

Resource Allocation in Cognitive Wireless Powered Communication Networks with Wireless Powered Secondary Users and Primary Users

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Appendix A Feasibility check of the problem (P1) with a given τ_0

In order to check the feasibility of the problem (P1) with a given τ_0 , we obtain the maximum PU rate by assuming that all the subcarriers are allocated to the PU in the WIT phase as given by

$$\max_{\{p_c^i \geq 0\}, \{p_p^i \geq 0\}} \frac{1 - \tau_0}{N} \sum_{i=1}^N \ln \left(1 + \frac{p_p^i h_p^i}{\sigma^2} \right) \quad (\text{A1})$$

$$\text{s.t.} \quad \sum_{i=1}^N p_c^i \leq P, \quad (\text{A2})$$

$$\sum_{i=1}^N p_p^i (1 - \tau_0) \leq \zeta \sum_{i=1}^N p_c^i h_{sp}^i \tau_0. \quad (\text{A3})$$

It is easy to see that the objective function in (A1) is maximized when the constraint in (A3) is satisfied with equality, and a higher value of $\sum_{i=1}^N p_c^i h_{sp}^i$ can achieve a higher objective function value in (A1). In order to maximize $\sum_{i=1}^N p_c^i h_{sp}^i$, it is apparent that assigning all the available power to the subcarrier with the maximum h_{sp}^i is optimal, i.e., $p_c^j = P, j = \arg \max_i h_{sp}^i$, and $p_c^i = 0, i \neq j$. Now, the problem in (A1) is simplified as

$$\max_{\{p_p^i \geq 0\}} \frac{1 - \tau_0}{N} \sum_{i=1}^N \ln \left(1 + \frac{p_p^i h_p^i}{\sigma^2} \right) \quad (\text{A4})$$

$$\text{s.t.} \quad \sum_{i=1}^N p_p^i (1 - \tau_0) = \zeta P h_{sp}^j \tau_0. \quad (\text{A5})$$

The above problem is convex and thus can be easily solved by the Lagrange multiplier method as $p_p^i = \left(\frac{1}{\lambda N} - \frac{\sigma^2}{h_p^i} \right)^+$, where $(\cdot)^+ = \max(\cdot, 0)$ and λ is numerically obtained from (A5). The problem (P1) with a given τ_0 is feasible only if the obtained maximum objective function value in (A4) is larger than or equal to R_{min} , otherwise the problem is infeasible.

Appendix B Proof of Proposition 1

The partial Lagrangian of the problem (P1) with a given τ_0 is given as

$$L(\{p_c^i\}, \{p_s^i\}, \{p_p^i\}, \mu_1, \mu_2) = \frac{1 - \tau_0}{N} \sum_{i=1}^N \ln \left(1 + \frac{p_s^i h_s^i}{\sigma^2} \right) - \mu_1 \left(\sum_{i=1}^N p_s^i (1 - \tau_0) - \zeta \sum_{i=1}^N p_c^i h_s^i \tau_0 \right)$$

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$$-\mu_2 \left(\sum_{i=1}^N p_p^i (1 - \tau_0) - \zeta \sum_{i=1}^N p_c^i h_{sp}^i \tau_0 \right), \quad (\text{B1})$$

where μ_1 and μ_2 are the non-negative dual variables. We can rewrite the Lagrangian as

$$L(\{p_c^i\}, \{p_s^i\}, \{p_p^i\}, \mu_1, \mu_2) = \zeta \tau_0 L_1(\{p_c^i\}, \mu_1, \mu_2) + (1 - \tau_0) \sum_{i=1}^N L_2(p_s^i, p_p^i, \mu_1, \mu_2), \quad (\text{B2})$$

where

$$L_1(\{p_c^i\}, \mu_1, \mu_2) = \sum_{i=1}^N (\mu_1 h_s^i + \mu_2 h_{sp}^i) p_c^i, \quad (\text{B3})$$

$$L_2(p_s^i, p_p^i, \mu_1, \mu_2) = \frac{1}{N} \ln \left(1 + \frac{p_s^i h_s^i}{\sigma^2} \right) - \mu_1 p_s^i - \mu_2 p_p^i. \quad (\text{B4})$$

The dual function $G(\mu_1, \mu_2)$ is then obtained as the maximum objective function value of the following problem as given by

$$\max_{\{p_c^i \geq 0\}, \{p_s^i \geq 0\}, \{p_p^i \geq 0\}} L(\{p_c^i\}, \{p_s^i\}, \{p_p^i\}, \mu_1, \mu_2) \quad (\text{B5})$$

$$\text{s.t. } \sum_{i=1}^N p_c^i \leq P, \quad (\text{B6})$$

$$p_s^i p_p^i = 0, i = 1, \dots, N, \quad (\text{B7})$$

$$R_p \geq R_{min}. \quad (\text{B8})$$

It is observed that $\{p_c^i\}$ is not coupled with $\{p_s^i\}$ and $\{p_p^i\}$ in the above problem. Thus, we can decouple the above problem into two problems as given by

$$\begin{aligned} \max_{\{p_c^i \geq 0\}} \quad & \zeta \tau_0 L_1(\{p_c^i\}, \mu_1, \mu_2) \\ \text{s.t.} \quad & \text{constraint (B6)}, \end{aligned} \quad (\text{B9})$$

and

$$\begin{aligned} \max_{\{p_s^i \geq 0\}, \{p_p^i \geq 0\}} \quad & (1 - \tau_0) \sum_{i=1}^N L_2(p_s^i, p_p^i, \mu_1, \mu_2) \\ \text{s.t.} \quad & \text{constraints (B7), (B8)}. \end{aligned} \quad (\text{B10})$$

The problem in (B9) is shown to belong to linear programming and the optimal solution can be easily obtained as $p_c^k = P, k = \arg \max_i \mu_1 h_s^i + \mu_2 h_{sp}^i$, and $p_c^i = 0, i \neq k$. According to [1], the duality gap of the problem in (B10) is negligible for a large number of subcarriers and thus we can solve the problem in (B10) in the dual domain. The Lagrangian of the problem in (B10) is given by

$$L'(\{p_s^i\}, \{p_p^i\}, \nu) = (1 - \tau_0) \sum_{i=1}^N L_2(p_s^i, p_p^i, \mu_1, \mu_2) + \nu \left(\frac{1 - \tau_0}{N} \sum_{i=1}^N \ln \left(1 + \frac{p_p^i h_p^i}{\sigma^2} \right) - R_{min} \right), \quad (\text{B11})$$

where ν is the non-negative dual variable associated with the constraint in (B8). Then, the dual function $G'(\nu)$ of the problem in (B10) is given by

$$\begin{aligned} G'(\nu) = \quad & \max_{\{p_s^i \geq 0\}, \{p_p^i \geq 0\}} L'(\{p_s^i\}, \{p_p^i\}, \nu) \\ \text{s.t.} \quad & \text{constraint (B7)}. \end{aligned} \quad (\text{B12})$$

It is observed that the optimization variables p_s^i and p_p^i for different subcarriers are decoupled in the above problem. Thus, the above problem is decoupled into subproblems, one for each subcarrier as given by

$$\max_{p_s^i \geq 0, p_p^i \geq 0} (1 - \tau_0) L_2(p_s^i, p_p^i, \mu_1, \mu_2) + \frac{(1 - \tau_0)\nu}{N} \ln \left(1 + \frac{p_p^i h_p^i}{\sigma^2} \right) \quad (\text{B13})$$

$$\text{s.t. } p_s^i p_p^i = 0, \quad (\text{B14})$$

for $i = 1, \dots, N$. According to the constraint (B14), either the SU or the PU occupies the subcarrier in the WIT phase. Thus, we can solve the above problem by first deriving the optimal p_s^i assuming that the SU occupies the subcarrier i and the optimal p_p^i assuming that the PU occupies the subcarrier i , and then selecting the one with a higher objective function value. It is easy to verify that the objective function in (B13) is convex with respect to p_s^i and p_p^i . Thus, by assuming that subcarrier i is allocated to the SU or the PU, the optimal p_s^i and p_p^i can be obtained as

$$\hat{p}_s^i = \left(\frac{1}{\mu_1 N} - \frac{\sigma^2}{h_s^i} \right)^+, \quad (\text{B15})$$

$$\hat{p}_p^i = \left(\frac{\nu}{\mu_2 N} - \frac{\sigma^2}{h_p^i} \right)^+, \quad (\text{B16})$$

respectively. Then, the solution to the problem in (B13) is $p_s^i = \hat{p}_s^i, p_p^i = 0$ if $L_2(\hat{p}_s^i, 0, \mu_1, \mu_2) \geq L_2(0, \hat{p}_p^i, \mu_1, \mu_2) + \frac{\nu}{N} \ln \left(1 + \frac{\hat{p}_p^i h_p^i}{\sigma^2} \right)$ and is $p_s^i = 0, p_p^i = \hat{p}_p^i$ if $L_2(\hat{p}_s^i, 0, \mu_1, \mu_2) < L_2(0, \hat{p}_p^i, \mu_1, \mu_2) + \frac{\nu}{N} \ln \left(1 + \frac{\hat{p}_p^i h_p^i}{\sigma^2} \right)$. The dual variable ν can be obtained by solving the dual problem as given by

$$\min_{\nu \geq 0} G'(\nu). \quad (\text{B17})$$

The above problem can be solved efficiently by the subgradient method [2]. Finally, the dual variables μ_1 and μ_2 can be obtained by solving the dual problem as given by

$$\min_{\mu_1 \geq 0, \mu_2 \geq 0} G(\mu_1, \mu_2). \quad (\text{B18})$$

The above problem can be also solved efficiently via the subgradient method.

Appendix C The heuristic design

In this appendix, we propose the heuristic design. In order to let the PU harvest more energy to satisfy its minimum rate constraint, the value of p_c^i is optimized to maximize the energy harvested by the PU as given by

$$\max_{\{p_c^i \geq 0\}} \zeta \sum_{i=1}^N p_c^i h_{sp}^i \tau_0 \quad (\text{C1})$$

$$\text{s.t.} \quad \sum_{i=1}^N p_c^i \leq P. \quad (\text{C2})$$

It can be verified easily that the optimal solution to the above problem is $p_c^j = P, j = \arg \max_i h_{sp}^i$, and $p_c^i = 0, i \neq j$.

Then, we assign subcarriers to the PU with high priority in the WIT phase for satisfying the minimum rate constraint. Let \mathbb{N}_p and \mathbb{N}_s denote the sets of subcarriers allocated to the PU and the SU in the WIT phase, respectively. We initialize \mathbb{N}_p as $\mathbb{N}_p = \{i : i = 1, \dots, N\}$ and optimize $\{p_p^i \geq 0, i \in \mathbb{N}_p\}$ as

$$\min_{\{p_p^i \geq 0, i \in \mathbb{N}_p\}} \frac{1 - \tau_0}{N} \sum_{i \in \mathbb{N}_p} \ln \left(1 + \frac{p_p^i h_p^i}{\sigma^2} \right) \quad (\text{C3})$$

$$\text{s.t.} \quad \sum_{i \in \mathbb{N}_p} p_p^i (1 - \tau_0) \leq \zeta P h_{sp}^j \tau_0. \quad (\text{C4})$$

The above problem is convex and the optimal solution is $p_p^i = \left(\frac{1}{\alpha N} - \frac{\sigma^2}{h_p^i} \right)^+, i \in \mathbb{N}_p$ where α is numerically obtained from $\sum_{i \in \mathbb{N}_p} \left(\frac{1}{\alpha N} - \frac{\sigma^2}{h_p^i} \right)^+ (1 - \tau_0) = \zeta P h_{sp}^j \tau_0$. The set \mathbb{N}_p is then updated as $\mathbb{N}_p = \{i : p_p^i > 0, i = 1, \dots, N\}$. If the achieved PU rate $\frac{1 - \tau_0}{N} \sum_{i \in \mathbb{N}_p} \ln \left(1 + \left(\frac{h_p^i}{\alpha N \sigma^2} - 1 \right)^+ \right)$ is larger than R_{min} , then we exclude the subcarrier with the minimum h_p^i from the set \mathbb{N}_p and solve the problem in (C3)-(C4) with the new set \mathbb{N}_p . The above procedure terminates until the achieved PU rate is no larger than R_{min} and the last subcarrier excluded from the set \mathbb{N}_p shall be included in the set \mathbb{N}_p if the achieved PU rate is smaller than R_{min} . Then, the set \mathbb{N}_s is obtained as $\mathbb{N}_s = \{i : i = 1, \dots, N\} \setminus \mathbb{N}_p$ and the value of $\{p_s^i, i \in \mathbb{N}_s\}$ is optimized as

$$\max_{\{p_s^i \geq 0, i \in \mathbb{N}_s\}} \frac{1 - \tau_0}{N} \sum_{i \in \mathbb{N}_s} \ln \left(1 + \frac{p_s^i h_s^i}{\sigma^2} \right) \quad (\text{C5})$$

$$\text{s.t.} \quad \sum_{i \in \mathbb{N}_s} p_s^i (1 - \tau_0) \leq \zeta P h_s^j \tau_0. \quad (\text{C6})$$

The above problem has similar structure as the problem in (C3)-(C4) and the optimal solution is $p_s^i = \left(\frac{1}{\beta N} - \frac{\sigma^2}{h_s^i} \right)^+, i \in \mathbb{N}_s$, where β is obtained numerically from $\sum_{i \in \mathbb{N}_s} \left(\frac{1}{\beta N} - \frac{\sigma^2}{h_s^i} \right)^+ (1 - \tau_0) = \zeta P h_s^j \tau_0$.

References

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