

EXperiment CL1 : MIMO Communications

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EX1

1. Set up basic Parameters

```
clear all; clc; close all;
```

```
% Basic Parameters
```

```
SNR_dB = 0:1:30; % SNR in dB
```

```
SNR_linear = power(10, SNR_dB / 10); % SNR in linear scale
```

```
N = 5:1:10; % Number of input and output antennas at the MIMO system
```

```
num_simulations = 10000; % Number of Monte Carlo simulations
```

```
C_ergodic = zeros(length(N), length(SNR_dB)); % Predefine Shannon capacity matrix
```

use matlab to set the SNR values ranging from 0 dB to 30 dB and number of antennas from 5 to 10. I also define the number of Monte Carlo simulation times and C_ergodic matrix to store result

2. Monte Carlo Simulation

For MIMO system, $N_R = N_T = 2$, the Rayleigh fading channel matrix can be presents as:

$$H_{ij} = \frac{X_{ij} + jY_{ij}}{2}$$

where X_{ij} , Y_{ij} are i.i.d Gaussian r.v. with zero mean and unit variance AND their size is $n_r \times n_t$

due to H are statistically independent over the time,
for each ergodic channels, the Shannon Capacity is equal to
the statistical average

$$C_{\text{ergodic}} = E[C_H]$$

It is hard to derive close-form expression of MIMO Capacity,
so I use two nested loop to iterate over each SNR
value and each number of antennas given by requirement.
For each SNR and antennas, I will run 10,000 times
Monte Carlo simulation to obtain the statistical
average Capacity

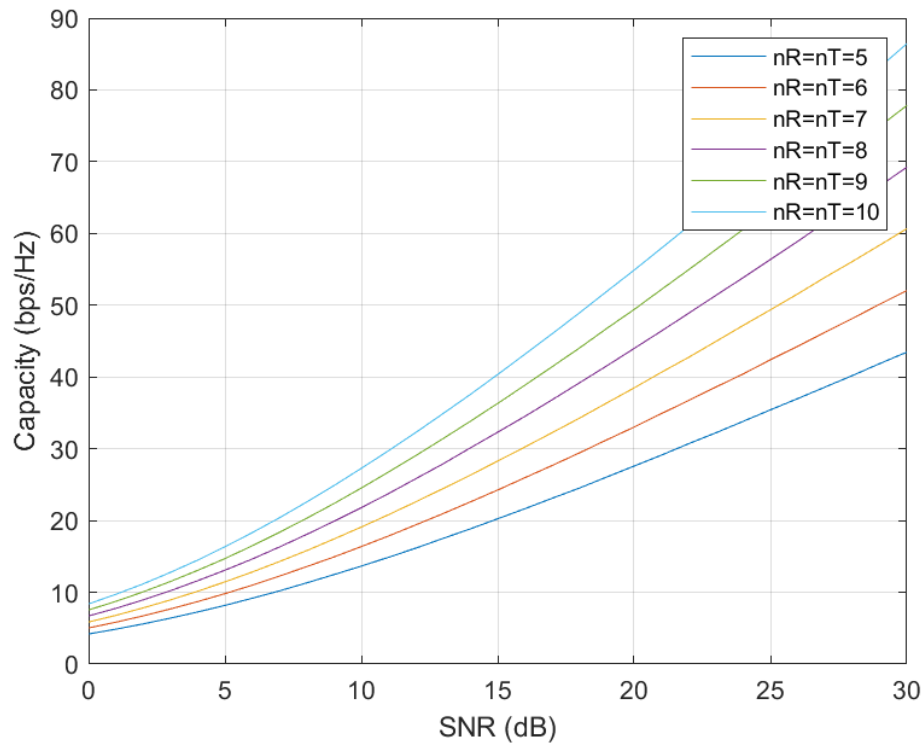
code shown below:

```
% for each defined SNR
for SNR_index = 1:length(SNR_dB)
    SNR = SNR_linear(SNR_index);

    % for each number of antennas at the MIMO system
    for N_index = 1:length(N)
        % define nR (number of receive antennas) = nT (number of transmit
        % antennas) in MIMO system
        nT = N(N_index);
        nR = nT;
        sum_capacity = 0; % Predefine Sum the capacity over all Monte Carlo simulations
        % Monte Carlo simulation
        for sim_loop = 1:num_simulations
            % Generate random Rayleigh fading channel matrix
            H = sqrt(1/2)*(randn(nR, nT) + 1i * randn(nR, nT));
            % Calculate instantaneous channel capacity in this simulation
            C_H = log2(det(eye(nR) + SNR/nT * (H' * H)));
            % Sum the capacity over all simulations
            sum_capacity = sum_capacity + C_H;
        end

        % Statistical average the capacity over all simulations
        C_ergodic(N_index, SNR_index) = sum_capacity / num_simulations;
    end
end
```

3. plot and results analysis



We can clearly find that as the increase of antenna number $n_R=n_T$, the capacity increases in different SNR.

EX2

1. Parameters setting up

Firstly, I initialize basic parameters like the number of antennas $N_R = N_T = 2$, measured SNR range, modulation order $M = 4$ [For QPSK], BER matrices etc.

AND the simulation data size has been defined here as well

$$\begin{cases} \text{num_bits} = 100 & \% \text{number of sended bits.} \\ H_rounds = 3000 & \% \text{Simulation times for each SNR.} \end{cases}$$

Noted:

Here, I choose bigger number of random matrices H than recommended 300 in remark 2 to obtain more precise and more smooth curve

Lastly, I generate random sended symbols sequence based on :

$$N_{\text{symbol}} = N_{\text{bits}} * \log_2 M$$

Code shown below

```
clear all;clc;close all;
```

% Basic Parameters

```
nT = 2; % Number of transmit antennas
```

```
nR = nT; % Number of receive antennas
```

```
SNR_dB_range= 0:5:35; % SNR in dB
```

```
step = 5;% SNR Iteration Step
```

```
M = 4; % QPSK modulation, so M=4
```

```
num_bits = 1e6;% Number of sended bits
```

```
H_rounds = 3000; % Number of random matrices H for each SNR
```

```
bit_symbol = log2(M); % bit per symbol
```

```
num_symbols = num_bits/ bit_symbol;% Number of sended symbols
```

```
AVG_BER_ZF = zeros(1,length(SNR_dB_range));%ZF detection Bit Error Rate
```

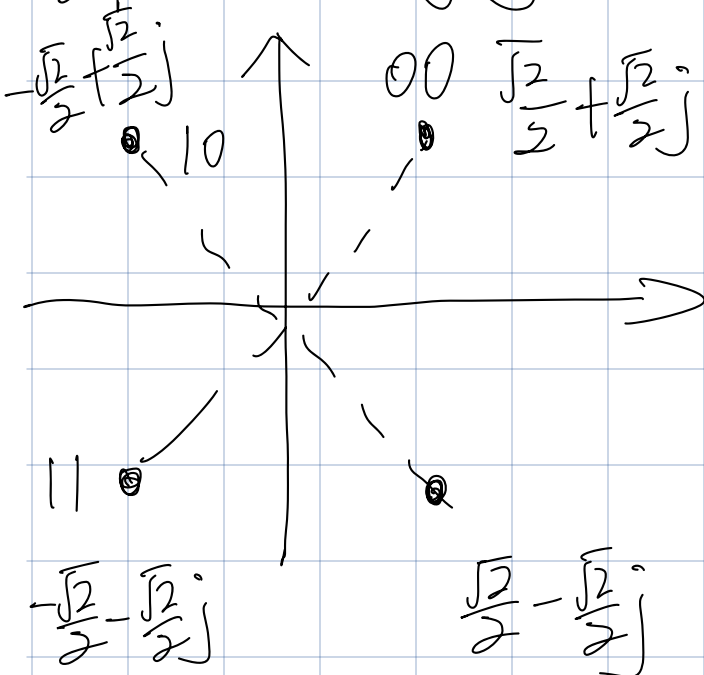
```
AVG_BER_ML = zeros(1,length(SNR_dB_range));%ML detection Bit Error Rate
```

```
AVG_BER_MMSE = zeros(1,length(SNR_dB_range));%MMSE detection Bit Error Rate
```

```
Tx_msg = randi([0, M-1], 1, num_symbols); % Generate random sended data
```

2. Simulation

QPSK with gray:



Due to 4PAM = QPSK.

I use qammod fuction to obtain gray mapping

```
function Rx= QPSK(M,num_symbols,SNR_dB,Tx_msg, H)
% Perform QPSK modulation using gray code
Tx= qammod(Tx_msg, M, 'gray','UnitAveragePower', true);

% Reshape symbols into 2xN matrix for 2x2 MIMO
Tx = reshape(Tx, 2, []);

% Noise Matrix
noise = sqrt(1/2) * (randn(2, num_symbols/2) + 1i*randn(2, num_symbols/2));

% Receive Matrix (Y = HX + N)
%Rx = H* Tx + noise;

SNR_linear= power(10,SNR_dB/10);
Rx = sqrt(SNR_linear/ 2) *H * Tx + noise;
```

$$y = \sqrt{\frac{P}{n_T}} Hx + n$$

ML Detection:

$$\hat{s} = \arg \min_{s \in \mathcal{C}} \|y - \sqrt{\frac{P}{n_T}} H x\|^2$$

\Rightarrow compare standard point to find nearest one

ZF Detection:

$$\hat{s} = \mathcal{Q} \left\{ \left(\sqrt{\frac{P}{n_T}} H \right)^H y \right\}$$

MMSE Detection:

$$\hat{s} = \mathcal{Q} \left\{ \left(\frac{P}{n_T} H^H H + I \right)^{-1} \sqrt{\frac{P}{n_T}} H^H y \right\}$$

Parents&Images

CL experiment.pdf

Exercise1.m

Exercise2.m

Exercise3.m

ML_Detection.m

ML_Detection_Alam.m

MMSE_Detection.m

QPSK.m

QPSK_Alam.m

Quantization.m

ZF_Detection.m

Here is my program structure, I define some functions:

ML detection

MMSE detection

QPSK modulation with gray mapping

ZF detection

Here is main simulation code:

```

parfor SNR_loop= 1: 8% Multithreaded parallel computing
    SUM_BER_ML = 0;
    SUM_BER_ZF = 0;
    SUM_BER_MMSE = 0;
    SNR_dB = (SNR_loop-1)*5
    for H_loop = 1:H_rounds
        % Generate random Rayleigh fading channel matrix
        H = sqrt(1/2)*(randn(nR, nT) + 1i * randn(nR, nT));

        %QPSK modulation
        Rx = QPSK(M,num_symbols,SNR_dB,Tx_msg, H);

        % ML Detection
        ML_symbols = ML_Detection(Rx,M,H,num_symbols,nT,SNR_dB);

        % ZF Detection
        ZF_symbols = ZF_Detection(Rx,SNR_dB,H,nT);

        % MMSE Detection
        MMSE_symbols = MMSE_Detection(Rx,SNR_dB,H,nT);

        % Demodulate using QAM demodulation
        ML_Rx_msg = qamdemod(reshape(ML_symbols, 1, []), M, 'gray', 'UnitAveragePower', true);
        ZF_Rx_msg = qamdemod(reshape(ZF_symbols, 1, []), M, 'gray', 'UnitAveragePower', true);
        MMSE_Rx_msg = qamdemod(reshape(MMSE_symbols, 1, []), M, 'gray', 'UnitAveragePower', true);

        % Calculate BER
        BER_ML= sum(Tx_msg ~= ML_Rx_msg) / num_symbols;
        SUM_BER_ML = SUM_BER_ML + BER_ML;

        BER_ZF= sum(Tx_msg ~= ZF_Rx_msg) / num_symbols;
        SUM_BER_ZF = SUM_BER_ZF + BER_ZF;

        BER_MMSE= sum(Tx_msg ~= MMSE_Rx_msg) / num_symbols;
        SUM_BER_MMSE = SUM_BER_MMSE + BER_MMSE;
    end
end
% Statistical average the BER
AVG_BER_ML(SNR_loop) = SUM_BER_ML/H_rounds;
AVG_BER_ZF(SNR_loop) = SUM_BER_ZF/H_rounds;
AVG_BER_MMSE(SNR_loop) = SUM_BER_MMSE/H_rounds;
end

```

→ use "parfor" than "for"
to open multithread pool
to speed up computation.

→ symbol₂ symbol modulation
with gray mapping

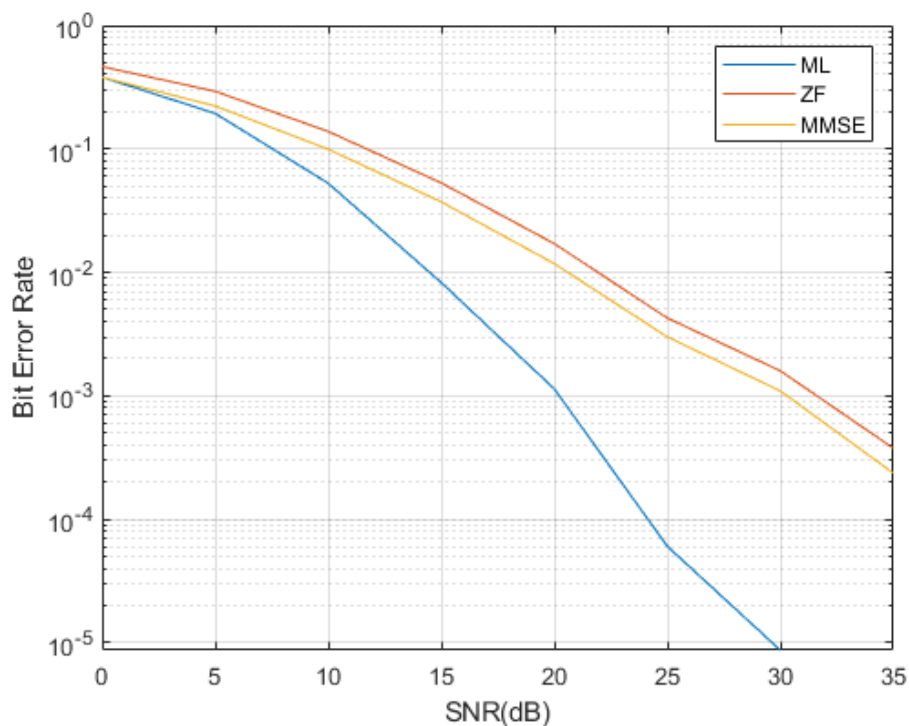
} collect each loop
results.

3. result analysis

In order to get more smooth and precise Curve,
I run 3000 H and 10⁶ bits

I find performance:

$$ML > MMSE > ZF$$



EX3

Alamouti coding for 2 antennas

$$X = \begin{bmatrix} x_1 & -x_2^* \\ x_2 & x_1^* \end{bmatrix}$$

```
function Rx= QPSK_Alam(M,num_symbols,SNR_dB,Tx_msg, H)
% Perform QPSK modulation using gray code
Tx= qammod(Tx_msg, M, 'gray','UnitAveragePower', true);

% Alamouti encoding
Tx_Alam = reshape(Tx, 2, []);
Tx_Alam = [Tx_Alam; [-conj(Tx_Alam(2,:)); conj(Tx_Alam(1,:))]];

% Convert the encoded symbols back to a single column for transmission
Tx_Alam = reshape(Tx_Alam, [], 1);

% Noise Matrix
noise = sqrt(1/2) * ((randn(2, num_symbols) + 1i*randn(2, num_symbols)));
SNR_linear= power(10,SNR_dB/10);
% Receive Matrix (Y = HX + N)
Rx = sqrt(SNR_linear/ 2) * H * reshape(Tx_Alam, 2, []) + noise;
```

modify QPSK in EX2

keep similar main simulation code structure

Also, run 3000 H and 10^6 bits:

I find performance:

ML with Alamouti coding better than

ML without Alamouti coding.

