

SECTION 3: COMPREHENSIVE MATHEMATICAL FORMULAS & CALCULATION ALGORITHMS

3.1 RSSI (Received Signal Strength Indicator) - Complete Analysis

3.1.1 RSSI Physics & Decibel Mathematics

Definition: RSSI is the logarithmic power of radio waves received at the antenna, measured in dBm (decibels relative to 1 milliwatt).

Mathematical Foundation:

$$\begin{aligned}\text{RSSI (dBm)} &= 10 \times \log (\text{Power in Watts} / 0.001 \text{ Watts}) \\ &= 10 \times \log (\text{Power_Watts}) + 30\end{aligned}$$

Inverse formula (convert RSSI back to power):

$$\text{Power (Watts)} = 10^{(\text{RSSI} - 30) / 10}$$

$$\text{Power (Watts)} = 0.001 \times 10^{(\text{RSSI}/10)}$$

Why Logarithmic?

Radio signals span ~15 orders of magnitude in power (from femtowatts to miliwatts). Linear representation would require handling numbers from 10^1 to 10^{15} . Logarithmic compression makes this humanly readable (-120 to -30 dBm) and aligns with human perception (10 dB = 10× power change).

Example Calculations:

Scenario 1: Victim very close to antenna

Physical power received = 1 W = 10 watts

$$\text{RSSI} = 10 \times \log (10) + 30 = 10 \times (-6) + 30 = -30 \text{ dBm}$$

Scenario 2: Victim moderate distance

Physical power = 1 nW = 10⁻⁹ watts

$$\text{RSSI} = 10 \times \log (10^{-9}) + 30 = 10 \times (-9) + 30 = -60 \text{ dBm}$$

Scenario 3: Victim far away

Physical power = 1 pW = 10^{-12} watts

$$\text{RSSI} = 10 \times \log (10^{-12}) + 30 = 10 \times (-12) + 30 = -90 \text{ dBm}$$

Scenario 4: Barely detectable (receiver sensitivity limit)

Physical power = 0.2 pW = 2×10^{-13} watts

$$\text{RSSI} = 10 \times \log (2 \times 10^{-13}) + 30 = 10 \times (-12.7) + 30 = -97 \text{ dBm}$$

LoRa Signal Strength Ranges for FalconResQ:

RSSI (dBm)	Description	Signal Quality	Packet Loss Rate
------------	-------------	----------------	------------------

-30	Extremely close	Excellent	0%
-50	Very close (<5m)	Excellent	<0.1%
-70	Close (5-50m)	Good	<1%
-85	Medium (50-200m)	Fair	5-10%
-100	Far (200-1000m)	Poor	20-40%
-120	Very far (1-10km)	Very Poor	60-95%
-137	Receiver limit	Unusable	>95%

3.1.2 Signal Strength Scoring & Thresholds (config.py)

FalconResQ Configuration:

```
RSSI_STRONG_THRESHOLD = -70 dBm # Green indicator - good communication
RSSI_WEAK_THRESHOLD = -85 dBm # Red indicator - critical signal
RSSI_RANGE = 15 dBm # Distinguishing three categories
```

Signal Quality Score Function:

```
def get_signal_quality_score(rssi: int,
                             rssi_strong: int = -70,
                             rssi_weak: int = -85) -> float:
    """
    Convert RSSI value to normalized score (0-100).
    
```

Mathematical model: Linear interpolation between thresholds

```
score(rssi) = {
    100,                                     if rssi  rssi_strong
    (rssi - rssi_weak) / (rssi_strong - rssi_weak) * 100,   if rssi_weak < rssi < rssi_strong
    0,                                         if rssi  rssi_weak
}

if rssi >= rssi_strong:
    return 100.0
elif rssi <= rssi_weak:
    return 0.0
else:
    # Linear interpolation in dBm space
    range_dbm = rssi_strong - rssi_weak # = 15 dBm
    distance_from_weak = rssi - rssi_weak
    score = (distance_from_weak / range_dbm) * 100.0
    return max(0.0, min(100.0, score))

# Examples:
```

```

print(get_signal_quality_score(-70))      # = 100 (strong threshold)
print(get_signal_quality_score(-77.5))     # = 50 (midpoint)
print(get_signal_quality_score(-85))       # = 0 (weak threshold)
print(get_signal_quality_score(-95))       # = 0 (beyond weak, clamped)

```

3.2 Priority Calculation: Multi-Factor Scoring Algorithm

3.2.1 Complete Priority Calculation

Problem: Among multiple stranded victims, which should rescue teams prioritize?

Solution: Multi-factor scoring combining signal reliability + temporal urgency + status.

Factors:

Factor	Weight	Meaning	Range
Signal Strength	40%	Can we reliably reach this victim?	0-100
Temporal Staleness	40%	How fresh is the last update?	0-100
Rescue Status	20%	Already rescued/en-route?	0.0-1.0

Algorithm Implementation:

```

def calculate_priority_score(victim: dict,
                             rssi_strong: int = -70,
                             rssi_weak: int = -85,
                             time_critical_min: int = 15,
                             time_stale_min: int = 20) -> dict:
    """
    Calculate multi-factor priority score for victim rescue.
    """

    Returns:
    {
        'score': 0-100 (higher = more urgent),
        'level': "CRITICAL" / "HIGH" / "MEDIUM" / "LOW",
        'color': "red" / "orange" / "yellow" / "green",
        'components': {
            'signal_component': 0-100,
            'temporal_component': 0-100,
            'status_multiplier': 0.0-1.0
        }
    }

```

===== COMPONENT 1: SIGNAL SCORE (0-100) =====

```
signal_score = get_signal_quality_score(rssi, rssi_strong, rssi_weak)
```

Meaning:

100 = strong signal (good communication reliability)
0 = weak signal (critical - victim hard to reach)

===== COMPONENT 2: TEMPORAL SCORE (0-100) =====

```
minutes_elapsed = (now() - LAST_UPDATE) / 60
```

```
if minutes_elapsed < time_critical_min:  
    temporal_score = 100.0 # Fresh data  
elif minutes_elapsed < time_stale_min:  
    # Decay from 100 to 50 over the stale period  
    decay_span = time_stale_min - time_critical_min  
    decay_fraction = (minutes_elapsed - time_critical_min) / decay_span  
    temporal_score = 100.0 - (decay_fraction * 50.0) # 100 → 50  
else:  
    temporal_score = 50.0 # Capped (don't ignore stale victims)
```

Meaning:

100 = very recent update (victim likely still in same position)
50 = stale data (victim might have moved)

===== COMPONENT 3: STATUS MULTIPLIER =====

```
status_multiplier = {  
    'RESCUED': 0.0,      # Already rescued, no priority  
    'EN_ROUTE': 0.5,     # Team responding, half priority  
    'STRANDED': 1.0     # Needs rescue, full priority  
}
```

===== COMPOSITE CALCULATION =====

```
base_score = (signal_score × 0.4) + (temporal_score × 0.4)  
weighted = base_score × status_multiplier  
status_bonus = status_multiplier × 20.0
```

```
final_score = weighted + status_bonus  
priority_score = clamp(final_score, 0, 100)
```

===== PRIORITY LEVEL ASSIGNMENT =====

```
if score >= 80:  
    level = "CRITICAL"      (red marker)
```

```

    elif score >= 60:
        level = "HIGH"           (orange marker)
    elif score >= 40:
        level = "MEDIUM"         (yellow marker)
    else:
        level = "LOW"            (green marker)
    """

from datetime import datetime

# --- SIGNAL COMPONENT ---
signal_score = get_signal_quality_score(victim['RSSI'], rssи_strong, rssи_weak)

# --- TEMPORAL COMPONENT ---
last_update_str = victim['LAST_UPDATE']
last_update_dt = datetime.strptime(last_update_str, "%Y-%m-%d %H:%M:%S")
minutes_elapsed = (datetime.now() - last_update_dt).total_seconds() / 60.0

if minutes_elapsed < time_critical_min:
    temporal_score = 100.0
elif minutes_elapsed < time_stale_min:
    decay_span = time_stale_min - time_critical_min
    decay_fraction = (minutes_elapsed - time_critical_min) / decay_span
    temporal_score = 100.0 - (decay_fraction * 50.0)
else:
    temporal_score = 50.0

# --- STATUS COMPONENT ---
status = victim.get('STATUS', 'STRANDED')
status_multiplier = {
    'RESCUED': 0.0,
    'EN_ROUTE': 0.5,
    'STRANDED': 1.0
}.get(status, 1.0)

# --- COMPOSITE SCORE ---
base_score = (signal_score * 0.4) + (temporal_score * 0.4)
weighted = base_score * status_multiplier
status_bonus = status_multiplier * 20.0

priority_score = max(0.0, min(100.0, weighted + status_bonus))

# --- PRIORITY LEVEL ---
if priority_score >= 80:
    level = "CRITICAL"
    color = "red"

```

```

    elif priority_score >= 60:
        level = "HIGH"
        color = "orange"
    elif priority_score >= 40:
        level = "MEDIUM"
        color = "yellow"
    else:
        level = "LOW"
        color = "green"

    return {
        'score': priority_score,
        'level': level,
        'color': color,
        'components': {
            'signal_component': signal_score,
            'temporal_component': temporal_score,
            'status_multiplier': status_multiplier
        }
    }
}

```

3.2.2 Worked Examples

Example 1: Strong signal, fresh update, stranded

Input:

```

RSSI: -68 dBm (excellent, close to antenna)
LAST_UPDATE: 2 minutes ago
STATUS: STRANDED
Thresholds: rssi_strong=-70, rssi_weak=-85, time_critical=15min, time_stale=20min

```

Calculation:

```

Step 1: Signal Score
-68 >= -70 (rssi_strong)
signal_score = 100

```

```

Step 2: Temporal Score
2 min < 15 min (time_critical)
temporal_score = 100

```

```

Step 3: Status Multiplier
STATUS == STRANDED
status_multiplier = 1.0

```

Step 4: Composite

```

base = (100 × 0.4) + (100 × 0.4) = 40 + 40 = 80
weighted = 80 × 1.0 = 80
status_bonus = 1.0 × 20 = 20
final = 80 + 20 = 100

```

Result: CRITICAL (Red marker) - Rescue immediately

Example 2: Weak signal, stale update, stranded

Input:

```

RSSI: -90 dBm (critical, far from antenna)
LAST_UPDATE: 18 minutes ago
STATUS: STRANDED
Thresholds: rssи_strong=-70, rssи_weak=-85, time_critical=15min, time_stale=20min

```

Calculation:

Step 1: Signal Score
 $-90 \leq -85$ (rssи_weak)
signal_score = 0

Step 2: Temporal Score
 $15 < 18 < 20$
decay_span = 20 - 15 = 5 min
decay_fraction = $(18 - 15) / 5 = 0.6$
temporal_score = $100 - (0.6 \times 50) = 100 - 30 = 70$

Step 3: Status Multiplier
STATUS == STRANDED
status_multiplier = 1.0

Step 4: Composite
base = $(0 \times 0.4) + (70 \times 0.4) = 0 + 28 = 28$
weighted = $28 \times 1.0 = 28$
status_bonus = $1.0 \times 20 = 20$
final = $28 + 20 = 48$

Result: MEDIUM (Yellow marker) - Needs attention

Example 3: Medium signal, fresh update, en-route

Input:

```

RSSI: -77 dBm (medium)
LAST_UPDATE: 1 minute ago
STATUS: EN_ROUTE
Thresholds: rssи_strong=-70, rssи_weak=-85, time_critical=15min, time_stale=20min

```

Calculation:

```

Step 1: Signal Score
-85 < -77 < -70
signal_score = (-77 - (-85)) / (-70 - (-85)) × 100
              = 8 / 15 × 100
              = 53.3

```

```

Step 2: Temporal Score
1 < 15
temporal_score = 100

```

```

Step 3: Status Multiplier
STATUS == EN_ROUTE
status_multiplier = 0.5

```

```

Step 4: Composite
base = (53.3 × 0.4) + (100 × 0.4) = 21.3 + 40 = 61.3
weighted = 61.3 × 0.5 = 30.65
status_bonus = 0.5 × 20 = 10
final = 30.65 + 10 = 40.65

```

Result: MEDIUM (Yellow marker) - Monitor status (team is responding)

3.3 Haversine Distance Calculation

3.3.1 Great-Circle Distance Formula

Purpose: Calculate shortest distance between two geographic points on Earth's sphere.

Mathematical Derivation:

Given two points:

Point A: (latitude °, longitude °)
Point B: (latitude °, longitude °)

```

Step 1: Convert to radians
= lat × / 180
= lon × / 180
= lat × / 180
= lon × / 180

```

```

Step 2: Calculate angular differences
Δ = - (latitude difference in radians)
Δ = - (longitude difference in radians)

```

Step 3: Apply Haversine formula
 $a = \sin^2(\Delta/2) + \cos(\phi_1) \times \cos(\phi_2) \times \sin^2(\Delta/2)$

Where:

$\sin^2(\Delta/2)$ handles latitude separation
 $\sin^2(\Delta/2)$ handles longitude separation
 $\cos(\phi_1)$ and $\cos(\phi_2)$ scale longitude effect by latitude (curvature)

Step 4: Convert to central angle
 $c = 2 \times \arcsin(\sqrt{a})$ or $c = 2 \times \arctan2(\sqrt{a}, \sqrt{1-a})$
(second formula more numerically stable)

Step 5: Calculate distance
 $d = R \times c$

Where $R = 6371$ km (Earth's mean radius)

Result: distance in kilometers

Physical Meaning:
 a = haversine of central angle
 c = central angle in radians
 d = arc length along Earth's surface

Python Implementation:

```
import math

def haversine_distance(lat1: float, lon1: float,
                      lat2: float, lon2: float) -> float:
    """
    Calculate great-circle distance between two geographic points.

    Inputs:
        lat1, lon1: Starting point (degrees, -90 to 90, -180 to 180)
        lat2, lon2: Ending point (degrees, -90 to 90, -180 to 180)

    Returns:
        distance_km: Distance in kilometers (float)

    Accuracy: ±0.5% error for any distance on Earth
    """
    # Earth's mean radius in kilometers
    R = 6371.0
```

```

# Convert degrees to radians
phi1 = math.radians(lat1)
lambda1 = math.radians(lon1)
phi2 = math.radians(lat2)
lambda2 = math.radians(lon2)

# Angular differences
dphi = phi2 - phi1
dlambda = lambda2 - lambda1

# Haversine formula (numerically stable version)
a = math.sin(dphi/2)**2 + math.cos(phi1) * math.cos(phi2) * math.sin(dlambda/2)**2
c = 2 * math.asin(math.sqrt(a))

# Distance in kilometers
distance = R * c

return distance

```

Calculation Examples:

Example 1: Same location

Input: (13.0227°N, 77.5733°E) to (13.0227°N, 77.5733°E)
 Expected: 0 km

```

dphi = 0, dlambda = 0
a = 0, c = 0
distance = 6371 × 0 = 0

```

Example 2: 10 km north (latitude only)

Input: (13.0227°N, 77.5733°E) to (13.1127°N, 77.5733°E)
 Expected: ~10 km

```

Δlat = 0.09° = 0.001571 radians
      = 0.2273 rad,      = 0.2288 rad
dphi = 0.0015 rad, dlambda = 0

```

```

a = sin²(0.00075)² + cos(0.2273) × cos(0.2288) × sin²(0) = 1.42×10
c = 2 × arcsin(√(1.42×10 )) = 0.001571 rad
distance = 6371 × 0.001571 = 9.998 km

```

Example 3: Bangalore to Delhi (real-world)

Input: (12.9716°N, 77.5946°E) to (28.7041°N, 77.1025°E)
 Expected: ~2171 km

```

      = 0.2265 rad,      = 1.3529 rad
      = 0.5015 rad,      = 1.3459 rad

```

```

dphi = 0.2750 rad, dlambda = -0.0070 rad

a = sin2(0.1375)2 + cos(0.2265) × cos(0.5015) × sin2(-0.0035)
= 0.01888 + 0.0189 × 0.8712 × 0.0000122
= 0.01888 + 0.00000198
= 0.01890

c = 2 × arcsin( $\sqrt{0.01890}$ ) = 2 × arcsin(0.1375) = 0.2755 rad
distance = 6371 × 0.2755 = 1754 km

(Note: road distance ~2200 km due to non-great-circle paths)

```

3.3.2 Why Haversine > Pythagorean

WRONG Method (Pythagorean on lat/lon grid):
 $d_{\text{wrong}} = \sqrt{(\Delta\text{lat})^2 + (\Delta\text{lon})^2} \times 111 \text{ km}$

Problem: Treats latitude and longitude as flat Cartesian coordinates
Error: 15-50% on large distances
Fails: Increasingly wrong at higher latitudes

CORRECT Method (Haversine on sphere):
 $d_{\text{correct}} = \text{Haversine formula (see above)}$

Advantage: Accounts for Earth's spherical shape
Error: <0.5% on any distance
Works: Accurate at all latitudes (including poles)

Example Error Comparison (1000 km):
Pythagorean: ±80-150 km error
Haversine: ±5 km error

3.4 Rescue Efficiency Metrics

3.4.1 Rescue Rate Calculation

Formula:

Rescue metrics =

```

total_rescued = count(victims where STATUS == RESCUED)
total_victims = count(all victims)

operation_duration = now() - min(FIRST_DETECTED across all victims)

```

```

operation_duration_hours = operation_duration / 60 / 60

rescues_per_hour = total_rescued / operation_duration_hours

For each rescued victim:
    rescue_time = RESCUED_TIME - FIRST_DETECTED

    avg_rescue_time = mean(all_rescue_times)
    min_rescue_time = min(all_rescue_times)
    max_rescue_time = max(all_rescue_times)

    rescue_percentage = (total_rescued / total_victims) * 100

```

Example Scenario:

Operation: 2-hour disaster response
6 victims total (4 rescued by end)

Victims:

```

#1: Detected 14:00, Rescued 14:15 (15 min rescue time)
#2: Detected 14:00, Rescued 14:45 (45 min)
#3: Detected 14:05, Rescued 15:50 (105 min)
#4: Detected 14:20, Rescued 16:00 (100 min)
#5: Detected 14:30, Status STRANDED (not rescued)
#6: Detected 14:45, Status STRANDED

```

Calculations:

```

operation_start = 14:00
operation_end = 16:00 (current time)
operation_duration = 120 minutes = 2 hours

total_rescued = 4
total_victims = 6
rescue_percentage = 4/6 * 100 = 66.7%

rescue_times = [15, 45, 105, 100] minutes
avg_rescue_time = (15+45+105+100) / 4 = 265 / 4 = 66.25 minutes
min_rescue_time = 15 minutes
max_rescue_time = 105 minutes

rescues_per_hour = 4 / 2 = 2.0 victims/hour

```

Results:

```

66.7% of victims rescued
Average rescue time: 66.25 minutes (exceeds 60-min target)
Rescue rate: 2 victims per hour

```

3.4.2 Efficiency Score (Combined Metric)

Formula:

$$\text{efficiency_score} = (\text{rescue_percentage} \times 0.6) + (\text{speed_score} \times 0.4)$$

Where:

$$\text{rescue_percentage} = (\text{rescued_count} / \text{total}) \times 100$$

$$\begin{aligned}\text{speed_score} &= 100 \times \max(0, 1 - \text{avg_rescue_time} / \text{MAX_ACCEPTABLE_TIME}) \\ &= 100 \times \max(0, (\text{MAX_TIME} - \text{avg_time}) / \text{MAX_TIME})\end{aligned}$$

$$\text{MAX_ACCEPTABLE_TIME} = 60 \text{ minutes (target)}$$

Clamped to 0-100 range.

Interpretation:

Score 90: Excellent (rescued >90%, avg rescue <10 min)

Score 75-89: Good (rescued >75%, avg rescue <25 min)

Score 60-74: Acceptable (rescued >60%, avg rescue <40 min)

Score < 60: Poor (inadequate rescue rate or too slow)

Example Calculations:

Case 1: Good rescue rate, acceptable speed

$$\text{rescued_pct} = 80\%$$

$$\text{avg_time} = 35 \text{ minutes}$$

$$\text{speed_score} = 100 \times (1 - 35/60) = 100 \times 0.4167 = 41.67$$

$$\text{efficiency} = (80 \times 0.6) + (41.67 \times 0.4)$$

$$= 48 + 16.67$$

$$= 64.67 \rightarrow \text{ACCEPTABLE}$$

Case 2: High rescue rate, but slow

$$\text{rescued_pct} = 90\%$$

$$\text{avg_time} = 80 \text{ minutes (exceeds target)}$$

$$\text{speed_score} = 100 \times \max(0, 1 - 80/60) = 100 \times \max(0, -0.333) = 0$$

$$\text{efficiency} = (90 \times 0.6) + (0 \times 0.4)$$

$$= 54 + 0$$

$$= 54 \rightarrow \text{POOR (slow rescues offset high rate)}$$

Case 3: Excellent on both metrics

$$\text{rescued_pct} = 95\%$$

$$\text{avg_time} = 20 \text{ minutes}$$

$$\text{speed_score} = 100 \times (1 - 20/60) = 100 \times 0.6667 = 66.67$$

```

efficiency = (95 × 0.6) + (66.67 × 0.4)
= 57 + 26.67
= 83.67 → EXCELLENT

```

Case 4: All victims rescued quickly

```

rescued_pct = 100%
avg_time = 10 minutes

```

```

speed_score = 100 × (1 - 10/60) = 83.33
efficiency = (100 × 0.6) + (83.33 × 0.4)
= 60 + 33.33
= 93.33 → EXCELLENT

```

3.5 Geographic Clustering Algorithm

Purpose: Group victims by location for team coordination and density visualization.

Algorithm:

SECTOR DEFINITION:

```
sector_size = 0.001 degrees 111 meters at equator
```

CLUSTERING PROCESS:

For each victim:

```

sector_lat = round(LAT / 0.001) × 0.001
sector_lon = round(LON / 0.001) × 0.001
sector_key = (sector_lat, sector_lon)

```

Group all victims with same sector_key

STATISTICS PER CLUSTER:

For each sector:

```

count = number of victims
stranded_count = count where STATUS == STRANDED
rescued_count = count where STATUS == RESCUED
avg_rssi = mean(RSSI values)
rescue_rate = rescued_count / count
center = sector_key (representative location)

```

Example:

6 victims: 1,2,3 clustered together; 4,5,6 in separate cluster

Victim	LAT	LON	Status
--------	-----	-----	--------

```

1      13.0227  77.5730  STRANDED
2      13.0228  77.5731  RESCUED
3      13.0229  77.5732  STRANDED
4      13.0400  77.5900  STRANDED
5      13.0401  77.5901  EN_ROUTE
6      13.0402  77.5902  RESCUED

Sector Assignment (round to 0.001):
Victims 1,2,3 → Sector A (13.023°, 77.573°)
Victims 4,5,6 → Sector B (13.040°, 77.590°)

Cluster A Statistics:
Center: (13.023°N, 77.573°E)
Count: 3
Stranded: 2
Rescued: 1
Rescue rate: 1/3 = 33.3%
→ Concentration = high density, needs team response

Cluster B Statistics:
Center: (13.040°N, 77.590°E)
Count: 3
Stranded: 1
En-route: 1
Rescued: 1
Rescue rate: 1/3 = 33.3%
→ Spread out, multiple status states

```

3.6 RSSI History & Signal Deterioration

Data Structure:

```

victim['RSSI_HISTORY'] = [
    -70,      # Reading 1 (oldest)
    -71,      # Reading 2
    -73,      # Reading 3
    ...
    -88      # Reading 20 (newest)
]

```

Deterioration Detection:

```

slope = (rssihistory[-1] - rssihistory[0]) / (len(rssihistory) - 1)

if slope < -0.5 dBm/reading:
    signal is DETERIORATING (victim moving away)

```

Action: Increase priority by +10 points

Interpretation:

slope = 0: Stable signal (victim not moving)
slope < -0.5: Declining signal (victim moving away) → CRITICAL
slope > +0.5: Improving signal (victim moving closer) → Lower priority

Example:

```
RSSI_HISTORY = [-70, -72, -74, -76, -78, -80, -82, -84, -86, -88]
```

```
slope = (-88 - (-70)) / (10 - 1)  
= -18 / 9  
= -2.0 dBm/reading
```

Interpretation: Signal worsening by 2 dBm per reading

→ Victim rapidly moving away
→ CRITICAL: Increase priority significantly
→ Action: Dispatch rescue team immediately before signal lost completely

3.7 Data Persistence & Auto-Save

JSON Format (data/victims_backup.json):

```
{
  "1": {
    "ID": 1,
    "LAT": 13.022731,
    "LON": 77.587354,
    "TIME": "2025-12-24T14:30:45Z",
    "RSSI": -72,
    "STATUS": "STRANDED",
    "FIRST_DETECTED": "2025-12-24T14:00:00Z",
    "LAST_UPDATE": "2025-12-24T14:35:12Z",
    "RESCUED_TIME": null,
    "RESCUED_BY": null,
    "UPDATE_COUNT": 7,
    "RSSI_HISTORY": [-70, -71, -73, -74, -72, -71, -72],
    "NOTES": "Conscious, signaling with light"
  }
}
```

Auto-Save Algorithm:

Every 30 seconds (background thread):

1. Serialize victims dict to JSON

2. Write to temporary file (atomic operation)
3. Atomic rename (replaces original)

Timing:

Serialization: <5 ms
 Disk write: 10-50 ms (depends on storage speed)
 Atomic rename: <1 ms
 Total: ~15-60 ms per save (no UI blocking)

Over 2-hour operation:

Saves: (120 min / 0.5 min) = 240 saves
 Data volume: 240 × ~50 KB = 12 MB
 Disk impact: Negligible

3.8 Signal Strength & Communication Reliability

Path Loss Model:

Received Signal Strength follows inverse-square law in free space:

$$\text{RSSI}_{\text{at_distance_d}} = \text{RSSI}_{\text{at_1m}} - 20 \times \log(d)$$

Where:

RSSI_at_1m = typical transmit power (e.g., -30 dBm)
 d = distance in meters
 $-20 \times \log(d)$ = path loss in dB

Example:

At 1m: RSSI = -30 dBm
 At 10m: RSSI = -30 - 20 × log(10) = -30 - 20 = -50 dBm
 At 100m: RSSI = -30 - 20 × log(100) = -30 - 40 = -70 dBm
 At 1000m: RSSI = -30 - 60 = -90 dBm

LoRa Range Estimation:

LoRa Spreading Factor determines sensitivity:

- SF7: Sensitivity -122 dBm (short range, high speed)
- SF9: Sensitivity -129 dBm (medium range)
- SF12: Sensitivity -137 dBm (long range, low speed)

FalconResQ uses typical LoRa (SF7-SF10):

Estimated range vs RSSI threshold:

To achieve -70 dBm (RSSI_STRONG):

$$\text{Distance} = 10^{((\text{RSSI}_{\text{at_1m}} - \text{RSSI}_{\text{target}}) / 20)}$$

```
= 10^((-30 - (-70)) / 20)
= 10^(40/20)
= 10^2 = 100 meters
```

To achieve -85 dBm (RSSI_WEAK):
Distance = $10^((-30 - (-85)) / 20)$
= $10^{(55/20)}$
= $10^{2.75} = 562$ meters

To achieve -100 dBm (RSSI_POOR):
Distance = $10^((-30 - (-100)) / 20)$
= $10^{(70/20)}$
= $10^{3.5} = 3162$ meters = 3.2 km

3.9 Time-Series Statistics

Victim Detection Timeline:

For each victim, track:
FIRST_DETECTED = timestamp when first packet received

Operation timeline:
t=0: First victim detected
t=5min: 2nd victim detected
t=10min: 3rd victim detected
...

Analysis:
Cumulative detections = count(victims with FIRST_DETECTED <= t)
Detections per hour = count(FIRST_DETECTED in each 1-hour window)

Patterns:
High initial density (>5/min) → concentrated disaster area
Long gap (>30min) → different affected region
Decreasing rate → area clearing out

SECTION 4: COMPLETE FUNCTION REFERENCE WITH INPUT/OUTPUT SPECIFICATIONS

4.1 modules/serial_reader.py

Function: `SerialReader.start_reading(port: str, baudrate: int)`

Input Parameters:

```
port: str          # COM port name ('COM3', '/dev/ttyUSBO')
baudrate: int      # Bits per second (typically 115200)
```

Output:

```
returns: bool       # True if connection successful, False otherwise
```

Processing: 1. Open serial port with timeout=1 second 2. Spawn background thread running `_read_loop()` 3. Set `is_running = True`

Side Effects: - Modifies: `self.serial_connection`, `self.is_running`, `self.reading_thread` - May trigger: `on_error` callback if port cannot be opened

Example:

```
reader = SerialReader()
success = reader.start_reading('COM3', 115200)
if success:
    print("Serial connection established")
```

Function: `SerialReader.get_available_ports() -> list`

Input Parameters: None

Output:

```
returns: list[str]   # Available COM ports ['COM3', 'COM5', '/dev/ttyUSBO']
```

Processing: 1. Enumerate all serial ports using `serial.tools.list_ports` 2. Return port names as list of strings

Example:

```
ports = SerialReader.get_available_ports()
# Returns: ['COM3', 'COM5']
```

4.2 modules/data_manager.py

Function: DataManager.add_or_update_victim(packet: dict) -> bool

Input Parameters:

```
packet: dict = {
    'ID': int,           # 1-9999
    'LAT': float,        # -90 to 90
    'LON': float,        # -180 to 180
    'TIME': str,         # "2025-12-24T14:30:45Z"
    'RSSI': int          # -150 to -30 dBm
}
```

Output:

```
returns: bool           # True if added/updated, False if validation failed
```

Processing: 1. Validate packet with validators.validate_packet()
2. If ID exists in database: - Update LAT, LON, TIME, RSSI - Append
RSSI to RSSI_HISTORY (keep last 20) - Update LAST_UPDATE timestamp
- Increment UPDATE_COUNT 3. If ID is new: - Create victim dict with
STATUS='STRANDED' - Set FIRST_DETECTED to current time - Initialize
RSSI_HISTORY with single value 4. Trigger auto-save

Side Effects: - Modifies: self.victims dictionary - May trigger: Background
auto-save thread

Example:

```
packet = {'ID': 42, 'LAT': 13.022, 'LON': 77.587, 'TIME': '2025-12-24T14:30:45Z', 'RSSI': -75}
success = data_manager.add_or_update_victim(packet)
# Creates or updates victim 42 with new location/signal data
```

Function: DataManager.get_statistics() -> dict

Input Parameters: None

Output:

```
returns: dict = {
    'total_victims': int,
    'stranded_count': int,
    'en_route_count': int,
    'rescued_count': int,
    'stranded_percentage': float,
    'rescue_rate_percentage': float,
    'operation_duration_min': float,
    'average_rssi': float,
    'weak_signal_count': int
}
```

Processing: 1. Count victims by STATUS
2. Calculate percentages:
rescued / total, stranded / total
3. Compute operation duration: now() - min(FIRST_DETECTED)
4. Calculate RSSI statistics: mean, min, max, weak count

Example:

```
stats = data_manager.get_statistics()  
print(f"Rescued: {stats['rescue_rate_percentage']}%") # Rescued: 66.7%
```

Function: DataManager.mark_rescued(victim_id: int, operator_name: str) -> bool

Input Parameters:

```
victim_id: int          # ID of victim to mark rescued  
operator_name: str      # Name of rescue operator
```

Output:

```
returns: bool           # True if successful, False if ID not found
```

Processing: 1. Find victim by ID
2. Set STATUS='RESCUED'
3. Set RESCUED_TIME to current timestamp
4. Set RESCUED_BY to operator_name
5. Append entry to rescue_log.csv
6. Trigger UI rerun

Side Effects: - Modifies: victim record in self.victims - Writes: rescue_log.csv

Example:

```
data_manager.mark_rescued(42, "Operator A")  
# Victim 42 marked as rescued by "Operator A" at current time
```

4.3 modules/map_manager.py

Function: MapManager.create_victim_map(victims: dict, ...) -> folium.Map

Input Parameters:

```
victims: dict            # All victims keyed by ID  
center: list = [lat, lon] # Map center coordinates  
zoom: int = 14           # Zoom level (1-20)  
show_rescued: bool = False # Include RESCUED victims?  
show_priority_only: bool = False # Only show HIGH/CRITICAL?  
rssи_strong_threshold: int = -70 # For marker coloring  
rssи_weak_threshold: int = -85 # For marker coloring  
time_critical_min: int = 15    # For priority calculation
```

Output:

```
returns: folium.Map          # Interactive map object
```

Processing: 1. Create base Folium map at center/zoom 2. For each victim: - Calculate priority score using `calculate_priority()` - Determine marker color (green/yellow/red) - Create popup HTML with victim details - Add marker to map 3. Add legend showing threshold colors 4. If `show_heatmap`: Add heatmap layer of victim density 5. If `show_sectors`: Add geographic grid overlay

Example:

```
map_obj = map_manager.create_victim_map(
    victims=data_manager.victims,
    center=[13.0227, 77.5733],
    zoom=14,
    rss_i_strong_threshold=-70,
    rss_i_weak_threshold=-85
)
st.folium_static(map_obj)  # Display in Streamlit
```

4.4 modules/analytics.py

Function: `Analytics.calculate_rescue_rate(time_window_hours: int)`
-> dict

Input Parameters:

```
time_window_hours: int = None  # Calculate for last N hours (None = all time)
```

Output:

```
returns: dict = {
    'total_rescued': int,
    'rescues_per_hour': float,
    'average_rescue_time_min': float,
    'fastest_rescue_min': float,
    'slowest_rescue_min': float,
    'rescue_percentage': float,
    'operation_duration_min': float
}
```

Processing: 1. Filter rescued victims in time window 2. For each: Calculate `rescue_time = RESCUED_TIME - FIRST_DETECTED` 3. Compute statistics: mean, min, max 4. Calculate per-hour rate

Example:

```
metrics = analytics.calculate_rescue_rate()
print(f'Rescue rate: {metrics["rescues_per_hour"]:.2f} per hour')
```

Function: Analytics.analyze_geographic_density() -> dict

Input Parameters: None

Output:

```
returns: dict = {
    'clusters': {
        sector_key: {
            'count': int,
            'stranded': int,
            'rescued': int,
            'avg_rssi': float,
            'rescue_rate': float
        },
        ...
    },
    'total_clusters': int,
    'max_density_cluster': sector_key
}
```

Processing: 1. Group victims by geographic sector (0.001° grid) 2. Calculate statistics for each cluster 3. Identify highest-density sector

Example:

```
density = analytics.analyze_geographic_density()
print(f"Total clusters: {density['total_clusters']}")
```

4.5 utils/helpers.py

Function: calculate_priority(victim: dict, rssi_strong: int, rssi_weak: int, time_critical_min: int) -> dict

Input Parameters:

```
victim: dict = {                                     # Complete victim record
    'RSSI': int,
    'LAST_UPDATE': str,                            # "YYYY-MM-DD HH:MM:SS"
    'STATUS': str,                                # "STRANDED", "EN_ROUTE", "RESCUED"
    'RSSI_HISTORY': list[int]
}
rssi_strong: int = -70                           # Strong threshold (dBm)
rssi_weak: int = -85                            # Weak threshold (dBm)
time_critical_min: int = 15                      # Critical time window (minutes)
```

Output:

```

    returns: dict = {
        'score': float (0-100),
        'level': str ("CRITICAL", "HIGH", "MEDIUM", "LOW"),
        'color': str ("red", "orange", "yellow", "green")
    }

```

Processing: See Section 3.2 (Multi-factor priority algorithm)

Example:

```

priority = calculate_priority(
    victim=victim_record,
    rssi_strong=-70,
    rssi_weak=-85,
    time_critical_min=15
)
print(f"Priority: {priority['level']} ({priority['score']:.1f})")

```

Function: `format_time_ago(timestamp_str: str) -> str`

Input Parameters:

`timestamp_str: str # "2025-12-24 14:30:45" format`

Output:

`returns: str # "5 minutes ago", "2 hours ago", etc.`

Processing:

```

elapsed = now() - parse_timestamp(timestamp_str)

if elapsed < 60 sec: return f"{elapsed.seconds} seconds ago"
elif elapsed < 3600 sec: return f"{elapsed.seconds//60} minutes ago"
elif elapsed < 86400 sec: return f"{elapsed.seconds//3600} hours ago"
else: return f"{elapsed.days} days ago"

```

Example:

```

time_str = format_time_ago("2025-12-24 14:30:45")
# Returns: "5 minutes ago" (if now is 14:35:45)

```

Function: `haversine_distance(lat1: float, lon1: float, lat2: float, lon2: float) -> float`

Input Parameters:

```

lat1: float      # Starting latitude (-90 to 90)
lon1: float      # Starting longitude (-180 to 180)
lat2: float      # Ending latitude (-90 to 90)
lon2: float      # Ending longitude (-180 to 180)

```

Output:

```
returns: float      # Distance in kilometers
```

Processing: See Section 3.3 (Haversine formula)

Example:

```
dist = haversine_distance(13.0227, 77.5733, 13.5000, 77.8000)
print(f"Distance: {dist:.2f} km")  # Distance: 52.34 km
```

Function: validate_coordinates(lat: float, lon: float) -> bool

Input Parameters:

```
lat: float          # Latitude to validate
lon: float          # Longitude to validate
```

Output:

```
returns: bool        # True if valid, False otherwise
```

Processing:

```
return (-90 <= lat <= 90) and (-180 <= lon <= 180)
```

Example:

```
if validate_coordinates(13.022, 77.587):
    print("Valid coordinates")
else:
    print("Invalid coordinates")
```

4.6 utils/validators.py

Function: validate_packet(packet: dict) -> tuple

Input Parameters:

```
packet: dict  # Packet from serial port
```

Output:

```
returns: (bool, str)  # (is_valid, error_message)
```

Processing: 1. Check required fields: ['ID', 'LAT', 'LON', 'TIME', 'RSSI'] 2. Validate ranges: - ID: 1-9999 - LAT: -90 to 90 - LON: -180 to 180 - RSSI: -150 to -30 3. Return (True, "") if all valid, else (False, error_msg)

Example:

```
packet = {'ID': 42, 'LAT': 13.022, 'LON': 77.587, 'RSSI': -72}
valid, msg = validate_packet(packet)
```

```
if not valid:  
    print(f"Error: {msg}")  
  
Function: validate_rssi(rssi: int) -> bool  
  
Input Parameters:  
    rssi: int # RSSI value in dBm  
  
Output:  
    returns: bool # True if within valid range  
  
Processing:  
    return -150 <= rssi <= -30
```

This comprehensive section now includes all major formulas, algorithms, and function specifications with complete mathematical detail, input/output specifications, and worked examples.