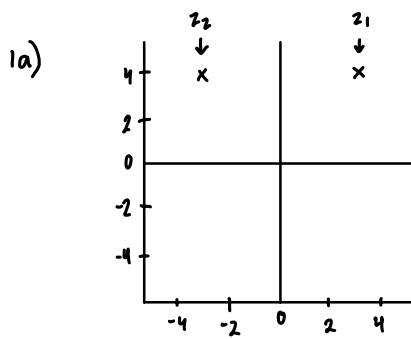


Complex Numbers and Signals



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HW#1

$$1b) z_1 + z_2 = (3 + 4j) + (-3 + 4j) = 8j$$

$$1c) z_1 - z_2 = (3 + 4j) - (-3 + 4j) = 6$$

$$1d) z_1 \cdot z_2 = (3 + 4j) \cdot (-3 + 4j) = -25$$

$$1e) z_1/z_2 = \frac{(3+4j)(-3-4j)}{(-3+4j)(-3-4j)} = \frac{-25-24j}{25} = 0.28 - 0.96j$$

$$1f) |z_1| = \sqrt{9+16} = 5$$

$$1g) |z_2| = \sqrt{9+16} = 5$$

$$1h) \phi_1 = \arg(z_1) = \tan^{-1}\left(\frac{4}{3}\right) = \approx 0.92 \text{ radians}$$

$$1i) \phi_2 = \arg(z_2) = \tan^{-1}\left(\frac{4}{-3}\right) = \approx 2.214 \text{ radians}$$

$$1j) z_1 = 5 \angle 0.927 \text{ radians}$$

$$z_2 = 5 \angle 2.214 \text{ radians}$$

$$2) \cos(2\pi ft) :$$

$$\cos(2\pi ft) = \frac{e^{2\pi ift} + e^{-2\pi ift}}{2}$$

$$e^{2\pi ift} = \cos(2\pi ft) + j\sin(2\pi ft)$$

$$e^{-2\pi ift} = \cos(2\pi ft) - j\sin(2\pi ft)$$

$$e^{2\pi ift} + e^{-2\pi ift} = 2\cos(2\pi ft)$$

$$\text{So, } \cos(2\pi ft) = \frac{e^{2\pi ift} + e^{-2\pi ift}}{2}$$

$$\sin(2\pi ft) :$$

$$\sin(2\pi ft) = \frac{e^{2\pi ift} - e^{-2\pi ift}}{2j}$$

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

$$e^{-2\pi i f t} = \cos(2\pi f t) - j \sin(2\pi f t)$$

$$e^{2\pi i f t} - e^{-2\pi i f t} = 2j \sin(2\pi f t)$$

$$\sin(2\pi f t) = \frac{e^{2\pi i f t} - e^{-2\pi i f t}}{2j}$$

$$3a) \cos(A) \cos(B) = \frac{1}{2} [\cos(A-B) + \cos(A+B)]$$

$$A = 2\pi f_1 t, \quad B = 2\pi f_2 t - \phi$$

$$P = v_1(t) v_2(t) = 4 \cos(2\pi f_1 t) \cdot 4 \cos(2\pi f_2 t - \phi)$$

$$P = 16 \cdot \frac{1}{2} [\cos((2\pi f_1 t) - (2\pi f_2 t - \phi)) + \cos((2\pi f_1 t) + (2\pi f_2 t - \phi))]]$$

$$(2\pi f_1 t) - (2\pi f_2 t - \phi) = 2\pi(f_1 - f_2)t + \phi$$

$$(2\pi f_1 t) + (2\pi f_2 t - \phi) = 2\pi(f_1 + f_2)t - \phi$$

Thus,

$$P = 8 [\cos(2\pi(f_1 - f_2)t + \phi) + \cos(2\pi(f_1 + f_2)t - \phi)]$$

P is a sum of two cosines:

- $f_+ = f_1 + f_2$ and $f_- = f_1 - f_2$
- phase shifts give an offset of 2ϕ

3b) If $\phi = 0$ and $f_1 = f_2$, P becomes:

$$P = 8 [\cos(2\pi(f_1 - f_2)t) + \cos(2\pi(f_1 + f_2)t)]$$

Since $f_1 = f_2$:

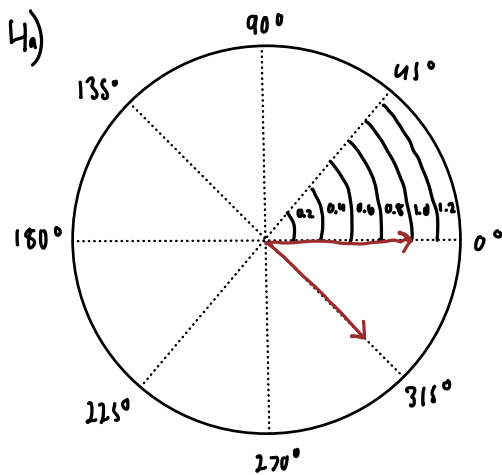
$$f_1 - f_2 = 0, \quad f_1 + f_2 = 2f_1$$

So:

$$P = 8 [\cos(0) + \cos(2\pi(2f_1)t)]$$

$$\cos(0) = 1, \quad P = 8 [1 + \cos(4\pi f_1 t)]$$

$$P_{max} = 8 \cdot (1+1) = 16$$



$$\sqrt{2}(1) = 1$$

$$v_1(t) = \exp(-j\psi)$$

$$4b) \quad z = e^{j0} + e^{-j\phi} = 1 + e^{-j\phi}$$

$$e^{-j\phi} = \cos(\phi) - j\sin(\phi)$$

$$z = 1 + \cos(\phi) - j\sin(\phi)$$

$$\text{magnitude} = |z| = \sqrt{(1 + \cos(\phi))^2 + (-\sin(\phi))^2}$$

$$\angle z = \tan^{-1}\left(\frac{-\sin(\phi)}{1 + \cos(\phi)}\right)$$

for $\phi = 0^\circ$:

• magnitude of the resultant signal is approximately 1.85

• the phase angle is -22.5°

4c) when $\phi = 0^\circ$:

$$e^{-j\phi} = 1, \text{ so } z = 1 + 1 = 2$$

$$\text{magnitude} = 2$$

$$\text{phase} = 0$$

when $\phi = 180^\circ$:

$$e^{-j\phi} = -1, \text{ so } z = 1 - 1 = 0$$

$$\text{magnitude} = 0$$

phase = undefined (zero amplitude)

3 Probability and Statistics, Noise

1(a)

This distribution is a bell-shaped curve, symmetric around $x=0$. The Octave code generates 10,000 random samples using the `randn` function, which also follows a standard normal distribution. The histogram closely matches the Gaussian curve, confirming that the samples generated by `randn` follow the expected probability density function. The histogram bars approximate the values of $p(x)$ for different bins, while the Gaussian curve provides the theoretical continuous distribution.

Octave code:

```
% gaussian curve (Eq. 6)
x = linspace(-4, 4, 1000); % range of x values
sigma = 1; % stdev
mu = 0; % mean
p_x = (1 / sqrt(2 * pi * sigma^2)) * exp(-((x - mu).^2) / (2 * sigma^2));

% generate random samples
samples = randn(10000, 1); % random samples from standard normal distribution

% plot gaussian curve
figure(1);
plot(x, p_x, 'LineWidth', 2);
hold on;

% plot histogram of samples
histogram(samples, 30, 'Normalization', 'pdf', 'FaceAlpha', 0.6);
title('Gaussian Distribution vs Histogram of Samples');
xlabel('x');
ylabel('Probability Density');
legend('Gaussian Distribution (Eq. 6)', 'Histogram of Samples');
grid on;
```

1(c)

Octave code:

```
% parameters for sine wave
```

```

fs = 1000; % sampling frequency
t = linspace(0, 1, fs); % time vector (1 second duration)
A = 1; % amplitude of sine wave
f = 5; % frequency of sine wave in Hz

% generate sine wave
sine_wave = A * sin(2 * pi * f * t);

% add gaussian noise
sigma = 0.5; % standard deviation of noise
noise = normrnd(0, sigma, size(t)); % gaussian noise with mean 0 and std 0.5
noisy_signal = sine_wave + noise;

% plot sine wave and noisy signal
figure(3);
plot(t, sine_wave, 'LineWidth', 2);
hold on;
plot(t, noisy_signal, 'LineWidth', 1);
title('Sine Wave with Added Gaussian Noise');
xlabel('Time (s)');
ylabel('Amplitude');
legend('Original Sine Wave', 'Sine Wave with Gaussian Noise');
grid on;

```

(2)

Octave Code:

```

% parameters
num_samples = 10000; % number of random numbers
num_sums = 12; % number of uniform random numbers to sum

% generate random numbers and compute sums
uniform_random_numbers = rand(num_samples, num_sums); % uniform random numbers
sums = sum(uniform_random_numbers, 2); % the sum across rows

% normalize the sums to create a standard gaussian
sums = (sums - mean(sums)) / std(sums);

% plot histogram of normalized sums

```

```

figure(1);
histogram(sums, 30, 'Normalization', 'pdf', 'FaceAlpha', 0.6);
hold on;

% plot theoretical gaussian curve for comparison
x = linspace(-4, 4, 1000);
p_x = (1 / sqrt(2 * pi)) * exp(-x.^2 / 2);
plot(x, p_x, 'LineWidth', 2);

% labels and title
title('Gaussian Distribution from Summed Uniform Random Numbers');
xlabel('Value');
ylabel('Probability Density');
legend('Summed Uniform Numbers', 'Gaussian Curve');
grid on;

```

4 ADC and DAC

```

% parameters
f = 5; % signal frequency in hz
T = 1; % duration of the signal in seconds
fs1 = 50; % high sampling frequency (10 times signal frequency)
fs2 = 10; % sampling frequency equal to 2 times signal frequency
fs3 = 7; % sampling frequency less than 2 times signal frequency

% generate time vectors for different sampling frequencies
t1 = 0:1/fs1:T; % high sampling rate
t2 = 0:1/fs2:T; % nyquist rate
t3 = 0:1/fs3:T; % below nyquist rate

% generate sine waves
sine_wave1 = sin(2 * pi * f * t1);
sine_wave2 = sin(2 * pi * f * t2);
sine_wave3 = sin(2 * pi * f * t3);

% plot sine waves
figure(1);
subplot(3, 1, 1);
stem(t1, sine_wave1, 'filled');

```

```
title('high sampling rate (fs >> 2f)');  
xlabel('time (s)');  
ylabel('amplitude');  
grid on;
```

```
subplot(3, 1, 2);  
stem(t2, sine_wave2, 'filled');  
title('nyquist sampling rate (fs = 2f)');  
xlabel('time (s)');  
ylabel('amplitude');  
grid on;
```

```
subplot(3, 1, 3);  
stem(t3, sine_wave3, 'filled');  
title('below nyquist sampling rate (fs < 2f)');  
xlabel('time (s)');  
ylabel('amplitude');  
grid on;
```

When the sampling frequency is much higher than twice the signal frequency ($f_s \gg 2f$), the signal graph accurately represents the sine wave with smooth transitions between the sampled points, allowing for precise reconstruction. At the Nyquist rate ($f_s = 2f$), the signal is still well sampled, but the number of points is minimal, which leads to a less smooth but still accurate representation. However, when the sampling frequency is below the Nyquist rate ($f_s < 2f$), aliasing occurs, causing the signal graph to appear distorted. This distortion results in a misrepresentation of the original sine wave, making accurate reconstruction impossible.