# The DSP Cookbook

## Pelle Juul Christensen

### October 2017

## Contents

| 1        | The  | ory                                      | 2  |
|----------|------|--|----|
|          | 1.1  | Sine Wave                                | 2  |
|          | 1.2  | Phasors                                  | 3  |
|          | 1.3  | Digital Signals                          | 3  |
|          | 1.4  | DSP Systems                              | 4  |
|          | 1.5  | Causal Systems                           | 4  |
|          | 1.6  | Difference Equations                     | 4  |
|          | 1.7  | Linear Time-Invariant System             | 4  |
|          | 1.8  | Impulse Response                         | 5  |
|          | 1.9  | Convolution                              | 6  |
|          | 1.10 | Digital Frequency                        | 6  |
|          | 1.11 | Frequency Response                       | 6  |
|          |      | Z-transform                              | 8  |
| <b>2</b> | Rec  | ipes                                     | 9  |
|          | 2.1  | Echo and Delay                           | 9  |
|          | 2.2  | Chorus                                   | 10 |
|          | 2.3  | Flanging                                 | 10 |
|          | 2.4  | Equal Power Panning and Fading           | 10 |
|          | 2.5  | Ring Modulation and Amplitude Modulation | 11 |
|          | 2.6  | Biquad Filters                           | 11 |
|          |      | 2.6.1 Low-pass Filter                    | 11 |
|          |      | 2.6.2 High-pass Filter                   | 12 |
| 3        | Refe | erences                                  | 12 |

### 1 Theory

#### 1.1 Sine Wave

The general equation of is

$$A\sin(2\pi ft + \theta) \tag{1}$$

where

A is the amplitude of the wave.

sin is the sin function.

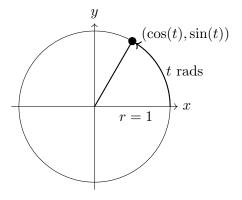
 $2\pi$  normalizes the period of the sin function to 1.

t is time.

f is the frequency of the wave.

 $\theta$  is the initial phase of the wave.

The sin and cos functions are defined using the unit circle. Given a variable t we go t radians along the unit circle  $^1$ . The point at t radians will have coordinates  $(\cos(t), \sin(t)) - \cos(t)$  is the projection of the point onto the x-axis,  $\sin(t)$  is the projection onto the y-axis.



Sinusoids have a period of  $2\pi$  (a full arc), thus if we multiply t by  $2\pi$  to get  $\sin(2\pi t)$ , we will get a period of 1. If t is time in seconds, we will have one period per second.

<sup>&</sup>lt;sup>1</sup>A unit circle is simply a circle with a radius of 1.

To get more or less periods per second we multiply t by that amount. For example  $\sin(2\pi 440t)$  has 440 periods per second.

Periods per second is measured in Hertz which has the unit 1/s, which is also called the frequency of a wave. To calculate the period of a sine wave at a specific frequency f, you use 1/f, e.g. a sine wave at 440 Hz will have a period of  $1/f = 0.002 \ s$ .

We can change the starting point of a wave by adding an initial phase. A sin wave with a phase of  $\pi/2$  is equivalent to a cos wave. The initial phase is usually indicated by  $\theta$ .

A sine wave swings between -1 and 1. To increase or decrease the amplitude, we multiply with a gain factor A.

#### 1.2Phasors

A *Phasor* is a complex sinusoid defined my Euler's equation

$$e^{jx} = \cos(x) + j\sin(x) \tag{2}$$

such that we have

$$\cos(x) = \operatorname{Re}\left\{e^{jx}\right\}$$

$$\sin(x) = \operatorname{Im}\left\{e^{jx}\right\}$$
(4)

$$\sin(x) = \operatorname{Im}\left\{e^{jx}\right\} \tag{4}$$

#### **Digital Signals** 1.3

Unlike real-world analog signals, which are continuous, digital signals are represented by discrete numbers at specific points in time – a list of numbers. Each number is separated by some amount of time called the sampling period.

If we wish to convert a analog signal to a digital one, we must sample it. This is done by measuring the analog signal one time each sampling period (1/samplerate). Given the analog signal x(t) then we get the digital signal

$$x[n] = x(nT) \tag{5}$$

where T is the sampling period and n is the sample index. Notice that it is convention to use square brackets for discrete signals and parentheses for continuous signals.

If we are measuring a real-world signal, we will use an Analog-to-Digital Converter (ADC). Since computers only have a finite amount of memory, we will need to choose how precise our measurements will be. If we represent out sample with b bits then we will have  $2^b$  levels to work with. The

process of rounding our sample to the nearest representable number is called quantization.

#### 1.4 DSP Systems

A DSP system (or just "a system") is a computer program – an algorithm – which takes one (or more) digital signals and returns one (or more) digital signals.

#### 1.5 Causal Systems

In systems with memory, we need to agree on what should happen when n < 0. Usually we will use x[n] = 0 for n < 0. Such a system is said to be causal.

If we have a system where  $x[n] \neq 0$  for n < 0, then the system is said to be non-causal.

#### 1.6 Difference Equations

The general form of a difference equation is

$$y[n] = a_1 y[n-1] + a_2 y[n-2] + \dots + a_N y[n-N] + b_0 x[n] + b_1 x[n-1] + b_2 x[n-2] + \dots + b_L [n-L]$$
(6)

Also written as

$$y[n] = \sum_{k=1}^{N} a_k y[n-k] + \sum_{k=0}^{L} b_k x[n-k]$$
 (7)

Where the  $y[\cdot]$  components determine the recursive (feedback) characteristic, and the  $x[\cdot]$  components the non-recursive characteristic.

### 1.7 Linear Time-Invariant System

A system T(x[n]) is linear time-invariant (LTI) if it satisfies both linearity and time-invariancy properties.

A system is linear if it supports the scalability property

$$T(\alpha x[n]) = \alpha T(x[n]). \tag{8}$$

As well as the superposition property

$$T(x_3[n]) = T(x_1[n] + x_2[n]) = T(x_1[n]) + T(x_2[n])$$
(9)

where  $x_3[n] = x_1[n] + x_2[n]$ .

A system is time-invariant if it meets the condition

$$x[n-L] = y[n-L] \tag{10}$$

for some delay L — a delay in the input signal will cause a corresponding delay in the output signal.

A system is LTI if all of its subcomponents are LTIs.

#### 1.8 Impulse Response

An impulse is a short spike in amplitude. The ideal impulse signal is defined by

$$\delta[n] = \begin{cases} 1, & \text{if } n = 0\\ 0, & \text{otherwise} \end{cases}$$
 (11)

The *impulse response* of a system is defined as the system output for an input of  $x[n] = \delta[n]$ . In this case the output of the system is normally denoted h[n].

LTI systems can be described completely by their impulse responses. To obtain the output of a system based on an impulse response, one should convolve the input signal with the impulse response

$$T(x[n]) = x[n] * h_T[n]$$
(12)

A system has a *finite impulse response* (FIR) if it does not contain any recursive components. In that case the impulse response is determined by

$$h[n] = T(\delta[n]) = \sum_{k=0}^{L} b_k \delta[n-k]$$
(13)

The impulse response of FIR systems will always converge to zero.

A system has an *infinite impulse reponse* (IIR) if it contains recursive components. In that case the impulse response is determined by

$$h[n] = T(\delta[n]) = \sum_{k=1}^{N} a_k h[n-k] + \sum_{k=0}^{L} b_k \delta[n-k]$$
 (14)

Some IIR systems will converge to a state of near-zero. Such systems are said to be *stable*. IIR systems can also be *unstable*, which means that their output will be divergent (i.e. will blow up to infinity).

#### 1.9 Convolution

The *convolution* of two signals x[n] and h[n] is defined as

$$x[n] * h[n] = \sum_{m=-\infty}^{m=\infty} x[n-m]h[m]$$
 (15)

In practice, m will be bounded by the sizes of the signals. If x[n] has size N and h[n] size M, then m should be within 0 and M, since no samples is defined before 0, and no samples in h[x] are defined after M. The equation is then

$$x[n] * h[n] = \sum_{m=0}^{M} x[n-m]h[m]$$
 (16)

The result will be defined for n in  $0 \le n < N + M - 1$ .

Convolution is commutative, meaning that

$$x[n] * h[n] = h[n] * x[n]$$

$$\tag{17}$$

It is also associative, which means that

$$g[x] * (x[n] * h[n]) = (g[x] * x[n]) * h[n]$$
(18)

as well as distributive

$$q[x] * (x[n] + h[n]) = (q[x] * x[n]) + (q[x] * h[n])$$
(19)

#### 1.10 Digital Frequency

Digital frequency is defined as

$$\Omega = 2\pi f T,\tag{20}$$

where f is the analog frequency in Hz and T is the sampling period. This means that  $\Omega=0$  denotes DC,  $\Omega=\pi$  the Nyquist frequency (1/2\*samplerate), and  $\Omega=2\pi$  the sampling frequency.

#### 1.11 Frequency Response

The frequency response of a system describes the frequency content of its impulse response in terms of magnitude and phase. The frequency response of a system is usually shown as a plot with frequency on the x-axis and

amplitude on the y-axis. If the plot is presented with a logarithmic frequency scale, it is sometimes called a *Bode* diagram.

Using the equation

$$H(e^{j\theta}) = \frac{\sum_{k=0}^{L} b_k e^{-j\theta k}}{\sum_{k=1}^{N} a_k e^{-j\theta k}}$$
(21)

we get a complex number which describes the phase and magnitude response for a signal at the digital frequency  $\theta$ . The magnitude  $||H(e^{j\theta})||$  is the gain at that frequency. The angle  $\angle H(e^{j\theta})$  is the phase shift at that frequency.

The  $H(e^{j\theta})$  term is derived by first selecting the input signal

$$x[n] = e^{j\theta n} \tag{22}$$

and then convolving with the impulse response of our system

$$y[n] = \sum_{m=-\infty}^{\infty} h[m]x[n-m] = \sum_{m=-\infty}^{\infty} h[m]e^{j\theta(n-m)}$$
 (23)

$$=e^{j\theta(n)}\sum_{m=-\infty}^{\infty}h[m]e^{j\theta m}=x[n]\sum_{m=-\infty}^{\infty}h[m]e^{j\theta m} \qquad (24)$$

$$=x[n]H(e^{j\theta}) \tag{25}$$

so we have

$$H(e^{j\theta}) = \sum_{m=-\infty}^{\infty} h[m]e^{j\theta m}$$
 (26)

and since

$$y[n] = x[n]H(e^{j\theta}) \tag{27}$$

we have

$$H(e^{j\theta}) = \frac{y[n]}{x[n]} \tag{28}$$

To get to our final definition of  $H(e^{j\theta})$  we first not that

$$y[n] = \sum_{m = -\infty}^{\infty} h[m]x[n - m]$$
(29)

$$y[n] = x[n]H(e^{j\theta}) = e^{j\theta n}H(e^{j\theta})$$
(30)

And then rearranging

$$e^{j\theta n}H(e^{j\theta}) = \sum_{k=1}^{N} a_k e^{j\theta(n-k)}H(e^{j\theta}) + \sum_{k=0}^{L} b_k e^{j\theta(n-k)}$$
 (31)

$$e^{j\theta n}H(e^{j\theta}) = e^{j\theta n} \left( \sum_{k=1}^{N} a_k e^{-j\theta k} H(e^{j\theta}) + \sum_{k=0}^{L} b_k e^{-j\theta k} \right)$$
(32)

$$H(e^{j\theta}) = \sum_{k=1}^{N} a_k e^{-j\theta k} H(e^{j\theta}) + \sum_{k=0}^{L} b_k e^{-j\theta k}$$
 (33)

$$1 = \sum_{k=1}^{n} a_k e^{-j\theta k} + \frac{\sum_{k=0}^{L} b_k e^{-j\theta k}}{H(e^{j\theta})}$$
 (34)

$$1 - \sum_{k=1}^{n} a_k e^{-j\theta k} = \frac{\sum_{k=0}^{L} b_k e^{-j\theta k}}{H(e^{j\theta})}$$
 (35)

$$\frac{1 - \sum_{k=1}^{n} a_k e^{-j\theta k}}{\sum_{k=0}^{L} b_k e^{-j\theta k}} = \frac{1}{H(e^{j\theta})}$$
(36)

$$H(e^{j\theta}) = \frac{\sum_{k=0}^{L} b_k e^{-j\theta k}}{1 - \sum_{k=1}^{n} a_k e^{-j\theta k}}$$
(37)

#### 1.12 Z-transform

The z-transform is a way of analysing the frequency response of a system. It is defined by

$$Z(x[n]) = X(z) = \sum_{n = -\infty}^{\infty} x[n]z^{-n},$$
(38)

and has the shifting property

$$Z(x[n-m]) = z^{-m} \sum_{n=-\infty}^{\infty} x[n]z^{-n} = z^{-m}X(z).$$
 (39)

If we apply the z-transform to the general difference equation we get

$$y[n] = \sum_{k=1}^{N} a_k y[n-k] + \sum_{k=0}^{L} b_k x[n-k]$$
 (40)

$$Y(z) = \sum_{k=1}^{N} a_k z^{-k} Y(z) + \sum_{k=0}^{L} b_k z^{-k} X(z)$$
(41)

(42)

from which we arrive at

$$\frac{Y(z)}{X(z)} = \frac{\sum_{k=0}^{L} b_k z^{-k}}{1 - \sum_{k=1}^{N} a_k z^{-k}},$$
(43)

which is called the *transfer function*. Note that setting  $z = e^{j\theta}$  makes the above equation equal to the frequency response (Eq. 37).

We use the above equation to solve for *poles* and *zeros* of the system. The zeros are found by solving the equation

$$\sum_{k=0}^{L} b_k z^{-k} = 0, (44)$$

which is the numerator of Eq. 43 — when this is zero, the system response will be zero.

To find the poles, we solve for the numerator like

$$1 - \sum_{k=1}^{N} a_k z^{-k} = 0, (45)$$

which gives us the places where the ratio explodes into infinity.

In reality the poles and zeros will cancel each other out by different amounts based on the input frequency. Therefore, the response will seldom actually be zero or infinity.

### 2 Recipes

#### 2.1 Echo and Delay

An *echo* effect, in its simplest form, is achieved by mixing a signal with a delayed version of itself

$$y[n] = x[n] + b_m x[n - m], (46)$$

where m is the delay amount in samples and  $b_m$  the amplitude of the echo.

The above equation will only give one reflection, if you want multiple reflections, you will need a feedback component:

$$y[n] = x[n] + a_m y[n - m], (47)$$

You can get creative with delay effects by adding more delay taps, or applying effects to the delayed components — a low-pass filter can be used to achieve a dampening effect.

#### 2.2 Chorus

The chorus effect is achieved by mixing a signal with a delayed version of itself, while varying the delay. The delay usually varies between 20 and 50 ms. This causes a doubling effect, like when multiple instruments are playing together.

The basic effect looks like

$$y[n] = x[n] + x[n - g[n]]$$
 (48)

where g[n] is the function determining the delay for sample n.

You can easily get creative with the chorus effect by using feedback instead of delay, and by adding multiple delays.

#### 2.3 Flanging

The flanger effect is achieved by mixing a signal with a delayed version of itself, while varying the delay. The delay is usually below 20 ms. This creates constructive and destructive interference at specific frequencies, resulting in a swooshing sound.

The basic effect looks like

$$y[n] = x[n] + x[n - g[n]]$$
(49)

where g[n] is the function determining the delay for sample n.

You can easily get creative with the flanger effect by using feedback instead of delay, and by adding multiple delays.

#### 2.4 Equal Power Panning and Fading

If you wish to do a fade between two signals or position a source in a stereo field, you cannot use a simple linear transition. Since humans perceive sound level in terms of *power*, we will need to pan with equal power, rather than equal amplitude. This is done by using the equation

$$y[n] = \sqrt{r} \cdot x_1[n] + \sqrt{1 - r} \cdot x_1[n],$$
 (50)

to fade between the signals  $x_1[n]$  and  $x_2[n]$  where r is the amount of fade, defined between 0 and 1.

To pan a signal between two outputs (one stereo channel) we use,

$$y_l[n] = \sqrt{r} \cdot x[n] \tag{51}$$

$$y_r[n] = \sqrt{1 - r} \cdot x[n], \tag{52}$$

where r = 0.5 is the center position.

#### 2.5 Ring Modulation and Amplitude Modulation

Ring modulation is a simple but powerful effect achieved by multiplying two signals,

$$y[n] = x_m[n] \cdot x_c[n], \tag{53}$$

where  $x_m[n]$  is the modulator signal,  $x_c[n]$  is the carrier signal.

Amplitude modulation (AM) is a related effect, achieved by mixing some of the modulated signal back into the output,

$$y[n] = A_m \cdot x_m[n] \cdot x_c[n] + A_c \cdot x_c[n], \tag{54}$$

where  $A_m$  is the amplitude of the modulated signal, and  $A_c$  is the amplitude of the carrier signal.

#### 2.6 Biquad Filters

Biquad filters are a general form of second-order two pole, two zeros filters. They have the difference equation

$$H(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2}}{1 + a_1 z_{-1} + a_2 z_{-2}}.$$
 (55)

If we multiply this by  $z^2/z^2$  we get

$$H(z) = \frac{b_0 z^2 + b_1 z + b_2}{z^2 + a_1 z + a_2 z},$$
(56)

and see that both the numerator and denominator are quadratic equations, hence the name.

Biquad filters can be used to implement many common types of filter including low-pass, high-pass, bandpass, band-reject and all-pass.

For calculating the coefficients of these filters we will need

$$K = \tan(\pi f_c/f_s),\tag{57}$$

where  $f_c$  is the filter center frequency and  $f_s$  is the sample rate.

#### 2.6.1 Low-pass Filter

Low-pass biquad filters are designed by selecting a cutoff frequency  $f_c$ , and a Q factor which controls the gain around  $f_c$ . When  $Q = 1/\sqrt{2}$  the filter will be flat around  $f_c$  — values above or below will cause amplification or attenuation.

The coefficients are then calculated by:

$$b_0 = \frac{K^2 Q}{K^2 Q + K + Q} \tag{58}$$

$$b_1 = \frac{2K^2Q}{K^2Q + K + Q} \tag{59}$$

$$b_2 = b_0 \tag{60}$$

$$a_1 = \frac{2Q(K^2 - 1)}{K^2Q + K + Q} \tag{61}$$

$$a_2 = \frac{K^2Q - K + Q}{K^2Q + K + Q} \tag{62}$$

[Zolzer, 2011]

#### 2.6.2 High-pass Filter

High-pass biquad filters are designed by selecting a cutoff frequency  $f_c$ , and a Q factor which controls the gain around  $f_c$ . When  $Q=1/\sqrt{2}$  the filter will be flat around  $f_c$ —values above or below will cause amplification or attenuation.

The coefficients are then calculated by:

$$b_0 = \frac{K^2 Q}{K^2 Q + K + Q} \tag{63}$$

$$b_1 = -\frac{2Q}{K^2 Q + K + Q} \tag{64}$$

$$b_2 = \frac{Q}{K^2 Q + K + Q} \tag{65}$$

$$a_1 = \frac{2Q(K^2 - 1)}{K^2Q + K + Q} \tag{66}$$

$$a_2 = \frac{K^2 Q - K + Q}{K^2 Q + K + Q} \tag{67}$$

[Zolzer, 2011]

### 3 References

### References

[Zolzer, 2011] Zolzer, U. (2011). DAFX: Digital Audio Effects. Wiley Publishing, 2nd edition.

### Index

ADC, 3 Panning, 10 AM, see Amplitude modulation period, 2 Amplitude, 3 Phase, 3 Amplitude modulation, 11 Phasor, 3 Poles, 9 Biquad filter, 11 Power, 10 Bode diagram, 6 Quantization, 3 Chorus effect, 10 Convolution, 6 Ring modulation, 11 Cos, 2 Sampling period, 3 Dampening, 9 Sampling rate, 3 Delay effect, 9 Second order filter, 11 Digital frequency, 6 Sin, 2 Digital signals, 3 Sine waves, 2 Stability, 5 Echo effect, 9 Stable, 5 Equal power panning, 10 Superposition property, 5 System, 4 Fading, 10 Finite impulse repsonse, 5 Time-invariant, 5 FIR, see Finite impulse repsonse Transfer function, 9 Flanging, 10 Frequency, 3 Unstable, 5 Frequency response, 6 Z-transform, 8 Hertz, 3 Zeros, 9 High-pass filter (biquad), 12 Hz, see Hertz IIR, see Infinite impulse response Impulse response, 5 Infinte impulse response, 5 Linear Time-Invariant System, 4 Linearity property, 4 Low-pass filter (biquad), 11

LTI, see Linear Time-Invariant Sys-

 $_{\rm tem}$