

Trust-based and social-aware coalition formation game for multihop data uploading in 5G systems



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ABSTRACT

In this paper a trust-based coalition formation game is proposed to design opportunistic hop-by-hop forwarding schemes, relying on cellular *Device-to-Device* (D2D) communications, to enhance content uploading services. The User Equipments (UEs) are sources of data to be uploaded to a cellular base station (eNodeB) and are assumed to be *rational self-interested* players as they aim at maximizing their own utility. To this aim, the UEs cooperate to opportunistically implement proximity-based data exchanges where the presence of *malicious nodes* in the network is a constant threat for the successful cooperation. To cope with this issue, *reliability* and *reputation* notions are considered to model the level of *trust* among the players. Taking inspiration from the recent Social Internet of Things (SIoT) paradigm, *social-awareness of the devices* is spotted as a key parameter to effectively define the wished trustworthiness. The effectiveness of the proposed solution is validated through a simulative analysis showing a relevant reduction in the data loss due to malicious behavior of a subset of the devices. In particular, up to 86% reduction in terms of data loss is obtained with respect to the case where the proposed trust model is not implemented. Moreover, the trust-based and social-aware solution also guarantees higher gains in terms of the uploading time for the devices taking part of the cooperative D2D-based content uploading.

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1. Introduction

Fifth generation (5G) systems are expected to introduce a revolution in the ICT domain with innovative networking features [1]. Among them, device-to-device (D2D) communications, whereby in-proximity devices communicate directly with each other to avoid routing the data paths through the network infrastructure, will play an undoubted key role [2]. The growing interest for this technology is driven by the possibilities it offers to overcome the limitations of conventional cellular systems. Among others, D2D communications, either over cellular resources or over alternative Wi-Fi/Bluetooth technologies, introduce benefits in terms of improved spectrum utilization, higher throughput, and lower energy consumption [3]. A large number of D2D-based applications have been investigated for future 5G wireless systems, such as mobile data offloading for proximity based applications, network coverage extension (also in case of network failure), content sharing support

among UEs, etc. [4]. In this context, the reference scenario for this paper considers groups of devices in close proximity and willing to upload some data to the Cloud. Typical example scenarios are small-scale environments at aggregation places (e.g., a stadiums, university campuses, music events, or fairs) where the devices can exploit opportunistic data forwarding over the devices in proximity [5]. A further example is disaster scenarios where D2D relaying may be important to send out emergency messages from disconnected areas and to support information sharing among people gathered in evacuation centers [6].

In a traditional cellular system, each User Equipment (UE) interested in uploading a content activates a unicast uplink communication to the eNodeB. In very crowded environments, where many UEs want to upload some content, the risk of spectrum crunch and poor service quality is high. Moreover, a UE located far from the eNodeB could suffer from low channel quality leading to out-of-coverage situations, which, in some cases, may be of high concern (in emergency situations, for instance). These limitations are overcome by solutions that exploit the enhanced capabilities and multiple network interfaces of modern smart devices. As an example, the research activities in the field of Mobile Networking in Proximity (MNP) [7] is very active. This new paradigm complements the classic scenario by adding continuous connectivity

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coverage through short-range communications, based on Bluetooth and WiFi Direct, whenever a loss of connectivity is observed due to obstacles, coverage hole, or bad channel quality. Based on this paradigm, a UE that is far from the base station may establish proximity-based communications with nearby UEs that are experiencing a higher-quality in the communication link to the eNodeB. The proximity communication can be implemented over multiple cellular D2D links to set-up opportunistic hop-by-hop forwarding towards the destination. A necessary condition for such a “cooperative” relaying solution to bring benefits compared to the non cooperative case, is that the link quality of the multihop D2D topology is higher than that one of the separate links. This condition is more likely to occur in non-isotropic propagation environments with obstacles where non line of sight (NLOS) conditions may cause partial and temporary out-of-coverage conditions [8].

A first analysis of a cooperative content uploading in LTE-A (Long Term Evolution-Advanced) scenarios has been proposed in a recent paper [9]. In this paper, we continue our research by introducing an analysis of a challenging aspect of utmost importance for an effective implementation of proximity communications, namely the need of *trustworthiness* [10,11]. In realistic scenarios, where human interactions and human behavior is to be considered, the presence of *malicious* nodes in the network is a constant threat for a successful cooperative interaction. Indeed, in opportunistic and random communications among UEs, the end-users may not be aware of the device and end-user they are going to be connected to. Malicious nodes may decide to drop the data packets they are expected to forward without informing the interested users or forward corrupted data. To cope with these threats, the objective of this paper is to model a trust-based and social-aware multihop D2D data uploading able to satisfy the rational users interested in reducing their content uploading time. To reach this goal, *reliability* and *reputation* notions will be considered to model the level of *trust* among the involved entities. By taking inspiration from recent Social Internet of Things (SIoT) models [12], in this paper we consider the sociality level of the devices to model the *reliability* of the communication [13]. The historical *reputation* of the cooperative users will also be considered to offer rational users the possibility to filter out untrusted users and avoid unsuccessful opportunistic hop-by-hop D2D interactions. The main contributions of this paper can be summarized as follows:

- We model the *trust* constraints for a successful D2D-based content uploading, where sociality among devices, as a measure of *reliability*, and historical *reputation* are included into the model;
- We define the content uploading time through a multihop D2D topology as a function of the number of UEs forming the topology and the links status;
- We define a *constrained coalition formation game* that forms the overlapping multihop D2D coalitions under the constraint of reciprocal UEs proximity for the direct links activation and a minimum *trust* level among the cooperating devices. The algorithm converges to a stable coalition structure, where all players are happy to join the formed network partition and do not have incentives to leave the coalition they are part of;
- We perform a simulative performance evaluation showing high reduction in the data loss due to malicious behavior of a subset of the involved devices. In particular, up to a 86% reduction in terms of data loss is obtained, in the best case, with respect to the case where the proposed trust model is not implemented. Moreover, the trust-based and social-aware solution also guarantees higher gains in terms of the uploading time for the devices taking part in the cooperative D2D-based content uploading.

The rest of the paper is organized as follows. In [Section 2](#) we browse the related work, whereas in [Section 3](#) we discuss the ref-

erence system and background for multihop D2D communications in LTE-A systems. The proposed trust model and the sociality concepts are introduced in [Section 4](#). [Section 5](#) introduces the constrained coalitional game adopted to model the D2D topology formation. Numerical results are provided in [Section 6](#), whereas conclusions are drawn in the last section.

2. Related work

In a cooperative context as the one studied in this paper, it is of utmost importance to build trust-based interactions to guarantee reliability in the communication and limit the negative effect of typical issues such as the presence of *malicious* nodes in network [13] or *free riders* [14]. Malicious nodes may try to modify data or manipulate the reputation of the devices through *ballot-stuffing* and *bad-mouthing attacks* to either increase its own reputation or decrease the reputation of other nodes. A free rider is a node that is taking part in a coalition to benefit from a reduced content uploading time, but is actually dropping all the incoming data from the upstreams in the cooperating topology. In this way, the node saves the energy of its battery while still benefiting from the cooperative behavior of the other nodes in the coalition. These behaviors, require solutions able to isolate the malicious nodes and build reliable reputation of the nodes.

In the past few years, with the advent of online social networks several methods to calculate trust and distrust between two persons have been proposed [10,11,15]. Generally, trust is defined as the quantified belief by a truster with respect to the competence, honesty, security and dependability of a trustee within a specified context [16]. When two users want to cooperate, one of them (the truster) assumes the role of a service requester and the other (the trustee) acts as the service provider. Specifically, in our cooperative coalition the node acting as relay/gateway for another node will be the trustee and the source node of the relayed data is the truster. The trustworthiness of the truster with respect to the trustee can be determined considering reliability and/or reputation. The former is a direct measure derived by subjective observations of the truster during its interactions with the trustee; the latter is an indirect measure based on the opinions that the other actors in the community have about the trustee.

In the literature, several trust models have been proposed to represent both reliability and reputation [16]. The mechanism we propose enhances classic trust models through the exploitation of *social relationships* among the involved devices (to improve device *reliability*) and of recommendation exchange (to the purpose of reputation definition). In particular, we consider the potential of the SIoT model defined in [12], to embrace the social networking concepts and build trustworthy relationships among the devices [13]. In particular, mobility patterns and relevant context can be considered to configure the appropriate forms of socialization among the UEs. Specifically, the so-called *co-location object relationships* (C-LOR) and *co-work object relationships* (C-WOR) are established between devices in a similar manner as among humans, when they share personal (e.g., cohabitation) or public (e.g., work) experiences. Another type of relationship may be defined for the objects owned by a single user, which is named *ownership object relationship* (OOR). The parental object relationship (POR) is defined among similar devices built in the same period by the same manufacturer, where the production batch is considered a family. Finally, the social object relationship (SOR) is established when objects come into contact, sporadically or continuously, for reasons related to relations among their owners.

In the definition of the social-aware trust-based cooperative data uploading topologies, we exploit a coalition formation game to form *stable* D2D multihop coalition of UEs. Stability implies that all players interested in their own payoff do not find moti-

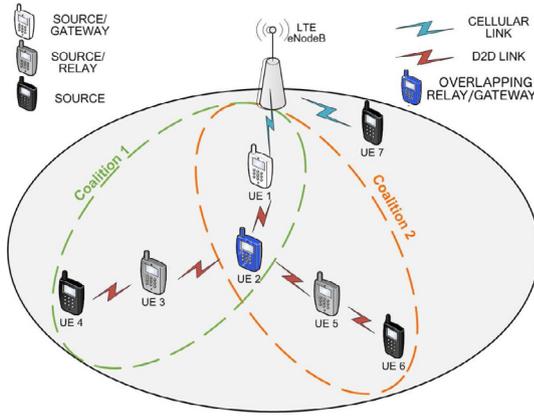


Fig. 1. Multihop D2D-based coalitions in tree topology.

vations to leave the coalition they are part of. The effectiveness of this model is witnessed by the large number of applications in wireless networking problems such as in [17,18], and many others. The main novelty of our proposal is the definition of a *constrained coalition formation game*, where only feasible coalitions are considered as dictated by D2D coverage and trustworthiness constraints. A trust-based coalition formation problem has been studied also in [19] and [20], but for service composition problems and multi-agent systems respectively. Besides the different problem setting, in our work we adopt a merge-and-split coalition formation algorithm that guarantees stability of the obtained coalition structure and we include social-awareness for the reputation, reliability and trust modeling.

3. Reference system and research background

We consider a single LTE-A cell with multiple UEs interested in uploading their content to the Internet. Data uploading according to the traditional *cellular-mode* is performed through the activation of separate links from each UE to the eNodeB. With the proposed *cooperative upload* instead, under the control of the eNodeB (i.e., network-assisted D2D), the UEs organize themselves to form a “logical multihop D2D topology” and cooperate in uploading the content generated by *all* of them to the eNodeB (see Fig. 1). In general, only the UEs in mutual coverage can establish direct links. In the formed cooperative topology, that we hereafter also call “coalition”, the UEs located farther from the base station relay their content to a nearby UE and only the UE at the head of the topology, the so-called *gateway*, is in charge of uploading all the contents received from the other UEs to the eNodeB. The gateway is the UE with the best link quality in the coalition and it may receive, if needed, all the radio resources that would have been separately allocated by the eNodeB to the UEs in the coalition. All intermediate UEs in the topology also act as *relays* for the contents received from the upstream UEs, thus benefiting of the higher quality of the short D2D links w.r.t. the direct cellular link. Actually, when overlapping coalitions are formed, any intermediate UE can receive content from multiple branches of the resulting tree topology. Thus, in the most general configuration, each relay has one or more links active to receive data from the preceding sources in the tree topology, and one single link active to relay data (its own generated traffic and the traffic from the incoming D2D links) to the subsequent UE in the topology.

Each UE operates in half-duplex mode; thus, it either receives or transmits in a given transmission time interval. We consider a reasonable assumption for *rational* self-interested devices, that each UE uploads its own generated content first and then the content received by the preceding UEs in the topology. In particu-

lar, the transmission starts only after that the generic UE has received the whole content (in other words, UEs use the decode-and-forward relaying protocol). Devices in the same coalition may share the same resources, whereas devices in different coalitions are always allocated to orthogonal frequency resources by the scheduler at the eNodeB, so that no mutual interference is caused by different coalitions (this is a reasonable assumption, used in other works [21]). Simultaneously transmitting UEs within the same coalition can use either the same or different frequencies, based on the decision of the eNodeB according to the interference level experienced on each direct link. In particular, in this paper we consider the two extreme cases, namely the *best-case* where the same radio resources can be reused on the D2D links, and the *worst-case*, where simultaneous transmissions interfere. In this latter case, to avoid interference on simultaneous D2D transmissions, orthogonal resources are used and we assume that the radio resources used on the D2D links are those allocated by the scheduler in the uplink toward the eNodeB.

The eNodeB manages the spectrum by assigning the adequate number of Resource Blocks¹ (RBs) to each scheduled user and by selecting the Modulation and Coding Scheme (MCS) for each RB. Scheduling procedures are based on the *Channel Quality Indicator* (CQI) feedback, transmitted by each UE to the eNodeB over dedicated control channels. The CQI is associated to a given maximum supported MCS as specified in [22]. Thus, given a link l and the allocated RBs, the channel data rate $R_{Ch,l}$ on the wireless link l follows Shannon’s capacity formula $R_{Ch,l} = B \log_2(1 + \gamma_l)$, where γ_l is the signal to interference plus noise ratio (SINR) value experienced on the link and B is the available bandwidth. For a user transmission, the available bandwidth B is determined by the radio resource allocation policy. In addition, in the LTE-A system generally the channel data rate for transmitting over cellular and D2D links, i.e., $R_{Ch,m}$ and $R_{Ch,d}$ is determined as a function of the CQI values. Therefore, for a given CQI value q , the attainable data rate can be represented as a function $f(q, n_q)$ ² where n_q are the assigned RBs.

When D2D coalitions are formed in the cell, we propose that the eNodeB only assigns to the gateway of each coalition a *pool* of uplink resources that can reach up to the sum of the radio resources separately requested by the UEs in the coalition. The eNodeB assists the users in the cooperative coalition formation process. In the first step, the eNodeB computes the radio resources allocated to the UEs as if they were transmitting separately on the uplink according to the adopted scheduling policy. These resources are “virtually” allocated since UEs may form a coalition and can be used as the *pool* of resources allocated to the gateway. Based on this initial information, the eNodeB implements the coalition formation algorithm (see Section 5). As a result, stable coalitions are formed in the cell, the roles of each node in the coalition is identified, and the best routing path is defined. In particular, when considering a potential coalition, a step-wise decision algorithm determines the *best path* that covers all the nodes in the topology. To this aim, the eNodeB first sorts the devices in a decreasing order of uplink CQI (first those with better channel quality) and then selects the first node in the list as the *gateway* for the coalition. This is important, so that the resource pooling will produce the highest throughput toward the eNodeB for the whole multihop topology. Once the gateway is selected, the *best path* over the set of nodes is computed with focus on the D2D link qualities. We consider a simple *greedy* approach where the next hop from the gateway is selected as the one in the one-hop vicinity with the best D2D link quality. Similarly, each node in the topology will select its neighbor

¹ The RB corresponds to the smallest time frequency resource that can be allocated to a user (12 sub-carriers) in LTE. For example, a channel bandwidth of 20Mhz corresponds to 100 RB.

² The admissible throughput values per MCS level follow Table 7.1.7.2.1-1 in [23].

based on the best CQI of the direct link to the remaining nodes in the coalition. Once the coalitions are formed in the cell, the eNodeB determines the radio resources assigned to the gateway and to each D2D link and transmits all the information to the UEs so that the transmissions can start.

4. The social-aware trust model for D2D-based cooperation

We propose a network-assisted model where the eNodeB serves as a trusted third party supporting the coalition formation among the interested players. To this aim, the eNodeB will store up-to-date information about the reliability, reputation and trust parameters relative to the users in the cell. In particular, given the higher computational and storage capabilities of the eNodeB compared to the single UEs, we assume that the eNodeB will store a so-called *player trust matrix (PTM)* containing reliability, reputation and trust information relevant to every couple of devices in the network. The information stored in the PTM will be used whenever a new trust-based coalition formation step is considered for cooperative content uploading by eNodeB (see details of the algorithm in Section 5). After each cooperative content uploading, the eNodeB will send an acknowledgment to the respective source nodes. Moreover, it will detect malicious behavior in the coalitions based on data loss levels and update the reliability level of the interested players. We assume the data amount for the information exchange between source nodes and the eNodeB to be small and the data to be sent over control channels. Compared to the main content size to be uploaded by the source nodes, the control messages are very small and the corresponding transmission time and energy consumption are assumed to be negligible. Let $i \rightarrow j$ be a generic D2D link in the coalition being considered at time t during the coalition formation algorithm, where node j is expected to act as relay/gateway for the data he receives from node i (its own and the preceding nodes in the topology); we consider i, j as the truster and the trustee respectively. The parameters the eNodeB will use to determine the level of trust for the link, and contained in the PTM, are:

- *Social player reliability* ($spr_{i,j}$): stored in the PTM, is the reliability that node i assigns to player j only based on the social relationship that links the two players and is a value in $[0, 1]$;
- *Player reliability* ($pr_{i,j}^t$): stored in the PTM, is the reliability that node i assigns to player j at time instant t . This is a subjective evaluation of the players which we consider to be influenced by the *social player reliability* and the outcome of past interactions where player j was expected to act as relay/gateway for player i . Specifically, the player reliability value is a real number ranging in $[0, 1]$ with the 0/1 values meaning that i judges player j as completely unreliable/reliable.
- *Recommendation reliability* ($rr_{i,j}$): stored in the PTM, is the reliability that node i assigns to the recommendations provided by player j about other players in the network. Also this parameter is based on a subjective evaluation of the interested player and in our model it is influenced by the social relationship between the interested UEs. It is a real number ranging in $[0, 1]$ with the 0/1 values meaning that i judges the recommendation received from j as completely unreliable/reliable.
- *Player reputation* ($pp_{i,j}^t$): stored in the PTM, is the reputation that player i assigns to player j for the specific service based on the recommendations provided by other players in the network at time instant t . Similar to the previous parameters, it is a real number ranging in $[0, 1]$ with the 0/1 values meaning that the reputation assigned by i to player j based on the recommendation of the other devices is minimum/maximum.
- *Player trust* ($pt_{i,j}^t$): stored in the PTM, is the trust level that player i associates to player j at time t , which is the final

parameter that determines whether player i is willing to entrust player j as relay/gateway node in a D2D-based cooperative coalition. This is a weighted combination of the reliability $pr_{i,j}^t$ and the reputation $pp_{i,j}^t$ for the player. The final player trust value is a real number ranging in $[0, 1]$ with the 0/1 values meaning that player j is considered as completely untrusted/trusted by player i .

Player reliability. The *player reliability* is updated over time, based on the past experience related to cooperative interactions where a player was expected to act as relay/gateway for the data sent by a precedent player in the formed D2D topology. To consider the past experience, we assume that for each cooperative interaction the eNodeB sends an acknowledgment to the source nodes in the coalition with information about the data being successfully received. However, based on this simple information, it is not possible for the eNodeB to determine which node in the cooperative topology has actually dropped the data. In our proposal we assume that the eNodeB will associate the outcome value δ_d to the node j that was entrusted by node i as relay/gateway forming a D2D link $i \rightarrow j$. Thus, at time $t = 0$ when no cooperation history exists, the only information the interested devices can exploit for judging the *player reliability* is *social player reliability* ($spr_{i,j}$). This is set according to predefined values as reported in Table 1.³ At subsequent time instants $t > 0$ the player reliability $pr_{i,j}^t$ is computed by taking into consideration the past interactions between i and j , with j acting as relay/gateway for data sent by i . We define with $\Delta_{i,j}^t = \{\delta_1, \dots, \delta_d \dots \delta_D\}$ the set of past interactions registered until time t , where the generic $\delta_d \in [0, 1] \in \mathbb{R}$ is a value measuring the outcome of the cooperative interaction. This is equal to the total percentage of data that has been successfully forwarded by node j and reached the eNodeB. Thus, the player reliability $pr_{i,j}^t$ is computed as follows:

$$pr_{i,j}^t = \begin{cases} spr_{i,j} & t = 0 \\ \alpha \cdot spr_{i,j} + (1 - \alpha) \cdot \frac{\sum_{d \in \Delta_{i,j}^t} \delta_d}{|\Delta_{i,j}^t|} & t > 0 \end{cases} \quad (1)$$

where $\alpha \in [0, 1]$ is a real number used as weighting factor to give more or less importance to the initial sociality relationship between the involved nodes.

Player reputation. The *player reputation* is based on the opinion of the community in the network, which is updated after each cooperative interaction. Let us consider a player i asking an opinion about player j and let $\mathcal{K} \subseteq \mathcal{N} \setminus \{i\}$ be the set of players which provide an opinion about player j to player i . The opinion player k will provide (i.e., the eNodeB consider the corresponding value in the PTM) is its own measure of trust about player j at time instant t , i.e., $pr_{k,j}^t$. The opinion received from the other players in the network is weighted by a confidence factor the requesting player has about the received recommendation. This weighting factor is the so-called *recommendation reliability* ($rr_{i,k}$). In our model the *recommendation reliability* is set according to the social relationship between the involved devices and its value is reported in Table 1. Noteworthy, we assume that the *recommendation reliability* has a lower value w.r.t. *social player reliability* in general. The reason for this choice is that the recommendation received by a socially related device may be influenced by the outcome of past cooperative iterations with other devices which affected the ability to provide an objective recommendation. Thus, the *player reputation* at time t

³ If two communicating entities are tied by two or more types of relationships, the strongest tie with the highest factor has to be considered [13].

Table 1

Player and recommendation reliability values associated to the social relationship between devices.

Relationship	Description	Social player reliability (spr_{ij})	Recomm. reliability (rr_{ij})
Ownership object relationship (OOR)	Objects owned by the same person	1	0.9
Co-location object relationship (C-LOR)	Objects sharing personal experiences	0.8	0.6
Co-work object relationship (C-WOR)	Objects sharing public experiences	0.7	0.5
Social object relationship (SOR)	Objects in contact for owner's relations	0.6	0.5
Parental object relationship (POR)	Objects with production relations	0.5	0.4
No relationship		0.1	0.1

is computed as follows:

$$pp_{i,j}^t = \frac{\sum_{k \in \mathcal{K}} rr_{i,k} \cdot pt_{k,j}^t}{\sum_{k \in \mathcal{K}} rr_{i,k}} \quad (2)$$

Player trust. Based on the notions introduced above, player i can finally determine the player trust value $pt_{i,j}^t$ it associates to player j at time instant t . This is a combination of the player reliability value at time t , ($pr_{i,j}^t$), and the player reputation ($pp_{i,j}^t$), suitably weighted by a real coefficient β ranging in $[0, 1] \in \mathbb{R}$:

$$pt_{i,j}^t = \begin{cases} 0.5 & t = 0 \\ \beta \cdot pr_{i,j}^t + (1 - \beta) \cdot pp_{i,j}^t & t > 0 \end{cases} \quad (3)$$

Note that for newly arrived nodes at time $t = 0$, the initial trust is set to 0.5 as suggested in [24] to contrast *whitewashing strategies* where a dishonest adviser is able to whitewash its low trustworthiness by starting a new account with the initial trustworthiness value.

5. A constrained coalition formation for trust-based cooperation

In this section we introduce a so-called constrained coalition formation (CCF) game for the cooperative D2D content uploading. A traditional coalitional game in cost form is defined by (\mathcal{N}, c) where $\mathcal{N} = \{p_1, \dots, p_N\}$ is the set of N players and c is the cost function that models the feasible cost-value $c(S)$ for every set of players (coalition) $S \subseteq \mathcal{N}$. In particular, it is an *overlapping coalition formation game* [25] when each player is able to cooperate and join multiple coalitions. In our problem, the cost for a player p_i in coalition S is expressed in terms of uploading time needed for its own data to reach the eNodeB. With this cost function, that is not transferable by definition, the studied game is a *non-transferable utility game*. In particular, for any singleton coalition $\{i\}$, the cost for player i alone is equal to the content uploading time in the cellular mode, i.e., when the UE uploads its content of size b_i over its cellular link having a data rate r_i^c : $c(\{i\}) = \frac{b_i}{r_i^c}$. For any coalition $S \subseteq \mathcal{N}$, with cardinality $|S| > 1$, the associated cost $c(S)$ is the data uploading time for the coalition as defined in Section 5.1. If the multihop D2D coalition cannot be formed due to coverage or trust constraints between the involved UEs, then we define: $c(S) = \sum_{i \in S} c(\{i\})$. In particular, the problem definition will lead to either a line or a tree topology for the coalitions.

We define a *collection* of coalitions \mathcal{C} as the set $\mathcal{C} = \{C_1, \dots, C_l\}$ of mutually disjoint coalitions $C_i \subset \mathcal{N}$ such that $C_i \cap C_{i'} = \emptyset$ for $i \neq i'$. If the collection contains all players in \mathcal{N} , i.e., $\bigcup_{i=1}^l C_i = \mathcal{N}$, then

the collection is a *partition* or *coalition structure (CS)*. The set of all possible *coalition structures* is identified by $\Pi(\mathcal{N})$. Formally we define the CCF game for the problem studied in this paper as $G = \langle \mathcal{N}, \mathcal{P}, \mathcal{Q}, \mathcal{Z}, c \rangle$ where \mathcal{N} is the set of UEs in the cell and $S \subseteq \mathcal{N}$ is any multihop D2D coalition, $\mathcal{P} \subseteq 2^{\mathcal{N}}$ is a set of *positive constraints* such that a coalition C satisfies a constraint $P \in \mathcal{P}$ if $P \subseteq C$, $\mathcal{Q} \subseteq 2^{\mathcal{N}}$ is a set of *negative constraints* such that a coalition C satisfies a constraint $Q \in \mathcal{Q}$ if $Q \not\subseteq C$, and \mathcal{Z} is a set that defines the constraints on the coalitions size (*size constraints*). In our setting, the constraints for the problem are only the negative constraints deriving from a combination of *trust* and *coverage* between pairs of UEs, whereas we set $\mathcal{P} = \emptyset$, $\mathcal{Z} = \emptyset$. For the exact definition of \mathcal{Q} , those coalitions for which the CQI value between two consequent UEs in the corresponding topology is zero should be automatically considered as *not feasible* and thus stored in \mathcal{Q} . Moreover, we introduce a *feasibility threshold FT* for each coalition, which indicates the minimum value of trust for each D2D link in a coalition. We say that coalition $S \subseteq \mathcal{N}$ is *not feasible* and thus included in \mathcal{Q} if the resulting topology foresees at least one link $i \rightarrow j$ that does not meet the following constraint:

$$pt_{i,j} \cdot d_{i,j} \geq FT \quad (4)$$

where $pt_{i,j} \rightarrow [0, 1]$ is the *player trust* that i associates to j as defined in Eq. (3), whereas the second term $d_{i,j}$ is a binary function taking the value of 0 if the users i and j are not in proximity, and taking the value of 1 otherwise.

Since $\mathcal{Q} \neq \emptyset$, it is implicitly said that the grand coalition is not formed as it is certainly not a *feasible* coalition. To characterize the *feasible coalitional structure* to form for the CCF game, we propose simple merge-and-split rules. The key mechanism is to enable players to join or leave a coalition based on well-defined preferences so that each player is able to compare and order its potential coalitions based on which coalition it prefers to belong to. To do this, we introduce a preference relation over coalitions. The preference order \succ_i for any player $p_i \in \mathcal{N}$, is defined as a complete, reflexive, and transitive binary relation over the set of all feasible coalitions that player p_i can possibly form, i.e., the set Π_i of coalitions containing p_i . A UE can decide to join or leave a coalition according to its preference order. In particular, for each player p_i , if $C \succ_i C'$, p_i prefers being a member of coalition C more than coalition C' . Noteworthy, under this preference order definition a tree topology for the overlapping coalitions can be formed, as the uploading time for the UEs that are part of multiple coalitions is not negatively affected. This is due to the forwarding priority given to its own data compared to the data from the incoming D2D links.

To correctly define the coalition formation game, the coalition preference relation has to be defined over all pairs of coalitions in

Π_i . In this paper, the preference order is defined according to its *individual cost*. Thus, for each UE $p_i \in \mathcal{N}$ and for all $C, C' \in \Pi_i$, we say that:

$$C \succ_i C' \Leftrightarrow c_i(C) < c_i(C') \wedge c_j(C') \leq c_j(C \setminus \{i\}), \quad (5)$$

$$\forall j \in \{C \setminus \{i\}\} \wedge c_j(C) \leq c_j(C \setminus \{i\}), \forall j \in \{C \setminus \{i\}\}$$

In words, with this definition, any UE i prefers being a member of coalition C over C' if it obtains a lower individual cost $c_i(C)$, without causing an increase in the cost for any other player in C and C' (also known as *Pareto order* preference). The so-defined individual preference order is adopted for two simple rules for the feasible coalition formation game.

Definition 1 (Merge rule). Merge any pair of coalitions C and C' in a unique *feasible* coalition $\{C \cup C'\} \Leftrightarrow [(\exists k \in C \text{ s.t. } \{C \cup C'\} \succ_k C) \vee (\exists k \in C' \text{ s.t. } \{C \cup C'\} \succ_k C')] \wedge \{C \cup C'\}$ is *feasible*.

Definition 2 (Split Rule). Split any coalition $\{C \cup C'\}$ in *feasible* coalitions $\{C, C'\} \Leftrightarrow [(\exists i \in C \text{ s.t. } C \succ_i \{C \cup C'\}) \vee (\exists j \in C' \text{ s.t. } C' \succ_j \{C \cup C'\})] \wedge \{C, C'\}$ are *feasible*.

The merge rule implies that two coalitions join to form a larger *feasible coalition* if operating all together strictly reduces the cost for at least one player while all the other involved players are not having a higher cost. The split rule implies that a coalition splits only if there exists at least one player that obtains a lower cost, under the constraint that this has no negative effect on the cost of other players and the resulting coalitions are both *feasible*.

Once the merge-and-split operations are defined, then the CCF game for the cooperative multihop D2D data uploading can be modeled. The objective of a UE is to find a coalition with the lowest uploading time through an iterative application of the merge and the split rules. By starting from an initial partition $\Pi^{ini}(N) = \mathcal{N} = \{p_1, p_2, \dots, p_N\}$, the eNodeB iteratively applies the merge-and-split rules to any pair of coalitions in the partition. In particular, by first considering the merge rule, for every pair of coalitions if the merged coalition is preferred w.r.t. the separated coalitions, then a new merged coalition is considered for the partition. The merging stops when no couple of coalitions exists in the current partition $\Pi^{cur}(N)$ that can be merged. Thus, the split rule is applied to every coalition in the partition. When no split occurs, the algorithm considers again the merging function. The algorithm terminates when no merging or splitting occurred in the last iteration. In this case, the final resulting partition $\Pi^{fin}(N)$ will be adopted by the eNodeB. It can be demonstrated (proof not shown due to length constraints) that the proposed coalition formation algorithm converges to a stable partition of disjoint coalitions of UEs.

5.1. Content uploading time model

Let \mathcal{N} be the set of UEs in the coalition, let each UE $i \in \mathcal{N}$ have some content of size $b_i \neq 0$ to upload and r_i^c be the data rate for user i when transmitting in traditional cellular-mode (i.e., directly to the eNodeB). The “virtual” radio resources for the UEs in the coalition are all allocated to the gateway. This implies having more resources available and achieving a consequent improved uplink data rate r_i^c for the gateway-eNodeB link. The detailed modeling of the uploading time for all UEs in a multihop D2D coalition is derived from the analysis conducted in a previous paper of ours [9]. For completeness in the presentation we report here the main findings.

To model the uploading time the channel occupation time for a generic UE i in the D2D coalition is computed as the time spent by the UE to transmit to the next hop its own data and the data

received from the previous UE (or multiple UEs) in the coalition. For the sake of notation simplicity, let us consider an N-hops D2D coalition with $i = 1$ being the gateway and $i = N$ be the last UE in the multi-hop coalition. Then, let r_i^d be the data rate for UE i on the D2D link that it uses to forward its data to the next UE. Let us consider UE N as the last UE in the coalition. This UE will occupy the channel for a time $T_N = b_N/r_N^d$ to forward its data b_N to UE $N - 1$ over the D2D link having data rate r_N^d . Considering UE $N - 1$, it will send its own data b_{N-1} and the data received from the previous UE b_N and occupy the channel for the time $T_{N-1} = b_{N-1}/r_{N-1}^d + (b_N/r_N^d + b_N/r_{N-1}^d)$. By repeating this reasoning for all UEs in a coalition, and considering that the gateway, UE 1, transmits to the eNodeB with a data rate r_1^c , the channel occupation time for UE 1 to upload all data from the coalition to the eNodeB is computed as a function of the number of UEs in the topology:

$$T_1(N) = \frac{b_1}{r_1^c} + \dots + \left(\frac{b_N}{r_N^d} + \frac{b_N}{r_{N-1}^d} + \dots + \frac{b_N}{r_1^c} \right)$$

$$= \sum_{i=1}^N \left(\frac{b_i}{r_1^c} + \sum_{j=2}^i \frac{b_i}{r_j^d} \right) \quad (6)$$

The formulation can be generalized to represent the channel occupation time for any UE $n = \{1, \dots, N\}$ in the multi-hop coalition. This time includes the time to forward to the next hop in the chain all data generated by UEs n and the data from its previous UEs in the chain (until the last UE N) and can be written as follows:

$$T_n(N) = \begin{cases} \sum_{i=n}^N \left(\frac{b_i}{r_1^c} + \sum_{j=2}^i \frac{b_i}{r_j^d} \right) & n = 1 \\ \sum_{i=n}^N \sum_{j=n}^i \frac{b_i}{r_j^d} & n > 1 \end{cases} \quad (7)$$

The content uploading time to the eNodeB for a coalition is given by the number of data frames needed, under the constraints posed by the data rate over the involved communication links and the communication assumptions in the topology. This corresponds to the cost $c(N)$ associated to the corresponding coalition in the game theoretic model. Moreover, this time is determined by the data frame organization where, given the UE half-duplex operation, the uplink sub-frames will be used either for D2D communications or for transmissions towards the eNodeB (the interested reader can find more details on the model in [9]). Moreover, the transmitting UEs will use either the same or separate RBs, according to the interference level experienced on each link.

The content uploading time has also a direct impact on the UEs' energy consumption, which is defined, as in the non-cooperative case, as: $E_i^c(b_i) = (P_{tx}^c + P_0) \cdot \frac{b_i}{r_i^c}$, where P_{tx}^c is the transmission power and P_0 the circuit power. When considering the cooperative data uploading, we have three cases: (1) the UE is the gateway; it consumes energy in receiving data from the second UE and in transmitting data to the eNodeB; (2) the UE is the last UE in the chain; it only consumes energy in transmitting its own data to the next UE in the D2D chain; (3) the UE is an intermediate UE in the chain; it consumes energy to receive data from the previous UE and to transmit data to the next UE in the chain. In all three cases, energy is also spent during the idle times on the channel. However, according to [26] the power consumption in idle times is as low as -50 dbm; therefore, this contribution can be neglected and only the transmitting and receiving power on the D2D links, P_{tx}^d and P_{rx}^d , are considered. The energy consumption for a generic UE i in the D2D chain will be the sum of the energy spent for transmission

Table 2
Main simulation parameters.

Parameter	Value
Cell radius	500 m
Maximum D2D link coverage	100 m
Frame Structure	Type 2 (TDD)
TTI	1 ms
TDD configuration	0
Carrier Frequency	2.1 GHz
Cellular transmission power consumption	23 dBm
D2D power consumption	−19 dBm
CQI-MCS mapping for D2D links	[27]
Noise power	−174 dBm/Hz
Cellular link model	Rayleigh fading channel
D2D link model	Rician fading channel [28]
Path loss (cell link)	$128.1 + 37.6 \log(d)$, $d[\text{km}]$
Path loss (D2D link, NLOS)	$40 \log(d) + 30 \log(f) + 49$, $d[\text{km}]$, $f[\text{Hz}]$
Path loss (D2D link, LOS)	$16.9 \log(d) + 20 \log(f/5) + 46.8$, $d[\text{m}]$, $f[\text{GHz}]$
Shadowing standard deviation	10 dB (cell mode); 12 dB (D2D mode)
Content size	50 MB
Weighting factors $\alpha = \beta$	0.5
# Malicious nodes	30% of UEs
Simulation time	100 s
# of Runs	500

and for reception: $E_i(N) = Et x_i^d(N) + Er x_i^d(N)$.

$$Et x_i^d(N) = \begin{cases} (P_{tx}^c + P_0) \sum_{j=1}^N \frac{b_j}{r_{ij}^d} & i = 1 \\ (P_{tx}^d + P_0) \sum_{j=1}^N \frac{b_j}{r_{ij}^d} & 1 < i \leq N \end{cases}$$

$$Er x_i^d(N) = \begin{cases} (P_{rx}^d + P_0) \sum_{j=i+1}^N \frac{b_j}{r_{ij}^d} & 1 \leq i < N \\ 0 & i = N \end{cases} \quad (8)$$

6. Performance evaluation

In this section we evaluate the ability of the proposed solution to cope with malicious nodes. The assessment campaign is conducted by following the system model guidelines in [23]. The main simulation parameters are listed in Table 2. A single cell where up to 20 UEs are uniformly distributed is considered and where the available radio resources are $RB = 50$.

The radio resources that can be used on a single D2D link of the multihop topology depend on the frequency reuse efficiency. For a complete investigation on this aspect, we consider the two extreme cases where either all the available RBs can be reused by all the D2D users within a coalition, the so-called *best-case*, or orthogonal RBs are to be adopted over the D2D links in a coalition, the so-called *worst-case*. In this latter case, the radio resources that can be used on a D2D transmission will be limited to the virtual resources allocated by the eNodeB to the involved pairs of UEs. All the other cases of radio resource re-use on the D2D links fall in-between these two extreme cases, which represent the lower and the upper bound for the system-level analysis. Further, we consider that 30% of the UEs are malicious, which means they drop all the incoming data from the upstreams in the cooperating topology. We compare our *trust-based* proposal with a so-called *basic* solution where the coalition formation game completely disregards the trustworthiness aspects. The analysis also evaluates the effects of the radio resource allocation policy implemented by the eNodeB, considering either a *maximum throughput (MT)* or a *proportional fair (PF)* scheduler. The performance parameters we focus on are: (i) *data loss*, (ii) *average data uploading time gain*, and (iii) *average energy consumption gain*. In particular, the latter two parameters

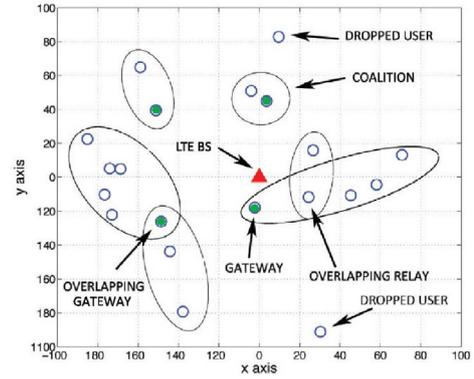


Fig. 2. Example of coalitions being formed: 20 UEs, 50 MB of data size, MT resource allocation policy.

represent the gain achieved by the cooperative upload w.r.t. a pure cellular upload modality.

In fact, as discussed in our previous work [9], all UEs in a multihop D2D data forwarding topology not only achieve uploading time gains, but in most of the cases, they also achieve energy consumption gains. In particular, also the gateway node will save energy in small coalitions when the total data in the chain is small and the transmission time on the cellular links is low. This result is interesting, since although the main objective is to achieve gain in the data uploading time, also energy saving is obtained because of the low power consumption on the D2D links.

Before illustrating the performance results, in Fig. 2 we show an example of coalitions, formed by applying the proposed scheme with 20 UEs uniformly deployed within the LTE cell (MT scheduler is adopted, the packet size is set to 50 MB and the best-case radio resource allocation is considered). As we can observe, two users are dropped by the packet scheduler, whereas the remaining UEs are grouped in coalitions. The UEs highlighted in green represent the gateway nodes appointed to upload the data from the UEs belonging to the respective coalitions. In addition, we observe that there are two overlapping coalitions sharing an overlapping relay and an overlapping gateway.

Next, we focus on the first analysis which shows the impact of the feasibility threshold in the coalition formation process in a study case with 20 UEs. In particular, we consider three different values for the FT parameter, which, we recall, gives the minimum level of trust required for each D2D link within a coalition. As we can observe from Fig. 3, the data loss strongly depends on this metric since for higher FT values the proposed solution is able to better filter out the malicious nodes in the network coalition. Noteworthy, in all cases our solution outperforms the basic solution. Focusing on the best-case, with a threshold equal to 0.6 we achieve a reduction of 45% and 34% w.r.t. the basic approach for the PF and MT schedulers respectively. This improvement reaches up to 86% and 68% if the FT value is set to 0.8. A similar trend is observed also when considering the worst-case analysis where, with an $FT = 0.6$ we are able to achieve a data loss reduction of 52% and 56% for the PT and MT scheduler, whereas a maximum gain of 80% and 79% is obtained when the FT is set to 0.8.

In Fig 4(a) and (b), instead, results in terms of content uploading time are reported. Also for this parameter, the gain for the devices in using the mobile networking in proximity paradigm w.r.t. to the cellular mode data uploading is higher for the proposed social-based trusted solution. In particular, compared to the basic solution this gain is up to 18% and 15% higher, for the best- and worst-case respectively, when $FT = 0.8$. The motivation of this result is that when malicious nodes are involved in a coalition, all the users within the coalition achieve an uploading time gain that

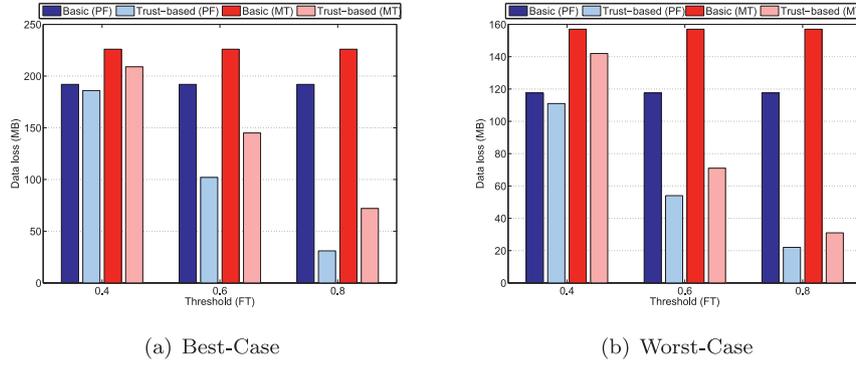


Fig. 3. Data loss in scenarios with 20 UEs.

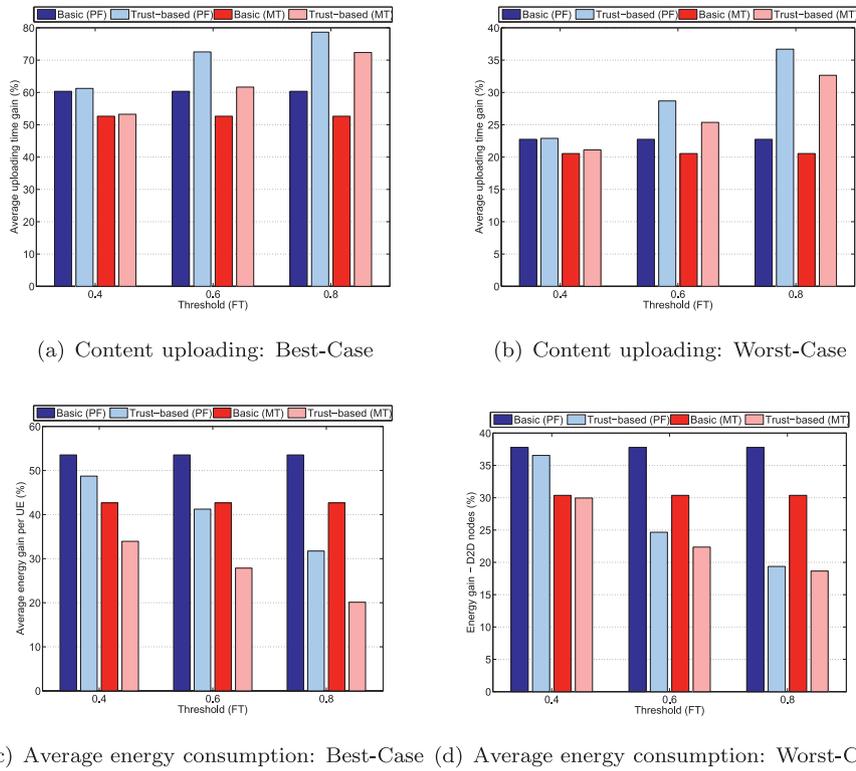


Fig. 4. Average gains in multihop content uploading (20 UEs).

is equal to zero since no data are uploaded to the eNodeB. Considering our trust-based approach, users tend to form coalition only among trusted devices increasing the probability for the content to successfully reach the eNodeB.

When considering the average energy consumption gain for the nodes in the scenario, we observe in Fig. 4(c) and (d) that the trust-based solution actually has lower energy savings w.r.t. to the basic solution. This result should not be surprising since, with the trust-based solution, the malicious nodes are isolated and need to upload their content using more energy demanding unicast transmissions towards the eNodeB. As a consequence, the positive effects of adopting low power D2D communications among the nodes is limited.

6.1. Analysis by varying the number of UEs

The next analysis shows the results by varying the number of UEs in the range of [2 – 20], when setting the FT parameter to a sample value of 0.6. When considering the data loss parameter, in Fig. 5 we can observe that the gap between the basic and the trust-

based solution increases linearly with the number of UEs for both the considered schedulers.

When considering the average uploading gain achieved by the users, the trust-based solution overcomes the basic solution in all cases. As plotted in Fig. 6(a) and (b), the gain increment for 20 UEs with the proposed solution considering the PF and MT scheduler is equal to 15% and 8% when considering the best-case, whereas it has a value of 6% and 5% for the worst-case analysis.

Finally, the energy consumption gain results plotted in Fig. 6(c) and (d) confirm the trend already observed in Fig. 4. In fact, the trust-based solution experiences lower energy consumption gains when the number of UEs is high. The motivation again is that the malicious nodes are isolated and work with the energy demanding cellular mode uploading, lowering the benefits of low power D2D communications in the coalitions.

7. Conclusion

In this paper a novel trust-based solution has been proposed to model effective cooperative content uploading in cellular environ-

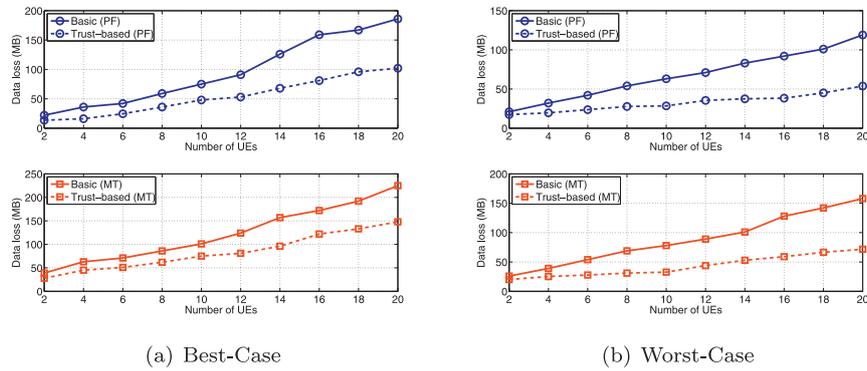


Fig. 5. Data loss by varying the number of UEs.

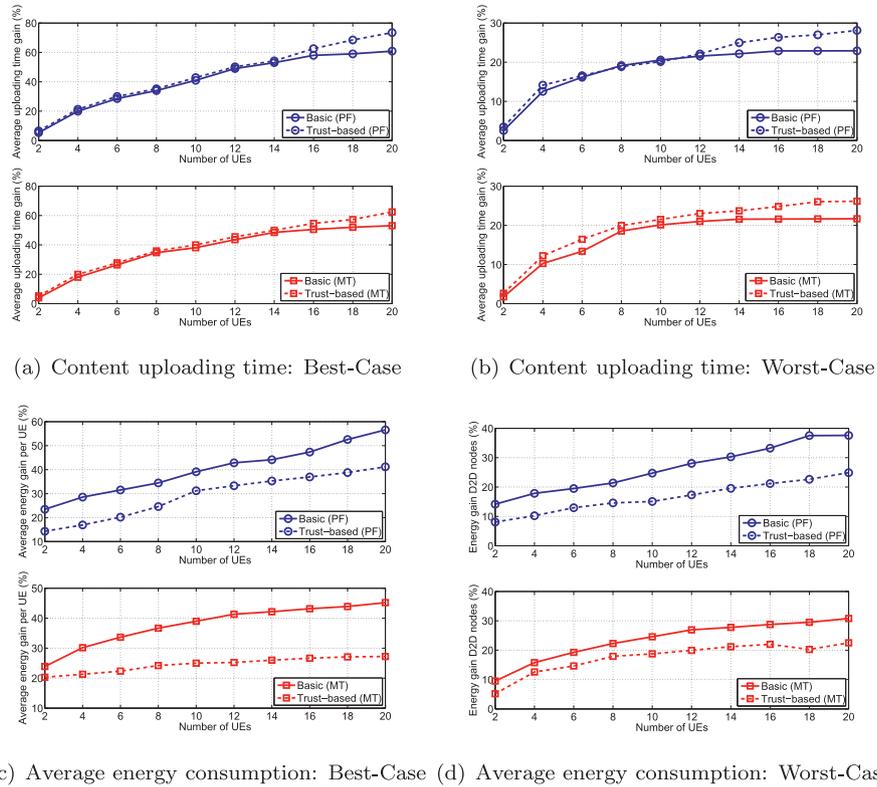


Fig. 6. Average gains in multihop content uploading (20 UEs).

ments based on D2D proximity communications. Social-awareness for the devices has been considered to evaluate the reliability for the nodes and to suitably weight the recommendations exchange for the reputation definition. A constrained coalition formation game has been studied where both coverage constraints and trust constraints are considered. This allowed implementation of a cooperative and opportunistic hop-by-hop forwarding exploiting low energy consuming and high data rate D2D links and at the same time to filter out malicious nodes in the network. The cooperating devices are assumed to be rational self-interested players aiming at maximizing their uploading time gain w.r.t. cellular mode transmissions. A simulative analysis validated the proposed solution in a variety of scenarios showing how the social based trusted solution guarantees higher gains in the content uploading time and has the ability to increase the amount of successful cooperative interactions reducing the amount of data losses in the network. In our future work we will investigate the proposed solution in other networking contexts, where D2D communications in unlicensed bands are used and where malicious nodes may also forward corrupted

data instead of simply dropping the data. In this context, innovative technologies such as *blockchain* [29], acting as a universal digital ledger to store the security and trustworthiness, may replace the centralized authority role played by the eNodeB in a cellular environment by relying, instead, on a decentralized consensus.

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