Joint Transmission in QoE-Driven Backhaul-Aware MC-NOMA Cognitive Radio Network

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Abstract—In this paper, we develop a resource allocation framework to optimize the downlink transmission of a backhaul-aware multi-cell cognitive radio network (CRN) which is enabled with multi-carrier non-orthogonal multiple access (MC-NOMA). The considered CRN is composed of a single macro base station (MBS) and multiple small BSs (SBSs) that are referred to as the primary and secondary tiers, respectively. For the primary tier, we consider orthogonal frequency division multiple access (OFDMA) scheme and also Quality of Service (QoS) to evaluate the user satisfaction. On the other hand in secondary tier, MC-NOMA is employed and the user satisfaction for web, video and audio as popular multimedia services is evaluated by Quality-of-Experience (QoE). Furthermore, each user in secondary tier can be served simultaneously by multiple SBSs over a subcarrier via Joint Transmission (JT). In particular, we formulate a joint optimization problem of power control and scheduling (i.e., user association and subcarrier allocation) in secondary tier to maximize total achievable QoE for the secondary users. An efficient resource allocation mechanism has been developed to handle the non-linear form interference and to overcome the non-convexity of QoE serving functions. The scheduling and power control policy leverage on Augmented Lagrangian Method (ALM). Simulation results reveal that proposed solution approach can control the interference and JT-NOMA improves total perceived QoE compared to the existing schemes.

I. INTRODUCTION

Provisioning multimedia services such as video streaming, audio applications, web browsing, file download and best-effort services in forthcoming wireless networks, prioritizes user-centric resource allocation over conventional network-centric approach. In contrast to Quality of Service (QoS) in conventional wireless networks, user-centric resource allocation considers the users’ Quality of Experience (QoE) which is defined as the service-based subjective perception of end users [1]. To enhance spectral reuse, multi-carrier non-orthogonal multiple access (MC-NOMA) is a potential technique that allows multiple users’ transmissions over the same subcarrier [2]. [3]. Nevertheless, it is demonstrated in [4] that MC-NOMA can initiate additional interference and necessitates efficient spectrum allocation. Recently, it is shown that the integration of joint transmission (JT) as a coordinated multi-point (CoMP) feature with MC-NOMA can achieve considerable performance gain [5]. JT stimulates the cooperation of base stations (BSs) to serve a user through multiple BSs over a specified subcarrier. However, to realize the JT, backhaul capacity [6] of BSs is a limiting factor for serving the users cooperatively [8].

Recently, QoE has been considered as the user-centric criterion in various research works [9]–[11]. The authors in [9] proposed a resource allocation scheme for web and video multimedia services using multi-antenna BSs. In [10], a resource allocation for satisfying web, video and audio multimedia services is carried out in Device-to-Device (D2D) networks. The authors in [11] employed MC-NOMA and performed a QoE-driven resource allocation, where user association and subcarrier allocation operations were treated over different phases and each user was served by only one BS.

Keeping in view the BS densification in future wireless networks, we consider JT-NOMA as a potential technique to enhance the multimedia user perception. To our best knowledge, this integration has not been investigated yet in the context of QoE-aware resource allocation.

In this paper, we develop a QoE-aware resource allocation framework to support web surfing, video streaming, and audio application as multimedia services. We propose integrating JT with MC-NOMA in a two-tier multi-cell cognitive radio network (CRN) with a single macro BS (MBS) and multiple small BSs (SBSs) with finite backhauling capacity as the primary and secondary tier respectively. To enhance spectrum reuse, we consider underlay (UDL) mode of CRN for sharing a subcarrier between a primary and secondary user and co-channel deployment (CCD), to further share a subcarrier among multiple secondary users. Specifically, to the aim of maximizing total perceived QoE for the secondary users, we perform power control and joint scheduling (i.e., the joint operation of subcarrier allocation and user association) in secondary tier. We ensure the power constraints for all the BSs, QoS and QoE requirement thresholds for the primary and secondary users respectively and backhaul capacity for the SBSs. Hence, the main contributions of this paper are summarized as follows:

- We investigate JT-NOMA in multi-cell backhaul limited CRN to enhance perceived QoE.
- We adopt an efficient joint scheduling in secondary tier among users/SBSs/subcarriers through binary linearization.

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†Backhaul capacity for a specific BS can be defined as the number of users associated with that BS i.e., the load factor [6] or as the total transmitting data rate through that BS [7].
technique. We propose an efficient power control mechanism by handling the quasi-concave non-linear JT interference based on the Augmented Lagrangian Method (ALM).

- Simulation results verify that the proposed resource allocation outperforms the existing literature in terms of perceived QoE.

II. SYSTEM MODEL AND ASSUMPTIONS

As shown in Fig. 1, we consider downlink CRN where users and multiple SBSs are equipped with single antenna and uniformly distributed within the coverage area of single MBS. The primary tier consists of a cell centered MBS with wide coverage area and unlimited backhaul capacity denoted by $s$ and set of primary user terminals (PUTs) denoted by $M = \{1, ..., M\}$. The assessment metric for the PUTs is assumed to be conventional QoS and the subcarrier allocation per subcarrier $n$ with corresponding matrices $p_{L,G}$ and $q_{m}^{n}$ as the downlink transmit power of SBS $l$ to SUT $g$ and $q_{m}^{n}$ as the downlink transmit power of the MBS to PUT $n$ both over the subcarrier $n$ with quasi-static, independent and identically distributed (i.i.d) Rayleigh model. Also we denote $p_{L,G}^{n}$ as the downlink transmit power of SBS $l$ to SUT $g$ and $q_{m}^{n}$ as the downlink transmit power of the MBS to PUT $n$ both over the subcarrier $n$ with corresponding matrices $p = \{p_{L,G}^{n}\}_{L \in \mathcal{L} \times \{G, ... \}}$ and $q = \{q_{m}^{n}\}_{M \times \{\mathcal{L} \times \{G, ... \}}$. Respectless to scheduling operation, the received signal of SUT $g$ from SBS $l$ over the subcarrier $n$ with corresponding matrices $p = \{p_{L,G}^{n}\}_{L \in \mathcal{L} \times \{G, ... \}}$ and $q = \{q_{m}^{n}\}_{M \times \{\mathcal{L} \times \{G, ... \}}$. Respectless to scheduling operation, the received signal of SUT $g$ from SBS $l$ over the subcarrier $n$ is given by (1), where $x_{a,t}^{c}$ indicates the transmitted signal from BS $a$ to SUT $b$ over the subcarrier

$$y_{n,l,g}^{\mu} = \sqrt{p_{L,G}^{n} h_{l,g}^{n} x_{l,g}^{n}} + \sum_{m \in M} \sqrt{q_{m}^{n} h_{l,m}^{n} x_{l,m}^{n}} + \omega_{n,l,g}$$

(1)

UDL Interference: results from sharing a subcarrier among a primary and a secondary user. The MBS and SBSs impose inter-tier UDL interference to the involving SUTs and PUTs, respectively.

CCD Interference: results from the reuse of a subcarrier among multiple secondary users associated with different SBSs. In particular, SUTs experience intra-tier inter-cell interference corresponding to adjacent interfering SBSs.

NOMA Interference: Since MC-NOMA is exploited in secondary tier, a subcarrier may be shared among multiple SUTs within a SBS yielding intra-tier intra-cell interference.

JT Interference: The secondary tier is also JT-enabled, resulting the joint interference from SBSs involving in JT operation to other SUTs sharing the same JT subcarrier. Finally, each SUT can be served simultaneously over multiple subcarriers within a cell using carrier aggregation. Let us denote $h_{l,g}^{n}$ as the channel gain between BS $x$ and user $y$ over the subcarrier $n$ with quasi-static, independent and identically distributed (i.i.d) Rayleigh model. Also, we denote $p_{L,G}^{n}$ as the downlink transmit power of SBS $l$ to SUT $g$ and $q_{m}^{n}$ as the downlink transmit power of the MBS to PUT $n$ both over the subcarrier $n$ with corresponding matrices $p = \{p_{L,G}^{n}\}_{L \in \mathcal{L} \times \{G, ... \}}$ and $q = \{q_{m}^{n}\}_{M \times \{\mathcal{L} \times \{G, ... \}}$. Respectless to scheduling operation, the received signal of SUT $g$ from SBS $l$ over the subcarrier $n$ is given by (1), where $x_{a,t}^{c}$ indicates the transmitted signal from BS $a$ to SUT $b$ over the subcarrier

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(1)

We assume that the full channel state information (CSI) is available for the BSs [10], and channel gains satisfy the order of $|h_{1,1}^{n}| \leq |h_{1,2}^{n}| \leq \ldots \leq |h_{L,G}^{n}|$. Let us denote the set of SUTs in NOMA cluster of subcarrier $n$ at the SBS $l$ by $\mathcal{G}_{l}^{n}$. For specific couple of SUTs in same NOMA cluster e.g., $(g, g') \in \mathcal{G}_{l}^{n}$, the transmitted signal for $g'$ is detectable in $g$ only if $g'$ is the precedent one in the order. Then, $g'$ decodes its own signal while treating $g$-th signal as interference. On the other side, $g$ with better channel condition, decodes $g'$-th superimposed signal first and removes it through Successive Interference Cancellation (SIC) before decoding its own message [2]. The procedure is supposed to be performed for every pair of SUTs in same NOMA cluster.

III. QoE MODEL AND PROBLEM FORMULATION

Based on what discussed so far, the downlink received signal-to-interference-plus-noise-ratio (SINR) for SUT $g$ associated with SBS $l$ over subcarrier $n$ is expressed as

$$\gamma_{l,g}^{n}(\theta, \varepsilon, p, q) = \frac{p_{L,G}^{n} h_{l,g}^{n} |x_{l,g}^{n}|^2}{I_{l,g}^{n}(UDL) + I_{l,g}^{n}(CCD) + I_{l,g}^{n}(NOMA) + I_{l,g}^{n}(JT) + \sigma_{g}^2}$$

(2)
where, 
\[ I_{1,R}^{p}(UDL) = \sum_{m \in M} \sum_{n \in N} \pi_{m,n}^{p} q_{m,n}^{p} |h_{m,n}^{p}|^2, \]  
(3) 
\[ I_{1,R}^{p}(CCD) = \sum_{l \in L} \sum_{g \in G} \sum_{g' \in G} \theta_{l,g,g'} p_{l,g}^{p} q_{l,g}^{p} |h_{l,g}^{p}|^2, \]  
(4) 
\[ I_{1,R}^{p}(NOMA) = \sum_{g \in G} \theta_{l,g} q_{l,g}^{p} |h_{l,g}^{p}|^2, \]  
(5) 
\[ I_{1,R}^{p}(JT) = \sum_{l \in L} \sum_{g \in G} \sum_{g' \in G} 2\theta_{l,g,g'} p_{l,g}^{p} q_{l,g}^{p} |h_{l,g}^{p}|^2 |h_{l,g'}^{p}|^2, \]  
(6) 
and \( \pi_{m,n}^{p} \) is the predefined primary tier binary subcarrier allocation variable with the value of 1 if the subcarrier \( n \) is allocated to PUT \( m \) and zero otherwise. Besides, the binary variables \( \theta_{l,g} = \mathbf{1}_{\{l \in L \mid |g| \in |G| \}} \) and \( \epsilon = \mathbf{1}_{\{l \in L \mid g \in |N| \}} \) indicate associating of user \( g \) with BS \( l \) and allocating the subcarrier \( n \) of BS \( l \) to user \( g \), respectively. Moreover, \( I_{1,R}^{p}(NOMA) \) reveals the precedence of SUTs with poor channel condition in decoding order through applying successful SIC. The procedure also stands for the special case where couple of JT-enabled and single-associated SUTs (i.e., the SUTs served only by one SBS) are in a NOMA cluster. In ordering of SUTs in a NOMA cluster for demonstrated two-tier CRN in Fig. 1, the cell-edge JT-NOMA SUTs are precedent to cell-centered non-JT-NOMA SUTs. In a NOMA cluster, neighbouring SUTs experience intra-tier intra-cell interference as the result of superposition coding in power domain of the transmitter. The achievable data rate of SUT \( g \) then is given by 
\[ R_{g}(\theta, \epsilon, p, q) = \sum_{l \in L} \sum_{n \in N} \theta_{l,g} \epsilon_{l,g}^{n} \log_2 (1 + q_{l,g}^{n}(\theta, \epsilon, p, q)). \]  
(7) 
In this paper, we adopt Mean Opinion Score (MOS) as the main criterion of QoE for the web surfing, video streaming, and audio application services with indices \( j \in \{1, 2, 3\} \), respectively [10].

Let us define the total perceived QoE for the SUTs [11]
\[ U_{QoE}^{j}(\theta, \epsilon, p, q) = \sum_{g \in G} \sum_{l \in L} \sum_{n \in N} \theta_{l,g} \epsilon_{l,g}^{n} \log_2 (1 + q_{l,g}^{n}(\theta, \epsilon, p, q)), \]  
(8)
as the objective function of optimization for the \( j \)-th QoE service with \( \theta, \epsilon \) and \( p, q \) as the decision variables. For the transmit power thresholds we have
\[ \sum_{m \in M} \sum_{n \in N} \pi_{m,n}^{p} q_{m,n}^{p} \leq q_{\max}, \]  
(9) 
\[ \sum_{g \in G} \sum_{n \in N} \theta_{l,g} \epsilon_{l,g}^{n} p_{l,g}^{p} \leq p_{\max} \quad \forall l, \]  
(10)
where \( q_{\max} \) and \( p_{\max} \) are the maximum transmit power for MBS and SBSs, respectively. The QoS requirement of PUTs can be ensured if 
\[ \tilde{R}_{m}(\theta, \epsilon, p, q) = \sum_{n \in N} \pi_{m,n}^{p} \log_2 (1 + q_{m}^{n}|h_{m,n}^{p}|^2) \geq R_{m}^{\min}. \]  
(11)

After achieving successful SIC and signal decoding operation in SUTs, the QoE requirement is conceivable if [10]
\[ MOS_{g}^{j}(\sum_{l \in L} \sum_{n \in N} \theta_{l,g} \epsilon_{l,g}^{n} \log_2 (1 + q_{l,g}^{n}(\theta, \epsilon, p, q))) \geq MOS_{g}^{\min,j}. \]  
(12)
The backhaul capacity limit for the SBS \( l \) can be ensured through
\[ C_{l}(\theta, \epsilon, p, q) = \sum_{g \in G} \sum_{n \in N} \theta_{l,g} \epsilon_{l,g}^{n} \log_2 (1 + q_{l,g}^{n}(\theta, \epsilon, p, q)) \leq C_{l}^{\max}. \]  
(13)

For the backhaul limit as the load of each SBS in terms of number of associated SUTs we have [6]
\[ \sum_{g \in G} \theta_{l,g} \leq Z_{l} \quad \forall l. \]  
(14)

Regarding the system feasibility, each SUT is imposed to be associated with one SBS and allocated one subcarrier at least given by
\[ \sum_{l \in L} \theta_{l,g} \geq 1 \quad \forall g, \]  
(15) 
\[ \sum_{l \in L} \epsilon_{l,g}^{n} \geq 1 \quad \forall g, \]  
(16) 
respectively. The complexity of SIC decoding over a specified subcarrier on receiver can be controlled through limiting the number of multiplexed users over that subcarrier as
\[ \sum_{l \in L \in G} \epsilon_{l,g}^{n} \leq \Omega_{n} \quad \forall n, \]  
(17)
with \( \Omega_{n} \) denoting the maximum number of multiplexed users over subcarrier \( n \). The precedence of the cell association procedure to the subcarrier allocation [13] can indicated by
\[ \epsilon_{l,g}^{n} \leq \theta_{l,g} \quad \forall l, g, n. \]  
(18)

The problem at hand for joint QoE-aware optimization of the \( j \)-th service is formally stated as
\[ P_{1} : \max_{\theta, \epsilon, p, q} U_{QoE}^{j}(\theta, \epsilon, p, q) \]  
\[ \text{s.t.} \quad (9)-(18), \]  
(19)
which is complex due to the objective function along with non-linear and non-convex constraints [10]-[13].

IV. POWER CONTROL AND JOINT SCHEDULING ALGORITHM

In this section, we provide an alternative decomposition method to address (19). In doing so, two sub-problems namely power control and joint scheduling are solved in a joint convergent manner based on Algorithm 1.
A. Power Control Policy

Under predetermined values of variables \([\theta^{(t-1)}, e^{(t-1)}]\), the joint primitive optimization problem [19] is transformed into power control sub-problem with decision variables \(p^{(t)}\) and \(q^{(t)}\) given by

\[
P_2: \max_{p^{(t)}, q^{(t)}} U_j^{QoE}(\theta^{(t-1)}, e^{(t-1)}, p^{(t)}, q^{(t)})
\]

s.t: \([9] - [13]\)

(20)

The reformulated sub-problem is non-linear and quasi-concave form because of JT interference [2]. For multiplication of two continuous variables \(p_{r', g'}^{n}(t)\) and \(p_{r, g}^{n}(t)\), the following convex inequality for any fixed \(\lambda > 0\) holds [19]

\[
p_{r', g'}^{n}(t) p_{r, g}^{n}(t) \leq \frac{\lambda}{2} (\frac{p_{r', g'}^{n}(t)}{p_{r, g}^{n}(t)})^2 + \frac{1}{2\lambda} (\frac{p_{r', g'}^{n}(t)}{p_{r, g}^{n}(t)})^2 \forall l', l'' , n, n' \neq l'' .
\]

(21)

Then, the convex form of JT interference can be given as

\[
T_n^{L_g}(JT) = \sum_{l' \in L} \sum_{l'' \in L} \sum_{g \in G} \sum_{g' \in G} \sum_{n} \frac{\lambda}{2} (\frac{p_{r', g'}^{n}(t)}{p_{r, g}^{n}(t)})^2 |h_{r', g'}^{n}|^2 + \frac{1}{2\lambda} (\frac{p_{r', g'}^{n}(t)}{p_{r, g}^{n}(t)})^2 |h_{r, g}^{n}|^2 .
\]

(22)

Due to incorporating interference terms in (11)-(13), the sub-problem (20) is still non-convex which initiates a duality gap between the solution of (20) and its dual problem. We employ a power control policy leveraging ALM [15] where \(\alpha\) is the augmenting Lagrangian multiplier and \(\Psi_f - \Psi_b\) are the Lagrangian multipliers for the corresponding constraints. Accordingly, the dual problem of (20) for the \(j\)-th service is expressed as following

\[
D^j_{\phi_2} = \min_{\alpha, \Psi_b, \ldots, \Psi_f} \max_{p, q} \phi_2(\alpha, \Psi_b, \ldots, \Psi_f, p, q).
\]

(23)

To address (23) through ALM, the Lagrangian function can be given by (24) on top of the next page. By updating the Lagrangian multipliers following (16) after some iterations, the duality gap will be zero, Karush–Kuhn–Tucker (KKT) conditions hold which yields a sub-optimal solution [17].

B. Joint scheduling

Given \([p^{(t)}, q^{(t)}]\) matrices obtained from power control policy, constraints (11)-(13) in (19) along with the objective function are in non-linear multiplicative form. We define new binary decision variable as \(\chi_{l, g}^{n}(t) = \theta_{l, g}^{n}(t)e_{l, g}^{n}(t)\) indicating the joint operation of scheduling. By including the following additional constraints [18],

\[
\chi_{l, g}^{n}(t) \leq \theta_{l, g}^{n}(t) \forall l, g, n,
\]

(25)

\[
\chi_{l, g}^{n}(t) \leq e_{l, g}^{n}(t) \forall l, g, n,
\]

(26)

\[
\theta_{l, g}^{n}(t) + e_{l, g}^{n}(t) - 1 \leq \chi_{l, g}^{n}(t) \forall l, g, n,
\]

(27)

\(^3\)Recall that for \(\lambda = \frac{p_{r', g'}^{n}(t)}{p_{r, g}^{n}(t)}\), the approximation is tight.

the non-linear form is derived from multiplication of binary variables is tackled hitherto [24]. The non-convexity due to interference terms incorporation nonetheless, is the challenging issue yet. Hence, the relaxation of binary decision variables into continuous domain is invoked so as to make a more tractable sub-problem. Further, binary-forcing constraints [19]-[22] given by

\[
\sum_{l \in L} \sum_{g \in G} \sum_{n} \left( (\epsilon_{l, g}^{n}(t)) - (e_{l, g}^{n}(t))^2 \right) \leq 0,
\]

(28)

\[
\sum_{l \in L} \sum_{g \in G} \left( (\theta_{l, g}^{n}(t)) - (e_{l, g}^{n}(t))^2 \right) \leq 0,
\]

(29)

\[
\sum_{l \in L} \sum_{g \in G} \sum_{n} \left( (\lambda_{l, g}^{n}(t)) - (e_{l, g}^{n}(t))^2 \right) \leq 0,
\]

(30)

can guarantee the binary domain of decision variables. The feasible region for the aforementioned constraints consists of corner critical points i.e., \([0,1]\). Thus, the joint optimization problem for the \(j\)-th service is restated as

\[
P_3: \max_{\chi^{(t)}, \theta^{(t)}, e^{(t)}, p^{(t)}, q^{(t)}} U_j^{QoE}(\theta^{(t-1)}, e^{(t-1)}, p^{(t)}, q^{(t)})
\]

s.t: \([10], [18], [25], [27]\)

(31)

Reformulated optimization problem (31) is still non-convex due to incorporating decision variable \(\chi\) as the indicator of interference. The remaining solution for the sub-problem is completely similar to power control policy henceforth where, ALM with respect to decision variables \(\chi^{(t)}, \theta^{(t)}, e^{(t)}\) is applied to obtain the sub-optimal solution. Then, dual problem will be indicated by

\[
D^j_{\phi_3} = \min_{\alpha, \Delta_b, \ldots, \Delta_v, \chi, \theta, e} \max_{p, q} \phi_3(\alpha, \Delta_b, \ldots, \Delta_v, \chi, \theta, e),
\]

(32)

where, \(\Delta_b, \ldots, \Delta_v\) are the corresponding Lagrangian multipliers. The Lagrangian multipliers can be updated as in [16] to obtain the sub-optimal solution.

V. Simulation Results

In this section, simulation results are presented to investigate the efficiency of proposed solutions of the joint optimization problem. The MBS covers 500 meters of area, coexisting with SBSs with 50 meters of coverage. Maximum transmit power of the MBS \((q^{\text{max}})\) and all the SBSs \((p_i^{\text{max}} \forall i)\) are 42dBm and 37dBm, respectively. The bandwidth of 15 kHz for each subcarrier is considered. For the backhaul capacity, the commercial optical fiber modem capable of supporting \(C^{\text{max}} = 11.183\) Mbps \(\forall l\) is considered [17]. Other simulation parameters are set as \(\sigma^2 = -117\) dBm, \(R_{\text{min}} = 2\) bps/Hz \(\forall m\), MOS\(_{g, j}^{\text{min}} = 1\) bps/Hz \(\forall g, j \in \{1, 2, 3\}, Z_l = 3\) Hz and \(\Omega_{\eta} = 2\) Hz, unless specified.

The convergence behavior of the proposed solution is demonstrated with average MOS and average data rate on left and right hand side of the figures, respectively. Also, the performance evaluation is demonstrated in figures through four cases, i.e., proposed JT-OMA, non-JT-OMA according to [11], as well as Orthogonal Multiple Access (OMA) schemes like JT-OMA and non-JT-OMA for benchmarking purposes.
\( \xi_{b}^{i}(\Psi_{b}, \ldots, \Psi_{f}, p, q) = U_{QoE}^{i}(\theta(t^{i}), e(t^{i-1})^{i}, p^{i}, q^{i}) + \frac{1}{2a^{(i)}} \left[ \left( \sum_{m \in \mathbb{M}} \sum_{n \in \mathbb{N}} q_{m}^{n} - q_{m}^{max} \right) \right]^{2} - \frac{1}{2} \Psi_{b}^{2} \) +

\[ \sum_{t \in \mathbb{L}} \left[ \left( \Psi_{c1}^{i} + a^{(i)} \sum_{g \in \mathbb{G}} \sum_{n \in \mathbb{N}} \left( \sum_{n \in \mathbb{N}} e_{l}^{n, (t-1)} - p_{l}^{(t)max} \right) \right] \right]^{2} - \Psi_{c1}^{2} \] +

\[ \sum_{m \in \mathbb{M}} \left[ \left( \Psi_{c2}^{i} + a^{(i)} \left( \sum_{g \in \mathbb{G}} \sum_{n \in \mathbb{N}} \left( \sum_{n \in \mathbb{N}} e_{l}^{n, (t-1)} - p_{l}^{(t)max} \right) \right) \right] \right]^{2} - \Psi_{c2}^{2} \]

Algorithm 1 Joint Resource Allocation

Initialization: \( E_{rr} = 10^{-3}, \theta^{(0)} \) and \( e^{(0)} \) are randomly initialized, transmit power matrices are initialized across subcarriers (i.e., \( p_{l}^{(t)max} = p_{l}^{(t)max} / N \forall m, n, q_{m}^{n} = q_{m}^{max} / N \forall m, n \)). Iteration numbers \( t \) and \( T \) are initialized to 0, ALM Convergence = Final Convergence = false, \( \alpha = 2 \) and other Lagrangian multipliers are set to 0.1.

1: \( \text{repeat} \)
2: \( \text{repeat} \)
3: Solve (20) to get \( p^{(t)} \) and \( q^{(t)} \).
4: if \( |p^{(t)} - p^{(t-1)}| < E_{rr} \) \( \text{then} \)
5: ALM Convergence = true.
6: else
7: \( t = t + 1 \).
8: Update Lagrangian multipliers \((\Psi_{b}) - (\Psi_{f})\).
9: end if
10: until ALM Convergence
11: \( t = 1 \).
12: \( \text{repeat} \)
13: Solve (21) to get \( \theta^{(t)} \) and \( \chi^{(t)} \).
14: if \( \chi^{(t)} = \chi^{(t-1)} \) \( \text{then} \)
15: ALM Convergence = true.
16: else
17: Update Lagrangian multipliers \((\Delta_{b}) - (\Delta_{c})\).
18: \( t = t + 1 \).
19: end if
20: until ALM Convergence
21: Calculate \( U_{QoE}^{(T)}(\theta, e, p, q) \).
22: if \( |U_{QoE}^{(T)}(\theta, e, p, q) - U_{QoE}^{(T-1)}(\theta, e, p, q)| < E_{rr} \) \( \text{then} \)
23: Final Convergence = true.
24: else
25: \( T = T + 1 \).
26: end if
27: until Final Convergence

For all the services, minimum target MOS i.e., \( MOS_{min}^{g} \) is mapped to \( R_{g}^{min} = 2 \) bit/s/Hz while the maximum MOS value, 5 for web service and 4.5 for video and audio services are mapped to \( R_{g}^{max} = 7 \) bit/s/Hz [9].

1) Web Service Assessment: The network dependent parameters consisting of maximum segment size, round trip time and web page size are set as in (11). The network configuration parameters are \( L = 10, N = 32, \) and \( M = 6 \). Fig. 2(a) is related to the convergence behavior of our proposed solution for different numbers of SUTs indicating the exact values of load latency (11) at convergence point. In Fig. 2(d) with increasing number of SUTs, ‘JT-NOMA’ scheme outperforms others in terms of average MOS. Also, CCD, NOMA and JT interference terms, result in decreasing trend of average MOS per SUT versus increasing the number of SUTs.

2) Video Service Assessment: For the video service with common video coding H.264/AVC, the network configuration parameters are \( L = 10, N = 16, \) and \( G = 10 \). In single-snapshot scenario shown in Fig. 2(b), convergence of the proposed algorithm and corresponding peak signal to noise ratio (PSNR) values (9) at convergence point are demonstrated for different values for PUTs. The multi-snapshot Fig. 2(c) demonstrates decreasing average MOS per SUT versus increasing the number of PUTs due to higher UDL interference from MBS on SUTs. It is also noticeable that integration of JT-NOMA can achieve higher foreseen QoE than non-JT-NOMA (11) due to providing opportunities for SUTs to be associated with multiple SBSs.

3) Audio Service Assessment: In this section, two cases of maximum and minimum value for Packet Loss Ratio (PLR) is considered as \( PLR = 20\% \) and minimum \( PLR = 0\% \) respectively (10). The network configuration parameters are set to \( L = 10, G = 8, \) and \( M = 4 \). Fig. 2(e) shows the convergence behavior of proposed algorithm for two values of the number of subcarriers with related rating factor (24) in each case is calculated. In Fig. 2(f), increasing the number of subcarriers let SUTs access more subcarriers based on carrier aggregation and also more reusing opportunity which leads to higher average MOS per SUT. Subcarriers as the key factor for JT let more scheduling opportunity for SUTs and yield more foreseen QoE for JT-enabled baselines in comparison with non-JT. Also, more subcarriers facilitate reuse of the spectrum with respect to NOMA and thus, superiority of the NOMA over OMA is clearly convincing.

VI. CONCLUSIONS

In this paper, we studied the integration of JT and MC-NOMA in the downlink of backhaul limited multi-cell two-tier CRN to enhance the foreseen MOS for web, video, and audio multimedia services. Inspired of various interference categories initiated by reusing the subcarriers by JT-NOMA in CRN context, adequate resource allocation techniques were invoked to the non-linear form of interference terms as well as to overcome the non-convexity of MOS serving functions. In simulation results, the convergence behaviour and superiority of the proposed algorithm is shown using different network parameters for multimedia services.

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