

## CT216-Lab06

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### 1 Analysis Exercise

Lab-6

#### \* Analysis Exercise

1. We applied WLLN for showing that probability of  $q^N$  success is :-

$$q^{qN} (1-q)^{N-qN}$$

For finding no. of such sequences we used :-

$$k \times q^{qN} (1-q)^{N-qN} = 1 \quad (\text{Prob. of whole typical set})$$

$$\therefore k = q^{-qN} (1-q)^{2N-N}$$

Taking  $\log_2()$  both sides  $\rightarrow$

$$\log_2 k = -qN \log_2 q + (qN - N) \log_2 (1-q)$$

$$= -qN \log_2 q + qN \log_2 (1-q) - N \log_2 (1-q)$$

$$= +N (-q \log_2 q - (1-q) \log_2 (1-q))$$

$$= NH_2(q)$$

$$\therefore k = 2^{NH_2(q)} = 2^N$$

Now, using Sterling's approx.  $\rightarrow$

$$\log_2 \binom{N}{r} \approx NH_2\left(\frac{r}{N}\right)$$

Here  $r = qN$

$$\therefore \log_2 \binom{N}{qN} \approx NH_2(q)$$

$$\therefore \binom{N}{qN} \approx 2^{NH_2(q) - O(1)}$$

Therefore from (i) & (ii) we get the same result that # of such binary sequences  $\binom{N}{qN}$  is  $2^{NH_2(q)}$  as shown directly from Shannon's Noiseless Channel Coding.

2.

(a) We are asked the # of times  $X_k$  occur, which is basically expectation of  $X_k$  in the sequence.

$$\therefore E[X_k] = N p_k$$

Here as  $N \rightarrow \infty$

$$\therefore E[X_k] = \lim_{N \rightarrow \infty} N p_k$$

(b) Probability of these sequences can be calculated as:

$$P_{\text{sequence}} = \lim_{N \rightarrow \infty} \prod_{k=1}^K p_k^{N p_k} \left[ \begin{array}{l} (N \rightarrow \infty, P_{\text{sequence}} \rightarrow 1) \\ \text{(We know } E[X] = N p_k) \end{array} \right]$$

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(c) Since our Universal set has been cutoff from  $\mathcal{Q}^N$ , so we wouldn't see any other sequence other than the typical sequence

Hence  $\rightarrow$

$$Z = \lim_{N \rightarrow \infty} \prod_{k=1}^K \frac{N \cdot p_k}{p_k} = 1 \quad (Z \text{ is \# of typical sequences})$$

Taking  $\log_2()$  both sides  $=$

$$\therefore Z = \lim_{N \rightarrow \infty} \prod_{k=1}^K p_k^{-N \cdot p_k}$$

$$\log_2(Z) = -N \sum_{k=1}^K p_k \log_2(p_k)$$

$$\text{Entropy of K-ary source} = H_2(N) = - \sum_{k=1}^K p_k \log_2(p_k)$$

$$\therefore \log_2(Z) = -N H_2(X)$$

$$\therefore \log_2(Z) = -N (-H_2(X))$$

$$\therefore Z = 2^{N H_2(X)}$$

\* Proof - 2

∴ Linearity Property says if  $c_1$  &  $c_2$  are valid codeword then so are  $c_1 + c_2$ .

$$\text{Let } c_1 = [c_{1,1}, c_{1,2}, \dots, c_{1,n}]^T$$

$$c_2 = [c_{2,1}, c_{2,2}, \dots, c_{2,n}]^T$$

$$\text{As } c_1, c_2 \rightarrow \text{codeword } Hc_1 = 0 \text{ \& } Hc_2 = 0$$

$$c_1 + c_2 = [c_{1,1} + c_{2,1}, c_{1,2} + c_{2,2}, \dots, c_{1,n} + c_{2,n}]^T$$

$$c_1 - c_2 = [c_{1,1} - c_{2,1}, c_{1,2} - c_{2,2}, \dots, c_{1,n} - c_{2,n}]^T$$

$$\begin{aligned} [H]_{n \times n} [c_1 + c_2]_{n \times 1} &= [H]_{n \times n} [c_1]_{n \times 1} + [H]_{n \times n} [c_2]_{n \times 1} \quad (\text{Using Linearity property of Matrix Multiplication}) \\ &= 0 + 0 \\ &= 0 \end{aligned}$$

Similarly

$$[H]_{n \times n} [c_1 - c_2]_{n \times 1} = 0 - 0 = 0$$

Hence as  $S_{c_1 + c_2} = 0 = S_{c_1 - c_2} \rightarrow c_1 + c_2$  &  $c_1 - c_2$  are also valid codeword.

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2. Syndrome Prop: I says that if code is linear  
 $S$  remains unchanged when transmitter sends  
 $c'$  or  $c$ .

Let

$$r = [r_1, r_2, \dots, r_n]^T, c = [c_1, c_2, \dots, c_n]^T$$

$$e = [e_1, e_2, \dots, e_n]^T$$

$$r = c + e \rightarrow r_1 = c_1 + e_1, r_2 = c_2 + e_2 \rightarrow r_n = c_n + e_n$$

As  $c$  is valid codeword

$$S_r = [H][r]^T$$

$$S_{r'} = [H][c + e]^T$$

$$= [H][c]^T + [H][e]^T$$

$$= 0 + S_e$$

If any other codeword  $c'$   
 then

$$S_{r'} = [H][r']^T$$

$$= [H][c' + e]^T$$

$$= [H][c']^T + [H][e]^T$$

$$= 0 + S_e$$

$$S_{r'} = S_{r_c}$$

$\therefore$  We showed that if code is linear  $S$  vector  
 depends only on  $e$  vector.

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3. Syndrome Prop II says that we can determine the error by calculating  $S$  (unless  $c = e$ )

Let us say there is no error

$$\therefore r = c$$

$$\therefore S = [H][r]$$

$$= [H][c]$$

$$S = 0$$

(as  $c \rightarrow$  valid codeword)

Now, introducing error  $\rightarrow r' = c + e$

$$\therefore S = [H][r']$$

$$= [H][c] + [H][e]$$

$$S = 0 + S_e \neq 0$$

As  $S \neq 0 \rightarrow$  error detected.

But if we take  $e = c'$  ( $c'$  is valid codeword)

$$r'' = c + e = c + c' = c'' \text{ (valid codeword)}$$

(As we proved that  $c + c'$  is also a valid codeword using linearity)

$$\therefore S = [H][r'']$$

$$= [H][c + c']$$

$$= [H][c'']$$

$$S = 0$$

hence passed

4. All zero  $(n \times 1)$  is always a valid codeword.  
(regardless of  $G$  &  $H$ )

$$\therefore z_{n \times 1} = [0, \dots, 0]^T$$

A codeword should satisfy  
 $z = [H][C] = 0$

Here

$$z_{n \times 1} = H_{m \times n} z_{n \times 1} = 0 \quad ([0] \text{ vector} \times \text{any vector is } [0])$$

$\therefore$  All Zero vector is always a codeword  
regardless what  $G$  &  $H$  are.

5. To show that  $d_{\min}^H = w_{\min}$  (let us prove  
through contradiction by taking  $d_{\min}^H < w_{\min}$ ).

$c_1 = z_{n \times 1}$  (all zero codeword  $\rightarrow$  valid)

$c_2 = \text{min-Hamming wt codeword}$

As we took  $d_{\min}^H < w_{\min}$ ,  $\therefore$  there exists a sequence  
whose distance is greater than  $w_{\min}$ .

But as we took  $c_1$  &  $c_2$  as  $c_1 = z_{n \times 1}$  &  
 $c_2 = C_{w_{\min}}$

The distance b/w them is equal to  $w_{\min}$   
which is less than what we assumed.

Hence  $d_{\min}^H < w_{\min}$  is false

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Now  $d_{\min}^H \leq w_{\min}$

This can also be proved false in the same way by taking  $c_1 = [0]_n$  &  $c_2 = c_{w_{\min}}$ .

We are assuming  $d_{\min}^H < w_{\min}$ , but for  $c_1$  &  $c_2$   $d^H = w_{\min}$  which in turn is  $d_{\min}^H$ .

Hence only case possible is  $d_{\min}^H = w_{\min}$  for linear codes.

G.  $r = \frac{16}{9} \therefore k = 4, n = 7, u = n - k = 3$

Let parity  $\rightarrow \left. \begin{aligned} p_1 &= m_0 + m_1 + m_2 \\ p_2 &= m_1 + m_2 + m_3 \\ p_3 &= m_2 + m_3 + m_4 \end{aligned} \right\} \begin{array}{l} \text{modulo-2} \\ \text{sum} \end{array}$

$\therefore H = \begin{bmatrix} 1 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}_{3 \times 7}$

Let us take 4 msg.  $m_1 = \begin{bmatrix} 1 & 0 & 1 & 1 \end{bmatrix}$  ( $p_1 = 0; p_2 = 0; p_3 = 0$ )  
 $c_1 = [1011000]^T$

$s = [H][c_1] = \begin{bmatrix} 1 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$   
 $= \begin{bmatrix} 2 \\ 2 \\ 2 \end{bmatrix} \text{ mod } 2 = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

Valid

Teacher's Signature.....



Now let us say 1 bit got flipped during transmission:

$$c_2 = [1001000]^T \text{ instead of } [1011000]^T$$

$$\therefore s = [H][c_2]$$

$$= \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} \text{ mod } 2$$

$$= \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

Error detected. And we can check "where" by taking help from  $P_1, P_2, P_3$ .

Since for 1 & 2 error is detected,  $P_1$  &  $P_2$  has some anomaly in it.

$\Rightarrow$  Since  $P_3$  is correct,  $m_1, m_2$  are considered to be correct, error lies in  $m_3$ . Hence can be corrected by.

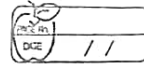
\* Now if we introduce 1 more error (2-bit error) i.e.:-

$$c_3 = [1101000]^T$$

$$s = [H][c_3]$$

$$= \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

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So new 122 are received correctly as  
1<sup>st</sup> 2 rows of 8 are 0 but 3<sup>rd</sup> row has 1 so  
some error regarding  $p$  or we can say in  
 $m_0, m_1, m_2$

But as  $p_1, p_2$  has no anomaly, so we wouldn't  
be able to make out which bit is in error.

→ Hence we were able to correct & detect for  
1 bit error but only detect for 2-bit error.

$\therefore t_2 = 1 \text{ bit}$

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```
disp('Question 1');
```

Question 1

```
disp('(7,4) Hamming Code');
```

(7,4) Hamming Code

```
n=7;  
k=4;  
u=n-k;
```

```
%{  
    We define parity as:-  
    p1=m0+m1+m2  
    p2=m1+m2+m3  
    p3=m0+m1+m3  
%}
```

Therefore first 4 rows are 4x4 identity matrix and next 3 rows are the parity checkers.

```
%}  
  
% Parity matrix  
P=[
```

```
    1 1 0 1;  
    1 0 1 1;  
    0 1 1 1;  
    ]
```

P = 3x4

1	1	0	1
1	0	1	1
0	1	1	1

```
identity_matrix=eye(k);  
G=[identity_matrix;P];  
disp('Generator Matrix:');
```

Generator Matrix:

```
disp(G);
```

1	0	0	0
0	1	0	0
0	0	1	0
0	0	0	1
1	1	0	1
1	0	1	1
0	1	1	1

```
identity_parity=eye(u);  
H=[P identity_parity];
```

```
disp('Parity Check Matrix:');
```

Parity Check Matrix:

```
disp(H);
```

1	1	0	1	1	0	0
1	0	1	1	0	1	0
0	1	1	1	0	0	1

```
decnum=0:(2^k-1);  
binnum=dec2bin(decnum);  
M=zeros(k,length(decnum));
```

```
for i=1:length(decnum)  
    M(:,i)=binnum(i,:)-'0';  
end
```

```
disp('Matrix representing all possible length k(here 4) sequences:');
```

Matrix representing all possible length k(here 4) sequences:

```
disp(M);
```

0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1
0	0	0	0	1	1	1	1	0	0	0	0	1	1	1	1
0	0	1	1	0	0	1	1	0	0	1	1	0	0	1	1
0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1

```
C=mod(G*M,2);  
disp('All possible Codeword Matrix(length 7):');
```

All possible Codeword Matrix(length 7):

```
disp(C);
```

0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1
0	0	0	0	1	1	1	1	0	0	0	0	1	1	1	1
0	0	1	1	0	0	1	1	0	0	1	1	0	0	1	1
0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1
0	1	0	1	1	0	1	0	1	0	1	0	0	1	0	1
0	1	1	0	0	1	1	0	1	0	0	1	1	0	0	1
0	1	1	0	1	0	0	1	0	1	1	0	1	0	0	1

```
w=sum(C);  
w_min=min(w(2:length(C)));  
disp(['Minimum Hamming Weight=',num2str(w_min)]);
```

Minimum Hamming Weight=3

```
d_hmin=k;  
for cdwr1=1:length(C)  
    for cdwr2=cdwr1+1:length(C)
```

```

        d_h=sum(xor(C(:,cdwrd1),C(:,cdwrd2)));
        d_hmin=min(d_h,d_hmin);
    end
end

disp(['Minimum Hamming Distance:',num2str(d_hmin)]);

```

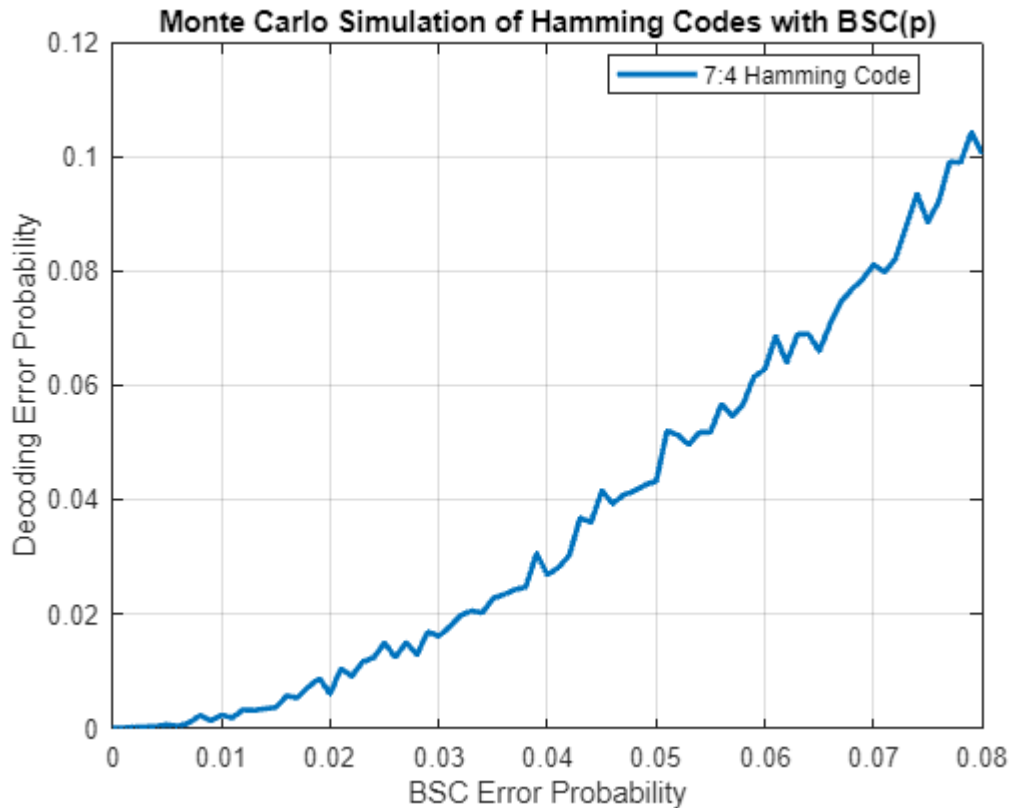
Minimum Hamming Distance:3

```

err=zeros(size(0:0.001:0.08));
nsim=1e4;

for p = 0:0.001:0.08 % Iterate over different error probabilities
    for ksim = 1:nsim
        m_in = randi([1, 2^k]); % Randomly select a message index
        m = M(:, m_in); % Get the message corresponding to the index
        c = C(:, m_in); % Get the codeword corresponding to the message
        % Introduce BSC noise
        bsc_noise = rand(size(c)) < p;
        r = mod(c + bsc_noise, 2); % Received codeword
        e=r-C;
        answer = sum(abs(e));
        [~,est_index] = min(answer);
        % Check for error
        if est_index ~= m_in
            err(round(p * 1000) + 1) = err(round(p * 1000) + 1) + 1; % Increment error count
        end
    end
end
% Convert error counts to error probabilities
err_prob= err / nsim;
% Plotting
p_values = 0:0.001:0.08;
plot(p_values, err_prob,'LineWidth',2);
xlabel('BSC Error Probability');
legend('7:4 Hamming Code','location','best');
ylabel('Decoding Error Probability');
title('Monte Carlo Simulation of Hamming Codes with BSC(p)');
grid("on");

```



```

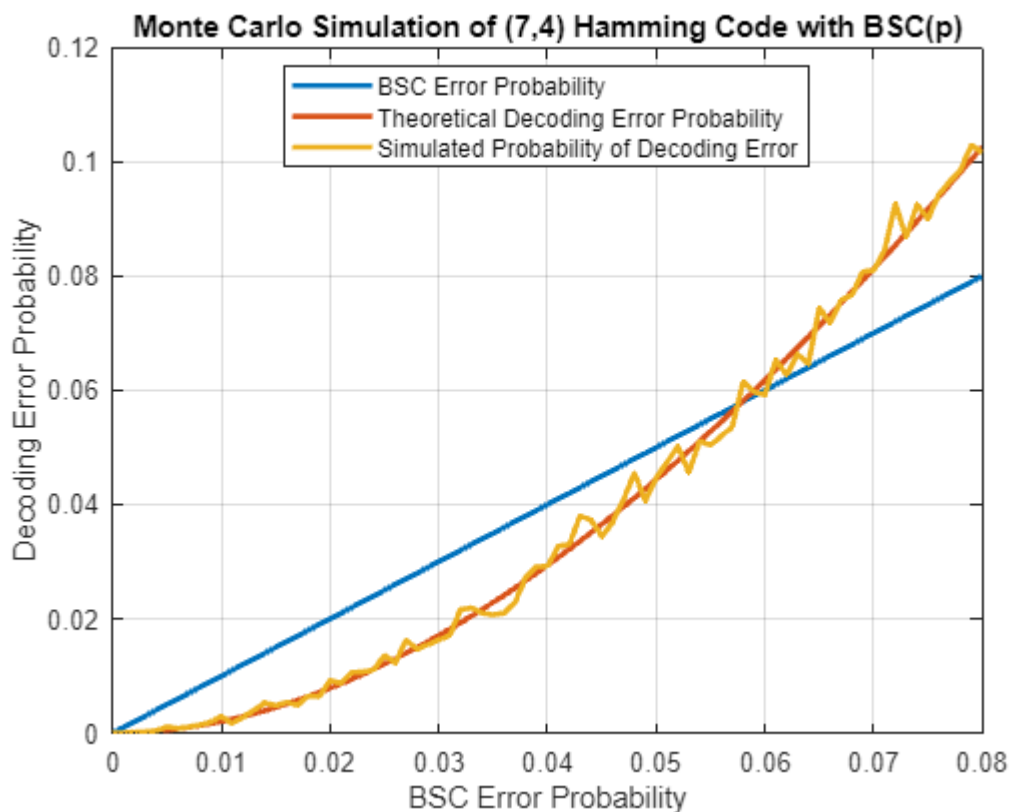
err = zeros(size(0:0.001:0.08)); % Simulated error count
theoretical_err = zeros(size(0:0.001:0.08)); % Theoretical error probability
nsim = 1e4; % Number of simulations
t_c=1;
for p = 0:0.001:0.08 % Iterate over different error probabilities
% Calculate theoretical probability of decoding error
% Hamming code minimum distance
d_min = 3;
% Theoretical probability of decoding error for Hamming code
for t = t_c+1:n
theoretical_err(round(p * 1000) + 1) = theoretical_err(round(p * 1000) + 1) +
nchoosek(n, t) * p^t * (1-p)^(n-t);
end
for ksim = 1:nsim
m_in = randi([1, 16]); % Randomly select a message index
m = M(:, m_in); % Get the message corresponding to the index
c = C(:, m_in); % Get the codeword corresponding to the message
% Introduce BSC noise
bsc_noise = rand(size(c)) < p;
r = mod(c + bsc_noise, 2); % Received codeword
e=r-C;
answer = sum(abs(e));
[~,est_index] = min(answer);
% Check for error
if est_index ~= m_in

```

```

err(round(p * 1000) + 1) = err(round(p * 1000) + 1) + 1; % Increment error count
end
end
end
% Convert error counts to error probabilities
err_prob = err / nsim;
% Plotting
p_values = 0:0.001:0.08;
plot(p_values, p_values, 'Color', '#0072BD', 'LineWidth', 2);
hold on;
plot(p_values, theoretical_err, 'Color', '#D95319', 'LineWidth', 2);
plot(p_values, err_prob, 'Color', '#EDB120', 'LineWidth', 2);
legend('BSC Error Probability', 'Theoretical Decoding Error Probability', 'Simulated
Probability of Decoding Error', Location='best');
xlabel('BSC Error Probability');
ylabel('Decoding Error Probability');
title('Monte Carlo Simulation of (7,4) Hamming Code with BSC(p)');
grid on;
hold off;

```



```
disp('Question 2');
```

Question 2

```
disp('(14,8) RPC Code');
```

```

n=14;
k=8;
u=n-k;

P=[
    1 0 0 0 1 0 0 0;
    0 1 0 0 0 1 0 0;
    0 0 1 0 0 0 1 0;
    0 0 0 1 0 0 0 1;
    1 1 1 1 0 0 0 0;
    0 0 0 0 1 1 1 1];

identity_matrix=eye(k);
G=[identity_matrix; P];
disp('Generator Matrix:');

```

Generator Matrix:

```
disp(G);
```

```

1  0  0  0  0  0  0  0
0  1  0  0  0  0  0  0
0  0  1  0  0  0  0  0
0  0  0  1  0  0  0  0
0  0  0  0  1  0  0  0
0  0  0  0  0  1  0  0
0  0  0  0  0  0  1  0
0  0  0  0  0  0  0  1
1  0  0  0  1  0  0  0
0  1  0  0  0  1  0  0
0  0  1  0  0  0  1  0
0  0  0  1  0  0  0  1
1  1  1  1  0  0  0  0
0  0  0  0  1  1  1  1

```

```

identity_parity=eye(u);
H=[P identity_parity];
disp('Parity Check Matrix:');

```

Parity Check Matrix:

```
disp(H);
```

```

1  0  0  0  1  0  0  0  1  0  0  0  0  0
0  1  0  0  0  1  0  0  0  1  0  0  0  0
0  0  1  0  0  0  1  0  0  0  1  0  0  0
0  0  0  1  0  0  0  1  0  0  0  1  0  0
1  1  1  1  0  0  0  0  0  0  0  0  1  0
0  0  0  0  1  1  1  1  0  0  0  0  0  1

```

```
decnum=0:(2^k-1);
```



```

binnum=dec2bin(decnum);
M=zeros(k,length(decnum));

for i=1:length(decnum)
    M(:,i)=binnum(i,:)-'0';
end

disp('Matrix representing all possible length k(here 8) sequences:');

```

Matrix representing all possible length k(here 8) sequences:

```
disp(M);
```

0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0	0
0	0	0	0	1	1	1	1	0	0	0	0	1	1	1	1	0	0
0	0	1	1	0	0	1	1	0	0	1	1	0	0	1	1	0	0
0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1

```

C=mod(G*M,2);
disp('All possible Codeword Matrix(length 14):');

```

All possible Codeword Matrix(length 14):

```
disp(C);
```

0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0	0
0	0	0	0	1	1	1	1	0	0	0	0	1	1	1	1	0	0
0	0	1	1	0	0	1	1	0	0	1	1	0	0	1	1	0	0
0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0	0
0	0	0	0	1	1	1	1	0	0	0	0	1	1	1	1	0	0
0	0	1	1	0	0	1	1	0	0	1	1	0	0	1	1	0	0
0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1	1	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1
0	1	1	0	1	0	0	1	1	0	0	1	0	1	1	0	0	1

```

w=sum(C);
w_min=min(w(2:length(C)));
disp(['Minimum Hamming Weight=',num2str(w_min)]);

```

Minimum Hamming Weight=3

```

d_hmin=k;
for cdwrd1=1:length(C)
    for cdwrd2=cdwrd1+1:length(C)
        d_h=sum(xor(C(:,cdwrd1),C(:,cdwrd2)));
        d_hmin=min(d_h,d_hmin);
    end
end

```

```

end
end

disp(['Minimum Hamming Distance:', num2str(d_hmin)]);

```

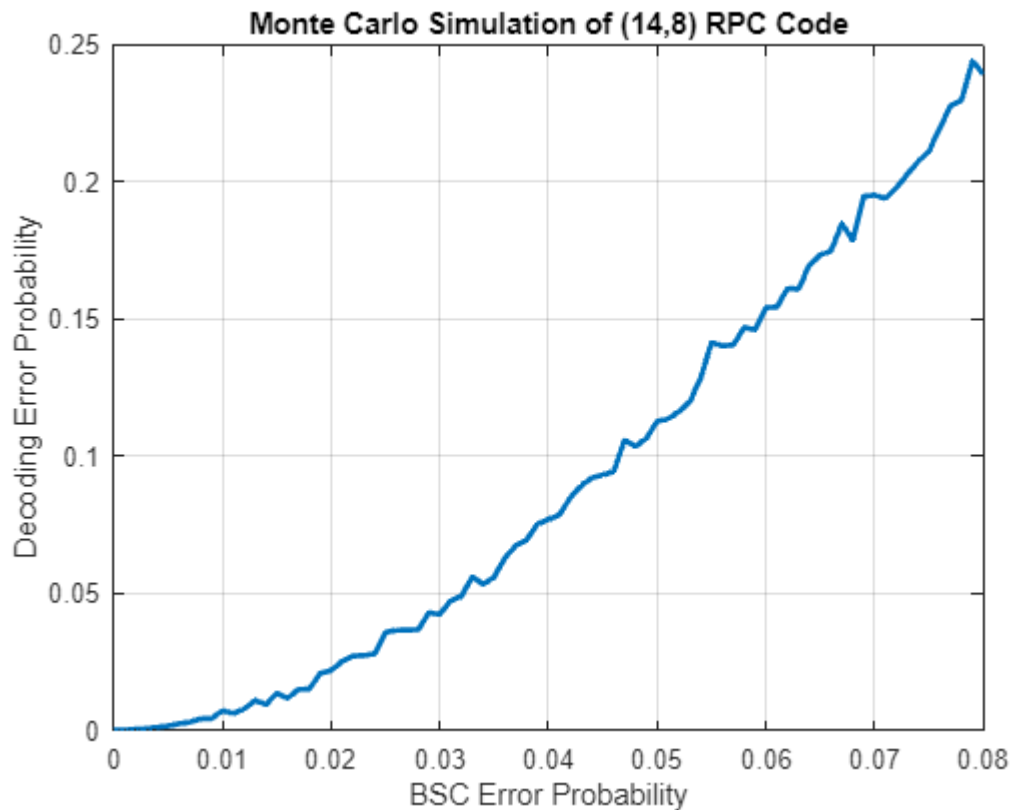
Minimum Hamming Distance:3

```

err=zeros(size(0:0.001:0.08));
nsim=1e4;

for p = 0:0.001:0.08 % Iterate over different error probabilities
for ksim = 1:nsim
m_in = randi([1, 2^k]); % Randomly select a message index
m = M(:, m_in); % Get the message corresponding to the index
c = C(:, m_in); % Get the codeword corresponding to the message
% Introduce BSC noise
bsc_noise = rand(size(c)) < p;
r = mod(c + bsc_noise, 2); % Received codeword
e=r-C;
answer = sum(abs(e));
[~,est_index] = min(answer);
% Check for error
if est_index ~= m_in
err(round(p * 1000) + 1) = err(round(p * 1000) + 1) + 1; % Increment error count
end
end
end
% Convert error counts to error probabilities
err_prob_1= err / nsim;
% Plotting
p_values = 0:0.001:0.08;
plot(p_values, err_prob_1,'LineWidth',2);
xlabel('BSC Error Probability');
ylabel('Decoding Error Probability');
title('Monte Carlo Simulation of (14,8) RPC Code');
grid("on");

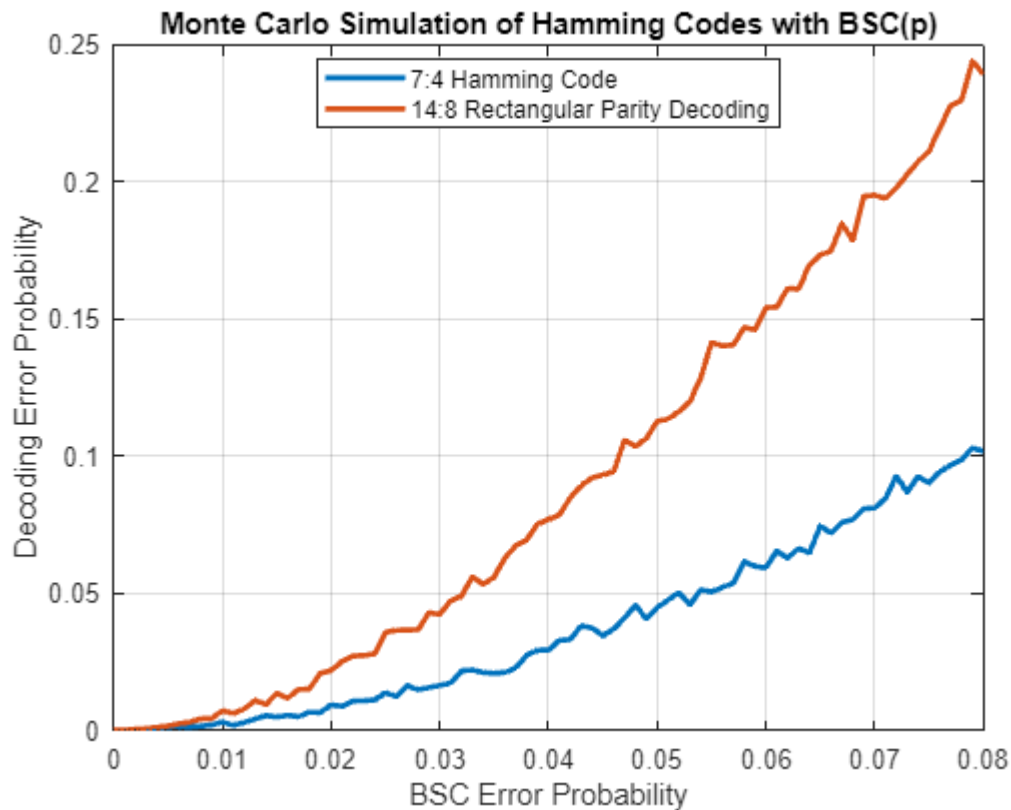
```



```
% Comparing RPC and Hamming Code
disp('Comparing Hamming Code with RPC Code.');
```

Comparing Hamming Code with RPC Code.

```
figure;
p_values = 0:0.001:0.08;
plot(p_values, err_prob, 'LineWidth', 2);
hold on;
plot(p_values, err_prob_1, 'LineWidth', 2);
xlabel('BSC Error Probability');
legend('7:4 Hamming Code', '14:8 Rectangular Parity Decoding', 'location', 'best');
ylabel('Decoding Error Probability');
title('Monte Carlo Simulation of Hamming Codes with BSC(p)');
grid("on");
hold off;
```



% From the above graph it is clear that the Hamming Code performs better  
 % though the rate are same for both.  
 % The reason behind this lies in the theory of perfect codes.

```
%{
r=4/7 and r=8/14(4/7) both have same rate but to check whether one is a
perfect code or not we show:
2^k.Volume(n,tc)=2^n or not where tc= # of correction bits and
Volume(n,tc)=sum(n choose i) { i=0 to tc}
```

For  $r=4/7$   
 there are 16 hamming spheres and  $16 \times 8$  binary sequences.  
 where  $2^n = 16 \times 8 = 128$   
 Hence a perfect Code

For  $r=4/7$  but with  $k=8$  and  $n=14$   
 $2^8 \times 15$  cannot be  $2^{14}$   
 Hence not a perfect code

Hence Hamming Code is better than RPC Code  
 %}

```
disp('Question 3');
```

### Question 3

```
disp('(9,4) Product Code');
```

(9,4) Product Code

```
n=9;
k=4;
u=n-k;

P=[
    1 0 1 0;
    0 1 0 1;
    1 1 0 0;
    0 0 1 1;
    1 1 1 1];

identity_matrix=eye(k);
identity_parity=eye(u);
G=[identity_matrix; P];
H=[P identity_parity];
disp('Generator Matrix:');
```

Generator Matrix:

```
disp(G);
```

```
1    0    0    0
0    1    0    0
0    0    1    0
0    0    0    1
1    0    1    0
0    1    0    1
1    1    0    0
0    0    1    1
1    1    1    1
```

```
disp('Parity Check Matrix:');
```

Parity Check Matrix:

```
disp(H);
```

```
1    0    1    0    1    0    0    0    0
0    1    0    1    0    1    0    0    0
1    1    0    0    0    0    1    0    0
0    0    1    1    0    0    0    1    0
1    1    1    1    0    0    0    0    1
```

```
decnum=0:(2^k-1);
binnum=dec2bin(decnum);
M=zeros(k,length(decnum));

for i=1:length(decnum)
```

```
M(:,i)=binnum(i,:)-'0';
end
```

```
disp('Matrix representing all possible length k(here 4) sequences:');
```

Matrix representing all possible length k(here 4) sequences:

```
disp(M);
```

```

0    0    0    0    0    0    0    0    1    1    1    1    1    1    1    1
0    0    0    0    1    1    1    1    0    0    0    0    1    1    1    1
0    0    1    1    0    0    1    1    0    0    1    1    0    0    1    1
0    1    0    1    0    1    0    1    0    1    0    1    0    1    0    1

```

```
C=mod(G*M,2);
disp('All possible Codeword Matrix(length 9):');
```

All possible Codeword Matrix(length 7):

```
disp(C);
```

```

0    0    0    0    0    0    0    0    1    1    1    1    1    1    1    1
0    0    0    0    1    1    1    1    0    0    0    0    1    1    1    1
0    0    1    1    0    0    1    1    0    0    1    1    0    0    1    1
0    1    0    1    0    1    0    1    0    1    0    1    0    1    0    1
0    0    1    1    0    0    1    1    1    1    0    0    1    1    0    0
0    1    0    1    1    0    1    0    0    1    0    1    1    0    1    0
0    0    0    0    1    1    1    1    1    1    1    1    0    0    0    0
0    1    1    0    0    1    1    0    0    1    1    0    0    1    1    0
0    1    1    0    1    0    0    1    1    0    0    1    0    1    1    0

```

```
w=sum(C);
w_min=min(w(2:length(C)));
disp(['Minimum Hamming Weight=',num2str(w_min)]);
```

Minimum Hamming Weight=4

```

d_hmin=k;
for cdwr1=1:length(C)
    for cdwr2=cdwr1+1:length(C)
        d_h=sum(xor(C(:,cdwr1),C(:,cdwr2)));
        d_hmin=min(d_h,d_hmin);
    end
end

disp(['Minimum Hamming Distance:',num2str(d_hmin)]);
```

Minimum Hamming Distance:4

```

err = zeros(size(0:0.1:1));% Initialize array to store error counts for each p
theoretical_prob = zeros(size(0:0.1:1));
nsim = 1000; % Number of simulations

```

```

for p = 0:0.1:1 % Iterate over different erasure probabilities
    for ksim = 1:nsim
        m_in = randi([1, 2^k]); % Randomly select a message index
        m = M(:, m_in); % Get the message corresponding to the index
        c = C(:, m_in); % Get the codeword corresponding to the message
        bec_noise = rand(size(c)) < p;
        % Apply BEC noise
        num = 5 * bec_noise; % Treat terms with more than 1 erasure as erasures
        r = num + c; % Received codeword
        % Set values greater than 4 to NaN
        r(r > 4) = NaN;
        theoretical_prob = binopdf(0, n, p_values) + binopdf(1, n, p_values) +
binopdf(2, n, p_values) + binopdf(3, n, p_values) + binopdf(4, n, p_values) +
binopdf(5, n, p_values) / 2;
        % Reshape the received codeword into a 3x3 matrix
        r_resaped = [r(1), r(2), r(5); r(3), r(4), r(6); r(7), r(8), r(9)];

        % Perform iterative decoding
        for iter = 1:3 % Iterations for both row and column decoding
            % Column decoding
            for col = 1:3
                nan_indices = find(isnan(r_resaped(:, col)));
                if length(nan_indices) == 1 % Exactly one NaN in the column
                    % Find non-NaN indices
                    non_nan_indices = find(~isnan(r_resaped(:, col)));
                    % XOR the other two elements
                    r_resaped(nan_indices, col) =
mod(sum(r_resaped(non_nan_indices, col)), 2);
                end
            end
            % Row decoding
            for row = 1:3
                nan_indices = find(isnan(r_resaped(row, :)));
                if length(nan_indices) == 1 % Exactly one NaN in the row
                    % Find non-NaN indices
                    non_nan_indices = find(~isnan(r_resaped(row, :)));
                    % XOR the other two elements
                    r_resaped(row, nan_indices) =
mod(sum(r_resaped(row, non_nan_indices)), 2);
                end
            end
            % Check for remaining NaNs
            if any(isnan(r_resaped), 'all')
                err(round(p * 10) + 1) = err(round(p * 10) + 1) + 1; % Increment error
count
            end
        end
    end
end
% Convert error counts to error probabilities

```

```

err_prob = err / (nsim);
succ_prob = zeros(size(0:0.1:1)); % Initialize array to store error counts for each
p
succ_prob = 1-err_prob;
% Plotting
p_values = 0:0.1:1;
plot(p_values, succ_prob, 'r--', 'LineWidth', 2);
hold on; % Hold the plot to add more elements
% Plotting the blue dots
scatter(p_values, succ_prob, 'b', 'filled');
plot(p_values, theoretical_prob, 'g', 'LineWidth', 2);
hold off; % Release the hold
xlabel('p');
ylabel('Prob');
title('Probability of Successful Decoding for (9,4) Product Code');
ylim([0 1.2]);
grid on;
% Adding legend
legend('Success Probability (Red Dotted Line)', 'Success Probability Points (Blue Dots)', 'Theoretical Analysis');

```

