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VEHICULAR COMMUNICATIONS FOR 5G COOPERATIVE SMALL CELL NETWORKS

Xiaohu Ge, Hui Cheng, Guoqiang Mao, Yang Yang, Song Tu

Abstract—The cooperative transmission is an effective approach for vehicular communications to improve the wireless transmission capacity and reliability in the fifth generation (5G) small cell networks. Based on distances between the vehicle and cooperative small cell BSs, the cooperative probability and the coverage probability have been derived for 5G cooperative small cell networks where small cell base stations (BSs) follow Poisson point process distributions. Furthermore, the vehicular handoff rate and the vehicular overhead ratio have been proposed to evaluate the vehicular mobility performance in 5G cooperative small cell networks. To balance the vehicular communication capacity and the vehicular handoff ratio, an optimal vehicular overhead ratio can be achieved by adjusting the cooperative threshold of 5G cooperative small cell networks.

Index Terms—Vehicular communications, small cell, cooperative transmission, mobility performance

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I. INTRODUCTION

IN the future fifth generation (5G) cellular networks, denser and smaller cells are expected to provide high transmission rate for users [1]–[3]. Different with traditional personal users, vehicles are sensitive to transmission scenarios in 5G cooperative small cell networks [4], [5]. Moreover, due to the mobility nature of vehicles and the related high vehicular speed, the channel characteristics of the vehicular communications scenario can be significantly different from those of conventional wireless communication scenarios [6], [7], and make the topology of vehicular wireless networks becomes highly dynamic and prone to recurrent link intermediate [8]–[10]. In this case, cooperative transmissions are recommended as a promising solution for vehicles in 5G cooperative small cell networks [11]. However, there still exist some problems, such as the frequent handoff and coverage problems for vehicles in 5G cooperative small cell networks [12]. Therefore, it is a great challenge to investigate vehicular communications for 5G cooperative small cell networks.

To meet the communication requirements from vehicles, some studies have been investigated for vehicular communications in cellular networks [13]–[17]. Congestion and awareness control techniques have been investigated for cooperative vehicular communications which is based on wireless communications between vehicles and with other infrastructure nodes [13]. To minimize the cost of transmission or alternatively transmission time in vehicular heterogeneous networks, performing vehicular handoff is an appreciate choice at lower speeds, whereas it would be better to avoid vertical handoff and stay in the cellular network at higher speeds [14]. Based on a traffic model of two-tier cellular networks composed of macro cells and small cells, the impact that the user traffic dynamics, the mobility of users and the capacity constraint of the small cell backhaul on the system performance has been evaluated in [15]. Accounting for the vehicular mobility and network load in cellular/802.11p heterogeneous networks, an analytical model was proposed for estimating the average achievable individual throughput and an optimal handoff threshold was derived in [16]. To resolve problems resulting from limited roadside units and insufficient resources in vehicular ad hoc networks, the vehicles were configured as special vehicular small cells which have been furtherly integrated into the layered heterogeneous networks [17]. Considering that small cell base stations (BSs) are deployed at vehicles, a closed

form outage probability was derived for evaluating the user gain in two-tier cellular networks [18].

Considering the coverage of small cell is smaller, the cooperative communication is widely used for small cell networks [11], [19]–[26]. In reference [19] multi-cell multiple-input multi-output (MIMO) cooperation concepts were examined from different perspectives, including an examination of the fundamental information-theoretic limits, a review of the coding and signal processing algorithmic developments, and consideration of scalability and system-level integration. Based on random cellular networks, a general methodology was proposed to treat problems of cooperation in cellular networks, in the case where the data exchange is allowed only between pairs of nodes [20]. Taking into account the irregular BS deployment typically encountered in practice, the signal-to-interference-plus-noise ratio (SINR) distribution with cooperation was precisely characterized in a generality-preserving form and a tractable model was furtherly proposed for analyzing noncoherent joint-transmission BS cooperations [21]. To mitigate the impact of the cross-tier interference in multi-tier wireless networks, a scheme was proposed for location-aware cross-tier cooperation between BSs in different tiers for downlink coordination multipoint (CoMP) transmission in two-tier cellular networks [22]. Utilizing the average user throughput under CoMP and non-CoMP transmission after taking into account the downlink training overhead, each user was allowed to select transmission model between coherent CoMP and non-CoMP to avoid the extra overhead outweighing the cooperative gain in cellular networks [23]. To fully exploit benefits of heterogeneous networks, a radio resource allocation scheme was proposed for cooperative relays where the relay nodes with in-band backhaul act as micro BSs and are able to serve users either independently or cooperatively with macro cell BSs [24]. By defining the cooperation region as a function of the user quality of service (QoS) requirements and network load, a QoS aware cooperative downlink scheduling approach was proposed for cell-edge and handoff users that offers more reliability and higher effective capacity [25]. Using stochastic geometry-based heterogeneous cellular networks, the coverage probability, the average achievable rate and the energy efficient were derived for K -tier heterogeneous wireless networks with different cooperative sleep models for small cells [26].

However, in all the aforementioned vehicular communication studies, only simple scenarios, such as two cooperative cells with single or multiple antennas, were considered and the underlying vehicular communications were limited to simple point-to-point wireless communications. Besides, the exact coverage probability of cooperative MIMO small cell networks with co-channel interference has not been investigated. Moreover, detailed investigation of the vehicle mobility performance used for 5G cooperative MIMO small cell networks is surprisingly rare in the open literature. Motivated by above gaps, in this paper we consider the scenarios of vehicular communications for vehicle-to-infrastructure (V2I) and for urban roads, we derive the vehicular handoff rate and the vehicular overhead ratio to evaluate the vehicular mobility performance in 5G cooperative MIMO small cell networks considering co-channel interference. The contributions and

novelties of this paper are summarized as follows.

- 1) Based on distances between the vehicle and cooperative small cells, the cooperative probability and the coverage probability of cooperative small cell networks have been derived for vehicles equipped with multiply antennas.
- 2) From the proposed cooperative probability and coverage probability, the vehicle handoff rate and the vehicle overhead ratio are proposed to evaluate the vehicle mobility performance in 5G cooperative MIMO small cell networks considering co-channel interference.
- 3) Numerical results imply that there exists a minimum vehicle overhead ratio for 5G cooperative MIMO small cell networks considering different cooperative thresholds. This result can be used for optimizing vehicular communications in 5G cooperative MIMO small cell networks.

The rest of this paper is organized as follows. Section II describes the system model of 5G cooperative MIMO small cell networks where small cell BSs follow Poisson point process distributions. In Section III, the cooperative probability has been derived for 5G cooperative MIMO small cell networks. Moreover, the coverage probability has been derived for 5G cooperative MIMO small cell networks considering co-channel interference from adjacent small cells in Section IV. Furthermore, in Section V the vehicular handoff rate and the vehicular overhead ratio have been proposed to evaluate the vehicular mobility performance in 5G cooperative MIMO small cell networks. Numerical results indicate that there exists a minimum vehicular overhead ratio for 5G cooperative MIMO small cell networks considering different cooperative thresholds. Finally, Section VI concludes this paper.

II. SYSTEM MODEL

Fig. 1 shows the system model of 5G cooperative small cell networks which is a two-tier cellular network including macro cell BSs and small cell BSs. Macro cell BSs take charge of control information for vehicles and small cell BSs. Small cell BSs transmit the desired data to vehicles. In this case, macro cells form the control plane (C-plane) and small cells form the user plane (U-plane) in 5G cooperative small cell networks. Without loss of generality, both the control zone signalling and the L1/L2/L3 signalling, carried by physical downlink shared channel (PDSCH) and scheduled by physical downlink control channel (PDCCH) in the data zone of downlink sub-frames, are assumed to belong to the C-plane information. Only the user traffic data is carried by U-plane sub-frames in the C/U plane split architecture [27]. Macro cells with the same radius are assumed to be regularly deployed in the infinite plane \mathbb{R}^2 . Small cell BSs are assumed to be randomly deployed in the infinite plane \mathbb{R}^2 . Moreover, the locations of small cell BSs follow an independent Poisson point processes Φ_s with the intensity λ_s . Every small cell BS has the same transmission power P_s . In this paper, the vehicle is assumed to be associated with the closest BS, which would suffer the least path loss during wireless transmissions. Every small cell is assumed to include only one BS and a few vehicles. Then, the cell boundary, which can be obtained

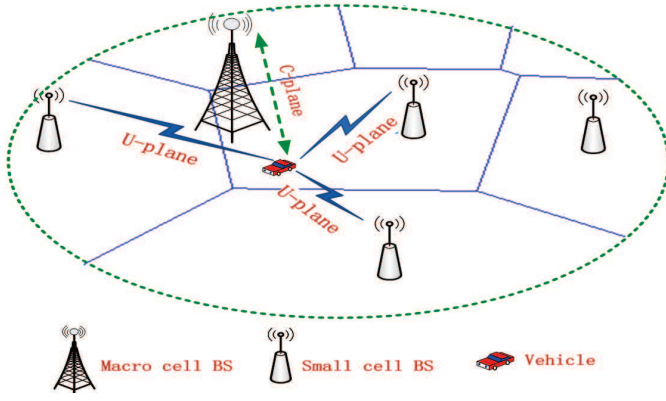


Fig. 1. System model.

through the Delaunay triangulation method by connecting the perpendicular bisector lines between each pair of small cell BSs [28], splits the plane \mathbb{R}^2 into irregular polygons that correspond to different small cell coverage areas. Such stochastic and irregular topology forms a so-called Poisson-Voronoi tessellation (PVT) [29]. An illustration of one macro cell scenario is depicted in Fig. 1, where each small cell is denoted as $\mathcal{E}_q (q = 1, 2, 3, \dots)$. Despite of its complexity, an outstanding property of PVT random small cell networks is that the geometric characteristics of any small cell \mathcal{E}_q coincide with that of a typical PVT small cell \mathcal{E}_1 , according to the Palm theory [30]. This feature implies that the analytical results for a typical PVT small cell \mathcal{E}_1 can be extended to the whole random small cell networks.

Without loss of generality, the initial location of the vehicle UE_0 located at \mathcal{E}_1 is assumed as the origin position. The distance between the UE_0 and the closest small cell BS BS_1 is denoted as R_1 . Moreover, the distance between the UE_0 and the i -th closest small cell BS BS_i is denoted as $R_i (i = 2, 3, 4, \dots)$. In this paper, adjacent small cell BSs can cooperatively transmit data to a specified vehicle. Moreover, this paper is focused on downlinks of 5G cooperative small cell networks.

III. COOPERATIVE PROBABILITY IN SMALL CELL NETWORKS

In this paper, cooperative small cell BSs is selected by distances between the vehicle UE_0 and adjacent small cell BSs. How to evaluate the distance distribution of cooperative small cell BSs is the basis for the cooperative transmission of small cell networks.

A. Distance distribution of cooperative small cell BSs

In a homogeneous M -dimensionality Poisson point process with intensity λ , the probability of finding N nodes in a bounded Borel space $\mathcal{A} \subset \mathbb{R}^M$ is expressed as

$$P_r[N \text{ nodes in } \mathcal{A}] = e^{-\lambda \mathcal{A}} \frac{(\lambda \mathcal{A})^N}{N!}. \quad (1)$$

For a homogeneous two-dimensionality Poisson point process with intensity λ and $\mathcal{A} = \pi r^2$, the distance R_n between a point

and its n -th closest point is governed by the generalized Gamma distribution

$$f_{R_n}(r) = e^{-\lambda \pi r^2} \frac{2(\lambda \pi r^2)^n}{r \Gamma(n)} \quad (2)$$

where $\Gamma(\cdot)$ is the Gamma function.

Corollary 1 [31]: Let $y \in \mathbb{R}^2$, and let $X_i \in \mathbb{R}^2$ be the points of a homogeneous point process of intensity λ in \mathbb{R}^2 plane ordered according to their Euclidean distance to y . Then $R_i := \|y - X_i\|^2$ has the same distribution as the one-dimensional Poisson process of intensity $\lambda \pi$, the expectation and cumulative distribution function (CDF) of R_i is expressed as

$$\mathbb{E}[R_i] = i/(\lambda \pi), \quad (3a)$$

$$F_{R_i}(r) = 1 - \frac{\Gamma_{ic}(i, \lambda \pi r^2)}{\Gamma(i)}, \quad (3b)$$

where $\Gamma_{ic}(\cdot, \cdot)$ is the incomplete Gamma function. When the differential is operated on (3b), the probability density function (PDF) of R_i is derived by

$$f_{R_i}(r) = \frac{2e^{-\lambda \pi r^2} (\lambda \pi r^2)^i}{r \Gamma(i)}. \quad (4)$$

B. Cooperative probability

Since the radius of 5G small cells is usually less than 100 meters (m), the vehicle has to frequently execute the handoff operation when the high speed vehicle is only associated with one small cell BS. Even so, it is a great challenge to keep the wireless link reliability for vehicular communications in 5G small cell networks. To solve these problems, the cooperative transmission based on adjacent small cell BSs is a promising candidate. In this paper, cooperative small cell BSs is selected according to the following cooperative scheme. Considering that the radius of small cell is much less than the radius of macro cells, the wireless link is assumed be line of sight (LOS) transmission in this study. To simplify derivations, the path loss and Rayleigh fading are considered but the shadowing effect is ignored in wireless channels, as commonly done in the area [32], [33].

Cooperative scheme: When the ratio of the distance R_i to the distance R_1 is less than or equal to the given cooperative threshold ρ , the small cell BS BS_i being R_i apart from the vehicle UE_0 is selected for cooperative transmissions, which can be expressed as

$$\frac{R_i}{R_1} \leq \rho. \quad (5)$$

Therefore, the cooperative probability of the BS BS_i is expressed as

$$\begin{aligned} P_r\left(\frac{R_i}{R_1} \leq \rho\right) &= \int_0^{+\infty} P_r(R_i \leq \rho y, R_1 = y) dy \\ &= \int_0^{+\infty} P_r(R_i \leq \rho y | R_1 = y) P_r(R_1 = y) dy, \quad (6) \\ &= \int_0^{+\infty} \left[\sum_{k=i-1}^{+\infty} e^{-\lambda_s \mathcal{D}} \cdot \frac{(\lambda_s \mathcal{D})^k}{k!} \right] f_{R_1}(y) dy \end{aligned}$$

where $\mathcal{D} = \pi(\rho y)^2 - \pi y^2$ is the area between circles with different radii of R_1 and ρR_1 . Substitute (4) into (6), the

$$P_r\left(\frac{R_i}{R_1} \leq \rho\right) = \int_0^\infty \left\{ 1 - e^{-\lambda_s[\pi(\rho y)^2 - \pi y^2]} \cdot \frac{\Gamma(i-1, \lambda_s[\pi(\rho y)^2 - \pi y^2])}{\Gamma(i-1)} \right\} \cdot 2\lambda_s \pi y e^{-\lambda_s \pi y^2} dy. \quad (7)$$

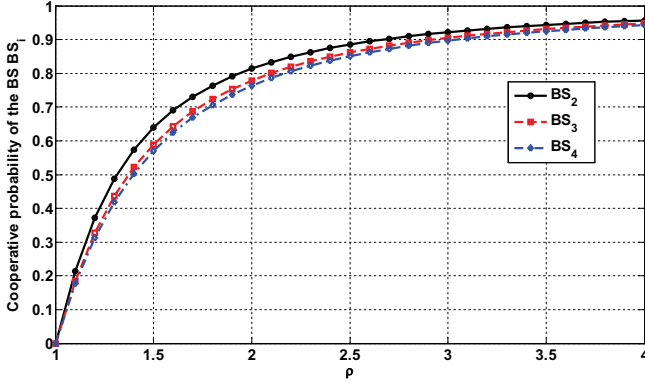


Fig. 2. Cooperative probability with respect to the cooperative threshold considering different cooperative small cell BSs.

cooperative probability of the BS BS_i is further derived by formula (7).

When k small cell BSs are closed with the vehicle UE_0 , the cooperative probability of k adjacent small cell BSs is derived by

$$\begin{aligned} \mathbb{P}_k &= P_r\left(\frac{R_k}{R_1} \leq \rho \cap \frac{R_{k+1}}{R_1} > \rho\right) \\ &= \int_0^{+\infty} P_r(R_k \leq \rho y, R_{k+1} > \rho y, R_1 = y) dy \\ &= \int_0^{+\infty} P_r(R_k \leq \rho y, R_{k+1} > \rho y | R_1 = y) P_r(R_1 = y) dy \\ &= \int_0^{+\infty} e^{-\lambda_s \pi y^2} \cdot \frac{(\lambda_s \pi)^{k-1}}{(k-1)!} \cdot f_{R_1}(y) dy \\ &= \int_0^{+\infty} e^{-\lambda_s [\pi(\rho y)^2 - \pi y^2]} \cdot \frac{[\lambda_s \pi (\rho y)^2 - \lambda_s \pi y^2]^{k-1}}{(k-1)!} 2\lambda_s \pi y e^{-\lambda_s \pi y^2} dy \\ &= \int_0^{+\infty} 2\lambda_s \pi y e^{-\lambda_s \pi (\rho y)^2} \cdot \frac{[\lambda_s \pi (\rho y)^2 - \lambda_s \pi y^2]^{k-1}}{(k-1)!} dy \end{aligned} \quad (8)$$

C. Performance analysis of cooperative probability

To validate the proposed cooperative probability, some performance analysis is simulated by numerical results in Fig. 2 and Fig. 3. The intensity of small cell BSs is configured as $\lambda_s = 1/(\pi \times 50^2)$. Fig. 2 shows the impact of the cooperative threshold ρ on the cooperative probability of the BS BS_i . When a small cell BSs is selected, the cooperative probability of the BS BS_i increases with the increase of the cooperative threshold ρ . In this paper, the cooperative small cell BSs are ordered by the distance between the BS BS_i and the vehicle UE_0 . When the threshold ρ is fixed, the cooperative probability of the BS BS_i decreases with the increase of the distance between the BS BS_i and the vehicle UE_0 . When the threshold ρ is larger than 3.5, the cooperative probability of the BS BS_i , ($i = 2, 3, 4$), approaches a saturated value.

Fig. 3 illustrates the impact of the number of cooperative small cell BSs and the cooperative threshold ρ on the cooperative probability of small cell BSs. When the number of cooperative small cell BSs is fixed as 1, i.e., only one small cell BS is selected for cooperative transmissions, the cooperative

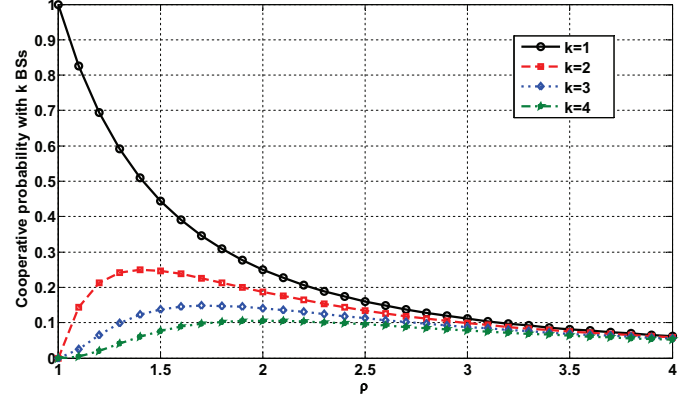


Fig. 3. Cooperative probability with respect to the number of cooperative small cell BSs and the cooperative threshold ρ .

probability monotonously decreases with the increase of the cooperative threshold ρ . When the number of cooperative small cell BSs is larger than 1, the cooperative probability first increases with the increase of the cooperative threshold ρ . When the cooperative probability achieves the maximum, the cooperative probability decreases with the increase of the cooperative threshold ρ . In the end, the cooperative probability approaches to a saturated value when the threshold ρ is larger than 4. When the threshold is fixed, the cooperative probability decreases with the increase of the number of cooperative small cell BSs.

IV. COVERAGE PROBABILITY OF COOPERATIVE SMALL CELL NETWORKS

A. Interference Model

When the orthogonal frequency division multiplexing (OFDM) scheme is assumed to be adopted by small cell BSs to support multi-user transmission in a small cell, there is not co-channel interference generated from the intra-cell in this paper. For the vehicle UE_0 , no more than one co-channel interfering vehicle is assumed to exist in each adjacent small cell. The vehicle UE_0 is interfered by downlinks of co-channel vehicles in the adjacent small cells, which is transmitted from their associated small cell BSs. The small cell BS is equipped with n_t antennas and the vehicle is equipped with n_r antennas. Hence, in this paper the vehicular communication is a type of MIMO communications. Without loss of generality, for the vehicle UE_0 , $\mathcal{C} \subset \Phi_s$ is the cooperative small cell BSs set which can simultaneously transmit data to a given vehicle, $\mathcal{B} \subset \Phi_s \setminus \mathcal{C}$ is the interfering small cell BSs set. Considering the cooperative transmission from adjacent small cell BSs, the received signal at the vehicle UE_0 is expressed as

$$y = \sum_{i \in \mathcal{C}} \frac{\sqrt{P_s}}{R_i^{\eta/2}} \mathbf{H}_{i0} \mathbf{x}_i + \sum_{j \in \mathcal{B}} \frac{\sqrt{P_s}}{R_j^{\eta/2}} \mathbf{H}_{j0} \mathbf{x}_j + \mathbf{Z}, \quad (9)$$

where $\mathbf{y} \in \mathbb{C}^{n_r \times 1}$ is the received signal vector at the vehicle UE_0 , R_i is the distance between the vehicle UE_0 and cooperative small cell BSs, R_j is the distance between the vehicle UE_0 and interfering small cell BSs, η is the path loss coefficient, $\mathbf{x}_i \in \mathbb{C}^{n_t \times 1}$ is the desired signal vector from the cooperative transmission small cell BS BS_i , $\mathbf{x}_j \in \mathbb{C}^{n_t \times 1}$ is the interfering signal vector from the adjacent interfering small cell BS BS_j , $\mathbf{Z} \in \mathbb{C}^{n_t \times 1}$ is the additive white Gaussian noise (AWGN) with variance σ^2 in wireless channels. $\mathbf{H}_{i0} \in \mathbb{C}^{n_r \times n_t}$ is the small scale fading channel matrix between the vehicle UE_0 and the cooperative small cell BS BS_i , $h_{i,m,n}$ ($m = 1, 2, \dots, n_r; n = 1, 2, \dots, n_t$) is the element of the channel matrix \mathbf{H}_{i0} and is governed by a complex Gaussian distribution, i.e., $h_{i,m,n} \sim \mathcal{CN}(0, 1)$, and its magnitude $|h_{i,m,n}|$ is a Rayleigh-distributed random variable [22], where $h_{i,m,n}$ is the channel coefficient between the m -th receiving antenna at the vehicle UE_0 and the n -th transmission antenna at the cooperative small cell BS BS_i ; $\mathbf{H}_{j0} \in \mathbb{C}^{n_r \times n_t}$ is the small scale fading channel matrix between the vehicle UE_0 and the interfering small cell BS BS_j , $h_{j,m,n}$ ($m = 1, 2, \dots, n_r; n = 1, 2, \dots, n_t$) is the element of the channel matrix \mathbf{H}_{j0} and is governed by a complex Gaussian distribution, i.e., $h_{j,m,n} \sim \mathcal{CN}(0, 1)$, and its magnitude $|h_{j,m,n}|$ is Rayleigh-distributed random variables, where $h_{j,m,n}$ is the channel coefficient between the m -th receiving antenna at the vehicle UE_0 and the n -th transmission antenna at the interfering small cell BS BS_j .

When k cooperative small cell BSs are assumed to jointly transmit data to the vehicle UE_0 , and considering the maximum ratio transmission /maximum ratio combining (MRT/MRC) [34], [35], the SINR received by the vehicle UE_0 is derived by

$$\begin{aligned} \text{SINR}^c &= \frac{P_s \sum_{i \in \mathcal{C}} R_i^{-\eta} |\mathbf{H}_{i,0}|^2}{\sigma^2 + \sum_{j \in \mathcal{B}} P_s R_j^{-\eta} |\mathbf{H}_{j,0}|^2} \\ &\approx \frac{\frac{P_s}{n_t} \sum_{i=1}^k R_i^{-\eta} \lambda_{\max}(\mathbf{H}_{i,0} \mathbf{H}_{i,0}^H)}{\sigma^2 + \frac{P_s}{n_t} \sum_{j=k+1}^{\infty} R_j^{-\eta} \left(\sum_{m=1}^{n_r} \sum_{n=1}^{n_t} |h_{j,m,n}|^2 \right)} \quad (10) \\ &\approx \frac{\frac{P_s}{n_t} \sum_{i=1}^k R_i^{-\eta} \left(\sum_{m=1}^{n_r} \sum_{n=1}^{n_t} |h_{i,m,n}|^2 \right)}{\sigma^2 + \frac{P_s}{n_t} \sum_{j=k+1}^{\infty} R_j^{-\eta} \left(\sum_{m=1}^{n_r} \sum_{n=1}^{n_t} |h_{j,m,n}|^2 \right)} \end{aligned}$$

Where $\lambda_{\max}(\mathbf{H}_{i,0} \mathbf{H}_{i,0}^H)$ is the maximum singular value of the matrix $\mathbf{H}_{i,0} \mathbf{H}_{i,0}^H$. Furthermore, the interference aggregated at the vehicle UE_0 can be expressed as

$$I_{agg} = \frac{P_s}{n_t} \sum_{j=k+1}^{\infty} R_j^{-\eta} \left(\sum_{m=1}^{n_r} \sum_{n=1}^{n_t} |h_{j,m,n}|^2 \right). \quad (11)$$

To simplify the derivation, let $g_i = \sum_{m=1}^{n_r} \sum_{n=1}^{n_t} |h_{i,m,n}|^2$ and $\Phi \triangleq \{R_j | j \in \mathcal{B}\}$. From the distribution of $h_{i,m,n}$, we can derive the PDF of g_i is $f_{g_i}(x) = \frac{x^{n_t n_r - 1}}{\Gamma(n_t n_r)} e^{-x}$. Based on the mapping theorem, Φ is an inhomogeneous Poisson point process with intensity $\lambda(r) = 2\pi\lambda_s r$ [36]. As a consequence, the Laplace transform of the aggregate interference at the

vehicle UE_0 is derived by

$$\begin{aligned} \mathcal{L}_{I_{agg}}(s) &= \mathbb{E}(e^{-s I_{agg}}) \\ &= \mathbb{E}[\exp(-s \sum_{j=k+1}^{\infty} \frac{P_s}{n_t} R_j^{-\eta} (\sum_{m=1}^{n_r} \sum_{n=1}^{n_t} |h_{j,m,n}|^2))] \\ &= \mathbb{E}_{\Phi, g_j} \left\{ \prod_{j>k} [\exp(-s \frac{P_s}{n_t} R_j^{-\eta} g_j)] \right\} \\ &\stackrel{(a)}{=} \exp[-2\pi\lambda_s \int_{r>R_k} \mathbb{E}_{g_j} (1 - e^{-s \frac{P_s}{n_t} r^{-\eta} g_j}) r dr] \\ &= \exp[-2\pi\lambda_s \int_{r>R_k} (\int_0^{\infty} \frac{g^{n_t n_r - 1}}{\Gamma(n_t n_r)} e^{-g} (1 - e^{-s \frac{P_s}{n_t} r^{-\eta} g}) dg) r dr] \\ &= \exp[\frac{-2\pi\lambda_s}{\Gamma(n_t n_r)} \int_{r>R_k} (1 - \int_0^{\infty} g^{n_t n_r - 1} e^{-(s \frac{P_s}{n_t} r^{-\eta} + 1)g} dg) r dr] \\ &= \exp[-2\pi\lambda_s \int_{r>R_k} (1 - \frac{1}{(1 + s \frac{P_s}{n_t} r^{-\eta})^{n_t n_r}}) r dr] \quad (12) \end{aligned}$$

where $\mathbb{E}(\cdot)$ is the expectation operation, (a) is due to the probability generating functional for a PPP.

B. Coverage probability

For cooperative transmissions of small cell BSs, the coverage of cooperative small cell BSs can be extended from every coverage of cooperative small cell BS. The extended coverage of cooperative small cell BSs can provide for a better reliable link service for vehicular communications. When the outage threshold is configured as ε for vehicle links, the coverage probability of k cooperative small cell BSs is expressed as

$$\begin{aligned} \mathbb{P}_c^k &= P_r(\text{SINR}^c > \varepsilon) \\ &= P_r \left(\frac{\frac{P_s}{n_t} \sum_{i=1}^k R_i^{-\eta} g_i}{\sigma^2 + I_{agg}} > \varepsilon \right). \quad (13) \end{aligned}$$

However, the analytical expression can not be derived for (13) when the distance R_i is a random variable. In the most cases, cooperative transmissions are related with the limited adjacent small cell BSs in 5G cooperative small cell networks. Moreover, the number of cooperative small cell BSs is less than or equal to 5 in realistic scenarios. When the vehicle UE_0 is assumed to be located at the edge of small cells, the distance between the vehicle UE_0 and cooperative small cell BSs is approximated to be equal. Therefore, the distance between the vehicle UE_0 and cooperative small cell BSs is configured as D in the following derivations. To simplify the derivation, the transmission power of small cell BS is normalized as 1 and the noise is ignored in wireless channels considering that the noise power is obviously less than the desired signal power and the interference power [37]. As a consequence, the coverage probability of k cooperative small cell BSs is further derived by

$$\begin{aligned} \mathbb{P}_c^k &= P_r \left(\sum_{i=1}^k R_i^{-\eta} g_i > \varepsilon I_{agg} \right) \\ &\stackrel{(a)}{=} P_r(D^{-\eta} \sum_{i=1}^k g_i > \varepsilon I_{agg}) \\ &\stackrel{(b)}{=} \mathbb{E}_D \left\{ \mathbb{E}_{I_{agg}} \left[\sum_{n=0}^{kn_t n_r - 1} \frac{(\varepsilon D^\eta)^n}{n!} I_{agg}^n e^{-\varepsilon D^\eta I_{agg}} \right] \right\} \quad (14) \\ &\stackrel{(c)}{=} \mathbb{E}_D \left[\sum_{n=0}^{kn_t n_r - 1} \frac{(-\varepsilon D^\eta)^n}{n!} \mathcal{L}_{I_{agg}}^{(n)}(\varepsilon D^\eta) \right] \end{aligned}$$

where the condition (a) is the assumption that the distance between the vehicle UE_0 and the cooperative small cell

BSs is equal, the condition (b) is based on the CDF of Gamma distribution, the condition (c) is the Laplace transform property $\mathbb{E}_{I_{agg}}[I_{agg}^n e^{-s I_{agg}}] = (-1)^n \frac{d^n}{ds^n} \mathcal{L}_{I_{agg}}(s)$. Let $x_n = \frac{(-\varepsilon D^\eta)^n}{n!} \mathcal{L}_{I_{agg}}^{(n)}(\varepsilon D^\eta)$, (14) is simply expressed by

$$\mathbb{P}_c^k = \mathbb{E}_D \left[\sum_{n=0}^{kn_t n_r - 1} x_n \right] = \mathbb{E}_D \left[x_0 + \sum_{n=1}^{kn_t n_r - 1} x_n \right]. \quad (15)$$

Based on (12), the differentiation of the aggregate interference at the vehicle UE_0 is derived by formula (16).

Furthermore, the n -th order derivative of (12) is derived by formula (17).

Let $v^{-\eta/2} = su^{-\eta/2}$, then $du = s^{\eta/2} dv$. (17) can be further derived by

$$\begin{aligned} \mathcal{L}_{I_{agg}}^{(n)}(s) &= \pi \lambda_s \sum_{i=0}^{n-1} C_{n-1}^i \cdot (-1)^{n-i} \cdot \frac{(n-i-1+n_t n_r)!}{(n_t n_r - 1)!} s^{2/\eta - n + i} \\ &\times \int_{\varepsilon^{-2/\eta}}^{\infty} \frac{(v^{-\eta/2})^{n-i}}{(1+v^{-\eta/2})^{n-i+n_t n_r}} dv \times \mathcal{L}_{I_{agg}}^{(i)}(s) \end{aligned} \quad (18)$$

Substitute $s = \varepsilon D^\eta$ into (12), the following expression is derived as

$$x_0 = \mathcal{L}_{I_{agg}}(s) = \exp(-\pi \lambda_s k_0 D^2), \quad (19a)$$

with

$$k_0 = \varepsilon^{2/\eta} \int_{\varepsilon^{-2/\eta}}^{\infty} \left(1 - \frac{1}{(1+v^{-\eta/2})^{n_t n_r}} \right) dv. \quad (19b)$$

When $n \geq 1$, the following expression is derived by formula (20a-b).

Based on (20a), a linear recurrence relation of x_n is derived for the explicit expression of the coverage probability via linear algebra. Let

$$\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r} = [x_1, x_2, \dots, x_{kn_t n_r}]^T, \quad (21a)$$

$$\begin{aligned} \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} &= [y_1, y_2, \dots, y_{kn_t n_r}]^T \\ &= [n_t n_r k_1, \frac{n_t n_r (1+n_t n_r)}{2} k_2, \dots, \frac{n_t n_r C_{kn_t n_r - 1}^{n_t n_r}}{kn_t n_r} k_{kn_t n_r}]^T, \end{aligned} \quad (21b)$$

then (20) can be represented in a matrix form as formula (22a-c). Since $\mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r}$ is a strictly lower triangular matrix, we have $\mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r}^n = 0, n \geq kn_t n_r$. According to this property, after iterating, $\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r}$ can be rewritten as

$$\begin{aligned} \mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r} &= a x_0 \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} + a \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r} \mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r} \\ &= a x_0 \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} + a \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r} (a x_0 \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} + a \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r} \mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r}) \\ &= a x_0 \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} + a^2 x_0 \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r} \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} + a^2 \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r}^2 \mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r} \\ &= \dots \\ &= \sum_{n=1}^{kn_t n_r} a^n x_0 \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r}^{n-1} \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} \end{aligned} \quad (23)$$

Similarly, denote $\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r - 1} = [x_1, x_2, \dots, x_{kn_t n_r - 1}]^T$, then

$$\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r - 1} = \sum_{n=1}^{kn_t n_r - 1} a^n x_0 \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r - 1}^{n-1} \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r - 1}. \quad (24)$$

Define $\text{sum}(\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r - 1}) = \sum_{n=1}^{kn_t n_r - 1} x_n$, which is the sum of elements in the vector $\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r - 1}$. In the end, the coverage

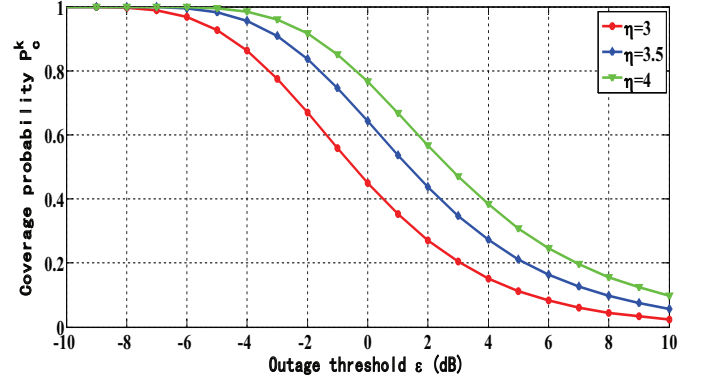


Fig. 4. Coverage probability with respect to the outage threshold considering different path loss coefficients.

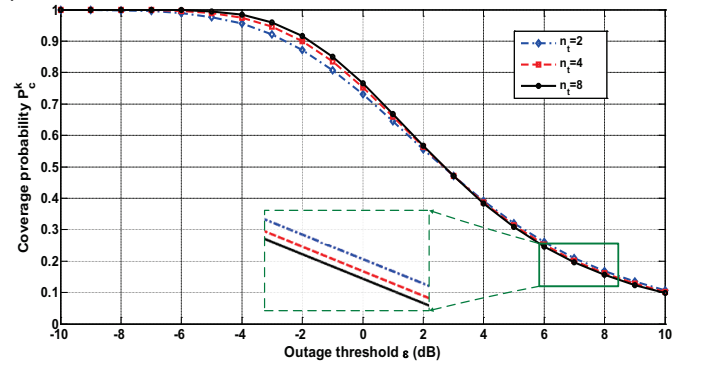


Fig. 5. Coverage probability with respect to the number of transmission antennas at cooperative small cell BSs considering different outage thresholds.

probability of k cooperative small cell BSs can be expressed as an explicit form as (25).

Without loss of generality, three small cell cooperative BSs scenario is considered as follows. In this case, the distance D between the vehicle UE_0 and cooperative small cell BSs is given by [37]

$$f_D(D) = 2(\lambda\pi)^2 D^3 e^{-\lambda\pi D^2}. \quad (26)$$

Substitute (26) into (25), the coverage probability of three cooperative small cell BSs is simply derived by (27).

C. Performance analysis of coverage probability

Based on the proposed coverage probability of three cooperative small cell BSs in (27), some performance evaluations are numerically analyzed in Fig. 4–Fig. 7. In the following coverage probability analysis, default parameters are configured as follows: the antenna number at the small cell BS is $n_t = 4$, the antenna number at the vehicle is $n_r = 2$, the small cell BS transmission power is normalized as 1, the path loss coefficient is $\eta = 4$, the intensity of small cell BSs is $\lambda_s = 1/(\pi \times 50^2)$, the outage threshold of vehicle links is $\varepsilon = 0$ dB [38], the radius of small cell is 50 m.

$$\begin{aligned}
\mathcal{L}_{I_{agg}}^{(1)}(s) &= \frac{d}{ds} L_{I_{agg}}(s) = \frac{d}{ds} \exp \left(-\pi \lambda_s \int_{R_k^2}^{\infty} \left(1 - \frac{1}{(su^{-\eta/2}+1)^{n_t n_r}} \right) du \right) \\
&= \exp \left(-\pi \lambda_s \int_{R_k^2}^{\infty} \left(1 - \frac{1}{(su^{-\eta/2}+1)^{n_t n_r}} \right) du \right) \cdot \left(-\pi \lambda_s \int_{R_k^2}^{\infty} \left(1 - \frac{1}{(su^{-\eta/2}+1)^{n_t n_r}} \right) du \right)_s^{(1)} \\
&= \exp \left(-\pi \lambda_s \int_{R_k^2}^{\infty} \left(1 - \frac{1}{(su^{-\eta/2}+1)^{n_t n_r}} \right) du \right) \cdot \left(-\pi \lambda_s \int_{R_k^2}^{\infty} \frac{n_t n_r u^{-\eta/2}}{(su^{-\eta/2}+1)^{n_t n_r+1}} du \right) \\
&= \underbrace{\left(-\pi \lambda_s \int_{R_k^2}^{\infty} \frac{n_t n_r u^{-\eta/2}}{(su^{-\eta/2}+1)^{n_t n_r+1}} du \right)}_{\omega(s)} \times \underbrace{\mathcal{L}_I(s)}_{v(s)}. \tag{16}
\end{aligned}$$

$$\begin{aligned}
\mathcal{L}_{I_{agg}}^{(n)}(s) &= \left(\mathcal{L}_I^{(1)}(s) \right)^{(n-1)} = \sum_{i=0}^{n-1} C_{n-1}^i \omega^{(n-1-i)} v^{(i)}(s) \\
&= \sum_{i=0}^{n-1} C_{n-1}^i \omega^{(n-1-i)} \times \mathcal{L}_I^{(i)}(s) \\
&= \pi \lambda_s n_t n_r \sum_{i=0}^{n-1} C_{n-1}^i \left(-\int_{R_k^2}^{\infty} \left(u^{-\eta/2} (1 + su^{-\eta/2})^{-n_t n_r - 1} \right) du \right)^{(n-1-i)} \times \mathcal{L}_I^{(i)}(s) \\
&= \pi \lambda_s n_t n_r \sum_{i=0}^{n-1} C_{n-1}^i \left(-\int_{R_k^2}^{\infty} \left(u^{-\eta/2} (-1)^{n-i-1} \frac{(n-i-1+n_t n_r)!}{(n_t n_r)!} \right. \right. \\
&\quad \left. \left. \times (1 + su^{-\eta/2})^{-n_t n_r - 1 - n + i + 1} (u^{-\eta/2})^{n-i-1} \right) du \right) \times \mathcal{L}_I^{(i)}(s) \\
&= \pi \lambda_s \sum_{i=0}^{n-1} C_{n-1}^i (-1)^{n-i} \frac{(n-i-1+n_t n_r)!}{(n_t n_r - 1)!} \left(\int_{R_k^2}^{\infty} \left(\frac{(u^{-\eta/2})^{n-i}}{(1 + su^{-\eta/2})^{n-i+n_t n_r}} \right) du \right) \times \mathcal{L}_I^{(i)}(s)
\end{aligned} \tag{17}$$

$$\begin{aligned}
x_n &= \frac{s^n}{n!} (-1)^n \mathcal{L}_{I_{agg}}^{(n)}(s) \\
&= \pi \lambda_s \sum_{i=0}^{n-1} C_{n-1}^i (-1)^i \frac{(n-i-1+n_t n_r)!}{n! (n_t n_r - 1)!} s^{\frac{2}{\eta} + i} \left(\int_{\varepsilon^{-2/\eta}}^{\infty} \frac{(\nu^{-\frac{\eta}{2}})^{n-i}}{(1 + \nu^{-\frac{\eta}{2}})^{n-i+n_t n_r}} dv \right) \mathcal{L}_{I_{agg}}^{(i)}(s) \\
&= \pi \lambda_s s^{\frac{2}{\eta}} \sum_{i=0}^{n-1} C_{n-1}^i \frac{i! (n-i-1+n_t n_r)!}{n! (n_t n_r - 1)!} \left(\int_{\varepsilon^{-2/\eta}}^{\infty} \frac{(\nu^{-\frac{\eta}{2}})^{n-i}}{(1 + \nu^{-\frac{\eta}{2}})^{n-i+n_t n_r}} dv \right) (-1)^i \frac{s^i}{i!} \mathcal{L}_{I_{agg}}^{(i)}(s) \\
&= \pi \lambda_s s^{\frac{2}{\eta}} \sum_{i=0}^{n-1} \frac{(n-1)!}{i! (n-i-1)!} \frac{i! (n-i-1+n_t n_r)!}{n! (n_t n_r - 1)!} \left(\int_{\varepsilon^{-2/\eta}}^{\infty} \frac{(\nu^{-\frac{\eta}{2}})^{n-i}}{(1 + \nu^{-\frac{\eta}{2}})^{n-i+n_t n_r}} dv \right) x_i \\
&\stackrel{s=\varepsilon D^\eta}{=} \pi \lambda_s D^2 \sum_{i=0}^{n-1} \frac{n_t n_r C_{n-1-i+n_t n_r}^{n_t n_r}}{n} \varepsilon^{2/\eta} \left(\int_{\varepsilon^{-2/\eta}}^{\infty} \frac{(\nu^{-\frac{\eta}{2}})^{n-i}}{(1 + \nu^{-\frac{\eta}{2}})^{n-i+n_t n_r}} dv \right) x_i \\
&= \pi \lambda_s D^2 \sum_{i=0}^{n-1} \frac{n_t n_r C_{n-1-i+n_t n_r}^{n_t n_r}}{n} k_{n-i} x_i
\end{aligned} \tag{20a}$$

with

$$k_i = \varepsilon^{2/\eta} \int_{\varepsilon^{-2/\eta}}^{\infty} \frac{1}{(1 + v^{\eta/2})^i (1 + v^{-\eta/2})^{n_t n_r}} dv \quad i \geq 1. \tag{20b}$$

$$\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r} = a x_0 \mathbf{y}_{\mathbf{k}n_t \mathbf{n}_r} + a \mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r} \mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r}, \tag{22a}$$

$$a = \pi \lambda_s D^2, \tag{22b}$$

$$\mathbf{G}_{\mathbf{k}n_t \mathbf{n}_r} = \begin{bmatrix} 0 & & & & \\ \frac{1}{2} n_t n_r k_1 & 0 & & & \\ \frac{n_t n_r (1+n_t n_r)}{3} k_2 & \frac{n_t n_r}{3} k_1 & 0 & & \\ \vdots & & & 0 & \\ \frac{n_t n_r C_{k n_t n_r + n_t n_r - 2}^{n_t n_r}}{k n_t n_r} k_{k n_t n_r - 1} & \frac{n_t n_r C_{k n_t n_r + n_t n_r - 3}^{n_t n_r}}{k n_t n_r} k_{k n_t n_r - 2} & \cdots & \frac{n_t n_r}{k n_t n_r} k_1 & 0 \end{bmatrix}. \tag{22c}$$

Fig. 4 illustrates the impact of the outage threshold and the path loss coefficient on the cooperative probability. When the path loss coefficient is fixed, the coverage probability decreases with the increase of the outage threshold. When the

$$\begin{aligned}
\mathbb{P}_c^k &= \mathbb{E}_D \left[x_0 + \sum_{n=1}^{kn_t n_r - 1} x_n \right] \\
&= \int_0^\infty f_D(D) \left(x_0 + \sum_{n=1}^{kn_t n_r - 1} x_n \right) dD \\
&= \int_0^\infty f_D(D) \left(\exp \left(-\pi \lambda_s D^2 \varepsilon^{2/\eta} \int_{\varepsilon^{-2/\eta}}^\infty \left(1 - \frac{1}{(1+v^{-\eta/2})^{n_t n_r}} \right) dv \right) + \text{sum}(\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r - 1}) \right) dD
\end{aligned} \tag{25}$$

$$\begin{aligned}
\mathbb{P}_c^3 &= \int_0^\infty 2(\lambda\pi)^2 D^3 e^{-\lambda\pi D^2} \left(\exp \left(-\pi \lambda_s D^2 \varepsilon^{2/\eta} \int_{\varepsilon^{-2/\eta}}^\infty \left(1 - \frac{1}{(1+v^{-\eta/2})^{n_t n_r}} \right) dv \right) + \text{sum}(\mathbf{x}_{\mathbf{k}n_t \mathbf{n}_r - 1}) \right) dD
\end{aligned} \tag{27}$$

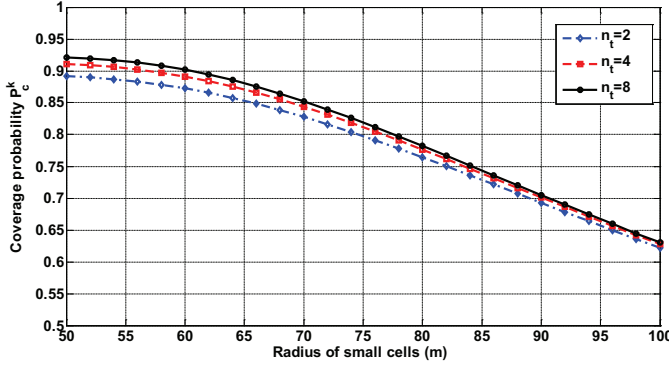


Fig. 6. Coverage probability with respect to the radius of small cells considering different transmission antennas at cooperative small cell BSs.

outage threshold is fixed, the coverage probability increases with the increase of the path loss coefficient. This result can be explained by the follows: the signal propagation attenuation is obviously increased with the increase of the propagation range when the path loss coefficient is increased. Compared with cooperative small cell BSs, the interfering small cell BSs is far away with the received vehicle. Therefore, the interference attenuation is larger than the desired signal attenuation. When the outage threshold and the transmission power of small cell BS are fixed, the coverage probability of three cooperative small cell BSs increases with the increase of the path loss coefficient.

Fig. 5 analyze the impact of the number of transmission antennas at cooperative small cell BSs on the coverage probability. When the outage threshold is less than 2.5 dB, the coverage probability increases with the increase of the number of transmission antennas at cooperative small cell BSs. When the outage threshold is larger than or equal to 2.5 dB, the coverage probability decreases with the increase of the number of transmission antennas at cooperative small cell BSs. Fig. 6 depicts the impact of the radius of small cells and the number of transmission antennas at cooperative small cell BSs on the coverage probability. When the number of transmission antennas is fixed, the coverage probability decreases with the increase of the radius of small cells. When the radius of small cell is fixed, the coverage probability increases with the increase of the number of transmission antennas at cooperative small cell BSs.

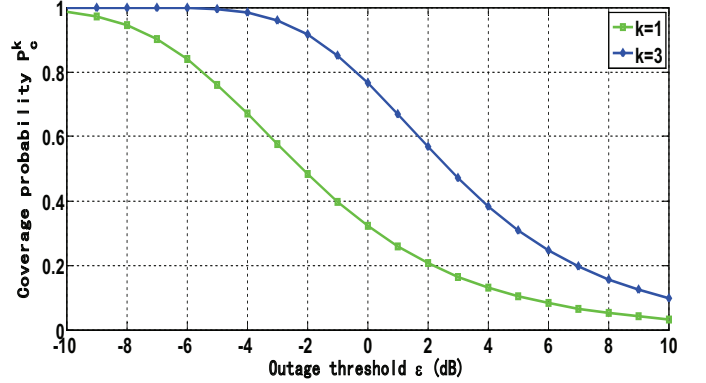


Fig. 7. Coverage probability with and without cooperative communication schemes.

Fig. 7 shows the coverage probability with and without cooperative transmission schemes in small cell networks. When the cooperative transmission scheme is adopted in small cell networks, the number of cooperative small cell BSs is configured as $k = 3$. When the cooperative communication scheme is not adopted in small cell networks, the number of cooperative small cell BSs is configured as $k = 1$. Compared with two curves in Fig. 7, the cooperative transmission scheme can improve the coverage probability in small cell networks and the gain of coverage probability with the cooperative transmission scheme achieve the maximum when the outage threshold is -1 dB.

V. VEHICLE MOBILITY ANALYSIS OF COOPERATIVE SMALL CELL NETWORKS

A. Vehicle handoff rate of cooperative small cell networks

When the initial position of the vehicle is assumed as the origin, the vehicle will arrive at a new position after a minimal slot τ . Without loss of generality, the Gauss-Markov Mobility model is adopted for the vehicle mobility in this study. Assume that the vehicle moves with the velocity of φ_{n-1} and the direction of ϑ_{n-1} at the $(n-1)^{th}$ time instant, then the velocity and direction of the n^{th} time instant are calculated by $\varphi_n = \alpha\varphi_{n-1} + (1-\alpha)\bar{\varphi} + \sqrt{(1-\alpha^2)}\varphi_{x_{n-1}}$, $\vartheta_n = \alpha\vartheta_{n-1} + (1-\alpha)\bar{\vartheta} + \sqrt{(1-\alpha^2)}\vartheta_{x_{n-1}}$ respectively, where $\bar{\varphi}$ and $\bar{\vartheta}$ represent the mean value of velocity and direction as $n \rightarrow \infty$, and $\varphi_{x_{n-1}}$ and $\vartheta_{x_{n-1}}$ are random variables

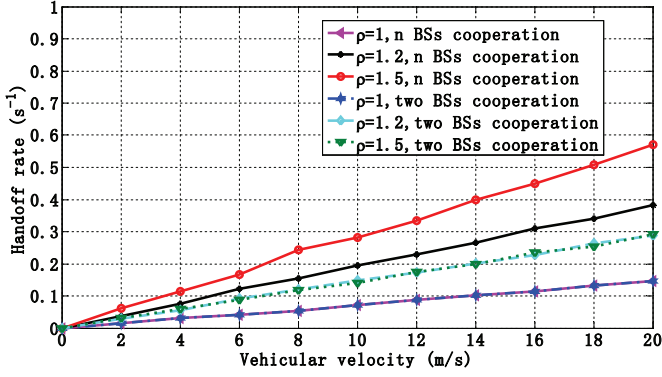


Fig. 8. Vehicular handoff rate with respect to the vehicular velocity and the cooperative threshold.

governed by Gaussian distribution. To simplify the derivation, we assume that $\alpha = 1$. In this case, the vehicle keeps the velocity φ and the direction ϑ constant. When the distance between the vehicle UE_0 and the i -th closest small cell BS BS_i is assumed as R_i in the last slot, the new distance between the vehicle UE_0 and the small cell BS BS_i in the current slot is expressed as

$$R_i^{new} = \sqrt{R_i^2 + (\varphi\tau)^2 + 2R_i\varphi\tau \cos \vartheta}. \quad (28)$$

Let $\xi = \arg \min_i R_i^{new}$, the distance between the vehicle UE_0 and the closest small cell BS BS_ξ is denoted as R_ξ^{new} in the current slot. The vehicle passes through different cooperative small cells when the vehicle moves for a slot τ . The trigger of vehicle handoff in a slot τ is expressed as (29).

where $1[\cdot]$ is the indicator function, which equals to 1 when the condition inside the bracket is satisfied and 0 otherwise. For a long time, e.g., $T = t\tau$, $t \gg 1$ and $t \in N^+$, the handoff number of the vehicle ΔH_T is the sum of (29) in the time T . Furthermore, the vehicle handoff rate in cooperative small cell networks is expressed by

$$HO = \frac{\Delta H_T}{T}. \quad (30)$$

Fig. 8 analyzes the impact of the vehicular velocity and the cooperative threshold on the vehicular handoff rate. Without loss of generality, the traditional fixed two small cells cooperative scenario is compared with the proposed multi-cell cooperative scenarios in Fig. 8. When the cooperative threshold is fixed, the vehicular handoff rate increases with the increase of the vehicular velocity in both scenarios. When $\rho = 1$, it means only one small cell transmits signals to the vehicle. In this case, the curve of multi-cell cooperative scenario is coincided with the curve of fixed two small cells cooperative scenario. when $\rho \neq 1$, the number of cooperative small cells is changed accounting for the distance between the vehicle and the small cell BSs for the proposed multi-cell cooperative scenarios. Therefore, the vehicular handoff rate of the proposed multi-cell cooperative scenarios is obviously larger than the vehicular handoff rate of traditional fixed two small cells cooperative scenario.

B. Vehicle overhead ratio of cooperative small cell networks

When the vehicle passes through different small cells, the cooperative status has to be changed. Based on small cell network architectures [39], the cooperative link of small cells is defined as the X2 link, which has been used for transmitting handoff and cooperative information in cooperative small cell networks. The X2 link of cooperative small cell networks is composed of two parts, i.e., the X2-U link and the X2-C link. In general, the X2-U link is used for transmitting handoff information and the X2-C link is used for transmitting cooperative control information among small cells. Therefore, the overhead traffic of X2-C links is expressed as

$$T_{X2-C} = \delta \cdot HO, \quad (31)$$

where δ is the average cooperative control data size per small cell when a handoff is triggered in small cell networks [40].

when the total L types traffic is assumed for vehicular communications, let ψ_l and ζ_l^{-1} , $1 \leq l \leq L$, are the traffic arrive rate and the average vehicle session duration of the $type-l$ vehicle traffic. Based on the queue theory, the active probability of the $type-l$ vehicle traffic is expressed as

$$p_A(l) = \psi_l / (\psi_l + \zeta_l). \quad (32)$$

Furthermore, the handoff rate of the $type-l$ vehicle traffic is expressed as

$$HO_l = p_A(l) \cdot HO. \quad (33)$$

Let a_l is the flow rate of $type-l$ vehicle traffic and the average handoff duration is χ , the overhead of the $type-l$ vehicle traffic generated by a handoff over the X2-U link is expressed as

$$\beta_l = a_l \cdot \chi. \quad (34)$$

When all handoff requests are assumed to be accepted in small cell networks, the overhead traffic of X2-U links is expressed as

$$T_{X2-U} = \sum_{l=1}^L \beta_l \cdot HO_l. \quad (35)$$

Based on (31) and (35), the total overhead traffic of X2 links in cooperative small cell networks can be expressed by

$$T_{X2} = T_{X2-C} + T_{X2-U}. \quad (36)$$

Without loss of generality, a change of the cooperative status is triggered by a vehicular handoff in cooperative small cell networks. In general, the change of vehicular cooperative status conduces to some overhead traffic in cooperative small cell networks. The size of overhead traffic is expressed as

$$C = k \cdot T_{X2}, \quad (37)$$

Therefore, the expectation of the size of overhead traffic is expressed as

$$\mathbb{E}[C] = \sum_k \mathbb{P}_k \cdot k \cdot T_{X2}. \quad (38)$$

$$\Delta H_\tau = 1 \left[\exists i \in N^+, \left(\frac{R_i^{new}}{R_\xi^{new}} > \rho \cap \frac{R_i}{R_1} \leq \rho \right) \cup \left(\frac{R_i^{new}}{R_\xi^{new}} \leq \rho \cap \frac{R_i}{R_1} > \rho \right) \right], \quad (29)$$

When the SINR of vehicular communication is configured as the outage threshold ε , the vehicular communication capacity is expressed as

$$\partial = \left[\sum_{k=1}^{\infty} \mathbb{P}_c^k \cdot P_k \right] \cdot B_w \cdot \log(1 + \varepsilon), \quad (39)$$

where B_w is the bandwidth for vehicle wireless links.

To evaluate the cooperative communication overhead for vehicular communications, the vehicular overhead ratio in cooperative small cell networks is defined as

$$\Omega = \frac{\mathbb{E}[C]}{\partial}. \quad (40)$$

Considering the cooperative transmission in small cell networks, the gain, i.e., the vehicular communication capacity, and the cost, i.e., the vehicular overhead ratio can be evaluated by (39) and (40).

C. Performance analysis

Based on the proposed vehicular communication capacity and vehicular overhead ratio, some performance evaluations are numerically analyzed in Fig. 9–Fig. 11. In the following analysis, the default parameters are configured as follows: the outage threshold is $\varepsilon = 0$ dB, the bandwidth is $B_w = 10$ Mbps, the average cooperative control data size is $\delta = 480$ bits, the path loss coefficient is $\eta = 4$, the time slot is $\tau = 15$ millisecond (ms), the radius of small cell is 50 m, the vehicular velocity is $\varphi = 10$ m/s, the handoff duration is $\chi = 0.05$ s. Without loss of generation, two types of traffic are considered in this paper. The *type-1* traffic has the following configuration parameters: $a_1 = 12.2$ Kbps, $\psi_1 = 1.5$ and $\zeta_1^{-1} = 0.03333$. The *type-2* traffic has the following configuration parameters: $a_2 = 353.8$ Kbps, $\psi_2 = 0.5$ and $\zeta_2^{-1} = 0.05$ [40].

Fig. 9 evaluates the vehicular communication capacity with respect to the radius of small cells considering different cooperative thresholds. When the cooperative threshold is fixed, the vehicular communication capacity decreases with the increase of the radius of small cells. When the radius of small cells is fixed, the vehicular communication capacity increases with the increase of the cooperative threshold. Fig. 10 illustrates the vehicular overhead ratio with respect to the vehicular velocity considering different cooperative thresholds. When the cooperative threshold is fixed, the vehicular overhead ratio increases with the increase of the vehicular velocity. When the vehicular velocity is fixed, the vehicular overhead ratio increases with the increase of the cooperative threshold. The capacity gain is increased when the cooperative threshold is less than 1.5. When the cooperative threshold is larger than or equal to 1.5, the cooperation probability of k small cells decreases with the increase of the cooperative threshold. This result is validated in Fig. 3. It implies that the number of

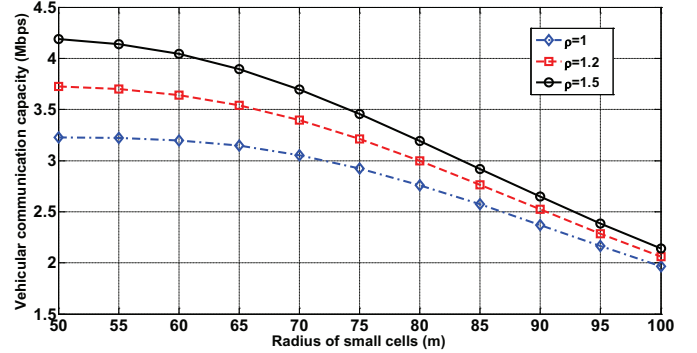


Fig. 9. Vehicular communication capacity with respect to the radius of small cells considering different cooperative thresholds.

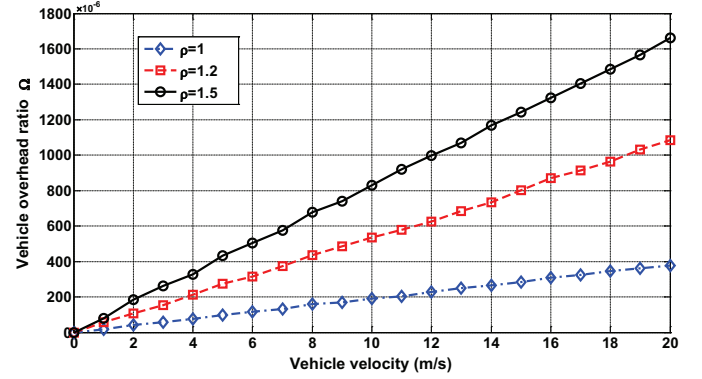


Fig. 10. Vehicular overhead ratio with respect to the vehicular velocity considering different cooperative thresholds.

cooperative small cells will be limited with the increase of the cooperative threshold in practical applications. Therefore, the capacity gain is limited with increase of the cooperative threshold. Fig. 11 analyzes the vehicular overhead ratio with respect to the radius of small cells considering different cooperative thresholds. When the radius of small cells is fixed, the vehicular overhead ratio increases with the increase of the cooperative threshold. When the cooperative threshold is fixed, the vehicular overhead ratio first decreases with the increase of the radius of small cells. Numerical results show that there exist turning points, i.e., the radius of small cells is 75, 80 and 85 m, corresponding to the cooperative threshold 1.5, 1.2 and 1. When the radius of small cells is larger than the turning points, the vehicular overhead ratio increases with the increase of the radius of small cells. Therefore, there exist a minimal value for the vehicular overhead ratio under different cooperative thresholds. The minimal vehicle overhead ratio is 4.5×10^{-4} , 3.4×10^{-4} and 1.3×10^{-4} , corresponding to the cooperative threshold of 1, 1.2 and 1.5. This result provide a guideline for optimizing the vehicular overhead ratio in 5G

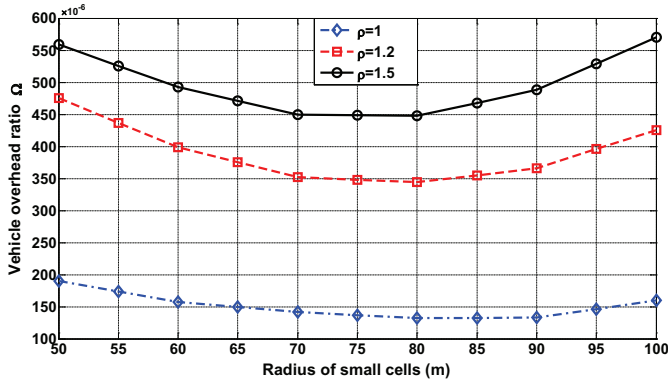


Fig. 11. Vehicular overhead ratio with respect to the radius of small cells considering different cooperative threshold.

cooperative small cell networks with different radii of small cells.

VI. CONCLUSIONS

In this paper, the vehicular mobility performance is analyzed for 5G cooperative MIMO small cell networks considering co-channel interference. Based on distances between the vehicle and cooperative small cell BSs, the cooperative probability and the coverage probability have been derived for 5G cooperative small cell networks where small cell BSs follow Poisson point process distributions. Furthermore, the vehicular handoff rate and the vehicular overhead ratio have been proposed to evaluate the vehicular mobility performance in 5G cooperative MIMO small cell networks. Numerical results indicate that the cooperative transmission scheme increases the vehicular communication capacity and the vehicular handoff rate in 5G cooperative MIMO small cell networks. Therefore, there exist a tradeoff between the vehicular communication capacity and the vehicular handoff ratio. By evaluating the vehicular overhead ratio, numerical results show that there exists a minimum vehicular overhead ratio for 5G cooperative MIMO small cell networks considering different cooperative thresholds. These results provide a guideline for optimizing vehicular communications in 5G cooperative MIMO small cell networks.

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