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Article

Optimization and Communication in UAV Networks

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1. Introduction

With new technological advances, UAVs are becoming a reality and attract more and more attention. UAVs or drones are flying devices that can be remotely controlled or, more recently, completely autonomous. They can be used alone or as a fleet, and in a large set of applications: from rescue operations to event coverage going through servicing other networks such as sensor networks for replacing, recharging, or data offloading. They are hardware-constrained since they cannot be too heavy and rely on batteries. Depending on their use (alone or in a swarm) and the targeted applications, they must evolve differently and meet different requirements (energy preservation, delay of covering an area, coverage, limited number of devices, etc.) with limited resources (energy, speed, etc.). Yet, their use still raises a large set of new exciting challenges, in terms of trajectory optimization, positioning, when they are used alone or in cooperation, coordination, and communication when they evolve in swarm, just to name a few. This Special Issue was calling for any new original submissions that deal with UAV or UAV swarm optimization or communication aspects. Among the numerous submissions, only twelve of them have been selected after a rigorous selection process. Main themes which arise from them are : (i) ground data collection from the air, (ii), (iii) control of UAV swarm UAV-based Mobile Edge Computing and (iv) application-driven UAV based measurements.

In the following, we sum up the contributions of the papers published to this Special Issue for each category to then conclude by drawing future challenges and still open issues.

2. Ground data collection from the air and path planning

It becomes more and more common to imagine having data sensed from ground wireless sensors collected by UAV to alleviate wireless peer-to-peer communications between ground sensors and reduce their energy consumption. But such a paradigm raises a set of new challenges such as how to prioritize the sensors to visit, how to optimize the time to collect all data by visiting all devices, etc. This is an exciting optimization problem. Works [1–3] and [4] propose different approaches to address this issue with different perspectives. Different criteria are considered to plan the trajectory of the UAV and different functions are optimized.

[1] proposes to visit the nodes in a given order and for a variable time that depend on a node priority while in [2], the authors aim to maximize the data collection utility by jointly optimizing the communication scheduling and trajectory of each UAV. The data collection utility is determined by the amount and value of the collected data and novel trajectory planning algorithm is design to maximize it. The author of [3] focus on the problem of minimizing the mission completion time (flying time and hovering time) for a multi-UAV system in a monitoring scenario while ensuring that the information of each sensor is collected. As for [4], the authors aim to improve the secrecy performance of cellular-enabled unmanned aerial vehicle communication networks through an aerial cooperative jamming scheme.

36 **3. Control of UAV swarm**

37 When more than a drone is required, a swarm of UAV can be deployed. Although bringing more
38 performances in terms of coverage and connectivity, new optimization challenges pop up due to the
39 difficulty to control and scale such swarms both in a distributed or centralized way. [5–8] tackle these
40 numerous challenges going from connectivity maintenance to swarm control.

41 [5] studies the different factors that may impact the accuracy and efficiency of an unmanned aerial
42 vehicle (UAV) swarm coordination. The authors propose a mathematical data model to demonstrate
43 fundamental properties of antenna arrays and study the performance of the data collection system
44 framework. Numerical examples and practical measurements are provided to demonstrate the
45 feasibility of the proposed data collection system framework using iterative-MUSIC algorithm and
46 benchmark theoretical expectations.

47 [6] deals with multi-UAV systems where the UAV autonomy is much smaller than the time
48 to complete their mission. The authors thus introduce a UAV replacement procedure as a way to
49 guarantee ground users' connectivity over time, formulating the practical UAV replacements problem
50 in moderately large multi-UAV swarms and proves it to be an NP-hard problem in which an optimal
51 solution has exponential complexity. [7] focuses on the maintenance formation with time-varying
52 shape of a swarm proposing a virtual leader approach while [8] investigates a stochastic model of the
53 UAV Swarm system with multiplicative noises.

54 **4. UAV enabled Mobile Edge Computing**

55 The potential offered by the abundance of sensors, actuators, and communications in the Internet
56 of Things (IoT) era is hindered by the limited computational capacity of local nodes. But these latter are
57 not necessarily always in capacity to offload data to an edge server. In such a case, mobile edge servers
58 can go to them thank to the deployment of UAV-assisted Multi-access Edge Computing systems, which
59 raises new challenging optimization and networking issues as addressed in [9,10].

60 [9] proposes to provide Unmanned Aerial Vehicle (UAV)-assisted Multi-access Edge Computing
61 (MEC) system based on a usage-based pricing policy for allowing the exploitation of the servers'
62 computing resources while the authors of [10] introduce the DRUID-NET perspective. aiming to adapt
63 to a rapidly varying demand by applying different tools from Automata and Graph theory, Machine
64 Learning, Modern Control Theory, and Network Theory are combined.

65 **5. Application-driven UAV based measurements**

66 In such cases, the application that has asked for UAV deployment comes with very specific
67 constraints and requirements and calls for specific optimization models. [11,12] give two of such
68 examples dedicated respectively for three-dimensional measurements and surveillance.

69 For instance, [11] aims to provide a comparative analysis of the precision of ground geodetic data
70 versus the three-dimensional measurements from unmanned aerial vehicles (UAV), while establishing
71 the impact of herbaceous vegetation on the UAV 3D model. A constraint to take into account in this
72 application is the fact that herbaceous vegetation can impede the establishment of the anthropogenic
73 roughness of the surface and deteriorates the identification of minor surface.

74 [12] focuses on UAV cooperative surveillance networks and introduces the use of complex field
75 network coding (CFNC) for this application. According to whether there is a direct communication link
76 between any source drone and the destination, the information transfer mechanism at the downlink is
77 set to one of two modes, either mixed or relay transmission, and two corresponding irregular topology
78 structures for CFNC-based networks are proposed. Theoretical analysis and simulation results with an
79 additive white Gaussian noise (AWGN) channel show that the CFNC obtains a throughput as high as
80 $1/2$ symbol per source per channel use. Results show that CFNC applied to the proposed irregular
81 structures under the two transmission modes can achieve better reliability due to full diversity gain as
82 compared to that based on the regular structure.

83 6. Conclusion

84 As you can notice, challenges in UAV networks are huge, numerous and heterogeneous. They
85 are concerning different aspects of the deployment of drones, from path trajectory to connectivity
86 maintenance going through energy management. Much of them have been addressed with
87 optimization tools but there remain a lot of open issues and research directions. Contributions
88 presented in this special issue are only a first step to pave the way towards even more exciting
89 investigations.

90

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