

## 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 milliwatts). Linear representation would require handling numbers from  $10^{-15}$  to  $10^{-3}$ . Logarithmic compression makes this humanly readable (-120 to -30 dBm) and aligns with human perception (10 dB =  $10\times$  power change).

**Example Calculations:**

Scenario 1: Victim very close to antenna

Physical power received = 1 W =  $10^0$  watts

$$\text{RSSI} = 10 \times \log (10^0) + 30 = 10 \times (-6) + 30 = -30 \text{ dBm}$$

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 =  $2 \times 10^{-13}$  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'], rssi_strong, rssi_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: rssi_strong=-70, rssi_weak=-85, time_critical=15min, time_stale=20min
```

Calculation:

Step 1: Signal Score

```
-90 <= -85 (rssi_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 × 50) = 100 - 30 = 70
```

Step 3: Status Multiplier

```
STATUS == STRANDED
status_multiplier = 1.0
```

Step 4: Composite

```
base = (0 × 0.4) + (70 × 0.4) = 0 + 28 = 28
weighted = 28 × 1.0 = 28
status_bonus = 1.0 × 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: rssi_strong=-70, rssi_weak=-85, time_critical=15min, time_stale=20min
```

Calculation:

Step 1: Signal Score

-85 < -77 < -70

$$\begin{aligned}\text{signal\_score} &= (-77 - (-85)) / (-70 - (-85)) \times 100 \\ &= 8 / 15 \times 100 \\ &= 53.3\end{aligned}$$

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

$$c = 2 \times \arcsin(\sqrt{a}) \quad \text{or} \quad c = 2 \times \arctan2(\sqrt{a}, \sqrt{1-a})$$

(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

$d\phi = 0$ ,  $d\lambda = 0$   
 $a = 0$ ,  $c = 0$   
 $distance = 6371 \times 0 = 0$

Example 2: 10 km north (latitude only)

Input: (13.0227°N, 77.5733°E) to (13.1127°N, 77.5733°E)

Expected: ~10 km

$\Delta lat = 0.09^\circ = 0.001571$  radians  
 $= 0.2273$  rad,  $= 0.2288$  rad  
 $d\phi = 0.0015$  rad,  $d\lambda = 0$

$a = \sin^2(0.00075)^2 + \cos(0.2273) \times \cos(0.2288) \times \sin^2(0) = 1.42 \times 10^{-6}$   
 $c = 2 \times \arcsin(\sqrt{1.42 \times 10^{-6}}) = 0.001571$  rad  
 $distance = 6371 \times 0.001571 = 9.998$  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, dlambd = -0.0070 rad

a =  $\sin^2(0.1375)^2 + \cos(0.2265) \times \cos(0.5015) \times \sin^2(-0.0035)$   
= 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\_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\_correct = 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

rescued\_pct = 80%

avg\_time = 35 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

rescued\_pct = 90%

avg\_time = 80 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

rescued\_pct = 95%

avg\_time = 20 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 = (rssi_history[-1] - rssi_history[0]) / (len(rssi_history) - 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]

$$\begin{aligned}\text{slope} &= (-88 - (-70)) / (10 - 1) \\ &= -18 / 9 \\ &= -2.0 \text{ dBm/reading}\end{aligned}$$

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:

$$\text{RSSI\_at\_distance\_d} = \text{RSSI\_at\_1m} - 20 \times \log(d)$$

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 =  $-30 - 20 \times \log(10) = -30 - 20 = -50$  dBm  
 At 100m: RSSI =  $-30 - 20 \times \log(100) = -30 - 40 = -70$  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):

$$\text{Distance} = 10^{((\text{RSSI\_at\_1m} - \text{RSSI\_target}) / 20)}$$

$$= 10^{((-30 - (-70)) / 20)}$$

$$= 10^{(40/20)}$$

$$= 10^2 = 100 \text{ meters}$$

To achieve -85 dBm (RSSI\_WEAK):

$$\text{Distance} = 10^{((-30 - (-85)) / 20)}$$

$$= 10^{(55/20)}$$

$$= 10^{2.75} \quad 562 \text{ meters}$$

To achieve -100 dBm (RSSI\_POOR):

$$\text{Distance} = 10^{((-30 - (-100)) / 20)}$$

$$= 10^{(70/20)}$$

$$= 10^{3.5} \quad 3162 \text{ meters} \quad 3.2 \text{ km}$$


---

### 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/ttyUSB0')*  
`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/ttyUSB0']*

**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': -70}
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?
rssi_strong_threshold: int = -70 # For marker coloring
rssi_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,
    rssi_strong_threshold=-70,
    rssi_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.