

# **Agreement of Gait Events Detection by Inertial Measurement Units and an open source Markerless End-to-End Workflow?**

Ahmed Yacine Lardjane, Mathias Blandeau, Laura Wallard, Christopher Paglia, Christophe Gillet, Sébastien Leteneur, Emilie Simoneau-Buessinger

# **To cite this version:**

Ahmed Yacine Lardjane, Mathias Blandeau, Laura Wallard, Christopher Paglia, Christophe Gillet, et al.. Agreement of Gait Events Detection by Inertial Measurement Units and an open source Markerless End-to-End Workflow?. 2024. hal-04737941

# **HAL Id: hal-04737941 <https://hal.science/hal-04737941v1>**

Preprint submitted on 15 Oct 2024

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



[Distributed under a Creative Commons Attribution 4.0 International License](http://creativecommons.org/licenses/by/4.0/)

# **Agreement of Gait Events Detection by Inertial Measurement Units and an open source Markerless End-to-End Workflow?**

#### Ahmed Yacine LARDJANE \*

*LAMIH, CNRS, UMR 8201, Univ. Polytechnique Hauts-de-France, Valenciennes, France.* 

*IRT Railenium, Valenciennes, France.*

#### Mathias BLANDEAU

*LAMIH, CNRS, UMR 8201, Univ. Polytechnique Hauts-de-France, Valenciennes, France*

#### Laura WALLARD

*LAMIH, CNRS, UMR 8201, Univ. Polytechnique Hauts-de-France, Valenciennes, France*

#### Christopher PAGLIA

*IRT Railenium, Valenciennes, France.*

#### Christophe GILLET

*LAMIH, CNRS, UMR 8201, Univ. Polytechnique Hauts-de-France, Valenciennes, France*

#### Sébastien LETENEUR

*LAMIH, CNRS, UMR 8201, Univ. Polytechnique Hauts-de-France, Valenciennes, France*

#### Emilie SIMONEAU-BUESSINGER

*LAMIH, CNRS, UMR 8201, Univ. Polytechnique Hauts-de-France, Valenciennes, France*

\***Corresponding author.** Email: [AhmedYacine.Lardjane@uphf.fr](mailto:AhmedYacine.Lardjane@uphf.fr) 

**Keywords**: Spatiotemporal ; IMU ; Markerless ; Gait ; Pose2Sim

#### **1. Introduction**

Precise spatiotemporal calculation and proper analysis of kinematic and kinetic variables during walking depend on gait cycle events, usually detected by force plates. Those variables are widely used by clinicians, but also in urban planning, particularly for pedestrian crossing where walking speed is investigated in different environments, such as roads or railways stations (Grison et al. 2022). Motion capture outside laboratories is challenging, leading to the rise of accelerometer-based Inertial Measurement Units (IMUs) as a popular alternative. Markerless motion analysis technologies have emerged, with numerous commercial systems available. However, these

systems can be costly and often rely on proprietary code and datasets for their training (Wade et al. 2022). Pose2Sim represents a recent breakthrough in markerless motion capture. This open source workflow uses 2D pose estimation data from videos, leveraging validated open source pose estimation models such as OpenPose and AlphaPose, as well as custom-trained models through open source tools like DeepLabCut. Pose2Sim identifies and associates individuals, accurately triangulates joint positions, and integrates with an OpenSim model (Pagnon et al. 2022). It supports any type of camera, broadening its applicability across various study settings. Despite extensive research in IMU and kinematics for walking event estimation, literature lacks a comparative analysis between open source, markerless motion capture solutions and gold standards. This study aims to evaluate the agreement of three open source algorithms for walking events detection from force plates, IMU and videos.

### **2. Methods**

#### *2.1 Protocol*

Four Kistler Type 9286 force plates (500Hz), 4 Delsys Trigno IMUs (370Hz) and 5 Miqus video cameras (25 Hz), mounted on the wall, ensuring a full view of the entire walkway were used in this study. All devices were synchronized by Qualisys software. IMUs were attached to the participants' first metatarsals and the lateral aspects of their heels. The study included one healthy young male (age: 31 yrs, 1.73 m, 71 kg) who walked on a 5-meters corridor equipped with the force platforms, turned around, and returned. A valid trial required a complete foot contact per force plate. The participant kept a forward gaze, avoiding force plate targeting. A total of 125 shoed and 125 barefoot walks were completed. Rest periods were given every 15 trials, and a 5-minutes break after 45 trials.

#### *2.2 Data analysis*

For both conditions, reference events were identified using force platform data on Matlab, with heel strikes (HS) and toe-offs (TO) when the vertical force exceeded or dropped below 20N, respectively (Blandeau et al. 2023). For the IMU, a first-order lowpass Butterworth filter with a 10Hz cutoff was applied to acceleration data, while a first-order band-pass Butterworth filter, spanning 0.001 to 5Hz was applied to gyroscope data. Then both acceleration and gyroscope data were smoothed using a Savitzky-Golay filter on the flexion axes. Maximum and minimum points of these filtered signals were identified as HS and TO, respectively (Gottlieb et al. 2020). For videos processing, body's keypoints were identified using the opensource BODY\_25B OpenPose model as it is the

most accessible pre-trained model (Stenum et al. 2023) and were processed using Pose2Sim end-to-end workflow. Events were then identified as proposed in Zeni, et al. (2008). All codes are open source and available on https://github.com/yacine1007/imupose2sim-fp

### **3. Results and discussion**

In total, 305 and 350 footstrikes were validated in barefoot and shoed conditions, respectively. The mean error results are presented in Table 1 and Figure 1.

**Table 1**. *Mean error and SD (in seconds) between IMU and video algorithms and force platforms*.

	Heel Strike		Toe-Off	
	IMU	Video	IMU	Video
<b>Barefoot</b>	$0.060*$	$0.005*$	$0.024*$	$-0.068*$
	$\pm 0.020$	$\pm 0.022$	$\pm 0.010$	$\pm 0.047$
Shoed	$0.030*$	$-0.006*$	$0.028*$	$-0.050*$
	$\pm 0.020*$	$\pm 0.018$	$\pm 0.006$	$\pm 0.050$

\* Represents a significant difference  $(p<0.05)$  in gait event detection between IMU or Video and force platforms



*Figure 1. Mean errors distributions (in seconds) between algorithms and force platforms*

A Shapiro-wilk test was applied to verify the normality of the data, followed by a non-parametric Friedman ANOVA in all conditions. For both barefoot and shoed conditions, the applied tests revealed a significant difference between gait event detection for both HS and TO obtained with IMU and video compared to data obtained with force platforms,  $(p < 0.001)$ . Our results are in line with the studies of Blandeau et al. (2023) and Stenum et al. (2023). In both shoed and barefoot conditions, IMUs errors were less dispersed than videos' errors (Figure1), this might be due to optimal IMU placement on the metatarsal and their higher

capture rate (Gottlieb et al. 2020). The distribution of video errors was mostly negative, suggesting a detection that occurred after the event happened. This could be attributed to a misdetection by the model and to the low recording rate. Improving the recording rate might enhance the model accuracy.

# **4. Conclusions**

This study aimed to evaluate the agreement among open source algorithms using force plate data, video analysis and IMU for gait event detection. All three methods detected walking events (HS and TO). Although errors were detected, results align with previous studies, encouraging the use of open source models. However, it must nevertheless be done carefully. Further investigations should be conducted on larger samples under different conditions.

# **5. Conflict of Interest Statement**

The authors declare no conflict of interest.

## **6. References**

- Blandeau M, Lesaffre R, Wallard L. 2023. Validation of walking events detecting algorithm during Nordic walking of healthy subjects. Computer Methods in Biomechanics and Biomedical Engineering: Imaging &Visualization https://doi.org/10.1080/10255842.2023.2246304
- Gottlieb U, Balasukumaran T, Hoffman JR, Springer S. 2020. Agreement of Gait Events Detection during Treadmill Backward Walking by Kinematic Data and Inertial Motion Units. Sensors. https://doi.org/10.3390/s20216331
- Grison E, Aupetit S, Escaich S, Morgagni S. 2022. Using virtual reality to evaluate the effects of two safety systems at pedestrian track crossings on human behaviour.
- Pagnon D, Domalain M, Reveret L. 2022. Pose2Sim: An open-source Python package for multiviewmarkerless kinematics. JOSS. https://doi.org/10.21105/joss.04362
- Stenum J, Hsu MM, Pantelyat AY, Roemmich RT. 2023. Clinical gait analysis using video-based pose estimation: multiple perspectives, clinical populations, and measuring change https://doi.org/10.1101/2023.01.26.23285007
- Wade L, Needham L, McGuigan P, Bilzon J. 2022. Applications and limitations of current markerless motion capture methods for clinical gait biomechanics. PeerJ.

https://doi.org/10.7717/peerj.12995

Zeni JA, Richards JG, Higginson Two simple methods for determining gait events during treadmill and overground walking using kinematic data. Gait & Posture.https://doi.org/10.1016/j.gaitpost.2007.07. 007

**Received date:06/04/2024 Accepted date: 28/06/2024 Published date: XX/XX/2024** Volume: 1 **Publication year: 2024**