# Simulating Different Receivers in a Rayleigh, SISO Environment Winter Intern Seminar (Project #1)

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모든 실험은 다음의 조건하에 진행되었다.

- Es/N0는 -2dB 20dB(2dB 간격)
- SISO; Single Input, Single Output

## 1 Binary Moduation

### 공통조건

```
NumberIteration = 10^4;
LengthBitSequence = 10^2;

EbN0_dB = -2 : 1 : 20;
EbN0 = db2pow(EbN0_dB);
```

## 1.1 ZF(Zero-forcing)

```
ErrorCount_ZF = zeros(1, length(EbN0_dB));
2
3
   for iTotal = 1 : NumberIteration
       BitSequence = randi([0 1], 1, LengthBitSequence); % Bit Generation (
4
          BitSequence = rand(1, LengthBitSequence) > 0.5;)
       % SymbolSequence = qammod(BitSequence, 2, 'bin'); % 1 and -1
       SymbolSequence = 2 * BitSequence - 1; % Symbol (s) Generation;
6
          consisting of 1 and -1
       NoiseSequence = (randn(1, length(SymbolSequence)) + 1j * randn(1,
          length(SymbolSequence))) ./ sqrt(2); % Noise (n) Generation
       H = (randn(1, length(SymbolSequence)) + 1j * randn(1, length(
8
          SymbolSequence))) ./ sqrt(2); % Channel (h) Generation
       for indx_EbN0 = 1 : length(EbN0)
9
           ReceivedSymbolSequence_H = H .* SymbolSequence + NoiseSequence *
              sqrt(1 / EbNO(indx_EbNO)); % Received Signal (y = hs + n)
              Generation
11
           % ZF Receiver
12
           DetectionSymbolSequence_ZF = ReceivedSymbolSequence_H ./ H; %
13
              Detection (Zero-Forcing: y / h)
14
           % MMSE Receiver
```

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```
16
           rho = EbNO(indx_EbNO);
17
           w_mmse = conj(H) ./ (H.*conj(H)+1/rho);
18
           DetectionSymbolSequence_MMSE = ReceivedSymbolSequence_H .* w_mmse;
19
20
           % MLD Receiver
21
           arg = ([1 1]' * ReceivedSymbolSequence_H) - ([-1 1]' * H);
22
           arg = arg .* conj(arg);
           DetectionSymbolSequence_MLD = (arg(1, :)-arg(2, :)) > 0; % TODO:
23
               could possibly simplify it more
24
25
           % Symbol Sequence -> Bit Sequence
26
           DetectionBitSequence_ZF = real(DetectionSymbolSequence_ZF)>0;
27
           DetectionBitSequence_MMSE = real(DetectionSymbolSequence_MMSE)>0;
               % not sure if this is right
28
           DetectionBitSequence_MLD = DetectionSymbolSequence_MLD;
29
30
           ErrorCount_ZF(1, indx_EbN0) = ErrorCount_ZF(1, indx_EbN0) + biterr
               (DetectionBitSequence_ZF, BitSequence);
           ErrorCount_MMSE(1, indx_EbN0) = ErrorCount_MMSE(1, indx_EbN0) +
31
               biterr(DetectionBitSequence_MMSE, BitSequence);
           ErrorCount_MLD(1, indx_EbN0) = ErrorCount_MLD(1, indx_EbN0) +
               biterr(DetectionBitSequence_MLD, BitSequence);
       end
34
   end
36 | scatterplot(SymbolSequence)
37
38 | BER_Simulation_ZF = ErrorCount_ZF / (LengthBitSequence * NumberIteration);
39
   BER_Simulation_MMSE = ErrorCount_MMSE / (LengthBitSequence *
      NumberIteration);
   BER_Simulation_MLD = ErrorCount_MLD / (LengthBitSequence * NumberIteration
40
      );
41
42 BER_Theory2_H = berfading(EbN0_dB, 'psk', 2, 1);
43
44 % Plot
45 | figure()
46 | semilogy(EbNO_dB, BER_Theory2_H, 'k--');
47 | hold on
48 | %hold on don't know what it does
49 | semilogy(EbNO_dB, BER_Simulation_ZF, 'bo');
50 | semilogy(EbNO_dB, BER_Simulation_MMSE, 'rx');
  semilogy(EbNO_dB, BER_Simulation_MLD, 'g^');
   axis([-2\ 20\ 10^-5\ 0.5]) % axis([a\ b\ c\ d]) x-axis from a to b, y-axis from
      c to d
   grid on
53
54 | legend('Theory', 'ZF', 'MMSE', 'MLD');
55 | xlabel('Eb/No [dB]');
56 | ylabel('BER');
  title('BER for Binary Modulation');
```

- 1.2 ZF(Zero-forcing)
- 1.3 MMSE(Minimum Mean Square Error)
- 1.4 MLD(Maximum Likelihood Detection)

## 2 M-ary QAM

 $M=2^2n\;(n=1,2,3,...)$ 의 상황을 가정하였다. 한 가지 생각해볼 만한 사항은 Normalization Factor이다. 이 Normalization Factor를 사용하여 평균 전력이 1W가 되게끔 둘 수 있다.

QAM의 일반적인 Constellation Diagram을 살펴보면 실수  $\sqrt{M}$ 개, 허수  $\sqrt{M}$ 개의 point를 갖는 것을 알 수 있다.

하나의 지점을 하나의 alphabet이라고하자. M개의 alphabet은 다음과 같다.

$$alphabet = \pm (2n-1) \pm j \cdot (2n-1) \qquad n \in 1, 2, ..., \sqrt{M}$$

$$\tag{1}$$

신호의 평균전력은 다음과 같이 일반화 가능하다.

$$E_{s} = E[|s|^{2}] = \frac{1}{M} \sum_{n=1}^{M} |s_{n}|^{2}$$

$$= \frac{1}{M} \sum_{n=1}^{M} \sum_{m=1}^{M} |(2m-1)^{2} + (2n-1)^{2}|$$

$$= \frac{1}{M} \cdot 4 \sum_{n=1}^{\frac{\sqrt{M}}{2}} [(2n-1)^{2} \cdot \sqrt{M}]$$

$$= \frac{4}{M} \sum_{n=1}^{\frac{\sqrt{M}}{2}} [4n^{2} - 4n + 1]$$

$$= \frac{2}{3}(M-1)$$
(2)

(2)에서의 결과를 토대로 normalization이 이뤄진 alphabet을 구할 수 있다.

$$normalized \quad alphabet = \left[ \pm \frac{2n-1}{\sqrt{\frac{2}{3}(M-1)}} \pm j \cdot \frac{2n-1}{\sqrt{\frac{2}{3}(M-1)}} \right] \qquad n \in \{1, 2, ..., \sqrt{M}\}$$
 (3)

해당 결과를 토대로 다시 평균 전력을 구한다면  $E_s$ 가 1W임을 확인할 수 있다.

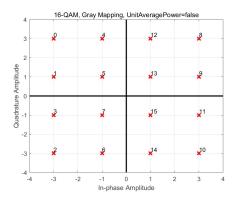
#### 참고자료

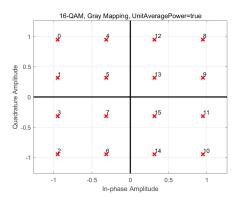
다음은 16-QAM의 Constellation이다.

- 2.1 ZF(Zero-forcing)
- 2.2 MMSE(Minimum Mean Square Error)
- 2.3 MLD(Maximum Likelihood Detection)

# 3 과제 외적 의문점

• 왜 M=4 일때의  $Error\ Count$ 는 ZF와 MMSE의 경우 다르지만, M=16일 때는 왜 모든 값이 같은가?





(a) Non-normalized Constellation

(b) Normalized Constellation

Figure 1: 16-QAM Constellation