An optimal relay number selection algorithm for balancing multiple performance in flying ad hoc networks

SUOPING LI*1,2, (Member, IEEE), FAN WANG2, JAAFAR GABER3, (Member, IEEE), YONGQIANG ZHOU1

1School of Science, Lanzhou University of Technology, Lanzhou,730000, China
2School of Electrical and Information Engineering, Lanzhou University of Technology, Lanzhou,730000, China
3Department of Computer Science and Computer Engineering, Universite de Technologie Belfort-Montbéliard, 90010 Belfort, France

Corresponding author: Suoping Li (Email: lsuop@163.com), Fan Wang (Email: wfl573@qq.com)

This work was supported in part by the National Natural Science Foundation of China under Grant 61663024, in part by the Erasmus+ Programme under Grant 573879-EPP-1-2016-1-FR-EPPKA2-CBHE-JP, and in part by the Hongliu First Class Discipline Development Project of Lanzhou University of Technology, China.

ABSTRACT The rapid development of aviation technology has made the application of unmanned aerial vehicles (UAVs) more popular in recent years. Due to the inadequate capability of a single UAV, flying ad hoc networks (FANETs), which consist of multiple different UAVs, not only break the limits of single UAV by introducing the cooperation between UAVs, but also complete more complex missions by extending the communication range at infrastructure-less areas. One of the most important design issues for FANETs is the communication and cooperation between UAVs, and the number of UAV relays has been proved to be closely related to the improvement of system performance. However, few FANET protocols or algorithms have been proposed from the perspective of scheduling the number of relays. Inspired by the idea, this paper proposes an optimal relay number selection algorithm based on a more realistic network model which includes a novel cooperative \((n+2)\)-node system model considering the distance metrics between relays and a Nakagami-\(m\) short-term static fading channel model more suitable for UAV operation environments. The system outage probability is calculated by introducing the Meijer-G function, and a three-dimensional discrete time Markov chain (DTMC) model is established and analyzed to derive the closed-form binary expressions of throughput, energy efficiency and average transmission delay. In order to balance the three performance so as to achieve a comprehensive system performance, a trade-off factor named \(EDT\) is further proposed and maximized to evaluate the optimal number of relays. Finally, simulation results assess the impact of network parameters on system performance, and verify the superiority of the proposed algorithm over its fixed relay number counterpart.

INDEX TERMS FANETs, outage probability, relay number selection, DTMC, system performance

I. INTRODUCTION

In the past decade, the rapid development of flight control technology and integrated circuit technology has greatly promoted the application of unmanned aerial vehicles (UAVs), making the mobile communication networks composed of UAVs, which act as both sensors and routers, become a powerful complement to the static wireless sensor networks (WSNs). UAVs have gained widespread attention in the emerging communication field for their excellent advantages of high mobility, low cost, strong survivability and no risk of casualties. They not only play an important role in modern military wars such as intelligence reconnaissance and military strike, but also have broad application prospect in future civilian fields such as traffic monitoring, managing wildfire and power line inspection.

Since the 21st century, with the increasing demand of wireless users for communication modes and quality of service (QoS), taking military communication as an example, the application of UAVs has developed from the early single and large UAV to the current UAV group...
This article has been accepted for publication in a future issue of this journal, but has not been fully edited. Content may change prior to final publication. Citation information: DOI 10.1109/ACCESS.2020.3044502, IEEE Access

introduction of U2U can effectively reduce the hardware equipped with cheap and simple hardware. Obviously, the mode U2U can be built between small UAVs that are only another mode, i.e., UAV-to-UAV (U2U), in FANETs. The communicate with the ground BS or satellite, there has equipped with an expensive and complicated hardware to single UAV system, besides the communication mode shown in Figure 2.

As everyone knows, cooperative communication (CC), as a virtual multiple input multiple output (MIMO) technology, is one of the effective anti-fading technologies because it can effectively use the diversity gain obtained by neighbor nodes to assist the communication between the source node and destination node, significantly improving the system performance. In recent years, CC has been widely used in many popular wireless mobile networks such as WSNs[1], vehicular ad hoc networks (VANETs)[2]; cognitive radio networks (CRNs)[3] and wireless local area networks (WLANs)[4]. CC can effectively improve the communication capacity and transmission rate by making neighbor nodes form a relay set and cooperate with each other to forward the data information buffered in them.

Obviously, FANETs stand as a fascinating blend of CC and UAVs. Figure 1 shows the network structure of FANETs.

As shown in Figure 1, different from the traditional single UAV system, besides the communication mode UAV-to-Infrastructure (U2I) where the large UAV must be equipped with an expensive and complicated hardware to communicate with the ground BS or satellite, there has another mode, i.e., UAV-to-UAV (U2U), in FANETs. The mode U2U can be built between small UAVs that are only equipped with cheap and simple hardware. Obviously, the introduction of U2U can effectively reduce the hardware load of network and the acquisition and maintenance cost of nodes, extend the scalability and improve the survivability, making cooperative system complete missions faster than single UAV system. An application example of FANETs is shown in Figure 2.

A. RELATED WORK

FANETs were mostly used in the military field in their early days. Since the 21st century, the flourish of UAV technology has gradually prompted FANETs become a new way in the field of civil communication. Until now, a lot of related work has been proposed to study and improve them.

FIGURE 1. The network structure of FANETs.

As shown in Figure 1, different from the traditional single UAV system, besides the communication mode UAV-to-Infrastructure (U2I) where the large UAV must be equipped with an expensive and complicated hardware to communicate with the ground BS or satellite, there has another mode, i.e., UAV-to-UAV (U2U), in FANETs. The mode U2U can be built between small UAVs that are only equipped with cheap and simple hardware. Obviously, the introduction of U2U can effectively reduce the hardware load of network and the acquisition and maintenance cost of nodes, extend the scalability and improve the survivability, making cooperative system complete missions faster than single UAV system. An application example of FANETs is shown in Figure 2.

VOLUME XX, 2017

FIGURE 2. A FANET scenario for aerial surveillance.

Mentioning the origin of FANETs, there were various names of wireless networks composed of UAVs before 2013. For example, the earliest name can be traced back to the UAV ad hoc network named by T X Brown et al. at the university of Colorado in 2004[5]. T X Brown et al. developed and built a wireless network test bed using IEEE 802.11b (WiFi) radio equipments mounted on small low-cost UAVs, and described the real ad hoc network behavior among UAVs by recording detailed data on network throughput, delay, range, and connectivity under different operating regimes. Aerial robot team is another name, but as a collaborative and autonomous multi-UAV system whose network architecture is ad hoc, its research mainly focuses on the collaborative coordination of multi-UAV systems, rather than the design and optimization of network structures, algorithms or protocols[6]. The third commonly used name is aerial sensor network proposed in [7], which presented a low-cost, miniature controlled-mobile aerial sensor network called Sensor Fly. [7] also demonstrated the hardware design, flight control and collaborative localization capabilities of the Sensor Fly system.

It was not until 2013 that Ilker Bekmezci et al. first formally defined the concept of FANETs in [8], discussed different FANETs application scenarios, and clarified the differences between FANETs, MANETs (mobile ad hoc networks) and VANETs from the perspective of node mobility, topology change, computational power, etc. Furthermore, [8] introduced the main FANETs design challenges and open research issues by providing a comprehensive review of the existing FANETs literatures in order to encourage more researchers to work for the solutions. Based on [8], by introducing the basic networking architecture and main channel characteristics, [9] provided an overview on FANETs with the help of three cases: UAV-aided ubiquitous coverage, UAV-aided relaying and UAV-aided information dissemination. The
key design considerations as well as the opportunities were further highlighted to pave the way for investigators to study FANETs system. Obviously, the early work mainly focused on the overview of network basic principles, application scenarios and open issues, the research on FANETs is still in their infancy[10].

However, the unique features of FANETs, such as high mobility, low node density, and high frequency of topology changes, introduce great challenges to the communication design. Recently, many research works[11]-[25] have been proposed, and they can be classified in Figure 3 according to the different layers of open system interconnect (OSI) model.

**FIGURE 3. The related work of FANETs.**

In terms of network layer, due to the rapid mobility and highly dynamic topology, designing a routing protocol for UAV networks is a challenging task[11]. As a result, a lot of work has been devoted to the design of FANETs routing protocols[11]-[17]. Among them, several overview articles[11]-[14] have been reported from the perspective of topology-based routing[12], position-based routing[13] and cluster-based routing[11]. For example, [11] extensively surveyed and qualitatively compared the cluster-based routing protocols in terms of outstanding features, competitive advantages and limitations. Some open research issues and challenges on cluster-based routing were also discussed in [11]. Meanwhile, many efforts[15]-[17] have been made to improve network performance. An adaptive density-based routing protocol (ADR P) was proposed in [15] to improve the forwarding efficiency by dynamically fine-tuning the rebroadcasting probability according to the number of neighbor nodes. NS-2 simulation results revealed that ADRP achieved better performance than ad hoc on demand distance vector (AODV) protocol in terms of the packet delivery fraction, average end-to-end delay, and normalized routing load, etc. In order to solve the two main problems of short flight time and inefficient routing, [16] saved the communication energy by anticipating the operational requirements of UAVs to adjust their transmission power, and reduced the routing overhead by using a variant of the K-means density clustering algorithm to select the optimal cluster heads. Simulation results evaluated the performance from the aspects of energy consumption, cluster lifetime etc. Based on the optimized link-state routing (OLSR) protocol, [17] designed an OLSR extension, i.e., predictive OLSR (P-OLSR), by taking advantage of the Global Positioning System (GPS) information available on board. A test bed was built to evaluate the link performance, communication range and routing performance. Furthermore, the superiority of P-OLSR over OLSR in the presence of frequent network topology changes was proved.

Although the existing routing protocols have revealed promising results, further investigations (quick recovery, effective dynamic topology formation and maintenance, etc.) are required to take full advantage of UAVs in various applications. For example, a majority of existing works do not take energy efficiency into account, which is a vital problem for commercial off-the-shelf battery powered UAVs, and the traffic characteristics (load balancing) exchanged among nodes.

On the other hand, some media access control (MAC) protocols[18]-[22] have been designed to coordinate and provide means for UAVs to access the medium in an efficient and fair manner. Among them, [18], [19] devoted to improve network performance by designing modified MAC protocol. For example, an adaptive MAC protocol scheme for UAVs (AMUAV) was proposed in [18]. AMUAV implemented a transmission scheme in which UAVs send control packages (Request-to-Send (RTS), Clear-to-Send (CTS), and Acknowledgement (ACK)) with omni-directional antennas and switch to directional antennas to send data packets. OPNET simulation results proved that AMUAV protocol can improve throughput, end-to-end delay and bit error rate for multi-UAV systems. [20], [21] have been proposed to realize a collision-free time slot allocation. [20] proposed a token-based MAC protocol with full-duplex (FD) and multi-packet reception (MPR) radios. This protocol eliminated the packet collision by frequently updating channel state information (CSI) to ensure UAVs can have the latest CSI at any time. Performance results showed the effectiveness of proposed MAC protocol. Moreover, in order to address the well known directional antenna deafness problem, Temel et al.[22] designed a MAC protocol called as LODMAC (location oriented directional MAC) by incorporating the utilization of directional antennas and location estimation of neighbor nodes. Performance results showed that LODMAC protocol outperforms the well-known DMac (directional MAC) protocol in terms of throughput, utilization, average network delay and fairness.

It is obvious that some existing MAC protocols have well investigated on eliminating packet collision, dealing with hidden exposure terminal problems and directional deafness issues. However, how to provide continuous and timely positioning information for directional antenna is still a difficult point. Moreover, MAC protocols which can not only adapt to dynamic UAV networks but also make full use of channel resources has not been well studied.
Comprehensively, some works [23]-[25] have been proposed from the perspective of cross-layer analysis. Considering the problem of spectrum scarcity, [23] proposed an integration of UAVs with cognitive radio (CR) technology, discussed the CR-UAV integration issues, and highlighted future research challenges in layer-wise manner. From the perspective of cross-layer design, an adaptive hybrid communication protocol including a position-prediction-based directional MAC protocol (PPMAC) and a self-learning routing protocol based on reinforcement learning (RLSRP) was proposed in [24]. Simulation results showed that the proposed PPMAC overcome the directional deafness problem with directional antennas, and RLSRP provided an automatically evolving and more effective routing scheme. Tian W et al. [25] firstly studied the issue of how to utilize CC to improve the connectivity recovery efficiency in FANETs, and designed a CC-based connectivity recovery algorithm named C3RUN, which not only uses CC to enlarge UAV’s communication range and thus achieve quick repair of network connectivity, but also enables UAVs to proactively move to better places for ensuring the establishment of relay links. Simulation results revealed that C3RUN not only can achieve connectivity recovery with less nodes and shorter distance to move, but also can achieve 100% success ratio for connectivity recovery compared with existing work.

The cross-layer design approach has received considerable attention in recent years. A cross-layer protocol, which allows exchanging useful information and feedback among layers for routing decisions, would be an interesting choice to achieve the reliability requirement of FANETs. In future design of cross-layer protocols, machine learning (ML) based algorithms (such as deep learning, neural networks, reinforcement learning, etc.) are considered as most promising techniques. Moreover, more low-complex algorithms for solving the cross-layer optimization problems need to be designed to make the cross-layer protocols more efficiently employed in practical applications.

For quick review, the development of FANETs and some state-of-the-art works are shown in Figure 4.

**Timetable: The development and research status of FANETs**

2004  -  T X Brown et al. developed and built a UAV ad hoc network test bed by installing WiFi radio equipment on small, low-cost UAVs [5].

2009  -  Purohit A et al. proposed a low-cost, miniature controlled-mobile aerial sensor network system called SensorFly [7].

2010  -  An adaptive MAC protocol scheme for UAVs (AMUAV) was proposed to switch the antenna mechanism used to transmit different packets [18].

2012  -  A token-based MAC layer scheme using FD and MPR radios was proposed to eliminate the packet collision [20].

2015  -  Tareque M H et al. classified the existing routing protocols for FANETs into six major categories and introduced them based on various performance criteria [14].

2015  -  Temel et al. designed a MAC protocol called as LOMDAC by incorporating the utilization of directional antennas and location estimation of neighbor nodes [22].

2016  -  Rosati S et al. proposed a predictive OLSR (P-OLSR) to improve the routing performance in the presence of frequent network topology changes [17].

2017  -  An adaptive density-based routing protocol (ADRP) was proposed to improve the forwarding efficiency [15].

2018  -  Aadil F et al. used a variant of the K-means density clustering algorithm to select the optimal cluster heads to reduce the routing overhead [16].

2018  -  An adaptive hybrid communication protocol including a PPMAC protocol and a RLSRP protocol was proposed in [24].

2019  -  Tian W et al. designed a CC-based connectivity recovery algorithm named C3RUN to improve the connectivity recovery efficiency [25].

2020  -  Li B et al. proposed a modified CSMA/CA protocol to deal with the communications heterogeneity to analyze the throughput of uplink [19].
To sum up, the current work of FANETs has the following deficiencies. Firstly, compared with the research of routing protocols, less work has been concentrated on the design of MAC layer schemes. Although some existing MAC protocols [17]-[21] have been proposed, most of them mainly devote to improve single performance or solve optimization problems such as position prediction and directional deafness. To the best of our knowledge, there is no MAC scheme has been proposed from the perspective of controlling the number of relays. As described in [8]: “in many applications, the performance enhancement is closely related with the number of UAVs. For example, the higher number of UAVs can complete a search and rescue operation faster. FANET protocols and algorithms should be designed so that any number of UAVs can operate together with minimal performance degradation”. Coincidentally, this idea coincides with us. In our finished work [26], an important conclusion that under a specific communication scenario (such as the communication distance between source nodes and destination nodes), there always exists an optimal number of relays in the system which can maximize the energy efficiency has been drawn. Based on this, it is possible to design an optimal relay number selection algorithm which can dynamically adjust the number of relays according to the changing communication distance?

Secondly, the majority of existing system models proposed to analyze multi-relay cooperative networks ignores the distance difference between relays. In FANETs, very low node density and very fast channel quality change lead to a fact that many existing theoretical assumptions considered in other networks cannot be well applied to FANETs. For example, [27]-[29] all studied the MAC schemes under multi-relay networks, but they all made an ideal assumption that the probability parameters such as outage probability, frame error rate (FER) or packet error rate (PER) from the source (S) to the destination (D), and that from Rr to the destination (D) were equal. Specifically, [29] has assumed that \( PER_{SR} \approx PER_{SR} \approx \cdots \approx PER_{SR} \approx PER_{SR} \) and \( PER_{sR} = PER_{sD} = \cdots = PER_{sD} = PER_{sD} \). That is, [29] equated multiple relays to a “super” single relay and thus ignored the distance metrics between different relays. It is reasonable in the low-speed mobile networks such as WSNs and WLANs. However, it is obviously unsuitable for FANETs.

Finally, most of the existing literatures only analyze single performance optimization without considering a trade-off between multiple performances. Until now, single performance optimization is the main object of current FANET research, such as improving the throughput[19], and enhancing the connectivity recovery efficiency[25]. However, it is insufficient in high-speed mobile communication scenarios that emphasize comprehensive performance. On the other hand, there always exist some contradictory performances in the system. How to balance them so as to achieve an optimal comprehensive performance is also a key problem that needs to be studied.

B. CONTRIBUTIONS AND PAPER STRUCTURE
Focusing on the above shortcomings, our main contributions are summarized as follows:

a) The proposal of a novel network model. Focusing on the problem that the distance difference caused by network characteristics of FANETs cannot be ignored, a novel \((n+2)\)-node system model considering the distance metrics between relays is proposed. Furthermore, a Nakagami-\(m\) short-term static fading channel model is selected for FANETs, and the outage probability of all communication links is analyzed by introducing the Meijer-G function.

b) The design of an optimal relay number selection algorithm. According to the inspiration that the improvement of FANET performance is closely related to the number of UAVs, a distance-based optimal relay number selection algorithm is designed. Moreover, the proposed strategy can balance multiple performances by introducing a trade-off factor which is called as EDT.

c) The establishment of a novel discrete time Markov chain (DTMC) model. Focusing on the mathematical description problem of the proposed algorithm, a 3-dimensional DTMC model is built and analyzed. Furthermore, the closed-form expressions of throughput, energy efficiency and average transmission delay are derived by calculating the transmission completion probability \(Pr(i,k,J)\).

d) Four simulation cases are performed to evaluate the influence of network parameters such as the maximum transmission number, communication distance, and the location of relays on system performance. Moreover, the superiority of the proposed strategy over its counterpart whose optimal number of relay is fixed are verified.

The rest of this paper is organized as follows: the system model and the proposed algorithm are introduced in Section II. Section III analyzes the system outage probability, builds a 3-D DTMC, and calculates the transmission completion probability \(Pr(i,k,J)\) by deriving one-step state transition probability. System performance analysis and the evaluation of optimal relay number are presented in Section IV and Section V, respectively. Section VI presents the simulation cases. Finally, the conclusion is drawn in Section VII.

II. SYSTEM MODEL AND OPTIMAL RELAY NUMBER SELECTION ALGORITHM
In this section, we build both system and channel models for multi-relay cooperative FANETs. Furthermore, the main assumptions used in this paper are described, and the optimal relay number selection algorithm is introduced.
A. SYSTEM MODEL

As shown in Figure 5, this paper considers a multi-relay cooperative FANET consisting of a UAV source node (S), a ground BS destination node (D), and n UAV neighbor nodes \( R_i, i = \{1, 2, \cdots, n\} \). It is assumed that the distance between any two nodes does not exceed the communication range of antennas and the hidden terminals do not exist.

![FIGURE 5. A FANET assisted by multiple UAV neighbor nodes.](image)

As described in Section I, it is the distinct characteristic of FANETs that makes many existing multi-relay system models which ignore the distance difference between relays significantly unsuitable for analyzing FANETs. Therefore, as shown in Figure 6, this paper builds a dedicated \((n+2)\)-node system model for multi-relay FANETs. In this model, the distance between S (D) and \( R_n \), i.e., \( d_{SR} ( d_{DR}) \), is different.

Each node in the network operates in half duplex and is equipped with the same switched beam antenna arrays [22] that can focus more energy on predetermined directions. That is, system can only conduct once transmission towards one of the \( M \) directions of the antennas within a time slot. Given that the main beam angle is \( \theta \). In radians, the beam number of the antenna can be calculated as:

\[
M = 2\pi/\theta, \quad (M > 1)
\]

In order to cover the entire communication range, antennas may transmit data continuously and the beams calculated in formula (1) are numbered from 1 to \( M \). Nodes can only transmit or receive data through one of these antenna beams at a given time slot. Obviously, in order to broadcast data, S must carry out sequential transmissions through all the beams. This paper also assumes that all nodes use the identical antenna patterns and can maintain the beam direction at any time. Moreover, the switching time between antenna beams is ignored which can be achieved by using some digital signal processing (DSP) technologies. A perfect CSI is considered, all nodes in the network remain fully synchronized, and the feedback channel is ideal.

![FIGURE 6. A \((n+2)\) node system model for multi-relay FANETs.](image)

As for the selection of channel models, according to the conclusion drawn in [30]: “the fading model should be chosen according to the operation environment. For example, Rayleigh fading can be more suitable for low altitude crowded area applications, while Nakagami-m and Weibull fading with high fading parameters best fit for high altitude open space missions.” It is known that Nakagami-\(m\) channel is equivalent to the Rayleigh when \( \text{m}=1 \). For generality, this paper adopts independent Nakagami-\(m\) fading. On the other hand, because the very high mobility is the most distinct characteristic between FANETs and traditional terrestrial networks, we further select the short-term static feature to characterize the complex FANETs time-varying channel. It is worth noting that in our finished work for typical terrestrial networks such as VANETs[26] and WSNs[31], the long-term static Rayleigh fading is chosen as the channel model. The difference between long-term and short-term static feature lies in the changing frequency of channel fading coefficient. Given that the duration for sending a specific data frame is an ARQ process, and each transmission in it refers to an ARQ round. A frame can be transmitted at most \( L \) times in one ARQ process. The channel fading coefficient of link \( i-j \) under the \( l \)th ARQ round, i.e., \( \text{h}_{ij,l} \), changes differently under the two channel features. In particular, \( \text{h}_{ij,l} \) keeps constant while \( l \) changes for the same node pair \( (i, j) \) under the long-term static channel, i.e., \( \text{h}_{ij,l} \) can be rewritten as \( \text{h}_{ij} \), and the channel fading coefficients are independent for different node pairs. However, for the short-term static channel, \( \text{h}_{ij,l} \) is independent and identically distributed (i.i.d) while \( l \) changes for the same node pair \( (i, j) \), and independent for different node pairs. It is obvious that the short-term static feature is more accurate to describe the frequent FANETs time-varying channel. Therefore, this paper adopts independent Nakagami-\(m\) short-term static fading as the channel model, and selects the additive white Gaussian noise (AWGN) as the noise model.

Furthermore, denoting the power intensity gain \( \omega_{ij} \triangleq \left| \text{h}_{ij} \right|^2 \), because of the adopted channel model, \( \omega_{ij} \) obeys the Gamma distribution with parameter of \( \left( m_j, \Omega_j / m_j \right) \), i.e., \( \omega_{ij} \sim \text{Ga} \left( m_j, \Omega_j / m_j \right), (i \in (S,R), j \in (R,D), i \neq j) \), and its probability density function (PDF) is:

\[
f_{\omega_{ij}}(v) = \left( \frac{m_j}{\Omega_j} \right)^{m_j} v^{(m_j-1)} \frac{1}{
\Gamma(m_j)} \exp \left( -\frac{m_j}{\Omega_j} v \right) \tag{2}
\]

where \( m_j > 0 \) is the Nakagami multi-path fading parameter of link \( i-j \), \( \Omega_j = E \left[ \omega_{ij} \right] > 0 \) is the average power of link \( i-j \) ( \( E \) is the expectation), and \( \Gamma(\cdot) \) is the Gamma function as shown in formula (3):
This paper assumes that relays work in the decode-and-forward (DF) mode, and thus after the first ARQ round, \(k \in \{0,1,2,\cdots,n\}\) out of \(n\) neighbor nodes successfully decode the data frame broadcast by S, and assist S in retransmitting the frame.

**B. OPTIMAL RELAY NUMBER SELECTION ALGORITHM**

In order to maximize the performance improvement in a specific communication scenario, an optimal relay number selection algorithm is proposed in Figure 7. It is worth noting that the detailed hardware implementation of the proposed algorithm is beyond the scope of this paper.

\[
\Gamma(x) = \int_0^{x^+} t^{-1}e^{-t}dt, \quad x > 0 \tag{3}
\]

Frames, for transmission. After D receives the handshake message broadcast from S, through advanced positioning technologies such as GPS, differential GPS (DGPS), assisted GPS (AGPS) and inertial measurement unit (IMU), D exchanges handshake messages, and calculates \(n^*\) relays to participate in the cooperation based on the communication distance between it and S. In the transmission module, S transmits data frames in the form of omnidirectional broadcast, if D decodes the data frame correctly; S will transmit a new data frame in the next ARQ round. Otherwise, D judges whether the ARQ round \(l\) is less than the maximum value \(L\). if yes, system enters retransmission phases. Otherwise, a NACK is fed back by D and \(l\) is reset to 0. In the retransmission phase where relays work in the DF mode, \(k\) out of \(n^*\) relays successfully decode the data frame broadcast from S, if \(k=0\), the data frame is retransmitted by S in the next ARQ round. Otherwise, \(k\) relays directionally transmit data frames to D. Similarly, \(k\) relays do not stop retransmission until they receive an ACK from D, or \(L\) is reached.

**III. SYSTEM OUTAGE PROBABILITY ANALYSIS AND MARKOV MODEL ESTABLISHMENT**

This section devotes to analyze system outage probability and to establish the corresponding Markov model.

**A. SYSTEM OUTAGE PROBABILITY ANALYSIS**

Supposing that the transmitted signal is \(x\) with a signal-to-noise ratio (SNR) of \(\gamma\), the received signal is \(y\), and the channel fading coefficient is \(h\), then the outage probability of the data link with signaling rate \(r\) bits/slot/Hz is:

\[
Pr_{out} = Pr\{I(x, y | h) < r\} = Pr\{\log_2 \left(1 + \gamma |h|^2\right) < r\} \tag{4}
\]

In this paper, the outage events that happen on link S-R, and S-D with and without successful reception at relays in the \(l\)th ARQ round are denoted as \(SR_{out,l}\), \(SRD_{out,l}\) and \(SD_{out,l}\), respectively. The assumption: \(d_{r,S} \neq d_{r,D} \neq \cdots \neq d_{r,D}\) but \(d_{r,S} = d_{SR} = \cdots = d_{SR} = d_{SR}\) is made to simplify the analysis. It is known that \(E[\omega_{ji}^\beta]\), is proportional to \(d_{ji}^{-\beta}\), where \(\beta\) is the path loss factor. Therefore, in the Nakagami-\(m\) short-term static fading channel with AWGN communication scenario, the probability of above outage events is expressed as follows:

\[
Pr\left(SR_{out, l}\right) = \prod_{a=1}^{l} Pr\left\{\log_2 \left(1 + \gamma h_{SR,a}^\sim\right) < r\right\} = \prod_{a=1}^{l} Pr\left\{\omega_{SR,a} < \varepsilon\right\} = \prod_{a=1}^{l} F_{\omega_{SR,a}} (\varepsilon)
\]

\[
Pr\left(SRD_{out, l}\right) = \prod_{a=1}^{l} Pr\left\{\log_2 \left(1 + \gamma h_{SRD,a}^\sim\right) < r\right\}
\]

\[
Pr\left(SD_{out, l}\right) = \prod_{a=1}^{l} Pr\left\{\log_2 \left(1 + \gamma h_{SD,a}^\sim\right) < r\right\}
\]
\[
\prod_{r=t_{l-1}+1}^{t_l} \Pr \left[ \log_2 \left( 1 + \sum_{i=1}^{\gamma} \|h_{r,D,i}\|^2 \right) < r \right] = \prod_{a=1}^{2^k} \Pr \left[ \omega_{SD,a} < \varepsilon \right] \cdot \prod_{a=1}^{2^k} \Pr \left[ \sum_{i=1}^{\gamma} \|h_{r,D,i}\|^2 < \varepsilon \right]
\]
(6)
\[
\Pr(SD_{out}) = \prod_{a=1}^{2^k} \Pr \left[ \omega_{SD,a} < \varepsilon \right]
= \prod_{a=1}^{2^k} \Pr \left[ \log_2 \left( 1 + \gamma \|h_{r,D,a}\|^2 \right) < r \right] = \prod_{a=1}^{2^k} F_{\omega_{SD,a}}(\varepsilon)
\]
(7)

where \( \varepsilon = \frac{2^r - 1}{\gamma} \), \( F_{\omega}(x) \) is the cumulative distribution function (CDF) of the random variable \( X \). \( \omega_{SR}, \omega_{SD}, \omega_{RD} \) are the elements of power intensity gain matrices
\[
W_{se} \triangleq \begin{bmatrix} \omega_{SR,1} & \omega_{SR,2} & \cdots & \omega_{SR,2^k} \\ \omega_{SD,1} & \omega_{SD,2} & \cdots & \omega_{SD,2^k} \\ \omega_{RD,1} & \omega_{RD,2} & \cdots & \omega_{RD,2^k} \\ \end{bmatrix}
\]
and \( W_{SN} \triangleq \begin{bmatrix} \omega_{SR,1} & \omega_{SR,2} & \cdots & \omega_{SR,2^k} \\ \omega_{SD,1} & \omega_{SD,2} & \cdots & \omega_{SD,2^k} \\ \omega_{RD,1} & \omega_{RD,2} & \cdots & \omega_{RD,2^k} \\ \end{bmatrix} \), respectively. \( T^k_R \)
represents that \( k \) out of \( n \) UAV neighbor nodes successfully decode the data frame for the first time in the \( T^k_R \) ARQ round. Equations (5) and (7) mean that the transmission of links S-R and S-D have failed (i.e., outage happens) for \( l \) ARQ rounds, while equation (6) represents the transmission of link S-D assisted by \( k \) relays has failed for \( l \) ARQ rounds. In particular, in the \( l \) failed transmissions seen by D, the first \( T^k_R \) times received from S, and the last \( (l - T^k_R) \) times sent by \( k \) relays. The probability of \( T^k_R \) can be expressed as:
\[
\Pr \left( T^k_R = t \right) = \binom{n}{k} \left[ \frac{m_t}{\Omega_k} \right] \left[ \frac{m_t}{\Omega_k} - 1 \right] \cdots \left[ \frac{m_t}{\Omega_k} - (t - 1) \right] \cdot \frac{m_t}{\Omega_k} - t \cdot \frac{m_t}{\Omega_k} \cdot \cdots \frac{m_t}{\Omega_k} - n \cdot \frac{m_t}{\Omega_k},
\]
(8)
which means that \( k \) out of \( n \) UAV relays do not correctly decode the frame transmitted from S until the \( n \)th ARQ round.

It is known that \( \omega_{h,t} \) is a Gamma random variable with parameter of \( (m_t, \Omega_t/m_t) \). Therefore, the solution of
\[
\Pr \left( \sum_{i=1}^{\gamma} \|h_{r,D,i}\|^2 < a \right) \text{ in formula (6) is essentially to calculate the CDF of the sum of } k \text{ independent but not necessarily identical (i.i.d) Gamma random variables. Denoting } \xi_1, \xi_2, \cdots, \xi_k \text{ as } k \text{ i.i.d Gamma random variables with parameters of } m_t \text{ and } \Omega_t (t = 1, 2, \cdots, k). \text{ Making } Z = \sum_{i=1}^{\gamma} \xi_i, \text{ the PDF and CDF of } Z \text{ can be obtained according to [32]:}
\]
\[
f_z(z) = \prod_{r=1}^{m_t} \left( \frac{m_t}{\Omega_t} \right) \left[ \frac{m_t}{\Omega_t} - 1 \right] \cdots \left[ \frac{m_t}{\Omega_t} - (t - 1) \right] \cdot \frac{m_t}{\Omega_t} - t \cdot \frac{m_t}{\Omega_t} \cdot \cdots \frac{m_t}{\Omega_t} - n \cdot \frac{m_t}{\Omega_t},
\]
(9)
\[
f_z(z) = \prod_{r=1}^{m_t} \left( \frac{m_t}{\Omega_t} \right) \left[ \frac{m_t}{\Omega_t} - 1 \right] \cdots \left[ \frac{m_t}{\Omega_t} - (t - 1) \right] \cdot \frac{m_t}{\Omega_t} - t \cdot \frac{m_t}{\Omega_t} \cdot \cdots \frac{m_t}{\Omega_t} - n \cdot \frac{m_t}{\Omega_t}
\]
(10)

where \( \kappa = \sum_{i=1}^{k} m_i \) is an integer, and \( G^{\alpha,\beta}_{\alpha,\beta} \) is the Meijer-G function shown in formula (11):
\[
G^{\alpha,\beta}_{\alpha,\beta}[x] = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha + \beta)} \int_0^1 \frac{\mu^\alpha \Gamma(\alpha) \Gamma(\beta)}{\mu^\alpha \Gamma(\alpha) \Gamma(\beta)} \cdot \Gamma(\alpha + \beta) \cdot \mu^d \cdot ds
\]
(11)

In particular, when \( m_t = 1 \) (i.e., the average power \( \Omega_t = 1 \) and \( t = 1, 2, \cdots, k \)), the PDF of the sum of \( k \) exponential random variables (i.e., the Rayleigh fading channel) can be obtained as follows based on equation (9):
\[
f_z^{\text{Rayleigh}}(z) = \prod_{r=1}^{m_t} \left( \frac{m_t}{\Omega_t} \right) \left[ \frac{m_t}{\Omega_t} - 1 \right] \cdots \left[ \frac{m_t}{\Omega_t} - (t - 1) \right] \cdot \frac{m_t}{\Omega_t} - t \cdot \frac{m_t}{\Omega_t} \cdot \cdots \frac{m_t}{\Omega_t} - n \cdot \frac{m_t}{\Omega_t},
\]
(12)
which can be simplified in equation (15) by using the Meijer-G identity:
\[
f_z^{\text{Rayleigh}}(z) = \prod_{r=1}^{m_t} \left( \frac{m_t}{\Omega_t} \right) \left[ \frac{m_t}{\Omega_t} - 1 \right] \cdots \left[ \frac{m_t}{\Omega_t} - (t - 1) \right] \cdot \frac{m_t}{\Omega_t} - t \cdot \frac{m_t}{\Omega_t} \cdot \cdots \frac{m_t}{\Omega_t} - n \cdot \frac{m_t}{\Omega_t}
\]
(13)

From above, we can obtain the expressions of outage probability for all data links.

B. MARKOV MODEL ESTABLISHMENT AND ANALYSIS

According to the system model and optimal relay number selection algorithm described in Section II, a 3-dimensional DTMC model is established in Figure 8.

In the DTMC, state S represents that D decodes the data frame successfully, state F represents that D fails to
decode the data frame until the $L$th ARQ round is reached. State $R_{i,k,j}$ represents that in the $i$th ARQ round, $k$ relays correctly decode the data frame $j$ times, but D fails to decode it, and $1 \leq i \leq L-1$, $0 \leq k \leq n$, $0 \leq j \leq i$.

![Diagram](Image)

**FIGURE 8.** The 3-dimensional DTMC model.

Obviously, there are totally $(L+1)+n\sum_{i=1}^{L-1}j$ states in this DTMC. From Figure 8, $j = 0$ when $k = 0$. $j \neq 0$ and $1 \leq j \leq i$ when $k \neq 0$ (i.e., $1 \leq k \leq n$).

Based on the obtained expressions of outage probability, the one-step transition probability of this 3-D DTMC model can be analyzed. There are mainly three types of transition:

1) The transition from the initial state $S(F)$.

The state $S$ (or $F$) that represents a successful (or failed) transmission of the previous data frame can only move to state $S$, $R_{i,0,0}$, or $R_{i,k,1}$ ($1 \leq k \leq n$), reflecting the transmission events that in the first ARQ round, D decodes the frame correctly; both D and n relays fail to decode the frame; $k$ out of $n$ relays successfully decode the frame for the first time, but D fails to decode the frame, respectively. Their corresponding one-step transition probability is:

$$Pr_{SS} = Pr_{SF} = Pr(SD_{out,1})$$

$$Pr_{SR_{i,0,0}} = Pr_{SR_{i,0,0}} = Pr(SD_{out,i}) \cdot Pr^*(SR_{out,i})$$

Summing $k$ in formula (18):

$$\sum_{k=1}^{n} Pr_{SR_{i,j}} = \sum_{k=1}^{n} Pr_{F_{j,k,i}}$$

$$= Pr(SD_{out,i}) \sum_{k=1}^{n} \left(\binom{n}{k} \left(1 - Pr(SR_{out,i})\right) \cdot Pr^*(SR_{out,i})\right)$$

$$= Pr(SD_{out,i}) \left[1 - Pr^*(SR_{out,i})\right]$$

Obviously, the sum of formulas (16)-(18) equals to one, which satisfies the basic rule that the sum of all the one-step transition probability of state $S(F)$ is one.

2) The transition from the general state $R_{i,0,0}$.

The general state $R_{i,0,0}$ that represents both D and n relays fail to decode the frame in the $i$th ARQ round can move to state $S$, $R_{i+1,0,0}$, $R_{i+1,k,1}$ ($1 \leq k \leq n$), or $F$ (only when $i = L-1$), reflecting the transmission events that in the $(i+1)$th ARQ round, D decodes the frame correctly; D and n relays still fail to decode the frame; $k$ out of $n$ relays successfully...
decode the frame for the first time, but D fails to decode the frame; D finally fails to decode the frame and discards it, respectively. Their transition probability can be expressed as:

\[
\Pr_{R_{t,0}}(i) = 1 - \Pr(SD_{out,i} | SD_{out,i}), \quad 1 \leq i \leq L-1 \tag{20}
\]

\[
\Pr_{R_{t,0}}(i) = \Pr(SD_{out,i} | SD_{out,i}), \quad 1 \leq i \leq L-2 \tag{21}
\]

\[
\Pr_{R_{t,0}}(i) = \Pr(SR_{out,i} | SR_{out,i}), \quad 1 \leq i \leq L-2 \tag{22}
\]

Similarly, formula (24) can be derived by summing \( k \) in formula (22), and the same conclusion can be drawn as that from formula (19).

\[
\sum_{k=0}^{n} \Pr_{R_{t,0}}(i) = \Pr(SD_{out,i} | SD_{out,i})
\]

\[
\sum_{k=0}^{n} \Pr_{R_{t,0}}(i) = \Pr(SR_{out,i} | SR_{out,i})
\]

\[
\Pr_{R_{t,0}}(i) = \Pr(SD_{out,i} | SD_{out,i}) \left[ 1 - \Pr(SR_{out,i} | SR_{out,i}) \right], \quad 1 \leq i \leq L-2
\]

3) The transition from the general state \( R_{t,k,j} \).

The general state \( R_{t,k,j} \) that represents in the \( i \)th ARQ round, \( k \) relays have successfully decoded the frame \( j \) times while D still fails to decode the frame can move to state \( S \), \( R_{t+1,k,j+1} \), or \( F \) (only when \( i = L-1 \)), reflecting the transmission events that in the \((i+1)\)th ARQ round, D decodes the frame correctly with the cooperation of \( k \) relays; D still fails to decode the frame correctly with the cooperation of \( k \) relays; and D finally fails to decode the frame and discards it when the maximum transmission number \( L \) is reached, respectively. Therefore, the transition probability expressions of state \( R_{t,k,j} \) can be expressed as:

\[
\Pr_{R_{t,k,j}} = 1 - \Pr(SRD_{out,j+1} | SRD_{out,j}), \quad 1 \leq j \leq L-1, \quad T_{t,j} = i - j + 1, \quad 1 \leq k \leq n \tag{25}
\]

\[
\Pr_{R_{t,k,j}} = \Pr(SRD_{out,j+1} | SRD_{out,j}), \quad 1 \leq j \leq L-2, \quad T_{t,j} = i - j + 1, \quad 1 \leq k \leq n \tag{26}
\]

\[
\Pr_{R_{t,k,j}} = \Pr(SRD_{out,L} | SRD_{out,L-1}), \quad 1 \leq j \leq L-1, \quad T_{t,j} = L - j, \quad 1 \leq k \leq n \tag{27}
\]

From above, all one-step transition probability expressions of the 3-D DTMC model can be obtained.

C. TRANSMISSION COMPLETION PROBABILITY

In order to analyze system performance in Section IV, it is necessary to obtain the probability that the ARQ process for transmitting a frame is completed through \( k \) relays cooperatively retransmit the frame \( J(0 \leq J \leq i-1) \) times in the \( i \)th ARQ round, i.e., \( \Pr(i,k,J) \), which is also called as the transmission completion probability in this paper. It is worth noting that all the possible values of \( J \) when \( i \geq 2 \) are presented in Figure 9. Obviously, when \( i = 1 \), it means that the transmission process ends in the first ARQ round, i.e., D successfully decodes the frame in the first round.

From Figure 9, the transmission completion probability \( \Pr(i,k,J) \) can be expressed as:

\[
\Pr(i,k,J) = \begin{cases} 
1 - \Pr(SD_{out,i}), & i = 1 \\
\Pr(\hat{i},k,J), & 2 \leq i = \hat{i} \leq L-1 \\
\Pr(L,k,J), & i = L \end{cases}
\]

where \( \Pr(\hat{i},k,J) \) represents the probability that the ARQ process is completed before reaching the maximum transmission number \( L \) (i.e., \( 2 \leq i = \hat{i} \leq L-1 \)), while \( \Pr(L,k,J) \) represents the probability that the process is not completed until \( L \) is reached (\( i = L \)). They are expressed as:

\[
\Pr(\hat{i},k,J) = \begin{cases} 
\Pr(\hat{i},0,0), & k = 0, \quad J = 0 \\
\Pr(\hat{i},\hat{k},J), & 1 \leq \hat{k} = \hat{n}, 1 \leq J = \hat{i} - 1 \end{cases}
\]

and

\[
\Pr(L,k,J) = \begin{cases} 
\Pr(L,0,0), & k = 0, \quad J = 0 \\
\Pr(L,\hat{k},J), & 1 \leq \hat{k} = \hat{n}, 1 \leq J = \hat{i} - 1 \end{cases}
\]

FIGURE 9 All possible values of cooperative retransmission times \( J \). The expressions of four sub-formulas in formulas (29) and (30) are respectively analyzed below:

1) The formula (29-a), i.e., \( \Pr(\hat{i},0,0) \).

\[
\Pr(\hat{i},0,0), \quad \text{which represents that the S-D link successfully completes the transmission in the } \hat{i} \text{ th ARQ round without any cooperation, can be expressed as follows according to the multiplication of one-step transition probability of DTMC model shown in Figure 8:}
\]
This article has been accepted for publication in a future issue of this journal, but has not been fully edited. Content may change prior to final publication. Citation information: DOI 10.1109/ACCESS.2020.3044502, IEEE Access

\[
\begin{align*}
P_r(\hat{L}, \hat{k}, \hat{j}) &= \prod_{i=0}^{k} \left[ P_{S-R-D}(\hat{L}_k, \hat{k}_x, \hat{k}_y) \right] \\
&= P_r(\hat{S}-R-D) \cdot P_r(\hat{S}-R-D) \cdot \ldots \cdot P_r(\hat{S}-R-D), \quad 1 \leq \hat{j} \leq \hat{L} - 2
\end{align*}
\]

where \( P_r(\hat{S}-R-D) \) represents the probability that the S-D link successfully completes transmission in the \( \hat{j} \) th ARQ round.  

2) The formula (29-b), i.e., \( P_r(\hat{i}, \hat{k}, \hat{j}) \).

\[
\begin{align*}
P_r(\hat{i}, \hat{k}, \hat{j}) &= P_r(\hat{S}-R-D) \cdot P_r(\hat{S}-R-D) \cdot \ldots \cdot P_r(\hat{S}-R-D), \quad 1 \leq \hat{j} \leq \hat{L} - 2
\end{align*}
\]

The complete expression of \( P_r(\hat{i}, \hat{k}, \hat{j}) \) can be obtained by substituting equations (31)-(32) and (34)-(35) into (28).

However, it is necessary to reorganize the formula (28) for energy efficiency analysis in Section IV. Specifically, combining formulas (29-a) and (30-a) into (36) (denoted as \( P_r(\hat{i}, 0, 0) \)), (29-b) and (30-b) into (37) (denoted as \( P_r(\hat{i}, \hat{k}, \hat{j}) \)) to represent the probability that the transmission is completed without (i.e., \( k = J = 0 \)) and with (i.e., \( 1 \leq \hat{k} \leq n, 1 \leq \hat{j} \leq \hat{L} - 1 \)) cooperation of relays, respectively.

IV. SYSTEM PERFORMANCE ANALYSIS

Based on the one-step transition probability and transmission completion probability obtained in Section III, this section mainly derives the closed-form expressions of system performance, which includes throughput, energy efficiency and average transmission delay, under arbitrary maximum transmission number \( L \) and relay number \( n \).

A. THROUGHPUT

This paper defines system throughput as the average number of data frames correctly received by D within one round of ARQ.
time slot, i.e., the average number of time slots consumed by the DTMC in the state $S$, which can be equivalent to the steady state probability of state $S$. Supposing that the steady state distribution of the 3-D DTMC model is $\pi = (\pi_1, \pi_2, \pi_3)$ where $\pi_3$ is the steady probability of state $S$. It is worth noting that throughout $\pi_3$ is a binary function of $L$ and $n$. Then $\pi_3$ can be derived according to the state transition probability matrix composed of all one-step transition probability in Section III. $\pi_3$ can be obtained as:

$$\pi_3 (L, n) = \frac{1 - FER(L, n)}{h(L, n)}$$ (39)

where $FER(L, n)$ is the system frame error rate (FER), $h(L, n)$ is a binary function of $L$ and $n$. They are shown as:

$$FER(L, n) \triangleq \begin{cases} \Pr_{SR,0,0} \cdot \Pr_{R,0,0} + \sum_{k=1}^{n} \Pr_{SR,k,1} \cdot \Pr_{R,k,1} & L = 2 \\ \text{equation (42)} & L \geq 3 \end{cases}$$ (40)

and

$$h(L, n) \triangleq \begin{cases} 1 + \Pr_{SR,0,0} + \sum_{k=1}^{n} \Pr_{SR,k,1} & L = 2 \\ \text{equation (43)} & L \geq 3 \end{cases}$$ (41)

where equations (42) and (43) are:

$$FER(L, n) \triangleq \Pr_{SR,0,0} \cdot \sum_{k=1}^{n} \left[ \prod_{i=1}^{L-1} \left[ \Pr_{R,0,0} \cdot \Pr_{R,i,1} \right] \cdot \Pr_{R,L-1,i,0} \cdot \Pr_{R,L-1,i,1} \right]$$

$$+ \sum_{k=1}^{n} \left[ \prod_{i=1}^{L-1} \left[ \Pr_{R,0,0} \cdot \Pr_{R,i,1} \right] \cdot \Pr_{R,L-1,i,1} \cdot \Pr_{R,L-1,i,F} \right]$$ (42)

$$\sum_{k=1}^{n} \left[ \prod_{i=1}^{L-1} \left[ \Pr_{R,0,0} \cdot \Pr_{R,i,1} \right] \cdot \Pr_{R,L-1,i,1} \cdot \Pr_{R,L-1,i,F} \right]$$

$$+ \sum_{i=L-2,j=L-1-k}^{L-k-2} \sum_{j=L-1-i}^{L-k-1-i} \left[ \Pr_{R,k,i} \cdot \Pr_{R,k,i+1,j} \cdot \Pr_{R,L-k-1,i+1,j} \cdot \Pr_{R,L-k-1,i+1,j,F} \right]$$ (43)

and

$$h(L, n) \triangleq h(L-1, n) + \Pr_{SR,0,0} \cdot \Pr_{R,0,0} \cdot \Pr_{R,L-1,i,0}$$

$$+ \sum_{k=1}^{n} \left[ \prod_{i=1}^{L-2} \left[ \Pr_{R,0,0} \cdot \Pr_{R,i,1} \right] \cdot \Pr_{R,L-1,i,1} \cdot \Pr_{R,L-1,i,F} \right]$$

$$+ \sum_{k=1}^{n} \left[ \prod_{i=1}^{L-2} \left[ \Pr_{R,0,0} \cdot \Pr_{R,i,1} \right] \cdot \Pr_{R,L-1,i,1} \cdot \Pr_{R,L-1,i,F} \right]$$ (43)

The throughput $\pi_3 (L, n)$ can be derived by substituting formulas (40)-(43) into (39).

**B. ENERGY EFFICIENCY**

Energy efficiency is the average number of data frames successfully transmitted by system within a time slot per energy (Joule). Similarly, the binary function of energy efficiency $\eta(L, n)$ can be expressed as:

$$\eta(L, n) = K \cdot \frac{1 - FER(L, n)}{P_{avg}(L, n)}$$ (44)

where $K$ (frames/s) is the average number of data frames transmitted in a time slot, $P_{avg}(L, n)$ is the average power consumption value per a frame under arbitrary $L$ and $n$, which are calculated as:

$$P_{avg}(L, n) = \sum_{i=1}^{L} \sum_{j=1}^{n} \Pr(i, k, j) \cdot P(i, k, j)$$ (45)

where $Pr(i,k,j)$ is the solved transmission completion probability shown in equation (38), $P(i,k,j)$ is its corresponding power consumption value. It is worth noting that this paper uses a general power consumption model which considers the power consumption of all the amplifiers $P_A$ and the power consumption of transmitter and receiver circuit blocks $P_c$, which can be expressed as:

$$P_{c} = P_{DAC} + P_{filr} + P_{mix} + P_{syn}$$

$$P_{c} = P_{DAC} + P_{filr} + P_{mix} + P_{syn} + P_{ADC} + P_{syn}$$ (46)

where $P_{DAC}$, $P_{filr}$, $P_{mix}$, $P_{syn}$, $P_{filr}$, $P_{mix}$, $P_{IF4}$ and $P_{ADC}$ are the power consumption values of the digital-to-analog converter (DAC), the filter at the transmitter, the mixer, the frequency synthesizer, the filter at the receiver, the low-noise amplifier (LNA).

Based on equation (38), (45) can be rewritten as:

$$P_{avg}(L, n) = \Pr(1,k,J) \cdot P(1,k,J) + \sum_{i=1}^{L} \sum_{j=1}^{n} \Pr(i,k,j) \cdot P(i,k,j)$$ (47)

where $P(1,k,J)$, $P(i,k,0)$ and $P(i,k,j)$ can be expressed respectively as:

$$P(1,k,J) = M \cdot (P_a + P_c) + (n+1) P_c$$ (48)

$$P(i,0,0) = \begin{bmatrix} M(P_a + P_c) + (n+1)(n+1) & P_c \\ 2 \leq i \leq L \end{bmatrix}$$ (49)

$$P(i,k,j) = \begin{bmatrix} M(i-J) + k & \begin{bmatrix} P_a + P_c \end{bmatrix} \\ 2 \leq i \leq L \end{bmatrix}$$ (50)

This article has been accepted for publication in a future issue of this journal, but has not been fully edited. Content may change prior to final publication. Citation information: DOI 10.1109/ACCESS.2020.3044502, IEEE Access
\( P_{\text{ave}}(L,n) \) can be obtained by calculating equations (48)-(50) and substituting the results into equation (47). Then, the energy efficiency \( \eta(L,n) \) can be derived.

C. AVERAGE TRANSMISSION DELAY

The average transmission delay in this paper is defined as the average total transmission time per a data frame. It is known that under arbitrary \( L \) and \( n \), one frame needs to be transmitted \( E[T_{\text{SD}}(L,n)] \) times on average before it is correctly received by D, where the first \( E[T_{\text{SD}}(L,n)] \) times are broadcast by S. Therefore, \( T(L,n) \) can be expressed as:

\[
T(L,n) = \frac{M \cdot E[T_{\text{SD}}(L,n)] + \{E[T_{\text{SD}}(L,n)] - E[T_{\text{SD}}(L,n)]\]}{K}
\]

where \( E[T_{\text{SD}}(L,n)] \) is calculated as:

\[
E[T_{\text{SD}}(L,n)] = \sum_{i=1}^{L} i \cdot \text{Pr}(T_{\text{SD}} = i)
\]

Based on equation (28), \( \text{Pr}(T_{\text{SD}} = i) \) can be rewritten as:

\[
\text{Pr}(T_{\text{SD}} = i) = \text{Pr}(\text{SD}_{\text{out}} = i) \cdot \text{Pr}(T_{\text{SD}} = i)
\]

and

\[
\text{Pr}(T_{\text{SD}} = i) = \text{Pr}(\text{SD}_{\text{out}} = i) \cdot \text{Pr}(T_{\text{SD}} = i)
\]

After calculating \( E[T_{\text{SD}}(L,n)] \), known that if \( n \) relays fail to decode the frame until \( L \), they will refuse to receive it in the \( L \)th ARQ round. Therefore, \( E[T_{\text{SD}}(L,n)] \) are derived as:

\[
E[T_{\text{SD}}(L,n)] = \sum_{i=1}^{L} i \cdot \text{Pr}(T_{\text{SD}} = i)
\]

where \( \text{Pr}(T_{\text{SD}} = i) \) has been given in equation (33).

D. SIMULATION EXPERIMENTS

In order to evaluate the optimal relay number \( n^* \) that can balance throughput, energy efficiency and average transmission delay from a general value \( n \).

It is reasonable that high throughput and energy efficiency, and low transmission delay are three of the most crucial considerations in the design of wireless network strategy. As a result, it is ideal to increase throughput and energy while reducing transmission delay in most wireless mobile networks including FANETs. However, these performance metrics are often contradictory to each other. For instance, reducing power consumption of battery-powered devices leads to a long packet delay in the network[34]. It is this trade-off relationship that makes it very difficult to improve multiple performance at the same time.

In order to determine the appropriate number of relays, this paper proposes a trade-off factor called EDT (energy, delay and throughput) according to the above optimization principle. That is:

\[
EDT = \eta(L,n) \cdot \frac{\pi_k(L,n)}{T(L,n)}
\]

It can be seen from equation (57) that the main goal of our proposed optimal relay number selection algorithm is to maximize \( EDT \) to balance the throughput, energy efficiency and average transmission delay. Based on the system performance in Section IV, because the number of relays \( n \) is the important parameter that affects these performances simultaneously (i.e., all the three performance metrics are a function of \( n \) when \( L \) is given), the optimization function of \( EDT \) can be obtained, which is shown in equation (58), by using \( n \) as the control knob parameter.

\[
\text{maximize } n^* \cdot EDT
\]

Obviously, the optimal value \( n^* \) can be obtained by rigorously searching a sufficient range for \( n \), thus solving the integer programming problem shown in formula (58). The acquisition of \( n^* \) means that the ground BS (i.e., D) can calculate and schedule optimal number of relays according to the different communication distance between S and D, achieving the highest comprehensive performance in the specific communication environment.

VI. SIMULATION EXPERIMENTS

In order to analyze the impact of network parameters on system performance and to verify the effectiveness of the proposed algorithm, this section performs four study cases by using MATLAB numerical simulations.

The system parameters adopted in the simulation are summarized in Table I. \( q_{sd} \) is the ratio of the distance between S and R (i.e., \( d_{sd} \)) to the distance between S and D (i.e., \( d_{sd} \)). That is, \( q_{sd} = d_{sd} / d_{sd} \). Then \( q_{R,D} = d_{R,D} / d_{sd} \). Then \( q_{R,D} = d_{R,D} / d_{sd} \). Then \( t = [1, 2, \ldots, n] \) is \( n \) random numbers in the range of \( [1 - q_{sd}] \), and the communication range of antennas is set to 1000 meters (m).

\[
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\]
A. STUDY CASE 1: THE IMPACT OF NETWORK PARAMETERS ON SYSTEM PERFORMANCE

To evaluate the impact of network parameters on system performance, Figures 10-12 simulate throughput, energy efficiency and average transmission delay as a function of \( n \) under the same \( d_{SD} \), different \( L \) and \( q_{SR} \), respectively.

It can be seen firstly from Figure 10 and 12 that with the increase of \( n \), both throughput and delay monotonically change (throughput increases but delay decreases) to a certain value and then keep constant, indicating that there indeed exists a suitable \( n \) in the system. Beyond this value, the increase of system cost (such as increasing the number of UAVs) will not improve the system performance.

Secondly, Figure 10-12 also show that the maximum transmission number \( L \) is directly proportional to both throughput and energy efficiency, but inversely proportional to the average transmission delay, which is realistic because the more the transmissions are performed, the more payloads the system will transmit, the throughput and energy

### TABLE I

| Parameter | Value          | Parameter | Value          |
|-----------|----------------|-----------|----------------|
| \( P_s \) | \( 1.3 \times 10^{-4} \) W | \( L \)  | 2,3            |
| \( P_w \) | \( 10^{-4} \) W | \( \theta \) | \( \pi \)    |
| \( P_r \) | \( 5 \times 10^{-4} \) W | \( d_{SO} \) | [100,1000] m |
| \( \gamma \) | 105 dB     | \( n \)  | [0,10]         |
| \( \beta \) | 4            | \( K \)  | 5 frames / s   |
| \( r \)  | 0.5 bits / slot / Hz | \( m \)  | 1              |

FIGURE 10. Case 1: the impact of parameters on throughput.

However, the trend of energy efficiency in Figure 11 which increases first and then decreases with \( n \), even reflecting that after exceeding an appropriate \( n \), the increase of system cost will lead to the decrease of energy efficiency! This also demonstrates the importance of choosing a suitable \( n \) for improving system performance. On the other hand, it can be easily seen from Figure 11 that the energy efficiency is maximal when \( n=1 \). It is because that the channel quality of link S-D deteriorates with the increase of \( d_{SD} \). In the medium distance communications (i.e., \( d_{SD}=500 \) in this case), its channel quality is worse than that of short distance communications, making the outage probability of link S-D gets higher and system performance becomes poorer. The relay cooperation can effectively compensate the path loss caused by increasing communication distance. Therefore, as shown in Figure 11, the energy efficiency performs optimal at \( n=1 \), which proves the effectiveness of introducing relay to improve system performance.

FIGURE 11. Case 1: the impact of parameters on energy efficiency.

FIGURE 12. Case 1: the impact of parameters on transmission delay.

Secondly, Figure 10-12 also show that the maximum transmission number \( L \) is directly proportional to both throughput and energy efficiency, but inversely proportional to the average transmission delay, which is realistic because the more the transmissions are performed, the more payloads the system will transmit, the throughput and energy
efficiency are naturally improved. However, it is obvious that this improvement is at the expense of the transmission delay.

Thirdly, it can be clearly seen from Figure 10-12 that no matter which performance is, it almost performs optimally under the curve of $q_{SD} = 0.6$, which proves that the closer the relays gets to the middle position of $d_{SD}$, the higher performance the system will achieve. Moreover, these relays should be preferred to the optimal relays.

Case 1 provides some theoretical references for the design and optimization of future multi-relay wireless networks from the perspective of parameter selection and setting.

B. STUDY CASE 2: THE IMPACT OF COMMUNICATION DISTANCE ON SYSTEM PERFORMANCE

To evaluate the impact of the communication distance between S and D (i.e., $d_{SD}$) on system performance, Figures 13-15 simulate throughput, energy efficiency and average transmission delay as a function of $n$ under the same $L$ and $q_{SD}$, different $d_{SD}$, respectively.

Throughput

![Figure 13](image)

Case 2: the impact of $d_{SD}$ on throughput.

Energy efficiency

![Figure 14](image)

Case 2: the impact of $d_{SD}$ on energy efficiency.

On the other hand, it can be clearly seen from Figure 13-15 that the longer the $d_{SD}$ is, the greater $n$ the system needs to achieve optimal performance. Taking the energy efficiency in Figure 14 as an example. When $d_{SD} = 200m$ (i.e., short distance communications), the energy efficiency performs best under $n = 0$. At this point, the cooperation of relays does not improve energy efficiency but leading to a waste of energy consumption. When $d_{SD} = 500m$ (i.e., medium distance communications), energy efficiency performs best under $n = 1$, which shows that the advantage of relays is gradually presented with the increase of $d_{SD}$. However, when $d_{SD} = 800m$ and $d_{SD} = 1000m$ (i.e., long distance communications), energy efficiency performs optimal only under multi-relay cooperation. Specifically, $n = 2$ when $d_{SD} = 800m$, $n = 3$ and when $d_{SD} = 1000m$.

This is also because that longer communication distance leads to greater path loss, and thus more relays are needed to assist source nodes in transmitting to ensure system performance.

Case 2 fully demonstrates the importance of using an appropriate number of relays to effectively compensate for the larger attenuation of system performance caused by longer communication distance.

C. STUDY CASE 3: EVALUATION OF THE OPTIMAL RELAY NUMBER $n^*$

In order to evaluate the optimal relay number $n^*$ analyzed in Section V, Figure 16 simulates EDT, which is shown in equation (57), as a function of $n$ under the same $L$ and $q_{SD}$, but different $d_{SD}$.

The following conclusions can be drawn from Figure 16 that: when S and D communicate in short distance ($d_{SD} < 400m$), it is not recommended to introduce any relay for cooperation. That is, system can achieve the
highest comprehensive performance at $n^* = 0$. However, if S and D communicate in medium distance ($400m < d_{SD} < 800m$), a single neighbor UAV is preferred to act as the relay, i.e., $n^* = 1$. Otherwise, if S and D communicate in long distance ($800m \leq d_{SD} \leq 1000m$) where S-D link is difficult to directly communicate successfully, system can achieve the best comprehensive performance only by increasing the number of relays (i.e., multi-relay cooperation). Specifically, it is recommended to use 2-3 UAVs as the relays for cooperation when $d_{SD} = 800m$, while 4-5 UAVs are recommended to assist S in transmission under $d_{SD} = 1000m$.

![Figure 16. Case 3: evaluation of $n^*$.](image)

Case 3 proves that the optimal algorithm proposed in this paper can effectively select the optimal number of relays according to different communication distances, so as to achieve the highest comprehensive system performance.

D. STUDY CASE 4: CONTRAST EXPERIMENT

For completeness, case 4 compares our optimal relay number selection algorithm with its fixed relay number counterpart. Figure 17 simulates EDT as a function of $d_{SD}$ under the same $L$ and $q_{SR}$, but different $n$. It can be clearly seen from Figure 17 that, compared with the four curves of fixed relay number schemes, our proposed algorithm can enable system to maintain the highest comprehensive performance under any circumstances by dynamically adjusting the number of relays according to the different communication distance.

To be specific, when $d_{SD} < 400m$, the curve of $n^*$ coincides with that of $n = 0$, indicating that relays are not introduced in the short distance communication scenario; when $400m < d_{SD} < 800m$, the curve of $n^*$ overlapped with that of $n = 1$, indicating that the system coordinates a UAV to participate in cooperation based on the increasing communication distance; when $800m < d_{SD} < 1000m$, the curve of $n^*$ coincides with that of $n = 2$ and $n = 4$ successively, indicating that the system can compensate for the attenuation effect caused by large path loss and space loss in the long distance communication scenario by introducing multiple UAVs to participate in the cooperation, so as to effectively guarantee the system QoS.

![Figure 17. Case 4: contrast experiment.](image)

Case 4 powerfully proves that the optimal relay number selection algorithm proposed in this paper can keep the highest comprehensive performance of multi-relay FANETs system at any communication distance.

VII. CONCLUSIONS

This paper proposes an optimal relay number selection algorithm which can balance multiple system performance for multi-relay FANETs. Based on a novel system model that adds the distance metrics between relays, a 3-D DTMC model considering the number of relays at any maximum transmission number is established by designing the algorithm flow. The one-step transition probability, steady-state distribution and transmission completion probability of the 3-D DTMC model are respectively solved by analyzing the system outage probability, and then the closed-form binary functions of system performance including throughput, energy efficiency and average transmission delay are derived.

Finally, MATLAB numerical simulation results not only analyze the influence of network parameters including the maximum transmission number, location of relays, number of relays and communication distance on system performance, but also evaluate the optimal relay number $n^*$ by introducing and maximizing a trade-off factor EDT. The effectiveness of the proposed algorithm compared with its fixed relay number counterpart is also verified.

In future work, we will further study the performance optimization of FANETs combining with the smart Internet of Things (IoT) based on the inspiration of [35] that proposes a novel routing protocol for FANETs using modified AntHocNet.
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Suoping Li, IEEE Member, is a professor of school of science at Lanzhou University of Technology (LUT), China. He received the Ph.D. degree in Signal and Information Processing from Beijing Jiaotong University, China, in 2004. He also hold a M.Sc. degree in Stochastic Model with Applications from Lanzhou University in 1996, and B.Sc. degree in Mathematics from Northwest Normal University in 1986. He was a visiting scholar at Swiss Federal Institute of Technology Zurich (ETH) from May 2007 to May 2008, and also a visiting professor at American East Texas Baptist University (ETBU) from Aug to Dec 2011. He completed short academic visits at Universite de Technologie Belfort-Montbeliard (UTBM) France in 2017 and at Universiti degli Studi dell’Aquila (UNIVAQ) Italy in 2018. His primary research interests include stochastic signal analysis, design of data communication protocol, stochastic modeling analysis of wireless communication network such as ad hoc network, cognitive radio network, and wireless sensor network. Email: bsopil@163.com

Fan Wang received her BS degree on Telecommunication Engineering from Tianjin University of Commerce, China, in 2014. She is currently a PhD candidate on Control Theory and Control Engineering at Lanzhou University of Technology, China. Her research orientation is in the areas of stochastic modeling analysis of wireless communication systems, and wireless sensor network. Email: w1573@qq.com
Jaafar Gaber received the Ph.D. degree in computer science from the University of Science and Technology of Lille, Lille, France, in 1998. He was a Research Scientist at the Institute of Computational Sciences and Informatics, George Mason University, Fairfax, VA. He is currently a full Professor of computational sciences and computer engineering at the University of Technology of Belfort-Montbeliard, France. His research interests include high-performance computing, parallel processing and distributed systems, and experimental performance evaluations. Email: gaber@utbm.fr.

Yongqiang Zhou is an associate professor of school of science at Lanzhou University of Technology (LUT), China. He received the Ph.D. degree in Control Theory and Control Engineering from LUT in 2014. He also holds a M.Sc. degree in Operational Research and Cybernetics from LUT in 2011, and B.Sc. degree in Mathematics and Applied Mathematics from Huanghuai University in 2008. His research interests are in the areas of stochastic modeling analysis of wireless sensor networks, error control theory and cooperative communication. Email: zhouypaper@126.com