Adaptive Mixed Bias Resource Allocation for Wireless Mesh Networks

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Overview

- Introduction
- Allocation Schemes
- Mixed Bias
- Adaptive Mixed Bias
- Performance Evaluation
- Results
- Conclusions & Future Work



Introduction

- Resource allocation
 - Focus on access to the medium
 - Could also be applied to other resources such as queues, CPU, frequency
- Wireless Mesh Network
 - Traffic to and from Gateway (GW)
 via Mesh Routers (MR)



Existing Allocation Schemes,

Proportional:

- Resources assigned proportionally to some characteristic
- Danger of starvation with strong proportionality (bias)

Max-min:

 Allocate resources such that we maximize the minimum of some characteristic of the network

Round-robin / Fair:

Resource split evenly



Existing Allocation Schemes

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Reference	Technique	Layers	Goal
 Ernst, 	Combined	MAC, PHY,	Scheduling
Denko	Mixed Bias	NET	
[2] Singh et al	Mixed Bias	MAC, NET	Scheduling
[3] José-	Genetic	PHY	Channel
Revuelta	Algorithm		Assignment
[4] Mahjoub et	Real-time	NET	Routing
al	Search		
	Algorithm		
[5] Choi and	Heuristic,	PHY, MAC	Scheduling with
Choi	Utility based		Service
			Differentiation
[7] Mandal et al	Constraint	PHY	Channel
	based heuristic		Assignment
[8] Beljadid et	Genetic	PHY	Channel
a1	Algorithm,		Assignment
	Tabu Search		
[10] Zhang et al	Diversity	PHY, MAC,	Scheduling, Queue
	weight adaptive	NET	Management
	scheduling		
[11] Hedayati et	Mixed-integer	PHY, MAC	Power, Rate
al	linear		Adaption
	programming		_
[12] Prashanth	Traffic	PHY, MAC,	Joint Scheduling
et al	Prediction	NET	and Routing

See the paper for more details

Mixed Bias

Resource assigned by:

$$R = \frac{\alpha}{c^{\beta_1}} + \frac{1-\alpha}{c^{\beta_2}} \alpha < 1, \beta_1, \beta_2 > 0$$

- α weight of competing biases
- β_1 , β_2 strength of bias
- c characteristic to bias against
- R probability a MR will transmit (Ernst, Denko), (Singh, et al)



Mixed Bias

 With c = distance from GW α,β₁,β₂ determined analytically (for static topology)

 With c = link quality or other characteristics, analysis becomes difficult, dependant on a model The Case for Adaptive Resource Allocation

Resource allocation often rigid

Wireless network conditions change rapidly

- Decisions made from limited information and assumptions
- Often disregard:
 - Interference, Mobility, Congestion, Distance ...



Motivation

- Difficult to determine objective function for complicated multi-hop heterogeneous networks
- Avoid long delays in making decisions
- Introduces autonomous aspect to WMN
- Capitalize on MR abundant abilities



Motivation

- Goals:
 - Maximize packet delivery ratio
 - Minimize end-to-end delay
 - Utility function, combination of two



Related Work

- Scheduling & Resource allocation with defined service levels (Choi et al)
 - Focus is on maximizing profit, not network performance in many solution
 - Service level weak point
- Genetic algorithms applied to channel assignment (José-Revuelta)
- Constraint-based routing (Mahjoub et al)
 - (Heuristic optimization approaches used in wireless often)



- Explore the parameter space of the mixed bias algorithm
- Real time changes to the mixed bias parameters (Tabu Move)
- Change in performance used to make next parameter change
- Moves restricted temporarily so that the same space is not searched repeatedly



 Chosen because other techniques may require "offline" simulation time (PSO, Genetic ...)

 While offline simulation occurs the conditions in network may change

 Only tabu search can react quickly to changing conditions



- Utility Function:
 - -U = 1 / delay + PDR
 - Simple to start, more factors taken into account later: ex) fairness index, congestion level etc
- Utility computed periodically
- Goal is for long term maximization of utility
- Adapt to changing network conditions



 Neighbourhood defined by adjusted one or more parameters (alpha or beta1, beta2)

 Tabu moves restrict the entire solution (alpha, beta1, beta2)

Tabu tenures (set to expire after 0.5 seconds of simulation time)



- Neighbourhood
 - Equal probability of changing each value
 - Value changes are restricted within a set range
 - 0.1 < alpha < 1,
 - 0.5 <= b1, b2 <= 10
 - b1 < b2
 - Small probability of a random value assignment within the range



- Aspiration Criteria
 - After a set number of optimization steps,
 we reset back to the best known
 solution, and search again from here
 - (do not know if a given tabu move will have a higher objective value than current so we can't use that)
- Tabu moves have set tenure



- Every 10 requests
 - Tabu Iteration
- Everyone 100 tabu iterations
 - Potentially reset back to best known solution
- Initial Condition
 - Alpha = 0.5, Beta1 = 2, Beta2 = 5(from original work by Singh et al)
 - Alpha -> 1, Beta1 -> 0.5, Beta2 -> 0.5



Performance Evaluation

HWMP Routing Algorithm used in this topology

Each MR running the AMB algorithm

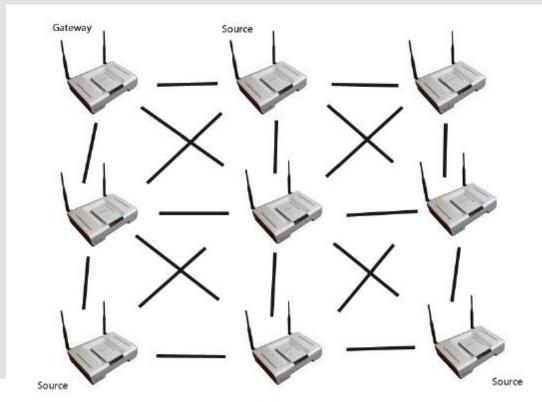


Figure 1. Example Network Topology



Performance Evaluation

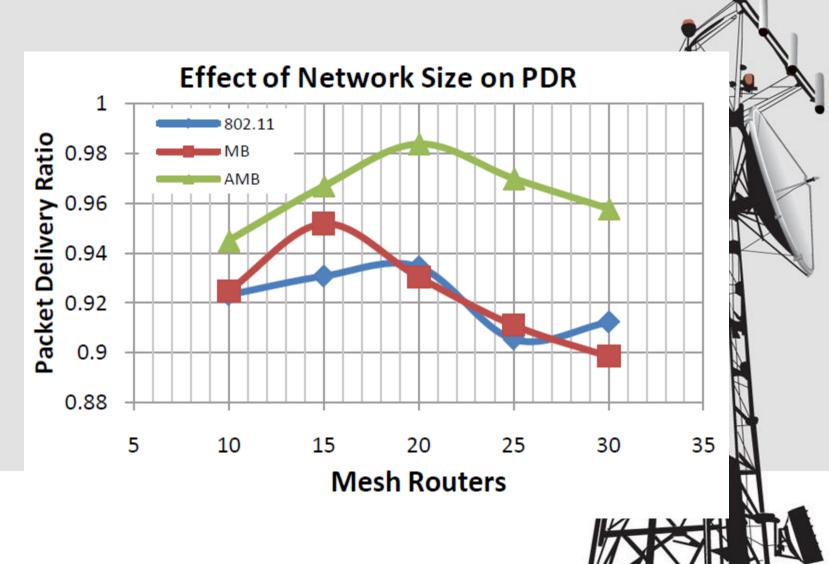
TABLE II.	SIMULATION PARAMETERS
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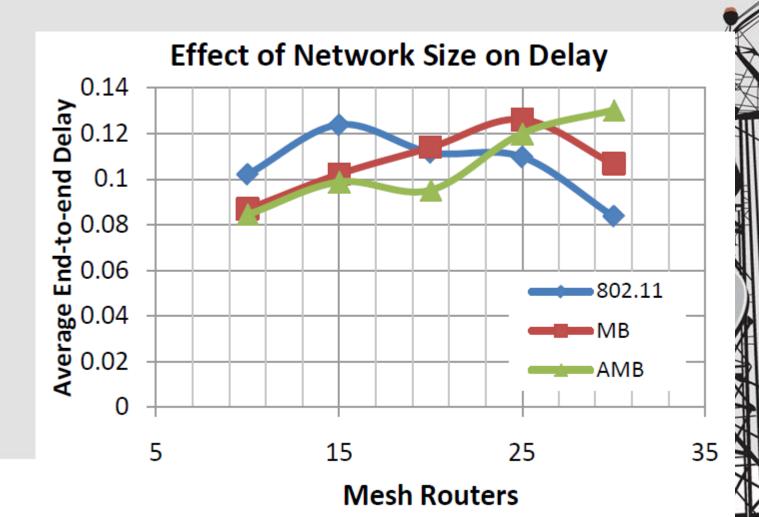
Parameter	Value
Interarrival rate	0.01
Packet Size	1024 bytes
MRs	10-30
Distance between MRs	100m
Source MRs	3
Simulation time	100s
Tabu tenure	0.5s
Packets between tabu move	15
Packets between tabu reset	300
Initial α	0.5
Initial β ₁	2
Intiali β ₂	5

Implemented in ns3 simulation tool

Performance Evaluation

- Important to note:
 - Every 300 packets, tabu resets to best known solution
 - Every 15 packets, tabu move
 - When a move is declared tabu, this lasts for 0.5 seconds
 - These numbers were determined empirically
 - $-\alpha, \beta_1, \beta_2$ taken from original work





- As network size grows
 - Adaptive Mixed Bias delivers more packets
 - Delay increases, likely due to packet drops in other solutions yielding an artificially lower delay
 - Mostly the delay is equivalent in all cases (Between 0.08 and 0.1s)



 Improvement in delay results may be seen with more emphasis on delay in utility (but may come at the cost of PDR)



Future Work

- Adjust utility function
 - Include fairness measure, congestion measure

- Adjust tabu search parameters
 - Neighbourhood function
 - Explore different rules for adjusting parameters
 - Tabu tenures
 - How long to keep a set of parameters tabu
 - Aspiration Criteria
 - More complex than return to best known



Future Work

- Experiment with other heuristic approaches:
 - Simulated annealing, temperature could be related to how congested the network
 - Fuzzy approach (rules translate well)
- Use an offline approach (GA, PSO) in order to determine a parameter set that "performs well" under many network conditions

Summary

- Overview of the Adaptive Mixed Bias (AMB) technique
- Comparison of AMB to IEEE 802.11 and MB
- Performs well, good starting point for further refinement
- Lots of opportunity for future research



Questions?

Thanks for listening!

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