

Resource Allocation in OFDMA Systems: Relay-Enhanced Cellular Networks

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Overview

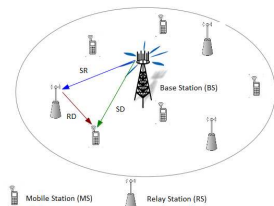
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Motivation

- Ubiquitous high data-rate coverage is the theme of next-generation wireless networks.
- Ubiquitous coverage demands that service has to reach users in the most unfavorable channel conditions.
- Shrink the size of cells, Cost-wise ineffective solution.
- Multihop relaying scheme is one of the attractive solution.
- Resource allocation (RA) becomes more complicated and challenging!

System Model

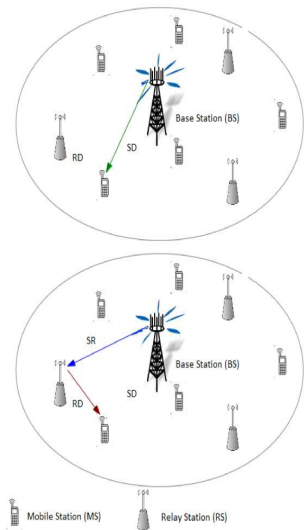
Consider an OFDMA relay-enhanced cellular network as shown below



Notations:

- $\mathcal{K} = \{0, 1, 2, \dots, K\}$ denotes the index of BS or a RS.
- $\mathcal{M} = \{1, 2, \dots, M\}$ denotes the index of a user.
- $\mathcal{N} = \{1, 2, \dots, N\}$ denotes the subchannel index.
- p_*^n denotes the power allocated to subchannel n on link ' $*'$ '
- $R_{k,m}^n$ denotes the achievable data rate of m^{th} user.
- γ_*^n denotes the CNRs of the link ' $*'$ ' on the n^{th} subchannel.

Problem Formulation



Achievable rates

Direct path:

$$R_{0,m}^n = R_{SD_m}^n = \log_2(1 + p_{SD_m}^n \gamma_{SD_m}^n)$$

Relaying path:

$$R_{k,m}^n = \frac{1}{2} \min \{ R_{SR_k}^n, R_{R_k D_m}^n \}$$

Where,

$$R_{SR_k}^n = \log_2(1 + p_{SR_k}^n \gamma_{SR_k}^n)$$
$$R_{R_k D_m}^n = \log_2(1 + p_{R_k D_m}^n \gamma_{R_k D_m}^n)$$

Problem Formulation

Mathematically, Resource Allocation (RA) problem is formulated as following

$$\begin{aligned} \max_{p_*^n} \quad & \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} R_{k,m}^n \\ \text{s.t.} \quad & \text{c1 : } \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \{p_{SD_m}^n + \sum_{k \in \mathcal{K}, k \neq 0} \frac{1}{2}(p_{SR_k}^n + p_{R_k D_m}^n)\} \leq P_T \\ & \text{or} \\ & \text{c1 : } \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} p_{R_k D_m}^n \leq P_R, \forall k \\ & \text{c2 : } p_{SD_m}^n, p_{SR_k}^n, p_{R_k M_m}^n \geq 0, \forall k, m, n \\ & \text{c3 : } p_{SD_m}^n p_{SR_k}^n = 0, p_{SD_m}^n p_{R_k M_m}^n = 0, \forall n \end{aligned} \tag{1}$$

Problem Re-Formulation

- Let us introduce $(K + 1)MN$ new variables $\{p_{k,m}^n \forall k, m, n\}$

$$p_{k,m}^n = \begin{cases} p_{SD_m}^n, & \text{if } k=0 \\ \frac{1}{2}(p_{SR_k}^n + p_{R_kD_m}^n), & \text{otherwise} \end{cases} \quad (2)$$

- The optimal power allocated to the first hop-link of a relaying path and that allocated to its second-hop link should satisfy

$$\frac{p_{SR_k}^n}{p_{R_kD_m}^n} = \frac{\gamma_{R_kD_m}^n}{\gamma_{SR_k}^n} \quad (3)$$

Problem Re-Formulation

With (2) and (3), we get the achievable rate of the k^{th} path of user m on the n^{th} subchannel as

$$R_{k,m}^n = C1_k \log_2(1 + C2_{k,m}^n p_{k,m}^n) \quad (4)$$

Where

$$C1_k = \begin{cases} 1 & \text{if } k = 0 \\ \frac{1}{2} & \text{otherwise} \end{cases}$$

$$C2_{k,m}^n = \begin{cases} \gamma_{SD_m}^n & \text{if } k = 0 \\ \frac{2\gamma_{SR_k}^n \gamma_{R_k D_m}}{\gamma_{SR_k}^n + \gamma_{R_k D_m}^n} & \text{otherwise} \end{cases}$$

Problem Re-Formulation

So the optimization problem (1) becomes

$$\begin{aligned} & \max_{\{p_{k,m}^n\}} \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} C1_k \log_2(1 + C2_{k,m}^n p_{k,m}^n) \\ \text{s.t. } & \text{c1 : } \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} p_{k,m}^n \leq P_T \\ & \text{c2 : } 0 \leq p_{k,m}^n \leq P^{\max}, \forall k, m, n \\ & \text{c3 : } p_{k,m}^n p_{r,s}^n = 0, \forall k \neq r \text{ and/or } m \neq s; \\ & \quad k, r \in \mathcal{K}; m, s \in \mathcal{M}; n \in \mathcal{N} \end{aligned} \tag{5}$$

Problem Re-Formulation

By introducing a group of binary variables $y_{k,m}^n$, problem (5) can be reformulated as

$$\begin{aligned} & \max_{\{p_{k,m}^n, y_{k,m}^n\}} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} C1_k \log_2(1 + C2_{k,m}^n p_{k,m}^n) \\ \text{s.t.} & \text{c1:} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} p_{k,m}^n \leq P_T \\ & \text{c2:} && 0 \leq p_{k,m}^n \leq y_{k,m}^n P^{\max}, \forall k, m, n \\ & \text{c3:} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n \leq 1, n \in \mathcal{N} \\ & \text{c4:} && y_{k,m}^n \in \{0, 1\}, m \in \mathcal{M}, k \in \mathcal{K}, n \in \mathcal{N} \end{aligned} \tag{6}$$

State-of-the-art Algorithms

- Method A : Relaxation, Divide and Conquer !
- Method B : Allocate channels to users with maximum gain and then solve for power.
- Method C : Relax, Dual optimization problem, solve for power and then solve for paths and assign subchannels.
- **Room for Improvement!**

Goal and Approach

- **Approach 1** : SBL for RA
- **Approach 2** : IRM Framework for RA

IRM Framework for RA

Goal : Solve the following optimization problem using IRM framework

$$\begin{aligned} & \max_{\{p_{k,m}^n, y_{k,m}^n\}} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} C1_k \log_2(1 + C2_{k,m}^n p_{k,m}^n) \\ \text{s.t. } & \text{c1 :} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} p_{k,m}^n \leq P_T \\ & \text{c2 :} && 0 \leq p_{k,m}^n \leq y_{k,m}^n P^{\max}, \forall k, m, n \\ & \text{c3 :} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n \leq 1, n \in \mathcal{N} \\ & \text{c4 :} && y_{k,m}^n \in \{0, 1\}, m \in \mathcal{M}, k \in \mathcal{K}, n \in \mathcal{N} \end{aligned} \tag{7}$$

Step 1 : Relaxation

Let $x_{k,m}^n = \frac{p_{k,m}^n}{P^{\max}}$, the above problem can be equivalently rewritten as following:

$$\begin{aligned} & \max_{\{x_{k,m}^n, y_{k,m}^n\}} && \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} C1_k \log_2(1 + C2_{k,m}^n P^{\max} x_{k,m}^n) \\ \text{s.t } & \text{c1 :} && \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} \sum_{m \in \mathcal{M}} P^{\max} x_{k,m}^n \leq P_T \quad , m \in \mathcal{M} \\ & \text{c2 :} && 0 \leq x_{k,m}^n \leq y_{k,m}^n \quad , \forall k, m, n \\ & \text{c3 :} && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n = 1 \quad , n \in \mathcal{N} \\ & \text{c4 :} && y_{k,m}^n \in \{0, 1\} \quad , m \in \mathcal{M}, k \in \mathcal{K}, n \in \mathcal{N} \end{aligned} \tag{8}$$

Step 1 : Relaxation Contd.

In particular, $y_{k,m}^n$ satisfies the last two constraints of the above problem if and only if $y_{k,m}^n$ solves the following minimization problem.

$$\begin{aligned} & \min_{\{y_{k,m}^n\}} \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} (y_{k,m}^n + \epsilon)^q \quad q \in (0, 1) \\ \text{subject to} \quad & \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n = 1 \quad , n \in \mathcal{N} \\ & y_{k,m}^n \geq 0 \quad , k \in \mathcal{K}, m \in \mathcal{M}, n \in \mathcal{N} \end{aligned} \tag{9}$$

Step 1 : Relaxation Contd.

Based on (9), we can relax (8) to

$$\begin{aligned} & - \min_{\{x_{k,m}^n, y_{k,m}^n\}} && - \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} C1_k \log_2(1 + C2_{k,m}^n P^{\max} x_{k,m}^n) + \\ & && \lambda \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} (y_{k,m}^n + \epsilon)^q \\ \text{s.t } & \text{c1 : } && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} P^{\max} x_{k,m}^n \leq P_T \quad , m \in \mathcal{M} \\ & \text{c2 : } && 0 \leq x_{k,m}^n \leq y_{k,m}^n \quad , \forall k, m, n \\ & \text{c3 : } && \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n = 1 \quad , n \in \mathcal{N} \end{aligned} \tag{10}$$

Here $\lambda < 0$.

Step 2: Iterative Re-weighted Minimization (IRM) Framework

Iterate : Until Convergence / Maximum iteration

$$\begin{aligned} - \min_{\{x_{k,m}^n, y_{k,m}^n\}} & - \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} C1_k \log_2(1 + C2_{k,m}^n P^{\max} x_{k,m}^n) + \\ & \lambda q \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} w_{k,m}^n y_{k,m}^n \end{aligned}$$

$$\text{s.t } c1 : \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} P^{\max} x_{k,m}^n \leq P_T$$

$$c2 : 0 \leq x_{k,m}^n \leq y_{k,m}^n, \forall k, m, n$$

$$c3 : \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n = 1, n \in \mathcal{N}$$

$$\text{Update: } W_{k,m}^n(t+1) = [y_{k,m}^n(t+1) + \epsilon(t+1)]^{q-1}$$

$\epsilon(t+1) = \text{Fix it}$ or update it adaptively

Algorithm : IRM Framework for RA

Input : CSI and maximum power allowed for per frame.

Iterate : Until Convergence / Maximum iteration

$$\begin{aligned} - \min_{\{x_{k,m}^n, y_{k,m}^n\}} & - \sum_{m \in \mathcal{M}} \sum_{n \in \mathcal{N}} \sum_{k \in \mathcal{K}} C1_k \log_2(1 + C2_{k,m}^n P^{\max} x_{k,m}^n) + \\ & \lambda q \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} w_{k,m}^n y_{k,m}^n \end{aligned}$$

$$\text{s.t } c1 : \sum_{k \in \mathcal{K}} \sum_{n \in \mathcal{N}} P^{\max} x_{k,m}^n \leq P_T$$

$$c2 : 0 \leq x_{k,m}^n \leq y_{k,m}^n, \forall k, m, n$$

$$c3 : \sum_{m \in \mathcal{M}} \sum_{k \in \mathcal{K}} y_{k,m}^n = 1, n \in \mathcal{N}$$

$$\begin{aligned} \text{Update: } W_{k,m}^n(t+1) &= [y_{k,m}^n(t+1) + \epsilon(t+1)]^{q-1} \\ \epsilon(t+1) &= \text{Fix it } \textit{or} \textit{ update it adaptively} \end{aligned}$$

Output : Power and channel allocation matrix, $x_{k,m}^n$ and $y_{k,m}^n$, respectively

Variants of the problem

- Different Relay Co-operative schemes (AF and CF).
- Partial/imperfect CSI Knowledge.
- Variable time length.
- Time-sharing of sub-channels, $y_{k,m}^n \in [0, 1]$

Summary

- The proposed algorithm uses IRM framework based on an effective relaxation of RA problem.
- The proposed algorithm jointly allocates the channel and power resources. Hence, different from state-of-the-art algorithms.