

Theory Problems

1. Consider a receiver with noise power -160 dBm (within the signal bandwidth of interest). Assume a simplified path-loss model with $d_0 = 1$ m, and K obtained from the free-space model path loss formula (given below with omnidirectional antennas) and $f_c = 1$ GHz, and $\gamma = 4$. For a transmitter power of $P_t = 10$ mW, find the maximum distance between the transmitter and receiver such that the received signal-to-noise power ratio is 20dB. The received power for the simplified path-loss model is given by the following:

$$P_r = P_t K \left[\frac{d_0}{d} \right]^\gamma$$

$$P_r \text{ dBm} = P_t \text{ dBm} + K \text{ dB} - 10\gamma \log_{10} \left[\frac{d}{d_0} \right]$$

where P_r is the received power, P_t is the transmit power, d is the distance in meters, d_0 is a reference distance for the antenna far field, and γ is the path-loss exponent with K given below:

$$K = \left(\frac{\lambda}{4\pi d_0} \right)^2$$

where λ is the wavelength of the carrier signal given by

$$\lambda = \frac{c}{f_c}$$

and c is the speed of light $c \approx 3.0 \times 10^8$ m/s. Values for γ depend on the environment and are given in the table below.

Environment	γ range
Urban macrocells	3.7-6.5
Urban microcells (up to 35km)	2.7-3.5
Office building (same floor)	1.6-3.5
Office building (multiple floors)	2-6
Store	1.8-2.2
Factory	1.6-3.3
Home	3

Table 1: Typical path-loss exponents

$$\lambda = (3 * 10^8 \text{ m/s}) / (1 * 10^9 / \text{s}) = 0.3 \text{ m}$$

$$K = (0.3 \text{ m} / 4\pi d_0)^2$$

$$\text{For } 20\text{dB SNR } P_r = -140\text{dBm} = 10^{-14} \text{ mW}$$

$$10^{-14} \text{ mW} = 10 \text{ mW} * (0.3 \text{ m} / 4\pi * 1^2)^2 \left[\frac{1 \text{ m}}{d(\text{m})} \right]^4$$

$$d^4 = (10^{-14} / 10^{-14}) * (0.3 \text{ m} / 4\pi)^2$$

$$= 10^{-15} * (0.3 \text{ m} / 4\pi)^2$$

$$d = 868.87 \text{ m}$$

2. Phase noise

A circularly gaussian probability density function $f(x, y)$ can be written as

$$f(x, y) = \left[\frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\langle x \rangle)^2 / 2\sigma^2} \right] \left[\frac{1}{\sqrt{2\pi}\sigma} e^{-(y-\langle y \rangle)^2 / 2\sigma^2} \right], \quad (1)$$

where $\langle x \rangle^2$ and $\langle y \rangle^2$ are the squares of the expected values of the independent gaussian

random variables x and y that have the same variance σ^2 . Let $A = \sqrt{\langle x \rangle^2 + \langle y \rangle^2}$.

Using the following change of variables, $x = A \cos \theta$, $y = A \sin \theta$, $dx dy = dr r d\theta$, $r = \sqrt{x^2 + y^2}$

$x = r \cos \phi$, $y = r \sin \phi$, and redefining ϕ with respect to the phase θ of the constant amplitude signal so that $\theta - \phi$ is replaced by ϕ . Using these substitutions (1) becomes

$$f_{r\phi}(r, \phi) = \frac{r}{2\pi\sigma^2} e^{-(r^2 - 2Ar \cos \phi + A^2) / 2\sigma^2}$$

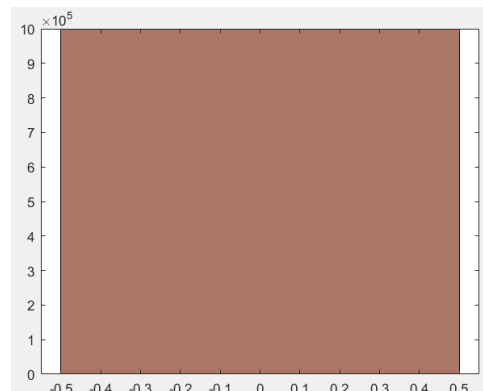
This is a circularly symmetric gaussian random variable in polar coordinates. The marginal phase distribution $f_\phi(\phi)$ is of joint distribution is (see Eq. 2.2.35 in the text)

$$f_\phi(\phi) = \frac{1}{2\pi} \left(e^{-F} + \sqrt{\pi F} \cos \phi e^{-F \sin^2 \phi} \left(1 + \text{erf} \left(\sqrt{F} \cos \phi \right) \right) \right)$$

where $F = A^2 / 2\sigma^2$, and erf is the error function.

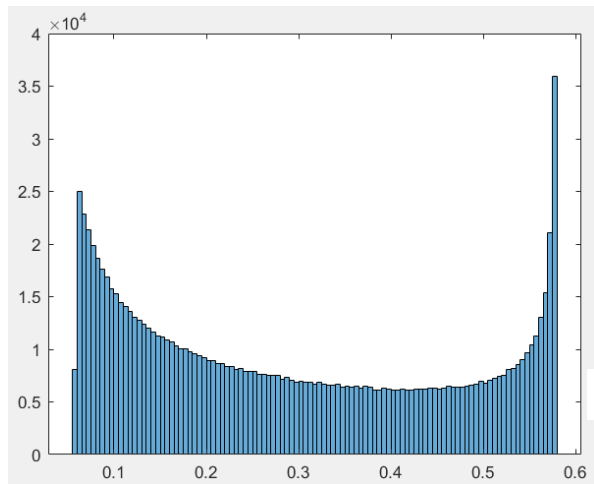
a) Plot $f_\phi(\phi)$ for the following values of F :

i. $F = 0$ ($\sigma = 2$, $u_x = u_y = 0$, x, y 1M samples each case)



```
sigma =2;
mux = sqrt(0)*sigma;
muy = sqrt(0)*sigma;
A = sqrt(mux^2+muy^2)
x = normrnd( mux, sigma, 1, 1000000);
y = normrnd( muy, sigma, 1, 1000000);
```

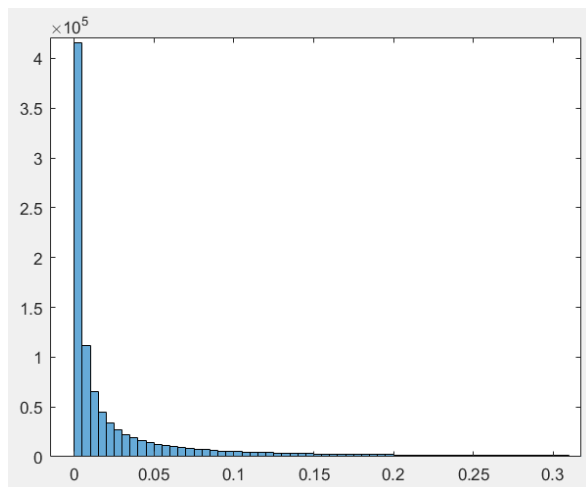
ii. $F = 1$ ($u_x=u_y=\sigma=2$, x, y 1M samples each case)



```
sigma =2;
mux = sqrt(1)*sigma;
muy = sqrt(1)*sigma;
A = sqrt(mux^2+muy^2)
x = normrnd( mux, sigma, 1, 1000000);
y = normrnd( muy, sigma, 1, 1000000);
```

(radian)

iii. $F = 10$ ($u_x=u_y=\sqrt{10}\sigma$, x, y 1M samples each case)



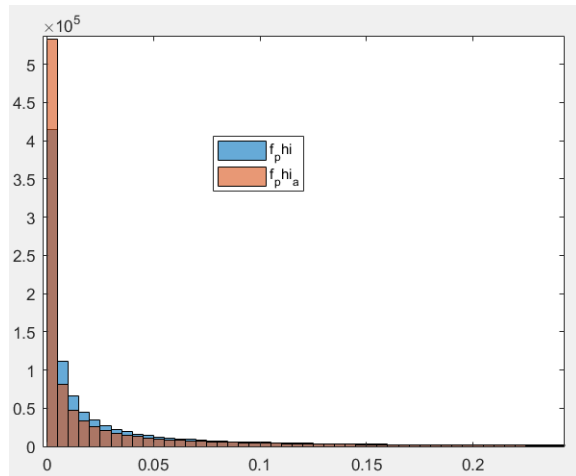
```
sigma =2;
mux = sqrt(10)*sigma;
muy = sqrt(10)*sigma;
A = sqrt(mux^2+muy^2)
x = normrnd( mux, sigma, 1, 1000000);
y = normrnd( muy, sigma, 1, 1000000);
```

(radian)

b) On the same curve as the plot for $F = 10$, plot the following approximate expression

$$f_{\phi}(\phi) \approx \sqrt{\frac{F}{\pi}} e^{-F\phi^2},$$

$F = 10$ ($u_x=u_y=\sqrt{10}\sigma$, x, y 1M samples each case)



```
sigma =2;
mux = sqrt(10)*sigma;
muy = sqrt(10)*sigma;
A = sqrt(mux^2+muy^2)
x = normrnd( mux, sigma, 1, 1000000);
y = normrnd( muy, sigma, 1, 1000000);
```

:approximation has higher at zero and lower around near zero phases. i.e., it's more optimistic phase noise approximation.

3. Fading

Using the analytic expression for a Rayleigh fade (Slide 17 Lecture 4a)

$$\bar{P}_e = \frac{1}{2} \left(1 - \sqrt{\frac{z}{1+z}} \right)$$

determine the slope on log-log scale for large values of z .

a) Compare this slope to the slope to the unfaded case and comment. (See Figure on Slide 18 of Lecture 4a.) You may solve this problem numerically.

<Rayleigh-Faded case>

$$\log \bar{P}_e = \log(1/2) + \log\left(1 - \sqrt{\frac{z}{1+z}}\right) = \log(1/2) + \log\left((1+z)^{\frac{1}{2}} - z^{\frac{1}{2}}\right) - \log(1+z)^{\frac{1}{2}}$$

$$\begin{aligned} \frac{d}{dz} \log \bar{P}_e &= \frac{d}{dz} \left(\log(1/2) + \log\left((1+z)^{\frac{1}{2}} - z^{\frac{1}{2}}\right) - \log(1+z)^{\frac{1}{2}} \right) \\ &= 0 + \frac{d}{dz} \log\left(z(1+z)^{\frac{1}{2}} - z^{\frac{1}{2}}\right) - \frac{1}{2} \frac{d}{dz} \log(1+z) \\ &\cong -\frac{1}{2} \frac{d}{dz} \log(z) = -\frac{1}{2} \frac{d}{dz} \ln(z) / \ln(10) = (1/4.6) * \ln(1/z) \quad , \text{ with large } z \end{aligned}$$

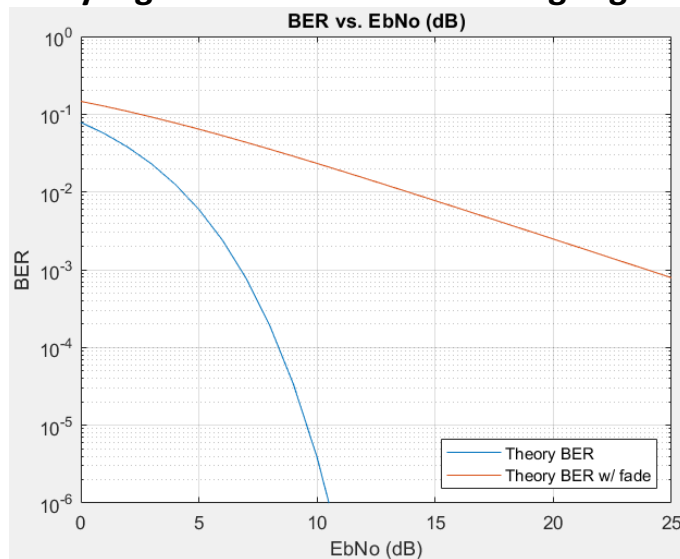
< unfaded case>

$$\begin{aligned} \bar{P}_e &= \frac{1}{2} \operatorname{erfc} \left(\left(\frac{Eb}{No} \right)^{\frac{1}{2}} \right) \\ &= \int_{\left(\frac{Eb}{No} \right)^{\frac{1}{2}}}^{\infty} \left(\frac{1}{\sqrt{2\pi}} e^{-r'^2/2} \right) dr', \quad r' = (r + A) / \sigma \end{aligned}$$

$$\begin{aligned}
\log \bar{P}_e &= \int_{A/\sigma}^{\infty} \left(\log \left(\frac{1}{\sqrt{2\pi}} \right) - \frac{r'^2}{2} \right) dr' \\
\frac{d}{dz} \log \bar{P}_e &= \int_{\sqrt{z}}^{\infty} \frac{d}{dz} \frac{r'^2}{2} dr' \\
&= 2 \int_{\sqrt{z}}^{\infty} r' \frac{dr'}{dz} dr' \\
&= 2 \int_{\sqrt{z}}^{\infty} \left(\frac{r}{\sigma} + z^{\frac{1}{2}} \right) \left(\frac{r}{2\sqrt{\sigma}} z + z^{-\frac{3}{2}} \right)^2 dz \\
&\cong 2 \int_{\sqrt{z}}^{\infty} z^{\frac{1}{2}} z^{-3} dz, \quad \text{with large } z \\
&= 2 \int_{\sqrt{z}}^{\infty} z^{-\frac{5}{2}} dz = -4/3 [z^{-\frac{3}{2}}]_{\sqrt{z}}^{\infty} = 4/3 (z^{-\frac{3}{4}})
\end{aligned}$$

Log-scaled Rayleigh faded P_e is linearly inverse proportional to $\log(E_b/N_0)$ while unfaded P_e is non-linearly more inverse proportional to E_b/N_0 as follows

< Rayleigh faded vs. unfaded in log-log scale >



```

z_db = 0:26;
z = 10 .^ (z_db/10);
theory_bpsk = 0.5*erfc(sqrt(z));
theory_bpsk_fade = 0.5*(1 - sqrt( z ./ (z
+ 1 ) ) );

figure(1)
semilogy( z_db, theory_bpsk, z_db,
theory_bpsk_fade);

grid on;
xlabel('EbNo (dB)');
ylabel('BER');
title('BER vs. EbNo (dB)');
legend('Theory BER','Theory BER w/
fade','location','SouthEast');
axis([0 26-1 1e-6 1]);

```

4. Multipath Fading and Equalization

Consider a received sequence of signal values $y(n)$ for a transmitted sequence of signal values $x(n)$. The received sequence $y(n)$ is the summation of multiple scaled and delayed copies of $x(n)$.

$$y(n) = \sum_{k=0}^{L-1} \alpha_k x(n-k)$$

The values of $\alpha_k = |\alpha_k| e^{j\phi(k)}$ are in general complex.

- a) Find the z-transform $Y(z)$ of $y(n)$. (You may need to review Z-transforms.)

$$\begin{aligned} y(n) &= \sum_{k=0}^{L-1} \alpha_k x(n-k) = \alpha_0 x(n) + \alpha_1 x(n-1) + \dots + \alpha_{L-1} x(n-L+1) \\ &= x(n) \sum_{k=0}^{L-1} \alpha_k \delta(n-k) \\ Y(z) &= \sum_{n=0}^N \sum_{k=0}^{L-1} \alpha_k x(n-k) z^{-n} \\ &= \sum_{n-k=0}^{N-k} \sum_{k=0}^{L-1} \alpha_k x(n-k) z^{-(n-k)} z^{-k} \\ &= X(z) \sum_{k=0}^{L-1} \alpha_k z^{-k} = X(z) \alpha(z) \\ &= X(z) [\alpha_0 + \alpha_1 z^{-1} + \dots + \alpha_{L-1} z^{-L+1}] \\ &= X(z) [\alpha_0 z^{L-1} + \alpha_1 z^{L-2} + \dots + \alpha_{L-1}] / z^{L-1} \end{aligned}$$

- b) Find the transfer function of the channel $H(z)$ by writing $Y(z)$ in terms of $X(z)$ (i.e.

$$\begin{aligned} y(n) &= x(n) \sum_{k=0}^{L-1} \alpha_k x(n-k) \\ Y(z) &= X(z) \sum_{k=0}^{L-1} \alpha_k z^{-k} \\ H(z) &= \sum_{k=0}^{L-1} \alpha_k z^{-k} \\ &= \alpha_0 + \alpha_1 z^{-1} + \dots + \alpha_{L-1} z^{-L+1} \\ &= [\alpha_0 z^{L-1} + \alpha_1 z^{L-2} + \dots + \alpha_{L-1}] / z^{L-1} \end{aligned}$$

- c) Answer the following questions about the impulse response $h(n)$ of the channel:

- i. Is the impulse response $h(n)$ linear?

From the above transfer function we can tell channel is linear. Gain or loss to input signal get's the same gain or loss after the transfer function. It's just like different harmonics with different coefficients.

- ii. Is $h(n)$ an FIR (finite-impulse-response) or IIR (infinite-impulse-response) filter?

IIR filter as the linear channel response is like FIR filter and it's inverse makes IIR.

- d) In general, $H(z)$ causes unwanted distortion in the received sequence of signal values $y(n)$. Suppose we wish to "undo" the effects of $H(z)$ by "equalizing" $Y(z)$. To do so, we want to find a equalization filter $G(z)$ such that

$$G(z) Y(z) = X(z)$$

Find the form of the equalization filter $G(z)$ in terms of the transfer function $H(z)$

$$G(z) = X(z)/Y(z) = 1/H(z) = (1 - z^{-1}) / (\sum_{k=0}^{L-1} \alpha_k z^k)$$

e) Answer the following questions about the impulse response $g(n)$ of the equalization filter:

i. Is the impulse response $g(n)$ linear?

Yes. As it's just inverse of the linear FIR filter.

ii. Is $g(n)$ an FIR (finite-impulse-response) or IIR (infinite-impulse-response) filter?

It's IIR filter with poles generated by inverse of FIR filter's zeros.

5. Two Component Fading Model and Equalization

Suppose you have the following difference equation for the received sequence of signal values

$$y(n): y(n) = a_0 x(n) + a_1 x(n - m)$$

$$m = 8$$

a) Find the z-transform for this specific form of $y(n)$, which consists of the "direct path" signal $x(n)$ and a single "image" path signal $x(n - m)$. The coefficients for each of these signals are $a_0 = 1$ and $a_1 = -A$ where $0 < A < 1$.

$$y(n) = x(n) - Ax(n - m)$$

$$Y(z) = \sum_{n=0}^{\infty} (x(n) - Ax(n - 8))z^{-n}$$

$$= X(z)(1 - Az^{-8})$$

b) Determine the transfer function $H(z)$ of the channel.

$$H(z) = 1 - Az^{-8} = (z^8 - Az^{-4})/z^4,$$

c) Evaluate $H(z)$ on the unit circle using ($|z| = |e^{jw}| = 1$) so that $z = e^{jw}$.

$$(* H(e^{jw}) = (2/e^{-j4w})(e^{j4w} - Ae^{-j4w})/2)$$

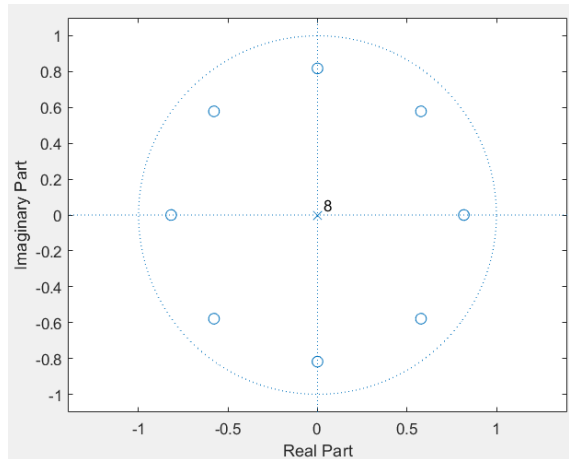
$$(\text{for simplicity if set } A = 1)$$

$$(H(e^{jw}) = (2/e^{-j4w})(e^{j4w} - e^{-j4w})/2 = 2e^{j4w} \sin(4w))$$

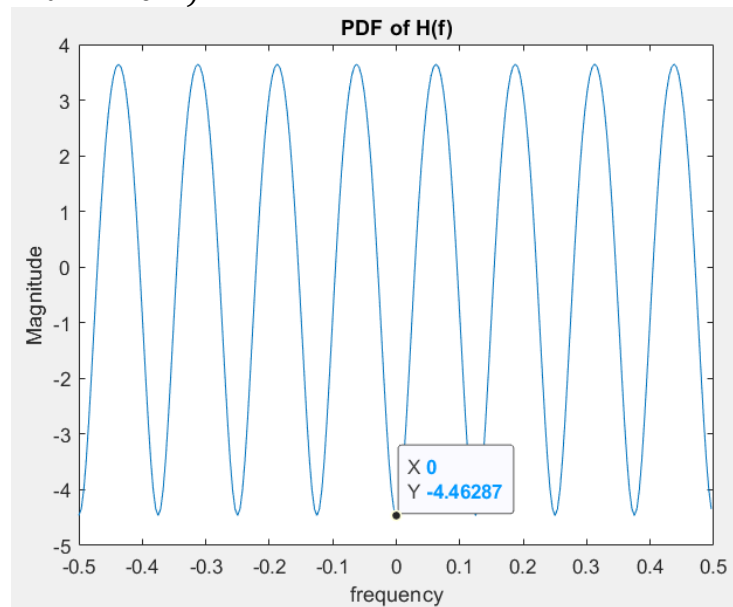
i. Sketch the pole-zeros plot for $H(z)$

$$H(z) = 1 - Az^{-8}$$

$$h(n) = [1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ -A] \text{ in MatLab. When set } A = 0.2$$



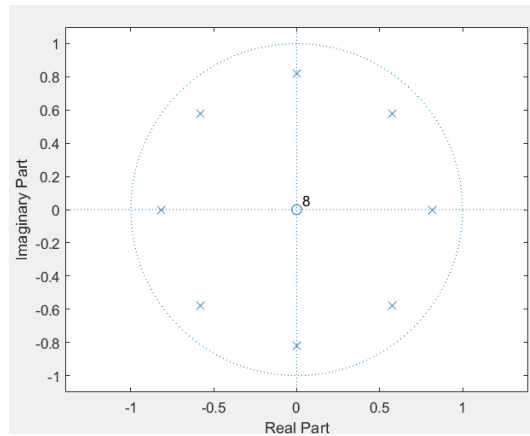
- ii. Sketch $|H(e^{j\omega})|^2$.
 (*with $A = 0.2$)



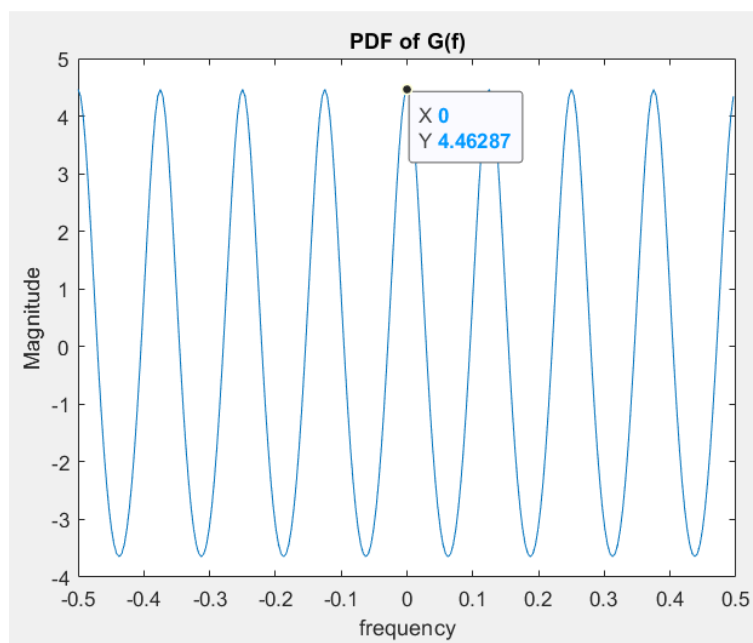
- d) Find an expression for the transfer function $G(e^{j\omega})$ of the equalization filter .

$$G(z) = 1/(1-Az^{-8}) = z^8/(z^8-A)$$

- i. Sketch the pole-zeros plot for $G(z)$
 $G(z) = 1/H(z)$, can be plotted by `zplane(1,H(z))`.

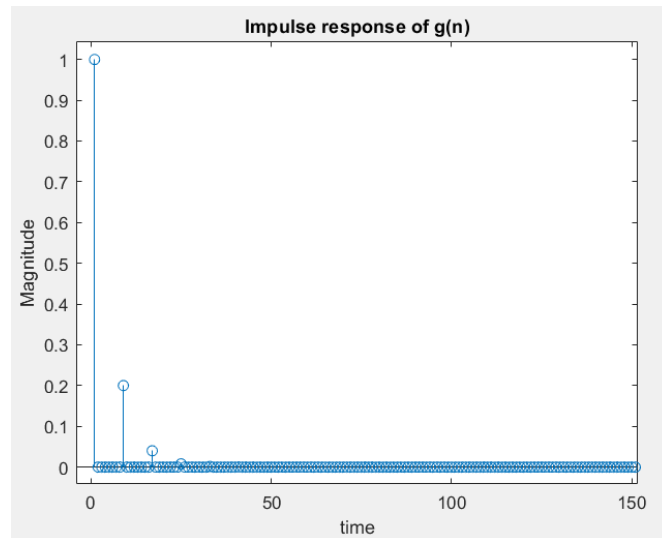


ii. Sketch $|G(e^{jw})|^2$.



e) Find an expression for the impulse response $g(n)$ of the equalization filter.

i. Sketch the impulse response $g(n)$



e) Comment on your results, how does A and m affect the frequency response of $H(e^{j\omega})$?

Find zero forcing equalizer works OK by $1/h(n)$.

A : affects the poles & zeros location. Smaller A makes poles and zeros close to $(0,0)$ while bigger A make them closer to circle of one.

n_0 : determines how many poles and zeros are in a Nyquist zone. Bigger n_0 makes more poles or zeros proportional to it.

1. Central Limit Theorem

In this matlab problem, you will be validating the central limit theorem in simulation. Understanding the central limit theorem will be useful for analyzing multipath fading channel models in Lab 4. A summary of the central limit theorem is as follows:

$$Y = \frac{1}{N} \sum_{i=1}^N X_i$$

For a set of random variables X_i that are i.i.d (independent identically distributed), the sum

random variable Y approaches a normal distribution. In the first part, we will be assuming

that the random variables X_i are distributed uniformly (Note: In actuality, we could use any type of random variable and the fundamental result won't change).

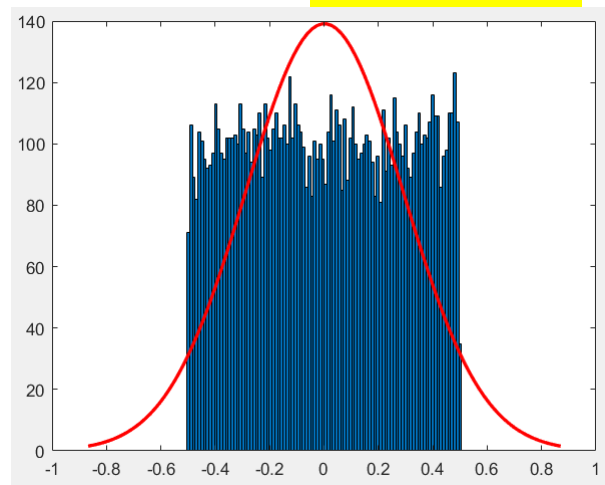
a) Generate a two dimensional random vector of uniform random variables $X \sim \mathcal{U}[-0.5, 0.5]$.

$X = \text{rand}(1000, 10000) - 0.5$

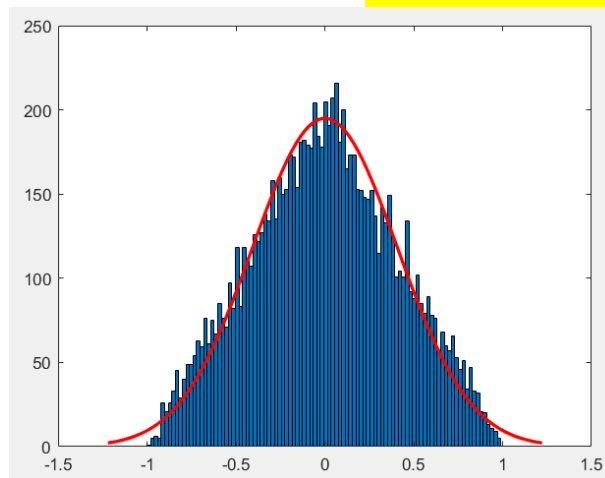
this will create a 1000 by 10000 matrix of uniform random numbers.

i. Plot the histogram (i.e. use the function `histfit(.)`) of one row of this random vector

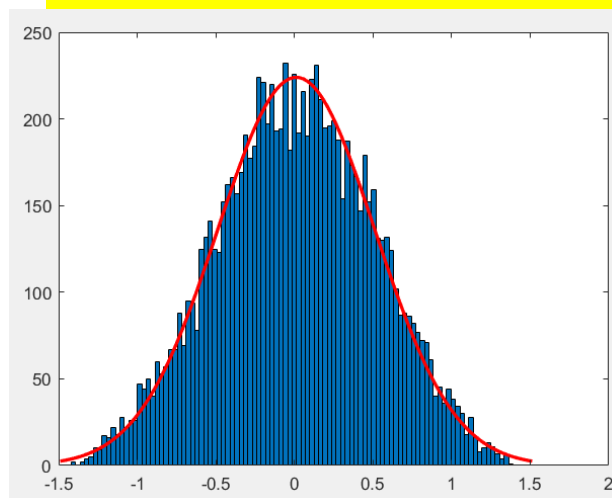
(i.e. $Y = X_1$). What kind of distribution is this? Uniform distribution.



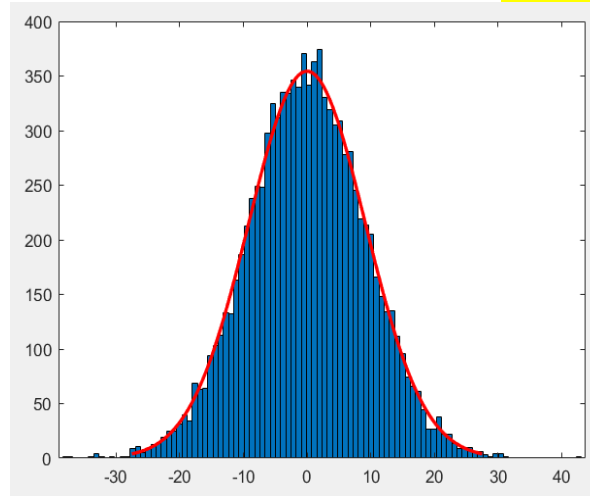
ii. Plot the histogram of the sum of any two rows of X (i.e. this matlab command models: $Y = X_1 + X_2$). What kind of distribution is this? Normal distribution.



iii. Plot the histogram of the sum of any three rows of X (i.e. this matlab command models: $Y = X_1 + X_2 + X_3$). More fit to normal distribution than the summation of two.



iv. Plot the histogram of the sum of all 1000 rows of X (i.e. this matlab command models: $Y = X_1 + X_2 + \dots + X_{1000}$). What kind of distribution this? **Normal distribution.**



v. Comment on your results (i.e. how well does the histogram conform to a normal distribution. And also, how fast does it converge?).

It converges very fast. It starts to show close to normal distribution from the summation of 2 X_i 's.

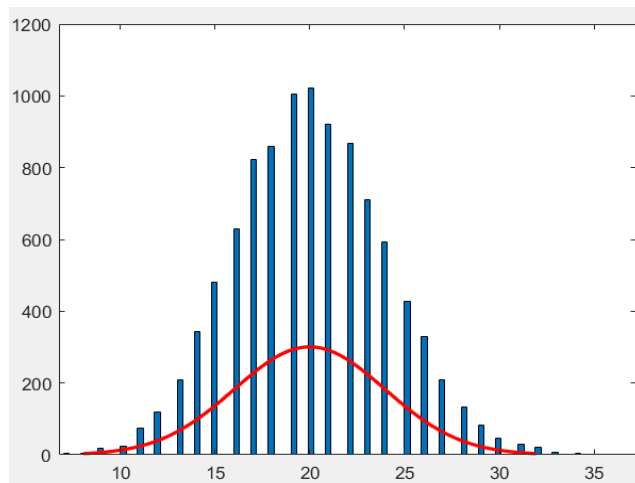
=====< used code>=====

```
X = rand(1000,10000)-0.5;
Y(1,:) = zeros(1,10000);
for n=1:1000
    Y(1,1:end) = Y(1,1:end)+X(n,1:1:end);
end
histfit(Y)
```

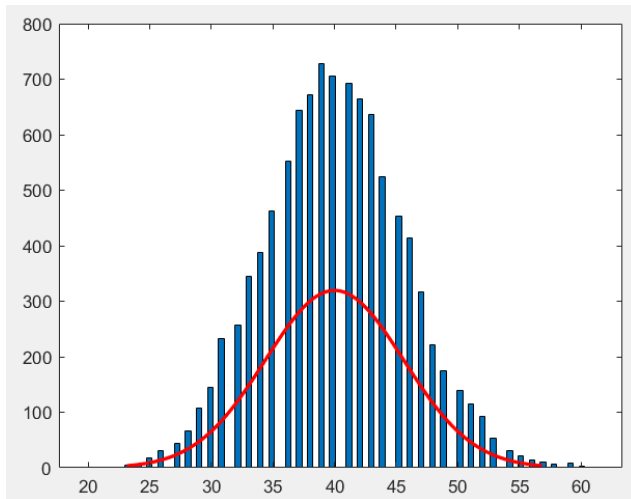
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b) Repeat part (a) for any random distribution of your choice (i.e. choose from *binornd*(.), *randsrc*(.), *round*(*rand*()), *raylrnd*(.), etc..).

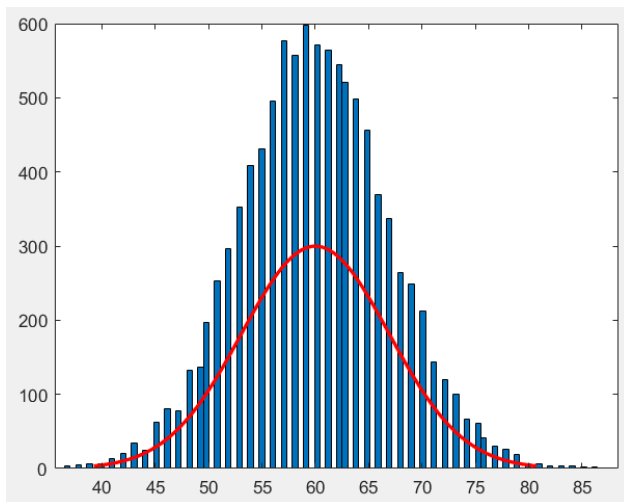
i.



ii.



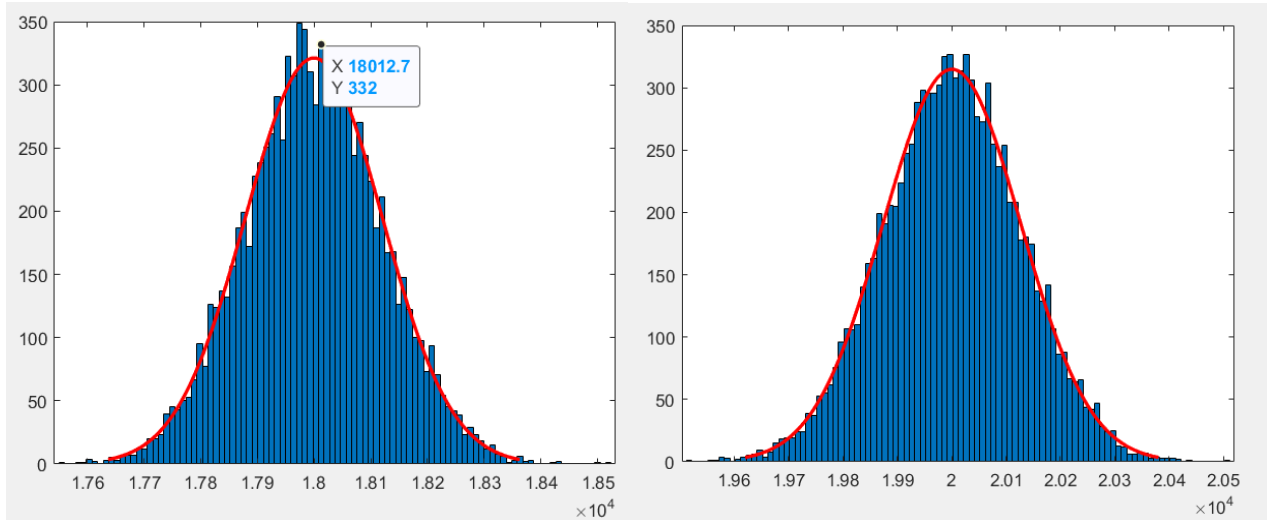
iii.



iv.

< summation of 900 Xi's >

< summation of 1000 Xi's >

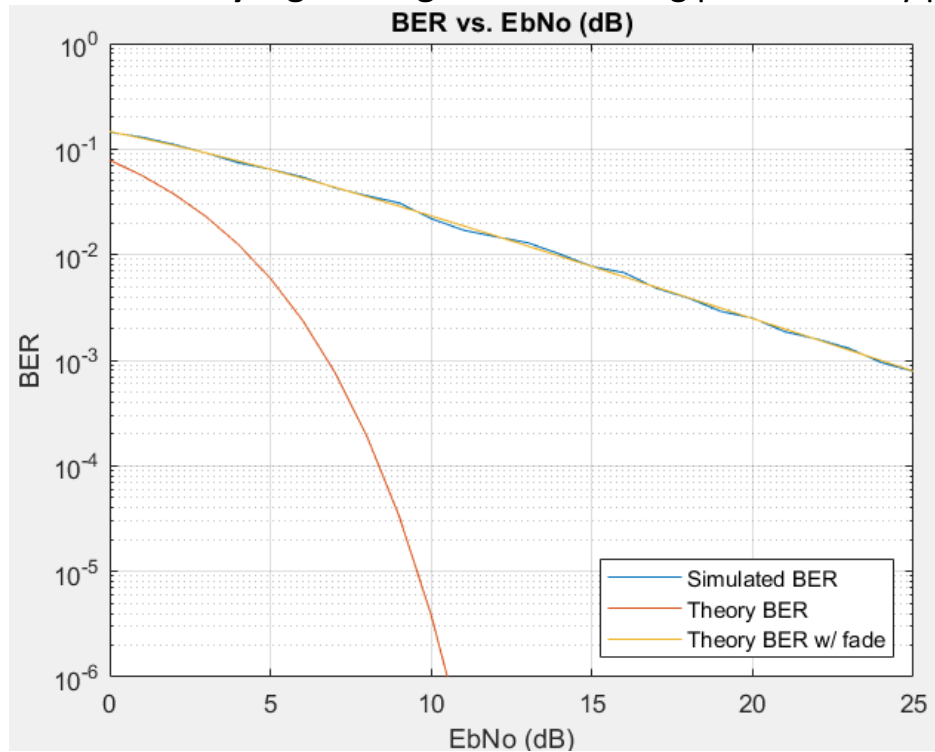


v.

It converges very slowly. It's not close to normal distribution even with summation of 900 X_i 's.

2. Fading simulation

< **BER with Rayleigh fading** >: added fading part to theory problem 3 code



3. Least Squares Channel Estimation

In this matlab exercise, you will be estimating the channel impulse response and performing

channel equalization on a vector of noisy received samples.

a) First, load the data file “p5_msg.mat”. Parameters of interest are as follows:

$$f_s = 20\text{MHz}$$

$$T_s = 1/f_s = 50\text{ns}$$

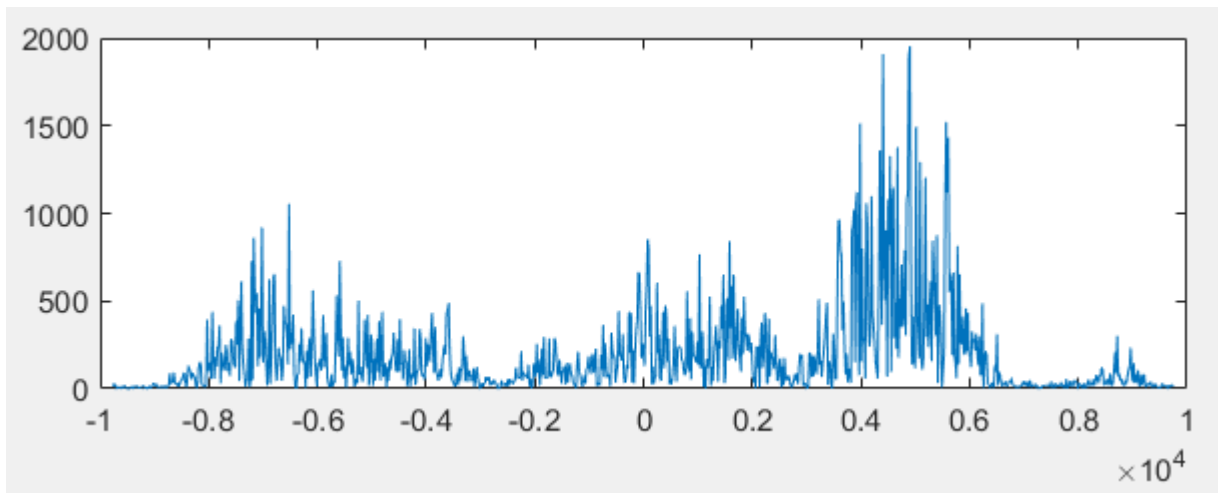
$$T_g = 1.6\mu\text{s}$$

$$T_t = 6.35\mu\text{s}$$

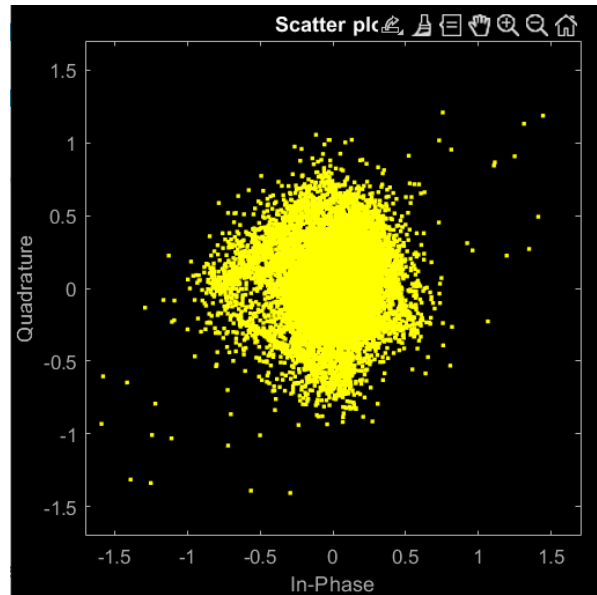
The message contained in variable y consists of a data packet sampled at $f_s = 20\text{MHz}$ that has been corrupted by a typical indoor fading channel (i.e. such as the WiFi channel). The packet consists of the following: a BPSK pseudo-noise preamble of length T_t followed by a guard interval of duration T_g and the modulated payload of length T_p . The order $M = 7$, PN sequence was generated using *pngen.m*, which can be found in the labs folder.

b) Since the data symbols have been corrupted by a fading channel, it is almost impossible to detect the symbols without applying any sort of equalization. Verify the effect of the channel response by plotting:

i. The power spectrum of the vector y .



ii. The constellation of vector y .



c) Using the linear model as discussed in *Lecture 4B* and given below:

$$\mathbf{r} = \mathbf{T}\mathbf{h} + \mathbf{n}$$

$$\mathbf{T} = \begin{bmatrix} t_0 & 0 & \dots & 0 \\ t_1 & t_0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ t_{2J} & t_{2J-1} & \dots & t_0 \\ \vdots & & & \vdots \\ t_{K-1} & t_{K-2} & \dots & t_{K-L} \end{bmatrix}.$$

$\mathbf{T}_{ij} = t_{i-j}$ for $i = 0, 1, \dots, K-1; j = 0, 1, \dots, L-1$.

find the least-squares estimate of the channel coefficients given by $\hat{\mathbf{h}}$:

$$\hat{\mathbf{h}} = \left(\mathbf{T}^\dagger \mathbf{T} \right)^{-1} \mathbf{T}^\dagger \mathbf{r}.$$

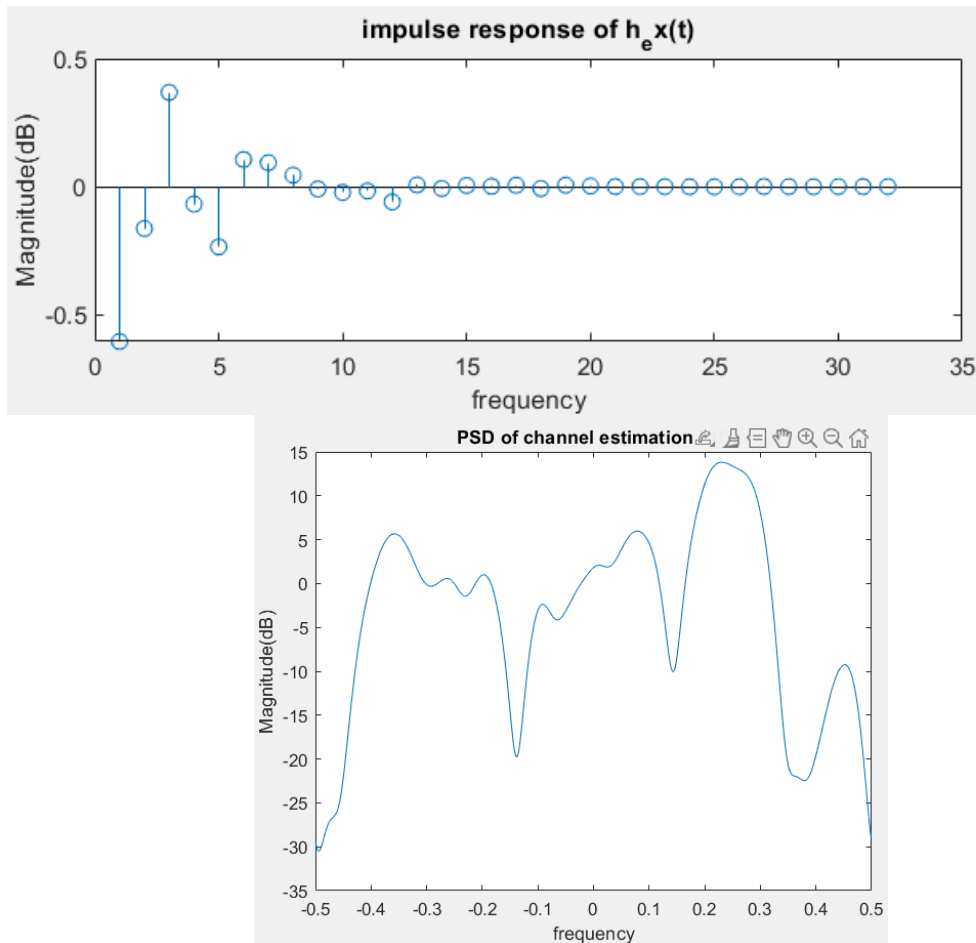
or

$$\hat{\mathbf{h}} = \text{pinv}(\mathbf{T}) \mathbf{r}$$

where $\text{pinv}(\cdot)$ denotes the pseudo-inverse. Note that, if \mathbf{T} is invertible, then $\text{pinv}(\mathbf{T}) = \mathbf{T}^{-1}$. You may assume that the guard interval T_g is equal to the length of the channel

response.

- i. Plot the time series $\hat{h}(n)$ and the power spectrum $\hat{H}(e^{j\omega})$.



d) While the previous approach gives us the least-squares estimate of the channel coefficients $h(n)$ it does not directly give us the coefficients of the equalization filter $g(n)$. One approach to finding the equalization filter coefficients $g(n)$ is by forming another linear equation which describes the convolution of the received samples \mathbf{r} and the equalization filter $g(n)$ as given below in matrix notation:

$$\mathbf{R}\mathbf{g} = \mathbf{t} + \mathbf{n}$$

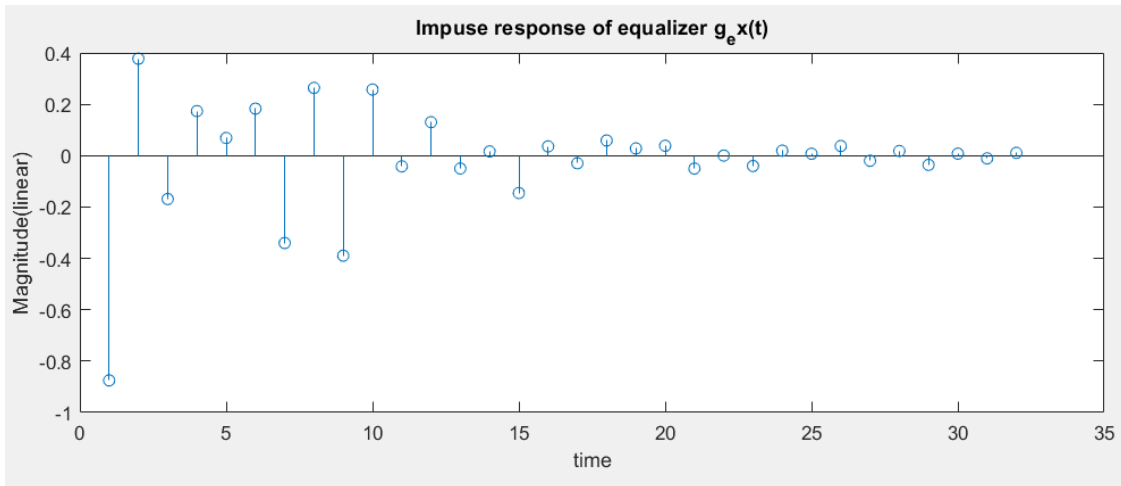
where \mathbf{g} is the equalization filter, \mathbf{R} is the matrix representation of the received signal vector \mathbf{r} and \mathbf{t} is training sequence vector.

- i. Find the least-squares estimate $\hat{\mathbf{g}}$, which is given by:

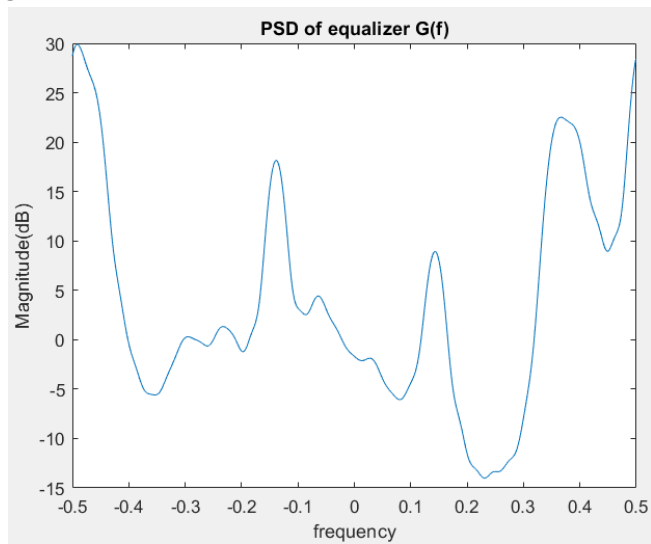
$$\hat{\mathbf{g}} = (\mathbf{R}^{\dagger}\mathbf{R})^{-1} \mathbf{R}^{\dagger}\mathbf{t}$$

or

$$\hat{\mathbf{g}} = \text{pinv}(\mathbf{R}) \mathbf{t}$$



- ii. Plot the time series $\hat{g}(n)$ and the power spectrum $\hat{G}(e^{j\omega})$.

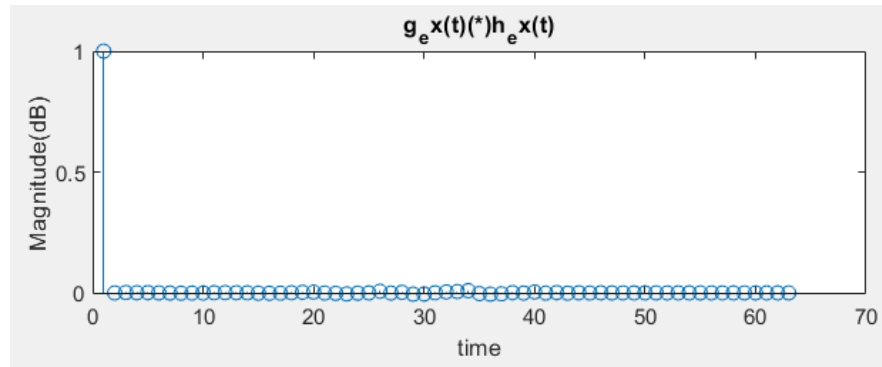


e) Recall that given a fading channel with impulse response $h(n)$ and frequency response $H(e^{j\omega})$, the goal of the equalization filter $g(n)$ is to remove the unwanted distortion caused by the channel $h(n)$. Therefore, one requirement for the ideal equalization filter is that:

$$h(n) * g(n) = \delta(n)$$

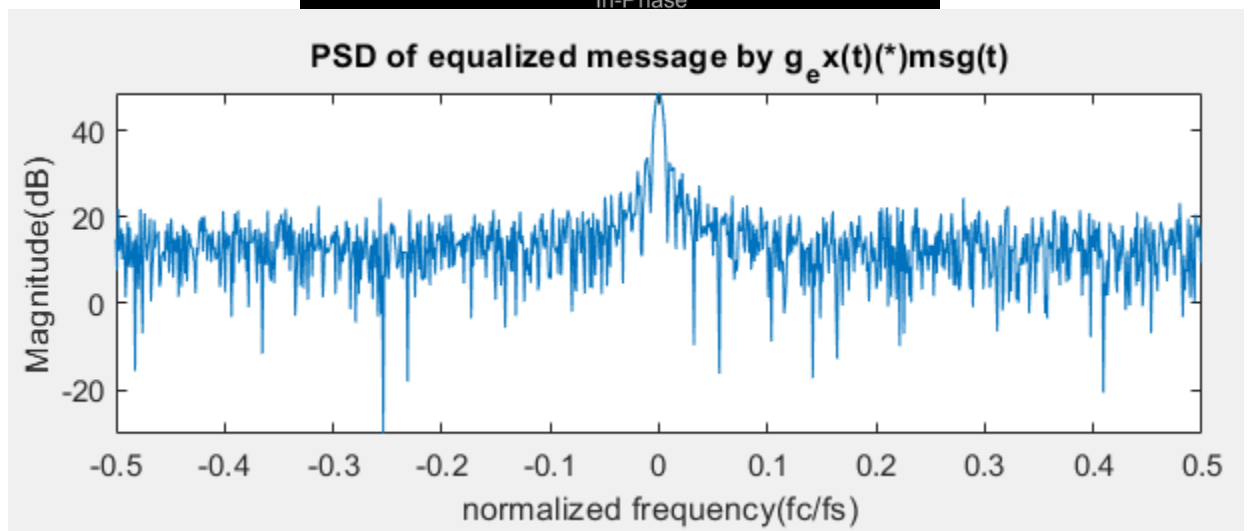
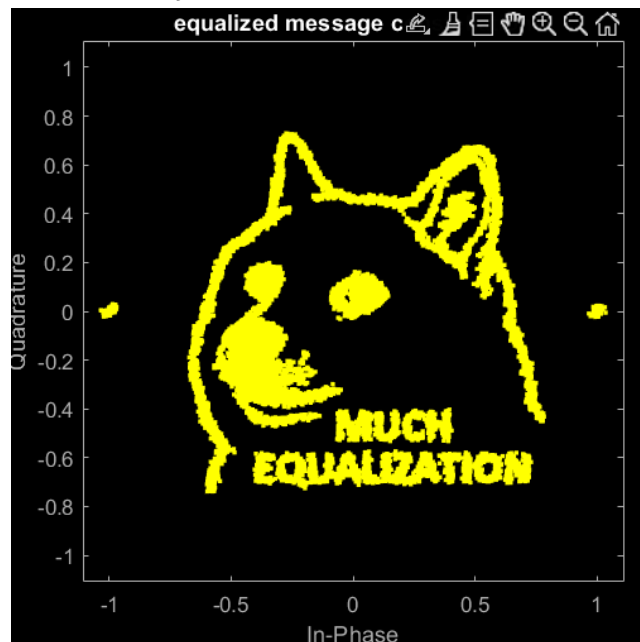
the convolution of $h(n)$ and $g(n)$ should give us a delta function $\delta(n)$.

- i. Plot the time-series $\hat{h}(n) * \hat{g}(n)$.



f) Finally, apply the equalization filter $\hat{g}(n)$ to the packet and verify that the equalization process has successfully removed the channel response $h(n)$.

- i. Plot the equalized constellation $y_{eq}(n)$ and the corresponding power spectrum $Y_{eq}(e^{j\omega})$.



- ii.

ii. Comment on your results (i.e. what type of modulation is used for the payload?
etc.).

: it's a dog or bear. As the constellation has big degree of freedom, it seems like big number of QAM is used.

6

<used code for theory problem 5>

```
fftlen = 256;
A = 0.2;
h_n = [ 1 0 0 0 0 0 0 0 -A];
[H_f,W]=freqz(h_n,1,fftlen,'whole');

figure(1)
zplane(h_n,1)

figure(2)
plot(-0.5:1/fftlen:0.5-1/fftlen,20*log(abs(H_f)))
title('PDF of H(f)')
ylabel('Magnitude')
xlabel('frequency')

figure(3)
zplane(1,h_n)

figure(4)
[G_f,W]=freqz(1, h_n, fftlen,'whole');
plot(-0.5:1/fftlen:0.5-1/fftlen, 20*log(abs(G_f)))
title('PDF of G(f)')
ylabel('Magnitude')
xlabel('frequency')

figure(5)
stem(g_n)
zplane(ifft(G_f,fftlen),1)

%%
fftlen = 1024;
n = 0:1:500;
fc = 1;
fs = fftlen/4;

figure(6)
[G_f,W]=freqz(1, h_n, fftlen, "whole");
g_n = ifft(G_f)
stem(g_n)
```

```

title('Impulse response of g(n)')
ylabel('Magnitude')
xlabel('time')

figure(7)
check_n = conv(h_n,g_n);
stem(check_n)
figure(8)
G_f = fftshift(20*log(abs(fft(g_n,fftlen)))));
plot(-0.5:1/fftlen:0.5-1/fftlen, G_f)

```

<Used code for Matlab problem 1>

```

X = binornd(100,0.2,1000,10000)
Y(1,:) = zeros(1,10000);
for n=1:900
    Y(1,1:1:end) = Y(1,1:1:end)+X(n,1:1:end);
end
histfit(Y)

```

<Used code for Matlab problem 2>

(: reused the prelab2 solution code with unnecessary codes removal)

```

Ls = 26; % Number of 1db Ebno steps to sweep
numSymbols = 10000; % data vector size
ebno_db = 0:Ls-1; % generate ebno
ebno = 10 .^ (ebno_db/10); % generate ebno in linear scale
BER = zeros(1,Ls); % pre-allocate BER vector

fading = 1; % apply fading
threshold = 0; % decision threshold
clock_offset = 0; % clock offset in samples
N = 1; % Samples Per Symbol
h_ps = ones(1,N); % Pulse Shaping Filter
h_mf = h_ps/N; % Matched Filter

% pre-allocate vectors
bits = zeros(1, numSymbols); % bits
rBits = zeros(1,numSymbols); % received bits
n = zeros(1,numSymbols*N); % noise vector
x = zeros(1,numSymbols*N); % data vector
e = zeros(1,numSymbols*N); % error vector
y = zeros(1,numSymbols*N); % received vector
r = y; % output of matched filter

```

```

% ensure clock_offset value does not go out of index
clock_offset = mod(abs(clock_offset),N);

```

```

%sweep ebno

```

```

for i=1:Ls

    numErrors = 0;
    numBits = 0;

```

```

% gen noise, note: Eb is N, sigma = No/2
sigma = sqrt(N/(2*ebno(i)));

% generate rayleigh sigma
sig_r = sqrt(N/2);

while( numErrors < 500 )
    % gen data bits
    bits = sign(randn(1,numSymbols));

    x = upsample(bits,N);      % upsample to N samples per symbol
    x = filter(h_ps,1,x);      % pulse shape

    % gen noise
    n = sigma*randn(1,N*numSymbols);

    % apply fading / noise
    if( fading )
        r1 = sig_r*randn(1,N*numSymbols);
        r2 = sig_r*randn(1,N*numSymbols);
        R = sqrt(r1.^2 + r2.^2);

        y = x .* R + n;
    else
        y = x + n;
    end

    % apply matched filter
    r = filter(h_mf,1,y);

    % clock offset
    rBits = r( N*(1:numSymbols) - clock_offset );

    % threshold and compare bits
    e = abs( (rBits > threshold) - (bits > 0) );

    % accumulate errors / bits
    numBits = numBits + numSymbols;
    numErrors = numErrors + sum(e);
end

BER(i) = numErrors/numBits;
end

%compute theory curves
theory_bpsk = 0.5*erfc(sqrt(ebno));
theory_bpsk_fade = 0.5*(1 - sqrt( ebno ./ (ebno + 1) ) );

theory_bpsk_offset = 0.5*0.5*erfc(sqrt(ebno)*(1-2*clock_offset/N))+0.5*0.5*erfc(sqrt(ebno));

% plots
figure(1);
semilogy( ebno_db,BER,ebno_db, theory_bpsk, ebno_db, theory_bpsk_fade);

```

```

grid on;
xlabel('EbNo (dB)');
ylabel('BER');
title('BER vs. EbNo (dB)');
legend('Simulated BER','Theory BER','Theory BER w/ fade','location','SouthEast');
axis([0 15 -1 1e-6 1]);

```

<Used code for Matlabg problem 4>

```

clear all;close all;clc;

fftlens = 1024;
fs = 20*10^6; %20MHz
Ts =1/fs ;%50nsec
Tg = 1.6*10^-6 ;%1.63us
Tt = 6.35*10^-6 ;%6.35us
msg = load("p5_msg_revised.mat");
MSG = fft(msg.y,fftlens);
P_MSG = fftshift(abs(MSG).^2);

figure(1)
subplot(211)
plot((-0.5:1/fftlens:0.5-1/fftlens)*fs/fftlens,P_MSG)
title('message spectrum')
ylabel('Magnitude')
xlabel('frequency')
subplot(212)
stem(1:1:127,msg.y(1:1:127));
axis([1 127 -2 2])
title('Message preambgle in time domain')
ylabel('Magnitude')
%figure(2)
scatterplot(msg.y,2)
title('Message constellation')
%%=====
%%< T matrix >
T = zeros(127,32);
dm = prng(7,127).*2-1;
T= toeplitz(dm(1:1:127),[dm(1) zeros(1,31)]);
%%=====
% <h_ex>
msg_127= msg.y(1:1:127)
h_ex = pinv(T)*msg_127.'
H_ex = fft(h_ex,fftlens);
figure(3)
plot(-0.5:1/fftlens:0.5-1/fftlens,fftshift(10*log(abs(H_ex))))
title('H(f)')
ylabel('Magnitude(dB)')
xlabel('frequency')
%%=====
% <check h_ex>
msg_127_check = T*h_ex;
figure(4)

```

```

subplot(211)
stem(msg_127)
title('original message preamble')
subplot(212)
stem(msg_127_check)
title('preambgle recovered by H(f)')
%%=====
%%< R matrix >
R= toeplitz(msg_127(1:1:127),[msg_127(1) zeros(1,31)]);
%%=====
% <g_ex>
dm = pngen(7,127)*2-1;
g_ex = pinv(R)*dm.'
G_ex = fft(g_ex,fftlen);
figure(5)
plot(-0.5:1/fftlen:0.5-1/fftlen,fftshift(10*log(abs(G_ex))))
title('G(f)')
ylabel('Magnitude(dB)')
xlabel('frequency')
%%=====q=====
% < h_ex*g_ex >
eq_check = conv(h_ex,g_ex);
EQ_check = fft(eq_check,fftlen);
figure(6)
subplot(211)
stem(eq_check)
title('g_ex(t)(*)h_ex(t)')
ylabel('Magnitude(dB)')
xlabel('time')
subplot(212)
plot(-0.5:1/fftlen:0.5-1/fftlen,fftshift(10*log(abs(EQ_check))))
title('G_ex(f)*H_ex(f)')
ylabel('Magnitude(dB)')
xlabel('frequency')
%%=====q=====
% < g_ex(*)msg.y >
msg_check = conv(g_ex,msg.y);
MSG_check = fft(msg_check,fftlen);
figure(7)
subplot(211)
stem(msg_check)
title('equalized message by g_ex(t)(*)msg(t)')
ylabel('Magnitude(dB)')
xlabel('time')
subplot(212)
plot(-0.5:1/fftlen:0.5-1/fftlen,fftshift(10*log(abs(EQ_check))))
title('equalized message spectrum by G_ex(f)*H_ex(f)')
ylabel('Magnitude(dB)')
xlabel('frequency')
%%=====q=====
% < equalized messagbe constellation >]
scatterplot(msg_check)

```