



Millimeter Wave Remote UAV Control and Communications for Public Safety Scenarios

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Abstract—Communication and video capture from unmanned aerial vehicles (UAVs) offer significant potential for assisting first responders in remote public safety settings. In such uses, millimeter wave (mmWave) wireless links can provide high throughput and low latency connectivity between the UAV and a remote command center. However, maintaining reliable aerial communication in the mmWave bands is challenging due to the need to support high speed beam tracking and overcome blockage. This paper provides a simulation study aimed at assessing the feasibility of public safety UAV connectivity through a 5G link at 28 GHz. Real flight motion traces are captured during maneuvers similar to those expected in public safety settings. The motions traces are then incorporated into a detailed mmWave network simulator that models the channel, blockage, beamforming and full 3GPP protocol stack. We show that 5G mmWave communications can deliver throughput up to 1 Gbps with consistent sub ms latency when the base station is located near the mission area, enabling remote offloading of the UAV control and perception algorithms.

Index Terms—UAV, mmWave, remote control, performance evaluation

I. INTRODUCTION

Unmanned Aerial Vehicles (UAVs) are being increasingly considered for use in public safety scenarios [1]–[3]. UAVs can provide valuable video and surveillance of emergency zones as well as wireless connectivity in regions where cellular infrastructure is either not available (such as wildfires in remote areas) or where that infrastructure is no longer operational. In such scenarios, high bandwidth, low-latency remote communication to the UAVs may be of value. For example, a UAV could provide wireless connectivity to a number of first responders in a disaster scenario and then backhaul that traffic to a command unit. Since the UAV would be aggregating traffic from several sources and will have to exchange uncompressed data from high bit-rate sensors, such as advanced hyper-spectral cameras and lidars, the total bandwidth requirement can be much larger than what traditional frequency bands can actually offer. More specifically, in order to push off-board the heavy computational burden of the artificial intelligence algorithms used to process the data, Gbps of uncompressed data¹ will have to be uploaded in real time to the cloud or to the edge of the network [5].

¹Lidars themselves can generate millions of points per second with over 50 channels each [4].

In addition, remote control of the drone may demand very low latency connectivity, particularly for high velocity and agile maneuvers of small scale vehicles navigating outside the line of sight [6], or scenarios with strong wind perturbations [7].

The massive bandwidth available in the millimeter wave (mmWave) bands offers the possibility of both very high throughput and low latency connectivity [8], [9]. Due to their enormous potential, communication in the mmWave bands along with other transmissions above 6 GHz have emerged as a central component of the 3GPP Fifth Generation (5G) NR standard [10]. UAV communication is also already being extensively studied for 4G LTE [11], [12] and 5G NR may offer significantly greater capabilities.

Nevertheless, there are several key challenges in mmWave communications for UAVs: due to high isotropic path loss, mmWave signals are transmitted in narrow, electrically steerable beams and reliable communication requires rapid beam tracking and steering. Such tracking can be a challenge in high velocity flight. Besides, for low height scenarios, typically in the range of 10 – 20 m, line-of-sight links may be blocked by buildings and other obstacles in the environments, leading to intermittent connectivity. The broad purpose of this paper is twofold: (1) to assess throughput and latency requirements for the offloading at the edge of the control of drones, considering a public safety setting; and (2) evaluate the feasibility of mmWave connections for this task.

To perform this evaluation, we have collected detailed motion traces for actual drone flights in settings that we expect are similar to the maneuvers in public safety settings. The motion traces include position, velocity and orientation.

We then simulate a hypothetical mmWave link in this flight using the detailed network simulator described in [13]. Importantly, the simulation includes the effects of beamforming – a key challenge in mmWave aerial communication. The full upper layer protocol stack of 5G cellular networks is also modeled. The simulation then provides throughput trends and the effect of the latency can be evaluated in the context of drone control offloading. The methodology of our approach is depicted in Fig. 1. To the best of our knowledge, this is the first comprehensive end-to-end simulation of 5G mmWave UAV communication and control.

Our initial results show that 5G mmWave communications

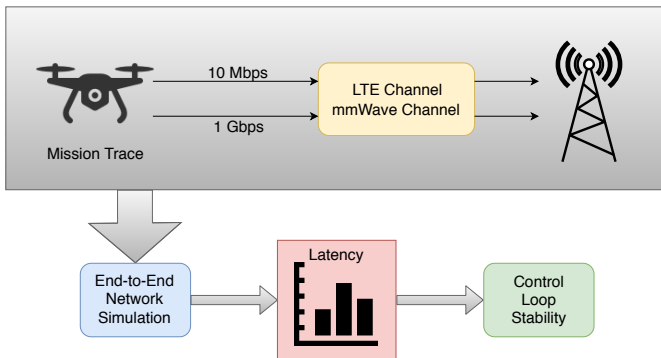


Fig. 1: Performance Evaluation Diagram.

can deliver throughput up to 1 Gbps with consistent sub ms latency when the base station is located near the mission area.

The remainder of the paper is organized as follows. We present a review of the state of the art in Sec. II, and discuss why mmWave communications can be beneficial for remote UAV control in Sec. III. Then, we describe the real flight traces and the associated missions in Sec. IV, and present our results in Sec. V. Finally, Sec. VI concludes the paper and presents possible extensions.

II. STATE OF THE ART

In this section, we discuss the advances and work done in the area of wireless communications for UAV at mmWave frequencies. An overview and presentation of the benefits of using UAV for cellular networks is discussed in [14], while [15] focuses on mmWave communications in aerial scenarios. The paper [3] explores the potentials of frequencies above 6 GHz for Public Safety Communication (PSC), and provides a specific use case for mmWave communications with a UAVs in a wildfire scenario.

The numerical characterization of the performance of airborne mmWave networks depends on the modeling of the UAV air-to-ground and air-to-air channel. [16] investigates the characteristics of the mmWave air-to-ground channels for UAV in two frequency bands (28 GHz and 60 GHz) through ray tracing and an experimental setup, showing that the received signal strength follows a two ray propagation model. Similarly, [17] analyzes the temporal and spatial characteristics of a 28 GHz air-to-ground channel through ray tracing, highlighting the impact of the scatterers on the overall channel behavior. [18] presents a study of mmWave Multiple User (MU)-Multiple Input, Multiple Output (MIMO) networks wherein atmospheric attenuation effects are modeled and single path channels are considered. Finally, channel tracking methods based on UAV movement state information and channel gain information are proposed by [19]. Utilization and study of use cases for UAVs are discussed in [20] and [21]. Both of these papers propose the use of UAVs as base stations and develop methods to optimize network coverage and spectral efficiency.

With respect to the state of the art, this paper evaluates the end-to-end performance of UAVs at mmWave frequencies for

real public safety flight mission scenarios, also considering the requirements for the offloading of the control algorithms of the UAV to the edge of the network. This scenario considers an uplink heavy connection to the UAVs, rather than the downlink heavy cases of the existing work related to the UAV deployment as base stations.

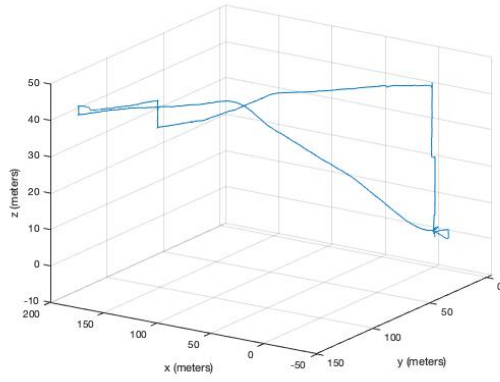
III. AUTONOMY OVER 5G

As mentioned in Sec. I, cellular or ad hoc mmWave deployments could be used to remotely control a UAV, and at the same time stream telemetry, video and user data from the UAV to a remote location. However, there exist bounds in the tolerable latency that should not be exceeded in order to maintain stability in the control of the system.

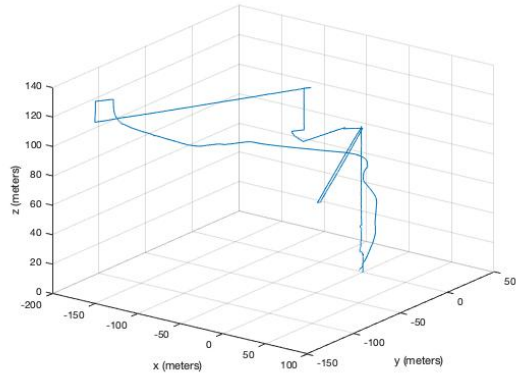
We discuss here the autonomy problem of a quadrotor vehicle in terms of control and perception, while the control loop is subject to delays in communication or computation. The main goal is to give the reader an overview about potential benefits of 5G and mmWave communications in this field.

The recent advances in computing, sensing, and perception algorithms provide the ability to run most of the autonomous components on board of the small scale aerial vehicles. Current controllers for quadrotor or, in general, Vertical Take-off and Landing (VTOL) commercial vehicles are composed of an inner loop and an outer loop. The inner loop is responsible for stabilizing the attitude dynamics of the platform, whereas the outer control loop is used to control the position of the robot. If the inner loop runs on the vehicle, then current communication technologies, such as Wi-Fi and LTE, already provide the ability to remotely run the position control on a ground station. Under these conditions, despite the system being slower, it is still possible to demonstrate the stability of the overall quadrotor system. To show this property, it is possible to model the system as a double integrator, with independent subsystems for each Cartesian component, and assuming that the inner attitude loop is faster than the outer one. The time scale separation between inner and outer loops is a common assumption in the literature [22]. In these conditions, considering modeling errors and algorithm delays due to perception processing algorithms (typically within 10 – 20 ms), the average available phase margin is around 40 deg.

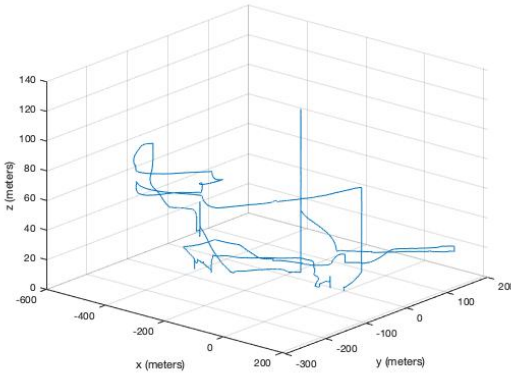
Nonetheless, by using 5G and mmWave communications, it is possible to obtain a better performance, particularly in terms of control system reactivity. The internal processing needed to solve the perception problem still takes around 10 to 20 ms [23], [24]. By exploiting the sub-ms latency and the high bandwidth of mmWave links, it is possible to efficiently offload the computation to edge-based systems, e.g., in the Base Stations (BSs), thus overcoming the limits in computational power of the on-board processing units and solving the perception problem in a shorter time interval. This provides more robustness and reactivity to the autonomous vehicle. The usage of off-board processors has also the potential to reduce the scale and size of the platforms with the inherent additional benefit to increase users' safety.



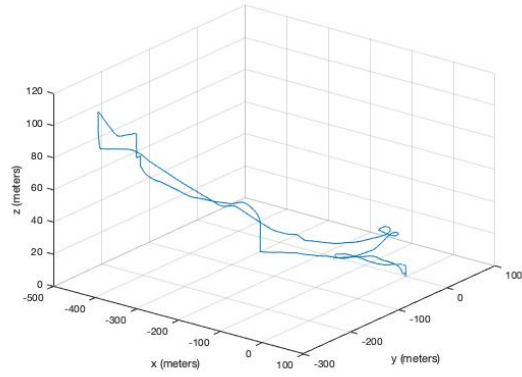
(a) Crowd Overwatch (mission 1)



(b) Missing Person (mission 2)



(c) Prescribed Burn (mission 3)



(d) Sonar Boat Training (mission 4)

Fig. 2: Plots of Drone Mission Flights.

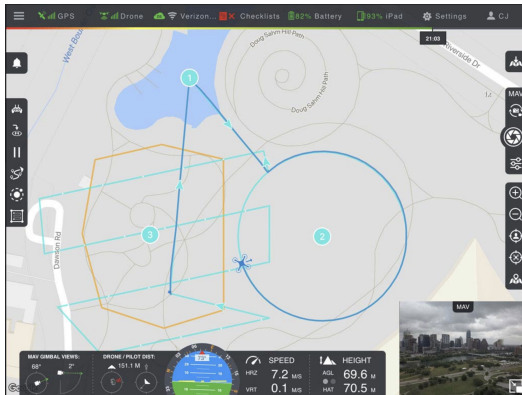


Fig. 3: Graphical interface of the software platform provided by DroneSense.

IV. DRONE MISSION FLIGHTS

The data we consider for our performance evaluation was provided by DroneSense, a company that partners up with our collaborators at Austin Fire Department (AFD) and develops a software platform for drones in public safety scenarios to expand situational awareness. The image in Fig. 3 represents the graphical interface provided by DroneSense's software platform, whereas the plots in Fig. 2 show the flight paths

each drone takes during its respective mission.

The traces incorporate four large scale critical missions:

- **Crowd Overwatch:** The drones are used to monitor crowd migration, evacuation corridors and the gathering of situational awareness for the Incident Commander during a large outdoor music festival.
- **Missing Person:** The AFD Robotics Emergency Deployment (RED) team participates in the search efforts for missing persons in both the green space and throughout various lake and river networks. The drone traces provide a virtual timeline of the aircraft being used to locate a victim that was reported missing in a remote woodland area.
- **Prescribed Burn:** The AFD regularly participates in the activity of prescription or controlled burns in order to reduce vegetative fuel loads and prevent extreme wildfires. The drones are used to track personnel, monitor fire behavior, and provide the Burn Boss/Incident Commander critical information from an elevated vantage.
- **Sonar Boat Training:** The AFD recently received a Unmanned Surface Vehicle (USV) through the Department of Navy. This system is used to search for victims and other points of interest located below the water surface. The AFD RED team acknowledges the value of machine

teaming to enhance the outcomes of using both the USV and Unmanned Aerial System (UAS) in both training and real world deployments. The UAS traces demonstrate how drones can be used to track, monitor and visualize surface operations of the USV aiding to the situational awareness required for emergency operations.

V. END-TO-END PERFORMANCE EVALUATION

We evaluated the performance of the UAV in terms of Signal to Noise Ratio (SNR), achievable throughput and latency using the open source ns-3 mmWave module described in [13]. This simulator features a full-stack implementation of 3GPP-like BSs and User Equipments (UEs), several channel models for mmWave propagation and fading, and procedures to handle beamforming and mobility events. Moreover, thanks to the integration with ns-3 [25], it is possible to integrate in the simulation the application, transport and network layers, and the modeling of the terminal mobility. This becomes particularly relevant for the simulation campaigns we present in this paper, where we consider a fixed BS, close to the Incident Command (IC) station, and the remote terminal installed in a UAV, which moves according to the traces of the missions described in Sec. IV.

A. Flight Data Processing and Mobility Model

During the flight missions, the flight controller in the UAV captures location data using a GPS-like format, i.e., with the latitude and the longitude of each visited point. We note that the geodesic curve on which any of these flight missions occur is nearly flat, given that the distance that the UAV spans is in the order of a few hundred meters, as shown in Fig. 2. Therefore, we convert the latitude and longitude measurements to 2-dimensional Cartesian coordinates by assuming a constant distance of 111 km between latitude lines. We proceed similarly for longitude lines, considering the average distance they have at the latitude at which the missions took place. This approximation is made since any warping to the flight patterns is negligible for the motivations of this work.

We use the ns-3 `WaypointMobilityModel` to integrate the mobility traces for the four scenarios in the simulations. The model accepts a 4-dimensional vector as a *waypoint*, with three spatial coordinates and a corresponding time stamp. Due to the complexity of the mobility model, the number of waypoints used for each simulation is only a fraction of the total coordinate points from the original extracted data, but is still representative of the mobility of the UAV during the flight. Notice that we do not account for the orientation of the vehicle; the modeling of this feature is left as a future work.

B. Simulation Parameters

For the mmWave links, the simulation uses the same channel model used for the evaluations in [3], [26], i.e., a single-ray Line of Sight (LOS) model with free space propagation, according to the 3GPP model in [27], shadowing, and the Doppler effect introduced by the mobility of the vehicle. We consider a bandwidth $B = 1$ GHz with a carrier frequency

TABLE I: Main simulation parameters

Parameter	Value
mmWave carrier frequency f_c	28 GHz
mmWave bandwidth B	1 GHz
LTE carrier frequency f_c	2.1 GHz
LTE bandwidth B	20 MHz
Beamforming update period	5 ms
Antenna combinations $A = N_{\text{BS}} \times N_{\text{UAV}}$	$16 \times 4, 64 \times 16$
Application source rate R	10, 1000 Mbps
BS height	25 m

$f_c = 28$ GHz, as reported in Table I. The physical layer model in the simulator implements an adaptive modulation and coding scheme which yields a throughput of 3.2 Gbps for a BS with the best channel conditions. We also compare the performance of the mmWave connection with a baseline architecture with LTE connectivity at frequency below 6 GHz (i.e., $f_c = 2.1$ GHz). While, in general, the LTE connection enjoys a reduced pathloss with respect to mmWave, it can only exploit a much smaller bandwidth, which generally yields a much lower throughput. Moreover, the frame design of LTE does not support the sub-ms latency of the mmWave connection [28].

When considering mmWave communications in the context of UAV mobility, a critical operation is the beam tracking, which makes it possible to maintain a fine alignment of the transmitter and receiver beams and obtain a high beamforming gain [15]. While beam management operations are fundamental also in cellular networks [29], efficient and precise tracking is harder to obtain with flying vehicles, which present more erratic mobility patterns. Therefore, when simulating operations with UAVs at mmWave frequencies, it is important to model a realistic beam tracking procedure. In this paper, following the approach presented in [3], we model the beam tracking by updating the beam pair to the optimal one with a fixed periodicity, and by keeping it fixed in the time interval between two consecutive updates. The periodicity we consider (5 ms) is compatible with that of the transmission of tracking reference signals (e.g., Channel State Information - Reference Signals (CSI-RSs)) in 3GPP NR [29].

For the application, we consider a source that periodically generates a payload of 1500 bytes, for a total source rate of 10 or 1000 Mbps, in order to model different use cases (e.g., sensor information, video at different rates). We consider UDP as transport protocol, and the transmissions are in uplink, i.e., from the UAV to the IC station.

C. Results

In this paragraph, we will report the main results of our preliminary performance evaluation, where we focus on three main metrics: (i) the SNR of the link; (ii) the throughput and (iii) the one-way latency. Both throughput and latency are measured at the Packet Data Convergence Protocol (PDCP) layer of the cellular protocol stack, given that we assume that the sink of the communications is at the IC station and thus

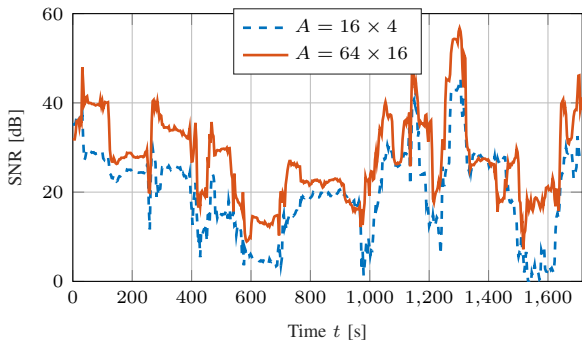


Fig. 4: SNR over time, for mission 3 (prescribed burn).

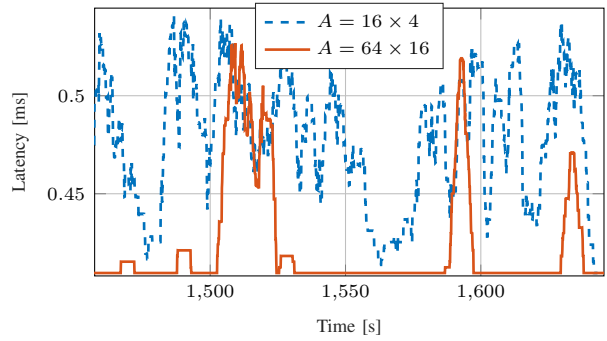


Fig. 5: Latency over time, for mission 3 (prescribed burn). The latency is averaged over fixed intervals of 5 seconds.

there is no additional delay introduced by forwarding over the public Internet.

Fig. 4 reports, as an example, the evolution of the SNR over time in mission 3, i.e., the prescribed burn, for the mmWave link and different combinations of the transmitter and receiver side antennas. Notice that, given that the BS is located in the area of the mission of the UAV, the SNR is high for most of the mission duration also with the antennas with fewer elements, which in general introduce a lower beamforming gain. Nonetheless, a higher number of antenna elements can help to sustain a better connection in particularly critical

conditions, e.g., when the UAV and the BS at the IC station are further away. This is shown in Fig. 4, where it can be seen that for the three time intervals roughly between $t = 580$ and 700 seconds, around $t = 1000$ seconds and between $t = 1500$ and 1600 seconds the SNR achievable with $A = 64 \times 16$ is approximately 10 dB higher than that with $A = 16 \times 4$. This has a direct impact on the latency of the connection, as shown in Fig. 5: a lower SNR translates into a higher number of errors at the physical layer, and, consequently, retransmissions at the higher layers.

In order to better understand the limitations of the LTE

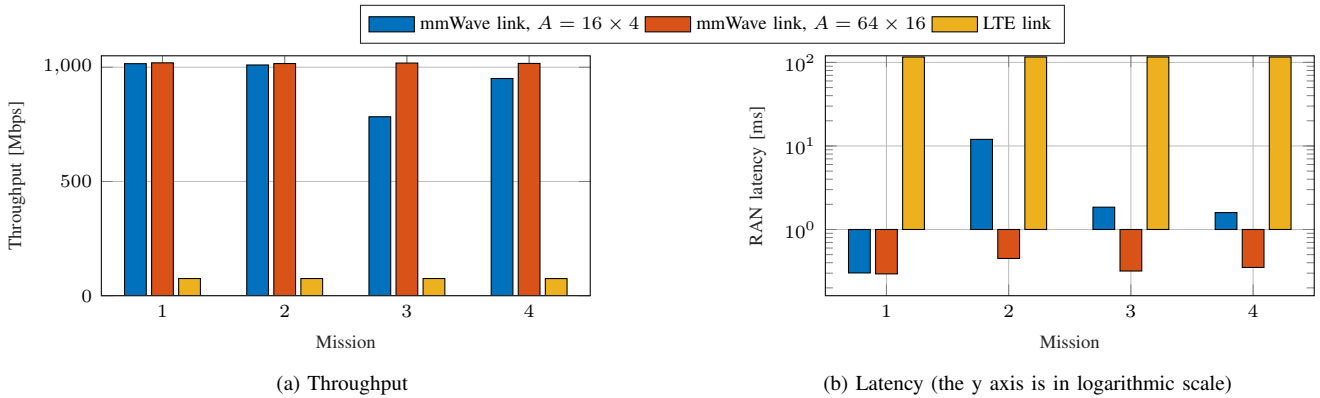


Fig. 6: Performance for different missions, with different antenna combinations and a source rate of 1000 Mbps.

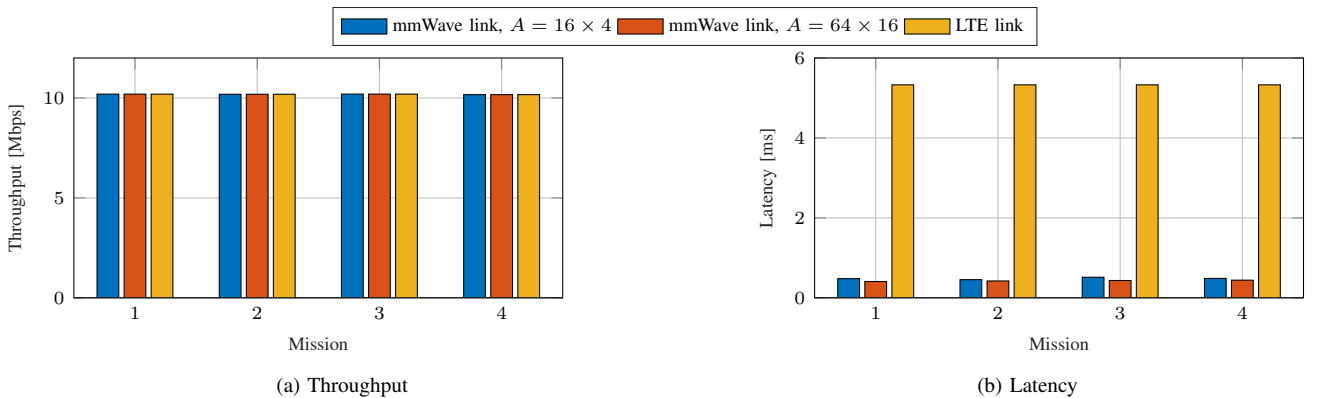


Fig. 7: Performance for different missions, with different antenna combinations and a source rate of 10 Mbps.

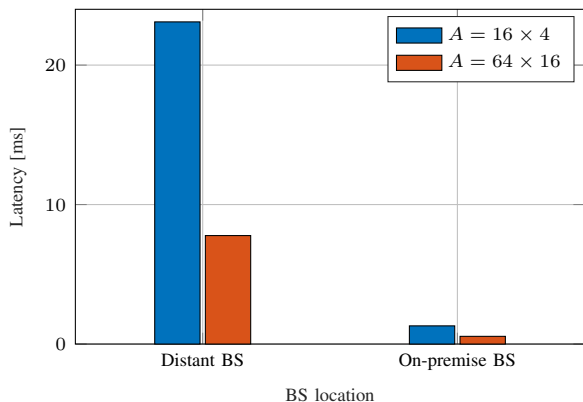


Fig. 8: Comparison of the average latency with different locations of the BS covering the scenario: the distant BS is 2 km away from the mission area, while the on-premise BS is in the mission area.

connection and the potentials of mmWave communications for remote UAV control and data exchange, we plot in Figs. 6 and 7 the average throughput and latency over the duration of each mission. In particular, we report results for the highest source rate we consider (i.e., 1000 Mbps) in Fig. 6, and for 10 Mbps in Fig. 7, and analyze the performance of the 2.1 GHz LTE link with that of the two antenna combinations for mmWave frequencies at 28 GHz. By comparing the two figures, it is possible to understand the trade-off between LTE and mmWaves. As expected, LTE can yield a low latency (i.e., below 10 ms, in line with the frame structure that is considered in this simulator [30]) only with low source rates, and cannot support rates higher than tens of Mbps in uplink, as shown by Fig. 6a. Therefore, the improved remote control of a UAV through offloading could be enabled only if the link is not used for the transmission of other kinds of data, and for a limited number of UAVs. Furthermore, this prevents for example the sharing of high-quality aerial video or other sensor data (e.g., LIDAR data) that cannot consequently be used to refine the performance of the control algorithms. The mmWave connection, instead, performs better in terms of both throughput and latency. Thanks to the high availability of bandwidth, the mmWave links manage to deliver the full 1000 Mbps throughput in almost all configurations. Moreover, the low-latency frame design that is coupled with mmWave communications in 5G cellular networks [29], [31] helps achieve a small latency, with a sub-ms one-way latency that can enable a prompt control of the remote UAV, also if the connection is shared with data-intensive streams from, for example, high-resolution cameras.

When comparing different missions, it can be seen that for missions 1 and 2 both the antenna configurations can support the source rate of 1000 Mbps, while the full throughput is reached for missions 3 and 4 only with the largest antenna arrays tested. This configuration also provides the best performance in terms of latency.

The same trend can be observed in Fig. 8, where we compare the average latency with a BS on-premise, i.e., in

the area of the mission (as for the previous results), and one placed 2 km apart. The rate we consider in this case is 1000 Mbps, which is not sustainable by the LTE connection. It can be seen that, despite the challenging communication condition introduced by the high propagation loss, the antenna configuration with $A = 64 \times 16$ is still capable of ensuring low-latency data exchange, and thus enable prompt and reactive UAV control also at larger distances. Moreover, as future work, we will explore the possibility of using an even larger number of antennas at the base station, to provide a higher beamforming gain and reduce even further the latency in the distant BS scenario.

VI. CONCLUSIONS

MmWave communications are a valuable resource in various realistic public safety scenarios, as described in Sec. IV. We offered an analysis of End-to-End (E2E) simulations and showed the potential of 5G networks to deliver throughput up to 1 Gbps with consistent sub ms latency when the base station is located near the mission area. The benefits derived from the advent of 5G technology will certainly not be limited better controllability and reactivity of the aerial vehicles. In general, 5G signal traces have the potential to be used in a cooperative localization framework to identify the spatial location of different vehicles in GPS denied environments out of the line of sight of an operator, thereby increasing autonomy, safety, and usability of these platforms. Finally, the ability to outsource most of the computation to the edge cloud will also allow to scale down these vehicles, increasing user and environment safety.

That said, there are several extensions to this work that require attention in order to present this emerging technology as a complete and viable solution in the context of PSCs. As mentioned in Sec. V, the orientation of the UAV is not modeled. This factor is expected to significantly affect the beam tracking and thus a more complex simulation framework is needed to evaluate and understand these effects. Furthermore, we note that the environment for many public safety scenarios will likely have buildings within the mission area as well as wind perturbations. This will require simulations that model these blockage effects and perturbations in the flight path of the UAV. The throughput, latency and subsequent implications for the platform's stability can then be explored in more detail.

VII. ACKNOWLEDGMENTS

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REFERENCES

- [1] A. Merwaday and I. Guvenc, "UAV assisted heterogeneous networks for public safety communications," in *IEEE Wireless Communications and Networking Conference Workshops (WCNCW)*, March 2015, pp. 329–334.

- [2] D. He, S. Chan, and M. Guizani, "Drone-assisted public safety networks: The security aspect," *IEEE Communications Magazine*, vol. 55, no. 8, pp. 218–223, 2017.
- [3] M. Mezzavilla, M. Polese, A. Zanella, A. Dhananjay, S. Rangan, C. Kessler, T. Rappaport, and M. Zorzi, "Public Safety Communications above 6 GHz: Challenges and Opportunities," *IEEE Access*, vol. 6, pp. 316–329, November 2017.
- [4] "Velodyne Lidar," <https://velodynelidar.com/hdl-64e.html>.
- [5] S. Barbarossa, S. Sardellitti, E. Ceci, and M. Merluzzi, "Chapter 16 - The Edge Cloud: A Holistic View of Communication, Computation, and Caching," in *Cooperative and Graph Signal Processing*, P. M. Djuric and C. Richard, Eds. Academic Press, 2018, pp. 419–444. [Online]. Available: <http://www.sciencedirect.com/science/article/pii/B978012813677500016X>
- [6] G. Loianno, C. Brunner, G. McGrath, and V. Kumar, "Estimation, Control, and Planning for Aggressive Flight With a Small Quadrotor With a Single Camera and IMU," *IEEE Robotics and Automation Letters*, vol. 2, no. 2, pp. 404–411, April 2017.
- [7] G. Loianno, V. Spurny, J. Thomas, T. Baca, D. Thakur, D. Hert, R. Penicka, T. Krajnik, A. Zhou, A. Cho, M. Saska, and V. Kumar, "Localization, Grasping, and Transportation of Magnetic Objects by a Team of MAVs in Challenging Desert-Like Environments," *IEEE Robotics and Automation Letters*, vol. 3, no. 3, pp. 1576–1583, July 2018.
- [8] S. Rangan, T. S. Rappaport, and E. Erkip, "Millimeter-Wave Cellular Wireless Networks: Potentials and Challenges," *Proceedings of the IEEE*, vol. 102, no. 3, pp. 366–385, March 2014.
- [9] T. S. Rappaport, S. Sun, R. Mayzus, H. Zhao, Y. Azar, K. Wang, G. N. Wong, J. K. Schulz, M. Samimi, and F. Gutierrez, "Millimeter wave mobile communications for 5G cellular: It will work!" *IEEE Access*, vol. 1, pp. 335–349, 2013.
- [10] 3GPP, "NR and NG-RAN Overall Description - Rel. 15," TS 38.300, 2018.
- [11] S. D. Muruganathan, X. Lin, H.-L. Maattanen, Z. Zou, W. A. Hapsari, and S. Yasukawa, "An overview of 3GPP release-15 study on enhanced LTE support for connected drones," *arXiv preprint arXiv:1805.00826*, 2018.
- [12] 3GPP, "Study on Enhanced LTE Support for Aerial Vehicles - Rel. 15," TS 36.777, 2017.
- [13] M. Mezzavilla, M. Zhang, M. Polese, R. Ford, S. Dutta, S. Rangan, and M. Zorzi, "End-to-End Simulation of 5G mmWave Networks," *IEEE Communications Surveys and Tutorials*, vol. 20, no. 3, pp. 2237–2263, Third quarter 2018.
- [14] Y. Zeng, J. Lyu, and R. Zhang, "Cellular-Connected UAV: Potential, Challenges, and Promising Technologies," *IEEE Wireless Communications*, vol. 26, no. 1, pp. 120–127, February 2019.
- [15] Z. Xiao, P. Xia, and X. Xia, "Enabling UAV cellular with millimeter-wave communication: potentials and approaches," *IEEE Communications Magazine*, vol. 54, no. 5, pp. 66–73, May 2016.
- [16] W. Khawaja, O. Ozdemir, and I. Guvenc, "UAV Air-to-Ground Channel Characterization for mmWave Systems," in *IEEE 86th Vehicular Technology Conference (VTC-Fall)*, Sep. 2017, pp. 1–5.
- [17] —, "Temporal and Spatial Characteristics of mm Wave Propagation Channels for UAVs," in *11th Global Symposium on Millimeter Waves (GSMM)*, May 2018, pp. 1–6.
- [18] T. Cuvelier and R. W. Heath, "MmWave MU-MIMO for Aerial Networks," in *15th International Symposium on Wireless Communication Systems (ISWCS)*, Aug 2018, pp. 1–6.
- [19] J. Zhao, F. Gao, L. Kuang, Q. Wu, and W. Jia, "Channel Tracking With Flight Control System for UAV mmWave MIMO Communications," *IEEE Communications Letters*, vol. 22, no. 6, pp. 1224–1227, June 2018.
- [20] Z. Khosravi, M. Gerasimenko, S. Andreev, and Y. Koucheryavy, "Performance Evaluation of UAV-Assisted mmWave Operation in Mobility-Enabled Urban Deployments," in *41st International Conference on Telecommunications and Signal Processing (TSP)*, July 2018, pp. 1–5.
- [21] N. Rupasinghe, Y. Yapici, I. Guvenc, and Y. Kakishima, "Non-orthogonal multiple access for mmWave drones with multi-antenna transmission," in *51st Asilomar Conference on Signals, Systems, and Computers*, Oct 2017, pp. 958–963.
- [22] N. Michael, D. Mellinger, Q. Lindsey, and V. Kumar, "The GRASP Multiple Micro-UAV Testbed," *IEEE Robotics Automation Magazine*, vol. 17, no. 3, pp. 56–65, Sep. 2010.
- [23] C. Forster, M. Pizzoli, and D. Scaramuzza, "SVO: Fast semi-direct monocular visual odometry," in *IEEE International Conference on Robotics and Automation (ICRA)*, May 2014, pp. 15–22.
- [24] G. Loianno, M. Watterson, and V. Kumar, "Visual inertial odometry for quadrotors on se(3)," in *IEEE International Conference on Robotics and Automation (ICRA)*, May 2016, pp. 1544–1551.
- [25] T. R. Henderson, M. Lacage, G. F. Riley, C. Dowell, and J. Kopena, "Network simulations with the ns-3 simulator," *SIGCOMM demonstration*, vol. 14, no. 14, p. 527, 2008.
- [26] M. Polese, M. Mezzavilla, S. Rangan, C. Kessler, and M. Zorzi, "mmWave for Future Public Safety Communications," in *Proceedings of the First CoNEXT Workshop on ICT Tools for Emergency Networks and Disaster Relief*, ser. I-TENDER '17. Incheon, Republic of Korea: ACM, 2017, pp. 44–49. [Online]. Available: <http://doi.acm.org/10.1145/3152896.3152905>
- [27] 3GPP, "Study on Channel Model for Frequency Spectrum Above 6 GHz," TR 38.900 (Rel. 14), 2017.
- [28] R. Ford, M. Zhang, M. Mezzavilla, S. Dutta, S. Rangan, and M. Zorzi, "Achieving Ultra-Low Latency in 5G Millimeter Wave Cellular Networks," *IEEE Communications Magazine*, vol. 55, no. 3, pp. 196–203, March 2017.
- [29] M. Giordani, M. Polese, A. Roy, D. Castor, and M. Zorzi, "A Tutorial on Beam Management for 3GPP NR at mmWave Frequencies," *IEEE Communications Surveys and Tutorials*, vol. 21, no. 1, pp. 173–196, First quarter 2019.
- [30] K. Takeda, L. H. Wang, and S. Nagata, "Latency Reduction toward 5G," *IEEE Wireless Communications*, vol. 24, no. 3, pp. 2–4, June 2017.
- [31] S. Dutta, M. Mezzavilla, R. Ford, M. Zhang, S. Rangan, and M. Zorzi, "Frame structure design and analysis for millimeter wave cellular systems," *IEEE Transactions on Wireless Communications*, vol. 16, no. 3, pp. 1508–1522, March 2017.