

Cross-layer communication over fading channels with adaptive decision feedback

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Motivation

Physical Layer Design Objective

- ① Reliable Communication
- ② Efficiency in Rate and Power

Network Layer Design Objective

- ① Quality of Service(delay, etc.)
- ② System Stability

Cross Layer Design Objective

- ① Physical and Network Layer consideration Reliable Communication
- ② Minimizing both Power and Delay in Reliable Communication

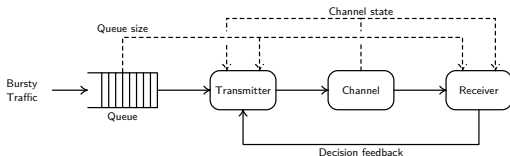
Features of Cross Layer Design in Wireless systems

1 Rate Adaptation in wireless systems:

- Traffic Load in Network
- Channel State

2 Optimization Objectives:

- Block Error Probability
- End to End Packet Delays
- Transmission Power



Motivating Question

Effect of Feedback on System Performance

Effect of an ACK/NACK feedback channel

- It can improve decoding error performance
- It can also increase queuing delay due to re-transmission

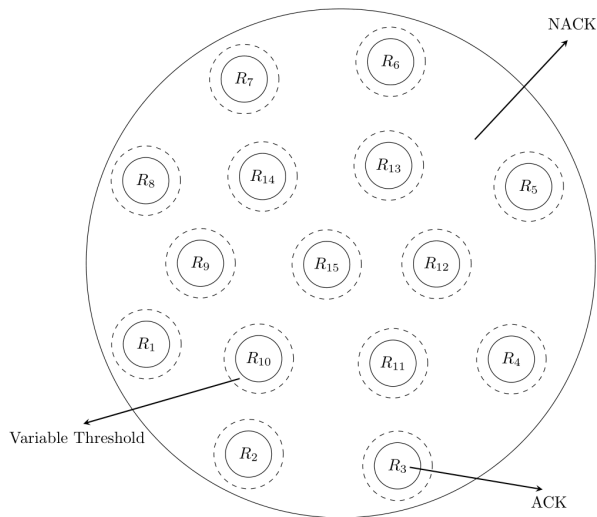
What controls the expected number of retransmissions?

- 1 Channel State (exogenous process)
- 2 The decoding threshold at the receiver.

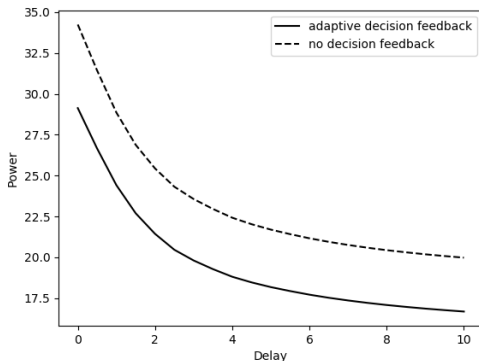
Important Question

- How can we control decoding threshold?

Our main idea: exploit adaptive decision feedback

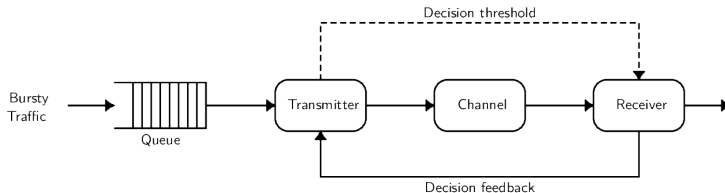


Adaptive Decision feedback implies significantly better power-delay trade-off



$$\bullet P(\alpha) = \min_g \{ \text{power of policy } (g) \mid \text{Delay of policy } (g) \leq \alpha \}$$

Communication System



System Variables

- Number of new packets: A_k
- Queue length: Q_k
- Channel state: S_k
- Number of transmitter packets : U_k
- Decoding threshold : T_k
- Fading gain : $H(S_k)$

Processes and Queue Dynamic

- A_k is an i.i.d process with decreasing and convex pdf
- S_k is an i.i.d process
- $Q_{k+1} = \begin{cases} Q_k - U_k + A_k, & \text{if } \mathfrak{E}_k = 1 \\ Q_k + A_k, & \text{if } \mathfrak{E}_k = 0 \end{cases}$

Solution Approach

Ensuring Reliable Communication

Performance Metrics

- Queuing delay
- Probability of Error
- Transmission power

Our Solution

- Buffering delay is $d(Q_{k+1} - A_k)$
- We upper bound P_e by ε
- How to find transmission power?

Probability of error and re-transmission

By [Forney 1968], we know error exponent for channels with feedback

$$P_e \leq \exp\left(-\frac{\rho}{2} M \log\left(1 + \frac{\pi H(s)}{(1+\rho)}\right) + \rho N u + \frac{1}{1+\rho} M t\right)$$

and

$$P_r = \exp(Mt) P_e$$

Solution Approach

Ensuring Reliable Communication

Finding Power needed to ensure desired probability of error

- ① Power Function is $\phi(u, t)h(s)$, where

$$\phi(u, t) = (1 + \rho) \left[\exp \left(-\frac{2 \log \varepsilon}{\rho M} + 2 \frac{N}{M} u - \frac{2}{1 + \rho} t \right) - 1 \right]$$

- ② probability of re-transmission $p(t)$

$$p_r(U_k, T_k, \Pi_k, S_k) = \exp(MT_k)\varepsilon =: p(T_k)$$

Imposing power constraint

- Maximum power constraint: $\phi(u, t)h(s) \leq \Pi_{\max}$

Solution Approach

Assumptions on the model

Communication cost at each time

$$c(q, s, u, t) = \lambda_{\pi} \phi(u, t) h(s) + \lambda_d [p(t) d(q) + (1 - p(t)) d(q - u)]$$

Assumptions on the Model

- $d(\cdot)$ increasing convex
- $\phi(\cdot, \cdot)$ increasing in u_k
- $\phi(\cdot, \cdot)$ decreasing in t_k
- $h(\cdot)$ increasing, convex
- $p(t)$ increasing in t_k

Solution Approach

MDP Formulation

Problem(1)

State Space:

$$(Q_k, S_k)$$

Action Space:

$$(U_k, T_k)$$

Find policy g which minimizes:

$$J(g) = \mathbb{E} \left[\sum_{k=1}^K c(Q_k, S_k, U_k, T_k) \right]$$

communication rule:

$$(U_k, T_k) = g_k(Q_k, S_k)$$

communication policy :

$$g = (g_1, g_2, \dots, g_N)$$

Question: What are some qualitative properties of policy g ?

Literature Review

Similar Models in the Literature

- Power-delay trade-off only with rate adaption.
[Berry and Gallager 2002],[Uysal-Biyikoglu,Prabhakar, and El Gamal 2002],[Zafer and Modiano 2009],[Bettesh and Shamai 2006],[Goyal, Kumar, and Sharma 2008],[Fu, Modiano, and Tsitsiklis 2006]
 - Dynamic Programming Decomposition/structure of optimal policy in 1 dimension.
- [Cao and Yeh 2008]: Investigated *Adaptive Decision Feedback*, and 2 dimensional structural Properties
 - Only for binary choice for rate and threshold

Key Contribution of This paper

Investigating qualitative structure of multi-dimensional optimal policies

Question

Question: Does the optimal policy have *monotonicity Properties*?

Why this question is important?

- 1 Understanding the physical system
- 2 Can be exploited in online-learning algorithms

Why it's hard to answer?

- 1 Both state and action spaces are multi-dimensional
- 2 Most of previous results in cross layer design are for 1 dimensional systems.

Solution and Main Results

Dynamic Programming Decomposition

Defining value functions and action-value functions

Proposition 1

For any q and s

$$V_{K+1}(q, s) = 0$$

and for $k \in \{K, K-1, \dots, 1\}$,

$$W_k(q, s, u, t) = c_k(q, s, u, t) + \sum_{(s,a)} P_S(s)P_A(a) [(1-p(t))V_{k+1}(q-u+a, s) + p(t)V_{k+1}(q+a, s)]$$

- Value function : $V_k(q, s) = \min_{u,t} W_k(q, s, u, t)$.
- Then: $g_k(q, s) \rightarrow \arg \min$ at stage $k \Rightarrow g = (g_1, \dots, g_K)$ is optimal for Problem.

Main Results

Structural Properties of Value Function

Theorem 1

For any time slot k , the value function $V_k(q, s)$ satisfies the following properties:

- ① For any $s \in \mathcal{S}$, $V_k(q, s)$ is weakly increasing in q .
- ② For any $q \in \mathcal{Q}$, $V_k(q, s)$ is weakly increasing in s .

- $1 \Rightarrow \uparrow \text{ number of packets} \Rightarrow \uparrow \text{ cost to go.}$
- $2 \Rightarrow \text{ channel deteriorates} \Rightarrow \uparrow \text{ cost to go.}$

Main Results

Structural Properties of Optimal policy

Theorem 2 (Informal Representation)

- \uparrow number of packets \Rightarrow (Rate \uparrow) Or (Threshold \downarrow).
- channel deteriorates \Rightarrow (Rate \downarrow) Or (Threshold \uparrow).

Main Results

Structural Properties of Optimal policy

Theorem 3

Suppose the cost function satisfies the following property:

(P) for any $(q, s) \in \mathcal{Q} \times \mathcal{S}$, and any $u_1, u_2 \in \mathcal{U}$ and $t_1, t_2 \in \mathcal{T}$ such that $u_1 \leq u_2$ and $t_1 \leq t_2$, we have

$$c(q, s, u_1, t_2) + c(q, s, u_2, t_1) \leq c(q, s, u_2, t_2) + c(q, s, u_1, t_1).$$

Then, the “or” in Theorem 2 can be replaced by “and”.

- Verify property (P) \rightarrow system parameters $(\lambda_\pi, \lambda_d, \Pi_{\max}, M)$
- 1 \Rightarrow \uparrow number of packets \Rightarrow (Rate \uparrow) And (Threshold \downarrow).
- 2 \Rightarrow channel deteriorates \Rightarrow (Rate \downarrow) And (Threshold \uparrow).

Key Idea of the Proof

Partial Order on Action Space

Definition

$\preceq_{\mathcal{A}}$: a partial order on $\mathcal{U} \times \mathcal{T}$

- we say $(u_1, t_1) \preceq_{\mathcal{A}} (u_2, t_2)$ if $u_1 \leq u_2$ and $t_1 \geq t_2$

- 1 Fixed $s \in \mathcal{S}$, $W_k(q, s, u, t)$ has decreasing differences on $\mathcal{Q} \times (\mathcal{U} \times \mathcal{T})$.
- 2 Fixed $q \in \mathcal{Q}$, $W_k(q, s, u, t)$ has increasing differences on $\mathcal{S} \times (\mathcal{U} \times \mathcal{T})$.
- 3 Given the sub-modularity on $(\mathcal{U} \times \mathcal{T}) \Rightarrow$ Theorem 3.

Conclusion and Future Directions

Conclusion

- 1 We investigate the impact of adaptive decision feedback on power-delay curve.
- 2 We establish *monotonicity* property of the optimal policy and value function.

Future Directions

- 1 Extend these results to *infinite horizon setup*
- 2 Develop online learning algorithms to utilize such properties.

Thank you!

Questions

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