

# Discrete-Time Systems, Frequency Response

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## Exercise 1

- *Stable/Unstable: every bounded input produces a bounded output for all time in a stable system (BIBO), meaning that from  $-\infty$  to  $+\infty$  the amplitude of the signal has to be finite.*
- *Causal/Anti-Causal: the output depends only on the current and past inputs, not on future inputs. An anti-causal system is one where the output depends only on future inputs.*
- *Linear/Non-Linear: obeys the superposition and scalability or in other words called the Law of Additivity and Law of Homogeneity.*
- *Time-Variant/Time-Invariant: the system's response to inputs do not change with time (a shift in the input produces a shift in the output).*
- *Memoryless/Not-Memoryless: the output at a given time is dependent only on the input at that same time.*

## System 1:

$$y[n] = e^{nx[n]}$$

1. **Unstable:** increasing/decreasing the  $n$  value causes the output amplitude to increase until  $\pm\infty$ .
2. **Causal:** the output at any time  $n$  depends on current and past values of  $x[n]$ .

3. **Non-Linear:** the system does not obey the superposition property. For example, setting  $x[n] = x_1[n] + x_2[n]$  and plugging into  $y[n]$ , it is expected to get  $y_1[n] + y_2[n]$ . The derivation shows that 
$$e^{n(x_1[n] + x_2[n])} = e^{nx_1[n] + nx_2[n]} = e^{nx_1[n]} * e^{nx_2[n]} \neq y_1[n] + y_2[n].$$
4. **Time Variant:** shift the  $y[n]$  by  $n_0$  such that  $y[n - n_0] = e^{(n - n_0)x[n - n_0]}$ . Shift the input of to  $x$  such that  $x[n - n_0]$  and plug back into equation  $e^{nx[n - n_0]}$ . The two shifts are not equivalent and the system is therefore time variant.
5. **Memoryless:** the output at any time  $n$  depends only on the input at that same time  $n$ . For example, at  $n = 0$ ,  $y[1] = e^{1x[1]}$  the output depends on the present value of  $x[n]$ .

## System 2:

$$y[n] = x[n] - u[n + 1]$$

1. **Stable:** since  $u[n + 1]$  is bounded, if the input  $x[n]$  is bounded, the output will also be bounded.
2. **Anti-Causal:** output at time  $n$  depends on the future value at  $n + 1$ .
3. **Non-Linear:** the system does not obey the superposition property. For example, setting  $x[n] = x_1[n] + x_2[n]$  and plugging into  $y[n]$ , it is expected to get 
$$y_1[n] + y_2[n] = (x_1[n] - u[n + 1]) + (x_2[n] - u[n + 1]) = x_1[n] + x_2[n] - 2u[n + 1].$$
 The derivation shows that  $(x_1[n] + x_2[n]) - u[n + 1] = x_1[n] + x_2[n] - u[n + 1] \neq y_1[n] + y_2[n]$ .
4. **Time Variant:** shift the  $y[n]$  by  $n_0$  such that  $y[n - n_0] = x[n - n_0] - u[(n - n_0) + 1]$ . Shift the input of to  $x$  such that  $x[n - n_0]$  and plug back into equation  $x[n - n_0] - u[n + 1]$ . The two shifts are not equivalent and the system is therefore time variant.
5. **Memoryless:** the output at any time  $n$  depends only on a present value since for  $u[n + 1]$  it is known that  $u[n]$  will be 1 when  $n \geq 0$  and will be 0 for  $n < 0$ . Therefore,  $u[n - 1]$  will be 1 when  $n \geq -1$  such that  $y[n] = x[n] - 1$  and  $u[n - 1]$  will be 0 when  $n < -1$  such that  $y[n] = x[n] - 0$ .

## System 3:

$$y[n] = ax[n] + b$$

1. **Stable:** if  $a, b$  are finite and  $x[n]$  is bounded.
2. **Causal:** the output at any time  $n$  depends on current and past values of  $x[n]$ .
3. **Non-Linear:** the system does not obey the superposition property. For example, setting  $x[n] = x_1[n] + x_2[n]$  and plugging into  $y[n]$ , it is expected to get  $y_1[n] + y_2[n] = (ax_1[n] + b) + (ax_2[n] + b)$ . The derivation shows that  $a(x_1[n] + x_2[n]) + b = ax_1[n] + ax_2[n] + b \neq y_1[n] + y_2[n]$ . HOWEVER, the system can be **Linear** if the constant  $b = 0$ .
4. **Time Invariant:** the system is performing time shifting so it will most likely be time invariant as a result. Firstly, shift the  $y[n]$  by  $n_0$  such that  $y[n - n_0] = ax[n - n_0] + b$ . Secondly, shift the inputs to  $x$  such that  $x[n - n_0]$  and plug back into equation  $x[n - n_0] + b$ . The two are equivalent and the system is therefore time invariant.

5. **Memoryless:** the output at any time  $n$  depends only on the input at that same time  $n$ . For example, at  $n = 0$ ,  $y[0] = ax[0] + b$  the output depends on the present value of  $x[n]$ .

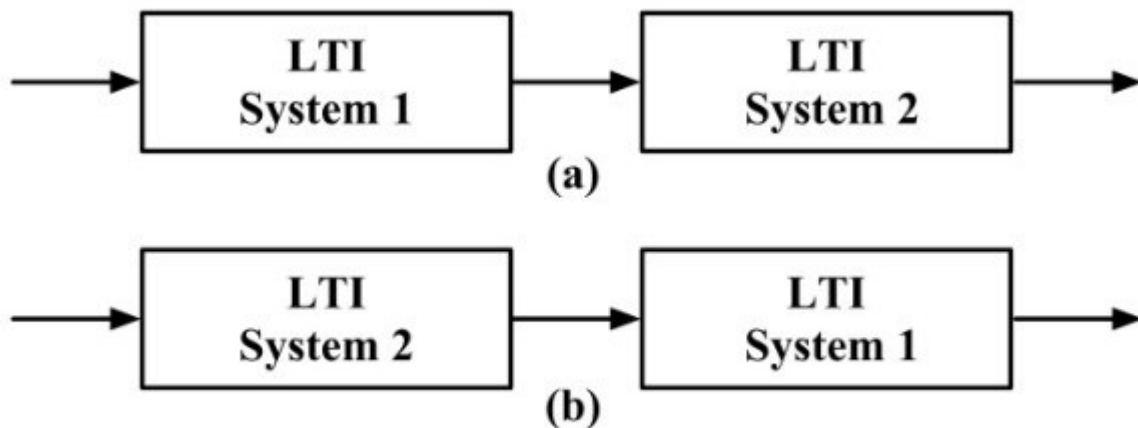
Sources used to complete this section:

- [https://www.youtube.com/watch?v=ngJ2QWvMPbl&ab\\_channel=NesoAcademy](https://www.youtube.com/watch?v=ngJ2QWvMPbl&ab_channel=NesoAcademy)
- [https://www.youtube.com/watch?v=ldwpoS7YeJA&ab\\_channel=NesoAcademy](https://www.youtube.com/watch?v=ldwpoS7YeJA&ab_channel=NesoAcademy)
- [tutorialspoint.com/signals-and-systems-stable-and-unstable-system](https://tutorialspoint.com/signals-and-systems-stable-and-unstable-system)
- <https://math.stackexchange.com/questions/3270918/the-system-yn-c-cdot-xn-d-is-given-is-the-system-linear-time-invariant>
- <https://www.tutorialspoint.com/signals-and-systems-causal-non-causal-and-anti-causal-signals>
- [https://www.youtube.com/watch?v=mqwUtn5cip8&ab\\_channel=NesoAcademy](https://www.youtube.com/watch?v=mqwUtn5cip8&ab_channel=NesoAcademy)
- [https://www.youtube.com/watch?v=wOQDGvCLOs8&ab\\_channel=NesoAcademy](https://www.youtube.com/watch?v=wOQDGvCLOs8&ab_channel=NesoAcademy)
- [https://www.youtube.com/watch?v=LezLNMznZm4&ab\\_channel=NesoAcademy](https://www.youtube.com/watch?v=LezLNMznZm4&ab_channel=NesoAcademy)
- [https://www.youtube.com/watch?v=0sRAMfPUQ58&ab\\_channel=NesoAcademy](https://www.youtube.com/watch?v=0sRAMfPUQ58&ab_channel=NesoAcademy)

## Exercise 2

### Part 1

Proving that the block diagram in Figure 1.a is equivalent to the one in Figure 1.b. In other words, proving the commutativity of the convolution sum.



System 1 =  $s_1[n]$

System 2 =  $s_2[n]$

$$(s_1 * s_2)[n] = s_1[n] * s_2[n] = \sum_{k=-\infty}^{\infty} s_1[k]s_2[n-k]$$

Using the following equivalences:

$$K = n - k \text{ such that } k = n - K$$

$$k = \infty \text{ then } K = -\infty$$

$$k = -\infty \text{ then } K = \infty$$

The original convolution sum can be rewritten by substituting  $k = n - K$ :

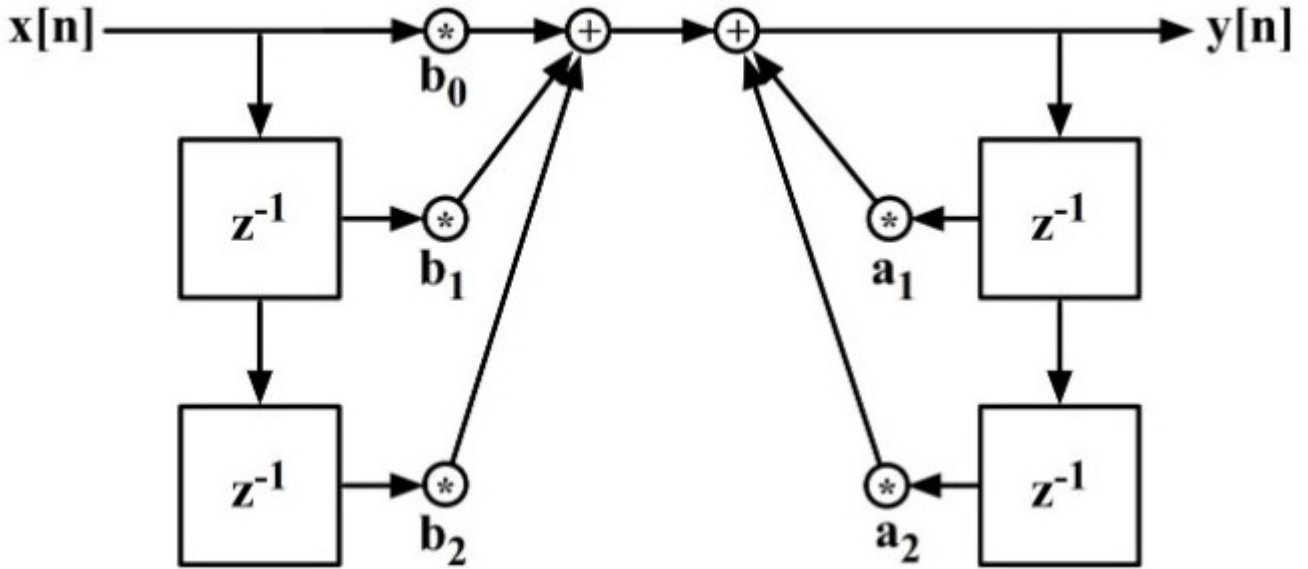
$$y[n] = \sum_{K=-\infty}^{\infty} s_1[n - K]s_2[n - (n - K)] = \sum_{K=-\infty}^{\infty} s_2[K]s_1[n - K]$$

Observing that the range of  $k$  and  $K$  are the same, the following equivalence can be derived giving that

$$s_1 * s_2 = s_2 * s_1:$$

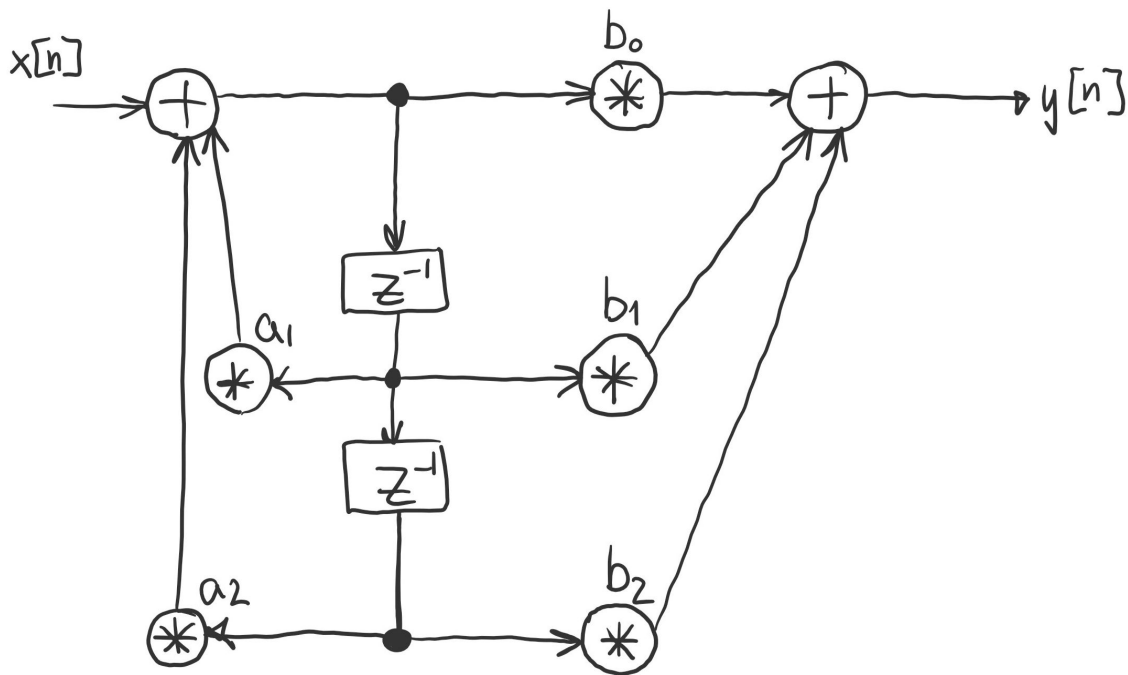
$$y[n] = \sum_{K=-\infty}^{\infty} s_2[K]s_1[n - K] = \sum_{k=-\infty}^{\infty} s_2[k]s_1[n - k] = s_2[n] * s_1[n]$$

## Part 2



$$\begin{aligned}
 y[n] &= b_0x[n] + b_1x[n-1] + b_2x[n-2] + a_1y[n-1] + a_2y[n-2] \\
 &= (x[n])b_0 + (x[n] * z^{-1})b_1 + (x[n] * z^{-1} * z^{-1})b_2 + (y[n] * z^{-1})a_1 + (y[n] * z^{-1} * z^{-1})a_2 \\
 &= (x[n])b_0 + (z^{-1} * x[n])b_1 + (z^{-1} * z^{-1} * x[n])b_2 + (z^{-1} * y[n])a_1 + (z^{-1} * z^{-1} * y[n])a_2 \\
 &= (x[n])b_0 + (z^{-1} * b_1x[n]) + (z^{-1} * z^{-1} * b_2x[n]) + (z^{-1} * a_1y[n]) + (z^{-1} * z^{-1} * a_2y[n]) \\
 &= (x[n])b_0 + (z^{-1} * (b_1x[n] + a_1y[n] + (z^{-1} * b_2x[n] + z^{-1} * a_2y[n]))) \\
 &= (x[n])b_0 + (z^{-1} * (b_1x[n] + a_1y[n] + z^{-1} * (b_2x[n] + a_2y[n]))) \\
 &= b_0x[n] + b_1x[n-1] + a_1y[n-1] + b_2x[n-2] + a_2y[n-2] \\
 &= b_0x[n] + b_1x[n-1] + b_2x[n-2] + a_1y[n-1] + a_2y[n-2]
 \end{aligned}$$

Applying the commutativity of the convolution sum, the simplified flow diagram that uses only two 1-sample delays is displayed below.



### Part 3

The frequency response  $H(\omega) = \frac{Y(\omega)}{X(\omega)}$  of the system

$y[n] = b_0x[n] + b_1x[n-1] + b_2x[n-2] - a_1y[n-1] - a_2y[n-2]$  is derived below:

$$y[n] = b_0x[n] + b_1x[n-1] + b_2x[n-2] - a_1y[n-1] - a_2y[n-2]$$

$$Y(\omega) - a_1e^{-j\omega}Y(\omega) - a_2e^{-j2\omega}Y(\omega) = b_0X[\omega] + b_1e^{-j\omega}X[\omega] + b_2e^{-j2\omega}X[\omega]$$

$$Y(\omega)(1 - a_1e^{-j\omega} - a_2e^{-j2\omega}) = X[\omega](b_0 + b_1e^{-j\omega} + b_2e^{-j2\omega})$$

$$H[\omega] = \frac{Y(\omega)}{X[\omega]} = \frac{(b_0 + b_1e^{-j\omega} + b_2e^{-j2\omega})}{(1 - a_1e^{-j\omega} - a_2e^{-j2\omega})}$$

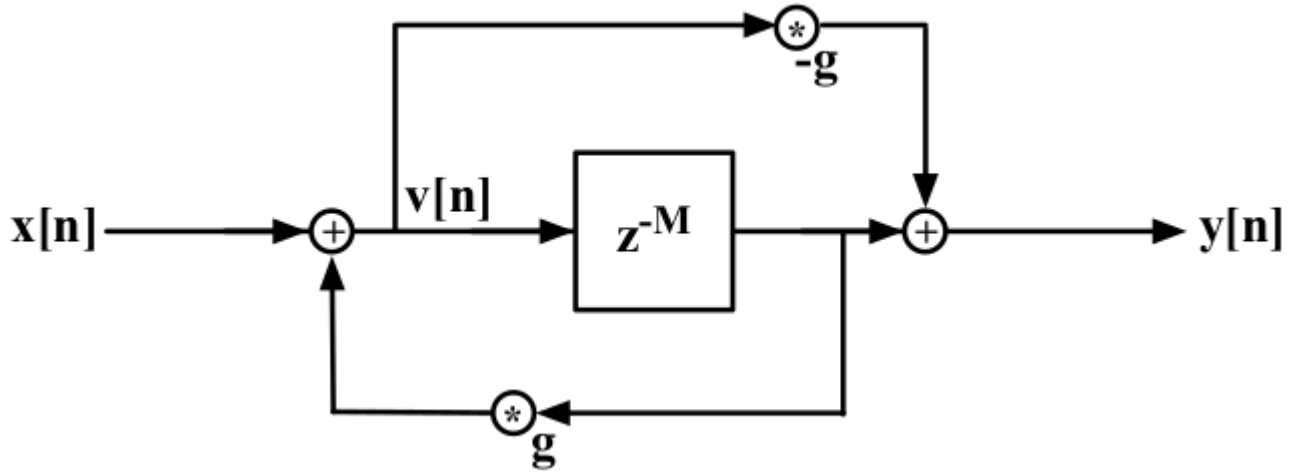
Sources used to complete this section:

- [https://www.dsprelated.com/freebooks/filters/Direct\\_Form\\_II.html](https://www.dsprelated.com/freebooks/filters/Direct_Form_II.html)
- [https://www.youtube.com/watch?v=ap1qXBTkU8g&list=PLbqhA-NKGP6Afr\\_KbPUuy\\_yIBpPR4jzWo&index=8&ab\\_channel=AkashMurthy](https://www.youtube.com/watch?v=ap1qXBTkU8g&list=PLbqhA-NKGP6Afr_KbPUuy_yIBpPR4jzWo&index=8&ab_channel=AkashMurthy)
- [https://www.youtube.com/watch?v=mRKPGt97ytw&ab\\_channel=Engg-Course-Made-Easy](https://www.youtube.com/watch?v=mRKPGt97ytw&ab_channel=Engg-Course-Made-Easy)
- [https://eng.libretexts.org/Bookshelves/Electrical\\_Engineering/Signal\\_Processing\\_and\\_Modeling/Signals\\_and\\_Systems\\_\(Baraniuk\\_et\\_al.\)/04%3A\\_Time\\_Domain\\_Analysis\\_of\\_Discrete\\_Time\\_Systems/4.04%3A\\_Properties\\_of\\_Discrete\\_Time\\_Convolution](https://eng.libretexts.org/Bookshelves/Electrical_Engineering/Signal_Processing_and_Modeling/Signals_and_Systems_(Baraniuk_et_al.)/04%3A_Time_Domain_Analysis_of_Discrete_Time_Systems/4.04%3A_Properties_of_Discrete_Time_Convolution)
- <https://httpsachinbagul.files.wordpress.com/2017/12/lec-10.pdf>
- [https://www.youtube.com/watch?v=yuS86\\_oimY5Y&ab\\_channel=ECACademy](https://www.youtube.com/watch?v=yuS86_oimY5Y&ab_channel=ECACademy)
- [https://www.youtube.com/watch?v=Jl6dapilt58&ab\\_channel=JohnBuck](https://www.youtube.com/watch?v=Jl6dapilt58&ab_channel=JohnBuck)

### Exercise 3

## Part 1: Computing the Frequency Response

The difference equations  $y[n] = v[n - M] - gv[n]$  and  $v[n] = x[n] + gv[n - M]$  can be used to describe the system in the figure below. Using the two equations, the frequency response  $H(\omega) = \frac{Y(\omega)}{X(\omega)}$  can be computed.



The transfer function  $H(\omega) = \frac{Y(\omega)}{X(\omega)}$ , describes the steady-state response of a system to sinusoidal inputs of varying frequencies. In other words it is how a sinusoidal signal of a given frequency is affected by the filter.

Rearranging  $v[n] = x[n] + gv[n - M]$  to get  $v[n] - gv[n - M] = x[n]$ . Taking the DTFT on both sides of the equation:

$$V(\omega) - ge^{-jM\omega}V(\omega) = X(\omega)$$

$$V(\omega)(1 - ge^{-jM\omega}) = X(\omega)$$

$$V(\omega) = \frac{X(\omega)}{(1 - ge^{-jM\omega})}$$

Taking the DTFT on both sides of the equation  $y[n] = v[n - M] - gv[n]$ :

$$Y(\omega) = e^{-jM\omega}V(\omega) - gV(\omega)$$

$$Y(\omega) = V(\omega)(e^{-jM\omega} - g)$$

Substituting  $V(\omega)$  into  $Y(\omega)$  to obtain  $H(\omega)$ :

$$Y(\omega) = \frac{X(\omega)}{(1 - ge^{-jM\omega})} (e^{-jM\omega} - g)$$

$$Y(\omega) = X(\omega) \frac{(e^{-jM\omega} - g)}{(1 - ge^{-jM\omega})}$$

$$H(\omega) = \frac{Y(\omega)}{X(\omega)} = \frac{(e^{-jM\omega} - g)}{(1 - ge^{-jM\omega})}$$

## Part 2: Plotting the Frequency and Phase Responses

The frequency response appears to have positive magnitude for all omega inputs which suggests this may be an all-pass filter.

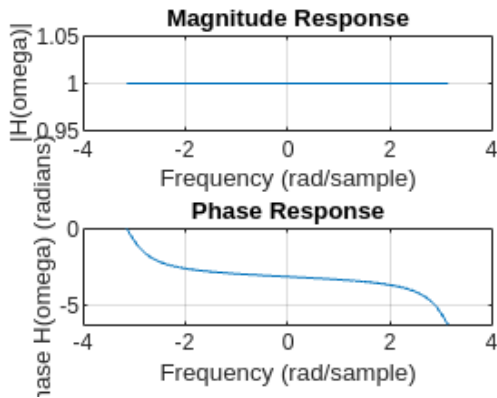
```
% Given values
g = 0.7;
M = 1000;
omega = linspace(-pi, pi, M); % the frequency range

% Calculate the frequency response H(omega) based on the computed transfer function
in Part 1
Y_omega = exp(-1j*M*omega) - g;
X_omega = 1 - g*exp(-1j*M*omega);
H_omega = Y_omega ./ X_omega;

% Calculate magnitude and phase response
magnitude_response = abs(H_omega);
phase_response = unwrap(angle(H_omega)); % Unwrap the phase response

% Plot magnitude response
figure;
subplot(2, 1, 1);
plot(omega, magnitude_response);
title('Magnitude Response');
xlabel('Frequency (rad/sample)');
ylabel('|H(omega)|');
grid on;
ylim([0.95 1.05]); % Adjust the y-axis range to better visualize the magnitude
response

% Plot phase response
subplot(2, 1, 2);
plot(omega, phase_response);
title('Phase Response');
xlabel('Frequency (rad/sample)');
ylabel('Phase H(omega) (radians)');
grid on;
```



### Part 3: Filter Implementation

The filter implementation function `a2_filter` can be found in the Appendix.

### Part 4: Impulse Response of Filter

Passing an impulse through the filter produces a sort of clicking sound that is slightly delayed (there are small silences) and also dampened with time. The best analogy is a small round object bouncing to a stop.

```
% Given values
g = 0.7;
M = 1000;

% Impulse
duration = get_impulse_duration(M, g);
x = get_impulse(duration);

% Evaluate the filter
y = a2_filter(x, M, g);
sound(y);
```

### Part 5: Cascading Filters

Passing an impulse through the filter produces a clicking sound that is damped and low pitched but has smaller silences between sounds. The best analogy is either a very thick rubber band being flicked or a Vargan instrument oscillating.

```
% Given values
g = 0.7;
Ms = [1051, 337, 113];

% Calculate durations
durations = arrayfun(@(M) get_impulse_duration(M, g), Ms);
duration = get_impulse_duration(Ms(1), g);

% Create impulse
x = get_impulse(duration);
```



```
% Cascade filters
y1 = a2_filter(x, Ms(1), g);
y2 = a2_filter(y1, Ms(2), g);
y3 = a2_filter(y2, Ms(3), g);
sound(y3);
```

## Part 6: Sound File Through Filter

Passing a *guitar.wav* sound file through the *a2\_filter* system produces a sound that is identical to the input. The system is therefore an **all-pass filter**.

```
% Given values
g = 0.7;
[x, Fs] = audioread('guitar.wav');
M = length(x);
sound(x, Fs); % original sound

y = a2_filter(x, M, g);
sound(y, Fs); % filtered sound
```

Sources used to complete this section:

- <https://www.mathworks.com/matlabcentral/answers/1756050-how-to-plot-frequency-response-phase-response-from-transfer-function>
- [https://www.projectrhea.org/rhea/index.php/Zachary\\_Curosh\\_-\\_Frequency\\_Response\\_and\\_Difference\\_Equations\\_ECE301Fall2008mboutin](https://www.projectrhea.org/rhea/index.php/Zachary_Curosh_-_Frequency_Response_and_Difference_Equations_ECE301Fall2008mboutin)
- [https://www.youtube.com/watch?v=cNwuwoGRuH8&ab\\_channel=StephenMendes](https://www.youtube.com/watch?v=cNwuwoGRuH8&ab_channel=StephenMendes)
- <https://www.mathworks.com/matlabcentral/answers/44770-logarithm-of-base-other-than-e-10-and-2>
- [https://www.youtube.com/watch?v=Utvvsr\\_d02c&ab\\_channel=EdD](https://www.youtube.com/watch?v=Utvvsr_d02c&ab_channel=EdD)

## Exercise 4

The goal of this exercise is to find the transient response of a causal Linear Time Invariant (LTI) system when the input is a causal signal, i.e. a signal that starts at time  $n = 0$ . The idea is to understand what happens during the attack of a sound in common situations where the sound is generated by a sudden activation of the production system.

The general strategy is to get the transient response  $y_{tr}[n]$  by comparing  $y_c[n]$  and  $y[n]$ , the causal and non-causal responses of the system respectively. Two strategies are going to be used: first strategy in parts 1-3, second strategy in parts 4 and 5.

### Part 1

Considering a causal LTI system  $H$  characterized by its impulse response  $h[n] = \frac{1}{3} \sum_{l=0}^{l=2} \delta[n-l]$ , then

$y_c[n]$  will be the response of system  $H$  to a ‘causal sinusoidal’ input  $x_c[n] = 2\cos(\frac{\pi}{3}n)u[n]$ , where  $u[n]$

is the unit-step signal. Using the *fold-shift-add* method of convolution, a mathematical expression for  $y_c[n] = x_c[n] * h[n]$  is obtained below.

Manually drawing each signal,  $h[n]$  looks like  $h[n] = \left\{\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right\}$  for  $n \in \mathbb{N}$  and  $x_c[n]$  looks like

$x_c[n] = \{2, 1, -1, -2, -1, 1, 2, 1, -1, -2, -1, 1, \dots\}$  for  $n \in \mathbb{N}$ . Convolving the two signals together:

$$\frac{1}{3}(2) = \frac{2}{3}$$

$$\frac{1}{3}(1) + \frac{1}{3}(2) = 1$$

$$\frac{1}{3}(-1) + \frac{1}{3}(1) + \frac{1}{3}(2) = \frac{2}{3}$$

$$\frac{1}{3}(-2) + \frac{1}{3}(-1) + \frac{1}{3}(1) = -\frac{2}{3}$$

$$\frac{1}{3}(-1) + \frac{1}{3}(-2) + \frac{1}{3}(-1) = -\frac{4}{3}$$

$$\frac{1}{3}(1) + \frac{1}{3}(-1) + \frac{1}{3}(-2) = -\frac{2}{3}$$

$$\frac{1}{3}(2) + \frac{1}{3}(1) + \frac{1}{3}(-1) = \frac{2}{3}$$

$$\frac{1}{3}(1) + \frac{1}{3}(2) + \frac{1}{3}(1) = \frac{4}{3}$$

$$\frac{1}{3}(-1) + \frac{1}{3}(1) + \frac{1}{3}(2) = \frac{2}{3}$$

$$\frac{1}{3}(-2) + \frac{1}{3}(-1) + \frac{1}{3}(1) = -\frac{2}{3}$$

...

Based on the above derivation, a general formula for the convolution  $y_c[n]$  can be derived:

$$y_c[n] = \begin{cases} \frac{1}{3} \left( 2 \cos \left( \frac{\pi}{3} n \right) + 2 \cos \left( \frac{\pi}{3} (n-1) \right) + 2 \cos \left( \frac{\pi}{3} (n-2) \right) \right), & \text{for } n \geq 2 \\ \frac{1}{3} \left( 2 \cos \left( \frac{\pi}{3} n \right) + 2 \cos \left( \frac{\pi}{3} (n-1) \right) \right), & \text{for } n = 1 \\ \frac{1}{3} \left( 2 \cos \left( \frac{\pi}{3} n \right) \right), & \text{for } n = 0 \\ 0, & \text{for } n < 0 \end{cases}$$

After simplifying the equations, the final convolution  $y_c[n]$  is obtained below:

$$y_c[n] = \begin{cases} \frac{2}{3} \left( \cos\left(\frac{\pi}{3}n\right) + \cos\left(\frac{\pi}{3}(n-1)\right) + \cos\left(\frac{\pi}{3}(n-2)\right) \right), & \text{for } n \geq 3 \\ \frac{2}{3}, & \text{for } n = 2 \\ 1, & \text{for } n = 1 \\ \frac{2}{3}, & \text{for } n = 0 \\ 0, & \text{for } n < 0 \end{cases}$$

## Part 2

Computing the frequency response  $H(\omega)$  of the LTI system  $H$  defined in the beginning of this section.

Deriving the specific expression of  $H(\omega)$  when  $\omega = \omega_0 = \frac{\pi}{3}$  to describe the action of  $H$  in terms of

delay and amplification. The expression for the response  $y[n]$  is expressed in terms of the signal

$x[n] = 2\cos(\frac{\pi}{3}n)$ . The part below computes the expression for  $y_{tr}[n]$  by comparing the corresponding

$y[n]$  to the already obtained expression of  $y_c[n]$  through  $y_{tr}[n] = y_c[n] - y[n]$ .

$$h[n] = \frac{1}{3}(\delta[n] + \delta[n-1] + \delta[n-2])$$

$$H(\omega) = \frac{1}{3}(1 + e^{-\omega j} + e^{-2\omega j})$$

$$H\left(\frac{\pi}{3}\right) = \frac{1}{3}\left(1 + e^{\frac{-\pi}{3}j} + e^{\frac{-2\pi}{3}j}\right)$$

$$H\left(\frac{\pi}{3}\right) = \frac{1}{3}\left(1 + \frac{1}{2} - \frac{\sqrt{3}}{2}j - \frac{1}{2} - \frac{\sqrt{3}}{2}j\right)$$

$$H\left(\frac{\pi}{3}\right) = \frac{1}{3}(1 - \sqrt{3}j)$$

$$H\left(\frac{\pi}{3}\right) = \frac{1}{3} - \frac{\sqrt{3}}{3}j$$

$$\angle H\left(\frac{\pi}{3}\right) = \tan\left(\frac{\frac{-\sqrt{3}}{3}}{\frac{1}{3}}\right) = \tan(-\sqrt{3}) = -\frac{\pi}{3}$$

$$|H\left(\frac{\pi}{3}\right)| = \sqrt{\left(\frac{1}{3}\right)^2 + \left(\frac{\sqrt{3}}{3}\right)^2} = \sqrt{\frac{1}{9} + \frac{3}{9}} = \sqrt{\frac{4}{9}} = \frac{2}{3}$$

From the above result the  $y[n]$  can be deduced to be  $\frac{4}{3}\cos\left(\frac{\pi}{3}n + \frac{\pi}{3}\right)$  which results in the following equation for

$y_{tr}[n]$ :

$$y_{tr}[n] = \begin{cases} \frac{2}{3} \left( \cos\left(\frac{\pi}{3}n\right) + \cos\left(\frac{\pi}{3}(n-1)\right) + \cos\left(\frac{\pi}{3}(n-2)\right) \right) - \frac{4}{3} \cos\left(\frac{\pi}{3}n - \frac{\pi}{3}\right), & \text{for } n \geq 2 \\ \frac{2}{3} \left( \cos\left(\frac{\pi}{3}n\right) + \cos\left(\frac{\pi}{3}(n-1)\right) \right) - \frac{4}{3} \cos\left(\frac{\pi}{3}n - \frac{\pi}{3}\right), & \text{for } n = 1 \\ \frac{2}{3} \left( \cos\left(\frac{\pi}{3}n\right) \right) - \frac{4}{3} \cos\left(\frac{\pi}{3}n - \frac{\pi}{3}\right), & \text{for } n = 0 \\ 0, & \text{for } n < 0 \end{cases}$$

After simplifying the equations, the final convolution  $y_{tr}[n]$  is obtained below:

$$y_{tr}[n] = \begin{cases} \frac{2}{3} \left( \cos\left(\frac{\pi}{3}n\right) + \cos\left(\frac{\pi}{3}(n-1)\right) + \cos\left(\frac{\pi}{3}(n-2)\right) \right) - \frac{4}{3} \cos\left(\frac{\pi}{3}n - \frac{\pi}{3}\right), & \text{for } n \geq 3 \\ 0, & \text{for } n = 2 \\ -\frac{1}{3}, & \text{for } n = 1 \\ 0, & \text{for } n = 0 \\ 0, & \text{for } n < 0 \end{cases}$$

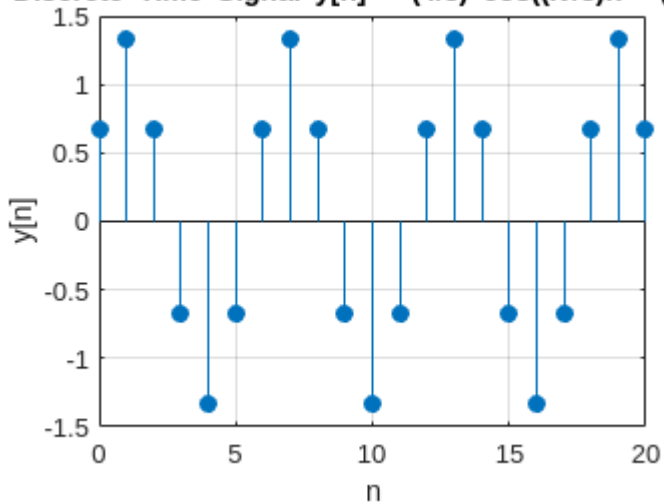
### Part 3

Plotting the steady-state response  $y[n]$ .

```
% Plotting the computed y[n]
n = 0:20; % Define the range
y_n = (4/3) * cos((pi/3) * n - (pi/3));

% Plot the signal
stem(n, y_n, 'filled');
title('Discrete Time Signal y[n] = (4/3) cos((\pi/3)n - (\pi/3))');
xlabel('n');
ylabel('y[n]');
grid on;
```

Discrete Time Signal  $y[n] = (4/3) \cos((\pi/3)n - (\pi/3))$

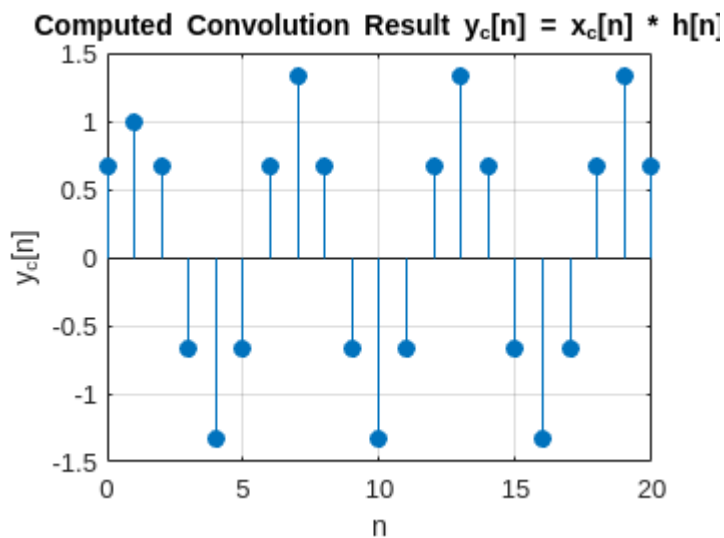


Verifying  $y_c[n]$  by first computing the convolution and plotting the result, then plotting the obtained mathematical expression from Part 1. The two graphs are the same.

```
% Plotting the computed y_c[n]
n = 0:20; % Define the range
% Computing the convolution y_c[n] = x_c[n] * h[n] for verification
h = (1/3)*ones(1,3); % Define the impulse response h[n]
x_c = 2*cos(pi/3*n); % Define the input signal x_c[n]

% Zero-pad x_c to make sure the convolution result is the same length as n
x_c_padded = [x_c, zeros(1, length(h) - 1)];
y_c = conv(x_c_padded, h ); % y_c[n] = x_c[n] * h[n]

% Plot the signal
figure;
stem(n, y_c(1:length(n)), 'filled');
title('Computed Convolution Result y_{c}[n] = x_{c}[n] * h[n]');
xlabel('n');
ylabel('y_{c}[n]');
grid on;
```



```
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Plotting the formula y_c[n] obtained in Part 1
x_c = 2*cos(pi/3*n);

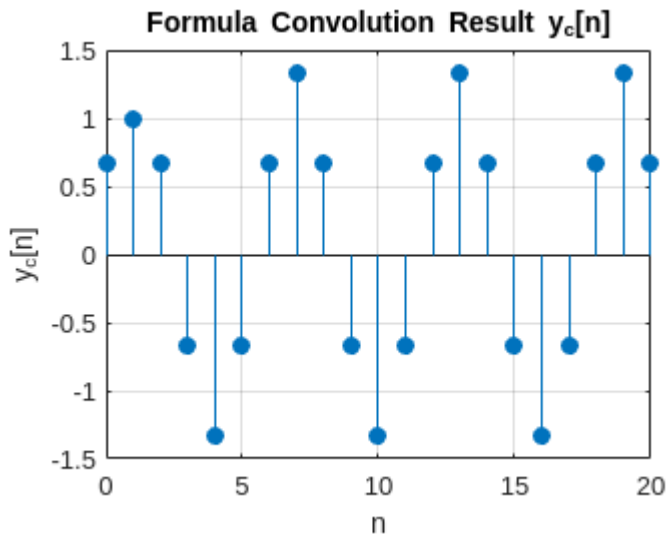
% Calculate y[n] for each value of n using the piecewise function
y = zeros(1, length(n));
for i = 1:length(n)
    if i >= 3
        y(i) = (1/3) * (x_c(i) + x_c(i-1) + x_c(i-2));
    elseif i == 2
        y(i) = (1/3) * (x_c(i) + x_c(i-1));
    elseif i == 1
```

```

        y(i) = (1/3) * x_c(i);
    end
end

% Plot the signal
stem(n, y, 'filled');
title('Formula Convolution Result y_{c}[n]');
xlabel('n');
ylabel('y_{c}[n]');
grid on;

```



Plotting the transient response  $y_{tr}[n] = y_c[n] - y[n]$ .

```

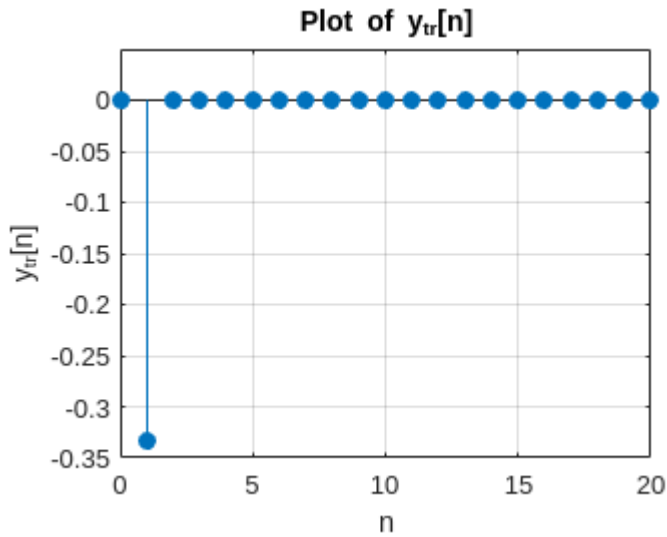
% Plotting the computed y_tr[n]
n = 0:20; % Assuming we want to plot from n=0 to n=20

% Calculate y_tr[n] for each value of n using the piecewise function
y_tr = zeros(1, length(n));
for i = 1:length(n)
    if n(i) >= 2
        % for n >= 2
        y_tr(i) = (2/3)*(cos(pi/3*n(i)) + cos(pi/3*(n(i)-1)) + cos(pi/3*(n(i)-2)))
        - (4/3)*cos(pi/3*n(i) - pi/3);
    elseif n(i) == 1
        % for n = 1
        y_tr(i) = (2/3)*(cos(pi/3*n(i)) + cos(pi/3*(n(i)-1))) - (4/3)*cos(pi/3*n(i)
        - pi/3);
    elseif n(i) == 0
        % for n = 0
        y_tr(i) = (2/3)*cos(pi/3*n(i)) - (4/3)*cos(pi/3*n(i) - pi/3);
    end
end

% Plot the signal
stem(n, y_tr, 'filled');

```

```
title('Plot of y_{tr}[n]');
xlabel('n');
ylabel('y_{tr}[n]');
grid on;
```



## Part 4

This section will now derive a general expression for the transient response  $y_{tr}[n]$  of a causal LTI system, for which the impulse response is  $h[n]$  when the input is a ‘causal sinusoid’  $x_c$ , i.e. a sinusoid that starts at time  $n = 0$ . Starting from the expression of the output of the LTI as a convolution sum  $\sum_{k=-\infty}^{k=+\infty} x[k]h[n-k]$ , it is shown that the transient response can be expressed as:

$$y_{tr}[n] = - \sum_{k=n+1}^{k=+\infty} h[k] \cos(\omega(n-k)) \text{ for } n \geq 0$$

Using the definition of convolution:

$$y_c[n] = \sum_{k=-\infty}^{\infty} h[k] \cos(\omega(n-k)) u[n-k]$$

Applying the  $u[n]$  step function:

$$y_c[n] = \sum_{k=0}^{\infty} h[k] \cos(\omega(n-k))$$

Assuming that the filter  $h$  has a length of  $L$  and using the formula for the convolution of infinite sequences with a finite filter:

$$y_c[n] = \sum_{k=0}^{\min(n,L)} h[k] \cos(\omega(n-k))$$

At steady state,  $n \geq L$  and the above expression can be simplified to express the steady state response using the following equation:

$$y[n] = \sum_{k=0}^L h[k] \cos(\omega(n-k))$$

Using the formula for the transient response  $y_{tr}[n] = y_c[n] - y[n]$ :

$$y_{tr}[n] = \sum_{k=0}^{\min(n,L)} h[k] \cos(\omega(n-k)) - \sum_{k=0}^L h[k] \cos(\omega(n-k))$$

Since both summations have the same  $h[k] \cos(\omega(n-k))$ , the first few terms cancel (depending on what is the  $n$ ) and the equation is left with the values at  $n+1$ , such the lower bound of the summation of the second term gets updated to  $k = n+1$  and the first term gets removed. Leaving the following equation:

$$y_{tr}[n] = - \sum_{k=n+1}^L h[k] \cos(\omega(n-k))$$

Since the length of the filter is not known, it can be assumed that  $L = \infty$ :

$$y_{tr}[n] = - \sum_{k=n+1}^{\infty} h[k] \cos(\omega(n-k))$$

## Part 5

$$\begin{aligned} y_{tr}[n] &= - \sum_{k=n+1}^{\infty} h[k] \cos(\omega(n-k)) \\ &= - \sum_{k=n+1}^{\infty} \sum_{l=0}^2 \frac{1}{3} \delta[k-l] 2 \cos\left(\frac{\pi}{3}(n-k)\right) \\ &= - \frac{2}{3} \sum_{k=n+1}^{\infty} \sum_{l=0}^2 \delta[k-l] \cos\left(\frac{\pi}{3}(n-k)\right) \\ &= - \frac{2}{3} \sum_{k=n+1}^2 \cos\left(\frac{\pi}{3}(n-k)\right) \end{aligned}$$

Sources used to complete this section:

- [https://www.reddit.com/r/DSP/comments/w1ajdp/found\\_this\\_super\\_easy\\_way\\_of\\_doing\\_convolution/](https://www.reddit.com/r/DSP/comments/w1ajdp/found_this_super_easy_way_of_doing_convolution/)
- <https://dsp.stackexchange.com/questions/23988/why-is-the-output-of-an-lti-system-expressed-as-the-convolution-of-the-input-wit>
- [https://eceweb1.rutgers.edu/Assuming that the filter h as a length of :u/~orfanidi/intro2sp/2e/orfanidis-isp2e-1up.pdf](https://eceweb1.rutgers.edu/Assuming%20that%20the%20filter%20h%20has%20a%20length%20of%20%22u%22/~orfanidi/intro2sp/2e/orfanidis-isp2e-1up.pdf)

## Appendix

### Exercise 3 Part 3



```

function y = a2_filter(x, M, g)
    % Args:
    %   x: Input signal
    %   M: Delay length
    %   g: Gain
    %
    % Returns:
    %   y: Output signal (vector) after processing through the filter.
    %
    % The filter implements the following difference equations:
    %    $v[n] = x[n] + g \cdot v[n-M]$ 
    %    $y[n] = v[n-M] - g \cdot v[n]$ 

    % Filter 1
    %  $v[n] = x[n] + g \cdot v[n-M]$ 
    b_1 = 1; % feedforward coefficients
    a_1 = [1, zeros(1, M-1), -g]; % feedback coefficient

    % Filter 2
    %  $y[n] = v[n-M] - g \cdot v[n]$ 
    b_2 = [-g, zeros(1, M-1), 1]; % feedforward coefficients
    a_2 = 1; % feedback coefficient

    v = filter(b_1, a_1, x);
    y = filter(b_2, a_2, v);
end

```

## Helper Functions

```

function duration = get_impulse_duration(M, g)
    duration = round(M*(1 - ((log10(1000*(1-g.^2))) ./ (log10(abs(g))))));
end

function impulse = get_impulse(duration)
    impulse = [1; zeros(duration-1, 1)];
end

```