

# UAV-Assisted Emergency Networks in Disasters

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## Abstract

Reliable and flexible emergency communication is a key challenge for search and rescue in the event of disasters, especially for the case when base stations (BSs) are no longer functioning. Unmanned aerial vehicle (UAV) assisted network is emerging as a promising method to establish emergency networks. In this article, a unified framework of UAV-assisted emergency network is established in disasters. First, the trajectory and scheduling of UAV are jointly optimized to provide wireless service to ground devices with surviving BSs. Then, the transceiver design of UAV and establishment of multi-hop ground device-to-device (D2D) communication are studied to extend the wireless coverage of UAV. In addition, multi-hop UAV relaying is added to realize information exchange between the disaster areas and outside through optimizing the hovering positions of UAVs. Simulation results are presented to show the effectiveness of these three schemes. Finally, open research issues and challenges are discussed.

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## Index Terms

Emergency wireless networks, multi-hop relaying, trajectory optimization, transceiver design, unmanned aerial vehicle.

## INTRODUCTION

Emergency communication network is crucial for emergency rescue in natural disasters, especially when the communications infrastructure, such as base stations (BSs), are destroyed due to damage. However, existing methods lack in flexibility, which are limited by environment and space. To overcome these challenges, unmanned aerial vehicles (UAVs) can be utilized by acting as flying BSs to provide wireless coverage to the ground devices in disasters, due to their inherent advantages of flexibility and mobility [1].

Recently, UAV-assisted communications and networking have attracted plenty of interest from both academia and industry [2]–[12]. The throughput of UAV-enabled mobile relaying was optimized by Zeng *et al.*, by jointly optimizing the transmit power and relay trajectory in [2]. In [3], Zhao *et al.* proposed a UAV-assisted secure transmission scheme in hyper-dense networks via caching. A blind beam tracking scheme was proposed for UAV-satellite communications by Zhao *et al.* in [4], using large-scale antenna array at the UAV. In [5], some excellent work was done by Wu *et al.* to maximize the minimum throughput of ground devices by jointly optimizing the UAV's trajectory, transmit power and scheduling. Energy trade-off was considered to achieve data collection from ground to the UAV by Yang *et al.* in [6], via two kinds of trajectory optimization. In [7], the channel models of UAV communications were characterized through practical measurements by Ahmed *et al.*. A novel UAV-enabled wireless power transfer system was proposed by Xu *et al.* in [8], in which a UAV-enabled energy transmitter delivered wireless energy to multiple ground energy receivers. In [9], Cheng *et al.* proposed a UAV trajectory optimization scheme to offload traffic for BSs at the edges of several adjacent cells. In [10], Wu *et al.* characterized the capacity region of UAV-enabled two-user broadcast channel, via jointly

optimizing trajectory and transmit power or rate. Two typical multi-UAV relaying schemes of a single multi-hop link and multiple dual-hop links were studied by Chen *et al.* in [11], in which the optimal hovering positions were derived. In [12], Menouar *et al.* demonstrated the possible applications of intelligent transportation systems based on UAVs, with the potential and challenges highlighted.

Although excellent research has been conducted on UAV communications, very few works have focused on the aspect of UAV-assisted emergency networks in disasters [13]–[15]. In [13], Erdelj and Natalizio demonstrated the disaster management applications of UAV networks and discussed some open research issues. Message wireless transmission systems with the assistance of UAVs in large-scale disasters were studied by Mase and Okada in [14]. UAV flying path was optimized by Christy *et al.* for D2D communication in disasters in [15]. A systematic study of UAV-assisted emergency networks is missing in the literature.

In this article, a unified framework of UAV-assisted emergency networks in disasters is established. First, the flight trajectory and communication scheduling of UAV are jointly optimized, to provide wireless service for mobile devices with the surviving ground BSs. Then, the establishment of multi-hop D2D and the transceiver design of UAV are discussed in the scenario without ground BSs, to effectively extend the wireless coverage of UAV. Furthermore, to realize the information exchange between disaster and outside areas in the above two scenarios, the multi-hop UAV relaying scheme is proposed to optimize the hovering positions of UAVs. Simulation results are presented to illustrate the proposed schemes, and some interesting open research issues and challenges are pointed out for UAV-assisted emergency networks.

The rest of this article is organized as follows. In the next section, the framework of UAV-assisted emergency networks in disasters is first presented. Then, the flight trajectory and communication scheduling of UAV are jointly optimized. In addition, the establishment of multi-hop D2D and transceiver design of UAV are studied. Furthermore, the multi-hop UAV relaying

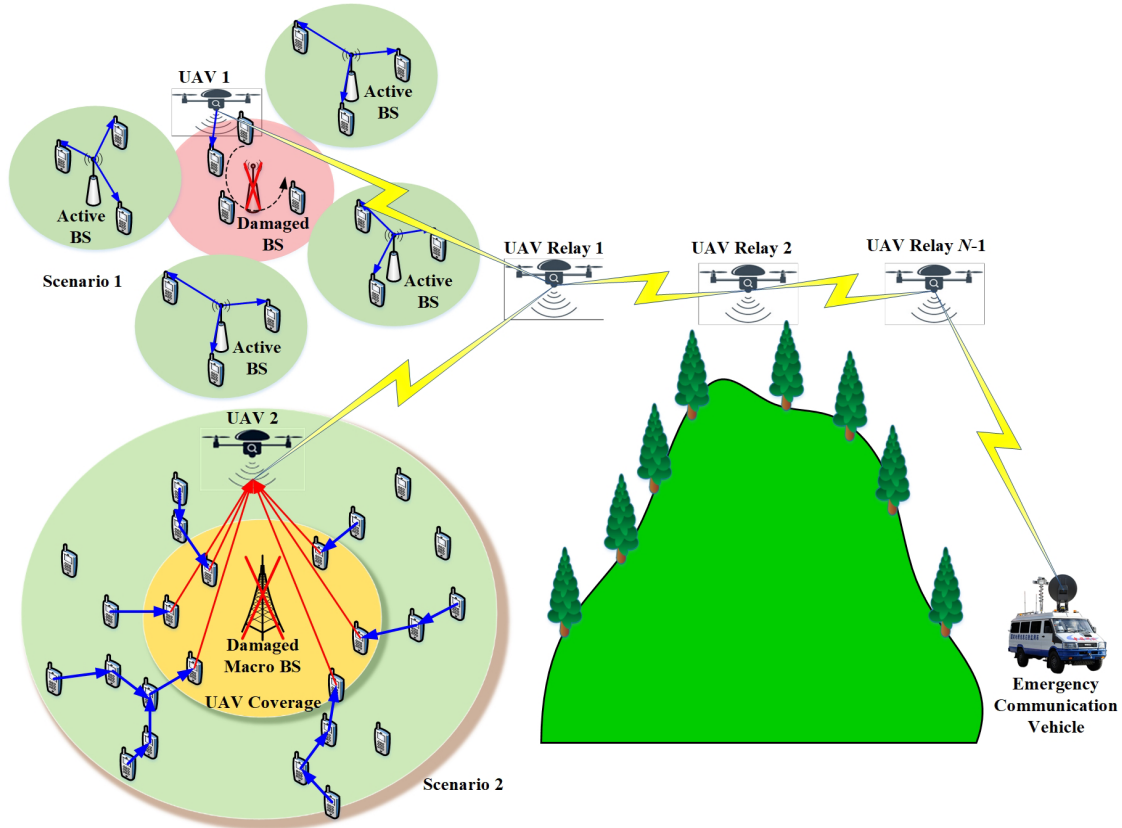


Fig. 1. Framework of UAV-assisted emergency networks in disasters.

scheme is proposed to optimize the positions of UAVs. Finally, open research issues and challenges are discussed, followed by conclusions in the final section.

### FRAMEWORK OF UAV-ASSISTED EMERGENCY NETWORKS

The framework of UAV-assisted emergency networks in disasters is shown in Fig. 1, which is described as follows:

- *Scenario 1*: In the scenario with active ground BSs, UAVs can cooperate with the surviving BSs to provide wireless service for the ground devices. In this case, the flight trajectory and communication scheduling can be jointly optimized to improve the performance.
- *Scenario 2*: In the scenario with no BSs, a large-scale UAV can act as flying BS to provide wireless connections, with the help of multi-hop D2D to extend its coverage area. In

addition, UAV transceiver design can be further utilized to improve the reliability.

- *Multi-hop UAV Relaying*: The information exchange between disaster areas and outside in both Scenario 1 and Scenario 2 can be realized via multi-hop UAV relaying, in which the optimal hovering positions of UAVs can be derived with low complexity.

In the following sections, these key scenarios for UAV-assisted emergency networks will be discussed in detail.

## JOINT TRAJECTORY AND SCHEDULING OPTIMIZATION

In disasters, victims and rescue workers are usually randomly distributed. It is hard to reach them when BSs are partially damaged. UAVs can be fully exploited to provide wireless connections to the ground devices via specific flight trajectories, due to their flexibility and mobility. In [5], some fundamental work has been conducted to jointly optimize the UAV's trajectory, transmit power and scheduling without considering any surviving ground BSs. Due to the mobility of UAV, it can fly close to the ground devices to achieve better performance; however, this may cause severe interference to the devices in other cells. When there exist some active surviving BSs as shown in Scenario 1 of Fig. 1, more complex situations should be considered, and the interference between the BS-served devices and UAV-served devices should be properly avoided to guarantee the reliable transmission [9], which will be discussed in this section.

### *Problem Formulation*

In a cellular network with several adjacent cells, the BS for the central cell is assumed to be damaged due to natural disasters, as shown in the Scenario 1 of Fig. 1. Thus, a UAV is deployed to provide wireless connections to the ground devices in the central cell. To guarantee reliable transmission and avoid severe interference to the BS-served devices, we assume that the UAV flies periodically at a fixed altitude  $H$ . Each flying cycle  $T$  can be further divided into  $N$  time

slots equally. Multiple antennas are equipped at each BS, while a single antenna is equipped at the UAV.

Define the UAV's horizontal position at the  $n$ th slot in a specific flying cycle  $\mathbf{w}(n)$  as  $[x(n), y(n)]^T$ , and its maximum speed as  $V$ . Thus, the starting and ending positions of the UAV in a specific cycle should be located at the same point, i.e.,  $\mathbf{w}(1) = \mathbf{w}(N)$ . In addition, the flying speed of UAV in each time slot should not exceed  $V$ , which requires that  $\|\mathbf{w}(n+1) - \mathbf{w}(n)\|^2$  be smaller than or equal to  $(VT/N)^2$ . To schedule the transmission of UAV, we define the binary parameter  $s_k(n)$ .  $s_k(n)$  equals 1 or 0 means that the  $k$ th device is served or not served by the UAV at the  $n$ th slot, respectively. Due to the limited capability of single-antenna UAV, at most one device is served by it at each time slot. Thus,  $\sum_{k \in \mathcal{K}} s_k(n)$  is equal to either 1 or 0, where  $\mathcal{K}$  is the set of UAV-served devices. This time division multiple access (TDMA) mode can achieve higher reliability with tolerant latency.

To improve the transmission efficiency of UAV and guarantee the quality of service (QoS) of the BS-served users, the sum rate of the UAV-served devices can be maximized by jointly optimizing the communication scheduling  $\mathbf{S}$  as  $\{s_k(n), \forall k, \forall n\}$  and flight trajectory  $\mathbf{W}$  as  $\{\mathbf{w}(n), \forall n\}$ , with the constraints of  $\mathbf{w}(n)$ ,  $s_k(n)$ , and the rate threshold for each BS-served and UAV-served device satisfied. The optimization is centralized at the UAV, with some necessary feedback and control from the surviving BSs to UAV. However, this optimization is extremely difficult to solve, due to the fact that it is a mixed-integer non-convex problem.

To handle this problem effectively, we first relax the binary variables  $s_k(n)$  into continuous ones  $\widehat{s}_k(n)$  that lie between 0 and 1. Thus, the suboptimal solutions can be calculated through solving the sub-problem with fixed trajectory  $\mathbf{W}$  and the sub-problem with fixed scheduling  $\mathbf{S}$ , iteratively, whose convergence can be guaranteed. When the trajectory  $\mathbf{W}$  is fixed, the problem becomes a linear programming, which can be solved easily through classic optimization algorithms, for example the interior-point method. When the scheduling  $\mathbf{S}$  is fixed, the problem

remains non-convex, which is still difficult to tackle. To handle this subproblem effectively, the constraints are first transformed into convex ones via successive convex optimization, and then, block coordinate descent is applied to change this subproblem into an approximately convex one, which can also be solved via classic algorithms. After convergence via iterations, the calculated continuous variables  $\widehat{s}_k(n)$  should be turned back into binary ones via comparing their values with 0.5. Thus, the suboptimal values of  $\mathbf{W}$  and  $\mathbf{S}$  can be achieved via this iterative algorithm with low computational complexity. In addition, power allocation of UAV at each time slot is not considered in this scheme to avoid complex controls. Instead, the UAV can fly close to the ground nodes to achieve optimal performance due to its flexibility.

Through the above joint optimization scheme for UAV, the average throughput of UAV-served devices can be maximized with the QoS of BS-served devices guaranteed. This is achieved by the joint consideration of flying close to its served devices to improve the throughput and staying away from the BS-served devices to avoid interference.

### *Simulation Results*

In Fig. 2, the trajectory of the UAV is demonstrated to maximize the sum rate of the UAV-served users. We set  $H = 50$  m,  $V = 50$  m/s,  $T = 120$  s and  $N = 60$ . The channel noise is assumed to be -110 dBm. The transmit power of each BS is set to 0.1W, while the transmit power of UAV is 0.05 W. The rate thresholds of the BS-served devices and UAV-served devices are set to be 1.5 bit/s/Hz and 0.5 bit/s/Hz, respectively. Due to the much better channel conditions in the upper left cell than the others, we can see that the blue curve tends to move towards the upper left cell in order to guarantee the QoS of devices in the other two cells. In addition, when the 2nd device served by the upper right BS moves towards UAV, the red curve will move even closer to the upper left cell than the blue one, which means that the UAV will fly away from this device to avoid generating strong interference to it. Thus, the sum rate of the UAV-served devices can be

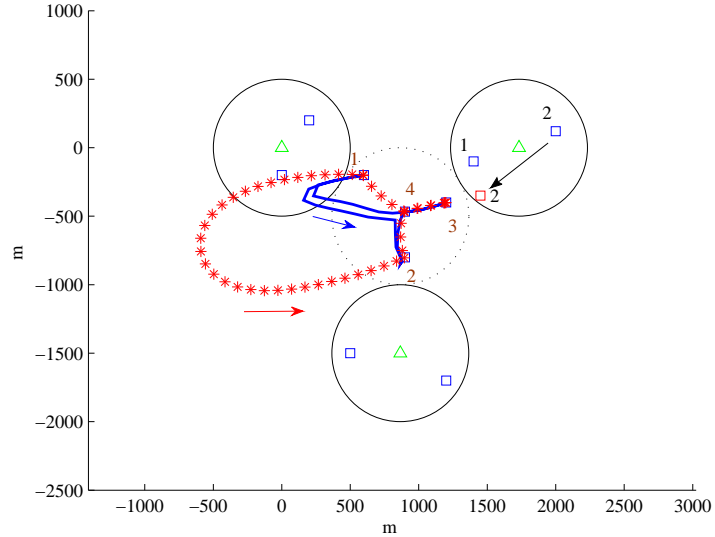


Fig. 2. Performance comparison of joint optimization of UAV scheduling and trajectory, when a specific user served by the 2nd BS moves towards the UAV.

optimized with the QoS of both the BS-served devices and the UAV-served devices guaranteed, through proper managing the scheduling and trajectory of the UAV.

## TRANSCEIVER DESIGN AND MULTI-HOP D2D ESTABLISHMENT

In Scenario 2 of Fig. 1, all the ground BSs have been damaged due to disasters, and a large-scale UAV with multiple antennas can be deployed as flying BS to provide wireless service, as shown in Fig. 3. Thus, the UAV transceiver can be carefully designed to guarantee the reliability in the downlink and uplink. In addition, the coverage area of UAV is limited due to its battery constraint, and thus, multi-hop D2D links can be established to extend its coverage.

### *Transceiver Design*

In a circular disaster area with radius  $R_1$  as shown in Fig. 3, all the BSs have been destroyed. Thus, a large-scale UAV equipped with  $M$  antennas is deployed at the center with altitude  $H$ , to provide wireless coverage to  $K$  ground single-antenna devices. According to the maximum angle between UAV and device, we can calculate the maximum transmission distance between



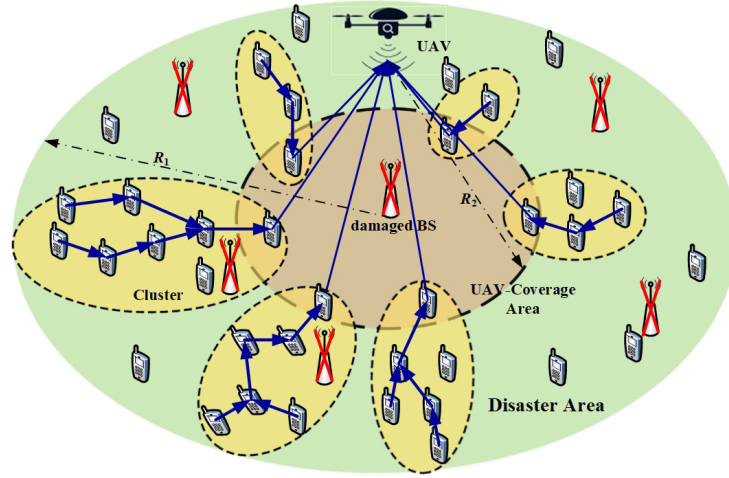


Fig. 3. Demonstration of wireless coverage via a large-scale UAV with ground multi-hop D2D links.

the UAV and a specific ground device  $R_2$  as  $H/\cos\phi$ . Due to the limited transmit power of UAV, the QoS of edge devices is difficult to guarantee. Thus, the multiple antennas at UAV should be fully exploited to achieve reliable transmission.

In the uplink, many devices may want to connect to the UAV simultaneously, and the decoding vectors at the UAV for each device should be carefully designed. In addition, the transmit power of all the devices is assumed to be equal, because the global channel state information (CSI) is difficult to obtain at each node without ground BSs for any optimization. Thus, we can maximize the throughput of all the devices by jointly optimizing the unit decoding vectors, with a constraint on the rate of each device. Although this optimization is non-convex, it can be transformed into a convex one, and its closed-form solution can be derived through maximizing the signal-to-interference-plus-noise ratio (SINR) of each link via its corresponding decoding vector.

On the other hand, the precoding design of UAV in the downlink is more complex. For simplicity, the power information can be integrated into the precoding vectors. Thus, we can maximize the throughput of all the devices by jointly optimizing the precoding vectors, with constraints on the rate of each device and the UAV transmission power. This optimization is non-

convex, which cannot be solved directly. First, some auxiliary variables are introduced, and the constraints can be converted into convex ones through first-order Taylor expansion approximately. Then, the objective function can also be converted to convex via second-order-cone programming (SOCP). Thus, suboptimal solutions can be obtained by solving a SOCP problem iteratively via classic optimization algorithms.

### *Multi-Hop D2D Establishment*

Although the reliability of wireless connections for ground devices can be enhanced through the transceiver design of UAV, its coverage is still limited as shown in Fig. 3, due to the constraint of transmit power. Thus, to increase UAV coverage effectively for the randomly distributed victims and rescue workers, multi-hop D2D links can be established to bridge the nodes within the direct coverage of UAV and the outside ones. In this scenario, we have to perform multi-hop D2D to extend the coverage of UAV. It can be deemed as a self-rescue behavior of the survivals, although this will cause the power consumption of their own devices.

To establish the multi-hop D2D links effectively in disasters, the number of hops should be minimized with reliable performance, due to the power limitation of each hop and the shortage in power supply. Thus, a shortest-path-routing (SPR) algorithm can be designed, in which a device can be selected as a relay node if it is closer to the destination than all the other available devices, within a coverage radius  $r$  of the current node to guarantee the reliability. In addition, the selected device should also be closer to the line from the source to destination. The SPR algorithm is a suitable scheme in establishing multi-hop D2D links with less number of hops to extend the coverage of UAV, although its performance is not the best.

Outage probability is a key measure of the reliability for the multi-hop D2D establishment, which is the probability that the received SINR  $\gamma$  falls below a predefined threshold  $\varepsilon$ . To derive the average outage probability, the successful transmission probability of a single hop is first

analyzed, based on which the single-hop outage probability can be obtained with a Poisson point process (PPP) density for the device distribution. Using single hop, the average number of hops  $J$  from the source to destination with distance  $R$  can be achieved according to the average distance of a single hop. Finally, the average outage probability of the multi-hop D2D link can be derived based on the single-hop outage probability and  $J$ .

### *Discussion*

In the above demonstration, a large-scale UAV with multiple antennas is considered. Nevertheless, the case when only single-antenna UAVs are available may happen in disasters. In this situation, non-orthogonal multiple access (NOMA) can be adopted to serve several ground devices simultaneously, when successive interference cancellation can be performed at the 5G-enabled devices or UAV in the downlink and uplink, respectively. To enhance the reliability of the NOMA-based UAV transmission, the hovering position of UAV and power allocation for each device can be jointly optimized.

### *Simulation Results*

The performance of downlink is analyzed in Fig. 4, in which  $K = 6$ ,  $H = 50$  m,  $R_1 = 200$  m,  $r = 10$  m and  $\phi = 60^\circ$ . First, the sum rate of UAV downlink is compared for different signal-to-noise ratio (SNR). The precoding vectors with randomly generated complex Gaussian entries are also compared. From the results, we can see that reliable UAV downlink transmission can be achieved with  $M \geq K = 6$  antennas equipped at the UAV. When  $M < K$ , the zero-forcing scheme cannot be solved, and the performance of the optimal precoding scheme will also degrade severely. In addition, the sum rate increases with antennas at UAV or and decreases with channel noise  $\sigma^2$ . Then, the overall outage probability of the link from the UAV to the destination in the multi-hop D2D is compared for different transmit power of each device, in

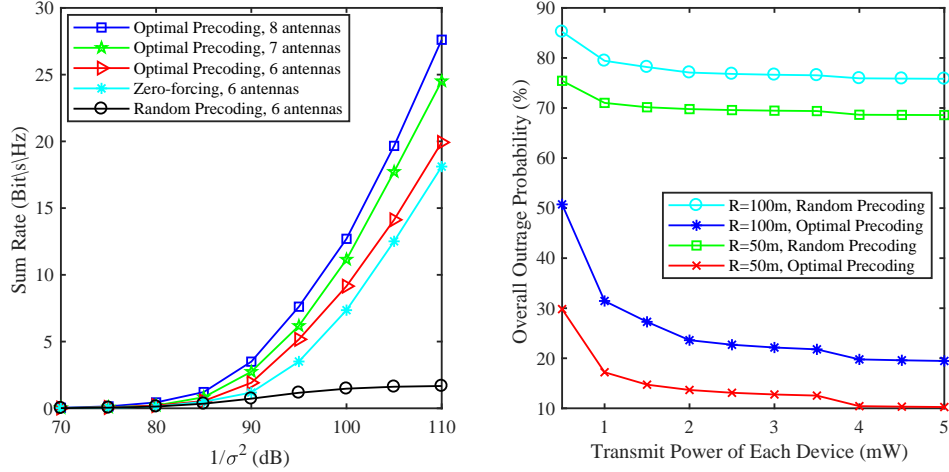


Fig. 4. (a) Downlink sum rate comparison of the optimal precoding design, zero-forcing and random precoding design. (b) Overall outage probability comparison of UAV and multi-hop D2D downlink.

which the transmit power of UAV is 10 mW, the number of available devices for establishing the multi-hop link is 100, the PPP density is 0.001,  $\sigma^2 = -110$  dBm,  $M = 6$  and  $\varepsilon = -6$  dB. From the results, we can see that reliable transmission from UAV to the devices out of its direct scope can be guaranteed with low outage probability through the proposed SPR algorithm and precoding design. In addition, the outage probability will become higher with larger distance between the source and destination of the multi-hop link.

## MULTI-HOP UAV RELAYING

Although wireless coverage can be achieved in disasters via UAVs as shown in Scenarios 1 and 2 of Fig. 1, the information bridge should be built between these UAVs and the outside emergency communication vehicles or core networks for these two scenarios effectively. To realize information exchange between disaster areas and outside, multi-hop UAV relaying can be established to overcome the space and environment limitations as shown in Fig. 5. In this section, the optimal hovering positions of UAV relaying systems are discussed [11].

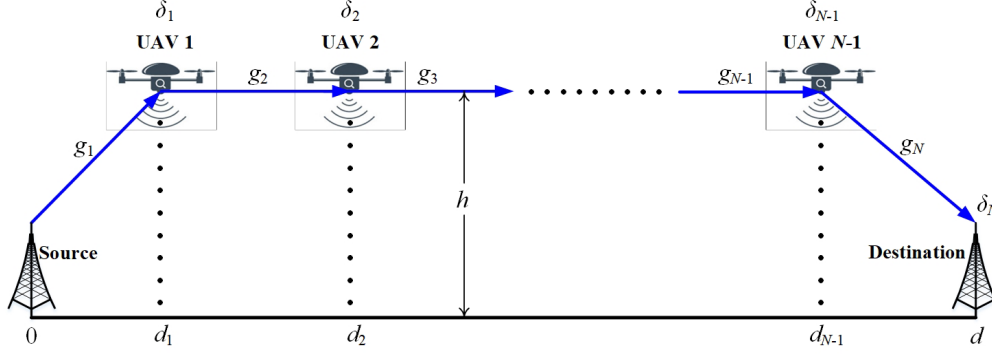


Fig. 5. Diagram for the multi-hop UAV relaying system.

### Optimal Hovering of UAV Relays

In a multi-hop UAV relaying system as shown in Fig. 5, the source and destination nodes are linked by  $N - 1$  UAV relays deployed in the same horizontal line at the same altitude  $h$ , according to which the number of UAVs can be minimized. The distance between the source and destination is  $d$ , and the  $n$ th UAV is  $d_n$  away from the source,  $n = 1, 2, \dots, N - 1$ .  $g_i$  is the fading coefficient of the  $i$ th hop, which includes both path loss and channel fading. Denote the air-to-air and air-to-ground path-loss models as  $\beta_1 r^{\alpha_1}$  and  $\beta_2 r^{\alpha_2}$ , respectively. In the existing literature, there are plenty of works focusing on the path loss for UAV communications, and in this article, we adopt the one in [7] to set the parameters as  $\alpha_1 = 2.05$ ,  $\alpha_2 = 2.32$ ,  $\beta_1 = \beta_2 = (4\pi f/c)^2$ , which were obtained through practical measurements. For the channel fading, we adopt the Nakagami  $m$  distribution.

Define the received SNR of the  $i$ th hop of the relaying system as  $\delta_i$ . When the amplify-and-forward (AF) relaying protocol is adopted, the received SNR at the destination of multi-hop relaying can be expressed as

$$\delta_{ee1} = \left( \prod_{i=1}^N (1 + 1/\delta_i) - 1 \right)^{-1}. \quad (1)$$

When the decode-and-forward (DF) relaying protocol is exploited, the received SNR at the

destination can be expressed as

$$\delta_{ee2} = \min\{\delta_1, \delta_2, \dots, \delta_N\}. \quad (2)$$

Taking the case of maximizing  $\delta_{ee1}$  for example, first, the optimal hovering altitude can be derived by taking the first-order derivative of the objective function, which is equal to 0. Then, the optimal distances of UAV relays from the source can be derived<sup>1</sup>. The result obtained will be time-varying, which is hard to implement. To fix this problem, the instantaneous CSI is replaced by the average CSI [11]. Thus, we can replace the instantaneous value of the fading coefficient  $g_i$  with its average value to make it more practical.

### *Discussion*

In the above demonstration, the optimal hovering positions of UAV relaying are derived with  $\delta_{ee1}$  when AF is adopted. For  $\delta_{ee2}$  with DF, the solutions of optimal positions are quite different, although they can be derived similarly as the case of  $\delta_{ee1}$ , which will not be presented here for simplicity [11]. In addition, when other types of channel modelling are considered, the results of the relaying optimization will also change obviously. For example, the optimal altitude  $\hat{h}$  will not always remain zero when other channel models are adopted. Moreover, the information exchange between disaster area and outside can also be achieved by multiple dual-hop UAV relaying, as indicated in [11]. When we aim to cover a long distance in disasters, the multi-hop single-link UAV relaying can achieve much better performance; on the other hand, if we aim to enhance the reliability of relaying within a shorter distance, the multiple dual-hop UAV relaying is more suitable [11].

### *Simulation Results*

In the simulation, the outage probability and the bit error rate (BER) for the derived optimal positions of multi-hop UAV relaying are compared in Fig. 6 for different  $d$ , in multi-hop UAV

<sup>1</sup>The derivation of the optimal distances can be referred to [11].

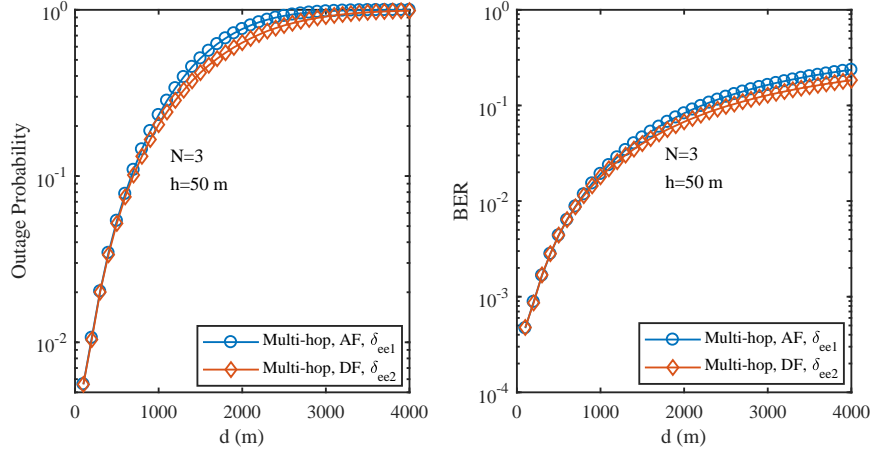


Fig. 6. Outage probability and BER comparison with different  $d$  when  $N = 3$  and  $h = 50$  m.

relaying with end-to-end SNR equal to  $\delta_{ee1}$  and  $\delta_{ee2}$ , respectively.  $f = 2$  GHz, the transmit power of each hop is 10 dBm, noise power at each UAV relay is -100 dBm, and the average value of  $|g_i|^2$  is 1.  $m$  is set to 1 for the Nakagami fading. From the results, we can see that the outage probability and BER both become higher with larger distance, due to the more severe path loss. In addition, we can see that both the outage and BER performance of the DF protocol is better than that of the AF protocol, although the computational complexity of DF is higher.

#### OPEN RESEARCH ISSUES AND CHALLENGES

Although some fundamental works have been conducted on the UAV-assisted emergency networks in disasters in this article, there still remain some open research issues and challenges to be addressed in the future.

**Multi-UAV Trajectory:** In this paper, only a single UAV is considered to provide wireless connections for the ground devices with surviving BSs. Nevertheless, in a larger disaster area with more devices to be served, multiple UAVs are needed. This will complicate trajectory optimization of all these UAVs, considering its influence on the ground BS-served devices. In the future, intelligent distributed algorithms for the trajectory and scheduling optimization for UAVs should be developed.

**Interference Management:** In this paper, a large-scale UAV is deployed to provide wireless service to ground devices with the help of multi-hop D2D links. Nevertheless, when more UAVs are deployed to provide service for a much larger area, the interference will appear among the devices served by different UAVs. In addition, if there still exist some Ad Hoc networks within the coverage area of UAV, the interference between them should be fully considered. Thus, interference management is a key challenge for UAV-assisted emergency networks in disasters.

**Channel Modelling:** The most distinctive characteristic of UAV communication is the channel modelling, especially between the UAV and ground. In this paper, the channel model in [7] is adopted to analyze the positions of multi-hop UAV relaying. Nevertheless, in practical systems, the channel models for UAV communication are quite different from case to case, and some other models can also be suitable to be used. Thus, the optimal hovering positioning of multi-hop UAV relaying should be further analyzed with other feasible channel models.

**Energy Supply:** Energy supply always remains a key challenge for UAV communication due to the battery limitation, especially for the disasters, in which it is difficult to provide stable energy supply. This challenge can be solved by energy harvesting, such as solar energy, however, it will become invalid during night time. Thus, more energy-efficient systems should be designed to prolong the operational time for UAVs.

## CONCLUSIONS

UAV can be utilized to establish emergency wireless networks to overcome the space and environment limitations in disasters, due to its flexibility and mobility. In this article, a unified framework of UAV-assisted emergency networks in disasters has been established. First, with the surviving BSs considered, the trajectory and scheduling of UAV have been jointly optimized to provide wireless connections for the ground devices. Then, the transceiver design of UAV and ground multi-hop D2D establishment have been studied, to extend the coverage scope of



the UAV BS. Furthermore, a multi-hop relaying scheme has been examined to exchange the information between disaster area and outside, in which the hovering positions of UAVs were optimized. Finally, some open research issues and challenges in the UAV-assisted emergency networks have been discussed.

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