

Dynamic Calibration of Nonlinear Temperature Sensors for Residual Limb Monitoring

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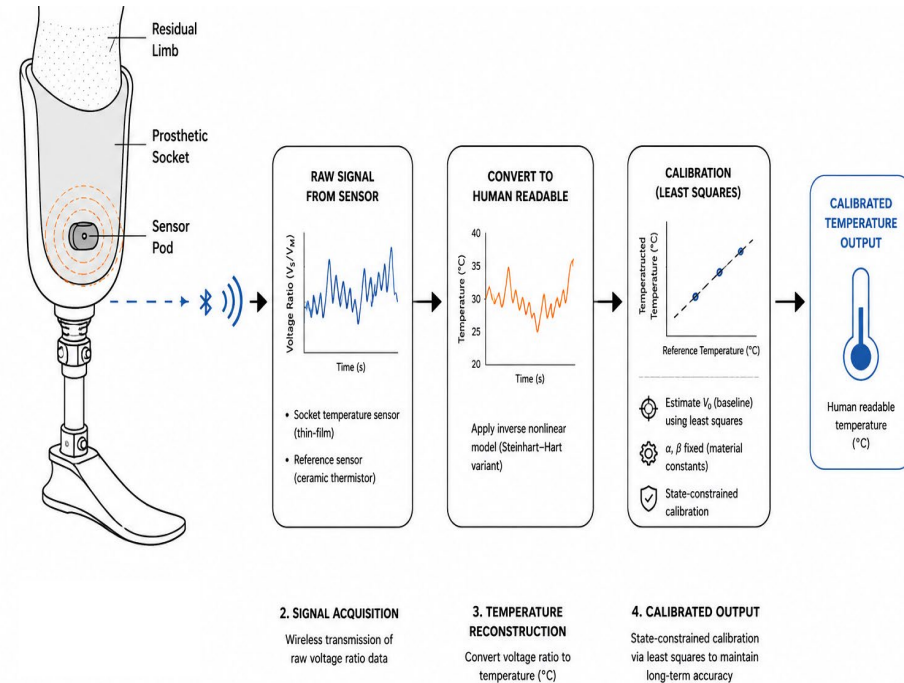
SAIP 2026 Oral Presentation
Track: F - Applied Physics

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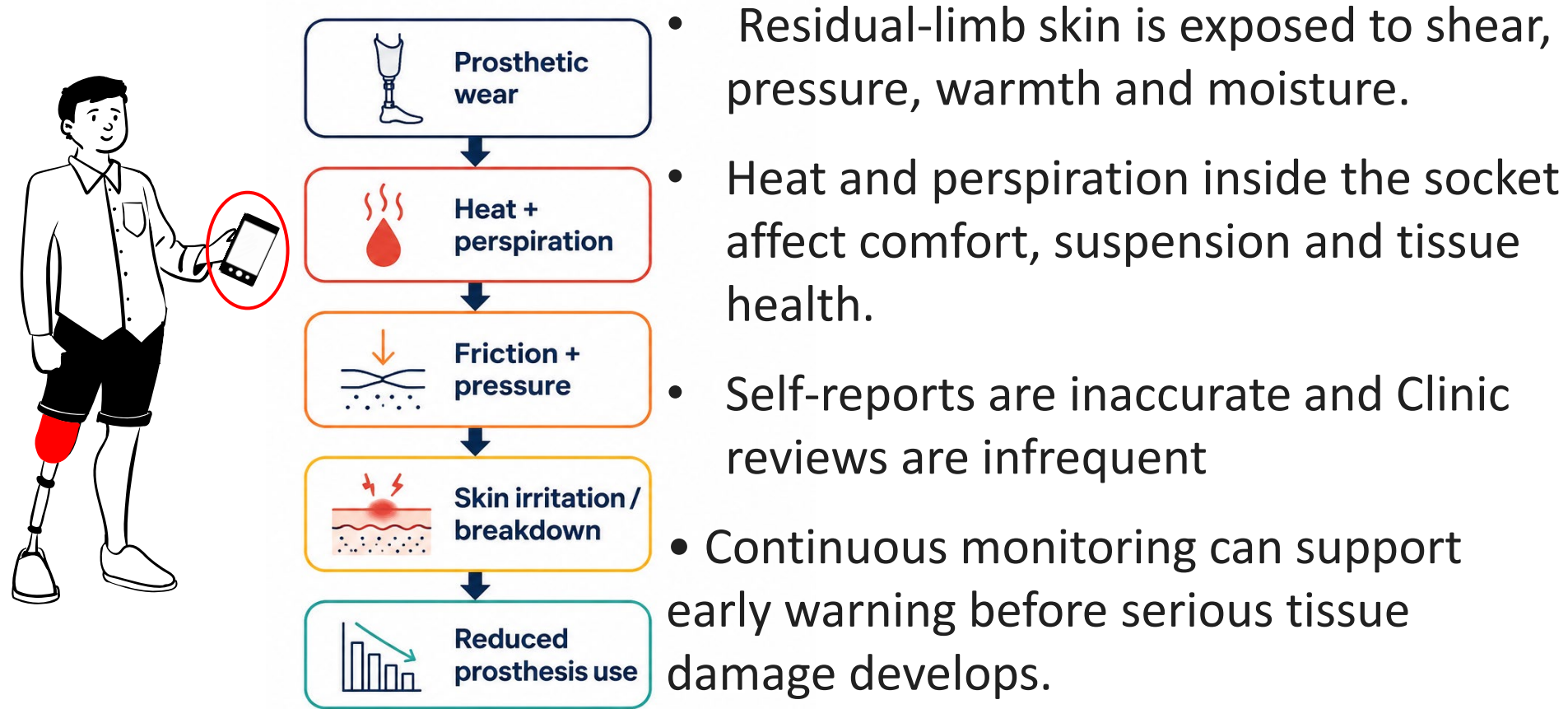
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Why Residual Limb Monitoring Matters

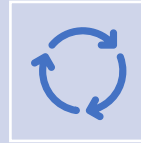


Research Question & Objectives

Can long-term sensor drift be compensated automatically while maintaining accurate residual limb temperature measurements?



- **Objective 1:** Develop dynamic calibration.



- **Objective 2:** Enable reliable long-term monitoring.



- **Approach:** Multi-scale temporal framework (morning equilibrium + sliding windows)



- **Objective 3:** Validate robustness across cohort and extended wear periods

The Physics: Nonlinear Temperature Reconstruction

$$T = \left[\frac{\beta}{\ln\left(\frac{V_S}{V_M} - (1 - \alpha(V_0 - 1))\right) - \ln\left(V_0 - (1 - \alpha(V_0 - 1))\right) + \frac{\beta}{T_0 + 273.15}} - 273.15 \right] ^\circ\text{C}$$

MEASURED: V_S/V_M (voltage ratio)

FIXED: α, β (material constants) $\leftarrow$ Do NOT drift

DRIFTS: V_0 (voltage reference) $\leftarrow$ What we estimate

$$\sum_{m=n-N}^n \left(T(V_0, V_S[m], V_M[m]) - T_{ref}[m] \right)^2 \text{ with respect to } V_0$$

Least-squares estimation: find V_0 that best matches reference sensor

Calibration function [N. Ndzamela, Honours Project 2020]

Experimental Sensor Platform: Dual-Sensor Design



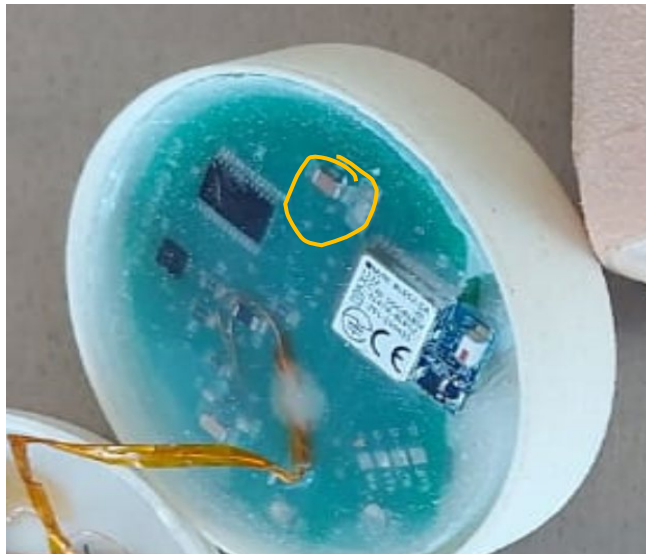
1st) generation pod



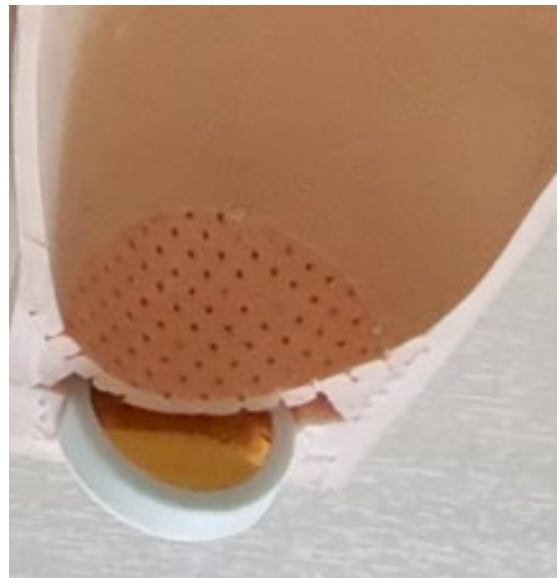
b) Placement



c) Prosthesis socket



d): Internals of the Pod



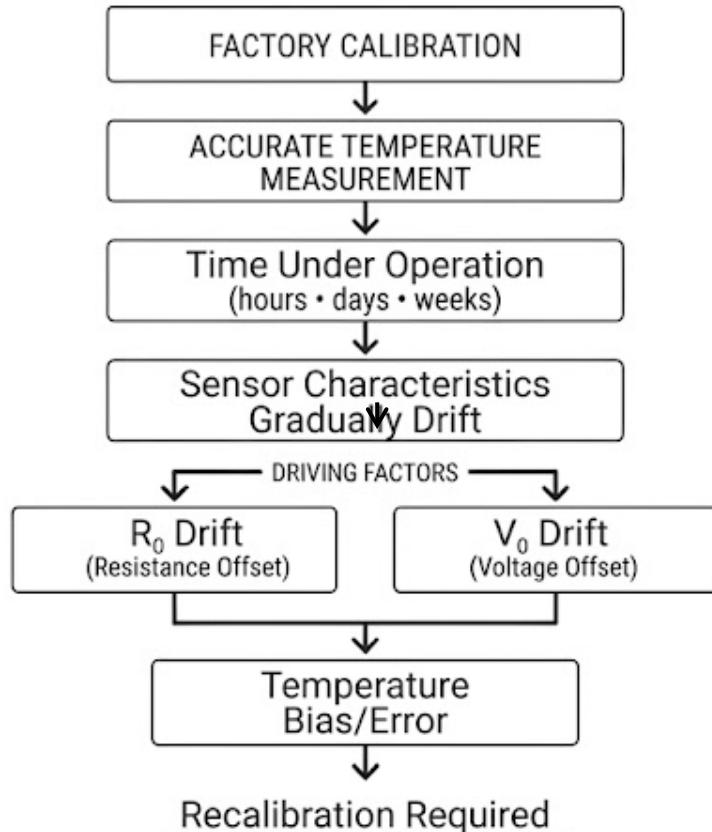
e) Placement



30 sec

f): The new Pod

The Calibration Problem: Why Static Fails



• Sensor drift occurs gradually due to:

- Aging of electronic components
- Repeated thermal cycling
- Mechanical stresses
- Environmental exposure
- Long-term sensor degradation

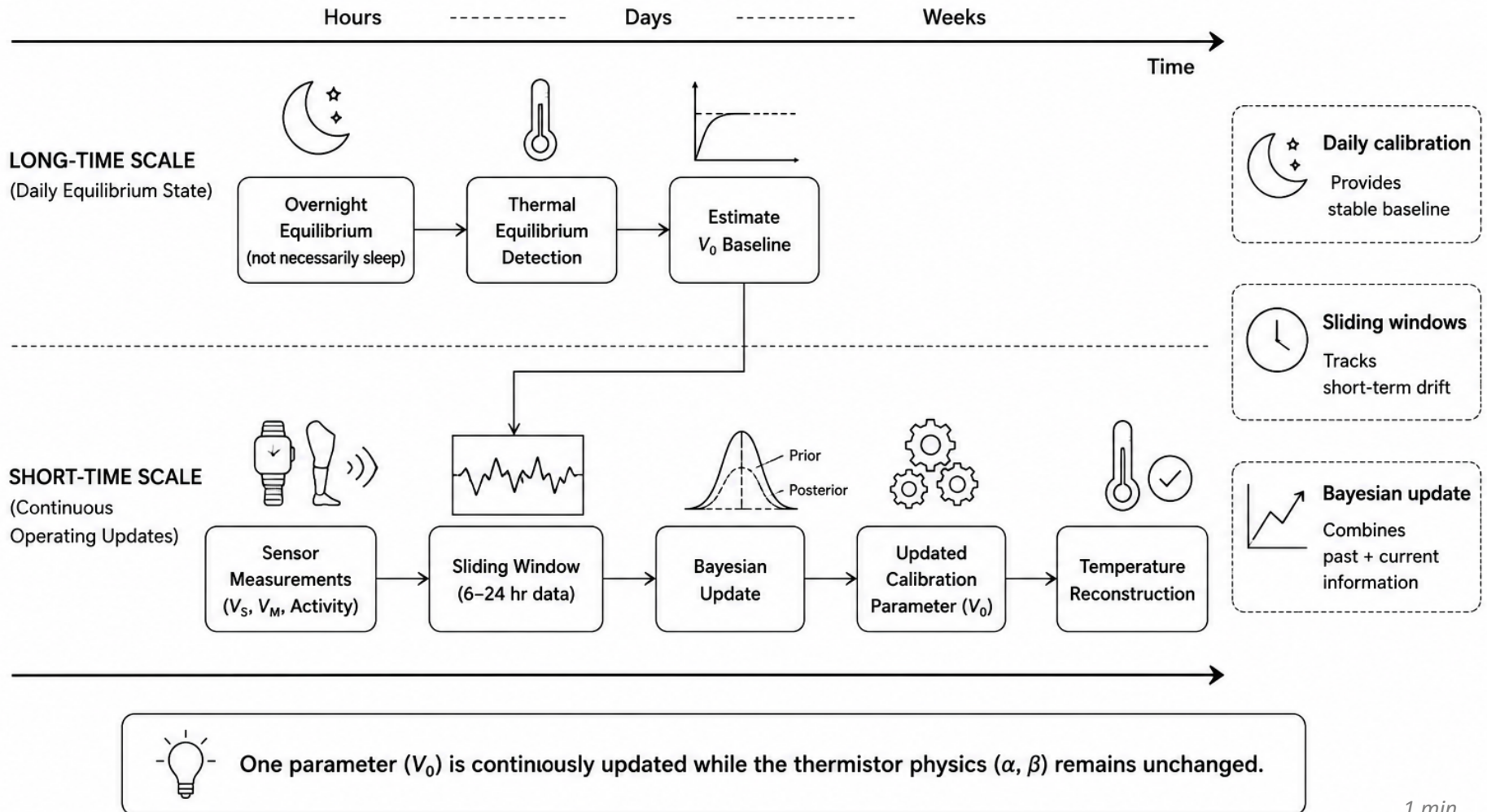
Result

Even when the residual limb temperature remains unchanged, the sensor output gradually deviates from the true value, reducing measurement accuracy.

Multi-Scale Dynamic Calibration Framework

Multi-Scale Calibration Framework

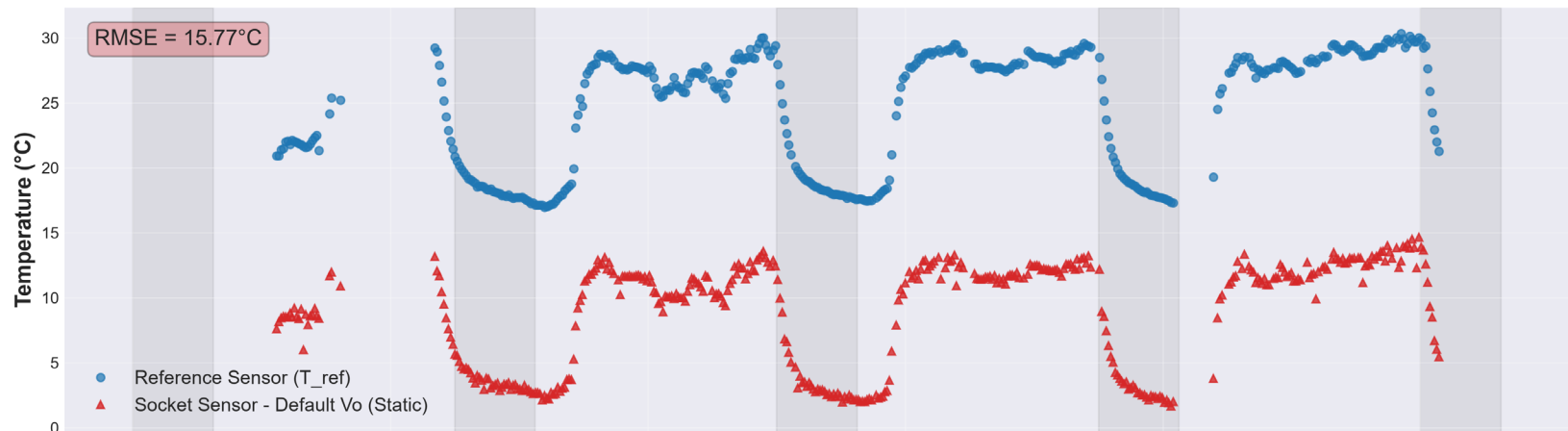
Calibration is continuously refined using both long-term equilibrium and short-term operating data.



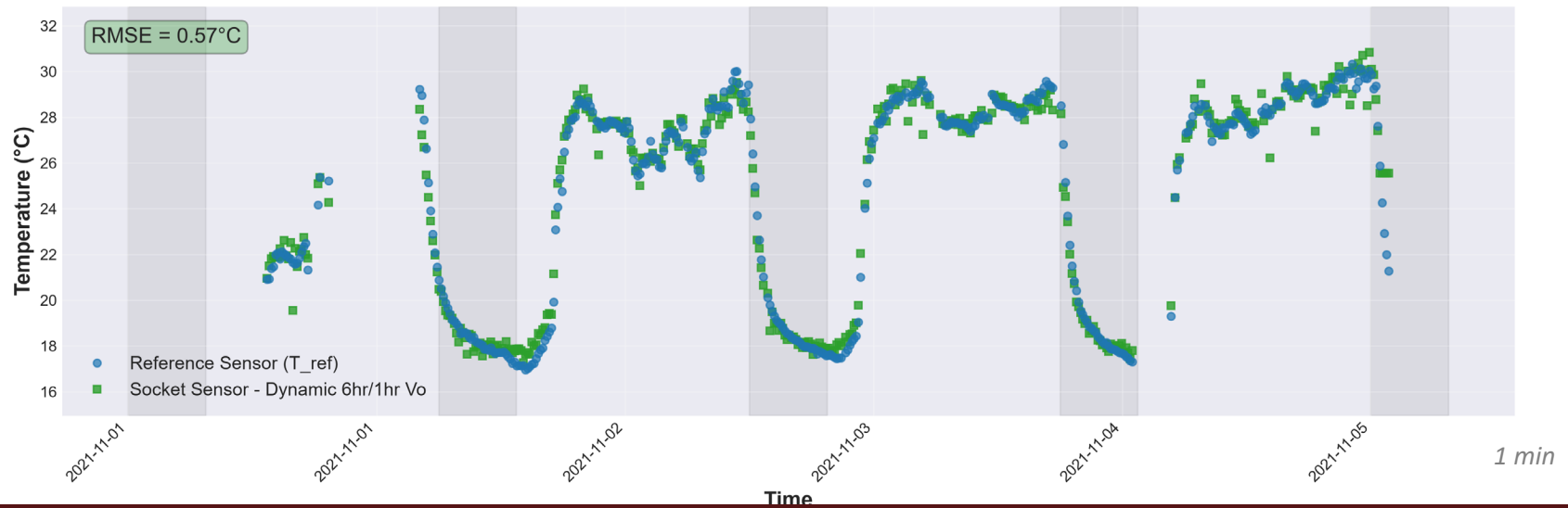
Results: Adaptive Calibration Performance

Participant P14_1

Static Calibration: Default Vo (Fixed Throughout Trial)



Dynamic Calibration: Recalibrated Every 6 Hours

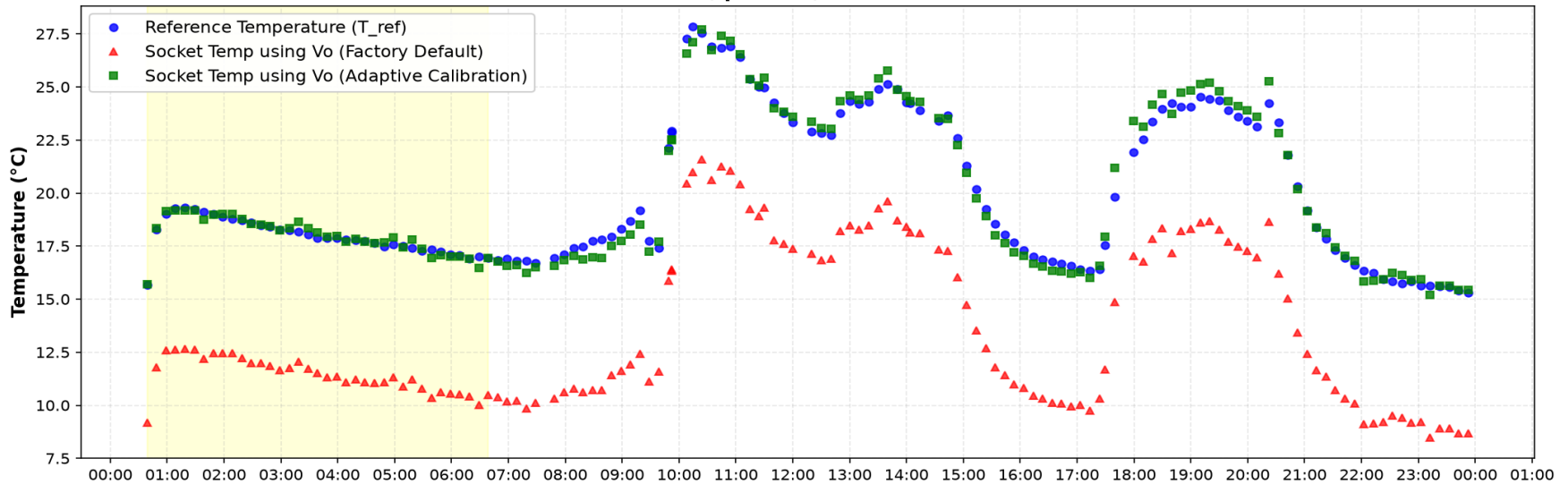


Results: Time Evolution – Temperature reconstruction

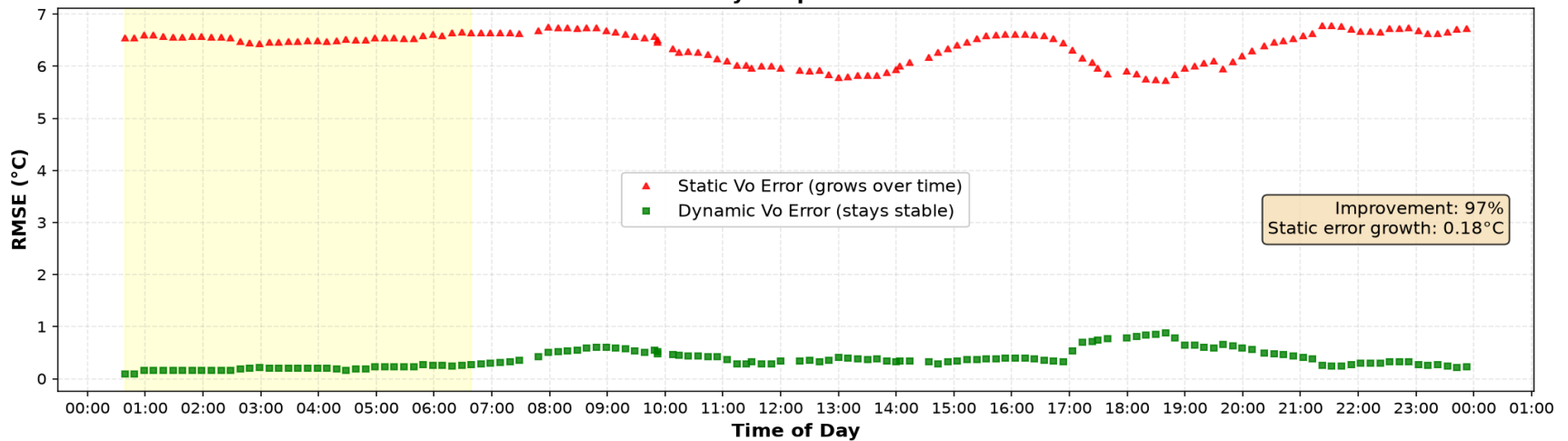
Participant P20 | 2022-02-24

P20 Time-Evolution: Temperature Reconstruction Over 2022-02-24

$\alpha=0.3870$, $\beta=1485$, $V_o(\text{static})=1.612$



Error Evolution: Why Adaptive Calibration Matters



Results: Extended Wear (2+ months — Participant P02)

P02: Extended Wear Validation (Full Trial Duration)

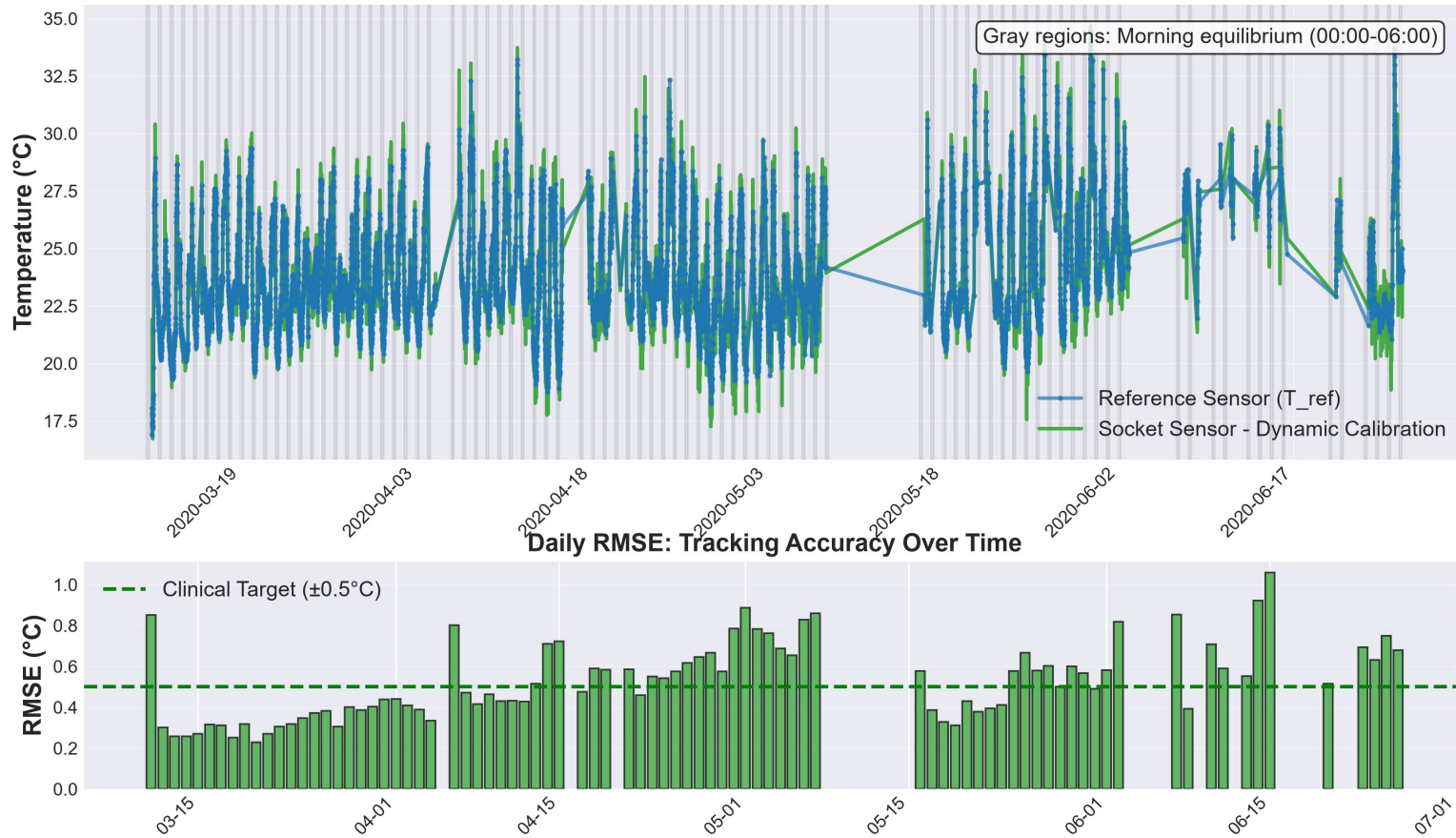
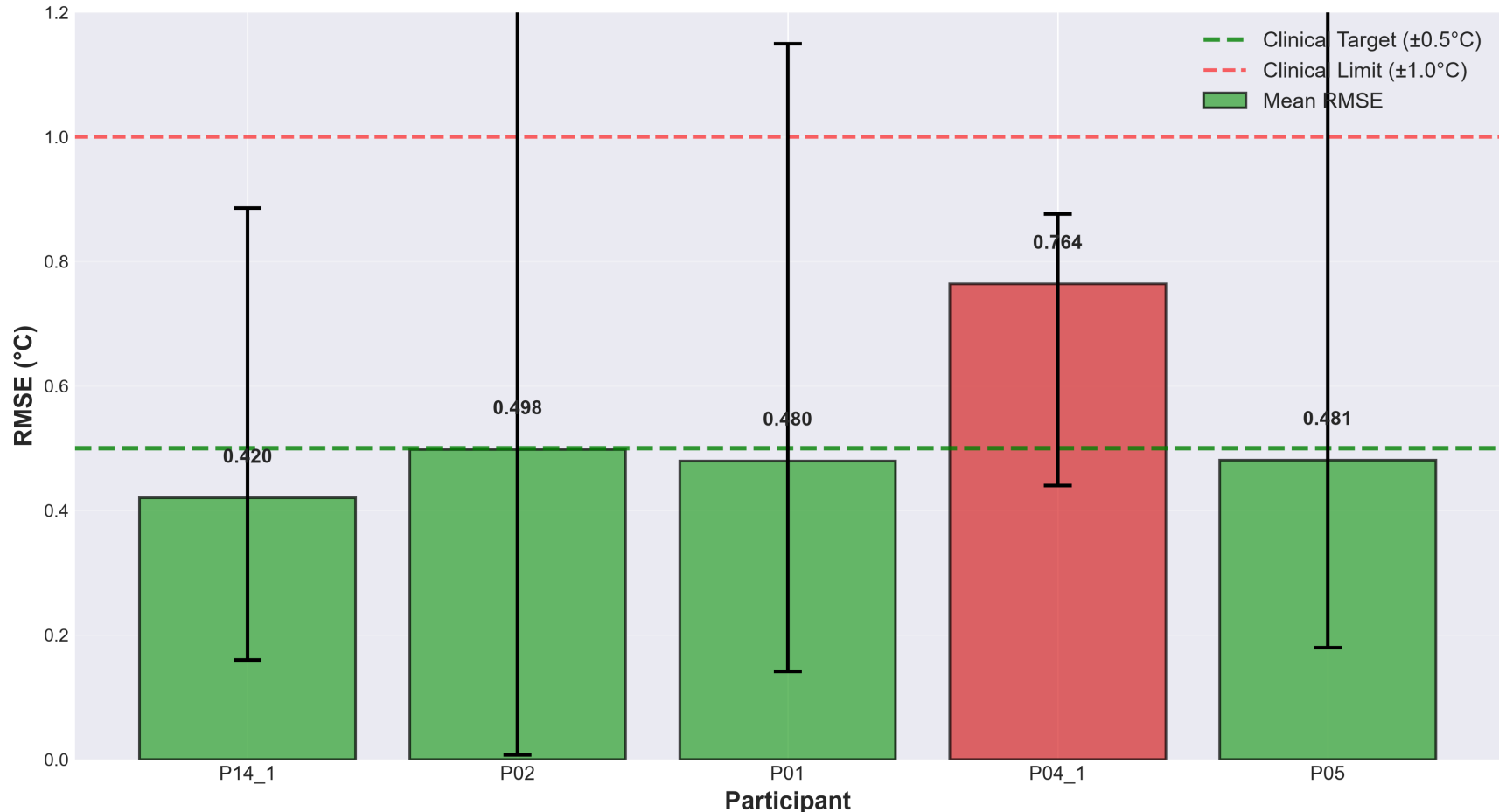


Figure 4:
Extended wear
validation

- Full trial duration: blue scatter (T_{ref}) vs. green line (adaptive T_{socket})
- Shaded morning regions (00:00–06:00): calibration windows
- Consistent $RMSE \leq 0.7^{\circ}C$ throughout extended wear — **no degradation**

Results: Error RMSE Comparison (5 Participants)

RMSE Cohort Summary: Dynamic Calibration Accuracy

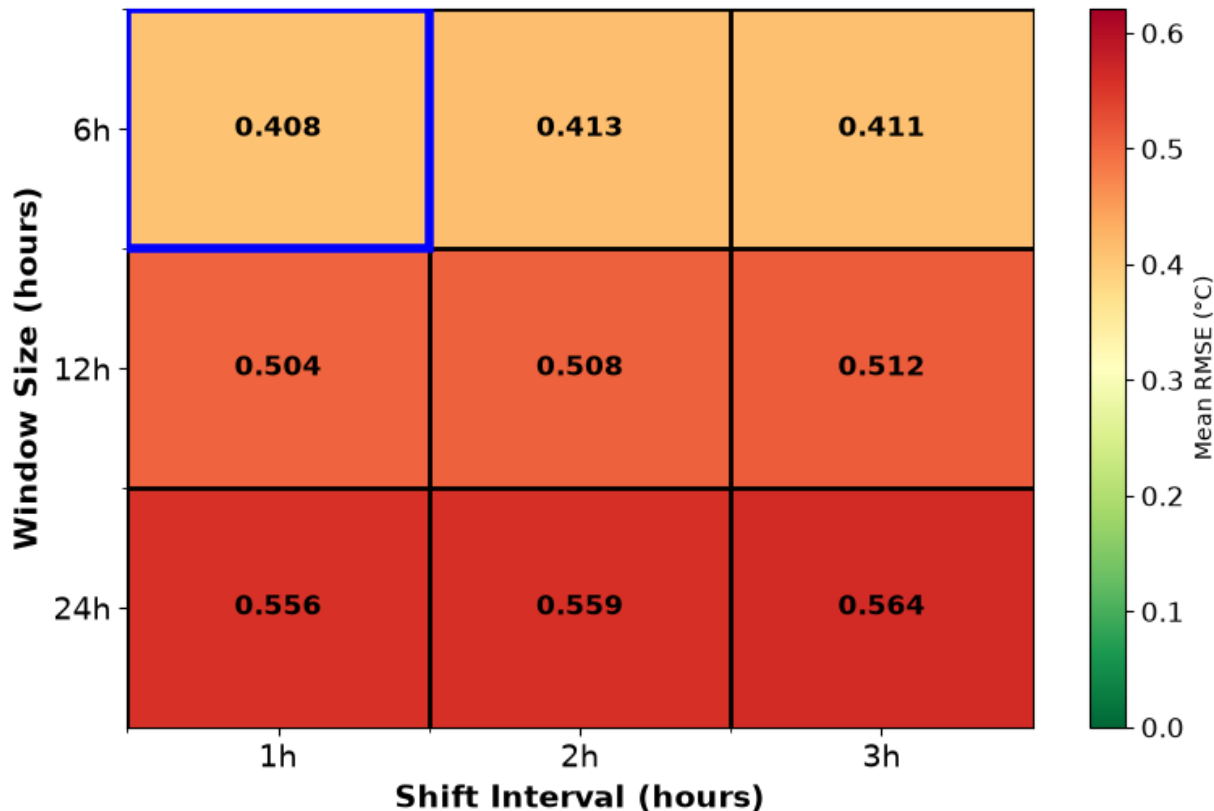


- Mean RMSE of 0.3–0.5°C across all participants (green zone)
- Sub-degree trend detection enables early warning of thermal changes
- Consistent performance: algorithm generalizes across morphologies and wear patterns

1 min

Results: Window Selection Optimization (Grid Search)

**Grid Search: Window/Shift Configuration Analysis
(6h/1h highlighted as optimal)**



- **9-combination grid:** window sizes (6/12/24hr) × shift intervals (1/2/3hr)
- **Cell colors:** green (low RMSE ~0.25°C) → yellow → red (high ~0.7°C)
- **6hr/1hr (highlighted):** optimal empirical trade-off

Discussion: Physics & Clinical Application

- **Six-hour calibration provides the best balance between responsiveness and stability.** Shorter windows are more sensitive to noise, whereas longer windows smooth genuine sensor drift.
- **For the operating lifetime investigated,** sensor drift is dominated by changes in the baseline voltage (V_0), while α and β remain effectively constant as intrinsic sensor properties.
- **Natural overnight calibration** enables continuous monitoring without external equipment or user intervention.
- **Sub-degree accuracy ($<0.5^\circ\text{C RMSE}$)** is sufficient for monitoring thermal trends associated with prolonged prosthesis wear.

Take-home Messages

- ✓ **Dynamic calibration maintains clinically useful temperature accuracy (<0.5 °C RMSE).**
- ✓ **Only the baseline parameter (V_0) requires periodic updating, while α and β remain stable over the operating lifetime investigated.**
- ✓ **The framework can be extended to other long-term wearable temperature sensing applications.**

Future Work

- Clinical validation in a larger amputee cohort

Thank You

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