

MATHEMATICAL QUESTIONS

Question 1

First-order hold (FOH) performs a linear interpolation to generate the sampled signal

$$y(t) = \sum_{k=-\infty}^{\infty} x(kT_s) \Lambda\left(\frac{t - kT_s}{T_s}\right)$$

from the band-limited signal $x(t)$ with bandwidth W .

(a) Find the spectrum of the sampled signal $Y(f)$.

We know that $\mathcal{F}[\Lambda(\frac{t}{T_s})] = T_s \text{sinc}^2(T_s f)$. So,

$$y(t) = \sum_{k=-\infty}^{\infty} x(kT_s) \Lambda\left(\frac{t - kT_s}{T_s}\right) = \Lambda\left(\frac{t}{T_s}\right) * \sum_{k=-\infty}^{\infty} x(kT_s) \delta(t - kT_s) = \Lambda\left(\frac{t}{T_s}\right) * x_\delta(t)$$

$$\mathcal{F}[y(t)] = Y(f) = \mathcal{F}[\Lambda(\frac{t}{T_s})] \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(f - \frac{k}{T_s})$$

$$Y(f) = \text{sinc}^2(T_s f) \sum_{k=-\infty}^{\infty} X(f - \frac{k}{T_s})$$

(b) Propose a condition on the sampling period T_s for the perfect reconstruction of the original signal $x(t)$ from the sampled signal $y(t)$.

To avoid aliasing, Nyquist condition should be held, i.e. $\frac{1}{T_s} > 2W$.

(c) Obtain an expression for the reconstruction filter $H(f)$.

Clearly,

$$X(f) = X_p(f)H(f), \quad X_p(f) = P(f)X_\delta(f)$$

$$H(f) = \frac{1}{\mathcal{F}[\Lambda(\frac{t}{T_s})]} T_s \Pi\left(\frac{f}{2W'}\right)$$

, which yields

$$H(f) = \frac{\Pi(\frac{f}{2W'})}{\text{sinc}^2(T_s f)}$$

, where

$$W < W' < \frac{1}{T_s} - W$$

Question 2

The frequency-domain sampling theorem says that if $x(t)$ is a time-limited signal such that $x(t) = 0$ for $|t| \geq T$, then $X(f)$ is completely determined by its sample values $X(nf_0)$ with $f_0 \leq 1/2T$. Prove this theorem.

$$X_\delta(f) = \sum_{n=-\infty}^{\infty} X(nf_0)\delta(f - nf_0) = X(f) \sum_{n=-\infty}^{\infty} \delta(f - nf_0)$$

Taking the inverse Fourier transform,

$$x_\delta(t) = \mathcal{F}^{-1}\{X_\delta(f)\} = x(t) * \left[\frac{1}{f_0} \sum_{n=-\infty}^{\infty} \delta(t - \frac{n}{f_0}) \right] = \frac{1}{f_0} \sum_{n=-\infty}^{\infty} x(t - \frac{n}{f_0})$$

Now, if $T \leq \frac{1}{f_0} - T \Rightarrow f_0 \leq \frac{1}{2T}$,

$$x(t) = x_\delta(t) f_0 \text{rect}\left(\frac{t}{2T'}\right), \quad T \leq T' \leq \frac{1}{f_0} - T$$

Or equivalently,

$$X(f) = X_\delta(f) * [2T' f_0 \text{sinc}(2T' f)]$$

Question 3

The analog signal $x(t)$ which takes the values of $[-x_m, +x_m]$ uniformly passes a uniform midrise quantizer with $N = 2^\nu$ levels. Find the SQNR of the quantizer.

If the interval $[-x_m, +x_m]$ is covered by the even number of quantization levels $N = 2^\nu$, the quantization width equals $\Delta = \frac{2x_m}{2^\nu}$. The quantization interval $[-x_m + k\Delta, -x_m + (k+1)\Delta]$ is mapped to the quantized value $-x_m + k\Delta + \Delta/2$. Hence,

$$\begin{aligned} E\{(X - Q(X))^2\} &= \int_{-x_m}^{x_m} (x - Q(x))^2 f_X(x) dx \\ &= \sum_{k=0}^{2^\nu-1} \int_{-x_m+k\Delta}^{-x_m+(k+1)\Delta} [x - (-x_m + k\Delta + \Delta/2)]^2 \frac{1}{2x_m} dx \\ &= \sum_{k=0}^{2^\nu-1} \int_{-x_m+k\Delta}^{-x_m+(k+1)\Delta} [x + x_m - k\Delta - \Delta/2]^2 \frac{1}{2x_m} dx \\ &= \frac{1}{2x_m} \sum_{k=0}^{2^\nu-1} \int_{-\Delta/2}^{\Delta/2} u^2 du = \frac{2^\nu}{2x_m} \int_{-\Delta/2}^{\Delta/2} u^2 du = \frac{1}{\Delta} \int_{-\Delta/2}^{\Delta/2} u^2 du = \frac{1}{\Delta} \frac{2\Delta^3}{24} = \frac{\Delta^2}{12} \end{aligned}$$

$$E\{X^2\} = \int_{-x_m}^{x_m} x^2 f_X(x) dx = \int_{-x_m}^{x_m} x^2 \frac{1}{2x_m} dx = \frac{1}{2x_m} 2 \frac{1}{3} x_m^3 = \frac{x_m^2}{3} = \frac{\Delta^2}{12} 2^{2\nu}$$

$$\text{SQNR} = \frac{E\{X^2\}}{E\{(X - Q(X))^2\}} = \frac{\frac{\Delta^2}{12} 2^{2\nu}}{\frac{\Delta^2}{12}} = 2^{2\nu}$$

Question 4

The lowpass signal $x(t)$ with a bandwidth of W is sampled with a sampling interval of T_s and the signal

$$x_p(t) = \sum_{n=-\infty}^{\infty} x(nT_s)p(t - nT_s)$$

is reconstructed from the samples, where $p(t)$ is an arbitrary-shaped pulse (not necessarily time limited to the interval $[0, T_s]$).

(a) Find the Fourier transform of $x_p(t)$.

We have

$$x_p(t) = \sum_{n=-\infty}^{\infty} x(nT_s)p(t - nT_s) = p(t) * \sum_{n=-\infty}^{\infty} x(nT_s)\delta(t - nT_s) = p(t) * x_\delta(t)$$

Applying the Fourier Transform to both sides and using the convolution property, we get

$$X_p(f) = P(f)X_\delta(f) = \frac{P(f)}{T_s} \sum_{n=-\infty}^{\infty} X(f - \frac{n}{T_s})$$

(b) Find the conditions for perfect reconstruction of $x(t)$ from $x_p(t)$.

In order to avoid aliasing $\frac{1}{T_s} > 2W$. Furthermore, the spectrum $P(f)$ should be invertible for $|f| < W$.

(c) Determine the required reconstruction filter.

Considering that the conditions of perfect reconstruction are satisfied, we have

$$X(f) = X_p(f)H(f) = X_p(f) \left[\frac{T_s}{P(f)} \cap \left(\frac{f}{2W'} \right) \right]$$

, where $W \leq W' \leq \frac{1}{T_s} - W$.

Question 5

The lowpass signal $x(t)$ with a bandwidth of W is sampled at the Nyquist rate and the signal

$$y(t) = \sum_{n=-\infty}^{\infty} (-1)^n x(nT_s)\delta(t - nT_s)$$

is generated.

(a) Find the Fourier transform of $y(t)$.

$$\begin{aligned}
 y(t) &= \sum_{n=-\infty}^{\infty} (-1)^n x(nT_s) \delta(t - nT_s) = x(t) \sum_{n=-\infty}^{\infty} (-1)^n \delta(t - nT_s) \\
 &= x(t) \left[\sum_{n=-\infty}^{\infty} \delta(t - 2nT_s) - \sum_{n=-\infty}^{\infty} \delta(t - T_s - 2nT_s) \right] \\
 \Rightarrow Y(f) &= \frac{1}{2T_s} X(f) * \left[\sum_{n=-\infty}^{\infty} \delta\left(f - \frac{n}{2T_s}\right) - \sum_{n=-\infty}^{\infty} \delta\left(f - \frac{n}{2T_s}\right) e^{-j2\pi f T_s} \right] \\
 &= \frac{1}{2T_s} \left[\sum_{n=-\infty}^{\infty} X\left(f - \frac{n}{2T_s}\right) - \sum_{n=-\infty}^{\infty} X\left(f - \frac{n}{2T_s}\right) (-1)^n \right] \\
 &= \frac{1}{T_s} \sum_{n=-\infty}^{\infty} X\left(f - \frac{2n+1}{2T_s}\right) = \frac{1}{T_s} \sum_{n=-\infty}^{\infty} X\left(f - \frac{n}{T_s} - \frac{1}{2T_s}\right)
 \end{aligned}$$

To have the spectrum copies separated, we should we,

$$\frac{1}{2T_s} + W \leq \frac{3}{2T_s} - W \Rightarrow 2W \leq \frac{1}{T_s}$$

, where W is the bandwidth of the signal $X(t)$.

(b) Can $x(t)$ be reconstructed from $y(t)$ by using a linear time-invariant system? Why?

The answer is NO. Here a copy of the spectrum of the original signal is not located at $f = 0$. So, we cannot separate the spectrum of the original signal using an LTI system which can shift the spectrum of its input.

(c) Can $x(t)$ be reconstructed from $y(t)$ by using a linear time-varying system? How?

Yes. First we use a bandpass filter to isolate $X_1(f) = X(f - \frac{1}{2T_s}) + X(f + \frac{1}{2T_s})$. Then, mix the filtered signal with $\cos(2\pi \frac{1}{2T_s} t)$ that gives

$$\frac{1}{2} [X_1(f + \frac{1}{2T_s}) + X_1(f - \frac{1}{2T_s})] = \frac{1}{2} [X(f) + X(f + \frac{2}{2T_s}) + X(f) + X(f - \frac{2}{2T_s})]$$

Applying a lowpass filter with bandwidth $W' \in [W, \frac{1}{T_s} - W]$, the original spectrum of $X(f)$ appears.

Question 6

A stationary source is distributed according to the triangle probability density function $f_X(x) = 0.5\Lambda(0.5x)$. This source is quantized using the four-level uniform quantizer

$$Q(x) = \begin{cases} 1.5, & 1 < x \leq 2 \\ 0.5, & 0 < x \leq 1 \\ -0.5, & -1 < x \leq 0 \\ -1.5, & -2 \leq x \leq -1 \end{cases}$$

Determine the probability density function of the random variable representing the quantizer error $X - Q(X)$.

Clearly, for $|e| > 0.5$, $f_E(e) = 0$. For $|e| \leq 0.5$,

$$\begin{aligned} F_E(e) &= P(E \leq e) = P(1 < X < 1.5 + e) + P(0 < X < 0.5 + e) \\ &\quad + P(-1 < X < -0.5 + e) + P(-2 < X < -1.5 + e) \\ &= F_X(1.5 + e) - F_X(1) + F_X(0.5 + e) - F_X(0) \\ &\quad + F_X(-0.5 + e) - F_X(-1) + F_X(-1.5 + e) - F_X(-2) \end{aligned}$$

$$f_E(e) = \frac{dF_E(e)}{de} = f_X(1.5 + e) + f_X(0.5 + e) + f_X(-0.5 + e) + f_X(-1.5 + e)$$

So,

$$f_E(e) = \frac{1}{2} \left[\Lambda\left(\frac{e + \frac{3}{2}}{2}\right) + \Lambda\left(\frac{e + \frac{1}{2}}{2}\right) + \Lambda\left(\frac{e - \frac{1}{2}}{2}\right) + \Lambda\left(\frac{e - \frac{3}{2}}{2}\right) \right] = 1$$

Finally,

$$f_E(e) = \begin{cases} 0, & |e| > 0.5 \\ 1, & |e| \leq 0.5 \end{cases}$$

SOFTWARE QUESTIONS

Question 7

A quantizer with 2^ν quantized levels working over the input range $[-1, 1]$ is fed with a zero-mean Gaussian random variable having the variance σ^2 . Develop a MATLAB/Python code to calculate the signal to quantization noise ratio when the quantization intervals are uniformly distributed and when the quantization intervals are nonuniformly distributed according to A-law companding method with the parameter A . Discuss the results for different values of ν , σ^2 , and A . Feel free to plot any suitable curve to better describe the observations.

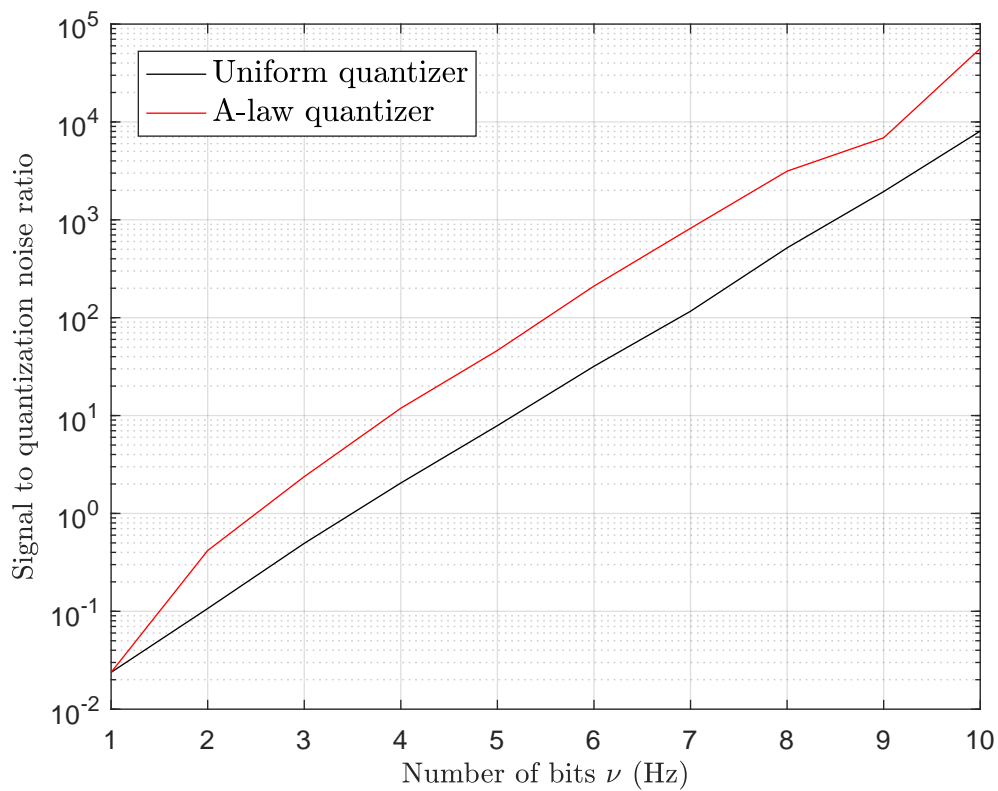


Figure 1: SQNR versus number of bits ν for $\sigma^2 = 0.01$ and $A = 9$.

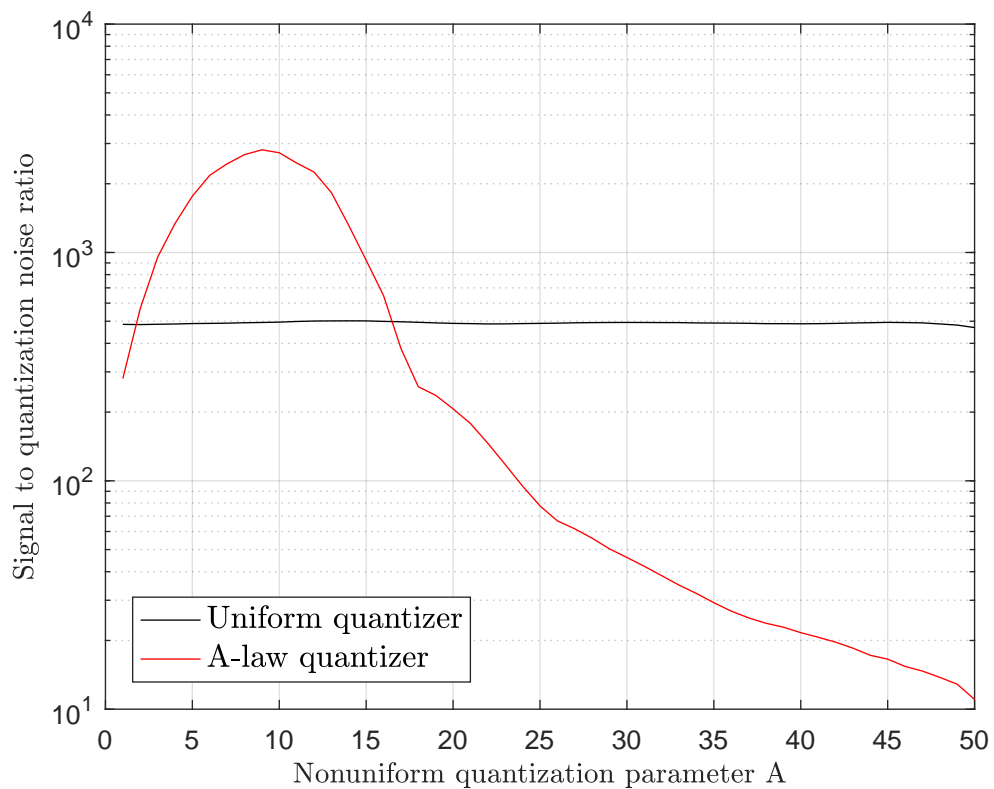


Figure 2: SQNR versus A -law parameter A for $\sigma^2 = 0.01$ and $\nu = 8$.

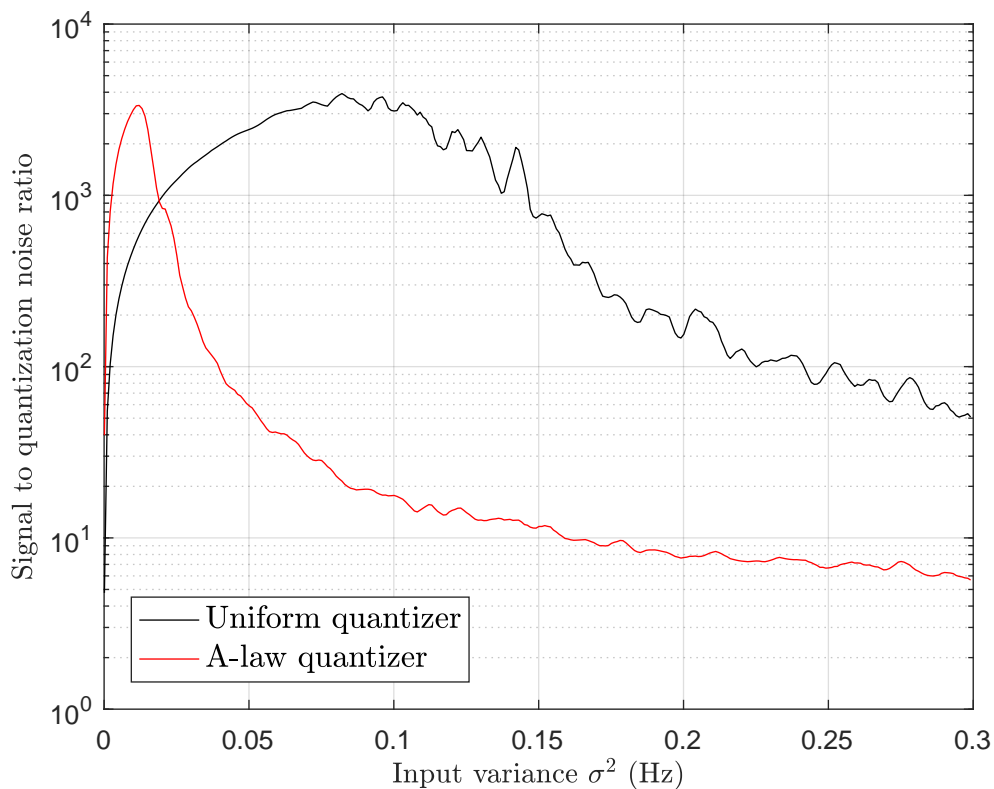


Figure 3: SQNR versus input variance σ^2 for $A = 9$ and $\nu = 8$.

The uniform quantizer can be implemented as.

```
1 function [value, level, bit] = uniform_quantizer(x, nu)
2 % quantization intervals
3 a=linspace(-1,1,2^nu+1);
4 a(1)=-Inf;
5 a(end)=[];
6 % quantization level
7 level=sum(x>=a)-1;
8 % quantization value
9 a(1)=-1;
10 value = a(level+1);
11 % quantization bit
12 bit=de2bi(bin2gray(level, 'pam', 2^nu), nu, 'left-msb');
13 end
```

A possible implementation of the A -law quantizer is

```
1 function [value, level, bit] = alaw_quantizer(x, nu, A)
2 % quantization intervals
3 a=linspace(-1,1,2^nu+1);
4 a(1)=-Inf;
5 a(end)=[];
6 b=zeros(size(a));
7 for m=1:(2^nu)
8     if (abs(a(m)) < 1/abs(1+log(a(m))))
9         b(m) = (abs(a(m))*(1+log(A))) / A.*sign(a(m));
10     else
11         b(m) = exp(-1+(abs(a(m))*(1+log(A)))) / A.*sign(a(m));
12     end
13 end
14 % quantization level
15 level=sum(x>=b)-1;
16 % quantization value
```

```

17 b(1) = -1;
18 value = b(level+1);
19 % quantization bit
20 bit=de2bi(bin2gray(level, 'pam', 2^nu), nu, 'left-msb');
21 end

```

The following mfile calls the quantizers and computes their signal to quantization noise ratio.

```

1 %%
2 clear
3 clc
4 close all
5 % settings
6 sigma = sqrt(0.01);
7 A = 9;
8 nu = 8;
9 N = 100;
10
11 % average noise calculation
12 uniform_err = 0;
13 Alaw_err = 0;
14 for n=1:N
15     x = sigma*randn;
16     uniform_err = ((n-1)*uniform_err + (x-uniform_quantizer(x,nu))^2)/n;
17     Alaw_err = ((n-1)*Alaw_err + (x-alaw_quantizer(x,nu,A))^2)/n;
18 end
19
20 % snr calculation
21 uniform_snr = sigma^2/uniform_err;
22 Alaw_snr = sigma^2/Alaw_err;
23
24 % snr gain
25 Alaw_snr/uniform_snr

```

As you can see in Fig. 1, increasing the number of bits ν improves SQNR. The SQNR of the A -law quantizer can be 7 times the SQNR of the uniform quantizer. According to Fig. 2, the SQNR of the A -law quantizer is maximized for a suitable value of A . Finally, if the input variance is high, meaning that the input of the quantizer is not concentrated around zero, the A -law quantizer may work worse than the uniform quantizer, as can be seen in Fig. 3.

BONUS QUESTIONS

Question 8

The DCT of an $N \times N$ picture with luminance function $x(m, n), 0 \leq m, n \leq N-1$ can be obtained as

$$X(0,0) = \frac{1}{N} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} x(k,l)$$

$$X(u,v) = \frac{2}{N} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} x(k,l) \cos \left[\frac{(2k+1)u\pi}{2N} \right] \cos \left[\frac{(2l+1)v\pi}{2N} \right], \quad u, v \neq 0$$

The $X(0, 0)$ coefficient is usually called the DC component, and the other coefficients are called the AC components. Find the DCT of a constant picture having $x(m, n) = C, 0 \leq m, n \leq N - 1$.

For the DC component,

$$X(0, 0) = \frac{1}{N} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} C = NC$$

And for the AC component,

$$\begin{aligned} X(u, v) &= \frac{2}{N} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} x(k, l) \cos \left[\frac{(2k+1)u\pi}{2N} \right] \cos \left[\frac{(2l+1)v\pi}{2N} \right], \quad u, v \neq 0 \\ &= \frac{2C}{N} \sum_{k=0}^{N-1} \cos \left[\frac{(2k+1)u\pi}{2N} \right] \sum_{l=0}^{N-1} \cos \left[\frac{(2l+1)v\pi}{2N} \right] \end{aligned}$$

But $\sum_{l=0}^{N-1} \cos \left[\frac{(2l+1)v\pi}{2N} \right] = 0$ since for integer v

$$\begin{aligned} \sum_{l=0}^{N-1} \cos \left[\frac{(2l+1)v\pi}{2N} \right] &= \sum_{l=0}^{N-1} \Re \{ e^{j \frac{(2l+1)v\pi}{2N}} \} \\ &= \Re \{ e^{j \frac{v\pi}{2N}} \sum_{l=0}^{N-1} e^{j \frac{lv\pi}{N}} \} = \Re \{ e^{j \frac{v\pi}{2N}} \frac{1 - (e^{j \frac{v\pi}{N}})^N}{1 - e^{j \frac{v\pi}{N}}} \} = \Re \{ e^{j \frac{v\pi}{2N}} \frac{1 - e^{jv\pi}}{1 - e^{j \frac{v\pi}{N}}} \} \\ &= \Re \{ e^{j \frac{v\pi}{2N}} \frac{e^{j \frac{v\pi}{2}} e^{-j \frac{v\pi}{2}} - e^{j \frac{v\pi}{2}}}{e^{j \frac{v\pi}{2N}} e^{-j \frac{v\pi}{2N}} - e^{j \frac{v\pi}{2N}}} \} = \Re \{ e^{j \frac{v\pi}{2}} \frac{\sin(\frac{v\pi}{2})}{\sin(\frac{v\pi}{2N})} \} = \cos\left(\frac{v\pi}{2}\right) \frac{\sin(\frac{v\pi}{2})}{\sin(\frac{v\pi}{2N})} = \frac{\sin(2v\pi)}{2 \sin(\frac{v\pi}{2N})} = 0 \end{aligned}$$

So, the AC component is zero.

Question 9

Return your answers by filling the \LaTeX template of the assignment.