

Reverse Engineering MAC: A Non-Cooperative Game Model

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Summary

Reverse engineering:

Given the solution, what is the problem?

Know what works, what doesn't, why it works, how to improve.

Summary

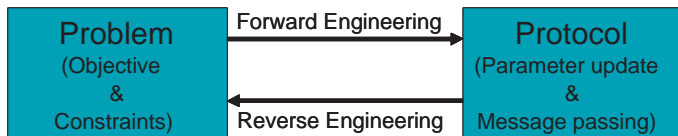
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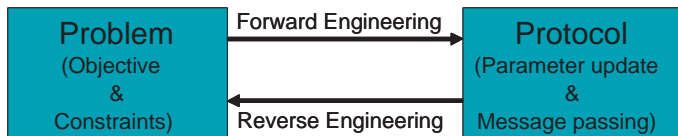
Know what works, what doesn't, why it works, how to improve.

Provide the **missing piece** (on MAC) for rigorous mathematical understanding of existing layers 2-4 protocols

Reverse Engineering



Reverse Engineering



- Related works:

- ▶ Layer 4: TCP/AQM [Kelly-Maulloo-Tan98, Low03, Kunniyur-Srikant03, ...] [NUM](#)
- ▶ Layer 3: BGP [Griffin-Shepherd-Wilfong02] [SPP](#)
- ▶ Layer 2: MAC (contention avoidance in random access) [This Talk]

Review: TCP/AQM

Network Utility Maximization Problem

$$\begin{aligned} & \text{maximize} && \sum_s U_s(x_s) \\ & \text{subject to} && \sum_{s:l \in L(s)} x_s \leq c_l, \quad \forall l, \\ & && \mathbf{x}^{\min} \preceq \mathbf{x} \preceq \mathbf{x}^{\max}. \end{aligned}$$

- $U_s(x_s)$: utility of each user depends on its **own** data rate
- **Adequate** feedback from the network

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- $U_s(x_s)$: utility of each user depends on its **own** data rate
- **Adequate** feedback from the network
- Reverse engineering provides
 - ▶ **Better understanding**: existence, uniqueness, optimality and stability, counter-intuitive behaviors
 - ▶ **Systematic design**: scalable price signal, control laws with better stability properties

MAC Reverse Engineering

- Utility of each link depends on transmission probabilities of **all** links
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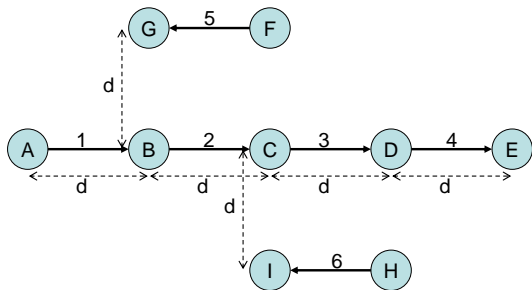
- Questions:
 - ▶ What are users' **utility functions**?
 - ▶ What does the MAC protocol do for the game?
 - ▶ What are the **properties** of the Nash Equilibrium (result of game)?

Different Work

Game to MAC:

- MacKenzie, Wicker 2003
 - Jin, Kesidis 2004
 - Altman et. al. 2005
 - Yuen, Marbach 2005
 - Wang, Krunz, Younis 2006
-
- This is different: **Reverse engineering**
 - Discover, **not** impose, utility and game

Sample Network



Persistence Probabilistic Model of Protocol

- Protocol parameters:

- ▶ p_i^{\max} : Maximum persistent probability (**politeness**)
- ▶ p_i^{\min} : Minimum persistent probability
- ▶ $\beta_i \in (0, 1)$: Backoff multiplier

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- ▶ p_l^{\max} : Maximum persistent probability (**politeness**)
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- ▶ $\beta_l \in (0, 1)$: Backoff multiplier

- Protocol description: link l transmits with a probability p_l

- ▶ If **success** (no collision), update $p_l = p_l^{\max}$
- ▶ If **failure** (collision), update $p_l = \max\{p_l^{\min}, \beta_l p_l\}$, where $0 < \beta_l < 1$

Persistence Probability Update

Persistence Probability **Stochastic** Update

$$\begin{aligned} p_l(t+1) = & \max\{p_l^{\min}, p_l^{\max} \mathbf{1}_{\{T_l(t)=1\}} \mathbf{1}_{\{C_l(t)=0\}} \\ & + \beta_l p_l(t) \mathbf{1}_{\{T_l(t)=1\}} \mathbf{1}_{\{C_l(t)=1\}} \\ & + p_l(t) \mathbf{1}_{\{T_l(t)=0\}}\} \end{aligned}$$

- $T_l(t)$: link l **transmits** at time slot t

$$\text{Prob}\{T_l(t) = 1 | \mathbf{p}(t)\} = p_l(t)$$

- $C_l(t)$: at least one link that can cause **collision** to link l transmits at t

$$\text{Prob}\{C_l(t) = 1 | \mathbf{p}(t)\} = 1 - \prod_{n \in L_{to}(l)} (1 - p_n(t))$$

Deterministic Approximation

Persistence Probability Update: Deterministic Approximation

$$\begin{aligned} p_l(t+1) = & \max\{p_l^{\min}, p_l^{\max} p_l(t) \prod_{n \in L_{to}(l)} (1 - p_n(t)) \\ & + \beta_l p_l(t) p_l(t) \left(1 - \prod_{n \in L_{to}(l)} (1 - p_n(t)) \right) \\ & + p_l(t)(1 - p_l(t))\}, \end{aligned}$$

- Links are playing a game
- Each link l tries to maximize its utility U_l based on other links' current transmission probabilities

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- Links are playing a game
- Each link l tries to maximize its utility U_l based on other links' current transmission probabilities
- **Key question:** what is the game model?

MAC Game

Definition

A MAC game is $[E, \times_{I \in E} A_I, \{U_I\}_{I \in E}]$

- E : set of players (links)
- $A_I = \{p_I | p_I^{\min} \leq p_I \leq p_I^{\max}\}$: action set of link I
- U_I : utility function of link I

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Theorem

Utility function turns out to be *expected net reward*:

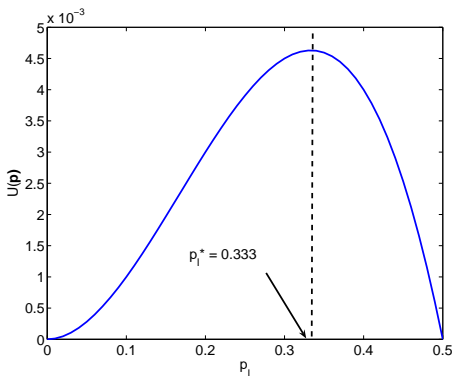
$$U_I(\mathbf{p}) = R(p_I)S(\mathbf{p}) - C(p_I)F(\mathbf{p})$$

where $R(p_I)$ is *reward* for transmission success,

$S(\mathbf{p})$ is *probability* of transmission success,

$C(p_I)$ is *cost* for transmission failure,

$F(\mathbf{p})$ is *probability* of transmission failure.



Dependence of a utility function on its own persistence probability

$$(\beta_l = 0.5, p_l^{max} = 0.5, \text{ and } \prod_{n \in L_{to}(l)} (1 - p_n) = 0.5)$$

Interpretation of MAC protocol: a stochastic subgradient algorithm

- Is it a gradient-based maximization of $U_I(\mathbf{p})$ over p_I ?
 - ▶ No, that requires explicit message passing among links

Interpretation of MAC protocol: a stochastic subgradient algorithm

- Is it a gradient-based maximization of $U_l(\mathbf{p})$ over p_l ?
 - ▶ No, that requires explicit message passing among links
- MAC maximizes U_l using **stochastic subgradient** ascent method (using only local information on success and collision):

$$p_l(t+1) = \max\{p_l^{min}, p_l(t) + v_l(t)\}$$

where

$$E\{v_l(t)|\mathbf{p}(t)\} = \frac{\partial U_l(\mathbf{p})}{\partial p_l} \Big|_{\mathbf{p}=\mathbf{p}(t)}$$

Existence of Nash Equilibrium

- Assume all links have the same $p^{\max} < 1$ and $p^{\min} = 0$

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Theorem

There always exists a Nash equilibrium in the MAC game, which can be characterized by

$$p_l^* = \frac{p^{\max} \prod_{n \in L_{to}(l)} (1 - p_n^*)}{1 - \beta_l (1 - \prod_{n \in L_{to}(l)} (1 - p_n^*))}, \quad \forall l.$$

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- **Proof:** Fixed point theorem in the compact strategy interval.
- The Nash equilibrium **may not be unique** in general.

Uniqueness and Convergence of Nash Equilibrium

- Define the best response function as

$$p_i^*(t+1) = \arg \max_{p_i^{\min} \leq p_i \leq p_i^{\max}} U_i(p_i, p_{-i}^*(t))$$

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Define maximum interference degree as $K = \max_I |L_{to}(I)|$, then if

$$\frac{p^{\max} K}{4\beta(1 - p^{\max})} < 1$$

- The Nash equilibrium is *unique*
- The best response iteration *globally* converges to the unique equilibrium

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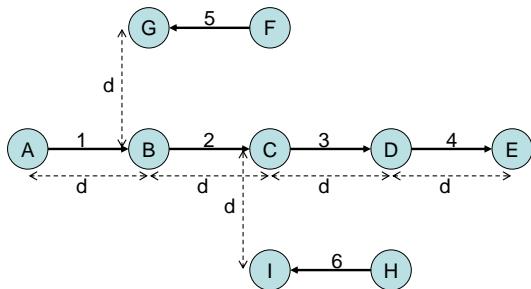
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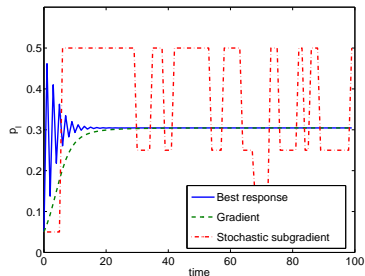
- The Nash equilibrium is *unique*
- The best response iteration *globally* converges to the unique equilibrium
- Proof:** Properly bounding the matrix norm of the Jacobian. Show it is a contraction mapping.
- How polite is necessary? Critical value: p_c^{\max}

Network Topology

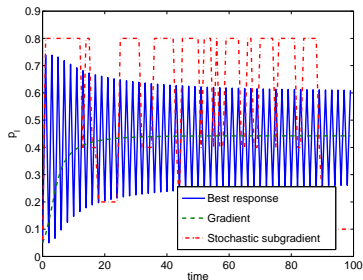


A network with Six Links

Convergence



$$p^{\max} = 0.5$$



$$p^{\max} = 0.8$$

Comparison of trajectories of $p_i(t)$ in the network

Summary

- **Topic:** reverse engineering of MAC protocol
- **Key idea:** a non-cooperative game model
- **Results:**
 - ▶ Utility function discovered: expected net reward
 - ▶ Current MAC algorithm corresponds to stochastic subgradient update
 - ▶ NE always exists. It is unique and stable if the protocol is polite enough and backoff smooth enough
- **Implications:**
 - ▶ Reverse engineering leads to deeper understanding of existing protocols
 - ▶ Insights are helpful for better forward engineering