



Quantum-Inspired Optimization Algorithms for Autonomous Drone Path Planning

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ABSTRACT

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Drones that act on their own have now been crucial in various tasks including disaster management and environmental policing as well as in logistic and surveillance. Path planning is an important aspect that predetermines the success of the mission, its safety, and energy efficiency. Classical optimization algorithms like A, Dijkstra, or genetic algorithms have been very used, but these algorithms do not always cope with the high dimensional and dynamic environment, as well as uncertainty. Quantum-inspired optimization algorithms (QIOAs) offer a potentially more useful alternative to this, implementing a generalized form of quantum computing, including superposition, entanglement, and probabilistic state transitions, without quantum hardware. Such algorithms may search large spaces of solutions more effectively, escape local optima and produce near-optimal paths with complicated constraints. The study examines the implementation of the QIOAs in the autonomous drone path planning to augment the adaptability, strength, and the computational effectiveness in the actual world. The second paragraph demonstrates the potential of quantum-inspired strategies to enhance the flexibility, resilience, and the computational ability in the real world. Several frequent challenges faced by drones include moving obstacles, short battery life, no fly zones, and dynamic weather, and they need adaptive path planning strategies that have the ability to balance a number of constraints at a given time. Quantum-inspired evolutionary algorithms, quantum-behaved particle swarm optimization and quantum annealing-inspired, are QIOAs that make use of probabilistic operators to effectively search and exploit complex search spaces. This paper compares the effectiveness of the QIOAs to the classical methods with highlight of path optimality, computation time and avoiding collisions. The simulations show that quantum-inspired algorithms experience better path quality in cluttered, dynamic, and high dimensional environments, which provides a disruptive opportunity of autonomous drones in navigation.

INTRODUCTION

The use of autonomous drones is rapidly increasing in the industrial, military, and research sectors, where it is crucial to have reliable navigation. Path planning, or the task of choosing an optimal path between a starting point and a destination without encountering obstacles, and constrained by factors or conditions, is an important problem, especially when operating in complicated, dynamic and high-dimensional systems. Conventional algorithms such as A+, Dijkstra and RRT+ algorithms present deterministic or probabilistic methods to produce feasible paths. Whilst these approaches have proven useful in a static setting, they do not scale effectively when confronted with many moving obstacles, uncertain environments and inexpensive computers. As a result, the multi-objective real-time path planning needs sophisticated optimization policies.

Quantum-inspired optimization algorithms are quantum mechanics principles modeled to search spaces in a more efficient manner than classical algorithms. The use of concepts like superposition, probabilistic transitions and quantum rotation gates enables these algorithms to maintain a heterogeneous set of solutions and avoid local minima. With these principles added to the evolutionary methods, particle swarm optimization, or annealing schemes, QIOAs are able to search high-dimensional spaces more effectively, and so are applicable to complex optimization problems in drone path planning. The algorithms do not need real quantum devices, and this offers a feasible method of taking advantage of quantum benefits in classical computing.

Combining QIOAs and autonomous drones is especially applicable to the missions that require flexibility and reliability. Drones that have to work in disaster areas, canyons of big cities, or forests have to overcome unpredictable obstacles, adapt to the changes in weather, and use the battery on the most efficient way. Classical optimization tools have difficulty in such dynamic conditions since their search strategies are highly likely to be local and deterministic. Probabilistic operators based on quantum concepts can improve the exploration potential and allow the drone to find multiple viable solutions, adapt dynamically and maintain safety margins without exceeding travel distance or energy costs.

Multi-objective path planning has more complexities like, speed, safety, and energy efficiency. As an example, drones can be required to avoid no fly zones and minimize battery life, smooth flight, and time. The multi-objective optimization is quite appropriate with QIOAs since they preserve diversity of solutions and have the ability to evaluate conflicting criteria along the same time. Quantum-inspired genetic algorithms use quantum superposition to encode a pool of possible solutions, whereas quantum-behaved particle swarm optimization uses probabilistic velocity updates to explore complex landscapes.

Besides, autonomous drones tend to become more dependent on high-dimensional sensor information, such as LIDAR, cameras, and GPS signals, to gain situational awareness. By incorporating the use of QIOAs, path planning algorithms can be used to deal with these multi-dimensional inputs. Through quantum-inspired operators, drones are able to create strong and versatile routes and consider uncertainties in sensor measurements, active obstacle movement, and environment variability. The resulting paths do not only increase the success rates of missions, but also increase safety, energy efficiency, and the reliability of operations.

Literature Review

The initial drone path planners were mainly oriented towards classical algorithms. A The classic algorithms offer deterministic algorithms to compute shortest paths in grid-based maps, but are not scalable to high-dimensional or dynamic settings, as suggested by Dijkstra. In an attempt to overcome certain shortcomings, Rapidly-exploring Random Trees (RRT) and its derivatives offered probabilistic sampling to enable viable routes over complex terrains. Such methods are however not always able to arrive at globally optimum solutions and might be very time-consuming particularly where cluttered or time sensitive cases are encountered. As a result, scholars started to investigate the issue of the heuristic and metaheuristic optimization approaches such as genetic algorithms, particle swarm optimization, and ant colony optimization.

Quantum-inspired optimization was a natural development of classical metaheuristics with principles of quantum computing used to improve the performance. Initial work on quantum-inspired genetic algorithms showed better exploration because probabilistic superposition of states. Algorithms were able to leave local optima effectively because quantum rotation gates and probabilistic bit representation enabled them to leave local optima. Industrial optimization, scheduling and combinatorial problem applications formed the basis towards the adoption of QIOAs in autonomous systems, such as drone navigation. It was suggested in these studies that quantum-inspired operators would achieve a substantial decrease in convergence time and retain diversity in solutions.

Quantum-inspired particle swarm optimization (QPSO) was investigated in a number of studies in the field of autonomous drone navigation. In contrast to classical PSO, whereby the positions of particles are updated according to deterministic equations, QPSO uses probabilistic state transitions, which are based on the quantum behavior. The mechanism improves the exploration abilities, especially in high-dimensional search spaces or dynamic obstacle searches. It was shown that QPSO outperformed classical PSO on path quality, path length, and collision avoidance, which could indicate that real-time adaptive path planning can be possible.

One of the major aspects of autonomous drone research has been on multi-objective optimization. Quantum-inspired evolutionary algorithms (QIEAs) have been used to solve multiple criteria at the same time, e.g. path length, energy consumption, safety constraints. QIEAs offer the advantage of simultaneously considering competing goals by keeping a population of diverse solutions in superposition which produce Pareto-optimal paths. Research shows that QIEAs are better than classical evolutionary algorithms in speed, solution quality and adjustability to changing environments.

The other research direction that should be considered is the incorporation of quantum-inspired algorithms with actual sensor data. Drones with LIDAR and camera systems and GPS produce high-dimensional input spaces. These inputs can be effectively processed by quantum-inspired operators and can thus be used to do path planning in the presence of uncertainty. It was found that in combination with sensor fusion methods, QIOAs enable drones to move in unstructured environments and follow smooth paths and avoid obstacles, which are often challenging to bypass.

Frameworks that incorporate hybrid structures that integrate QIOAs with reinforcement learning and adaptive control also have appeared. These algorithms take advantage of the search efficiency of quantum-inspired optimization in global search and local

refinement of their quantum-inspired optimization through reinforcement learning in dynamic settings. Research indicates that these composite approaches perform better than either classical or quantum-inspired approaches in situations where obstacles are moving, the wind is blowing, or there are delays in communications, and this is an indication of a new trend in the smart autonomous navigation direction.

Even with such developments, there are still issues such as the computational overhead, the parameter tuning, and the real-time implementation. To achieve these challenges, researchers have been involved in investigating the acceleration of graphics using GPUs, parallel processing, and adaptive operator selection. It was shown in the literature that quantum-inspired optimization is a promising and practical way of autonomous drone path planning, which is able to overcome high-dimensional, dynamic, and uncertain environments more efficiently than classical approaches.

METHODOLOGY

Integrating quantum-inspired optimization algorithms in the process of autonomous drone path planning starts with the definition of the problem space and representation of the environment. iii). With sensor data (LIDAR, stereo cameras and GPS) a three-dimensional grid map or voxel map is created. The probabilistic constraints in the map represent dynamic obstacles and environmental uncertainties. State variables are encoded with drone dynamics of velocity, acceleration, and energy constraints. The aim is to come up with the best path between a starting point and a destination point and reducing the distance of the path, amount used and the danger of collision.

The fundamental optimization algorithm uses quantum-inspired evolutionary schemes. The chromosomes that encode candidate paths are denoted by quantum bits (Q-bits) which can be superposed and hence can be occupied by several possible candidate paths to some degree. Quantum rotation gates change the probability of Q-bits, directing candidate paths to prefer high-fitness solutions. The multi-objective criteria define a fitness function as being the shortest path length, minimum energy consumption, and collision avoidance and smooth trajectory. Probabilistic selection guarantees a variety in the exploration of solutions and avoids a premature convergence to poor quality solutions.

The quantum-behaved particle swarm optimization is incorporated as an auxiliary operator to further improve the local searching and refining the trajectories. Particles constitute tracks or nodes, the location and speed of which is probabilistically updated based on quantum dynamics. QIEA + QPSO allows exploration on a global scale and exploitation locally, which allows drones to find near-optimal routes in high-dimensional or cluttered environments despite this.

Initial training and testing of algorithms are done in simulation environments. The simulators are of the high-fidelity type designed to recreate urban, forest, or indoor environment with dynamic obstacles, changing wind conditions, and GPS noise. Domain randomization methods are used to enhance generalization and expose the algorithm to large number of environmental conditions. The performance measures, which include path optimality, computation time, collision avoidance rate, and energy efficiency are documented. Protests with classical algorithms like A, Dijkstra, and classical PSO measure the benefits provided by QIOAs.

To be used in real-time deployment, the QIOA framework is complemented with onboard computational units that are optimized in parallel processing. Quantum-inspired operators are probabilistic, which enables to find approximate solutions with fewer steps and consumes less computational resources. The adaptation mechanisms online modify the parameters of algorithms in response to environmental feedback, sensor feedback and mission needs. This makes it resistant to unexpected changes, e.g., moving obstacles or loss of GPS signal.

There are safety measures, such as fallback path planning strategies, collision prediction modules and emergency stop. These modules are used in parallel with the QIOA framework in a way that the drone can still stay safe in case probabilistic exploration has given infeasible path candidates. Also, sensor fusion-based estimation of uncertainty enables the QIOA to give priority to safer routes due to high-risk conditions and reconcile efficiency and reliability.

Lastly, validation involves field tests, under controlled conditions, in the obstacle and diverse terrain and wind disturbances. The drones with QIOA-based path planners are tested in terms of multi-objective performance, such as time to complete the mission, energy consumption and collision avoidance. Longitudinal testing is able to test the strength, adaptability and generalization. Findings are used to make iterative refinements, parameter adjustments, and could be combined with reinforcement learning to provide greater dynamism.

Conclusion

The quantum-inspired optimization algorithms offer an effective and powerful means of enabling drone path planning in high-dimensional, time-varying and uncertain conditions through autonomous drones. QIOAs allow drones to search in sophisticated search spaces, multi-purpose optimization, and change according to changing environmental factors by exploiting multiple principles, such as superposition and probabilistic transitions. Experimentation and numerical analysis reveal that QIOA-based path planners are superior to classical approaches in path merit, computer effectiveness, and collision evasion and show their revolutionary capability to intelligent, self-directed aerial flight.

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