

## Scalability Discussion

To handle real-world data from tens of thousands of commercial drones, the current system would need to evolve significantly in terms of architecture, efficiency, and reliability. Below are key points for scaling the system:

### 1. Distributed Computing

- Use distributed systems like Apache Spark or Dask to process data in parallel across multiple machines.
- Each machine could handle a subset of drones, and global coordination logic could resolve cross-region conflicts.

### 2. Real-Time Data Ingestion

- Integrate streaming frameworks like Apache Kafka or Apache Flink to ingest telemetry and mission data from drones in real-time.
- Buffer incoming data streams and manage them with fault-tolerant pipelines.

### 3. Efficient Conflict Checking

- Instead of brute-force  $O(n^2)$  comparisons, use spatial indexing techniques such as k-d trees, R-trees, or geohashing.
- Use time-binned indexing to reduce temporal conflict checks.

### 4. Scalable Storage

- Use cloud storage solutions (e.g., AWS S3, Google Cloud Storage) with time-series databases like InfluxDB or TimescaleDB for efficient querying of historical data.

### 5. Fault Tolerance and Redundancy

- Build the system with fault-tolerant services that can recover from node failure or lost drone signals.
- Use container orchestration tools like Kubernetes for automated recovery and scaling.

## 6. UI and Visualization

- Offload visualization to web-based dashboards that subscribe to the drone telemetry service.
- Use downsampled or region-specific data to prevent overloading the front end.

## 7. AI Integration (Optional)

- Use AI to predict potential conflicts ahead of time based on historical flight patterns.
- Implement adaptive safety buffers based on drone velocity and airspace density.

## 8. Testing Strategy

- Simulate thousands of drones in parallel using synthetic data generators.
- Implement stress tests to measure system throughput, latency, and conflict resolution success rates.

## Conclusion:

A scalable drone conflict detection system requires robust architecture, distributed computing, real-time processing, and efficient spatial/temporal indexing to ensure reliable performance at scale.

## Purpose

The purpose of the collision detection logic is to identify when two or more drones come too close to each other during their missions, violating a defined minimum safe distance (also known as the safety buffer).

This helps ensure spatial deconfliction and supports safe multi-drone operations.

## Where It Is Used

The logic is implemented in the method:

```
def _check_collision(self, pos1, pos2, threshold=0.1):
```

This method is called repeatedly within the animation update loop as each drone's position changes over time.

## Logic Breakdown

### 1. Distance-Based Collision Check

We use the Euclidean distance formula to measure how far apart two drones are at any point in time:

```
np.linalg.norm(np.array(pos1) - np.array(pos2))
```

### 2. Threshold Comparison

A collision is detected if the distance is less than a defined threshold (e.g., 0.1 meters):

```
return np.linalg.norm(np.array(pos1) - np.array(pos2)) < threshold
```

### 3. Checking All Pairs of Drones

Within each update step (in `_update_loop()`), we:

- Compute current positions of all drones.

- Iterate over all unique pairs of drones (using a nested loop).
- Call `_check_collision(pos1, pos2)` for each pair.

If a collision is found, we:

- Log the conflict time
- Log the collision location
- Identify the two involved drones
- Optionally trigger a callback function (to update UI or report table)

### Example

If Drone A is at (1.0, 0.0, 0.0) and Drone B is at (1.05, 0.0, 0.0), and the threshold is 0.1, the system detects a collision, because their distance is 0.05.

### Code Snippet in Context

```
def _check_collision(self, pos1, pos2, threshold=0.1):
    return np.linalg.norm(np.array(pos1) - np.array(pos2)) < threshold
```

Called from `_update_loop()`:

```
for i in range(len(positions)):
    for j in range(i + 1, len(positions)):
        if self._check_collision(positions[i], positions[j]):
            if self.collision_callback:
                self.collision_callback(
                    self.trajectory_names[i],
                    self.trajectory_names[j],
                    positions[i],
                    elapsed
                )
```

## Benefits

- Efficient: Only checks at current time step, no unnecessary future predictions.
- Simple: Easy to understand, modify, or extend.
- Real-Time Ready: Can be called every frame in an animation or during live simulation.