



Heriot-Watt University
Research Gateway

Energy Efficiency Optimization for UAV-assisted Backscatter Communications

Citation for published version:

Yang, S, Deng, Y, Tang, X, Ding, Y & Zhou, J 2019, 'Energy Efficiency Optimization for UAV-assisted Backscatter Communications', *IEEE Communications Letters*, vol. 23, no. 11, pp. 2041-2045.
<https://doi.org/10.1109/LCOMM.2019.2931900>

Digital Object Identifier (DOI):

[10.1109/LCOMM.2019.2931900](https://doi.org/10.1109/LCOMM.2019.2931900)

Link:

[Link to publication record in Heriot-Watt Research Portal](#)

Document Version:

Peer reviewed version

Published In:

IEEE Communications Letters

Publisher Rights Statement:

© 2019 IEEE. Personal use is permitted, but republication/redistribution requires IEEE permission. See http://www.ieee.org/publications_standards/publications/rights/index.html for more information.

General rights

Copyright for the publications made accessible via Heriot-Watt Research Portal is retained by the author(s) and / or other copyright owners and it is a condition of accessing these publications that users recognise and abide by the legal requirements associated with these rights.

Take down policy

Heriot-Watt University has made every reasonable effort to ensure that the content in Heriot-Watt Research Portal complies with UK legislation. If you believe that the public display of this file breaches copyright please contact open.access@hw.ac.uk providing details, and we will remove access to the work immediately and investigate your claim.

Energy Efficiency Optimization for UAV-assisted Backscatter Communications

Shengzhi Yang, Yansha Deng*, Xuanxuan Tang, Yuan Ding, and Jianming Zhou

Abstract—Future Internet-of-Things (IoT) has high demand for energy-saving communications, especially in remote areas and smart cities. To meet this demand, we propose novel Unmanned Aerial Vehicle-assisted backscatter communications, where a UAV first collects data from multiple terrestrial backscattering tags via time division multiple access, and then flies into the coverage region of a terrestrial base station to upload its collected data to its associated base station. To determine the optimal UAV data collection location, we first analyze the system average outage probability, and then optimize the energy efficiency with the optimal backscattering location through Golden Section method under UAV energy constraint. Our analytical and simulation results illustrate that there is a trade-off between UAV data collection location and the outage probability, and the optimal UAV data collection location to achieve maximum energy efficiency needs to be closer to the tags for lower UAV transmit power.

Index Terms—UAV, backscatter communications, energy efficiency, optimal UAV data collection location.

I. INTRODUCTION

Backscatter communication is a promising technology for future IoT networks to link a huge number of smart devices in various applications, including industrial automation, precision agriculture, and smart cities [1, 2]. In ambient backscatter systems [3], ambient radio frequency (RF) energy, such as TV, WiFi and cellular signals, is harvested as the only power source for tag operations. Since the available ambient energy is limited, its communication range is commonly in the range of a few meters, hindering its extensive field applications [4]. In order to extend its communication range, bistatic architecture with dedicated RF power sources [4-6] is proposed, where a nearby signal generator is exploited to create a RF carrier that, after being modulated by the tags, is able to convey information to readers located hundreds or even thousands of meters away. Specifically, in [4], their experimental results illustrated when the backscattering device is close to an RF source, the system transmission distance can reach 2.8km.

Thanks to the high maneuverability, ease of deployment, hovering ability and low cost, Unmanned Aerial Vehicles (UAVs) are more attractive to provide wireless connectivity [7], especially for applications in remote areas and restricted regions, such as intelligent agriculture in large farms and fauna and flora protection in national parks [8]. In these scenarios, UAVs can act as information collectors and uploaders from IoT devices to the nearest base station (BS), while powering the

IoT nodes simultaneously. A UAV-based prototype wireless power transfer (WPT)-BS in [9] demonstrated the ability of UAVs both as the WPT and communication platforms. Besides, the UAV relay location [10] and the UAV power consumption [11, 12] for energy efficiency communications can be optimized.

Motivated by above, the mobility of the UAV can be optimized compared with backscattering communications at fixed locations to investigate the optimal system performance. We propose a UAV-assisted backscatter communications, which combines the advantages of both UAVs and backscattering communications. A UAV acts as a data collector from multiple terrestrial backscattering tags via time division multiple access (TDMA). Then the UAV deposits the collected data into a far-away BS after a period of flight. We first analyze the outage probability of this UAV-assisted backscatter communication, and then optimize the UAV data collection location for maximum energy efficiency under UAV energy constraint. Our results show that lower UAV transmit power leads to closer UAV data collection location to the tags so as to decrease the outage probability and improve the energy efficiency.

The rest of the paper is organized as below. Section II describes the system model, and then provides analysis of the system average outage probability. The UAV data collection location is optimized to achieve maximum energy efficiency in Section III, and validated via simulation in Section IV. Conclusions are drawn in Section V.

Notation: A modified incomplete gamma function is denoted by $\Gamma(m, x) = \frac{1}{\Gamma(m)} \int_0^x t^{m-1} \cdot e^{-t} dt$. $\Gamma(x)$ denotes the Gamma function. $f_A(x)$ and $F_A(x)$ denotes a probability density function (PDF) of A, and the corresponding cumulative distribution function (CDF), respectively.

II. SYSTEM MODEL AND OUTAGE PROBABILITY

A. System Model

As illustrated in Fig. 1, we consider a UAV-assisted backscatter communication system, where a UAV with the limited total energy E_{total} collects data from M terrestrial tags. We assume that these tags U_m ($1 \leq m \leq M$) are randomly scattered with uniform distribution within a range of 20 meters. The entire operation consists of three stages: information collection via backscattering communications, UAV flight to nearby BS, and UAV uplink information to BS.

The backscatter communication period T_B is equally divided into M time slots, and the time sequence is given. Each terrestrial tag U_m at $(x_{U_m}, 0)$ is chosen one slot to deliver their collected information to the UAV via backscattering. In the other $M - 1$ time slots, the tag only harvests energy. The UAV hovers at (x_1, h) to collect data with its fixed transmit power P_V from the backscattering tags via TDMA. As shown in Fig. 1, the distance between the UAV data collection location and

Manuscript received June 26, 2019; accepted July 25, 2019. The associate editor coordinating the review of this paper and approving it for publication was D. Ciuonzo. The work of Shengzhi Yang was supported by the China Scholarship Council. (Corresponding author: Yansha Deng.)

Shengzhi Yang and Jianming Zhou are with School of Information and Electronics, Beijing Institute of Technology, Beijing 100081, China. (E-mail: shengzhi.yang@kcl.ac.uk; E-mail:zhoujm@bit.edu.cn)

Yansha Deng is with King's College London, London WC2B 4BG, UK. (E-mail: yansha.deng@kcl.ac.uk)

Xuanxuan Tang is with Air Force Early Warning Academy, Wuhan 430019, China. (E-mail: tang_xx@126.com)

Yuan Ding is with Heriot-Watt University, Edinburgh EH14 4AS, UK. (E-mail: yuan.ding@hw.ac.uk)

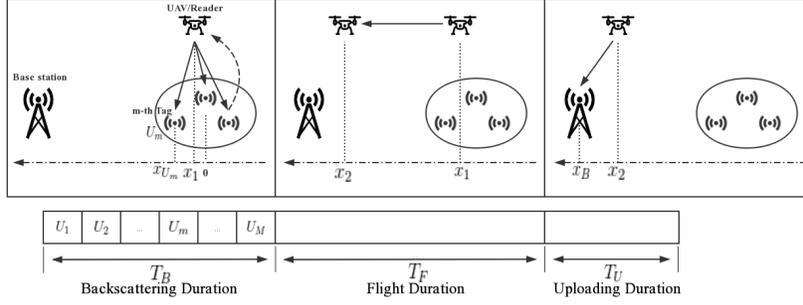


Fig. 1: System model

the m -th terrestrial tag is $d_{VU_m} = \sqrt{h^2 + (x_1 - x_{U_m})^2}$ with an elevation angle $\theta_{VU_m} = \frac{180^\circ}{\pi} \times \arcsin\left(\frac{h}{d_{VU_m}}\right)$.

After collecting the data from all the tags, the UAV flies to the location (x_2, h) near a BS at location $(x_B, 0)$ with a flight duration T_F using a constant velocity of v . In the uploading duration T_U , the UAV uploads collected data to the BS. As shown in Fig. 1, the distance for uploading from the UAV to the BS is $d_{VB} = \sqrt{h^2 + (x_2 - x_B)^2}$ with an elevation angle $\theta_{VB} = \frac{180^\circ}{\pi} \times \arcsin\left(\frac{h}{d_{VB}}\right)$.

Let us use VB to represent UAV to BS and VU_m to represent UAV to tag U_m , and denote $a \in \{VB, VU_m\}$ and $b \in \{\text{LoS}, \text{NLoS}\}$. We assume that all the channels are Nakagami- m fading channels (i.i.d) with shape factor $k_{a,b}$, which is a generalized channel model [13]. Therefore, the PDF of the channel power gain $|g_{a,b}|^2$ follows Gamma distribution,

$$f_{|g_{a,b}|^2}(x) = \frac{1}{\Gamma(k_{a,b})} \cdot \left(\frac{k_{a,b}}{\Omega}\right)^{k_{a,b}} \cdot x^{k_{a,b}-1} e\left(\frac{-k_{a,b}}{\Omega}x\right), \quad (1)$$

where Ω represents the mean value, i.e. $\Omega = \omega\beta_0 d_a^{-\alpha}$ if path loss is considered, where the path loss exponent $\alpha = 2$ [14], $\omega = 1$ for LoS propagation and $\omega = \eta_a$ for NLoS propagation, and β_0 denotes the channel power gain at the reference distance $d_0 = 1\text{m}$. In addition, the line-of-sight (LoS) probability [15] is given as

$$p_{a,\text{LoS}} = \frac{1}{1 + c \cdot \exp[-q(\theta_a - c)]}, \quad (2)$$

where c and q are constant values depending on the environment. Accordingly, non line-of-sight (NLoS) probability is

$$p_{a,\text{NLoS}} = 1 - p_{a,\text{LoS}}. \quad (3)$$

We assume the UAV works in a full duplex mode¹. Within the m -th time slot of the backscattering duration, considering a discrete-time signal model in the baseband, the received signal at the UAV is given by $y_{U_mV} = g_{VU_m} g'_{VU_m} s_1 s_3 + g_{VU_m} n_{U_m} + n_V$ [16], where s_1 is the signal transmitted by the UAV, s_3 is the tag's information signal, n_V denotes the additive white Gaussian noise (AWGN) with its power σ_V^2 at the UAV, n_{U_m} denotes the AWGN with its power $\sigma_{U_m}^2$ at the tag U_m . Besides, the received power at the UAV is given by $P'_V = \eta_R P_V |g_{VU_m}|^2 |g'_{VU_m}|^2$. Therefore, its signal-to-noise-ratio (SNR) can be derived as

¹This assumption is commonly used in most researches [16, 17]. With this assumption, perfect signal separation can be operated at the UAV. Thus, there is no interference at the UAV.

$$\gamma_{U_mV} = \frac{\eta_R P_V |g_{VU_m}|^2 |g'_{VU_m}|^2}{|g_{VU_m}|^2 \sigma_{U_m}^2 + \sigma_V^2}, \quad (4)$$

where η_R is the fraction of reflected power.

By the end of M time slots, the total data collected by UAV is $T_U R_U = \sum_{m=1}^M \frac{T_U}{M} R_m$, where R_m and R_U are the information rate of the m -th tag and the UAV, respectively. In addition, $\gamma_{\text{th}}^m \triangleq 2^{R_m} - 1$ and $\gamma_{\text{th}}^U \triangleq 2^{R_U} - 1$ are the SNR thresholds corresponding to their Shannon capacity per Hz. Besides, the signal from the UAV to the BS can be characterized by $y_{VB} = \sqrt{P_V} \cdot g_{VB} \cdot s_2 + n_B$, where s_2 is the transmit signal from the UAV and n_B is the additive white Gaussian noise of uploading channel with its power σ_B^2 . Thus, the SNR of the UAV-BS signal can be written as

$$\gamma_{VB} = \frac{P_V \cdot |g_{VB}|^2}{\sigma_B^2}. \quad (5)$$

B. Energy Outage Probability

For the m -th terrestrial tag U_m , the total energy received from the UAV consists of the energy received during backscattering single time slot and the energy harvested during other $M - 1$ time slot, which can be given as

$$\begin{aligned} \varepsilon_m &= \frac{(1 - \eta_R) \eta_C P_V |g_{VU_m}|^2}{M} T_B + \frac{(m - 1) \eta_C P_V |g_{VU_m}|^2}{M} T_B \\ &= \frac{(m - \eta_R) \eta_C P_V |g_{VU_m}|^2 T_B}{M}, \end{aligned} \quad (6)$$

where η_R is the fraction of reflected power in the backscattering period, and η_C is the circuit conversion efficiency [1, 2]. The energy outage events occur if the energy received by the m -th terrestrial tag from the UAV ε_m in Eq. (6) is less than the circuit power loss during the backscattering time slot, which can be formulated as $P_{E,m} = \Pr(\varepsilon_m < \frac{T_B}{M} P_C)$, where P_C is the circuit power loss at each terrestrial tag.

Lemma 1 (Energy Outage Probability): The energy outage probability of the m -th tag can be derived as

$$\begin{aligned} P_{E,m} &= p_{VU_m,\text{LoS}} \cdot \Gamma\left(k_{VU_m,\text{LoS}}, \frac{P_C \cdot d_{VU_m}^2 \cdot k_{VU_m,\text{LoS}}}{(m - \eta_R) \eta_C P_V \beta_0}\right) \\ &+ p_{VU_m,\text{NLoS}} \cdot \Gamma\left(k_{VU_m,\text{NLoS}}, \frac{P_C \cdot d_{VU_m}^2 \cdot k_{VU_m,\text{NLoS}}}{(m - \eta_R) \eta_C P_V \beta_0 \eta_{VU_m}}\right), \end{aligned} \quad (7)$$

where $p_{VU_m,\text{LoS}}$ and $p_{VU_m,\text{NLoS}}$ are given in Eq. (2) and Eq. (3), respectively.

Proof: See Appendix. A. ■

C. System Average Outage Probability

Note that there are two main factors causing information outage, which are the energy outage, and that the SNR of the backscattered signals at the UAV to the BS is lower than a given threshold. Thus, the outage probability between the m -th tag and the BS can be defined as

$$P_{in,m} = [1 - \Pr(\gamma_{U_{m,v}} \geq \gamma_{th}^m, \gamma_{VB} \geq \gamma_{th}^U)](1 - P_{E,m}) + P_{E,m} \\ = 1 - [1 - F_{\gamma_{U_{m,v}}}(\gamma_{th}^m)] [1 - F_{\gamma_{VB}}(\gamma_{th}^U)] [1 - P_{E,m}], \quad (8)$$

where $P_{E,m}$ is given in Eq. (7) in Lemma 1. Meanwhile, the system average outage probability can be defined as

$$P_{in} = \frac{1}{M} \sum_{m=1}^M P_{in,m}. \quad (9)$$

Then, by substituting Eq. (7), Eq. (14) with $x = \gamma_{th}^m$ and Eq. (15) with $x = \gamma_{th}^U$ into Eq. (8), the system average outage probability is finally derived in the following theorem.

Theorem 1 (The System Average Outage Probability): The closed-form expression for the system average outage probability of the UAV-assisted backscattering communications in Nakagami- m fading is derived as

$$P_{in} = 1 - \frac{1}{M} \sum_{m=1}^M [1 - p_{VB,LoS} \cdot \Gamma(k_{VB,LoS}, \frac{\gamma_{th}^U \sigma_B^2 d_{VB}^2 \cdot k_{VB,LoS}}{P_V \beta_0}) \\ - p_{VB,NLoS} \cdot \Gamma(k_{VB,NLoS}, \frac{\gamma_{th}^U \sigma_B^2 d_{VB}^2 \cdot k_{VB,NLoS}}{P_V \eta_{VB} \beta_0})] \\ \times [1 - p_{VU_m,LoS} \cdot \Gamma(k_{VU_m,LoS}, \frac{P_C \cdot d_{VU_m}^2 \cdot k_{VU_m,LoS}}{(m - \eta_R) \eta_C P_V \beta_0}) \\ - p_{VU_m,NLoS} \cdot \Gamma(k_{VU_m,NLoS}, \frac{P_C \cdot d_{VU_m}^2 \cdot k_{VU_m,NLoS}}{(m - \eta_R) \eta_C P_V \beta_0 \eta_{VU_m}})] \\ \times [1 - \int_0^\infty (p_{VU_m,LoS} \cdot \Gamma(k_{VU_m,LoS}, \frac{\gamma_{th}^m (y \sigma_{U_m}^2 + \sigma_V^2) d_{VU_m}^2}{\eta_R P_V y \beta_0}) \\ + p_{VU_m,NLoS} \cdot \Gamma(k_{VU_m,NLoS}, \frac{\gamma_{th}^m (y \sigma_{U_m}^2 + \sigma_V^2) d_{VU_m}^2}{\eta_R P_V \cdot y \cdot \eta_{VU_m} \beta_0}) \\ \cdot (p_{VU_m,LoS} \cdot f_{|g_{VU_m,LoS}|^2}(y) + p_{VU_m,NLoS} \cdot f_{|g_{VU_m,NLoS}|^2}(y)) dy], \quad (10)$$

with $p_{a,b}$ given in Eq. (2) and Eq. (3), and $f_{|g_{a,b}|^2}(y)$ given in Eq. (1), $a \in \{VB, VU_m\}$ and $b \in \{LoS, NLoS\}$.

III. UAV DATA COLLECTION LOCATION OPTIMIZATION

The average capacity of the proposed system is $\frac{1}{M} \sum_{m=1}^M R_m(1 - P_{in,m})$ with total energy cost of $T_F P_F + (T_B + T_U) P_V$, where P_F is the consumed power of UAV during fly. Thus, the optimization problem of energy efficiency under energy constraint can be formulated as

$$\max_{T_F} \eta_{en} = \frac{\frac{1}{M} \sum_{m=1}^M R_m(1 - P_{in,m})}{T_F P_F + (T_B + T_U) P_V} \quad (11) \\ s.t. \quad T_F P_F + (T_B + T_U) P_V \leq E_{total}.$$

The problem (11) can be rewritten as

$$\max_{x_1} \eta_{en}(x_1) = \frac{\frac{1}{M} \sum_{m=1}^M R_m(1 - P_{in,m}(x_1))}{\frac{x_2 - x_1}{v} P_F + (T_B + T_U) P_V} \quad (12) \\ s.t. \quad \frac{x_2 - x_1}{v} P_F + (T_B + T_U) P_V \leq E_{total}.$$

Due to the complexity of $P_{in,m}(x_1)$, the first- and the second-order derivatives cannot be derived readily. Thus, one-dimensional linear searching method is utilized. We first derive the feasible region $\{x_1 | x_1 \geq \frac{E_{total} - (T_B + T_U) P_V}{P_F} \cdot v\}$. Then, the search region of one-dimensional linear searching method to find the optimal solution of $\eta_{en}(x_1)$ in the problem (12) is $\{x_1 | x_1 \geq \frac{E_{total} - (T_B + T_U) P_V}{P_F} \cdot v\} \cap \{x_1 | x_1 \leq x_2\}$. Golden Section Method is used to find the optimal solution x_1^* of Problem (12) within the aforementioned searching region [18]. Due to the limit of the space of the paper, we do not give the details.

IV. SIMULATION RESULTS

In this section, we validate our derived analytical expressions and conduct performance analysis based on numerical simulations with the following parameters: $M = 3$; $P_C = 0.001W$; $\eta_R = 0.5$; $\eta_C = 0.5$; $\eta_{VB} = 0.5$; $\eta_{VU_m} = 0.5$; $h = 50m$; $v = 10m/s$, which are common assumptions in previously reported works. Other parameters are set as follows: $T_U = 1s$; $T_B = 1s$; $x_2 = 300$; $x_B = 500$; $c = 11.95$; $q = 0.136$; $\sigma_{U_m}^2 = \sigma_V^2 = 10^{-9}W$; $P_F = 100W$; $R_m = 1$; $\beta_0 = 1$. The shape-factor $k_{a,b}$ is 2.

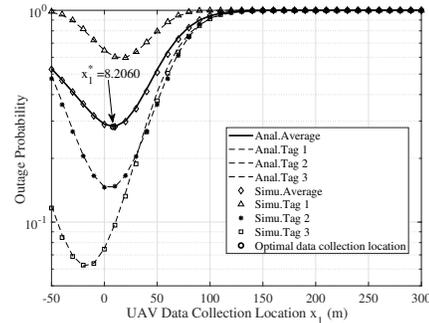


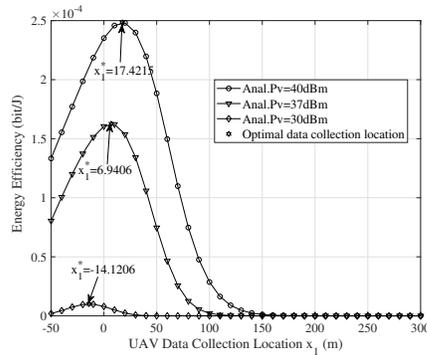
Fig. 2: System average outage probability P_{in} v.s. UAV data collection location x_1 .

Fig. 2 plots the impact of UAV data collection location on the system average outage probability P_{in} given in Eq. (10) and the outage probability of each tag $P_{in,m}$ given in Eq. (8). We see that the theoretical results match well with the Monte-Carlo simulations. It is obvious that the problem in Eq. (10) is convex, and there exists an optimal collection location leading to a minimum outage probability. The terrestrial tag with larger backscattering distance to the UAV suffers higher outage probability.

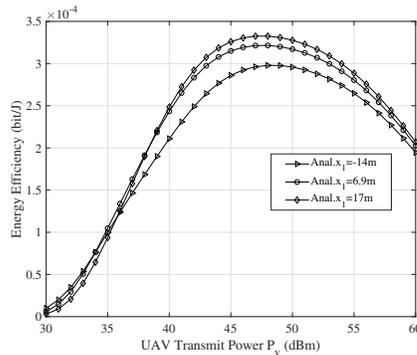
Fig. 3 plots the energy efficiency of the UAV versus (v.s.) its data collection locations and its transmit power. We can observe from Fig. 3(a) that higher UAV transmit power leads to closer UAV data collection location to the BS, and lower flight energy consumed by the UAV, contributing to enhanced energy efficiency. In addition, higher transmit power leads to lower average information outage probability, which also contributes to improved energy efficiency. Meanwhile, we can observe from Fig. 3(b) that there is an optimal transmit power leading to the maximum energy efficiency. However, the optimal transmit power on board should be feasible.

V. CONCLUSION

In this paper, a UAV-assisted backscatter communications via TDMA was studied, where a UAV acts as a data collector from multiple terrestrial backscattering tags. We derived the



(a) Energy efficiency v.s. UAV data collection location x_1 under the energy constraint with different UAV transmit power P_V .



(b) Energy efficiency v.s. UAV transmit power P_V under the energy constraint with different UAV data collection location x_1^* .

Fig. 3: Energy efficiency optimization with different variables

closed-form expression of the system average outage probability, and optimized the UAV data collection location under the energy constraint in order to achieve maximum energy efficiency. Simulation results verified our derived analysis and illustrated that there is a tradeoff between the UAV data collection location and the energy efficiency.

APPENDIX A ENERGY OUTAGE PROBABILITY

Proof The energy outage probability can be formulated as

$$\begin{aligned} P_{E,m} &= \Pr(\varepsilon_m < \frac{T_B}{M} P_C) = \Pr(|g_{VU_m}|^2 < \frac{P_C}{(m - \eta_R)\eta_C P_V}) \\ &= p_{VU_m,LOS} \cdot \Pr(|g_{VU_m,LOS}|^2 < \frac{P_C}{\omega(m - \eta_R)\eta_C P_V}) \\ &\quad + p_{VU_m,NLOS} \cdot \Pr(|g_{VU_m,NLOS}|^2 < \frac{P_C}{\omega(m - \eta_R)\eta_C P_V}). \end{aligned} \quad (13)$$

Here in Eq. (13), $\Pr(|g_{VU_m,LOS}|^2 < \frac{P_C}{\omega(m - \eta_R)\eta_C P_V})$ and $\Pr(|g_{VU_m,NLOS}|^2 < \frac{P_C}{\omega(m - \eta_R)\eta_C P_V})$ can be both calculated by the CDF corresponding to the PDF in Eq. (1), respectively. This completes the proof of Eq. (7). ■

APPENDIX B

PROOF OF $F_{\gamma_{U_m V}}(x)$ AND $F_{\gamma_{VB}}(x)$

Note that $\gamma_{U_m V}$ is given in Eq. (4), thus the CDF of $\gamma_{U_m V}$ can be derived as

$$\begin{aligned} F_{\gamma_{U_m V}}(x) &= \Pr(\frac{\eta_R P_V |g_{VU_m}|^2 |g'_{VU_m}|^2}{|g_{VU_m}|^2 \sigma_{U_m}^2 + \sigma_V^2} < x) \\ &= \int_0^\infty \Pr(|g'_{VU_m}|^2 < \frac{x(y\sigma_{U_m}^2 + \sigma_V^2)}{\eta_R P_V y}) \cdot f_{|g_{VU_m}|^2}(y) dy. \end{aligned} \quad (14)$$

where $\Pr(|g'_{VU_m}|^2 < \frac{x(y\sigma_{U_m}^2 + \sigma_V^2)}{\eta_R P_V y})$ can be calculated similarly with Eq. (13), $f_{|g_{VU_m}|^2}(y)$ is given in Eq. (1).

Similarly, the CDF of γ_{VB} can be derived as

$$\begin{aligned} F_{\gamma_{VB}}(x) &= \Pr(\gamma_{VB} < x) = p_{VB,LOS} \cdot \Gamma(k_{VB,LOS}, \frac{x\sigma_B^2 d_{VB}^2 \cdot k_{VB,LOS}}{P_V \beta_0}) \\ &\quad + p_{VB,NLOS} \cdot \Gamma(k_{VB,NLOS}, \frac{x\sigma_B^2 d_{VB}^2 \cdot k_{VB,NLOS}}{P_V \eta_{VB} \beta_0}). \end{aligned} \quad (15)$$

This completes the proof. ■

REFERENCES

- [1] K. Han and K. Huang, "Wirelessly powered backscatter communication networks: Modeling, coverage, and capacity," *IEEE Trans. Wireless Commun.*, vol. 16, no. 4, pp. 2548–2561, Apr. 2017.
- [2] X. Lu, H. Jiang, D. Niyato, D. I. Kim, and Z. Han, "Wireless-powered device-to-device communications with ambient backscattering: Performance modeling and analysis," *IEEE Trans. Wireless Commun.*, vol. 17, no. 3, pp. 1528–1544, Mar. 2018.
- [3] B. Kellogg, A. Parks, S. Gollakota, J. R. Smith, and D. Wetherall, "Wi-Fi backscatter: Internet connectivity for RF-powered devices," in *Proc. ACM SIGCOMM 2014*, ser. SIGCOMM '14. New York, NY, USA: ACM, Aug. 2014, pp. 607–618.
- [4] "LoRa backscatter: Enabling the vision of ubiquitous connectivity," in *Proc. ACM Interact. Mob. Wearable Ubiquitous Technol.*, vol. 1, no. 3. New York, NY, USA: ACM, Sep. 2017, pp. 105:1–105:24.
- [5] X. Lu, P. Wang, D. Niyato, D. I. Kim, and Z. Han, "Wireless charging technologies: Fundamentals, standards, and network applications," *IEEE Commun. Surveys Tuts.*, vol. 18, no. 2, pp. 1413–1452, 2016.
- [6] A. Varshney, O. Harms, C. Perez-Penichet, C. Rohner, F. Hermans, and T. Voigt, "LoRea: A backscatter architecture that achieves a long communication range," in *ACM SENSYS 2017*, Nov. 2017, pp. 1–14.
- [7] S. Hayat, E. Yanmaz, and R. Muzaffar, "Survey on unmanned aerial vehicle networks for civil applications: A communications viewpoint," *IEEE Commun. Surveys Tuts.*, vol. 18, no. 4, pp. 2624–2661, 2016.
- [8] Y. Zeng, J. Lyu, and R. Zhang, "Cellular-connected UAV: Potential, challenges, and promising technologies," *IEEE Wireless Commun.*, vol. 26, no. 1, pp. 120–127, Feb. 2019.
- [9] X. He, J. Bito, and M. M. Tentzeris, "A drone-based wireless power transfer and communications platform," in *2017 IEEE WPTC*, May 2017, pp. 1–4.
- [10] S. Zhang, H. Zhang, Q. He, K. Bian, and L. Song, "Joint trajectory and power optimization for UAV relay networks," *IEEE Commun. Lett.*, vol. 22, no. 1, pp. 161–164, Jan. 2018.
- [11] M. Hua, Y. Wang, Z. Zhang, C. Li, Y. Huang, and L. Yang, "Power-efficient communication in UAV-aided wireless sensor networks," *IEEE Communications Letters*, vol. 22, no. 6, pp. 1264–1267, June 2018.
- [12] M. Hua, Y. Wang, Q. Wu, H. Dai, Y. Huang, and L. Yang, "Energy-efficient cooperative secure transmission in multi-uav enabled wireless networks," *IEEE Transactions on Vehicular Technology*, pp. 1–1, 2019.
- [13] H. Liu, S. Yoo, and K. S. Kwak, "Opportunistic relaying for low-altitude UAV swarm secure communications with multiple eavesdroppers," *J. Commun. Netw.*, vol. 20, no. 5, pp. 496–508, Oct 2018.
- [14] M. Mozaffari, W. Saad, M. Bennis, and M. Debbah, "Mobile internet of things: Can UAVs provide an energy-efficient mobile architecture?" in *2016 IEEE GLOBECOM*, Dec. 2016, pp. 1–6.
- [15] A. Al-Hourani, S. Kandeepan, and S. Lardner, "Optimal LAP altitude for maximum coverage," *IEEE Wireless Commun. Lett.*, vol. 3, no. 6, pp. 569–572, Dec. 2014.
- [16] W. Saad, X. Zhou, Z. Han, and H. V. Poor, "On the physical layer security of backscatter wireless systems," *IEEE Trans. Wireless Commun.*, vol. 13, no. 6, pp. 3442–3451, June 2014.
- [17] W. Liu, K. Huang, X. Zhou, and S. Durrani, "Full-duplex backscatter interference networks based on time-hopping spread spectrum," *IEEE Trans. Wireless Commun.*, vol. 16, no. 7, pp. 4361–4377, Jul. 2017.
- [18] W. H. Press, S. A. Teukolsky, W. T. Vetterling, and B. P. Flannery, "Section 10.2. golden section search in one dimension," *Numerical recipes: the art of scientific computing*, pp. 492–496, 2007.