

Patient Agent - Personalized Baseline System

Overview

The Patient Agent learns what's "normal" for each individual patient and detects meaningful deviations from their personal baseline. This enables personalized cardiac risk assessment.

Key Philosophy

- **Stable, Not Reactive:** Uses slow adaptation (EMA with $\alpha=0.05$) to prevent single bad days from shifting the baseline
- **No Deep Learning:** Uses interpretable statistical methods only
- **Conservative:** Requires sustained changes (3+ days) before escalating behavioral state
- **Explainable:** Every assessment can be traced to specific deviations from baseline

What It Tracks

Cardiac Baseline

- **Heart Rate:** Mean and std deviation
- **HRV (RMSSD):** Short-term variability baseline
- **HRV (SDNN):** Overall variability baseline
- **Arrhythmia Rate:** Normal percentage of time with arrhythmia detected
- **Physio Risk:** Average daily risk score

Activity Baseline

- Time spent in each activity (WALKING, SITTING, STANDING, LAYING, etc.)
- Activity pattern disruptions indicate behavioral changes

Behavioral State

- **STABLE:** Normal patterns, sensitivity = 1.0x
- **DEGRADED:** Minor deviations detected, sensitivity = 1.15x
- **AT_RISK:** Sustained abnormalities (3+ days), sensitivity = 1.3x

How It Works

1. Daily Aggregation

At end of each day (automatically detected at midnight):

- Collects all sensor outputs from the day
- Computes daily statistics (mean HR, mean HRV, arrhythmia frequency, etc.)

2. Baseline Update (Online Learning)

Uses **Exponential Moving Average** for stable adaptation:

$$\text{new_baseline} = (1 - \alpha) \times \text{old_baseline} + \alpha \times \text{today's_value}$$

- **First 7 days:** $\alpha = 0.20$ (faster learning)
- **Day 8+:** $\alpha = 0.05$ (very stable, 95% weight on history)

This prevents:

- Overreacting to single bad days
- Baseline drift from outliers
- Unstable alert thresholds

3. Risk Factor Detection

Checks 5 key indicators daily:

1. **HR Elevation:** >1.5 std above personal baseline
2. **HRV Reduction:** >1.5 std below baseline (lower HRV = worse)
3. **Arrhythmia Increase:** $>10\%$ absolute increase from baseline rate
4. **Physio Risk Elevation:** >2.0 std above baseline (double weighted)
5. **Activity Disruption:** $>30\%$ change in daily activity patterns

4. Behavioral State Logic

```
risk_factors = count of triggered indicators
```

```
if risk_factors >= 3:  
    degrading_days += 1  
elif risk_factors == 0:  
    improving_days += 1  
  
if degrading_days >= 3:  
    state = AT_RISK  
elif degrading_days >= 1:  
    state = DEGRADED  
else:  
    state = STABLE
```

Conservative by design: Requires 3 consecutive degrading days for AT_RISK.

5. Sensitivity Adjustment

Sensitivity factor modulates alert thresholds in Decision Agent:

- **STABLE:** 1.0x (normal thresholds)
- **DEGRADED:** 1.15x (15% more sensitive)
- **AT_RISK:** 1.3x (30% more sensitive)

Early in tracking (low confidence), adjustments are dampened.

API Endpoints

GET /api/patient/baseline

Returns learned baseline values:

json

```
{
  "baseline": {
    "hr_mean": 72.5,
    "hr_std": 12.3,
    "hrv_rmssd_mean": 38.2,
    "arrhythmia_rate": 0.04,
    "physio_risk_mean": 0.18
  },
  "activity_profile": {
    "SITTING": 0.42,
    "WALKING": 0.18,
    "STANDING": 0.22,
    "LAYING": 0.18
  },
  "days_seen": 15,
  "confidence": 0.5
}
```

GET /api/patient/state

Returns current behavioral assessment:

json

```
{
  "behavioral_state": "STABLE",
  "sensitivity_factor": 1.0,
  "confidence": 0.5,
  "days_seen": 15,
  "recent_trend": {
    "degrading_days": 0,
    "stable_days": 12,
    "improving_days": 3
  }
}
```

GET /api/patient/risk_context

Compares current metrics to baseline:

json

```
{
  "deviations": {
    "hr_zscore": 1.2,
    "hr_status": "normal",
    "hrv_zscore": -0.5,
    "hrv_status": "normal",
    "arrhythmia_above_baseline": false
  },
  "behavioral_state": "STABLE",
  "baseline_hr": 72.5,
  "baseline_hrv": 38.2
}
```

POST /api/patient/trigger_update

Manually trigger end-of-day update (normally automatic at midnight):

json

```
{
  "message": "Processed 432 sensor outputs",
  "patient_output": {
    "day": "2025-12-27",
    "sensitivity_factor": 1.0,
    "behavioral_state": "STABLE",
    "confidence": 0.5
  }
}
```

Confidence Building

The Patient Agent builds confidence over time:

- **Days 1-7:** Confidence = 0-0.23 (Learning mode, no sensitivity adjustments)
- **Days 8-30:** Confidence = 0.23-1.0 (Calibration, gradual adjustments)
- **Day 30+:** Confidence = 1.0 (Full personalization active)

Formula: $\text{confidence} = \min(1.0, \text{days_seen} / 30.0)$

Persistent State

Stored in `data/patient_state.json`:

json

```
{
  "patient_id": "patient_001",
  "created_date": "2025-12-01T00:00:00",
  "days_seen": 15,
  "last_update": "2025-12-15T23:59:59",
  "baseline": { ... },
  "activity_profile": { ... },
  "behavioral_state": "STABLE",
  "sensitivity_factor": 1.0,
  "confidence": 0.5,
  "recent_trend": { ... }
}
```

Integration with System

1. **Sensor Agent** → Outputs every 2 seconds
2. **Flask App** → Collects outputs throughout the day
3. **Midnight** → Automatic daily_update() triggered
4. **Patient Agent** → Updates baseline, assesses state
5. **Decision Agent** → Uses sensitivity_factor for alert thresholds

Example Scenario

Day 1-7: Patient walks 5000 steps/day, HR avg 70, HRV 40

- Agent learns this is "normal" for them

Day 15: Patient walks 5200 steps, HR 72, HRV 38

- Within baseline → STABLE state

Day 20: Patient walks 2000 steps, HR 85, HRV 25

- Multiple risk factors triggered (activity disruption, HR elevation, HRV reduction)
- degrading_days = 1 → DEGRADED state → sensitivity 1.15x

Day 21-22: Pattern continues

- degrading_days = 3 → AT_RISK state → sensitivity 1.3x
- Alerts now trigger at lower thresholds

Day 23: Patient resumes normal activity

- improving_days begins
- After 3 days of improvement → back to STABLE

Design Principles

- ✓ **Stable:** EMA prevents baseline jumps
- ✓ **Conservative:** Requires sustained changes (3 days)
- ✓ **Transparent:** Every decision traceable to metrics
- ✓ **Adaptive:** Learns individual norms over time
- ✓ **Interpretable:** No black-box models

This is the **intelligence** that makes the system personalized without being unpredictable.