

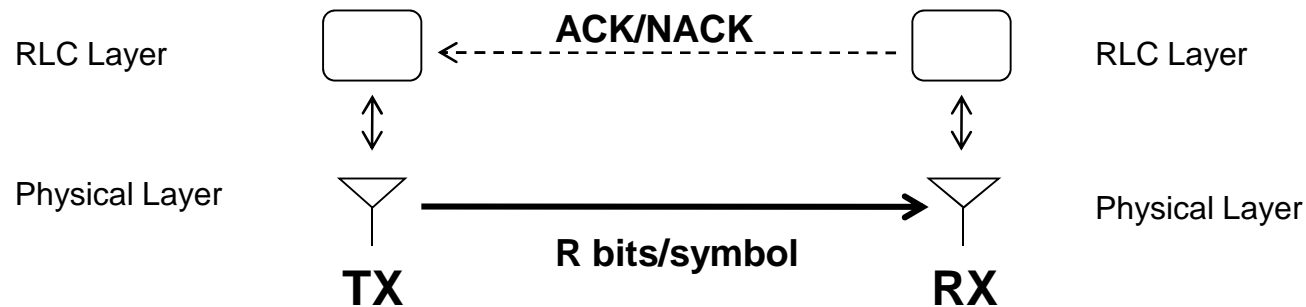


Coding vs. ARQ in Fading Channels: How reliable should the PHY be?

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Introduction



- Two different and competing reliability mechanisms:
 - Coding: Reducing transmitted rate R (bits/sym) reduces packet error probability ϵ
 - ARQ: Retransmit packets not received correctly
- What is the optimal balance for reliable communication?



Main Assumptions

- Basic assumptions:
 - Perfect CSI at receiver, transmitter only knows channel statistics (high mobility scenario)
 - Packets retransmitted until successfully decoded
 - Error free ACK/NACK
 - Perfect error detection
 - Errors independent across packet transmissions
 - At PHY, either of HARQ and no HARQ can be used

- Important performance measure: long-term avg. successful rate

$$\eta = \frac{R}{E[X]} \text{ bits/symbol}$$

- X: number of PHY transmission attempts for each packet
- η is also referred as goodput/successful throughput



PHY Reliability without PHY HARQ

- Number of packet TX's ~ Geometric (1- ϵ) :

$$E[\# \text{ of packet TX' s}] = \frac{1}{1 - \epsilon}$$

- Successful throughput/goodput :

$$R(1 - \epsilon) \text{ bits/sym}$$

- Error prob. increases in TX rate R
- A reliable PHY layer requires the sacrifice of TX rate
- Objective: Characterize ϵ^* as a function of channel parameters (SNR, time/frequency selectivity)

$$\epsilon^* = \arg \max_{\epsilon} R(1 - \epsilon)$$

- Same issue arises for optimization of application layer fountain codes



Channel Model without PHY HARQ

- Fast fading: channel varies too rapidly to be tracked at TX
- Block-fading (order L time/freq selectivity): $y_t = \sqrt{SNR} h_k x_t + z_t$
 - h_k : channel in k -th fading block, iid across blocks
 - Channels are complex Gaussian (Rayleigh fading), known at RX
 - SNR: average received signal-to-noise ratio

- With infinite blocklength and strong channel code, successfully decode iff:

$$\frac{1}{L} \sum_{i=1}^L \log_2 (1 + |h_i|^2 SNR) > R$$

- Errors due to bad channel realizations
- Packet error probability = mutual information outage probability

$$\varepsilon = \mathbb{P} \left[\frac{1}{L} \sum_{i=1}^L \log_2 (1 + |h_i|^2 SNR) < R \right]$$

Gaussian Approximation

- Approximate RX mutual information by Gaussian:

$$\varepsilon = \mathbb{P} \left[\underbrace{\frac{1}{L} \sum_{i=1}^L \log_2(1 + |h_i|^2 \text{SNR})}_{\sim N(\mu(\text{SNR}), \sigma^2(\text{SNR})/L)} < R \right] \approx Q \left(\frac{\mu(\text{SNR}) - R}{\sigma(\text{SNR})/\sqrt{L}} \right)$$

$$\Rightarrow R \approx \mu(\text{SNR}) - Q^{-1}(\varepsilon) \frac{\sigma(\text{SNR})}{\sqrt{L}}$$

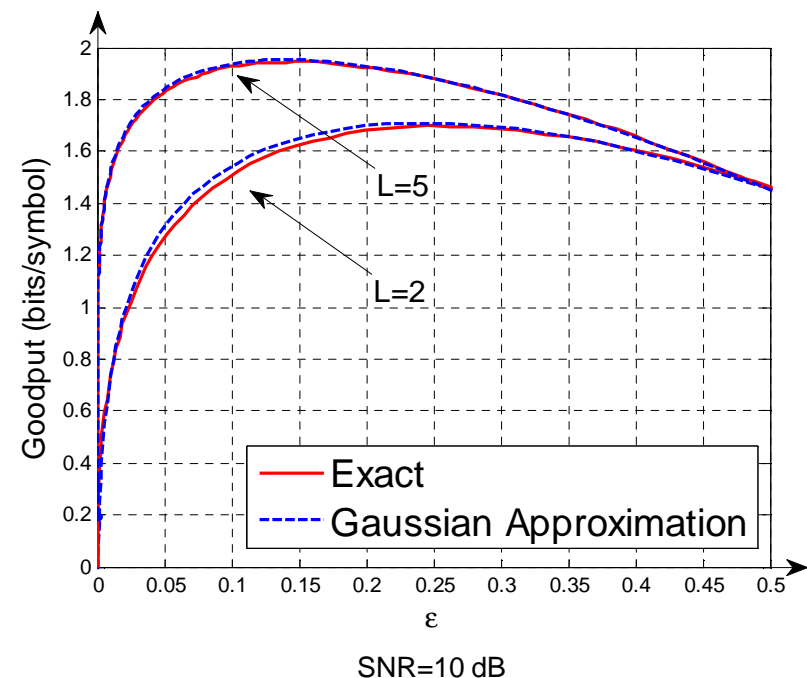
- Goodput approximation:

$$\eta_g = \mu(\text{SNR}) (1 - \kappa \cdot Q^{-1}(\varepsilon)) (1 - \varepsilon)$$

- κ : μ -normalized standard deviation

$$\kappa = \frac{\sigma(\text{SNR})}{\mu(\text{SNR})/\sqrt{L}}$$

- κ decreases in both SNR and L
- η_g is strictly concave in ε



Goodput Optimization

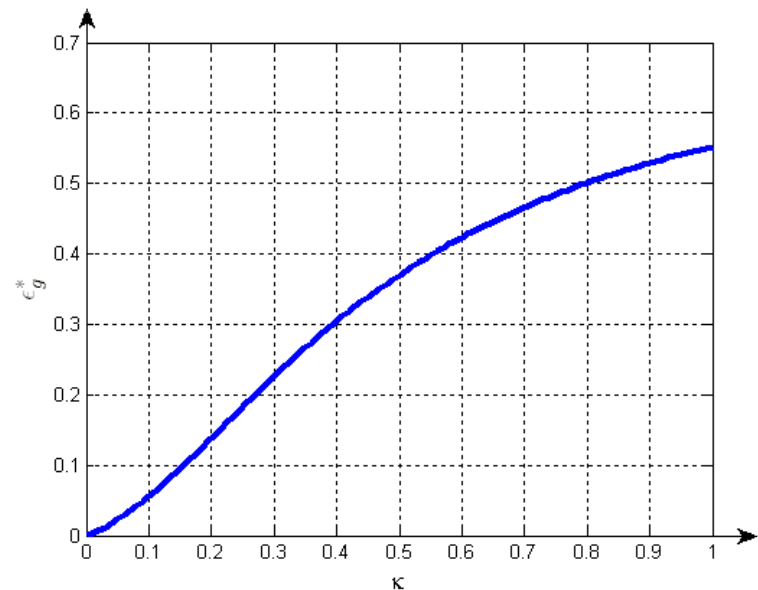
- η_g - maximizing error probability:

$$\varepsilon_g^*(SNR, L) = \arg \max_{\varepsilon} (1 - \kappa \cdot Q^{-1}(\varepsilon))(1 - \varepsilon)$$

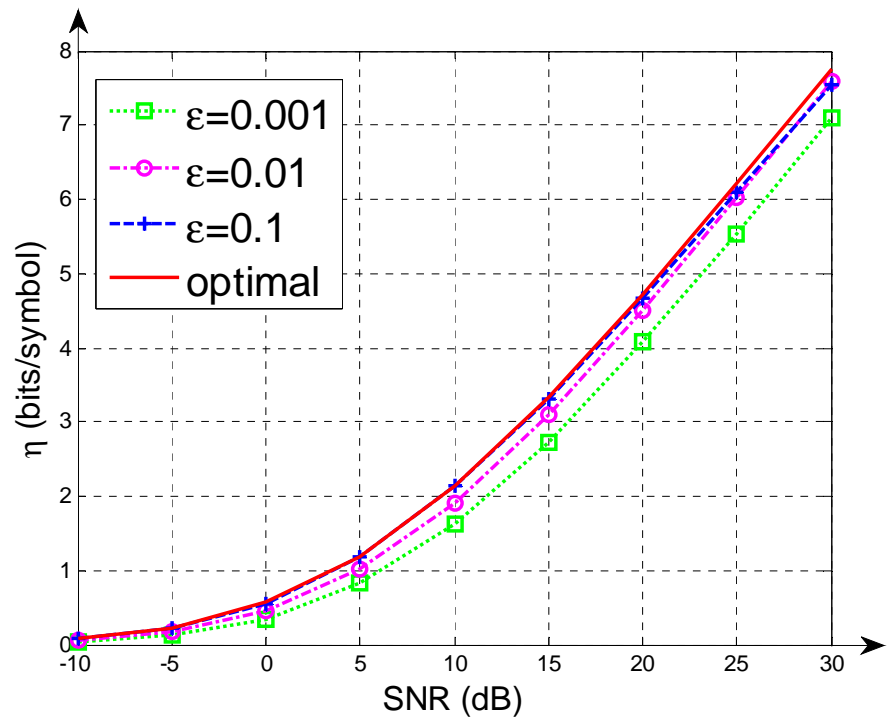
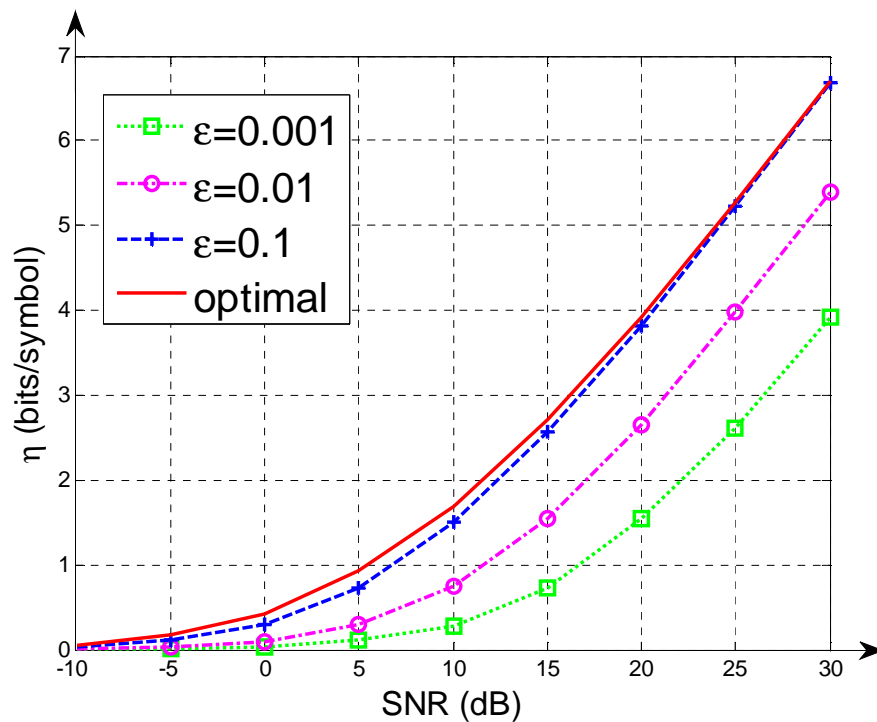
- ε_g^* satisfies

$$\left(Q^{-1}(\varepsilon_g^*) - (1 - \varepsilon_g^*) \frac{dQ^{-1}(\varepsilon)}{d\varepsilon} \Big|_{\varepsilon = \varepsilon_g^*} \right)^{-1} = \kappa$$

- ε_g^* is only determined by κ
 - Basic analysis can be easily extended to MIMO and other fading distributions
 - ε_g^* monotonically increases in κ (decreases in SNR and L)



Goodput vs. SNR



- 10% near optimal for wide range of selectivity (L)
- Making PHY too reliable incurs significant penalty in open loop systems

Incremental Redundancy (IR)

- Outage if cannot decode after L IR rounds-> triggers RLC ARQ retransmission
- Frequency-flat channel during each IR round, i.i.d. fading across rounds
- If initial transmission rate is RL bits/sym, # of IR rounds is smallest \hat{x} s.t.

$$\sum_{i=1}^{\hat{x}} \log_2 (1 + |h_i|^2 SNR) > RL$$

- \hat{X} : random variable describing # of IR rounds

$$P[\hat{X} \leq k] = \begin{cases} P\left[\sum_{i=1}^k \log_2 (1 + |h_i|^2 SNR) > RL\right] & k = 1, \dots, L-1 \\ 1 & k = L \end{cases}$$

- Error probability identical to expression without IR
- Relationship between R and ε unaffected by IR

$$\varepsilon = P\left[\sum_{i=1}^L \log_2 (1 + |h_i|^2 SNR) < RL\right]$$

- Expectation of \hat{X} is $E[\hat{X}] = 1 + \sum_{k=1}^{L-1} P\left[\sum_{i=1}^k \log_2 (1 + |h_i|^2 SNR) < RL\right]$

Goodput Optimization with IR

- Successful throughput (i.e., goodput):

$$\frac{RL}{E[\hat{X}]/(1 - \varepsilon)} = \frac{RL}{E[\hat{X}]} (1 - \varepsilon) = \underbrace{R(1 - \varepsilon)}_{\text{PHY Transmitted Rate}} \underbrace{\frac{L}{E[\hat{X}]}}_{\text{Goodput without IR}}$$

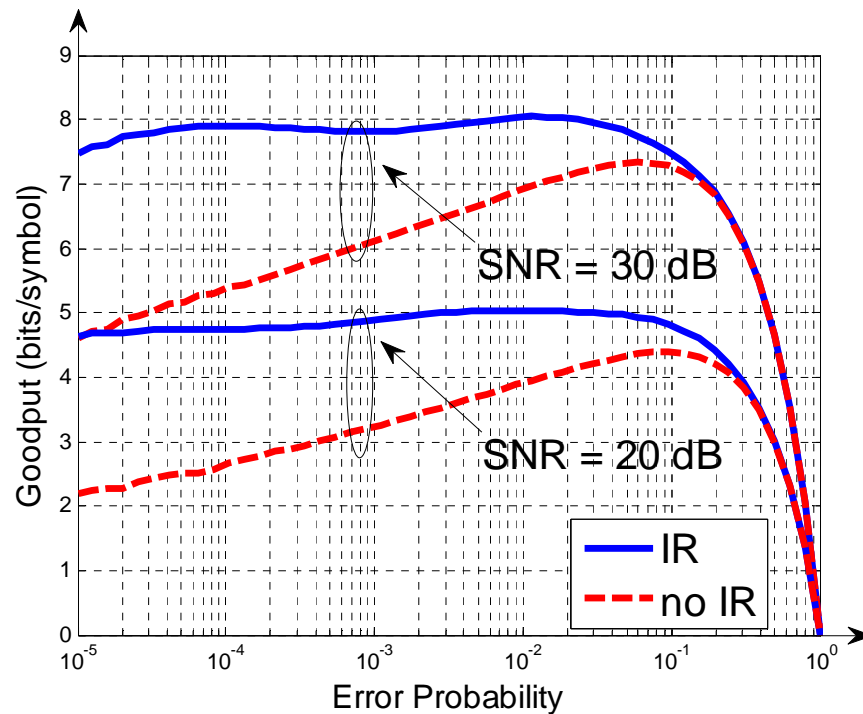
- Optimal operating point:

$$\varepsilon_{IR}^* (SNR, L) = \arg \max_{\varepsilon} R(1 - \varepsilon) \frac{L}{E[\hat{X}]}$$

- $E[\hat{X}]$: an increasing function of R and ε
- Result: The optimizing error probability with IR is less than or equal to the optimizing error probability without IR:

$$\varepsilon_{IR}^* \leq \varepsilon_{no-IR}^*$$

Goodput (IR) vs. Error Prob.



- If error prob. (or TX rate) too large, no early termination
- If error prob. (or TX rate) too small, always terminate in one IR round
- IR makes goodput less sensitive to the choice of error probability if parameters are carefully chosen



Conclusion and Future Work

- Don't make PHY reliable if it is difficult (e.g., lacking diversity) to do so, let ARQ achieve reliability instead
 - Increase PHY reliability only when it is easy (e.g., plenty of diversity, closed loop system) to do so
 - Same argument applies to rateless coding
- HARQ can achieve lower optimum PHY error probability
 - Insensitivity of goodput to choice of reliability makes PHY optimization almost pointless
- Future work:
 - The effect of HARQ feedback on the goodput/optimal reliability
 - Taking RLC delay into account