A Tactical Communication Protocol Link32 and Skynet Reference Implementation for Affordable LTE 4G-based Drone Swarm Coordination in Cluster Hybrid Topologies

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Abstract

Coordinating large-scale drone swarms in battlefield or demonstration scenarios is a complex challenge, particularly under constraints of cost, scalability, and communication reliability. This article addresses the problem of designing an affordable, LTE 4G-based swarm coordination framework for micro air vehicles (MAVs) using cluster hybrid topologies, with low-resolution video streams from cluster lead drones for command-and-control (C2C) monitoring. We propose a hybrid architecture combining terrestrial and UAV-based base stations, leveraging low-bandwidth payloads (100 bytes at 10 Hz) and minimal video streaming (200 kbps per cluster lead). Key algorithms for path planning, collision avoidance, task allocation, formation control, and communication optimization are reviewed, drawing from bio-inspired and AI-driven methods. The solution supports approximately 5,000 drones over a 30 km radius, balancing cost, scalability, and performance.

Link32 is a tactical communication protocol designed for low-latency, secure, and scalable data exchange in contested environments, enabling swarm drone coordination, real-time position location information (PLI), command and control (C2), and tactical chat over UDP-based multicast networks. Skynet, its reference implementation, is a lightweight C99 framework inspired by LTE's QoS Class Identifier (QCI) framework, tailored for military applications and resource-constrained devices such as drones and embedded systems. With minimal dependencies and a focus on portability, Skynet supports dynamic TDMA slot management, AES-256-GCM encryption, and lock-free concurrency, ensuring robust performance in mission-critical scenarios.

Contents

1 Introduction

Link32 is a tactical communication protocol inspired by standards such as VMF, TSM, SRW, and MQTT [4], LTE (HSS, PDCP, MME, QCI), and TDL [1] with [2], tailored for military applications requiring robust, low-latency, and secure data exchange in contested environments. It facilitates swarm drone coordination, real-time position location information (PLI), command and control (C2), and tactical chat using UDP-based multicast networks.

Skynet is the reference server implementation of Link32, developed in C99 with minimal dependencies to ensure portability and performance on resource-constrained devices. Its convergence layer draws inspiration from LTE's QoS Class Identifier QCI framework [3], enabling granular QoS control and dynamic resource allocation for mission-critical communications.

The proliferation of micro air vehicles (MAVs) has enabled applications such as battlefield surveillance, search and rescue, and large-scale demonstrations. Coordinating thousands of drones requires robust architectures and algorithms to manage communication, navigation, and task allocation in dynamic environments. Traditional centralized systems face scalability issues, while high-cost 5G solutions are impractical for affordable deployments. This article proposes an affordable, LTE 4G-based swarm coordination framework using cluster hybrid topologies, where each drone transmits a 100-byte payload (position, velocity, tactical data) at 10 Hz, and 5–10 cluster lead drones stream low-resolution video (200 kbps) for C2C monitoring. We focus on a 30 km radius operational area, addressing the challenges of bandwidth, latency, and cost through hybrid architectures and optimized algorithms.

2 Problem Statement

The problem is to design a cost-effective LTE 4G-based communication and coordination system for a swarm of approximately 5,000 MAVs operating over a 30 km radius in a battlefield or demonstration setting. Each drone sends a 100-byte payload (position, velocity, tactical data) at 10 Hz, requiring approximately 9.6 kbps per drone, while 5–10 cluster lead drones stream low-resolution video at 200 kbps each for C2C monitoring. The total uplink bandwidth is approximately 50 Mbps. The system must:

- Ensure reliable communication within LTE 4G constraints (50 Mbps uplink, 20–100 ms latency).

- Achieve 30 km coverage using minimal, low-cost equipment.

- Support scalable coordination for path planning, collision avoidance, task allocation, and formation control.

- Maintain affordability through open-source or commercial off-the-shelf (COTS) components.

Challenges include limited bandwidth, latency constraints, signal interference in battlefield environments, and the need for decentralized control to scale to thousands of drones.

3 Cluster Hybrid Topology

To address the problem, we propose a cluster hybrid topology combining a terrestrial LTE 4G base station with one or more UAV-based aerial base stations (ABSs). The architecture is designed for affordability and scalability, leveraging open-source solutions and COTS hardware.

3.1 Terrestrial Base Station

A single terrestrial base station, implemented using srsRAN with a LimeSDR Mini (cost: ∼$1,500), provides core coverage of 10–15 km. A high-gain omnidirectional antenna (e.g., Laird FG24008, ∼$150) and a 20 W RF amplifier (∼$500) extend the range toward 20 km. The base station connects to an open-source Evolved Packet Core (EPC) like Open5GS on a low-cost server (∼$500), handling up to 2,000 simultaneous connections.

3.2 UAV-Based Aerial Base Station

One custom-built UAV (cost: ∼$2,000) equipped with a LimeSDR Mini (∼$300) and a lightweight antenna (e.g., Taoglas GW.26, ∼$50) serves as an aerial base station at 100–120 m altitude, extending coverage to 20–30 km via line-of-sight (LoS) propagation. The UAV rotates with spares to ensure continuous operation, supported by a portable generator (∼$1,100) and battery swap system (∼$300).

3.3 Cluster-Based Swarm Organization

The swarm is organized into clusters of 50–100 drones, each led by a cluster head responsible for intra-cluster coordination and LTE communication with the base station. Cluster heads transmit 100-byte payloads (9.6 kbps) and, for 5–10 designated leaders, low-resolution video streams (200 kbps). Drone-to-drone (D2D) communication via LTE sidelink reduces network load, enabling scalability to 5,000 drones within the 50 Mbps uplink capacity.

3.4 Backhaul and Ground Control

A point-to-point microwave link (e.g., Ubiquiti airFiber 24, ∼$1,500) or satellite terminal (e.g., Starlink, ∼$600) provides backhaul to a command center. A rugged laptop with open-source software like QGroundControl (∼$500) manages UAV flight and network configuration. Total cost is approximately $8,400, making the system affordable for battlefield or demonstration use.

4 Swarm Coordination

The coordination of 5,000 MAVs requires algorithms for path planning, collision avoidance, task allocation, formation control, and communication optimization. These algorithms are optimized for low computational overhead and LTE 4G constraints, drawing from bio-inspired and AI-driven approaches. Below, we describe each algorithm, highlighting its essence and relevance to the proposed architecture.

4.1 Path Planning and Navigation

Effective path planning ensures drones navigate efficiently in complex environments, balancing computational simplicity with adaptability to dynamic conditions.

4.1.1 A* Algorithm

The A* algorithm is a cornerstone of path planning, leveraging a graph-based approach to find the shortest path from a drone's current position to its target. By combining heuristic estimates with actual costs, A* ensures optimal trajectories while integrating with genetic algorithms for real-time optimization in cluttered environments, such as battlefields with obstacles. Its computational efficiency suits micro drones with limited processing power [9, 10].

4.1.2 Particle Swarm Optimization (PSO)

Inspired by the flocking behavior of birds, PSO optimizes drone trajectories by treating each drone as a particle exploring a solution space. Particles adjust their paths based on local (individual best) and global (swarm best) solutions, making PSO ideal for target tracking and search missions in dynamic settings. Its distributed nature aligns with cluster-based topologies [11, 12].

4.1.3 Ant Colony Optimization (ACO)

ACO mimics the pheromone trails of ants to discover optimal paths in Flying Ad Hoc Networks (FANETs). Drones share virtual pheromone data to guide routing and task allocation, enabling efficient navigation in large-scale swarms. Its robustness to dynamic changes suits battlefield scenarios with intermittent connectivity [13, 14].

4.1.4 Differential Evolution

Differential Evolution dynamically adjusts swarm coordination parameters, such as trajectory weights, by evolving a population of candidate solutions. This adaptability ensures drones respond to environmental changes, optimizing paths in real-time for tasks like formation control. Its lightweight computation is suitable for micro drones [15, 16].

4.1.5 Kalman Filter

The Kalman Filter estimates drone positions and velocities by fusing noisy sensor data, critical for navigation in GPS-denied environments like urban canyons or forests. Its recursive nature minimizes computational load, enabling robust localization in large swarms [17, 18].

4.1.6 Spatial-Temporal Joint Optimization

This approach synchronizes trajectory shapes and timing to optimize navigation in cluttered environments. By jointly considering spatial paths and temporal constraints, it ensures efficient coordination for tasks like surveillance, minimizing energy use and collisions [19, 20].

4.2 Collision Avoidance

Collision avoidance is critical for dense swarms, ensuring safe navigation without excessive computational or communication demands.

4.2.1 Artificial Potential Field (APF)

APF treats drones as particles in a potential field, repelled by obstacles and neighbors while attracted to targets. This intuitive method generates smooth trajectories with low computational cost, though it risks local minima where drones become static. It suits resource-constrained MAVs [21, 53].

4.2.2 Reactive Collision Avoidance

Using onboard sensors like time-of-flight, reactive collision avoidance enables drones to detect and avoid obstacles in real-time. Its simplicity and low power requirements make it ideal for micro drones in dynamic battlefield environments [23, 24].

4.2.3 Flocking Algorithms

Based on Reynolds' rules (cohesion, separation, alignment), flocking algorithms ensure drones maintain safe distances while moving cohesively. This bio-inspired approach scales well for large swarms, supporting cluster-based coordination [25, 26].

4.2.4 Vision-Based Avoidance

Vision-based avoidance uses cameras to detect neighbors and obstacles, offering precise spatial awareness. Despite limitations from field-of-view and lighting, it enhances safety in dense swarms when paired with lightweight sensors [27, 28].

4.3 Task Allocation and Coordination

Task allocation ensures drones efficiently distribute mission responsibilities, adapting to dynamic conditions with minimal communication.

4.3.1 Deep Reinforcement Learning (DRL)

DRL, using algorithms like Proximal Policy Optimization (PPO) and Deep Q-Network (DQN), enables drones to learn optimal task allocation and path planning through environmental interaction. Its adaptability excels in dynamic settings like battlefield patrolling [29, 30].

4.3.2 Stigmergy

Inspired by social insects, stigmergy allows drones to leave virtual "pheromones" (data markers) to guide others toward tasks, reducing communication overhead. This decentralized approach suits LTE-constrained swarms [31, 32].

4.3.3 Consensus Algorithms

Consensus algorithms ensure drones agree on shared goals or states, even under poor LTE connectivity. They are critical for merging swarms or maintaining coordination in disrupted environments [33, 34].

4.3.4 Virtual Navigator Model

This model dynamically adjusts patrol paths based on environmental changes, enhancing flexibility for tasks like surveillance. It integrates with cluster-based topologies for scalable coordination [35, 36].

4.4 Formation Control

Formation control maintains swarm geometry, balancing precision with robustness to disruptions.

4.4.1 Leader-Follower

In leader-follower formations, a cluster head guides followers, maintaining relative positions. Its simplicity suits small clusters but is vulnerable to leader failure [37, 38].

4.4.2 Virtual Structure

The virtual structure approach treats the swarm as a rigid geometric shape, assigning each drone a fixed position. It ensures precise formations but lacks flexibility in dynamic environments [39, 40].

4.4.3 Behavior-Based

Behavior-based methods use simple rules (e.g., avoid collisions, stay near neighbors) to achieve emergent formations. Their scalability makes them ideal for large swarms [41, 42].

4.4.4 Consensus-Based

Consensus-based methods share state information to maintain formations, robust to communication disruptions. They suit LTE 4G environments with variable connectivity [43, 44].

4.4.5 Graph-Based Models

Graph-based models represent drones as vertices and communication links as edges, enabling flexible formation adjustments. They support dynamic reconfiguration in cluster-based swarms [45, 46].

4.5 Communication Optimization

Communication optimization ensures efficient data exchange within LTE 4G constraints, minimizing latency and energy use.

4.5.1 Reactive-Greedy-Reactive (RGR) Protocol

RGR combines Ad Hoc On-Demand Distance Vector (AODV) with Greedy Geographic Forwarding (GGF) to reduce latency in FANETs. It adapts to dynamic topologies, ensuring reliable packet delivery [47, 48].

4.5.2 Bee Colony Optimization (BCO)

BCO mimics bee foraging to optimize routing in FANETs, balancing exploration and exploitation for efficient communication in large swarms [49, 50].

4.5.3 EPOS

EPOS is a decentralized multi-agent learning algorithm that optimizes energy consumption and task allocation for spatio-temporal sensing, ideal for LTE-constrained swarms [51, 52].

4.5.4 Graph Attention-Based Decentralized Actor-Critic

This method uses graph neural networks to enable dual-objective control (task completion, energy efficiency), enhancing communication efficiency in cluster-based topologies [53, 54].

4.6 Principles

Link32 adheres to the following design principles:

- **Threat Model**: Confidentiality and integrity, no non-repudiation.

- **Latency**: Microsecond-precision timings .

- **Implementation**: Written in C99 for portability and performance.

- **Key Provisioning**: Key distribution for controlled setup.

- **Mandatory Encryption**: All messages encrypted with AES-256-GCM.

- **Node Identification**: Node names hashed to 32-bit using FNV-1a.

- **Lock-Free Design**: Uses atomic compare-and-swap for concurrency.

- **Topic Architecture**: Topics map to IP multicast groups.

- **Queue Management**: Global network queue, per-topic subscribes.

- **Key Storage**: Separate key stores per executable.

4.7 Properties

Link32 and Skynet are designed with the following properties:

- **Message Size**: 32-byte header (48 bytes payload with GCM tag).

- **Security**: ECDH key exchange SECP384R1, Mandatory AES-256-GCM.

- **Non-blocking**: Using monotonic clocks and non-blocking I/O.

- **Footprint**: ∼64KB L1 cache usage, ∼2000 lines of code (LOC).

- **Swarm Scalability**: Supports large-scale drone swarms.

- **Dependencies**: Single dependency on OpenSSL for cryptography.

5 Link32 Protocol

5.1 S-Message Format

The Link32 message structure, `SkyNetMessage`, is compact for large-scale swarm communication:

```
typedef struct {
  uint8_t version : 4;    // Protocol version (current: 1)
  uint8_t type : 4;       // Message type (0-6)
  uint8_t qos : 4;        // Quality of Service (0-3)
  uint8_t hop_count : 4;  // Hop count for routing (0-15)
  uint32_t npg_id;        // Topic identifier (1-103)
  uint32_t node_id;       // Sender node ID (FNV-1a hash)
  uint32_t seq_no;        // Sequence number for deduplication
  uint8_t iv[16];         // AES-256-GCM initialization vector
  uint16_t payload_len;   // Payload length (0-32767)
  uint8_t payload[MAX_BUFFER];    // GCM-TAG
} SkyNetMessage;
```

- **Header Size**: 32 bytes.

- **Minimum Payload Size**: 48 bytes minimum (32-byte + GCM tag).

- **Maximum Payload Size**: Up to 32720 bytes.

5.2 Message Types

The protocol defines seven message types, as shown in Table 1.

Table 1: Link32 Message Types

Type	Name	Description
0	Key Exchange	Exchanges ECC public keys for ECDH session setup.
1	Slot Request	Requests a TDMA slot from the server.
2	Chat	Sends tactical chat messages.
3	Ack	Acknowledges slot assignments or control messages.
4	Waypoint	Specifies navigation waypoints for C2.
5	Status	Reports position, velocity, or sensor data (e.g., PLI).
6	Formation	Coordinates swarm formations.

5.3 Multicast Topics

Link32 uses multicast topics mapped to IP multicast groups, as in Table 2.

Table 2: Multicast Topics

NPG	Name	Multicast	Purpose
1	npg_control	239.255.0.1	Kkey exchange, Slot requests.
6	npg_pli	239.255.0.6	Position information (status).
7	npg_surveillance	239.255.0.7	Sensor data forwarding.
29	npg_chat	239.255.0.29	Tactical chat and acks.
100	npg_c2	239.255.0.100	C2 (waypoints, formations).
101	npg_alerts	239.255.0.101	Network alerts and self-healing.
102	npg_logistics	239.255.0.102	Logistical coordination.
103	npg_coord	239.255.0.103	Swarm Coordination.

5.4 Slot Management

Link32 employs a Time Division Multiple Access (TDMA)-like slot manager to minimize collisions:

- **Slot Array**: Fixed-size (256) array `slots`.

- **Dynamic Topics**: Each slot creates a temporary multicast group.

- **Allocation**: First-come, first-serve with no timeouts.

- **Timing**: Slots cycle every `TIME_SLOT_INTERVAL_US=1000`μ`s`.

Clients send `SKYNET_MSG_SLOT_REQUEST` to NPG 1, and the server assigns slots via `SKYNET_MSG_ACK`.

5.5 Deduplication

A fixed-size circular buffer (`seq_cache`) prevents message loops:

- **Structure**: Stores `node_id`, `seq_no`.

- **Memory**: ~16KB (1024 × 16 bytes).

- **Complexity**: O(1) lookup using FNV-1a hashing.

- **Threshold**: Discards duplicates within 1 second.

5.6 Security

Security mechanisms include:

- **Key Exchange**: ECDH over secp384r1 for 256-bit AES keys.

- **Encryption**: AES-256-GCM with 16-byte IV and 16-byte tag.

- **Key Storage**: Per Process in `ec_priv` and `ec_pub`.

- **Key Derivation**: HKDF-SHA256 for AES keys.

- **Self-messages**: Skips (drops) messages.

5.7 Subscriptions

Nodes subscribe to topics based on roles, as shown in Table 3:

Table 3: Role-Based Subscriptions

Role	NPGs	Purpose
Infantry	1, 29	Network control and tactical chat.
Drone	1, 6, 7, 100, 101	C2, PLI, surveillance, alerts.
Air	1, 6, 7, 100, 101, 103	C2, PLI, surveillance, alerts, coord.
Sea	1, 7, 29, 102, 103	C2, surveillance, chat, logistics, coord.
Ground	1, 7, 29, 102	C2, surveillance, chat, logistics.
Relay	1, 6, 101	C2, PLI, alerts for relaying.
Controller	1, 6, 100, 101	C2, PLI, alerts for command posts.

6 Skynet Implementation

6.1 Dependencies

- **OpenSSL**: For ECC, ECDH, and AES-256-GCM.

- **C99 Compiler**: GCC or equivalent.

- **POSIX Environment**: For threading, epoll, and timerfd.

6.2 Build

To build Skynet:

```
# git clone git@github.com:BitEdits/skynet
# cd skynet
# cc -o skynet_client skynet_client.c skynet_proto.c -lcrypto
# cc -o skynet skynet.c skynet_proto.c skynet_conv.c -lcrypto
```

6.3   Installation

The provisioning script `skynet.sh` generates ECC key pairs:

```
# ./skynet.sh
Generated keys for node npg_control (hash: 06c5bc52) in /secp/
Generated keys for node npg_pli (hash: c9aef284) in /secp/
Generated keys for node npg_surveillance (hash: 4d128cdc) in /secp/
Generated keys for node npg_chat (hash: 9c69a767) in /secp/
Generated keys for node npg_c2 (hash: 89f28794) in /secp/
Generated keys for node npg_alerts (hash: 9f456bca) in /secp/
Generated keys for node npg_logistics (hash: 542105cc) in /secp/
Generated keys for node npg_coord (hash: e46c0c22) in /secp/
Generated keys for node server (hash: 40ac3dd2) in /secp/
Generated keys for node client (hash: 8f929c1e) in /client/secp/
# cp /secp/*.ec_pub /client/secp/
```

6.4   Server Operation

The server binds to UDP port 6566, joins multicast groups, and processes messages using a global queue. Example output:

```
# skynet server
Node 40ac3dd2 bound to 0.0.0.0:6566.
Joined multicast group 239.255.0.1 (NPG 1: control).
Joined multicast group 239.255.0.6 (NPG 6: PLI).
Message received, from=8f929c1e, to=1, size=231.
Decryption successful, from=8f929c1e, to=1, size=215.
Saved public key for client 8f929c1e.
Assigned slot 0 to node 8f929c1e.
Message received, from=8f929c1e, to=6, size=40.
Decryption successful, from=8f929c1e, to=6, size=24.
Message sent from=8f929c1e, to=6, seq=3, multicast=239.255.1.0.
```

6.5   Client Operation

The client joins topic-specific multicast groups and sends key exchange, slot requests, and status messages. Example output:

```
# skynet_client client
Node 8f929c1e connecting to port 6566.
Joined multicast group 239.255.0.1 (NPG 1).
Joined multicast group 239.255.0.6 (NPG 6).
Sent key exchange message to server.
Sent slot request message to server.
Received slot assignment: slot=0.
Joined slot group 2399540.1.
Sent status message: multicast=239.255.1.0,
                     pos=[0.1, 0.1, 0.1],
                     vel=[0.0, 0.0, 0.0],
                     seq=2.
```

6.6 Usage

Skynet includes five utilities:

Keys Provisioning

Generates ECC key pairs.

```
skynet_keygen <node> [--server|--client]
```

Message Encryption

Encrypts a test message to `<npg_id>.sky`.

```
skynet_encrypt <sender> <recipient> <file>
```

Message Decryption

Decrypts `<file.sky>`.

```
skynet_decoder <sender> <recipient> <file.sky>
```

Skynet Server

Runs the server with FNV-1a hashed `<node>`.

```
skynet <node>
```

Skynet Client

Runs the client with FNV-1a hashed `<node>`.

```
skynet_client <node>
```

6.7 Limitations

- **Slot Scalability**: Fixed `SLOT_COUNT=256` limits nodes to 256.

- **No Retransmission**: Dropped messages are not retransmitted.

- **Key Management**: Manual public key copying required.

- **Deduplication**: `SEQ_CACHE_SIZE=1024` may cause collisions.

7 Convergence Architecture

The Skynet system's convergence layer employs a 3-level structural hierarchy comprising Quality of Service (QoS), Bearer, and Entity components. This design is driven by the need for granular QoS control, per-node resource management, reliable packet delivery, dynamic slot allocation, and scalability in a Time Division Multiple Access (TDMA)-based tactical network. Inspired by the Long-Term Evolution (LTE) QoS Class Identifier (QCI) framework, the hierarchy ensures military-grade performance, including latency below 50ms for command-and-control (C2) traffic and reliable delivery for critical communications, such as swarm drone coordination. Compared to alternative designs, such as flat or centralized structures, this approach excels in dynamic, resilient scenarios, providing robust and scalable QoS management.

7.1 Flat QoS Structure

A flat QoS structure assigns slots directly to Network Protocol Groups (NPGs) based on their QoS profiles, without intermediate bearer or entity layers. For example, an NPG like SKYNET_NPG_C2 might be statically allocated three slots with QoS level 3. While simpler and requiring less memory, this approach lacks per-node isolation, making it unsuitable for dynamic networks with multiple nodes sharing the same NPG. It also complicates reliable delivery, as there is no mechanism for per-flow reordering or sequence tracking. Additionally, static slot assignments cannot adapt to node arrivals or departures, leading to inefficient resource utilization. The 3-level hierarchy overcomes these limitations by introducing bearers for flow isolation and entities for node-level coordination, enabling dynamic and scalable QoS management.

```
typedef struct {
    uint32_t npg_id;
    uint8_t qos;
    uint32_t slot_count;
    uint32_t slot_ids[MAX_QOS_SLOTS];
    uint8_t priority;
} QoSSlotAssignment;
```

7.2 Hierarchical QoS Structure

The 3-level hierarchy separates concerns into QoS, Bearer, and Entity layers, each addressing specific aspects of network management. The SkyNetBearerQoS structure defines QoS parameters such as priority (1–15, where 1 is highest), delay budget (in milliseconds), reliability (best-effort or reliable), and minimum TDMA slots, allowing precise service differentiation. The SkyNetBearer represents a logical communication channel for a node-NPG pair, encapsulating QoS parameters, assigned slots, and state for reordering and reliability. The SkyNetConvergenceEntity aggregates bearers for a single node, managing slot requests and coordinating resource allocation. This modular design ensures

scalability, flexibility, and robustness, outperforming flat structures by providing flow isolation and dynamic adaptation.

```
typedef struct {
    uint8_t priority;        // 1-15 (1 = highest)
    uint32_t delay_budget_ms;// Delay tolerance (ms)
    uint8_t reliability;     // 0 (best-effort), 1 (reliable)
    uint32_t min_slots;      // Minimum TDMA slots
} SkyNetBearerQoS;

typedef struct {
    uint32_t bearer_id;      // Unique bearer ID
    SkyNetBearerQoS qos;     // QoS parameters
    uint32_t node_id;        // Owning node
    uint32_t npg_id;         // Associated NPG
    uint32_t assigned_slots[SKYNET_MAX_SLOTS]; // Assigned slot IDs
    uint32_t slot_count;     // Number of assigned slots
    SkyNetMessage reorder_queue[SKYNET_REORDER_SIZE];
    uint32_t expected_seq_no;// Next expected sequence number
    uint32_t last_delivered; // Last delivered sequence number
    uint64_t last_reorder_time_us; // Last reorder check
} SkyNetBearer;

typedef struct {
    SkyNetBearer bearers[SKYNET_MAX_BEARERS]; // Active bearers
    uint32_t bearer_count;   // Number of active bearers
    atomic_uint slot_requests_pending; // Pending slot requests
} SkyNetConvergenceEntity;
```

7.3 Granular QoS Control

Tactical networks handle diverse traffic types, such as C2, position location information (PLI), and chat, each with distinct latency, reliability, and bandwidth requirements. A uniform QoS approach fails to meet these needs. The SkyNetBearerQoS structure enables granular control by defining specific parameters for each bearer. For instance, SKYNET_NPG_CONTROL (NPG 1, QoS 3) is assigned three slots with a low delay budget to ensure timely C2 delivery, while SKYNET_NPG_CHAT (NPG 103, QoS 0) receives one slot for best-effort traffic. This differentiation, inspired by LTE QCI, guarantees that high-priority traffic meets stringent military requirements, such as sub-50ms latency for C2, while optimizing resource allocation for lower-priority flows.

7.4 Per-Node and Per-Flow Resource Management

Each node in the network may support multiple concurrent flows (e.g., C2, PLI, control) with varying QoS needs. Without isolation, these flows compete for resources, risking contention and degraded performance. The SkyNetBearer structure provides flow-level isolation by associating each bearer with a specific node-NPG pair, tracking its assigned slots and QoS parameters. The SkyNetConvergenceEntity groups all bearers for a node, enabling centralized

resource management and preventing one flow from starving others. This per-node and per-flow approach ensures efficient slot allocation, supports multiple simultaneous communications, and scales to accommodate dynamic network topologies.

7.5 Reliability and Reordering

TDMA networks may deliver packets out of order due to slot scheduling or retransmissions, particularly for reliable traffic (e.g., `reliability=1`). The bearer includes a `reorder_queue` and tracks `expected_seq_no` and `last_delivered` to reorder packets and ensure reliable delivery. The entity coordinates reorder checks across bearers using `last_reorder_time_us`, minimizing overhead. This mechanism is critical for applications requiring guaranteed delivery, such as C2 or control messages, and enhances robustness compared to flat structures, which lack per-flow reordering capabilities.

7.6 Dynamic Slot Allocation

Tactical networks operate in dynamic environments where nodes join or leave, and traffic patterns shift. Static slot assignments are inefficient and inflexible. The entity tracks `slot_requests_pending` and manages bearer slot assignments via `assigned_slots` and `slot_count`. The bearer parameter `min_slots` ensures minimum resource guarantees, while the entity facilitates dynamic re-allocation through the `skynet_convergence_schedule_slots` function. Weighted Fair Queuing, driven by bearer priorities, optimizes slot distribution, ensuring high-QoS traffic receives preferential treatment. This adaptability is a key advantage over static or centralized designs.

7.7 Scalability and Modularity

A flat QoS structure becomes unwieldy as the number of nodes and NPGs increases, complicating scheduling and state management. The 3-level hierarchy addresses this by separating concerns: QoS defines service requirements, bearers manage flow-specific state, and entities coordinate node-level convergence. This modular design scales to support large networks, simplifies debugging, and facilitates maintenance. The entity layer aggregates bearer state, reducing scheduling complexity from $O(n)$ for n bearers to $O(m)$ for m nodes. The hierarchy also supports future enhancements, such as preemption or adaptive QoS, without requiring a system overhaul.

7.8 LTE QCI Compatibility

The 3-level hierarchy draws inspiration from LTE's QCI framework, which uses bearers with QoS profiles to manage diverse traffic types (e.g., VoIP, video, best-effort) per User Equipment (UE). By adapting this model to TDMA, Skynet replaces LTE's EPS bearers with TDMA slot-based bearers, leveraging telecom

best practices. The bearer parameters mirror QCI attributes, such as priority and delay budget, ensuring compatibility with established standards. This alignment reduces design risk, enhances interoperability with telecom systems, and provides a familiar framework for engineers, making it easier to develop and maintain the system.

7.9 MQTT Compatibility

Skynet's convergence layer maps its QoS levels to MQTT's QoS semantics (ISO/IEC 20922:2016) and LTE's QCI framework (3GPP TS 23.203, Release 15), enabling compatibility with both standards for tactical communications. MQTT defines three QoS levels:

- **QoS 0 (At Most Once)**: Best-effort delivery with no acknowledgment or retransmission. Suitable for non-critical data like tactical chat where packet loss is tolerable.

- **QoS 1 (At Least Once)**: Guarantees at least one delivery with acknowledgment (ACK), allowing duplicates. Used for data requiring delivery, such as position location information (PLI), where duplicates are acceptable.

- **QoS 2 (Exactly Once)**: Ensures exactly one delivery via a four-way handshake (`PUBLISH`, `PUBREC`, `PUBREL`, `PUBCOMP`). Critical for command and control (C2) messages requiring no loss or duplication.

Skynet implements these using its `SkyNetBearerQoS` and `SkyNetBearer` structures, configured via `skynet_convergence_init`. QoS 0 skips `reorder_queue` and ACKs, QoS 1 uses ACKs (`SKYNET_MSG_ACK`) with retransmission on timeout (100ms), and QoS 2 employs a handshake with strict ordering and deduplication via `seq_no`. All Network Protocol Groups (NPGs) and message types are mapped to QoS levels, as shown in Table 4.

Table 4: Skynet QoS Mapping to MQTT and LTE QCI

QoS	Priority	NPG	QCI	Budget	Slots	Type
0	15	29	QCI 9	300ms	1	2
0	14	102	QCI 9	300ms	1	2, 5
0	13	103	QCI 9	300ms	1	2, 4
1	7	6	QCI 7	100ms	2	5
1	6	7	QCI 7	100ms	2	5
2	3	1	QCI 5	50ms	3	0, 1, 3
2	2	100	QCI 3	50ms	3	3, 4, 6
2	1	101	QCI 5	50ms	3	3, 6

This mapping ensures military-grade performance: QoS 2 meets sub-50ms latency for C2 (`npg_c2`, NPG 100) and control (`npg_control`, NPG 1), QoS 1 supports reliable PLI (`npg_pli`, NPG 6) with 100ms latency, and QoS 0 optimizes

bandwidth for chat (`npg_chat`, NPG 29). The `SkyNetConvergenceEntity` dynamically allocates slots via Weighted Fair Queuing, prioritizing high-QoS bearers, while `reorder_queue` ensures reliability for QoS 1 and 2, aligning with LTE QCI's packet error loss rates (e.g., 10^{-6} for QCI 5).

In conclusion, the 3-level hierarchy, inspired by LTE QCI, enables Skynet to replicate MQTT's QoS semantics, ensuring scalable, low-latency, and reliable communication for swarm drone coordination and other tactical applications.

8 Dedication

The Skynet is dedicated to Vitalii Karvatskyi.

9 Conclusion

Link32 and Skynet provide a robust framework for tactical communication, combining low-latency, security, and scalability for military applications, including swarm drone coordination. Future improvements could address slot scalability beyond 256 nodes, automated key distribution to replace manual provisioning, retransmission mechanisms for dropped messages, and enhanced deduplication to support larger networks.

The proposed LTE 4G-based cluster hybrid topology provides an affordable, scalable solution for coordinating 5,000 MAVs over a 30 km radius, with each drone transmitting a 100-byte payload at 10 Hz and 5–10 cluster leads streaming 200 kbps video for C2C monitoring. The hybrid architecture, combining a terrestrial base station and UAV relay, achieves full coverage at a cost of approximately \$8,400, leveraging open-source srsRAN and COTS hardware. Key algorithms, including PSO, DRL, and consensus-based methods, enable efficient path planning, collision avoidance, task allocation, formation control, and communication optimization. Future work should explore 5G integration for higher capacity and anti-jamming techniques for battlefield reliability.

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