

# Estimation of Lower-limb Joint Kinematics with Inertial Measurement Units

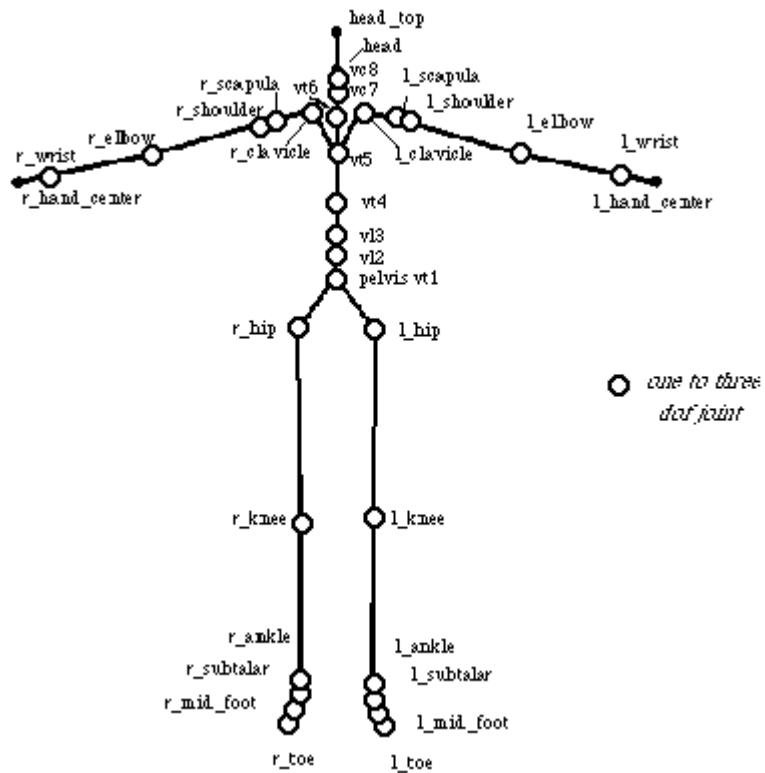
Team 1

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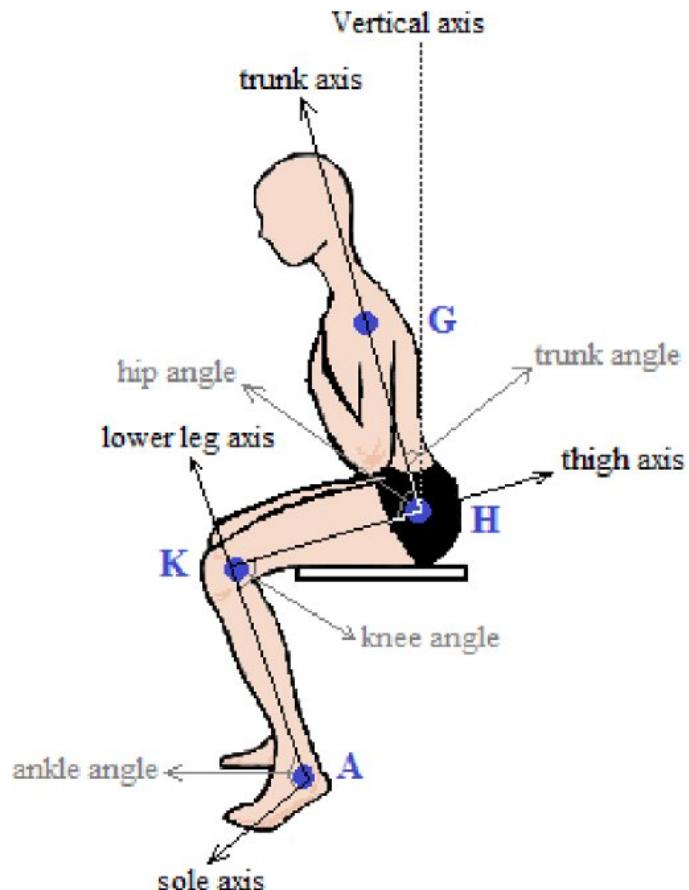
Final Presentation



# What is joint kinematics?

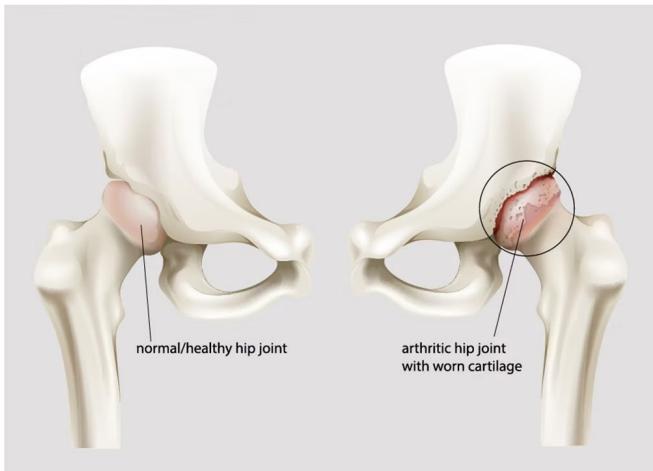


<https://www.euclideanspace.com/physics/kinematics/joints/index.htm>

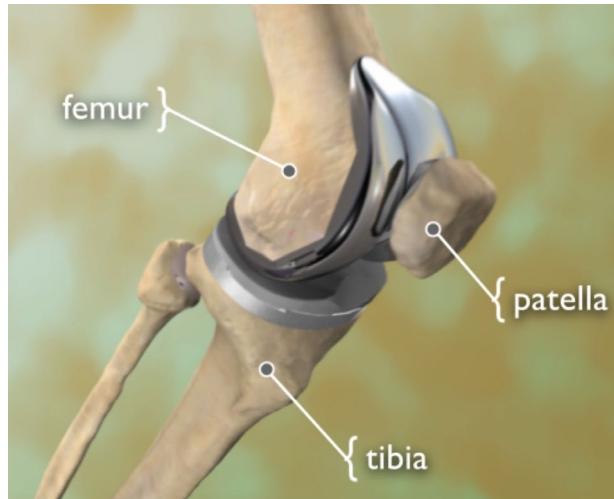


[https://www.researchgate.net/publication/312422406\\_Influence\\_of\\_time\\_restriction\\_20\\_minutes\\_and\\_946\\_months\\_of\\_visual\\_information\\_on\\_an\\_gular\\_displacement\\_during\\_the\\_sit-to-stand\\_STS\\_task\\_in\\_three\\_planes/figures?lo=1](https://www.researchgate.net/publication/312422406_Influence_of_time_restriction_20_minutes_and_946_months_of_visual_information_on_an_gular_displacement_during_the_sit-to-stand_STS_task_in_three_planes/figures?lo=1)

# Why do we need joint kinematics estimation?



<https://stiwell.medel.com/orthopaedics/hip-arthrosis>



<https://orthoinfo.aaos.org/en/treatment/total-knee-replacement-animation/>



[https://en.wikipedia.org/wiki/Achilles\\_tendinitis](https://en.wikipedia.org/wiki/Achilles_tendinitis)

Traditional approaches use cumbersome and expensive marker-based systems for motion capture



01

## Accuracy of Full-Body Kinematics

State Estimation Filters: Madgwick, Mahony, Complementary, EKF, Xsens

Biomechanical Modelling: OpenSense

02

## Sensitivity to Sensor-to-Body Calibration Errors

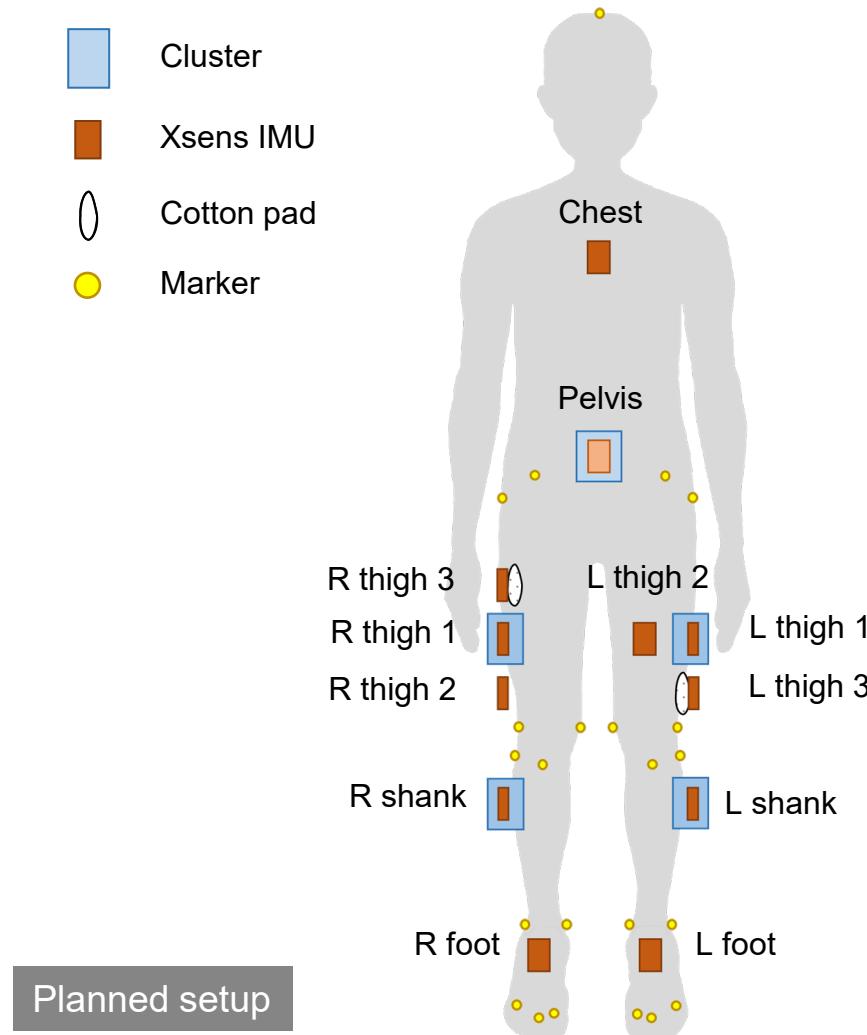
Evaluate the impact of sensor placement errors on estimation accuracy

03

## Sensitivity to Skin-motion Artifact

Investigate the influence of skin-motion artifact on the quality of kinematic estimations

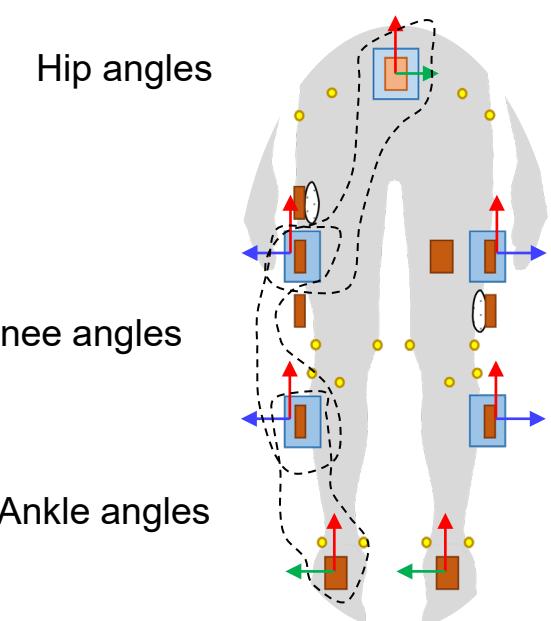
# Experimental setup



# Different sensor configurations for different aims of this study

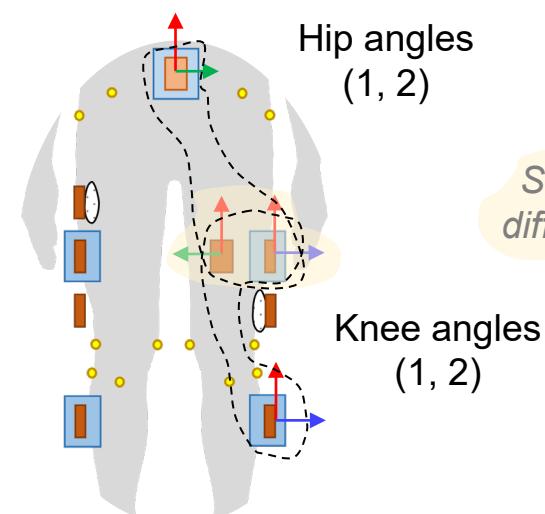
Aim 1

Accuracy of lower-body kinematics using (1) state- estimation filters, and (2) combination of state-estimation filters and biomechanical model



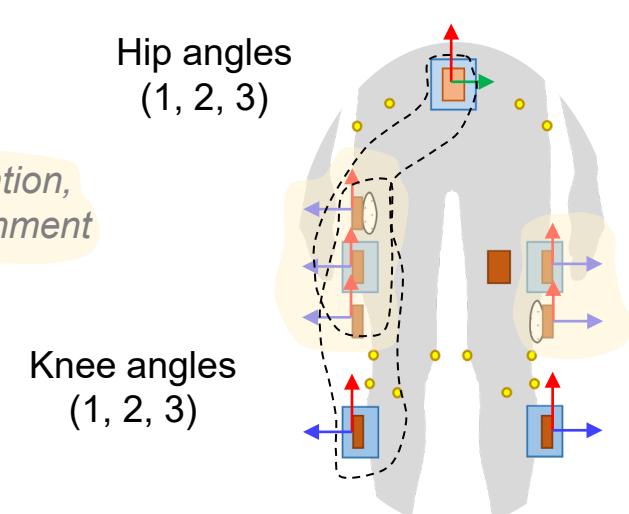
Aim 2

Sensitivity analysis of sensor-to-body calibration errors

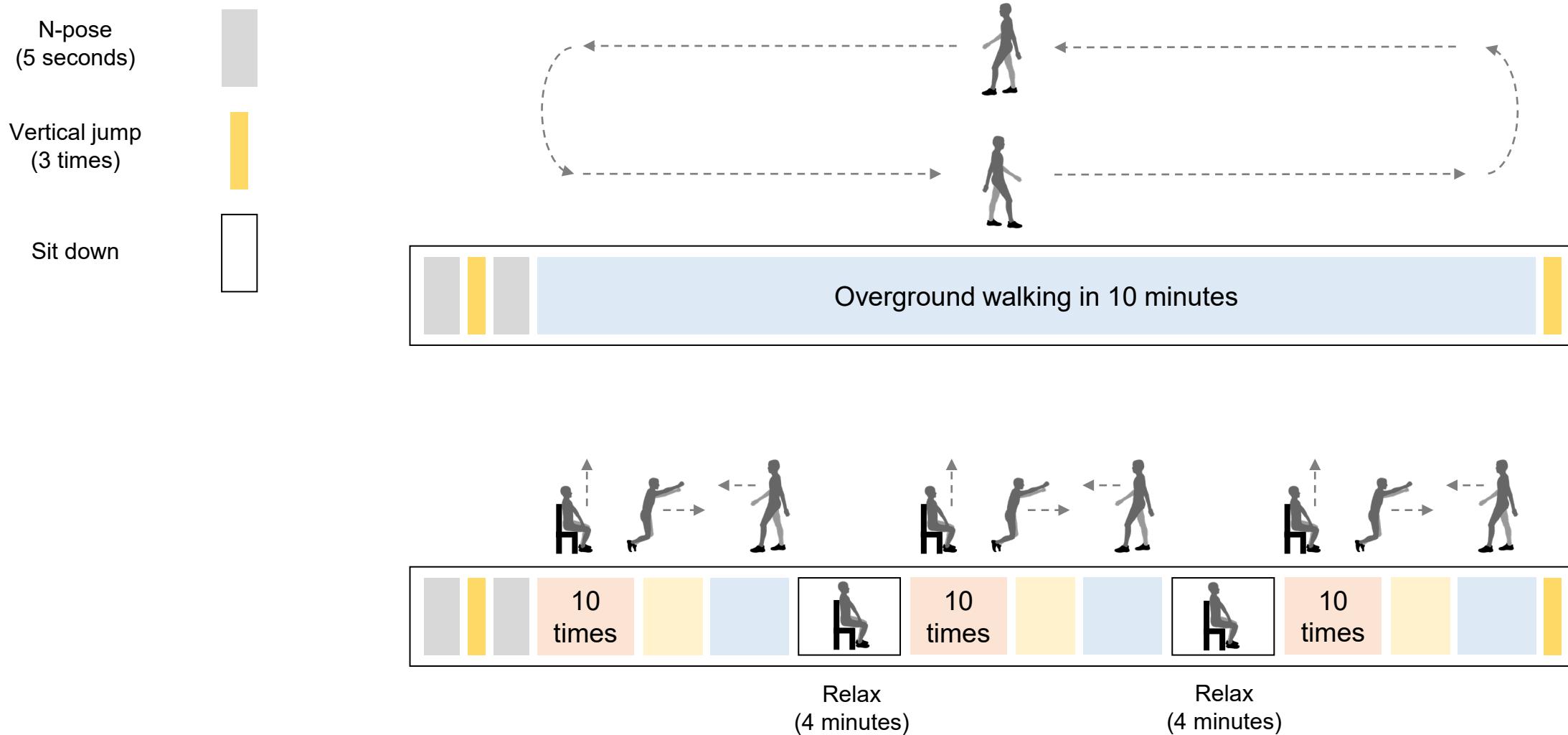


Aim 3

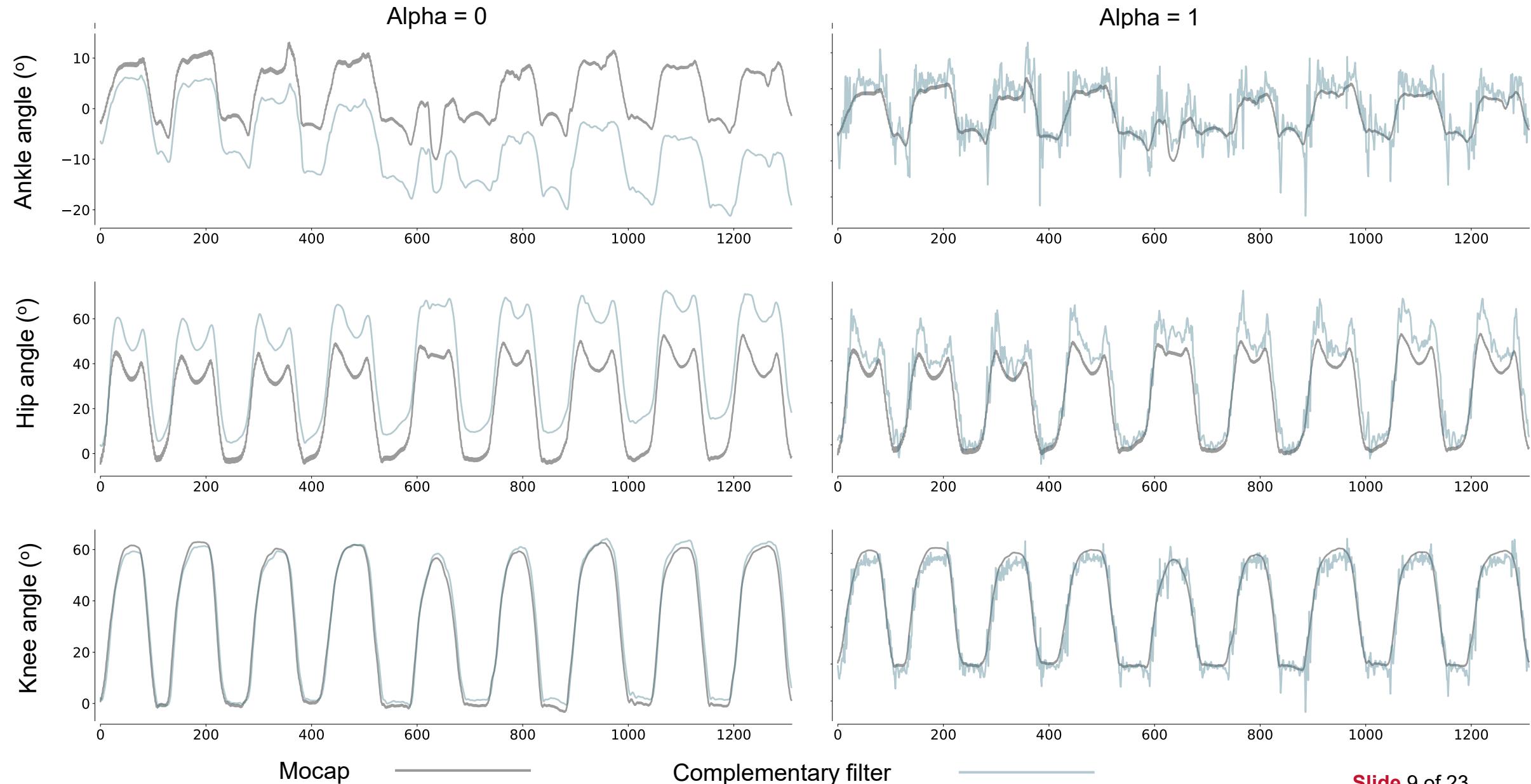
Sensitivity analysis of skin motion artifacts in different tasks



# Various tasks were performed, but only sit-to-stand provided high-quality data



# Different sensor fusion methods were manually tuned



# How generalizable is tuning?

## **Preliminary observation**

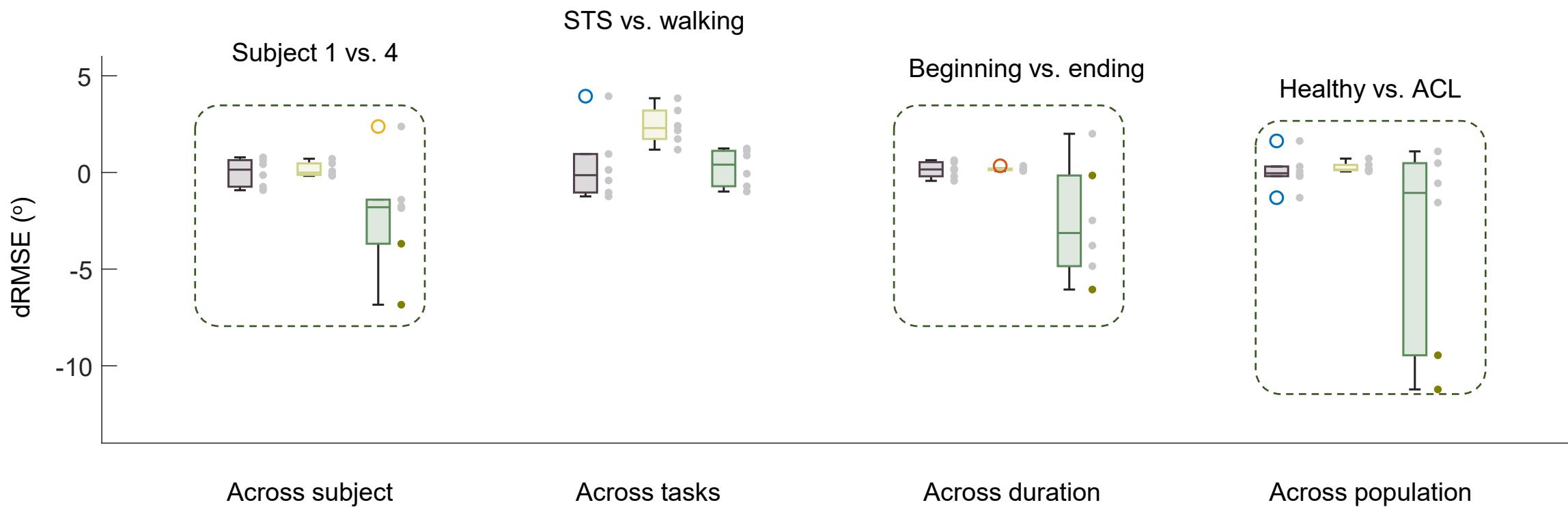
Offset on the estimation of EKF that can be resolved by tuning its parameters

## **Decision**

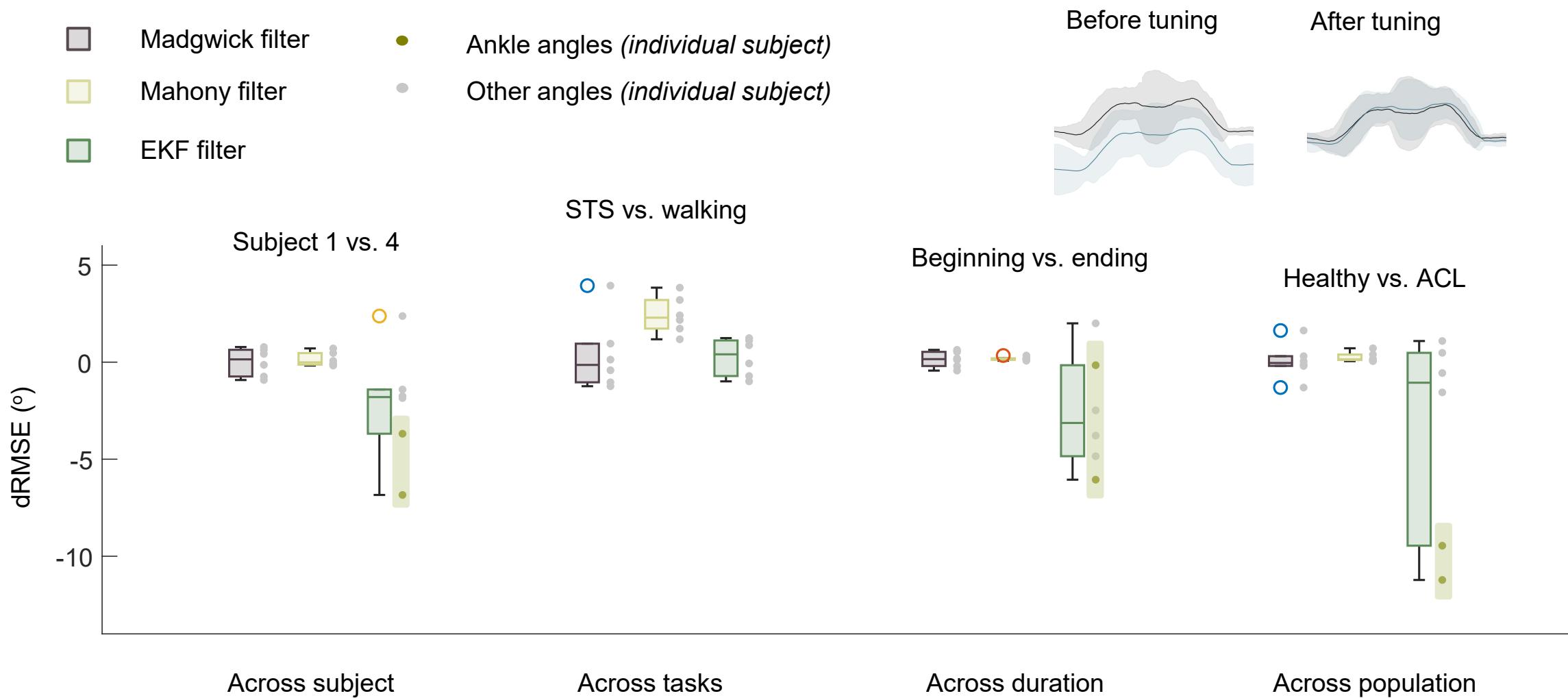
Tuned all filters before performance evaluation

EKF tuning may be generalizable to different scenarios, e.g., across subject, duration, and population

- █ Madgwick filter
  - █ Mahony filter
  - █ EKF filter
- Ankle angles (*individual subject*)
  - Other angles (*individual subject*)



# Tuning may be beneficial for specific joint angles



Which sensor fusion algorithm perform the best?

Would the biomechanical model improve the estimation?

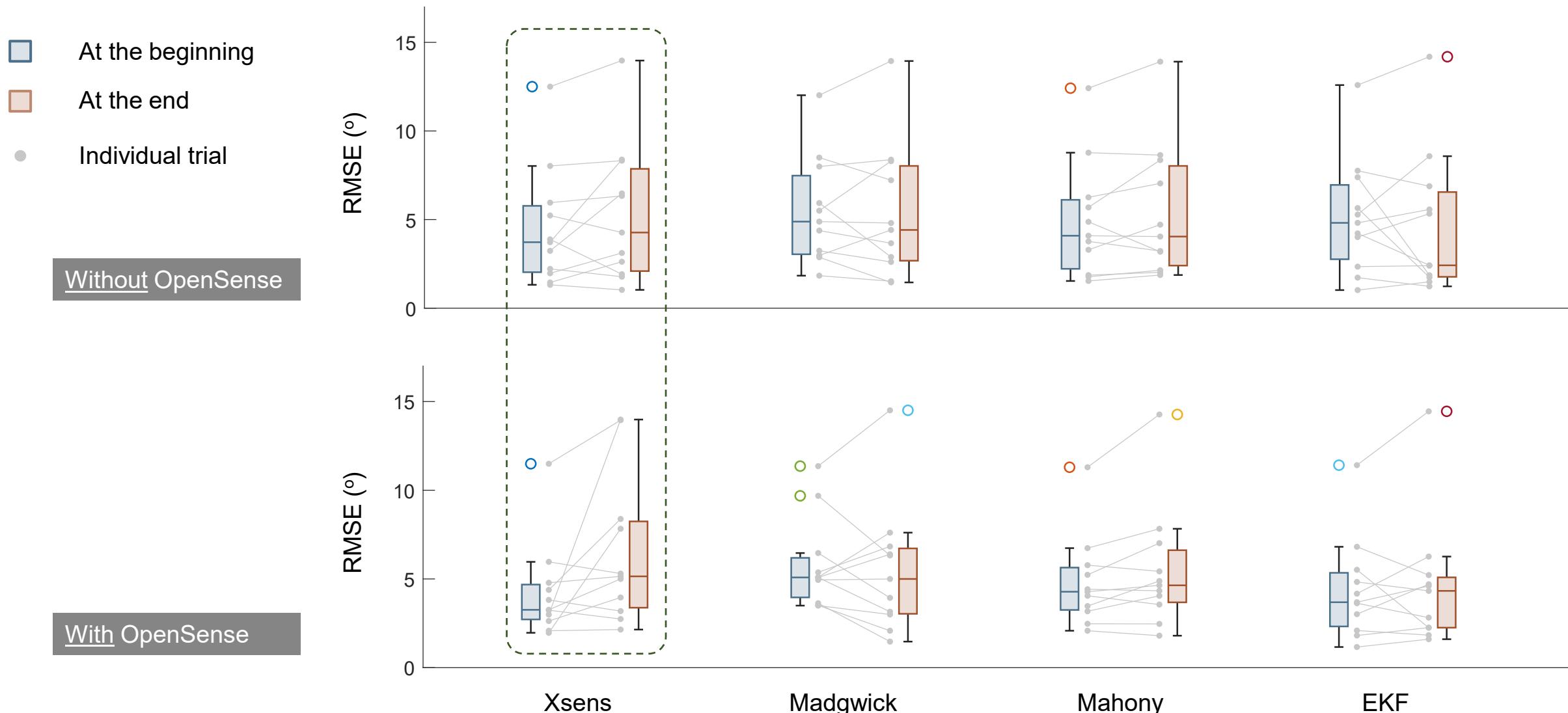
**Preliminary observation**

Traditional complementary filter resulted in very poor estimation

**Decision**

Removed from the analysis

# Sensor fusion methods have similar performance, while Xsens built-in filter may not be tuned for long-duration recordings



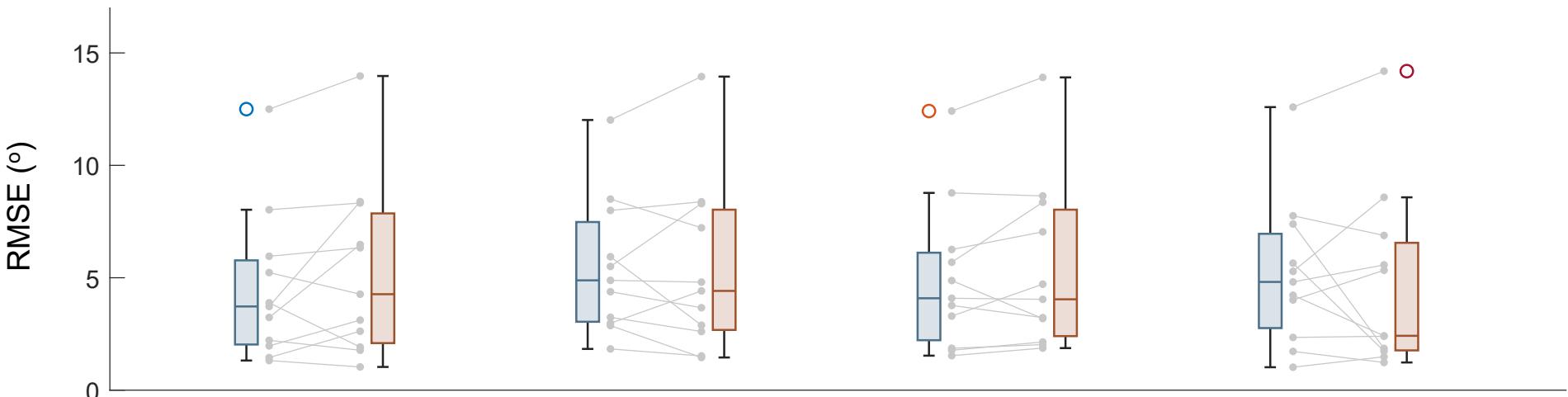
# No clear evidence of using biomechanical model to improve estimation performance

At the beginning

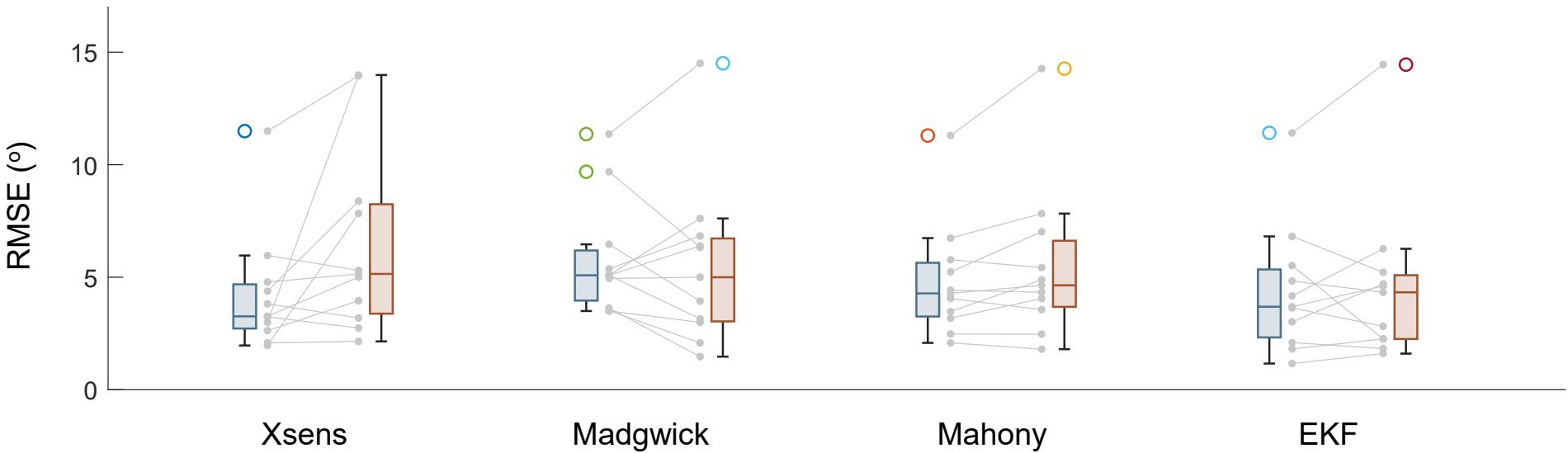
At the end

Individual trial

Without OpenSense



With OpenSense



How does **sensor misplacement** and **skin/soft motion artifacts** affect the estimation accuracy?

Would the biomechanical model (from OpenSense) be helpful?

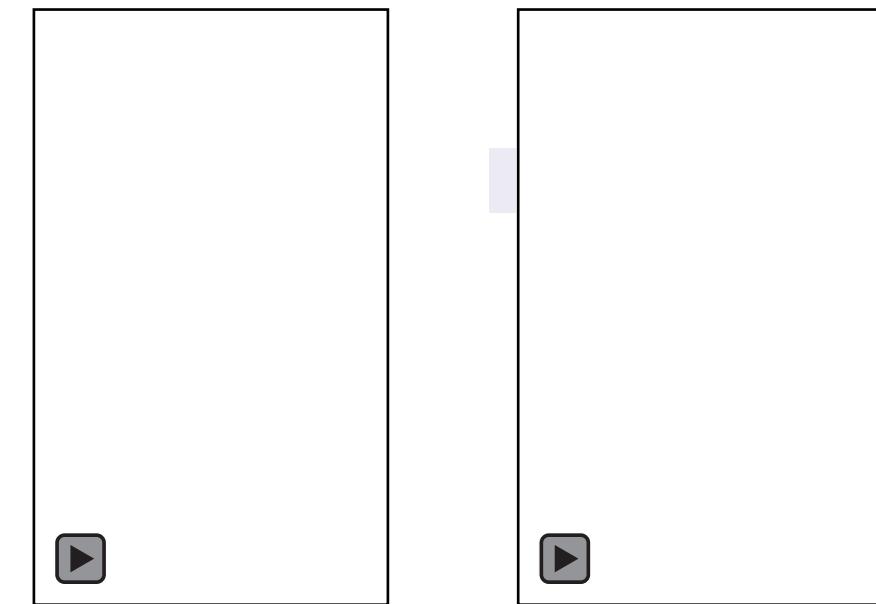
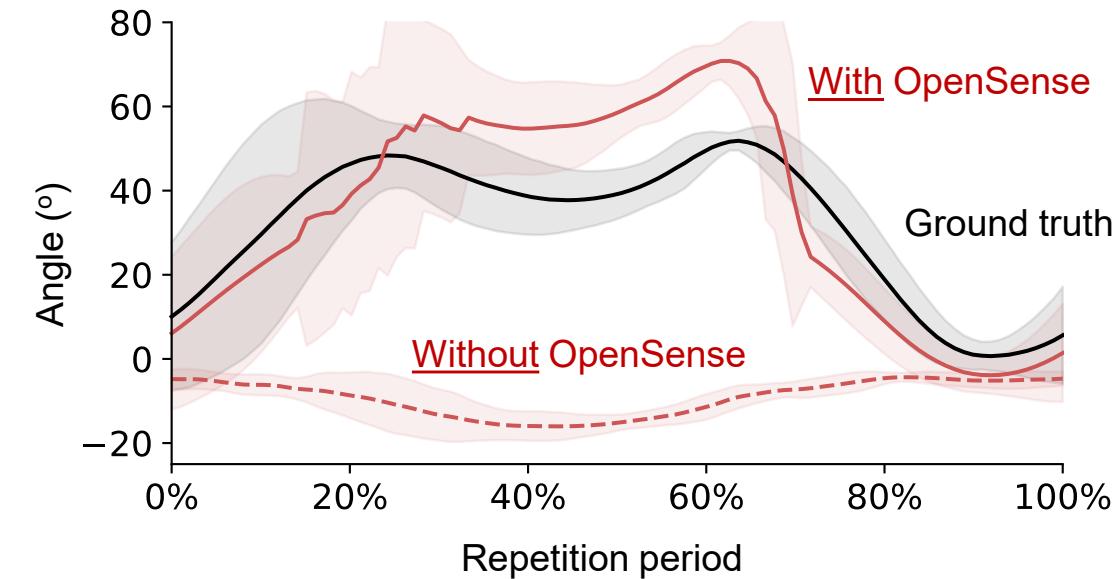
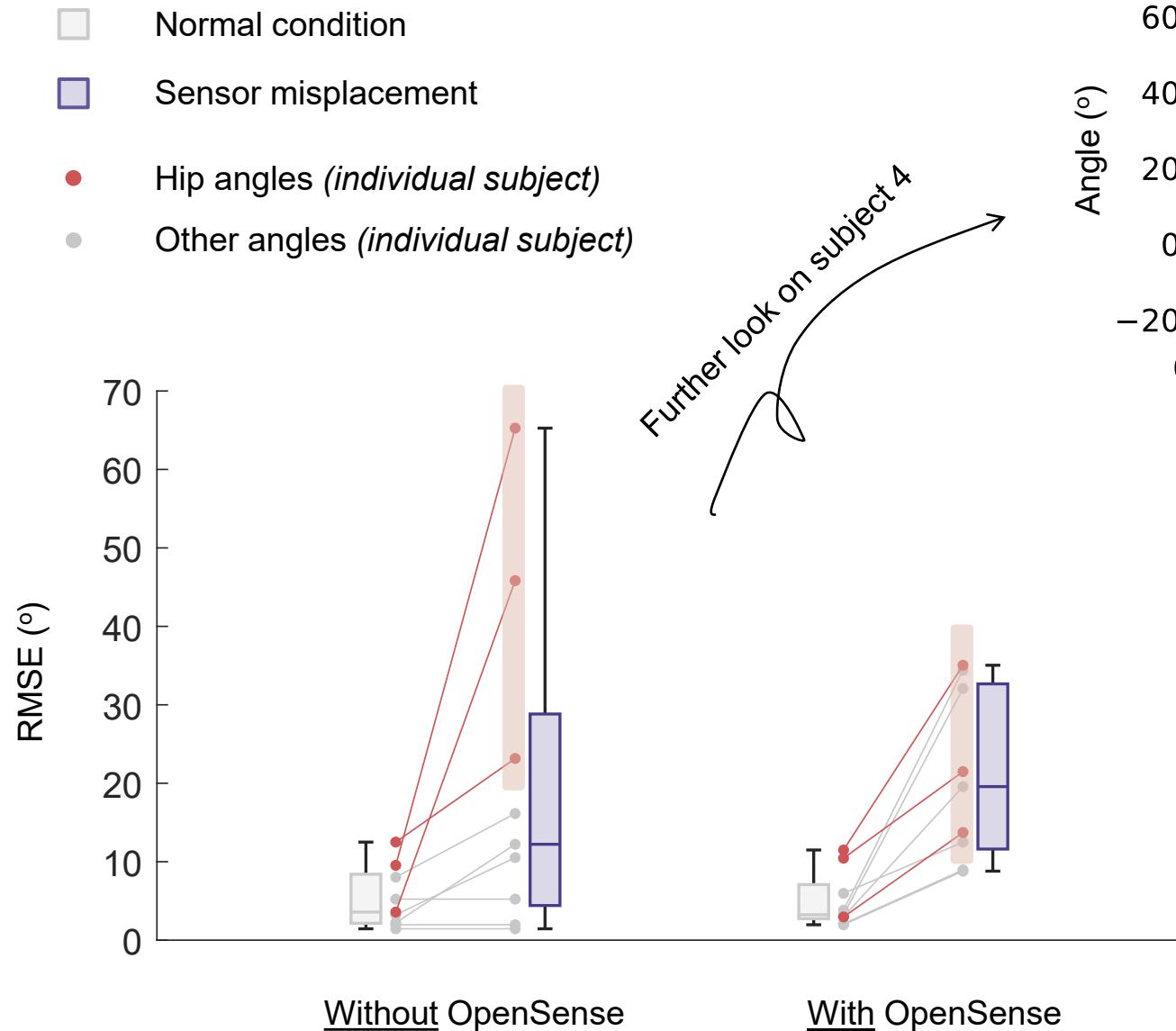
### **Preliminary observation**

Trend across sensor fusion methods (e.g., increase in error as a sensor is misplaced, etc.)

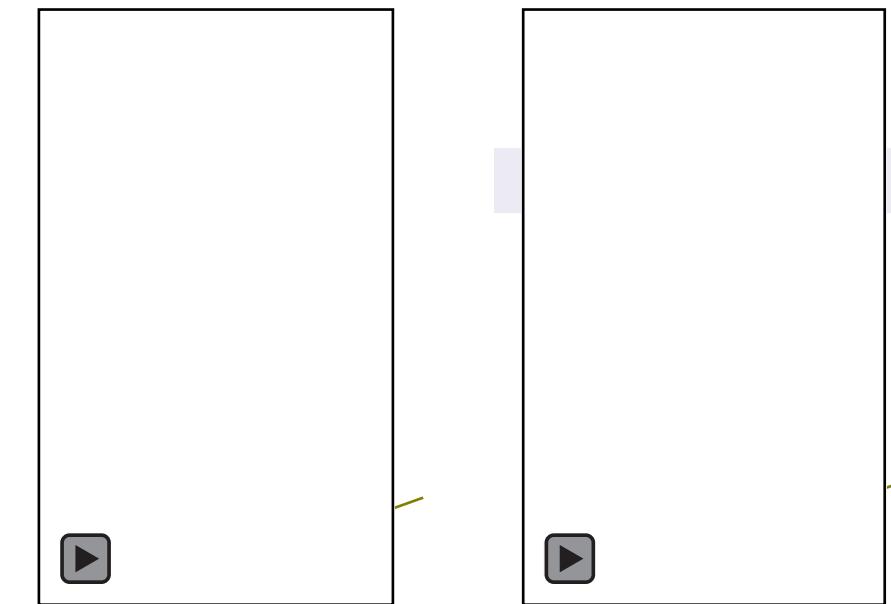
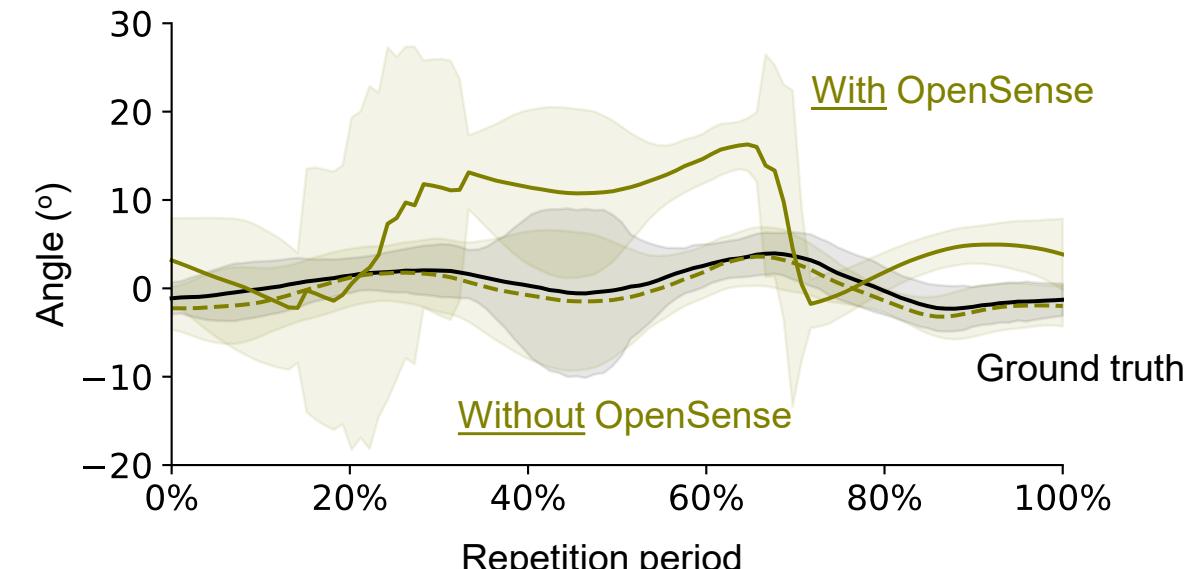
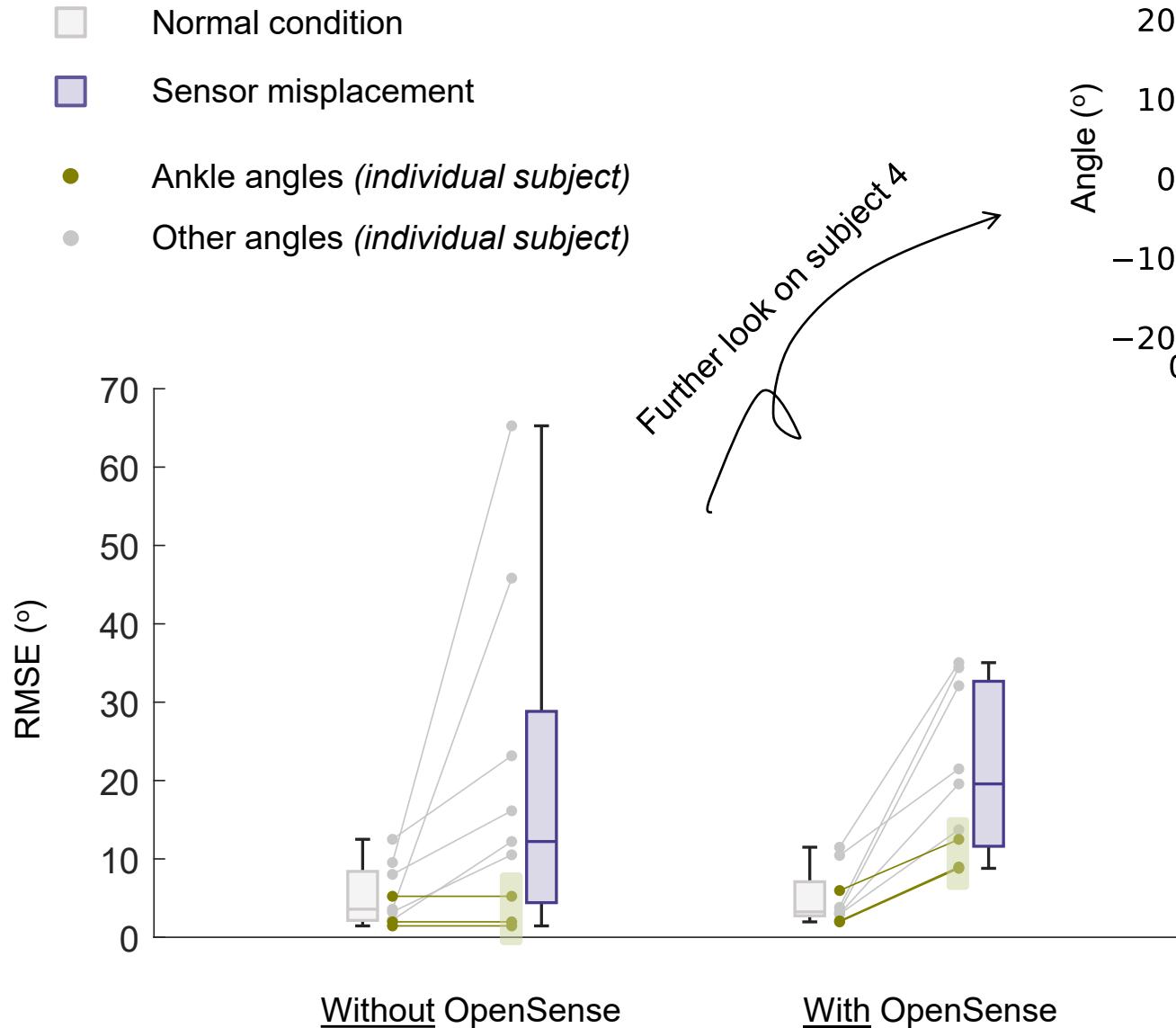
### **Decision**

Selected Xsens built-in filter (i.e., standard, commonly used, etc.)

# OpenSense improves the estimation of **hip angles** given sensor misplacement



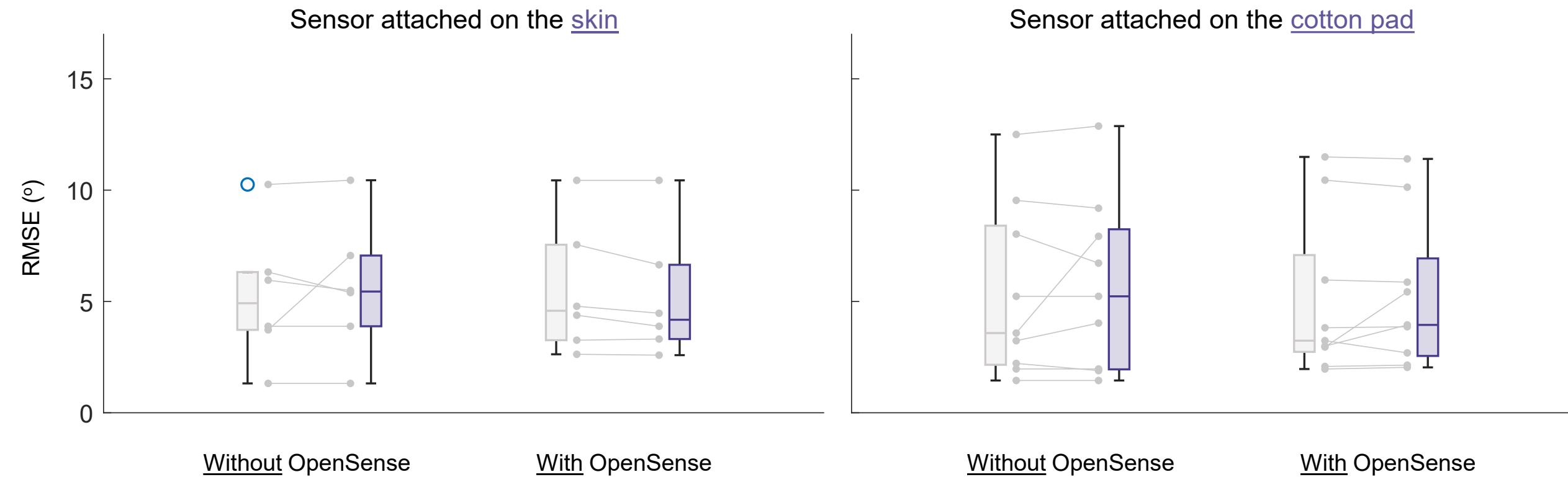
# However, estimation of ankle angles was compensated



# Tight wrapping of sensors could remove the skin/soft tissue motion artifacts

- Normal condition
- Sensor misplacement
- Angles (*individual subject*)

Note: Sensors were wrapped tightly on the body segments.

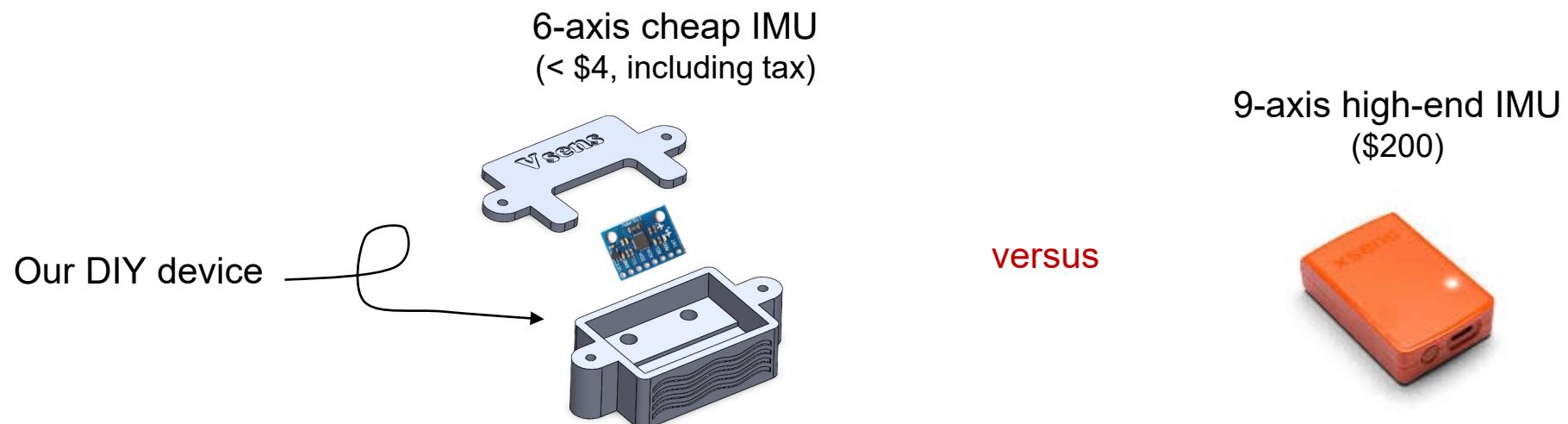


\*One subject was removed from this analysis due to poor-quality motion capture data

## How would these knowledge be valid and translational to real life?

*“All theory, dear friend, is grey, but the golden tree of actual life springs ever green”*

- Johann Wolfgang von Goethe

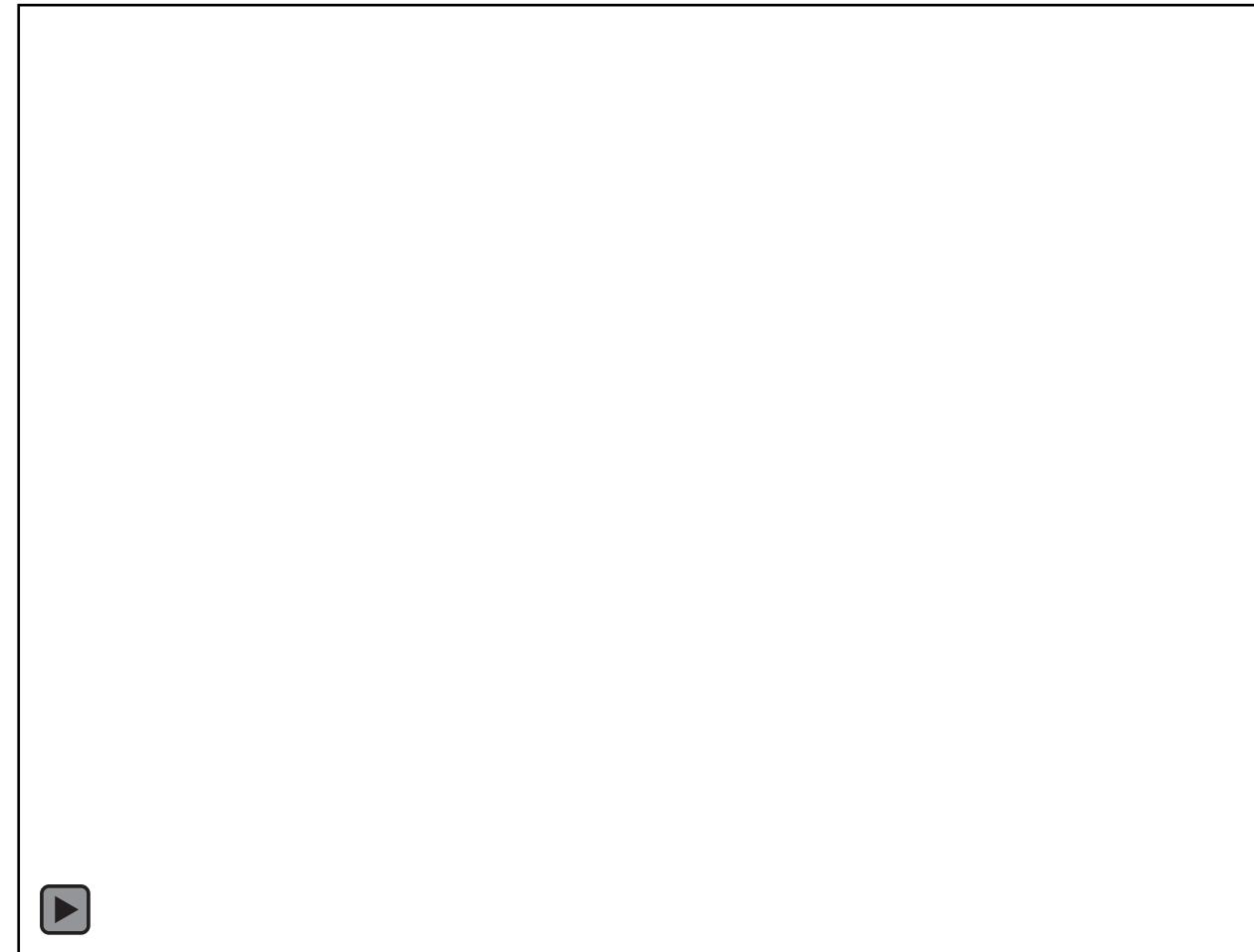
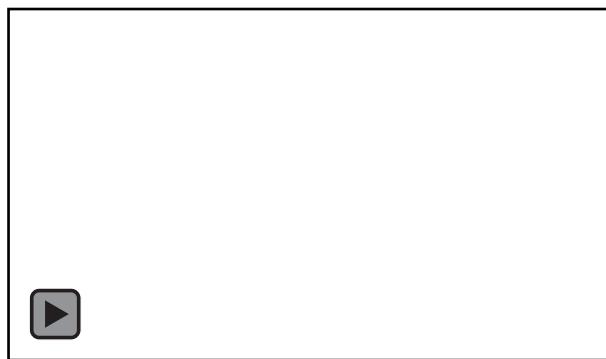


# Implementation perspectives

Lack of magnetometer

Low robustness

User experience



# Takeaway messages

*(Tuning generalizability)* The Kalman filter family tuning may be generalizable across various scenarios and tuning may be more generalizable for certain joints, specifically the ankle.

*(Filter generalizability)* The complimentary filter family may be more generalizable across fields (i.e., no improvement after tuning)

No significant difference between complementary and Kalman filter family given sufficient information (i.e., accelerometer, gyroscope, and magnetometer)

No clear evidence that OpenSense improves the joint kinematics estimation

OpenSense may improve estimation accuracy of a specific joint angle but with some cost on the other angles

Tight wrapping of sensors could remove the skin/soft tissue motion artifacts

## Limitations

Limited sample size

Only sit-to-stand task

Thank you for your listening!  
Q&A?

