

Digital Communication Systems

ECS 452

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3 Discrete Memoryless Channel (DMC)



Office Hours:

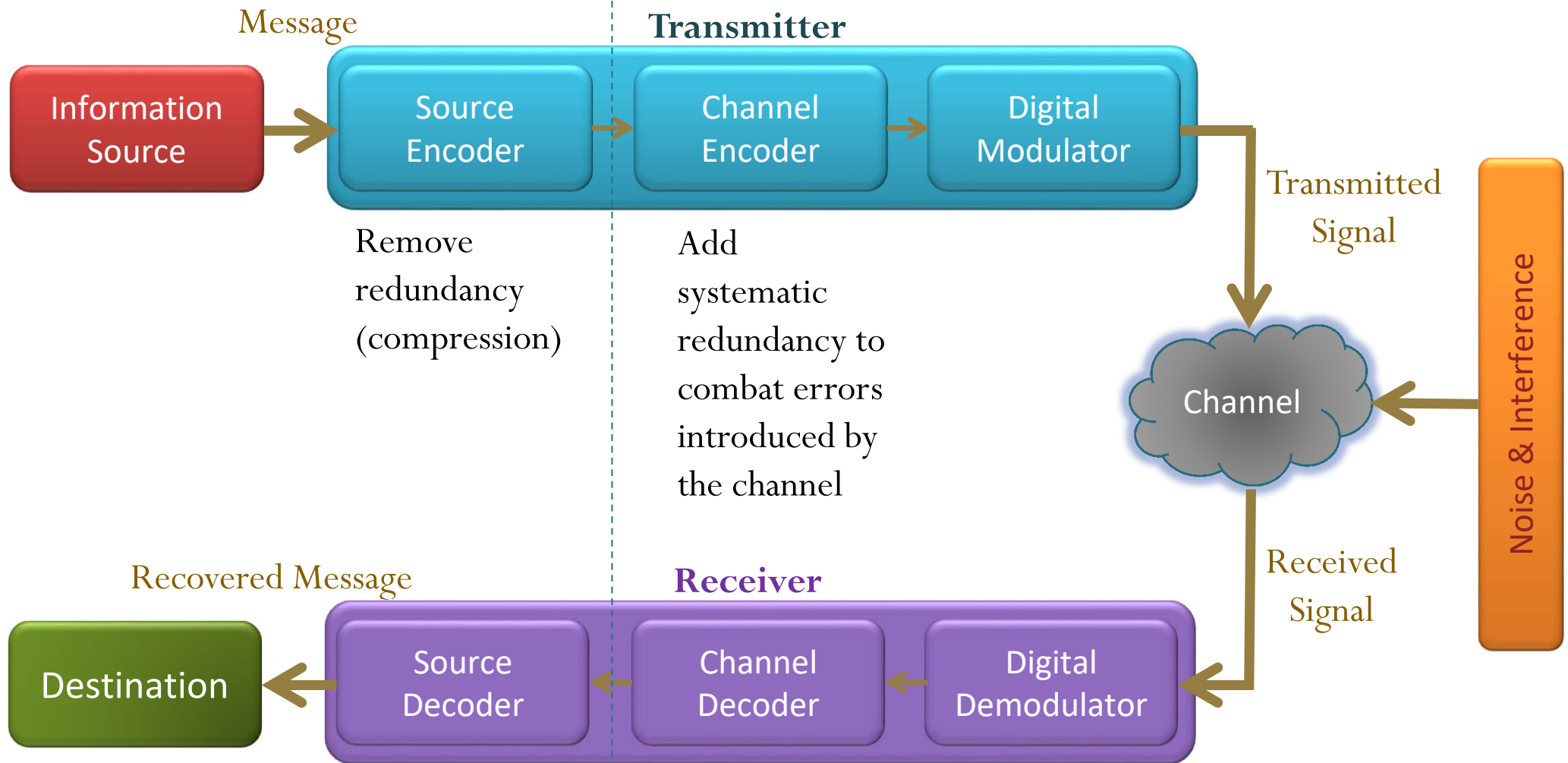
BKD, 6th floor of Sirindhralai building

Tuesday **14:20-15:20**

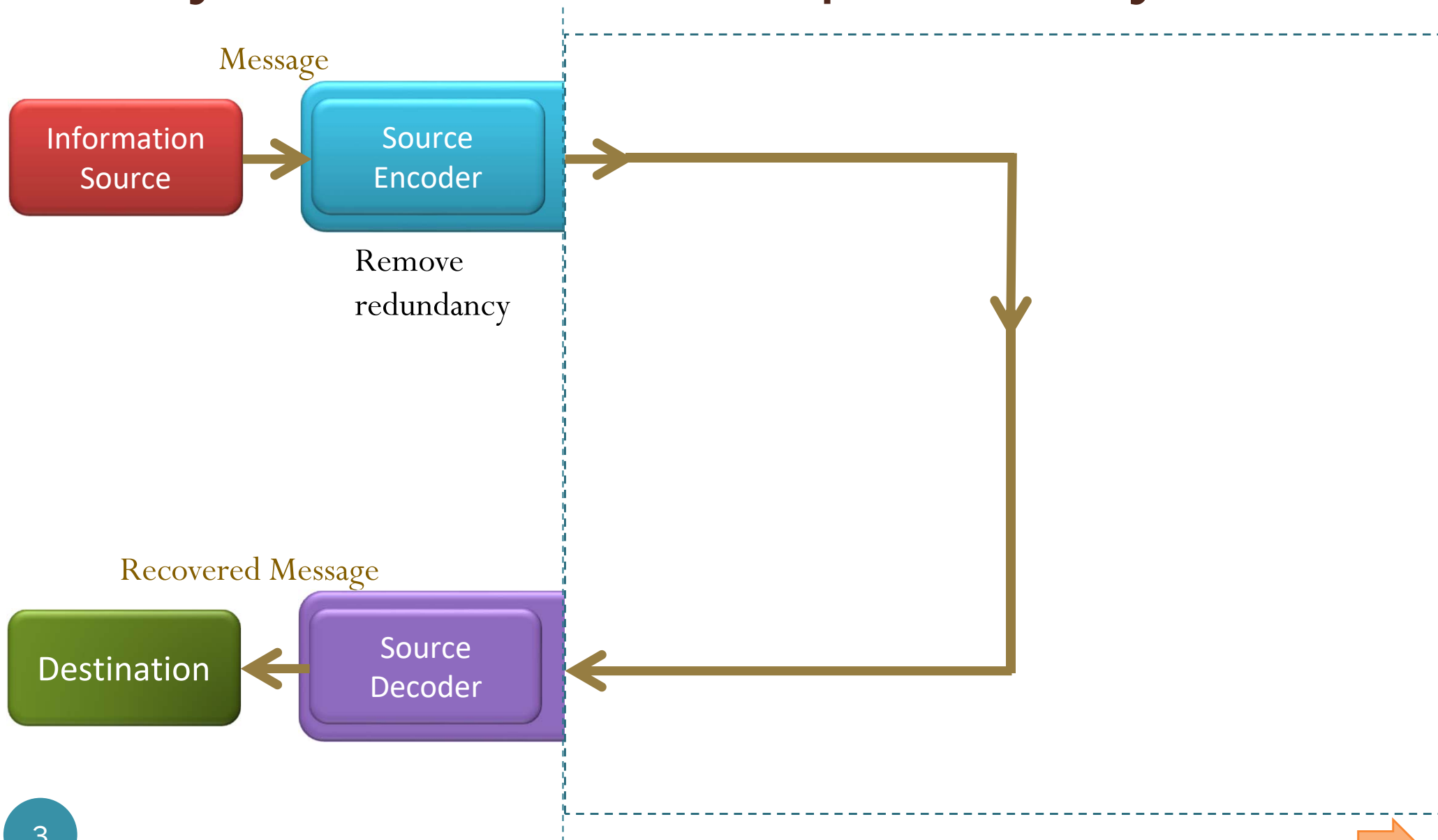
Wednesday **14:20-15:20**

Friday **9:15-10:15**

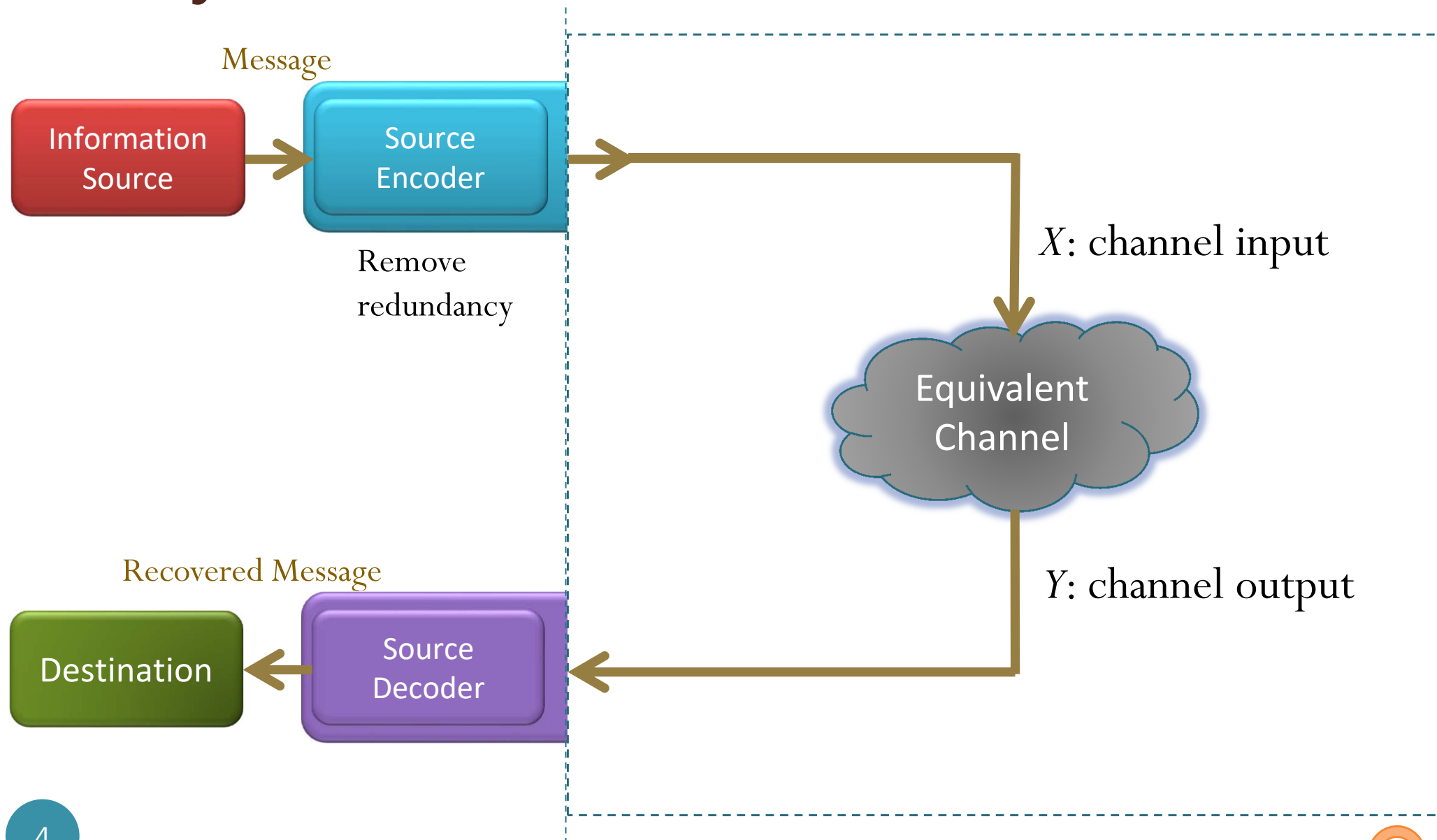
Elements of digital commu. sys.



System considered previously



System considered in this section



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3.1 DMC Models

MATLAB

```
%% Generating the channel input x  
x = randsrc(1,n,[S_X;p_X]); % channel input
```

```
%% Applying the effect of the channel to create the channel output y  
y = DMC_Channel_sim(x,S_X,S_Y,Q); % channel output
```

```
function y = DMC_Channel_sim(x,S_X,S_Y,Q)  
%% Applying the effect of the channel to create the channel output y  
y = zeros(size(x)); % preallocation  
for k = 1:length(x)  
    % Look at the channel input one by one. Choose the corresponding row  
    % from the Q matrix to generate the channel output.  
    y(k) = randsrc(1,1,[S_Y;Q(find(S_X == x(k)),:)]);  
end
```

[DMC_Channel_sim.m]

[Example 3.2]

Ex: BSC

```
>> BSC_demo
```

```
ans =
```

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

```
ans =
```

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

```
p_X =
```

```
0.3000 0.7000
```

```
Q =
```

```
0.9000 0.1000
```

```
0.1000 0.9000
```

```
q =
```

```
0.3400 0.6600
```

```
%% Simulation parameters
% The number of symbols to be transmitted
n = 20;
% Channel Input
S_X = [0 1]; S_Y = [0 1];
p_X = [0.3 0.7];
% Channel Characteristics
p = 0.1; Q = [1-p p; p 1-p];
```

Rel. freq. from the simulation

```
%% Statistical Analysis
% The probability values for the channel inputs
p_X          % Theoretical probability
p_X_sim = hist(x,S_X)/n % Relative frequencies from the simulation
% The probability values for the channel outputs
q = p_X*Q    % Theoretical probability
q_sim = hist(y,S_Y)/n % Relative frequencies from the simulation
% The channel transition probabilities from the simulation
Q_sim = [];
for k = 1:length(S_X)
    I = find(x==S_X(k)); LI = length(I);
    rel_freq_Xk = LI/n;
    yc = y(I);
    cond_rel_freq = hist(yc,S_Y)/LI; Q_sim = [Q_sim; cond_rel_freq];
end
Q          % Theoretical probability
Q_sim     % Relative frequencies from the simulation
```


[Example 3.2]

Ex: BSC

>> BSC_demo

ans =

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

ans =

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

p_X =

0.3000 0.7000

p_X_sim =

0.1500 0.8500

q =

0.3400 0.6600

q_sim =

0.1500 0.8500

Q =

0.9000 0.1000

0.1000 0.9000

Q_sim =

0.6667 0.3333

0.0588 0.9412

```
%% Simulation parameters
% The number of symbols to be transmitted
n = 20;
% Channel Input
S_X = [0 1]; S_Y = [0 1];
p_X = [0.3 0.7];
% Channel Characteristics
p = 0.1; Q = [1-p p; p 1-p];
```

Because there are only 20 samples, we can't expect the relative freq. from the simulation to match the specified or calculated probabilities.



Ex: BSC


```
>> BSC_demo
```

```
p_X =  
    0.3000    0.7000
```

```
p_X_sim =  
    0.3037    0.6963
```

```
q =  
    0.3400    0.6600
```

```
q_sim =  
    0.3407    0.6593
```



```
%% Simulation parameters  
% The number of symbols to be transmitted  
n = 1e4;  
% Channel Input  
S_X = [0 1]; S_Y = [0 1];  
p_X = [0.3 0.7];  
% Channel Characteristics  
p = 0.1; Q = [1-p p; p 1-p];
```

```
Q =  
    0.9000    0.1000
```

```
Q_sim =  
    0.1000    0.9000
```

```
    0.9078    0.0922  
    0.0934    0.9066
```

Elapsed time is 0.922728 seconds.

Ex: DMC

```

%% Simulation parameters
% The number of symbols to be transmitted
n = 20;
% General DMC
% Ex. 3.6 amd 3.12 in lecture note
% Channel Input
S_X = [0 1]; S_Y = [1 2 3];
p_X = [0.2 0.8];
% Channel Characteristics
Q = [0.5 0.2 0.3; 0.3 0.4 0.3];

```

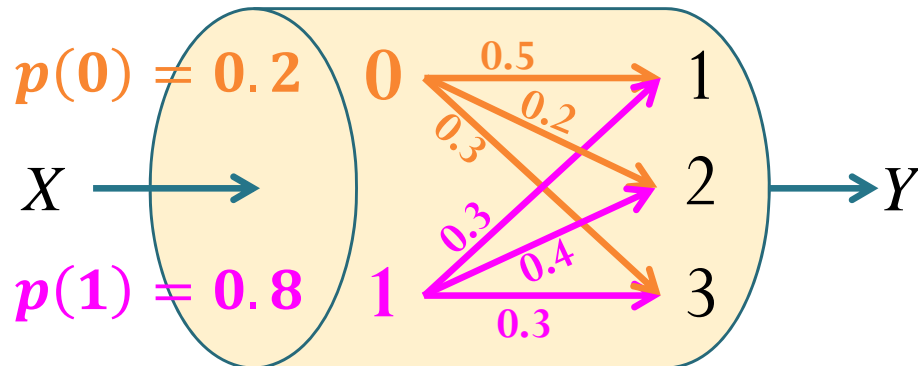
>> DMC_demo

ans =

x: 1 1 1 1 1 1 1 1 1 1 0 0 1 1 1 0 1 1 0 1

ans =

y: 1 3 2 2 1 2 1 2 2 3 1 1 1 3 1 3 2 3 1 2



$p_X =$

0.2000 0.8000

$p_{X_sim} =$

0.2000 0.8000

$q =$

0.3400 0.3600 0.3000

$q_sim =$

0.4000 0.3500 0.2500

$Q =$

0.5000 0.2000 0.3000

0.3000 0.4000 0.3000

$Q_sim =$

0.7500 0 0.2500

0.3125 0.4375 0.2500

>> sym(Q_sim)

ans =

[3/4, 0, 1/4]

[5/16, 7/16, 1/4]



Ex: DMC

```
>> p = [0.2 0.8]
```

```
p =
```

```
    0.2000    0.8000
```

```
>> p = [0.2 0.8];
```

```
>> Q = [0.75 0 0.25; 0.3125 0.4375 0.25];
```

```
>> p*Q
```

```
ans =
```

```
    0.4000    0.3500    0.2500
```



Block Matrix Multiplications

$$\begin{pmatrix} 10 & 6 \\ 9 & 7 \end{pmatrix} \begin{matrix} \text{A} \\ \text{B} \end{matrix} \times \begin{pmatrix} \begin{matrix} 2 & 5 \\ 3 & 4 \end{matrix} \text{C} & \begin{matrix} 10 & 2 \\ 5 & 10 \end{matrix} \text{D} \\ \begin{matrix} 3 & 4 \\ 7 & 5 \\ 8 & 6 \end{matrix} \text{E} & \begin{matrix} 1 & 5 \\ 3 & 6 \\ 9 & 3 \end{matrix} \text{F} \end{pmatrix}$$

$$= \begin{pmatrix} 108 & 73 & 136 & 175 & 150 & 193 & 126 & 149 \\ 155 & 85 & 164 & 224 & 213 & 197 & 158 & 165 \end{pmatrix}$$

$AC+BE$
 $AD+BF$

$$\begin{pmatrix} 10 & 6 \\ 9 & 7 \end{pmatrix} \text{X} \begin{pmatrix} 6 & 4 & 3 \\ 3 & 5 & 9 \end{pmatrix} \times \begin{pmatrix} \begin{matrix} 2 & 5 & 10 \\ 3 & 4 & 1 \\ 7 & 3 & 9 \end{matrix} \text{G} & \begin{matrix} 2 & 2 & 5 \\ 10 & 5 & 3 \\ 1 & 5 & 5 \\ 10 & 6 & 10 \\ 8 & 3 & 6 \end{matrix} \text{H} \end{pmatrix}$$

$$= \begin{pmatrix} 108 & 73 & 136 & 175 & 150 & 193 & 126 & 149 \\ 155 & 85 & 164 & 224 & 213 & 197 & 158 & 165 \end{pmatrix}$$

XG
 XH



Review: Evaluation of Probability from the Joint PMF Matrix

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|ccccc} & \begin{array}{c} y \\ \hline 2 \quad 3 \quad 4 \quad 5 \quad 6 \end{array} \\ \begin{array}{c} x \\ \hline 1 \\ 3 \\ 4 \\ 6 \end{array} & \left[\begin{array}{ccccc} 0.1 & 0.1 & 0 & 0 & 0 \\ 0.1 & 0 & 0 & 0.1 & 0 \\ 0 & 0.1 & 0.2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.3 \end{array} \right] \end{array}$$

- Find $P[X + Y < 7]$

Step 1: Find the pairs (x,y) that satisfy the condition “ $x+y < 7$ ”

One way to do this is to first construct the matrix of $x+y$.

$$x + y = \begin{array}{c|ccccc} & \begin{array}{c} y \\ \hline 2 \quad 3 \quad 4 \quad 5 \quad 6 \end{array} \\ \begin{array}{c} x \\ \hline 1 \\ 3 \\ 4 \\ 6 \end{array} & \left[\begin{array}{ccccc} 3 & 4 & 5 & 6 & 7 \\ 5 & 6 & 7 & 8 & 9 \\ 6 & 7 & 8 & 9 & 10 \\ 8 & 9 & 10 & 11 & 12 \end{array} \right] \end{array}$$


Review: Evaluation of Probability from the Joint PMF Matrix

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|ccccc} & y & 2 & 3 & 4 & 5 & 6 \\ \hline x & 1 & 0.1 & 0.1 & 0 & 0 & 0 \\ & 3 & 0.1 & 0 & 0 & 0.1 & 0 \\ & 4 & 0 & 0.1 & 0.2 & 0 & 0 \\ & 6 & 0 & 0 & 0 & 0 & 0.3 \end{array}$$

- Find $P[X + Y < 7]$

Step 2: Add the corresponding probabilities from the joint pmf (matrix)

$$P[X + Y < 7] = 0.1 + 0.1 + 0.1 = 0.3$$

$$x + y = \begin{array}{c|ccccc} & y & 2 & 3 & 4 & 5 & 6 \\ \hline x & 1 & 3 & 4 & 5 & 6 & 7 \\ & 3 & 5 & 6 & 7 & 8 & 9 \\ & 4 & 6 & 7 & 8 & 9 & 10 \\ & 6 & 8 & 9 & 10 & 11 & 12 \end{array}$$


Review: Evaluation of Probability from the Joint PMF Matrix

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|cccccc} & \begin{array}{c} y \\ \hline 2 \quad 3 \quad 4 \quad 5 \quad 6 \end{array} \\ \begin{array}{c} x \\ \hline 1 \\ 3 \\ 4 \\ 6 \end{array} & \begin{bmatrix} 0.1 & 0.1 & 0 & 0 & 0 \\ 0.1 & 0 & 0 & 0.1 & 0 \\ 0 & 0.1 & 0.2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.3 \end{bmatrix} \end{array}$$

- Find $P[X = Y]$

$$P[X = Y] = 0 + 0.2 + 0.3 = 0.5$$



Review: Sum of two discrete RVs

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|ccccc} & y & 2 & 3 & 4 & 5 & 6 \\ \hline x & 1 & 0.1 & 0.1 & 0 & 0 & 0 \\ & 3 & 0.1 & 0 & 0 & 0.1 & 0 \\ & 4 & 0 & 0.1 & 0.2 & 0 & 0 \\ & 6 & 0 & 0 & 0 & 0 & 0.3 \end{array}$$

- Find $P[X + Y = 7]$

$$P[X + Y = 7] = 0.1$$

$$x + y = \begin{array}{c|ccccc} & y & 2 & 3 & 4 & 5 & 6 \\ \hline x & 1 & 3 & 4 & 5 & 6 & 7 \\ & 3 & 5 & 6 & 7 & 8 & 9 \\ & 4 & 6 & 7 & 8 & 9 & 10 \\ & 6 & 8 & 9 & 10 & 11 & 12 \end{array}$$


Ex: DMC

```
>> p = [0.2 0.8];  
>> Q = [0.5 0.2 0.3; 0.3 0.4 0.3];  
>> p*Q  
ans =  
    0.3400    0.3600    0.3000  
>> P = (diag(p))*Q  
P =  
    0.1000    0.0400    0.0600  
    0.2400    0.3200    0.2400  
>> sum(P)  
ans =  
    0.3400    0.3600    0.3000
```



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3.2 Decoder and $P(\mathcal{E})$

$P(\mathcal{E})$ for Naïve Decoder

```
%% Naive Decoder
```

```
x_hat = y;
```

```
%% Error Probability
```

```
PE_sim = 1-sum(x==x_hat)/n % Error probability from the simulation
```

```
% Calculation of the theoretical error probability
```

```
PC = 0;
```

```
for k = 1:length(S_X)
```

```
    t = S_X(k);
```

```
    i = find(S_Y == t);
```

```
    if length(i) == 1
```

```
        PC = PC+ p_X(k)*Q(k,i);
```

```
    end
```

```
end
```

```
PE_theretical = 1-PC
```

} Formula derived in 3.19 of lecture notes

Ex: Naïve Decoder and BAC [Ex. 3.18]

```

%% Simulation parameters
% The number of symbols to be transmitted
n = 20;
% Binary Assymmetric Channel (BAC)
% Ex 3.8 in lecture note (11.3 in [Z&T, 2010])
% Channel Input
S_X = [0 1]; S_Y = [0 1];
p_X = [0.5 0.5];
% Channel Characteristics
Q = [0.7 0.3; 0.4 0.6];

```

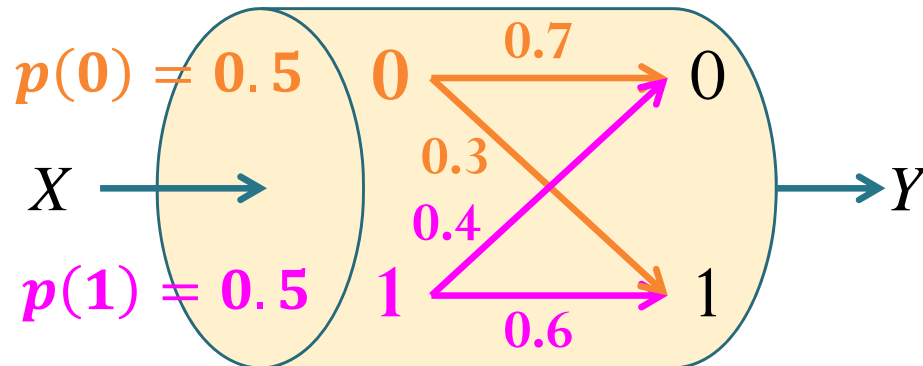
>> BAC_demo

ans =

x: 0 0 0 1 1 0 0 1 0 0 0 0 1 0 0 1 0 1 0 0

ans =

y: 0 0 1 1 0 0 0 1 1 1 0 0 1 0 0 0 0 0 1 0



p_X =
0.5000 0.5000

p_X_sim =
0.7000 0.3000

q =
0.5500 0.4500

q_sim =
0.6500 0.3500

Q =
0.7000 0.3000

0.4000 0.6000

Q_sim =
0.7143 0.2857

0.5000 0.5000

$\frac{7}{20}$ → PE_sim =
0.3500

PE_theretical =
0.3500

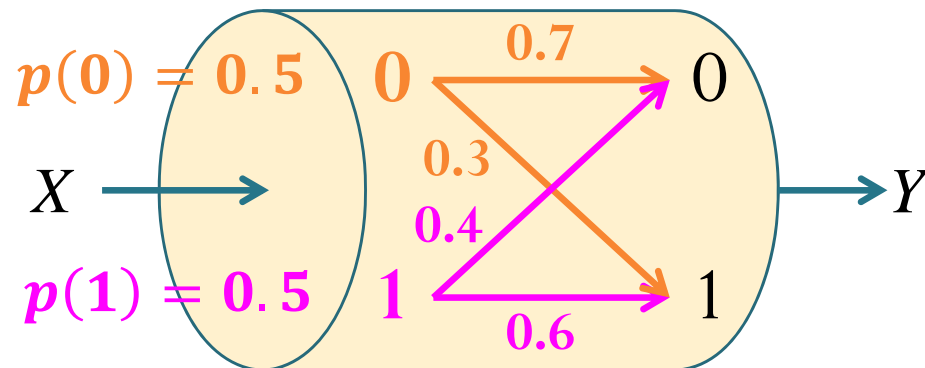
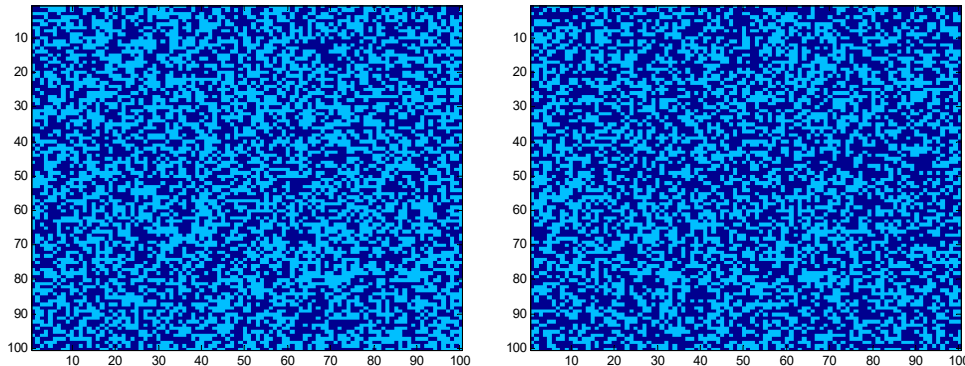
[BAC_demo.m] →

Ex: Naïve Decoder and BAC [Ex. 3.18]

```

%% Simulation parameters
% The number of symbols to be transmitted
n = 1e4;
% Binary Assymmetric Channel (BAC)
% Ex 3.8 in lecture note (11.3 in [Z&T, 2010])
% Channel Input
S_X = [0 1]; S_Y = [0 1];
p_X = [0.5 0.5];
% Channel Characteristics
Q = [0.7 0.3; 0.4 0.6];

```



p_X =
0.5000 0.5000

p_X_sim =
0.5043 0.4957

q =
0.5500 0.4500


q_sim =
0.5532 0.4468

Q =
0.7000 0.3000
0.4000 0.6000

Q_sim =
0.7109 0.2891
0.3928 0.6072

PE_sim =
0.3405

PE_theretical =
0.3500

[BAC_demo.m] 

Ex: Naïve Decoder and DMC [Ex. 3.21]

```

%% Simulation parameters
% The number of symbols to be transmitted
n = 20;
% General DMC
% Ex. 3.16 in lecture note
% Channel Input
S_X = [0 1]; S_Y = [1 2 3];
p_X = [0.2 0.8];
% Channel Characteristics
Q = [0.5 0.2 0.3; 0.3 0.4 0.3];

```

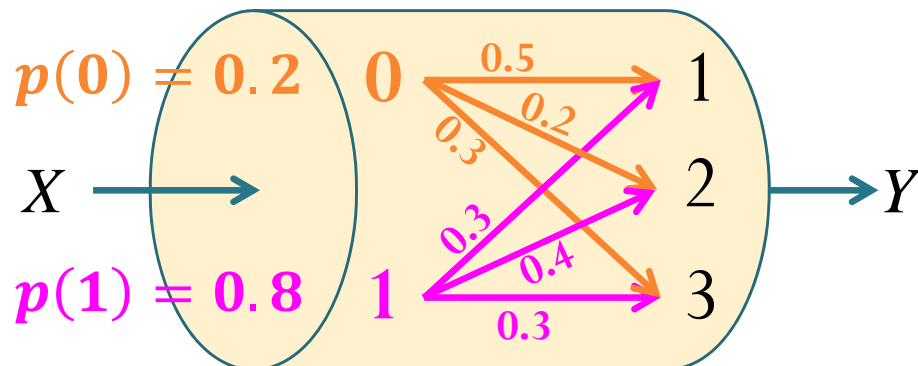
>> DMC_demo [Same samples as in Ex. 3.6]

ans =

x: 1 1 1 1 1 1 1 1 1 1 0 0 1 1 1 0 1 1 0 1

ans =

y: 1 3 2 2 1 2 1 2 2 3 1 1 1 3 1 3 2 3 1 2



$$\frac{20 - 4}{20}$$

PE_sim =

0.7500

PE_theretical =

0.7600

p_X =

0.2000 0.8000

p_X_sim =

0.2000 0.8000

q =

0.3400 0.3600 0.3000

q_sim =

0.4000 0.3500 0.2500

Q =

0.5000 0.2000 0.3000

0.3000 0.4000 0.3000

Q_sim =

0.7500 0 0.2500

0.3125 0.4375 0.2500

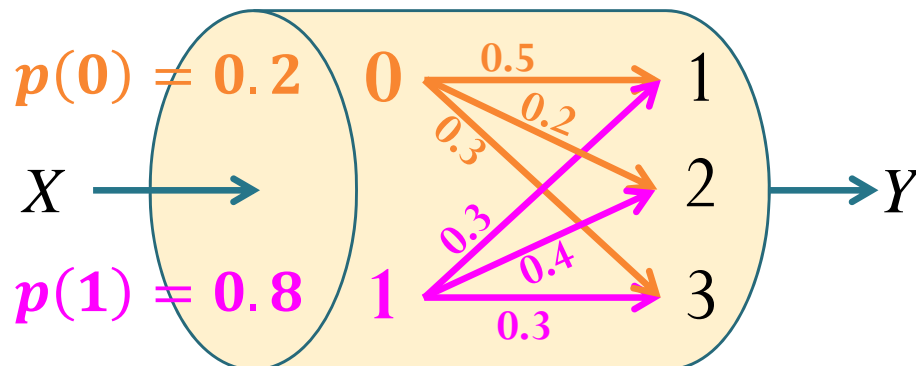
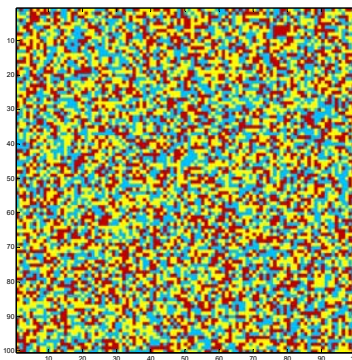
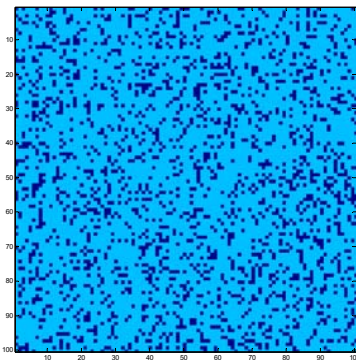


Ex: Naïve Decoder and DMC [Ex. 3.21]

```

%% Simulation parameters
% The number of symbols to be transmitted
n = 1e4;
% General DMC
% Ex. 3.16 in lecture note
% Channel Input
S_X = [0 1]; S_Y = [1 2 3];
p_X = [0.2 0.8];
% Channel Characteristics
Q = [0.5 0.2 0.3; 0.3 0.4 0.3];

```



```

p_X =
    0.2000    0.8000
p_X_sim =
    0.2011    0.7989
q =
    0.3400    0.3600    0.3000
q_sim =
    0.3387    0.3607    0.3006
Q =
    0.5000    0.2000    0.3000
    0.3000    0.4000    0.3000
Q_sim =
    0.4943    0.1914    0.3143
    0.2995    0.4033    0.2972

```

PE_sim =
0.7607

PE_theretical =
0.7600



DIY Decoder [Ex. 3.22]

```
>> DMC_decoder_DIY_demo
ans =
X 1 0 1 1 1 1 1 0 1 1 0 1 1 1 1 0 0 1 0 1
ans =
Y 2 1 1 3 3 1 2 2 1 2 1 2 3 1 1 3 1 3 1 1
ans =
 $\hat{X}$  1 0 0 0 0 0 1 1 0 1 0 1 0 0 0 0 0 0 0 0
PE_sim =
    0.5500
PE_theretical =
    0.5200
Elapsed time is 0.081161 seconds.
```

```
%% Simulation parameters
% The number of symbols to be transmitted
n = 20;
% General DMC
% Ex. 3.16 in lecture note
% Channel Input
S_X = [0 1]; S_Y = [1 2 3];
p_X = [0.2 0.8];
% Channel Characteristics
Q = [0.5 0.2 0.3; 0.3 0.4 0.3];
```

```
%% DIY Decoder
Decoder_Table = [0 1 0]; % The decoded
values corresponding to the received Y
```

DIY Decoder [Ex. 3.22]

```
%% DIY Decoder  
Decoder_Table = [0 1 0]; % The decoded values corresponding to the received Y
```

```
% Decode according to the decoder table  
x_hat = y; % preallocation  
for k = 1:length(S_Y)  
    I = (y==S_Y(k));  
    x_hat(I) = Decoder_Table(k);  
end  
  
PE_sim = 1-sum(x==x_hat)/n % Error probability from the simulation
```

```
% Calculation of the theoretical error probability  
PC = 0;  
for k = 1:length(S_X)  
    I = (Decoder_Table == S_X(k));  
    q = Q(k,:);  
    PC = PC+ p_X(k)*sum(q(I));  
end  
PE_theoretical = 1-PC
```

DIY Decoder [Ex. 3.22]

```
>> DMC_decoder_DIY_demo
```

```
PE_sim =
```

```
0.5213
```

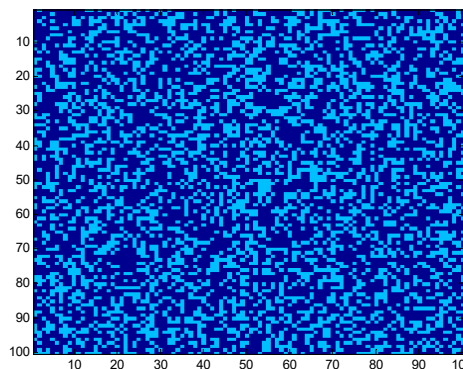
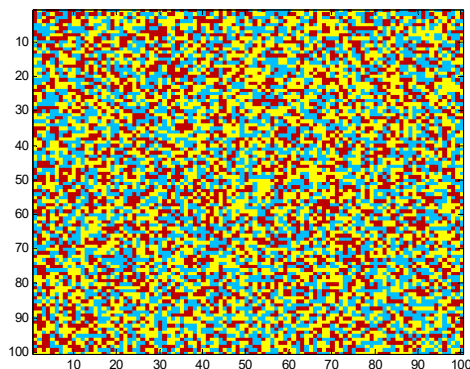
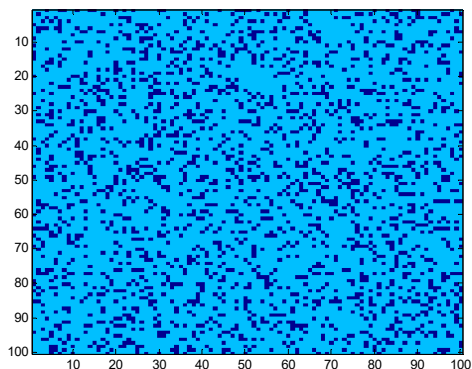
```
PE_theretical =
```

```
0.5200
```

```
Elapsed time is 2.154024 seconds.
```

```
%% Simulation parameters
% The number of symbols to be transmitted
n = 1e4;
% General DMC
% Ex. 3.16 in lecture note
% Channel Input
S_X = [0 1]; S_Y = [1 2 3];
p_X = [0.2 0.8];
% Channel Characteristics
Q = [0.5 0.2 0.3; 0.3 0.4 0.3];
```

```
%% DIY Decoder
Decoder_Table = [0 1 0]; % The decoded
values corresponding to the received Y
```



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3.3 Optimal Decoder

Searching for the Optimal Detector

```
>> DMC_decoder_ALL_demo
```

```
ans =    $\hat{x}(1)$         $\hat{x}(2)$         $\hat{x}(3)$         $P(\mathcal{E})$ 
        0           0           0           0.8000
        0           0           1.0000       0.6200
        0           1.0000          0           0.5200
        0           1.0000       1.0000       0.3400
        1.0000          0           0           0.6600
        1.0000          0           1.0000       0.4800
        1.0000       1.0000          0           0.3800
        1.0000       1.0000       1.0000       0.2000
```

Ex. 3.22

Ex. 3.23

```
Min_PE =
```

```
0.2000
```

```
Optimal_Detector =
```

```
1 1 1
```

```
Elapsed time is 0.003351 seconds.
```



Review: ECS315 (2016)

6.4. Interpretation: It is sometimes useful to interpret $P(A)$ as our knowledge of the occurrence of event A *before* the experiment takes place. Conditional probability²⁴ $P(A|B)$ is the **updated probability** of the event A given that we now know that B occurred (but we still do not know which particular outcome in the set B did occur).

Definition 6.5. Sometimes, we refer to $P(A)$ as

- a priori probability, or
- the prior probability of A , or
- the unconditional probability of A .


in which case, we refer to $P(A|B)$ as
a posteriori probability
the posterior probability
conditional probability



Guessing Game 1

- There are 15 cards.
 - Each have a number on it.
 - Here are the 15 cards:

1 2 2 3 3 3 4 4 4 4 5 5 5 5 5

- One card is randomly selected from the 15 cards.
- You need to guess the number on the card.
- Have to pay 1 Baht for incorrect guess. 
- The game is to be repeated $n = 10,000$ times.
- What should be your guess value?



```
close all; clear all;

n = 5; % number of time to play this game

D = [1 2 2 3 3 3 4 4 4 4 5 5 5 5 5];
X = D(randi(length(D),1,n));

if n <= 10
    X
end

g = 1
cost = sum(X ~= g)

if n > 1
    averageCostPerGame = cost/n
end
```

```
>> GuessingGame_4_1_1
X =
     3     5     1     2     5
g =
     1
cost =
     4
averageCostPerGame =
     0.8000
```




```
close all; clear all;

n = 5; % number of time to play this game

D = [1 2 2 3 3 3 4 4 4 4 5 5 5 5 5];
X = D(randi(length(D),1,n));

if n <= 10
    X
end

g = 3.3
cost = sum(X ~= g)

if n > 1
    averageCostPerGame = cost/n
end
```

```
>> GuessingGame_4_1_1
X =
     5     3     2     4     1
g =
     3.3000
cost =
     5
averageCostPerGame =
     1
```



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

```
n = 1e4; % number of time to play this game
```

```
D = [1 2 2 3 3 3 4 4 4 4 5 5 5 5 5];
```

```
X = D(randi(length(D),1,n));
```

```
if n <= 10
```

```
    X
```

```
end
```

```
g = ?
```

```
cost = sum(X ~= g)
```

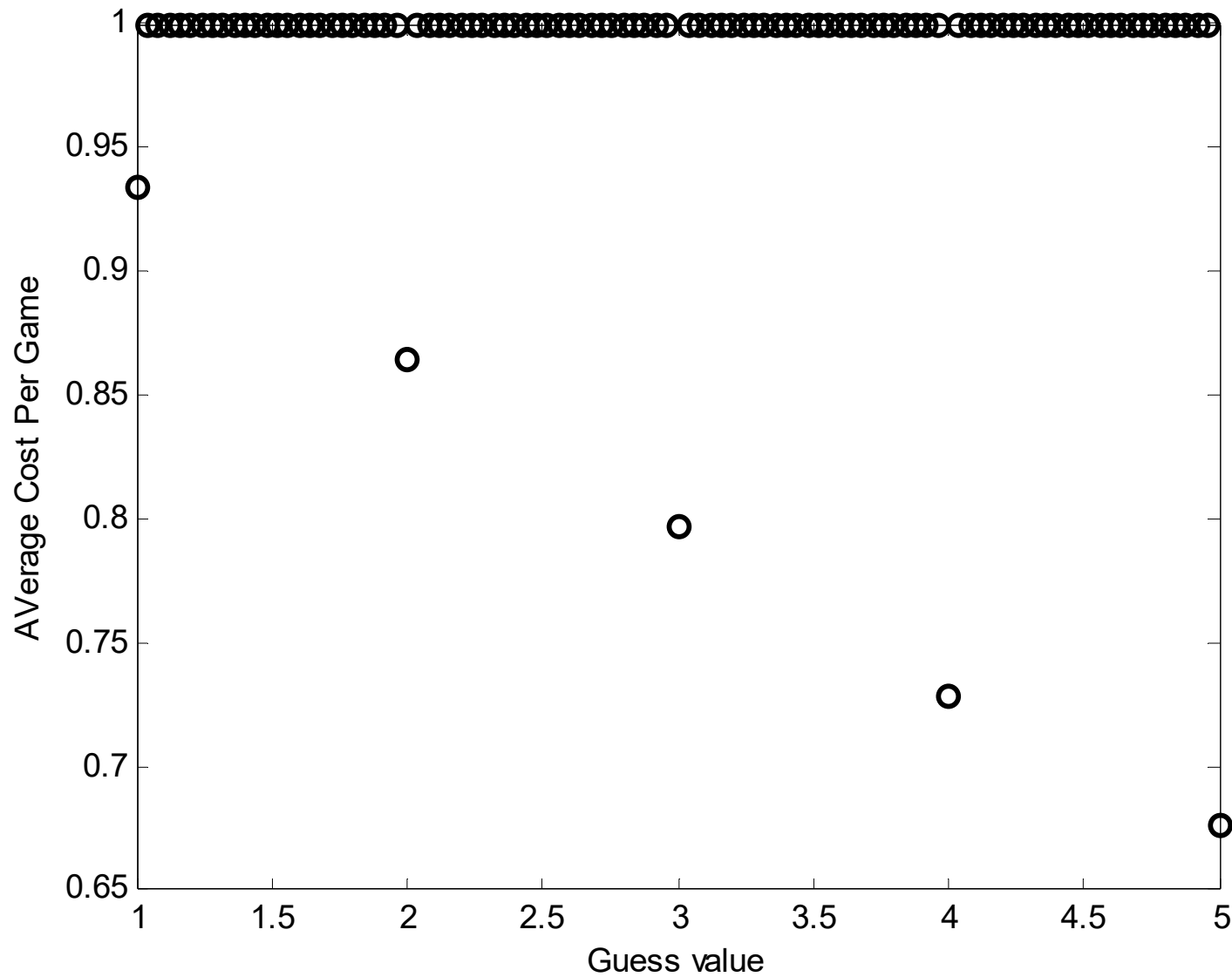
```
if n > 1
```

```
    averageCostPerGame = cost/n
```

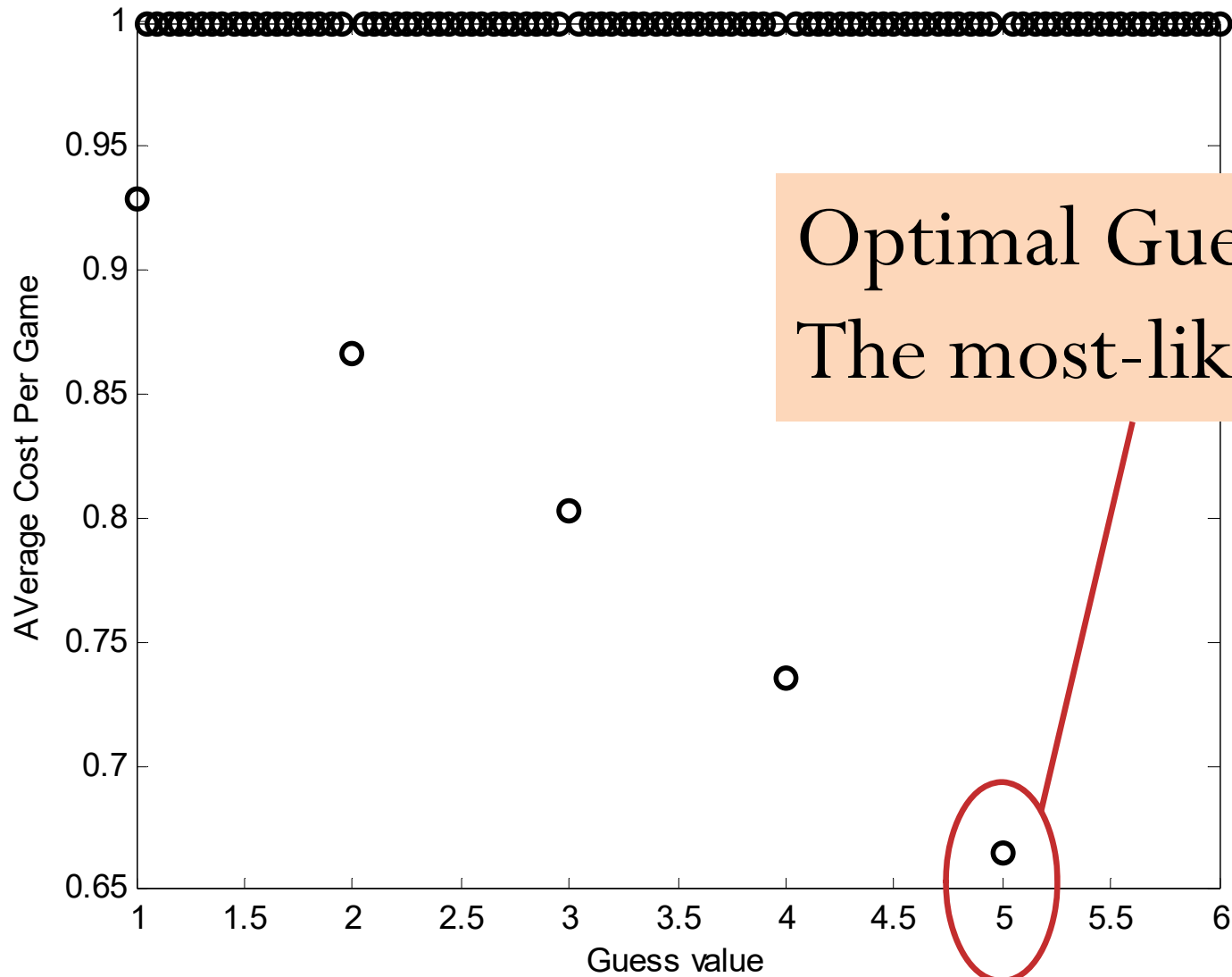
```
end
```



Guessing Game 1



Guessing Game 1



Optimal Guess:
The most-likely value



MAP Decoder

```
%% MAP Decoder
P = diag(p_X)*Q; % Weight the channel transition probability by the
                % corresponding prior probability.
[V I] = max(P); % For I, the default MATLAB behavior is that when there are
                % multiple max, the index of the first one is returned.
Decoder_Table = S_X(I) % The decoded values corresponding to the received Y
```

```
%% Decode according to the decoder table
x_hat = y; % preallocation
for k = 1:length(S_Y)
    I = (y==S_Y(k));
    x_hat(I) = Decoder_Table(k);
end

PE_sim = 1-sum(x==x_hat)/n % Error probability from the simulation
```

```
%% Calculation of the theoretical error probability
PC = 0;
for k = 1:length(S_X)
    I = (Decoder_Table == S_X(k));
    Q_row = Q(k,:);
    PC = PC+ p_X(k)*sum(Q_row(I));
end
PE_theoretical = 1-PC
```



ML Decoder

```
%% ML Decoder
[V I] = max(Q); % For I, the default MATLAB behavior is that when there are
                % multiple max, the index of the first one is returned.
Decoder_Table = S_X(I) % The decoded values corresponding to the received Y
```

```
%% Decode according to the decoder table
x_hat = y; % preallocation
for k = 1:length(S_Y)
    I = (y==S_Y(k));
    x_hat(I) = Decoder_Table(k);
end

PE_sim = 1-sum(x==x_hat)/n % Error probability from the simulation
```

```
%% Calculation of the theoretical error probability
PC = 0;
for k = 1:length(S_X)
    I = (Decoder_Table == S_X(k));
    Q_row = Q(k,:);
    PC = PC+ p_X(k)*sum(Q_row(I));
end
PE_theretical = 1-PC
```