

# S-72.333 Postgraduate course in radio communications

Channel Capacity, MIMO capacity

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#### **Outline**

- Introduction
- Definitions of Channel and Channel Capacity
- Capacity of Discrete Memoryless Channel
- Capacity of Gaussian Channel
- MIMO capacity
- Summary and conclusions



#### Introduction

- Channel capacity is a fundamental limit on the rate of error free messages that can take place over a channel
- Advantage of Multiple-Input Multiple-Output (MIMO) systems is increased capacity

## Definitions (Shannon)



- Channel is defined by conditional distribution of channel output given input
  - Channel input and output are random variables where distribution of the input determines distribution of the output
  - Mutual information, I(X;Y), can be considered as amount of the information conveyed through channel. It depends on the input distribution.
- Channel capacity is maximum mutual information of input and output

$$C = \max_{p (x)} I(X;Y)$$

## Example: Binary Symmetric Channel

In Binary Symmetric Channel (BSC)

output = input + error mod 2

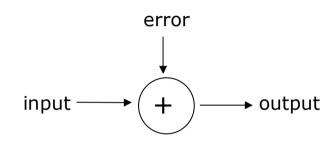
where input, output and error are

binary random variables taking

values 0 or 1.

Bit error probability is

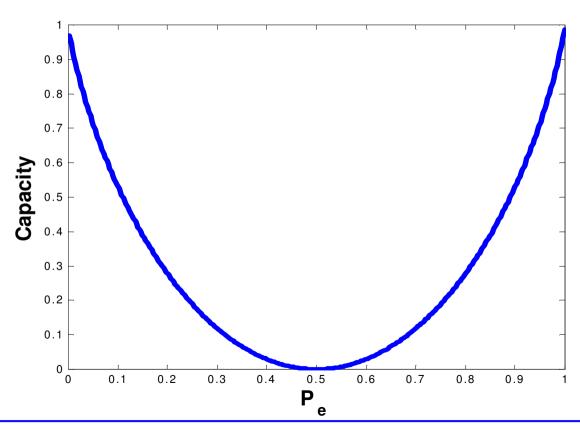
$$P_e = P(error = 1)$$



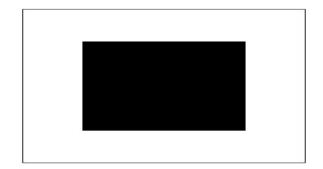
$$P(output \mid input) = \begin{bmatrix} 1 - p_e & p_e \\ p_e & 1 - p_e \end{bmatrix}$$

## Example: Binary Symmetric Channel

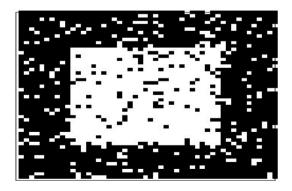
$$C_{BSC} = 1 + p_e \log_2 p_e + (1 - p_e) \log_2 (1 - p_e)$$



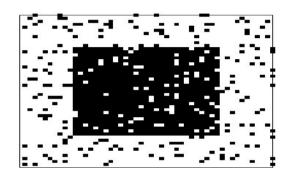
## Example: Binary Symmetric Channel



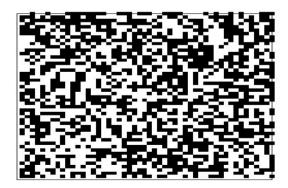
Pe = 0, C = 1



$$Pe = 0.9, C = 0.5$$

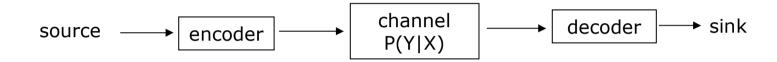


$$Pe = 0.1, C = 0.5$$



$$Pe = 0.5, C = 0$$

## **Channel Coding Theorem**



- Channel coding theorem (Shannon 1948) says that reliable information rates up to channel capacity are possible.
- Channel coding theorem do not say how to achieve these reliable rates. Efforts on designing efficient channel codes has been going on ever since.
- Theorem do not promise non-erroneous transmissions when application has delay constraints. Result of the theorem is asymptotic: non-erroneus transmissions require unlimited block lengths.

## Capacity of Gaussian Channel

Gaussian channel is time discrete continuous alphabet channel.
 Output is sum of the input and a Gaussian noise, where input and noise are independent.

$$Y = X + Z$$
  $Z \sim \mathbf{N}(0, N)$ 

 Since input and noise are independent, mutual information of output and input is differential entropy of output minus diffrerential entropy of noise.

$$I(X;Y) = h(Y) - h(Y \mid X)$$
$$= h(Y) - h(Z \mid X)$$
$$= h(Y) - h(Z)$$



## Capacity of Gaussian Channel

- Without restrictions on input, capacity would be infinity.
- With constraint E (X²) ≤ P
  mutual information is maximized when input is
  Gaussian with variance P. Then output is a sum of
  two independent Gaussians and so it too is
  Gaussian, with variance N+P.
- Differential entropy of Gaussian random variable with varianse  $\sigma^2$  is  $\frac{1}{2}\log_2(2\pi e\sigma^2)$  , thus

$$C_G = \frac{1}{2}\log_2(2\pi e(N+P)) - \frac{1}{2}\log_2(2\pi eN)$$
$$= \frac{1}{2}\log_2(1+\frac{P}{N})$$

## Capacity of bandlimited Gaussian Channel

Bandlimited Gaussian channel is a continuous time channel

$$Y = (X + Z) * h_w$$

where where Y is output, X is input, Z is white noise and  $h_{W}$  is impulse response of perfect bandpass filter.

 Output can be recconstructed from samples taken at 1/2W interval. Capacity of a sample is achieved when input is Gaussian

$$C_{sample} = \frac{1}{2} \log_2 \left(1 + \frac{P}{WN_0}\right)$$
 bits/sample

 2W samples are taken per second, so channel capacity per second is is 2W times sample capacity

$$C = 2WC_{sample} = W \log_2 (1 + \frac{P}{2WN_0})$$
 bits/second

## Capacity of RF channel

In general we can approximate capacity of the RF channel with equation

$$C = \log_2(1 + \rho h^2) \text{ bps/Hz}$$

where *p* is *SNR* and *h* is channel impulse responce.



#### MIMO channel

General formula for MIMO channel is

$$Y = H S + n$$

- H is NXM matrix where M (N) is the number of input (output) antennas. Elements  $h_{ij}$  of H repsesent channel impulse responces between ith input and jth ouptut antenna
- S is Mx1 random variable representing input antennas
- Y is Nx1 random variable representing output antennas

$$H(\tau,t) = \begin{bmatrix} h_{1,1}(\tau,t) & h_{1,2}(\tau,t) & \cdots & h_{1,M}(\tau,t) \\ h_{2,1}(\tau,t) & h_{2,2}(\tau,t) & \cdots & h_{2,M}(\tau,t) \\ \vdots & \vdots & \ddots & \vdots \\ h_{N,1}(\tau,t) & h_{N,2}(\tau,t) & \cdots & h_{N,M}(\tau,t) \end{bmatrix}$$

## Capacity of deterministic MIMO channel

- In deterministic MIMO channel matrix is assumed to be non-random
- n is additive temporally white complex zero-mean Gaussian noise with covariance matrix

$$E(nn^H) = N_0 I_M$$

- S is power constrained  $\sum_{i=1}^{M} E(s_i^2) \leq P$
- Mutual information of output and input is

$$I(Y;S) = h(Y) - h(Y \mid S)$$
$$= h(Y) - h(n \mid S)$$
$$= h(Y) - h(n)$$

Covariance matrix of Y is

$$E(YY^{H}) = E(HS)E(n)^{H} + E(HS(HS)^{H}) + E(n)E(HS)^{H} + E(nn^{H})$$

$$= E(HSS^{H}H^{H}) + N_{0}I_{M}$$

$$= HR_{SS}H^{H} + N_{0}I_{M}$$

where R is the covariance matrix of S

## Capacity of deterministic MIMO channel

- Mutual information of input and output is maximized when input is Gaussian => output is Gaussian.
- Differential entropy of multivariate Gaussian is  $\frac{1}{2}\log_2((2\pi e)^n |K|)$  where |K| is determinant of the covariance matrix

$$C^{D} = \log_{2}(|I_{M} + \frac{\rho}{M}HR_{SS}H^{H}|) \text{ bps/Hz}$$

• In the abcense of channel state information at the tranmsitter, it is reasonable to choose *S* to be spatially white i.e. R = I. Then we have

$$C^{D} = \log_{2}(|I + \frac{\rho}{M}HH^{H}|)$$
$$= \sum_{i=1}^{r} \log_{2}(1 + \frac{\rho}{M}\lambda_{i}) \text{ bps/Hz}$$

where r is the rank of and  $\lambda_i$  denote positive eigenvalues of  $H\!H^{H}$  .



## SISO, MISO, SIMO and MIMO capacities

 For a deterministic channel matrix without channel knowledge at the transmitter we have

$$C_{SISO}^{D} = \log_{2}(1 + \rho h^{2}) \text{ bps/Hz}$$
 $C_{MISO}^{D} = \log_{2}(1 + \frac{\rho}{M} \sum_{i=1}^{M} h_{i}^{2}) \text{ bps/Hz}$ 
 $C_{SIMO}^{D} = \log_{2}(1 + \rho \sum_{i=1}^{N} h_{i}^{2}) \text{ bps/Hz}$ 
 $C_{MIMO}^{D} = \sum_{i=1}^{r} \log_{2}(1 + \frac{\rho}{M} \lambda_{i}) \text{ bps/Hz}$ 

 SIMO capacity increases more rapidly with number of receiving antennas than MISO capacity with number of transmitting antennas. This is because MISO has no spatial array gain at the receiver.



## MIMO capacity

- At it's best MIMO capacity increase linearly with min(M,N)
- Advantageous of MIMO depends on H; the larger rank and eigenvalues HH have, the more MIMO capacity we have.



## Ergodic Capacity of Gaussian MIMO channel with Rayleigh fading

- Additive noise is Gaussian and entries of channel matrix are complex gaussian i.i.d zero mean unit variance random variables.
- Channel matrix is a random variable => mutual information of input and output is a random variable=> "channel capacity" is a random variable C.
- Definition: **ergodic capacity** eC is a mean of C, eC = E(C).
- **ergodic capacity** is a function of *E*(*HH*)
- By choosing S to be i.i.d Gaussian and noting that entries of HH are  $\chi^2_{2M}$  at the diagonal and zero-mean else, we get

$$eC = C^{D}$$

where  $C^D$  is capacity of the deterministic MIMO channel whose channel matrix has  $HH^H = E\left(\chi^2_{2M}\right)I_N$ 

## Outage Probability

- Channel coding theorem do not apply with ergodic capacity i.e. ergodic capacity do not express maximal zero-error information rate.
- Definition: Outage probability for a given rate R is probability that mutual information falls below that rate,

$$P_{out}(R) = P(C \le R)$$

Outage probability can be interpreted as PER.

#### Capacity of frequency-selective fading MIMO channels

- MIMO capacity of frequency flat channel is high
- Frequency band can be divided into subchannels so that bandwith of subchannels is less than coherence bandwith => subchannel is flat
- Assuming that transmit power is allocated uniformly across transmit antennas, mutual information is given by

$$I_{FS} = \frac{1}{K} \sum_{i=1}^{K} \log_2 |I_M| + \frac{\rho}{N} H_i H_i^H |bps/Hz|$$

and ergodic capacity is  $E(I_{FS})$ 

 Outage probability will in general be lower when frequency band is divided

## Summary and conclusions

- Main ideas about deriving channel capacities for discrete, Gaussian, RF and MIMO channels were presented.
- Capacity of RF channel can be approximated by logarithmic function of SNR
- MIMO capacity depends on channel matrix. At it's best MIMO capacity may increase linearly with number of antennas.

#### References

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#### Homeworks

- (1) Is it possible to increase channel capacity by channel coding?
- (2) Find channel capacities for deterministic multiple antenna channels when SNR = 10dB (=10) and channel matrices are

(a) 
$$H = [1]$$
 (SISO)

(b) 
$$H = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$
 (SIMO)

(c) 
$$H = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$$
 (MISO)

(d) 
$$H = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$
 (MIMO) (rank(H) = 3, all three eigenvalues are 1)