Benchmarking Practical RRM Algorithms for D2D Communications in LTE Advanced

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Published online: 23 December 2014
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Abstract Device-to-device (D2D) communication integrated into cellular networks is an advanced tool to take advantage of the proximity of devices and allow for reusing cellular resources and thereby to increase the user bitrates and the system capacity. However, the introduction of D2D in legacy long term evolution (LTE) cellular spectrum requires to revisit and modify the existing radio resource management and power control (PC) techniques in order to fully realize the potential of the proximity and reuse gains and to limit the interference to the cellular layer. In this paper, we examine the performance of the legacy LTE PC tool box and benchmark it against an utility optimal iterative scheme. We find that the open loop PC scheme of LTE performs well for cellular users both in terms of the used transmit power levels and the achieved signal-to-interference-and-noise-ratio distribution. However, the performance of the D2D users as well as the overall system throughput can be boosted.

The work of G. Fodor has been partially performed in the framework of the FP7 project ITC 317669 METIS. G. Fodor has also been supported by the Swedish Foundation for Strategic Research Strategic Mobility SM13-0008 Matthew Project.

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by the utility optimal scheme, by taking better advantage of both the proximity and the reuse gains. Therefore, in this paper we propose a hybrid PC scheme, in which cellular users employ the legacy LTE open loop PC, while D2D users exploits the utility optimizing distributed PC scheme. We also recognize that the hybrid scheme is not only nearly optimal, and can balance between spectral and energy efficiency, but it also allows for a distributed implementation at the D2D users, while preserving the LTE PC scheme for the cellular users.

1 Introduction

Device-to-device (D2D) communication underlaying a cellular infrastructure, also known as LTE Direct, has the potential of increasing spectrum and energy efficiency as well as allowing new peer-to-peer services by taking advantage of the so called proximity, hop and reuse gains [1–6]. In fact, D2D communication in cellular spectrum is currently studied under the framework of 3rd Generation Partnership Project (3GPP) to facilitate proximity aware internetworking services [7], national security and public safety applications [8] and machine type communications [9,10].

Despite the promising advantages, D2D communications utilizing cellular spectrum pose new challenges, because relative to cellular communication scenarios, the system needs to cope with new interference situations. For example, in an orthogonal frequency division multiplexing (OFDM) system in which user equipments (UE) are D2D-capable in the LTE network, D2D communication links may reuse some of the OFDM time–frequency physical resource blocks (RB) in order to better capitalize on the available bandwidth. Due to the reuse, intracell orthogonality is lost and intracell interference can become severe due to the random positions of the D2D transmitters and receivers as well as of the cellular UEs communicating with their respective serving base stations (BS) [11,12]. To realize the promises of D2D communications and to deal with intra- and intercell interference, the research community has proposed a number of important radio resource management (RRM) algorithms (see Fig. 1).

Although the objectives of such algorithms may be different (including enhancing the network capacity [13], improving the reliability [14], minimizing the sum transmission power

![Fig. 1](image_url) A D2D candidate pair consists of a D2D transmitter and a D2D receiver that are in the proximity of each other. The mode selection (MS) algorithm needs to decide on one of 3 possible communication modes: cellular mode, D2D mode with dedicated resources, or D2D mode with reused resources. This latter case involves a decision on which D2D pair(s) is (are) sharing resources with which cellular UE (pairing).
[4], ensuring quality of service [15] or protecting the cellular layer (i.e. the cellular UEs) from harmful interference caused by the D2D layer [16]), there seems to be a consensus that the key RRM techniques include:

1. **Mode Selection (MS):** MS algorithms determine whether D2D candidates in the proximity of each other should communicate in **direct mode** using the D2D link or in **cellular mode** (i.e. via the BS) [6,17–19], see Fig. 1;

2. **Resource Allocation (RA):** Surprisingly, resource allocation in the sense of selecting particular OFDM RBs or frequency channels out of a set of available ones for each transmit–receive pair (cellular or D2D) is seldom addressed in the literature [11,20–22];

3. **Pairing:** In the D2D context, pairing refers to selecting the D2D pair(s) and at most one cellular UE that share (reuse) the same OFDM RB, similarly to multiuser MIMO techniques. Pairing is a key technique to achieve high reuse gains [4];

4. **Multiple Input Multiple Output (MIMO) Schemes:** Interference avoiding MIMO schemes have been proposed by [23]. Such schemes can be applied, for example for the cellular transmissions to avoid generating interference to a D2D receiver.

Apart from mode selection and resource allocation (i.e. RB selection), power control (PC) is a key technique to deal with intra- and intercell interference [15,16,24–26]. References [15] and [24] analyze the single (isolated) cell scenario and provide some basic insights into the impact of PC and RA. The authors in [16] study a multi-cell system focusing on a PC scheme that minimizes the interference from the D2D layer to the cellular users assuming that D2D users that operate in D2D mode reuse the cellular resources. The work reported in [25] evaluates the LTE PC scheme for the hybrid cellular and D2D system and concludes that PC needs to be complemented by mode selection, resource scheduling and link adaptation to properly handle intra- and intercell interference. Recently, the work in [26] has devised a power control mechanism to combat cross-tier interference from D2D transmissions and cellular transmissions, and analyzed link coverage probabilities.

In this paper, we examine the performance of the LTE power control scheme when applied to the hybrid cellular D2D system and compare it with the performance of a distributed power control scheme based on utility maximization, where dynamic resource allocation and mode selection are also exercised by the network.

As such, in this paper we recognize the importance of power control, mode selection, resource allocation and pairing and try to tackle the challenging problem of D2D communications in underlying cellular networks by jointly considering these key factors when evaluating performances. To the best of authors knowledge this is a novel step in current literature. Further, a purpose of this examination is to gain insight into the applicability of LTE PC for D2D communications by quantifying its performance with respect to a utility optimal scheme in a realistic multi-cellular system.

We structure the paper as follows. Section 2 contains a brief overview of the LTE PC toolkit that provides various options for D2D PC. Section 3 describes the system model and states some basic assumptions. Next, Sect. 4 elaborates on the signal-to-noise-and-interference-ratio (SINR) target setting and PC problem in the integrated cellular and D2D environment. Section 5 proposes a solution approach to the PC problem based on the convexification and decomposition of the problem. Section 6 describes the mode selection and resource allocation problem, while Sect. 7 develops a heuristic aiming at reducing intracell interference based on full path gain matrix knowledge at the base station, and two other heuristics that are applicable in real systems. The numerical results are presented and discussed in Sect. 8. Finally, Sect. 9 concludes the paper.
2 Power Control Options Based on LTE Mechanisms

It is natural to base a PC strategy for D2D communications underlaying an LTE network on the LTE standard uplink PC mechanisms [2]. Building on the already standardized and widely deployed schemes facilitates not only a smooth introduction of D2D-capable user equipment (UE), but would also help to develop inter-operable solutions between different devices and network equipments. However, due to intracell interference and new intercell interference scenarios, the question that naturally arises is whether the available LTE PC is suitable for D2D communications integrated in an LTE network. Also, the ad-hoc networking community has proposed efficient distributed schemes suitable for D2D communications, including situations with or without the availability of a cellular infrastructure (see e.g. [24,25,27,28]). Such schemes can also serve as a basis for D2D PC design [29,30].

The LTE PC scheme can be seen as a ‘toolkit’ from which different PC strategies can be selected depending on the deployment scenario and operator preference [31]. It employs a combination of open-loop (OL) and closed-loop power control to set the UE transmit power (up to a maximum level of $P_{MAX} = 24$ dBm) as follows:

$$P_{UE} = \min \left[ P_{MAX}, \frac{P_0 - \alpha \cdot G}{\text{OL operating point}} + \frac{\Delta_{TF} + f(\Delta_{TPC}) + 10 \cdot \log_{10} M}{\text{dynamic offset}} + \frac{\text{BW factor}}{M} \right], \quad (1)$$

where $G$ is the path gain between the UE and the BS. The OL operating point allows for path loss (PL) compensation and the dynamic offset can further adjust the transmit power taking into account the current modulation and coding scheme (MCS) and explicit transmit power control (TPC) commands from the network. The bandwidth factor takes into account the number of scheduled RBs ($M$). For the OL operating point, $P_0$ is a base power level used to control the SNR target and it is calculated as [32]:

$$P_0 = \alpha \cdot (\gamma_{tgt}^{\text{gt}} + P_{IN}) + (1 - \alpha) \cdot (P_{MAX} - 10 \cdot \log_{10} M), \quad (2)$$

where $\alpha$ is the PL compensation factor and $P_{IN}$ is the estimated noise and interference power. For the dynamic offset, $\Delta_{TF}$ is the transport format (MCS) dependent component, $f(\Delta_{TPC})$ represents the explicit TPC commands.

For the integrated D2D communications scenario, we consider the following options:

- No Power Control (NPC), reference case: With NPC, there is no fixed $\gamma_{tgt}^{\text{gt}}$ and the transmit power of the cellular UEs and D2D transmitters is set to some fixed value that is equal to or less than $P_{MAX}$ according to (1).\(^1\) For $M = 1$ this can be obtained by setting $\alpha = 0$ and $P_{MAX} = P_{fix}$ in (2).

- Fixed SNR target (FST): FST fully utilizes the LTE path loss compensation capability by setting $\alpha = 1$ and $P_0 = \gamma_{tgt}^{\text{gt}} + P_{IN}$, where $\gamma_{tgt}^{\text{gt}}$ is a predefined SNR target and $P_{IN}$ is the interference plus noise power (in practice, for simplicity, $P_{IN}$ can be set to a fixed value, e.g. $P_{IN} \approx -121 \ldots -116$ dBm).

- Open Loop with Fractional Path Loss Compensation (OFPC): The OFPC scheme allows users to transmit with variable power levels, depending on their path loss. In contrast to the FST-case, the OFPC compensates for the fraction of the path loss by setting $\alpha$ to some suitable value in the range $[0, 1]$, e.g. 0.4 …0.9.

- Closed Loop PC: Such scheme extends the FST scheme by adding the dynamic offset or tuning step $f(\Delta_{TPC})$ in (1) in order to compensate the measured SINR ($\hat{\gamma}$) at the receiver.

\(^1\) Note that (2) is valid only in the case when a $\gamma_{tgt}^{\text{gt}}$ value exists.
Fig. 2 The UE communicating in cellular mode with its serving BS uses the de facto standard LTE open loop with fractional path loss compensation (OFPC). For the D2D link, we study various power control strategies, including no power control (i.e., fixed transmit power), fixed SNR target, OFPC, and closed loop that can all be easily deployed using the flexible standard LTE PC tool box. In our hybrid scheme, the cellular UEs use LTE OFPC, while the D2D pairs use utility maximizing PC, as in [30]. For benchmarking purposes, both the cellular and the D2D users use utility maximization with the desired SNR target value. The tuning step can be computed as follows [25]:

\[
 f(\Delta_{TPC}) = \begin{cases} 
 |\gamma^{tgt} - \hat{\gamma}| / 2 & \text{if } |\gamma^{tgt} - \hat{\gamma}| > 2 \text{ dB} \\
 1 \text{ dB} & \text{otherwise}
\end{cases}
\] (3)

For UEs communicating in cellular mode with their respective serving base stations, OFPC provides a well-proven alternative, typically used in practice. It avoids the complexity and overhead associated with the dynamic offset of the closed loop scheme, but makes use of the fractional path loss compensation balancing between overall spectrum efficiency and cell edge performance [31]. Figure 2 illustrates the PC options for the D2D link, while we assume that the cellular link employs the de facto standard LTE fractional path loss compensating power control scheme.

The PC options available in the integrated cellular and D2D environment are summarized by Fig. 2. For cellular users, the LTE OFPC scheme is a viable option, while for D2D users we are interested in the performance of various PC alternatives, including those based on the LTE 'tool box' and utility maximization. We use the term 'hybrid power control' for the case when cellular UEs use LTE OFPC, while D2D users use the distributed PC scheme. As we will elaborate later, for benchmarking purposes, we employ the utility maximizing scheme for all (cellular and D2D) users.

3 System Model

In order to derive a benchmarking scheme for network assisted D2D communications, we model the hybrid cellular-D2D network as a set of \( L \) transmitter–receiver pairs. A transmitter–receiver pair can be a cellular UE transmitting data to its serving BS or a D2D pair communicating in cellular uplink spectrum. D2D candidates are source–destination pairs in the...
proximity of each other that may communicate in direct mode, depending on the MS decision that is part of the RRM algorithm that is investigated in Sect. 6.

The network topology is represented by a directed graph with links labelled with \( l = 1, \ldots, L \) indexing the transmitter–receiver pairs in the network. Any transmitter, i.e. either cellular or D2D transmitter, operating in the link \( l \) is assumed to have always data to send to the corresponding receiver at a transmission rate \( s_l \). Associated with each link \( l \) is a function \( u_l(\cdot) \), which describes the utility of communicating at rate \( s_l \). The utility function \( u_l \) is assumed to be increasing and strictly concave, with \( u_l \to -\infty \) as \( s_l \to 0^+ \). We let \( \mathbf{c} = [c_l] \) denote the vector of link capacities, which depend on the system bandwidth \( W \), the achieved SNR of the links \( (\gamma_l) \) as well as the specific modulation and coding schemes used for the communication. A feasible rate vector \( \mathbf{s} \) must fulfill the following set of constraints:

\[
\mathbf{s} \preceq \mathbf{c}(\mathbf{p}), \quad \mathbf{s} \succeq \mathbf{0}.
\]

In this formulation, it is convenient to look at the \( \mathbf{s} \) vector as the vector of the rate targets directly derived from a corresponding vector of SINR targets, while the capacity vector \( \mathbf{c} \) depends on the specific powers \( \mathbf{p} \) selected by the transmitters. Specifically, each link can be seen as a Gaussian channel with Shannon-like capacity

\[
\gamma_l(\mathbf{p}) = W \log_2 \left( 1 + K \gamma_l(\mathbf{p}) \right)
\]

which represents the maximum rate that can be achieved on link \( l \), where \( K \) models the SINR-gap reflecting a specific modulation and coding scheme and \( \gamma_l(\mathbf{p}) \) represents the SNR perceived at the receiver of link \( l \). With no loss of generality, we assume \( K = 1 \) in the following.

Let \( G_{lm} \) denotes the effective link gain between the transmitter of pair \( m \) and the receiver of pair \( l \) (including the effects of path-loss and shadowing\(^2\)) and let \( \sigma_l \) be the thermal noise power at the receiver of link \( l \), and \( P_l \) be the transmission power. The SNR of link \( l \) is

\[
\gamma_l(\mathbf{p}) = \frac{G_{ll} P_l}{\sigma_l + \sum_{m \neq l} G_{lm} P_m}
\]

where \( \mathbf{p} = [P_1, \ldots, P_L] \) is the power allocation vector, and \( \sum_{m \neq l} G_{lm} P_m \) is the interference experienced at the receiver of link \( l \). Equation (5) can also be written as

\[
\gamma_l(P_{tot}^{Rx}, P_l, G_{ll}) = \frac{G_{ll} P_l}{(P_{tot}^{Rx} - G_{ll} P_l)}
\]

where \( P_{tot}^{Rx} \) represents the total received power (including \( \sigma_l \)) measured by the receiver of link \( l \). Hence, the SINR in (6) can be computed by the receiver-l without direct knowledge of any of the channel gains, except the one related to its corresponding transmitter-l. For the ease of notation, in the following we adopt \( \gamma_l(\mathbf{p}) \) to indicate the SINR measured at receiver-l.

4 The SINR Target Setting and Power Control Problem

4.1 Formulating the D2D Power Control Problem

In this section, we assume that MS has already been performed for the D2D candidates and a RA algorithm has already assigned to cellular and D2D links certain RBs for communication.

\(^2\) We assume that the G matrix is obtained after Layer-1 filtering, that is typically used for open loop power control, mobility management and other purposes in LTE [31].
From the concept of D2D communications reusing cellular spectrum, a given RB may be used by multiple cellular and D2D transmitters even within the same cell, thus causing intracell interference. In this section we focus on handling such interference by properly setting the SINR targets and allocating transmit powers, while the MS and RA problems that determine the specific cellular and D2D transmitters that share a given RB are considered to be already been performed. MS and RA are investigated in more details in Sects. 6 and 7.

The framework that we adopt in this paper has been introduced in [33] and successively extended to the field of radio resource management in multi-hop wireless networks in [34] and [35]. However, previous works develop optimization algorithms that require extensive signaling to piggyback local UE information to the other UEs in the network. This would rapidly impair network performances and make it difficult to apply such approaches in a real-world scenario. In contrast, without compromising the mathematical soundness, we derive a practical optimization algorithm that is tailored for D2D cellular scenarios since it is completely distributed and does not require any coordination or message flooding scheme between UEs. A further benefit of the proposed algorithm is that it maintains the logical separation between layers in the protocol stack, i.e. a fast power control is executed at physical layer to adapt to the variations in traffic and channel conditions, while at the medium access control layer the transmitting rates are adjusted at a slower time-scale.

For the set of interfering links sharing the same RB and thereby causing interference to one another, we formulate the problem of target rate setting and PC as:

$$\text{maximize } \sum_{l} u_{l}(s_{l}) - \omega \sum_{l} P_{l}$$
subject to $s_{l} \leq c_{l}(p)$, $\forall l$, $p, s \geq 0$ (7)

which aims at maximizing the utility while taking into account the transmit powers by means of a predefined weight $\omega \in (0, +\infty)$ [35], so as to both increase spectrum efficiency and reduce the sum power consumption over all transmitters sharing a specific RB. Constraints of Problem (7) formally ensure that the rate allocation does not exceed the link capacities that in turn depends on the transmit powers on the given RB. As shown in Sect. 5, Problem (7) can be decomposed into two separate problems (Problem-I and Problem-II) that need to be executed recursively until convergence to the optimum of Problem (7). Specifically, Problem-I selects the transmit rate target, while Problem-II selects the transmit power that fulfills the desired transmit rate target, i.e. the SINR target. As such, Problem-I and Problem-II resemble an outer-loop and an inner-loop mechanism respectively, where the inner-loop power allocation ensures that the target rate $s$ reduces to the optimal capacity vectors ($c$) at convergence of the outer–inner loop routine.

4.2 Convexifying the Problem of Eq. (7)

Before presenting the decomposition approach, it is important to note that Problem (7) is not convex in its original formulation. However, by appealing to the results presented in [35] and [36], Problem (7) can be converted into the following equivalent form:

$$\text{maximize } \sum_{l} u_{l}(e^{\tilde{s}_{l}}) - \omega \sum_{l} e^{\tilde{P}_{l}}$$
subject to $\log(e^{\tilde{s}_{l}}) \leq \log(c_{l}(e^{\tilde{p}}))$, $\forall l$, (8)

where $s_{l} \leftarrow e^{\tilde{s}_{l}}$ and $P_{l} \leftarrow e^{\tilde{P}_{l}}$. The transformed Problem (8) is proved to be convex (now in the $\tilde{s}_{l}$-s and $\tilde{P}_{l}$-s) since the utility functions $u_{l}(\cdot)$ are selected to be $(\log, x)$-concave.
over their domains [35]. In this paper we use \( u_l(x) \triangleq \ln(x), \forall l \). Under this condition, we can solve Problem (8) to optimality by means of an iterative algorithm where the \( \tilde{s}_l \)-s (or equivalently the SINR targets) are set by an outer-loop. The transmit powers \( \tilde{P}_l \)-s that meet the particular SINR targets (set in each outer-loop cycle) are in turn set by a Zander type iterative SINR target following inner-loop [37]. This separation of the setting of the SINR targets and corresponding power levels are detailed in the next section.

5 A Decomposition Approach to the SINR Target Setting and Power Control Problem

5.1 Formulating the Decomposed Problem

We now reformulate Problem (8) as a problem in the user rates \( \tilde{s} \) (Problem-I), which, due to the convexification, can be solved for a given power allocation (\( \tilde{p} \)). Note that the target rate vector \( \tilde{s} \) can be uniquely mapped to a target SINR vector \( \gamma_{tgt} \) as it will be shown later. We define **Problem-I** as:

\[
\begin{align*}
\text{maximize} & \quad \nu(\tilde{s}) \\
\text{subject to} & \quad \tilde{s} \in \tilde{S} \quad (9)
\end{align*}
\]

where \( \tilde{S} = \{ \tilde{s} | \log(e^{\tilde{s}_l}) \leq \log(c_l(e^{\tilde{p}})), \forall l \} \) represents the set of feasible rate vectors that, for a given power vector \( \tilde{p} \), fulfill the constraints of Problem (8).

Comparing (8) and (9), it follows that the objective function in (9) is defined as \( \nu(\tilde{s}) \triangleq \sum_l u_l(e^{\tilde{s}_l}) - \varphi(\tilde{p}) \), where \( \varphi(\tilde{p}) \triangleq \omega \sum_l e^{\tilde{P}_l} \) represents the cost in terms of the total transmit power for realizing a given target rate \( \tilde{s} \). Accordingly, we denote with \( \varphi^*(\tilde{p}) \triangleq \omega \sum_l e^{\tilde{P}^*_l} \) the cost of achieving the optimum rates \( \tilde{s}^* \) that solve the utility maximization Problem (9).

Therefore, **Problem-II**, for a given \( \tilde{s} \) vector, can be formulated as

\[
\begin{align*}
\text{minimize} & \quad \omega \sum_l e^{\tilde{P}_l} \\
\text{subject to} & \quad \log(e^{\tilde{s}_l}) \leq \log(c_l(e^{\tilde{p}})) \quad \forall l.
\end{align*}
\]

Solution approaches to Problem-I and Problem-II are proposed in the next subsection.

5.2 Solving the Rate (SINR Target) Allocation Problem

Provided that the objective function \( \nu(\tilde{s}) \) in (9) is concave and differentiable we can determine the optimal \( \tilde{s}^* \) by means of **projected gradient** iterations, with a fixed predefined step \( \epsilon \):

\[
\tilde{s}^{(k+1)}_l = \max \left[ 0, \tilde{s}^{(k)}_l + \epsilon \nabla_l \nu(\tilde{s}^{(k)}) \right] \quad \forall l,
\]

where

\[
\nabla_l \nu(\tilde{s}) = \frac{\partial}{\partial \tilde{s}_l} \left[ \sum_l u_l(e^{\tilde{s}_l}) - \varphi^*(\tilde{p}) \right] = u_l'(e^{\tilde{s}_l})e^{\tilde{s}_l} - \frac{\partial}{\partial \tilde{s}_l} \left[ \varphi^*(\tilde{p}) \right].
\]

To compute (12), we first need to find \( \varphi^*(\tilde{p}) \) by solving the primal Problem-II (10). Since it is convex in \( \tilde{p} \), it can be conveniently solved by **Lagrangian Decomposition** as follows. Let \( \lambda \) be the Lagrange multipliers (dual variables) for the constraints in (10) and form the Lagrangian function:

\[
\mathcal{L}(\lambda, \tilde{p}) \triangleq \omega \sum_l e^{\tilde{P}_l} + \sum_l \lambda_l \left[ \log(e^{\tilde{s}_l}) - \log(c_l(e^{\tilde{p}})) \right].
\]

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The Lagrangian dual problem of Problem-II is given by:

\[
\text{maximize } \mathcal{L}(\lambda) = \min_{\tilde{p}} \mathcal{L}(\lambda, \tilde{p}) \\
\text{subject to } \lambda \succeq 0.
\] (14)

Since the original problem is convex, if regularity conditions hold the solution of Problem (14) corresponds to the solution of Problem (10), i.e. \( \mathcal{L}(\lambda^*) = \varphi^*(\tilde{p}) \). Assuming that \((\lambda^*, \tilde{p}^*)\) represents the optimum solution of Problem-II (10), we are now in the position to calculate \( \varphi^*(\tilde{p}) \) from (13):

\[
\varphi^*(\tilde{p}) = \sum_l \left[ \omega e^{\tilde{p}_l^\ast} - \lambda^*_l \log \left( c_l (e^{\tilde{p}_l^\ast}) \right) \right] + \sum_l \lambda^*_l \log (e^{\tilde{\gamma}_l}) \quad \text{and} \quad \frac{\partial}{\partial \tilde{s}_i} [\varphi^*(\tilde{p})] = \lambda^*_i.
\]

Recalling (12), we have:

\[
\nabla_i \psi(\tilde{s}) = u_i'(e^{\tilde{s}_i}) e^{\tilde{s}_i} - \lambda^*_i = e^{\tilde{s}_i} \left[ u_i'(e^{\tilde{s}_i}) - \frac{\lambda^*_i}{e^{\tilde{s}_i}} \right] = s_i \left[ u_i'(s_i) - \frac{\lambda^*_i}{s_i} \right],
\] (15)

The final target rate update is:

\[
s_i^{(k+1)} = e^{\tilde{s}^{(k+1)}_i} = s_i^{(k)} \exp \left( \epsilon \nabla_i \psi(\tilde{s}^{(k)}) \right).
\]

Combining the above with (12), we can write the SINR target setting rule in the following form:

\[
s_i^{(k+1)} = s_i^{(k)} \exp \left( \epsilon s_i^{(k)} \left[ u_i'(s_i^{(k)}) - \frac{\lambda^*_i(s_i^{(k)})}{s_i^{(k)}} \right] \right)
\] (16)

Equation (16) dictates the outer-loop mechanism for a certain transmitter \( i \). Specifically, at any iteration \((k + 1)\), Equation (16) determines the rate (and hence the SINR) that should be targeted during the next inner-loop PC. Following the decomposition approach, Eq. (16) requires the knowledge of the Lagrange multipliers \( \lambda^*_i \) associated with Problem-II, which can be found by solving the PC problem associated with the \((k)\)-th outer-loop iteration. We consider this specific problem in the next section.

5.3 Solving the Power Allocation Problem for a Given SINR Target

The inner-loop PC problem (Problem-II) takes as an input a certain SINR target that can be easily derived from Eq. (16). Given \( \tilde{s}^{(k)} \in \tilde{S} \), the constraints in (10) correspond to require that the SINR-s of the links exceed a target value, i.e.

\[
\log \left( e^{\tilde{s}_l} \right) \leq \log \left( c_l (e^{\tilde{p}}) \right) \quad \Leftrightarrow \quad \gamma_l^t(p) \geq \gamma_l^{tgt}(\tilde{s}^{(k)}) \quad \forall l,
\]

where \( \gamma_l(p) \) is defined in (5), and

\[
\gamma_l^{tgt}(\tilde{s}^{(k)}) \triangleq \frac{2}{e^{\tilde{s}_l}} - 1.
\] (17)

Therefore, Problem (10) can be rewritten as:

\[
\text{minimize } \omega \sum_l e^{\tilde{p}_l} \\
\text{subject to } \gamma_l(e^{\tilde{p}}) \geq \gamma_l^{tgt}(\tilde{s}^{(k)}) \quad \forall l,
\] (18)

3 We draw a box around equations that need to be implemented by a receiver or transmitter node, as will be summarized in Fig. 3.
and solved with an iterative closed-loop PC scheme [37]:

\[
P_l^{(t+1)} = \frac{\gamma_l^{tgt}(\tilde{s}_i)}{\gamma_l(p^{(t)})} P_l^{(t)}.
\]

(19)

Thus, for a given \(\gamma_l^{tgt}(\tilde{s}_i)\), the PC inner-loop (19) sets the transmit powers for each transmitter at step \((t+1)\), provided that the transmitter is aware of the SINR \(\gamma_l(p^{(t)})\) measured at the receiver in the previous step.

5.4 Determining the \(\lambda^*_l\)-s

We can now determine the \(\lambda^*_l\)-s for the outer-loop update (16) by exploiting the relationship between the optimal \(p^*_l\) and the associated Lagrange multipliers \(\lambda^*_l\)-s. To this end, we rewrite the constraints in (18) as:

\[
\frac{G_{ll} P_l}{\sigma_l + \sum_{m \neq l} G_{lm} P_m} - \gamma_l^{tgt} \geq 0 \Rightarrow P_l - \gamma_l^{tgt} \sum_{m \neq l} \frac{G_{im} G_{ll}}{G_{ll}} P_m - \gamma_l^{tgt} \frac{\sigma_l}{G_{ll}} \geq 0 \forall l.
\]

(20)

Furthermore, let \(H \in \mathbb{R}^{L \times L}\) and \(\eta \in \mathbb{R}^L\) be defined as follows:

\[
H = [h_{lm}] \triangleq \begin{cases} 
-1 & \text{if } l = m \\
\gamma_l^{tgt} G_{lm} / G_{ll} & \text{if } l \neq m
\end{cases}
\]

(21)

\[
\eta = [\eta_l] \triangleq \left[ \gamma_l^{tgt} \sigma_l / G_{ll} \right].
\]

Using this notation, we can reformulate Problem (18) as the following Linear Programming (LP) problem:

\[
\begin{align*}
\text{minimize} & \quad \omega^T p \\
\text{subject to} & \quad Hp \preceq -\eta; \quad p \succeq 0,
\end{align*}
\]

(22)

with the corresponding Dual Problem

\[
\begin{align*}
\text{maximize} & \quad \eta^T \lambda^{(LP)} \\
\text{subject to} & \quad H^T \lambda^{(LP)} \succeq -\omega \mathbf{1}; \quad \lambda^{(LP)} \succeq 0
\end{align*}
\]

(23)

which is necessary to compute the Lagrange multipliers in Eq. (16) for the rate update.

As it is shown in “Appendix 1”, the inequality constraints in (23) can be rewritten explicitly as:

\[
\frac{\lambda_j^{(LP)}}{\omega} - \sum_{k \neq l} \frac{G_{kl}}{G_{kk}} \frac{\gamma_k^{tgt}}{\omega} \frac{\lambda_k^{(LP)}}{\omega} \leq 1, \quad \forall l.
\]

(24)

As it is shown in “Appendix 2”, by defining

\[
\mu_l \triangleq \frac{\lambda_j^{(LP)}}{\omega} \frac{\gamma_l^{tgt} \sigma_l}{G_{ll}} = \frac{\lambda_j^{(LP)}}{\omega} \eta_l,
\]

(25)

Equation (24) can be interpreted as an SINR requirement, i.e.

\[
\gamma_l^{cc}(\mu) \triangleq \frac{\mu_l G_{ll}}{\sigma_l + \sum_{k \neq l} \frac{G_{kl}}{\sigma_k} \mu_k} \leq \gamma_l^{tgt}, \quad \forall l.
\]

(26)
Therefore, Problem (23) can be reformulated as:

$$\text{maximize}_{\mu} \omega \mathbf{1}^T \mu$$

subject to \( \gamma_{\ell}^{cc} \leq \gamma_{\ell}^{tgt}, \forall \ell; \mu \geq 0 \) (27)

where the solution \( \mu \) can be computed according to the following distributed closed-loop PC similarly to Eq. (19)

$$\mu_{\ell}^{(t+1)} = \frac{\gamma_{\ell}^{tgt}}{\gamma_{\ell}^{cc}(\mu^{(t)})} \mu_{\ell}^{(t)} \forall \ell.$$ (28)

Equation (28) can be interpreted as a reverse link PC problem that is executed in the control channel between the receiver and the transmitter of link \( \ell \). Specifically, the receiver-\( \ell \)-adapted its transmitting power \( \mu_{\ell} \) according to Eq. (28), while the transmitter-\( \ell \)-measures the experienced SINR \( \gamma_{\ell}^{cc} \) in the corresponding control channel.

Once the iterative procedure (28) converges to the optimum \( \mu^* \), the optimal dual variables \( \lambda^{*(LP)}_{\ell} \) can be retrieved from Eq. (25) as

$$\lambda^*_{\ell} = \omega \mu^*_{\ell} \eta_{\ell}^{-1}, \forall \ell.$$ (29)

The original nonlinear PC problem (10) and the corresponding LP formulation (22) are equivalent in the sense that there exists the following specific relation between their optimal solutions \( (\hat{p^*}, \lambda^*) \) and \( (p^*, \lambda^{*(LP)}) \):

$$P_{\ell}^* = e^{\hat{p}_{\ell}^*} \forall \ell$$

$$\lambda_{\ell}^* = \log(1 + \gamma_{\ell}^{tgt}) \frac{1 + \gamma_{\ell}^{tgt}}{\gamma_{\ell}^{tgt}} P_{\ell}^* \log(2) \lambda^{*(LP)}_{\ell} \forall \ell.$$ (30)

Hence, once both \( P_{\ell}^* \) and \( \mu_{\ell}^* \) are achieved by means of Eqs. (29) and (30), we are able to compute \( \lambda^* \) as

$$\lambda_{\ell}^* = \log(1 + \gamma_{\ell}^{tgt}) \frac{1 + \gamma_{\ell}^{tgt}}{\gamma_{\ell}^{tgt}} P_{\ell}^* \log(2) \omega \mu_{\ell}^* \frac{G_{ll}}{\sigma_{\ell} \gamma_{\ell}^{tgt}} \forall \ell.$$ (31)

Equation (31) is then used to update the user rates in Eq. (16).

5.5 Summary

In this section, we have explored an outer–inner loop iterative solution for the convex optimization Problem (8). The basic idea is to decompose Problem (8) into separate subproblems in \( \hat{s} \) (Problem-I (9)) and \( \hat{p} \) (Problem-II (10)). For each link \( \ell \), Problem-I and Problem-II operate in concert as show in Fig. 3. Problem-I is in charge of the outer-loop iterations, while Problem-II deals with the inner-loop PC. More specifically, the solution of Problem-I at step \( (k) \), i.e. \( \gamma_{\ell}^{tgt} \left( s_{\ell}^{(k)} \right) \), serves as input of Problem-II that is executed until convergence to \( \mu_{\ell}^* \) and \( P_{\ell}^* \). In turn, Problem-II outputs \( \lambda_{\ell}^* \) that is used by a new instance of Problem-I at step \( (k+1) \). It is important to note that given the constraints of Problem-I, Problem-II is always provided with a set a feasible SINR target that can be achieved in a finite number of iterations by finite values of \( \mathbf{p} \). In other words, the solutions of Problem-II in Eq. (16) always move within the rate feasibility region \( \hat{S} \), provided that the step size \( \epsilon \) is small enough. In a setting at step \( (k = 0) \), the outer-loop can be initiated with a low feasible SINR target vector that
allows the inner-loop to determine in a finite number of iterations the finite transmit power levels $p^*$ and the corresponding $\lambda^*$.

6 The Mode Selection (MS) and Resource Allocation (RA) Problem

6.1 Basic Considerations for Mode Selection and Resource Allocation

While cellular UEs communicate with their respective serving BS, D2D-capable UEs may communicate either in direct mode with their respective D2D pairs or in cellular mode with the serving BS. In the direct mode case, D2D transmitters are allowed either to reuse cellular RBs, i.e. D2D reuse mode, or allocated orthogonal (dedicated) RBs, i.e. D2D dedicated mode [4,6]. In the latter case, the reuse gain of D2D communications is not harvested.

On the other hand, when a D2D-capable UE communicates in cellular mode, D2D communication reduces to the ordinary cellular communication and RA follows the legacy OFDMA allocation strategy, i.e. RBs are allocated orthogonally between all UEs. Therefore, three different communication modes can be considered for D2D communications: D2D mode with dedicated resources, D2D mode reusing cellular resources and cellular mode. We note that when the D2D candidate pairs communicate in cellular mode, downlink resources need to be allocated for the BS-D2D receiver link. For the sake of ease, downlink resource usage is not modeled in this paper.

We now consider a cellular system with $N$ cellular UEs and $M$ D2D transmitters and corresponding $M$ D2D receivers belonging to the sets $\mathcal{N}$ and $\mathcal{M}$, respectively such that the total number of users in a cell is $L = N + M$. We denote with $x_{l,j}(q)$ that indicate whether a transmitter–receiver pair $l$ is assigned to RB-$j$ in communication mode $q$, where $(q = 0)$ denotes cellular mode and $(q = 1)$ the D2D direct mode. By definition, any cellular UE $n \in \mathcal{N}$ always transmits in the cellular mode $(q = 0)$, while a D2D candidate can be forced either to operate using the direct link $(q = 1)$, or the cellular mode $(q = 0)$, or adaptively switch between the direct and cellular link according to a specific MS algorithm. With this terminology at hand, we can formulate the resource constraints as follows:
– Forced D2D mode:
\[ x_{m,j}(q) = x_{m,j}(1), \quad \forall m \in \mathcal{M} \quad \text{and} \quad \sum_{n \in \mathcal{N}} x_{n,j}(0) \leq 1, \quad \forall j \]

– Forced cellular mode:
\[ x_{m,j}(q) = x_{m,j}(0), \quad \forall m \in \mathcal{M} \quad \text{and} \quad \sum_{n \in \mathcal{N}} x_{n,j}(0) + \sum_{m \in \mathcal{M}} x_{m,j}(0) \leq 1, \quad \forall j \]

– Adaptive MS:
\[ \sum_{q} x_{m,j}(q) \leq 1, \quad \forall m \in \mathcal{M} \quad \text{and} \quad \sum_{n \in \mathcal{N}} x_{n,j}(0) + \sum_{m \in \mathcal{M}} x_{m,j}(0) \leq 1, \quad \forall j \]

where the last inequality indicates that a specific D2D candidate pair \( m \) can only be either in D2D or cellular mode when using RB- \( j \). Note that formally a specific D2D candidate pair \( m \) is allowed to use some RBs in D2D mode and other RBs in cellular mode.

6.2 Formulating the Mode Selection and Resource Allocation Problem

We now formulate the problem of allocating RBs to users (both cellular UEs and D2D pairs) \( l \in \mathcal{L} \), and selecting the appropriate communication mode (\( q \)) for the D2D pairs in order to take advantage of the potential proximity. More specifically, the RA task is formulated as a single cell-based optimization problem that maximizes the overall spectral efficiency for a given power allocation vector. The spectral efficiency for transmitter–receiver pair \( l \) on a given RB- \( j \) can be defined as \( \eta_{l,j} = \log_2 \left( 1 + \frac{G_{l,l,j} P_l}{\sigma + I_{l,j}} \right) \). Hence, it depends on the path gain \( G_{k,l,j} \) between transmitter- \( k \) and receiver- \( l \) on the RB- \( j \) and the intracell interference \( I_{l,j} = \sum_{k \neq l} P_l \cdot G_{k,l,j} \), due to the possible RB sharing between D2D pairs and cellular-UEs.

Thus, the user assignment task becomes:

\[
\begin{align*}
\text{maximize} & \quad \sum_{l} \sum_{j} \log_2 \left( 1 + \frac{G_{l,l,j} P_l \cdot x_{l,j}(q)}{\sigma + I_{l,j}} \right) \\
\text{subject to} & \quad \sum_{l} x_{l,j}(0) \leq 1 \quad \forall j & \text{(C1)} \\
& \quad \sum_{q} x_{l,j}(q) \leq 1, \quad \forall l, j & \text{(C2)} \\
& \quad x_{n,j}(1) = 0 \quad \forall n \in \mathcal{N}, j & \text{(C3)} \\
& \quad x_{l,j}(q) \in \{0, 1\} & \text{(C4)}
\end{align*}
\]

The constraints (C1) indicate that each RB can be allocated to at most one user in cellular mode due to the orthogonality constraint. Constraints (C2) ensure that to each user is assigned only one of the two possible modes. By definition, cellular UEs must not be assigned to mode \( (q = 1) \) (C3).

7 Heuristic Algorithms to Solve the Resource Allocation Problem

7.1 The MinInterf Algorithm

Given the NP-hardness of Problem (32) even in legacy OFDMA cellular communications [38], in order to obtain benchmarking results, we first propose a centralized procedure based on the full knowledge of the path loss measurements between all transmitters and receivers within the cell [39,40]. This scheme, that we call MinInterf, exploits the proximity between D2D candidates for MS, and performs RA that aims at reducing the intracell interference by minimizing the sum of the harmful path gains as will be shown in Eq. (33) and in Algorithm 1.
An example of a D2D transmission with dedicated resource. The D2D Tx node selects the transmission-mode (Cellular Mode or D2D Mode) according to the shadowed path loss measurements towards the D2D Rx node and towards the BS. If the channel gain between the D2D pair is higher than the one towards the BS, then the D2D Mode is preferred.

MinInterf involves two steps. Firstly, orthogonal resources are allocated to cellular UEs employing legacy RA schemes. Next, for each D2D candidate in the cell, MinInterf considers two possible cases:

- **D2D transmission with dedicated resource.** If there are orthogonal resources left, they can be assigned to the D2D candidate so that the D2D transmission does not affect others within the same cell. In this case, the D2D transmitter selects the best communication mode (i.e. Cellular Mode or D2D Mode) on the basis of the path gains both towards the D2D receiver \(G_{d2d_{Mode}}\) and the BS \(G_{CellularMode}\). Specifically, if \(G_{CellularMode} \leq G_{d2d_{Mode}}\), then the direct mode is preferred. (See Fig. 4.)

- **D2D transmission with resource reuse** (Fig. 5). When there are no unused RBs in the cell, the D2D pair must communicate in direct mode (D2D Mode) and reuse RBs. Sharing resources with other users within the same cell produces intracell interference. To reduce intracell interference, for each resource- \(j\) MinInterf considers the sum

\[
S(j) = G_{2T_{x,1Rx,j}} + G_{1T_{x,2Rx,j}} \quad [\text{dB}]
\]

(33)

as a measure of the potential interference that assigning the D2D-pair to resource- \(j\) causes. Here \(G_{2T_{x,1Rx,j}}\) represents the path gain between the D2D transmitter and the receiver of link(s) already allocated to resource- \(j\), which may be the cellular BS and/or other D2D receiver(s). \(G_{2T_{x,1Rx,j}}\) takes into account the interference that the D2D pair produces transmitting on RB- \(j\). \(G_{1T_{x,2Rx,j}}\), on the other hand, is the path gain between the transmitter(s) already allocated to RB- \(j\) (which can be both a cellular-UE and/or other D2D transmitters) and the receiver of the new D2D pair to be allocated. \(G_{1T_{x,2Rx,j}}\) is therefore related to the interference that the D2D receiver will experience due to the reuse. Once expression (33) is computed for each available resource- \(j\), the D2D pair is assigned to that resource corresponding to the minimum value. (See Fig. 5.)
Fig. 5 An example of a D2D transmission with resource reuse. D2D Tx node communicates directly to its D2D Rx node sharing a resource block (RB) with the cellular user UE. The shared RB is selected in such a way to minimize an estimate (Eq. (33)) of the intracell interference that D2D communication might perceive (related to the gain $G_{1Tx\_2Rx}$ between the UE and the D2D Rx node) and produce (related to the gain $G_{2Tx\_1Rx\_RB2}$ between the D2D Tx node and the BS).

It is worth noting that the final RA achieved by MinInterf represents a suboptimal solution of Problem (32), nevertheless numerical results show that its interplay with the iterative PC procedure allows to attain good performance in terms of spectrum and energy efficiency. **Algorithm 1** summarizes the main steps of the MinInterf scheme.

**Algorithm 1** MinInterf

Allocate orthogonal resources (RB) to cellular-UEs (using legacy algorithms)

for Each D2D candidate do
  if there is an orthogonal resource-$l$ left then
    if $G_{Cellular\_Mode} \leq G_{d2d\_Mode}$ then
      D2D candidate transmits in $D2D\_Mode$ on resource-$l$
    else
      D2D candidate transmits in $Cellular\_Mode$ on resource-$l$
    end if
  else /* Resource Reuse */
    for Each available resource-$j$ do
      $S(j) = [G_{2Tx\_1Rx\_j} + G_{1Tx\_2Rx\_j}]$
    end for
    D2D candidate transmits in $D2D\_Mode$ on resource-$j$ corresponding to the minimum value of $S$
  end if
end for
Algorithm 2 Balanced Random Allocation (BRA) and Cellular Protection Allocation (CPA)

BS maintains variables $\rho_j$ and $\rho_{MIN}$ as explained below

BS maintains (from e.g. D2D discovery algorithms) the D2D candidates for mode selection and the value of large scale fading (path gain) $G_{d2dMode}$ between each D2D candidate, which can be updated on the slow time scale (e.g. 500 ms).

$\rho_j = 0$ for all resource blocks RB-$j$ ($j = 1 \ldots R$)

if there are cellular UEs in the cell then
  Allocate orthogonal resources (RB) to cellular-UEs (using legacy algorithms)
  Set $\rho_j = 1$ for RB:s assigned to UEs
/* For CPA: Store $g(j)$, where $g(j)$ is the path gain between cellular UE using RB-$j$ and the serving BS */
end if

for each D2D candidate do

$\rho_{MIN} := \min_{j=1 \ldots R} \rho_j$ where $R$ is the total number of resource blocks

if $\rho_{MIN} == 0$ then
/* there is an orthogonal resource-$l$ left: Schedule D2D on orthogonal resource-$l$ */
  if $G_{CellularMode} \leq G_{d2dMode}$ then
    D2D candidate transmits in D2D-Mode on resource-$l$
  else
    D2D candidate transmits in Cellular-Mode on resource-$l$
/* For CPA: Store $g(l)$, where $g(l)$ is the path gain between the D2D transmitter in cellular mode using RB-$l$ and the serving BS */
  end if
else
/* Resource Reuse */
  Select a resource-$j$ out of the resources for which $\rho_j == \rho_{MIN}$ (e.g. based on channel state/quality information)
/* For the CPA algorithm: Substitute the above by: Pick the resource-$j$ out of the resources for which $\rho_j == \rho_{MIN}$ for which $j = \arg \max g(j)$, where $g(j)$ is the path gain between the cellular transmitter using RB-$j$ and the BS */
  D2D candidate transmits in D2D-Mode on resource-$j$
end if

Increment $\rho_j$

end for

7.2 Practical MS and RA Algorithms with Limited or No Channel State Information: BRA and CPA

While MinInterf can serve as a tool to benchmark RA algorithms, it cannot be employed in practice because it relies on a full $G$ matrix knowledge in the “Resource Reuse” branch of the Algorithm 1. Therefore we seek viable alternatives to MinInterf. Our first proposed algorithm operates without any path loss knowledge but keeps track of the reuse factors $\rho_j$ of each RB as described by the pseudo code of the “Balanced Random Allocation” (BRA, Algorithm 2). $\rho_j$ is a counter associated with resource-$j$ that counts the number of intracell transmitters using that resource.4

Our second proposed practical algorithm is called “Cellular Protection Allocation” (CPA). CPA takes advantage of the knowledge of the path gains between any cellular transmitter (i.e. cellular UE or D2D candidate operating in cellular mode) and the BS, that is available in practice due to measurement reports by the UE. As indicated in the pseudo code of Algorithm 2, a D2D transmitter that reuses a cellular RB is assigned to the particular RB used by a cellular UE that has the strongest cellular link. The rationale for this heuristic is that a cellular UE

4 We note that BRA can be made completely distributed by skipping the usage of $\rho_j$ in the algorithm. Simulation results (not shown here) indicate that the impact of skipping $\rho_j$ in BRA is not significant.
with a strong cellular connection with its serving BS can be expected to tolerate intracell interference caused by D2D resource reuse.

8 Numerical Results

8.1 Simulation Setup and Parameter Setting

In this section we consider the uplink of a 7-cells system, in which the number of uplink (UL) physical RB is 8 per cell. We perform Monte Carlo experiments to build performance statistics when employing the MinInterf, CPA and BRA resource allocations together with the utility maximizing, as well as with the LTE based PC or hybrid PC. In the hybrid scheme, the cellular UEs use the LTE de-facto standard OFPC while the D2D users use the utility maximizing scheme. In the hybrid scheme, the D2D transmitters are assumed to know their path gains towards the cellular BS (using LTE legacy reference signals for channel measurement) and limit their transmit power levels such that the interference perceived at the BS remains under the parameter $I^\star$.

In each cell, we drop 6 cellular UEs that communicate with their serving BSs using 1 UL RB. In addition, 6 D2D candidate pairs are also dropped in the coverage area of each cell. For the D2D users the system may select the D2D mode or cellular mode to communicate, as described in Sect. 6. When a D2D pair uses the cellular mode, the D2D transmitter sends data to the BS in the UL spectrum, and the BS relays such data to the D2D receiver in the DL band. In our study, we do not model the DL transmission, essentially assuming that the DL resources are in abundance so that we can focus on the UL performance. Instead, when the D2D pair communicates in the direct mode, the D2D transmitter sends data to the D2D receiver using UL resources. This case is referred to ‘MS’ to emphasize the role of the mode selection for the D2D candidates.

Since 6 D2D pairs are dropped in addition to the 6 cellular users, at most 2 of them can use direct mode with no reuse while the remaining 4 D2D candidates will employ direct mode and reuse resources with either other D2D direct mode users or with cellular users. This is because we assume 8 RBs per cell accommodating 12 transmitters and we assume that cellular users and D2D candidates in cellular mode must remain orthogonal within a cell as in legacy LTE OFDMA framework. Intuitively, we expect some SINR degradation on the reused resources, but an increase in the total rate, i.e. spectrum efficiency, due to a better exploitation of available bandwidth. This allows to distinguish the reuse gain achieved by D2D pairs operating in D2D direct mode while reusing cellular resources from the proximity gain achieved by D2D pairs operating in D2D direct mode with no reuse.

When the utility maximizing PC is used, all (cellular and D2D) transmitters employ the outer–inner loop based power control described in Fig. 3. In contrast, when the LTE PC is used, the cellular UEs use standard LTE OFPC method, whereas we test fixed SNR target, fixed transmit power and the closed loop based method for the D2D link.

The main simulation parameters are given in Table 1.

8.2 Numerical Results

Figures 6 and 7 compare the SINR performance of MinInterf, CPA and BRA for the cellular UEs and D2D pairs when using the utility maximizing and the LTE OFPC, respectively.

When using the utility maximizing PC (both for the cellular UEs and D2D pairs), the SINR performance of the MinInterf, CPA and BRA resource allocation schemes is very
Table 1  Parameters of the 7-cell system under study

| Parameter                                      | Value                                      |
|------------------------------------------------|--------------------------------------------|
| System bandwidth                              | 5 MHz                                      |
| Carrier frequency                             | 2 GHz                                      |
| Gain at 1 m distance                          | $-37$ dB                                   |
| Thermal noise $\sigma$                        | $-114$ dBm                                 |
| Path loss coefficient                         | 3.5                                        |
| Lognormal shadow fading                       | 6 dB                                       |
| Cell radius                                    | 500 m                                      |
| Number of cells                                | 7                                          |
| Max Tx power                                   | 200 mW                                     |
| Min Tx power                                   | $5 \times 10^{-6}$ W                      |
| $P_{IN}$                                       | $-116$ dBm                                 |
| $\alpha$                                       | 0.8                                        |
| Number of RB’s requested by users              | 1                                          |
| Max. number of outer-loop iterations           | 100                                        |
| Max. number of inner-loop iterations           | 10                                         |
| Number of MonteCarlo simulations              | 100                                        |
| Initial power                                  | 0.01 W                                     |
| Initial $\gamma_{tg}$                         | 0.2                                        |
| Initial $\mu$                                 | 0.01                                       |
| $\epsilon$                                    | 0.05                                       |
| $\omega$                                      | 0.01 ... 10                                |
| Distance between D2D pairs                    | 50–100 m                                   |
| Maximum interference caused by D2D users at the BS | $I^* = 0.02 \cdot N_0 \ldots 500 \cdot N_0$ |

similar, except in the high SINR region of the cellular users, that could gain a 6–8 dB SINR increase with MinInterf as compared to CPA. Somewhat surprisingly, BRA gets closer to the performance of MinInterf in this region at the expense of performing a bit worse for cell edge users than CPA. The reason for this is that CPA tends to reuse the RBs of strong (cell center) cellular users.

With the LTE PC, MinInterf shows gains over CPA and BRA for cellular UEs in the low SINR regime (up to 5 dB gain), essentially protecting the cell edge UEs from excessive interference from the D2D traffic. For example, the gain of MinInterf at the 50% percentile is only around 1–2 dB (see Fig. 7). In scenarios in which the cellular UEs are further away from their respective serving BSs, the gain of MinInterf is greater, but still typically remains under 3 dB difference. Also, somewhat surprisingly, there is no notable difference between the performance of the CPA and BRA allocations. However, comparing the D2D SINR distributions of Figs. 6 and 7, we observe a significant gain obtained by the utility maximizing scheme in the range of 5–8 dB throughout the CDF. These results encourage us to use the simple balanced random—BRA—resource allocation scheme in the remaining of the numerical section and focus on comparing the performance of different PC approaches.
Fig. 6 Comparing the performance of the MinInterf, CPA and BRA resource allocations in terms of SINR distribution of the cellular UEs and D2D pairs when employing the utility maximizing PC scheme for all (i.e. cellular UE and D2D) users. MinInterf is clearly superior in terms of protecting the cellular UEs in the entire SINR region. Since CPA tends to reuse resources used by the strong (cell center) UEs, its SINR performance is somewhat worse in the high SINR region than the BRA

Fig. 7 Comparing the performance of the MinInterf, CPA and BRA resource allocations in terms of SINR distribution of the cellular UEs and D2D pairs when using the LTE open loop PC. The 3 RA schemes perform quite similarly, but somewhat worse than with the utility optimal PC, especially in the high SINR region (5–8 dB performance loss for the 80%-ile and stronger users)

Figure 8 compares the power consumption and the achieved SINR of the cellular UEs when employing different PC strategies in the system. The cellular UE power consumption is only affected by the PC algorithms used by the cellular UE, i.e. utility maximization or LTE OFPC), as shown by the left hand side figure. We can see that setting $\omega$ to 1 results in similar power levels for the utility and LTE based PC schemes, while setting $\omega$ to 0.01 significantly increases the transmit power level. On the other hand, $\omega = 10$ leads to significant power saving for the cellular UEs as compared to the LTE PC scheme. The achieved SINR by the
Fig. 8  Comparing the performance of the LTE based OFPCL with that of the utility optimizing scheme in terms of transmit power and achieved SINR for the cellular users. The utility based scheme can be tuned to boost the SINR performance of the cellular UEs in the high SINR regime at the expense of higher power consumption by setting the $\omega$ to a lower value. For cellular UEs, the hybrid LTE-utility maximizing scheme performs similarly to the LTE open loop PC.

cellular UEs depends not only on their own power control scheme, but also on the power used by the D2D pairs, as it is shown by the right hand side figure. We can see that in the utility maximization case, setting $\omega$ to a low value (e.g. when $\omega \leq 1$) can significantly boost the achieved peak SINR values. Apart from this high SINR regime, the SINR performance of the hybrid PC scheme (i.e. LTE open loop for the cellular UEs and utility maximization with a low $I^\ast$ cap for the D2D pairs) shows very good performance, showing for example up to 5–8 dB gains in the lower SINR regime compared to the pure utility maximization schemes (depending on the setting of the $\omega$). This result shows that setting the $I^\ast$ interference limit to a proper value is an important tool for protecting the cellular traffic from the interference caused by the D2D layer.

Figure 9 shows the distribution of the transmit power and SINR levels of the D2D pairs. Similarly to the cellular UEs, the D2D transmit power levels can be tuned by setting the $\omega$ (here within the range of $\omega = 0.01 \ldots 10$). We can also see that the different LTE based schemes perform quite differently both in terms of power consumption and achieved SINR. In terms of SINR, the LTE open loop power control yields an acceptable performance (close to the LTE closed loop scheme except in the low SINR region), but this SINR performance can be significantly improved by employing the hybrid scheme when setting $\omega$ and $I^\ast$ to proper values (e.g. $\omega = 1, I^\ast = 500 \cdot N_0$). Recall from the previous figure, that the performance punishment for the cellular UEs when using this more aggressive setting for the D2D pairs is negligible. When $I^\ast$ is set to a low value, the LTE PC scheme performs better in the low and medium SINR regime. In Fig. 9 it is interesting to observe the distribution of the transmit power level when using the “Fixed Tx Power” method for the D2D pairs. Since this figure shows the transmit power distribution prior to mode selection, the actual power
level set for the D2D candidates can be different from the predetermined fixed transmit power level if the cellular mode is selected for a D2D candidate. This is because when using cellular mode, the transmit power level is set by the open loop path loss compensation method.

Figure 10 examines the trade off between the performances of the cellular and the D2D layers when using the hybrid PC scheme under various settings. Here we can see that setting $\omega = 1$ and $I^*$ to a high value with respect to the noise floor (sigma) boosts the D2D performance (solid blue line) with some moderate and acceptable negative impact on the cellular layer.

Figures 11 and 12 offer an insight into the mode selection and reuse gains of D2D communications. Recall that the mode selection gain is due to selecting the direct communication link rather than using cellular transmission, as is shown in the figure. When, in addition, resource sharing is possible between D2D and cellular transmitters, the overall system throughput further increases at a cost of higher total transmit power. This total transmit power increase depends on the geometry of the system, that is exemplified by the right hand side of the figure (i.e. total system rate). We can also observe that the utility maximization significantly improves the total system rate performance and at same time reducing the average power level in the system. (The hybrid scheme (not shown here) performs close to the utility maximization scheme when $\omega$ and $I^*$ are properly set.)

Finally, Fig. 13 shows the correlation between the used transmit power and achieved SINR levels for cellular UEs and D2D pairs when using the utility based, the LTE based and the hybrid PC algorithms. When the LTE PC targeting a fixed SNR level is employed, the resulting SINR levels are rather similar throughout the simulations. For the D2D pairs, the fixed Tx power yields a large variation in the achieved SINR values. The other LTE based
Fig. 10 Hybrid PC scheme: cellular UEs use the legacy LTE power control, while D2D users use the distributed utility maximizing scheme. By setting the $I^*$ threshold to some suitable value, (e.g. $\omega = 1, I^* = 500 \cdot N_0$) the D2D performance can be boosted at the expense of a minimal impact on the cellular layer.

Fig. 11 Comparing the mode selection gain and the resource reuse gain in terms of total system throughput and used power when both cellular UEs and D2D pairs use the LTE open loop PC scheme. When the D2D candidates are forced to operate in cellular mode (“UE Mode”), they can neither take advantage of the short communication distance nor the reuse gain. When the D2D candidates can select D2D (direct) mode, proper mode selection (MS) yields increased total system rate and reduced power. Finally, when D2D pairs may reuse cellular resources (i.e. a RB can be used by multiple transmitters), the total system rate can be further improved (“MS Reuse”). However, this total system rate improvement, that is the reuse gain, may come at the expense of some increase of the total system power (see the left hand side plot, especially when the D2D distance is set to 100 m).
Fig. 12 Comparing the mode selection gain and the resource reuse gain in terms of total system throughput and used power when both the cellular UEs and the D2D pairs use the utility maximization scheme. The overall gain of the utility optimal scheme (as compared with the LTE based PC) is quite large both in terms of overall power consumption and achieved user bit rates, as visible by comparing this figure with Fig. 11.

Fig. 13 Comparing the transmit power and the SINR performance for cellular UEs (left) and D2D pairs (right) when D2D pairs employ the LTE and utility PC based algorithms. For cellular UEs, the utility based scheme (applied to the D2D pairs) with $\omega = 0.1$ tends to trigger higher power values (set by the LTE OFPC power control) and thereby to reach higher SINR values. For D2D pairs, the utility based scheme is clearly superior to legacy LTE PC when comparing the achieved SINR with the same transmit power levels. The hybrid scheme clearly pushes the D2D performance towards higher SINR values, while essentially keeping the benefits of the utility maximizing scheme for the cellular UEs, without requiring a new PC scheme for the cellular UEs.

schemes as well as the utility function method perform in between these two extremes, the utility based PC providing the best performance in terms of achieved SINR but tending to consume somewhat higher power both for cellular UEs and D2D transmitters.
9 Conclusions

In this work, we examined the performance of practical radio resource management algorithms for D2D communications integrated in cellular networks. The main motivation for this examination is to gain an understanding of how well LTE friendly power control and resource allocation schemes perform as compared to optimization based approaches. Specifically, we developed a distributed power control algorithm that maximizes a utility function that is capable of balancing between maximizing spectral efficiency and minimizing the sum transmit power for a given set of interfering D2D and cellular links. We used this algorithm as a benchmarking tool with respect to practical PC schemes based on the LTE PC toolkit, including no PC, PC with fixed SINR target, OFPC and closed loop PC. For mode selection and resource allocation, we developed a heuristic algorithm (MinInterf) that attempts to reduce the intra-cell interference introduced by D2D communications assuming full path loss knowledge. Using MinInterf as a benchmark, we then examined the performance of two practically feasible MS and RA algorithms in a realistic system simulator.

The numerical results indicate that the LTE PC gets close to the utility-based scheme, both in terms of used transmit power levels by the cellular as well as the D2D users and the resulting SINR values. However, there is a significant gain when employing the optimization-based approach in terms of the SINR obtained by the D2D users. On the other hand, the LTE OFPC scheme, (depending on the $\omega$ parameter of the utility-based method) can produce somewhat higher SINR values for the cellular UEs. These results tend to suggest that the flexible LTE power control scheme is well prepared for network assisted D2D communications, especially for the cellular UEs. However, for the D2D pairs, the utility based scheme can provide gains in terms of SINR distribution and total transmit power consumption. These gains can be harvested by a hybrid scheme, in which cellular UEs use the LTE PC scheme, whereas D2D users rely on a distributed scheme, whose parameters in practice can be controlled by the cellular network. In future work we plan to investigate methods to set the value of $I^\star$.

Appendix 1: Derivation of Inequality (24)

Constraints in Problem (23) can be elaborated by appealing to the definition of matrix $H^T$ in Eq. (21) as follows:

$$H^T\lambda^{(LP)} \geq -\omega 1$$
$$\sum_k h_{kl}\lambda_k^{(LP)} \geq -\omega, \quad \forall l;$$
$$-h_{ll}\lambda_l^{(LP)} + \sum_{k \neq l} h_{kl}\lambda_k^{(LP)} \geq -\omega, \quad \forall l$$
$$\frac{\lambda_l^{(LP)}}{\omega} - \sum_{k \neq l} \frac{G_{kl}}{G_{kk}} \gamma_{l}^{(LP)} \leq 1, \quad \forall l.$$

Appendix 2: Derivation of Inequality (26)

Inequality (26) can be derived by appealing to Eq. (25) as follows:
\[
\frac{\gamma_i^{(LP)}}{\omega} - \sum_{k \neq l} \frac{G_{kl}}{G_{kk}} \gamma_k^{tgt} \frac{\gamma_k^{(LP)}}{\omega} \leq 1
\]

\[
\frac{\gamma_i^{(LP)}}{\omega} \frac{\gamma_i^{tgt} \sigma_i}{G_{ll}} - \sum_{k \neq l} \frac{G_{kl}}{G_{kk}} \gamma_k^{tgt} \frac{\gamma_k^{(LP)}}{\omega} \sigma_k \leq 1
\]

\[
\frac{\mu_i G_{ll}}{\gamma_i^{tgt} \sigma_i} \leq \sum_{k \neq l} \frac{G_{kl}}{\sigma_k} \mu_k + 1
\]

\[
\gamma_i^{cc} (\mu) \triangleq \frac{\mu_i G_{ll}}{\sigma_i} + \sum_{k \neq l} \frac{G_{kl}}{\sigma_k} \mu_k \leq \gamma_i^{tgt}, \quad \forall l.
\]

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