



Project Component:02

Introduction to Communication Systems

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Submission by

S U Swakath

180020036

Course Instructor

Prof. Naveen M B

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Introduction to Communication Systems.

Project Components : 2.

Name: S.W. Swakath.

Roll no: 180020036.

2. Given x is the transmitted signal and y is the received signal ; $y = h * x + n$, where h is gaussian random variable with zero mean and unit variance and n is gaussian random variable with zero mean and variance changing according to the Signal to Noise Ratio. (for calculation purposes we will take it as σ^2).

Let us take $x \in \{s_1, s_2\}$ where $s_1 = [a, 0]$ $s_2 = [0, a]$
where $a > 0, a \in \mathbb{R}^+$

s_1 and s_2 are 2 different symbols with 2 dimension. Let s_1 represent bit 1 and s_2 represent bit 0.

Since the symbols are 2 dimension we can model our h (the channel) as 2 dimension.

$h = [h_1, h_2]$ where h_1 and h_2 are gaussian random variables with zero mean and variance $\sigma^2/2$.

Lets take the noise to be n , as expressed before the noise is also a RV with mean 0 and variance σ^2 .

$$y = h * x + n ; x \in \{s_1, s_2\}. \quad n = [n_1, n_2]$$

Y Given s_1 was transmitted is.

$$Y_{s_1} = [ah_1 + n_1, n_2]$$

Similarly Y Given s_2 was transmitted is.

$$Y_{s_2} = [n_1, ah_2 + n_2]$$

n_1 and n_2 are mean 0 and variance $\sigma^2/2$

Since we have a two dimension symbol.

$$P(Y/s_1) = P(Y/s_1, \text{indimension 1}) * P(Y/s_1, \text{indimension 2}) \\ = P(a h_1 + h_{n1}) * P(h_{n2}).$$

$$P(Y/s_2) = P(h_{n1}) * P(a h_2 + h_{n2}).$$

as stastically (h_1, h_2) and (h_{n1}, h_{n2}) pairs. are same that is with mean 0 and variance 1/2

$$P(Y/s_2) = P(Y/s_1) \quad \text{hence}$$

$$\log \left(\frac{P(Y/s_1)}{P(Y/s_2)} \right) = 0.$$

So we cannot demodulate using conventional ML.

We can observe this in another method also.

• Y when s_1 is passed is

$$[a h_1 + h_{n1}, h_{n2}]$$

the mean of this random variable is $[0, 0]$
and variance is $[\frac{a^2 + \sigma^2}{2}, \frac{\sigma^2}{2}]$

• this is because $a h_1 + h_{n1}$ and h_{n2} are random variables by it self and h_1 has mean = 0, variance = 1
 h_{n1} and h_{n2} are noise with mean = 0 and variance σ^2

• Similarly Y when s_2 is passed is

$$[h_{n1}, a h_2 + h_{n2}]$$

the mean of this RV is $[0, 0]$ and variance is

$$[\frac{\sigma^2}{2}, \frac{a^2 + \sigma^2}{2}]$$

from above we can conclude χ when s_1 or s_2 is passed the received RV is from a gaussian ~~var~~ distribution of mean $[0, 0]$ in both case. ML rule can be implemented if and only if s_1 and s_2 are received RV are from gaussian distribution with 2 different means.

Hence, we now need to some other method to demodulate the received signal.

\Rightarrow if x is a RV with mean μ_x and variance σ_x^2 the expected power / average power $= E(x \cdot x^*)$

let us take x is always real

so expected power $= E(x^2)$.

$$\sigma_x^2 = E((x - \mu_x)^2) = E(x^2 + \mu_x^2 - 2\mu_x x)$$

$$= E(x^2) + \mu_x^2 - 2\mu_x^2$$

$$E(x^2) = \sigma_x^2 + \mu_x^2$$

by this we can conclude the expected power of a RV x is $\sigma_x^2 + \mu_x^2$.

Applying this observation regarding power on the received random variable (χ).

χ Given $s_1 : [ah_1 + hn_1, hn_2] \Rightarrow$ expected power $\left[\frac{a^2 + \sigma^2}{2}, \frac{\sigma^2}{2} \right]$

χ Given $s_2 : [hn_1, ah_2 + hn_2] \Rightarrow$ expected power $\left[\frac{\sigma^2}{2}, \frac{a^2 + \sigma^2}{2} \right]$

From this we can conclude that if s_1 was transmitted. my received signal's power in dimension 1 will be most likely greater than power in dimension 2. Similarly when s_2 is transmitted then received signal's power is

dimension 2 is most likely to be greater than power in dimension 1.

from this we can write the following expression.

if received signal $\chi = [\alpha, \beta]$

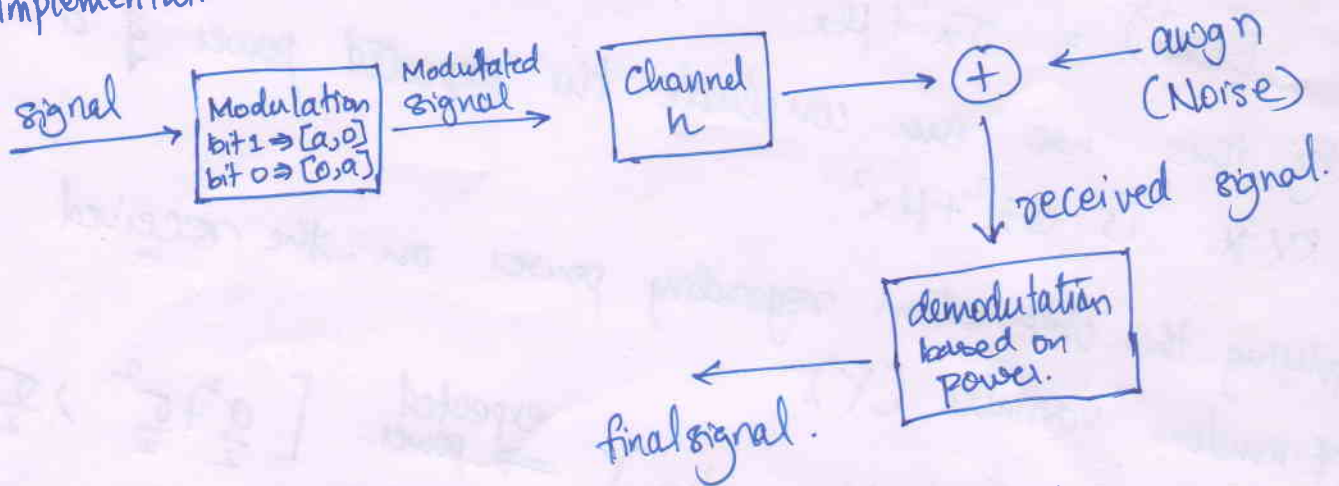
$$\alpha^2 \sum_{s_2}^{s_1} \beta^2 \Rightarrow \frac{\alpha^2}{\beta^2} \sum_{s_2}^{s_1} 1$$

from computation simplification

$$\frac{|\alpha|}{|\beta|} \sum_{s_2}^{s_1} 1$$

The ~~the~~ above discussed. demodulation is implemented in the second part of the question.

(b) Implementation.



We are implementing the above block diagram in Matlab.
The following functions ~~are used~~ coded and used :-

- channel.m
- demodulation.m
- modulation.m
- Bit_Error_Rate.m.

* modulation.m

• This function takes a signal and a scalar value 'a' as input and maps bit 1 $\Rightarrow [a, 0]$ and bit 0 $\Rightarrow [0, a]$ and returns the modulated signal.

* channel.m

• This function takes modulated signal (x) and SNR as input and computes the receive signal (Y) using.
$$Y = h * x + n$$
, where $h = [h_{-c1}, h_{-c2}]$
 h_{-c1} and h_{-c2} are gaussian random variable with zero mean and variance 1. n is noise, the values are determined based on SNR, and returns Y.

* ~~Demod~~ demodulation.m

• This function takes the received signal (Y) and maps $[x, \beta]$ to bit 1 if $|x| > |\beta|$ else it maps to bit 0.
• the function returns the final signal.

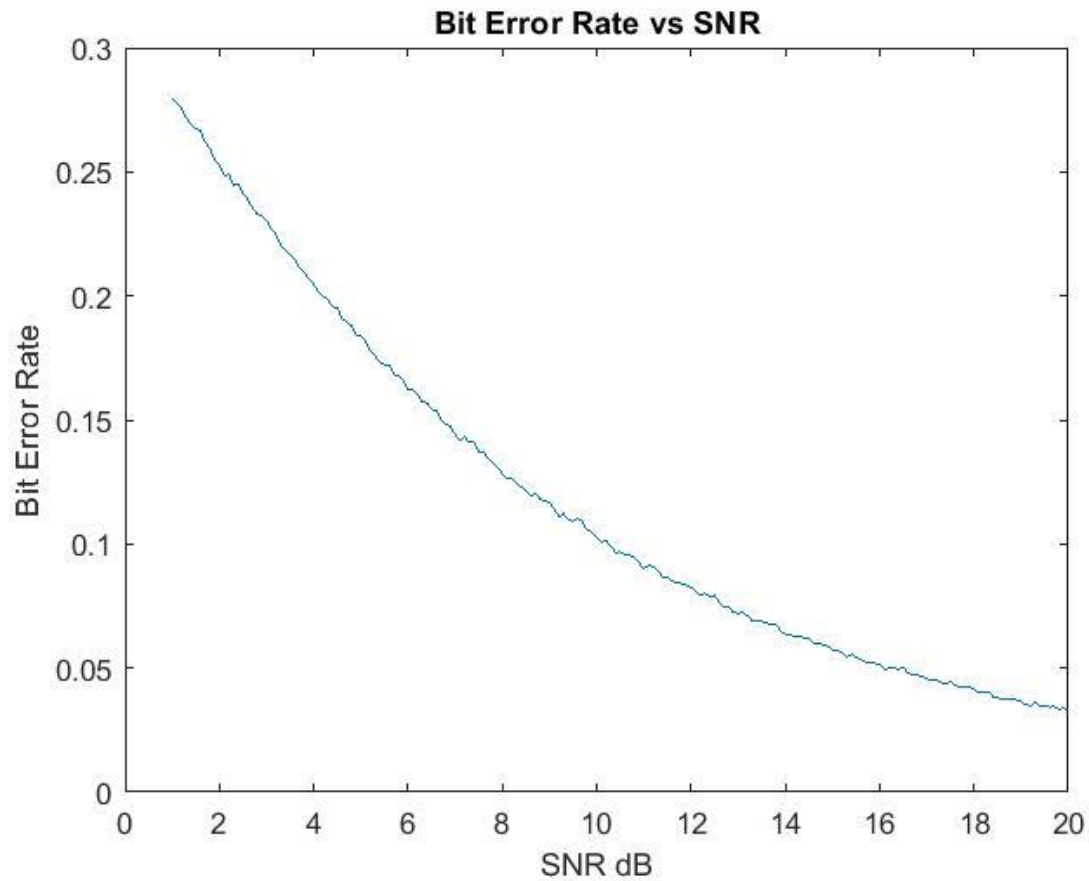
* Bit-Error-Rate.m

• the code ~~also~~ creates a signal of length N (=1000), modulates the signal with (a=2). Then computes the Bit Error rate for each value of SNR, where SNR is varying from 1 to 20 in steps of 0.1. To ~~error~~ calculate the BER we take 1000 samples for each SNR value.

The plot obtained and matlab code is attached below.

Plots and Inferences

Bit Error Rate vs SNR



from the above graph we can conclude that the Bit Error Rate (BER) decreases as the Signal to Noise Ratio (SNR) increases. This trend is as expected, that is when the power of the signal is larger than that of the noise, the probability that a change of bit occurs will reduce.


```
1 %This function takes a signal and value "a" as inputs and maps
2 %bit 0 => [0 a] and bit 1 => [a 0] and returns the modulatedSignal
3 function [modulatedSignal] = modulation(signal,a)
4     signalLen = size(signal,2);
5     modulatedSignal = [];
6     for i = 1:signalLen
7         if(signal(i) == 0)
8             modulatedSignal = [modulatedSignal ; [0 a]];
9         else
10            modulatedSignal = [modulatedSignal ; [a 0]];
11        end
12    end
13 end
```

```
1 %This fuction takes a modulatedSignal (represented x) and returns received
2 %signal (represented y), where  $y = h*x + n$  , h is gaussian random variable
3 %with zero mean and unit variance, n is gaussian noise according to the
4 %given signal to noise ratio.
5 function [receivedSignal] = channel(modulatedSignal,snr)
6     receivedSignal = [];
7     modulatedSignalLen = size(modulatedSignal,1);
8
9     h_c1 = rand(modulatedSignalLen,1);
10    h_c2 = rand(modulatedSignalLen,1);
11
12    %finding (h*x)
13    receivedSignal = modulatedSignal;
14    receivedSignal(:,1) = h_c1.*receivedSignal(:,1);
15    receivedSignal(:,2) = h_c2.*receivedSignal(:,2);
16
17    %adding awgn
18    receivedSignal = awgn(receivedSignal,snr,'measured');
19
20 end
```

```
1 %This function takes a receivedSignal and according to the energy of the
2 %received symbol it maps it to bit1 or bit0 and returns the final signal
3
4 function [finalSignal] = demodulation(receivedSignal)
5     finalSignal = [];
6     receivedSignalLen = size(receivedSignal,1);
7     absValues = abs(receivedSignal);
8
9     for i = 1:receivedSignalLen
10         temp = absValues(i,:);
11         if(temp(1)>temp(2))
12             finalSignal = [finalSignal 1];
13         else
14             finalSignal = [finalSignal 0];
15         end
16     end
17 end
```

```
1 %Code to find Bit Error Rate vs SNR graph
2 clear;clc;
3 %Creating a signal of length equal to N
4 a = 2; % a value for modulation
5 N = 1000; % length of the signal
6 signal = randi([0 1],1,N); %creating random signal containing bit1 and bit0
7
8 modulatedSignal = modulation(signal,a); %finding the modulated signal
9 SNR = 1:0.1:20; %different values of SNR for the plot
10 errorSNR = []; %array containing BER corresponding to each SNR
11 sampleSize = 1000; %Number of sample taken to calculate the error
12
13 %Finding the BER for each SNR value
14 for i = 1:0.1:20
15     fprintf("SNR %0.2f ....\n",i);
16     error = 0;
17     for j = 1:sampleSize
18         receivedSignal = channel(modulatedSignal,i);
19         finalSignal = demodulation(receivedSignal);
20         allError = finalSignal~=signal;
21         error = error + sum(allError)/size(allError,2);
22     end
23     error = error/sampleSize;
24     errorSNR = [errorSNR error];
25 end
26
27 %Plotting the BER vs SNR graph
28 plot(SNR,errorSNR);
29 xlabel("SNR dB");
30 ylabel("Bit Error Rate")
31 title("Bit Error Rate vs SNR");
```