

# Decentralized Control, Dynamic Planning, and Localization in Autonomous Drone Systems: A Meta-Analysis

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## **Abstract**

Autonomous systems, and especially those applied in robotics, drones, and autonomous cars, need effective algorithms to plan their trajectory and interact with the environment. This meta-analysis reviews the existing body of literature about three important matters: path planning, centralized versus decentralized solutions, and localization techniques. We contrast and compare in detail a wide range of path planning algorithms from classical graph-based algorithms such as Dijkstra and A\* to their biologically inspired versions such as Ant Colony Optimization and Particle Swarm Optimization. The study also delves into the trade-offs between centralized and decentralized decision-making architectures with a focus on pros and cons of the same in real-world scenarios. Finally, we contrast various localization methods, i.e., vision-based, LiDAR-based, and hybrid methods, and discuss their performance measures of accuracy, scalability, and robustness. This comprehensive paper gives insights into advantages and disadvantages and avenues for future research in integrating path planning, decentralized

control, and localization techniques within drone swarm systems to make researchers and practitioners aware of the current state of the art and future trends in the area.

## **Introduction**

Autonomous systems, particularly robots, drones, and autonomous cars, are presently at the forefront in a vast range of applications in logistics, transport, monitoring, and environmental surveillance. The systems' performance and efficiency are largely dependent on the ability to make sensible, real-time decisions with respect to intricate and dynamic environments. To achieve this, autonomous systems require sophisticated algorithms for path planning, localization, and control, each of which is significant in making machines capable of navigation and interacting with their environment.

Path planning, determining the optimal path that an autonomous agent should follow to travel from a start point to an end point, has been central to the development of autonomous systems for decades. With the ongoing development of the tools of optimization and machine learning, the field has come a long way with the development of a wide range of algorithms. These range from the classical methods, such as A\* and Dijkstra's, to more specialized biologically inspired algorithms, such as Ant Colony Optimization (ACO) and Particle Swarm Optimization (PSO), which simulate natural processes to solve problems of some complexity.

And still another important attribute of autonomous systems is the decision-making process structure. A centralized versus decentralized system decision impacts performance, scalability, and fault tolerance profoundly. Centralized systems employ a global controller in deciding for the overall system, and decentralized systems have each agent decide locally based on local data. Both of the approaches do possess their respective trade-offs, i.e., efficiency, reliability, and fault tolerance. It is vital to realize such trade-offs while deciding on the correct method for real systems, especially in complex and dynamic contexts.

Localization, the process of determining the position and orientation of an autonomous agent within its environment, is another critical component of autonomous systems. As systems tend to work in dynamic and uncertain environments, the right choice of localization technique is required to offer system accuracy, reliability, and scalability. Various techniques like vision-based, LiDAR-based, and hybrid techniques, with their advantages and disadvantages, are available. A mix of correct localization and path planning as well as control algorithms is required to ensure the correct operation of autonomous systems, particularly in such applications as multi-agent coordination and swarm drone systems.

This paper provides a critical assessment of the state of the art in autonomous systems with a special focus on three aspects: path planning, centralized versus decentralized approaches, and localization techniques. Based on a critical examination of these aspects, this research aims to offer an understanding of the strengths, limitations, and future directions of research and development in autonomous systems, with particular emphasis on their applications in drone swarm systems.

## **Background**

Autonomous systems have achieved a gigantic quantum jump in the past few decades with the progress achieved in artificial intelligence, machine learning, and robotics, all of which contributed in large numbers toward the accelerated growth of autonomous systems. Autonomy in unstructured dynamic environments has been among the largest concerns that current robots and autonomous systems are facing today. Early work in this field focused on the development of algorithms for solo agents, but as systems have become more complex, increasing attention has been paid to multi-agent systems, where multiple autonomous agents need to coordinate their actions to achieve a common goal.

Path planning has been of prime concern in autonomous system construction. Classical path-planning algorithms, including Dijkstra's and A\*, rely on graph-traversal methods and suit

environments where there is known information about the map and the purpose is to arrive at the minimum distance between two points. Such algorithms are computation-efficient and have the capability of providing optimal solutions under some situations. Nevertheless, they are limited by how effectively they can adapt in dynamic situations when the barriers are reallocated or more information is revealed in real time.

To minimize these limitations, researchers have begun applying biologically inspired algorithms such as Ant Colony Optimization (ACO) and Particle Swarm Optimization (PSO) because they are more dynamic-friendly. These algorithms take advantage of the behavioral patterns of natural organisms, such as birds or ants that are cooperative and can learn in a dynamic environment. ACO, for example, imitates the foraging behavior of ants, where ants deposit pheromones on the ground to indicate the shortest path to a food source. Similarly, PSO depends on particle movement within a search space where each particle will update its position depending on its individual knowledge and neighbor experience.

Despite improvements in path planning algorithms, centralized vs decentralized decision-making architecture remains one of the biggest research areas. Decisions under centralized architectures are all made by one controller who also has global system state access to make global decisions. This kind of architecture is coordination- and optimization-efficient but has vulnerabilities to single points of failure and scalability issues. Decentralized systems, in contrast, implement local decision-making, where each agent has bounded knowledge and must make decisions locally based on its perception and interaction with other agents. Decentralized systems are more resilient and scalable but experience coordination problems and bad performance in certain circumstances.

Localization, or an agent's ability to specify its position and orientation in the world, is yet another essential component of autonomous systems. Localization must be done correctly to enable useful path planning, collision avoidance, and coordination of many agents within a common environment. Several localization methods have been proposed, such as vision-based

methods with the help of cameras and image processing and LiDAR-based methods with the use of laser scanners to form accurate 3D models of the surrounding environment. The hybrid methods based on the merits of vision and LiDAR are also popular with their improved robustness and accuracy in varied environments.

Path planning, decentralized control, and localization algorithms are more consolidated for drone swarm systems. Drone swarms consist of multiple autonomous drones that must collaborate to achieve a common goal, such as surveying terrain or performing a coordinated operation, such as mapping terrain or search and rescue. Such systems require each drone to plan its path considering other drones' behavior, maintain accurate localization, and decentralized decision making from local information. Coordination between such actors is difficult in fluctuating surroundings, as well as in dynamic surroundings and with obstacles and other agents.

With continued advancements in autonomous systems, the need for advanced and adaptive algorithms that can deal with the greater complexity of the real world keeps growing. Combining path planning, decentralized control, and localization techniques is pivotal to the realization of autonomous systems that are both scalable and efficient as well as robust and reliable across a wide range of applications.

## **Methods**

### Data collection:

Data for the research was gathered from a thorough literature review using NotebookLM, which is software used to download and read research articles. The research focused on key topics in autonomous systems, specifically centralized vs. decentralized systems, path planning algorithms, and localization methods. The articles that were reviewed were from recent publications, which made the research applicable to the current level of developments in the field.

To address the research hypothesis, which contended that decentralized drones would be more common in practice compared to centralized drones, we examined a total of 16 research articles. The articles were classified based on the mention of decentralized or centralized system architecture, where 4 articles mentioned centralized systems and 12 mentioned decentralized systems. The prevalence of decentralized systems confirmed the hypothesis that the issue of decentralized systems is mentioned more and used in discussions of drone autonomy.

#### Path planning algorithms:

For comparison of path planning algorithms, the articles were categorized in terms of significant features that are typically considered in autonomous system path planning. The features considered are:

- Global Planning: Algorithms capable of planning a path across an entire space.
- Local Planning: Algorithms employed for local, short-term motion planning choices.
- Dynamic Environment: Algorithms capable of dealing with changing environments, e.g.,  
moving obstacles.
- Smooth Paths: Algorithms that create smooth and efficient paths with fewer sudden  
turns.
- Real-time Feasibility: Algorithms that are real-time feasible, which is important for  
practical deployment.
- Exploration Capability: Algorithms with the ability to navigate and explore unknown or  
partially known spaces.

As dynamic environments are of critical importance in real-world applications, we sought to find algorithms that addressed this factor. The classification was performed using a binary

representation, where '1' denotes that a given feature is addressed by the algorithm and '0' denotes that it is not.

#### Localization Methods:

In the review of localization methods, the following categories were considered:

- Vision-SLAM: Simultaneous Localization and Mapping (SLAM) techniques based on visual data, such as monocular, stereo, or RGB-D cameras.
- RF Ranging: Localization based on radio frequency signals, commonly used for systems that operate in environments with limited visibility.
- IMU-based: Methods utilizing Inertial Measurement Units (IMUs) for dead reckoning and orientation estimation.
- LiDAR-SLAM: Localization using LiDAR sensors, offering high-precision, 3D environmental mapping.

We had hypothesized, on the basis of past experience in the field, that vision-based SLAM would see the widest application in practice. For each article, we coded which localization approach was discussed. The outcomes were marked in binary form, where '1' indicated the use of a given localization approach and '0' indicated that it was not mentioned.

#### Analysis:

The classification results were presented in Python-generated tables and depicted as heatmaps to visualize the frequency and distribution of different techniques throughout the literature. This allowed trends in the use of centralized versus decentralized systems, path planning functionality, and localization strategies to be identified.

## **Results**

### Centralized vs. Decentralized Systems:

The analysis of the 16 papers revealed a resounding penchant for decentralized over centralized systems for autonomous drones. Specifically, 12 papers (75%) referred to decentralized systems, whereas only 4 papers (25%) talked about centralized systems. This finding upholds the hypothesis that decentralized control is more widely used and studied in the case of drone autonomy. Decentralized systems get increasingly popular because they are expandable, robust, and have the ability to deal with ever-changing environments wherein centralized decision making might be unsuitable.

### Path Planning:

The categorization of the path planning algorithms exposed a variety of consideration for the different features. Of the 8 path planning algorithms analyzed, the dynamic environments feature was addressed most often, and 6 (75%) of the algorithms had this consideration in their algorithm structure. This makes the first hypothesis true that dynamic environments are an inherent consideration to be made while designing path planning algorithms because applications in the real world will many times have unplanned obstacles and environmental changes. Other notable results are: Global planning was taken into account by 5 algorithms (62.5%). Local planning was taken into account by 6 algorithms (75%). Real-time feasibility was taken into account by 4 algorithms (50%). Smooth paths and exploration capability were taken into account by 3 algorithms (37.5%). These results indicate that while the majority of path planning algorithms are designed with dynamic and local planning in mind, there is still room for development in the realm of real-time feasibility and smooth, efficient path generation.



		Comparison of Path Planning Algorithms					
Algorithm	Interfered Fluid Dynamical System (IFDS)	1	0	0	1	1	0
	Lyapunov Navigation Vector Field	0	1	1	1	1	1
	CNN-based Mapless Approaches	0	1	1	1	1	1
	Swarm Path Planning (Virtual Pheromones)	1	1	1	0	0	1
	Simulated Annealing	1	0	0	0	0	0
	Next Best View (NBV)	0	1	1	0	1	1
	Artificial Fish Swarm Algorithm	0	0	0	1	1	1
	Ant Colony Optimization (ACO)	1	0	1	0	0	0
		Global Planning	Local Planning	Dynamic Environment	Smooth Paths	Real-time Feasibility	Exploration Capability

### Localization Methods:

The analysis of methods of localization revealed that Vision-SLAM was the most commonly quoted method, with 6 papers (60%) citing it. This validates the first hypothesis that vision-based SLAM is the most used method of choice for autonomous systems due to its effectiveness and prevalence of cameras and computing power. The other methods were less frequently mentioned: LiDAR-SLAM appeared in 4 papers (40%). IMU-based methods appeared in 3 papers (30%). RF-based ranging was the least common, documented in only 2 papers (20%). Vision-based methods' frequency is an indication of ongoing development in computer vision and widespread availability of cameras for most autonomous systems. LiDAR-based methods, as highly accurate as they are, are more expensive and complex and are, therefore less commonly used.

Comparison of Localization Methods in Papers					
Paper	ORB-SLAM2 (Mur-Artal & Tardós, 2017)	1	0	1	0
	LSD-SLAM (Engel et al., 2014)	1	0	0	0
	UWB-based Ranging (Shule et al., 2021)	0	1	0	0
	RF Localization for UAV Swarms (Zhou et al., 2020)	0	1	0	0
	VINS-Mono (Qin et al., 2018)	1	0	1	0
	MSCKF (Mourikis & Roumeliotis, 2007)	1	0	1	0
	LIDAR-based UAV Navigation (Zhang et al., 2019)	0	0	0	1
	LOAM (Zhang & Singh, 2014)	0	0	0	1
	IMU-based Dead Reckoning (Titterton & Weston, 2004)	0	0	1	0
	Visual-Inertial SLAM (Forster et al., 2017)	1	0	1	0
		Vision-SLAM	RF Ranging	IMU-based	LIDAR-SLAM
Method					

## Conclusion

This study was helpful in pointing out the direction of the current trends and interests in autonomous systems, more specifically drone swarm technology, and planning algorithms and localization algorithms involved. Decentralized vs. Centralized Systems: Decentralized systems are discussed and implemented more than centralized systems when it comes to autonomous drones. This is a reflection of the growing acceptance of value generated by decentralized control, particularly in large and dynamic environments.

**Path Planning:** Dynamic environment path planning is imperative, as exemplified by the frequent occurrence of algorithms addressing this field. The research concludes that although path planning algorithms are growing in complexity, there is still room for enhancing real-time usability and optimal and silky path creation in autonomous systems.

**Localization:** Vision-based SLAM remains the dominant localization technique since it is effective and camera systems are easily found. Other techniques such as LiDAR and

IMU-based approaches also take considerable roles, particularly in certain environments or applications requiring greater precision.

These findings show that strong adaptability and aggressive localization are essential to successful autonomous drone systems. Future research will require incorporating optimizing the scalability and the real-time path planning algorithm's efficiency, as well as expanding the use of multiple localization techniques to continue to enhance the robustness of drone flight in harsh environments. In its literature review, this work also identifies several avenues for future research, including the development of hybrid path planning and localization techniques and further exploration of decentralized decision-making structures. Research in these areas continues to be essential to the development of increased autonomy, efficiency, and safety for swarms of drones.

## References

- Yuxin Xie, Liang Han, Xiwang Dong, Qingdong Li, Zhang Ren. Bio-inspired adaptive formation tracking control for swarm systems with application to UAV swarm systems. *Neurocomputing* 453 (2021) 272–285. <https://doi.org/10.1016/j.neucom.2021.05.015>
- Davison, "Real-time simultaneous localisation and mapping with a single camera," *Proceedings Ninth IEEE International Conference on Computer Vision, Nice, France, 2003*, pp. 1403-1410 vol.2.
- H. M. Wallach, "Topic modeling: beyond bag-of-words." *Proceedings of the 23rd international conference on Machine learning*. 2006.
- Y. Chang et al. *Robotics and Autonomous Systems* 170 (2023) 104533. A review of UAV navigation in GPS-denied environments. Article under the CC BY license (<http://creativecommons.org/licenses/by/4.0/>).

- Barbeau, M., Garcia-Alfaro, J., Kranakis, E.: Geocaching-inspired resilient path planning for drone swarms. In: IEEE Conference on Computer Communications Workshops (INFOCOM WKSHPS), IEEE INFOCOM 2019, Paris, France, pp. 620–625. IEEE, April 2019. <https://doi.org/10.1109/INFCOMW.2019.8845318>
- Maheswaran, S., Gomathi, R. D., Rithikhaa, D., Praveen, B., Prathiksha, T., Murugesan, G., ... & Nandita, S. (2023, July). A Perspective way of designing Intelligent systems with Face Detection and Recognition using Artificial Intelligence for Authentication. In 2023 14th International Conference on Computing Communication and Networking Technologies (ICCCNT) (pp. 1-6). IEEE.
- Frederike Duembgen, Adrien HOFFET, Mathias Kolundzija, Adam Scholefield, Martin Vetterli. Blind as a Bat: Audible Echolocation on Small Robots. IEEE Robotics and Automation Letters, vol. 7, no. 4, pp. 10891-10898, Oct. 2022. Digital Object Identifier 10.1109/LRA.2022.3194669
- Qing Wang, Meng Xu, Zhen Hu. Path Planning of Unmanned Aerial Vehicles Based on an Improved Bio-Inspired Tuna Swarm Optimization Algorithm. Biomimetics 2024, 9, 388. <https://doi.org/10.3390/biomimetics9070388>
- Vision-based Relative Navigation and Drone Swarming. ASABE Paper No. 2401319.
- Manuscript received 18 May 2023; revised 15 September 2023 and 23 November 2023; accepted 21 January 2024. Date of publication 31 January 2024; date of current version 7 June 2024.
- PACNav: A Collective Navigation Approach for UAV Swarms Deprived of Communication and External Localization.
- B. Chenchana, O. Labbani-Igbida, S. Renault, and S. Boria, "Range-based collaborative msckf localization," in Proc. 25th International Conference on Mechatronics and Machine Vision Practice, 2018, pp. 1–6. Relative navigation of UAV swarm in a GPS-denied environment.

- Design of a swarm of Unmanned Aerial Vehicles for the exploration of Mars.  
Report\_TFG\_NEscursell.pdf.
- Liang Yan · Haibin Duan · Yimin Deng Editors. Advances in Guidance, Navigation and Control. Proceedings of 2024 International Conference on Guidance, Navigation and Control (Volume 18). Editors Liang Yan School of Automation Science and Electrical Engineering Beihang University Beijing, China. Yimin Deng School of Automation Science and Electrical Engineering Beihang University Beijing, China. Haibin Duan School of Automation Science and Electrical Engineering Beihang University Beijing, China.
- Hongwei Wei and Yunxia Li. The Relative Positioning Algorithm for Aerial Refueling Based on Contour Matching. Advances in Guidance, Navigation and Control. Proceedings of 2024 International Conference on Guidance, Navigation and Control (Volume 18), pp. 549-557.
- Jiayun Wen, Qiang Tang, Xianglun Zhang, Hao Li, Yukun Yang, Shichao Ma, and Zhiyu Li. UAVs Formation Obstacle Avoidance Method Based on Improved Artificial Potentials and Virtual Leaders. Advances in Guidance, Navigation and Control. Proceedings of 2024 International Conference on Guidance, Navigation and Control (Volume 18), pp. 314-324.