

Problem 1

Sampling a Periodic Signal: Write a MATLAB code to generate a sinusoid signal with an amplitude of 1 a.u. (arbitrary unit) and frequency of 0.5 Hz. Plot the analog signal as a function of time. Sample the signal with a sampling time, Δt of 0.1 second. Plot the sampled signal, $x[n]$ as a function of sample number, n . [repeat question from assignment 1]

(4 points) Generate a noise signal using 'rand' function in MATLAB within the range $[-0.5, 0.5]$.

Plot the noise signal.

(4 Points) Add the noise to $x[n]$ to generate a noisy signal, $x_{\text{noise}}[n]$.

(9 Points) Apply a moving average transformation on $x_{\text{noise}}[n]$, with a symmetric window when

(i) window size=3 samples ($M1 = 1, M2=1$)

(ii) window size=5 samples ($M1 = 2, M2=2$)

(iii) window size=7 samples ($M1 = 3, M2=3$)

Comment on the optimal window size that recovers the trend of the original signal.

```
clear
clc
% Parameters
A = 1; % Amplitude in arbitrary units
f = 0.5; % Frequency in Hz
w = f*2*pi; % Angular frequency
Ts = 0.1; % Sampling Period
N = (2*pi)/(w*Ts);
T = 1/f; % Period

% Generate sinusoid signal
t = 0:0.1:T;
signal = A*sin(w*t);

% Generate sinusoid signal
n = 0:N-1;
t_sampled = n*Ts;
sampled_signal = A*sin(w*t_sampled);

% Generate noise signal
noise_signal = (rand(1, length(n)) - 0.5);

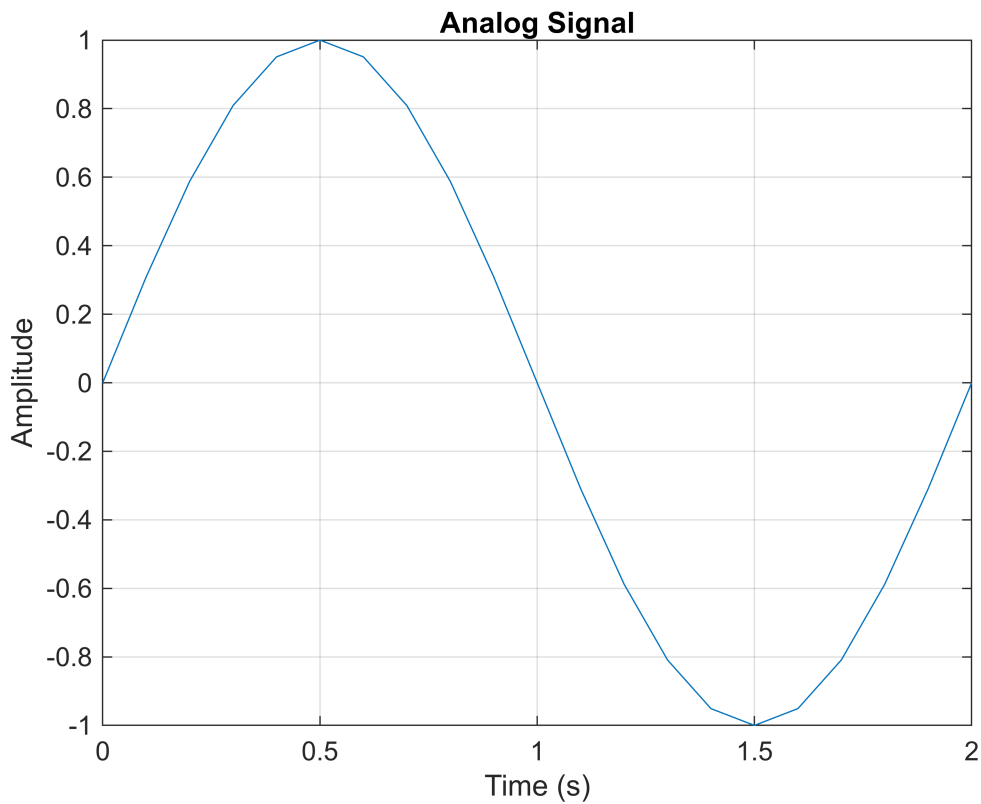
% Add the original signal to the noise signal
x_noise = sampled_signal + noise_signal;

% Moving Average for a window size of 3, 5, and 7
window_3_output = movmean(x_noise, 3);
window_5_output = movmean(x_noise, 5);
window_7_output = movmean(x_noise, 7);
```

```

% Plot the original analog signal
figure(1);
plot(t, signal);
xlabel('Time (s)');
ylabel('Amplitude');
title('Analog Signal');
grid on;

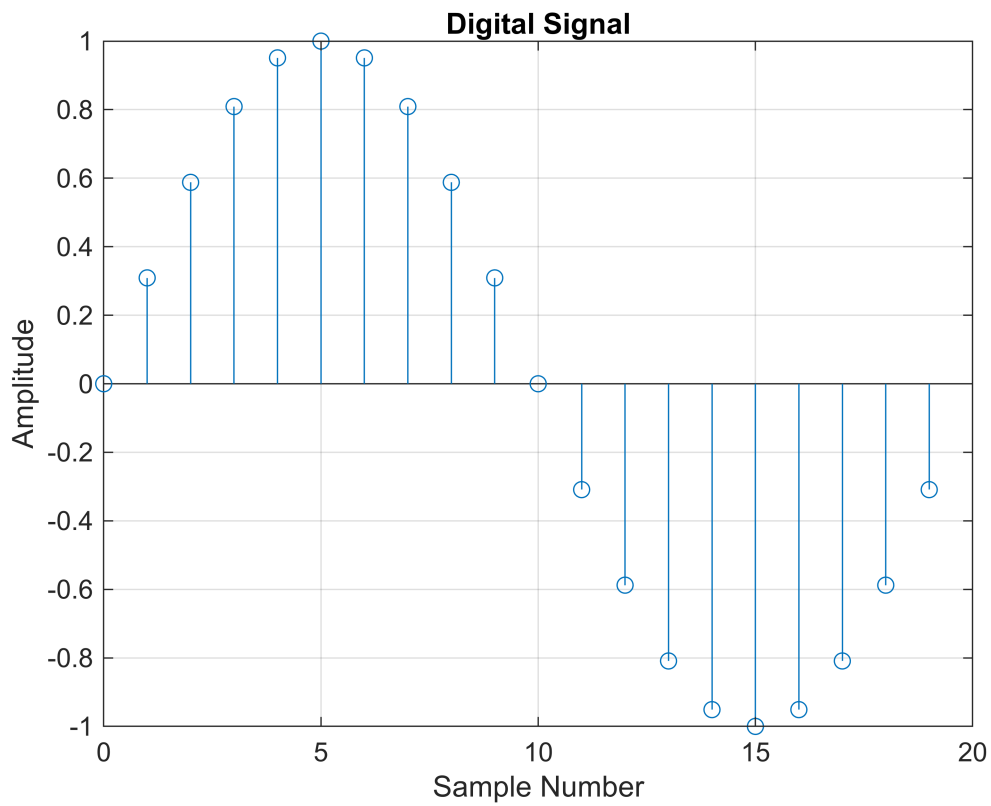
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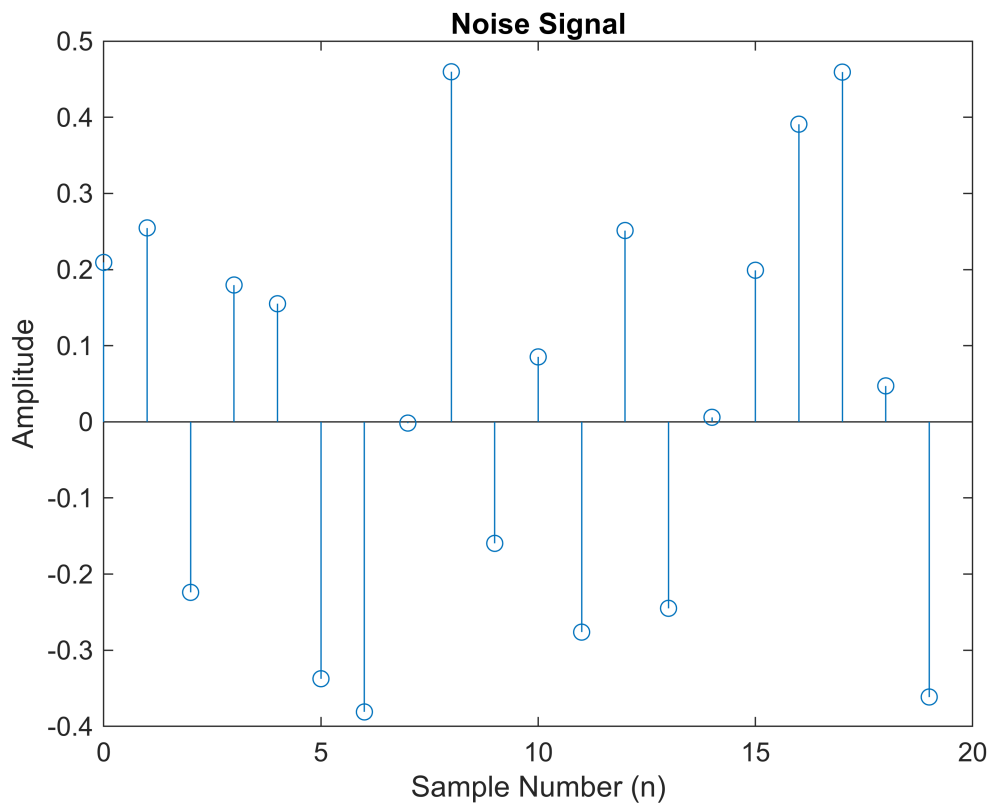
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% Plot the original digital signal
figure(2);
stem(n, sampled_signal, 'Marker', 'o');
xlabel('Sample Number');
ylabel('Amplitude');
title('Digital Signal');
grid on;

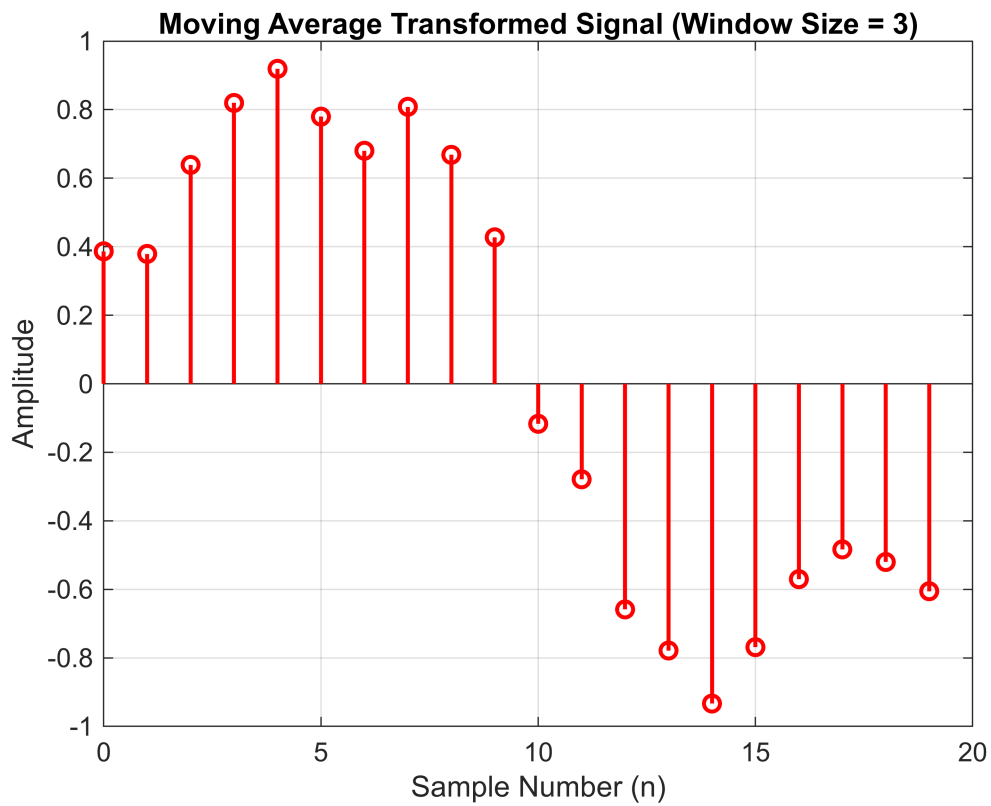
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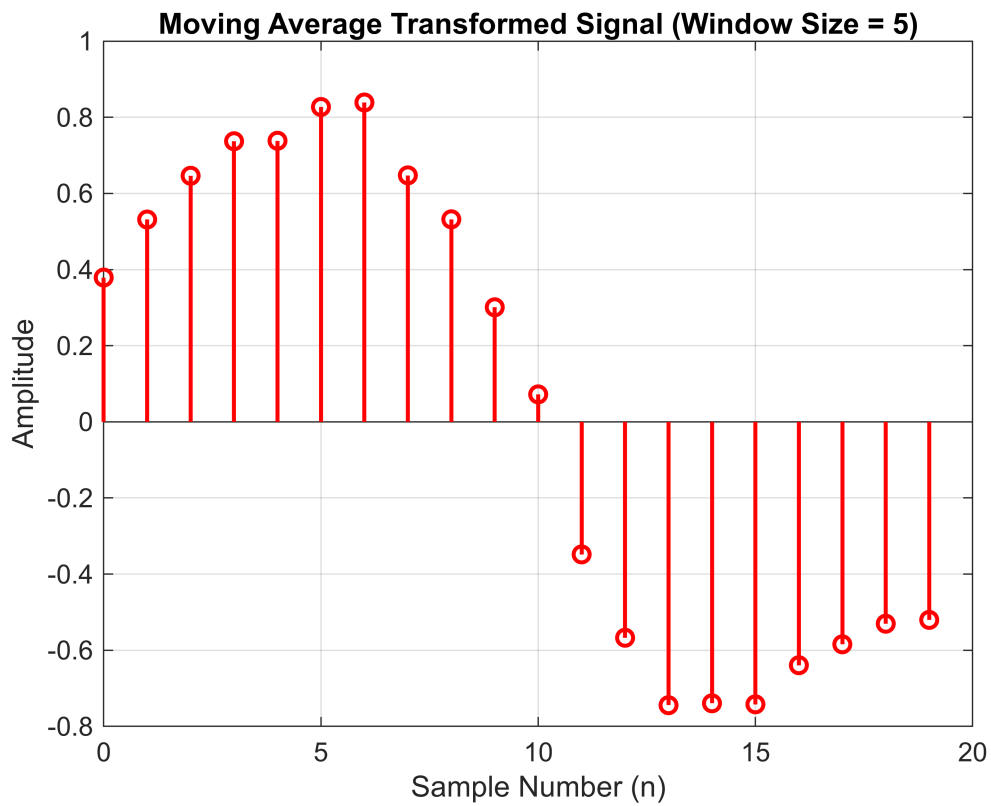
```
% Plot the noise signal
figure(3);
stem(n, noise_signal);
title('Noise Signal');
xlabel('Sample Number (n)');
ylabel('Amplitude');
```



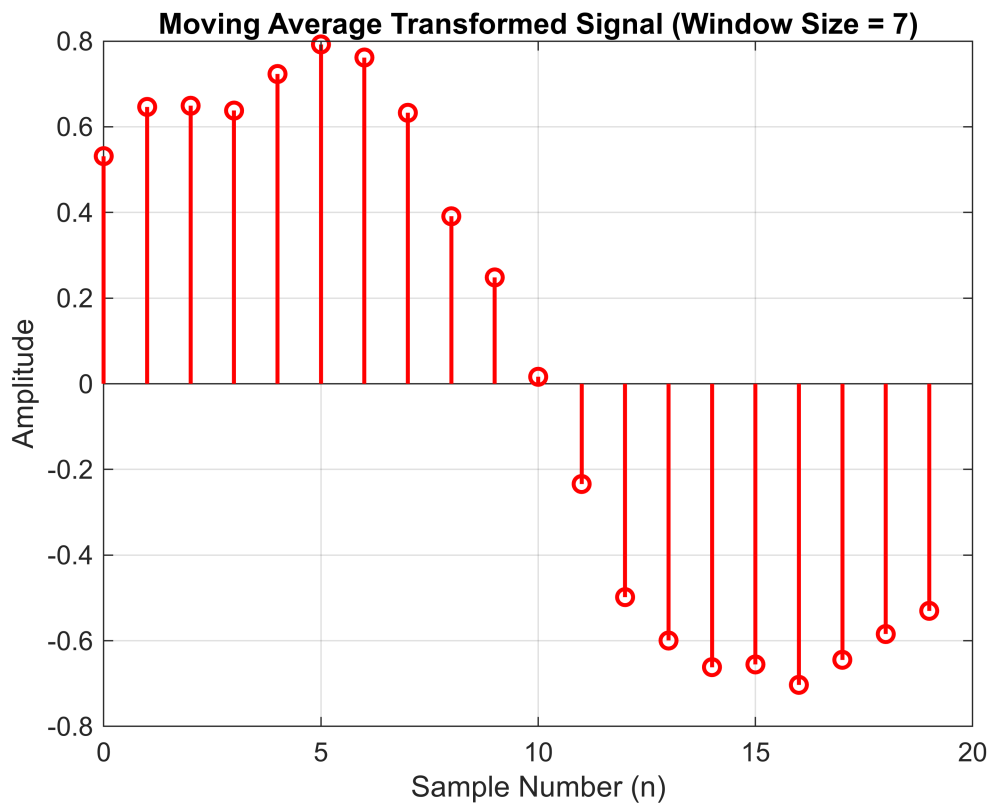
```
% Plot the moving avg for a window of 3
figure(4);
stem(n, window_3_output, 'r', 'LineWidth', 1.5);
title('Moving Average Transformed Signal (Window Size = 3)');
xlabel('Sample Number (n)');
ylabel('Amplitude');
grid on;
```



```
% Plot the moving avg for a window of 5
figure(5);
stem(n, window_5_output, 'r', 'LineWidth', 1.5);
title('Moving Average Transformed Signal (Window Size = 5)');
xlabel('Sample Number (n)');
ylabel('Amplitude');
grid on;
```



```
% Plot the moving avg for a window of 7
figure(6);
stem(n, window_7_output, 'r', 'LineWidth', 1.5);
title('Moving Average Transformed Signal (Window Size = 7)');
xlabel('Sample Number (n)');
ylabel('Amplitude');
grid on;
```



1

$$2. a. \quad T(x[n]) = \sum_{k=n-n_0}^{n+n_0} x[k]$$

$$1. \text{ assume } |x[n]| \leq M_x < \infty \quad \forall n$$

since the input is bounded from $n-n_0$ to $n+n_0$, the system is BIBO stable

2. non-causal, since $x[n+n_0]$ is a part of the system and this would rely on a future input if n_0 is positive

$$3. \quad y[n] = \sum_{k=n-n_0}^{n+n_0} x[k]$$

$$\alpha x_1[n] + \beta x_2[n] \mapsto \sum_{k=n-n_0}^{n+n_0} \alpha x[k] + \beta x[k]$$

$$= \alpha \sum_{k=n-n_0}^{n+n_0} x[k] + \beta \sum_{k=n-n_0}^{n+n_0} x[k] \rightarrow \text{linear}$$

$$4. \text{ shifted output: } y_0[n] = y[n-m] = \sum_{k=n-m-n_0}^{n-m+n_0} x[k]$$

output due to a

$$\text{shifted input: } x_1[n] = x[n-m]$$

$$y_1[n] = \sum_{k=n-n_0}^{n+n_0} x[k-m] \rightarrow \sum_{l=n-n_0-m}^{n+n_0-m} x[l] = y_0[n]$$

change of variables:

$$l = k-m \rightarrow k = l+m$$

so time-invariant

5. has memory since $x[k]$ depends on past or future inputs of $n+n_0$ and $n-n_0$

2

$$b. T(x[n]) = x[n-n_0]$$

$$1. \text{ assume } |x[n]| \leq M_x < \infty \quad \forall n$$

$$|y[n]| = |x[n-n_0]| = M_x < \infty \text{ so it's } \textcircled{\text{stable}}$$

2. causal, if $n_0 \geq 0$, but non-causal if $n_0 < 0$

$$3. \underbrace{\alpha x_1[n] + \beta x_2[n]}_{\textcircled{\text{linear}}} \mapsto \underbrace{\alpha x_1[n-n_0]}_{y_1[n]} + \underbrace{\beta x_2[n-n_0]}_{y_2[n]}$$

$$4. \text{ shifted output: } y_0[n] = y[n-m] = x[n-m-n_0]$$

output of shifted input:

$$y_1[n] = x_1[n-n_0] = x[n-n_0-m]$$

$$y_0[n] = y_1[n] \text{ so } \textcircled{\text{time invariant}}$$

5. has memory since it relies on past or future input depending on the value of n_0

$$c. T(x[n]) = e^{\{x[n]\}}$$

$$1. \text{ assume } |x[n]| \leq M_x < \infty \quad \forall n$$

$$|y[n]| = e^{x[n]} = e^{M_x} < \infty \text{ so } y[n] \text{ is } \textcircled{\text{stable}}$$

2. causal since it only depends on present values

$$3. \bar{x}[n] = \alpha x_1[n] + \beta x_2[n]$$

$$\mapsto e^{\{\alpha x_1[n] + \beta x_2[n]\}} = e^{\alpha x_1[n]} e^{\beta x_2[n]} \neq e^{\alpha x_1[n]} + e^{\beta x_2[n]}$$

so non-linear

$$4. \text{ shifted output: } y_0[n] = y[n-m] = e^{\{x[n-m]\}}$$

$$\text{or due to shifted input: } y_1[n] = e^{\{x[n-m]\}} \\ \text{so } \textcircled{\text{time invariant}}$$

3

5. output only depends on the present input
so memoryless

d. $T(x[n]) = x[-n]$

1. assume $|x[n]| \leq \mu_x < \infty \forall n$

$|y[n]| = |x[-n]| = \mu_x < \infty$ so it's stable

2. not causal \rightarrow when n is positive, the output depends on a future input

ex. $n=2 \rightarrow x[-2]$

3. $\bar{x}[n] = \alpha x_1[n] + \beta x_2[n] \mapsto \underbrace{\alpha x_1[-n]}_{y_1[n]} + \underbrace{\beta x_2[-n]}_{y_2[n]}$

so it's linear

4. shifted output: $y_0[n] = y[n-m] = x[-n+m]$

or of a shifted input: $y_1[n] = x[-n-m]$
 $x_1[n] = x[n-m]$

since $y_0[n] \neq y_1[n]$, it's time variant

5. has memory since it depends on past values as n increases

e. $T(x[n]) = x[n] + 2u[n+1]$

1. assume $|x[n]| \leq \mu_x < \infty \forall n$

$|y[n]| = |x[n] + 2u[n+1]| = |x[n]| + |2u[n+1]| \rightarrow 1$

$|x[n]| + 2 = \mu_x + 2 < \infty$ so it's stable

2. causal since $x[n]$ only depends on present inputs

3. $\alpha x_1[n] + \beta x_2[n] \mapsto \underbrace{\alpha x[n]}_{\text{only part of } y_1} + \underbrace{\beta x[n]}_{y_2} + 2u[n+1]$

not linear

4

4. shifted output: $y_0[n] = y[n-m] = x[n-m] - 2u[n-m+1]$
 op of a shifted input:

$$y_1[n] = x_1[n] + 2u[n+1] = x_1[n-m] + 2u[n+1] \neq y_0[n]$$

so time variant

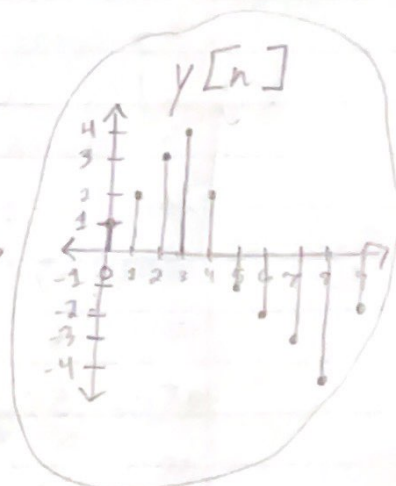
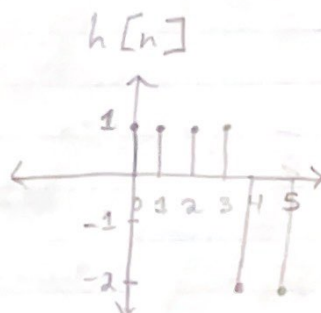
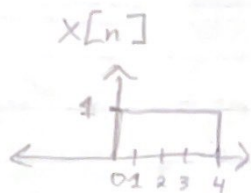
5. memoryless since $x[n]$ only depends on present inputs

5

3.

$$h[n] = \begin{cases} 0, & n < 0 \\ 1, & n = 0, 1, 2, 3 \\ -2, & n = 4, 5 \\ 0, & n > 5 \end{cases}$$

$$x[n] = u[n] - u[n-4]$$



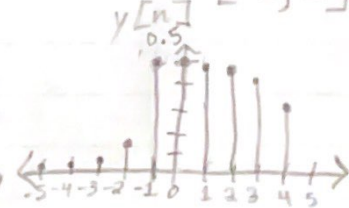
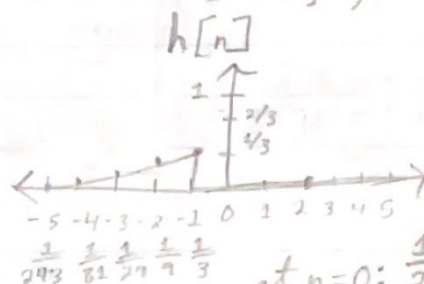
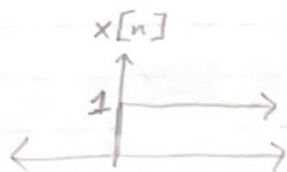
- shift 0: 1
 shift 1: 1+1=2
 shift 2: 1+1+1=3
 shift 3: 1+1+1+1=4
 shift 4: 1+1+1+1-2=2
 shift 5: 1+1+1-2-2=-1
 shift 6: 1+1-2-2=-2
 shift 7: 1-2-2=-3
 shift 8: -2-2=-4
 shift 9: -2

4.

$$x[n] = u[n]$$

$$h[n] = a^n u[-n-1], \quad a = 3$$

$$n = [-5, 5]$$



at $n = -5$: $\frac{1}{243} = 0.0041$

at $n = -4$: $\frac{1}{243} + \frac{1}{81} = 0.0164$

at $n = -3$: $\frac{1}{243} + \frac{1}{81} + \frac{1}{27} = 0.0534$

at $n = -2$: $\frac{1}{243} + \frac{1}{81} + \frac{1}{27} + \frac{1}{9} = 0.165$

at $n = -1$: $\frac{1}{243} + \frac{1}{81} + \frac{1}{27} + \frac{1}{9} + \frac{1}{3} = 0.5$

at $n = 0$: $\frac{1}{243} + \frac{1}{81} + \frac{1}{27} + \frac{1}{9} + \frac{1}{3} = 0.5$

at $n = 1$: $\frac{1}{81} + \frac{1}{27} + \frac{1}{9} + \frac{1}{3} = 0.4939$

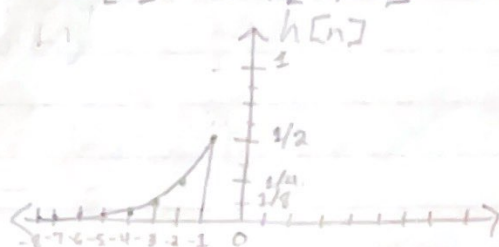
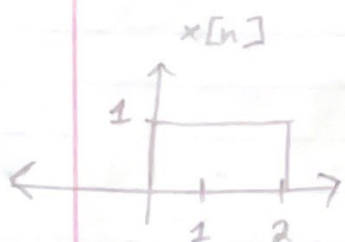
at $n = 2$: $\frac{1}{27} + \frac{1}{9} + \frac{1}{3} = 0.4815$

at $n = 3$: $\frac{1}{9} + \frac{1}{3} = 0.444$

at $n = 4$: $\frac{1}{3} = 0.333$

6

5. $x[n] = u[n] - u[n-2]$ $h[n] = 2^n u[-n-1]$ $n = [-8, 8]$



at $n = -8$: $\frac{1}{256} = 0.0039$ at $n = -2$: $\frac{1}{8} + \frac{1}{4} = 0.375$

at $n = -7$: $\frac{1}{256} + \frac{1}{128} = 0.0117$ at $n = -4$: $\frac{1}{4} + \frac{1}{2} = 0.75$

at $n = -6$: $\frac{1}{128} + \frac{1}{64} = 0.0234$ at $n = 0$: $\frac{1}{2} = 0.5$

at $n = -5$: $\frac{1}{64} + \frac{1}{32} = 0.0469$

at $n = -4$: $\frac{1}{32} + \frac{1}{16} = 0.0938$

at $n = -3$: $\frac{1}{16} + \frac{1}{8} = 0.1875$

