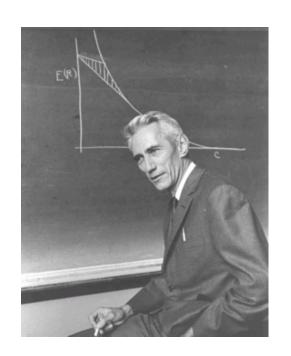
Claude E. Shannon (1916-2001),

 A Mathematical Theory of Communication, Bell System Technical Journal, Vol. 27, pp. 379–423, 623–656, 1948



# **Channel Capacity: A Review**

#### **Outline**

- Introduction
- Capacity and Achievable Rates
- Entropy and Mutual Information
- Capacity of AWGN channels
- Capacity of Frequency-Flat Fading Channels
- Capacity of Frequency-Selective Fading Channels

#### **Capacity and Achievable Rates**

Before Shannon, reliable communication over noisy channel

Repetition coding
101 111110000011111
(info) (rate approaches zero)

- Reliable communication: It is possible to communicate at a strictly positive transmission rate with an arbitrary small error probability
- Maximum rate for which this is possible is called the capacity of the channel
- Any rate below the capacity is an achievable rate
- It is impossible to drive error probability to zero for rates higher than capacity

- Consider a discrete RV with PMF  $P_X(\cdot)$  taking on values from set  $\mathcal{X}$
- Define entropy

$$H(X) = -\sum_{x \in \mathcal{X}} P_X(x) \log(P_X(x)) = -\mathbb{E}[\log(P_X(X))]$$

- Self-information of event X=x:  $I(x) = -\log(P_X(x))$
- Entropy expected value of self-info: average amount of information produced per symbol by discrete source
- Also, entropy is a measure of uncertainty of RV: number of bits on average to describe RV
- Example: Bernoulli RV with prob. p

$$H(p) = -p \log(p) - (1-p) \log(1-p)$$

Note: All logs base 2

- It can be shown that  $0 \le H(X) \le \log |\mathcal{X}|$ 
  - **Zero** when event is certain  $\exists x \in \mathcal{X} \text{ s.t. } P_X(x) = 1$
  - Max when equiprobable  $P_X(x) = 1/|\mathcal{X}|, \ \forall x \in \mathcal{X}$
- Consider now 2 RVs X and Y
- Entropy of RV Y conditioned on a given realization of X

$$H(Y|X = x_0) = -\sum_{y \in \mathcal{Y}} P_{Y|X}(y|x_0) \log(P_{Y|X}(y|x_0))$$

Entropy of RV Y conditioned on RV X: conditional entropy

$$H(Y|X) = \mathbb{E}_X[H(Y|X=x)] = -\sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} P_{X,Y}(x,y) \log(P_{Y|X}(y|x))$$

Average amount of uncertainty in Y after observing X

The reduction of uncertainty of one RV due to another is called mutual information

$$I(X,Y) = I(Y,X) = H(Y) - H(Y|X) = H(X) - H(X|Y)$$

- Consider a communication channel with input X, output Y
- Output depends probabilistically on input with transition probability
   P(Y|X), i.e., discrete memoryless channel

$$X \xrightarrow{P(Y| \rightarrow Y)} Y$$

Capacity of the DMC

$$C = \max_{P_X(\cdot)} I(X, Y)$$

#### Examples:

1.-Noiseless binary channel

$$X$$
  $Y$   $I(X,Y) = H(Y) - H(Y|X)$ 
 $0 \longrightarrow 0 = H(Y) \le 1$ 
 $1 \longrightarrow 1$   $C = 1$  achieved by equiprobable inputs

2.-Binary symmetric channel

$$X$$
  $Y$   $I(X,Y) = H(Y) - H(Y|X)$   $= H(Y) - H(p)$   $\leq 1 - H(p)$   $C = 1 - H(p)$  achieved by equiprobable inputs

#### **Entropy and Mutual Information: Continuous sources**

- How about for cts RVs with PDF  $f_X(x)$ ?
- Differential entropy

$$h(X) = -\int f_X(u) \log(f_X(u)) \, du$$

• Example: Gaussian RV  $w \sim \mathcal{N}(\mu, \sigma^2)$ 

$$h(w) = [1/2] \cdot \log(2\pi e \sigma^2)$$

Conditional entropy and mutual info defined similarly

$$h(Y|X) = -\int f_{X,Y}(u,v) \log(f_{Y|X}(v,u)) dudv$$

$$I(X,Y) = h(Y) - h(Y|X)$$

# **Entropy and Mutual Information: Continuous sources**

• For cts RV with second moment constraint  $\mathbb{E}[x^2] \leq \sigma^2$ 

$$h(x) \le [1/2] \cdot \log(2\pi e \sigma^2)$$

with equality when  $x \sim \mathcal{N}(0, \sigma^2)$ 

Gaussian RVs are entropy maximizers for variance constraint

Complex AWGN channel

$$y = hx + n$$

- Complex input x, output y, noise  $n \sim \mathcal{CN}(0, N_0)$
- Deterministic, time-invariant channel  $\,h=1\,$
- Power-constrained capacity

$$C = \max_{f_X(x)} \quad I(x,y) \quad \text{s.t.} \quad \mathbb{E}[|x|^2] \leq P$$

- Mutual information between input and output I(x,y)
- Optimization over all PDFs  $f_X(x)$  satisfying power constraint

$$\int |u|^2 f_X(u) \, du \le P$$

$$I(x,y) = h(y) - h(y|x)$$
 $= h(y) - h(n)$ 
 $= h(y) - \log(\pi e N_0)$ 
 $\leq \log(\pi e (P + N_0)) - \log(\pi e N_0)$ 

Achieved with equality when y is circular symmetric Gaussian RV

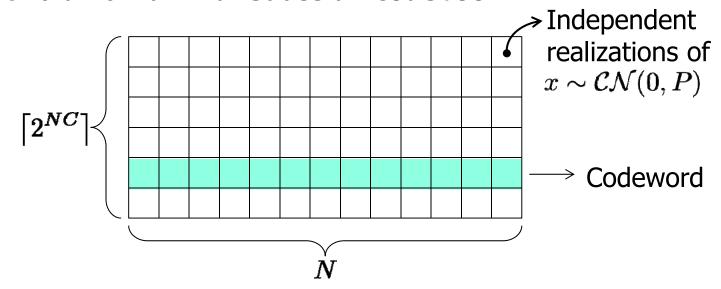
$$y \sim \mathcal{CN}(0, P + N_0)$$

- Circular symmetric Gaussian RV maximizes differential entropy for second moment constraint  $\mathbb{E}[|y|^2] \leq P + N_0$
- Therefore

$$C = \log(\pi e(P+N_0)) - \log(\pi e N_0) = \log\left(1 + rac{P}{N_0}
ight)$$

Note: still need to prove achievability and converse

- Given  $y \sim \mathcal{CN}(0, P + N_0)$  , then  $x \sim \mathcal{CN}(0, P)$
- Capacity achieved by transmitting with full power a randomly selected codeword from an i.i.d. Gaussian codebook



- Since input x is independent, does it mean coding is not needed?
  - No, asymptotically independent for large N
  - Coding is crucial to approach capacity

- Note: Capacity is a channel characteristic (independent of Tx/Rx technique or limitation)
- For practical systems with finite constellations (QPSK,QAM)
  - Need to find codes that approach the capacity
  - Codes long enough to average noise
  - Code should be easily encodable/decodable
- Advanced techniques (turbo, LDPC) perform close to capacity (capacity approaching codes)

Consider cts-time AWGN with bandwidth W

$$y(t) = x(t) + n(t)$$

Discrete-time complex baseband channel (sampled at Nyquist 1/W)

$$y_i = x_i + n_i$$

where 
$$a_i = a(i/W), \quad n_i \sim \mathcal{CN}(0, N_0W)$$

Capacity:

$$C = \log \left(1 + \frac{P}{WN_0}\right)$$
 [bits/symbol], log(.): log<sub>2</sub>(.)

- Maximum symbol rate: W symbols/second in W [Hz]= 1 symbol/s/Hz
  - Why? Nyquist criterion for band-limited signals (no ISI)

Capacity/unit time: 
$$C = W \log \left( 1 + rac{P}{W N_0} 
ight)$$
 [b/s]

• Spectral efficiency: 
$$C/W = \log\left(1 + \frac{P}{WN_0}
ight)$$
 [b/s/Hz]

- Key resources: Power and bandwidth
- Power in term of  $SNR=P/(N_oW)=x$ :

$$\log(1+x) \approx x/\ln(2), \quad x \approx 0$$
  
 $\log(1+x) \approx \log(x), \quad x \gg 1$ 

- For low SNR ( $\approx$ 0), C increases *linearly* with P (twice P → twice C)
- For high SNR (>>1), C increases *logarithmically* (twice  $P \rightarrow C+1$ )

- C can be shown to be increasing/concave function of W
  - When W small, SNR large: Linear increase in W compensates for log decrease in SNR. Increasing W yield rapid increase in C (bandwidth-limited regime)
  - When W large, SNR small: Increase in W cannot compensate for log decrease in SNR. Increasing W has small impact on C (power-limited regime)

$$W \log \left(1 + \frac{P}{WN_0}\right) \approx W \left(\frac{P}{WN_0 \ln(2)}\right) = \frac{P}{N_0 \ln(2)}$$

• When W in infinite 
$$\lim_{W \to \infty} W \log \left( 1 + \frac{P}{W N_0} \right) \to \frac{P}{N_0 \ln(2)}$$

$$y(k) = x(k) + n(k), n(k) : AWGN$$

$$\operatorname{Capacity:} \left(\frac{C}{B}\right) = \log_2\left(1 + \frac{E_b}{N_0}\left(\frac{C}{B}\right)\right) \quad b/s/Hz, \frac{E_b}{N_0}\left(\frac{C}{B}\right) = \frac{P}{N_0B} = SNR$$

$$\Rightarrow C = B\log_2\left(1 + SNR\right) \approx \begin{cases} \frac{P}{N_0}\log_2 e, \ for \ SNR \approx 0 \\ B\log_2 SNR, \ for \ SNR >> 1 \end{cases}$$

$$C \leq \frac{P}{N_0}\log_2 e \rightarrow \\ \text{Inear in power,} \quad \text{insensitive to bandwidth:} \quad \text{Capacity is finite even if W is not} \quad \text{BANDWIDTH-LIMITED REGION:} \quad \text{Bandwidth} \quad \text{Capacity is finite even in bandwidth} \quad \text{Bandwidth} \quad \text{Bandwidt$$

- How about minimizing energy per bit  $E_b = P/C$
- Operate at most power-efficient regime
- Minimum bit SNR required for reliable communications

$$\left(\frac{E_b}{N_0}\right)_{\min} = \lim_{P \to 0} \frac{P}{CN_0} = \ln(2) = -1.59 \text{ dB}$$

Consider flat-fading channel

$$y = hx + n$$

where noise  $n \sim \mathcal{CN}(0, N_0)$ 

- Without loss of generality, fading process  $\mathbb{E}[|h|^2] = 1$
- Capacity depends on how fast channel changes
  - Fast fading: Codeword spans several coherence periods
  - Slow fading: Codeword spans a single coherence interval
- Depends also on that is known about the channel (CSI)
  - Channel distribution known at Tx/Rx (hard)
  - Rx/Tx know distribution, and Rx knows channel gain (CSIR)
  - Rx/Tx know distribution and channel gain (CSIT)

#### Fast Fading – CSIR

- Block fading: h constant for T<sub>c</sub> symbol periods and changes after
- CSIR: Rx tracks channel value
- Channel can be modeled as having two outputs  $x \longrightarrow f(y|x) \xrightarrow{b} h$
- Capacity

$$C = \max_{f_X(x)} \quad I(x, \{y, h\}) \quad \text{s.t.} \quad \mathbb{E}[|x|^2] \leq P$$

From chain rule of mutual info and from independence of x and h

$$I(x, \{y, h\}) = I(x, h) + I(x, y|h) = I(x, y|h)$$

where conditional mutual info defined similar to conditional entropy

$$I(x,y|h) = h(y|h) - h(y|x,h) = \mathbb{E}_h[\underbrace{I(x,y|h=\mathrm{h})}]$$

Mutual info btw input & output conditioned on realization of h

#### Fast Fading – CSIR

- Condition on a realization of h, channel is AWGN
  - Gaussian inputs with full power optimal

$$I(x,y|h=\mathrm{h})=\log\left(1+rac{P|\mathrm{h}|^2}{N_0}
ight)$$

CSIR capacity

$$C = \mathbb{E}_h \left[ \log \left( 1 + rac{P|h|^2}{N_0} 
ight) 
ight]$$

 This is known as ergodic capacity: codeword must be long enough to capture ergodicity of channel

# Fast Flat-Fading Channel: Fading Known at Receiver only

$$y(k) = hx(k) + n(k), n(k) : AWGN, \quad a = |h|^2 : \text{ power fading with pdf } p(a), E\{a\} = 1$$
  
instantaneous:  $P_s(a) \approx A_M Q\left(\sqrt{B_M^2(aSNR)}\right), C(a) = \log_2\left(1 + aSNR\right) \text{ b/s/Hz},$ 

#### For Fast Fading, $T_{\text{symbol}} \approx T_{\text{coherence}}$ or $> T_{\text{coherence}}$

- introduced random phase can remove correlation between symbol phases, and hence leads to an irreducible error floor for differential modulation/demodulation.
- For coherent demodulation, if fading is known by receiver only, taking average over many independent fades h, we have

Average (symbol) error probability: 
$$\overline{P}_s = \int_0^s P_s(a)p(a)da$$

ergodic:
$$C = E\{C(a)\} = \int_0^\infty \log_2(1 + aSNR) p(a) da \le \log_2(1 + SNR)$$

i.e., If FADING KNOWN AT RECEIVER ONLY, at best, ergodic C approaches C<sub>AWGN</sub>

(using Jensen's inequality, i.e.,  $E\{f(u)\}\le f(E\{u\})$  for a strictly concave f(u))

#### Fast Fading – CSIR

Define average received

$$SNR = \mathbb{E}_h[P|h|^2/N_0] = P/N_0$$

- How to achieve this capacity?
  - 1.- Use **single-rate** Gaussian codebook with rate=C over multiple h
  - 2.- Use **multi-rate** Gaussian codebook: for each fading state, rate= $log(1 + |h|^2SNR)$  (need CSIT, can do better)
- How does it compare to AWGN?
  - From Jensen's inequality  $\mathbb{E}[f(u)] \leq f(\mathbb{E}[u])$  for strictly concave f

$$C = \mathbb{E}_h[\log(1+|h|^2\mathrm{SNR})] \le \log(1+\mathrm{SNR}) = C_{\mathrm{awgn}}$$

- Equality when h is deterministic
- Peaks cannot compensate for valleys

#### Fast Fading – CSIR

Low SNR: fading approaches AWGN

$$C = \mathbb{E}[\log(1+|h|^2\mathrm{SNR})] pprox \mathbb{E}\left[rac{|h|^2\mathrm{SNR}}{\ln(2)}
ight] = rac{\mathrm{SNR}}{\ln(2)} pprox C_{\mathrm{awgn}}$$

High SNR: constant difference

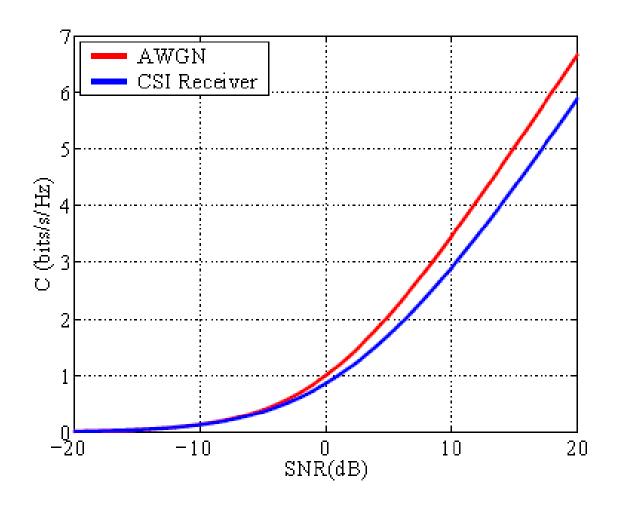
$$C pprox \mathbb{E}[\log(|h|^2 \mathrm{SNR})]$$

$$= \log(\mathrm{SNR}) + \mathbb{E}[\log(|h|^2)]$$

$$\approx C_{\mathrm{awgn}} + \mathbb{E}[\log(|h|^2)]$$

For Rayleigh, -0.83 b/s/Hz (around 2.5dB)

# Fast Fading: Capacity if Channel is Known at Receiver only



■ If fading is known at receiver only, at best, C<sub>CSIR</sub> approaches C<sub>AWGN</sub>

#### **Capacity of Flat-Fading Channels**

- If fading is also known at transmitter, can we do better?
- Consider a family of codes, one for each possible fading state a, and build a variable-rate coding scheme that adaptively selects code with appropriate rate depending on a:
  - We achieve log(1 + SNR<sub>a</sub>) for each state with fixed SNR
  - Therefore, at least we can achieve C = E[log(1+SNRa)].
  - Can we do better?
- We can actually do better by adapting the transmitted power for each state A(a)SNR to achieve

instantaneous: 
$$C[A(a)] = \log_2[1 + aA(a)SNR]$$

$$\Rightarrow$$
 ergodic: $C_A = \int_0^\infty \log_2 [1 + aA(a)SNR] p(a) da$ 

- can have a capacity greater than that of the AWGN channel, i.e., fading can provide more opportunities for performance enhancement in an opportunistic communication approach.
- How to choose A(a)?

#### **Adaptive Channel Inversion?**

• P: average Tx power, B: Tx bandwidth,  $E\{a\}=1$  for avg Rx SNR=  $E\{a\}P/BN_o=P/BN_o$ 

Channel Inversion: With a know, find an appropriate constant  $\alpha$  and set  $A(a)=\alpha/a$  to keep the instantaneous Rx SNR= $A(a)[aP/(BN_o)]$ = SNR<sub>ZO</sub> (= $\alpha P/(BN_o)$ : constant) for zero outage under average Tx power constraint.

instantaneous Tx power:  $A(a)P = \alpha P/a$   $\rightarrow$  zero-outage:  $C_{ZO} = C[A(a)] = \log_2 [1 + SNR_{ZO}]$ 

but with Tx power constraint: 
$$\int_0^\infty [\alpha P/a] p(a) da = P \to \alpha = [E\{1/a\}]^{-1}$$

$$\Rightarrow$$
 ergodic, zero-outage: $C_{ZO} = \log_2 \left[ 1 + SNR_{ZO} \right]$  with  $SNR_{ZO} = \frac{P/(BN_o)}{E\{1/a\}}$ 

for Rayleigh fading channel:  $E\{a^{-1}\} = \int_{a_0}^{\infty} a^{-1} e^{-a} da \approx e^{-a_0} \ln(1 + a_0^{-1}) \rightarrow \infty \text{ as } a_0 \rightarrow 0 \Rightarrow C_{ZO} = 0$ 

- simplifies design (i.e., fixed rate at all channel states) but is power-inefficient since for very small a,  $A(a) = \alpha/a$  is very large.
- achieves a delay-limited capacity, but with a greatly reduced capacity.

**Truncated inversion:**  $A(a) = \alpha/\max(a, a^*)$  only if a is above cutoff fade depth  $a^*$ 

- to maintain constant SNR (and hence fixed rate) above cutoff  $a^*$
- to increase capacity with appropriate choice of cutoff  $a^*$ : Close to optimal

# Fast Fading – CSIT: Opportunistic transmission

- CSIT: Tx/Rx track channel
  - How Tx knows channel? Feedback or channel reciprocity in TDD
- Opportunistic transmission: Tx can adapt to channel conditions
- Assume codeword spans L coherence intervals
- modelled as a parallel channel

$$y_l = h_l x_l + n_l \quad (l = 1, \dots, L)$$

- As shown before, for given channel, Gaussian optimal
- Average rate

$$\frac{1}{L} \sum_{l=1}^{L} I(x_l, y_l | h = h_l) = \frac{1}{L} \sum_{l=1}^{L} \log \left( 1 + \frac{P_l |h_l|^2}{N_0} \right)$$

power allocated to sub-channel l:  $P_l$ 

#### Fast Fading – CSIT

Capacity

$$\max_{P_l} \frac{1}{L} \sum_{l=1}^{L} \log \left( 1 + \frac{P_l |h_l|^2}{N_0} \right) \quad \text{s.t.} \quad \frac{1}{L} \sum_{l=1}^{L} P_l \le P$$

Optimal solution (using KKT): waterfilling

$$P_l^{\star} = \left(\frac{1}{\lambda} - \frac{N_0}{|h_l|^2}\right)^+$$

where lambda satisfies full power

$$\frac{1}{L} \sum_{l=1}^{L} \left( \frac{1}{\lambda} - \frac{N_0}{|h_l|^2} \right) = P$$

Problem?

#### Fast Fading – CSIT

However, when  $L \to \infty$ 

$$\frac{1}{L} \sum_{l=1}^{L} \left( \frac{1}{\lambda} - \frac{N_0}{|h_l|^2} \right) \to \mathbb{E} \left[ \left( \frac{1}{\lambda} - \frac{N_0}{|h|^2} \right)^+ \right] = P$$

$$\frac{1}{L} \sum_{l=1}^{L} \log \left( 1 + \frac{P_l^{\star} |h_l|^2}{N_0} \right) \to \mathbb{E} \left[ \log \left( 1 + \frac{P^{\star}(h)|h|^2}{N_0} \right) \right]$$

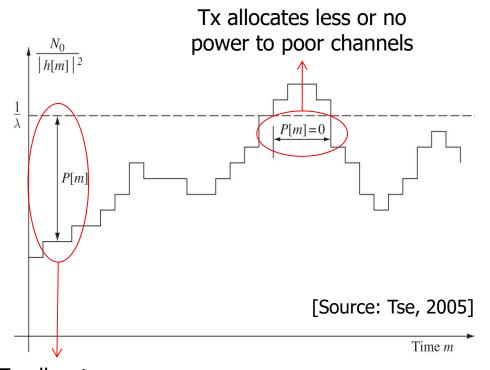
Capacity

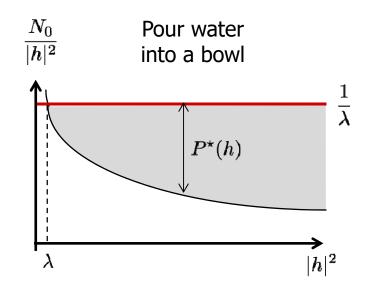
$$C = \mathbb{E}\left[\log\left(1 + \frac{P^{\star}(h)|h|^2}{N_0}\right)\right] \qquad P^{\star}(h) = \left(\frac{1}{\lambda} - \frac{N_0}{|h|^2}\right)^{+}$$

$$P^{\star}(h) = \left(\frac{1}{\lambda} - \frac{N_0}{|h|^2}\right)^{+}$$

Depends on present value of h and statistics

#### Waterfilling



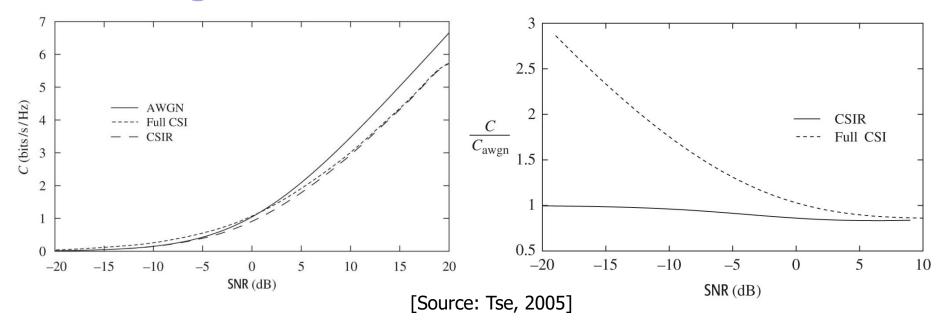


Tx allocates more power to good channels

#### **Fast Fading – CSIT**

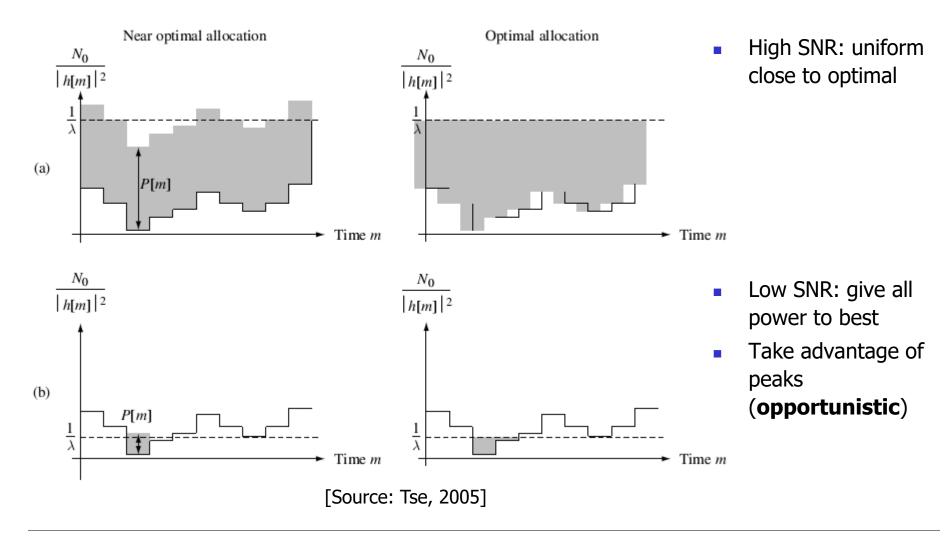
- How to achieve?
  - 1.- Multi-rate variable-power Gaussian scheme: for each fading state, adapt rate and power to  $\log(1 + P^*(h)|h|^2/N_0)$
  - 2.- Single-rate variable-power Gaussian scheme: adapt power but keep constant rate  $\mathbb{E}[\log(1+P^{\star}(h)|h|^2/N_0)]$
- 1<sup>st</sup> strategy simpler (no need to code across channel states)
- How does it compare to AWGN or CSIR?

#### **Fast Fading – CSIT**



- Low SNR: CSIT better than CSIR, and even better than AWGN!
- High SNR: CSIT and CSIR comparable performance (AWGN best)
  - Still good since it simplifies coding
- Capacity very sensitive to dynamic power allocation in low SNRs but insensitive in high SNRs

# **Fast Fading – CSIT**



# Fast Fading – CSIT: Channel inversion

- suboptimal solution
- Invert fading: Convert channel into AWGN  $P(h) = \sigma/|h|^2$

$$\mathbb{E}[P(h)] = \mathbb{E}\left[\frac{\sigma}{|h|^2}\right] = P \ \Rightarrow \ \sigma = \frac{P}{\mathbb{E}[1/|h|^2]}$$

$$C = \log\left(1 + \frac{P}{N_0 \mathbb{E}[1/|h|^2]}\right)$$

- Exact opposite to waterfilling
- Simplifies design (fixed rates for all channel states, zero outage)
- Power inefficient (hard to invert poor channels)
- In Rayleigh,  $\mathbb{E}[1/|h|^2] \to \infty$  so capacity is zero

# Fast Fading - CSIT: Truncated channel inversion

Invert fading only when channel above cut-off fade

$$\begin{split} P(h) &= \left\{ \begin{array}{l} \sigma/|h|^2, & |h|^2 \geq |h_0|^2 \\ 0, & |h|^2 < |h_0|^2 \end{array} \right. \\ \sigma &= \frac{P}{\mathbb{E}_{h_0}[1/|h|^2]}, \quad \mathbb{E}_{h_0}[1/|h|^2] = \int_{|h_0|^2}^{\infty} \frac{1}{|h|^2} f(|h|^2) \, d|h|^2 \\ C &= \log \left( 1 + \frac{P}{N_0 \mathbb{E}_{h_0}[1/|h|^2]} \right) \cdot \mathbb{P}(|h|^2 \geq |h_0|^2) \end{split}$$

- Constant rate for good channels, no rate for poor ones (outage)
- Still need to find optimal value of cut-off

#### Slow Fading – CSIR

- In slow fading, codeword spans single realization
- Conditioned on realization, maximum rate is  $\log(1 + |h|^2 \text{SNR})$ 
  - Max rate is random
- Regardless of selected rate, there is a probability that

$$\log(1 + |h|^2 \text{SNR}) < R$$

- Error rate cannot be made arbitrarily small
- Ergodic capacity theoretically zero
- System is said to be in outage

$$P_{\text{out}}(R) = \mathbb{P}\{\log(1+|h|^2\text{SNR}) < R\}$$

 Reliable communication possible as long as not in outage (deep fade)

# Slow Fading – CSIR

Example: Rayleigh fading

$$P_{ ext{out}}(R) = 1 - \exp\left(rac{-(2^R-1)}{ ext{SNR}}
ight)$$

At high SNR

$$P_{
m out}(R)pprox rac{(2^R-1)}{
m SNR}$$

- Decays as 1/SNR: coding cannot improve significantly error performance over slow fading
- ε-outage capacity: alternative performance measure
- Largest rate such that  $P_{\mathrm{out}}(R) < \epsilon$

$$C_{\epsilon} = \log(1 + F^{-1}(1 - \epsilon) \text{SNR})$$

where complementary CDF  $F(x) = \mathbb{P}\{|h|^2 > x\}$ 

#### Slow Fading – CSIR

- How does it compare to AWGN?
  - At high SNRs: relative loss gets smaller at high SNRs (constant difference)

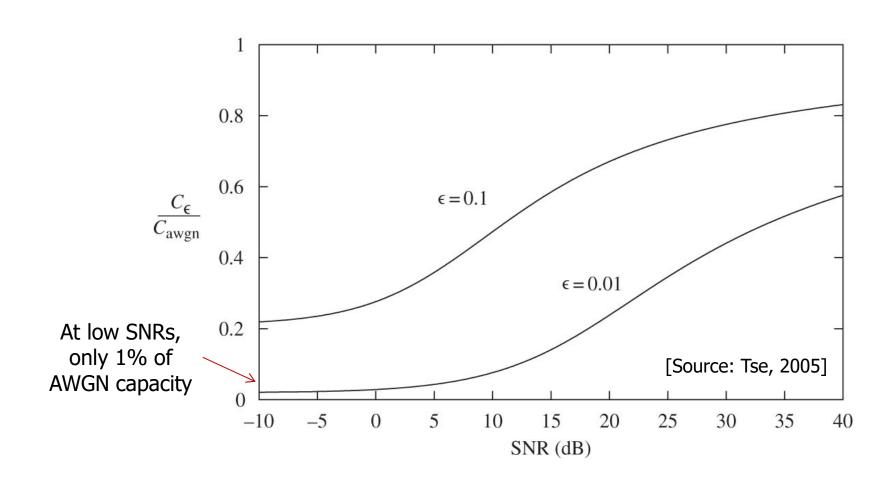
$$C_{\epsilon} \approx \log(\text{SNR}) + \log(F^{-1}(1 - \epsilon))$$
  
=  $C_{\text{awgn}} - \log(1/F^{-1}(1 - \epsilon))$ 

At low SNRs: only a small fraction of AWGN

$$C_{\epsilon} \approx F^{-1}(1 - \epsilon) \cdot \text{SNR}/\ln(2)$$
  
=  $F^{-1}(1 - \epsilon) \cdot C_{\text{awgn}}$ 

• For Rayleigh with small  $\epsilon$ :  $F^{-1}(1-\epsilon) \approx \epsilon$ 

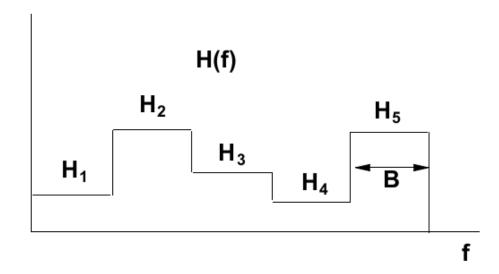
# Slow Fading – CSIR



#### **Capacity of Frequency-Selective Channels**

#### Time-invariant

- Consider time-invariant channel with frequency response H(f)
- Assume H(f) is block fading with N subcarriers B=W/N



This can be modeled as a parallel channel with

$$SNR = \frac{|H_n|^2 P_n}{N_0 B}$$

#### **Capacity of Frequency-Selective Channels**

#### **Time-invariant**

Capacity

$$\max_{P_n} \sum_{n=1}^{N} \log \left( 1 + \frac{P_n |H_n|^2}{N_0 B} \right) \quad \text{s.t.} \quad \frac{1}{N} \sum_{n=1}^{N} P_n \le P$$

- Similar to parallel channel before
- Waterfilling over frequency optimal

$$P_n^{\star} = \left(\frac{1}{\lambda} - \frac{N_0 B}{|H_n|^2}\right)^{+} \qquad \frac{1}{N} \sum_{n=1}^{N} \left(\frac{1}{\lambda} - \frac{N_0 B}{|H_n|^2}\right) = P$$

- How to achieve?
  - Separate coding/power for each subcarrier

#### **Capacity of Frequency-Selective Channels**

#### **Time-invariant**

As N approaches infinity, for cts H(f)

$$P^\star(f) = \left(rac{1}{\lambda} - rac{N_0}{|H(f)|^2}
ight)^+ \qquad \int P^\star(f) \, df = P$$

$$C = \int \log \left( 1 + \frac{P^{\star}(f)|H(f)|^2}{N_0} \right) df$$

#### References

- A. Goldsmith, Wireless Communications, Cambridge University Press, 2005, Chapter 4D.
- Tse, P. Viswanath, Fundamentals of Wireless Communication, Cambridge University Press, 2005, Chapter 5
- and materials from various sources