

# Multicast Scheduling with Cooperation and Network Coding in Cognitive Radio Networks

(INFOCOM 2010 Paper)

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- 3 Problem Formulation and Solution
  - Centralized Greedy Algorithm
  - Distributed Online Algorithm
- 4 Performance Evaluation
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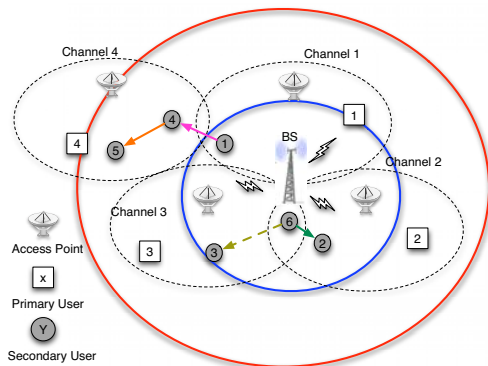
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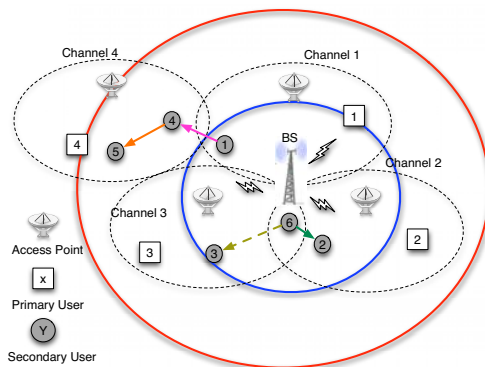
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# What is Cognitive Radio Network?



**Figure:** An illustrative example on multicast scheduling in CRNs.

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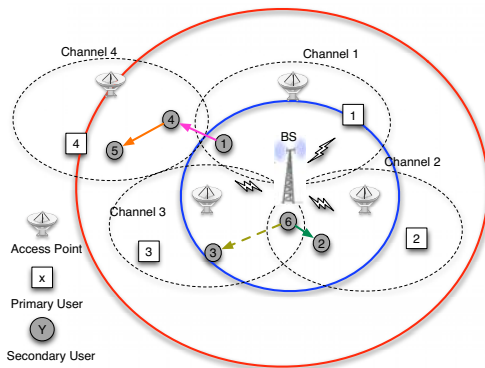


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- Access point and Primary user:  
Licensed channels

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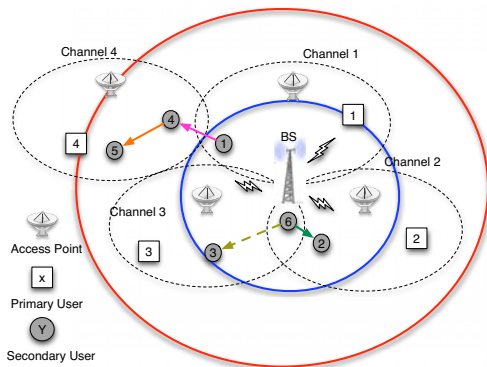


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**Resource:** Orthogonal channels, e.g. *FDM*, *CDM*.

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- The potential benefits provided by **user and channel diversities** are overlooked.

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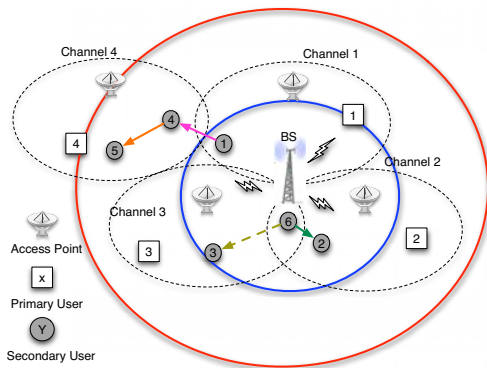
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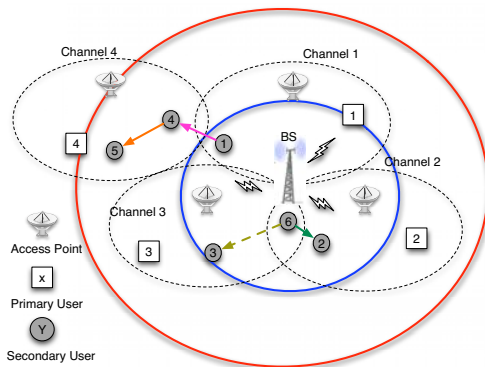


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The same reason for the need of a novel multicast algorithm.

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- 2 Encode the data blocks into  $n$  coded blocks with random linear coefficients.
- 3 Since all linearly independent blocks are equally innovative, certain number of coded blocks can recover the original packet.



The tasks:

- Power control for Base Station (BS).
- Channel assignment for Secondary Users (SU).

- $C$  PUs and orthogonal channels,  $N$  SUs and one BS.

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- $H(t) = \{h_n^c(t)\}_{N \times C}$ : Channel accessibility depending on channel state and SU location.
  - $h_n^c(t) = 1$ : SU  $n$  can access channel  $c$
  - $h_n^c(t) = 0$ : Otherwise

# Optimization Problem

Objective function: Maximize utility; Proportional fairness (can find a good balance between utilization and fairness)

$$\max_{P_{BS}, \Theta} \sum_{n=1}^N \frac{U_n}{\bar{r}_n} \quad (1)$$

$U_n$  represents throughput on SU  $n$ ;

$\bar{r}_n$  is the average throughput that SU  $n$  obtains over previous time slots;

$P_{BS} = \{P_{BS}^c\}_C$  denotes the multicast power used on each channel from BS;

$\Theta = \{\mu_{mn}^c\}_{N \times N \times C}$  is the set of feasible channel assignments.

# Constraints

Constraints on power control:

$$\begin{aligned} P_{BS}^c &\leq P_{max}^c \quad \forall c \\ P_{BS}^c \cdot g^c \cdot S_c &\leq \beta \quad \forall c \end{aligned} \tag{2}$$

# Constraints (Cont.)

Constraints on channel availability for cooperative communication.

$$\begin{aligned}\mu_{mn}^c &\leq h_m^c, \mu_{mn}^c \leq h_n^c \quad \forall m, n, c \\ \mu_{mn}^c &\leq l_m^c, \mu_{mn}^c \leq l_n^c \quad \forall m, n, c\end{aligned}\tag{3}$$

## Constraints (Cont.)

Constraints on potential interference by cooperative communication.

$$\begin{aligned}0 &\leq \sum_{m=1}^N \mu_{mn}^c \leq 1 \quad \forall n, c \\0 &\leq \sum_{c=1}^C \mu_{mn}^c \leq 1 \quad \forall m, n \\0 &\leq \sum_{m=1}^N \mu_{mn}^c + \sum_{m'=1}^N \mu_{nm'}^c \leq 1 \quad \forall n, c\end{aligned}\tag{4}$$



## Constraints (Cont.)

Constraints on utility calculation.

$$U_n = \sum_{c=1}^C BW \cdot \log_2 \left( 1 + \frac{P_{BS}^c \cdot g_n^c}{N_n^c} \right) + \sum_{c=1}^C \sum_{m=1}^N \mu_{mn}^c \omega_{mn}^c \quad (5)$$

$$\omega_{mn}^c = \min \left\{ \omega_{max}, \max \left\{ 0, \frac{B_m - B_n}{T} \right\} \right\}$$

$BW$  and  $N_n^c$  denote channel bandwidth and noise respectively;  $\omega_{mn}^c$  represents the achievable cooperative transmission rate from SU  $m$  to SU  $n$  on channel  $c$  with upper bound  $\omega_{max}$ ; it is also limited by amount of innovative data that SU  $m$  is able to contributed to  $n$ :  $B_m - B_n$ .

A centralized greedy algorithm can be derived using *branch-and-bound* algorithm.

Revise the problem with stochastic network model.

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# Buffer dynamics

$$B_n(t+1) = \max\{B_n(t) - R_n(t), 0\} + M_n(t) + \sum_{c=1}^C \sum_{m=1}^N \mu_{mn}^c(t) S_n^c(t) \omega_{mn}^c(t) \quad (6)$$

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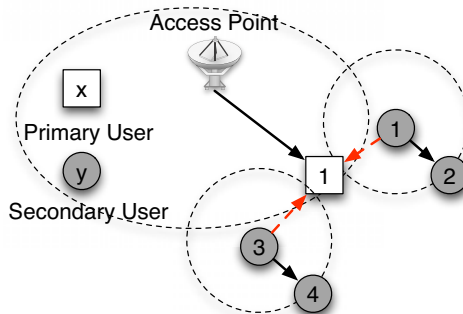
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- $M_n(t)$  is the multicast throughput obtained directly from BS at time slot  $t$  for SU  $n$ .

## Average throughput of SU $n$

$$r_n = \lim_{t \rightarrow \infty} \frac{1}{t} \sum_{\tau=0}^{t-1} R_n(\tau) \quad (7)$$

# Collisions by cooperative transmissions



**Figure:** Cooperative communication may generate interference to PUs.

## Collisions by cooperative transmissions (Cont.)

Total number of collisions caused by cooperative transmission for each PU:

$$E_c(t) = \sum_{m=1}^N \sum_{n=1}^N \mu_{mn}^c(t) I_m^c(t) (1 - S_c(t)) \quad (8)$$

Here,  $I_m^c(t)$  is the binary variable indicating whether the cooperative communication issued by SU  $m$  may generate interference to PU  $c$  at time slot  $t$ .

## Collisions by cooperative transmissions (Cont.)

Time average rate of interference for PU  $c$ :

$$e_c = \lim_{t \rightarrow \infty} \frac{1}{t} \sum_{\tau=0}^{t-1} E_c(\tau) \quad (9)$$

## Collisions by cooperative transmissions (Cont.)

Interference queue dynamics:

$$X_c(t+1) = \max\{X_c(t) - \rho_c, 0\} + E_c(t) \quad (10)$$

Here,  $\rho_c$  is the time average interference tolerant rate for channel PU  $c$ .

# Optimal online scheduling policies



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- Power control policy: BS chooses multicast power as the solution of the following problem

$$\begin{aligned} \max \quad & \sum_{n=1}^N R_n(t) M_n(t) \\ \text{s.t.} \quad & (2) \end{aligned}$$

Rationale: maximize aggregate throughput on all SUs with regard to direct multicast from BS.

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- Channel allocation policy: Channel resources are allocated as the solution of the following

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Stochastic Lyapunov Optimization is applied to solve this problem.

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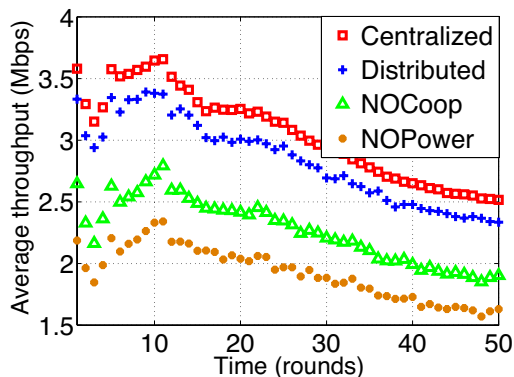
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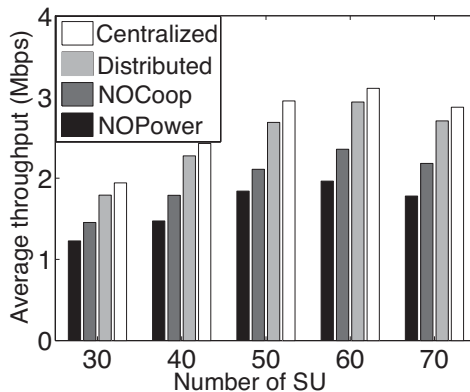
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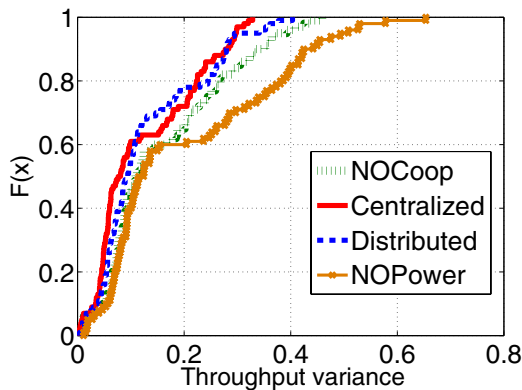




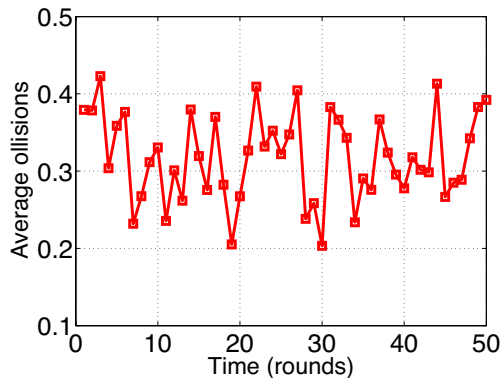
**Figure:** Average throughput performance of all protocols in realistic CRN scenarios.



**Figure:** Throughput performance of all protocols with different numbers of SUs, which represent the degree of possible cooperation among SUs.



**Figure:** CDF of all SUs with respect to throughput variance, which indicates fairness performance.



**Figure:** Evaluation of interference on PUs, which remains stable and bounded in our protocols.

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# Discussion



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- 1 Protocol interference model is utilized. What about the difference to system design and performance under physical interference model?
- 2 Both algorithms are derived with optimization tools and only multicast is addressed. How about extend the problem into unicast, anycast, broadcast, convergecast? Also, we can try graph-theory on the problems.

**Thank You!**