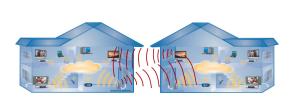
Distributed Spectrum Assignment for Home WLANs

Julien Herzen (EPFL) Ruben Merz (Swisscom) Patrick Thiran (EPFL)

April 17th, 2013

Context

Interfering neighboring wi-fi home/office networks





- Several possible channels (center frequencies)
- \bullet Variable bandwidth (5 \rightarrow 20 \rightarrow 40 \rightarrow 160 MHz), limited spectrum
- Non-heterogeneous density
- No central control

Goal

Joint allocation of channel center frequencies and bandwidths

Conflicting goals:

- Bandwidth

 → Capacity
- Bandwidth \nearrow \Rightarrow Interference likelihood \nearrow

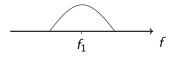
Goal

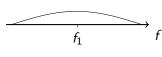
Joint allocation of channel center frequencies and bandwidths

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- Bandwidth

 → Capacity
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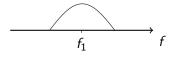
Capacity *→*

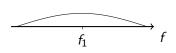
Goal

Joint allocation of channel center frequencies and bandwidths

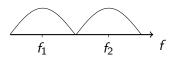
Conflicting goals:

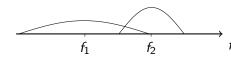
- Bandwidth \nearrow \Rightarrow Capacity \nearrow
- Bandwidth $\nearrow \Rightarrow$ Interference likelihood \nearrow





 $Capacity \nearrow$





Capacity →?

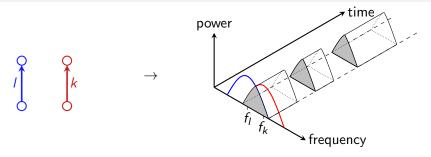
Design Goals

- Decentralized algorithm
- Global convergence guarantees
- Online for adaptivity to time-varying conditions
- Transparent to user traffic
- Practical for implementation on off-the-shelf 802.11 hardware

Main contribution

The first decentralized algorithm for joint center frequency and bandwidth adaptation with global convergence guarantees

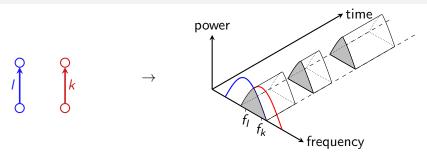
Interference Model



Interference produced by *k* on neighbor *l*:

$$I_l(k) = \operatorname{airtime}(k) \cdot \operatorname{overlap}(k, l)$$

Interference Model



Interference produced by k on neighbor l:

$$I_l(k) = \operatorname{airtime}(k) \cdot \operatorname{overlap}(k, l)$$

For two BSSs A and B:

$$I_A(B) = \sum_{I \in A} \sum_{k \in B} I_I(k)$$

Optimization Objective

Explicit interference vs. bandwidth trade-off:

$$\text{minimize} \quad \mathcal{E} := \underbrace{\sum_{A} \sum_{B \in \mathcal{N}_{\mathcal{A}}} I_{A}(B)}_{\text{Total interference}} \quad + \quad \underbrace{\sum_{A} \text{cost}_{A}(b_{A})}_{\text{Sum of bandwidth "costs"}}$$

- ullet cost_A(b_A) is the cost that BSS A attributes to using bandwidth b_A
- E.g., $cost_A(b_A) \propto 1/b_A$

Algorithm at BSS A

Initialization:

Pick a random configuration (f_A, b_A)

After random (exp. distributed) time intervals:

Pick a random configuration (f_{new}, b_{new})

Measure
$$e_1 := \sum_{B \in \mathcal{N}_A} (I_A(B) + I_B(A)) + \operatorname{cost}_A(b_A)$$
 if A uses (f_A, b_A)

Measure
$$e_2 := \sum_{B \in \mathcal{N}_A} (I_A(B) + I_B(A)) + \operatorname{cost}_A(b_{\text{new}})$$
 if A uses $(f_{\text{new}}, b_{\text{new}})$

Compute

$$eta_T = egin{cases} 1 & ext{if } e_2 < e_1 \ ext{exp} rac{e_1 - e_2}{T} & ext{else} \end{cases}$$

Set $(f_A, b_A) = (f_{new}, b_{new})$ with probability β_T

Convergence

Metropolis sampling for center frequency and bandwidth

Theorem

Denote X_n the global state of the network after the n-th iteration. Consider a network where all the BSSs run our algorithm using a given parameter T. Then X_n is a Markov chain, and it converges in distribution to

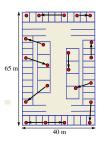
$$\pi(X) \propto e^{-\mathcal{E}(X)/T}$$

where X is a global state.

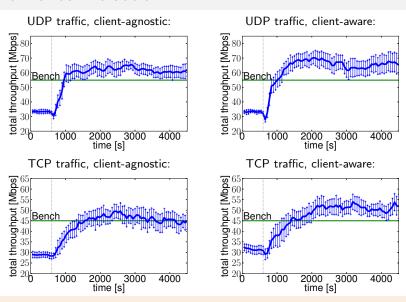
- State gets arbitrarily close to optimal for T small enough
- T encodes a trade-off between likelihood of local optima and asymptotic efficiency

Implementation

- 802.11g with 5, 10 and 20 MHz channel widths
- Interference measured by spending \leq 50 ms. out-of-band
- Optional client collaboration for interference measurement
- C++ implementation using *Click* in userspace
- $\operatorname{cost}_A(b_A) = 1/b_A$



Performance Evaluation



"Bench" line: centralized graph-coloring for fixed-width channels

- Random distribution of BSSs on the plane
- Capacity of link $I = b_l \cdot \log(1 + SINR)$
- $cost_A(b_A) = c/b_A$, optimization objective becomes:

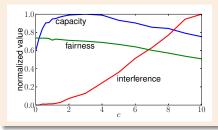
minimize
$$\sum_{A} \sum_{B \in \mathcal{N}_A} I_A(B) + c \cdot \sum_{A} 1/b_A$$

- c = 0: minimize interference
- $c \to \infty$: use largest bandwidth, irrespective of interference

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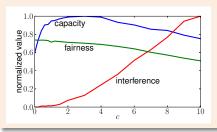
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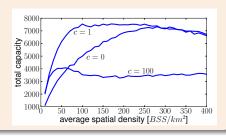


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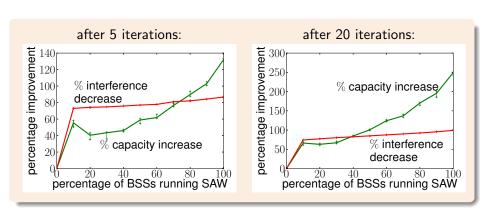
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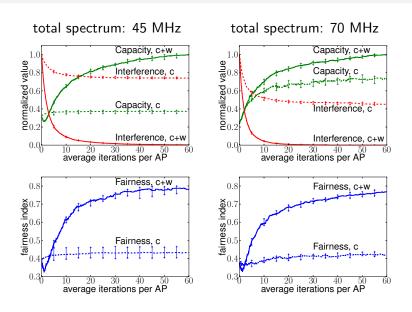
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Improvement with respect to random allocations





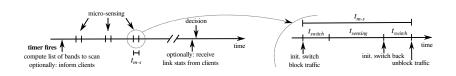
Conclusion

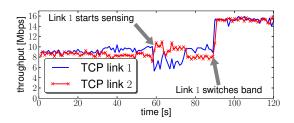
- Distributed, joint allocation of center frequencies and bandwidths
- Bandwidth influences both capacity and interference; ideal spectrum consumption should depend on network density
- Optimization of an explicit trade-off between interference mitigation and use of advantageous bandwidths
- Simple optimization objectives yield best results irrespective of network density
- Large capacity improvements, even when not all BSSs run the algorithm
- Testbed implementation shows feasibility and improvements compared to fixed-width graph coloring

Some Related Work

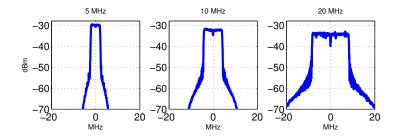
- Channel allocation / graph coloring, e.g., [Akella et al. 2005, Kauffmann et al. 2007, Duffy et al. 2011, Leith et al. 2012]
 - ► Main goal: minimize interference (no variable bandwidth)
- Variable bandwidth / white spaces, e.g., [Chandra et al. 2008, Bahl et al. 2009, Rayanchu et al. 2011]
 - Heuristics, no focus on self-organization

Micro-sensing

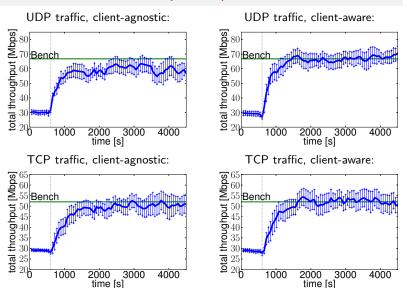




Channel widths



Performance Evaluation (uplink)



"Bench" line: centralized graph-coloring for fixed-width channels