

# Cooperation in Wireless Ad Hoc Networks

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# Introduction

## Wireless Ad Hoc Network

- ▶ Nodes can forward traffic for one another
- ▶ Energy constraints, like battery life, prevent nodes from forwarding all traffic
- ▶ Assuming that nodes are rational, how can traffic be routed to minimize throughput and maximize node lifetime?

# Wireless Ad Hoc Network Model

- ▶ A finite set  $N$  of nodes(laptops, PDAs, etc)
- ▶ Power constraints on each node
- ▶ One session per time slot, one time slot per session
- ▶ Trade-off between routing traffic and battery-life
- ▶ Nodes route in a manner that is selfish, rational

Normalized Acceptance Rate(NAR)=

$$\frac{\text{successful relay requests made}}{\text{relay requests made}}$$

# Contributions

- ▶ Assumes selfish nodes with known energy constraints and lifetime expectation
- ▶ Determines feasible set of NARs
- ▶ Identifies set of strategies that obtains a Pareto optimal Nash Equilibrium
  - ▶ Nash Equilibrium: No change in an individual node's strategy will improve it's NAR
  - ▶ Pareto optimal: No change in strategy will improve one node's NAR without decreasing another's
- ▶ GTFT algorithm that attains Nash equilibrium and Pareto optimality

# Wireless Ad Hoc Network Model

- ▶  $N$  nodes distributed among  $K$  classes
- ▶ Nodes in class  $i$  have energy constraint  $E_i$  and lifetime expectation  $L_i$
- ▶ Average power constraint  $\rho_i = E_i/L_i$  and  $\rho_1 > \rho_2 > \dots > \rho_k$
- ▶ In each timeslot, source chosen from uniform distribution on nodes
- ▶ Maximum number of relays  $M$  to reach destination
- ▶ Distribution on the number of relays used,  $q(l)$ .  $q(0) = 0$ , i.e. no direct transmissions.

# Wireless Ad Hoc Network Model

- ▶ Relays randomly and uniformly selected from non-source nodes
- ▶ Relays accept or reject with ACKs
- ▶ If a single relay reject, session is blocked
- ▶ A session is type  $j$  if at least one node belongs to class  $j$  and all other nodes belong to class less than or equal to  $j$ .
- ▶ Node with smallest power constraint dominates a session's interaction
- ▶ Energy in transmit mode dominates energy for receiving and processing, so only transmission energy is considered

# Wireless Ad Hoc Network Model

$B_h^j(k)$  number of relay requests made by node  $h$  for session type  $j$  until time  $k$

$A_h^j(k)$  number of relay requests made by node  $h$  for session type  $j$  until time  $k$  that have been accepted

$$\phi_h^j(k) = A_h^j(k)/B_h^j(k)$$

$D_h^j(k)$  number of relay requests made to node  $h$  for session type  $j$  until time  $k$

$C_h^j(k)$  number of relay requests made to node  $h$  for session type  $j$  until time  $k$  that have been accepted

$$\psi_h^j(k) = C_h^j(k)/D_h^j(k)$$

# Rational and Pareto Optimal Operating Point

$$\lim_{k \rightarrow \infty} \phi_h^j(k) = \lim_{k \rightarrow \infty} A_h^i(k)/B_h^i(k) = \text{NAR}$$

This is the throughput for node  $h$  and session type  $j$   
NAR values

1. Energy constraints on each node met
2. Pareto optimal values attained
3. All rational agents find allocation fair

# Rational and Pareto Optimal Operating Point

Average energy per session spent by a node  $p$  as a source

$$e_{pj}^{(s)} = \frac{1}{N} \times \text{NAR}$$
$$= \frac{1}{N} \sum_{l=1}^M \sum_{h_1, \dots, h_j} q(l) \Gamma(l; h_1, \dots, h_j) \tau_{1j}^{h_1} \dots \tau_{jj}^{h_j}$$

- ▶  $1/N$  is the probability the node  $p$  is the source
- ▶  $\Gamma(l; h_1, \dots, h_K)$  is a multivariate probability function conditioned on the fact that the session belongs to type  $j$  with  $l$  relays.  $h_i$  refers to the number of relays of class  $i$  participating in the session
- ▶  $\tau_{1j}^{h_1} \dots \tau_{jj}^{h_j}$  represents the probability that all the relay nodes accept the request

# Rational and Pareto Optimal Operating Point

Average energy per session spent by a node  $p$  as a relay

$$e_{pj}^{(r)} = \frac{1}{N} \sum_{l=1}^M l q(l) \sum_{h_1, \dots, h_j} \Gamma(l-1; h_1, \dots, h_j) \tau_{1j}^{h_1} \dots \tau_{jj}^{h_j} \tau_{class(p)j}$$

The feasible region for the  $\tau_{ij}$  is defined by

$$\sum_{j=1}^K (e_{pj}^{(s)} + e_{pj}^{(r)}) \leq \rho_{class(p)} \quad 1 \leq p \leq N$$

$$\tau_{class(p)j} \in [0, 1] \quad 1 \leq j \leq K; 1 \leq p \leq N$$

# Rational and Pareto Optimal Operating Point

- ▶ Feasible points given by the triangular region
- ▶ Pareto optimal values occur on the boundary line between  $(0, 2\rho)$  and  $(2\rho, 0)$
- ▶ Increasing one node's NAR decreases another's NAR
- ▶ By symmetry,  $(\rho, \rho)$  is the unique Pareto optimal Nash equilibrium

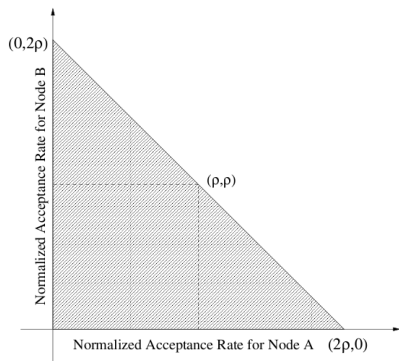


Fig. 1. Feasible Region for  $N = 2$ ,  $K = 1$ ,  $\rho = 0.5$ .

# Rational and Pareto Optimal Operating Point

## Lemma

$$\tau_{ij} = \tau_{ji} \quad 1 \leq i \leq j \leq K$$

**One Class** If all nodes are in the same class, then they will have the same NAR.

**Multiple Classes** If nodes are in multiple classes in the same session, then nodes with higher power have no incentive to accept with a higher probability than the lower powered nodes which are constraining the session.

# Rational and Pareto Optimal Operating Point - Example

For  $M = 1, q(1) = 1$ :

$$e_i^{(s)} = \frac{1}{N(N-1)} \left[ \sum_{k=1}^{i-1} n_k \tau_i + (n_i - 1) \tau_i + \sum_{l=i+1}^K n_l \tau_l \right]$$

$$e_i^{(r)} = \frac{1}{N(N-1)} \left[ \sum_{k=1}^{i-1} n_k \tau_i + (n_i - 1) \tau_i + \sum_{l=i+1}^K n_l \tau_l \right]$$

$$e_i^{(s)} + e_i^{(r)} = \rho_i$$

For  $K = 1$ :

$$2 \cdot \left[ \frac{\tau(N-1)}{N(N-1)} \right] = \rho$$

$$\tau = N\rho/2$$

# Routing Strategy

Given that we can compute the optimal NAR values for all nodes, how can we determine when to except or reject relay requests?

- ▶ Stationary strategies are dominated by the *always-deny* selfish strategy
- ▶ Addaptive history-based strategies can prevent exploitation by selfish nodes
- ▶ Pareto optimal solution should be obtained and maintained across iterative turns

# Prisoner's Dilemma

	Confess	Not Confess
Confess	$(-5, -5)$	$(0, -10)$
Not Confess	$(-10, 0)$	$(-1, -1)$

Single turn Nash Equilibrium (C,C) is Pareto dominated by (NC,NC)

GTFT Pareto optimal solution (NC,NC) is obtained by repeatedly playing the opponents previous move (occasionally cooperating regardless of previous move)

# GTFT

Each node maintains

1.  $\phi_h^i(k)$  ratio of accepted relay requests from  $h$
2.  $\psi_h^j(k)$  ratio of accepted relay requests to  $h$

Reject iff

$$\psi_h^j(k) > \tau_j \text{ or } \phi_h^j(k) < \psi_h^j(k) - \epsilon$$

1. node  $h$  has relayed more traffic for type  $j$  sessions than what it should based on its optimal NAR
2. node  $h$  has relayed more traffic for type  $j$  than other nodes in type  $j$  have relayed for  $h$

$\epsilon$  is the generosity parameter

## m-GTFT

If the number of relays in a path is greater than 1, then the amount of cooperation provided is always greater than the amount of cooperation received. GTFT can be modified using a normalization factor

$$L_{ij} = \frac{\text{Prob}(h \text{ is served in a type } j \text{ session})}{\text{Prob}(h \text{ accepts to relay a type } j \text{ session})}$$

Reject iff

$$\psi_h^j(k) > \tau_j \text{ or } \phi_h^j(k) < L_{ij}\psi_h^j(k) - \epsilon$$

# m-GTFT

## Theorem

Consider a system with  $N$  nodes,  $K$  classes,  $M > 1$ ,  $q(l) > 0$ ,  $l = 1, \dots, M$ ,  $n_i$  nodes in class  $i$ ,  $i = 1, \dots, K$ , and energy constraints  $\rho_1 > \rho_2 > \dots > \rho_K$ . Then,

1. If all nodes except node  $h$  are employing m-GTFT, then  $\limsup_{k \rightarrow \infty} \phi_h^j(k) \leq \tau_j$
2. If all nodes employ m-GTFT, then all  $\phi_h^j(k)$  converge to  $\tau_j$  ( $h = 1, \dots, N$ ;  $i, j = 1, \dots, K$ )

GTFT and m-GTFT constitute a Nash Equilibrium and converge to the rational Pareto optimal operating points.

# Results

## GTFT Validated using simulations

- ▶ 25 nodes, 5 classes of nodes
- ▶  $\rho_1 = 0.03, \rho_2 = 0.025, \rho_3 = 0.02, \rho_4 = 0.015, \rho_5 = 0.01$
- ▶ Single relay sessions

## m-GTFT Validated using simulations

- ▶ 12 nodes, 2 classes of nodes
- ▶  $\rho_1 = 0.03, \rho_2 = 0.015$
- ▶ One and two relay sessions

# Results

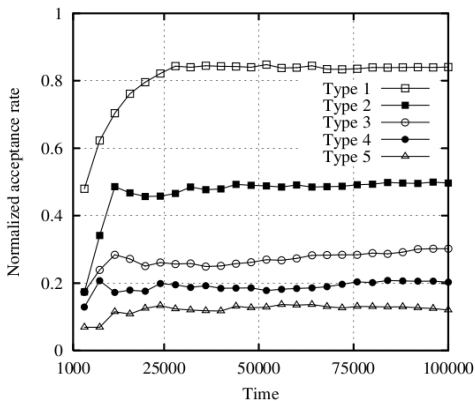


Fig. 2. NAR versus time when  $N = 25$ ,  $K = 5$ ,  $q(1) = 1$ ,  $M = 1$ , and all nodes employ GTFT. NAR values converge to the optimal operating point.

# Results - Generosity

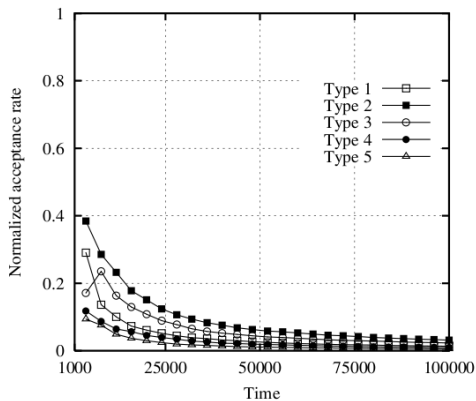


Fig. 3. NAR versus time when  $N = 25$ ,  $K = 5$ ,  $q(1) = 1$ ,  $M = 1$ , all nodes employ GTFT, and  $\epsilon < 0$ . If nodes are not slightly generous ( $\epsilon > 0$ ), GTFT fails to reach the optimal operating point.

# Results - Parasitic Nodes

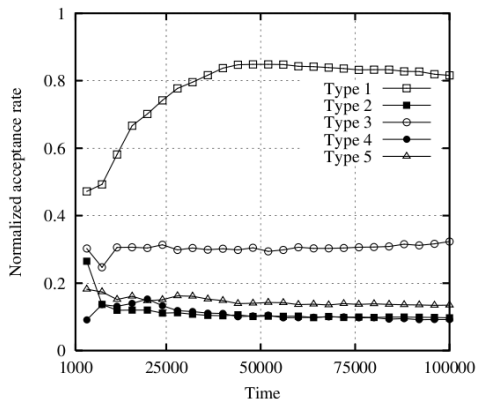


Fig. 4. NAR versus time when  $N = 25$ ,  $K = 5$ ,  $q(1) = 1$ ,  $M = 1$ , and one node in class 2 and one node in class 4 are parasites while all other nodes employ GTFT. Performance of nodes in type 2 and type 4 sessions degrade showing that GTFT prevents parasitic behavior in rational users.

# Results - m-GTFT

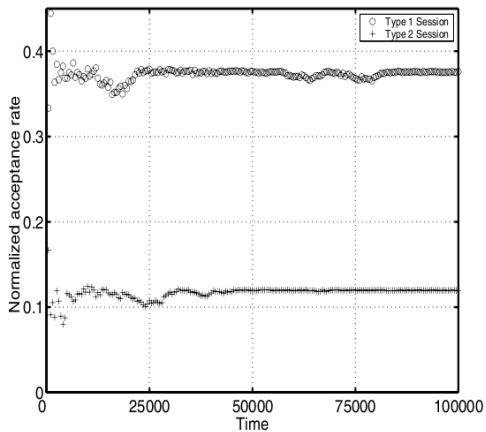


Fig. 5. Convergence of m-GTFT for  $N = 12$ ,  $K = 2$ ,  $\rho_1 = 0.03$ ,  $\rho_2 = 0.015$ ,  $q(1) = q(2) = 0.5$ , and  $M = 2$ .

# Discussion

- ▶ GTFT adapts based on the optimal NAR values, but these must be computed, requiring every node to have knowledge of other nodes and their energy constraints
- ▶ Distributed mechanism necessary to disseminate information to all users
- ▶ Malicious irrational users can reduce the performance of networks by denying all requests, watchdog mechanism necessary to identify such nodes