrix W_r is invertible.

3.3 Reachable Canonical Form

It is often convenient to change coordinates and write the dynamics of the system in the transformed coordinates

$$z = Tx \quad (29)$$

One application of a change of coordinates is to convert a system into a canonical form in which it is easy to perform certain types of analysis.

Remark 3.5. Reachable Canonical Form

A linear state space system is in reachable canonical form, if its dynamics are given by

$$\frac{dz}{dt} = Az + Bu \quad (30)$$

$$y = Cz + du \quad (31)$$

The characteristic polynomial for a system in reachable canonical form is given

$$\lambda(s) = s^n + \alpha_1 s^{n-1} + \cdots + \alpha_{n-1} s + \alpha_n \quad (32)$$

The reachability matrix also has a relatively simple structure:

We now consider the problem of changing coordinates such that the dynamics of a system can be written in reachable canonical form. Let A, B represent the dynamics of a given system and \tilde{A}, \tilde{B} be the dynamics in reachable canonical form. Suppose that we wish to transform the original system into reachable canonical form using a coordinate transformation $z = Tx$. As shown in the last chapter, the dynamics matrix and the control matrix for the transformed system are

$$\tilde{A} = TAT^{-1}, \quad \tilde{B} = TB \quad (33)$$

The reachability matrix for the transformed system then becomes

$$\tilde{W}_r = [\tilde{B} \quad \tilde{A}\tilde{B} \quad \cdots \quad \tilde{A}^{n-1}\tilde{B}] \quad (34)$$

We can transform each element individually and thus get the reachability matrix from the transformed system

$$\tilde{W}_r = TW_r \quad (35)$$

However, we know that W_r is invertible. Therefore we can solve for the transformation T that takes the system into reachable canonical form:

$$T = \tilde{W}_r W_r^{-1} \quad (36)$$

Theorem 3.2. *Reachable canonical form*

Let A and B be the dynamics and control matrices for a reachable system. Then there exists a transformation

$$z = Tx$$

such that in the transformed coordinates the dynamics and control matrices are in the reachable canonical form and the characteristic polynomial for A is given by

$$\det(sI - A) = 0$$

One important implication of this theorem is that for any reachable system, we can assume without loss of generality that the coordinates are chosen such that the system is in reachable canonical form. This is particularly useful for proofs, as we shall see later in this chapter. However, for high-order systems, small changes in the coefficients a_i (that is the coefficients of the characteristic polynomial) can give large changes in the eigenvalues. Hence, the reachable canonical form is not always well conditioned and must be used with some care.

4 Stabilization by State Feedback

In this section, we will assume that the system to be controlled is described by a linear state model and has a single input (for simplicity). The feedback control law will be developed step by step using a single idea: the positioning of closed loop eigenvalues in desired locations.

Figure 3 has already been shown at the beginning of this chapter. It shows a diagram of a typical control system using state feedback.

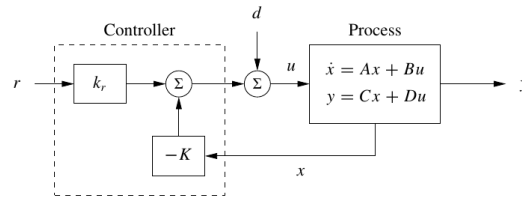


Fig. 3: A feedback control system with state feedback. The controller uses the system state x and the reference input r to command the process through its input u . We model disturbances via the additive input d .

The full system consists of the process dynamics, which we take to be linear, the controller elements K and k_r , the reference input (or command signal) r and process disturbances d . The goal of the feedback controller is to regulate the output of the system y such that it tracks the reference input in the presence of disturbances and also uncertainty in the process dynamics.

An important element of the control design is the performance specification. The simplest performance specification is that of stability: in the absence of any disturbances, we would like the equilibrium point of the system to be asymptotically stable. More sophisticated performance specifications typically involve giving desired properties of the step or frequency response of the system, such as specifying the desired rise time, overshoot and settling time of the step response. Finally, we are often concerned with the disturbance attenuation properties of the system: to what extent can we experience disturbance inputs d and still hold the output y near the desired value?

Consider a system described by the linear differential equation

$$\frac{dx}{dt} = Ax + Bu \quad y = Cx + Du \quad (37)$$

5 Example: Vehicle Steering

Let's sum up the major topics we discussed above by considering vehicle steering. A common problem in motion control is to control the trajectory of a vehicle through an actuator that causes a change in the orientation.

52

CHAPTER 2. SYSTEM MODELING

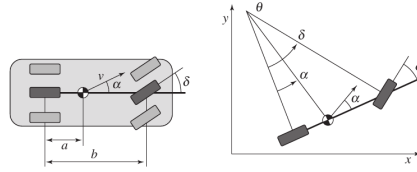


Fig. 4: Vehicle steering dynamics. The left figure shows an overhead view of a vehicle with four wheels. The wheel base is b and the center of mass at a distance a forward of the rear wheels. By approximating the motion of the front and rear pairs of wheels by a single front wheel and a single rear wheel, we obtain an abstraction called the bicycle model, shown on the right. The steering angle is δ and the velocity at the center of mass has the angle α relative the length axis of the vehicle. The position of the vehicle is given by (x, y) and the orientation (heading) by θ .

Let's consider a vehicle with two wheels. We are interested in how the velocity of the steering depends on the steering angle δ . Specifically, consider the velocity v at the center of mass, a distance w from the rear wheel, and let b be the wheel base, as shown in Figure 4. Let (x, y) be the coordinates of the center of mass, θ the heading angle and α the angle between the velocity vector v and the centerline of the vehicle. Since $b = r_w \tan(\delta)$ and $w = r_w \tan(\alpha)$, it follows that $\tan(\alpha) = (w/b) \tan(\delta)$ and we get the following relation between α and the steering angle δ :

$$\alpha(\delta) = \arctan\left(\frac{w \tan(\delta)}{b}\right) \quad (38)$$

The response in the dynamics is shown in the figures below

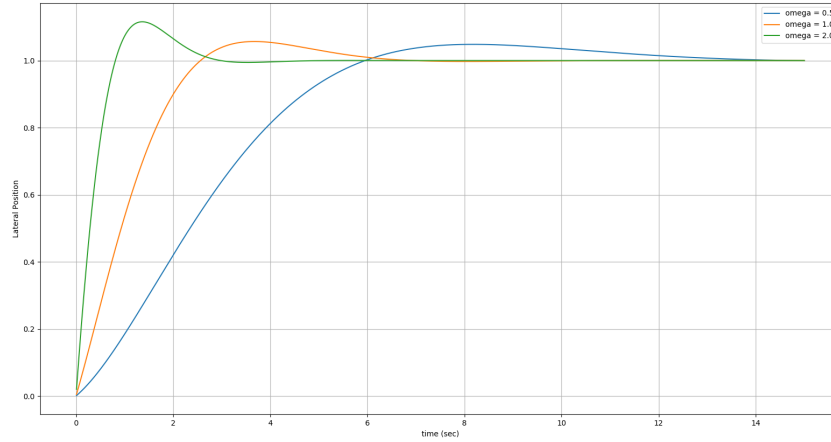


Fig. 5: Response of the vehicle lateral position with respect to ω_n .

6 Exercises

6.0.1 Exercise 1

Consider the system

$$\dot{x} = Ax + Bu \quad (39)$$

$$y = Cx \quad (40)$$

controlled by a state feedback controller:

$$u = -Kx \quad (41)$$

What are the sizes of the matrices A , B , C , and K if we have one input signal, three states and two output signals.

Answer

Since the state has three components then A must be 3×3 . As we have only one input but the result of Bu must be added to the result of Ax then B must be 3×1 . Since the output has two components then C must be 2×3 . Finally, K must be 1×3 .

6.0.2 Exercise 2

Consider again the system in the previous question. Assume now that but now the system has two input signals, three states and one output signal. What are the sizes of the involved matrices?

Answer

Using the same line of reasoning, since the state has three components then A must be 3×3 . As we have only two input but the result of Bu must be added to the result of Ax then B must be 3×2 . Since the output has two components then C must be 1×3 . Finally, K must be 2×3 .

6.0.3 Exercise 3

Consider again the system in the previous question. Assume now that but now the system has two input signals, four states and one output signal. What are the sizes of the involved matrices?

Answer

Using the same line of reasoning, since the state has three components then A must be 4×4 . As we have only two input but the result of Bu must be added to the result of Ax then B must be 4×2 . Since the output has two components then C must be 1×4 . Finally, K must be 2×4 .

6.0.4 Exercise 4

Consider the mechanical system shown in Figure 6

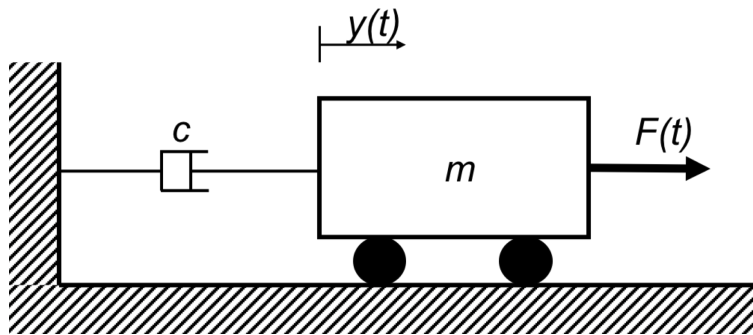


Fig. 6: Schematic of mechanical system

The system is described by the following state-space model

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & -c/m \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1/m \end{bmatrix} u \quad (42)$$

$$y = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad (43)$$

where x_1 is the position of the mass, x_2 is the velocity of the mass and u is the input signal, the force. The output y is the position. Assume that the desired closed loop characteristic polynomial is

$$p(s) = s^2 + 2\zeta\omega_n s + \omega_n^2 \quad (44)$$

and the state feedback controller is

$$u = -Kx \quad (45)$$

Determine the elements of $K = [k_{11}, k_{12}]$

Answer

By substituting the expression for u into the equation describing the state dynamics, we can write

$$\dot{x} = (A - BK)x \quad (46)$$

We can now calculate the characteristic polynomial via

$$\det(sI - (A - BK)) = 0 \quad (47)$$

This gives us a second order polynomial

$$s^2 + ((c + k_{12})/m)s + k_{11}/m \quad (48)$$

by matching the coefficients with the desired closed loop characteristic polynomial, we get

$$k_{12} = 2\zeta m\omega_n - c, \quad k_{11} = m\omega_n^2 \quad (49)$$

6.0.5 Exercise 5

Consider the mechanical system shown in Figure 6 described by the same steady-state model. Assume that the feedback controller is

$$u = -Kx + K_r r \quad (50)$$

where r is the reference signal. We want to determine the steady state reference gain K_r such that the steady-state level is 1.

Answer

The reference gain K_r is given by

$$K_r = -(C(A - BK)^{-1}B)^{-1} \quad (51)$$

Thus, by substituting the values for k_{11} and k_{12} and performing the calculations, we arrive at

$$K_r = m\omega_n^2 \quad (52)$$

6.0.6 Exercise 6

Consider the closed loop system with the controller designed as in the previous exercise. Determine the transfer function for closed loop system, from the reference signal to the output. How many zeros does the system have?

Answer

The transfer function can be computed as

$$G_{ry}(s) = C(sI - A + BK)^{-1}BK_r$$

substituting the relevant quantities we arrive at

$$G_{ry}(s) = \frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

the number of zeros is 0 as the numerator is constant.

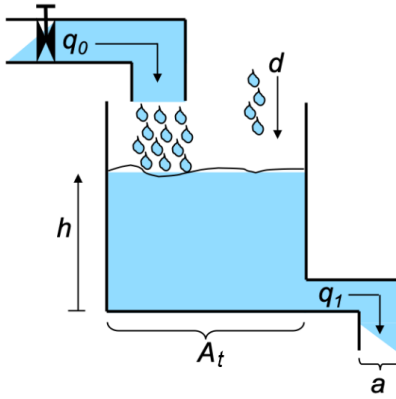


Fig. 7: Watertank schematics.

6.0.7 Exercise 7

Consider the tank system as shown in figure 7. The watertank is modeled by

$$\begin{aligned}\dot{h} &= \frac{1}{A_t}q_0 + \frac{1}{A_t}d - \frac{\alpha}{A_t}\sqrt{2gh} \\ q_1 &= \alpha\sqrt{2gh}\end{aligned}$$

where h is the level in the tank, q_0 is the inflow, d is the disturbance and q_1 is the outflow. Linearize the tank system around the equilibrium point $(h_e, q_{0e}, d_e) = (h_0, q_0, 0)$. Determine the matrices A , B , C and H , and corresponding to the model:

$$\begin{aligned}\Delta\dot{h} &= A\Delta h + B\Delta q_0 + H\Delta d \\ \Delta q_1 &= C\Delta h\end{aligned}$$

Answer

Let's consider the following functions

$$\begin{aligned}f(h, q_0, d) &= \frac{1}{A_t}q_0 + \frac{1}{A_t}d - \frac{\alpha}{A_t}\sqrt{2gh} \\ l(h, q_0, d) &= \alpha\sqrt{2gh}\end{aligned}$$

The linearized system can be determined as:

$$\begin{aligned}\Delta \dot{h} &= \frac{\partial f}{\partial h} \Delta h + \frac{\partial f}{\partial q_0} \Delta q_0 + \frac{\partial f}{\partial d} \Delta d \\ \Delta q_1 &= \frac{\partial l}{\partial h} \Delta h\end{aligned}$$

The partial derivatives can be expressed as

$$\frac{\partial f}{\partial h} = -\frac{\alpha\sqrt{2g}}{2A_t\sqrt{h}}, \quad \frac{\partial f}{\partial q_0} = \frac{1}{A_t}, \quad \frac{\partial f}{\partial d} = \frac{1}{A_t}, \quad \frac{\partial l}{\partial h} = \frac{\alpha\sqrt{2g}}{2\sqrt{h}}$$

Inserting the equilibrium point gives:

$$\begin{aligned}\Delta \dot{h} &= -\frac{\alpha\sqrt{2g}}{2A_t\sqrt{h_0}} \Delta h + \frac{1}{A_t} \Delta q_0 + \frac{1}{A_t} \Delta d \\ \Delta q_1 &= \frac{\alpha\sqrt{2g}}{2\sqrt{h_0}} \Delta h\end{aligned}$$

Thus, the matrices are

$$A = -\frac{\alpha\sqrt{2g}}{2A_t\sqrt{h_0}}, \quad B = \frac{1}{A_t}, \quad H = \frac{1}{A_t}, \quad C = \frac{\alpha\sqrt{2g}}{2\sqrt{h_0}}$$

6.0.8 Exercise 8

A leaf spring system is shown in figure 8

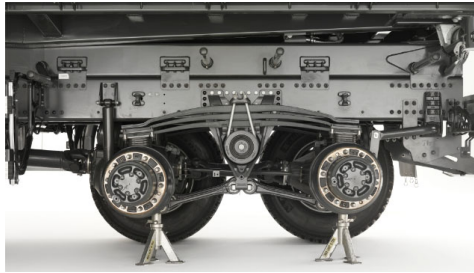


Fig. 8: Leaf spring system.

The system can be described using the following nonlinear model:

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ -\frac{k_1 x_1}{m} - \frac{k_2 x_1^3}{m} \end{bmatrix}$$

Three equilibrium points was considered for x_{1e} namely 0 and $\pm\sqrt{-k_1/k_2}$. Linearize the model around the equilibrium point

$$(x_{1e}, x_{2e}) = (\sqrt{-k_1/k_2}, 0)$$

Determine system matrix A Express the matrix elements using the system parameter symbols k_1, k_2 and m .

Answer

We can use the functions f_1, f_2 to condense the right hand side.

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ -\frac{k_1 x_1}{m} - \frac{k_2 x_1^3}{m} \end{bmatrix} = \begin{bmatrix} f_1(x_1, x_2) \\ f_2(x_1, x_2) \end{bmatrix}$$

The linearized system can be determined as:

$$\begin{bmatrix} \Delta \dot{x}_1 \\ \Delta \dot{x}_2 \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\frac{k_1}{m} - \frac{3k_2 x_{1e}^2}{m} & 0 \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \end{bmatrix}$$

Evaluation of the partial derivatives at the equilibrium point gives

$$\begin{bmatrix} \Delta \dot{x}_1 \\ \Delta \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ \frac{2k_1}{m} & 0 \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \end{bmatrix}$$

Thus,

$$A = \begin{bmatrix} 0 & 1 \\ \frac{2k_1}{m} & 0 \end{bmatrix}$$

6.0.9 Exercise 9

Consider the following longitudinal vehicle nonlinear model :

$$\begin{aligned} \frac{dx_1}{dt} &= \frac{1}{\tau} x_1 + \frac{ki}{\tau r_w} u \\ \frac{dx_2}{dt} &= \frac{1}{m} (x_1 - mgf \cos d_1 - \frac{1}{2} \rho C_D a_f x_2^2 - mgsind_1) \\ y &= x_2 \end{aligned}$$

where x_1 is the force at the wheels, x_2 is the vehicle velocity, u is the input signal, the accelerator pedal position, and d_1 is the disturbance, the road slope.

This problem is about determining the equilibrium point $(x_{1e}, x_{2e}, u_e, d_{1e})$ that corresponds to the condition when vehicle is traveling with a velocity of $15m/s$ on a flat road.

Answer

The equilibrium point is determined by setting the derivatives equal to zero:

$$\begin{aligned} 0 &= -\frac{1}{\tau}x_1 + \frac{ki}{\tau r_w}u \\ 0 &= \frac{1}{m} \left(x_1 - mgf \cos d_1 - \frac{1}{2}\rho C_D A_f x_2^2 - mg \sin d_1 \right) \\ y &= x_2 \end{aligned}$$

From the specification of the exercise we get $(x_{2e} = 15)$ and $(d_{1e} = 0)$. Using the algebraic equations the other two equilibrium points can be determined as $x_{1e} = \frac{1}{2}\rho C_D A_f x_{2e}^2 + mgf$ and $u_e = \frac{r_w}{ki}$, $x_{1e} = \frac{r_w \rho C_D A_f x_{2e}^2}{2ki}$. If we insert numerical values we get: $(x_{1e}, x_{2e}, u_e, d_{1e}) = (3216, 15, 0.2010, 0)$.

6.0.10 Exercise 10

Determine the partial derivatives needed for linearizing the the system

$$\begin{aligned} \frac{dx_1}{dt} &= \frac{1}{\tau}x_1 + \frac{ki}{\tau r_w}u = f_1(x_1, x_2, u, d_1) \\ \frac{dx_2}{dt} &= \frac{1}{m}(x_1 - mgf \cos d_1 - \frac{1}{2}\rho C_D a_f x_2^2 - mgsind_1) = f_2(x_1, x_2, u, d_1) \\ y &= x_2 = h(x_1, x_2, u, d_1) \end{aligned}$$

around the equilibrium point $(x_{1e}, x_{2e}, u_e, d_{1e})$. The linearized system can be determined as

$$\begin{aligned} \begin{bmatrix} \Delta \dot{x}_1 \\ \Delta \dot{x}_2 \end{bmatrix} &= \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \end{bmatrix} + \begin{bmatrix} \frac{\partial f_1}{\partial u} \\ \frac{\partial f_2}{\partial u} \end{bmatrix} \Delta u + \begin{bmatrix} \frac{\partial f_1}{\partial d_1} \\ \frac{\partial f_2}{\partial d_1} \end{bmatrix} \Delta d_1 \\ \Delta y &= \begin{bmatrix} \frac{\partial h}{\partial x_1} & \frac{\partial h}{\partial x_2} \end{bmatrix} \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \end{bmatrix} + \begin{bmatrix} \frac{\partial h}{\partial u} \end{bmatrix} \Delta u + \begin{bmatrix} \frac{\partial h}{\partial d_1} \end{bmatrix} \Delta d_1 \end{aligned} \quad (53)$$

Answer

We will have:

$$\begin{aligned}
\frac{\partial f_1}{\partial x_1} &= \frac{\partial}{\partial x_1} \left(-\frac{1}{\tau} x_1 + \frac{ki}{\tau r_w} u \right) = -\frac{1}{\tau} \\
\frac{\partial f_1}{\partial x_2} &= \frac{\partial}{\partial x_2} \left(-\frac{1}{\tau} x_1 + \frac{ki}{\tau r_w} u \right) = 0 \\
\frac{\partial f_1}{\partial u} &= \frac{\partial}{\partial u} \left(-\frac{1}{\tau} x_1 + \frac{ki}{\tau r_w} u \right) = \frac{ki}{\tau r_w} \\
\frac{\partial f_1}{\partial d_1} &= \frac{\partial}{\partial d_1} \left(-\frac{1}{\tau} x_1 + \frac{ki}{\tau r_w} u \right) = 0 \\
\frac{\partial f_2}{\partial x_1} &= \frac{\partial}{\partial x_1} \frac{1}{m} \left(x_1 - mgf \cos d_1 - \frac{1}{2} \rho C_D A_f x_2^2 - mg \sin d_1 \right) = \frac{1}{m} \\
\frac{\partial f_2}{\partial x_2} &= \frac{\partial}{\partial x_2} \frac{1}{m} \left(x_1 - mgf \cos d_1 - \frac{1}{2} \rho C_D A_f x_2^2 - mg \sin d_1 \right) = -\frac{\rho C_D A_f x_2}{m} \\
\frac{\partial f_2}{\partial u} &= \frac{\partial}{\partial u} \frac{1}{m} \left(x_1 - mgf \cos d_1 - \frac{1}{2} \rho C_D A_f x_2^2 - mg \sin d_1 \right) = 0 \\
\frac{\partial f_2}{\partial d_1} &= \frac{\partial}{\partial d_1} \frac{1}{m} \left(x_1 - mgf \cos d_1 - \frac{1}{2} \rho C_D A_f x_2^2 - mg \sin d_1 \right) = gf \sin d_{1e} - g \cos d_{1e} \\
\frac{\partial h}{\partial x_1} &= \frac{\partial}{\partial x_1} x_2 = 0 \\
\frac{\partial h}{\partial x_2} &= \frac{\partial}{\partial x_2} x_2 = 1 \\
\frac{\partial h}{\partial u} &= \frac{\partial}{\partial u} x_2 = 0 \\
\frac{\partial h}{\partial d_1} &= \frac{\partial}{\partial d_1} x_2 = 0
\end{aligned}$$

6.0.11 Exercise 11

The linearized longitudinal vehicle dynamics model for a vehicle traveling at 20m/sec on a flat road is given by the following state-space model

The number of zeros is 0 as the numerator is constant.

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -1.25 & 0 \\ 0.000005 & -0.0024 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 20000 \\ 0 \end{bmatrix} u + \begin{bmatrix} 0 \\ -9.82 \end{bmatrix} d_1 \quad (54)$$

$$y = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad (55)$$

where x_1 is the force at the wheels, x_2 is the vehicle velocity, u is the input signal and d_1 is the disturbance of the road.

Design a state feedback cruise controller using pole placement on the form

$$u = -Kx + k_r r \quad (56)$$

such that the closed loop system's characteristic polynomial becomes

$$p(s) = s^2 + 2\zeta\omega_n s + \omega_n^2 \quad (57)$$

where $\omega_n = 0.6$ and $\zeta = 1/\sqrt{2}$. The reference gain k_r should be chosen such that the steady state gain is 1. The disturbance d_1 can be considered to be 0.

References

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- [2] Philip , Florent Althel, Brigitte dAndrea-Novet, and Arnaud de La Fortelle *The Kinematic Bicycle Model: a Consistent Model for Planning Feasible Trajectories for Autonomous Vehicles?* HAL Id: hal-01520869, <https://hal-polytechnique.archives-ouvertes.fr/hal-01520869>
- [3] Marcos R. O., A. Maximo *Model Predictive Controller for Trajectory Tracking by Differential Drive Robot with Actuation constraints*