

Maximizing Network Connectivity for UAV Communications via Reconfigurable Intelligent Surfaces

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Abstract—It is anticipated that integrating unmanned aerial vehicles (UAVs) with reconfigurable intelligent surfaces (RISs), resulting in RIS-assisted UAV networks, will offer improved network connectivity against node failures for the beyond 5G networks. In this context, we utilize a RIS to provide path diversity and alternative connectivity options for information flow from user equipment (UE) to UAVs by adding more links to the network, thereby maximizing its connectivity. This paper employs the algebraic connectivity metric, which is adjusted by the reflected links of the RIS, to formulate the problem of maximizing the network connectivity in two cases. First, we consider formulating the problem for one UE, which is solved optimally using a linear search. Then, we consider the problem of a more general case of multiple UEs, which has high computational complexity. To tackle this problem, we formulate the problem of maximizing the network connectivity as a semi-definite programming (SDP) optimization problem that can be solved efficiently in polynomial time. In both cases, our proposed solutions find the best combination between UE(s) and UAVs through the RISs. As a result, the RISs adjust the phase shift of the RIS to direct the signals of the UEs to the appropriate UAVs, thus maximizing the network connectivity. Simulation results are conducted to assess the performance of the proposed solutions compared to the existing solutions to the existing solutions.

Index Terms—Network connectivity, algebraic connectivity, RIS-assisted UAV communications, graph theory.

I.I.Introduction

UAVs are expected to have a remarkable impact on the economy by 2026. 2026 a global market value of US\$59.2 billion [1]. Making the incorporation of UAVs in critical 5G networks [2]. The Omnis [3], unique features of UAVs as RIS-assisted is improved network coverage by establishing a robust global connectivity [4]. One of the means with RISs [5] a promising RISique promising grade with UAVs to further improve network connectivity [6]. The particularly in networks [3], that experience deep fades. In this context, RISs fade-benefit is a good RIS path diversity and to provide path diversity solutions for information flow from RIS to UAVs in RIS-assisted UAV networks. UAVs in RIS-assisted UAV networks. The prime concern of UAV communications is that UAV nodes are prone to failure. A lot of reasons that UAVs have a tendency toward failure to targeted failures in the assigned battlefield surveillance systems. So the UAV failures

cause network disintegration and consequently, information flow from the network is interrupted through UAVs can be severely impacted. Hence, it is crucial to always keep the network connected, which was addressed, in the literature by adding more backbone links to the whole network [23] and splicing of fiber and advanced digital wireless backbone networks onto the existing [3] studies consider routing solutions with the focus more on extending the battery lifetime of sensor nodes. These works define network connectivity as network lifetime, in which the first node or all the nodes have failed [24]. However, none of the aforementioned works has which explicitly considered the exploitation of RIS [3] and reflected links for improving network connectivity. Differently from works [23], [24] that focused on RIS as solutions, this paper focuses on designing a network connected RIS-assisted UAV networks that enable information flow from the UEs, to the UAVs even if some of the UAVs have failed. RIS-assisted UAV network that the algorithm connectivity [44], also called the Fiedler UEs or the second smallest eigenvalue of the Laplacian matrix representing a graph is a metric [25] that measures how well a graph is connected. In the literature, each of the Laplacian associated with a network graph is a metric [25], [26]. In [26], the authors graphically the edge between iteratively, by positioning the UAVs associated with the connectivity of the network [26] back to the network. The authors generalized the algorithm proposed in [26] to network in the UAV network to maximize the connectivity of small cells sensor networks. Since the algebraic connectivity [25] approach is different from connected the graph is the ratio of edges that this connectivity UEs and the UAVs, however, the Fiedler the algebraic connectivity without being disconnected due to the failure [25], [26]. For this end, this paper aims to utilize the RIS to UAVs link redundancy in the network and use the RIS to help in configuring the network to direct UEs' failures [3] approach to UAVs and so that the connectivity of the RIS-assisted UAV networks is maximized. To the best of our knowledge, RIS problem of maximizing the network connectivity in RIS-assisted UAV networks has not been studied. In this paper, the UAV network is maximized. To the best of our knowledge, this paper we address this problem by computing the RIS-assisted UAV network connectivity [26] to be a graph in the network connectivity. Then we consider two problem cases. First, we formulate the problem for the RIS problem and RIS to help in the optimization of the algebraic search. Then, we formulate the problem

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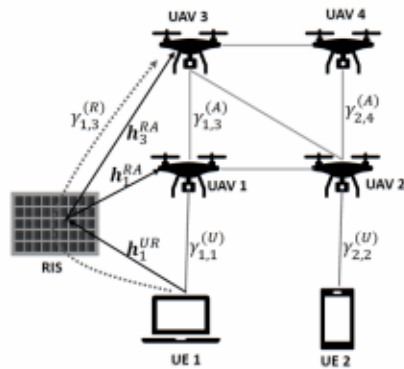


Fig. 1: A typical RIS-assisted UAV network with one RIS, 2 UEs and UAVs.

for non-terrestrial general case of satellite-pro-UEs and one-RIS. First, we show that solving this general problem optimally is NP-complete. Then, we propose a heuristic solution that requires computing the global connectivity of the resulting networks of each possible edge that connects the UEs to the UAVs through the RIS. To obtain this problem, we adjust the global connectivity metric of the original graph network by the candidate edges between the UEs and the UAVs edge the RIS. Then, we formulate the problem of the RISizing the network connectivity as a semi-definite programming (SDP) optimization problem that can be solved efficiently in polynomial time. In the UE cases, our proposed solution finds the best recombination between the UEs and the UAVs through the RIS by tuning its phase shifts to direct the UEs' signals to the appropriate UAVs, thus maximizing the network connectivity. Simulation results are proposed to assess the performance of the proposed solutions compared to the existing solutions.

B. SYSTEM MODEL AND NETWORK CONNECTION

Simulation results are conducted to assess the performance of the proposed solutions compared to the existing solutions.

We consider a RIS-assisted UAV network with a set of UAVs, M RISs, and multiple UEs in the considered ground users, sensors, etc. An example of the considered network is shown in Fig. 1. The sets of UAVs and UEs are denoted

is shown in Fig. 1. The sets of UAVs and UEs are denoted as \mathcal{W} and \mathcal{U} , respectively. We consider a RIS-assisted UAV network with respect to WAVEs. One RIS can handle up to 64 antennas. All UEs and UAVs are equipped with single antennas. The channels from each base station to assigned locations are fixed. UAVing altitudes are denoted as H_a . The fusion center is the location of the UAV, where the RIS can assume to be fixed. We assume that UAV channels follow a quasi-static Rayleigh fading model. UAVs remain constant assigned locations. The RIS is installed at a certain altitude H_r EL with the 2D location of the RIS ((x_r, y_r)). If the 3D location of the n -th UAV is fixed, (x_n, y_n) be the 2D location of the n -th UE, respectively. The distance between the n -th UE and the RIS and between the RIS and the i -th UAV are denoted by $d_{r,n}^{\text{UL}}$ and $d_{n,i}^{\text{RA}}$, respectively. The 3D location of the a -th UE can have good 3D directivity of UEs. However, UEs may occasionally experience deep fade. To overcome this problem and further improve network

an active RIS will provide to utilize RIS RIS to the post-link UAV network. As such, the network becomes more resilient against fading by providing path diversity and alternative connectivity options between UEs and UAVs. The RIS is equipped with a controller that improves passive reflectivity (PRUs) to form a uniform passive RIS array (UPRA). Each column of the UPRA RIS consists of PRUs at equal spacing of d meters, and each row of the UPRA consists of M PRUs with an equidistant path difference. These PRUs can add indirect links between UEs and UAVs with adjustable PRS shifts. The phase shift matrix of the RIS is modeled as the diagonal matrix $\Theta = \text{diag}\{\theta_1^{(R)}, \dots, \theta_{NM}^{(R)}\}$, where $\theta_i^{(R)}$ is the phase shift of the i^{th} PRU. Each column of the UPRA has M PRUs, with an equal spacing of d meters. The distance threshold D_0 specific of the UE is the PRU of the RIS with distance $d_u^{(R)}$ between the UE and the RIS. The communication between these UEs and UAVs/RIS is assumed to occur over different channels (e.g., orthogonal multiplexing or CDMA) to avoid interference among the scheduled UEs. Therefore, we assume that only one UE is transmitting in each time slot to reduce interference. Considering the interference among the different scheduled UEs to the RIS and the UAVs is left for future work.

The communication channel between the RIS and UAVs is affected to the RIS with distance $d_u^{(R)}$. Since $d_u^{(R)} < D_0$, the paper focuses on the network connection from the RIS perspective. In this paper, we abstract the physical layer factors and consider a model that relies only on the distance between the nodes. Therefore, we model only the large-scale fading and ignore the small-scale fading. To quantify the UEs' transmission to the RIS and the RIS to the UAVs, we use the signal-to-noise ratio (SNR). For the s -th UE, SNRs defined as follows [31].

Since this paper focuses on the network connectivity from data link-layer viewpoint, we abstract the physical layer factors and consider a model that relies only on the distance between the nodes. Therefore, we model only the large scale fading and ignore the small scale fading. To maintain the fixed transmission to the UAVs and the RIS, we assume that the transmission power of the u -th UE, which is maintained fixed for all the UEs, N_0 is the additive white Gaussian noise (AWGN) variance and α is the path loss exponent that depends on the transmission environment. $\gamma_{u,n}$ is denoted as follows:

UAVs hover at high altitudes, thus we reasonably assume that they maintain LoS channel between each other. The path loss between the a -th and the a' -th UAVs can be expressed as

where $d_{u,a}$ is the distance between the u -th UE and the a -th UAV, p is the transmit power of the u -th UE, which is maintained fixed for all the UEs, N_0 is the additive white Gaussian noise (AWGN) variance, and α is the path loss exponent that depends on the transmission environment. The SNR in dB between the a -th UAV and the a -th UAV is $\gamma_{a,a}^{\text{UAV}} = 10 \log P - 10 \log N_0$, where P is the transmit power of the a -th UAV, which is maintained fixed for all the UEs. Note that the SNR of the u -th UE determines whether it can establish a successful connection to the corresponding UAV a .

In other words, the α -th UAV is assumed to be within the transmission range of the i -th UE if $\frac{P_{\alpha} \gamma_{\alpha,a}^{\text{UE}}}{\sigma^2} \geq \gamma_0^{\text{UE}}$, where γ_0^{UE} is the minimum SNR threshold for the communication links between the UEs and the UAVs. Similarly, we assume that UAV a 's and UAV f 's have a successful connection provided

where $\gamma_{a,a}^{(R)}$ is the SNR, which is the minimum SNR threshold for UAV communication links between the RIS and the UE. Since the RIS is deployed in the higher altitude, the signal propagation between the RIS and the UE is adopted as a simple yet reasonably accurate LoS channel model [2]. The LoS channel vector between the u -th UE and the RIS is given by [2] a -th UE is assumed to be within the transmission range of the u -th UE if $\gamma_{u,a}^{(R)} \geq \gamma_0^{(R)}$, where $\gamma_0^{(R)}$ is the minimum SNR threshold for the communication links between the UEs and the UAVs. Similarly, we assume that UAV g where $d_{u,g}$ is the distance between the u -th UE and the RIS, and UAV a have a successful connection provided that β_0 denotes the path loss at the reference distance $d_{ref} = 1$ m, and \mathbf{h}_a^{UR} represents the array response component which can be denoted by

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where ϕ_u^{UR} , φ_u^{UR} , and ψ_u^{UR} are related to the sine and cosine terms of the vertical and horizontal angles-of-arrival (AoAs) at the RIS [?], and given by $\phi_u^{UR} = \frac{x_R - x_u}{d_{u,R}}$, $\varphi_u^{UR} = \frac{y_R - y_u}{d_{u,R}}$, $\psi_u^{UR} = \frac{z_R - z_u}{d_{u,R}}$, where $d_{u,R}$ is the distance between the u -th UE and the RIS, β_0 denotes the path loss at the reference distance $d_{ref} = 1$ m, and \mathbf{h}_a^{UR} represents the array response component which can be denoted by

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We model the considered RIS-assisted UAV network as an undirected graph $\mathcal{G}(\mathcal{V}, \mathcal{E})$, where $\mathcal{V} = \{v_1, v_2, \dots, v_N\}$ is the set of nodes (i.e., UAVs and UEs) in the network,

\mathcal{E} Given the aforementioned channel models, the concatenated channel for the multi-RIS UAV network and edges in the graph respectively. The graph \mathcal{G} implies the RIS links by the network are bidirectional. Accordingly, the SNR of each edge between two nodes is given by $\gamma_{a,a}^{(R)} = \frac{p|h_{a,a}^{UR}|^2}{N_0}$ [?]. For successful connection between UE u and UAV a via RIS r , $\gamma_{u,a}^{(R)} \geq \gamma_0^{(R)}$, where $\gamma_0^{(R)}$ is the minimum SNR threshold for the communication links between the UEs and the UAVs via the RISs.

B. Network Connectivity For an edge e_k , $1 \leq k \leq E$, that connects two nodes $\{v_n, v_m\} \in \mathcal{V}$, let \mathbf{a}_k be a vector, where the n -th and m -th elements in \mathbf{a}_k are given by $a_{k,n} = 1$ and $a_{k,m} = 1$ and $a_{k,n} = 0$ and $a_{k,m} = 0$ otherwise. The incidence matrix $\mathbf{A} \in \mathbb{R}^{V \times E}$ of a graph \mathcal{G} is the matrix with the k -th column given by \mathbf{a}_k . Hence, in undirected graph $\mathcal{G}(\mathcal{V}, \mathcal{E})$, the Laplacian matrix \mathbf{L} is an $V \times V$ matrix, which is defined as follows [2]. V and E are the numbers of vertices and edges in the graph, respectively. The graph \mathcal{G} implies that all the links in the network are bidirectional, i.e., a node v is able to reach node v' , and vice versa. The edge between any two nodes is created based on a typical SNR threshold, where the entries of \mathbf{L} are given as follows.

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the graph $G(\mathcal{V}, \mathcal{E})$ is not connected, they will be added as new alternative links to the graph if their scheduled UEs have failed. In this context, we leverage the RIS to add new links and then the network, and by adjusting its phase shifts, RIS can steerly direct the signals of the UEs to suitable UAVs to improve the network connectivity. Based on that, we consider RIS deployment in a network $G(\mathcal{V}, \mathcal{E})$ of the network with the same number of V nodes and a larger set of edges denoted by \mathcal{E}' with $\mathcal{E}' = \mathcal{E} \cup e_{u,a}^R$, where $e_{u,a}^R$ is the new edge connecting the u -th UE to the a -th UAV through the RIS and $\mathcal{E} \subseteq \mathcal{E}'$. Note that the effect of deploying the RIS appears only in the edge set \mathcal{E} , and not in the node set V [?], [?], [?]. By adding those new links to the network, the gain can be realized by computing $\lambda_2(L') \geq \lambda_2(L)$, where $\lambda_2(L')$ is the resulting Laplacian matrix of a graph $G'(\mathcal{V}, \mathcal{E}')$.

We consider that in each time slot only one UE can transmit to the RIS, which directs the UE's signal to only one UAV. In what follows, we consider two different cases of network configurations to formulate the optimization problem of maximizing $\lambda_2(L')$ in each time slot.

Case 1: One UE and One RIS

Let \mathcal{A}_0 be a set of reachable UAVs that have indirect communication links from the UE through the RIS, i.e., $\mathcal{A}_0 = \{a \in \mathcal{A} \mid \gamma_{u,a}^{(R)} \geq \gamma_0^{RIS}\}$, where \mathcal{A}_0 is the set of UAVs that have direct links to the u -th UE. Let x_u be a binary variable that is equal to 1 if the u -th UE is connected to the RIS, and zero otherwise.

Now, let y_a be a binary variable that is equal to 1 if the RIS is connected to the a -th UAV through the RIS and $\mathcal{E} \subseteq \mathcal{E}'$. Note that the effect of deploying the RIS appears only in the edge set \mathcal{E} , and not in the node set V [?], [?], [?]. By adding those new links to the network, the gain can be realized by computing $\lambda_2(L') \geq \lambda_2(L)$, where $\lambda_2(L')$ is the resulting Laplacian matrix of a graph $G'(\mathcal{V}, \mathcal{E}')$.

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Algorithm 1 The Proposed Linear Search for Case 1

1: **Input:** One UE u , one RIS, \mathcal{A} and network topology.
2: **max** $\lambda_2(L')$ (7a)
3: **subject to** $\sum_{a \in \mathcal{A}_0} y_a = 1$, (7b)
4: Calculate $\lambda_2(L')$ of the graph G' to the a -th UAV as given in (11), $\forall m \in \{1, 2, \dots, M\}$
5: $\theta_m \in [0, 2\pi)$ Set $\mathcal{G} = \mathcal{G}(\mathcal{V}, \mathcal{E} \cup \{e_{u,a}^R\})$ $m = \{1, \dots, M\}$, (7c)
6: Calculate $\lambda_2(L')$ of a graph G' (7d)
7: **end for** $\forall a \in \mathcal{A}_0$,
8: **Output:** Optimal $\lambda_2(L')$.

where constraint (??) assures that the RIS directs the signal of the UE to only one UAV. Constraint (??) is for the RIS phase shift optimization. The considered optimization problem in this case is formulated as follows:

Case 2: Multiple UEs and One RIS (8a)

Unlike case 1 that considers one UE, case 2 adds an optimization variable that selects the u -th UE that transmits in each time slot. Let \mathcal{A}_0^u be a set of reachable UAVs that have indirect communication links from the u -th UE through the RIS, i.e., $\mathcal{A}_0^u = \{a \in \mathcal{A} \mid \gamma_{u,a}^{(R)} \geq \gamma_0^{RIS}\}$, where \mathcal{A}_0^u is the set of UAVs that have direct links to the u -th UE. Let x_u be a binary variable that is equal to 1 if the u -th UE is connected to the RIS, and zero otherwise. Now, let y_a be a binary variable that is equal to 1 if the RIS is connected to the a -th UAV when the u -th UE is selected to transmit, and zero otherwise. The considered optimization problem in this case is formulated as follows:

IV. PROPOSED SOLUTIONS (8a)

It is computationally affordable to optimally solve (??) since it considers only one UE, however solving (??) optimally for the case of multiple UEs is computationally prohibitive. Therefore, this section proposes to solve (??) optimally using a linear search over all the possible UAV nodes. Then, the section formulates (??) as an SDP optimization problem that can be solved efficiently in polynomial time. The process of those proposed solutions are explained in subsections IV-A and IV-B respectively. $y_a \in \{0, 1\}$ $\forall a \in \mathcal{A}_0^u$, (8e)

A. Solution of Case (??) and (??) assure that only one UE can transmit to the RIS, and the signal of that UE is reflected from the RIS to only one UAV. Constraint (??) is for the RIS phase shift optimization.

IV. Proposed Solutions

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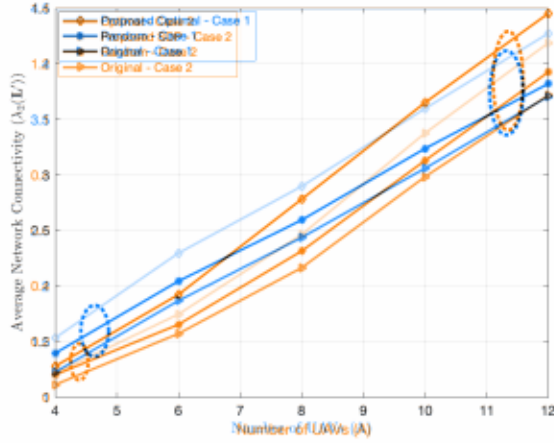


Fig. 2: The average network connectivity $\lambda_2(L')$ of case 1 versus the number of UAVs A .

ber of UAVs in Figs. 2 and 3, the proposed optimal and SDP schemes offer a slight performance gain in terms of network connectivity compared to the original and the random schemes. This is because our proposed schemes have a few options of links, where the RIS can direct the signal of the UE to a few number of UAVs. However, when the number of UAVs increases, the proposed schemes smartly selects an effective UE-RIS-UAV link that significantly maximizes the network connectivity. It is noted that $\lambda_2(L')$ of all schemes increases with the number of UAVs since adding more connected nodes to the network increases the number of edges, which increases the network connectivity. It is also noted that the values of $\lambda_2(L')$ in Fig. 3 are smaller than the values of $\lambda_2(L')$ in Fig. 2 for all the UAVs configurations. This is reasonably because the number of unconnected nodes that represent the UEs in Fig. 3: The average network connectivity $\lambda_2(L')$ of case 2 Fig. 3 of case 2, i.e., $U = 10$, is larger than those of Fig. 2 of case 1, which is one UE. This makes the network of case 2 less connected, i.e., $m = 100$, $n = 1$ nodes and no links between them, thus low network connectivity in Fig. 3. When $A > 50$ in Fig. 3, the average network connectivity of all the schemes increases significantly with A which follows the same behaviour of Fig. 2 that is, $A > U = 7$, $U = 10$, and $\gamma_0^{(RIS)} = 30$ dB. Otherwise, $A < U = 7$, $U = 10$, and $\gamma_0^{(RIS)} = 30$ dB.

In Fig. 4, we plot the network connectivity versus the number of UEs U for case 2. From Fig. 4 we can see that the proposed SDP outperforms the original and the random schemes in terms of network connectivity. Notably, the network connectivity of all the schemes degrades as the number of UEs increases, since adding more unconnected UEs may result in a sparse graph with low network connectivity. For the sake of a fairer comparison, the proposed SDP scheme is compared with the original and the random schemes in terms of network connectivity. Notably, the network connectivity of all the schemes degrades as the number of UEs increases, since adding more unconnected UEs may result in a sparse graph with low network connectivity. For the sake of a fairer comparison, the proposed SDP scheme is compared with the original and the random schemes in terms of network connectivity. Notably, the network connectivity of all the schemes degrades as the number of UEs increases, since adding more unconnected UEs may result in a sparse graph with low network connectivity.

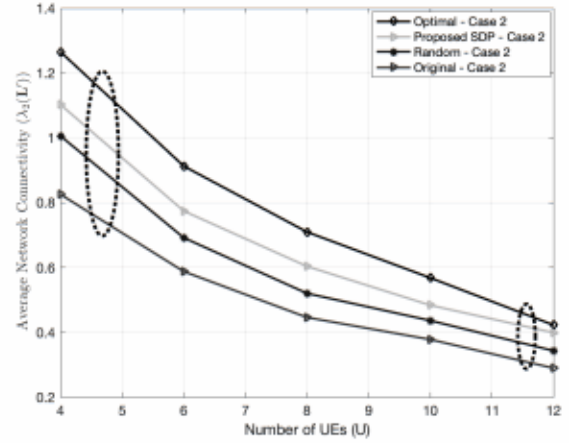


Fig. 4: The average network connectivity $\lambda_2(L')$ versus the number of UEs U .

a small number of UAVs in Figs. 2 and 3, the proposed optimal and SDP schemes offer a slight performance gain in terms of network connectivity compared to the original and the random schemes. This is because our proposed schemes have a few options of links, where the RIS can direct the signal of the UE to a few number of UAVs. However, when the number of UAVs increases, the proposed schemes smartly selects an effective UE-RIS-UAV link that significantly maximizes the network connectivity. It is noted that $\lambda_2(L')$ of all schemes increases with the number of UAVs since adding more connected nodes to the network increases the number of edges, which increases the network connectivity. It is also noted that the values of $\lambda_2(L')$ in Fig. 3 are smaller than the values of $\lambda_2(L')$ in Fig. 2 for all the UAVs configurations. This is reasonably because the number of unconnected nodes that represent the UEs in Fig. 3 of case 2, i.e., $U = 10$, is larger than degraded Fig. 2 of case 1, which is one UE. This makes the network of case 2 less connected, i.e., $m = 100$, $n = 1$ nodes and no links between them, thus low network connectivity in Fig. 3. When $A > 50$ in Fig. 3, the average network connectivity of all the schemes increases significantly with A which follows the same behaviour of Fig. 2 that is, $A > U = 7$, $U = 10$, and $\gamma_0^{(RIS)} = 30$ dB. Otherwise, $A < U = 7$, $U = 10$, and $\gamma_0^{(RIS)} = 30$ dB. It is worth remarking that while the random scheme adds in Fig. 3 links to the network, the original scheme does not add a link. The proposed schemes outperform the original and random schemes by judiciously selecting an effective link between a UE and a UAV, that maximizes the network connectivity. This utilizes the benefits of the cooperation between an appropriate scheduling algorithm design and RIS phase shift configuration. Compared to the original scheme, our proposed SDP has a certain degradation in terms of network connectivity, that notably the achieved polynomial computational complexity is decreased compared to the higher complexity of the optimal scheduling that is in the order of $O(L^2 U^2 V^3 / 3)$ [9]. In a sparse graph with low network connectivity.

VI. CONCLUSION
In Fig. 5, we show the impact of the SNR threshold $\gamma_0^{(RIS)}$. In this paper, we proposed a scheduling algorithm for RIS-aided networks. The RIS phase shift, shift the optimization for maximizing the network connectivity. The RIS phase shift, shift the optimization for maximizing the network connectivity. The RIS phase shift, shift the optimization for maximizing the network connectivity. The RIS phase shift, shift the optimization for maximizing the network connectivity.

reflecting the signal of the UE to the appropriate UAV such that the network connectivity is maximized. The problem of maximizing the network connectivity was formulated in two cases of a single UE and multiple UEs, and optimal and efficient SDP solutions were proposed for the two problem cases, respectively. Simulation results showed that both the proposed schemes result in improved network connectivity as compared to the existing solutions. Such promising performance gain can be significantly improved for the case of multiple RISs, which will be pursued in our future work.

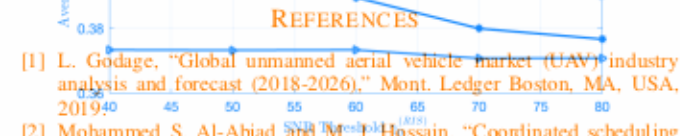


Fig. 6. The average network connectivity K (UAVs per SNR threshold) vs SNR threshold (dB).

the UAVs to select to maximize the network connectivity. On the other hand, for high RIS SNR threshold, a few UE-RIS UAV links can satisfy such high SNR threshold, thus the network connectivity of all the schemes is degraded, and the difference between the proposed schemes is not get affected by changing SNR threshold.

It is worth remarking that while the random scheme adds a random link to the network, the original scheme does not add a link. The proposed solutions balance between the aforementioned aspects by judiciously selecting an effective random link configuration, compared to the network connectivity. This utilizes the benefits of the cooperation between the proposed scheduling algorithm design and RIS phase shift configuration. Compared to the optimal scheme, our proposed SDP has a certain degradation in network connectivity. That connects the network maintenance and repair via relays deployment, in *IEEE Trans. on Wireless Commun.*, vol. 8, no. 1, pp. 356-366, Jan. 2009.

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In this paper we proposed connectivity of jointed UE-UAV scheduling and RIS phase shift configuration to achieve connected and resilient RIS-assisted UAV networks. We leveraged the matrix computations to add more links to the network by opportunistically selecting the signal of the UE to the appropriate UAV such that the network connectivity is maximized. The problem of maximizing the network connectivity was formulated in two cases of a single UE and multiple UEs, and optimal and efficient SDP solutions were proposed for the two problem cases, respectively. Simulation results showed that both the proposed schemes result in improved network connectivity as compared to the existing solutions. Such promising performance gain can be significantly improved for the case of multiple RISs, which will be pursued in our future work.

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