A New Multiple Access Technique for 5G: Power Domain Sparse Code Multiple Access (PSMA)

MOHAMMAD MOLTAFET¹, NADER MOKARI², MOHAMMAD.R JAVAN³, HAMID SAEEDI⁴, and HOSSEIN PISHRO-NIK⁵
¹ECE department, Tarbiat Modares University, Tehran, Iran
²ECE department, Tarbiat Modares University, Tehran, Iran
³EE Department, Sharood University of Technology, Shahrood, Iran
⁴ECE department, Tarbiat Modares University, Tehran, Iran
⁵University of Massachusetts, Amherst, MA, USA

Corresponding author: Nader Mokari (e-mail: nader.mokari@modares.ac.ir)

ABSTRACT In this paper, a new approach for multiple access (MA) in fifth generation (5G) of cellular networks called power domain sparse code multiple access (PSMA) is proposed. In PSMA, we adopt both the power domain and the code domain to transmit multiple users’ signals over a subcarrier simultaneously. In such a model, the same sparse code multiple access (SCMA) codebook can be used by multiple users where, for these users, power domain non-orthogonal multiple access (PD-NOMA) technique is used to send signals non-orthogonally. Although the signal of different SCMA codebooks can be detected orthogonally, the same codebook used by multiple users produces interference over these users. With PSMA, a codebook can be reused in the coverage area of each base station more than one time, which can improve the spectral efficiency. We investigate the signal model as well as the receiver and transmitter of the PSMA method. In the receiver side, we propose a message passing algorithm (MPA) based successive interference cancellation (SIC) detector to detect the signal of each user. To evaluate the performance of PSMA, we consider a heterogeneous cellular network (HetNet). In this case our design objective is to maximize the system sum rate of the network subject to some system level and QoS constraints such as transmit power constraints. We formulate the proposed resource allocation problem as an optimization problem and solve it by successive convex approximation (SCA) techniques. Moreover, we compare PSMA with SCMA and PD-NOMA from the performance and computational complexity perspective. Finally, the effectiveness of the proposed approach is investigated using numerical results. We show that by a reasonable increase in complexity, PSMA can improve the spectral efficiency about 50% compared to SCMA and PD-NOMA.

INDEX TERMS Non-orthogonal multiple access (NOMA), sparse code multiple access SCMA, power domain coded sparse code multiple access (PSMA), resource allocation, successive convex approximation.

I. INTRODUCTION

MULTIPLE access (MA) techniques have an essential role on the performance improvement of cellular networks. In the fourth generation (4G) of cellular networks, orthogonal frequency division multiple access (OFDMA) is proposed as an efficient MA technique to address the upcoming challenges. As the statistical data shows, mobile data traffic will grow several folds in the next decade [1], [2]. Therefore, the next generations of cellular networks should be designed to address existing challenges like spectral efficiency and energy efficiency.

Due to the demand for high data rate services and the limitations of the available bandwidth for cellular networks, applying new techniques and methods to improve spectral efficiency in 5G is very important. For MA techniques in 5G, some non-orthogonal techniques such as power domain non-orthogonal multiple access (PD-NOMA) [3] and sparse code multiple access (SCMA) [4] are proposed. By applying superimposed coding on the transmitter side, PD-NOMA assigns a subcarrier to multiple users simultaneously, while on the receiver side, by using successive interference cancellation (SIC) method, the signals of users are detected. SCMA is a codebook based MA technique in which each subcarrier can be used in different codebooks on the transmitter side, and on
the receiver side, users’ signals are detected by applying the message passing algorithm (MPA).

Recently, PD-NOMA and SCMA have received significant attention as appropriate candidates for MA technique for 5G [5]- [19]. In [5], the authors study user pairing in a PD-NOMA based system. They show that the system throughput can be improved by pairing users enjoying good channel situations with users suffering from poor channel conditions. The authors of [6] study the joint power allocation and pre-coding design in a multiuser multiple input multiple output (MIMO) PD-NOMA based system in order to maximize the system sum rate. In [12], by considering a minimum data rate requirement for the users with bad channel situations, the authors propose a power allocation problem. To solve the corresponding problem, they apply two algorithms: an optimal algorithm using the bisection search method with high computational complexity and a suboptimal one based on the SIC approach. In [13], the authors propose different resource allocation problems. Based on the proposed method, they study the effect of the fairness among users in a PD-NOMA based system. The authors of [14] propose a resource allocation method which maximizes the energy efficiency in an SCMA-based system. In [15], PD-NOMA and SCMA as the pioneer candidates of MA in 5G are compared from performance and receiver complexity points of view. In [16], the authors, by considering imperfect channel state information at transmitter, investigate radio resource allocation in a multicarrier PD-NOMA based system. They formulate an optimization problem which minimizes the total transmit power with minimum rate requirement constraint. The authors of [17] study sum rate maximization in a full-duplex PD-NOMA based system. They solve the proposed optimization problem by exploiting both the optimal solution with high complexity and the suboptimal solution with low complexity.

In the core of PD-NOMA lies the SIC technique which makes it possible to allocate one subcarrier to more than one user by removing the interference resulting from non-exclusive utilization of subcarriers. On the other hand, SCMA employs exclusive codebooks for each user where each codebook is assigned non-exclusively to certain number of subcarriers. Instead of SIC, SCMA uses MPA to remove the resulting interference. The core idea of PSMA is to take advantage of SCMA ability to remove the inference using MPA while letting the users to utilize the codebooks non-exclusively, i.e., each codebook is assigned to more than one user. The SIC method is then applied to remove the interference resulting from non-exclusive use of the codebooks after applying the MPA.

The main contributions of this paper can be summarized as follows:

- We propose a multiple access method called PSMA at which a codebook can be reused more than one time in the coverage area of a BS. Reusing a codebook more than one time can improve the system spectral efficiency.
- We also propose a MPA based SIC detector to detect the signal of each user in the considered PSMA-based system. Moreover, we investigate the SIC ordering in this system.
- We investigate the signal model and detection techniques of PSMA together with PD-NOMA and SCMA for comparison. Moreover, we compare the receiver complexity for the three techniques.
- We propose a novel resource allocation problem for PSMA-based heterogeneous networks (HetNets) in which we maximize the sum rate with certain constraints. To solve the corresponding problem, we apply advanced optimization techniques to address the non-convexity of the problem.
- For practical scenarios, we consider similar resource allocation problems for PD-NOMA and SCMA to compare their performance and complexity with that of PSMA.

The results indicate that by a reasonable increase in the receiver complexity, PSMA can improve the spectral efficiency as high as 50% compared to SCMA and PD-NOMA.

The remainder of this paper is organized as follows: system model and signal model for transmitter and receiver of PD-NOMA and SCMA are studied in Section II and Section III, respectively. The PSMA technique and its signaling on transmitter and receiver sides are investigated in Section IV. The complexity of the receiver for different MA techniques is studied in Section V. To evaluate the performance of PSMA, resource allocation problems are proposed in Section VI. In Section VII, the solution algorithms of the proposed resource allocation problems are presented. Numerical results are presented in Section VIII. Finally, the paper is concluded in Section IX.

II. MULTIPLE ACCESS TECHNIQUE: PD-NOMA

In a PD-NOMA system, each subcarrier can be assigned to multiple users simultaneously by applying superimposed coding (SC), and each user removes the signals of other users by exploiting SIC. Based on the PD-NOMA approach, each user on the receiver side removes the signals of the users with worse channel, and considers the signals of other users as noise.

A. PD-NOMA SYSTEM MODEL

We consider a downlink PD-NOMA-based HetNet with \( F \) BSs, \( M \) users, and \( N \) subcarriers. In this system model, \( F \) indicates the set of BSs with \( f = 1 \) denoting the macro base station (MBS). \( M_f \) denotes the set of users of cell \( f \) with \( \cup_{f \in F} M_f = M \) denoting the set of all users. \( N \) demonstrates the set of subcarriers. In addition, \( h_{m,n}^f \) is the channel coefficient between user \( m \) and BS \( f \) on subcarrier \( n \) and \( p_{m,n}^f \) is the transmit power of BS \( f \) to user \( m \) over subcarrier \( n \). We define the subcarrier assignment indicator \( \beta_{m,n}^f \) with \( \beta_{m,n}^f = 1 \) if subcarrier \( n \) is assigned to user \( m \) over BS \( f \), and otherwise \( \beta_{m,n}^f = 0 \).
B. PD-NOMA SIGNAL MODEL: RECEIVED SIGNAL

With parameter definitions in the previous subsection, the received signal at user \( m \) in BS \( f \) over subcarrier \( n \) is given by:

\[
Y_{m,n}^f = h_{m,n}^f \sqrt{p_{m,n}^f} s_{m,n}^f + \sum_{j \in M_f, j \neq m} \beta_{j,n}^f h_{m,n}^j \sqrt{p_{j,n}^f} s_{j,n}^f + \sum_{i \in F_1 \times \bar{F} \times \bar{F}} \sum_{j \in M_i} \beta_{j,n}^{i,f} h_{m,n}^{i,f} \sqrt{p_{j,n}^{i,f}} s_{j,n}^{i,f} + w_{m,n}^f,
\]

where \( s_{j,n}^f \) indicates the message which BS \( i \) sends to user \( j \) on subcarrier \( n \) and \( w_{m,n}^f \) shows the noise. The first term of (1) indicates the signal of user \( m \) on subcarrier \( n \) over BS \( f \), the second term presents the NOMA interference term, and the third term shows the inter-cell interference.

C. PD-NOMA SIGNAL: RECEIVER SIDE

In the PD-NOMA approach, on the receiver side for user \( m \), the signal of users with better channel situation are considered as noise and the signals of other users are detected and removed. For example, if we consider a worse user \( m' \), i.e., \( |h_{m,n}^f| > |h_{m',n}^f| \), the signal to interference plus noise ratio (SINR) of user \( m' \) at user \( m \), i.e., \( \gamma_{m,n}^f(m) \), can be obtained by equation (2) shown on the next page. An important issue in this procedure is that we assume user \( m \) is able to decode the message of user \( m' \). This is possible if the SINR of user \( m' \) at user \( m \) is larger than that of user \( m' \) at its receiver, i.e., \( \gamma_{m,n}^f(m') \geq \gamma_{m,n}^f(m) \). In single cell PD-NOMA, the strategy is to sort users based on their channel gains and let users with better channels decode the signals of users with worse channels. This strategy ensures that the condition \( \gamma_{m,n}^f(m') \geq \gamma_{m,n}^f(m) \) for single cell PD-NOMA is always satisfied. However, for multicell PD-NOMA, we must adopt appropriate policy to guarantee this condition.

After decoding the signal of user \( m' \), user \( m \) will decode it and subtract it from the received signal. User \( m \) will apply this SIC procedure for all users worse than itself to finally obtain the following signal:

\[
\tilde{Y}_{m,n}^f = h_{m,n}^f \sqrt{p_{m,n}^f} s_{m,n}^f + \sum_{j \in M_f, |h_{j,n}^f|^2 > |h_{m,n}^f|^2} \beta_{j,n}^f h_{m,n}^j \sqrt{p_{j,n}^f} s_{j,n}^f + \sum_{i \in F_1 \times \bar{F} \times \bar{F}} \sum_{j \in M_i} \beta_{j,n}^{i,f} h_{m,n}^{i,f} \sqrt{p_{j,n}^{i,f}} s_{j,n}^{i,f} + w_{m,n}^f,
\]

which leads to SINR \( \gamma_{m,n}^f(m) \). This can be obtained from (2) using the corresponding parameters. The transmitter and receiver block diagram of a PD-NOMA-based system with minimum mean square error (MMSE) detector is shown in Fig. 1, in which \( L_T \) shows the number of users which are superimposed in each subcarrier.

III. MULTIPLE ACCESS TECHNIQUE: SCMA

An SCMA encoder is a mapping from \( \log_2(J) \) bits to an \( N \)-dimensional codebook of size \( J \) [4]. The \( N \)-dimensional codewords of a codebook are sparse vectors with \( U \) \( (U < N) \) non-zero entries that refer to \( U \) specific subcarriers. Based on the SCMA approach, codebooks that are composed of subcarriers are the basic resource units in networks [4], [19]. If each codebook consists of \( U \) subcarriers, there are \( C(N,U) = \frac{N!}{(N-U)U!} \) codebooks in the considered system. The transmitter and receiver block diagrams of an SCMA-based system are shown in Fig. 2.

A. SCMA SYSTEM MODEL

In SCMA, the BS sends the signals of all users simultaneously. Since users use different codes that occupy different subcarriers, the received signal over each subcarrier has components from signals of other users’ codes. Note that in the considered HetNet model, users in each cell are assumed to use different codes meaning that a code is reused in each cell only once. We consider a downlink SCMA-based HetNet with \( F \) BSs, \( M \) users, and \( C \) codebooks. In this system...
\[ \gamma_{m',n}^f (m) = \frac{|h_{m,n}^f|^2 p_{m',n}^f}{\sum_{j \in \mathcal{M}_f} |h_{j,n}^f|^2 |h_{m',n}^f|^2 p_{j,n}^f + \sum_{i \in \mathcal{F}, i \neq f} \sum_{j \in \mathcal{M}_f} |h_{m,n}^i|^2 p_{j,n}^i + E\{|w_{m,n}^f|^2\}}. \] (2)

\[ Y_{m,n}^f = \sum_{i \in \mathcal{F}} \sum_{j \in \mathcal{M}_i} \sum_{c \in \mathcal{C}} q_{i,j,c}^f h_{m,n}^i p_{n,c}^i \eta_{n,c}^f \sqrt{\rho_{c}^f} s_{n,c}^i + w_{m,n}^f, \] (4)

where \( s_{n,c}^i \) is the message which BS \( f \) sends on subcarrier \( n \) in codebook \( c \) and \( w_{m,n}^f \) indicates the noise that user \( m \) experiences on subcarrier \( n \).

We note that over each subcarrier, each user receives the signals of all other users which use codebooks that contain subcarrier \( n \).

**C. SCMA SIGNAL MODEL: RECEIVER SIDE**

At the receiver side of user \( m \) that uses code \( c \), the MPA is performed. The algorithm naturally cancels the interference from all other codes different from code \( c \). Note that since code \( c \) is reused, we have interference from code \( c \) used in other cells over the considered user \( m \) in cell \( f \). With parameters defined in the system model subsection, in an SCMA based HetNet system, the SINR of user \( m \) on codebook \( c \) over BS \( f \) is given by equation (5), where \( \sigma_{m,c}^f \) is the noise that user \( m \) achieves from codebook \( c \) over BS \( f \) and \( |E\{\sigma_{m,c}^f\}|^2 \) is the power of it.

**IV. MULTIPLE ACCESS TECHNIQUE: PSMA**

In the proposed PSMA system, we assume that each codebook can be assigned to more than one user in a BS simultaneously as shown in Fig. 3. Based on this new approach, each codebook is assigned to more than one user by applying the SC method. On the receiver side, the users’ signals are detected by using MPA and SIC. Based on the PSMA approach, each user can detect and remove the signals of users with worse average channel gain by applying MPA and SIC while considering the signals of the users with better average channel gain as noise. The transmitter and receiver block diagrams of the proposed PSMA-based system are shown in Fig. 4.

**A. PSMA SYSTEM MODEL**

1) PSMA-based single cell system model

We consider a downlink PSMA-based system with one BS, \( M \) users, \( N \) subcarriers, and \( C \) codebooks. In this system model, \( h_{m,n} \) is the channel coefficient between user \( m \) and BS on subcarrier \( n \), \( p_{m,c} \) is the transmit power of BS to user \( m \) on codebook \( c \), and \( q_{m,c} \) demonstrates the codebook assignment between user \( m \) and codebook \( c \) with \( q_{m,c} = 1 \) if codebook \( c \) is allocated to user \( m \) and otherwise \( q_{m,c} = 0 \). In addition, \( \rho_{n,c} \) indicates the mapping between subcarriers and codebooks with \( \rho_{n,c} = 1 \) if codebook \( c \) consists of subcarrier \( n \), otherwise \( \rho_{n,c} = 0 \). We assume that \( \rho \) is a known parameter. Also, we assume that \( p_{m,c} \) is allocated to...
\[
\gamma_{m,c}^f = \frac{q_{m,c}^f p_{m,c}^f \sum_{n \in N} h_{m,n}^f |h_{m,n}^f|^2}{\sum_{f' \in F \setminus \{f\}} \sum_{m' \in M_{f'}} \sum_{n \in N} q_{m',c}^{f'} p_{m',c}^{f'} \rho_{n,c}^{f'} |h_{m,n}^{f'}|^2 + E\{|\sigma_{m,c}^f|^2\}}.
\]

&  
\sum_{f' \in F} \sum_{m' \in M_{f'}} \sum_{n \in N} q_{m',c}^{f'} p_{m',c}^{f'} \rho_{n,c}^{f'} |h_{m,n}^{f'}|^2 + E\{|\sigma_{m,c}^f|^2\}

\text{FIGURE 3: Signal model of PSMA-based system. As can be seen, by applying superimposed coding, each codebook can be assigned to more than one user simultaneously. For example, the signals of UE2 and UE3 are transmitted on codebook 1.}

subcarrier \( n \) in codebook \( c \) based on \( \eta_{n,c} \) where \( 0 \leq \eta_{n,c} \leq 1 \) is determined based on the codebook design and satisfies \( \sum_{n \in N} \eta_{n,c} = 1 \) \( \forall c \) [4], [19].

Note that in contrast to the SCMA model, a code can be reused by more than one user in the cell, i.e., for a code \( c \), we may have \( \sum_{m \in M_f} q_{m,c}^f \geq 1 \).

2) PSMA-Based HetNet System Model

We consider a downlink PSMA-based HetNet with \( F \) BSs, \( M \) users, and \( C \) codebooks. Here, we again emphasize that in contrast to the SCMA model, a code can be reused by more than one user in each cell.

B. PSMA SIGNAL MODEL: RECEIVED SIGNAL

In PSMA, a codebook can be reused more than once in each cell. The received signal at user \( m \) over subcarrier \( n \) in cell \( f \) (for a general case of multicell scenarios) is given by:

\[
Y_{m,n}^f = \sum_{f' \in F} \sum_{m' \in M_{f'}} \sum_{c \in C} q_{m',c}^{f'} h_{m,n}^{f'} \rho_{n,c}^{f'} |h_{m,n}^{f'}|^2 + w_{m,n}^{f'}
\]

where \( s_{m,k}^i \) shows the message which is sent on subcarrier \( k \) in codebook \( c \).

In SCMA, we assume the other signals coming from the different BSs as noise. In PSMA, however, in addition to the different BSs signals, we accept some controllable interference to improve the system sum rate. In SCMA, there is the constraint \( \sum_{m \in M_f} q_{m,c}^f \leq 1, \forall c \in C, f \in F \) which indicates that in each BS, a codebook can be assigned to at most one user. However, in PSMA we have \( \sum_{m \in M_f} q_{m,c}^f \leq L_T, \forall c \in C, f \in F \) which shows that in each BS codebook can be assigned to \( L_T \) users simultaneously.
C. PSMA SIGNAL MODEL: RECEIVER SIDE

Since the PSMA takes advantage of both PD-NOMA and SCMA techniques, the receiver performs the detection techniques of these schemes in an appropriate way. In the PSMA method, on the transmitter side, users are sorted based on certain criteria (in our case, channel quality on the codebooks). On the receiver side, a user detects the signals of all other users that are worse than itself according to this criterion and removes them from the received signal using SIC. This user treats the signal of all users better than itself as noise. Note that in PSMA, to detect the signals of users, the message passing algorithm is run. Therefore, users with different codebooks do not interfere with each other and only users with the same codebook produce interference over each other. Therefore by sorting users, we mean sorting users that are using the same codebook.

In a system with SIC based receivers, SIC ordering in each receiver to determine the order of detection and to cancel the signal of other users, has essential role in the performance of the system. Finding, an optimal SIC ordering in a system such as the considered PSMA based system is very complicated. In PSMA based systems, signal of different users are transmitted using the same codebook. As we know a codebook consists of several subcarriers, therefore, proposing of a criterion for SIC ordering in PSMA based systems to consider the effects of all subcarriers is very challenging. Therefore, we may consider a criterion based on which we sort the users and better users are able to decode the signals of worse users. The direct channel gain is used in single-carrier systems is not suitable for the PSMA as it is a multicarrier system. We must take into account all the subcarriers used by users for the sorting purpose. Here, the proposed criterion for sorting users is based on their average channel gains in a codebook and given by

$$\hat{h}_{m,c} = \frac{\sum_{n=1}^{N} \rho_{n,c} |h_{m,n}|^2}{\sum_{n=1}^{N} \rho_{n,c}},$$

where we note that for each code $c$, we have $\sum_{n=1}^{N} \rho_{n,c} = U$. Also note that for two users $m$ and $m'$ using the same codebook $c$, we say user $m$ is better than user $m'$ if we have $\hat{h}_{m,c} \geq \hat{h}_{m',c}$.

On the receiver side, user $m$, which is using codebook $c$, tries to decode the signals of all users using the same codebook $c$ with worse average channel gains. Assume we have a single cell PSMA network and that user $m$ is better than user $m'$. Assuming that user $m$ has decoded the signals of all users worse than user $m'$ and subtracted them from the received signal, the available signal is given by

$$Y_{m,c} = \sum_{n \in N} h_{m,n} \rho_{n,c} \eta_{n,c} \sqrt{p_m} s_{m,n}$$

$$+ \sum_{j \in M, |j,c| > |m,c|} \sum_{n \in N} q_{j,c} h_{m,n} \rho_{n,c} \eta_{n,c} \sqrt{p_j} s_{j,n} + \sigma_{m,c},$$

and the SINR of user $m'$ at user $m$ is given by (9). Note that the SINR of user $m'$ at its receiver is given by (10). We emphasize that for user $m$ to correctly decode the signal of user $m'$, we must have $\gamma_{m',c}(m) \geq \gamma_{m',c}(m')$. For various power allocation strategies for the codebooks, ensuring $\hat{h}_{m,c} \geq \hat{h}_{m',c}$ does not guarantee the requirement $\gamma_{m',c}(m) \geq \gamma_{m',c}(m)$. Therefore, in some cases, we must explicitly declare this condition as an optimization constraint.

Subtracting the detected signals of all users worse than user $m$ from the received signal, the corresponding SINR is given by (11).

For multicell scenario, the same procedure is applied. However, the signals received from other cells are treated as noise. After detecting the signals of all users worse than user $m$, the available signal for user $m$ is given by:

$$Y_{m,c} = \sum_{k \in N} h_{m,k} \rho_{n,c} \eta_{k,c} \sqrt{p_{m,k}} s_{m,k}$$

$$+ \sum_{j \in M, |j,c| > |m,c|} \sum_{k \in N} q_{j,c} h_{m,k} \rho_{n,c} \eta_{k,c} \sqrt{p_j} s_{j,k} + \sigma_{m,c}.$$
where the first term of (12) represents the signal of user $m$ on codebook $c$ over BS $f$. The second term shows the interference that comes from non-orthogonality in power domain. The third term indicates inter-cell interference. The corresponding SINR is given by

$$\gamma_{m,c} = \frac{q_m c P_{m,c} \sum_{n \in N} \eta_{n,c} \rho_{n,c} h_{m,n}^2}{\sum_{i \in M, h_{i,c} \geq h_{m,n}} q_i c P_{i,c} \sum_{n \in N} \eta_{n,c} \rho_{n,c} h_{i,n}^2 + E\{\sigma_{m,c}^2\}}.$$  \hspace{1cm} (9)

$$\gamma_{m',c} = \frac{q_m c P_{m',c} \sum_{n \in N} \eta_{n,c} \rho_{n,c} h_{m',n}^2}{\sum_{i \in M, h_{i,c} \geq h_{m',n}} q_i c P_{i,c} \sum_{n \in N} \eta_{n,c} \rho_{n,c} h_{i,n}^2 + E\{\sigma_{m,c}^2\}}.$$  \hspace{1cm} (10)

$$\gamma_{m,c} = \frac{q_m c P_{m,c} \sum_{n \in N} \eta_{n,c} \rho_{n,c} h_{m,n}^2}{\sum_{i \in M, h_{i,c} \geq h_{m,n}} q_i c P_{i,c} \sum_{n \in N} \eta_{n,c} \rho_{n,c} h_{i,n}^2 + E\{\sigma_{m,c}^2\}}.$$  \hspace{1cm} (11)

Algorithm 1: PSMA based receiver

I: Set $b = 1$ as iteration number,
II: $H_{f,m,c} = \min H_c$
III: Apply MPA on $Y_j^f$
Output: $\sum_{k \in N} \eta_{k,c}^f s^f_{m',k}$ (signal of user $m'$ on codebook $c$ over BS $f$)
V: Apply SIC:
$Y = Y - \sum_{k \in N} p_{m',c,k}^f s^f_{m',k}$
VI: If $b = L_T$, apply MPA on $Y$ as signal of user $m$
else
set $b = b + 1$
set $H_c = H_c - \{H_{f,m,c}\}$
go back to II.

V. RECEIVER COMPLEXITY

In this section, the complexity of PD-NOMA, SCMA, and PSMA are investigated. The computational complexity of PD-NOMA and SCMA receiver is investigated in [15]. If we assume that in PSMA each codebook is assigned to $L_T$ users simultaneously, each user should be able to apply MPA $L_T$ times to detect and decode the transmitted data (Fig. 4). Therefore, if $G$ codebooks are assigned to a user, the complexity order of PSMA is approximately given by

$$O(I_T |\pi|^d(G)(L_T)),$$  \hspace{1cm} (16)

where $\pi$ shows the codebook set size, $I_T$ denotes the total number of iterations, and $d$ represents the non-zero elements in each row of the matrix $X$ where $X = (x_1, \ldots, x_n)$ is the factor graph matrix.

The receiver complexity order for PD-NOMA, SCMA and PSMA are summarized in Fig. 5. By setting the system parameters as demonstrated in Fig. 6, we can see that the PSMA complexity increases about an order of magnitude with respect to SCMA. However, as will be shown in simulation results, this comes with significant improvement in system throughput. In this table, $L_T$, and $G'$ show the total number of users that can be assigned to each user and the total number of subcarriers assigned to each subcarrier in PD-NOMA, respectively.

VI. RESOURCE ALLOCATION PROBLEMS FOR PSMA BASED CELLULAR NETWORKS

In this section, we consider resource allocation problems for the proposed PSMA framework to assess the performance compared to SCMA and PD-NOMA. To get a better insight, we first consider the simple case of a single cell system with equal proportion of power for each subcarrier. Then the unequal proportion of power is considered. Finally, a comprehensive HetNet system is analyzed.
### A. PSMA-BASED SINGLE CELL SYSTEM WITH EQUAL PROPORTION OF POWER FOR EACH SUBCARRIER

Here, we assume that $\eta_{m,c}$ for each subcarrier in codebook $c$ has equal value. We use the following Remark:

**Remark 1.** The interference cancellation (IC) of a candidate user (worse user) at user $m$ (better user) can be successfully performed if the (corresponding) SINR of the candidate user with less average channel gain, which is measured at user $m$, is more than its SINR on its receiver (17) chapter 6). For more clarification, suppose $m$ and $m'$ are two users with $\hat{h}_{m,c} \geq \hat{h}_{m',c}$. To achieve successful IC at user $m$, we should have $\gamma_{m',c}(m) \geq g_{m',c} \gamma_{m',c}(m')$ where $\gamma_{m',c}(m)$ is the SINR of user $m'$ at user $m$ on codebook $c$ and $\gamma_{m',c}(m')$ are, respectively, given by (9) and (10).

**Lemma 1.** With uniform power allocation, $(\eta_{m,c} = 1/U, \forall n \in N_c)$, the SIC constraint of Remark 1 is always satisfied.

**Proof.** For $\hat{h}_{m,c} > \hat{h}_{m',c}$, we can write $\gamma_{m',c}(m)$ and $\gamma_{m',c}(m')$ as (17) and (18), respectively. where $|w_{m,c}|^2$ and $|w_{m,c}''|^2$ are, respectively, given by

$$|w_{m,c}'|^2 = \frac{E[|\sigma_{m,c}|^2]}{\sum_{n \in N} p_{n,c} |\hat{h}_{m,n}|^2},$$

$$|w_{m,c}''|^2 = \frac{E[|\sigma_{m,c}|^2]}{\sum_{n \in N} p_{n,c} |\hat{h}_{m,n}|^2}.$$

As can be seen from (7) and $\hat{h}_{m,c} \geq \hat{h}_{m',c}$, we can conclude that $\gamma_{m',c}(m) \geq g_{m',c} \gamma_{m',c}(m')$. 

To formulate the optimization problem, we note that the rate of user $m$ on codebook $c$ is given by

$$r_{m,c}(P', Q') = \log(1 + \gamma_{m,c}), \quad (19)$$

where $\gamma_{m,c}$ indicates the SINR of user $m$ on codebook $c$ which is given by (11), $Q' = [q_{m,c}] \forall m \in M, c \in \mathcal{C}$ and $P' = [p_{m,c}] \forall m \in M, c \in \mathcal{C}$.

The problem formulation of this system model is formulated as follows:

$$\max_{Q', P} \sum_{m \in M} \sum_{c \in \mathcal{C}} r_{m,c}(P', Q') \quad (20a)$$

subject to

$$\sum_{m \in M} \sum_{c \in \mathcal{C}} q_{m,c} p_{m,c} \leq P_{\text{max}}, \quad (20b)$$

$$\sum_{m \in M} \sum_{c \in \mathcal{C}} q_{m,c} \rho_{m,c} \leq K, \forall n \in N', \quad (20c)$$

$$\sum_{m \in M} q_{m,c} \leq L_T, \forall c \in \mathcal{C}, \quad (20d)$$

$$p_{m,c} \geq 0, \forall m \in M, c \in \mathcal{C}, \quad (20e)$$

$$q_{m,c} \in \{0, 1\}, \forall m \in M, n \in N', \quad (20f)$$

where (20b) indicates the maximum available transmit power in BS, (20c) shows that each subcarrier can be reused at most $K$ times, and (20d) demonstrates that each codebook can be assigned to $L_T$ users simultaneously.

### B. PSMA BASED SINGLE CELL SYSTEM WITH UNEQUAL PROPORTION OF POWER FOR EACH SUBCARRIER

If we consider unequal power proportion for different subcarriers in a codebook, to achieve successful IC, we cannot use the concept used in the previous system model. Based on Remark 1, to achieve successful IC, in this system model, the following constraint should be applied:

$$\gamma_{i,c}(m) \geq q_{m,c} \gamma_{i,c}(i) \quad \forall i, m \in M, \hat{h}_{i,c} \geq \hat{h}_{m,c}. \quad (21)$$

The resource allocation problem is formulated as:

$$\max_{Q,P} \sum_{m \in M} \sum_{c \in \mathcal{C}} r_{m,c}(P, Q) \quad (22a)$$

subject to

$$(20b) - (20f), \quad (22b)$$

$$\gamma_{i,c}(m) \geq q_{m,c} \gamma_{i,c}(i) \quad \forall i, m \in M, \hat{h}_{i,c} > \hat{h}_{m,c}. \quad (22b)$$

### C. PSMA-BASED HETNET SYSTEM MODEL

Considering Remark 1, to achieve successful IC, the following constraint is applied:

---

![FIGURE 5](image_url) Receiver complexity order for PD-NOMA, SCMA and PSMA

<table>
<thead>
<tr>
<th>NOMA-complexity order</th>
<th>SCMA-complexity order</th>
<th>PSMA-complexity order</th>
</tr>
</thead>
<tbody>
<tr>
<td>$O((2L_T^d) + (2L_T^d)(G')(L_T^T - 1))$</td>
<td>$O((I_T(</td>
<td>I'</td>
</tr>
</tbody>
</table>

![FIGURE 6](image_url) Comparison between PD-NOMA, SCMA and PSMA receiver complexity ($N$ is the number of subcarriers, $I_T$ is the iteration number, $d$ is the non-zero elements in each row of the matrix $X$, $U$ is the number of subcarriers of each codebook, $L_T$ is the total number of users that can be assigned to each codebook in PSMA approach, $G$ is the total number of codebook assigned to each user in PSMA approach, $L_T'$ is the total number of users that can be assigned to each subcarrier in PD-NOMA approach, $G'$ is the total number of subcarriers assigned to each user in PD-NOMA approach)
and after simplifying, the linear constraint is obtained by
\begin{align}
\gamma_{m',c}(m) &= \frac{q_{m,c}P_{m,c} \sum_{n \in N} \eta_{m,c} \rho_{m,c} |h_{m,n}|^2}{\sum_{i \in M, h_{i,c} > h_{m,c}} q_{i,c} P_{i,c} \sum_{n \in N} \eta_{m,c} \rho_{m,c} |h_{m,n}|^2} + E\{|\sigma_{m,c}|^2\} \\
&= \frac{q_{m,c}1/U_{p,m,c}}{\sum_{i \in M, h_{i,c} > h_{m,c}} q_{i,c}1/U_{p,i,c} \sum_{n \in N} \rho_{m,c} |h_{m,n}|^2 + E\{|\sigma_{m,c}|^2\}} = \frac{q_{m,c}1/U_{p,m,c}}{\sum_{i \in M, h_{i,c} > h_{m,c}} q_{i,c}1/U_{p,i,c} + |w_{m,c}|^2}.
\end{align}

\begin{align}
\gamma_{m',c}(m') &= \frac{q_{m,c}P_{m,c} \sum_{n \in N} \eta_{m,c} \rho_{m,c} |h_{m,n}|^2}{\sum_{i \in M, h_{i,c} > h_{m,c}} q_{i,c} P_{i,c} \sum_{n \in N} \eta_{m,c} \rho_{m,c} |h_{m,n}|^2} + E\{|\sigma_{m,c}|^2\} \\
&= \frac{q_{m,c}1/U_{p,m,c}}{\sum_{i \in M, h_{i,c} > h_{m,c}} q_{i,c}1/U_{p,i,c} \sum_{n \in N} \rho_{m,c} |h_{m,n}|^2 + E\{|\sigma_{m,c}|^2\}} = \frac{q_{m,c}1/U_{p,m,c}}{\sum_{i \in M, h_{i,c} > h_{m,c}} q_{i,c}1/U_{p,i,c} + |w_{m,c}|^2}.
\end{align}

\[\gamma_{f,m,c}(j) \geq q_{f,j,c}^\prime \gamma_{m,c}(m) \quad \forall j, m \in M_f, f \in \mathcal{F}, h_{m,c}^\prime > h_{j,c}^\prime,\]  
where \(\gamma_{f,m,c}(m)\) shows the SINR of user \(i\) at user \(m\) in codebook \(c\) over BS \(f\).

The rate of user \(m\) on codebook \(c\) over BS \(f\) is given by
\[r_{m,c}(P, Q) = \log(1 + \gamma_{f,m,c}),\]
where \(\gamma_{f,m,c}\) indicates the SINR of user \(m\) on codebook \(c\) over BS \(f\) which is given by (13), \(Q = [q_{f,m,c}]\) \(\forall m \in M_f, f \in \mathcal{F}, c \in \mathcal{C}\) and \(P = [P_{f,m,c}]\) \(\forall m \in M_f, f \in \mathcal{F}, c \in \mathcal{C}\).

The proposed optimization problem formulation of the PSMA based HetNet system is written as:
\[\max_{Q, P} \sum_{f \in F} \sum_{m \in M_f} \sum_{c \in C} r_{m,c}(P, Q) = \sum_{f \in F} \sum_{m \in M_f} \sum_{c \in C} q_{f,m,c}^\delta P_{m,c} \leq p_{\text{max}} \quad \forall f \in \mathcal{F},\]  
\[\sum_{m \in M_f} \sum_{c \in C} q_{f,m,c}^\delta P_{m,c} \leq K, \quad \forall n \in N, f \in \mathcal{F},\]  
\[\sum_{m \in M_f} q_{f,m,c} \leq L_T, \quad \forall c \in \mathcal{C}, f \in \mathcal{F},\]  
\[\gamma_{f,m,c}(j) \geq q_{f,j,c}^\prime \gamma_{m,c}(m), \quad \forall j, m \in M_f, f \in \mathcal{F},\]  
\[h_{m,c}^\prime > h_{j,c}^\prime,\]  
\[p_{f,m,c} \geq 0, \quad \forall m \in M_f, c \in \mathcal{C}, f \in \mathcal{F},\]  
\[q_{f,m,c} \in \{0, 1\}, \quad \forall m \in M_f, n \in N, f \in \mathcal{F},\]

where (25b) indicates the maximum available transmit power in each BS, (25c) shows that each subcarrier can be reused at most \(K\) times, and (25d) demonstrates that each codebook can be assigned to \(L_T\) users simultaneously in each BS.

Remark 2. Constraint (25e) is a linear constraint in terms of power variables. To show the linearity of constraint (25e), we obtain equation (26) shown at the top of the next page and after simplifying, the linear constraint is obtained by (27) also shown at the top of the next page.

It should be mentioned that the problem formulation and solution methods of SCMA and PD-NOMA based systems are investigated in [15].

VII. SOLUTION ALGORITHM OF PROPOSED RESOURCE ALLOCATION PROBLEMS

We propose the solution for the problem corresponding to the PSMA-based HetNet which is the most general one. The solution for the the first two problems can be considered as special cases.

A. SOLUTION ALGORITHM

Problem (25) is non-convex and contains both integer and continuous variables. Therefore, the available methods to solve convex problems cannot be applied directly. To solve the proposed problem, an iterative algorithm based on the SCA method is applied and power and codebook are assigned separately in each iteration. The power allocation problem is non-convex, and the codebook allocation problem is integer non-linear programing (INLP). To solve the power allocation problem first, successive convex approximation with low complexity (SCALE) [20] and difference of two concave functions (DC) [22] is applied to approximate the problem by a convex one, then the dual method is used. To solve the codebook assignment problem, the mesh adaptive direct search (MADS) [25] algorithm is exploited. An overview of the iterative algorithm is shown in Algorithm 2.

Algorithm 2 Overview of the solution algorithm

I: Initialize \(Q(0), P(0)\) and set \(t = 0\) (iteration number).
II: Repeat:
III: Set \(Q = Q(t)\) and find a solution for problem (25) by applying the SCA approach and assign it to \(P(t + 1)\).
IV: Find \(Q(t + 1)\) by solving (25) with \(P = P(t + 1)\).
V: When \(\|P(t) - P(t - 1)\| \leq \delta\) stop.
Otherwise, set \(t = t + 1\) and go back to III.
Output:
\(P(k)\) and \(Q(k)\) are adopted for the considered system.
\[ \gamma_{m,c}^f(j) \geq q^f_{j,c} \gamma_{m,c}^f(m) \Rightarrow \]

\[ q^f_{j,c} \sum_{n \in N} \rho^f_{n,c} \eta^f_{n,c} h_{j,m}^f |h_{j,m}^f|^2 \]

\[ \sum_{j' \in \mathcal{F} / \{f\}} \sum_{m \in \mathcal{M}_f} \sum_{n \in N} q^f_{j,c} \rho^f_{m,c} \rho_{m,c} n,c h_{j,m}^f |h_{j,m}^f|^2 + \sum_{i \in \mathcal{M}_f} \sum_{h_{i,c} > h_{m,c}} \sum_{n \in N} q^f_{i,c} \rho^f_{m,c} \rho_{m,c} n,c h_{m,n}^f |h_{m,n}^f|^2 + E\{|\sigma_{m,c}^f|^2\} \]

\[ - (q^f_{j,c} \eta_{m,c}^f) \left( \sum_{n \in N} \sum_{j' \in \mathcal{F} / \{f\}} \sum_{m \in \mathcal{M}_f} \sum_{n \in N} q^f_{j,c} \rho^f_{m,c} \rho_{m,c} n,c h_{j,m}^f |h_{j,m}^f|^2 + \sum_{i \in \mathcal{M}_f} \sum_{h_{i,c} > h_{m,c}} \sum_{n \in N} q^f_{i,c} \rho^f_{m,c} \rho_{m,c} n,c h_{m,n}^f |h_{m,n}^f|^2 + E\{|\sigma_{m,c}^f|^2\} \right) \]

\[ + \sum_{i \in \mathcal{M}_f} \sum_{h_{i,c} > h_{m,c}} \sum_{n \in N} q^f_{i,c} \rho^f_{m,c} \rho_{m,c} n,c h_{m,n}^f |h_{m,n}^f|^2 + E\{|\sigma_{m,c}^f|^2\} \]  

\[ = (q^f_{j,c} \eta_{m,c}^f) \left( \sum_{n \in N} \sum_{j' \in \mathcal{F} / \{f\}} \sum_{m \in \mathcal{M}_f} \sum_{n \in N} q^f_{j,c} \rho^f_{m,c} \rho_{m,c} n,c h_{j,m}^f |h_{j,m}^f|^2 + \sum_{i \in \mathcal{M}_f} \sum_{h_{i,c} > h_{m,c}} \sum_{n \in N} q^f_{i,c} \rho^f_{m,c} \rho_{m,c} n,c h_{m,n}^f |h_{m,n}^f|^2 + E\{|\sigma_{m,c}^f|^2\} \right) \]

\[ + \sum_{i \in \mathcal{M}_f} \sum_{h_{i,c} > h_{m,c}} \sum_{n \in N} q^f_{i,c} \rho^f_{m,c} \rho_{m,c} n,c h_{m,n}^f |h_{m,n}^f|^2 + E\{|\sigma_{m,c}^f|^2\} \]  

\[ + 1 \log(1 + \lambda) \right), \]

\[ \xi = \frac{\lambda_0}{\lambda_0 + 1}, \psi_c = \log(1 + \lambda_0) - \frac{\lambda_0}{\lambda_0 + 1} \log(\lambda_0). \]

By exploiting this inequality, the objective function of problem (28) is written by:

\[ \sum_{f \in \mathcal{F}} \sum_{m \in \mathcal{M}_f} \sum_{c \in \mathcal{C}} \xi_{m,c}^f \log(\gamma_{m,c}^f) + \eta_{m,c}^f. \]

The objective function is still non-convex. By transforming \( p_{m,c} = \exp(\tilde{p}_{m,c}^f) \), the convex form of the objective function is achieved. However, using the mentioned transformation, constraint (25e) becomes a non-convex constraint. To tackle this issue, an approximation method such as DC [22] is applied to approximate (25e) by a convex function. Therefore, based on DC, approximation of constraint (25e) is written as (30a) < 0.

Consequently, a convex optimization problem in standard form with variables \( P \) is achieved as follows:

\[ \max_{P} \sum_{f \in \mathcal{F}} \sum_{m \in \mathcal{M}_f} \sum_{c \in \mathcal{C}} r_{m,c}^f \left( P_{m,c}^f \right) \]  

\[ \text{s.t.: } (25b), (25c). \]

In order to tackle the non-convexity issue of this problem, the following inequality (known as SCALE inequality) is applied [20]:

\[ \xi \log(\lambda^f) + \psi_c \leq \log(1 + \lambda^f), \]

where

\[ \lambda = \frac{\lambda_0}{\lambda_0 + 1}, \psi_c = \log(1 + \lambda_0) - \frac{\lambda_0}{\lambda_0 + 1} \log(\lambda_0). \]

To show the concavity of the objective function, we rewrite it as follows:

\[ 1 \]  

\[ = \sum_{f \in \mathcal{F}} \sum_{m \in \mathcal{M}_f} \sum_{c \in \mathcal{C}} \xi_{m,c}^f \left( \log(q_{m,c}^f \sum_{n \in N} \rho_{n,c}^f \eta_{n,c}^f |h_{m,n}^f|^2) + \tilde{p}_{m,c}^f \right) \]

\[ - \log \left( \sum_{f \in \mathcal{F} / \{f\}} \sum_{m \in \mathcal{M}_f} \sum_{n \in N} q_{m,c}^f \rho_{n,c}^f \eta_{n,c}^f \exp(\tilde{p}_{m,c}^f) |h_{m,n}^f|^2 + \sum_{i \in \mathcal{M}_f} \sum_{h_{i,c} > h_{m,c}} \sum_{n \in N} q_{i,c}^f \rho_{m,c}^f \rho_{m,c} n,c h_{m,n}^f |h_{m,n}^f|^2 \]

\[ + E\{|\sigma_{m,c}^f|^2\} \right) \]  

Each term in (32) is concave, and therefore, the new objective function is concave. We note that the log-sum-exp function is convex [21].

To deploy the SCALE algorithm in Algorithm 2, we use an iterative power allocation algorithm to find a power allocation not worse than \( P^t \) when \( Q = Q^t \). This procedure is presented as in Algorithm 3 where \( z \) indicates the iteration number and \( P^{t,z} \) shows the power allocation in iteration \( z \). In each iteration \( t \), the power allocation problem is solved. Moreover, \( \xi^z \) and \( \psi^z \) are updated as \( \xi^{z+1} \) and \( \psi^{z+1} \), respectively. The algorithm is initialized by \( \xi^0 = 1 \) and \( \psi^0 = 0 \), and continued until \( \|P^{t,z} - P^{t,z-1}\| \leq \epsilon \).
we have:

\[ \tilde{p}_{m,c} = \text{arg} \max \left\{ \sum_{n \in N} \eta_{n,c} |h_{m,n}|^2 \left( \sum_{f \in F} \sum_{m \in M_f} \sum_{n \in N} q_{f,c}^p (\exp \tilde{p}_{m',c}^{f,t-1} + \exp \tilde{p}_{m',c}^{f,t-1} - \tilde{p}_{m',c}^{f,t-1}) ) \rho_{n,c}^{f,t} \tilde{p}_{m,c}^{f,t-1} |h_{j,n}|^2 \right) \right\} \]  

(30a)

+ \sum_{i \in M_J, |h_{j,n}|^2 > |h_{m,n}|^2} q_{i,c}^p (\exp \tilde{p}_{i,c}^{f,t-1} + \exp \tilde{p}_{i,c}^{f,t-1} - \tilde{p}_{i,c}^{f,t-1} ) ) \rho_{n,c}^{f,t} \eta_{n,c} |h_{j,n}|^2 + E\{\sigma_{j,c}^f \} + ( \sum_{n \in N} \rho_{n,c}^{f,t} |h_{j,n}|^2 \}

(30b)

\[ L(\tilde{p}, \delta, \beta) = \sum_{f \in F} \sum_{m \in M_J} \sum_{c \in C} \xi_{f,c}^p \log(\gamma_{f,c}^p (\exp(\tilde{p}_{m,c}))) + \psi_{f,c}^p + \sum_{f \in F} \delta_f (\tilde{p}_{m,c}^f - \sum_{m \in M_J} \sum_{n \in N} \exp(\tilde{p}_{m,c}^f)) \]

(33)

Algorithm 3 ALGORITHM TO IMPROVE THE SIC METHOD

I: Set \( z = 0 \) and initialize \( \xi^0 = 1 \) and \( \psi^0 = 0 \),
II: Repeat:
III: Solve (31) then give the solution to \( P_{t,z} \),
IV: Update \( \xi^{z+1} \) and \( \psi^{z+1} \) with \( P_{t,z} \),
V: When \( \|P_{t,z} - P_{t,z-1}\| \leq \epsilon \) stop.
otherwise,
set \( z = z + 1 \) and go back to III.

To solve problem (32), we use the dual method. Therefore, the corresponding Lagrangian function is given by (33), where \( \delta \) and \( \beta \) are the Lagrange multipliers. The dual objective function is given by

\[ g(\delta) = \max_{\tilde{p}} L(\tilde{p}, \delta, \beta). \]  

(34)

To solve the dual problem, we should find the stationary point of (33) with respect to \( \tilde{p} \) where \( \delta \) and \( \beta \) are fixed. Therefore, we have:

\[ \frac{\partial L(\tilde{p}, \delta, \beta)}{\partial \tilde{p}_{m,c}} = 0. \]  

(35)

By simplifying (35), \( p_{m,c}^f \) is given by

\[ p_{m,c}^f = \left[ \frac{\xi_{f,c}^p + G_{m,c}^f}{\delta_f + A_{m,c}^f + B_{m,c}^f + C_{m,c}^f} \right]^+, \]  

(36)

where \([.]^+ = \max(\cdot, 0)\) and

\[ A_{m,c}^f = \sum_{i=m+1}^{M_J} \xi_{i,c}^f \gamma_{i,c}^{f}(p_{i,c}^f), \]

\[ B_{m,c}^f = \sum_{k \in F} \sum_{i \in M_J} \xi_{k} \sum_{n \in N} \rho_{n,c}^{f} \eta_{n,c} |h_{m,n}|^2 + \sum_{j \in M_J} \sum_{n \in N} \rho_{n,c}^{f} \eta_{n,c} |h_{j,n}|^2, \]

\[ C_{m,c}^f = \sum_{f \in F} \sum_{s \in M_J} \sum_{j \in M_J} \beta_{1s} \sum_{n \in N} \rho_{n,c}^{f} \eta_{n,c} |h_{m,n}|^2, \]

\[ G_{m,c}^f = \sum_{f \in F} \sum_{s \in M_J} \sum_{j \in M_J} \beta_{1s} \left( \sum_{n \in N} \rho_{n,c}^{f} \eta_{n,c} |h_{m,n}|^2 \right), \]

\[ \|P_{t,z} - P_{t,z-1}\| \leq \epsilon \]
To update the dual multipliers, the subgradient method is applied as follows:

\[
\delta_{f}^{u+1} = [\delta_{f}^{u} - \nu_{1}(p_{max}^{f} - \sum_{m \in M_{f}} \sum_{n \in N} q_{m,n}^{f} p_{m,n}^{f})]^{+},
\]

and

\[
\beta_{f,s}^{u+1} = [\beta_{f,s}^{u} - \nu_{2}((30a))]^{+},
\]

where \( u \) indicates the iteration number of updating dual multipliers, and \( \nu_{1} \) and \( \nu_{2} \) are small updating step-size.

The final procedure in the power allocation sub-problem is applied as follows:

1. Set \( u = 0 \) and initialize \( \delta^{0} \) and \( \beta^{0} \).
2. Repeat:
   a. Compute \( \mathbf{p} \) by applying (36).
   b. Update \( \delta \) and \( \beta \) by using (37) and (38), respectively,
   c. When \( \| \mathbf{P}^{u} - \mathbf{P}^{u-1} \| \leq \epsilon \) stop, otherwise, set \( u = u + 1 \) and go back to III.

### Algorithm 4 ALGORITHM TO FIND STATIONARY POINT

I: Set \( u = 0 \) and initialize \( \delta^{0} \) and \( \beta^{0} \),
II: Repeat:
III: Compute \( \mathbf{p} \) by applying (36).
IV: Update \( \delta \) and \( \beta \) by using (37) and (38), respectively,
V: When \( \| \mathbf{P}^{u} - \mathbf{P}^{u-1} \| \leq \epsilon \) stop,
otherwise, set \( u = u + 1 \) and go back to III.

1) Convergence

The SCA method with the SCLAE approximation algorithm generates a sequence of improved solution which converges to a locally optimal solution. It can be shown that solution of each iteration is a feasible solution for the next iteration and value of objective function either improves or stays unchanged as the previous iteration.

### C. CODEBOOK ASSIGNMENT

The problem of codebook assignment is formulated as:

\[
\max_{\mathbf{Q}} \sum_{f \in F} \sum_{m \in M_{f}} \sum_{c \in C} r_{m,c}^{f}(\mathbf{Q})
\]

\[
\text{s.t. : } (25b), (25e), (25g).
\]

Problem (39) is an INLP which can be solved using MADS algorithm. To apply MADS algorithm, available optimization software such as NOMAD solver [25] can be used.

### VIII. SIMULATION RESULTS

In this section, the system sum rate for PSMA, SCMA, and PD-NOMA is evaluated under different number of users and small cells. In the numerical results, the system parameters are set as follows: MBS radius is 1 Km, the SBSs radius is 20 m, \( N = 8 \), \( h_{m,c} = 1/2 \), \( f, c, m, S = 2 \), \( K = 6 \), \( h_{m,c}^{f} = x_{m,n}^{f}(d_{m}^{f})^{\mu} \) where \( \mu \) indicates the path loss exponent and \( \mu = -2 \), \( x_{m,n}^{f} \) indicates the Rayleigh fading coefficient, and \( d_{m}^{f} \) demonstrates the distance between user \( m \) and BS \( f \).

Fig. 7 shows the system sum rate versus the total number of users for PSMA, PD-NOMA, and SCMA where we assume that \( P_{max}^{f} = 30 \) Watts, \( P_{max}^{f} = 2 \) Watts for \( f \in \{2, \ldots, F\} \), and for PD-NOMA and PSMA, \( L_{T} = 3 \) users. As can be seen, PSMA significantly improves the system performance compared to the other NOMA approaches. For small number of users, PSMA provides about 30% more throughput. This increases by 50% for large number of users.

Fig. 8 shows the system sum rate versus the total transmit power for PSMA, PD-NOMA, and SCMA where we considered that the total number of users is 12, and for PD-NOMA and PSMA, we set \( L_{T} = 3 \) users. Again PSMA outperforms SCMA and PD-NOMA by a great margin.

Finally in Fig. 9, we investigated the effect of \( L_{T} \), i.e., the total number of users that can be assigned to a codebook simultaneously, on PSMA performance. As can be seen, PSMA with \( L_{T} = 1 \) exhibits a performance similar to SCMA. As \( L_{T} \) increases, the performance improves and we have roughly 25% improvement per each unit increase in \( L_{T} \). We, however, expect that this saturates at some point.

![Figure 7: System sum rate versus number of users.](image1)

![Figure 8: System sum rate versus total transmit power.](image2)

![Figure 9: System sum rate versus total transmit power.](image3)
IX. CONCLUSION

In this paper, we proposed a new MA technique for 5G which uses code and power domain to send multiple users’ signals in a subcarrier. We investigated the PSMA transmitter and receiver and compared it to other NOMA approaches from the aspect of receiver complexity and system performance. To this end, we proposed a novel resource allocation problem. To solve the proposed problem, we used an iterative algorithm based on the SCA approach where in each iteration, codebook assignment was solved by applying the MADS algorithm and power allocation was solved based on the SCALE and DC methods. Moreover, from simulation results, we concluded that the PSMA technique significantly outperforms other NOMA techniques while imposing a reasonable increase in complexity to the system. Future works include improving the robustness of PSMA decoders by taking into account factors such as CSI uncertainty, and developing more robust SIC ordering techniques. As a future work, we study the link level performance of PSMA based systems and compare it to PD-NOMA and SCMA based systems.

REFERENCES


FIGURE 9: System sum rate versus $L_T$ (total number of users that can be assign to a codebook.)
M. Moltafet et al.: A New Multiple Access Technique for 5G

NADER MOKARI received the Ph.D. degree in electrical engineering from Tarbiat Modares University, Tehran, Iran, in 2014. He joined the Department of Electrical and Computer Engineering, Tarbiat Modares University, as an Assistant Professor, in 2015. He has been involved in a number of large scale network design and consulting projects in the telecom industry. His research interests include design, analysis, and optimization of communication networks.

MOHAMMAD REZA JAVAN received his B.Sc. degree from Shahid Beheshti University, Tehran, Iran, and his M.Sc. degrees from Sharif University of Technology, Tehran, Iran, and his Ph.D. degree from Tarbiat Modares University, Tehran, Iran, all in Electrical Engineering in 2003, 2006, and 2013, respectively. At present, he is a faculty member of department of electrical engineering, Shahrood University, Shahrood, Iran. His research interests include design and analysis of wireless communication networks with emphasis on the application of optimization theory.

HAMID SAEEDI received the B.Sc. and M.Sc. degrees from the Sharif University of Technology, Tehran, Iran, in 1999 and 2001, respectively, and the Ph.D. degree from Carleton University, Ottawa, ON, Canada, in 2007, all in electrical engineering. From 2008 to 2009, he was a Post-doctoral Fellow with the Department of Electrical and Computer Engineering, University of Massachusetts, Amherst, MA, USA. In 2010, he joined the Department of Electrical and Computer Engineering, Tarbiat Modares University, Tehran. His current research interests include coding and information theory, wireless communications, and resource allocation in conventional, heterogeneous, and cognitive radio networks. Dr. Saeedi has been a technical program committee member of some international conferences, including the Vehicular Technology Conference 2012, the Wireless Communication and Networking Conference 2014, 2015, and 2016, and the International Conference on Communications 2015. He was a recipient of some awards, including the Carleton University Senate Medal for Outstanding Academic Achievement, the Natural Sciences and Engineering Council of Canada Industrial Research and Development Fellowship, and the Ontario Graduate Scholarship. He is currently an Editor of the IEEE COMMUNICATIONS LETTERS.

HOSSEIN PISHRO-NIK received the B.S. degree from the Sharif University of Technology, and the M.Sc. and Ph.D. degrees from the Georgia Institute of Technology, all in electrical and computer engineering. He is currently an Associate Professor of Electrical and Computer Engineering with the University of Massachusetts, Amherst, MA, USA. His awards include the NSF Faculty Early Career Development (CAREER) Award, the Outstanding Junior Faculty Award from UMass, and the Outstanding Graduate Research Award from the Georgia Institute of Technology.