

WIRELESS COMMUNICATION PRACTISE

EXPERIMENT - 01 (BER of BPSK in fading channel)

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1 Aim:

To analyze the fading channel. To evaluate performance of BPSK from BER (probability of error) vs SNR ratio in fading channel. Explore the other possibilities with BPSK in fading channel.

2 Software:

To perform this experiment, c++ language and gnuplot (open source) have been used. Data have been generated by the program in c++ language and plots are made by gnuplot.

3 Theory

3.1 Wireless channel and model

3.1.1 Wireless and wired

From basic understanding of electrical signal and electromagnetic waves, data (*modulated signal*) propagate in wired medium, that implies it always has certain path and direction inside the medium. Unlike wired medium, in wireless, data (*modulated signal*) radiates into the space (free space/air), that implies it do not have fixed path and direction inside the medium. This issue creates problems for receiver.

3.1.2 Multipath

Consider no noise case, if transmitter transmits $x(t)$ signal in wireless channel, then receiver receives $y(t)$ as

$$y(t) = \sum_i \alpha_i(t) x(t - \tau_i(t))$$

where, α_i and τ_i are amplitudes and time delays respectively, of the received signal through i^{th} path. This is the basic model for multi-path issue. (*Note:- Here, transmitter and receiver are assumed to be in non-motion, hence, no Doppler shift has been considered here*)

3.1.3 Multipath modelling

Consider transmitted signal $x(t)$ as cosine signal (most basic signal).

$$\begin{aligned}
 x(t) &= \cos(\omega t) \\
 x(t) &= \text{Re}\{e^{j\omega t}\} \quad (\text{Re implies real part}) \\
 \Rightarrow y(t) &= \sum_i \text{Re}\{\alpha_i(t) e^{j\omega(t-\tau_i(t))}\} \\
 &= \sum_i \text{Re}\{\alpha_i(t) e^{-j\omega\tau_i(t)} e^{j\omega t}\} \quad (1)
 \end{aligned}$$

Here, $\alpha_i e^{-j\omega\tau_i(t)}$ can be considered as the overall impairment to the transmitted signal by wireless channel. Consider channel is slow changing channel, that implies environment of channel is changing slowly with respect to change in signal. In other words, bit duration is very small as compare to rate of change in channel. For example if data rate is 1Mbps implies 10^{-6} sec is bit duration (after that bit may change hence signal may change) but dynamic elements in channel (so called reflectors) can not even move in 10^{-6} sec. This type of channel is known as slow fading channel.

Due to slow fading channel assumption, $\alpha_i(t)$ and $\tau_i(t)$ can be considered as constant in every fix sized time slot.

$$\begin{aligned}
 y(t) &= \sum_i \text{Re}\{\alpha_i(t) e^{-j\omega\tau_i(t)} e^{j\omega t}\} \quad (\text{from 1}) \\
 y_k(t) &= \sum_i \text{Re}\{\alpha_i e^{-j\omega\tau_i} e^{j\omega t}\} \quad (\text{due to slow fading assumption}) \\
 \text{consider, } -j\omega\tau_i &= \phi_i \quad (\phi_i \in [-\pi, \pi]) \\
 &= \sum_i \text{Re}\{\alpha_i e^{\phi_i} e^{j\omega t}\} \\
 &= \text{Re}\{\sum_i \alpha_i e^{\phi_i} e^{j\omega t}\}
 \end{aligned}$$

In practical case, there are infinite number of reflectors, that implies i goes from zero to infinity. If $\alpha_i e^{\phi_i}$ consider as a independent and identical random variable for every sample of $y(t)$ like y_k , then it would converge to the Gaussian random variable. (Central limit theorem).

$$\sum_i \alpha_i e^{\phi_i} = R + jI \quad (2)$$

Since, in equation-2 LHS is complex quantity, therefore RHS is modelled as complex random variable, where real and imaginary part is gaussian random variable.

$$\begin{aligned}
 R &= \sum_i \alpha_i \cos(\phi_i) \\
 I &= \sum_i \alpha_i \sin(\phi_i) \\
 R &\sim \mathcal{N}(0, \sigma^2) \\
 I &\sim \mathcal{N}(0, \sigma^2)
 \end{aligned}$$

Finally, received signal in k^{th} sample of $y(t)$ can be written as

$$\begin{aligned} y_k(t) &= Re\left\{\sum_i \alpha_i e^{j\phi_i} e^{j\omega t}\right\} \\ y_k(t) &= Re\{(R + jI) e^{j\omega t}\} \\ y_k(t) &= R \cos(\omega t) - I \sin(\omega t) \end{aligned}$$

By complex analysis, one can write $y_k(t)$ is polar form.

$$\begin{aligned} y_k(t) &= Re\{M_k e^{j\theta_k} e^{j\omega t}\} \\ y_k(t) &= M_k Re\{e^{j\theta_k} e^{j\omega t}\} \end{aligned}$$

To remove the phase term, there is process called co-phasing. After co-phasing.

$$\begin{aligned} y_k(t) &= M_k Re\{e^{j\omega t}\} \\ y_k(t) &= M_k x_k(t) & (k^{th} \text{ sample of } x(t)) \\ Y_k &= M_k X_k & (3) \\ \text{where, } M_k &= \sqrt{R^2 + I^2} \end{aligned}$$

Now, problem is to find statistical properties of M_k

3.1.4 Rayleigh distribution

Consider X and Y are two independent zero mean but same variance gaussian random variables.

$$\begin{aligned} X &\sim \mathcal{N}(0, \sigma^2) \\ Y &\sim \mathcal{N}(0, \sigma^2) \\ X &\perp\!\!\!\perp Y \\ \text{let } Z &= X + jY = M e^{j\theta} \\ \text{where, } M &= \sqrt{X^2 + Y^2} \\ \theta &= \tan^{-1}\left(\frac{Y}{X}\right) \\ \because X &\perp\!\!\!\perp Y \\ \therefore f_{X,Y}(x,y) &= f_X(x) f_Y(y) \\ &= \frac{1}{2\pi\sigma^2} \exp\left(\frac{-1}{2\sigma^2}(x^2 + y^2)\right) \end{aligned}$$

Convert the $f_{X,Y}(x,y)$ into $f_{M,\theta}(m,\Theta)$, for that use concepts of transformation of random variables, jacobian matrix.

$$\begin{aligned}
& \text{consider } \theta = \tan^{-1} \left(\frac{Y}{X} \right) \\
& \implies Y = X \tan(\theta) \\
& \therefore, M = \sqrt{X^2 + Y^2} \quad (M \geq 0) \\
& \therefore, M = X \sec(\theta) \quad (X \text{ and } \sec(\theta) \geq 0 \text{ or } X \text{ and } \sec(\theta) \leq 0) \\
& \implies X = M \cos(\theta) \\
& \implies Y = M \sin(\theta)
\end{aligned}$$

Note:- This is only valid of principle values of $\tan^{-1}(x) \forall x$ i.e. $(-\pi/2, \pi/2)$ and $M \geq 0$, But X and Y can be any value positive and negative. This gives two solutions.

First as $X = M \cos(\theta)$ and $Y = M \sin(\theta)$

second as $X = -M \cos(\theta)$ and $Y = -M \sin(\theta)$

Now jacobian

$$\begin{aligned}
|J| &= \begin{vmatrix} \frac{\partial X}{\partial M} & \frac{\partial X}{\partial \theta} \\ \frac{\partial Y}{\partial M} & \frac{\partial Y}{\partial \theta} \end{vmatrix} \\
|J| &= \begin{vmatrix} \cos(\theta) & -M \sin(\theta) \\ \sin(\theta) & M \cos(\theta) \end{vmatrix} \\
|J| &= M
\end{aligned}$$

Both solution leads to same jacobian value. That mean, jacobian for entire principle value of θ is $2M$.

Transformation of random variable

$$\begin{aligned}
f_{M,\theta}(m, \Theta) &= |J| f_{X,Y}(x, y) |_{X=M \cos(\theta), Y=M \sin(\theta)} \\
f_{M,\theta}(m, \Theta) &= \frac{2M}{2\pi\sigma^2} \exp\left(\frac{-1}{2\sigma^2}(M^2)\right) \\
f_M(m) &= \int_{-\pi/2}^{\pi/2} \frac{2M}{2\pi\sigma^2} \exp\left(\frac{-1}{2\sigma^2}(M^2)\right) d\theta \quad M \geq 0 \\
f_M(m) &= \frac{M}{\sigma^2} \exp\left(\frac{-M^2}{2\sigma^2}\right) \quad M \geq 0
\end{aligned}$$

This distribution is known as Rayleigh distribution of parameter σ^2 . Similarly,

$$\begin{aligned}
f_\theta(\Theta) &= \int_0^\infty \frac{2M}{2\pi\sigma^2} \exp\left(\frac{-1}{2\sigma^2}(M^2)\right) dM \\
&= \frac{1}{\pi\sigma^2} \int_0^\infty M \exp\left(\frac{-M^2}{2\sigma^2}\right) dM \\
&\text{let } u = \frac{M^2}{2\sigma^2} \implies du = M/\sigma^2 \\
&= \frac{1}{\pi} \int_0^\infty \exp(-u) du \\
&= \frac{1}{\pi}
\end{aligned}$$

That means $\theta \sim \text{uniform}(-\pi/2, \pi/2)$

Since $\theta \in (-\pi, \pi)$, extend the uniformity from $(-\pi/2, \pi/2)$ to $(-\pi, \pi)$.

$$f_{\theta}(\Theta) = \frac{1}{2\pi} \quad ; \quad \theta \in (-\pi, \pi)$$

Warning:

As a responsible engineer and author, i must confess that i do not know whether can we extend the uniform random variable. Maybe whatever i did for θ in last line is wrong. i will not use this in further parts of this report.

$$\begin{aligned} f_{M,\theta}(m, \Theta) &= \left(\frac{1}{\pi}\right) \frac{M}{\sigma^2} \exp\left(\frac{-M^2}{2\sigma^2}\right) \quad ; \quad M \geq 0 \\ &= f_{\theta}(\Theta) f_M(m) \\ \therefore, \theta &\perp\!\!\!\perp M \end{aligned}$$

For wireless channel, M is Rayleigh with parameter 1/2.

$$\begin{aligned} Y_k &= M_k X_k \\ M_k &\sim \text{Rayleigh}(1/2) \end{aligned}$$

3.2 BPSK in Rayleigh Fading channel

3.2.1 ML detector

$$\begin{aligned} f_{Y|B,M}(y|0, m) &\geq_{B^o=1}^{B^o=0} f_{Y|B,M}(y|1, m) \\ f_Y(m\sqrt{E} + \eta = y) &\geq_{B^o=1}^{B^o=0} f_Y(-m\sqrt{E} + \eta = y) \\ f_N(\eta = y - m\sqrt{E}) &\geq_{B^o=1}^{B^o=0} f_N(\eta = y + m\sqrt{E}) \\ \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left(\frac{-1}{2\sigma^2}(y - m\sqrt{E})^2\right) &\geq_{B^o=1}^{B^o=0} \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left(\frac{-1}{2\sigma^2}(y + m\sqrt{E})^2\right) \\ (y - m\sqrt{E})^2 &\geq_{B^o=0}^{B^o=1} (y + m\sqrt{E})^2 \\ 2y m\sqrt{E} &\geq_{B^o=1}^{B^o=0} 0 \\ y &\geq_{B^o=1}^{B^o=0} 0 \end{aligned}$$

ML detector is same as normal BPSK detector.

3.2.2 Probability of error

$$\begin{aligned}
P(\text{Error}) &= \sum_{i=0}^1 P(\text{Error}, B = i) \\
&= \sum_{i=0}^1 P(\text{Error}|B = i)P(B = i) \\
\text{consider } P(\text{Error}|B = 0) &= P(B^o = 1|B = 0) \\
&= \int_0^\infty P(B^o = 1|B = 0, M = m) f_M(m) dm \\
&= \int_0^\infty P(y < 0|B = 0, M = m) f_M(m) dm \\
&= \int_0^\infty P(m\sqrt{E} + \eta < 0) f_M(m) dm \\
&= \int_0^\infty P(\eta < -m\sqrt{E}) f_M(m) dm \\
&= \int_{m=0}^\infty \int_{x=m\sqrt{E}}^\infty \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right) dx 2m \exp(-m^2) dm \\
\text{Change the variables} &= \int_{x=0}^\infty \int_{m=0}^{x/\sqrt{E}} \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right) 2m \exp(-m^2) dm dx \\
&= \int_{x=0}^\infty \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right) 2 \int_{m=0}^{x/\sqrt{E}} m \exp(-m^2) dm dx \\
&= \int_{x=0}^\infty \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right) \left[1 - \exp\left(-\frac{x^2}{E}\right)\right] dx \\
&= \frac{1}{2} - \int_{x=0}^\infty \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2} - \frac{x^2}{E}\right) dx \\
&= \frac{1}{2} - \int_{x=0}^\infty \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-x^2 \left[\frac{1}{2\sigma^2} + \frac{1}{E}\right]\right) dx \\
\text{let } \frac{s}{\sqrt{2}} &= x \sqrt{\frac{1}{2\sigma^2} + \frac{1}{E}} \\
dx &= \frac{ds}{\sqrt{2} \sqrt{\frac{1}{2\sigma^2} + \frac{1}{E}}} \\
&= \frac{1}{2} - \int_{s=0}^\infty \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{s^2}{2}\right) \frac{ds}{\sqrt{2} \sqrt{\frac{1}{2\sigma^2} + \frac{1}{E}}} \\
&= \frac{1}{2} - \frac{1}{2\sqrt{2}\sigma \left(\sqrt{\frac{1}{2\sigma^2} + \frac{1}{E}}\right)} \\
&= \frac{1}{2} \left[1 - \frac{1}{\left(\sqrt{1 + \frac{2\sigma^2}{E}}\right)}\right]
\end{aligned}$$

Now assume, $\gamma = \frac{E}{\sigma^2}$ Hence,

$$P(B^o = 1|B = 0) = \frac{1}{2} \left[1 - \frac{1}{\left(\sqrt{1 + \frac{2}{\gamma}} \right)} \right]$$

$$\text{similarly, } P(B^o = 0|B = 1) = \frac{1}{2} \left[1 - \frac{1}{\left(\sqrt{1 + \frac{2}{\gamma}} \right)} \right]$$

$$\therefore P(\text{Error}) = \frac{1}{2} \left[1 - \frac{1}{\left(\sqrt{1 + \frac{2}{\gamma}} \right)} \right]$$

For large values of γ , approximation is possible in $P(\text{Error})$

$$\left(\sqrt{1 + \frac{2}{\gamma}} \right) \approx 1 + \frac{1}{\gamma} \quad \text{(Using Taylor series expansion)}$$

$$P(\text{Error}) \approx \frac{1}{2} \left(1 - \frac{1}{1 + \frac{1}{\gamma}} \right)$$

$$\approx \frac{1}{2} \left(1 - \frac{\gamma}{\gamma + 1} \right)$$

$$\approx \frac{1}{2(1 + \gamma)} \quad \text{(a simple approximation for large } \gamma \text{)}$$

$$\approx \frac{1}{2\gamma} \quad \text{(One more simple approximation)}$$

4 Pseudo code

- S1. Generate random bit stream of length equal to million.
- S2. Map 0 to $-\sqrt{E}$ and 1 to \sqrt{E} .
- S3. Generate two gaussian noise vector of zero mean and variance as half, treat as real part and imaginary part of the complex random variable.
- S4. Compute Rayleigh coefficient by using above random variable.
- S5. Generate gaussian noise component.
- S6. Apply channel model equation $Y_k = H_k X_k + N_k$ and get the received signal.
- S7. Apply ML rule and decode the received signal.
- S8. Count errors
- S9. Repeat S1 to S8 for many SNR values.
- S10. Store data (SNR, Error count) in .dat file.
- S11. Compute three theoretical probability of error (one exact and two approx value). Store into .dat file
- S12. Plot the result using gnuplot in log scale.

5 Code and Results

Main code for Rayleigh fading channel, for bpsk a header file has been designed to reduce the length of the program.

5.1 Code for Rayleigh fading channel

```

1  //////////////////////////////////////
2  //////////////////////////////////////
3  // Author:- MANAS KUMAR MISHRA
4  // Organisation:- IIITDM KANCHEEPURAM
5  // Topic:- Performance of BPSK in Rayleigh fading channel
6  //////////////////////////////////////
7  //////////////////////////////////////
8
9  #include <iostream>
10 #include <cmath>
11 #include <iterator>
12 #include <random>
13 #include <chrono>
14 #include <time.h>
15 #include <fstream>
16 #include "BPSK.h"
17
18 #define one_million 1000000
19
20 using namespace std;
21
22
23 // Function for printing the vector on the console output.
24 void PrintVectorDouble(vector<double> vectr)
25 {
26     std::copy(begin(vectr), end(vectr), std::ostream_iterator<double>(std::cout, "
27 ));
28     cout<<endl;
29 }
30
31 // Function for generating binary bits at source side. Each bit is equiprobable.
32 // Input is nothing
33 // Output is a vector that contains the binary bits of length one_million*1.
34 vector<double> sourceVector()
35 {
36     vector<double> sourceBits;
37
38     // Use current time as seed for random generator
39     srand(time(0));
40
41     for(int i = 0; i<one_million; i++){
42         sourceBits.insert(sourceBits.end(), rand()%2);
43     }
44
45     return sourceBits;

```



```

46 }
47
48
49 // Function for Rayleigh fading coefficients
50 // Inputs are two vectors one is the gaussian noise as real
51 // part and second as Gaussian noise as imaginary part
52 // Output is a vector that contains rayleigh noise coeff, sqrt(real_part^2 + imz_part^2)
53 vector<double> RayleighFadingCoff(vector<double> realGaussian,
54                                 vector<double> ImziGaussian)
55 {
56     vector<double> rayleighNoise;
57     double temp;
58
59     for(int times=0; times<realGaussian.size(); times++){
60
61         temp = sqrt(pow(realGaussian[times], 2)+pow(ImziGaussian[times], 2));
62         rayleighNoise.insert(rayleighNoise.end(), temp);
63     }
64
65     return rayleighNoise;
66 }
67
68
69 // function for modelling wireless channel
70 // Inputs are the transmitted signal (X_k), rayleigh
71 // coefficient (H_k) and AWGN (N_k) component
72 // Output is the H_k*X_k+N_k
73 vector<double> ChannelOperation(vector<double> TransSignal,
74                                vector<double> RayleighCoff,
75                                vector<double> gnoise)
76 {
77     vector<double> channelResult;
78     double temp;
79     for(int i =0; i<TransSignal.size(); i++){
80         temp = (RayleighCoff[i]*TransSignal[i]) + gnoise[i];
81         channelResult.insert(channelResult.end(), temp);
82     }
83
84     return channelResult;
85 }
86
87
88 // Function to count number of errors in the received bits.
89 // Inputs are the sourcebits and decodedbits
90 // Output is the number of error in received bits.
91 // error: if sourcebit != receivebit
92 double errorCalculation (vector<double> sourceBits, vector<double> decodedBits)
93 {
94     double countError =0;
95     for(int i =0; i<sourceBits.size(); i++){
96         if(sourceBits[i] != decodedBits[i]){
97             countError++;
98         }
99     }
100
101     return countError;

```

```

102 }
103
104
105 // Function to store the data in the file (.dat)
106 // Input is the SNR per bit in dB and calculated probability of error
107 // Output is the nothing but in processing it is creating a file and writing data into it
108 void datafile(vector<double> xindB, vector<double> Prob_error)
109 {
110     ofstream outfile;
111
112     outfile.open("FaddingChan1.dat");
113
114     if(!outfile.is_open()){
115         cout<<"File opening error !!!"<<endl;
116         return;
117     }
118
119     for(int i =0; i<xindB.size(); i++){
120         outfile<< xindB[i] << " "<<"\t"<<" "<< Prob_error[i]<< endl;
121     }
122
123     outfile.close();
124 }
125
126
127
128 vector<double> Errorfunction(vector <double> SNR_dB)
129 {
130     vector <double> Qvalue;
131
132     double po, normalValue, inter;
133     for (int k =0; k<SNR_dB.size(); k++){
134         normalValue = pow(10, (SNR_dB[k]/10));
135         // inter = 1+(2/normalValue);
136         // inter = sqrt(inter);
137         // po = 0.5*(1-(1/inter));
138         po = 1/(2*(normalValue));
139         Qvalue.insert(Qvalue.end(), po);
140     }
141
142     return Qvalue;
143 }
144
145 // Function to store the data in the file (.dat)
146 // Input is the SNR per bit in dB and calculated Qfunction values
147 // Output is the nothing but in processing it is creating a file and writing data into it.
148 void ErrorValueInFile(vector <double> SNR, vector <double> Qvalue)
149 {
150     ofstream outfile;
151
152     outfile.open("FaddingChan.Qvalue1.dat");
153
154     if(!outfile.is_open()){
155         cout<<"File opening error !!!"<<endl;
156         return;
157     }

```

```

158
159     for(int i =0; i<SNR.size(); i++){
160         outfile<< SNR[i] << " "<<"\t"<<" "<< Qvalue[i]<< endl;
161     }
162
163     outfile.close();
164 }
165
166
167 int main()
168 {
169     // source defination
170     vector<double> sourceBits;
171
172     // Mapping of bits to symbols;
173     vector<double> transmittedSymbol;
174
175     // Noise definition
176     vector<double> gnoise;
177
178     vector<double> realGaussian;
179     vector<double> imziGaussian;
180
181     vector<double> RayleighNoise;
182
183     vector<double> receiveSignal;
184
185     vector<double> decodedBits;
186
187     double sigmaSquare = 0.5;
188     double stddevRayleigh = sqrt(sigmaSquare);
189     double N_o =4;
190     double p, stdnoise;
191     double countererror, P_error;
192
193     vector<double> SNR_dB;
194     for(float i =0; i<=35; i=i+0.5)
195     {
196         SNR_dB.insert(SNR_dB.end(), i);
197     }
198
199     vector<double> energyOfSymbol;
200     vector<double> Prob_error;
201     double normalValue;
202
203     for(int i =0; i<SNR_dB.size(); i++){
204
205         normalValue = pow(10, (SNR_dB[i]/10));
206         energyOfSymbol.insert(energyOfSymbol.end(), N_o*normalValue);
207     }
208
209
210     for(int step =0; step <energyOfSymbol.size(); step++){
211
212         sourceBits = sourceVector();
213

```

```

214     transmittedSymbol=bit_maps_to_symbol_of_energy_E(sourceBits, energyOfSymbol[step],
215                                                         one_million);
216
217     realGaussian = GnoiseVector(0.0, stddevRayleigh, one_million);
218     imziGaussian = GnoiseVector(0.0, stddevRayleigh, one_million);
219
220     RayleighNoise = RayleighFaddingCoff(realGaussian, imziGaussian);
221
222     stdnoise = sqrt(N_o);
223     gnoise = GnoiseVector(0.0, stdnoise, one_million);
224
225     receiveSignal = ChannelOperation(transmittedSymbol, RayleighNoise, gnoise);
226
227     decodedBits = decisionBlock(receiveSignal);
228     countererror = errorCalculation(sourceBits, decodedBits);
229
230     P_error = countererror/one_million;
231
232     Prob_error.insert(Prob_error.end(), P_error);
233
234     cout<<endl;
235
236     cout<<"Energy of symbol      : "<<energyOfSymbol[step]<<endl;
237     cout<<"Count errors          : "<<countererror<<endl;
238     cout<<"Probability of error : "<<P_error<<endl;
239
240     cout<<endl;
241 }
242
243 datafile(SNR_dB, Prob_error);
244 vector<double> qvalue = Errorfunction(SNR_dB);
245 ErrorValueInFile(SNR_dB, qvalue);
246
247 return 0;
248 }

```

Code for BPSK header file

```

1  //////////////////////////////////////
2  //////////////////////////////////////
3  // Author:- MANAS KUMAR MISHRA
4  // Organisation:- IIITDM KANCHEEPURAM
5  // Topic:- header file BPSK scheme
6  //////////////////////////////////////
7  //////////////////////////////////////
8  #include <cmath>
9  #include <iterator>
10 #include <random>
11 #include <chrono>
12 #include <time.h>
13
14 using namespace std;
15
16 // Function for mapping bits to symbol.
17 // Input is a binary bit vector. Here 0---> -(sqrt(Energy)) and 1---> (sqrt(Energy))
18 // Output is a vector that contains transmitted symbols.

```

```

19 vector<double> bit_maps_to_symbol_of_energy_E(vector<double> sourceBits,
20                                             double energyOfSymbol,
21                                             const int one_million)
22 {
23     vector<double> transmittedSymbol;
24
25     for(int i=0; i<one_million; i++){
26         if(sourceBits[i]== 0){
27             transmittedSymbol.insert(transmittedSymbol.end(), -sqrt(energyOfSymbol));
28         }
29         else{
30             transmittedSymbol.insert(transmittedSymbol.end(), sqrt(energyOfSymbol));
31         }
32     }
33 }
34
35 return transmittedSymbol;
36 }
37
38 // Function for generating random noise based on gaussian distribution N(mean, variance).
39 // Input mean and standard deviation.
40 // Output is the vector that contain gaussian noise as an element.
41 vector<double> GnoiseVector(double mean, double stddev, const int one_million)
42 {
43     std::vector<double> data;
44
45     // construct a trivial random generator engine from a time-based seed:
46     unsigned seed = std::chrono::system_clock::now().time_since_epoch().count();
47     std::default_random_engine generator (seed);
48
49     std::normal_distribution<double> dist(mean, stddev);
50
51     // Add Gaussian noise
52     for (int i =0; i<one_million; i++) {
53         data.insert(data.end(), dist(generator));
54     }
55
56     return data;
57 }
58
59 // Function for modeling additive channel. Here gaussian noise adds to the transmitted bit.
60 // Inputs are the transmitted bit and gaussian noise with mean 0 and variance 1.
61 // Output is the receive bits.
62 vector<double> receiveBits(vector<double> transBit, vector<double> gnoise)
63 {
64     vector<double> recievebits;
65
66     for(int j =0; j<transBit.size(); j++){
67         recievebits.insert(recievebits.end(), transBit[j]+gnoise[j]);
68     }
69
70     return recievebits;
71 }
72
73 }
74

```

```

75
76
77 // Function for deciding the bit value from the received bits
78 // Input is the received bits.
79 // Output is the decoded bits.
80 // Decision rule :- if receiveBit > 0 then 1 otherwise 0 (simple Binary detection)
81 vector<double> decisionBlock(vector<double> receiveBits)
82 {
83     vector<double> decodedBits;
84
85     for(int i = 0; i < receiveBits.size(); i++){
86         if(receiveBits[i] > 0){
87             decodedBits.insert(decodedBits.end(), 1);
88         }
89         else{
90             decodedBits.insert(decodedBits.end(), 0);
91         }
92     }
93
94     return decodedBits;
95 }

```

5.2 Results

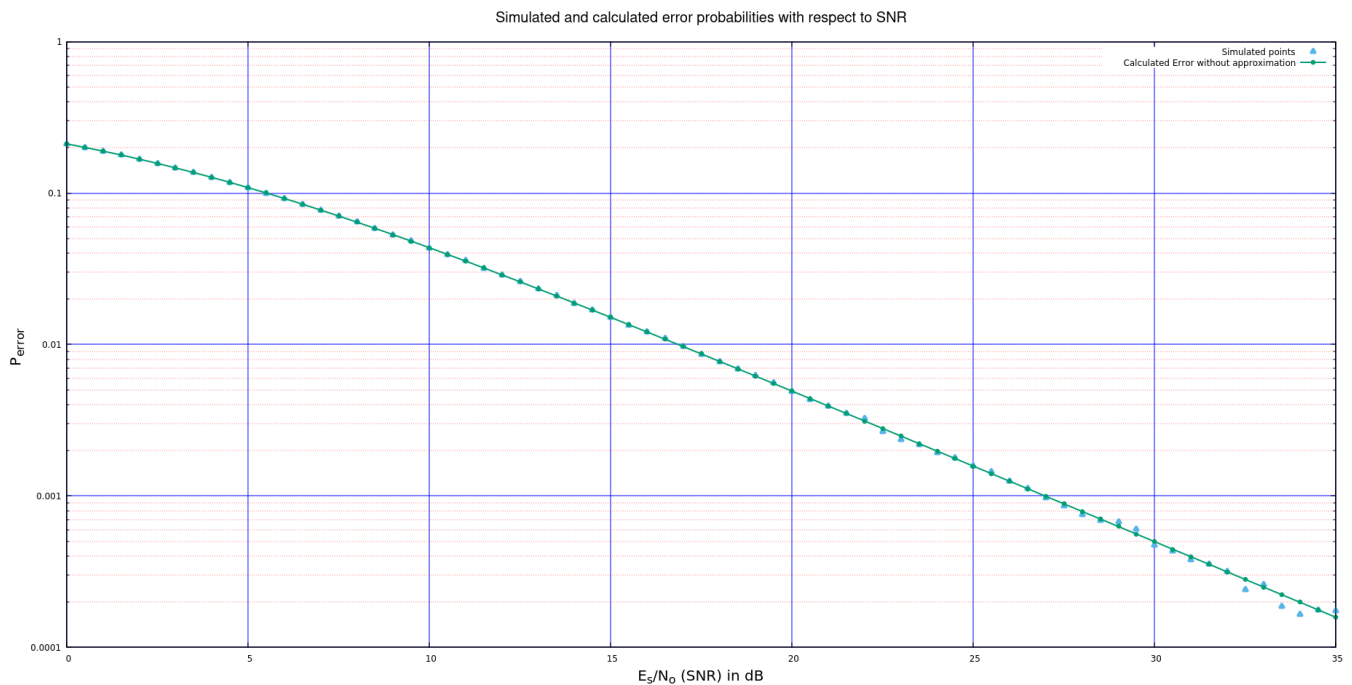


Figure 1: Simulated and calculated (theoretical) results

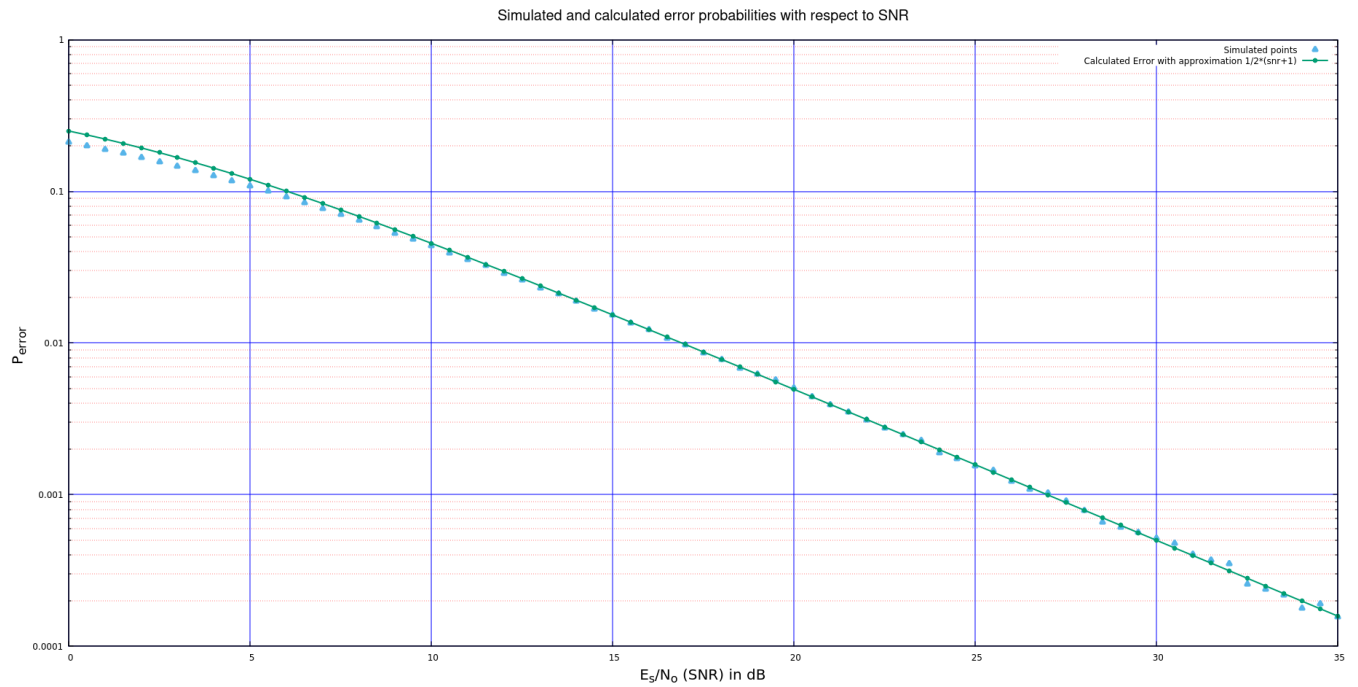


Figure 2: Simulated and calculated results (using approximation)

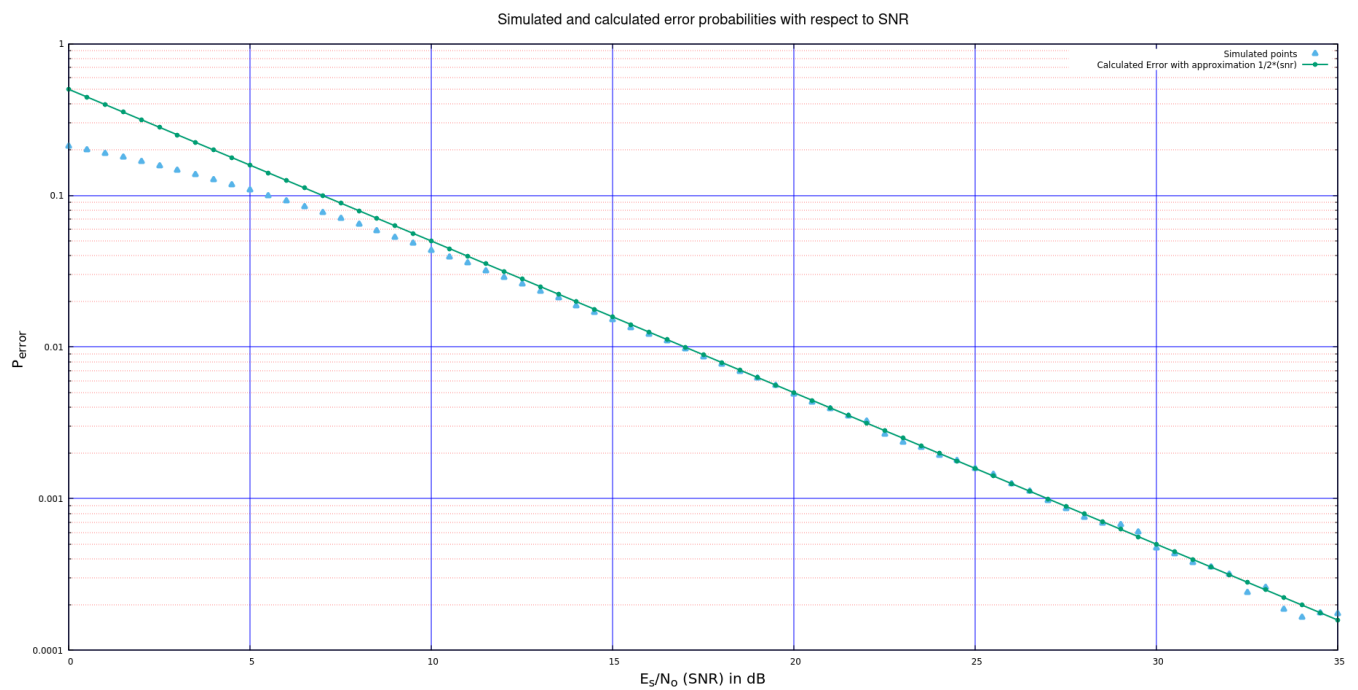


Figure 3: Simulated and calculated results (using approximation)

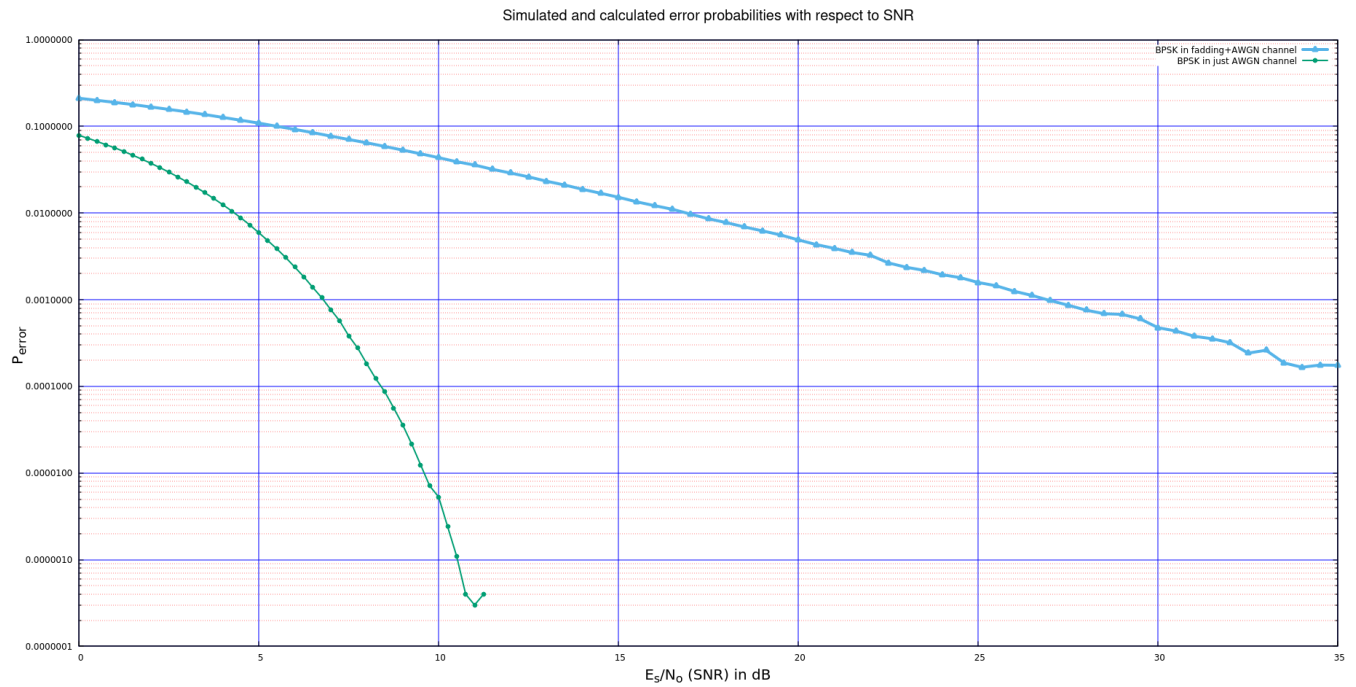


Figure 4: BPSK in just AWGN and BPSK in fading channel

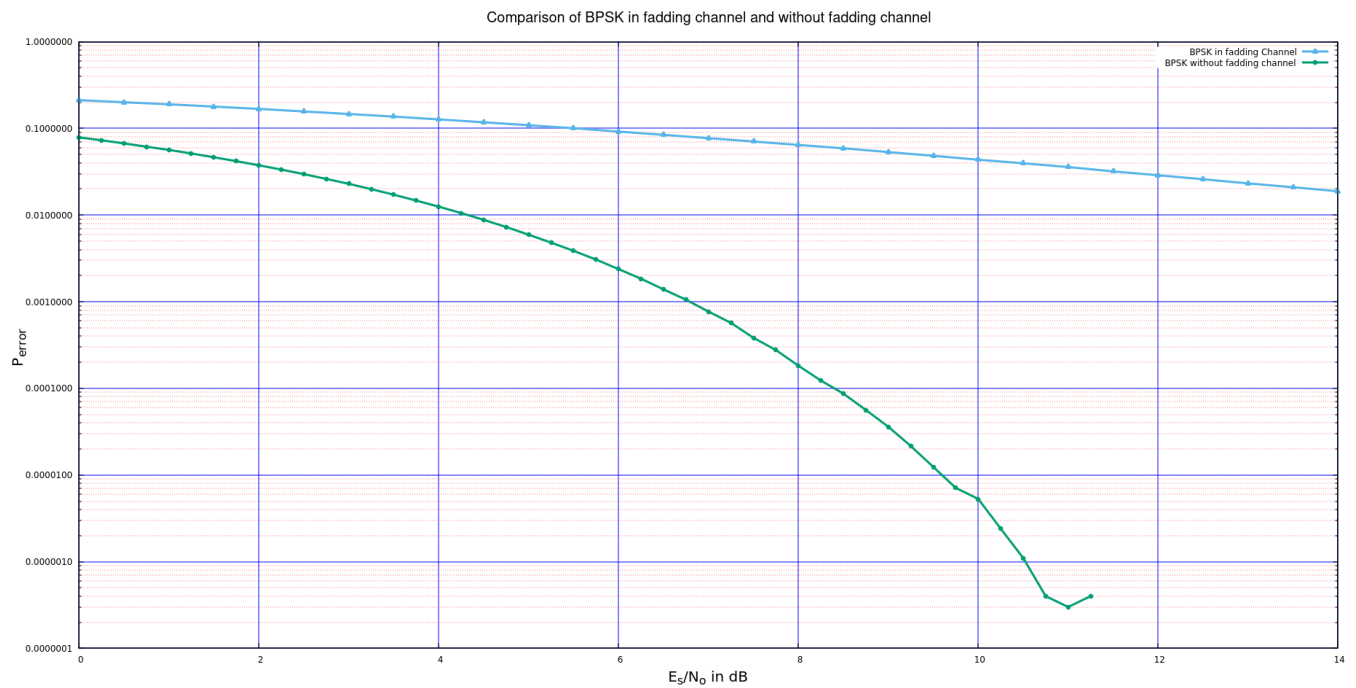


Figure 5: BPSK in just AWGN and BPSK in fading channel for low values of SNR

5.3 Something extra

As anyone can see, fading channel has very poor probability of error for given SNR value, as compare to normal AWGN channel. Now curiosity is that how to go close to the AWGN channel values in fading

channel. In other words, is there any way to shift fading channel probability of error close to AWGN channel value.

Answer is yes, it is possible with loss of data rate. If transmitter transmits same bit L consecutive times (it need not be consecutive, rather transmitter can transmit sequence of bits L times repeatedly), then receiver receive the same data on different signals. Due to that receiver can improve the probability of error.

Based on above hypothesis, given code has been modified for same consecutive two and three bits ($L = 2$ and $L = 3$). Results are given below. And it satisfies the above hypothesis.

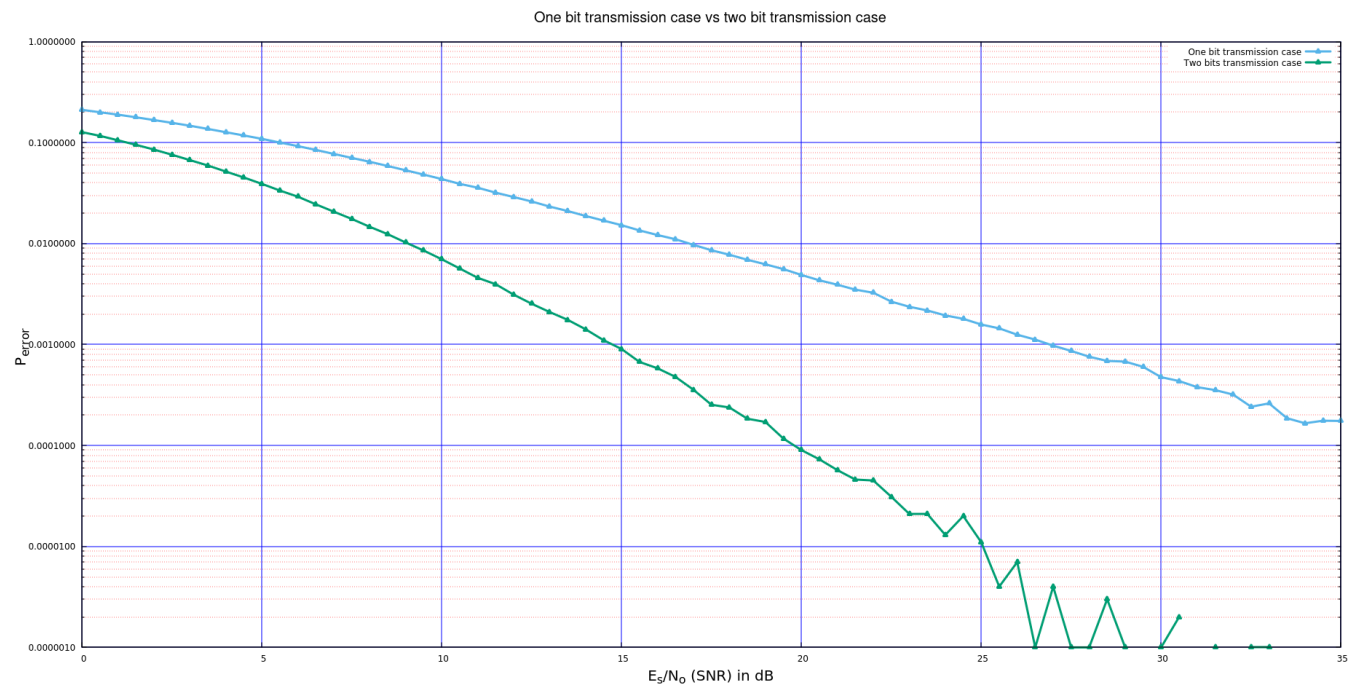


Figure 6: Only one bit and two same bits transmission

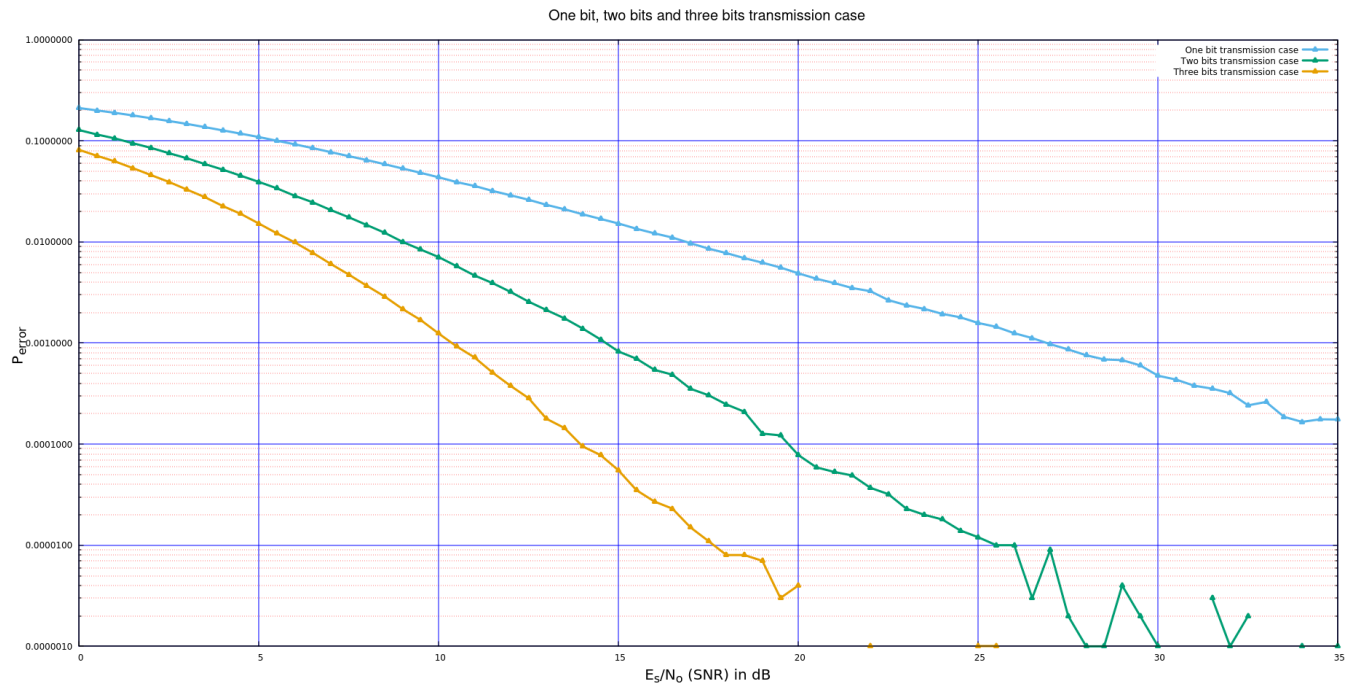


Figure 7: Only one bit, two same bits and three same bits transmission

6 Inference

Inferences list:

Based on the Comparison of AWGN and fading

In fading channel, probability of error approximately follows linear trend in downward direction (in log scale), but in AWGN channel falling trend is approximately exponential. That mean, as we increase SNR, AWGN channel improves a lot but fading channel improve slowly.

Based on the approximation

In fading channel case, probability of error for BPSK is a very complex term (relative to BPSK in AWGN), but it can be approximated by Taylor series, result as pretty well figure 2, and figure 3. For large SNR, actual data and approximated probability of error are indistinguishable. Hence, one can use approximated result for BPSK in fading channel.

Based on the equal gain combining

As we can see in figure 6 and figure 7, by transmitting same bit multiple times, one can reduce the probability of error drastically for a given SNR in fading channel.

7 Result/Conclusion

7.1 What did I learn?

1. I derived the fading channel model.
2. I derived the BPSK scheme in fading channel model (ML rule and Probability of error). I tested that results through c++ program.
3. I tested with same bit multiple times to improve performance.
4. I learned about overall performance of fading channel through BPSK scheme.