

E-Week Engineering Showcase 2025 CPE496/498 Capstone Design Course

SPARTN Drone Detection Network [Sensor Platform for **Acoustic and Radio Tracking Network**]

Tyler Briggs, Nicholas Brownfield, Dawson Hampton, Jeff Ikeda, Kentrell Martin, Mikhail Ramirez, Dr. David Coe, Department of Electrical and Computer Engineering **Overview**

- Rapid drone technology advancement and ineffective geofencing create new challenges.
- There is a need for a simple, rapid deployment system with real-time detection, identification, and tracking of hobbyist-sized drones.

Requirements

- The system should reliably detect small, hobbyistsized drones, anything under 250 grams.
- The system should detect drones using sound and radio-frequency signatures. The system should use localization and report the exact position of drones in real-time. • The system should be able to characterize the type of drone detected.
- Our project addresses this with real time detection, identification, and tracking of hobbyist-sized drones.



System Design

The SPARTN drone detection network is comprised of four microphones, four HackRF SDRs, a central control tower, and a tablet.



Figure 2. Training the drone classification Al





Figure 4. SPARTN System hardware Mic stands, HackRFs, battery, usb splitters, References tablet

	0 0			
drone	Tower 4 (from config)	Mic Connected 🕥	(A)	
	XY	Send Location	Î	
	Listening on 10.4.153.44:39439			No [

Results/Conclusion

- The most important results for our project are drone locating and identification through real time captured audio data.
- The amplitude shifting localization method estimates position within a 16x16 foot area.
- The AI model is operating at up to 90% accuracy for drone classification.
- Future work would consist of greater integration of RF systems, better hardware for TDOA, expanded range and more robust encryption.

1. Shi, Z., Chang, X., Yang, C., Wu, Z., & Wu, J. (2020). An acoustic-based surveillance system for amateur drones detection and localization. IEEE Transactions on Vehicular Technology, 69(3), 2731–2739. https://doi.org/10.1109/TVT.2020.2964110 2. Wang, L., & Cavallaro, A. (2022). Deep-learning-assisted sound source localization from a flying drone. IEEE Sensors Journal, 22(21), 20828–20838. https://doi.org/10.1109/JSEN.2022.3207660

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