

Testable linear shift-invariant systems

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Chapter 1

Signals and systems

It is assumed the reader is familiar with the concept of a function! That is, a map from the elements in a set X to the elements in another set Y . Consider sets

$$X = \left\{ \begin{array}{c} \text{Mario} \\ \text{Link} \\ \text{Ness} \end{array} \right\} \quad Y = \left\{ \begin{array}{c} \text{Freeman} \\ \text{Ryu} \\ \text{Sephiroth} \\ \text{Conker} \\ \text{Ness} \end{array} \right\}.$$

An example of function from X to Y is

$$f(x) = \begin{cases} \text{Conker} & x = \text{Mario} \\ \text{Sephiroth} & x = \text{Link} \\ \text{Sephiroth} & x = \text{Ness}. \end{cases}$$

The function f maps Mario to Conker, Link to Sephiroth, and Ness to Sephiroth. The set X is called the **domain** of the function f and the set Y a **range**. The value of f for input x is denoted by $f(x)$ so, for example, $f(\text{Mario}) = \text{Conker}$ and $f(\text{Link}) = \text{Sephiroth}$. The set of all functions mapping X to Y is denoted by $X \rightarrow Y$ and so $f \in X \rightarrow Y$ in the example above.

A **signal** is a function with domain the set of real numbers \mathbb{R} and range the set of complex numbers \mathbb{C} , that is, a signal is a function from the set $\mathbb{R} \rightarrow \mathbb{C}$. For example

$$\sin(\pi t), \quad \frac{1}{2}t^3, \quad e^{-t^2}$$

all represent signals with $t \in \mathbb{R}$. These signals are plotted in Figure 1.1. Many physical phenomena, such as sound, light, weather, and motion, can be modelled using signals. In this text we primarily focus on examples from

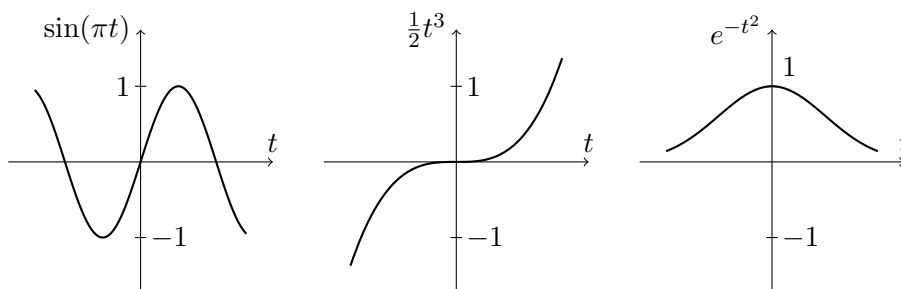


Figure 1.1: Plots of three signals.

electrical and mechanical engineering where signals are used to model changes in voltage, current, position, and angle over time. In these examples, the independent variable t represents “time”. However, there is no fundamental reason for this and the techniques developed here can be applied equally well when t represents a quantity other than time. An example where this occurs is image processing.

1.1 Properties of signals

A signal x is **bounded** if there exists a real number M such that

$$|x(t)| < M \quad \text{for all } t \in \mathbb{R}$$

where $|\cdot|$ denotes the (complex) magnitude. Both $\sin(\pi t)$ and e^{-t^2} are examples of bounded signals because $|\sin(\pi t)| \leq 1$ and $|e^{-t^2}| \leq 1$ for all $t \in \mathbb{R}$. However, $\frac{1}{2}t^3$ is not bounded because its magnitude grows indefinitely as t moves away from the origin.

A signal x is **periodic** if there exists a positive real number T such that

$$x(t) = x(t + kT) \quad \text{for all } k \in \mathbb{Z} \text{ and } t \in \mathbb{R}.$$

The smallest such positive T is called the **fundamental period** or simply the **period**. For example, the signal $\sin(\pi t)$ is periodic with period $T = 2$. Neither $\frac{1}{2}t^3$ or e^{-t^2} are periodic.

A signal x is **right sided** if there exists a $T \in \mathbb{R}$ such that $x(t) = 0$ for all $t < T$. Correspondingly x is **left sided** if $x(t) = 0$ for all $T > t$. For example, the **step function**

$$u(t) = \begin{cases} 1 & t \geq 0 \\ 0 & t < 0 \end{cases} \quad (1.1.1)$$

is right-sided. Its horizontal reflection $u(-t)$ is left sided (Figure 1.2). A signal x is said to be **finite** or to have **finite support** if it is both left and

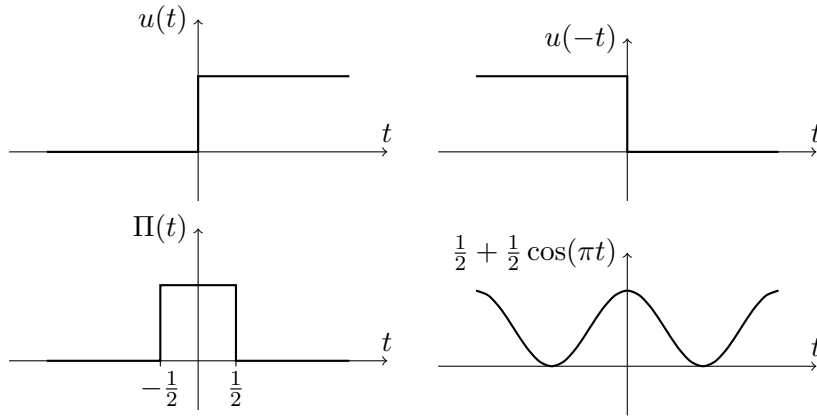


Figure 1.2: The right sided step function $u(t)$, its left sided reflection $u(-t)$, the finite rectangular pulse $\Pi(t)$ and the signal $\frac{1}{2} + \frac{1}{2} \cos(x)$ that is not finite.

right sided, that is, if there exists a $T \in \mathbb{R}$ such that $x(t) = x(-t) = 0$ for all $t > T$. The signals $\sin(\pi t)$ and e^{-t^2} do not have finite support, but the **rectangular pulse**

$$\Pi(t) = \begin{cases} 1 & |t| < \frac{1}{2} \\ 0 & \text{otherwise} \end{cases} \quad (1.1.2)$$

does.

A signal x is **even** (or **symmetric**) if $x(t) = x(-t)$ for all $t \in \mathbb{R}$ and **odd** (or **antisymmetric**) if $x(t) = -x(-t)$ for all $t \in \mathbb{R}$. For example, $\sin(\pi t)$ and $\frac{1}{2}t^3$ are odd and e^{-t^2} is even. A signal x is **conjugate symmetric** if $x(t) = x(-t)^*$ for all $t \in \mathbb{R}$ and **conjugate antisymmetric** if $x(t) = -x(-t)^*$ for all $t \in \mathbb{R}$, where $*$ denotes the complex conjugate of a complex number. Equivalently, x is conjugate symmetric if its real part $\text{Re}(x)$ is an even signal and its imaginary part $\text{Im}(x)$ is an odd signal, and x is conjugate antisymmetric if its real part is odd and its imaginary part is even. For example, the signal $e^{-t^2} + j \sin(\pi t)$ where $j = \sqrt{-1}$ is conjugate symmetric and the signal $\frac{1}{2}t^3 + j e^{-t^2}$ is conjugate antisymmetric.

A signal x is **continuous at** $t \in \mathbb{R}$ if

$$\lim_{h \rightarrow 0} x(t+h) = \lim_{h \rightarrow 0} x(t-h)$$

and x is said to be **continuous** if it is continuous at all $t \in \mathbb{R}$. The signals $\sin(\pi t)$, $\frac{1}{2}t^3$, and e^{-t^2} are continuous, but the step function u is not continuous at zero because

$$\lim_{h \rightarrow 0} u(h) = 1 \neq 0 = \lim_{h \rightarrow 0} u(-h).$$

The set of continuous signals is denoted by $C^0(\mathbb{R})$ or just C^0 . A signal x is

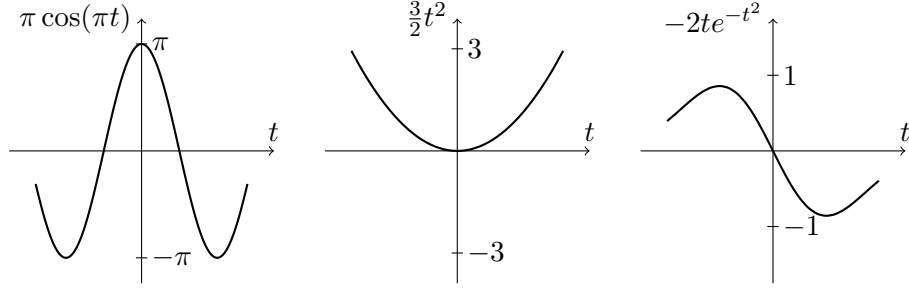


Figure 1.3: Derivatives of the signals $\sin(\pi t)$, $\frac{1}{2}t^3$, e^{-t^2} from Figure 1.1.

continuously differentiable or just **differentiable** if

$$\lim_{h \rightarrow 0} \frac{x(t+h) - x(t)}{h} = \lim_{h \rightarrow 0} \frac{x(t) - x(t-h)}{h} \quad \text{for all } t \in \mathbb{R}.$$

Considered as a function of t this limit is called the **derivative** of x at t and is typically denoted by $\frac{d}{dt}x(t)$. For example, the signals $\sin(\pi t)$, $\frac{1}{2}t^3$, e^{-t^2} , and t^2 are differentiable with derivatives

$$\pi \cos(\pi t), \quad \frac{3}{2}t^2, \quad -2te^{-t^2}, \quad 2t,$$

but the step function u and the rectangular pulse Π are not differentiable (Exercise 1.7). The set of differentiable signals is denoted by C^1 or $C^1(\mathbb{R})$. A signal is **k -times differentiable** if its $k-1$ th derivative is differentiable. The set of k -times differentiable signals is denoted by C^k or $C^k(\mathbb{R})$.

A signal x is **locally integrable** if

$$\int_a^b |x(t)| dt < \infty$$

for all finite constants a and b , where $< \infty$ means that the integral evaluates to a finite number. The signals $\sin(\pi t)$, $\frac{1}{2}t^3$, and e^{-t^2} are all locally integrable. An example of a signal that is not locally integrable is $x(t) = \frac{1}{t}$ (Exercise 1.3). The set of locally integrable signals is denoted by L_{loc} or $L_{\text{loc}}(\mathbb{R})$.

A signal x is **absolutely integrable** if

$$\|x\|_1 = \int_{-\infty}^{\infty} |x(t)| dt < \infty. \quad (1.1.3)$$

Here we introduce the notation $\|x\|_1$ called the **L^1 -norm** of x . For example $\sin(\pi t)$ and $\frac{1}{2}t^3$ are not absolutely integrable, but e^{-t^2} is because [Nicholas and Yates, 1950]

$$\int_{-\infty}^{\infty} |e^{-t^2}| dt = \int_{-\infty}^{\infty} e^{-t^2} dt = \sqrt{\pi}. \quad (1.1.4)$$

It is common to denote the set of absolutely integrable signals by L^1 or $L^1(\mathbb{R})$. So, $e^{-t^2} \in L^1$ and $\frac{1}{2}t^3 \notin L^1$. A signal x is **square integrable** if

$$\|x\|_2^2 = \int_{-\infty}^{\infty} |x(t)|^2 dt < \infty.$$

The real number $\|x\|_2$ is called the L^2 -**norm** of x . Square integrable signals are also called **energy signals** and the squared L^2 -norm $\|x\|_2^2$ is called the **energy** of x . For example, $\sin(\pi t)$ and $\frac{1}{2}t^3$ are not energy signals, but e^{-t^2} is. It has energy $\|e^{-t^2}\|_2^2 = \sqrt{\pi/2}$ (Exercise 1.6). The set of square integrable signals is denoted by L^2 or $L^2(\mathbb{R})$.

We write $x = y$ to indicate that two signals x and y are **equal pointwise**, that is, $x(t) = y(t)$ for all $t \in \mathbb{R}$. This definition of equality is often stronger than we desire. For example, the step function u and the signal

$$z(t) = \begin{cases} 1 & t > 0 \\ 0 & t \leq 0 \end{cases}$$

are not equal pointwise because they are not equal at $t = 0$ since $u(0) = 1$ and $z(0) = 0$. It is useful to identify signals that differ only at isolated points and for this we use a weaker definition of equality. We say that two signals x and y are equal **almost everywhere** if

$$\int_a^b |x(t) - y(t)| dt = 0$$

for all finite constants a and b . So, in the previous example, while $u \neq z$ pointwise we do have $u = z$ almost everywhere. Typically the term almost everywhere is abbreviated to a.e. and one writes

$$x = y \text{ a.e.} \quad \text{or} \quad x(t) = y(t) \text{ a.e.}$$

to indicate that the signals x and y are equal almost everywhere.

1.2 Spaces of signals

We will regularly be interested in subsets of the set of all signals $\mathbb{R} \rightarrow \mathbb{C}$. Two important families of subsets are the **linear spaces** and the **shift-invariant spaces**.

Let x and y be signals. We denote by $x + y$ the signal that takes the value $x(t) + y(t)$ for each $t \in \mathbb{R}$, that is, the signal that results from adding x and y . For a a complex number we denote by ax the signal that takes the value $ax(t)$ for each $t \in \mathbb{R}$, that is, the signal that results from multiplying x by a (Figure 1.4). For signals x and y and complex numbers a and b the signal

$$ax + by$$

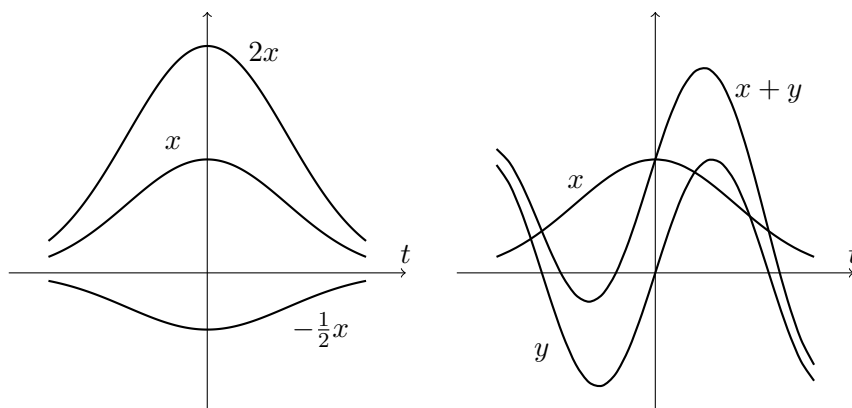


Figure 1.4: The signal $x(t) = e^{-t^2}$ and the signals $2x$ and $-\frac{1}{2}x$ (left). The signals $x(t) = e^{-t^2}$ and $y(t) = \sin(\pi t)$ and the signal $x + y$.

is called a **linear combination** of x and y .

Let $X \subseteq \mathbb{R} \rightarrow \mathbb{C}$ be a set of signals. The set X is a **linear space** (or **vector space**) if for all signals x and y from X and all complex numbers a and b the linear combination $ax + by$ is also in X . The set of all signals $\mathbb{R} \rightarrow \mathbb{C}$ is a linear space. Another example is the set of differentiable signals, because, if x and y are differentiable, then the linear combination $ax + by$ is differentiable. The derivative is $aDx + bDy$. The set of even signals is another example of a linear space because if x and y are even then $ax(t) + by(t) = ax(-t) + by(-t)$ and so the linear combination $ax + by$ is even. The set of locally integrable signals L_{loc} , the set of absolutely integrable signals L^1 , and the set of square integrable signals L^2 are linear spaces (Exercise 1.9). The set of periodic signals is not a linear space (Exercise 1.10).

For a real number τ the signal $x(t - \tau)$ is called a *time-shift* or just *shift* of the signal $x(t)$. Figure 1.5 depicts the shift $x(t - \tau)$ for different values of τ in the case that $x(t) = e^{-t^2}$. A set of signals $X \subseteq \mathbb{R} \rightarrow \mathbb{C}$ is a **shift-invariant space** if for all $x \in X$ and all $\tau \in \mathbb{R}$ the shift $x(t - \tau)$ is also in X . Examples of shift-invariant spaces are the set of differentiable signals, the set of periodic signals (Exercise 1.10), and L_{loc} , L^1 , and L^2 (1.9). The set of even signals and the set of odd signals are not shift-invariant spaces.

1.3 Systems (functions of signals)

A **system** is a function that maps a signal to another signal. For example,

$$x(t) + 3x(t - 1), \quad \int_0^1 x(t - \tau) d\tau, \quad \frac{1}{x(t)}, \quad \frac{d}{dt}x(t)$$

represent systems, each mapping the signal x to another signal. Consider the electric circuit in Figure 1.6 called a **voltage divider**. If the voltage at

time t is $x(t)$ then, by Ohm's law, the current at time t satisfies

$$i(t) = \frac{1}{R_1 + R_2} x(t),$$

and the voltage over the resistor R_2 is

$$y(t) = R_2 i(t) = \frac{R_2}{R_1 + R_2} x(t). \quad (1.3.1)$$

The circuit can be considered as a system mapping the signal x representing the voltage to the signal $i = \frac{1}{R_1 + R_2} x$ representing the current, or a system mapping x to the signal $y = \frac{R_2}{R_1 + R_2} x$ representing the voltage over resistor R_2 .

Let $X \subseteq \mathbb{R} \rightarrow \mathbb{C}$ and $Y \subseteq \mathbb{R} \rightarrow \mathbb{C}$ be sets of signals. We denote systems with capital letters such as H and G . A system is a function $H \in X \rightarrow Y$ that maps each signal from the domain X to a signal from the range Y . Given **input signal** $x \in X$ the **output signal** of the system is denoted by $H(x)$. The output signal is often called the **response** of system H to signal x . We will often drop the brackets and write simply Hx for the response of H to x ¹. The value of the output signal Hx at $t \in \mathbb{R}$ is denoted by $Hx(t)$ or $H(x)(t)$ or $H(x, t)$ and we do not distinguish between these notations. It is sometimes useful to depict systems with a block diagram as in Figure 1.7. The electric circuit in Figure 1.6 corresponds with the system

$$Hx = \frac{R_2}{R_1 + R_2} x = y.$$

This system multiplies the input signal x by $\frac{R_2}{R_1 + R_2}$. This brings us to our first practical test.

Test 1 (Voltage divider) In this test we construct the voltage divider from Figure 1.6 on a breadboard with resistors $R_1 \approx 100\Omega$ and $R_2 \approx 470\Omega$ with values accurate to within 5%. Using a computer soundcard (an approximation of) the voltage signal

$$x(t) = \sin(2\pi f_1 t) \quad \text{with} \quad f_1 = 100$$

is passed through the circuit. The approximation is generated by sampling $x(t)$ at rate $F = \frac{1}{P} = 44100\text{Hz}$ to generate samples

$$x(nP) \quad n = 0, \dots, 2F$$

¹In the literature it is customary to drop the brackets only when H is a **linear** system (Section 1.5). In this text we occasionally drop the brackets even when H is not linear. Since we deal primarily with linear systems this Faux Pas will occur rarely.

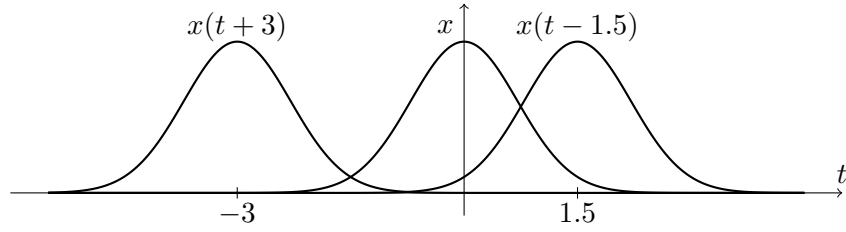


Figure 1.5: The signal $x(t) = e^{-t^2}$ and shift $x(t - \tau) = e^{-(t-\tau)^2}$ for $\tau = 1.5$ and -3 .

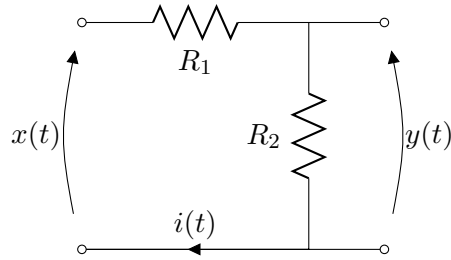


Figure 1.6: A **voltage divider** circuit.

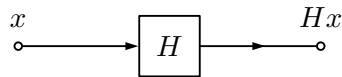


Figure 1.7: System block diagram with input signal x and output signal $H(x)$.

corresponding to approximately 2 seconds of signal. These samples are passed to the soundcard which starts playback. The voltage over resistor R_2 is recorded (also using the soundcard) that returns a list of samples y_1, \dots, y_L taken at rate F . The voltage over R_2 can be (approximately) reconstructed from these samples as

$$\tilde{y}(t) = \sum_{\ell=1}^L y_{\ell} \operatorname{sinc}(Ft - \ell) \quad (1.3.2)$$

where

$$\operatorname{sinc}(t) = \frac{\sin(\pi t)}{\pi t} \quad (1.3.3)$$

is called the **sinc function** and is plotted in Figure 5.1. We will justify this reconstruction in Section 5.4. Simultaneously the (stereo) soundcard is used to record the input voltage x producing samples x_1, \dots, x_L taken at rate F . An approximation of the input signal is

$$\tilde{x}(t) = \sum_{\ell=1}^L x_{\ell} \operatorname{sinc}(Ft - \ell). \quad (1.3.4)$$

In view of (1.3.1) we would expect the approximate relationship

$$\tilde{y} \approx \frac{R_2}{R_1 + R_2} \tilde{x} = \frac{47}{57} \tilde{x}.$$

A plot of \tilde{y} , \tilde{x} and $\frac{47}{57} \tilde{x}$ over a 20ms period from 1s to 1.02s is given in Figure 1.8. The hypothesised output signal $\frac{47}{57} \tilde{x}$ does not match the observed output signal \tilde{y} . A primary reason is that the circuitry inside the soundcard itself cannot be ignored. When deriving the equation for the voltage divider we implicitly assumed that current flows through the output of the soundcard without resistance (a short circuit), and that no current flows through the input device of the soundcard (an open circuit). These assumptions are not realistic. Modelling the circuitry in the sound card won't be attempted here. In Section 2.2 we will construct circuits that contain external sources of power (active circuits). These are less sensitive to the circuitry inside the soundcard.

When specifying a system we are free to choose the domain X and range Y at our convenience. In cases such as the voltage divider it is reasonable to choose the domain $X = \mathbb{R} \rightarrow \mathbb{C}$, that is, the domain can contain *all* signals. However, this is not always convenient or possible. For example, the system

$$Hx(t) = \frac{1}{x(t)}$$

is not defined at those t where $x(t) = 0$ because we cannot divide by zero. To avoid this we might choose the domain as the set of signals $x(t)$ that are not zero for any $t \in \mathbb{R}$.

Another example is the system I_∞ defined by

$$I_\infty x(t) = \int_{-\infty}^t x(\tau) d\tau, \quad (1.3.5)$$

called an **integrator**. The signal $x(t) = 1$ cannot be input to the integrator because the integral $\int_{-\infty}^t dt$ is not finite for any t . However, the integrator I_∞ can operate on absolutely integrable signals because, if x is absolutely integrable, then

$$I_\infty x(t) = \int_{-\infty}^t x(\tau) d\tau \leq \int_{-\infty}^t |x(\tau)| d\tau < \int_{-\infty}^{\infty} |x(\tau)| d\tau = \|x\|_1 < \infty$$

for all $t \in \mathbb{R}$. We might then choose a domain for I_∞ as the set of absolutely integrable signals L^1 . The integrator can also be applied to signals that are right sided and locally integrable because, for any right sided signal x there exists $T \in \mathbb{R}$ such that $x(t) = 0$ for all $t < T$ and so,

$$I_\infty x(t) = \int_{-\infty}^t x(\tau) d\tau = \int_T^t x(\tau) d\tau < \infty$$

for all $t \in \mathbb{R}$ if x is locally integrable. So another possible domain for I_∞ is the set of right sided locally integrable signals. A final possible domain is the subset of locally integrable signals for which $\int_{-\infty}^0 |x(t)| dt$ is finite. This last example will be the domain we usually choose for the integrator I_∞ .

1.4 Some important systems

The system

$$T_\tau x(t) = x(t - \tau)$$

is called a **time-shifter** or simply **shifter**. This system shifts the input signal along the t axis (“time” axis) by τ . When τ is positive T_τ delays the input signal by τ . The shifter will appear so regularly that we use the special notation T_τ to represent it. Figure 1.5 depicts the action of shifters $T_{1.5}$ and T_{-3} on the signal $x(t) = e^{-t^2}$. When $\tau = 0$ the shifter is the **identity system** $T_0 x = x$ that maps a signal to itself. Another important system is the **time-scaler** that has the form

$$Hx(t) = x(\alpha t), \quad \alpha \in \mathbb{R}.$$

Figure 1.9 depicts the action of time-scalers with different values for α . When $\alpha = -1$ the time-scaler reflects the input signal in the t axis. When

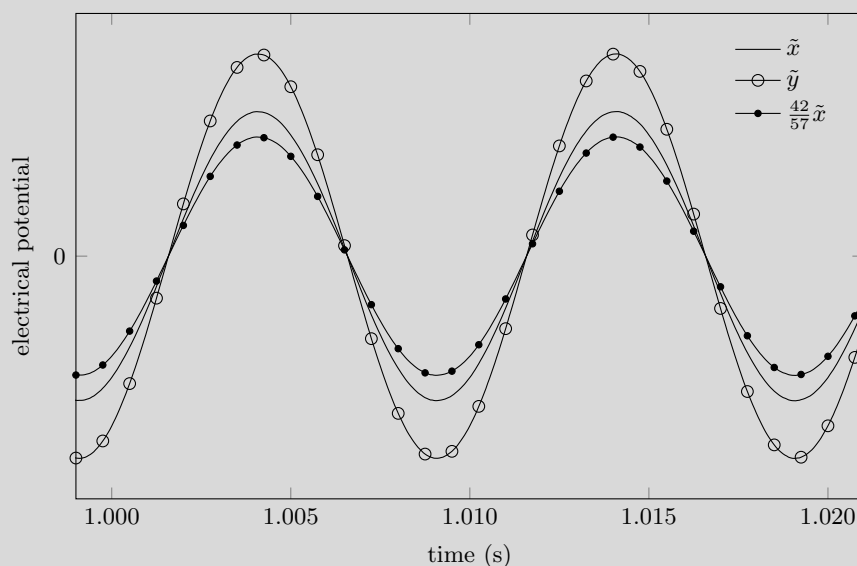


Figure 1.8: Plot of reconstructed input signal \tilde{x} (solid line), output signal \tilde{y} (solid line with circle) and hypothesised output signal $\frac{42}{57}\tilde{x}$ (solid line with dot) for the voltage divider circuit in Figure 1.6. The hypothesised signal does not match \tilde{y} . One reason is that the model does not take account of the circuitry inside the soundcard.

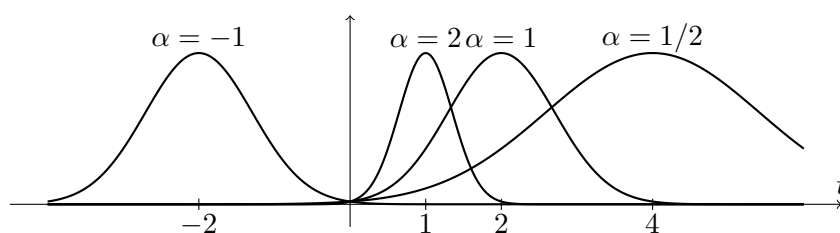


Figure 1.9: Time-scaler system $Hx(t) = x(\alpha t)$ for $\alpha = -1, \frac{1}{2}, 1$ and 2 acting on the signal $x(t) = e^{-(t-2)^2}$.

$\alpha = 1$ the time-scaler is the identity system T_0 . Both the shifter and time-scaler are well defined for all signals and so it is reasonable to choose their domains to be the entire set of signals $\mathbb{R} \rightarrow \mathbb{C}$. We always assume this is the case unless otherwise stated.

Another regularly encountered system is the **differentiator**

$$Dx(t) = \frac{d}{dt}x(t)$$

that returns the derivative of the input signal. We also define a k th differentiator

$$D^k x(t) = \frac{d^k}{dt^k} x(t)$$

that returns the k th derivative of the input signal. The differentiator is only defined for differentiable signals. The largest possible domain for D is the set of differentiable signals C^1 and the largest possible domain for D^k is the set of k -times differentiable signals C^k . Unless otherwise stated we will always assume the domain of D^k to be C^k .

Another important system is the **integrator**

$$I_a x(t) = \int_{-a}^t x(\tau) d\tau.$$

The parameter a describes the lower bound of the integral. In this course it will often be that $a = \infty$. For example, the response of the integrator I_∞ to the signal $tu(t)$ where u is the step function (1.1.1) is

$$\int_{-\infty}^t \tau u(\tau) d\tau = \begin{cases} \int_0^t \tau d\tau = \frac{t^2}{2} & t > 0 \\ 0 & t \leq 0. \end{cases}$$

Observe that the integrator I_∞ cannot be applied to the signal $x(t) = t$ because $\int_{-\infty}^t \tau d\tau$ is not finite for any t . A domain for I_∞ cannot contain the signal $x(t) = t$. Unless otherwise stated we will assume the domain of I_∞ to be the subset of locally integrable signals L_{loc} for which $\int_{-\infty}^0 |x(t)| dt < \infty$ (Exercise 1.15). For finite a we will assume, unless otherwise stated, that the domain of I_a is the set of locally integrable signals L_{loc} .

1.5 Properties of systems

A system $H \in X \rightarrow Y$ is called **memoryless** if, for all input signals $x \in X$, the output signal Hx at time t depends only on x at time t . For example $\frac{1}{x(t)}$ and the identity system T_0 are memoryless, but

$$x(t) + 3x(t-1) \quad \text{and} \quad \int_0^1 x(t-\tau) d\tau$$

are not. A shifter T_τ with $\tau \neq 0$ is not memoryless.

A system $H \in X \rightarrow Y$ is **causal** if, for all input signals $x \in X$, the output signal Hx at time t depends on x at times less than or equal to t . Memoryless systems such as $\frac{1}{x(t)}$ and T_0 are also causal. The shifter T_τ is causal when $\tau \geq 0$, but is not causal when $\tau < 0$. The systems

$$x(t) + 3x(t-1) \quad \text{and} \quad \int_0^1 x(t-\tau) d\tau$$

are causal, but the systems

$$x(t) + 3x(t+1) \quad \text{and} \quad \int_0^1 x(t+\tau) d\tau$$

are not causal.

A system $H \in X \rightarrow Y$ is called **bounded-input-bounded-output (BIBO) stable** or just **stable** if the output signal Hx is bounded whenever the input signal x is bounded. That is, H is stable if for every positive real number M there exists a positive real number K such that for all input signals $x \in X$ bounded below M , that is,

$$|x(t)| < M \quad \text{for all } t \in \mathbb{R},$$

it holds that the output signal Hx is bounded below K , that is,

$$|Hx(t)| < K \quad \text{for all } t \in \mathbb{R}.$$

For example, the system $x(t) + 3x(t-1)$ is stable with $K = 4M$ since if $|x(t)| < M$, then

$$|x(t) + 3x(t-1)| \leq |x(t)| + 3|x(t-1)| < 4M = K.$$

The integrator I_a for any $a \in \mathbb{R}$ and differentiator D are not stable (Exercises 1.16 and 1.17).

Let $H \in X \rightarrow Y$ be a system with domain X and range Y being linear spaces. The system H is **linear** if

$$H(ax + by) = aHx + bHy$$

for all signals $x, y \in X$ and all complex numbers a and b . That is, a linear system has the property: If the input consists of a weighted sum of signals, then the output consists of the same weighted sum of the responses of the system to those signals. Figure 1.10 indicates the linearity property using a block diagram. For example, the differentiator is linear because

$$D(ax + by)(t) = \frac{d}{dt}(ax(t) + by(t)) = a\frac{d}{dt}x(t) + b\frac{d}{dt}y(t) = aDx(t) + bDy(t)$$

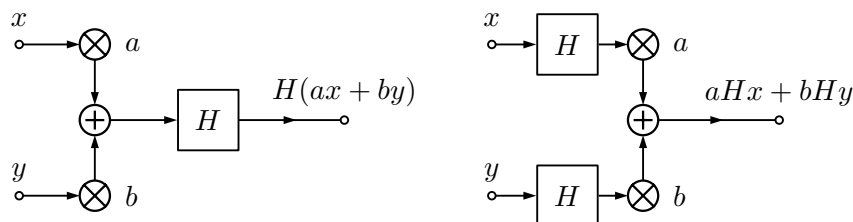


Figure 1.10: If H is a linear system the outputs of these two diagrams are the same signal, i.e. $H(ax + by) = aHx + bHy$.

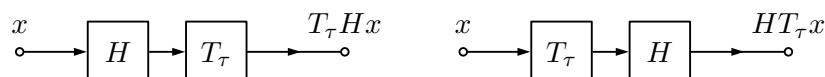


Figure 1.11: If H is a shift-invariant system the outputs of these two diagrams are the same signal, i.e. $HT_\tau x = T_\tau Hx$.

whenever both x and y are differentiable. However, the system $Hx(t) = \frac{1}{x(t)}$ is not linear because

$$H(ax + by)(t) = \frac{1}{ax(t) + by(t)} \neq \frac{a}{x(t)} + \frac{b}{y(t)} = aHx(t) + bHy(t)$$

in general.

Let $H \in X \rightarrow Y$ be a system with domain X and range Y being shift-invariant spaces. The system H is **shift-invariant** (or **time-invariant**) if

$$HT_\tau x(t) = Hx(t - \tau)$$

for all signals $x \in X$ and all shifts $\tau \in \mathbb{R}$. That is, a system is shift-invariant if shifting the input signal results in the same shift of the output signal. Equivalently, H is shift-invariant if it commutes with the shifter T_τ , that is, if

$$HT_\tau x = T_\tau Hx$$

for all $\tau \in \mathbb{R}$ and all signals $x \in X$. Figure 1.11 represents the property of shift-invariance with a block diagram.

Exercises

- 1.1. How many distinct functions from the set $X = \{\text{Mario}, \text{Link}\}$ to the set $Y = \{\text{Freeman}, \text{Ryu}, \text{Sephiroth}\}$ exist? Write down each function, that is, write down all functions from the set $X \rightarrow Y$.
- 1.2. State whether the step function $u(t)$ is bounded, periodic, absolutely integrable, an energy signal.

1.3. Show that the signal t^2 is locally integrable, but that the signal $\frac{1}{t^2}$ is not.

1.4. Plot the signal

$$x(t) = \begin{cases} \frac{1}{t+1} & t > 0 \\ \frac{1}{t-1} & t \leq 0. \end{cases}$$

State whether it is: bounded, locally integrable, absolutely integrable, square integrable.

1.5. Plot the signal

$$x(t) = \begin{cases} \frac{1}{\sqrt{t}} & 0 < t \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

Show that x is absolutely integrable, but not square integrable.

1.6. Compute the energy of the signal $e^{-\alpha^2 t^2}$ (Hint: use equation (1.1.4) on page 4 and a change of variables).

1.7. Show that the signal t^2 is differentiable, but the step function u and rectangular pulse Π are not.

1.8. Plot the signal $\sin(t) + \sin(\pi t)$. Show that this signal is not periodic.

1.9. Show that the set of locally integrable signals L_{loc} , the set of absolutely integrable signals L^1 , and the set of square integrable signals L^2 are linear shift-invariant spaces.

1.10. Show that the set of periodic signals is a shift-invariant space, but not a linear space.

1.11. Show that the set of bounded signals is a linear shift-invariant space.

1.12. Let $K > 0$ be a fixed real number. Show that the set of signals bounded below K is a shift invariant space, but not a linear space.

1.13. Show that the set of even signals and the set of odd signals are not shift invariant spaces.

1.14. Show that the integrator I_c with finite $c \in \mathbb{R}$ is not stable.

1.15. Show that if the signal x is locally integrable and $\int_{-\infty}^0 |x(t)| dt < \infty$ then $I_{\infty}x(t) = \int_{-\infty}^t x(t)dt < \infty$ for all $t \in \mathbb{R}$.

1.16. Show that the integrator I_{∞} is not stable.

1.17. Show that the differentiator system D is not stable.

1.18. Show that the shifter T_{τ} is linear and shift-invariant and that the time-scaler is linear, but not time invariant.

- 1.19. Show that the integrator I_c with finite $c \in \mathbb{R}$ is linear, but not shift-invariant.
- 1.20. Show that the integrator I_∞ is linear and shift-invariant.
- 1.21. State whether the system $Hx = x + 1$ is linear, shift-invariant, stable.
- 1.22. State whether the system $Hx = 0$ is linear, shift-invariant, stable.
- 1.23. State whether the system $Hx = 1$ is linear, shift-invariant, stable.
- 1.24. Let x be a signal with period T that is not equal to zero almost everywhere. Show that x is neither absolutely integrable nor square integrable.

Chapter 2

Systems modelled by differential equations

Systems of particular interest are those where the input signal x and output signal y are related by a linear differential equation with constant coefficients, that is, an equation of the form

$$\sum_{\ell=0}^m a_{\ell} \frac{d^{\ell}}{dt^{\ell}} x(t) = \sum_{\ell=0}^k b_{\ell} \frac{d^{\ell}}{dt^{\ell}} y(t),$$

where a_0, \dots, a_m and b_0, \dots, b_k are real or complex numbers. In what follows we use the differentiator system D rather than the notation $\frac{d}{dt}$ to represent differentiation. To represent the ℓ th derivative we write D^{ℓ} instead of $\frac{d^{\ell}}{dt^{\ell}}$. Using this notation the differential equation above is

$$\sum_{\ell=0}^m a_{\ell} D^{\ell} x = \sum_{\ell=0}^k b_{\ell} D^{\ell} y. \quad (2.0.1)$$

Equations of this form can be used to model a large number of electrical, mechanical and other real world devices. For example, consider the resistor and capacitor (RC) circuit in Figure 2.1. Let the signal v_R represent the voltage over the resistor and i the current through both resistor and capacitor. The voltage signals satisfy

$$x = y + v_R,$$

and the current satisfies both

$$v_R = Ri \quad \text{and} \quad i = CDy.$$

Combining these equations,

$$x = y + RCDy \quad (2.0.2)$$

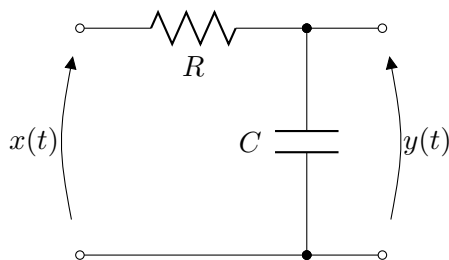


Figure 2.1: An electrical circuit with resistor and capacitor in series, otherwise known as an **RC circuit**.

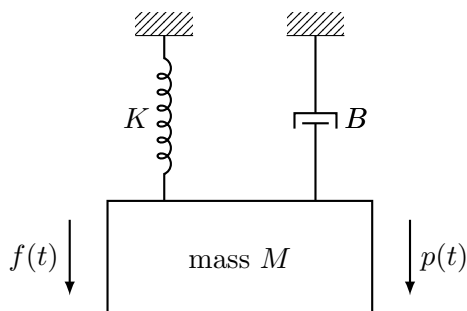


Figure 2.2: A mechanical mass-spring-damper system

that is in the form of (2.0.1).

As another example, consider the mass-spring-damper in Figure 2.2. A force represented by the signal f is externally applied to the mass, and the position of the mass is represented by the signal p . The spring exerts force $-Kp$ that is proportional to the position of the mass, and the damper exerts force $-BDp$ that is proportional to the velocity of the mass. The cumulative force exerted on the mass is

$$f_m = f - Kp - BDp$$

and by Newton's law the acceleration of the mass D^2p satisfies

$$MD^2p = f_m = f - Kp - BDp.$$

We obtain the differential equation

$$f = Kp + BDp + MD^2p \quad (2.0.3)$$

that is in the form of (2.0.1) if we put $x = f$ and $y = p$. Given p we can readily solve for the corresponding force f . As a concrete example, let the spring constant, damping constant and mass be $K = B = M = 1$. If the position satisfies $p(t) = e^{-t^2}$, then the corresponding force satisfies

$$f(t) = e^{-t^2}(4t^2 - 2t - 1).$$

Figure 2.3: A solution to the mass-spring-damper system with $K = B = M = 1$. The position is $p(t) = e^{-t^2}$ with corresponding force $f(t) = e^{-t^2}(4t^2 - 2t - 1)$.

Figure 2.3 depicts these signals.

What happens if a particular force signal f is applied to the mass? For example, say we apply the force

$$f(t) = \Pi(t - \tfrac{1}{2}) = \begin{cases} 1 & 0 < t \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

What is the corresponding position signal p ? We are not yet ready to answer this question, but will be later (Exercise 4.14).

In both the mechanical mass-spring-damper system in Figure 2.2 and the electrical RC circuit in Figure 2.1 we obtain a differential equation relating the input signal x with the output signal y . The equations do not specify the output signal y explicitly in terms of the input signal x , that is, they do not explicitly define a system H such $y = Hx$. As they are, the differential equations do not provide as much information about the behaviour of the system as we would like. For example, is the system stable? Much more information about these systems will be obtained when the **Laplace transform** is introduced in Chapter 4. The remainder of this chapter details the construction of differential equations that model various mechanical, electrical, and electro-mechanical systems. The systems constructed will be used as examples throughout the text.

2.1 Passive electrical circuits

Passive electrical circuits require no sources of power other than the input signal itself. For example, the voltage divider in Figure 1.6 and the RC circuit in Figure 2.1 are passive circuits. Another common passive electrical circuit is the resistor, capacitor and inductor (RLC) circuit depicted in Figure 2.4. In this circuit we let the output signal y be the voltage over the resistor. Let v_C represent the voltage over the capacitor and v_L the voltage over the inductor and let i be the current. We have

$$y = Ri, \quad i = CDv_C, \quad v_L = LDi,$$

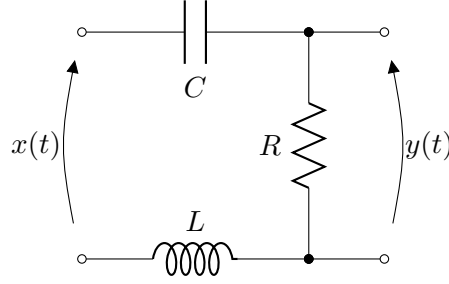


Figure 2.4: An electrical circuit with resistor, capacitor and inductor in series, otherwise known as an **RLC circuit**.

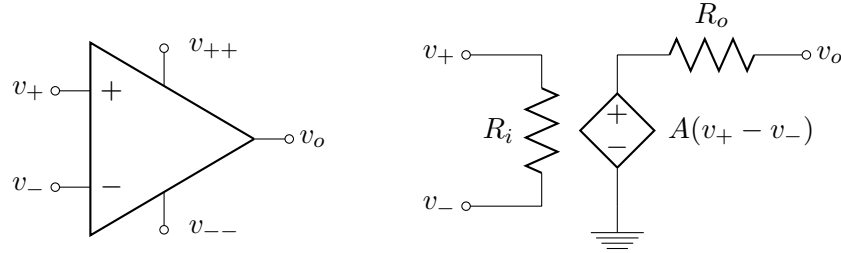


Figure 2.5: Left: triangular component diagram of an **operational amplifier**. The v_{++} and v_{--} connectors indicate where an external voltage source can be connected to the amplifier. These connectors will usually be omitted. Right: model for an operational amplifier including input resistance R_i , output resistance R_o , and open loop gain A . The diamond shaped component is a dependent voltage source. This model is usually only useful when the operational amplifier is in a negative feedback circuit.

leading to the following relationships between y , v_C and v_L ,

$$y = RCDv_C, \quad Rv_L = LDy.$$

Kirchhoff's voltage law gives $x = y + v_C + v_L$ and by differentiating both sides

$$Dx = Dy + Dv_C + Dv_L.$$

Substituting the equations relating y , v_C and v_L leads to

$$RCDx = y + RCDy + LCD^2y. \quad (2.1.1)$$

We can similarly find equations relating the input voltage with v_C and v_L .

2.2 Active electrical circuits

Unlike passive electrical circuits, an **active electrical circuit** requires a source of power external to the input signal. Active circuits can be modelled

and constructed using **operational amplifiers** as depicted in Figure 2.5. The left hand side of Figure 2.5 shows a triangular circuit diagram for an operational amplifier, and the right hand side of Figure 2.5 shows a circuit that can be used to model the behaviour of the amplifier. The v_{++} and v_{--} connectors indicate where an external voltage source can be connected to the amplifier. These connectors will usually be omitted. The diamond shaped component is a dependent voltage source with voltage $A(v_+ - v_-)$ that depends on the difference between the voltage at the **non-inverting input** v_+ and the voltage at the **inverting input** v_- . The dimensionless constant A is called the **open loop gain**. Most operational amplifiers have large open loop gain A , large **input resistance** R_i and small **output resistance** R_o . As we will see, it can be convenient to consider the behaviour as $A \rightarrow \infty$, $R_i \rightarrow \infty$ and $R_o \rightarrow 0$, resulting in an **ideal operational amplifier**.

As an example, an operational amplifier configured as a **multiplier** is depicted in Figure 2.6. This circuit is an example of an operational amplifier configured with **negative feedback**, meaning that the output of the amplifier is connected (in this case by a resistor) to the inverting input v_- . The horizontal wire at the bottom of the plot is considered to be ground (zero volts) and is connected to the negative terminal of the dependent voltage source of the operational amplifier depicted in Figure 2.5. An equivalent circuit for the multiplier using the model in Figure 2.5 is shown in Figure 2.7. Solving this circuit (Exercise 2.1) yields the following relationship between the input voltage signal x and the output voltage signal y ,

$$y = \frac{R_i(R_o - AR_2)}{R_i(R_2 + R_o) + R_1(R_2 + R_i + AR_i + R_o)}x. \quad (2.2.1)$$

For an ideal operational amplifier we let $A \rightarrow \infty$, $R_i \rightarrow \infty$ and $R_o \rightarrow 0$. In this case terms involving the product AR_i dominate and we are left with the simpler equation

$$y = -\frac{R_2}{R_1}x. \quad (2.2.2)$$

Thus, assuming an ideal operational amplifier, the circuit acts as a multiplier with constant $-\frac{R_2}{R_1}$.

The equation relating x and y is much simpler for the ideal operational amplifier. Fortunately this equation can be obtained directly using the following two rules:

1. the voltage at the inverting and non-inverting inputs are equal,
2. no current flows through the inverting and non-inverting inputs.

These rules are only useful for analysing circuits with negative feedback. Let us now rederive (2.2.2) using these rules. Because the non-inverting input is connected to ground, the voltage at the inverting input is zero. So, the voltage over resistor R_2 is $y = R_2i$. Because no current flows through the

inverting input the current through R_1 is also i and $x = -R_1 i$. Combining these results, the input voltage x and the output voltage y are related by

$$y = -\frac{R_2}{R_1}x.$$

In Test 2 the inverting amplifier circuit is constructed and the relationship above is tested using a computer soundcard.

We now consider another circuit consisting of an operational amplifier, two resistors and two capacitors depicted in Figure 2.8. Assuming an ideal operational amplifier, the voltage at the inverting terminal is zero because the non-inverting terminal is connected to ground. Thus, the voltage over capacitor C_2 and resistor R_2 is equal to y and, by Kirchoff's current law,

$$i = \frac{y}{R_2} + C_2 Dy.$$

Similarly, since no current flows through the inverting terminal,

$$i = -\frac{x}{R_1} - C_1 Dx.$$

Combining these equations yields

$$-\frac{x}{R_1} - C_1 Dx = \frac{y}{R_2} + C_2 Dy. \quad (2.2.3)$$

Observe the similarity between this equation and that for the passive RC circuit (2.0.2) when $R_1 = R_2$ and $C_1 = 0$ (an open circuit). In this case

$$x = -y - R_1 C_2 Dy. \quad (2.2.4)$$

We call this the **active RC circuit**. This circuit is tested in Test 3.

Consider the circuit in Figure 2.9. Assuming an ideal operational amplifier, the input voltage x satisfies

$$-i = \frac{x}{R_1} + C_1 Dx.$$

The voltage over the capacitor C_2 is $y - R_2 i$ and so the current satisfies

$$i = C_2 D(y - R_2 i).$$

Combining these equations gives

$$-\frac{x}{R_1} - C_1 Dx = C_2 Dy + \frac{R_2 C_2}{R_1} Dx + R_2 C_2 C_1 D^2 x,$$

and after rearranging,

$$Dy = -\frac{1}{R_1 C_2} x - \left(\frac{R_2}{R_1} + \frac{C_1}{C_2} \right) Dx - R_2 C_1 D^2 x.$$

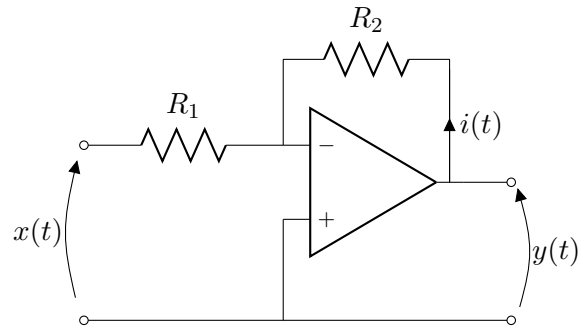


Figure 2.6: Inverting amplifier

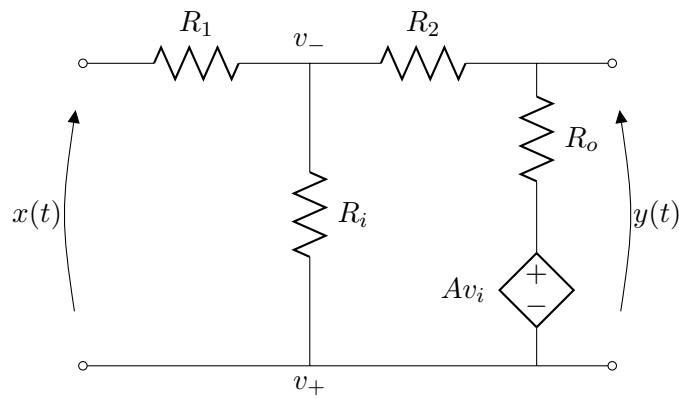


Figure 2.7: An equivalent circuit for the inverting amplifier from Figure 2.6 using the model for an operational amplifier in Figure 2.5. The symbol $v_i = v_+ - v_-$ is the voltage over resistor R_i .

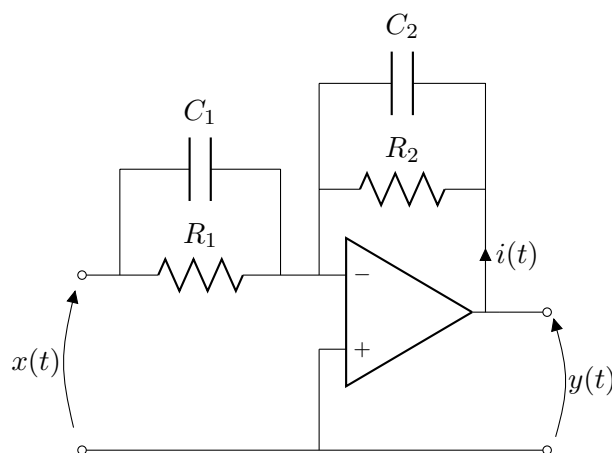


Figure 2.8: Operational amplifier configured with two capacitors and two resistors.

Test 2 (Inverting amplifier) In this test we construct the inverting amplifier circuit from Figure 2.6 with $R_2 \approx 22\text{k}\Omega$ and $R_1 \approx 12\text{k}\Omega$ that are accurate to within 5% of these values. The operational amplifier used is the Texas Instruments LM358P. Using a computer soundcard (an approximation of) the voltage signal

$$x(t) = \frac{1}{3} \sin(2\pi f_1 t) + \frac{1}{3} \sin(2\pi f_2 t)$$

with $f_1 = 100$ and $f_2 = 233$ is passed through the circuit. As in previous tests, the soundcard is used to sample the input signal x and the output signal y . Approximate reconstructions of the input signal \tilde{x} and output signal \tilde{y} are given according to (1.3.4) and (1.3.2). According to (2.1.1) we expect the approximate relationship

$$\tilde{y} \approx -\frac{R_2}{R_1} \tilde{x} = -\frac{11}{6} \tilde{x}.$$

Each of \tilde{y} , \tilde{x} and $-\frac{11}{6} \tilde{x}$ are plotted in Figure 2.9.

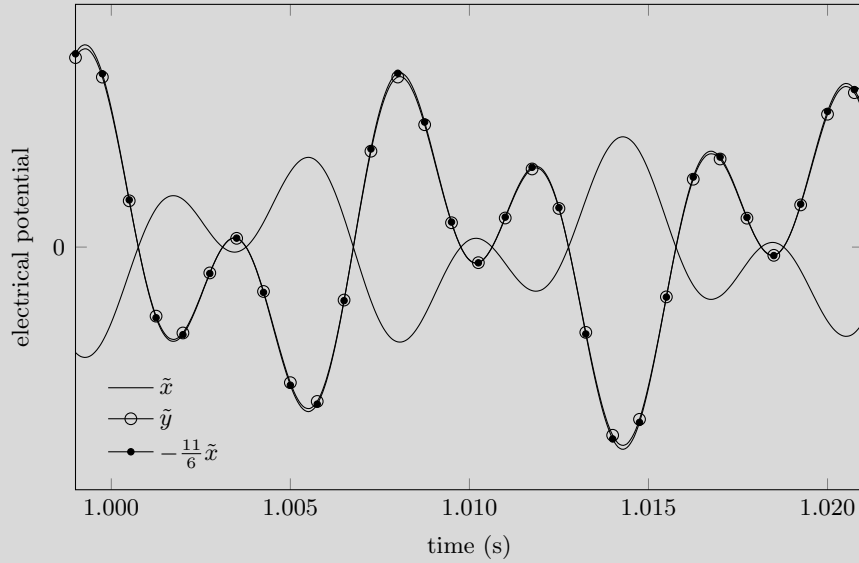


Figure 2.9: Plot of reconstructed input signal \tilde{x} (solid line), output signal \tilde{y} (solid line with circle) and hypothesised output signal $-\frac{11}{6} \tilde{x}$ (solid line with dot).

Test 3 (Active RC circuit) In this test we construct the circuit from Figure 2.8 with $R_1 \approx R_2 \approx 27\text{k}\Omega$ and $C_2 \approx 10\text{nF}$ accurate to within 5% of these values and $C_1 = 0$ (an open circuit). The operational amplifier used is a Texas Instruments LM358P. Using a computer soundcard (an approximation of) the voltage signal

$$x(t) = \frac{1}{3} \sin(2\pi f_1 t) + \frac{1}{3} \sin(2\pi f_2 t)$$

with $f_1 = 500$ and $f_2 = 1333$ is passed through the circuit. As in previous tests, the soundcard is used to sample the input signal x and the output signal y and approximate reconstructions \tilde{x} and \tilde{y} are given according to (1.3.4) and (1.3.2). According to (2.2.4) we expect the approximate relationship

$$\tilde{x} \approx -\frac{R_1}{R_2} \tilde{y} - R_1 C D(\tilde{y}) = -\tilde{y} - \frac{27}{10^5} D(\tilde{y}).$$

The derivative of the sinc function is

$$D \text{sinc}(t) = \frac{d}{dt} \text{sinc}(t) = \frac{1}{\pi t^2} (\pi t \cos(\pi t) - \sin(\pi t)), \quad (2.2.5)$$

and so,

$$D\tilde{y}(t) = \frac{d}{dt} \left(\sum_{\ell=1}^L y_\ell \text{sinc}(Ft - \ell) \right) = F \sum_{\ell=1}^L y_\ell D \text{sinc}(Ft - \ell). \quad (2.2.6)$$

Each of \tilde{y} , \tilde{x} and $-\tilde{y} - \frac{27}{10^5} D\tilde{y}$ are plotted in Figure 2.9.

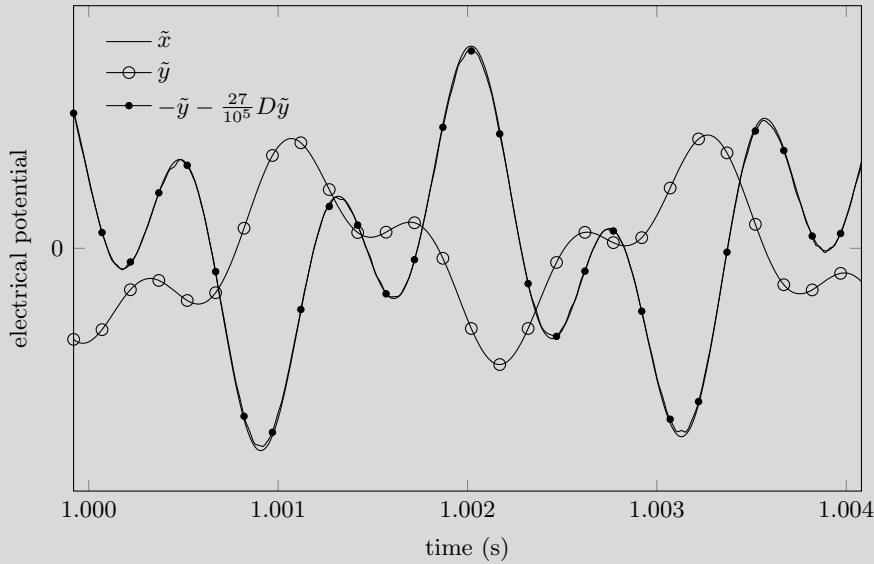


Figure 2.9: Plot of reconstructed input signal \tilde{x} (solid line with circle), output signal \tilde{y} (solid line), and hypothesised input signal $-\tilde{y} - \frac{27}{10^5} D\tilde{y}$ (solid line with dot).

Put

$$K_i = \frac{1}{R_1 C_2}, \quad K_p = \frac{R_2}{R_1} + \frac{C_1}{C_2}, \quad K_d = R_2 C_1$$

and now

$$Dy = -K_i x - K_p Dx - K_d D^2 x. \quad (2.2.7)$$

This equation models what is called a **proportional-integral-derivative controller** or **PID controller**. The coefficients K_i, K_p and K_d are called the **integral gain**, **proportional gain**, and **derivative gain**.

The final active circuit we consider is called a **Sallen-Key** [Sallen and Key, 1955] and is depicted in Figure 2.10. Observe that the output of the amplifier is connected directly to the inverting input and is also connected to the noninverting input by a capacitor and resistor. This circuit has both negative *and* positive feedback. It is not immediately apparent that we can use the simplifying assumptions for an ideal operational amplifier with negative feedback. However, we will do so and will find that it works in this case.

Let v_{R1}, v_{R2}, v_{C1} , and v_{C2} be the voltages over the components R_1, R_2, C_1 , and C_2 . Kirchoff's voltage law leads to the equations

$$x = v_{R1} + v_{R2} + v_{C2}, \quad y = v_{C1} + v_{R2} + v_{C2}.$$

The voltage at the inverting and noninverting terminals is y and so the voltage over the capacitor C_2 is y , that is, $y = v_{C2}$. Using this, the equations above simplify to

$$x = v_{R1} + v_{R2} + y, \quad v_{C1} = -v_{R2}.$$

The current i_2 through capacitor C_2 satisfies $i_2 = C_2 Dv_{C2} = C_2 Dy$. Because no current flows into the inverting terminal of the amplifier the current through R_2 is also i_2 and so $v_{R2} = R_2 i_2 = R_2 C_2 Dy$. Substituting this into the equations above gives

$$x = v_{R1} + R_2 C_2 Dy + y, \quad v_{C1} = -R_2 C_2 Dy. \quad (2.2.8)$$

Kirchoff's current law asserts that $i + i_1 = i_2$. The current i through capacitor C_1 satisfies $i = C_1 Dv_{C1} = -R_2 C_1 C_2 D^2 y$ and the current through resistor R_1 satisfies

$$v_{R1} = R_1 i_1 = R_1 (i_2 - i) = R_1 C_2 Dy + R_1 R_2 C_1 C_2 D^2 y.$$

Substituting this into the equation on the left of (2.2.8) gives

$$x = y + C_2 (R_1 + R_2) Dy + R_1 R_2 C_1 C_2 D^2 y. \quad (2.2.9)$$

The Sallen-Key will be useful when we consider the design of analogue electrical filters in Section 5.2.

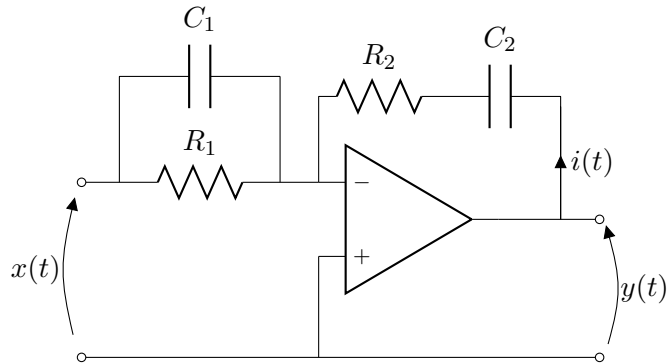


Figure 2.9: Operational amplifier implementing a **proportional-integral-derivative controller**.

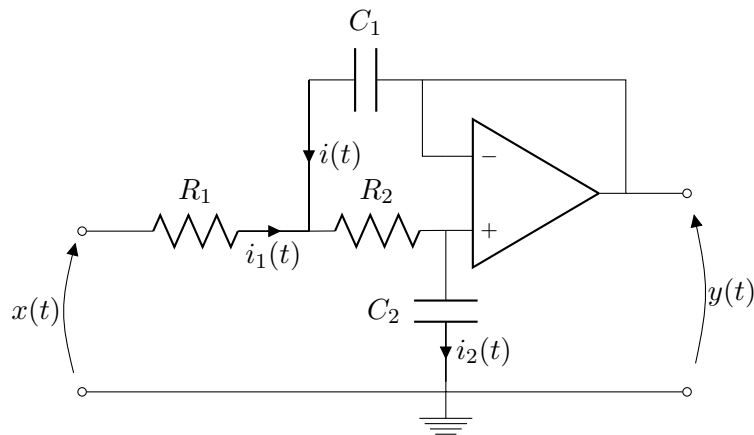


Figure 2.10: Operational amplifier implementing a **Sallen-Key**.

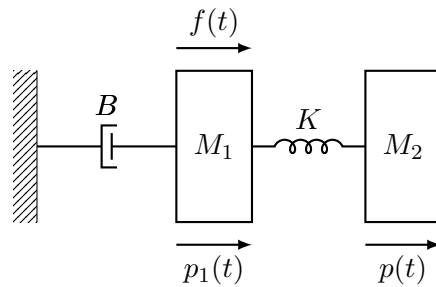


Figure 2.11: Two masses, a spring and a damper

2.3 Masses, springs, and dampers

A mechanical mass-spring-damper system was described in Section 2 and Figure 2.2. We now consider another mechanical system depicted in Figure 2.11 involving two masses, a spring and a damper. A mass M_1 is connected to a wall by a damper with constant B , and to another mass M_2 by a spring with constant K . A force represented by the signal f is applied to the first mass. We will derive a differential equation relating f with the position p of the second mass. Assume that the spring applies no force (is in equilibrium) when the masses are distance d apart. The forces due to the spring satisfy

$$f_{s1} = -f_{s2} = K(p - p_1 - d)$$

where f_{s1} and f_{s2} are signals representing the force due to the spring on mass M_1 and M_2 respectively. It is convenient to define the signal $g = p_1 + d$ so that forces due to spring satisfy the simpler equation

$$f_{s1} = -f_{s2} = K(p - g).$$

The only force applied to M_2 is by the spring and so, by Newton's law, the acceleration of M_2 satisfies

$$M_2 D^2 p = f_{s2} = Kg - Kp. \quad (2.3.1)$$

The force applied by the damper on mass M_1 is given by the signal

$$f_d = -BDp_1 = -BDg$$

where the replacement of p_1 by g is justified because differentiation will remove the constant d . The cumulative force on M_1 is given by the signal

$$f_1 = f + f_d + f_{s1} = f + f_d - f_{s2} = f - M_2 D^2 p - BDg$$

and by Newton's law the acceleration of M_1 satisfies

$$M_1 D^2 p_1 = M_1 D^2 g = f_1 = f - M_2 D^2 p - BDg.$$

Combining this equation with (2.3.1) we obtain a fourth order differential equation relating the position p and force f ,

$$f = BDp + (M_1 + M_2)D^2 p + \frac{BM_2}{K}D^3 p + \frac{M_1 M_2}{K}D^4 p. \quad (2.3.2)$$

Given the position of the second mass p we can readily solve for the corresponding force f and position of the first mass p_1 . For example, if the constants $B = K = 1$ and $M_1 = M_2 = \frac{1}{2}$ and $d = \frac{5}{2}$, and if the position of the second mass satisfies

$$p(t) = e^{-t^2}$$

then, by application of (2.3.2) and (2.3.1),

$$f(t) = e^{-t^2}(1 + 4t - 8t^2 - 4t^3 + 4t^4), \quad \text{and} \quad p_1(t) = 2e^{-t^2}t^2 - \frac{5}{2}.$$

This solution is plotted in Figure 2.12.

Figure 2.12: Solution of the system describing two masses with a spring and damper where $B = K = 1$ and $M_1 = M_2 = \frac{1}{2}$ and the position of the second mass is $p(t) = e^{-t^2}$.

2.4 Direct current motors

Direct current (DC) motors convert electrical energy, in the form of a voltage, into rotary kinetic energy [Nise, 2007, page 76]. We derive a differential equation relating the input voltage v to the angular position of the motor θ . Figure 2.13 depicts the components of a DC motor.

The voltages over the resistor and inductor satisfy

$$v_R = Ri, \quad v_L = LDi,$$

and the motion of the motor induces a voltage called the **back electromotive force** (EMF),

$$v_b = K_b D\theta$$

that we model as being proportional to the angular velocity of the motor. The input voltage now satisfies

$$v = v_R + v_L + v_b = Ri + LDi + K_b D\theta.$$

The torque τ applied by the motor is modelled as being proportional to the current i ,

$$\tau = K_\tau i.$$

A load with inertia J is attached to the motor. Two forces are assumed to act on the load, the torque τ applied by the current, and a torque $\tau_d = -BD\theta$

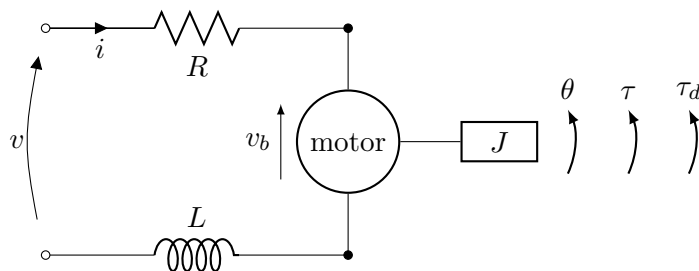


Figure 2.13: Diagram for a rotary direct current (DC) motor

modelling a damper that acts proportionally against the angular velocity of the motor. By Newton's law, the angular acceleration of the load satisfies

$$JD^2\theta = \tau + \tau_d = K_\tau i - BD\theta.$$

Combining these equations we obtain the 3rd order differential equation

$$v = \left(\frac{RB}{K_\tau} + K_b \right) D\theta + \frac{RJ + LB}{K_\tau} D^2\theta + \frac{LJ}{K_\tau} D^3\theta$$

relating voltage and motor position. In many DC motors the inductance L is small and can be ignored, leaving the simpler second order equation

$$v = \left(\frac{RB}{K_\tau} + K_b \right) D\theta + \frac{RJ}{K_\tau} D^2\theta. \quad (2.4.1)$$

Given the position signal θ we can find the corresponding voltage signal v . For example, put the constants $K_b = K_\tau = B = R = J = 1$ and assume that

$$\theta(t) = 2\pi(1 + \text{erf}(t))$$

where $\text{erf}(t) = \frac{2}{\pi} \int_{-\infty}^t e^{-\tau^2} d\tau$ is the **error function**. The corresponding angular velocity $D\theta$ and voltage v satisfy

$$D\theta(t) = 4\sqrt{\pi}e^{-t^2}, \quad v(t) = 8\sqrt{\pi}e^{-t^2}(1 - t).$$

These signals are depicted in Figure 2.14. This voltage signal is sufficient to make the motor perform two revolutions and then come to rest.

Exercises

- 2.1. Analyse the inverting amplifier circuit in Figure 2.7 to obtain the relationship between input voltage x and output voltage y given by (2.2.1). You may wish to use a symbolic programming language (for example Maxima, Sage, Mathematica, or Maple).

Figure 2.14: Voltage and corresponding angle for a DC motor with constants $K_b = K_\tau = B = R = J = 1$.

- 2.2. Figure 2.15 depicts a mechanical system involving two masses, two springs, and a damper connected between two walls. Suppose that the spring K_2 is at rest when the mass M_2 is at position $p(t) = 0$. A force, represented by the signal f , is applied to mass M_1 . Derive a differential equation relating the force f and the position p of mass M_2 .
- 2.3. Consider the electromechanical system in Figure 2.16. A direct current motor is connected to a potentiometer in such a way that the voltage at the output of the potentiometer is equal to the angle of the motor θ . This voltage is fed back to the input terminal of the motor. An input voltage v is applied to the other terminal on the motor. Find the differential equation relating v and θ . What is the input voltage v if the motor angle satisfies $\theta(t) = \frac{\pi}{2}(1 + \text{erf}(t))$? Plot θ and v in this case when the motor coefficients satisfy $L = 0$, $R = \frac{3}{4}$, and $K_b = K_\tau = B = J = 1$.

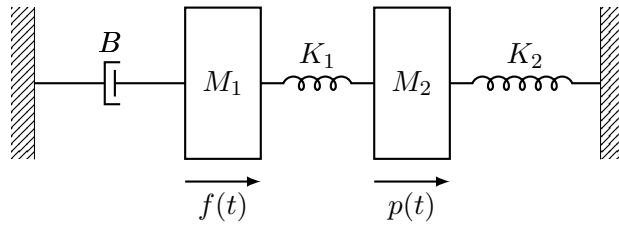


Figure 2.15: Two masses, a spring, and a damper connect between two walls for Exercise 2.2.

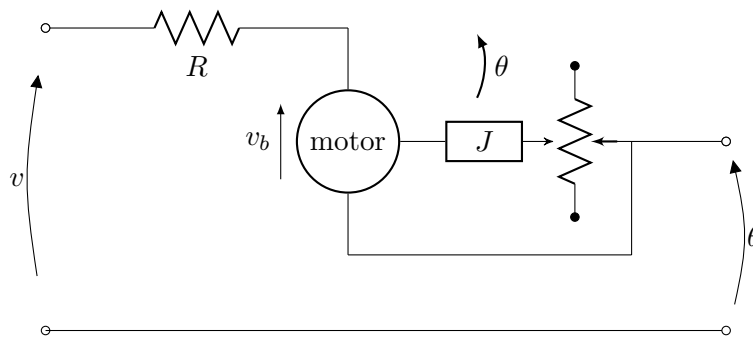


Figure 2.16: Diagram for a rotary direct current (DC) with potentiometer feedback for Exercise 2.3.

Chapter 3

Linear shift-invariant systems

In the previous section we derived differential equations that model mechanical, electrical, and electro-mechanical systems. The equations themselves often do not provide sufficient information. For example, we were able to find a signal p representing the position of the mass-spring-damper in Figure 2.2 given a particular force signal f is applied to the mass. However, it is not immediately obvious how to find the force signal f given a particular position signal p . We will be able to solve this problem and, more generally, to describe properties of systems modelled by linear differential equations with constant coefficient, if we make the added assumptions that the systems are **linear** and **shift-invariant**. We study linear shift-invariant systems in this chapter. Throughout this chapter H will denote a linear shift-invariant system.

3.1 Convolution, regular systems and the delta “function”

A large number of linear shift-invariant systems can be represented by a signal called the **impulse response**. The impulse response of a system H is a locally integrable signal h such that

$$Hx(t) = \int_{-\infty}^{\infty} h(\tau)x(t - \tau)d\tau,$$

that is, the response of H to input signal x can be represented as an integral equation involving x and the impulse response h . The integral is called a **convolution** and appears so often that a special notation is used for it. We write $h * x$ to indicate the signal that results from convolution of signals h and x , that is, $h * x$ is the signal

$$h * x = \int_{-\infty}^{\infty} h(\tau)x(t - \tau)d\tau.$$

Those systems that have an impulse response we call **regular systems**¹. Unless otherwise stated the domain of the regular system H is assumed to be the set of signals x such that the integral

$$\int_{-\infty}^{\infty} |h(\tau)x(t-\tau)| d\tau < \infty \quad \text{for all } t \in \mathbb{R}.$$

This set of signals is denoted by $\text{dom}(h)$ and so $H \in \text{dom}(h) \rightarrow \mathbb{R} \rightarrow \mathbb{C}$. Observe that for all $x \in \text{dom}(h)$ the response

$$Hx(t) = \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau \leq \int_{-\infty}^{\infty} |h(\tau)x(t-\tau)| d\tau < \infty$$

for all $t \in \mathbb{R}$. It can be shown that $\text{dom}(h)$ is a linear shift-invariance space (Exercise 3.1).

Regular systems are linear because, for all $x, y \in \text{dom}(h)$ and all $a, b \in \mathbb{C}$,

$$\begin{aligned} H(ax + by) &= h * (ax + by) \\ &= \int_{-\infty}^{\infty} h(\tau)(ax(t-\tau) + by(t-\tau))d\tau \\ &= a \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau + b \int_{-\infty}^{\infty} h(\tau)y(t-\tau)d\tau \quad (3.1.1) \\ &= a(h * x) + b(h * y) \\ &= aHx + bHy. \end{aligned}$$

The above equation shows that convolution commutes with scalar multiplication and distributes with addition, that is,

$$h * (ax + by) = a(h * x) + b(h * y).$$

Regular systems are also shift-invariant because for all $x \in \text{dom}(h)$

$$\begin{aligned} T_{\kappa}Hx &= T_{\kappa}(h * x) \\ &= \int_{-\infty}^{\infty} h(\tau)x(t-\kappa-\tau)d\tau \\ &= \int_{-\infty}^{\infty} h(\tau)T_{\kappa}x(t-\tau)d\tau \\ &= h * (T_{\kappa}x) \\ &= HT_{\kappa}x. \end{aligned}$$

The impulse response of a regular system H can be found in the following way. First define the signal

$$p_{\gamma}(t) = \begin{cases} \gamma, & 0 < t \leq \frac{1}{\gamma} \\ 0, & \text{otherwise,} \end{cases}$$

¹The name **regular system** is motivated by the term **regular distribution** [Zem-
nian, 1965]

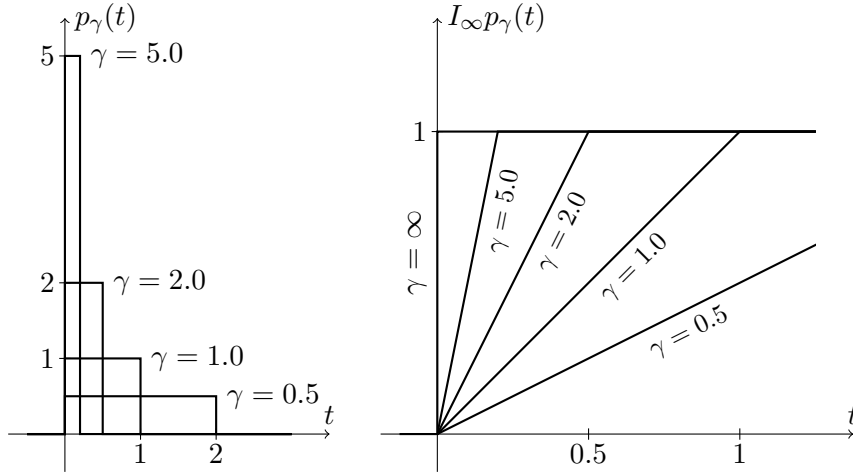


Figure 3.1: The rectangular shaped pulse p_γ for $\gamma = 0.5, 1, 2, 5$ and the response of the integrator I_∞ to p_γ for $\gamma = 0.5, 1, 2, 5, \infty$.

that is, a rectangular shaped pulse of height γ and width $\frac{1}{\gamma}$. The signal p_γ is plotted in Figure 3.1 for $\gamma = \frac{1}{2}, 1, 2, 5$. As γ increases the pulse gets thinner and higher so as to keep the area under p_γ equal to one. Consider the response of the regular system H to the signal p_γ ,

$$Hp_\gamma = h * p_\gamma = \int_{-\infty}^{\infty} h(\tau)p_\gamma(t - \tau)d\tau = \gamma \int_{t-1/\gamma}^t h(\tau)d\tau.$$

Taking $\gamma \rightarrow \infty$ we find that

$$\lim_{\gamma \rightarrow \infty} Hp_\gamma = \lim_{\gamma \rightarrow \infty} \gamma \int_{t-1/\gamma}^t h(\tau)d\tau = h \text{ a.e..}$$

As an example, consider the integrator system I_∞ described in Section 1.4. This response of I_∞ to p_γ is

$$I_\infty p_\gamma(t) = \int_{-\infty}^t p_\gamma(\tau)d\tau = \begin{cases} 0, & t \leq 0 \\ \gamma t, & 0 < t \leq \frac{1}{\gamma} \\ 1, & t > \frac{1}{\gamma} \end{cases}$$

The response is plotted in Figure 3.1. Taking the limit as $\gamma \rightarrow \infty$ we find that the impulse response of the integrator is the step function

$$u(t) = \lim_{\gamma \rightarrow \infty} I_\infty p_\gamma(t) = \begin{cases} 0 & t \leq 0 \\ 1 & t > 0. \end{cases} \quad \text{a.e.} \quad (3.1.2)$$

Some important systems do not have an impulse response and are not regular. For example, the identity system T_0 is not regular. Similarly, the

shifter T_τ and differentiators D^k are not regular. However, it is common to pretend that T_0 *does* have an impulse response and this is typically denoted by the symbol δ called the **delta function**. The idea is to assign δ the property

$$\int_{-\infty}^{\infty} x(t)\delta(t)dt = x(0)$$

so that convolution of x and δ satisfies

$$\delta * x = \int_{-\infty}^{\infty} \delta(\tau)x(t-\tau)d\tau = x(t) = T_0x.$$

We now treat δ as if it were a signal. So $\delta(t-\tau)$ will represent the impulse response of the shifter T_τ because

$$\begin{aligned} T_\tau x &= \delta(t-\tau) * x \\ &= \int_{-\infty}^{\infty} \delta(\kappa-\tau)x(t-\kappa)d\kappa \\ &= \int_{-\infty}^{\infty} \delta(k)x(t-\tau-k)dk \quad (\text{change variable } k = \kappa - \tau) \\ &= x(t-\tau). \end{aligned}$$

For $a \in \mathbb{R}$ it is common to plot $a\delta(t-\tau)$ using an arrow of height a at $t = \tau$ as indicated in Figure 3.2. It is important to realise that δ is not actually a signal. It is not a function. However, it can be convenient to treat δ as if it were a function. The manipulations in the last set of equations, such as the change of variables, are not formally justified, but they do lead to the desired result $T_\tau x = x(t-\tau)$ in this case. In general, there is no guarantee that mechanical mathematical manipulations involving δ will lead to sensible results.

The only other non regular systems that we have use of are differentiators D^k and it is common to define a similar notation for pretending that these systems have an impulse response. In this case, the symbol δ^k is assigned the property

$$\int_{-\infty}^{\infty} x(t)\delta^k(t)dt = D^k x(0),$$

so that convolution of x and δ^k is

$$\delta^k * x = \int_{-\infty}^{\infty} \delta^k(\tau)x(t-\tau)d\tau = D^k x(t).$$

As with the delta function the symbol δ^k must be treated with care. This notation can be useful, but purely formal manipulations with δ^k may not always lead to sensible results.

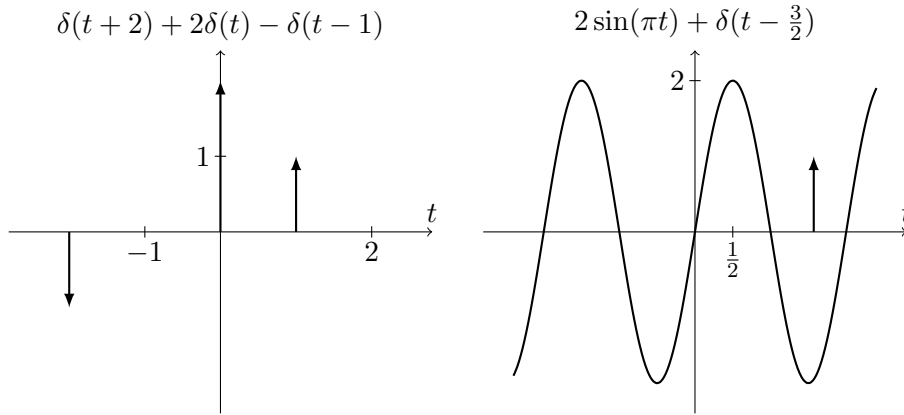


Figure 3.2: Plot of the “signal” $\delta(t+2) + 2\delta(t) - \delta(t-1)$ (left) and the “signal” $2\sin(\pi t) + \delta(t - \frac{3}{2})$ (right).

The impulse response h immediately yields some properties of the corresponding system H . For example, if $h(t) = 0$ for all $t < 0$, then H is causal because

$$Hx(t) = \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau = \int_0^{\infty} h(\tau)x(t-\tau)d\tau$$

only depends on values of the input signal x at $t-\tau$ with $\tau > 0$. The system H is stable if and only if h is absolutely integrable (Exercise 3.5).

Another related important signal is the **step response** defined as the response of the system to the step function u . For example, the step response of the shifter T_τ is the shifted step function $T_\tau u(t) = u(t-\tau)$. The step response of the integrator I_∞ is

$$I_\infty u(t) = \int_{-\infty}^t u(\tau)d\tau = \begin{cases} \int_0^t d\tau = t & t > 0 \\ 0 & t \leq 0. \end{cases}$$

This signal is often called the **ramp function**. Not all systems have a step response. For example, the system with impulse response $u(-t)$ does not because convolution of the step $u(t)$ and its reflection $u(-t)$ is not possible. If a system H has both an impulse response h and a step response Hu , then these two signals are related. To see this, observe that the step response is

$$Hu = h * u = \int_{-\infty}^{\infty} h(\tau)u(t-\tau)d\tau = \int_{-\infty}^t h(\tau)d\tau = I_\infty h. \quad (3.1.3)$$

Thus, the step response can be obtained by applying the integrator I_∞ to the impulse response in the case that both of these signals exist.

3.2 Properties of convolution

The convolution $x * y$ of two signals x and y does not always exist. For example, if $x(t) = u(t)$ and $y(t) = 1$, then

$$x * y = \int_{-\infty}^{\infty} x(\tau)y(t - \tau)d\tau = \int_{-\infty}^{\infty} u(\tau)d\tau = \int_0^{\infty} d\tau$$

is not finite for any t . We cannot convolve the step function u and the signal that takes the constant value 1. On the other hand, if $x(t) = y(t) = u(t)$, then

$$x * y = \int_{-\infty}^{\infty} u(\tau)u(t - \tau)d\tau = \begin{cases} \int_0^t d\tau = \tau & t > 0 \\ 0 & t \leq 0 \end{cases}$$

if finite for all t .

We have already shown in (3.1.1) that convolution commutes with scalar multiplication and is distributive with addition, that is, for signals x, y, w and complex numbers a, b ,

$$a(x * w) + b(y * w) = (ax + by) * w.$$

Convolution is commutative, that is, $x * y = y * x$ whenever these convolutions exist. To see this, write

$$\begin{aligned} x * y &= \int_{-\infty}^{\infty} x(\tau)y(t - \tau)d\tau \\ &= \int_{-\infty}^{\infty} x(t - \kappa)y(\kappa)d\kappa \quad (\text{change variable } \kappa = t - \tau) \\ &= y * x. \end{aligned}$$

Convolution is also associative, that is, for signals x, y, z ,

$$(x * y) * z = x * (y * z). \quad (\text{Exercise 3.4})$$

By combining the associative and commutative properties we find that the order in which the convolutions in $x * y * z$ are performed does not matter, that is

$$x * y * z = y * z * x = z * x * y = y * x * z = x * z * y = z * y * x$$

provided that all the convolutions involved exist. More generally, the order in which any sequence of convolutions is performed does not change the final result.

3.3 Linear combination and composition

Let $H_1 \in X \rightarrow Y$ and $H_2 \in X \rightarrow Y$ be linear shift-invariant systems with domain $X \subseteq \mathbb{R} \rightarrow \mathbb{C}$ and range $Y \subseteq \mathbb{R} \rightarrow \mathbb{C}$. For complex numbers c and d , let $H \in X \rightarrow Y$ be the system satisfying

$$Hx = cH_1x + dH_2x.$$

The system H is said to be a **linear combination** of H_1 and H_2 . The system H is linear because for all signals $x, y \in X$ and $a, b \in \mathbb{C}$,

$$\begin{aligned} H(ax + by) &= cH_1(ax + by) + dH_2(ax + by) \\ &= acH_1x + bcH_1y + adH_2x + bdH_2y \quad (\text{linearity } H_1, H_2) \\ &= a(cH_1x + dH_2x) + b(cH_1y + dH_2y) \\ &= aHx + bHy. \end{aligned}$$

The system H is also shift-invariant because

$$\begin{aligned} HT_\tau x &= cH_1T_\tau x + dH_2T_\tau x \\ &= cT_\tau H_1x + dT_\tau H_2x \quad (\text{shift-invariance } H_1, H_2) \\ &= T_\tau(cH_1x + dH_2x) \quad (\text{linearity } T_\tau) \\ &= T_\tau Hx. \end{aligned}$$

So, linear shift-invariant systems can be constructed as linear combinations of other linear shift-invariant systems. If H_1 and H_2 are regular systems with impulse responses h_1 and h_2 then

$$\begin{aligned} Hx &= aH_1x + bH_2x \\ &= ah_1 * x + bh_2 * x \\ &= (ah_1 + bh_2) * x \quad (\text{distributivity of convolution}) \\ &= h * x, \end{aligned}$$

and so, H is a regular system with impulse response $h = ah_1 + bh_2$.

Another way to construct linear shift-invariant systems is by **composition**. Let X, Y, Z be linear shift-invariant spaces of signals. Let $H_1 \in X \rightarrow Y$ and $H_2 \in Y \rightarrow Z$ be linear shift-invariant systems and let $H \in X \rightarrow Z$ be the system satisfying

$$Hx = H_2H_1x,$$

that is, H first applies H_1 and then applies H_2 . The system H is said to be the **composition** of H_1 and H_2 . The system H is linear because, for signals $x, y \in X$ and complex numbers a, b ,

$$\begin{aligned} H(ax + by) &= H_2H_1(ax + by) \\ &= H_2(aH_1x + bH_1y) \quad (\text{linearity } H_1) \\ &= aH_2H_1x + bH_2H_1y \quad (\text{linearity } H_2) \\ &= aHx + bHy. \end{aligned}$$

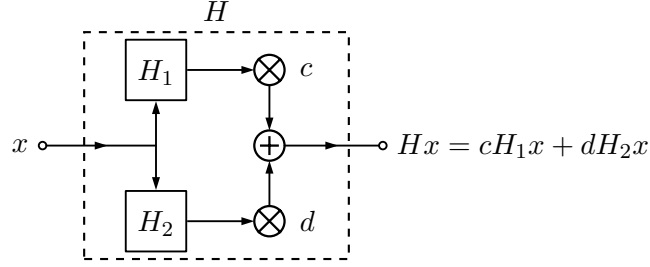


Figure 3.3: Block diagram depicting the linear combination of linear shift-invariant systems. The system $cH_1x + dH_2x$ can be expressed as a single linear shift-invariant system Hx .

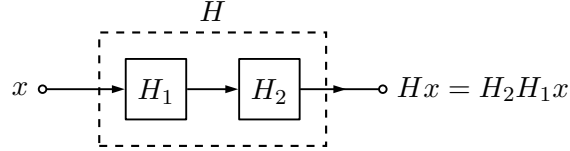


Figure 3.4: Block diagram depicting composition of linear shift-invariant systems. The system H_2H_1 can be expressed as a single linear shift-invariant system H .

The system is also shift-invariant because

$$\begin{aligned}
 HT_\tau x &= H_2H_1T_\tau x \\
 &= H_2T_\tau H_1x && \text{(shift-invariance } H_1) \\
 &= T_\tau H_2H_1x && \text{(shift-invariance } H_2) \\
 &= T_\tau Hx.
 \end{aligned}$$

If H_1 and H_2 are regular systems the composition property can be expressed using their impulse responses h_1 and h_2 . It follows that

$$\begin{aligned}
 Hx &= H_2H_1x \\
 &= h_2 * (h_1 * x) \\
 &= (h_2 * h_1) * x && \text{(associativity of convolution)} \\
 &= h * x,
 \end{aligned}$$

and so, H is a regular system with impulse response $h = h_2 * h_1$.

A wide variety of linear shift-invariant systems can be constructed by linear combination and composition of simpler systems.

3.4 Eigenfunctions and the transfer function

Let $s = \sigma + j\omega \in \mathbb{C}$ where $j = \sqrt{-1}$. A signal of the form

$$e^{st} = e^{\sigma t} e^{j\omega t} = e^{\sigma t} (\cos(\omega t) + j \sin(\omega t))$$

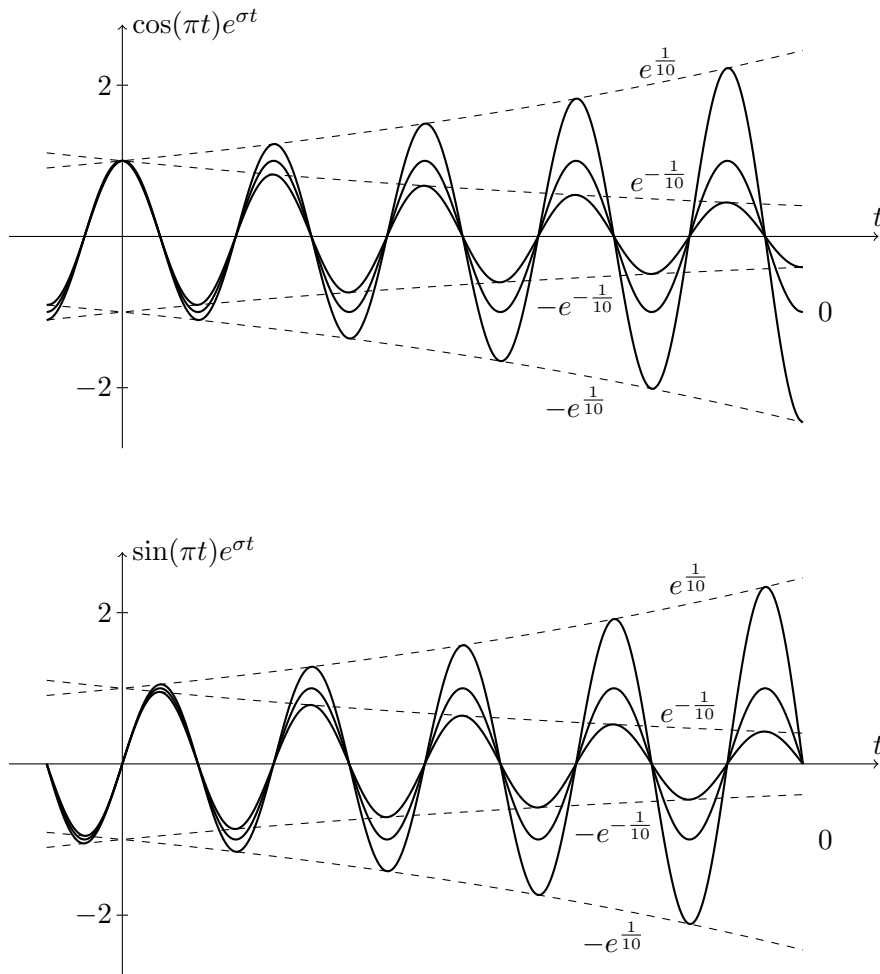


Figure 3.5: The function $\cos(\pi t)e^{\sigma t}$ (top) and $\sin(\pi t)e^{\sigma t}$ (bottom) for $\sigma = -\frac{1}{10}, 0, \frac{1}{10}$.

is called a **complex exponential signal**. Complex exponential signals play an important role in the study of linear shift-invariant systems. The real and imaginary parts of $e^{(\sigma+j\pi)t}$ are plotted in Figure 3.5 for $\sigma = -\frac{1}{10}, 0, \frac{1}{10}$. The signal is oscillatory when $\omega \neq 0$. The signal converges to zero as $t \rightarrow \infty$ when $\sigma < 0$ and diverges as $t \rightarrow \infty$ when $\sigma > 0$.

Let $H \in X \rightarrow Y$ be a linear shift-invariant system. Suppose that $y = He^{st}$ is the response of H to the complex exponential signal $e^{st} \in X$. Consider the response of H to the shifted signal $T_\tau e^{st} = e^{s(t-\tau)}$ for $\tau \in \mathbb{R}$. By shift-invariance

$$HT_\tau e^{st} = T_\tau He^{st} = y(t - \tau)$$

and by linearity

$$He^{s(t-\tau)} = e^{-s\tau} He^{st} = e^{-s\tau} y(t).$$

Combining these equations we obtain

$$y(t - \tau) = e^{-s\tau} y(t) \quad \text{for all } t, \tau \in \mathbb{R}.$$

This equation is satisfied by signals of the form $y(t) = \lambda e^{st}$ where λ is a complex number. That is, the response of a linear shift-invariant system H to a complex exponential signal e^{st} is the same signal e^{st} multiplied by some constant complex number λ . Due to this property complex exponential signals are called **eigenfunctions** of linear shift-invariant systems. The constant λ does not depend on t , but it does usually depend on the complex number s and the system H . To highlight this dependence on H and s we write $\lambda H(s)$ or $\lambda(H)(s)$ or $\lambda(H, s)$ and do not distinguish between these notations. Considered as a function of s , the expression λH is called the **transfer function** of the system H . Observe that the transfer function λH maps a complex number to a complex number. Letting S be the set of complex number s such $e^{st} \in X$ we have $\lambda H \in S \rightarrow \mathbb{C}$. The transfer function satisfies

$$He^{st} = \lambda H(s) e^{st} \tag{3.4.1}$$

whenever $e^{st} \in X$. This is an important equation. Stated in words: the response of a linear shift-invariant system $H \in X \rightarrow Y$ to a complex exponential signal $e^{st} \in X$ is the transfer function λH evaluated at s , that is, $\lambda H(s)$, multiplied by the same complex exponential signal e^{st} .

We can use these eigenfunctions to better understand the properties of systems modelled by differential equations, such as those in Section 2. Consider the active electrical circuit from Figure 2.8. In the case that the resistors $R_1 = R_2$, and the capacitor $C_1 = 0$ (an open circuit) the differential equation relating the input voltage x and output voltage y is

$$x = -y - R_1 C_2 D y.$$

We called this the **active RC** circuit. To simplify notation put $R = R_1$ and $C = C_2$ so that $x = -y - RCD(y)$. Suppose that H is a linear shift-invariant system that maps the input voltage x to the output voltage y ,

that is, H is a system that describes the active RC circuit. If the input voltage is a complex exponential signal $x = e^{st}$, then the output voltage is the same complex exponential signal multiplied by the transfer function, that is, $y = Hx = \lambda H(s)e^{st}$. Substituting this into the differential equation for the active RC circuit we obtain

$$e^{st} = -\lambda e^{st} - RCD(\lambda e^{st}) = -\lambda e^{st}(1 - RCs)$$

where, to simplify notation, we have written simply λ for $\lambda H(s)$ above. Solving for λ to obtain the transfer function of the system H describing the active RC circuit

$$\lambda H(s) = -\frac{1}{1 + RCs}. \quad (3.4.2)$$

3.5 The spectrum

It is of interest to focus on the transfer function when s is purely imaginary, that is, when $s = j\omega$ with $\omega \in \mathbb{R}$. In this case the complex exponential signal takes the form

$$e^{j\omega t} = e^{j2\pi f t} = \cos(2\pi f t) + j \sin(2\pi f t)$$

where $\omega = 2\pi f$. This signal is oscillatory when $f \neq 0$ and does not decay or explode as $|t| \rightarrow \infty$. Let $H \in X \rightarrow Y$ be a linear shift-invariant system with domain X containing the signal $e^{j2\pi f t}$. We denote by ΛH the function satisfying

$$\Lambda H(f) = \lambda H(j2\pi f)$$

called the **spectrum** of H . It follows from (3.4.1) that the response of H to $e^{j2\pi f t}$ satisfies

$$He^{j2\pi f t} = \lambda H(j2\pi f)e^{j2\pi f t} = \Lambda H(f)e^{j2\pi f t}$$

whenever $e^{j2\pi f t} \in X$. It is of interest to consider the **magnitude spectrum** $|\Lambda H(f)|$ and the **phase spectrum** $\angle \Lambda H(f)$ separately. The notation \angle denotes the **argument** (or **phase**) of a complex number. We have,

$$\Lambda H(f) = |\Lambda H(f)| e^{j\angle \Lambda H(f)}$$

and correspondingly

$$He^{j2\pi f t} = |\Lambda H(f)| e^{j(2\pi f t + \angle \Lambda H(f))}.$$

By taking real and imaginary parts we obtain the pair of real valued solutions

$$H \cos(2\pi f t) = |\Lambda H(f)| \cos(2\pi f t + \angle \Lambda H(f)),$$

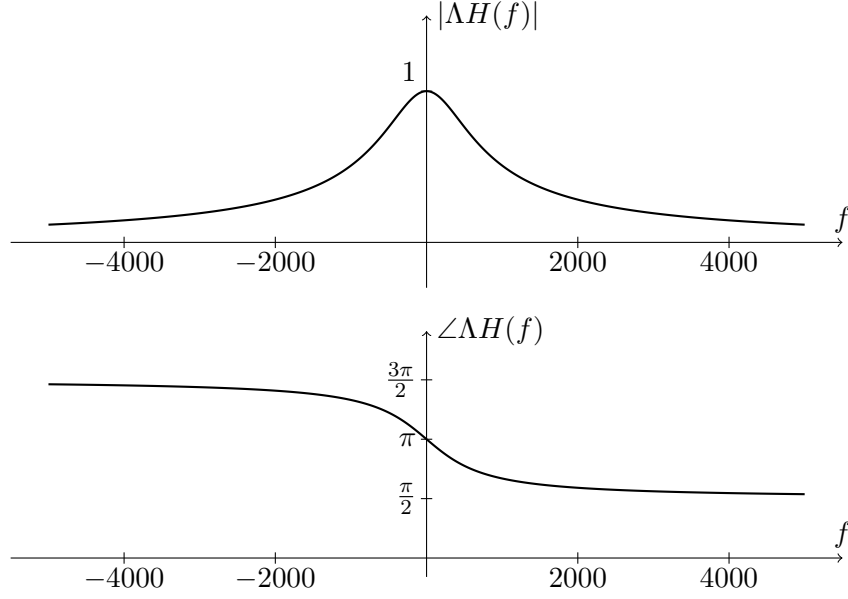


Figure 3.6: Magnitude spectrum (top) and phase spectrum (bottom) of the active RC circuit with $R = 27 \times 10^3$ and $C = 10 \times 10^{-9}$.

$$H \sin(2\pi ft) = |\Lambda H(f)| \sin(2\pi ft + \angle \Lambda H(f)). \quad (3.5.1)$$

Consider again the active RC circuit with H the system mapping the input voltage x to the output voltage y . According to (3.4.2) the spectrum of H is

$$\Lambda H(f) = -\frac{1}{1 + 2\pi RC f j}. \quad (3.5.2)$$

The magnitude and phase spectrum is

$$|\Lambda H(f)| = (1 + 4\pi^2 R^2 C^2 f^2)^{-\frac{1}{2}}, \quad \angle \Lambda H(f) = \pi - \text{atan}(2\pi RC f).$$

These plotted in Figure 3.6 when $R = 27 \times 10^3$ and $C = 10 \times 10^{-9}$. Observe from the plot of the magnitude spectrum that a low frequency sinusoidal signal, say 100Hz or less, input to the active RC circuit results in a sinusoidal output signal with the same frequency and approximately the same amplitude. However, a high frequency sinusoidal signal, say greater than 1000Hz, input to the circuit results in a sinusoidal output signal with the same frequency, but smaller amplitude. For this reason RC circuits are called **low pass filters**.

Test 4 (Spectrum of the active RC circuit) We test the hypothesis that the active RC circuit satisfies (3.5.1). To do this sinusoidal signals at

varying frequencies of the form

$$x_k(t) = \sin(2\pi f_k t), \quad f_k = \left\lceil 110 \times 2^{k/2} \right\rceil, \quad k = 0, 1, \dots, 12$$

are input to the active RC circuit constructed as in Test 3 with $R = R_1 = 27\text{k}\Omega$ and $C = C_2 = 10\text{nF}$. The notation $\lceil \cdot \rceil$ denotes rounding to the nearest integer with half integers rounded up. In view of (3.5.1) the expected output signals are of the form

$$y_k(t) = |\Lambda H(f_k)| \sin(2\pi f_k t + \angle \Lambda H(f_k)), \quad k = 0, 1, \dots, 12.$$

This equality can also be shown directly using the differential equation for the active RC circuit.

Using the soundcard each signal x_k is played for a period of approximately 1 second and approximately $F = 44100$ samples are obtained. On the soundcard hardware used for this test samples near the beginning and end of playback are distorted. This appears to be an unavoidable feature of the soundcard. To alleviate this we discard the first 10^4 samples and use only the $L = 8820$ samples that follow (corresponding to 200ms of signal). After this process we have samples $x_{k,1}, \dots, x_{k,L}$ and $y_{k,1}, \dots, y_{k,L}$ of the input and output signals corresponding with the k th signal x_k . The samples are expected to take the form

$$x_{k,\ell} \approx x_k(P\ell - \tau) = \rho \sin(2\pi f_k P\ell - \theta)$$

and

$$y_{k,\ell} \approx y_k(\ell P - \tau) = |\Lambda H(f_k)| \rho \sin(2\pi f_k P\ell - \theta + \angle \Lambda H(f_k))$$

where $P = \frac{1}{F}$ is the sample period, the positive real number ρ corresponds with the gain on the input and output of the soundcard, and $\theta = 2\pi f_k \tau$ corresponds with delays caused by discarding the first 10^4 samples and also unavoidable delays that occur when starting soundcard playback and recording.

We will not measure the gain ρ nor the delay θ , but will be able to test the properties of the circuit without knowledge of these. To simplify notation put $\gamma = 2\pi f_k P$. From the samples of the input signal $x_{k,1}, \dots, x_{k,L}$ compute the complex number

$$A = \frac{2j}{L} \sum_{\ell=1}^L x_{k,\ell} e^{-j\gamma\ell} \approx \frac{2j}{L} \sum_{\ell=1}^L \rho \sin(\gamma\ell - \theta) e^{-j\gamma\ell} = \alpha + \alpha^* C$$

where $\alpha = \rho e^{-j\theta}$ and α^* denotes the complex conjugate of α and

$$C = e^{-\gamma(L+1)} \frac{\sin(\gamma L)}{L \sin(\gamma)} \quad (\text{Excercise 3.8}).$$

Similarly, from the samples of the output signal $y_{k,1}, \dots, y_{k,L}$ we compute the complex number

$$B = \frac{2j}{L} \sum_{\ell=1}^L y_{k,\ell} e^{-j\gamma\ell} \approx \beta + \beta^* C$$

where $\beta = \rho e^{-j\theta} \Lambda H(f_k) = \alpha \Lambda H(f_k)$. Now compute the quotient

$$Q_k = \frac{B - B^* C}{A - A^* C} \approx \frac{\beta(1 + |C|^2)}{\alpha(1 + |C|^2)} = \frac{\beta}{\alpha} = \Lambda H(f_k).$$

Thus, we expect the quotient Q_k to be close to the spectrum of the active RC circuit evaluated at frequency f_k . We will test this hypothesis by observing the magnitude and phase of Q_k individually, that is, we will test the expected relationships

$$|Q_k| \approx |\Lambda H(f_k)| = \sqrt{\frac{1}{1 + 4\pi^2 R^2 C^2 f_k^2}}$$

and

$$\angle Q_k \approx \angle \Lambda H(f_k) = \pi - \text{atan}(2\pi RC f_k)$$

for each $k = 0, \dots, 12$. Figure 3.7 plots the hypothesised magnitude and phase spectrum alongside the measurements Q_k for $k = 0, \dots, 12$.

Exercises

- 3.1. Let h be a locally integrable signal. Show that the set $\text{dom}(h)$ defined in Section 3.1 on page 33 is a linear shift-invariant space.
- 3.2. Show that $\text{dom}(u)$ where u is the step function is the subset of locally integrable signals such that $\int_{-\infty}^0 |x(t)| dt < \infty$.
- 3.3. Show that convolution distributes with addition and commutes with scalar multiplication, that is, show that $a(x*w) + b(y*w) = (ax + by)*w$.
- 3.4. Show that convolution is associative. That is, if x, y, z are signals then $x * (y * z) = (x * y) * z$.
- 3.5. Show that a regular system is stable if and only if its impulse response is absolutely integrable.
- 3.6. Show that the system $H(x) = \int_{-1}^1 \sin(\pi\tau) x(t + \tau) d\tau$ is linear time invariant and regular. Find and sketch the impulse response and the step response.

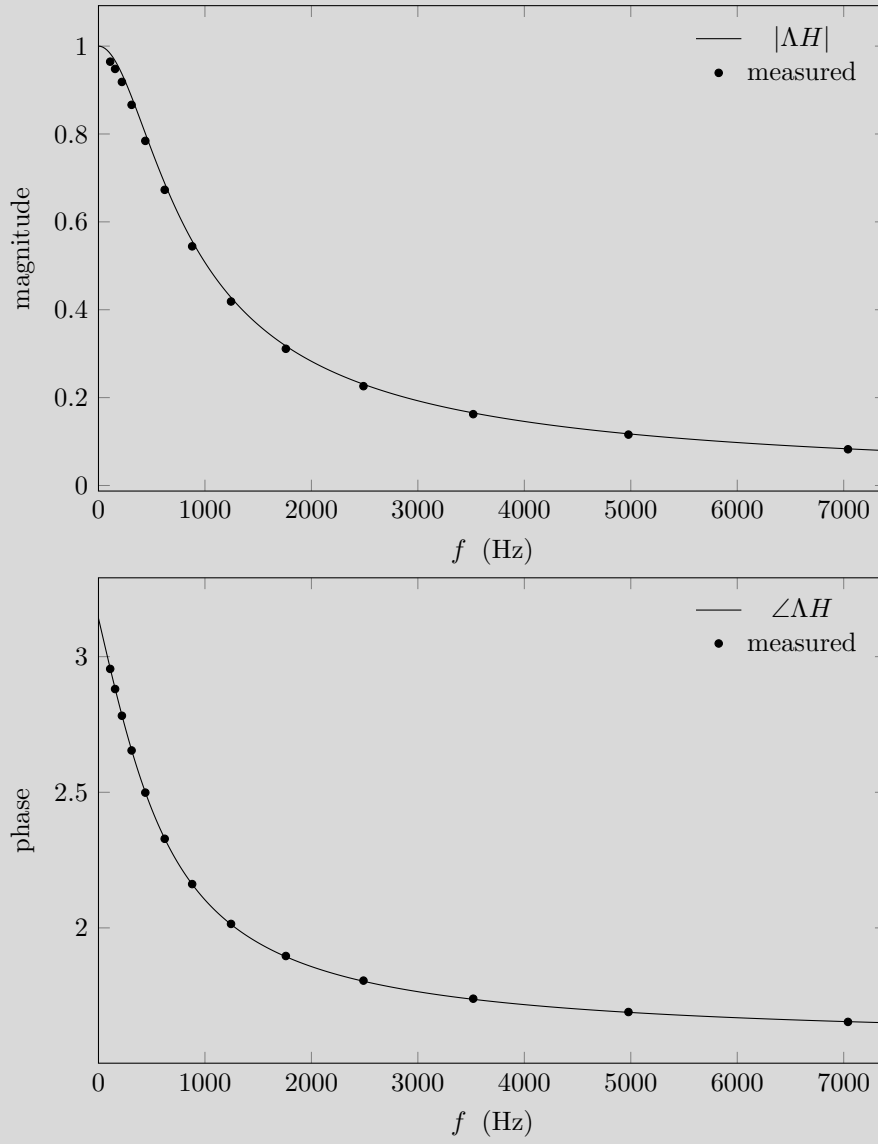


Figure 3.7: Hypothesised magnitude spectrum $|\Lambda H(f)|$ (top) and phase spectrum $\angle \Lambda H(f)$ (bottom) and the measured magnitude and phase spectrum $|Q_k|$ and $\angle Q_k$ for $k = 0, \dots, 12$ (dots).

3.7. Show that $\sum_{\ell=1}^L e^{\beta\ell} = \frac{e^{\beta(L+1)} - e^{\beta}}{e^{\beta} - 1}$ (Hint: sum a geometric progression).

3.8. Show that

$$\frac{2j}{L} \sum_{\ell=1}^L \sin(\gamma\ell - \theta) e^{-j\gamma\ell} = \alpha + \alpha^* C$$

where $\alpha = e^{-j\theta}$ and $C = e^{-j\gamma(L+1)} \frac{\sin(\gamma L)}{L \sin(\gamma)}$. (Hint: solve Exercise 3.7 first and then use the formula $2j \sin(x) = e^{jx} - e^{-jx}$).

3.9. State whether each of the following systems are: causal, linear, shift-invariant, or stable. Plot the impulse and step response of the systems whenever they exist.

(a) $Hx(t) = 3x(t-1) - 2x(t+1)$

(b) $Hx(t) = \sin(2\pi x(t))$

(c) $Hx(t) = t^2 x(t)$

(d) $Hx(t) = \int_{-1/2}^{1/2} \cos(\pi\tau) x(t+\tau) d\tau$

Bibliography

- Bluestein, L. I. [1968]. A linear filtering approach to the computation of the discrete Fourier transform. In *Northeast Electronics Research and Engineering Meeting Record*, volume 10, pp. 218–219.
- Butterworth, S. [1930]. On the theory of filter amplifiers. *Experimental Wireless and the Wireless Engineer*, pp. 536–541.
- Cooley, J. W. and Tukey, J. W. [1965]. An Algorithm for the Machine Calculation of Complex Fourier Series. *Mathematics of Computation*, 19(90), 297–301.
- Fine, B. and Rosenberger, G. [1997]. *The Fundamental Theorem of Algebra*. Undergraduate Texts in Mathematics. Springer-Verlag, Berlin.
- Frigo, M. and Johnson, S. G. [2005]. The Design and Implementation of FFTW3. *Proceedings of the IEEE*, 93(2), 216–231.
- Graham, R. L., Knuth, D. E. and Patashnik, O. [1994]. *Concrete Mathematics: A Foundation for Computer Science*. Addison-Wesley, Reading, MA, 2nd edition.
- Nicholas, C. B. and Yates, R. C. [1950]. The Probability Integral. *Amer. Math. Monthly*, 57, 412–413.
- Nise, N. S. [2007]. *Control systems engineering*. Wiley, 5th edition.
- Oppenheim, A. V., Willsky, A. S. and Nawab, S. H. [1996]. *Signals and Systems*. Prentice Hall, 2nd edition.
- Papoulis, A. [1977]. *Signal analysis*. McGraw-Hill.
- Proakis, J. G. [2007]. *Digital communications*. McGraw-Hill, 5th edition.
- Quinn, B. G. and Hannan, E. J. [2001]. *The Estimation and Tracking of Frequency*. Cambridge University Press, New York.
- Quinn, B. G., McKilliam, R. G. and Clarkson, I. V. L. [2008]. Maximizing the Periodogram. In *IEEE Global Communications Conference*, pp. 1–5.

- Rader, C. M. [1968]. Discrete Fourier transforms when the number of data samples is prime. *Proceedings of the IEEE*, 56, 1107–1108.
- Rudin, W. [1986]. *Real and complex analysis*. McGraw-Hill.
- Sallen, R. and Key, E. [1955]. A practical method of designing RC active filters. *Circuit Theory, IRE Transactions on*, 2(1), 74–85.
- Soliman, S. S. and Srinath, M. D. [1990]. *Continuous and discrete signals and systems*. Prentice-Hall Information and Systems Series. Prentice-Hall.
- Stewart, I. and Tall, D. O. [2004]. *Complex Analysis*. Cambridge University Press.
- Vetterli, M., Kovačević, J. and Goyal, V. K. [2014]. *Foundations of Signal Processing*. Cambridge University Press, 1st edition.
- Zemanian, A. H. [1965]. *Distribution theory and transform analysis*. Dover books on mathematics. Dover.