Outage Analysis for Intelligent Reflecting Surface Assisted Vehicular Communication Networks

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Abstract—Vehicular communication is an important application of the fifth generation of mobile communication systems (5G). Due to its low cost and energy efficiency, intelligent reflecting surface (IRS) has been envisioned as a promising technique that can enhance the coverage performance significantly by passive beamforming. In this paper, we analyze the outage probability performance in IRS-assisted vehicular communication networks. We derive the expression of outage probability by utilizing series expansion and central limit theorem. Numerical results show that the IRS can significantly reduce the outage probability for vehicles in its vicinity. The outage probability is closely related to the vehicle density and the number of IRS elements, and better performance is achieved with more reflecting elements.

Index Terms—Vehicular communication networks, intelligent reflecting surface, outage probability, series expansion, central limit theorem.

I. INTRODUCTION

Vehicular communication networks have been extensively investigated to realize the concept of intelligent transportation systems (ITS) [1]. As an important application of the fifth generation of mobile communication systems (5G), vehicular communication networks require high data-rate, low-latency transmission and high reliability. Traditional vehicular communication has been supported by the dedicated short-range communication (DSRC) standard. However, DSRC may not effectively meet the requirements of high data rate due to its limited coverage and capacity [2].

Therefore, to improve the data rate, the millimeter wave (mmWave) communications have been introduced into vehicular communication networks. With larger available bandwidth, mmWave is able to support the high data rate and low latency communications for the emerging vehicular applications [1]. Nevertheless, due to its various problems such as high path loss and vulnerability to blockage [3], the beam-tracking and transmission outage remain a challenge for mmWave communications [1]. Therefore, high outage probability and low coverage rate are still unsolved problems for mmWave vehicular communications.

Among the latest technologies for green and effective wireless communication, the intelligent reflecting surface (IRS) has emerged as an innovative and cost-effective paradigm for improving the transmission coverage and signal quality via passive reflecting arrays [4]. Owing to the tunable phase shifts of all reflecting elements, signal enhancement and interference suppression can be achieved by the IRS without the use of active transmitters [4]. The authors of [5] provide a detailed overview and historical perspective on state-of-the-art solutions of IRS-aided communication systems. Various aspects of IRS have recently been investigated in terms of channel estimation [6], [7], data rate maximization [8], [9], physical layer security [10], etc.

To the best of the authors’ knowledge, there are a paucity of contributions studying the integration of IRS in vehicular network systems. In [11], the physical layer security of vehicular networks employing IRS was studied and two IRS-based vehicular network system models were designed. However, the work only focused on the secrecy capacity and the outage performance of IRS-assisted vehicular networks has not been studied. Given the blockage and vehicle density in practical road conditions, the difficulty of analysis would increase, which were not considered in [11].

In this paper, we propose to use IRS in vehicular networks to improve the coverage and outage performance. We establish a model for an IRS-assisted downlink vehicular communication system with Poisson Point Process (PPP) distribution and derive the expression of outage probability based on series expansion and central limit theorem. The outage performance of traditional vehicular networks and IRS-assisted vehicular networks are also compared.

The main contributions of this work are listed as follows.

• We establish an IRS-assisted downlink vehicular communication model by considering a more practical scenario with vehicle blockings and density of vehicles.
• Accurate approximations of the outage probability are derived based on series expansion and central limit theorem. Lower vehicle density and larger size of IRS will improve the outage probability performance.
• Simulation analysis is conducted and it is shown that the outage probability can be reduced significantly by using large IRS. Besides, IRS can effectively improve the coverage performance of vehicular communication networks.

The rest of this paper is summarized as follows. In Section II, the IRS-assisted downlink channel model is given and the problem for outage analysis is formulated. In Section III, the expression for outage probability is derived based on series expansion and central limit theorem. The numerical results are
Poisson Point Process (PPP) is very useful for modeling the distribution of vehicles on a lane [12]. We use two PPPs, i.e., $\Phi_0$ with density $\lambda_0$ and $\Phi_1$ with density $\lambda_1$, to model the vehicle distribution in the obstacle lane and the user lane, respectively. The density value is related to the average speed and density of vehicles. For a PPP with density $\lambda$, let us denote $N([a, b])$ as the number of points (events) within the interval of $[a, b)$, the distribution of which is derived as

$$P[N([a, b)) = k] = e^{-\lambda(b-a)} \frac{(\lambda(b-a))^k}{k!}. \quad (1)$$

The probabilities of the RSU-IRS channel being unblocked (LOS) and blocked (NLOS) are denoted by $p_{IR, L}$ and $p_{IR, N}$, respectively, while for the RSU-user channel, the probabilities are denoted by $p_{UR, L}$ and $p_{UR, N} = 1 - p_{UR, L}$, respectively. By some mathematical manipulations, we obtain

$$p_{IR, L} = e^{-\lambda_0 \sigma^2} e^{-\lambda_1 \tau}, \quad p_{IR, N} = 1 - e^{-\lambda_0 \sigma} e^{-\lambda_1 \tau}; \quad (2a)$$
$$p_{UR, L} = e^{-\lambda_0 \sigma}, \quad p_{UR, N} = 1 - e^{-\lambda_0 \sigma}. \quad (2b)$$

NLOS and LOS scenarios result in different path loss and small-scale fading. The path loss of a specific link at a distance of $r$ is modeled as

$$l(r) = \delta_L C_L r^{-\alpha_L} + (1 - \delta_L) C_{NL} r^{-\alpha_N}, \quad (3)$$

where $\alpha_L$ and $\alpha_N$ are the path loss exponents, while $C_L$ and $C_N$ are the path loss intercept factors of the LOS and NLOS cases, respectively. In (3), $\delta_L$ is equal to one for LOS and zero for NLOS.

Regarding small-scale fading, Rayleigh and Rician fading channels are used for modeling NLOS and LOS cases, respectively. The probability density function (PDF) of Rayleigh and Rician distribution are given, respectively, by

$$p_{\text{Ray}}(x) = \begin{cases} \frac{2}{\sigma^2} x e^{-\frac{x^2}{2\sigma^2}}, & x > 0 \\ 0, & \text{otherwise} \end{cases}, \quad (4a)$$
$$p_{\text{Ric}}(x) = \begin{cases} p(x) = \frac{1}{\sigma_m} \exp\left(-\frac{x^2 + A^2}{2\sigma_m^2}\right) I_0 \left(\frac{x A}{\sigma_m}\right), & x > 0 \\ 0, & \text{otherwise} \end{cases}, \quad (4b)$$

where $\sigma_0^2$ denotes the variance of the components in a Rayleigh variable; $A$ denotes the amplitude of the main signal component of a Rician variable; $\sigma_m^2$ denotes the power of the multi-path signal component of a Rician variable; $I_0(\cdot)$ is the modified zero-order Bessel function.

### B. Problem Formulation

Assume that the transmitted symbol $x$ is expressed as

$$x = \sqrt{P_{tx}} s, \quad (5)$$

where $P_{tx}$ is the transmission power, $s$ is the information-carrying symbol which follows the distribution of $CN(0, 1)$. Hence, the received signal at the target user can be written as

$$y = \left[H \sqrt{l(r_{UR})} + h_{UL}^T \Theta u_{IR} \sqrt{\sqrt{l(r_{UL})}}(r_{UL})\right] x + n, \quad (6)$$

where $\Theta = \text{diag}(e^{-j\theta_1}, e^{-j\theta_2}, \ldots, e^{-j\theta_M})$ contains the phase shifting coefficients of the IRS; $n$ denotes the received white noise.
Gaussian noise which follows the distribution of \(CN(0, \sigma_n^2)\); \(r_{UR}, r_{IR}\) and \(r_{UI}\) denote the distance of RSU-user, RSU-IRS and IRS-User links, respectively.

According to (6), the SNR of the received signal is given by
\[
\text{SNR} = \frac{P_0}{\sigma_n^2} |H(\theta) + h_{UI} \Theta h_{IR} \sqrt{L(r_{IR})/L(r_{UI})}|^2.
\]

Note that the SNR will be different when \(l(r)\) and \(l(r_{UI})\) change under different road conditions. For simplicity, we denote \(L_D = \sqrt{l(r)}\) and \(L_R = \sqrt{l(r_{IR})/l(r_{UI})}\).

The channel coefficients can be expressed in detail as
\[
H = |H(\theta)| e^{j\varphi},
\]
\[
h_{UI} = [|h_{UI,1}| e^{j\varphi_{UI,1}}, |h_{UI,2}| e^{j\varphi_{UI,2}}, \ldots, |h_{UL,M}| e^{j\varphi_{UI,M}}]^T,
\]
\[
h_{IR} = [|h_{IR,1}| e^{j\varphi_{IR,1}}, |h_{IR,2}| e^{j\varphi_{IR,2}}, \ldots, |h_{IR,M}| e^{j\varphi_{IR,M}}]^T.
\]

According to [5], the SNR is maximized when the phase shift of the direct and the reflection paths are the same, which means the optimal phase shift is given by \(\theta_i = \varphi_{UI,i} + \varphi_{IR,i} - \varphi\). Therefore, (7) becomes
\[
\text{SNR} = \frac{P_0}{\sigma_n^2} |L_D|H| + L_R \times \sum_{i=1}^{M} |h_{UI,i}|^2 h_{IR,i}|^2.
\]

The outage probability of a given threshold \(t\) is given by
\[
P_O(t) = \Pr(\text{SNR} < t).
\]

In (10), the SNR is a function of the channel coefficients between the RSU and the user. Due to the vehicle distribution and the complicated channel distribution which involves both Rayleigh and Rician distributions, a closed-form expression for (10) is very difficult to obtain.

### III. OUTAGE PROBABILITY ANALYSIS

To simplify the analysis of outage probability, let us first examine the summation term \(G = \sum_{i=1}^{M} |h_{UI,i}|^2 h_{IR,i}|^2\) in (9). For a specific \(i\), denote the mean and variance of \(Z_i = |h_{UI,i}|^2 h_{IR,i}|^2\) as \(\xi_i\) and \(\nu_i\), respectively. According to CLT, since \(Z_i\)'s are independently identically distributed (i.i.d.), we have \(G \sim \mathcal{N}(M\xi_0, M\nu_0)\) when \(M\) is large. Substituting (9) into (10) yields
\[
P_O(t) = \Pr(\text{SNR} < t) = \Pr\left(\frac{P_0}{\sigma_n^2} |L_D|H| + L_R G < Z\right),
\]
where \(Z = \sigma_n \sqrt{\frac{P_0}{\sigma_n^2}}\) is the modified threshold. In the following derivation, we use \(P_O(Z)\) to represent the outage probability.

The analysis focuses on deriving the distribution of the random variable \(Z' = L_D|H| + L_R G\) in (11). Then the outage probability can be obtained by integration. Since a closed-form of (11) is very difficult to derive, we provide accurate approximations based on series expansion based approximation (SEA) and central limit theorem based approximation (CLA).

### A. Analysis based on SEA

The distribution of \(Z'\) varies in different channel conditions. Both the RSU-user link and RSU-IRS link could be LOS or NLOS channels. We consider the following four situations.

1) **NLOS for both RSU-user and RSU-IRS links:** Since the RSU-user and RSU-IRS links are both NLOS channels, both \(|H|\) and \(|h_{IR,i}|\) follow Rayleigh distribution in (4a). Note that \(|h_{UI,i}|\) is always a Rician distributed variable, which follows the distribution in (4b). In this case, denote the mean and variance of \(G\) as \(\xi_N\) and \(\nu_N\), which are given by
\[
\xi_N = \frac{P_0}{\sigma_n^2} |L_D|H| + L_R \sum_{i=1}^{M} |h_{UI,i}|^2 h_{IR,i}|^2.
\]
\[
\nu_N = \frac{1}{2} M \sigma_n \sigma_0 \exp\left(-\frac{\alpha(x)}{2\sigma_0^2}\right) \times \left\{\frac{1}{2} \alpha(x) \sqrt{\frac{\pi}{\beta_N}} \left[1 + \operatorname{erf}\left(\sqrt{\beta_N}(\alpha(x))^2\right]\right] + \frac{1}{2\beta_N} e^{-\beta_N(\alpha(x))^2}\right\},
\]
where \(\alpha(x) = \frac{L_D\sigma_0^2}{M\nu_N L_D \sigma_0^2 + \frac{2}{M^2} \nu_N L_D \sigma_0^2}\) and \(\beta_N = \frac{M\nu_N L_D^2 + \frac{2}{M^2} \nu_N L_D \sigma_0^2}{2M\nu_N L_D \sigma_0^2}\).

Note that \(P_O(Z)\) is obtained by integration of (13) over \(Z\). However, the integral is very difficult since (13) contains error function \(\operatorname{erf}(.1)\) to derive in closed form. In this work, we provide series expansion based approximation to solve the integral. In (13), \(\operatorname{erf}(x)\) can be expressed in the following series as
\[
\operatorname{erf}(x) \approx \frac{2}{\sqrt{\pi}} \sum_{i=0}^{n} \frac{(-1)^i}{(2i+1) \cdot i!} x^{2i+1}
\]
where \(n\) is the series order.

Applying (14) to (13), we obtain the expression of \(P_O(Z)\).

**Proposition 1:** When the RSU-user link and the RSU-IRS link are both NLOS, the outage probability can be expressed as
\[
P_{O,NN}(Z) = C(\xi_N) \times \left\{\frac{\sqrt{\pi}}{2\beta_N \nu_N} \operatorname{erf}\left(D_N \left(Z - B_N\right)\right) + \operatorname{erf}\left(D_N B_N\right)\right\} - \frac{1}{2} \sqrt{\frac{\pi}{\beta_N}} L_D \sigma_0^2 \left(e^{-\beta_N(x)^2} - 2F_0(0) - F_0(B_N) - F_0(\alpha_N(x))\right),
\]
where \(B_N = \frac{L_D}{}\) and \(\alpha_N = (Z - B_N)^2 / (2\beta_N)\) with \(\alpha_N = \frac{M\nu_N L_D + \frac{2}{M^2} \nu_N L_D \sigma_0^2}{2M\nu_N L_D \sigma_0^2}\) and \(C(x) = \left(\frac{2\pi M\nu_N L_D \sigma_0^2}{x^2}\right)^{-1}\) and \(F_0(x) = \sum_{i=0}^{n} \frac{(-1)^i}{(2i+1) \cdot i!} \left(M\nu_N L_D^2 + \frac{2}{M^2} \nu_N L_D \sigma_0^2\right)^{i+0.5} \Gamma(i + 1, x)\) with \(\Gamma(\alpha, x) = \int_x^{\infty} e^{-t} t^{\alpha-1} dt\).

When \(n\) goes to infinity, (16) is an accurate expression. However, finite items can guarantee the accuracy of approximation in practical use.
2) LOS for both RSU-user and RSU-IRS links: In this case, $|h_{\text{L,1}}||h_{\text{R,1}}|$ is the product of two Rician variables and we denote the mean and variance of $G$ as $\varepsilon_L$ and $\nu_L$, respectively, which can be expressed as

$$\varepsilon_L = \frac{\pi}{2} \sigma^2_m e^{-K \left[ (1+K) L_0 \left( \frac{K}{2} \right) + K L_1 \left( \frac{K}{2} \right) \right]^2}, \quad (18a)$$

$$\nu_L = M \left( 2\sigma^2_m + A^2 \right) - \varepsilon_L^2 / M. \quad (18b)$$

In this case $|H|$ is a Rician variable and it is very difficult to derive the PDF of $Z'$. Therefore, we convert the problem into a double integration based on the PDF of $|H|$ and $G$.

In the derivation, the Taylor expansion of Bessel function $J_n$ is given by (15) at the bottom of this page.

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have been provided in (12) and (18), which are expressed as \( \mathcal{E}_N, \mathcal{V}_N \) and \( \mathcal{E}_L, \mathcal{V}_L \). For \( |I_l| \), the mean and variance when the RSU-user link is LOS or NLOS are given, respectively, by \( \mathcal{E}_{I_l,LL} = \sqrt{\pi/2} \sigma_{\eta}\alpha^{K/2} [(1 + K) I_O (K/2) + K I_1 (K/2)], \) \( \mathcal{V}_{I_l,LL} = \sigma_{\eta}^2 + \alpha^2 \) and \( \mathcal{E}_{I_l,NN} = \sqrt{\pi/2} \sigma_0, \mathcal{V}_{I_l,NN} = (2\pi/2) \sigma_0^2 \). By integration over \( Z' \) and assuming that it is Gaussian distributed, the approximated outage probability is obtained in the following theorem.

**Theorem 2:** Based on CLT, a simplified approximated expression of outage probability is given by

\[
P_O(Z) = \frac{1}{2} \text{erf} \left( \sqrt{\frac{1}{2\sigma_0^2}} (Z - \epsilon) \right) + \frac{1}{2} \text{erf} \left( \sqrt{\frac{1}{2\sigma_0^2}} \epsilon \right).
\]

(25)

where \( Z \) is the modified threshold in (11).

**C. Discussion**

To gain more insights, we analyze the impact of the vehicle density and the number of IRS elements on the outage probability.

**Corollary 1:** The outage probability is a monotonically increasing function of \( \lambda_0 \) or \( \lambda_u \). With higher vehicle density, the probability of channel blockages increases, leading to the increase of outage probability.

**Proof:** For any fixed \( Z \), use \( P_O, P_{O,NN}, P_{O,LL}, P_{O,NN} \) and \( P_{O,LL} \) to represent \( P_O(Z), P_{O,NN}(Z), P_{O,LL}(Z), P_{O,NN}(Z) \) and \( P_{O,LL}(Z) \) for simplicity. According to (23), taking the first-order derivation of \( P_O \) with respect to \( \lambda_0 \), we obtain

\[
\frac{dP_O}{d\lambda_0} = (P_{O,NN} - P_{O,LL}) \left( e^{-(\lambda_0 + \lambda_u)\tau} - e^{-(2\lambda_0 + \lambda_u)\tau} \right) \nu + (P_{O,NN} - P_{O,LL}) \left( e^{-(\lambda_0 - \lambda_u)\tau} - e^{-(2\lambda_0 + \lambda_u)\tau} \right) \nu + (P_{O,LL} - 2P_{O,LL}) e^{-(2\lambda_0 + \lambda_u)\tau}. \]

(26)

Deriving the expressions of \( P_{O,NN} - P_{O,LL}, P_{O,NN} - P_{O,NN} \) and \( P_{O,LL} - 2P_{O,LL} \), and taking the first-order derivation of these expressions with respect to \( Z \), we can prove that \( P_{O,NN} > P_{O,LL} > P_{O,NN} > P_{O,NN} \) and \( P_{O,LL} > P_{O,NN} > P_{O,LL} \). Then we have \( \frac{dP_O}{d\lambda_0} > 0 \). Similarly, we have \( \frac{dP_O}{d\lambda_u} > 0 \). Thus, the proof is completed.

**Corollary 2:** According to (25), when \( M = 0 \), the IRS is out of service and the outage probability reaches a peak value. However, when \( M \to \infty \), the outage probability will approach 0.

**Proof:** With the increase of \( M \), the values of \( \epsilon \) and \( \nu \) will decrease. In (24), when the value of \( M \to \infty \), we have \( \epsilon \to \infty \) and \( \nu \to \infty \). Since \( \epsilon^2 \) and \( \nu \) are both second-order infinity of \( M \), we have \( \sqrt{2\sigma_0^2} \epsilon \to C_0 \) and \( \sqrt{2\sigma_0^2} \nu \to 0 \), where \( C_0 \) is a constant value. Therefore, \( P_O(Z) \) will approach 0.

Note that in reality, it is impossible for \( M \) to be infinity and the outage probability would never become 0. In general, larger value of \( M \) will reduce the outage probability.

**IV. NUMERICAL RESULTS**

In this section, we carry out Monte-Carlo simulations to evaluate the performance of IRS-assisted vehicular communication systems and the accuracy of analysis in previous sections. The simulation parameters are summarized in Table I.

As shown in Fig. 3, the width of each lane is 4m and the vehicle length is 5m. The coordinates of RSU and IRS are \((10,0)\) and \((22,8)\), respectively. The transmit power of RSU is 27dBm. Parameters related to path loss are given in Table I.

In Fig. 4, the outage probabilities calculated using SEA and CLA are compared, where the number of IRS elements is fixed at 100 and \( \lambda_0 \) and \( \lambda_u \) are both 1/10, which means the average distance between two adjacent vehicles is 5m. This implies a very crowded scenario, where outage would be severe. As can be seen in Fig. 4, the outage probabilities differ greatly at different locations of the target user. For a given threshold \( t \), the SEA and CLA proposed in this paper are very accurate, the results of which almost coincide with simulation results. SEA with \( n = 10 \) is the most accurate, but the performance of \( n = 3 \) is fine as well. CLA has excellent performance and its expression is concise, which makes it very helpful for analysis.

**TABLE I**

| Parameter         | Value | Parameter | Value |
|-------------------|-------|-----------|-------|
| Lane width \( \tau \) | 4 m   | \( \sigma_N \) | 4 [14] |
| Vehicle length \( \tau \) | 5 m   | \( \sigma_N \) | 4 [14] |
| Carrier frequency \( f \) | 28 GHz | \( \sigma_N^2 \) | -39dBm [12] |
| \( C_L, C_N \) | \(-20\log(4\pi f/c)\) | \( P_{tx} \) | 27dBm [12] |

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\[1\] It is the free space path loss in dB at a distance of 1m and \( c \) is the speed of light [15].
The vicinity of the IRS decrease significantly. When two lanes is identical. The results with roadside to improve the coverage performance of all regions. it is feasible to deploy multiple collaborative IRSs along the where IRS could work effectively is quite small. Therefore, increases for 3 times when IRS. The reason is that large IRS can greatly small coverage range and suffers from severe blockage, IRS signals. Although mmWave vehicular communication has a large IRS can greatly improve the coverage performance of the RSU by reflecting signals. Although mmWave vehicular communication has a small coverage range and suffers from severe blockage, IRS could help to increase the outage performance significantly. Compared with $M = 0$ (no IRS used), the range for $P_O < 0.3$ increases for 3 times when $M = 500$. Note that the range where IRS could work effectively is quite small. Therefore, it is feasible to deploy multiple collaborative IRSs along the roadside to improve the coverage performance of all regions.

To evaluate the influence of vehicle distance $d$ on the outage probability, we assume that the average vehicle distance of the two lanes is identical. The results with $d = 5\text{m}$ ($\tau$), $d = 25\text{m}$ ($5\tau$) and $d = 50\text{m}$ ($10\tau$) are depicted in Fig. 6. As the user location changes, the outage probability reaches its minimum at $x = 22\text{m}$, where the IRS locates. It can be concluded that the outage probability decreases with respect to $d$ due to the reduced probability of blockage.

V. CONCLUSION

In this work, we studied the outage probability for IRS-assisted vehicular communication systems. Accurate approx-imate expressions of outage probability were provided by means of SEA and CLA. Numerical results validate the analysis. It is shown that the outage probability can be reduced significantly by using large IRS even for very crowded road.

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