

# An explicit solution of a two user dynamic team

Aditya Mahajan

Dept of ECE  
McGill University

September 30, 2010  
Allerton

# Is dynamic team theory useful?

# Is dynamic team theory useful?

Hlyuchj and Gallager, 1981

Although the notion of a dynamic team problem has been around for over 25 years, the class of problems is of sufficient complexity that little progress has been made toward a general solution technique or even in finding general properties of optimal solutions.

Hence its value to the multi-access problem does not go much beyond a conceptual level.

# Is dynamic team theory useful?

## Hlyuchj and Gallager, 1981

Although the notion of a dynamic team problem has been around for over 25 years, the class of problems is of sufficient complexity that little progress has been made toward a general solution technique or even in finding general properties of optimal solutions.

Hence its value to the multi-access problem does not go much beyond a conceptual level.

## What is the state of the art after 30 years?

Have we made any progress toward a general solution technique to be of any value to the problem that Hlyuchj and Gallager were interested in?

# Problem Setup: Two-user multiple access broadcast

## Two-users with single slot buffer

- $x_{i,t} \in \{0, 1\}$  : # packets in buffer
- $a_{i,t} \in \{0, 1\}$  : # new packet arrivals

$$a_{i,t} \sim \text{Ber}(p_i)$$

- $u_{i,t} \in \{0, 1\}$  : # transmitted packets



# Problem Setup: Two-user multiple access broadcast

## Two-users with single slot buffer

- $x_{i,t} \in \{0, 1\}$  : # packets in buffer
- $a_{i,t} \in \{0, 1\}$  : # new packet arrivals

$$a_{i,t} \sim \text{Ber}(p_i)$$

- $u_{i,t} \in \{0, 1\}$  : # transmitted packets



## Multiple access channel

Indicator of successful decoding:  $z_t = u_{1,t} \oplus u_{2,t}$

$$x_{i,t+1} = (x_{i,t} - u_{i,t}z_t) \vee a_{i,t}$$

# Problem Setup: Two-user multiple access broadcast

## Two-users with single slot buffer

- $x_{i,t} \in \{0, 1\}$  : # packets in buffer
- $a_{i,t} \in \{0, 1\}$  : # new packet arrivals

$$a_{i,t} \sim \text{Ber}(p_i)$$

- $u_{i,t} \in \{0, 1\}$  : # transmitted packets



## Multiple access channel

Indicator of successful decoding:  $z_t = u_{1,t} \oplus u_{2,t}$

$$x_{i,t+1} = (x_{i,t} - u_{i,t}z_t) \vee a_{i,t}$$

## Broadcast channel

$z_t$  is available to the users after unit delay

# Problem Setup: Two user multiple access broadcast

## Problem (P1)

- **Given:** arrival rates  $p_1$  and  $p_2$
- **Choose:** Transmission policies  $(\mathbf{g}_1, \mathbf{g}_2)$  where  $\mathbf{g}_i = (g_{i,1}, g_{i,2}, \dots, g_{i,T})$  and

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

- **Objective:** Maximize

$$\mathbb{E}^{\mathbf{g}_1, \mathbf{g}_2} \left\{ \sum_{t=1}^T u_{1,t} \oplus u_{2,t} \right\} \quad \text{or} \quad \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}^{\mathbf{g}_1, \mathbf{g}_2} \left\{ \sum_{t=1}^T u_{1,t} \oplus u_{2,t} \right\}$$





Simplest canonical problem in multi-access networks.

- **Slotted ALOHA and variants:** Provide approximately optimal performance when the number of users is large. Huge literature . . .
- **Collision incurs a cost but does not affect the dynamics**  
Schoute, 76, Walrand Varaiya, 79,
- We are interested in the two-user problem in which collision affects the dynamics

# Literature overview for Problem (P1)

Symmetric arrival rates

Asymmetric arrival rates

# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  
  
  
  
  
  
  
  
  
  
- Ooi Wornell 96:

## Asymmetric arrival rates

# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  - Restrict to **window protocols**
  - Analytic soln.
  - **lower bound**
- Ooi Wornell 96:

## Asymmetric arrival rates

# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  - Restrict to **window protocols**
  - Analytic soln.
  - **lower bound**
- Ooi Wornell 96:
  - Genie reveals buffer state after a delay
  - Numerical soln
  - **upper bound**

## Asymmetric arrival rates

# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  - Analytic
  - lower bound
- Ooi Wornell 96:
  - Numerical
  - upper bound

## Asymmetric arrival rates

lower and upper bounds match

# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  - Analytic
  - lower bound
- Ooi Wornell 96:
  - Numerical
  - upper bound

## Asymmetric arrival rates

- Lot of AI literature ...
- Hansen et. al. 04
  
- Bernstein et. al. 05
  
- Szer Charpillet 06

lower and upper bounds match

# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  - Analytic
  - lower bound
- Ooi Wornell 96:
  - Numerical
  - upper bound

lower and upper bounds match

## Asymmetric arrival rates

- Lot of AI literature ...
- Hansen et. al. 04
  - Numerical algorithm to find optimal soln
  - Out of memory for  $T=5$
- Bernstein et. al. 05
  - Heuristic algorithm
  - Controller for size=8
- Szer Charpillet 06
  - Approx. algorithm
  - Out of memory for  $T=5$



# Literature overview for Problem (P1)

## Symmetric arrival rates

- Hlyuchj Gallager 81:
  - Analytic
  - lower bound
- Ooi Wornell 96:
  - Numerical
  - upper bound

lower and upper bounds match

## Asymmetric arrival rates

- Lot of AI literature ...

Approx algorithms ...  
but can only solve the system  
until  $T = 4$

# Questions?

Symmetric arrival rates

Asymmetric arrival rates

## Symmetric arrival rates

- Optimal soln is known
- The proof is numerical
- Can we provide an analytic proof?

## Asymmetric arrival rates

# Questions?

## Symmetric arrival rates

- Optimal soln is known
- The proof is numerical
- Can we provide an analytic proof?

## Asymmetric arrival rates

- Approx algorithms only work for small horizon
- Can we find algorithms that can solve large or infinite horizon problem?

# Contributions of this paper

- Provide a dynamic programming decomposition
- The DP has countable state space and finite action space.  
**Easy to use existing algorithms** to find numerical solution for large or infinite horizon setups
- For symmetric arrival rates, **find an analytic soln to the DP.**

# Problem Setup: Two user multiple access broadcast

## Problem (P1)

- **Given:** arrival rates  $p_1$  and  $p_2$
- **Choose:** Transmission policies  $(\mathbf{g}_1, \mathbf{g}_2)$  where  $\mathbf{g}_i = (g_{i,1}, g_{i,2}, \dots, g_{i,T})$  and

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

- **Objective:** Maximize

$$\mathbb{E}^{\mathbf{g}_1, \mathbf{g}_2} \left\{ \sum_{t=1}^T u_{1,t} \oplus u_{2,t} \right\} \quad \text{or} \quad \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}^{\mathbf{g}_1, \mathbf{g}_2} \left\{ \sum_{t=1}^T u_{1,t} \oplus u_{2,t} \right\}$$



## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## Feedback $\equiv$ control sharing

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$



## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## $x_{i,1:t-1}$ is redundant

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Feedback $\equiv$ control sharing

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## $x_{i,1:t-1}$ is redundant

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Feedback $\equiv$ control sharing

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Suff statistic for common info

$$u_{i,t} = g_{i,t}(x_{i,t}, \pi_{1,t}, \pi_{2,t})$$

where

$$\pi_{i,t} = \Pr(x_{i,t} = 1 | u_{1,1:t-1}, u_{2,1:t-1})$$

# Solution Outline (cont)

## Dynamic Program

$$V_{T+1}(\pi_1, \pi_2) = 0$$

and for  $t = T, T - 1, \dots, 1$

$$V_t(\pi_1, \pi_2) = \max\{W_{10,t}(\pi_1, \pi_2), W_{01,t}(\pi_1, \pi_2), W_{11,t}(\pi_1, \pi_2)\}$$

# Solution Outline (cont)

## Dynamic Program

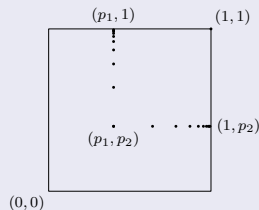
$$V_{T+1}(\pi_1, \pi_2) = 0$$

and for  $t = T, T - 1, \dots, 1$

$$V_t(\pi_1, \pi_2) = \max\{W_{10,t}(\pi_1, \pi_2), W_{01,t}(\pi_1, \pi_2), W_{11,t}(\pi_1, \pi_2)\}$$

## Reachability Analysis

The reachable set of  $(\pi_1, \pi_2)$  is countable.



## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## $x_{i,1:t-1}$ is redundant

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Feedback $\equiv$ control sharing

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Suff statistic for common info

$$u_{i,t} = g_{i,t}(x_{i,t}, \pi_{1,t}, \pi_{2,t})$$

where

$$\pi_{i,t} = \Pr(x_{i,t} = 1 | u_{1,1:t-1}, u_{2,1:t-1})$$

# Feedback $\equiv$ control sharing

- $z_t = u_{1,t} \oplus u_{2,t}$
- Thus,

$$u_{1,t} = z_t \oplus u_{2,t} \quad \text{and} \quad u_{2,t} = z_t \oplus u_{1,t}$$

- $z_t = u_{1,t} \oplus u_{2,t}$
- Thus,

$$u_{1,t} = z_t \oplus u_{2,t} \quad \text{and} \quad u_{2,t} = z_t \oplus u_{1,t}$$

- Hence,

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1}, z_{1:t-1})$$

- $z_t = u_{1,t} \oplus u_{2,t}$
- Thus,

$$u_{1,t} = z_t \oplus u_{2,t} \quad \text{and} \quad u_{2,t} = z_t \oplus u_{1,t}$$

- Hence,

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1}, z_{1:t-1})$$

- Since  $z_t = u_{1,t} \oplus u_{2,t}$ ,

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$



## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## $x_{i,1:t-1}$ is redundant

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Feedback $\equiv$ control sharing

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Suff statistic for common info

$$u_{i,t} = g_{i,t}(x_{i,t}, \pi_{1,t}, \pi_{2,t})$$

where

$$\pi_{i,t} = \Pr(x_{i,t} = 1 | u_{1,1:t-1}, u_{2,1:t-1})$$

## $x_{i,1:t-1}$ is redundant

- Arbitrarily fix the transmission policy of user 2
- $(x_{1,t}, u_{1,1:t-1}, u_{2,1:t-1})$  is a **controlled Markov chain** with control action  $u_{1,t}$

## $x_{i,1:t-1}$ is redundant

- Arbitrarily fix the transmission policy of user 2
- $(x_{1,t}, u_{1,1:t-1}, u_{2,1:t-1})$  is a **controlled Markov chain** with control action  $u_{1,t}$
- Conditioned on the controls, the dynamics are independent

$$x_{1,1:t} \leftrightarrow (u_{1,1:t-1}, u_{2,1:t-1}) \leftrightarrow x_{2,1:t}$$

## $x_{i,1:t-1}$ is redundant

- Arbitrarily fix the transmission policy of user 2
- $(x_{1,t}, u_{1,1:t-1}, u_{2,1:t-1})$  is a **controlled Markov chain** with control action  $u_{1,t}$
- Conditioned on the controls, the dynamics are independent

$$x_{1,1:t} \leftrightarrow (u_{1,1:t-1}, u_{2,1:t-1}) \leftrightarrow x_{2,1:t}$$

- Thus, conditional expected reward

$$\begin{aligned} \mathbb{E}[u_{1,t} \oplus u_{2,t} | x_{1,1:t}, u_{1,1:t-1}, u_{2,1:t-}] \\ = \mathbb{E}[u_{1,t} \oplus u_{2,t} | x_{1,t}, u_{1,1:t-1}, u_{2,1:t-}] \end{aligned}$$

## $x_{i,1:t-1}$ is redundant

- Arbitrarily fix the transmission policy of user 2
- $(x_{1,t}, u_{1,1:t-1}, u_{2,1:t-1})$  is a **controlled Markov chain** with control action  $u_{1,t}$
- Conditioned on the controls, the dynamics are independent

$$x_{1,1:t} \leftrightarrow (u_{1,1:t-1}, u_{2,1:t-1}) \leftrightarrow x_{2,1:t}$$

- Thus, conditional expected reward

$$\begin{aligned} \mathbb{E}[u_{1,t} \oplus u_{2,t} | x_{1,1:t}, u_{1,1:t-1}, u_{2,1:t-1}] \\ = \mathbb{E}[u_{1,t} \oplus u_{2,t} | x_{1,t}, u_{1,1:t-1}, u_{2,1:t-1}] \end{aligned}$$

- Thus,

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Transmission policy

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{i,1:t-1}, z_{1:t-1})$$

## $x_{i,1:t-1}$ is redundant

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Feedback $\equiv$ control sharing

$$u_{i,t} = g_{i,t}(x_{i,1:t}, u_{1,1:t-1}, u_{2,1:t-1})$$

## Suff statistic for common info

$$u_{i,t} = g_{i,t}(x_{i,t}, \pi_{1,t}, \pi_{2,t})$$

where

$$\pi_{i,t} = \Pr(x_{i,t} = 1 | u_{1,1:t-1}, u_{2,1:t-1})$$

# Sufficient statistic for common information

$$u_{i,t} = g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1})$$

- Common information:  $(u_{1,1:t-1}, u_{2,1:t-1})$
- Private information:  $x_{i,t}$

A special case of Mahajan, Nayyar, Teneketzis, 2008

Same solution approach (using the notion of a coordinator) applies

# Sufficient statistic for common information (cont)

## Coordinator of the two users

- Observation of coordinator: common information

$$(u_{1,1:t-1}, u_{2,1:t-1})$$

- Action of the coordinator: **partial functions**  $(\gamma_{1,t}, \gamma_{2,t})$  s.t.

$$u_{i,t} = \gamma_{i,t}(x_{i,t})$$



# Sufficient statistic for common information (cont)

## Coordinator of the two users

- Observation of coordinator: common information

$$(u_{1,1:t-1}, u_{2,1:t-1})$$

- Action of the coordinator: **partial functions**  $(\gamma_{1,t}, \gamma_{2,t})$  s.t.

$$u_{i,t} = \gamma_{i,t}(x_{i,t})$$

- For ease of notation, let  $\varphi_{i,t} = \gamma_{i,t}(1)$ . Then

$$u_{i,t} = \varphi_{i,t} x_{i,t}$$

# Sufficient statistic for common information (cont)

## Coordinator of the two users

- Observation of coordinator: common information

$$(u_{1,1:t-1}, u_{2,1:t-1})$$

- Action of the coordinator: **partial functions**  $(\gamma_{1,t}, \gamma_{2,t})$  s.t.

$$u_{i,t} = \gamma_{i,t}(x_{i,t})$$

- For ease of notation, let  $\varphi_{i,t} = \gamma_{i,t}(1)$ . Then

$$u_{i,t} = \varphi_{i,t} x_{i,t}$$

- Think of  $(\varphi_{1,t}, \varphi_{2,t})$  as the control action of the coordinator.

## Problem (P2)

- **Given:** arrival rates  $p_1$  and  $p_2$
- **Choose:** Coordination policy  $\mathbf{h} = (h_1, h_2, \dots, h_T)$  where

$$(\varphi_{1,t}, \varphi_{2,t}) = h_t(u_{1,1:t-1}, u_{2,1:t-1}, \varphi_{1,1:t-1}, \varphi_{2,1:t-1})$$

- **Objective:** Maximize

$$\mathbb{E}^{\mathbf{h}} \left\{ \sum_{t=1}^T u_{1,t} \oplus u_{2,t} \right\} \quad \text{or} \quad \lim_{T \rightarrow \infty} \frac{1}{T} \mathbb{E}^{\mathbf{h}} \left\{ \sum_{t=1}^T u_{1,t} \oplus u_{2,t} \right\}$$

# Sufficient statistic for common information (cont)

## Proposition

Problem (P1) and (P2) are equivalent.

## Proof.

- Any transmission policy  $(\mathbf{g}_1, \mathbf{g}_2)$  for (P1) can be implemented in (P2) by choosing

$$\varphi_{i,t} = g_{i,t}(\mathbf{1}, u_{1,1:t-1}, u_{2,1:t-1})$$

resulting in identical realization of all system variables.

- Any coordination policy  $\mathbf{h}$  for (P2) can be implemented in (P1) by choosing

$$g_{i,t}(x_{i,t}, u_{1,1:t-1}, u_{2,1:t-1}) = \varphi_{i,t} x_{i,t}$$

where  $\varphi_{i,t}$  is recursively chosen according to

$$(\varphi_{1,t}, \varphi_{2,t}) = h_t(u_{1,1:t-1}, u_{2,1:t-1}, \varphi_{1,1:t-1}, \varphi_{2,1:t-1})$$

# Sufficient statistic for common information (cont)

## Definition

$$\pi_{i,t} = \Pr \left( x_{i,t} = 1 \mid \begin{array}{l} u_{1,1:t-1}, u_{2,1:t-1} \\ \varphi_{1,1:t-1}, \varphi_{2,1:t-1} \end{array} \right)$$

## Proposition

In (P2), restricting attention to coordination policies of the form

$$(\varphi_{1,t}, \varphi_{2,t}) = h_t(\pi_{1,t}, \pi_{2,t})$$

is without loss. Therefore, in (P1) restricting attention to transmission policies of the form

$$u_{i,t} = g_{i,t}(x_{i,t}, \pi_{1,t}, \pi_{2,t})$$

is without loss.

# Sufficient statistic for common information (cont)

## Proof.

- $(\pi_{1,t}, \pi_{2,t})$  is a controlled Markov process with control action  $(\varphi_{1,t}, \varphi_{2,t})$ .
- Expected conditional reward

$$\begin{aligned}\mathbb{E}[u_{1,t} \oplus u_{2,t} | u_{1,1:t-1}, u_{2,1:t-1}, \varphi_{1,1:t}, \varphi_{2,1:t}] \\ &= \pi_{1,t} \varphi_{1,t} (1 - \pi_{2,t} \varphi_{2,t}) + (1 - \pi_{1,t} \varphi_{1,t}) \pi_{2,t} \varphi_{2,t} \\ &= \mathbb{E}[u_{1,t} \oplus u_{2,t} | \pi_{1,t}, \pi_{2,t}, \varphi_{1,t}, \varphi_{2,t}]\end{aligned}$$



# Solution Outline (cont)

## Dynamic Program

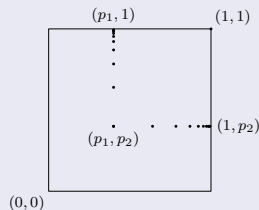
$$V_{T+1}(\pi_1, \pi_2) = 0$$

and for  $t = T, T - 1, \dots, 1$

$$V_t(\pi_1, \pi_2) = \max\{W_{10,t}(\pi_1, \pi_2), W_{01,t}(\pi_1, \pi_2), W_{11,t}(\pi_1, \pi_2)\}$$

## Reachability Analysis

The reachable set of  $(\pi_1, \pi_2)$  is countable.



- DP follows immediately from the fact that  $(\pi_{1,t}, \pi_{2,t})$  is a controlled Markov process.
- By the same argument, the DP naturally extends to infinite horizon setup.



- Let  $A_i$  be an operator from  $[0, 1]$  to  $[0, 1]$  such that for any  $\pi \in [0, 1]$

$$A_i\pi = 1 - (1 - p_i)(1 - \pi)$$

## Evolution of info state

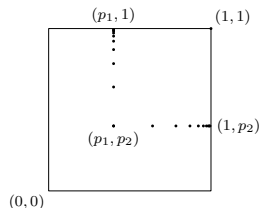
- When  $(\varphi_{1,t}, \varphi_{2,t}) = (0, 0)$ ,  $(\pi_{1,t+1}, \pi_{2,t+1}) = (A_1\pi_{1,t}, A_2\pi_{2,t})$ .
- When  $(\varphi_{1,t}, \varphi_{2,t}) = (1, 0)$ ,  $(\pi_{1,t+1}, \pi_{2,t+1}) = (p_1, A_2\pi_{2,t})$ .
- When  $(\varphi_{1,t}, \varphi_{2,t}) = (0, 1)$ ,  $(\pi_{1,t+1}, \pi_{2,t+1}) = (A_1\pi_{1,t}, p_2)$ .
- When  $(\varphi_{1,t}, \varphi_{2,t}) = (1, 1)$ ,  
$$(\pi_{1,t+1}, \pi_{2,t+1}) = \begin{cases} (1, 1) & \text{if } x_{1,t} = x_{2,t} = 1 \\ (p_1, p_2) & \text{otherwise} \end{cases}$$

# Reachability Analysis (cont)

## Reachable Set

Suppose the system starts in state  $(\pi_1, \pi_2) = (p_1, p_2)$ . Then the reachable set of  $(\pi_1, \pi_2)$  is

$$S = \{(1, 1), (p_1, 1), (1, p_2), (p_1, p_2)\} \\ \cup \{(A_1^n p_1, p_2), (p_1, A_2^n p_2), : n \in \mathbb{N}\}$$



## Reachability Analysis (cont)

- The reachable set of  $(\pi_{1,t}, \pi_{2,t})$  is countable.
- Thus, the infinite horizon DP has countable state space and finite action space
- Standard techniques to numerically solve such DP (e.g. Sennot, 97 , Leizarowitz Schwartz, 07)
- Contrast this with earlier attempt to obtain a numerical solution for this problem.

- Optimal coordination policy is symmetric  $h(\pi_1, \pi_2) = h(\pi_2, \pi_1)$

## Some definitions

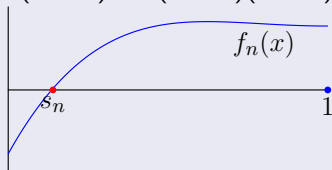
- Let  $\tau \approx 0.38196$  be the root of  $x = (1 - x)^2$  that lies in  $[0, 1]$ .

# Symmetric arrivals

- Optimal coordination policy is symmetric  $h(\pi_1, \pi_2) = h(\pi_2, \pi_1)$

## Some definitions

- Let  $\tau \approx 0.38196$  be the root of  $x = (1 - x)^2$  that lies in  $[0, 1]$ .
- Let  $f_n(x) = 1 + (1 - x)^2 - (3 + x)(1 - x)^{n+1}$



and  $s_n$  denote the root of  $f_n(x)$  that is between  $[0, 1]$ .

- $s_0 > \tau > s_1 > s_2 > \dots > 0$

## Theorem

An optimal policy of the infinite horizon variant of (P2) is:

- **round-robin policy** for  $p \geq \tau$

$$h^*(\pi_1, \pi_2) = \begin{cases} (1, 0) & \text{if } \pi_1 > \pi_2, \\ (0, 1) & \text{if } \pi_1 < \pi_2, \\ (1, 0) \text{ or } (0, 1) & \text{if } \pi_1 = \pi_2. \end{cases}$$

- **transmit if you have a packet policy** for  $p < \tau$

$$h^*(\pi_1, \pi_2) = \begin{cases} (1, 1) & \text{if } \pi_1 \leq A^m p, \pi_2 \leq A^m p, \\ (1, 0) & \text{if } \pi_1 > \pi_2, \pi_1 > A^m p \\ (0, 1) & \text{if } \pi_1 < \pi_2, \pi_2 > A^m p \\ (1, 0) \text{ or } (0, 1) & \text{if } \pi_1 = \pi_2 = 1. \end{cases}$$

where  $m$  is s.t.  $s_{m+1} \leq p \leq s_m$ .

## Theorem

The average reward per unit time for the infinite horizon variant of (P2) is

$$J^* = \begin{cases} p[1 - (2p^2 - 1)/D(p)] & \text{if } p \leq s_1, \\ (1 - \bar{p}^2) & \text{if } s_1 \leq p; \end{cases}$$

where  $\bar{p} = 1 - p$  and  $D(p) = 1 + p^2 + p^3$ .

## Proof

Guess the form of the value function and verify!

1. When  $p \geq s_1$ ,

$$v(p, A^n p) = v(A^n p, p) = (1 - \bar{p}^{n+1}), \quad n > 1$$

$$v(p, 1) = v(1, p) = 1,$$

$$v(1, 1) = (1 + \bar{p}^2),$$

$$v(p, p) = p$$



## Proof (cont)

Guess the form of the value function and verify!

2. When  $s_{m+1} \leq p < s_m$ ,  $m \in \mathbb{N}$

$$v(p, 1) = v(1, p) = p[1 - f_0(p)/D(p)],$$

$$v(1, 1) = 1,$$

$$v(p, p) = f_1(p)/D(p),$$

$$v(A^n p, p) = v(p, A^n p) = \begin{cases} c_*(n) & \text{if } n \leq m, \\ c^*(n) & \text{if } n > m \end{cases}$$

where

$$c_*(n) = \frac{\bar{p}}{p}(1 - \bar{p}^n)J^* + \bar{p}^{n+1} - \bar{p} + v(p, p),$$

$$c^*(n) = (1 - \bar{p}^{n+1}) + c_*(1) - v(1, p)$$

## Proof

Guess the form of the value function and verify!

- Rest is just a matter of elementary (but tedious) algebra.
- The important point is that once we have a dynamic program, optimality of a particular policy can be checked systematically.

## Proof

Guess the form of the value function and verify!

- Rest is just a matter of elementary (but tedious) algebra.
- The important point is that once we have a dynamic program, optimality of a particular policy can be checked systematically.
- We also need to guess the differential reward functions for the non-optimal actions. In general, this can be difficult. But, we exploit the symmetry and the fact that state space is countable.

## Contributions

- An interesting example of two-user dynamic team that can be solved explicitly.
- For symmetric arrivals, identified the optimal policy analytically. The previous proof of optimality involved numerically solving a genie aided upper bound.
- For asymmetric arrivals, identified a DP with countable state space and finite action space. Earlier attempts for a numerical solution could only solve finite horizon problems with  $T = 4$ .

## Future work

- We are missing a structural result:  
Each user gets a transmission opportunity  $\varphi_{i,t} = 1$ , at least once in two consecutive time slots
- The optimal policy satisfies this property.
- If we can prove this upfront, the DP will be much simpler (finite state and finite action spaces).

Thank You